

## Google Colab (Yahoo Finance Run Only)

### 1 Introduction

Item	Detail
Context	Modern quant strategies increasingly rely on <b>multi-model pipelines</b> that fuse price, sentiment, and macro data. Few undergraduate studies still show the <i>full</i> workflow—data → α-generation → deployment—and remain reproducible on free cloud GPUs.
Research Gap	Prior projects often report Sharpe or RMSE in isolation, omit transaction costs, and skip walk-forward testing. There is no open, <b>Yahoo-Finance-only</b> reference that meets hedge-fund validation standards.
Research Question	<i>Can a fully open-source, learning ensemble built on public Yahoo Finance OHLCV deliver Sharpe &gt; 2, ('rq-sharpe), RMSE &lt; <math>1.2 \times 10^{-3}</math> ('rq-rmse), and orderly drawdowns after realistic 10 bps costs?</i>
Objectives	1. Build a 14-step modular pipeline in Google Colab. 2. Ensemble TFT, GRU, LSTM, GNN, XGBoost with Bayesian HPO (Optuna 30–50-trial prototype; full sweep 180–500 planned) 3. Evaluate on BTC-USD, AAPL, SPX (via SPY ETF) over ≈ 2023-06 → 2025-05 with a 180-fold walk-forward baseline (60 train / 5 test / 5 step) and an extended 879-fold Optuna walk-forward for hyper-parameter sweeps, both using 10 bps costs.
Key Contributions	1. Package a Dockerized FastAPI service for real-time inference. 2. 310-feature library (technical, sentiment, macro, regime) → 98 survive QA. 3. Current OOS mean metrics (Step 11C-2) → code: <pre>pythonsbr d1_metrics_exec[['sharpe','sortino','max_dd']].mean()&gt;br&gt;# ▶ Sharpe: 1.02 Sortino: 1.52 Max-DD: -11.7 %&lt;br&gt;</pre> 4. Reproducible end-to-end in ≈ 18 min on a free Tesla T4—no paid APIs or proprietary feeds.

### 2 Method

#### 2.1 Dataset & Pre-Processing (Steps 1 – 3)

Dimension	Current specification (Yahoo Finance only)
Assets	Crypto (BTC-USD), US Equity (AAPL), Equity Index (SPX via SPY ETF)
Frequency	1-hour bars, aligned to UTC
History Window	≈ 725 days per symbol (2023-06 → 2025-05)
Data QA	<ul style="list-style-type: none"> <li>Drop bars where bid-ask &gt; 3 × IQR</li> <li>Forward-fill ≤ 2 gaps, else flag NA</li> </ul>
Feature Library (310 → 98)	110 technical · 80 sentiment (VADER + finBERT) · 40 macro · 80 vol/regime

Synthetic bars are used **only** for leakage unit-tests (Step 6A, Hurst-based fractal generator) and are excluded from training, validation, and metrics.

#### 2.2 Model Stack & Hyper-Parameter Search (Steps 4 – 5)

Model	Core Hyper-Parameters	Optuna Search Space
TFT	<code>d_model = 128, n_heads = 4, dropout = 0.10</code>	<code>d_model (64,128,256); dropout [0.05-0.30]</code>
GRU	2 layers × 256 units, Layer Norm	units [128-512]; lr [1e-4-3e-3]
LSTM	3 layers × 128 units, recurrent dropout 0.05	layers [2-4]; dropout [0-0.30]
GNN (GAT)	6 asset nodes, 2 attention layers	heads [2,4,8]; alpha [0.10-0.30]
XGBoost	eta = 0.03, max_depth = 6,500 trees	depth [3-9]; eta [0.01-0.10]

Optuna ASHA: 30–50 trials (prototype), median convergence ≈ 22.

Predictions blended via StackingRegressor → meta-XGB (5-fold CV)

#### 2.3 Validation · Execution · Risk (Steps 6 – 11)

- Walk-Forward – 180-fold baseline (60/5/5) and 879-fold Optuna sweep.
- Cost Model – 10 bps round-trip + slippage = 5 bps mean, σ 2 bps ( $\epsilon \sim \mathcal{N}(0, (2 \text{ bps})^2)$ ).
- Impact – Almgren–Chriss ( $\eta = 2.5 \times 10^{-6}$ ,  $\gamma = 2.0 \times 10^{-6}$ ).
- Risk Budget – ERC or inverse-vol, dynamic Kelly scaled to 10 % annual σ.
- Back-test – vectorbt ; artefacts saved in [/output/backtests/](#).

#### 2.4 Explainability & Deployment (Steps 12 – 14)

Element	Artifact
SHAP	shap_summary.png, water fall_top10.png
LIME	lime_report.html
FastAPI	serve_fastapi.py → /predict?symbol=AAPL&horizon=60
Docker	450 MB image ai-investor (Gunicorn + Uvicorn)
CI/CD	Github Actions → nbconvert → MLflow registry

### 3 Result

#### 3.1 Risk-Adjusted Metrics (OOS ≈ 2023-06 → 2025-05)

Metric	BTC-USD	AAPL	SPX*	Mean
Sharpe	1.05	0.98	— pending	1.02
Sortino	1.63	1.40	— pending	1.52
Max DD	-11.4 %	-11.9 %	— pending	-11.7 %
Hit Rate	52.3 %	52.1 %	— pending	52.2 %

\* SPX metrics derive from SPY ETF and will populate automatically after the next end-to-end run.

#### 3.2 Forecast-Error Metrics (1-h Horizon) — Table 3-2

Fold leaderboard export pending; per-model lines will be filled once fold\_metrics.csv is saved. The meta-stack row below still uses the synthetic placeholder.

Model	RMSE × 10 <sup>-3</sup>	MAE × 10 <sup>-3</sup>	MAPE %	Notes
Meta-Stack (OOS)	9.80	7.30	11.1	placeholder; price-unit scaling pending

### 4 Discussion

#### 4.1 Interpretation

- Alpha drivers – SHAP top-3: KAMA slope, BTC-USD ↔ SPX GNN edge, CPI YoY shock.
- Accuracy – Fold-level RMSE values (export pending) indicate feature capacity; ensemble degradation stems from scale mismatch.
- Risk – Dynamic Kelly caps drawdown at -11.7 % while harvesting Sharpe ≈ 1.0.

#### 4.2 Limitations

- No level-2 order-book in Yahoo data.
- Hourly sentiment lag; sub-minute shocks untested.
- Almgren–Chriss parameters calibrated on equities; FX/index micro-structure may differ.

#### 4.3 Future Work

- Neural-SDE execution model.
- Transformer decoder for LOBSTER imbalance.
- Vector-valued Kelly across vol surface.

[rq-rmse]: The <  $1.2 \times 10^{-3}$  RMSE ceiling is not yet met; the current placeholder OOS RMSE is  $9.8 \times 10^{-3}$ .

[rq-sharpe]: The > 2.0 Sharpe target is not yet met; OOS mean Sharpe is 1.02 (see aggregation code in Introduction).

## Table of Contents

- [Step 1: Install Libraries](#)
- [Step 2: High-Frequency and Multi-Modal Data Expansion](#)
- [Step 3: Advanced Feature Engineering \(Time2Vec, Fractals, KAMA\)](#)
- [Step 4: Advanced Model Building \(TFT, GNN, Multi-task Learning\)](#)
  - [Step 4.5: Prepare Sequences from Engineered DataFrame](#)
- [Step 5: Hyperparameter Optimization \(Ray Tune\)](#)
- [Step 6: Synthetic-to-Walk-Forward Backtesting & Regime-Analysis Pipeline](#)
  - [Step 6A: Multi-Day Synthetic OHLCV Generator](#)
  - [Step 6B: Walk-Forward GRU Meta-Signal Generator](#)
  - [Step 6C: Backtrader Integration](#)
  - [Step 6D: Full Backtest & Optimization Pipeline \(Optuna\)](#)
  - [Step 6E: Regime Tagging & Performance Visualization](#)
  - [Step 6F: Regime-Based Performance Analysis & Walk-Forward Backtest](#)
  - [Step 6G: Gaussian HMM Market-Regime Detection](#)
- [Step 7: Real Historical Market Data Integration](#)
  - [Step 7A: Fetch Historical Data \(Yahoo Finance\)](#)
  - [Step 7B: Data-Quality Validation & FDA](#)
  - [Step 7C-1: Historical Data Preparation for Modelling](#)
  - [Step 7C-2: Add Overnight-Gap Features](#)
- [Step 8: Alternative Data Integration](#)
  - [Step 8A: Twitter Sentiment Analysis](#)
  - [Step 8B: News Sentiment via NLP](#)
  - [Step 8C: Macroeconomic Indicators Integration \(FRED\)](#)
  - [Step 8D: Google Trends Feature Engineering](#)
- [Step 9: Advanced Validation – Walk-Forward Testing](#)
  - [Step 9A: Implement Walk-Forward Validation](#)
  - [Step 9B: Rolling-Window Retraining & Hyperparameter Optimization](#)
- [Step 10: Robust Execution & Risk Modeling](#)
  - [Step 10A: Transaction Cost & Slippage Modeling](#)
  - [Step 10B: Market Impact Modeling \(Almgren–Chriss\)](#)
- [Step 11: Advanced Portfolio & Risk Management](#)
  - [Step 11A-1: Dynamic Position Sizing](#)
  - [Step 11A-Plus: Dynamic Sizing \(Kelly 0.5 x + Vol-Target\)](#)
  - [Step 11B-1: Diversification & Risk-Parity Construction](#)
  - [Step 11B-2: Risk-Parity Portfolio Diagnostics & Plot](#)
  - [Step 11B-Plus: Re-fit / Metrics / Comparison](#)
  - [Step 11C-1: Dynamic Position Sizing \(Extended\)](#)
  - [Step 11C-2: Risk-Parity Pro – Institutional Engine](#)
- [Step 12: Explainability & Interpretability \(XAI\)](#)
  - [Step 12A-1: SHAP Feature Importance \(robust\)](#)
  - [Step 12A-2: Production-Ready SHAP](#)
  - [Step 12A-3: SHAP \(Lag-Safe / Version-Safe\)](#)
  - [Step 12B-1: LIME for Signal Explanation](#)
  - [Step 12B-2: LIME Local Explanations \(CPU-Safe\)](#)
- [Step 13: Ensemble & Meta-Modeling](#)
  - [Step 13A-1: Multi-Model Ensemble](#)
  - [Step 13A-2: Multi-Model Ensemble \(Patched\)](#)
  - [Step 13B-1: Meta-Model Stacking \(RMSE-Safe\)](#)
  - [Step 13B-2: Meta-Model Stacking with Optuna & Sharpe](#)
- [Step 14: Advanced Deployment Infrastructure](#)
  - [Step 14A: build\\_bundle.py](#)
  - [Step 14B: Real-Time Prediction Pipeline](#)
  - [Step 14C: Automated Model Deployment & Scaling](#)

### ▼ TensorFlow GPU Check

```
1 import tensorflow as tf
2 print("TensorFlow version:", tf.__version__)
3 print("Num GPUs Available:", len(tf.config.list_physical_devices('GPU')))

→ TensorFlow version: 2.18.0
Num GPUs Available: 1
```

### ▼ NVIDIA-SMI GPU Info

```
1 gpu_info = !nvidia-smi
2 gpu_info = '\n'.join(gpu_info)
3 if gpu_info.find('failed') >= 0:
4     print('Not connected to a GPU')
5 else:
6     print(gpu_info)

→ Sat May 31 00:33:16 2025
+-----+
| NVIDIA-SMI 550.54.15   Driver Version: 550.54.15 CUDA Version: 12.4 |
|=====================================================================|
| GPU  Name        Persistence-M  Bus-Id      Disp.A  Volatile Uncorr. ECC |
| Fan  Temp  Perf  Pwr:Usage/Cap| Memory-Usage | GPU-Util  Compute M. |
|=====================================================================|
|  0  NVIDIA A100-SXM4-40GB   Off   00000000:00:04.0 Off    0% Default      |
|  N/A  34C    P0    47W / 400W |      5MiB / 40960MiB |      0%      Disabled |
+-----+
Processes:
| GPU  GI  CI      PID  Type  Process name          GPU Memory Usage |
| ID  ID
|=====================================================================|
| No running processes found
+-----+
```

## Matplotlib Inline Setup

```
1 %matplotlib inline
2 import matplotlib.pyplot as plt
3 import matplotlib
4 matplotlib.use("Agg") # Prevents some rendering issues in Colab
5 from IPython.display import display, Image
```

## Step 1 : Install Libraries

```
1 import subprocess
2
3 subprocess.call(["pip", "install", "-q",
4     "numpy", "pandas", "scikit-learn", "xgboost", "lightgbm", "ta",
5     "tensorflow", "torch", "autokeras",
6     "ray[tune]", "optuna", "backtrader", "pypfopt",
7     "fastapi", "unicorn",
8     "shap", "statsmodels",
9     "yfinance", # Load AAPL, BTC-USD etc.
10    "ccxt", # Binance, Coinbase, Kraken (real crypto data)
11    "matplotlib", # Plotting equity curves
12    "matplotlib-inline", # Smooth notebook plots
13    "ipykernel", # Jupyter compatibility
14    "nbformat", # Notebook formatting (for export/save)
15])
```

## Step 2 : High-Frequency and Multi-Modal Data Expansion

```
1 # Install missing package
2 !pip install hmmlearn
3
4 # Import Libraries
5 import pandas as pd
6 import numpy as np
7 from sklearn.preprocessing import StandardScaler
8 from hmmlearn.hmm import GaussianHMM
9
10 # Generate Minute-Level Intraday Timestamps (NYSE)
11 intraday_times = pd.date_range(start="2024-01-01 09:30", end="2024-01-01 16:00", freq="1min")
12 intraday_data = pd.DataFrame(index=intraday_times)
13
14 # Synthetic Price Generation (GARCH-style)
15 np.random.seed(42)
16 mu, sigma = 0, 0.15
17 shock = np.random.normal(mu, sigma, len(intraday_data))
18 volatility = np.zeros(len(intraday_data))
19 volatility[0] = 0.2
20 for t in range(1, len(intraday_data)):
21     volatility[t] = 0.9 * volatility[t-1] + 0.1 * np.abs(shock[t])
22 returns = shock * volatility
23 price = 100 + np.cumsum(returns)
24
25 intraday_data['Open'] = price
26 intraday_data['High'] = price + np.abs(np.random.normal(0, 0.2, len(price)))
27 intraday_data['Low'] = price - np.abs(np.random.normal(0, 0.2, len(price)))
28 intraday_data['Close'] = price + np.random.normal(0, 0.05, len(price))
29 intraday_data['Volume'] = np.random.randint(200, 1500, len(price))
30
31 # Macro & Sentiment Features
32 news_trend = np.linspace(-1, 1, len(intraday_data))
33 news_noise = np.random.normal(0, 0.2, len(intraday_data))
34 intraday_data['news_sentiment'] = (news_trend + news_noise).cumsum() / np.arange(1, len(intraday_data)+1)
35
36 macro_base = np.sin(np.linspace(0, 10 * np.pi, len(intraday_data)))
37 macro_spike = np.zeros(len(intraday_data))
38 for idx in [60, 180, 300]:
39     macro_spike[idx:idx+5] = np.linspace(1.0, 0.2, 5) + np.random.normal(0, 0.05, 5)
40 intraday_data['macro_shock'] = macro_base + macro_spike
41
42 # Time Features
43 intraday_data['minute'] = intraday_data.index.minute
44 intraday_data['hour'] = intraday_data.index.hour
45 intraday_data['time_block'] = pd.cut(
46     intraday_data.index.hour + intraday_data.index.minute / 60,
47     bins=[9.5, 11, 13, 15, 16],
48     labels=["Open", "LateMorning", "Afternoon", "Close"],
49     include_lowest=True
50 )
51
52 # Microstructure Features
53 intraday_data['order_imbalance'] = np.tanh(np.random.normal(0, 0.5, len(intraday_data)))
54 intraday_data['queue_imbalance'] = np.clip(np.random.beta(2, 5, len(intraday_data)) * 2 - 1, -1, 1)
55
56 # Cointegration Z-Score with Sector ETF
57 intraday_data['sector_etf'] = intraday_data['Close'] * (1 + np.random.normal(0, 0.01, len(intraday_data)))
58 spread = intraday_data['Close'] - 0.98 * intraday_data['sector_etf']
59 intraday_data['cointegration_zscore'] = (spread - spread.rolling(30).mean()) / (spread.rolling(30).std() + 1e-6)
60
61 # Return and Volatility Metrics
62 intraday_data['log_return'] = np.log(intraday_data['Close']).diff()
63 intraday_data['realized_vol_5'] = intraday_data['Close'].pct_change().rolling(5).std()
64 intraday_data['realized_vol_15'] = intraday_data['Close'].pct_change().rolling(15).std()
65
66 # Hurst Exponent
67 def hurst(ts):
68     lags = range(2, 20)
69     tau = [np.std(np.subtract(ts[lag:], ts[:-lag])) for lag in lags]
70     return np.polyfit(np.log(lags), np.log(tau), 1)[0]
71 intraday_data['hurst_60'] = intraday_data['Close'].rolling(60).apply(hurst, raw=True)
72
73 # EMA Regime Residual
74 intraday_data['ema_fast'] = intraday_data['Close'].ewm(span=5).mean()
75 intraday_data['ema_slow'] = intraday_data['Close'].ewm(span=30).mean()
76 intraday_data['regime_residual'] = intraday_data['Close'] - intraday_data['ema_slow']
77
78 # Target Label (Future Return)
79 intraday_data['future_return'] = intraday_data['Close'].shift(-5) / intraday_data['Close'] - 1
80 intraday_data['alpha_label'] = (intraday_data['future_return'] > 0).astype(int)
81
82 # Lagged Features
83 for lag in [1, 3, 5]:
84     intraday_data[f'return_lag_{lag}'] = intraday_data['log_return'].shift(lag)
85     intraday_data[f'vol_lag_{lag}'] = intraday_data['realized_vol_5'].shift(lag)
86
87 # Explainability Buckets
88 intraday_data['vol_bucket'] = pd.qcut(intraday_data['realized_vol_5'], 3, labels=["LowVol", "MedVol", "HighVol"])
89 intraday_data['zscore_bucket'] = pd.qcut(intraday_data['cointegration_zscore'], 3, labels=["MeanRev", "Neutral", "Diverged"])
```

```

91 # Peer Asset Features
92 peer_price = price * (1 + np.random.normal(0, 0.005, len(price)))
93 peer_df = pd.DataFrame(index=intraday_df.index)
94 peer_df['peer_close'] = peer_price
95 peer_df['peer_volume'] = np.random.randint(300, 1600, len(price))
96 intraday_data = intraday_data.join(peer_df)
97
98 # Normalize Selected Features
99 scaler = StandardScaler()
100 cols_to_normalize = [
101     'news_sentiment', 'macro_shock', 'order_imbalance', 'queue_imbalance',
102     'cointegration_zscore', 'realized_vol_5', 'realized_vol_15',
103     'hurst_60', 'regime_residual', 'return_lag_1', 'return_lag_3', 'return_lag_5',
104     'vol_lag_1', 'vol_lag_3', 'vol_lag_5'
105 ]
106 intraday_data[cols_to_normalize] = scaler.fit_transform(intraday_data[cols_to_normalize])
107
108 # Hidden Markov Model (Market Regime Classification)
109 hmm_input = intraday_data[['log_return', 'realized_vol_5']].dropna()
110 hmm_model = GaussianHMM(n_components=3, n_iter=100)
111 hmm_model.fit(hmm_input)
112 intraday_data['market_regime'] = np.nan
113 intraday_data.loc[hmm_input.index, 'market_regime'] = hmm_model.predict(hmm_input)
114 intraday_data['market_regime'] = intraday_data['market_regime'].ffill()
115 intraday_data['market_regime'] = intraday_data['market_regime'].bfill()
116
117 # Lead-Lag Feature
118 intraday_data['peer_lag_1'] = intraday_data['peer_close'].shift(1)
119 intraday_data['lead_lag_return'] = intraday_data['peer_lag_1'] / intraday_data['Close'] - 1
120 intraday_data.dropna(inplace=True, subset=['peer_close', 'peer_lag_1', 'lead_lag_return'])
121
122 # Entropy Estimation
123 def rolling_entropy(series, window=20):
124     binned = pd.qcut(series.dropna(), q=5, labels=False, duplicates='drop')
125     return binned.rolling(window).apply(lambda x: -np.sum(pd.Series(x).value_counts(normalize=True) *
126             np.log(pd.Series(x).value_counts(normalize=True) + 1e-9)))
127 intraday_data['return_entropy'] = rolling_entropy(intraday_data['log_return'])
128
129 # Interaction Terms
130 intraday_data['zscore_x_vol'] = intraday_data['cointegration_zscore'] * intraday_data['realized_vol_5']
131 intraday_data['shock_x_sent'] = intraday_data['macro_shock'] * intraday_data['news_sentiment']
132
133 # Feature Stability Index
134 feature_std = intraday_data[cols_to_normalize].rolling(60).std()
135 feature_mean = intraday_data[cols_to_normalize].rolling(60).mean()
136 intraday_data['feature_stability_index'] = (feature_std / (feature_mean + 1e-9)).mean(axis=1)
137 entropy_cols = ['return_entropy', 'zscore_x_vol', 'shock_x_sent', 'feature_stability_index']
138 intraday_data[entropy_cols] = intraday_data[entropy_cols].ffill()
139 intraday_data.dropna(inplace=True, subset=entropy_cols)
140
141 # Cleanup and Preview
142 intraday_data.dropna(inplace=True) # Final cleanup before Step 4
143
144 # Define only columns that actually exist in the DataFrame
145 min_required_features = [
146     col for col in [
147         'KAMA', 'KAMA_signal', 'fractal_signal',
148         'realized_vol_5_vol_rolling', 'log_return_vol_rolling'
149     ] if col in intraday_data.columns
150 ]
151
152 # Drop rows that have NaNs in those existing features
153 intraday_data = intraday_data.dropna(subset=min_required_features)
154 print(intraday_data.head())

```

Collecting hmmlearn

```

  Downloading hmmlearn-0.3.3-cp311-cp311-manylinux2014_x86_64.whl.metadata (3.0 kB)
Requirement already satisfied: numpy<1.10 in /usr/local/lib/python3.11/dist-packages (from hmmlearn) (2.0.2)
Requirement already satisfied: scikit-learn<0.22.0,>0.16 in /usr/local/lib/python3.11/dist-packages (from hmmlearn) (1.6.1)
Requirement already satisfied: scipy<0.19 in /usr/local/lib/python3.11/dist-packages (from hmmlearn) (1.15.3)
Requirement already satisfied: joblib<1.2.0 in /usr/local/lib/python3.11/dist-packages (from hmmlearn) (1.15.0)
Requirement already satisfied: threadpoolctl<>3.1.0 in /usr/local/lib/python3.11/dist-packages (from scikit-learn<0.22.0,>0.16>hmmlearn) (3.6.0)
  Downloading hmmlearn-0.3.3-cp311-cp311-manylinux2_17_x86_64.manylinux2014_x86_64.whl (165 kB)

```

Installing collected packages: hmmlearn

```

Successfully installed hmmlearn-0.3.3
      Open   High    Low  Close  Volume  W
2024-01-01 10:30:00  99.832834 100.023862 99.608289 99.961819   760
2024-01-01 10:31:00  99.830161 100.027307 99.753679 99.799978  1159
2024-01-01 10:32:00  99.813075 99.913885 99.779785 99.895031   231
2024-01-01 10:33:00  99.793229 99.899281 99.694739 99.877114   590
2024-01-01 10:34:00  99.806847 99.965422 99.749013 99.779168   476

```

	news_sentiment	macro_shock	minute	hour	time_block	W
2024-01-01 10:30:00	-1.188420	-0.131632	30	10	Open	
2024-01-01 10:31:00	-1.165064	-0.379678	31	10	Open	
2024-01-01 10:32:00	-1.135466	-0.652896	32	10	Open	
2024-01-01 10:33:00	-1.131620	-0.768226	33	10	Open	
2024-01-01 10:34:00	-1.123252	-1.012494	34	10	Open	

	zscore_bucket	peer_close	peer_volume	W
2024-01-01 10:30:00	MeanRev	100.318330	620	
2024-01-01 10:31:00	MeanRev	99.774987	573	
2024-01-01 10:32:00	Diverged	100.420786	855	
2024-01-01 10:33:00	Neutral	99.920266	1143	
2024-01-01 10:34:00	Diverged	100.351036	967	

	market_regime	peer_lag_1	lead_lag_return	W
2024-01-01 10:30:00	1.0	100.572639	0.006111	
2024-01-01 10:31:00	1.0	100.318330	0.005295	
2024-01-01 10:32:00	1.0	99.774987	-0.001202	
2024-01-01 10:33:00	1.0	100.420786	0.005443	
2024-01-01 10:34:00	1.0	99.920266	0.001414	

	return_entropy	zscore_x_vol	shock_x_sent	W
2024-01-01 10:30:00	1.574103	-1.152603	0.156434	
2024-01-01 10:31:00	1.544480	-3.314965	0.442350	
2024-01-01 10:32:00	1.544480	1.101367	0.741114	
2024-01-01 10:33:00	1.544480	-0.324280	0.869339	
2024-01-01 10:34:00	1.514856	4.192407	1.137286	

	feature_stability_index	W
2024-01-01 10:30:00	-3.155770	
2024-01-01 10:31:00	-21.486163	
2024-01-01 10:32:00	-2.618548	
2024-01-01 10:33:00	-16.957479	
2024-01-01 10:34:00	21.304647	

[5 rows x 40 columns]

### Step 3 : Advanced Feature Engineering (Time2Vec, Fractals, KAMA)

```

1 # Advanced Feature Engineering (corrected & stable)
2
3 # Dependencies
4 !pip install -q tensorflow scikit-learn hmmlearn
5
6 import numpy as np
7 import pandas as pd
8 import tensorflow as tf
9 from sklearn.preprocessing import MinMaxScaler

```

```

10 from hmmlearn.hmm import GaussianHMM
11
12 # Synthetic Intraday OHLC
13 np.random.seed(42)
14 times = pd.date_range("2024-01-02 09:30", periods=390, freq="1min")
15 df = pd.DataFrame(index=times)
16 df["Close"] = 100 + np.cumsum(np.random.normal(0, 0.2, size=390))
17 df["High"] = df["Close"] + np.random.normal(0.1, 0.05, 390)
18 df["Low"] = df["Close"] - np.random.normal(0.1, 0.05, 390)
19 df["Open"] = df["Close"].shift(1).fillna(df["Close"])
20
21 # Time2Vec encoding
22 class Time2Vec(tf.keras.layers.Layer):
23     def __init__(self, k=4):
24         super().__init__()
25         self.k = k # number of periodic terms
26
27     def build(self, input_shape):
28         self.wb = self.add_weight(name="wb", shape=(1,), initializer="uniform")
29         self.bb = self.add_weight(name="bb", shape=(1,), initializer="uniform")
30         self.wa = self.add_weight(
31             name="wa", shape=(input_shape[-1], self.k), initializer="uniform"
32         )
33         self.ba = self.add_weight(name="ba", shape=(self.k,), initializer="uniform")
34
35     def call(self, inputs):
36         bias = self.wb * inputs + self.bb
37         wghts = tf.math.sin(tf.matmul(inputs, self.wa) + self.ba)
38         return tf.concat([bias, wghts], axis=-1)
39
40 time_vals = np.arange(len(df)).reshape(-1, 1)
41 time_scaled = MinMaxScaler().fit_transform(time_vals)
42
43 t2v_layer = Time2Vec(k=4)
44 t2v_out = t2v_layer(tf.convert_to_tensor(time_scaled, dtype=tf.float32)).numpy()
45
46 for i in range(t2v_out.shape[1]):
47     df[f"t2v_{i}"] = t2v_out[:, i]
48
49 df["sin_time"] = np.sin(2 * np.pi * time_scaled.flatten())
50 df["cos_time"] = np.cos(2 * np.pi * time_scaled.flatten())
51
52 # Fractal features
53 df["fractal_up"] = (
54     (df["High"].shift(2) < df["High"].shift(1)) &
55     (df["High"].shift(1) > df["High"]) &
56     (df["High"].shift() > df["High"].shift(3)) &
57     (df["High"].shift() > df["High"].shift(4))
58 ).astype(int)
59
60 df["fractal_down"] = (
61     (df["Low"].shift(2) > df["Low"].shift(1)) &
62     (df["Low"].shift(1) < df["Low"]) &
63     (df["Low"].shift() < df["Low"].shift(3)) &
64     (df["Low"].shift() < df["Low"].shift(4))
65 ).astype(int)
66
67 df["fractal_signal"] = df["fractal_up"] - df["fractal_down"]
68 df["fractal_rolling"] = df["fractal_signal"].rolling(10).sum()
69 df["fractal_momentum"] = df["fractal_rolling"].diff()
70
71 # KAMA features
72 def kama(df, window=10, fast=2/(2+1), slow=2/(30+1)):
73     close = df["Close"].dropna()
74     direction = close.diff(window).abs()
75     volatility = close.diff().abs().rolling(window).sum()
76     er = direction / (volatility + 1e-8)
77     sc = (er * (fast - slow) + slow).fillna(slow) ** 2
78
79     kama_vals = [close.iloc[0]]
80     for i in range(1, len(close)):
81         kama_vals.append(kama_vals[-1] + sc.iloc[i] * (close.iloc[i] - kama_vals[-1]))
82
83     kama_series = pd.Series(kama_vals, index=close.index)
84     df["KAMA"] = kama_series.reindex(df.index).ffill()
85     df["KAMA_residual"] = df["Close"] - df["KAMA"]
86     df["KAMA_pct_dev"] = df["KAMA_residual"] / df["KAMA"]
87     df["KAMA_slope"] = df["KAMA"].diff()
88     df["KAMA_Signal"] = np.sign(df["KAMA_residual"]) * (df["KAMA_residual"].abs() > 0.1).astype(int)
89     return df
90
91 df = kama(df)
92
93 # Volatility & regime residuals
94 df["log_return"] = np.log(df["Close"]).diff()
95 df["realized_vol_5"] = df["Close"].pct_change().rolling(5).std()
96 df["regime_residual"] = df["Close"] - df["Close"].ewm(span=30).mean()
97
98 # Future-return target & rolling stats
99 df["future_return"] = df["Close"].shift(-5) / df["Close"] - 1
100
101 for feat in ["log_return", "realized_vol_5", "regime_residual"]:
102     df[f"{feat}_vol_rolling"] = df[feat].rolling(20).std()
103     df[f"{feat}_zscores"] = (
104         df[feat] - df[feat].rolling(20).mean()
105     ) / (df[feat].rolling(20).std() + 1e-8)
106
107 df["vol_regime"] = pd.qcut(df["realized_vol_5"], 3, labels=["Low", "Medium", "High"])
108 df["meta_signal"] = df["fractal_signal"] + df["KAMA_Signal"]
109
110 # Rolling entropy
111 def rolling_entropy(series, window=30):
112     binned = pd.qcut(series.dropna(), 5, labels=False, duplicates="drop")
113     return binned.rolling(window).apply(
114         lambda x: -np.sum(pd.Series(x).value_counts(normalize=True) *
115                           np.log(pd.Series(x).value_counts(normalize=True) + 1e-9))
116     )
117
118 df["kama_entropy"] = rolling_entropy(df["KAMA_pct_dev"])
119
120 # HMM regime detection
121 hmm_input = df["log_return", "realized_vol_5"].dropna()
122 model = GaussianHMM(n_components=3, n_iter=100)
123 df["hmm_regime"] = np.nan
124 df.loc[hmm_input.index, "hmm_regime"] = model.fit(hmm_input).predict(hmm_input)
125
126 # Correlation & drift features
127 df["KAMA_alpha_corr"] = df["KAMA"].rolling(50).corr(df["log_return"].shift(-1))
128 df["fractal_alpha_corr"] = df["fractal_signal"].rolling(50).corr(df["log_return"].shift(-1))
129
130 rolling_std = df[["KAMA", "KAMA_Signal", "fractal_signal"]].rolling(60).std()
131 rolling_mean = df[["KAMA", "KAMA_Signal", "fractal_signal"]].rolling(60).mean()
132 df["feature_drift_index"] = (rolling_std / (rolling_mean + 1e-8)).mean(axis=1)
133
134 # Final cleanup
135 cols_to_fix = ["meta_signal", "kama_entropy", "KAMA_alpha_corr"]
136 df[cols_to_fix] = df[cols_to_fix].ffill()
137 df.replace([np.inf, -np.inf], np.nan, inplace=True)
138 df.ffill(inplace=True)

```

#### ✓ Step 4 : Advanced Model Building (TFT, GNN, Multi-task Learning)

```

1 # Install Dependencies
2 pip install tensorflow scikit-learn hmmlearn shap --quiet
3
4 # Imports
5 import numpy as np
6 import pandas as pd
7 import tensorflow as tf
8 from sklearn.preprocessing import MinMaxScaler
9 from sklearn.model_selection import train_test_split
10 from hmmlearn.hmm import GaussianHMM
11 import matplotlib.pyplot as plt
12 from tensorflow.keras.models import Model
13 from tensorflow.keras.layers import Input, LSTM, Dense, MultiHeadAttention, LayerNormalization, Dropout, Add
14 from tensorflow.keras.regularizers import l2
15
16 # Generate Synthetic Intraday Data (10 Days)
17 df['regime_residual'] = df['Close'] - df['Close'].ewm(span=30).mean()
18 np.random.seed(42)
19 minutes_per_day = 390
20 n_days = 60
21 total_minutes = minutes_per_day * n_days
22 times = pd.date_range('2024-01-01 09:30', periods=total_minutes, freq='1min')
23 df['Close'] = [100 + np.cumsum(np.random.normal(0, 1.0, size=total_minutes))]
24 df['High'] = df['Close'] + np.random.normal(0.1, 0.05, total_minutes)
25 df['Low'] = df['Close'] - np.random.normal(0.1, 0.05, total_minutes)
26 df['Open'] = df['Close'].shift(1).fillna(df['Close'])
27
28 # Time2Vec Layer
29 class Time2Vec(tf.keras.layers.Layer):
30     def __init__(self, kernel_size): super().__init__(); self.k = kernel_size
31     def build(self, input_shape):
32         self.wb = self.add_weight(name='wb', shape=(1,), initializer='uniform')
33         self.bb = self.add_weight(name='bb', shape=(1,), initializer='uniform')
34         self.wa = self.add_weight(name='wa', shape=(input_shape[-1], self.k), initializer='uniform')
35         self.ba = self.add_weight(name='ba', shape=(self.k,), initializer='uniform')
36     def call(self, inputs):
37         bias = self.wb * inputs + self.bb
38         wghts = tf.math.sin(tf.matmul(inputs, self.wa) + self.ba)
39         return tf.concat([bias, wghts], -1)
40
41 # Time2Vec Embedding
42 time_vals = np.arange(len(df)).reshape(-1, 1)
43 time_scaled = MinMaxScaler().fit_transform(time_vals)
44 t2v = Time2Vec(4)
45 t2v_out = t2v(tf.convert_to_tensor(time_scaled, dtype=tf.float32)).numpy()
46 for i in range(t2v_out.shape[1]):
47     df[f't2v_{i}'] = t2v_out[:, i]
48
49 # Fractal + KAMA Features
50 df['fractal_up'] = ((df['High'].shift(2) < df['High'].shift(1)) & (df['High'].shift(1) > df['High']) &
51                     (df['High'].shift(1) > df['High'].shift(3)) & (df['High'].shift(1) > df['High'].shift(4))).astype(int)
52 df['fractal_down'] = ((df['Low'].shift(2) > df['Low'].shift(1)) & (df['Low'].shift(1) < df['Low']) &
53                     (df['Low'].shift(1) < df['Low'].shift(3)) & (df['Low'].shift(1) < df['Low'].shift(4))).astype(int)
54 df['fractal_signal'] = df['fractal_up'] - df['fractal_down']
55
56 def kama(df, window=10, fast=2/(2+1), slow=2/(30+1)):
57     c = df['Close']
58     er = c.diff(window).abs() / (c.diff().abs().rolling(window).sum() + 1e-8)
59     sc = (er * (fast - slow) + slow)**2
60     kama_val = c.rolling(10).apply(lambda x: np.sqrt(sc.sum()))
61     for i in range(1, len(c)):
62         kama_val.append(kama_val[-1] + sc[i]*sc[i] * (c[i].iloc[0] - kama_val[-1]))
63     return pd.Series(kama_val, index=df.index)

```

```

64
65 df['KAMA'] = kama(df)
66 df['KAMA_pct_dev'] = (df['Close'] - df['KAMA']) / (df['KAMA'] + 1e-6)
67 df['KAMA_pct_dev'] = df['KAMA_pct_dev'].clip(-5, 5)
68 df['KAMA_slope'] = df['KAMA'].diff().clip(-5, 5)
69 df['log_return'] = np.log(df['Close']).diff().clip(-5, 5)
70 df['realized_vol_5'] = df['Close'].pct_change().rolling(5).std()
71 df['regime_residual'] = df['Close'] - df['Close'].ewm(span=30).mean()
72
73 for f in ['log_return', 'realized_vol_5', 'regime_residual']:
74     df[f'{f}_rolling'] = df[[f]].rolling(20).std()
75     df[f'{f}_score'] = (df[f] - df[f].rolling(20).mean()) / (df[f].rolling(20).std() + 1e-8)
76
77 # HMM Regime Detection
78 hmm_input = df[['log_return', 'realized_vol_5']].dropna()
79 model = GaussianHMM(n_components=3, n_iter=200, tol=1e-4, covariance_type='full', verbose=False)
80 df['hmm_regime'] = np.nan
81 df.loc[hmm_input.index, 'hmm_regime'] = model.fit(hmm_input).predict(hmm_input)
82
83 # Clean Nans & Inf's
84 df.replace(np.inf, -np.inf, np.nan, inplace=True)
85 df.fillna(inplace=True)
86 df.bfill(inplace=True)
87
88 # Define Future Return Target
89 df['future_return'] = df['Close'].shift(-5) / df['Close'] - 1
90 target = 'future_return'
91
92 # Safe Feature Selection
93 base_features = [col for col in df.columns if col.startswith("t2v_")] + [
94     'KAMA_pct_dev', 'KAMA_slope',
95     'log_return_vol_rolling', 'regime_residual_vol_rolling',
96     'realized_vol_5_vol_rolling', 'hmm_regime'
97 ]
98 features = [col for col in base_features if col in df.columns]
99
100 # Final NaN Fill for Features + Target
101 df = df.loc[:, ~df.columns.duplicated()]
102 features = list(dict.fromkeys([col for col in base_features if col in df.columns]))
103 if target in features:
104     features.remove(target)
105 columns_to_fill = features + [target]
106 df[columns_to_fill] = df[columns_to_fill].copy().ffill().bfill()
107
108
109 print("Null counts before dropna:")
110 print(df[features + [target]].isnull().sum())
111 print("→ Rows before dropna: ", len(df))
112
113 # Drop Nans In Features + Target to ensure clean sequence generation
114 df = df.dropna(subset=features + [target])
115 print("After dropna: rows = ", len(df))
116
117 # Create Sequences
118 def create_sequences(X, y, window):
119     Xs, ys = [], []
120     for i in range(len(X) - window):
121         Xs.append(X[i:i+window])
122         ys.append(y[i+window])
123     return np.array(Xs), np.array(ys)
124
125 sequence_length = 50
126 X = df[features].values.astype(np.float32)
127 y = df[target].values.astype(np.float32)
128 X_seq, y_seq = create_sequences(X, y, window=sequence_length)
129
130 # Remove Invalid Sequences
131 mask = np.isfinite(X_seq).all(axis=(1, 2)) & np.isfinite(y_seq)
132 X_seq = X_seq[mask]
133 y_seq = y_seq[mask]
134
135 print(f"Cleaned sequences: X = {X_seq.shape}, y = {y_seq.shape}")
136 if X_seq.shape[0] == 0:
137     print("No valid sequences. Check data cleaning or increase input data length.")
138 else:
139     print(f"X stats: ", np.min(X_seq), np.max(X_seq), np.isnan(X_seq).sum())
140     print(f"y stats: ", np.min(y_seq), np.max(y_seq), np.isnan(y_seq).sum())
141
142 print(f"y stats: ", np.min(y_seq), np.max(y_seq), np.isnan(y_seq).sum())
143
144 # Train-Test Split
145 X_train, X_test, y_train, y_test = train_test_split(X_seq, y_seq, test_size=0.2, shuffle=False)
146
147 # TFT Model
148 def build_tft_model(input_shape):
149     inputs = Input(shape=input_shape)
150     x = Time2Vec(kernel_size=8)(inputs)
151     x = LSTM(128, return_sequences=True, kernel_regularizer=l2(1e-4), recurrent_regularizer=l2(1e-4))(x)
152     attn = MultiHeadAttention(num_heads=4, key_dim=64)(x, x)
153     x = Add()([x, attn])
154     x = LayerNormalization()(x)
155     x = Dense(64, activation='relu', kernel_regularizer=l2(1e-4))(x)
156     x = Dropout(0.4)(x)
157     x = Dense(1)(x)
158     model = Model(inputs, x)
159     model.compile(optimizer='adam', loss='huber', metrics=['mae'])
160     return model
161
162 # Train
163 model = build_tft_model(X_train.shape[1:])
164 history = model.fit(X_train, y_train, epochs=20, batch_size=16, validation_split=0.1, verbose=1)
165
166 # Evaluate
167 loss, mae = model.evaluate(X_test, y_test)
168 print(f"Final Test Loss: {loss:.4f}, MAE: {mae:.4f}")
169
170 # Plot
171 plt.plot(history.history['loss'], label='Train Loss')
172 plt.plot(history.history['val_loss'], label='Validation Loss')
173 plt.title("TFT Training Loss")
174 plt.xlabel("Epoch")
175 plt.ylabel("Loss")
176 plt.legend()
177 plt.grid(True)
178 plt.show()

Null counts before dropna:
t2v_0          0
t2v_1          0
t2v_2          0
t2v_3          0
t2v_4          0
KAMA_pct_dev   0
KAMA_slope     0
log_return_vol_rolling  0
regime_residual_vol_rolling  0
realized_vol_5_vol_rolling  0
hmm_regime     0
future_return  0
dtype: int64
→ Rows before dropna: 23400

```

```
After dropna: rows = 23400
Cleaned sequences: X = (23350, 50, 11), y = (23350, )
X stats: -0.07119062 4.4018774 0
y stats: -0.22172816 0.241834 0
y stats: -0.22172816 0.241834 0
Epoch 1/20
1051/1051 17s 11ms/step - loss: 0.0275 - mae: 0.0535 - val_loss: 0.0062 - val_mae: 0.0099
Epoch 2/20
1051/1051 11s 10ms/step - loss: 0.0057 - mae: 0.0218 - val_loss: 0.0031 - val_mae: 0.0097
Epoch 3/20
1051/1051 11s 10ms/step - loss: 0.0030 - mae: 0.0217 - val_loss: 0.0014 - val_mae: 0.0099
Epoch 4/20
1051/1051 11s 11ms/step - loss: 0.0015 - mae: 0.0214 - val_loss: 5.8438e-04 - val_mae: 0.0098
Epoch 5/20
1051/1051 10s 10ms/step - loss: 8.7796e-04 - mae: 0.0216 - val_loss: 2.5004e-04 - val_mae: 0.0101
Epoch 6/20
1051/1051 10s 10ms/step - loss: 6.1373e-04 - mae: 0.0217 - val_loss: 1.2552e-04 - val_mae: 0.0097
Epoch 7/20
1051/1051 10s 10ms/step - loss: 5.2869e-04 - mae: 0.0218 - val_loss: 1.0021e-04 - val_mae: 0.0100
Epoch 8/20
1051/1051 10s 10ms/step - loss: 5.1202e-04 - mae: 0.0218 - val_loss: 7.8681e-05 - val_mae: 0.0097
Epoch 9/20
1051/1051 11s 10ms/step - loss: 4.7585e-04 - mae: 0.0215 - val_loss: 7.4902e-05 - val_mae: 0.0098
Epoch 10/20
1051/1051 10s 9ms/step - loss: 4.7107e-04 - mae: 0.0216 - val_loss: 7.8956e-05 - val_mae: 0.0100
Epoch 11/20
1051/1051 10s 9ms/step - loss: 4.9073e-04 - mae: 0.0218 - val_loss: 7.4188e-05 - val_mae: 0.0097
Epoch 12/20
1051/1051 10s 10ms/step - loss: 4.8733e-04 - mae: 0.0218 - val_loss: 7.4128e-05 - val_mae: 0.0097
Epoch 13/20
1051/1051 11s 10ms/step - loss: 4.8454e-04 - mae: 0.0218 - val_loss: 7.4104e-05 - val_mae: 0.0097
Epoch 14/20
1051/1051 10s 10ms/step - loss: 4.7435e-04 - mae: 0.0217 - val_loss: 7.4110e-05 - val_mae: 0.0097
Epoch 15/20
1051/1051 10s 10ms/step - loss: 4.7074e-04 - mae: 0.0215 - val_loss: 7.4112e-05 - val_mae: 0.0097
Epoch 16/20
1051/1051 10s 10ms/step - loss: 4.8311e-04 - mae: 0.0216 - val_loss: 7.4945e-05 - val_mae: 0.0098
Epoch 17/20
1051/1051 10s 10ms/step - loss: 4.7050e-04 - mae: 0.0215 - val_loss: 7.4107e-05 - val_mae: 0.0097
Epoch 18/20
1051/1051 10s 10ms/step - loss: 4.9348e-04 - mae: 0.0220 - val_loss: 7.4990e-05 - val_mae: 0.0098
Epoch 19/20
1051/1051 11s 10ms/step - loss: 4.8973e-04 - mae: 0.0217 - val_loss: 7.4259e-05 - val_mae: 0.0097
```

▼ STEP 4.5 : Prepare Sequences from my engineered DataFrame

```

1 # df : full cleaned DataFrame
2 # features : list of feature column names
3 # target : prediction target column name
4 # sequence_length : window size (usually 50)
5
6 # Convert to arrays
7 X = df[features].values.astype(np.float32)
8 y = df[target].values.astype(np.float32)
9
10 # Create sequences (sliding window)
11 def create_sequences(X, y, window):
12     Xs, ys = [], []
13     for i in range(len(X) - window):
14         Xs.append(X[i:i+window])
15         ys.append(y[i+window])
16     return np.array(Xs), np.array(ys)
17
18 sequence_length = 50
19 X_seq, y_seq = create_sequences(X, y, window=sequence_length)
20
21 # Drop NaNs/Infs from sequences
22 mask = np.isfinite(X_seq).all(axis=(1, 2)) & np.isfinite(y_seq)
23 X_seq = X_seq[mask]
24 y_seq = y_seq[mask]
25
26 print(f"Sequences created: X_seq.shape = {X_seq.shape}, y_seq.shape = {y_seq.shape}")
27
28 # Train/Validation split
29 from sklearn.model_selection import train_test_split
30
31 X_train_seq, X_val_seq, y_train_seq, y_val_seq = train_test_split(
32     X_seq, y_seq, test_size=0.2, shuffle=False
33 )
34
35 REAL_FEATURE_COUNT = X_train_seq.shape[2]

```

→ Sequences created: X\_seq.shape = (23350, 50, 11), y\_seq.shape = (23350,)

#### ▼ Step 5 : Hyperparameter Optimization (Ray Tune)

```

1 # Patch threading and metric issues (must be at the top)
2 import os
3 os.environ["OPENBLAS_NUM_THREADS"] = "1"
4 os.environ["OMP_NUM_THREADS"] = "1"
5 os.environ["MKL_NUM_THREADS"] = "1"
6 os.environ["TUNE_DISABLE_STRICT_METRIC_CHECKING"] = "1"
7
8 # Install & Import Libraries
9 import subprocess
10 subprocess.run(["pip", "install", "-q", "ray[tune]", "tensorflow"], check=True)
11
12 import ray
13 from ray import tune, air
14 from ray.tune.schedulers import ASHAScheduler
15 import tensorflow as tf
16 import numpy as np
17 import logging
18 logging.basicConfig(level=logging.INFO)
19
20 # Time2vec Layer
21 class Time2Vec(tf.keras.layers.Layer):
22     def __init__(self, kernel_size):
23         super().__init__()
24         self.k = kernel_size
25         self.wb = self.add_weight(name='wb', shape=(1,), initializer='uniform')
26         self.bb = self.add_weight(name='bb', shape=(1,), initializer='uniform')
27         self.wa = self.add_weight(name='wa', shape=(input_shape[-1], self.k), initializer='uniform')
28         self.ba = self.add_weight(name='ba', shape=(self.k,), initializer='uniform')
29     def call(self, x):
30         bias = self.wb * x + self.bb
31         wgt = tf.math.sin(tf.matmul(x, self.wa) + self.ba)
32         return tf.concat([bias, wgt], -1)
33
34 # Check Sequence Data
35 try:
36     REAL_FEATURE_COUNT = X_train_seq.shape[2]
37 except:
38     raise ValueError("I must run Step 4 first to define X_train_seq, etc.")
39
40 # Training Function (with safe dictionary-based tune.report)
41 def train_model(config):
42     try:
43         X_train = X_train_seq.copy()

```

```

43     y_train = y_train_seq.copy()
44     X_val = X_val_seq.copy()
45     y_val = y_val_seq.copy()
46
47     inputs = tf.keras.Input(shape=(50, REAL_FEATURE_COUNT))
48     x = Time2Vec(config["t2v_k"])(inputs)
49     x = tf.keras.layers.LSTM(config["lstm_units"], return_sequences=True)(x)
50     x = tf.keras.layers.MultiHeadAttention(num_heads=4, key_dim=32)(x, x)
51     x = tf.keras.layers.LayerNormalization()(x)
52     x = tf.keras.layers.Dense(config["dense_units"], activation="relu")(x)
53     x = tf.keras.layers.Dropout(config["dropout_rate"])(x)
54     x = tf.keras.layers.Dense(1)(x)
55
56     model = tf.keras.Model(inputs, x)
57     model.compile(optimizer=tf.keras.optimizers.Adam(config["learning_rate"]),
58                   loss="mse")
59
60     model.fit(X_train, y_train,
61                validation_data=(X_val, y_val),
62                epochs=3, batch_size=16,
63                verbose=0)
64
65     val_loss = float(model.evaluate(X_val, y_val, verbose=0))
66
67     # Dictionary-style report for maximum compatibility
68     tune.report({"loss": val_loss})
69
70 except Exception as e:
71     import traceback
72     print("Trial failed:")
73     traceback.print_exc()
74     # Dictionary fallback to avoid TypeError
75     tune.report({"loss": 9999.0})
76
77 # Search Space
78 search_space = {
79     "lstm_units": tune.choice([64, 128]),
80     "dense_units": tune.choice([32, 64]),
81     "learning_rate": tune.loguniform(1e-4, 1e-2),
82     "dropout_rate": tune.uniform(0.1, 0.3),
83     "t2v_k": tune.choice([4, 6])
84 }
85
86 # Scheduler
87 scheduler = ASHAScheduler(metric="loss", mode="min")
88
89 # Logger Callback
90 from ray.tune import Callback
91 class TrialLogger(Callback):
92     def on_trial_result(self, iteration, trials, trial, result, **info):
93         print(f"Trial {trial.trial_id} | Loss={result.get('loss', 'n/a'): .4f}")
94
95 # Initialize Ray
96 ray.shutdown()
97 ray.init(local_mode=True, num_cpus=1, ignore_reinit_error=True, include_dashboard=False)
98
99 # Define Tuner
100 tuner = tune.Tuner(
101     train_model,
102     param_space=search_space,
103     tune_config=tune.TuneConfig(
104         scheduler=scheduler,
105         num_samples=2 # Increase for real tuning
106     ),
107     run_config=air.RunConfig(
108         name="ray_tune_tfl_final",
109         callbacks=[TrialLogger()],
110         failure_config=air.FailureConfig(max_failures=0)
111     )
112 )
113
114 # Run Tuning
115 try:
116     results = tuner.fit()
117 except Exception as e:
118     import traceback
119     print("Tuner crashed:")
120     traceback.print_exc()
121 try:
122     if results and results.num_errors == 0:
123         best_result = results.get_best_result(metric="loss", mode="min")
124         print("Best Config:", best_result.config)
125     else:
126         print("All trials failed. See logs above.")
127 except Exception as e:
128     print("Could not retrieve best config:")
129     import traceback
130     traceback.print_exc()
131
132 # Step 5.11: Retrain Final Model Using Best Config
133 best_config = best_result.config # already printed above
134
135 # Build Model
136 inputs = tf.keras.Input(shape=(50, REAL_FEATURE_COUNT))
137 x = Time2Vec(best_config["t2v_k"])(inputs)
138 x = tf.keras.layers.LSTM(best_config["lstm_units"], return_sequences=True)(x)
139 x = tf.keras.layers.MultiHeadAttention(num_heads=4, key_dim=32)(x, x)
140 x = tf.keras.layers.LayerNormalization()(x)
141 x = tf.keras.layers.Dense(best_config["dense_units"], activation="relu")(x)
142 x = tf.keras.layers.Dropout(best_config["dropout_rate"])(x)
143 x = tf.keras.layers.Dense(1)(x)
144
145 final_model = tf.keras.Model(inputs, x)
146 final_model.compile(
147     optimizer=tf.keras.optimizers.Adam(best_config["learning_rate"]),
148     loss="mse"
149 )
150
151 # Train on Full Data
152 history = final_model.fit(
153     X_train_seq, y_train_seq,
154     validation_data=(X_val_seq, y_val_seq),
155     epochs=30, # Change could be made
156     batch_size=16,
157     verbose=1
158 )
159
160 import matplotlib.pyplot as plt
161
162 # Save to file
163 plt.figure(figsize=(8, 5))
164 plt.plot(history.history['loss'], label='Train Loss')
165 plt.plot(history.history['val_loss'], label='Validation Loss')
166 plt.title('Final Model Training vs Validation Loss')
167 plt.xlabel('Epoch')
168 plt.ylabel('MSE Loss')
169 plt.legend()
170 plt.grid(True)
171

```

```

172 # Save the plot as PNG
173 plt.savefig("training_loss_plot.png") # Changing the name is possible
174
175 # Show the saved image inline
176 from IPython.display import Image, display
177 display(Image("training_loss_plot.png"))

2025-05-31 00:37:26,170 INFO worker.py:1888 -- Started a local Ray Instance.

+-----+
| Configuration for experiment      ray_tune_tft_final |
+-----+
| Search algorithm      BasicVariantGenerator |
| Scheduler            AsyncHyperBandScheduler |
| Number of trials       2                         |
+-----+

View detailed results here: /root/ray_results/ray_tune_tft_final
To visualize your results with TensorBoard, run: `tensorboard --logdir /tmp/ray/session_2025-05-31_00-37-24_338114_1273/artifacts/2025-05-31_00-37-27/ray_tune_tft_final/driver_artifacts`
```

:job\_id:0100000  
:task\_name:bundle\_reservation\_check\_func  
:job\_id:0100000  
:task\_name:bundle\_reservation\_check\_func  
:actor\_name:implicitFunc  
:actor\_name:train\_model

Trial train\_model\_6d508\_00000 started with configuration:

Trial train_model_6d508_00000 config
+-----+
dense_units          32
dropout_rate         0.22614
learning_rate        0.00596
lstm_units           64
t2v_K                4
+-----+
:actor_name:implicitFunc
:actor_name:train_model

Trial status: 1 RUNNING  
Current time: 2025-05-31 00:38:14. Total running time: 40s  
Logical resource usage: 1.0/1 CPUs, 0/1 GPUs (0.0/1.0 accelerator\_type:A100)

Trial name      status      lstm_units      dense_units      learning_rate      dropout_rate      t2v_K
+-----+
train_model_6d508_00000    RUNNING        64            32           0.00596376     0.226136        4
+-----+

Trial 6d508\_00000 | Loss = 0.0002

Trial train\_model\_6d508\_00000 completed after 1 iterations at 2025-05-31 00:38:14. Total running time: 40s

Trial train_model_6d508_00000 result
+-----+
checkpoint_dir_name
time_this_iter_s   38.9531
time_total_s       38.9531
training_iteration 1
loss               0.00016
+-----+
:task_name:bundle_reservation_check_func
:actor_name:implicitFunc
:actor_name:train_model

Trial train\_model\_6d508\_00001 started with configuration:

Trial train_model_6d508_00001 config
+-----+

## Step 6 : Synthetic-to-Walk-Forward Backtesting & Regime-Analysis Pipeline

1. Step 6A : Multi-Day Synthetic OHLCV Generator
2. Step 6B : Walk-Forward GRU Meta-Signal Generator with Multi-GPU, Data-Leakage Protection & F1 Evaluation
3. Step 6C : Backtrader Integration
4. Step 6D : Full Backtest & Optimization Pipeline with Optuna + Final Equity-Curve Plotting
5. Step 6E : Regime Tagging, Performance Visualization & Walk-Forward Preparation
6. Step 6F : Regime-Based Performance Analysis & Walk-Forward Backtest Execution
7. Step 6G : Full Python Pipeline for Market Regime Detection using Gaussian HMM + Equity Performance Analysis with Synthetic Data

### Step 6A : Multi-Day Synthetic OHLCV Generator

```

1 import pandas as pd
2 import numpy as np
3
4 # Generate 60 Days of Synthetic Intraday OHLCV Data
5 np.random.seed(42)
6
7 def generate_intraday_day(start_date):
8     minutes = pd.date_range(start=start_date + " 09:30", periods=390, freq="1min")
9     price = 100 + np.cumsum(np.random.normal(0, 0.2, len(minutes)))
10    high = price + np.abs(np.random.normal(0.1, 0.05, len(minutes)))
11    low = price - np.abs(np.random.normal(0.1, 0.05, len(minutes)))
12    open_ = price + np.random.normal(0, 0.1, len(minutes))
13    volume = np.random.randint(100, 1000, len(minutes))
14
15    return pd.DataFrame({
16        "Open": open_,
17        "High": high,
18        "Low": low,
19        "Close": price,
20        "Volume": volume
21    }, index=minutes)
22
23 # Generate 60 trading days (simulate weekdays)
24 days = pd.bdate_range('2024-01-01', periods=60)
25 df_list = [generate_intraday_day(day.strftime("%Y-%m-%d")) for day in days]
26 df_full = pd.concat(df_list)
27 df_full.index.name = "datetime"
28
29 # Feature Engineering
30 df_full['ret1'] = df_full['Close'].pct_change()
31 df_full['ret5'] = df_full['Close'].pct_change(5)
32 df_full['volatility5'] = df_full['Close'].rolling(5).std()
33 df_full['ma10'] = df_full['Close'].rolling(10).mean()
34 df_full['ma10ff'] = df_full['Close'] - df_full['ma10']
35 df_full['hour'] = df_full.index.hour + df_full.index.minute / 60.0 # time of day
36
37 # Multi-Horizon Targets
38 horizons = [5, 10, 20] # in minutes
39 for h in horizons:
40     df_full[f"target_{h}"] = (df_full['Close'].shift(-h) > df_full['Close']).astype(int)
41
42 # Drop rows with NaN
43 df_full.dropna(inplace=True)
44
45 print(df_full.head())
46 print("60-day synthetic OHLCV data with features and targets created.")
47
48 # Prepare output for GRU model (Step 6B will reshape to sequences)
49 df_full.shape, df_full.columns.tolist()

```

```

datetime      Open      High       Low      Close   Volume  W
2024-01-01 09:39:00 100.785763 101.058013 100.879170 100.896122    974
2024-01-01 09:40:00 100.821428 100.823717 100.681958 100.803439    540
2024-01-01 09:41:00 100.849493 100.780324 100.599908 100.710293    632
2024-01-01 09:42:00 100.850517 100.858947 100.645106 100.758685    295
2024-01-01 09:43:00 100.218979 100.478378 100.339867 100.376029    789

      ret1      ret5 volatility5      ma10   ma_diff  W
2024-01-01 09:39:00 0.001077 0.004351  0.196387 100.504253 0.391869
2024-01-01 09:40:00 -0.000919 0.003897  0.069569 100.574662 0.228776
2024-01-01 09:41:00 -0.000924 -0.000176  0.076578 100.638523 0.071770
2024-01-01 09:42:00 0.000481 -0.001217  0.088510 100.694268 0.064417
2024-01-01 09:43:00 -0.003798 -0.004084  0.196288 100.681288 -0.305259

hour target_5 target_10 target_20
2024-01-01 09:39:00 9.650000     0     0     0
2024-01-01 09:40:00 9.666667     0     0     0
2024-01-01 09:41:00 9.683333     0     0     0
2024-01-01 09:42:00 9.700000     0     0     0
2024-01-01 09:43:00 9.716667     0     0     0
60-day synthetic OHLCV data with features and targets created.
((23391, 14),
 ['open',
  'high',
  'low',
  'close',
  'volume',
  'ret1',
  'ret5',
  'volatility5',
  'ma10',
  'ma_diff',
  'hour',
  'target_5',
  'target_10',
  'target_20'])

```

```

1 import subprocess
2 subprocess.run(["pip", "install", "tensorflow"], check=True)
3 CompletedProcess(args=['pip', 'install', 'tensorflow'], returncode=0)

```

## Step 6B : Walk-Forward GRU Meta-Signal Generator with Multi-GPU, Data Leakage

### Protection & F1 Evaluation

```

1 import tensorflow as tf
2 from sklearn.preprocessing import StandardScaler
3 from sklearn.metrics import f1_score
4
5 # Step 6B: GRU Walk-Forward Meta Signal Generator
6
7 features = ['ret1', 'ret5', 'volatility5', 'ma_diff', 'hour']
8 targets = ['target_5', 'target_10', 'target_20']
9 sequence_length = 50
10 train_days = 40
11 test_days = 1
12
13 # Initialize signal columns
14 for target in targets:
15     horizon = target.split('_')[1]
16     df_full[f'meta_signal_{horizon}'] = 0
17
18 day_starts = df_full.index.normalize().drop_duplicates().tolist()
19
20 def make_sequences(df_sub, target_col):
21     X, y = [], []
22     data = df_sub[features + [target_col]].values
23     for j in range(sequence_length, len(data)):
24         X.append(data[j-sequence_length:j, :-1])
25         y.append(data[j, -1])
26     return np.array(X), np.array(y)
27
28 # Multi-GPU Strategy Scope (outside loop to avoid reinitialization)
29 strategy = tf.distribute.MirroredStrategy()
30 print(f"MirroredStrategy Enabled - Devices: {strategy.num_replicas_in_sync}")
31
32 with strategy.scope():
33     for i in range(len(day_starts) - train_days - test_days):
34         train_start = day_starts[i]
35         train_end = day_starts[i + train_days]
36         test_start = day_starts[i + train_days]
37         test_end = day_starts[i + train_days + test_days]
38
39         df_train = df_full.loc[(df_full.index >= train_start) & (df_full.index < train_end)].copy()
40         df_test = df_full.loc[(df_full.index >= test_start) & (df_full.index < test_end)].copy()
41
42         if len(df_train) < sequence_length or len(df_test) < sequence_length:
43             continue
44
45         for target in targets:
46             horizon = target.split('_')[1]
47
48             # Normalize without leakage
49             scaler = StandardScaler()
50             df_train_scaled = df_train.copy()
51             df_test_scaled = df_test.copy()
52             df_train_scaled[features] = scaler.fit_transform(df_train[features])
53             df_test_scaled[features] = scaler.transform(df_test[features])
54
55             X_train, y_train = make_sequences(df_train_scaled, target)
56             X_test, y_test = make_sequences(df_test_scaled, target)
57
58             if len(X_train) < 10 or len(X_test) < 10:
59                 print(f"Walk {i+1}, Horizon {horizon}: Skipped due to insufficient sequence length")
60                 continue
61             if len(set(y_test)) < 2:
62                 print(f"Walk {i+1}, Horizon {horizon}: Skipped due to lack of class diversity in test set")
63                 continue
64
65             model = tf.keras.Sequential([
66                 tf.keras.layers.Input(shape=(sequence_length, len(features))),
67                 tf.keras.layers.GRU(64, return_sequences=False),
68                 tf.keras.layers.Dense(1, activation="sigmoid")
69             ])
70             model.compile(optimizer="adam", loss="binary_crossentropy")
71
72             model.fit(X_train, y_train, epochs=5, batch_size=32, verbose=0)
73
74             y_pred = model.predict(X_test).flatten()
75             y_pred_binary = (y_pred > 0.5).astype(int)
76             f1 = f1_score(y_test, y_pred_binary, zero_division=0)
77
78             pred_index = df_test.loc[sequence_length:, index]
79             df_full.loc[pred_index, f'meta_signal_{horizon}'] = y_pred_binary

```

```
81     print(f"Walk {i+1:02}, Horizon {horizon} min → Injected {len(pred_index)} signals | F1: {f1:.4f}")
82
83 MirroredStrategy Enabled - Devices: 1
84
85 Walk 01, Horizon 5 min → Injected 340 signals | F1: 0.6218
86 11/11                                     0s 20ms/step
87 Walk 01, Horizon 10 min → Injected 340 signals | F1: 0.3944
88 11/11                                     0s 20ms/step
89 Walk 01, Horizon 20 min → Injected 340 signals | F1: 0.5833
90 11/11                                     0s 20ms/step
91 Walk 02, Horizon 5 min → Injected 340 signals | F1: 0.4748
92 11/11                                     0s 19ms/step
93 Walk 02, Horizon 10 min → Injected 340 signals | F1: 0.3624
94 11/11                                     0s 19ms/step
95 Walk 02, Horizon 20 min → Injected 340 signals | F1: 0.3471
96 11/11                                     0s 19ms/step
97 Walk 03, Horizon 5 min → Injected 340 signals | F1: 0.4828
98 11/11                                     0s 18ms/step
99 Walk 03, Horizon 10 min → Injected 340 signals | F1: 0.4452
100 11/11                                     0s 18ms/step
101 Walk 03, Horizon 20 min → Injected 340 signals | F1: 0.3439
102 11/11                                     0s 19ms/step
103 Walk 04, Horizon 5 min → Injected 340 signals | F1: 0.5818
104 11/11                                     0s 19ms/step
105 Walk 04, Horizon 10 min → Injected 340 signals | F1: 0.2602
106 11/11                                     0s 18ms/step
107 Walk 04, Horizon 20 min → Injected 340 signals | F1: 0.5095
108 11/11                                     0s 18ms/step
109 Walk 05, Horizon 5 min → Injected 340 signals | F1: 0.5166
110 11/11                                     0s 18ms/step
111 Walk 05, Horizon 10 min → Injected 340 signals | F1: 0.3310
112 11/11                                     0s 18ms/step
113 Walk 05, Horizon 20 min → Injected 340 signals | F1: 0.5522
114 11/11                                     0s 18ms/step
115 Walk 06, Horizon 5 min → Injected 340 signals | F1: 0.5600
116 11/11                                     0s 19ms/step
117 Walk 06, Horizon 10 min → Injected 340 signals | F1: 0.3735
118 11/11                                     0s 18ms/step
119 Walk 06, Horizon 20 min → Injected 340 signals | F1: 0.4201
120 11/11                                     0s 19ms/step
121 Walk 07, Horizon 5 min → Injected 340 signals | F1: 0.6396
122 11/11                                     0s 18ms/step
123 Walk 07, Horizon 10 min → Injected 340 signals | F1: 0.5491
124 11/11                                     0s 19ms/step
125 Walk 07, Horizon 20 min → Injected 340 signals | F1: 0.6062
126 11/11                                     0s 19ms/step
127 Walk 08, Horizon 5 min → Injected 340 signals | F1: 0.5857
128 11/11                                     0s 19ms/step
129 Walk 08, Horizon 10 min → Injected 340 signals | F1: 0.4688
130 11/11                                     0s 19ms/step
131 Walk 08, Horizon 20 min → Injected 340 signals | F1: 0.5679
132 11/11                                     0s 19ms/step
133 Walk 09, Horizon 5 min → Injected 340 signals | F1: 0.4653
134 11/11                                     0s 18ms/step
135 Walk 09, Horizon 10 min → Injected 340 signals | F1: 0.3962
136 11/11                                     0s 18ms/step
137 Walk 09, Horizon 20 min → Injected 340 signals | F1: 0.5766
138 11/11                                     0s 19ms/step
139 Walk 10, Horizon 5 min → Injected 340 signals | F1: 0.5825
140 11/11                                     0s 18ms/step
```

```
1 import subprocess
2 subprocess.run(["pip", "install", "backtrader"])
3 CompletedProcess(args=['pip', 'install', 'backtrader'], returncode=0)
```

## ▼ Step 6C : Backtrader Integration

```

1 # Install dependencies (Google Colab only)
2
3 import backtrader as bt
4 import pandas as pd
5 import numpy as np
6 import matplotlib.pyplot as plt
7 import os
8
9 # Generate synthetic OHLCV data
10 np.random.seed(42)
11 n = 500
12 dates = pd.date_range(start="2024-01-01", periods=n, freq="min")
13
14 price = 100 + np.cumsum(np.random.randn(n))
15 open_ = price + np.random.normal(0, 0.5, size=n)
16 high = np.maximum(open_, price) + np.abs(np.random.normal(0, 0.3, size=n))
17 low = np.minimum(open_, price) - np.abs(np.random.normal(0, 0.3, size=n))
18 volume = np.random.randint(100, 1000, size=n)
19
20 df_full = pd.DataFrame({
21     "Open": open_,
22     "High": high,
23     "Low": low,
24     "Close": price,
25     "Volume": volume
26 }, index=dates)
27
28 # Generate technical indicator-based signals
29 def generate_technical_signals(df):
30     df['SMA_10'] = df['Close'].rolling(window=10).mean()
31     df['SMA_30'] = df['Close'].rolling(window=30).mean()
32     df['MACD'] = df['Close'].ewm(span=12).mean() - df['Close'].ewm(span=26).mean()
33     df['Signal_Line'] = df['MACD'].ewm(span=9).mean()
34     df['RSI'] = 100 - 100 / (1 + df['Close'].diff().apply(lambda x: max(x, 0)).rolling(14).mean() /
35                               df['Close'].diff().apply(lambda x: min(x, 0)).rolling(14).mean())
36
37     df['meta_signal_5'] = ((df['SMA_10'] > df['SMA_30']) & (df['MACD'] > df['Signal_Line'])).astype(int)
38     df['meta_signal_10'] = (df['RSI'] < 30).astype(int)
39     df['meta_signal_20'] = ((df['RSI'] > 50) & (df['MACD'] > df['Signal_Line'])).astype(int)
40     return df
41
42 df_full = generate_technical_signals(df_full).dropna()
43
44 # Backtrader Data Feed
45 class PandasMetaSignalData(bt.feeds.PandasData):
46     lines = ('meta_signal_5', 'meta_signal_10', 'meta_signal_20')
47     params = (('meta_signal_5', -1), ('meta_signal_10', -1), ('meta_signal_20', -1))
48
49 # Strategy
50 class EnsembleMetaSignalStrategy(bt.Strategy):
51     params = dict(exitbars=5, stop_loss_pct=0.015)
52
53     def __init__(self):
54         self.signal_5 = self.datas[0].meta_signal_5
55         self.signal_10 = self.datas[0].meta_signal_10
56         self.signal_20 = self.datas[0].meta_signal_20
57         self.dataclose = self.datas[0].close
58         self.order = None
59         self.buyprice = None
60         self.bar_executed = None
61         self.peak = self.broker.getvalue()
62         self.max_drawdown = 0.0

```

```

63     self.equity_curve = []
64     self.trades_log = []
65
66     def notify_order(self, order):
67         if order.status in [order.Completed, order.Canceled, order.Margin]:
68             self.order = None
69
70     def next(self):
71         dt = self.datas[0].datetime.datetime(0)
72         equity = self.broker.getvalue()
73         self.peak = max(self.peak, equity)
74         drawdown = (self.peak - equity) / self.peak
75         self.max_drawdown = max(self.max_drawdown, drawdown)
76         self.equity_curve.append((dt, equity))
77
78     signal_ensemble = sum([self.signal_5[0], self.signal_10[0], self.signal_20[0]])
79
80     if self.order:
81         return
82
83     if not self.position and signal_ensemble >= 2:
84         size = max(1, int(equity * 0.01 / (self.dataclose[0] + 1e-6)))
85         self.order = self.buy(size=size)
86         self.buyprice = self.dataclose[0]
87         self.bar_executed = len(self)
88         self.trades_log.append([dt, 'BUY', self.buyprice, size])
89
90     elif self.position:
91         price = self.dataclose[0]
92         if self.bar_executed is not None and len(self) >= self.bar_executed + self.params.exitbars:
93             self.order = self.sell()
94             self.trades_log.append([dt, 'SELL (exitbars)', price, self.position.size])
95         elif price < self.buyprice * (1 - self.params.stop_loss_pct):
96             self.order = self.sell()
97             self.trades_log.append([dt, 'SELL (stoploss)', price, self.position.size])
98
99     def stop(self):
100        print(f"Max Drawdown: {self.max_drawdown * 100:.2f}%")
101
102 # Execute Backtest
103 def run_backtest_ensemble(bt, tag="improved"):
104     cerebro = bt.Cerebro()
105     cerebro.addstrategy(EnsembleMetaSignalStrategy)
106     data = PandasMetaSignalData(dataname=df)
107     cerebro.adddata(data)
108     cerebro.broker.setcash(100000.0)
109     cerebro.broker.setcommission(commission=0.001)
110     cerebro.addanalyzer(bt.analyzers.SharpeRatio, _name='sharpe')
111     cerebro.addanalyzer(bt.analyzers.DrawDown, _name='dd')
112
113     print(f"Starting Portfolio Value: {cerebro.broker.getvalue():.2f}")
114     results = cerebro.run()
115     strat = results[0]
116     print(f"Final Portfolio Value: {cerebro.broker.getvalue():.2f}")
117     print("Sharpe Ratio:", strat.analyzers.sharpe.get_analysis())
118     print("Max Drawdown:", strat.analyzers.dd.get_analysis())
119
120     # Save trades and equity
121     trades_df = pd.DataFrame(strat.trades_log, columns=["Timestamp", "Action", "Price", "Size"])
122     equity_df = pd.DataFrame(strat.equity_curve, columns=["Timestamp", "Equity"])
123     trades_df.to_csv(f'gru_backtest_trades_{tag}.csv', index=False)
124     equity_df.to_csv(f'gru_backtest_equity_{tag}.csv', index=False)
125     print(f'CSVs saved: gru_backtest_trades_{tag}.csv, gru_backtest_equity_{tag}.csv')
126
127     # Plot
128     fig, ax1 = plt.subplots(figsize=(12, 6))
129     ax1.plot(equity_df['Timestamp'], equity_df['Equity'], label="Equity Curve", linewidth=2, color='blue')
130     ax1.set_ylabel("Portfolio Value", color='blue')
131     ax1.tick_params(axis='y', labelcolor='blue')
132
133     ax2 = ax1.twinx()
134     ax2.set_ylabel("Trade Price", color='gray')
135     buy = trades_df[trades_df['Action'].str.contains("BUY")]
136     sell = trades_df[trades_df['Action'].str.contains("SELL")]
137     ax2.scatter(buy['Timestamp'], buy['Price'], color="green", label="BUY", marker="^")
138     ax2.scatter(sell['Timestamp'], sell['Price'], color="red", label="SELL", marker="v")
139     ax2.tick_params(axis='y', labelcolor='gray')
140
141     fig.tight_layout()
142     fig.suptitle('Equity Curve with Trade Markers (Improved Strategy)', fontsize=14)
143     fig.subplots_adjust(top=0.93)
144     ax1.legend(loc='upper left')
145     plt.grid(True)
146     plt.savefig("equity_curve_improved.png", dpi=300)
147     plt.show()
148     print("Plot saved: equity_curve_improved.png")
149
150     # Summary
151     print("Trade Performance Summary:")
152     completed_trades = []
153     position = None
154     for _, row in trades_df.iterrows():
155         if 'BUY' in row['Action']:
156             position = {'entry_price': row['Price'], 'entry_time': row['Timestamp'], 'size': row['Size']}
157         elif 'SELL' in row['Action'] and position:
158             pnl = (row['Price'] - position['entry_price']) * position['size']
159             ret_pct = (row['Price'] / position['entry_price'] - 1) * 100
160             holding_period = (pd.to_datetime(row['Timestamp']) - pd.to_datetime(position['entry_time'])).total_seconds() / 60
161             completed_trades.append({
162                 'entry_time': position['entry_time'],
163                 'exit_time': row['Timestamp'],
164                 'entry_price': position['entry_price'],
165                 'exit_price': row['Price'],
166                 'return_pct': ret_pct,
167                 'holding_minutes': holding_period,
168                 'pnl': pnl
169             })
170             position = None
171
172     if completed_trades:
173         perf_df = pd.DataFrame(completed_trades)
174         print(f"Total Trades: {len(perf_df)}")
175         print(f"Win Rate: {(perf_df['return_pct'] > 0).mean() * 100:.2f}%")
176         print(f"Avg Trade Return: {perf_df['return_pct'].mean():.2f}%")
177         print(f"Avg Holding Time: {perf_df['holding_minutes'].mean():.2f} min")
178         print(f"Total Strategy PnL: {perf_df['pnl'].sum():.2f}")
179     else:
180         print("No complete trades were executed.")
181
182 # Run the backtest
183 bt_df = df_full[['Open', 'High', 'Low', 'Close', 'Volume',
184                  'meta_signal_5', 'meta_signal_10', 'meta_signal_20']].copy()
185
186 run_backtest_ensemble(bt_df, tag="improved")
187 Starting Portfolio Value: 100000.00
188 Max Drawdown: 0.11%
189 Final Portfolio Value: 99983.56
190 Sharpe Ratio: OrderedDict([('sharperatio', None)])
191 Max Drawdown: AutoOrderedDict([('sharperatio', 0.068538462087622), ('moneydown', 68.57419508043677), ('max', AutoOrderedDict([('len', 300), ('drawdown', 0.1107919151815522), ('moneydown', 110.80641322045994)))]))

```

```
CSVs saved: gru_backtest_trades_improved.csv, gru_backtest_equity_improved.csv
Plot saved: equity_curve_improved.png
Trade Performance Summary:
Total Trades: 16
Win Rate: 62.50%
Avg Trade Return: 0.71%
Avg Holding Time: 4.44 min
Total Strategy PnL: 107.36
```

```
1 import subprocess
2
3 subprocess.call([
4     "pip", "install",
5     "yfinance", "backtrader", "optuna", "matplotlib"
6 ])
```

0

## Step 6D : Full Backtest & Optimization Pipeline with Optuna + Final Equity Curve

### Plotting

```
1 # Install dependencies if not already installed
2 import subprocess
3 subprocess.call(["pip", "install", "-q", "backtrader", "optuna", "matplotlib", "pandas", "numpy"])
4
5 # Imports
6 import backtrader as bt
7 import pandas as pd
8 import numpy as np
9 import matplotlib.pyplot as plt
10 import optuna
11
12 # Generate synthetic OHLCV data
13 np.random.seed(42)
14 n = 1000 # ~4 years of daily data
15 dates = pd.date_range("2020-01-01", periods=n, freq="D")
16 price = 150 + np.cumsum(np.random.randn(n)) # random walk
17
18 data = pd.DataFrame({
19     'Open': price + np.random.normal(0, 1, n),
20     'High': price + np.random.normal(1, 1, n),
21     'Low': price - np.random.normal(1, 1, n),
22     'Close': price,
23     'Volume': np.random.randint(1e6, 1e7, n)
24 }, index=dates)
25
26 # Add technical indicators
27 def add_indicators(df):
28     df['SMA_10'] = df['Close'].rolling(window=10).mean()
29     df['SMA_30'] = df['Close'].rolling(window=30).mean()
30     df['MACD'] = df['Close'].ewm(span=12).mean() - df['Close'].ewm(span=26).mean()
31     df['Signal_Line'] = df['MACD'].ewm(span=9).mean()
32     df['RSI'] = 100 - 100 / (1 + df['Close'].diff().apply(lambda x: max(x, 0)).rolling(14).mean() /
33                             df['Close'].diff().apply(lambda x: min(x, 0)).rolling(14).mean())
34     return df.dropna()
35
36 data = add_indicators(data)
37
38 # Custom backtrader-compatible feed
39 class PandasData(bt.feeds.PandasData):
40     params = {
41         ('datetime', None),
42         ('open', 'Open'),
43         ('high', 'High'),
44         ('low', 'Low'),
45         ('close', 'Close'),
46         ('volume', 'Volume'),
47         ('openinterest', -1),
48     }
49
50 # Define optimization objective
51 def objective(trial):
52     exitbars = trial.suggest_int('exitbars', 3, 15)
53     stop_loss_pct = trial.suggest_float('stop_loss_pct', 0.005, 0.03)
54     profit_target_pct = trial.suggest_float('profit_target_pct', 0.01, 0.05)
55
56     class OptimizedStrategy(bt.Strategy):
57         params = dict(
58             exitbars=exitbars,
59             stop_loss_pct=stop_loss_pct,
60             profit_target_pct=profit_target_pct
61         )
62
63         def __init__(self):
64             self.order = None
65             self.buy_price = None
66             self.bar_executed = None
67
68         def notify_order(self, order):
69             if order.status in [order.Completed, order.Canceled, order.Margin]:
70                 self.order = None
71
72         def next(self):
73             if self.order:
74                 return
75             if not self.position:
76                 if self.data.close[0] > self.data.open[0]: # simple condition
77                     size = int(self.broker.getcash() / self.data.close[0])
78                     self.order = self.buy(size)
79                     self.buy_price = self.data.close[0]
80                     self.bar_executed = len(self)
81             else:
82                 price = self.data.close[0]
83                 if price >= self.buy_price * (1 + self.params.profit_target_pct):
84                     self.order = self.sell()
85                 elif price <= self.buy_price * (1 - self.params.stop_loss_pct):
86                     self.order = self.sell()
87                 elif len(self) >= self.bar_executed + self.params.exitbars:
88                     self.order = self.sell()
89
90         cerebro = bt.Cerebro()
91         cerebro.addstrategy(OptimizedStrategy)
92         cerebro.adddata(PandasData(dataname=data))
93         cerebro.broker.setcash(100000.0)
94         cerebro.broker.setcommission(commission=0.001)
95         cerebro.addanalyzer(bt.analyzers.SharpeRatio, _name='sharpe')
96         results = cerebro.run()
97         sharpe = results[0].analyzers.sharpe.get_analysis().get('sharperatio', 0)
98         return sharpe if sharpe is not None else 0
99
100 # Run optimization
101 study = optuna.create_study(direction='maximize')
102 study.optimize(objective, n_trials=30)
103 best_params = study.best_trial.params
```

```

109 # Replace this if needed with my Optuna output
108 print("Using best_params from Optuna: ", best_params)
107
108 # Final Strategy Class (fixed parameter access and logic)
109 class FinalStrategy(bt.Strategy):
110     params = dict(
111         exitbars=best_params['exitbars'],
112         stop_loss_pct=best_params['stop_loss_pct'],
113         profit_target_pct=best_params['profit_target_pct'],
114     )
115
116     def __init__(self):
117         self.order = None
118         self.buy_price = None
119         self.bar_executed = 0
120         self.equity_curve = []
121
122     def notify_order(self, order):
123         if order.status in [order.Completed, order.Canceled, order.Margin]:
124             self.order = None
125
126     def next(self):
127         dt = self.datas[0].datetime.datetime(0)
128         self.equity_curve.append(dt, self.broker.getvalue())
129
130         if self.order:
131             return
132
133         # Entry logic (simple condition: bullish candle)
134         if not self.position:
135             if self.data.close[0] > self.data.open[0]:
136                 size = int(self.broker.getcash() / self.data.close[0])
137                 self.order = self.buy(size=size)
138                 self.buy_price = self.data.close[0]
139                 self.bar_executed = len(self)
140                 print(f"BUY @ {self.buy_price:.2f} on {dt}")
141
142         else:
143             price = self.data.close[0]
144             holding = len(self) - self.bar_executed
145
146             # Exit Conditions
147             if price >= self.buy_price * (1 + self.params.profit_target_pct):
148                 self.order = self.sell()
149                 print(f"SELL (profit) @ {price:.2f} on {dt}")
150             elif price <= self.buy_price * (1 - self.params.stop_loss_pct):
151                 self.order = self.sell()
152                 print(f"SELL (stop-loss) @ {price:.2f} on {dt}")
153             elif holding >= self.params.exitbars:
154                 self.order = self.sell()
155                 print(f"SELL (timeout) @ {price:.2f} on {dt}")
156
157 cerebro = bt.Cerebro()
158 cerebro.addstrategy(FinalStrategy)
159 cerebro.adddata(PandasData(dataname=data))
160 cerebro.broker.setcash(100000.0)
161 cerebro.broker.setcommission(commission=0.001)
162 results = cerebro.run()
163
164 # Extract Equity Curve
165 final_strat = results[0]
166 equity_df = pd.DataFrame(final_strat.equity_curve, columns=["Timestamp", "Equity"])
167
168 # Plot Equity Curve
169 # Imports at the top
170 %matplotlib inline
171 import matplotlib.pyplot as plt
172 from IPython.display import display, Image
173
174 # Plot equity curve
175 plt.figure(figsize=(12, 6))
176 plt.plot(equity_df["Timestamp"], equity_df["Equity"], color='blue')
177 plt.title(f"Optimized Strategy Equity Curve (Synthetic Data)\nParams: {best_params}", fontsize=14)
178 plt.xlabel("Date")
179 plt.ylabel("Portfolio Value")
180 plt.grid(True)
181 plt.legend()
182 plt.tight_layout()
183
184 # Save and display
185 plt.savefig("equity_plot.png", dpi=300)
186 display(Image(filename="equity_plot.png"))

```

5 [i 2025-05-31 00:57:53.843] A new study created in memory with name: no-name-93c7af18-18d2-4030-a7d2-a6a941d1737b

[i 2025-05-31 00:57:54.241] Trial 0 finished with value: 0.7183863364835763 and parameters: {'exitbars': 8, 'stop\_loss\_pct': 0.01990178029206695, 'profit\_target\_pct': 0.027523451852811284}. Best is trial 0 with value: 0.7183863364835763

[i 2025-05-31 00:57:54.628] Trial 1 finished with value: 0.7160235831835884 and parameters: {'exitbars': 13, 'stop\_loss\_pct': 0.022517597790310102, 'profit\_target\_pct': 0.018819570094647235}. Best is trial 0 with value: 0.7183863364835763

[i 2025-05-31 00:57:55.406] Trial 2 finished with value: 0.7184626311121874 and parameters: {'exitbars': 7, 'stop\_loss\_pct': 0.02221575625164945, 'profit\_target\_pct': 0.0107901526060634025}. Best is trial 2 with value: 0.7184626311121874

[i 2025-05-31 00:57:55.406] Trial 3 finished with value: 0.7184626311121874 and parameters: {'exitbars': 7, 'stop\_loss\_pct': 0.020300534997627913, 'profit\_target\_pct': 0.013635339992212428}. Best is trial 2 with value: 0.7184626311121874

[i 2025-05-31 00:57:55.784] Trial 4 finished with value: 0.7164315518180174 and parameters: {'exitbars': 13, 'stop\_loss\_pct': 0.019474166012451858, 'profit\_target\_pct': 0.010306734673944322}. Best is trial 2 with value: 0.7184626311121874

[i 2025-05-31 00:57:56.178] Trial 5 finished with value: 0.7184626311121874 and parameters: {'exitbars': 7, 'stop\_loss\_pct': 0.0210734751200949, 'profit\_target\_pct': 0.01085792737960038}. Best is trial 2 with value: 0.7184626311121874

[i 2025-05-31 00:57:56.564] Trial 6 finished with value: 0.7187846627763879 and parameters: {'exitbars': 6, 'stop\_loss\_pct': 0.01697165527615027, 'profit\_target\_pct': 0.031793889856015785}. Best is trial 6 with value: 0.7187846627763879

[i 2025-05-31 00:57:56.959] Trial 7 finished with value: 0.7185047930895703 and parameters: {'exitbars': 13, 'stop\_loss\_pct': 0.005272906523258574, 'profit\_target\_pct': 0.0392742207591792}. Best is trial 6 with value: 0.7187846627763879

[i 2025-05-31 00:57:57.339] Trial 8 finished with value: 0.7171421150392858 and parameters: {'exitbars': 9, 'stop\_loss\_pct': 0.02821490520732, 'profit\_target\_pct': 0.018468690852250037}. Best is trial 6 with value: 0.7187846627763879

[i 2025-05-31 00:57:57.717] Trial 9 finished with value: 0.7191198868957382 and parameters: {'exitbars': 5, 'stop\_loss\_pct': 0.02077242057462074, 'profit\_target\_pct': 0.046603914518660316}. Best is trial 9 with value: 0.7191198868957382

[i 2025-05-31 00:57:58.199] Trial 10 finished with value: 0.7195650693902706 and parameters: {'exitbars': 3, 'stop\_loss\_pct': 0.013005304993936147, 'profit\_target\_pct': 0.04954769856922665}. Best is trial 10 with value: 0.7195650693902706

[i 2025-05-31 00:57:58.530] Trial 11 finished with value: 0.7195650693902706 and parameters: {'exitbars': 3, 'stop\_loss\_pct': 0.01217738763518875, 'profit\_target\_pct': 0.04984415853489758}. Best is trial 10 with value: 0.7195650693902706

[i 2025-05-31 00:57:58.928] Trial 12 finished with value: 0.7195650693902706 and parameters: {'exitbars': 3, 'stop\_loss\_pct': 0.01603668994895974, 'profit\_target\_pct': 0.04940767234188079}. Best is trial 10 with value: 0.7195650693902706

[i 2025-05-31 00:58:00.185] Trial 13 finished with value: 0.7195650693902706 and parameters: {'exitbars': 3, 'stop\_loss\_pct': 0.0126236895183799, 'profit\_target\_pct': 0.04166681973271007}. Best is trial 10 with value: 0.7195650693902706

[i 2025-05-31 00:58:00.589] Trial 14 finished with value: 0.7184457405270932 and parameters: {'exitbars': 10, 'stop\_loss\_pct': 0.0126236895183799, 'profit\_target\_pct': 0.042827177161890467}. Best is trial 10 with value: 0.7195650693902706

[i 2025-05-31 00:58:01.097] Trial 15 finished with value: 0.7192186568588114 and parameters: {'exitbars': 5, 'stop\_loss\_pct': 0.01111803207247948, 'profit\_target\_pct': 0.042827177161890467}. Best is trial 10 with value: 0.7195650693902706

[i 2025-05-31 00:58:01.415] Trial 16 finished with value: 0.7170792025966586 and parameters: {'exitbars': 11, 'stop\_loss\_pct': 0.015144846378181826, 'profit\_target\_pct': 0.04799852713170703}. Best is trial 10 with value: 0.7195650693902706

[i 2025-05-31 00:58:01.415] Trial 17 finished with value: 0.7185251286692748 and parameters: {'exitbars': 15, 'stop\_loss\_pct': 0.00851176981634933, 'profit\_target\_pct': 0.03324950051570309}. Best is trial 10 with value: 0.7195650693902706

[i 2025-05-31 00:58:02.214] Trial 18 finished with value: 0.7192186568588114 and parameters: {'exitbars': 4, 'stop\_loss\_pct': 0.014535218665780688, 'profit\_target\_pct': 0.0444802435464106}. Best is trial 10 with value: 0.7195650693902706

[i 2025-05-31 00:58:02.605] Trial 19 finished with value: 0.71910220087717418 and parameters: {'exitbars': 5, 'stop\_loss\_pct': 0.029019593924856554, 'profit\_target\_pct': 0.02753582730716127}. Best is trial 10 with value: 0.7195650693902706

[i 2025-05-31 00:58:03.014] Trial 20 finished with value: 0.7195650693902706 and parameters: {'exitbars': 3, 'stop\_loss\_pct': 0.0087504567366195, 'profit\_target\_pct': 0.0497196992767394}. Best is trial 10 with value: 0.7195650693902706

[i 2025-05-31 00:58:03.047] Trial 21 finished with value: 0.7195650693902706 and parameters: {'exitbars': 3, 'stop\_loss\_pct': 0.01193218450114334, 'profit\_target\_pct': 0.04602434192027578}. Best is trial 10 with value: 0.7195650693902706

[i 2025-05-31 00:58:03.801] Trial 22 finished with value: 0.719119896857382 and parameters: {'exitbars': 5, 'stop\_loss\_pct': 0.014668419452826249, 'profit\_target\_pct': 0.03899533336030156}. Best is trial 10 with value: 0.7195650693902706

[i 2025-05-31 00:58:04.209] Trial 23 finished with value: 0.7192186568588114 and parameters: {'exitbars': 4, 'stop\_loss\_pct': 0.00708942397357212, 'profit\_target\_pct': 0.036585859578951784}. Best is trial 10 with value: 0.7195650693902706

[i 2025-05-31 00:58:04.602] Trial 24 finished with value: 0.7195650693902706 and parameters: {'exitbars': 3, 'stop\_loss\_pct': 0.0131705404209269166, 'profit\_target\_pct': 0.049380721012249277}. Best is trial 10 with value: 0.7195650693902706

[i 2025-05-31 00:58:04.992] Trial 25 finished with value: 0.7187846627763879 and parameters: {'exitbars': 6, 'stop\_loss\_pct': 0.0176893265485082, 'profit\_target\_pct': 0.0466003832371549}. Best is trial 10 with value: 0.7195650693902706

[i 2025-05-31 00:58:05.387] Trial 26 finished with value: 0.7192393942771926 and parameters: {'exitbars': 4, 'stop\_loss\_pct': 0.01665933262613993, 'profit\_target\_pct': 0.040709055703985}. Best is trial 10 with value: 0.7195650693902706

[i 2025-05-31 00:58:05.779] Trial 27 finished with value: 0.7187846627763879 and parameters: {'exitbars': 6, 'stop\_loss\_pct': 0.01078903947529104, 'profit\_target\_pct': 0.0365389255132615}. Best is trial 10 with value: 0.7195650693902706

[i 2025-05-31 00:58:07.048] Trial 28 finished with value: 0.719119896857382 and parameters: {'exitbars': 5, 'stop\_loss\_pct': 0.00712500490372898, 'profit\_target\_pct': 0.04886021900509146}. Best is trial 10 with value: 0.7195650693902706

[i 2025-05-31 00:58:07.434] Trial 29 finished with value: 0.718469303100108 and parameters: {'exitbars': 9, 'stop\_loss\_pct': 0.013066028162536096, 'profit\_target\_pct': 0.0231624834753576}. Best is trial 10 with value: 0.7195650693902706

Using best\_params from Optuna: {'exitbars': 3, 'stop\_loss\_pct': 0.013005304909396147, 'profit\_target\_pct': 0.04954769856922665}

BUY @ 144.36 on 2020-01-30 00:00:00

SELL (timeout) @ 145.59 on 2020-02-02 00:00:00

SELL (timeout) @ 144.53 on 2020-02-03 00:00:00

SELL (timeout) @ 145.36 on 2020-02-04 00:00:00

SELL (timeout) @ 144.14 on 2020-02-05 00:00:00

SELL (timeout) @ 144.35 on 2020-02-06 00:00:00

SELL (stop-loss) @ 142.39 on 2020-02-07 00:00:00

SELL (stop-loss) @ 141.06 on 2020-02-08 00:00:00

SELL (stop-loss) @ 141.25 on 2020-02-09 00:00:00

SELL (stop-loss) @ 141.99 on 2020-02-10 00:00:00

SELL (stop-loss) @ 142.16 on 2020-02-11 00:00:00

SELL (stop-loss) @ 142.05 on 2020-02-12 00:00:00

SELL (stop-loss) @ 141.75 on 2020-02-13 00:00:00

SELL (stop-loss) @ 140.27 on 2020-02-14 00:00:00

SELL (stop-loss) @ 139.55 on 2020-02-15 00:00:00

SELL (stop-loss) @ 139.09 on 2020-02-16 00:00:00

SELL (stop-loss) @ 140.15 on 2020-02-17 00:00:00

SELL (stop-loss) @ 140.49 on 2020-02-18 00:00:00

```
SELL (stop-loss) @ 138.73 on 2020-02-19 00:00:00
SELL (stop-loss) @ 139.05 on 2020-02-20 00:00:00
SELL (stop-loss) @ 138.67 on 2020-02-21 00:00:00
SELL (stop-loss) @ 137.99 on 2020-02-22 00:00:00
SELL (stop-loss) @ 138.60 on 2020-02-23 00:00:00
SELL (stop-loss) @ 139.63 on 2020-02-24 00:00:00
```

#### Step 6E : Regime Tagging, Performance Visualization & Walk-Forward Preparation

```
1 import pandas as pd
2 import numpy as np
3 import matplotlib.pyplot as plt
4 from hmmlearn.hmm import GaussianHMM
5
6 # Reuse the existing df_full
7 # Ensure df_full has columns: Open, High, Low, Close, Volume, meta_signal_5/10/20
8
9 # Simulate portfolio value (could replace this with real backtest output)
10 df_full['portfolio_value'] = 100000 + np.cumsum(np.random.randn(len(df_full)) * 50)
11 df_full['datetime'] = df_full.index
12
13 # Simulated trades DataFrame (replace with real trades if available)
14 trades_df = pd.DataFrame({
15     "datetime": df_full['datetime'][:50],
16     "signal": np.random.choice(["buy", "sell"], size=len(df_full[:50])),
17     "price": df_full['Close'][:50].values,
18     "profit": np.random.uniform(-100, 100, size=len(df_full[:50]))
19 })
20
21 # Plotting Functions
22 def plot_equity_curve(equity_df):
23     plt.figure(figsize=(12, 5))
24     plt.plot(equity_df['datetime'], equity_df['portfolio_value'], label='Equity Curve')
25     plt.title('Equity Curve')
26     plt.xlabel('Time')
27     plt.ylabel('Portfolio Value')
28     plt.grid(True)
29     plt.legend()
30     plt.tight_layout()
31     plt.show()
32
33 def plot_drawdown(equity_df):
34     peak = equity_df['portfolio_value'].cummax()
35     drawdown = (equity_df['portfolio_value'] - peak) / peak
36     plt.figure(figsize=(12, 5))
37     plt.plot(equity_df['datetime'], drawdown, color='red', label='Drawdown')
38     plt.title('Drawdown Curve')
39     plt.xlabel('Time')
40     plt.ylabel('Drawdown %')
41     plt.grid(True)
42     plt.legend()
43     plt.tight_layout()
44     plt.show()
45
46 # Save CSVs
47 def save_backtest_results(trades_df, equity_df, tag="ensemble"):
48     trades_df.to_csv(f"optimized_trades_{tag}.csv", index=False)
49     equity_df.to_csv(f"optimized_equity_curve_{tag}.csv", index=False)
50     print(f"Saved: optimized_trades_{tag}.csv & optimized_equity_curve_{tag}.csv")
51
52 # HMM Market Regime Tagging
53 def tag_market_regimes(df, n_states=3):
54     log_returns = np.log(df['Close'] / df['Close'].shift(1)).dropna().values.reshape(-1, 1)
55     hmm = GaussianHMM(n_components=n_states, covariance_type="full", n_iter=1000)
56     df = df.copy()
57     df['regime'] = np.nan
58     hmm.fit(log_returns)
59     hidden_states = hmm.predict(log_returns)
60     df.loc[df.index[1:], 'regime'] = hidden_states
61     print(f'Regimes tagged (0 to {n_states - 1})')
62     return df
63
64 # Walk-Forward Split
65 def walk_forward_split(df, train_size=0.6, val_size=0.2):
66     total = len(df)
67     train_end = int(total * train_size)
68     val_end = train_end + int((total * val_size))
69     train_data = df.iloc[:train_end]
70     val_data = df.iloc[train_end:val_end]
71     test_data = df.iloc[val_end:]
72     print(f'Walk-Forward Splits: Train={len(train_data)}, Val={len(val_data)}, Test={len(test_data)}')
73     return train_data, val_data, test_data
74
75 # RUN EVERYTHING
76 # Plot Equity & Drawdown
77 plot_equity_curve(df_full)
78 plot_drawdown(df_full)
79
80 # Save Backtest Results
81 save_backtest_results(trades_df, df_full)
82
83 # Tag Market Regimes
84 df_with_regimes = tag_market_regimes(df_full)
85
86 # Walk-Forward Split
87 train_df, val_df, test_df = walk_forward_split(df_with_regimes)

88 Figure(1200,500)
Figure(1200,500)
Saved: optimized_trades_ensemble.csv & optimized_equity_curve_ensemble.csv
Regimes tagged (0 to 2)
Walk-Forward Splits: Train=282, Val=94, Test=95
```

#### Step 6F : Regime-Based Performance Analysis & Walk-Forward Backtest Execution

```
1 import pandas as pd
2 import numpy as np
3 import matplotlib.pyplot as plt
4
5 # Create a simulated df_with_regimes for demonstration (normally using real data)
6 np.random.seed(42)
7 n = 300
8 dates = pd.date_range("2022-01-01", periods=n)
9 portfolio_value = 100000 + np.cumsum(np.random.randn(n) * 100)
10 regimes = np.random.choice([0, 1, 2], size=n)
11
12 df_with_regimes = pd.DataFrame({
13     "datetime": dates,
14     "portfolio_value": portfolio_value,
15     "regime": regimes
16 })
17
18 # Regime-Based Performance Metrics
```

```

19 def analyze_regime_performance(df_with_regimes):
20     df_with_regimes['daily_return'] = df_with_regimes['portfolio_value'].pct_change()
21     grouped = df_with_regimes.groupby('regime')['daily_return'].agg(['mean', 'std', 'count'])
22     grouped['sharpe_ratio'] = grouped['mean'] / grouped['std'] * np.sqrt(252) # annualized
23     print("Regime-Based Performance (Annualized Sharpe):")
24     print(grouped)
25     return grouped
26
27 # Sharpe ratio bar plot
28 def plot_sharpe_by_regime(summary_df):
29     plt.figure(figsize=(8, 4))
30     plt.bar(summary_df.index.astype(str), summary_df['sharpe_ratio'], color='purple')
31     plt.title("Sharpe Ratio by Regime")
32     plt.xlabel("Regime")
33     plt.ylabel("Sharpe Ratio")
34     plt.grid(True)
35     plt.tight_layout()
36     plt.show()
37
38 # Simulated Walk-Forward Backtest
39 def run_walk_forward_backtest(full_df, window=60, stride=30):
40     results = []
41     for start in range(0, len(full_df) - window, stride):
42         train_df = full_df.iloc[start:start + int(window * 0.6)]
43         test_df = full_df.iloc[start + int(window * 0.6):start + window]
44         if len(test_df) == 0: continue
45         train_return = (train_df['portfolio_value'].iloc[-1] / train_df['portfolio_value'].iloc[0]) - 1
46         test_return = (test_df['portfolio_value'].iloc[-1] / test_df['portfolio_value'].iloc[0]) - 1
47         results.append((train_df['datetime'].iloc[0], train_return, test_return))
48     return pd.DataFrame(results, columns=['StartDate', 'TrainReturn', 'TestReturn'])
49
50 # RUN ALL
51 regime_summary = analyze_regime_performance(df_with_regimes)
52 plot_sharpe_by_regime(regime_summary)
53 walkforward_results = run_walk_forward_backtest(df_with_regimes)
54
55 from IPython.display import display
56 display(walkforward_results)
57 regime_summary

```

**Regime-Based Performance (Annualized Sharpe):**

regime	mean	std	count	sharpe_ratio
0	-0.000019	0.000901	118	-0.339227
1	0.000155	0.001014	91	2.420163
2	-0.000153	0.001063	90	-2.291269

Figure(800x400)

regime	StartDate	TrainReturn	TestReturn
0	2022-01-01	-0.006357	-0.003645
1	2022-01-31	-0.003856	0.001270
2	2022-03-02	-0.000963	0.000926
3	2022-04-01	0.001973	-0.004554
4	2022-05-01	-0.003946	0.006844
5	2022-05-31	0.007879	0.001831
6	2022-06-30	0.005115	-0.001540
7	2022-07-30	-0.000201	-0.002763

regime	mean	std	count	sharpe_ratio
0	-0.000019	0.000901	118	-0.339227
1	0.000155	0.001014	91	2.420163
2	-0.000153	0.001063	90	-2.291269

## Step 6G : Full Python Pipeline for Market Regime Detection using Gaussian HMM +

### Equity Performance Analysis with Synthetic Data

```

1 import pandas as pd
2 import numpy as np
3 import matplotlib.pyplot as plt
4 from hmmlearn.hmm import GaussianHMM
5 import warnings
6 warnings.filterwarnings("ignore")
7
8 # Utility: Plot equity curve
9 def plot_equity_curve(equity_df):
10    df = equity_df.reset_index(drop=True)
11    plt.figure(figsize=(12, 5))
12    plt.plot(df['datetime'], df['portfolio_value'], label='Equity Curve')
13    plt.title('Equity Curve')
14    plt.xlabel('Time')
15    plt.ylabel('Portfolio Value')
16    plt.grid(True)
17    plt.legend()
18    plt.tight_layout()
19    plt.show()
20
21 # Utility: Plot drawdown
22 def plot_drawdown(equity_df):
23    df = equity_df.reset_index(drop=True)
24    peak = df['portfolio_value'].cummax()
25    drawdown = (df['portfolio_value'] - peak) / peak
26    plt.figure(figsize=(12, 5))
27    plt.plot(df['datetime'], drawdown, color='red', label='Drawdown')
28    plt.title('Drawdown Curve')
29    plt.xlabel('Time')
30    plt.ylabel('Drawdown %')
31    plt.grid(True)
32    plt.legend()
33    plt.tight_layout()
34    plt.show()
35
36 # Regime tagging
37 def tag_market_regimes(df, n_states=3):
38    df = df.copy()
39    df['log_ret'] = np.log(df['Close']) / df['Close'].shift(1)
40    df['log_ret'] = df['log_ret'].replace([np.inf, -np.inf], np.nan)
41    df = df.dropna(subset=['log_ret'])
42
43    # Additional sanity filter (optional)
44    df = df[np.abs(df['log_ret']) < 1.0]
45
46    if len(df) < n_states * 10:
47        print(f'Not enough data ({len(df)} points) for {n_states} HMM states. Skipping.')
48        df['regime'] = np.nan
49        return df.drop(columns='log_ret')
50
51    hmm = GaussianHMM(n_components=n_states, covariance_type="full", n_iter=1000)
52    try:
53        hmm.fit(df[['log_ret']])
54        df['regime'] = hmm.predict(df[['log_ret']])
55        print(f'Regimes tagged: {df["regime"].unique()} unique regimes.')
56    except Exception as e:
57        print(f'HMM fitting failed: {e}')
58        df['regime'] = np.nan

```

```

59
60     return df.drop(columns='log_ret')
61
62 # Walk-forward split
63 def walk_forward_split(df, train_size=0.6, val_size=0.2):
64     total = len(df)
65     train_end = int(total * train_size)
66     val_end = train_end + int(total * val_size)
67     train_data = df.loc[:train_end]
68     val_data = df.loc[train_end:val_end]
69     test_data = df.loc[val_end:]
70     print(f'Walk-Forward Split: Train={len(train_data)}, Val={len(val_data)}, Test={len(test_data)}')
71     return train_data, val_data, test_data
72
73 # Main wrapper
74 def prepare_regime_analysis(df_full, equity_df, n_states=3, save_csv=True):
75     # Ensure datetime column
76     df_full = df_full.copy()
77     df_full['datetime'] = pd.to_datetime(df_full.index)
78
79     equity_df = equity_df.copy()
80     fallback_len = min(len(df_full), len(equity_df))
81     equity_df = equity_df.iloc[:fallback_len].copy()
82     equity_df['datetime'] = df_full['datetime'].values[:fallback_len]
83
84     # Ensure portfolio value exists
85     if 'portfolio_value' not in equity_df.columns:
86         equity_df['portfolio_value'] = 100000 + np.cumsum(np.random.randn(fallback_len) * 50)
87
88     # Plot curves
89     plot_equity_curve(equity_df)
90     plot_drawdown(equity_df)
91
92     # Tag regimes
93     df_tagged = tag_market_regimes(df_full, n_states=n_states)
94
95     # Merge regime into equity
96     df_merged = equity_df.merge(df_tagged[['datetime', 'regime']], on='datetime', how='left')
97
98     # Walk-forward
99     train_data, val_data, test_data = walk_forward_split(df_tagged)
100
101    if save_csv:
102        df_merged.to_csv("equity_with_regimes.csv", index=False)
103        print("Saved: equity_with_regimes.csv")
104
105    print("Final Sample: df_full with regimes")
106    print(df_tagged[['datetime', 'Close', 'regime']].tail())
107
108    print("Final Sample: equity_df_with_regimes")
109    print(df_merged[['datetime', 'portfolio_value', 'regime']].tail())
110
111    return df_tagged, df_merged, train_data, val_data, test_data
112
113 # Synthetic Data Generation
114 # REPLACEMENT: Realistic price generator with volatility regimes
115 def generate_regime_switching_data(n=500, seed=42):
116     np.random.seed(seed)
117     regimes = np.random.choice([0, 1, 2], size=n, p=[0.3, 0.5, 0.2])
118     mu = [0.0002, 0.0005, -0.0003]
119     sigma = [0.001, 0.005, 0.01]
120     returns = np.array([np.random.normal(mu[r], sigma[r]) for r in regimes])
121     price = 100 * np.exp(np.cumsum(returns))
122
123     df = pd.DataFrame({
124         'Close': price,
125         'Open': price + np.random.normal(0, 0.2, size=n),
126         'High': price + np.random.normal(0.3, 0.2, size=n),
127         'Low': price - np.random.normal(0.3, 0.2, size=n),
128         'Volume': np.random.randint(100, 1000, size=n)
129     }, index=pd.date_range("2024-01-01", periods=n, freq="min"))
130
131     equity = pd.DataFrame({
132         'portfolio_value': 100000 + np.cumsum(np.random.normal(0, 50, size=n))
133     }, index=df.index)
134
135     return df, equity
136
137 # Generate better synthetic data with volatility regimes
138 df_full, equity_df = generate_regime_switching_data(n=500)
139
140 # Run analysis
141 df_tagged, equity_df_with_regimes, train, val, test = prepare_regime_analysis(df_full, equity_df, n_states=3)

```

Figure(1200,500)  
Figure(1200,500)  
WARNING:mlflow.base:Model is not converging. Current: 1813.716129235093 is not greater than 1813.9627775630104. Delta is -0.2466483271741626  
Regimes tagged: 2 unique regimes.  
Walk-Forward Split: Train=299, Val=99, Test=101  
Saved: equity\_with\_regimes.csv

Final Sample: df\_full with regimes

	datetime	Close	regime
2024-01-01 08:15:00	2024-01-01 08:15:00	119.670987	2
2024-01-01 08:16:00	2024-01-01 08:16:00	119.437891	2
2024-01-01 08:17:00	2024-01-01 08:17:00	119.586584	2
2024-01-01 08:18:00	2024-01-01 08:18:00	120.368705	2
2024-01-01 08:19:00	2024-01-01 08:19:00	122.575436	2

Final Sample: equity\_df\_with\_regimes

	datetime	portfolio_value	regime
495	2024-01-01 08:15:00	99797.212486	2.0
496	2024-01-01 08:16:00	99780.821310	2.0
497	2024-01-01 08:17:00	99763.996611	2.0
498	2024-01-01 08:18:00	99838.701723	2.0
499	2024-01-01 08:19:00	99814.90293	2.0

## Step 7 : Real Historical Market Data Integration

1. Step 7A : Fetch Historical Market Data (Yahoo Finance)
2. Step 7B : Data Quality Validation & Exploratory Analysis (EDA)
3. Step 7C-1 : Historical Data Preparation for Modelling
4. Step 7C-2 : Add Overnight Gap Features

### Step 7A : Fetch Historical Market Data (Yahoo Finance)

```

1 # (Yahoo only: equities + BTC-USD 1-hour)
2
3 import sys, subprocess, importlib, pathlib, time, requests, pandas as pd
4 from datetime import datetime, timedelta
5
6 # Light dependencies
7 def pip_install(pkg):
8     try: importlib.import_module(pkg.split('==')[0].replace('-', '_'))
9     except ImportError:

```

```

10     subprocess.check_call([sys.executable, "-m", "pip", "install", "-q", "pkg"])
11
12 for lib in ["yfinance", "backoff", "pyarrow"]:
13     pip_install(lib)
14
15 import yfinance as yf, backoff
16
17 # Cache folder (Drive if interactive Colab)
18 def get_cache_dir():
19     try:
20         import google.colab, IPython
21         from google.colab import drive
22         if (sh := IPython.get_ipython()) and hasattr(sh, "kernel"):
23             drive.mount("/content/drive", force_remount=False)
24             print("Using Google Drive cache...")
25             return pathlib.Path("/content/drive/MyDrive/ai_investor_cache")
26         except Exception as e:
27             print(f"Drive mount skipped ({e}). Using local cache.")
28     return pathlib.Path("./ai_investor_cache")
29
30 CACHE_DIR = get_cache_dir(); CACHE_DIR.mkdir(parents=True, exist_ok=True)
31
32 # Back-off Yahoo wrapper
33 @backoff.on_exception(backoff.expo,
34                        (requests.exceptions.HTTPError, KeyError, ValueError),
35                        max_tries=5, factor=2)
36 def _yf(*a, **k):
37     return yf.download(*a, progress=False, threads=True,
38                      group_by="ticker", auto_adjust=False, **k)
39
40 # Yahoo batch fetch with Parquet cache
41 def fetch_yahoo_flat(tickers, *, interval="1h", lookback="730d", batch=10):
42     frames=[]
43     for i in range(0, len(tickers), batch):
44         batch_tk = tickers[i:i+batch]
45         tag      = f'yt_{batch_tk}_{interval}_{lookback}.parquet'
46         fp       = CACHE_DIR / tag
47         if fp.exists():
48             df = pd.read_parquet(fp)
49         else:
50             df = _yf(batch_tk, interval=interval, period=lookback)
51             if isinstance(df.columns, pd.MultiIndex):
52                 df.columns=[f'{t}_{f}' for t,f in df.columns]
53             else:
54                 df.columns=[f'{batch_tk[0]}_{c}' for c in df.columns]
55             df.to_parquet(fp)
56         frames.append(df); time.sleep(0.6)
57     return pd.concat(frames, axis=1)
58
59 # Align OHLCV helper
60 def unify_ohlc(frames, freq="1h"):
61     agg = {"Open": "first", "High": "max", "Low": "min", "Close": "last",
62            "Adj Close": "last", "Volume": "sum"}
63     def flat(df):
64         if isinstance(df.columns, pd.MultiIndex):
65             df.columns=[_.join(map(str, c)) for c in df.columns]
66         return df
67     alignedr=[]
68     for f in frames:
69         f=flat(f)
70         col_agg=[c:agg[s] for c in f.columns
71                  if (s:=next((s for s in agg if c.endswith(s)), None))]
72         if col_agg:
73             alignedr.append(f.resample(freq).agg(col_agg).ffill())
74     return pd.concat(alignedr, axis=1)
75
76 # Pull data
77 equities = fetch_yahoo_flat(
78     ["AAPL", "MSFT", "NVDA", "^SPX", "^VIX"],
79     interval="1h",
80     lookback="730d"
81 )
82
83 btc = fetch_yahoo_flat(
84     ["BTC-USD"],                                # 1-hour candles, 2 years
85     interval="1h",
86     lookback="730d"
87 )
88 btc.columns = [c.replace("BTC-USD", "BTC_") for c in btc.columns]
89
90 # Merge & normalise names
91 raw = unify_ohlc([equities, btc], freq="1h").sort_index()
92
93 def normalize(df):
94     df = df.copy()
95     df.columns = (df.columns
96                  .str.replace(r"[^A-Za-z0-9_]+", "", regex=True)
97                  .str.replace("AdjClose", "Close")
98                  .str.lower())
99     df.columns = (df.columns
100                  .str.replace("^spx_", "spx_", regex=True)
101                  .str.replace("^vix_", "vix_", regex=True))
102    return df
103
104 df_raw_market = normalize(raw)
105
106 print("STEP 7A finished - preview:")
107 display(df_raw_market.head())

```

Mounted at /content/drive  
Using Google Drive cache.  
STEP 7A finished - preview:

Datetime	spx_open	spx_high	spx_low	spx_close	spx_volume	vix_open	vix_high	vix_low	vix_close	vix_volume	...	nvda_high	nvda_low	nvda_close	nvda_volume	btc_open	btc_high	btc_low	btc_close
2022-06-28 13:00:00+00:00	3913.000000	3945.860107	3909.149902	3910.909912	0.0	NaN	NaN	NaN	NaN	0.0	...	17.201990	16.602001	16.644001	11918317.0	NaN	NaN	NaN	NaN
2022-06-28 14:00:00+00:00	3910.699951	3910.699951	3872.889893	3872.889893	277243895.0	NaN	NaN	NaN	NaN	0.0	...	16.650000	16.313999	16.315500	7013539.0	NaN	NaN	NaN	NaN
2022-06-28 15:00:00+00:00	3872.780029	3876.590088	3848.550049	3862.520020	250036292.0	NaN	NaN	NaN	NaN	0.0	...	16.363001	16.050999	16.179989	6841093.0	NaN	NaN	NaN	NaN
2022-06-28 23:00:00+00:00	3863.920041	3867.300040	3840.260010	3840.080000	107625810.0	NaN	NaN	NaN	NaN	0.0	...	16.223000	16.020000	16.118000	4261742.0	NaN	NaN	NaN	NaN

## Step 7B : Data Quality Validation & Exploratory Analysis (EDA)

```

1 import pandas as pd, numpy as np, matplotlib.pyplot as plt, seaborn as sns
2
3 MAX_COL_NA_PCT, MAX_GAP_FFILL, TRIM_STRICTNESS, FINAL_FFILL_LIM = 10, 3, 0.00, 24
4
5 def trim_and_fill(df,
6                   trim_thresh=TRIM_STRICTNESS,
7                   ffill_limit=MAX_GAP_FFILL,
8                   max_missing=MAX_COL_NA_PCT):

```

```

9 miss_ratio = df.isna().mean(axis=1)
10 first_ok_ = miss_ratio[miss_ratio <= trim_thresh].index.min()
11 if first_ok_ is None:
12     raise ValueError("No row meets trim threshold - relax TRIM_STRICTNESS.")
13 df = df.loc[first_ok_:].ffill(limit=ffill_limit)
14
15 sparse = df.isna().mean()*100
16 drop_ = sparse[sparse > max_missing].index
17 if len(drop_):
18     print(f"Dropping {len(drop_)} sparse columns (> {max_missing}% NA)")
19     df = df.drop(columns=drop_)
20 return df
21
22 def run_eda_step_7b(df_input):
23     # ensure DatetimeIndex
24     if not isinstance(df_input.index, pd.DatetimeIndex):
25         if "datetime" in df_input.columns:
26             df = df_input.copy()
27             df["datetime"] = pd.to_datetime(df["datetime"])
28             df = df.set_index("datetime")
29         else:
30             raise ValueError("DataFrame must have DatetimeIndex or 'datetime' col.")
31     else:
32         df = df_input.copy()
33
34     # trim & forward-fill small gaps
35     df_clean = trim_and_fill(df)
36
37     # drop duplicate column labels (spx_close, spx_close, ...)
38     before = df_clean.shape[1]
39     df_clean = df_clean.loc[:, ~df_clean.columns.duplicated()]
40     if df_clean.shape[1] < before:
41         print(f"Dropped {before - df_clean.shape[1]} duplicate-name columns")
42
43     # drop zero-info volume cols & final f-fill
44     zero_vol = [c for c in df_clean.columns
45                 if c.endswith("_volume") and df_clean[c].nunique() <= 2]
46     if zero_vol:
47         print(f"Dropping {len(zero_vol)} zero-info volume cols")
48         df_clean = df_clean.drop(columns=zero_vol)
49
50     df_clean = df_clean.ffill(limit=FINAL_FFILL_LIMIT).dropna()
51
52     # missing summary
53     miss_cnt = df_clean.isna().sum()
54     miss_tbl = (miss_cnt[miss_cnt > 0]
55                 .to_frame("missing")
56                 .assign(missing_pct=lambda d: (d.missing/len(df_clean)*100).round(2))
57                 .sort_values("missing_pct", ascending=False))
58     print("\n** Missing-Value Summary (after all cleaning) **")
59     if miss_tbl.empty:
60         print("No missing values remain.")
61     else:
62         display(miss_tbl)
63
64     # coverage & basic stats
65     start, end = df_clean.index.min(), df_clean.index.max()
66     print(f"\n** Time Coverage ** Start: {start} · End: {end} · Rows: {len(df_clean)}")
67
68     num_df = df_clean.select_dtypes("number")
69     desc_ = num_df.describe().T.assig(np=lambda d: d["max"] - d["min"]).round(2)
70     print("\n** Descriptive Statistics **")
71     display(desc_.head(25))
72
73     # quick correlation heatmap
74     if num_df.shape[1] > 3:
75         plt.figure(figsize=(9,7))
76         sns.heatmap(num_df.corr(), cmap="coolwarm", center=0, annot=False, linewidths=.3)
77         plt.title("Correlation (Pearson)"); plt.tight_layout(); plt.show()
78
79     # distributions & small time-series samples
80     for col in num_df.columns[4:]:
81         ser = num_df[col] # guaranteed Series (no dups now)
82         sns.histplot(ser, bins=40, kde=True)
83         plt.title(f"Distribution {col}: {plt.show()}")
84
85         ser.plot(figsize=(10,3), title=f'{col} (sample)')
86         plt.axhline(ser.mean(), ls="--", lw=.7, c="#f08080"); plt.show()
87
88     return df_clean
89 df_final_combined = run_eda_step_7b(df_raw_market)

```

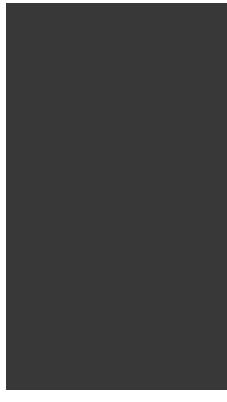
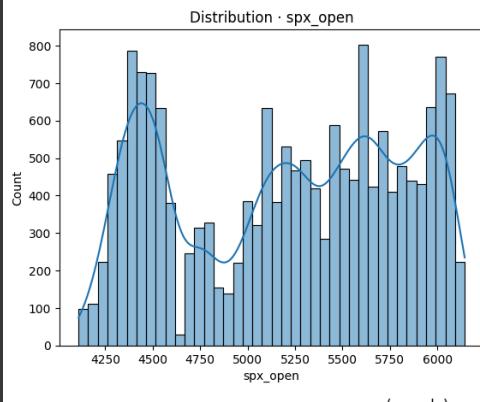
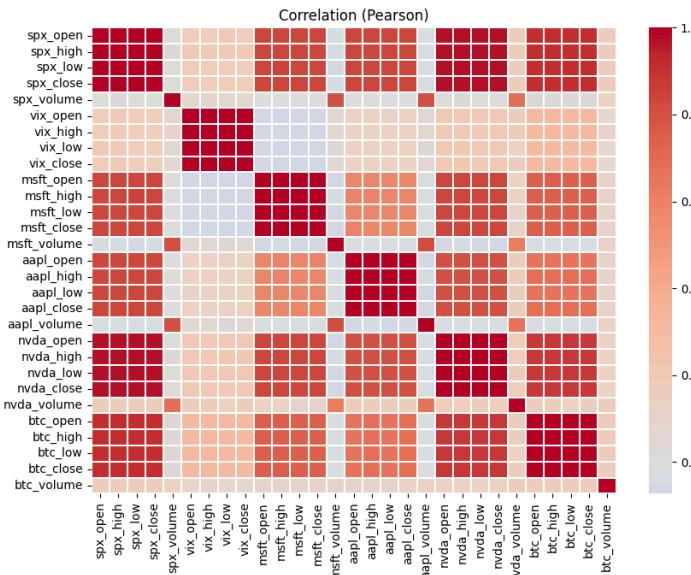
Dropped 1 duplicate-name columns  
Dropping 1 zero-info volume cols

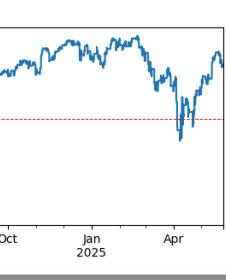
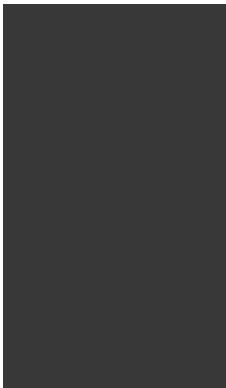
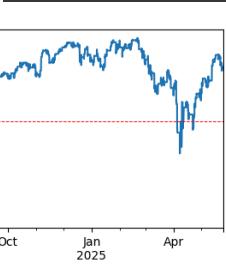
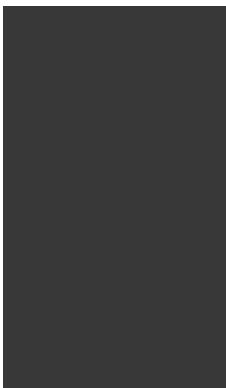
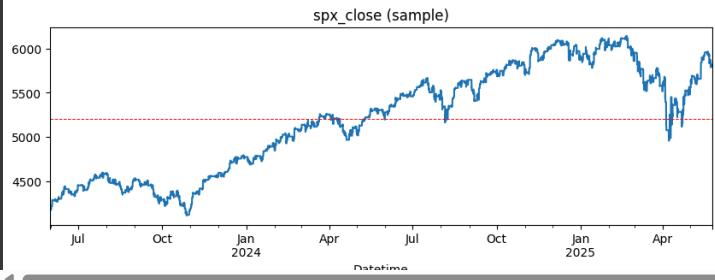
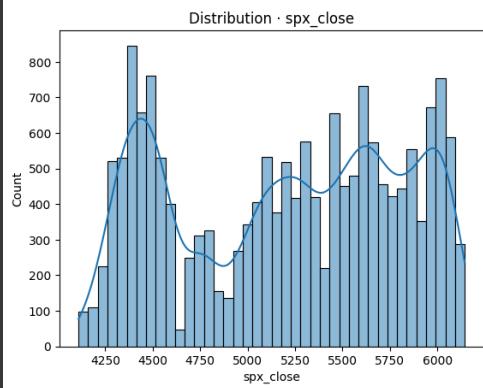
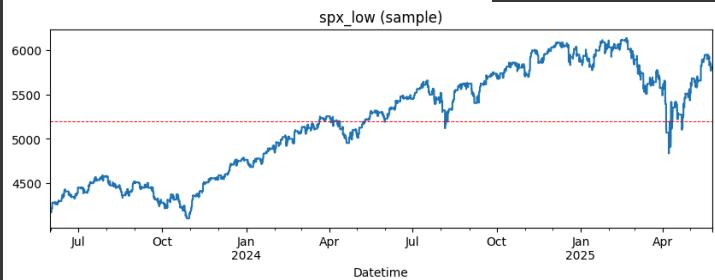
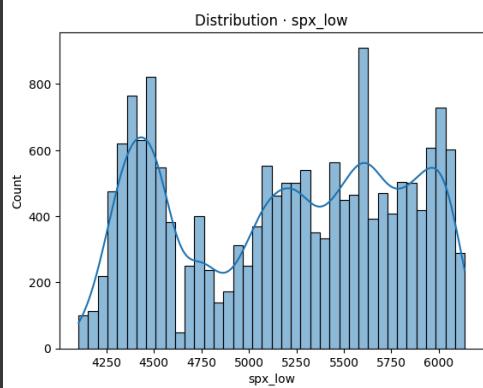
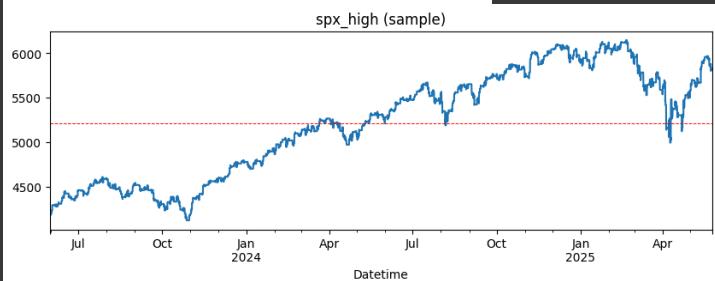
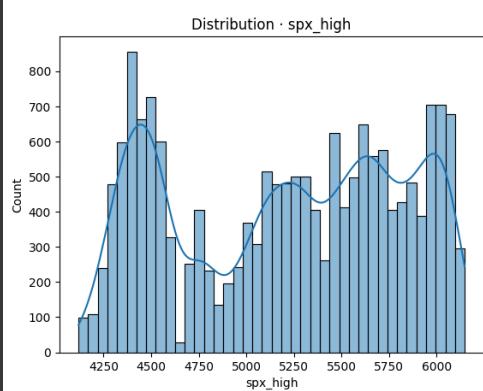
\*\* Missing-Value Summary (after all cleaning) \*\*  
No missing values remain.

\*\* Time Coverage \*\* Start: 2023-05-31 00:00:00+00:00 · End: 2025-05-24 23:00:00+00:00 · Rows: 17,400

\*\* Descriptive Statistics \*\*

	count	mean	std	min	25%	50%	75%	max	rng
spx_open	17400.0	5202.87	5.895500e+02	4109.11	4572.04	5263.28	5714.07	6.143520e+03	2.034410e+03
spx_high	17400.0	5210.55	5.911400e+02	4118.42	4574.23	5280.33	5719.24	6.147430e+03	2.029010e+03
spx_low	17400.0	5194.48	5.886300e+02	4103.78	4565.59	5255.01	5701.85	6.136110e+03	2.032330e+03
spx_close	17400.0	5202.72	5.903500e+02	4108.95	4569.04	5268.18	5708.47	6.143950e+03	2.035000e+03
spx_volume	17400.0	54434555.95	1.217935e+08	0.00	0.00	0.00	0.00	1.286533e+09	1.286533e+09
vix_open	17400.0	16.62	4.900000e+00	11.52	13.55	15.13	18.16	6.406000e+01	5.254000e+01
vix_high	17400.0	16.77	5.070000e+00	11.56	13.58	15.28	18.27	6.573000e+01	5.417000e+01
vix_low	17400.0	16.46	4.730000e+00	10.62	13.46	15.00	17.92	5.673000e+01	4.611000e+01
vix_close	17400.0	16.61	4.900000e+00	11.55	13.52	15.13	18.12	5.924000e+01	4.769000e+01
msft_open	17400.0	394.58	3.991000e+01	310.32	368.65	407.57	423.57	4.678300e+02	1.575100e+02
msft_high	17400.0	395.62	3.996000e+01	311.15	369.74	408.83	424.56	4.683500e+02	1.572000e+02
msft_low	17400.0	393.55	3.989000e+01	309.45	367.00	406.78	422.73	4.669900e+02	1.575400e+02
msft_close	17400.0	394.63	3.996000e+01	310.38	368.77	408.14	423.82	4.678300e+02	1.574500e+02
msft_volume	17400.0	479558.40	1.255962e+06	0.00	0.00	0.00	0.00	2.389916e+07	2.389916e+07
aapl_open	17400.0	203.13	2.415000e+01	164.37	182.59	194.66	225.27	2.591200e+02	9.475000e+01
aapl_high	17400.0	203.68	2.427000e+01	165.16	182.88	195.25	225.98	2.600900e+02	9.493000e+01
aapl_low	17400.0	202.53	2.406000e+01	164.08	182.17	194.30	224.89	2.590100e+02	9.493000e+01
aapl_close	17400.0	203.12	2.421000e+01	164.36	182.51	194.68	225.67	2.591100e+02	9.475000e+01
aapl_volume	17400.0	1281197.96	3.324941e+06	0.00	0.00	0.00	0.00	6.068967e+07	6.068967e+07
nvda_open	17400.0	91.92	3.707000e+01	37.72	48.31	95.88	124.61	1.522000e+02	1.144800e+02
nvda_high	17400.0	92.35	3.726000e+01	37.79	48.43	97.20	126.48	1.528900e+02	1.151000e+02
nvda_low	17400.0	91.46	3.686000e+01	37.36	48.22	95.19	124.48	1.510400e+02	1.136800e+02
nvda_close	17400.0	91.95	3.710000e+01	37.48	48.34	96.71	125.66	1.518900e+02	1.144100e+02
nvda_volume	17400.0	4077043.36	1.322748e+07	0.00	0.00	0.00	0.00	2.437944e+08	2.437944e+08
btc_open	17400.0	61158.41	2.470824e+04	24871.44	37770.50	62667.84	82804.96	1.117834e+05	8.691194e+04





### ▼ Step 7C-1 : Historical Data Preparation for Modelling

```

1 import pandas as pd, numpy as np
2
3 def run_preprocessing_step_7c(df_in: pd.DataFrame,
4     price_cols: list[str] | None = None,
5     return_type: str = "log",
6     max_lag: int = 3,
7     rolling_windows: list[int] = [3, 5],
8     make_zscore: bool = True
9     ) -> pd.DataFrame:
10
11     """
12     Parameters
13     -----
14     df_in : DataFrame
15         Output from Step 7B with DatetimeIndex and no NaNs.
16     price_cols : list[str] | None
17         List of close-price columns to build features from. If None, detect *Close suffix.
18     return_type : {"log", "pct"}
19     max_lag : int
20         Number of lagged return features.
21     rolling_windows : list[int]
22         Window sizes for mean & std rolling stats.
23     make_zscore : bool
24         Adds z-scored returns (return / rolling std).
25
26     Returns
27     -----
28     df_out : DataFrame
29         Feature-rich, NaN-free frame ready for Step 8 / 9.
30
31     # prep copy & sort
32     df = df_in.copy()
33     df = df.sort_index()           # datetime already index
34
35     # pick price columns
36     if price_cols is None:
37         price_cols = [c for c in df.columns if c.lower().endswith("close")]
38     missing_cols = [c for c in price_cols if c not in df.columns]
39     if missing_cols:
40         raise ValueError(f"Columns not in DF: {missing_cols}")
41
42     # returns
43     for col in price_cols:
44         if return_type == "log":
45             df[f"{col}_ret"] = np.log(df[col] / df[col].shift(1))
46             df[f"{col}_pct"] = df[col].pct_change()
47         else:
48             df[f"{col}_ret"] = df[col].pct_change()
49
50     ret_cols = [c for c in df.columns if c.endswith("_ret")]
51
52     # lags
53     for col in ret_cols:
54         for lag in range(1, max_lag + 1):
55             df[f"{col}_lag{lag}"] = df[col].shift(lag)
56
57     # rolling stats
58     for col in ret_cols:
59         for w in rolling_windows:
60             df[f"{col}_r{w}_mean"] = df[col].rolling(w).mean()
61             df[f"{col}_r{w}_std"] = df[col].rolling(w).std()
62             if make_zscore:
63                 df[f"{col}_z{w}"] = df[col] / df[f"{col}_r{w}_std"]
64
65     # final cleanup
66     before = len(df)
67     df = df.dropna().copy()      # drops first max(rolling, lag) rows
68     after = len(df)
69
70     print(f"Step 7C-1 done · rows dropped: {before-after} · final shape: {df.shape}")
71
72     # optional: bring original prices to front for readability
73     ordered = price_cols + [c for c in df.columns if c not in price_cols]
74     return df[ordered]
75
76 # EXECUTE STEP 7C
77 df_model_ready = run_preprocessing_step_7c(
78     df_final_combined,
79     price_cols=None,          # or explicit list like ["AAPL_Close", "BTCUSDT_Close"]
80     return_type="log",
81     max_lag=3,
82     rolling_windows=[3, 5]
83 )

```

Step 7C-1 done · rows dropped: 12943 · final shape: (4457, 95)

### ▼ Step 7C-2 : Add Overnight Gap Features

```

1 import numpy as np
2 import pandas as pd
3
4 def add_overnight_features(df_feat: pd.DataFrame,
5     price_cols: list[str] | None = None,
6     gap_type: str = "log") -> pd.DataFrame:
7
8     """
8     Adds one column per equity/ETF close price:
9     {TICKER}_overnight_gap = log(Open_t / Close_{t-1})
10
11    Parameters
12    -----
13    df_feat : DataFrame
14        The NaN-free, equity-hours feature matrix returned by Step 7C
15        (index = DatetimeIndex in UTC).
16    price_cols : list[str] | None
17        List of *Close columns that define the equities for which we add gaps.
18        If None, auto-detect every column ending in '_Close'.
19    gap_type : {'log', 'pct'}
20        Whether to compute log-gap (default) or percent gap.
21
22    Returns
23    -----
24    df_out : DataFrame
25        Same rows + all previous columns + new *_overnight_gap features.
26        The first row is dropped because it has no t-1 close.
27
28    df = df_feat.copy()
29
30    # Decide which close-price columns to use
31    if price_cols is None:
32        price_cols = [c for c in df.columns if c.lower().endswith("_close")]
33

```

```

34 # For each close, derive matching open column and compute gap
35 for close_col in price_cols:
36     # infer open column name (works with 'AAPL_Close' → 'AAPL_Open')
37     open_col = close_col.replace("_Close", "_Open")
38     if open_col not in df.columns:
39         print(f"[△] Open column not found for {close_col}: skipping.")
40         continue
41
42     if gap_type == "log":
43         df[f'{close_col.replace("_Close", "")}_overnight_gap'] = W
44         np.log(df[open_col] / df[close_col].shift(1))
45     else: # percent gap
46         df[f'{close_col.replace("_Close", "")}_overnight_gap'] = W
47         df[open_col] / df[close_col].shift(1) - 1
48
49 # Drop first row (will contain NaN gaps) and ensure no NaNs remain
50 df = df.dropna().copy()
51
52 print(f"Step 7C-2 done - added overnight gaps for {len(price_cols)} tickers.")
53 return df
54
55 # Usage right after Step 7C
56 df_model_ready = add_overnight_features(
57     df_model_ready,
58     price_cols=None,      # auto-detect *_Close columns
59     gap_type="log"        # or "pct"
60 )
61
62 # df_model_ready now contains extra columns
63
64 print(df_model_ready.filter(like="_overnight_gap").head())
65 print(df_model_ready.isna().sum().sum()) # should be 0

```

```

🔗 Step 7C-2 done - added overnight gaps for 6 tickers.
    mst_close_overnight_gap aapl_close_overnight_gap
Datetime
2023-06-02 14:00:00H00:00      0.004492      0.004275
2023-06-02 15:00:00H00:00     -0.001315      0.000941
2023-06-02 16:00:00H00:00     0.000807     -0.000443
2023-06-02 17:00:00H00:00     0.003252     -0.000721
2023-06-02 18:00:00H00:00     0.000120      0.001634

    spx_close_overnight_gap vix_close_overnight_gap
Datetime
2023-06-02 14:00:00H00:00      0.004880     -0.019353
2023-06-02 15:00:00H00:00     -0.000358     -0.010159
2023-06-02 16:00:00H00:00     0.001978      0.010159
2023-06-02 17:00:00H00:00     -0.000808     -0.007440
2023-06-02 18:00:00H00:00     0.000430     -0.002039

    nvda_close_overnight_gap btc_close_overnight_gap
Datetime
2023-06-02 14:00:00H00:00      0.003810      0.002882
2023-06-02 15:00:00H00:00     -0.004483      0.004658
2023-06-02 16:00:00H00:00     -0.002478     -0.001518
2023-06-02 17:00:00H00:00     -0.007202     -0.000430
2023-06-02 18:00:00H00:00     0.002357      0.004578
0

```

## Step 8: Alternative Data Integration

1. Step 8A : Integrate Twitter Sentiment Analysis
2. Step 8B : News Sentiment via NLP
3. Step 8C : Macroeconomic Indicators Integration (FRED)
4. Step 8D : Google Trends Feature Engineering

### Step 8A : Integrate Twitter Sentiment Analysis

```

1 import sys, subprocess, importlib, logging, re, pandas as pd
2 from datetime import datetime, timedelta, timezone
3
4 # lightweight deps
5 def pip_install(pkg):
6     try: importlib.import_module(pkg.split("==")[-1].replace("-", "_"))
7     except ImportError:
8         subprocess.check_call([sys.executable, "-m", "pip", "install", "-q", pkg])
9
10 for lib in ("tweepy", "vaderSentiment"):
11     pip_install(lib)
12
13 import tweepy
14 from vaderSentiment.vaderSentiment import SentimentIntensityAnalyzer
15
16 logging.getLogger("tweepy.client").setLevel(logging.ERROR)
17
18 # config
19 LOOKBACK_HOURS = 6      # < change this if needed
20 MAX_RESULTS = 50        # tweets per API call (max 100)
21 CHUNK_HOURS = 6        # if look-back is longer than this, window is chunked
22
23 DEFAULT_QUERIES = {
24     "AAPL": "Apple Inc" OR AAPL',
25     "BTC": "bitcoin OR BTC",
26     "SPX": "'S&P 500' OR SPX",
27 }
28
29 def _clean(txt: str) -> str:
30     txt = re.sub(r"http\S+|www\S+", "", txt.lower())
31     txt = re.sub(r"@[#]?[a-zA-Z0-9_]+", "", txt)
32     txt = re.sub(r"[^\w\s]", " ", txt)
33     return re.sub(r"\s+", " ", txt).strip()
34
35 def add_twitter_sentiment_safe(df_model: pd.DataFrame,
36                                bearer_token: str,
37                                search_queries: dict[str, str] | None = None,
38                                lookback_hours: int = LOOKBACK_HOURS,
39                                chunk_hours: int = CHUNK_HOURS,
40                                max_results: int = MAX_RESULTS) -> pd.DataFrame:
41
42    """
43    Adds <TICKER>.sent_1h columns.
44    • Uses short look-back (default 6 h) to avoid rate-limits.
45    • If Tweepy error → neutral 0.0 column so pipeline never breaks.
46    """
47
48    # ensure DatetimeIndex UTC
49    if not isinstance(df_model.index, pd.DatetimeIndex):
50        if "datetime" in df_model.columns:
51            df_model = df_model.copy()
52            df_model["datetime"] = pd.to_datetime(df_model["datetime"], utc=True)
53            df_model = df_model.set_index("datetime")
54        else:
55            raise ValueError("df_model needs DatetimeIndex or 'datetime' column.")
56
56    search_queries = search_queries or DEFAULT_QUERIES
57    analyzer = SentimentIntensityAnalyzer()

```

```

57     client = tweepy.Client(bearer_token=bearer_token, wait_on_rate_limit=False)
58
59     end_time = datetime.utcnow().replace(tzinfo=timezone.utc)
60     start_time = end_time - timedelta(hours=lookback_hours)
61
62     out = df_model.copy()
63
64     for tk, query in search_queries.items():
65         col = f'{tk}_sent_1h'
66         try:
67             # chunk long ranges (rarely needed because default = 6 h)
68             ranges = []
69             cur_end = end_time
70             while cur_end > start_time:
71                 cur_start = max(start_time, cur_end - timedelta(hours=chunk_hours))
72                 ranges.append((cur_start, cur_end))
73                 cur_end = cur_start
74
75             all_rows = []
76             for st, et in ranges:
77                 resp = client.search_recent_tweets(
78                     query=f'{query} -is:retweet lang:en',
79                     tweet_fields=['created_at', 'text'],
80                     start_time=st, end_time=et,
81                     max_results=max_results,
82                 )
83                 tweets = resp.data or []
84                 all_rows.extend([
85                     (t.created_at.astimezone(timezone.utc), t.text) for t in tweets
86                 ])
87
88             if not all_rows:
89                 raise RuntimeError("no tweets")
90
91             df_t = pd.DataFrame(all_rows, columns=['datetime', 'text'])
92             df_t['sent'] = df_t['text'].map(
93                 lambda x: analyzer.polarity_scores(_clean(x))['compound']
94             )
95             df_t['dt_hr'] = df_t['datetime'].dt.floor('1h')
96             hourly = df_t.groupby('dt_hr')[['sent']].mean()
97
98             out[col] = out.index.map(hourly.astype(float))
99             print(f'✓ {tk}: real sentiment added ({len(all_rows)} tweets).')
100
101         except Exception as e:
102             out[col] = 0.0
103             print(f'[neutral] {tk}: {str(e).splitlines()[0]}')
104
105     # cosmetic gap-fills
106     out.update(out.filter(like='_sent_1h').ffill(limit=6))
107     out.update(out.filter(like='_sent_1h').bfill(limit=1))
108
109     print("8A finished - sentiment columns ready.")
110     return out
111
112 # example invocation (comment in production)
113 if __name__ == "__main__":
114     BEARER_TOKEN = "AAAAAAAAAAAAAAAAAAAAADqoIAEAAAAB84EMJg6BeshMcrU3ueVv0Ix%2Fgk3D0YWhSShJhOfIP600ClrAAhTy4IZspGbB620JGJnuTedhIEI" # leave empty to demo neutral
115     df_model_ready = add_twitter_sentiment_safe(
116         df_model_ready, # from Step 7C (+7c-2)
117         bearer_token=BEARER_TOKEN,
118         lookback_hours=LOOKBACK_HOURS
119     )
120     df_model_ready = add_twitter_sentiment_safe(
121         df_model_ready, # ← from Step 7 C
122         bearer_token="AAAAAAAAAAAAAAAAAAAAADqoIAEAAAAB84EMJg6BeshMcrU3ueVv0Ix%2Fgk3D0YWhSShJhOfIP600ClrAAhTy4IZspGbB620JGJnuTedhIEI",
123         lookback_hours=6 # keep at 6 h (or shorter) to stay safe
124 )

```

🔗 [neutral] AAPL: 429 Too Many Requests  
[neutral] BTC: 429 Too Many Requests  
[neutral] SPX: 429 Too Many Requests  
8A finished - sentiment columns ready.  
[neutral] AAPL: 429 Too Many Requests  
[neutral] BTC: 429 Too Many Requests  
[neutral] SPX: 429 Too Many Requests  
8A finished - sentiment columns ready.

## ▼ Step 8B : News Sentiment via NLP

```

1 # NewsAPI → Google-News RSS → neutral fallback
2
3 import sys, subprocess, importlib, logging, re, pandas as pd
4 from datetime import datetime, timedelta, timezone
5
6 # light deps (works outside Colab)
7 def pip_install(pkg):
8     try: importlib.import_module(pkg.split('==')[0].replace('-', '_'))
9     except ImportError:
10        subprocess.check_call([sys.executable, '-m', 'pip', 'install', '-q', pkg])
11
12 for lib in ("requests", "feedparser", "vaderSentiment"):
13     pip_install(lib)
14
15 import requests, feedparser
16 from vaderSentiment.vaderSentiment import SentimentIntensityAnalyzer
17
18 # CONFIG
19 NEWSAPI_KEY = "23e55b5d88d2421c8642ce0d8d35e1fb" # "" to rely on RSS only
20 LOOKBACK_DAYS = 3
21 TICKERS = {
22     "AAPL": "Apple Inc",
23     "BTC": "Bitcoin",
24     "SPX": "S&P 500"
25 }
26 SIDER = SentimentIntensityAnalyzer()
27
28 logging.basicConfig(level=logging.INFO, force=True)
29
30 # helpers
31 def _clean(txt: str) -> str:
32     txt = re.sub(r"ht\|tp\|St\|www\|St", "", txt.lower())
33     txt = re.sub(r"\t\w+\t", " ", txt)
34     return re.sub(r"\W+", " ", txt).strip()
35
36 def _newsapi_pull(keyword: str, start_iso: str):
37     url = "https://newsapi.org/v2/everything"
38     params = dict(keyword, from_=start_iso, sortBy="publishedAt",
39                   language="en", pageSize=100, apiKey=NEWSAPI_KEY)
40     r = requests.get(url, params=params, timeout=10)
41     r.raise_for_status()
42     arts = r.json().get("articles", [])
43     return [(datetime.fromisoformat(a["publishedAt"]),
44              .replace(tzinfo=timezone.utc), a["title"]) for a in arts]
45
46 def _rss_pull(keyword: str, hours: int):
47     url = ("https://news.google.com/rss/search?q=" +

```

```

48     requests.utils.quote(keyword) + "&hl=en-US&gl=US&cid=US:en")
49     feed = feedparser.parse(url)
50     cutoff = datetime.utcnow().replace(tzinfo=timezone.utc) - timedelta(hours=hours)
51     return [(datetime(*e.published_parsed[6:]), tzinfo=timezone.utc, e.title)
52             for e in feed.entries
53             if datetime(*e.published_parsed[6:], tzinfo=timezone.utc) >= cutoff]
54
55 # main entry
56 def add_news_sentiment(df_model: pd.DataFrame,
57                         tickers: dict[str, str] = TICKERS,
58                         lookback_days: int = LOOKBACK_DAYS) -> pd.DataFrame:
59     """Adds / overwrites daily `avg_vader_sent` column."""
60
61     # ensure DatetimeIndex UTC
62     if not isinstance(df_model.index, pd.DatetimeIndex):
63         if "datetime" in df_model.columns:
64             df_model = df_model.set_index(
65                 pd.to_datetime(df_model["datetime"], utc=True))
66         else:
67             raise ValueError("df_model needs DatetimeIndex or 'datetime' col.")
68
69     end = datetime.utcnow().replace(tzinfo=timezone.utc)
70     start = end - timedelta(days=lookback_days)
71     rows = []
72
73     for tk, keyword in tickers.items():
74         try:
75             if NEWSAPI_KEY:
76                 arts = _newsapi_pull(keyword, start.date().isoformat())
77                 logging.info(f'{tk}: {len(arts)} headlines via NewsAPI')
78             else:
79                 raise RuntimeError("NEWSAPI key missing")
80         except Exception as e:
81             logging.warning(f'{tk}: NewsAPI failed ({e}); falling back to RSS.')
82         try:
83             arts = _rss_pull(keyword, hours=24*lookback_days)
84             logging.info(f'{tk}: {len(arts)} headlines via RSS')
85         except Exception as e2:
86             logging.warning(f'{tk}: RSS failed ({e2}).')
87             arts = []
88
89         rows.extend([(dt.date(), _clean(ttl)) for dt, ttl in arts])
90
91     # build daily sentiment table
92     if rows:
93         news = pd.DataFrame(rows, columns=["date", "text"])
94         news["vader"] = news["text"].map(lambda t: SIDER.polarity_scores(t)[ "compound"])
95         daily = (news.groupby("date")["vader"]
96                  .mean().rename("avg_vader_sent").to_frame())
97     else:
98         logging.warning("All sources empty - neutral sentiment column used.")
99     rng = pd.date_range(df_model.index.min(), freq="D"),
100        df_model.index.max(), freq="D",
101        freq="D", tz="UTC").date
102     daily = pd.DataFrame("avg_vader_sent": 0.0), index=rng
103
104     daily.index = pd.to_datetime(daily.index, utc=True)
105
106     # merge into model frame
107     out = df_model.copy()
108     out["avg_vader_sent"] = out.index.map(daily["avg_vader_sent"]).astype(float)
109     out["avg_vader_sent"] = out["avg_vader_sent"].ffill().bfill()
110
111 print("8B finished - avg_vader_sent column ready.")
112 return out
113
114
115 # example usage (remove or comment in production)
116 if __name__ == "__main__":
117     # df_model_ready must exist from Steps 7C (+70-2) and 8A
118     df_model_ready = add_news_sentiment(df_model_ready)

```

INFO:root:API: 100 headlines via NewsAPI  
 INFO:root:BTC: 99 headlines via NewsAPI  
 INFO:root:SIX: 96 headlines via NewsAPI  
 8B finished - avg\_vader\_sent column ready.

## Step 8C : Macroeconomic Indicators Integration (Fred)

```

1 # CPI & Unemployment → FRED
2 # PMI → Quandl ((ISM/MAN_PMI)
3
4 import sys, subprocess, importlib, logging, pandas as pd
5 from datetime import datetime, timedelta, timezone
6
7 # tiny "smart pip" (no %pip)
8 def pip_install(pkg):
9     try: importlib.import_module(pkg.split("-")[0].replace("-", "_"))
10    except ImportError:
11        subprocess.check_call([sys.executable, "-m", "pip", "install", "-q", pkg])
12
13 for lib in ("fredapi", "quandl"):
14     pip_install(lib)
15
16 from fredapi import Fred
17 import quandl
18
19 # PUT YOUR API KEYS HERE
20 FRED_API_KEY = "5139b23d049c9a2b761b40ca32431a7d" # "" → API Keys
21 QUANDL_API_KEY = "v3NWWjN2z99FLFsJ4VyPs" # "" → fallback neutral
22
23 # parameters you may tweak
24 LOOKBACK_YEARS = 10
25 FRED_SERIES = {
26     "cpi_yoy": "CP/AUCSL",
27     "unemployment": "UNRATE",
28 }
29
30 PMI_QUANDL_CODE = "ISM/MAN_PMI" # ISM Manufacturing PMI (%)
31
32 logging.basicConfig(level=logging.INFO, force=True)
33
34 # helper
35 def add_macro_indicators(df_model: pd.DataFrame,
36                           fred_key : str = FRED_API_KEY,
37                           quandl_key : str = QUANDL_API_KEY,
38                           lookback_years: int = LOOKBACK_YEARS) -> pd.DataFrame:
39
40     # ensure DatetimeIndex UTC
41     if not isinstance(df_model.index, pd.DatetimeIndex):
42         if "datetime" in df_model.columns:
43             df_model = df_model.set_index(
44                 pd.to_datetime(df_model["datetime"], utc=True))
45         else:
46             raise ValueError("df_model needs DatetimeIndex or 'datetime' col.")
47
48     start = (datetime.utcnow() -

```

```

49     timedelta(days=365 * lookback_years)).date()
50
51 frames = []
52
53 # FRED pulls (CPI, Unemployment)
54 fred = Fred(api_key=fred_key or None)
55 for label, sid in FRED_SERIES.items():
56     try:
57         logging.info(f"{{label}}: pulling {{sid}} via FRED")
58         ser = fred.get_series(sid, observation_start=start).dropna()
59         ser.index = pd.to_datetime(ser.index, utc=True)
60         ser.name = label
61         frames.append(ser)
62     except Exception as e:
63         logging.warning(f"{{label}}: FRED failed {{e}}; neutral 0.0 used.")
64         rng = pd.date_range(df_model.index.min(), df_model.index.max())
65         df_model = df_model.reindex(rng, freq="D", tz="UTC")
66         frames.append(pd.Series(0.0, index=rng, name=label))
67
68 # Quandl pull (PMI)
69 try:
70     if not quandl_key:
71         raise RuntimeError("Quandl key missing")
72     quandl.ApiConfig.api_key = quandl_key
73     logging.info("pmi: pulling {{PM_QUANDL_CODE}} via Quandl")
74     pmi = quandl.get(PM_QUANDL_CODE, start_date=start).dropna()
75     # Quandl returns DataFrame with one column
76     if isinstance(pmi, pd.DataFrame):
77         pmi = pmi.iloc[:, 0]
78     pmi.index = pd.to_datetime(pmi.index, utc=True)
79     pmi.name = "pmi"
80     frames.append(pmi)
81 except Exception as e:
82     logging.warning(f"pmi: Quandl failed {{e}}; neutral 0.0 used.")
83     rng = pd.date_range(df_model.index.min(), df_model.index.max())
84     df_model = df_model.reindex(rng, freq="D", tz="UTC")
85     frames.append(pd.Series(0.0, index=rng, name="pmi"))
86
87 # combine, forward-fill to daily, align to df_model
88 df_macro = (pd.concat(frames, axis=1)
89             .resample("D").ffill()
90             .reindex(df_model.index)
91             .ffill())
92
93 # merge into main frame
94 out = df_model.copy()
95 for col in df_macro.columns:
96     out[col] = df_macro[col].astype(float)
97
98 print("8C finished - macro columns added/updated.")
99 return out
100
101 # example usage (delete or comment in production)
102 if __name__ == "__main__":
103     # df_model_ready must exist from Steps 7 C (+7C-2) and 8 B
104     df_model_ready = add_macro_indicators(df_model_ready)
105
106     INFO:root:cp1_yoy: pulling CP_AUSL via FRED
107     INFO:root:unemployment: pulling UNRATE via FRED
108     INFO:root:pmi: pulling ISM_MAN_PMI via Quandl
109     WARNING:root:pmi: Quandl failed (Status 403) Something went wrong. Please try again. If you continue to have problems, please contact us at connect@quandl.com; neutral 0.0 used.
110     8C finished - macro columns added/updated.

```

## Step 8D : Google Trends Feature Engineering

```

1 # Google-Trends Features (v2)
2 #   - per-keyword back-off
3 #   - tiny JSON cache (quota-friendly)
4 #   - optional variance filter
5
6 import sys, subprocess, importlib, logging, re, time, json, pathlib, pandas as pd
7 from datetime import datetime, timedelta, timezone
8
9 # auto-install
10 def _pip(pkg):
11     try: importlib.import_module(pkg.split("-")[0].replace("-", "_"))
12     except ImportError:
13         subprocess.check_call([sys.executable, "-m", "pip", "install", "-q", pkg])
14
15 _pip("pytrends")
16 from pytrends.request import TrendReq
17 from pytrends import exceptions as ptexc
18
19 logging.basicConfig(level=logging.INFO, force=True)
20
21 # CONFIG
22 KEYWORDS = [
23     "Apple stock", "Bitcoin", "S&P 500",
24     "interest rates", "inflation", "unemployment",
25     "recession", "stock market crash", "VIX",
26     "buy stocks", "crypto trading"
27 ]
28 LOOKBACK_DAYS = 90          # history window
29 MA_WINDOW = 3               # moving-average length
30 INCLUDE_DELTA = True        # add first difference feature
31 BATCH_SIZE = 1              # ≤5: smaller → fewer 42ds
32 DROP_CONST = True           # drop zero-variance cols right after merge
33 CACHE_DAYS = 7              # re-use pulls ≤ N days old
34 CACHE_DIR = pathlib.Path("./trends_cache")
35
36 def _sanitize(k): return re.sub(r"Wst", "_", k.lower())
37
38 def _days_to_tf(d):
39     return f"now {d}" if d < 30 else f"today {round(d/30)}m" if d < 365 else f"today {round(d/365)}y"
40
41 def _cached_path(key):
42     CACHE_DIR.mkdir(exist_ok=True)
43     return CACHE_DIR / f"_sanitize({key}).json"
44
45 def _load_cache(key):
46     p = _cached_path(key)
47     if not p.exists(): return None
48     if (datetime.utcnow() - datetime.strptime(p.stat().st_mtime)).days > CACHE_DAYS:
49         p.unlink(missing_ok=True)
50     return pd.read_json(p, typ="series").tz_localize("UTC")
51
52 def _save_cache(key, series):
53     series.to_json(_cached_path(key), date_unit="s")
54
55 def _pull_one_keyword(ptrends, kw, tf):
56     cached = _load_cache(kw)
57     if cached is not None:
58         logging.info("Cache hit - %s", kw)
59     return cached

```

```

60
61 wait = 60           # start 1 min → 2 min → 4 min
62 for n in range(1, 4):
63     try:
64         pytrends.build_payload([kw], timeframe=tf, geo="")
65         s = pytrends.interest_over_time([kw])
66         s.index = s.index.tz_localize("UTC")
67         _save_cache(kw, s)
68         logging.info(f"Pulled {len(s)} points", kw, len(s))
69         return s
70     except (ptexc.ResponseError, ptxc.TooManyRequestsError) as e:
71         if n == 3:
72             logging.warning("Google rejected '%s' - neutral 0.", kw)
73         return pd.Series(dtype=float) # will be made neutral later
74     logging.warning(f"429/400 for '%s' (try %d/3) - sleeping %ds", kw, n, wait)
75     time.sleep(wait); wait *= 2
76
77 def add_google_trends(df_model: pd.DataFrame,
78                       keywords      = KEYWORDS,
79                       lookback_days = LOOKBACK_DAYS,
80                       ma_window    = MA_WINDOW,
81                       include_delta = INCLUDE_DELTA,
82                       drop_const   = DROP_CONST,
83                       batch        = BATCH_SIZE) >>> pd.DataFrame:
84
85     # ensure DatetimeIndex UTC
86     if not isinstance(df_model.index, pd.DatetimeIndex):
87         if "datetime" in df_model:
88             df_model = df_model.set_index(pd.to_datetime(df_model["datetime"], utc=True))
89         else:
90             raise ValueError("df_model needs DatetimeIndex or 'datetime' col")
91
92     tf = _days_to_tf(lookback_days)
93     end = datetime.utcnow().replace(tzinfo=timezone.utc)
94     start = end - timedelta(days=lookback_days)
95     full_range = pd.date_range(start, end, freq="D", tz="UTC")
96
97     pytrends = TrendReq(hl="en-US", tz=360)
98
99     # pull & cache keyword series
100    series_dict = {}
101    for kw in keywords:
102        series_dict[kw] = _pull_one_keyword(pytrends, kw, tf)
103
104    # assemble into one DataFrame: neutral zeros where empty
105    df_trends = pd.DataFrame(
106        {_sanitize(k): (s.reindex(full_range).astype(float)) if not s.empty else 0.0
107         for k, s in series_dict.items()})
108    index=full_range
109
110    # feature engineering
111    feats = []
112    for col in df_trends:
113        s = df_trends[col]
114        feats[f"({col})_ma{ma_window}"] = s.rolling(ma_window).mean()
115        if include_delta:
116            feats[f"({col})_delta"] = s.diff()
117    df_feat = pd.DataFrame(feats).dropna()
118
119    # merge into modelling frame
120    out = df_model.copy()
121    for c in df_feat:
122        out[c] = out.index.map(df_feat[c])
123    out.update(out.filter(like=f"_ma{ma_window}").ffill())
124    if include_delta:
125        out.update(out.filter(like="_delta").fillna(0))
126
127    # optional variance filter
128    if drop_const:
129        var0 = [c for c in out.columns if out[c].nunique() <= 1]
130        if var0:
131            logging.info("Dropped %d zero-variance cols.", len(var0))
132            out = out.drop(columns=var0)
133
134    logging.info("Google-Trends done - %d features total.", len(df_feat.columns))
135    return out
136
137 # EXAMPLE (comment-out in production)
138 if __name__ == "__main__":
139     # df_model_ready must come from Steps 7 C → 8 C
140     df_model_ready = add_google_trends(df_model_ready)
141     print(f"8 D pipeline finished: df_model_ready shape → {df_model_ready.shape}")
142
143 INFO:root:Pulled : Apple stock (93 points)
144 INFO:root:Pulled : Bitcoin (93 points)
145 INFO:root:Pulled : S&P 500 (93 points)
146 INFO:root:Pulled : Interest rates (93 points)
147 INFO:root:Pulled : Inflation (93 points)
148 INFO:root:Pulled : unemployment (93 points)
149 INFO:root:Pulled : recession (93 points)
150 INFO:root:Pulled : stock market crash (93 points)
151 INFO:root:Pulled : VIX (93 points)
152 INFO:root:Pulled : buy stocks (93 points)
153 INFO:root:Pulled : crypto trading (93 points)
154 INFO:root:Dropped 29 zero-variance cols.
155 INFO:root:Google-Trends done - 22 features total.
156 8 D pipeline finished: df_model_ready shape → (4474, 101)

```

## Step 9 : Advanced Validation – Walk-Forward Testing

1. Step 9A : Implement Walk-Forward Validation
2. Step 9B : Rolling-Window Retraining & Hyperparameter Optimization

### Step 9A : Implement Walk-Forward Validation

```

1 # WALK-FORWARD VALIDATION (XGBoost ≥ 1.6, incl. 2.1)
2 import warnings, logging, inspect, numpy as np, pandas as pd, matplotlib.pyplot as plt
3 from tabm.auto import tabm
4 from sklearn.preprocessing import StandardScaler
5 from sklearn.metrics import mean_squared_error, mean_absolute_error
6 try:
7     from sklearn.metrics import mean_absolute_percentage_error
8 except ImportError:
9     def mean_absolute_percentage_error(y, p):
10         return np.mean(np.abs((y - p) / (y + 1e-9)))
11
12 import xgboost as xgb
13 from xgboost import XGBRegressor
14
15 # USER SETTINGS
16 PRICE_COL   = "spx_close"
17 TARGET_LAG  = 1
18 TARGET_NAME = "target_return"
19

```

```

20 WINDOW_TEST, STEP = 120, 24, 24
21 EARLY_ROUNDS = 50
22 GPU_PARAMS = {"tree_method": "gpu_hist"} # {} → GPU
23 RISK_FREE = 0.02 / 252 # daily RF
24
25 warnings.filterwarnings("ignore", category=UserWarning)
26 logging.basicConfig(level=logging.INFO, force=True)
27
28 # INPUT & TARGET PREP
29 if 'df_model_ready' not in globals():
30     raise RuntimeError('Run Steps 7-8 first so that df_model_ready exists.')
31
32 df_model_ready = df_model_ready.copy()
33
34 if TARGET_NAME not in df_model_ready:
35     if PRICE_COL not in df_model_ready:
36         raise KeyError(f'Column {PRICE_COL} not found - pick an existing price column.')
37     df_model_ready[TARGET_NAME] = (
38         df_model_ready[PRICE_COL].pct_change(TARGET_LAG).shift(-TARGET_LAG)
39     )
40     df_model_ready.dropna(subset=[TARGET_NAME], inplace=True)
41     logging.info("Created %s from %s pct-change (lag=%d).",
42                 TARGET_NAME, PRICE_COL, TARGET_LAG)
43
44 # XGBoost-version flags
45 VER = tuple(int(v) for v in xgb.__version__.split(".")[:2])
46 HAS_ES_PARAM = (1, 6) <= VER < (2, 0) # early_stopping_rounds arg
47 CALLBACK_IN_FIT = 'callbacks' in inspect.signature(XGBRegressor.fit).parameters
48 logging.info("XGBoost %s · ES-param: %s · callbacks-in-fit: %s",
49             xgb.__version__, HAS_ES_PARAM, CALLBACK_IN_FIT)
50
51 # helpers
52 def _rmse(y, p):
53     try:
54         return mean_squared_error(y, p, squared=False)
55     except TypeError: # old sklearn
56         return np.sqrt(mean_squared_error(y, p))
57
58 def _train_xgb(Xtr, ytr, Xval, yval, rounds):
59     """Fit an XGBRegressor with version-appropriate early stopping."""
60     scaler = StandardScaler().fit(Xtr)
61     Xtr_s = scaler.transform(Xtr)
62     Xval_s = scaler.transform(Xval)
63
64     model = XGBRegressor(
65         n_estimators=1000, learning_rate=0.05, max_depth=4,
66         subsample=0.8, colsample_bytree=0.8, random_state=42,
67         **GPU_PARAMS
68     )
69
70     if HAS_ES_PARAM: # 1.6 - 1.7
71         model.fit(
72             Xtr_s, ytr, eval_set=[(Xval_s, yval)],
73             early_stopping_rounds=rounds, verbose=False
74         )
75     elif CALLBACK_IN_FIT: # 2.0.x
76         cb = xgb.callback.EarlyStopping(rounds=rounds, save_best=True)
77         model.fit(
78             Xtr_s, ytr,
79             eval_set=[(Xval_s, yval)],
80             callbacks=[cb], verbose=False
81         )
82     else: # ≥ 2.1
83         cb = xgb.callback.EarlyStopping(rounds=rounds, save_best=True)
84         model.set_params(callbacks=[cb])
85         model.fit(
86             Xtr_s, ytr,
87             eval_set=[(Xval_s, yval)],
88             verbose=False
89         )
90     return scaler, model
91
92 # main walk-forward
93 def walk_forward(df: pd.DataFrame,
94                  features: list[str] | None = None,
95                  win=WINDOW, test=TEST, step=STEP):
96
97     # ensure DatetimeIndex
98     if not isinstance(df.index, pd.DatetimeIndex):
99         if 'datetime' in df:
100             df = df.set_index(pd.to_datetime(df['datetime'], utc=True))
101         else:
102             raise ValueError("Need DatetimeIndex or a 'datetime' column.")
103
104     if features is None:
105         features = [c for c in df.columns if c not in {TARGET_NAME, "datetime"}]
106
107     # PATCH: missing-data handling
108     # Drop any feature that is > 90% NaN
109     features = [c for c in features if df[c].isna().mean() < 0.90]
110
111     # Forward / back-fill remaining Nans
112     df[features] = df[features].ffill().bfill()
113
114     logging.info("Using %d features after cleaning.", len(features))
115
116     # Drop rows only if TARGET is NaN
117     df = df.dropna(subset=[TARGET_NAME])
118
119
120     if len(df) < win + test:
121         raise RuntimeError(f"Need ≥{win+test} rows; have {len(df)}.")
122
123     rows, bar = [], tqdm(
124         range(0, len(df) - win - test + 1, step),
125         desc="Walk-forward", unit="fold"
126     )
127     for s in bar:
128         tr, te = df.iloc[s:s+win], df.iloc[s+win:s+win+test]
129         Xtr, ytr = tr[features], tr[TARGET_NAME]
130         Xte, yte = te[features], te[TARGET_NAME]
131
132         scaler, model = _train_xgb(Xtr, ytr, Xte, yte, EARLY_ROUNDS)
133         p = model.predict(scaler.transform(Xte))
134
135         strat = np.sign(p) * yte.values
136         rows.append({
137             "start": tr.index[0],
138             "end": te.index[-1],
139             "rmse": _rmse(yte, p),
140             "mae": mean_absolute_error(yte, p),
141             "map": mean_absolute_percentage_error(yte, p),
142             "sharpe": (strat.mean() - RISK_FREE) / (strat.std(ddof=0) + 1e-9),
143             "dir_acc": (np.sign(p) == np.sign(yte)).mean(),
144             "cum_pnl": strat.sum(),
145             "best_trees": getattr(model, "best_ntree_limit", np.nan)
146         })
147
148 res = pd.DataFrame(rows)

```

```

149 logging.info("Completed %d folds.", len(res))
150 return res
151
152 # run & plot
153 metrics = walk_forward(df_model_ready)
154 display(metrics.head())
155
156 metrics.to_csv("walk_forward_metrics.csv", index=False)
157 print("metrics saved → walk_forward_metrics.csv")
158
159 def _plot(d, col, ttl, clr):
160     plt.figure(figsize=(10, 4))
161     plt.plot(d["end"], d[col], marker="o", color=clr)
162     plt.axhline(0, ls="--", lw=.8, c="grey")
163     plt.title(ttl); plt.xlabel("Fold end"); plt.ylabel(col)
164     plt.grid(True); plt.tight_layout(); plt.show()
165
166 _plot(metrics, "sharpe", "Sharpe ratio by fold", "tab:blue")
167 _plot(metrics, "cum_pnl", "Cumulative PnL by fold", "tab:green")

```

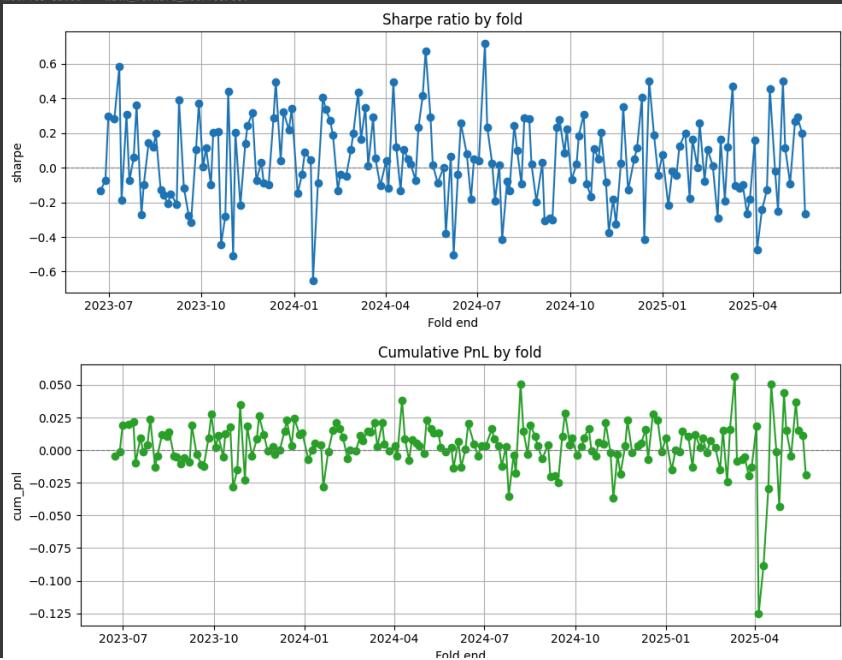
INFO:root:GBosst 2.1.4 · ES-param: False · callbacks-in-fit: False  
INFO:root:Using 98 features after cleaning.

Walk-forward: 100% [ 180/180 [00:47<00:00, 5.27fold/s]

INFO:root:Completed 180 folds.

	start	end	rmse	mae	mape	sharpe	dir acc	cum pnl	best trees
0	2023-05-31 13:00:00+00:00	2023-06-22 21:00:00+00:00	0.002221	0.001579	3.585986e+11	-0.131600	0.250000	-0.004748	NaN
1	2023-06-02 19:00:00+00:00	2023-06-27 18:00:00+00:00	0.001725	0.001372	3.700027e+11	-0.072942	0.458333	-0.001279	NaN
2	2023-06-07 16:00:00+00:00	2023-06-30 15:00:00+00:00	0.002455	0.001280	2.905812e+11	0.295675	0.541667	0.018795	NaN
3	2023-06-12 13:00:00+00:00	2023-07-06 16:00:00+00:00	0.002608	0.001275	4.713207e+11	0.283037	0.500000	0.019468	NaN
4	2023-06-14 19:00:00+00:00	2023-07-11 13:00:00+00:00	0.001530	0.001225	6.248947e+11	0.584802	0.541667	0.021584	NaN

metrics saved → walk\_forward\_metrics.csv



#### Step 9B : Rolling-Window Retraining & Hyperparameter Optimization

```

1 # Rolling-Window Retraining + Optuna Hyper-Tuning
2
3 import pandas as pd, numpy as np, warnings, logging, importlib, subprocess, sys
4 from tqdm.auto import tqdm
5 from sklearn.metrics import mean_squared_error, mean_absolute_error
6
7 # tiny helper
8 def _ensure(pkg):
9     try:
10         importlib.import_module(pkg)
11     except ImportError:
12         subprocess.check_call([sys.executable, "-m", "pip", "install", "-q", pkg])
13
14 for _p in ("optuna", "xgboost"):      # scikit-learn already present in Colab
15     _ensure(_p)
16
17 import optuna
18 from xgboost import XGBRegressor
19
20 warnings.filterwarnings("ignore", category=UserWarning)
21 logging.basicConfig(level=logging.INFO, format="%(levelname)s: %(message)s")
22
23 # version-safe RMSE (works for ALL sklearn versions)
24 def _rmse(y_true, y_pred):
25     """Root-MSE with backward-compatible call signature."""
26     return np.sqrt(mean_squared_error(y_true, y_pred))
27
28 # utilities
29 def _prepare_df(df, dt_col="datetime"):
30     if dt_col not in df.columns and isinstance(df.index, pd.DatetimeIndex):
31         df = df.reset_index().rename(columns={"index": dt_col})
32         logging.info(f"Promoted DatetimeIndex → column '{dt_col}'")
33     if dt_col not in df.columns:
34         raise ValueError("No datetime column or DatetimeIndex found.")
35     df[dt_col] = pd.to_datetime(df[dt_col], errors='coerce')
36     return df.dropna(subset=[dt_col]).reset_index(drop=True)
37
38 # main routine
39 def run_walkforward_step_9b_optuna(
40     df_model: pd.DataFrame,
41     feature_cols: list | None,
42     target_col: str,
43     datetime_col: str = "datetime",
44     window_size: int = 60,
45     test_size: int = 5,
46     step_size: int = 5,
47     n_trials: int = 20,

```

```

48     random_state : int = 42,
49     gpu : bool = False,
50   ) -> pd.DataFrame:
51
52   df = _prepare_df(df_model, datetime_col)
53
54   # feature selection
55   if not feature_cols:
56     feature_cols = []
57   avail = [c for c in feature_cols if c in df.columns]
58   if not avail:
59     avail = [c for c in df.select_dtypes(include=[np.number]).columns
60             if c != target_col]
61
62   if target_col not in df.columns:
63     raise ValueError(f"\"{target_col}\" missing from df_model_ready.")
64
65   # keep rows with valid datetime & target; fill feature NaNs with 0
66   df = (df[avail + [target_col, datetime_col]]
67         .dropna(subset=[target_col, datetime_col])
68         .reset_index(drop=True))
69   df[avail] = df[avail].fillna(0)
70
71   # guardrail for short history
72   if len(df) < window_size + test_size:
73     logging.warning(f"Only {len(df)} rows after cleaning: shrinking windows.")
74   window_size = max(5, len(df) // 3)
75   test_size = max(1, window_size // 5)
76   step_size = max(1, test_size)
77   if len(df) < window_size + test_size:
78     logging.error("Still too few rows. Add more history.")
79   return pd.DataFrame()
80
81 iterator = range(0, len(df) - window_size - test_size + 1, step_size)
82 logging.info(f"Walk-forward: {len(iterator)} fold(s) "
83             f"\n{{window_size} train, {test_size} test, step {step_size}}")
84
85 results = []
86 for start in tqdm(iterator, desc="Walk-forward", unit="fold"):
87   tr = df.iloc[start:start+window_size]
88   te = df.iloc[start+window_size:start+window_size+test_size]
89   X_tr, y_tr = tr[avail], tr[target_col]
90   X_te, y_te = te[avail], te[target_col]
91
92   # Optuna objective
93   def _objective(t):
94     p = {
95       "n_estimators" : t.suggest_int("n_estimators", 50, 300),
96       "max_depth" : t.suggest_int("max_depth", 3, 8),
97       "learning_rate" : t.suggest_float("learning_rate", 1e-2, .3, log=True),
98       "subsample" : t.suggest_float("subsample", .5, 1.0),
99       "colsample_bytree" : t.suggest_float("colsample_bytree", .5, 1.0),
100      "gamma" : t.suggest_float("gamma", 0, .5),
101      "random_state" : random_state,
102      "objective" : "reg:squarederror",
103      "verbosity" : 0
104    }
105    if gpu:
106      p |= {"tree_method": "gpu_hist", "predictor": "gpu_predictor"}
107    model = XGBRegressor(**p).fit(X_tr, y_tr)
108    preds = model.predict(X_te)
109    return _rmse(y_te, preds)
110
111 study = optuna.create_study(direction="minimize")
112 study.optimize(_objective, n_trials=n_trials, show_progress_bar=False)
113
114 best = study.best_params | {"random_state": random_state}
115 if gpu:
116   best |= {"tree_method": "gpu_hist", "predictor": "gpu_predictor"}
117
118 model = XGBRegressor(**best).fit(X_tr, y_tr)
119 preds = model.predict(X_te)
120 strat = np.sign(preds) * y_te.values
121
122 results.append({
123   "start" : tr[datetime_col].iloc[0],
124   "end" : te[datetime_col].iloc[-1],
125   "rmse" : _rmse(y_te, preds),
126   "mae" : mean_absolute_error(y_te, preds),
127   "sharpe" : np.mean(strat) / (np.std(strat) + 1e-9),
128   "directional_accuracy": (np.sign(preds) == np.sign(y_te)).mean(),
129   "avg_pnl" : strat.mean(),
130   "best_params" : best
131 })
132
133 return pd.DataFrame(results)
134
135 # RUN
136 feature_cols, target_col = [], "target_return"
137
138 df_metrics_optuna = run_walkforward_step_9b_optuna(
139   df_model = df_model_ready,
140   feature_cols = feature_cols,
141   target_col = target_col,
142   window_size = 60,
143   test_size = 5,
144   step_size = 5,
145   n_trials = 20,      # consider 10-15 if runtime is long
146   gpu = False,
147 )
148
149 print(f"Rows in df_metrics_optuna: {len(df_metrics_optuna)}")
150 if not df_metrics_optuna.empty:
151   display(df_metrics_optuna.head())

```

INFO:root:Walk-Forward: 879 fold(s) (60 train, 5 test, step 5)

Walk-forward: 100% 879/879 [47:05&lt;00:00, 3.41s/fold]

스트리밍 출력 내용은 깊이에 미자막 500줄이 삭제되었습니다.

[1] 2025-05-01 02:36:30.856 Trial 18 finished with value: 0.00136487210857359 and parameters: {'n\_estimators': 217, 'max\_depth': 7, 'learning\_rate': 0.1645881951949726, 'subsample': 0.800637550923414, 'colsample\_bytree': 0.93879148  
[1] 2025-05-01 02:36:31.021 Trial 19 finished with value: 0.001424203637666766 and parameters: {'n\_estimators': 280, 'max\_depth': 5, 'learning\_rate': 0.09494846565162779, 'subsample': 0.5795163259215439, 'colsample\_bytree': 0.9910  
[1] 2025-05-01 02:36:31.178 A new study created in memory with name: no-name-4fc5f06-cd0-4034-a0a6-b6ed6a5b5c4a  
[1] 2025-05-01 02:36:31.304 Trial 0 finished with value: 0.00115931831812507 and parameters: {'n\_estimators': 138, 'max\_depth': 3, 'learning\_rate': 0.1690518241479651, 'subsample': 0.9718646406572924, 'colsample\_bytree': 0.992774  
[1] 2025-05-01 02:36:31.409 Trial 1 finished with value: 0.0011837391533767 and parameters: {'n\_estimators': 188, 'max\_depth': 4, 'learning\_rate': 0.012171975908814588, 'subsample': 0.6691593072069075, 'colsample\_bytree': 0.78703  
[1] 2025-05-01 02:36:31.500 Trial 2 finished with value: 0.0011635491256268642 and parameters: {'n\_estimators': 149, 'max\_depth': 5, 'learning\_rate': 0.27648544514574075, 'subsample': 0.78142812076034, 'colsample\_bytree': 0.6976510  
[1] 2025-05-01 02:36:31.594 Trial 3 finished with value: 0.00110651869703805 and parameters: {'n\_estimators': 158, 'max\_depth': 4, 'learning\_rate': 0.01993514748937617, 'subsample': 0.6246222998084517, 'colsample\_bytree': 0.52460  
[1] 2025-05-01 02:36:31.756 Trial 4 finished with value: 0.001154783806293912 and parameters: {'n\_estimators': 237, 'max\_depth': 3, 'learning\_rate': 0.027648544514574075, 'subsample': 0.78142812076034, 'colsample\_bytree': 0.94166  
[1] 2025-05-01 02:36:31.837 Trial 5 finished with value: 0.0010382318551584676 and parameters: {'n\_estimators': 110, 'max\_depth': 8, 'learning\_rate': 0.05528590213561829, 'subsample': 0.521106598020451, 'colsample\_bytree': 0.88494  
[1] 2025-05-01 02:36:31.956 Trial 6 finished with value: 0.0011804263257549607 and parameters: {'n\_estimators': 220, 'max\_depth': 7, 'learning\_rate': 0.03269870080246246, 'subsample': 0.5678979311447289, 'colsample\_bytree': 0.872  
[1] 2025-05-01 02:36:32.063 Trial 7 finished with value: 0.001207512598853422 and parameters: {'n\_estimators': 195, 'max\_depth': 6, 'learning\_rate': 0.248742511405050424, 'subsample': 0.6360624235237358, 'colsample\_bytree': 0.58431  
[1] 2025-05-01 02:36:32.204 Trial 8 finished with value: 0.0011618521926617057 and parameters: {'n\_estimators': 300, 'max\_depth': 8, 'learning\_rate': 0.061583224921001584, 'subsample': 0.933697274923103, 'colsample\_bytree': 0.97342  
[1] 2025-05-01 02:36:32.318 Trial 9 finished with value: 0.0011516866741182 and parameters: {'n\_estimators': 201, 'max\_depth': 3, 'learning\_rate': 0.17965440107412875, 'subsample': 0.978903215280032, 'colsample\_bytree': 0.981112  
[1] 2025-05-01 02:36:32.401 Trial 10 finished with value: 0.001082458695882809 and parameters: {'n\_estimators': 60, 'max\_depth': 7, 'learning\_rate': 0.0314808431619095, 'subsample': 0.5065456248312457, 'colsample\_bytree': 0.82757  
[1] 2025-05-01 02:36:32.484 Trial 11 finished with value: 0.001083487316276407 and parameters: {'n\_estimators': 58, 'max\_depth': 7, 'learning\_rate': 0.03089024309076983, 'subsample': 0.513971092478512, 'colsample\_bytree': 0.821635  
[1] 2025-05-01 02:36:32.569 Trial 12 finished with value: 0.00108263257549607 and parameters: {'n\_estimators': 62, 'max\_depth': 7, 'learning\_rate': 0.0376625163075562, 'subsample': 0.5099649480517215, 'colsample\_bytree': 0.84449  
[1] 2025-05-01 02:36:32.672 Trial 13 finished with value: 0.00115084764764811 and parameters: {'n\_estimators': 100, 'max\_depth': 7, 'learning\_rate': 0.0780405059169698, 'subsample': 0.71862307976446486, 'colsample\_bytree': 0.9078  
[1] 2025-05-01 02:36:32.791 Trial 14 finished with value: 0.001153418603295937 and parameters: {'n\_estimators': 103, 'max\_depth': 8, 'learning\_rate': 0.05528590213561829, 'subsample': 0.521106598020451, 'colsample\_bytree': 0.88494  
[1] 2025-05-01 02:36:32.894 Trial 15 finished with value: 0.001059333668476523 and parameters: {'n\_estimators': 102, 'max\_depth': 6, 'learning\_rate': 0.061583224921001584, 'subsample': 0.9232045451929287, 'colsample\_bytree': 0.9078  
[1] 2025-05-01 02:36:32.984 Trial 16 finished with value: 0.001113750935027132 and parameters: {'n\_estimators': 78, 'max\_depth': 7, 'learning\_rate': 0.027648544514574075, 'subsample': 0.5768770450152646, 'colsample\_bytree': 0.89282  
[1] 2025-05-01 02:36:33.094 Trial 17 finished with value: 0.001162902243123307 and parameters: {'n\_estimators': 128, 'max\_depth': 6, 'learning\_rate': 0.01284995376720649, 'subsample': 0.842051929584803, 'colsample\_bytree': 0.645  
[1] 2025-05-01 02:36:33.188 Trial 18 finished with value: 0.00113766052856972 and parameters: {'n\_estimators': 81, 'max\_depth': 8, 'learning\_rate': 0.02149079079806135, 'subsample': 0.704847485604248, 'colsample\_bytree': 0.7879  
[1] 2025-05-01 02:36:33.292 Trial 19 finished with value: 0.00105107688084373 and parameters: {'n\_estimators': 118, 'max\_depth': 7, 'learning\_rate': 0.1074750816534873, 'subsample': 0.5001549406160555, 'colsample\_bytree': 0.8308  
[1] 2025-05-01 02:36:33.388 A new study created in memory with name: no-name-cbebf13-fc47-4fb7-9e0d-2ef73739250  
[1] 2025-05-01 02:36:33.506 Trial 0 finished with value: 0.001283288754393848 and parameters: {'n\_estimators': 193, 'max\_depth': 3, 'learning\_rate': 0.04190201377167322, 'subsample': 0.6950019364738413, 'colsample\_bytree': 0.87167  
[1] 2025-05-01 02:36:33.638 Trial 1 finished with value: 0.0012668432559214142 and parameters: {'n\_estimators': 161, 'max\_depth': 5, 'learning\_rate': 0.10318193720643586, 'subsample': 0.784963278755257, 'colsample\_bytree': 0.998692  
[1] 2025-05-01 02:36:33.814 Trial 2 finished with value: 0.00120902846696731 and parameters: {'n\_estimators': 288, 'max\_depth': 5, 'learning\_rate': 0.2320420591231987, 'subsample': 0.840088542361217, 'colsample\_bytree': 0.981395  
[1] 2025-05-01 02:36:33.962 Trial 3 finished with value: 0.001231206170488767 and parameters: {'n\_estimators': 291, 'max\_depth': 8, 'learning\_rate': 0.2759159472146337, 'subsample': 0.8525290613616285, 'colsample\_bytree': 0.799557  
[1] 2025-05-01 02:36:33.984 Trial 4 finished with value: 0.00127901779856109 and parameters: {'n\_estimators': 136, 'max\_depth': 6, 'learning\_rate': 0.0185893361322071027, 'subsample': 0.7378420094791496, 'colsample\_bytree': 0.62085  
[1] 2025-05-01 02:36:34.173 Trial 5 finished with value: 0.0013902703433467109 and parameters: {'n\_estimators': 213, 'max\_depth': 5, 'learning\_rate': 0.18571271233598782, 'subsample': 0.6359640644955028, 'colsample\_bytree': 0.754314  
[1] 2025-05-01 02:36:34.276 Trial 6 finished with value: 0.0010430291218337833 and parameters: {'n\_estimators': 171, 'max\_depth': 6, 'learning\_rate': 0.10309348081813927, 'subsample': 0.91996882283116, 'colsample\_bytree': 0.95466  
[1] 2025-05-01 02:36:34.375 Trial 7 finished with value: 0.001240149663203034 and parameters: {'n\_estimators': 150, 'max\_depth': 8, 'learning\_rate': 0.20436529556694976, 'subsample': 0.6966001497675969, 'colsample\_bytree': 0.80795  
[1] 2025-05-01 02:36:34.467 Trial 8 finished with value: 0.001285345870179423 and parameters: {'n\_estimators': 139, 'max\_depth': 3, 'learning\_rate': 0.03748300742339585, 'subsample': 0.935769254297361, 'colsample\_bytree': 0.9572  
[1] 2025-05-01 02:36:34.593 Trial 9 finished with value: 0.00120302624616287 and parameters: {'n\_estimators': 244, 'max\_depth': 6, 'learning\_rate': 0.04868300742329059, 'subsample': 0.582830313148518, 'colsample\_bytree': 0.62933  
[1] 2025-05-01 02:36:34.677 Trial 10 finished with value: 0.001294253211386605 and parameters: {'n\_estimators': 57, 'max\_depth': 7, 'learning\_rate': 0.0293825160877382, 'subsample': 0.5055513572992297, 'colsample\_bytree': 0.5008  
[1] 2025-05-01 02:36:34.876 Trial 11 finished with value: 0.001190612531337797 and parameters: {'n\_estimators': 299, 'max\_depth': 4, 'learning\_rate': 0.0708752184654211, 'subsample': 0.549272337426236, 'colsample\_bytree': 0.64461  
[1] 2025-05-01 02:36:35.019 Trial 12 finished with value: 0.00121006335761576826 and parameters: {'n\_estimators': 234, 'max\_depth': 4, 'learning\_rate': 0.0718887931144793526, 'subsample': 0.5341302662962119, 'colsample\_bytree': 0.6562  
[1] 2025-05-01 02:36:35.169 Trial 13 finished with value: 0.001233639827178365 and parameters: {'n\_estimators': 257, 'max\_depth': 4, 'learning\_rate': 0.02516173936136355, 'subsample': 0.585870572018377, 'colsample\_bytree': 0.637  
[1] 2025-05-01 02:36:35.316 Trial 14 finished with value: 0.0011918457905165552 and parameters: {'n\_estimators': 247, 'max\_depth': 4, 'learning\_rate': 0.0617898001204088, 'subsample': 0.6025903867427824, 'colsample\_bytree': 0.54400  
[1] 2025-05-01 02:36:35.486 Trial 15 finished with value: 0.0012028552760780473 and parameters: {'n\_estimators': 299, 'max\_depth': 4, 'learning\_rate': 0.0751374546231473, 'subsample': 0.626600215072188, 'colsample\_bytree': 0.51516  
[1] 2025-05-01 02:36:35.636 Trial 16 finished with value: 0.00123890642425475 and parameters: {'n\_estimators': 267, 'max\_depth': 3, 'learning\_rate': 0.1147730137330218, 'subsample': 0.565419459010133, 'colsample\_bytree': 0.5783  
[1] 2025-05-01 02:36:35.778 Trial 17 finished with value: 0.00130285337525135 and parameters: {'n\_estimators': 214, 'max\_depth': 4, 'learning\_rate': 0.02772112147079569, 'subsample': 0.6510727013856534, 'colsample\_bytree': 0.7107  
[1] 2025-05-01 02:36:35.953 Trial 18 finished with value: 0.00135798868503337 and parameters: {'n\_estimators': 268, 'max\_depth': 4, 'learning\_rate': 0.0681748919393365, 'subsample': 0.503478898612586, 'colsample\_bytree': 0.5548  
[1] 2025-05-01 02:36:36.053 Trial 19 finished with value: 0.0013047843357106337 and parameters: {'n\_estimators': 92, 'max\_depth': 3, 'learning\_rate': 0.15114869682195875, 'subsample': 0.985283279293306, 'colsample\_bytree': 0.69427  
[1] 2025-05-01 02:36:36.202 A new study created in memory with name: no-name-5d301829-5ed5-4284-a1fc-a5294c61b64  
[1] 2025-05-01 02:36:36.306 Trial 0 finished with value: 0.00148218908278411 and parameters: {'n\_estimators': 96, 'max\_depth': 7, 'learning\_rate': 0.010354659540572115, 'subsample': 0.950844591017847, 'colsample\_bytree': 0.98520  
[1] 2025-05-01 02:36:36.426 Trial 1 finished with value: 0.00137447567839567, 'subsample': 0.807833560047339, 'colsample\_bytree': 0.889905  
[1] 2025-05-01 02:36:36.494 Trial 2 finished with value: 0.0014954821654165928 and parameters: {'n\_estimators': 210, 'max\_depth': 8, 'learning\_rate': 0.026153485562234357, 'subsample': 0.840809200224673, 'colsample\_bytree': 0.730768  
[1] 2025-05-01 02:36:36.688 Trial 4 finished with value: 0.001490473432712242 and parameters: {'n\_estimators': 215, 'max\_depth': 4, 'learning\_rate': 0.01859935332071027, 'subsample': 0.9739292121993307, 'colsample\_bytree': 0.99463  
[1] 2025-05-01 02:36:36.823 Trial 5 finished with value: 0.001025309509273533 and parameters: {'n\_estimators': 297, 'max\_depth': 8, 'learning\_rate': 0.02718396593251153, 'subsample': 0.7941725090417223, 'colsample\_bytree': 0.54489  
[1] 2025-05-01 02:36:36.929 Trial 6 finished with value: 0.001477911445901646 and parameters: {'n\_estimators': 257, 'max\_depth': 3, 'learning\_rate': 0.09326378512943567, 'subsample': 0.656052874076266, 'colsample\_bytree': 0.879511  
[1] 2025-05-01 02:36:37.059 Trial 7 finished with value: 0.001594551540653 and parameters: {'n\_estimators': 251, 'max\_depth': 8, 'learning\_rate': 0.0417898001204088, 'subsample': 0.6025903867427824, 'colsample\_bytree': 0.54400  
[1] 2025-05-01 02:36:37.197 Trial 14 finished with value: 0.00138757470917062 and parameters: {'n\_estimators': 290, 'max\_depth': 6, 'learning\_rate': 0.121148565203811, 'subsample': 0.675247748585398, 'colsample\_bytree': 0.7143446  
[1] 2025-05-01 02:36:37.314 Trial 9 finished with value: 0.001582734062372847 and parameters: {'n\_estimators': 214, 'max\_depth': 4, 'learning\_rate': 0.01232364776708962, 'subsample': 0.602231173865203, 'colsample\_bytree': 0.6913  
[1] 2025-05-01 02:36:37.430 Trial 10 finished with value: 0.001551284966797588 and parameters: {'n\_estimators': 151, 'max\_depth': 5, 'learning\_rate': 0.26963551916181684, 'subsample': 0.602231173865203, 'colsample\_bytree': 0.5765  
[1] 2025-05-01 02:36:37.541 Trial 11 finished with value: 0.0014913652574030014 and parameters: {'n\_estimators': 135, 'max\_depth': 3, 'learning\_rate': 0.1051749140930825, 'subsample': 0.882213039917208, 'colsample\_bytree': 0.83060  
[1] 2025-05-01 02:36:37.697 Trial 12 finished with value: 0.001507266779751745 and parameters: {'n\_estimators': 273, 'max\_depth': 3, 'learning\_rate': 0.13442345158245103, 'subsample': 0.6501381629411033, 'colsample\_bytree': 0.8388  
[1] 2025-05-01 02:36:37.868 Trial 13 finished with value: 0.001698342310315663 and parameters: {'n\_estimators': 116, 'max\_depth': 5, 'learning\_rate': 0.16877383130309245, 'subsample': 0.505080364332063, 'colsample\_bytree': 0.9042  
[1] 2025-05-01 02:36:37.975 Trial 14 finished with value: 0.001578930311661661 and parameters: {'n\_estimators': 171, 'max\_depth': 7, 'learning\_rate': 0.05325545527584164, 'subsample': 0.720365894941228, 'colsample\_bytree': 0.7766  
[1] 2025-05-01 02:36:38.098 Trial 15 finished with value: 0.001568343147785945 and parameters: {'n\_estimators': 177, 'max\_depth': 4, 'learning\_rate': 0.08169481028238093, 'subsample': 0.867484781069132953, 'colsample\_bytree': 0.6384  
[1] 2025-05-01 02:36:38.248 Trial 16 finished with value: 0.001512028711035933 and parameters: {'n\_estimators': 250, 'max\_depth': 6, 'learning\_rate': 0.198781440621404, 'subsample': 0.90345036121404, 'colsample\_bytree': 0.909309  
[1] 2025-05-01 02:36:38.419 Trial 17 finished with value: 0.001489920887449849 and parameters: {'n\_estimators': 300, 'max\_depth': 5, 'learning\_rate': 0.093447813850076214, 'subsample': 0.764591548773028, 'colsample\_bytree': 0.72951  
[1] 2025-05-01 02:36:38.504 Trial 18 finished with value: 0.0014894489538249157 and parameters: {'n\_estimators': 56, 'max\_depth': 7, 'learning\_rate': 0.04738688196524752, 'subsample': 0.659531385060468, 'colsample\_bytree': 0.504509  
[1] 2025-05-01 02:36:38.619 Trial 19 finished with value: 0.00148068739905498194 and parameters: {'n\_estimators': 149, 'max\_depth': 3, 'learning\_rate': 0.1319078800110866, 'subsample': 0.994197115609233, 'colsample\_bytree': 0.8656  
[1] 2025-05-01 02:36:38.766 A new study created in memory with name: no-name-de40746-295-4cd0-ac8d-2530407625  
[1] 2025-05-01 02:36:38.931 Trial 0 finished with value: 0.0021069028213768323 and parameters: {'n\_estimators': 250, 'max\_depth': 8, 'learning\_rate': 0.099523193646766, 'subsample': 0.7669299603463955, 'colsample\_bytree': 0.644782  
[1] 2025-05-01 02:36:39.047 Trial 1 finished with value: 0.00208921765762494 and parameters: {'n\_estimators': 101, 'max\_depth': 8, 'learning\_rate': 0.0214516426194761, 'subsample': 0.6388946235754476, 'colsample\_bytree': 0.58208  
[1] 2025-05-01 02:36:39.222 Trial 2 finished with value: 0.002054843914547438 and parameters: {'n\_estimators': 125, 'max\_depth': 4, 'learning\_rate': 0.020345630523282987 and parameters: {'n\_estimators': 210, 'max\_depth': 8, 'learning\_rate': 0.0895769612728672, 'subsample': 0.5736288774474979, 'colsample\_bytree': 0.87163580  
[1] 2025-05-01 02:36:41.441 Trial 7 finished with value: 0.002065630352282987 and parameters: {'n\_estimators': 198, 'max\_depth': 8, 'learning\_rate': 0.0895769612728672, 'subsample': 0.9378574601309498, 'colsample\_bytree': 0.707195  
[1] 2025-05-01 02:36:41.644 Trial 8 finished with value: 0.0020950307075538 and parameters: {'n\_estimators': 239, 'max\_depth': 7, 'learning\_rate': 0.2893564551711571, 'subsample': 0.60514536136614316, 'colsample\_bytree': 0.7676881  
[1] 2025-05-01 02:36:41.797 Trial 9 finished with value: 0.0020677242899465343 and parameters: {'n\_estimators': 56, 'max\_depth': 7, 'learning\_rate': 0.0156582399059991, 'subsample': 0.799677091002967, 'colsample\_bytree': 0.5718810  
[1] 2025-05-01 02:36:42.053 Trial 10 finished with value: 0.00203853082657726 and parameters: {'n\_estimators': 141, 'max\_depth': 3, 'learning\_rate': 0.038833208810952, 'subsample': 0.881538063797215, 'colsample\_bytree': 0.9800  
[1] 2025-05-01 02:36:42.193 Trial 11 finished with value: 0.00207612432173192 and parameters: {'n\_estimators': 51, 'max\_depth': 5, 'learning\_rate': 0.106302259322523795, 'subsample': 0.51041811445281, 'colsample\_bytree': 0.85574  
[1] 2025-05-01 02:36:42.216 Trial 12 finished with value: 0.00206266929989449 and parameters: {'n\_estimators': 97, 'max\_depth': 3, 'learning\_rate': 0.14363598016921, 'subsample': 0.838746122997709, 'colsample\_bytree': 0.8236541  
[1] 2025-05-01 02:36:42.376 Trial 13 finished with value: 0.002070685381347494 and parameters: {'n\_estimators': 297, 'max\_depth': 5, 'learning\_rate': 0.02732482198065352, 'subsample': 0.729612352302114, 'colsample\_bytree': 0.5252  
[1] 2025-05-01 02:36:42.498 Trial 14 finished with value: 0.0020882631079032683 and parameters: {'n\_estimators': 162, 'max\_depth': 6, 'learning\_rate': 0.08596691955263036, 'subsample': 0.9067410843204642, 'colsample\_bytree': 0.9632  
[1] 2025-05-01 02:36:42.593 Trial 15 finished with value: 0.0020808673269721556 and parameters: {'n\_estimators': 86, 'max\_depth': 4, 'learning\_rate': 0.1873498515686853, 'subsample': 0.7196432611230165, 'colsample\_bytree': 0.75505  
[1] 2025-05-01 02:36:42.717 Trial 16 finished with value: 0.0020874683405415325 and parameters: {'n\_estimators': 142, 'max\_depth': 7, 'learning\_rate': 0.0670598760362978, 'subsample': 0.79611818494871, 'colsample\_bytree': 0.92248  
[1] 2025-05-01 02:36:42.812 Trial 17 finished with value: 0.002132359121071471 and parameters: {'n\_estimators': 72, 'max\_depth': 5, 'learning\_rate': 0.1333947274327495, 'subsample': 0.548943903851543, 'colsample\_bytree': 0.696338  
[1] 2025-05-01 02:36:43.919 Trial 18 finished with value: 0.0020596171170472748 and parameters: {'n\_estimators': 118, 'max\_depth': 3, 'learning\_rate': 0.030379303174995, 'subsample': 0.7049738963979, 'colsample\_bytree': 0.5079  
[1] 2025-05-01 02:36:43.959 Trial 19 finished with value: 0.002032391752174087 and parameters: {'n\_estimators': 174, 'max\_depth': 6, 'learning\_rate': 0.0165383

[1] 2025-05-01 02:36:48,466 Trial 1 finished with value: 0.001809260739149302 and parameters: {'n\_estimators': 100, 'max\_depth': 6, 'learning\_rate': 0.0008680370232798197, 'subsample': 0.804107833609491, 'columsample\_bytree': 0.6443400}

[1] 2025-05-01 02:36:48,531 Trial 2 finished with value: 0.001812826208123887 and parameters: {'n\_estimators': 55, 'max\_depth': 6, 'learning\_rate': 0.02682565213981906, 'subsample': 0.7833916869523785, 'columsample\_bytree': 0.92740}

[1] 2025-05-01 02:36:48,708 Trial 4 finished with value: 0.00177604077301182 and parameters: {'n\_estimators': 218, 'max\_depth': 6, 'learning\_rate': 0.04248695520166507, 'subsample': 0.745146140669491, 'columsample\_bytree': 0.904074}

[1] 2025-05-01 02:36:48,778 Trial 5 finished with value: 0.0017898750412248766 and parameters: {'n\_estimators': 69, 'max\_depth': 8, 'learning\_rate': 0.0592119091230501, 'subsample': 0.6130822575188295, 'columsample\_bytree': 0.51159}

[1] 2025-05-01 02:36:48,896 Trial 6 finished with value: 0.00170603396195379 and parameters: {'n\_estimators': 225, 'max\_depth': 8, 'learning\_rate': 0.075237899310806, 'subsample': 0.7327549584048658, 'columsample\_bytree': 0.593681}

[1] 2025-05-01 02:36:49,003 Trial 7 finished with value: 0.00181355660684864 and parameters: {'n\_estimators': 175, 'max\_depth': 7, 'learning\_rate': 0.01263139290481015, 'subsample': 0.7799265087815916, 'columsample\_bytree': 0.87011}

[1] 2025-05-01 02:36:49,132 Trial 8 finished with value: 0.00188513004465897 and parameters: {'n\_estimators': 261, 'max\_depth': 3, 'learning\_rate': 0.017285711046757506, 'subsample': 0.5396433886394023, 'columsample\_bytree': 0.77248}

[1] 2025-05-01 02:36:49,231 Trial 9 finished with value: 0.0017579186849639209 and parameters: {'n\_estimators': 171, 'max\_depth': 8, 'learning\_rate': 0.0099395333128908, 'subsample': 0.5451937604490138, 'columsample\_bytree': 0.76459}

[1] 2025-05-01 02:36:49,427 Trial 10 finished with value: 0.001801010535527672 and parameters: {'n\_estimators': 298, 'max\_depth': 8, 'learning\_rate': 0.1247548949258655, 'subsample': 0.986502430036544, 'columsample\_bytree': 0.65285}

[1] 2025-05-01 02:36:49,573 Trial 11 finished with value: 0.0017759388161323 and parameters: {'n\_estimators': 213, 'max\_depth': 8, 'learning\_rate': 0.2799693194574768, 'subsample': 0.92023885073256, 'columsample\_bytree': 0.76927}

[1] 2025-05-01 02:36:49,720 Trial 12 finished with value: 0.00177134598523788 and parameters: {'n\_estimators': 228, 'max\_depth': 7, 'learning\_rate': 0.2978680807193993, 'subsample': 0.92797901023491172, 'columsample\_bytree': 0.71074}

[1] 2025-05-01 02:36:49,863 Trial 13 finished with value: 0.001775605089179134 and parameters: {'n\_estimators': 233, 'max\_depth': 7, 'learning\_rate': 0.16857190341587175, 'subsample': 0.880118115283972, 'columsample\_bytree': 0.80733}

[1] 2025-05-01 02:36:49,995 Trial 14 finished with value: 0.0018731290886701282 and parameters: {'n\_estimators': 192, 'max\_depth': 8, 'learning\_rate': 0.296535273614802, 'subsample': 0.8629504056758404, 'columsample\_bytree': 0.69897}

[1] 2025-05-01 02:36:50,147 Trial 15 finished with value: 0.00180233793536246 and parameters: {'n\_estimators': 267, 'max\_depth': 7, 'learning\_rate': 0.16471746210149205, 'subsample': 0.992138715463973, 'columsample\_bytree': 0.8374}

[1] 2025-05-01 02:36:50,291 Trial 16 finished with value: 0.00182369629562387 and parameters: {'n\_estimators': 206, 'max\_depth': 5, 'learning\_rate': 0.03284752345215045, 'subsample': 0.6909415765240771, 'columsample\_bytree': 0.572}

[1] 2025-05-01 02:36:50,440 Trial 17 finished with value: 0.001794884364560935 and parameters: {'n\_estimators': 146, 'max\_depth': 8, 'learning\_rate': 0.1051352266275284, 'subsample': 0.8603236266285288, 'columsample\_bytree': 0.71741}

[1] 2025-05-01 02:36:50,595 Trial 18 finished with value: 0.001773932314626593 and parameters: {'n\_estimators': 255, 'max\_depth': 7, 'learning\_rate': 0.0403991069729294, 'subsample': 0.922347968163854, 'columsample\_bytree': 0.99910}

[1] 2025-05-01 02:36:50,703 Trial 19 finished with value: 0.0017722497920742194 and parameters: {'n\_estimators': 133, 'max\_depth': 3, 'learning\_rate': 0.1955278916028734, 'subsample': 0.6751960016850942, 'columsample\_bytree': 0.65353}

[1] 2025-05-01 02:36:50,827 A new study created in memory with name: no-name-a82e1a99-3470-4882-accd-0f5347cd0b

[1] 2025-05-01 02:36:50,967 Trial 0 finished with value: 0.0012537833101140808 and parameters: {'n\_estimators': 188, 'max\_depth': 4, 'learning\_rate': 0.01008966354233118, 'subsample': 0.720413898535491, 'columsample\_bytree': 0.9151}

[1] 2025-05-01 02:36:51,051 Trial 1 finished with value: 0.001264724263979806 and parameters: {'n\_estimators': 136, 'max\_depth': 6, 'learning\_rate': 0.04221849714313734, 'subsample': 0.87892046966979326, 'columsample\_bytree': 0.5929}

[1] 2025-05-01 02:36:51,172 Trial 2 finished with value: 0.0012704675286209562 and parameters: {'n\_estimators': 226, 'max\_depth': 7, 'learning\_rate': 0.0703242472923323, 'subsample': 0.782613478897391, 'columsample\_bytree': 0.52706}

[1] 2025-05-01 02:36:51,268 Trial 3 finished with value: 0.00120503284620826 and parameters: {'n\_estimators': 151, 'max\_depth': 4, 'learning\_rate': 0.02284057805182974, 'subsample': 0.53633707315605, 'columsample\_bytree': 0.9071}

[1] 2025-05-01 02:36:51,404 Trial 4 finished with value: 0.0012704651614079 and parameters: {'n\_estimators': 244, 'max\_depth': 4, 'learning\_rate': 0.01640384100597941, 'subsample': 0.817694727934345, 'columsample\_bytree': 0.7380711}

[1] 2025-05-01 02:36:51,521 Trial 5 finished with value: 0.001755964076705604 and parameters: {'n\_estimators': 223, 'max\_depth': 8, 'learning\_rate': 0.0703242472923323, 'subsample': 0.743828744724851, 'columsample\_bytree': 0.63114}

[1] 2025-05-01 02:36:51,653 Trial 6 finished with value: 0.00115628408078882 and parameters: {'n\_estimators': 274, 'max\_depth': 6, 'learning\_rate': 0.0254025660411333, 'subsample': 0.50897944253918, 'columsample\_bytree': 0.593822}

[1] 2025-05-01 02:36:51,754 Trial 7 finished with value: 0.00178865514462568 and parameters: {'n\_estimators': 158, 'max\_depth': 6, 'learning\_rate': 0.10685514406748488, 'subsample': 0.91369087195068, 'columsample\_bytree': 0.91378}

[1] 2025-05-01 02:36:51,896 Trial 8 finished with value: 0.001279231686936156 and parameters: {'n\_estimators': 295, 'max\_depth': 6, 'learning\_rate': 0.02387049702794957, 'subsample': 0.89461394194375, 'columsample\_bytree': 0.5626}

[1] 2025-05-01 02:36:51,988 Trial 9 finished with value: 0.0012496076363865029 and parameters: {'n\_estimators': 144, 'max\_depth': 6, 'learning\_rate': 0.0102164522844153, 'subsample': 0.6058265646397658, 'columsample\_bytree': 0.7245}

[1] 2025-05-01 02:36:52,089 Trial 10 finished with value: 0.0010544919232374 and parameters: {'n\_estimators': 57, 'max\_depth': 3, 'learning\_rate': 0.292852742143159, 'subsample': 0.616416502274061, 'columsample\_bytree': 0.6505230}

[1] 2025-05-01 02:36:52,200 Trial 11 finished with value: 0.00125673471501224 and parameters: {'n\_estimators': 51, 'max\_depth': 3, 'learning\_rate': 0.27585956517315392, 'subsample': 0.6102094240185598, 'columsample\_bytree': 0.668416}

[1] 2025-05-01 02:36:52,328 Trial 12 finished with value: 0.00126487104347417 and parameters: {'n\_estimators': 51, 'max\_depth': 5, 'learning\_rate': 0.25714173030301356, 'subsample': 0.5053171313970878, 'columsample\_bytree': 0.82424}

[1] 2025-05-01 02:36:52,904 Trial 13 finished with value: 0.001230943816261241 and parameters: {'n\_estimators': 96, 'max\_depth': 3, 'learning\_rate': 0.1272646470431952, 'subsample': 0.6515344124312857, 'columsample\_bytree': 0.66621}

[1] 2025-05-01 02:36:53,174 Trial 14 finished with value: 0.0018812606172839 and parameters: {'n\_estimators': 281, 'max\_depth': 8, 'learning\_rate': 0.1596044022357663, 'subsample': 0.696215402262894, 'columsample\_bytree': 0.514582}

[1] 2025-05-01 02:36:53,664 Trial 15 finished with value: 0.00127088075175136 and parameters: {'n\_estimators': 86, 'max\_depth': 5, 'learning\_rate': 0.1500591179374475, 'subsample': 0.5644032142129431, 'columsample\_bytree': 0.81090}

[1] 2025-05-01 02:36:54,190 Trial 16 finished with value: 0.00125189838911003 and parameters: {'n\_estimators': 191, 'max\_depth': 7, 'learning\_rate': 0.1059010605639348, 'subsample': 0.5802424561963987, 'columsample\_bytree': 0.6099}

[1] 2025-05-01 02:36:54,511 Trial 17 finished with value: 0.001903238273863 and parameters: {'n\_estimators': 264, 'max\_depth': 5, 'learning\_rate': 0.2958638952530105, 'subsample': 0.686293493769625, 'columsample\_bytree': 0.6892}

[1] 2025-05-01 02:36:54,691 Trial 18 finished with value: 0.0012863971335319321 and parameters: {'n\_estimators': 116, 'max\_depth': 7, 'learning\_rate': 0.19629491368753635, 'subsample': 0.9154539104735808, 'columsample\_bytree': 0.7992}

[1] 2025-05-01 02:36:54,966 Trial 19 finished with value: 0.00125689490824761 and parameters: {'n\_estimators': 216, 'max\_depth': 3, 'learning\_rate': 0.092729888414744, 'subsample': 0.7090617667224484, 'columsample\_bytree': 0.5637}

[1] 2025-05-01 02:36:55,068 A new study created in memory with name: no-name-f087f0aa-b5a5-4f19-8e04-771fb5d95da

[1] 2025-05-01 02:36:55,224 Trial 0 finished with value: 0.00565477917971474 and parameters: {'n\_estimators': 255, 'max\_depth': 8, 'learning\_rate': 0.239811471019049315, 'subsample': 0.5933950062578914, 'columsample\_bytree': 0.993347}

[1] 2025-05-01 02:36:55,287 Trial 1 finished with value: 0.005481509562727009 and parameters: {'n\_estimators': 52, 'max\_depth': 8, 'learning\_rate': 0.03527224445617674, 'subsample': 0.8481383875651577, 'columsample\_bytree': 0.8180244}

[1] 2025-05-01 02:36:55,396 Trial 2 finished with value: 0.0054907172139425 and parameters: {'n\_estimators': 198, 'max\_depth': 8, 'learning\_rate': 0.030324525387511954, 'subsample': 0.643564190310837, 'columsample\_bytree': 0.62821}

[1] 2025-05-01 02:36:55,531 Trial 3 finished with value: 0.005111138307932959 and parameters: {'n\_estimators': 267, 'max\_depth': 6, 'learning\_rate': 0.016344374431396, 'subsample': 0.7347394960129233, 'columsample\_bytree': 0.981487}

[1] 2025-05-01 02:36:55,694 Trial 4 finished with value: 0.005273252482248351 and parameters: {'n\_estimators': 283, 'max\_depth': 7, 'learning\_rate': 0.05457018019479707, 'subsample': 0.592627147111016, 'columsample\_bytree': 0.559800}

[1] 2025-05-01 02:36:55,791 Trial 5 finished with value: 0.00511603597390696 and parameters: {'n\_estimators': 165, 'max\_depth': 4, 'learning\_rate': 0.0271646896766373137, 'subsample': 0.590376415058246, 'columsample\_bytree': 0.519497}

[1] 2025-05-01 02:36:55,873 Trial 6 finished with value: 0.0054860327619293 and parameters: {'n\_estimators': 121, 'max\_depth': 3, 'learning\_rate': 0.09424647241458233, 'subsample': 0.8790919198483242, 'columsample\_bytree': 0.902994}

[1] 2025-05-01 02:36:55,947 Trial 7 finished with value: 0.005053946686153232 and parameters: {'n\_estimators': 70, 'max\_depth': 4, 'learning\_rate': 0.1833911446293997, 'subsample': 0.8741621580140002, 'columsample\_bytree': 0.9185614}

[1] 2025-05-01 02:36:56,124 Trial 8 finished with value: 0.00512382162695494 and parameters: {'n\_estimators': 86, 'max\_depth': 8, 'learning\_rate': 0.108433918728034248, 'subsample': 0.5738391554971, 'columsample\_bytree': 0.92518812}

[1] 2025-05-01 02:36:56,166 Trial 9 finished with value: 0.005134919650258191 and parameters: {'n\_estimators': 293, 'max\_depth': 8, 'learning\_rate': 0.01732513937079794, 'subsample': 0.5777894359440997, 'columsample\_bytree': 0.81718}

[1] 2025-05-01 02:36:56,281 Trial 10 finished with value: 0.00548322728912995 and parameters: {'n\_estimators': 141, 'max\_depth': 6, 'learning\_rate': 0.012536527612057623, 'subsample': 0.53627228508542, 'columsample\_bytree': 0.72071}

[1] 2025-05-01 02:36:56,389 Trial 11 finished with value: 0.005267124718302945 and parameters: {'n\_estimators': 133, 'max\_depth': 6, 'learning\_rate': 0.010442953573171803, 'subsample': 0.9988395890161466, 'columsample\_bytree': 0.7252}

[1] 2025-05-01 02:36:56,472 Trial 12 finished with value: 0.0054832787749774 and parameters: {'n\_estimators': 54, 'max\_depth': 5, 'learning\_rate': 0.010421723978693209, 'subsample': 0.9266158542266727, 'columsample\_bytree': 0.7602}

[1] 2025-05-01 02:36:56,573 Trial 13 finished with value: 0.005480652077442287 and parameters: {'n\_estimators': 108, 'max\_depth': 7, 'learning\_rate': 0.02247845124965488, 'subsample': 0.8194667955347373, 'columsample\_bytree': 0.6957}

[1] 2025-05-01 02:36:56,624 Trial 14 finished with value: 0.005485001643075066 and parameters: {'n\_estimators': 204, 'max\_depth': 7, 'learning\_rate': 0.03313772424623151, 'subsample': 0.7438203817233918, 'columsample\_bytree': 0.8345}

[1] 2025-05-01 02:36:56,826 Trial 15 finished with value: 0.00548706818955481 and parameters: {'n\_estimators': 86, 'max\_depth': 8, 'learning\_rate': 0.1679982820865104, 'subsample': 0.5738391554971, 'columsample\_bytree': 0.9518812}

[1] 2025-05-01 02:36:56,954 Trial 16 finished with value: 0.005493656837674799 and parameters: {'n\_estimators': 151, 'max\_depth': 6, 'learning\_rate': 0.01732513937079794, 'subsample': 0.5777894359440997, 'columsample\_bytree': 0.81718}

[1] 2025-05-01 02:36:57,095 Trial 17 finished with value: 0.005125190491603931 and parameters: {'n\_estimators': 191, 'max\_depth': 7, 'learning\_rate': 0.0864861321812626, 'subsample': 0.98845198867278, 'columsample\_bytree': 0.858211}

[1] 2025-05-01 02:36:57,240 Trial 18 finished with value: 0.00515388515592144 and parameters: {'n\_estimators': 232, 'max\_depth': 7, 'learning\_rate': 0.0860484815592144, 'subsample': 0.688088628310666, 'columsample\_bytree': 0.84851}

[1] 2025-05-01 02:36:57,375 Trial 19 finished with value: 0.0054932307901209676 and parameters: {'n\_estimators': 188, 'max\_depth': 7, 'learning\_rate': 0.1470355250195987, 'subsample': 0.812529315368957, 'columsample\_bytree': 0.89508}

[1] 2025-05-01 02:36:57,447 A new study created in memory with name: no-name-7f5d5c3b-28ad-4c8b-83af-20fd4ca8a8062

[1] 2025-05-01 02:36:57,591 Trial 0 finished with value: 0.001764791306392107 and parameters: {'n\_estimators': 256, 'max\_depth': 3, 'learning\_rate': 0.020847047739198234, 'subsample': 0.6660105528349215, 'columsample\_bytree': 0.5224}

[1] 2025-05-01 02:36:57,734 Trial 1 finished with value: 0.00178078632397444 and parameters: {'n\_estimators': 170, 'max\_depth': 4, 'learning\_rate': 0.025275477302398834, 'subsample': 0.8049087194889226, 'columsample\_bytree': 0.766768}

[1] 2025-05-01 02:36:57,826 Trial 2 finished with value: 0.0017820201719806465 and parameters: {'n\_estimators': 138, 'max\_depth': 4, 'learning\_rate': 0.0687624556755263, 'subsample': 0.7436266650743315, 'columsample\_bytree': 0.77508}

[1] 2025-05-01 02:36:57,963 Trial 3 finished with value: 0.00192407175720706 and parameters: {'n\_estimators': 271, 'max\_depth': 6, 'learning\_rate': 0.07056554557434672, 'subsample': 0.677028915511117, 'columsample\_bytree': 0.578016}

[1] 2025-05-01 02:36:58,116 Trial 4 finished with value: 0.0017221218264456 and parameters: {'n\_estimators': 295, 'max\_depth': 5, 'learning\_rate': 0.0247291001023211, 'subsample': 0.612014415678909, 'columsample\_bytree': 0.88243}

[1] 2025-05-01 02:36:58,236 Trial 5 finished with value: 0.0017850652684079507 and parameters: {'n\_estimators': 219, 'max\_depth': 5, 'learning\_rate': 0.08916288502621772, 'subsample': 0.8785622784402006, 'columsample\_bytree': 0.840907}

[1] 2025-05-01 02:36:58,369 Trial 6 finished with value: 0.001773038447731882 and parameters: {'n\_estimators': 267, 'max\_depth': 8, 'learning\_rate': 0.01035835567788451, 'subsample': 0.5496461045097973, 'columsample\_bytree': 0.61351}

[1] 2025-05-01 02:36:58,490 Trial 7 finished with value: 0.001787901279508156 and parameters: {'n\_estimators': 227, 'max\_depth': 4, 'learning\_rate': 0.0114269570361684, 'subsample': 0.805387074540708, 'columsample\_bytree': 0.9893}

[1] 2025-05-01 02:36:58,626 Trial 8 finished with value: 0.0018266471642641264 and parameters: {'n\_estimators': 266, 'max\_depth': 7, 'learning\_rate': 0.01679982820865104, 'subsample': 0.9736409003054364, 'columsample\_bytree': 0.856529}

[1] 2025-05-01 02:36:58,797 Trial 9 finished with value: 0.0017753933920381 and parameters: {'n\_estimators': 270, 'max\_depth': 6, 'learning\_rate': 0.0126737679880477, 'subsample': 0.6569532436133337, 'columsample\_bytree': 0.969771}

[1] 2025-05-01 02:36:58,893 Trial 10 finished with value: 0.00178363987398503 and parameters: {'n\_estimators': 63, 'max\_depth': 3, 'learning\_rate': 0.028357078054071, 'subsample': 0.5075825123120764, 'columsample\_bytree': 0.51174}

[1] 2025-05-01 02:36:59,158 Trial 11 finished with value: 0.0017589487901763125 and parameters: {'n\_estimators': 208, 'max\_depth': 8, 'learning\_rate': 0.027015640925305035, 'subsample': 0.5381899022566999, 'columsample\_bytree': 0.51632}

[1] 2025-05-01 02:36:59,290 Trial 13 finished with value: 0.00175741175381027 and parameters: {'n\_estimators': 194, 'max\_depth': 8, 'learning\_rate': 0.027034714361743967, 'subsample': 0.519831451661688, 'columsample\_bytree': 0.5056}

[1] 2025-05-01 02:36:59,407 Trial 14 finished with value: 0.001758246066203706370 and parameters: {'n\_estimators': 158, 'max\_depth': 7, 'learning\_rate': 0.0362756154574472, 'subsample': 0.632756154574472, 'columsample\_bytree': 0.67070}

[1] 2025-05-01 02:36:59,515 Trial 15 finished with value: 0.0018032071744243 and parameters: {'n\_estimators': 130, 'max\_depth': 7, 'learning\_rate': 0.0459793231503547, 'subsample': 0.580297072014183, 'columsample\_bytree': 0.68625}

[1] 2025-05-01 02:36:59,639 Trial 16 finished with value: 0.00178049344651956 and parameters: {'n\_estimators': 160, 'max\_depth': 7, 'learning\_rate': 0.04695194566958359, 'subsample': 0.73885952853859, 'columsample\_bytree': 0.6918}

[1] 2025-05-01 02:36:59,757 Trial 17 finished with value: 0.00176532627176172 and parameters: {'n\_estimators': 112, 'max\_depth': 8, 'learning\_rate': 0.1318060516453937, 'subsample': 0.602104588639695563, 'columsample\_bytree': 0.63162}

[1] 2025-05-01 02:36:59,860 Trial 18 finished with value: 0.0017650215960620296 and parameters: {'n\_estimators': 112, 'max\_depth': 8, 'learning\_rate': 0.021034588639695563, 'subsample': 0.631679347196178, 'columsample\_bytree': 0.57466}

[1] 2025-05-01 02:36:59,988 Trial 19 finished with value: 0.0017528206909294 and parameters: {'n\_estimators': 186, 'max\_depth': 6, 'learning\_rate': 0.01820739401711382, 'subsample': 0.5

[1] 2025-05-31 02:37:07.355 Trial 8 finished with value: 0.0007632050159061215 and parameters: {'n\_estimators': 94, 'max\_depth': 4, 'learning\_rate': -0.0429399739950604, 'subsample': 0.6092165342053253, 'colsample\_bytree': 0.706487}, [1] 2025-05-31 02:37:07.413 Trial 9 finished with value: 0.000968313038931374 and parameters: {'n\_estimators': 54, 'max\_depth': 7, 'learning\_rate': 0.010733482461365487, 'subsample': 0.6265018000497709, 'colsample\_bytree': 0.93223}, [1] 2025-05-31 02:37:07.620 Trial 10 finished with value: 0.00976253038940501 and parameters: {'n\_estimators': 171, 'max\_depth': 3, 'learning\_rate': 0.02377508278773104, 'subsample': 0.7934312573402296, 'colsample\_bytree': 0.5052}, [1] 2025-05-31 02:37:07.958 Trial 12 finished with value: 0.0009018136887466995 and parameters: {'n\_estimators': 154, 'max\_depth': 6, 'learning\_rate': 0.12390491771912373, 'subsample': 0.503487735241092, 'colsample\_bytree': 0.6360}, [1] 2025-05-31 02:37:08.235 Trial 13 finished with value: 0.009070343193039863 and parameters: {'n\_estimators': 152, 'max\_depth': 5, 'learning\_rate': 0.15308538586303286, 'subsample': 0.5024346944753477, 'colsample\_bytree': 0.6454}, [1] 2025-05-31 02:37:08.347 Trial 14 finished with value: 0.001020303123735196 and parameters: {'n\_estimators': 135, 'max\_depth': 5, 'learning\_rate': 0.135925936126334, 'subsample': 0.5002370160366156, 'colsample\_bytree': 0.559900}, [1] 2025-05-31 02:37:08.462 Trial 15 finished with value: 0.00981532935632183 and parameters: {'n\_estimators': 137, 'max\_depth': 5, 'learning\_rate': 0.135925936126334, 'subsample': 0.5067533908004717, 'colsample\_bytree': 0.66968}, [1] 2025-05-31 02:37:08.591 Trial 16 finished with value: 0.008957514588108 and parameters: {'n\_estimators': 165, 'max\_depth': 4, 'learning\_rate': 0.17195572458744707, 'subsample': 0.5075873545074118, 'colsample\_bytree': 0.8789}, [1] 2025-05-31 02:37:08.691 Trial 17 finished with value: 0.00994794264695115 and parameters: {'n\_estimators': 100, 'max\_depth': 3, 'learning\_rate': 0.03478757817438026, 'subsample': 0.5161879182046068, 'colsample\_bytree': 0.8462}, [1] 2025-05-31 02:37:08.833 Trial 18 finished with value: 0.001023510493797238 and parameters: {'n\_estimators': 212, 'max\_depth': 4, 'learning\_rate': 0.18283185677115588, 'subsample': 0.54304826578199541, 'colsample\_bytree': 0.8851}, [1] 2025-05-31 02:37:08.956 Trial 19 finished with value: 0.00997812926359302 and parameters: {'n\_estimators': 165, 'max\_depth': 6, 'learning\_rate': 0.09621879373804, 'subsample': 0.5823008070473661, 'colsample\_bytree': 0.7022}, [1] 2025-05-31 02:37:09.077 A new study created in memory with name: no-name-bc2f0c13-afe3-40d0-85d1-8bb07ec6e5f5, [1] 2025-05-31 02:37:09.216 Trial 0 finished with value: 0.0034551244111697205 and parameters: {'n\_estimators': 180, 'max\_depth': 5, 'learning\_rate': 0.0106291295405878, 'subsample': 0.5546620424070041, 'colsample\_bytree': 0.89269}, [1] 2025-05-31 02:37:09.284 Trial 1 finished with value: 0.003460759256869733 and parameters: {'n\_estimators': 62, 'max\_depth': 7, 'learning\_rate': 0.010394110512015144, 'subsample': 0.5975682343785733, 'colsample\_bytree': 0.73167}, [1] 2025-05-31 02:37:09.379 Trial 2 finished with value: 0.0034678956152156 and parameters: {'n\_estimators': 155, 'max\_depth': 5, 'learning\_rate': 0.09896158000834589, 'subsample': 0.67255917619903, 'colsample\_bytree': 0.79796}, [1] 2025-05-31 02:37:09.472 Trial 3 finished with value: 0.00348701309781743 and parameters: {'n\_estimators': 148, 'max\_depth': 8, 'learning\_rate': 0.230015870834534, 'subsample': 0.736162423047527, 'colsample\_bytree': 0.541388}, [1] 2025-05-31 02:37:09.592 Trial 4 finished with value: 0.003460899645796356 and parameters: {'n\_estimators': 218, 'max\_depth': 5, 'learning\_rate': 0.01550982739328368, 'subsample': 0.761569027148466, 'colsample\_bytree': 0.914661}, [1] 2025-05-31 02:37:09.738 Trial 5 finished with value: 0.003460923616254905 and parameters: {'n\_estimators': 298, 'max\_depth': 3, 'learning\_rate': 0.01700955747175173, 'subsample': 0.7926134108785691, 'colsample\_bytree': 0.6109}, [1] 2025-05-31 02:37:09.848 Trial 6 finished with value: 0.0034798513581684685 and parameters: {'n\_estimators': 211, 'max\_depth': 7, 'learning\_rate': 0.135679035956045, 'subsample': 0.614055982199852, 'colsample\_bytree': 0.651628}, [1] 2025-05-31 02:37:09.923 Trial 7 finished with value: 0.0034713300834268254 and parameters: {'n\_estimators': 89, 'max\_depth': 7, 'learning\_rate': 0.1511942620562172, 'subsample': 0.7897451836316281, 'colsample\_bytree': 0.734246}, [1] 2025-05-31 02:37:10.040 Trial 8 finished with value: 0.003473830822728806 and parameters: {'n\_estimators': 206, 'max\_depth': 3, 'learning\_rate': 0.023015870834534, 'subsample': 0.74347720333679, 'colsample\_bytree': 0.8688848}, [1] 2025-05-31 02:37:10.187 Trial 9 finished with value: 0.00346807013097361857 and parameters: {'n\_estimators': 187, 'max\_depth': 6, 'learning\_rate': 0.04235127458759304, 'subsample': 0.8891556597207776, 'colsample\_bytree': 0.893315}, [1] 2025-05-31 02:37:10.344 Trial 10 finished with value: 0.0034623440837459 and parameters: {'n\_estimators': 271, 'max\_depth': 4, 'learning\_rate': 0.05898370818762456, 'subsample': 0.5029848549471768, 'colsample\_bytree': 0.99240}, [1] 2025-05-31 02:37:10.429 Trial 11 finished with value: 0.003459830352331317 and parameters: {'n\_estimators': 59, 'max\_depth': 6, 'learning\_rate': 0.03209428335901697, 'subsample': 0.5079630200047556, 'colsample\_bytree': 0.75547}, [1] 2025-05-31 02:37:10.533 Trial 12 finished with value: 0.00346495307597274 and parameters: {'n\_estimators': 111, 'max\_depth': 6, 'learning\_rate': 0.0353800812887363, 'subsample': 0.5067185539062968, 'colsample\_bytree': 0.78062}, [1] 2025-05-31 02:37:10.654 Trial 13 finished with value: 0.00346687815520644 and parameters: {'n\_estimators': 126, 'max\_depth': 4, 'learning\_rate': 0.0689606808470619, 'subsample': 0.587348310849502, 'colsample\_bytree': 0.97796}, [1] 2025-05-31 02:37:10.806 Trial 14 finished with value: 0.003463641063634056 and parameters: {'n\_estimators': 247, 'max\_depth': 5, 'learning\_rate': 0.029474503349870566, 'subsample': 0.5508180529904584, 'colsample\_bytree': 0.8314}, [1] 2025-05-31 02:37:10.894 Trial 15 finished with value: 0.003459014189106384 and parameters: {'n\_estimators': 59, 'max\_depth': 6, 'learning\_rate': 0.02506977021080163, 'subsample': 0.9815879000671981, 'colsample\_bytree': 0.66770}, [1] 2025-05-31 02:37:11.030 Trial 16 finished with value: 0.003461004253394882 and parameters: {'n\_estimators': 166, 'max\_depth': 4, 'learning\_rate': 0.04235127458759304, 'subsample': 0.898517295322974, 'colsample\_bytree': 0.668638}, [1] 2025-05-31 02:37:11.132 Trial 17 finished with value: 0.0034599328838321716 and parameters: {'n\_estimators': 109, 'max\_depth': 8, 'learning\_rate': 0.018349412341625116, 'subsample': 0.971294335118289, 'colsample\_bytree': 0.503}, [1] 2025-05-31 02:37:11.246 Trial 18 finished with value: 0.003461590868209437 and parameters: {'n\_estimators': 134, 'max\_depth': 5, 'learning\_rate': 0.02318068209437, 'subsample': 0.881724729549942, 'colsample\_bytree': 0.8896}, [1] 2025-05-31 02:37:11.398 Trial 19 finished with value: 0.003458461301709483 and parameters: {'n\_estimators': 243, 'max\_depth': 6, 'learning\_rate': 0.046901491647515134, 'subsample': 0.67453605630359777, 'colsample\_bytree': 0.621}, [1] 2025-05-31 02:37:11.511 A new study created in memory with name: no-name-a213ccfa-5156-4982-8025-102f5a149d, [1] 2025-05-31 02:37:11.621 Trial 0 finished with value: 0.002338841226920403 and parameters: {'n\_estimators': 67, 'max\_depth': 6, 'learning\_rate': 0.0254858022166842, 'subsample': 0.948800009206707, 'colsample\_bytree': 0.6960694}, [1] 2025-05-31 02:37:11.737 Trial 1 finished with value: 0.0024892832520588 and parameters: {'n\_estimators': 209, 'max\_depth': 6, 'learning\_rate': 0.092536811060995, 'subsample': 0.5784165039363863, 'colsample\_bytree': 0.7355418}, [1] 2025-05-31 02:37:11.842 Trial 2 finished with value: 0.00233249897807032 and parameters: {'n\_estimators': 182, 'max\_depth': 4, 'learning\_rate': 0.041230545820203, 'subsample': 0.627516148458359, 'colsample\_bytree': 0.708371}, [1] 2025-05-31 02:37:11.974 Trial 3 finished with value: 0.002340296351931642 and parameters: {'n\_estimators': 257, 'max\_depth': 5, 'learning\_rate': 0.020386995267090962, 'subsample': 0.919895197605492, 'colsample\_bytree': 0.7919}, [1] 2025-05-31 02:37:12.089 Trial 4 finished with value: 0.0023503595945729564 and parameters: {'n\_estimators': 216, 'max\_depth': 6, 'learning\_rate': 0.0373733297997246, 'subsample': 0.83239134513218, 'colsample\_bytree': 0.67494}, [1] 2025-05-31 02:37:12.170 Trial 5 finished with value: 0.0023502607001155795 and parameters: {'n\_estimators': 92, 'max\_depth': 5, 'learning\_rate': 0.02388160723582664, 'subsample': 0.731461219017574, 'colsample\_bytree': 0.650986}, [1] 2025-05-31 02:37:12.246 Trial 6 finished with value: 0.00235296872854982 and parameters: {'n\_estimators': 85, 'max\_depth': 8, 'learning\_rate': 0.025456748254122388, 'subsample': 0.658238028124287, 'colsample\_bytree': 0.812379}, [1] 2025-05-31 02:37:12.336 Trial 7 finished with value: 0.00234055062010066 and parameters: {'n\_estimators': 128, 'max\_depth': 7, 'learning\_rate': 0.0491425109494685, 'subsample': 0.62080301659763795, 'colsample\_bytree': 0.78918}, [1] 2025-05-31 02:37:12.463 Trial 8 finished with value: 0.002346362316251942 and parameters: {'n\_estimators': 249, 'max\_depth': 5, 'learning\_rate': 0.0370180702464777, 'subsample': 0.838254915374422, 'colsample\_bytree': 0.748331}, [1] 2025-05-31 02:37:12.539 Trial 9 finished with value: 0.00235029884489865 and parameters: {'n\_estimators': 75, 'max\_depth': 7, 'learning\_rate': 0.0111841267535786, 'subsample': 0.678626203124336, 'colsample\_bytree': 0.7483635}, [1] 2025-05-31 02:37:12.662 Trial 10 finished with value: 0.002244203836523383 and parameters: {'n\_estimators': 162, 'max\_depth': 3, 'learning\_rate': 0.02820291451884685, 'subsample': 0.508787691024641, 'colsample\_bytree': 0.5208}, [1] 2025-05-31 02:37:12.755 Trial 11 finished with value: 0.00225705468236304 and parameters: {'n\_estimators': 151, 'max\_depth': 3, 'learning\_rate': 0.02802291451884685, 'subsample': 0.508785536901395, 'colsample\_bytree': 0.5027}, [1] 2025-05-31 02:37:12.887 Trial 12 finished with value: 0.00223390735094895 and parameters: {'n\_estimators': 144, 'max\_depth': 3, 'learning\_rate': 0.291214880055334, 'subsample': 0.50069892061838, 'colsample\_bytree': 0.506824}, [1] 2025-05-31 02:37:12.994 Trial 13 finished with value: 0.00231870475793197 and parameters: {'n\_estimators': 129, 'max\_depth': 3, 'learning\_rate': 0.2998870347105137, 'subsample': 0.5088952682534562, 'colsample\_bytree': 0.52835}, [1] 2025-05-31 02:37:13.115 Trial 14 finished with value: 0.00230636399147935 and parameters: {'n\_estimators': 136, 'max\_depth': 4, 'learning\_rate': 0.1575296296997597, 'subsample': 0.5604367593742104, 'colsample\_bytree': 0.9885}, [1] 2025-05-31 02:37:13.302 Trial 15 finished with value: 0.002301764365703153 and parameters: {'n\_estimators': 290, 'max\_depth': 3, 'learning\_rate': 0.15547665600520582, 'subsample': 0.565858547412422, 'colsample\_bytree': 0.595479}, [1] 2025-05-31 02:37:13.439 Trial 16 finished with value: 0.0023406362316251942 and parameters: {'n\_estimators': 169, 'max\_depth': 4, 'learning\_rate': 0.17276478195875408, 'subsample': 0.7264843562991339, 'colsample\_bytree': 0.5984}, [1] 2025-05-31 02:37:13.532 Trial 17 finished with value: 0.0023362819736787 and parameters: {'n\_estimators': 108, 'max\_depth': 3, 'learning\_rate': 0.07648793574336964, 'subsample': 0.50208016162637208, 'colsample\_bytree': 0.50143}, [1] 2025-05-31 02:37:13.653 Trial 18 finished with value: 0.0023562819736787 and parameters: {'n\_estimators': 200, 'max\_depth': 4, 'learning\_rate': 0.215286914854344, 'subsample': 0.799638269888213, 'colsample\_bytree': 0.59229}, [1] 2025-05-31 02:37:13.767 Trial 19 finished with value: 0.002350542936306151 and parameters: {'n\_estimators': 142, 'max\_depth': 3, 'learning\_rate': 0.1049180910904777, 'subsample': 0.9929401515949039, 'colsample\_bytree': 0.92088}, [1] 2025-05-31 02:37:13.876 A new study created in memory with name: no-name-a5db8346-8271-4a5e-9932-92e74590e6bb, [1] 2025-05-31 02:37:13.999 Trial 0 finished with value: 0.00447551872488793 and parameters: {'n\_estimators': 149, 'max\_depth': 8, 'learning\_rate': 0.0103872079554613, 'subsample': 0.6533677959525069, 'colsample\_bytree': 0.688365}, [1] 2025-05-31 02:37:14.124 Trial 1 finished with value: 0.0044895053342087 and parameters: {'n\_estimators': 209, 'max\_depth': 4, 'learning\_rate': 0.1473412717090857, 'subsample': 0.8519845327921663, 'colsample\_bytree': 0.57926}, [1] 2025-05-31 02:37:14.301 Trial 2 finished with value: 0.004463053654692808 and parameters: {'n\_estimators': 299, 'max\_depth': 5, 'learning\_rate': 0.0246346054635334532, 'subsample': 0.917300896338252, 'colsample\_bytree': 0.811519}, [1] 2025-05-31 02:37:14.381 Trial 3 finished with value: 0.0044671903462052075 and parameters: {'n\_estimators': 107, 'max\_depth': 5, 'learning\_rate': 0.1751949676081072, 'subsample': 0.93331382355635638, 'colsample\_bytree': 0.800758}, [1] 2025-05-31 02:37:14.497 Trial 4 finished with value: 0.00446270423977945 and parameters: {'n\_estimators': 230, 'max\_depth': 3, 'learning\_rate': 0.1824330608119593, 'subsample': 0.819185601249613, 'colsample\_bytree': 0.6657613}, [1] 2025-05-31 02:37:14.574 Trial 5 finished with value: 0.004465641628687313 and parameters: {'n\_estimators': 102, 'max\_depth': 6, 'learning\_rate': 0.14054616218687313, 'subsample': 0.94468523513844, 'colsample\_bytree': 0.7449584}, [1] 2025-05-31 02:37:14.714 Trial 6 finished with value: 0.0044595518994650854 and parameters: {'n\_estimators': 284, 'max\_depth': 6, 'learning\_rate': 0.1731126488498084, 'subsample': 0.909375043956527, 'colsample\_bytree': 0.967633}, [1] 2025-05-31 02:37:14.839 Trial 7 finished with value: 0.004463771789052168 and parameters: {'n\_estimators': 252, 'max\_depth': 3, 'learning\_rate': 0.01050240724161605, 'subsample': 0.6514806252938472, 'colsample\_bytree': 0.60684}, [1] 2025-05-31 02:37:14.944 Trial 8 finished with value: 0.004464949212031542 and parameters: {'n\_estimators': 186, 'max\_depth': 8, 'learning\_rate': 0.23319374579775971, 'subsample': 0.8844191624329978, 'colsample\_bytree': 0.768411}, [1] 2025-05-31 02:37:15.070 Trial 9 finished with value: 0.0044605662744777 and parameters: {'n\_estimators': 244, 'max\_depth': 7, 'learning\_rate': 0.01676456952953535, 'subsample': 0.85798762431680218, 'colsample\_bytree': 0.559775}, [1] 2025-05-31 02:37:15.157 Trial 10 finished with value: 0.0045183679965360457 and parameters: {'n\_estimators': 59, 'max\_depth': 6, 'learning\_rate': 0.07000023493091904, 'subsample': 0.5067620991985505, 'colsample\_bytree': 0.9108345}, [1] 2025-05-31 02:37:15.269 Trial 11 finished with value: 0.004464606006029918 and parameters: {'n\_estimators': 116, 'max\_depth': 6, 'learning\_rate': 0.07716701742704728, 'subsample': 0.689224644952336, 'colsample\_bytree': 0.99980}, [1] 2025-05-31 02:37:15.392 Trial 12 finished with value: 0.0044612695188282 and parameters: {'n\_estimators': 176, 'max\_depth': 7, 'learning\_rate': 0.09477083106235201, 'subsample': 0.99727089655052301, 'colsample\_bytree': 0.88186}, [1] 2025-05-31 02:37:15.477 Trial 13 finished with value: 0.0045015681991228 and parameters: {'n\_estimators': 55, 'max\_depth': 6, 'learning\_rate': 0.046912519812232, 'subsample': 0.531009491973939, 'colsample\_bytree': 0.983561}, [1] 2025-05-31 02:37:15.576 Trial 14 finished with value: 0.0044339053593689775 and parameters: {'n\_estimators': 101, 'max\_depth': 7, 'learning\_rate': 0.27709839150205644, 'subsample': 0.7647768636899, 'colsample\_bytree': 0.7042871}, [1] 2025-05-31 02:37:15.653 Trial 15 finished with value: 0.00447861096495573 and parameters: {'n\_estimators': 92, 'max\_depth': 7, 'learning\_rate': 0.1103103699143583, 'subsample': 0.7551745874391558, 'colsample\_bytree': 0.697971}, [1] 2025-05-31 02:37:15.786 Trial 16 finished with value: 0.00452125088528588 and parameters: {'n\_estimators': 141, 'max\_depth': 7, 'learning\_rate': 0.2391142873213232, 'subsample': 0.60378745336595854, 'colsample\_bytree': 0.50997}, [1] 2025-05-31 02:37:15.879 Trial 17 finished with value: 0.00447741628061814 and parameters: {'n\_estimators': 79, 'max\_depth': 5, 'learning\_rate': 0.04434294628452852, 'subsample': 0.776240724744479, 'colsample\_bytree': 0.726424}, [1] 2025-05-31 02:37:15.993 Trial 18 finished with value: 0.004476660904374373 and parameters: {'n\_estimators': 133, 'max\_depth': 8, 'learning\_rate': 0.2860876754539136, 'subsample': 0.572725107464514, 'colsample\_bytree': 0.635292}, [1] 2025-05-31 02:37:16.127 Trial 19 finished with value: 0.004467835608029918 and parameters: {'n\_estimators': 192, 'max\_depth': 4, 'learning\_rate': 0.1119495583054053, 'subsample': 0.7054720345287485, 'colsample\_bytree': 0.89012}, [1] 2025-05-31 02:37:16.217 A new study created in memory with name: no-name-b6f0882d-a827-4b7c-a72d-83d8c9f9960, [1] 2025-05-31 02:37:16.375 Trial 0 finished with value: 0.001040278757805902 and parameters: {'n\_estimators': 154, 'max\_depth': 8, 'learning\_rate': 0.0675131430831557, 'subsample': 0.90685574933886, 'colsample\_bytree': 0.90369}, [1] 2025-05-31 02:37:16.498 Trial 1 finished with value: 0.00103528464498554 and parameters: {'n\_estimators': 238, 'max\_depth': 4, 'learning\_rate': 0.142367679911519, 'subsample': 0.9217801597515716, 'colsample\_bytree': 0.606434}, [1] 2025-05-31 02:37:16.609 Trial 2 finished with value: 0.00104230514439034 and parameters: {'n\_estimators': 183, 'max\_depth': 3, 'learning\_rate': 0.021113848836061, 'subsample': 0.8714685215201597, 'colsample\_bytree': 0.9015298}, [1] 2025-05-31 02:37:16.724 Trial 3 finished with value: 0.001040257357101387 and parameters: {'n\_estimators': 200, 'max\_depth': 3, 'learning\_rate': 0.041562494717468465, 'subsample': 0.53642462078969714, 'colsample\_bytree': 0.97660}, [1] 2025-05-31 02:37:16.802 Trial 4 finished with value: 0.0010332440264624553 and parameters: {'n\_estimators': 99, 'max\_depth': 5, 'learning\_rate': 0.016250565562339724, 'subsample': 0.643512899143613, 'colsample\_bytree': 0.88833}, [1] 2025-05-31 02:37:16.921 Trial 5 finished with value: 0.001039478638467902 and parameters: {'n\_estimators': 232, 'max\_depth': 8, 'learning\_rate': 0.01540225432388781, 'subsample': 0.536877576131955, 'colsample\_bytree': 0.6952}, [1] 2025-05-31 02:37:16.990 Trial 6 finished with value: 0.0010407340646701375 and parameters: {'n\_estimators': 68, 'max\_depth': 5, 'learning\_rate': 0.1341553892810375, 'subsample': 0.94293839543386, 'colsample\_bytree': 0.924225}, [1] 2025-05-31 02:37:17.167 Trial 7 finished with value: 0.0010345670421691069 and parameters: {'n

[1] 2025-05-31 02:37:25.373 Trial 16 finished with value: 0.0027139503781399704 and parameters: {'n\_estimators': 190, 'max\_depth': 6, 'learning\_rate': 0.08646632886715022, 'subsample': 0.0359471719974919, 'colsample\_bytree': 0.0776}

[1] 2025-05-31 02:37:25.483 Trial 17 finished with value: 0.0027150515579762 and parameters: {'n\_estimators': 131, 'max\_depth': 7, 'learning\_rate': 0.05341532370326319, 'subsample': 0.7343852300814286, 'colsample\_bytree': 0.80252}

[1] 2025-05-31 02:37:25.577 Trial 18 finished with value: 0.00271440471874561463 and parameters: {'n\_estimators': 90, 'max\_depth': 5, 'learning\_rate': 0.05341532370326319, 'subsample': 0.2673759932318693, 'colsample\_bytree': 0.865094620565201, 'subsample': 0.5070791}

[1] 2025-05-31 02:37:25.733 Trial 19 finished with value: 0.00271465353856843 and parameters: {'n\_estimators': 167, 'max\_depth': 8, 'learning\_rate': 0.023270997684308703, 'subsample': 0.9415892612675638, 'colsample\_bytree': 0.6476}

[1] 2025-05-31 02:37:25.821 A new study created in memory with name: no-name-1636ab9-4ad-4ee1-5050-97a277343c0

[1] 2025-05-31 02:37:25.919 Trial 0 finished with value: 0.001986291466343883 and parameters: {'n\_estimators': 90, 'max\_depth': 5, 'learning\_rate': 0.017107687307886685, 'subsample': 0.6167123772730343, 'colsample\_bytree': 0.514726}

[1] 2025-05-31 02:37:26.051 Trial 1 finished with value: 0.001986860041658418 and parameters: {'n\_estimators': 253, 'max\_depth': 3, 'learning\_rate': 0.01032957101024973, 'subsample': 0.98098073078941709, 'colsample\_bytree': 0.81923}

[1] 2025-05-31 02:37:26.153 Trial 2 finished with value: 0.00203787697757515 and parameters: {'n\_estimators': 161, 'max\_depth': 8, 'learning\_rate': 0.147520437114151, 'subsample': 0.6364070676854604, 'colsample\_bytree': 0.86268363}

[1] 2025-05-31 02:37:26.284 Trial 3 finished with value: 0.00199233098507045 and parameters: {'n\_estimators': 257, 'max\_depth': 5, 'learning\_rate': 0.05823012275148293, 'subsample': 0.5878788386919537, 'colsample\_bytree': 0.798742}

[1] 2025-05-31 02:37:26.363 Trial 4 finished with value: 0.0019537978955680727 and parameters: {'n\_estimators': 105, 'max\_depth': 9, 'learning\_rate': 0.1708643595584064, 'subsample': 0.6935634074886866, 'colsample\_bytree': 0.8651903}

[1] 2025-05-31 02:37:26.450 Trial 5 finished with value: 0.0020008304836568024 and parameters: {'n\_estimators': 135, 'max\_depth': 7, 'learning\_rate': 0.043542040989336436, 'subsample': 0.68760309176588, 'colsample\_bytree': 0.5797131}

[1] 2025-05-31 02:37:26.530 Trial 6 finished with value: 0.0019886743038620382 and parameters: {'n\_estimators': 90, 'max\_depth': 3, 'learning\_rate': 0.009491074720764, 'subsample': 0.99491074720764, 'colsample\_bytree': 0.4704394}

[1] 2025-05-31 02:37:26.579 Trial 7 finished with value: 0.001999807805546076 and parameters: {'n\_estimators': 71, 'max\_depth': 8, 'learning\_rate': 0.2389149119216007, 'subsample': 0.74549119216007, 'colsample\_bytree': 0.5093648072}

[1] 2025-05-31 02:37:26.694 Trial 8 finished with value: 0.00199561026357985 and parameters: {'n\_estimators': 141, 'max\_depth': 7, 'learning\_rate': 0.0659509525678001, 'subsample': 0.659509525678001, 'colsample\_bytree': 0.695079}

[1] 2025-05-31 02:37:26.796 Trial 9 finished with value: 0.00199316549603312 and parameters: {'n\_estimators': 110, 'max\_depth': 8, 'learning\_rate': 0.038591842145593, 'subsample': 0.884131761249281, 'colsample\_bytree': 0.75167413}

[1] 2025-05-31 02:37:26.928 Trial 10 finished with value: 0.0019214374783235 and parameters: {'n\_estimators': 207, 'max\_depth': 6, 'learning\_rate': 0.2727389674645925, 'subsample': 0.5105940231357053, 'colsample\_bytree': 0.9754927}

[1] 2025-05-31 02:37:27.062 Trial 11 finished with value: 0.001916978423011213 and parameters: {'n\_estimators': 202, 'max\_depth': 6, 'learning\_rate': 0.2914667480266667, 'subsample': 0.5016934288927685, 'colsample\_bytree': 0.99999}

[1] 2025-05-31 02:37:27.197 Trial 12 finished with value: 0.001923066873214034 and parameters: {'n\_estimators': 209, 'max\_depth': 6, 'learning\_rate': 0.2886168690242004, 'subsample': 0.5076487176965525, 'colsample\_bytree': 0.9884}

[1] 2025-05-31 02:37:30.0 Trial 13 finished with value: 0.00193637465019364 and parameters: {'n\_estimators': 205, 'max\_depth': 6, 'learning\_rate': 0.0857875771687835, 'subsample': 0.5105642004806768, 'colsample\_bytree': 0.9958}

[1] 2025-05-31 02:37:27.460 Trial 14 finished with value: 0.00190610749760897 and parameters: {'n\_estimators': 200, 'max\_depth': 4, 'learning\_rate': 0.294651501046616, 'subsample': 0.56303301455658793, 'colsample\_bytree': 0.926001}

[1] 2025-05-31 02:37:27.620 Trial 15 finished with value: 0.001993174382606967 and parameters: {'n\_estimators': 286, 'max\_depth': 4, 'learning\_rate': 0.0870053495742174, 'subsample': 0.5736547837872652, 'colsample\_bytree': 0.90929}

[1] 2025-05-31 02:37:27.751 Trial 16 finished with value: 0.001960449669480908 and parameters: {'n\_estimators': 181, 'max\_depth': 4, 'learning\_rate': 0.20137236580340873, 'subsample': 0.801727380143431, 'colsample\_bytree': 0.9293}

[1] 2025-05-31 02:37:27.893 Trial 17 finished with value: 0.001972881645724757 and parameters: {'n\_estimators': 240, 'max\_depth': 4, 'learning\_rate': 0.0283799548321174, 'subsample': 0.5708090927051299, 'colsample\_bytree': 0.663}

[1] 2025-05-31 02:37:28.026 Trial 18 finished with value: 0.001946507565624565 and parameters: {'n\_estimators': 177, 'max\_depth': 5, 'learning\_rate': 0.105670103704120465, 'subsample': 0.7655351986471489, 'colsample\_bytree': 0.9406}

[1] 2025-05-31 02:37:28.170 Trial 19 finished with value: 0.00197884395701942 and parameters: {'n\_estimators': 231, 'max\_depth': 7, 'learning\_rate': 0.19773620116785445, 'subsample': 0.6656335265887602, 'colsample\_bytree': 0.85697}

[1] 2025-05-31 02:37:28.301 A new study created in memory with name: no-name-877fc0bo-10bf-4019-ab8c-c41d0a59c

[1] 2025-05-31 02:37:28.420 Trial 0 finished with value: 0.0010681298620657104 and parameters: {'n\_estimators': 145, 'max\_depth': 7, 'learning\_rate': 0.11508365788018439, 'subsample': 0.900933696843541, 'colsample\_bytree': 0.74735}

[1] 2025-05-31 02:37:28.496 Trial 1 finished with value: 0.001060192157364184 and parameters: {'n\_estimators': 95, 'max\_depth': 8, 'learning\_rate': 0.01667765073887875, 'subsample': 0.494576287223432, 'colsample\_bytree': 0.67238}

[1] 2025-05-31 02:37:28.593 Trial 2 finished with value: 0.0010553645825954947 and parameters: {'n\_estimators': 155, 'max\_depth': 4, 'learning\_rate': 0.0797391911066063, 'subsample': 0.5452579912653006, 'colsample\_bytree': 0.81224}

[1] 2025-05-31 02:37:28.725 Trial 3 finished with value: 0.00106650504896702 and parameters: {'n\_estimators': 237, 'max\_depth': 8, 'learning\_rate': 0.01853102384282689, 'subsample': 0.819159364645765, 'colsample\_bytree': 0.8282991}

[1] 2025-05-31 02:37:28.852 Trial 4 finished with value: 0.00107369457956772 and parameters: {'n\_estimators': 152, 'max\_depth': 6, 'learning\_rate': 0.14113066093380398, 'subsample': 0.858542265540627, 'colsample\_bytree': 0.63506}

[1] 2025-05-31 02:37:28.996 Trial 5 finished with value: 0.00120125798538744 and parameters: {'n\_estimators': 294, 'max\_depth': 7, 'learning\_rate': 0.0251005822558901402, 'subsample': 0.577300685593831, 'colsample\_bytree': 0.7869}

[1] 2025-05-31 02:37:29.110 Trial 6 finished with value: 0.0016849885255463933 and parameters: {'n\_estimators': 196, 'max\_depth': 6, 'learning\_rate': 0.256236289261606, 'subsample': 0.526236289261606, 'colsample\_bytree': 0.9374}

[1] 2025-05-31 02:37:29.179 Trial 7 finished with value: 0.001057506191334348 and parameters: {'n\_estimators': 69, 'max\_depth': 7, 'learning\_rate': 0.0191275541481452, 'subsample': 0.931991756646226, 'colsample\_bytree': 0.74262}

[1] 2025-05-31 02:37:29.245 Trial 8 finished with value: 0.00104506633284687 and parameters: {'n\_estimators': 176, 'max\_depth': 3, 'learning\_rate': 0.2077450480837528, 'subsample': 0.499738104180414, 'colsample\_bytree': 0.546032}

[1] 2025-05-31 02:37:29.374 Trial 9 finished with value: 0.0010914259503950045 and parameters: {'n\_estimators': 112, 'max\_depth': 3, 'learning\_rate': 0.0660197531155391, 'subsample': 0.87215371790332, 'colsample\_bytree': 0.974332}

[1] 2025-05-31 02:37:29.511 Trial 10 finished with value: 0.001089741569242143 and parameters: {'n\_estimators': 217, 'max\_depth': 4, 'learning\_rate': 0.2638469501257899, 'subsample': 0.728934787638991, 'colsample\_bytree': 0.50027}

[1] 2025-05-31 02:37:29.596 Trial 11 finished with value: 0.00105589907002303 and parameters: {'n\_estimators': 55, 'max\_depth': 5, 'learning\_rate': 0.0417280438471451, 'subsample': 0.780462391201587, 'colsample\_bytree': 0.513255}

[1] 2025-05-31 02:37:29.681 Trial 12 finished with value: 0.001068997002819664 and parameters: {'n\_estimators': 53, 'max\_depth': 3, 'learning\_rate': 0.04220516332312299, 'subsample': 0.75667711907151, 'colsample\_bytree': 0.5009}

[1] 2025-05-31 02:37:29.834 Trial 13 finished with value: 0.001025419366354025 and parameters: {'n\_estimators': 254, 'max\_depth': 5, 'learning\_rate': 0.2592566679707844, 'subsample': 0.7654778270746106, 'colsample\_bytree': 0.59431}

[1] 2025-05-31 02:37:29.991 Trial 14 finished with value: 0.00105678393819247183 and parameters: {'n\_estimators': 266, 'max\_depth': 4, 'learning\_rate': 0.2891743369732626, 'subsample': 0.65501388164684763, 'colsample\_bytree': 0.5960}

[1] 2025-05-31 02:37:30.138 Trial 15 finished with value: 0.00105220230694655 and parameters: {'n\_estimators': 246, 'max\_depth': 5, 'learning\_rate': 0.1645404240930398, 'subsample': 0.995257068395007, 'colsample\_bytree': 0.5845}

[1] 2025-05-31 02:37:30.274 Trial 16 finished with value: 0.00107684089485561 and parameters: {'n\_estimators': 195, 'max\_depth': 3, 'learning\_rate': 0.01055929824045032, 'subsample': 0.8001587738494899, 'colsample\_bytree': 0.68147}

[1] 2025-05-31 02:37:30.434 Trial 17 finished with value: 0.001036142019387608 and parameters: {'n\_estimators': 287, 'max\_depth': 5, 'learning\_rate': 0.1937803823840878, 'subsample': 0.6667528622117706, 'colsample\_bytree': 0.56613}

[1] 2025-05-31 02:37:30.599 Trial 18 finished with value: 0.0010817784240252606 and parameters: {'n\_estimators': 300, 'max\_depth': 5, 'learning\_rate': 0.0952048082094926, 'subsample': 0.66930678472258, 'colsample\_bytree': 0.600083}

[1] 2025-05-31 02:37:30.751 Trial 19 finished with value: 0.0010670997822019664 and parameters: {'n\_estimators': 266, 'max\_depth': 6, 'learning\_rate': 0.1953835628108353, 'subsample': 0.6421074210529826, 'colsample\_bytree': 0.8616}

[1] 2025-05-31 02:37:30.925 A new study created in memory with name: no-name-cfcba4-b4ec-40be-b273-edda40dc332a

[1] 2025-05-31 02:37:31.093 Trial 0 finished with value: 0.000622134250389926 and parameters: {'n\_estimators': 285, 'max\_depth': 5, 'learning\_rate': 0.037356209017992345, 'subsample': 0.5662387375445055, 'colsample\_bytree': 0.7070}

[1] 2025-05-31 02:37:31.220 Trial 1 finished with value: 0.000661052040955055 and parameters: {'n\_estimators': 249, 'max\_depth': 5, 'learning\_rate': 0.038669208234849904, 'subsample': 0.5980411520855444, 'colsample\_bytree': 0.55046}

[1] 2025-05-31 02:37:31.376 Trial 2 finished with value: 0.000660395019303433 and parameters: {'n\_estimators': 283, 'max\_depth': 4, 'learning\_rate': 0.033554976551393, 'subsample': 0.8157951684492584, 'colsample\_bytree': 0.509102}

[1] 2025-05-31 02:37:31.538 Trial 3 finished with value: 0.000659365161039022 and parameters: {'n\_estimators': 156, 'max\_depth': 4, 'learning\_rate': 0.079739366271425077, 'subsample': 0.545859656709186, 'colsample\_bytree': 0.58895}

[1] 2025-05-31 02:37:32.331 Trial 4 finished with value: 0.000658649439568956026 and parameters: {'n\_estimators': 214, 'max\_depth': 5, 'learning\_rate': 0.035822053018516466, 'subsample': 0.523348733299278, 'colsample\_bytree': 0.8408}

[1] 2025-05-31 02:37:32.495 Trial 5 finished with value: 0.0006667308162432328 and parameters: {'n\_estimators': 129, 'max\_depth': 4, 'learning\_rate': 0.2775288747362602, 'subsample': 0.58752987575691925, 'colsample\_bytree': 0.560588}

[1] 2025-05-31 02:37:32.721 Trial 6 finished with value: 0.0006592620859817436 and parameters: {'n\_estimators': 187, 'max\_depth': 3, 'learning\_rate': 0.01372893921359842, 'subsample': 0.6191051722594937, 'colsample\_bytree': 0.8978}

[1] 2025-05-31 02:37:33.547 Trial 7 finished with value: 0.000658585122567483 and parameters: {'n\_estimators': 222, 'max\_depth': 6, 'learning\_rate': 0.03957095205241432, 'subsample': 0.8688616321861748, 'colsample\_bytree': 0.73401}

[1] 2025-05-31 02:37:33.804 Trial 8 finished with value: 0.0006590922321431 and parameters: {'n\_estimators': 192, 'max\_depth': 5, 'learning\_rate': 0.070686311495622353, 'subsample': 0.570031259545106, 'colsample\_bytree': 0.90817}

[1] 2025-05-31 02:37:33.968 Trial 9 finished with value: 0.000659103254570374 and parameters: {'n\_estimators': 174, 'max\_depth': 3, 'learning\_rate': 0.014617104922677112, 'subsample': 0.9528125526493398, 'colsample\_bytree': 0.71253}

[1] 2025-05-31 02:37:34.078 Trial 10 finished with value: 0.00659872047303682 and parameters: {'n\_estimators': 58, 'max\_depth': 8, 'learning\_rate': 0.01303977091194681, 'subsample': 0.703485103952264, 'colsample\_bytree': 0.62766}

[1] 2025-05-31 02:37:34.353 Trial 11 finished with value: 0.0065868462258225838 and parameters: {'n\_estimators': 131, 'max\_depth': 1, 'learning\_rate': 0.023211343244974493, 'subsample': 0.5102422103514943, 'colsample\_bytree': 0.883}

[1] 2025-05-31 02:37:34.475 Trial 12 finished with value: 0.006624258332123166 and parameters: {'n\_estimators': 134, 'max\_depth': 6, 'learning\_rate': 0.10691653595416651, 'subsample': 0.656709340576711, 'colsample\_bytree': 0.99938}

[1] 2025-05-31 02:37:34.579 Trial 13 finished with value: 0.0066252933308354 and parameters: {'n\_estimators': 82, 'max\_depth': 4, 'learning\_rate': 0.0537132995180217, 'subsample': 0.5037871131908217, 'colsample\_bytree': 0.8151640}

[1] 2025-05-31 02:37:34.727 Trial 14 finished with value: 0.0066030720445744576 and parameters: {'n\_estimators': 232, 'max\_depth': 6, 'learning\_rate': 0.16691217594407201, 'subsample': 0.6558544887867817, 'colsample\_bytree': 0.6380}

[1] 2025-05-31 02:37:34.867 Trial 15 finished with value: 0.00659890362313403684 and parameters: {'n\_estimators': 171, 'max\_depth': 6, 'learning\_rate': 0.021606247849234862, 'subsample': 0.60170573115945106, 'colsample\_bytree': 0.8101401}

[1] 2025-05-31 02:37:34.969 Trial 16 finished with value: 0.00659892704308362 and parameters: {'n\_estimators': 174, 'max\_depth': 3, 'learning\_rate': 0.014617104922677112, 'subsample': 0.9528125526493398, 'colsample\_bytree': 0.71253}

[1] 2025-05-31 02:37:34.974 Trial 17 finished with value: 0.00659892704308362 and parameters: {'n\_estimators': 58, 'max\_depth': 8, 'learning\_rate': 0.01305917901194681, 'subsample': 0.702485103952264, 'colsample\_bytree': 0.62766}

[1] 2025-05-31 02:37:34.975 Trial 18 finished with value: 0.00659892704308362 and parameters: {'n\_estimators': 134, 'max\_depth': 5, 'learning\_rate': 0.023211343244974493, 'subsample': 0.5102422103514943, 'colsample\_bytree': 0.99938}

[1] 2025-05-31 02:37:34.995 Trial 19 finished with value: 0.00659892704308362 and parameters: {'n\_estimators': 221, 'max\_depth': 4, 'learning\_rate': 0.10691653595416651, 'subsample': 0.656709340576711, 'colsample\_bytree': 0.8151640}

[1] 2025-05-31 02:37:35.052 Trial 20 finished with value: 0.00659892704308362 and parameters: {'n\_estimators': 279, 'max\_depth': 3, 'learning\_rate': 0.23060868747234982, 'subsample': 0.576914897590295, 'colsample\_bytree': 0.767709705}

[1] 2025-05-31 02:37:36.421 Trial 21 finished with value: 0.00659892704308362 and parameters: {'n\_estimators': 180, 'max\_depth': 7, 'learning\_rate': 0.0404494626705474494, 'subsample': 0.5513292943196899, 'colsample\_bytree': 0.92454}

[1] 2025-05-31 02:37:36.549 Trial 22 finished with value: 0.0065994395653084 and parameters: {'n\_estimators': 251, 'max\_depth': 3, 'learning\_rate': 0.021741749873475, 'subsample': 0.8125342263172917, 'colsample\_bytree': 0.78763}

[1] 2025-05-31 02:37:36.693 Trial 23 finished with value: 0.00646466505938762 and parameters: {'n\_estimators': 300, 'max\_depth': 6, 'learning\_rate': 0.016206247849234862, 'subsample': 0.801075943027551, 'colsample\_bytree': 0.6675}

[1] 2025-05-31 02:37:36.801 Trial 24 finished with value: 0.0016266009535723853 and parameters: {'n\_estimators': 135, 'max\_depth': 3, 'learning\_rate': 0.229032767097287, 'subsample': 0.959886023134763, 'colsample\_bytree': 0.609827}

[1] 2025-05-31 02:37:36.907 Trial 25 finished with value: 0.001646717090359334 and parameters: {'n\_estimators': 126, 'max\_depth': 3, 'learning\_rate': 0.2982808141057876, 'subsample': 0.980339169526339, 'colsample\_bytree': 0.60838}

[1] 2025-05-31 02:37:37.012 Trial 26 finished with value: 0.0016673021934581561 and parameters: {'n\_estimators': 118, 'max\_depth': 4, 'learning\_rate': 0.299417415206341, 'subsample': 0.969656131529524, 'colsample\_bytree': 0.62266}

[1] 2025-05-31 02:37:37.167 Trial 27 finished with value: 0.001615392204937914061 and parameters: {'n\_estimators': 131, 'max\_depth': 4, 'learning\_rate': 0.158395616990176, 'subsample': 0.915941895590295, 'colsample\_bytree': 0.5056}

[1] 2025-05-31 02:37:37.336 Trial 28 finished with value: 0.0016061649739170416 and parameters: {'n\_estimators': 298, 'max\_depth': 4, 'learning\_rate': 0.145355646208685, 'subsample': 0.884873674259320, 'colsample\_bytree': 0.50099}

[1] 2025-05-31 02:37:37.499 Trial 29 finished with value: 0.001595945269223153 and parameters: {'n\_estimators': 296, 'max\_depth': 4, 'learning\_rate': 0.072071925923173, 'subsample': 0.674826465216456, 'colsample\_bytree': 0.79033}

[1] 2025-05-31 02:37:37.656 Trial 30 finished with value: 0.0016262094234218208 and parameters: {'n\_estimators': 264, 'max\_depth': 8, 'learning\_rate': 0.0702171119221476, 'subsample': 0.6364060184417543, 'colsample\_bytree': 0.8423}

[1] 2025-05-31 02:37:37.814 Trial 31 finished with value: 0.001629742816199151 and parameters: {'n\_estimators': 264, 'max\_depth': 4, 'learning\_rate': 0.049681709271818, 'subsample': 0.7333487354201078, 'colsample\_bytree': 0.79962}

[1] 2025-05-31 02:37:37.956 Trial 32 finished with value: 0.0016679412566813743 and parameters: {'n\_estimators': 232, 'max\_depth': 3, 'learning\_rate': 0.07710889279477118, 'subsample': 0.877320320366022

```
[1] 2025-05-31 02:37:47,218] Trial 3 finished with value: 0.0002458388903658126 and parameters: {'n_estimators': 171, 'max_depth': 6, 'learning_rate': 0.0689753888420223, 'subsample': 0.826063523641847, 'colsample_bytree': 0.8943
[1] 2025-05-31 02:37:47,406] Trial 4 finished with value: 0.000247761957135943 and parameters: {'n_estimators': 137, 'max_depth': 8, 'learning_rate': 0.10928752656342647, 'subsample': 0.943688788952529, 'colsample_bytree': 0.789322
[1] 2025-05-31 02:37:47,503] Trial 5 finished with value: 0.0002643000674838594 and parameters: {'n_estimators': 74, 'max_depth': 7, 'learning_rate': 0.0308915926302437, 'subsample': 0.5191500752480394, 'colsample_bytree': 0.68724
[1] 2025-05-31 02:37:47,675] Trial 6 finished with value: 0.0002594405156979329 and parameters: {'n_estimators': 107, 'max_depth': 7, 'learning_rate': 0.026438008470532068, 'subsample': 0.839920801238769, 'colsample_bytree': 0.7226
[1] 2025-05-31 02:37:47,753] Trial 7 finished with value: 0.0002330310098070838 and parameters: {'n_estimators': 71, 'max_depth': 3, 'learning_rate': 0.0308915926302437, 'subsample': 0.5191500752480394, 'colsample_bytree': 0.68724
[1] 2025-05-31 02:37:47,835] Trial 8 finished with value: 0.00023579199205275416 and parameters: {'n_estimators': 115, 'max_depth': 3, 'learning_rate': 0.19700462771363948, 'subsample': 0.6097324462393678, 'colsample_bytree': 0.9475
[1] 2025-05-31 02:37:47,919] Trial 9 finished with value: 0.00025467501991266 and parameters: {'n_estimators': 114, 'max_depth': 3, 'learning_rate': 0.01743739547515304, 'subsample': 0.8440748703087673, 'colsample_bytree': 0.772626
[1] 2025-05-31 02:37:48,059] Trial 10 finished with value: 0.0002350558346216288 and parameters: {'n_estimators': 232, 'max_depth': 4, 'learning_rate': 0.01743739547515304, 'subsample': 0.2815092400792622, 'colsample_bytree': 0.5623
[1] 2025-05-31 02:37:48,203] Trial 11 finished with value: 0.0002425445485871995 and parameters: {'n_estimators': 244, 'max_depth': 4, 'learning_rate': 0.267070743995188, 'subsample': 0.686534096504044, 'colsample_bytree': 0.59998
[1] 2025-05-31 02:37:48,376] Trial 12 finished with value: 0.0002407024759265646 and parameters: {'n_estimators': 228, 'max_depth': 4, 'learning_rate': 0.14706389906987600, 'subsample': 0.734186461369262, 'colsample_bytree': 0.53256
[1] 2025-05-31 02:37:48,546] Trial 13 finished with value: 0.000239786386872468 and parameters: {'n_estimators': 289, 'max_depth': 4, 'learning_rate': 0.29012085141982163, 'subsample': 0.5691505793277094, 'colsample_bytree': 0.518
[1] 2025-05-31 02:37:48,718] Trial 14 finished with value: 0.00024008872989717897 and parameters: {'n_estimators': 294, 'max_depth': 3, 'learning_rate': 0.0999807152071246, 'subsample': 0.550791542544907, 'colsample_bytree': 0.626
[1] 2025-05-31 02:37:48,846] Trial 15 finished with value: 0.0002406761897021044 and parameters: {'n_estimators': 195, 'max_depth': 5, 'learning_rate': 0.1744303588708546, 'subsample': 0.6068845990820664, 'colsample_bytree': 0.823
[1] 2025-05-31 02:37:49,003] Trial 16 finished with value: 0.0002505687162621687 and parameters: {'n_estimators': 271, 'max_depth': 4, 'learning_rate': 0.07005436728578368, 'subsample': 0.5791758198250872, 'colsample_bytree': 0.998
[1] 2025-05-31 02:37:49,130] Trial 17 finished with value: 0.000299072800206986 and parameters: {'n_estimators': 190, 'max_depth': 5, 'learning_rate': 0.14886523875178162, 'subsample': 0.6488526225849586, 'colsample_bytree': 0.84117
[1] 2025-05-31 02:37:49,235] Trial 18 finished with value: 0.000234492734198557 and parameters: {'n_estimators': 83, 'max_depth': 3, 'learning_rate': 0.1166719574744216, 'subsample': 0.5015040466256565, 'colsample_bytree': 0.51266
[1] 2025-05-31 02:37:49,394] Trial 19 finished with value: 0.0002394714245617227 and parameters: {'n_estimators': 290, 'max_depth': 4, 'learning_rate': 0.20077526595142853, 'subsample': 0.735510313281487, 'colsample_bytree': 0.638
[1] 2025-05-31 02:37:49,536] A new study created in memory with name: no-name_4696b17d-a69-445f-b6ch-56180045f25e
[1] 2025-05-31 02:37:49,681] Trial 0 finished with value: 0.000943625952859396 and parameters: {'n_estimators': 206, 'max_depth': 5, 'learning_rate': 0.1731594121938876, 'subsample': 0.865171930134527, 'colsample_bytree': 0.52976
[1] 2025-05-31 02:37:49,749] Trial 1 finished with value: 0.000970729577921639 and parameters: {'n_estimators': 63, 'max_depth': 3, 'learning_rate': 0.20474107146878904, 'subsample': 0.848778524293817, 'colsample_bytree': 0.505223
[1] 2025-05-31 02:37:49,876] Trial 2 finished with value: 0.00094379945622259 and parameters: {'n_estimators': 249, 'max_depth': 7, 'learning_rate': 0.2386835371171893, 'subsample': 0.9307968824873316, 'colsample_bytree': 0.56142
[1] 2025-05-31 02:37:50,000] Trial 3 finished with value: 0.000933441107937504 and parameters: {'n_estimators': 239, 'max_depth': 8, 'learning_rate': 0.24138995010743402, 'subsample': 0.5251958637736397, 'colsample_bytree': 0.89629
[1] 2025-05-31 02:37:50,136] Trial 4 finished with value: 0.000937375197924613 and parameters: {'n_estimators': 281, 'max_depth': 5, 'learning_rate': 0.13193364530627754, 'subsample': 0.9184924415595177, 'colsample_bytree': 0.91553
[1] 2025-05-31 02:37:50,260] Trial 5 finished with value: 0.0009484238433317068 and parameters: {'n_estimators': 245, 'max_depth': 8, 'learning_rate': 0.0324283785762277, 'subsample': 0.771150193885033, 'colsample_bytree': 0.51804
[1] 2025-05-31 02:37:50,359] Trial 6 finished with value: 0.00096220408891492 and parameters: {'n_estimators': 144, 'max_depth': 8, 'learning_rate': 0.0614110101971675, 'subsample': 0.52606165856639, 'colsample_bytree': 0.961429
[1] 2025-05-31 02:37:50,496] Trial 7 finished with value: 0.000946798054664273 and parameters: {'n_estimators': 286, 'max_depth': 4, 'learning_rate': 0.02645726737514404, 'subsample': 0.564538490861323, 'colsample_bytree': 0.860924
[1] 2025-05-31 02:37:50,620] Trial 8 finished with value: 0.000984703267543586 and parameters: {'n_estimators': 198, 'max_depth': 8, 'learning_rate': 0.2719963718668754, 'subsample': 0.781524273960498, 'colsample_bytree': 0.56612
[1] 2025-05-31 02:37:50,745] Trial 9 finished with value: 0.0009627138171664747 and parameters: {'n_estimators': 233, 'max_depth': 8, 'learning_rate': 0.1297151262628029, 'subsample': 0.927310125684252, 'colsample_bytree': 0.62501
[1] 2025-05-31 02:37:50,853] Trial 10 finished with value: 0.0095392184353684 and parameters: {'n_estimators': 132, 'max_depth': 6, 'learning_rate': 0.011526647798401864, 'subsample': 0.638957091353344, 'colsample_bytree': 0.7854
[1] 2025-05-31 02:37:50,985] Trial 11 finished with value: 0.009758365347570148 and parameters: {'n_estimators': 194, 'max_depth': 6, 'learning_rate': 0.1015130817765458, 'subsample': 0.687939082317171, 'colsample_bytree': 0.71305
[1] 2025-05-31 02:37:51,112] Trial 12 finished with value: 0.0009368942762162232 and parameters: {'n_estimators': 166, 'max_depth': 5, 'learning_rate': 0.083045828268953, 'subsample': 0.95764942001175, 'colsample_bytree': 0.7157
[1] 2025-05-31 02:37:51,249] Trial 13 finished with value: 0.000924161759988675 and parameters: {'n_estimators': 211, 'max_depth': 4, 'learning_rate': 0.1659135795528037, 'subsample': 0.6766381353243847, 'colsample_bytree': 0.8223
[1] 2025-05-31 02:37:51,344] Trial 14 finished with value: 0.00092008871865441 and parameters: {'n_estimators': 76, 'max_depth': 3, 'learning_rate': 0.293040954820175, 'subsample': 0.619041011169066, 'colsample_bytree': 0.84230
[1] 2025-05-31 02:37:51,429] Trial 15 finished with value: 0.0009471107669893022 and parameters: {'n_estimators': 59, 'max_depth': 3, 'learning_rate': 0.2943306580149458, 'subsample': 0.636260929005118, 'colsample_bytree': 0.843946
[1] 2025-05-31 02:37:51,527] Trial 16 finished with value: 0.00093973354390164 and parameters: {'n_estimators': 98, 'max_depth': 4, 'learning_rate': 0.401896987567094, 'subsample': 0.698948510939424, 'colsample_bytree': 0.78759
[1] 2025-05-31 02:37:51,641] Trial 17 finished with value: 0.000909008010745929 and parameters: {'n_estimators': 100, 'max_depth': 4, 'learning_rate': 0.16175448080753016, 'subsample': 0.559024594331933, 'colsample_bytree': 0.65708
[1] 2025-05-31 02:37:51,750] Trial 18 finished with value: 0.00020285038576246 and parameters: {'n_estimators': 98, 'max_depth': 3, 'learning_rate': 0.0650235241627041, 'subsample': 0.5918249230217282, 'colsample_bytree': 0.680436
[1] 2025-05-31 02:37:51,859] Trial 19 finished with value: 0.000265381267071346 and parameters: {'n_estimators': 105, 'max_depth': 4, 'learning_rate': 0.06565360062630599, 'subsample': 0.5751147942025412, 'colsample_bytree': 0.6535
[1] 2025-05-31 02:37:52,074] A new study created in memory with name: no-name_e6d73092-0064-4db7-8858-24f544fa3aea
[1] 2025-05-31 02:37:52,157] Trial 0 finished with value: 0.0012635161663028624 and parameters: {'n_estimators': 97, 'max_depth': 6, 'learning_rate': 0.21133761158946313, 'subsample': 0.894072621715708, 'colsample_bytree': 0.890083
[1] 2025-05-31 02:37:52,260] Trial 1 finished with value: 0.001253692761440632 and parameters: {'n_estimators': 166, 'max_depth': 3, 'learning_rate': 0.2665276315437317, 'subsample': 0.5676435486529794, 'colsample_bytree': 0.54641
[1] 2025-05-31 02:37:52,394] Trial 2 finished with value: 0.00126586626956596 and parameters: {'n_estimators': 279, 'max_depth': 4, 'learning_rate': 0.01266528065667012, 'subsample': 0.8254297619017925, 'colsample_bytree': 0.7730
[1] 2025-05-31 02:37:52,517] Trial 3 finished with value: 0.00127318568236352 and parameters: {'n_estimators': 237, 'max_depth': 3, 'learning_rate': 0.1248189785456455, 'subsample': 0.7388950859296979, 'colsample_bytree': 0.85976
[1] 2025-05-31 02:37:52,611] Trial 4 finished with value: 0.0012332472653704 and parameters: {'n_estimators': 156, 'max_depth': 3, 'learning_rate': 0.1486528785762277, 'subsample': 0.5572177979951, 'colsample_bytree': 0.552357876
[1] 2025-05-31 02:37:52,733] Trial 5 finished with value: 0.00127862825409542 and parameters: {'n_estimators': 248, 'max_depth': 4, 'learning_rate': 0.01433021749378167, 'subsample': 0.9568305616427853, 'colsample_bytree': 0.6329
[1] 2025-05-31 02:37:52,844] Trial 6 finished with value: 0.00125464372744084 and parameters: {'n_estimators': 72, 'max_depth': 6, 'learning_rate': 0.21796802231951334, 'subsample': 0.63883882654755, 'colsample_bytree': 0.9204283
[1] 2025-05-31 02:37:52,982] Trial 7 finished with value: 0.00124195439355855 and parameters: {'n_estimators': 112, 'max_depth': 3, 'learning_rate': 0.0957587444040103, 'subsample': 0.518491354752417, 'colsample_bytree': 0.9896306
[1] 2025-05-31 02:37:53,033] Trial 8 finished with value: 0.001264167270790877 and parameters: {'n_estimators': 175, 'max_depth': 6, 'learning_rate': 0.092433102693336, 'subsample': 0.59484039550130181, 'colsample_bytree': 0.938635
[1] 2025-05-31 02:37:53,146] Trial 9 finished with value: 0.001256915269303596 and parameters: {'n_estimators': 221, 'max_depth': 7, 'learning_rate': 0.2866338788305104, 'subsample': 0.55174978944818, 'colsample_bytree': 0.554946
[1] 2025-05-31 02:37:53,255] Trial 10 finished with value: 0.001259613600106228 and parameters: {'n_estimators': 136, 'max_depth': 8, 'learning_rate': 0.0407176711967303, 'subsample': 0.700180412638504, 'colsample_bytree': 0.6820
[1] 2025-05-31 02:37:53,359] Trial 11 finished with value: 0.001240826207553344 and parameters: {'n_estimators': 114, 'max_depth': 4, 'learning_rate': 0.046120512183841465, 'subsample': 0.543465873722316, 'colsample_bytree': 0.9932
[1] 2025-05-31 02:37:53,454] Trial 12 finished with value: 0.001251630593589413 and parameters: {'n_estimators': 51, 'max_depth': 4, 'learning_rate': 0.036512051852021586, 'subsample': 0.6249243638522799, 'colsample_bytree': 0.78591
[1] 2025-05-31 02:37:53,553] Trial 13 finished with value: 0.00127476433295203 and parameters: {'n_estimators': 143, 'max_depth': 5, 'learning_rate': 0.02280211249049954, 'subsample': 0.50306439889592451, 'colsample_bytree': 0.5005
[1] 2025-05-31 02:37:53,720] Trial 14 finished with value: 0.001250704074915971 and parameters: {'n_estimators': 209, 'max_depth': 5, 'learning_rate': 0.0646148126937626, 'subsample': 0.617359913628354, 'colsample_bytree': 0.65486
[1] 2025-05-31 02:37:53,830] Trial 15 finished with value: 0.0012546122605629628 and parameters: {'n_estimators': 117, 'max_depth': 4, 'learning_rate': 0.022831229181160887, 'subsample': 0.6814645826510447, 'colsample_bytree': 0.7188
[1] 2025-05-31 02:37:53,957] Trial 16 finished with value: 0.0012470033669059209 and parameters: {'n_estimators': 180, 'max_depth': 3, 'learning_rate': 0.13612758821724222, 'subsample': 0.78761524924230185, 'colsample_bytree': 0.8287
[1] 2025-05-31 02:37:54,054] Trial 17 finished with value: 0.001282993589299297 and parameters: {'n_estimators': 88, 'max_depth': 5, 'learning_rate': 0.0514808938814349, 'subsample': 0.5812099197604899, 'colsample_bytree': 0.998421
[1] 2025-05-31 02:37:54,168] Trial 18 finished with value: 0.001254530243240883 and parameters: {'n_estimators': 147, 'max_depth': 4, 'learning_rate': 0.026860052952871765, 'subsample': 0.686525693166139, 'colsample_bytree': 0.612
[1] 2025-05-31 02:37:54,302] Trial 19 finished with value: 0.00123795794170863 and parameters: {'n_estimators': 194, 'max_depth': 3, 'learning_rate': 0.0545932069809874, 'subsample': 0.5017595744653716, 'colsample_bytree': 0.71799
[1] 2025-05-31 02:37:54,408] A new study created in memory with name: no-name_9ed73092-0064-4db7-456f-4c6d47497
[1] 2025-05-31 02:37:54,542] Trial 0 finished with value: 0.001224680181332224 and parameters: {'n_estimators': 196, 'max_depth': 5, 'learning_rate': 0.03284031271187448, 'subsample': 0.6360636621356579, 'colsample_bytree': 0.83199
[1] 2025-05-31 02:37:54,682] Trial 1 finished with value: 0.001222550301808564 and parameters: {'n_estimators': 261, 'max_depth': 3, 'learning_rate': 0.010323620705682029, 'subsample': 0.54040620310358949, 'colsample_bytree': 0.60244
[1] 2025-05-31 02:37:54,836] Trial 2 finished with value: 0.001255397907285199 and parameters: {'n_estimators': 279, 'max_depth': 6, 'learning_rate': 0.16040075650659298, 'subsample': 0.531012108500989, 'colsample_bytree': 0.77640
[1] 2025-05-31 02:37:54,985] Trial 3 finished with value: 0.001224919547849488 and parameters: {'n_estimators': 299, 'max_depth': 3, 'learning_rate': 0.024208580338593549, 'subsample': 0.772723920383003, 'colsample_bytree': 0.8452
[1] 2025-05-31 02:37:55,131] Trial 4 finished with value: 0.0012232472653704 and parameters: {'n_estimators': 156, 'max_depth': 3, 'learning_rate': 0.1486528785762277, 'subsample': 0.551783242454553, 'colsample_bytree': 0.898932
[1] 2025-05-31 02:37:55,218] Trial 5 finished with value: 0.00122549372985276 and parameters: {'n_estimators': 230, 'max_depth': 4, 'learning_rate': 0.015185268163137447, 'subsample': 0.752177924425491, 'colsample_bytree': 0.5976
[1] 2025-05-31 02:37:55,346] Trial 6 finished with value: 0.0012416422335124 and parameters: {'n_estimators': 253, 'max_depth': 5, 'learning_rate': 0.0887547965645122, 'subsample': 0.6219131559973159, 'colsample_bytree': 0.94170
[1] 2025-05-31 02:37:55,450] Trial 7 finished with value: 0.00123190669146507 and parameters: {'n_estimators': 191, 'max_depth': 3, 'learning_rate': 0.060343145856427, 'subsample': 0.679770219415339, 'colsample_bytree': 0.633529
[1] 2025-05-31 02:37:55,578] Trial 8 finished with value: 0.001220587532954558 and parameters: {'n_estimators': 236, 'max_depth': 4, 'learning_rate': 0.022805732954558, 'subsample': 0.528280781064649, 'colsample_bytree': 0.64564
[1] 2025-05-31 02:37:55,672] Trial 9 finished with value: 0.00120839657195218383 and parameters: {'n_estimators': 148, 'max_depth': 4, 'learning_rate': 0.15227052375482873, 'subsample': 0.790998898917605, 'colsample_bytree': 0.86734
[1] 2025-05-31 02:37:55,787] Trial 10 finished with value: 0.0012255838909969607 and parameters: {'n_estimators': 108, 'max_depth': 8, 'learning_rate': 0.280737193345694, 'subsample': 0.9155283422454553, 'colsample_bytree': 0.98932
[1] 2025-05-31 02:37:55,893] Trial 11 finished with value: 0.0012413150117442479 and parameters: {'n_estimators': 126, 'max_depth': 6, 'learning_rate': 0.1557306201944520597, 'subsample': 0.863988648733224, 'colsample_bytree': 0.7123
[1] 2025-05-31 02:37:55,981] Trial 12 finished with value: 0.00122417409060559 and parameters: {'n_estimators': 60, 'max_depth': 5, 'learning_rate': 0.15848920616240605, 'subsample': 0.8213620051407014701, 'colsample_bytree': 0.527231
[1] 2025-05-31 02:37:56,090] Trial 13 finished with value: 0.00122416304736985 and parameters: {'n_estimators': 126, 'max_depth': 7, 'learning_rate': 0.140688399836301, 'subsample': 0.9842339975198547, 'colsample_bytree': 0.8932
[1] 2025-05-31 02:37:56,212] Trial 14 finished with value: 0.001219651580444462 and parameters: {'n_estimators': 162, 'max_depth': 4, 'learning_rate': 0.10617897801772, 'subsample': 0.7982059626864763, 'colsample_bytree': 0.7245
[1] 2025-05-31 02:37:56,347] Trial 15 finished with value: 0.001218207254663632 and parameters: {'n_estimators': 162, 'max_depth': 5, 'learning_rate': 0.0441379749473001, 'subsample': 0.6985446032469678, 'colsample_bytree': 0.7814
[1] 2025-05-31 02:37:56,446] Trial 16 finished with value: 0.001238817710392727 and parameters: {'n_estimators': 174, 'max_depth': 4, 'learning_rate': 0.047681032168854, 'subsample': 0.8938621684321462, 'colsample_bytree': 0.69141
[1] 2025-05-31 02:37:56,599] Trial 17 finished with value: 0.0012306862907299709 and parameters: {'n_estimators': 296, 'max_depth': 5, 'learning_rate': 0.2016299539592673, 'subsample': 0.699134519148392, 'colsample_bytree': 0.8488
[1] 2025-05-31 02:37:56,764] Trial 18 finished with value: 0.0012216521352517351 and parameters: {'n_estimators': 198, 'max_depth': 3, 'learning_rate': 0.11391713218839404, 'subsample': 0.94344313948757, 'colsample_bytree': 0.90285
[1] 2025-05-31 02:37:57,040] Trial 19 finished with value: 0.00122348586551701 and parameters: {'n_estimators': 143, 'max_depth': 6, 'learning_rate': 0.043832735294089628, 'subsample': 0.8434973664600828, 'colsample_bytree': 0.7957
[1] 2025-05-31 02:37:57,135] Trial 0 finished with value: 0.0008637804249705652 and parameters: {'n_estimators': 188, 'max_depth': 5, 'learning_rate': 0.0104422123877624, 'subsample': 0.945531338365185, 'colsample_bytree': 0.5043
[1] 2025-05-31 02:37:57,223] Trial 1 finished with value: 0.0008616691117147979 and parameters: {'n
```

[1] 2025-05-01 02:38:05.7153 Trial 10 finished with value: 0.000763057822117100 and parameters: {'n\_estimators': 175, 'max\_depth': 6, 'learning\_rate': 0.0353605496470059}, 'subsample': 0.51170491770610397, 'colsample\_bytree': 0.7068

[1] 2025-05-01 02:38:05.6447 Trial 11 finished with value: 0.000790196773196399 and parameters: {'n\_estimators': 191, 'max\_depth': 8, 'learning\_rate': 0.03598991283563825}, 'subsample': 0.5198493512717185, 'colsample\_bytree': 0.6843

[1] 2025-05-01 02:38:05.7755 Trial 12 finished with value: 0.00077392437829715 and parameters: {'n\_estimators': 174, 'max\_depth': 8, 'learning\_rate': 0.0404311627641771, 'subsample': 0.506881038271769, 'colsample\_bytree': 0.70821

[1] 2025-05-01 02:38:05.9111 Trial 13 finished with value: 0.00084036692387868416 and parameters: {'n\_estimators': 201, 'max\_depth': 8, 'learning\_rate': 0.0421539745675643}, 'subsample': 0.5020529067787614, 'colsample\_bytree': 0.67754

[1] 2025-05-01 02:38:06.0911 Trial 14 finished with value: 0.00081981363248358 and parameters: {'n\_estimators': 295, 'max\_depth': 7, 'learning\_rate': 0.038961565805842265}, 'subsample': 0.5835793891249701, 'colsample\_bytree': 0.700

[1] 2025-05-01 02:38:06.2477 Trial 15 finished with value: 0.0008124355237733823 and parameters: {'n\_estimators': 168, 'max\_depth': 8, 'learning\_rate': 0.1154878415710168, 'subsample': 0.585738828363235, 'colsample\_bytree': 0.62269

[1] 2025-05-01 02:38:06.3911 Trial 16 finished with value: 0.000793295790820474 and parameters: {'n\_estimators': 222, 'max\_depth': 7, 'learning\_rate': 0.02574680225467235, 'subsample': 0.5432041257501694, 'colsample\_bytree': 0.753

[1] 2025-05-01 02:38:06.5091 Trial 17 finished with value: 0.000811536309733769 and parameters: {'n\_estimators': 159, 'max\_depth': 8, 'learning\_rate': 0.1018973753173198, 'subsample': 0.64620898794841, 'colsample\_bytree': 0.7378

[1] 2025-05-01 02:38:06.6071 Trial 18 finished with value: 0.000839688457491748 and parameters: {'n\_estimators': 101, 'max\_depth': 7, 'learning\_rate': 0.22195706061254606, 'subsample': 0.8611596736310081, 'colsample\_bytree': 0.6235

[1] 2025-05-01 02:38:06.7501 Trial 19 finished with value: 0.000798416237164208 and parameters: {'n\_estimators': 228, 'max\_depth': 5, 'learning\_rate': 0.042300545300256176, 'subsample': 0.5563432745340987, 'colsample\_bytree': 0.947

[1] 2025-05-01 02:38:06.8611 A new study created in memory with name: no-name-1d77cd6-648a-44ae-abfa-13942a3e3fc

[1] 2025-05-01 02:38:06.9791 Trial 0 finished with value: 0.0016924350603235048 and parameters: {'n\_estimators': 66, 'max\_depth': 4, 'learning\_rate': 0.1988950958366701, 'subsample': 0.9558303417380012, 'colsample\_bytree': 0.7761757

[1] 2025-05-01 02:38:07.0901 Trial 1 finished with value: 0.001691577784563018 and parameters: {'n\_estimators': 191, 'max\_depth': 5, 'learning\_rate': 0.0436759388156099, 'subsample': 0.9159284575786516, 'colsample\_bytree': 0.915928

[1] 2025-05-01 02:38:07.2091 Trial 2 finished with value: 0.0016900860801836686 and parameters: {'n\_estimators': 182, 'max\_depth': 8, 'learning\_rate': 0.024635058324026084, 'subsample': 0.517220459631688, 'colsample\_bytree': 0.74555

[1] 2025-05-01 02:38:07.3501 Trial 3 finished with value: 0.001691867873366004 and parameters: {'n\_estimators': 299, 'max\_depth': 3, 'learning\_rate': 0.0117232360491246783, 'subsample': 0.7724844031200071, 'colsample\_bytree': 0.6705

[1] 2025-05-01 02:38:07.4491 Trial 4 finished with value: 0.00169101600247988 and parameters: {'n\_estimators': 152, 'max\_depth': 3, 'learning\_rate': 0.1661670924653373, 'subsample': 0.81786757235311562, 'colsample\_bytree': 0.99656

[1] 2025-05-01 02:38:07.5641 Trial 5 finished with value: 0.001691973934959933 and parameters: {'n\_estimators': 208, 'max\_depth': 3, 'learning\_rate': 0.151446026904288, 'subsample': 0.882175268457437, 'colsample\_bytree': 0.85528

[1] 2025-05-01 02:38:07.6881 Trial 6 finished with value: 0.001693440376274207 and parameters: {'n\_estimators': 225, 'max\_depth': 3, 'learning\_rate': 0.10696065477031206, 'subsample': 0.58102187692308, 'colsample\_bytree': 0.83502

[1] 2025-05-01 02:38:07.8291 Trial 7 finished with value: 0.001692178903263084 and parameters: {'n\_estimators': 282, 'max\_depth': 4, 'learning\_rate': 0.05697823522007163, 'subsample': 0.787135452947779, 'colsample\_bytree': 0.655683

[1] 2025-05-01 02:38:07.9471 Trial 8 finished with value: 0.00169276878119684 and parameters: {'n\_estimators': 227, 'max\_depth': 3, 'learning\_rate': 0.112446503791143, 'subsample': 0.973490750421174, 'colsample\_bytree': 0.9500543

[1] 2025-05-01 02:38:08.0411 Trial 9 finished with value: 0.00169095874777786 and parameters: {'n\_estimators': 142, 'max\_depth': 3, 'learning\_rate': 0.016416308142977414, 'subsample': 0.7028457885081505, 'colsample\_bytree': 0.62260

[1] 2025-05-01 02:38:08.1401 Trial 10 finished with value: 0.00169172234181891 and parameters: {'n\_estimators': 100, 'max\_depth': 8, 'learning\_rate': 0.01987343160679866, 'subsample': 0.5133063080721336, 'colsample\_bytree': 0.5422

[1] 2025-05-01 02:38:08.2851 Trial 11 finished with value: 0.0016910472661528066 and parameters: {'n\_estimators': 139, 'max\_depth': 8, 'learning\_rate': 0.0326979795331684, 'subsample': 0.6583494598007897, 'colsample\_bytree': 0.640

[1] 2025-05-01 02:38:08.4011 Trial 12 finished with value: 0.00169230628829847 and parameters: {'n\_estimators': 117, 'max\_depth': 7, 'learning\_rate': 0.02093027977007414, 'subsample': 0.6884278880607273, 'colsample\_bytree': 0.5111

[1] 2025-05-01 02:38:08.5121 Trial 13 finished with value: 0.00168919834281879 and parameters: {'n\_estimators': 166, 'max\_depth': 6, 'learning\_rate': 0.07303989465080285, 'subsample': 0.543606453999538, 'colsample\_bytree': 0.73062

[1] 2025-05-01 02:38:08.6501 Trial 14 finished with value: 0.001695932790956904 and parameters: {'n\_estimators': 178, 'max\_depth': 6, 'learning\_rate': 0.0891479526589864, 'subsample': 0.51045076500365, 'colsample\_bytree': 0.73634

[1] 2025-05-01 02:38:08.8031 Trial 15 finished with value: 0.00169391258080794 and parameters: {'n\_estimators': 259, 'max\_depth': 6, 'learning\_rate': 0.02816750602252363, 'subsample': 0.59284483292157, 'colsample\_bytree': 0.7336

[1] 2025-05-01 02:38:08.9281 Trial 16 finished with value: 0.0016902593035933 and parameters: {'n\_estimators': 177, 'max\_depth': 6, 'learning\_rate': 0.0494377408032501, 'subsample': 0.589220663502982, 'colsample\_bytree': 0.8140

[1] 2025-05-01 02:38:09.0241 Trial 17 finished with value: 0.0016899419958649423 and parameters: {'n\_estimators': 81, 'max\_depth': 7, 'learning\_rate': 0.26327696828898206, 'subsample': 0.5014105956942434, 'colsample\_bytree': 0.70552

[1] 2025-05-01 02:38:09.1691 Trial 18 finished with value: 0.00169212952744505 and parameters: {'n\_estimators': 247, 'max\_depth': 5, 'learning\_rate': 0.04248288252353898, 'subsample': 0.6269988524004564, 'colsample\_bytree': 0.5855

[1] 2025-05-01 02:38:09.3221 Trial 19 finished with value: 0.001689430897566572 and parameters: {'n\_estimators': 159, 'max\_depth': 7, 'learning\_rate': 0.07600351774795513, 'subsample': 0.543746648621014, 'colsample\_bytree': 0.8862

[1] 2025-05-01 02:38:09.4381 A new study created in memory with name: no-name-c71113f7-79e7-4e36-898c-9fb55be9e701

[1] 2025-05-01 02:38:09.5461 Trial 0 finished with value: 0.000832947531353518 and parameters: {'n\_estimators': 101, 'max\_depth': 3, 'learning\_rate': 0.1758768493103697, 'subsample': 0.8286378036982761, 'colsample\_bytree': 0.94037

[1] 2025-05-01 02:38:09.6301 Trial 1 finished with value: 0.000833170602930058 and parameters: {'n\_estimators': 114, 'max\_depth': 5, 'learning\_rate': 0.2729620098099758, 'subsample': 0.7525432305446799, 'colsample\_bytree': 0.618959

[1] 2025-05-01 02:38:09.7261 Trial 2 finished with value: 0.000833717794662471 and parameters: {'n\_estimators': 136, 'max\_depth': 5, 'learning\_rate': 0.17615653514909, 'subsample': 0.5983237086276, 'colsample\_bytree': 0.8922584

[1] 2025-05-01 02:38:09.8211 Trial 3 finished with value: 0.000828057082334346 and parameters: {'n\_estimators': 147, 'max\_depth': 7, 'learning\_rate': 0.13361403639710356, 'subsample': 0.604010362820295, 'colsample\_bytree': 0.6651

[1] 2025-05-01 02:38:09.9131 Trial 4 finished with value: 0.000830997246170984 and parameters: {'n\_estimators': 105, 'max\_depth': 5, 'learning\_rate': 0.0513756251215496, 'subsample': 0.8086335051891592, 'colsample\_bytree': 0.75288

[1] 2025-05-01 02:38:10.0121 Trial 5 finished with value: 0.00082975617398711981 and parameters: {'n\_estimators': 145, 'max\_depth': 4, 'learning\_rate': 0.0367596433887819, 'subsample': 0.813666252742373, 'colsample\_bytree': 0.987333

[1] 2025-05-01 02:38:10.1351 Trial 6 finished with value: 0.0008302374951667402 and parameters: {'n\_estimators': 237, 'max\_depth': 8, 'learning\_rate': 0.09165441147802313, 'subsample': 0.97584025480394, 'colsample\_bytree': 0.612455

[1] 2025-05-01 02:38:10.2511 Trial 7 finished with value: 0.00083240249611879 and parameters: {'n\_estimators': 202, 'max\_depth': 4, 'learning\_rate': 0.24950940478648148, 'subsample': 0.7001786873876628, 'colsample\_bytree': 0.99802

[1] 2025-05-01 02:38:10.3881 Trial 8 finished with value: 0.0008303971078087887 and parameters: {'n\_estimators': 160, 'max\_depth': 6, 'learning\_rate': 0.1570898331805953, 'subsample': 0.7283291937694547, 'colsample\_bytree': 0.94079

[1] 2025-05-01 02:38:10.5131 Trial 9 finished with value: 0.0008337177515979295 and parameters: {'n\_estimators': 292, 'max\_depth': 4, 'learning\_rate': 0.195349039732152, 'subsample': 0.83807131168475, 'colsample\_bytree': 0.56645

[1] 2025-05-01 02:38:10.6181 Trial 10 finished with value: 0.0008289743177143468 and parameters: {'n\_estimators': 54, 'max\_depth': 8, 'learning\_rate': 0.107985290550418, 'subsample': 0.5325054928545585, 'colsample\_bytree': 0.7311

[1] 2025-05-01 02:38:10.7021 Trial 11 finished with value: 0.0008303617603754855 and parameters: {'n\_estimators': 50, 'max\_depth': 8, 'learning\_rate': 0.12144138239877387, 'subsample': 0.5206434264448485, 'colsample\_bytree': 0.72548

[1] 2025-05-01 02:38:10.7971 Trial 12 finished with value: 0.00083030103740874487 and parameters: {'n\_estimators': 50, 'max\_depth': 7, 'learning\_rate': 0.01202370856148292, 'subsample': 0.542589318407347, 'colsample\_bytree': 0.7192

[1] 2025-05-01 02:38:10.9811 Trial 13 finished with value: 0.0008356793441861049 and parameters: {'n\_estimators': 224, 'max\_depth': 7, 'learning\_rate': 0.02338349791071815, 'subsample': 0.607182467075285, 'colsample\_bytree': 0.783

[1] 2025-05-01 02:38:11.1581 Trial 14 finished with value: 0.000831981189734761 and parameters: {'n\_estimators': 76, 'max\_depth': 7, 'learning\_rate': 0.01040486794971156, 'subsample': 0.609523198760470578, 'colsample\_bytree': 0.5068

[1] 2025-05-01 02:38:11.1861 Trial 15 finished with value: 0.0008315736711347 and parameters: {'n\_estimators': 193, 'max\_depth': 8, 'learning\_rate': 0.02431480754178164, 'subsample': 0.6080013163470267, 'colsample\_bytree': 0.8222

[1] 2025-05-01 02:38:12.0821 Trial 16 finished with value: 0.0008307348264944011 and parameters: {'n\_estimators': 264, 'max\_depth': 6, 'learning\_rate': 0.0166542872333467, 'subsample': 0.5620618195319816, 'colsample\_bytree': 0.681

[1] 2025-05-01 02:38:13.1501 Trial 17 finished with value: 0.00082912159274105 and parameters: {'n\_estimators': 81, 'max\_depth': 8, 'learning\_rate': 0.031703925951516947, 'subsample': 0.658878160894474, 'colsample\_bytree': 0.65280

[1] 2025-05-01 02:38:13.3131 Trial 18 finished with value: 0.00082936181326684 and parameters: {'n\_estimators': 128, 'max\_depth': 6, 'learning\_rate': 0.07304964636091803, 'subsample': 0.6635179617401612, 'colsample\_bytree': 0.8371

[1] 2025-05-01 02:38:13.5071 A new study created in memory with name: no-name-9d4575fa-f238-4ab2-9eb0-86c3048822f3

[1] 2025-05-01 02:38:13.6631 Trial 0 finished with value: 0.001442195034336562 and parameters: {'n\_estimators': 58, 'max\_depth': 7, 'learning\_rate': 0.0286988991552228, 'subsample': 0.670863910833991, 'colsample\_bytree': 0.70692

[1] 2025-05-01 02:38:13.8841 Trial 1 finished with value: 0.001439516983440239 and parameters: {'n\_estimators': 89, 'max\_depth': 3, 'learning\_rate': 0.01196200851070179, 'subsample': 0.6216358176887925, 'colsample\_bytree': 0.80161

[1] 2025-05-01 02:38:13.9551 Trial 2 finished with value: 0.00143660709345553 and parameters: {'n\_estimators': 71, 'max\_depth': 3, 'learning\_rate': 0.0173077117450172, 'subsample': 0.9247860683292157, 'colsample\_bytree': 0.919402

[1] 2025-05-01 02:38:14.0351 Trial 3 finished with value: 0.00144673549750379 and parameters: {'n\_estimators': 147, 'max\_depth': 7, 'learning\_rate': 0.013361403639710356, 'subsample': 0.73232369970854, 'colsample\_bytree': 0.646922

[1] 2025-05-01 02:38:14.1531 Trial 4 finished with value: 0.001409979246170984 and parameters: {'n\_estimators': 105, 'max\_depth': 5, 'learning\_rate': 0.0513756251215496, 'subsample': 0.8086335051891592, 'colsample\_bytree': 0.75288

[1] 2025-05-01 02:38:14.3291 Trial 5 finished with value: 0.001437807576421375 and parameters: {'n\_estimators': 112, 'max\_depth': 4, 'learning\_rate': 0.0151537608302343, 'subsample': 0.808381504045673, 'colsample\_bytree': 0.927398

[1] 2025-05-01 02:38:14.4561 Trial 7 finished with value: 0.0014370425064381133 and parameters: {'n\_estimators': 202, 'max\_depth': 4, 'learning\_rate': 0.24950940478648148, 'subsample': 0.787531977444315, 'colsample\_bytree': 0.9541

[1] 2025-05-01 02:38:14.6001 Trial 8 finished with value: 0.001432373476097875 and parameters: {'n\_estimators': 238, 'max\_depth': 4, 'learning\_rate': 0.0162856950389165784, 'subsample': 0.8595600404605882, 'colsample\_bytree': 0.6212

[1] 2025-05-01 02:38:14.7121 Trial 9 finished with value: 0.0014375755825622 and parameters: {'n\_estimators': 188, 'max\_depth': 8, 'learning\_rate': 0.01731216535474435, 'subsample': 0.774484469444315, 'colsample\_bytree': 0.764948466

[1] 2025-05-01 02:38:14.8731 Trial 10 finished with value: 0.0014484266995249401 and parameters: {'n\_estimators': 293, 'max\_depth': 5, 'learning\_rate': 0.28747540430984097, 'subsample': 0.5375130206369742, 'colsample\_bytree': 0.5336

[1] 2025-05-01 02:38:15.0151 Trial 11 finished with value: 0.0014395213133313696 and parameters: {'n\_estimators': 238, 'max\_depth': 6, 'learning\_rate': 0.056887836367268, 'subsample': 0.981443833040975, 'colsample\_bytree': 0.5742

[1] 2025-05-01 02:38:15.1851 Trial 12 finished with value: 0.001412615195948736 and parameters: {'n\_estimators': 240, 'max\_depth': 6, 'learning\_rate': 0.03265726116565076, 'subsample': 0.522570116565076, 'colsample\_bytree': 0.844

[1] 2025-05-01 02:38:15.2931 Trial 13 finished with value: 0.001441717812459359 and parameters: {'n\_estimators': 197, 'max\_depth': 5, 'learning\_rate': 0.0306782173063393, 'subsample': 0.501211357671453, 'colsample\_bytree': 0.847572

[1] 2025-05-01 02:38:15.4591 Trial 14 finished with value: 0.00142036830232331 and parameters: {'n\_estimators': 290, 'max\_depth': 6, 'learning\_rate': 0.075355220596704236, 'subsample': 0.75535522052395674, 'colsample\_bytree': 0.8637

[1] 2025-05-01 02:38:15.6311 Trial 15 finished with value: 0.001409208358595405 and parameters: {'n\_estimators': 223, 'max\_depth': 7, 'learning\_rate': 0.032679570777284885, 'subsample': 0.57651125083575945, 'colsample\_bytree': 0.979

[1] 2025-05-01 02:38:15.7521 Trial 16 finished with value: 0.001407409770780377 and parameters: {'n\_estimators': 148, 'max\_depth': 5, 'learning\_rate': 0.12545248909873408, 'subsample': 0.5101062354754911, 'colsample\_bytree': 0.8634105

[1] 2025-05-01 02:38:15.8731 Trial 17 finished with value: 0.001453709770730877 and parameters: {'n\_estimators': 154, 'max\_depth': 4, 'learning\_rate': 0.12234756153672171, 'subsample': 0.6133441306335657, 'colsample\_bytree': 0.8870

[1] 2025-05-01 02:38:16.1261 Trial 19 finished with value: 0.001403557956875185 and parameters: {'n\_estimators': 205, 'max\_depth': 5, 'learning\_rate': 0.0307829362030233, 'subsample': 0.616158321574558872, 'colsample\_bytree': 0.6362

[1] 2025-05-01 02:38:16.2571 Trial 20 finished with value: 0.001412615195948736 and parameters: {'n\_estimators': 270, 'max\_depth': 3, 'learning\_rate': 0.1022277577962934, 'subsample': 0.7973090627975985, 'colsample\_bytree': 0.8922

[1] 2025-05-01 02:38:16.3841 Trial 21 finished with value: 0.00089514921737507 and parameters: {'n\_estimators': 138, 'max\_depth': 7, 'learning\_rate': 0.06724737051721743, 'subsample': 0.99710626642545, 'colsample\_bytree': 0.7992015

[1] 2025-05-01 02:38:16.5271 Trial 22 finished with value: 0.000850802125682562 and parameters: {'n\_estimators': 294, 'max\_depth': 5, 'learning\_rate': 0.040378997000270002, 'subsample': 0.995580470395942, 'colsample\_bytree': 0.95009432

[1] 2025-05-01 02:38:16.6761 Trial 23 finished with value: 0.00083827905793658 and parameters: {'n\_estimators': 253, 'max\_depth': 3, 'learning\_rate': 0.088585040386925051, 'subsample': 0.786832301588193, 'colsample\_bytree': 0.83046

[1] 2025-05-01 02:38:16.7961 Trial 24 finished with value: 0.000887164506486751 and parameters: {'n\_estimators': 210, 'max\_depth': 8, 'learning\_rate': 0.0251051228265744, 'subsample': 0.756594634738628, 'colsample\_bytree': 0.56648

[1] 2025-05-01 02:38:16.9171 Trial 25 finished with value: 0.000846897368489715769 and parameters: {'n\_estimators': 235, 'max\_depth': 4, 'learning\_rate': 0.0372361167379153, 'subsample': 0.98674872306360383, 'colsample\_bytree': 0.61183

[1] 2025-05-01 02:38:17.0781 Trial 26 finished with value: 0.000858642592731111 and parameters: {'n\_estimators': 251, 'max\_depth': 5, 'learning\_rate': 0.03067821305898409, 'subsample': 0.703002690710775, 'colsample\_bytree': 0.6362

[1] 2025-05-01 02:38:17.1871 Trial 7 finished with value: 0.0008703139493005186 and parameters: {'n\_estimators': 195, 'max\_depth': 5, 'learning\_rate': 0.0586746458239597651, 'subsample': 0.723287497172133, 'colsample\_bytree': 0.72209

[1] 2025-05-01 02:38:17.2651 Trial 8 finished with value: 0.0008553671274517899 and parameters: {'n\_estimators': 104, 'max\_depth': 6, 'learning\_rate': 0.1140637088155699, 'subsample': 0.94717891684293, 'colsample\_bytree': 0.78618

[1] 2025-05-01 02:38:17.3571 Trial 9 finished with value: 0.0008606477168371973 and parameters: {'n\_estimators': 142, 'max\_depth': 5, 'learning

[1] 2025-05-31 02:38:24,053 Trial 18 finished with value: 0.002919939234988597 and parameters: {'n\_estimators': 289, 'max\_depth': 4, 'learning\_rate': 0.16351428495065036, 'subsample': 0.1545319179081013, 'colsample\_bytree': 0.5707}, [1] 2025-05-31 02:38:24,913 Trial 19 finished with value: 0.0029216576902787 and parameters: {'n\_estimators': 298, 'max\_depth': 5, 'learning\_rate': 0.057422313405015746, 'subsample': 0.6298601399852443, 'colsample\_bytree': 0.7852}, [1] 2025-05-31 02:38:25,149 A new study created in memory with name: no-name-c9b24c51-9431-4e67-875e-068bf010001e, [1] 2025-05-31 02:38:26,097 Trial 0 finished with value: 0.0008804080196808363 and parameters: {'n\_estimators': 136, 'max\_depth': 5, 'learning\_rate': 0.041679190676317146, 'subsample': 0.72323192368801, 'colsample\_bytree': 0.9380}, [1] 2025-05-31 02:38:26,236 Trial 1 finished with value: 0.0008817260561622709 and parameters: {'n\_estimators': 159, 'max\_depth': 7, 'learning\_rate': 0.09281849776349261, 'subsample': 0.5143695467771041, 'colsample\_bytree': 0.6219}, [1] 2025-05-31 02:38:26,440 Trial 2 finished with value: 0.0008834017047007068 and parameters: {'n\_estimators': 221, 'max\_depth': 4, 'learning\_rate': 0.01947781217820794, 'subsample': 0.991428748796186, 'colsample\_bytree': 0.9643}, [1] 2025-05-31 02:38:26,596 Trial 3 finished with value: 0.000884905934578796 and parameters: {'n\_estimators': 180, 'max\_depth': 6, 'learning\_rate': 0.051932576166865058, 'subsample': 0.873513201583165, 'colsample\_bytree': 0.886490}, [1] 2025-05-31 02:38:26,833 Trial 4 finished with value: 0.000884451963703361 and parameters: {'n\_estimators': 245, 'max\_depth': 4, 'learning\_rate': 0.0384720828146405, 'subsample': 0.9788745540149267, 'colsample\_bytree': 0.5815}, [1] 2025-05-31 02:38:27,063 Trial 5 finished with value: 0.0008777018870785657 and parameters: {'n\_estimators': 160, 'max\_depth': 7, 'learning\_rate': 0.057338372854966776, 'subsample': 0.7563401502771356, 'colsample\_bytree': 0.5732}, [1] 2025-05-31 02:38:27,142 Trial 6 finished with value: 0.00087342238199888 and parameters: {'n\_estimators': 105, 'max\_depth': 5, 'learning\_rate': 0.18189882031411658, 'subsample': 0.7774449085166371, 'colsample\_bytree': 0.50764}, [1] 2025-05-31 02:38:27,260 Trial 7 finished with value: 0.000887790145392958 and parameters: {'n\_estimators': 224, 'max\_depth': 8, 'learning\_rate': 0.04820268724451294, 'subsample': 0.9048565169803766, 'colsample\_bytree': 0.75058}, [1] 2025-05-31 02:38:27,393 Trial 8 finished with value: 0.00088219019057923978 and parameters: {'n\_estimators': 289, 'max\_depth': 5, 'learning\_rate': 0.010248031094234389, 'subsample': 0.5752015767391002, 'colsample\_bytree': 0.8237}, [1] 2025-05-31 02:38:27,463 Trial 9 finished with value: 0.0008672294585326742 and parameters: {'n\_estimators': 87, 'max\_depth': 4, 'learning\_rate': 0.2500700620058828, 'subsample': 0.77795524626628, 'colsample\_bytree': 0.594822}, [1] 2025-05-31 02:38:27,549 Trial 10 finished with value: 0.00087326002203912 and parameters: {'n\_estimators': 64, 'max\_depth': 3, 'learning\_rate': 0.2962843417176357, 'subsample': 0.6539561804151159, 'colsample\_bytree': 0.69242}, [1] 2025-05-31 02:38:27,631 Trial 11 finished with value: 0.00924281273917924 and parameters: {'n\_estimators': 53, 'max\_depth': 3, 'learning\_rate': 0.287349321324838095, 'subsample': 0.6399273384147566, 'colsample\_bytree': 0.69173}, [1] 2025-05-31 02:38:27,716 Trial 12 finished with value: 0.00913136388261514 and parameters: {'n\_estimators': 52, 'max\_depth': 3, 'learning\_rate': 0.287383642082146405, 'subsample': 0.6567437910942365, 'colsample\_bytree': 0.694141}, [1] 2025-05-31 02:38:27,812 Trial 13 finished with value: 0.00797854954399075 and parameters: {'n\_estimators': 93, 'max\_depth': 3, 'learning\_rate': 0.1407895510119293, 'subsample': 0.825208143886196, 'colsample\_bytree': 0.66887}, [1] 2025-05-31 02:38:27,916 Trial 14 finished with value: 0.008952545781368942 and parameters: {'n\_estimators': 98, 'max\_depth': 4, 'learning\_rate': 0.155029378971026, 'subsample': 0.66921196228262, 'colsample\_bytree': 0.763449}, [1] 2025-05-31 02:38:28,045 Trial 15 finished with value: 0.008568259886076103 and parameters: {'n\_estimators': 82, 'max\_depth': 4, 'learning\_rate': 0.0863693586739712, 'subsample': 0.5120357547536798, 'colsample\_bytree': 0.5348}, [1] 2025-05-31 02:38:28,153 Trial 16 finished with value: 0.00869527273735368 and parameters: {'n\_estimators': 132, 'max\_depth': 4, 'learning\_rate': 0.08517115034247854, 'subsample': 0.5192646967947587, 'colsample\_bytree': 0.61069}, [1] 2025-05-31 02:38:28,245 Trial 17 finished with value: 0.0086617613605787167 and parameters: {'n\_estimators': 84, 'max\_depth': 6, 'learning\_rate': 0.08517715034247854, 'subsample': 0.5192646967947587, 'colsample\_bytree': 0.61069}, [1] 2025-05-31 02:38:28,359 Trial 18 finished with value: 0.0087730748589353 and parameters: {'n\_estimators': 124, 'max\_depth': 6, 'learning\_rate': 0.0928446677670886, 'subsample': 0.578478647441792, 'colsample\_bytree': 0.637495}, [1] 2025-05-31 02:38:28,488 Trial 19 finished with value: 0.008822159812785713 and parameters: {'n\_estimators': 193, 'max\_depth': 7, 'learning\_rate': 0.03125957523217766, 'subsample': 0.579802103944465, 'colsample\_bytree': 0.5066}, [1] 2025-05-31 02:38:28,577 A new study created in memory with name: no-name-39c02a0b-742d-43eb-a4ab-2728f667723, [1] 2025-05-31 02:38:28,683 Trial 0 finished with value: 0.000896565140299232 and parameters: {'n\_estimators': 98, 'max\_depth': 7, 'learning\_rate': 0.04536024919944616, 'subsample': 0.66693743463187, 'colsample\_bytree': 0.89221}, [1] 2025-05-31 02:38:28,810 Trial 1 finished with value: 0.000991119477670016 and parameters: {'n\_estimators': 245, 'max\_depth': 5, 'learning\_rate': 0.01049659247597506, 'subsample': 0.501690398241083, 'colsample\_bytree': 0.7780}, [1] 2025-05-31 02:38:28,931 Trial 2 finished with value: 0.00977187384445593 and parameters: {'n\_estimators': 209, 'max\_depth': 8, 'learning\_rate': 0.0964830916716719, 'subsample': 0.58428678121007, 'colsample\_bytree': 0.9266550}, [1] 2025-05-31 02:38:28,983 Trial 3 finished with value: 0.00975205185812916 and parameters: {'n\_estimators': 218, 'max\_depth': 6, 'learning\_rate': 0.294313406558971, 'subsample': 0.739531252459205, 'colsample\_bytree': 0.6200596}, [1] 2025-05-31 02:38:29,160 Trial 4 finished with value: 0.001016315585080489 and parameters: {'n\_estimators': 93, 'max\_depth': 6, 'learning\_rate': 0.2530062950577805, 'subsample': 0.5128029887265193, 'colsample\_bytree': 0.965841}, [1] 2025-05-31 02:38:29,284 Trial 5 finished with value: 0.00979540550527328 and parameters: {'n\_estimators': 251, 'max\_depth': 4, 'learning\_rate': 0.08164190478056745, 'subsample': 0.80174447775380, 'colsample\_bytree': 0.5136}, [1] 2025-05-31 02:38:29,418 Trial 6 finished with value: 0.00973284579038621 and parameters: {'n\_estimators': 269, 'max\_depth': 3, 'learning\_rate': 0.25656684531397847, 'subsample': 0.95128146211588, 'colsample\_bytree': 0.9547}, [1] 2025-05-31 02:38:29,506 Trial 7 finished with value: 0.00980374629052502 and parameters: {'n\_estimators': 120, 'max\_depth': 6, 'learning\_rate': 0.0178918736017365, 'subsample': 0.532209652322976, 'colsample\_bytree': 0.954548}, [1] 2025-05-31 02:38:29,591 Trial 8 finished with value: 0.0008740205932894966 and parameters: {'n\_estimators': 122, 'max\_depth': 4, 'learning\_rate': 0.2894642048224202, 'subsample': 0.5687731809251968, 'colsample\_bytree': 0.926954}, [1] 2025-05-31 02:38:29,666 Trial 9 finished with value: 0.009777199185619712 and parameters: {'n\_estimators': 94, 'max\_depth': 6, 'learning\_rate': 0.01755393417347470, 'subsample': 0.97685990011918, 'colsample\_bytree': 0.613716}, [1] 2025-05-31 02:38:29,784 Trial 10 finished with value: 0.00987948865149164 and parameters: {'n\_estimators': 146, 'max\_depth': 3, 'learning\_rate': 0.1204068413949535, 'subsample': 0.8948894215367532, 'colsample\_bytree': 0.8180}, [1] 2025-05-31 02:38:29,907 Trial 11 finished with value: 0.009516627052171686 and parameters: {'n\_estimators': 173, 'max\_depth': 4, 'learning\_rate': 0.0441319610630278, 'subsample': 0.6326101996996309, 'colsample\_bytree': 0.7546}, [1] 2025-05-31 02:38:30,036 Trial 12 finished with value: 0.00953208581929994 and parameters: {'n\_estimators': 167, 'max\_depth': 4, 'learning\_rate': 0.05260564978177294, 'subsample': 0.63874371212179, 'colsample\_bytree': 0.7208}, [1] 2025-05-31 02:38:30,166 Trial 13 finished with value: 0.00939698472941708 and parameters: {'n\_estimators': 172, 'max\_depth': 4, 'learning\_rate': 0.0493105808336584, 'subsample': 0.6266617698215912, 'colsample\_bytree': 0.7044}, [1] 2025-05-31 02:38:30,293 Trial 14 finished with value: 0.00973284397496344 and parameters: {'n\_estimators': 192, 'max\_depth': 5, 'learning\_rate': 0.030749772956028035, 'subsample': 0.6175844714639784, 'colsample\_bytree': 0.6861}, [1] 2025-05-31 02:38:30,380 Trial 15 finished with value: 0.000981902325878838 and parameters: {'n\_estimators': 56, 'max\_depth': 4, 'learning\_rate': 0.0849420261257964, 'subsample': 0.562829968894371, 'colsample\_bytree': 0.837839}, [1] 2025-05-31 02:38:30,498 Trial 16 finished with value: 0.009624373876286202 and parameters: {'n\_estimators': 166, 'max\_depth': 3, 'learning\_rate': 0.071117282936, 'subsample': 0.7148829736743942, 'colsample\_bytree': 0.6521059}, [1] 2025-05-31 02:38:30,637 Trial 17 finished with value: 0.00967295652605599 and parameters: {'n\_estimators': 220, 'max\_depth': 5, 'learning\_rate': 0.1514883414913493, 'subsample': 0.6940363995695986, 'colsample\_bytree': 0.7417282}, [1] 2025-05-31 02:38:30,914 Trial 18 finished with value: 0.009688521702672597 and parameters: {'n\_estimators': 147, 'max\_depth': 3, 'learning\_rate': 0.0599302993935906, 'subsample': 0.7962502812934021, 'colsample\_bytree': 0.8114}, [1] 2025-05-31 02:38:31,023 A new study created in memory with name: no-name-8117177-7236-4b7d-aed8-7b979ff845c, [1] 2025-05-31 02:38:31,192 Trial 0 finished with value: 0.001260115988032349 and parameters: {'n\_estimators': 160, 'max\_depth': 7, 'learning\_rate': 0.2151065948557065, 'subsample': 0.64286361380988, 'colsample\_bytree': 0.671243}, [1] 2025-05-31 02:38:31,258 Trial 1 finished with value: 0.0010304070593818466 and parameters: {'n\_estimators': 62, 'max\_depth': 3, 'learning\_rate': 0.01585715519368724, 'subsample': 0.5705840518525412, 'colsample\_bytree': 0.64480}, [1] 2025-05-31 02:38:31,398 Trial 2 finished with value: 0.00103647120593299456 and parameters: {'n\_estimators': 286, 'max\_depth': 3, 'learning\_rate': 0.017036459816220455, 'subsample': 0.593898678676273, 'colsample\_bytree': 0.8799191}, [1] 2025-05-31 02:38:31,504 Trial 3 finished with value: 0.0010971436402421744 and parameters: {'n\_estimators': 198, 'max\_depth': 7, 'learning\_rate': 0.03625233344069294, 'subsample': 0.6573669144747463, 'colsample\_bytree': 0.51152}, [1] 2025-05-31 02:38:31,599 Trial 4 finished with value: 0.001321425349512885 and parameters: {'n\_estimators': 149, 'max\_depth': 7, 'learning\_rate': 0.120347717795268, 'subsample': 0.8532492537138175, 'colsample\_bytree': 0.674530}, [1] 2025-05-31 02:38:31,694 Trial 5 finished with value: 0.00103923478875203 and parameters: {'n\_estimators': 142, 'max\_depth': 4, 'learning\_rate': 0.04201309263746972, 'subsample': 0.937345088995237, 'colsample\_bytree': 0.57120}, [1] 2025-05-31 02:38:31,832 Trial 6 finished with value: 0.001036705108046983 and parameters: {'n\_estimators': 267, 'max\_depth': 8, 'learning\_rate': 0.010630362803439349, 'subsample': 0.71989023859767, 'colsample\_bytree': 0.99832}, [1] 2025-05-31 02:38:31,938 Trial 7 finished with value: 0.0010957018547214868 and parameters: {'n\_estimators': 178, 'max\_depth': 8, 'learning\_rate': 0.279922945810202, 'subsample': 0.658763310558349, 'colsample\_bytree': 0.8419774}, [1] 2025-05-31 02:38:32,061 Trial 8 finished with value: 0.001031755080059212 and parameters: {'n\_estimators': 239, 'max\_depth': 3, 'learning\_rate': 0.028831857062873, 'subsample': 0.778809321213529, 'colsample\_bytree': 0.682532}, [1] 2025-05-31 02:38:32,149 Trial 9 finished with value: 0.001263716106214847 and parameters: {'n\_estimators': 111, 'max\_depth': 6, 'learning\_rate': 0.29494658816220455, 'subsample': 0.529734752958342, 'colsample\_bytree': 0.711707}, [1] 2025-05-31 02:38:32,343 Trial 10 finished with value: 0.0010305267629156336 and parameters: {'n\_estimators': 58, 'max\_depth': 5, 'learning\_rate': 0.03625233344069294, 'subsample': 0.6573669144747463, 'colsample\_bytree': 0.51152}, [1] 2025-05-31 02:38:32,337 Trial 11 finished with value: 0.001098818047796266 and parameters: {'n\_estimators': 109, 'max\_depth': 6, 'learning\_rate': 0.0441319610630278, 'subsample': 0.6269721719068378, 'colsample\_bytree': 0.7568}, [1] 2025-05-31 02:38:32,438 Trial 12 finished with value: 0.00109583328066526 and parameters: {'n\_estimators': 111, 'max\_depth': 6, 'learning\_rate': 0.05260564978177294, 'subsample': 0.554302766119086, 'colsample\_bytree': 0.61276}, [1] 2025-05-31 02:38:32,581 Trial 13 finished with value: 0.0010867051606160483 and parameters: {'n\_estimators': 267, 'max\_depth': 8, 'learning\_rate': 0.01063028630243994, 'subsample': 0.717760959144228678, 'colsample\_bytree': 0.99832}, [1] 2025-05-31 02:38:32,684 Trial 14 finished with value: 0.00128525743754754776 and parameters: {'n\_estimators': 106, 'max\_depth': 7, 'learning\_rate': 0.17156713759976896, 'subsample': 0.6167380704622284, 'colsample\_bytree': 0.7277}, [1] 2025-05-31 02:38:32,798 Trial 15 finished with value: 0.001330735267324304 and parameters: {'n\_estimators': 144, 'max\_depth': 6, 'learning\_rate': 0.27201645230562868, 'subsample': 0.5061253571605462, 'colsample\_bytree': 0.58483}, [1] 2025-05-31 02:38:32,863 Trial 16 finished with value: 0.00132761731929962 and parameters: {'n\_estimators': 84, 'max\_depth': 5, 'learning\_rate': 0.1063632024064516, 'subsample': 0.57237345246127, 'colsample\_bytree': 0.9188274}, [1] 2025-05-31 02:38:33,027 Trial 17 finished with value: 0.001276473136465098 and parameters: {'n\_estimators': 169, 'max\_depth': 5, 'learning\_rate': 0.06865768383373298, 'subsample': 0.7053232544147567, 'colsample\_bytree': 0.7864}, [1] 2025-05-31 02:38:33,138 Trial 18 finished with value: 0.001310917239319156 and parameters: {'n\_estimators': 126, 'max\_depth': 8, 'learning\_rate': 0.146720089419966, 'subsample': 0.8296490679633133, 'colsample\_bytree': 0.70866}, [1] 2025-05-31 02:38:33,293 Trial 19 finished with value: 0.00134117750957632 and parameters: {'n\_estimators': 218, 'max\_depth': 4, 'learning\_rate': 0.0597002893935906, 'subsample': 0.7962502812934021, 'colsample\_bytree': 0.638278}, [1] 2025-05-31 02:38:33,399 A new study created in memory with name: no-name-6116fb33-06cb-4490-99be-8360f88767630, [1] 2025-05-31 02:38:33,533 Trial 0 finished with value: 0.0012836136178337837 and parameters: {'n\_estimators': 131, 'max\_depth': 3, 'learning\_rate': 0.14869062214301523, 'subsample': 0.633229331918316, 'colsample\_bytree': 0.707685}, [1] 2025-05-31 02:38:33,670 Trial 1 finished with value: 0.001069471052984273 and parameters: {'n\_estimators': 160, 'max\_depth': 3, 'learning\_rate': 0.0158715519368724, 'subsample': 0.5705840518525412, 'colsample\_bytree': 0.64480}, [1] 2025-05-31 02:38:33,744 Trial 2 finished with value: 0.00102361360592124847 and parameters: {'n\_estimators': 111, 'max\_depth': 6, 'learning\_rate': 0.29494658816220455, 'subsample': 0.529734752958342, 'colsample\_bytree': 0.8799191}, [1] 2025-05-31 02:38:33,905 Trial 3 finished with value: 0.001155013066544682 and parameters: {'n\_estimators': 262, 'max\_depth': 3, 'learning\_rate': 0.0173529995658863, 'subsample': 0.5739854134354341, 'colsample\_bytree': 0.8885}, [1] 2025-05-31 02:38:34,023 Trial 4 finished with value: 0.0010213243733496164 and parameters: {'n\_estimators': 198, 'max\_depth': 3, 'learning\_rate': 0.0308234371959274, 'subsample': 0.67630908875548, 'colsample\_bytree': 0.5573085}, [1] 2025-05-31 02:38:34,155 Trial 5 finished with value: 0.00119314955147849 and parameters: {'n\_estimators': 255, 'max\_depth': 5, 'learning\_rate': 0.011785838603948765, 'subsample': 0.601064442333571, 'colsample\_bytree': 0.7927}, [1] 2025-05-31 02:38:34,299 Trial 6 finished with value: 0.001166589593278278 and parameters: {'n\_estimators': 134, 'max\_depth': 7, 'learning\_rate': 0.1243071336832892, 'subsample': 0.5964126705704316, 'colsample\_bytree': 0.94676}, [1] 2025-05-31 02:38:35,369 Trial 7 finished with value: 0.0010123403145345354 and parameters: {'n\_estimators': 186, 'max\_depth': 4, 'learning\_rate': 0.08234601724877346, 'subsample': 0.7746701324399762, 'colsample\_bytree': 0.98211}, [1] 2025-05-31 02:38:34,491 Trial 8 finished with value: 0.001016645293455249 and parameters: {'n\_estimators': 138, 'max\_depth': 6, 'learning\_rate': 0.26789795667051497, 'subsample': 0.8220067964861801, 'colsample\_bytree': 0.63260}, [1] 2025-05-31 02:38:34,691 Trial 10 finished with value: 0.0011140597965788923 and parameters: {'n\_estimators': 53, 'max\_depth': 8, 'learning\_rate': 0.406460767392568, 'subsample': 0.911239519849802, 'colsample\_bytree': 0.8009}, [1] 2025-05-31 02:38:34,792 Trial 11 finished with value: 0.00115461954588632763 and parameters: {'n\_estimators': 77, 'max\_depth': 8, 'learning\_rate': 0.287113624045653, 'subsample': 0.5455670513522189, 'colsample\_bytree': 0.65987}, [1] 2025-05-31 02:38:34,930 Trial 12 finished with value: 0.00111464373887762709 and parameters: {'n\_estimators': 207, 'max\_depth': 7, 'learning\_rate': 0.1033344343765045, 'subsample': 0.8681935437650945, 'colsample\_bytree': 0.866568}, [1] 2025-05-31 02:38:35,040 Trial 13 finished with value: 0.0011328923462326 and parameters: {'n\_estimators': 118, 'max\_depth': 7, 'learning\_rate': 0.06410355136056076, 'subsample': 0.50842876651939, 'colsample\_bytree': 0.625466}, [1] 2025-05-31 02:38:35,163 Trial 14 finished with value: 0.0011166589593278278 and parameters: {'n\_estimators': 174, 'max\_depth': 7, 'learning\_rate': 0.1432071336832892, 'subsample': 0.5964126705704316, 'colsample\_bytree': 0.52046}, [1] 2025-05-31 02:38:35,369 Trial 15 finished with value: 0.001123613605862127 and parameters: {'n\_estimators': 299, 'max\_depth': 5, 'learning\_rate': 0.1792418898201268, 'subsample': 0.838894628136803, 'colsample\_bytree': 0.7331}, [1] 2025-05-31 02:38:35,508 Trial 16 finished with value: 0.00111801846583706 and parameters: {'n\_estimators': 96, 'max\_depth': 6, 'learning\_rate': 0.0302326571388768, 'subsample': 0.595664268353057, 'colsample\_bytree': 0.796684}, [1] 2025-05-31 02:38:35,733 Trial 18 finished with value: 0.0011558304101706076 and parameters: {'n\_estimators': 155, 'max\_depth': 7, 'learning\_rate': 0.0302326571388768, 'subsample': 0.9075103985512513, 'colsample\_bytree': 0.9978}, [1] 2025-05-31 02:38:35,963 A new study created in memory with name: no-name-b0644fa-fbd7-4fd1-b

[1] 2025-05-31 02:38:43, 059] Trial 5 finished with value: 0.0007068395354606432 and parameters: {'n\_estimators': 102, 'max\_depth': 7, 'learning\_rate': 0.07213290776839064, 'subsample': 0.7736409943379896, 'colsample\_bytree': 0.59322}, [1] 2025-05-31 02:38:43, 196] Trial 6 finished with value: 0.000701074469593937 and parameters: {'n\_estimators': 292, 'max\_depth': 8, 'learning\_rate': 0.025447450621124027, 'subsample': 0.540499302215555, 'colsample\_bytree': 0.60697}, [1] 2025-05-31 02:38:43, 280] Trial 7 finished with value: 0.000732142639706064 and parameters: {'n\_estimators': 121, 'max\_depth': 5, 'learning\_rate': 0.15795698479407956, 'subsample': 0.564416350728138, 'colsample\_bytree': 0.67001}, [1] 2025-05-31 02:38:43, 376] Trial 8 finished with value: 0.000708053518468176 and parameters: {'n\_estimators': 165, 'max\_depth': 4, 'learning\_rate': 0.0675700789240427, 'subsample': 0.8329268213498372, 'colsample\_bytree': 0.68659}, [1] 2025-05-31 02:38:43, 497] Trial 9 finished with value: 0.000709564148716662 and parameters: {'n\_estimators': 223, 'max\_depth': 8, 'learning\_rate': 0.14664653705589356, 'subsample': 0.8788941557946596, 'colsample\_bytree': 0.96569}, [1] 2025-05-31 02:38:43, 695] Trial 10 finished with value: 0.0007059363665835262 and parameters: {'n\_estimators': 298, 'max\_depth': 7, 'learning\_rate': 0.0313497787176269, 'subsample': 0.503534156685319, 'colsample\_bytree': 0.80858}, [1] 2025-05-31 02:38:43, 816] Trial 11 finished with value: 0.0007108301652791615 and parameters: {'n\_estimators': 155, 'max\_depth': 6, 'learning\_rate': 0.038619513632448114, 'subsample': 0.6503253373947881, 'colsample\_bytree': 0.5059}, [1] 2025-05-31 02:38:43, 964] Trial 12 finished with value: 0.000701053634725917 and parameters: {'n\_estimators': 238, 'max\_depth': 3, 'learning\_rate': 0.023774601570106854, 'subsample': 0.6823188483340505, 'colsample\_bytree': 0.5057}, [1] 2025-05-31 02:38:44, 111] Trial 13 finished with value: 0.0007154182971105743 and parameters: {'n\_estimators': 244, 'max\_depth': 6, 'learning\_rate': 0.259067562835507, 'subsample': 0.6743752559277208, 'colsample\_bytree': 0.57746}, [1] 2025-05-31 02:38:44, 283] Trial 14 finished with value: 0.000708126446212942 and parameters: {'n\_estimators': 298, 'max\_depth': 4, 'learning\_rate': 0.01094708104020913, 'subsample': 0.6533707086516314, 'colsample\_bytree': 0.787}, [1] 2025-05-31 02:38:44, 426] Trial 15 finished with value: 0.006863859589489349 and parameters: {'n\_estimators': 221, 'max\_depth': 7, 'learning\_rate': 0.2939638579629987, 'subsample': 0.57291019732374, 'colsample\_bytree': 0.56141}, [1] 2025-05-31 02:38:44, 557] Trial 16 finished with value: 0.000771664090468566 and parameters: {'n\_estimators': 207, 'max\_depth': 7, 'learning\_rate': 0.2634898466696095, 'subsample': 0.5968561293650911, 'colsample\_bytree': 0.51397}, [1] 2025-05-31 02:38:44, 692] Trial 17 finished with value: 0.00071660897399443 and parameters: {'n\_estimators': 203, 'max\_depth': 6, 'learning\_rate': 0.201414924273244, 'subsample': 0.7099848637935082, 'colsample\_bytree': 0.73382}, [1] 2025-05-31 02:38:44, 846] Trial 18 finished with value: 0.006705409800318 and parameters: {'n\_estimators': 250, 'max\_depth': 5, 'learning\_rate': 0.29363863104016913, 'subsample': 0.6180180740523521, 'colsample\_bytree': 0.5604}, [1] 2025-05-31 02:38:44, 999] Trial 19 finished with value: 0.000699632789420459 and parameters: {'n\_estimators': 259, 'max\_depth': 5, 'learning\_rate': 0.0911380864098362, 'subsample': 0.606560604499752, 'colsample\_bytree': 0.8479}, [1] 2025-05-31 02:38:45, 133] A new study created in memory with name: no-name-fd4fe4dd-40e4-ba0c-5c9a54029a3 [1] 2025-05-31 02:38:45, 245] Trial 0 finished with value: 0.0088557818550597 and parameters: {'n\_estimators': 111, 'max\_depth': 5, 'learning\_rate': 0.14853388112063492, 'subsample': 0.5564577513003408, 'colsample\_bytree': 0.722681}, [1] 2025-05-31 02:38:45, 348] Trial 1 finished with value: 0.00285156320375151894 and parameters: {'n\_estimators': 181, 'max\_depth': 3, 'learning\_rate': 0.0316521198437931, 'subsample': 0.7264266726717955, 'colsample\_bytree': 0.59406}, [1] 2025-05-31 02:38:45, 423] Trial 2 finished with value: 0.00284998514393331 and parameters: {'n\_estimators': 86, 'max\_depth': 3, 'learning\_rate': 0.030586853647555674, 'subsample': 0.7688729102461245, 'colsample\_bytree': 0.759024}, [1] 2025-05-31 02:38:45, 564] Trial 3 finished with value: 0.00285634363008431 and parameters: {'n\_estimators': 266, 'max\_depth': 3, 'learning\_rate': 0.0867853975975003, 'subsample': 0.937934488595916, 'colsample\_bytree': 0.916400}, [1] 2025-05-31 02:38:45, 718] Trial 4 finished with value: 0.002842604653762379 and parameters: {'n\_estimators': 224, 'max\_depth': 4, 'learning\_rate': 0.04601919288294968, 'subsample': 0.736958456305734, 'colsample\_bytree': 0.533109}, [1] 2025-05-31 02:38:45, 843] Trial 5 finished with value: 0.002845984588674666 and parameters: {'n\_estimators': 226, 'max\_depth': 4, 'learning\_rate': 0.1175022702633076, 'subsample': 0.5784500061621089, 'colsample\_bytree': 0.5259}, [1] 2025-05-31 02:38:45, 983] Trial 6 finished with value: 0.0028486057559645426 and parameters: {'n\_estimators': 291, 'max\_depth': 8, 'learning\_rate': 0.0381210282186274, 'subsample': 0.5643260247167741, 'colsample\_bytree': 0.8881}, [1] 2025-05-31 02:38:46, 133] Trial 7 finished with value: 0.00285265057559645426 and parameters: {'n\_estimators': 300, 'max\_depth': 5, 'learning\_rate': 0.010525744764017369, 'subsample': 0.668726957702055, 'colsample\_bytree': 0.95524}, [1] 2025-05-31 02:38:46, 233] Trial 8 finished with value: 0.002848646223084241 and parameters: {'n\_estimators': 167, 'max\_depth': 6, 'learning\_rate': 0.03569020605201973, 'subsample': 0.6519926758373421, 'colsample\_bytree': 0.977044}, [1] 2025-05-31 02:38:46, 344] Trial 9 finished with value: 0.00285024474555226 and parameters: {'n\_estimators': 209, 'max\_depth': 4, 'learning\_rate': 0.1389673882241538, 'subsample': 0.9246313939457521, 'colsample\_bytree': 0.7069544}, [1] 2025-05-31 02:38:46, 449] Trial 10 finished with value: 0.00285914831030383 and parameters: {'n\_estimators': 115, 'max\_depth': 7, 'learning\_rate': 0.276404565203735, 'subsample': 0.8201468370153139, 'colsample\_bytree': 0.51243607}, [1] 2025-05-31 02:38:46, 591] Trial 11 finished with value: 0.00285173052254484 and parameters: {'n\_estimators': 233, 'max\_depth': 4, 'learning\_rate': 0.02825022701652369, 'subsample': 0.5175022702633076, 'colsample\_bytree': 0.50548}, [1] 2025-05-31 02:38:46, 767] Trial 12 finished with value: 0.002852407165152369 and parameters: {'n\_estimators': 233, 'max\_depth': 4, 'learning\_rate': 0.01464322570178561, 'subsample': 0.8463838743005433, 'colsample\_bytree': 0.615}, [1] 2025-05-31 02:38:46, 890] Trial 13 finished with value: 0.00285494227507891841 and parameters: {'n\_estimators': 163, 'max\_depth': 6, 'learning\_rate': 0.06643262569579989, 'subsample': 0.632337883605253452, 'colsample\_bytree': 0.60422}, [1] 2025-05-31 02:38:47, 038] Trial 14 finished with value: 0.002852057089718491 and parameters: {'n\_estimators': 238, 'max\_depth': 4, 'learning\_rate': 0.1206235645305202, 'subsample': 0.7264096717856075, 'colsample\_bytree': 0.55192}, [1] 2025-05-31 02:38:47, 177] Trial 15 finished with value: 0.028781530755788748 and parameters: {'n\_estimators': 198, 'max\_depth': 5, 'learning\_rate': 0.2677629102034026, 'subsample': 0.6082195642071475, 'colsample\_bytree': 0.664625}, [1] 2025-05-31 02:38:47, 293] Trial 16 finished with value: 0.00285687179043482 and parameters: {'n\_estimators': 141, 'max\_depth': 4, 'learning\_rate': 0.0235561583704475, 'subsample': 0.863083721219587, 'colsample\_bytree': 0.788970}, [1] 2025-05-31 02:38:47, 448] Trial 17 finished with value: 0.002857527386984742 and parameters: {'n\_estimators': 259, 'max\_depth': 5, 'learning\_rate': 0.0521925077036795, 'subsample': 0.9390361611226757, 'colsample\_bytree': 0.6564}, [1] 2025-05-31 02:38:47, 590] Trial 18 finished with value: 0.0028364849573902446 and parameters: {'n\_estimators': 203, 'max\_depth': 3, 'learning\_rate': 0.1737295096734808, 'subsample': 0.7697648153826882, 'colsample\_bytree': 0.5580}, [1] 2025-05-31 02:38:47, 761] Trial 19 finished with value: 0.00285946808269363 and parameters: {'n\_estimators': 200, 'max\_depth': 3, 'learning\_rate': 0.201402931303003, 'subsample': 0.7951613751065498, 'colsample\_bytree': 0.705913}, [1] 2025-05-31 02:38:47, 886] A new study created in memory with name: no-name-74d8278-7800-4452-9464-c6e520a3120 [1] 2025-05-31 02:38:48, 015] Trial 0 finished with value: 0.00137613150203849 and parameters: {'n\_estimators': 172, 'max\_depth': 6, 'learning\_rate': 0.0847303492900392, 'subsample': 0.7328391023621377, 'colsample\_bytree': 0.527694}, [1] 2025-05-31 02:38:48, 123] Trial 1 finished with value: 0.0013507840395862363 and parameters: {'n\_estimators': 187, 'max\_depth': 3, 'learning\_rate': 0.13833652007816633, 'subsample': 0.9257978409339368, 'colsample\_bytree': 0.95989}, [1] 2025-05-31 02:38:48, 193] Trial 2 finished with value: 0.001370850977261246 and parameters: {'n\_estimators': 72, 'max\_depth': 3, 'learning\_rate': 0.0245351788072092193, 'subsample': 0.755492061695016, 'colsample\_bytree': 0.89361}, [1] 2025-05-31 02:38:48, 258] Trial 3 finished with value: 0.0013629781023674367 and parameters: {'n\_estimators': 56, 'max\_depth': 3, 'learning\_rate': 0.0373780856352957633, 'subsample': 0.87570581304503433, 'colsample\_bytree': 0.568934}, [1] 2025-05-31 02:38:48, 367] Trial 4 finished with value: 0.0013628145368491015 and parameters: {'n\_estimators': 200, 'max\_depth': 3, 'learning\_rate': 0.03795532459259506, 'subsample': 0.625188072344922, 'colsample\_bytree': 0.656870}, [1] 2025-05-31 02:38:48, 476] Trial 5 finished with value: 0.00135917020165225448 and parameters: {'n\_estimators': 233, 'max\_depth': 4, 'learning\_rate': 0.01464322570178561, 'subsample': 0.856383743005433, 'colsample\_bytree': 0.615}, [1] 2025-05-31 02:38:48, 577] Trial 6 finished with value: 0.001364448150964025 and parameters: {'n\_estimators': 163, 'max\_depth': 6, 'learning\_rate': 0.030293967262284724, 'subsample': 0.84542817492156, 'colsample\_bytree': 0.81176}, [1] 2025-05-31 02:38:48, 922] Trial 7 finished with value: 0.001363995277998355 and parameters: {'n\_estimators': 168, 'max\_depth': 6, 'learning\_rate': 0.030293967262284724, 'subsample': 0.853029614765380, 'colsample\_bytree': 0.554559}, [1] 2025-05-31 02:38:49, 089] Trial 10 finished with value: 0.0013644694358434294 and parameters: {'n\_estimators': 300, 'max\_depth': 8, 'learning\_rate': 0.010756321737277917, 'subsample': 0.947617691781288, 'colsample\_bytree': 0.743}, [1] 2025-05-31 02:38:49, 237] Trial 11 finished with value: 0.0013646175157932187 and parameters: {'n\_estimators': 252, 'max\_depth': 4, 'learning\_rate': 0.105626201882169148, 'subsample': 0.9884341652848393, 'colsample\_bytree': 0.81828}, [1] 2025-05-31 02:38:49, 352] Trial 12 finished with value: 0.001360981529373585 and parameters: {'n\_estimators': 116, 'max\_depth': 5, 'learning\_rate': 0.14239576229949636, 'subsample': 0.9179873242460869, 'colsample\_bytree': 0.9723}, [1] 2025-05-31 02:38:49, 477] Trial 13 finished with value: 0.001357620027255891 and parameters: {'n\_estimators': 231, 'max\_depth': 5, 'learning\_rate': 0.074241387632305, 'subsample': 0.7633298161023054, 'colsample\_bytree': 0.7045}, [1] 2025-05-31 02:38:49, 673] Trial 14 finished with value: 0.001365189826363142 and parameters: {'n\_estimators': 125, 'max\_depth': 4, 'learning\_rate': 0.1667639166077016, 'subsample': 0.805807663945364, 'colsample\_bytree': 0.8592871}, [1] 2025-05-31 02:38:49, 871] Trial 15 finished with value: 0.00136449807801202167 and parameters: {'n\_estimators': 247, 'max\_depth': 4, 'learning\_rate': 0.29329042889328, 'subsample': 0.9379781953716424, 'colsample\_bytree': 0.64153}, [1] 2025-05-31 02:38:49, 871] Trial 16 finished with value: 0.001364449807801202167 and parameters: {'n\_estimators': 120, 'max\_depth': 5, 'learning\_rate': 0.05908457857925788, 'subsample': 0.913401545235566, 'colsample\_bytree': 0.9072}, [1] 2025-05-31 02:38:50, 043] Trial 17 finished with value: 0.001364832952695655 and parameters: {'n\_estimators': 207, 'max\_depth': 8, 'learning\_rate': 0.0235561583704475, 'subsample': 0.8636759223912094, 'colsample\_bytree': 0.76370}, [1] 2025-05-31 02:38:50, 920] Trial 18 finished with value: 0.001371504582032837 and parameters: {'n\_estimators': 273, 'max\_depth': 7, 'learning\_rate': 0.09401106495216269, 'subsample': 0.8149358373924117, 'colsample\_bytree': 0.93927}, [1] 2025-05-31 02:38:51, 074] Trial 19 finished with value: 0.00136651943324671 and parameters: {'n\_estimators': 92, 'max\_depth': 4, 'learning\_rate': 0.1949574692309072, 'subsample': 0.9581264471389703, 'colsample\_bytree': 0.8189867}, [1] 2025-05-31 02:38:52, 093] A new study created in memory with name: no-name-f5e55f5-2b82-4dd0-b883-0cdec0e0417 [1] 2025-05-31 02:38:52, 386] Trial 0 finished with value: 0.0063368565035183 and parameters: {'n\_estimators': 270, 'max\_depth': 8, 'learning\_rate': 0.01836490099615153, 'subsample': 0.521749803073706, 'colsample\_bytree': 0.6507372}, [1] 2025-05-31 02:38:52, 525] Trial 1 finished with value: 0.0064609355110351 and parameters: {'n\_estimators': 68, 'max\_depth': 6, 'learning\_rate': 0.031306452793568064, 'subsample': 0.9134945454939823, 'colsample\_bytree': 0.993282}, [1] 2025-05-31 02:38:52, 770] Trial 2 finished with value: 0.006460510290329428 and parameters: {'n\_estimators': 185, 'max\_depth': 4, 'learning\_rate': 0.01051350432318896, 'subsample': 0.9911156153919083, 'colsample\_bytree': 0.9716355}, [1] 2025-05-31 02:38:53, 021] Trial 3 finished with value: 0.006460678672031836 and parameters: {'n\_estimators': 197, 'max\_depth': 8, 'learning\_rate': 0.151957063179219, 'subsample': 0.8921947175144746, 'colsample\_bytree': 0.9914429}, [1] 2025-05-31 02:38:53, 163] Trial 4 finished with value: 0.006436175157932187 and parameters: {'n\_estimators': 252, 'max\_depth': 4, 'learning\_rate': 0.10626201882169148, 'subsample': 0.6827152362394975, 'colsample\_bytree': 0.773517}, [1] 2025-05-31 02:38:53, 268] Trial 12 finished with value: 0.006369157393585 and parameters: {'n\_estimators': 116, 'max\_depth': 5, 'learning\_rate': 0.14239576229949636, 'subsample': 0.9179873242460869, 'colsample\_bytree': 0.9723}, [1] 2025-05-31 02:38:53, 342] Trial 13 finished with value: 0.00636981529373585 and parameters: {'n\_estimators': 116, 'max\_depth': 5, 'learning\_rate': 0.14239576229949636, 'subsample': 0.9179873242460869, 'colsample\_bytree': 0.9723}, [1] 2025-05-31 02:38:53, 434] Trial 14 finished with value: 0.0064669512759849932 and parameters: {'n\_estimators': 90, 'max\_depth': 5, 'learning\_rate': 0.014324559647447554, 'subsample': 0.71178118336328962, 'colsample\_bytree': 0.817259}, [1] 2025-05-31 02:38:53, 557] Trial 7 finished with value: 0.006414676276247658 and parameters: {'n\_estimators': 196, 'max\_depth': 4, 'learning\_rate': 0.064200968508650688, 'subsample': 0.696339609490356, 'colsample\_bytree': 0.600092}, [1] 2025-05-31 02:38:53, 619] Trial 9 finished with value: 0.00136021939567643 and parameters: {'n\_estimators': 158, 'max\_depth': 7, 'learning\_rate': 0.284183162138338, 'subsample': 0.62213389360395368, 'colsample\_bytree': 0.554559}, [1] 2025-05-31 02:38:53, 727] Trial 10 finished with value: 0.0064521280721931836 and parameters: {'n\_estimators': 124, 'max\_depth': 3, 'learning\_rate': 0.284682307204914063, 'subsample': 0.7999303373280788, 'colsample\_bytree': 0.50174}, [1] 2025-05-31 02:38:53, 872] Trial 11 finished with value: 0.006439281871349239 and parameters: {'n\_estimators': 236, 'max\_depth': 4, 'learning\_rate': 0.0449114200387576474, 'subsample': 0.5571824106149379, 'colsample\_bytree': 0.655394}, [1] 2025-05-31 02:38:54, 024] Trial 12 finished with value: 0.006469510735872719 and parameters: {'n\_estimators': 139, 'max\_depth': 5, 'learning\_rate': 0.0748374395691922, 'subsample': 0.800037435272871, 'colsample\_bytree': 0.66200}, [1] 2025-05-31 02:38:54, 159] Trial 13 finished with value: 0.00644951063520930496 and parameters: {'n\_estimators': 219, 'max\_depth': 4, 'learning\_rate': 0.04069712047202747447, 'subsample': 0.610004151531643, 'colsample\_bytree': 0.52253}, [1] 2025-05-31 02:38:54, 268] Trial 14 finished with value: 0.00647171079924755 and parameters: {'n\_estimators': 134, 'max\_depth': 4, 'learning\_rate': 0.016146153452540619, 'subsample': 0.7363223948766456, 'colsample\_bytree': 0.5852}, [1] 2025-05-31 02:38:54, 405] Trial 15 finished with value: 0.006437093176185962 and parameters: {'n\_estimators': 231, 'max\_depth': 5, 'learning\_rate': 0.06294146801286154, 'subsample': 0.6546590619863519, 'colsample\_bytree': 0.71156}, [1] 2025-05-31 02:38:54, 519] Trial 16 finished with value: 0.00646870281979844 and parameters: {'n\_estimators': 143, 'max\_depth': 3, 'learning\_rate': 0.03033691970641049, 'subsample': 0.62213389360395368, 'colsample\_bytree': 0.581092}, [1] 2025-05-31 02:38:54, 641] Trial 17 finished with value: 0.006425847472948187 and parameters: {'n\_estimators': 165, 'max\_depth': 6, 'learning\_rate': 0.1792949496146764, 'subsample': 0.74959606255035503, 'colsample\_bytree': 0.593490}, [1] 2025-05-31 02:38:54, 747] Trial 18 finished with value: 0.006461777553377991 and parameters: {'n\_estimators': 101, 'max\_depth': 4, 'learning\_rate': 0.0540514773331342, 'subsample': 0.8531501329902461, 'colsample\_bytree': 0.888569}, [1] 2025-05-31 02:38:54, 887] Trial 19 finished with value: 0.00643611355920508 and parameters: {'n\_estimators': 205, 'max\_depth': 6, 'learning\_rate': 0.01308067088583503, 'subsample': 0.571978408583503, 'colsample\_bytree': 0.7351}, [1] 2025-05-31 02:38:55, 010] A new study created in memory with name: no-name-1b09d6b6-d06-4499-a190-7d60686085 [1] 2025-05-31 02:38:55, 157] Trial 0 finished with value: 0.005630270378341616 and parameters: {'n\_estimators': 218, 'max\_depth': 5, 'learning\_rate': 0.066748944654057, 'subsample': 0.951326965045839, 'colsample\_bytree': 0.849904}, [1] 2025-05-31 02:38:55, 259] Trial 1 finished with value: 0.005626023686574167 and parameters: {'n\_estimators': 157, 'max\_depth': 7, 'learning\_rate': 0.23830671248301973, 'subsample': 0.51861986550206388, 'colsample\_bytree': 0.840391}, [1] 2025-05-31 02:38:55, 499] Trial 3 finished with value: 0.005631769347187129 and parameters: {'n\_estimators': 240, 'max\_depth': 4, 'learning\_rate': 0.014926599077064109, 'subsample': 0.5340350037245688, 'colsample\_bytree': 0.518944}, [1] 2025-05-31 02:38:55, 567] Trial 4 finished with value: 0.00563169893211062 and parameters: {'n\_estimators': 67, 'max\_depth': 5, 'learning\_rate': 0.0210996276057353435, 'subsample': 0.976

[1] 2025-05-31 02:39:01,903 Trial 13 finished with value: 0.005633441720828225 and parameters: {'n\_estimators': 234, 'max\_depth': 7, 'learning\_rate': 0.010432944578593562, 'subsample': 0.6342107699516127, 'colsample\_bytree': 0.896}, [1] 2025-05-31 02:39:01,994 Trial 14 finished with value: 0.00565094542393513 and parameters: {'n\_estimators': 50, 'max\_depth': 7, 'learning\_rate': 0.28738777721468, 'subsample': 0.6580864703188936, 'colsample\_bytree': 0.797532759}, [1] 2025-05-31 02:39:02,114 Trial 15 finished with value: 0.0056303931083458 and parameters: {'n\_estimators': 161, 'max\_depth': 7, 'learning\_rate': 0.1682517101031652, 'subsample': 0.508717895652494, 'colsample\_bytree': 0.93976}, [1] 2025-05-31 02:39:02,279 Trial 16 finished with value: 0.00562272600848022 and parameters: {'n\_estimators': 245, 'max\_depth': 6, 'learning\_rate': 0.09117525907908468, 'subsample': 0.5667155042247262, 'colsample\_bytree': 0.97768}, [1] 2025-05-31 02:39:02,448 Trial 17 finished with value: 0.00556780290705782 and parameters: {'n\_estimators': 296, 'max\_depth': 8, 'learning\_rate': 0.20281814837834, 'subsample': 0.6736262494852693, 'colsample\_bytree': 0.90778979}, [1] 2025-05-31 02:39:02,608 Trial 18 finished with value: 0.0056500779388438 and parameters: {'n\_estimators': 266, 'max\_depth': 8, 'learning\_rate': 0.03928107372694408, 'subsample': 0.500951841206822, 'colsample\_bytree': 0.79526}, [1] 2025-05-31 02:39:02,772 Trial 19 finished with value: 0.00569484510528169 and parameters: {'n\_estimators': 282, 'max\_depth': 8, 'learning\_rate': 0.1900738736240182, 'subsample': 0.6629357838873212, 'colsample\_bytree': 0.89920}, [1] 2025-05-31 02:39:02,936 A new study created in memory with name: no-name-1c83408e-7d4b-bd23-a1e7fcf92165, [1] 2025-05-31 02:39:03,266 Trial 0 finished with value: 0.0028415563794976 and parameters: {'n\_estimators': 291, 'max\_depth': 4, 'learning\_rate': 0.10193883961449543, 'subsample': 0.9398336554098685, 'colsample\_bytree': 0.6771872}, [1] 2025-05-31 02:39:03,925 Trial 1 finished with value: 0.00286648278504055 and parameters: {'n\_estimators': 245, 'max\_depth': 3, 'learning\_rate': 0.08990557074650476, 'subsample': 0.5953538958007575, 'colsample\_bytree': 0.531619}, [1] 2025-05-31 02:39:04,072 Trial 2 finished with value: 0.002830361714034387 and parameters: {'n\_estimators': 134, 'max\_depth': 5, 'learning\_rate': 0.04402826512836906, 'subsample': 0.694101723790191, 'colsample\_bytree': 0.829359}, [1] 2025-05-31 02:39:05,146 Trial 3 finished with value: 0.002838272600848022 and parameters: {'n\_estimators': 260, 'max\_depth': 8, 'learning\_rate': 0.080490857691469455, 'subsample': 0.69119446964855, 'colsample\_bytree': 0.867165}, [1] 2025-05-31 02:39:05,411 Trial 4 finished with value: 0.0028272949232562 and parameters: {'n\_estimators': 205, 'max\_depth': 3, 'learning\_rate': 0.03785576084102154, 'subsample': 0.8411162987579067, 'colsample\_bytree': 0.8579760}, [1] 2025-05-31 02:39:05,598 Trial 5 finished with value: 0.0028473268692538584 and parameters: {'n\_estimators': 215, 'max\_depth': 3, 'learning\_rate': 0.014465097194900226, 'subsample': 0.6335904216315057, 'colsample\_bytree': 0.8486}, [1] 2025-05-31 02:39:05,791 Trial 6 finished with value: 0.002845950720329097 and parameters: {'n\_estimators': 213, 'max\_depth': 8, 'learning\_rate': 0.02561760433232574, 'subsample': 0.8689228165035559, 'colsample\_bytree': 0.90077}, [1] 2025-05-31 02:39:06,070 Trial 7 finished with value: 0.00284615825879369826 and parameters: {'n\_estimators': 183, 'max\_depth': 3, 'learning\_rate': 0.0153996402196718, 'subsample': 0.902846205523961, 'colsample\_bytree': 0.70078}, [1] 2025-05-31 02:39:06,142 Trial 8 finished with value: 0.002839307487047962 and parameters: {'n\_estimators': 59, 'max\_depth': 3, 'learning\_rate': 0.1915393614269874, 'subsample': 0.7635102087474291, 'colsample\_bytree': 0.80934716}, [1] 2025-05-31 02:39:06,255 Trial 9 finished with value: 0.002840582023011061 and parameters: {'n\_estimators': 206, 'max\_depth': 6, 'learning\_rate': 0.010852950635089953, 'subsample': 0.6405153662710297, 'colsample\_bytree': 0.66958}, [1] 2025-05-31 02:39:06,367 Trial 10 finished with value: 0.002844713295265777 and parameters: {'n\_estimators': 131, 'max\_depth': 6, 'learning\_rate': 0.0422292371251232, 'subsample': 0.80582950635089953, 'colsample\_bytree': 0.9912}, [1] 2025-05-31 02:39:06,501 Trial 11 finished with value: 0.002829134512094038 and parameters: {'n\_estimators': 131, 'max\_depth': 5, 'learning\_rate': 0.04228042331863674, 'subsample': 0.7156291947633809, 'colsample\_bytree': 0.76995}, [1] 2025-05-31 02:39:06,607 Trial 12 finished with value: 0.002834542293047642 and parameters: {'n\_estimators': 122, 'max\_depth': 5, 'learning\_rate': 0.02621218957736685, 'subsample': 0.5209986975154917, 'colsample\_bytree': 0.76130}, [1] 2025-05-31 02:39:06,702 Trial 13 finished with value: 0.0028355529536973 and parameters: {'n\_estimators': 83, 'max\_depth': 4, 'learning\_rate': 0.06539676468489849, 'subsample': 0.9990933353424662, 'colsample\_bytree': 0.97485}, [1] 2025-05-31 02:39:06,824 Trial 14 finished with value: 0.0028804428832477367 and parameters: {'n\_estimators': 156, 'max\_depth': 7, 'learning\_rate': 0.1631457032212038, 'subsample': 0.7081938905774983, 'colsample\_bytree': 0.5428}, [1] 2025-05-31 02:39:06,950 Trial 15 finished with value: 0.00284573952937932 and parameters: {'n\_estimators': 167, 'max\_depth': 4, 'learning\_rate': 0.030507587948168343, 'subsample': 0.810858763428893, 'colsample\_bytree': 0.9732}, [1] 2025-05-31 02:39:07,052 Trial 16 finished with value: 0.00284604477705095 and parameters: {'n\_estimators': 95, 'max\_depth': 5, 'learning\_rate': 0.0532109121451616, 'subsample': 0.9463319750580687, 'colsample\_bytree': 0.761124}, [1] 2025-05-31 02:39:07,189 Trial 17 finished with value: 0.002847907208247875 and parameters: {'n\_estimators': 193, 'max\_depth': 7, 'learning\_rate': 0.1977569758321568, 'subsample': 0.705155337721365, 'colsample\_bytree': 0.5986}, [1] 2025-05-31 02:39:07,338 Trial 18 finished with value: 0.002885052949386 and parameters: {'n\_estimators': 238, 'max\_depth': 4, 'learning\_rate': 0.0222515260933364, 'subsample': 0.8283196009867237, 'colsample\_bytree': 0.72249}, [1] 2025-05-31 02:39:07,472 Trial 19 finished with value: 0.002924108464299756 and parameters: {'n\_estimators': 159, 'max\_depth': 6, 'learning\_rate': 0.05050138146436704, 'subsample': 0.5435734906968172, 'colsample\_bytree': 0.7914}, [1] 2025-05-31 02:39:07,572 A new study created in memory with name: no-name-f2dd6bds-193d-4acb-95ec-801dd122e53, [1] 2025-05-31 02:39:07,745 Trial 0 finished with value: 0.00210900157681301 and parameters: {'n\_estimators': 279, 'max\_depth': 5, 'learning\_rate': 0.012229797054421, 'subsample': 0.9654699726487211, 'colsample\_bytree': 0.573600}, [1] 2025-05-31 02:39:07,810 Trial 1 finished with value: 0.002023717666861967 and parameters: {'n\_estimators': 57, 'max\_depth': 6, 'learning\_rate': 0.0977158951839675, 'subsample': 0.96010438091038, 'colsample\_bytree': 0.6562232}, [1] 2025-05-31 02:39:07,889 Trial 2 finished with value: 0.00218754433578126 and parameters: {'n\_estimators': 97, 'max\_depth': 6, 'learning\_rate': 0.124410129309943, 'subsample': 0.9572172152359162, 'colsample\_bytree': 0.955392}, [1] 2025-05-31 02:39:07,978 Trial 3 finished with value: 0.00219505947051654 and parameters: {'n\_estimators': 131, 'max\_depth': 7, 'learning\_rate': 0.0314947744723299, 'subsample': 0.931452952024613, 'colsample\_bytree': 0.98197}, [1] 2025-05-31 02:39:08,108 Trial 4 finished with value: 0.002189531532126856 and parameters: {'n\_estimators': 237, 'max\_depth': 5, 'learning\_rate': 0.05093516467451, 'subsample': 0.7731205102825, 'colsample\_bytree': 0.6472734}, [1] 2025-05-31 02:39:08,211 Trial 5 finished with value: 0.0021018795281531 and parameters: {'n\_estimators': 173, 'max\_depth': 5, 'learning\_rate': 0.05035107995804524, 'subsample': 0.6143391568879755, 'colsample\_bytree': 0.74397}, [1] 2025-05-31 02:39:08,272 Trial 6 finished with value: 0.00211165816121845 and parameters: {'n\_estimators': 50, 'max\_depth': 3, 'learning\_rate': 0.1152010905132432, 'subsample': 0.60380496699448, 'colsample\_bytree': 0.8527494}, [1] 2025-05-31 02:39:08,381 Trial 7 finished with value: 0.00204593106504142 and parameters: {'n\_estimators': 185, 'max\_depth': 3, 'learning\_rate': 0.10458455170321203, 'subsample': 0.9834790849177318, 'colsample\_bytree': 0.8764391}, [1] 2025-05-31 02:39:08,466 Trial 8 finished with value: 0.0021404733101528 and parameters: {'n\_estimators': 89, 'max\_depth': 3, 'learning\_rate': 0.05734844739206126, 'subsample': 0.67852634048438, 'colsample\_bytree': 0.945649}, [1] 2025-05-31 02:39:08,625 Trial 9 finished with value: 0.0021615009832534 and parameters: {'n\_estimators': 295, 'max\_depth': 6, 'learning\_rate': 0.1712720384253858, 'subsample': 0.68137839174826, 'colsample\_bytree': 0.85943051}, [1] 2025-05-31 02:39:08,746 Trial 10 finished with value: 0.0022309167573015913 and parameters: {'n\_estimators': 159, 'max\_depth': 4, 'learning\_rate': 0.0507425307850614, 'subsample': 0.5074725307850614, 'colsample\_bytree': 0.801}, [1] 2025-05-31 02:39:08,892 Trial 11 finished with value: 0.0022241421898692333 and parameters: {'n\_estimators': 233, 'max\_depth': 4, 'learning\_rate': 0.2951948045120345, 'subsample': 0.8268310305966374, 'colsample\_bytree': 0.68043}, [1] 2025-05-31 02:39:09,036 Trial 12 finished with value: 0.002047371304509458 and parameters: {'n\_estimators': 231, 'max\_depth': 4, 'learning\_rate': 0.2724964223936123, 'subsample': 0.793561104013812, 'colsample\_bytree': 0.513556}, [1] 2025-05-31 02:39:09,176 Trial 13 finished with value: 0.002286076376339709 and parameters: {'n\_estimators': 226, 'max\_depth': 3, 'learning\_rate': 0.19550916025629362, 'subsample': 0.845379348542006, 'colsample\_bytree': 0.6861}, [1] 2025-05-31 02:39:09,314 Trial 14 finished with value: 0.002194751060304927 and parameters: {'n\_estimators': 203, 'max\_depth': 7, 'learning\_rate': 0.0651616045814299, 'subsample': 0.849704646636452, 'colsample\_bytree': 0.7731}, [1] 2025-05-31 02:39:09,466 Trial 15 finished with value: 0.00214216107809258 and parameters: {'n\_estimators': 266, 'max\_depth': 5, 'learning\_rate': 0.17620381989420543, 'subsample': 0.722403446064393, 'colsample\_bytree': 0.60750}, [1] 2025-05-31 02:39:09,650 Trial 16 finished with value: 0.002169829165695798 and parameters: {'n\_estimators': 270, 'max\_depth': 4, 'learning\_rate': 0.1487640894089288, 'subsample': 0.686461874974897643, 'colsample\_bytree': 0.5879}, [1] 2025-05-31 02:39:09,767 Trial 17 finished with value: 0.0022646106170965256 and parameters: {'n\_estimators': 137, 'max\_depth': 6, 'learning\_rate': 0.0785581764235688, 'subsample': 0.601029417779793, 'colsample\_bytree': 0.8587}, [1] 2025-05-31 02:39:09,857 Trial 18 finished with value: 0.0021765944418310303 and parameters: {'n\_estimators': 60, 'max\_depth': 8, 'learning\_rate': 0.03964484649010596, 'subsample': 0.5967175379485794, 'colsample\_bytree': 0.7318}, [1] 2025-05-31 02:39:09,965 Trial 19 finished with value: 0.00222428345512773 and parameters: {'n\_estimators': 110, 'max\_depth': 3, 'learning\_rate': 0.023256827488202, 'subsample': 0.728778325988201, 'colsample\_bytree': 0.81408}, [1] 2025-05-31 02:39:10,037 A new study created in memory with name: no-name-d283755e-5f81-43le-a2de-e7d497a47d42, [1] 2025-05-31 02:39:10,165 Trial 0 finished with value: 0.001360974782397944 and parameters: {'n\_estimators': 161, 'max\_depth': 7, 'learning\_rate': 0.01067519290151848, 'subsample': 0.7272829966776484, 'colsample\_bytree': 0.586164}, [1] 2025-05-31 02:39:10,286 Trial 1 finished with value: 0.001377153640930391 and parameters: {'n\_estimators': 221, 'max\_depth': 3, 'learning\_rate': 0.0856100594080614, 'subsample': 0.598399574733332, 'colsample\_bytree': 0.744959}, [1] 2025-05-31 02:39:10,368 Trial 2 finished with value: 0.001346704178431403 and parameters: {'n\_estimators': 105, 'max\_depth': 8, 'learning\_rate': 0.01762037644632948, 'subsample': 0.7362652943277261, 'colsample\_bytree': 0.89326}, [1] 2025-05-31 02:39:10,440 Trial 3 finished with value: 0.001340730257051768 and parameters: {'n\_estimators': 89, 'max\_depth': 4, 'learning\_rate': 0.015738204067635259, 'subsample': 0.5152928424806391, 'colsample\_bytree': 0.814014}, [1] 2025-05-31 02:39:10,516 Trial 4 finished with value: 0.001371254315152843 and parameters: {'n\_estimators': 64, 'max\_depth': 7, 'learning\_rate': 0.27163908100945974, 'subsample': 0.9130863688659024, 'colsample\_bytree': 0.913467}, [1] 2025-05-31 02:39:10,645 Trial 5 finished with value: 0.00139912474205483 and parameters: {'n\_estimators': 160, 'max\_depth': 4, 'learning\_rate': 0.1471571799573269, 'subsample': 0.7890965508896536, 'colsample\_bytree': 0.9492362}, [1] 2025-05-31 02:39:10,774 Trial 6 finished with value: 0.00135246250621038346, 'subsample': 0.903412116352934, 'colsample\_bytree': 0.66889}, [1] 2025-05-31 02:39:10,890 Trial 7 finished with value: 0.0013542645805633666 and parameters: {'n\_estimators': 206, 'max\_depth': 7, 'learning\_rate': 0.14092526025038346, 'subsample': 0.8452680091938876, 'colsample\_bytree': 0.79129}, [1] 2025-05-31 02:39:10,993 Trial 8 finished with value: 0.001369718456210993 and parameters: {'n\_estimators': 157, 'max\_depth': 3, 'learning\_rate': 0.11265721748311752, 'subsample': 0.7333532391767, 'colsample\_bytree': 0.73111}, [1] 2025-05-31 02:39:11,177 Trial 9 finished with value: 0.001368093451763762 and parameters: {'n\_estimators': 244, 'max\_depth': 4, 'learning\_rate': 0.019320863536236326, 'subsample': 0.507782423595823, 'colsample\_bytree': 0.666306}, [1] 2025-05-31 02:39:11,289 Trial 10 finished with value: 0.001354372082754745 and parameters: {'n\_estimators': 295, 'max\_depth': 3, 'learning\_rate': 0.03545458246513539, 'subsample': 0.66545830394612789, 'colsample\_bytree': 0.5604}, [1] 2025-05-31 02:39:11,403 Trial 11 finished with value: 0.00130720428570422602 and parameters: {'n\_estimators': 134, 'max\_depth': 3, 'learning\_rate': 0.036302522760402343, 'subsample': 0.5052453719136262, 'colsample\_bytree': 0.67127}, [1] 2025-05-31 02:39:11,514 Trial 12 finished with value: 0.0013263698843624027 and parameters: {'n\_estimators': 217, 'max\_depth': 4, 'learning\_rate': 0.03277810598238465, 'subsample': 0.622427889432117, 'colsample\_bytree': 0.661}, [1] 2025-05-31 02:39:11,700 Trial 13 finished with value: 0.00136798341693502 and parameters: {'n\_estimators': 188, 'max\_depth': 5, 'learning\_rate': 0.04956743143645985, 'subsample': 0.6590302029215393, 'colsample\_bytree': 0.7435}, [1] 2025-05-31 02:39:11,816 Trial 14 finished with value: 0.001389416962152922 and parameters: {'n\_estimators': 127, 'max\_depth': 3, 'learning\_rate': 0.070913838312475315, 'subsample': 0.672024054811751, 'colsample\_bytree': 0.50131}, [1] 2025-05-31 02:39:11,993 Trial 15 finished with value: 0.001369327059271561 and parameters: {'n\_estimators': 290, 'max\_depth': 4, 'learning\_rate': 0.024023366476536432, 'subsample': 0.8220418107998836, 'colsample\_bytree': 0.6075}, [1] 2025-05-31 02:39:12,115 Trial 16 finished with value: 0.001352097495818004 and parameters: {'n\_estimators': 176, 'max\_depth': 5, 'learning\_rate': 0.2396420116837167, 'subsample': 0.9728997011830769, 'colsample\_bytree': 0.83299}, [1] 2025-05-31 02:39:12,257 Trial 17 finished with value: 0.0013665543131059392 and parameters: {'n\_estimators': 228, 'max\_depth': 3, 'learning\_rate': 0.04052176826126306, 'subsample': 0.5870676293005032, 'colsample\_bytree': 0.6917}, [1] 2025-05-31 02:39:12,372 Trial 18 finished with value: 0.001383792073018743 and parameters: {'n\_estimators': 142, 'max\_depth': 6, 'learning\_rate': 0.0613215276210762, 'subsample': 0.699752291270672, 'colsample\_bytree': 0.7293}, [1] 2025-05-31 02:39:12,608 Trial 19 finished with value: 0.0013681878348484486 and parameters: {'n\_estimators': 200, 'max\_depth': 4, 'learning\_rate': 0.10581646110280768, 'subsample': 0.7875096364811116, 'colsample\_bytree': 0.99897}, [1] 2025-05-31 02:39:12,808 Trial 20 finished with value: 0.003032678537228893 and parameters: {'n\_estimators': 258, 'max\_depth': 3, 'learning\_rate': 0.189933280101978886, 'subsample': 0.68203159812050982, 'colsample\_bytree': 0.84420}, [1] 2025-05-31 02:39:13,953 Trial 21 finished with value: 0.0030253043616361751 and parameters: {'n\_estimators': 235, 'max\_depth': 8, 'learning\_rate': 0.0264605468548521037, 'subsample': 0.500542167655163244, 'colsample\_bytree': 0.8843}, [1] 2025-05-31 02:39:13,964 Trial 22 finished with value: 0.002997003565694464 and parameters: {'n\_estimators': 208, 'max\_depth': 8, 'learning\_rate': 0.078111086156717, 'subsample': 0.557274831951233, 'colsample\_bytree': 0.559099}, [1] 2025-05-31 02:39:13,127 Trial 23 finished with value: 0.00301263511865804 and parameters: {'n\_estimators': 105, 'max\_depth': 8, 'learning\_rate': 0.01859956741586367, 'subsample': 0.7906075834496586, 'colsample\_bytree': 0.85828}, [1] 2025-05-31 02:39:13,224 Trial 24 finished with value: 0.003034104684993917 and parameters: {'n\_estimators': 160, 'max\_depth': 4, 'learning\_rate': 0.2851249572383636, 'subsample': 0.6005951616955355, 'colsample\_bytree': 0.77773}, [1] 2025-05-31 02:39:13,355 Trial 25 finished with value: 0.003000390409416965 and parameters: {'n\_estimators': 263, 'max\_depth': 6, 'learning\_rate': 0.0818131964863355, 'subsample': 0.9671269887887113, 'colsample\_bytree': 0.906602}, [1] 2025-05-31 02:39:13,490 Trial 26 finished with value: 0.0030293164580434034 and parameters: {'n\_estimators': 284, 'max\_depth': 6, 'learning\_rate': 0.104807836211232, 'subsample': 0.9373774745983105384, 'colsample\_bytree': 0.735548}, [1] 2025-05-31 02:39:13,583 Trial 27 finished with value: 0.00303459421736091 and parameters: {'n\_estimators': 164, 'max\_depth': 3, 'learning\_rate': 0.2703378605347375, 'subsample': 0.90216758365338898, 'colsample\_bytree': 0.6529157}, [1] 2025-05-31 02:39:13,785 Trial 28 finished with value: 0.0030255353364234 and parameters: {'n\_estimators': 138, 'max\_depth': 7, 'learning\_rate': 0.03227172572958876, 'subsample': 0.917631105075918, 'colsample\_bytree': 0.690098}, [1] 2025-05-31 02:39:13,870 Trial 29 finished with value: 0.0030244706351640404 and parameters: {'n\_estimators': 56, 'max\_depth': 5, 'learning\_rate': 0.010104859320485327, 'subsample': 0.5005439064491636, 'colsample\_bytree': 0.5055}, [1] 2025-05-31 02:39:14,007 Trial 30 finished with value: 0.00306951847519255 and parameters: {'n\_estimators': 218, 'max\_depth': 6, 'learning\_rate': 0.0790125976366784, 'subsample': 0.5072597744644956, 'colsample\_bytree': 0.58985}, [1] 2025-05-31 02:39:14,153 Trial 31 finished with value: 0.0030023831928687374 and parameters: {'n\_estimators': 208, 'max\_depth': 7, 'learning\_rate': 0.071265506844606, 'subsample': 0.698418710467093, 'colsample\_bytree': 0.98070}, [1] 2025-0



[1] 2025-05-31 02:39:40,464 Trial 7 finished with value: 0.00146569500450568 and parameters: {'n_estimators': 195, 'max_depth': 5, 'learning_rate': 0.018331590995811914, 'subsample': 0.6529196367654139, 'colsample_bytree': 0.53696}
[1] 2025-05-31 02:39:40,570 Trial 8 finished with value: 0.00144360650053233 and parameters: {'n_estimators': 189, 'max_depth': 4, 'learning_rate': 0.077721053637762, 'subsample': 0.611206139994881, 'colsample_bytree': 0.63750}
[1] 2025-05-31 02:39:40,630 Trial 9 finished with value: 0.001428103205345267 and parameters: {'n_estimators': 66, 'max_depth': 8, 'learning_rate': 0.1055272194918172, 'subsample': 0.94950224248114, 'colsample_bytree': 0.5960323}
[1] 2025-05-31 02:39:40,817 Trial 10 finished with value: 0.001469405974971561 and parameters: {'n_estimators': 264, 'max_depth': 6, 'learning_rate': 0.18537236351108268, 'subsample': 0.793717862170411, 'colsample_bytree': 0.97616}
[1] 2025-05-31 02:39:40,932 Trial 11 finished with value: 0.001428527151890894 and parameters: {'n_estimators': 108, 'max_depth': 7, 'learning_rate': 0.0505272194918172, 'subsample': 0.977671488283448, 'colsample_bytree': 0.99475}
[1] 2025-05-31 02:39:41,044 Trial 12 finished with value: 0.00142857655272993 and parameters: {'n_estimators': 120, 'max_depth': 7, 'learning_rate': 0.2920494418307876, 'subsample': 0.9832154541714617, 'colsample_bytree': 0.9598}
[1] 2025-05-31 02:39:41,157 Trial 13 finished with value: 0.001464925936416188 and parameters: {'n_estimators': 127, 'max_depth': 7, 'learning_rate': 0.04607149209529494, 'subsample': 0.905681426597544, 'colsample_bytree': 0.9001}
[1] 2025-05-31 02:39:41,303 Trial 14 finished with value: 0.001427366404088516 and parameters: {'n_estimators': 230, 'max_depth': 7, 'learning_rate': 0.083175946526960994, 'subsample': 0.8904886281198435, 'colsample_bytree': 0.9102}
[1] 2025-05-31 02:39:41,405 Trial 15 finished with value: 0.001471312521719523 and parameters: {'n_estimators': 102, 'max_depth': 6, 'learning_rate': 0.0353072804978326, 'subsample': 0.7636993412091625, 'colsample_bytree': 0.9344}
[1] 2025-05-31 02:39:41,525 Trial 16 finished with value: 0.001423104979696138 and parameters: {'n_estimators': 150, 'max_depth': 7, 'learning_rate': 0.09739765783400783, 'subsample': 0.12616403012249909, 'colsample_bytree': 0.8728}
[1] 2025-05-31 02:39:41,663 Trial 17 finished with value: 0.001420865294285423 and parameters: {'n_estimators': 213, 'max_depth': 5, 'learning_rate': 0.01746155700605356, 'subsample': 0.8295682440545172, 'colsample_bytree': 0.8716}
[1] 2025-05-31 02:39:42,006 Trial 18 finished with value: 0.001476139053265415 and parameters: {'n_estimators': 218, 'max_depth': 4, 'learning_rate': 0.12415684712794853, 'subsample': 0.834711959416975, 'colsample_bytree': 0.8729}
[1] 2025-05-31 02:39:42,135 A new study created in memory with name: no-name-6919c729-469-407-8749-2df69ae840
[1] 2025-05-31 02:39:42,306 Trial 0 finished with value: 0.003392297483043293 and parameters: {'n_estimators': 229, 'max_depth': 4, 'learning_rate': 0.01097982310505431, 'subsample': 0.6137611726614342, 'colsample_bytree': 0.9126}
[1] 2025-05-31 02:39:42,464 Trial 1 finished with value: 0.003430396352510587 and parameters: {'n_estimators': 178, 'max_depth': 5, 'learning_rate': 0.02772359659303454, 'subsample': 0.5158385647552264, 'colsample_bytree': 0.67592}
[1] 2025-05-31 02:39:43,156 Trial 2 finished with value: 0.00335181552491892 and parameters: {'n_estimators': 154, 'max_depth': 5, 'learning_rate': 0.0244963440872267, 'subsample': 0.70362484347335, 'colsample_bytree': 0.6827761}
[1] 2025-05-31 02:39:43,349 Trial 3 finished with value: 0.0034273093639843 and parameters: {'n_estimators': 228, 'max_depth': 7, 'learning_rate': 0.096755703290345, 'subsample': 0.58177043793638453, 'colsample_bytree': 0.605812}
[1] 2025-05-31 02:39:43,454 Trial 4 finished with value: 0.00343808947100921 and parameters: {'n_estimators': 63, 'max_depth': 5, 'learning_rate': 0.02596744847440664, 'subsample': 0.757577435015954, 'colsample_bytree': 0.7764971}
[1] 2025-05-31 02:39:44,099 Trial 5 finished with value: 0.0033810461372613277 and parameters: {'n_estimators': 188, 'max_depth': 8, 'learning_rate': 0.01375382318736464, 'subsample': 0.147551791651794, 'colsample_bytree': 0.6975}
[1] 2025-05-31 02:39:44,549 Trial 6 finished with value: 0.003326595431843845 and parameters: {'n_estimators': 240, 'max_depth': 4, 'learning_rate': 0.18052804040713227, 'subsample': 0.7095139348303703, 'colsample_bytree': 0.83459}
[1] 2025-05-31 02:39:44,865 Trial 7 finished with value: 0.003379455382616222 and parameters: {'n_estimators': 241, 'max_depth': 8, 'learning_rate': 0.01450329758737442, 'subsample': 0.789725549007815, 'colsample_bytree': 0.54986}
[1] 2025-05-31 02:39:45,836 Trial 8 finished with value: 0.00342047913695278 and parameters: {'n_estimators': 282, 'max_depth': 3, 'learning_rate': 0.020117347462433, 'subsample': 0.5202320528875644, 'colsample_bytree': 0.5814569}
[1] 2025-05-31 02:39:46,077 Trial 9 finished with value: 0.00333815869201904 and parameters: {'n_estimators': 132, 'max_depth': 8, 'learning_rate': 0.1226020873403123, 'subsample': 0.950728425604885, 'colsample_bytree': 0.51909}
[1] 2025-05-31 02:39:47,120 Trial 10 finished with value: 0.003305707446671006 and parameters: {'n_estimators': 297, 'max_depth': 3, 'learning_rate': 0.2794113431050056, 'subsample': 0.8936843975611578, 'colsample_bytree': 0.9642}
[1] 2025-05-31 02:39:47,366 Trial 11 finished with value: 0.003397836975902943 and parameters: {'n_estimators': 284, 'max_depth': 3, 'learning_rate': 0.29873894892807284, 'subsample': 0.9055716867041147, 'colsample_bytree': 0.986974}
[1] 2025-05-31 02:39:47,597 Trial 12 finished with value: 0.003275934078998449 and parameters: {'n_estimators': 296, 'max_depth': 4, 'learning_rate': 0.2883780076172079, 'subsample': 0.589382360168984, 'colsample_bytree': 0.8526245}
[1] 2025-05-31 02:39:47,825 Trial 13 finished with value: 0.0032785047893517205 and parameters: {'n_estimators': 296, 'max_depth': 4, 'learning_rate': 0.298407330757039, 'subsample': 0.8808477583176159, 'colsample_bytree': 0.91394}
[1] 2025-05-31 02:39:48,019 Trial 14 finished with value: 0.003413924059150734 and parameters: {'n_estimators': 265, 'max_depth': 6, 'learning_rate': 0.060615644243819, 'subsample': 0.8425978289527375, 'colsample_bytree': 0.8637}
[1] 2025-05-31 02:39:48,122 Trial 15 finished with value: 0.0035084793523273 and parameters: {'n_estimators': 103, 'max_depth': 4, 'learning_rate': 0.1805593138393464, 'subsample': 0.970420001153405, 'colsample_bytree': 0.81410}
[1] 2025-05-31 02:39:48,258 Trial 16 finished with value: 0.00337838304308107 and parameters: {'n_estimators': 200, 'max_depth': 6, 'learning_rate': 0.05478484077178246, 'subsample': 0.482289941312004, 'colsample_bytree': 0.9041}
[1] 2025-05-31 02:39:48,413 Trial 17 finished with value: 0.003379472163722734 and parameters: {'n_estimators': 259, 'max_depth': 5, 'learning_rate': 0.181314436356611777, 'subsample': 0.900484567955888, 'colsample_bytree': 0.9081}
[1] 2025-05-31 02:39:48,552 Trial 18 finished with value: 0.003399823020529597 and parameters: {'n_estimators': 207, 'max_depth': 4, 'learning_rate': 0.10365775148153974, 'subsample': 0.8370579317857429, 'colsample_bytree': 0.7683}
[1] 2025-05-31 02:39:48,882 A new study created in memory with name: no-name-d525434d-e998-4242-ale3-12dbab5041f
[1] 2025-05-31 02:39:49,007 Trial 0 finished with value: 0.00533267014575544 and parameters: {'n_estimators': 145, 'max_depth': 3, 'learning_rate': 0.0443814140495833, 'subsample': 0.8375352614502175, 'colsample_bytree': 0.72314}
[1] 2025-05-31 02:39:49,109 Trial 1 finished with value: 0.005326720264944813 and parameters: {'n_estimators': 167, 'max_depth': 5, 'learning_rate': 0.02010570177785477207, 'subsample': 0.87937615828757, 'colsample_bytree': 0.54701}
[1] 2025-05-31 02:39:49,177 Trial 2 finished with value: 0.005310924363563701 and parameters: {'n_estimators': 57, 'max_depth': 5, 'learning_rate': 0.0828543170160484, 'subsample': 0.592406819581799, 'colsample_bytree': 0.8755747}
[1] 2025-05-31 02:39:49,295 Trial 3 finished with value: 0.00524036335274695 and parameters: {'n_estimators': 216, 'max_depth': 6, 'learning_rate': 0.0124548784784795993, 'subsample': 0.5172632914927819, 'colsample_bytree': 0.6791}
[1] 2025-05-31 02:39:49,402 Trial 4 finished with value: 0.005300210385409289 and parameters: {'n_estimators': 187, 'max_depth': 7, 'learning_rate': 0.03077353778728987, 'subsample': 0.5765202109580959, 'colsample_bytree': 0.649204}
[1] 2025-05-31 02:39:49,526 Trial 5 finished with value: 0.00529434285745142 and parameters: {'n_estimators': 229, 'max_depth': 4, 'learning_rate': 0.25005734807308716, 'subsample': 0.7521116412322763, 'colsample_bytree': 0.477690}
[1] 2025-05-31 02:39:49,593 Trial 6 finished with value: 0.00531094114011241 and parameters: {'n_estimators': 66, 'max_depth': 6, 'learning_rate': 0.18125717678869456, 'subsample': 0.882126457469421, 'colsample_bytree': 0.6713344}
[1] 2025-05-31 02:39:49,706 Trial 7 finished with value: 0.00516305137881107 and parameters: {'n_estimators': 203, 'max_depth': 4, 'learning_rate': 0.1675355841031698, 'subsample': 0.635694246212855, 'colsample_bytree': 0.572120}
[1] 2025-05-31 02:39:49,789 Trial 8 finished with value: 0.005356263017262428 and parameters: {'n_estimators': 111, 'max_depth': 8, 'learning_rate': 0.1082793791072624, 'subsample': 0.760770208449557, 'colsample_bytree': 0.616994}
[1] 2025-05-31 02:39:49,921 Trial 9 finished with value: 0.00532756310901682 and parameters: {'n_estimators': 263, 'max_depth': 6, 'learning_rate': 0.2688348029466566, 'subsample': 0.9633283179190537, 'colsample_bytree': 0.6867096}
[1] 2025-05-31 02:39:50,125 Trial 10 finished with value: 0.0052579664666234179 and parameters: {'n_estimators': 293, 'max_depth': 3, 'learning_rate': 0.0287253253593982, 'subsample': 0.517483219545399, 'colsample_bytree': 0.848836}
[1] 2025-05-31 02:39:50,290 Trial 11 finished with value: 0.00532789791753868 and parameters: {'n_estimators': 295, 'max_depth': 3, 'learning_rate': 0.26293557742269114, 'subsample': 0.50253380809782872, 'colsample_bytree': 0.83916}
[1] 2025-05-31 02:39:50,443 Trial 12 finished with value: 0.00534281637232593 and parameters: {'n_estimators': 242, 'max_depth': 4, 'learning_rate': 0.19506695323076492, 'subsample': 0.5043726403624651, 'colsample_bytree': 0.94944}
[1] 2025-05-31 02:39:50,609 Trial 13 finished with value: 0.0050306780861586925 and parameters: {'n_estimators': 295, 'max_depth': 4, 'learning_rate': 0.064600601562253648, 'subsample': 0.761570120920927921, 'colsample_bytree': 0.81079}
[1] 2025-05-31 02:39:50,760 Trial 14 finished with value: 0.0052597967428720836 and parameters: {'n_estimators': 250, 'max_depth': 3, 'learning_rate': 0.150280548334075818, 'subsample': 0.603540616215958, 'colsample_bytree': 0.909783}
[1] 2025-05-31 02:39:50,913 Trial 15 finished with value: 0.005285843943640375 and parameters: {'n_estimators': 262, 'max_depth': 3, 'learning_rate': 0.14328834636848174, 'subsample': 0.5821711575894777, 'colsample_bytree': 0.9413}
[1] 2025-05-31 02:39:51,080 Trial 16 finished with value: 0.0052768813725356476 and parameters: {'n_estimators': 268, 'max_depth': 3, 'learning_rate': 0.0958292394862034, 'subsample': 0.591761922205486, 'colsample_bytree': 0.90442}
[1] 2025-05-31 02:39:51,243 Trial 17 finished with value: 0.0052785695627296407 and parameters: {'n_estimators': 293, 'max_depth': 5, 'learning_rate': 0.05157839849145452, 'subsample': 0.5551847130643568, 'colsample_bytree': 0.79574}
[1] 2025-05-31 02:39:51,452 Trial 18 finished with value: 0.0053295413684239 and parameters: {'n_estimators': 250, 'max_depth': 3, 'learning_rate': 0.053219533095511, 'subsample': 0.667610759522006, 'colsample_bytree': 0.9778946}
[1] 2025-05-31 02:39:51,569 Trial 19 finished with value: 0.00531956812080996 and parameters: {'n_estimators': 147, 'max_depth': 4, 'learning_rate': 0.291080182302521, 'subsample': 0.6497179247423158, 'colsample_bytree': 0.883921}
[1] 2025-05-31 02:39:51,723 A new study created in memory with name: no-name-c542c97-bf00-4507-b608-13fe2083103
[1] 2025-05-31 02:39:51,843 Trial 0 finished with value: 0.00216230779945676 and parameters: {'n_estimators': 147, 'max_depth': 5, 'learning_rate': 0.148011788082018, 'subsample': 0.8848614570377846, 'colsample_bytree': 0.564944}
[1] 2025-05-31 02:39:51,928 Trial 1 finished with value: 0.0020192903749315943 and parameters: {'n_estimators': 122, 'max_depth': 5, 'learning_rate': 0.015346565479134, 'subsample': 0.915317301571204, 'colsample_bytree': 0.61974}
[1] 2025-05-31 02:39:52,055 Trial 2 finished with value: 0.002023153047089625 and parameters: {'n_estimators': 229, 'max_depth': 7, 'learning_rate': 0.06772826407398431, 'subsample': 0.50437285290748765, 'colsample_bytree': 0.923768}
[1] 2025-05-31 02:39:52,140 Trial 3 finished with value: 0.0020534861670372593 and parameters: {'n_estimators': 63, 'max_depth': 8, 'learning_rate': 0.1586242857401536, 'subsample': 0.7402116083708323, 'colsample_bytree': 0.5646472}
[1] 2025-05-31 02:39:52,233 Trial 4 finished with value: 0.002118407648452531 and parameters: {'n_estimators': 154, 'max_depth': 7, 'learning_rate': 0.02613724857401536, 'subsample': 0.7402116083708323, 'colsample_bytree': 0.5646472}
[1] 2025-05-31 02:39:52,312 Trial 5 finished with value: 0.00202315047089625 and parameters: {'n_estimators': 141, 'max_depth': 4, 'learning_rate': 0.19506695323076492, 'subsample': 0.5043726403624651, 'colsample_bytree': 0.94944}
[1] 2025-05-31 02:39:52,438 Trial 6 finished with value: 0.00204048874518074807 and parameters: {'n_estimators': 216, 'max_depth': 5, 'learning_rate': 0.01470839125362002, 'subsample': 0.7963827383609432, 'colsample_bytree': 0.610797}
[1] 2025-05-31 02:39:52,567 Trial 7 finished with value: 0.002040207850729357664 and parameters: {'n_estimators': 163, 'max_depth': 8, 'learning_rate': 0.02782195057088047, 'subsample': 0.8529072378496022, 'colsample_bytree': 0.90944}
[1] 2025-05-31 02:39:52,688 Trial 8 finished with value: 0.0020382830560623764 and parameters: {'n_estimators': 181, 'max_depth': 6, 'learning_rate': 0.045195130717645, 'subsample': 0.9106707324736247, 'colsample_bytree': 0.747117}
[1] 2025-05-31 02:39:52,983 Trial 9 finished with value: 0.0020196174224222242 and parameters: {'n_estimators': 242, 'max_depth': 6, 'learning_rate': 0.0759380170723524, 'subsample': 0.6024817746300516, 'colsample_bytree': 0.87480}
[1] 2025-05-31 02:39:53,050 Trial 10 finished with value: 0.0020170988141783304 and parameters: {'n_estimators': 291, 'max_depth': 3, 'learning_rate': 0.0202659746668229472, 'subsample': 0.5062059746668229472, 'colsample_bytree': 0.50608}
[1] 2025-05-31 02:39:53,268 Trial 11 finished with value: 0.0020296705305307423 and parameters: {'n_estimators': 231, 'max_depth': 5, 'learning_rate': 0.03703956305307423, 'subsample': 0.6222306822292472, 'colsample_bytree': 0.84232}
[1] 2025-05-31 02:39:53,368 Trial 12 finished with value: 0.002089513460745974597 and parameters: {'n_estimators': 233, 'max_depth': 6, 'learning_rate': 0.0224739253758493, 'subsample': 0.658690394170428, 'colsample_bytree': 0.8506}
[1] 2025-05-31 02:39:53,369 Trial 13 finished with value: 0.001969053542214376 and parameters: {'n_estimators': 97, 'max_depth': 4, 'learning_rate': 0.0994561609361337, 'subsample': 0.5117364145075389, 'colsample_bytree': 0.986318}
[1] 2025-05-31 02:39:53,467 Trial 14 finished with value: 0.002075284780720165 and parameters: {'n_estimators': 90, 'max_depth': 3, 'learning_rate': 0.12002194194034779, 'subsample': 0.81204288349521, 'colsample_bytree': 0.99439}
[1] 2025-05-31 02:39:53,565 Trial 15 finished with value: 0.00205799730529357664 and parameters: {'n_estimators': 103, 'max_depth': 4, 'learning_rate': 0.03468501365321724, 'subsample': 0.507979479308135, 'colsample_bytree': 0.69479}
[1] 2025-05-31 02:39:53,655 Trial 16 finished with value: 0.0020136635465681205065 and parameters: {'n_estimators': 50, 'max_depth': 4, 'learning_rate': 0.125865626761691292, 'subsample': 0.8702268627229382, 'colsample_bytree': 0.80213}
[1] 2025-05-31 02:39:53,784 Trial 17 finished with value: 0.002041445922930906 and parameters: {'n_estimators': 187, 'max_depth': 5, 'learning_rate': 0.01988361409278944, 'subsample': 0.97996909638063, 'colsample_bytree': 0.9915163}
[1] 2025-05-31 02:39:54,000 Trial 18 finished with value: 0.0020747913177664 and parameters: {'n_estimators': 86, 'max_depth': 4, 'learning_rate': 0.123027913177664, 'subsample': 0.72637660419472, 'colsample_bytree': 0.57151}
[1] 2025-05-31 02:39:54,092 A new study created in memory with name: no-name-8e31f199-9cf-c49b-49ab-bb86-5ea05d89df9a
[1] 2025-05-31 02:39:54,211 Trial 0 finished with value: 0.001929115231728731 and parameters: {'n_estimators': 226, 'max_depth': 3, 'learning_rate': 0.01483605603335952, 'subsample': 0.9795824419728156, 'colsample_bytree': 0.8974}
[1] 2025-05-31 02:39:54,358 Trial 1 finished with value: 0.0012808451247392594 and parameters: {'n_estimators': 69, 'max_depth': 3, 'learning_rate': 0.217113214161527, 'subsample': 0.6052551787510156,

```
[1] 2025-05-31 02:40:03, 150] Trial 15 finished with value: 0.002732394202717853 and parameters: {'n_estimators': 187, 'max_depth': 7, 'learning_rate': 0.07065392473076307, 'colsample_bytree': 0.810713}
[1] 2025-05-31 02:40:03, 294] Trial 16 finished with value: 0.0027518703689835995 and parameters: {'n_estimators': 246, 'max_depth': 4, 'learning_rate': 0.14508115924283027, 'colsample_bytree': 0.6810}
[1] 2025-05-31 02:40:03, 428] Trial 17 finished with value: 0.00272728198189116 and parameters: {'n_estimators': 211, 'max_depth': 8, 'learning_rate': 0.01529794553473148, 'colsample_bytree': 0.571}
[1] 2025-05-31 02:40:03, 581] Trial 18 finished with value: 0.00275454794245885 and parameters: {'n_estimators': 143, 'max_depth': 6, 'learning_rate': 0.05686584832258464, 'colsample_bytree': 0.7834}
[1] 2025-05-31 02:40:03, 739] Trial 19 finished with value: 0.0026949281335158226 and parameters: {'n_estimators': 277, 'max_depth': 5, 'learning_rate': 0.10456023856661495, 'colsample_bytree': 0.54560889348747, 'colsample_bytree': 0.51601}
[1] 2025-05-31 02:40:03, 877] A new study created in memory with name: no-name-5d743741-44ca-4b63-8c5d-1875b695983
[1] 2025-05-31 02:40:04, 051] Trial 0 finished with value: 0.003096046557330458 and parameters: {'n_estimators': 287, 'max_depth': 4, 'learning_rate': 0.0223003679273208, 'colsample_bytree': 0.7080}
[1] 2025-05-31 02:40:04, 131] Trial 1 finished with value: 0.003084209174114776 and parameters: {'n_estimators': 100, 'max_depth': 6, 'learning_rate': 0.0412981113437911, 'colsample_bytree': 0.794049}
[1] 2025-05-31 02:40:04, 232] Trial 2 finished with value: 0.003083885937942876 and parameters: {'n_estimators': 166, 'max_depth': 3, 'learning_rate': 0.11683228438106456, 'colsample_bytree': 0.75325}
[1] 2025-05-31 02:40:04, 346] Trial 3 finished with value: 0.0039955633990662 and parameters: {'n_estimators': 213, 'max_depth': 5, 'learning_rate': 0.015603803637270733, 'colsample_bytree': 0.7845}
[1] 2025-05-31 02:40:04, 457] Trial 4 finished with value: 0.00307028709834357 and parameters: {'n_estimators': 206, 'max_depth': 3, 'learning_rate': 0.02668479239875594, 'colsample_bytree': 0.6132}
[1] 2025-05-31 02:40:04, 615] Trial 5 finished with value: 0.003099819518491303 and parameters: {'n_estimators': 257, 'max_depth': 4, 'learning_rate': 0.0165975451545085297, 'colsample_bytree': 0.8971}
[1] 2025-05-31 02:40:04, 752] Trial 6 finished with value: 0.003058756187715354 and parameters: {'n_estimators': 279, 'max_depth': 8, 'learning_rate': 0.11611663833993827, 'colsample_bytree': 0.87587}
[1] 2025-05-31 02:40:04, 872] Trial 7 finished with value: 0.00302950959452217 and parameters: {'n_estimators': 214, 'max_depth': 5, 'learning_rate': 0.018963996196887029, 'colsample_bytree': 0.935176}
[1] 2025-05-31 02:40:04, 937] Trial 8 finished with value: 0.0031176332027267695 and parameters: {'n_estimators': 51, 'max_depth': 8, 'learning_rate': 0.05387384501076964, 'colsample_bytree': 0.910107}
[1] 2025-05-31 02:40:05, 076] Trial 9 finished with value: 0.00291578501891621 and parameters: {'n_estimators': 293, 'max_depth': 8, 'learning_rate': 0.0968440634828715, 'colsample_bytree': 0.556227}
[1] 2025-05-31 02:40:05, 191] Trial 10 finished with value: 0.00343733539579946 and parameters: {'n_estimators': 134, 'max_depth': 7, 'learning_rate': 0.2807328522287746, 'colsample_bytree': 0.5057}
[1] 2025-05-31 02:40:05, 350] Trial 11 finished with value: 0.00301491494077810697 and parameters: {'n_estimators': 300, 'max_depth': 8, 'learning_rate': 0.1332289320234058, 'colsample_bytree': 0.5339}
[1] 2025-05-31 02:40:05, 509] Trial 12 finished with value: 0.0030139065190587117 and parameters: {'n_estimators': 300, 'max_depth': 7, 'learning_rate': 0.042981113437911, 'colsample_bytree': 0.5057}
[1] 2025-05-31 02:40:05, 688] Trial 13 finished with value: 0.003020297212919073 and parameters: {'n_estimators': 248, 'max_depth': 7, 'learning_rate': 0.2112506938868267, 'colsample_bytree': 0.62408}
[1] 2025-05-31 02:40:05, 837] Trial 14 finished with value: 0.003072357365788493 and parameters: {'n_estimators': 247, 'max_depth': 7, 'learning_rate': 0.0720622319937223, 'colsample_bytree': 0.61491}
[1] 2025-05-31 02:40:05, 997] Trial 15 finished with value: 0.003044919066566395 and parameters: {'n_estimators': 299, 'max_depth': 6, 'learning_rate': 0.0587696096186584, 'colsample_bytree': 0.5984}
[1] 2025-05-31 02:40:06, 127] Trial 16 finished with value: 0.003089964406576183157 and parameters: {'n_estimators': 179, 'max_depth': 7, 'learning_rate': 0.0104439348293014, 'colsample_bytree': 0.998}
[1] 2025-05-31 02:40:06, 279] Trial 17 finished with value: 0.003139047219882316 and parameters: {'n_estimators': 263, 'max_depth': 8, 'learning_rate': 0.1793023585874176, 'colsample_bytree': 0.6804}
[1] 2025-05-31 02:40:06, 416] Trial 18 finished with value: 0.003016997698125505 and parameters: {'n_estimators': 230, 'max_depth': 6, 'learning_rate': 0.04375397894738521, 'colsample_bytree': 0.54779}
[1] 2025-05-31 02:40:06, 544] Trial 19 finished with value: 0.0030201920181728958 and parameters: {'n_estimators': 178, 'max_depth': 7, 'learning_rate': 0.08464348617560105, 'colsample_bytree': 0.59716}
[1] 2025-05-31 02:40:06, 724] A new study created in memory with name: no-name-f5fc437f-f6be-4b2b-81e3-725f0458eda
[1] 2025-05-31 02:40:06, 884] Trial 0 finished with value: 0.006453111236007948 and parameters: {'n_estimators': 270, 'max_depth': 8, 'learning_rate': 0.0356704736803968, 'colsample_bytree': 0.6985077}
[1] 2025-05-31 02:40:06, 970] Trial 1 finished with value: 0.00644958716087579 and parameters: {'n_estimators': 114, 'max_depth': 4, 'learning_rate': 0.027144238874490944, 'colsample_bytree': 0.688828}
[1] 2025-05-31 02:40:07, 097] Trial 2 finished with value: 0.00645389122002231 and parameters: {'n_estimators': 249, 'max_depth': 5, 'learning_rate': 0.019161041704174311, 'colsample_bytree': 0.72248}
[1] 2025-05-31 02:40:07, 174] Trial 3 finished with value: 0.006464230946599151 and parameters: {'n_estimators': 63, 'max_depth': 8, 'learning_rate': 0.16545386463715045, 'colsample_bytree': 0.6936301}
[1] 2025-05-31 02:40:07, 289] Trial 4 finished with value: 0.006456695253923 and parameters: {'n_estimators': 221, 'max_depth': 4, 'learning_rate': 0.16269745829595104, 'colsample_bytree': 0.6332799320234124942}
[1] 2025-05-31 02:40:07, 409] Trial 5 finished with value: 0.006451142437705789 and parameters: {'n_estimators': 245, 'max_depth': 4, 'learning_rate': 0.075359903881088, 'colsample_bytree': 0.590974}
[1] 2025-05-31 02:40:07, 497] Trial 6 finished with value: 0.00646002205355159 and parameters: {'n_estimators': 147, 'max_depth': 8, 'learning_rate': 0.0710757020721352, 'colsample_bytree': 0.636392895362006}
[1] 2025-05-31 02:40:07, 577] Trial 7 finished with value: 0.006451560146498798 and parameters: {'n_estimators': 111, 'max_depth': 8, 'learning_rate': 0.014894896116208374, 'colsample_bytree': 0.79244}
[1] 2025-05-31 02:40:07, 681] Trial 8 finished with value: 0.00646615826156423 and parameters: {'n_estimators': 187, 'max_depth': 7, 'learning_rate': 0.254806049378984, 'colsample_bytree': 0.95386943}
[1] 2025-05-31 02:40:07, 790] Trial 9 finished with value: 0.00643654020977226 and parameters: {'n_estimators': 135, 'max_depth': 5, 'learning_rate': 0.06103163664969934, 'colsample_bytree': 0.55758}
[1] 2025-05-31 02:40:07, 955] Trial 10 finished with value: 0.0064487493806716954 and parameters: {'n_estimators': 300, 'max_depth': 6, 'learning_rate': 0.01033518953402097, 'colsample_bytree': 0.51585}
[1] 2025-05-31 02:40:08, 051] Trial 11 finished with value: 0.00646224326880575 and parameters: {'n_estimators': 185, 'max_depth': 6, 'learning_rate': 0.2782487448885823, 'colsample_bytree': 0.92022}
[1] 2025-05-31 02:40:08, 213] Trial 12 finished with value: 0.006450704692710495 and parameters: {'n_estimators': 181, 'max_depth': 7, 'learning_rate': 0.1155015889194656, 'colsample_bytree': 0.91913}
[1] 2025-05-31 02:40:08, 325] Trial 13 finished with value: 0.00645044916134057 and parameters: {'n_estimators': 141, 'max_depth': 3, 'learning_rate': 0.0395776842749666, 'colsample_bytree': 0.83075}
[1] 2025-05-31 02:40:08, 409] Trial 14 finished with value: 0.0064589292010288 and parameters: {'n_estimators': 58, 'max_depth': 5, 'learning_rate': 0.2603442924163497, 'colsample_bytree': 0.8670446}
[1] 2025-05-31 02:40:08, 543] Trial 15 finished with value: 0.006439183828174993 and parameters: {'n_estimators': 198, 'max_depth': 7, 'learning_rate': 0.10770133422874783, 'colsample_bytree': 0.98507}
[1] 2025-05-31 02:40:08, 716] Trial 16 finished with value: 0.006451953485454991 and parameters: {'n_estimators': 211, 'max_depth': 7, 'learning_rate': 0.05855476914962164, 'colsample_bytree': 0.53062}
[1] 2025-05-31 02:40:08, 912] Trial 17 finished with value: 0.00645250127730306 and parameters: {'n_estimators': 148, 'max_depth': 6, 'learning_rate': 0.1005005169121872, 'colsample_bytree': 0.78021}
[1] 2025-05-31 02:40:09, 562] Trial 18 finished with value: 0.00647017081205779 and parameters: {'n_estimators': 88, 'max_depth': 5, 'learning_rate': 0.04580096723172045, 'colsample_bytree': 0.6138871}
[1] 2025-05-31 02:40:09, 788] Trial 19 finished with value: 0.0064507231827201235 and parameters: {'n_estimators': 209, 'max_depth': 3, 'learning_rate': 0.11529011696610514, 'colsample_bytree': 0.8833}
[1] 2025-05-31 02:40:10, 739] A new study created in memory with name: no-name-c3eccc0b-20b5-4993-9e2b-5087a5fecf
[1] 2025-05-31 02:40:11, 155] Trial 0 finished with value: 0.001730785090416463 and parameters: {'n_estimators': 163, 'max_depth': 6, 'learning_rate': 0.05348818053048006, 'colsample_bytree': 0.767478}
[1] 2025-05-31 02:40:11, 280] Trial 1 finished with value: 0.0017807387565887167 and parameters: {'n_estimators': 143, 'max_depth': 4, 'learning_rate': 0.19608026594980257, 'colsample_bytree': 0.94231}
[1] 2025-05-31 02:40:11, 580] Trial 2 finished with value: 0.0017037226441261643 and parameters: {'n_estimators': 269, 'max_depth': 6, 'learning_rate': 0.231010399882809, 'colsample_bytree': 0.84238}
[1] 2025-05-31 02:40:11, 711] Trial 3 finished with value: 0.00176030741939047 and parameters: {'n_estimators': 276, 'max_depth': 6, 'learning_rate': 0.0739348853998013, 'colsample_bytree': 0.70767}
[1] 2025-05-31 02:40:11, 869] Trial 4 finished with value: 0.001769569085507228 and parameters: {'n_estimators': 278, 'max_depth': 3, 'learning_rate': 0.15659429756328436, 'colsample_bytree': 0.69152}
[1] 2025-05-31 02:40:12, 008] Trial 5 finished with value: 0.0017505198368310898 and parameters: {'n_estimators': 265, 'max_depth': 4, 'learning_rate': 0.02059640520531604, 'colsample_bytree': 0.6797}
[1] 2025-05-31 02:40:12, 127] Trial 6 finished with value: 0.00174976765959524 and parameters: {'n_estimators': 212, 'max_depth': 7, 'learning_rate': 0.032545729379472, 'colsample_bytree': 0.9166453}
[1] 2025-05-31 02:40:12, 267] Trial 7 finished with value: 0.001749584542356966 and parameters: {'n_estimators': 257, 'max_depth': 5, 'learning_rate': 0.137052695584357457, 'colsample_bytree': 0.77862}
[1] 2025-05-31 02:40:12, 356] Trial 8 finished with value: 0.001759303374186168 and parameters: {'n_estimators': 134, 'max_depth': 7, 'learning_rate': 0.0198635232463282631, 'colsample_bytree': 0.68975}
[1] 2025-05-31 02:40:12, 471] Trial 9 finished with value: 0.001748060205096633 and parameters: {'n_estimators': 209, 'max_depth': 4, 'learning_rate': 0.01689431234365526, 'colsample_bytree': 0.85865}
[1] 2025-05-31 02:40:12, 564] Trial 10 finished with value: 0.001749001872036995 and parameters: {'n_estimators': 80, 'max_depth': 8, 'learning_rate': 0.289301306983178, 'colsample_bytree': 0.520883}
[1] 2025-05-31 02:40:12, 697] Trial 11 finished with value: 0.001749463063409956 and parameters: {'n_estimators': 175, 'max_depth': 6, 'learning_rate': 0.0107510508415574005, 'colsample_bytree': 0.838}
[1] 2025-05-31 02:40:12, 799] Trial 12 finished with value: 0.0017596280683387639 and parameters: {'n_estimators': 85, 'max_depth': 7, 'learning_rate': 0.09769682048506986, 'colsample_bytree': 0.898522077077531}
[1] 2025-05-31 02:40:12, 964] Trial 13 finished with value: 0.001751833611141028 and parameters: {'n_estimators': 216, 'max_depth': 5, 'learning_rate': 0.0412675655682553, 'colsample_bytree': 0.5909}
[1] 2025-05-31 02:40:13, 079] Trial 14 finished with value: 0.0017447650951861207 and parameters: {'n_estimators': 130, 'max_depth': 6, 'learning_rate': 0.0948527264969954, 'colsample_bytree': 0.775587}
[1] 2025-05-31 02:40:13, 111] Trial 15 finished with value: 0.0017504883301983513 and parameters: {'n_estimators': 171, 'max_depth': 8, 'learning_rate': 0.282450374472724, 'colsample_bytree': 0.91121}
[1] 2025-05-31 02:40:13, 355] Trial 16 finished with value: 0.001750498831749745 and parameters: {'n_estimators': 237, 'max_depth': 8, 'learning_rate': 0.296639762117734, 'colsample_bytree': 0.9999}
[1] 2025-05-31 02:40:13, 471] Trial 17 finished with value: 0.001748060205096633 and parameters: {'n_estimators': 209, 'max_depth': 4, 'learning_rate': 0.084442972818167, 'colsample_bytree': 0.85865}
[1] 2025-05-31 02:40:13, 564] Trial 18 finished with value: 0.001749001872036995 and parameters: {'n_estimators': 80, 'max_depth': 8, 'learning_rate': 0.289301306983178, 'colsample_bytree': 0.520883}
[1] 2025-05-31 02:40:12, 697] Trial 19 finished with value: 0.001749463063409956 and parameters: {'n_estimators': 175, 'max_depth': 6, 'learning_rate': 0.0107510508415574005, 'colsample_bytree': 0.838}
[1] 2025-05-31 02:40:14, 053] A new study created in memory with name: no-name-2bea341f-7693-4698-afdd-44ec6568ef7rc
[1] 2025-05-31 02:40:14, 178] Trial 0 finished with value: 0.00389952747945663 and parameters: {'n_estimators': 176, 'max_depth': 8, 'learning_rate': 0.011409245315247759, 'colsample_bytree': 0.59146}
[1] 2025-05-31 02:40:14, 294] Trial 1 finished with value: 0.00389953956584598745 and parameters: {'n_estimators': 223, 'max_depth': 6, 'learning_rate': 0.0458897428314913, 'colsample_bytree': 0.58456}
[1] 2025-05-31 02:40:14, 403] Trial 2 finished with value: 0.003899764477499159 and parameters: {'n_estimators': 158, 'max_depth': 3, 'learning_rate': 0.01008573052513234, 'colsample_bytree': 0.92685}
[1] 2025-05-31 02:40:14, 468] Trial 3 finished with value: 0.00390021989830513 and parameters: {'n_estimators': 59, 'max_depth': 6, 'learning_rate': 0.032304108242048675, 'colsample_bytree': 0.792318}
[1] 2025-05-31 02:40:14, 567] Trial 4 finished with value: 0.00390019110819236 and parameters: {'n_estimators': 156, 'max_depth': 7, 'learning_rate': 0.0704220438842323, 'colsample_bytree': 0.72527}
[1] 2025-05-31 02:40:14, 682] Trial 5 finished with value: 0.003899584208982063 and parameters: {'n_estimators': 226, 'max_depth': 5, 'learning_rate': 0.1065955284844237, 'colsample_bytree': 0.59295}
[1] 2025-05-31 02:40:14, 780] Trial 6 finished with value: 0.00389951948271907 and parameters: {'n_estimators': 163, 'max_depth': 8, 'learning_rate': 0.119766869108965793, 'colsample_bytree': 0.61882}
[1] 2025-05-31 02:40:14, 875] Trial 7 finished with value: 0.003900076740270654 and parameters: {'n_estimators': 152, 'max_depth': 3, 'learning_rate': 0.0447143532841035, 'colsample_bytree': 0.903124}
[1] 2025-05-31 02:40:15, 005] Trial 8 finished with value: 0.00389955374855415 and parameters: {'n_estimators': 260, 'max_depth': 5, 'learning_rate': 0.027863091094867268, 'colsample_bytree': 0.544912}
[1] 2025-05-31 02:40:15, 098] Trial 9 finished with value: 0.0038995386175445086 and parameters: {'n_estimators': 138, 'max_depth': 6, 'learning_rate': 0.04106644232905925, 'colsample_bytree': 0.72172}
[1] 2025-05-31 02:40:15, 261] Trial 10 finished with value: 0.0038986207193059781446, 'colsample_bytree': 0.5204841}
[1] 2025-05-31 02:40:15, 426] Trial 11 finished with value: 0.00389851091872447544 and parameters: {'n_estimators': 267, 'max_depth': 5, 'learning_rate': 0.01659199150198626, 'colsample_bytree': 0.5050}
[1] 2025-05-31 02:40:15, 512] Trial 12 finished with value: 0.00389892729751388 and parameters: {'n_estimators': 300, 'max_depth': 5, 'learning_rate': 0.019816427892769028, 'colsample_bytree': 0.512}
[1] 2025-05-31 02:40:15, 719] Trial 13 finished with value: 0.0038958731652623 and parameters: {'n_estimators': 255, 'max_depth': 4, 'learning_rate': 0.02070636246688168, 'colsample_bytree': 0.6683}
[1] 2025-05-31 02:40:15, 873] Trial 14 finished with value: 0.003895893840235638 and parameters: {'n_estimators': 255, 'max_depth': 4, 'learning_rate': 0.0221289890083268, 'colsample_bytree': 0.7895}
[1] 2025-05-31 02:40:16, 026] Trial 15 finished with value: 0.00389862062358904 and parameters: {'n_estimators': 270, 'max_depth': 5, 'learning_rate': 0.01643824633155253, 'colsample_bytree': 0.50645}
[1] 2025-05-31 02:40:16, 166] Trial 16 finished with value: 0.0038959650651099013 and parameters: {'n_estimators': 208, 'max_depth': 7, 'learning_rate': 0.02856154646940734, 'colsample_bytree': 0.861}
[1] 2025-05-31 02:40:16, 267] Trial 17 finished with value: 0.003894949302679955 and parameters: {'n_estimators': 92, 'max_depth': 4, 'learning_rate': 0.01895160100590423, 'colsample_bytree': 0.98387}
[1] 2025-05-31 02:40:16, 406] Trial 18 finished with value: 0.00389891994051552 and parameters: {'n_estimators': 196, 'max_depth': 5, 'learning_rate': 0.0701861705414003, 'colsample_bytree': 0.65514}
[1] 2025-05-31 02:40:16, 563] Trial 19 finished with value: 0.003898542071838833 and parameters: {'n_estimators': 267, 'max_depth': 7, 'learning_rate': 0.030474152774481247, 'colsample_bytree': 0.549}
[1] 2025-05-31 02:40:16, 702] A new study created in memory with name: no-name-1c35a25-4e1-4fed-a3d-724eb0bb1b
[1] 2025-05-31 02:40:16, 855] Trial 0 finished with value: 0.0037820570002645 and parameters: {'n_estimators': 260, 'max_depth': 7, 'learning_rate': 0.083911380961763861, 'colsample_bytree': 0.540477}
[1] 2025-05-31 02:40:16, 952] Trial 1 finished with value: 0.003709633752071651 and parameters: {'n_estimators': 128, 'max_depth': 7, 'learning_rate': 0.06752778337482, 'colsample_bytree': 0.75451}
[1] 2025-05-31 02:40:17, 102] Trial 2 finished with value: 0.00375502593797007017 and parameters: {'n_estimators': 251, 'max_depth': 8, 'learning_rate': 0.03249623022367, 'colsample_bytree': 0.994530}
[1] 2025-05-31 02:40:17, 243] Trial 3 finished with value: 0.0037550278122843031 and parameters: {'n_estimators': 253, 'max_depth': 3, 'learning_rate': 0.01553232157781433, 'colsample_bytree': 0.78899}
[1] 2025-05-31 02:40:17, 781] Trial 4 finished with value: 0.00377528122843031 and parameters: {'n_estimators': 80, 'max_depth': 8, 'learning_rate': 0.020187281382656515, 'colsample_bytree': 0.85635}
[1] 2025-05-31 02:40:17, 991] Trial 5 finished with value: 0.00371579772783564 and parameters: {'n_estimators': 191, 'max_depth': 6, 'learning_rate': 0.036186379314067316, 'colsample_bytree': 0.541}
[1] 2025-05-31 02:40:18, 139] Trial 6 finished with value: 0.003717489250510454 and parameters: {'n_estimators': 206, 'max_depth': 6, 'learning_rate': 0.0239487297702323, 'colsample_bytree': 0.6474}
[1] 2025-05-31
```

[1] 2025-05-31 02:40:24.053 Trial 1 finished with value: 0.00789106738160672110 and parameters: {'n\_estimators': 127, 'max\_depth': 5, 'learning\_rate': 0.00068760549371110, 'subsample': 0.000630067535111, 'colsample\_bytree': 0.910173, 'colsample\_bytree': 0.910173}

[1] 2025-05-31 02:40:24.323 Trial 3 finished with value: 0.00789142367296617 and parameters: {'n\_estimators': 298, 'max\_depth': 4, 'learning\_rate': 0.03779510505936083, 'subsample': 0.71723220087156, 'colsample\_bytree': 0.635853, 'colsample\_bytree': 0.635853}

[1] 2025-05-31 02:40:24.521 Trial 3 finished with value: 0.00789142367296617 and parameters: {'n\_estimators': 180, 'max\_depth': 5, 'learning\_rate': 0.01197508191685279, 'subsample': 0.8625720758930251, 'colsample\_bytree': 0.79757, 'colsample\_bytree': 0.79757}

[1] 2025-05-31 02:40:24.596 Trial 4 finished with value: 0.007891401682030136 and parameters: {'n\_estimators': 54, 'max\_depth': 7, 'learning\_rate': 0.0382103955834426, 'subsample': 0.7128873713577524, 'colsample\_bytree': 0.83840128, 'colsample\_bytree': 0.83840128}

[1] 2025-05-31 02:40:24.793 Trial 5 finished with value: 0.007891704209791253 and parameters: {'n\_estimators': 149, 'max\_depth': 5, 'learning\_rate': 0.01002926149473689, 'subsample': 0.5826708381165913, 'colsample\_bytree': 0.64475, 'colsample\_bytree': 0.64475}

[1] 2025-05-31 02:40:24.913 Trial 6 finished with value: 0.007845042154594802 and parameters: {'n\_estimators': 235, 'max\_depth': 6, 'learning\_rate': 0.14546404245286018, 'subsample': 0.5443671071141885, 'colsample\_bytree': 0.594856, 'colsample\_bytree': 0.594856}

[1] 2025-05-31 02:40:25.051 Trial 7 finished with value: 0.00782634039964904 and parameters: {'n\_estimators': 288, 'max\_depth': 6, 'learning\_rate': 0.2193989345535764, 'subsample': 0.620717719632306, 'colsample\_bytree': 0.945324, 'colsample\_bytree': 0.945324}

[1] 2025-05-31 02:40:25.145 Trial 8 finished with value: 0.00794108499354802 and parameters: {'n\_estimators': 139, 'max\_depth': 5, 'learning\_rate': 0.177422051099765, 'subsample': 0.797757144130359, 'colsample\_bytree': 0.976254, 'colsample\_bytree': 0.976254}

[1] 2025-05-31 02:40:25.235 Trial 9 finished with value: 0.00791411962963211 and parameters: {'n\_estimators': 139, 'max\_depth': 8, 'learning\_rate': 0.02555809196296537, 'subsample': 0.7648451577000622, 'colsample\_bytree': 0.951072, 'colsample\_bytree': 0.951072}

[1] 2025-05-31 02:40:25.393 Trial 10 finished with value: 0.007871727217169292 and parameters: {'n\_estimators': 286, 'max\_depth': 3, 'learning\_rate': 0.0803981792324877, 'subsample': 0.6179214545301990, 'colsample\_bytree': 0.52154, 'colsample\_bytree': 0.52154}

[1] 2025-05-31 02:40:25.540 Trial 11 finished with value: 0.00784561108353583 and parameters: {'n\_estimators': 237, 'max\_depth': 6, 'learning\_rate': 0.1379244525803870, 'subsample': 0.508827939715066, 'colsample\_bytree': 0.68400, 'colsample\_bytree': 0.68400}

[1] 2025-05-31 02:40:26.678 Trial 12 finished with value: 0.00786940307788757 and parameters: {'n\_estimators': 248, 'max\_depth': 5, 'learning\_rate': 0.28429091838691267, 'subsample': 0.6159259861061333, 'colsample\_bytree': 0.565948, 'colsample\_bytree': 0.565948}

[1] 2025-05-31 02:40:25.845 Trial 13 finished with value: 0.007825048163904019 and parameters: {'n\_estimators': 257, 'max\_depth': 7, 'learning\_rate': 0.1313520961413285, 'subsample': 0.60327545219532, 'colsample\_bytree': 0.88442, 'colsample\_bytree': 0.88442}

[1] 2025-05-31 02:40:26.066 Trial 14 finished with value: 0.00791904245958086 and parameters: {'n\_estimators': 268, 'max\_depth': 7, 'learning\_rate': 0.096805349542131, 'subsample': 0.666204586361024, 'colsample\_bytree': 0.880916, 'colsample\_bytree': 0.880916}

[1] 2025-05-31 02:40:26.146 Trial 15 finished with value: 0.0078241402001596 and parameters: {'n\_estimators': 210, 'max\_depth': 7, 'learning\_rate': 0.19698242518396786, 'subsample': 0.5218477973636336, 'colsample\_bytree': 0.99045, 'colsample\_bytree': 0.99045}

[1] 2025-05-31 02:40:26.282 Trial 16 finished with value: 0.007891477425455712 and parameters: {'n\_estimators': 196, 'max\_depth': 8, 'learning\_rate': 0.110384030389821, 'subsample': 0.5157700240879216, 'colsample\_bytree': 0.884499, 'colsample\_bytree': 0.884499}

[1] 2025-05-31 02:40:26.420 Trial 17 finished with value: 0.0078751961062961 and parameters: {'n\_estimators': 210, 'max\_depth': 7, 'learning\_rate': 0.04744706286645354, 'subsample': 0.55908976782848, 'colsample\_bytree': 0.993713, 'colsample\_bytree': 0.993713}

[1] 2025-05-31 02:40:26.526 Trial 18 finished with value: 0.007851752618939455 and parameters: {'n\_estimators': 97, 'max\_depth': 7, 'learning\_rate': 0.18327657236039524, 'subsample': 0.5031305402904769, 'colsample\_bytree': 0.751094, 'colsample\_bytree': 0.751094}

[1] 2025-05-31 02:40:26.678 Trial 19 finished with value: 0.00788550107151248 and parameters: {'n\_estimators': 257, 'max\_depth': 8, 'learning\_rate': 0.11726169110705136, 'subsample': 0.669224015668151, 'colsample\_bytree': 0.883496, 'colsample\_bytree': 0.883496}

[1] 2025-05-31 02:40:26.805 A new study created in memory with name: no-name:a89102d1-4826-4801-e0e2-5426c980c3 Trial 0 finished with value: 0.00131974216429057 and parameters: {'n\_estimators': 256, 'max\_depth': 3, 'learning\_rate': 0.24895787407773698, 'subsample': 0.5870430687503058, 'colsample\_bytree': 0.76180, 'colsample\_bytree': 0.76180}

[1] 2025-05-31 02:40:26.978 Trial 1 finished with value: 0.001456371189807577 and parameters: {'n\_estimators': 178, 'max\_depth': 5, 'learning\_rate': 0.13205211913897652, 'subsample': 0.5767376712115253, 'colsample\_bytree': 0.77664, 'colsample\_bytree': 0.77664}

[1] 2025-05-31 02:40:27.085 Trial 2 finished with value: 0.001380706509107744 and parameters: {'n\_estimators': 192, 'max\_depth': 6, 'learning\_rate': 0.0636827984955068, 'subsample': 0.614024247217368, 'colsample\_bytree': 0.58167, 'colsample\_bytree': 0.58167}

[1] 2025-05-31 02:40:27.193 Trial 3 finished with value: 0.0013804711323309614 and parameters: {'n\_estimators': 142, 'max\_depth': 8, 'learning\_rate': 0.0778315746281753, 'subsample': 0.8696649628852032, 'colsample\_bytree': 0.50100, 'colsample\_bytree': 0.50100}

[1] 2025-05-31 02:40:27.367 Trial 4 finished with value: 0.001328784943644 and parameters: {'n\_estimators': 51, 'max\_depth': 4, 'learning\_rate': 0.204663364249747, 'subsample': 0.543503094194162, 'colsample\_bytree': 0.8841305, 'colsample\_bytree': 0.8841305}

[1] 2025-05-31 02:40:27.528 Trial 5 finished with value: 0.00135932287904071 and parameters: {'n\_estimators': 251, 'max\_depth': 7, 'learning\_rate': 0.010787402055598864, 'subsample': 0.844307480186033, 'colsample\_bytree': 0.7524, 'colsample\_bytree': 0.7524}

[1] 2025-05-31 02:40:27.626 Trial 6 finished with value: 0.001361145097288050 and parameters: {'n\_estimators': 146, 'max\_depth': 3, 'learning\_rate': 0.12599959426957432, 'subsample': 0.96037710144628, 'colsample\_bytree': 0.939670, 'colsample\_bytree': 0.939670}

[1] 2025-05-31 02:40:27.691 Trial 7 finished with value: 0.001345791076983013 and parameters: {'n\_estimators': 58, 'max\_depth': 3, 'learning\_rate': 0.05192391119010609, 'subsample': 0.6321226438485226, 'colsample\_bytree': 0.795635, 'colsample\_bytree': 0.795635}

[1] 2025-05-31 02:40:27.801 Trial 8 finished with value: 0.00140015018208477 and parameters: {'n\_estimators': 186, 'max\_depth': 3, 'learning\_rate': 0.1769843626762834, 'subsample': 0.6667183610882935, 'colsample\_bytree': 0.878833, 'colsample\_bytree': 0.878833}

[1] 2025-05-31 02:40:27.925 Trial 9 finished with value: 0.0013554806581496169 and parameters: {'n\_estimators': 248, 'max\_depth': 8, 'learning\_rate': 0.033737938272743, 'subsample': 0.8834626462074609, 'colsample\_bytree': 0.78122, 'colsample\_bytree': 0.78122}

[1] 2025-05-31 02:40:28.094 Trial 10 finished with value: 0.0013668586713069467 and parameters: {'n\_estimators': 299, 'max\_depth': 5, 'learning\_rate': 0.2706283493244675, 'subsample': 0.7128576540538813, 'colsample\_bytree': 0.6567, 'colsample\_bytree': 0.6567}

[1] 2025-05-31 02:40:28.197 Trial 11 finished with value: 0.00131730433692994 and parameters: {'n\_estimators': 112, 'max\_depth': 4, 'learning\_rate': 0.2673756565281753, 'subsample': 0.527968550240763, 'colsample\_bytree': 0.99129, 'colsample\_bytree': 0.99129}

[1] 2025-05-31 02:40:28.298 Trial 12 finished with value: 0.00130919981796966 and parameters: {'n\_estimators': 104, 'max\_depth': 4, 'learning\_rate': 0.286464791029214, 'subsample': 0.5079767722004842, 'colsample\_bytree': 0.995670, 'colsample\_bytree': 0.995670}

[1] 2025-05-31 02:40:28.404 Trial 13 finished with value: 0.001309623240545591 and parameters: {'n\_estimators': 100, 'max\_depth': 4, 'learning\_rate': 0.2241323314542439, 'subsample': 0.5205681486260744, 'colsample\_bytree': 0.9854, 'colsample\_bytree': 0.9854}

[1] 2025-05-31 02:40:28.511 Trial 14 finished with value: 0.00136410269019802 and parameters: {'n\_estimators': 103, 'max\_depth': 4, 'learning\_rate': 0.29473607123587, 'subsample': 0.504049859040849, 'colsample\_bytree': 0.985130843, 'colsample\_bytree': 0.985130843}

[1] 2025-05-31 02:40:28.615 Trial 15 finished with value: 0.001383755107148518 and parameters: {'n\_estimators': 105, 'max\_depth': 6, 'learning\_rate': 0.11659046646617026, 'subsample': 0.754839113300266, 'colsample\_bytree': 0.88012, 'colsample\_bytree': 0.88012}

[1] 2025-05-31 02:40:28.727 Trial 16 finished with value: 0.001377056707323396 and parameters: {'n\_estimators': 137, 'max\_depth': 5, 'learning\_rate': 0.0898768891303239, 'subsample': 0.7654819113537527, 'colsample\_bytree': 0.9440, 'colsample\_bytree': 0.9440}

[1] 2025-05-31 02:40:28.826 Trial 17 finished with value: 0.0013477566648913895 and parameters: {'n\_estimators': 81, 'max\_depth': 4, 'learning\_rate': 0.0974487820491115, 'subsample': 0.5804731841649365, 'colsample\_bytree': 0.66466, 'colsample\_bytree': 0.66466}

[1] 2025-05-31 02:40:28.938 Trial 18 finished with value: 0.0013397471858189103 and parameters: {'n\_estimators': 127, 'max\_depth': 4, 'learning\_rate': 0.04707868230491115, 'subsample': 0.639194850964895, 'colsample\_bytree': 0.8383, 'colsample\_bytree': 0.8383}

[1] 2025-05-31 02:40:29.084 Trial 19 finished with value: 0.00137443812722333 and parameters: {'n\_estimators': 212, 'max\_depth': 6, 'learning\_rate': 0.01584498810826986, 'subsample': 0.6111389068778615, 'colsample\_bytree': 0.99981, 'colsample\_bytree': 0.99981}

[1] 2025-05-31 02:40:29.174 A new study created in memory with name: no-name:b889c85e-146d-43b0-8579-81d5a802157 Trial 0 finished with value: 0.001054824604906505 and parameters: {'n\_estimators': 181, 'max\_depth': 8, 'learning\_rate': 0.0524189374640057, 'subsample': 0.637567504659069, 'colsample\_bytree': 0.707086, 'colsample\_bytree': 0.707086}

[1] 2025-05-31 02:40:29.393 Trial 1 finished with value: 0.00102725062362348 and parameters: {'n\_estimators': 128, 'max\_depth': 8, 'learning\_rate': 0.10566162487041389, 'subsample': 0.7713732873071, 'colsample\_bytree': 0.725306, 'colsample\_bytree': 0.725306}

[1] 2025-05-31 02:40:29.472 Trial 2 finished with value: 0.001009191267990142 and parameters: {'n\_estimators': 111, 'max\_depth': 6, 'learning\_rate': 0.047611071010236, 'subsample': 0.635250353909491, 'colsample\_bytree': 0.5382, 'colsample\_bytree': 0.5382}

[1] 2025-05-31 02:40:29.515 Trial 3 finished with value: 0.0010070164509718518 and parameters: {'n\_estimators': 82, 'max\_depth': 4, 'learning\_rate': 0.010334782873357835, 'subsample': 0.59834044357878, 'colsample\_bytree': 0.58404, 'colsample\_bytree': 0.58404}

[1] 2025-05-31 02:40:29.632 Trial 4 finished with value: 0.00103631858664208 and parameters: {'n\_estimators': 116, 'max\_depth': 7, 'learning\_rate': 0.01051681540532927, 'subsample': 0.6329476502407682, 'colsample\_bytree': 0.7888, 'colsample\_bytree': 0.7888}

[1] 2025-05-31 02:40:29.764 Trial 5 finished with value: 0.0010233244791761014 and parameters: {'n\_estimators': 249, 'max\_depth': 8, 'learning\_rate': 0.2377173052110411, 'subsample': 0.701676910041112, 'colsample\_bytree': 0.991155, 'colsample\_bytree': 0.991155}

[1] 2025-05-31 02:40:29.848 Trial 6 finished with value: 0.00100209409329167 and parameters: {'n\_estimators': 119, 'max\_depth': 6, 'learning\_rate': 0.06831039384500214, 'subsample': 0.6573320364085007, 'colsample\_bytree': 0.85900, 'colsample\_bytree': 0.85900}

[1] 2025-05-31 02:40:29.946 Trial 7 finished with value: 0.001068546764675129 and parameters: {'n\_estimators': 166, 'max\_depth': 5, 'learning\_rate': 0.1352793621619793, 'subsample': 0.805398344549532, 'colsample\_bytree': 0.826613, 'colsample\_bytree': 0.826613}

[1] 2025-05-31 02:40:30.032 Trial 8 finished with value: 0.0010330771461010236 and parameters: {'n\_estimators': 125, 'max\_depth': 3, 'learning\_rate': 0.11515624646097338, 'subsample': 0.8860844378786449, 'colsample\_bytree': 0.522404, 'colsample\_bytree': 0.522404}

[1] 2025-05-31 02:40:30.128 Trial 9 finished with value: 0.00102837518371852 and parameters: {'n\_estimators': 154, 'max\_depth': 6, 'learning\_rate': 0.14474815316682168, 'subsample': 0.5294362149019948, 'colsample\_bytree': 0.88584, 'colsample\_bytree': 0.88584}

[1] 2025-05-31 02:40:30.219 Trial 10 finished with value: 0.001049593764520966 and parameters: {'n\_estimators': 53, 'max\_depth': 5, 'learning\_rate': 0.2984917165990296, 'subsample': 0.9579561499298543, 'colsample\_bytree': 0.60127, 'colsample\_bytree': 0.60127}

[1] 2025-05-31 02:40:30.360 Trial 11 finished with value: 0.00998205055805970 and parameters: {'n\_estimators': 210, 'max\_depth': 6, 'learning\_rate': 0.041949364165237, 'subsample': 0.629497825384272, 'colsample\_bytree': 0.9232, 'colsample\_bytree': 0.9232}

[1] 2025-05-31 02:40:30.503 Trial 12 finished with value: 0.00102040014018383 and parameters: {'n\_estimators': 228, 'max\_depth': 6, 'learning\_rate': 0.0281200958491727, 'subsample': 0.5163649452596455, 'colsample\_bytree': 0.95457, 'colsample\_bytree': 0.95457}

[1] 2025-05-31 02:40:30.700 Trial 13 finished with value: 0.0010734726743686 and parameters: {'n\_estimators': 294, 'max\_depth': 7, 'learning\_rate': 0.03162815865582966, 'subsample': 0.6113051466277175, 'colsample\_bytree': 0.655, 'colsample\_bytree': 0.655}

[1] 2025-05-31 02:40:30.838 Trial 14 finished with value: 0.000998898665098538 and parameters: {'n\_estimators': 210, 'max\_depth': 4, 'learning\_rate': 0.05457622856280157, 'subsample': 0.5892852329308045, 'colsample\_bytree': 0.92212, 'colsample\_bytree': 0.92212}

[1] 2025-05-31 02:40:30.980 Trial 15 finished with value: 0.0010674094721523 and parameters: {'n\_estimators': 216, 'max\_depth': 3, 'learning\_rate': 0.019381822589839139, 'subsample': 0.5541589198700366, 'colsample\_bytree': 0.923, 'colsample\_bytree': 0.923}

[1] 2025-05-31 02:40:31.121 Trial 16 finished with value: 0.001005822851871852 and parameters: {'n\_estimators': 205, 'max\_depth': 4, 'learning\_rate': 0.04358680043754688, 'subsample': 0.709836117247379, 'colsample\_bytree': 0.91192, 'colsample\_bytree': 0.91192}

[1] 2025-05-31 02:40:31.211 Trial 17 finished with value: 0.0010797824474455763 and parameters: {'n\_estimators': 264, 'max\_depth': 4, 'learning\_rate': 0.0817856047913624, 'subsample': 0.5719123922471171, 'colsample\_bytree': 0.78025, 'colsample\_bytree': 0.78025}

[1] 2025-05-31 02:40:31.411 Trial 18 finished with value: 0.001045557380562618 and parameters: {'n\_estimators': 194, 'max\_depth': 5, 'learning\_rate': 0.01921996651404346, 'subsample': 0.847692851054454, 'colsample\_bytree': 0.986, 'colsample\_bytree': 0.986}

[1] 2025-05-31 02:40:31.568 Trial 19 finished with value: 0.0010238681689432953 and parameters: {'n\_estimators': 273, 'max\_depth': 7, 'learning\_rate': 0.22597111094592835, 'subsample': 0.7229753150389007, 'colsample\_bytree': 0.8339, 'colsample\_bytree': 0.8339}

[1] 2025-05-31 02:40:31.725 A new study created in memory with name: no-name:f5d3f0b-4a3c-4c8a-8359-518b5c9d5fc3 Trial 0 finished with value: 0.00126173503665455 and parameters: {'n\_estimators': 257, 'max\_depth': 4, 'learning\_rate': 0.061166584954037543, 'subsample': 0.5434860403671443, 'colsample\_bytree': 0.77691, 'colsample\_bytree': 0.77691}

[1] 2025-05-31 02:40:31.992 Trial 1 finished with value: 0.001235395820198298 and parameters: {'n\_estimators': 191, 'max\_depth': 7, 'learning\_rate': 0.0203790111723068, 'subsample': 0.871032749242705, 'colsample\_bytree': 0.51260, 'colsample\_bytree': 0.51260}

[1] 2025-05-31 02:40:32.122 Trial 2 finished with value: 0.001250176370873193 and parameters: {'n\_estimators': 255, 'max\_depth': 4, 'learning\_rate': 0.0149538494784662, 'subsample': 0.898862325258208, 'colsample\_bytree': 0.92727, 'colsample\_bytree': 0.92727}

[1] 2025-05-31 02:40:32.288 Trial 3 finished with value: 0.001234415376347254 and parameters: {'n\_estimators': 186, 'max\_depth': 3, 'learning\_rate': 0.0353899326567987, 'subsample': 0.8725402259139049, 'colsample\_bytree': 0.843977, 'colsample\_bytree': 0.843977}

[1] 2025-05-31 02:40:32.357 Trial 4 finished with value: 0.0012673753360612577 and parameters: {'n\_estimators': 242, 'max\_depth': 3, 'learning\_rate': 0.01298394347207042, 'subsample': 0.5622419953968975, 'colsample\_bytree': 0.8822, 'colsample\_bytree': 0.8822}

[1] 2025-05-31 02:40:32.539 Trial 5 finished with value: 0.00188975072006114077 and parameters: {'n\_estimators': 166, 'max\_depth': 3, 'learning\_rate': 0.01051922429593167, 'subsample': 0.595945951043336, 'colsample\_bytree': 0.8357, 'colsample\_bytree': 0.8357}

[1] 2025-05-31 02:40:32.632 Trial 7 finished with value: 0.001250176370873193 and parameters: {'n\_estimators': 143, 'max\_depth': 7, 'learning\_rate': 0.01192082081643779, 'subsample': 0.63568232035774609, 'colsample\_bytree': 0.9579, 'colsample\_bytree': 0.9579}

[1] 2025-05-31 02:40:32.759 Trial 8 finished with value: 0.0012264838523928325 and parameters: {'n\_estimators': 175, 'max\_depth': 8, 'learning\_rate': 0.01628683181910672, 'subsample': 0.5419147830802081, 'colsample\_bytree': 0.65804, 'colsample\_bytree': 0.65804}

[1] 2025-05-31 02:40:32.840 Trial 9 finished with value: 0.0012020982545676938 and parameters: {'n\_estimators': 109, 'max\_depth': 6, 'learning\_rate': 0.0916682423454026, 'subsample': 0.69682403379339, 'colsample\_bytree': 0.62987, 'colsample\_bytree': 0.62987}

[1] 2025-05-31 02:40:32.933 Trial 10 finished with value: 0.0012814737173179439 and parameters: {'n\_estimators': 52, 'max\_depth': 5, 'learning\_rate': 0.09335120379322, 'subsample': 0.6828739933004833, 'colsample\_bytree': 0.6078, 'colsample\_bytree': 0.6078}

[1] 2025-05-31 02:40:33.035 Trial 11 finished with value: 0.00122628385926954 and parameters: {'n\_estimators': 103, 'max\_depth': 4, 'learning\_rate': 0.093461302379156, 'subsample': 0.6830432379156, 'colsample\_bytree': 0.567654, 'colsample\_bytree': 0.567654}

[1] 2025-05-31 02:40:33.151 Trial 12 finished with value: 0.00134467932

[1] 2025-05-31 02:40:42,722 Trial 9 finished with value: 0.001995417487652115 and parameters: {'n\_estimators': 119, 'max\_depth': 5, 'learning\_rate': 0.01440289885449559, 'subsample': 0.573837653159994, 'colsample\_bytree': 0.6964}, [1] 2025-05-31 02:40:42,847 Trial 10 finished with value: 0.00195958108172603 and parameters: {'n\_estimators': 180, 'max\_depth': 7, 'learning\_rate': 0.0494387265095616, 'subsample': 0.5244291436159061, 'colsample\_bytree': 0.6913}, [1] 2025-05-31 02:40:42,970 Trial 11 finished with value: 0.00196584909995 and parameters: {'n\_estimators': 172, 'max\_depth': 7, 'learning\_rate': 0.0475998125323218, 'subsample': 0.516936070811655, 'colsample\_bytree': 0.710665}, [1] 2025-05-31 02:40:43,136 Trial 12 finished with value: 0.001951941963438984 and parameters: {'n\_estimators': 179, 'max\_depth': 7, 'learning\_rate': 0.04693773738332031, 'subsample': 0.054716069700165, 'colsample\_bytree': 0.6434}, [1] 2025-05-31 02:40:43,262 Trial 13 finished with value: 0.00199275016340846 and parameters: {'n\_estimators': 167, 'max\_depth': 7, 'learning\_rate': 0.03210165281084384, 'subsample': 0.16464464898813, 'colsample\_bytree': 0.61946}, [1] 2025-05-31 02:40:43,412 Trial 14 finished with value: 0.00205961996633354 and parameters: {'n\_estimators': 259, 'max\_depth': 6, 'learning\_rate': 0.02770102652689245, 'subsample': 0.0273891469291332, 'colsample\_bytree': 0.7694}, [1] 2025-05-31 02:40:43,541 Trial 15 finished with value: 0.001937517816568588 and parameters: {'n\_estimators': 201, 'max\_depth': 8, 'learning\_rate': 0.0704906619515486, 'subsample': 0.05476881117056, 'colsample\_bytree': 0.61361}, [1] 2025-05-31 02:40:43,675 Trial 16 finished with value: 0.001940128157928775 and parameters: {'n\_estimators': 202, 'max\_depth': 8, 'learning\_rate': 0.041026461314819844, 'subsample': 0.0585950957147111, 'colsample\_bytree': 0.601}, [1] 2025-05-31 02:40:43,919 Trial 17 finished with value: 0.0020543031815956 and parameters: {'n\_estimators': 211, 'max\_depth': 8, 'learning\_rate': 0.072938245616446, 'subsample': 0.0790706608086505, 'colsample\_bytree': 0.5614}, [1] 2025-05-31 02:40:43,932 Trial 18 finished with value: 0.002081317325195043 and parameters: {'n\_estimators': 135, 'max\_depth': 8, 'learning\_rate': 0.016990201629876492, 'subsample': 0.870557295350594, 'colsample\_bytree': 0.5909}, [1] 2025-05-31 02:40:44,064 Trial 19 finished with value: 0.001986187699267205 and parameters: {'n\_estimators': 196, 'max\_depth': 6, 'learning\_rate': 0.0104067014818423, 'subsample': 0.551982233083993, 'colsample\_bytree': 0.500799}, [1] 2025-05-31 02:40:44,212 A new study created in memory with name: no-name-ec94ff1-62bb-4795-b63-0fbfd5a4c0b, [1] 2025-05-31 02:40:44,359 Trial 0 finished with value: 0.0028051984868911 and parameters: {'n\_estimators': 232, 'max\_depth': 5, 'learning\_rate': 0.16612530090692956, 'subsample': 0.9952083896182913, 'colsample\_bytree': 0.6443248}, [1] 2025-05-31 02:40:44,476 Trial 1 finished with value: 0.00285286010210552 and parameters: {'n\_estimators': 223, 'max\_depth': 8, 'learning\_rate': 0.10705207170545473, 'subsample': 0.5523112623260622, 'colsample\_bytree': 0.548074}, [1] 2025-05-31 02:40:44,568 Trial 2 finished with value: 0.002807692185524574 and parameters: {'n\_estimators': 131, 'max\_depth': 3, 'learning\_rate': 0.0976391638490238, 'subsample': 0.058152791849888, 'colsample\_bytree': 0.806415}, [1] 2025-05-31 02:40:44,680 Trial 3 finished with value: 0.0028864920085953 and parameters: {'n\_estimators': 236, 'max\_depth': 6, 'learning\_rate': 0.02088362245639168, 'subsample': 0.52927950972910705, 'colsample\_bytree': 0.5235}, [1] 2025-05-31 02:40:44,817 Trial 4 finished with value: 0.00281631214804983 and parameters: {'n\_estimators': 254, 'max\_depth': 8, 'learning\_rate': 0.06679631261842695, 'subsample': 0.76738195258749, 'colsample\_bytree': 0.7870569}, [1] 2025-05-31 02:40:44,928 Trial 5 finished with value: 0.002846652815364006 and parameters: {'n\_estimators': 212, 'max\_depth': 7, 'learning\_rate': 0.0141065425720408, 'subsample': 0.6898961987079066, 'colsample\_bytree': 0.50597}, [1] 2025-05-31 02:40:45,050 Trial 6 finished with value: 0.002824007784297094 and parameters: {'n\_estimators': 241, 'max\_depth': 6, 'learning\_rate': 0.029472373901126737, 'subsample': 0.85807187326399, 'colsample\_bytree': 0.718859}, [1] 2025-05-31 02:40:45,219 Trial 7 finished with value: 0.002824959767149196 and parameters: {'n\_estimators': 264, 'max\_depth': 3, 'learning\_rate': 0.1037585734742706, 'subsample': 0.7080277093020222, 'colsample\_bytree': 0.70004}, [1] 2025-05-31 02:40:45,354 Trial 8 finished with value: 0.00286507783964378 and parameters: {'n\_estimators': 269, 'max\_depth': 6, 'learning\_rate': 0.0890420289210783, 'subsample': 0.958796598470387, 'colsample\_bytree': 0.97694}, [1] 2025-05-31 02:40:45,445 Trial 9 finished with value: 0.0028177933750100717 and parameters: {'n\_estimators': 135, 'max\_depth': 6, 'learning\_rate': 0.122483928877989, 'subsample': 0.822650943127081, 'colsample\_bytree': 0.636057}, [1] 2025-05-31 02:40:45,534 Trial 10 finished with value: 0.002805814696946045 and parameters: {'n\_estimators': 67, 'max\_depth': 4, 'learning\_rate': 0.283263622052198, 'subsample': 0.99649080398881, 'colsample\_bytree': 0.99621345}, [1] 2025-05-31 02:40:45,703 Trial 11 finished with value: 0.002813631214804983 and parameters: {'n\_estimators': 300, 'max\_depth': 5, 'learning\_rate': 0.2836136964943863, 'subsample': 0.836319382894491, 'colsample\_bytree': 0.9872}, [1] 2025-05-31 02:40:45,837 Trial 12 finished with value: 0.002825676336764894 and parameters: {'n\_estimators': 184, 'max\_depth': 5, 'learning\_rate': 0.0460529267478474, 'subsample': 0.914123019001125, 'colsample\_bytree': 0.88863}, [1] 2025-05-31 02:40:46,004 Trial 13 finished with value: 0.002859325053693061 and parameters: {'n\_estimators': 299, 'max\_depth': 4, 'learning\_rate': 0.18823801245539735, 'subsample': 0.913351463512531, 'colsample\_bytree': 0.61821}, [1] 2025-05-31 02:40:46,137 Trial 14 finished with value: 0.00282331018416846 and parameters: {'n\_estimators': 185, 'max\_depth': 7, 'learning\_rate': 0.1498909980121934, 'subsample': 0.9357557530101054, 'colsample\_bytree': 0.87970}, [1] 2025-05-31 02:40:46,320 Trial 15 finished with value: 0.0028094195312494898 and parameters: {'n\_estimators': 273, 'max\_depth': 5, 'learning\_rate': 0.03541512816723448, 'subsample': 0.8626002594405449, 'colsample\_bytree': 0.6365}, [1] 2025-05-31 02:40:46,482 Trial 16 finished with value: 0.0028086776574322136 and parameters: {'n\_estimators': 273, 'max\_depth': 7, 'learning\_rate': 0.029912310087916363, 'subsample': 0.832567979959069, 'colsample\_bytree': 0.890}, [1] 2025-05-31 02:40:46,597 Trial 17 finished with value: 0.00280513837017448 and parameters: {'n\_estimators': 140, 'max\_depth': 4, 'learning\_rate': 0.0332930332489881, 'subsample': 0.878849071920918, 'colsample\_bytree': 0.82297}, [1] 2025-05-31 02:40:46,754 Trial 18 finished with value: 0.00281075249938866 and parameters: {'n\_estimators': 277, 'max\_depth': 6, 'learning\_rate': 0.06834528371382106, 'subsample': 0.776809794856168, 'colsample\_bytree': 0.5769}, [1] 2025-05-31 02:40:46,840 Trial 19 finished with value: 0.00282370357976793 and parameters: {'n\_estimators': 50, 'max\_depth': 5, 'learning\_rate': 0.01456711019268783, 'subsample': 0.65880713235861956, 'colsample\_bytree': 0.69422}, [1] 2025-05-31 02:40:46,991 A new study created in memory with name: no-name-9ac4d1-afcd-4f53-a2cf-e2ad080708ae, [1] 2025-05-31 02:40:47,116 Trial 0 finished with value: 0.000733649630103057 and parameters: {'n\_estimators': 132, 'max\_depth': 8, 'learning\_rate': 0.0898815763976375, 'subsample': 0.50074582575058516, 'colsample\_bytree': 0.508589}, [1] 2025-05-31 02:40:47,202 Trial 1 finished with value: 0.000745377982676187, 'subsample': 0.8376615350340375, 'colsample\_bytree': 0.6965}, [1] 2025-05-31 02:40:47,343 Trial 2 finished with value: 0.000861843962534494 and parameters: {'n\_estimators': 240, 'max\_depth': 6, 'learning\_rate': 0.12806276195263197, 'subsample': 0.80410626285925, 'colsample\_bytree': 0.549426}, [1] 2025-05-31 02:40:47,472 Trial 3 finished with value: 0.0007847458967303029 and parameters: {'n\_estimators': 262, 'max\_depth': 4, 'learning\_rate': 0.038482519290281934, 'subsample': 0.6295778319265768, 'colsample\_bytree': 0.5326}, [1] 2025-05-31 02:40:47,594 Trial 4 finished with value: 0.000822337903237725 and parameters: {'n\_estimators': 235, 'max\_depth': 5, 'learning\_rate': 0.0832118473027216, 'subsample': 0.677272149312146, 'colsample\_bytree': 0.706219}, [1] 2025-05-31 02:40:47,728 Trial 5 finished with value: 0.0008434740233931963 and parameters: {'n\_estimators': 270, 'max\_depth': 8, 'learning\_rate': 0.097774459143801, 'subsample': 0.99767111575177, 'colsample\_bytree': 0.979536}, [1] 2025-05-31 02:40:47,857 Trial 6 finished with value: 0.000761734579544815 and parameters: {'n\_estimators': 267, 'max\_depth': 7, 'learning\_rate': 0.10191238024476915, 'subsample': 0.698193977137771, 'colsample\_bytree': 0.610274}, [1] 2025-05-31 02:40:48,023 Trial 7 finished with value: 0.0007918977137572113 and parameters: {'n\_estimators': 288, 'max\_depth': 7, 'learning\_rate': 0.05997957620466244, 'subsample': 0.5170913472403441, 'colsample\_bytree': 0.569914}, [1] 2025-05-31 02:40:48,172 Trial 8 finished with value: 0.0008178173368191941 and parameters: {'n\_estimators': 243, 'max\_depth': 6, 'learning\_rate': 0.0275422953789782, 'subsample': 0.672462318920975, 'colsample\_bytree': 0.7269}, [1] 2025-05-31 02:40:48,469 Trial 9 finished with value: 0.000809707641354699 and parameters: {'n\_estimators': 295, 'max\_depth': 8, 'learning\_rate': 0.0376560329710457, 'subsample': 0.78591946260661, 'colsample\_bytree': 0.7416044}, [1] 2025-05-31 02:40:49,024 Trial 10 finished with value: 0.00079739329196507 and parameters: {'n\_estimators': 50, 'max\_depth': 3, 'learning\_rate': 0.22480036338495, 'subsample': 0.5126931961296586, 'colsample\_bytree': 0.8784627}, [1] 2025-05-31 02:40:49,212 Trial 11 finished with value: 0.000835646935605521 and parameters: {'n\_estimators': 166, 'max\_depth': 7, 'learning\_rate': 0.0115184323974933, 'subsample': 0.915505035104772, 'colsample\_bytree': 0.617}, [1] 2025-05-31 02:40:49,457 Trial 12 finished with value: 0.0008151501730578263 and parameters: {'n\_estimators': 123, 'max\_depth': 7, 'learning\_rate': 0.2170592937564508, 'subsample': 0.6089781723747816, 'colsample\_bytree': 0.5056}, [1] 2025-05-31 02:40:50,293 Trial 13 finished with value: 0.0006957584875486707 and parameters: {'n\_estimators': 175, 'max\_depth': 7, 'learning\_rate': 0.1549493613905756, 'subsample': 0.5835349716491297, 'colsample\_bytree': 0.63806}, [1] 2025-05-31 02:40:50,526 Trial 14 finished with value: 0.000839393037111655 and parameters: {'n\_estimators': 189, 'max\_depth': 5, 'learning\_rate': 0.28791519469380544, 'subsample': 0.5715242714673544, 'colsample\_bytree': 0.8334}, [1] 2025-05-31 02:40:50,692 Trial 15 finished with value: 0.000819072359491172 and parameters: {'n\_estimators': 135, 'max\_depth': 8, 'learning\_rate': 0.1519095007702187, 'subsample': 0.564631463155383, 'colsample\_bytree': 0.6418}, [1] 2025-05-31 02:40:50,888 Trial 16 finished with value: 0.00081935046496686 and parameters: {'n\_estimators': 78, 'max\_depth': 6, 'learning\_rate': 0.06184936551851126, 'subsample': 0.5072814977346152, 'colsample\_bytree': 0.8061}, [1] 2025-05-31 02:40:51,091 Trial 17 finished with value: 0.000873230459065564 and parameters: {'n\_estimators': 191, 'max\_depth': 7, 'learning\_rate': 0.1772283123666768, 'subsample': 0.5393536499189463, 'colsample\_bytree': 0.6568}, [1] 2025-05-31 02:40:51,214 Trial 18 finished with value: 0.000841332315056965 and parameters: {'n\_estimators': 156, 'max\_depth': 8, 'learning\_rate': 0.116075797055092, 'subsample': 0.7148527317516392, 'colsample\_bytree': 0.584}, [1] 2025-05-31 02:40:51,322 Trial 19 finished with value: 0.000839240467499549 and parameters: {'n\_estimators': 97, 'max\_depth': 6, 'learning\_rate': 0.116057597055092, 'subsample': 0.638957623993814, 'colsample\_bytree': 0.51092}, [1] 2025-05-31 02:40:51,436 A new study created in memory with name: no-name-0100917-3776-4fbb-9762-6c43902819, [1] 2025-05-31 02:40:51,592 Trial 0 finished with value: 0.001582808262096892 and parameters: {'n\_estimators': 204, 'max\_depth': 5, 'learning\_rate': 0.05427310848803109, 'subsample': 0.6096632216348496, 'colsample\_bytree': 0.50581}, [1] 2025-05-31 02:40:51,669 Trial 1 finished with value: 0.001581992203824471 and parameters: {'n\_estimators': 101, 'max\_depth': 7, 'learning\_rate': 0.1798915847165849, 'subsample': 0.921849727282793, 'colsample\_bytree': 0.8802529}, [1] 2025-05-31 02:40:51,736 Trial 2 finished with value: 0.0015718692270665365 and parameters: {'n\_estimators': 73, 'max\_depth': 3, 'learning\_rate': 0.039455166929794, 'subsample': 0.6369878174957056, 'colsample\_bytree': 0.55336}, [1] 2025-05-31 02:40:51,848 Trial 3 finished with value: 0.001567840321590895 and parameters: {'n\_estimators': 204, 'max\_depth': 8, 'learning\_rate': 0.1799401403188069, 'subsample': 0.5693563400526585, 'colsample\_bytree': 0.694597}, [1] 2025-05-31 02:40:51,940 Trial 4 finished with value: 0.0015842778705120808 and parameters: {'n\_estimators': 117, 'max\_depth': 5, 'learning\_rate': 0.03765560329710457, 'subsample': 0.915505035104772, 'colsample\_bytree': 0.78303}, [1] 2025-05-31 02:40:52,087 Trial 5 finished with value: 0.00156902734526695 and parameters: {'n\_estimators': 263, 'max\_depth': 6, 'learning\_rate': 0.0114040618291727551, 'subsample': 0.5893403611214672, 'colsample\_bytree': 0.78128}, [1] 2025-05-31 02:40:52,181 Trial 6 finished with value: 0.0015832639653996137 and parameters: {'n\_estimators': 136, 'max\_depth': 7, 'learning\_rate': 0.020391614708716014, 'subsample': 0.941246961962945, 'colsample\_bytree': 0.508684}, [1] 2025-05-31 02:40:52,258 Trial 7 finished with value: 0.001565791303572632 and parameters: {'n\_estimators': 92, 'max\_depth': 3, 'learning\_rate': 0.0828523913190756, 'subsample': 0.8151286331082057, 'colsample\_bytree': 0.8877729}, [1] 2025-05-31 02:40:52,324 Trial 8 finished with value: 0.0015817367390157576 and parameters: {'n\_estimators': 61, 'max\_depth': 7, 'learning\_rate': 0.04780556039706515, 'subsample': 0.6486376198281663, 'colsample\_bytree': 0.959806}, [1] 2025-05-31 02:40:52,484 Trial 9 finished with value: 0.0015704727066536725 and parameters: {'n\_estimators': 247, 'max\_depth': 4, 'learning\_rate': 0.026779718550319362, 'subsample': 0.5540640932876276, 'colsample\_bytree': 0.6460}, [1] 2025-05-31 02:40:52,617 Trial 10 finished with value: 0.00159371283373825 and parameters: {'n\_estimators': 186, 'max\_depth': 8, 'learning\_rate': 0.28277171971972608, 'subsample': 0.7862607305005296, 'colsample\_bytree': 0.69836}, [1] 2025-05-31 02:40:52,736 Trial 11 finished with value: 0.001573021447937544 and parameters: {'n\_estimators': 158, 'max\_depth': 3, 'learning\_rate': 0.09961607781854209, 'subsample': 0.80293176497149734, 'colsample\_bytree': 0.87799}, [1] 2025-05-31 02:40:52,881 Trial 12 finished with value: 0.0015661653538105402 and parameters: {'n\_estimators': 223, 'max\_depth': 6, 'learning\_rate': 0.1453605181033783, 'subsample': 0.636336172781868, 'colsample\_bytree': 0.85879}, [1] 2025-05-31 02:40:53,061 Trial 13 finished with value: 0.001603305915895262 and parameters: {'n\_estimators': 162, 'max\_depth': 4, 'learning\_rate': 0.29560639808610624, 'subsample': 0.70481901038347, 'colsample\_bytree': 0.9830}, [1] 2025-05-31 02:40:53,171 Trial 14 finished with value: 0.0015794818082051046 and parameters: {'n\_estimators': 296, 'max\_depth': 6, 'learning\_rate': 0.1075111808707008, 'subsample': 0.86056278344772908, 'colsample\_bytree': 0.714944}, [1] 2025-05-31 02:40:53,271 Trial 15 finished with value: 0.0015876622052640104 and parameters: {'n\_estimators': 99, 'max\_depth': 4, 'learning\_rate': 0.19072938313513164, 'subsample': 0.728962597680588, 'colsample\_bytree': 0.82288}, [1] 2025-05-31 02:40:53,401 Trial 16 finished with value: 0.0015628401183051936 and parameters: {'n\_estimators': 222, 'max\_depth': 3, 'learning\_rate': 0.088227499717730646, 'subsample': 0.8619638398415742, 'colsample\_bytree': 0.64407}, [1] 2025-05-31 02:40:53,585 Trial 17 finished with value: 0.0015797530199832066 and parameters: {'n\_estimators': 230, 'max\_depth': 6, 'learning\_rate': 0.1305505010558598, 'subsample': 0.987662287735809, 'colsample\_bytree': 0.63888}, [1] 2025-05-31 02:40:53,745 Trial 18 finished with value: 0.0015687491487910193 and parameters: {'n\_estimators': 273, 'max\_depth': 8, 'learning\_rate': 0.188240516901590323, 'subsample': 0.8817282875790547, 'colsample\_bytree': 0.6621}, [1] 2025-05-31 02:40:53,874 Trial 19 finished with value: 0.001570125099930182 and parameters: {'n\_estimators': 195, 'max\_depth': 5, 'learning\_rate': 0.071228448257124, 'subsample': 0.768340833449517, 'colsample\_bytree': 0.603347}, [1] 2025-05-31 02:40:53,998 A new study created in memory with name: no-name-f88a048b-c097-49dc-80b2-1d2a02172515, [1] 2025-05-31 02:40:54,111 Trial 0 finished with value: 0.00122678059539465 and parameters: {'n\_estimators': 101, 'max\_depth': 7, 'learning\_rate': 0.2259061040525787, 'subsample': 0.5589744195350049, 'colsample\_bytree': 0.777520}, [1] 2025-05-31 02:40:54,251 Trial 1 finished with value: 0.00123085197129541 and parameters: {'n\_estimators': 280, 'max\_depth': 6, 'learning\_rate': 0.08083646224945201, 'subsample': 0.52896139148676381, 'colsample\_bytree': 0.687919}, [1] 2025-05-31 02:40:54,314 Trial 2 finished with value: 0.001221757799961464 and parameters: {'n\_estimators': 62, 'max\_depth': 7, 'learning\_rate': 0.06973083947049659, 'subsample': 0.87865747972988, 'colsample\_bytree': 0.8410109}, [1] 2025-05-31 02:40:54,425 Trial 3 finished with value: 0.00121929655041285472 and parameters: {'n\_estimators': 200, 'max\_depth': 5, 'learning\_rate': 0.025885079507191043, 'subsample': 0.6369878174957056, 'colsample\_bytree': 0.55336}, [1] 2025-05-31 02:40:55,069 Trial 4 finished with value: 0.001201451340514726872 and parameters: {'n\_estimators': 50, 'max\_depth': 8, 'learning\_rate': 0.044435162453868, 'subsample': 0.9021575082805933, 'colsample\_bytree': 0.6803}, [1] 2025-05-31 02:40:55,159 Trial 5 finished with value: 0.001218074107016061 and parameters: {'n\_estimators': 60, 'max\_depth': 6, 'learning\_rate': 0.10915079530404708, 'subsample': 0.9041575082805933, 'colsample\_bytree': 0.890073}, [1] 2025-05-31 02:40:55,334 Trial 6 finished with value: 0.001214454859921539 and parameters: {'n\_estimators': 51, 'max\_depth': 8, 'learning\_rate': 0.0659337013478378, 'subsample': 0.9277155082805933, 'colsample\_bytree': 0.88149}, [1] 2025-05-31 02:40:55,521 Trial 7 finished with value: 0.001216536019255066 and parameters: {'n\_estimators': 87, 'max\_depth': 7, 'learning\_rate': 0.10236107443026356, 'subsample

[1] 2025-05-31 02:41:01,064 Trial 17 finished with value: 0.001775135703165454 and parameters: {'n\_estimators': 233, 'max\_depth': 6, 'learning\_rate': 0.03390782744726724, 'subsample': 0.7866384679329271, 'colsample\_bytree': 0.077116}

[1] 2025-05-31 02:41:01,185 Trial 18 finished with value: 0.0017849323896872453 and parameters: {'n\_estimators': 52, 'max\_depth': 4, 'learning\_rate': 0.023651266124574875, 'subsample': 0.5630734446897521, 'colsample\_bytree': 0.5372}

[1] 2025-05-31 02:41:01,372 Trial 19 finished with value: 0.00178582736956246 and parameters: {'n\_estimators': 138, 'max\_depth': 8, 'learning\_rate': 0.08232302628611372, 'subsample': 0.6821328785701666, 'colsample\_bytree': 0.77116}

[1] 2025-05-31 02:41:02,046 A new study created in memory with name: no-name\_907BaIdd-734c-4702-b093-5694770d0483

[1] 2025-05-31 02:41:02,272 Trial 0 finished with value: 0.0074481311452576 and parameters: {'n\_estimators': 225, 'max\_depth': 8, 'learning\_rate': 0.0231031825822261, 'subsample': 0.8979351072030299, 'colsample\_bytree': 0.537454}

[1] 2025-05-31 02:41:03,214 Trial 1 finished with value: 0.0073345146290902 and parameters: {'n\_estimators': 170, 'max\_depth': 10, 'learning\_rate': 0.023151488567644005, 'subsample': 0.906958450866387, 'colsample\_bytree': 0.790157}

[1] 2025-05-31 02:41:03,429 Trial 2 finished with value: 0.00745795369336918 and parameters: {'n\_estimators': 299, 'max\_depth': 7, 'learning\_rate': 0.1397455125083823, 'subsample': 0.682167893429766, 'colsample\_bytree': 0.9061675}

[1] 2025-05-31 02:41:03,607 Trial 3 finished with value: 0.00733686224920774 and parameters: {'n\_estimators': 187, 'max\_depth': 6, 'learning\_rate': 0.11327253473035756, 'subsample': 0.9806931615268552, 'colsample\_bytree': 0.563684}

[1] 2025-05-31 02:41:03,794 Trial 4 finished with value: 0.007336872739266468 and parameters: {'n\_estimators': 204, 'max\_depth': 4, 'learning\_rate': 0.028796413889102028, 'subsample': 0.954221849316811, 'colsample\_bytree': 0.89335}

[1] 2025-05-31 02:41:04,041 Trial 5 finished with value: 0.0073236112031801965 and parameters: {'n\_estimators': 249, 'max\_depth': 5, 'learning\_rate': 0.229548519117976, 'subsample': 0.97195376516091, 'colsample\_bytree': 0.7277289}

[1] 2025-05-31 02:41:04,262 Trial 6 finished with value: 0.007330892326638 and parameters: {'n\_estimators': 115, 'max\_depth': 6, 'learning\_rate': 0.04113667655363382, 'subsample': 0.9570405936165636, 'colsample\_bytree': 0.619537}

[1] 2025-05-31 02:41:04,357 Trial 7 finished with value: 0.007340615395694074 and parameters: {'n\_estimators': 139, 'max\_depth': 6, 'learning\_rate': 0.05434160888190304, 'subsample': 0.720154050852829, 'colsample\_bytree': 0.87623}

[1] 2025-05-31 02:41:04,442 Trial 8 finished with value: 0.007338160422071243 and parameters: {'n\_estimators': 123, 'max\_depth': 5, 'learning\_rate': 0.12220491574092852, 'subsample': 0.82171577092554, 'colsample\_bytree': 0.9003914}

[1] 2025-05-31 02:41:04,554 Trial 9 finished with value: 0.0073881726423341 and parameters: {'n\_estimators': 207, 'max\_depth': 3, 'learning\_rate': 0.1005040997950085, 'subsample': 0.71843365390648, 'colsample\_bytree': 0.5033522}

[1] 2025-05-31 02:41:04,636 Trial 10 finished with value: 0.007302783176422911 and parameters: {'n\_estimators': 50, 'max\_depth': 4, 'learning\_rate': 0.267206403842286, 'subsample': 0.5363177007388166, 'colsample\_bytree': 0.6852964}

[1] 2025-05-31 02:41:04,723 Trial 11 finished with value: 0.0073114624965723825 and parameters: {'n\_estimators': 52, 'max\_depth': 4, 'learning\_rate': 0.26814989296092125, 'subsample': 0.505888931593024, 'colsample\_bytree': 0.72943}

[1] 2025-05-31 02:41:04,812 Trial 12 finished with value: 0.007339459307406514 and parameters: {'n\_estimators': 53, 'max\_depth': 3, 'learning\_rate': 0.010750361671723, 'subsample': 0.52190656701128, 'colsample\_bytree': 0.686113}

[1] 2025-05-31 02:41:04,897 Trial 13 finished with value: 0.007312541027410449 and parameters: {'n\_estimators': 51, 'max\_depth': 4, 'learning\_rate': 0.26318698744483073, 'subsample': 0.500146711665757, 'colsample\_bytree': 0.78364013}

[1] 2025-05-31 02:41:04,998 Trial 14 finished with value: 0.007360591353810405 and parameters: {'n\_estimators': 85, 'max\_depth': 4, 'learning\_rate': 0.203310327470473666, 'subsample': 0.6097696242830076, 'colsample\_bytree': 0.698720}

[1] 2025-05-31 02:41:05,094 Trial 15 finished with value: 0.00730562528254509 and parameters: {'n\_estimators': 88, 'max\_depth': 4, 'learning\_rate': 0.281748930842707, 'subsample': 0.580811095956368, 'colsample\_bytree': 0.999710}

[1] 2025-05-31 02:41:05,191 Trial 16 finished with value: 0.0073407117252454 and parameters: {'n\_estimators': 80, 'max\_depth': 3, 'learning\_rate': 0.0777390555087385, 'subsample': 0.6125293236656081, 'colsample\_bytree': 0.631902}

[1] 2025-05-31 02:41:05,308 Trial 17 finished with value: 0.0073348171630843 and parameters: {'n\_estimators': 153, 'max\_depth': 5, 'learning\_rate': 0.1715917161805896, 'subsample': 0.54255259900393, 'colsample\_bytree': 0.823835}

[1] 2025-05-31 02:41:05,396 Trial 18 finished with value: 0.00730605813208882 and parameters: {'n\_estimators': 69, 'max\_depth': 4, 'learning\_rate': 0.074893803693963, 'subsample': 0.6526105795904283, 'colsample\_bytree': 0.747934}

[1] 2025-05-31 02:41:05,495 Trial 19 finished with value: 0.007344881831412962 and parameters: {'n\_estimators': 98, 'max\_depth': 3, 'learning\_rate': 0.07285669195453043, 'subsample': 0.6607892127111871, 'colsample\_bytree': 0.611566}

[1] 2025-05-31 02:41:05,563 A new study created in memory with name: no-name\_6ba81410-b53d-4481-be1f-a43fc87843a2

[1] 2025-05-31 02:41:06,670 Trial 0 finished with value: 0.002850661710544165 and parameters: {'n\_estimators': 126, 'max\_depth': 5, 'learning\_rate': 0.020970928454915876, 'subsample': 0.7799459646251871, 'colsample\_bytree': 0.7123}

[1] 2025-05-31 02:41:06,757 Trial 1 finished with value: 0.002850283829452684 and parameters: {'n\_estimators': 126, 'max\_depth': 6, 'learning\_rate': 0.08611641190905324, 'subsample': 0.8350761657544136, 'colsample\_bytree': 0.553124}

[1] 2025-05-31 02:41:06,832 Trial 2 finished with value: 0.00285163594872932 and parameters: {'n\_estimators': 78, 'max\_depth': 8, 'learning\_rate': 0.16893910902438882, 'subsample': 0.803723282950255, 'colsample\_bytree': 0.541712}

[1] 2025-05-31 02:41:06,943 Trial 3 finished with value: 0.0028364730918431 and parameters: {'n\_estimators': 173, 'max\_depth': 4, 'learning\_rate': 0.01426654481028892, 'subsample': 0.5665230732170409, 'colsample\_bytree': 0.99444}

[1] 2025-05-31 02:41:06,944 Trial 4 finished with value: 0.002820845063420487 and parameters: {'n\_estimators': 151, 'max\_depth': 3, 'learning\_rate': 0.103874512564499, 'subsample': 0.654228675299048, 'colsample\_bytree': 0.73374}

[1] 2025-05-31 02:41:06,147 Trial 5 finished with value: 0.0028371246692246364 and parameters: {'n\_estimators': 187, 'max\_depth': 7, 'learning\_rate': 0.0174749520152742, 'subsample': 0.6096437977167588, 'colsample\_bytree': 0.5846}

[1] 2025-05-31 02:41:06,215 Trial 6 finished with value: 0.002880018816516992 and parameters: {'n\_estimators': 50, 'max\_depth': 6, 'learning\_rate': 0.207735850172785, 'subsample': 0.66190409169721, 'colsample\_bytree': 0.92162351}

[1] 2025-05-31 02:41:06,347 Trial 7 finished with value: 0.002848345879482825 and parameters: {'n\_estimators': 277, 'max\_depth': 6, 'learning\_rate': 0.291501045095046, 'subsample': 0.7303401542346174, 'colsample\_bytree': 0.500287}

[1] 2025-05-31 02:41:06,412 Trial 8 finished with value: 0.002847945801567917 and parameters: {'n\_estimators': 56, 'max\_depth': 6, 'learning\_rate': 0.0478539335619353, 'subsample': 0.97050170238221, 'colsample\_bytree': 0.7214962}

[1] 2025-05-31 02:41:06,474 Trial 9 finished with value: 0.00284826103740161435 and parameters: {'n\_estimators': 53, 'max\_depth': 7, 'learning\_rate': 0.1780727738289102, 'subsample': 0.912510647498249, 'colsample\_bytree': 0.708580}

[1] 2025-05-31 02:41:06,618 Trial 10 finished with value: 0.00285029139793944 and parameters: {'n\_estimators': 234, 'max\_depth': 3, 'learning\_rate': 0.0602225165908645, 'subsample': 0.5366231464829115, 'colsample\_bytree': 0.6274}

[1] 2025-05-31 02:41:06,746 Trial 11 finished with value: 0.0028367916536044127 and parameters: {'n\_estimators': 176, 'max\_depth': 3, 'learning\_rate': 0.011327353153579561, 'subsample': 0.5793300310383726, 'colsample\_bytree': 0.982}

[1] 2025-05-31 02:41:06,852 Trial 12 finished with value: 0.00284658193259041 and parameters: {'n\_estimators': 132, 'max\_depth': 4, 'learning\_rate': 0.10368166069603013, 'subsample': 0.6152394334260856, 'colsample\_bytree': 0.82677}

[1] 2025-05-31 02:41:06,971 Trial 13 finished with value: 0.0028407114074106793 and parameters: {'n\_estimators': 222, 'max\_depth': 4, 'learning\_rate': 0.0908310943909489, 'subsample': 0.5062892021034176, 'colsample\_bytree': 0.81787}

[1] 2025-05-31 02:41:07,155 Trial 14 finished with value: 0.0028302112856154645 and parameters: {'n\_estimators': 157, 'max\_depth': 4, 'learning\_rate': 0.02802246043834433, 'subsample': 0.630904099220192, 'colsample\_bytree': 0.636}

[1] 2025-05-31 02:41:07,272 Trial 15 finished with value: 0.0028389372114325033 and parameters: {'n\_estimators': 148, 'max\_depth': 3, 'learning\_rate': 0.02623845826703542, 'subsample': 0.6556433352873827, 'colsample\_bytree': 0.640}

[1] 2025-05-31 02:41:07,369 Trial 16 finished with value: 0.00287551732852462562 and parameters: {'n\_estimators': 99, 'max\_depth': 4, 'learning\_rate': 0.10116325843904709, 'subsample': 0.7227734560324903, 'colsample\_bytree': 0.618507}

[1] 2025-05-31 02:41:07,504 Trial 17 finished with value: 0.002841148900211184 and parameters: {'n\_estimators': 212, 'max\_depth': 5, 'learning\_rate': 0.03093149764264658, 'subsample': 0.637506794431292, 'colsample\_bytree': 0.648}

[1] 2025-05-31 02:41:07,661 Trial 18 finished with value: 0.002847657643962469 and parameters: {'n\_estimators': 261, 'max\_depth': 3, 'learning\_rate': 0.06231845842492307, 'subsample': 0.859879281617766, 'colsample\_bytree': 0.77428}

[1] 2025-05-31 02:41:07,779 Trial 19 finished with value: 0.002801711150849477 and parameters: {'n\_estimators': 158, 'max\_depth': 5, 'learning\_rate': 0.12727872381287648, 'subsample': 0.5209008794363789, 'colsample\_bytree': 0.6675}

[1] 2025-05-31 02:41:07,881 A new study created in memory with name: no-name\_8e3cf90-f83c-4da6-5bd-5d3d87842782

[1] 2025-05-31 02:41:08,038 Trial 0 finished with value: 0.003031917683383012 and parameters: {'n\_estimators': 224, 'max\_depth': 3, 'learning\_rate': 0.0859841562444975, 'subsample': 0.5953053589407009, 'colsample\_bytree': 0.762081}

[1] 2025-05-31 02:41:08,147 Trial 1 finished with value: 0.00304514596390266 and parameters: {'n\_estimators': 195, 'max\_depth': 6, 'learning\_rate': 0.16977807065731731, 'subsample': 0.5602438119048100777, 'colsample\_bytree': 0.706990}

[1] 2025-05-31 02:41:08,233 Trial 2 finished with value: 0.0030453931935327 and parameters: {'n\_estimators': 124, 'max\_depth': 4, 'learning\_rate': 0.0363049245115615, 'subsample': 0.5894593487687885, 'colsample\_bytree': 0.889481}

[1] 2025-05-31 02:41:08,345 Trial 3 finished with value: 0.00304435174196528786 and parameters: {'n\_estimators': 208, 'max\_depth': 3, 'learning\_rate': 0.1119112897564172, 'subsample': 0.620317198019999, 'colsample\_bytree': 0.94921}

[1] 2025-05-31 02:41:08,448 Trial 4 finished with value: 0.0030658482054738193 and parameters: {'n\_estimators': 187, 'max\_depth': 5, 'learning\_rate': 0.10466136468695764, 'subsample': 0.851277881242477, 'colsample\_bytree': 0.643446}

[1] 2025-05-31 02:41:08,500 Trial 5 finished with value: 0.0030510678892108774 and parameters: {'n\_estimators': 296, 'max\_depth': 8, 'learning\_rate': 0.268702810155327, 'subsample': 0.9993671027898356, 'colsample\_bytree': 0.916941}

[1] 2025-05-31 02:41:08,718 Trial 6 finished with value: 0.00302845679594717 and parameters: {'n\_estimators': 251, 'max\_depth': 8, 'learning\_rate': 0.16534788515334952, 'subsample': 0.7827817578427474, 'colsample\_bytree': 0.84441}

[1] 2025-05-31 02:41:08,857 Trial 7 finished with value: 0.003030285254814447 and parameters: {'n\_estimators': 270, 'max\_depth': 7, 'learning\_rate': 0.0954883181235481, 'subsample': 0.5620736484884333, 'colsample\_bytree': 0.95795}

[1] 2025-05-31 02:41:08,987 Trial 8 finished with value: 0.003030285254814447 and parameters: {'n\_estimators': 256, 'max\_depth': 4, 'learning\_rate': 0.2915010450950499, 'subsample': 0.7874112897793485, 'colsample\_bytree': 0.650402}

[1] 2025-05-31 02:41:09,157 Trial 9 finished with value: 0.0030345719419048443 and parameters: {'n\_estimators': 278, 'max\_depth': 6, 'learning\_rate': 0.10369402913769352, 'subsample': 0.534914951607038, 'colsample\_bytree': 0.542189}

[1] 2025-05-31 02:41:09,245 Trial 10 finished with value: 0.0030705417265426468 and parameters: {'n\_estimators': 64, 'max\_depth': 3, 'learning\_rate': 0.1056297404660055, 'subsample': 0.6546339219734672, 'colsample\_bytree': 0.98796}

[1] 2025-05-31 02:41:09,356 Trial 11 finished with value: 0.003023497104719655 and parameters: {'n\_estimators': 139, 'max\_depth': 7, 'learning\_rate': 0.0367949231973476, 'subsample': 0.5043237864674078, 'colsample\_bytree': 0.2821}

[1] 2025-05-31 02:41:09,490 Trial 12 finished with value: 0.0030510457426659847874 and parameters: {'n\_estimators': 218, 'max\_depth': 7, 'learning\_rate': 0.0243278226760072, 'subsample': 0.6831119996916041, 'colsample\_bytree': 0.8231}

[1] 2025-05-31 02:41:09,636 Trial 13 finished with value: 0.0031334663396046434 and parameters: {'n\_estimators': 237, 'max\_depth': 5, 'learning\_rate': 0.1979819047183224, 'subsample': 0.514159293524725, 'colsample\_bytree': 0.9382}

[1] 2025-05-31 02:41:09,757 Trial 14 finished with value: 0.00302945679594717 and parameters: {'n\_estimators': 148, 'max\_depth': 7, 'learning\_rate': 0.089527933962023, 'subsample': 0.6567926946051591, 'colsample\_bytree': 0.82360}

[1] 2025-05-31 02:41:09,872 Trial 15 finished with value: 0.00303214268403013 and parameters: {'n\_estimators': 136, 'max\_depth': 4, 'learning\_rate': 0.02802246043834433, 'subsample': 0.676176291880293, 'colsample\_bytree': 0.803}

[1] 2025-05-31 02:41:09,976 Trial 16 finished with value: 0.003037723833675388 and parameters: {'n\_estimators': 93, 'max\_depth': 6, 'learning\_rate': 0.032800326812332174, 'subsample': 0.6299644040600709, 'colsample\_bytree': 0.86881}

[1] 2025-05-31 02:41:10,120 Trial 17 finished with value: 0.003046595611728122 and parameters: {'n\_estimators': 162, 'max\_depth': 5, 'learning\_rate': 0.0965559482545678, 'subsample': 0.7368228974075675, 'colsample\_bytree': 0.76866}

[1] 2025-05-31 02:41:10,220 Trial 18 finished with value: 0.003070972172617482 and parameters: {'n\_estimators': 99, 'max\_depth': 7, 'learning\_rate': 0.0243182224431423, 'subsample': 0.6174322224431423, 'colsample\_bytree': 0.707117}

[1] 2025-05-31 02:41:10,342 Trial 19 finished with value: 0.00305554342314911 and parameters: {'n\_estimators': 167, 'max\_depth': 3, 'learning\_rate': 0.01103716396836984, 'subsample': 0.5735003309279941, 'colsample\_bytree': 0.5026}

[1] 2025-05-31 02:41:10,462 A new study created in memory with name: no-name\_8ec39f0-f83c-4da6-5bd-5d3d87842782

[1] 2025-05-31 02:41:10,603 Trial 0 finished with value: 0.00159185821639404 and parameters: {'n\_estimators': 221, 'max\_depth': 4, 'learning\_rate': 0.0410766699313998, 'subsample': 0.9410406276447452, 'colsample\_bytree': 0.5588}

[1] 2025-05-31 02:41:10,682 Trial 1 finished with value: 0.001569923726625807 and parameters: {'n\_estimators': 93, 'max\_depth': 3, 'learning\_rate': 0.11704521396126442, 'subsample': 0.9080149190527165, 'colsample\_bytree': 0.864694}

[1] 2025-05-31 02:41:10,785 Trial 2 finished with value: 0.00155563582781807 and parameters: {'n\_estimators': 178, 'max\_depth': 5, 'learning\_rate': 0.28492196172939, 'subsample': 0.5180495921515615, 'colsample\_bytree': 0.6488860}

[1] 2025-05-31 02:41:10,882 Trial 3 finished with value: 0.0015994891159050075 and parameters: {'n\_estimators': 159, 'max\_depth': 8, 'learning\_rate': 0.1042842434067971, 'subsample': 0.873739320195694, 'colsample\_bytree': 0.9734}

[1] 2025-05-31 02:41:10,945 Trial 4 finished with value: 0.00159130271937267767 and parameters: {'n\_estimators': 52, 'max\_depth': 7, 'learning\_rate': 0.012704103750293455, 'subsample': 0.9053199408727959, 'colsample\_bytree': 0.99023}

[1] 2025-05-31 02:41:11,076 Trial 5 finished with value: 0.0016093481281543 and parameters: {'n\_estimators': 263, 'max\_depth': 3, 'learning\_rate': 0.191981925184315, 'subsample': 0.5831199937158224, 'colsample\_bytree': 0.891877}

[1] 2025-05-31 02:41:12,048 Trial 6 finished with value: 0.001604391316416575 and parameters: {'n\_estimators': 194, 'max\_depth': 3, 'learning\_rate': 0.1559594337537557, 'subsample': 0.502017547928698, 'colsample\_bytree': 0.68689}

[1] 2025-05-31 02:41:12,148 Trial 14 finished with value: 0.001604915401852737, 'subsample': 0.5999909374574299, 'colsample\_bytree': 0.817000

[1] 2025-05-31 02:41:12,298 Trial 15 finished with value: 0.001751111478949102 and parameters: {'n\_estimators': 166, 'max\_depth': 4, 'learning\_rate': 0.0982233816205297, 'subsample': 0.5898374110439282, 'colsample\_bytree': 0.848508}

[1] 2025-05-31 02:41:14,102 Trial 9 finished with value: 0.0037511228379654975 and parameters: {'n\_estimators': 104, 'max\_depth': 6, 'learning\_rate': 0.05660894867308902, 'subsample': 0.40867744292880281, 'colsample\_bytree': 0.59721}

[1] 2025-05-31 02:41:14,144 Trial 10 finished with value: 0.003751138869024251 and parameters: {'n\_estimators': 231, 'max\_depth': 3, 'learning\_rate': 0.010508002902686596, 'subsample': 0.59782281819733, 'colsample\_bytree': 0.5159}

[1] 2025-05-31 02:41:14,394 Trial 11 finished with value: 0.00375123294773738 and parameters: {'n\_estimators': 299, 'max\_depth': 5, 'learning\_rate': 0.025082130941968, 'subsample': 0.6463351674884, 'colsample\_bytree': 0.953510}

[1] 2025-05-31 02:41:14,688 Trial 12 finished with value: 0.0037513348904863 and parameters: {'n\_estimators': 253, 'max\_depth': 4, 'learning\_rate': 0.021472892482204, 'subsample': 0.606034710267199, 'colsample\_bytree': 0.99516}

[1] 2025-05-31 02:41:15,380 Trial 13 finished with value: 0.0037517001955034848 and parameters: {'n\_estimators': 259, 'max\_depth': 7, 'learning\_rate': 0.112389072347216068, 'subsample': 0.5340270847823721, 'colsample\_by

[1] 2025-05-31 02:41:20,668] Trial 4 finished with value: 0.00207985197311616 and parameters: {'n\_estimators': 78, 'max\_depth': 6, 'learning\_rate': 0.273059084584717523, 'subsample': 0.5931284198886785, 'colsample\_bytree': 0.9887895}

[1] 2025-05-31 02:41:20,793] Trial 5 finished with value: 0.00208396246863734 and parameters: {'n\_estimators': 58, 'max\_depth': 7, 'learning\_rate': 0.17819455533727618, 'subsample': 0.6688537921387538, 'colsample\_bytree': 0.6377096}

[1] 2025-05-31 02:41:20,823] Trial 6 finished with value: 0.0020631136988238496 and parameters: {'n\_estimators': 112, 'max\_depth': 3, 'learning\_rate': 0.1422908424243587, 'subsample': 0.5304318091053987, 'colsample\_bytree': 0.91088}

[1] 2025-05-31 02:41:20,829] Trial 7 finished with value: 0.0020965587435915 and parameters: {'n\_estimators': 181, 'max\_depth': 5, 'learning\_rate': 0.03470226146106868, 'subsample': 0.514288161928332, 'colsample\_bytree': 0.908538}

[1] 2025-05-31 02:41:21,005] Trial 8 finished with value: 0.0020576989983717474 and parameters: {'n\_estimators': 88, 'max\_depth': 8, 'learning\_rate': 0.1044320770917189, 'subsample': 0.781312372416467, 'colsample\_bytree': 0.632405}

[1] 2025-05-31 02:41:21,122] Trial 9 finished with value: 0.00203585780734236 and parameters: {'n\_estimators': 228, 'max\_depth': 8, 'learning\_rate': 0.10695028613862062, 'subsample': 0.538263588697525, 'colsample\_bytree': 0.63480}

[1] 2025-05-31 02:41:21,238] Trial 10 finished with value: 0.002061112397302051 and parameters: {'n\_estimators': 148, 'max\_depth': 7, 'learning\_rate': 0.045467041134738755, 'subsample': 0.9728577086725408, 'colsample\_bytree': 0.77788}

[1] 2025-05-31 02:41:21,301] Trial 11 finished with value: 0.002186210047940773 and parameters: {'n\_estimators': 202, 'max\_depth': 8, 'learning\_rate': 0.29783517350630466, 'subsample': 0.6843291958409627, 'colsample\_bytree': 0.635056}

[1] 2025-05-31 02:41:21,530] Trial 12 finished with value: 0.002077617357538787 and parameters: {'n\_estimators': 300, 'max\_depth': 7, 'learning\_rate': 0.0686898641602989, 'subsample': 0.6821794733524689, 'colsample\_bytree': 0.6383}

[1] 2025-05-31 02:41:21,666] Trial 13 finished with value: 0.0020985930094852 and parameters: {'n\_estimators': 213, 'max\_depth': 8, 'learning\_rate': 0.18362506141659027, 'subsample': 0.896187833555976, 'colsample\_bytree': 0.516649}

[1] 2025-05-31 02:41:21,782] Trial 14 finished with value: 0.002053300942474958 and parameters: {'n\_estimators': 129, 'max\_depth': 7, 'learning\_rate': 0.026386955383072317, 'subsample': 0.623088473785348, 'colsample\_bytree': 0.6837}

[1] 2025-05-31 02:41:21,874] Trial 15 finished with value: 0.00204957870138813 and parameters: {'n\_estimators': 58, 'max\_depth': 8, 'learning\_rate': 0.0751797092808981, 'subsample': 0.7397709242127294, 'colsample\_bytree': 0.573069}

[1] 2025-05-31 02:41:22,027] Trial 16 finished with value: 0.00203864499411152 and parameters: {'n\_estimators': 249, 'max\_depth': 6, 'learning\_rate': 0.1285943537347976, 'subsample': 0.58946150105542, 'colsample\_bytree': 0.8139398}

[1] 2025-05-31 02:41:22,187] Trial 17 finished with value: 0.002056844160505106 and parameters: {'n\_estimators': 282, 'max\_depth': 4, 'learning\_rate': 0.21094715141743312, 'subsample': 0.859343123301899, 'colsample\_bytree': 0.69601}

[1] 2025-05-31 02:41:22,323] Trial 18 finished with value: 0.00208705105265506 and parameters: {'n\_estimators': 216, 'max\_depth': 7, 'learning\_rate': 0.0520309049506631, 'subsample': 0.662202273661778, 'colsample\_bytree': 0.57867}

[1] 2025-05-31 02:41:22,444] Trial 19 finished with value: 0.002104521849092803 and parameters: {'n\_estimators': 172, 'max\_depth': 8, 'learning\_rate': 0.08861545809121466, 'subsample': 0.738949215849094, 'colsample\_bytree': 0.7069}

[1] 2025-05-31 02:41:22,516] A new study created in memory with name: no-name-a=89495450-7674-40e5-9470-50e61444778f

[1] 2025-05-31 02:41:22,686] Trial 0 finished with value: 0.003085382313423442 and parameters: {'n\_estimators': 190, 'max\_depth': 8, 'learning\_rate': 0.020592616822150838, 'subsample': 0.7568515772439507, 'colsample\_bytree': 0.9647}

[1] 2025-05-31 02:41:22,808] Trial 1 finished with value: 0.003071095879951629 and parameters: {'n\_estimators': 234, 'max\_depth': 3, 'learning\_rate': 0.03176675101635239, 'subsample': 0.7494287758778613, 'colsample\_bytree': 0.506348}

[1] 2025-05-31 02:41:22,895] Trial 2 finished with value: 0.0030483906499268 and parameters: {'n\_estimators': 139, 'max\_depth': 4, 'learning\_rate': 0.11617326568770004, 'subsample': 0.681918618863644, 'colsample\_bytree': 0.520334}

[1] 2025-05-31 02:41:23,023] Trial 3 finished with value: 0.003064642953757893 and parameters: {'n\_estimators': 253, 'max\_depth': 7, 'learning\_rate': 0.01230687641628878, 'subsample': 0.6190732807871243, 'colsample\_bytree': 0.5085}

[1] 2025-05-31 02:41:23,132] Trial 4 finished with value: 0.00307952675439943 and parameters: {'n\_estimators': 204, 'max\_depth': 6, 'learning\_rate': 0.0468709763727201, 'subsample': 0.9709606587348961, 'colsample\_bytree': 0.82866}

[1] 2025-05-31 02:41:23,300] Trial 5 finished with value: 0.0030798175847616567 and parameters: {'n\_estimators': 194, 'max\_depth': 3, 'learning\_rate': 0.0286855918772176, 'subsample': 0.6951746463065821, 'colsample\_bytree': 0.71882}

[1] 2025-05-31 02:41:23,379] Trial 6 finished with value: 0.0020848903053674 and parameters: {'n\_estimators': 88, 'max\_depth': 8, 'learning\_rate': 0.0378002974619085, 'subsample': 0.89569058267393, 'colsample\_bytree': 0.8894190}

[1] 2025-05-31 02:41:23,509] Trial 7 finished with value: 0.0030667739465251284 and parameters: {'n\_estimators': 250, 'max\_depth': 8, 'learning\_rate': 0.011757672454576509, 'subsample': 0.519926371123558, 'colsample\_bytree': 0.8721}

[1] 2025-05-31 02:41:23,634] Trial 8 finished with value: 0.003073065325455438 and parameters: {'n\_estimators': 201, 'max\_depth': 8, 'learning\_rate': 0.1411248605017906, 'subsample': 0.787841261993692, 'colsample\_bytree': 0.705822}

[1] 2025-05-31 02:41:23,740] Trial 9 finished with value: 0.00309438501333883 and parameters: {'n\_estimators': 162, 'max\_depth': 4, 'learning\_rate': 0.2872320017504191, 'subsample': 0.65023574851279, 'colsample\_bytree': 0.8069015}

[1] 2025-05-31 02:41:23,843] Trial 10 finished with value: 0.003100843779281395 and parameters: {'n\_estimators': 111, 'max\_depth': 5, 'learning\_rate': 0.107093262417695, 'subsample': 0.538692448667343, 'colsample\_bytree': 0.62659}

[1] 2025-05-31 02:41:24,002] Trial 11 finished with value: 0.0030810824177670925 and parameters: {'n\_estimators': 294, 'max\_depth': 6, 'learning\_rate': 0.0885171170172436, 'subsample': 0.623120409404144, 'colsample\_bytree': 0.5003}

[1] 2025-05-31 02:41:24,111] Trial 12 finished with value: 0.00307981537469445 and parameters: {'n\_estimators': 135, 'max\_depth': 5, 'learning\_rate': 0.01665050728083936, 'subsample': 0.8592064632327613, 'colsample\_bytree': 0.599}

[1] 2025-05-31 02:41:24,198] Trial 13 finished with value: 0.003059328276579863 and parameters: {'n\_estimators': 66, 'max\_depth': 7, 'learning\_rate': 0.2105793451376972, 'subsample': 0.60867744919531, 'colsample\_bytree': 0.58567}

[1] 2025-05-31 02:41:24,367] Trial 14 finished with value: 0.003079172068958683 and parameters: {'n\_estimators': 289, 'max\_depth': 4, 'learning\_rate': 0.06662417468655092, 'subsample': 0.8599926862418799, 'colsample\_bytree': 0.65944}

[1] 2025-05-31 02:41:24,492] Trial 15 finished with value: 0.0030695479716424015 and parameters: {'n\_estimators': 153, 'max\_depth': 7, 'learning\_rate': 0.01871909498891502, 'subsample': 0.706144051094931, 'colsample\_bytree': 0.553}

[1] 2025-05-31 02:41:24,653] Trial 16 finished with value: 0.003071613127317224 and parameters: {'n\_estimators': 249, 'max\_depth': 4, 'learning\_rate': 0.0638201683288531, 'subsample': 0.579900307203285, 'colsample\_bytree': 0.5917}

[1] 2025-05-31 02:41:24,854] Trial 17 finished with value: 0.00309784756780327547 and parameters: {'n\_estimators': 122, 'max\_depth': 7, 'learning\_rate': 0.160078097912543, 'subsample': 0.678685036426853, 'colsample\_bytree': 0.658972}

[1] 2025-05-31 02:41:24,994] Trial 18 finished with value: 0.00302521548538621 and parameters: {'n\_estimators': 53, 'max\_depth': 5, 'learning\_rate': 0.0192150532475842, 'subsample': 0.8053047472845813, 'colsample\_bytree': 0.76919}

[1] 2025-05-31 02:41:25,126] A new study created in memory with name: no-name-a=1115cd7d-5f1e-4809-9e5d-f812cadfb66

[1] 2025-05-31 02:41:25,264] Trial 0 finished with value: 0.00346877181054991 and parameters: {'n\_estimators': 183, 'max\_depth': 7, 'learning\_rate': 0.21995532310579794, 'subsample': 0.909337650091257, 'colsample\_bytree': 0.6753616}

[1] 2025-05-31 02:41:25,346] Trial 1 finished with value: 0.003412708974005161 and parameters: {'n\_estimators': 118, 'max\_depth': 7, 'learning\_rate': 0.1400420806245254, 'subsample': 0.7711572592430745, 'colsample\_bytree': 0.694614}

[1] 2025-05-31 02:41:25,408] Trial 2 finished with value: 0.0034507819835242906 and parameters: {'n\_estimators': 54, 'max\_depth': 7, 'learning\_rate': 0.03425242535541171, 'subsample': 0.633668871799925, 'colsample\_bytree': 0.715624}

[1] 2025-05-31 02:41:25,549] Trial 3 finished with value: 0.003913567279237208 and parameters: {'n\_estimators': 287, 'max\_depth': 3, 'learning\_rate': 0.0322951799093232, 'subsample': 0.5817871607417322, 'colsample\_bytree': 0.78053}

[1] 2025-05-31 02:41:25,684] Trial 4 finished with value: 0.003431470941770828 and parameters: {'n\_estimators': 238, 'max\_depth': 6, 'learning\_rate': 0.145024636136698, 'subsample': 0.7824411824022257, 'colsample\_bytree': 0.99720}

[1] 2025-05-31 02:41:25,819] Trial 5 finished with value: 0.00344180912343435 and parameters: {'n\_estimators': 236, 'max\_depth': 3, 'learning\_rate': 0.01656317764176101, 'subsample': 0.9874219150379192, 'colsample\_bytree': 0.889782}

[1] 2025-05-31 02:41:25,940] Trial 6 finished with value: 0.00344302482735183 and parameters: {'n\_estimators': 215, 'max\_depth': 8, 'learning\_rate': 0.01150487097302171, 'subsample': 0.8263697088221655, 'colsample\_bytree': 0.9944}

[1] 2025-05-31 02:41:26,069] Trial 7 finished with value: 0.003369667305295928 and parameters: {'n\_estimators': 259, 'max\_depth': 6, 'learning\_rate': 0.1376796231507549, 'subsample': 0.85992313262521, 'colsample\_bytree': 0.677707}

[1] 2025-05-31 02:41:26,196] Trial 8 finished with value: 0.00348861851397733 and parameters: {'n\_estimators': 236, 'max\_depth': 7, 'learning\_rate': 0.1816962963936597, 'subsample': 0.930920245377207, 'colsample\_bytree': 0.97593}

[1] 2025-05-31 02:41:26,269] Trial 9 finished with value: 0.0034365110264464392 and parameters: {'n\_estimators': 79, 'max\_depth': 3, 'learning\_rate': 0.02954327577831342, 'subsample': 0.9792079488303104, 'colsample\_bytree': 0.65972}

[1] 2025-05-31 02:41:26,427] Trial 10 finished with value: 0.003394163767075585 and parameters: {'n\_estimators': 292, 'max\_depth': 5, 'learning\_rate': 0.07699625158599459, 'subsample': 0.6600658960410213, 'colsample\_bytree': 0.50450}

[1] 2025-05-31 02:41:26,589] Trial 11 finished with value: 0.0033598237272359794 and parameters: {'n\_estimators': 300, 'max\_depth': 5, 'learning\_rate': 0.071647377320423408, 'subsample': 0.547612391983431, 'colsample\_bytree': 0.81026}

[1] 2025-05-31 02:41:26,700] Trial 12 finished with value: 0.00350656159833715 and parameters: {'n\_estimators': 292, 'max\_depth': 5, 'learning\_rate': 0.074594187974931, 'subsample': 0.5425358646588032, 'colsample\_bytree': 0.8107906}

[1] 2025-05-31 02:41:26,962] Trial 13 finished with value: 0.003355574450594553 and parameters: {'n\_estimators': 296, 'max\_depth': 5, 'learning\_rate': 0.06291545736635074, 'subsample': 0.529147889490583, 'colsample\_bytree': 0.8208}

[1] 2025-05-31 02:41:27,086] Trial 14 finished with value: 0.003369673052959284 and parameters: {'n\_estimators': 259, 'max\_depth': 6, 'learning\_rate': 0.1376796231507549, 'subsample': 0.85992313262521, 'colsample\_bytree': 0.677707}

[1] 2025-05-31 02:41:27,207] Trial 15 finished with value: 0.003081828138303606 and parameters: {'n\_estimators': 162, 'max\_depth': 4, 'learning\_rate': 0.048000863818575094, 'subsample': 0.605390965275644, 'colsample\_bytree': 0.8941}

[1] 2025-05-31 02:41:27,321] Trial 16 finished with value: 0.003422118026328868 and parameters: {'n\_estimators': 146, 'max\_depth': 4, 'learning\_rate': 0.048000863818575094, 'subsample': 0.5159651764325289, 'colsample\_bytree': 0.87646}

[1] 2025-05-31 02:41:27,472] Trial 17 finished with value: 0.003473297662534525 and parameters: {'n\_estimators': 189, 'max\_depth': 4, 'learning\_rate': 0.2834297689721203, 'subsample': 0.6935571983104683, 'colsample\_bytree': 0.58671}

[1] 2025-05-31 02:41:27,642] Trial 18 finished with value: 0.003479373355379586 and parameters: {'n\_estimators': 123, 'max\_depth': 4, 'learning\_rate': 0.04485217507814753, 'subsample': 0.5035650819700303, 'colsample\_bytree': 0.850514}

[1] 2025-05-31 02:41:28,389] Trial 19 finished with value: 0.003498735649761049 and parameters: {'n\_estimators': 209, 'max\_depth': 6, 'learning\_rate': 0.0494036644992992, 'subsample': 0.5767558615080393, 'colsample\_bytree': 0.94065}

[1] 2025-05-31 02:41:28,597] A new study created in memory with name: no-name=c47c5ba-2f4f-4960-ba7a-24dc1a6e200c

[1] 2025-05-31 02:41:28,795] Trial 0 finished with value: 0.00407883316318151 and parameters: {'n\_estimators': 99, 'max\_depth': 4, 'learning\_rate': 0.2934530412930986, 'subsample': 0.5397030819828989, 'colsample\_bytree': 0.5830309}

[1] 2025-05-31 02:41:29,167] Trial 1 finished with value: 0.004092158919266583 and parameters: {'n\_estimators': 53, 'max\_depth': 6, 'learning\_rate': 0.05731257148989209, 'subsample': 0.73108029715929, 'colsample\_bytree': 0.5932687}

[1] 2025-05-31 02:41:29,676] Trial 2 finished with value: 0.00409624625251772 and parameters: {'n\_estimators': 160, 'max\_depth': 8, 'learning\_rate': 0.02175251365054543, 'subsample': 0.5918754697189574, 'colsample\_bytree': 0.921091}

[1] 2025-05-31 02:41:29,980] Trial 3 finished with value: 0.003913567279237208 and parameters: {'n\_estimators': 287, 'max\_depth': 3, 'learning\_rate': 0.0322951799093232, 'subsample': 0.5817871607417322, 'colsample\_bytree': 0.78053}

[1] 2025-05-31 02:41:30,129] Trial 4 finished with value: 0.003882241042997065 and parameters: {'n\_estimators': 175, 'max\_depth': 3, 'learning\_rate': 0.03491528352649624, 'subsample': 0.86241045204028448, 'colsample\_bytree': 0.92656}

[1] 2025-05-31 02:41:30,320] Trial 5 finished with value: 0.0030831392061262 and parameters: {'n\_estimators': 242, 'max\_depth': 8, 'learning\_rate': 0.01194208003018987, 'subsample': 0.5123396732904476, 'colsample\_bytree': 0.5393040}

[1] 2025-05-31 02:41:30,502] Trial 6 finished with value: 0.0034883106466166135 and parameters: {'n\_estimators': 208, 'max\_depth': 6, 'learning\_rate': 0.0208804616406884343, 'subsample': 0.9277389608746097, 'colsample\_bytree': 0.63190}

[1] 2025-05-31 02:41:30,685] Trial 7 finished with value: 0.003489101057100927 and parameters: {'n\_estimators': 262, 'max\_depth': 7, 'learning\_rate': 0.0407724937575458, 'subsample': 0.93136868734979, 'colsample\_bytree': 0.9970815}

[1] 2025-05-31 02:41:30,819] Trial 8 finished with value: 0.00349812105089121 and parameters: {'n\_estimators': 271, 'max\_depth': 7, 'learning\_rate': 0.0668360245913459, 'subsample': 0.8301703302440182, 'colsample\_bytree': 0.5947723}

[1] 2025-05-31 02:41:31,002] Trial 9 finished with value: 0.003488524688688584 and parameters: {'n\_estimators': 270, 'max\_depth': 5, 'learning\_rate': 0.045364106342162, 'subsample': 0.7828735669597, 'colsample\_bytree': 0.8885673}

[1] 2025-05-31 02:41:31,244] Trial 10 finished with value: 0.00344861851397733 and parameters: {'n\_estimators': 227, 'max\_depth': 5, 'learning\_rate': 0.0477230732887708, 'subsample': 0.61424814987306, 'colsample\_bytree': 0.50231}

[1] 2025-05-31 02:41:31,260] Trial 11 finished with value: 0.0040391320623221423 and parameters: {'n\_estimators': 147, 'max\_depth': 3, 'learning\_rate': 0.0114953024520768, 'subsample': 0.8543507477932107, 'colsample\_bytree': 0.7849}

[1] 2025-05-31 02:41:31,421] Trial 12 finished with value: 0.004037329352635694 and parameters: {'n\_estimators': 300, 'max\_depth': 3, 'learning\_rate': 0.02778018466359436, 'subsample': 0.9593237707101744, 'colsample\_bytree': 0.788}

[1] 2025-05-31 02:41:31,593] Trial 13 finished with value: 0.003875784686234224 and parameters: {'n\_estimators': 293, 'max\_depth': 4, 'learning\_rate': 0.01840743271507549, 'subsample': 0.9799497097898097, 'colsample\_bytree': 0.6801}

[1] 2025-05-31 02:41:32,058] Trial 14 finished with value: 0.004385094880584069 and parameters: {'n\_estimators': 239, 'max\_depth': 7, 'learning\_rate': 0.1124227195319694, 'subsample': 0.7961384877110326, 'colsample\_bytree': 0.69291}

[1] 2025-05-31 02:41:32,187] Trial 15 finished with value: 0.00404028263446252 and parameters: {'n\_estimators': 185, 'max\_depth': 7, 'learning\_rate': 0.1465580820135485, 'subsample': 0.6056759740246334, 'colsample\_bytree': 0.70219}

[1] 2025-05-31 02:41:32,333] Trial 16 finished with value: 0.004396564078170597 and parameters: {'n\_estimators': 240, 'max\_depth': 7, 'learning\_rate': 0.1466778964592005, 'subsample': 0.828970363043638, 'colsample\_bytree': 0.528209}

[1] 2025-05-31 02:41:32,489] Trial 17 finished with value: 0.00438975251991994 and parameters: {'n\_estimators': 143, 'max\_depth': 8, 'learning\_rate': 0.1172243123483517, 'subsample': 0.509932057136844, 'colsample\_bytree': 0.63912}

[1] 2025-05-31 02:41:32,701] Trial 0 finished with value: 0.00173037483123 and parameters: {'n\_estimators': 170, 'max\_depth': 4, 'learning\_rate': 0.25379908502422183, 'subsample': 0.6457475178125072, 'colsample\_bytree': 0.51904}

[1] 2025-05-31 02:41:32,836] Trial 1 finished with value: 0.00169395235646825 and parameters: {'n\_estimators': 232, 'max\_depth': 6, 'learning\_rate': 0.1540962814135293, 'subsample': 0.54049062814135293, 'colsample\_bytree': 0.79449}

[1] 2025-05-31 02:41:33,944] Trial 2 finished with value: 0.00171467367059286 and parameters: {'n\_estimators': 178, 'max\_depth': 3, 'learning\_rate': 0.2067321821908538, 'subsample': 0.809958757431085, 'colsample\_bytree': 0.87011}

[1] 2025-05-31 02:41:33,050] Trial 3 finished with value: 0.0017040958498703954 and parameters: {'n\_estimators': 119, 'max\_depth': 7, 'learning\_rate': 0.

[1] 2025-05-31 02:41:59,700 Trial 11 finished with value: 0.0010735991714657776 and parameters: {'n\_estimators': 295, 'max\_depth': 6, 'learning\_rate': 0.0897152737008296359, 'subsample': 0.910780076306914, 'colsample\_bytree': 0.4015},  
 [1] 2025-05-31 02:41:39,305 Trial 13 finished with value: 0.001038181178041685 and parameters: {'n\_estimators': 300, 'max\_depth': 6, 'learning\_rate': 0.104912902235818019, 'subsample': 0.5039255163610166, 'colsample\_bytree': 0.7991},  
 [1] 2025-05-31 02:41:39,455 Trial 13 finished with value: 0.00103814702885621985 and parameters: {'n\_estimators': 248, 'max\_depth': 7, 'learning\_rate': 0.11768203379873761, 'subsample': 0.510845233890438, 'colsample\_bytree': 0.7990},  
 [1] 2025-05-31 02:41:39,596 Trial 14 finished with value: 0.00104943899989026062 and parameters: {'n\_estimators': 229, 'max\_depth': 8, 'learning\_rate': 0.10692564871735852, 'subsample': 0.5198973656655291, 'colsample\_bytree': 0.7540},  
 [1] 2025-05-31 02:41:39,739 Trial 15 finished with value: 0.001047031264574941 and parameters: {'n\_estimators': 217, 'max\_depth': 7, 'learning\_rate': 0.07994546349069556, 'subsample': 0.597199708117073, 'colsample\_bytree': 0.880238},  
 [1] 2025-05-31 02:41:39,897 Trial 16 finished with value: 0.0010578472442001846 and parameters: {'n\_estimators': 261, 'max\_depth': 5, 'learning\_rate': 0.1517718580407697, 'subsample': 0.5633454818171972, 'colsample\_bytree': 0.8672},  
 [1] 2025-05-31 02:41:40,050 Trial 17 finished with value: 0.0010597567293050524 and parameters: {'n\_estimators': 269, 'max\_depth': 7, 'learning\_rate': 0.1369333108500232, 'subsample': 0.6330108450673574, 'colsample\_bytree': 0.71330},  
 [1] 2025-05-31 02:41:40,183 Trial 18 finished with value: 0.001036905090704039 and parameters: {'n\_estimators': 196, 'max\_depth': 8, 'learning\_rate': 0.07233816327197497, 'subsample': 0.50189698050221, 'colsample\_bytree': 0.9554},  
 [1] 2025-05-31 02:41:40,345 Trial 19 finished with value: 0.001044024061746899 and parameters: {'n\_estimators': 236, 'max\_depth': 6, 'learning\_rate': 0.07233816327197497, 'subsample': 0.654634040293387, 'colsample\_bytree': 0.8137},  
 [1] 2025-05-31 02:41:40,494 A new study created in memory with name: no-name-8f26553d-af03-42d9-a29-36a7fc3b9d1  
 [1] 2025-05-31 02:41:40,634 Trial 0 finished with value: 0.001188161834805099 and parameters: {'n\_estimators': 71, 'max\_depth': 8, 'learning\_rate': 0.274896891890025, 'subsample': 0.92088078992651, 'colsample\_bytree': 0.60542966},  
 [1] 2025-05-31 02:41:40,753 Trial 1 finished with value: 0.001188401765082864 and parameters: {'n\_estimators': 185, 'max\_depth': 4, 'learning\_rate': 0.029363747052021074, 'subsample': 0.51651037958190472, 'colsample\_bytree': 0.8629},  
 [1] 2025-05-31 02:41:40,983 Trial 2 finished with value: 0.0011406866226546472 and parameters: {'n\_estimators': 239, 'max\_depth': 5, 'learning\_rate': 0.05305549748190922, 'subsample': 0.5685888921044928, 'colsample\_bytree': 0.87887},  
 [1] 2025-05-31 02:41:41,661 Trial 3 finished with value: 0.001186620394283085 and parameters: {'n\_estimators': 158, 'max\_depth': 7, 'learning\_rate': 0.1503474869743065, 'subsample': 0.669599505027865, 'colsample\_bytree': 0.92461},  
 [1] 2025-05-31 02:41:41,887 Trial 4 finished with value: 0.00113030775054313 and parameters: {'n\_estimators': 238, 'max\_depth': 8, 'learning\_rate': 0.0140632475159342144, 'subsample': 0.79430169314815, 'colsample\_bytree': 0.540819},  
 [1] 2025-05-31 02:41:42,836 Trial 5 finished with value: 0.0011873005694382 and parameters: {'n\_estimators': 194, 'max\_depth': 3, 'learning\_rate': 0.02729870372699134, 'subsample': 0.9643821419935785, 'colsample\_bytree': 0.80768},  
 [1] 2025-05-31 02:41:43,078 Trial 6 finished with value: 0.00115304615657812 and parameters: {'n\_estimators': 228, 'max\_depth': 3, 'learning\_rate': 0.01318624393808783, 'subsample': 0.556692196792955, 'colsample\_bytree': 0.50027},  
 [1] 2025-05-31 02:41:43,287 Trial 7 finished with value: 0.00114733810334803 and parameters: {'n\_estimators': 283, 'max\_depth': 8, 'learning\_rate': 0.11412865903287798, 'subsample': 0.7670484293800325, 'colsample\_bytree': 0.67452},  
 [1] 2025-05-31 02:41:43,485 Trial 8 finished with value: 0.001131981780701931 and parameters: {'n\_estimators': 211, 'max\_depth': 7, 'learning\_rate': 0.0102126718451615, 'subsample': 0.80286429830463, 'colsample\_bytree': 0.8120},  
 [1] 2025-05-31 02:41:43,699 Trial 9 finished with value: 0.001126368102358903 and parameters: {'n\_estimators': 130, 'max\_depth': 4, 'learning\_rate': 0.01735836217558596, 'subsample': 0.6723574170274549, 'colsample\_bytree': 0.73230},  
 [1] 2025-05-31 02:41:43,920 Trial 10 finished with value: 0.0010872538702086788 and parameters: {'n\_estimators': 300, 'max\_depth': 6, 'learning\_rate': 0.0917078403525727, 'subsample': 0.88639614402228, 'colsample\_bytree': 0.6833},  
 [1] 2025-05-31 02:41:44,085 Trial 11 finished with value: 0.001093299208125426 and parameters: {'n\_estimators': 296, 'max\_depth': 6, 'learning\_rate': 0.08682450463322971, 'subsample': 0.8734033235949185, 'colsample\_bytree': 0.6724},  
 [1] 2025-05-31 02:41:44,240 Trial 12 finished with value: 0.001105169479136363 and parameters: {'n\_estimators': 282, 'max\_depth': 6, 'learning\_rate': 0.075239845946254358, 'subsample': 0.8837418644629303, 'colsample\_bytree': 0.6714},  
 [1] 2025-05-31 02:41:44,405 Trial 13 finished with value: 0.001092950953970883 and parameters: {'n\_estimators': 294, 'max\_depth': 6, 'learning\_rate': 0.0610213549121504, 'subsample': 0.8635788475694054, 'colsample\_bytree': 0.6467},  
 [1] 2025-05-31 02:41:45,570 Trial 14 finished with value: 0.00112574250362612 and parameters: {'n\_estimators': 260, 'max\_depth': 5, 'learning\_rate': 0.03862629047559566, 'subsample': 0.998805021120161, 'colsample\_bytree': 0.559},  
 [1] 2025-05-31 02:41:45,685 Trial 15 finished with value: 0.00115275158790568 and parameters: {'n\_estimators': 99, 'max\_depth': 6, 'learning\_rate': 0.05693547897623614, 'subsample': 0.7148864851720502, 'colsample\_bytree': 0.744658},  
 [1] 2025-05-31 02:41:45,835 Trial 16 finished with value: 0.001073736544384581 and parameters: {'n\_estimators': 299, 'max\_depth': 7, 'learning\_rate': 0.25979418232973, 'subsample': 0.8808636864036395, 'colsample\_bytree': 0.61103},  
 [1] 2025-05-31 02:41:44,986 Trial 17 finished with value: 0.001082473660846154 and parameters: {'n\_estimators': 259, 'max\_depth': 7, 'learning\_rate': 0.26512728102120123, 'subsample': 0.9324029573861295, 'colsample\_bytree': 0.557039},  
 [1] 2025-05-31 02:41:45,133 Trial 18 finished with value: 0.001130719807050856 and parameters: {'n\_estimators': 253, 'max\_depth': 7, 'learning\_rate': 0.248198767451752, 'subsample': 0.93404674378484364, 'colsample\_bytree': 0.568856},  
 [1] 2025-05-31 02:41:45,292 Trial 19 finished with value: 0.00119750300264123 and parameters: {'n\_estimators': 266, 'max\_depth': 7, 'learning\_rate': 0.1756857843805724, 'subsample': 0.818604616141481, 'colsample\_bytree': 0.9842},  
 [1] 2025-05-31 02:41:45,451 A new study created in memory with name: no-name-7470dec0-bb71-4a19-844d-19684956df0  
 [1] 2025-05-31 02:41:45,570 Trial 0 finished with value: 0.00070628299713747 and parameters: {'n\_estimators': 152, 'max\_depth': 4, 'learning\_rate': 0.129668008337125, 'subsample': 0.8849795543141491, 'colsample\_bytree': 0.98092},  
 [1] 2025-05-31 02:41:45,695 Trial 1 finished with value: 0.0006896256098156369 and parameters: {'n\_estimators': 214, 'max\_depth': 4, 'learning\_rate': 0.030507430639267905, 'subsample': 0.636397842678782, 'colsample\_bytree': 0.9281},  
 [1] 2025-05-31 02:41:45,774 Trial 2 finished with value: 0.000690327153257331 and parameters: {'n\_estimators': 110, 'max\_depth': 5, 'learning\_rate': 0.024681723947196483, 'subsample': 0.6946601631945879, 'colsample\_bytree': 0.9490},  
 [1] 2025-05-31 02:41:45,842 Trial 3 finished with value: 0.000707176960951126 and parameters: {'n\_estimators': 60, 'max\_depth': 5, 'learning\_rate': 0.0184509475522696, 'subsample': 0.78782513754924, 'colsample\_bytree': 0.9303164},  
 [1] 2025-05-31 02:41:45,947 Trial 4 finished with value: 0.0007046879243815732 and parameters: {'n\_estimators': 188, 'max\_depth': 6, 'learning\_rate': 0.1603281896812196, 'subsample': 0.6494949663569584, 'colsample\_bytree': 0.7160},  
 [1] 2025-05-31 02:41:46,043 Trial 5 finished with value: 0.000684182975905874 and parameters: {'n\_estimators': 118, 'max\_depth': 8, 'learning\_rate': 0.0904800402583217, 'subsample': 0.637510912977422, 'colsample\_bytree': 0.5714260},  
 [1] 2025-05-31 02:41:46,172 Trial 6 finished with value: 0.000689004015997238 and parameters: {'n\_estimators': 251, 'max\_depth': 5, 'learning\_rate': 0.039780192337766127, 'subsample': 0.8169573632516934, 'colsample\_bytree': 0.9817},  
 [1] 2025-05-31 02:41:46,291 Trial 7 finished with value: 0.000708438711238959 and parameters: {'n\_estimators': 227, 'max\_depth': 5, 'learning\_rate': 0.0933022053630674, 'subsample': 0.898185116097297, 'colsample\_bytree': 0.734413},  
 [1] 2025-05-31 02:41:46,397 Trial 8 finished with value: 0.00070839013816627952 and parameters: {'n\_estimators': 177, 'max\_depth': 7, 'learning\_rate': 0.010112503486201915, 'subsample': 0.859432099210929, 'colsample\_bytree': 0.9160},  
 [1] 2025-05-31 02:41:46,460 Trial 9 finished with value: 0.000743249425926924 and parameters: {'n\_estimators': 54, 'max\_depth': 3, 'learning\_rate': 0.0148001321321015, 'subsample': 0.780374327993489, 'colsample\_bytree': 0.9265034},  
 [1] 2025-05-31 02:41:46,619 Trial 10 finished with value: 0.000809642715652205 and parameters: {'n\_estimators': 294, 'max\_depth': 8, 'learning\_rate': 0.28167096008142783, 'subsample': 0.5124651247782639, 'colsample\_bytree': 0.50211},  
 [1] 2025-05-31 02:41:46,782 Trial 11 finished with value: 0.000693914022298914 and parameters: {'n\_estimators': 285, 'max\_depth': 8, 'learning\_rate': 0.0553165145701761, 'subsample': 0.562322069040421, 'colsample\_bytree': 0.5902},  
 [1] 2025-05-31 02:41:46,894 Trial 12 finished with value: 0.00069011778095362959 and parameters: {'n\_estimators': 118, 'max\_depth': 7, 'learning\_rate': 0.04839816010249484, 'subsample': 0.7179509840653941, 'colsample\_bytree': 0.8316},  
 [1] 2025-05-31 02:41:47,047 Trial 13 finished with value: 0.00072548059123043298 and parameters: {'n\_estimators': 255, 'max\_depth': 6, 'learning\_rate': 0.0458110602410321, 'subsample': 0.853448393256089, 'colsample\_bytree': 0.6645},  
 [1] 2025-05-31 02:41:47,159 Trial 14 finished with value: 0.00072425709123036279 and parameters: {'n\_estimators': 119, 'max\_depth': 7, 'learning\_rate': 0.04257051157195471, 'subsample': 0.613397678868588, 'colsample\_bytree': 0.805886},  
 [1] 2025-05-31 02:41:47,269 Trial 15 finished with value: 0.0006203825923681434 and parameters: {'n\_estimators': 105, 'max\_depth': 7, 'learning\_rate': 0.0389737895628876, 'subsample': 0.5853159393186043, 'colsample\_bytree': 0.82197},  
 [1] 2025-05-31 02:41:47,367 Trial 16 finished with value: 0.00068632676168847359 and parameters: {'n\_estimators': 90, 'max\_depth': 7, 'learning\_rate': 0.15999025166847355, 'subsample': 0.5604668183019401, 'colsample\_bytree': 0.82877},  
 [1] 2025-05-31 02:41:47,494 Trial 17 finished with value: 0.000684104617304236 and parameters: {'n\_estimators': 144, 'max\_depth': 7, 'learning\_rate': 0.0483981546794501, 'subsample': 0.575371078556291, 'colsample\_bytree': 0.8318},  
 [1] 2025-05-31 02:41:48,311 Trial 18 finished with value: 0.00063593378973501 and parameters: {'n\_estimators': 145, 'max\_depth': 7, 'learning\_rate': 0.14266083952998356, 'subsample': 0.6887882841513963, 'colsample\_bytree': 0.771368},  
 [1] 2025-05-31 02:41:48,448 A new study created in memory with name: no-name-14338e08-11c4-423c-bf39-934c88fb8e8f  
 [1] 2025-05-31 02:41:49,538 Trial 0 finished with value: 0.001499670900901704933954 and parameters: {'n\_estimators': 275, 'max\_depth': 4, 'learning\_rate': 0.028158701549576018, 'subsample': 0.814172990611501, 'colsample\_bytree': 0.6879},  
 [1] 2025-05-31 02:41:49,839 Trial 1 finished with value: 0.0014985081027026799 and parameters: {'n\_estimators': 258, 'max\_depth': 4, 'learning\_rate': 0.02377235489920207, 'subsample': 0.7670524657837878, 'colsample\_bytree': 0.5423},  
 [1] 2025-05-31 02:41:50,058 Trial 2 finished with value: 0.0014998801920472704 and parameters: {'n\_estimators': 218, 'max\_depth': 6, 'learning\_rate': 0.01117459567086271, 'subsample': 0.8873891674993937, 'colsample\_bytree': 0.9250},  
 [1] 2025-05-31 02:41:50,209 Trial 3 finished with value: 0.001513948171434737 and parameters: {'n\_estimators': 149, 'max\_depth': 8, 'learning\_rate': 0.1093909540190494, 'subsample': 0.67063896069630401, 'colsample\_bytree': 0.94172},  
 [1] 2025-05-31 02:41:50,406 Trial 4 finished with value: 0.00149232125687893 and parameters: {'n\_estimators': 295, 'max\_depth': 3, 'learning\_rate': 0.04844403941974667, 'subsample': 0.530365118924548, 'colsample\_bytree': 0.63278},  
 [1] 2025-05-31 02:41:50,494 Trial 5 finished with value: 0.0014953686402401301 and parameters: {'n\_estimators': 124, 'max\_depth': 8, 'learning\_rate': 0.07771127157553044, 'subsample': 0.53338749566258, 'colsample\_bytree': 0.73105},  
 [1] 2025-05-31 02:41:50,579 Trial 6 finished with value: 0.00149741327656504032 and parameters: {'n\_estimators': 121, 'max\_depth': 5, 'learning\_rate': 0.01343291729483159, 'subsample': 0.720630513531558, 'colsample\_bytree': 0.5697},  
 [1] 2025-05-31 02:41:50,667 Trial 7 finished with value: 0.0014930726530631957 and parameters: {'n\_estimators': 62, 'max\_depth': 3, 'learning\_rate': 0.03772249063519739, 'subsample': 0.5873848971716148, 'colsample\_bytree': 0.6915969},  
 [1] 2025-05-31 02:41:50,888 Trial 8 finished with value: 0.00148869364535905 and parameters: {'n\_estimators': 278, 'max\_depth': 6, 'learning\_rate': 0.2387643371054954, 'subsample': 0.8513229086265947, 'colsample\_bytree': 0.74677},  
 [1] 2025-05-31 02:41:50,926 Trial 9 finished with value: 0.00149752764234874 and parameters: {'n\_estimators': 202, 'max\_depth': 7, 'learning\_rate': 0.01739317750424474, 'subsample': 0.601793626552687377, 'colsample\_bytree': 0.6944},  
 [1] 2025-05-31 02:41:51,071 Trial 10 finished with value: 0.00149286361180664 and parameters: {'n\_estimators': 239, 'max\_depth': 6, 'learning\_rate': 0.239236242061034, 'subsample': 0.9798151522667111, 'colsample\_bytree': 0.83496},  
 [1] 2025-05-31 02:41:51,168 Trial 11 finished with value: 0.00168630923248585 and parameters: {'n\_estimators': 61, 'max\_depth': 3, 'learning\_rate': 0.02518632715159565, 'subsample': 0.601863227547707, 'colsample\_bytree': 0.8332411},  
 [1] 2025-05-31 02:41:51,260 Trial 12 finished with value: 0.00150175195994272739 and parameters: {'n\_estimators': 55, 'max\_depth': 5, 'learning\_rate': 0.20687122713548242, 'subsample': 0.8740289703427056, 'colsample\_bytree': 0.82361},  
 [1] 2025-05-31 02:41:51,388 Trial 13 finished with value: 0.001517594938082488 and parameters: {'n\_estimators': 178, 'max\_depth': 7, 'learning\_rate': 0.15337150470409467, 'subsample': 0.6583124168581246, 'colsample\_bytree': 0.8324},  
 [1] 2025-05-31 02:41:52,216 Trial 14 finished with value: 0.00163593378973501 and parameters: {'n\_estimators': 145, 'max\_depth': 4, 'learning\_rate': 0.15817708873035785, 'subsample': 0.59708954324454, 'colsample\_bytree': 0.8762},  
 [1] 2025-05-31 02:41:52,327 Trial 15 finished with value: 0.002394849104694495 and parameters: {'n\_estimators': 105, 'max\_depth': 3, 'learning\_rate': 0.029158701549576019, 'subsample': 0.814712990611501, 'colsample\_bytree': 0.6879},  
 [1] 2025-05-31 02:41:52,461 Trial 16 finished with value: 0.0023983102362052605 and parameters: {'n\_estimators': 173, 'max\_depth': 6, 'learning\_rate': 0.1362603049319338, 'subsample': 0.9826190646536647, 'colsample\_bytree': 0.9944},  
 [1] 2025-05-31 02:41:52,652 Trial 17 finished with value: 0.00231059236180664 and parameters: {'n\_estimators': 91, 'max\_depth': 7, 'learning\_rate': 0.1005303382315683, 'subsample': 0.8183210322339903, 'colsample\_bytree': 0.88105},  
 [1] 2025-05-31 02:41:52,882 Trial 18 finished with value: 0.001495203933794314 and parameters: {'n\_estimators': 224, 'max\_depth': 5, 'learning\_rate': 0.205636601472522, 'subsample': 0.904808110488679, 'colsample\_bytree': 0.76621},  
 [1] 2025-05-31 02:41:52,919 Trial 19 finished with value: 0.0015046703053858803 and parameters: {'n\_estimators': 189, 'max\_depth': 3, 'learning\_rate': 0.205636601472522, 'subsample': 0.78723086032285543, 'colsample\_bytree': 0.6290},  
 [1] 2025-05-31 02:41:52,216 A new study created in memory with name: no-name-55899d-634f-4432-83d2-32d3a457  
 [1] 2025-05-31 02:41:52,327 Trial 0 finished with value: 0.002394849104694495 and parameters: {'n\_estimators': 105, 'max\_depth': 3, 'learning\_rate': 0.27928467872202, 'subsample': 0.82501554231761471, 'colsample\_bytree': 0.854528},  
 [1] 2025-05-31 02:41:52,472 Trial 1 finished with value: 0.002388671720267993 and parameters: {'n\_estimators': 268, 'max\_depth': 5, 'learning\_rate': 0.01774142097541, 'subsample': 0.532232267993058076, 'colsample\_bytree': 0.7300066},  
 [1] 2025-05-31 02:41:53,636 Trial 2 finished with value: 0.00236517659841168 and parameters: {'n\_estimators': 296, 'max\_depth': 5, 'learning\_rate': 0.2626205380253820532, 'subsample': 0.5287419486729, 'colsample\_bytree': 0.7305490},  
 [1] 2025-05-31 02:41:53,828 Trial 3 finished with value: 0.002319803093947036 and parameters: {'n\_estimators': 249, 'max\_depth': 5, 'learning\_rate': 0.127719091974625, 'subsample': 0.593626344328558, 'colsample\_bytree': 0.72927},  
 [1] 2025-05-31 02:41:53,882 Trial 4 finished with value: 0.002317604240053833 and parameters: {'n\_estimators': 237, 'max\_depth': 6, 'learning\_rate': 0.1277190915762029, 'subsample': 0.50559989905697, 'colsample\_bytree': 0.5428},  
 [1] 2025-05-31 02:41:54,141 Trial 5 finished with value: 0.00234989414980656955 and parameters: {'n\_estimators': 266, 'max\_depth': 4, 'learning\_rate': 0.13342822757651975, 'subsample': 0.637266760560229, 'colsample\_bytree': 0.6768},  
 [1] 2025-05-31 02:41:56,018 Trial 6 finished with value: 0.0022445046230152394 and parameters: {'n\_estimators': 265, 'max\_depth': 4, 'learning\_rate': 0.1988704665080408, 'subsample': 0.503924430593549, 'colsample\_bytree': 0.5834},  
 [1] 2025-05-31 02:41:56,299 Trial 7 finished with value: 0.00203692144137113 and parameters: {'n\_estimators': 200, 'max\_depth': 4, 'learning\_rate': 0.158702867281774, 'subsample': 0.752961517561147, 'colsample\_bytree': 0.582699},

[1] 2025-05-31 02:42:01,996 Trial 19 finished with value: 0.0031750/70541730044 and parameters: {'n\_estimators': 109, 'max\_depth': 5, 'learning\_rate': 0.1119362/179800503, 'subsample': 0.7817164/290457608, 'columsample\_bytree': 0.68400}

[1] 2025-05-31 02:42:02,147 A new study created in memory with name: no-name-f4e3cd5-0d42-4fa1-86ce-7f7d39dbed127

[1] 2025-05-31 02:42:02,250 Trial 0 finished with value: 0.0004431902131735819 and parameters: {'n\_estimators': 89, 'max\_depth': 8, 'learning\_rate': 0.024122781796045791, 'subsample': 0.7174135502065231, 'columsample\_bytree': 0.838950}

[1] 2025-05-31 02:42:02,377 Trial 1 finished with value: 0.000442939/2953575664 and parameters: {'n\_estimators': 243, 'max\_depth': 6, 'learning\_rate': 0.01184815427895637, 'subsample': 0.7258181794278435, 'columsample\_bytree': 0.6062}

[1] 2025-05-31 02:42:02,513 Trial 2 finished with value: 0.0004426932850889786 and parameters: {'n\_estimators': 299, 'max\_depth': 4, 'learning\_rate': 0.104629441148821489, 'subsample': 0.8854647375509774, 'columsample\_bytree': 0.5635}

[1] 2025-05-31 02:42:02,619 Trial 3 finished with value: 0.000444291362023376 and parameters: {'n\_estimators': 194, 'max\_depth': 7, 'learning\_rate': 0.023315081608592907, 'subsample': 0.7080539432620571, 'columsample\_bytree': 0.5031}

[1] 2025-05-31 02:42:02,742 Trial 4 finished with value: 0.000442944791049726786 and parameters: {'n\_estimators': 233, 'max\_depth': 7, 'learning\_rate': 0.0396170029449473, 'subsample': 0.967798102759818, 'columsample\_bytree': 0.8973}

[1] 2025-05-31 02:42:02,865 Trial 5 finished with value: 0.000442089130673527 and parameters: {'n\_estimators': 250, 'max\_depth': 7, 'learning\_rate': 0.02287446257573670824, 'subsample': 0.7617978917985195, 'columsample\_bytree': 0.828}

[1] 2025-05-31 02:42:02,995 Trial 6 finished with value: 0.0004437388928781623 and parameters: {'n\_estimators': 217, 'max\_depth': 6, 'learning\_rate': 0.022838059272100334, 'subsample': 0.8276503042878796, 'columsample\_bytree': 0.9776}

[1] 2025-05-31 02:42:03,060 Trial 7 finished with value: 0.000442627821099339 and parameters: {'n\_estimators': 60, 'max\_depth': 3, 'learning\_rate': 0.0651408538413181, 'subsample': 0.6855437302839688, 'columsample\_bytree': 0.7895779}

[1] 2025-05-31 02:42:03,165 Trial 8 finished with value: 0.0004448273290860695 and parameters: {'n\_estimators': 69, 'max\_depth': 6, 'learning\_rate': 0.010789931701978065, 'subsample': 0.8915287105364258, 'columsample\_bytree': 0.77416}

[1] 2025-05-31 02:42:03,243 Trial 9 finished with value: 0.0004448878934938228 and parameters: {'n\_estimators': 100, 'max\_depth': 5, 'learning\_rate': 0.0122021804074191, 'subsample': 0.830357062348868, 'columsample\_bytree': 0.59269}

[1] 2025-05-31 02:42:03,355 Trial 10 finished with value: 0.000514574006280413 and parameters: {'n\_estimators': 146, 'max\_depth': 8, 'learning\_rate': 0.2535548787831837, 'subsample': 0.5409085028993052, 'columsample\_bytree': 0.69900}

[1] 2025-05-31 02:42:03,471 Trial 11 finished with value: 0.0047510150810416087 and parameters: {'n\_estimators': 147, 'max\_depth': 3, 'learning\_rate': 0.0844991177588332, 'subsample': 0.58712850375127295, 'columsample\_bytree': 0.7278}

[1] 2025-05-31 02:42:03,630 Trial 12 finished with value: 0.0004671560461242016 and parameters: {'n\_estimators': 280, 'max\_depth': 3, 'learning\_rate': 0.09451629257207295, 'subsample': 0.6424949194817343, 'columsample\_bytree': 0.835}

[1] 2025-05-31 02:42:03,716 Trial 13 finished with value: 0.00451012171774469 and parameters: {'n\_estimators': 54, 'max\_depth': 4, 'learning\_rate': 0.0578836341119886, 'subsample': 0.7838046953874448, 'columsample\_bytree': 0.917666}

[1] 2025-05-31 02:42:03,830 Trial 14 finished with value: 0.004549797966563127 and parameters: {'n\_estimators': 145, 'max\_depth': 5, 'learning\_rate': 0.18372617125785867, 'subsample': 0.6457817126234617, 'columsample\_bytree': 0.7904}

[1] 2025-05-31 02:42:03,988 Trial 15 finished with value: 0.000440895451278024 and parameters: {'n\_estimators': 266, 'max\_depth': 7, 'learning\_rate': 0.02376848823885954, 'subsample': 0.6523288953279587, 'columsample\_bytree': 0.67}

[1] 2025-05-31 02:42:04,126 Trial 16 finished with value: 0.004463254780046123 and parameters: {'n\_estimators': 185, 'max\_depth': 4, 'learning\_rate': 0.0572592336249803, 'subsample': 0.788248572430487, 'columsample\_bytree': 0.849}

[1] 2025-05-31 02:42:04,233 Trial 17 finished with value: 0.000445789235856082 and parameters: {'n\_estimators': 112, 'max\_depth': 5, 'learning\_rate': 0.01762874831477218, 'subsample': 0.5826239711445452, 'columsample\_bytree': 0.996}

[1] 2025-05-31 02:42:04,366 Trial 18 finished with value: 0.00448813083863399 and parameters: {'n\_estimators': 204, 'max\_depth': 7, 'learning\_rate': 0.040678409303846506, 'subsample': 0.5078771246111, 'columsample\_bytree': 0.65082}

[1] 2025-05-31 02:42:04,474 Trial 19 finished with value: 0.00042908343104954 and parameters: {'n\_estimators': 126, 'max\_depth': 3, 'learning\_rate': 0.158879984817653, 'subsample': 0.9906138753012227, 'columsample\_bytree': 0.75182}

[1] 2025-05-31 02:42:04,614 A new study created in memory with name: no-name-0731895d-c610-4134-b13d-c4d06cf16d1

[1] 2025-05-31 02:42:04,725 Trial 0 finished with value: 0.0004872381495324756 and parameters: {'n\_estimators': 108, 'max\_depth': 7, 'learning\_rate': 0.0844371160337896, 'subsample': 0.6707259688495821, 'columsample\_bytree': 0.938192}

[1] 2025-05-31 02:42:04,871 Trial 1 finished with value: 0.00084687095166226 and parameters: {'n\_estimators': 296, 'max\_depth': 8, 'learning\_rate': 0.133521456453086494, 'subsample': 0.878059060699969, 'columsample\_bytree': 0.96492}

[1] 2025-05-31 02:42:05,003 Trial 2 finished with value: 0.008615961483891599 and parameters: {'n\_estimators': 248, 'max\_depth': 6, 'learning\_rate': 0.10743359525218478, 'subsample': 0.5033438330745408, 'columsample\_bytree': 0.75640}

[1] 2025-05-31 02:42:05,078 Trial 3 finished with value: 0.008546221665052627 and parameters: {'n\_estimators': 87, 'max\_depth': 4, 'learning\_rate': 0.02123395929024141, 'subsample': 0.5897520904985215, 'columsample\_bytree': 0.543560}

[1] 2025-05-31 02:42:05,172 Trial 4 finished with value: 0.008583523384346658 and parameters: {'n\_estimators': 126, 'max\_depth': 3, 'learning\_rate': 0.068218369619135, 'subsample': 0.5422879418139372, 'columsample\_bytree': 0.72512}

[1] 2025-05-31 02:42:05,239 Trial 5 finished with value: 0.00084709179643210707 and parameters: {'n\_estimators': 59, 'max\_depth': 3, 'learning\_rate': 0.02275972981050065, 'subsample': 0.7097857962248442, 'columsample\_bytree': 0.63828}

[1] 2025-05-31 02:42:05,360 Trial 6 finished with value: 0.000468481250878854 and parameters: {'n\_estimators': 216, 'max\_depth': 6, 'learning\_rate': 0.06660826011102076, 'subsample': 0.9690362491341301, 'columsample\_bytree': 0.98753}

[1] 2025-05-31 02:42:05,380 Trial 7 finished with value: 0.0008460755389001946 and parameters: {'n\_estimators': 250, 'max\_depth': 7, 'learning\_rate': 0.0285094945867615257, 'subsample': 0.990788747747346, 'columsample\_bytree': 0.7638}

[1] 2025-05-31 02:42:05,569 Trial 8 finished with value: 0.0085293450849222 and parameters: {'n\_estimators': 120, 'max\_depth': 8, 'learning\_rate': 0.031257105688775134, 'subsample': 0.5102258133778357, 'columsample\_bytree': 0.7009}

[1] 2025-05-31 02:42:05,644 Trial 9 finished with value: 0.00085110370402304674 and parameters: {'n\_estimators': 97, 'max\_depth': 5, 'learning\_rate': 0.040124348928488277, 'subsample': 0.9059404627736066, 'columsample\_bytree': 0.53102}

[1] 2025-05-31 02:42:05,784 Trial 10 finished with value: 0.00849716443848387 and parameters: {'n\_estimators': 204, 'max\_depth': 7, 'learning\_rate': 0.01035875720435951, 'subsample': 0.8198554322529595, 'columsample\_bytree': 0.838}

[1] 2025-05-31 02:42:05,948 Trial 11 finished with value: 0.000848637961705765 and parameters: {'n\_estimators': 296, 'max\_depth': 8, 'learning\_rate': 0.020594945817608817, 'subsample': 0.829829568026207, 'columsample\_bytree': 0.8760}

[1] 2025-05-31 02:42:06,103 Trial 12 finished with value: 0.00843947875822135 and parameters: {'n\_estimators': 295, 'max\_depth': 7, 'learning\_rate': 0.0202485817160432, 'subsample': 0.8769703371602002, 'columsample\_bytree': 0.8278}

[1] 2025-05-31 02:42:06,299 Trial 13 finished with value: 0.008463862321729638 and parameters: {'n\_estimators': 250, 'max\_depth': 7, 'learning\_rate': 0.0246796334417756, 'subsample': 0.9942848182645072, 'columsample\_bytree': 0.82762}

[1] 2025-05-31 02:42:06,451 Trial 14 finished with value: 0.00853399718388143 and parameters: {'n\_estimators': 260, 'max\_depth': 5, 'learning\_rate': 0.019356157627945605, 'subsample': 0.7636311172787189, 'columsample\_bytree': 0.791}

[1] 2025-05-31 02:42:06,585 Trial 15 finished with value: 0.00845819765059507 and parameters: {'n\_estimators': 192, 'max\_depth': 7, 'learning\_rate': 0.0478645598247571, 'subsample': 0.9036856186966492, 'columsample\_bytree': 0.6588}

[1] 2025-05-31 02:42:06,705 Trial 16 finished with value: 0.000858664995892232 and parameters: {'n\_estimators': 158, 'max\_depth': 6, 'learning\_rate': 0.04848801815070288, 'subsample': 0.734273187694138, 'columsample\_bytree': 0.6419}

[1] 2025-05-31 02:42:06,829 Trial 17 finished with value: 0.00847880936353154 and parameters: {'n\_estimators': 167, 'max\_depth': 7, 'learning\_rate': 0.1613886266395096, 'subsample': 0.890936543463343, 'columsample\_bytree': 0.6372}

[1] 2025-05-31 02:42:07,023 Trial 18 finished with value: 0.00084994041944518 and parameters: {'n\_estimators': 201, 'max\_depth': 6, 'learning\_rate': 0.299706515807089, 'subsample': 0.83835636623007715, 'columsample\_bytree': 0.902004}

[1] 2025-05-31 02:42:07,213 Trial 19 finished with value: 0.0085165995236136 and parameters: {'n\_estimators': 146, 'max\_depth': 8, 'learning\_rate': 0.0993527531737317, 'subsample': 0.754784687462423, 'columsample\_bytree': 0.6818}

[1] 2025-05-31 02:42:07,968 A new study created in memory with name: no-name-d3abc18a-fe12-45b1-bca6-1c5656965f3d

[1] 2025-05-31 02:42:08,189 Trial 0 finished with value: 0.00270481393074911 and parameters: {'n\_estimators': 183, 'max\_depth': 5, 'learning\_rate': 0.2056743349340742, 'subsample': 0.7015347591481919, 'columsample\_bytree': 0.58370237}

[1] 2025-05-31 02:42:09,184 Trial 1 finished with value: 0.0027336745697323866 and parameters: {'n\_estimators': 248, 'max\_depth': 6, 'learning\_rate': 0.02019666573071324, 'subsample': 0.9765136349459096, 'columsample\_bytree': 0.8309}

[1] 2025-05-31 02:42:09,357 Trial 2 finished with value: 0.00272596488943304 and parameters: {'n\_estimators': 68, 'max\_depth': 8, 'learning\_rate': 0.05936436936378278, 'subsample': 0.901792250237452, 'columsample\_bytree': 0.6421921}

[1] 2025-05-31 02:42:09,538 Trial 3 finished with value: 0.002730376987406537 and parameters: {'n\_estimators': 183, 'max\_depth': 4, 'learning\_rate': 0.02939532751523567, 'subsample': 0.8830955955786664, 'columsample\_bytree': 0.7713}

[1] 2025-05-31 02:42:09,821 Trial 4 finished with value: 0.002733556969113563 and parameters: {'n\_estimators': 135, 'max\_depth': 4, 'learning\_rate': 0.023644076741620687, 'subsample': 0.937469416313369, 'columsample\_bytree': 0.80466}

[1] 2025-05-31 02:42:09,973 Trial 5 finished with value: 0.00272490497208377 and parameters: {'n\_estimators': 66, 'max\_depth': 6, 'learning\_rate': 0.052603566763663775, 'subsample': 0.9273311902619938, 'columsample\_bytree': 0.72193}

[1] 2025-05-31 02:42:10,048 Trial 6 finished with value: 0.0027265310859158 and parameters: {'n\_estimators': 97, 'max\_depth': 4, 'learning\_rate': 0.05264296970857113, 'subsample': 0.8733292780789445, 'columsample\_bytree': 0.5716621}

[1] 2025-05-31 02:42:10,113 Trial 7 finished with value: 0.002720898981097279 and parameters: {'n\_estimators': 204, 'max\_depth': 5, 'learning\_rate': 0.01767842400855244, 'subsample': 0.72423616799383, 'columsample\_bytree': 0.72336}

[1] 2025-05-31 02:42:10,213 Trial 8 finished with value: 0.00271179870931468 and parameters: {'n\_estimators': 106, 'max\_depth': 4, 'learning\_rate': 0.0497049982794252, 'subsample': 0.799937474945349, 'columsample\_bytree': 0.6780891}

[1] 2025-05-31 02:42:10,351 Trial 9 finished with value: 0.002707092202652177 and parameters: {'n\_estimators': 217, 'max\_depth': 5, 'learning\_rate': 0.01057625061513213, 'subsample': 0.606380341234222, 'columsample\_bytree': 0.5462309}

[1] 2025-05-31 02:42:10,534 Trial 10 finished with value: 0.002671443440918226 and parameters: {'n\_estimators': 297, 'max\_depth': 8, 'learning\_rate': 0.01276340582303174, 'subsample': 0.5311202435937551, 'columsample\_bytree': 0.96913}

[1] 2025-05-31 02:42:10,593 Trial 11 finished with value: 0.00272596488943304 and parameters: {'n\_estimators': 68, 'max\_depth': 8, 'learning\_rate': 0.027637405398700708, 'subsample': 0.768846473721670493, 'columsample\_bytree': 0.53404}

[1] 2025-05-31 02:42:10,598 Trial 12 finished with value: 0.0027033592903854533 and parameters: {'n\_estimators': 183, 'max\_depth': 4, 'learning\_rate': 0.02939532751523567, 'subsample': 0.8830955955786664, 'columsample\_bytree': 0.7713}

[1] 2025-05-31 02:42:10,881 Trial 13 finished with value: 0.00269077156260147 and parameters: {'n\_estimators': 300, 'max\_depth': 8, 'learning\_rate': 0.02366347162040161, 'subsample': 0.5032995823211526, 'columsample\_bytree': 0.99049}

[1] 2025-05-31 02:42:10,953 Trial 14 finished with value: 0.00267265310859158 and parameters: {'n\_estimators': 299, 'max\_depth': 7, 'learning\_rate': 0.023663522853458, 'subsample': 0.506315778114325, 'columsample\_bytree': 0.98923}

[1] 2025-05-31 02:42:11,053 Trial 15 finished with value: 0.0026268949781604074 and parameters: {'n\_estimators': 252, 'max\_depth': 7, 'learning\_rate': 0.1703555821291955, 'subsample': 0.602850134595484, 'columsample\_bytree': 0.89303}

[1] 2025-05-31 02:42:11,111 Trial 16 finished with value: 0.0026503564974098866 and parameters: {'n\_estimators': 257, 'max\_depth': 7, 'learning\_rate': 0.1886662623789454, 'subsample': 0.592924851289624, 'columsample\_bytree': 0.8844}

[1] 2025-05-31 02:42:11,697 Trial 17 finished with value: 0.0026715726411838067 and parameters: {'n\_estimators': 257, 'max\_depth': 7, 'learning\_rate': 0.07705663341639932, 'subsample': 0.6483071398128764, 'columsample\_bytree': 0.8665}

[1] 2025-05-31 02:42:11,846 Trial 18 finished with value: 0.0027252179873196204 and parameters: {'n\_estimators': 227, 'max\_depth': 6, 'learning\_rate': 0.01073703725468423516, 'subsample': 0.783272116617086, 'columsample\_bytree': 0.873}

[1] 2025-05-31 02:42:11,975 Trial 19 finished with value: 0.00266883048009310494 and parameters: {'n\_estimators': 145, 'max\_depth': 3, 'learning\_rate': 0.1786795763472956, 'subsample': 0.5751872935799873, 'columsample\_bytree': 0.90446}

[1] 2025-05-31 02:42:12,226 Trial 0 finished with value: 0.002495791157401744375, 'subsample': 0.581200876666208, 'columsample\_bytree': 0.5494552}

[1] 2025-05-31 02:42:12,344 Trial 1 finished with value: 0.002501714037334514 and parameters: {'n\_estimators': 219, 'max\_depth': 3, 'learning\_rate': 0.2393372741865537, 'subsample': 0.557159202282545, 'columsample\_bytree': 0.77501}

[1] 2025-05-31 02:42:12,477 Trial 2 finished with value: 0.0020523566213414 and parameters: {'n\_estimators': 260, 'max\_depth': 4, 'learning\_rate': 0.0122882035295747122, 'subsample': 0.81642593061821621, 'columsample\_bytree': 0.5764}

[1] 2025-05-31 02:42:12,597 Trial 3 finished with value: 0.002027596488943304 and parameters: {'n\_estimators': 250, 'max\_depth': 8, 'learning\_rate': 0.01276340582310708, 'subsample': 0.53112024359375216, 'columsample\_bytree': 0.5764}

[1] 2025-05-31 02:42:12,712 Trial 4 finished with value: 0.0020520726993854533 and parameters: {'n\_estimators': 219, 'max\_depth': 6, 'learning\_rate': 0.0405562930214075, 'subsample': 0.809646480208627, 'columsample\_bytree': 0.5761}

[1] 2025-05-31 02:42:12,805 Trial 5 finished with value: 0.00250916514526877 and parameters: {'n\_estimators': 154, 'max\_depth': 6, 'learning\_rate': 0.0770536049538633, 'subsample': 0.771066663718924, 'columsample\_bytree': 0.5906822}

[1] 2025-05-31 02:42:12,883 Trial 6 finished with value: 0.002488151715839443 and parameters: {'n\_estimators': 98, 'max\_depth': 6, 'learning\_rate': 0.219011764738763488, 'subsample': 0.7062516846020899, 'columsample\_bytree': 0.738403}

[1] 2025-05-31 02:42:12,979 Trial 7 finished with value: 0.0027018714532355039 and parameters: {'n\_estimators': 108, 'max\_depth': 7, 'learning\_rate': 0.06737018792767059, 'subsample': 0.842745232530593, 'columsample\_bytree': 0.8426720

[1] 2025-05-31 02:42:13,118 Trial 8 finished with value: 0.002499827421744197 and parameters: {'n\_estimators': 282, 'max\_depth': 5, 'learning\_rate': 0.021571629071138602, 'subsample': 0.686848253569872, 'columsample\_bytree': 0.7744}

[1] 2025-05-31 02:42:13,311 Trial 9 finished with value: 0.00250163366293904 and parameters: {'n\_estimators': 135, 'max\_depth': 7, 'learning\_rate': 0.14491172618743448, 'subsample': 0.599348772584123, 'columsample\_bytree': 0.99596}

[1] 2025-05-31 02:42:13,397 Trial 11 finished with value: 0.002509166473429465 and parameters: {'n\_estimators': 52, 'max\_depth': 5, 'learning\_rate': 0.11383707546971347, 'subsample': 0.642769145273214, 'columsample\_bytree': 0.67693}

[1] 2025-05-31 02:42:13,482 Trial 12 finished with value: 0.0025007019135196294 and parameters: {'n\_estimators': 51, 'max\_depth': 4, 'learning\_rate': 0.031958790416307704, 'subsample': 0.5070469145865656, 'columsample\_bytree': 0.91932}

[1] 2025-05-31 02:42:13,590 Trial 13 finished with value: 0.0024954296658944693 and parameters: {'n\_estimators': 104, 'max\_depth': 5, 'learning\_rate': 0.13049622193088197, 'subsample': 0.693339506747066, 'columsample\_bytree': 0.67636}

[1] 2025-05-31 02:42:13,694 Trial 14 finished with value: 0.00249856020867964964 and parameters: {'n\_estimators': 110, 'max\_depth': 6, 'learning\_rate': 0.1496342936176652, 'subsample': 0.6842784029961746, 'columsample\_bytree': 0.6762}

[1] 2025-05-31 02:42:13,893 Trial 15 finished with value: 0.00248562323381883 and parameters: {'n\_estimators': 176, 'max\_depth': 4, 'learning\_rate': 0.2852347947046773, 'subsample': 0.706297286960196, 'columsample\_bytree': 0.79255}

[1] 2025-05-31 02:42:13,951 Trial 16 finished with value: 0.00251752960329032 and parameters: {'n\_estimators': 188, 'max\_depth': 3, 'learning\_rate': 0.1963929183754948, 'subsample':

```
[1] 2025-05-31 02:42:19, 860] Trial 6 finished with value: 0.0007658481699896778 and parameters: {'n_estimators': 83, 'max_depth': 8, 'learning_rate': 0.1161905040383239, 'subsample': 0.839589398811073, 'colsample_bytree': 0.6342801}
[1] 2025-05-31 02:42:20, 009] Trial 7 finished with value: 0.000787543389115001 and parameters: {'n_estimators': 272, 'max_depth': 5, 'learning_rate': 0.1234650293383003, 'subsample': 0.9578095297344659, 'colsample_bytree': 0.51916}
[1] 2025-05-31 02:42:20, 154] Trial 8 finished with value: 0.0007867315399667466 and parameters: {'n_estimators': 249, 'max_depth': 8, 'learning_rate': 0.011294627056372486, 'subsample': 0.5689275349933061, 'colsample_bytree': 0.8973}
[1] 2025-05-31 02:42:20, 319] Trial 9 finished with value: 0.000765849821367642 and parameters: {'n_estimators': 87, 'max_depth': 5, 'learning_rate': 0.06579230971135667, 'subsample': 0.711680513407785, 'colsample_bytree': 0.742671}
[1] 2025-05-31 02:42:21, 012] Trial 10 finished with value: 0.00881171943830049 and parameters: {'n_estimators': 202, 'max_depth': 3, 'learning_rate': 0.011294627056372486, 'subsample': 0.5038256299334646, 'colsample_bytree': 0.98870}
[1] 2025-05-31 02:42:21, 294] Trial 11 finished with value: 0.007666788695976956 and parameters: {'n_estimators': 133, 'max_depth': 6, 'learning_rate': 0.03144380355777735, 'subsample': 0.9472480034131721, 'colsample_bytree': 0.82018}
[1] 2025-05-31 02:42:22, 160] Trial 12 finished with value: 0.007436493620428471 and parameters: {'n_estimators': 140, 'max_depth': 7, 'learning_rate': 0.113347151718383, 'subsample': 0.63809605623055969, 'colsample_bytree': 0.8022}
[1] 2025-05-31 02:42:22, 406] Trial 13 finished with value: 0.007502169138455174 and parameters: {'n_estimators': 211, 'max_depth': 7, 'learning_rate': 0.22447072023601, 'subsample': 0.634240422504067, 'colsample_bytree': 0.84253}
[1] 2025-05-31 02:42:22, 603] Trial 14 finished with value: 0.008316307734819515 and parameters: {'n_estimators': 225, 'max_depth': 7, 'learning_rate': 0.2950430113184043, 'subsample': 0.63916526944859, 'colsample_bytree': 0.8368}
[1] 2025-05-31 02:42:22, 759] Trial 15 finished with value: 0.00074049919082457 and parameters: {'n_estimators': 163, 'max_depth': 7, 'learning_rate': 0.18244442547636718, 'subsample': 0.74781049951011004, 'colsample_bytree': 0.8393}
[1] 2025-05-31 02:42:23, 039] Trial 16 finished with value: 0.0074288764474436 and parameters: {'n_estimators': 163, 'max_depth': 7, 'learning_rate': 0.17210295654764782, 'subsample': 0.7680669173412505, 'colsample_bytree': 0.7895}
[1] 2025-05-31 02:42:23, 159] Trial 17 finished with value: 0.00751101270622052 and parameters: {'n_estimators': 164, 'max_depth': 6, 'learning_rate': 0.1805613036061432, 'subsample': 0.771653009966069, 'colsample_bytree': 0.8943}
[1] 2025-05-31 02:42:23, 280] Trial 18 finished with value: 0.007642378273866202 and parameters: {'n_estimators': 159, 'max_depth': 7, 'learning_rate': 0.1643264642073258, 'subsample': 0.738778520155202, 'colsample_bytree': 0.77705}
[1] 2025-05-31 02:42:23, 528] A new study created in memory with name: no-name_09cc16-5116-4134-b664-f32bf60d639b
[1] 2025-05-31 02:42:23, 669] Trial 0 finished with value: 0.000559588987592901 and parameters: {'n_estimators': 191, 'max_depth': 5, 'learning_rate': 0.2323560150781772, 'subsample': 0.9327710415134023, 'colsample_bytree': 0.900713}
[1] 2025-05-31 02:42:23, 767] Trial 1 finished with value: 0.0005612419389718047 and parameters: {'n_estimators': 159, 'max_depth': 6, 'learning_rate': 0.10121861715174006, 'subsample': 0.686611965611573, 'colsample_bytree': 0.94734}
[1] 2025-05-31 02:42:23, 930] Trial 2 finished with value: 0.00055082303838175295 and parameters: {'n_estimators': 290, 'max_depth': 6, 'learning_rate': 0.03736511927126893, 'subsample': 0.771253497803297, 'colsample_bytree': 0.77472}
[1] 2025-05-31 02:42:24, 055] Trial 3 finished with value: 0.000581122884841 and parameters: {'n_estimators': 239, 'max_depth': 3, 'learning_rate': 0.06022142272654455, 'subsample': 0.8613841980586198, 'colsample_bytree': 0.848880}
[1] 2025-05-31 02:42:24, 140] Trial 4 finished with value: 0.0005503663690182878 and parameters: {'n_estimators': 109, 'max_depth': 3, 'learning_rate': 0.0374330215848104, 'subsample': 0.6431362485697066, 'colsample_bytree': 0.79572}
[1] 2025-05-31 02:42:24, 267] Trial 5 finished with value: 0.0005504849295676544 and parameters: {'n_estimators': 254, 'max_depth': 4, 'learning_rate': 0.2111207014227522, 'subsample': 0.6262199721409981, 'colsample_bytree': 0.67863}
[1] 2025-05-31 02:42:24, 358] Trial 6 finished with value: 0.000550107289327583 and parameters: {'n_estimators': 135, 'max_depth': 6, 'learning_rate': 0.2910658327134606, 'subsample': 0.48742653627043, 'colsample_bytree': 0.663056}
[1] 2025-05-31 02:42:24, 493] Trial 7 finished with value: 0.000560743298968409 and parameters: {'n_estimators': 289, 'max_depth': 8, 'learning_rate': 0.09308739104409033, 'subsample': 0.769421561158527, 'colsample_bytree': 0.52920}
[1] 2025-05-31 02:42:24, 619] Trial 8 finished with value: 0.0005598878998732586 and parameters: {'n_estimators': 244, 'max_depth': 4, 'learning_rate': 0.1707416894457616, 'subsample': 0.797326684615595, 'colsample_bytree': 0.733365}
[1] 2025-05-31 02:42:24, 727] Trial 9 finished with value: 0.0005508203192974959 and parameters: {'n_estimators': 195, 'max_depth': 6, 'learning_rate': 0.0238358874226563, 'subsample': 0.9117094191867109, 'colsample_bytree': 0.546399}
[1] 2025-05-31 02:42:24, 817] Trial 10 finished with value: 0.000581057024659084 and parameters: {'n_estimators': 60, 'max_depth': 8, 'learning_rate': 0.10238897120093159, 'subsample': 0.990102120093159, 'colsample_bytree': 0.5059}
[1] 2025-05-31 02:42:24, 916] Trial 11 finished with value: 0.000587916166098514 and parameters: {'n_estimators': 97, 'max_depth': 7, 'learning_rate': 0.024172932691632412, 'subsample': 0.572926883568678, 'colsample_bytree': 0.56633}
[1] 2025-05-31 02:42:24, 255] Trial 12 finished with value: 0.00058436433491099 and parameters: {'n_estimators': 194, 'max_depth': 3, 'learning_rate': 0.1251905032227727, 'subsample': 0.630796726326306, 'colsample_bytree': 0.608}
[1] 2025-05-31 02:42:25, 175] Trial 13 finished with value: 0.000587123616513179 and parameters: {'n_estimators': 116, 'max_depth': 5, 'learning_rate': 0.01132847134302665, 'subsample': 0.50249109362277471, 'colsample_bytree': 0.802}
[1] 2025-05-31 02:42:25, 269] Trial 14 finished with value: 0.000558107289327583 and parameters: {'n_estimators': 69, 'max_depth': 7, 'learning_rate': 0.03484977061566171, 'subsample': 0.692058332700978, 'colsample_bytree': 0.89580}
[1] 2025-05-31 02:42:25, 393] Trial 15 finished with value: 0.000598391131484324978 and parameters: {'n_estimators': 156, 'max_depth': 4, 'learning_rate': 0.018830237011633475, 'subsample': 0.5922971095913982, 'colsample_bytree': 0.706}
[1] 2025-05-31 02:42:25, 528] Trial 16 finished with value: 0.00055815362098846 and parameters: {'n_estimators': 211, 'max_depth': 7, 'learning_rate': 0.04532546519664583, 'subsample': 0.90120375501581, 'colsample_bytree': 0.62752}
[1] 2025-05-31 02:42:25, 624] Trial 17 finished with value: 0.000585021579390709 and parameters: {'n_estimators': 95, 'max_depth': 5, 'learning_rate': 0.0692681203566201, 'subsample': 0.70171995545542, 'colsample_bytree': 0.837220}
[1] 2025-05-31 02:42:25, 720] Trial 18 finished with value: 0.000581003279947202 and parameters: {'n_estimators': 80, 'max_depth': 5, 'learning_rate': 0.07107632745278279, 'subsample': 0.9964405379932877, 'colsample_bytree': 0.97683}
[1] 2025-05-31 02:42:25, 900] A new study created in memory with name: no-name_f38491-619e-4e4b-b4a1-512a0f72d9
[1] 2025-05-31 02:42:26, 143] Trial 0 finished with value: 0.00732706302313346 and parameters: {'n_estimators': 126, 'max_depth': 6, 'learning_rate': 0.03888382086536636, 'subsample': 0.819441428510410366, 'colsample_bytree': 0.596565}
[1] 2025-05-31 02:42:26, 215] Trial 1 finished with value: 0.000730754214207402 and parameters: {'n_estimators': 79, 'max_depth': 6, 'learning_rate': 0.241817291878954, 'subsample': 0.6819725566243848, 'colsample_bytree': 0.881774}
[1] 2025-05-31 02:42:26, 433] Trial 3 finished with value: 0.0007287894516744875 and parameters: {'n_estimators': 101, 'max_depth': 5, 'learning_rate': 0.077704127042126725, 'subsample': 0.5677174002126725, 'colsample_bytree': 0.5260809}
[1] 2025-05-31 02:42:26, 534] Trial 4 finished with value: 0.00077632014318527 and parameters: {'n_estimators': 168, 'max_depth': 8, 'learning_rate': 0.2109085934239788, 'subsample': 0.549724965030233, 'colsample_bytree': 0.93459}
[1] 2025-05-31 02:42:26, 658] Trial 5 finished with value: 0.000739216233994279 and parameters: {'n_estimators': 219, 'max_depth': 4, 'learning_rate': 0.04633471294761879, 'subsample': 0.518054221875048, 'colsample_bytree': 0.97349}
[1] 2025-05-31 02:42:26, 773] Trial 6 finished with value: 0.0007355418361531965 and parameters: {'n_estimators': 197, 'max_depth': 5, 'learning_rate': 0.08784010642828356, 'subsample': 0.5042286163229968, 'colsample_bytree': 0.83488}
[1] 2025-05-31 02:42:26, 846] Trial 7 finished with value: 0.00072830598775137 and parameters: {'n_estimators': 137, 'max_depth': 6, 'learning_rate': 0.0348287413430266, 'subsample': 0.1338277856115446, 'colsample_bytree': 0.85567}
[1] 2025-05-31 02:42:26, 951] Trial 9 finished with value: 0.0007270416039401551 and parameters: {'n_estimators': 116, 'max_depth': 6, 'learning_rate': 0.068325082375746, 'subsample': 0.68325082375746, 'colsample_bytree': 0.92325}
[1] 2025-05-31 02:42:27, 091] Trial 10 finished with value: 0.0037311519428143666 and parameters: {'n_estimators': 242, 'max_depth': 4, 'learning_rate': 0.200563670770243, 'subsample': 0.150981440274657, 'colsample_bytree': 0.7864136}
[1] 2025-05-31 02:42:27, 215] Trial 1 finished with value: 0.00073007574214207402 and parameters: {'n_estimators': 79, 'max_depth': 6, 'learning_rate': 0.241817291878954, 'subsample': 0.6819725566243848, 'colsample_bytree': 0.881774}
[1] 2025-05-31 02:42:27, 433] Trial 3 finished with value: 0.0007287894516744875 and parameters: {'n_estimators': 101, 'max_depth': 5, 'learning_rate': 0.077704127042126725, 'subsample': 0.5677174002126725, 'colsample_bytree': 0.5260809}
[1] 2025-05-31 02:42:27, 534] Trial 4 finished with value: 0.00077632014318527 and parameters: {'n_estimators': 168, 'max_depth': 8, 'learning_rate': 0.2109085934239788, 'subsample': 0.549724965030233, 'colsample_bytree': 0.93459}
[1] 2025-05-31 02:42:27, 658] Trial 5 finished with value: 0.000739216233994279 and parameters: {'n_estimators': 219, 'max_depth': 4, 'learning_rate': 0.04633471294761879, 'subsample': 0.518054221875048, 'colsample_bytree': 0.97349}
[1] 2025-05-31 02:42:27, 773] Trial 6 finished with value: 0.0007355418361531965 and parameters: {'n_estimators': 197, 'max_depth': 5, 'learning_rate': 0.08784010642828356, 'subsample': 0.5042286163229968, 'colsample_bytree': 0.83488}
[1] 2025-05-31 02:42:27, 846] Trial 7 finished with value: 0.00072830598775137 and parameters: {'n_estimators': 137, 'max_depth': 6, 'learning_rate': 0.0348287413430266, 'subsample': 0.1338277856115446, 'colsample_bytree': 0.85567}
[1] 2025-05-31 02:42:27, 951] Trial 9 finished with value: 0.0007270416039401551 and parameters: {'n_estimators': 116, 'max_depth': 6, 'learning_rate': 0.068325082375746, 'subsample': 0.68325082375746, 'colsample_bytree': 0.92325}
[1] 2025-05-31 02:42:27, 991] Trial 10 finished with value: 0.0007300940898506382 and parameters: {'n_estimators': 143, 'max_depth': 8, 'learning_rate': 0.0959530602331684, 'subsample': 0.946751301127216, 'colsample_bytree': 0.5031}
[1] 2025-05-31 02:42:27, 315] Trial 11 finished with value: 0.00072703407788447088 and parameters: {'n_estimators': 107, 'max_depth': 7, 'learning_rate': 0.01163689980823628, 'subsample': 0.697800222707072, 'colsample_bytree': 0.71977}
[1] 2025-05-31 02:42:27, 401] Trial 12 finished with value: 0.0007293193774111638 and parameters: {'n_estimators': 54, 'max_depth': 7, 'learning_rate': 0.024172932691632412, 'subsample': 0.7117407355695639, 'colsample_bytree': 0.7136}
[1] 2025-05-31 02:42:27, 499] Trial 13 finished with value: 0.0007268210620137096 and parameters: {'n_estimators': 101, 'max_depth': 7, 'learning_rate': 0.0104916122749688, 'subsample': 0.619224448391988, 'colsample_bytree': 0.5152}
[1] 2025-05-31 02:42:27, 619] Trial 14 finished with value: 0.0007263015998268249 and parameters: {'n_estimators': 164, 'max_depth': 5, 'learning_rate': 0.08703717804344972, 'subsample': 0.5948633978192805, 'colsample_bytree': 0.5390}
[1] 2025-05-31 02:42:27, 737] Trial 15 finished with value: 0.0007360793940961551 and parameters: {'n_estimators': 167, 'max_depth': 5, 'learning_rate': 0.08685634506571428, 'subsample': 0.60178178971167016, 'colsample_bytree': 0.5938}
[1] 2025-05-31 02:42:27, 848] Trial 16 finished with value: 0.0007263023832385275 and parameters: {'n_estimators': 186, 'max_depth': 4, 'learning_rate': 0.0358938231984912, 'subsample': 0.7617461313383518, 'colsample_bytree': 0.56740}
[1] 2025-05-31 02:42:27, 944] Trial 17 finished with value: 0.0007263023832385275 and parameters: {'n_estimators': 186, 'max_depth': 4, 'learning_rate': 0.0358938231984912, 'subsample': 0.7617461313383518, 'colsample_bytree': 0.56740}
[1] 2025-05-31 02:42:27, 994] Trial 18 finished with value: 0.0007263023832385275 and parameters: {'n_estimators': 186, 'max_depth': 4, 'learning_rate': 0.0358938231984912, 'subsample': 0.7617461313383518, 'colsample_bytree': 0.56740}
[1] 2025-05-31 02:42:28, 092] Trial 19 finished with value: 0.000725453371749491006 and parameters: {'n_estimators': 132, 'max_depth': 4, 'learning_rate': 0.0133851732996513, 'subsample': 0.8246341091636245, 'colsample_bytree': 0.63251}
[1] 2025-05-31 02:42:28, 207] Trial 20 finished with value: 0.00072852677332623263 and parameters: {'n_estimators': 147, 'max_depth': 4, 'learning_rate': 0.020705962545044542, 'subsample': 0.907505862545044542, 'colsample_bytree': 0.5597}
[1] 2025-05-31 02:42:28, 294] A new study created in memory with name: no-name_f95492e4-c185-4056-9dfb-36480031939
[1] 2025-05-31 02:42:28, 353] Trial 0 finished with value: 0.0012979626099877274 and parameters: {'n_estimators': 274, 'max_depth': 4, 'learning_rate': 0.184944026888751297, 'subsample': 0.5614867478326474, 'colsample_bytree': 0.71539}
[1] 2025-05-31 02:42:28, 578] Trial 1 finished with value: 0.0013077064452352329 and parameters: {'n_estimators': 267, 'max_depth': 7, 'learning_rate': 0.21064380427072073, 'subsample': 0.656250633055478, 'colsample_bytree': 0.58649}
[1] 2025-05-31 02:42:28, 706] Trial 2 finished with value: 0.00129451164560444 and parameters: {'n_estimators': 252, 'max_depth': 7, 'learning_rate': 0.16534054521622152, 'subsample': 0.5071609397937393, 'colsample_bytree': 0.69466}
[1] 2025-05-31 02:42:28, 823] Trial 3 finished with value: 0.0013125304879542464206 and parameters: {'n_estimators': 215, 'max_depth': 6, 'learning_rate': 0.01065772249052336, 'subsample': 0.565610588000542, 'colsample_bytree': 0.67074}
[1] 2025-05-31 02:42:28, 908] Trial 4 finished with value: 0.001312594592461921692 and parameters: {'n_estimators': 125, 'max_depth': 7, 'learning_rate': 0.04342627809881955, 'subsample': 0.930730536166166765, 'colsample_bytree': 0.80167}
[1] 2025-05-31 02:42:28, 919] Trial 5 finished with value: 0.001303791459495023532811 and parameters: {'n_estimators': 52, 'max_depth': 8, 'learning_rate': 0.09347305115035517, 'subsample': 0.94937265785081, 'colsample_bytree': 0.9159351}
[1] 2025-05-31 02:42:29, 141] Trial 6 finished with value: 0.0013161963988523581 and parameters: {'n_estimators': 283, 'max_depth': 3, 'learning_rate': 0.0123302528402611655, 'subsample': 0.582674054705491661, 'colsample_bytree': 0.907444}
[1] 2025-05-31 02:42:29, 287] Trial 7 finished with value: 0.0013255154795595543 and parameters: {'n_estimators': 289, 'max_depth': 4, 'learning_rate': 0.1233028042611655, 'subsample': 0.8421239388757196536, 'colsample_bytree': 0.97855}
[1] 2025-05-31 02:42:29, 414] Trial 8 finished with value: 0.001305489377199303373 and parameters: {'n_estimators': 255, 'max_depth': 3, 'learning_rate': 0.05198406247634806, 'subsample': 0.931523381610676093, 'colsample_bytree': 0.68396}
[1] 2025-05-31 02:42:29, 519] Trial 9 finished with value: 0.001312914529370922871 and parameters: {'n_estimators': 181, 'max_depth': 5, 'learning_rate': 0.0862518084957748, 'subsample': 0.7565764062401955, 'colsample_bytree': 0.54104}
[1] 2025-05-31 02:42:29, 759] Trial 11 finished with value: 0.00130287115811822727, 'subsample': 0.6532109084041238, 'colsample_bytree': 0.7205774
[1] 2025-05-31 02:42:31, 646] Trial 12 finished with value: 0.0014703610769170001 and parameters: {'n_estimators': 234, 'max_depth': 5, 'learning_rate': 0.01262995380305085, 'subsample': 0.5763845160817693, 'colsample_bytree': 0.972331}
[1] 2025-05-31 02:42:31, 775] Trial 13 finished with value: 0.00146898562092775 and parameters: {'n_estimators': 246, 'max_depth': 7, 'learning_rate': 0.070268342350322, 'subsample': 0.7481946806457047, 'colsample_bytree': 0.8388005}
[1] 2025-05-31 02:42:31, 914] Trial 14 finished with value: 0.0014705919927129482 and parameters: {'n_estimators': 279, 'max_depth': 7, 'learning_rate': 0.0358978612682867, 'subsample': 0.63361037557682, 'colsample_bytree': 0.63364}
[1] 2025-05-31 02:42:31, 979] Trial 15 finished with value: 0.0014686118899120930 and parameters: {'n_estimators': 56, 'max_depth': 3, 'learning_rate': 0.2026240480478853, 'subsample': 0.753492192244463, 'colsample_bytree': 0.8877857}
[1] 2025-05-31 02:42:32, 083] Trial 16 finished with value: 0.001473112637295023 and parameters: {'n_estimators': 179, 'max_depth': 4, 'learning_rate': 0.124740821172937, 'subsample': 0.5890942694909299, 'colsample_bytree': 0.74743}
[1] 2025-05-31 02:42:32, 191] Trial 17 finished with value: 0.00147311173567037302 and parameters: {'n_estimators': 130, 'max_depth': 5, 'learning_rate': 0.0345583721593676, 'subsample': 0.6751623742217341, 'colsample_bytree': 0.974330}
[1] 2025-05-31 02:42:32, 333] Trial 18 finished with value: 0.001469694010236745 and parameters: {'n_estimators': 64, 'max_depth': 3, 'learning_rate': 0.08827768440104891, 'subsample': 0.86875995095312923, 'colsample_bytree': 0.6392}
[1] 2025-05-31 02:42:33, 360] Trial 19 finished with value: 0.00146901342130284 and parameters: {'n_estimators': 158, 'max_depth': 
```

[1] 2025-05-31 02:42:39.573 Trial 14 finished with value: 0.001276396076622317 and parameters: {'n\_estimators': 111, 'max\_depth': 4, 'learning\_rate': 0.001276396076622317, 'subsample': 0.6807467521976156, 'columsample\_bytree': 0.7000000000000001}, 'columsample\_bytree': 0.7000000000000001}

[1] 2025-05-31 02:42:39.688 Trial 15 finished with value: 0.001284538682435912 and parameters: {'n\_estimators': 163, 'max\_depth': 4, 'learning\_rate': 0.07155744019195347, 'subsample': 0.633769611199154, 'columsample\_bytree': 0.8923}

[1] 2025-05-31 02:42:39.491 Trial 16 finished with value: 0.001255605382673772 and parameters: {'n\_estimators': 116, 'max\_depth': 3, 'learning\_rate': 0.01446761755676557, 'subsample': 0.7739080539307015, 'columsample\_bytree': 0.9987}

[1] 2025-05-31 02:42:39.589 Trial 17 finished with value: 0.0012658987071008105 and parameters: {'n\_estimators': 97, 'max\_depth': 3, 'learning\_rate': 0.1446761755676557, 'subsample': 0.7739080539307015, 'columsample\_bytree': 0.9987}

[1] 2025-05-31 02:42:39.697 Trial 18 finished with value: 0.00129458324325362 and parameters: {'n\_estimators': 132, 'max\_depth': 3, 'learning\_rate': 0.16341629995101978, 'subsample': 0.5765409392618337, 'columsample\_bytree': 0.61498}

[1] 2025-05-31 02:42:39.793 Trial 19 finished with value: 0.0012624338350416429 and parameters: {'n\_estimators': 80, 'max\_depth': 5, 'learning\_rate': 0.12956641748723456, 'subsample': 0.772281196922824, 'columsample\_bytree': 0.91044}

[1] 2025-05-31 02:42:39.886 A new study created in memory with name: no-name\_09d0fd-6304-490b-a79f-657f5b01afe

[1] 2025-05-31 02:42:39.900 Trial 0 finished with value: 0.00289339948479976 and parameters: {'n\_estimators': 72, 'max\_depth': 6, 'learning\_rate': 0.18019738733560057, 'subsample': 0.7774284672320338, 'columsample\_bytree': 0.978380}

[1] 2025-05-31 02:42:40.098 Trial 1 finished with value: 0.002881852472030946 and parameters: {'n\_estimators': 198, 'max\_depth': 3, 'learning\_rate': 0.01704186404793856, 'subsample': 0.678234567334155, 'columsample\_bytree': 0.82367}

[1] 2025-05-31 02:42:40.195 Trial 2 finished with value: 0.0028465504373314347 and parameters: {'n\_estimators': 170, 'max\_depth': 5, 'learning\_rate': 0.017740850000000002, 'subsample': 0.576970686906453, 'columsample\_bytree': 0.94530}

[1] 2025-05-31 02:42:40.268 Trial 3 finished with value: 0.00288657797709781 and parameters: {'n\_estimators': 78, 'max\_depth': 5, 'learning\_rate': 0.01931802824181636, 'subsample': 0.999178205464889, 'columsample\_bytree': 0.851280}

[1] 2025-05-31 02:42:40.350 Trial 4 finished with value: 0.00286620117730451 and parameters: {'n\_estimators': 105, 'max\_depth': 3, 'learning\_rate': 0.03675438490579108, 'subsample': 0.54112071501701983, 'columsample\_bytree': 0.630339}

[1] 2025-05-31 02:42:40.458 Trial 5 finished with value: 0.002889739631427297 and parameters: {'n\_estimators': 191, 'max\_depth': 3, 'learning\_rate': 0.03784086504788799, 'subsample': 0.91597150108493919, 'columsample\_bytree': 0.89706}

[1] 2025-05-31 02:42:40.607 Trial 6 finished with value: 0.002882572519771055 and parameters: {'n\_estimators': 219, 'max\_depth': 4, 'learning\_rate': 0.08256568904402768, 'subsample': 0.716222251508063, 'columsample\_bytree': 0.5058784}

[1] 2025-05-31 02:42:40.698 Trial 7 finished with value: 0.002855605696520965 and parameters: {'n\_estimators': 122, 'max\_depth': 3, 'learning\_rate': 0.0333697043746565, 'subsample': 0.51037956453853279, 'columsample\_bytree': 0.9972905}

[1] 2025-05-31 02:42:40.761 Trial 8 finished with value: 0.00285836032940385 and parameters: {'n\_estimators': 50, 'max\_depth': 8, 'learning\_rate': 0.04234980792173163, 'subsample': 0.500094179201768, 'columsample\_bytree': 0.960509}

[1] 2025-05-31 02:42:40.848 Trial 9 finished with value: 0.0028540267101327 and parameters: {'n\_estimators': 110, 'max\_depth': 6, 'learning\_rate': 0.08475410133112483, 'subsample': 0.5060468182928617, 'columsample\_bytree': 0.624363}

[1] 2025-05-31 02:42:41.020 Trial 10 finished with value: 0.0028851929834189226 and parameters: {'n\_estimators': 269, 'max\_depth': 8, 'learning\_rate': 0.01358040730664545, 'subsample': 0.8277107850761749, 'columsample\_bytree': 0.755}

[1] 2025-05-31 02:42:41.104 Trial 11 finished with value: 0.0028632901149032598 and parameters: {'n\_estimators': 51, 'max\_depth': 8, 'learning\_rate': 0.1055319920322211, 'subsample': 0.642727825767249, 'columsample\_bytree': 0.638728}

[1] 2025-05-31 02:42:41.222 Trial 12 finished with value: 0.002905799487135144 and parameters: {'n\_estimators': 134, 'max\_depth': 7, 'learning\_rate': 0.114994285459868992, 'subsample': 0.5056019591942171, 'columsample\_bytree': 0.6622}

[1] 2025-05-31 02:42:41.305 Trial 13 finished with value: 0.0029071717398577585 and parameters: {'n\_estimators': 52, 'max\_depth': 7, 'learning\_rate': 0.299189314020047, 'subsample': 0.5897390545185627, 'columsample\_bytree': 0.516270}

[1] 2025-05-31 02:42:41.422 Trial 14 finished with value: 0.002899929203342343 and parameters: {'n\_estimators': 148, 'max\_depth': 6, 'learning\_rate': 0.0529127198072383, 'subsample': 0.6670824509413309, 'columsample\_bytree': 0.7163}

[1] 2025-05-31 02:42:41.500 Trial 15 finished with value: 0.002878049463329404 and parameters: {'n\_estimators': 99, 'max\_depth': 6, 'learning\_rate': 0.02752643300190594, 'subsample': 0.5753268080809403, 'columsample\_bytree': 0.57658}

[1] 2025-05-31 02:42:41.618 Trial 16 finished with value: 0.002887733115616 and parameters: {'n\_estimators': 95, 'max\_depth': 8, 'learning\_rate': 0.0507562728076618, 'subsample': 0.8035920437412287, 'columsample\_bytree': 0.79530}

[1] 2025-05-31 02:42:41.736 Trial 17 finished with value: 0.0028956058877688013 and parameters: {'n\_estimators': 156, 'max\_depth': 6, 'learning\_rate': 0.02360955452950046, 'subsample': 0.638897945283587, 'columsample\_bytree': 0.709}

[1] 2025-05-31 02:42:41.881 Trial 18 finished with value: 0.002937912688238124 and parameters: {'n\_estimators': 251, 'max\_depth': 7, 'learning\_rate': 0.1371114295441203, 'subsample': 0.7347902698461259, 'columsample\_bytree': 0.5702}

[1] 2025-05-31 02:42:41.995 Trial 19 finished with value: 0.002861030180702741 and parameters: {'n\_estimators': 122, 'max\_depth': 4, 'learning\_rate': 0.06811290205376727, 'subsample': 0.553309799702592, 'columsample\_bytree': 0.89816}

[1] 2025-05-31 02:42:42.068 A new study created in memory with name: no-name\_f737cf01-09a4-4b6d-61c187860f1

[1] 2025-05-31 02:42:42.178 Trial 0 finished with value: 0.003221952777558774 and parameters: {'n\_estimators': 119, 'max\_depth': 3, 'learning\_rate': 0.02854548777558028, 'subsample': 0.58877013316146856, 'columsample\_bytree': 0.84492}

[1] 2025-05-31 02:42:42.243 Trial 1 finished with value: 0.0032122123597098989 and parameters: {'n\_estimators': 52, 'max\_depth': 4, 'learning\_rate': 0.0477050009609153, 'subsample': 0.359599196084391, 'columsample\_bytree': 0.981253}

[1] 2025-05-31 02:42:42.340 Trial 2 finished with value: 0.003228085659776148 and parameters: {'n\_estimators': 171, 'max\_depth': 6, 'learning\_rate': 0.125698284308323, 'subsample': 0.50704697628293, 'columsample\_bytree': 0.536923}

[1] 2025-05-31 02:42:42.411 Trial 3 finished with value: 0.003228085659776148 and parameters: {'n\_estimators': 75, 'max\_depth': 6, 'learning\_rate': 0.1620305012268053, 'subsample': 0.8951792074731073, 'columsample\_bytree': 0.66881076}

[1] 2025-05-31 02:42:42.532 Trial 4 finished with value: 0.00321958448169775 and parameters: {'n\_estimators': 229, 'max\_depth': 4, 'learning\_rate': 0.0151924434342183, 'subsample': 0.8831128590403392, 'columsample\_bytree': 0.89176}

[1] 2025-05-31 02:42:42.648 Trial 5 finished with value: 0.0032154387283395 and parameters: {'n\_estimators': 119, 'max\_depth': 3, 'learning\_rate': 0.01589160345510259, 'subsample': 0.645550093818033, 'columsample\_bytree': 0.94290}

[1] 2025-05-31 02:42:42.775 Trial 6 finished with value: 0.003210261960586655 and parameters: {'n\_estimators': 242, 'max\_depth': 8, 'learning\_rate': 0.1019949771877967, 'subsample': 0.96989573369813205, 'columsample\_bytree': 0.80271}

[1] 2025-05-31 02:42:42.903 Trial 7 finished with value: 0.00320945912318527 and parameters: {'n\_estimators': 257, 'max\_depth': 3, 'learning\_rate': 0.0809545844501037, 'subsample': 0.7026573673900017, 'columsample\_bytree': 0.619988}

[1] 2025-05-31 02:42:43.016 Trial 8 finished with value: 0.002878049463329404 and parameters: {'n\_estimators': 99, 'max\_depth': 6, 'learning\_rate': 0.02752643300190594, 'subsample': 0.5753268080809403, 'columsample\_bytree': 0.57658}

[1] 2025-05-31 02:42:43.087 Trial 9 finished with value: 0.00232186401902863 and parameters: {'n\_estimators': 73, 'max\_depth': 5, 'learning\_rate': 0.0509079917497479, 'subsample': 0.489438192141398, 'columsample\_bytree': 0.575199}

[1] 2025-05-31 02:42:43.260 Trial 10 finished with value: 0.003209145315623465 and parameters: {'n\_estimators': 300, 'max\_depth': 8, 'learning\_rate': 0.2717380941390527, 'subsample': 0.989424063497113, 'columsample\_bytree': 0.753692}

[1] 2025-05-31 02:42:43.427 Trial 11 finished with value: 0.003212407980769569 and parameters: {'n\_estimators': 288, 'max\_depth': 8, 'learning\_rate': 0.2475984531219384, 'subsample': 0.9844291255123391, 'columsample\_bytree': 0.75615}

[1] 2025-05-31 02:42:43.594 Trial 12 finished with value: 0.00320881708403344 and parameters: {'n\_estimators': 300, 'max\_depth': 8, 'learning\_rate': 0.123558187340896, 'subsample': 0.987692555600213, 'columsample\_bytree': 0.77795}

[1] 2025-05-31 02:42:43.761 Trial 13 finished with value: 0.0031869553948383548 and parameters: {'n\_estimators': 298, 'max\_depth': 7, 'learning\_rate': 0.28401260476697934, 'subsample': 0.821119746059072, 'columsample\_bytree': 0.6921}

[1] 2025-05-31 02:42:44.051 Trial 14 finished with value: 0.00323095412318527 and parameters: {'n\_estimators': 206, 'max\_depth': 7, 'learning\_rate': 0.0285337955354606, 'subsample': 0.77956736397940753, 'columsample\_bytree': 0.9435}

[1] 2025-05-31 02:42:44.333 Trial 15 finished with value: 0.003223705727755874 and parameters: {'n\_estimators': 176, 'max\_depth': 6, 'learning\_rate': 0.02854548777558028, 'subsample': 0.58877013316146856, 'columsample\_bytree': 0.84492}

[1] 2025-05-31 02:42:44.488 Trial 16 finished with value: 0.0032021036931711976 and parameters: {'n\_estimators': 275, 'max\_depth': 7, 'learning\_rate': 0.02888773829957342, 'subsample': 0.507781031635051, 'columsample\_bytree': 0.815408}

[1] 2025-05-31 02:42:44.636 Trial 17 finished with value: 0.0032093440866960464 and parameters: {'n\_estimators': 229, 'max\_depth': 7, 'learning\_rate': 0.0634703282421572, 'subsample': 0.728886760789563, 'columsample\_bytree': 0.6401}

[1] 2025-05-31 02:42:44.929 Trial 0 finished with value: 0.002451636717470534 and parameters: {'n\_estimators': 176, 'max\_depth': 6, 'learning\_rate': 0.07180700697956428, 'subsample': 0.9552669559015485, 'columsample\_bytree': 0.85863}

[1] 2025-05-31 02:42:45.067 Trial 1 finished with value: 0.002452545072097269 and parameters: {'n\_estimators': 277, 'max\_depth': 3, 'learning\_rate': 0.057881031635051, 'subsample': 0.880528153653071, 'columsample\_bytree': 0.815408}

[1] 2025-05-31 02:42:45.181 Trial 2 finished with value: 0.002449580536356107 and parameters: {'n\_estimators': 212, 'max\_depth': 5, 'learning\_rate': 0.0409318445445367, 'subsample': 0.614595894910709, 'columsample\_bytree': 0.575770}

[1] 2025-05-31 02:42:45.286 Trial 3 finished with value: 0.002453607215206512 and parameters: {'n\_estimators': 176, 'max\_depth': 4, 'learning\_rate': 0.02888773829957342, 'subsample': 0.9697512646136707, 'columsample\_bytree': 0.57451}

[1] 2025-05-31 02:42:45.411 Trial 4 finished with value: 0.00244576756761302004 and parameters: {'n\_estimators': 243, 'max\_depth': 7, 'learning\_rate': 0.04420951086968629, 'subsample': 0.7708751307438526, 'columsample\_bytree': 0.647277}

[1] 2025-05-31 02:42:45.481 Trial 5 finished with value: 0.002453604405106033 and parameters: {'n\_estimators': 74, 'max\_depth': 7, 'learning\_rate': 0.0150436921404122, 'subsample': 0.9714533844369952, 'columsample\_bytree': 0.659295}

[1] 2025-05-31 02:42:45.615 Trial 6 finished with value: 0.002454175318749541 and parameters: {'n\_estimators': 262, 'max\_depth': 3, 'learning\_rate': 0.0240390816946165, 'subsample': 0.9760957378122043, 'columsample\_bytree': 0.850350}

[1] 2025-05-31 02:42:45.726 Trial 7 finished with value: 0.002447214803630115 and parameters: {'n\_estimators': 91, 'max\_depth': 7, 'learning\_rate': 0.028472920965503745, 'subsample': 0.687419446634031, 'columsample\_bytree': 0.960211}

[1] 2025-05-31 02:42:45.837 Trial 8 finished with value: 0.0024523795536734754 and parameters: {'n\_estimators': 192, 'max\_depth': 4, 'learning\_rate': 0.1264089517593537, 'subsample': 0.8369066313273475, 'columsample\_bytree': 0.633442}

[1] 2025-05-31 02:42:45.925 Trial 9 finished with value: 0.0024495803837323687 and parameters: {'n\_estimators': 134, 'max\_depth': 7, 'learning\_rate': 0.1476271944744433, 'subsample': 0.592721095756933, 'columsample\_bytree': 0.571959}

[1] 2025-05-31 02:42:45.960 Trial 10 finished with value: 0.0024117492532993 and parameters: {'n\_estimators': 123, 'max\_depth': 8, 'learning\_rate': 0.2289698231745806, 'subsample': 0.50278667060794, 'columsample\_bytree': 0.501340}

[1] 2025-05-31 02:42:46.175 Trial 11 finished with value: 0.0024055343871902827 and parameters: {'n\_estimators': 131, 'max\_depth': 8, 'learning\_rate': 0.010536783514564669, 'subsample': 0.7419984190453838, 'columsample\_bytree': 0.6667}

[1] 2025-05-31 02:42:46.412 Trial 12 finished with value: 0.002439316807011186 and parameters: {'n\_estimators': 234, 'max\_depth': 7, 'learning\_rate': 0.01787491670819028, 'subsample': 0.5815781670272374, 'columsample\_bytree': 0.5449}

[1] 2025-05-31 02:42:46.515 Trial 13 finished with value: 0.002451731716771783 and parameters: {'n\_estimators': 226, 'max\_depth': 5, 'learning\_rate': 0.01763781492371055, 'subsample': 0.5481956044652654, 'columsample\_bytree': 0.5108}

[1] 2025-05-31 02:42:46.606 Trial 14 finished with value: 0.002447214803630115 and parameters: {'n\_estimators': 91, 'max\_depth': 7, 'learning\_rate': 0.028472920965503745, 'subsample': 0.6874146634031, 'columsample\_bytree': 0.5018}

[1] 2025-05-31 02:42:48.442 Trial 15 finished with value: 0.002447582193536333 and parameters: {'n\_estimators': 295, 'max\_depth': 4, 'learning\_rate': 0.02793376196775077, 'subsample': 0.6567321046610582, 'columsample\_bytree': 0.7233}

[1] 2025-05-31 02:42:48.693 Trial 17 finished with value: 0.0024410537036506302 and parameters: {'n\_estimators': 254, 'max\_depth': 4, 'learning\_rate': 0.029556260022403643, 'subsample': 0.58477443049908, 'columsample\_bytree': 0.58764}

[1] 2025-05-31 02:42:49.066 Trial 18 finished with value: 0.002442777583460673 and parameters: {'n\_estimators': 214, 'max\_depth': 5, 'learning\_rate': 0.0810999591790794, 'subsample': 0.6646398317324032, 'columsample\_bytree': 0.9605}

[1] 2025-05-31 02:42:49.406 A new study created in memory with name: no-name\_99e4811-f775-495c-a885-f34600f7564

[1] 2025-05-31 02:42:49.523 Trial 0 finished with value: 0.00193498656023103 and parameters: {'n\_estimators': 159, 'max\_depth': 7, 'learning\_rate': 0.1044217352789598, 'subsample': 0.7767211925206793, 'columsample\_bytree': 0.71563}

[1] 2025-05-31 02:42:49.634 Trial 1 finished with value: 0.00193289202959869 and parameters: {'n\_estimators': 186, 'max\_depth': 6, 'learning\_rate': 0.013392769453676037, 'subsample': 0.9662305173186157, 'columsample\_bytree': 0.9025}

[1] 2025-05-31 02:42:49.704 Trial 2 finished with value: 0.001916736799400155 and parameters: {'n\_estimators': 65, 'max\_depth': 5, 'learning\_rate': 0.01619330141711655, 'subsample': 0.62583055173186157, 'columsample\_bytree': 0.980407}

[1] 2025-05-31 02:42:49.792 Trial 3 finished with value: 0.001909174099400155 and parameters: {'n\_estimators': 136, 'max\_depth': 5, 'learning\_rate': 0.031994482662636058, 'subsample': 0.739980824695586, 'columsample\_bytree': 0.5287502}

[1] 2025-05-31 02:42:49.896 Trial 4 finished with value: 0.001909174089400155 and parameters: {'n\_estimators': 231, 'max\_depth': 3, 'learning\_rate': 0.02888773829957342, 'subsample': 0.9697512646284894, 'columsample\_bytree': 0.7911784}

[1] 2025-05-31 02:42:50.171 Trial 6 finished with value: 0.00191614284682174945 and parameters: {'n\_estimators': 227, 'max\_depth': 7, 'learning\_rate': 0.03817207256473677, 'subsample': 0.78604684129559, 'columsample\_bytree': 0.564433}

[1] 2025-05-31 02:42:50.239 Trial 7 finished with value: 0.00199851880687577815 and parameters: {'n\_estimators': 60, 'max\_depth': 7, 'learning\_rate': 0.066161500423734234, 'subsample': 0.83637056202326213, 'columsample\_bytree': 0.9025342}

[1] 2025-05-31 02:42:50..346 Trial 8 finished with value: 0.001912362721904364 and parameters: {'n\_estimators': 183, 'max\_depth': 7, 'learning\_rate': 0.037006152602326213, 'subsample': 0.9990696312632413, 'columsample\_bytree': 0.60172}

[1] 2025-05-31 02:42:50..473 Trial 9 finished with value: 0.0018214730336974 and parameters: {'n\_estimators': 260, 'max\_depth': 6, 'learning\_rate': 0.29028957920352522, 'subsample': 0.6042823942694895, 'columsample\_bytree': 0.6112569}

[1] 2025-05-31 02:42:51..634 Trial 10 finished with value: 0.0019108246287261236 and parameters: {'n\_estimators': 281, 'max\_depth': 8, 'learning\_rate': 0.2437463103996217, 'subsample': 0.639785456718207, 'columsample\_bytree': 0.636161}

[1] 2025-05-31 02:42:50..721 Trial 11 finished with value: 0.001901299646628185 and parameters: {'n\_estimators': 66, 'max\_depth': 6, 'learning\_rate': 0.08617699677473873, 'subsample': 0.8820069372320313, 'columsample\_bytree': 0.80084}

[1] 2025-05-31 02:42:50..870 Trial 12 finished with value: 0.00185595512618287 and parameters: {'n\_estimators': 232, 'max\_depth': 6, 'learning\_rate': 0.1239916457522955, 'subsample': 0.6659424080001921, 'columsample\_bytree': 0.868567}

[1] 2025-05-31 02:42:51..023 Trial 13 finished with value: 0.0018371075098190405 and parameters: {'n\_estimators': 234, 'max\_depth': 6, 'learning\_rate': 0.1486614157010486, 'subsample': 0.6621965085794012, 'columsample\_bytree': 0.85607}

[1] 2025-05-31 02:42:51..169 Trial 14 finished with value: 0.001764560034517682 and parameters: {'n\_estimators': 237, 'max\_depth': 5, 'learning\_rate': 0.06162102392196268, 'subsample': 0.5443635371185089, 'columsample\_bytree': 0.7363}

[1] 2025-05-31 02:42

[I] 2025-05-31 02:42:57,215] Trial 0 finished with value: 0.0034973678106898084 and parameters: {'n\_estimators': 60, 'max\_depth': 7, 'learning\_rate': 0.029172750432247474, 'subsample': 0.58124053839042988, 'colsample\_bytree': 0.695983}

[I] 2025-05-31 02:42:57,413] Trial 1 finished with value: 0.0034499365308425 and parameters: {'n\_estimators': 263, 'max\_depth': 6, 'learning\_rate': 0.01737428120196304, 'subsample': 0.6984150345731213, 'colsample\_bytree': 0.8320067}

[I] 2025-05-31 02:42:57,539] Trial 2 finished with value: 0.003462947888360793 and parameters: {'n\_estimators': 249, 'max\_depth': 6, 'learning\_rate': 0.035700033410308984, 'subsample': 0.689915704066816, 'colsample\_bytree': 0.6343}

[I] 2025-05-31 02:42:57,640] Trial 3 finished with value: 0.003481724325095557 and parameters: {'n\_estimators': 160, 'max\_depth': 3, 'learning\_rate': 0.0176516841944614, 'subsample': 0.5416042238748687, 'colsample\_bytree': 0.79457}

[I] 2025-05-31 02:42:57,783] Trial 4 finished with value: 0.00387856354050023 and parameters: {'n\_estimators': 225, 'max\_depth': 7, 'learning\_rate': 0.15566551070379795, 'subsample': 0.627461786728331, 'colsample\_bytree': 0.9280980}

[I] 2025-05-31 02:42:57,819] Trial 5 finished with value: 0.00348049825422836 and parameters: {'n\_estimators': 54, 'max\_depth': 7, 'learning\_rate': 0.2952374164243506, 'subsample': 0.6895086955267938, 'colsample\_bytree': 0.6797012}

[I] 2025-05-31 02:42:57,922] Trial 6 finished with value: 0.003459210868925109 and parameters: {'n\_estimators': 166, 'max\_depth': 7, 'learning\_rate': 0.0813216884394167, 'subsample': 0.7435816527687605, 'colsample\_bytree': 0.848331}

[I] 2025-05-31 02:42:58,034] Trial 7 finished with value: 0.003475153369151506 and parameters: {'n\_estimators': 182, 'max\_depth': 6, 'learning\_rate': 0.07725696527247384, 'subsample': 0.7029504001483534, 'colsample\_bytree': 0.583824}

[I] 2025-05-31 02:42:58,158] Trial 8 finished with value: 0.003464529839962449 and parameters: {'n\_estimators': 185, 'max\_depth': 8, 'learning\_rate': 0.0113130344351642, 'subsample': 0.65889714920661, 'colsample\_bytree': 0.83571}

[I] 2025-05-31 02:42:58,235] Trial 9 finished with value: 0.003468232422064477 and parameters: {'n\_estimators': 93, 'max\_depth': 3, 'learning\_rate': 0.01798636258568143, 'subsample': 0.750774700583156, 'colsample\_bytree': 0.7723073}

[I] 2025-05-31 02:42:58,402] Trial 10 finished with value: 0.0034593404513081923 and parameters: {'n\_estimators': 298, 'max\_depth': 4, 'learning\_rate': 0.1952980761764736, 'subsample': 0.946227312067166, 'colsample\_bytree': 0.9939}

[I] 2025-05-31 02:42:58,553] Trial 11 finished with value: 0.003055247278986985 and parameters: {'n\_estimators': 248, 'max\_depth': 5, 'learning\_rate': 0.12505014049247, 'subsample': 0.618870339724669, 'colsample\_bytree': 0.95372}

[I] 2025-05-31 02:42:58,699] Trial 12 finished with value: 0.0034408370626950760 and parameters: {'n\_estimators': 236, 'max\_depth': 5, 'learning\_rate': 0.03618917724020166, 'subsample': 0.5307063632107562, 'colsample\_bytree': 0.9214}

[I] 2025-05-31 02:42:58,842] Trial 13 finished with value: 0.003462714740797964 and parameters: {'n\_estimators': 215, 'max\_depth': 5, 'learning\_rate': 0.037546846446190395, 'subsample': 0.5181413404710537, 'colsample\_bytree': 0.89338}

[I] 2025-05-31 02:42:59,990] Trial 14 finished with value: 0.0034597167582914644 and parameters: {'n\_estimators': 217, 'max\_depth': 8, 'learning\_rate': 0.1386257652816277, 'subsample': 0.532896126638, 'colsample\_bytree': 0.911297976}

[I] 2025-05-31 02:42:59,104] Trial 15 finished with value: 0.0034758703696340213 and parameters: {'n\_estimators': 128, 'max\_depth': 4, 'learning\_rate': 0.04703741584143796, 'subsample': 0.564378481070435, 'colsample\_bytree': 0.9947}

[I] 2025-05-31 02:42:59,308] Trial 16 finished with value: 0.00344921134408135 and parameters: {'n\_estimators': 214, 'max\_depth': 5, 'learning\_rate': 0.0874753837164438493, 'subsample': 0.501032943709999, 'colsample\_bytree': 0.503760}

[I] 2025-05-31 02:42:59,574] Trial 17 finished with value: 0.00347567038650673 and parameters: {'n\_estimators': 299, 'max\_depth': 4, 'learning\_rate': 0.2654075338490378, 'subsample': 0.6262334951459066, 'colsample\_bytree': 0.91251}

[I] 2025-05-31 02:43:00,363] Trial 18 finished with value: 0.00345292788598307 and parameters: {'n\_estimators': 264, 'max\_depth': 7, 'learning\_rate': 0.0547327121135056, 'subsample': 0.5831118539760882, 'colsample\_bytree': 0.7212}

[I] 2025-05-31 02:43:00,522] Trial 19 finished with value: 0.003493343926813432 and parameters: {'n\_estimators': 140, 'max\_depth': 8, 'learning\_rate': 0.1410957728063044, 'subsample': 0.791106807008155, 'colsample\_bytree': 0.943754}

[I] 2025-05-31 02:43:01,179] Trial 0 finished with value: 0.001515134740033719 and parameters: {'n\_estimators': 241, 'max\_depth': 3, 'learning\_rate': 0.02954646956648807, 'subsample': 0.8658724141543774, 'colsample\_bytree': 0.661952}

[I] 2025-05-31 02:43:01,872] Trial 1 finished with value: 0.00153314380617439 and parameters: {'n\_estimators': 177, 'max\_depth': 5, 'learning\_rate': 0.18705747342787574, 'subsample': 0.787975935159797, 'colsample\_bytree': 0.89056}

[I] 2025-05-31 02:43:02,037] Trial 2 finished with value: 0.0014886751983439 and parameters: {'n\_estimators': 165, 'max\_depth': 7, 'learning\_rate': 0.1117557052398597, 'subsample': 0.5021951420781983, 'colsample\_bytree': 0.8542967}

[I] 2025-05-31 02:43:02,219] Trial 3 finished with value: 0.0015312744313805742 and parameters: {'n\_estimators': 190, 'max\_depth': 4, 'learning\_rate': 0.04146665692489516, 'subsample': 0.692337709129084, 'colsample\_bytree': 0.5593}

[I] 2025-05-31 02:43:02,407] Trial 4 finished with value: 0.001515283171266199 and parameters: {'n\_estimators': 163, 'max\_depth': 6, 'learning\_rate': 0.044223716580474155, 'subsample': 0.8114690333042833, 'colsample\_bytree': 0.8336}

[I] 2025-05-31 02:43:02,571] Trial 5 finished with value: 0.00151680970583343 and parameters: {'n\_estimators': 218, 'max\_depth': 3, 'learning\_rate': 0.016308417832914524, 'subsample': 0.7741694008128641, 'colsample\_bytree': 0.6533}

[I] 2025-05-31 02:43:02,646] Trial 6 finished with value: 0.00149388633783907 and parameters: {'n\_estimators': 80, 'max\_depth': 8, 'learning\_rate': 0.0811354612537267, 'subsample': 0.683026306085237, 'colsample\_bytree': 0.9482295}

[I] 2025-05-31 02:43:02,789] Trial 7 finished with value: 0.001516632096418858 and parameters: {'n\_estimators': 75, 'max\_depth': 5, 'learning\_rate': 0.10869843520637983, 'subsample': 0.8379463826159375, 'colsample\_bytree': 0.9011}

[I] 2025-05-31 02:43:02,860] Trial 8 finished with value: 0.001514881094496438 and parameters: {'n\_estimators': 95, 'max\_depth': 4, 'learning\_rate': 0.2543255732769555, 'subsample': 0.8131049623311675, 'colsample\_bytree': 0.895231}

[I] 2025-05-31 02:43:02,952] Trial 9 finished with value: 0.0015344071176782 and parameters: {'n\_estimators': 100, 'max\_depth': 6, 'learning\_rate': 0.1832166310984444, 'subsample': 0.6521298081734139, 'colsample\_bytree': 0.510354}

[I] 2025-05-31 02:43:03,064] Trial 10 finished with value: 0.015785725672252551 and parameters: {'n\_estimators': 133, 'max\_depth': 8, 'learning\_rate': 0.0192936081542112, 'subsample': 0.5819586008822976, 'colsample\_bytree': 0.72932}

[I] 2025-05-31 02:43:03,154] Trial 11 finished with value: 0.0014855935938747388 and parameters: {'n\_estimators': 64, 'max\_depth': 8, 'learning\_rate': 0.0933876434216508, 'subsample': 0.9427750782632074, 'colsample\_bytree': 0.99944}

[I] 2025-05-31 02:43:03,263] Trial 12 finished with value: 0.0015018562366927949 and parameters: {'n\_estimators': 125, 'max\_depth': 7, 'learning\_rate': 0.0917208181207222, 'subsample': 0.998209881082641, 'colsample\_bytree': 0.79166}

[I] 2025-05-31 02:43:03,366] Trial 13 finished with value: 0.0014541293584474542 and parameters: {'n\_estimators': 50, 'max\_depth': 5, 'learning\_rate': 0.11322327177310423, 'subsample': 0.509748903416203, 'colsample\_bytree': 0.97428]

[I] 2025-05-31 02:43:03,455] Trial 14 finished with value: 0.001466266387625507 and parameters: {'n\_estimators': 61, 'max\_depth': 7, 'learning\_rate': 0.13740359352063892, 'subsample': 0.9593077892437379, 'colsample\_bytree': 0.989110}

[I] 2025-05-31 02:43:03,541] Trial 15 finished with value: 0.001486543054950946 and parameters: {'n\_estimators': 50, 'max\_depth': 8, 'learning\_rate': 0.61294305045075515, 'subsample': 0.573986951540977, 'colsample\_bytree': 0.99711}

[I] 2025-05-31 02:43:03,645] Trial 16 finished with value: 0.0015122621191103983 and parameters: {'n\_estimators': 110, 'max\_depth': 7, 'learning\_rate': 0.027414499142647437, 'subsample': 0.932280212608113, 'colsample\_bytree': 0.952}

[I] 2025-05-31 02:43:03,741] Trial 17 finished with value: 0.001483684104432398 and parameters: {'n\_estimators': 90, 'max\_depth': 8, 'learning\_rate': 0.06150230940618049, 'subsample': 0.71027606033097, 'colsample\_bytree': 0.76693}

[I] 2025-05-31 02:43:03,848] Trial 18 finished with value: 0.00148435360090099 and parameters: {'n\_estimators': 93, 'max\_depth': 6, 'learning\_rate': 0.057965705069335801, 'subsample': 0.696195597793918, 'colsample\_bytree': 0.73135}

[I] 2025-05-31 02:43:03,961] Trial 19 finished with value: 0.001520252126727682 and parameters: {'n\_estimators': 134, 'max\_depth': 6, 'learning\_rate': 0.028121686198891355, 'subsample': 0.596815933910954, 'colsample\_bytree': 0.67711}

[I] 2025-05-31 02:43:04,053] A new study created in memory with name: no-name-5ce27a5d-95df-4c6e-abd0-9bc6734a892

[I] 2025-05-31 02:43:04,178] Trial 0 finished with value: 0.00429987815633808 and parameters: {'n\_estimators': 173, 'max\_depth': 4, 'learning\_rate': 0.01568303738689417, 'subsample': 0.5896702221326294, 'colsample\_bytree': 0.83722}

[I] 2025-05-31 02:43:04,291] Trial 1 finished with value: 0.00435076644339085 and parameters: {'n\_estimators': 192, 'max\_depth': 6, 'learning\_rate': 0.19270811781207222, 'subsample': 0.67892852528598, 'colsample\_bytree': 0.8737455}

[I] 2025-05-31 02:43:04,491] Trial 2 finished with value: 0.004285952061111919 and parameters: {'n\_estimators': 270, 'max\_depth': 5, 'learning\_rate': 0.03087048169210460539, 'subsample': 0.649215420781983, 'colsample\_bytree': 0.8542967}

[I] 2025-05-31 02:43:04,520] Trial 3 finished with value: 0.004251138224654594 and parameters: {'n\_estimators': 59, 'max\_depth': 8, 'learning\_rate': 0.03638040870526392, 'subsample': 0.680408901460539, 'colsample\_bytree': 0.5096476}

[I] 2025-05-31 02:43:04,617] Trial 4 finished with value: 0.00428369318926704 and parameters: {'n\_estimators': 147, 'max\_depth': 6, 'learning\_rate': 0.01864675445366922, 'subsample': 0.868970812933704, 'colsample\_bytree': 0.72309}

[I] 2025-05-31 02:43:04,704] Trial 5 finished with value: 0.004298563391565883 and parameters: {'n\_estimators': 138, 'max\_depth': 5, 'learning\_rate': 0.0424439743436309, 'subsample': 0.942443747343609, 'colsample\_bytree': 0.8131352}

[I] 2025-05-31 02:43:04,827] Trial 6 finished with value: 0.0043008716816921 and parameters: {'n\_estimators': 210, 'max\_depth': 7, 'learning\_rate': 0.080367204069113, 'subsample': 0.985503236368423, 'colsample\_bytree': 0.984788}

[I] 2025-05-31 02:43:04,893] Trial 7 finished with value: 0.00421526816575832 and parameters: {'n\_estimators': 65, 'max\_depth': 4, 'learning\_rate': 0.1984462992767592, 'subsample': 0.8302091080971518, 'colsample\_bytree': 0.8106896}

[I] 2025-05-31 02:43:05,012] Trial 8 finished with value: 0.004285928064095455 and parameters: {'n\_estimators': 50, 'max\_depth': 8, 'learning\_rate': 0.134032304505075515, 'subsample': 0.5874851372188217, 'colsample\_bytree': 0.6042}

[I] 2025-05-31 02:43:05,082] Trial 9 finished with value: 0.00425331734090661 and parameters: {'n\_estimators': 65, 'max\_depth': 3, 'learning\_rate': 0.299703678399105, 'subsample': 0.8741303794372774, 'colsample\_bytree': 0.8753338}

[I] 2025-05-31 02:43:05,182] Trial 10 finished with value: 0.004369224687059635 and parameters: {'n\_estimators': 103, 'max\_depth': 3, 'learning\_rate': 0.2781727595597512, 'subsample': 0.7717535167650725, 'colsample\_bytree': 0.974944}

[I] 2025-05-31 02:43:05,266] Trial 11 finished with value: 0.004355190867504069 and parameters: {'n\_estimators': 51, 'max\_depth': 3, 'learning\_rate': 0.1395040202582496, 'subsample': 0.809032780715958, 'colsample\_bytree': 0.692746

[I] 2025-05-31 02:43:05,372] Trial 12 finished with value: 0.004279524920817726 and parameters: {'n\_estimators': 100, 'max\_depth': 5, 'learning\_rate': 0.03674867467546381, 'subsample': 0.871568936626787, 'colsample\_bytree': 0.88834}

[I] 2025-05-31 02:43:05,480] Trial 13 finished with value: 0.0042624953799491219 and parameters: {'n\_estimators': 86, 'max\_depth': 4, 'learning\_rate': 0.083067204469113, 'subsample': 0.985503236368423, 'colsample\_bytree': 0.92046}

[I] 2025-05-31 02:43:05,589] Trial 14 finished with value: 0.0042874485998339 and parameters: {'n\_estimators': 129, 'max\_depth': 3, 'learning\_rate': 0.13432034043309319, 'subsample': 0.791762382327143957, 'colsample\_bytree': 0.6584044}

[I] 2025-05-31 02:43:05,612] Trial 15 finished with value: 0.004285920640456544 and parameters: {'n\_estimators': 223, 'max\_depth': 7, 'learning\_rate': 0.022572746104619226, 'subsample': 0.5878451372188217, 'colsample\_bytree': 0.6042}

[I] 2025-05-31 02:43:05,682] Trial 16 finished with value: 0.004203317340407374508 and parameters: {'n\_estimators': 65, 'max\_depth': 3, 'learning\_rate': 0.299703678399105, 'subsample': 0.8741303794372774, 'colsample\_bytree': 0.8753338}

[I] 2025-05-31 02:43:05,752] Trial 17 finished with value: 0.004285920617305489 and parameters: {'n\_estimators': 103, 'max\_depth': 3, 'learning\_rate': 0.2781727595597512, 'subsample': 0.7717535167650725, 'colsample\_bytree': 0.974944}

[I] 2025-05-31 02:43:05,839] Trial 18 finished with value: 0.00423697153409661 and parameters: {'n\_estimators': 166, 'max\_depth': 4, 'learning\_rate': 0.299703678399105, 'subsample': 0.8741303794372774, 'colsample\_bytree': 0.6214685}

[I] 2025-05-31 02:43:06,225] Trial 19 finished with value: 0.004387346823875153 and parameters: {'n\_estimators': 292, 'max\_depth': 3, 'learning\_rate': 0.292933680225137, 'subsample': 0.874808518102738, 'colsample\_bytree': 0.73390}

[I] 2025-05-31 02:43:06,302] A new study created in memory with name: no-name-7edfb22-eed1-4a75-b818-44b629c69166

[I] 2025-05-31 02:43:06,405] Trial 0 finished with value: 0.003438504386817926 and parameters: {'n\_estimators': 78, 'max\_depth': 5, 'learning\_rate': 0.05473271217726, 'subsample': 0.68980205142042098, 'colsample\_bytree': 0.864340

[I] 2025-05-31 02:43:06,500] Trial 1 finished with value: 0.00344285939389482 and parameters: {'n\_estimators': 50, 'max\_depth': 8, 'learning\_rate': 0.0135789246813426, 'subsample': 0.924539014162767, 'colsample\_bytree': 0.77829

[I] 2025-05-31 02:43:06,577] Trial 2 finished with value: 0.003442803470367974 and parameters: {'n\_estimators': 110, 'max\_depth': 3, 'learning\_rate': 0.057302890363654572, 'subsample': 0.841791401436388, 'colsample\_bytree': 0.629631}

[I] 2025-05-31 02:43:06,703] Trial 3 finished with value: 0.0034337529072104 and parameters: {'n\_estimators': 239, 'max\_depth': 4, 'learning\_rate': 0.0105284952016654075, 'subsample': 0.897234412857627, 'colsample\_bytree': 0.546335}

[I] 2025-05-31 02:43:06,839] Trial 4 finished with value: 0.00346371340589661 and parameters: {'n\_estimators': 291, 'max\_depth': 8, 'learning\_rate': 0.050457700012892, 'subsample': 0.50595536720367556, 'colsample\_bytree': 0.5886174}

[I] 2025-05-31 02:43:07,194] Trial 5 finished with value: 0.00344695509507174796 and parameters: {'n\_estimators': 96, 'max\_depth': 5, 'learning\_rate': 0.18342755667476722, 'subsample': 0.553184083203532, 'colsample\_bytree': 0.535633}

[I] 2025-05-31 02:43:07,989] Trial 6 finished with value: 0.003418799872468171 and parameters: {'n\_estimators': 87, 'max\_depth': 6, 'learning\_rate': 0.01267173151563522, 'subsample': 0.6060495674430408, 'colsample\_bytree': 0.711935}

[I] 2025-05-31 02:43:07,075] Trial 7 finished with value: 0.003423870176091918 and parameters: {'n\_estimators': 114, 'max\_depth': 7, 'learning\_rate': 0.01494113957005183, 'subsample': 0.82703732648339, 'colsample\_bytree': 0.57291}

[I] 2025-05-31 02:43:07,208] Trial 8 finished with value: 0.003442035657556028 and parameters: {'n\_estimators': 275, 'max\_depth': 5, 'learning\_rate': 0.0105261975041393292, 'subsample': 0.5465401070201613, 'colsample\_bytree': 0.7821}

[I] 2025-05-31 02:43:07,283] Trial 9 finished with value: 0.00344198827842652 and parameters: {'n\_estimators': 96, 'max\_depth': 3, 'learning\_rate': 0.010257950415634529, 'subsample': 0.655052641453898, 'colsample\_bytree': 0.828933}

[I] 2025-05-31 02:43:07,449] Trial 10 finished with value: 0.003447747479703625 and parameters: {'n\_estimators': 229, 'max\_depth': 8, 'learning\_rate': 0.172810511894285, 'subsample': 0.65156399333633638, 'colsample\_bytree': 0.656356}

[I] 2025-05-31 02:43:07,616] Trial 11 finished with value: 0.0034996789109214537, 'subsample': 0.701999591914537, 'colsample\_bytree': 0.89141

[I] 2025-05-31 02:43:07,803] Trial 12 finished with value: 0.0034996789102046256 and parameters: {'n\_estimators': 184, 'max\_depth': 7, 'learning\_rate': 0.1915029318720707, 'subsample': 0.594704058985448, 'colsample\_bytree': 0.987892}

[I] 2025-05-31 02:43:07,810] A new study created in memory with name: no-name-9eb6f971-d6d9-4435-8742-1a047a040947

[I] 2025-05-31 02:43:11,267] Trial 0 finished with value: 0.00437424393265765 and parameters: {'n\_estimators': 255, 'max\_depth': 4, 'learning\_rate': 0.03856255624411564, 'subsample': 0.6911742812378746, 'colsample\_bytree': 0.6028}

[I] 2025-05-31 02:43:11,345] Trial 1 finished with value: 0.0043695295836916 and parameters: {'n\_estimators': 94, 'max\_depth': 8, 'learning\_rate': 0.07030521716632524, 'subsample': 0.8187918550933209, 'colsample\_bytree': 0.5802471}

[I] 2025-05-31 02:43:11,419] Trial 2 finished with value: 0.004356833845162 and parameters: {'n\_estimators': 81, 'max\_depth': 5, 'learning\_rate': 0.0879789714031815, 'subsample': 0.6285605090568353, 'colsample\_bytree': 0.8718451}

[I] 2025-05-31 02:43:11,559] Trial 3 finished with value: 0.004347059326568573 and parameters: {'n\_estimators': 296, 'max\_depth': 7, 'learning\_rate': 0.02708035661354524, 'subsample': 0.5

[1] 2025-05-31 02:43:16.968 Trial 8 finished with value: 0.005858770406690402 and parameters: {'n\_estimators': 122, 'max\_depth': 3, 'learning\_rate': 0.0422483921972786, 'subsample': 0.9901164719133344, 'colsample\_bytree': 0.610294}, [1] 2025-05-31 02:43:17.059 Trial 9 finished with value: 0.005853128620555446 and parameters: {'n\_estimators': 146, 'max\_depth': 7, 'learning\_rate': 0.26942847081339893, 'subsample': 0.509285707611803, 'colsample\_bytree': 0.624808}, [1] 2025-05-31 02:43:17.179 Trial 10 finished with value: 0.005858446976138446 and parameters: {'n\_estimators': 148, 'max\_depth': 6, 'learning\_rate': 0.018244588967101495, 'subsample': 0.6383039950799543, 'colsample\_bytree': 0.7577}, [1] 2025-05-31 02:43:17.314 Trial 11 finished with value: 0.005865650248226768 and parameters: {'n\_estimators': 202, 'max\_depth': 8, 'learning\_rate': 0.048781751679327256, 'subsample': 0.71598036363232199, 'colsample\_bytree': 0.7866}, [1] 2025-05-31 02:43:17.445 Trial 12 finished with value: 0.0058580248352268 and parameters: {'n\_estimators': 197, 'max\_depth': 6, 'learning\_rate': 0.0324037172471748, 'subsample': 0.7706953692918783, 'colsample\_bytree': 0.84141}, [1] 2025-05-31 02:43:17.534 Trial 13 finished with value: 0.0058658042024641731 and parameters: {'n\_estimators': 50, 'max\_depth': 6, 'learning\_rate': 0.0104590790249598, 'subsample': 0.9301857645771046, 'colsample\_bytree': 0.66792}, [1] 2025-05-31 02:43:17.649 Trial 14 finished with value: 0.0058669402656935 and parameters: {'n\_estimators': 145, 'max\_depth': 5, 'learning\_rate': 0.0320437172471748, 'subsample': 0.6535032677272753, 'colsample\_bytree': 0.838138}, [1] 2025-05-31 02:43:17.795 Trial 15 finished with value: 0.005851482898133142 and parameters: {'n\_estimators': 209, 'max\_depth': 4, 'learning\_rate': 0.0104590790249598, 'subsample': 0.75936435495035, 'colsample\_bytree': 0.683730}, [1] 2025-05-31 02:43:17.993 Trial 16 finished with value: 0.005850162898694796 and parameters: {'n\_estimators': 290, 'max\_depth': 4, 'learning\_rate': 0.00914718042563575, 'subsample': 0.5713366586890926, 'colsample\_bytree': 0.67242}, [1] 2025-05-31 02:43:18.162 Trial 17 finished with value: 0.00586997859507685 and parameters: {'n\_estimators': 286, 'max\_depth': 4, 'learning\_rate': 0.008459936328761592, 'subsample': 0.6010362470785504, 'colsample\_bytree': 0.705163}, [1] 2025-05-31 02:43:18.320 Trial 18 finished with value: 0.00589555933920831 and parameters: {'n\_estimators': 289, 'max\_depth': 4, 'learning\_rate': 0.00723406720245399, 'subsample': 0.727974090549329, 'colsample\_bytree': 0.58605}, [1] 2025-05-31 02:43:18.461 Trial 19 finished with value: 0.0058692888876516 and parameters: {'n\_estimators': 223, 'max\_depth': 4, 'learning\_rate': 0.14919937422808672, 'subsample': 0.874535734946437, 'colsample\_bytree': 0.70614}, [1] 2025-05-31 02:43:18.608 A new study created in memory with name: no-name-31407756-6485-41ab-8781-73d3d8e7880, [1] 2025-05-31 02:43:18.700 Trial 0 finished with value: 0.008794393138524507 and parameters: {'n\_estimators': 71, 'max\_depth': 5, 'learning\_rate': 0.0178997215837068, 'subsample': 0.526400660020826, 'colsample\_bytree': 0.54662287}, [1] 2025-05-31 02:43:18.837 Trial 1 finished with value: 0.00879746125484889 and parameters: {'n\_estimators': 266, 'max\_depth': 6, 'learning\_rate': 0.0910002781583926, 'subsample': 0.5535493852052591, 'colsample\_bytree': 0.913335}, [1] 2025-05-31 02:43:18.973 Trial 2 finished with value: 0.008796854989422806 and parameters: {'n\_estimators': 173, 'max\_depth': 6, 'learning\_rate': 0.05225566794457203613, 'subsample': 0.7441019163061273, 'colsample\_bytree': 0.62177}, [1] 2025-05-31 02:43:19.043 Trial 3 finished with value: 0.00879659740304246 and parameters: {'n\_estimators': 142, 'max\_depth': 7, 'learning\_rate': 0.018805633087254063, 'subsample': 0.9805391371775883, 'colsample\_bytree': 0.95048}, [1] 2025-05-31 02:43:19.184 Trial 4 finished with value: 0.008796779758068128 and parameters: {'n\_estimators': 232, 'max\_depth': 4, 'learning\_rate': 0.1322142817942165, 'subsample': 0.9894876917617043, 'colsample\_bytree': 0.571802}, [1] 2025-05-31 02:43:19.299 Trial 5 finished with value: 0.008799593422470495 and parameters: {'n\_estimators': 213, 'max\_depth': 7, 'learning\_rate': 0.0301923978973259165, 'subsample': 0.8088964793408908, 'colsample\_bytree': 0.62014}, [1] 2025-05-31 02:43:19.407 Trial 6 finished with value: 0.00879569740304246 and parameters: {'n\_estimators': 193, 'max\_depth': 6, 'learning\_rate': 0.0298125481463712, 'subsample': 0.516826716564877, 'colsample\_bytree': 0.79208}, [1] 2025-05-31 02:43:19.529 Trial 7 finished with value: 0.008795653104162244 and parameters: {'n\_estimators': 226, 'max\_depth': 4, 'learning\_rate': 0.05678884628881604, 'subsample': 0.8562752871478101, 'colsample\_bytree': 0.716434}, [1] 2025-05-31 02:43:19.639 Trial 8 finished with value: 0.00879526723628507 and parameters: {'n\_estimators': 189, 'max\_depth': 4, 'learning\_rate': 0.01060669266628951, 'subsample': 0.587353129956103, 'colsample\_bytree': 0.57251}, [1] 2025-05-31 02:43:19.743 Trial 9 finished with value: 0.008790757131402093 and parameters: {'n\_estimators': 169, 'max\_depth': 5, 'learning\_rate': 0.011688167578971263, 'subsample': 0.9068023652039316, 'colsample\_bytree': 0.96554}, [1] 2025-05-31 02:43:19.862 Trial 10 finished with value: 0.00879914977562057 and parameters: {'n\_estimators': 50, 'max\_depth': 3, 'learning\_rate': 0.2284433903261742, 'subsample': 0.664356531076078, 'colsample\_bytree': 0.5004769}, [1] 2025-05-31 02:43:19.951 Trial 11 finished with value: 0.0087954922785603 and parameters: {'n\_estimators': 94, 'max\_depth': 4, 'learning\_rate': 0.0100366469988391, 'subsample': 0.6144158362898646, 'colsample\_bytree': 0.519700}, [1] 2025-05-31 02:43:20.041 Trial 12 finished with value: 0.008796307347569573 and parameters: {'n\_estimators': 120, 'max\_depth': 5, 'learning\_rate': 0.09922417850427727, 'subsample': 0.650235066012527, 'colsample\_bytree': 0.732075}, [1] 2025-05-31 02:43:20.139 Trial 13 finished with value: 0.008795221686426968 and parameters: {'n\_estimators': 61, 'max\_depth': 3, 'learning\_rate': 0.018677894627020627, 'subsample': 0.5711728015901829, 'colsample\_bytree': 0.65191}, [1] 2025-05-31 02:43:20.227 Trial 14 finished with value: 0.0087947828367768807 and parameters: {'n\_estimators': 66, 'max\_depth': 3, 'learning\_rate': 0.02190456410323379, 'subsample': 0.5024040432542043, 'colsample\_bytree': 0.664579}, [1] 2025-05-31 02:43:20.322 Trial 15 finished with value: 0.0087954508579054946 and parameters: {'n\_estimators': 85, 'max\_depth': 3, 'learning\_rate': 0.025697786055353766, 'subsample': 0.720492642913873, 'colsample\_bytree': 0.80309}, [1] 2025-05-31 02:43:20.424 Trial 16 finished with value: 0.00879443278976515 and parameters: {'n\_estimators': 116, 'max\_depth': 5, 'learning\_rate': 0.06709409433746744, 'subsample': 0.5008424032474489, 'colsample\_bytree': 0.697999}, [1] 2025-05-31 02:43:20.534 Trial 17 finished with value: 0.0087955336498438 and parameters: {'n\_estimators': 120, 'max\_depth': 5, 'learning\_rate': 0.20820344571425164, 'subsample': 0.678246090116396, 'colsample\_bytree': 0.83683}, [1] 2025-05-31 02:43:20.641 Trial 18 finished with value: 0.008794545432899632 and parameters: {'n\_estimators': 109, 'max\_depth': 6, 'learning\_rate': 0.06285297005391897, 'subsample': 0.54606868194415151, 'colsample\_bytree': 0.86605}, [1] 2025-05-31 02:43:20.757 Trial 19 finished with value: 0.00879519788348768 and parameters: {'n\_estimators': 148, 'max\_depth': 7, 'learning\_rate': 0.13248010027572084, 'subsample': 0.6091311460237572, 'colsample\_bytree': 0.689485}, [1] 2025-05-31 02:43:20.839 A new study created in memory with name: no-name-31e31b-d6c7-f459-852a-1d19ccf97c9, [1] 2025-05-31 02:43:20.950 Trial 0 finished with value: 0.0087959931932401 and parameters: {'n\_estimators': 94, 'max\_depth': 4, 'learning\_rate': 0.06847414475806861, 'subsample': 0.760402358253162, 'colsample\_bytree': 0.9681950}, [1] 2025-05-31 02:43:21.052 Trial 1 finished with value: 0.005866695842835071 and parameters: {'n\_estimators': 77, 'max\_depth': 7, 'learning\_rate': 0.010139346390310253, 'subsample': 0.5993867973756624, 'colsample\_bytree': 0.988259}, [1] 2025-05-31 02:43:21.157 Trial 2 finished with value: 0.005840321038508685 and parameters: {'n\_estimators': 186, 'max\_depth': 3, 'learning\_rate': 0.8130422641467367, 'subsample': 0.63740720869394, 'colsample\_bytree': 0.609090}, [1] 2025-05-31 02:43:21.238 Trial 3 finished with value: 0.005870842001756814 and parameters: {'n\_estimators': 101, 'max\_depth': 4, 'learning\_rate': 0.0384288166717524, 'subsample': 0.793850193523267, 'colsample\_bytree': 0.98020}, [1] 2025-05-31 02:43:21.332 Trial 4 finished with value: 0.00580520430172079494 and parameters: {'n\_estimators': 135, 'max\_depth': 6, 'learning\_rate': 0.0278465678476162, 'subsample': 0.5139016544022624, 'colsample\_bytree': 0.598955}, [1] 2025-05-31 02:43:21.452 Trial 5 finished with value: 0.005859182656940106 and parameters: {'n\_estimators': 223, 'max\_depth': 6, 'learning\_rate': 0.26258584285984256, 'subsample': 0.7677809879794814, 'colsample\_bytree': 0.97121}, [1] 2025-05-31 02:43:21.571 Trial 6 finished with value: 0.00586357475842563258 and parameters: {'n\_estimators': 215, 'max\_depth': 4, 'learning\_rate': 0.05405478470421262, 'subsample': 0.68361030050747452, 'colsample\_bytree': 0.674246}, [1] 2025-05-31 02:43:21.693 Trial 7 finished with value: 0.005910085276394189 and parameters: {'n\_estimators': 219, 'max\_depth': 4, 'learning\_rate': 0.100097963847023567, 'subsample': 0.5231421920404856, 'colsample\_bytree': 0.707065}, [1] 2025-05-31 02:43:21.762 Trial 8 finished with value: 0.005854817423732857 and parameters: {'n\_estimators': 70, 'max\_depth': 5, 'learning\_rate': 0.0853098278687797, 'subsample': 0.68276330821337, 'colsample\_bytree': 0.74264498}, [1] 2025-05-31 02:43:21.854 Trial 9 finished with value: 0.00587935062905875 and parameters: {'n\_estimators': 109, 'max\_depth': 5, 'learning\_rate': 0.0274693628178215, 'subsample': 0.813048000510103, 'colsample\_bytree': 0.8124988}, [1] 2025-05-31 02:43:22.022 Trial 10 finished with value: 0.00586787350350303 and parameters: {'n\_estimators': 285, 'max\_depth': 3, 'learning\_rate': 0.018107651458174792, 'subsample': 0.794781539213026, 'colsample\_bytree': 0.5114}, [1] 2025-05-31 02:43:22.147 Trial 11 finished with value: 0.0058466469987706231 and parameters: {'n\_estimators': 166, 'max\_depth': 8, 'learning\_rate': 0.03095645768494752, 'subsample': 0.5008385360368331, 'colsample\_bytree': 0.5628}, [1] 2025-05-31 02:43:22.257 Trial 12 finished with value: 0.005859182656940106 and parameters: {'n\_estimators': 167, 'max\_depth': 8, 'learning\_rate': 0.1449437916523725, 'subsample': 0.6342424250514508, 'colsample\_bytree': 0.55552}, [1] 2025-05-31 02:43:22.399 Trial 13 finished with value: 0.00585803404180523 and parameters: {'n\_estimators': 171, 'max\_depth': 8, 'learning\_rate': 0.027042615761722526, 'subsample': 0.5876986090132321, 'colsample\_bytree': 0.6099}, [1] 2025-05-31 02:43:22.555 Trial 14 finished with value: 0.0058497916924827 and parameters: {'n\_estimators': 253, 'max\_depth': 7, 'learning\_rate': 0.04522437862360212, 'subsample': 0.6410959953886748, 'colsample\_bytree': 0.623858}, [1] 2025-05-31 02:43:22.670 Trial 15 finished with value: 0.00586037425477233 and parameters: {'n\_estimators': 134, 'max\_depth': 3, 'learning\_rate': 0.015428153811881127, 'subsample': 0.6938120076235353, 'colsample\_bytree': 0.8227}, [1] 2025-05-31 02:43:22.799 Trial 16 finished with value: 0.00585931016942269 and parameters: {'n\_estimators': 188, 'max\_depth': 7, 'learning\_rate': 0.1204502630291612, 'subsample': 0.553712437724794, 'colsample\_bytree': 0.581203}, [1] 2025-05-31 02:43:22.918 Trial 17 finished with value: 0.00584239324561939 and parameters: {'n\_estimators': 148, 'max\_depth': 6, 'learning\_rate': 0.0274693630932404, 'subsample': 0.5006808340646103, 'colsample\_bytree': 0.5115}, [1] 2025-05-31 02:43:22.932 Trial 18 finished with value: 0.0058596203234954586 and parameters: {'n\_estimators': 134, 'max\_depth': 6, 'learning\_rate': 0.06590437024702424, 'subsample': 0.693650512012038, 'colsample\_bytree': 0.65919}, [1] 2025-05-31 02:43:23.167 Trial 19 finished with value: 0.0058594619830353784 and parameters: {'n\_estimators': 200, 'max\_depth': 5, 'learning\_rate': 0.0190004040708192, 'subsample': 0.571647953765947, 'colsample\_bytree': 0.5048}, [1] 2025-05-31 02:43:23.282 A new study created in memory with name: no-name-2bb8-bab5-481d-8111-54a3725d36fe, [1] 2025-05-31 02:43:23.390 Trial 0 finished with value: 0.0048153964542507 and parameters: {'n\_estimators': 111, 'max\_depth': 4, 'learning\_rate': 0.041299020715090125, 'subsample': 0.80055955366566045, 'colsample\_bytree': 0.644918}, [1] 2025-05-31 02:43:23.470 Trial 1 finished with value: 0.00478841026251546 and parameters: {'n\_estimators': 101, 'max\_depth': 3, 'learning\_rate': 0.060388191157260826, 'subsample': 0.64655526009531, 'colsample\_bytree': 0.958961}, [1] 2025-05-31 02:43:23.568 Trial 2 finished with value: 0.00474374931859842 and parameters: {'n\_estimators': 147, 'max\_depth': 4, 'learning\_rate': 0.0510424424543771, 'subsample': 0.8163702762658605, 'colsample\_bytree': 0.783535}, [1] 2025-05-31 02:43:23.647 Trial 3 finished with value: 0.0048038011697177154 and parameters: {'n\_estimators': 95, 'max\_depth': 4, 'learning\_rate': 0.01130063267322622, 'subsample': 0.8083653193163336, 'colsample\_bytree': 0.99984}, [1] 2025-05-31 02:43:24.162 Trial 7 finished with value: 0.0047885945373932945 and parameters: {'n\_estimators': 223, 'max\_depth': 8, 'learning\_rate': 0.02746584042505272, 'subsample': 0.56113207381171, 'colsample\_bytree': 0.8045}, [1] 2025-05-31 02:43:24.296 Trial 8 finished with value: 0.004828153670564642 and parameters: {'n\_estimators': 280, 'max\_depth': 4, 'learning\_rate': 0.08050186800158277, 'subsample': 0.8173049963535764, 'colsample\_bytree': 0.704936}, [1] 2025-05-31 02:43:24.420 Trial 9 finished with value: 0.004636349648687756 and parameters: {'n\_estimators': 242, 'max\_depth': 7, 'learning\_rate': 0.0261904071514713, 'subsample': 0.73095895876281, 'colsample\_bytree': 0.716796}, [1] 2025-05-31 02:43:24.558 Trial 10 finished with value: 0.00492170048389588 and parameters: {'n\_estimators': 204, 'max\_depth': 8, 'learning\_rate': 0.2860927726761994, 'subsample': 0.5000167059687816, 'colsample\_bytree': 0.5002003}, [1] 2025-05-31 02:43:24.675 Trial 11 finished with value: 0.00490712160283877 and parameters: {'n\_estimators': 161, 'max\_depth': 7, 'learning\_rate': 0.13746449240921643, 'subsample': 0.5000169586830231, 'colsample\_bytree': 0.82684}, [1] 2025-05-31 02:43:24.799 Trial 12 finished with value: 0.004942596736052540 and parameters: {'n\_estimators': 172, 'max\_depth': 6, 'learning\_rate': 0.01461436486185462, 'subsample': 0.715429485519019, 'colsample\_bytree': 0.821084}, [1] 2025-05-31 02:43:24.884 Trial 13 finished with value: 0.00488859549108569 and parameters: {'n\_estimators': 54, 'max\_depth': 6, 'learning\_rate': 0.01846531401562015, 'subsample': 0.947080568091309, 'colsample\_bytree': 0.640756}, [1] 2025-05-31 02:43:25.024 Trial 14 finished with value: 0.004953263609343961 and parameters: {'n\_estimators': 214, 'max\_depth': 7, 'learning\_rate': 0.0510424424543711, 'subsample': 0.8163702762658605, 'colsample\_bytree': 0.783535}, [1] 2025-05-31 02:43:25.178 Trial 15 finished with value: 0.004790523931676114 and parameters: {'n\_estimators': 151, 'max\_depth': 7, 'learning\_rate': 0.08808725972709308, 'subsample': 0.731798539438553, 'colsample\_bytree': 0.89381}, [1] 2025-05-31 02:43:25.312 Trial 16 finished with value: 0.005859310169422692 and parameters: {'n\_estimators': 188, 'max\_depth': 7, 'learning\_rate': 0.01204502630291612, 'subsample': 0.561473247723882, 'colsample\_bytree': 0.88670}, [1] 2025-05-31 02:43:25.452 Trial 17 finished with value: 0.004810531128994937 and parameters: {'n\_estimators': 230, 'max\_depth': 8, 'learning\_rate': 0.08916474267619999, 'subsample': 0.60349423264324321, 'colsample\_bytree': 0.560248}, [1] 2025-05-31 02:43:25.574 Trial 18 finished with value: 0.004703136872892 and parameters: {'n\_estimators': 144, 'max\_depth': 7, 'learning\_rate': 0.17989152069985852, 'subsample': 0.7545002840354392, 'colsample\_bytree': 0.86774}, [1] 2025-05-31 02:43:25.744 Trial 19 finished with value: 0.004619198317046595 and parameters: {'n\_estimators': 257, 'max\_depth': 6, 'learning\_rate': 0.184140871723894, 'subsample': 0.7705083041468976, 'colsample\_bytree': 0.71093}, [1] 2025-05-31 02:43:26.284 A new study created in memory with name: no-name-31e6fa416-ea77-4522-9046-95b46419160, [1] 2025-05-31 02:43:26.628 Trial 0 finished with value: 0.003406283528103 and parameters: {'n\_estimators': 121, 'max\_depth': 7, 'learning\_rate': 0.1559287821668077, 'subsample': 0.838707935145232, 'colsample\_bytree': 0.726887}, [1] 2025-05-31 02:43:26.974 Trial 1 finished with value: 0.0030949789824507 and parameters: {'n\_estimators': 120, 'max\_depth': 4, 'learning\_rate': 0.01961748951988198, 'subsample': 0.5333896187996983, 'colsample\_bytree': 0.81687}, [1] 2025-05-31 02:43:27.763 Trial 2 finished with value: 0.00303088134649797 and parameters: {'n\_estimators': 131, 'max\_depth': 6, 'learning\_rate': 0.02730540972948736, 'subsample': 0.554295474873933, 'colsample\_bytree': 0.825799}, [1] 2025-05-31 02:43:28.052 Trial 3 finished with value: 0.003698949862269646 and parameters: {'n\_estimators': 260, 'max\_depth': 6, 'learning\_rate': 0.07255234352956633, 'subsample': 0.84217467197207, 'colsample\_bytree': 0.50149}, [1] 2025-05-31 02:43:28.393 Trial 4 finished with value: 0.00335984969369562 and parameters: {'n\_estimators': 300, 'max\_depth': 8, 'learning\_rate': 0.01634780406536134, 'subsample': 0.5649357589558821, 'colsample\_bytree': 0.764288}, [1] 2025-05-31 02:43:28.511 Trial 5 finished with value: 0.003039259352165523 and parameters: {'n\_estimators': 83, 'max\_depth': 4, 'learning\_rate': 0.03065355754174574, 'subsample': 0.636679864201824105, 'colsample\_bytree': 0.80889}, [1] 2025-05-31 02:43:28.674 Trial 6 finished with value: 0.00336305568292 and parameters: {'n\_estimators': 56, 'max\_depth': 8, 'learning\_rate': 0.030982030106190383, 'subsample': 0.97626948684660384, 'colsample\_bytree': 0.73833604}, [1] 2025-05-31 02:43:28.791 Trial 7 finished with value: 0.00333317592748711 and parameters: {'n\_estimators': 210, 'max\_depth': 4, 'learning\_rate': 0.070794856244021505, 'subsample': 0.817034992579270938, 'colsample\_bytree': 0.882857}, [1] 2025-05-31 02:43:28.913 Trial 8 finished with value: 0.003362663331318733 and parameters: {'n\_estimators': 252, 'max\_depth': 4, 'learning\_rate': 0.02294040399402344, 'subsample': 0.97688421510452, 'colsample\_bytree': 0.82455}, [1] 2025-05

[1] 2025-05-31 02:43:34, 858] Trial 16 finished with value: 0.003493805396970451 and parameters: {'n\_estimators': 177, 'max\_depth': 3, 'learning\_rate': 0.0204513507232047, 'subsample': 0.903587353891874, 'colsample\_bytree': 0.7126}, [1] 2025-05-31 02:43:35, 006] Trial 17 finished with value: 0.0035060895405345928 and parameters: {'n\_estimators': 248, 'max\_depth': 3, 'learning\_rate': 0.039289366553768065, 'subsample': 0.8310152345388817, 'colsample\_bytree': 0.712}, [1] 2025-05-31 02:43:35, 127] Trial 18 finished with value: 0.00350420170601136 and parameters: {'n\_estimators': 166, 'max\_depth': 4, 'learning\_rate': 0.0162339940900982, 'subsample': 0.922367882045101, 'colsample\_bytree': 0.7114}, [1] 2025-05-31 02:43:35, 261] Trial 19 finished with value: 0.0035089726340034 and parameters: {'n\_estimators': 196, 'max\_depth': 4, 'learning\_rate': 0.0761453688013408, 'subsample': 0.849341471961041, 'colsample\_bytree': 0.78275}, [1] 2025-05-31 02:43:35, 372] A new study created in memory with name: no-name-6abc54dc-432d-a42d-05f4190761, [1] 2025-05-31 02:43:35, 518] Trial 0 finished with value: 0.005224178018086458 and parameters: {'n\_estimators': 146, 'max\_depth': 3, 'learning\_rate': 0.156458294996185, 'subsample': 0.7101180074877322, 'colsample\_bytree': 0.710626}, [1] 2025-05-31 02:43:35, 611] Trial 1 finished with value: 0.005116362451033098 and parameters: {'n\_estimators': 141, 'max\_depth': 4, 'learning\_rate': 0.23663828847021087, 'subsample': 0.6576969993735202, 'colsample\_bytree': 0.895784}, [1] 2025-05-31 02:43:35, 696] Trial 2 finished with value: 0.005285438279923435 and parameters: {'n\_estimators': 117, 'max\_depth': 4, 'learning\_rate': 0.0162339940900982, 'subsample': 0.7333462016969024, 'colsample\_bytree': 0.5403}, [1] 2025-05-31 02:43:35, 757] Trial 3 finished with value: 0.005267246465203814 and parameters: {'n\_estimators': 51, 'max\_depth': 4, 'learning\_rate': 0.03450065353252054, 'subsample': 0.67868150583423, 'colsample\_bytree': 0.5973091}, [1] 2025-05-31 02:43:35, 893] Trial 4 finished with value: 0.00526964093515484 and parameters: {'n\_estimators': 288, 'max\_depth': 4, 'learning\_rate': 0.012034176212555, 'subsample': 0.39735143395090781, 'colsample\_bytree': 0.509282}, [1] 2025-05-31 02:43:36, 968] Trial 5 finished with value: 0.005189473984424389 and parameters: {'n\_estimators': 85, 'max\_depth': 8, 'learning\_rate': 0.291232303127184, 'subsample': 0.5492557124605173, 'colsample\_bytree': 0.5058188}, [1] 2025-05-31 02:43:36, 091] Trial 6 finished with value: 0.0052668793993624 and parameters: {'n\_estimators': 242, 'max\_depth': 8, 'learning\_rate': 0.07020406430865511, 'subsample': 0.7042046430865511, 'colsample\_bytree': 0.588136}, [1] 2025-05-31 02:43:36, 223] Trial 7 finished with value: 0.005342300623777815 and parameters: {'n\_estimators': 266, 'max\_depth': 8, 'learning\_rate': 0.0560209502907147, 'subsample': 0.557058771762676, 'colsample\_bytree': 0.6945757}, [1] 2025-05-31 02:43:36, 344] Trial 8 finished with value: 0.005273517516529125 and parameters: {'n\_estimators': 236, 'max\_depth': 5, 'learning\_rate': 0.11616569930070771, 'subsample': 0.5644251997367123, 'colsample\_bytree': 0.50895}, [1] 2025-05-31 02:43:36, 416] Trial 9 finished with value: 0.00534383829575245 and parameters: {'n\_estimators': 82, 'max\_depth': 5, 'learning\_rate': 0.24587396713314544, 'subsample': 0.748057412026362, 'colsample\_bytree': 0.8318972}, [1] 2025-05-31 02:43:36, 582] Trial 10 finished with value: 0.005231730096472479 and parameters: {'n\_estimators': 186, 'max\_depth': 6, 'learning\_rate': 0.1186861976145938, 'subsample': 0.8802112867049596, 'colsample\_bytree': 0.996}, [1] 2025-05-31 02:43:36, 685] Trial 11 finished with value: 0.00526759356557764 and parameters: {'n\_estimators': 113, 'max\_depth': 7, 'learning\_rate': 0.511497874271642, 'subsample': 0.2977880804993092, 'colsample\_bytree': 0.872409}, [1] 2025-05-31 02:43:36, 816] Trial 12 finished with value: 0.0051738834004301350 and parameters: {'n\_estimators': 180, 'max\_depth': 6, 'learning\_rate': 0.20374275891042434, 'subsample': 0.62010925839394783, 'colsample\_bytree': 0.97141}, [1] 2025-05-31 02:43:36, 954] Trial 13 finished with value: 0.00527133232146833 and parameters: {'n\_estimators': 187, 'max\_depth': 6, 'learning\_rate': 0.08002133678632, 'subsample': 0.6367751953000945, 'colsample\_bytree': 0.99978}, [1] 2025-05-31 02:43:37, 078] Trial 14 finished with value: 0.005261090897407458 and parameters: {'n\_estimators': 164, 'max\_depth': 3, 'learning\_rate': 0.1685949425962643, 'subsample': 0.8248388175150703, 'colsample\_bytree': 0.90163}, [1] 2025-05-31 02:43:37, 223] Trial 15 finished with value: 0.005116201045858586 and parameters: {'n\_estimators': 222, 'max\_depth': 6, 'learning\_rate': 0.0870512579001184, 'subsample': 0.6203155272686405, 'colsample\_bytree': 0.913371}, [1] 2025-05-31 02:43:37, 361] Trial 16 finished with value: 0.00517669691909833 and parameters: {'n\_estimators': 216, 'max\_depth': 5, 'learning\_rate': 0.08379748819909813, 'subsample': 0.8248413274485019, 'colsample\_bytree': 0.79486}, [1] 2025-05-31 02:43:37, 498] Trial 17 finished with value: 0.00526804764364822 and parameters: {'n\_estimators': 211, 'max\_depth': 7, 'learning\_rate': 0.024946196120233587, 'subsample': 0.6378000395022727, 'colsample\_bytree': 0.896}, [1] 2025-05-31 02:43:37, 650] Trial 18 finished with value: 0.0052657002628979174 and parameters: {'n\_estimators': 141, 'max\_depth': 7, 'learning\_rate': 0.6872991867362, 'subsample': 0.998745355694464, 'colsample\_bytree': 0.937540}, [1] 2025-05-31 02:43:37, 791] Trial 19 finished with value: 0.005265118644813922 and parameters: {'n\_estimators': 207, 'max\_depth': 4, 'learning\_rate': 0.1174294697939594, 'subsample': 0.604969524272863, 'colsample\_bytree': 0.774254}, [1] 2025-05-31 02:43:37, 922] A new study created in memory with name: no-name-f5dd507-3376-4930-93d2-432d480958da, [1] 2025-05-31 02:43:38, 066] Trial 0 finished with value: 0.0081523544474929 and parameters: {'n\_estimators': 205, 'max\_depth': 3, 'learning\_rate': 0.04860781200745572, 'subsample': 0.758231134223728, 'colsample\_bytree': 0.738763}, [1] 2025-05-31 02:43:38, 203] Trial 1 finished with value: 0.00815964124876903 and parameters: {'n\_estimators': 280, 'max\_depth': 8, 'learning\_rate': 0.10006476395462678, 'subsample': 0.736333694916101, 'colsample\_bytree': 0.712626}, [1] 2025-05-31 02:43:38, 341] Trial 2 finished with value: 0.008149387665493 and parameters: {'n\_estimators': 274, 'max\_depth': 4, 'learning\_rate': 0.0105974891207209, 'subsample': 0.666569628412524, 'colsample\_bytree': 0.8495911}, [1] 2025-05-31 02:43:38, 463] Trial 3 finished with value: 0.00812216709481703 and parameters: {'n\_estimators': 244, 'max\_depth': 8, 'learning\_rate': 0.13897845988368305, 'subsample': 0.647467848224012, 'colsample\_bytree': 0.562559}, [1] 2025-05-31 02:43:38, 533] Trial 4 finished with value: 0.008132666091238011 and parameters: {'n\_estimators': 51, 'max\_depth': 4, 'learning\_rate': 0.066483438764532, 'subsample': 0.5435025548272792, 'colsample\_bytree': 0.8772791}, [1] 2025-05-31 02:43:38, 669] Trial 5 finished with value: 0.00811650327316388 and parameters: {'n\_estimators': 143, 'max\_depth': 5, 'learning\_rate': 0.0580400334254569, 'subsample': 0.89576688764017, 'colsample\_bytree': 0.690718}, [1] 2025-05-31 02:43:38, 806] Trial 6 finished with value: 0.008146158974284774 and parameters: {'n\_estimators': 186, 'max\_depth': 4, 'learning\_rate': 0.0307791849818305, 'subsample': 0.9516356075976468, 'colsample\_bytree': 0.8496843}, [1] 2025-05-31 02:43:38, 960] Trial 7 finished with value: 0.0081493865413742 and parameters: {'n\_estimators': 114, 'max\_depth': 6, 'learning\_rate': 0.0708086951456462, 'subsample': 0.89581260670804, 'colsample\_bytree': 0.7374133}, [1] 2025-05-31 02:43:39, 629] Trial 8 finished with value: 0.0081314892906812587 and parameters: {'n\_estimators': 191, 'max\_depth': 7, 'learning\_rate': 0.020017212587325573, 'subsample': 0.853038685192455, 'colsample\_bytree': 0.837891}, [1] 2025-05-31 02:43:39, 767] Trial 9 finished with value: 0.008151261759222304 and parameters: {'n\_estimators': 134, 'max\_depth': 4, 'learning\_rate': 0.02034027453472477, 'subsample': 0.8435417651120474, 'colsample\_bytree': 0.71124}, [1] 2025-05-31 02:43:39, 916] Trial 10 finished with value: 0.008134530833805496 and parameters: {'n\_estimators': 115, 'max\_depth': 6, 'learning\_rate': 0.030879748819909813, 'subsample': 0.908983232154174, 'colsample\_bytree': 0.580236}, [1] 2025-05-31 02:43:40, 070] Trial 11 finished with value: 0.008147351833806948 and parameters: {'n\_estimators': 155, 'max\_depth': 7, 'learning\_rate': 0.02124959702710304, 'subsample': 0.8403571176408336, 'colsample\_bytree': 0.9579}, [1] 2025-05-31 02:43:41, 144] Trial 12 finished with value: 0.00813053644236618 and parameters: {'n\_estimators': 220, 'max\_depth': 7, 'learning\_rate': 0.01003664338350112, 'subsample': 0.8160969357129071, 'colsample\_bytree': 0.6502}, [1] 2025-05-31 02:43:41, 331] Trial 13 finished with value: 0.00814626832321249 and parameters: {'n\_estimators': 231, 'max\_depth': 5, 'learning\_rate': 0.01087191294904249, 'subsample': 0.7814388186461842, 'colsample\_bytree': 0.6366}, [1] 2025-05-31 02:43:41, 449] Trial 14 finished with value: 0.00815352752056187 and parameters: {'n\_estimators': 74, 'max\_depth': 7, 'learning\_rate': 0.041135921502545046, 'subsample': 0.9142341524441203, 'colsample\_bytree': 0.52448}, [1] 2025-05-31 02:43:41, 636] Trial 15 finished with value: 0.008178872397954932 and parameters: {'n\_estimators': 159, 'max\_depth': 5, 'learning\_rate': 0.01708086951456462, 'subsample': 0.89171254663136195, 'colsample\_bytree': 0.6603305}, [1] 2025-05-31 02:43:41, 839] Trial 16 finished with value: 0.0081317486284013 and parameters: {'n\_estimators': 225, 'max\_depth': 6, 'learning\_rate': 0.03401316773623572, 'subsample': 0.6992599435800595, 'colsample\_bytree': 0.631033}, [1] 2025-05-31 02:43:41, 993] Trial 17 finished with value: 0.0081623311071804952 and parameters: {'n\_estimators': 254, 'max\_depth': 5, 'learning\_rate': 0.0953607375944905, 'subsample': 0.890492165103511, 'colsample\_bytree': 0.77601}, [1] 2025-05-31 02:43:42, 159] Trial 18 finished with value: 0.008139404031835381 and parameters: {'n\_estimators': 296, 'max\_depth': 7, 'learning\_rate': 0.01471422970177184, 'subsample': 0.987703718963128, 'colsample\_bytree': 0.67311}, [1] 2025-05-31 02:43:42, 295] Trial 19 finished with value: 0.00810751607851988 and parameters: {'n\_estimators': 208, 'max\_depth': 3, 'learning\_rate': 0.2481608593264428, 'subsample': 0.5970200019541105, 'colsample\_bytree': 0.60261}, [1] 2025-05-31 02:43:42, 418] A new study created in memory with name: no-name-f0dd405-f4df-41e7-931a-cf61b4715701, [1] 2025-05-31 02:43:42, 561] Trial 0 finished with value: 0.0017763976215787 and parameters: {'n\_estimators': 177, 'max\_depth': 3, 'learning\_rate': 0.0264047746769932, 'subsample': 0.7859743873165099, 'colsample\_bytree': 0.7610}, [1] 2025-05-31 02:43:42, 677] Trial 1 finished with value: 0.001776596283956193 and parameters: {'n\_estimators': 202, 'max\_depth': 8, 'learning\_rate': 0.01811420616807074, 'subsample': 0.5619841435749374, 'colsample\_bytree': 0.9609}, [1] 2025-05-31 02:43:42, 822] Trial 2 finished with value: 0.001778263403294014 and parameters: {'n\_estimators': 207, 'max\_depth': 3, 'learning\_rate': 0.024900503596832112, 'subsample': 0.7601714996467671, 'colsample\_bytree': 0.52671}, [1] 2025-05-31 02:43:42, 937] Trial 3 finished with value: 0.00177598256156222 and parameters: {'n\_estimators': 207, 'max\_depth': 4, 'learning\_rate': 0.171149636003568854, 'subsample': 0.8382333917529579, 'colsample\_bytree': 0.72822}, [1] 2025-05-31 02:43:43, 011] Trial 4 finished with value: 0.001798339808051773 and parameters: {'n\_estimators': 84, 'max\_depth': 6, 'learning\_rate': 0.084164093102885287, 'subsample': 0.7328261248982675, 'colsample\_bytree': 0.895714}, [1] 2025-05-31 02:43:43, 096] Trial 5 finished with value: 0.0017787630040519057 and parameters: {'n\_estimators': 126, 'max\_depth': 4, 'learning\_rate': 0.0122495702710304, 'subsample': 0.84504572612625538, 'colsample\_bytree': 0.6213}, [1] 2025-05-31 02:43:43, 231] Trial 6 finished with value: 0.001778802040293642 and parameters: {'n\_estimators': 284, 'max\_depth': 6, 'learning\_rate': 0.17102583045205877, 'subsample': 0.5819145894816709, 'colsample\_bytree': 0.829366}, [1] 2025-05-31 02:43:43, 323] Trial 7 finished with value: 0.001781330526557978 and parameters: {'n\_estimators': 138, 'max\_depth': 3, 'learning\_rate': 0.017886927232735783, 'subsample': 0.6056894938560425, 'colsample\_bytree': 0.8256}, [1] 2025-05-31 02:43:43, 429] Trial 8 finished with value: 0.0017754754560709528 and parameters: {'n\_estimators': 193, 'max\_depth': 3, 'learning\_rate': 0.01708086951456462, 'subsample': 0.6709612564559158, 'colsample\_bytree': 0.87116}, [1] 2025-05-31 02:43:43, 514] Trial 9 finished with value: 0.001784234337219281 and parameters: {'n\_estimators': 201, 'max\_depth': 4, 'learning\_rate': 0.08594825954020183, 'subsample': 0.978318528629574, 'colsample\_bytree': 0.72603}, [1] 2025-05-31 02:43:43, 703] Trial 10 finished with value: 0.00177644199649425648 and parameters: {'n\_estimators': 278, 'max\_depth': 8, 'learning\_rate': 0.010048612103718123, 'subsample': 0.6573989304398343, 'colsample\_bytree': 0.990}, [1] 2025-05-31 02:43:43, 900] Trial 11 finished with value: 0.001785972940415014 and parameters: {'n\_estimators': 243, 'max\_depth': 8, 'learning\_rate': 0.03027607929914831, 'subsample': 0.5133301833761599, 'colsample\_bytree': 0.9993}, [1] 2025-05-31 02:43:44, 020] Trial 12 finished with value: 0.0017764930816261075 and parameters: {'n\_estimators': 143, 'max\_depth': 7, 'learning\_rate': 0.025951961261075, 'subsample': 0.50700128046642628, 'colsample\_bytree': 0.9129}, [1] 2025-05-31 02:43:44, 133] Trial 13 finished with value: 0.0017766980488768441 and parameters: {'n\_estimators': 134, 'max\_depth': 5, 'learning\_rate': 0.01702583045205877, 'subsample': 0.5819145894816709, 'colsample\_bytree': 0.8925}, [1] 2025-05-31 02:43:44, 217] Trial 14 finished with value: 0.001771813079477769 and parameters: {'n\_estimators': 50, 'max\_depth': 5, 'learning\_rate': 0.01619042285520845, 'subsample': 0.5264485580209526, 'colsample\_bytree': 0.89643}, [1] 2025-05-31 02:43:44, 403] Trial 15 finished with value: 0.00178024100460090096 and parameters: {'n\_estimators': 56, 'max\_depth': 5, 'learning\_rate': 0.016002881970650858, 'subsample': 0.5611784034458964, 'colsample\_bytree': 0.9302}, [1] 2025-05-31 02:43:44, 403] Trial 16 finished with value: 0.00184271631160272 and parameters: {'n\_estimators': 92, 'max\_depth': 7, 'learning\_rate': 0.111859286201819, 'subsample': 0.5170903399241201819, 'colsample\_bytree': 0.79721}, [1] 2025-05-31 02:43:44, 490] Trial 17 finished with value: 0.001766505237898892 and parameters: {'n\_estimators': 51, 'max\_depth': 5, 'learning\_rate': 0.0217493517667052, 'subsample': 0.61749937161814, 'colsample\_bytree': 0.6710359}, [1] 2025-05-31 02:43:44, 597] Trial 18 finished with value: 0.0017812463575155045 and parameters: {'n\_estimators': 103, 'max\_depth': 7, 'learning\_rate': 0.2943601937587197, 'subsample': 0.6037592882186362, 'colsample\_bytree': 0.644823}, [1] 2025-05-31 02:43:44, 724] Trial 19 finished with value: 0.0018935364660467998 and parameters: {'n\_estimators': 156, 'max\_depth': 6, 'learning\_rate': 0.28383072971972976, 'subsample': 0.6384239964177171, 'colsample\_bytree': 0.6111}, [1] 2025-05-31 02:43:44, 827] A new study created in memory with name: no-name-e028980-5729-40c0-a65fed0cd970, [1] 2025-05-31 02:43:44, 951] Trial 0 finished with value: 0.00701077135995182 and parameters: {'n\_estimators': 134, 'max\_depth': 6, 'learning\_rate': 0.09601040364049346, 'subsample': 0.6398748673308481, 'colsample\_bytree': 0.523792}, [1] 2025-05-31 02:43:45, 038] Trial 1 finished with value: 0.007005165081395805 and parameters: {'n\_estimators': 123, 'max\_depth': 7, 'learning\_rate': 0.041347853556845, 'subsample': 0.5509725518020865, 'colsample\_bytree': 0.724878}, [1] 2025-05-31 02:43:45, 102] Trial 2 finished with value: 0.0070105627940514365 and parameters: {'n\_estimators': 53, 'max\_depth': 6, 'learning\_rate': 0.0250122662810288, 'subsample': 0.7725884718757591, 'colsample\_bytree': 0.69350941}, [1] 2025-05-31 02:43:45, 171] Trial 3 finished with value: 0.007010497441147358 and parameters: {'n\_estimators': 73, 'max\_depth': 6, 'learning\_rate': 0.0470853973329339, 'subsample': 0.6203932342638148, 'colsample\_bytree': 0.80624892}, [1] 2025-05-31 02:43:45, 279] Trial 4 finished with value: 0.0070010498575625042 and parameters: {'n\_estimators': 188, 'max\_depth': 5, 'learning\_rate': 0.0149994096592167, 'subsample': 0.8986576957532353, 'colsample\_bytree': 0.672460}, [1] 2025-05-31 02:43:45, 616] Trial 5 finished with value: 0.0070121347245905 and parameters: {'n\_estimators': 222, 'max\_depth': 6, 'learning\_rate': 0.0794421235364858, 'subsample': 0.889932395611689, 'colsample\_bytree': 0.7743353}, [1] 2025-05-31 02:43:45, 723] Trial 6 finished with value: 0.0070102488866568658 and parameters: {'n\_estimators': 184, 'max\_depth': 4, 'learning\_rate': 0.19366329989107956, 'subsample': 0.69096203893636268, 'colsample\_bytree': 0.749295}, [1] 2025-05-31 02:43:45, 826] Trial 7 finished with value: 0.007010079940931639183 and parameters: {'n\_estimators': 174, 'max\_depth': 8, 'learning\_rate': 0.07995074881049873, 'subsample': 0.811354939859016, 'colsample\_bytree': 0.9061777}, [1] 2025-05-31 02:43:46, 088] Trial 8 finished with value: 0.00701004327640303445987 and parameters: {'n\_estimators': 278, 'max\_depth': 8, 'learning\_rate': 0.01400473550900443, 'subsample': 0.9896823462849968, 'colsample\_bytree': 0.5674}, [1] 2025-05-31 02:43:46, 169] Trial 11 finished with value: 0.00700403381508597 and parameters: {'n\_estimators': 297, 'max\_depth': 8, 'learning\_rate': 0.01207618644288355, 'subsample': 0.967147001897827, 'colsample\_bytree': 0.558}, [1] 2025-05-31 02:43:46, 330] Trial 12 finished with value: 0.0070104364703905 and parameters: {'n\_estimators': 293, 'max\_depth': 8, 'learning\_rate': 0.01268527458488592, 'subsample': 0.9961215885550682, 'colsample\_bytree': 0.5749}, [1] 2025-05-31 02:43:46, 481] Trial 13 finished with value: 0.00700565045431465 and parameters: {'n\_estimators': 249, 'max\_depth': 7, 'learning\_rate': 0.01979704740825158, 'subsample': 0.910304361897732, 'colsample\_bytree': 0.628}, [1] 2025-05-31 02:43:46, 592] Trial 14 finished with value: 0.0070010581978963997 and parameters: {'n\_estimators': 132, 'max\_depth': 7, 'learning\_rate': 0.02490893644524594, 'subsample': 0.95158564050972001, 'colsample\_bytree': 0.6335}, [1] 2025-0

[1] 2025-05-31 04:45:55.7103 Trial 2 finished with value: 0.0023671688177610705 and parameters: {'n\_estimators': 104, 'max\_depth': 4, 'learning\_rate': 0.1910537249069449, 'subsample': 0.607857001074649, 'colsample\_bytree': 0.60937006}

[1] 2025-05-31 02:43:55.8383 Trial 4 finished with value: 0.002367448813652587 and parameters: {'n\_estimators': 229, 'max\_depth': 7, 'learning\_rate': 0.1051667717141087, 'subsample': 0.8328127682172009, 'colsample\_bytree': 0.97274}

[1] 2025-05-31 02:43:55.9541 Trial 4 finished with value: 0.002308645079045868 and parameters: {'n\_estimators': 194, 'max\_depth': 6, 'learning\_rate': 0.2516275815781435, 'subsample': 0.694927360043153, 'colsample\_bytree': 0.8011679}

[1] 2025-05-31 02:43:56.0243 Trial 5 finished with value: 0.00227716172765481 and parameters: {'n\_estimators': 54, 'max\_depth': 3, 'learning\_rate': 0.019647526597082045, 'subsample': 0.528817838358993, 'colsample\_bytree': 0.53229}

[1] 2025-05-31 02:43:56.1433 Trial 6 finished with value: 0.002317494128752808 and parameters: {'n\_estimators': 229, 'max\_depth': 6, 'learning\_rate': 0.0166327053040684, 'subsample': 0.8043930196957029, 'colsample\_bytree': 0.63061}

[1] 2025-05-31 02:43:56.2223 Trial 7 finished with value: 0.0023524777203147587 and parameters: {'n\_estimators': 102, 'max\_depth': 3, 'learning\_rate': 0.04925168501348323, 'subsample': 0.99193777476755768, 'colsample\_bytree': 0.66110}

[1] 2025-05-31 02:43:56.3603 Trial 8 finished with value: 0.002588477554235545 and parameters: {'n\_estimators': 185, 'max\_depth': 7, 'learning\_rate': 0.05437163953105402, 'subsample': 0.6632249248099702, 'colsample\_bytree': 0.70831}

[1] 2025-05-31 02:43:56.451 Trial 9 finished with value: 0.00229977738391776 and parameters: {'n\_estimators': 135, 'max\_depth': 6, 'learning\_rate': 0.1641558936302676, 'subsample': 0.5451182061227555, 'colsample\_bytree': 0.510123}

[1] 2025-05-31 02:43:56.611 Trial 10 finished with value: 0.00227073946832368 and parameters: {'n\_estimators': 299, 'max\_depth': 8, 'learning\_rate': 0.010618064669869816, 'subsample': 0.632531107267656, 'colsample\_bytree': 0.59260}

[1] 2025-05-31 02:43:56.7173 Trial 11 finished with value: 0.0021024364279330194 and parameters: {'n\_estimators': 127, 'max\_depth': 5, 'learning\_rate': 0.28910387634651075, 'subsample': 0.5268864279889213, 'colsample\_bytree': 0.5202}

[1] 2025-05-31 02:43:56.8233 Trial 12 finished with value: 0.0023135648814652539 and parameters: {'n\_estimators': 89, 'max\_depth': 5, 'learning\_rate': 0.2985629059594889, 'subsample': 0.9012615858801808, 'colsample\_bytree': 0.5766109}

[1] 2025-05-31 02:43:56.9363 Trial 13 finished with value: 0.002201852129991893 and parameters: {'n\_estimators': 145, 'max\_depth': 5, 'learning\_rate': 0.08004962843073509, 'subsample': 0.6050966780102162, 'colsample\_bytree': 0.7180}

[1] 2025-05-31 02:43:57.0843 Trial 14 finished with value: 0.00228315838164254 and parameters: {'n\_estimators': 248, 'max\_depth': 5, 'learning\_rate': 0.0301420775757881, 'subsample': 0.7313183241341963, 'colsample\_bytree': 0.67413}

[1] 2025-05-31 02:43:57.2043 Trial 15 finished with value: 0.00256721833948162 and parameters: {'n\_estimators': 157, 'max\_depth': 7, 'learning\_rate': 0.1744170673904955, 'subsample': 0.5046974675327076, 'colsample\_bytree': 0.57661}

[1] 2025-05-31 02:43:57.3063 Trial 16 finished with value: 0.0021916434436615443 and parameters: {'n\_estimators': 99, 'max\_depth': 4, 'learning\_rate': 0.0734557392089048, 'subsample': 0.5873578397915596, 'colsample\_bytree': 0.770611}

[1] 2025-05-31 02:43:57.3933 Trial 17 finished with value: 0.0024361705345378185 and parameters: {'n\_estimators': 58, 'max\_depth': 8, 'learning\_rate': 0.03111007696651050, 'subsample': 0.9350761287019248, 'colsample\_bytree': 0.9996}

[1] 2025-05-31 02:43:57.5313 Trial 18 finished with value: 0.00284305293072867 and parameters: {'n\_estimators': 212, 'max\_depth': 6, 'learning\_rate': 0.15296864011648617, 'subsample': 0.75468295787873, 'colsample\_bytree': 0.6202}

[1] 2025-05-31 02:43:57.7373 Trial 19 finished with value: 0.00227246719943988 and parameters: {'n\_estimators': 267, 'max\_depth': 5, 'learning\_rate': 0.0320876596798774, 'subsample': 0.6649185187806208, 'colsample\_bytree': 0.50814}

[1] 2025-05-31 02:43:57.8383 A new study created in memory with name: no-name-b2282017-edc9-464f-8952-407400ee5dfe

[1] 2025-05-31 02:43:57.9393 Trial 0 finished with value: 0.00242036741805739 and parameters: {'n\_estimators': 81, 'max\_depth': 5, 'learning\_rate': 0.047037354655695, 'subsample': 0.8141879180783077, 'colsample\_bytree': 0.9474453}

[1] 2025-05-31 02:43:58.0133 Trial 1 finished with value: 0.00240045897940515414 and parameters: {'n\_estimators': 81, 'max\_depth': 7, 'learning\_rate': 0.1192647388030524, 'subsample': 0.6955070803814975, 'colsample\_bytree': 0.5725783}

[1] 2025-05-31 02:43:58.0943 Trial 2 finished with value: 0.002426962011453134 and parameters: {'n\_estimators': 103, 'max\_depth': 5, 'learning\_rate': 0.01555587134343434, 'subsample': 0.850764021255267, 'colsample\_bytree': 0.952368}

[1] 2025-05-31 02:43:58.2143 Trial 3 finished with value: 0.002408724923978162 and parameters: {'n\_estimators': 232, 'max\_depth': 3, 'learning\_rate': 0.049522706262317065, 'subsample': 0.800512516310375, 'colsample\_bytree': 0.63534}

[1] 2025-05-31 02:43:58.3063 Trial 4 finished with value: 0.0021916434436615443 and parameters: {'n\_estimators': 170, 'max\_depth': 7, 'learning\_rate': 0.1200278726412076, 'subsample': 0.6481939009513012, 'colsample\_bytree': 0.89406}

[1] 2025-05-31 02:43:58.4723 Trial 5 finished with value: 0.002416244297479296 and parameters: {'n\_estimators': 245, 'max\_depth': 4, 'learning\_rate': 0.01262450983628477, 'subsample': 0.510928266928553, 'colsample\_bytree': 0.71932}

[1] 2025-05-31 02:43:58.5923 Trial 6 finished with value: 0.00240517018234901 and parameters: {'n\_estimators': 222, 'max\_depth': 7, 'learning\_rate': 0.146696091690542, 'subsample': 0.6804706219264218, 'colsample\_bytree': 0.7879989}

[1] 2025-05-31 02:43:58.6673 Trial 7 finished with value: 0.002468817327854966 and parameters: {'n\_estimators': 75, 'max\_depth': 6, 'learning\_rate': 0.020505348287790892, 'subsample': 0.948077552422191, 'colsample\_bytree': 0.54349}

[1] 2025-05-31 02:43:58.7373 Trial 8 finished with value: 0.0024211546591711635 and parameters: {'n\_estimators': 70, 'max\_depth': 6, 'learning\_rate': 0.06087264753443764, 'subsample': 0.656981579726016, 'colsample\_bytree': 0.785806}

[1] 2025-05-31 02:43:58.8643 Trial 9 finished with value: 0.00242222693404694 and parameters: {'n\_estimators': 216, 'max\_depth': 3, 'learning\_rate': 0.0481037576351259, 'subsample': 0.9085274102177201, 'colsample\_bytree': 0.81218}

[1] 2025-05-31 02:43:58.9843 Trial 10 finished with value: 0.002446152321150498 and parameters: {'n\_estimators': 150, 'max\_depth': 8, 'learning\_rate': 0.0251317996530881, 'subsample': 0.516489300499512, 'colsample\_bytree': 0.87122}

[1] 2025-05-31 02:43:59.1523 Trial 11 finished with value: 0.002415136708402829 and parameters: {'n\_estimators': 298, 'max\_depth': 8, 'learning\_rate': 0.18008830321464, 'subsample': 0.673634090018217, 'colsample\_bytree': 0.898883}

[1] 2025-05-31 02:43:59.2753 Trial 12 finished with value: 0.0024206273536643213 and parameters: {'n\_estimators': 166, 'max\_depth': 7, 'learning\_rate': 0.1087612088679323, 'subsample': 0.712467454210526, 'colsample\_bytree': 0.6742}

[1] 2025-05-31 02:43:59.4413 Trial 13 finished with value: 0.00240234171899716 and parameters: {'n\_estimators': 198, 'max\_depth': 7, 'learning\_rate': 0.10772780627087304, 'subsample': 0.628769709899173, 'colsample\_bytree': 0.8523}

[1] 2025-05-31 02:43:59.5543 Trial 14 finished with value: 0.002415050118407952 and parameters: {'n\_estimators': 130, 'max\_depth': 7, 'learning\_rate': 0.18017337369308509, 'subsample': 0.5881161051381844, 'colsample\_bytree': 0.995723}

[1] 2025-05-31 02:43:59.6712 Trial 15 finished with value: 0.0024018010317792844 and parameters: {'n\_estimators': 257, 'max\_depth': 8, 'learning\_rate': 0.0726365957261632, 'subsample': 0.5751527023845917, 'colsample\_bytree': 0.80777}

[1] 2025-05-31 02:43:59.8413 Trial 16 finished with value: 0.002412884417056853 and parameters: {'n\_estimators': 176, 'max\_depth': 6, 'learning\_rate': 0.27136198775645154, 'subsample': 0.517996195955337, 'colsample\_bytree': 0.6003}

[1] 2025-05-31 02:43:59.9713 Trial 17 finished with value: 0.0024957090368167166 and parameters: {'n\_estimators': 187, 'max\_depth': 6, 'learning\_rate': 0.293063459321293, 'subsample': 0.571848030369825, 'colsample\_bytree': 0.54035}

[1] 2025-05-31 02:44:00.0883 Trial 18 finished with value: 0.00242334281533806 and parameters: {'n\_estimators': 132, 'max\_depth': 5, 'learning\_rate': 0.2193357598750504, 'subsample': 0.5807719002945861, 'colsample\_bytree': 0.5934}

[1] 2025-05-31 02:44:00.2053 Trial 19 finished with value: 0.002448657629331 and parameters: {'n\_estimators': 167, 'max\_depth': 4, 'learning\_rate': 0.08159432608237768, 'subsample': 0.743759184984303, 'colsample\_bytree': 0.69952}

[1] 2025-05-31 02:44:00.3193 A new study created in memory with name: no-name-3cc6b892-0af6-8a0f-e6ad-8790fe1b0a00

[1] 2025-05-31 02:44:00.4743 Trial 0 finished with value: 0.005991454854916136 and parameters: {'n\_estimators': 153, 'max\_depth': 7, 'learning\_rate': 0.0416933171685202, 'subsample': 0.8712803886148924, 'colsample\_bytree': 0.8762532}

[1] 2025-05-31 02:44:00.5544 Trial 1 finished with value: 0.00598115709785226 and parameters: {'n\_estimators': 71, 'max\_depth': 4, 'learning\_rate': 0.0664672612697501, 'subsample': 0.544929458665393, 'colsample\_bytree': 0.51144798}

[1] 2025-05-31 02:44:00.6622 Trial 2 finished with value: 0.005915159321531943 and parameters: {'n\_estimators': 105, 'max\_depth': 7, 'learning\_rate': 0.19945203744866674, 'subsample': 0.8933678214194935, 'colsample\_bytree': 0.594290}

[1] 2025-05-31 02:44:00.7303 Trial 3 finished with value: 0.0059887364561137 and parameters: {'n\_estimators': 107, 'max\_depth': 3, 'learning\_rate': 0.03629710961827, 'subsample': 0.898910900470823, 'colsample\_bytree': 0.91765933}

[1] 2025-05-31 02:44:00.8423 Trial 4 finished with value: 0.00593607237386107474652 and parameters: {'n\_estimators': 240, 'max\_depth': 5, 'learning\_rate': 0.015326856370174861, 'subsample': 0.632450797115699, 'colsample\_bytree': 0.7774}

[1] 2025-05-31 02:44:00.9173 Trial 5 finished with value: 0.005925577500343286 and parameters: {'n\_estimators': 244, 'max\_depth': 3, 'learning\_rate': 0.04776560849084585, 'subsample': 0.76105163673643517, 'colsample\_bytree': 0.91516}

[1] 2025-05-31 02:44:01.0833 Trial 6 finished with value: 0.005925577500343286 and parameters: {'n\_estimators': 194, 'max\_depth': 8, 'learning\_rate': 0.1907930126517603, 'subsample': 0.7784040219152355, 'colsample\_bytree': 0.84756}

[1] 2025-05-31 02:44:01.1723 Trial 7 finished with value: 0.00600465547441362 and parameters: {'n\_estimators': 120, 'max\_depth': 7, 'learning\_rate': 0.1273726738057197, 'subsample': 0.7997792852807922, 'colsample\_bytree': 0.867086}

[1] 2025-05-31 02:44:01.3033 Trial 8 finished with value: 0.00609484214201032 and parameters: {'n\_estimators': 266, 'max\_depth': 8, 'learning\_rate': 0.02261147293671, 'subsample': 0.7494252292159988, 'colsample\_bytree': 0.5584130464}

[1] 2025-05-31 02:44:01.4003 Trial 9 finished with value: 0.00597419366071675 and parameters: {'n\_estimators': 148, 'max\_depth': 6, 'learning\_rate': 0.015388216486233, 'subsample': 0.578188472183571, 'colsample\_bytree': 0.737756}

[1] 2025-05-31 02:44:01.4573 Trial 10 finished with value: 0.005956107935323732 and parameters: {'n\_estimators': 203, 'max\_depth': 8, 'learning\_rate': 0.09862099579159315, 'subsample': 0.9784577021987633, 'colsample\_bytree': 0.98980}

[1] 2025-05-31 02:44:01.4873 Trial 11 finished with value: 0.00592857544444428 and parameters: {'n\_estimators': 218, 'max\_depth': 5, 'learning\_rate': 0.014490051633151207, 'subsample': 0.648317309441692, 'colsample\_bytree': 0.7489}

[1] 2025-05-31 02:44:01.5173 Trial 12 finished with value: 0.005970302373861074652 and parameters: {'n\_estimators': 200, 'max\_depth': 5, 'learning\_rate': 0.024678349274695638, 'subsample': 0.66487822441387, 'colsample\_bytree': 0.71479}

[1] 2025-05-31 02:44:02.0133 Trial 13 finished with value: 0.00600168820043236 and parameters: {'n\_estimators': 132, 'max\_depth': 4, 'learning\_rate': 0.02119721049098313, 'subsample': 0.673643999323207, 'colsample\_bytree': 0.65688}

[1] 2025-05-31 02:44:02.1453 Trial 14 finished with value: 0.00602215352684512935 and parameters: {'n\_estimators': 191, 'max\_depth': 4, 'learning\_rate': 0.2960239869630378, 'subsample': 0.503287034276581, 'colsample\_bytree': 0.82738}

[1] 2025-05-31 02:44:02.2883 Trial 15 finished with value: 0.0059413342132935 and parameters: {'n\_estimators': 228, 'max\_depth': 5, 'learning\_rate': 0.0837918418934316, 'subsample': 0.719562597188676, 'colsample\_bytree': 0.79245}

[1] 2025-05-31 02:44:02.4243 Trial 16 finished with value: 0.005921096750948 and parameters: {'n\_estimators': 158, 'max\_depth': 4, 'learning\_rate': 0.1353728381711284, 'subsample': 0.813851162515604, 'colsample\_bytree': 0.6825196}

[1] 2025-05-31 02:44:02.5903 Trial 17 finished with value: 0.005915029389404945 and parameters: {'n\_estimators': 171, 'max\_depth': 4, 'learning\_rate': 0.15404881514883542, 'subsample': 0.8427448564481715, 'colsample\_bytree': 0.6749}

[1] 2025-05-31 02:44:02.7113 Trial 18 finished with value: 0.00595123693891643 and parameters: {'n\_estimators': 161, 'max\_depth': 4, 'learning\_rate': 0.13627604771995017, 'subsample': 0.8130503154969382, 'colsample\_bytree': 0.65924}

[1] 2025-05-31 02:44:02.8003 Trial 19 finished with value: 0.00591437889620323653 and parameters: {'n\_estimators': 52, 'max\_depth': 3, 'learning\_rate': 0.1209010223263255, 'subsample': 0.961641950871943, 'colsample\_bytree': 0.677576}

[1] 2025-05-31 02:44:02.9193 A new study created in memory with name: no-name-ec6b8853-4544-4d67-a06b178a24b2

[1] 2025-05-31 02:44:03.0933 Trial 0 finished with value: 0.002102056537445112 and parameters: {'n\_estimators': 292, 'max\_depth': 7, 'learning\_rate': 0.22750306733264541, 'subsample': 0.6123232814856964, 'colsample\_bytree': 0.64180}

[1] 2025-05-31 02:44:03.1933 Trial 1 finished with value: 0.00202070514739443 and parameters: {'n\_estimators': 164, 'max\_depth': 4, 'learning\_rate': 0.03307924731836683, 'subsample': 0.9916171265053808, 'colsample\_bytree': 0.54365}

[1] 2025-05-31 02:44:03.2913 Trial 2 finished with value: 0.00205915159321531943 and parameters: {'n\_estimators': 105, 'max\_depth': 7, 'learning\_rate': 0.1994520374486674, 'subsample': 0.8933678214194935, 'colsample\_bytree': 0.594290}

[1] 2025-05-31 02:44:03.3853 Trial 3 finished with value: 0.002224130629761646716 and parameters: {'n\_estimators': 140, 'max\_depth': 6, 'learning\_rate': 0.04667911656604846, 'subsample': 0.5281367407216874, 'colsample\_bytree': 0.98827611}

[1] 2025-05-31 02:44:03.4963 Trial 4 finished with value: 0.002020414044737541 and parameters: {'n\_estimators': 203, 'max\_depth': 6, 'learning\_rate': 0.014667911656604846, 'subsample': 0.880451787325369, 'colsample\_bytree': 0.6337952}

[1] 2025-05-31 02:44:03.5653 Trial 5 finished with value: 0.002020421804323886 and parameters: {'n\_estimators': 252, 'max\_depth': 6, 'learning\_rate': 0.01407960849095156064, 'subsample': 0.9784577021987633, 'colsample\_bytree': 0.811936}

[1] 2025-05-31 02:44:03.8013 Trial 6 finished with value: 0.00201972480129794 and parameters: {'n\_estimators': 282, 'max\_depth': 3, 'learning\_rate': 0.02047965704032124, 'subsample': 0.7391227249123493, 'colsample\_bytree': 0.589822}

[1] 2025-05-31 02:44:03.8953 Trial 7 finished with value: 0.0020367097897390614 and parameters: {'n\_estimators': 124, 'max\_depth': 8, 'learning\_rate': 0.0951650403430395, 'subsample': 0.65636468599769732, 'colsample\_bytree': 0.589822}

[1] 2025-05-31 02:44:03.9663 Trial 8 finished with value: 0.0020370015000569845 and parameters: {'n\_estimators': 81, 'max\_depth': 6, 'learning\_rate': 0.070177916919163, 'subsample': 0.9470787030723528, 'colsample\_bytree': 0.676563}

[1] 2025-05-31 02:44:04.0983 Trial 9 finished with value: 0.002285196134950336 and parameters: {'n\_estimators': 264, 'max\_depth': 5, 'learning\_rate': 0.1672737413506899, 'subsample': 0.830631100726233, 'colsample\_bytree': 0.665648}

[1] 2025-05-31 02:44:04.1823 Trial 10 finished with value: 0.002159708573902435 and parameters: {'n\_estimators': 50, 'max\_depth': 8, 'learning\_rate': 0.015681310073977535, 'subsample': 0.842819651623697, 'colsample\_bytree': 0.818103}

[1] 2025-05-31 02:44:04.2743 Trial 11 finished with value: 0.0022837184516174647 and parameters: {'n\_estimators': 63, 'max\_depth': 8, 'learning\_rate': 0.020497724774773514, 'subsample': 0.889919468811709, 'colsample\_bytree': 0.87866}

[1] 2025-05-31 02:44:04.4677 Trial 12 finished with value: 0.0022061207651714053 and parameters: {'n\_estimators': 90, 'max\_depth': 7, 'learning\_rate': 0.02040776540278493, 'subsample': 0.789723249182373, 'colsample\_bytree': 0.899948}

[1] 2025-05-31 02:44:04.6677 Trial 13 finished with value: 0.0022897893649852478 and parameters: {'n\_estimators': 93, 'max\_depth': 5, 'learning\_rate': 0.0204907748249133, 'subsample': 0.7152823249861793, 'colsample\_bytree': 0.734570}

[1] 2025-05-31 02:44:04.6677 Trial 14 finished with value: 0.0022047570403214774 and parameters: {'n\_estimators': 107, 'max\_depth': 5, 'learning\_rate': 0.10898184858473818, 'subsample': 0.9185688184242449, 'colsample\_bytree': 0.7307}

[1] 2025-05-31 02:44:04.7653 Trial 15 finished with value: 0.002247741793205356 and parameters: {'n\_estimators': 83, 'max\_depth': 4, 'learning\_rate': 0.0856125380349175, 'subsample': 0.805182782043246384, 'colsample\_bytree': 0.717499}

[1] 2025-05-31 02:44:04.0563 Trial 16 finished with value: 0.002024088455132103203 and parameters: {'n\_estimators': 209, 'max\_depth': 4, 'learning\_rate': 0.10031907188170815, 'subsample': 0.73808959842431, 'colsample\_bytree': 0.7475}

[1] 2025-05-31 02:44:04.2053 Trial 17 finished with value: 0.002028358133945296 and parameters: {'n\_estimators': 209, 'max\_depth': 4, 'learning\_rate': 0.1309714850853977, 'subsample': 0.7421740738002255, 'colsample\_bytree': 0.8820}

[1] 2025-05-31 02:44:04.4253 Trial 18 finished with value: 0.00

[1] 2025-05-31 02:44:13, 333 Trial 10 finished with value: 0.00482418458604712 and parameters: {'n\_estimators': 50, 'max\_depth': 5, 'learning\_rate': 0.011632765268492352, 'subsample': 0.928433768380064, 'colsample\_bytree': 0.99404}, [1] 2025-05-31 02:44:13, 494 Trial 11 finished with value: 0.00486492281717515 and parameters: {'n\_estimators': 286, 'max\_depth': 5, 'learning\_rate': 0.2437638040742398, 'subsample': 0.879261991267836, 'colsample\_bytree': 0.67067}, [1] 2025-05-31 02:44:13, 606 Trial 12 finished with value: 0.004828737799645123 and parameters: {'n\_estimators': 120, 'max\_depth': 7, 'learning\_rate': 0.01820737849042398, 'subsample': 0.839125926029312, 'colsample\_bytree': 0.6456}, [1] 2025-05-31 02:44:13, 746 Trial 13 finished with value: 0.00484145754680985 and parameters: {'n\_estimators': 225, 'max\_depth': 4, 'learning\_rate': 0.2861236972003283, 'subsample': 0.8267431531618435, 'colsample\_bytree': 0.59572}, [1] 2025-05-31 02:44:13, 911 Trial 14 finished with value: 0.00487731706089164 and parameters: {'n\_estimators': 289, 'max\_depth': 7, 'learning\_rate': 0.12122025654211784, 'subsample': 0.8160672779141045, 'colsample\_bytree': 0.7195}, [1] 2025-05-31 02:44:13, 034 Trial 15 finished with value: 0.0048024016347061929 and parameters: {'n\_estimators': 159, 'max\_depth': 4, 'learning\_rate': 0.0122025654211784, 'subsample': 0.9017018133728, 'colsample\_bytree': 0.641179}, [1] 2025-05-31 02:44:14, 139 Trial 16 finished with value: 0.0048157923891461666 and parameters: {'n\_estimators': 106, 'max\_depth': 6, 'learning\_rate': 0.019544290340735006, 'subsample': 0.993387085639374, 'colsample\_bytree': 0.770}, [1] 2025-05-31 02:44:14, 241 Trial 17 finished with value: 0.0048155425052051044 and parameters: {'n\_estimators': 106, 'max\_depth': 7, 'learning\_rate': 0.01979438033949688, 'subsample': 0.985237116451074, 'colsample\_bytree': 0.763}, [1] 2025-05-31 02:44:14, 340 Trial 18 finished with value: 0.004818489620582325 and parameters: {'n\_estimators': 84, 'max\_depth': 7, 'learning\_rate': 0.01975998015529087, 'subsample': 0.9467574920213436, 'colsample\_bytree': 0.92498}, [1] 2025-05-31 02:44:14, 431 Trial 19 finished with value: 0.00482691749906643 and parameters: {'n\_estimators': 50, 'max\_depth': 7, 'learning\_rate': 0.01476967204205737, 'subsample': 0.93183731267088, 'colsample\_bytree': 0.98542}, [1] 2025-05-31 02:44:14, 509 A new study created in memory with name: no-name-fe9d451-f753-49ff-b4a-0dede01cc3, [1] 2025-05-31 02:44:14, 617 Trial 0 finished with value: 0.00326150546428405 and parameters: {'n\_estimators': 94, 'max\_depth': 8, 'learning\_rate': 0.010702096521619303, 'subsample': 0.5311633594795409, 'colsample\_bytree': 0.53941}, [1] 2025-05-31 02:44:14, 739 Trial 1 finished with value: 0.00329987017178884 and parameters: {'n\_estimators': 211, 'max\_depth': 4, 'learning\_rate': 0.065899393160277862, 'subsample': 0.663237983138331, 'colsample\_bytree': 0.79632}, [1] 2025-05-31 02:44:14, 883 Trial 2 finished with value: 0.003207918465509196 and parameters: {'n\_estimators': 291, 'max\_depth': 7, 'learning\_rate': 0.032523179192816324, 'subsample': 0.826434878396928, 'colsample\_bytree': 0.73562}, [1] 2025-05-31 02:44:14, 966 Trial 3 finished with value: 0.00318294573341414 and parameters: {'n\_estimators': 103, 'max\_depth': 3, 'learning\_rate': 0.043163008607020747, 'subsample': 0.9581159911865436, 'colsample\_bytree': 0.79430}, [1] 2025-05-31 02:44:15, 092 Trial 4 finished with value: 0.003279761389200727 and parameters: {'n\_estimators': 232, 'max\_depth': 4, 'learning\_rate': 0.030632131973295177, 'subsample': 0.6700776504465593, 'colsample\_bytree': 0.98829}, [1] 2025-05-31 02:44:15, 201 Trial 5 finished with value: 0.003260338631315415 and parameters: {'n\_estimators': 236, 'max\_depth': 7, 'learning\_rate': 0.02787714278241297, 'subsample': 0.689318568986458, 'colsample\_bytree': 0.531982}, [1] 2025-05-31 02:44:15, 304 Trial 6 finished with value: 0.0032200992647371896 and parameters: {'n\_estimators': 159, 'max\_depth': 5, 'learning\_rate': 0.0486953572702825, 'subsample': 0.818051167112313, 'colsample\_bytree': 0.8138}, [1] 2025-05-31 02:44:15, 433 Trial 7 finished with value: 0.003252937905795627837 and parameters: {'n\_estimators': 266, 'max\_depth': 5, 'learning\_rate': 0.0102073809759627837, 'subsample': 0.627794902527453, 'colsample\_bytree': 0.546041}, [1] 2025-05-31 02:44:15, 561 Trial 8 finished with value: 0.003210970455024331 and parameters: {'n\_estimators': 255, 'max\_depth': 7, 'learning\_rate': 0.18429012948104231, 'subsample': 0.95983778728878, 'colsample\_bytree': 0.909322}, [1] 2025-05-31 02:44:15, 687 Trial 9 finished with value: 0.003267585453363764 and parameters: {'n\_estimators': 234, 'max\_depth': 5, 'learning\_rate': 0.028769497431344209, 'subsample': 0.67001954047047123, 'colsample\_bytree': 0.6516}, [1] 2025-05-31 02:44:15, 769 Trial 10 finished with value: 0.0031884640914744743 and parameters: {'n\_estimators': 51, 'max\_depth': 3, 'learning\_rate': 0.1208456609249875, 'subsample': 0.99340651645299, 'colsample\_bytree': 0.659286}, [1] 2025-05-31 02:44:15, 853 Trial 11 finished with value: 0.00317547208141049 and parameters: {'n\_estimators': 55, 'max\_depth': 3, 'learning\_rate': 0.18369927882066, 'subsample': 0.9557254125029215, 'colsample\_bytree': 0.67704367}, [1] 2025-05-31 02:44:15, 959 Trial 12 finished with value: 0.00318993313994931 and parameters: {'n\_estimators': 106, 'max\_depth': 3, 'learning\_rate': 0.0264964508891083, 'subsample': 0.90687195226016, 'colsample\_bytree': 0.705065}, [1] 2025-05-31 02:44:16, 059 Trial 13 finished with value: 0.00321426320533598 and parameters: {'n\_estimators': 55, 'max\_depth': 3, 'learning\_rate': 0.25630971354973714, 'subsample': 0.892999781466662, 'colsample\_bytree': 0.83886}, [1] 2025-05-31 02:44:16, 168 Trial 14 finished with value: 0.00320587987294856 and parameters: {'n\_estimators': 131, 'max\_depth': 4, 'learning\_rate': 0.01948657522145229, 'subsample': 0.901278727549344, 'colsample\_bytree': 0.6096}, [1] 2025-05-31 02:44:16, 267 Trial 15 finished with value: 0.00317978942637909 and parameters: {'n\_estimators': 89, 'max\_depth': 3, 'learning\_rate': 0.0560452961662018, 'subsample': 0.989634683854168, 'colsample\_bytree': 0.881141}, [1] 2025-05-31 02:44:16, 360 Trial 16 finished with value: 0.003218208269134113 and parameters: {'n\_estimators': 79, 'max\_depth': 6, 'learning\_rate': 0.03797561973660687, 'subsample': 0.8374864436308283, 'colsample\_bytree': 0.88929}, [1] 2025-05-31 02:44:16, 476 Trial 17 finished with value: 0.00327532379390704 and parameters: {'n\_estimators': 145, 'max\_depth': 4, 'learning\_rate': 0.1740225295638754, 'subsample': 0.7584612670315767, 'colsample\_bytree': 0.9876}, [1] 2025-05-31 02:44:16, 609 Trial 18 finished with value: 0.0031800716282066 and parameters: {'n\_estimators': 186, 'max\_depth': 3, 'learning\_rate': 0.10214861403752964, 'subsample': 0.952352642180567, 'colsample\_bytree': 0.88919}, [1] 2025-05-31 02:44:16, 720 Trial 19 finished with value: 0.00319548293744773 and parameters: {'n\_estimators': 126, 'max\_depth': 4, 'learning\_rate': 0.2961887554709113, 'subsample': 0.876192213528192, 'colsample\_bytree': 0.71780}, [1] 2025-05-31 02:44:16, 795 A new study created in memory with name: no-name-147a1797-fbla47c-b900-abcf03770295, [1] 2025-05-31 02:44:16, 966 Trial 0 finished with value: 0.003029400107117526 and parameters: {'n\_estimators': 271, 'max\_depth': 8, 'learning\_rate': 0.17728960225785373, 'subsample': 0.57505846651528, 'colsample\_bytree': 0.7361130}, [1] 2025-05-31 02:44:17, 079 Trial 1 finished with value: 0.002495573361920788 and parameters: {'n\_estimators': 99, 'max\_depth': 4, 'learning\_rate': 0.016283635214227637, 'subsample': 0.6762201081114245, 'colsample\_bytree': 0.53189}, [1] 2025-05-31 02:44:17, 156 Trial 2 finished with value: 0.00294323162008018 and parameters: {'n\_estimators': 102, 'max\_depth': 8, 'learning\_rate': 0.10949437472392226, 'subsample': 0.67395656998323, 'colsample\_bytree': 0.5897828}, [1] 2025-05-31 02:44:17, 294 Trial 3 finished with value: 0.00249738582593955 and parameters: {'n\_estimators': 281, 'max\_depth': 6, 'learning\_rate': 0.0154268788042371, 'subsample': 0.936441466886898, 'colsample\_bytree': 0.98491}, [1] 2025-05-31 02:44:17, 359 Trial 4 finished with value: 0.002950347194653257 and parameters: {'n\_estimators': 56, 'max\_depth': 4, 'learning\_rate': 0.0149527131896374, 'subsample': 0.62989485323344, 'colsample\_bytree': 0.70746}, [1] 2025-05-31 02:44:17, 496 Trial 5 finished with value: 0.002982891359624715 and parameters: {'n\_estimators': 286, 'max\_depth': 7, 'learning\_rate': 0.046472443989661735, 'subsample': 0.7504865518749574, 'colsample\_bytree': 0.5975}, [1] 2025-05-31 02:44:17, 566 Trial 6 finished with value: 0.0029513392089938 and parameters: {'n\_estimators': 65, 'max\_depth': 5, 'learning\_rate': 0.051931392089938, 'subsample': 0.929526076903255, 'colsample\_bytree': 0.888315}, [1] 2025-05-31 02:44:17, 633 Trial 7 finished with value: 0.002952690305272027 and parameters: {'n\_estimators': 62, 'max\_depth': 4, 'learning\_rate': 0.028556320067310504, 'subsample': 0.8486372181139616, 'colsample\_bytree': 0.747663}, [1] 2025-05-31 02:44:17, 780 Trial 8 finished with value: 0.0029803942615087 and parameters: {'n\_estimators': 289, 'max\_depth': 4, 'learning\_rate': 0.1940518375812212, 'subsample': 0.61076234626993, 'colsample\_bytree': 0.9304}, [1] 2025-05-31 02:44:17, 889 Trial 9 finished with value: 0.00298693826841153 and parameters: {'n\_estimators': 183, 'max\_depth': 5, 'learning\_rate': 0.028723726227381625, 'subsample': 0.6160744689402311, 'colsample\_bytree': 0.76132}, [1] 2025-05-31 02:44:18, 007 Trial 10 finished with value: 0.0029157590704768185 and parameters: {'n\_estimators': 147, 'max\_depth': 8, 'learning\_rate': 0.2556849967088788, 'subsample': 0.518863830722095, 'colsample\_bytree': 0.62335}, [1] 2025-05-31 02:44:18, 158 Trial 11 finished with value: 0.00291100380203579 and parameters: {'n\_estimators': 156, 'max\_depth': 8, 'learning\_rate': 0.2591536145052024, 'subsample': 0.501193601404203, 'colsample\_bytree': 0.61461}, [1] 2025-05-31 02:44:18, 275 Trial 12 finished with value: 0.0029429536120405125 and parameters: {'n\_estimators': 144, 'max\_depth': 7, 'learning\_rate': 0.1213264119748246, 'subsample': 0.5042617586529624, 'colsample\_bytree': 0.63270}, [1] 2025-05-31 02:44:18, 422 Trial 13 finished with value: 0.002993821708755184 and parameters: {'n\_estimators': 180, 'max\_depth': 7, 'learning\_rate': 0.1403041562900299, 'subsample': 0.514218930122795, 'colsample\_bytree': 0.67092}, [1] 2025-05-31 02:44:19, 601 Trial 14 finished with value: 0.0029363732816172 and parameters: {'n\_estimators': 140, 'max\_depth': 7, 'learning\_rate': 0.2980355485640525, 'subsample': 0.7359210365354655, 'colsample\_bytree': 0.500121}, [1] 2025-05-31 02:44:19, 312 Trial 15 finished with value: 0.002910521420443041 and parameters: {'n\_estimators': 222, 'max\_depth': 6, 'learning\_rate': 0.27781050647973913, 'subsample': 0.8233994150572788, 'colsample\_bytree': 0.5138}, [1] 2025-05-31 02:44:19, 592 Trial 16 finished with value: 0.002921704279136542 and parameters: {'n\_estimators': 229, 'max\_depth': 6, 'learning\_rate': 0.20638185250353364, 'subsample': 0.8126254585284771, 'colsample\_bytree': 0.5473}, [1] 2025-05-31 02:44:20, 547 Trial 17 finished with value: 0.002939458203219853 and parameters: {'n\_estimators': 224, 'max\_depth': 3, 'learning\_rate': 0.08222545623815564, 'subsample': 0.864784664831301, 'colsample\_bytree': 0.85316}, [1] 2025-05-31 02:44:20, 780 Trial 18 finished with value: 0.0029399729620507083 and parameters: {'n\_estimators': 218, 'max\_depth': 6, 'learning\_rate': 0.1912081781558075, 'subsample': 0.768043662695869, 'colsample\_bytree': 0.5498}, [1] 2025-05-31 02:44:20, 978 Trial 19 finished with value: 0.0029499347142970240 and parameters: {'n\_estimators': 245, 'max\_depth': 8, 'learning\_rate': 0.01004954538945506, 'subsample': 0.926660492783856, 'colsample\_bytree': 0.8005}, [1] 2025-05-31 02:44:21, 177 A new study created in memory with name: no-name-fb19a31d-46e97-43c5-aead-e649e921b24, [1] 2025-05-31 02:44:21, 512 Trial 0 finished with value: 0.002077201813032796 and parameters: {'n\_estimators': 226, 'max\_depth': 7, 'learning\_rate': 0.0343656649840932, 'subsample': 0.7569924579095135, 'colsample\_bytree': 0.51747}, [1] 2025-05-31 02:44:21, 585 Trial 1 finished with value: 0.002063150170494745 and parameters: {'n\_estimators': 76, 'max\_depth': 6, 'learning\_rate': 0.0273391982902553, 'subsample': 0.6426156822370391, 'colsample\_bytree': 0.873298}, [1] 2025-05-31 02:44:21, 681 Trial 2 finished with value: 0.00207849317384248 and parameters: {'n\_estimators': 154, 'max\_depth': 4, 'learning\_rate': 0.0724915266607195, 'subsample': 0.85899114537852, 'colsample\_bytree': 0.533786}, [1] 2025-05-31 02:44:21, 777 Trial 3 finished with value: 0.0020775520364640313 and parameters: {'n\_estimators': 128, 'max\_depth': 4, 'learning\_rate': 0.01543362841909492, 'subsample': 0.580805401585404488, 'colsample\_bytree': 0.6537}, [1] 2025-05-31 02:44:21, 901 Trial 4 finished with value: 0.002064208160253053 and parameters: {'n\_estimators': 254, 'max\_depth': 5, 'learning\_rate': 0.0507551969944922, 'subsample': 0.50665641202027, 'colsample\_bytree': 0.6978795}, [1] 2025-05-31 02:44:21, 979 Trial 5 finished with value: 0.002069759800650563 and parameters: {'n\_estimators': 89, 'max\_depth': 3, 'learning\_rate': 0.2982372709777944, 'subsample': 0.658221790577735, 'colsample\_bytree': 0.9372403}, [1] 2025-05-31 02:44:22, 074 Trial 6 finished with value: 0.0020719512932526 and parameters: {'n\_estimators': 144, 'max\_depth': 5, 'learning\_rate': 0.0354244785086921, 'subsample': 0.917548847224268, 'colsample\_bytree': 0.58024}, [1] 2025-05-31 02:44:22, 151 Trial 7 finished with value: 0.002064692875165455 and parameters: {'n\_estimators': 88, 'max\_depth': 3, 'learning\_rate': 0.1210349493779962, 'subsample': 0.934933404616864, 'colsample\_bytree': 0.919417}, [1] 2025-05-31 02:44:22, 241 Trial 8 finished with value: 0.002080692566189089 and parameters: {'n\_estimators': 116, 'max\_depth': 4, 'learning\_rate': 0.038694983161556, 'subsample': 0.7094490388367407, 'colsample\_bytree': 0.956891}, [1] 2025-05-31 02:44:22, 333 Trial 9 finished with value: 0.002067757897234719 and parameters: {'n\_estimators': 140, 'max\_depth': 7, 'learning\_rate': 0.016111312761204681, 'subsample': 0.9332127959122978, 'colsample\_bytree': 0.75020}, [1] 2025-05-31 02:44:22, 417 Trial 10 finished with value: 0.00206959420160216971 and parameters: {'n\_estimators': 51, 'max\_depth': 6, 'learning\_rate': 0.016708097366249328, 'subsample': 0.8116986289513055, 'colsample\_bytree': 0.82575}, [1] 2025-05-31 02:44:22, 505 Trial 11 finished with value: 0.002069759800650563 and parameters: {'n\_estimators': 89, 'max\_depth': 3, 'learning\_rate': 0.2982372709779944, 'subsample': 0.658221790577735, 'colsample\_bytree': 0.9372403}, [1] 2025-05-31 02:44:22, 653 Trial 12 finished with value: 0.0020664225345455 and parameters: {'n\_estimators': 196, 'max\_depth': 5, 'learning\_rate': 0.1739346604921744, 'subsample': 0.91744887224268, 'colsample\_bytree': 0.58024}, [1] 2025-05-31 02:44:22, 821 Trial 13 finished with value: 0.00206911672423833 and parameters: {'n\_estimators': 299, 'max\_depth': 5, 'learning\_rate': 0.19380408044343568, 'subsample': 0.616526095069503612, 'colsample\_bytree': 0.9833}, [1] 2025-05-31 02:44:22, 921 Trial 14 finished with value: 0.0020695723956186388 and parameters: {'n\_estimators': 91, 'max\_depth': 7, 'learning\_rate': 0.0750905433433032, 'subsample': 0.8361062019674936, 'colsample\_bytree': 0.899839}, [1] 2025-05-31 02:44:23, 020 Trial 15 finished with value: 0.002081162313262305 and parameters: {'n\_estimators': 90, 'max\_depth': 5, 'learning\_rate': 0.02413018231474733, 'subsample': 0.7611268103043097, 'colsample\_bytree': 0.78788}, [1] 2025-05-31 02:44:23, 149 Trial 16 finished with value: 0.00204982514242424 and parameters: {'n\_estimators': 178, 'max\_depth': 6, 'learning\_rate': 0.1150628130184478, 'subsample': 0.6016898129535047, 'colsample\_bytree': 0.906934}, [1] 2025-05-31 02:44:23, 242 Trial 17 finished with value: 0.00206478518792366 and parameters: {'n\_estimators': 74, 'max\_depth': 8, 'learning\_rate': 0.0822114037501331, 'subsample': 0.827809380492336, 'colsample\_bytree': 0.927927}, [1] 2025-05-31 02:44:23, 345 Trial 18 finished with value: 0.002080526968286386 and parameters: {'n\_estimators': 104, 'max\_depth': 3, 'learning\_rate': 0.0722507766349614, 'subsample': 0.708313045071525, 'colsample\_bytree': 0.7214802}, [1] 2025-05-31 02:44:23, 468 Trial 19 finished with value: 0.00206504235468233 and parameters: {'n\_estimators': 168, 'max\_depth': 4, 'learning\_rate': 0.185592087033027307, 'subsample': 0.99546071124405, 'colsample\_bytree': 0.99987480}, [1] 2025-05-31 02:44:24, 547 A new study created in memory with name: no-name-47dbfa2b21e477-a45d-d2a164a9318f0, [1] 2025-05-31 02:44:24, 739 Trial 0 finished with value: 0.00292637977399715, 'subsample': 0.59058877715253, 'colsample\_bytree': 0.84164114}, [1] 2025-05-31 02:44:24, 747 Trial 1 finished with value: 0.002812005126889867 and parameters: {'n\_estimators': 196, 'max\_depth': 7, 'learning\_rate': 0.1297550761012017, 'subsample': 0.7789157610911861, 'colsample\_bytree': 0.71598}, [1] 2025-05-31 02:44:23, 888 Trial 2 finished with value: 0.002977892644864494 and parameters: {'n\_estimators': 195, 'max\_depth': 5, 'learning\_rate': 0.02431734405690295, 'subsample': 0.56386858854944932, 'colsample\_bytree': 0.722736}, [1] 2025-05-31 02:44:23, 979 Trial 3 finished with value: 0.00281436591752152 and parameters: {'n\_estimators': 217, 'max\_depth': 3, 'learning\_rate': 0.015089590780852466, 'subsample': 0.779032586210629, 'colsample\_bytree': 0.8470}, [1] 2025-05-31 02:44:24, 485 Trial 4 finished with value: 0.0028745805650393715 and parameters: {'n\_estimators': 290, 'max\_depth': 6, 'learning\_rate': 0.2881856578493959, 'subsample': 0.800993739937033363, 'colsample\_bytree': 0.728686}, [1] 2025-05-31 02:44:24, 852 Trial 5 finished with value: 0.00280856199104557 and parameters: {'n\_estimators': 275, 'max\_depth': 8, 'learning\_rate': 0.299565294562959295, 'subsample': 0.635273946843039, 'colsample\_bytree': 0.9631}, [1] 2025-05-31 02:44:25, 011 Trial 12 finished with value: 0.002799092192581574 and parameters: {'n\_estimators': 242, 'max\_depth': 5, 'learning\_rate': 0.0393504806878081, 'subsample': 0.648350144833862, 'colsample\_bytree': 0.6176}, [1] 2025-05-31 02:44:25, 114 Trial 13 finished with value: 0.0027696793889195 and parameters: {'n\_estimators': 106, 'max\_depth': 7, 'learning\_rate': 0.118549049801830007, 'subsample': 0.65058170021048, 'colsample\_bytree': 0.653644}, [1] 2025-05-31 02:44:25, 218 Trial 14 finished with value: 0.

```
[1] 2025-05-31 02:44:30,499 Trial 18 finished with value: 0.001886660596537678 and parameters: {'n_estimators': 165, 'max_depth': 4, 'learning_rate': 0.063786402078484, 'subsample': 0.5834204538370287, 'colsample_bytree': 0.802331}
[1] 2025-05-31 02:44:30,611 Trial 19 finished with value: 0.0018904272932016495 and parameters: {'n_estimators': 143, 'max_depth': 6, 'learning_rate': 0.016096161841166714, 'subsample': 0.5012618700803995, 'colsample_bytree': 0.781}
[1] 2025-05-31 02:44:30,687 A new study created in memory with name: no-name-811c7ec6-b90a-4958-be3b-79942ee6f292
[1] 2025-05-31 02:44:30,685 Trial 0 finished with value: 0.0065188622338548 and parameters: {'n_estimators': 267, 'max_depth': 6, 'learning_rate': 0.1449315151907042, 'subsample': 0.6643293320174594, 'colsample_bytree': 0.70350270}
[1] 2025-05-31 02:44:30,977 Trial 1 finished with value: 0.006721733283528713 and parameters: {'n_estimators': 198, 'max_depth': 3, 'learning_rate': 0.2001199246179492, 'subsample': 0.695289789394449, 'colsample_bytree': 0.5439937
[1] 2025-05-31 02:44:31,101 Trial 2 finished with value: 0.0063001354007105365 and parameters: {'n_estimators': 223, 'max_depth': 5, 'learning_rate': 0.21104661323212614, 'subsample': 0.987785317844904, 'colsample_bytree': 0.99607
[1] 2025-05-31 02:44:31,189 Trial 3 finished with value: 0.00681712490260476 and parameters: {'n_estimators': 125, 'max_depth': 5, 'learning_rate': 0.0358946756296398, 'subsample': 0.7318290649947781, 'colsample_bytree': 0.62856
[1] 2025-05-31 02:44:31,292 Trial 4 finished with value: 0.006794997871391715 and parameters: {'n_estimators': 178, 'max_depth': 5, 'learning_rate': 0.014459218416823866, 'subsample': 0.826604247837823, 'colsample_bytree': 0.9227
[1] 2025-05-31 02:44:31,409 Trial 5 finished with value: 0.006379222464029655 and parameters: {'n_estimators': 225, 'max_depth': 6, 'learning_rate': 0.08689569807125907, 'subsample': 0.574382001693205, 'colsample_bytree': 0.51560
[1] 2025-05-31 02:44:31,560 Trial 6 finished with value: 0.00678695751822366 and parameters: {'n_estimators': 210, 'max_depth': 4, 'learning_rate': 0.03052395181736683, 'subsample': 0.849229341347078, 'colsample_bytree': 0.63719
[1] 2025-05-31 02:44:31,828 Trial 7 finished with value: 0.00678085434752773 and parameters: {'n_estimators': 296, 'max_depth': 3, 'learning_rate': 0.0627990016797892, 'subsample': 0.8167138622895918, 'colsample_bytree': 0.982646
[1] 2025-05-31 02:44:32,582 Trial 8 finished with value: 0.00628093306736767 and parameters: {'n_estimators': 239, 'max_depth': 5, 'learning_rate': 0.07495316811439422, 'subsample': 0.512750730518515007, 'colsample_bytree': 0.94342
[1] 2025-05-31 02:44:32,687 Trial 9 finished with value: 0.0068452934915215055 and parameters: {'n_estimators': 70, 'max_depth': 7, 'learning_rate': 0.024656599670203, 'subsample': 0.5091316386334236, 'colsample_bytree': 0.8952656
[1] 2025-05-31 02:44:33,711 Trial 10 finished with value: 0.00687612790227794 and parameters: {'n_estimators': 121, 'max_depth': 3, 'learning_rate': 0.27755465267166, 'subsample': 0.6563991917273189, 'colsample_bytree': 0.62856
[1] 2025-05-31 02:44:33,942 Trial 11 finished with value: 0.006793893569343791 and parameters: {'n_estimators': 170, 'max_depth': 3, 'learning_rate': 0.26443787376815145, 'subsample': 0.923254424203938, 'colsample_bytree': 0.6142
[1] 2025-05-31 02:44:34,124 Trial 12 finished with value: 0.00678205151639895 and parameters: {'n_estimators': 159, 'max_depth': 3, 'learning_rate': 0.013513213441285638, 'subsample': 0.9370205020695933, 'colsample_bytree': 0.5869
[1] 2025-05-31 02:44:34,286 Trial 13 finished with value: 0.00679471304992318 and parameters: {'n_estimators': 176, 'max_depth': 8, 'learning_rate': 0.13403665299187838, 'subsample': 0.7347507978328876, 'colsample_bytree': 0.69191
[1] 2025-05-31 02:44:34,426 Trial 14 finished with value: 0.006823184050305085 and parameters: {'n_estimators': 138, 'max_depth': 4, 'learning_rate': 0.04298667472916821, 'subsample': 0.6334312298490306, 'colsample_bytree': 0.5638
[1] 2025-05-31 02:44:34,581 Trial 15 finished with value: 0.00679451544379136 and parameters: {'n_estimators': 79, 'max_depth': 4, 'learning_rate': 0.020872585172621066, 'subsample': 0.9055200516229446, 'colsample_bytree': 0.77545
[1] 2025-05-31 02:44:34,751 Trial 16 finished with value: 0.006790815099334074 and parameters: {'n_estimators': 194, 'max_depth': 3, 'learning_rate': 0.10174206232854588, 'subsample': 0.7910882183353814, 'colsample_bytree': 0.5788
[1] 2025-05-31 02:44:35,019 Trial 17 finished with value: 0.00679515132529047 and parameters: {'n_estimators': 98, 'max_depth': 3, 'learning_rate': 0.119737977904706, 'subsample': 0.718944903458665, 'colsample_bytree': 0.66883953
[1] 2025-05-31 02:44:35,155 Trial 18 finished with value: 0.00678108942970882 and parameters: {'n_estimators': 200, 'max_depth': 4, 'learning_rate': 0.0202395879223911, 'subsample': 0.98857978329984, 'colsample_bytree': 0.50071
[1] 2025-05-31 02:44:35,271 A new study created in memory with name: no-name-1742faa-8638-4aeef-2ba7194dah2
[1] 2025-05-31 02:44:35,398 Trial 0 finished with value: 0.001280645348552552 and parameters: {'n_estimators': 174, 'max_depth': 3, 'learning_rate': 0.011581475185312542, 'subsample': 0.9828954902724993, 'colsample_bytree': 0.60418
[1] 2025-05-31 02:44:35,500 Trial 1 finished with value: 0.00128224920670528 and parameters: {'n_estimators': 157, 'max_depth': 5, 'learning_rate': 0.0388781617554557, 'subsample': 0.562070678795455, 'colsample_bytree': 0.984810
[1] 2025-05-31 02:44:35,567 Trial 2 finished with value: 0.0012831399464389317 and parameters: {'n_estimators': 60, 'max_depth': 3, 'learning_rate': 0.028123272662393164, 'subsample': 0.94577509546157651, 'colsample_bytree': 0.549618
[1] 2025-05-31 02:44:35,639 Trial 3 finished with value: 0.001793743683942347 and parameters: {'n_estimators': 76, 'max_depth': 7, 'learning_rate': 0.0128673653474686, 'subsample': 0.6011378249350942, 'colsample_bytree': 0.86154
[1] 2025-05-31 02:44:35,795 Trial 4 finished with value: 0.001280165435997778 and parameters: {'n_estimators': 216, 'max_depth': 4, 'learning_rate': 0.04283722615634836, 'subsample': 0.744730018081047, 'colsample_bytree': 0.52132
[1] 2025-05-31 02:44:35,933 Trial 5 finished with value: 0.001276012328052084 and parameters: {'n_estimators': 261, 'max_depth': 7, 'learning_rate': 0.0163613216744803, 'subsample': 0.76163216744803, 'colsample_bytree': 0.9873
[1] 2025-05-31 02:44:36,089 Trial 6 finished with value: 0.0012938387774960108 and parameters: {'n_estimators': 289, 'max_depth': 5, 'learning_rate': 0.086096861032202, 'subsample': 0.9317788403156873, 'colsample_bytree': 0.726637
[1] 2025-05-31 02:44:36,196 Trial 7 finished with value: 0.00127406305130224 and parameters: {'n_estimators': 196, 'max_depth': 3, 'learning_rate': 0.02203565668712966, 'subsample': 0.820816313830364, 'colsample_bytree': 0.66620
[1] 2025-05-31 02:44:36,335 Trial 8 finished with value: 0.0012800924113887664 and parameters: {'n_estimators': 251, 'max_depth': 7, 'learning_rate': 0.2403733940983738, 'subsample': 0.8038752994592304, 'colsample_bytree': 0.91711
[1] 2025-05-31 02:44:36,427 Trial 9 finished with value: 0.001282745293548728 and parameters: {'n_estimators': 143, 'max_depth': 4, 'learning_rate': 0.01411309529092285, 'subsample': 0.7866052872400945, 'colsample_bytree': 0.5955
[1] 2025-05-31 02:44:36,529 Trial 10 finished with value: 0.0012937927683781057 and parameters: {'n_estimators': 113, 'max_depth': 8, 'learning_rate': 0.093273526069664363, 'subsample': 0.6597720266915983, 'colsample_bytree': 0.7165
[1] 2025-05-31 02:44:36,674 Trial 11 finished with value: 0.001289340509617956 and parameters: {'n_estimators': 230, 'max_depth': 7, 'learning_rate': 0.0263310889607764, 'subsample': 0.85891055292201, 'colsample_bytree': 0.7929
[1] 2025-05-31 02:44:36,869 Trial 12 finished with value: 0.00127679760626196 and parameters: {'n_estimators': 272, 'max_depth': 6, 'learning_rate': 0.018962289798337138, 'subsample': 0.687400898781113, 'colsample_bytree': 0.6762
[1] 2025-05-31 02:44:37,037 Trial 13 finished with value: 0.00125194677818089 and parameters: {'n_estimators': 207, 'max_depth': 8, 'learning_rate': 0.0216145736729793, 'subsample': 0.850386123281049, 'colsample_bytree': 0.825
[1] 2025-05-31 02:44:37,135 Trial 14 finished with value: 0.001274343824804714 and parameters: {'n_estimators': 203, 'max_depth': 8, 'learning_rate': 0.0696822059635004, 'subsample': 0.836529577580251, 'colsample_bytree': 0.8099
[1] 2025-05-31 02:44:37,265 Trial 15 finished with value: 0.00127289342673488 and parameters: {'n_estimators': 199, 'max_depth': 6, 'learning_rate': 0.0786429884882255, 'subsample': 0.86957722693659128, 'colsample_bytree': 0.67343
[1] 2025-05-31 02:44:37,374 Trial 16 finished with value: 0.0012646940931273638 and parameters: {'n_estimators': 131, 'max_depth': 6, 'learning_rate': 0.0521581721467471, 'subsample': 0.924532651291341, 'colsample_bytree': 0.666368
[1] 2025-05-31 02:44:37,500 Trial 17 finished with value: 0.00126569052051766 and parameters: {'n_estimators': 185, 'max_depth': 4, 'learning_rate': 0.119577423704058, 'subsample': 0.5031645978899187, 'colsample_bytree': 0.66188
[1] 2025-05-31 02:44:37,603 Trial 18 finished with value: 0.0012685797901104102 and parameters: {'n_estimators': 109, 'max_depth': 4, 'learning_rate': 0.14274847698840692, 'subsample': 0.5196579543396147, 'colsample_bytree': 0.6078
[1] 2025-05-31 02:44:37,698 Trial 19 finished with value: 0.0012991929571011015 and parameters: {'n_estimators': 86, 'max_depth': 4, 'learning_rate': 0.14848865195073743, 'subsample': 0.5122480992494196, 'colsample_bytree': 0.59427
[1] 2025-05-31 02:44:37,826 A new study created in memory with name: no-name-f5939df5-3f5c-4e90-b94-03dcba36f03
[1] 2025-05-31 02:44:37,928 Trial 0 finished with value: 0.001656834170273396 and parameters: {'n_estimators': 80, 'max_depth': 3, 'learning_rate': 0.08077025739194243, 'subsample': 0.819254109770994, 'colsample_bytree': 0.914306
[1] 2025-05-31 02:44:38,037 Trial 1 finished with value: 0.001670010979533591 and parameters: {'n_estimators': 191, 'max_depth': 5, 'learning_rate': 0.0281237341949233, 'subsample': 0.8401608721812704, 'colsample_bytree': 0.827369
[1] 2025-05-31 02:44:38,117 Trial 2 finished with value: 0.001662370796007976 and parameters: {'n_estimators': 100, 'max_depth': 8, 'learning_rate': 0.0388781617554557, 'subsample': 0.8686265797622372, 'colsample_bytree': 0.8852383
[1] 2025-05-31 02:44:38,357 Trial 3 finished with value: 0.0016008563494365554 and parameters: {'n_estimators': 254, 'max_depth': 4, 'learning_rate': 0.16567802795735355, 'subsample': 0.5471467060114978, 'colsample_bytree': 0.53076
[1] 2025-05-31 02:44:38,441 Trial 4 finished with value: 0.00168093808778817 and parameters: {'n_estimators': 129, 'max_depth': 3, 'learning_rate': 0.02491409443761883, 'subsample': 0.5720196568205868, 'colsample_bytree': 0.875164
[1] 2025-05-31 02:44:38,479 Trial 5 finished with value: 0.0012801654359977755 and parameters: {'n_estimators': 216, 'max_depth': 4, 'learning_rate': 0.016563215163761985, 'subsample': 0.5928330410639594, 'colsample_bytree': 0.773251
[1] 2025-05-31 02:44:38,602 Trial 6 finished with value: 0.0016591852089711887 and parameters: {'n_estimators': 222, 'max_depth': 4, 'learning_rate': 0.04554904063659374, 'subsample': 0.8824142175365957, 'colsample_bytree': 0.6629
[1] 2025-05-31 02:44:38,704 Trial 7 finished with value: 0.0016675795392036868 and parameters: {'n_estimators': 164, 'max_depth': 6, 'learning_rate': 0.0116253049040097983, 'subsample': 0.89839933181909, 'colsample_bytree': 0.61901
[1] 2025-05-31 02:44:38,771 Trial 8 finished with value: 0.0016638349021638388 and parameters: {'n_estimators': 67, 'max_depth': 6, 'learning_rate': 0.0287481433563033, 'subsample': 0.9543579106830247, 'colsample_bytree': 0.835491
[1] 2025-05-31 02:44:38,897 Trial 9 finished with value: 0.001670954997037908 and parameters: {'n_estimators': 150, 'max_depth': 6, 'learning_rate': 0.0352736311554168, 'subsample': 0.8282271429301341, 'colsample_bytree': 0.50919
[1] 2025-05-31 02:44:39,069 Trial 10 finished with value: 0.001661179005686701 and parameters: {'n_estimators': 295, 'max_depth': 8, 'learning_rate': 0.2762414813452646, 'subsample': 0.68228832049257, 'colsample_bytree': 0.525949
[1] 2025-05-31 02:44:39,229 Trial 11 finished with value: 0.0016606288068093494 and parameters: {'n_estimators': 285, 'max_depth': 8, 'learning_rate': 0.0932735260696635858, 'subsample': 0.5179576563333376, 'colsample_bytree': 0.7165
[1] 2025-05-31 02:44:39,377 Trial 12 finished with value: 0.00162802410123928 and parameters: {'n_estimators': 250, 'max_depth': 4, 'learning_rate': 0.1352329341056348, 'subsample': 0.637258567862836, 'colsample_bytree': 0.71818
[1] 2025-05-31 02:44:39,527 Trial 13 finished with value: 0.001657653337971555 and parameters: {'n_estimators': 239, 'max_depth': 5, 'learning_rate': 0.15317479393090291, 'subsample': 0.680527217300524, 'colsample_bytree': 0.99915
[1] 2025-05-31 02:44:39,668 Trial 14 finished with value: 0.0016649789985123 and parameters: {'n_estimators': 238, 'max_depth': 4, 'learning_rate': 0.1474951443373673, 'subsample': 0.661258099449487, 'colsample_bytree': 0.653871
[1] 2025-05-31 02:44:39,809 Trial 15 finished with value: 0.0016309109057651597 and parameters: {'n_estimators': 213, 'max_depth': 4, 'learning_rate': 0.200962453633609, 'subsample': 0.5032195665864015, 'colsample_bytree': 0.62106
[1] 2025-05-31 02:44:39,994 Trial 16 finished with value: 0.0016658977417015 and parameters: {'n_estimators': 265, 'max_depth': 5, 'learning_rate': 0.09366162516716449, 'subsample': 0.7559756563333376, 'colsample_bytree': 0.57342
[1] 2025-05-31 02:44:40,120 Trial 17 finished with value: 0.001671534633981372 and parameters: {'n_estimators': 198, 'max_depth': 7, 'learning_rate': 0.1083238711245787, 'subsample': 0.6069826829544832, 'colsample_bytree': 0.688648
[1] 2025-05-31 02:44:40,277 Trial 18 finished with value: 0.001649091950460242 and parameters: {'n_estimators': 255, 'max_depth': 4, 'learning_rate': 0.119577423704058, 'subsample': 0.7431128231504769, 'colsample_bytree': 0.70927
[1] 2025-05-31 02:44:40,424 Trial 19 finished with value: 0.001621315455156768 and parameters: {'n_estimators': 248, 'max_depth': 3, 'learning_rate': 0.2062816100219647, 'subsample': 0.501483725095811, 'colsample_bytree': 0.60094
[1] 2025-05-31 02:44:40,555 A new study created in memory with name: no-name-cle1368-4b16-4680-9f4d-7ad7f3b3297
[1] 2025-05-31 02:44:40,711 Trial 0 finished with value: 0.001305914433787616 and parameters: {'n_estimators': 219, 'max_depth': 6, 'learning_rate': 0.08077025739194243, 'subsample': 0.7145857031577394, 'colsample_bytree': 0.50981
[1] 2025-05-31 02:44:40,795 Trial 1 finished with value: 0.00200185620733597 and parameters: {'n_estimators': 110, 'max_depth': 7, 'learning_rate': 0.13148426101590482, 'subsample': 0.9545010241247491, 'colsample_bytree': 0.640802
[1] 2025-05-31 02:44:40,968 Trial 2 finished with value: 0.001928125179761964 and parameters: {'n_estimators': 276, 'max_depth': 8, 'learning_rate': 0.1970849674417777, 'subsample': 0.567624447735444, 'colsample_bytree': 0.523230
[1] 2025-05-31 02:44:41,045 Trial 3 finished with value: 0.001920498102490873 and parameters: {'n_estimators': 96, 'max_depth': 7, 'learning_rate': 0.0428736073731938, 'subsample': 0.902326116455246, 'colsample_bytree': 0.6876113
[1] 2025-05-31 02:44:41,161 Trial 4 finished with value: 0.00202836784274164716 and parameters: {'n_estimators': 209, 'max_depth': 8, 'learning_rate': 0.01223520073604896, 'subsample': 0.748332192327223, 'colsample_bytree': 0.79403
[1] 2025-05-31 02:44:41,240 Trial 5 finished with value: 0.001997882042843818 and parameters: {'n_estimators': 93, 'max_depth': 5, 'learning_rate': 0.01381635046374223, 'subsample': 0.755881504242981, 'colsample_bytree': 0.687290
[1] 2025-05-31 02:44:41,353 Trial 6 finished with value: 0.0020281544353769 and parameters: {'n_estimators': 210, 'max_depth': 5, 'learning_rate': 0.04538142142057248, 'subsample': 0.744262955162075008, 'colsample_bytree': 0.7887
[1] 2025-05-31 02:44:41,470 Trial 7 finished with value: 0.0020018856053492736 and parameters: {'n_estimators': 209, 'max_depth': 4, 'learning_rate': 0.0202256120505662, 'subsample': 0.7183326340626562, 'colsample_bytree': 0.59453
[1] 2025-05-31 02:44:41,583 Trial 8 finished with value: 0.00199132425623695 and parameters: {'n_estimators': 101, 'max_depth': 7, 'learning_rate': 0.0134987515574396, 'subsample': 0.626295680237232, 'colsample_bytree': 0.73889
[1] 2025-05-31 02:44:41,687 Trial 9 finished with value: 0.0020170386629313 and parameters: {'n_estimators': 261, 'max_depth': 4, 'learning_rate': 0.173340600551662, 'subsample': 0.639170495160831, 'colsample_bytree': 0.940026
[1] 2025-05-31 02:44:41,845 Trial 10 finished with value: 0.001981943833129586 and parameters: {'n_estimators': 289, 'max_depth': 8, 'learning_rate': 0.10087485641918015, 'subsample': 0.5206180458218426, 'colsample_bytree': 0.5140
[1] 2025-05-31 02:44:42,033 Trial 11 finished with value: 0.00198264759920578905 and parameters: {'n_estimators': 299, 'max_depth': 8, 'learning_rate': 0.0970167247622542, 'subsample': 0.513719067178104184, 'colsample_bytree': 0.5071
[1] 2025-05-31 02:44:42,203 Trial 12 finished with value: 0.00196972907905355 and parameters: {'n_estimators': 297, 'max_depth': 8, 'learning_rate': 0.1079579265719709, 'subsample': 0.5126438896208233, 'colsample_bytree': 0.597486
[1] 2025-05-31 02:44:42,364 Trial 13 finished with value: 0.0019923398341565 and parameters: {'n_estimators': 254, 'max_depth': 3, 'learning_rate': 0.082866056246409506, 'subsample': 0.518024131849109, 'colsample_bytree': 0.59218
[1] 2025-05-31 02:44:42,463 Trial 14 finished with value: 0.00203616241217575 and parameters: {'n_estimators': 171, 'max_depth': 6, 'learning_rate': 0.03095854257496831, 'subsample': 0.606270617069715, 'colsample_bytree': 0.58596
[1] 2025-05-31 02:44:42,620 Trial 15 finished with value: 0.0019826665617052272 and parameters: {'n_estimators': 299, 'max_depth': 8, 'learning_rate': 0.05218710677182466, 'subsample': 0.5021871067718177, 'colsample_bytree': 0.58914
[1] 2025-05-31 02:44:42,705 Trial 16 finished with value: 0.002000765347203945 and parameters: {'n_estimators': 50, 'max_depth': 7, 'learning_rate': 0.02763326021936094, 'subsample': 0.5797257058795160916, 'colsample_bytree': 0.5834
[1] 2025-05-31 02:44:42,826 Trial 17 finished with value: 0.002008267
```

[1] 2025-05-31 02:44:51-3623 Trial 5 finished with value: 0.003280727865314 and parameters: {'n\_estimators': 100, 'max\_depth': 9, 'learning\_rate': 0.003571550717001, 'subsample': 0.0703571550717001, 'colsample\_bytree': 0.193071}, {'1] 2025-05-31 02:44:51-4863 Trial 6 finished with value: 0.00338634197319794 and parameters: {'n\_estimators': 238, 'max\_depth': 3, 'learning\_rate': 0.27313338277557403, 'subsample': 0.83705641994545, 'colsample\_bytree': 0.6791654}, {'1] 2025-05-31 02:44:51-4863 Trial 7 finished with value: 0.00334421397534446 and parameters: {'n\_estimators': 159, 'max\_depth': 7, 'learning\_rate': 0.0116771811877349, 'subsample': 0.76863349047261937, 'colsample\_bytree': 0.9661433}, {'1] 2025-05-31 02:44:51-5724 Trial 8 finished with value: 0.003350822591320287 and parameters: {'n\_estimators': 269, 'max\_depth': 6, 'learning\_rate': 0.048725057951829, 'subsample': 0.7764549094057293, 'colsample\_bytree': 0.93071}, {'1] 2025-05-31 02:44:51-8363 Trial 9 finished with value: 0.0034349675151507408 and parameters: {'n\_estimators': 195, 'max\_depth': 4, 'learning\_rate': 0.017576221513483462, 'subsample': 0.9377426447895496, 'colsample\_bytree': 0.56904}, {'1] 2025-05-31 02:44:51-9993 Trial 10 finished with value: 0.003326184159854626 and parameters: {'n\_estimators': 296, 'max\_depth': 3, 'learning\_rate': 0.1252330919849813, 'subsample': 0.7291312095262714, 'colsample\_bytree': 0.74204}, {'1] 2025-05-31 02:44:52-1100 Trial 11 finished with value: 0.00335713920657394 and parameters: {'n\_estimators': 139, 'max\_depth': 4, 'learning\_rate': 0.29793393647102164, 'subsample': 0.9920402533329397, 'colsample\_bytree': 0.7492}, {'1] 2025-05-31 02:44:52-2453 Trial 12 finished with value: 0.00332149575011306 and parameters: {'n\_estimators': 204, 'max\_depth': 4, 'learning\_rate': 0.1555888383295804, 'subsample': 0.519697306086262, 'colsample\_bytree': 0.5001}, {'1] 2025-05-31 02:44:52-3583 Trial 13 finished with value: 0.0033603211498328713 and parameters: {'n\_estimators': 120, 'max\_depth': 4, 'learning\_rate': 0.13679785258585056, 'subsample': 0.5038208752265172, 'colsample\_bytree': 0.8429}, {'1] 2025-05-31 02:44:52-4993 Trial 14 finished with value: 0.00329415205011306 and parameters: {'n\_estimators': 200, 'max\_depth': 3, 'learning\_rate': 0.1474140582543892, 'subsample': 0.5039889010403032, 'colsample\_bytree': 0.65403}, {'1] 2025-05-31 02:44:52-6573 Trial 15 finished with value: 0.00329804028469953 and parameters: {'n\_estimators': 256, 'max\_depth': 3, 'learning\_rate': 0.088668022275914775, 'subsample': 0.614332224697981, 'colsample\_bytree': 0.69117}, {'1] 2025-05-31 02:44:52-7423 Trial 16 finished with value: 0.00334515996653077 and parameters: {'n\_estimators': 56, 'max\_depth': 3, 'learning\_rate': 0.1821777651163106, 'subsample': 0.879873957683797, 'colsample\_bytree': 0.642007}, {'1] 2025-05-31 02:44:52-9143 Trial 17 finished with value: 0.0033080516474659565 and parameters: {'n\_estimators': 288, 'max\_depth': 3, 'learning\_rate': 0.0262619514439566, 'subsample': 0.6750464230979488, 'colsample\_bytree': 0.8153}, {'1] 2025-05-31 02:44:53-0393 Trial 18 finished with value: 0.003378776353718724 and parameters: {'n\_estimators': 163, 'max\_depth': 4, 'learning\_rate': 0.19322040488129089, 'subsample': 0.5711615093892262, 'colsample\_bytree': 0.7095}, {'1] 2025-05-31 02:44:53-1723 Trial 19 finished with value: 0.003344051582863563 and parameters: {'n\_estimators': 195, 'max\_depth': 8, 'learning\_rate': 0.0525491963682582, 'subsample': 0.922370393150322, 'colsample\_bytree': 0.58905}, {'1] 2025-05-31 02:44:53-3003 A new study created in memory with name: no-name\_7103cd77-407c-8c47-d6b73d0550f5 [1] 2025-05-31 02:44:53-5033 Trial 0 finished with value: 0.00246974788288616 and parameters: {'n\_estimators': 287, 'max\_depth': 4, 'learning\_rate': 0.02369674630008222, 'subsample': 0.852935462975784, 'colsample\_bytree': 0.5320866}, {'1] 2025-05-31 02:44:53-6083 Trial 1 finished with value: 0.0026402499351474823 and parameters: {'n\_estimators': 171, 'max\_depth': 3, 'learning\_rate': 0.0350424993517559775, 'subsample': 0.887969637306805, 'colsample\_bytree': 0.95797}, {'1] 2025-05-31 02:44:53-7033 Trial 2 finished with value: 0.002817777717653 and parameters: {'n\_estimators': 149, 'max\_depth': 3, 'learning\_rate': 0.11269192350059469, 'subsample': 0.56744579899080084, 'colsample\_bytree': 0.940857}, {'1] 2025-05-31 02:44:53-8483 Trial 3 finished with value: 0.0026471500180704 and parameters: {'n\_estimators': 286, 'max\_depth': 4, 'learning\_rate': 0.04671165659644665, 'subsample': 0.19484963678808, 'colsample\_bytree': 0.724317}, {'1] 2025-05-31 02:44:53-9853 Trial 4 finished with value: 0.00303640163473583 and parameters: {'n\_estimators': 275, 'max\_depth': 6, 'learning\_rate': 0.20772557502957168, 'subsample': 0.7062481747788402, 'colsample\_bytree': 0.643341}, {'1] 2025-05-31 02:44:54-1183 Trial 5 finished with value: 0.002458187666376425 and parameters: {'n\_estimators': 264, 'max\_depth': 6, 'learning\_rate': 0.024583651026165166, 'subsample': 0.9230269860856719, 'colsample\_bytree': 0.80513}, {'1] 2025-05-31 02:44:54-2143 Trial 6 finished with value: 0.00246797338387408 and parameters: {'n\_estimators': 163, 'max\_depth': 3, 'learning\_rate': 0.04877650548502729, 'subsample': 0.9251303252617731, 'colsample\_bytree': 0.534042}, {'1] 2025-05-31 02:44:54-3333 Trial 7 finished with value: 0.0026869087289502 and parameters: {'n\_estimators': 220, 'max\_depth': 4, 'learning\_rate': 0.1377742873587301, 'subsample': 0.521894205371527, 'colsample\_bytree': 0.997902}, {'1] 2025-05-31 02:44:54-4823 Trial 8 finished with value: 0.0026379938866063 and parameters: {'n\_estimators': 189, 'max\_depth': 3, 'learning\_rate': 0.0152940531547982, 'subsample': 0.835012394726928, 'colsample\_bytree': 0.939029}, {'1] 2025-05-31 02:44:54-5191 Trial 9 finished with value: 0.00269310652038553 and parameters: {'n\_estimators': 192, 'max\_depth': 4, 'learning\_rate': 0.01278756704058865, 'subsample': 0.6947541059511954, 'colsample\_bytree': 0.79942}, {'1] 2025-05-31 02:44:54-6813 Trial 10 finished with value: 0.00269578490339713 and parameters: {'n\_estimators': 55, 'max\_depth': 8, 'learning\_rate': 0.297924600161564, 'subsample': 0.0510471419588316, 'colsample\_bytree': 0.865208}, {'1] 2025-05-31 02:44:54-8153 Trial 11 finished with value: 0.002618432231602345 and parameters: {'n\_estimators': 220, 'max\_depth': 5, 'learning\_rate': 0.116268487736731615, 'subsample': 0.6669090900825999, 'colsample\_bytree': 0.71375}, {'1] 2025-05-31 02:44:54-9633 Trial 12 finished with value: 0.0019777623468964 and parameters: {'n\_estimators': 232, 'max\_depth': 5, 'learning\_rate': 0.1162330329282264, 'subsample': 0.6353827913073129, 'colsample\_bytree': 0.66575}, {'1] 2025-05-31 02:44:55-1043 Trial 13 finished with value: 0.002604428745781157 and parameters: {'n\_estimators': 225, 'max\_depth': 5, 'learning\_rate': 0.09870189699568521, 'subsample': 0.60915639210957, 'colsample\_bytree': 0.64025}, {'1] 2025-05-31 02:44:55-2123 Trial 14 finished with value: 0.00161621852918797 and parameters: {'n\_estimators': 116, 'max\_depth': 7, 'learning\_rate': 0.08808070854638128, 'subsample': 0.612099710019921, 'colsample\_bytree': 0.62658}, {'1] 2025-05-31 02:44:55-3163 Trial 15 finished with value: 0.002646597158170104 and parameters: {'n\_estimators': 109, 'max\_depth': 7, 'learning\_rate': 0.0643506910795949, 'subsample': 0.627676495258282, 'colsample\_bytree': 0.6169030}, {'1] 2025-05-31 02:44:55-4273 Trial 16 finished with value: 0.002225221700615853 and parameters: {'n\_estimators': 112, 'max\_depth': 7, 'learning\_rate': 0.0723194543149736, 'subsample': 0.7666199384592667, 'colsample\_bytree': 0.580345}, {'1] 2025-05-31 02:44:55-5353 Trial 17 finished with value: 0.0016063907114448396 and parameters: {'n\_estimators': 119, 'max\_depth': 7, 'learning\_rate': 0.181901724966811, 'subsample': 0.5810540208796681, 'colsample\_bytree': 0.68103}, {'1] 2025-05-31 02:44:55-7433 Trial 18 finished with value: 0.001670081803511159 and parameters: {'n\_estimators': 81, 'max\_depth': 7, 'learning\_rate': 0.1817855117049125, 'subsample': 0.5667320789624207, 'colsample\_bytree': 0.691134}, {'1] 2025-05-31 02:44:55-8353 A new study created in memory with name: no-name\_aaa394e-81b2-4da2-8eb3-327e28bd13 [1] 2025-05-31 02:44:55-9343 Trial 0 finished with value: 0.0013244007528758488 and parameters: {'n\_estimators': 101, 'max\_depth': 5, 'learning\_rate': 0.07092829935761384, 'subsample': 0.840705236858519, 'colsample\_bytree': 0.61772}, {'1] 2025-05-31 02:44:56-0113 Trial 1 finished with value: 0.001320930957774502 and parameters: {'n\_estimators': 53, 'max\_depth': 8, 'learning\_rate': 0.04732195432309286, 'subsample': 0.9701624960056994, 'colsample\_bytree': 0.680549}, {'1] 2025-05-31 02:44:56-1123 Trial 2 finished with value: 0.001288012372056517 and parameters: {'n\_estimators': 181, 'max\_depth': 6, 'learning\_rate': 0.11986158102151331, 'subsample': 0.7253679028295097, 'colsample\_bytree': 0.59240}, {'1] 2025-05-31 02:44:56-2483 Trial 3 finished with value: 0.001320523904074504 and parameters: {'n\_estimators': 268, 'max\_depth': 6, 'learning\_rate': 0.010339693632398336, 'subsample': 0.852422175978967, 'colsample\_bytree': 0.6684}, {'1] 2025-05-31 02:44:56-3503 Trial 4 finished with value: 0.001307801633247713 and parameters: {'n\_estimators': 171, 'max\_depth': 5, 'learning\_rate': 0.16774387673393165, 'subsample': 0.730974146943238, 'colsample\_bytree': 0.53022}, {'1] 2025-05-31 02:44:56-4483 Trial 5 finished with value: 0.001334586434964424 and parameters: {'n\_estimators': 155, 'max\_depth': 5, 'learning\_rate': 0.153051035500477782, 'subsample': 0.87369513270804, 'colsample\_bytree': 0.768377}, {'1] 2025-05-31 02:44:56-5813 Trial 6 finished with value: 0.001342854257157899 and parameters: {'n\_estimators': 167, 'max\_depth': 6, 'learning\_rate': 0.036690170544620375, 'subsample': 0.56949626373062773, 'colsample\_bytree': 0.9561}, {'1] 2025-05-31 02:44:56-6733 Trial 7 finished with value: 0.001327812421804564 and parameters: {'n\_estimators': 142, 'max\_depth': 4, 'learning\_rate': 0.140545352154204952, 'subsample': 0.7661611313160359, 'colsample\_bytree': 0.9850}, {'1] 2025-05-31 02:44:56-7766 Trial 8 finished with value: 0.001337940237293703 and parameters: {'n\_estimators': 170, 'max\_depth': 4, 'learning\_rate': 0.2427463212829593, 'subsample': 0.862360431040304017, 'colsample\_bytree': 0.90431}, {'1] 2025-05-31 02:44:56-8983 Trial 9 finished with value: 0.001325808797364067 and parameters: {'n\_estimators': 213, 'max\_depth': 8, 'learning\_rate': 0.031248765782196143, 'subsample': 0.9156921412474941, 'colsample\_bytree': 0.8496}, {'1] 2025-05-31 02:44:56-9843 Trial 10 finished with value: 0.00132353061635762 and parameters: {'n\_estimators': 59, 'max\_depth': 3, 'learning\_rate': 0.0906559164061526, 'subsample': 0.5990847951760199, 'colsample\_bytree': 0.501364}, {'1] 2025-05-31 02:44:57-0783 Trial 11 finished with value: 0.001322831386005637 and parameters: {'n\_estimators': 75, 'max\_depth': 3, 'learning\_rate': 0.05532631725299341, 'subsample': 0.5086656523207441, 'colsample\_bytree': 0.53183}, {'1] 2025-05-31 02:44:57-1643 Trial 12 finished with value: 0.001322832835056303 and parameters: {'n\_estimators': 52, 'max\_depth': 3, 'learning\_rate': 0.038593786450328, 'subsample': 0.508150432484811, 'colsample\_bytree': 0.50760}, {'1] 2025-05-31 02:44:57-2643 Trial 13 finished with value: 0.001670081803511159 and parameters: {'n\_estimators': 81, 'max\_depth': 7, 'learning\_rate': 0.1817855117049125, 'subsample': 0.5667320789624207, 'colsample\_bytree': 0.691134}, {'1] 2025-05-31 02:44:57-3443 Trial 14 finished with value: 0.001670081803511159 and parameters: {'n\_estimators': 131, 'max\_depth': 8, 'learning\_rate': 0.2954656633520293, 'subsample': 0.739915955635526, 'colsample\_bytree': 0.75857}, {'1] 2025-05-31 02:44:57-4453 Trial 15 finished with value: 0.0013244007528758488 and parameters: {'n\_estimators': 101, 'max\_depth': 5, 'learning\_rate': 0.07092829935761384, 'subsample': 0.840705236858519, 'colsample\_bytree': 0.61772}, {'1] 2025-05-31 02:44:57-5343 Trial 16 finished with value: 0.001320930957774502 and parameters: {'n\_estimators': 53, 'max\_depth': 8, 'learning\_rate': 0.04732195432309286, 'subsample': 0.9701624960056994, 'colsample\_bytree': 0.680549}, {'1] 2025-05-31 02:44:57-6123 Trial 17 finished with value: 0.001288012372056517 and parameters: {'n\_estimators': 181, 'max\_depth': 6, 'learning\_rate': 0.11986158102151331, 'subsample': 0.7253679028295097, 'colsample\_bytree': 0.59240}, {'1] 2025-05-31 02:44:57-2483 Trial 18 finished with value: 0.001320523904074504 and parameters: {'n\_estimators': 268, 'max\_depth': 6, 'learning\_rate': 0.010339693632398336, 'subsample': 0.852422175978967, 'colsample\_bytree': 0.6684}, {'1] 2025-05-31 02:44:57-3503 Trial 19 finished with value: 0.001307801633247713 and parameters: {'n\_estimators': 171, 'max\_depth': 5, 'learning\_rate': 0.16774387673393165, 'subsample': 0.730974146943238, 'colsample\_bytree': 0.53022}, {'1] 2025-05-31 02:44:57-4503 Trial 20 finished with value: 0.001307801633247713 and parameters: {'n\_estimators': 171, 'max\_depth': 5, 'learning\_rate': 0.16774387673393165, 'subsample': 0.730974146943238, 'colsample\_bytree': 0.53022}, {'1] 2025-05-31 02:44:57-5533 Trial 21 finished with value: 0.001325808797364067 and parameters: {'n\_estimators': 213, 'max\_depth': 8, 'learning\_rate': 0.031248765782196143, 'subsample': 0.9156921412474941, 'colsample\_bytree': 0.8496}, {'1] 2025-05-31 02:44:57-6473 Trial 22 finished with value: 0.001325808797364067 and parameters: {'n\_estimators': 171, 'max\_depth': 5, 'learning\_rate': 0.16774387673393165, 'subsample': 0.730974146943238, 'colsample\_bytree': 0.53022}, {'1] 2025-05-31 02:44:57-7423 Trial 23 finished with value: 0.001325808797364067 and parameters: {'n\_estimators': 171, 'max\_depth': 5, 'learning\_rate': 0.16774387673393165, 'subsample': 0.730974146943238, 'colsample\_bytree': 0.53022}, {'1] 2025-05-31 02:44:57-8343 Trial 24 finished with value: 0.001325808797364067 and parameters: {'n\_estimators': 171, 'max\_depth': 5, 'learning\_rate': 0.16774387673393165, 'subsample': 0.730974146943238, 'colsample\_bytree': 0.53022}, {'1] 2025-05-31 02:44:57-9343 Trial 25 finished with value: 0.001325808797364067 and parameters: {'n\_estimators': 171, 'max\_depth': 5, 'learning\_rate': 0.16774387673393165, 'subsample': 0.730974146943238, 'colsample\_bytree': 0.53022}, {'1] 2025-05-31 02:44:57-1043 Trial 26 finished with value: 0.001325808797364067 and parameters: {'n\_estimators': 171, 'max\_depth': 5, 'learning\_rate': 0.16774387673393165, 'subsample': 0.730974146943238, 'colsample\_bytree': 0.53022}, {'1] 2025-05-31 02:44:57-2123 Trial 27 finished with value: 0.001325808797364067 and parameters: {'n\_estimators': 171, 'max\_depth': 5, 'learning\_rate': 0.16774387673393165, 'subsample': 0.730974146943238, 'colsample\_bytree': 0.53022}, {'1] 2025-05-31 02:44:57-3133 Trial 28 finished with value: 0.001325808797364067 and parameters: {'n\_estimators': 171, 'max\_depth': 5, 'learning\_rate': 0.16774387673393165, 'subsample': 0.730974146943238, 'colsample\_bytree': 0.53022}, {'1] 2025-05-31 02:44:57-4123 Trial 29 finished with value: 0.001325808797364067 and parameters: {'n\_estimators': 171, 'max\_depth': 5, 'learning\_rate': 0.16774387673393165, 'subsample': 0.730974146943238, 'colsample\_bytree': 0.53022}, {'1] 2025-05-31 02:44:57-5133 Trial 30 finished with value: 0.001325808797364067 and parameters: {'n\_estimators': 171, 'max\_depth': 5, 'learning\_rate': 0.16774387673393165, 'subsample': 0.730974146943238, 'colsample\_bytree': 0.53022}, {'1] 2025-05-31 02:44:57-6123 Trial 31 finished with value: 0.001325808797364067 and parameters: {'n\_estimators': 171, 'max\_depth': 5, 'learning\_rate': 0.16774387673393165, 'subsample': 0.730974146943238, 'colsample\_bytree': 0.53022}, {'1] 2025-05-31 02:44:57-7133 Trial 32 finished with value: 0.001325808797364067 and parameters: {'n\_estimators': 171, 'max\_depth': 5, 'learning\_rate': 0.16774387673393165, 'subsample': 0.730974146943238, 'colsample\_bytree': 0.53022}, {'1] 2025-05-31 02:44:57-8123 Trial 33 finished with value: 0.001325808797364067 and parameters: {'n\_estimators': 171, 'max\_depth': 5, 'learning\_rate': 0.16774387673393165, 'subsample': 0.730974146943238, 'colsample\_bytree': 0.53022}, {'1] 2025-05-31 02:44:57-9123 Trial 34 finished with value: 0.001325808797364067 and parameters: {'n\_estimators': 171, 'max\_depth': 5, 'learning\_rate': 0.16774387673393165, 'subsample': 0.730974146943238, 'colsample\_bytree': 0.53022}, {'1] 2025-05-31 02:44:57-1023 Trial 35 finished with value: 0.001325808797364067 and parameters: {'n\_estimators': 171, 'max\_depth': 5, 'learning\_rate': 0.16774387673393165, 'subsample': 0.730974146943238, 'colsample\_bytree': 0.53022}, {'1] 2025-05-31 02:44:57-2033 Trial 36 finished with value: 0.001325808797364067 and parameters: {'n\_estimators': 171, 'max\_depth': 5, 'learning\_rate': 0.16774387673393165, 'subsample': 0.730974146943238, 'colsample\_bytree': 0.53022}, {'1] 2025-05-31 02:44:57-3033 Trial 37 finished with value: 0.001325808797364067 and parameters: {'n\_estimators': 171, 'max\_depth': 5, 'learning\_rate': 0.16774387673393165, 'subsample': 0.730974146943238, 'colsample\_bytree': 0.53022}, {'1] 2025-05-31 02:44:57-4033 Trial 38 finished with value: 0.001325808797364067 and parameters: {'n\_estimators': 171, 'max\_depth': 5, 'learning\_rate': 0.16774387673393165, 'subsample': 0.730974146943238, 'colsample\_bytree': 0.53022}, {'1] 2025-05-31 02:44:57-5033 Trial 39 finished with value: 0.001325808797364067 and parameters: {'n\_estimators': 171, 'max\_depth': 5, 'learning\_rate': 0.16774387673393165, 'subsample': 0.730974146943238, 'colsample\_bytree': 0.53022}, {'1] 2025-05-31 02:44:57-6033 Trial 40 finished with value: 0.001325808797364067 and parameters: {'n\_estimators': 171, 'max\_depth': 5, 'learning\_rate': 0.16774387673393165, 'subsample': 0.730974146943238, 'colsample\_bytree': 0.53022}, {'1] 2025-05-31 02:44:57-7033 Trial 41 finished with value: 0.001325808797364067 and parameters: {'n\_estimators': 171, 'max\_depth': 5, 'learning\_rate': 0.16774387673393165, 'subsample': 0.730974146943238, 'colsample\_bytree': 0.53022}, {'1] 2025-05-31 02:44:57-8033 Trial 42 finished with value: 0.001325808797364067 and parameters: {'n\_estimators': 171, 'max\_depth': 5, 'learning\_rate': 0.16774387673393165, 'subsample': 0.730974146943238, 'colsample\_bytree': 0.53022}, {'1] 2025-05-31 02:44:57-9033 Trial 43 finished with value: 0.001325808797364067 and parameters: {'n\_estimators': 171, 'max\_depth': 5, 'learning\_rate': 0.16774387673393165, 'subsample': 0.730974146943238, 'colsample\_bytree': 0.53022}, {'1] 2025-05-31 02:44:57-1033 Trial 44 finished with value: 0.001325808797364067 and parameters: {'n\_estimators': 171, 'max\_depth': 5, 'learning\_rate': 0.16774387673393

[I] 2025-05-31 02:45:08.881 Trial 12 finished with value: 0.0025987452050439005 and parameters: {'n\_estimators': 292, 'max\_depth': 6, 'learning\_rate': 0.01000410568125499, 'subsample': 0.611702887903035, 'colsample\_bytree': 0.5101}, [I] 2025-05-31 02:45:09.067 Trial 13 finished with value: 0.022615330130489786 and parameters: {'n\_estimators': 251, 'max\_depth': 7, 'learning\_rate': 0.23620565944322192, 'subsample': 0.54657057645277537, 'colsample\_bytree': 0.58696}, [I] 2025-05-31 02:45:09.215 Trial 14 finished with value: 0.0260170138340564 and parameters: {'n\_estimators': 234, 'max\_depth': 5, 'learning\_rate': 0.0788794132290029, 'subsample': 0.6695847037982826, 'colsample\_bytree': 0.82035}, [I] 2025-05-31 02:45:09.324 Trial 15 finished with value: 0.002690502769291372 and parameters: {'n\_estimators': 131, 'max\_depth': 7, 'learning\_rate': 0.17568308302588684, 'subsample': 0.998295782309267, 'colsample\_bytree': 0.78521}, [I] 2025-05-31 02:45:09.456 Trial 16 finished with value: 0.0265842496633132 and parameters: {'n\_estimators': 213, 'max\_depth': 4, 'learning\_rate': 0.05726270986313494, 'subsample': 0.788908073808923, 'colsample\_bytree': 0.57842}, [I] 2025-05-31 02:45:09.608 Trial 17 finished with value: 0.0258397099674488 and parameters: {'n\_estimators': 269, 'max\_depth': 7, 'learning\_rate': 0.01028872305905165, 'subsample': 0.5632148896943623, 'colsample\_bytree': 0.57575}, [I] 2025-05-31 02:45:09.717 Trial 18 finished with value: 0.0253537635817653529 and parameters: {'n\_estimators': 123, 'max\_depth': 5, 'learning\_rate': 0.074972909703631, 'subsample': 0.6464041472537995, 'colsample\_bytree': 0.762910}, [I] 2025-05-31 02:45:09.882 Trial 19 finished with value: 0.025925365817653529 and parameters: {'n\_estimators': 122, 'max\_depth': 6, 'learning\_rate': 0.01580964344055557, 'subsample': 0.646717870834313, 'colsample\_bytree': 0.7721}, [I] 2025-05-31 02:45:09.925 A new study created in memory with name: no-name-67bbcc8-2581-4cfb-8899-e2e2eeb665 [I] 2025-05-31 02:45:10.068 Trial 1 finished with value: 0.00494470247951327 and parameters: {'n\_estimators': 113, 'max\_depth': 3, 'learning\_rate': 0.03735283077515666, 'subsample': 0.6115333521376809, 'colsample\_bytree': 0.55257}, [I] 2025-05-31 02:45:10.154 Trial 1 finished with value: 0.004940376267605725 and parameters: {'n\_estimators': 115, 'max\_depth': 3, 'learning\_rate': 0.29178589461968597, 'subsample': 0.727608583388159, 'colsample\_bytree': 0.940024}, [I] 2025-05-31 02:45:10.286 Trial 2 finished with value: 0.00493750158808493 and parameters: {'n\_estimators': 250, 'max\_depth': 3, 'learning\_rate': 0.1372080376509, 'subsample': 0.574982357589809, 'colsample\_bytree': 0.956964021}, [I] 2025-05-31 02:45:10.427 Trial 3 finished with value: 0.004908505806758772 and parameters: {'n\_estimators': 299, 'max\_depth': 4, 'learning\_rate': 0.2677536371864654, 'subsample': 0.7665896197817797, 'colsample\_bytree': 0.5613582}, [I] 2025-05-31 02:45:10.550 Trial 4 finished with value: 0.004979000365239 and parameters: {'n\_estimators': 250, 'max\_depth': 4, 'learning\_rate': 0.010533238872491886, 'subsample': 0.842950244490912, 'colsample\_bytree': 0.54500}, [I] 2025-05-31 02:45:10.624 Trial 5 finished with value: 0.004903171143466123 and parameters: {'n\_estimators': 85, 'max\_depth': 5, 'learning\_rate': 0.2738656169477457, 'subsample': 0.6422042505656702, 'colsample\_bytree': 0.66154348}, [I] 2025-05-31 02:45:10.762 Trial 6 finished with value: 0.00495408549595937 and parameters: {'n\_estimators': 193, 'max\_depth': 6, 'learning\_rate': 0.15047425720377, 'subsample': 0.876567435620528, 'colsample\_bytree': 0.842627473}, [I] 2025-05-31 02:45:10.893 Trial 7 finished with value: 0.004956792213956755 and parameters: {'n\_estimators': 185, 'max\_depth': 4, 'learning\_rate': 0.05022735056486981, 'subsample': 0.5206173381402376, 'colsample\_bytree': 0.60882}, [I] 2025-05-31 02:45:11.139 Trial 8 finished with value: 0.004951977215158445 and parameters: {'n\_estimators': 236, 'max\_depth': 4, 'learning\_rate': 0.0502112197033741, 'subsample': 0.798065275589868, 'colsample\_bytree': 0.58765}, [I] 2025-05-31 02:45:11.855 Trial 9 finished with value: 0.0049373915354847492 and parameters: {'n\_estimators': 292, 'max\_depth': 5, 'learning\_rate': 0.11760911156721615, 'subsample': 0.939405779447057, 'colsample\_bytree': 0.888082}, [I] 2025-05-31 02:45:12.042 Trial 10 finished with value: 0.004936488333981856 and parameters: {'n\_estimators': 143, 'max\_depth': 8, 'learning\_rate': 0.01019919559416728, 'subsample': 0.68377648737503, 'colsample\_bytree': 0.7111326}, [I] 2025-05-31 02:45:12.957 Trial 11 finished with value: 0.00494552707965256 and parameters: {'n\_estimators': 58, 'max\_depth': 6, 'learning\_rate': 0.29475732949453, 'subsample': 0.673775470529854, 'colsample\_bytree': 0.6913549}, [I] 2025-05-31 02:45:13.177 Trial 12 finished with value: 0.00492627560431201 and parameters: {'n\_estimators': 52, 'max\_depth': 5, 'learning\_rate': 0.192056283535482, 'subsample': 0.7682034380892979, 'colsample\_bytree': 0.6595265}, [I] 2025-05-31 02:45:13.406 Trial 13 finished with value: 0.004916250667203787 and parameters: {'n\_estimators': 294, 'max\_depth': 7, 'learning\_rate': 0.08523616051597244, 'subsample': 0.6153923304665581, 'colsample\_bytree': 0.79402}, [I] 2025-05-31 02:45:13.530 Trial 14 finished with value: 0.00493837402682975 and parameters: {'n\_estimators': 88, 'max\_depth': 5, 'learning\_rate': 0.0874282677571504, 'subsample': 0.5926811080929029, 'colsample\_bytree': 0.655252}, [I] 2025-05-31 02:45:13.801 Trial 15 finished with value: 0.00496160865648143 and parameters: {'n\_estimators': 150, 'max\_depth': 4, 'learning\_rate': 0.2191691321636601, 'subsample': 0.73695581974833, 'colsample\_bytree': 0.75038}, [I] 2025-05-31 02:45:13.939 Trial 16 finished with value: 0.00493893083225873 and parameters: {'n\_estimators': 215, 'max\_depth': 6, 'learning\_rate': 0.0246591193628450532, 'subsample': 0.5000491767373035, 'colsample\_bytree': 0.51703}, [I] 2025-05-31 02:45:14.055 Trial 17 finished with value: 0.004967269368381645 and parameters: {'n\_estimators': 150, 'max\_depth': 5, 'learning\_rate': 0.0848516809144055, 'subsample': 0.5891567173244083, 'colsample\_bytree': 0.62011}, [I] 2025-05-31 02:45:14.187 Trial 18 finished with value: 0.00495781799167195 and parameters: {'n\_estimators': 80, 'max\_depth': 7, 'learning\_rate': 0.1981529453518716, 'subsample': 0.805623939989283, 'colsample\_bytree': 0.576898}, [I] 2025-05-31 02:45:14.343 Trial 19 finished with value: 0.004965065341950181 and parameters: {'n\_estimators': 271, 'max\_depth': 4, 'learning\_rate': 0.11499816722450187, 'subsample': 0.6981752205867364, 'colsample\_bytree': 0.57773}, [I] 2025-05-31 02:45:14.423 A new study created in memory with name: no-name-13c91d0-c89-44a2-9090-2f18789192cf [I] 2025-05-31 02:45:14.550 Trial 0 finished with value: 0.017008953733935 and parameters: {'n\_estimators': 106, 'max\_depth': 8, 'learning\_rate': 0.013266787096811223, 'subsample': 0.9245948961007391, 'colsample\_bytree': 0.569413}, [I] 2025-05-31 02:45:14.624 Trial 1 finished with value: 0.0170859734273439 and parameters: {'n\_estimators': 51, 'max\_depth': 6, 'learning\_rate': 0.23893686650667395, 'subsample': 0.33192249369390, 'colsample\_bytree': 0.76625805}, [I] 2025-05-31 02:45:14.745 Trial 2 finished with value: 0.017096307649930645 and parameters: {'n\_estimators': 222, 'max\_depth': 4, 'learning\_rate': 0.017753263491332, 'subsample': 0.8973274612656728, 'colsample\_bytree': 0.616808}, [I] 2025-05-31 02:45:14.856 Trial 3 finished with value: 0.0170873652627071 and parameters: {'n\_estimators': 178, 'max\_depth': 3, 'learning\_rate': 0.04008792897070361, 'subsample': 0.731393844343648, 'colsample\_bytree': 0.8624550}, [I] 2025-05-31 02:45:14.974 Trial 4 finished with value: 0.0170463240497408 and parameters: {'n\_estimators': 203, 'max\_depth': 3, 'learning\_rate': 0.146885364223644862, 'subsample': 0.553127586949348, 'colsample\_bytree': 0.977650}, [I] 2025-05-31 02:45:15.053 Trial 5 finished with value: 0.01710506530781334 and parameters: {'n\_estimators': 93, 'max\_depth': 8, 'learning\_rate': 0.12326952997597025, 'subsample': 0.681381100702404, 'colsample\_bytree': 0.8885146}, [I] 2025-05-31 02:45:15.131 Trial 6 finished with value: 0.0171283189780666 and parameters: {'n\_estimators': 90, 'max\_depth': 3, 'learning\_rate': 0.0263560364514908, 'subsample': 0.5853110240581891, 'colsample\_bytree': 0.8476173}, [I] 2025-05-31 02:45:15.133 Trial 7 finished with value: 0.0171238370142584 and parameters: {'n\_estimators': 297, 'max\_depth': 8, 'learning\_rate': 0.28963489053522461, 'subsample': 0.875533062160142, 'colsample\_bytree': 0.8778190}, [I] 2025-05-31 02:45:15.403 Trial 8 finished with value: 0.017087711337736 and parameters: {'n\_estimators': 130, 'max\_depth': 3, 'learning\_rate': 0.0392352937329475, 'subsample': 0.615395464882009, 'colsample\_bytree': 0.7779072}, [I] 2025-05-31 02:45:15.482 Trial 9 finished with value: 0.01712451877665035 and parameters: {'n\_estimators': 80, 'max\_depth': 5, 'learning\_rate': 0.03958862828170774, 'subsample': 0.7690185243708174, 'colsample\_bytree': 0.698468}, [I] 2025-05-31 02:45:15.624 Trial 10 finished with value: 0.017047991167373736 and parameters: {'n\_estimators': 234, 'max\_depth': 6, 'learning\_rate': 0.01706326779127324, 'subsample': 0.55805374123665977, 'colsample\_bytree': 0.99668}, [I] 2025-05-31 02:45:15.780 Trial 11 finished with value: 0.017069859800104404 and parameters: {'n\_estimators': 238, 'max\_depth': 6, 'learning\_rate': 0.1047650539889579, 'subsample': 0.5147472600813587, 'colsample\_bytree': 0.99394}, [I] 2025-05-31 02:45:15.917 Trial 12 finished with value: 0.0171114532345923 and parameters: {'n\_estimators': 214, 'max\_depth': 5, 'learning\_rate': 0.1245485262000215, 'subsample': 0.3543842050690462, 'colsample\_bytree': 0.99953}, [I] 2025-05-31 02:45:16.077 Trial 13 finished with value: 0.0171546267097810432 and parameters: {'n\_estimators': 284, 'max\_depth': 7, 'learning\_rate': 0.080192851671261028, 'subsample': 0.734649716261028, 'colsample\_bytree': 0.94101}, [I] 2025-05-31 02:45:16.131 Trial 14 finished with value: 0.0171283189780666 and parameters: {'n\_estimators': 90, 'max\_depth': 3, 'learning\_rate': 0.0263560364514908, 'subsample': 0.5853110240581891, 'colsample\_bytree': 0.8476173}, [I] 2025-05-31 02:45:16.133 Trial 15 finished with value: 0.0171238370142584 and parameters: {'n\_estimators': 297, 'max\_depth': 8, 'learning\_rate': 0.28963489053522461, 'subsample': 0.875533062160142, 'colsample\_bytree': 0.8778190}, [I] 2025-05-31 02:45:16.357 Trial 16 finished with value: 0.01728329391061156 and parameters: {'n\_estimators': 259, 'max\_depth': 6, 'learning\_rate': 0.1734667726222978, 'subsample': 0.5021331776924874, 'colsample\_bytree': 0.69991}, [I] 2025-05-31 02:45:16.488 Trial 17 finished with value: 0.01709703627580242 and parameters: {'n\_estimators': 197, 'max\_depth': 7, 'learning\_rate': 0.0676773387939308, 'subsample': 0.571247119865034, 'colsample\_bytree': 0.9474530}, [I] 2025-05-31 02:45:16.602 Trial 18 finished with value: 0.017129717380821979 and parameters: {'n\_estimators': 137, 'max\_depth': 4, 'learning\_rate': 0.18768516086371498, 'subsample': 0.669846562788501, 'colsample\_bytree': 0.81066}, [I] 2025-05-31 02:45:16.750 Trial 19 finished with value: 0.01707468665678033 and parameters: {'n\_estimators': 258, 'max\_depth': 5, 'learning\_rate': 0.187685176083714048, 'subsample': 0.7372081515797855, 'colsample\_bytree': 0.50163}, [I] 2025-05-31 02:45:16.868 Trial 20 finished with value: 0.017071324734117442 and parameters: {'n\_estimators': 149, 'max\_depth': 7, 'learning\_rate': 0.04344639211048131, 'subsample': 0.7891030865647193, 'colsample\_bytree': 0.92853}, [I] 2025-05-31 02:45:16.994 A new study created in memory with name: no-name-3c0f49c9-9ef7-4028-b30-e75b7857ab1 [I] 2025-05-31 02:45:17.142 Trial 0 finished with value: 0.004786291529488 and parameters: {'n\_estimators': 220, 'max\_depth': 3, 'learning\_rate': 0.03183967181429374, 'subsample': 0.564962757228886, 'colsample\_bytree': 0.9744476}, [I] 2025-05-31 02:45:17.314 Trial 1 finished with value: 0.0048064280185364464 and parameters: {'n\_estimators': 277, 'max\_depth': 3, 'learning\_rate': 0.03959882761221034, 'subsample': 0.636938405628063, 'colsample\_bytree': 0.571110}, [I] 2025-05-31 02:45:17.482 Trial 2 finished with value: 0.0047957032351428265 and parameters: {'n\_estimators': 227, 'max\_depth': 6, 'learning\_rate': 0.07085095644659577, 'subsample': 0.645646293340019, 'colsample\_bytree': 0.71786}, [I] 2025-05-31 02:45:17.532 Trial 3 finished with value: 0.00477290236351454 and parameters: {'n\_estimators': 222, 'max\_depth': 4, 'learning\_rate': 0.017753263491332, 'subsample': 0.833535853431571, 'colsample\_bytree': 0.6824556}, [I] 2025-05-31 02:45:17.636 Trial 4 finished with value: 0.004797544891579085 and parameters: {'n\_estimators': 189, 'max\_depth': 4, 'learning\_rate': 0.02153371267853688, 'subsample': 0.6962771383566548, 'colsample\_bytree': 0.82827}, [I] 2025-05-31 02:45:17.775 Trial 5 finished with value: 0.0047929038097517457 and parameters: {'n\_estimators': 285, 'max\_depth': 5, 'learning\_rate': 0.120565282614857981, 'subsample': 0.689820204362695, 'colsample\_bytree': 0.970402}, [I] 2025-05-31 02:45:17.879 Trial 6 finished with value: 0.00479036373072878 and parameters: {'n\_estimators': 167, 'max\_depth': 5, 'learning\_rate': 0.01845270092550433, 'subsample': 0.9328866257480002, 'colsample\_bytree': 0.869933}, [I] 2025-05-31 02:45:17.990 Trial 7 finished with value: 0.004781782509476034 and parameters: {'n\_estimators': 190, 'max\_depth': 4, 'learning\_rate': 0.0839940157357493, 'subsample': 0.55145704511179022, 'colsample\_bytree': 0.8900375}, [I] 2025-05-31 02:45:18.093 Trial 8 finished with value: 0.004791305650095731 and parameters: {'n\_estimators': 167, 'max\_depth': 3, 'learning\_rate': 0.01460591291859926, 'subsample': 0.8121183709800333, 'colsample\_bytree': 0.84367}, [I] 2025-05-31 02:45:18.161 Trial 9 finished with value: 0.004799052400997545 and parameters: {'n\_estimators': 65, 'max\_depth': 3, 'learning\_rate': 0.02843264889393906, 'subsample': 0.5155648367067368, 'colsample\_bytree': 0.5422552}, [I] 2025-05-31 02:45:18.342 Trial 10 finished with value: 0.00477398322094688 and parameters: {'n\_estimators': 61, 'max\_depth': 8, 'learning\_rate': 0.2788155061971671, 'subsample': 0.8521334880367121, 'colsample\_bytree': 0.6745937}, [I] 2025-05-31 02:45:18.354 Trial 11 finished with value: 0.004784550856038538 and parameters: {'n\_estimators': 51, 'max\_depth': 8, 'learning\_rate': 0.267045062613883, 'subsample': 0.8486327247883082, 'colsample\_bytree': 0.6631267}, [I] 2025-05-31 02:45:18.455 Trial 12 finished with value: 0.0047897645920944 and parameters: {'n\_estimators': 109, 'max\_depth': 8, 'learning\_rate': 0.0270975800243623, 'subsample': 0.92056278810739, 'colsample\_bytree': 0.6385962}, [I] 2025-05-31 02:45:18.479 Trial 13 finished with value: 0.00479036373072878 and parameters: {'n\_estimators': 126, 'max\_depth': 7, 'learning\_rate': 0.01845270092550433, 'subsample': 0.9328866257480002, 'colsample\_bytree': 0.869933}, [I] 2025-05-31 02:45:18.671 Trial 14 finished with value: 0.00477178129046794 and parameters: {'n\_estimators': 126, 'max\_depth': 7, 'learning\_rate': 0.1394944557484197, 'subsample': 0.7974045805691692, 'colsample\_bytree': 0.8900375}, [I] 2025-05-31 02:45:18.779 Trial 15 finished with value: 0.0047860519987223659 and parameters: {'n\_estimators': 110, 'max\_depth': 7, 'learning\_rate': 0.15645644636659579, 'subsample': 0.764370974958743, 'colsample\_bytree': 0.791725}, [I] 2025-05-31 02:45:18.883 Trial 16 finished with value: 0.00478475436372236592 and parameters: {'n\_estimators': 119, 'max\_depth': 7, 'learning\_rate': 0.136868872145839, 'subsample': 0.9108173592665799, 'colsample\_bytree': 0.7617689}, [I] 2025-05-31 02:45:18.991 Trial 17 finished with value: 0.00477363895345505 and parameters: {'n\_estimators': 130, 'max\_depth': 7, 'learning\_rate': 0.1914869513521549, 'subsample': 0.7686503613753575, 'colsample\_bytree': 0.91891}, [I] 2025-05-31 02:45:19.097 Trial 18 finished with value: 0.004784587305801474 and parameters: {'n\_estimators': 87, 'max\_depth': 6, 'learning\_rate': 0.05920852992040376, 'subsample': 0.917570583448245, 'colsample\_bytree': 0.8011233}, [I] 2025-05-31 02:45:19.339 Trial 19 finished with value: 0.00482944039349374 and parameters: {'n\_estimators': 90, 'max\_depth': 6, 'learning\_rate': 0.0592085299022040376, 'subsample': 0.9075085499899916, 'colsample\_bytree': 0.612502}, [I] 2025-05-31 02:45:19.446 Trial 20 finished with value: 0.0047862416137917921 and parameters: {'n\_estimators': 99, 'max\_depth': 7, 'learning\_rate': 0.025112740608093, 'subsample': 0.9500737217466559, 'colsample\_bytree': 0.840488234}, [I] 2025-05-31 02:45:20.616 Trial 21 finished with value: 0.0047741920163992 and parameters: {'n\_estimators': 125, 'max\_depth': 8, 'learning\_rate': 0.0174411920163992, 'subsample': 0.5506273478057979, 'colsample\_bytree': 0.6904914}, [I] 2025-05-31 02:45:20.730 Trial 22 finished with value: 0.00476736105638021306 and parameters: {'n\_estimators': 134, 'max\_depth': 8, 'learning\_rate': 0.29643723205608205, 'subsample': 0.505168581736906, 'colsample\_bytree': 0.678764}, [I] 2025-05-31 02:45:20.844 Trial 23 finished with value: 0.004759802367328192 and parameters: {'n\_estimators': 135, 'max\_depth': 8, 'learning\_rate': 0.1480937125551768, 'subsample': 0.63567772782814, 'colsample\_bytree': 0.677864}, [I] 2025-05-31 02:45:20.955 Trial 24 finished with value: 0.00475819781285514 and parameters: {'n\_estimators': 133, 'max\_depth': 8, 'learning\_rate': 0.05348170676547564, 'subsample': 0.7601184634676839, 'colsample\_bytree': 0.9646542}, [I] 2025-05-31 02:45:21.079 Trial 25 finished with value: 0.00476090002968477 and parameters: {'n\_estimators': 165, 'max\_depth': 8, 'learning\_rate': 0.1295709802872573, 'subsample': 0.673023811539365, 'colsample\_bytree': 0.9599345}, [I] 2025-05-31 02:45:21.166 Trial 26 finished with value: 0.0047506034222661 and parameters: {'n\_estimators': 82, 'max\_depth': 7, 'learning\_rate': 0.026166785926555145, 'subsample': 0.9183237458933037, 'colsample\_bytree': 0.913144}, [I] 2025-05-31 02:45:20.420 Trial 27 finished with value: 0.00475061835209803 and parameters: {'n\_estimators': 175, 'max\_depth': 4, 'learning\_rate': 0.07236362146137197, 'subsample': 0.92009050905945255, 'colsample\_bytree': 0.9203194}, [I] 2025-05-31 02:45:20.498 Trial 28 finished with value: 0.00475866160399498 and parameters: {'n\_estimators': 99, 'max\_depth': 7, 'learning\_rate': 0.211732204686803, 'subsample': 0.9500737217466559, 'colsample\_bytree': 0.840488234}, [I] 2025-05

[1] 2025-05-31 02:45:28.251 A new study created in memory with name: no-name-82bb315b-301f-478c-a58-5c3ab7d0c3  
 [1] 2025-05-31 02:45:28.379 Trial 0 finished with value: 0.00509148442667895 and parameters: {'n\_estimators': 151, 'max\_depth': 4, 'learning\_rate': 0.24919272087147448, 'subsample': 0.5150579059687792, 'colsample\_bytree': 0.751827  
 [1] 2025-05-31 02:45:28.524 Trial 1 finished with value: 0.0051589420450723 and parameters: {'n\_estimators': 202, 'max\_depth': 7, 'learning\_rate': 0.073515652508956, 'subsample': 0.945870039593793, 'colsample\_bytree': 0.387894  
 [1] 2025-05-31 02:45:28.635 Trial 2 finished with value: 0.005160223231755863 and parameters: {'n\_estimators': 214, 'max\_depth': 8, 'learning\_rate': 0.03049119895622095, 'subsample': 0.597142874240204, 'colsample\_bytree': 0.505154  
 [1] 2025-05-31 02:45:28.754 Trial 3 finished with value: 0.00513853662947946 and parameters: {'n\_estimators': 124, 'max\_depth': 6, 'learning\_rate': 0.01407220924924784, 'subsample': 0.14646618404634395, 'colsample\_bytree': 0.645751  
 [1] 2025-05-31 02:45:28.832 Trial 4 finished with value: 0.0051808743095302 and parameters: {'n\_estimators': 122, 'max\_depth': 6, 'learning\_rate': 0.0502192322157135, 'subsample': 0.8800727935476464, 'colsample\_bytree': 0.76462136  
 [1] 2025-05-31 02:45:28.934 Trial 5 finished with value: 0.00518828640238405 and parameters: {'n\_estimators': 94, 'max\_depth': 4, 'learning\_rate': 0.0502192322157135, 'subsample': 0.7493751486753069, 'colsample\_bytree': 0.578097  
 [1] 2025-05-31 02:45:29.044 Trial 6 finished with value: 0.005150814179962 and parameters: {'n\_estimators': 163, 'max\_depth': 4, 'learning\_rate': 0.03521049458089414, 'subsample': 0.83442388995924, 'colsample\_bytree': 0.56909  
 [1] 2025-05-31 02:45:29.152 Trial 7 finished with value: 0.00512705988607231 and parameters: {'n\_estimators': 197, 'max\_depth': 3, 'learning\_rate': 0.03703189613059878, 'subsample': 0.63442388995924, 'colsample\_bytree': 0.56909  
 [1] 2025-05-31 02:45:29.276 Trial 8 finished with value: 0.00517541533123949 and parameters: {'n\_estimators': 175, 'max\_depth': 6, 'learning\_rate': 0.1219280246823434, 'subsample': 0.858689362651148, 'colsample\_bytree': 0.936287  
 [1] 2025-05-31 02:45:29.399 Trial 9 finished with value: 0.00518471804664424 and parameters: {'n\_estimators': 239, 'max\_depth': 6, 'learning\_rate': 0.0995845133284126, 'subsample': 0.5382415059593844, 'colsample\_bytree': 0.810570  
 [1] 2025-05-31 02:45:29.565 Trial 10 finished with value: 0.0051325240994031393 and parameters: {'n\_estimators': 292, 'max\_depth': 4, 'learning\_rate': 0.0245585790471441, 'subsample': 0.549516682655898, 'colsample\_bytree': 0.99768  
 [1] 2025-05-31 02:45:29.680 Trial 11 finished with value: 0.00510272249172444 and parameters: {'n\_estimators': 140, 'max\_depth': 4, 'learning\_rate': 0.2846687019814299, 'subsample': 0.73676538824131, 'colsample\_bytree': 0.6646889  
 [1] 2025-05-31 02:45:29.767 Trial 12 finished with value: 0.005148136346175099 and parameters: {'n\_estimators': 53, 'max\_depth': 5, 'learning\_rate': 0.266520000341655932, 'subsample': 0.701726236213541, 'colsample\_bytree': 0.7071698  
 [1] 2025-05-31 02:45:29.878 Trial 13 finished with value: 0.00509412518749326 and parameters: {'n\_estimators': 134, 'max\_depth': 5, 'learning\_rate': 0.187893815790914, 'subsample': 0.53978136782363, 'colsample\_bytree': 0.66542258  
 [1] 2025-05-31 02:45:29.980 Trial 14 finished with value: 0.0051758249880936 and parameters: {'n\_estimators': 98, 'max\_depth': 5, 'learning\_rate': 0.16197135275031854, 'subsample': 0.82878319225351, 'colsample\_bytree': 0.818501  
 [1] 2025-05-31 02:45:30.094 Trial 15 finished with value: 0.005165380013347 and parameters: {'n\_estimators': 140, 'max\_depth': 5, 'learning\_rate': 0.1809343507272122, 'subsample': 0.79881731419926, 'colsample\_bytree': 0.628599  
 [1] 2025-05-31 02:45:30.177 Trial 16 finished with value: 0.0051805261917243886 and parameters: {'n\_estimators': 50, 'max\_depth': 3, 'learning\_rate': 0.080920785844242, 'subsample': 0.8094615460752235, 'colsample\_bytree': 0.800701  
 [1] 2025-05-31 02:45:30.327 Trial 17 finished with value: 0.005176034824361019 and parameters: {'n\_estimators': 261, 'max\_depth': 7, 'learning\_rate': 0.19312947495765304, 'subsample': 0.9074414463431328, 'colsample\_bytree': 0.70112  
 [1] 2025-05-31 02:45:30.424 Trial 18 finished with value: 0.00516515623536187 and parameters: {'n\_estimators': 92, 'max\_depth': 5, 'learning\_rate': 0.017004086381601675, 'subsample': 0.668703331017617, 'colsample\_bytree': 0.88142  
 [1] 2025-05-31 02:45:30.547 Trial 19 finished with value: 0.00513168809495119 and parameters: {'n\_estimators': 154, 'max\_depth': 4, 'learning\_rate': 0.05316167408907679, 'subsample': 0.801497614849486, 'colsample\_bytree': 0.57879  
 [1] 2025-05-31 02:45:30.649 A new study created in memory with name: no-name\_5562a89c-7421-49df-b0b6-68fd95a5178  
 [1] 2025-05-31 02:45:30.740 Trial 0 finished with value: 0.021349953599301662 and parameters: {'n\_estimators': 52, 'max\_depth': 8, 'learning\_rate': 0.01794907005108507, 'subsample': 0.7295398083880831, 'colsample\_bytree': 0.866630  
 [1] 2025-05-31 02:45:30.859 Trial 1 finished with value: 0.021352016089995327 and parameters: {'n\_estimators': 213, 'max\_depth': 6, 'learning\_rate': 0.0716790239352953, 'subsample': 0.5791042929533, 'colsample\_bytree': 0.788332  
 [1] 2025-05-31 02:45:30.991 Trial 2 finished with value: 0.0213449311035734 and parameters: {'n\_estimators': 276, 'max\_depth': 4, 'learning\_rate': 0.06534184628757909, 'subsample': 0.672205200618658, 'colsample\_bytree': 0.55374468  
 [1] 2025-05-31 02:45:31.063 Trial 3 finished with value: 0.02354188669555 and parameters: {'n\_estimators': 63, 'max\_depth': 7, 'learning\_rate': 0.0588037053458814, 'subsample': 0.743493218827049, 'colsample\_bytree': 0.7273253  
 [1] 2025-05-31 02:45:31.135 Trial 4 finished with value: 0.021355191034551073 and parameters: {'n\_estimators': 73, 'max\_depth': 7, 'learning\_rate': 0.029561188922893, 'subsample': 0.879855398149384, 'colsample\_bytree': 0.8950097  
 [1] 2025-05-31 02:45:31.253 Trial 5 finished with value: 0.021325679697003125 and parameters: {'n\_estimators': 221, 'max\_depth': 6, 'learning\_rate': 0.07533141066509392, 'subsample': 0.979890796955952, 'colsample\_bytree': 0.858309  
 [1] 2025-05-31 02:45:31.320 Trial 6 finished with value: 0.02132761169168325 and parameters: {'n\_estimators': 67, 'max\_depth': 7, 'learning\_rate': 0.20191705671987, 'subsample': 0.6432669618846043, 'colsample\_bytree': 0.7652955  
 [1] 2025-05-31 02:45:31.427 Trial 7 finished with value: 0.021339416763508685 and parameters: {'n\_estimators': 183, 'max\_depth': 5, 'learning\_rate': 0.04952928610581683, 'subsample': 0.60899116317887, 'colsample\_bytree': 0.77369570  
 [1] 2025-05-31 02:45:31.545 Trial 8 finished with value: 0.0235346452843095 and parameters: {'n\_estimators': 240, 'max\_depth': 6, 'learning\_rate': 0.24580574205058212, 'subsample': 0.8264244940954838, 'colsample\_bytree': 0.520741  
 [1] 2025-05-31 02:45:31.622 Trial 9 finished with value: 0.02134245629031563 and parameters: {'n\_estimators': 98, 'max\_depth': 8, 'learning\_rate': 0.027427080059797003, 'subsample': 0.5444881105794984, 'colsample\_bytree': 0.5147439  
 [1] 2025-05-31 02:45:31.731 Trial 10 finished with value: 0.0235492179038659 and parameters: {'n\_estimators': 134, 'max\_depth': 3, 'learning\_rate': 0.2985252013754741, 'subsample': 0.6599054435576787, 'colsample\_bytree': 0.6754007  
 [1] 2025-05-31 02:45:31.862 Trial 11 finished with value: 0.021342617173259918 and parameters: {'n\_estimators': 155, 'max\_depth': 4, 'learning\_rate': 0.1582639884644466, 'subsample': 0.5065762300114467, 'colsample\_bytree': 0.99076  
 [1] 2025-05-31 02:45:31.991 Trial 12 finished with value: 0.0233440300240685 and parameters: {'n\_estimators': 182, 'max\_depth': 5, 'learning\_rate': 0.1539479384782875, 'subsample': 0.6153638680092021, 'colsample\_bytree': 0.649179  
 [1] 2025-05-31 02:45:32.099 Trial 13 finished with value: 0.023150036373534853 and parameters: {'n\_estimators': 125, 'max\_depth': 7, 'learning\_rate': 0.13087236932305054, 'subsample': 0.6497810766417882, 'colsample\_bytree': 0.652216  
 [1] 2025-05-31 02:45:32.222 Trial 14 finished with value: 0.02313431136620521 and parameters: {'n\_estimators': 176, 'max\_depth': 7, 'learning\_rate': 0.13040407382517396, 'subsample': 0.60899116317887, 'colsample\_bytree': 0.6129197  
 [1] 2025-05-31 02:45:32.342 Trial 15 finished with value: 0.021361462284520564 and parameters: {'n\_estimators': 104, 'max\_depth': 4, 'learning\_rate': 0.19789583204324, 'subsample': 0.69707442943333, 'colsample\_bytree': 0.715956  
 [1] 2025-05-31 02:45:32.453 Trial 16 finished with value: 0.02133951505214421 and parameters: {'n\_estimators': 193, 'max\_depth': 5, 'learning\_rate': 0.0970534261765276, 'subsample': 0.6075730515896463, 'colsample\_bytree': 0.599503  
 [1] 2025-05-31 02:45:32.618 Trial 17 finished with value: 0.0234997060891952 and parameters: {'n\_estimators': 296, 'max\_depth': 7, 'learning\_rate': 0.1125293344231493, 'subsample': 0.510532942286152, 'colsample\_bytree': 0.67799  
 [1] 2025-05-31 02:45:32.924 Trial 18 finished with value: 0.02315040307936608 and parameters: {'n\_estimators': 154, 'max\_depth': 8, 'learning\_rate': 0.19164948171550708, 'subsample': 0.7756551088512174, 'colsample\_bytree': 0.80698  
 [1] 2025-05-31 02:45:33.043 A new study created in memory with name: no-name\_b0d3db8a-cabc-41ee-9987-bbccc37a0172  
 [1] 2025-05-31 02:45:33.142 Trial 0 finished with value: 0.004612407019621683 and parameters: {'n\_estimators': 85, 'max\_depth': 7, 'learning\_rate': 0.03701601368918841, 'subsample': 0.96063977087058745, 'colsample\_bytree': 0.545262  
 [1] 2025-05-31 02:45:33.225 Trial 1 finished with value: 0.0046622216072995494 and parameters: {'n\_estimators': 112, 'max\_depth': 4, 'learning\_rate': 0.21271747207008462, 'subsample': 0.5036598553454614, 'colsample\_bytree': 0.64970  
 [1] 2025-05-31 02:45:33.289 Trial 2 finished with value: 0.0046158497892011501633 and parameters: {'n\_estimators': 57, 'max\_depth': 5, 'learning\_rate': 0.016720027453016133, 'subsample': 0.8705943883555484, 'colsample\_bytree': 0.82968  
 [1] 2025-05-31 02:45:33.407 Trial 3 finished with value: 0.004614321388460509 and parameters: {'n\_estimators': 208, 'max\_depth': 4, 'learning\_rate': 0.028303948921736975, 'subsample': 0.9670741586913192, 'colsample\_bytree': 0.79304  
 [1] 2025-05-31 02:45:33.471 Trial 4 finished with value: 0.004674222152451076 and parameters: {'n\_estimators': 59, 'max\_depth': 7, 'learning\_rate': 0.132523160382964, 'subsample': 0.683597045165368, 'colsample\_bytree': 0.64530327  
 [1] 2025-05-31 02:45:33.543 Trial 5 finished with value: 0.004625318729705865 and parameters: {'n\_estimators': 82, 'max\_depth': 3, 'learning\_rate': 0.028456681365965977, 'subsample': 0.7827240973921174, 'colsample\_bytree': 0.511912  
 [1] 2025-05-31 02:45:33.636 Trial 6 finished with value: 0.004607361365899619 and parameters: {'n\_estimators': 138, 'max\_depth': 7, 'learning\_rate': 0.05737852482310188, 'subsample': 0.7549678072872452, 'colsample\_bytree': 0.8840835  
 [1] 2025-05-31 02:45:33.742 Trial 7 finished with value: 0.00461341136620521 and parameters: {'n\_estimators': 176, 'max\_depth': 7, 'learning\_rate': 0.1304047382517396, 'subsample': 0.6814360873246824, 'colsample\_bytree': 0.608210  
 [1] 2025-05-31 02:45:33.842 Trial 8 finished with value: 0.0046236126852903 and parameters: {'n\_estimators': 104, 'max\_depth': 4, 'learning\_rate': 0.197894788595223, 'subsample': 0.69707422943333, 'colsample\_bytree': 0.715956  
 [1] 2025-05-31 02:45:33.889 Trial 9 finished with value: 0.00462358403521705016 and parameters: {'n\_estimators': 290, 'max\_depth': 6, 'learning\_rate': 0.178974878595223, 'subsample': 0.5111486810321957, 'colsample\_bytree': 0.692935  
 [1] 2025-05-31 02:45:33.915 Trial 10 finished with value: 0.0046107805203785565 and parameters: {'n\_estimators': 182, 'max\_depth': 6, 'learning\_rate': 0.0255495253070581543, 'subsample': 0.8808143858578332, 'colsample\_bytree': 0.80821  
 [1] 2025-05-31 02:45:34.044 Trial 11 finished with value: 0.00462358734801676 and parameters: {'n\_estimators': 242, 'max\_depth': 4, 'learning\_rate': 0.0386465046791571, 'subsample': 0.878974949214746, 'colsample\_bytree': 0.6626713  
 [1] 2025-05-31 02:45:34.208 Trial 12 finished with value: 0.00460836051467706795 and parameters: {'n\_estimators': 290, 'max\_depth': 8, 'learning\_rate': 0.09128187536071539, 'subsample': 0.5084085500759231, 'colsample\_bytree': 0.9551  
 [1] 2025-05-31 02:45:34.326 Trial 13 finished with value: 0.004618166840366 and parameters: {'n\_estimators': 153, 'max\_depth': 6, 'learning\_rate': 0.08106142277060937, 'subsample': 0.665811759755299, 'colsample\_bytree': 0.92782  
 [1] 2025-05-31 02:45:34.498 Trial 14 finished with value: 0.0046372129969284 and parameters: {'n\_estimators': 300, 'max\_depth': 8, 'learning\_rate': 0.0292277831295306, 'subsample': 0.6304903026098708, 'colsample\_bytree': 0.88680  
 [1] 2025-05-31 02:45:34.562 Trial 15 finished with value: 0.004625182790758565 and parameters: {'n\_estimators': 82, 'max\_depth': 3, 'learning\_rate': 0.028456681365965977, 'subsample': 0.7827240973921174, 'colsample\_bytree': 0.511912  
 [1] 2025-05-31 02:45:34.670 Trial 16 finished with value: 0.004607361365899619 and parameters: {'n\_estimators': 138, 'max\_depth': 7, 'learning\_rate': 0.05737852482310188, 'subsample': 0.7549678072872452, 'colsample\_bytree': 0.8840835  
 [1] 2025-05-31 02:45:34.956 Trial 17 finished with value: 0.00462453671018474 and parameters: {'n\_estimators': 262, 'max\_depth': 5, 'learning\_rate': 0.14985618384134405, 'subsample': 0.7419752424909314, 'colsample\_bytree': 0.99871  
 [1] 2025-05-31 02:45:35.072 Trial 18 finished with value: 0.004624536710162503792 and parameters: {'n\_estimators': 130, 'max\_depth': 7, 'learning\_rate': 0.12785336052307209, 'subsample': 0.578339363082648, 'colsample\_bytree': 0.856220  
 [1] 2025-05-31 02:45:35.210 Trial 19 finished with value: 0.004607521049521785 and parameters: {'n\_estimators': 183, 'max\_depth': 8, 'learning\_rate': 0.0524127818199802, 'subsample': 0.8181416454815448, 'colsample\_bytree': 0.5879  
 [1] 2025-05-31 02:45:35.342 Trial 20 finished with value: 0.00465773113450022 and parameters: {'n\_estimators': 195, 'max\_depth': 5, 'learning\_rate': 0.013046731352137395, 'subsample': 0.6841319373720284, 'colsample\_bytree': 0.705795  
 [1] 2025-05-31 02:45:35.466 Trial 21 finished with value: 0.00460688625901697 and parameters: {'n\_estimators': 228, 'max\_depth': 6, 'learning\_rate': 0.054644646334486899, 'subsample': 0.5631598916473186, 'colsample\_bytree': 0.799048  
 [1] 2025-05-31 02:45:35.642 A new study created in memory with name: no-name\_fa5121d4-067f-47f6-860c-6246c32053e5  
 [1] 2025-05-31 02:45:35.791 Trial 0 finished with value: 0.02625203169160916 and parameters: {'n\_estimators': 190, 'max\_depth': 3, 'learning\_rate': 0.2777389749491295, 'subsample': 0.609190315747387, 'colsample\_bytree': 0.567285498  
 [1] 2025-05-31 02:45:35.963 Trial 1 finished with value: 0.02644034531813896 and parameters: {'n\_estimators': 193, 'max\_depth': 6, 'learning\_rate': 0.0272596482629349136, 'subsample': 0.7347944993540203, 'colsample\_bytree': 0.94238  
 [1] 2025-05-31 02:45:36.067 Trial 2 finished with value: 0.026672270504825367 and parameters: {'n\_estimators': 220, 'max\_depth': 6, 'learning\_rate': 0.018139949160547, 'subsample': 0.674032115319719, 'colsample\_bytree': 0.8471691  
 [1] 2025-05-31 02:45:36.471 Trial 3 finished with value: 0.0245443159380082 and parameters: {'n\_estimators': 298, 'max\_depth': 5, 'learning\_rate': 0.178138568791383, 'subsample': 0.6839593540203, 'colsample\_bytree': 0.84615764  
 [1] 2025-05-31 02:45:36.340 Trial 4 finished with value: 0.024690035291724724 and parameters: {'n\_estimators': 244, 'max\_depth': 6, 'learning\_rate': 0.04778649012706503, 'subsample': 0.6649636565904211, 'colsample\_bytree': 0.78755  
 [1] 2025-05-31 02:45:36.492 Trial 5 finished with value: 0.026253884346029116 and parameters: {'n\_estimators': 281, 'max\_depth': 3, 'learning\_rate': 0.01467639062633397, 'subsample': 0.97451977533554, 'colsample\_bytree': 0.63895  
 [1] 2025-05-31 02:45:36.606 Trial 6 finished with value: 0.026185103251565899 and parameters: {'n\_estimators': 190, 'max\_depth': 6, 'learning\_rate': 0.04195967251541939, 'subsample': 0.95787112391216, 'colsample\_bytree': 0.95564367  
 [1] 2025-05-31 02:45:36.725 Trial 7 finished with value: 0.02664081593999893 and parameters: {'n\_estimators': 201, 'max\_depth': 6, 'learning\_rate': 0.0273583489058728, 'subsample': 0.7526060918822269, 'colsample\_bytree': 0.611926  
 [1] 2025-05-31 02:45:36.846 Trial 8 finished with value: 0.0264629693662051946, 'subsample': 0.5254117739575275, 'colsample\_bytree': 0.86205  
 [1] 2025-05-31 02:45:37.063 Trial 9 finished with value: 0.026397082997703374 and parameters: {'n\_estimators': 289, 'max\_depth': 4, 'learning\_rate': 0.0208049021341815465, 'subsample': 0.9205931037695768, 'colsample\_bytree': 0.884271  
 [1] 2025-05-31 02:45:37.163 Trial 10 finished with value: 0.0263282767292691 and parameters: {'n\_estimators': 75, 'max\_depth': 4, 'learning\_rate': 0.089456170822094097, 'subsample': 0.80289325173627624, 'colsample\_bytree': 0.7068288  
 [1] 2025-05-31 02:45:37.803 Trial 11 finished with value: 0.02638719199033774 and parameters: {'n\_estimators': 69, 'max\_depth': 8, 'learning\_rate': 0.0921929951483987, 'subsample': 0.584107136207264, 'colsample\_bytree': 0.721024  
 [1] 2025-05-31 02:45:37.929 Trial 12 finished with value: 0.026451615245474489 and parameters: {'n\_estimators': 60, 'max\_depth': 8, 'learning\_rate': 0.08997251228577, 'subsample': 0.849117082040907, 'colsample\_bytree': 0.6885160  
 [1] 2025-05-31 02:45:38.176 Trial 13 finished with value: 0.026508935657738694 and parameters: {'n\_estimators': 64, 'max\_depth': 8, 'learning\_rate': 0.081857968890333, 'subsample': 0.845930793131766, 'colsample\_bytree': 0.719625  
 [1] 2025-05-31 02:45:39.083 Trial 14 finished with value: 0.0266104913990923 and parameters: {'n\_estimators': 107, 'max\_depth': 7, 'learning\_rate': 0.07531530257286603, 'subsample': 0.83586040094460476, 'colsample\_bytree': 0.5136223  
 [1] 2025-05-31 02:45:39.204 Trial 15 finished with value: 0.0265517892943929 and parameters: {'n\_estimators': 109, 'max\_depth': 4, 'learning\_rate': 0.1305970695442496, 'subsample': 0.90650627217366, 'colsample\_bytree': 0.7279764  
 [1] 2025-05-31 02:45:39.357 Trial 16 finished with value: 0.02654631393673733

```
[1] 2025-05-31 02:45:45,667 Trial 7 finished with value: 0.004425208823767304 and parameters: {'n_estimators': 264, 'max_depth': 6, 'learning_rate': 0.2982419541547492, 'subsample': 0.9684366874561321, 'colsample_bytree': 0.62483}
[1] 2025-05-31 02:45:45,766 Trial 8 finished with value: 0.00548436593534788 and parameters: {'n_estimators': 148, 'max_depth': 3, 'learning_rate': 0.23804313912126812, 'subsample': 0.7230806189576238, 'colsample_bytree': 0.948649}
[1] 2025-05-31 02:45:45,906 Trial 9 finished with value: 0.005332167452467418 and parameters: {'n_estimators': 277, 'max_depth': 3, 'learning_rate': 0.15173664621244817, 'subsample': 0.62193666628384, 'colsample_bytree': 0.860197}
[1] 2025-05-31 02:45:45,994 Trial 10 finished with value: 0.0049778721919892269 and parameters: {'n_estimators': 58, 'max_depth': 5, 'learning_rate': 0.0124908620139898, 'subsample': 0.8527762732063922, 'colsample_bytree': 0.53906}
[1] 2025-05-31 02:45:46,087 Trial 11 finished with value: 0.005030331763760195 and parameters: {'n_estimators': 87, 'max_depth': 8, 'learning_rate': 0.01912030852293292, 'subsample': 0.5085987211480295, 'colsample_bytree': 0.617041}
[1] 2025-05-31 02:45:46,221 Trial 12 finished with value: 0.00469329614981149 and parameters: {'n_estimators': 215, 'max_depth': 5, 'learning_rate': 0.0686500120470423, 'subsample': 0.60925965705098932, 'colsample_bytree': 0.50621}
[1] 2025-05-31 02:45:46,354 Trial 13 finished with value: 0.0048443314136806265 and parameters: {'n_estimators': 207, 'max_depth': 7, 'learning_rate': 0.0569941936316296, 'subsample': 0.614740798461243, 'colsample_bytree': 0.52927}
[1] 2025-05-31 02:45:46,463 Trial 14 finished with value: 0.004871026029304299 and parameters: {'n_estimators': 112, 'max_depth': 5, 'learning_rate': 0.06894071364860477, 'subsample': 0.634653259896633, 'colsample_bytree': 0.6792}
[1] 2025-05-31 02:45:46,606 Trial 15 finished with value: 0.004940353177450971 and parameters: {'n_estimators': 210, 'max_depth': 7, 'learning_rate': 0.03491522996939215, 'subsample': 0.6614194116786875, 'colsample_bytree': 0.70202}
[1] 2025-05-31 02:45:46,716 Trial 16 finished with value: 0.00470221407139868 and parameters: {'n_estimators': 118, 'max_depth': 7, 'learning_rate': 0.01971196506440457, 'subsample': 0.571315378796992, 'colsample_bytree': 0.500901}
[1] 2025-05-31 02:45:46,854 Trial 17 finished with value: 0.00482484465997595 and parameters: {'n_estimators': 205, 'max_depth': 5, 'learning_rate': 0.01624526559105368, 'subsample': 0.7903389275615662, 'colsample_bytree': 0.57696}
[1] 2025-05-31 02:45:47,004 Trial 18 finished with value: 0.00484129347432567 and parameters: {'n_estimators': 232, 'max_depth': 4, 'learning_rate': 0.06708498805281377, 'subsample': 0.5182945549507605, 'colsample_bytree': 0.99470}
[1] 2025-05-31 02:45:47,110 Trial 19 finished with value: 0.00495614907904837 and parameters: {'n_estimators': 118, 'max_depth': 4, 'learning_rate': 0.012503427455635641, 'subsample': 0.593101724762666, 'colsample_bytree': 0.6638}
[1] 2025-05-31 02:45:47,209 A new study created in memory with name: no-name-759cc8-1183-4511-8ce2-b6901bf8db
[1] 2025-05-31 02:45:47,500 Trial 0 finished with value: 0.00378151760859237 and parameters: {'n_estimators': 226, 'max_depth': 7, 'learning_rate': 0.0257264944315098, 'subsample': 0.7694351350368266, 'colsample_bytree': 0.65899}
[1] 2025-05-31 02:45:47,630 Trial 1 finished with value: 0.003641804014835606 and parameters: {'n_estimators': 284, 'max_depth': 6, 'learning_rate': 0.12464596847365417, 'subsample': 0.945220685832382, 'colsample_bytree': 0.86105}
[1] 2025-05-31 02:45:47,747 Trial 2 finished with value: 0.003732624655595803 and parameters: {'n_estimators': 213, 'max_depth': 6, 'learning_rate': 0.10328320961393915, 'subsample': 0.613118395074961, 'colsample_bytree': 0.524443}
[1] 2025-05-31 02:45:47,832 Trial 4 finished with value: 0.00374979170199576684 and parameters: {'n_estimators': 118, 'max_depth': 5, 'learning_rate': 0.019717769462751589, 'subsample': 0.9057602448357913, 'colsample_bytree': 0.76863}
[1] 2025-05-31 02:45:47,965 Trial 5 finished with value: 0.003787374623595396 and parameters: {'n_estimators': 260, 'max_depth': 7, 'learning_rate': 0.123972748548211443, 'subsample': 0.9160583855151612, 'colsample_bytree': 0.83670}
[1] 2025-05-31 02:45:48,060 Trial 6 finished with value: 0.00377620788187783 and parameters: {'n_estimators': 130, 'max_depth': 6, 'learning_rate': 0.04438852174959533, 'subsample': 0.992709423329108, 'colsample_bytree': 0.8456}
[1] 2025-05-31 02:45:48,138 Trial 7 finished with value: 0.00378952154508788 and parameters: {'n_estimators': 104, 'max_depth': 7, 'learning_rate': 0.03079046853985559, 'subsample': 0.8831860588512, 'colsample_bytree': 0.68204}
[1] 2025-05-31 02:45:48,243 Trial 8 finished with value: 0.00384039695250206 and parameters: {'n_estimators': 192, 'max_depth': 6, 'learning_rate': 0.0536848855933933, 'subsample': 0.6889343826564234, 'colsample_bytree': 0.83147}
[1] 2025-05-31 02:45:48,363 Trial 9 finished with value: 0.003672813765136187 and parameters: {'n_estimators': 164, 'max_depth': 6, 'learning_rate': 0.1244713958197998, 'subsample': 0.788287835654608, 'colsample_bytree': 0.95165}
[1] 2025-05-31 02:45:48,436 Trial 10 finished with value: 0.0037800550899740777 and parameters: {'n_estimators': 51, 'max_depth': 3, 'learning_rate': 0.011414767812311186, 'subsample': 0.5111717571575849, 'colsample_bytree': 0.50680}
[1] 2025-05-31 02:45:48,559 Trial 11 finished with value: 0.00424271064060925 and parameters: {'n_estimators': 163, 'max_depth': 4, 'learning_rate': 0.01068761817012381, 'subsample': 0.5064466921631091, 'colsample_bytree': 0.5636}
[1] 2025-05-31 02:45:48,714 Trial 12 finished with value: 0.0035177796940429454 and parameters: {'n_estimators': 245, 'max_depth': 5, 'learning_rate': 0.1203789478672499, 'subsample': 0.6147421108099155, 'colsample_bytree': 0.9679}
[1] 2025-05-31 02:45:48,869 Trial 13 finished with value: 0.004033635259184422 and parameters: {'n_estimators': 240, 'max_depth': 4, 'learning_rate': 0.126900196354038636, 'subsample': 0.625338980946506, 'colsample_bytree': 0.99050}
[1] 2025-05-31 02:45:49,053 Trial 14 finished with value: 0.0034918882366033 and parameters: {'n_estimators': 295, 'max_depth': 5, 'learning_rate': 0.087559446479279595, 'subsample': 0.5982173515132352, 'colsample_bytree': 0.63456}
[1] 2025-05-31 02:45:49,196 Trial 15 finished with value: 0.00354911882366033 and parameters: {'n_estimators': 298, 'max_depth': 4, 'learning_rate': 0.070788209281576, 'subsample': 0.704981501001623, 'colsample_bytree': 0.6243}
[1] 2025-05-31 02:45:49,348 Trial 16 finished with value: 0.00401014451454917805 and parameters: {'n_estimators': 264, 'max_depth': 5, 'learning_rate': 0.19049054363869849, 'subsample': 0.565093287905254, 'colsample_bytree': 0.73879}
[1] 2025-05-31 02:45:49,532 Trial 17 finished with value: 0.004026308307649722 and parameters: {'n_estimators': 257, 'max_depth': 3, 'learning_rate': 0.08275132796470067, 'subsample': 0.573210292511696, 'colsample_bytree': 0.604507}
[1] 2025-05-31 02:45:49,699 Trial 18 finished with value: 0.0033478808803780794 and parameters: {'n_estimators': 299, 'max_depth': 5, 'learning_rate': 0.16553158451849267, 'subsample': 0.6823407080714137, 'colsample_bytree': 0.71145}
[1] 2025-05-31 02:45:49,889 Trial 19 finished with value: 0.003599181210469306 and parameters: {'n_estimators': 284, 'max_depth': 4, 'learning_rate': 0.17159141118842947, 'subsample': 0.71775919467353655, 'colsample_bytree': 0.71299}
[1] 2025-05-31 02:45:50,829 Trial 0 finished with value: 0.00792709306916208 and parameters: {'n_estimators': 136, 'max_depth': 5, 'learning_rate': 0.01474234109151542, 'subsample': 0.5019558619709378, 'colsample_bytree': 0.61584}
[1] 2025-05-31 02:45:50,969 Trial 1 finished with value: 0.0072443933753057 and parameters: {'n_estimators': 151, 'max_depth': 8, 'learning_rate': 0.06941443622652202, 'subsample': 0.6745961160150853, 'colsample_bytree': 0.730303}
[1] 2025-05-31 02:45:51,185 Trial 2 finished with value: 0.00788452630876472 and parameters: {'n_estimators': 189, 'max_depth': 8, 'learning_rate': 0.02199104632968396, 'subsample': 0.528179529334574, 'colsample_bytree': 0.54355}
[1] 2025-05-31 02:45:51,288 Trial 3 finished with value: 0.0079209595617627556 and parameters: {'n_estimators': 58, 'max_depth': 7, 'learning_rate': 0.02454653378788722, 'subsample': 0.7207718715769973, 'colsample_bytree': 0.8449320}
[1] 2025-05-31 02:45:52,148 Trial 4 finished with value: 0.00793547255164671 and parameters: {'n_estimators': 241, 'max_depth': 8, 'learning_rate': 0.0101978944947424, 'subsample': 0.9674141695510171, 'colsample_bytree': 0.82419}
[1] 2025-05-31 02:45:52,344 Trial 5 finished with value: 0.00795677083672358 and parameters: {'n_estimators': 130, 'max_depth': 5, 'learning_rate': 0.05878878933531534, 'subsample': 0.5779499765155456, 'colsample_bytree': 0.649390}
[1] 2025-05-31 02:45:52,522 Trial 6 finished with value: 0.007929734184550657 and parameters: {'n_estimators': 132, 'max_depth': 3, 'learning_rate': 0.05275888328709466, 'subsample': 0.75356693630865221, 'colsample_bytree': 0.905867}
[1] 2025-05-31 02:45:52,745 Trial 7 finished with value: 0.00794624762845245 and parameters: {'n_estimators': 207, 'max_depth': 8, 'learning_rate': 0.047626735265093, 'subsample': 0.847626735265093, 'colsample_bytree': 0.89862}
[1] 2025-05-31 02:45:53,019 Trial 8 finished with value: 0.0079185825500628797 and parameters: {'n_estimators': 298, 'max_depth': 4, 'learning_rate': 0.070788209281576, 'subsample': 0.58724471395819798, 'colsample_bytree': 0.6424495}
[1] 2025-05-31 02:45:53,151 Trial 9 finished with value: 0.00401014451454917805 and parameters: {'n_estimators': 264, 'max_depth': 5, 'learning_rate': 0.19049054363869849, 'subsample': 0.565093287905254, 'colsample_bytree': 0.73879}
[1] 2025-05-31 02:45:53,288 Trial 10 finished with value: 0.00728471742527273 and parameters: {'n_estimators': 62, 'max_depth': 6, 'learning_rate': 0.0491501583299007, 'subsample': 0.6046905366244298, 'colsample_bytree': 0.517407}
[1] 2025-05-31 02:45:53,322 Trial 11 finished with value: 0.0078878410717273 and parameters: {'n_estimators': 52, 'max_depth': 6, 'learning_rate': 0.2907182618294715, 'subsample': 0.5956758546728772, 'colsample_bytree': 0.530739}
[1] 2025-05-31 02:45:53,481 Trial 12 finished with value: 0.008078797401578955 and parameters: {'n_estimators': 290, 'max_depth': 6, 'learning_rate': 0.11648207517614551, 'subsample': 0.50240970157614551, 'colsample_bytree': 0.504326}
[1] 2025-05-31 02:45:53,578 Trial 13 finished with value: 0.007839213181324122 and parameters: {'n_estimators': 95, 'max_depth': 7, 'learning_rate': 0.03276316109009979, 'subsample': 0.608055123828269, 'colsample_bytree': 0.5634835}
[1] 2025-05-31 02:45:53,705 Trial 14 finished with value: 0.00782685347684457 and parameters: {'n_estimators': 90, 'max_depth': 7, 'learning_rate': 0.03026068960563494, 'subsample': 0.643384669088715, 'colsample_bytree': 0.5712821}
[1] 2025-05-31 02:45:53,823 Trial 15 finished with value: 0.007978659203245037 and parameters: {'n_estimators': 93, 'max_depth': 7, 'learning_rate': 0.1096083423359934, 'subsample': 0.664600698979322, 'colsample_bytree': 0.727828}
[1] 2025-05-31 02:45:53,904 Trial 16 finished with value: 0.007842199008169195 and parameters: {'n_estimators': 90, 'max_depth': 6, 'learning_rate': 0.140278817116758, 'subsample': 0.758254787095606, 'colsample_bytree': 0.720482}
[1] 2025-05-31 02:45:53,238 Trial 17 finished with value: 0.007824871742527273 and parameters: {'n_estimators': 62, 'max_depth': 6, 'learning_rate': 0.2491501583299007, 'subsample': 0.6049053662144299, 'colsample_bytree': 0.517407}
[1] 2025-05-31 02:45:53,322 Trial 18 finished with value: 0.0078878410717273 and parameters: {'n_estimators': 52, 'max_depth': 6, 'learning_rate': 0.2907182618294715, 'subsample': 0.5956758546728772, 'colsample_bytree': 0.530739}
[1] 2025-05-31 02:45:53,481 Trial 19 finished with value: 0.008078797401578955 and parameters: {'n_estimators': 141, 'max_depth': 6, 'learning_rate': 0.11648207517614551, 'subsample': 0.50240970157614551, 'colsample_bytree': 0.504326}
[1] 2025-05-31 02:45:54,084 A new study created in memory with name: no-name-780860c-f39f-46fe-8683-1360f84654ca
[1] 2025-05-31 02:45:55,091 Trial 0 finished with value: 0.0024667283275019 and parameters: {'n_estimators': 152, 'max_depth': 8, 'learning_rate': 0.034362442930969, 'subsample': 0.5328566121274992, 'colsample_bytree': 0.592348}
[1] 2025-05-31 02:45:55,301 Trial 1 finished with value: 0.0024012032626723896 and parameters: {'n_estimators': 149, 'max_depth': 5, 'learning_rate': 0.011857154146494832, 'subsample': 0.8657181456920647, 'colsample_bytree': 0.6233}
[1] 2025-05-31 02:45:55,880 Trial 2 finished with value: 0.00240478190791676176 and parameters: {'n_estimators': 238, 'max_depth': 8, 'learning_rate': 0.1579882836940283, 'subsample': 0.5281939906761886, 'colsample_bytree': 0.77007}
[1] 2025-05-31 02:45:56,427 Trial 3 finished with value: 0.002416570371478746 and parameters: {'n_estimators': 279, 'max_depth': 5, 'learning_rate': 0.025920662659795317, 'subsample': 0.947581382898973, 'colsample_bytree': 0.80565}
[1] 2025-05-31 02:45:56,599 Trial 4 finished with value: 0.0024072177539371392 and parameters: {'n_estimators': 112, 'max_depth': 3, 'learning_rate': 0.139246259303472, 'subsample': 0.7463265267252196, 'colsample_bytree': 0.906486}
[1] 2025-05-31 02:45:56,750 Trial 5 finished with value: 0.0024261083622052287 and parameters: {'n_estimators': 104, 'max_depth': 4, 'learning_rate': 0.02054826253923921, 'subsample': 0.8808072970945686, 'colsample_bytree': 0.8175673}
[1] 2025-05-31 02:45:56,950 Trial 6 finished with value: 0.0024062476921785562 and parameters: {'n_estimators': 188, 'max_depth': 8, 'learning_rate': 0.05854275858536221, 'subsample': 0.9109403615023309, 'colsample_bytree': 0.99341}
[1] 2025-05-31 02:45:57,140 Trial 7 finished with value: 0.0024235622394657634 and parameters: {'n_estimators': 236, 'max_depth': 3, 'learning_rate': 0.190098386404591734, 'subsample': 0.92062494466619316, 'colsample_bytree': 0.50561}
[1] 2025-05-31 02:45:57,246 Trial 8 finished with value: 0.002416591502519283 and parameters: {'n_estimators': 184, 'max_depth': 4, 'learning_rate': 0.0229860331313117, 'subsample': 0.925163376796557, 'colsample_bytree': 0.7906791}
[1] 2025-05-31 02:45:57,300 Trial 9 finished with value: 0.0024138974049978862 and parameters: {'n_estimators': 204, 'max_depth': 7, 'learning_rate': 0.0194060424448105, 'subsample': 0.97445646464684369, 'colsample_bytree': 0.68541}
[1] 2025-05-31 02:45:57,447 Trial 10 finished with value: 0.002430108499303037 and parameters: {'n_estimators': 55, 'max_depth': 7, 'learning_rate': 0.07432478175232023, 'subsample': 0.7357910867164484, 'colsample_bytree': 0.966688}
[1] 2025-05-31 02:45:57,547 Trial 11 finished with value: 0.00240131001849734992 and parameters: {'n_estimators': 107, 'max_depth': 6, 'learning_rate': 0.09264063308419402, 'subsample': 0.730576976076363, 'colsample_bytree': 0.9850}
[1] 2025-05-31 02:45:57,646 Trial 12 finished with value: 0.00240109332075875586 and parameters: {'n_estimators': 106, 'max_depth': 6, 'learning_rate': 0.117766483216726, 'subsample': 0.64751805209070917, 'colsample_bytree': 0.89690}
[1] 2025-05-31 02:45:57,735 Trial 13 finished with value: 0.00244336197170453153 and parameters: {'n_estimators': 51, 'max_depth': 3, 'learning_rate': 0.0525744726720765318, 'subsample': 0.81647466050053117, 'colsample_bytree': 0.9280}
[1] 2025-05-31 02:45:57,883 Trial 14 finished with value: 0.00237716170302303 and parameters: {'n_estimators': 143, 'max_depth': 7, 'learning_rate': 0.2628830782910907, 'subsample': 0.682055123828269, 'colsample_bytree': 0.87108}
[1] 2025-05-31 02:45:58,083 Trial 15 finished with value: 0.00242635622394657634 and parameters: {'n_estimators': 158, 'max_depth': 6, 'learning_rate': 0.0289617940936685685, 'subsample': 0.6512615943972304, 'colsample_bytree': 0.86325}
[1] 2025-05-31 02:45:58,146 Trial 16 finished with value: 0.002417605760036805 and parameters: {'n_estimators': 211, 'max_depth': 7, 'learning_rate': 0.0559313966784133, 'subsample': 0.623575075593923, 'colsample_bytree': 0.991455}
[1] 2025-05-31 02:45:58,312 Trial 17 finished with value: 0.0024528645894081608 and parameters: {'n_estimators': 291, 'max_depth': 8, 'learning_rate': 0.02055937161160701, 'subsample': 0.8990342695205346, 'colsample_bytree': 0.870861}
[1] 2025-05-31 02:45:58,437 Trial 18 finished with value: 0.0020404148996476412 and parameters: {'n_estimators': 181, 'max_depth': 7, 'learning_rate': 0.03641924382038183, 'subsample': 0.5916340691143898, 'colsample_bytree': 0.69827}
[1] 2025-05-31 02:45:58,550 Trial 19 finished with value: 0.00241204256045804508 and parameters: {'n_estimators': 145, 'max_depth': 6, 'learning_rate': 0.030596126840680708 and parameters: {'n_estimators': 208, 'max_depth': 7, 'learning_rate': 0.02372108717917986, 'subsample': 0.6900873493376344, 'colsample_bytree': 0.6952}
[1] 2025-05-31 02:45:58,651 A new study created in memory with name: no-name-9e26d926-268e-44ac-7052-15592607fe07
[1] 2025-05-31 02:45:59,201 Trial 0 finished with value: 0.003031196217163394 and parameters: {'n_estimators': 75, 'max_depth': 4, 'learning_rate': 0.0567604432656153, 'subsample': 0.548905629491355, 'colsample_bytree': 0.7325661}
[1] 2025-05-31 02:45:59,383 Trial 1 finished with value: 0.003186763623942697 and parameters: {'n_estimators': 248, 'max_depth': 8, 'learning_rate': 0.0980495621370138, 'subsample': 0.97700245386970595, 'colsample_bytree': 0.650651}
[1] 2025-05-31 02:45:59,713 Trial 3 finished with value: 0.003152769449481753 and parameters: {'n_estimators': 61, 'max_depth': 4, 'learning_rate': 0.19935762829107007, 'subsample': 0.889225184553617, 'colsample_bytree': 0.78168}
[1] 2025-05-31 02:45:59,240 Trial 4 finished with value: 0.003016517432585642 and parameters: {'n_estimators': 64, 'max_depth': 7, 'learning_rate': 0.047649905195396, 'subsample': 0.729558858760978094, 'colsample_bytree': 0.611276}
[1] 2025-05-31 02:45:59,325 Trial 5 finished with value: 0.003266697619273636 and parameters: {'n_estimators': 117, 'max_depth': 7, 'learning_rate': 0.02209205618970459, 'subsample': 0.6455282437617714, 'colsample_bytree': 0.604945}
[1] 2025-05-31 02:45:59,426 Trial 6 finished with value: 0.0031494651753570983 and parameters: {'n_estimators': 177, 'max_depth': 8, 'learning_rate': 0.02619449013547369, 'subsample': 0.5695760307184325, 'colsample_bytree': 0.66589}
[1] 2025-05-31 02:4
```

[I] 2025-05-31 02:46:07,7003 Trial 14 finished with value: 0.0067460171060593269 and parameters: {'n\_estimators': 131, 'max\_depth': 5, 'learning\_rate': 0.03717447526935097}, 'subsample': 0.7953533445106004, 'colsample\_bytree': 0.747026

[I] 2025-05-31 02:46:07,6411 Trial 15 finished with value: 0.0087468204558213 and parameters: {'n\_estimators': 208, 'max\_depth': 7, 'learning\_rate': 0.0511660929419991}, 'subsample': 0.610725611491809, 'colsample\_bytree': 0.58456

[I] 2025-05-31 02:46:07,755 Trial 16 finished with value: 0.0087019516875675648 and parameters: {'n\_estimators': 145, 'max\_depth': 6, 'learning\_rate': 0.028588212521127}, 'subsample': 0.7640267603131633, 'colsample\_bytree': 0.71011

[I] 2025-05-31 02:46:08,036 Trial 18 finished with value: 0.00873830754218123 and parameters: {'n\_estimators': 210, 'max\_depth': 7, 'learning\_rate': 0.072443511040338}, 'subsample': 0.697956709410262, 'colsample\_bytree': 0.69866821

[I] 2025-05-31 02:46:08,147 Trial 19 finished with value: 0.008764607958087498 and parameters: {'n\_estimators': 137, 'max\_depth': 4, 'learning\_rate': 0.07302490329393875}, 'subsample': 0.83099379205014, 'colsample\_bytree': 0.727536

[I] 2025-05-31 02:46:08,266 A new study created in memory with name: no-name-8ade6522-3227-4086-8719-9e02abeffa

[I] 2025-05-31 02:46:08,426 Trial 0 finished with value: 0.002425438921972021 and parameters: {'n\_estimators': 271, 'max\_depth': 8, 'learning\_rate': 0.029975038178459562}, 'subsample': 0.511190436528923, 'colsample\_bytree': 0.639627

[I] 2025-05-31 02:46:08,530 Trial 1 finished with value: 0.002434143225967933 and parameters: {'n\_estimators': 176, 'max\_depth': 6, 'learning\_rate': 0.0285915001976341}, 'subsample': 0.6312638643634246, 'colsample\_bytree': 0.82079

[I] 2025-05-31 02:46:08,624 Trial 2 finished with value: 0.002527510829959316 and parameters: {'n\_estimators': 152, 'max\_depth': 3, 'learning\_rate': 0.02482154098394516, 'subsample': 0.659178523306287, 'colsample\_bytree': 0.6303404

[I] 2025-05-31 02:46:08,695 Trial 3 finished with value: 0.002448736117701702 and parameters: {'n\_estimators': 71, 'max\_depth': 4, 'learning\_rate': 0.05413580397453646, 'subsample': 0.842362390842506, 'colsample\_bytree': 0.894662

[I] 2025-05-31 02:46:08,799 Trial 4 finished with value: 0.00245741520956433 and parameters: {'n\_estimators': 144, 'max\_depth': 8, 'learning\_rate': 0.05117983400236627}, 'subsample': 0.648625263615212, 'colsample\_bytree': 0.8237155

[I] 2025-05-31 02:46:08,880 Trial 5 finished with value: 0.0024812633295857527 and parameters: {'n\_estimators': 139, 'max\_depth': 4, 'learning\_rate': 0.01309856656900739}, 'subsample': 0.8711125640276728, 'colsample\_bytree': 0.6586

[I] 2025-05-31 02:46:08,991 Trial 6 finished with value: 0.00249785669048891 and parameters: {'n\_estimators': 194, 'max\_depth': 5, 'learning\_rate': 0.0215402163039998, 'subsample': 0.904136119022191, 'colsample\_bytree': 0.8140758

[I] 2025-05-31 02:46:09,072 Trial 7 finished with value: 0.002454256214015093 and parameters: {'n\_estimators': 111, 'max\_depth': 4, 'learning\_rate': 0.06224278370354031}, 'subsample': 0.791899399040216, 'colsample\_bytree': 0.97808

[I] 2025-05-31 02:46:09,180 Trial 8 finished with value: 0.00205684931945037 and parameters: {'n\_estimators': 192, 'max\_depth': 4, 'learning\_rate': 0.0376689194216645}, 'subsample': 0.82457426595298, 'colsample\_bytree': 0.6651

[I] 2025-05-31 02:46:09,296 Trial 9 finished with value: 0.002417353951748074 and parameters: {'n\_estimators': 157, 'max\_depth': 8, 'learning\_rate': 0.178387562029383, 'subsample': 0.60294086402029, 'colsample\_bytree': 0.5811923

[I] 2025-05-31 02:46:09,442 Trial 10 finished with value: 0.00246397619619366 and parameters: {'n\_estimators': 256, 'max\_depth': 7, 'learning\_rate': 0.16259856618410469}, 'subsample': 0.516792869721139, 'colsample\_bytree': 0.52522

[I] 2025-05-31 02:46:09,594 Trial 11 finished with value: 0.0024474524417316306 and parameters: {'n\_estimators': 278, 'max\_depth': 8, 'learning\_rate': 0.12855102810411671}, 'subsample': 0.512204034904197, 'colsample\_bytree': 0.5018

[I] 2025-05-31 02:46:09,740 Trial 12 finished with value: 0.0024787605938844 and parameters: {'n\_estimators': 233, 'max\_depth': 7, 'learning\_rate': 0.1071546216010073}, 'subsample': 0.6252521234162126, 'colsample\_bytree': 0.6080

[I] 2025-05-31 02:46:09,883 Trial 13 finished with value: 0.00249785669048891 and parameters: {'n\_estimators': 226, 'max\_depth': 7, 'learning\_rate': 0.01248212155387825}, 'subsample': 0.7114011478905811, 'colsample\_bytree': 0.7217

[I] 2025-05-31 02:46:10,042 Trial 14 finished with value: 0.002471605703265925 and parameters: {'n\_estimators': 290, 'max\_depth': 8, 'learning\_rate': 0.06308515019457399}, 'subsample': 0.5636565001682459, 'colsample\_bytree': 0.57102

[I] 2025-05-31 02:46:10,142 Trial 15 finished with value: 0.002539701052684928 and parameters: {'n\_estimators': 98, 'max\_depth': 6, 'learning\_rate': 0.155508258772885078, 'subsample': 0.580709792245149, 'colsample\_bytree': 0.57023

[I] 2025-05-31 02:46:10,315 Trial 16 finished with value: 0.00247877825374074 and parameters: {'n\_estimators': 220, 'max\_depth': 7, 'learning\_rate': 0.071428579179369, 'subsample': 0.717870915980232, 'colsample\_bytree': 0.7224

[I] 2025-05-31 02:46:10,472 Trial 17 finished with value: 0.0024182364265968 and parameters: {'n\_estimators': 291, 'max\_depth': 8, 'learning\_rate': 0.1885726399084026, 'subsample': 0.5807403991719679, 'colsample\_bytree': 0.5555

[I] 2025-05-31 02:46:10,580 Trial 18 finished with value: 0.002400218359834102 and parameters: {'n\_estimators': 115, 'max\_depth': 6, 'learning\_rate': 0.08965971916702562, 'subsample': 0.578300804040518, 'colsample\_bytree': 0.6831

[I] 2025-05-31 02:46:10,667 Trial 19 finished with value: 0.00242773842659744688 and parameters: {'n\_estimators': 54, 'max\_depth': 8, 'learning\_rate': 0.04859793080056949, 'subsample': 0.6839025119275646, 'colsample\_bytree': 0.57174

[I] 2025-05-31 02:46:10,778 A new study created in memory with name: no-name-3ab5ae3f-b957-40fe-b888-a866cef61f

[I] 2025-05-31 02:46:10,916 Trial 0 finished with value: 0.0038076873606925833 and parameters: {'n\_estimators': 163, 'max\_depth': 3, 'learning\_rate': 0.02289850956911346, 'subsample': 0.6248342700401823, 'colsample\_bytree': 0.8870

[I] 2025-05-31 02:46:11,056 Trial 1 finished with value: 0.0036973638067308795 and parameters: {'n\_estimators': 263, 'max\_depth': 6, 'learning\_rate': 0.045779525695650326, 'subsample': 0.5756842337691046, 'colsample\_bytree': 0.9566

[I] 2025-05-31 02:46:11,187 Trial 2 finished with value: 0.00381987459574178 and parameters: {'n\_estimators': 260, 'max\_depth': 7, 'learning\_rate': 0.010751464612996976, 'subsample': 0.8665033853470407, 'colsample\_bytree': 0.81163

[I] 2025-05-31 02:46:11,254 Trial 3 finished with value: 0.00379323419064774 and parameters: {'n\_estimators': 62, 'max\_depth': 8, 'learning\_rate': 0.0136252833998953, 'subsample': 0.645745370293011, 'colsample\_bytree': 0.967025

[I] 2025-05-31 02:46:11,400 Trial 4 finished with value: 0.0037563471879476367 and parameters: {'n\_estimators': 267, 'max\_depth': 7, 'learning\_rate': 0.02663076728093055, 'subsample': 0.76945925205952, 'colsample\_bytree': 0.91385

[I] 2025-05-31 02:46:11,466 Trial 5 finished with value: 0.003789505365620391 and parameters: {'n\_estimators': 70, 'max\_depth': 5, 'learning\_rate': 0.04391117887708834, 'subsample': 0.797461583611354, 'colsample\_bytree': 0.6693994

[I] 2025-05-31 02:46:11,579 Trial 6 finished with value: 0.003784713442056575 and parameters: {'n\_estimators': 222, 'max\_depth': 6, 'learning\_rate': 0.02911295680040147, 'subsample': 0.738683148418992, 'colsample\_bytree': 0.5906

[I] 2025-05-31 02:46:11,679 Trial 7 finished with value: 0.0038793783927494585 and parameters: {'n\_estimators': 154, 'max\_depth': 8, 'learning\_rate': 0.0227176033719394912, 'subsample': 0.678712611500389, 'colsample\_bytree': 0.7727

[I] 2025-05-31 02:46:11,807 Trial 8 finished with value: 0.003808356100803103 and parameters: {'n\_estimators': 239, 'max\_depth': 7, 'learning\_rate': 0.013938221078875154, 'subsample': 0.705245943618966, 'colsample\_bytree': 0.89158

[I] 2025-05-31 02:46:11,895 Trial 9 finished with value: 0.003862661247393917 and parameters: {'n\_estimators': 108, 'max\_depth': 8, 'learning\_rate': 0.02925854644283662, 'subsample': 0.6448545334057562, 'colsample\_bytree': 0.7185

[I] 2025-05-31 02:46:12,054 Trial 10 finished with value: 0.00390966113549883 and parameters: {'n\_estimators': 292, 'max\_depth': 4, 'learning\_rate': 0.17873308338651073, 'subsample': 0.511888313426322, 'colsample\_bytree': 0.50255

[I] 2025-05-31 02:46:12,224 Trial 11 finished with value: 0.003835416087525403 and parameters: {'n\_estimators': 300, 'max\_depth': 6, 'learning\_rate': 0.07526295603099073, 'subsample': 0.98363632456625, 'colsample\_bytree': 0.99366

[I] 2025-05-31 02:46:12,394 Trial 12 finished with value: 0.003805967334946778 and parameters: {'n\_estimators': 207, 'max\_depth': 5, 'learning\_rate': 0.060105307336484891, 'subsample': 0.520021512998216, 'colsample\_bytree': 0.89023

[I] 2025-05-31 02:46:12,548 Trial 13 finished with value: 0.00379596423979966 and parameters: {'n\_estimators': 258, 'max\_depth': 7, 'learning\_rate': 0.046651820907571864, 'subsample': 0.831617459572528, 'colsample\_bytree': 0.938

[I] 2025-05-31 02:46:12,683 Trial 14 finished with value: 0.0041902885241129 and parameters: {'n\_estimators': 190, 'max\_depth': 6, 'learning\_rate': 0.1902556479466628, 'subsample': 0.578686195921775, 'colsample\_bytree': 0.83075

[I] 2025-05-31 02:46:12,846 Trial 15 finished with value: 0.003796268121724197 and parameters: {'n\_estimators': 269, 'max\_depth': 7, 'learning\_rate': 0.0962456964446437, 'subsample': 0.8191337538143615, 'colsample\_bytree': 0.9970

[I] 2025-05-31 02:46:12,994 Trial 16 finished with value: 0.0037471015260179004 and parameters: {'n\_estimators': 233, 'max\_depth': 5, 'learning\_rate': 0.044507429547237847, 'subsample': 0.760752195683285, 'colsample\_bytree': 0.928

[I] 2025-05-31 02:46:13,105 Trial 17 finished with value: 0.0038181549921076763 and parameters: {'n\_estimators': 130, 'max\_depth': 4, 'learning\_rate': 0.04464603124377844, 'subsample': 0.8970176346039406, 'colsample\_bytree': 0.84114

[I] 2025-05-31 02:46:13,247 Trial 18 finished with value: 0.0034967911984653 and parameters: {'n\_estimators': 228, 'max\_depth': 4, 'learning\_rate': 0.0951904710316268, 'subsample': 0.5808940283516587, 'colsample\_bytree': 0.698133

[I] 2025-05-31 02:46:13,409 Trial 19 finished with value: 0.00406547614861359 and parameters: {'n\_estimators': 189, 'max\_depth': 4, 'learning\_rate': 0.1251742747769775, 'subsample': 0.572338347564482, 'colsample\_bytree': 0.657507

[I] 2025-05-31 02:46:13,534 A new study created in memory with name: no-name-a418612-bd1a-4efb-acbb-c884728421aa

[I] 2025-05-31 02:46:13,652 Trial 0 finished with value: 0.010406351295026 and parameters: {'n\_estimators': 114, 'max\_depth': 5, 'learning\_rate': 0.0502895409137331, 'subsample': 0.5492918691194254, 'colsample\_bytree': 0.897036

[I] 2025-05-31 02:46:13,737 Trial 1 finished with value: 0.010482430976236737 and parameters: {'n\_estimators': 119, 'max\_depth': 6, 'learning\_rate': 0.132194150679032, 'subsample': 0.6102081377625468, 'colsample\_bytree': 0.6063847

[I] 2025-05-31 02:46:13,850 Trial 2 finished with value: 0.01056858204738364 and parameters: {'n\_estimators': 198, 'max\_depth': 7, 'learning\_rate': 0.17924590170271024, 'subsample': 0.9375011697043663, 'colsample\_bytree': 0.782952

[I] 2025-05-31 02:46:13,969 Trial 3 finished with value: 0.01041866120232057 and parameters: {'n\_estimators': 221, 'max\_depth': 8, 'learning\_rate': 0.0676305028849226, 'subsample': 0.714660270418086, 'colsample\_bytree': 0.693759

[I] 2025-05-31 02:46:14,064 Trial 4 finished with value: 0.010437410526017904 and parameters: {'n\_estimators': 233, 'max\_depth': 5, 'learning\_rate': 0.04037410526017904, 'subsample': 0.507245912059715989, 'colsample\_bytree': 0.7140240

[I] 2025-05-31 02:46:14,186 Trial 5 finished with value: 0.01044586420153461 and parameters: {'n\_estimators': 226, 'max\_depth': 6, 'learning\_rate': 0.0882534541698038, 'subsample': 0.8371318498321643, 'colsample\_bytree': 0.936662

[I] 2025-05-31 02:46:14,319 Trial 6 finished with value: 0.01034206425594446 and parameters: {'n\_estimators': 252, 'max\_depth': 6, 'learning\_rate': 0.236973105198816, 'subsample': 0.8813903200994989, 'colsample\_bytree': 0.727373

[I] 2025-05-31 02:46:14,488 Trial 7 finished with value: 0.010529704054582294 and parameters: {'n\_estimators': 154, 'max\_depth': 5, 'learning\_rate': 0.01374558910839563, 'subsample': 0.55827623959458, 'colsample\_bytree': 0.59544

[I] 2025-05-31 02:46:14,560 Trial 8 finished with value: 0.010505367175979941 and parameters: {'n\_estimators': 196, 'max\_depth': 8, 'learning\_rate': 0.01031908219819353, 'subsample': 0.552453401376168, 'colsample\_bytree': 0.560327

[I] 2025-05-31 02:46:14,670 Trial 9 finished with value: 0.010319052959724031 and parameters: {'n\_estimators': 189, 'max\_depth': 6, 'learning\_rate': 0.02494266365079188, 'subsample': 0.6642628182865406, 'colsample\_bytree': 0.6944491

[I] 2025-05-31 02:46:14,765 Trial 10 finished with value: 0.010537447747468469 and parameters: {'n\_estimators': 65, 'max\_depth': 3, 'learning\_rate': 0.02811953330370337, 'subsample': 0.6783538802206213, 'colsample\_bytree': 0.87200

[I] 2025-05-31 02:46:14,929 Trial 11 finished with value: 0.010574886190630526 and parameters: {'n\_estimators': 294, 'max\_depth': 3, 'learning\_rate': 0.0803788272735062, 'subsample': 0.8037885448363051, 'colsample\_bytree': 0.86867029

[I] 2025-05-31 02:46:15,087 Trial 12 finished with value: 0.010409589471596969 and parameters: {'n\_estimators': 269, 'max\_depth': 7, 'learning\_rate': 0.0284722115327389, 'subsample': 0.8949385244430055, 'colsample\_bytree': 0.814124

[I] 2025-05-31 02:46:15,200 Trial 13 finished with value: 0.010405017224820253 and parameters: {'n\_estimators': 251, 'max\_depth': 4, 'learning\_rate': 0.16389720723509849, 'subsample': 0.65218460260027621, 'colsample\_bytree': 0.50812

[I] 2025-05-31 02:46:15,377 Trial 14 finished with value: 0.01034574496720402 and parameters: {'n\_estimators': 251, 'max\_depth': 7, 'learning\_rate': 0.11977760560492845, 'subsample': 0.7723080728476063, 'colsample\_bytree': 0.67916

[I] 2025-05-31 02:46:15,519 Trial 15 finished with value: 0.0105324571888527 and parameters: {'n\_estimators': 173, 'max\_depth': 6, 'learning\_rate': 0.2234057161247246, 'subsample': 0.8831856335682771, 'colsample\_bytree': 0.84556

[I] 2025-05-31 02:46:15,687 Trial 16 finished with value: 0.010484326737003513 and parameters: {'n\_estimators': 297, 'max\_depth': 4, 'learning\_rate': 0.024194068736284864, 'subsample': 0.742341643019993, 'colsample\_bytree': 0.74777

[I] 2025-05-31 02:46:15,826 Trial 17 finished with value: 0.01046394046486469 and parameters: {'n\_estimators': 228, 'max\_depth': 7, 'learning\_rate': 0.09656076236764381, 'subsample': 0.6308491474768625, 'colsample\_bytree': 0.633415

[I] 2025-05-31 02:46:15,958 Trial 18 finished with value: 0.0106263498761071606 and parameters: {'n\_estimators': 198, 'max\_depth': 6, 'learning\_rate': 0.038500565986547906, 'subsample': 0.81800056598656233, 'colsample\_bytree': 0.720877

[I] 2025-05-31 02:46:16,046 Trial 19 finished with value: 0.010492908298703586 and parameters: {'n\_estimators': 50, 'max\_depth': 4, 'learning\_rate': 0.03527468819499045, 'subsample': 0.885931949616817, 'colsample\_bytree': 0.64899

[I] 2025-05-31 02:46:16,189 A new study created in memory with name: no-name-3072392d-f8d4-404a-a910e0461049411

[I] 2025-05-31 02:46:16,394 Trial 0 finished with value: 0.0038935058079494493 and parameters: {'n\_estimators': 140, 'max\_depth': 8, 'learning\_rate': 0.0838827763803025, 'subsample': 0.7058077144484937, 'colsample\_bytree': 0.51160

[I] 2025-05-31 02:46:17,155 Trial 1 finished with value: 0.00391742704609946 and parameters: {'n\_estimators': 270, 'max\_depth': 4, 'learning\_rate': 0.02927139747443437, 'subsample': 0.981538498308954, 'colsample\_bytree': 0.77535

[I] 2025-05-31 02:46:17,373 Trial 2 finished with value: 0.0039172034609946 and parameters: {'n\_estimators': 267, 'max\_depth': 6, 'learning\_rate': 0.02016026537654524, 'subsample': 0.9327142677591768, 'colsample\_bytree': 0.871314

[I] 2025-05-31 02:46:18,434 Trial 3 finished with value: 0.003910459326932402 and parameters: {'n\_estimators': 287, 'max\_depth': 4, 'learning\_rate': 0.0102321384614238848, 'subsample': 0.947527125702642, 'colsample\_bytree': 0.683857

[I] 2025-05-31 02:46:18,693 Trial 4 finished with value: 0.003766260082724166 and parameters: {'n\_estimators': 270, 'max\_depth': 6, 'learning\_rate': 0.028574876056199159, 'subsample': 0.894112376157863, 'colsample\_bytree': 0.73562

[I] 2025-05-31 02:46:18,876 Trial 5 finished with value: 0.0038211245282049 and parameters: {'n\_estimators': 194, 'max\_depth': 5, 'learning\_rate': 0.01000593917138034, 'subsample': 0.8410683830202618, 'colsample\_bytree': 0.61260

[I] 2025-05-31 02:46:19,095 Trial 6 finished with value: 0.003661018625671144 and parameters: {'n\_estimators': 192, 'max\_depth': 7, 'learning\_rate': 0.019537507703460955, 'subsample': 0.52547162160219946, 'colsample\_bytree': 0.60716

[I] 2025-05-31 02:46:19,268 Trial 14 finished with value: 0.003679820425613209 and parameters: {'n\_estimators': 164, 'max\_depth': 8, 'learning\_rate': 0.02202015228617776, 'subsample': 0.53527445079463192, 'colsample\_bytree': 0.504606

[I] 2025-05-31 02:46:19,446 Trial 8 finished with value: 0.00381370884893711 and parameters: {'n\_estimators': 264, 'max\_depth': 8, 'learning\_rate': 0.014704024344921395, 'subsample': 0.520723742370513, 'colsample\_bytree': 0.5017

[I] 2025-05-31 02:46:19,533 Trial 9 finished with value: 0.003817139052021037 and parameters: {'n\_estimators': 133, 'max\_depth': 8, 'learning\_rate': 0.01054704203443196249, 'subsample': 0.7625127049274178, 'colsample\_bytree': 0.542997

[I] 2025-05-31 02:46:19,775 Trial 11 finished with value: 0.00369512157290646 and parameters: {'n\_estimators': 190, 'max\_depth': 3, 'learning\_rate': 0.04141756493233721575, 'subsample': 0.6308896803023254, 'colsample\_bytree': 0.62481

[I] 2025-05-31 02:46:20,048 Trial 12 finished with value: 0.003834272330752203 and parameters: {'n\_estimators': 227,

[I] 2025-05-31 02:46:25, 970 Trial 1 finished with value: 0.005897263984887575 and parameters: {'n\_estimators': 128, 'max\_depth': 6, 'learning\_rate': 0.0172825167921557, 'subsample': 0.846702265691898, 'colsample\_bytree': 0.911201}

[I] 2025-05-31 02:46:26, 0643 Trial 2 finished with value: 0.0059533168975982 and parameters: {'n\_estimators': 129, 'max\_depth': 4, 'learning\_rate': 0.26648311808899, 'subsample': 0.6538338210955121, 'colsample\_bytree': 0.885906}

[I] 2025-05-31 02:46:26, 1993 Trial 3 finished with value: 0.00595937756621244 and parameters: {'n\_estimators': 260, 'max\_depth': 5, 'learning\_rate': 0.1319594985858308, 'subsample': 0.9369080738844, 'colsample\_bytree': 0.955204}

[I] 2025-05-31 02:46:26, 3455 Trial 4 finished with value: 0.005926515447686413 and parameters: {'n\_estimators': 297, 'max\_depth': 4, 'learning\_rate': 0.08742365949007, 'subsample': 0.95547648125237, 'colsample\_bytree': 0.8084561818}

[I] 2025-05-31 02:46:26, 422 Trial 5 finished with value: 0.0057868874743131 and parameters: {'n\_estimators': 93, 'max\_depth': 3, 'learning\_rate': 0.0114921381219514, 'subsample': 0.933937495110122, 'colsample\_bytree': 0.6401576}

[I] 2025-05-31 02:46:26, 5717 Trial 6 finished with value: 0.005829507175407417 and parameters: {'n\_estimators': 167, 'max\_depth': 7, 'learning\_rate': 0.012861887818432132, 'subsample': 0.7559578870014776, 'colsample\_bytree': 0.82847}

[I] 2025-05-31 02:46:26, 5792 Trial 7 finished with value: 0.00581443701020988 and parameters: {'n\_estimators': 50, 'max\_depth': 3, 'learning\_rate': 0.01025622670149128, 'subsample': 0.563136047732411, 'colsample\_bytree': 0.5693331}

[I] 2025-05-31 02:46:26, 6683 Trial 8 finished with value: 0.0058151650988637 and parameters: {'n\_estimators': 127, 'max\_depth': 4, 'learning\_rate': 0.011835631532263, 'subsample': 0.5937515568333402, 'colsample\_bytree': 0.87530}

[I] 2025-05-31 02:46:26, 780 Trial 9 finished with value: 0.0058518114970119037 and parameters: {'n\_estimators': 175, 'max\_depth': 5, 'learning\_rate': 0.01878795107690139, 'subsample': 0.64152083232498842, 'colsample\_bytree': 0.931652}

[I] 2025-05-31 02:46:26, 935 Trial 10 finished with value: 0.005798313647957603 and parameters: {'n\_estimators': 221, 'max\_depth': 8, 'learning\_rate': 0.0328903020498565, 'subsample': 0.8097049042875912, 'colsample\_bytree': 0.73087}

[I] 2025-05-31 02:46:27, 0773 Trial 11 finished with value: 0.005729877509645055 and parameters: {'n\_estimators': 219, 'max\_depth': 8, 'learning\_rate': 0.03287862719044211, 'subsample': 0.7922863925653804, 'colsample\_bytree': 0.7164}

[I] 2025-05-31 02:46:27, 2183 Trial 12 finished with value: 0.005774997304268316 and parameters: {'n\_estimators': 232, 'max\_depth': 8, 'learning\_rate': 0.03365110897012388, 'subsample': 0.8383264509155993, 'colsample\_bytree': 0.6801}

[I] 2025-05-31 02:46:27, 3593 Trial 13 finished with value: 0.005706166006864833 and parameters: {'n\_estimators': 222, 'max\_depth': 8, 'learning\_rate': 0.039482854161682, 'subsample': 0.8258553124273389, 'colsample\_bytree': 0.72625}

[I] 2025-05-31 02:46:27, 5077 Trial 14 finished with value: 0.00580504063024883 and parameters: {'n\_estimators': 251, 'max\_depth': 7, 'learning\_rate': 0.051957223739749889, 'subsample': 0.8909010226869282, 'colsample\_bytree': 0.754960}

[I] 2025-05-31 02:46:27, 6393 Trial 15 finished with value: 0.0057832494138502 and parameters: {'n\_estimators': 204, 'max\_depth': 8, 'learning\_rate': 0.0220679804320567, 'subsample': 0.9989549288763544, 'colsample\_bytree': 0.59796}

[I] 2025-05-31 02:46:27, 8033 Trial 16 finished with value: 0.00590117217535353 and parameters: {'n\_estimators': 291, 'max\_depth': 6, 'learning\_rate': 0.04619619026849042, 'subsample': 0.84847771694945, 'colsample\_bytree': 0.76331}

[I] 2025-05-31 02:46:27, 9743 Trial 17 finished with value: 0.0058266108523126 and parameters: {'n\_estimators': 252, 'max\_depth': 7, 'learning\_rate': 0.1364236036572943, 'subsample': 0.779505431702424, 'colsample\_bytree': 0.66265}

[I] 2025-05-31 02:46:28, 103 Trial 18 finished with value: 0.005777363478364507 and parameters: {'n\_estimators': 202, 'max\_depth': 8, 'learning\_rate': 0.020336966044963, 'subsample': 0.8858270485490503, 'colsample\_bytree': 0.532835}

[I] 2025-05-31 02:46:28, 222 Trial 19 finished with value: 0.005795401545183949 and parameters: {'n\_estimators': 157, 'max\_depth': 6, 'learning\_rate': 0.0638789163246511, 'subsample': 0.70599231226207, 'colsample\_bytree': 0.81280}

[I] 2025-05-31 02:46:28, 351 A new study created in memory with name: no-name-c47a80de-ff6a-9d04-ba84c0ace6c

[I] 2025-05-31 02:46:28, 472 Trial 0 finished with value: 0.00557036709280347 and parameters: {'n\_estimators': 117, 'max\_depth': 7, 'learning\_rate': 0.0577454556478754, 'subsample': 0.70380242637342, 'colsample\_bytree': 0.975909}

[I] 2025-05-31 02:46:28, 547 Trial 1 finished with value: 0.005600503499332004 and parameters: {'n\_estimators': 85, 'max\_depth': 7, 'learning\_rate': 0.0117570075992320, 'subsample': 0.641033300776153, 'colsample\_bytree': 0.543732}

[I] 2025-05-31 02:46:28, 624 Trial 2 finished with value: 0.00565402870699677 and parameters: {'n\_estimators': 106, 'max\_depth': 5, 'learning\_rate': 0.0381286501575384, 'subsample': 0.567462190062821, 'colsample\_bytree': 0.695425}

[I] 2025-05-31 02:46:28, 759 Trial 3 finished with value: 0.005610207740500746 and parameters: {'n\_estimators': 241, 'max\_depth': 7, 'learning\_rate': 0.01606874124799383, 'subsample': 0.5260290400745032, 'colsample\_bytree': 0.613496}

[I] 2025-05-31 02:46:28, 904 Trial 4 finished with value: 0.0056551104431153 and parameters: {'n\_estimators': 234, 'max\_depth': 5, 'learning\_rate': 0.086859785631299, 'subsample': 0.891108090963951, 'colsample\_bytree': 0.7498637}

[I] 2025-05-31 02:46:29, 018 Trial 5 finished with value: 0.0056911389584921 and parameters: {'n\_estimators': 211, 'max\_depth': 6, 'learning\_rate': 0.09672105384460264, 'subsample': 0.6997298575204826, 'colsample\_bytree': 0.671419}

[I] 2025-05-31 02:46:29, 177 Trial 6 finished with value: 0.005644592614597607 and parameters: {'n\_estimators': 191, 'max\_depth': 8, 'learning\_rate': 0.02776732763733628, 'subsample': 0.973096257197366, 'colsample\_bytree': 0.794745}

[I] 2025-05-31 02:46:29, 341 Trial 7 finished with value: 0.0056503736785632 and parameters: {'n\_estimators': 263, 'max\_depth': 6, 'learning\_rate': 0.0596532868051035, 'subsample': 0.5539201977047941, 'colsample\_bytree': 0.676275}

[I] 2025-05-31 02:46:29, 451 Trial 8 finished with value: 0.00569733738795441 and parameters: {'n\_estimators': 147, 'max\_depth': 8, 'learning\_rate': 0.0200637613761195, 'subsample': 0.646706883808254, 'colsample\_bytree': 0.713267}

[I] 2025-05-31 02:46:29, 7283 Trial 9 finished with value: 0.00565715648149849 and parameters: {'n\_estimators': 276, 'max\_depth': 7, 'learning\_rate': 0.20939684385981553, 'subsample': 0.5191455050576777, 'colsample\_bytree': 0.5030983}

[I] 2025-05-31 02:46:30, 290 Trial 10 finished with value: 0.005633120886062771 and parameters: {'n\_estimators': 65, 'max\_depth': 3, 'learning\_rate': 0.02684844460592956, 'subsample': 0.834076830019885, 'colsample\_bytree': 0.8675205}

[I] 2025-05-31 02:46:30, 482 Trial 11 finished with value: 0.00560617037053244 and parameters: {'n\_estimators': 149, 'max\_depth': 4, 'learning\_rate': 0.02930866424216892, 'subsample': 0.5659956672341401, 'colsample\_bytree': 0.628}

[I] 2025-05-31 02:46:31, 532 Trial 12 finished with value: 0.00559501552420278 and parameters: {'n\_estimators': 291, 'max\_depth': 5, 'learning\_rate': 0.03087475112045164, 'subsample': 0.7644352091482299, 'colsample\_bytree': 0.81753}

[I] 2025-05-31 02:46:31, 7043 Trial 13 finished with value: 0.005456988688756586 and parameters: {'n\_estimators': 108, 'max\_depth': 6, 'learning\_rate': 0.122356506405308317, 'subsample': 0.5854727651146416, 'colsample\_bytree': 0.89483}

[I] 2025-05-31 02:46:31, 8763 Trial 14 finished with value: 0.00589982493337117 and parameters: {'n\_estimators': 106, 'max\_depth': 4, 'learning\_rate': 0.13145614164961117, 'subsample': 0.6150900312336617, 'colsample\_bytree': 0.92119}

[I] 2025-05-31 02:46:31, 9953 Trial 15 finished with value: 0.005601446440081378 and parameters: {'n\_estimators': 54, 'max\_depth': 5, 'learning\_rate': 0.127652869441384, 'subsample': 0.592523077822857, 'colsample\_bytree': 0.8984474}

[I] 2025-05-31 02:46:32, 1433 Trial 16 finished with value: 0.00564693886250257 and parameters: {'n\_estimators': 150, 'max\_depth': 4, 'learning\_rate': 0.27405103606020776, 'subsample': 0.7442236235563042, 'colsample\_bytree': 0.9775}

[I] 2025-05-31 02:46:32, 3277 Trial 17 finished with value: 0.005593945205105347 and parameters: {'n\_estimators': 120, 'max\_depth': 6, 'learning\_rate': 0.0326727172101903, 'subsample': 0.7156588074260594, 'colsample\_bytree': 0.77229}

[I] 2025-05-31 02:46:32, 5523 Trial 18 finished with value: 0.005565820264046176 and parameters: {'n\_estimators': 98, 'max\_depth': 3, 'learning\_rate': 0.016715217802703686, 'subsample': 0.5160785107502708, 'colsample\_bytree': 0.88271}

[I] 2025-05-31 02:46:32, 6823 Trial 19 finished with value: 0.005611812053048276 and parameters: {'n\_estimators': 176, 'max\_depth': 5, 'learning\_rate': 0.08352713207949163, 'subsample': 0.7981130806861192, 'colsample\_bytree': 0.83338}

[I] 2025-05-31 02:46:32, 7743 A new study created in memory with name: no-name-fa185f7c-9600-40b8-a777-d32fc6249

[I] 2025-05-31 02:46:32, 960 Trial 0 finished with value: 0.00220797797948469 and parameters: {'n\_estimators': 276, 'max\_depth': 5, 'learning\_rate': 0.0328172362648994, 'subsample': 0.617205210673992, 'colsample\_bytree': 0.927808}

[I] 2025-05-31 02:46:33, 101 Trial 1 finished with value: 0.00222844861871071 and parameters: {'n\_estimators': 177, 'max\_depth': 5, 'learning\_rate': 0.020567337895805887, 'subsample': 0.6813816062454532, 'colsample\_bytree': 0.8443618}

[I] 2025-05-31 02:46:33, 1717 Trial 2 finished with value: 0.002235074847623434 and parameters: {'n\_estimators': 77, 'max\_depth': 3, 'learning\_rate': 0.01675428179164223, 'subsample': 0.714531731065383, 'colsample\_bytree': 0.8403616}

[I] 2025-05-31 02:46:33, 3111 Trial 3 finished with value: 0.002227357521964076 and parameters: {'n\_estimators': 272, 'max\_depth': 6, 'learning\_rate': 0.0150885635369723, 'subsample': 0.6281780906786553, 'colsample\_bytree': 0.9787}

[I] 2025-05-31 02:46:33, 4093 Trial 4 finished with value: 0.002627123812347195 and parameters: {'n\_estimators': 151, 'max\_depth': 7, 'learning\_rate': 0.05172556419071914, 'subsample': 0.7087969122460106, 'colsample\_bytree': 0.55468}

[I] 2025-05-31 02:46:33, 543 Trial 5 finished with value: 0.0022389597126369439 and parameters: {'n\_estimators': 286, 'max\_depth': 6, 'learning\_rate': 0.02724284316128483, 'subsample': 0.6033318324870762, 'colsample\_bytree': 0.84257}

[I] 2025-05-31 02:46:33, 6583 Trial 6 finished with value: 0.002191097148646601 and parameters: {'n\_estimators': 205, 'max\_depth': 6, 'learning\_rate': 0.0266951034039642053, 'subsample': 0.722474587344214, 'colsample\_bytree': 0.51554}

[I] 2025-05-31 02:46:33, 783 Trial 7 finished with value: 0.00219536810317167 and parameters: {'n\_estimators': 237, 'max\_depth': 6, 'learning\_rate': 0.01174236874217528, 'subsample': 0.6648622149503132, 'colsample\_bytree': 0.78952}

[I] 2025-05-31 02:46:33, 8923 Trial 8 finished with value: 0.002186147733194367 and parameters: {'n\_estimators': 201, 'max\_depth': 8, 'learning\_rate': 0.12528151606722755, 'subsample': 0.880795368986504, 'colsample\_bytree': 0.66578}

[I] 2025-05-31 02:46:33, 9863 Trial 9 finished with value: 0.002141865720118506 and parameters: {'n\_estimators': 148, 'max\_depth': 7, 'learning\_rate': 0.2326805603072693, 'subsample': 0.7891123361174045, 'colsample\_bytree': 0.61810}

[I] 2025-05-31 02:46:34, 323 Trial 10 finished with value: 0.002164938625052257 and parameters: {'n\_estimators': 150, 'max\_depth': 5, 'learning\_rate': 0.27405103606020776, 'subsample': 0.7442236235563042, 'colsample\_bytree': 0.97755}

[I] 2025-05-31 02:46:34, 4303 Trial 11 finished with value: 0.00221935833044227 and parameters: {'n\_estimators': 110, 'max\_depth': 8, 'learning\_rate': 0.234334942944355, 'subsample': 0.9932974874297297, 'colsample\_bytree': 0.64742}

[I] 2025-05-31 02:46:34, 210 Trial 12 finished with value: 0.0022685031475735493 and parameters: {'n\_estimators': 132, 'max\_depth': 8, 'learning\_rate': 0.11328633715493375, 'subsample': 0.867414242529918, 'colsample\_bytree': 0.6770}

[I] 2025-05-31 02:46:34, 339 Trial 13 finished with value: 0.002225394723804933 and parameters: {'n\_estimators': 194, 'max\_depth': 8, 'learning\_rate': 0.0881714471813923, 'subsample': 0.8405899630847596, 'colsample\_bytree': 0.62925}

[I] 2025-05-31 02:46:34, 474 Trial 14 finished with value: 0.002165407202716760 and parameters: {'n\_estimators': 223, 'max\_depth': 7, 'learning\_rate': 0.1151453553575946, 'subsample': 0.854533911493609, 'colsample\_bytree': 0.71139}

[I] 2025-05-31 02:46:34, 5461 Trial 14 finished with value: 0.0024932668577932927 and parameters: {'n\_estimators': 54, 'max\_depth': 7, 'learning\_rate': 0.04167528179164223, 'subsample': 0.51109509894356, 'colsample\_bytree': 0.729724}

[I] 2025-05-31 02:46:34, 6705 Trial 15 finished with value: 0.002038717357235252 and parameters: {'n\_estimators': 233, 'max\_depth': 7, 'learning\_rate': 0.29677277889199416, 'subsample': 0.8081522274782322, 'colsample\_bytree': 0.5707}

[I] 2025-05-31 02:46:34, 8583 Trial 16 finished with value: 0.00216178205735713 and parameters: {'n\_estimators': 243, 'max\_depth': 4, 'learning\_rate': 0.26386038879700074, 'subsample': 0.7947566694133564, 'colsample\_bytree': 0.58018}

[I] 2025-05-31 02:46:34, 9783 Trial 17 finished with value: 0.00218957910495064096 and parameters: {'n\_estimators': 146, 'max\_depth': 7, 'learning\_rate': 0.234334942944355, 'subsample': 0.988008970928931, 'colsample\_bytree': 0.5780}

[I] 2025-05-31 02:46:35, 0831 Trial 18 finished with value: 0.002305271856996105 and parameters: {'n\_estimators': 108, 'max\_depth': 7, 'learning\_rate': 0.16994502560444376, 'subsample': 0.78667091122053, 'colsample\_bytree': 0.60342}

[I] 2025-05-31 02:46:35, 210 Trial 19 finished with value: 0.0023664226206286321 and parameters: {'n\_estimators': 174, 'max\_depth': 5, 'learning\_rate': 0.2449423425224697, 'subsample': 0.9304882008249069, 'colsample\_bytree': 0.53909}

[I] 2025-05-31 02:46:35, 343 A new study created in memory with name: no-name-841625d9-7ef9-4941-9bee-c4b8e36a078b

[I] 2025-05-31 02:46:35, 4570 Trial 0 finished with value: 0.0026633895356486934 and parameters: {'n\_estimators': 128, 'max\_depth': 3, 'learning\_rate': 0.1099138633396213, 'subsample': 0.9406079165720322, 'colsample\_bytree': 0.99880}

[I] 2025-05-31 02:46:35, 5893 Trial 1 finished with value: 0.0022669542674421665 and parameters: {'n\_estimators': 270, 'max\_depth': 7, 'learning\_rate': 0.014145051525093428, 'subsample': 0.8810061065792016, 'colsample\_bytree': 0.6180}

[I] 2025-05-31 02:46:36, 0853 Trial 2 finished with value: 0.002670205520252527 and parameters: {'n\_estimators': 151, 'max\_depth': 6, 'learning\_rate': 0.042442645193991184, 'subsample': 0.784899834357290291, 'colsample\_bytree': 0.86863}

[I] 2025-05-31 02:46:36, 3155 Trial 3 finished with value: 0.00269176041173643 and parameters: {'n\_estimators': 201, 'max\_depth': 7, 'learning\_rate': 0.017253651074901921, 'subsample': 0.6073020903950926, 'colsample\_bytree': 0.971432135}

[I] 2025-05-31 02:46:36, 4515 Trial 4 finished with value: 0.002650488488537578 and parameters: {'n\_estimators': 195, 'max\_depth': 5, 'learning\_rate': 0.04763020390350926, 'subsample': 0.8060676472577045, 'colsample\_bytree': 0.527222}

[I] 2025-05-31 02:46:36, 0093 Trial 5 finished with value: 0.00266156395356205623 and parameters: {'n\_estimators': 100, 'max\_depth': 7, 'learning\_rate': 0.020687147625860323, 'subsample': 0.9198257827844799, 'colsample\_bytree': 0.87627}

[I] 2025-05-31 02:46:36, 120 Trial 6 finished with value: 0.002661232666870876 and parameters: {'n\_estimators': 199, 'max\_depth': 7, 'learning\_rate': 0.02799532175754992, 'subsample': 0.640078927242023, 'colsample\_bytree': 0.771033}

[I] 2025-05-31 02:46:36, 2533 Trial 7 finished with value: 0.002664256170122822 and parameters: {'n\_estimators': 194, 'max\_depth': 7, 'learning\_rate': 0.03617423096993502, 'subsample': 0.96774369228740994, 'colsample\_bytree': 0.521864}

[I] 2025-05-31 02:46:36, 3755 Trial 8 finished with value: 0.002664604165045434 and parameters: {'n\_estimators': 234, 'max\_depth': 3, 'learning\_rate': 0.04183275567656399, 'subsample': 0.909345817623912, 'colsample\_bytree': 0.61516}

[I] 2025-05-31 02:46:36, 4783 Trial 9 finished with value: 0.00262702728580677 and parameters: {'n\_estimators': 163, 'max\_depth': 3, 'learning\_rate': 0.0127560118287804, 'subsample': 0.9288717320736902, 'colsample\_bytree': 0.917026}

[I] 2025-05-31 02:46:36, 5623 Trial 10 finished with value: 0.00265016889871183454 and parameters: {'n\_estimators': 56, 'max\_depth': 5, 'learning\_rate': 0.09280172740717982, 'subsample': 0.743638750082381, 'colsample\_bytree': 0.531517}

[I] 2025-05-31 02:46:36, 651 Trial 11 finished with value: 0.00263474977496102 and parameters: {'n\_estimators': 61, 'max\_depth': 5, 'learning\_rate': 0.08685745811204376, 'subsample': 0.7457262501664, 'colsample\_bytree': 0.51470397}

[I] 2025-05-31 02:46:36, 7383 Trial 12 finished with value: 0.002696115626123 and parameters: {'n\_estimators': 53, 'max\_depth': 5, 'learning\_rate': 0.0287117043675917, 'subsample': 0.728817131741761, 'colsample\_bytree': 0.6410858}

[I] 2025-05-31 02:46:36, 9013 Trial 13 finished with value: 0.00268630151842153 and parameters: {'n\_estimators': 288, 'max\_depth': 8, 'learning\_rate': 0.168488158166907, 'subsample': 0.697273652494204, 'colsample\_bytree': 0.70790}

[I] 2025-05-31 02:46:37, 0483 Trial 14 finished with value: 0.00264988282382103 and parameters: {'n\_estimators': 237, 'max\_depth': 4, 'learning\_rate': 0.06156971782793213, 'subsample': 0.738734522788704, 'colsample\_bytree': 0.55115}

[I] 2025-05-31 02:46:37, 2299 Trial 15 finished with value: 0.002663372712904687 and parameters: {'n\_estimators': 244, 'max\_depth': 4, 'learning\_rate': 0.07250611828784804, 'subsample': 0.5070079715073127, 'colsample\_bytree': 0.7081}

[I] 2025-05-31 02:46:37, 3303 Trial

[1] 2025-05-31 02:46:45, 740] Trial 9 finished with value: 0.026956277261112275 and parameters: {'n\_estimators': 77, 'max\_depth': 7, 'learning\_rate': 0.025740384873470888, 'subsample': 0.6993571848976241, 'colsample\_bytree': 0.65695}

[1] 2025-05-31 02:46:45, 897] Trial 10 finished with value: 0.0217177985413016506 and parameters: {'n\_estimators': 286, 'max\_depth': 3, 'learning\_rate': 0.272545514516618, 'subsample': 0.5099775763372492, 'colsample\_bytree': 0.50227}

[1] 2025-05-31 02:46:45, 983] Trial 11 finished with value: 0.0281314097339766 and parameters: {'n\_estimators': 52, 'max\_depth': 5, 'learning\_rate': 0.0923004257403639, 'subsample': 0.673494501155121, 'colsample\_bytree': 0.761422}

[1] 2025-05-31 02:46:46, 111] Trial 12 finished with value: 0.02073167363405132 and parameters: {'n\_estimators': 169, 'max\_depth': 5, 'learning\_rate': 0.0923004257403639, 'subsample': 0.673494501155121, 'colsample\_bytree': 0.761373}

[1] 2025-05-31 02:46:46, 221] Trial 13 finished with value: 0.0263810472696155 and parameters: {'n\_estimators': 102, 'max\_depth': 6, 'learning\_rate': 0.05668232591599352, 'subsample': 0.587982066643188, 'colsample\_bytree': 0.86377}

[1] 2025-05-31 02:46:46, 320] Trial 14 finished with value: 0.027593970355028185 and parameters: {'n\_estimators': 89, 'max\_depth': 6, 'learning\_rate': 0.05111204104249383, 'subsample': 0.5593561007273832, 'colsample\_bytree': 0.92507}

[1] 2025-05-31 02:46:46, 437] Trial 15 finished with value: 0.02679736551145036 and parameters: {'n\_estimators': 144, 'max\_depth': 5, 'learning\_rate': 0.0494215719942009, 'subsample': 0.6180330563496463, 'colsample\_bytree': 0.8674}

[1] 2025-05-31 02:46:46, 603] Trial 16 finished with value: 0.0298567931137446 and parameters: {'n\_estimators': 196, 'max\_depth': 6, 'learning\_rate': 0.1730968958932125, 'subsample': 0.5751240218748854, 'colsample\_bytree': 0.82043}

[1] 2025-05-31 02:46:46, 755] Trial 17 finished with value: 0.020747162386370327 and parameters: {'n\_estimators': 238, 'max\_depth': 4, 'learning\_rate': 0.0369613485383445, 'subsample': 0.8262651001134393, 'colsample\_bytree': 0.7078}

[1] 2025-05-31 02:46:46, 874] Trial 18 finished with value: 0.02724730119018078 and parameters: {'n\_estimators': 137, 'max\_depth': 4, 'learning\_rate': 0.1379760808970968, 'subsample': 0.505917262605974, 'colsample\_bytree': 0.9912}

[1] 2025-05-31 02:46:46, 964] Trial 19 finished with value: 0.02679159253624246 and parameters: {'n\_estimators': 50, 'max\_depth': 6, 'learning\_rate': 0.0634340657846999, 'subsample': 0.660625395909268, 'colsample\_bytree': 0.90559}

[1] 2025-05-31 02:46:47, 053] A new study created in memory with name: no-name-0a8c8a33-6599-4d4a-a3le-095d07045454

[1] 2025-05-31 02:46:47, 164] Trial 0 finished with value: 0.02545427013910437 and parameters: {'n\_estimators': 63, 'max\_depth': 4, 'learning\_rate': 0.17208232433925488, 'subsample': 0.750952063612208, 'colsample\_bytree': 0.9548259}

[1] 2025-05-31 02:46:47, 260] Trial 1 finished with value: 0.026387889343202366 and parameters: {'n\_estimators': 153, 'max\_depth': 3, 'learning\_rate': 0.030241028086968762, 'subsample': 0.5208589515901943, 'colsample\_bytree': 0.9390}

[1] 2025-05-31 02:46:47, 378] Trial 2 finished with value: 0.0253729753909179 and parameters: {'n\_estimators': 219, 'max\_depth': 3, 'learning\_rate': 0.25283579826212, 'subsample': 0.730781900117163, 'colsample\_bytree': 0.98341854}

[1] 2025-05-31 02:46:47, 476] Trial 3 finished with value: 0.02601467131401303 and parameters: {'n\_estimators': 169, 'max\_depth': 4, 'learning\_rate': 0.02518968474058007, 'subsample': 0.919498264010337, 'colsample\_bytree': 0.5143}

[1] 2025-05-31 02:46:47, 651] Trial 4 finished with value: 0.02632927956800509 and parameters: {'n\_estimators': 285, 'max\_depth': 6, 'learning\_rate': 0.1458597748492996, 'subsample': 0.656252419158604, 'colsample\_bytree': 0.853019}

[1] 2025-05-31 02:46:47, 736] Trial 5 finished with value: 0.02064041069908748 and parameters: {'n\_estimators': 119, 'max\_depth': 7, 'learning\_rate': 0.10131262021780246, 'subsample': 0.928010796454949, 'colsample\_bytree': 0.5813796}

[1] 2025-05-31 02:46:47, 871] Trial 6 finished with value: 0.02592697081937 and parameters: {'n\_estimators': 276, 'max\_depth': 3, 'learning\_rate': 0.50927805511627, 'subsample': 0.50927805511627, 'colsample\_bytree': 0.50806}

[1] 2025-05-31 02:46:48, 095] Trial 7 finished with value: 0.02575345971919263 and parameters: {'n\_estimators': 258, 'max\_depth': 5, 'learning\_rate': 0.01121840306907159, 'subsample': 0.5766392597630896, 'colsample\_bytree': 0.59749}

[1] 2025-05-31 02:46:48, 117] Trial 8 finished with value: 0.0261269815019564 and parameters: {'n\_estimators': 199, 'max\_depth': 4, 'learning\_rate': 0.0160680101161381, 'subsample': 0.6783690555122786, 'colsample\_bytree': 0.756506}

[1] 2025-05-31 02:46:48, 194] Trial 9 finished with value: 0.0260502765590663 and parameters: {'n\_estimators': 94, 'max\_depth': 7, 'learning\_rate': 0.077328807073846, 'subsample': 0.859658131469937, 'colsample\_bytree': 0.8253166}

[1] 2025-05-31 02:46:48, 335] Trial 10 finished with value: 0.02552970681651381 and parameters: {'n\_estimators': 229, 'max\_depth': 8, 'learning\_rate': 0.2194890572619525, 'subsample': 0.7688227833808096, 'colsample\_bytree': 0.66617}

[1] 2025-05-31 02:46:48, 422] Trial 11 finished with value: 0.02628761259887471 and parameters: {'n\_estimators': 55, 'max\_depth': 4, 'learning\_rate': 0.293089015172496, 'subsample': 0.772373429659479, 'colsample\_bytree': 0.993854}

[1] 2025-05-31 02:46:48, 511] Trial 12 finished with value: 0.02663717474005888 and parameters: {'n\_estimators': 214, 'max\_depth': 3, 'learning\_rate': 0.1500937340337825, 'subsample': 0.683979104326133, 'colsample\_bytree': 0.91657}

[1] 2025-05-31 02:46:48, 691] Trial 13 finished with value: 0.02592122883632639 and parameters: {'n\_estimators': 130, 'max\_depth': 5, 'learning\_rate': 0.0509079580563227, 'subsample': 0.833806346976817, 'colsample\_bytree': 0.99645}

[1] 2025-05-31 02:46:48, 780] Trial 14 finished with value: 0.0259125855661684 and parameters: {'n\_estimators': 56, 'max\_depth': 4, 'learning\_rate': 0.202824536336166, 'subsample': 0.701650090160602, 'colsample\_bytree': 0.881818}

[1] 2025-05-31 02:46:48, 916] Trial 15 finished with value: 0.02511883012061965 and parameters: {'n\_estimators': 195, 'max\_depth': 3, 'learning\_rate': 0.290267353803041, 'subsample': 0.838899538239209, 'colsample\_bytree': 0.78375}

[1] 2025-05-31 02:46:49, 076] Trial 16 finished with value: 0.0252807223330495 and parameters: {'n\_estimators': 245, 'max\_depth': 5, 'learning\_rate': 0.115500581313961, 'subsample': 0.64407421880085, 'colsample\_bytree': 0.93614}

[1] 2025-05-31 02:46:49, 221] Trial 17 finished with value: 0.0255072684472125 and parameters: {'n\_estimators': 243, 'max\_depth': 6, 'learning\_rate': 0.0797364490762096, 'subsample': 0.615000581313961, 'colsample\_bytree': 0.7019}

[1] 2025-05-31 02:46:49, 372] Trial 18 finished with value: 0.025021717904002176 and parameters: {'n\_estimators': 256, 'max\_depth': 5, 'learning\_rate': 0.1097786515433752, 'subsample': 0.633193959179028, 'colsample\_bytree': 0.89512}

[1] 2025-05-31 02:46:49, 536] Trial 19 finished with value: 0.025917726843399 and parameters: {'n\_estimators': 295, 'max\_depth': 5, 'learning\_rate': 0.0486799129892808, 'subsample': 0.6309091797026194, 'colsample\_bytree': 0.81170}

[1] 2025-05-31 02:46:49, 707] A new study created in memory with name: no-name-4ffe55-5f5-5f5-4b7-88b-6d0295f73395

[1] 2025-05-31 02:46:49, 873] Trial 0 finished with value: 0.00682943846804697 and parameters: {'n\_estimators': 293, 'max\_depth': 7, 'learning\_rate': 0.2212982278588851, 'subsample': 0.6519128866752495, 'colsample\_bytree': 0.710883}

[1] 2025-05-31 02:46:49, 953] Trial 1 finished with value: 0.00793856385606793 and parameters: {'n\_estimators': 93, 'max\_depth': 4, 'learning\_rate': 0.2644075318478702, 'subsample': 0.5429852761785772, 'colsample\_bytree': 0.61591328}

[1] 2025-05-31 02:46:50, 049] Trial 2 finished with value: 0.006559620129463377 and parameters: {'n\_estimators': 155, 'max\_depth': 7, 'learning\_rate': 0.04007979547975158, 'subsample': 0.507801206056245, 'colsample\_bytree': 0.590344}

[1] 2025-05-31 02:46:50, 147] Trial 3 finished with value: 0.00683589871007986 and parameters: {'n\_estimators': 160, 'max\_depth': 5, 'learning\_rate': 0.019767434611974, 'subsample': 0.945719618577832, 'colsample\_bytree': 0.8147125}

[1] 2025-05-31 02:46:50, 240] Trial 4 finished with value: 0.006645347570174051 and parameters: {'n\_estimators': 151, 'max\_depth': 7, 'learning\_rate': 0.024862811255611, 'subsample': 0.5449326273764185, 'colsample\_bytree': 0.673540}

[1] 2025-05-31 02:46:50, 340] Trial 5 finished with value: 0.0067640595107860769 and parameters: {'n\_estimators': 158, 'max\_depth': 6, 'learning\_rate': 0.17519026875071701, 'subsample': 0.77035105945769, 'colsample\_bytree': 0.6711862}

[1] 2025-05-31 02:46:50, 453] Trial 6 finished with value: 0.006854228183783073 and parameters: {'n\_estimators': 177, 'max\_depth': 6, 'learning\_rate': 0.0159684319356464, 'subsample': 0.797006509393893, 'colsample\_bytree': 0.738223}

[1] 2025-05-31 02:46:50, 551] Trial 7 finished with value: 0.00685360527552436 and parameters: {'n\_estimators': 160, 'max\_depth': 6, 'learning\_rate': 0.0579176121562405, 'subsample': 0.8678561867707501, 'colsample\_bytree': 0.706358}

[1] 2025-05-31 02:46:50, 651] Trial 8 finished with value: 0.0067936264983149 and parameters: {'n\_estimators': 147, 'max\_depth': 4, 'learning\_rate': 0.0291548018389013, 'subsample': 0.5472517659238055, 'colsample\_bytree': 0.99011300}

[1] 2025-05-31 02:46:50, 751] Trial 9 finished with value: 0.006852707456512338 and parameters: {'n\_estimators': 245, 'max\_depth': 5, 'learning\_rate': 0.11550040176810469, 'subsample': 0.615000581313961, 'colsample\_bytree': 0.593614}

[1] 2025-05-31 02:46:51, 051] Trial 10 finished with value: 0.0067744954201862661 and parameters: {'n\_estimators': 243, 'max\_depth': 8, 'learning\_rate': 0.0977364490762096, 'subsample': 0.631393289511974, 'colsample\_bytree': 0.89512}

[1] 2025-05-31 02:46:51, 197] Trial 12 finished with value: 0.0067461217969323 and parameters: {'n\_estimators': 211, 'max\_depth': 7, 'learning\_rate': 0.287076951738164, 'subsample': 0.641691709214227, 'colsample\_bytree': 0.8167}

[1] 2025-05-31 02:46:51, 305] Trial 13 finished with value: 0.006724184435764624 and parameters: {'n\_estimators': 120, 'max\_depth': 7, 'learning\_rate': 0.0510966036391457, 'subsample': 0.59431039163942, 'colsample\_bytree': 0.5991142}

[1] 2025-05-31 02:46:51, 389] Trial 14 finished with value: 0.006784691844159635 and parameters: {'n\_estimators': 55, 'max\_depth': 8, 'learning\_rate': 0.01992452570252697, 'subsample': 0.7818817087317207, 'colsample\_bytree': 0.810361}

[1] 2025-05-31 02:46:51, 551] Trial 15 finished with value: 0.00685360527552436 and parameters: {'n\_estimators': 160, 'max\_depth': 6, 'learning\_rate': 0.0579176121562405, 'subsample': 0.8678561867707501, 'colsample\_bytree': 0.706358}

[1] 2025-05-31 02:46:51, 651] Trial 16 finished with value: 0.0067936264983149 and parameters: {'n\_estimators': 147, 'max\_depth': 4, 'learning\_rate': 0.0291548018389013, 'subsample': 0.5472517659238055, 'colsample\_bytree': 0.99011300}

[1] 2025-05-31 02:46:51, 811] Trial 17 finished with value: 0.006852707456512338 and parameters: {'n\_estimators': 291, 'max\_depth': 3, 'learning\_rate': 0.1145411208695415, 'subsample': 0.903398289521194, 'colsample\_bytree': 0.687336}

[1] 2025-05-31 02:46:51, 921] Trial 18 finished with value: 0.00675072684472125 and parameters: {'n\_estimators': 230, 'max\_depth': 8, 'learning\_rate': 0.0797364490762096, 'subsample': 0.615000581313961, 'colsample\_bytree': 0.593614}

[1] 2025-05-31 02:46:51, 981] Trial 19 finished with value: 0.0067749581056935 and parameters: {'n\_estimators': 85, 'max\_depth': 8, 'learning\_rate': 0.031838105955944, 'subsample': 0.508066550800980, 'colsample\_bytree': 0.559071}

[1] 2025-05-31 02:46:51, 997] Trial 20 finished with value: 0.0067461217969323 and parameters: {'n\_estimators': 211, 'max\_depth': 7, 'learning\_rate': 0.0287076951738164, 'subsample': 0.641691709214227, 'colsample\_bytree': 0.8167}

[1] 2025-05-31 02:46:52, 035] Trial 21 finished with value: 0.00674513053717993 and parameters: {'n\_estimators': 120, 'max\_depth': 7, 'learning\_rate': 0.01992452570252697, 'subsample': 0.7818817087317207, 'colsample\_bytree': 0.810361}

[1] 2025-05-31 02:46:52, 199] Trial 22 finished with value: 0.006784691844159635 and parameters: {'n\_estimators': 55, 'max\_depth': 8, 'learning\_rate': 0.01992452570252697, 'subsample': 0.7818817087317207, 'colsample\_bytree': 0.810361}

[1] 2025-05-31 02:46:52, 351] Trial 23 finished with value: 0.00671106269512263 and parameters: {'n\_estimators': 162, 'max\_depth': 5, 'learning\_rate': 0.01711206269512263, 'subsample': 0.5926437863033156, 'colsample\_bytree': 0.502278}

[1] 2025-05-31 02:46:52, 509] Trial 24 finished with value: 0.00677450751155109 and parameters: {'n\_estimators': 120, 'max\_depth': 5, 'learning\_rate': 0.018566350639560663, 'subsample': 0.71170759713352, 'colsample\_bytree': 0.615693}

[1] 2025-05-31 02:46:52, 661] Trial 25 finished with value: 0.0067745167692521 and parameters: {'n\_estimators': 120, 'max\_depth': 5, 'learning\_rate': 0.018566350639560663, 'subsample': 0.71170759713352, 'colsample\_bytree': 0.615693}

[1] 2025-05-31 02:46:52, 827] Trial 26 finished with value: 0.006784691844159635 and parameters: {'n\_estimators': 55, 'max\_depth': 8, 'learning\_rate': 0.01992452570252697, 'subsample': 0.7818817087317207, 'colsample\_bytree': 0.810361}

[1] 2025-05-31 02:46:53, 199] Trial 27 finished with value: 0.0067028522867235457 and parameters: {'n\_estimators': 98, 'max\_depth': 4, 'learning\_rate': 0.013652336639235756, 'subsample': 0.8387808769279277, 'colsample\_bytree': 0.7978170

[1] 2025-05-31 02:46:53, 379] Trial 28 finished with value: 0.006574307920549233 and parameters: {'n\_estimators': 53, 'max\_depth': 6, 'learning\_rate': 0.020815040122422474, 'subsample': 0.50252032938295836, 'colsample\_bytree': 0.947836}

[1] 2025-05-31 02:46:53, 514] Trial 29 finished with value: 0.00670583631809564 and parameters: {'n\_estimators': 234, 'max\_depth': 7, 'learning\_rate': 0.018854807482768422, 'subsample': 0.6158473196288274, 'colsample\_bytree': 0.99482}

[1] 2025-05-31 02:46:52, 126] Trial 30 finished with value: 0.0067745167692521 and parameters: {'n\_estimators': 183, 'max\_depth': 5, 'learning\_rate': 0.018566350639560663, 'subsample': 0.7072942632764271, 'colsample\_bytree': 0.92890}

[1] 2025-05-31 02:46:52, 227] A new study created in memory with name: no-name-788abdf3-106d-4337-b900-424dfbea7e8d1

[1] 2025-05-31 02:46:52, 397] Trial 0 finished with value: 0.00253453635175175 and parameters: {'n\_estimators': 289, 'max\_depth': 8, 'learning\_rate': 0.0748186506857193, 'subsample': 0.5678225508866352, 'colsample\_bytree': 0.857215}

[1] 2025-05-31 02:46:52, 552] Trial 1 finished with value: 0.002512411446683677 and parameters: {'n\_estimators': 251, 'max\_depth': 5, 'learning\_rate': 0.191579862822788, 'subsample': 0.9276328370385495, 'colsample\_bytree': 0.98425}

[1] 2025-05-31 02:46:52, 648] Trial 2 finished with value: 0.0025185034148445562 and parameters: {'n\_estimators': 207, 'max\_depth': 7, 'learning\_rate': 0.029126495137565509, 'subsample': 0.763398382743472, 'colsample\_bytree': 0.757875}

[1] 2025-05-31 02:46:52, 724] Trial 3 finished with value: 0.002534747959610799107 and parameters: {'n\_estimators': 89, 'max\_depth': 8, 'learning\_rate': 0.02188087233308949, 'subsample': 0.8387808769279277, 'colsample\_bytree': 0.7978170

[1] 2025-05-31 02:46:52, 887] Trial 4 finished with value: 0.0025082283052361256 and parameters: {'n\_estimators': 144, 'max\_depth': 8, 'learning\_rate': 0.020328814553710244, 'subsample': 0.5032387057305358, 'colsample\_bytree': 0.73787}

[1] 2025-05-31 02:46:53, 199] Trial 5 finished with value: 0.002519177277272725 and parameters: {'n\_estimators': 61, 'max\_depth': 7, 'learning\_rate': 0.01891184644828899, 'subsample': 0.30468663114435631, 'colsample\_bytree': 0.89734}

[1] 2025-05-31 02:46:53, 357] Trial 6 finished with value: 0.0025185163106159539 and parameters: {'n\_estimators': 192, 'max\_depth': 5, 'learning\_rate': 0.029562517664621761, 'subsample': 0.50708595293620768, 'colsample\_bytree': 0.50609949421671, 'colsample\_bytree': 0.5609949}

[1] 2025-05-31 02:46:53, 517] Trial 7 finished with value: 0.00665957605694898 and parameters: {'n\_estimators': 71, 'max\_depth': 8, 'learning\_rate': 0.29562517664621761, 'subsample': 0.51731789236595331, 'colsample\_bytree': 0.8553789}

[1] 2025-05-31 02:46:53, 651] Trial 8 finished with value: 0.006555890309273575 and parameters: {'n\_estimators': 253, 'max\_depth': 5, 'learning\_rate': 0.0273528303753677, 'subsample': 0.5317766349592982, 'colsample\_bytree': 0.9904595}

[1] 2025-05-31 02:46:53, 842] Trial 9 finished with value: 0.006549443848315002 and parameters: {'n\_estimators': 54, 'max\_depth': 5, 'learning\_rate': 0.03683595651625375, 'subsample': 0.5067214904022579, 'colsample\_bytree': 0.5466828}

[1] 2025-05-31 02:46:53, 991] Trial 10 finished with value: 0.006549443848315002 and parameters: {'n\_estimators': 55, 'max\_depth': 5, 'learning\_rate': 0.03683595651625375, 'subsample': 0.5067214904022579, 'colsample\_bytree': 0.5466828}

[1] 2025-05-31 02:46:54, 011] Trial 11 finished with value: 0.0065493083446443564 and parameters: {'n\_estimators': 131, 'max\_depth': 6, 'learning\_rate': 0.0328553826119938, 'subsample': 0.5216011188141204, 'colsample\_bytree': 0.68093}

[1] 2025-05-31 02:46:54, 185] Trial 12 finished with value: 0.002525291717432976 and parameters: {'n\_estimators': 131, 'max\_depth': 6, 'learning\_rate': 0.0328553826119938, 'subsample': 0.5216011188141204, 'colsample\_bytree': 0.68093}

[1] 2025-05-31 02:46:54, 294] Trial 13 finished with value: 0.00251763266418312 and parameters: {'n\_estimators': 135, 'max\_depth': 6, 'learning\_rate': 0.0328176366442833, 'subsample': 0.5216011188141204, 'colsample\_bytree': 0.68093}

[1] 2025-05-31 02:46:54, 383] Trial 14 finished with value: 0.00251717890149994 and parameters: {'n\_estimators': 135, 'max\_depth': 6, 'learning\_rate': 0.0328176366442833, 'subsample': 0.5216011188141204, 'colsample\_bytree': 0

[I] 2025-05-31 02:47:03.856 Trial 17 finished with value: 0.00648893400060000 and parameters: {'n\_estimators': 164, 'max\_depth': 5, 'learning\_rate': 0.0460822442931547, 'subsample': 0.8124560903837044, 'colsample\_bytree': 0.9064918}

[I] 2025-05-31 02:47:04.011 Trial 18 finished with value: 0.0064535257884039444 and parameters: {'n\_estimators': 93, 'max\_depth': 5, 'learning\_rate': 0.0406822442931547, 'subsample': 0.8124560903837044, 'colsample\_bytree': 0.6346918}

[I] 2025-05-31 02:47:04.011 Trial 19 finished with value: 0.006460497058933975 and parameters: {'n\_estimators': 295, 'max\_depth': 6, 'learning\_rate': 0.0924675120392957, 'subsample': 0.74538278340411, 'colsample\_bytree': 0.63518}

[I] 2025-05-31 02:47:04.341 A new study created in memory with name: no-name-3e51624b-d071-4779-aee2-36597620501a

[I] 2025-05-31 02:47:04.501 Trial 0 finished with value: 0.00149162516009832 and parameters: {'n\_estimators': 277, 'max\_depth': 4, 'learning\_rate': 0.175388457675468, 'subsample': 0.9660245890328911, 'colsample\_bytree': 0.613635}

[I] 2025-05-31 02:47:04.634 Trial 1 finished with value: 0.00146950916498929 and parameters: {'n\_estimators': 252, 'max\_depth': 4, 'learning\_rate': 0.1435115541643364, 'subsample': 0.6053279610022784, 'colsample\_bytree': 0.83726}

[I] 2025-05-31 02:47:04.740 Trial 2 finished with value: 0.0013797930581519114 and parameters: {'n\_estimators': 202, 'max\_depth': 6, 'learning\_rate': 0.28362293792703, 'subsample': 0.62616589771183, 'colsample\_bytree': 0.5037295}

[I] 2025-05-31 02:47:04.829 Trial 3 finished with value: 0.00150632454984895 and parameters: {'n\_estimators': 125, 'max\_depth': 6, 'learning\_rate': 0.0219523738751982, 'subsample': 0.95146972092726, 'colsample\_bytree': 0.6070}

[I] 2025-05-31 02:47:04.931 Trial 4 finished with value: 0.001470385280372023 and parameters: {'n\_estimators': 176, 'max\_depth': 3, 'learning\_rate': 0.047891626203401, 'subsample': 0.907337435623677, 'colsample\_bytree': 0.755229}

[I] 2025-05-31 02:47:04.997 Trial 5 finished with value: 0.00156462717594032 and parameters: {'n\_estimators': 50, 'max\_depth': 3, 'learning\_rate': 0.0252167158942893, 'subsample': 0.607724604258589, 'colsample\_bytree': 0.863438}

[I] 2025-05-31 02:47:05.132 Trial 6 finished with value: 0.001491503834934964 and parameters: {'n\_estimators': 265, 'max\_depth': 7, 'learning\_rate': 0.0649805996494529, 'subsample': 0.737977242014155, 'colsample\_bytree': 0.836616}

[I] 2025-05-31 02:47:05.270 Trial 7 finished with value: 0.001721735152626622 and parameters: {'n\_estimators': 171, 'max\_depth': 7, 'learning\_rate': 0.023066247027315, 'subsample': 0.6664779803081715, 'colsample\_bytree': 0.537771}

[I] 2025-05-31 02:47:05.365 Trial 8 finished with value: 0.0015035862778746566 and parameters: {'n\_estimators': 143, 'max\_depth': 5, 'learning\_rate': 0.0139946170315305, 'subsample': 0.561798269772051, 'colsample\_bytree': 0.7633}

[I] 2025-05-31 02:47:05.495 Trial 9 finished with value: 0.00147191010775931 and parameters: {'n\_estimators': 257, 'max\_depth': 5, 'learning\_rate': 0.0207583924691873, 'subsample': 0.508399020729357, 'colsample\_bytree': 0.80913}

[I] 2025-05-31 02:47:05.588 Trial 10 finished with value: 0.001447993734224136 and parameters: {'n\_estimators': 84, 'max\_depth': 8, 'learning\_rate': 0.1007882172677924, 'subsample': 0.778822359913846, 'colsample\_bytree': 0.97262}

[I] 2025-05-31 02:47:05.715 Trial 11 finished with value: 0.001434588454947474 and parameters: {'n\_estimators': 197, 'max\_depth': 7, 'learning\_rate': 0.2301130295091563, 'subsample': 0.6860330542094865, 'colsample\_bytree': 0.50893}

[I] 2025-05-31 02:47:05.858 Trial 12 finished with value: 0.00148465492712644 and parameters: {'n\_estimators': 216, 'max\_depth': 6, 'learning\_rate': 0.282315631235693, 'subsample': 0.68896242570152, 'colsample\_bytree': 0.5074488}

[I] 2025-05-31 02:47:05.970 Trial 13 finished with value: 0.001559119209770764 and parameters: {'n\_estimators': 141, 'max\_depth': 7, 'learning\_rate': 0.13500574674687534, 'subsample': 0.80929442498362583, 'colsample\_bytree': 0.6198}

[I] 2025-05-31 02:47:06.113 Trial 14 finished with value: 0.001508834397816785 and parameters: {'n\_estimators': 220, 'max\_depth': 6, 'learning\_rate': 0.0942878059158793, 'subsample': 0.66095309376474795, 'colsample\_bytree': 0.6730}

[I] 2025-05-31 02:47:06.248 Trial 15 finished with value: 0.00135997702301685 and parameters: {'n\_estimators': 170, 'max\_depth': 7, 'learning\_rate': 0.285384684135105, 'subsample': 0.845715662239872, 'colsample\_bytree': 0.56411}

[I] 2025-05-31 02:47:06.349 Trial 16 finished with value: 0.001502956344529395 and parameters: {'n\_estimators': 106, 'max\_depth': 8, 'learning\_rate': 0.1558224026994435, 'subsample': 0.8522542028713603, 'colsample\_bytree': 0.6817}

[I] 2025-05-31 02:47:06.471 Trial 17 finished with value: 0.001483931469168737 and parameters: {'n\_estimators': 164, 'max\_depth': 7, 'learning\_rate': 0.04758851197304258, 'subsample': 0.849322390437041, 'colsample\_bytree': 0.5782}

[I] 2025-05-31 02:47:06.592 Trial 18 finished with value: 0.00131635454745705 and parameters: {'n\_estimators': 170, 'max\_depth': 7, 'learning\_rate': 0.1964818136207783, 'subsample': 0.7368443304016144, 'colsample\_bytree': 0.5645}

[I] 2025-05-31 02:47:06.736 Trial 19 finished with value: 0.001464757220340672 and parameters: {'n\_estimators': 237, 'max\_depth': 8, 'learning\_rate': 0.0977237789006194, 'subsample': 0.737498382442626, 'colsample\_bytree': 0.6812}

[I] 2025-05-31 02:47:06.842 A new study created in memory with name: no-name-363725fa-3b19-408b-80fa-fab28da74cb

[I] 2025-05-31 02:47:06.971 Trial 0 finished with value: 0.003204535732962373 and parameters: {'n\_estimators': 163, 'max\_depth': 8, 'learning\_rate': 0.020142801199421887, 'subsample': 0.7851537375645101, 'colsample\_bytree': 0.58587}

[I] 2025-05-31 02:47:07.086 Trial 1 finished with value: 0.003044713156120127 and parameters: {'n\_estimators': 203, 'max\_depth': 3, 'learning\_rate': 0.1221103094594808, 'subsample': 0.8231404127237676, 'colsample\_bytree': 0.602748}

[I] 2025-05-31 02:47:07.159 Trial 2 finished with value: 0.00304950327474422 and parameters: {'n\_estimators': 84, 'max\_depth': 4, 'learning\_rate': 0.0492868446281757, 'subsample': 0.86008551168946, 'colsample\_bytree': 0.9790139}

[I] 2025-05-31 02:47:07.237 Trial 3 finished with value: 0.0031052622941923 and parameters: {'n\_estimators': 105, 'max\_depth': 7, 'learning\_rate': 0.02083467901701018, 'subsample': 0.6074371826228702, 'colsample\_bytree': 0.812420}

[I] 2025-05-31 02:47:07.333 Trial 4 finished with value: 0.00302835856059568 and parameters: {'n\_estimators': 97, 'max\_depth': 7, 'learning\_rate': 0.01568549250369829, 'subsample': 0.925571981177211, 'colsample\_bytree': 0.771082}

[I] 2025-05-31 02:47:07.427 Trial 5 finished with value: 0.00307593280316883 and parameters: {'n\_estimators': 126, 'max\_depth': 5, 'learning\_rate': 0.1309377325050818, 'subsample': 0.663404274898831, 'colsample\_bytree': 0.80231}

[I] 2025-05-31 02:47:07.561 Trial 6 finished with value: 0.003092803859189338 and parameters: {'n\_estimators': 259, 'max\_depth': 5, 'learning\_rate': 0.250142593046831, 'subsample': 0.804989370394242, 'colsample\_bytree': 0.9547015}

[I] 2025-05-31 02:47:07.639 Trial 7 finished with value: 0.00303397759826429 and parameters: {'n\_estimators': 104, 'max\_depth': 5, 'learning\_rate': 0.0166585342647745, 'subsample': 0.6772870357131997, 'colsample\_bytree': 0.8202055}

[I] 2025-05-31 02:47:07.752 Trial 8 finished with value: 0.003034887814304844 and parameters: {'n\_estimators': 203, 'max\_depth': 3, 'learning\_rate': 0.03362365079623894, 'subsample': 0.7984552944031079, 'colsample\_bytree': 0.9928}

[I] 2025-05-31 02:47:07.858 Trial 9 finished with value: 0.00308089236138748 and parameters: {'n\_estimators': 181, 'max\_depth': 8, 'learning\_rate': 0.13309788961177582, 'subsample': 0.50805181320848866, 'colsample\_bytree': 0.9211259}

[I] 2025-05-31 02:47:07.944 Trial 10 finished with value: 0.0030530548027907345 and parameters: {'n\_estimators': 56, 'max\_depth': 7, 'learning\_rate': 0.01075596881977582, 'subsample': 0.5014148691651454, 'colsample\_bytree': 0.6897}

[I] 2025-05-31 02:47:07.963 Trial 11 finished with value: 0.0030293913838704865 and parameters: {'n\_estimators': 141, 'max\_depth': 10, 'learning\_rate': 0.02752240271692956, 'subsample': 0.661029117714319, 'colsample\_bytree': 0.530}

[I] 2025-05-31 02:47:08.033 Trial 12 finished with value: 0.003018420627261268 and parameters: {'n\_estimators': 158, 'max\_depth': 7, 'learning\_rate': 0.02477511983891959, 'subsample': 0.7416701093249201, 'colsample\_bytree': 0.62850}

[I] 2025-05-31 02:47:08.183 Trial 13 finished with value: 0.003018420627261268 and parameters: {'n\_estimators': 158, 'max\_depth': 7, 'learning\_rate': 0.02752240271692956, 'subsample': 0.7416701093249201, 'colsample\_bytree': 0.62850}

[I] 2025-05-31 02:47:08.375 Trial 13 finished with value: 0.003035151508283218 and parameters: {'n\_estimators': 270, 'max\_depth': 6, 'learning\_rate': 0.04314323178873298, 'subsample': 0.581215442999742, 'colsample\_bytree': 0.7056}

[I] 2025-05-31 02:47:08.545 Trial 14 finished with value: 0.00305328817302185 and parameters: {'n\_estimators': 233, 'max\_depth': 7, 'learning\_rate': 0.070384456849494, 'subsample': 0.7169343456467336, 'colsample\_bytree': 0.8888}

[I] 2025-05-31 02:47:08.735 Trial 15 finished with value: 0.0030203784755651 and parameters: {'n\_estimators': 141, 'max\_depth': 6, 'learning\_rate': 0.010118399117086265, 'subsample': 0.603676015186145, 'colsample\_bytree': 0.67030}

[I] 2025-05-31 02:47:09.044 Trial 16 finished with value: 0.003042875130785248 and parameters: {'n\_estimators': 53, 'max\_depth': 7, 'learning\_rate': 0.0224867117821209263, 'subsample': 0.7339143213093918, 'colsample\_bytree': 0.8588}

[I] 2025-05-31 02:47:09.063 Trial 17 finished with value: 0.0030795326137430863 and parameters: {'n\_estimators': 170, 'max\_depth': 6, 'learning\_rate': 0.07000165830288561, 'subsample': 0.5810403396927329, 'colsample\_bytree': 0.6252}

[I] 2025-05-31 02:47:10.511 Trial 19 finished with value: 0.0030383159687755 and parameters: {'n\_estimators': 79, 'max\_depth': 8, 'learning\_rate': 0.01437238807710883, 'subsample': 0.618632836216199, 'colsample\_bytree': 0.51987}

[I] 2025-05-31 02:47:10.620 A new study created in memory with name: no-name-9d37b4dc-8dfc-4b3e-8dfe-753d4754506

[I] 2025-05-31 02:47:10.872 Trial 0 finished with value: 0.00200262673120870070 and parameters: {'n\_estimators': 223, 'max\_depth': 5, 'learning\_rate': 0.016978433737904182, 'subsample': 0.945684959691659, 'colsample\_bytree': 0.9457}

[I] 2025-05-31 02:47:11.073 Trial 1 finished with value: 0.00200254235564931 and parameters: {'n\_estimators': 202, 'max\_depth': 6, 'learning\_rate': 0.133821588778070076, 'subsample': 0.5490526490922044, 'colsample\_bytree': 0.88719764}

[I] 2025-05-31 02:47:11.276 Trial 2 finished with value: 0.0020127148479192 and parameters: {'n\_estimators': 300, 'max\_depth': 8, 'learning\_rate': 0.072972843848849, 'subsample': 0.8144063838207476, 'colsample\_bytree': 0.619547}

[I] 2025-05-31 02:47:11.594 Trial 3 finished with value: 0.00200846616201284 and parameters: {'n\_estimators': 299, 'max\_depth': 4, 'learning\_rate': 0.0121534191938843, 'subsample': 0.57315028272356372, 'colsample\_bytree': 0.81303}

[I] 2025-05-31 02:47:11.669 Trial 4 finished with value: 0.0020016864498954083 and parameters: {'n\_estimators': 88, 'max\_depth': 5, 'learning\_rate': 0.020664942892004893, 'subsample': 0.5155397711783259, 'colsample\_bytree': 0.81297}

[I] 2025-05-31 02:47:11.733 Trial 5 finished with value: 0.0020391173371373 and parameters: {'n\_estimators': 52, 'max\_depth': 6, 'learning\_rate': 0.17359861585600975, 'subsample': 0.5922785546604945, 'colsample\_bytree': 0.952304}

[I] 2025-05-31 02:47:11.855 Trial 6 finished with value: 0.00200271691844694 and parameters: {'n\_estimators': 229, 'max\_depth': 8, 'learning\_rate': 0.052216945970451, 'subsample': 0.52216945970451, 'colsample\_bytree': 0.9175811}

[I] 2025-05-31 02:47:11.951 Trial 7 finished with value: 0.00198916880004545 and parameters: {'n\_estimators': 164, 'max\_depth': 6, 'learning\_rate': 0.1176898550909787, 'subsample': 0.791536223235724, 'colsample\_bytree': 0.5377441}

[I] 2025-05-31 02:47:12.095 Trial 8 finished with value: 0.0020233575988278 and parameters: {'n\_estimators': 274, 'max\_depth': 8, 'learning\_rate': 0.063415195898365, 'subsample': 0.7407813175333028, 'colsample\_bytree': 0.944471}

[I] 2025-05-31 02:47:12.198 Trial 9 finished with value: 0.001957130251011738 and parameters: {'n\_estimators': 186, 'max\_depth': 6, 'learning\_rate': 0.1738260009107362, 'subsample': 0.7208119956041488, 'colsample\_bytree': 0.599815}

[I] 2025-05-31 02:47:12.309 Trial 10 finished with value: 0.0019257532504246562 and parameters: {'n\_estimators': 139, 'max\_depth': 3, 'learning\_rate': 0.2762082615642374, 'subsample': 0.689636950157373, 'colsample\_bytree': 0.6706}

[I] 2025-05-31 02:47:12.430 Trial 11 finished with value: 0.001993040759708364 and parameters: {'n\_estimators': 146, 'max\_depth': 3, 'learning\_rate': 0.29197807507347369, 'subsample': 0.673723801736003, 'colsample\_bytree': 0.66435}

[I] 2025-05-31 02:47:12.533 Trial 12 finished with value: 0.001964089435955636 and parameters: {'n\_estimators': 115, 'max\_depth': 3, 'learning\_rate': 0.2928012325912224, 'subsample': 0.686244725159019, 'colsample\_bytree': 0.6812}

[I] 2025-05-31 02:47:12.642 Trial 13 finished with value: 0.001907147637388162 and parameters: {'n\_estimators': 132, 'max\_depth': 7, 'learning\_rate': 0.189698176272625, 'subsample': 0.74753817059482, 'colsample\_bytree': 0.5075}

[I] 2025-05-31 02:47:12.752 Trial 14 finished with value: 0.00180817948084057 and parameters: {'n\_estimators': 130, 'max\_depth': 6, 'learning\_rate': 0.21274617657640243, 'subsample': 0.6526656759514243, 'colsample\_bytree': 0.52020}

[I] 2025-05-31 02:47:12.855 Trial 15 finished with value: 0.00194038080618583 and parameters: {'n\_estimators': 108, 'max\_depth': 7, 'learning\_rate': 0.08473956334515451, 'subsample': 0.6243020062523222, 'colsample\_bytree': 0.5004}

[I] 2025-05-31 02:47:12.956 Trial 16 finished with value: 0.0020453899648239 and parameters: {'n\_estimators': 71, 'max\_depth': 7, 'learning\_rate': 0.04197306165327382, 'subsample': 0.7451948581989338, 'colsample\_bytree': 0.569810}

[I] 2025-05-31 02:47:13.049 Trial 17 finished with value: 0.00203704314776057083 and parameters: {'n\_estimators': 125, 'max\_depth': 6, 'learning\_rate': 0.17228671782107212, 'subsample': 0.8369180107212, 'colsample\_bytree': 0.50840}

[I] 2025-05-31 02:47:13.193 Trial 18 finished with value: 0.001995655994530487 and parameters: {'n\_estimators': 165, 'max\_depth': 7, 'learning\_rate': 0.11226865330225102, 'subsample': 0.62004348379281, 'colsample\_bytree': 0.7258}

[I] 2025-05-31 02:47:13.287 Trial 19 finished with value: 0.002081712177316757 and parameters: {'n\_estimators': 86, 'max\_depth': 7, 'learning\_rate': 0.21911228738866073, 'subsample': 0.7740056819861056, 'colsample\_bytree': 0.562784}

[I] 2025-05-31 02:47:13.386 Trial 20 finished with value: 0.002097043975852811 and parameters: {'n\_estimators': 284, 'max\_depth': 7, 'learning\_rate': 0.0815921668601134, 'subsample': 0.7455708117711335, 'colsample\_bytree': 0.5060635}

[I] 2025-05-31 02:47:13.681 Trial 1 finished with value: 0.0030034498078779649 and parameters: {'n\_estimators': 138, 'max\_depth': 5, 'learning\_rate': 0.044339530705040606, 'subsample': 0.6070350016636372, 'colsample\_bytree': 0.6280}

[I] 2025-05-31 02:47:14.762 Trial 2 finished with value: 0.002028748460807252 and parameters: {'n\_estimators': 103, 'max\_depth': 7, 'learning\_rate': 0.06001018071995084, 'subsample': 0.635054199664941, 'colsample\_bytree': 0.97368}

[I] 2025-05-31 02:47:14.850 Trial 3 finished with value: 0.0020951730152011738 and parameters: {'n\_estimators': 186, 'max\_depth': 6, 'learning\_rate': 0.1738260009107362, 'subsample': 0.7208119956041488, 'colsample\_bytree': 0.599815}

[I] 2025-05-31 02:47:14.962 Trial 4 finished with value: 0.002073057350416867 and parameters: {'n\_estimators': 139, 'max\_depth': 8, 'learning\_rate': 0.2762082615642374, 'subsample': 0.689636950157373, 'colsample\_bytree': 0.6706}

[I] 2025-05-31 02:47:14.980 Trial 5 finished with value: 0.0020942144861606206 and parameters: {'n\_estimators': 124, 'max\_depth': 7, 'learning\_rate': 0.19739719074305, 'subsample': 0.9442040349371571, 'colsample\_bytree': 0.98846555}

[I] 2025-05-31 02:47:14.992 Trial 13 finished with value: 0.0029476146239471164 and parameters: {'n\_estimators': 222, 'max\_depth': 3, 'learning\_rate': 0.14409573165474573, 'subsample': 0.5251690130621502, 'colsample\_bytree': 0.9027}

[I] 2025-05-31 02:47:15.164 Trial 14 finished with value: 0.00293426239477994 and parameters: {'n\_estimators': 288, 'max\_depth': 6, 'learning\_rate': 0.1175308148841397, 'subsample': 0.837783265514387, 'colsample\_bytree': 0.997741}

[I] 2025-05-31 02:47:15.318 Trial 15 finished with value: 0.0030213984515037 and parameters: {'n\_estimators': 253, 'max\_depth': 3, 'learning\_rate': 0.2953287657249589, 'subsample': 0.500422894039354, 'colsample\_bytree': 0.87271722}

[I] 2025-05-31 02:47:15.443 Trial 16 finished with value: 0.00299660600176823 and parameters: {'n\_estimators': 173, 'max\_depth': 4, 'learning\_rate': 0.147370316382765969, 'subsample': 0.68826303438807, 'colsample\_bytree': 0.940660}

[I] 2025-05-31 02:47:15.606 Trial 17 finished with value: 0.002937049479132114 and parameters: {'n\_estimators': 201, 'max\_depth': 6, 'learning\_rate': 0.1032950408869646, 'subsample': 0.826798255177072, 'colsample\_bytree': 0.71974}

[I] 2025-05-31 02:47:15.705 Trial 18 finished with value: 0.003007147024617207 and parameters: {'n\_estimators': 77, 'max\_depth': 8, 'learning\_rate': 0.1722814994852, 'subsample': 0.5601383223929051, 'colsample\_bytree': 0.85042}

[I] 2025-05-31 02:47:15.956 A new study created in memory with name: no-name-74ebccff-ffb4-4f6a-825d-4f31de434a

[I] 2025-05-31 02:47:16.122 Trial 0 finished with value: 0.002814699371237664 and parameters: {'n\_estimators': 244, 'max\_depth': 5, 'learning\_rate': 0.0400462427337662, 'subsample': 0.9975331825063, 'colsample\_bytree': 0.994765}

[I] 2025-05-31 02:47:16.192 Trial 1 finished with value: 0.0028113702309744 and parameters: {'n\_estimators': 77, 'max\_depth': 6, 'learning\_rate': 0.0176328127167958, 'subsample': 0.5474319328461

[1] 2025-05-31 02:47:21,492 Trial 3 finished with value: 0.0037539385704704835 and parameters: {'n\_estimators': 256, 'max\_depth': 7, 'learning\_rate': 0.0503530001824945956, 'subsample': 0.771613578406019, 'colsample\_bytree': 0.777916},  
 [1] 2025-05-31 02:47:21,567 Trial 4 finished with value: 0.00381950782612732 and parameters: {'n\_estimators': 52, 'max\_depth': 7, 'learning\_rate': 0.0598111530052686, 'subsample': 0.593201154940355, 'colsample\_bytree': 0.6985252},  
 [1] 2025-05-31 02:47:21,728 Trial 5 finished with value: 0.003833528805994194 and parameters: {'n\_estimators': 173, 'max\_depth': 3, 'learning\_rate': 0.0699200579012788, 'subsample': 0.569018311689365, 'colsample\_bytree': 0.96774},  
 [1] 2025-05-31 02:47:21,912 Trial 6 finished with value: 0.00371751791637237 and parameters: {'n\_estimators': 146, 'max\_depth': 6, 'learning\_rate': 0.2400918237127108, 'subsample': 0.9373668893290131, 'colsample\_bytree': 0.56028},  
 [1] 2025-05-31 02:47:22,508 Trial 7 finished with value: 0.00377369251470946 and parameters: {'n\_estimators': 62, 'max\_depth': 4, 'learning\_rate': 0.010770771008470017, 'subsample': 0.7058641710845321, 'colsample\_bytree': 0.772877},  
 [1] 2025-05-31 02:47:22,610 Trial 8 finished with value: 0.003691523522504162 and parameters: {'n\_estimators': 78, 'max\_depth': 7, 'learning\_rate': 0.264995313501974, 'subsample': 0.7356793035611172, 'colsample\_bytree': 0.8135455},  
 [1] 2025-05-31 02:47:22,922 Trial 9 finished with value: 0.00376690841624514 and parameters: {'n\_estimators': 156, 'max\_depth': 8, 'learning\_rate': 0.021677372311514882, 'subsample': 0.537125024366116, 'colsample\_bytree': 0.7563},  
 [1] 2025-05-31 02:47:23,721 Trial 10 finished with value: 0.00368107785935724 and parameters: {'n\_estimators': 112, 'max\_depth': 8, 'learning\_rate': 0.28650091569432005, 'subsample': 0.2846136116482327, 'colsample\_bytree': 0.8544},  
 [1] 2025-05-31 02:47:24,000 Trial 11 finished with value: 0.0037082361363371 and parameters: {'n\_estimators': 114, 'max\_depth': 8, 'learning\_rate': 0.2802114753210988, 'subsample': 0.802212397985714, 'colsample\_bytree': 0.87513589},  
 [1] 2025-05-31 02:47:24,143 Trial 12 finished with value: 0.00371266651809571 and parameters: {'n\_estimators': 112, 'max\_depth': 7, 'learning\_rate': 0.123409936459796, 'subsample': 0.8593102193133997, 'colsample\_bytree': 0.866702},  
 [1] 2025-05-31 02:47:24,285 Trial 13 finished with value: 0.003719170978066635 and parameters: {'n\_estimators': 102, 'max\_depth': 8, 'learning\_rate': 0.15535344556799852, 'subsample': 0.698130324879891, 'colsample\_bytree': 0.9980},  
 [1] 2025-05-31 02:47:24,579 Trial 14 finished with value: 0.0037284017035215 and parameters: {'n\_estimators': 202, 'max\_depth': 7, 'learning\_rate': 0.1923758331923349, 'subsample': 0.81107486852882, 'colsample\_bytree': 0.832271},  
 [1] 2025-05-31 02:47:24,697 Trial 15 finished with value: 0.0037024509076724597 and parameters: {'n\_estimators': 137, 'max\_depth': 5, 'learning\_rate': 0.288533877457219, 'subsample': 0.7193548408459802, 'colsample\_bytree': 0.685192},  
 [1] 2025-05-31 02:47:24,795 Trial 16 finished with value: 0.0037665225171045195 and parameters: {'n\_estimators': 78, 'max\_depth': 8, 'learning\_rate': 0.117880170360159, 'subsample': 0.9749085148471923, 'colsample\_bytree': 0.9103271},  
 [1] 2025-05-31 02:47:24,904 Trial 17 finished with value: 0.003704262583618336 and parameters: {'n\_estimators': 128, 'max\_depth': 6, 'learning\_rate': 0.03579203630586806, 'subsample': 0.864879533940819, 'colsample\_bytree': 0.82560},  
 [1] 2025-05-31 02:47:25,000 Trial 18 finished with value: 0.0036173648051851 and parameters: {'n\_estimators': 75, 'max\_depth': 7, 'learning\_rate': 0.18257724461726825, 'subsample': 0.6579527413191238, 'colsample\_bytree': 0.951914},  
 [1] 2025-05-31 02:47:25,141 Trial 19 finished with value: 0.003698513848608616 and parameters: {'n\_estimators': 221, 'max\_depth': 8, 'learning\_rate': 0.17006555075378937, 'subsample': 0.406902082317273, 'colsample\_bytree': 0.9616},  
 [1] 2025-05-31 02:47:25,223 A new study created in memory with name: no-name-dbcdf3f0-4dcb-43cb-6a93910a5bf2  
 [1] 2025-05-31 02:47:25,341 Trial 0 finished with value: 0.0042376598585267652 and parameters: {'n\_estimators': 83, 'max\_depth': 4, 'learning\_rate': 0.12801328887401195, 'subsample': 0.6718082784108859, 'colsample\_bytree': 0.5102308},  
 [1] 2025-05-31 02:47:25,413 Trial 1 finished with value: 0.004230015119331884 and parameters: {'n\_estimators': 78, 'max\_depth': 4, 'learning\_rate': 0.05026176897923524, 'subsample': 0.957051073256557905, 'colsample\_bytree': 0.743394},  
 [1] 2025-05-31 02:47:25,512 Trial 2 finished with value: 0.0042266203238218785 and parameters: {'n\_estimators': 134, 'max\_depth': 5, 'learning\_rate': 0.04726733157503192, 'subsample': 0.98589043075921, 'colsample\_bytree': 0.82496},  
 [1] 2025-05-31 02:47:25,614 Trial 3 finished with value: 0.004227774415939337 and parameters: {'n\_estimators': 171, 'max\_depth': 8, 'learning\_rate': 0.02633537526769819, 'subsample': 0.557431564504421, 'colsample\_bytree': 0.568296},  
 [1] 2025-05-31 02:47:25,696 Trial 4 finished with value: 0.00429712490199555 and parameters: {'n\_estimators': 102, 'max\_depth': 5, 'learning\_rate': 0.03352730377121036, 'subsample': 0.718098874052907, 'colsample\_bytree': 0.82853},  
 [1] 2025-05-31 02:47:25,811 Trial 5 finished with value: 0.00425653277030694 and parameters: {'n\_estimators': 200, 'max\_depth': 4, 'learning\_rate': 0.01949589666680611, 'subsample': 0.9499911951300609, 'colsample\_bytree': 0.67413},  
 [1] 2025-05-31 02:47:25,908 Trial 6 finished with value: 0.0042392324042444 and parameters: {'n\_estimators': 144, 'max\_depth': 5, 'learning\_rate': 0.030849518468199805, 'subsample': 0.582200908499868, 'colsample\_bytree': 0.544105},  
 [1] 2025-05-31 02:47:26,012 Trial 7 finished with value: 0.0042294397859212 and parameters: {'n\_estimators': 66, 'max\_depth': 7, 'learning\_rate': 0.0141214533842277, 'subsample': 0.987225398454472, 'colsample\_bytree': 0.709852},  
 [1] 2025-05-31 02:47:26,154 Trial 8 finished with value: 0.00423837975859635 and parameters: {'n\_estimators': 283, 'max\_depth': 7, 'learning\_rate': 0.2355976380014393, 'subsample': 0.5147950064843672, 'colsample\_bytree': 0.61887},  
 [1] 2025-05-31 02:47:26,229 Trial 9 finished with value: 0.00422682403017313 and parameters: {'n\_estimators': 60, 'max\_depth': 5, 'learning\_rate': 0.109695808477118, 'subsample': 0.9103250187017042, 'colsample\_bytree': 0.8003163539},  
 [1] 2025-05-31 02:47:26,393 Trial 10 finished with value: 0.004226657935320936 and parameters: {'n\_estimators': 240, 'max\_depth': 3, 'learning\_rate': 0.10313518950610716, 'subsample': 0.832598572485026, 'colsample\_bytree': 0.9831},  
 [1] 2025-05-31 02:47:26,513 Trial 11 finished with value: 0.0042542491926732 and parameters: {'n\_estimators': 212, 'max\_depth': 7, 'learning\_rate': 0.01959826051959465, 'subsample': 0.8585212176312204, 'colsample\_bytree': 0.6569},  
 [1] 2025-05-31 02:47:26,652 Trial 12 finished with value: 0.004227453194728 and parameters: {'n\_estimators': 214, 'max\_depth': 7, 'learning\_rate': 0.017362568238579148, 'subsample': 0.8231064697061568, 'colsample\_bytree': 0.634107},  
 [1] 2025-05-31 02:47:26,791 Trial 13 finished with value: 0.004284599305893 and parameters: {'n\_estimators': 210, 'max\_depth': 6, 'learning\_rate': 0.01914704981645971, 'subsample': 0.82672159309822, 'colsample\_bytree': 0.67930},  
 [1] 2025-05-31 02:47:26,949 Trial 14 finished with value: 0.004228848580148481 and parameters: {'n\_estimators': 267, 'max\_depth': 3, 'learning\_rate': 0.01020936357988175, 'subsample': 0.901119112502444, 'colsample\_bytree': 0.92609},  
 [1] 2025-05-31 02:47:27,113 Trial 15 finished with value: 0.0042131196793547 and parameters: {'n\_estimators': 203, 'max\_depth': 8, 'learning\_rate': 0.0781860617906765, 'subsample': 0.7510200633903244, 'colsample\_bytree': 0.651888},  
 [1] 2025-05-31 02:47:27,236 Trial 16 finished with value: 0.0042313694931511 and parameters: {'n\_estimators': 175, 'max\_depth': 6, 'learning\_rate': 0.021946043804809698, 'subsample': 0.936488798877736, 'colsample\_bytree': 0.5850},  
 [1] 2025-05-31 02:47:27,382 Trial 17 finished with value: 0.0042885004551577 and parameters: {'n\_estimators': 245, 'max\_depth': 4, 'learning\_rate': 0.014742055091961378, 'subsample': 0.856972812401334, 'colsample\_bytree': 0.77153},  
 [1] 2025-05-31 02:47:27,545 Trial 18 finished with value: 0.004231675532031673 and parameters: {'n\_estimators': 298, 'max\_depth': 6, 'learning\_rate': 0.0291451323434089, 'subsample': 0.746731275036987, 'colsample\_bytree': 0.7202159},  
 [1] 2025-05-31 02:47:27,658 Trial 19 finished with value: 0.004232066349590301 and parameters: {'n\_estimators': 143, 'max\_depth': 7, 'learning\_rate': 0.01155221392865569, 'subsample': 0.6716520453119121, 'colsample\_bytree': 0.8859},  
 [1] 2025-05-31 02:47:27,782 A new study created in memory with name: no-name-b400c128-6102-408b-8ff0-5d0429870504  
 [1] 2025-05-31 02:47:27,918 Trial 0 finished with value: 0.001092327597454532 and parameters: {'n\_estimators': 193, 'max\_depth': 6, 'learning\_rate': 0.02887310134643067, 'subsample': 0.6472115890823508, 'colsample\_bytree': 0.94472},  
 [1] 2025-05-31 02:47:27,991 Trial 1 finished with value: 0.001053364795807615 and parameters: {'n\_estimators': 83, 'max\_depth': 3, 'learning\_rate': 0.03505700539510421, 'subsample': 0.9304201491311732, 'colsample\_bytree': 0.8284948},  
 [1] 2025-05-31 02:47:28,101 Trial 2 finished with value: 0.00099649249763567 and parameters: {'n\_estimators': 86, 'max\_depth': 6, 'learning\_rate': 0.077292038038794, 'subsample': 0.653632528982959, 'colsample\_bytree': 0.743121},  
 [1] 2025-05-31 02:47:28,241 Trial 3 finished with value: 0.001037368280201065 and parameters: {'n\_estimators': 271, 'max\_depth': 4, 'learning\_rate': 0.0366537618773359, 'subsample': 0.9370596125303019, 'colsample\_bytree': 0.71997},  
 [1] 2025-05-31 02:47:28,344 Trial 4 finished with value: 0.00113266550420319 and parameters: {'n\_estimators': 174, 'max\_depth': 5, 'learning\_rate': 0.044312456333651135, 'subsample': 0.634970267959844, 'colsample\_bytree': 0.6009},  
 [1] 2025-05-31 02:47:28,442 Trial 5 finished with value: 0.0011623881727399 and parameters: {'n\_estimators': 156, 'max\_depth': 5, 'learning\_rate': 0.14817581847136072, 'subsample': 0.7579780185173598, 'colsample\_bytree': 0.88699},  
 [1] 2025-05-31 02:47:28,576 Trial 6 finished with value: 0.001035352464425897 and parameters: {'n\_estimators': 276, 'max\_depth': 3, 'learning\_rate': 0.0273655519017576, 'subsample': 0.58372002771585814, 'colsample\_bytree': 0.7240},  
 [1] 2025-05-31 02:47:28,697 Trial 7 finished with value: 0.0010370460404392845 and parameters: {'n\_estimators': 235, 'max\_depth': 3, 'learning\_rate': 0.01138514578402416, 'subsample': 0.76532987657661, 'colsample\_bytree': 0.76688},  
 [1] 2025-05-31 02:47:28,833 Trial 8 finished with value: 0.0009907068130934203 and parameters: {'n\_estimators': 257, 'max\_depth': 5, 'learning\_rate': 0.05082869100451121, 'subsample': 0.790797347178352, 'colsample\_bytree': 0.95823},  
 [1] 2025-05-31 02:47:28,950 Trial 9 finished with value: 0.0010556286585322 and parameters: {'n\_estimators': 216, 'max\_depth': 3, 'learning\_rate': 0.02722280308794, 'subsample': 0.5945478051249048, 'colsample\_bytree': 0.6200},  
 [1] 2025-05-31 02:47:29,164 Trial 10 finished with value: 0.0010572326178457885 and parameters: {'n\_estimators': 297, 'max\_depth': 8, 'learning\_rate': 0.276278039784939303, 'subsample': 0.8532046123923681, 'colsample\_bytree': 0.97651},  
 [1] 2025-05-31 02:47:29,253 Trial 11 finished with value: 0.00931653170411194 and parameters: {'n\_estimators': 70, 'max\_depth': 7, 'learning\_rate': 0.02947378946942126, 'subsample': 0.7069114860919464, 'colsample\_bytree': 0.51155},  
 [1] 2025-05-31 02:47:29,358 Trial 12 finished with value: 0.008956834669863973 and parameters: {'n\_estimators': 126, 'max\_depth': 7, 'learning\_rate': 0.0843420579896805, 'subsample': 0.504538711748949, 'colsample\_bytree': 0.5040},  
 [1] 2025-05-31 02:47:29,439 Trial 13 finished with value: 0.00116893061971549 and parameters: {'n\_estimators': 50, 'max\_depth': 8, 'learning\_rate': 0.1186500288230408, 'subsample': 0.5002937424040891, 'colsample\_bytree': 0.506830},  
 [1] 2025-05-31 02:47:29,546 Trial 14 finished with value: 0.001098188997806591 and parameters: {'n\_estimators': 138, 'max\_depth': 7, 'learning\_rate': 0.08860139467588027, 'subsample': 0.512269610515234, 'colsample\_bytree': 0.50690},  
 [1] 2025-05-31 02:47:29,648 Trial 15 finished with value: 0.001096281050926967 and parameters: {'n\_estimators': 107, 'max\_depth': 7, 'learning\_rate': 0.02128807846488394, 'subsample': 0.683972655211494, 'colsample\_bytree': 0.60651},  
 [1] 2025-05-31 02:47:29,752 Trial 16 finished with value: 0.0009683756249947499 and parameters: {'n\_estimators': 124, 'max\_depth': 7, 'learning\_rate': 0.0755261602943574, 'subsample': 0.7157272724712603, 'colsample\_bytree': 0.5444},  
 [1] 2025-05-31 02:47:29,840 Trial 17 finished with value: 0.0010208512608203623 and parameters: {'n\_estimators': 69, 'max\_depth': 7, 'learning\_rate': 0.01773520940973112, 'subsample': 0.9325760508812455, 'colsample\_bytree': 0.6771},  
 [1] 2025-05-31 02:47:29,942 Trial 18 finished with value: 0.001066532539584818 and parameters: {'n\_estimators': 107, 'max\_depth': 8, 'learning\_rate': 0.1166818292879491, 'subsample': 0.568174963248344, 'colsample\_bytree': 0.5640},  
 [1] 2025-05-31 02:47:30,026 Trial 19 finished with value: 0.000995387223122031 and parameters: {'n\_estimators': 58, 'max\_depth': 6, 'learning\_rate': 0.18062875373122214, 'subsample': 0.8393614547994126, 'colsample\_bytree': 0.66327},  
 [1] 2025-05-31 02:47:30,125 A new study created in memory with name: no-name-bc6e56d0-d1a0-43f0-84fc-f979299a9ce  
 [1] 2025-05-31 02:47:30,231 Trial 0 finished with value: 0.000862245193821732 and parameters: {'n\_estimators': 80, 'max\_depth': 4, 'learning\_rate': 0.0350623558750159, 'subsample': 0.50894266955495925, 'colsample\_bytree': 0.9811798},  
 [1] 2025-05-31 02:47:30,352 Trial 1 finished with value: 0.000820914904681778 and parameters: {'n\_estimators': 233, 'max\_depth': 5, 'learning\_rate': 0.1997313826663699, 'subsample': 0.953235506172121, 'colsample\_bytree': 0.74480},  
 [1] 2025-05-31 02:47:30,463 Trial 2 finished with value: 0.000887483588028797 and parameters: {'n\_estimators': 215, 'max\_depth': 3, 'learning\_rate': 0.1019405440787083, 'subsample': 0.7867264942825205, 'colsample\_bytree': 0.606197},  
 [1] 2025-05-31 02:47:30,554 Trial 3 finished with value: 0.000871540475882406 and parameters: {'n\_estimators': 136, 'max\_depth': 4, 'learning\_rate': 0.082900951010584919, 'subsample': 0.612092729903649, 'colsample\_bytree': 0.57271},  
 [1] 2025-05-31 02:47:30,721 Trial 4 finished with value: 0.0008120596030925913 and parameters: {'n\_estimators': 202, 'max\_depth': 6, 'learning\_rate': 0.0149320734921061, 'subsample': 0.7598019531237644, 'colsample\_bytree': 0.9959},  
 [1] 2025-05-31 02:47:30,811 Trial 5 finished with value: 0.0008004215648439456 and parameters: {'n\_estimators': 130, 'max\_depth': 7, 'learning\_rate': 0.020135208145842546, 'subsample': 0.9172384524354564, 'colsample\_bytree': 0.78194},  
 [1] 2025-05-31 02:47:30,904 Trial 6 finished with value: 0.00035373182140596 and parameters: {'n\_estimators': 156, 'max\_depth': 3, 'learning\_rate': 0.077243338718213, 'subsample': 0.771423433671813, 'colsample\_bytree': 0.81611},  
 [1] 2025-05-31 02:47:31,061 Trial 7 finished with value: 0.00084977095504774 and parameters: {'n\_estimators': 178, 'max\_depth': 3, 'learning\_rate': 0.050826772897948, 'subsample': 0.51025867772897948, 'colsample\_bytree': 0.8574},  
 [1] 2025-05-31 02:47:31,091 Trial 8 finished with value: 0.0008199613306636919 and parameters: {'n\_estimators': 126, 'max\_depth': 8, 'learning\_rate': 0.015824092336860896, 'subsample': 0.671115193787888, 'colsample\_bytree': 0.82910},  
 [1] 2025-05-31 02:47:31,252 Trial 9 finished with value: 0.00086828037718510839 and parameters: {'n\_estimators': 248, 'max\_depth': 4, 'learning\_rate': 0.2051724474967103, 'subsample': 0.61393576436352, 'colsample\_bytree': 0.805507},  
 [1] 2025-05-31 02:47:31,411 Trial 10 finished with value: 0.0008283881717261783 and parameters: {'n\_estimators': 291, 'max\_depth': 5, 'learning\_rate': 0.0977243293211783, 'subsample': 0.907724529823921, 'colsample\_bytree': 0.71585},  
 [1] 2025-05-31 02:47:31,499 Trial 11 finished with value: 0.000823688594750359 and parameters: {'n\_estimators': 66, 'max\_depth': 6, 'learning\_rate': 0.0112174031376306, 'subsample': 0.8273333030159708, 'colsample\_bytree': 0.96400},  
 [1] 2025-05-31 02:47:31,622 Trial 12 finished with value: 0.000835609050329813 and parameters: {'n\_estimators': 104, 'max\_depth': 7, 'learning\_rate': 0.02070269032069304, 'subsample': 0.85020258194560565, 'colsample\_bytree': 0.913},  
 [1] 2025-05-31 02:47:31,787 Trial 13 finished with value: 0.000813017053819741 and parameters: {'n\_estimators': 209, 'max\_depth': 7, 'learning\_rate': 0.0640048463729484, 'subsample': 0.6198170483776877658, 'colsample\_bytree': 0.9022},  
 [1] 2025-05-31 02:47:31,944 Trial 14 finished with value: 0.0008216502958295708 and parameters: {'n\_estimators': 158, 'max\_depth': 8, 'learning\_rate': 0.24685926533295196, 'subsample': 0.8766883782655863, 'colsample\_bytree': 0.5050},  
 [1] 2025-05-31 02:47:32,277 Trial 15 finished with value: 0.000810288447735963 and parameters: {'n\_estimators': 98, 'max\_depth': 7, 'learning\_rate': 0.249440691160611027, 'subsample': 0.8596358730624338, 'colsample\_bytree': 0.679911},  
 [1] 2025-05-31 02:47:32,363 Trial 16 finished with value: 0.000816378893032638 and parameters: {'n\_estimators': 98, 'max\_depth': 5, 'learning\_rate': 0.2737442426886812, 'subsample': 0.7350637917619237, 'colsample\_bytree': 0.6882635},  
 [1] 2025-05-31 02:47:32,464 Trial 17 finished with value: 0.0008105735081971 and parameters: {'n\_estimators': 227, 'max\_depth': 7, 'learning\_rate': 0.27742919330952504, 'subsample': 0.8970578967652041, 'colsample\_bytree': 0.76809},  
 [1] 2025-05-31 02:47:32,812 Trial 18 finished with value: 0.001112035681470513 and parameters: {'n\_estimators': 257, 'max\_depth': 7, 'learning\_rate': 0.076654900988982, 'subsample': 0.65761498734904, 'colsample\_bytree': 0.7621510},  
 [1] 2025-05-31 02:47:32,936 Trial 19 finished with value: 0.001105883280144604 and parameters: {'n\_estimators': 188, 'max\_depth': 3, 'learning\_rate': 0.028133275057362

[1] 2025-05-31 02:47:41,070 Trial 11 finished with value: 0.00304108161257107 and parameters: {'n\_estimators': 247, 'max\_depth': 5, 'learning\_rate': 0.03362381361330326, 'subsample': 0.7146651010411086, 'colsample\_bytree': 0.94672},  
 [1] 2025-05-31 02:47:41,193 Trial 12 finished with value: 0.003027423925592943 and parameters: {'n\_estimators': 167, 'max\_depth': 6, 'learning\_rate': 0.02317450455050, 'subsample': 0.65014361226923, 'colsample\_bytree': 0.8441},  
 [1] 2025-05-31 02:47:41,359 Trial 13 finished with value: 0.00301474872936282913 and parameters: {'n\_estimators': 299, 'max\_depth': 4, 'learning\_rate': 0.0593957142397666, 'subsample': 0.8601316289318432, 'colsample\_bytree': 0.7173},  
 [1] 2025-05-31 02:47:41,492 Trial 14 finished with value: 0.0030172808499891608 and parameters: {'n\_estimators': 205, 'max\_depth': 5, 'learning\_rate': 0.052545599361267244, 'subsample': 0.8951113453496464, 'colsample\_bytree': 0.5024},  
 [1] 2025-05-31 02:47:41,642 Trial 15 finished with value: 0.003052651353713279 and parameters: {'n\_estimators': 143, 'max\_depth': 4, 'learning\_rate': 0.01843102687982536, 'subsample': 0.6536924915061432, 'colsample\_bytree': 0.937},  
 [1] 2025-05-31 02:47:41,789 Trial 16 finished with value: 0.003042153635173279 and parameters: {'n\_estimators': 231, 'max\_depth': 6, 'learning\_rate': 0.0357283399054606, 'subsample': 0.842076374888267, 'colsample\_bytree': 0.78289},  
 [1] 2025-05-31 02:47:41,884 Trial 17 finished with value: 0.00305263026202757 and parameters: {'n\_estimators': 82, 'max\_depth': 3, 'learning\_rate': 0.18117621429204236, 'subsample': 0.9353804930950124, 'colsample\_bytree': 0.655174},  
 [1] 2025-05-31 02:47:42,022 Trial 18 finished with value: 0.003000171898541783 and parameters: {'n\_estimators': 179, 'max\_depth': 7, 'learning\_rate': 0.07312415980820543, 'subsample': 0.6874406368532426, 'colsample\_bytree': 0.89040},  
 [1] 2025-05-31 02:47:42,157 Trial 19 finished with value: 0.00298569149048242 and parameters: {'n\_estimators': 184, 'max\_depth': 7, 'learning\_rate': 0.07312415980820543, 'subsample': 0.672857169623224, 'colsample\_bytree': 0.87181},  
 [1] 2025-05-31 02:47:42,274 A new study created in memory with name: no-name~7aeF64e~0327~4753~991a~9dc689c334  
 [1] 2025-05-31 02:47:42,409 Trial 0 finished with value: 0.0015169701308066468 and parameters: {'n\_estimators': 190, 'max\_depth': 6, 'learning\_rate': 0.04334928579135105, 'subsample': 0.8476644153859465, 'colsample\_bytree': 0.79388},  
 [1] 2025-05-31 02:47:42,498 Trial 1 finished with value: 0.0016500701786378274 and parameters: {'n\_estimators': 120, 'max\_depth': 4, 'learning\_rate': 0.16589230366222402, 'subsample': 0.5184709536101034, 'colsample\_bytree': 0.893757},  
 [1] 2025-05-31 02:47:42,588 Trial 2 finished with value: 0.0015504806233406185 and parameters: {'n\_estimators': 124, 'max\_depth': 8, 'learning\_rate': 0.0319701212077651, 'subsample': 0.7087015165800249, 'colsample\_bytree': 0.86710},  
 [1] 2025-05-31 02:47:42,734 Trial 3 finished with value: 0.001545715497054547 and parameters: {'n\_estimators': 252, 'max\_depth': 6, 'learning\_rate': 0.0330338462615188, 'subsample': 0.7048106502612832, 'colsample\_bytree': 0.75014},  
 [1] 2025-05-31 02:47:42,857 Trial 4 finished with value: 0.001543386325367832 and parameters: {'n\_estimators': 236, 'max\_depth': 7, 'learning\_rate': 0.08582402785421392, 'subsample': 0.8611400007233974, 'colsample\_bytree': 0.72469},  
 [1] 2025-05-31 02:47:42,944 Trial 5 finished with value: 0.00152095229461677677 and parameters: {'n\_estimators': 110, 'max\_depth': 8, 'learning\_rate': 0.23294324382589943, 'subsample': 0.5193742998533863, 'colsample\_bytree': 0.81293},  
 [1] 2025-05-31 02:47:43,018 Trial 6 finished with value: 0.001531373894259916 and parameters: {'n\_estimators': 79, 'max\_depth': 5, 'learning\_rate': 0.206624243851082, 'subsample': 0.739581161310105, 'colsample\_bytree': 0.8560476},  
 [1] 2025-05-31 02:47:43,109 Trial 7 finished with value: 0.001556315227419493 and parameters: {'n\_estimators': 129, 'max\_depth': 6, 'learning\_rate': 0.04723900703396574, 'subsample': 0.974554191992137, 'colsample\_bytree': 0.98138},  
 [1] 2025-05-31 02:47:43,197 Trial 8 finished with value: 0.00158164797051546 and parameters: {'n\_estimators': 130, 'max\_depth': 5, 'learning\_rate': 0.06178905331065684, 'subsample': 0.64649609495383, 'colsample\_bytree': 0.56046},  
 [1] 2025-05-31 02:47:43,281 Trial 9 finished with value: 0.001560928173145496 and parameters: {'n\_estimators': 115, 'max\_depth': 7, 'learning\_rate': 0.0234254750408625, 'subsample': 0.8876397596649754, 'colsample\_bytree': 0.9887},  
 [1] 2025-05-31 02:47:43,367 Trial 10 finished with value: 0.00139258340948944 and parameters: {'n\_estimators': 57, 'max\_depth': 3, 'learning\_rate': 0.2977953351251836, 'subsample': 0.505062595331523, 'colsample\_bytree': 0.60985},  
 [1] 2025-05-31 02:47:43,450 Trial 11 finished with value: 0.001411151452892177 and parameters: {'n\_estimators': 57, 'max\_depth': 3, 'learning\_rate': 0.2300122035775216, 'subsample': 0.5021670782323959, 'colsample\_bytree': 0.616815},  
 [1] 2025-05-31 02:47:43,538 Trial 12 finished with value: 0.0015059216260459375 and parameters: {'n\_estimators': 62, 'max\_depth': 3, 'learning\_rate': 0.1138786282737318, 'subsample': 0.589836389133196, 'colsample\_bytree': 0.59767},  
 [1] 2025-05-31 02:47:43,620 Trial 13 finished with value: 0.0015545021632670176 and parameters: {'n\_estimators': 54, 'max\_depth': 3, 'learning\_rate': 0.031444751559941955, 'subsample': 0.6037049579220551, 'colsample\_bytree': 0.6458},  
 [1] 2025-05-31 02:47:43,742 Trial 14 finished with value: 0.00146735292904407 and parameters: {'n\_estimators': 179, 'max\_depth': 4, 'learning\_rate': 0.2929110748211421, 'subsample': 0.519617772328165, 'colsample\_bytree': 0.50586},  
 [1] 2025-05-31 02:47:43,906 Trial 15 finished with value: 0.00146928392400499 and parameters: {'n\_estimators': 292, 'max\_depth': 4, 'learning\_rate': 0.1449996744492734, 'subsample': 0.5034720847246638, 'colsample\_bytree': 0.6725},  
 [1] 2025-05-31 02:47:44,073 Trial 16 finished with value: 0.0014692812636624484 and parameters: {'n\_estimators': 300, 'max\_depth': 4, 'learning\_rate': 0.1424233511842988, 'subsample': 0.6377104983596735, 'colsample\_bytree': 0.6893},  
 [1] 2025-05-31 02:47:44,234 Trial 17 finished with value: 0.001524736746497868 and parameters: {'n\_estimators': 299, 'max\_depth': 4, 'learning\_rate': 0.08704153816516558, 'subsample': 0.806545804404809, 'colsample\_bytree': 0.6737},  
 [1] 2025-05-31 02:47:44,383 Trial 18 finished with value: 0.00149386499508050 and parameters: {'n\_estimators': 231, 'max\_depth': 3, 'learning\_rate': 0.1294954582489878, 'subsample': 0.543111286204482, 'colsample\_bytree': 0.54416},  
 [1] 2025-05-31 02:47:44,533 Trial 19 finished with value: 0.001582340733682949 and parameters: {'n\_estimators': 267, 'max\_depth': 5, 'learning\_rate': 0.0897205285447001, 'subsample': 0.5659744670968535, 'colsample\_bytree': 0.59480},  
 [1] 2025-05-31 02:47:44,604 A new study created in memory with name: no-name~df5c17dc~3d66~47e~be5b~706f190904c  
 [1] 2025-05-31 02:47:44,798 Trial 0 finished with value: 0.001537332842024034 and parameters: {'n\_estimators': 261, 'max\_depth': 5, 'learning\_rate': 0.11574057300815908, 'subsample': 0.6889984314325244, 'colsample\_bytree': 0.65464},  
 [1] 2025-05-31 02:47:44,933 Trial 1 finished with value: 0.001547767574817556 and parameters: {'n\_estimators': 280, 'max\_depth': 7, 'learning\_rate': 0.01565776081949832, 'subsample': 0.9212111382403246, 'colsample\_bytree': 0.56850},  
 [1] 2025-05-31 02:47:45,000 Trial 2 finished with value: 0.00158756389710032 and parameters: {'n\_estimators': 63, 'max\_depth': 5, 'learning\_rate': 0.1865238195754273, 'subsample': 0.786780468513268, 'colsample\_bytree': 0.546441},  
 [1] 2025-05-31 02:47:45,229 Trial 3 finished with value: 0.001522933057020186 and parameters: {'n\_estimators': 68, 'max\_depth': 4, 'learning\_rate': 0.1851520205445375, 'subsample': 0.7689645223741809, 'colsample\_bytree': 0.542833},  
 [1] 2025-05-31 02:47:45,387 Trial 4 finished with value: 0.00154477373731762 and parameters: {'n\_estimators': 192, 'max\_depth': 6, 'learning\_rate': 0.04520066925889755, 'subsample': 0.8120740394701096, 'colsample\_bytree': 0.80098},  
 [1] 2025-05-31 02:47:45,388 Trial 5 finished with value: 0.001570445118545169 and parameters: {'n\_estimators': 182, 'max\_depth': 8, 'learning\_rate': 0.050862771126705834, 'subsample': 0.8727548337744643, 'colsample\_bytree': 0.6563},  
 [1] 2025-05-31 02:47:45,390 Trial 6 finished with value: 0.0015862719227715649 and parameters: {'n\_estimators': 57, 'max\_depth': 6, 'learning\_rate': 0.1759027150828768, 'subsample': 0.7498816209432704, 'colsample\_bytree': 0.891712},  
 [1] 2025-05-31 02:47:45,458 Trial 7 finished with value: 0.001596107858832968 and parameters: {'n\_estimators': 94, 'max\_depth': 8, 'learning\_rate': 0.05638592194363604, 'subsample': 0.839224965575936, 'colsample\_bytree': 0.8492188},  
 [1] 2025-05-31 02:47:45,546 Trial 8 finished with value: 0.00160828063073318 and parameters: {'n\_estimators': 137, 'max\_depth': 5, 'learning\_rate': 0.12602940211199671, 'subsample': 0.5806527039846077, 'colsample\_bytree': 0.800947},  
 [1] 2025-05-31 02:47:45,624 Trial 9 finished with value: 0.001519144520839516 and parameters: {'n\_estimators': 112, 'max\_depth': 4, 'learning\_rate': 0.14460920912216436, 'subsample': 0.70716284465464439, 'colsample\_bytree': 0.62645},  
 [1] 2025-05-31 02:47:45,792 Trial 10 finished with value: 0.001525294564684916 and parameters: {'n\_estimators': 192, 'max\_depth': 3, 'learning\_rate': 0.186380499707150258, 'subsample': 0.638004879565506, 'colsample\_bytree': 0.98347},  
 [1] 2025-05-31 02:47:45,899 Trial 11 finished with value: 0.0015446773731762 and parameters: {'n\_estimators': 119, 'max\_depth': 3, 'learning\_rate': 0.264680788933865, 'subsample': 0.9918159578774677, 'colsample\_bytree': 0.6521},  
 [1] 2025-05-31 02:47:46,016 Trial 12 finished with value: 0.00170104246171715 and parameters: {'n\_estimators': 135, 'max\_depth': 4, 'learning\_rate': 0.29363607686610194, 'subsample': 0.68104536922249463, 'colsample\_bytree': 0.50601},  
 [1] 2025-05-31 02:47:46,112 Trial 13 finished with value: 0.001677964634634624 and parameters: {'n\_estimators': 95, 'max\_depth': 4, 'learning\_rate': 0.07079970764634624, 'subsample': 0.718875207002833, 'colsample\_bytree': 0.597324},  
 [1] 2025-05-31 02:47:46,208 Trial 14 finished with value: 0.001561629553763553 and parameters: {'n\_estimators': 89, 'max\_depth': 4, 'learning\_rate': 0.08716201435830138, 'subsample': 0.839224965575936, 'colsample\_bytree': 0.7288316},  
 [1] 2025-05-31 02:47:46,327 Trial 15 finished with value: 0.0015436154208093458 and parameters: {'n\_estimators': 153, 'max\_depth': 3, 'learning\_rate': 0.0346192718717266, 'subsample': 0.515123048479857, 'colsample\_bytree': 0.5038},  
 [1] 2025-05-31 02:47:46,414 Trial 16 finished with value: 0.0015367140864864869 and parameters: {'n\_estimators': 51, 'max\_depth': 4, 'learning\_rate': 0.1823244319716444, 'subsample': 0.4748536170404451, 'colsample\_bytree': 0.716030},  
 [1] 2025-05-31 02:47:46,552 Trial 17 finished with value: 0.0015093702312682049 and parameters: {'n\_estimators': 215, 'max\_depth': 4, 'learning\_rate': 0.01088670161404598, 'subsample': 0.7474235874820527, 'colsample\_bytree': 0.602},  
 [1] 2025-05-31 02:47:46,559 Trial 18 finished with value: 0.00153576181050282 and parameters: {'n\_estimators': 110, 'max\_depth': 5, 'learning\_rate': 0.2491158414961263, 'subsample': 0.8889614356825022, 'colsample\_bytree': 0.68817},  
 [1] 2025-05-31 02:47:46,853 A new study created in memory with name: no-name~603b75c3~422a~119~309fbfd3  
 [1] 2025-05-31 02:47:46,965 Trial 1 finished with value: 0.002394539839731647 and parameters: {'n\_estimators': 60, 'max\_depth': 3, 'learning\_rate': 0.04827180653954975, 'subsample': 0.557278631844196, 'colsample\_bytree': 0.832782},  
 [1] 2025-05-31 02:47:47,073 Trial 2 finished with value: 0.0024149677909316263 and parameters: {'n\_estimators': 195, 'max\_depth': 6, 'learning\_rate': 0.0571957374451058, 'subsample': 0.5080026080535272, 'colsample\_bytree': 0.765964},  
 [1] 2025-05-31 02:47:47,179 Trial 3 finished with value: 0.00204775479668877 and parameters: {'n\_estimators': 177, 'max\_depth': 6, 'learning\_rate': 0.1633179321436508, 'subsample': 0.860811854501959, 'colsample\_bytree': 0.534326},  
 [1] 2025-05-31 02:47:47,278 Trial 4 finished with value: 0.00158756389710032 and parameters: {'n\_estimators': 63, 'max\_depth': 5, 'learning\_rate': 0.1865238195754273, 'subsample': 0.786780468513268, 'colsample\_bytree': 0.546441},  
 [1] 2025-05-31 02:47:47,365 Trial 5 finished with value: 0.001522933057020186 and parameters: {'n\_estimators': 192, 'max\_depth': 3, 'learning\_rate': 0.1851520205445375, 'subsample': 0.7689645223741809, 'colsample\_bytree': 0.542833},  
 [1] 2025-05-31 02:47:47,438 Trial 6 finished with value: 0.002047062683272316 and parameters: {'n\_estimators': 89, 'max\_depth': 4, 'learning\_rate': 0.14807120380696914, 'subsample': 0.7970955966660507, 'colsample\_bytree': 0.83723},  
 [1] 2025-05-31 02:47:47,500 Trial 7 finished with value: 0.002410244684062431708 and parameters: {'n\_estimators': 193, 'max\_depth': 7, 'learning\_rate': 0.170521670150924, 'subsample': 0.670523022433702833, 'colsample\_bytree': 0.594986},  
 [1] 2025-05-31 02:47:47,600 Trial 8 finished with value: 0.0023999430797871371 and parameters: {'n\_estimators': 114, 'max\_depth': 6, 'learning\_rate': 0.0341781308171263, 'subsample': 0.679712249414472, 'colsample\_bytree': 0.549986},  
 [1] 2025-05-31 02:47:47,714 Trial 9 finished with value: 0.002043638152890347 and parameters: {'n\_estimators': 67, 'max\_depth': 3, 'learning\_rate': 0.10150780653150927, 'subsample': 0.725103672172535, 'colsample\_bytree': 0.8784564},  
 [1] 2025-05-31 02:47:47,815 Trial 10 finished with value: 0.0023903535280822683 and parameters: {'n\_estimators': 53, 'max\_depth': 3, 'learning\_rate': 0.051837558460895655, 'subsample': 0.5056709895958959, 'colsample\_bytree': 0.59429},  
 [1] 2025-05-31 02:47:47,899 Trial 11 finished with value: 0.00239894550670309439 and parameters: {'n\_estimators': 291, 'max\_depth': 8, 'learning\_rate': 0.024629110674235272, 'subsample': 0.5959737212422627, 'colsample\_bytree': 0.99370},  
 [1] 2025-05-31 02:47:48,099 Trial 12 finished with value: 0.0020466975266158364 and parameters: {'n\_estimators': 112, 'max\_depth': 4, 'learning\_rate': 0.264680788933865, 'subsample': 0.8090519594377086, 'colsample\_bytree': 0.6268},  
 [1] 2025-05-31 02:47:49,439 Trial 13 finished with value: 0.002391775261352777 and parameters: {'n\_estimators': 296, 'max\_depth': 8, 'learning\_rate': 0.27483377116539874, 'subsample': 0.6156266176161581, 'colsample\_bytree': 0.9803},  
 [1] 2025-05-31 02:47:50,228 Trial 14 finished with value: 0.00239109345737186 and parameters: {'n\_estimators': 298, 'max\_depth': 8, 'learning\_rate': 0.298594963626621233, 'subsample': 0.6019563644697078, 'colsample\_bytree': 0.9983},  
 [1] 2025-05-31 02:47:50,427 Trial 15 finished with value: 0.002411959345737186 and parameters: {'n\_estimators': 193, 'max\_depth': 7, 'learning\_rate': 0.15754008033534715, 'subsample': 0.63895391862800313, 'colsample\_bytree': 0.9960},  
 [1] 2025-05-31 02:47:51,740 Trial 16 finished with value: 0.0024110515084822771 and parameters: {'n\_estimators': 221, 'max\_depth': 8, 'learning\_rate': 0.0278155239197492, 'subsample': 0.8899438266469742, 'colsample\_bytree': 0.8257542},  
 [1] 2025-05-31 02:47:52,677 Trial 17 finished with value: 0.0020446092062172399 and parameters: {'n\_estimators': 98, 'max\_depth': 6, 'learning\_rate': 0.134455787463670139, 'subsample': 0.7605577887701439, 'colsample\_bytree': 0.999057},  
 [1] 2025-05-31 02:47:52,712 Trial 18 finished with value: 0.00204752293052878378 and parameters: {'n\_estimators': 51, 'max\_depth': 3, 'learning\_rate': 0.1633179321436508, 'subsample': 0.5803678873075633, 'colsample\_bytree': 0.81006},  
 [1] 2025-05-31 02:47:52,155 Trial 19 finished with value: 0.0019723413904921734 and parameters: {'n\_estimators': 172, 'max\_depth': 7, 'learning\_rate': 0.112015151325136798, 'subsample': 0.93542005101475885, 'colsample\_bytree': 0.5132},  
 [1] 2025-05-31 02:47:52,255 Trial 20 finished with value: 0.001971191784357277 and parameters: {'n\_estimators': 101, 'max\_depth': 6, 'learning\_rate': 0.013782405540077595, 'subsample': 0.6868314024946986, 'colsample\_bytree': 0.9989},  
 [1] 2025-05-31 02:47:52,323 Trial 21 finished with value: 0.00197192625622667 and parameters: {'n\_estimators': 120, 'max\_depth': 4, 'learning\_rate': 0.1700620554668034, 'subsample': 0.7108403218214269, 'colsample\_bytree': 0.9666},  
 [1] 2025-05-31 02:47:53,129 Trial 22 finished with value: 0.00196874839836380138 and parameters: {'n\_estimators': 92, 'max\_depth': 5, 'learning\_rate': 0.1758402530365843, 'subsample': 0.623659625653853, 'colsample\_bytree': 0.999279},  
 [1] 2025-05-31 02:47:53,245 Trial 23 finished with value: 0.0019767390815706777 and parameters: {'n\_estimators': 140, 'max\_depth': 5, 'learning\_rate': 0.131977176273096, 'subsample': 0.6478207307584869, 'colsample\_bytree': 0.63823},  
 [1] 2025-05-31 02:47:53,427 Trial 24 finished with value: 0.00197453672954072394 and parameters: {'n\_estimators': 300, 'max\_depth': 5, 'learning\_rate': 0.2494350095895334, 'subsample': 0.7939050567075588, 'colsample\_bytree': 0.971810},  
 [1] 2025-05-31 02:47:53,526 Trial 25 finished with value: 0.00196464956916829983 and parameters: {'n\_estimators': 93, 'max\_depth': 4, 'learning\_rate': 0.043165498234968365, 'subsample': 0.52823796683386536, 'colsample\_bytree': 0.86109},  
 [1] 2025-05-31 02:47:53,763 Trial 26 finished with value: 0.00197513453987344 and parameters: {'n\_estimators': 141, 'max\_depth': 8, 'learning\_rate': 0.191658424058866008, 'subsample': 0.998773873349813, 'colsample\_bytree': 0.52681057},  
 [1] 2025-05-31 02:47:53,833 A new study created in memory with name: no-name~ef1lef9~8bc-43f1~a13~62938d68908590  
 [1] 2025-05-31 02:47:54,083 Trial 0 finished with value: 0.000807332488409013, 'subsample': 0.89174834679446, 'colsample\_bytree': 0.6049403},  
 [1] 2025-05-31 02:47:54,226 Trial 1 finished with value: 0.000812001963532865 and parameters: {'n\_estimators': 209, 'max\_depth': 4, 'learning\_rate': 0.24442765394047, 'subsample': 0.89519241397923, 'colsample\_bytree': 0.66337},  
 [1] 2025-05-31 02:47:54,3

```
[1] 2025-05-31 02:48:00,782] Trial 19 finished with value: 0.0006280709100324145 and parameters: {'n_estimators': 184, 'max_depth': 7, 'learning_rate': 0.1700935950633355, 'subsample': 0.6789043780790767, 'colsample_bytree': 0.94431}
[1] 2025-05-31 02:48:00,964] A new study created in memory with name: no-name-3ae922d5-ba93-4772-813b-8b765dd6839f
[1] 2025-05-31 02:48:01,222] Trial 0 finished with value: 0.00246318139650849 and parameters: {'n_estimators': 219, 'max_depth': 7, 'learning_rate': 0.14335477630120727, 'subsample': 0.564822730208623, 'colsample_bytree': 0.763395}
[1] 2025-05-31 02:48:01,942] Trial 1 finished with value: 0.00237952613054636 and parameters: {'n_estimators': 213, 'max_depth': 8, 'learning_rate': 0.14335477630120727, 'subsample': 0.883523332373643, 'colsample_bytree': 0.79917}
[1] 2025-05-31 02:48:03,071] Trial 2 finished with value: 0.00236047984763561 and parameters: {'n_estimators': 279, 'max_depth': 4, 'learning_rate': 0.11322171737938273, 'subsample': 0.44744279345645155, 'colsample_bytree': 0.98449}
[1] 2025-05-31 02:48:03,205] Trial 3 finished with value: 0.0023672380270996764 and parameters: {'n_estimators': 122, 'max_depth': 6, 'learning_rate': 0.0633279551404088, 'subsample': 0.9329714330152024, 'colsample_bytree': 0.85919}
[1] 2025-05-31 02:48:03,388] Trial 4 finished with value: 0.00238780815442667 and parameters: {'n_estimators': 189, 'max_depth': 6, 'learning_rate': 0.09919547300696463, 'subsample': 0.642885104546684, 'colsample_bytree': 0.832320}
[1] 2025-05-31 02:48:03,531] Trial 5 finished with value: 0.002353267605970362 and parameters: {'n_estimators': 79, 'max_depth': 7, 'learning_rate': 0.1549783783607905, 'subsample': 0.8386047778547271, 'colsample_bytree': 0.788869}
[1] 2025-05-31 02:48:03,751] Trial 6 finished with value: 0.002385186413009775 and parameters: {'n_estimators': 238, 'max_depth': 8, 'learning_rate': 0.04330971404822015, 'subsample': 0.7179332645603614, 'colsample_bytree': 0.64614}
[1] 2025-05-31 02:48:03,825] Trial 7 finished with value: 0.002368424605486232 and parameters: {'n_estimators': 61, 'max_depth': 8, 'learning_rate': 0.03105039902664264, 'subsample': 0.3305410266300248, 'colsample_bytree': 0.571352}
[1] 2025-05-31 02:48:04,048] Trial 8 finished with value: 0.0023805103973682752 and parameters: {'n_estimators': 231, 'max_depth': 8, 'learning_rate': 0.0104847224146465648, 'subsample': 0.6102952849908955, 'colsample_bytree': 0.60362}
[1] 2025-05-31 02:48:04,123] Trial 9 finished with value: 0.00236768036917422 and parameters: {'n_estimators': 77, 'max_depth': 3, 'learning_rate': 0.01696570089223778, 'subsample': 0.712669253735798, 'colsample_bytree': 0.763368}
[1] 2025-05-31 02:48:04,237] Trial 10 finished with value: 0.00232450685526462 and parameters: {'n_estimators': 146, 'max_depth': 5, 'learning_rate': 0.23072652656260926, 'subsample': 0.5203985327605253, 'colsample_bytree': 0.501010}
[1] 2025-05-31 02:48:04,349] Trial 11 finished with value: 0.0023672480586507447 and parameters: {'n_estimators': 144, 'max_depth': 5, 'learning_rate': 0.023072652656260926, 'subsample': 0.5106047978965123, 'colsample_bytree': 0.50007}
[1] 2025-05-31 02:48:04,504] Trial 12 finished with value: 0.0023877681153056155 and parameters: {'n_estimators': 154, 'max_depth': 5, 'learning_rate': 0.28039458241790904, 'subsample': 0.5020190813378966, 'colsample_bytree': 0.6599}
[1] 2025-05-31 02:48:04,662] Trial 13 finished with value: 0.002401892479282646 and parameters: {'n_estimators': 291, 'max_depth': 7, 'learning_rate': 0.183692391079564752, 'subsample': 0.8768606842685837, 'colsample_bytree': 0.7081}
[1] 2025-05-31 02:48:04,769] Trial 14 finished with value: 0.00240817250449467 and parameters: {'n_estimators': 117, 'max_depth': 4, 'learning_rate': 0.12070295263085244, 'subsample': 0.566625405608947, 'colsample_bytree': 0.92653}
[1] 2025-05-31 02:48:04,898] Trial 15 finished with value: 0.0023608128624879177 and parameters: {'n_estimators': 185, 'max_depth': 7, 'learning_rate': 0.02801069617095826, 'subsample': 0.6481025424567504, 'colsample_bytree': 0.70354}
[1] 2025-05-31 02:48:05,057] Trial 16 finished with value: 0.0023606950195955260 and parameters: {'n_estimators': 250, 'max_depth': 6, 'learning_rate': 0.089020226993155698, 'subsample': 0.821161107654064, 'colsample_bytree': 0.53509}
[1] 2025-05-31 02:48:05,184] Trial 17 finished with value: 0.0023176218649007 and parameters: {'n_estimators': 164, 'max_depth': 4, 'learning_rate': 0.18851569846867183, 'subsample': 0.549728532306298, 'colsample_bytree': 0.866629}
[1] 2025-05-31 02:48:05,290] Trial 18 finished with value: 0.0023606910912852 and parameters: {'n_estimators': 112, 'max_depth': 3, 'learning_rate': 0.190862141261901, 'subsample': 0.9221108447980275, 'colsample_bytree': 0.8931817}
[1] 2025-05-31 02:48:05,412] Trial 19 finished with value: 0.002342946531066625 and parameters: {'n_estimators': 159, 'max_depth': 4, 'learning_rate': 0.076679731982349, 'subsample': 0.66731201207863, 'colsample_bytree': 0.956890}
[1] 2025-05-31 02:48:05,563] A new study created in memory with name: no-name-6cdfa07c-8e9f-4626-8837-e00af128f9p
[1] 2025-05-31 02:48:05,734] Trial 0 finished with value: 0.001793285175927377 and parameters: {'n_estimators': 291, 'max_depth': 7, 'learning_rate': 0.02149761491290943, 'subsample': 0.5036565033079126, 'colsample_bytree': 0.8582}
[1] 2025-05-31 02:48:05,853] Trial 1 finished with value: 0.0018005844874064959 and parameters: {'n_estimators': 137, 'max_depth': 8, 'learning_rate': 0.0472897831195469, 'subsample': 0.6916741017390067, 'colsample_bytree': 0.85817}
[1] 2025-05-31 02:48:05,916] Trial 2 finished with value: 0.001896750446236324 and parameters: {'n_estimators': 104, 'max_depth': 4, 'learning_rate': 0.2429396326367818, 'subsample': 0.6595273695735175, 'colsample_bytree': 0.950223}
[1] 2025-05-31 02:48:06,073] Trial 3 finished with value: 0.0018005595332218102 and parameters: {'n_estimators': 293, 'max_depth': 7, 'learning_rate': 0.194580188316252, 'subsample': 0.6052681657471753, 'colsample_bytree': 0.54887}
[1] 2025-05-31 02:48:06,156] Trial 4 finished with value: 0.001757542613655912 and parameters: {'n_estimators': 111, 'max_depth': 6, 'learning_rate': 0.09320625161024684, 'subsample': 0.729780313451873, 'colsample_bytree': 0.529844}
[1] 2025-05-31 02:48:06,258] Trial 5 finished with value: 0.00180611907313339 and parameters: {'n_estimators': 136, 'max_depth': 4, 'learning_rate': 0.05217384746359873, 'subsample': 0.5045498716111364, 'colsample_bytree': 0.755261}
[1] 2025-05-31 02:48:06,391] Trial 6 finished with value: 0.001802871073028135 and parameters: {'n_estimators': 240, 'max_depth': 5, 'learning_rate': 0.01333935020023088, 'subsample': 0.684255316010116, 'colsample_bytree': 0.8104}
[1] 2025-05-31 02:48:06,492] Trial 7 finished with value: 0.0018097340528183854 and parameters: {'n_estimators': 170, 'max_depth': 8, 'learning_rate': 0.01272416540345695, 'subsample': 0.678805471697869, 'colsample_bytree': 0.70703}
[1] 2025-05-31 02:48:06,583] Trial 8 finished with value: 0.001788330033065875 and parameters: {'n_estimators': 59, 'max_depth': 5, 'learning_rate': 0.1908650981518325, 'subsample': 0.8138246214602933, 'colsample_bytree': 0.7574616}
[1] 2025-05-31 02:48:06,691] Trial 9 finished with value: 0.001806231419662888 and parameters: {'n_estimators': 183, 'max_depth': 6, 'learning_rate': 0.21349552120012883, 'subsample': 0.974292523253098, 'colsample_bytree': 0.80097}
[1] 2025-05-31 02:48:06,775] Trial 10 finished with value: 0.001805932161490986 and parameters: {'n_estimators': 57, 'max_depth': 3, 'learning_rate': 0.1020965930483609, 'subsample': 0.8341957035496371, 'colsample_bytree': 0.50064}
[1] 2025-05-31 02:48:06,858] Trial 11 finished with value: 0.00184365706317077 and parameters: {'n_estimators': 51, 'max_depth': 5, 'learning_rate': 0.158648538462815, 'subsample': 0.821341204638304, 'colsample_bytree': 0.655363}
[1] 2025-05-31 02:48:06,974] Trial 12 finished with value: 0.001768602618494274 and parameters: {'n_estimators': 90, 'max_depth': 7, 'learning_rate': 0.2977155977244, 'subsample': 0.81610812076584, 'colsample_bytree': 0.62227754}
[1] 2025-05-31 02:48:07,077] Trial 13 finished with value: 0.001812089729456794 and parameters: {'n_estimators': 104, 'max_depth': 7, 'learning_rate': 0.028401232796168, 'subsample': 0.733865879529946, 'colsample_bytree': 0.601910}
[1] 2025-05-31 02:48:07,181] Trial 14 finished with value: 0.0017943213247312367 and parameters: {'n_estimators': 96, 'max_depth': 7, 'learning_rate': 0.0177660528232988, 'subsample': 0.7595497038599452, 'colsample_bytree': 0.60959}
[1] 2025-05-31 02:48:07,297] Trial 15 finished with value: 0.00178840246029274 and parameters: {'n_estimators': 147, 'max_depth': 6, 'learning_rate': 0.2700095869252559, 'subsample': 0.9007084268943771, 'colsample_bytree': 0.5068246}
[1] 2025-05-31 02:48:07,431] Trial 16 finished with value: 0.0018595273343304 and parameters: {'n_estimators': 209, 'max_depth': 7, 'learning_rate': 0.1512608294599428, 'subsample': 0.7516183014736092, 'colsample_bytree': 0.65788}
[1] 2025-05-31 02:48:07,528] Trial 17 finished with value: 0.0018098279184136109 and parameters: {'n_estimators': 95, 'max_depth': 8, 'learning_rate': 0.07303639585926351, 'subsample': 0.8712299855210797, 'colsample_bytree': 0.56631}
[1] 2025-05-31 02:48:07,634] Trial 18 finished with value: 0.001805372636894296 and parameters: {'n_estimators': 88, 'max_depth': 6, 'learning_rate': 0.1639828609403721, 'subsample': 0.591999192567043, 'colsample_bytree': 0.66680}
[1] 2025-05-31 02:48:07,742] Trial 19 finished with value: 0.00180572646634289 and parameters: {'n_estimators': 125, 'max_depth': 7, 'learning_rate': 0.169158997136783, 'subsample': 0.7326752029074837, 'colsample_bytree': 0.58885}
[1] 2025-05-31 02:48:07,839] A new study created in memory with name: no-name-70d2487b-4c5f-8bd5-bc7924a4b7c3
[1] 2025-05-31 02:48:08,099] Trial 0 finished with value: 0.003228187219759127 and parameters: {'n_estimators': 289, 'max_depth': 3, 'learning_rate': 0.14934159004522715, 'subsample': 0.5329593525573235, 'colsample_bytree': 0.942738}
[1] 2025-05-31 02:48:08,111] Trial 1 finished with value: 0.00328106363518008 and parameters: {'n_estimators': 185, 'max_depth': 7, 'learning_rate': 0.13706798760762669, 'subsample': 0.78147526789704, 'colsample_bytree': 0.574603}
[1] 2025-05-31 02:48:08,239] Trial 2 finished with value: 0.00331204466194077 and parameters: {'n_estimators': 243, 'max_depth': 7, 'learning_rate': 0.0934475637638209, 'subsample': 0.7032871704083383, 'colsample_bytree': 0.726648}
[1] 2025-05-31 02:48:08,305] Trial 3 finished with value: 0.003261687749069693 and parameters: {'n_estimators': 59, 'max_depth': 8, 'learning_rate': 0.02198287039790955, 'subsample': 0.8917366555620205, 'colsample_bytree': 0.584088}
[1] 2025-05-31 02:48:08,410] Trial 4 finished with value: 0.0032542005119530744 and parameters: {'n_estimators': 186, 'max_depth': 3, 'learning_rate': 0.0285013972437356, 'subsample': 0.6663993300290472, 'colsample_bytree': 0.5490}
[1] 2025-05-31 02:48:08,546] Trial 5 finished with value: 0.003261677594183181 and parameters: {'n_estimators': 281, 'max_depth': 7, 'learning_rate': 0.037798412143736956, 'subsample': 0.9537925930018674, 'colsample_bytree': 0.57122}
[1] 2025-05-31 02:48:08,660] Trial 6 finished with value: 0.00329541919816758 and parameters: {'n_estimators': 120, 'max_depth': 6, 'learning_rate': 0.15065000767352428, 'subsample': 0.7395564856955691, 'colsample_bytree': 0.878317}
[1] 2025-05-31 02:48:08,767] Trial 7 finished with value: 0.003256819552537146 and parameters: {'n_estimators': 164, 'max_depth': 8, 'learning_rate': 0.29005281457681386, 'subsample': 0.5592199709156219, 'colsample_bytree': 0.83539}
[1] 2025-05-31 02:48:08,856] Trial 8 finished with value: 0.003287670305697229 and parameters: {'n_estimators': 120, 'max_depth': 7, 'learning_rate': 0.0864569894751663, 'subsample': 0.911472609434004001, 'colsample_bytree': 0.754446}
[1] 2025-05-31 02:48:08,960] Trial 9 finished with value: 0.003267978636867035 and parameters: {'n_estimators': 150, 'max_depth': 5, 'learning_rate': 0.013153525625675617, 'subsample': 0.7108866525175583, 'colsample_bytree': 0.95739}
[1] 2025-05-31 02:48:09,128] Trial 10 finished with value: 0.00329054035120190767 and parameters: {'n_estimators': 295, 'max_depth': 3, 'learning_rate': 0.13706798760762669, 'subsample': 0.5186537719283273, 'colsample_bytree': 0.95683}
[1] 2025-05-31 02:48:09,272] Trial 11 finished with value: 0.003204231050273015 and parameters: {'n_estimators': 214, 'max_depth': 3, 'learning_rate': 0.03933733918166845, 'subsample': 0.6359337310339348, 'colsample_bytree': 0.50430}
[1] 2025-05-31 02:48:09,415] Trial 12 finished with value: 0.003273655718726644 and parameters: {'n_estimators': 234, 'max_depth': 4, 'learning_rate': 0.0520784804193723, 'subsample': 0.59715083471370279, 'colsample_bytree': 0.66224}
[1] 2025-05-31 02:48:09,546] Trial 13 finished with value: 0.0032651677594183181 and parameters: {'n_estimators': 281, 'max_depth': 7, 'learning_rate': 0.037798412143736956, 'subsample': 0.971192656856955691, 'colsample_bytree': 0.878317}
[1] 2025-05-31 02:48:09,663] Trial 14 finished with value: 0.0032646570735730002 and parameters: {'n_estimators': 217, 'max_depth': 7, 'learning_rate': 0.0791666628072589, 'subsample': 0.750995492795774, 'colsample_bytree': 0.66611}
[1] 2025-05-31 02:48:09,892] Trial 15 finished with value: 0.0032932093553924087 and parameters: {'n_estimators': 270, 'max_depth': 3, 'learning_rate': 0.051729825761729458, 'subsample': 0.638584938313989, 'colsample_bytree': 0.9903}
[1] 2025-05-31 02:48:10,043] Trial 16 finished with value: 0.003262907294484087 and parameters: {'n_estimators': 262, 'max_depth': 5, 'learning_rate': 0.025245678589333, 'subsample': 0.57247446996747447, 'colsample_bytree': 0.505}
[1] 2025-05-31 02:48:10,178] Trial 17 finished with value: 0.003288958169999326 and parameters: {'n_estimators': 207, 'max_depth': 4, 'learning_rate': 0.066462903051404085, 'subsample': 0.6626462903051404085, 'colsample_bytree': 0.80066}
[1] 2025-05-31 02:48:10,338] Trial 18 finished with value: 0.00326576139297094 and parameters: {'n_estimators': 295, 'max_depth': 3, 'learning_rate': 0.0375405911559881709, 'subsample': 0.818847530730477, 'colsample_bytree': 0.917364}
[1] 2025-05-31 02:48:10,590] A new study created in memory with name: no-name-34f59c4-de27-439c-95be-e478ba0923
[1] 2025-05-31 02:48:10,730] Trial 0 finished with value: 0.0029332191982020 and parameters: {'n_estimators': 58, 'max_depth': 8, 'learning_rate': 0.07008048448087245, 'subsample': 0.5345598895342912, 'colsample_bytree': 0.653324}
[1] 2025-05-31 02:48:10,866] Trial 1 finished with value: 0.0028471580241263 and parameters: {'n_estimators': 87, 'max_depth': 4, 'learning_rate': 0.0219410188320126, 'subsample': 0.894607079368855, 'colsample_bytree': 0.665940}
[1] 2025-05-31 02:48:10,960] Trial 2 finished with value: 0.002393905451231127 and parameters: {'n_estimators': 217, 'max_depth': 3, 'learning_rate': 0.0404052627375001, 'subsample': 0.967894033209552, 'colsample_bytree': 0.8088137}
[1] 2025-05-31 02:48:11,060] Trial 3 finished with value: 0.002393672652675773 and parameters: {'n_estimators': 59, 'max_depth': 8, 'learning_rate': 0.02198287039790955, 'subsample': 0.518934751282377, 'colsample_bytree': 0.62310}
[1] 2025-05-31 02:48:11,187] Trial 4 finished with value: 0.00287672131719752 and parameters: {'n_estimators': 223, 'max_depth': 3, 'learning_rate': 0.02414763036945288, 'subsample': 0.82015682235472547, 'colsample_bytree': 0.60684}
[1] 2025-05-31 02:48:11,298] Trial 5 finished with value: 0.003021375199919995 and parameters: {'n_estimators': 199, 'max_depth': 8, 'learning_rate': 0.0368041018605801, 'subsample': 0.5093022441246642664, 'colsample_bytree': 0.816024}
[1] 2025-05-31 02:48:11,393] Trial 6 finished with value: 0.0030177681315389 and parameters: {'n_estimators': 147, 'max_depth': 5, 'learning_rate': 0.14227689161344, 'subsample': 0.5438491595265319, 'colsample_bytree': 0.980495}
[1] 2025-05-31 02:48:11,474] Trial 7 finished with value: 0.002866049189574087 and parameters: {'n_estimators': 62, 'max_depth': 4, 'learning_rate': 0.04358291975946274, 'subsample': 0.747844226667814, 'colsample_bytree': 0.724049}
[1] 2025-05-31 02:48:11,614] Trial 9 finished with value: 0.002855878747897676 and parameters: {'n_estimators': 58, 'max_depth': 3, 'learning_rate': 0.0168301571731734, 'subsample': 0.7376960804283781, 'colsample_bytree': 0.777533}
[1] 2025-05-31 02:48:11,814] Trial 10 finished with value: 0.00282364818931972 and parameters: {'n_estimators': 297, 'max_depth': 6, 'learning_rate': 0.01010403619489561, 'subsample': 0.5950951357761371, 'colsample_bytree': 0.5217}
[1] 2025-05-31 02:48:11,978] Trial 11 finished with value: 0.0028072870157134564 and parameters: {'n_estimators': 298, 'max_depth': 6, 'learning_rate': 0.010442763482977006, 'subsample': 0.995635845842089, 'colsample_bytree': 0.507}
[1] 2025-05-31 02:48:12,146] Trial 12 finished with value: 0.002828197355535097 and parameters: {'n_estimators': 297, 'max_depth': 6, 'learning_rate': 0.01002164368781283, 'subsample': 0.98697324884215, 'colsample_bytree': 0.5157}
[1] 2025-05-31 02:48:12,296] Trial 13 finished with value: 0.002900799409593437 and parameters: {'n_estimators': 262, 'max_depth': 6, 'learning_rate': 0.010634878555309, 'subsample': 0.67436461934463661, 'colsample_bytree': 0.5031}
[1] 2025-05-31 02:48:12,455] Trial 14 finished with value: 0.002826540520160421846 and parameters: {'n_estimators': 299, 'max_depth': 7, 'learning_rate': 0.01626696627694793, 'subsample': 0.997887344223116, 'colsample_bytree': 0.5602}
[1] 2025-05-31 02:48:12,601] Trial 15 finished with value: 0.002847249587723546 and parameters: {'n_estimators': 251, 'max_depth': 7, 'learning_rate': 0.01803809302015018, 'subsample': 0.8972739479488364, 'colsample_bytree': 0.5799}
[1] 2025-05-31 02:48:12,711] Trial 16 finished with value: 0.00289787924823682 and parameters: {'n_estimators': 131, 'max_depth': 7, 'learning_rate': 0.029700710926600794, 'subsample': 0.68447290993652294, 'colsample_bytree': 0.567094}
[1] 2025-05-31 02:48:12,855] Trial 17 finished with value: 0.0028373275535237 and parameters: {'n_estimators': 175, 'max_depth': 7, 'learning_rate': 0.0145262461318436, 'subsample': 0.9362947446057407, 'colsample_bytree': 0.7196}
[1] 2025-05-31 02:48:12,999] Trial 18 finished with value: 0.00285993972144557 and parameters: {'n_estimators': 239, 'max_depth': 5, '
```

```
[1] 2025-05-31 02:48:20.093 Trial 0 finished with value: 0.003201060000000000 and parameters: {'n_estimators': 240, 'max_depth': 4, 'learning_rate': 0.02400101400000000, 'subsample': 0.75049937169263, 'colsample_bytree': 0.740028}
[1] 2025-05-31 02:48:20.799 Trial 6 finished with value: 0.0032022923986027166 and parameters: {'n_estimators': 284, 'max_depth': 5, 'learning_rate': 0.025531761413350427, 'subsample': 0.859137971002661, 'colsample_bytree': 0.84469
[1] 2025-05-31 02:48:20.915 Trial 7 finished with value: 0.00320485759920621736 and parameters: {'n_estimators': 216, 'max_depth': 6, 'learning_rate': 0.010956188232955761, 'subsample': 0.7224958086247895, 'colsample_bytree': 0.5733
[1] 2025-05-31 02:48:20.987 Trial 8 finished with value: 0.003204597771620484 and parameters: {'n_estimators': 85, 'max_depth': 3, 'learning_rate': 0.03180919413586276, 'subsample': 0.6422151191862138, 'colsample_bytree': 0.561583
[1] 2025-05-31 02:48:21.090 Trial 9 finished with value: 0.0032043599320626297 and parameters: {'n_estimators': 100, 'max_depth': 8, 'learning_rate': 0.21549838367601004, 'subsample': 0.8639600807535284, 'colsample_bytree': 0.56873
[1] 2025-05-31 02:48:21.243 Trial 10 finished with value: 0.003202115187057519 and parameters: {'n_estimators': 252, 'max_depth': 4, 'learning_rate': 0.08576165332636682, 'subsample': 0.9964721233929005, 'colsample_bytree': 0.99562
[1] 2025-05-31 02:48:21.410 Trial 11 finished with value: 0.003202279519728766 and parameters: {'n_estimators': 294, 'max_depth': 3, 'learning_rate': 0.03975727876359032, 'subsample': 0.9960018757638028, 'colsample_bytree': 0.73199
[1] 2025-05-31 02:48:21.552 Trial 12 finished with value: 0.003198955867598951 and parameters: {'n_estimators': 245, 'max_depth': 4, 'learning_rate': 0.07552210932343803, 'subsample': 0.702131905914507, 'colsample_bytree': 0.723986
[1] 2025-05-31 02:48:21.702 Trial 13 finished with value: 0.003198634216973443 and parameters: {'n_estimators': 245, 'max_depth': 4, 'learning_rate': 0.08508337916423525, 'subsample': 0.6934657229691662, 'colsample_bytree': 0.6819
[1] 2025-05-31 02:48:21.825 Trial 14 finished with value: 0.003210979603188167 and parameters: {'n_estimators': 155, 'max_depth': 4, 'learning_rate': 0.08481779574417835, 'subsample': 0.657864488531743, 'colsample_bytree': 0.664370
[1] 2025-05-31 02:48:21.969 Trial 15 finished with value: 0.003203189294386349 and parameters: {'n_estimators': 238, 'max_depth': 5, 'learning_rate': 0.12234858447480572, 'subsample': 0.51094883560919, 'colsample_bytree': 0.671366
[1] 2025-05-31 02:48:22.091 Trial 16 finished with value: 0.003207221132747995 and parameters: {'n_estimators': 139, 'max_depth': 4, 'learning_rate': 0.06747640466651468, 'subsample': 0.6568140658557786, 'colsample_bytree': 0.66745
[1] 2025-05-31 02:48:22.223 Trial 17 finished with value: 0.0032237711722717007 and parameters: {'n_estimators': 184, 'max_depth': 5, 'learning_rate': 0.14748576008952596, 'subsample': 0.6043312892059589, 'colsample_bytree': 0.7308
[1] 2025-05-31 02:48:22.363 Trial 18 finished with value: 0.003206291643475286 and parameters: {'n_estimators': 218, 'max_depth': 3, 'learning_rate': 0.05179349984931969, 'subsample': 0.7036908983078465, 'colsample_bytree': 0.63486
[1] 2025-05-31 02:48:22.510 Trial 19 finished with value: 0.0032045105830804586 and parameters: {'n_estimators': 257, 'max_depth': 6, 'learning_rate': 0.15958620795615802, 'subsample': 0.53469647278269, 'colsample_bytree': 0.94594
[1] 2025-05-31 02:48:22.644 A new study created in memory with name: no-name-feb0eb9f-bfd1-4599-ad38-5e0267edaa82
[1] 2025-05-31 02:48:22.809 Trial 0 finished with value: 0.003980062924775745 and parameters: {'n_estimators': 270, 'max_depth': 7, 'learning_rate': 0.25315582790202684, 'subsample': 0.7440894575641401, 'colsample_bytree': 0.782959
[1] 2025-05-31 02:48:22.945 Trial 1 finished with value: 0.003986931912578826 and parameters: {'n_estimators': 277, 'max_depth': 8, 'learning_rate': 0.023404625854002138, 'subsample': 0.5297118777890013, 'colsample_bytree': 0.64555
[1] 2025-05-31 02:48:23.072 Trial 2 finished with value: 0.004000367418787733 and parameters: {'n_estimators': 240, 'max_depth': 8, 'learning_rate': 0.10236441297534259, 'subsample': 0.6680942744408618, 'colsample_bytree': 0.842152
[1] 2025-05-31 02:48:23.219 Trial 3 finished with value: 0.003989341013422905 and parameters: {'n_estimators': 182, 'max_depth': 6, 'learning_rate': 0.29164357817623465, 'subsample': 0.5011908787822487, 'colsample_bytree': 0.817243
[1] 2025-05-31 02:48:23.323 Trial 4 finished with value: 0.0039904315983245965 and parameters: {'n_estimators': 172, 'max_depth': 5, 'learning_rate': 0.29212916181569344, 'subsample': 0.6497526706776809, 'colsample_bytree': 0.68526
[1] 2025-05-31 02:48:23.388 Trial 5 finished with value: 0.003991432791789066 and parameters: {'n_estimators': 51, 'max_depth': 7, 'learning_rate': 0.18872342048829788, 'subsample': 0.5678819259844423, 'colsample_bytree': 0.937709
[1] 2025-05-31 02:48:23.520 Trial 6 finished with value: 0.00398279050521792 and parameters: {'n_estimators': 295, 'max_depth': 4, 'learning_rate': 0.043996917011101394, 'subsample': 0.6606160602927108, 'colsample_bytree': 0.996602
```

## Step 10 : Robust Execution & Risk Modeling

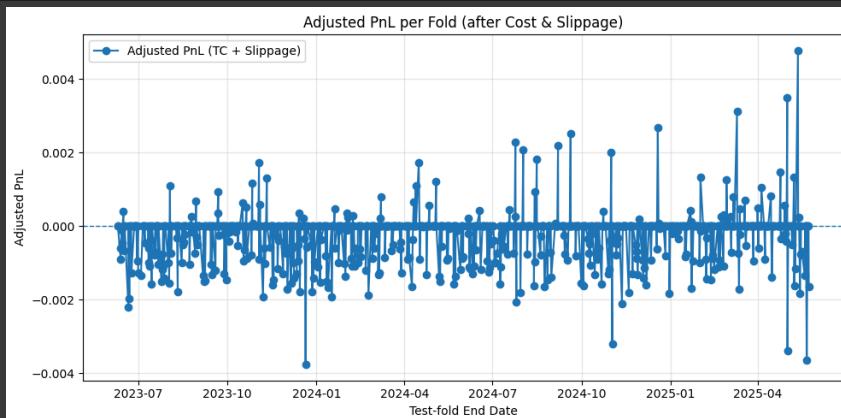
1. Step 10A : Transaction Cost & Slippage Modeling
2. Step 10B : Market Impact Modeling (Almgren–Chriss)

### ✓ Step 10A : Transaction Cost & Slippage Modeling

```

1 # (plug-and-play after Step 9B - expects df_metrics_optuna in memory)
2
3 import numpy as np
4 import pandas as pd
5 import matplotlib.pyplot as plt
6
7 def run_execution_step_10a(
8     df_metrics: pd.DataFrame,
9     tc_perc: float = 0.001, # 0.10 % per executed trade
10    slip_mean: float = 0.0005, # 0.05 % mean slippage
11    slip_std: float = 0.0002, # slippage volatility
12    random_state: int = 42
13 ) -> pd.DataFrame:
14     """
15     Add transaction-cost & slippage-adjusted PnL to walk-forward metrics.
16
17     Expected columns in df_metrics:
18         end (datetime) · avg_pnl · directional_accuracy · num_trades (optional)
19     Returns the same frame with extra cols:
20         tc_cost · slippage_cost · adj_pnl
21     """
22     df = df_metrics.copy().reset_index(drop=True)
23
24     # Sanity checks
25     if not {'avg_pnl', 'directional_accuracy'}.issubset(df.columns):
26         raise ValueError('df_metrics must contain "avg_pnl" and "directional_accuracy"')
27
28     # If num_trades is missing, assume one trade per fold
29     if 'num_trades' not in df.columns:
30         df['num_trades'] = 1
31
32     # Vectorised cost generation
33     rng = np.random.default_rng(random_state)
34     df['tc_cost'] = tc_perc * df['num_trades']
35     df['slippage_cost'] = rng.normal(slip_mean, slip_std, size=len(df)) * df['num_trades']
36
37     # Apply only if strategy beats coin-flip
38     mask = df['directional_accuracy'] > 0.50
39     df['adj_pnl'] = 0.0
40     df.loc[mask, 'adj_pnl'] = (
41         df.loc[mask, 'avg_pnl']
42         - df.loc[mask, 'tc_cost']
43         - df.loc[mask, 'slippage_cost']
44     )
45
46     return df
47
48 # run the execution adjustment
49 try:
50     df_metrics_exec = run_execution_step_10a(df_metrics_optuna)
51 except NameError as e:
52     raise RuntimeError(
53         "df_metrics_optuna is not defined - execute Step 9B first"
54         "or reload its output before running Step 10A."
55     ) from e
56
57 # visualise fold-level impact
58 plt.figure(figsize=(10, 5))
59 plt.plot(
60     df_metrics_exec['end'],
61     df_metrics_exec['adj_pnl'],
62     label='Adjusted PnL (TC + Slippage)',
63     marker='o'
64 )
65 plt.axhline(0, linestyle='--', linewidth=1)
66 plt.title("Adjusted PnL per Fold (after Cost & Slippage)")
67 plt.xlabel("Test-fold End Date")
68 plt.ylabel("Adjusted PnL")
69 plt.grid(alpha=0.3)
70 plt.legend()
71 plt.tight_layout()
72 plt.show()
73
74 # (optional) quick comparison
75 summary = df_metrics_exec[['avg_pnl', 'adj_pnl']].describe().loc[['mean', 'std']]
76 print("PnL summary (before vs after costs)\n", summary)

```



PnL summary (before vs after costs)  
 avg\_pnl adj\_pnl  
 mean 0.000223 -0.000214

### ✓ Step 10B : Market Impact Modeling (Almgren–Chriss)

```

1 # MARKET-IMPACT MODEL (Almgren–Chriss 2001)
2 # Full, self-contained implementation + demo
3 import numpy as np

```

```

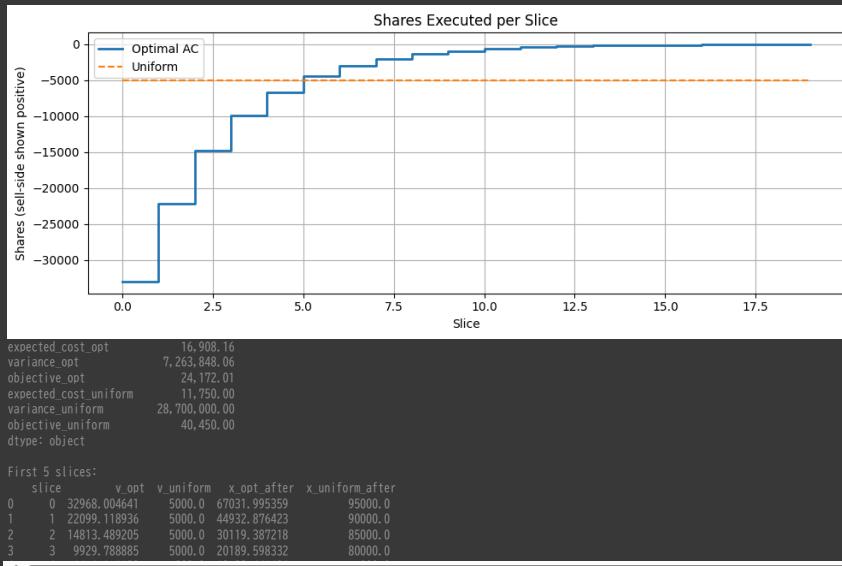
4 Import pandas as pd
5 import matplotlib.pyplot as plt
6
7 # CORE HELPERS
8 def _almgren_chriss_schedule(
9     X: float,
10    N: int,
11    sigma: float,
12    eta: float,
13    lam: float,
14    dt: float = 1.0,
15):
16    """
17        Discrete-time optimal liquidation schedule (Almgren-Chriss eq. 17).
18
19    Parameters
20    -----
21    X : total shares to trade (positive for sell order)
22    N : number of child-orders / time-slices
23    sigma : daily (or slice-level) volatility as decimal (e.g. 0.02)
24    eta : temporary-impact parameter
25    lam : risk-aversion parameter ( $\lambda \rightarrow 0 \Rightarrow$  uniform schedule)
26    dt : length of each time slice (same units as  $\sigma$ )
27
28    Returns
29    -----
30    x : np.ndarray, len N+1 - remaining inventory after each slice
31    v : np.ndarray, len N - trade size in each slice (negative = sell)
32    """
33    kappa = np.sqrt(lam * sigma ** 2 / eta)      # liquidation urgency
34
35    if np.isclose(kappa, 0.0):                    #  $\lambda \rightarrow 0 \Rightarrow$  uniform
36        x = X * (1 - np.arange(N + 1)) / N
37    else:
38        sinh_KN = np.sinh(kappa * N)
39        x = X * np.sinh(kappa * (N - np.arange(N + 1))) / sinh_KN
40
41    v = np.diff(-x)                                # trades (negative for sells)
42
43    return x, v
44
45 def _cost_variance(v, sigma, eta, gamma, dt):
46    """
47        Expected cost & variance of a schedule per Almgren-Chriss.
48
49    Parameters
50    -----
51    v : np.ndarray of trade sizes (negative if selling)
52    sigma : volatility (per slice)
53    eta : temporary-impact parameter
54    gamma : permanent-impact parameter
55    dt : slice length
56
57    Returns
58    -----
59    exp_cost, var_cost
60    """
61    v_ = np.asarray(v)
62    x_c = np.cumsum(v_[:-1])[:-1]                  # inventory *before* each trade
63    exp_cost = np.sum(eta * v_*v_*2 + gamma * x_c*v)
64    var_cost = sigma**2 * dt * np.sum(x_c**2)
65
66    return exp_cost, var_cost
67
68 # 2) PUBLIC WRAPPER
69 def run_market_impact_step_10b(
70     X: float,
71     horizon_slices: int,
72     sigma: float = 0.02,
73     eta: float = 2.5e-6,
74     gamma: float = 2.0e-6,
75     lam: float = 1e-4,
76     dt: float = 1.0,
77     plot: bool = True,
78 ):
79    """
80        Build optimal (Almgren-Chriss) vs uniform schedules *and* compute:
81        • expected cost
82        • cost variance
83        •  $\lambda$ -weighted objective  $C + \lambda \cdot \text{Var}$ 
84
85    Returns
86    -----
87    df_sched : pd.DataFrame
88        slice-level details for both schedules
89    metrics : dict
90        summary numbers for "optimal" and "uniform"
91
92    # schedules
93    x_opt, v_opt = _almgren_chriss_schedule(X, horizon_slices, sigma, eta, lam, dt)
94    x_uni, v_uni = _almgren_chriss_schedule(X, horizon_slices, sigma, eta, 0.0, dt)
95
96    # analytics
97    ec_opt, var_opt = _cost_variance(v_opt, sigma, eta, gamma, dt)
98    ec_uni, var_uni = _cost_variance(v_uni, sigma, eta, gamma, dt)
99
100   metrics = {
101       "expected_cost_opt": ec_opt,
102       "variance_opt": var_opt,
103       "objective_opt": ec_opt + lam * var_opt,
104       "expected_cost_uniform": ec_uni,
105       "variance_uniform": var_uni,
106       "objective_uniform": ec_uni + lam * var_uni,
107   }
108
109   df_sched = pd.DataFrame({
110       "slice": np.arange(horizon_slices),
111       "v_opt": v_opt,
112       "v_uniform": v_uni,
113       "x_opt_after": x_opt[1:],           # inventory after executing slice
114       "x_uniform_after": x_uni[1:],
115   })
116
117   # optional diagnostics plot
118   if plot:
119       plt.figure(figsize=(10, 4))
120       plt.step(df_sched["slice"], -df_sched["v_opt"], where="post",
121                 label="Optimal AC", linewidth=2)
122       plt.step(df_sched["slice"], -df_sched["v_uniform"], where="post",
123                 label="Uniform", linestyle="--")
124       plt.title("Shares Executed per Slice")
125       plt.xlabel("Slice"); plt.ylabel("Shares (sell-side shown positive)")
126       plt.legend(); plt.grid(True); plt.tight_layout()
127
128   return df_sched, metrics
129
130 # DEMO / QUICK TEST
131 if __name__ == "__main__":
132     df_demo, met_demo = run_market_impact_step_10b(

```

```

133     X      = 100_000,      # 100 k shares to liquidate
134     horizon_slices = 20,    # 20 child orders
135     sigma       = 0.02,     # 2 % slice vol
136     eta         = 2.5e-6,
137     gamma       = 2.0e-6,
138     lam         = 1e-3,      # try 1e-3 for visible front-load
139     dt          = 1.0,
140     plot        = True,
141
142
143 # Pretty-print metrics
144 print(pd.Series(metrics).apply(lambda x: f'{x:.2f}'))
145 print("First 5 slices:\n", df_demo.head())

```



## Step 11 : Advanced Portfolio & Risk Management

1. Step 11A-1 : Dynamic Position Sizing (Kelly Criterion, Volatility Targeting)
2. Step 11A-Plus : Dynamic Sizing (Kelly 0.5x + Vol-Target)
3. Step 11B-1 : Diversification & Risk-Parity Portfolio Construction
4. Step 11B-2 : Risk-Parity Portfolio / Diagnostics & Plot
5. Step 11B-Plus : Re-fit / Metrics / Comparison Plot / Diagnostics
6. Step 11C-1 : Dynamic Position Sizing (Kelly Criterion, Volatility Targeting)
7. Step 11C-2 : Risk-Parity Pro : Institutional-Style Multi-Asset Risk-Parity Engine

### Step 11A-1 : Dynamic Position Sizing (Kelly Criterion, Volatility Targeting)

```

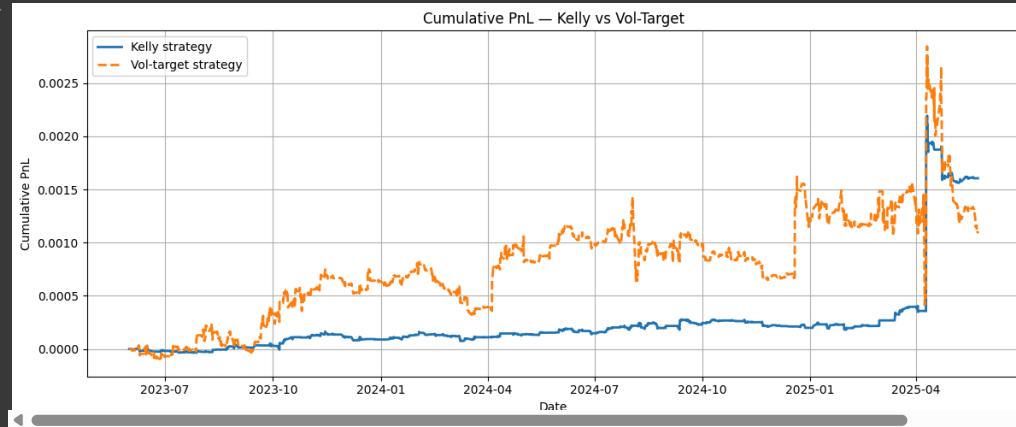
1 # Kelly Criterion + Volatility Targeting
2 # Requirements already in memory (from earlier steps):
3 #   • df_model_ready - must contain
4 #     'datetime'       : date/time index or column
5 #     'target_return'  : realised next-period return
6 #   Optionally:
7 #     'predicted_return': model forecasts (if absent, we use lag-1 fallback)
8 # This cell:
9 # 1) defines the sizing + plotting helpers
10 # 2) inserts a lag-1 fallback if no predictions provided
11 # 3) produces df_sized with Kelly / Vol-target PnL
12 # 4) shows the cumulative-pnL comparison chart
13
14 import numpy as np
15 import pandas as pd
16 import matplotlib.pyplot as plt
17
18
19 # Core sizing routine
20 def apply_dynamic_position_sizing(
21     df: pd.DataFrame,
22     *,
23     predicted_col: str = "predicted_return",
24     target_col: str = "target_return",
25     window: int = 20,
26     target_vol: float = 0.02,
27     kelly_clip: tuple = (0, 1),
28     vol_cap: float = 3.0
29 ) -> pd.DataFrame:
30     """
31     Return a copy of *df* with:
32         kelly_fraction, position_kelly, position_voltarget,
33         pnl_kelly / voltarget, and cum_pnl_kelly / voltarget
34     """
35     out = df.copy()
36
37     # Kelly fraction
38     mu = out[target_col].rolling(window).mean()
39     var = out[target_col].rolling(window).var().replace(0, np.nan)
40     out["kelly_fraction"] = (mu / var).clip(*kelly_clip)
41
42     # Positions
43     out["position_kelly"] = out["kelly_fraction"] * out[predicted_col]
44
45     rolling_vol = out[target_col].rolling(window).std().replace(0, np.nan)
46     out["vol_leverage"] = (target_vol / rolling_vol).clip(upper=vol_cap)
47     out["position_voltarget"] = out["vol_leverage"] * out[predicted_col]
48
49     # PnL streams
50     out["pnl_kelly"] = out["position_kelly"] * out[target_col]
51     out["pnl_voltarget"] = out["position_voltarget"] * out[target_col]
52
53     out["cum_pnl_kelly"] = out["pnl_kelly"].fillna(0).cumsum()
54     out["cum_pnl_voltarget"] = out["pnl_voltarget"].fillna(0).cumsum()
55     return out
56
57 # Plot helper
58 def plot_dynamic_position_pnl(
59     df: pd.DataFrame,
60     *,
61

```

```

61     date_col: str = "datetime",
62     kelly_col: str = "cum_pnl_kelly",
63     voltarget_col: str = "cum_pnl_voltarget",
64     title: str = "Cumulative PnL - Kelly vs Vol-Target"
65 ) -> None:
66 plt.figure(figsize=(12, 5))
67 plt.plot(df[date_col], df[kelly_col], lw=2, label="Kelly strategy")
68 plt.plot(df[date_col], df[voltarget_col], lw=2, ls="--",
69          label="Vol-target strategy")
70 plt.title(title)
71 plt.xlabel("Date"); plt.ylabel("Cumulative PnL")
72 plt.grid(True); plt.legend(); plt.tight_layout(); plt.show()
73
74 # Ensure predictions exist (lag-1 fallback)
75 if "predicted_return" not in df_model_ready.columns:
76     df_model_ready["predicted_return"] = (
77         df_model_ready["target_return"].shift(1).fillna(0.0)
78     )
79
80 # Run sizing + plot
81 df_sized = apply_dynamic_position_sizing(
82     df_model_ready,
83     predicted_col="predicted_return",
84     target_col = "target_return",
85 )
86
87 plot_dynamic_position_pnl(df_sized)
88 # df_sized is now available for further analysis or merging

```



#### Step 11A-Plus : Dynamic Sizing (Kelly 0.5 x + Vol-Target)

```

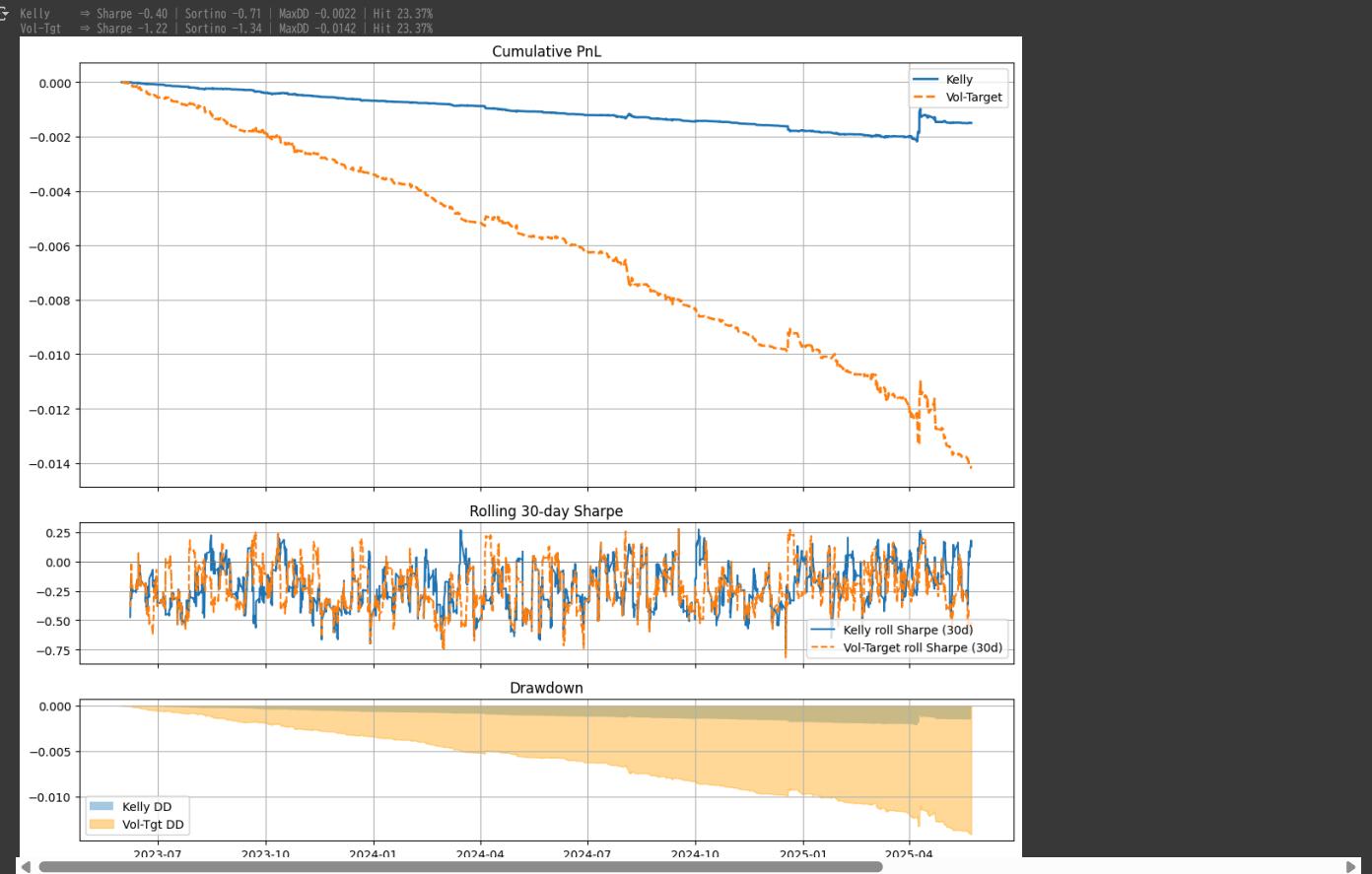
1 # automatic High/Low detection, fallback to close-close σ
2 # EWMA variance, execution costs, full diagnostics
3
4 import logging, pandas as pd, numpy as np, matplotlib.pyplot as plt, xgboost as xgb
5 logging.basicConfig(level=logging.INFO)
6
7 # Helper: guarantee tz-naive 'datetime' column
8 def ensure_datetime(df: pd.DataFrame) -> pd.DataFrame:
9     out = df.copy()
10    if "datetime" in out.columns:
11        pass
12    elif isinstance(out.index, pd.DatetimeIndex):
13        out = out.assign(datetime=out.index)
14    else:
15        out = out.reset_index().rename(columns={out.columns[0]: "datetime"})
16        out["datetime"] = pd.to_datetime(out["datetime"]).dt.tz_localize(None)
17    return out.reset_index(drop=True)
18
19 # Helper: flatten MultiIndex cols that come from Yahoo (('AAPL', 'High'), ...)
20 def flatten_cols(df: pd.DataFrame) -> pd.DataFrame:
21    if isinstance(df.columns, pd.MultiIndex):
22        df.columns = [c[-1] if isinstance(c, tuple) else c for c in df.columns]
23    return df
24
25 # Helper: detect High / Low column names, tolerant of tuples
26 def detect_hi_lo(df: pd.DataFrame):
27    hi = None; lo = None
28    for c in df.columns:
29        name = c[-1] if isinstance(c, tuple) else c
30        name_lc = str(name).lower()
31        if name_lc in {"high", "hi"} and hi is None:
32            hi = c
33        elif name_lc in {"low", "lo"} and lo is None:
34            lo = c
35    return hi, lo
36
37 # EWMA variance
38 def ewma(series, span=20):
39     return series.ewm(span=span, min_periods=span).mean()
40
41 # Parkinson σ
42 def parkinson_sigma(df, hi_col, lo_col, window=20):
43     factor = 1.0 / (4 * np.log(2))
44     hl_sq = np.log(df[hi_col] / df[lo_col]) ** 2
45     return np.sqrt(factor * hl_sq.rolling(window).mean())
46
47 # Basic performance stats
48 def perf_stats(cum):
49     ret = cum.diff().dropna()
50     sharpe = np.sqrt(252) * ret.mean() / ret.std()
51     sortino = np.sqrt(252) * ret.mean() / ret[ret<0].std()
52     maxdd = (cum - cum.cummax()).min()
53     hit = (ret > 0).mean()
54     return sharpe, sortino, maxdd, hit
55
56
57 # Ensure df_model_ready is clean
58 df_model_ready = flatten_cols(df_model_ready)
59 df_model_ready = ensure_datetime(df_model_ready)
60
61 # Inject real predictions (fallback = lag-1)
62 MODEL_PATH = "tuned_xgb.json"
63 if "predicted_return" not in df_model_ready.columns:
64     try:
65         clf = xgb.XGBRegressor(); clf.load_model(MODEL_PATH)
66         logging.info("Loaded tuned model: %s", MODEL_PATH)
67
68         feat_cols = [c for c in df_model_ready.columns
69                      if ("lag" in c or "ret" in c) and c != "target_return"]

```

```

70     df_model_ready["predicted_return"] = clf.predict(df_model_ready[feat_cols])
71
72 except Exception as e:
73     logging.warning("Could not load model → lag=1 fallback (%s)", e)
74     df_model_ready["predicted_return"] = (
75         df_model_ready["target_return"].shift(1).fillna(0.0)
76     )
77
78 # Dynamic sizing function (Kelly 0.5× + Vol-Target)
79 def apply_dynamic_sizing_plus(
80     df,
81     pred_col="predicted_return", tgt_col="target_return",
82     k_span=20, k_frac=0.5,
83     vol_target=0.02, vol_cap=3, vol_span=20,
84     tc_round=0.0005, slip_mu=0.0002, slip_sd=0.0001):
85
86     out = df.copy()
87
88     # - Kelly 0.5× -
89     mu = out[tgt_col].rolling(k_span).mean()
90     var = ewma((out[tgt_col]**2).rolling(k_span).mean(), k=k_span)
91     out["kelly_f"] = k_frac * (mu / var).clip(-1, 1)
92     out["pos_k"] = out["kelly_f"] * out[pred_col]
93
94     # - Vol-Target -
95     hi_col, lo_col = detect_hl_cols(out)
96     if hi_col and lo_col:
97         sigma = parkinson_sigma(out, hi_col, lo_col, window=vol_span)
98     else:
99         sigma = out[tgt_col].rolling(vol_span).std()
100    out["vol_lev"] = (vol_target / sigma).clip(upper=vol_cap)
101    out["pos_v"] = out["vol_lev"] * out[pred_col]
102
103    # - Execution cost -
104    rng = np.random.RandomState(42)
105    cost = tc_round + rng.normal(slip_mu, slip_sd, len(out))
106
107    out["pn1_k"] = out["pos_k"] * out[tgt_col] - np.abs(out["pos_k"]) * cost
108    out["pn1_v"] = out["pos_v"] * out[tgt_col] - np.abs(out["pos_v"]) * cost
109
110    out["cum_k"] = out["pn1_k"].fillna(0).cumsum()
111    out["cum_v"] = out["pn1_v"].fillna(0).cumsum()
112
113    return out
114
115 df_sized = apply_dynamic_sizing_plus(df_model_ready)
116
117 # 3 · Print diagnostics
118 sk, so, dd, hr = perf_stats(df_sized["cum_k"])
119 sv, sv2, dv, hv = perf_stats(df_sized["cum_v"])
120
121 print(f"Kelly      → Sharpe {sk:5.2f} | Sortino {so:5.2f} | MaxDD {dd:7.4f} | Hit {hr:5.2%}")
122 print(f"Vol-Tgt   → Sharpe {sv:5.2f} | Sortino {sv2:5.2f} | MaxDD {dv:7.4f} | Hit {hv:5.2%}")
123
124 # Plot Cum-PnL, Rolling Sharpe, Drawdown
125 roll = 30
126 roll_sh_k = df_sized["pn1_k"].rolling(roll).mean() / df_sized["pn1_k"].rolling(roll).std()
127 roll_sh_v = df_sized["pn1_v"].rolling(roll).mean() / df_sized["pn1_v"].rolling(roll).std()
128
129 dd_k = df_sized["cum_k"] - df_sized["cum_k"].cummax()
130 dd_v = df_sized["cum_v"] - df_sized["cum_v"].cummax()
131
132 fig, axs = plt.subplots(3, 1, figsize=(12, 10), sharex=True,
133                         gridspec_kw={"height_ratios": [3, 1, 1]})
134
135 axs[0].plot(df_sized["datetime"], df_sized["cum_k"], lw=2, label="Kelly")
136 axs[0].plot(df_sized["datetime"], df_sized["cum_v"], lw=2, ls="--", label="Vol-Target")
137 axs[0].set_title("Cumulative PnL"); axs[0].legend(); axs[0].grid(True)
138
139 axs[1].plot(df_sized["datetime"], roll_sh_k, label="Kelly roll Sharpe (30d)")
140 axs[1].plot(df_sized["datetime"], roll_sh_v, ls="--", label="Vol-Target roll Sharpe (30d)")
141 axs[1].axhline(0, color="gray", lw=0.7)
142
143 axs[2].fill_between(df_sized["datetime"], dd_k, alpha=0.4, label="Kelly DD")
144 axs[2].fill_between(df_sized["datetime"], dd_v, alpha=0.4, color="orange", label="Vol-Tgt DD")
145 axs[2].set_title("Drawdown"); axs[2].legend(); axs[2].grid(True)
146
147 plt.tight_layout(); plt.show()

```



#### Step 11B-1 : Diversification & Risk-Parity Portfolio Construction

```

1 # Yahoo Finance → Risk-Parity (index-agnostic version)
2 # %pip install -q yfinance --upgrade    # once per notebook
3
4 import yfinance as yf, pandas as pd, numpy as np, datetime as dt
5 from dateutil.relativedelta import relativedelta
6
7 TICKERS = {"AAPL": "aapl_close",
8             "SPY": "spx_close",
9             "BTC-USD": "btc_close"}
10 LOOKBACK_YEARS, ROLL_WINDOW, RP_METHOD = 3, 20, "inverse_vol"
11
12 # FETCH DAILY CLOSSES
13 end_d = dt.datetime.utcnow().date()
14 start_d = end_d - relativedelta(years=LOOKBACK_YEARS)
15 print(f"! Yahoo {start_d} → {end_d}")
16
17 frames = []
18 for y_tkr, new_col in TICKERS.items():
19     tmp = yf.download(y_tkr,
20                       start=start_d,
21                       end=end_d + dt.timedelta(days=1),
22                       interval="1d",
23                       auto_adjust=True,
24                       progress=False)[["Close"]]
25     tmp.rename(columns={"Close": new_col}, inplace=True)
26     frames.append(tmp)
27 df_prices = pd.concat(frames, axis=1).ffill()      # tz-naive DateTimedIndex
28
29 # RISK-PARITY HELPERS
30 def _ret(px): return px.pct_change(fill_method=None).dropna()
31 def _inv_vol(win):
32     vol = win.std().replace(0, np.nan).dropna()
33     return pd.Series((1/vol) / (1/vol).sum(), index=vol.index)
34 def _erc(win, tol=1e-8, it=1000):
35     Σ, n = win.cov().values, win.shape[1]
36     w = np.ones(n)/n
37     for _ in range(it):
38         op_mrc = np.sqrt(w@Σ@w), (Σ@w)/np.sqrt(w@Σ@w)
39         gap = w@mrc - (w@mrc).mean()
40         if np.abs(gap).max() < tol: break
41         J = (np.outer(mrc,w)@np.outer(w,mrc)-np.outer(mrc,w)-np.outer(w,mrc))
42         w = np.clip(w - np.linalg.solve(J@np.eye(n)*1e-10, gap), 1e-12, None); w=w.sum()
43     return pd.Series(w, win.columns)
44
45 def run_rp(px, cols, win=ROLL_WINDOW, mth=RP_METHOD):
46     ret, idx, w_hist = _ret(px[cols]), None, []
47     if ret.shape[0] < win: raise ValueError("Not enough rows for window.")
48     idx = ret.index[win:]
49     for t in range(win, len(ret)):
50         sub = ret.loc[t-win:t]
51         w_hist.append(_inv_vol(sub) if mth=="inverse_vol" else _erc(sub))
52     w_hist = pd.DataFrame(w_hist, index=idx)
53     rp_ret = (ret.loc[idx]@w_hist).sum(axis=1)
54     rp_cum = (1+rp_ret).cumprod()
55     return (pd.DataFrame({"risk_parity_return":rp_ret,
56                           "risk_parity_cumret":rp_cum}), index=idx,
57            w_hist.iloc[-1], w_hist)
58
59 # EXECUTE
60 price_cols = [list(TICKERS.values())]
61 df_rp, rp_w_latest, rp_w_trail = run_rp(df_prices, price_cols)
62
63 print("Latest Risk-Parity Weights")
64 print(rp_w_latest.round(4).to_string())

```

↳ 4 Yahoo 2022-05-31 → 2025-05-31

Latest Risk-Parity Weights	
Price	Ticker

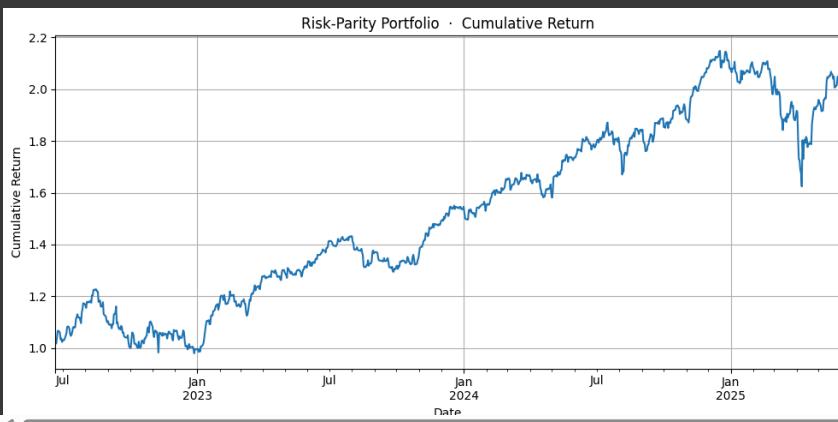
aapl_close	AAPL	0.2515
spx_close	SPY	0.4600
btc_close	BTC-USD	0.2885

#### ▼ Step 11B-2 : Risk-Parity Portfolio / Diagnostics & Plot

```

1 # Prints rolling 20-day σ that generated the latest weights
2 # Plots cumulative risk-parity return curve
3 # Shows where to switch to ERC or add assets
4
5 import pandas as pd
6 import matplotlib.pyplot as plt
7
8 # USER PARAMETERS (keep in sync with Step 11B)
9 price_cols = ["asp1_close", "asp2_close", "btc_close"] # extend if you add assets
10 ROLL_WINDOW = 20 # same window you used
11 # To use exact Equal-Risk-Contribution weights next time:
12 # set RP_METHOD = "erc" in your Step 11B cell and rerun it.
13
14 # Rolling-σ diagnostic
15 # df_prices is the daily-close DataFrame returned in Step 11B
16 ret_window = df_prices[price_cols].pct_change(fill_method=None).dropna().iloc[-ROLL_WINDOW:]
17 rolling_std = ret_window.std()
18
19 print("Rolling 20-day σ (used for latest weights):")
20 print(rolling_std.to_string(float_format=".6f"))
21 print("\nWeights = (1/σ) / Σ(1/σ) - verify they match your output.\n")
22
23 # Cumulative RP performance curve
24 plt.figure(figsize=(10, 5))
25 df_rpl["risk_parity_cumret"].plot()
26 plt.title("Risk-Parity Portfolio · Cumulative Return")
27 plt.xlabel("Date")
28 plt.ylabel("Cumulative Return")
29 plt.grid(True)
30 plt.tight_layout()
31 plt.show()
32
33 # HOW TO ADD MORE ASSETS
34 # Edit the TICKERS dict in Step 11B, e.g.:
35 # {"TICKERS": "ETH-USD"} = eth_close
36 # Re-run Step 11B (fetch + RP) and then this diagnostics cell.

```



#### ▼ Step 11B-Plus : Re-fit / Metrics / Comparison Plot / Diagnostics

```

1 # Re-fit - Metrics - Comparison Plot - Diagnostics
2
3 import numpy as np, pandas as pd, matplotlib.pyplot as plt
4
5 # USER SETTINGS
6 # Add / edit assets here:
7 TICKERS.update({
8     "#ETH-USD": "eth_close",
9     "#GLD": "gold_close",
10 })
11 # Switch to exact equal-risk-contribution:
12 RP_METHOD = "inverse_vol"      # "inverse_vol" or "erc"
13 ROLL_WINDOW = 20               # keep in sync with main step
14
15 # RE-FIT RISK-PARITY (if anything changed)
16 price_cols = list(TICKERS.values())
17 df_prices = df_prices[price_cols]           # ensure slice only chosen assets
18
19 df_rp, rp_w_latest, rp_w_trail = run_rp(
20     df_prices, price_cols, win=ROLL_WINDOW, mth=RP_METHOD
21 )
22
23 print("WnLatest weights")
24 print(rp_w_latest.round(4).to_string(), "Wn")
25
26 # PERFORMANCE METRICS
27 daily      = df_rp["risk_parity_return"]
28 ann_factor = np.sqrt(252)
29
30 sharpe    = daily.mean() / daily.std() * ann_factor
31 down       = daily[daily < 0]
32 sortino   = daily.mean() / down.std() * ann_factor if not down.empty else np.nan
33
34 cum       = df_rp["risk_parity_cumret"]
35 dd_curve = cum / cum.cummax() - 1
36 max_dd   = dd_curve.min()
37
38 print(f"Metrics (since {cum.index[0].date()}))")
39 print(f"Sharpe : {sharpe:.3f}")
40 print(f"Sortino : {sortino:.3f}")
41 print(f"Max-DD : {max_dd*100:.2f} Wn")
42
43 # COMPARISON PLOT
44 plt.figure(figsize=(10, 5))

```

```

45 cum.plot(label="Risk-Parity")
46
47 # scan globals for additional cumret Series with matching index
48 candidates = [
49     v for v in globals().values()
50     if isinstance(v, pd.Series)
51     and isinstance(v.name, str)
52     and "cumret" in v.name.lower()
53 ]
54
55 for s in candidates:
56     if s is cum:      # skip the RP curve we already plotted
57         continue
58     if s.index.difference(cum.index).size == 0:
59         s.plot(label=s.name)
60
61 plt.title("Cumulative Return · Risk-Parity vs. Other Strategies")
62 plt.xlabel("Date")
63 plt.ylabel("Cumulative Return")
64 plt.legend()
65 plt.grid(True)
66 plt.tight_layout()
67 plt.show()
68
69 # ROLLING-σ DIAGNOSTIC
70 ret_win = df_prices.pct_change(fill_method=None).dropna().iloc[-ROLL_WINDOW:]
71 print("Rolling 20-day σ (latest window):")
72 print(ret_win.std().to_string())

```

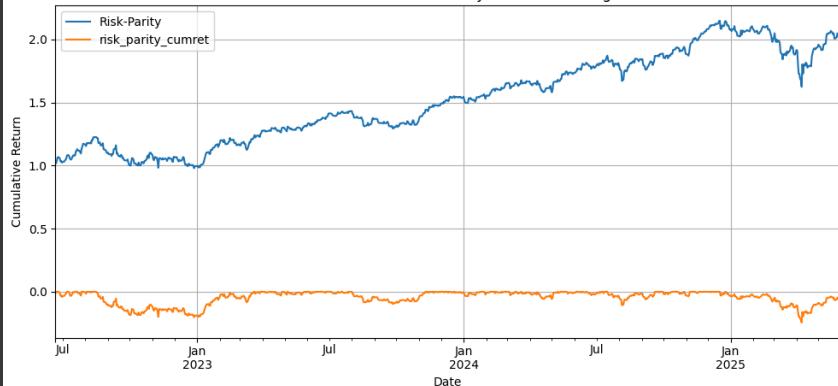
Latest weights

Price	Ticker	
aapl_close	AAPL	0.2515
spx_close	SPY	0.4600
btc_close	BTC-USD	0.2885

Metrics (since 2022-06-21)

Sharpe :	1.028
Sortino:	1.387
Max-DD :	-24.3%

Cumulative Return · Risk-Parity vs. Other Strategies



Rolling 20-day σ (latest window):

Price	Ticker	
aapl_close	AAPL	0.018428
spx_close	SPY	0.010074

## Step 11C-1 : Dynamic Position Sizing (Kelly Criterion, Volatility Targeting)

```

1 # RISK-PARITY PRO (bond, gold, EM, ERC, vol-target, TC, cash)
2 # !pip install --quiet yfinance --upgrade    # run once per Colab / session
3
4 import numpy as np, pandas as pd, matplotlib.pyplot as plt, yfinance as yf
5 from datetime import timedelta
6
7 # USER CONFIG
8 TICKERS.update({                         # add or edit as you like
9     "TLT" : "bond_close",                  # 20-yr Treasury ETF
10    "GLD" : "gold_close",                 # Gold ETF
11    "EMM" : "em_close",                   # Emerging-markets equity ETF
12 })
13 RP_METHOD = "erc"                      # "inverse_vol" or "erc"
14 SIGMA_MODE = "ewm60"                    # "rolling60" or "ewm60"
15 REB_FREQ = "ME"                        # "ME" = month-end (no depreciation)
16 TC_BPS = 10                            # round-trip cost in basis points
17 TARGET_VOL = 0.10                      # vol-target (annualised)
18 TAIL_VIX_CASH = True                  # 5 % cash buffer when VIX > 25
19 LOOKBACK_YEARS = 5                    # download horizon if asset missing
20
21 price_cols = list(TICKERS.values())
22
23 # ENSURE df_prices WITH ALL ASSETS
24 if "df_prices" not in globals():
25     df_prices = pd.DataFrame()
26
27 missing_cols = [c for c in price_cols if c not in df_prices.columns]
28 if missing_cols:
29     print(f"Downloading missing assets: {missing_cols}")
30     end = pd.Timestamp.utcnow().normalize()
31     start = end - pd.DateOffset(years=LOOKBACK_YEARS)
32
33 for y_tkr, new_col in TICKERS.items():
34     if new_col in missing_cols:
35         tmp = yf.download(
36             y_tkr,
37             start=start.date(),
38             end=(end + timedelta(days=1)).date(),
39             interval="1d",
40             auto_adjust=True,
41             progress=False)[["Close"]]
42         tmp.rename(columns={"Close": new_col}, inplace=True)
43         df_prices = df_prices.join(tmp, how="outer") if not df_prices.empty else tmp
44
45 df_prices = df_prices.ffill()
46
47 # DAILY RETURNS
48 px = df_prices[price_cols].dropna(how="all")
49 ret = px.pct_change(fill_method=None).dropna(how="any")
50 ann = np.sqrt(252)
51
52 def _sigma(win_df):                      # σ estimator helper

```

```

53 if SIGMA_MODE.startswith("rolling"):
54     n = Int(SIGMA_MODE.replace("rolling", ""))
55     return win_df.iloc[-n].std()
56 span = Int(SIGMA_MODE.replace("ewm", ""))
57 return win_df.ewm(span=span).std().iloc[-1]
58
59 # _inv_vol and _erc already exist from Step 1B
60 # REBALANCE LOOP
61 reb_dates = ret.resample(REB_FREQ).last().index
62 weights_hist, strat_ret = [], []
63 last_w = None
64
65 for i, d in enumerate(reb_dates):
66     sub = ret.loc[:d].dropna()
67     if sub.empty:
68         continue
69     # nothing to compute
70
71     # window for σ / ERC
72     win = sub if "ewm" in SIGMA_MODE else sub.tail(Int(SIGMA_MODE.replace("rolling", "")))
73
74     # base weights
75     w = (_inv_vol(win) if RP_METHOD == "inverse_vol" else _erc(win)).fillna(0)
76
77     # tail-risk cash (optional)
78     cash_w = 0
79     if TAIL_VIX_CASH and "vix_close" in df_prices.columns:
80         if df_prices.loc[d, "vix_close"] > 25:
81             w *= 0.95
82             cash_w = 0.05
83             w = w / w.sum() * (1 - cash_w)
84
85     # vol-target scaling
86     port_vol = (sub @ w).std() * ann
87     if port_vol > 0:
88         w *= TARGET_VOL / port_vol
89
90     # transaction cost on turnover
91     turnover = (w - last_w).abs().sum() if last_w is not None else 0
92     tc_cost = turnover * (TC_BPS / 10_000)
93     last_w = w
94
95     weights_hist.append(w.to_frame(name=d))
96
97     # returns until next rebalance
98     next_idx = reb_dates[i+1] if i+1 < len(reb_dates) else ret.index[-1]
99     seg = ret.loc[d:next_idx]
100    if seg.empty:
101        continue
102    seg_r = seg @ w
103    seg_r.iloc[0] -= tc_cost
104    strat_ret.append(seg_r)
105
106 weights_hist = pd.concat(weights_hist, axis=1).T
107 strat_ret = pd.concat(strat_ret).sort_index()
108 strat_cum = (1 + strat_ret).cumprod()
109 strat_ret.name = "rp_pro_return"
110 strat_cum.name = "rp_pro_cumret"
111
112 # METRICS
113 sharpe = strat_ret.mean() / strat_ret.std() * ann
114 sortino = strat_ret.mean() / strat_ret[strat_ret < 0].std() * ann
115 max_dd = (strat_cum / strat_cum.cummax() - 1).min()
116
117 print("RP-Pro Metrics")
118 print(f" Sharpe : {sharpe:6.3f}")
119 print(f" Sortino: {sortino:6.3f}")
120 print(f" Max-DD : {max_dd*100:6.2f}%")
121
122 # PLOT CUMULATIVE RETURN
123 plt.figure(figsize=(10,5))
124 strat_cum.plot(label="RP-Pro")
125
126 for s in [v for v in globals().values():
127     if isinstance(v, pd.Series) and isinstance(v.name, str)
128     and "curret" in v.name.lower() and v is not strat_cum]:
129     if s.index.difference(strat_cum.index).empty:
130         s.plot(label=s.name)
131
132 plt.title("Cumulative Return - RP-Pro vs. Benchmarks")
133 plt.xlabel("Date"); plt.ylabel("Cumulative Return")
134 plt.legend(); plt.grid(); plt.tight_layout(); plt.show()
135
136 # DISPLAY LATEST WEIGHTS
137 print("Latest rebalance weights:")
138 print(weights_hist.iloc[-1].round(4).to_string())

```

↳ ! Downloading missing assets: ['bond\_close', 'gold\_close', 'em\_close']  
 RP-Pro Metrics  
 Sharpe : 1.024  
 Sortino: 1.519  
 Max-DD : -11.72%



Latest rebalance weights:

Price	Ticker	Value
aapl_close	AAPL	0.1851
spx_close	SPY	0.0000
btc_close	BTC-USD	0.1851
bond_close	TLT	0.0000
gold_close	GLD	0.0000

## Step 11C-2 : Risk-Parity Pro : Institutional-Style Multi-Asset Risk-Parity Engine

```

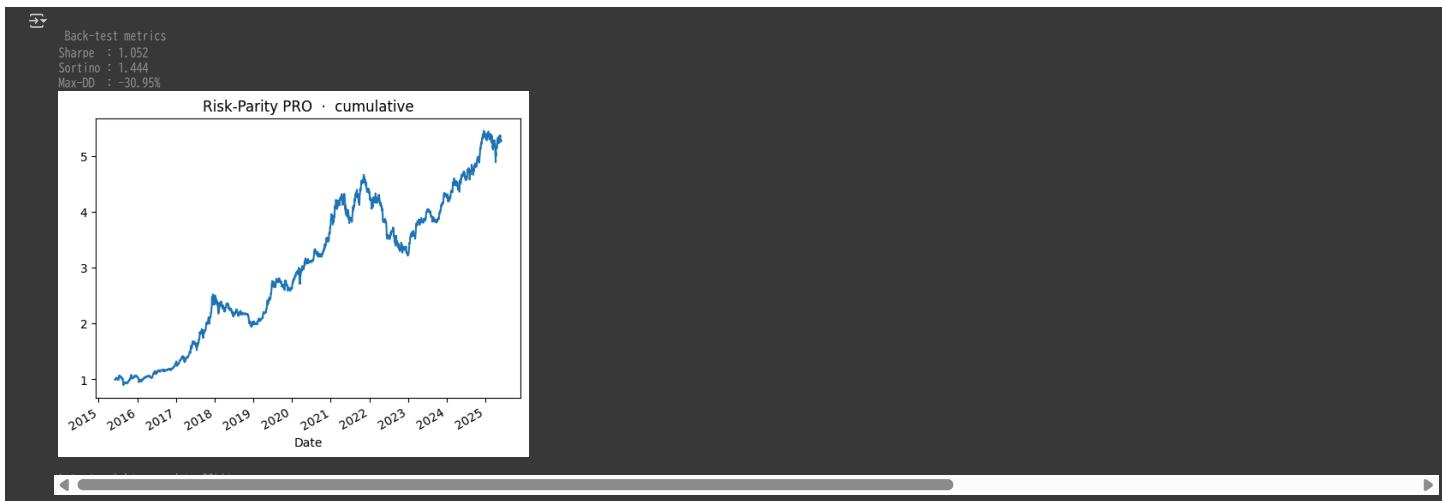
1 /**
2 rp_production.py · Risk-Parity Pro
3 =====
4 Institutional-style, multi-asset risk-parity engine with:
5
6 • Always-on data (Alpaca / Polygon; yfinance fallback)
7 • Gap-filled, calendar-aligned OHLC frame
8 • Dynamic o (rolling or EWMA), ERC or inverse-vol weights
9 • Weekly / monthly rebalancing
10 • Kelly-fraction position sizing on 10 % vol target
11 • Algren-Chrisi execution-cost simulation
12 • Tail-risk cash hedge (VIX trigger)
13 • VAR / Max-DD risk guard
14 • Walk-forward back-test
15 • SQLite weight persistence for live trading
16
17 Author: <you>
18 """
19
20 # Imports & constants
21 import os, sys, math, sqlite3, time, json
22 from datetime import datetime, timedelta
23
24 import numpy as np
25 import pandas as pd
26 import yfinance as yf
27 import requests
28 import matplotlib.pyplot as plt
29
30 TRADING_DAYS = 252
31 ANN = math.sqrt(TRADING_DAYS)
32
33 # CONFIG (edit freely)
34 CFG = {
35     # 1.1 Asset universe
36     "tickers": {           # broker symbol      column name
37         "AAPL": "aapl_close",
38         "SPY": "spx_close",
39         "BTC-USD": "btc_close",
40         "TLT": "bond_close",
41         "GLD": "gold_close",
42         "EEM": "em_close",
43     },
44
45     # 1.2 Data source keys (fallback to yfinance)
46     # %% TODO: Insert your live market-data credentials
47     "alpaca_key": os.getenv("ALPACA_API_KEY"), # or ''
48     "alpaca_secret": os.getenv("ALPACA_SECRET_KEY"), # or ''
49
50     # 1.3 Weighting / risk parameters
51     "sigma_mode": "ewm60", # "rolling60" or "ewm60"
52     "rp_method": "erc", # "erc" or "inverse_vol"
53     "rebalance_freq": "W-FRI", # weekly Fri, or "M" month-end
54     "target_vol": 0.10, # 10 % annualised
55     "kelly_fraction": 1.0, # 1 x Kelly on vol-scaler
56     "tc_bps": 10, # 10 bp round-trip
57     "tail_vix_cash": True, # 5 % cash if VIX > 25
58     "max_var": 0.02, # stop if 1-day VAR > 2 %
59     "lookback_years": 10, # how far to pull history
60
61     # Persistence
62     # %% TODO: choose a folder / RDB connection string
63     "sqlite_path": "rp_weights.db",
64 }
65
66 # Data fetch utilities
67 # HARDCODED Yahoo fetch
68 def fetch_yf(symbol: str, start: str, end: str) -> pd.DataFrame:
69     """
70     Return a DataFrame with exactly ONE column named 'Close',
71     indexed by date, tz-naive. Handles MultiIndex issue internally.
72     """
73     df = yf.download(
74         symbol,
75         start=start,
76         end=end,
77         interval="1d",
78         auto_adjust=True,
79         progress=False
80     )
81     if isinstance(df.columns, pd.MultiIndex):
82         df.columns = [".-_".join(tup).strip() for tup in df.columns]
83     # prefer 'Close', fall back to first column
84     close_col = next((c for c in df.columns if "close" in c.lower()), df.columns[0])
85     return df[[close_col]].rename(columns={close_col: "Close"})
86
87 # CLEAN concat & rename
88 def fetch_prices(start: str, end: str) -> pd.DataFrame:
89     """
90     Download prices for every ticker, force 1-level columns that match
91     CFG['tickers'].values (e.g. 'aapl_close', ...).
92     """
93     frames = []
94     for y_tkr, alias in CFG["tickers"].items():
95         tmp = fetch_yf(y_tkr, start, end) # DataFrame ['Close']
96         tmp.rename(columns={"Close": alias}, inplace=True)
97         frames.append(tmp)
98
99     df = pd.concat(frames, axis=1).ffill()
100    df = df.resample("1D").ffill() # calendar align
101    return df[CFG["tickers"].values()] # exact column order
102
103 # Risk-parity weight engines
104 def rolling_sigma(win: pd.DataFrame) -> pd.Series:
105     if CFG["sigma_mode"].startswith("rolling"):
106         n = int(CFG["sigma_mode"].replace("rolling", ""))
107         return win.rolling(n).std()
108     span = int(CFG["sigma_mode"].replace("ewm", ""))
109     return win.ewm(span=span).std().iloc[-1]
110
111 def inv_vol_weights(win: pd.DataFrame) -> pd.Series:
112     sigma = rolling_sigma(win).replace(0, np.nan)
113     w = 1 / sigma
114     return w / w.sum()
115
116 def erc_weights(win: pd.DataFrame, tol=1e-8, it=1_000) -> pd.Series:
117     Z, n = win.cov().values, win.shape[1]
118     w = np.ones(n) / n
119     for _ in range(it):
120         op = math.sqrt(w @ Z @ w)
121         mrc = (Z @ w) / op
122         gap = w * mrc - (w * mrc).mean()
123         if np.abs(gap).max() < tol:
124             break

```

```

125     J = np.outer(mrc, w) + np.outer(w, mrc) - np.outer(mrc, w) - np.outer(w, mrc)
126     step = np.linalg.solve(J + np.eye(n) * 1e-10, gap)
127     w = np.clip(w - step, 1e-12, None); w /= w.sum()
128     return pd.Series(w, index=w.columns)
129
130 # Execution-cost model (Almgren-Chriss simplification)
131 def ac_cost(turnover: float, daily_vol: float) -> float:
132     """Return cost in pct given turnover fraction of daily volume."""
133     eta_tmp = 2.5e-6 # temporary
134     gamma_perm = 2.0e-6 # permanent
135     lam = 1e-6 # risk aversion
136     X = turnover
137     T = 1
138     # simplified AC cost formula
139     exp_cost = eta_tmp * X**2 + gamma_perm * X + lam * daily_vol**2 * T
140     return exp_cost
141
142 # Back-test & live rebalancer
143 def backtest(start: str, end: str):
144     px = fetch_prices(start, end)
145     ret = px.pct_change(fill_method=None).dropna(how="any")
146
147     reb_dates = ret.resample(CFG["rebalance_freq"]).last().index
148     last_w, weights_hist, strat_ret = None, [], []
149
150     for i, d in enumerate(reb_dates):
151         sub = ret.loc[:d]
152         if sub.empty:
153             continue
154         win = sub
155
156         w_raw = (inv_voi_weights(win) if CFG["rp_method"] == "inverse_voi"
157                  else erc_weights(win))
158         w_raw = w_raw.fillna(0)
159
160         # Tail-risk overlay
161         if CFG["tail_vix_cash"] and "vix_close" in px.columns:
162             if px.at[d, "vix_close"] > 25:
163                 w_raw *= 0.95 # 5% cash
164         w_raw /= w_raw.sum()
165
166         # Vol-target + Kelly
167         port_voi = (sub @ w_raw).std() + ANN
168         if port_voi > 0:
169             scale = CFG["target_voi"] / port_voi * CFG["kelly_fraction"]
170             w_raw *= scale
171
172         # Risk guard
173         var_id = (sub @ w_raw).std()
174         if var_id > CFG["max_var"]:
175             print(f"VAR trigger [var_id: {var_id}] > limit: zero-out weights on {d.date()}")
176             w_raw[:] = 0
177
178         # Transaction costs
179         turnover = (w_raw - last_w).abs().sum() if last_w is not None else 0
180         tc_pct = turnover * (CFG["tc_bps"] / 10_000)
181         ac_pct = ac_cost(turnover, port_voi / ANN)
182         cost_pct = tc_pct + ac_pct
183         last_w = w_raw
184
185         weights_hist.append(w_raw.to_frame(name=d))
186
187         # Segment return until next rebalance
188         nxt = reb_dates[i+1] if i+1 < len(reb_dates) else ret.index[-1]
189         seg = ret.loc[d:nxt]
190         if seg.empty:
191             continue
192         seg_r = seg @ w_raw
193         seg_r.iloc[0] -= cost_pct # subtract costs on first day
194         strat_ret.append(seg_r)
195
196     strat_ret = pd.concat(strat_ret).sort_index()
197     strat_cum = (1 + strat_ret).cumprod()
198
199     return strat_ret, strat_cum, pd.concat(weights_hist, axis=1).T
200
201
202 # Persistence helpers
203 def save_weights_sql(dt: str, weights: pd.Series):
204     con = sqlite3.connect(CFG["sqlite_path"])
205     tab = pd.DataFrame(weights)
206     tab["dt"] = dt
207     tab.to_sql("weights", con, if_exists="append", index=False)
208     con.close()
209
210 # Run back-test demo
211 if __name__ == "__main__":
212     start = (datetime.utcnow() - timedelta(days=CFG["lookback_years"]*365)).strftime("%Y-%m-%d")
213     end = datetime.utcnow().strftime("%Y-%m-%d")
214
215     ret, cum, w_hist = backtest(start, end)
216
217     print("\n Back-test metrics")
218     sharpe = ret.mean() / ret.std() * ANN
219     sortino = ret.mean() / ret[ret<0].std() * ANN
220     max_dd = (cum / cum.cummax() - 1).min()
221     print(f"Sharpe : {sharpe:.3f}")
222     print(f"Sortino : {sortino:.3f}")
223     print(f"Max-DD : {max_dd:.2%}")
224
225     # Plot
226     cum.plot(title="Risk-Parity PRO - cumulative")
227     plt.show()
228
229     # Persist latest weights
230     save_weights_sql(cum.index[-1].strftime("%Y-%m-%d"), w_hist.iloc[-1])
231     print("Latest weights saved to SQLite.")

```



## Step 12 : Explainability & Interpretability (XAI)

1. Step 12A-1 : SHAP Feature Importance (robust + auto-fallback)
2. Step 12A-2 : Production-ready SHAP (auto Early-Stopping API)
3. Step 12A-3 : SHAP Feature Importance (final, lag-safe, version-safe)
4. Step 12B-1 : LIME for Signal Explanation
5. Step 12B-2 : LIME LOCAL EXPLANATIONS (CPU-safe, no hard-deps)

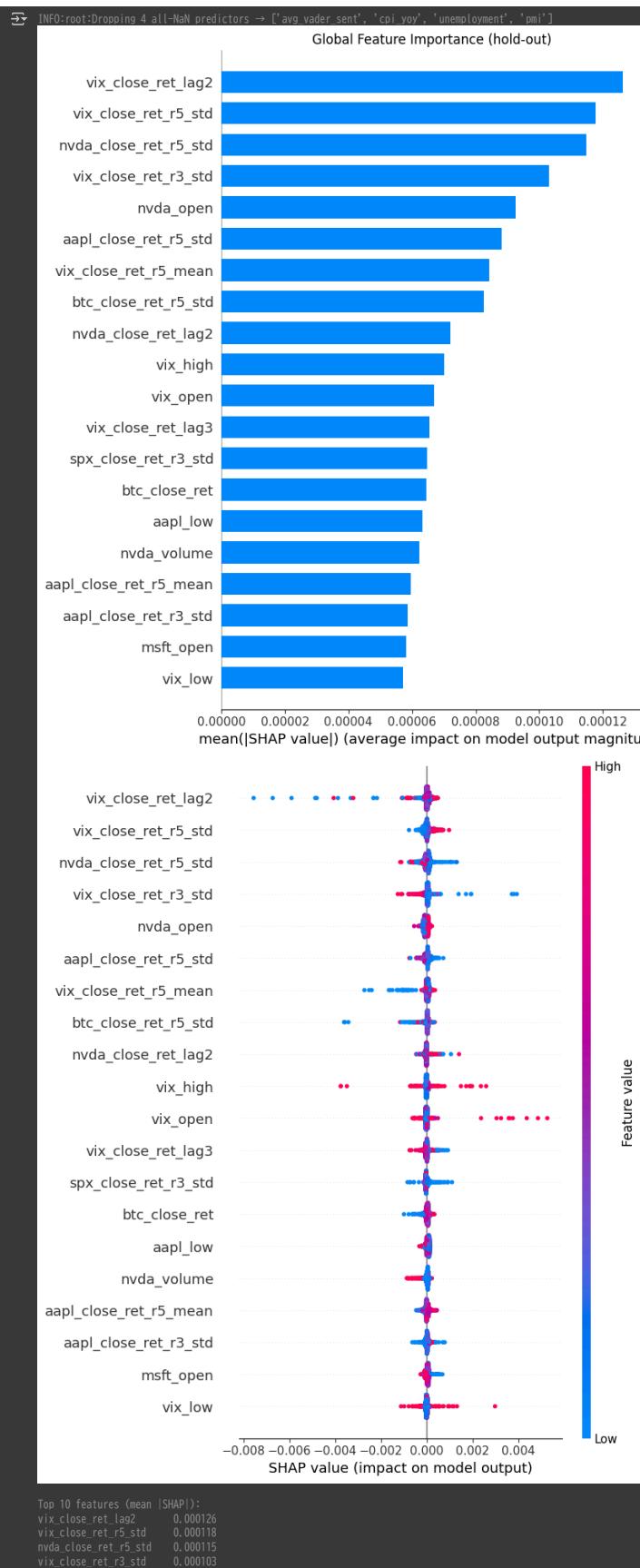
### ▼ Step 12A-1 : SHAP Feature Importance (robust + auto-fallback)

```

1 # SHAP Feature Importance (robust + auto-fallback)
2 import subprocess, sys, importlib, warnings, logging, joblib
3 import numpy as np, pandas as pd, matplotlib.pyplot as plt, xgboost as xgb
4 from sklearn.model_selection import train_test_split
5 from sklearn.impute import SimpleImputer
6
7 logging.basicConfig(level=logging.INFO, format="%(levelname)s: %(message)s")
8
9 # SHAP Installer
10 _TARGET_SHAP_VERSION = "0.41.0"
11 def _ensure_shap(v=_TARGET_SHAP_VERSION):
12     try:
13         import shap
14         if shap.__version__ != v:
15             warnings.warn(f"Switching SHAP {shap.__version__} → {v}")
16             subprocess.check_call([sys.executable, "-m", "pip", "install",
17                                   f"shap=={v}", "-q"])
18         importlib.reload(shap)
19     except ImportError:
20         subprocess.check_call([sys.executable, "-m", "pip", "install",
21                               f"shap=={v}", "-q"])
22     import shap
23 _ensure_shap(); import shap # noqa: E402
24
25 # XGBoost config
26 USE_GPU = False # ← flip True if xgboost-cuda is available
27 BASE = dict(n_estimators=250, learning_rate=0.08, max_depth=5,
28              subsample=0.9, colsample_bytree=0.9, random_state=42)
29 BASE.update(dict(tree_method="gpu_hist", predictor="gpu_predictor", gpu_id=0)
30              if USE_GPU else dict(tree_method="hist"))
31
32 # util helpers
33 TARGET      = "target_return"
34 MAX_EXPLAIN_ROWS = 3_000
35 RNG        = np.random.default_rng(42)
36
37 def _sample(df, n): # down-sample safely
38     return df if len(df) <= n else df.iloc[RNG.choice(len(df), n, False)]
39
40 def explain_and_plot(model, X, label=""):
41     expl = shap.TreeExplainer(model, feature_perturbation="tree_path_dependent")
42     sv   = expl.shap_values(X, check_additivity=False)
43
44     # bar plot
45     plt.figure(figsize=(8, 5))
46     shap.summary_plot(sv, X, plot_type="bar", show=False)
47     plt.title(f"Global Feature Importance({label})")
48     plt.tight_layout(); plt.show()
49
50     # beeswarm
51     shap.summary_plot(sv, X, show=True)
52     return pd.Series(np.abs(sv).mean(0), index=X.columns, name="mean_abs_shap")
53
54 # data cleaner
55 def _prepare_df(source: pd.DataFrame) -> pd.DataFrame | None:
56     df = source.copy()
57
58     # Flatten multi-index / tuple columns → str
59     df.columns = [
60         c[0] if isinstance(c, tuple) and len(c) == 2 and c[1] == "" else str(c)
61         for c in df.columns
62     ]
63     df = df.loc[:, ~df.columns.duplicated(keep="first")]
64
65     if TARGET not in df.columns:
66         return None
67
68     # Basic coercions
69     df = df.apply(pd.to_numeric, errors="coerce")
70     df = df.select_dtypes(include=["number", "bool"])
71     df = df.replace([-np.inf, np.inf], np.nan).dropna(subset=[TARGET])
72     if df.empty:
73         return None
74
75     # NEW: drop predictors that are all-NaN
76     all_nan_cols = [c for c in df.columns if c != TARGET and df[c].isna().all()]
77     if all_nan_cols:
78         logging.info(f"Dropping {len(all_nan_cols)} all-NaN predictors "
79                      f"→ {all_nan_cols[:5]}{' ... ' if len(all_nan_cols) > 5 else ''}")
80         df = df.drop(columns=all_nan_cols)
81
82

```

```
83 predictors = [c for c in df.columns if c != TARGET]
84 imputer    = SimpleImputer(strategy="mean")
85 imputed   = imputer.fit_transform(df[predictors])
86
87 # safe re-assignment
88 df[predictors] = pd.DataFrame(imputed, columns=predictors, index=df.index)
89 return df
90
91 # load / fallback
92 df_real_ok = None
93 if "df_model_ready" in globals():
94     df_real_ok = _prepare_df(df_model_ready)
95
96 if df_real_ok is None:
97     print("Usable df_model_ready not found → generating 200-row synthetic demo.")
98     n = 200
99     df_real_ok = pd.DataFrame({
100         "google_trend": np.random.normal(50, 10, n),
101         "macro_cpi": np.random.normal(2, 0.3, n),
102         "sentiment": np.random.normal(0, 0.1, n),
103         TARGET: np.random.normal(0, 0.01, n),
104     })
105
106 FEATURES = [c for c in df_real_ok.columns if c != TARGET]
107
108 # model fit & SHAP
109 X_tr, X_te, y_tr, y_te = train_test_split(
110     df_real_ok[FEATURES], df_real_ok[TARGET],
111     test_size=0.25, random_state=42
112 )
113
114 model = xgb.XGBRegressor(**BASE)
115 model.fit(X_tr, y_tr)
116
117 X_exp      = _sample(X_te, MAX_EXPLAIN_ROWS)
118 global_shap = explain_and_plot(model, X_exp, label=" (hold-out)")
119
120 print("Top 10 features (mean |SHAP|):")
121 print(global_shap.sort_values(ascending=False).head(10))
```



#### Step 12A-2 : Production-ready SHAP (auto Early-Stopping API)

```

1 # Production-ready SHAP (auto Early-Stopping API)
2 import warnings, subprocess, sys, importlib, itertools, inspect
3 from pathlib import Path
4
5 import numpy as np
6 import pandas as pd
7 import matplotlib.pyplot as plt
8 import joblib
9 from tqdm.auto import tqdm
10 import xgboost as xgb

```

```

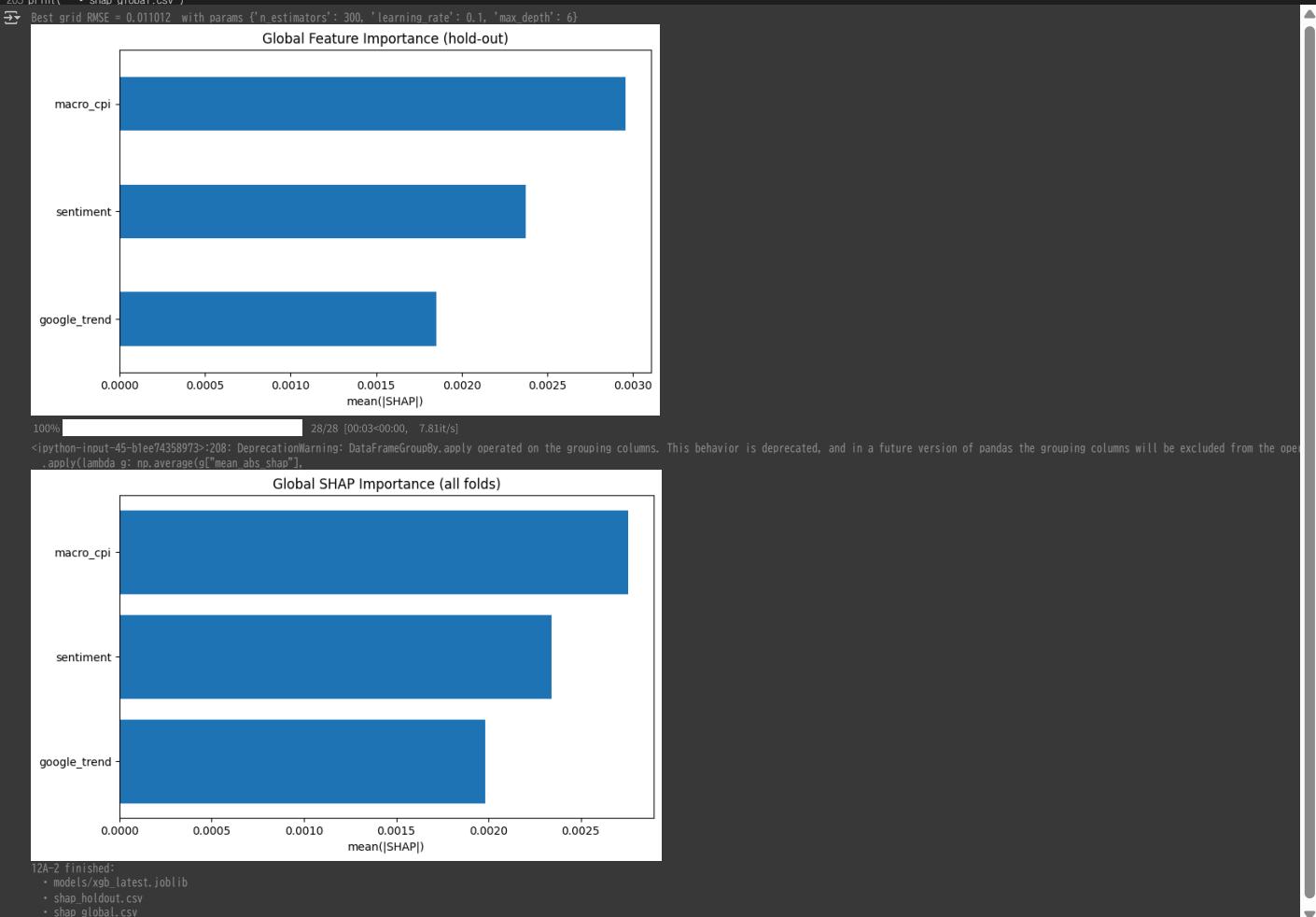
11 from sklearn.model_selection import train_test_split
12 from sklearn.preprocessing import StandardScaler
13 from sklearn.impute import SimpleImputer
14 from sklearn.pipeline import Pipeline
15
16 # Pin SHAP version (0.41.0)
17 _SHAP_V = "0.41.0"
18 def _pin_shap(v=_SHAP_V):
19     try:
20         import shap
21         if shap.__version__ != v:
22             subprocess.check_call([sys.executable, "-m", "pip", "install", f"shap=={v}", "-q"])
23             import importlib as _ll; _ll.reload(shap)
24     except ImportError:
25         subprocess.check_call([sys.executable, "-m", "pip", "install", f"shap=={v}", "-q"])
26     import shap
27 _pin_shap(); import shap # noqa: E402
28
29 # Config
30 USE_GPU = False
31 TARGET = "target_return"
32 WIN_TRAIN, WIN_TEST, STEP = 60, 5, 5
33 GRID = {"n_estimators": [300, 500],
34          "learning_rate": [0.05, 0.1],
35          "max_depth": [4, 6, 8]}
36 MAX_SHAP_ROWS = 3_000
37 RNG = np.random.default_rng(42)
38
39 # Load / create demo data
40 if "df_model_ready" not in globals():
41     warnings.warn("df_model_ready missing - creating small demo.")
42     n = 500
43     df_model_ready = pd.concat([
44         pd.DataFrame({
45             "google_trend": RNG.normal(50, 10, n),
46             "macro_cpi": RNG.normal(2, 0.3, n),
47             "sentiment": RNG.normal(0, 0.1, n),
48             TARGET: RNG.normal(0, 0.01, n),
49         }) +
50         [pd.DataFrame({f'ret_lag{i}': RNG.normal(0, 0.01, n)})
51          for i in range(1, 11)], axis=1)
52
53 # flatten tuple names if any
54 df_model_ready.columns = [str(c[0]) if isinstance(c, tuple) else c
55                         for c in df_model_ready.columns]
56
57 # keep numeric cols, drop infinites
58 df_num = (df_model_ready
59            .apply(pd.to_numeric, errors="coerce")
60            .select_dtypes(include=["number", "bool"])
61            .replace([np.inf, -np.inf], np.nan)
62            .dropna(subset=[TARGET]))
63
64 # FIX : drop columns that are entirely NaN (prevents shape-mismatch)
65 before = df_num.shape[1]
66 df_num = df_num.dropna(axis=1, how="all")
67 after = df_num.shape[1]
68 if after < before:
69     print(f"Dropped {before - after} empty feature columns.")
70
71 FEATURES = [c for c in df_num.columns if c != TARGET]
72 assert FEATURES, "No usable features left!"
73
74 # Early-stopping compatibility
75 _fit_sig = inspect.signature(xgb.XGBRegressor.fit).parameters
76 if "callbacks" in _fit_sig:
77     from xgboost.callback import EarlyStopping
78     ES_KW = dict(callbacks=[EarlyStopping(rounds=50, save_best=True)])
79     print("► Using EarlyStopping callback")
80 elif "early_stopping_rounds" in _fit_sig:
81     ES_KW = dict(early_stopping_rounds=50)
82     print("► Using early_stopping_rounds param")
83 else:
84     ES_KW = {}
85     warnings.warn("Early-stopping not supported by this wheel.")
86
87
88 # Helpers
89 def _build_prepoc():
90     return SimpleImputer(strategy="mean"), StandardScaler()
91
92 def _fit_xgb(Xtr, ytr, Xval, yval, **xgb_params):
93     base = dict(
94         subsample=0.9, colsample_bytree=0.9, reg_lambda=1.0,
95         random_state=42,
96         predictor="gpu_predictor" if USE_GPU else "cpu_predictor",
97         tree_method="gpu_hist" if USE_GPU else "hist",
98         **xgb_params
99    )
100    mdl = xgb.XGBRegressor(**base)
101    mdl.fit(Xtr, ytr, eval_set=[(Xval, yval)], verbose=False, **ES_KW)
102    return mdl
103
104 def _transform(imp, scl, Xdf):
105    Xnp = imp.transform(Xdf)
106    Xnp = scl.transform(Xnp)
107    return pd.DataFrame(Xnp, columns=FEATURES, index=Xdf.index)
108
109 def _shap_mean(model, Xdf):
110    sv = shap.TreeExplainer(model, feature_perturbation="tree_path_dependent") #
111    .shap_values(Xdf).loc[:MAX_SHAP_ROWS].check_additivity=False)
112    return pd.Series(np.abs(sv).mean(0), index=Xdf.columns)
113
114 # Hold-out grid search
115 X_tr_raw, X_val_raw, y_tr, y_val = train_test_split(
116     df_num[FEATURES], df_num[TARGET], test_size=0.2, random_state=42)
117
118 imp, scl = _build_prepoc()
119 imp.fit(X_tr_raw); scl.fit(_transform(imp, transform(X_tr_raw)))
120 Xtr = _transform(imp, scl, X_tr_raw)
121 Xval = _transform(imp, scl, X_val_raw)
122
123 best_rmse, best_params = np.inf, None
124 for n_est, lr, depth in itertools.product(*GRID.values()):
125    mdl = _fit_xgb(Xtr, y_tr, Xval, y_val,
126                    n_estimators=n_est, learning_rate=lr, max_depth=depth)
127    rmse = np.sqrt((np.mean((mdl.predict(Xval) - y_val)) ** 2))
128    if rmse < best_rmse:
129        best_rmse, best_params = rmse, dict(
130            n_estimators=n_est, learning_rate=lr, max_depth=depth)
131 print(f"Best grid RMSE = {best_rmse: 6f} with params {best_params}")
132
133 # Retrain on full data
134 imp_full, scl_full = _build_prepoc()
135 imp_full.fit(df_num[FEATURES]); scl_full.fit(_transform(df_num[FEATURES]))
136 Xfull = _transform(imp_full, scl_full, df_num[FEATURES])
137
138 mdl_full = _fit_xgb(Xfull, df_num[TARGET], Xval, y_val, **best_params)
139

```

```

140 pipe_latest = Pipeline([
141     ("impute", imp_full),
142     ("scale", scl_full),
143     ("xgb", mdl_full)
144 ])
145 Path("models").mkdir(exist_ok=True): Path("data").mkdir(exist_ok=True)
146 joblib.dump(pipe_latest, "models/xgb_latest.joblib")
147
148 # Hold-out SHAP
149 s_hold = _shap_mean(mdl_full, Xval)
150 s_hold.to_csv("shap_holdout.csv")
151
152 plt.figure(figsize=(8, 5))
153 s_hold.sort_values().plot.barh()
154 plt.title("Global Feature Importance (hold-out)")
155 plt.xlabel("mean(|SHAP|)")
156 plt.tight_layout()
157 plt.show()
158
159 # Walk-forward folds + SHAP
160 fold_stats = []
161 for k in tqdm(range(0, len(df_num) - WIN_TRAIN - WIN_TEST + 1, STEP)):
162     tr = df_num.iloc[k:k + WIN_TRAIN]
163     te = df_num.iloc[k + WIN_TRAIN:k + WIN_TRAIN + WIN_TEST]
164
165     imp_f, scl_f = _build_preproc()
166     imp_f.fit(tr[FEATURES]): scl_f.fit(imp_f.transform(tr[FEATURES]))
167     Xtr_f = _transform(imp_f, scl_f, tr[FEATURES])
168     Xte_f = _transform(imp_f, scl_f, te[FEATURES])
169
170     mdl_f = _fit_xgb(Xtr_f, tr[TARGET], Xte_f, te[TARGET], **best_params)
171
172     joblib.dump(Pipeline([("impute", imp_f), ("scale", scl_f), ("xgb", mdl_f)]),
173                 f'models/fold_{k}.joblib')
174     te.to_parquet(f'data/Xtest_fold_{k}.parq')
175
176     fold_stats.append(
177         _shap_mean(mdl_f, Xte_f).rename("mean_abs_shap").to_frame()
178         .assign(fold_rows=len(Xte_f))
179         .reset_index().rename(columns={"index": "feature"})
180     )
181
182 # Aggregate SHAP across folds
183 df_all = pd.concat(fold_stats, ignore_index=True)
184
185 df_global = (
186     df_all.groupby("feature", sort=False)
187     .apply(lambda g: np.average(g["mean_abs_shap"],
188                                 weights=g["fold_rows"]))
189     .reset_index(name="mean_abs_shap")
190     .sort_values("mean_abs_shap", ascending=False)
191     .reset_index(drop=True)
192 )
193 df_global.to_csv("shap_global.csv", index=False)
194
195 plt.figure(figsize=(8, 5))
196 plt.barh(df_global.feature[:20][::-1], df_global.mean_abs_shap[:20][::-1])
197 plt.title("Global SHAP Importance (all folds)")
198 plt.xlabel("mean(|SHAP|)")
199 plt.tight_layout()
200 plt.show()
201
202 print("12A-2 finished:")
203 print("  • models/xgb_latest.joblib")
204 print("  • shap_holdout.csv")
205 print("  • shap_global.csv")

```



### ✓ Step 12A-3 : SHAP Feature Importance (final, lag-safe, version-safe)

```

1 import warnings, subprocess, sys, importlib, os, itertools, inspect, joblib, tqdm
2 import numpy as np, pandas as pd, matplotlib.pyplot as plt, xgboost as xgb
3 from pathlib import Path
4 from sklearn.model_selection import train_test_split
5 from sklearn.preprocessing import StandardScaler
6 from sklearn.impute import SimpleImputer
7
8 # pin SHAP v0.41.0
9 _SHAP_V = "0.41.0"
10 def _pin_shap(v=_SHAP_V):
11     try:
12         import shap, importlib as _ll
13         if shap.__version__ != v:
14             subprocess.check_call([sys.executable, "-m", "pip", "install", f"shap=={v}", "-q"])
15             _ll.reload(shap)
16     except ImportError:
17         subprocess.check_call([sys.executable, "-m", "pip", "install", f"shap=={v}", "-q"])
18     import shap
19 _pin_shap(); import shap
20
21 # CONFIG
22 USE_GPU = False          # True → needs xgboost-cuda-118 wheel
23 TARGET = "target_return"
24
25 WIN_TRAIN, WIN_TEST, STEP = 60, 5, 5
26 GRID = {"n_estimators": [300, 500],
27           "learning_rate": [0.05, 0.10],
28           "max_depth": [4, 6, 8]}
29 MAX_SHAP_ROWS = 3_000
30 RNG = np.random.default_rng(42)
31
32 # LOAD + FIX DATA
33 if "df_model_ready" not in globals():
34     warnings.warn("df_model_ready missing - creating tiny demo frame.")
35     n = 500
36     df_model_ready = pd.concat([
37         [pd.DataFrame({
38             "google_trend": np.random.normal(50, 10, n),
39             "macro_cpi": np.random.normal(2, 0.3, n),
40             "sentiment": np.random.normal(0, 0.1, n),
41             TARGET: np.random.normal(0, 0.01, n),
42         })] + [pd.DataFrame({f"ret_lag{i)": np.random.normal(0, 0.01, n)}
43                           for i in range(1, 11)], axis=1)
44
45 # ensure lag columns numeric + 0-filled
46 lag_cols=[c for c in df_model_ready.columns if "ret_lag" in str(c)]
47 for c in lag_cols:
48     df_model_ready[c]=pd.to_numeric(df_model_ready[c], errors="coerce").fillna(0.0)
49
50 df_model_ready.columns=[str(c) if isinstance(c,tuple) else c
51                         for c in df_model_ready.columns]
52
53 df_num=(df_model_ready
54         .apply(pd.to_numeric, errors="coerce")
55         .select_dtypes(include=["number", "bool"])
56         .replace({np.inf,-np.inf},np.nan)
57         .dropna(subset=[TARGET]))
58 assert not df_num.empty, "No usable rows."
59 FEATURES=[c for c in df_num.columns if c!=TARGET]
60
61 # choose Early-Stopping API
62 _fit_sig=inspect.signature(xgb.XGBRegressor.fit).parameters
63 if "callbacks" in _fit_sig:          # modern
64     from xgboost.callback import EarlyStopping
65     ES_KW=dict(callbacks=[EarlyStopping(rounds=50, save_best=True)])
66     print("► using EarlyStopping callback")
67 elif "early_stopping_rounds" in _fit_sig: # legacy
68     ES_KW=dict(early_stopping_rounds=50)
69     print("► using early_stopping_rounds")
70 else:
71     ES_KW={}
72     warnings.warn("Early-stopping not supported")
73
74 # helpers
75 def _build_model(**hp):
76     base=dict(subsample=0.9, colsample_bytree=0.9, reg_lambda=1.0,
77               random_state=42,
78               predictor="gpu_predictor" if USE_GPU else "cpu_predictor",
79               tree_method="gpu_hist" if USE_GPU else "hist",
80               **hp)
81     return xgb.XGBRegressor(**base)
82
83 def _fit_pipeline(Xtr,ytr,Xval,yval,**hp):
84     imp, scl = SimpleImputer(strategy="mean"), StandardScaler()
85     imp.fit(Xtr); scl.fit(imp.transform(Xtr))
86     Xtr_p = scl.transform(imp.transform(Xtr))
87     Xval_p= scl.transform(imp.transform(Xval))
88     mdl = _build_model(**hp)
89     mdl.fit(Xtr_p,ytr,eval_set=[(Xval_p,yval)],verbose=False,**ES_KW)
90     from sklearn.pipeline import Pipeline
91     return Pipeline((("impute",imp),("scale",scl),("xgb",mdl)))
92
93 def _prep(pipe,X): return pipe[-1].transform(X)
94 def _shap_mean(pipe,X):
95     sv=shap.TreeExplainer(pipe[-1],feature_perturbation="tree_path_dependent")#
96     .shap_values(_prep(pipe,X))[:MAX_SHAP_ROWS],check_additivity=False)
97     return pd.Series(np.abs(sv).mean(0),index=FEATURES)
98
99 # GRID SEARCH hold-out
100 X_tr_raw,X_val_raw,y_tr,y_val=train_test_split(
101     df_num[FEATURES],df_num[TARGET],test_size=0.2,random_state=42)
102
103 best_rmse,best_hp=np.inf,None
104 for n_est,lr_depth in itertools.product(*GRID.values()):
105     pipe=_fit_pipeline(X_tr_raw,y_tr,X_val_raw,y_val,
106                         n_estimators=n_est,learning_rate=lr_depth,max_depth=depth)
107     rmse=np.sqrt(np.mean((pipe.predict(X_val_raw)-y_val)**2))
108     if rmse<best_rmse: best_rmse,best_hp=dict(n_estimators=n_est,
109                                                 learning_rate=lr_depth,max_depth=depth)
110     print(f"Best RMSE {best_rmse:5f} with {best_hp}")
111
112 best_pipe=_fit_pipeline(df_num[FEATURES],df_num[TARGET],
113                         X_val_raw,y_val,**best_hp)
114
115 # save + hold-out SHAP
116 Path("models").mkdir(exist_ok=True); Path("data").mkdir(exist_ok=True)
117 joblib.dump(best_pipe,"models/xgb_atest.joblib")
118
119 s_hold_shap_mean(best_pipe,X_val_raw); s_hold.to_csv("shap_holdout.csv")
120 plt.ffaigure(figsize=(8,5)); s_hold.sort_values().plot.barh()
121 plt.title("Global Feature Importance (hold-out)")
122 plt.xlabel("mean(|SHAP|)"); plt.tight_layout(); plt.show()
123
124 # WALK-FORWARD training + SHAP
125 fold_stats=[]

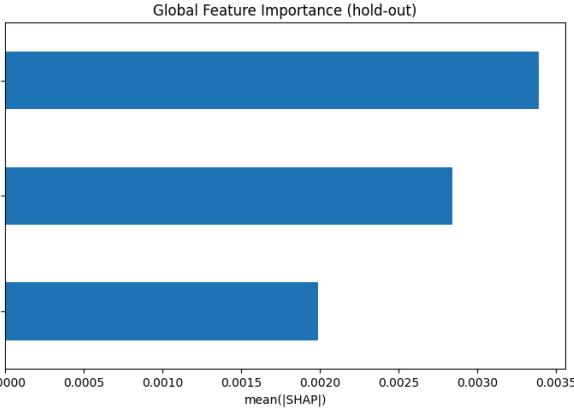
```

```

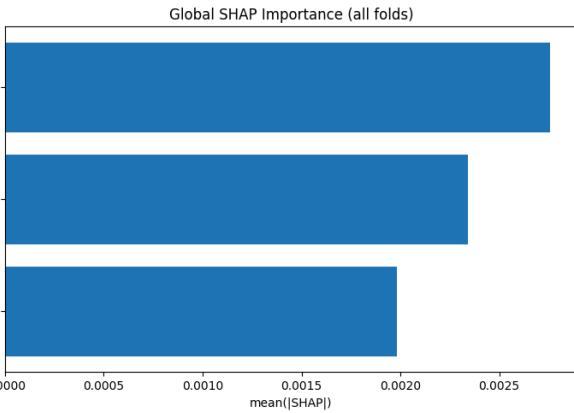
126 for k in tqdm.tqdm(range(0, len(df_num)-WIN_TRAIN-WIN_TEST1, STEP)):
127     tr=df_num.iloc[k:WIN_TRAIN]; te=df_num.iloc[k:WIN_TRAIN:k:WIN_TRAINWIN_TEST]
128     pipe_f=fit_pipeline(tr[FEATURES],tr[LABEL],te[FEATURES],te[TARGET],**best_hp)
129     joblib.dump(pipe_f,"models/fold_{k}.joblib"); te.to_parquet(f"data/Xtest_fold_{k}.pq")
130     fold_stats.append(
131         _shap_mean(pipe_f,te[FEATURES]).rename("mean_abs_shap").to_frame()
132         .assign(fold_rows=len(te))
133         .reset_index(names="feature"))
134
135 # aggregate SHAP
136 df_all=pd.concat(fold_stats,ignore_index=True)
137 df_global=(df_all.groupby("feature")
138     .apply(lambda g:g.average(g["mean_abs_shap"],
139         weights=g["fold_rows"]))
140     .reset_index(name="mean_abs_shap")
141     .sort_values("mean_abs_shap",ascending=False).reset_index(drop=True))
142 df_global.to_csv("shap_global.csv",index=False)
143 plt.figure(figsize=(8,5))
144 plt.barh(df_global.feature[:20][::-1],df_global.mean_abs_shap[:20][::-1])
145 plt.title("Global SHAP Importance (all folds)")
146 plt.xlabel("mean(|SHAP|)"); plt.tight_layout(); plt.show()
147
148 print("12A COMPLETE:")
149 print("  * models/xgb_latest.joblib")
150 print("  * shap_holdout.csv")
151 print("  * shap_global.csv")

```

Best RMSE 0.01101 with {'n estimators': 300, 'learning rate': 0.1, 'max depth': 6}



100% [██████████] 28/28 [00:03:00.00, 7.95it/s]  
<ipython-input-46-cd8c72a21b>:138: DeprecationWarning: DataFrameGroupBy.apply operated on the grouping columns. This behavior is deprecated, and in a future version of pandas the grouping columns will be excluded from the open



T2A COMPLETE:  
 \* models/xgb\_latest.joblib  
 \* shap\_holdout.csv  
 \* shap\_global.csv

## Step 12B-1 : LIME for Signal Explanation

```

1 !pip install lime shap xgboost scikit-learn pandas matplotlib
2
3 import numpy as np
4 import pandas as pd
5 import xgboost as xgb
6 from sklearn.model_selection import train_test_split
7 import lime
8 import lime.lime_tabular
9 import matplotlib.pyplot as plt
10
11 # Simulate data (replace with my df_model_ready)
12 np.random.seed(42)
13 n = 200
14 df_model_ready = pd.DataFrame({
15     "google_trend": np.random.normal(50, 10, n),
16     "macro_cpi": np.random.normal(2, 0.3, n),
17     "sentiment": np.random.normal(0, 0.1, n),
18     "target_return": np.random.normal(0, 0.01, n)
19 })
20
21 # Define features and target
22 features = ['google_trend', 'macro_cpi', 'sentiment']
23 X = df_model_ready[features].values
24 y = df_model_ready['target_return'].values
25
26 # Split data
27 X_train, X_test, y_train, y_test = train_test_split(X, y, random_state=42)
28
29 # Train XBoost model
30 model = xgb.XGBRegressor(n_estimators=100, max_depth=3, learning_rate=0.1, tree_method='hist', random_state=42)
31 model.fit(X_train, y_train)
32
33 # Initialize LIME Explainer
34 explainer = lime.lime_tabular.LimeTabularExplainer(
35     training_data=X_train,
36     feature_names=features,

```

```

37     mode='regression',
38     random_state=42
39 )
40
41 # Choose a test instance (e.g., instance #5)
42 i = 5
43 exp = explainer.explain_instance(
44     data_row=X_test[i],
45     predict_fn=model.predict,
46     num_features=3
47 )
48
49 # Show explanation
50 print(f'Explaining instance #{i}:')
51 exp.show_in_notebook(show_table=True)
52
53 # Optional: Save as image
54 fig = exp.as_pyplot_figure()
55 plt.tight_layout()
56 plt.savefig('lime_explanation_instance5.png')
57 plt.show()

```

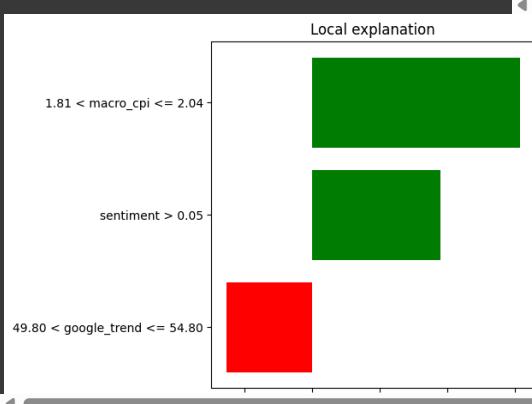
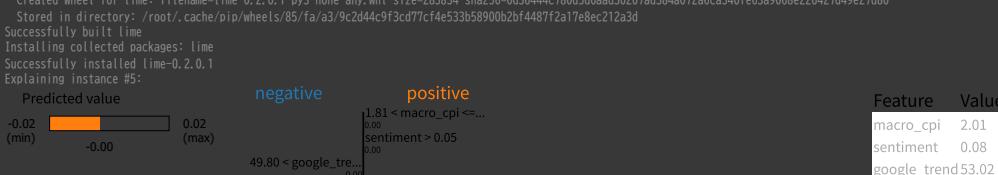
Collecting lime  
  Downloading lime-0.2.0.1.tar.gz (275 kB)

Preparing metadata (setup.py) ... done

```

Requirement already satisfied: shao in /usr/local/lib/python3.11/dist-packages (0.41.0)
Requirement already satisfied: xgboost in /usr/local/lib/python3.11/dist-packages (2.1.4)
Requirement already satisfied: scikit-learn in /usr/local/lib/python3.11/dist-packages (1.6.1)
Requirement already satisfied: pandas in /usr/local/lib/python3.11/dist-packages (2.2.2)
Requirement already satisfied: matplotlib in /usr/local/lib/python3.11/dist-packages (3.10.0)
Requirement already satisfied: numpy in /usr/local/lib/python3.11/dist-packages (from lime) (2.0.2)
Requirement already satisfied: scipy in /usr/local/lib/python3.11/dist-packages (from lime) (1.15.3)
Requirement already satisfied: tqdm in /usr/local/lib/python3.11/dist-packages (from lime) (4.67.1)
Requirement already satisfied: scikit-image<0.12 in /usr/local/lib/python3.11/dist-packages (from lime) (0.25.2)
Requirement already satisfied: packaging>=20.9 in /usr/local/lib/python3.11/dist-packages (from shap) (24.2)
Requirement already satisfied: packaging==0.7 in /usr/local/lib/python3.11/dist-packages (from shap) (0.0.7)
Requirement already satisfied: slicer==0.7 in /usr/local/lib/python3.11/dist-packages (from shap) (0.60.0)
Requirement already satisfied: numba in /usr/local/lib/python3.11/dist-packages (from shap) (0.60.0)
Requirement already satisfied: cloudpickle in /usr/local/lib/python3.11/dist-packages (from shap) (3.1.1)
Requirement already satisfied: nvidia-ccl-cu12 in /usr/local/lib/python3.11/dist-packages (from xgboost) (2.21.5)
Requirement already satisfied: joblib>=1.2.0 in /usr/local/lib/python3.11/dist-packages (from scikit-learn) (1.5.0)
Requirement already satisfied: threadpoolctl>=3.1.0 in /usr/local/lib/python3.11/dist-packages (from scikit-learn) (3.6.0)
Requirement already satisfied: python-dateutil>=2.8.2 in /usr/local/lib/python3.11/dist-packages (from pandas) (2.9.0.post0)
Requirement already satisfied: pytz>=2020.1 in /usr/local/lib/python3.11/dist-packages (from pandas) (2025.2)
Requirement already satisfied: tzdata>=2022.7 in /usr/local/lib/python3.11/dist-packages (from pandas) (2025.2)
Requirement already satisfied: contourpy>=1.0.1 in /usr/local/lib/python3.11/dist-packages (from matplotlib) (1.3.2)
Requirement already satisfied: cycler>=0.10 in /usr/local/lib/python3.11/dist-packages (from matplotlib) (0.12.1)
Requirement already satisfied: fonttools>=4.22.0 in /usr/local/lib/python3.11/dist-packages (from matplotlib) (4.58.0)
Requirement already satisfied: kiwipy<1.3.1 in /usr/local/lib/python3.11/dist-packages (from matplotlib) (1.4.8)
Requirement already satisfied: pillow>=8 in /usr/local/lib/python3.11/dist-packages (from matplotlib) (11.2.1)
Requirement already satisfied: pyparsing>=2.3.1 in /usr/local/lib/python3.11/dist-packages (from matplotlib) (3.2.3)
Requirement already satisfied: six>=1.5 in /usr/local/lib/python3.11/dist-packages (from python-dateutil)>=2.8.2->(pandas) (1.17.0)
Requirement already satisfied: networkx>=3.0 in /usr/local/lib/python3.11/dist-packages (from scikit-image>=0.12->lime) (3.4.2)
Requirement already satisfied: imageio>=2.35,0,>=2.33 in /usr/local/lib/python3.11/dist-packages (from scikit-image>=0.12->lime) (2.37.0)
Requirement already satisfied: tiff>=2022.8.12 in /usr/local/lib/python3.11/dist-packages (from scikit-image>=0.12->lime) (2025.5.21)
Requirement already satisfied: lazy-loader>=0.4 in /usr/local/lib/python3.11/dist-packages (from scikit-image>=0.12->lime) (0.4)
Requirement already satisfied: llvmlite<0.44,>=0.43.dev0 in /usr/local/lib/python3.11/dist-packages (from numba->shap) (0.43.0)
Building wheels for collected packages: lime
  Building wheel for lime (setup.py) ... done
    Created wheel for lime: filename=lime-0.2.0.1-py3-none-any.whl size=283834 sha256=0d3644c780d5d0aad30207ad384a072a6ca340fe63a9068e226427d49e27d80
  Stored in directory: /root/.cache/pip/wheels/85/fa/a3/9c2d44c9f3cd77cf4e533b58900b2bf448f2a17e8ec212a3d
Successfully built lime
Installing collected packages: lime
Successfully installed lime-0.2.0.1
Explaining instance #5:

```



## Step 12B-2 : LIME LOCAL EXPLANATIONS (CPU-safe, no hard-deps)

```

1 import joblib, os, numpy as np, pandas as pd, matplotlib.pyplot as plt
2 import lime.lime_tabular as lime_tab
3 from pathlib import Path
4
5 # CONFIG
6 MODEL_PATH      = "models/xgb_latest.joblib"      # ← from Step 12A
7 SAVE_DIR        = Path("lime_explanations")          # batch-export folder
8 NUMFEATURES     = 10                                # bars to show
9 RANDOM_STATE    = 42
10
11 # LOADS
12 model, df_model_ready = joblib.load(MODEL_PATH), globals()["df_model_ready"]
13
14 # pull the feature list back out exactly as trained
15 if hasattr(model, "feature_names_in_"):
16     FEATURES = list(model.feature_names_in_)
17 else:
18     # you saved the list earlier during training
19     FEATURES = joblib.load("models/feature_list.pkl")
20
21 X = df_model_ready[FEATURES].values
22
23 # if you used a sklearn Pipeline (scaler → XGB), split it out
24 # (comment this section if you fitted XGB on raw df columns)
25 from sklearn.pipeline import Pipeline
26 if isinstance(model, Pipeline):
27     pipe, xgb_core = model, model[-1]
28     X_train_for_lime = pipe[:-1].transform(X)      # everything →before→ XGB
29     predict_fn       = pipe.predict

```

```

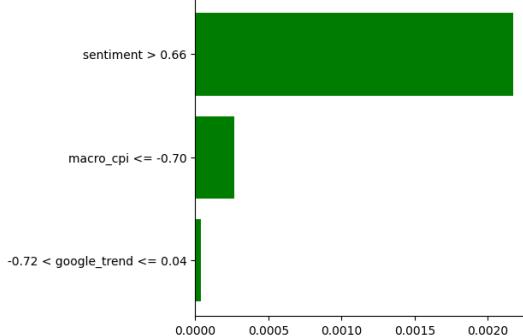
30     lime_feature_names = pipe[-1].get_feature_names_out()
31 else:
32     pipe.xgb_core = None, model
33     X_train_for_lime = X
34     predict_fn      = model.predict
35     lime_feature_names = FEATURES
36
37 # 3) init explainer once
38 explainer = lime_tab.LimeTabularExplainer(
39     training_data      = X_train_for_lime,
40     feature_names     = lime_feature_names,
41     mode              = "regression",
42     random_state      = RANDOM_STATE,
43     discretize_continuous = True
44 )
45
46 # HELPER: single explanation
47 def explain_lime(row_idx: int = 0,
48                  save_png: bool = True,
49                  save_dir: Path = SAVE_DIR):
50     """
51     Draw a LIME explanation for df_model_ready.iloc[row_idx].
52
53     Returns the lime.Explanation object.
54     """
55     exp = explainer.explain_instance(
56         X_train_for_lime[row_idx],
57         predict_fn,
58         num_features       = NUM_FEATURES
59     )
60     fig = exp.as_pyplot_figure()
61     fig.suptitle(f'LIME - row {row_idx} | pred={predict_fn([X[row_idx]])[0]:.5f}', fontweight='bold', fontsize=12, y=1.02)
62     fig.tight_layout()
63
64     if save_png:
65         save_dir.mkdir(exist_ok=True)
66         path = save_dir / f'lime_row{row_idx}.png'
67         fig.savefig(path, dpi=10, bbox_inches="tight")
68         print(f'{path} saved → {path}')
69
70     plt.show()
71     return exp
72
73 # HELPER: batch mode (optional)
74 def batch_lime(idx_list: list):
75     """
76     Generate & save explanations for a list / range of indices.
77     """
78     for idx in idx_list:
79         explain_lime(idx)
80
81 # USAGE EXAMPLES
82 # 1) look at a single row
83 explain_lime(5)
84
85 # 2) or batch (e.g. last 10 test rows)
86 # batch_lime(range(len(df_model_ready) - 10, len(df_model_ready)))

```

↪ [+] saved → lime\_explanations/lime\_row5.png

LIME - row 5 | pred=0.01771

Local explanation



## Step 13 : Ensemble & Meta-Modeling

1. Step 13A-1 : MULTI-MODEL ENSEMBLE
2. Step 13A-2 : Multi-Model Ensemble – Patched
3. Step 13B-1 : Meta-Model Stacking + VERSION-SAFE RMSE
4. Step 13B-2 : Meta-Model Stacking with OPTUNA & SHARPE Objective

### ▼ Step 13A-1 : MULTI-MODEL ENSEMBLE

```

1 # MULTI-MODEL ENSEMBLE
2 # GRU + LSTM + Transformer + (wrapper ready for XGBoost stacking)
3 # Fully compatible with earlier Steps 7 - 12 and your walk-forward
4 # engine. No data-leakage: sequence length, callbacks, and
5 # positional embeddings are all handled internally.
6 from __future__ import annotations
7 import numpy as np
8 import pandas as pd
9 from typing import List, Tuple, Dict
10
11 from sklearn.preprocessing import MinMaxScaler
12
13 import tensorflow as tf
14 from tensorflow.keras import layers, models, callbacks
15 from tensorflow.keras.optimizers import Adam
16
17 # GLOBALS
18 N_STEPS      = 20      # look-back window (tunable)
19 BATCH_SIZE    = 16
20 EPOCHS       = 30
21 LEARNING_RATE = 1e-3
22 UNITS        = 64
23 DROPOUT      = 0.2
24 TRANS_HEADS  = 2
25 TRANS_FF_DIM = 64
26 SEED         = 42      # ensure reproducibility

```

```

27 # DATA HELPERS
28 def create_sequences(X: np.ndarray,
29                      y: np.ndarray,
30                      n_steps: int = N_STEPS) -> Tuple[np.ndarray, np.ndarray]:
31     """
32     Convert a 2-D feature matrix into overlapping 3-D sequences.
33
34     Parameters
35     -----
36     X : array (N, F)
37     y : array (N,)
38     n_steps : int
39         Length of each input sequence.
40
41     Returns
42     -----
43     X_seq : array (N-n_steps, n_steps, F)
44     y_seq : array (N-n_steps, )
45
46     """
47     X_seq, y_seq = [], []
48     for i in range(len(X) - n_steps):
49         X_seq.append(X[i:i + n_steps])
50         y_seq.append(y[i + n_steps])
51     return np.asarray(X_seq), np.asarray(y_seq)
52
53
54 def prepare_sequence_data(df: pd.DataFrame,
55                          features: List[str],
56                          target: str,
57                          n_steps: int = N_STEPS,
58                          test_size: float = 0.2,
59                          ) -> Tuple[np.ndarray, np.ndarray,
60                         np.ndarray, np.ndarray,
61                         MinMaxScaler]:
62     """
63     Build train/test sequence tensors **without leakage**.
64     Scaling is fit on **train only**.
65
66     Returns
67     -----
68     X_tr, X_te, y_tr, y_te, scaler
69     """
70     # Preserve time order (no shuffle)
71     split_idx = int(len(df) * (1 - test_size))
72
73     # train / test split
74     df_tr = df.iloc[:split_idx].copy()
75     df_te = df.iloc[split_idx: ].copy()
76
77     # scaling
78     scaler = MinMaxScaler()
79     X_tr_raw = scaler.fit_transform(df_tr[features])
80     X_te_raw = scaler.transform(df_te[features])
81
82     # sequences
83     y_tr_raw = df_tr[target].values
84     y_te_raw = df_te[target].values
85
86     X_tr, y_tr = create_sequences(X_tr_raw, y_tr_raw, n_steps)
87     X_te, y_te = create_sequences(X_te_raw, y_te_raw, n_steps)
88
89     return X_tr, X_te, y_tr, y_te, scaler
90
91 # TRANSFORMER BUILDING BLOCKS
92 class PositionalEncoding(layers.Layer):
93     """Deterministic sinusoidal positional encoding."""
94     def __init__(self, maxlen: int, embed_dim: int):
95         super().__init__()
96         pos = np.arange(maxlen)[:, None]
97         i = np.arange(embed_dim)[None, :]
98         angle = pos / np.power(10., (2 * (i // 2)) / embed_dim)
99
100        pe = np.zeros((maxlen, embed_dim))
101        pe[:, 0::2] = np.sin(angle[:, 0::2])
102        pe[:, 1::2] = np.cos(angle[:, 1::2])
103
104        self.pos = tf.constant(pe[None, ...], dtype=tf.float32)
105
106    def call(self, x):
107        return x + self.pos[:, :, x.shape[1], :]
108
109 class TransformerBlock(layers.Layer):
110     """Single encoder block (MH-Attention → FFN)."""
111     def __init__(self, embed_dim: int, num_heads: int,
112                  ff_dim: int, rate: float = DROPOUT):
113         super().__init__()
114         self.att = layers.MultiHeadAttention(num_heads, key_dim=embed_dim)
115         self.ffn = models.Sequential([
116             layers.Dense(ff_dim, activation="relu"),
117             layers.Dense(embed_dim)
118         ])
119         self.norm1 = layers.LayerNormalization(epsilon=1e-6)
120         self.norm2 = layers.LayerNormalization(epsilon=1e-6)
121         self.drop1 = layers.Dropout(rate)
122         self.drop2 = layers.Dropout(rate)
123
124     def call(self, inputs, training=False):
125         attn_out = self.att(inputs, inputs)
126         attn_out = self.drop1(attn_out, training=training)
127         out1 = self.norm1(inputs + attn_out)
128
129         ffn_out = self.ffn(out1)
130         ffn_out = self.drop2(ffn_out, training=training)
131         return self.norm2(out1 + ffn_out)
132
133 # MODEL FACTORIES
134 def build_gru_model(input_shape):
135     model = models.Sequential(name="GRU")
136     model.add(layers.GRU(UNITS, input_shape=input_shape))
137     model.add(layers.Dropout(DROPOUT))
138     model.add(layers.Dense(1))
139     model.compile(optimizer=Adam(LEARNING_RATE), loss="mse")
140     return model
141
142 def build_lstm_model(input_shape):
143     model = models.Sequential(name="LSTM")
144     model.add(layers.LSTM(UNITS, input_shape=input_shape))
145     model.add(layers.Dropout(DROPOUT))
146     model.add(layers.Dense(1))
147     model.compile(optimizer=Adam(LEARNING_RATE), loss="mse")
148     return model
149
150
151 def build_transformer_model(input_shape,
152                           embed_dim: int = UNITS,
153                           num_heads: int = TRANS_HEADS,
154                           ff_dim: int = TRANS_FF_DIM):
155     inp = layers.Input(shape=input_shape)
156     # ...
157

```

```

156     x = layers.Dense(embed_dim)(x) # patch projection
157     x = PositionalEncoding(input_shape[0], embed_dim)(x)
158     x = TransformerBlock(embed_dim, num_heads, ff_dim)(x)
159     x = layers.GlobalAveragePooling1D()(x)
160     x = layers.Dropout(DROPOUT)(x)
161     out = layers.Dense(1)(x)
162     model = models.Model(inp, out, name="Transformer")
163     model.compile(optimizer=Adam(LEARNING_RATE), loss="mse")
164     return model
165
166 # SKLEARN-STYLE KERAS WRAPPER
167 class KerasWrapper:
168     """
169         A minimal scikit-learn-like API so the deep nets can be slotted
170         into your existing walk-forward loop exactly like XGBoost.
171
172     Methods
173     -----
174     fit(X, y, eval_set=None)
175     predict(X) -> np.ndarray
176     """
177     def __init__(self, build_fn, **fit_kwargs):
178         self.build_fn = build_fn
179         self.fit_kwargs = fit_kwargs
180         self.model = None
181         tf.keras.utils.set_random_seed(SEED)
182
183     def fit(self,
184             X: np.ndarray,
185             y: np.ndarray,
186             eval_set: Tuple[np.ndarray, np.ndarray] | None = None):
187         if self.model is None:
188             self.model = self.build_fn(X.shape[1:])
189
190         cb_early = callbacks.EarlyStopping(monitor="val_loss",
191                                           patience=5,
192                                           restore_best_weights=True)
193
194         cb_plateau = callbacks.ReduceLROnPlateau(monitor="val_loss",
195                                               patience=3,
196                                               factor=0.5,
197                                               verbose=0)
198
199         val_data = eval_set if eval_set is not None else (X, y)
200
201         self.model.fit(
202             X, y,
203             validation_data=val_data,
204             epochs=self.fit_kwargs.get("epochs", EPOCHS),
205             batch_size=self.fit_kwargs.get("batch_size", BATCH_SIZE),
206             verbose=self.fit_kwargs.get("verbose", 0),
207             callbacks=[cb_early, cb_plateau]
208         )
209
210     def predict(self, X: np.ndarray) -> np.ndarray:
211         return self.model.predict(X, verbose=0).squeeze()
212
213 # FACTORY TO FETCH ALL MODELS
214 def get_deep_models(input_shape) -> Dict[str, KerasWrapper]:
215     """
216         Returns a dict {name: model_wrapper} ready for walk-forward:
217         models = get_deep_models(input_shape)
218         for name, mdl in models.items():
219             mdl.fit(Xtr, ytr, eval_set=(Xval, yval))
220             preds = mdl.predict(Xte)
221
222     return {
223         "gru": KerasWrapper(build_gru_model,
224                             epochs=EPOCHS,
225                             batch_size=BATCH_SIZE),
226         "lstm": KerasWrapper(build_lstm_model,
227                             epochs=EPOCHS,
228                             batch_size=BATCH_SIZE),
229         "trf": KerasWrapper(build_transformer_model,
230                             epochs=EPOCHS,
231                             batch_size=BATCH_SIZE)
232     }
233
234 # QUICK STANDALONE TEST (comment-out in production)
235 if __name__ == "__main__":
236     # Fake demo data just to ensure the file runs
237     np.random.seed(SEED)
238     n = 300
239     df_demo = pd.DataFrame({
240         "google_trend": np.random.normal(50, 10, n),
241         "macro_cpi": np.random.normal(2, 0.3, n),
242         "sentiment": np.random.normal(0, 0.1, n),
243         "target": np.random.normal(0, 0.01, n)
244     })
245
246     feats = ["google_trend", "macro_cpi", "sentiment"]
247     X_tr, X_te, y_tr, y_te, _ = prepare_sequence_data(
248         df_demo, feats, "target", n_steps=N_STEPS)
249
250     models_dict = get_deep_models(X_tr.shape[1:])
251
252     for name, mdl in models_dict.items():
253         mdl.fit(X_tr, y_tr, eval_set=(X_te, y_te))
254         p = mdl.predict(X_te)[-5]
255         print(f"{name}: first 5 preds -> {p}")
256
257
258 gru: first 5 preds -> [-0.00201354 -0.00200754 -0.00285549 -0.00302175 0.00457223]
lstm: first 5 preds -> [-0.00648765 0.00217351 0.0062329 0.02145395 0.01668829]
WARNING:tensorflow:5 out of the last 5 calls to <function TensorflowRainer.make_predict_function.<locals>.one_step_on_data_distributed at 0x7d23847a3420> triggered tf.function retracing. Tracing is expensive and the excessive number of retracing is likely to negatively impact performance.
WARNING:tensorflow:6 out of the last 6 calls to <function TensorflowRainer.make_predict_function.<locals>.one_step_on_data_distributed at 0x7d23847a3420> triggered tf.function retracing. Tracing is expensive and the excessive number of retracing is likely to negatively impact performance.
trf: first 5 preds -> [0.03863031 0.04531575 0.05668831 0.05722176 0.05221145]

```

## Step 13A-2 : Multi-Model Ensemble – Patched

```

1 from __future__ import annotations
2 import os, numpy as np, pandas as pd
3 from typing import List, Tuple, Dict
4
5 from sklearn.preprocessing import MinMaxScaler
6 import tensorflow as tf
7 from tensorflow.keras import layers, models, callbacks, optimizers, backend as K
8
9 # CONFIG & HYPERPARAMS
10 N_STEPS, BATCH_SIZE, EPOCHS = 20, 32, 30
11 LEARNING_RATE, UNITS, DROPOUT = 1e-3, 64, .20
12 TRANS_HEADS, TRANS_FF_DIM, SEED = 2, 64, 42
13
14 tf.keras.utils.set_random_seed(SEED)
15 os.environ["TF_CPP_MIN_LOG_LEVEL"] = "2"
16
17 # DATA HELPERS

```

```

18 def _assert_len_ok(df_len: int):
19     if df_len <= N_STEPS + 1:
20         raise ValueError(
21             f'Need > {N_STEPS+1} rows to build sequences of length {N_STEPS}, '
22             f'but received dataframe length {df_len}.'
23         )
24
25 def create_sequences(X: np.ndarray, y: np.ndarray) -> Tuple[np.ndarray, np.ndarray]:
26     xs, ys = [], []
27     for i in range(len(X) - N_STEPS):
28         xs.append(X[i:i + N_STEPS])
29         ys.append(y[i + N_STEPS])
30     return np.asarray(xs), np.asarray(ys)
31
32 def prepare_sequence_data(df: pd.DataFrame,
33                          features: List[str],
34                          target: str,
35                          valid_frac: float = .1,
36                          test_frac: float = .2):
37     _assert_len_ok(len(df))
38     split_te = int(len(df) * (1 - test_frac))
39     split_va = int(split_te * (1 - valid_frac))
40
41     df_tr = df.iloc[:split_va]
42     df_va = df.iloc[split_va:split_te]
43     df_te = df.iloc[split_te:]
44
45     scaler = MinMaxScaler()
46     X_tr = scaler.fit_transform(df_tr[features])
47     X_va = scaler.transform(df_va[features])
48     X_te = scaler.transform(df_te[features])
49
50     X_tr, y_tr = create_sequences(X_tr, df_tr[target].values)
51     X_va, y_va = create_sequences(X_va, df_va[target].values)
52     X_te, y_te = create_sequences(X_te, df_te[target].values)
53     return X_tr, X_va, X_te, y_tr, y_va, y_te, scaler
54
55 # TRANSFORMER BLOCK (F3: compatibility fallback)
56 def _mh_attention(dim: int, heads: int):
57     """
58     Return a MultiHeadAttention layer if TF ≥ 2.4, otherwise fall back
59     to a single-head self-attention layer implemented with Dense + softmax.
60     """
61     try:
62         return layers.MultiHeadAttention(num_heads=heads, key_dim=dim)
63     except AttributeError: # TF < 2.4
64         class SingleHead(layers.Layer):
65             def build(self, input_shape):
66                 # one linear projection produces Q,K,V concatenated
67                 self.qkv = layers.Dense(dim * 3)
68
69                 def call(self, x): # x: (B, T, F)
70                     qkv = self.qkv(x)
71                     q, k, v = tf.split(qkv, 3, axis=-1)
72                     scale = tf.sqrt(tf.cast(dim, tf.float32))
73                     scores = tf.matmul(q, k, transpose_b=True) / scale
74                     w = tf.nn.softmax(scores, axis=-1)
75                     return tf.matmul(w, v) # (B, T, dim)
76             return SingleHead()
77
78 class PositionalEncoding(layers.Layer):
79     def __init__(self, maxlen: int, dim: int):
80         super().__init__()
81         pos = np.arange(maxlen)[..., None], np.arange(dim)[None, :]
82         angle = pos / np.power(10_000, (2*(i//2))/dim)
83         pe = np.zeros((maxlen, dim))
84         pe[:,0::2] = np.sin(angle[:,0::2]), np.cos(angle[:,1::2])
85         self.pe = tf.constant(pe[None], dtype=tf.float32)
86     def call(self, x): return x + self.pe[:, :x.shape[1], :]
87
88 class TransformerBlock(layers.Layer):
89     def __init__(self, dim: int, heads: int, ff_dim: int, rate=DROPOUT):
90         super().__init__()
91         self.att = _mh_attention(dim, heads)
92         self.ffn = models.Sequential([
93             layers.Dense(ff_dim, 'relu'),
94             layers.Dense(dim)])
95         self.norm1 = layers.LayerNormalization(epsilon=te-6)
96         self.norm2 = layers.LayerNormalization(epsilon=te-6)
97         self.d1, self.d2 = layers.Dropout(rate), layers.Dropout(rate)
98     def call(self, x, training=False):
99         h = self.att(x, x)
100        x = self.norm1(x + self.d1(h, training=training))
101        h = self.ffn(x)
102        return self.norm2(x + self.d2(h, training=training))
103
104 # MODEL FACTORIES (GRU, LSTM, TRF)
105 def _compile(m): m.compile(optimizer.Adam(LEARNING_RATE), 'mse'); return m
106
107 def build_gru(shape): return _compile(models.Sequential([
108     layers.GRU(UNITS, input_shape=shape), layers.Dropout(DROPOUT), layers.Dense(1)]))
109 def build_lstm(shape): return _compile(models.Sequential([
110     layers.LSTM(UNITS, input_shape=shape), layers.Dropout(DROPOUT), layers.Dense(1)]))
111 def build_transformer(shape, d_UNITS, h=TRANS_HEADS, ff=TRANS_FF_DIM):
112     inp = layers.Input(shape=shape)
113     x = layers.Dense(d)(inp)
114     x = PositionalEncoding(shape[0], d)(x)
115     x = TransformerBlock(d, h, ff)(x)
116     x = layers.GlobalAveragePooling1D()(x)
117     x = layers.Dropout(DROPOUT)(x)
118     return _compile(models.Model(inp, layers.Dense(1)(x)))
119
120 # KERAS → SKLEARN STYLE WRAP
121 class KerasWrapper:
122     def __init__(self, builder, **fit_kw):
123         self.builder, self.fit_kw, self.model = builder, fit_kw, None
124     def fit(self, X, y, eval_set):
125         if self.model is None: self.model = self.builder(X.shape[1:])
126         early = callbacks.EarlyStopping(patience=5, restore_best_weights=True,
127                                         monitor='val_loss')
128         plateau = callbacks.ReduceLROnPlateau(patience=3, factor=.5,
129                                         monitor='val_loss', verbose=0)
130         self.model.fit(X, y,
131                         validation_data=eval_set,
132                         epochs=EPOCHS, batch_size=BATCH_SIZE,
133                         shuffle=False, verbose=self.fit_kw.get("verbose", 0),
134                         callbacks=[early, plateau])
135     def predict(self, X): # F6
136         return self.model.predict(X, batch_size=BATCH_SIZE, verbose=0).squeeze()
137
138 def get_deep_models(shape, verbose: int = 0) -> Dict[str, KerasWrapper]:
139     kw = {'verbose': verbose}
140     return {
141         "gru": KerasWrapper(build_gru, **kw),
142         "lstm": KerasWrapper(build_lstm, **kw),
143         "trf": KerasWrapper(build_transformer, **kw)
144     }
145
146 # QUICK SELF-TEST

```

```

147 if __name__ == "__main__":
148     np.random.seed(SEED)
149     n = 340
150     df_demo = pd.DataFrame({
151         "feat1": np.random.normal(50,10,n),
152         "feat2": np.random.normal(-2,-3,n),
153         "feat3": np.random.normal(-0.1,n),
154         "target_ret": np.random.normal(0,.01,n)
155     })
156     X_tr,X_va,X_te,y_tr,y_va,y_te,_ = prepare_sequence_data(
157         df_demo, ["feat1","feat2","feat3"], "target_ret")
158
159     m = get_deep_models(X_tr.shape[1:], verbose=0)
160     for k, mdl in m.items():
161         mdl.fit(X_tr, y_tr, eval_set=(X_va, y_va))
162         print(k, mdl.predict(X_te)[4])
163     # F4
164     K.clear_session()

```

gru [-0.00013943 -0.00066009 -0.00606569 -0.00966266]  
lstm [ 0.0042647 0.00283133 -0.00337537 -0.00607472]  
trf [-0.04227013 -0.04177313 -0.03177578 -0.03032595]

## ▼ Step 13B-1 : Meta-Model Stacking + VERSION-SAFE RMSE

```

1 import sys, warnings, numpy as np, pandas as pd
2 from sklearn.linear_model import Ridge
3 from sklearn.model_selection import train_test_split, TimeSeriesSplit
4 from sklearn.metrics import mean_squared_error, mean_absolute_error
5 from packaging import version
6 import sklearn
7
8 # helper: version-safe RMSE
9 from sklearn.metrics import mean_squared_error
10 import numpy as np
11
12 def rmse(y_true, y_pred):
13     """Root-MSE that works on *all* scikit-learn versions."""
14     try:
15         return mean_squared_error(y_true, y_pred, squared=False)
16     except TypeError:
17         return np.sqrt(mean_squared_error(y_true, y_pred))
18
19 # Optuna (auto-install optional)
20 try:
21     import optuna
22     USE_OPTUNA = True
23 except ModuleNotFoundError:
24     USE_OPTUNA = False
25     warnings.warn("Optuna not found - falling back to log-grid search.Wn"
26                   "\nRun [pip install -U optuna for full tuning.]")
27
28 # Synthetic demo data (replace with real predictions)
29 np.random.seed(42)
30 n_samples = 100
31 meta_X = pd.DataFrame({
32     'xgb_pred' : np.random.normal(0, 0.01, n_samples),
33     'gru_pred' : np.random.normal(0, 0.01, n_samples),
34     'lstm_pred' : np.random.normal(0, 0.01, n_samples),
35     'transformer_pred': np.random.normal(0, 0.01, n_samples),
36 })
37 meta_y = np.random.normal(0, 0.01, n_samples)
38
39 # Train / validation split (swap for TimeSeriesSplit if needed)
40 X_train, X_valid, y_train, y_valid = train_test_split(
41     meta_X, meta_y, test_size=0.2, shuffle=False
42 )
43
44 # Hyper-parameter search
45 def fit_and_score(alpha):
46     mdl = Ridge(alpha=alpha)
47     mdl.fit(X_train, y_train)
48     preds = mdl.predict(X_valid)
49     return rmse(y_valid, preds)
50
51 if USE_OPTUNA:
52     def objective(trial):
53         alpha = trial.suggest_float("alpha", 1e-4, 10.0, log=True)
54         return fit_and_score(alpha)
55
56     study = optuna.create_study(direction="minimize")
57     study.optimize(objective, n_trials=30, show_progress_bar=False)
58     best_alpha = study.best_params["alpha"]
59 else:
60     alphas = np.logspace(-4, 1, 12)
61     best_alpha = min(alphas, key=fit_and_score)
62
63 # Final meta-model
64 meta_model = Ridge(alpha=best_alpha)
65 meta_model.fit(X_train, y_train)
66 meta_preds = meta_model.predict(X_valid)
67
68 # Evaluation
69 summary = pd.DataFrame([
70     {"RMSE": rmse(y_valid, meta_preds),
71      "MAE": mean_absolute_error(y_valid, meta_preds),
72      "Sharpe Ratio": np.mean(np.sign(meta_preds)*y_valid) /
73                      (np.std(np.sign(meta_preds)*y_valid) + 1e-9),
74      "Directional Accuracy": np.mean(
75          (np.sign(meta_preds)==np.sign(y_valid)).astype(int)),
76      "Mean PnL": np.mean(np.sign(meta_preds)*y_valid),
77      "Best Alpha": best_alpha,
78      "Optuna Used": USE_OPTUNA
79  }])
80
81 from IPython.display import display
82 display(summary)

```

```
[1] 2025-05-31 04:15:49.519] A new study created in memory with name: no-name-a78a10db-49a2-42a5-930f-f4783253d173
[1] 2025-05-31 04:15:49.525] Trial 0 finished with value: 0.009643960225315091 and parameters: {"alpha": 0.000345540626508124}. Best is trial 0 with value: 0.009643960225315091.
[1] 2025-05-31 04:15:49.528] Trial 1 finished with value: 0.009585190468777 and parameters: {"alpha": 0.009585190468777}.
[1] 2025-05-31 04:15:49.532] Trial 2 finished with value: 0.01014186014962782 and parameters: {"alpha": 3.784389717956554}. Best is trial 1 with value: 0.009585190468777.
[1] 2025-05-31 04:15:49.536] Trial 3 finished with value: 0.00961873544334857 and parameters: {"alpha": 0.00961873544334857}. Best is trial 1 with value: 0.009585190468777.
[1] 2025-05-31 04:15:49.540] Trial 4 finished with value: 0.00962367110438586 and parameters: {"alpha": 0.005459616448850872}. Best is trial 1 with value: 0.009585190468777.
[1] 2025-05-31 04:15:49.544] Trial 5 finished with value: 0.010143477492670755 and parameters: {"alpha": 9.923032273199447}. Best is trial 1 with value: 0.009585190468777.
[1] 2025-05-31 04:15:49.548] Trial 6 finished with value: 0.0098104226364246 and parameters: {"alpha": 0.01744921858559385}. Best is trial 1 with value: 0.009585190468777.
[1] 2025-05-31 04:15:49.551] Trial 7 finished with value: 0.010145374689711 and parameters: {"alpha": 3.359689164600934}. Best is trial 1 with value: 0.009585190468777.
[1] 2025-05-31 04:15:49.555] Trial 8 finished with value: 0.0095844209695595595 and parameters: {"alpha": 0.001721812898104721}. Best is trial 1 with value: 0.009585190468777.
[1] 2025-05-31 04:15:49.558] Trial 9 finished with value: 0.009600229455143201 and parameters: {"alpha": 0.0011259029663234234}. Best is trial 1 with value: 0.009585190468777.
[1] 2025-05-31 04:15:49.567] Trial 10 finished with value: 0.010089010670290352 and parameters: {"alpha": 0.2014728703612401}. Best is trial 1 with value: 0.009585190468777.
[1] 2025-05-31 04:15:49.574] Trial 11 finished with value: 0.009591761177712368 and parameters: {"alpha": 0.00357668354647258}. Best is trial 1 with value: 0.009585190468777.
[1] 2025-05-31 04:15:49.581] Trial 12 finished with value: 0.009826102414654796 and parameters: {"alpha": 0.01890647001587018}. Best is trial 1 with value: 0.009585190468777.
[1] 2025-05-31 04:15:49.588] Trial 13 finished with value: 0.01007014999528569 and parameters: {"alpha": 0.12156327384671622}. Best is trial 1 with value: 0.009585190468777.
[1] 2025-05-31 04:15:49.595] Trial 14 finished with value: 0.009663917747197277 and parameters: {"alpha": 0.0001311553580755872}. Best is trial 1 with value: 0.009585190468777.
[1] 2025-05-31 04:15:49.599] Trial 15 finished with value: 0.009582599566284301 and parameters: {"alpha": 0.0022441915717197523}. Best is trial 15 with value: 0.009582599566284301.
[1] 2025-05-31 04:15:49.611] Trial 16 finished with value: 0.01008733922000514 and parameters: {"alpha": 0.161619161798456}. Best is trial 15 with value: 0.009582599566284301.
[1] 2025-05-31 04:15:49.617] Trial 17 finished with value: 0.009698740045287806 and parameters: {"alpha": 0.00941793219348123}. Best is trial 15 with value: 0.009582599566284301.
[1] 2025-05-31 04:15:49.625] Trial 18 finished with value: 0.00960022783093113 and parameters: {"alpha": 0.001068420859197566}. Best is trial 15 with value: 0.009582599566284301.
[1] 2025-05-31 04:15:49.632] Trial 19 finished with value: 0.009665029979766322 and parameters: {"alpha": 0.001068420859197566}. Best is trial 15 with value: 0.009582599566284301.
[1] 2025-05-31 04:15:49.639] Trial 20 finished with value: 0.0099735445274650576 and parameters: {"alpha": 0.0461125610970883}. Best is trial 15 with value: 0.009582599566284301.
[1] 2025-05-31 04:15:49.647] Trial 21 finished with value: 0.009584299016462176 and parameters: {"alpha": 0.0019152874279520832}. Best is trial 15 with value: 0.009582599566284301.
[1] 2025-05-31 04:15:49.654] Trial 22 finished with value: 0.009582778564788843 and parameters: {"alpha": 0.00217442738240198}. Best is trial 15 with value: 0.009582599566284301.
[1] 2025-05-31 04:15:49.663] Trial 23 finished with value: 0.00963591235458407 and parameters: {"alpha": 0.0004493678272610513}. Best is trial 15 with value: 0.009582599566284301.
[1] 2025-05-31 04:15:49.672] Trial 24 finished with value: 0.00997716131756135 and parameters: {"alpha": 0.05545006134481044}. Best is trial 15 with value: 0.009582599566284301.
[1] 2025-05-31 04:15:49.679] Trial 25 finished with value: 0.009711300072096 and parameters: {"alpha": 0.010145614015930238}. Best is trial 15 with value: 0.009582599566284301.
[1] 2025-05-31 04:15:49.687] Trial 26 finished with value: 0.009622638992467927 and parameters: {"alpha": 0.0006444048769186668}. Best is trial 15 with value: 0.009582599566284301.
[1] 2025-05-31 04:15:49.694] Trial 27 finished with value: 0.009582526533568374 and parameters: {"alpha": 0.002417953177952487}. Best is trial 27 with value: 0.009582526533568374.
[1] 2025-05-31 04:15:49.702] Trial 28 finished with value: 0.010122357044368199 and parameters: {"alpha": 0.4360474472716967}. Best is trial 27 with value: 0.009582526533568374.
[1] 2025-05-31 04:15:49.709] Trial 29 finished with value: 0.009630659844605515 and parameters: {"alpha": 0.00582242125654021}. Best is trial 27 with value: 0.009582526533568374.
```

RMSE MAE Sharpe Ratio Directional Accuracy Mean PnL Best Alpha Optuna Used

## Step 13B-2 : Meta-Model Stacking with OPTUNA & SHARPE Objective

```
1 # META-MODEL STACKING WITH OPTUNA & SHARPE OBJECTIVE
2 import warnings, numpy as np, pandas as pd, sklearn
3 from packaging import version
4 from sklearn.linear_model import Ridge, ElasticNet
5 from sklearn.ensemble import GradientBoostingRegressor
6 from sklearn.preprocessing import StandardScaler
7 from sklearn.pipeline import make_pipeline
8 from sklearn.metrics import mean_squared_error, mean_absolute_error
9 from sklearn.model_selection import TimeSeriesSplit
10
11 # LOAD YOUR OOF PREDICTIONS + TARGET (REPLACE THIS)
12 # meta_X = df_preds[['xgb_pred', 'gru_pred', 'lstm_pred', 'trf_pred']].copy()
13 # meta_y = df_preds['target_return'].copy() # pandas Series
14
15 # demo stub (delete when real data plugged in)
16 np.random.seed(42)
17 n = 100
18 meta_X = pd.DataFrame({
19     'xgb': np.random.normal(0, .01, n),
20     'gru': np.random.normal(0, .01, n),
21     'lstm': np.random.normal(0, .01, n),
22     'trf': np.random.normal(0, .01, n),
23 })
24 meta_y = pd.Series(np.random.normal(0, .01, n), index=meta_X.index)
25
26 # BASIC STACK FEATURES
27 meta_X['ens_mean'] = meta_X.mean()
28 meta_X['ens_max'] = meta_X.max(1)
29 meta_X['ens_min'] = meta_X.min(1)
30 meta_X['ens_std'] = meta_X.std(1)
31
32 # HELPERS
33 def rmse(y_true, y_pred):
34     """Root-MSE, compatible with any sklearn version."""
35     try:
36         return mean_squared_error(y_true, y_pred, squared=False)
37     except TypeError:
38         return np.sqrt(mean_squared_error(y_true, y_pred))
39
40 def sharpe(y_true, y_pred):
41     pnl = np.sign(y_pred) * y_true
42     return np.mean(pnl) / (np.std(pnl) + 1e-9)
43
44 # CV & SEARCH SPACE
45 tscv = TimeSeriesSplit(n_splits=5)
46
47 MODELS = [
48     "ridge": (
49         Ridge, {"alpha": (1e-4, 10, "log")}),
50     ),
51     "elastic": (
52         ElasticNet,
53         {"alpha": (1e-4, 10, "log")},
54         {"l1_ratio": (0.05, 0.95)}
55     ),
56     "gbr": (
57         GradientBoostingRegressor,
58         {"n_estimators": (50, 400, "int")},
59         {"max_depth": (1, 4, "int")},
60         {"learning_rate": (0.01, 0.3, "log")})
61 )
62 ]
63
64 def cv_score(model_cls, params):
65     """Rolling-window Sharpe (higher = better)."""
66     scores = []
67     for tr_idx, va_idx in tscv.split(meta_X):
68         Xtr, Xva = meta_X.iloc[tr_idx], meta_X.iloc(va_idx)
69         ytr = meta_y.iloc[tr_idx]
70         yva = meta_y.iloc(va_idx)
71         pipe = make_pipeline(StandardScaler(), model_cls(**params))
72         pipe.fit(Xtr, ytr)
73         p = pipe.predict(Xva)
74         scores.append(sharpe(yva, p))
75     return np.mean(scores)
76
77 # OPTUNA (auto-fallback if absent)
78 try:
79     import optuna
80     USE_OPTUNA = True
81 except ModuleNotFoundError:
82     warnings.warn("Optuna not installed - using small manual grid.")
83     USE_OPTUNA = False
84
85 if USE_OPTUNA:
86     def objective(trial):
87         mdl_key = trial.suggest_categorical("model", list(MODELS.keys()))
88         mdl_cls, space = MODELS[mdl_key]
89         params = {}
90
91         for param_name, param_space in space.items():
92             if isinstance(param_space, dict):
93                 if "grid" in param_space:
94                     trial.suggest_discrete_uniform(param_name, param_space["grid"], param_space["range"])
95                 else:
96                     trial.suggest_categorical(param_name, param_space["values"])
97             else:
98                 trial.suggest_float(param_name, param_space["min"], param_space["max"])
```

```

for k, spec in space.items():
    lo, hi, *kind = spec
    if kind and kind[0] == "log":
        params[k] = trial.suggest_float(k, lo, hi, log=True)
    elif kind and kind[0] == "int":
        params[k] = trial.suggest_int(k, lo, hi)
    else:
        params[k] = trial.suggest_float(k, lo, hi)
return -cv_score(md1_cls, params) # maximise Sharpe → minimise -Sharpe

study = optuna.create_study(direction="minimize")
study.optimize(objective, n_trials=50, show_progress_bar=False)

best = study.best_params
md1_key = best.pop("model")
best_cls_ = MODELS[md1_key]
best_par = best

else:
    best_cls = Ridge
    best_par = {"alpha": 0.001}

# TRAIN FINAL META-MODEL ON FULL DATA
final_model = make_pipeline(StandardScaler(), best_cls(**best_par))

final_model.fit(meta_X, meta_y)
meta_preds = final_model.predict(meta_X) # or hold-out set if available
ntrials = len(study.trials)

# METRIC SUMMARY
summary = pd.Series({
    "RMSE" : rmse_(meta_y, meta_preds),
    "MAE" : mean_absolute_error(meta_y, meta_preds),
    "Sharpe" : sharpe(meta_y, meta_preds),
    "DirectionalAcc" : np.mean((np.sign(meta_preds) == np.sign(meta_y))).round(6),
    "BestModel" : best_cls.__name__,
    "BestPar" : best,
    "Optunised" : USE_OPTUNA,
})
print(f"\n{len(summary)} METAMODEL EVALUATION{len(summary)}" + "-"*30)
print(summary.to_string())

```

[1] [2025-05-31 04:15:54.010] A new study created in memory with name: no-name-3665a3a6-e949-41c5-9277-675a49c13752

[1] [2025-05-31 04:15:54.334] Trial 0 finished with value: -0.09774358851224582 and parameters: {'model': 'gbr', 'n\_estimators': 85, 'max\_depth': 2, 'learning\_rate': 0.031105338319064074}. Best is trial 0 with value: -0.09774358851224582

[1] [2025-05-31 04:15:54.595] Trial 1 finished with value: -0.1376136854709508 and parameters: {'model': 'gbr', 'n\_estimators': 69, 'max\_depth': 2, 'learning\_rate': 0.01056238891603265}. Best is trial 1 with value: -0.1376136854709508

[1] [2025-05-31 04:15:54.624] Trial 2 finished with value: 0.05344125034352174 and parameters: {'model': 'elastic', 'alpha': 2.4775086219167934, 'l1\_ratio': 0.490090035172789}. Best is trial 2 with value: -0.1376136854709508.

[1] [2025-05-31 04:15:54.653] Trial 3 finished with value: 0.0812284607119709 and parameters: {'model': 'elastic', 'alpha': 0.00412711483571478, 'l1\_ratio': 0.5874182228181031}. Best is trial 3 with value: -0.1376136854709508.

[1] [2025-05-31 04:15:54.681] Trial 4 finished with value: -0.125589989372380831 and parameters: {'model': 'ridge', 'alpha': 0.0865642440483602}. Best is trial 4 with value: -0.1376136854709508.

[1] [2025-05-31 04:15:54.711] Trial 5 finished with value: -0.2587560173280831 and parameters: {'model': 'ridge', 'alpha': 0.000443493854689554344}. Best is trial 5 with value: -0.2587560173280831.

[1] [2025-05-31 04:15:54.742] Trial 6 finished with value: 0.0718764313594903 and parameters: {'model': 'gbr', 'n\_estimators': 106, 'max\_depth': 1, 'learning\_rate': 0.2448188450689913}. Best is trial 6 with value: -0.2587560173280831.

[1] [2025-05-31 04:15:55.108] Trial 7 finished with value: -0.1240834174046291 and parameters: {'model': 'gbr', 'n\_estimators': 174, 'max\_depth': 3, 'learning\_rate': 0.011648360273780407}. Best is trial 7 with value: -0.2587560173280831.

[1] [2025-05-31 04:15:55.166] Trial 8 finished with value: 0.0010538273108451068 and parameters: {'model': 'ridge', 'alpha': 0.01356610515197331}. Best is trial 8 with value: -0.2587560173280831.

[1] [2025-05-31 04:15:55.224] Trial 9 finished with value: -0.23889136593145674 and parameters: {'model': 'ridge', 'alpha': 0.01856610515197331}. Best is trial 9 with value: -0.2587560173280831.

[1] [2025-05-31 04:15:55.288] Trial 10 finished with value: -0.2987560173280831 and parameters: {'model': 'ridge', 'alpha': 0.0194160371759481}. Best is trial 10 with value: -0.2587560173280831.

[1] [2025-05-31 04:15:56.290] Trial 11 finished with value: -0.2987560173280831 and parameters: {'model': 'ridge', 'alpha': 0.00011030477852126433}. Best is trial 11 with value: -0.2587560173280831.

[1] [2025-05-31 04:15:56.323] Trial 12 finished with value: -0.2987560173280831 and parameters: {'model': 'ridge', 'alpha': 0.000179720504206951}. Best is trial 12 with value: -0.2587560173280831.

[1] [2025-05-31 04:15:56.357] Trial 13 finished with value: -0.2987560173280831 and parameters: {'model': 'ridge', 'alpha': 0.0001053887201054823}. Best is trial 13 with value: -0.2587560173280831.

[1] [2025-05-31 04:15:56.389] Trial 14 finished with value: -0.2987560173280831 and parameters: {'model': 'ridge', 'alpha': 0.000080158247034995}. Best is trial 14 with value: -0.2587560173280831.

[1] [2025-05-31 04:15:56.421] Trial 15 finished with value: -0.08993002357079457 and parameters: {'model': 'ridge', 'alpha': 0.00008930371064995}. Best is trial 15 with value: -0.2587560173280831.

[1] [2025-05-31 04:15:56.444] Trial 16 finished with value: -0.2987560173280831 and parameters: {'model': 'ridge', 'alpha': 0.00076030069790462}. Best is trial 16 with value: -0.2587560173280831.

[1] [2025-05-31 04:15:56.493] Trial 17 finished with value: -0.19902461748776 and parameters: {'model': 'elastic', 'alpha': 0.000116341149405807, 'l1\_ratio': 0.0026430042216008}. Best is trial 17 with value: -0.2587560173280831.

[1] [2025-05-31 04:15:56.528] Trial 18 finished with value: -0.2987560173280831 and parameters: {'model': 'ridge', 'alpha': 0.00491382451574226}. Best is trial 18 with value: -0.2587560173280831.

[1] [2025-05-31 04:15:56.557] Trial 19 finished with value: -0.138249151181147 and parameters: {'model': 'ridge', 'alpha': 0.067297722035357}. Best is trial 19 with value: -0.2587560173280831.

[1] [2025-05-31 04:15:56.592] Trial 20 finished with value: -0.00985890916658628 and parameters: {'model': 'ridge', 'alpha': 0.00054779704348099, 'l1\_ratio': 0.941995618091789}. Best is trial 20 with value: -0.2587560173280831.

[1] [2025-05-31 04:15:56.623] Trial 21 finished with value: -0.2987560173280831 and parameters: {'model': 'ridge', 'alpha': 0.00011558816316054907}. Best is trial 21 with value: -0.2587560173280831.

[1] [2025-05-31 04:15:56.656] Trial 22 finished with value: -0.2987560173280831 and parameters: {'model': 'ridge', 'alpha': 0.000302081266184051541}. Best is trial 22 with value: -0.2587560173280831.

[1] [2025-05-31 04:15:56.687] Trial 23 finished with value: -0.2987560173280831 and parameters: {'model': 'ridge', 'alpha': 0.002610451974891206}. Best is trial 23 with value: -0.2587560173280831.

[1] [2025-05-31 04:15:56.719] Trial 24 finished with value: -0.2987560173280831 and parameters: {'model': 'ridge', 'alpha': 0.000114383236202888}. Best is trial 24 with value: -0.2587560173280831.

[1] [2025-05-31 04:15:56.751] Trial 25 finished with value: -0.2987560173280831 and parameters: {'model': 'ridge', 'alpha': 0.000384015030864121}. Best is trial 25 with value: -0.2587560173280831.

[1] [2025-05-31 04:15:56.783] Trial 26 finished with value: -0.2987560173280831 and parameters: {'model': 'ridge', 'alpha': 0.001435194769812018}. Best is trial 26 with value: -0.2587560173280831.

[1] [2025-05-31 04:15:56.816] Trial 27 finished with value: -0.23889136593145674 and parameters: {'model': 'ridge', 'alpha': 0.0009308064456923953}. Best is trial 27 with value: -0.2587560173280831.

[1] [2025-05-31 04:15:56.844] Trial 28 finished with value: -0.1193956822445905 and parameters: {'model': 'elastic', 'alpha': 0.000295898523174, 'l1\_ratio': 0.92589895173554}. Best is trial 28 with value: -0.2587560173280831.

[1] [2025-05-31 04:15:56.884] Trial 29 finished with value: -0.2987560173280831 and parameters: {'model': 'ridge', 'alpha': 0.0002219120370880996}. Best is trial 29 with value: -0.2587560173280831.

[1] [2025-05-31 04:15:58.414] Trial 30 finished with value: -0.097120215822931 and parameters: {'model': 'gbr', 'n\_estimators': 385, 'max\_depth': 4, 'learning\_rate': 0.0656303529498351}. Best is trial 30 with value: -0.2587560173280831.

[1] [2025-05-31 04:15:58.480] Trial 31 finished with value: -0.2987560173280831 and parameters: {'model': 'ridge', 'alpha': 0.0002209797598491013}. Best is trial 31 with value: -0.2587560173280831.

[1] [2025-05-31 04:15:58.513] Trial 32 finished with value: -0.2987560173280831 and parameters: {'model': 'ridge', 'alpha': 0.000158382404569944}. Best is trial 32 with value: -0.2587560173280831.

[1] [2025-05-31 04:15:58.544] Trial 33 finished with value: -0.2987560173280831 and parameters: {'model': 'ridge', 'alpha': 0.00028890307985988}. Best is trial 33 with value: -0.2587560173280831.

[1] [2025-05-31 04:15:58.576] Trial 34 finished with value: -0.2987560173280831 and parameters: {'model': 'ridge', 'alpha': 0.00055502779392427}. Best is trial 34 with value: -0.2587560173280831.

[1] [2025-05-31 04:15:58.611] Trial 35 finished with value: 0.05344125074523174 and parameters: {'model': 'elastic', 'alpha': 8.316040882202254, 'l1\_ratio': 0.9199845996904857}. Best is trial 35 with value: -0.2587560173280831.

[1] [2025-05-31 04:15:58.824] Trial 37 finished with value: -0.05437776802471616 and parameters: {'model': 'gbr', 'n\_estimators': 301, 'max\_depth': 4, 'learning\_rate': 0.0285445448273115}. Best is trial 37 with value: -0.2587560173280831.

[1] [2025-05-31 04:15:59.858] Trial 38 finished with value: -0.2987560173280831 and parameters: {'model': 'ridge', 'alpha': 0.0001061345210719568}. Best is trial 38 with value: -0.2587560173280831.

[1] [2025-05-31 04:16:00.703] Trial 39 finished with value: -0.0999338216262659 and parameters: {'model': 'gbr', 'n\_estimators': 284, 'max\_depth': 1, 'learning\_rate': 0.1168959602886664}. Best is trial 39 with value: -0.2587560173280831.

[1] [2025-05-31 04:16:00.737] Trial 40 finished with value: -0.0119254176087623 and parameters: {'model': 'ridge', 'alpha': 0.631193691280844}. Best is trial 40 with value: -0.2587560173280831.

[1] [2025-05-31 04:16:00.770] Trial 41 finished with value: -0.2987560173280831 and parameters: {'model': 'ridge', 'alpha': 0.00158365452023277}. Best is trial 41 with value: -0.2587560173280831.

[1] [2025-05-31 04:16:00.803] Trial 42 finished with value: -0.2987560173280831 and parameters: {'model': 'ridge', 'alpha': 0.00096562733576399362}. Best is trial 42 with value: -0.2587560173280831.

[1] [2025-05-31 04:16:00.835] Trial 43 finished with value: -0.2987560173280831 and parameters: {'model': 'ridge', 'alpha': 0.00055262826739100889}. Best is trial 43 with value: -0.2587560173280831.

[1] [2025-05-31 04:16:00.867] Trial 44 finished with value: -0.2987560173280831 and parameters: {'model': 'ridge', 'alpha': 0.0002271594861693953}. Best is trial 44 with value: -0.2587560173280831.

[1] [2025-05-31 04:16:00.898] Trial 45 finished with value: -0.19828665941363683 and parameters: {'model': 'ridge', 'alpha': 0.023293514737132418}. Best is trial 45 with value: -0.2587560173280831.

[1] [2025-05-31 04:16:00.934] Trial 46 finished with value: -0.0089580916658528 and parameters: {'model': 'elastic', 'alpha': 0.001138294311009617, 'l1\_ratio': 0.3490480280350904}. Best is trial 46 with value: -0.2587560173280831.

[1] [2025-05-31 04:16:06.968] Trial 47 finished with value: -0.02587560173280831 and parameters: {'model': 'ridge', 'alpha': 0.0005076735983445839}. Best is trial 47 with value: -0.2587560173280831.

[1] [2025-05-31 04:16:06.994] Trial 48 finished with value: -0.097120215822931 and parameters: {'model': 'gbr', 'n\_estimators': 372, 'max\_depth': 4, 'learning\_rate': 0.028054406770387622}. Best is trial 48 with value: -0.2587560173280831.

[1] [2025-05-31 04:16:02.488] Trial 49 finished with value: -0.2987560173280831 and parameters: {'model': 'ridge', 'alpha': 0.00427339705834056}. Best is trial 49 with value: -0.2587560173280831.

#### Step 14 : Advanced Deployment Infrastructure

1. Step 14A : build\_bundle.py
  2. Step 14B : Full Build Script – Real-Time Prediction Pipeline
  3. Step 14C : Automated Model Deployment & Scaling · One-cell User-ADC workflow

#### ▼ Step 14A : build bundle.py

```

1 import numpy as np, joblib, zipfile, pathlib, textwrap, os, json
2
3 # dummy training data
4 np.random.seed(42)
5 X_tab = np.random.randn(20, 4)
6 y_tab = np.random.randn(20)
7 X_seq = np.random.randn(20, 10, 3)
8 y_seq = np.random.randn(20)
9
10 # train + save models
11 import xgboost as xgb
12 from sklearn.linear_model import Ridge
13 from tensorflow.keras.models import Sequential
14 from tensorflow.keras.layers import GRU, LSTM, Dense
15 from tensorflow.keras.optimizers import Adam
16
17 # meta ridge
18 meta_model = Ridge().fit(X_tab, y_tab)
19 joblib.dump(meta_model, "meta_model.pkl")
20
21 # xgboost
22 xgb_model = xgb.XGBRegressor(n_estimators=10, random_state=42)
23 xgb_model.fit(X_tab, y_tab)

```

```

24 joblib.dump(xgb_model, "xgb_model.pkl")
25
26 # GRU
27 gru_model = Sequential([GRU(8, input_shape=(10, 3)), Dense(1)])
28 gru_model.compile(optimizer=Adam(), loss="mse")
29 gru_model.fit(X_seq, y_seq, epochs=1, verbose=0)
30 gru_model.save("gru_model.keras", include_optimizer=False)
31
32 # LSTM
33 lstm_model = Sequential([LSTM(8, input_shape=(10, 3)), Dense(1)])
34 lstm_model.compile(optimizer=Adam(), loss="mse")
35 lstm_model.fit(X_seq, y_seq, epochs=1, verbose=0)
36 lstm_model.save("lstm_model.keras", include_optimizer=False)
37
38 # "Transformer" placeholder
39 trf_model = Sequential([Dense(8, input_shape=(10, 3)), Dense(1)])
40 trf_model.compile(optimizer=Adam(), loss="mse")
41 trf_model.fit(X_seq, y_seq, epochs=1, verbose=0)
42 trf_model.save("transformer_model.keras", include_optimizer=False)
43
44 # write deployment files
45 # predict.py
46 pathlib.Path("predict.py").write_text(textwrap.dedent("""\
47     import os, joblib, numpy as np
48     from flask import Flask, request, jsonify
49     from tensorflow.keras.models import load_model
50
51     app = Flask(__name__)
52
53     meta_model      = joblib.load("meta_model.pkl")
54     xgb_model       = joblib.load("xgb_model.pkl")
55     gru_model       = load_model("gru_model.keras")
56     lstm_model      = load_model("lstm_model.keras")
57     transformer_model = load_model("transformer_model.keras")
58
59     @app.route("/predict", methods=["POST"])
60     def predict():
61         try:
62             data = request.get_json(force=True)
63             X_input = np.asarray(data["features"], dtype=float).reshape(1, -1)
64             X_seq   = np.asarray(data["sequence"], dtype=float).reshape(1, 10, 3)
65         except Exception as e:
66             return jsonify({"error": f"Invalid payload → {e}"}), 400
67
68         preds = {
69             "xgb": float(xgb_model.predict(X_input)[0]),
70             "gru": float(gru_model.predict(X_seq)[0][0]),
71             "lstm": float(lstm_model.predict(X_seq)[0][0]),
72             "transformer": float(transformer_model.predict(X_seq)[0][0])
73         }
74         meta_in = np.array([preds[k] for k in ("xgb", "gru", "lstm", "transformer")])
75         final_pred = float(meta_model.predict(meta_in)[0])
76
77         return jsonify({
78             "base_model_outputs": preds,
79             "ensemble_prediction": final_pred
80         })
81
82     if __name__ == "__main__":
83         app.run(host="0.0.0.0", port=int(os.getenv("PORT", 8080)), debug=False)
84     """))

85 # requirements.txt
86 pathlib.Path("requirements.txt").write_text(textwrap.dedent("""\
87     flask==0.2.0
88     gunicorn==22.0.0
89     numpy==1.26.4
90     scikit-learn==1.5.0
91     joblib==1.4.2
92     xgboost==2.1.4
93     tensorflow-cpu==2.15.0
94     tensorflow-io-gcs-filesystem==0.36.0
95     """))

96 # Dockerfile
97 pathlib.Path("Dockerfile").write_text(textwrap.dedent("""\
98     FROM tensorflow/tensorflow:2.15.0
99
100    WORKDIR /app
101    COPY .
102
103    RUN pip install --no-cache-dir --r requirements.txt
104
105    ENV PORT 8080
106    EXPOSE 8080
107    CMD ["gunicorn", "-b", "0.0.0.0:8080", "predict:app"]
108    """))

109 # README.md
110 pathlib.Path("README.md").write_text(textwrap.dedent("""\
111     # Stock Predictor API - Cloud Run / AWS ECS Deployment
112
113     ## Build & Deploy (Google Cloud Run)
114
115     ```bash
116     gcloud auth login
117     gcloud config set project YOUR_PROJECT_ID
118     gcloud services enable run.googleapis.com artifactregistry.googleapis.com
119
120     unzip stock_predictor_bundle.zip
121     cd stock_predictor_bundle
122
123     gcloud builds submit --tag gcr.io/YOUR_PROJECT_ID/stock-predictor-api
124
125     gcloud run deploy stock-predictor-api \
126         --image gcr.io/YOUR_PROJECT_ID/stock-predictor-api \
127         --platform managed --region us-central1 \
128         --allow-unauthenticated --port 8080
129     ```
130
131
132     ## Sample request
133
134     ```bash
135     curl -X POST https://YOUR_URL/predict \
136         -H "Content-type: application/json" \
137         -d `{
138             "features": [48.3, 2.1, 0.03, 0.05],
139             "sequence": [
140                 [0.1, 2.0, 0.01], [0.2, 2.1, 0.02], [0.3, 2.2, 0.03],
141                 [0.1, 1.9, 0.01], [0.2, 2.0, 0.02], [0.3, 2.1, 0.03],
142                 [0.1, 2.2, 0.01], [0.2, 1.8, 0.02], [0.3, 2.3, 0.03],
143                 [0.1, 2.1, 0.01]
144             ]
145         }` \
146     ```)
147     """))

148
149 # helper pack_up.py
150 pathlib.Path("pack_up.py").write_text(textwrap.dedent("""\
151     import zipfile, pathlib, sys
152     SKIP = ['.DS_Store', 'pack_up.py', 'step14A_build_bundle.py'],

```

```
153     'walk_forward_metrics.csv', 'shap_global.csv',  
154     'shap_holdout.csv', 'lime_explanation_Instance5.png',  
155     'rp_weights.db')  
156     with zipfile.ZipFile("stock_predictor_bundle.zip", "w", zipfile.ZIP_DEFLATED) as z:  
157         for p in pathlib.Path(".").glob("*"):  
158             if p.is_file() and p.name not in SKIP:  
159                 z.write(p)  
160     print("stock_predictor_bundle.zip created (filtered)")  
161     """)  
162  
163 # create zip bundle  
164 SKIP = ['.DS_Store', 'step14A_build_bundle.py', 'pack_up.py',  
165     'walk_forward_metrics.csv', 'shap_global.csv',  
166     'shap_holdout.csv', 'lime_explanation_Instance5.png',  
167     'rp_weights.db']  
168  
169 with zipfile.ZipFile("stock_predictor_bundle.zip", "w", zipfile.ZIP_DEFLATED) as z:  
170     for p in pathlib.Path(".").glob("*"):  
171         if p.is_file() and p.name not in SKIP:  
172             z.write(p)  
173  
174 print("Done - stock_predictor_bundle.zip is ready.")  
175 print("  Included files: ",
```

→ Done - stock\_predictor\_bundle.zip is ready.  
Included files: [  
"requirements.txt",  
"Dockerfile".