#AIT511: Course Project 1 Notebook Author:- Keyur Sanjaykumar Padiya

Roll no:- MT2025065

This is a report of Course Project 1 of Machine Learning. My objective is to create predictive models that can reveal hidden trends in lifestyle choices and advance knowledge of the risk factors for obesity and overweight.

For this i'll be using many manchine learning techniques, methods and concepts to get the required predictions with a good amount of accuracy.

1. Comprehensive EDA: Obesity Risk & Lifestyle Habits

1.1 Data Structure: Initial inspection, data types, and missing values.

The initial inspection revealed a high-quality dataset of 15,533 samples with zero missing values, simplifying the entire preprocessing phase. The data consists of a rich, mixed set of 17 features (8 numerical and 9 categorical), with the target variable, WeightCategory, confirming the task as multi-class classification.

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
import warnings
from sklearn.preprocessing import StandardScaler
from sklearn.decomposition import PCA
sns.set theme(style="whitegrid")
warnings.filterwarnings('ignore')
print("Libraries loaded and settings applied.")
df = pd.read csv("/content/drive/MyDrive/Obesity Data/train.csv")
print("Data loaded successfully.")
print(df.head())
print("\n--- Data Info ---")
df.info()
print(df.isnull().sum())
if 'id' in df.columns:
    df = df.drop('id', axis=1)
Libraries loaded and settings applied.
Data loaded successfully.
   id Gender
                            Height
                                        Weight
                     Age
family history with overweight \
```

```
0
   0
        Male 24.443011 1.699998
                                    81.669950
yes
1
   1 Female 18.000000
                         1.560000
                                    57.000000
yes
2
   2 Female 18.000000 1.711460
                                    50.165754
yes
   3 Female 20.952737 1.710730 131.274851
3
yes
        Male 31.641081 1.914186
4
                                    93.798055
yes
           FCVC
                       NCP
                                 CAEC SMOKE
                                                 CH20 SCC
  FAVC
FAF \
0 yes 2.000000 2.983297
                            Sometimes
                                             2.763573
                                                            0.000000
                                          no
                                                        no
  yes
       2.000000 3.000000
                            Frequently
                                         no
                                             2.000000
                                                        no
                                                            1.000000
2
                            Sometimes
                                             1.910378
  yes 1.880534 1.411685
                                                           0.866045
                                          no
                                                        no
  yes 3.000000 3.000000
                             Sometimes
                                          no
                                             1.674061
                                                        no
                                                           1.467863
  yes 2.679664 1.971472
                            Sometimes
                                         no
                                             1.979848
                                                        no 1.967973
                  CALC
                                       MTRANS
                                                    WeightCategory
       TUE
   0.976473
                        Public_Transportation
                                               Overweight Level II
            Sometimes
  1.000000
                                                     Normal Weight
1
                    no
                                   Automobile
                                              Insufficient Weight
2
  1.673584
                        Public Transportation
                    no
3
                        Public_Transportation
                                                 Obesity_Type_III
  0.780199
            Sometimes
  0.931721
            Sometimes
                        Public Transportation
                                              Overweight Level II
--- Data Info ---
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 15533 entries, 0 to 15532
Data columns (total 18 columns):
#
    Column
                                    Non-Null Count
                                                     Dtype
- - -
     -----
                                                     ----
0
    id
                                     15533 non-null
                                                     int64
    Gender
1
                                     15533 non-null
                                                     object
 2
    Aae
                                     15533 non-null
                                                     float64
 3
    Height
                                     15533 non-null
                                                     float64
 4
    Weight
                                    15533 non-null
                                                     float64
 5
                                    15533 non-null
                                                     obiect
    family history with overweight
 6
                                                     object
    FAVC
                                     15533 non-null
 7
    FCVC
                                     15533 non-null
                                                     float64
 8
    NCP
                                     15533 non-null
                                                     float64
 9
    CAEC
                                     15533 non-null
                                                     object
                                     15533 non-null
 10
    SMOKE
                                                     object
 11
    CH20
                                     15533 non-null
                                                     float64
 12
    SCC
                                     15533 non-null
                                                     object
```

```
13 FAF
                                       15533 non-null
                                                        float64
 14 TUE
                                                        float64
                                       15533 non-null
15 CALC
                                       15533 non-null
                                                        object
 16 MTRANS
                                       15533 non-null
                                                        object
17 WeightCategory
                                       15533 non-null
                                                        object
dtypes: float64(8), int64(1), object(9)
memory usage: 2.1+ MB
id
                                    0
                                    0
Gender
                                    0
Age
                                    0
Height
Weight
                                    0
family history with overweight
                                    0
                                    0
FAVC
FCVC
                                    0
NCP
                                    0
                                    0
CAEC
SM0KE
                                    0
                                    0
CH20
SCC
                                    0
                                    0
FAF
TUE
                                    0
CALC
                                    0
                                    0
MTRANS
WeightCategory
                                    0
dtype: int64
```

1.2: Target Variable Analysis

Now, let's focus on the variable we are trying to predict: **WeightCategory**. We'll plot its distribution to understand how many categories there are and how balanced they are.

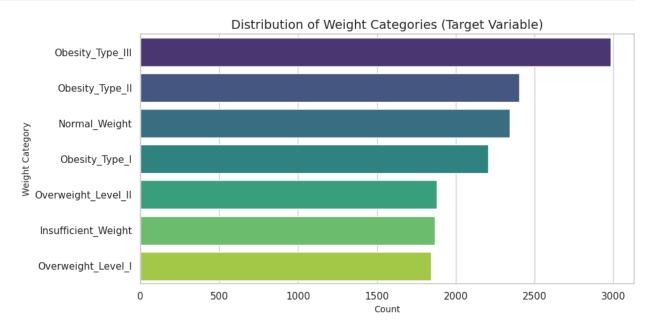
This confirms the task as a **multi-class classification problem**, immediately justifying the exclusion of regression models like **Linear Regression**. The target variable is adequately balanced, with categories ranging from $\approx 11.9\%$ to $\approx 19.2\%$; this even distribution ensures all models have sufficient data to learn each class, making advanced sampling techniques unnecessary.

```
print("\n--- Target Variable Analysis ---")

# Print the value counts
print("Normalized Target Variable Value Counts:")
print(df['WeightCategory'].value_counts(normalize=True) * 50)

# Plot the distribution
plt.figure(figsize=(10, 5))
sns.countplot(data=df, y='WeightCategory',
order=df['WeightCategory'].value_counts().index,palette="viridis")
plt.title('Distribution of Weight Categories (Target Variable)',
fontsize=14)
```

```
plt.xlabel('Count', fontsize=10)
plt.ylabel('Weight Category', fontsize=10)
plt.tight layout()
plt.savefig('weight category distribution.png')
print("Saved plot: weight category distribution.png")
# plt.show() # Uncomment this line if running in a local environment
--- Target Variable Analysis ---
Normalized Target Variable Value Counts:
WeightCategory
Obesity Type III
                       9.602137
Obesity Type II
                       7.735145
Normal Weight
                       7.548445
Obesity Type I
                       7.104230
Overweight Level II
                       6.054851
Insufficient Weight
                       6.019442
Overweight Level I
                       5.935750
Name: proportion, dtype: float64
Saved plot: weight category distribution.png
```



1.3: Numerical Feature Analysis

Let's explore the 8 numerical features. We will:

- 1. Get descriptive statistics (mean, min, max, etc.).
- 2. Plot histograms to see their individual distributions.
- 3. Create box plots to see how their distributions vary across each WeightCategory.

The numerical feature analysis confirmed that Weight is an extremely strong predictor, with its distributions showing minimal overlap across the seven target categories. While features like

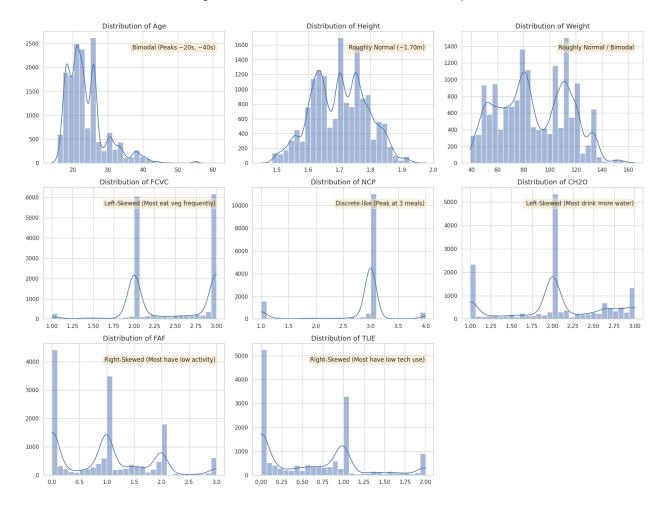
Age and Height exhibit clear trends, the analysis also revealed complex, non-linear relationships in lifestyle factors like FAF (e.g., the U-shaped pattern observed). This complexity, along with varied feature shapes (normal vs. skewed), strongly validates using Decision Tree and Boosting models, as they inherently capture these non-linear rules without relying on distributional assumptions.

```
print("\n--- Numerical Feature Analysis ---")
# Get descriptive statistics
numerical features = df.select dtypes(include=[np.number]).columns
print("--- Descriptive Statistics (Numerical Features) ---")
print(df[numerical features].describe())
# Plot distributions (Histograms)
fig, axes = plt.subplots(3, 3, figsize=(18, 14)) # Slightly larger
axes = axes.flatten() # Flatten the 2D array of axes for easy
iteration
# Dictionary of descriptions
descriptions = {
    'Age': 'Bimodal (Peaks ~20s, ~40s)',
    'Height': 'Roughly Normal (~1.70m)'
    'Weight': 'Roughly Normal / Bimodal',
    'FCVC': 'Left-Skewed (Most eat veg frequently)',
    'NCP': 'Discrete-like (Peak at 3 meals)',
    'CH20': 'Left-Skewed (Most drink more water)'
    'FAF': 'Right-Skewed (Most have low activity)',
    'TUE': 'Right-Skewed (Most have low tech use)'
}
# Loop through each numerical feature and plot it
for i, col in enumerate(numerical features):
    ax = axes[i] # Select the appropriate subplot
    # Plot the histogram with Kernel Density Estimate (KDE)
    sns.histplot(df[col], kde=True, bins=30, ax=ax)
    ax.set_title(f'Distribution of {col}', fontsize=14)
    ax.set_xlabel('') # Remove x-axis label for cleaner look
    ax.set ylabel('') # Remove y-axis label for cleaner look
    # --- Add text annotation ---
    # Get the current plot limits to position the text dynamically
    xlim = ax.get xlim()
    ylim = ax.get ylim()
    # Calculate position for the text box (top-right corner)
    x pos = x lim[0] + (x lim[1] - x lim[0]) * 0.95 # 95% across x-axis
    y_pos = ylim[1] * 0.9 # 90% up y-axis
    # Get the description text for the current column
    description = descriptions.get(col, '') # Get description or empty
```

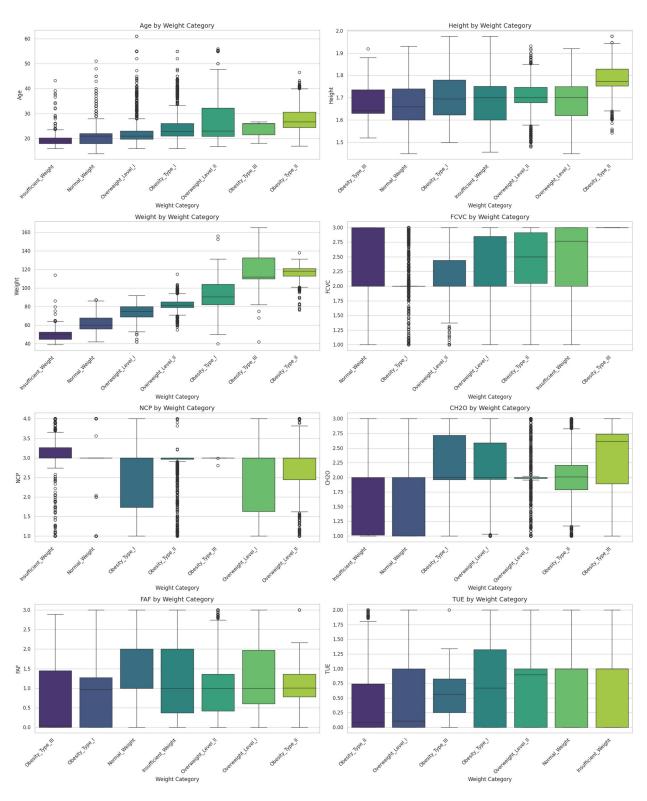
```
string if not found
    # Add the text to the plot
    ax.text(x pos, y pos, description,
            horizontalalignment='right', # Align text to the right
edge of x pos
            verticalalignment='top', # Align text to the top edge
of y_pos
            fontsize=12,
            # Add a background box for better readability
            bbox=dict(boxstyle='round,pad=0.3', fc='wheat',
alpha=0.5)
# Remove any unused subplots (if number of features is less than 9)
for j in range(len(numerical features), len(axes)):
    fig.delaxes(axes[j])
# Add a main title for the entire figure
fig.suptitle('Histograms of Numerical Features with Distribution
Descriptions', fontsize=20, y=1.02)
# Adjust layout to prevent title overlap
plt.tight layout(rect=[0, 0, 1, 1])
# Save the figure to a file
plt.savefig('numerical feature histograms annotated.png')
print("Saved annotated plot:
numerical_feature_histograms_annotated.png")
# plt.show() # Uncomment to display plot if running locally
# Plot box plots vs. WeightCategory
plt.figure(figsize=(20, 25))
for i, col in enumerate(numerical features):
    plt.subplot(4, 2, i + 1)
    order = df.groupby('WeightCategory')
[col].median().sort values().index
    sns.boxplot(data=df, x='WeightCategory', y=col,
order=order,palette="viridis")
    plt.title(f'{col} by Weight Category', fontsize=14)
    plt.xlabel('Weight Category', fontsize=12)
    plt.ylabel(col, fontsize=12)
    plt.xticks(rotation=45, ha='right')
plt.suptitle('Numerical Features by Weight Category', fontsize=24,
y=1.03)
plt.tight layout()
plt.savefig('numerical features vs target boxplots.png')
print("Saved plot: numerical features vs target boxplots.png")
# plt.show() # Uncomment this line
--- Numerical Feature Analysis ---
--- Descriptive Statistics (Numerical Features) ---
                                                         FCVC
                Age
                     Height
                                         Weight
```

NCP \					
count 15533.000000		15533.000000	15533.000000	15533.000000	
15533.00					
mean	23.816308	1.699918	87.785225	2.442917	
2.760425 std		0 007670	26.369144	0.530895	
0.706463			20.309144	0.550895	
min			39.000000	1.000000	
1.000000					
25%	20.000000	1.630927	66.000000	2.000000	
3.000000					
50%	22.771612	1.700000	84.000000	2.342220	
3.000000 75%	26.000000	1.762921	111.600553	3.000000	
3.000000		1.702921	111.000555	3.000000	
max	61.000000	1.975663	165.057269	3.000000	
4.000000					
	CH20	FAF			
	.5533.000000 2.027626	15533.000000 0.976968			
mean std	0.607733				
min	1.000000				
25%	1.796257	0.007050			
50%	2.000000	1.000000			
75%	2.531456	1.582675			
max 3.000000		3.000000		ma ammakakad	
			ature_nistogra target boxplot	ms_annotated.png	
Javeu pt	oc. Hullette	c_reacures_vs_	carget_boxptot	3 . ping	

Histograms of Numerical Features with Distribution Descriptions



Numerical Features by Weight Category

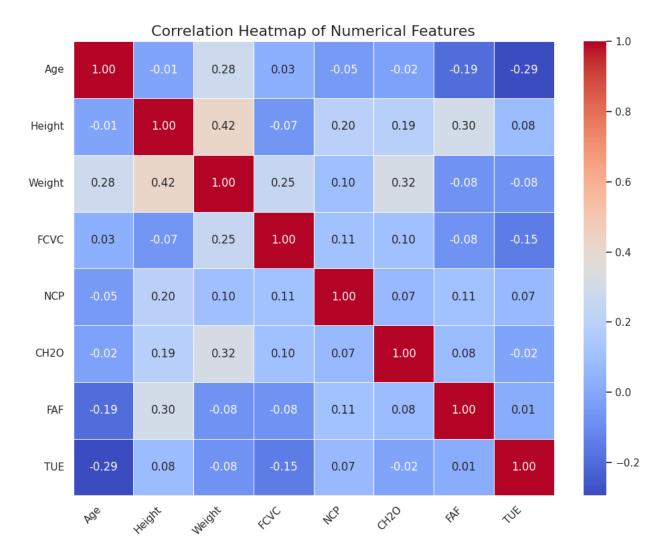


1.4: Correlation and Multicollinearity

Now, let's check for correlations among the numerical features. High correlation (multicollinearity) can be problematic for some models and suggests feature redundancy. This step is key to justifying the use of PCA.

The numerical features exhibit moderate correlations (e.g., 0.45 between Height and Weight) that violate the independence assumption of Naive Bayes, explaining its poor performance. While PCA was initially justified for KNN to address dimensionality, its transformation of features was ultimately unhelpful, as it diluted the clear, direct signal of the dominant predictor, Weight, which is better handled by tree-based models.

```
print("\n--- Correlation Analysis ---")
# Calculate the correlation matrix
corr matrix = df[numerical features].corr()
# Plot the heatmap
plt.figure(figsize=(10, 8))
sns.heatmap(corr matrix, annot=True, cmap='coolwarm', fmt='.2f',
linewidths=0.5)
plt.title('Correlation Heatmap of Numerical Features', fontsize=16)
plt.xticks(rotation=45, ha='right')
plt.yticks(rotation=0)
plt.tight_layout()
plt.savefig('correlation heatmap.png')
print("Saved plot: correlation heatmap.png")
plt.show()
--- Correlation Analysis ---
Saved plot: correlation heatmap.png
```



Why PCA won't be Helpful for This Specific Problem?

PCA's job is to create new, simplified features by combining the original ones based on variance. However, this process often dilutes the most critical information when one feature is overwhelmingly powerful, as is the case here with Weight. The EDA suggests tree-based models (which excel at using dominant features for splitting) are most suitable, PCA is logically counterproductive here.

1.5: Categorical Feature Analysis

Next, we'll analyze the categorical features to see how they relate to WeightCategory. We'll use stacked or grouped count plots to visualize this.

The categorical features, particularly family_history_with_overweight, MTRANS, and CAEC, proved highly predictive, revealing strong correlations with obesity categories in the plots. This provides compelling evidence for using tree-based models (like Random Forest and XGBoost), as they excel at learning the explicit rules suggested by these distributions. While these patterns also suit Naive Bayes conceptually, the requirement for one-hot encoding drastically increases

dimensionality, making distance-based models like KNN less suitable due to the curse of dimensionality, unless techniques like PCA are employed first.

```
print("\n--- Categorical Feature Analysis ---")
categorical features =
df.select dtypes(include=['object']).columns.drop('WeightCategory')
print(f"Categorical Features: {list(categorical features)}")
# Plot count plots for each categorical feature, grouped by
WeightCategory
plt.figure(figsize=(15, 20))
for i, col in enumerate(categorical features):
   plt.subplot(4, 2, i + 1)
    sns.countplot(data=df, x=col, hue='WeightCategory',
order=df[col].value counts().index)
   plt.title(f'{col} by Weight Category', fontsize=11)
   plt.xlabel(col, fontsize=12)
   plt.ylabel('Count', fontsize=12)
   plt.xticks(rotation=45, ha='right')
   # Place legend outside the plot
   plt.legend(title='Weight Category', loc='upper right',
bbox to anchor=(1.3, 1)
plt.suptitle('Categorical Features by Weight Category', fontsize=24,
y=1.03)
plt.tight layout()
plt.savefig('categorical features vs target.png')
print("Saved plot: categorical features vs target.png")
plt.show()
--- Categorical Feature Analysis ---
Saved plot: categorical features vs target.png
```

Categorical Features by Weight Category



1.6: Multivariate Analysis (PCA Visualization)

Let's directly visualize the justification for PCA and K-NN. We will:

- 1. Scale the numerical data (a requirement for PCA).
- 2. Apply PCA to reduce the 8 numerical features down to 2 principal components.
- 3. Plot these 2 components in a scatter plot, coloring the points by their WeightCategory. This shows us if the classes are "separable" in a lower-dimensional space.

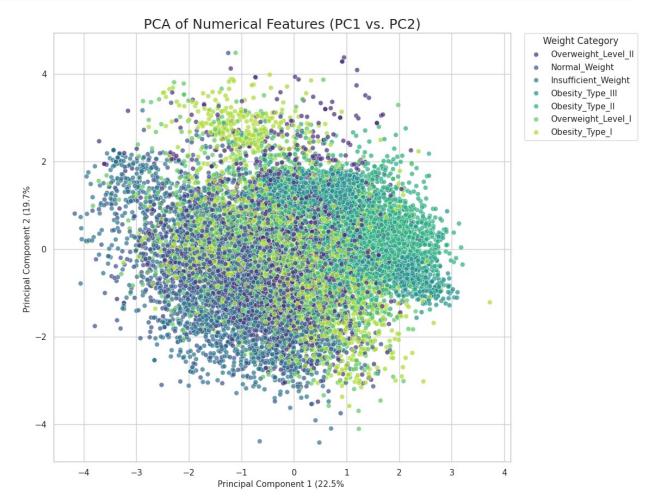
```
print("\n--- PCA Visualization ---")
# Separate numerical features and scale them
scaler = StandardScaler()
# Ensure only numerical features are selected if 'WeightCategory' was
accidentally included
df numerical = df[numerical features]
df scaled = scaler.fit transform(df numerical)
# Apply PCA
pca = PCA(n components=2)
principal components = pca.fit transform(df scaled)
df pca = pd.DataFrame(data=principal components, columns=['PC1',
'PC2'1)
df pca['WeightCategory'] = df['WeightCategory']
# --- Plot the PCA components---
plt.figure(figsize=(13, 9)) #
sns.scatterplot(
    data=df pca,
    x='PC1',
    y='PC2',
    hue='WeightCategory',
    palette='viridis',
    s = 40.
    alpha=0.7
)
plt.title('PCA of Numerical Features (PC1 vs. PC2)', fontsize=18) #
Slightly larger title
plt.xlabel(f'Principal Component 1
({pca.explained variance ratio [0]:.1%} ', fontsize=11)
plt.ylabel(f'Principal Component 2
({pca.explained variance ratio [1]:.1%} ', fontsize=11)
# Adjust legend placement to ensure it's outside
plt.legend(title='Weight Category', bbox_to_anchor=(1.03, 1),
loc='upper left', borderaxespad=0.)
plt.grid(True) # Keep grid lines
plt.tight_layout(rect=[0, 0, 0.9, 1]) # Adjust layout slightly to make
```

```
space for legend if outside
plt.savefig('pca_plot_tweaked.png')
print("Saved tweaked plot: pca_plot_tweaked.png")
# plt.show()

print(f"\nExplained variance by PC1:
{pca.explained_variance_ratio_[0]:.2%}")
print(f"Explained variance by PC2:
{pca.explained_variance_ratio_[1]:.2%}")
print(f"Total explained variance by 2 components:
{np.sum(pca.explained_variance_ratio_):.2%}")

--- PCA Visualization ---
Saved tweaked plot: pca_plot_tweaked.png

Explained variance by PC1: 22.48%
Explained variance by PC2: 19.69%
Total explained variance by 2 components: 42.17%
```



The PCA visualization of numerical features shows distinct clustering, with obesity types grouping together and Insufficient_Weight separating clearly . However, the first two components only capture ~42% of the variance, and significant overlap exists between Normal_Weight and Overweight categories. This visual clustering confirms the suitability of K-NN (as neighborhoods exist), while the extensive overlap simultaneously highlights the need for powerful, non-linear models like Random Forest and Gradient Boosting to effectively capture the complex decision boundaries.

2. Data Preprocessing steps

Data preprocessing encompasses a series of techniques used to clean, transform, and organize raw data before feeding it into machine learning models. Quality preprocessing directly impacts model performance, accuracy, and reliability. Data preprocessing is a critical step in machine learning pipelines that directly impacts model performance.

2.1: Data Loading & Exploration

By examining the shape and types, you can check if the dataset is imbalanced, contains enough observations for learning, and if the data types are compatible with your model.

The raw training and test datasets were successfully loaded, and non-predictive id columns were removed. Features (X) and the target (y) were cleanly separated, establishing distinct datasets for training the preprocessors and the final model, and for generating predictions.

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
import warnings
from sklearn.preprocessing import StandardScaler
from sklearn.decomposition import PCA
df = pd.read csv("/content/drive/MyDrive/Obesity Data/train.csv")
df test = pd.read csv('/content/drive/MyDrive/Obesity Data/test.csv')
print(df.snape)
print(df.info())
print(df.describe())
print(df.shape)
                               # Dimensions
                           # Data types
# Statistical summary
# Sample rows
print(df.head())
print(df.dtypes)
                               # Column types
(15533, 18)
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 15533 entries, 0 to 15532
Data columns (total 18 columns):
     Column
                                        Non-Null Count
                                                          Dtype
     -----
- - -
 0
     id
                                        15533 non-null
                                                          int64
 1
                                        15533 non-null
     Gender
                                                          object
```

```
2
                                      15533 non-null
                                                       float64
     Age
 3
                                      15533 non-null
                                                       float64
     Height
 4
     Weight
                                      15533 non-null
                                                       float64
 5
     family history with overweight
                                      15533 non-null
                                                       object
 6
                                      15533 non-null
                                                       object
 7
                                      15533 non-null
     FCVC
                                                       float64
 8
     NCP
                                      15533 non-null
                                                       float64
 9
     CAEC
                                      15533 non-null
                                                       obiect
 10
                                      15533 non-null
    SM0KE
                                                       object
 11
    CH20
                                      15533 non-null
                                                       float64
 12
    SCC
                                      15533 non-null
                                                       object
 13
    FAF
                                      15533 non-null
                                                       float64
 14
    TUE
                                      15533 non-null
                                                       float64
 15
    CALC
                                      15533 non-null
                                                       object
16 MTRANS
                                      15533 non-null
                                                       object
    WeightCategory
                                      15533 non-null
 17
                                                       object
dtypes: float64(8), int64(1), object(9)
memory usage: 2.1+ MB
None
                 id
                                          Height
                                                         Weight
                               Age
FCVC
                     15533.000000 15533.000000
count 15533.000000
                                                  15533.000000
15533.000000
                         23.816308
mean
        7766.000000
                                        1.699918
                                                      87.785225
2.442917
        4484.135201
                          5.663167
                                        0.087670
                                                      26.369144
std
0.530895
                         14.000000
                                        1.450000
                                                      39.000000
min
           0.000000
1.000000
25%
        3883.000000
                         20.000000
                                        1.630927
                                                      66.000000
2.000000
50%
        7766.000000
                         22.771612
                                        1.700000
                                                      84.000000
2.342220
75%
       11649.000000
                         26.000000
                                        1.762921
                                                     111.600553
3.000000
                         61.000000
max
       15532.000000
                                        1.975663
                                                     165.057269
3.000000
                NCP
                              CH20
                                              FAF
                                                            TUE
       15533.000000
                      15533.000000
                                    15533.000000
                                                   15533.000000
count
           2.760425
                          2.027626
                                        0.976968
                                                       0.613813
mean
                                                       0.602223
std
           0.706463
                          0.607733
                                        0.836841
min
           1.000000
                          1.000000
                                        0.000000
                                                       0.000000
25%
           3.000000
                          1.796257
                                        0.007050
                                                       0.000000
           3,000000
                          2.000000
                                        1.000000
                                                       0.566353
50%
           3.000000
                          2.531456
                                        1.582675
75%
                                                       1.000000
           4.000000
                          3.000000
                                        3.000000
                                                       2.000000
max
   id Gender
                     Age
                             Height
                                         Weight
family history with overweight \
```

```
0
    0
         Male 24.443011
                          1.699998
                                      81.669950
yes
1
    1 Female 18.000000
                          1.560000
                                      57.000000
yes
2
    2
       Female 18.000000
                          1.711460
                                      50.165754
yes
    3 Female 20.952737 1.710730 131.274851
3
yes
         Male 31.641081 1.914186
                                      93.798055
4
    4
yes
            FCVC
                       NCP
                                   CAEC SMOKE
                                                   CH20 SCC
  FAVC
FAF \
        2.000000 2.983297
                             Sometimes
                                               2.763573
                                                             0.000000
0 yes
                                           no
                                                         no
1
  yes
        2.000000 3.000000
                            Frequently
                                           no
                                               2.000000
                                                         no
                                                             1.000000
2
       1.880534 1.411685
                             Sometimes
                                               1.910378
                                                             0.866045
  yes
                                           no
                                                         no
  yes
        3.000000 3.000000
                             Sometimes
                                           no
                                               1.674061
                                                         no
                                                             1.467863
   yes 2.679664 1.971472
                             Sometimes
                                               1.979848
                                                             1.967973
                                           no
                                                         no
        TUE
                  CALC
                                        MTRANS
                                                     WeightCategory
   0.976473
                        Public_Transportation
                                                0verweight_Level_II
             Sometimes
  1.000000
                                    Automobile
                                                      Normal Weight
1
                    no
                        Public Transportation
                                                Insufficient Weight
2
   1.673584
                    no
3
   0.780199
             Sometimes
                        Public_Transportation
                                                   Obesity_Type_III
4
   0.931721
             Sometimes
                        Public Transportation
                                                Overweight Level II
id
                                     int64
Gender
                                    object
                                   float64
Age
                                   float64
Height
Weight
                                   float64
family history with overweight
                                    object
FAVC
                                    object
FCVC
                                   float64
NCP
                                   float64
CAEC
                                    object
SMOKE
                                    object
CH20
                                   float64
SCC
                                    object
FAF
                                   float64
                                   float64
TUE
                                    object
CALC
MTRANS
                                    object
WeightCategory
                                    object
dtype: object
```

2.2: Identify Feature Types

Before applying specific transformations, we must identify which columns are numerical and which are categorical. This allows us to apply the correct preprocessing step (scaling or encoding) to each type.

The script correctly identified 8 numerical and 9 categorical features based on their data types, enabling the application of appropriate, distinct transformations (scaling vs. encoding) in the subsequent steps.

```
X = df.drop(['id', 'WeightCategory'], axis=1)
y = df['WeightCategory']

categorical_cols = X.select_dtypes(include=['object']).columns
numerical_cols = X.select_dtypes(include=['float64', 'int64']).columns

print(f"Identified {len(categorical_cols)} categorical columns:
{list(categorical_cols)}")
print(f"Identified {len(numerical_cols)} numerical columns:
{list(numerical_cols)}")

Identified 8 categorical columns: ['Gender',
'family_history_with_overweight', 'FAVC', 'CAEC', 'SMOKE', 'SCC',
'CALC', 'MTRANS']
Identified 8 numerical columns: ['Age', 'Height', 'Weight', 'FCVC',
'NCP', 'CH2O', 'FAF', 'TUE']
```

2.3: Handle Categorical Features (One-Hot Encoding)

The EDA identified 9 categorical features. Tree-based models like XGBoost require numerical input. One-Hot Encoding converts these text-based categories into binary (0/1) columns, allowing the model to process them. Using drop_first=True avoids creating perfectly collinear features. Crucially, the columns must be aligned between the training and test sets after encoding to ensure they have the exact same structure.

```
test_ids = df_test['id']
X_test_official = df_test.drop('id', axis=1)
print("Applying One-Hot Encoding...")

X = pd.get_dummies(X, columns=categorical_cols, drop_first=True,
dtype=int)
print(f"Training data shape after OHE: {X.shape}")

X_test_official = pd.get_dummies(X_test_official,
columns=categorical_cols, drop_first=True, dtype=int)
print(f"Test data shape after OHE: {X_test_official.shape}")

print("Aligning columns between train and test sets...")
X, X_test_official = X.align(X_test_official, join='inner', axis=1,
fill_value=0)
```

```
print("Columns aligned.")
print(f"Shape after alignment: X={X.shape},
X_test_official={X_test_official.shape}")

Applying One-Hot Encoding...
Training data shape after OHE: (15533, 22)
Test data shape after OHE: (5225, 22)
Aligning columns between train and test sets...
Columns aligned.
Shape after alignment: X=(15533, 22), X_test_official=(5225, 22)
```

2.4: Handle Numerical Features (Standard Scaling)

The 8 numerical features have different scales. While XGBoost is somewhat robust to this, applying Standard Scaling standardizes the features (mean=0, std dev=1). This can sometimes help the gradient boosting process converge more smoothly. The scaler is fitted only on the training data and then used to transform both train and test sets to prevent data leakage.

```
print("Applying Standard Scaling...")
num features to scale =
['Age','Height','Weight','FCVC','NCP','CH20','FAF','TUE']
num features to scale = [col for col in num_features_to_scale if col
in X.columns]
scaler = StandardScaler()
scaler.fit(X[num features to scale])
print("Scaler fitted on training data.")
X[num_features_to_scale] = scaler.transform(X[num_features_to_scale])
X_test_official[num_features_to_scale] =
scaler.transform(X test official[num_features_to_scale])
print("Scaling applied to both datasets.")
print("\nSample scaled training data (first 5 rows):")
print(X.head())
Applying Standard Scaling...
Scaler fitted on training data.
Scaling applied to both datasets.
Sample scaled training data (first 5 rows):
              Height Weight
                                                NCP
                                                         CH20
        Age
                                     FCVC
FAF \
0 0.110667 0.000910 -0.231918 -0.834311 0.315486 1.211010 -
1.167485
1 -1.027075 -1.596015 -1.167509 -0.834311 0.339130 -0.045459
0.027524
2 -1.027075  0.131655 -1.426693 -1.059346 -1.909206 -0.192933 -
```

```
0.132554
3 -0.505665
            0.586624
4 1.381740
                      0.228033 0.445954 -1.116801 -0.078619
            2.444105
1.184260
       TUE
            Gender_Male
                          family_history_with_overweight_yes
   0.602221
1
  0.641290
                      0
                                                          1
                      0
2
  1.759822
                                                          1
                      0
3
  0.276295
                                                          1
                       1
4 0.527908
   CAEC_Sometimes
                  CAEC no
                                               CALC Sometimes
                           SMOKE yes
                                     SCC yes
CALC no \
                1
                        0
                                   0
                                            0
                                                            1
0
1
               0
                        0
                                   0
                                            0
                                                            0
1
2
                                   0
                                            0
                                                            0
                1
                        0
1
3
                        0
                                   0
                                            0
                                                            1
0
4
                        0
                                   0
                                            0
                                                            1
0
               MTRANS Motorbike MTRANS Public Transportation
   MTRANS Bike
MTRANS Walking
0
                               0
                                                            1
0
1
            0
                               0
                                                            0
0
2
             0
                               0
                                                            1
0
3
            0
                               0
                                                            1
0
4
             0
                               0
                                                            1
0
[5 rows x 22 columns]
```

2.5: Handle Target Variable (Label Encoding)

The target variable WeightCategory needs to be converted from text labels to integers (0, 1, 2... 6) for XGBoost's multi:softmax objective. Label Encoding achieves this mapping.

Label Encoding successfully transformed the 7 distinct text labels of the WeightCategory target variable into the required integer format (0-6) for XGBoost's multi-class classification objective (multi:softmax).

```
from sklearn.preprocessing import LabelEncoder
print("Applying Label Encoding to the target variable...")
le = LabelEncoder()
y enc = le.fit transform(y)
print("Target variable Label Encoded.")
print("\nLabel Encoding Mapping:")
for i, class name in enumerate(le.classes ):
    print(f"{class name} -> {i}")
print("\nPreprocessing pipeline complete. Data is ready for model
training.")
Applying Label Encoding to the target variable...
Target variable Label Encoded.
Label Encoding Mapping:
Insufficient Weight -> 0
Normal Weight -> 1
Obesity_Type_I -> 2
Obesity_Type_II -> 3
Obesity_Type_III -> 4
Overweight Level I -> 5
Overweight Level II -> 6
Preprocessing pipeline complete. Data is ready for model training.
```

3. Data Preprocessing steps

#Naive Bayes Classifier

Best Output/Test Accuracy: - .63305

```
import pandas as pd
import numpy as np
from sklearn.preprocessing import StandardScaler
from sklearn.naive_bayes import GaussianNB
from google.colab import drive

df = pd.read_csv("/content/drive/MyDrive/Obesity Data/train.csv")
df_test = pd.read_csv("/content/drive/MyDrive/Obesity Data/test.csv")

df = df.drop('id', axis=1)

X = df.drop('WeightCategory', axis=1)
y = df['WeightCategory']

categorical_cols =
```

```
X.select dtypes(include=['object']).columns.tolist()
numerical cols = X.select dtypes(include=['int64',
'float64']).columns.tolist()
X train full = X.copy()
y train full = y.copy()
test_df_ids = df_test['id']
X_test_official = df_test.drop('id', axis=1)
X train cat = pd.get dummies(X train full[categorical cols],
drop first=True, dtype=int)
X test cat = pd.get dummies(X test official[categorical cols],
drop first=True, dtype=int)
X train cat, X test cat = X train cat.align(X test cat, join='left',
axis=1, fill value=0)
X test cat = X test cat.fillna(0)
X train manual =
pd.concat([X train full.drop(columns=categorical cols).reset index(dro
p=True), X train cat.reset index(drop=True)], axis=1)
X test manual =
pd.concat([X test official.drop(columns=categorical cols).reset index(
drop=True), X test cat.reset index(drop=True)], axis=1)
scaler = StandardScaler()
scaler.fit(X train manual[numerical cols])
X train manual[numerical cols] =
scaler.transform(X train manual[numerical cols])
X test manual[numerical cols] =
scaler.transform(X test manual[numerical cols])
gnb = GaussianNB()
gnb.fit(X train manual, y train full)
y pred official = gnb.predict(X test manual)
submission df = pd.DataFrame({'id': test df ids, 'WeightCategory':
y pred official})
submission df.to csv('weight naive bayes submission MT2025065.csv',
index=False)
print(submission df.head())
      id
               WeightCategory
  15533
             Obesity Type III
1
  15534
               Obesity_Type_I
2
  15535
              Obesity Type \overline{I}I
              Obesity_Type_II
  15536
4 15537
          Insufficient Weight
```

#KNN with PCA Best Output/Test Accuracy:-.71763

This code justifies why PCA gives low accuracy, the code applies PCA on one-hot encoded and scaled features and uses only 10 principal components for KNN classification. PCA reduces dimensionality based purely on global variance without considering class labels or separability. This can lead to loss of important discriminative information because the principal components maximizing variance may not align with directions that separate classes well. Hence, PCA might discard features useful for classification, limiting accuracy improvement. In datasets where class-distinguishing features contribute less to total variance, PCA can actually degrade classification performance rather than help.

```
import pandas as pd
from sklearn.preprocessing import StandardScaler
from sklearn.neighbors import KNeighborsClassifier
from sklearn.decomposition import PCA
from sklearn.metrics import accuracy score, classification report,
confusion matrix
import seaborn as sns
import matplotlib.pyplot as plt
import numpy as np
import warnings
warnings.filterwarnings("ignore")
file path base = '/content/drive/MyDrive/Obesity Data/'
df train = pd.read csv(file path base + 'train.csv')
df test = pd.read csv(file path base + 'test.csv')
df train = df train.drop('id', axis=1)
X train full = df train.drop('WeightCategory', axis=1)
y train full = df train['WeightCategory']
test df ids = df test['id']
if 'WeightCategory' in df test.columns:
    X_test_official = df_test.drop(['id', 'WeightCategory'], axis=1)
    y test official = df test['WeightCategory']
else:
    X_test_official = df_test.drop('id', axis=1)
    y test official = None
numerical cols =
X train full.select dtypes(include=['float64']).columns
categorical cols =
X train full.select dtypes(include=['object']).columns
X train cat = pd.get dummies(X train full[categorical cols],
drop first=True, dtype=int)
X test cat = pd.get dummies(X test official[categorical cols],
drop first=True, dtype=int)
```

```
X train cat, X test cat = X train cat.align(X test cat, join='left',
axis=1, fill value=0)
X test cat = X test cat.fillna(0)
X train manual = X train full.drop(columns=categorical cols)
X test manual = X test official.drop(columns=categorical cols)
X train manual = pd.concat([X train manual.reset index(drop=True),
X train cat.reset index(drop=True)], axis=1)
X test manual = pd.concat([X test manual.reset index(drop=True),
X test cat.reset index(drop=True)], axis=1)
scaler = StandardScaler()
scaler.fit(X train manual)
X train scaled = scaler.transform(X train manual)
X test scaled = scaler.transform(X test manual)
n components = 10
pca = PCA(n components = n components, random state = 42)
pca.fit(X train scaled)
X train pca = pca.transform(X train scaled)
X test pca = pca.transform(X test scaled)
knn = KNeighborsClassifier(n neighbors=5)
knn.fit(X train pca, y train full)
y_pred_official = knn.predict(X test pca)
if y test official is not None:
    accuracy = accuracy score(y test official, y pred official)
    print(f"Accuracy on Test Set: {accuracy:.4f}")
    print(classification report(y test official, y pred official,
zero division=0))
    plt.figure(figsize=(10, 8))
    labels = np.unique(y test official)
    sns.heatmap(confusion_matrix(y_test_official, y_pred_official),
                annot=True, fmt='d', cmap='Blues',
                xticklabels=labels, yticklabels=labels)
    plt.title('Confusion Matrix for KNN/PCA on Test Data')
    plt.show()
submission df = pd.DataFrame({
    'id': test df ids,
    'WeightCategory': y_pred_official
})
submission df.to csv('weight knn submission MT2025065.csv',
```

```
index=False)
print(df.head())
  Gender
                Age
                       Height
                                  Weight
family history with overweight
    Male 24.443011 1.699998
                               81.669950
yes
  Female 18.000000 1.560000
                               57.000000
1
yes
  Female 18.000000 1.711460
                               50.165754
yes
3
  Female 20.952737 1.710730 131.274851
ves
    Male 31.641081 1.914186
                               93.798055
4
yes
 FAVC
           FCVC
                      NCP
                                CAEC SMOKE
                                                CH20 SCC
FAF \
      2.000000 2.983297
                            Sometimes
                                            2.763573
                                                          0.000000
0 yes
                                        no
                                                      no
1 yes 2.000000 3.000000
                           Frequently
                                        no
                                            2.000000
                                                          1.000000
                                                      no
2 yes 1.880534 1.411685
                            Sometimes
                                            1.910378
                                                         0.866045
                                        no
                                                      no
  yes 3.000000 3.000000
                            Sometimes
                                            1.674061
                                                          1.467863
                                        no
                                                      no
  yes 2.679664 1.971472
                            Sometimes
                                            1.979848
                                                      no 1.967973
                                        no
       TUE
                 CALC
                                     MTRANS
                                                  WeightCategory
  0.976473
            Sometimes
                       Public Transportation
                                             Overweight Level II
                                                   Normal Weight
1
  1.000000
                                 Automobile
                   no
                                             Insufficient Weight
2
  1.673584
                       Public Transportation
                   no
            Sometimes
3
  0.780199
                       Public Transportation
                                                Obesity Type III
                       Public Transportation
  0.931721
            Sometimes
                                             Overweight Level II
```

#Bagging Best Output/Test Accuracy: - .90688

```
import pandas as pd
import numpy as np
import time
import warnings
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler
from sklearn.tree import DecisionTreeClassifier
from sklearn.ensemble import BaggingClassifier
from sklearn.metrics import accuracy_score, classification_report,
confusion_matrix
import seaborn as sns
import matplotlib.pyplot as plt
```

```
warnings.filterwarnings("ignore")
file path train = '/content/drive/MyDrive/Obesity Data/train.csv'
df train = pd.read_csv(file_path_train)
df = df train.drop('id', axis=1)
X = df.drop('WeightCategory', axis=1)
y = df['WeightCategory']
X_train, X_test, y_train, y_test = train_test_split(
    X, y, test_size=0.2, random_state=42, stratify=y
numerical cols = X train.select dtypes(include=['float64',
'int64']).columns
categorical cols = X train.select dtypes(include=['object']).columns
X train cat = pd.get dummies(X train[categorical cols],
drop first=True, dtype=int)
X test cat = pd.get dummies(X test[categorical cols], drop first=True,
dtvpe=int)
X train cat, X test cat = X train cat.align(X test cat, join='left',
axis=1, fill value=0)
X_train_manual = X_train.drop(columns=categorical cols)
X test manual = X test.drop(columns=categorical cols)
X train manual = pd.concat([X_train_manual.reset_index(drop=True),
X_train_cat.reset_index(drop=True)], axis=1)
X test manual = pd.concat([X test manual.reset index(drop=True),
X test cat.reset index(drop=True)], axis=1)
scaler = StandardScaler()
numerical cols to scale = [col for col in numerical cols if col in
X train manual.columns and col not in categorical cols]
scaler.fit(X train manual[numerical cols to scale])
X train manual[numerical cols to scale] =
scaler.transform(X train manual[numerical cols to scale])
X test manual[numerical cols to scale] =
scaler.transform(X_test_manual[numerical cols to scale])
base estimator = DecisionTreeClassifier(max depth=30,
min samples leaf=3, random state=50)
bagging model = BaggingClassifier(
    estimator=base estimator,
    n estimators=1000,
    max features=0.6,
    random state=50,
    n jobs=-1
bagging model.fit(X train manual, y train)
y pred = bagging model.predict(X test manual)
accuracy = accuracy score(y test, y pred)
print(f"Validation Accuracy: {accuracy:.5f}")
print("Validation Classification Report:")
print(classification report(y test, y pred, zero division=0))
```

Validation Accuracy: Validation Classific				
	precision	recall	f1-score	support
Insufficient_Weight Normal_Weight Obesity_Type_I Obesity_Type_II Obesity_Type_III Overweight_Level_I Overweight_Level_II	0.93 0.87 0.89 0.95 0.99 0.82 0.80	0.94 0.90 0.86 0.98 1.00 0.73 0.82	0.93 0.89 0.87 0.96 1.00 0.77 0.81	374 469 441 481 597 369 376
accuracy macro avg weighted avg	0.89 0.90	0.89 0.90	0.90 0.89 0.90	3107 3107 3107

#Random Forest Best Output/Test Accuracy: - .90578

```
import pandas as pd
from sklearn.model selection import train test split
from sklearn.preprocessing import StandardScaler
from sklearn.ensemble import RandomForestClassifier
from sklearn.metrics import accuracy score, classification report
import warnings
import os
warnings.filterwarnings("ignore")
df = pd.read csv("/content/drive/MyDrive/Obesity Data/train.csv")
df = df.drop('id', axis=1)
X = df.drop('WeightCategory', axis=1)
y = df['WeightCategory']
X internal train, X_internal_val, y_internal_train, y_internal_val =
train test split(
    Х, у,
    test size=0.2,
    random state=42,
    stratify=y
X internal train = X internal train.reset index(drop=True)
X internal val = X internal val.reset index(drop=True)
y internal train = y internal train.reset index(drop=True)
y internal val = y internal val.reset index(drop=True)
numerical_cols_internal =
X internal train.select dtypes(include=['float64', 'int64']).columns
```

```
categorical cols internal =
X internal train.select dtypes(include=['object']).columns
X train cat internal =
pd.get dummies(X internal train[categorical cols internal],
drop first=True, dtype=int)
X val cat internal =
pd.get_dummies(X_internal val[categorical cols internal],
drop first=True, dtype=int)
X train cat internal, X val cat internal =
X train cat internal.align(X val cat internal, join='left', axis=1,
fill value=0)
X train proc internal =
X internal train.drop(columns=categorical cols internal)
X val proc internal =
X internal val.drop(columns=categorical cols internal)
X train proc internal = pd.concat([X train proc internal,
X train cat internal], axis=1)
X val proc internal = pd.concat([X val proc internal,
X_val_cat_internal], axis=1)
scaler internal = StandardScaler()
numerical cols to scale internal = [col for col in
numerical cols internal if col in X train proc internal.columns]
scaler internal.fit(X train proc internal[numerical cols to scale inte
rnall)
X train proc internal[numerical cols to scale internal] =
scaler internal.transform(X train proc internal[numerical cols to scal
e_internal])
X val proc internal[numerical cols to scale internal] =
scaler internal.transform(X val proc internal[numerical cols to scale
internall)
model_internal = RandomForestClassifier(
    n estimators=1000,
    \max depth=15,
    min samples leaf=5,
    max features=0.7,
    random state=50,
    n jobs=-1
)
model internal.fit(X train proc internal, y internal train)
y pred internal val = model internal.predict(X val proc internal)
```

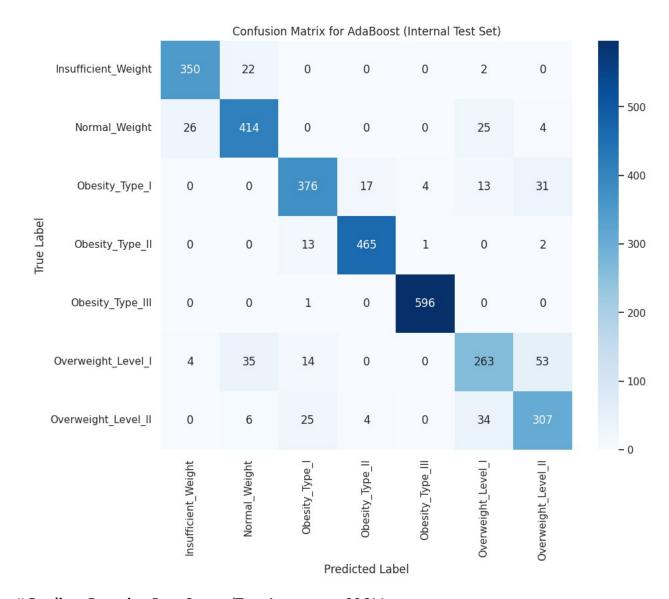
```
validation accuracy = accuracy score(y internal val,
y pred internal val)
print(f"Validation Accuracy: {validation accuracy:.5f}")
print(classification report(y internal val, y pred internal val,
zero division=0))
Validation Accuracy: 0.89411
                      precision
                                    recall f1-score
                                                        support
Insufficient Weight
                           0.92
                                      0.93
                                                 0.93
                                                            374
      Normal Weight
                           0.87
                                      0.89
                                                 0.88
                                                            469
     Obesity Type I
                           0.88
                                      0.85
                                                 0.86
                                                            441
    Obesity Type \overline{II}
                                      0.97
                                                 0.96
                           0.95
                                                            481
   Obesity Type III
                           0.99
                                      1.00
                                                 0.99
                                                            597
 Overweight Level I
                           0.79
                                      0.73
                                                 0.76
                                                            369
Overweight Level II
                                      0.81
                                                            376
                           0.79
                                                 0.80
           accuracy
                                                 0.89
                                                           3107
          macro avq
                           0.88
                                      0.88
                                                 0.88
                                                           3107
       weighted avg
                           0.89
                                      0.89
                                                 0.89
                                                           3107
```

#AdaBoost Best Output/Test Accuracy :- .89614

```
import pandas as pd
import numpy as np
import time
import warnings
from sklearn.model selection import train test split
from sklearn.preprocessing import StandardScaler
from sklearn.tree import DecisionTreeClassifier
from sklearn.ensemble import AdaBoostClassifier
from sklearn.metrics import accuracy score, classification report,
confusion matrix
import seaborn as sns
import matplotlib.pyplot as plt
warnings.filterwarnings("ignore")
file path = '/content/drive/MyDrive/Obesity Data/train.csv'
df train = pd.read_csv(file_path)
df = df train.drop('id', axis=1)
X = df.drop('WeightCategory', axis=1)
y = df['WeightCategory']
X train, X test, y train, y test = train test split(
    Χ, γ,
    test_size=0.2,
```

```
random state=42,
    stratify=y
)
numerical cols = ['Age', 'Height', 'Weight', 'FCVC', 'NCP', 'CH2O',
'FAF', 'TUE']
categorical cols = X train.select dtypes(include=['object']).columns
X train cat = pd.get dummies(X train[categorical cols],
drop first=True, dtype=int)
X test cat = pd.get dummies(X test[categorical cols], drop first=True,
dtype=int)
X train cat, X test cat = X train cat.align(X test cat, join='left',
axis=1, fill value=0)
X test cat = X test cat.fillna(0)
X_train_manual = X_train.drop(columns=categorical_cols)
X test manual = X test.drop(columns=categorical cols)
X train manual = pd.concat([X train manual.reset index(drop=True),
X train cat.reset index(drop=True)], axis=1)
X test manual = pd.concat([X test manual.reset index(drop=True),
X test cat.reset index(drop=True)], axis=1)
scaler = StandardScaler()
scaler.fit(X train manual[numerical cols])
X train manual[numerical cols] =
scaler.transform(X train manual[numerical cols])
X test manual[numerical cols] =
scaler.transform(X test manual[numerical cols])
base estimator = DecisionTreeClassifier(
    \max depth=15,
    min samples leaf=10,
    random state=50
)
ada model = AdaBoostClassifier(
    estimator=base estimator,
    n estimators=1000,
    learning rate=0.1,
    random state=50
)
ada model.fit(X train manual, y train)
y pred = ada model.predict(X test manual)
```

```
accuracy = accuracy_score(y_test, y_pred)
print(f"Validation Accuracy: {accuracy:.5f}")
print(classification_report(y_test, y_pred, zero_division=0))
plt.figure(figsize=(10, 8))
labels = np.unique(y test)
sns.heatmap(confusion_matrix(y_test, y_pred), annot=True, fmt='d',
cmap='Blues', xticklabels=labels, yticklabels=labels)
plt.title('Confusion Matrix for AdaBoost (Internal Test Set)')
plt.xlabel('Predicted Label')
plt.ylabel('True Label')
plt.show()
Validation Accuracy: 0.89186
                      precision
                                   recall f1-score
                                                       support
Insufficient Weight
                                     0.94
                                                0.93
                                                           374
                           0.92
      Normal Weight
                           0.87
                                     0.88
                                                0.88
                                                           469
     Obesity_Type I
                                     0.85
                                                0.86
                                                           441
                           0.88
    Obesity_Type_II
                           0.96
                                     0.97
                                                0.96
                                                           481
   Obesity Type \overline{I}II
                           0.99
                                     1.00
                                                0.99
                                                           597
 Overweight Level I
                           0.78
                                     0.71
                                                0.75
                                                           369
Overweight Level II
                           0.77
                                     0.82
                                                0.79
                                                           376
                                                0.89
                                                          3107
           accuracy
          macro avg
                           0.88
                                     0.88
                                                0.88
                                                          3107
                           0.89
                                     0.89
                                                0.89
                                                          3107
       weighted avg
```



#Gradient Boosting Best Output/Test Accuracy :- .89614

GridSearchCV on gradient boosting range

```
import pandas as pd
import numpy as np
import time
import warnings
from sklearn.model_selection import GridSearchCV
from sklearn.preprocessing import StandardScaler
from sklearn.ensemble import GradientBoostingClassifier
from sklearn.metrics import accuracy_score, classification_report,
confusion_matrix
import seaborn as sns
import matplotlib.pyplot as plt
```

```
warnings.filterwarnings("ignore")
print("Loading data from Google Drive...")
file path base = '/content/drive/MyDrive/Obesity Data/'
df train = pd.read csv(file path base + 'train.csv')
df test = pd.read csv(file path base + 'test.csv')
print("
    Both train.csv and test.csv loaded successfully.")
print("Preparing 100% of training data...")
df = df train.drop('id', axis=1)
X train full = df.drop('WeightCategory', axis=1)
y train full = df['WeightCategory']
print("Preparing test data for prediction...")
test df ids = df test['id']
X test official = df test.drop('id', axis=1)
y test official = None
if 'WeightCategory' in X_test_official.columns:
    y test_official = X_test_official['WeightCategory']
    X test official = X test official.drop('WeightCategory', axis=1)
    print("Test labels found. Model will be evaluated.")
else:
    print("No test labels found. Skipping evaluation.")
print("Starting preprocessing (OHE & Scaling)...")
numerical cols = ['Age', 'Height', 'Weight', 'FCVC', 'NCP', 'CH2O',
'FAF', 'TUE']
categorical cols =
X train full.select dtypes(include=['object']).columns
X train cat = pd.get dummies(X train full[categorical cols],
drop first=True, dtype=int)
X test cat = pd.get dummies(X test official[categorical cols],
drop first=True, dtype=int)
X train cat, X test cat = X train cat.align(X test cat, join='left',
axis=1, fill value=0)
X train manual = X train full.drop(columns=categorical cols)
X test manual = X test official.drop(columns=categorical cols)
X train manual = pd.concat([X train manual.reset index(drop=True),
X train cat.reset index(drop=True)], axis=1)
X test manual = pd.concat([X test manual.reset index(drop=True),
X test cat.reset index(drop=True)], axis=1)
scaler = StandardScaler()
scaler.fit(X train manual[numerical cols])
X train manual[numerical cols] =
scaler.transform(X train manual[numerical cols])
```

```
X test manual[numerical cols] =
scaler.transform(X test manual[numerical cols])
print("□ Preprocessing complete for both datasets.")
print("\n--- Starting Focused GridSearchCV for
GradientBoostingClassifier ---")
print("This may take some time...")
start time = time.time()
gb model = GradientBoostingClassifier(random state=42)
param grid = {
    'n estimators': [1000, 1200],
    'learning rate': [0.03, 0.05],
    'max depth': [5, 6],
    'subsample': [0.8, 0.9]
}
grid search = GridSearchCV(
    estimator=gb model,
    param grid=param grid,
    cv=3,
    scoring='accuracy',
    verbose=2,
    n iobs=-1
)
grid search.fit(X train manual, y train full)
end time = time.time()
print(f" GridSearchCV complete. Time taken: {end time -
start time:.2f} seconds.")
print("\n--- Best Parameters Found by GridSearch ---")
print(grid_search.best_params_)
print(f"\nBest internal cross-validation accuracy:
{grid search.best score :.5f}")
print("\n--- Generating predictions on test.csv using BEST model ---")
y pred official = grid search.predict(X test manual)
if y test official is not None:
   print("\n--- Evaluating Best Model on Test Set (Labels Found)
- - - " )
    accuracy = accuracy_score(y_test_official, y_pred_official)
    print(f"--- FINAL ACCURACY ON TEST SET: {accuracy:.5f} ---")
    print("\nClassification Report:")
    print(classification_report(y_test_official, y_pred_official,
zero division=0))
```

```
else:
    print("\nNo labels found in test.csv to evaluate against.")
print("\n--- Creating submission output file ---")
submission df = pd.DataFrame({
    'id': test df ids,
    'WeightCategory': y pred official
})
output path = '/content/drive/MyDrive/Obesity
Data/gridsearch gb submission.csv'
submission df.to csv(output path, index=False)
print(f"□ Success! Submission file saved to: {output path}")
print(submission df.head())
Loading data from Google Drive...
□ Both train.csv and test.csv loaded successfully.
Preparing 100% of training data...
Preparing test data for prediction...
No test labels found. Skipping evaluation.
Starting preprocessing (OHE & Scaling)...

  □ Preprocessing complete for both datasets.

--- Starting Focused GridSearchCV for GradientBoostingClassifier ---
This may take some time...
Fitting 3 folds for each of 16 candidates, totalling 48 fits
☐ GridSearchCV complete. Time taken: 10638.57 seconds.
--- Best Parameters Found by GridSearch ---
{'learning_rate': 0.03, 'max_depth': 5, 'n_estimators': 1000,
'subsample': 0.8}
Best internal cross-validation accuracy: 0.90047
--- Generating predictions on test.csv using BEST model ---
No labels found in test.csv to evaluate against.
--- Creating submission output file ---
☐ Success! Submission file saved to: /content/drive/MyDrive/Obesity
Data/gridsearch_gb_submission.csv
      id
               WeightCategory
  15533
             Obesity_Type_III
1
  15534
           Overweight Level I
  15535 Overweight Level II
  15536
              Obesity_Type_II
  15537
                Normal Weight
```

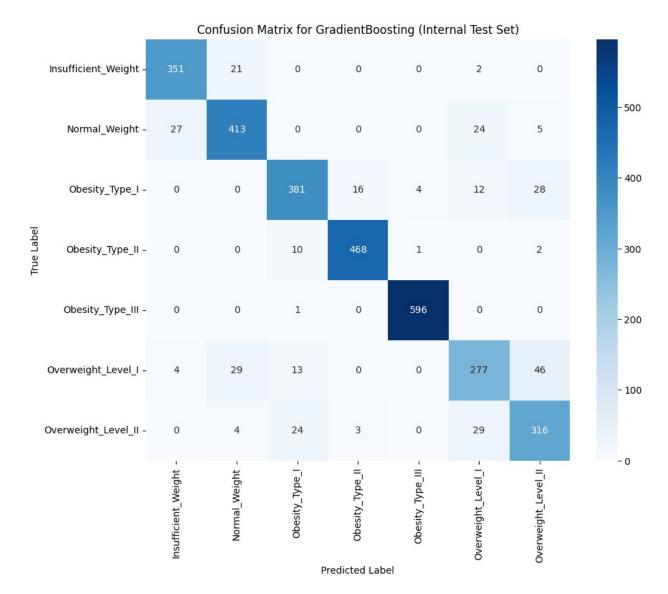
```
import pandas as pd
import numpy as np
import time
import warnings
from sklearn.metrics import accuracy score, classification report,
confusion matrix
import seaborn as sns
import matplotlib.pyplot as plt
from sklearn.preprocessing import StandardScaler
from sklearn.model selection import train test split
from sklearn.ensemble import GradientBoostingClassifier
warnings.filterwarnings("ignore")
print("Loading data from Google Drive...")
file path = '/content/drive/MyDrive/Obesity Data/'
df train = pd.read csv(file path + 'train.csv')
print("[] Training data loaded successfully.")
print("Preparing and splitting data (80% train, 20% test)...")
df = df_train.drop('id', axis=1)
X = df.drop('WeightCategory', axis=1)
y = df['WeightCategory']
X train, X test, y train, y test = train test split(
    Χ, γ,
    test size=0.2,
    random state=42,
    stratify=y
)
print(f"Internal training set size: {len(X train)} samples")
print(f"Internal testing set size: {len(X test)} samples")
print("Starting preprocessing (OHE & Scaling)...")
numerical cols = ['Age', 'Height', 'Weight', 'FCVC', 'NCP', 'CH2O',
'FAF', 'TUE']
categorical cols = X train.select dtypes(include=['object']).columns
X train cat = pd.get dummies(X train[categorical cols],
drop first=True, dtype=int)
X test cat = pd.get dummies(X test[categorical cols], drop first=True,
dtype=int)
X train cat, X test cat = X train cat.align(X test cat, join='left',
axis=1, fill value=0)
X_train_manual = X_train.drop(columns=categorical cols)
X test manual = X test.drop(columns=categorical cols)
X train manual = pd.concat([X train manual.reset index(drop=True),
```

```
X train cat.reset index(drop=True)], axis=1)
X test manual = pd.concat([X test manual.reset index(drop=True),
X test cat.reset index(drop=True)], axis=1)
scaler = StandardScaler()
scaler.fit(X train manual[numerical cols])
X train manual[numerical cols] =
scaler.transform(X train manual[numerical cols])
X test manual[numerical cols] =
scaler.transform(X test manual[numerical cols])
print("□ Preprocessing complete.")
print("\n--- Training GradientBoostingClassifier ---")
start time = time.time()
gb params = {
    'n_estimators': 1000,
    'learning rate': 0.018570273514045024,
    'max depth': 5,
    'subsample': 0.7549164768308696,
    'min_samples_split': 3,
    'min samples leaf': 2,
    'max features': 'sqrt',
    'random state': 42
}
gb_model = GradientBoostingClassifier(**gb_params) # Define gb_model
gb model.fit(X train manual, y train)
end time = time.time()
print(f"[] Training complete. Time taken: {end_time - start time:.2f}
seconds.")
print("\n--- Evaluating Model on Internal Test Set (20%) ---")
y pred = gb model.predict(X test manual)
accuracy = accuracy score(y test, y pred)
print(f"--- FINAL ACCURACY ON TEST SET (GradientBoosting):
{accuracy:.5f} ---")
print("\nClassification Report:")
print(classification report(y test, y pred, zero division=0))
print("\nConfusion Matrix:")
plt.figure(figsize=(10, 8))
labels = np.unique(y test)
sns.heatmap(confusion matrix(y test, y pred),
            annot=True, fmt='d', cmap='Blues',
```

```
xticklabels=labels, yticklabels=labels)
plt.title('Confusion Matrix for GradientBoosting (Internal Test Set)')
plt.xlabel('Predicted Label')
plt.ylabel('True Label')
plt.show()
Loading data from Google Drive...
□ Training data loaded successfully.
Preparing and splitting data (80% train, 20% test)...
Internal training set size: 12426 samples
Internal testing set size: 3107 samples
Starting preprocessing (OHE & Scaling)...

    □ Preprocessing complete.

--- Training GradientBoostingClassifier ---
☐ Training complete. Time taken: 270.47 seconds.
--- Evaluating Model on Internal Test Set (20%) ---
--- FINAL ACCURACY ON TEST SET (GradientBoosting): 0.90183 ---
Classification Report:
                     precision recall f1-score
                                                      support
                                     0.94
                                               0.93
Insufficient Weight
                          0.92
                                                           374
      Normal Weight
                           0.88
                                     0.88
                                               0.88
                                                           469
                                                           441
     Obesity Type I
                          0.89
                                     0.86
                                               0.88
    Obesity_Type_II
                          0.96
                                     0.97
                                               0.97
                                                           481
   Obesity_Type_III
                                               0.99
                          0.99
                                     1.00
                                                           597
 Overweight Level I
                                                           369
                          0.81
                                     0.75
                                               0.78
Overweight Level II
                                     0.84
                                               0.82
                          0.80
                                                           376
           accuracy
                                               0.90
                                                          3107
                                     0.89
                                               0.89
                                                          3107
                          0.89
          macro avq
       weighted avg
                          0.90
                                     0.90
                                               0.90
                                                          3107
Confusion Matrix:
```



```
import pandas as pd
import numpy as np
import time
import warnings
from sklearn.metrics import accuracy_score, classification_report
from sklearn.preprocessing import StandardScaler
from sklearn.model_selection import train_test_split
from sklearn.ensemble import GradientBoostingClassifier
import seaborn as sns
import matplotlib.pyplot as plt

warnings.filterwarnings("ignore")

print("Loading data from Google Drive...")
file_path_base = '/content/drive/MyDrive/Obesity Data/'
df_train = pd.read_csv(file_path_base + 'train.csv')
```

```
df test = pd.read csv(file path base + 'test.csv')
print("□ Data loaded successfully.")
print("Preparing and splitting data (80% train, 20% test)...")
df = df train.drop('id', axis=1)
X = df.drop('WeightCategory', axis=1)
y = df['WeightCategory']
X train, X test, y train, y test = train test split(
    Х, у,
    test size=0.2,
    random state=42,
    stratify=y
)
print(f"Internal training set size: {len(X train)} samples")
print(f"Internal testing set size: {len(X test)} samples")
print("Starting preprocessing...")
numerical cols = ['Age', 'Height', 'Weight', 'FCVC', 'NCP', 'CH2O',
'FAF', 'TUE']
categorical cols = X train.select dtypes(include=['object']).columns
X train cat = pd.get dummies(X train[categorical cols],
drop first=True, dtype=int)
X test cat = pd.get dummies(X test[categorical cols], drop first=True,
dtype=int)
X train cat, X test cat = X train cat.align(X test cat, join='left',
axis=1, fill value=0)
X train manual = X train.drop(columns=categorical cols)
X test manual = X_test.drop(columns=categorical_cols)
X train manual = pd.concat([X train manual.reset index(drop=True),
X train cat.reset index(drop=True)], axis=1)
X test manual = pd.concat([X test manual.reset index(drop=True),
X test cat.reset index(drop=True)], axis=1)
scaler = StandardScaler()
scaler.fit(X train manual[numerical cols])
X train manual[numerical cols] =
scaler.transform(X train manual[numerical cols])
X test manual[numerical cols] =
scaler.transform(X test manual[numerical cols])
print("□ Preprocessing complete.")
```

```
print("\n--- Training GradientBoostingClassifier ---")
start time = time.time()
gb_params = {
    'n estimators': 1000,
    'learning rate': 0.018570273514045024,
    'max depth': 5,
    'subsample': 0.7549164768308696,
    'min_samples_split': 3,
    'min samples leaf': 2,
    'max features': 'sqrt',
    'random state': 42
}
gb model = GradientBoostingClassifier(**gb params)
gb model.fit(X train manual, y train)
end time = time.time()
print(f"[] Training complete. Time taken: {end_time - start time:.2f}
seconds.")
print("\n--- Evaluating Model on Internal Test Set (20%) ---")
y pred = gb model.predict(X test manual)
accuracy = accuracy_score(y_test, y_pred)
print(f"--- FINAL ACCURACY ON TEST SET (GradientBoosting):
{accuracy:.5f} ---")
print("\nClassification Report:")
print(classification_report(y_test, y_pred, zero_division=0))
Loading data from Google Drive...
□ Data loaded successfully.
Preparing and splitting data (80% train, 20% test)...
Internal training set size: 12426 samples
Internal testing set size: 3107 samples
Starting preprocessing...
□ Preprocessing complete.
--- Training GradientBoostingClassifier ---
☐ Training complete. Time taken: 105.41 seconds.
--- Evaluating Model on Internal Test Set (20%) ---
--- FINAL ACCURACY ON TEST SET (GradientBoosting): 0.89990 ---
Classification Report:
                     precision recall f1-score
                                                     support
Insufficient Weight
                                    0.94
                          0.92
                                              0.93
                                                         374
```

Normal_Weight	0.89	0.88	0.88	469
Obesity_Type_I	0.89	0.86	0.88	441
Obesity_Type_II	0.96	0.98	0.97	481
Obesity_Type_III	0.99	1.00	1.00	597
Overweight_Level_I	0.78	0.75	0.77	369
Overweight_Level_II	0.79	0.83	0.81	376
accuracy			0.90	3107
macro avg	0.89	0.89	0.89	3107
weighted avg	0.90	0.90	0.90	3107

#XG Boosting Best Output/Test Accuracy: - .89614

Train-Test split file for checking validation accuracy

```
import pandas as pd
import numpy as np
import time
import warnings
from sklearn.metrics import accuracy score, classification report
from sklearn.preprocessing import StandardScaler, LabelEncoder
from sklearn.model selection import train test split
import xgboost as xgb
warnings.filterwarnings("ignore")
print("Loading train.csv...")
file path base = '/content/drive/MyDrive/Obesity Data/'
df = pd.read csv(file path base + 'train.csv')
print("□ train.csv loaded.")
X = df.drop(['id', 'WeightCategory'], axis=1)
v = df['WeightCategory']
categorical cols = X.select dtypes(include=['object']).columns
X = pd.get dummies(X, columns=categorical_cols, drop_first=True)
X_train, X_val, y_train, y_val = train_test_split(
    X, y, test_size=0.20, stratify=y, random_state=42
numerical cols =
['Age','Height','Weight','FCVC','NCP','CH20','FAF','TUE']
numerical cols present = [col for col in numerical cols if col in
X train.columnsl
scaler = StandardScaler()
X_train[numerical_cols_present] =
scaler.fit transform(X train[numerical cols present])
X val[numerical cols present] =
```

```
scaler.transform(X val[numerical cols present])
le = LabelEncoder()
y_train_encoded = le.fit_transform(y_train)
v val encoded = le.transform(y_val)
print("\nTraining XGBoost on 80% of data...")
start = time.time()
model = xqb.XGBClassifier(
    objective='multi:softmax',
    num class=len(v.unique()),
    n estimators=1000,
    learning rate=0.01,
    max depth=6,
    subsample=0.9,
    colsample_bytree=0.8,
    use label encoder=False,
    eval metric='mlogloss',
    random state=42,
    early stopping rounds=50
#.90473 basic settings, actual accuracy:- 0.90358
#.90666 0.01 LR
#.90763 0.01 LR 0.9 subsample actual accuracy:- 0.90936
#.90666 0.01 LR 0.9 subsample 0.9 colsample bytree
#.90505 0.01 LR 0.9 subsample 7 max depth
#.90183 0.01 LR 0.9 subsample 5 max depth
#.90602 2000 trees 0.01 LR 0.9 subsample
#.90602 1500 trees 0.01 LR 0.9 subsample
#.90602 1200 trees 0.01 LR 0.9 subsample
#.90005 0.005 LR 0.9 subsample
#.90537 0.01 LR 0.9 subsample 50 random state
#.90634 0.02 LR 0.9 subsample
#.90505 0.015 LR 0.9 subsample
#.90763 0.01 LR 0.95 subsample
#.90666 0.01 LR 1 subsample
#.90537 0.01 LR 0.9 subsample 0.7 colsample bytree
model.fit(X train, y train encoded,
          eval_set=[(X_val, y_val_encoded)],
          verbose=False)
end = time.time()
print(f"Training completed in {(end - start):.2f} seconds.")
y pred encoded = model.predict(X val)
y pred = le.inverse transform(y pred encoded)
acc = accuracy_score(y_val, y_pred)
print(f"\nValidation Accuracy (20% split): {acc:.5f}")
```

```
print("\nClassification Report:")
print(classification report(y val, y pred, zero division=0))
Loading train.csv...

    □ train.csv loaded.

Training XGBoost on 80% of data...
Training completed in 45.59 seconds.
Validation Accuracy (20% split): 0.90602
Classification Report:
                      precision recall f1-score
                                                        support
Insufficient Weight
                           0.92
                                      0.94
                                                 0.93
                                                             374
      Normal Weight
                            0.89
                                      0.89
                                                 0.89
                                                             469
     Obesity Type I
                           0.90
                                      0.87
                                                 0.88
                                                             441
   Obesity_Type_II
Obesity_Type_III
                           0.96
                                      0.97
                                                 0.97
                                                             481
                           0.99
                                      1.00
                                                 0.99
                                                             597
 Overweight Level I
                                                 0.78
                                                             369
                           0.81
                                      0.76
Overweight Level II
                                      0.84
                                                 0.83
                           0.81
                                                             376
           accuracy
                                                 0.91
                                                            3107
                                                 0.90
                                                            3107
                            0.90
                                      0.90
          macro avq
       weighted avg
                            0.91
                                      0.91
                                                 0.91
                                                            3107
```

Code for generating CSV file to be checked for output/test accuracy

```
import pandas as pd
import numpy as np
import time
import warnings
from sklearn.metrics import accuracy score, classification report
from sklearn.preprocessing import StandardScaler, LabelEncoder
from sklearn.model selection import train test split
import xgboost as xgb
warnings.filterwarnings("ignore")
print("Loading train.csv...")
file path base = '/content/drive/MyDrive/Obesity Data/'
df = pd.read csv(file path base + 'train.csv')
print("□ train.csv loaded.")
X = df.drop(['id', 'WeightCategory'], axis=1)
y = df['WeightCategory']
categorical cols = X.select dtypes(include=['object']).columns
X = pd.get dummies(X, columns=categorical cols, drop first=True)
```

```
X_train, X_val, y_train, y_val = train_test_split(
    X, y, test size=0.20, stratify=y, random state=42
numerical cols =
['Age','Height','Weight','FCVC','NCP','CH20','FAF','TUE']
numerical cols present = [col for col in numerical cols if col in
X train.columns]
scaler = StandardScaler()
X train[numerical cols present] =
scaler.fit transform(X train[numerical cols present])
X_val[numerical_cols_present] =
scaler.transform(X val[numerical cols present])
le = LabelEncoder()
y_train_encoded = le.fit_transform(y_train)
y val encoded = le.transform(y val)
print("\nTraining XGBoost on 80% of data...")
start = time.time()
model = xgb.XGBClassifier(
    objective='multi:softmax',
    num class=len(y.unique()),
    n_estimators=1000,
    learning rate=0.01,
    max_depth=6,
    subsample=0.9,
    colsample bytree=0.8,
    use label encoder=False,
    eval metric='mlogloss',
    random state=42,
    early stopping rounds=50
)
model.fit(X_train, y_train_encoded,
          eval set=[(X val, y val encoded)],
          verbose=False)
end = time.time()
print(f"Training completed in {(end - start):.2f} seconds.")
y pred encoded = model.predict(X val)
y_pred = le.inverse_transform(y_pred_encoded)
acc = accuracy score(y val, y pred)
print(f"\nValidation Accuracy (20% split): {acc:.5f}")
print("\nClassification Report:")
print(classification report(y val, y pred, zero division=0))
```

```
Loading train.csv...

  □ train.csv loaded.

Training XGBoost on 80% of data...
Training completed in 95.35 seconds.
Validation Accuracy (20% split): 0.90602
Classification Report:
                                    recall f1-score
                      precision
                                                        support
Insufficient Weight
                           0.92
                                      0.94
                                                 0.93
                                                             374
      Normal Weight
                           0.89
                                      0.89
                                                 0.89
                                                             469
     Obesity_Type_I
                           0.90
                                      0.87
                                                 0.88
                                                             441
    Obesity_Type_TI
                           0.96
                                      0.97
                                                 0.97
                                                             481
   Obesity Type \overline{I}II
                           0.99
                                      1.00
                                                 0.99
                                                             597
Overweight Level I
                           0.81
                                      0.76
                                                 0.78
                                                             369
Overweight Level II
                           0.81
                                      0.84
                                                 0.83
                                                             376
                                                 0.91
                                                            3107
           accuracy
                           0.90
                                      0.90
                                                 0.90
                                                            3107
          macro avg
                           0.91
                                      0.91
                                                 0.91
                                                            3107
       weighted avg
```

RandomSearchCV on xgboost

would result in .90661 output accuracy

```
import pandas as pd
import warnings
import time
from sklearn.preprocessing import StandardScaler, LabelEncoder
from sklearn.model selection import RandomizedSearchCV,
StratifiedKFold
import xgboost as xgb
from scipy.stats import uniform, randint
warnings.filterwarnings("ignore")
print("Loading train.csv and test.csv...")
file path base = '/content/drive/MyDrive/Obesity Data/'
df train = pd.read csv(file path base + 'train.csv')
df test = pd.read csv(file path base + 'test.csv')
print("□ Data loaded.")
X train = df train.drop(['id', 'WeightCategory'], axis=1)
y_train = df_train['WeightCategory']
categorical cols = X train.select dtypes(include=['object']).columns
X train = pd.get dummies(X train, columns=categorical cols,
```

```
drop first=True)
numerical cols =
['Age','Height','Weight','FCVC','NCP','CH20','FAF','TUE']
numerical cols present = [c for c in numerical cols if c in
X train.columnsl
scaler = StandardScaler()
X train[numerical cols present] =
scaler.fit transform(X train[numerical cols present])
le = LabelEncoder()
y train encoded = le.fit transform(y train)
param dist = {
    'n estimators': [800, 900, 1000, 1100, 1200],
    'learning rate': uniform(0.001, 0.099),
    'max depth': randint(4, 11),
    'subsample': uniform(0.6, 0.4),
    'colsample bytree': uniform(0.6, 0.4),
    'min child weight': randint(1, 11),
    'reg alpha': uniform(0.0, 0.6),
    'reg lambda': uniform(0.5, 3.5),
    'gamma': uniform(0.0, 0.5),
    'objective': ['multi:softmax'],
    'num class': [len(le.classes )],
    'use label_encoder': [False],
    'eval metric': ['mlogloss'],
    'random state': [42],
    'n jobs': [-1],
    'verbosity': [0]
}
print("\n" + "="*80)
print("STARTING BROADENED RANDOMIZED SEARCH - ALL HYPERPARAMETERS")
print("="*80)
print("Tuning: n estimators (800-1200), learning rate (0.001-0.1),")
print("
               max depth (4-10), subsample (0.6-1.0), colsample bytree
(0.6-1.0),")
               min child weight (1-10), reg alpha (0.0-0.6),
print("
reg_lambda (0.5-4.0),")
print("
               gamma (0.0-0.5)")
print("="*80)
xqb clf = xqb.XGBClassifier()
cv = StratifiedKFold(n splits=5, shuffle=True, random state=42)
random search = RandomizedSearchCV(
    estimator=xqb clf,
    param distributions=param dist,
    n iter=200,
```

```
cv=cv,
    verbose=2,
    random state=42,
    n jobs=-1,
    scoring='accuracy',
    return train score=True
random search.fit(X train, y train encoded)
best model = xgb.XGBClassifier(**random search.best params )
best model.fit(X_train, y_train_encoded)
print("□ Final model trained.")
X test = df test.drop(['id'], axis=1)
X test = pd.get dummies(X test, columns=categorical cols,
drop first=True)
X test = X test.reindex(columns=X train.columns, fill value=0)
X test[numerical cols present] =
scaler.transform(X test[numerical cols present])
y test pred encoded = best model.predict(X test)
y test pred = le.inverse transform(y test pred encoded)
results df = pd.DataFrame(random search.cv results )
submission = pd.DataFrame({
    'id': df test['id'],
    'WeightCategory': y_test_pred
})
output path = file path base + 'randomsearch broadened full.csv'
submission.to csv(output path, index=False)
print(f"\n□ Submission saved to: {output path}")
print(submission.head())
results df.to csv(file path base +
'xg_randomsearch_regularization_3.csv', index=False)
print(f" Full search results saved.")
import pandas as pd
import numpy as np
import time
import warnings
from sklearn.preprocessing import StandardScaler, LabelEncoder
from sklearn.model selection import StratifiedKFold, cross val score
import xgboost as xgb
import optuna
warnings.filterwarnings("ignore")
```

```
print("Loading train.csv and test.csv...")
file path base = '/content/drive/MyDrive/Obesity Data/'
df train = pd.read csv(file path base + 'train.csv')
df test = pd.read csv(file path base + 'test.csv')
print("□ Data loaded.")
X_train = df_train.drop(['id', 'WeightCategory'], axis=1)
y train = df train['WeightCategory']
categorical cols = X train.select dtypes(include=['object']).columns
X train = pd.get dummies(X train, columns=categorical cols,
drop first=True)
numerical cols =
['Age','Height','Weight','FCVC','NCP','CH20','FAF','TUE']
numerical cols present = [col for col in numerical cols if col in
X train.columnsl
scaler = StandardScaler()
X train[numerical_cols_present] =
scaler.fit transform(X train[numerical cols present])
le = LabelEncoder()
y train encoded = le.fit_transform(y_train)
def objective(trial):
    params = {
        'n estimators': trial.suggest int('n estimators', 900, 1200),
        'learning rate': trial.suggest float('learning rate', 0.009,
0.011),
        'max depth': trial.suggest int('max depth', 5, 7),
        'subsample': trial.suggest float('subsample', 0.90, 0.95),
        'colsample bytree': trial.suggest float('colsample bytree',
0.90, 0.95),
        'reg alpha': trial.suggest float('reg alpha', 0.8, 1.0),
        'reg lambda': trial.suggest float('reg lambda', 0.8, 1.2),
        'objective': 'multi:softmax',
        'num_class': len(le.classes_),
        'use label encoder': False,
        'eval_metric': 'mlogloss',
        'random state': 42,
        'n jobs': -1,
        'verbosity': 0
    }
    model = xqb.XGBClassifier(**params)
    cv = StratifiedKFold(n splits=3, shuffle=True, random state=42)
    scores = cross_val_score(model, X_train, y_train_encoded, cv=cv,
scoring='accuracy', n jobs=-1)
    return scores.mean()
print("\nStarting Optuna hyperparameter optimization...")
```

```
study = optuna.create study(direction='maximize',
sampler=optuna.samplers.TPESampler(seed=42))
start time = time.time()
study.optimize(objective, n trials=50, show progress bar=True)
end time = time.time()
print(f"\nOptimization finished in {(end_time - start_time) / 60:.2f}
minutes")
print(f"Best Accuracy: {study.best value:.5f}")
print(f"Best Parameters: {study.best_params}")
print("\nTraining final model on full training data...")
best params = study.best params
best params['objective'] = 'multi:softmax'
best params['num class'] = len(le.classes )
best_params['use_label_encoder'] = False
best params['eval metric'] = 'mlogloss'
best params['random state'] = 42
best params['n jobs'] = -1
best params['verbosity'] = 0
model = xgb.XGBClassifier(**best params)
model.fit(X_train, y_train_encoded)
print("Training complete.")
X test = df test.drop('id', axis=1)
X test = pd.get dummies(X test, columns=categorical cols,
drop first=True)
X test = X test.reindex(columns=X train.columns, fill value=0)
X test[numerical cols present] =
scaler.transform(X test[numerical cols present])
print("Predicting test data...")
y test pred encoded = model.predict(X test)
y test pred = le.inverse transform(y test pred encoded)
submission = pd.DataFrame({'id': df test['id'], 'WeightCategory':
y test pred})
output path = file path base + 'optuna xqb submission.csv'
submission.to csv(output path, index=False)
print(f"\n[ Submission saved to: {output path}")
print(submission.head())
Loading data from Google Drive...
□ Both train.csv and test.csv loaded successfully.
Preparing 100% of training data...
Preparing test data for prediction...
No test labels found. Skipping evaluation.
Starting preprocessing (OHE & Scaling)...

  □ Preprocessing complete for both datasets.
```

```
--- Starting Focused GridSearchCV for GradientBoostingClassifier ---
This may take some time...
Fitting 3 folds for each of 16 candidates, totalling 48 fits
☐ GridSearchCV complete. Time taken: 10638.57 seconds.
--- Best Parameters Found by GridSearch ---
{'learning rate': 0.03, 'max depth': 5, 'n estimators': 1000,
'subsample': 0.8}
Best internal cross-validation accuracy: 0.90047
--- Generating predictions on test.csv using BEST model ---
No labels found in test.csv to evaluate against.
--- Creating submission output file ---

☐ Success! Submission file saved to: /content/drive/MyDrive/Obesity

Data/gridsearch gb submission.csv
      id
               WeightCategory
  15533
             Obesity_Type_III
  15534 Overweight Level I
2 15535 Overweight Level II
3 15536
              Obesity Type II
                Normal Weight
4 15537
import pandas as pd
import numpy as np
import warnings
import time
from sklearn.preprocessing import StandardScaler, LabelEncoder
from sklearn.model selection import StratifiedKFold, cross val score
import xgboost as xgb
import optuna
warnings.filterwarnings("ignore")
print("Loading train.csv...")
file path base = '/content/drive/MyDrive/Obesity Data/'
df train = pd.read csv(file path base + 'train.csv')
df test = pd.read csv(file path base + 'test.csv')
print("□ Data loaded.")
X train = df train.drop(['id', 'WeightCategory'], axis=1)
y train = df train['WeightCategory']
categorical cols = X train.select dtypes(include=['object']).columns
X_train = pd.get_dummies(X_train, columns=categorical cols,
drop_first=True)
numerical cols =
```

```
['Age','Height','Weight','FCVC','NCP','CH20','FAF','TUE']
numerical cols present = [c for c in numerical cols if c in
X train.columns]
scaler = StandardScaler()
X train[numerical cols present] =
scaler.fit transform(X train[numerical cols present])
le = LabelEncoder()
y_train_encoded = le.fit_transform(y_train)
def objective(trial):
    params = {
        'n estimators': 1000,
        'learning rate': 0.01,
        'max depth': 6,
        'subsample': 0.8,
        'colsample_bytree': 0.8,
        'reg alpha': trial.suggest float('reg alpha', 0.15, 0.55),
        'reg_lambda': trial.suggest_float('reg_lambda', 0.8, 3.2),
        'objective': 'multi:softmax',
        'num class': len(le.classes ),
        'use label encoder': False,
        'eval metric': 'mlogloss',
        'random state': 42,
        'n jobs': -1,
        'verbosity': 0
    }
    model = xqb.XGBClassifier(**params)
    cv = StratifiedKFold(n_splits=5, shuffle=True, random_state=42)
    scores = cross val score(model, X train, y train encoded, cv=cv,
scoring='accuracy', n_jobs=-1)
    return scores.mean()
print("\n" + "="*80)
print("STARTING OPTUNA REGULARIZATION TUNING")
print("="*80)
print("Fixed parameters: n estimators=1000, lr=0.01, max depth=6,
subsample=0.8")
print("Tuning: reg alpha (0.15-0.55), reg lambda (0.8-3.2)")
print("="*80)
study = optuna.create study(direction='maximize',
sampler=optuna.samplers.TPESampler(seed=42))
start time = time.time()
study.optimize(objective, n_trials=100, show_progress_bar=True)
end time = time.time()
print(f"\n[ Optuna finished in {(end time - start time)/60:.2f}
minutes")
print(f"□ Best CV Accuracy: {study.best value:.5f}")
```

```
print(f"□ Best Parameters:")
for k, v in study.best params.items():
    print(f"
              \{k\}: \{v:.4f\}"\}
print("\n□ Top 10 Trials:")
trials df = study.trials dataframe().sort values('value',
ascending=False).head(10)
for idx, row in trials df.iterrows():
    print(f"Trial {int(row['number'])}: Accuracy={row['value']:.5f}, "
          f"alpha={row['params reg alpha']:.4f},
lambda={row['params reg lambda']:.4f}")
print("\n" + "="*80)
print("TRAINING FINAL MODEL ON FULL DATA")
print("="*80)
best params = {
    'n estimators': 1000,
    'learning rate': 0.01,
    'max depth': 6,
    'subsample': 0.8,
    'colsample bytree': 0.8,
    'reg alpha': study.best params['reg alpha'],
    'reg_lambda': study.best_params['reg_lambda'],
'objective': 'multi:softmax',
    'num class': len(le.classes ),
    'use label encoder': False,
    'eval metric': 'mlogloss',
    'random state': 42,
    'n jobs': -1,
    'verbosity': 0
}
final model = xqb.XGBClassifier(**best params)
final model.fit(X train, y train encoded)
print("□ Final model trained.")
X test = df test.drop(['id'], axis=1)
X test = pd.get dummies(X test, columns=categorical cols,
drop first=True)
X test = X test.reindex(columns=X train.columns, fill value=0)
X test[numerical cols present] =
scaler.transform(X test[numerical cols present])
y test pred encoded = final model.predict(X test)
y test pred = le.inverse transform(y test pred encoded)
submission = pd.DataFrame({
    'id': df test['id'],
    'WeightCategory': y test pred
```

```
})
output path = file path base + 'optuna regularization 2.csv'
submission.to csv(output path, index=False)
print(f"\n□ Submission saved to: {output path}")
print(submission.head())
Loading train.csv...
[I 2025-10-22 02:29:36,198] A new study created in memory with name:
no-name-fadflac1-36b1-4991-b66c-a463ed592a78

□ Data loaded.

STARTING OPTUNA REGULARIZATION TUNING
Fixed parameters: n estimators=1000, lr=0.01, max depth=6,
subsample=0.8
Tuning: reg alpha (0.15-0.55), reg lambda (0.8-3.2)
{"model id":"0c46d14ed30447dfa5fb38e73e5ff2d0","version major":2,"vers
ion minor":0}
[I 2025-10-22 02:31:10,620] Trial 0 finished with value:
0.9051055392648261 and parameters: {'reg_alpha': 0.299816047538945,
'reg lambda': 3.081714335383799}. Best is trial 0 with value:
0.9051055392648261.
[I 2025-10-22 02:32:38,203] Trial 1 finished with value:
0.9049767355395281 and parameters: {'reg_alpha': 0.442797576724562,
'req lambda': 2.236780362072888}. Best is trial 0 with value:
0.9051055392648261.
[I 2025-10-22 02:34:07,836] Trial 2 finished with value:
0.9049123647638602 and parameters: {'req alpha': 0.2124074561769746,
'reg lambda': 1.1743868488068865}. Best is trial 0 with value:
0.9051055392648261.
[I 2025-10-22 02:35:37,617] Trial 3 finished with value:
0.9052342600915075 and parameters: {'reg_alpha': 0.1732334448672798,
'reg lambda': 2.878822749859845}. Best is trial 3 with value:
0.9052342600915075.
[I 2025-10-22 02:37:03,621] Trial 4 finished with value:
0.9049767355395281 and parameters: {'reg_alpha': 0.3904460046972835,
'reg lambda': 2.4993741867105097}. Best is trial 3 with value:
0.9052342600915075.
[I 2025-10-22 02:38:34,041] Trial 5 finished with value:
0.9048480147128464 and parameters: {'reg alpha': 0.15823379771832097,
'reg lambda': 3.127783645188787}. Best is trial 3 with value:
```

```
0.9052342600915075.
[I 2025-10-22 02:40:04,174] Trial 6 finished with value:
0.9047192524368567 and parameters: {'reg alpha': 0.48297705632016874,
'reg lambda': 1.3096138656278629}. Best is trial 3 with value:
0.9052342600915075.
[I 2025-10-22 02:41:32,852] Trial 7 finished with value:
0.9049123854885144 and parameters: {'reg alpha': 0.22272998688284024,
'reg lambda': 1.2401708236482412}. Best is trial 3 with value:
0.9052342600915075.
[I 2025-10-22 02:43:03,744] Trial 8 finished with value:
0.9054274138678193 and parameters: {'reg alpha': 0.27169689718381507,
'reg lambda': 2.059415435917371}. Best is trial 8 with value:
0.9054274138678193.
[I 2025-10-22 02:44:32,307] Trial 9 finished with value:
0.9050411063151959 and parameters: {'reg_alpha': 0.3227780074568463,
'reg lambda': 1.4989499364753007}. Best is trial 8 with value:
0.9054274138678193.
[I 2025-10-22 02:45:59,984] Trial 10 finished with value:
0.9050411892138124 and parameters: {'req alpha': 0.27453813392581927,
'reg lambda': 1.7382532428172006}. Best is trial 8 with value:
0.9054274138678193.
[I 2025-10-22 02:47:31,553] Trial 11 finished with value:
0.9053630223674973 and parameters: {'reg alpha': 0.15301892772651723,
'reg lambda': 2.642380674738459}. Best is trial 8 with value:
0.9054274138678193.
[I 2025-10-22 02:49:00,105] Trial 12 finished with value:
0.9048480354375006 and parameters: {'reg_alpha': 0.26210293005372065,
'reg lambda': 2.5114140457967493}. Best is trial 8 with value:
0.9054274138678193.
[I 2025-10-22 02:50:28,247] Trial 13 finished with value:
0.9051698893158397 and parameters: {'reg alpha': 0.36890713244797957,
'reg lambda': 2.050917006142413}. Best is trial 8 with value:
0.9054274138678193.
[I 2025-10-22 02:51:58,012] Trial 14 finished with value:
0.9049123025898979 and parameters: {'req alpha': 0.2216406757884701,
'reg lambda': 0.8316241579120172}. Best is trial 8 with value:
0.9054274138678193.
[I 2025-10-22 02:53:24,828] Trial 15 finished with value:
0.9049766940902197 and parameters: {'reg_alpha': 0.5370145737504872,
'req lambda': 2.621807916323065}. Best is trial 8 with value:
0.9054274138678193.
[I 2025-10-22 02:54:50,202] Trial 16 finished with value:
0.9052986515918295 and parameters: {'reg_alpha': 0.16553548735476498,
'reg lambda': 1.842455052215387}. Best is trial 8 with value:
0.9054274138678193.
[I 2025-10-22 02:56:20,025] Trial 17 finished with value:
0.9051698893158399 and parameters: {'req alpha': 0.3390856026631795,
'reg lambda': 2.2370222216577917}. Best is trial 8 with value:
0.9054274138678193.
```

```
[I 2025-10-22 02:57:47,941] Trial 18 finished with value:
0.9050411477645041 and parameters: {'reg alpha': 0.2525270962814858,
'reg lambda': 2.822476150612655}. Best is trial 8 with value:
0.9054274138678193.
[I 2025-10-22 02:59:16,943] Trial 19 finished with value:
0.904526140109853 and parameters: {'reg_alpha': 0.40513596450126765,
'req lambda': 2.1966588322117837}. Best is trial 8 with value:
0.9054274138678193.
[I 2025-10-22 03:00:44,197] Trial 20 finished with value:
0.9049767977134904 and parameters: {'req alpha': 0.19796902397837812,
'reg lambda': 1.6302259106638672}. Best is trial 8 with value:
0.9054274138678193.
[I 2025-10-22 03:02:10,822] Trial 21 finished with value:
0.9049123854885144 and parameters: {'reg alpha': 0.1559755702293382,
'reg lambda': 1.8442937850715577}. Best is trial 8 with value:
0.9054274138678193.
[I 2025-10-22 03:03:39,682] Trial 22 finished with value:
0.9048480147128464 and parameters: {'reg_alpha': 0.19348359250801003,
'reg lambda': 1.9382784383972989}. Best is trial 8 with value:
0.9054274138678193.
[I 2025-10-22 03:05:06,836] Trial 23 finished with value:
0.9050411477645041 and parameters: {'req alpha': 0.2400163756541936,
'reg lambda': 2.3869022651158964}. Best is trial 8 with value:
0.9054274138678193.
[I 2025-10-22 03:06:34,145] Trial 24 finished with value:
0.9050411477645042 and parameters: {'reg alpha': 0.28686001610547185,
'reg lambda': 2.035928993650264}. Best is trial 8 with value:
0.9054274138678193.
[I 2025-10-22 03:08:03,237] Trial 25 finished with value:
0.9053630845414598 and parameters: {'reg_alpha': 0.15086383189459862,
'reg lambda': 2.735627787519728}. Best is trial 8 with value:
0.9054274138678193.
[I 2025-10-22 03:09:28,844] Trial 26 finished with value:
0.9047192524368566 and parameters: {'reg alpha': 0.3050503893703481,
'req lambda': 2.750190493460984}. Best is trial 8 with value:
0.9054274138678193.
[I 2025-10-22 03:10:56,086] Trial 27 finished with value:
0.9051055185401718 and parameters: {'reg alpha': 0.19168071535641254,
'reg lambda': 2.9214173723171606}. Best is trial 8 with value:
0.9054274138678193.
[I 2025-10-22 03:12:26,350] Trial 28 finished with value:
0.9052343015408159 and parameters: {'reg alpha': 0.2467826397932537,
'reg lambda': 2.6402156410342155}. Best is trial 8 with value:
0.9054274138678193.
[I 2025-10-22 03:13:56,495] Trial 29 finished with value:
0.9047835610385621 and parameters: {'reg_alpha': 0.30461744579007527,
'req lambda': 3.1941657746246666}. Best is trial 8 with value:
0.9054274138678193.
[I 2025-10-22 03:15:24,118] Trial 30 finished with value:
```

```
0.9049123647638602 and parameters: {'reg alpha': 0.1837947442654918,
'reg lambda': 2.96093370581489}. Best is trial 8 with value:
0.9054274138678193.
[I 2025-10-22 03:16:53,848] Trial 31 finished with value:
0.9054274345924735 and parameters: {'reg alpha': 0.1570786799191149,
'reg lambda': 2.365194826128868}. Best is trial 31 with value:
0.9054274345924735.
[I 2025-10-22 03:18:22,051] Trial 32 finished with value:
0.9054273931431653 and parameters: {'reg alpha': 0.16203711646550578,
'reg lambda': 2.381668651541538}. Best is trial 31 with value:
0.9054274345924735.
[I 2025-10-22 03:19:47,769] Trial 33 finished with value:
0.9051699307651478 and parameters: {'reg_alpha': 0.21524045955539853,
'req lambda': 2.353806275062844}. Best is trial 31 with value:
0.9054274345924735.
[I 2025-10-22 03:21:14,370] Trial 34 finished with value:
0.9051055599894801 and parameters: {'reg alpha': 0.18160849688081246,
'reg_lambda': 2.1720182326930337}. Best is trial 31 with value:
0.9054274345924735.
[I 2025-10-22 03:22:39,599] Trial 35 finished with value:
0.9052343015408159 and parameters: {'req alpha': 0.20683368095163612,
'reg lambda': 2.385677561104611}. Best is trial 31 with value:
0.9054274345924735.
[I 2025-10-22 03:24:04,870] Trial 36 finished with value:
0.9057493299201209 and parameters: {'reg alpha': 0.15074824565543687,
'reg lambda': 2.495546593721333}. Best is trial 36 with value:
0.9057493299201209.
[I 2025-10-22 03:25:33,528] Trial 37 finished with value:
0.9050411684891582 and parameters: {'reg alpha': 0.2288973449866798,
'reg lambda': 2.4629769917066944}. Best is trial 36 with value:
0.9057493299201209.
[I 2025-10-22 03:26:59,321] Trial 38 finished with value:
0.9051055392648261 and parameters: {'reg alpha': 0.17546922196135586,
'reg lambda': 2.3023950195835243}. Best is trial 36 with value:
0.9057493299201209.
[I 2025-10-22 03:28:24,779] Trial 39 finished with value:
0.904461727884877 and parameters: {'reg alpha': 0.43858799654617925,
'reg lambda': 2.144676497053553}. Best is trial 36 with value:
0.9057493299201209.
[I 2025-10-22 03:29:51,317] Trial 40 finished with value:
0.9053630223674975 and parameters: {'reg alpha': 0.2709767368294028,
'reg lambda': 2.5358170507459756}. Best is trial 36 with value:
0.9057493299201209.
[I 2025-10-22 03:31:18,143] Trial 41 finished with value:
0.905620546919477 and parameters: { 'reg alpha': 0.15087976054989324,
'reg lambda': 2.8150242860827874}. Best is trial 36 with value:
0.9057493299201209.
[I 2025-10-22 03:32:45,504] Trial 42 finished with value:
0.9047836232125244 and parameters: {'reg alpha': 0.16966088660292494,
```

```
'reg lambda': 3.0363603683528027}. Best is trial 36 with value:
0.9057493299201209.
[I 2025-10-22 03:34:12,892] Trial 43 finished with value:
0.9048479939881924 and parameters: {'reg alpha': 0.19969944046638907,
'reg lambda': 2.7839069971414}. Best is trial 36 with value:
0.9057493299201209.
[I 2025-10-22 03:35:40,010] Trial 44 finished with value:
0.9052986930411379 and parameters: {'reg alpha': 0.17389060804729375,
'reg lambda': 2.5678931373545293}. Best is trial 36 with value:
0.9057493299201209.
[I 2025-10-22 03:37:09,213] Trial 45 finished with value:
0.9051055599894802 and parameters: {'reg alpha': 0.15244169026119697,
'req lambda': 2.0887805641106274}. Best is trial 36 with value:
0.9057493299201209.
[I 2025-10-22 03:38:33,362] Trial 46 finished with value:
0.9049767769888362 and parameters: {'reg alpha': 0.23183678548587855,
'reg lambda': 2.4359177631102864}. Best is trial 36 with value:
0.9057493299201209.
[I 2025-10-22 03:39:59,782] Trial 47 finished with value:
0.9047836232125246 and parameters: {'reg_alpha': 0.3702277783201279,
'reg lambda': 2.272846663485144}. Best is trial 36 with value:
0.9057493299201209.
[I 2025-10-22 03:41:26,791] Trial 48 finished with value:
0.9050411684891582 and parameters: {'reg alpha': 0.21270216727747768,
'reg lambda': 2.7121064793679004}. Best is trial 36 with value:
0.9057493299201209.
[I 2025-10-22 03:42:54,885] Trial 49 finished with value:
0.9052343222654701 and parameters: {'reg alpha': 0.16776135635435802,
'reg lambda': 1.9387824529711926}. Best is trial 36 with value:
0.9057493299201209.
[I 2025-10-22 03:44:22,064] Trial 50 finished with value:
0.9054274138678196 and parameters: {'reg_alpha': 0.32924742587638145,
'reg lambda': 1.4561647770440007}. Best is trial 36 with value:
0.9057493299201209.
[I 2025-10-22 03:45:51,140] Trial 51 finished with value:
0.9049124062131684 and parameters: {'reg alpha': 0.3216111605400239,
'reg lambda': 0.9458111912892329}. Best is trial 36 with value:
0.9057493299201209.
[I 2025-10-22 03:47:17,688] Trial 52 finished with value:
0.9048480768868089 and parameters: {'reg alpha': 0.34302146505510456,
'reg lambda': 1.6492246413871434}. Best is trial 36 with value:
0.9057493299201209.
[I 2025-10-22 03:48:45,542] Trial 53 finished with value:
0.9049767562641822 and parameters: {'reg alpha': 0.40368183503657196,
'reg lambda': 1.15136749051824}. Best is trial 36 with value:
0.9057493299201209.
[I 2025-10-22 03:50:14,805] Trial 54 finished with value:
0.9051055392648261 and parameters: {'reg alpha': 0.4892063913134844,
'reg lambda': 1.4824644484453366}. Best is trial 36 with value:
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0.9057493299201209.
[I 2025-10-22 03:51:41,599] Trial 55 finished with value:
0.9051055185401718 and parameters: {'reg alpha': 0.36749244810426585,
'reg lambda': 1.8481941887278848}. Best is trial 36 with value:
0.9057493299201209.
[I 2025-10-22 03:53:07,619] Trial 56 finished with value:
0.9047192731615107 and parameters: {'req alpha': 0.3250942069208007,
'reg lambda': 2.310957880096248}. Best is trial 36 with value:
0.9057493299201209.
[I 2025-10-22 03:54:36,527] Trial 57 finished with value:
0.9049123854885144 and parameters: {'reg_alpha': 0.16313933293960584,
'req lambda': 1.418959649559617}. Best is trial 36 with value:
0.9057493299201209.
[I 2025-10-22 03:56:03,805] Trial 58 finished with value:
0.9047836024878704 and parameters: {'reg_alpha': 0.26167919173414966,
'reg lambda': 1.9572227977890824}. Best is trial 36 with value:
0.9057493299201209.
[I 2025-10-22 03:57:32,351] Trial 59 finished with value:
0.9049123854885142 and parameters: {'req alpha': 0.2888095415218006,
'reg lambda': 2.5724696209187763}. Best is trial 36 with value:
0.9057493299201209.
[I 2025-10-22 03:58:58,171] Trial 60 finished with value:
0.9050411477645041 and parameters: {'reg alpha': 0.1904930832907839,
'reg lambda': 2.8472989589753896}. Best is trial 36 with value:
0.9057493299201209.
[I 2025-10-22 04:00:25,536] Trial 61 finished with value:
0.9053630430921515 and parameters: {'reg_alpha': 0.16430481812385214,
'reg lambda': 3.0179438290943126}. Best is trial 36 with value:
0.9057493299201209.
[I 2025-10-22 04:01:56,040] Trial 62 finished with value:
0.9052986515918295 and parameters: {'reg alpha': 0.1526264069614121,
'req lambda': 2.6887081913179705}. Best is trial 36 with value:
0.9057493299201209.
[I 2025-10-22 04:03:23,937] Trial 63 finished with value:
0.9056205676441312 and parameters: {'req alpha': 0.18360889218265553,
'reg lambda': 2.626875970366356}. Best is trial 36 with value:
0.9057493299201209.
[I 2025-10-22 04:04:54,031] Trial 64 finished with value:
0.9053630845414599 and parameters: {'reg_alpha': 0.18755422612518302,
'reg lambda': 2.475260867877542}. Best is trial 36 with value:
0.9057493299201209.
[I 2025-10-22 04:06:23,753] Trial 65 finished with value:
0.9049767769888362 and parameters: {'reg_alpha': 0.17934245095743895,
'req lambda': 2.100899407943524}. Best is trial 36 with value:
0.9057493299201209.
[I 2025-10-22 04:07:53,389] Trial 66 finished with value:
0.9052343429901241 and parameters: {'reg alpha': 0.20154169768642718,
'reg lambda': 2.6199381842368052}. Best is trial 36 with value:
0.9057493299201209.
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[I 2025-10-22 04:09:21,963] Trial 67 finished with value:
0.9051055599894802 and parameters: {'reg alpha': 0.2174141193850344,
'reg lambda': 2.238974154540568}. Best is trial 36 with value:
0.9057493299201209.
[I 2025-10-22 04:10:48,634] Trial 68 finished with value:
0.9048479939881924 and parameters: {'reg_alpha': 0.1599772231955033,
'reg lambda': 2.8994033640053476}. Best is trial 36 with value:
0.9057493299201209.
[I 2025-10-22 04:12:15,167] Trial 69 finished with value:
0.9049123647638602 and parameters: {'req alpha': 0.23802939113886962,
'reg lambda': 1.7598425813472527}. Best is trial 36 with value:
0.9057493299201209.
[I 2025-10-22 04:13:45,103] Trial 70 finished with value:
0.9050411892138124 and parameters: {'reg alpha': 0.2524477558268357,
'reg lambda': 2.3853456855288098}. Best is trial 36 with value:
0.9057493299201209.
[I 2025-10-22 04:15:13,259] Trial 71 finished with value:
0.9052342808161618 and parameters: {'reg_alpha': 0.18828903425408908,
'reg lambda': 2.468436770498672}. Best is trial 36 with value:
0.9057493299201209.
[I 2025-10-22 04:16:40,371] Trial 72 finished with value:
0.9051055392648261 and parameters: {'reg alpha': 0.18094362417098955,
'req lambda': 2.4956985119572606}. Best is trial 36 with value:
0.9057493299201209.
[I 2025-10-22 04:18:10,409] Trial 73 finished with value:
0.9048480354375006 and parameters: {'reg alpha': 0.20483894237205447,
'reg lambda': 2.812250687639952}. Best is trial 36 with value:
0.9057493299201209.
[I 2025-10-22 04:19:37,178] Trial 74 finished with value:
0.9054274345924735 and parameters: {'reg_alpha': 0.15127325625468657,
'reg lambda': 2.6647971325112088}. Best is trial 36 with value:
0.9057493299201209.
[I 2025-10-22 04:21:04,349] Trial 75 finished with value:
0.9052342808161618 and parameters: {'reg alpha': 0.17028424373748574,
'reg lambda': 2.666094918651514}. Best is trial 36 with value:
0.9057493299201209.
[I 2025-10-22 04:22:33,648] Trial 76 finished with value:
0.9053630430921515 and parameters: {'reg alpha': 0.15201100161136843,
'reg lambda': 2.5859978371778873}. Best is trial 36 with value:
0.9057493299201209.
[I 2025-10-22 04:24:01,745] Trial 77 finished with value:
0.9053630845414598 and parameters: {'reg alpha': 0.16374856659353107,
'reg_lambda': 2.33817547851136}. Best is trial 36 with value:
0.9057493299201209.
[I 2025-10-22 04:25:32,815] Trial 78 finished with value:
0.9052987137657919 and parameters: {'reg_alpha': 0.1502992235387438,
'req lambda': 2.4129834795961624}. Best is trial 36 with value:
0.9057493299201209.
[I 2025-10-22 04:26:59,758] Trial 79 finished with value:
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0.9052343222654701 and parameters: {'reg alpha': 0.17632838062305037,
'reg lambda': 2.185749196384961}. Best is trial 36 with value:
0.9057493299201209.
[I 2025-10-22 04:28:24,649] Trial 80 finished with value:
0.9048479732635382 and parameters: {'reg alpha': 0.28983071931563575,
'reg lambda': 2.978648253721915}. Best is trial 36 with value:
0.9057493299201209.
[I 2025-10-22 04:29:53,390] Trial 81 finished with value:
0.9050411270398501 and parameters: {'reg alpha': 0.1951987359928039,
'reg lambda': 2.753450226587804}. Best is trial 36 with value:
0.9057493299201209.
[I 2025-10-22 04:31:19,770] Trial 82 finished with value:
0.9052342808161618 and parameters: {'reg_alpha': 0.18772575179502132,
'req lambda': 2.5086711196395886}. Best is trial 36 with value:
0.9057493299201209.
[I 2025-10-22 04:32:44,648] Trial 83 finished with value:
0.9054274553171278 and parameters: {'reg alpha': 0.17248678627707345,
'reg lambda': 2.626149776960389}. Best is trial 36 with value:
0.9057493299201209.
[I 2025-10-22 04:34:14,002] Trial 84 finished with value:
0.9047836024878704 and parameters: {'reg alpha': 0.17165892147786452,
'reg lambda': 3.1082615922010617}. Best is trial 36 with value:
0.9057493299201209.
[I 2025-10-22 04:35:41,707] Trial 85 finished with value:
0.9051698893158397 and parameters: {'reg alpha': 0.2094207616951899,
'reg lambda': 2.6481879033162605}. Best is trial 36 with value:
0.9057493299201209.
[I 2025-10-22 04:37:09,727] Trial 86 finished with value:
0.9049767562641821 and parameters: {'reg alpha': 0.15907126372121153,
'reg lambda': 2.875217174997706}. Best is trial 36 with value:
0.9057493299201209.
[I 2025-10-22 04:38:39,409] Trial 87 finished with value:
0.9053630223674975 and parameters: {'reg alpha': 0.16230996520808846,
'reg lambda': 2.555531288623627}. Best is trial 36 with value:
0.9057493299201209.
[I 2025-10-22 04:40:07,184] Trial 88 finished with value:
0.9049123025898979 and parameters: {'reg alpha': 0.42777183465473345,
'reg lambda': 2.7987069456830067}. Best is trial 36 with value:
0.9057493299201209.
[I 2025-10-22 04:41:37,236] Trial 89 finished with value:
0.9051055185401718 and parameters: {'reg alpha': 0.2258961323242087,
'reg lambda': 2.7070472490770117}. Best is trial 36 with value:
0.9057493299201209.
[I 2025-10-22 04:43:03,467] Trial 90 finished with value:
0.9049767148148741 and parameters: {'reg alpha': 0.17807440275823555,
'reg lambda': 1.1244388106273306}. Best is trial 36 with value:
0.9057493299201209.
[I 2025-10-22 04:44:31,080] Trial 91 finished with value:
0.9049767355395281 and parameters: {'reg alpha': 0.3161333513097684,
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'reg lambda': 2.478954659130061}. Best is trial 36 with value:
0.9057493299201209.
[I 2025-10-22 04:46:04,356] Trial 92 finished with value:
0.9054918468174495 and parameters: {'reg alpha': 0.18380086883912042,
'reg lambda': 2.264425909168491}. Best is trial 36 with value:
0.9057493299201209.
[I 2025-10-22 04:47:31,593] Trial 93 finished with value:
0.9051055392648261 and parameters: {'reg alpha': 0.1971630445942012,
'reg lambda': 2.002571159620905}. Best is trial 36 with value:
0.9057493299201209.
[I 2025-10-22 04:48:59,697] Trial 94 finished with value:
0.9052986723164839 and parameters: {'reg_alpha': 0.16984802366991494,
'req lambda': 2.2624482081868176}. Best is trial 36 with value:
0.9057493299201209.
[I 2025-10-22 04:50:27,059] Trial 95 finished with value:
0.9050411477645041 and parameters: {'reg alpha': 0.35577149176735107,
'reg lambda': 2.1053294386863564}. Best is trial 36 with value:
0.9057493299201209.
[I 2025-10-22 04:51:53,632] Trial 96 finished with value:
0.9052986930411377 and parameters: {'reg_alpha': 0.15882588359688904,
'reg lambda': 2.362814436699104}. Best is trial 36 with value:
0.9057493299201209.
[I 2025-10-22 04:53:21,020] Trial 97 finished with value:
0.9049767562641821 and parameters: {'reg alpha': 0.2746591720901031,
'reg lambda': 2.4259903387461907}. Best is trial 36 with value:
0.9057493299201209.
[I 2025-10-22 04:54:48,904] Trial 98 finished with value:
0.9053631052661139 and parameters: {'reg alpha': 0.15013013430377153,
'reg lambda': 2.2074655686079225}. Best is trial 36 with value:
0.9057493299201209.
[I 2025-10-22 04:56:15,529] Trial 99 finished with value:
0.9052343015408159 and parameters: {'reg_alpha': 0.18252514090727825,
'reg lambda': 2.3015563963326358}. Best is trial 36 with value:
0.9057493299201209.
□ Optuna finished in 146.66 minutes
☐ Best CV Accuracy: 0.90575

  □ Best Parameters:

   reg alpha: 0.1507
   reg lambda: 2.4955
□ Top 10 Trials:
Trial 36: Accuracy=0.90575, alpha=0.1507, lambda=2.4955
Trial 63: Accuracy=0.90562, alpha=0.1836, lambda=2.6269
Trial 41: Accuracy=0.90562, alpha=0.1509, lambda=2.8150
Trial 92: Accuracy=0.90549, alpha=0.1838, lambda=2.2644
Trial 83: Accuracy=0.90543, alpha=0.1725, lambda=2.6261
Trial 31: Accuracy=0.90543, alpha=0.1571, lambda=2.3652
Trial 74: Accuracy=0.90543, alpha=0.1513, lambda=2.6648
```

```
Trial 50: Accuracy=0.90543, alpha=0.3292, lambda=1.4562
Trial 8: Accuracy=0.90543, alpha=0.2717, lambda=2.0594
Trial 32: Accuracy=0.90543, alpha=0.1620, lambda=2.3817
TRAINING FINAL MODEL ON FULL DATA
\sqcap Final model trained.
☐ Submission saved to: /content/drive/MyDrive/Obesity
Data/optuna regularization 2.csv
               WeightCategory
      id
0
   15533
             Obesity Type III
1
  15534
           Overweight Level I
  15535
         Overweight Level II
3
  15536
              Obesity Type II
4 15537
                Normal Weight
pip install optuna
Collecting optuna
  Downloading optuna-4.5.0-py3-none-any.whl.metadata (17 kB)
Requirement already satisfied: alembic>=1.5.0 in
/usr/local/lib/python3.12/dist-packages (from optuna) (1.17.0)
Collecting colorlog (from optuna)
  Downloading colorlog-6.10.1-py3-none-any.whl.metadata (11 kB)
Requirement already satisfied: numpy in
/usr/local/lib/python3.12/dist-packages (from optuna) (2.0.2)
Requirement already satisfied: packaging>=20.0 in
/usr/local/lib/python3.12/dist-packages (from optuna) (25.0)
Requirement already satisfied: sqlalchemy>=1.4.2 in
/usr/local/lib/python3.12/dist-packages (from optuna) (2.0.44)
Requirement already satisfied: tgdm in /usr/local/lib/python3.12/dist-
packages (from optuna) (4.67.1)
Requirement already satisfied: PyYAML in
/usr/local/lib/python3.12/dist-packages (from optuna) (6.0.3)
Requirement already satisfied: Mako in /usr/local/lib/python3.12/dist-
packages (from alembic>=1.5.0->optuna) (1.3.10)
Requirement already satisfied: typing-extensions>=4.12 in
/usr/local/lib/python3.12/dist-packages (from alembic>=1.5.0->optuna)
(4.15.0)
Requirement already satisfied: greenlet>=1 in
/usr/local/lib/python3.12/dist-packages (from sqlalchemy>=1.4.2-
>optuna) (3.2.4)
Requirement already satisfied: MarkupSafe>=0.9.2 in
/usr/local/lib/python3.12/dist-packages (from Mako->alembic>=1.5.0-
>optuna) (3.0.3)
Downloading optuna-4.5.0-py3-none-any.whl (400 kB)
```

```
- 400.9/400.9 kB 4.1 MB/s eta
0:00:00
import pandas as pd
import numpy as np
import time
import warnings
from sklearn.preprocessing import StandardScaler, LabelEncoder
from sklearn.model_selection import RepeatedStratifiedKFold
import xgboost as xgb
import optuna
warnings.filterwarnings("ignore")
file path = '/content/drive/MyDrive/Obesity Data/'
df train = pd.read csv(file path + 'train.csv')
df test = pd.read csv(file path + 'test.csv')
X train full = df train.drop(['id','WeightCategory'], axis=1)
y train full = df train['WeightCategory']
print("Starting preprocessing...")
cat features = X train full.select dtypes(include=['object']).columns
X train full = pd.get dummies(X train full, columns=cat features,
drop first=True)
num features =
['Age','Height','Weight','FCVC','NCP','CH20','FAF','TUE']
scaler = StandardScaler()
X train full[num features] =
scaler.fit transform(X train full[num features])
le = LabelEncoder()
y train full enc = le.fit transform(y train full)
print("□ Preprocessing and Encoding complete.")
center params = {
    'n estimators': 1403,
    'learning_rate': 0.011519841358664365,
    'max depth': 13,
    'subsample': 0.8389925410061727,
    'colsample bytree': 0.4289205756686294,
    'min child weight': 1,
    'reg alpha': 0.21720626221992917,
    'reg lambda': 3.274796369552523,
    'gamma': 0.38490059483369904
}
PRUNING THRESHOLD = 0.90473
```

```
def objective(trial):
    params = {
        'n estimators': trial.suggest int('n estimators',
center params['n estimators']-200,
center params['n estimators']+200),
        'learning rate': trial.suggest float('learning rate',
center params['learning rate']/5,
center params['learning rate']*5,
                                              log=True),
        'max depth': trial.suggest int('max depth',
                                        max(3,
center params['max depth']-3),
                                        center params['max depth']+3),
        'subsample': trial.suggest float('subsample',
                                          \max(0.5,
center params['subsample']-0.2),
                                          min(1.0,
center params['subsample']+0.2)),
        'colsample bytree': trial.suggest float('colsample bytree',
                                                 max(0.4,
center params['colsample bytree']-0.2),
                                                 min(1.0,
center params['colsample bytree']+0.2)),
        'min child_weight': trial.suggest_int('min_child_weight',
                                               max(1,
center params['min child weight']-3),
center params['min child weight']+3),
        'reg alpha': trial.suggest float('reg alpha',
                                          center params['reg alpha']/5,
                                          center params['reg alpha']*5,
                                          log=True),
        'reg_lambda': trial.suggest_float('reg_lambda',
center params['reg lambda']/5,
center_params['reg_lambda']*5,
                                           log=True),
        'gamma': trial.suggest float('gamma', 0.3, 0.7),
        'objective': 'multi:softmax',
        'num class': len(le.classes ),
        'use label encoder': False,
        'eval_metric': 'mlogloss',
        'random state': 42,
        'verbosity': 0,
        'n jobs': -1,
```

```
'early stopping rounds': 50
    }
    cv = RepeatedStratifiedKFold(n splits=5, n repeats=3,
random state=42)
    scores = []
    fold count = 0
    for train idx, val idx in cv.split(X train full,
y train full enc):
        X tr, X val = X train full.iloc[train idx],
X train full.iloc[val idx]
        y_tr, y_val = y_train_full enc[train idx],
y train full enc[val idx]
        model = xgb.XGBClassifier(**params)
        model.fit(X_tr, y_tr, eval_set=[(X_val, y_val)],
verbose=False)
        score = model.score(X val, y val)
        scores.append(score)
        fold count += 1
        intermediate mean = np.mean(scores)
        trial.report(intermediate mean, step=fold count)
        if trial.should prune():
            raise optuna.TrialPruned()
    return np.mean(scores)
print(f"\nStarting Optuna optimization with pruning threshold:
{PRUNING THRESHOLD}")
print("Trials with avg accuracy lower than threshold might be stopped
early\n")
pruner = optuna.pruners.MedianPruner(n startup trials=10,
n warmup steps=5)
optuna.logging.set verbosity(optuna.logging.WARNING)
study = optuna.create study(
    direction='maximize',
    sampler=optuna.samplers.TPESampler(seed=42),
    pruner=pruner
)
start = time.time()
study.optimize(objective, n trials=100, show progress bar=True)
print(f"Best Accuracy: {study.best value:.5f}")
print(f"Number of completed trials: {len([t for t in study.trials if
t.state == optuna.trial.TrialState.COMPLETE()) ")
```

```
print(f"Number of pruned trials: {len([t for t in study.trials if
t.state == optuna.trial.TrialState.PRUNED])}")
print("\nBest params found by Optuna:", study.best_params)
```

best optuna for lambda, alpha

```
import pandas as pd
import numpy as np
import time
import warnings
from sklearn.metrics import accuracy score
from sklearn.preprocessing import StandardScaler, LabelEncoder
from sklearn.model_selection import StratifiedKFold
import xgboost as xgb
import optuna
warnings.filterwarnings("ignore")
print("Loading train.csv and test.csv...")
file path base = '/content/drive/MyDrive/Obesity Data/'
try:
    df train = pd.read csv(file path base + 'train.csv')
    df_test = pd.read_csv(file_path_base + 'test.csv')
    print("□ Data loaded.")
except FileNotFoundError:
    print(f"☐ ERROR: Could not find files in {file path base}. Please
check path.")
    raise
X = df train.drop(['id', 'WeightCategory'], axis=1)
y = df train['WeightCategory']
print("Starting preprocessing...")
categorical cols = X.select dtypes(include=['object']).columns
X = pd.get dummies(X, columns=categorical cols, drop first=True)
numerical cols =
['Age','Height','Weight','FCVC','NCP','CH20','FAF','TUE']
numerical cols present train = [col for col in numerical cols if col
in X.columns1
scaler = StandardScaler()
X[numerical_cols_present_train] =
scaler.fit transform(X[numerical cols present train])
le = LabelEncoder()
y_enc = le.fit_transform(y)
print("[ Preprocessing and Encoding complete.")
center params = {
    'n estimators': 1390,
```

```
'learning rate': 0.00680,
    'max depth': 8,
    'subsample': 0.85676,
    'colsample bytree': 0.50965,
    'min child weight': 7,
    'reg_alpha': 0.15512,
    'reg lambda': 3.51511,
}
def objective(trial):
    params = {
        'n estimators': trial.suggest int('n estimators', max(100,
center params['n estimators'] - 400), center params['n estimators'] +
400).
        'learning rate': trial.suggest float('learning rate',
center_params['learning_rate'] / 5, center_params['learning_rate'] *
5, log=True),
        'max depth': trial.suggest int('max depth', max(3,
center params['max depth'] - 3), center params['max depth'] + 3),
        'subsample': trial.suggest_float('subsample', max(0.5,
center params['subsample'] - 0.2), min(1.0, center params['subsample']
+ (0.1)),
        'colsample_bytree': trial.suggest_float('colsample_bytree',
\max(0.4, \text{ center params}['\text{colsample bytree'}] - 0.2), \min(0.8,
center params['colsample_bytree'] + 0.2)),
        'min child weight': trial.suggest int('min child weight',
max(1, center params['min child weight'] - 4),
center params['min child weight'] + 4),
        'reg alpha': trial.suggest float('reg alpha', max(0.0,
center params['reg alpha'] / 5), center params['reg alpha'] * 5,
log=True),
        'reg lambda': trial.suggest float('reg lambda', max(0.0,
center params['reg lambda'] / 5), center params['reg lambda'] * 5,
log=True),
        'objective': 'multi:softmax'
        'num class': len(le.classes ),
        'use_label_encoder': False,
        'eval metric': 'mlogloss',
        'random state': 42,
        'verbosity': 0,
        'n jobs': -1,
        'early stopping rounds': 50
    }
    cv = StratifiedKFold(n splits=5, shuffle=True, random state=42)
    scores = []
    for train idx, val idx in cv.split(X, y enc):
        X tr, X val = X.iloc[train idx], X.iloc[val idx]
        y tr, y val = y enc[train idx], y enc[val idx]
```

```
model = xgb.XGBClassifier(**params)
        model.fit(X tr, y tr, eval set=[(X val, y val)],
verbose=False)
        score = model.score(X val, v val)
        scores.append(score)
        trial.report(score, step=len(scores))
        if trial.should prune():
            raise optuna.TrialPruned()
    return np.mean(scores)
print("\nStarting refined Optuna tuning with early stopping...\n")
study = optuna.create study(direction='maximize',
sampler=optuna.samplers.TPESampler(seed=42))
start = time.time()
study.optimize(objective, n trials=100, show progress bar=True)
end = time.time()
print(f"\nCompleted Optuna tuning in {(end - start)/60:.2f} minutes")
print(f"Best CV accuracy: {study.best value:.5f}")
print("Best hyperparameters found by Optuna:")
best params from optuna = study.best params.copy()
print(best params from optuna)
print("\nTraining final model with best parameters found by
0ptuna...")
start final train = time.time()
best params final = best params from optuna
best params final.update({
    objective': 'multi:softmax',
    'num class': len(le.classes ),
    'use label encoder': False,
    'eval metric': 'mlogloss',
    'random state': 42,
    'verbosity': 0,
    'n jobs': -1
})
final model = xgb.XGBClassifier(**best params final)
final model.fit(X, y enc, verbose=False)
end final train = time.time()
print(f" Final model training complete in {(end final train -
start final train):.2f} seconds.")
print("Preparing and preprocessing test data...")
df test = pd.read csv(file path base + 'test.csv')
```

```
X test = df test.drop('id', axis=1)
original train cols = df train.drop(['id','WeightCategory'],
axis=1).columns
for col in categorical cols:
    if col not in X test.columns:
        X test[col] = 'Missing'
    train categories = df train[col].astype('category').cat.categories
    X test[col] = pd.Categorical(X test[col],
categories=train categories)
X test = pd.get dummies(X test, columns=categorical cols,
drop first=True)
X test = X test.reindex(columns=X.columns, fill value=0)
numerical cols present test = [col for col in numerical cols if col in
X test.columns]
cols to scale test = [col for col in numerical cols present train if
col in numerical cols present test]
X test[cols to scale test] =
scaler.transform(X test[cols to scale test])
print("[] Test data preprocessing complete.")
print("Generating test predictions...")
y test pred enc = final model.predict(X test)
y test pred = le.inverse transform(y test pred enc)
print("[] Predictions decoded.")
submission = pd.DataFrame({
    'id': df test['id'],
    'WeightCategory': y test pred
})
out path = file path base + 'weight xg boosting optuna final.csv'
submission.to csv(out path, index=False)
print(f"\n□ Submission saved to: {out path}")
print(submission.head())
```

From the best parameters I found above, I got a jump in accuracy, but still i was stuck after trying multiple tunings, then I manually tuned the values of new hyperparameter gamma from 0.1 to 1.0, and I found gamma=0.5 giving the most accurate results. This value of gamma helped me reach **0.91239** as the final accuracy.

below code tuned value of gamma based on centering at the values of parameters which resulted to the last most accurate model.

```
import pandas as pd
import numpy as np
import time
import warnings
from sklearn.preprocessing import StandardScaler, LabelEncoder
```

```
from sklearn.model selection import RepeatedStratifiedKFold
import xgboost as xgb
import optuna
warnings.filterwarnings("ignore")
file path = '/content/drive/MyDrive/Obesity Data/'
df = pd.read csv(file path + 'train.csv')
df test = pd.read csv(file path + 'test.csv')
X = df.drop(['id','WeightCategory'], axis=1)
y = df['WeightCategory']
categorical cols = X.select dtypes(include=['object']).columns
X = pd.get dummies(X, columns=categorical cols, drop first=True)
numerical cols =
['Age','Height','Weight','FCVC','NCP','CH20','FAF','TUE']
scaler = StandardScaler()
X[numerical cols] = scaler.fit transform(X[numerical cols])
le = LabelEncoder()
y enc = le.fit transform(y)
center params = {
    'n estimators': 1244,
    'learning rate': 0.012037573993022807,
    'max depth': 11,
    'subsample': 0.6732768958671234,
    'colsample bytree': 0.530161020060218,
    'min child weight': 3,
    'reg alpha': 0.13850139721638935,
    'reg lambda': 13.95559742202856
}
PRUNING THRESHOLD = 0.90473
def objective(trial):
    params = {
        'n estimators': trial.suggest int('n estimators',
center params['n estimators']-200,
center params['n estimators']+200),
        'learning rate': trial.suggest float('learning rate',
center params['learning rate']/5,
center params['learning rate']*5,
                                              log=True),
        'max_depth': trial.suggest_int('max_depth',
```

```
max(3,
center params['max depth']-3),
                                        center params['max depth']+3),
        'subsample': trial.suggest float('subsample',
                                          \max(0.5)
center params['subsample']-0.2),
                                          min(1.0,
center params['subsample']+0.2)),
        'colsample bytree': trial.suggest float('colsample bytree',
                                                 max(0.4,
center params['colsample bytree']-0.2),
                                                 min(1.0,
center params['colsample bytree']+0.2)),
        'min child weight': trial.suggest int('min child weight',
                                               max(1,
center_params['min child weight']-3),
center params['min child weight']+3),
        'reg alpha': trial.suggest float('reg alpha',
                                          center params['reg alpha']/5,
                                          center params['reg alpha']*5,
                                          log=True),
        'reg_lambda': trial.suggest_float('reg lambda',
center params['reg lambda']/5,
center params['reg lambda']*5,
                                           log=True).
        'gamma': trial.suggest float('gamma', 0.3, 0.7),
        'objective': 'multi:softmax',
        'num_class': len(le.classes ),
        'use label encoder': False,
        'eval metric': 'mlogloss',
        'random state': 42,
        'verbosity': 0,
        'n jobs': -1,
        'early stopping rounds': 50
    }
    cv = RepeatedStratifiedKFold(n splits=5, n repeats=3,
random state=42)
    scores = []
    fold count = 0
    for train idx, val idx in cv.split(X, y enc):
        X_tr, X_val = X.iloc[train_idx], X.iloc[val_idx]
        y tr, y val = y enc[train idx], y enc[val idx]
        model = xgb.XGBClassifier(**params)
        model.fit(X tr, y tr, eval set=[(X val, y val)],
verbose=False)
```

```
score = model.score(X val, y val)
        scores.append(score)
        fold count += 1
        intermediate mean = np.mean(scores)
        trial.report(intermediate mean, step=fold count)
        if trial.should prune():
            print(f"□ Trial pruned at fold {fold count}/15 with avg
accuracy: {intermediate mean:.5f}")
            raise optuna.TrialPruned()
    return np.mean(scores)
print(f"Starting Optuna optimization with pruning threshold:
{PRUNING THRESHOLD}")
print("Trials with avg accuracy < 0.90473 will be stopped early\n")</pre>
pruner = optuna.pruners.MedianPruner(n startup trials=10,
n warmup steps=5)
study = optuna.create study(
    direction='maximize',
    sampler=optuna.samplers.TPESampler(seed=42),
    pruner=pruner
)
start = time.time()
study.optimize(objective, n trials=100, show progress bar=True)
end = time.time()
print(f"\n□ Done in {(end-start)/60:.2f} min")
print(f"Best CV accuracy: {study.best value:.5f}")
print(f"Number of completed trials: {len([t for t in study.trials if
t.state == optuna.trial.TrialState.COMPLETE])}")
print(f"Number of pruned trials: {len([t for t in study.trials if
t.state == optuna.trial.TrialState.PRUNED])}")
print("\nBest params:", study.best params)
best = study.best params
best.update({
    'objective': 'multi:softmax',
    'num class':<mark>len</mark>(le.classes ),
    'use label encoder': False,
    'eval metric': 'mlogloss',
    'random state':42,
    'verbosity':0,
    'n jobs':-1
})
final model = xgb.XGBClassifier(**best)
final_model.fit(X, y_enc, verbose=False)
X test = df test.drop(['id'], axis=1)
```

```
X test = pd.get dummies(X test, columns=categorical_cols,
drop first=True)
X test = X test.reindex(columns=X.columns, fill value=0)
X test[numerical cols] = scaler.transform(X test[numerical cols])
print("\nGenerating test predictions...")
y_test_pred_enc = final model.predict(X test)
y test pred = le.inverse transform(y test pred enc)
print("[] Predictions decoded.")
submission = pd.DataFrame({
    'id': df test['id'],
    'WeightCategory': y_test_pred
})
out_path = file_path + 'xg_gamma optuna pruned.csv'
submission.to csv(out path, index=False)
print(f"□ Submission saved to: {out path}")
print(submission.head())
pip install optuna
Collecting optuna
  Downloading optuna-4.5.0-py3-none-any.whl.metadata (17 kB)
Requirement already satisfied: alembic>=1.5.0 in
/usr/local/lib/python3.12/dist-packages (from optuna) (1.17.0)
Collecting colorlog (from optuna)
  Downloading colorlog-6.10.1-py3-none-any.whl.metadata (11 kB)
Requirement already satisfied: numpy in
/usr/local/lib/python3.12/dist-packages (from optuna) (2.0.2)
Requirement already satisfied: packaging>=20.0 in
/usr/local/lib/python3.12/dist-packages (from optuna) (25.0)
Requirement already satisfied: sqlalchemy>=1.4.2 in
/usr/local/lib/python3.12/dist-packages (from optuna) (2.0.44)
Requirement already satisfied: tqdm in /usr/local/lib/python3.12/dist-
packages (from optuna) (4.67.1)
Requirement already satisfied: PyYAML in
/usr/local/lib/python3.12/dist-packages (from optuna) (6.0.3)
Requirement already satisfied: Mako in /usr/local/lib/python3.12/dist-
packages (from alembic>=1.5.0->optuna) (1.3.10)
Requirement already satisfied: typing-extensions>=4.12 in
/usr/local/lib/python3.12/dist-packages (from alembic>=1.5.0->optuna)
(4.15.0)
Requirement already satisfied: greenlet>=1 in
/usr/local/lib/python3.12/dist-packages (from sqlalchemy>=1.4.2-
>optuna) (3.2.4)
Requirement already satisfied: MarkupSafe>=0.9.2 in
/usr/local/lib/python3.12/dist-packages (from Mako->alembic>=1.5.0-
>optuna) (3.0.3)
Downloading optuna-4.5.0-py3-none-any.whl (400 kB)
```

```
- 400.9/400.9 kB 12.1 MB/s eta
0:00:00
import pandas as pd
import numpy as np
import time
import warnings
from sklearn.preprocessing import StandardScaler, LabelEncoder
from sklearn.model selection import RepeatedStratifiedKFold
import xgboost as xgb
import optuna
optuna.logging.set_verbosity(optuna.logging.INFO)
warnings.filterwarnings("ignore")
print("Loading Training and Test data...")
file path base = '/content/drive/MyDrive/Obesity Data/'
try:
    df train = pd.read csv(file path base + 'train.csv')
    df test = pd.read csv(file path base + 'test.csv')
    print("Data loaded successfully.")
except FileNotFoundError:
    print(f"☐ ERROR: Could not find files in {file path base}. Please
check the file path.")
    raise
X = df train.drop(['id', 'WeightCategory'], axis=1)
y = df train['WeightCategory']
print("Starting Data Preprocessing...")
categorical cols = X.select dtypes(include=['object']).columns
X = pd.get dummies(X, columns=categorical cols, drop first=True)
num features =
['Age','Height','Weight','FCVC','NCP','CH20','FAF','TUE']
num features present train = [col for col in num features if col in
X.columns1
scaler = StandardScaler()
X[num_features_present_train] =
scaler.fit_transform(X[num features present train])
le = LabelEncoder()
y enc = le.fit transform(y)
print("Preprocessing and Encoding complete.")
params 1 = \{
    'n estimators': 1244, 'learning rate': 0.012037573993022807,
'max depth': 11,
    'subsample': 0.6732768958671234, 'colsample bytree':
0.530161020060218.
    'min child weight': 3, 'reg alpha': 0.13850139721638935,
    'reg lambda': 13.95559742202856, 'gamma': 0.5
}
```

```
params 2 = {
    'n estimators': 1403, 'learning rate': 0.011519841358664365,
'max depth': 13,
    'subsample': 0.8389925410061727, 'colsample bytree':
0.4289205756686294,
    'min_child_weight': 1, 'reg alpha': 0.21720626221992917.
    'reg lambda': 3.274796369552523, 'gamma': 0.38490059483369904
}
print("\nAdjusting parameters for n estimators range [500-900]...\n")
def objective(trial):
    params = {
        'n estimators': trial.suggest int('n estimators', 500, 900),
        'max depth': trial.suggest int('max depth', 6, 10),
        'learning rate': trial.suggest float('learning rate', 0.015,
0.030, log=True),
        'reg alpha': trial.suggest float('reg alpha', 0.3, 1.0,
log=True),
        'reg lambda': trial.suggest float('reg lambda', <mark>20.0, 50.0</mark>,
log=True),
         gamma': trial.suggest float('gamma', 0.5, 1.0),
        'min child weight': trial.suggest int('min child weight', 3,
8),
        'subsample': trial.suggest float('subsample', 0.70, 0.85),
        'colsample bytree': trial.suggest float('colsample bytree',
0.50, 0.70),
        'objective': 'multi:softmax',
        'num class': len(le.classes ),
        'use label encoder': False,
        'eval metric': 'mlogloss',
        'random state': 42,
        'verbosity': 0,
        'n jobs': -1,
        'early stopping rounds': 30,
        'tree method': 'hist'
    }
    cv = RepeatedStratifiedKFold(n_splits=5, n_repeats=2,
random state=42)
    scores = []
    for train_idx, val_idx in cv.split(X, y_enc):
        X tr, X val = X.iloc[train idx], X.iloc[val idx]
        y tr, y val = y enc[train idx], y enc[val idx]
        model = xgb.XGBClassifier(**params)
        model.fit(X tr, y tr, eval set=[(X val, y val)],
verbose=False)
        score = model.score(X val, y val)
        scores.append(score)
    return np.mean(scores)
```

```
print("Starting OPTUNA tuning for LOW-COMPLEXITY models (500-900)
estimators)...\n")
study = optuna.create study(direction='maximize',
sampler=optuna.samplers.TPESampler(seed=42))
study.optimize(objective, n trials=50, show progress bar=True)
print(f"Best CV accuracy: {study.best_value:.5f}")
print("\nBest hyperparameters found:")
best_params_from_optuna = study.best_params.copy()
for k, v in best_params_from_optuna.items():
    print(f" {k}: {v}")
print("\nTraining final model with best parameters...")
start final train = time.time()
best params final = best params from optuna.copy()
best params final.update({
    'objective': 'multi:softmax',
    'num class': len(le.classes ),
    'use_label_encoder': False,
    'eval metric': 'mlogloss',
    'random state': 42,
    'verbosity': 0,
    'n jobs': -1,
    'tree method': 'hist'
})
final model = xgb.XGBClassifier(**best params final)
final model.fit(X, y enc, verbose=False)
end final train = time.time()
print("\nPreparing and preprocessing test data...")
X_test = df_test.drop('id', axis=1)
test ids = df test['id']
original train df = df train.drop(['id','WeightCategory'], axis=1)
for col in categorical cols:
    if col not in X test.columns:
        X test[col] = 'Missing'
    train categories =
original train df[col].astype('category').cat.categories
    X test[col] = pd.Categorical(X_test[col],
categories=train categories)
X test = pd.get dummies(X test, columns=categorical cols,
drop first=True)
X test = X test.reindex(columns=X.columns, fill value=0)
num features present test = [col for col in num features if col in
X test.columns]
cols to scale test = [col for col in num features present train if col
in num features present test]
X test[cols to scale test] =
scaler.transform(X test[cols to scale test])
print("Test data preprocessing complete.")
```

```
print("Generating test predictions...")
y test pred enc = final model.predict(X test)
y test pred = le.inverse_transform(y_test_pred_enc)
submission = pd.DataFrame({'id': test ids, 'WeightCategory':
v test pred})
out path = file path base + 'weight xgb 500 900 estimators.csv'
submission.to_csv(out path, index=False)
print(f"\nSubmission saved to: {out path}")
print("Sample Submission:")
print(submission.head())
Loading Training and Test data...
[I 2025-10-25 03:40:34,671] A new study created in memory with name:
no-name-d558dbc8-4ed5-436e-b63b-0905fb77c8a8
Data loaded successfully.
Starting Data Preprocessing...
Preprocessing and Encoding complete.
Adjusting parameters for n estimators range [500-900]...
Starting OPTUNA tuning for LOW-COMPLEXITY models (500-900
estimators)...
{"model id": "9a91a5d6423e45be8089a7c106d7abaa", "version major": 2, "vers
ion minor":0}
[I 2025-10-25 03:42:57,631] Trial 0 finished with value:
0.9060707278560699 and parameters: {'n estimators': 650, 'max depth':
10, 'learning rate': 0.024913996103247533, 'reg alpha':
0.6168038158199004, 'reg lambda': 23.073636940063274, 'gamma':
0.5779972601681014, 'min child weight': 3, 'subsample':
0.8299264218662402, 'colsample bytree': 0.6202230023486417}. Best is
trial 0 with value: 0.9060707278560699.
[I 2025-10-25 03:45:03,596] Trial 1 finished with value:
0.9059096869313026 and parameters: {'n estimators': 783, 'max depth':
6, 'learning rate': 0.0293807729915014, 'reg alpha':
0.8173118924709636, 'reg lambda': 24.295633253176874, 'gamma':
0.5909124836035503, 'min_child_weight': 4, 'subsample': 0.7456363364439306, 'colsample_bytree': 0.6049512863264476}. Best is
trial 0 with value: 0.9060707278560699.
[I 2025-10-25 03:46:58,855] Trial 2 finished with value:
0.9059419448554259 and parameters: {'n estimators': 673, 'max depth':
7, 'learning rate': 0.022923310303024545, 'reg alpha':
0.3548621092076139, 'reg lambda': 26.13882370496371, 'gamma':
0.6831809216468459, 'min child weight': 5, 'subsample':
0.817776394208952, 'colsample bytree': 0.5399347564316719}. Best is
trial 0 with value: 0.9060707278560699.
```

```
[I 2025-10-25 03:49:11,141] Trial 3 finished with value:
0.9046221677946751 and parameters: {'n estimators': 706, 'max depth':
8, 'learning rate': 0.015490813545105597, 'reg alpha':
0.6234383941568222, 'reg lambda': 23.38236130297505, 'gamma':
0.5325257964926398, 'min child weight': 8, 'subsample':
0.8448448049611839, 'colsample bytree': 0.6616794696232922}. Best is
trial 0 with value: 0.9060707278560699.
[I 2025-10-25 03:50:55,727] Trial 4 finished with value:
0.905362566425107 and parameters: {'n estimators': 622, 'max depth':
6, 'learning rate': 0.02410271269872124, 'reg alpha':
0.509644848106954, 'reg_lambda': 22.366286923412634, 'gamma':
0.7475884550556351, 'min child weight': 3, 'subsample':
0.8363980603118173, 'colsample_bytree': 0.5517559963200034}. Best is
trial 0 with value: 0.9060707278560699.
[I 2025-10-25 03:52:56,789] Trial 5 finished with value:
0.9046865592949969 and parameters: {'n estimators': 765, 'max depth':
7, 'learning rate': 0.021510342880766087, 'reg alpha':
0.57940793638336, 'reg_lambda': 23.691413854920985, 'gamma':
0.9847923138822793, 'min_child_weight': 7, 'subsample':
0.8409248412346284, 'colsample_bytree': 0.6789654700855298}. Best is
trial 0 with value: 0.9060707278560699.
[I 2025-10-25 03:55:25,044] Trial 6 finished with value:
0.905233783424463 and parameters: {'n estimators': 739, 'max depth':
10, 'learning rate': 0.015948878757113025, 'reg alpha':
0.37983635344012184, 'reg lambda': 20.846240478620484, 'gamma':
0.6626651653816322, 'min child weight': 5, 'subsample':
0.7407023547660844, 'colsample bytree': 0.6657475018303858}. Best is
trial 0 with value: 0.9060707278560699.
[I 2025-10-25 03:57:21,735] Trial 7 finished with value:
0.9047510129692814 and parameters: {'n_estimators': 643, 'max_depth':
7, 'learning rate': 0.0218503832414944, 'reg_alpha':
0.35547375076141013, 'reg lambda': 41.711544297437776, 'gamma':
0.5372753218398854, 'min_child_weight': 8, 'subsample':
0.8158367153944985, 'colsample bytree': 0.5397431363068345}. Best is
trial 0 with value: 0.9060707278560699.
[I 2025-10-25 03:58:54,191] Trial 8 finished with value:
0.9038496873996797 and parameters: {'n estimators': 502, 'max depth':
10, 'learning rate': 0.024483670277647855, 'reg alpha':
0.7216117696718226, 'reg lambda': 40.54612212253375, 'gamma':
0.5370223258670452, 'min child weight': 5, 'subsample':
0.7173803589287694, 'colsample bytree': 0.6726206851751186}. Best is
trial 0 with value: 0.9060707278560699.
[I 2025-10-25 04:01:03,554] Trial 9 finished with value:
0.9040106557881575 and parameters: {'n estimators': 749, 'max depth':
7, 'learning rate': 0.01567560202380831, 'reg alpha':
0.4362418243639405, 'reg_lambda': 26.94222443886185, 'qamma':
0.864803089169032, 'min child weight': 6, 'subsample':
0.8330819113864489, 'colsample bytree': 0.5944429850323898}. Best is
trial 0 with value: 0.9060707278560699.
```

```
[I 2025-10-25 04:03:27,231] Trial 10 finished with value:
0.9045257152544439 and parameters: {'n estimators': 878, 'max depth':
9, 'learning rate': 0.01856591877886313, 'reg alpha':
0.9217137344038946, 'reg lambda': 32.247202476965995, 'gamma':
0.8451235367845725, 'min child weight': 3, 'subsample':
0.7935064777944197, 'colsample_bytree': 0.6141095729763366}. Best is
trial 0 with value: 0.9060707278560699.
[I 2025-10-25 04:05:10,772] Trial 11 finished with value:
0.9066500855617345 and parameters: {'n estimators': 592, 'max depth':
9, 'learning rate': 0.027440622336573413, 'reg alpha':
0.3275711769380865, 'reg lambda': 29.77678013534258, 'gamma':
0.6712579968688783, 'min_child_weight': 4, 'subsample':
0.7910776411049838, 'colsample bytree': 0.5018093854636312}. Best is
trial 11 with value: 0.9066500855617345.
[I 2025-10-25 04:06:46,515] Trial 12 finished with value:
0.9065213854597068 and parameters: {'n estimators': 561, 'max depth':
9, 'learning rate': 0.028223154051390028, 'reg alpha':
0.304637703607719, 'reg_lambda': 32.150717536003434, 'gamma':
0.6372424866412404, 'min child weight': 4, 'subsample':
0.7861135797155597, 'colsample_bytree': 0.5011678503134699}. Best is
trial 11 with value: 0.9066500855617345.
[I 2025-10-25 04:08:22,135] Trial 13 finished with value:
0.9066179934348441 and parameters: {'n estimators': 548, 'max depth':
9, 'learning rate': 0.029829108609252496, 'reg alpha':
0.3006796839487717, 'reg lambda': 32.7748444556718, 'gamma':
0.662195899262473, 'min_child_weight': 4, 'subsample':
0.777957328049646, 'colsample_bytree': 0.5036892539507432}. Best is
trial 11 with value: 0.9066500855617345.
[I 2025-10-25 04:10:00,128] Trial 14 finished with value:
0.9061673254688799 and parameters: {'n_estimators': 569, 'max_depth':
9, 'learning rate': 0.027114940133267828, 'reg alpha':
0.3005959535771838, 'reg_lambda': 35.658552095556566, 'gamma':
0.7633599273058994, 'min_child_weight': 4, 'subsample':
0.7599971049075572, 'colsample_bytree': 0.5051221654613188}. Best is
trial 11 with value: 0.9066500855617345.
[I 2025-10-25 04:11:26,158] Trial 15 finished with value:
0.9051372272609612 and parameters: {'n estimators': 507, 'max depth':
8, 'learning_rate': 0.026500687756892675, 'reg alpha':
0.470615523232326, 'reg_lambda': 28.920215600113224, 'gamma':
0.7673463019630581, 'min child weight': 6, 'subsample':
0.79543017666551, 'colsample bytree': 0.568441560666624}. Best is
trial 11 with value: 0.9066500855617345.
[I 2025-10-25 04:13:03,100] Trial 16 finished with value:
0.9060063674427289 and parameters: {'n estimators': 568, 'max depth':
9, 'learning rate': 0.029827919188324527, 'reg alpha':
0.4163830612469085, 'reg_lambda': 36.63606874564723, 'gamma':
0.7105925797534106, 'min child weight': 4, 'subsample':
0.7666753501990495, 'colsample_bytree': 0.5187413823660614}. Best is
trial 11 with value: 0.9066500855617345.
```

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[I 2025-10-25 04:14:52,324] Trial 17 finished with value:
0.9047509300706649 and parameters: {'n estimators': 603, 'max depth':
8, 'learning rate': 0.02021483718177374, 'reg alpha':
0.34080030445903825, 'reg_lambda': 29.754758607483506, 'gamma':
0.8255044162944254, 'min_child_weight': 4, 'subsample': 0.8079689779534471, 'colsample_bytree': 0.578019486309862}. Best is
trial 11 with value: 0.9066500855617345.
[I 2025-10-25 04:16:25,378] Trial 18 finished with value:
0.9053303603126188 and parameters: {'n estimators': 528, 'max depth':
9, 'learning rate': 0.026555341635958976, 'reg alpha':
0.3001417417639796, 'reg lambda': 36.02969013261386, 'gamma':
0.6092379396242038, 'min_child_weight': 5, 'subsample':
0.7783290377283562, 'colsample bytree': 0.5235464057837369}. Best is
trial 11 with value: 0.9066500855617345.
[I 2025-10-25 04:18:40,124] Trial 19 finished with value:
0.9045579317292589 and parameters: {'n estimators': 819, 'max depth':
10, 'learning rate': 0.019805879107149135, 'reg alpha':
0.4035755304685789, 'reg_lambda': 46.31008664801553, 'gamma':
0.9340183831703528, 'min_child_weight': 3, 'subsample':
0.7246093428703273, 'colsample_bytree': 0.5590857595985498}. Best is
trial 11 with value: 0.9066500855617345.
[I 2025-10-25 04:20:22,596] Trial 20 finished with value:
0.9041716138143083 and parameters: {'n estimators': 599, 'max depth':
9, 'learning rate': 0.02796767660274991, 'reg alpha':
0.46747188542625184, 'reg_lambda': 49.56156337330534, 'gamma':
0.7058037716383934, 'min_child_weight': 6, 'subsample':
0.701743144244133, 'colsample_bytree': 0.6433438152801837}. Best is
trial 11 with value: 0.9066500855617345.
[I 2025-10-25 04:21:59,988] Trial 21 finished with value:
0.9066179416232087 and parameters: {'n_estimators': 555, 'max_depth':
9, 'learning rate': 0.028344762395586504, 'reg_alpha':
0.32576271413337793, 'reg_lambda': 32.354376458939626, 'gamma':
0.6399823782044075, 'min_child_weight': 4, 'subsample':
0.7853835217331906, 'colsample_bytree': 0.501789023714044}. Best is
trial 11 with value: 0.9066500855617345.
[I 2025-10-25 04:23:32,617] Trial 22 finished with value:
0.9062960255709072 and parameters: {'n estimators': 549, 'max depth':
8, 'learning rate': 0.029984478466628342, 'reg alpha':
0.33450777265611076, 'reg_lambda': 29.90648496936411, 'gamma':
0.6426166044326023, 'min_child_weight': 4, 'subsample':
0.766181387800847, 'colsample_bytree': 0.5251108781921068}. Best is
trial 11 with value: 0.9066500855617345.
[I 2025-10-25 04:25:17,312] Trial 23 finished with value:
0.905941986304734 and parameters: {'n estimators': 586, 'max depth':
9, 'learning rate': 0.02562828796735913, 'reg alpha':
0.32579068805141603, 'reg lambda': 34.395055241768944, 'qamma':
0.7302990497731223, 'min_child_weight': 5, 'subsample':
0.8040985494671934, 'colsample_bytree': 0.5047135146455393}. Best is
trial 11 with value: 0.9066500855617345.
```

```
[I 2025-10-25 04:26:53,622] Trial 24 finished with value:
0.905781007553929 and parameters: {'n estimators': 535, 'max depth':
10, 'learning rate': 0.02799114770141638, 'reg alpha':
0.37623198362228655, 'reg lambda': 39.1889013219618, 'gamma':
0.6213138892355491, 'min child weight': 3, 'subsample':
0.7801308535166759, 'colsample_bytree': 0.5380295744402516}. Best is
trial 11 with value: 0.9066500855617345.
[I 2025-10-25 04:28:48,011] Trial 25 finished with value:
0.9064570250463662 and parameters: {'n estimators': 696, 'max depth':
9, 'learning rate': 0.023435455901029266, 'reg alpha':
0.38525152467014057, 'reg_lambda': 27.942587044804228, 'gamma':
0.8049579237960481, 'min_child_weight': 4, 'subsample':
0.7593090542509797, 'colsample bytree': 0.5224984696129451}. Best is
trial 11 with value: 0.9066500855617345.
[I 2025-10-25 04:30:32,163] Trial 26 finished with value:
0.9064247774845698 and parameters: {'n estimators': 621, 'max depth':
8, 'learning rate': 0.025629300257736225, 'reg alpha':
0.32857643256819624, 'reg_lambda': 33.177766052851815, 'gamma':
0.6719485339323757, 'min child weight': 4, 'subsample':
0.7927835191703553, 'colsample_bytree': 0.5129193711050242}. Best is
trial 11 with value: 0.9066500855617345.
[I 2025-10-25 04:32:04,778] Trial 27 finished with value:
0.9060706864067616 and parameters: {'n estimators': 532, 'max depth':
8, 'learning rate': 0.028282664171211475, 'reg alpha':
0.4523139815379569, 'reg_lambda': 30.791398572769133, 'gamma':
0.5709986796051028, 'min child weight': 5, 'subsample':
0.7471794953251831, 'colsample_bytree': 0.5362782684383005}. Best is
trial 11 with value: 0.9066500855617345.
[I 2025-10-25 04:34:02,127] Trial 28 finished with value:
0.9048797856075981 and parameters: {'n estimators': 644, 'max depth':
9, 'learning rate': 0.01733118447823402, 'reg_alpha':
0.5406143965837826, 'reg_lambda': 25.550887501961533, 'gamma':
0.6891495823365336, 'min child weight': 7, 'subsample':
0.7705496497494498, 'colsample_bytree': 0.5764865906043057}. Best is
trial 11 with value: 0.9066500855617345.
[I 2025-10-25 04:36:03,095] Trial 29 finished with value:
0.9064247463975889 and parameters: {'n estimators': 671, 'max depth':
10, 'learning rate': 0.026193010379065323, 'reg alpha':
0.6396280946375229, 'reg_lambda': 27.88211988409453, 'gamma':
0.5018302718679613, 'min child weight': 3, 'subsample':
0.8178208430505992, 'colsample bytree': 0.5007792255157338}. Best is
trial 11 with value: 0.9066500855617345.
[I 2025-10-25 04:37:49,427] Trial 30 finished with value:
0.9052338974100607 and parameters: {'n estimators': 585, 'max depth':
9, 'learning rate': 0.024982249697679013, 'reg alpha':
0.3641923522774178, 'reg_lambda': 38.02043999991283, 'gamma':
0.5786021866960233, 'min_child_weight': 3, 'subsample':
0.8039823170019557, 'colsample_bytree': 0.6325683450492714}. Best is
trial 11 with value: 0.9066500855617345.
```

```
[I 2025-10-25 04:39:25,852] Trial 31 finished with value:
0.9066501580980239 and parameters: {'n estimators': 556, 'max depth':
9, 'learning rate': 0.02871848916417165, 'reg alpha':
0.3105780396405383, 'reg lambda': 32.29259545356618, 'gamma':
0.6507389294704313, 'min child weight': 4, 'subsample':
0.78273958979123, 'colsample bytree': 0.5002456370845217}. Best is
trial 31 with value: 0.9066501580980239.
[I 2025-10-25 04:40:55,286] Trial 32 finished with value:
0.9060063570804019 and parameters: {'n estimators': 522, 'max depth':
9, 'learning rate': 0.0290260631087539, 'reg alpha':
0.3230258190071752, 'reg lambda': 34.12930949956639, 'gamma':
0.6523569188077599, 'min_child_weight': 4, 'subsample':
0.7859700243745744, 'colsample bytree': 0.5119089951897215}. Best is
trial 31 with value: 0.9066501580980239.
[I 2025-10-25 04:42:30,467] Trial 33 finished with value:
0.9061673047442256 and parameters: {'n estimators': 553, 'max depth':
10, 'learning rate': 0.027390345686138667, 'reg alpha':
0.31478801443728177, 'reg_lambda': 31.01941923387742, 'gamma':
0.6134962941784908, 'min child weight': 5, 'subsample':
0.7765407568667437, 'colsample_bytree': 0.5302297332302667}. Best is
trial 31 with value: 0.9066501580980239.
[I 2025-10-25 04:44:17,463] Trial 34 finished with value:
0.9061994486827514 and parameters: {'n estimators': 618, 'max depth':
8, 'learning rate': 0.02908845504887923, 'reg alpha':
0.3493914520118413, 'reg_lambda': 33.364302843841145, 'gamma':
0.5938392953949548, 'min child weight': 4, 'subsample':
0.7536882250529917, 'colsample_bytree': 0.5456333378116934}. Best is
trial 31 with value: 0.9066501580980239.
[I 2025-10-25 04:46:04,082] Trial 35 finished with value:
0.9066823227612038 and parameters: {'n_estimators': 574, 'max_depth':
10, 'learning rate': 0.023079408485393083, 'reg_alpha':
0.39830031179361597, 'reg_lambda': 24.83362446335252, 'gamma':
0.7013920740060814, 'min_child_weight': 3, 'subsample':
0.8227129483717687, 'colsample bytree': 0.5140386624198232}. Best is
trial 35 with value: 0.9066823227612038.
[I 2025-10-25 04:48:01,582] Trial 36 finished with value:
0.9051372376232883 and parameters: {'n estimators': 660, 'max depth':
10, 'learning rate': 0.023242585525807834, 'reg alpha':
0.40584862965364754, 'reg_lambda': 25.315180886974098, 'gamma':
0.7898662926804407, 'min child weight': 3, 'subsample':
0.8243189690703979, 'colsample bytree': 0.6997281865118572}. Best is
trial 35 with value: 0.9066823227612038.
[I 2025-10-25 04:49:50,013] Trial 37 finished with value:
0.9061672840195716 and parameters: {'n_estimators': 584, 'max_depth':
10, 'learning rate': 0.022460496129851407, 'reg alpha':
0.35599786554293117, 'reg_lambda': 22.17605513573723, 'gamma':
0.731608639211617, 'min child weight': 3, 'subsample':
0.8472781094719573, 'colsample bytree': 0.5534278297709496}. Best is
trial 35 with value: 0.9066823227612038.
```

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[I 2025-10-25 04:51:57,933] Trial 38 finished with value:
0.9066501270110428 and parameters: {'n estimators': 708, 'max depth':
10, 'learning rate': 0.02068802189254348, 'reg alpha':
0.49807075090593816, 'reg lambda': 24.712129062867263, 'gamma':
0.6805529026435361, 'min child weight': 3, 'subsample':
0.7999189840729743, 'colsample_bytree': 0.5148192299565391}. Best is
trial 35 with value: 0.9066823227612038.
[I 2025-10-25 04:54:03,258] Trial 39 finished with value:
0.9064891482602379 and parameters: {'n estimators': 713, 'max depth':
10, 'learning rate': 0.02065096432279291, 'reg alpha':
0.5479283377060836, 'reg lambda': 20.56747490891688, 'gamma':
0.6983187756033097, 'min child weight': 3, 'subsample':
0.8252333909691628, 'colsample bytree': 0.516214551918953}. Best is
trial 35 with value: 0.9066823227612038.
[I 2025-10-25 04:56:22,442] Trial 40 finished with value:
0.9064891586225649 and parameters: {'n estimators': 808, 'max depth':
10, 'learning rate': 0.01898913135322602, 'reg alpha':
0.7146430849580342, 'reg_lambda': 22.28602024853672, 'gamma':
0.7195729319309041, 'min child weight': 3, 'subsample':
0.8123150232933687, 'colsample_bytree': 0.5309884797318123}. Best is
trial 35 with value: 0.9066823227612038.
[I 2025-10-25 04:58:25,674] Trial 41 finished with value:
0.9065213025610905 and parameters: {'n estimators': 691, 'max depth':
10, 'learning rate': 0.02094567063468085, 'reg alpha':
0.5045721553705783, 'reg_lambda': 24.53402599578761, 'gamma':
0.6736369859562744, 'min child weight': 3, 'subsample':
0.7984861495838593, 'colsample_bytree': 0.5127309142343901}. Best is
trial 35 with value: 0.9066823227612038.
[I 2025-10-25 05:00:18,122] Trial 42 finished with value:
0.9064892415211814 and parameters: {'n estimators': 726, 'max depth':
6, 'learning rate': 0.02418900073327826, 'reg alpha':
0.38770354321604267, 'reg_lambda': 26.818621240981887, 'gamma':
0.6816506741743824, 'min_child_weight': 4, 'subsample':
0.8017629350775075, 'colsample bytree': 0.5164538076121177}. Best is
trial 35 with value: 0.9066823227612038.
[I 2025-10-25 05:02:17,248] Trial 43 finished with value:
0.9053625353381258 and parameters: {'n estimators': 629, 'max depth':
10, 'learning rate': 0.01762162480906808, 'reg alpha':
0.4198125906267699, 'reg_lambda': 24.20253343519827, 'gamma':
0.7387914881539491, 'min child weight': 3, 'subsample':
0.7890869050595454, 'colsample bytree': 0.5444753746006015}. Best is
trial 35 with value: 0.9066823227612038.
[I 2025-10-25 05:03:53,476] Trial 44 finished with value:
0.9059419655800799 and parameters: {'n estimators': 505, 'max depth':
9, 'learning rate': 0.022299527525843233, 'reg alpha':
0.4967463487705302, 'reg_lambda': 21.404571761528494, 'gamma':
0.659811264748781, 'min child weight': 4, 'subsample':
0.8104375906434764, 'colsample bytree': 0.5279809589795919}. Best is
trial 35 with value: 0.9066823227612038.
```

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[I 2025-10-25 05:06:07,606] Trial 45 finished with value:
0.9063925713720818 and parameters: {'n estimators': 770, 'max depth':
10, 'learning rate': 0.01944883160254376, 'reg alpha':
0.3629884098074598, 'reg_lambda': 23.169694238689665, 'qamma':
0.762259579310665, 'min child weight': 5, 'subsample':
0.8367949936769459, 'colsample_bytree': 0.5104240692989952}. Best is
trial 35 with value: 0.9066823227612038.
[I 2025-10-25 05:07:56,214] Trial 46 finished with value:
0.906649971576137 and parameters: {'n estimators': 602, 'max depth':
9, 'learning rate': 0.02714971552688047, 'reg alpha':
0.31682991457033266, 'reg_lambda': 28.4350892741885, 'gamma':
0.5590449615355797, 'min_child_weight': 3, 'subsample':
0.8232372681355544, 'colsample bytree': 0.500625816545753}. Best is
trial 35 with value: 0.9066823227612038.
[I 2025-10-25 05:10:04,071] Trial 47 finished with value:
0.9051372168986342 and parameters: {'n estimators': 677, 'max depth':
9, 'learning rate': 0.02144372334585939, 'reg alpha':
0.9965000165234202, 'reg_lambda': 28.415644830071862, 'gamma':
0.5598818751630361, 'min child weight': 3, 'subsample':
0.8231119251298024, 'colsample_bytree': 0.5992772644250852}. Best is
trial 35 with value: 0.9066823227612038.
[I 2025-10-25 05:11:55,005] Trial 48 finished with value:
0.9066501788226781 and parameters: {'n estimators': 603, 'max depth':
8, 'learning rate': 0.0246965362624253, 'reg alpha':
0.43698062182953085, 'reg lambda': 25.043319881955032, 'gamma':
0.5495648717681827, 'min child weight': 3, 'subsample':
0.8401676513683018, 'colsample_bytree': 0.5629904695963326}. Best is
trial 35 with value: 0.9066823227612038.
[I 2025-10-25 05:13:48,129] Trial 49 finished with value:
0.9059097387429379 and parameters: {'n_estimators': 638, 'max_depth':
7, 'learning rate': 0.023999575389928585, 'reg alpha':
0.432578904051451, 'reg_lambda': 26.41592719570022, 'gamma':
0.6072159161207752, 'min_child_weight': 7, 'subsample':
0.8393160896631675, 'colsample_bytree': 0.5638179698634069}. Best is
trial 35 with value: 0.9066823227612038.
Best CV accuracy: 0.90668
Best hyperparameters found:
  n estimators: 574
  max depth: 10
  learning rate: 0.023079408485393083
  reg alpha: 0.39830031179361597
  reg lambda: 24.83362446335252
  gamma: 0.7013920740060814
  min child weight: 3
  subsample: 0.8227129483717687
  colsample bytree: 0.5140386624198232
Training final model with best parameters...
```

```
Preparing and preprocessing test data...
Test data preprocessing complete.
                                          Traceback (most recent call
NameError
last)
/tmp/ipython-input-3267767786.py in <cell line: 0>()
    155 X test[cols to scale test] =
scaler.transform(X test[cols to scale test])
    156 print("Test data preprocessing complete.")
--> 157 d
    158 # --- 8. PREDICT ON TEST SET ---
    159 print("Generating test predictions...")
NameError: name 'd' is not defined
import pandas as pd
import numpy as np
import time
import warnings
from sklearn.preprocessing import StandardScaler, LabelEncoder
from sklearn.model selection import RepeatedStratifiedKFold
import xgboost as xgb
import optuna
optuna.logging.set verbosity(optuna.logging.INFO)
warnings.filterwarnings("ignore")
print("Loading Training and Test data...")
file path base = '/content/drive/MyDrive/Obesity Data/'
try:
    df train = pd.read csv(file path base + 'train.csv')
    df test = pd.read csv(file path base + 'test.csv')
    print("Data loaded successfully.")
except FileNotFoundError:
    print(f"☐ ERROR: Could not find files in {file path base}. Please
check the file path.")
    raise
X = df train.drop(['id', 'WeightCategory'], axis=1)
y = df train['WeightCategory']
print("Starting Data Preprocessing...")
categorical_cols = X.select dtypes(include=['object']).columns
X = pd.get dummies(X, columns=categorical cols, drop first=True)
num features =
['Age','Height','Weight','FCVC','NCP','CH20','FAF','TUE']
num features present train = [col for col in num features if col in
```

```
X.columns1
scaler = StandardScaler()
X[num features present train] =
scaler.fit transform(X[num features present train])
le = LabelEncoder()
y enc = le.fit transform(y)
print("Preprocessing and Encoding complete.")
print("\nOptimizing around max depth=8 (your best performer:
0.90183)...\n")
def objective(trial):
    params = {
        'max depth': trial.suggest int('max depth', 7, 12),
        'n estimators': trial.suggest int('n estimators', <mark>550</mark>, 700),
        'learning rate': trial.suggest float('learning rate', 0.005,
0.035, log=True),
        'reg alpha': trial.suggest float('reg alpha', 0.15, 0.35,
log=True),
         reg lambda': trial.suggest float('reg lambda', 12.0, 22.0,
log=True),
         'gamma': trial.suggest_float('gamma', <mark>0.40, 0.65</mark>),
        'min child weight': trial.suggest int('min child weight', 1,
3),
        'subsample': trial.suggest_float('subsample', 0.80, 0.90),
        'colsample bytree': trial.suggest float('colsample bytree',
0.50, 0.65),
        'objective': 'multi:softmax',
        'num class': len(le.classes ),
        'use label encoder': False,
        'eval metric': 'mlogloss',
        'random state': 42,
        'verbosity': 0,
        'n jobs': -1,
        'early stopping rounds': 30,
        'tree method': 'hist'
    }
    cv = RepeatedStratifiedKFold(n splits=5, n repeats=2,
random state=42)
    scores = []
    for train_idx, val_idx in cv.split(X, y_enc):
        X tr, X val = X.iloc[train idx], X.iloc[val idx]
        y_tr, y_val = y_enc[train_idx], y_enc[val_idx]
        model = xgb.XGBClassifier(**params)
        model.fit(X tr, y tr, eval set=[(X val, y val)],
verbose=False)
        score = model.score(X val, y val)
        scores.append(score)
```

```
return np.mean(scores)
print("Starting OPTUNA search with max depth=8 as anchor...\n")
study = optuna.create study(direction='maximize',
sampler=optuna.samplers.TPESampler(seed=42))
study.optimize(objective, n trials=80, show progress bar=True)
print(f"\nBest CV accuracy: {study.best value:.5f}")
print("\nBest hyperparameters found:")
best params from optuna = study.best params.copy()
best params from optuna['max depth'] = 8
for k, v in best_params_from_optuna.items():
    print(f" {k}: {v}")
print("\nTraining final model with best parameters...")
best params final = best params from optuna.copy()
best_params_final.update({
    'objective': 'multi:softmax',
    'num class': len(le.classes ),
    'use label encoder': False,
    'eval metric': 'mlogloss',
    'random state': 42,
    'verbosity': 0,
    'n jobs': -1,
    'tree method': 'hist'
})
final model = xgb.XGBClassifier(**best params final)
final model.fit(X, y enc, verbose=False)
print("\nPreparing and preprocessing test data...")
X_test = df_test.drop('id', axis=1)
test ids = df test['id']
original train df = df train.drop(['id','WeightCategory'], axis=1)
for col in categorical cols:
    if col not in X test.columns:
        X_test[col] = 'Missing'
    train categories =
original train df[col].astype('category').cat.categories
    X test[col] = pd.Categorical(X test[col],
categories=train categories)
X test = pd.get dummies(X test, columns=categorical cols,
drop first=True)
X test = X test.reindex(columns=X.columns, fill value=0)
num features present test = [col for col in num features if col in
X test.columns]
cols to scale test = [col for col in num features present train if col
in num features present test]
X test[cols to scale test] =
```

```
scaler.transform(X test[cols to scale test])
print("Test data preprocessing complete.")
print("Generating test predictions...")
v test pred enc = final model.predict(X test)
y test pred = le.inverse transform(y test pred enc)
submission = pd.DataFrame({'id': test ids, 'WeightCategory':
y test pred})
out path = file path base + 'xgb maxdepth8 optimized.csv'
submission.to csv(out path, index=False)
print(f"\n Submission saved to: {out path}")
print("Sample Submission:")
print(submission.head())
[I 2025-10-25 05:47:28,062] A new study created in memory with name:
no-name-b1fd8f7f-f398-433a-9393-f1fb4b95eb1e
Loading Training and Test data...
Data loaded successfully.
Starting Data Preprocessing...
Preprocessing and Encoding complete.
Optimizing around max depth=8 (your best performer: 0.90183)...
Starting OPTUNA search with max depth=8 as anchor...
{"model id":"46e46fb6ae1742f48305ee78f5ca567d","version major":2,"vers
ion minor":0}
[I 2025-10-25 05:50:02,661] Trial 0 finished with value:
0.9077765534112677 and parameters: {'max depth': 9, 'n estimators':
693, 'learning_rate': 0.02077670387146178, 'reg_alpha':
0.24910571058105346, 'reg lambda': 13.190213245183196, 'gamma':
0.4389986300840507, 'min_child_weight': 1, 'subsample':
0.8866176145774936, 'colsample_bytree': 0.5901672517614813}. Best is
trial 0 with value: 0.9077765534112677.
[I 2025-10-25 05:51:53,947] Trial 1 finished with value:
0.9077444716467044 and parameters: {'max depth': 11, 'n estimators':
553, 'learning rate': 0.033009497973205286, 'reg alpha':
0.3036762487005602, 'reg_lambda': 13.648272915220799, 'gamma':
0.4454562418017752, 'min child weight': 1, 'subsample':
0.8304242242959539, 'colsample bytree': 0.5787134647448356}. Best is
trial 0 with value: 0.9077765534112677.
[I 2025-10-25 05:54:03,424] Trial 2 finished with value:
0.906617817275284 and parameters: {'max depth': 9, 'n estimators':
593, 'learning rate': 0.01644544715158596, 'reg alpha':
0.1688191702869054, 'reg_lambda': 14.324709470256131, 'gamma':
0.49159046082342295, 'min_child_weight': 2, 'subsample':
0.8785175961393014, 'colsample_bytree': 0.529951067323754}. Best is
```

```
trial 0 with value: 0.9077765534112677.
[I 2025-10-25 05:57:04,477] Trial 3 finished with value:
0.9010492063390085 and parameters: {'max depth': 10, 'n estimators':
639, 'learning rate': 0.005472996332062147, 'reg alpha':
0.2509884060932996, 'reg_lambda': 13.306696789696266, 'gamma': 0.4162628982463199, 'min_child_weight': 3, 'subsample':
0.896563203307456, 'colsample bytree': 0.6212596022174692}. Best is
trial 0 with value: 0.9077765534112677.
[I 2025-10-25 05:59:04,107] Trial 4 finished with value:
0.9066499094021747 and parameters: {'max depth': 8, 'n estimators':
564, 'learning_rate': 0.018932762804558884, 'reg_alpha':
0.21779964058391643, 'reg_lambda': 12.921316550182928, 'gamma':
0.5237942275278176, 'min_child_weight': 1, 'subsample':
0.8909320402078782, 'colsample_bytree': 0.5388169972400025}. Best is
trial 0 with value: 0.9077765534112677.
[I 2025-10-25 06:01:22,643] Trial 5 finished with value:
0.9050405985611703 and parameters: {'max depth': 10, 'n estimators':
597, 'learning rate': 0.013755566149109682, 'reg alpha':
0.23837897615555623, 'reg_lambda': 13.422783894473495, 'qamma':
0.6423961569411396, 'min_child_weight': 3, 'subsample':
0.8939498941564189, 'colsample_bytree': 0.6342241025641473}. Best is
trial 0 with value: 0.9077765534112677.
[I 2025-10-25 06:04:21,403] Trial 6 finished with value:
0.9024333645377542 and parameters: {'max depth': 10, 'n estimators':
689, 'learning_rate': 0.005939567803015175, 'reg alpha':
0.17709585755685345, 'reg lambda': 12.333517162037408, 'gamma':
0.48133258269081614, 'min_child_weight': 2, 'subsample':
0.8271349031773896, 'colsample bytree': 0.6243106263727894}. Best is
trial 0 with value: 0.9077765534112677.
[I 2025-10-25 06:06:34,382] Trial 7 finished with value:
0.9053303395879648 and parameters: {'max depth': 9, 'n estimators':
592, 'learning rate': 0.014374786415262011, 'reg alpha':
0.1690238938055383, 'reg_lambda': 19.514288900142493, 'gamma':
0.4186376609199427, 'min_child_weight': 3, 'subsample':
0.8772244769296658, 'colsample bytree': 0.5298073522301259}. Best is
trial 0 with value: 0.9077765534112677.
[I 2025-10-25 06:08:50,445] Trial 8 finished with value:
0.9066823020365498 and parameters: {'max depth': 7, 'n estimators':
673, 'learning_rate': 0.019784895474618876, 'reg_alpha':
0.2781947777606777, 'req lambda': 19.15188647361643, 'qamma':
0.41851116293352264, 'min child weight': 2, 'subsample':
0.811586905952513, 'colsample bytree': 0.629465513881339}. Best is
trial 0 with value: 0.9077765534112677.
[I 2025-10-25 06:11:20,240] Trial 9 finished with value:
0.8993110088740897 and parameters: {'max depth': 10, 'n_estimators':
599, 'learning rate': 0.00565826184943575, 'reg alpha':
0.19522055094233765, 'reg_lambda': 14.614466641579256, 'gamma':
0.5824015445845161, 'min_child_weight': 2, 'subsample':
0.8887212742576327, 'colsample_bytree': 0.5708322387742923}. Best is
```

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trial 0 with value: 0.9077765534112677.
[I 2025-10-25 06:13:56,469] Trial 10 finished with value:
0.9036887293735291 and parameters: {'max depth': 12, 'n estimators':
653, 'learning rate': 0.009099170006632542, 'reg alpha':
0.33048553987251267, 'reg_lambda': 16.45960992834629, 'gamma':
0.5725617683922862, 'min_child_weight': 1, 'subsample':
0.8623376518629466, 'colsample_bytree': 0.5855821797322525}. Best is
trial 0 with value: 0.9077765534112677.
[I 2025-10-25 06:15:44,150] Trial 11 finished with value:
0.9077766777591924 and parameters: {'max depth': 12, 'n estimators':
552, 'learning rate': 0.03282463863128617, 'reg alpha':
0.313593239150\overline{2}71, 'reg lambda': 16.177953024746373, 'gamma':
0.4642454341421508, 'min child weight': 1, 'subsample':
0.8399934012767569, 'colsample bytree': 0.5840374625454836}. Best is
trial 11 with value: 0.9077766777591924.
[I 2025-10-25 06:17:47,896] Trial 12 finished with value:
0.9072616182929061 and parameters: {'max depth': 12, 'n estimators':
626, 'learning_rate': 0.03216591388842767, 'reg_alpha':
0.34887518000952134, 'reg lambda': 16.998185207385877, 'gamma':
0.46843595490529355, 'min_child_weight': 1, 'subsample':
0.8506764259519106, 'colsample bytree': 0.5984174597871295}. Best is
trial 11 with value: 0.9077766777591924.
[I 2025-10-25 06:20:00,965] Trial 13 finished with value:
0.9076800697840554 and parameters: {'max depth': 8, 'n estimators':
698, 'learning_rate': 0.02436648204640415, 'reg alpha':
0.27392758607380197, 'reg_lambda': 15.450163445504144, 'gamma':
0.5242241657697266, 'min_child_weight': 1, 'subsample':
0.8400955969682045, 'colsample bytree': 0.5627211142441652}. Best is
trial 11 with value: 0.9077766777591924.
[I 2025-10-25 06:22:19,068] Trial 14 finished with value:
0.9072293810934369 and parameters: {'max_depth': 11, 'n_estimators':
664, 'learning rate': 0.025358102924802756, 'reg alpha':
0.2133934971611933, 'reg_lambda': 21.950639745161865, 'qamma':
0.4496692673051263, 'min child weight': 1, 'subsample':
0.8608307724547658, 'colsample bytree': 0.6057390623902266}. Best is
trial 11 with value: 0.9077766777591924.
[I 2025-10-25 06:24:07,892] Trial 15 finished with value:
0.9073260097932281 and parameters: {'max depth': 8, 'n estimators':
570, 'learning_rate': 0.024914080063535623, 'reg alpha':
0.2814726660750121, 'req lambda': 17.79963988304119, 'gamma':
0.49741826803821837, 'min_child_weight': 2, 'subsample': 0.8015510381994227, 'colsample_bytree': 0.5559381112062448}. Best is
trial 11 with value: 0.9077766777591924.
[I 2025-10-25 06:26:45,596] Trial 16 finished with value:
0.9045578591929695 and parameters: {'max depth': 11, 'n estimators':
615, 'learning_rate': 0.009670361349275526, 'reg_alpha':
0.3046117458228617, 'reg lambda': 12.055470418843898, 'gamma':
0.44697251074052136, 'min_child_weight': 1, 'subsample':
0.8602832256763634, 'colsample bytree': 0.5019768821571385}. Best is
```

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trial 11 with value: 0.9077766777591924.
[I 2025-10-25 06:28:37,202] Trial 17 finished with value:
0.9065856422497772 and parameters: {'max_depth': 7, 'n_estimators':
641, 'learning rate': 0.033316896254005544, 'reg alpha':
0.24583988949928673, 'reg lambda': 15.440049187671598, 'gamma':
0.5607118576476388, 'min_child_weight': 1, 'subsample':
0.8251154623606999, 'colsample bytree': 0.6486647890121541}. Best is
trial 11 with value: 0.9077766777591924.
[I 2025-10-25 06:31:09,532] Trial 18 finished with value:
0.903077144830722 and parameters: {'max depth': 9, 'n estimators':
616, 'learning rate': 0.010412824777274886, 'reg alpha':
0.15057751967828242, 'reg lambda': 18.21158267715644, 'gamma':
0.40749266361030134, 'min_child_weight': 1, 'subsample':
0.8460755104263068, 'colsample bytree': 0.5988244210473692}. Best is
trial 11 with value: 0.9077766777591924.
[I 2025-10-25 06:33:31,618] Trial 19 finished with value:
0.9077121722732727 and parameters: {'max depth': 12, 'n estimators':
674, 'learning_rate': 0.02040711315435472, 'reg_alpha':
0.3104326508280235, 'reg lambda': 15.387506828428595, 'gamma':
0.4601151687640659, 'min child weight': 2, 'subsample':
0.8704654344673929, 'colsample bytree': 0.5500693830180028}. Best is
trial 11 with value: 0.9077766777591924.
[I 2025-10-25 06:35:23,849] Trial 20 finished with value:
0.9069397540522399 and parameters: {'max depth': 11, 'n estimators':
575, 'learning_rate': 0.027381627181028344, 'reg alpha':
0.25780547625221695, 'reg lambda': 21.909587762190903, 'qamma':
0.502250267767581, 'min_child_weight': 2, 'subsample':
0.8379685076854719, 'colsample bytree': 0.5890398887498067}. Best is
trial 11 with value: 0.9077766777591924.
[I 2025-10-25 06:37:17,833] Trial 21 finished with value:
0.9077122240849083 and parameters: {'max_depth': 11, 'n_estimators':
556, 'learning rate': 0.02959919203233837, 'reg alpha':
0.30587942879952934, 'reg_lambda': 14.242003033857086, 'qamma':
0.45021885550094864, 'min child weight': 1, 'subsample':
0.8245917580944304, 'colsample bytree': 0.5780599190557069}. Best is
trial 11 with value: 0.9077766777591924.
[I 2025-10-25 06:39:10,031] Trial 22 finished with value:
0.9075513904066821 and parameters: {'max depth': 12, 'n estimators':
552, 'learning_rate': 0.034919414052455584, 'reg_alpha':
0.32447075977755074, 'reg lambda': 12.83691991014207, 'gamma':
0.43709187465481303, 'min_child_weight': 1, 'subsample':
0.8373373015106534, 'colsample bytree': 0.6123732454301051}. Best is
trial 11 with value: 0.9077766777591924.
[I 2025-10-25 06:41:15,356] Trial 23 finished with value:
0.9075190703085964 and parameters: {'max depth': 11, 'n estimators':
578, 'learning rate': 0.022492633605856304, 'reg alpha':
0.29063159664675053, 'reg_lambda': 13.78251583725749, 'gamma':
0.4340061104793146, 'min_child_weight': 1, 'subsample': 0.8156841177074695, 'colsample_bytree': 0.5706459492594185}. Best is
```

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trial 11 with value: 0.9077766777591924.
[I 2025-10-25 06:43:09,008] Trial 24 finished with value:
0.9074869160077436 and parameters: {'max depth': 12, 'n estimators':
552, 'learning rate': 0.02842292942301311, 'reg alpha':
0.3481113385254543, 'reg_lambda': 14.786862254705976, 'gamma': 0.46988916885757026, 'min_child_weight': 1, 'subsample':
0.8525362555539477, 'colsample bytree': 0.5864306453643248}. Best is
trial 11 with value: 0.9077766777591924.
[I 2025-10-25 06:45:33,599] Trial 25 finished with value:
0.9073258647206492 and parameters: {'max depth': 11, 'n estimators':
582, 'learning rate': 0.016812718012264895, 'reg alpha':
0.2662775039624314, 'reg_lambda': 12.488354342857185, 'gamma':
0.40022101754446626, 'min_child_weight': 1, 'subsample':
0.8282513797528969, 'colsample bytree': 0.5792891879216953}. Best is
trial 11 with value: 0.9077766777591924.
[I 2025-10-25 06:47:34,883] Trial 26 finished with value:
0.9071972475172382 and parameters: {'max depth': 9, 'n estimators':
564, 'learning_rate': 0.02199098254806664, 'reg alpha':
0.2262003953171724, 'reg lambda': 13.769971568991279, 'gamma':
0.5393794128932019, 'min child weight': 1, 'subsample':
0.8146043630671075, 'colsample bytree': 0.5992386289844998}. Best is
trial 11 with value: 0.9077766777591924.
[I 2025-10-25 06:49:55,601] Trial 27 finished with value:
0.904107170502351 and parameters: {'max depth': 10, 'n estimators':
613, 'learning rate': 0.01141014238867511, 'reg alpha':
0.29023999394402983, 'reg_lambda': 16.71990360344484, 'gamma':
0.4320065467877798, 'min_child_weight': 2, 'subsample':
0.83459035734124, 'colsample bytree': 0.5633924847994792}. Best is
trial 11 with value: 0.9077766777591924.
[I 2025-10-25 06:51:54,896] Trial 28 finished with value:
0.9076478533092402 and parameters: {'max_depth': 12, 'n_estimators':
586, 'learning rate': 0.029078792615753926, 'reg alpha':
0.3251132148201935, 'reg_lambda': 15.831107951605633, 'gamma':
0.471481820063784, 'min_child_weight': 1, 'subsample': 0.882457046942987, 'colsample_bytree': 0.5484689727278169}. Best is
trial 11 with value: 0.9077766777591924.
[I 2025-10-25 06:54:06,200] Trial 29 finished with value:
0.9068108881529795 and parameters: {'max depth': 9, 'n estimators':
633, 'learning rate': 0.01741200190894966, 'reg alpha':
0.18784478118409612, 'reg lambda': 14.19783638646906, 'gamma':
0.49616340423963334, 'min child weight': 2, 'subsample':
0.8471511671436528, 'colsample bytree': 0.5206446640342581}. Best is
trial 11 with value: 0.9077766777591924.
[I 2025-10-25 06:56:31,027] Trial 30 finished with value:
0.9064246945859535 and parameters: {'max depth': 8, 'n estimators':
656, 'learning rate': 0.015703865281000318, 'reg alpha':
0.23395897210127894, 'reg_lambda': 12.999302635297232, 'gamma':
0.5092001257375669, 'min_child_weight': 1, 'subsample':
0.8699638025472248, 'colsample bytree': 0.6138626324149455}. Best is
```

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trial 11 with value: 0.9077766777591924.
[I 2025-10-25 06:58:21,933] Trial 31 finished with value:
0.9080019133000675 and parameters: {'max depth': 11, 'n estimators':
561, 'learning rate': 0.030674455216771257, 'reg alpha':
0.3006180391529826, 'reg_lambda': 14.253617181349707, 'gamma': 0.45240079942853084, 'min_child_weight': 1, 'subsample':
0.819683870778105, 'colsample bytree': 0.5773993818832872}. Best is
trial 31 with value: 0.9080019133000675.
[I 2025-10-25 07:00:15,844] Trial 32 finished with value:
0.9077444820090314 and parameters: {'max depth': 11, 'n estimators':
562, 'learning_rate': 0.03161978859949837, 'reg alpha':
0.25983015161516354, 'reg_lambda': 14.835844837463998, 'gamma':
0.4839493078907537, 'min_child_weight': 1, 'subsample':
0.8058877036179068, 'colsample bytree': 0.5918946010101149}. Best is
trial 31 with value: 0.9080019133000675.
[I 2025-10-25 07:02:14,794] Trial 33 finished with value:
0.9075835239828807 and parameters: {'max depth': 11, 'n estimators':
572, 'learning rate': 0.030083877622823062, 'reg alpha':
0.25719043268305714, 'reg lambda': 14.939452758224508, 'gamma':
0.48431880203436073, 'min_child_weight': 1, 'subsample':
0.8048890926121655, 'colsample bytree': 0.5912150868966515}. Best is
trial 31 with value: 0.9080019133000675.
[I 2025-10-25 07:04:09,720] Trial 34 finished with value:
0.9072937415067777 and parameters: {'max depth': 10, 'n estimators':
566, 'learning rate': 0.026228739346824174, 'reg alpha':
0.2905810269713846, 'req lambda': 16.02496386106218, 'qamma':
0.45411788280204995, 'min_child_weight': 1, 'subsample':
0.81814870543741, 'colsample bytree': 0.5742357211642635}. Best is
trial 31 with value: 0.9080019133000675.
[I 2025-10-25 07:06:21,868] Trial 35 finished with value:
0.9076478429469133 and parameters: {'max_depth': 12, 'n_estimators':
560, 'learning_rate': 0.023149161566648955, 'reg_alpha':
0.2653831240808197, 'reg_lambda': 14.012172332074933, 'gamma':
0.4279634922091168, 'min_child_weight': 1, 'subsample':
0.807434144490633, 'colsample bytree': 0.6118814656193448}. Best is
trial 31 with value: 0.9080019133000675.
[I 2025-10-25 07:08:16,942] Trial 36 finished with value:
0.9079376668723242 and parameters: {'max depth': 11, 'n estimators':
606, 'learning rate': 0.03138262475678284, 'reg alpha':
0.21084241451514502, 'reg lambda': 13.315490102053365, 'gamma':
0.48036942149782097, 'min_child_weight': 3, 'subsample':
0.8190998377815791, 'colsample bytree': 0.5620625406334134}. Best is
trial 31 with value: 0.9080019133000675.
[I 2025-10-25 07:09:58,107] Trial 37 finished with value:
0.9072937933184131 and parameters: {'max depth': 10, 'n estimators':
605, 'learning rate': 0.03491311918958484, 'reg alpha':
0.19684196885297664, 'reg_lambda': 13.200500053371183, 'gamma':
0.6298887534073297, 'min_child_weight': 3, 'subsample':
0.8217056242771874, 'colsample bytree': 0.540536081812291}. Best is
```

```
trial 31 with value: 0.9080019133000675.
[I 2025-10-25 07:12:08,150] Trial 38 finished with value:
0.9071971231693136 and parameters: {'max depth': 10, 'n estimators':
589, 'learning rate': 0.018523857936931103, 'reg alpha':
0.2143068782569056, 'reg_lambda': 12.474636799445996, 'gamma': 0.46656481924213195, 'min_child_weight': 3, 'subsample':
0.8993869386119541, 'colsample bytree': 0.5632776275474294}. Best is
trial 31 with value: 0.9080019133000675.
[I 2025-10-25 07:14:55,557] Trial 39 finished with value:
0.9029162075292254 and parameters: {'max depth': 12, 'n estimators':
684, 'learning_rate': 0.007204284910352362, 'reg alpha':
0.22387428617463154, 'reg_lambda': 13.323568287182, 'gamma':
0.5363145996893269, 'min_child_weight': 3, 'subsample':
0.8337845867969467, 'colsample bytree': 0.57911780345701}. Best is
trial 31 with value: 0.9080019133000675.
[I 2025-10-25 07:17:07,266] Trial 40 finished with value:
0.9077122966211976 and parameters: {'max depth': 9, 'n estimators':
645, 'learning_rate': 0.021495628535436594, 'reg alpha':
0.2081251697647601, 'reg lambda': 12.662387130268861, 'gamma':
0.5087545609163029, 'min child weight': 3, 'subsample':
0.8441425263311085, 'colsample bytree': 0.5555789387612682}. Best is
trial 31 with value: 0.9080019133000675.
[I 2025-10-25 07:18:59,150] Trial 41 finished with value:
0.9073904323805312 and parameters: {'max depth': 11, 'n estimators':
560, 'learning rate': 0.03152049793182594, 'reg alpha':
0.23912437241255924, 'reg lambda': 14.520270371474979, 'gamma':
0.48209909861567146, 'min_child_weight': 1, 'subsample':
0.8098161765801741, 'colsample bytree': 0.5970630933100485}. Best is
trial 31 with value: 0.9080019133000675.
[I 2025-10-25 07:20:57,211] Trial 42 finished with value:
0.9080663358873705 and parameters: {'max depth': 11, 'n estimators':
602, 'learning rate': 0.031195089393552138, 'reg alpha':
0.2460091470352259, 'reg lambda': 13.62247629443974, 'gamma':
0.48605466284091947, 'min_child_weight': 2, 'subsample':
0.8193843879387073, 'colsample bytree': 0.5701408647413307}. Best is
trial 42 with value: 0.9080663358873705.
[I 2025-10-25 07:22:59,842] Trial 43 finished with value:
0.9078410485348604 and parameters: {'max depth': 10, 'n estimators':
603, 'learning_rate': 0.027982455866065337, 'reg alpha':
0.2458075554626403, 'reg lambda': 13.539209624000724, 'gamma':
0.44119071713453967, 'min_child_weight': 2, 'subsample':
0.8192419049816344, 'colsample bytree': 0.5714224224183675}. Best is
trial 42 with value: 0.9080663358873705.
[I 2025-10-25 07:25:00,378] Trial 44 finished with value:
0.907744430197396 and parameters: {'max depth': 11, 'n estimators':
605, 'learning rate': 0.027410851347189714, 'reg alpha':
0.24604376236903275, 'reg lambda': 13.575870459429554, 'gamma':
0.4599490561046012, 'min_child_weight': 2, 'subsample':
0.8200269290029351, 'colsample bytree': 0.5673646448904011}. Best is
```

```
trial 42 with value: 0.9080663358873705.
[I 2025-10-25 07:27:04,347] Trial 45 finished with value:
0.9078410692595144 and parameters: {'max depth': 10, 'n estimators':
623, 'learning rate': 0.027002221022745784, 'reg alpha':
0.20406582112895838, 'reg_lambda': 12.023511495735665, 'gamma': 0.41802186158009996, 'min_child_weight': 2, 'subsample':
0.828880673669941, 'colsample bytree': 0.5424071440844247}. Best is
trial 42 with value: 0.9080663358873705.
[I 2025-10-25 07:29:11,539] Trial 46 finished with value:
0.9077123173458517 and parameters: {'max depth': 10, 'n estimators':
621, 'learning rate': 0.02643347085868387, 'reg alpha':
0.20413700995903986, 'reg lambda': 12.2631809371444, 'gamma':
0.4192936588256685, 'min child weight': 2, 'subsample':
0.8313057544186089, 'colsample bytree': 0.5409183973231356}. Best is
trial 42 with value: 0.9080663358873705.
[I 2025-10-25 07:31:22,063] Trial 47 finished with value:
0.9071650414047502 and parameters: {'max depth': 10, 'n estimators':
599, 'learning_rate': 0.023552073841192525, 'reg alpha':
0.1946474074155541, 'reg lambda': 13.047911297950948, 'gamma':
0.4420333714824629, 'min child weight': 2, 'subsample':
0.8129070290096827, 'colsample bytree': 0.5562174310370064}. Best is
trial 42 with value: 0.9080663358873705.
[I 2025-10-25 07:33:24,286] Trial 48 finished with value:
0.9072616286552332 and parameters: {'max depth': 11, 'n estimators':
632, 'learning_rate': 0.02977908564709348, 'reg alpha':
0.1815904976504848, 'req lambda': 12.03241209042288, 'qamma':
0.4235150948139719, 'min_child_weight': 2, 'subsample':
0.8210054330181455, 'colsample bytree': 0.5333062897662957}. Best is
trial 42 with value: 0.9080663358873705.
[I 2025-10-25 07:35:31,280] Trial 49 finished with value:
0.9079054711221636 and parameters: {'max depth': 10, 'n estimators':
608, 'learning_rate': 0.025593926069678283, 'reg_alpha':
0.2244865711035126, 'reg lambda': 13.438828816583314, 'gamma':
0.41445747243364145, 'min_child_weight': 2, 'subsample':
0.8293759630608158, 'colsample bytree': 0.5475519263572838}. Best is
trial 42 with value: 0.9080663358873705.
[I 2025-10-25 07:37:35,720] Trial 50 finished with value:
0.9076477704106237 and parameters: {'max depth': 10, 'n estimators':
625, 'learning_rate': 0.024599660178505064, 'reg_alpha':
0.22313106825295487, 'reg lambda': 12.82032714752227, 'gamma':
0.4752434641542141, 'min_child_weight': 2, 'subsample': 0.8300621837662086, 'colsample_bytree': 0.5240530704489271}. Best is
trial 42 with value: 0.9080663358873705.
[I 2025-10-25 07:39:35,114] Trial 51 finished with value:
0.9077766777591924 and parameters: {'max depth': 10, 'n estimators':
606, 'learning_rate': 0.031428937945221655, 'reg_alpha':
0.20419737901838797, 'req lambda': 13.506469563358555, 'qamma':
0.4029186480169, 'min child weight': 2, 'subsample':
0.824757212090764, 'colsample bytree': 0.5473894549302011}. Best is
```

```
trial 42 with value: 0.9080663358873705.
[I 2025-10-25 07:41:41,270] Trial 52 finished with value:
0.9080019547493758 and parameters: {'max depth': 11, 'n estimators':
611, 'learning rate': 0.02785344980698432, 'reg alpha':
0.2361197034126573, 'reg_lambda': 13.898773756856004, 'gamma': 0.41220195600698206, 'min_child_weight': 2, 'subsample':
0.8175836278195951, 'colsample bytree': 0.5575890350668292}. Best is
trial 42 with value: 0.9080663358873705.
[I 2025-10-25 07:43:45,170] Trial 53 finished with value:
0.9074224934204406 and parameters: {'max depth': 11, 'n estimators':
594, 'learning rate': 0.02560999679976465, 'reg alpha':
0.2322506994614165, 'reg lambda': 13.96149283332096, 'gamma':
0.41570329197013384, 'min_child_weight': 3, 'subsample':
0.8153051070071936, 'colsample bytree': 0.5554515785544147}. Best is
trial 42 with value: 0.9080663358873705.
[I 2025-10-25 07:46:00,599] Trial 54 finished with value:
0.9077765844982488 and parameters: {'max depth': 11, 'n estimators':
611, 'learning rate': 0.020060980606179606, 'reg alpha':
0.215166979797605, 'req lambda': 14.452561154769537, 'gamma':
0.41178699767946225, 'min_child_weight': 2, 'subsample':
0.801379561710889, 'colsample bytree': 0.5344400667161776}. Best is
trial 42 with value: 0.9080663358873705.
[I 2025-10-25 07:47:57,398] Trial 55 finished with value:
0.9078088631470262 and parameters: {'max depth': 11, 'n estimators':
618, 'learning_rate': 0.034646464418638355, 'reg alpha':
0.22230118267811702, 'reg lambda': 13.23421113199199, 'qamma':
0.4244164718521905, 'min_child_weight': 2, 'subsample':
0.8107417605938473, 'colsample bytree': 0.5463300405982594}. Best is
trial 42 with value: 0.9080663\overline{3}58873705.
[I 2025-10-25 07:50:22,829] Trial 56 finished with value:
0.904976186336194 and parameters: {'max depth': 11, 'n estimators':
631, 'learning rate': 0.012793511474506365, 'reg alpha':
0.2379919933238532, 'reg_lambda': 15.179657933561483, 'qamma':
0.4099408236908283, 'min child weight': 3, 'subsample':
0.8238231806313936, 'colsample bytree': 0.5599482595674089}. Best is
trial 42 with value: 0.9080663358873705.
[I 2025-10-25 07:52:57,483] Trial 57 finished with value:
0.9022402107614423 and parameters: {'max depth': 11, 'n estimators':
610, 'learning rate': 0.007745621975040991, 'reg alpha':
0.17456568120245688, 'reg lambda': 14.047984086582765, 'gamma':
0.45546909624697657, 'min_child_weight': 2, 'subsample':
0.8297592407358606, 'colsample bytree': 0.5676741791390799}. Best is
trial 42 with value: 0.9080663358873705.
[I 2025-10-25 07:54:55,199] Trial 58 finished with value:
0.9074225452320757 and parameters: {'max depth': 10, 'n estimators':
621, 'learning_rate': 0.030621450117877883, 'reg_alpha':
0.1990021660851675, 'reg lambda': 12.549635611476539, 'gamma':
0.43285911332829635, 'min child weight': 2, 'subsample':
0.8178259085665496, 'colsample bytree': 0.5104665640801154}. Best is
```

```
trial 42 with value: 0.9080663358873705.
[I 2025-10-25 07:56:41,129] Trial 59 finished with value:
0.9073903909312229 and parameters: {'max_depth': 9, 'n_estimators':
600, 'learning rate': 0.033108753601876706, 'reg alpha':
0.18888564938234867, 'reg_lambda': 13.832459281486255, 'gamma':
0.5156405043913123, 'min_child_weight': 3, 'subsample':
0.8266367725786224, 'colsample bytree': 0.5525770314593279}. Best is
trial 42 with value: 0.9080663358873705.
[I 2025-10-25 07:58:42,627] Trial 60 finished with value:
0.9076156368344253 and parameters: {'max depth': 11, 'n estimators':
594, 'learning rate': 0.02406284949602339, 'reg alpha':
0.20912526401353435, 'reg lambda': 12.84102879048461, 'gamma':
0.4903741886583368, 'min_child_weight': 2, 'subsample':
0.8335655108876728, 'colsample bytree': 0.5436273619393318}. Best is
trial 42 with value: 0.9080663358873705.
[I 2025-10-25 08:00:44,040] Trial 61 finished with value:
0.9077122551718892 and parameters: {'max depth': 10, 'n estimators':
601, 'learning rate': 0.027776246117063155, 'reg alpha':
0.24668931972448088, 'reg lambda': 14.30016109134842, 'gamma':
0.4429532004305743, 'min_child_weight': 2, 'subsample': 0.8181584400842757, 'colsample_bytree': 0.5674346415776705}. Best is
trial 42 with value: 0.9080663358873705.
[I 2025-10-25 08:02:40,817] Trial 62 finished with value:
0.9076478843962213 and parameters: {'max depth': 10, 'n estimators':
606, 'learning_rate': 0.0282631478645281, 'reg alpha':
0.21915721433908691, 'reg lambda': 13.50190680102787, 'gamma':
0.444310210650178, 'min_child_weight': 2, 'subsample':
0.8226941034063092, 'colsample bytree': 0.5748799461936797}. Best is
trial 42 with value: 0.9080663358873705.
[I 2025-10-25 08:04:41,713] Trial 63 finished with value:
0.9075513282327197 and parameters: {'max_depth': 10, 'n_estimators':
585, 'learning rate': 0.027660090542247016, 'reg alpha':
0.2295091441661764, 'reg lambda': 13.662998211755538, 'gamma':
0.43798491817243235, 'min_child_weight': 2, 'subsample':
0.8092694733844165, 'colsample bytree': 0.5829936004196232}. Best is
trial 42 with value: 0.9080663358873705.
[I 2025-10-25 08:06:56,889] Trial 64 finished with value:
0.9074869367323977 and parameters: {'max depth': 11, 'n estimators':
627, 'learning_rate': 0.025578191453595456, 'reg_alpha':
0.2513113988789573, 'reg_lambda': 13.137863543280197, 'gamma':
0.4258379677241409, 'min child weight': 2, 'subsample':
0.8141288178024975, 'colsample bytree': 0.561540796720919}. Best is
trial 42 with value: 0.9080663358873705.
[I 2025-10-25 08:08:59,663] Trial 65 finished with value:
0.9077766984838463 and parameters: {'max depth': 9, 'n estimators':
637, 'learning_rate': 0.032516502997551544, 'reg_alpha':
0.2415207568264026, 'reg lambda': 12.30910893810\overline{4}396, 'qamma':
0.4078960252909845, 'min_child_weight': 2, 'subsample':
0.8281274058434309, 'colsample bytree': 0.5709281732482866}. Best is
```

```
trial 42 with value: 0.9080663358873705.
[I 2025-10-25 08:11:11,285] Trial 66 finished with value:
0.9074547098952556 and parameters: {'max_depth': 11, 'n_estimators':
619, 'learning rate': 0.021196236949773142, 'reg alpha':
0.26868501586626126, 'reg lambda': 13.383765790689377, 'qamma':
0.4513382493019402, 'min_child_weight': 2, 'subsample':
0.8401679795881903, 'colsample bytree': 0.5271410013247315}. Best is
trial 42 with value: 0.9080663358873705.
[I 2025-10-25 08:13:12,997] Trial 67 finished with value:
0.9075512971457383 and parameters: {'max depth': 10, 'n estimators':
610, 'learning rate': 0.02934305834237829, 'reg alpha':
0.28336952852574837, 'reg_lambda': 14.626698866969393, 'gamma':
0.4191651873790483, 'min_child_weight': 2, 'subsample':
0.817431946341585, 'colsample bytree': 0.5344774510486572}. Best is
trial 42 with value: 0.9080663358873705.
[I 2025-10-25 08:15:15,318] Trial 68 finished with value:
0.9075513075080656 and parameters: {'max depth': 10, 'n estimators':
580, 'learning_rate': 0.022683451461595377, 'reg_alpha':
0.1634036040782201, 'reg lambda': 14.137030768541806, 'gamma':
0.475031567970217, 'min child weight': 2, 'subsample':
0.803816318416277, 'colsample bytree': 0.5604321170956216}. Best is
trial 42 with value: 0.9080663358873705.
[I 2025-10-25 08:17:26,860] Trial 69 finished with value:
0.9082595414753177 and parameters: {'max depth': 12, 'n estimators':
590, 'learning rate': 0.026442178717374906, 'reg alpha':
0.22773759320643863, 'reg lambda': 12.698132398371067, 'gamma':
0.4021718785956999, 'min_child_weight': 2, 'subsample': 0.8568135817261427, 'colsample_bytree': 0.5818264816483791}. Best is
trial 69 with value: 0.9082595414753177.
[I 2025-10-25 08:19:33,881] Trial 70 finished with value:
0.9082273146381755 and parameters: {'max depth': 12, 'n estimators':
590, 'learning_rate': 0.026577254714110408, 'reg alpha':
0.2109740974977536, 'reg lambda': 12.202720549946353, 'gamma':
0.4115544026392923, 'min child weight': 2, 'subsample':
0.8627632573104782, 'colsample bytree': 0.5512906164779097}. Best is
trial 69 with value: 0.9082595414753177.
[I 2025-10-25 08:21:45,187] Trial 71 finished with value:
0.9077767088461736 and parameters: {'max depth': 12, 'n estimators':
590, 'learning_rate': 0.02616575254194039, 'reg_alpha':
0.2094702276518697, 'reg lambda': 12.187237501156943, 'gamma':
0.4002745958412632, 'min_child_weight': 2, 'subsample':
0.864161610906011, 'colsample bytree': 0.5515565916607229}. Best is
trial 69 with value: 0.9082595414753177.
[I 2025-10-25 08:23:46,068] Trial 72 finished with value:
0.9082594896636824 and parameters: {'max depth': 12, 'n estimators':
596, 'learning rate': 0.030698701547765647, 'reg alpha':
0.20309762450051652, 'reg_lambda': 12.738646591066333, 'gamma':
0.4125485510339032, 'min_child_weight': 2, 'subsample': 0.8541459214533714, 'colsample_bytree': 0.5808384487399275}. Best is
trial 69 with value: 0.9082595414753177.
```

```
[I 2025-10-25 08:26:28,141] Trial 73 finished with value:
0.8998581604672664 and parameters: {'max depth': 12, 'n estimators':
597, 'learning rate': 0.0051282507091826475, 'reg alpha':
0.21839577465315727, 'reg_lambda': 12.579488596905728, 'qamma':
0.410404901715178, 'min child weight': 2, 'subsample':
0.8576297706612442, 'colsample_bytree': 0.5845811877781516}. Best is
trial 69 with value: 0.9082595414753177.
[I 2025-10-25 08:28:27,783] Trial 74 finished with value:
0.908130789561655 and parameters: {'max depth': 12, 'n estimators':
587, 'learning rate': 0.030535067409386656, 'reg alpha':
0.23060610529765269, 'reg_lambda': 12.7836928550274, 'gamma':
0.42837180847462575, 'min child weight': 2, 'subsample':
0.8533780182058618, 'colsample bytree': 0.5807610727252992}. Best is
trial 69 with value: 0.9082595414753177.
[I 2025-10-25 08:30:24,780] Trial 75 finished with value:
0.9076801112333636 and parameters: {'max depth': 12, 'n estimators':
574, 'learning rate': 0.03211178941076371, 'reg alpha':
0.23483833392341752, 'reg_lambda': 12.863379653705037, 'gamma':
0.4295735207042843, 'min_child_weight': 2, 'subsample':
0.8544275694860042, 'colsample_bytree': 0.5785597953295201}. Best is
trial 69 with value: 0.9082595414753177.
[I 2025-10-25 08:32:18,170] Trial 76 finished with value:
0.9079375943360348 and parameters: {'max depth': 12, 'n estimators':
587, 'learning rate': 0.030551341494947325, 'reg alpha':
0.19132783064859712, 'reg lambda': 12.682344575780801, 'gamma':
0.46093925283617565, 'min_child_weight': 3, 'subsample':
0.8686609799348528, 'colsample bytree': 0.593400883625728}. Best is
trial 69 with value: 0.9082595414753177.
[I 2025-10-25 08:34:14,489] Trial 77 finished with value:
0.9075190910332503 and parameters: {'max_depth': 12, 'n_estimators':
578, 'learning rate': 0.034469957763186905, 'reg alpha':
0.21563436195633043, 'reg_lambda': 20.592020422194672, 'gamma':
0.4052176006691495, 'min_child_weight': 2, 'subsample':
0.8551772276827085, 'colsample bytree': 0.5813834209921134}. Best is
trial 69 with value: 0.9082595414753177.
[I 2025-10-25 08:36:25,719] Trial 78 finished with value:
0.9071649688684609 and parameters: {'max depth': 12, 'n estimators':
569, 'learning rate': 0.018990924129276183, 'reg alpha':
0.22927992509621367, 'reg_lambda': 13.050150833746645, 'gamma':
0.43566203018489347, 'min_child_weight': 2, 'subsample':
0.8661060555218771, 'colsample bytree': 0.5748619406405694}. Best is
trial 69 with value: 0.9082595414753177.
[I 2025-10-25 08:38:14,276] Trial 79 finished with value:
0.9071650621294044 and parameters: {'max depth': 12, 'n estimators':
594, 'learning rate': 0.028939792144467638, 'reg_alpha':
0.2543926157715564, 'reg_lambda': 12.364834361870141, 'gamma':
0.6071339351588716, 'min_child_weight': 2, 'subsample':
0.8498677865728134, 'colsample_bytree': 0.5641274507069189}. Best is
trial 69 with value: 0.9082595414753177.
```

```
Best CV accuracy: 0.90826
Best hyperparameters found:
  max depth: 8
  n estimators: 590
  learning rate: 0.026442178717374906
  reg alpha: 0.22773759320643863
  reg lambda: 12.698132398371067
  gamma: 0.4021718785956999
  min child weight: 2
  subsample: 0.8568135817261427
  colsample bytree: 0.5818264816483791
Training final model with best parameters...
Preparing and preprocessing test data...
Test data preprocessing complete.
Generating test predictions...
 Submission saved to: /content/drive/MyDrive/Obesity
Data/xgb maxdepth8 optimized.csv
Sample Submission:
      id
               WeightCategory
  15533
             Obesity_Type_III
1
  15534
           Overweight Level I
2 15535 Overweight Level II
3 15536
              Obesity_Type_II
4 15537
                Normal Weight
```

optuna search on best 2 accuracies till date(.91239)

```
import pandas as pd
import numpy as np
import time
import warnings
from sklearn.preprocessing import StandardScaler, LabelEncoder
from sklearn.model_selection import RepeatedStratifiedKFold
import xgboost as xgb
import optuna

warnings.filterwarnings("ignore")

file_path = '/home/iiitb/Desktop/IIITB/ML/project_first_half/'
df = pd.read_csv(file_path + 'train.csv')
df_test = pd.read_csv(file_path + 'test.csv')
X = df.drop(['id','WeightCategory'], axis=1)
y = df['WeightCategory']

categorical_cols = X.select_dtypes(include=['object']).columns
```

```
X = pd.get dummies(X, columns=categorical cols, drop first=True)
numerical cols =
['Age','Height','Weight','FCVC','NCP','CH20','FAF','TUE']
scaler = StandardScaler()
X[numerical cols] = scaler.fit transform(X[numerical cols])
le = LabelEncoder()
y enc = le.fit transform(y)
center1 = {'n_estimators': 1244, 'learning rate': 0.01203757,
'max_depth': 11, 'subsample': 0.6732769, 'colsample_bytree': 0.5301610, 'min_child_weight': 3, 'reg_alpha': 0.1385014,
'reg_lambda': 13.9555974, 'gamma': 0.5} center2 = {'n_estimators': 1403, 'learning_rate': 0.01151984,
'max depth': 13, 'subsample': 0.8389925, 'colsample bytree':
0.4289206, 'min child weight': 1, 'reg alpha': 0.2172063,
'reg lambda': 3.2747964, 'gamma': 0.3849006}
def objective(trial):
    ne lo = int(min(center1['n estimators'], center2['n estimators'])
* 0.95)
    ne hi = int(max(center1['n estimators'], center2['n estimators'])
* 1.05)
    lr lo = min(center1['learning rate'], center2['learning rate']) *
    lr hi = max(center1['learning rate'], center2['learning rate']) *
1.2
    ss_lo = min(center1['subsample'], center2['subsample']) - 0.05
    ss hi = max(center1['subsample'], center2['subsample']) + 0.05
    cs lo = min(center1['colsample bytree'],
center2['colsample bytree']) - 0.05
    cs hi = max(center1['colsample bytree'],
center2['colsample bytree']) + 0.05
    ra_lo = min(center1['reg_alpha'], center2['reg_alpha']) * 0.5
    ra_hi = max(center1['reg_alpha'], center2['reg_alpha']) * 1.5
    rl lo = min(center1['reg lambda'], center2['reg lambda']) * 0.8
    rl hi = max(center1['reg_lambda'], center2['reg_lambda']) * 1.2
    gm_lo = min(center1['gamma'], center2['gamma']) - 0.1
    gm hi = max(center1['gamma'], center2['gamma']) + 0.1
    params = {
        'n estimators': trial.suggest int('n estimators', ne lo,
ne hi),
        'learning rate': trial.suggest float('learning rate', lr lo,
lr hi, log=True),
         'max depth': trial.suggest int('max depth',
min(center1['max depth'], center2['max depth']),
max(center1['max_depth'], center2['max_depth'])),
         'subsample': trial.suggest float('subsample', max(0.5, ss lo),
```

```
min(1.0, ss hi)),
        'colsample bytree': trial.suggest float('colsample bytree',
\max(0.4, cs_{lo}), \min(1.0, cs_{hi})),
        'min child weight': trial.suggest int('min child weight',
min(center1['min_child_weight'], center2['min_child_weight']),
max(center1['min_child_weight'], center2['min_child_weight'])),
        'req alpha': trial.suggest float('req alpha', ra lo, ra hi,
log=True),
         reg lambda': trial.suggest float('reg lambda', rl lo, rl hi,
log=True),
         gamma': trial.suggest float('gamma', max(0.0, gm lo), gm hi),
        'objective': 'multi:softmax',
        'num class': len(le.classes ),
        'use label encoder': False,
        'eval metric': 'mlogloss',
        'random state': 42,
        'verbosity': 0,
        'n_jobs': -1,
        'early stopping rounds': 50
    }
    cv = RepeatedStratifiedKFold(n splits=5, n repeats=3,
random state=42)
    scores = []
    for train idx, val_idx in cv.split(X, y_enc):
        X tr, X val = X.iloc[train idx], X.iloc[val idx]
        y tr, y val = y enc[train idx], y enc[val idx]
        model = xqb.XGBClassifier(**params)
        model.fit(X_tr, y_tr, eval_set=[(X val, y val)],
verbose=False)
        scores.append(model.score(X val, y val))
    return np.mean(scores)
pruner = optuna.pruners.MedianPruner(n startup trials=10,
n warmup steps=5)
study = optuna.create study(direction='maximize',
sampler=optuna.samplers.TPESampler(seed=42), pruner=pruner)
start = time.time()
study.optimize(objective, n trials=100, show progress bar=True)
end = time.time()
print(f"Done in {(end - start)/60:.2f} min")
print(f"Best CV accuracy: {study.best value:.5f}")
print("Best params:", study.best_params)
best = study.best params.copy()
best.update({'objective':'multi:softmax','num class':len(le.classes ),
'use_label_encoder':False,'eval_metric':'mlogloss','random_state':42,'
verbosity':0,'n jobs':-1})
```

```
final model = xgb.XGBClassifier(**best)
final model.fit(X, y enc, verbose=False)
X test = df test.drop(['id'], axis=1)
X test = pd.get dummies(X test, columns=categorical cols,
drop first=True)
X test = X test.reindex(columns=X.columns, fill value=0)
X test[numerical cols] = scaler.transform(X test[numerical cols])
y test pred enc = final model.predict(X test)
y test pred = le.inverse transform(y test pred enc)
submission = pd.DataFrame({'id': df test['id'], 'WeightCategory':
y test pred})
submission.to csv(file path + 'xq optuna best2.csv', index=False)
print("Submission saved.")
[I 2025-10-23 20:25:51,122] A new study created in memory with name:
no-name-8e411c47-f418-44be-b839-47b0b76020d1
Best trial: 0. Best value: 0.908431:
                                                    | 1/100
[04:18<7:06:42, 258.61s/it]
[I 2025-10-23 20:30:09,725] Trial 0 finished with value:
0.9084311899688804 and parameters: {'n estimators': 1290,
'learning rate': 0.014128638416865078, 'max depth': 13, 'subsample':
0.7823497983235062, 'colsample bytree': 0.42810847428074983,
'min child weight': 1, 'reg alpha': 0.07576832591389868, 'reg lambda':
13.065131015793716, 'gamma': 0.474311579531278}. Best is trial 0 with
value: 0.9084311899688804.
Best trial: 0. Best value: 0.908431: 2%| | 2/100
[09:03<7:27:55, 274.24s/it]
[I 2025-10-23 20:34:54,910] Trial 1 finished with value:
0.9081951568832137 and parameters: {'n estimators': 1388,
'learning_rate': 0.00930152570615882, 'max_depth': 13, 'subsample':
0.8444698957658686, 'colsample bytree': 0.4382552265189089,
'min_child_weight': 1, 'reg_alpha': 0.0919964241115056, 'reg_lambda':
4.606686872823836, 'gamma': 0.4502510367534591}. Best is trial 0 with
value: 0.9084311899688804.
Best trial: 0. Best value: 0.908431: 3%|| | 3/100
[13:48<7:31:04, 279.01s/it]
[I 2025-10-23 20:39:39,601] Trial 2 finished with value:
0.9074654487201937 and parameters: {'n_estimators': 1307,
'learning rate': 0.010504601870873372, 'max_depth': 12, 'subsample':
0.6603425948794737, 'colsample bytree': 0.45263307202475345,
'min child weight': 2, 'reg alpha': 0.1403295882644004, 'reg lambda':
11.242309423353674, 'gamma': 0.3478176889538298}. Best is trial 0 with
value: 0.9084311899688804.
```

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Best trial: 0. Best value: 0.908431: 4%
                                            | 4/100
[18:07<7:13:53, 271.18s/it]
[I 2025-10-23 20:43:58,739] Trial 3 finished with value:
0.9073581640940808 and parameters: {'n estimators': 1331,
'learning rate': 0.012027242419088454, 'max depth': 11, 'subsample':
0.7847110448499018, 'colsample bytree': 0.43072179664762617,
'min child weight': 1, 'reg alpha': 0.30101433697764957, 'reg lambda':
15.712342613832865, 'gamma': 0.539626119353088}. Best is trial 0 with
value: 0.9084311899688804.
[21:31<6:30:47, 246.81s/it]
[I 2025-10-23 20:47:22,384] Trial 4 finished with value:
0.9072723225767542 and parameters: {'n estimators': 1270,
'learning_rate': 0.009629426722965774, 'max_depth': 13, 'subsample':
0.7402322839655144, 'colsample bytree': 0.4219865304278702,
'min_child_weight': 2, 'reg_alpha': 0.0730384934981502, 'reg_lambda': 14.153803817255655, 'gamma': 0.36644201693417633}. Best is trial 0
with value: 0.9084311899688804.
[24:26<5:48:42, 222.58s/it]
[I 2025-10-23 20:50:17,923] Trial 5 finished with value:
0.907122151733068 and parameters: {'n estimators': 1375,
'learning_rate': 0.010601744554806291, 'max_depth': 12, 'subsample': 0.7685463499018672, 'colsample_bytree': 0.4333035635619345,
'min child weight': 3, 'reg alpha': 0.23000200194869183, 'reg lambda':
14.968789273686847, 'gamma': 0.5668601612233419}. Best is trial 0 with
value: 0.9084311899688804.
[27:52<5:36:39, 217.20s/it]
[I 2025-10-23 20:53:44.039] Trial 6 finished with value:
0.9081522982985129 and parameters: {'n estimators': 1356,
'learning rate': 0.013946691700207918, 'max depth': 11, 'subsample':
0.6753526038774206, 'colsample_bytree': 0.40814819359741145, 
'min_child_weight': 1, 'reg_alpha': 0.1264226825312147, 'reg_lambda':
4.333991951476359, 'gamma': 0.5460352918912674}. Best is trial 0 with
value: 0.9084311899688804.
[31:59<5:47:33, 226.67s/it]
[I 2025-10-23 20:57:50,983] Trial 7 finished with value:
0.9080234185828165 and parameters: {'n estimators': 1285,
'learning rate': 0.010456112653721188, 'max depth': 12, 'subsample':
0.660722664993704, 'colsample bytree': 0.5445246102496286,
```

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'min_child_weight': 1, 'reg_alpha': 0.31926008395124267, 'reg_lambda':
10.975831968122215, 'gamma': 0.3475157920220088}. Best is trial 0 with
value: 0.9084311899688804.
Best trial: 0. Best value: 0.908431: 9%| | 9/100
[35:35<5:38:23, 223.12s/it]
[I 2025-10-23 21:01:26,292] Trial 8 finished with value:
0.9084311347031362 and parameters: {'n_estimators': 1182,
'learning_rate': 0.013295392473336972, 'max_depth': 13, 'subsample':
0.8169854770603118, 'colsample bytree': 0.5389528369292866.
'min child weight': 1, 'reg alpha': 0.1206442235692169, 'reg lambda':
3.2480658342780093, 'gamma': 0.5568639716313439}. Best is trial 0 with
value: 0.9084311899688804.
[39:18<5:34:46, 223.18s/it]
[I 2025-10-23 21:05:09,629] Trial 9 finished with value:
0.9073152433354175 and parameters: {'n estimators': 1363,
'learning_rate': 0.010693560092632599, 'max depth': 11, 'subsample':
0.7059097542040702, 'colsample bytree': 0.4585853524796608,
'min child weight': 3, 'reg alpha': 0.18586891186976448, 'reg lambda':
13.585074258111558, 'gamma': 0.4336952395895751}. Best is trial 0 with
value: 0.9084311899688804.
Best trial: 10. Best value: 0.908539: 11%| | 11/100
[43:23<5:41:03, 229.93s/it]
[I 2025-10-23 21:09:14,846] Trial 10 finished with value:
0.9085385022278658 and parameters: {'n estimators': 1457,
'learning_rate': 0.012534348968406635, 'max_depth': 13, 'subsample':
0.8710009875752958, 'colsample bytree': 0.49392513787719616,
'min_child_weight': 2, 'reg_alpha': 0.07089396483154492, 'reg_lambda':
8.327237314152747, 'gamma': 0.4646798898954994}. Best is trial 10 with
value: 0.9085385022278658.
[47:40<5:49:23, 238.22s/it]
[I 2025-10-23 21:13:32,048] Trial 11 finished with value:
0.908238077641877 and parameters: {'n_estimators': 1472,
'learning rate': 0.012632951325921798, 'max depth': 13, 'subsample':
0.8787196084573634, 'colsample_bytree': 0.501053512305816, 
'min_child_weight': 2, 'reg_alpha': 0.0704428896299714, 'reg_lambda':
7.588383231847572, 'gamma': 0.4713397586713802}. Best is trial 10 with
value: 0.9085385022278658.
Best trial: 12. Best value: 0.90886: 13%| ■ | 13/100
[51:52<5:51:05, 242.13s/it]
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[I 2025-10-23 21:17:43,177] Trial 12 finished with value:
0.9088604113719493 and parameters: {'n estimators': 1455,
'learning_rate': 0.014232230486412676, 'max_depth': 13, 'subsample': 0.888193160538898, 'colsample_bytree': 0.48445935290771,
'min child weight': 2, 'reg_alpha': 0.09560946226529322, 'reg_lambda':
7.762268378604656, 'gamma': 0.4916424835229907}. Best is trial 12 with
value: 0.9088604113719493.
Best trial: 12. Best value: 0.90886: 14%
[56:17<5:57:01, 249.08s/it]
[I 2025-10-23 21:22:08,319] Trial 13 finished with value:
0.908645814486851 and parameters: {'n estimators': 1472,
'learning rate': 0.012868511858213439, 'max depth': 13, 'subsample':
0.8888230900011032, 'colsample_bytree': 0.49726531044385175, 
'min_child_weight': 2, 'reg_alpha': 0.1006652099679006, 'reg_lambda':
7.593695610567005, 'gamma': 0.4083971321714596}. Best is trial 12 with
value: 0.9088604113719493.
[59:12<5:21:24, 226.87s/it]
[I 2025-10-23 21:25:03,713] Trial 14 finished with value:
0.907830340796903 and parameters: {'n estimators': 1422,
'learning_rate': 0.013341203769949976, 'max_depth': 12, 'subsample':
0.832003600794303, 'colsample_bytree': 0.5792136575151189, 
'min_child_weight': 3, 'reg_alpha': 0.10219761037031092, 'reg_lambda':
5.73566660637862, 'gamma': 0.41416301705437225}. Best is trial 12 with
value: 0.9088604113719493.
Best trial: 12. Best value: 0.90886: 16% | ■ | 16/100
[1:01:53<4:49:39, 206.90s/it]
[I 2025-10-23 21:27:44,250] Trial 15 finished with value:
0.9084097399518761 and parameters: {'n estimators': 1428,
'learning rate': 0.014434746878564834, 'max depth': 12, 'subsample':
0.8751939724241603, 'colsample_bytree': 0.4745402739339987,
'min child weight': 2, 'reg alpha': 0.09998494611716283, 'reg lambda':
9.051196660241171, 'gamma': 0.5091474539122197}. Best is trial 12 with
value: 0.9088604113719493.
[1:05:30<4:50:33, 210.04s/it]
[I 2025-10-23 21:31:21,577] Trial 16 finished with value:
0.9084956021938566 and parameters: {'n estimators': 1422,
'learning rate': 0.011561029166370527, 'max depth': 13, 'subsample':
0.8099084619714647, 'colsample bytree': 0.5190142207277009,
'min_child_weight': 2, 'reg_alpha': 0.15907152737621752, 'reg_lambda': 6.249712060954249, 'gamma': 0.2890533625439003}. Best is trial 12 with
value: 0.9088604113719493.
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Best trial: 12. Best value: 0.90886: 18%
[1:07:46<4:16:49, 187.93s/it]
[I 2025-10-23 21:33:38,018] Trial 17 finished with value:
0.9084312037853166 and parameters: {'n estimators': 1243,
'learning rate': 0.01326348329793426, 'max depth': 13, 'subsample':
0.8495641420107123, 'colsample bytree': 0.47611832860422315,
'min child weight': 2, 'reg alpha': 0.08928078441964953, 'reg lambda':
4.857534244905753, 'gamma': 0.5980102651941028}. Best is trial 12 with
value: 0.9088604113719493.
Best trial: 12. Best value: 0.90886: 19%| | 19/100
[1:10:39<4:07:19, 183.20s/it]
[I 2025-10-23 21:36:30,219] Trial 18 finished with value:
0.9078304236955196 and parameters: {'n estimators': 1453,
'learning_rate': 0.012446319049333247, 'max_depth': 13, 'subsample':
0.6240488141776346, 'colsample bytree': 0.5158564301492268,
'min_child_weight': 3, 'reg_alpha': 0.11436981218597324, 'reg_lambda': 7.265052385458937, 'gamma': 0.3973800102493195}. Best is trial 12 with
value: 0.9088604113719493.
Best trial: 12. Best value: 0.90886: 20%
[1:13:04<3:49:10, 171.89s/it]
[I 2025-10-23 21:38:55,741] Trial 19 finished with value:
0.9087746043957129 and parameters: {'n estimators': 1404,
'learning_rate': 0.013578265953954885, 'max_depth': 12, 'subsample': 0.8872531689406447, 'colsample_bytree': 0.4739057215541127,
'min child weight': 3, 'reg alpha': 0.1706046031422878, 'reg lambda':
3.588381127915649, 'gamma': 0.5072376233768641}. Best is trial 12 with
value: 0.9088604113719493.
Best trial: 12. Best value: 0.90886: 21%| | 21/100
[1:15:42<3:40:38, 167.58s/it]
[I 2025-10-23 21:41:33,275] Trial 20 finished with value:
0.9084097952176203 and parameters: {'n_estimators': 1400,
'learning rate': 0.013744475998946211, 'max depth': 12, 'subsample':
0.8524441353624777, 'colsample bytree': 0.47281544693194133,
'min_child_weight': 3, 'reg_alpha': 0.20413700523536582, 'reg_lambda':
2.676928562651908, 'gamma': 0.5056329986613246}. Best is trial 12 with
value: 0.9088604113719493.
Best trial: 12. Best value: 0.90886: 22%
[1:18:13<3:31:26, 162.65s/it]
[I 2025-10-23 21:44:04,432] Trial 21 finished with value:
0.9082809776758861 and parameters: {'n estimators': 1438,
'learning rate': 0.012992889070227475, 'max depth': 12, 'subsample':
0.885760423650374, 'colsample bytree': 0.5098866805151898,
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'min_child_weight': 2, 'reg_alpha': 0.16055545555734074, 'reg_lambda':
9.31034157086877, 'gamma': 0.5073160042224985}. Best is trial 12 with
value: 0.9088604113719493.
Best trial: 12. Best value: 0.90886: 23%| | 23/100
[1:20:50<3:26:45, 161.12s/it]
[I 2025-10-23 21:46:41,968] Trial 22 finished with value:
0.9085385713100462 and parameters: {'n_estimators': 1403,
'learning_rate': 0.013630581428086904, 'max_depth': 13, 'subsample':
0.887487034639951, 'colsample_bytree': 0.48037869751852685,
'min child weight': 2, 'reg alpha': 0.13907475885853762, 'reg lambda':
3.51080170968229, 'gamma': 0.3981284459153621}. Best is trial 12 with
value: 0.9088604113719493.
Best trial: 12. Best value: 0.90886: 24%
[1:23:18<3:18:58, 157.09s/it]
[I 2025-10-23 21:49:09,656] Trial 23 finished with value:
0.9086672575956374 and parameters: {'n estimators': 1470,
'learning_rate': 0.011863085981238666, 'max depth': 12, 'subsample':
0.8148870685374686, 'colsample_bytree': 0.5335416834780862, 
'min_child_weight': 3, 'reg_alpha': 0.0855746633052795, 'reg_lambda':
6.163449314594693, 'gamma': 0.4908656873642666}. Best is trial 12 with
value: 0.9088604113719493.
Best trial: 12. Best value: 0.90886: 25%
[1:25:55<3:16:11, 156.95s/it]
[I 2025-10-23 21:51:46,245] Trial 24 finished with value:
0.9083239053427674 and parameters: {'n estimators': 1447,
'learning_rate': 0.011181335487315947, 'max_depth': 12, 'subsample':
0.8093411392704601, 'colsample bytree': 0.5359094506328079,
'min child weight': 3, 'reg alpha': 0.08492410557910744, 'reg lambda':
5.507498077946424, 'gamma': 0.49338641863549787}. Best is trial 12
with value: 0.9088604113719493.
Best trial: 12. Best value: 0.90886: 26%
[1:28:15<3:07:20, 151.90s/it]
[I 2025-10-23 21:54:06,398] Trial 25 finished with value:
0.9082166897988347 and parameters: {'n_estimators': 1415,
'learning_rate': 0.012004196415642368, 'max_depth': 12, 'subsample':
0.7412343847498281, 'colsample_bytree': 0.5666717095546758,
'min_child_weight': 3, 'reg_alpha': 0.22680770924455074, 'reg_lambda':
3.622436645516942, 'gamma': 0.522889901189784}. Best is trial 12 with
value: 0.9088604113719493.
Best trial: 12. Best value: 0.90886: 27%
[1:30:38<3:01:47, 149.42s/it]
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[I 2025-10-23 21:56:30,041] Trial 26 finished with value:
0.9084312383264069 and parameters: {'n estimators': 1346,
'learning_rate': 0.014313172677245519, 'max_depth': 11, 'subsample': 0.8546105358745213, 'colsample_bytree': 0.4588060417713068,
'min child weight': 3, 'reg alpha': 0.18491724486069708, 'reg lambda':
2.745872771154202, 'gamma': 0.5804785925467719}. Best is trial 12 with
value: 0.9088604113719493.
Best trial: 12. Best value: 0.90886: 28% | | 28/100
[1:32:58<2:55:55, 146.60s/it]
[I 2025-10-23 21:58:50,054] Trial 27 finished with value:
0.9082595345670995 and parameters: {'n estimators': 1472,
'learning rate': 0.012064289705957363, 'max depth': 12, 'subsample':
0.8302780764498894, 'colsample_bytree': 0.5289571919755939, 
'min_child_weight': 3, 'reg_alpha': 0.1111435008413371, 'reg_lambda':
4.030329535711683, 'gamma': 0.48320946328304987}. Best is trial 12
with value: 0.9088604113719493.
Best trial: 12. Best value: 0.90886: 29%
[1:35:13<2:49:14, 143.02s/it]
[I 2025-10-23 22:01:04,732] Trial 28 finished with value:
0.9081951776078678 and parameters: {'n estimators': 1389,
'learning_rate': 0.011251465318119023, 'max_depth': 12, 'subsample':
0.8627753358353705, 'colsample bytree': 0.5502397543508873,
'min_child_weight': 3, 'reg_alpha': 0.08405610391452331, 'reg_lambda':
5.355032319895842, 'gamma': 0.528085950832093}. Best is trial 12 with
value: 0.9088604113719493.
Best trial: 12. Best value: 0.90886: 30% | | 30/100
[1:37:42<2:49:01, 144.88s/it]
[I 2025-10-23 22:03:33,951] Trial 29 finished with value:
0.9085170453026432 and parameters: {'n_estimators': 1443,
'learning rate': 0.009933716628593213, 'max depth': 11, 'subsample':
0.7934155810442843, 'colsample bytree': 0.48943167346008115,
'min child weight': 3, 'reg alpha': 0.08098317107987935, 'reg lambda':
6.2827562284119445, 'gamma': 0.4413501115755767}. Best is trial 12
with value: 0.9088604113719493.
Best trial: 12. Best value: 0.90886: 31%| | 31/100
[1:39:44<2:38:30, 137.83s/it]
[I 2025-10-23 22:05:35,339] Trial 30 finished with value:
0.9080664153318782 and parameters: {'n estimators': 1405,
'learning_rate': 0.01405980073567287, 'max_depth': 12, 'subsample':
0.7579544541729204, 'colsample bytree': 0.5565209781459456.
'min_child_weight': 3, 'reg_alpha': 0.17340480920661103, 'reg_lambda':
3.0921363506094486, 'qamma': 0.47176995289846396}. Best is trial 12
with value: 0.9088604113719493.
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Best trial: 12. Best value: 0.90886: 32%| | 32/100
[1:42:07<2:38:01, 139.44s/it]
[I 2025-10-23 22:07:58,522] Trial 31 finished with value:
0.9080878446242284 and parameters: {'n estimators': 1473,
'learning rate': 0.012774014877808477, 'max_depth': 13, 'subsample':
0.8378028119867051, 'colsample_bytree': 0.5030338690544467,
'min child weight': 2, 'reg alpha': 0.09827487380027798, 'reg lambda':
7.283459848721934, 'gamma': 0.4164471263371983}. Best is trial 12 with
value: 0.9088604113719493.
Best trial: 12. Best value: 0.90886: 33%
[1:44:19<2:33:21, 137.34s/it]
[I 2025-10-23 22:10:10,966] Trial 32 finished with value:
0.908302448417545 and parameters: {'n estimators': 1459,
'learning_rate': 0.01364829717091467, 'max_depth': 13, 'subsample':
0.8642224000728843, 'colsample bytree': 0.5258876378056191,
'min_child_weight': 2, 'reg_alpha': 0.10658608566307844, 'reg_lambda': 11.469221163839803, 'gamma': 0.4516258048704246}. Best is trial 12
with value: 0.9088604113719493.
Best trial: 12. Best value: 0.90886: 34%
[1:46:27<2:28:01, 134.57s/it]
[I 2025-10-23 22:12:19,066] Trial 33 finished with value:
0.9085814575276193 and parameters: {'n estimators': 1437,
'learning_rate': 0.01293975259805337, 'max_depth': 13, 'subsample':
0.8280113596238537, 'colsample_bytree': 0.4885536608447348,
'min child weight': 2, 'reg alpha': 0.1294224271789106, 'reg lambda':
9.09\overline{7}10586\overline{6}878215, 'gamma': 0.4891889074313301}. Best is trial 12 with
value: 0.9088604113719493.
Best trial: 12. Best value: 0.90886: 35%
[1:48:44<2:26:26, 135.18s/it]
[I 2025-10-23 22:14:35,662] Trial 34 finished with value:
0.908645800670415 and parameters: {'n estimators': 1456,
'learning rate': 0.012347592079386268, 'max depth': 12, 'subsample':
0.8863398311146401, 'colsample_bytree': 0.46690339828569105, 
'min_child_weight': 2, 'reg_alpha': 0.0932654259588829, 'reg_lambda':
8.06\overline{1642669131682}, 'gamma': 0.45643445033579655\). Best is trial 12
with value: 0.9088604113719493.
Best trial: 35. Best value: 0.908925: 36%
[1:51:07<2:26:41, 137.52s/it]
[I 2025-10-23 22:16:58,655] Trial 35 finished with value:
0.9089248097804894 and parameters: {'n estimators': 1382,
'learning_rate': 0.011734625785894707, 'max_depth': 12, 'subsample':
0.7931303088723458, 'colsample bytree': 0.4434522811997975,
```

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'min_child_weight': 1, 'reg_alpha': 0.14155673913516692, 'reg_lambda':
6.7056579332850985, 'gamma': 0.37140078584173314}. Best is trial 35
with value: 0.9089248097804894.
Best trial: 36. Best value: 0.909268: 37%
[1:53:29<2:25:44, 138.80s/it]
[I 2025-10-23 22:19:20,453] Trial 36 finished with value:
0.9092682103908858 and parameters: {'n_estimators': 1333,
'learning_rate': 0.011633586166366573, 'max depth': 12, 'subsample':
0.7938389735094136, 'colsample bytree': 0.4410016887165864.
'min child weight': 1, 'reg alpha': 0.259749812869647, 'reg lambda':
4.979243873230984, 'gamma': 0.35896863319759553}. Best is trial 36
with value: 0.9092682103908858.
Best trial: 36. Best value: 0.909268: 38%
[1:56:05<2:28:46, 143.97s/it]
[I 2025-10-23 22:21:56,488] Trial 37 finished with value:
0.9085385920347003 and parameters: {'n estimators': 1320,
'learning_rate': 0.010891085480812572, 'max depth': 12, 'subsample':
0.7895442881142876, 'colsample_bytree': 0.44419161002537844, 
'min_child_weight': 1, 'reg_alpha': 0.254627926441613, 'reg_lambda':
4.803070732765152, 'gamma': 0.32413461888192335}. Best is trial 36
with value: 0.9092682103908858.
Best trial: 36. Best value: 0.909268: 39%
[1:58:25<2:25:03, 142.68s/it]
[I 2025-10-23 22:24:16,165] Trial 38 finished with value:
0.9085385920347002 and parameters: {'n estimators': 1381,
'learning_rate': 0.011617233892698387, 'max_depth': 11, 'subsample':
0.7211552472750795, 'colsample bytree': 0.42296273652836364,
'min_child_weight': 1, 'reg_alpha': 0.2761770417655088, 'reg_lambda':
4.047912885715004, 'gamma': 0.3738524569507248}. Best is trial 36 with
value: 0.9092682103908858.
Best trial: 36. Best value: 0.909268: 40%
[2:00:49<2:23:13, 143.22s/it]
[I 2025-10-23 22:26:40,631] Trial 39 finished with value:
0.9081308137404182 and parameters: {'n_estimators': 1337,
'learning_rate': 0.010192777907940107, 'max depth': 12, 'subsample':
0.7641387745733491, 'colsample_bytree': 0.4432645178613215,
'min_child_weight': 1, 'reg_alpha': 0.14084083315609033, 'reg lambda':
6.748971384387662, 'gamma': 0.3408748023332511}. Best is trial 36 with
value: 0.9092682103908858.
Best trial: 36. Best value: 0.909268: 41%
[2:03:03<2:18:06, 140.44s/it]
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[I 2025-10-23 22:28:54,594] Trial 40 finished with value:
0.9083239467920756 and parameters: {'n estimators': 1302,
'learning_rate': 0.011183322146525836, 'max_depth': 12, 'subsample': 0.7146582630504785, 'colsample_bytree': 0.41108325561806397,
'min child weight': 1, 'reg alpha': 0.231848325574267, 'reg lambda':
5.07\overline{2}82942\overline{0}100\overline{2}34, 'gamma': 0.3775811972427718}. Best is trial 36 with
value: 0.9092682103908858.
Best trial: 36. Best value: 0.909268: 42%
[2:06:09<2:28:55, 154.06s/it]
[I 2025-10-23 22:32:00,426] Trial 41 finished with value:
0.9084741798097242 and parameters: {'n estimators': 1368,
'learning rate': 0.011787234433182319, 'max depth': 12, 'subsample':
0.7753147040303301, 'colsample_bytree': 0.4516904729212446, 
'min_child_weight': 1, 'reg_alpha': 0.07747842484692355, 'reg_lambda':
6.11\overline{540733897031}, 'gamma': 0.32354236872085257}. Best is trial 36 with
value: 0.9092682103908858.
Best trial: 36. Best value: 0.909268: 43%
[2:10:56<3:04:18, 194.01s/it]
[I 2025-10-23 22:36:47,630] Trial 42 finished with value:
0.9085814575276193 and parameters: {'n estimators': 1262,
'learning_rate': 0.011919958856874063, 'max_depth': 12, 'subsample':
0.7856498419724799, 'colsample bytree': 0.46157835729520913,
'min_child_weight': 1, 'reg_alpha': 0.20044658691442127, 'reg_lambda':
4.27\overline{2}57977\overline{4}154031, 'gamma': 0.5380561296920076}. Best is trial 36 with
value: 0.9092682103908858.
Best trial: 36. Best value: 0.909268: 44%
[2:17:21<3:54:36, 251.36s/it]
[I 2025-10-23 22:43:12,838] Trial 43 finished with value:
0.9086887766948224 and parameters: {'n estimators': 1352,
'learning rate': 0.012171122189237141, 'max depth': 12, 'subsample':
0.7972715888574476, 'colsample_bytree': 0.432601228853531,
'min child weight': 1, 'reg alpha': 0.27374342507158106, 'reg lambda':
6.604299990539631, 'gamma': 0.4280698286550289}. Best is trial 36 with
value: 0.9092682103908858.
Best trial: 36. Best value: 0.909268: 45%
[2:21:58<3:57:30, 259.09s/it]
[I 2025-10-23 22:47:49,973] Trial 44 finished with value:
0.9078302578982865 and parameters: {'n estimators': 1350,
'learning rate': 0.009294173983554498, 'max depth': 12, 'subsample':
0.7991504978004298, 'colsample bytree': 0.42966045522586893,
'min_child_weight': 1, 'reg_alpha': 0.2884858322991756, 'reg_lambda':
10.266881672859906, 'qamma': 0.3840596165682757}. Best is trial 36
with value: 0.9092682103908858.
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Best trial: 36. Best value: 0.909268: 46%
[2:24:06<3:17:49, 219.80s/it]
[I 2025-10-23 22:49:58,085] Trial 45 finished with value:
0.9078732615555664 and parameters: {'n estimators': 1325,
'learning rate': 0.012216564453606778, 'max depth': 11, 'subsample':
0.804209776962602, 'colsample_bytree': 0.40071628629096323,
'min child weight': 1, 'reg alpha': 0.3196622445249398, 'reg lambda':
6.855987861938438, 'gamma': 0.4271433055962178}. Best is trial 36 with
value: 0.9092682103908858.
Best trial: 36. Best value: 0.909268: 47%
[2:26:33<2:54:41, 197.77s/it]
[I 2025-10-23 22:52:24,437] Trial 46 finished with value:
0.9079805461816794 and parameters: {'n estimators': 1386,
'learning_rate': 0.011321498250238406, 'max_depth': 12, 'subsample':
0.7377948629139162, 'colsample bytree': 0.4402685510656001,
'min_child_weight': 1, 'reg_alpha': 0.25569397326034066, 'reg_lambda': 8.217245317045462, 'gamma': 0.36663332981723373}. Best is trial 36
with value: 0.9092682103908858.
Best trial: 36. Best value: 0.909268: 48%
[2:29:28<2:45:30, 190.98s/it]
[I 2025-10-23 22:55:19,577] Trial 47 finished with value:
0.9086458835690314 and parameters: {'n estimators': 1367,
'learning_rate': 0.010959099244425643, 'max_depth': 12, 'subsample':
0.7729095581137353, 'colsample bytree': 0.4145324025148479,
'min child weight': 1, 'reg_alpha': 0.25797335383509445, 'reg_lambda':
4.449069705597836, 'gamma': 0.35431597469287884}. Best is trial 36
with value: 0.9092682103908858.
Best trial: 36. Best value: 0.909268: 49%| 49/100
[2:34:56<3:17:12, 232.02s/it]
[I 2025-10-23 23:00:47,356] Trial 48 finished with value:
0.9087102405282631 and parameters: {'n estimators': 1310,
'learning rate': 0.013884910627156646, 'max depth': 12, 'subsample':
0.8185792944203133, 'colsample bytree': 0.4476929712790958,
'min child weight': 1, 'reg alpha': 0.2163001546449248, 'reg lambda':
5.5161320197022325, 'gamma': 0.29475907963180553}. Best is trial 36
with value: 0.9092682103908858.
Best trial: 36. Best value: 0.909268: 50%
[2:39:03<3:17:16, 236.73s/it]
[I 2025-10-23 23:04:55,069] Trial 49 finished with value:
0.908774597487495 and parameters: {'n estimators': 1310,
'learning_rate': 0.013982260708899253, 'max_depth': 12, 'subsample':
0.8247822440113828, 'colsample bytree': 0.45000611189381245,
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'min child weight': 1, 'reg alpha': 0.20894788131180597, 'reg lambda':
5.183377545848782, 'gamma': 0.2870019067448188}. Best is trial 36 with
value: 0.9092682103908858.
Best trial: 36. Best value: 0.909268: 51%| 51/100
[2:43:00<3:13:18, 236.71s/it]
[I 2025-10-23 23:08:51,741] Trial 50 finished with value:
0.9088604528212575 and parameters: {'n_estimators': 1295,
'learning_rate': 0.013320997762023674, 'max_depth': 11, 'subsample':
0.8582542766579662, 'colsample_bytree': 0.45150166027426963,
'min child weight': 1, 'reg alpha': 0.1756076184362804, 'reg lambda':
3.711627768856635, 'gamma': 0.31859785871880947}. Best is trial 36
with value: 0.9092682103908858.
Best trial: 36. Best value: 0.909268: 52%
[2:47:22<3:15:22, 244.21s/it]
[I 2025-10-23 23:13:13,443] Trial 51 finished with value:
0.9088175251543762 and parameters: {'n estimators': 1284,
'learning_rate': 0.013362123386435932, 'max depth': 11, 'subsample':
0.8676862696273725, 'colsample_bytree': 0.4544250775634968, 
'min_child_weight': 1, 'reg_alpha': 0.17263937969468768, 'reg_lambda':
3.933171363025703, 'gamma': 0.31302558449983725}. Best is trial 36
with value: 0.9092682103908858.
Best trial: 36. Best value: 0.909268: 53%
[2:51:13<3:08:12, 240.26s/it]
[I 2025-10-23 23:17:04,494] Trial 52 finished with value:
0.9084956021938566 and parameters: {'n estimators': 1274,
'learning_rate': 0.01344005516009276, 'max_depth': 11, 'subsample':
0.8688374209415851, 'colsample bytree': 0.46704828544096416,
'min_child_weight': 1, 'reg_alpha': 0.17001883810462878, 'reg_lambda':
3.7726708624933436, 'gamma': 0.3152512781215923}. Best is trial 36
with value: 0.9092682103908858.
Best trial: 36. Best value: 0.909268: 54%
[2:55:09<3:03:15, 239.03s/it]
[I 2025-10-23 23:21:00,657] Trial 53 finished with value:
0.908302448417545 and parameters: {'n_estimators': 1247,
'learning_rate': 0.013111889388278607, 'max_depth': 11, 'subsample':
0.8428121774861383, 'colsample bytree': 0.48081034023823715,
'min_child_weight': 1, 'reg_alpha': 0.1514754167858945, 'reg_lambda':
3.0275398209586637, 'gamma': 0.3066326690040963}. Best is trial 36
with value: 0.9092682103908858.
Best trial: 36. Best value: 0.909268: 55%| | 55/100
[2:59:07<2:59:05, 238.78s/it]
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[I 2025-10-23 23:24:58,854] Trial 54 finished with value:
0.9085814644358372 and parameters: {'n estimators': 1286,
'learning_rate': 0.01418983659568801, 'max_depth': 11, 'subsample':
0.8765724931432579, 'colsample bytree': 0.4349981339691219,
'min child weight': 1, 'reg alpha': 0.1884207802803456, 'reg lambda':
3.406159519050959, 'gamma': 0.34339713595926913}. Best is trial 36
with value: 0.9092682103908858.
Best trial: 36. Best value: 0.909268: 56%
[3:03:06<2:55:07, 238.81s/it]
[I 2025-10-23 23:28:57,724] Trial 55 finished with value:
0.9085814022618749 and parameters: {'n estimators': 1211,
'learning_rate': 0.013469315531986442, 'max depth': 11, 'subsample':
0.86286920726447, 'colsample bytree': 0.4661277471301234,
'min child weight': 1, 'reg alpha': 0.126446115329956, 'reg lambda':
3.9473455437996754, 'gamma': 0.33144080106693097}. Best is trial 36
with value: 0.9092682103908858.
Best trial: 36. Best value: 0.909268: 57%
[3:07:26<2:55:39, 245.11s/it]
[I 2025-10-23 23:33:17,526] Trial 56 finished with value:
0.9085170522108612 and parameters: {'n estimators': 1336,
'learning_rate': 0.012656673042716227, 'max_depth': 11, 'subsample':
0.8558022123819448, 'colsample bytree': 0.45622937191789237,
'min_child_weight': 1, 'reg_alpha': 0.17102266896542048, 'reg_lambda':
3.270256000074742, 'gamma': 0.3035225114664474}. Best is trial 36 with
value: 0.9092682103908858.
Best trial: 36. Best value: 0.909268: 58%
[3:11:13<2:47:47, 239.69s/it]
[I 2025-10-23 23:37:04,590] Trial 57 finished with value:
0.9090321358559106 and parameters: {'n estimators': 1294,
'learning rate': 0.013193446786908553, 'max depth': 11, 'subsample':
0.8783097\overline{2}56709833, 'colsample_bytree': 0.4\overline{1}813354184411144,
'min child weight': 1, 'reg alpha': 0.1442029156176727, 'reg lambda':
4.659513128884289, 'gamma': 0.35790072624798974}. Best is trial 36
with value: 0.9092682103908858.
Best trial: 36. Best value: 0.909268: 59%| | 59/100
[3:15:17<2:44:46, 241.14s/it]
[I 2025-10-23 23:41:09,093] Trial 58 finished with value:
0.9087745698546228 and parameters: {'n estimators': 1297,
'learning_rate': 0.013165519534748487, 'max_depth': 11, 'subsample':
0.8758440607728813, 'colsample bytree': 0.4222006804712295,
'min_child_weight': 1, 'reg_alpha': 0.13888390967744732, 'reg_lambda':
4.5712051672136385, 'gamma': 0.357902113416277}. Best is trial 36 with
value: 0.9092682103908858.
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Best trial: 36. Best value: 0.909268: 60%
[3:19:37<2:44:22, 246.57s/it]
[I 2025-10-23 23:45:28,330] Trial 59 finished with value:
0.9089892081890293 and parameters: {'n estimators': 1273,
'learning_rate': 0.01378766416325947, 'max_depth': 11, 'subsample':
0.8419158065051857, 'colsample bytree': 0.42526788915420577,
'min_child_weight': 1, 'reg_alpha': 0.15257090796338094, 'reg_lambda':
5.844754673105757, 'gamma': 0.33450071583688007}. Best is trial 36
with value: 0.9092682103908858.
Best trial: 36. Best value: 0.909268: 61%
[3:22:55<2:30:53, 232.14s/it]
[I 2025-10-23 23:48:46,809] Trial 60 finished with value:
0.9089892012808113 and parameters: {'n estimators': 1268,
'learning_rate': 0.014363365465822921, 'max_depth': 11, 'subsample':
0.8522410420976122, 'colsample_bytree': 0.41713354477915277,
'min_child_weight': 1, 'reg_alpha': 0.11667957637699129, 'reg_lambda': 5.692682276015727, 'gamma': 0.39156508440642296}. Best is trial 36
with value: 0.9092682103908858.
Best trial: 36. Best value: 0.909268: 62%
[3:26:50<2:27:33, 232.99s/it]
[I 2025-10-23 23:52:41,796] Trial 61 finished with value:
0.9089463357878923 and parameters: {'n estimators': 1255,
'learning_rate': 0.014419697492031572, 'max_depth': 11, 'subsample':
0.8399539139844312, 'colsample_bytree': 0.4187355669530909,
'min child weight': 1, 'reg alpha': 0.11707076333532751, 'reg lambda':
5.65\overline{8}56990\overline{7}574986, 'gamma': 0.3905283966743046}. Best is trial 36 with
value: 0.9092682103908858.
Best trial: 36. Best value: 0.909268: 63%| | 63/100
[3:30:52<2:25:17, 235.62s/it]
[I 2025-10-23 23:56:43.531] Trial 62 finished with value:
0.9086887490619503 and parameters: {'n estimators': 1254,
'learning rate': 0.013821637553156624, 'max depth': 11, 'subsample':
0.8413980049374119, 'colsample_bytree': 0.41846291832072247,
'min_child_weight': 1, 'reg_alpha': 0.11712251146082252, 'reg_lambda':
5.670454902892824, 'gamma': 0.3927060378334095}. Best is trial 36 with
value: 0.9092682103908858.
Best trial: 36. Best value: 0.909268: 64%
[3:34:14<2:15:17, 225.48s/it]
[I 2025-10-24 00:00:05,361] Trial 63 finished with value:
0.9083453691762083 and parameters: {'n estimators': 1230,
'learning_rate': 0.014395812645605001, 'max_depth': 11, 'subsample':
0.8482487608472795, 'colsample bytree': 0.40525716349620516,
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'min child weight': 1, 'reg alpha': 0.13299106618714487, 'reg lambda':
5.849844803977218, 'gamma': 0.3638054637367358}. Best is trial 36 with
value: 0.9092682103908858.
Best trial: 36. Best value: 0.909268: 65%| 65%| 65/100
[3:37:53<2:10:28, 223.67s/it]
[I 2025-10-24 00:03:44,818] Trial 64 finished with value:
0.9086029420857141 and parameters: {'n_estimators': 1276,
'learning rate': 0.014075576920871, 'max depth': 11, 'subsample':
0.8221005965963957, 'colsample bytree': 0.4280641044421402.
'min child weight': 1, 'reg alpha': 0.1212605519636035, 'reg lambda':
4.783836559199675, 'gamma': 0.33642397184145184}. Best is trial 36
with value: 0.9092682103908858.
Best trial: 36. Best value: 0.909268: 66%
[3:41:50<2:09:00, 227.65s/it]
[I 2025-10-24 00:07:41,748] Trial 65 finished with value:
0.9089033804881389 and parameters: {'n estimators': 1234,
'learning_rate': 0.014436565561322435, 'max depth': 11, 'subsample':
0.8374240890391097, 'colsample_bytree': 0.4257757169981658, 
'min_child_weight': 1, 'reg_alpha': 0.1494065795592453, 'reg_lambda':
5.125763469237197, 'gamma': 0.3498261843381374}. Best is trial 36 with
value: 0.9092682103908858.
Best trial: 36. Best value: 0.909268: 67%
[3:45:13<2:01:07, 220.22s/it]
[I 2025-10-24 00:11:04,634] Trial 66 finished with value:
0.9089892150972474 and parameters: {'n estimators': 1235,
'learning_rate': 0.014417344182409144, 'max_depth': 11, 'subsample':
0.8358113139105245, 'colsample bytree': 0.416664451941766,
'min_child_weight': 1, 'reg_alpha': 0.1534997627677792, 'reg_lambda':
4.990189761489708, 'gamma': 0.3512366756756833}. Best is trial 36 with
value: 0.9092682103908858.
Best trial: 36. Best value: 0.909268: 68%
[3:48:16<1:51:29, 209.06s/it]
[I 2025-10-24 00:14:07,647] Trial 67 finished with value:
0.9086243921027186 and parameters: {'n_estimators': 1198,
'learning_rate': 0.01419249220101124, 'max_depth': 11, 'subsample':
0.8090625916496127, 'colsample_bytree': 0.4137037024166879,
'min_child_weight': 1, 'reg_alpha': 0.1510609307007287, 'reg lambda':
6.011407639005337, 'gamma': 0.38930396962852337}. Best is trial 36
with value: 0.9092682103908858.
Best trial: 36. Best value: 0.909268: 69%| 69/100
[3:51:50<1:48:41, 210.38s/it]
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[I 2025-10-24 00:17:41,123] Trial 68 finished with value:
0.9089677236309345 and parameters: {'n estimators': 1226,
'learning_rate': 0.013679668992940007, 'max_depth': 11, 'subsample': 0.8334920415972789, 'colsample_bytree': 0.41669565281286364,
'min child weight': 1, 'reg alpha': 0.1605901526125892, 'reg lambda':
4.250639689324391, 'gamma': 0.40899180260150136}. Best is trial 36
with value: 0.9092682103908858.
Best trial: 36. Best value: 0.909268: 70%| | 70/100
[3:54:58<1:41:56, 203.88s/it]
[I 2025-10-24 00:20:49,842] Trial 69 finished with value:
0.9073153124175979 and parameters: {'n_estimators': 1219,
'learning rate': 0.00951049812978112, 'max depth': 11, 'subsample':
0.8327707332691404, 'colsample_bytree': 0.40734600414834204, 
'min_child_weight': 1, 'reg_alpha': 0.1605685413202516, 'reg_lambda':
4.229857524133561, 'gamma': 0.40322544045160097}. Best is trial 36
with value: 0.9092682103908858.
Best trial: 36. Best value: 0.909268: 71%
[3:57:06<1:27:27, 180.97s/it]
[I 2025-10-24 00:22:57,330] Trial 70 finished with value:
0.9089677374473708 and parameters: {'n estimators': 1257,
'learning_rate': 0.013816614833484042, 'max_depth': 11, 'subsample':
0.8512225674773805, 'colsample_bytree': 0.415419800487465,
'min_child_weight': 1, 'reg_alpha': 0.10858341022972987, 'reg_lambda':
4.892271334102184, 'gamma': 0.41497521560458694}. Best is trial 36
with value: 0.9092682103908858.
Best trial: 36. Best value: 0.909268: 72%
[3:59:10<1:16:34, 164.08s/it]
[I 2025-10-24 00:25:02,027] Trial 71 finished with value:
0.9090106927471241 and parameters: {'n estimators': 1265,
'learning rate': 0.013730382510104342, 'max depth': 11, 'subsample':
0.8508457834792416, 'colsample_bytree': 0.41936228545943427,
'min child weight': 1, 'reg alpha': 0.11238830347523215, 'reg lambda':
4.882463998028918, 'gamma': 0.4169063124396941}. Best is trial 36 with
value: 0.9092682103908858.
Best trial: 36. Best value: 0.909268: 73%| 73/100
[4:01:16<1:08:34, 152.39s/it]
[I 2025-10-24 00:27:07,129] Trial 72 finished with value:
0.9092037774412556 and parameters: {'n_estimators': 1236,
'learning_rate': 0.013732815110659354, 'max_depth': 11, 'subsample':
0.8523725497483511, 'colsample bytree': 0.417727429278456,
'min_child_weight': 1, 'reg_alpha': 0.10828106151694686, 'reg_lambda': 4.656581503121197, 'gamma': 0.414290078116593}. Best is trial 36 with
value: 0.9092682103908858.
```

```
Best trial: 36. Best value: 0.909268: 74%
[4:03:28<1:03:28, 146.49s/it]
[I 2025-10-24 00:29:19,859] Trial 73 finished with value:
0.9083025105915072 and parameters: {'n estimators': 1244,
'learning_rate': 0.013902237101079509, 'max_depth': 11, 'subsample':
0.8494851756428977, 'colsample bytree': 0.4005183928034034,
'min child weight': 1, 'reg alpha': 0.10724690671305374, 'reg lambda':
4.855103787506017, 'gamma': 0.4190480396124684}. Best is trial 36 with
value: 0.9092682103908858.
Best trial: 36. Best value: 0.909268: 75%| | 75/100
[4:05:39<59:06, 141.87s/it]
[I 2025-10-24 00:31:30,940] Trial 74 finished with value:
0.9091393997573695 and parameters: {'n estimators': 1265,
'learning_rate': 0.013756298204873629, 'max_depth': 11, 'subsample':
0.8587094801470007, 'colsample_bytree': 0.43741610032615025,
'min_child_weight': 1, 'reg_alpha': 0.1042830356143744, 'reg_lambda': 4.590754239839903, 'gamma': 0.37924098470479284}. Best is trial 36
with value: 0.9092682103908858.
Best trial: 36. Best value: 0.909268: 76%
[4:08:34<1:00:43, 151.83s/it]
[I 2025-10-24 00:34:26,017] Trial 75 finished with value:
0.9087316974534859 and parameters: {'n estimators': 1268,
'learning_rate': 0.014098419176695819, 'max_depth': 11, 'subsample':
0.8761839505050597, 'colsample_bytree': 0.4363186392260127,
'min child weight': 1, 'reg_alpha': 0.1033910553062228, 'reg_lambda':
4.569480026415737, 'gamma': 0.37881867502611405}. Best is trial 36
with value: 0.9092682103908858.
Best trial: 36. Best value: 0.909268: 77%
[4:10:50<56:22, 147.05s/it]
[I 2025-10-24 00:36:41,903] Trial 76 finished with value:
0.908152319023167 and parameters: {'n estimators': 1237,
'learning rate': 0.013584234440835142, 'max depth': 11, 'subsample':
0.6818203686828438, 'colsample_bytree': 0.40947343958047067, 
'min_child_weight': 1, 'reg_alpha': 0.09059513169440753, 'reg_lambda':
5.23092798671672, 'gamma': 0.4406801483843672}. Best is trial 36 with
value: 0.9092682103908858.
Best trial: 77. Best value: 0.909311: 78%
[4:13:09<52:59, 144.54s/it]
[I 2025-10-24 00:39:00,577] Trial 77 finished with value:
0.9093111311495488 and parameters: {'n estimators': 1267,
'learning_rate': 0.013034614108249017, 'max_depth': 11, 'subsample':
0.8583465135661814, 'colsample bytree': 0.4261762035998442,
```

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'min_child_weight': 1, 'reg_alpha': 0.09678563447756561, 'reg_lambda':
5.327675418687854, 'gamma': 0.3616069534480977}. Best is trial 77 with
value: 0.9093111311495488.
Best trial: 77. Best value: 0.909311: 79%| 79/100
[4:15:37<50:59, 145.68s/it]
[I 2025-10-24 00:41:28,928] Trial 78 finished with value:
0.9090965411726687 and parameters: {'n_estimators': 1278,
'learning rate': 0.01309691119651875, 'max depth': 11, 'subsample':
0.8589505848125705, 'colsample bytree': 0.4259437709596556.
'min child weight': 1, 'reg alpha': 0.09347262525963745, 'reg lambda':
5.414996037694247, 'gamma': 0.36138272076659167}. Best is trial 77
with value: 0.9093111311495488.
Best trial: 77. Best value: 0.909311: 80%| 80/100 | 80/100
[4:18:02<48:29, 145.47s/it]
[I 2025-10-24 00:43:53,921] Trial 79 finished with value:
0.908881895930044 and parameters: {'n estimators': 1202,
'learning rate': 0.012849538618257645, 'max depth': 11, 'subsample':
0.861728860640994, 'colsample_bytree': 0.4383826403425533, 
'min_child_weight': 1, 'reg_alpha': 0.10263606082643607, 'reg_lambda':
4.648532242779011, 'gamma': 0.35836600678352065}. Best is trial 77
with value: 0.9093111311495488.
Best trial: 77. Best value: 0.909311: 81%| 81/100
[4:20:23<45:38, 144.13s/it]
[I 2025-10-24 00:46:14,904] Trial 80 finished with value:
0.9091609050401186 and parameters: {'n estimators': 1283,
'learning_rate': 0.012626595333171838, 'max_depth': 11, 'subsample':
0.8706443831877445, 'colsample bytree': 0.42974951701437114,
'min_child_weight': 1, 'reg_alpha': 0.09730237399266735, 'reg_lambda':
5.3876802696492785, 'gamma': 0.3685535010343425}. Best is trial 77
with value: 0.9093111311495488.
Best trial: 77. Best value: 0.909311: 82%
[4:22:45<43:03, 143.51s/it]
[I 2025-10-24 00:48:36,983] Trial 81 finished with value:
0.9083453000940278 and parameters: {'n_estimators': 1286,
'learning_rate': 0.013060087061158413, 'max_depth': 11, 'subsample':
0.8713189300698599, 'colsample bytree': 0.43104976621407864,
'min child weight': 1, 'reg alpha': 0.09728952106774787, 'reg lambda':
16.497760537954452, 'gamma': 0.36696646150774465}. Best is trial 77
with value: 0.9093111311495488.
Best trial: 77. Best value: 0.909311: 83%| 83/100
[4:25:01<39:57, 141.03s/it]
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[I 2025-10-24 00:50:52,212] Trial 82 finished with value:
0.908667340494254 and parameters: {'n estimators': 1278,
'learning_rate': 0.01248118600985989, 'max_depth': 11, 'subsample':
0.8826733516250296, 'colsample bytree': 0.4221540866514421,
'min child weight': 1, 'reg alpha': 0.09491146315542089, 'reg lambda':
5.3994988084785644, 'gamma': 0.34970712632095907}. Best is trial 77
with value: 0.9093111311495488.
Best trial: 77. Best value: 0.909311: 84%
[4:27:45<39:28, 148.06s/it]
[I 2025-10-24 00:53:36,666] Trial 83 finished with value:
0.9088604251883852 and parameters: {'n estimators': 1264,
'learning rate': 0.012789440801670807, 'max depth': 11, 'subsample':
0.8574382910559808, 'colsample_bytree': 0.42859577861314285, 
'min_child_weight': 1, 'reg_alpha': 0.08768868327515321, 'reg_lambda':
4.379643359462329, 'gamma': 0.38134094631242543}. Best is trial 77
with value: 0.9093111311495488.
Best trial: 77. Best value: 0.909311: 85%| 85/100
[4:30:31<38:21, 153.46s/it]
[I 2025-10-24 00:56:22,741] Trial 84 finished with value:
0.9081308482815083 and parameters: {'n estimators': 1251,
'learning_rate': 0.012610773000450591, 'max_depth': 11, 'subsample':
0.8692697824870125, 'colsample bytree': 0.4039869586159632,
'min_child_weight': 1, 'reg_alpha': 0.08031624809420913, 'reg_lambda':
4.978485799772308, 'gamma': 0.3597583365100518}. Best is trial 77 with
value: 0.9093111311495488.
Best trial: 77. Best value: 0.909311: 86%
[4:33:05<35:51, 153.66s/it]
[I 2025-10-24 00:58:56,875] Trial 85 finished with value:
0.9084741798097242 and parameters: {'n estimators': 1304,
'learning rate': 0.013230796912603355, 'max depth': 11, 'subsample':
0.8824192929678571, 'colsample_bytree': 0.4110848437160135,
'min child weight': 1, 'reg alpha': 0.11105325397620858, 'reg lambda':
6.367274790408576, 'gamma': 0.36980929640328336}. Best is trial 77
with value: 0.9093111311495488.
Best trial: 77. Best value: 0.909311: 87%| 87/100
[4:35:40<33:20, 153.90s/it]
[I 2025-10-24 01:01:31,325] Trial 86 finished with value:
0.9087746458450212 and parameters: {'n estimators': 1321,
'learning rate': 0.012978691695841095, 'max depth': 11, 'subsample':
0.8457966388623135, 'colsample bytree': 0.4340364863083581,
'min_child_weight': 1, 'reg_alpha': 0.0910463949427835, 'reg_lambda': 5.349470936013335, 'gamma': 0.401562374334008}. Best is trial 77 with
value: 0.9093111311495488.
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Best trial: 77. Best value: 0.909311: 88%
[4:38:13<30:43, 153.60s/it]
[I 2025-10-24 01:04:04,233] Trial 87 finished with value:
0.9089248443215795 and parameters: {'n estimators': 1239,
'learning_rate': 0.013502938986534267, 'max_depth': 11, 'subsample':
0.865219502497379, 'colsample_bytree': 0.44627152744562115,
'min child weight': 1, 'reg alpha': 0.10008538103619978, 'reg lambda':
4.671190495766151, 'gamma': 0.3440609330028251}. Best is trial 77 with
value: 0.9093111311495488.
Best trial: 77. Best value: 0.909311: 89%
[4:40:45<28:06, 153.31s/it]
[I 2025-10-24 01:06:36,872] Trial 88 finished with value:
0.9092467258327909 and parameters: {'n estimators': 1262,
'learning_rate': 0.012335024389040841, 'max_depth': 11, 'subsample':
0.8594103049611129, 'colsample bytree': 0.42272712816144753,
'min_child_weight': 1, 'reg_alpha': 0.07429502538007034, 'reg_lambda': 4.178914267340856, 'gamma': 0.3836676254427426}. Best is trial 77 with
value: 0.9093111311495488.
Best trial: 77. Best value: 0.909311: 90%
[4:43:18<25:31, 153.13s/it]
[I 2025-10-24 01:09:09,575] Trial 89 finished with value:
0.9087316767288317 and parameters: {'n estimators': 1293,
'learning_rate': 0.012346464672298363, 'max_depth': 11, 'subsample':
0.8722588488041784, 'colsample bytree': 0.4389713191555283,
'min child weight': 1, 'reg alpha': 0.07134170836681439, 'reg_lambda':
4.149336374770273, 'gamma': 0.38287868745719406}. Best is trial 77
with value: 0.9093111311495488.
Best trial: 77. Best value: 0.909311: 91%| 91/100
[4:45:44<22:39, 151.05s/it]
[I 2025-10-24 01:11:35.771] Trial 90 finished with value:
0.9088604735459117 and parameters: {'n estimators': 1281,
'learning rate': 0.012659465117519106, 'max depth': 11, 'subsample':
0.8804813591171209, 'colsample_bytree': 0.425979533954202, 
'min_child_weight': 1, 'reg_alpha': 0.08170881381736822, 'reg_lambda':
3.837028571539937, 'gamma': 0.4268684261555109}. Best is trial 77 with
value: 0.9093111311495488.
Best trial: 77. Best value: 0.909311: 92%
[4:48:23<20:26, 153.32s/it]
[I 2025-10-24 01:14:14,401] Trial 91 finished with value:
0.9087101990789551 and parameters: {'n estimators': 1261,
'learning_rate': 0.011428670099235224, 'max_depth': 11, 'subsample':
0.8574217707062661, 'colsample bytree': 0.4203351660738926,
```

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'min_child_weight': 1, 'reg_alpha': 0.07497346092109705, 'reg_lambda':
5.067165766812146, 'gamma': 0.37394908955258693}. Best is trial 77
with value: 0.9093111311495488.
Best trial: 77. Best value: 0.909311: 93%| 93/100
[4:50:49<17:38, 151.16s/it]
[I 2025-10-24 01:16:40,504] Trial 92 finished with value:
0.9091823619653411 and parameters: {'n_estimators': 1249,
'learning_rate': 0.013116362003863222, 'max_depth': 11, 'subsample':
0.8598886165838623, 'colsample bytree': 0.43222286761383677,
'min child weight': 1, 'reg alpha': 0.09434388443596735, 'reg lambda':
4.482506297625118, 'gamma': 0.39734086322931467}. Best is trial 77
with value: 0.9093111311495488.
Best trial: 77. Best value: 0.909311: 94%
[4:53:24<15:13, 152.31s/it]
[I 2025-10-24 01:19:15,496] Trial 93 finished with value:
0.9088819028382621 and parameters: {'n estimators': 1247,
'learning_rate': 0.01318527916439002, 'max depth': 11, 'subsample':
0.8605459315504085, 'colsample bytree': 0.43102515530992813,
'min_child_weight': 1, 'reg_alpha': 0.08774740655645989, 'reg_lambda':
4.448613625298562, 'gamma': 0.38424696830953725}. Best is trial 77
with value: 0.9093111311495488.
Best trial: 77. Best value: 0.909311: 95%| 95/100
[4:55:50<12:31, 150.39s/it]
[I 2025-10-24 01:21:41,395] Trial 94 finished with value:
0.9087960336880634 and parameters: {'n estimators': 1267,
'learning_rate': 0.012778383796608972, 'max_depth': 11, 'subsample':
0.8472437169748941, 'colsample bytree': 0.42393089393478023,
'min child weight': 1, 'reg alpha': 0.0933793442861245, 'reg lambda':
4.434218428578332, 'gamma': 0.40865057974946145}. Best is trial 77
with value: 0.9093111311495488.
Best trial: 77. Best value: 0.909311: 96%
[4:58:21<10:02, 150.64s/it]
[I 2025-10-24 01:24:12,641] Trial 95 finished with value:
0.9090965135397966 and parameters: {'n_estimators': 1291,
'learning_rate': 0.013370060371908129, 'max depth': 11, 'subsample':
0.8663177898272738, 'colsample_bytree': 0.4406759751026435,
'min_child_weight': 1, 'reg_alpha': 0.12125460744587803, 'reg_lambda':
4.707128345572929, 'gamma': 0.42053225902218594}. Best is trial 77
with value: 0.9093111311495488.
Best trial: 77. Best value: 0.909311: 97%| 97/100
[5:00:56<07:35, 151.94s/it]
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[I 2025-10-24 01:26:47,620] Trial 96 finished with value:
0.9087316974534858 and parameters: {'n estimators': 1316,
'learning_rate': 0.013419452389747082, 'max_depth': 11, 'subsample': 0.8888652341168112, 'colsample_bytree': 0.4406945114948987,
'min child weight': 1, 'reg alpha': 0.12089253626405916, 'reg lambda':
4.11\overline{0}23639\overline{3}96471, 'gamma': \overline{0}.3990555592048463}. Best is trial 77 with
value: 0.9093111311495488.
Best trial: 77. Best value: 0.909311: 98%
[5:03:28<05:04, 152.03s/it]
[I 2025-10-24 01:29:19,842] Trial 97 finished with value:
0.9084956160102926 and parameters: {'n estimators': 1330,
'learning rate': 0.012955523679663503, 'max depth': 11, 'subsample':
0.8723274739102919, 'colsample_bytree': 0.4351985914549134, 
'min_child_weight': 1, 'reg_alpha': 0.10567754858278258, 'reg_lambda':
4.693222270406685, 'gamma': 0.4327611743816258}. Best is trial 77 with
value: 0.9093111311495488.
Best trial: 77. Best value: 0.909311: 99%| 99/100
[5:06:10<02:34, 154.94s/it]
[I 2025-10-24 01:32:01,589] Trial 98 finished with value:
0.9091609395812087 and parameters: {'n estimators': 1298,
'learning_rate': 0.012338515045132881, 'max_depth': 11, 'subsample':
0.8665798146289544, 'colsample bytree': 0.4416070347178628,
'min_child_weight': 1, 'reg_alpha': 0.09608257318845949, 'reg_lambda':
5.3105567098143664, 'gamma': 0.37478996338886267}. Best is trial 77
with value: 0.9093111311495488.
Best trial: 77. Best value: 0.909311: 100% | 100/100 | 100/100
[5:08:56<00:00, 185.36s/it]
[I 2025-10-24 01:34:47,171] Trial 99 finished with value:
0.9092252827240045 and parameters: {'n estimators': 1302,
'learning rate': 0.012400823073576645, 'max depth': 11, 'subsample':
0.8671164146793225, 'colsample bytree': 0.44391109256011196,
'min child weight': 1, 'reg alpha': 0.0965491356761557, 'reg lambda':
5.343183919972633, 'gamma': 0.37313371582931043}. Best is trial 77
with value: 0.9093111311495488.
Done in 308.93 min
Best CV accuracy: 0.90931
Best params: {'n estimators': 1267, 'learning rate':
0.013034614108249017, 'max_depth': 11, 'subsample':
0.8583465135661814, 'colsample_bytree': 0.4261762035998442,
'min child weight': 1, 'reg alpha': 0.09678563447756561, 'reg lambda':
5.327675418687854, 'gamma': 0.3616069534480977}
Submission saved.
import pandas as pd
import numpy as np
```

```
import time
import warnings
from sklearn.preprocessing import StandardScaler, LabelEncoder
from sklearn.model selection import RepeatedStratifiedKFold
import xgboost as xgb
import optuna
warnings.filterwarnings("ignore")
file path = '/home/iiitb/Desktop/IIITB/ML/project first half/'
df = pd.read csv(file path + 'train.csv')
df test = pd.read csv(file path + 'test.csv')
X = df.drop(['id','WeightCategory'], axis=1)
y = df['WeightCategory']
categorical cols = X.select dtypes(include=['object']).columns
X = pd.get dummies(X, columns=categorical cols, drop first=True)
numerical cols =
['Age','Height','Weight','FCVC','NCP','CH20','FAF','TUE']
scaler = StandardScaler()
X[numerical cols] = scaler.fit transform(X[numerical cols])
le = LabelEncoder()
y enc = le.fit transform(y)
center1 = {'n estimators': 1244, 'learning rate': 0.01203757,
'max_depth': 11, 'subsample': 0.6732769, 'colsample bytree':
0.5301610, 'min_child_weight': 3, 'reg_alpha': 0.1385014,
'reg_lambda': 13.9555974, 'gamma': 0.5}
center2 = {'n_estimators': 1403, 'learning_rate': 0.01151984,
'max depth': 13, 'subsample': 0.8389925, 'colsample bytree':
0.4289206, 'min child weight': 1, 'reg alpha': 0.2172063,
'reg_lambda': 3.2747964, 'gamma': 0.3849006}
def objective(trial):
    ne lo = int(min(center1['n estimators'], center2['n estimators'])
* 0.95)
    ne hi = int(max(center1['n estimators'], center2['n estimators'])
* 1.05)
    lr lo = min(center1['learning rate'], center2['learning rate']) *
0.8
    lr hi = max(center1['learning rate'], center2['learning rate']) *
1.2
    ss lo = min(center1['subsample'], center2['subsample']) - 0.05
    ss hi = max(center1['subsample'], center2['subsample']) + 0.05
    cs lo = min(center1['colsample bytree'],
center2['colsample bytree']) - 0.05
    cs hi = max(center1['colsample bytree'],
center2['colsample bytree']) + 0.05
```

```
ra_lo = min(center1['reg_alpha'], center2['reg_alpha']) * 0.5
    ra hi = max(center1['reg alpha'], center2['reg alpha']) * 1.5
    rl_lo = min(center1['reg_lambda'], center2['reg_lambda']) * 0.8
    rl hi = max(center1['reg lambda'], center2['reg lambda']) * 1.2
    gm lo = min(center1['gamma'], center2['gamma']) - 0.1
    gm hi = max(center1['gamma'], center2['gamma']) + 0.1
    params = {
        'n estimators': trial.suggest int('n estimators', ne lo,
ne_hi),
        'learning rate': trial.suggest float('learning_rate', lr_lo,
lr hi, log=True),
        'max depth': trial.suggest int('max depth',
min(center1['max depth'], center2['max depth']),
max(center1['max depth'], center2['max depth'])),
        'subsample': trial.suggest_float('subsample', max(0.5, ss_lo),
min(1.0, ss hi)),
        'colsample bytree': trial.suggest float('colsample bytree',
\max(0.4, cs_{lo}), \min(1.0, cs_{hi}),
        'min child weight': trial.suggest int('min child weight',
min(center1['min_child_weight'], center2['min_child_weight']),
max(center1['min_child_weight'], center2['min_child_weight'])),
        'reg alpha': trial.suggest float('reg alpha', ra lo, ra hi,
log=True),
         reg lambda': trial.suggest float('reg lambda', rl lo, rl hi,
log=True),
         qamma': trial.suggest float('qamma', max(0.0, gm lo), gm hi),
        'objective': 'multi:softmax'
        'num class': len(le.classes ),
        'use label encoder': False,
        'eval metric': 'mlogloss',
        'random state': 42,
        'verbosity': 0,
        'n jobs': -1,
        'early stopping rounds': 50
    }
    cv = RepeatedStratifiedKFold(n splits=5, n repeats=3,
random state=42)
    scores = []
    for train idx, val idx in cv.split(X, y enc):
        X tr, X val = X.iloc[train idx], X.iloc[val_idx]
        y tr, y val = y enc[train idx], y enc[val idx]
        model = xgb.XGBClassifier(**params)
        model.fit(X tr, y tr, eval set=[(X val, y val)],
verbose=False)
        scores.append(model.score(X val, y val))
    return np.mean(scores)
pruner = optuna.pruners.MedianPruner(n startup trials=10,
```

```
n warmup steps=5)
study = optuna.create study(direction='maximize',
sampler=optuna.samplers.TPESampler(seed=42), pruner=pruner)
start = time.time()
study.optimize(objective, n trials=100, show progress bar=True)
end = time.time()
print(f"Done in {(end - start)/60:.2f} min")
print(f"Best CV accuracy: {study.best value:.5f}")
print("Best params:", study.best params)
best = study.best params.copy()
best.update({'objective':'multi:softmax','num class':len(le.classes ),
'use label encoder': False, 'eval metric': 'mlogloss', 'random state': 42, '
verbosity':0,'n jobs':-1})
final model = xgb.XGBClassifier(**best)
final model.fit(X, y enc, verbose=False)
X test = df test.drop(['id'], axis=1)
X test = pd.get dummies(X test, columns=categorical cols,
drop first=True)
X test = X test.reindex(columns=X.columns, fill value=0)
X test[numerical cols] = scaler.transform(X test[numerical cols])
v test pred enc = final model.predict(X test)
y test pred = le.inverse transform(y test pred enc)
submission = pd.DataFrame({'id': df test['id'], 'WeightCategory':
y test pred})
submission.to csv(file path + 'xg optuna best2.csv', index=False)
print("Submission saved.")
/home/iiitb/Desktop/New Folder/MT2025065/.venv/lib/python3.12/site-
packages/tqdm/auto.py:21: TqdmWarning: IProgress not found. Please
update jupyter and ipywidgets. See
https://ipywidgets.readthedocs.io/en/stable/user install.html
  from .autonotebook import tqdm as notebook tqdm
[I 2025-10-25 11:30:43,222] A new study created in memory with name:
no-name-602a86a4-e7e9-44e9-8e7d-7f3c8ddd5e54
Loading Training and Test data...
Data loaded successfully.
Starting Data Preprocessing...
Preprocessing and Encoding complete.
Optimizing around max depth=8 (your best performer: 0.90183)...
Starting OPTUNA search with max depth searchable (7-12)...
```

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Best trial: 0. Best value: 0.907262:
                                         1% | 1/100
[00:36<1:00:51, 36.89s/it]
[I 2025-10-25 11:31:20,112] Trial 0 finished with value:
0.9072616701045414 and parameters: {'max depth': 9, 'n estimators':
693, 'learning rate': 0.09454306819536169, 'reg alpha':
0.3084019079099502, 'reg_lambda': 11.536818470031642, 'gamma':
0.362397808134481, 'min child weight': 1, 'subsample':
0.8598528437324806, 'colsample bytree': 0.5803345035229626\. Best is
trial 0 with value: 0.9072616701045414.
Best trial: 0. Best value: 0.907262: 2%|| | 2/100
[00:46<34:29, 21.12s/it]
[I 2025-10-25 11:31:30,195] Trial 1 finished with value:
0.9052661035225487 and parameters: {'max depth': 11, 'n estimators':
553, 'learning_rate': 0.41472250004816347, 'reg_alpha':
0.40865594623548174, 'reg_lambda': 12.14781662658844, 'gamma':
0.3727299868828402, 'min_child_weight': 1, 'subsample': 0.6912726728878613, 'colsample_bytree': 0.5574269294896713}. Best is
trial 0 with value: 0.9072616701045414.
Best trial: 2. Best value: 0.907712: 3%|| | 3/100
[01:50<1:05:30, 40.52s/it]
[I 2025-10-25 11:32:33,791] Trial 2 finished with value:
0.9077122758965434 and parameters: {'max depth': 9, 'n estimators':
593, 'learning rate': 0.044809759182149515, 'reg alpha':
0.17743105460380698, 'reg lambda': 13.06941185248186, 'gamma':
0.4465447373174767, 'min_child_weight': 2, 'subsample': 0.8355527884179041, 'colsample_bytree': 0.45990213464750795}. Best is
trial 2 with value: 0.9077122758965434.
Best trial: 2. Best value: 0.907712: 4%| | 4/100
[04:00<2:01:05, 75.68s/it]
[I 2025-10-25 11:34:43,378] Trial 3 finished with value:
0.8914566240243091 and parameters: {'max depth': 10, 'n estimators':
639, 'learning rate': 0.0013346527038305934, 'reg alpha':
0.3117191970784111, 'reg_lambda': 11.691180651487528, 'gamma': 0.3260206371941118, 'min_child_weight': 3, 'subsample':
0.8896896099223679, 'colsample bytree': 0.6425192044349384}. Best is
trial 2 with value: 0.9077122758965434.
Best trial: 4. Best value: 0.907938: 5%| | 5/100
[04:49<1:44:53, 66.24s/it]
[I 2025-10-25 11:35:32,890] Trial 4 finished with value:
0.9079375943360348 and parameters: {'max depth': 8, 'n estimators':
564, 'learning rate': 0.07026263205443048, 'reg alpha':
0.254822424053477, 'reg lambda': 11.18314346170632, 'gamma':
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0.498070764044508, 'min_child_weight': 1, 'subsample':
0.8727961206236347, 'colsample_bytree': 0.4776339944800051}. Best is
trial 4 with value: 0.9079375943360348.
[06:14<1:53:36, 72.52s/it]
[I 2025-10-25 11:36:57,588] Trial 5 finished with value:
0.9070041662772159 and parameters: {'max_depth': 10, 'n_estimators':
597, 'learning rate': 0.02533074654001447, 'reg_alpha':
0.28970396819168, 'reg lambda': 11.845706927460496, 'gamma':
0.6878338511058234, 'min_child_weight': 3, 'subsample':
0.8818496824692568, 'colsample bytree': 0.6684482051282946}. Best is
trial 4 with value: 0.9079375943360348.
[08:14<2:16:34, 88.12s/it]
[I 2025-10-25 11:38:57,817] Trial 6 finished with value:
0.892519000881005 and parameters: {'max depth': 10, 'n estimators':
689, 'learning rate': 0.0017331598058558703, 'reg alpha':
0.18991817672006095, 'reg lambda': 10.423120239310245, 'gamma':
0.4301321323053057, 'min_child_weight': 2, 'subsample':
0.6814047095321688, 'colsample bytree': 0.6486212527455788}. Best is
trial 4 with value: 0.9079375943360348.
[09:19<2:03:51, 80.78s/it]
[I 2025-10-25 11:40:02,892] Trial 7 finished with value:
0.9077121722732727 and parameters: {'max depth': 9, 'n estimators':
592, 'learning rate': 0.029155497059176992, 'reg alpha':
0.1777368753807051, 'reg lambda': 20.855772148718884, 'gamma':
0.3298202574719083, 'min child weight': 3, 'subsample':
0.8316734307889972, 'colsample bytree': 0.4596147044602517}. Best is
trial 4 with value: 0.9079375943360348.
Best trial: 4. Best value: 0.907938: 9%| | 9/100
[09:54<1:40:50, 66.49s/it]
[I 2025-10-25 11:40:37,965] Trial 8 finished with value:
0.9047830325598823 and parameters: {'max_depth': 7, 'n_estimators':
673, 'learning rate': 0.08086987436021253, 'reg alpha':
0.36080588483591136, 'reg_lambda': 20.273061061266873, 'gamma': 0.32961786069363613, 'min_child_weight': 2, 'subsample':
0.6347607178575388, 'colsample_bytree': 0.6589310277626781}. Best is
trial 4 with value: 0.9079375943360348.
Best trial: 4. Best value: 0.907938: 10%| ■ | 10/100
[11:21<1:49:19, 72.88s/it]
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[I 2025-10-25 11:42:05,154] Trial 9 finished with value:
0.8870787584522912 and parameters: {'max depth': 10, 'n estimators':
599, 'learning rate': 0.0014843697010415793, 'reg alpha':
0.21812091218197027, 'reg_lambda': 13.471112219430925, 'gamma':
0.5918424713352256, 'min_child_weight': 2, 'subsample':
0.866163822772898, 'colsample_bytree': 0.5416644775485848}. Best is
trial 4 with value: 0.9079375943360348.
[12:25<1:43:39, 69.88s/it]
[I 2025-10-25 11:43:08,219] Trial 10 finished with value:
0.8941284671569152 and parameters: {'max depth': 7, 'n estimators':
550, 'learning rate': 0.006767954762923837, 'reg alpha':
0.23643626561361977, 'reg_lambda': 16.344471855462476, 'gamma':
0.5429521169792498, 'min_child_weight': 1, 'subsample':
0.7840686658512219, 'colsample_bytree': 0.41421773065075795}. Best is
trial 4 with value: 0.9079375943360348.
[12:49<1:22:17, 56.11s/it]
[I 2025-10-25 11:43:32,825] Trial 11 finished with value:
0.9074870714426494 and parameters: {'max depth': 8, 'n estimators':
575, 'learning rate': 0.12406252498920857, 'reg alpha':
0.15401984858106837, 'reg_lambda': 14.888390067671295, 'qamma':
0.46239358524520957, 'min_child_weight': 1, 'subsample': 0.8027571855210263, 'colsample_bytree': 0.48217735311554505}. Best is
trial 4 with value: 0.9079375943360348.
Best trial: 4. Best value: 0.907938: 13%
[14:06<1:30:29, 62.40s/it]
[I 2025-10-25 11:44:49,719] Trial 12 finished with value:
0.8986348255844196 and parameters: {'max depth': 8, 'n estimators':
628, 'learning rate': 0.008061526676599168, 'reg_alpha':
0.24435456585995394, 'reg_lambda': 17.435091762344154, 'gamma':
0.5206301390946335, 'min_child_weight': 2, 'subsample':
0.7530182254524145, 'colsample bytree': 0.48935336512257066}. Best is
trial 4 with value: 0.9079375943360348.
[14:19<1:07:58, 47.42s/it]
[I 2025-10-25 11:45:02,533] Trial 13 finished with value:
0.9058131514924549 and parameters: {'max_depth': 12, 'n_estimators':
575, 'learning rate': 0.3221873115687971, 'reg alpha':
0.4996163439989345, 'reg_lambda': 10.03354901049281, 'gamma': 0.4371303839678691, 'min_child_weight': 1, 'subsample':
0.8242283019211265, 'colsample bytree': 0.41119095327822264}. Best is
trial 4 with value: 0.9079375943360348.
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Best trial: 4. Best value: 0.907938: 15%| | 15/100
[15:15<1:10:53, 50.04s/it]
[I 2025-10-25 11:45:58,625] Trial 14 finished with value:
0.9076156990083873 and parameters: {'max depth': 8, 'n estimators':
610, 'learning rate': 0.04759058309950689, 'reg alpha':
0.15183059080919273, 'reg_lambda': 14.002151024775975, 'gamma':
0.6006453699263155, 'min child weight': 2, 'subsample':
0.7505228771706659, 'colsample bytree': 0.515625654232895}. Best is
trial 4 with value: 0.9079375943360348.
[16:38<1:24:01, 60.02s/it]
[I 2025-10-25 11:47:21,839] Trial 15 finished with value:
0.9036565128987137 and parameters: {'max depth': 9, 'n estimators':
577, 'learning rate': 0.012134871009790911, 'reg_alpha':
0.19534669658014717, 'reg_lambda': 13.061913545468197, 'gamma': 0.47059925068704495, 'min_child_weight': 2, 'subsample': 0.8995222385654719, 'colsample_bytree': 0.44711218688904153}. Best is
trial 4 with value: 0.9079375943360348.
[16:48<1:02:14, 45.00s/it]
[I 2025-10-25 11:47:31,900] Trial 16 finished with value:
0.9069077240993118 and parameters: {'max depth': 8, 'n estimators':
646, 'learning rate': 0.18445537108293755, 'reg alpha':
0.2596546776956449, 'reg lambda': 17.80061695302859, 'gamma':
0.4031378292870405, 'min child weight': 1, 'subsample':
0.8326673858879777, 'colsample_bytree': 0.5999606917923157}. Best is
trial 4 with value: 0.9079375943360348.
Best trial: 4. Best value: 0.907938: 18%
[17:17<54:44, 40.06s/it]
[I 2025-10-25 11:48:00.447] Trial 17 finished with value:
0.9076478843962213 and parameters: {'max depth': 7, 'n estimators':
564, 'learning rate': 0.04916403939842982, 'reg alpha':
0.2067026416409015, 'reg_lambda': 10.984308065487593, 'gamma':
0.5604035330355402, 'min child weight': 3, 'subsample':
0.7784227274070701, 'colsample_bytree': 0.5078626886500156}. Best is
trial 4 with value: 0.9079375943360348.
Best trial: 4. Best value: 0.907938: 19%| | 19/100
[18:10<59:16, 43.91s/it]
[I 2025-10-25 11:48:53,339] Trial 18 finished with value:
0.8915853759379718 and parameters: {'max depth': 11, 'n estimators':
608, 'learning rate': 0.003560847005574074, 'reg alpha':
0.17136972625974958, 'reg lambda': 12.808426576879743, 'gamma':
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0.4888490300704192, 'min_child_weight': 1, 'subsample':
0.8447189511928629, 'colsample bytree': 0.4493190212829957}. Best is
trial 4 with value: 0.9079375943360348.
Best trial: 4. Best value: 0.907938: 20%| | 20/100
[18:53<58:13, 43.67s/it]
[I 2025-10-25 11:49:36,444] Trial 19 finished with value:
0.9052338041491172 and parameters: {'max_depth': 9, 'n_estimators':
584, 'learning rate': 0.01567452598445484, 'reg alpha':
0.3606436133897752, 'reg lambda': 15.069054172385574, 'gamma':
0.6517539045363733, 'min_child_weight': 2, 'subsample':
0.7217760796227951, 'colsample bytree': 0.5312717733967197}. Best is
trial 4 with value: 0.9079375943360348.
Best trial: 4. Best value: 0.907938: 21%
[19:03<44:20, 33.67s/it]
[I 2025-10-25 11:49:46,812] Trial 20 finished with value:
0.9057810179162562 and parameters: {'max depth': 8, 'n estimators':
615, 'learning rate': 0.20854981842516834, 'reg alpha':
0.21840526226234408, 'reg lambda': 23.40611117008207, 'gamma':
0.5169214340286176, 'min_child_weight': 2, 'subsample':
0.6034862884882659, 'colsample_bytree': 0.6981528355149126}. Best is
trial 4 with value: 0.9079375943360348.
Best trial: 4. Best value: 0.907938: 22%
[19:49<48:37, 37.40s/it]
[I 2025-10-25 11:50:32,907] Trial 21 finished with value:
0.9070040315669642 and parameters: {'max depth': 9, 'n estimators':
589, 'learning rate': 0.03638450400809701, 'reg alpha':
0.17762714934835896, 'reg lambda': 24.398683917929162, 'gamma':
0.4065916059622793, 'min child weight': 3, 'subsample':
0.8152006552468911, 'colsample bytree': 0.4576913355991772}. Best is
trial 4 with value: 0.9079375943360348.
Best trial: 4. Best value: 0.907938: 23%
[20:36<51:34, 40.19s/it]
[I 2025-10-25 11:51:19,592] Trial 22 finished with value:
0.9066499715761369 and parameters: {'max_depth': 9, 'n_estimators':
565, 'learning rate': 0.021094887010302554, 'reg alpha':
0.16749901388901225, 'reg_lambda': 21.574661877309772, 'gamma':
0.3652347869183132, 'min child weight': 3, 'subsample':
0.8523012013749908, 'colsample bytree': 0.4690968633724988}. Best is
trial 4 with value: 0.9079375943360348.
Best trial: 4. Best value: 0.907938: 24%
[21:05<46:42, 36.87s/it]
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[I 2025-10-25 11:51:48,733] Trial 23 finished with value:
0.9076478533092402 and parameters: {'max depth': 8, 'n estimators':
591, 'learning rate': 0.06355333466035192, 'reg alpha':
0.18617934738546185, 'reg_lambda': 18.351591490254794, 'qamma':
0.4503785446646292, 'min_child_weight': 3, 'subsample': 0.7947084771164531, 'colsample_bytree': 0.43219474398322844}. Best is
trial 4 with value: 0.9079375943360348.
Best trial: 4. Best value: 0.907938: 25%
[21:43<46:19, 37.06s/it]
[I 2025-10-25 11:52:26,232] Trial 24 finished with value:
0.9078409863608978 and parameters: {'max_depth': 9, 'n_estimators':
562, 'learning rate': 0.03129586877178342, 'reg alpha':
0.2346229082274662, 'reg_lambda': 19.72377790348703, 'gamma':
0.3126486621390768, 'min child weight': 3, 'subsample':
0.8708232037438176, 'colsample bytree': 0.4977652761441269}. Best is
trial 4 with value: 0.9079375943360348.
Best trial: 4. Best value: 0.907938: 26%
[21:58<37:33, 30.45s/it]
[I 2025-10-25 11:52:41,255] Trial 25 finished with value:
0.9076802355812882 and parameters: {'max depth': 11, 'n estimators':
562, 'learning_rate': 0.12872896680032264, 'reg_alpha':
0.2528311883336203, 'reg_lambda': 14.353518178991987, 'gamma':
0.305560896879458, 'min_child_weight': 2, 'subsample':
0.8711570123770287, 'colsample bytree': 0.49792489078277524}. Best is
trial 4 with value: 0.9079375943360348.
Best trial: 4. Best value: 0.907938: 27%| | 27/100
[22:36<40:06, 32.97s/it]
[I 2025-10-25 11:53:20,104] Trial 26 finished with value:
0.9025619921034922 and parameters: {'max depth': 7, 'n estimators':
563, 'learning rate': 0.01561956072736252, 'reg alpha':
0.2262975901575149, 'reg_lambda': 19.53862911827419, 'gamma': 0.49506704748454455, 'min_child_weight': 2, 'subsample':
0.8745727215750252, 'colsample bytree': 0.43152110056120235}. Best is
trial 4 with value: 0.9079375943360348.
Best trial: 4. Best value: 0.907938: 28% | | 28/100
[23:12<40:20, 33.62s/it]
[I 2025-10-25 11:53:55,248] Trial 27 finished with value:
0.9077767088461733 and parameters: {'max depth': 8, 'n estimators':
573, 'learning rate': 0.04479364005505688, 'reg alpha':
0.2757059957499572, 'reg_lambda': 15.730096923220845, 'gamma': 0.39773093401276494, 'min_child_weight': 3, 'subsample':
0.8983286879882807, 'colsample_bytree': 0.5252836940783759}. Best is
trial 4 with value: 0.9079375943360348.
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Best trial: 4. Best value: 0.907938: 29%| | 29/100
[24:05<46:43, 39.49s/it]
[I 2025-10-25 11:54:48,438] Trial 28 finished with value:
0.9003731681219174 and parameters: {'max depth': 8, 'n estimators':
550, 'learning rate': 0.00819422426242818, 'reg alpha':
0.2757651894620551, 'reg_lambda': 15.921449512633412, 'gamma': 0.39685743919914723, 'min_child_weight': 3, 'subsample':
0.8972183879671085, 'colsample bytree': 0.5724725802888885}. Best is
trial 4 with value: 0.9079375943360348.
Best trial: 4. Best value: 0.907938: 30%| | 30/100
[24:29<40:39, 34.84s/it]
[I 2025-10-25 11:55:12,433] Trial 29 finished with value:
0.9068433533236439 and parameters: {'max depth': 7, 'n estimators':
577, 'learning_rate': 0.08648716652183838, 'reg alpha':
0.3091979627871436, 'reg_lambda': 22.742785740518045, 'gamma': 0.30101289745572085, 'min_child_weight': 3, 'subsample':
0.856143509437081, 'colsample bytree': 0.6028894330026388}. Best is
trial 4 with value: 0.9079375943360348.
Best trial: 4. Best value: 0.907938: 31%
[25:01<39:08, 34.04s/it]
[I 2025-10-25 11:55:44,608] Trial 30 finished with value:
0.9072617219161767 and parameters: {'max depth': 8, 'n estimators':
658, 'learning rate': 0.06883017680923502, 'reg_alpha':
0.32765730219067646, 'reg_lambda': 16.75047311838786, 'gamma':
0.38165617032900273, 'min_child_weight': 3, 'subsample':
0.8593632102614747, 'colsample bytree': 0.5125379063979899}. Best is
trial 4 with value: 0.9079375943360348.
Best trial: 31. Best value: 0.908002: 32%
[26:00<47:05, 41.56s/it]
[I 2025-10-25 11:56:43,695] Trial 31 finished with value:
0.9080020480103193 and parameters: {'max depth': 9, 'n estimators':
569, 'learning rate': 0.040386180597475355, 'reg alpha':
0.2721168377389818, 'reg_lambda': 10.918588180951035, 'gamma':
0.4256939348254053, 'min child weight': 3, 'subsample':
0.8407277274415939, 'colsample bytree': 0.4805003655588821}. Best is
trial 31 with value: 0.9080020480103193.
Best trial: 32. Best value: 0.908485: 33%
[26:40<45:44, 40.96s/it]
[I 2025-10-25 11:57:23,254] Trial 32 finished with value:
0.9084848391901552 and parameters: {'max depth': 9, 'n estimators':
560, 'learning rate': 0.03200260990011846, 'reg alpha':
0.27650449377776626, 'reg lambda': 10.780818570572933, 'gamma':
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0.34868377217101126, 'min child weight': 3, 'subsample':
0.8784773465054003, 'colsample bytree': 0.5562973076751221}. Best is
trial 32 with value: 0.9084848391901552.
Best trial: 32. Best value: 0.908485: 34%| 34/100
[27:28<47:22, 43.07s/it]
[I 2025-10-25 11:58:11,241] Trial 33 finished with value:
0.9075513178703926 and parameters: {'max depth': 9, 'n estimators':
557, 'learning rate': 0.021554969190879205, 'reg_alpha':
0.2885384326163285, 'reg_lambda': 10.798367332712527, 'gamma':
0.34875116649979454, 'min_child_weight': 3, 'subsample':
0.8735312845912004, 'colsample_bytree': 0.5601319397198061}. Best is
trial 32 with value: 0.9084848391901552.
Best trial: 32. Best value: 0.908485: 35%
[28:10<46:34, 42.99s/it]
[I 2025-10-25 11:58:54,040] Trial 34 finished with value:
0.9076479569325109 and parameters: {'max depth': 10, 'n estimators':
570, 'learning rate': 0.03221807695778123, 'reg alpha':
0.3394022338368488, 'reg_lambda': 11.226233252292584, 'gamma': 0.33422780045821887, 'min_child_weight': 3, 'subsample': 0.8118551797582709, 'colsample_bytree': 0.47892458107538116}. Best is
trial 32 with value: 0.9084848391901552.
Best trial: 32. Best value: 0.908485: 36%
[28:26<37:13, 34.90s/it]
[I 2025-10-25 11:59:10,059] Trial 35 finished with value:
0.9062960255709072 and parameters: {'max depth': 9, 'n estimators':
556, 'learning rate': 0.11951133430347405, 'reg alpha':
0.2608602157953146, 'reg lambda': 10.024476187312924, 'gamma':
0.3611375571950897, 'min child weight': 3, 'subsample':
0.8397602735263449, 'colsample bytree': 0.5974113118702501}. Best is
trial 32 with value: 0.9084848391901552.
Best trial: 32. Best value: 0.908485: 37%
[28:57<35:26, 33.76s/it]
[I 2025-10-25 11:59:41,167] Trial 36 finished with value:
0.9074870092686871 and parameters: {'max_depth': 10, 'n_estimators':
559, 'learning_rate': 0.05962752923747867, 'reg_alpha':
0.2360807046786588, 'reg_lambda': 12.253530723059182, 'gamma': 0.38126366675045087, 'min_child_weight': 3, 'subsample': 0.8811231084360922, 'colsample_bytree': 0.5382451332272343}. Best is
trial 32 with value: 0.9084848391901552.
Best trial: 32. Best value: 0.908485: 38%
[29:50<40:50, 39.52s/it]
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[I 2025-10-25 12:00:34,137] Trial 37 finished with value:
0.9048153112086597 and parameters: {'max depth': 9, 'n estimators':
583, 'learning rate': 0.011591202242870191, 'reg alpha':
0.2933130790028229, 'reg lambda': 12.132700637480953, 'gamma':
0.42750865779935565, 'min child weight': 1, 'subsample':
0.8500672074592062, 'colsample_bytree': 0.5589060239791996}. Best is
trial 32 with value: 0.9084848391901552.
Best trial: 32. Best value: 0.908485: 39%
[30:41<43:39, 42.94s/it]
[I 2025-10-25 12:01:25,038] Trial 38 finished with value:
0.90806640842366 and parameters: {'max depth': 11, 'n estimators':
600, 'learning rate': 0.03647744383473826, 'reg alpha':
0.429933105631015, 'reg_lambda': 11.600316419930733, 'gamma': 0.3535855206246623, 'min_child_weight': 3, 'subsample':
0.6908581442184776, 'colsample bytree': 0.4957156374522839}. Best is
trial 32 with value: 0.9084848391901552.
Best trial: 32. Best value: 0.908485: 40%
[31:43<48:36, 48.61s/it]
[I 2025-10-25 12:02:26,905] Trial 39 finished with value:
0.9065856318874502 and parameters: {'max depth': 12, 'n estimators':
604, 'learning rate': 0.019027254214004845, 'reg alpha':
0.4206785297799769, 'req lambda': 11.61434224284003, 'qamma':
0.42613077142170974, 'min_child_weight': 3, 'subsample':
0.6779871853179118, 'colsample bytree': 0.615747665968695}. Best is
trial 32 with value: 0.9084848391901552.
Best trial: 32. Best value: 0.908485: 41%
[31:50<35:33, 36.16s/it]
[I 2025-10-25 12:02:34,004] Trial 40 finished with value:
0.9043322402459932 and parameters: {'max depth': 11, 'n estimators':
617, 'learning rate': 0.4921266312937146, 'reg alpha':
0.4110250394712402, 'reg_lambda': 10.655157710607, 'gamma': 0.35140157942318334, 'min_child_weight': 3, 'subsample':
0.6953462820883662, 'colsample_bytree': 0.47571765740510485}. Best is
trial 32 with value: 0.9084848391901552.
Best trial: 32. Best value: 0.908485: 42%| 42/100
[32:42<39:23, 40.75s/it]
[I 2025-10-25 12:03:25,479] Trial 41 finished with value:
0.9077122551718892 and parameters: {'max depth': 10, 'n estimators':
583, 'learning rate': 0.029374028983161916, 'reg alpha':
0.4589875331515143, 'reg lambda': 11.317620667697922, 'qamma':
0.313555976131678, 'min_child_weight': 3, 'subsample': 0.6480711596687364, 'colsample bytree': 0.49516427971041865}. Best is
trial 32 with value: 0.9084848391901552.
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Best trial: 32. Best value: 0.908485: 43%
[33:37<42:58, 45.24s/it]
[I 2025-10-25 12:04:21,198] Trial 42 finished with value:
0.9080985730868397 and parameters: {'max depth': 9, 'n estimators':
569, 'learning rate': 0.03435958017573791, 'reg alpha':
0.20690807398473646, 'reg_lambda': 10.51618878381535, 'gamma':
0.3401691543928093, 'min child weight': 3, 'subsample':
0.7687343695961995, 'colsample bytree': 0.548220834726337}. Best is
trial 32 with value: 0.9084848391901552.
Best trial: 32. Best value: 0.908485: 44%
[34:26<43:11, 46.27s/it]
[I 2025-10-25 12:05:09,867] Trial 43 finished with value:
0.9075191946565209 and parameters: {'max depth': 10, 'n estimators':
596, 'learning_rate': 0.044161005022946835, 'reg_alpha':
0.3628280232373648, 'reg lambda': 10.506662061346118, 'qamma':
0.3431340639477099, 'min_child_weight': 3, 'subsample': 0.7249459440568092, 'colsample_bytree': 0.5495809901124876}. Best is
trial 32 with value: 0.9084848391901552.
Best trial: 32. Best value: 0.908485: 45%
[34:50<36:22, 39.69s/it]
[I 2025-10-25 12:05:34,189] Trial 44 finished with value:
0.9072938347677212 and parameters: {'max depth': 10, 'n estimators':
570, 'learning rate': 0.09250880645044561, 'reg alpha':
0.20348378828655808, 'reg lambda': 12.416450303407444, 'gamma':
0.4125534161990071, 'min_child_weight': 3, 'subsample': 0.7122803622413336, 'colsample_bytree': 0.5751928428085551}. Best is
trial 32 with value: 0.9084848391901552.
Best trial: 32. Best value: 0.908485: 46%
[35:45<39:38, 44.05s/it]
[I 2025-10-25 12:06:28.411] Trial 45 finished with value:
0.9079055643831069 and parameters: {'max depth': 11, 'n estimators':
629, 'learning rate': 0.0379911764861185, 'reg alpha':
0.3273577391195896, 'reg_lambda': 11.84290726971009, 'gamma': 0.37195920407297817, 'min_child_weight': 3, 'subsample':
0.775623276354313, 'colsample_bytree': 0.5218727466546925}. Best is
trial 32 with value: 0.9084848391901552.
Best trial: 32. Best value: 0.908485: 47%
[36:18<36:10, 40.95s/it]
[I 2025-10-25 12:07:02,119] Trial 46 finished with value:
0.9070363205780685 and parameters: {'max depth': 9, 'n estimators':
691, 'learning rate': 0.05973747451700418, 'reg alpha':
0.2088988462053125, 'reg lambda': 10.333394198509296, 'gamma':
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0.46890109615561115, 'min child weight': 1, 'subsample':
0.7411904255438531, 'colsample bytree': 0.6206026986336131}. Best is
trial 32 with value: 0.9084848391901552.
Best trial: 32. Best value: 0.908485: 48%| 48/100
[37:23<41:33, 47.94s/it]
[I 2025-10-25 12:08:06,392] Trial 47 finished with value:
0.9073904012935501 and parameters: {'max_depth': 12, 'n_estimators':
584, 'learning rate': 0.025132188193697724, 'reg_alpha':
0.37777415191621244, 'reg lambda': 11.129885456170555, 'gamma':
0.3286440103922827, 'min_child_weight': 2, 'subsample': 0.6597348083734487, 'colsample_bytree': 0.5459537226914817}. Best is
trial 32 with value: 0.9084848391901552.
Best trial: 32. Best value: 0.908485: 49%
[37:49<35:15, 41.48s/it]
[I 2025-10-25 12:08:32,787] Trial 48 finished with value:
0.9062959426722909 and parameters: {'max depth': 10, 'n estimators':
569, 'learning rate': 0.14265136212461135, 'reg alpha':
0.2910426328574746, 'reg_lambda': 13.601116783993778, 'gamma': 0.3860847580968524, 'min_child_weight': 3, 'subsample':
0.6167077729827729, 'colsample bytree': 0.43391884877598785}. Best is
trial 32 with value: 0.9084848391901552.
Best trial: 32. Best value: 0.908485: 50%
[39:36<50:50, 61.02s/it]
[I 2025-10-25 12:10:19,387] Trial 49 finished with value:
0.896542661389617 and parameters: {'max depth': 9, 'n estimators':
554, 'learning rate': 0.003975662952647683, 'reg alpha':
0.24696268526637902, 'reg lambda': 11.594704908334311, 'gamma':
0.5587058506340858, 'min child weight': 2, 'subsample':
0.8214672685432955, 'colsample bytree': 0.5846468057755588}. Best is
trial 32 with value: 0.9084848391901552.
Best trial: 32. Best value: 0.908485: 51%
[39:49<38:05, 46.65s/it]
[I 2025-10-25 12:10:32,502] Trial 50 finished with value:
0.9055235037266037 and parameters: {'max_depth': 8, 'n_estimators':
601, 'learning rate': 0.29412875136223354, 'reg alpha':
0.4810605831036458, 'reg_lambda': 10.272214417548842, 'gamma':
0.5205459642997539, 'min_child_weight': 1, 'subsample':
0.7557004034791013, 'colsample_bytree': 0.4691539368411108}. Best is
trial 32 with value: 0.9084848391901552.
Best trial: 32. Best value: 0.908485: 52%
[41:15<46:53, 58.62s/it]
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[I 2025-10-25 12:11:59,075] Trial 51 finished with value:
0.9073904634675124 and parameters: {'max depth': 11, 'n estimators':
627, 'learning rate': 0.038193786968788425, 'reg alpha':
0.32437828423456566, 'reg lambda': 11.960059288752003, 'gamma':
0.36661923241341915, 'min_child_weight': 3, 'subsample':
0.7728108438858813, 'colsample_bytree': 0.5210810375984971}. Best is
trial 32 with value: 0.9084848391901552.
Best trial: 32. Best value: 0.908485: 53%
[41:46<39:27, 50.38s/it]
[I 2025-10-25 12:12:30,211] Trial 52 finished with value:
0.9066501062863888 and parameters: {'max depth': 11, 'n estimators':
637, 'learning rate': 0.07391942720227289, 'reg alpha':
0.2667450281782692, 'reg_lambda': 10.852191855611308, 'gamma': 0.3429128745457222, 'min_child_weight': 3, 'subsample':
0.7638889788401917, 'colsample bytree': 0.5098341965150155}. Best is
trial 32 with value: 0.9084848391901552.
Best trial: 32. Best value: 0.908485: 54%
[42:43<40:05, 52.30s/it]
[I 2025-10-25 12:13:26,989] Trial 53 finished with value:
0.9072295054413615 and parameters: {'max depth': 11, 'n estimators':
638, 'learning rate': 0.05318479710346607, 'reg alpha':
0.43220241650928704, 'reg lambda': 12.688914160897312, 'gamma':
0.3742026959805025, 'min child weight': 3, 'subsample':
0.7874437400144461, 'colsample bytree': 0.5346794280510786}. Best is
trial 32 with value: 0.9084848391901552.
Best trial: 32. Best value: 0.908485: 55%
[43:30<37:58, 50.64s/it]
[I 2025-10-25 12:14:13,754] Trial 54 finished with value:
0.906456807437498 and parameters: {'max depth': 12, 'n estimators':
621, 'learning rate': 0.016578935254680502, 'reg_alpha':
0.2789390983918678, 'reg_lambda': 11.638228091280695, 'gamma': 0.44698284761941703, 'min_child_weight': 3, 'subsample':
0.7378885129854517, 'colsample bytree': 0.48414565356243366}. Best is
trial 32 with value: 0.9084848391901552.
[44:13<35:23, 48.26s/it]
[I 2025-10-25 12:14:56,457] Trial 55 finished with value:
0.9078087802484098 and parameters: {'max depth': 10, 'n estimators':
648, 'learning rate': 0.027111046233822638, 'reg alpha':
0.39352999182389425, 'reg lambda': 10.97529793651663, 'gamma':
0.4166788549590792, 'min_child_weight': 3, 'subsample':
0.7965897646728793, 'colsample bytree': 0.5024295090478783}. Best is
trial 32 with value: 0.9084848391901552.
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Best trial: 32. Best value: 0.908485: 57%
[45:08<36:03, 50.32s/it]
[I 2025-10-25 12:15:51,578] Trial 56 finished with value:
0.9054912354401534 and parameters: {'max depth': 11, 'n estimators':
680, 'learning_rate': 0.012096495804165113, 'reg alpha':
0.34675514870862606, 'reg_lambda': 13.15056032939283, 'gamma':
0.3210856722462355, 'min child weight': 3, 'subsample':
0.7017963931202591, 'colsample bytree': 0.5616523940644064}. Best is
trial 32 with value: 0.9084848391901552.
Best trial: 32. Best value: 0.908485: 58%
[45:46<32:38, 46.63s/it]
[I 2025-10-25 12:16:29,601] Trial 57 finished with value:
0.9082917786747867 and parameters: {'max depth': 9, 'n estimators':
632, 'learning_rate': 0.03432432185399957, 'reg_alpha':
0.3054977589462564, 'reg_lambda': 10.50110764151904, 'gamma': 0.35721434057899976, 'min_child_weight': 3, 'subsample':
0.8888051799018377, 'colsample bytree': 0.5225824480602327}. Best is
trial 32 with value: 0.9084848391901552.
Best trial: 32. Best value: 0.908485: 59%
[46:05<26:09, 38.28s/it]
[I 2025-10-25 12:16:48,387] Trial 58 finished with value:
0.9074225866813839 and parameters: {'max depth': 9, 'n estimators':
578, 'learning rate': 0.10126776854202654, 'reg alpha':
0.2973869156610779, 'reg_lambda': 10.408659771098357, 'gamma':
0.6135846627815915, 'min_child_weight': 3, 'subsample':
0.882231354574486, 'colsample_bytree': 0.4619253182503328}. Best is
trial 32 with value: 0.9084848391901552.
Best trial: 32. Best value: 0.908485: 60%| 60/100
[46:49<26:47, 40.19s/it]
[I 2025-10-25 12:17:33.039] Trial 59 finished with value:
0.9083238811640044 and parameters: {'max depth': 9, 'n estimators':
661, 'learning rate': 0.02442242653896824, 'reg alpha':
0.1606134482947865, 'reg_lambda': 10.011552236626676, 'gamma': 0.35763246860453574, 'min_child_weight': 2, 'subsample': 0.8881324826281654, 'colsample_bytree': 0.48888297486168114}. Best is
trial 32 with value: 0.9084848391901552.
Best trial: 32. Best value: 0.908485: 61%
[47:35<27:13, 41.88s/it]
[I 2025-10-25 12:18:18,859] Trial 60 finished with value:
0.9084847562915387 and parameters: {'max depth': 9, 'n estimators':
661, 'learning rate': 0.023153490046554523, 'reg alpha':
0.16024583874220955, 'reg lambda': 10.00799542475368, 'gamma':
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0.3544900549153444, 'min_child_weight': 2, 'subsample':
0.8646364404218517, 'colsample bytree': 0.4901598333688622}. Best is
trial 32 with value: 0.9084848391901552.
Best trial: 61. Best value: 0.908581: 62%| 62/100
[48:21<27:11, 42.95s/it]
[I 2025-10-25 12:19:04,295] Trial 61 finished with value:
0.9085813746290029 and parameters: {'max depth': 9, 'n estimators':
660, 'learning rate': 0.02395726967342789, 'reg alpha':
0.160495237653443, 'reg lambda': 10.036834473396658, 'gamma':
0.3539289774766382, 'min_child_weight': 2, 'subsample':
0.8886038190313954, 'colsample_bytree': 0.4851336278434492}. Best is
trial 61 with value: 0.9085813746290029.
Best trial: 61. Best value: 0.908581: 63%
[49:08<27:19, 44.30s/it]
[I 2025-10-25 12:19:51,773] Trial 62 finished with value:
0.9079698729848124 and parameters: {'max depth': 9, 'n estimators':
666, 'learning rate': 0.022856839390730885, 'reg alpha':
0.15790685657520467, 'reg lambda': 10.14688095303086, 'gamma':
0.3562681470657958, 'min_child_weight': 2, 'subsample':
0.8880613232031223, 'colsample bytree': 0.5274511349524147}. Best is
trial 61 with value: 0.9085813746290029.
Best trial: 61. Best value: 0.908581: 64%
[50:00<27:53, 46.48s/it]
[I 2025-10-25 12:20:43,335] Trial 63 finished with value:
0.9073580811954643 and parameters: {'max depth': 9, 'n estimators':
662, 'learning rate': 0.014434221880692504, 'reg alpha':
0.16386086769002586, 'reg_lambda': 10.015392526217681, 'gamma':
0.33933559935134694, 'min child weight': 2, 'subsample':
0.8863932017766651, 'colsample bytree': 0.5048889014919622}. Best is
trial 61 with value: 0.9085813746290029.
Best trial: 61. Best value: 0.908581: 65%
[50:52<28:09, 48.27s/it]
[I 2025-10-25 12:21:35,790] Trial 64 finished with value:
0.9049118362851803 and parameters: {'max depth': 9, 'n estimators':
678, 'learning_rate': 0.009699566914819041, 'reg alpha':
0.18580287027412193, 'reg_lambda': 10.56085339249279, 'gamma':
0.3203504947191421, 'min child weight': 2, 'subsample':
0.8637234047482578, 'colsample_bytree': 0.4860771407148885}. Best is
trial 61 with value: 0.9085813746290029.
Best trial: 65. Best value: 0.908646: 66%
[51:42<27:33, 48.62s/it]
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[I 2025-10-25 12:22:25,230] Trial 65 finished with value:
0.9086457143176896 and parameters: {'max depth': 9, 'n estimators':
700, 'learning_rate': 0.019662200708138185, 'reg_alpha':
0.15783509135465426, 'reg lambda': 11.298508405688557, 'gamma':
0.3862661624452438, 'min_child_weight': 2, 'subsample':
0.895124702021648, 'colsample_bytree': 0.49215008300291513}. Best is
trial 65 with value: 0.9086457143176896.
Best trial: 65. Best value: 0.908646: 67%
[52:29<26:37, 48.41s/it]
[I 2025-10-25 12:23:13,139] Trial 66 finished with value:
0.9081306755760572 and parameters: {'max_depth': 9, 'n_estimators':
654, 'learning rate': 0.01683410259814006, 'reg alpha':
0.1587155343487758, 'reg_lambda': 10.603675976441396, 'gamma': 0.38627907512820997, 'min_child_weight': 2, 'subsample': 0.8931101325587198, 'colsample_bytree': 0.5151352156965179}. Best is
trial 65 with value: 0.9086457143176896.
Best trial: 65. Best value: 0.908646: 68%
[53:25<27:01, 50.68s/it]
[I 2025-10-25 12:24:09,120] Trial 67 finished with value:
0.90001910813109 and parameters: {'max_depth': 9, 'n estimators': 700,
'learning rate': 0.005978401302033651, 'reg alpha':
0.15215370049768176, 'reg lambda': 11.2466605970274, 'gamma':
0.3933619506429424, 'min child weight': 2, 'subsample':
0.8938133383436943, 'colsample bytree': 0.44725488880595643}. Best is
trial 65 with value: 0.9086457143176896.
Best trial: 65. Best value: 0.908646: 69%
[54:11<25:20, 49.06s/it]
[I 2025-10-25 12:24:54,388] Trial 68 finished with value:
0.9078087698860827 and parameters: {'max depth': 8, 'n estimators':
656, 'learning rate': 0.018971882065907275, 'reg_alpha':
0.17320525304125242, 'reg_lambda': 10.66907983150382, 'gamma':
0.3848913379983909, 'min child weight': 2, 'subsample':
0.8784802375890315, 'colsample bytree': 0.5137874493593642}. Best is
trial 65 with value: 0.9086457143176896.
Best trial: 65. Best value: 0.908646: 70%| | 70/100
[54:54<23:40, 47.35s/it]
[I 2025-10-25 12:25:37,751] Trial 69 finished with value:
0.9083238500770232 and parameters: {'max depth': 9, 'n estimators':
647, 'learning rate': 0.02611749898485379, 'reg alpha':
0.1596929435175385, 'reg_lambda': 10.263197117328204, 'gamma': 0.37344520025379335, 'min_child_weight': 2, 'subsample':
0.8662232575593309, 'colsample bytree': 0.4902960239622941}. Best is
trial 65 with value: 0.9086457143176896.
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Best trial: 65. Best value: 0.908646: 71%
[55:38<22:25, 46.41s/it]
[I 2025-10-25 12:26:21,960] Trial 70 finished with value:
0.9082595000260094 and parameters: {'max depth': 9, 'n estimators':
669, 'learning rate': 0.024427953199945996, 'reg alpha':
0.161245939045596, 'reg_lambda': 10.252680733329488, 'gamma': 0.3640799621197786, 'min_child_weight': 2, 'subsample':
0.8660734416697464, 'colsample bytree': 0.471180560646383}. Best is
trial 65 with value: 0.9086457143176896.
Best trial: 65. Best value: 0.908646: 72%
[56:23<21:27, 45.99s/it]
[I 2025-10-25 12:27:06,993] Trial 71 finished with value:
0.908130820648636 and parameters: {'max depth': 9, 'n estimators':
671, 'learning_rate': 0.025973884950344243, 'reg_alpha':
0.16368696147321793, 'reg_lambda': 10.23293393080099, 'gamma': 0.36825665208233194, 'min_child_weight': 2, 'subsample':
0.8599473802352052, 'colsample bytree': 0.47217519993550966}. Best is
trial 65 with value: 0.9086457143176896.
Best trial: 65. Best value: 0.908646: 73%| | 73/100
[57:13<21:14, 47.19s/it]
[I 2025-10-25 12:27:56,988] Trial 72 finished with value:
0.9061026749103813 and parameters: {'max depth': 9, 'n estimators':
646, 'learning rate': 0.013175703453581862, 'reg alpha':
0.17810711115749342, 'reg lambda': 10.263600230491488, 'gamma':
0.35547723010674, 'min child weight': 2, 'subsample':
0.8640475388050353, 'colsample bytree': 0.45929469461143096}. Best is
trial 65 with value: 0.9086457143176896.
Best trial: 65. Best value: 0.908646: 74%
[58:13<22:05, 50.97s/it]
[I 2025-10-25 12:28:56,774] Trial 73 finished with value:
0.8858876814935679 and parameters: {'max depth': 9, 'n estimators':
685, 'learning rate': 0.001084246507051776, 'reg alpha':
0.15822184632011585, 'reg_lambda': 10.01093908777242, 'gamma': 0.32678250327766045, 'min_child_weight': 2, 'subsample':
0.8997182874560639, 'colsample bytree': 0.49012822007104667}. Best is
trial 65 with value: 0.9086457143176896.
Best trial: 65. Best value: 0.908646: 75%
[58:57<20:20, 48.83s/it]
[I 2025-10-25 12:29:40,616] Trial 74 finished with value:
0.9082594482143742 and parameters: {'max depth': 9, 'n estimators':
651, 'learning rate': 0.022058759282821763, 'reg alpha':
0.16783150411792347, 'reg lambda': 10.910129790310773, 'gamma':
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0.3998228132885914, 'min_child_weight': 2, 'subsample':
0.8489636498928504, 'colsample bytree': 0.4391619250209322}. Best is
trial 65 with value: 0.9086457143176896.
Best trial: 65. Best value: 0.908646: 76%| | 76/100
[59:43<19:13, 48.08s/it]
[I 2025-10-25 12:30:26,931] Trial 75 finished with value:
0.9077121515486187 and parameters: {'max depth': 8, 'n estimators':
667, 'learning rate': 0.01826023990299992, 'reg alpha':
0.15456914241281874, 'reg lambda': 11.355755289910453, 'gamma':
0.3618677283660502, 'min_child_weight': 2, 'subsample':
0.8701071179869885, 'colsample_bytree': 0.4677165064758215}. Best is
trial 65 with value: 0.9086457143176896.
Best trial: 65. Best value: 0.908646: 77%
[1:00:28<18:00, 46.97s/it]
[I 2025-10-25 12:31:11,327] Trial 76 finished with value:
0.9082917372254785 and parameters: {'max depth': 9, 'n estimators':
633, 'learning rate': 0.028251443092805903, 'reg alpha':
0.1846880139791147, 'reg_lambda': 10.362023573012502, 'gamma': 0.3099717464400763, 'min_child_weight': 2, 'subsample':
0.8287148879595306, 'colsample bytree': 0.49015255813166647}. Best is
trial 65 with value: 0.9086457143176896.
Best trial: 65. Best value: 0.908646: 78%
[1:01:10<16:40, 45.46s/it]
[I 2025-10-25 12:31:53,261] Trial 77 finished with value:
0.9081308931849253 and parameters: {'max depth': 9, 'n estimators':
633, 'learning rate': 0.03162064940491306, 'reg alpha':
0.18277417499935278, 'reg_lambda': 10.741086594553874, 'gamma':
0.30006425855963814, 'min child weight': 2, 'subsample':
0.8307521043323196, 'colsample bytree': 0.42167121731313256}. Best is
trial 65 with value: 0.9086457143176896.
Best trial: 65. Best value: 0.908646: 79%
[1:01:43<14:39, 41.89s/it]
[I 2025-10-25 12:32:26,812] Trial 78 finished with value:
0.9077767088461733 and parameters: {'max depth': 8, 'n estimators':
640, 'learning rate': 0.0495594678448331, 'reg alpha':
0.17176309945479531, 'reg_lambda': 11.155072280342555, 'gamma':
0.3096203727614467, 'min child weight': 2, 'subsample':
0.8865240629984337, 'colsample_bytree': 0.490645603263449}. Best is
trial 65 with value: 0.9086457143176896.
Best trial: 65. Best value: 0.908646: 80%| 80/100
[1:02:40<15:25, 46.27s/it]
```

```
[I 2025-10-25 12:33:23,307] Trial 79 finished with value:
0.9054268128528502 and parameters: {'max depth': 10, 'n estimators':
643, 'learning_rate': 0.009499383114749259, 'reg_alpha':
0.1684523794397773, 'reg_lambda': 10.398127734123397, 'qamma':
0.318183206599427, 'min child weight': 2, 'subsample':
0.8766473410356742, 'colsample_bytree': 0.5032713420106371}. Best is
trial 65 with value: 0.9086457143176896.
Best trial: 65. Best value: 0.908646: 81%
[1:03:24<14:27, 45.66s/it]
[I 2025-10-25 12:34:07,527] Trial 80 finished with value:
0.908613601466145 and parameters: {'max depth': 9, 'n estimators':
662, 'learning rate': 0.02816873678663498, 'reg alpha':
0.1518756655294079, 'reg_lambda': 10.966183243829207, 'gamma': 0.33555965118892084, 'min_child_weight': 2, 'subsample': 0.8539883707998688, 'colsample_bytree': 0.4512503549631769}. Best is
trial 65 with value: 0.9086457143176896.
Best trial: 65. Best value: 0.908646: 82%
[1:04:07<13:28, 44.90s/it]
[I 2025-10-25 12:34:50,676] Trial 81 finished with value:
0.9084204684144874 and parameters: {'max depth': 9, 'n estimators':
661, 'learning rate': 0.028981870843272924, 'reg alpha':
0.15001325861522258, 'reg_lambda': 10.77687667853844, 'gamma':
0.33232766321402796, 'min child weight': 2, 'subsample':
0.854220484188128, 'colsample bytree': 0.4556315890065618}. Best is
trial 65 with value: 0.9086457143176896.
Best trial: 65. Best value: 0.908646: 83%
[1:04:55<13:00, 45.89s/it]
[I 2025-10-25 12:35:38,875] Trial 82 finished with value:
0.908227211014905 and parameters: {'max depth': 9, 'n estimators':
661, 'learning rate': 0.020044125424808063, 'reg_alpha':
0.15017477308105281, 'reg_lambda': 10.995203774779544, 'gamma':
0.33087734737566404, 'min_child_weight': 2, 'subsample':
0.8523948689900851, 'colsample bytree': 0.4583987206872346}. Best is
trial 65 with value: 0.9086457143176896.
Best trial: 65. Best value: 0.908646: 84%| 84/100
[1:05:38<11:57, 44.86s/it]
[I 2025-10-25 12:36:21,317] Trial 83 finished with value:
0.908034212673499 and parameters: {'max depth': 9, 'n estimators':
675, 'learning rate': 0.031208139832268153, 'reg alpha':
0.15395710996690187, 'reg_lambda': 10.76022080658729, 'gamma': 0.34722346747901084, 'min_child_weight': 2, 'subsample':
0.8841877333364909, 'colsample bytree': 0.4500286489366654}. Best is
trial 65 with value: 0.9086457143176896.
```

```
Best trial: 84. Best value: 0.908775: 85%| 85/100
[1:06:15<10:40, 42.71s/it]
[I 2025-10-25 12:36:59,028] Trial 84 finished with value:
0.9087746734778934 and parameters: {'max depth': 9, 'n estimators':
663, 'learning_rate': 0.0418335406255629, 'reg alpha':
0.17693694205214475, 'reg_lambda': 11.426826345687571, 'gamma':
0.3771891976408574, 'min child weight': 2, 'subsample':
0.8440084303923681, 'colsample bytree': 0.4800273186370416}. Best is
trial 84 with value: 0.9087746734778934.
Best trial: 84. Best value: 0.908775: 86%
[1:06:52<09:34, 41.01s/it]
[I 2025-10-25 12:37:36,058] Trial 85 finished with value:
0.9085815093392545 and parameters: {'max depth': 9, 'n estimators':
662, 'learning_rate': 0.04352836950313393, 'reg_alpha':
0.17379643279178741, 'reg_lambda': 11.926222547744798, 'gamma': 0.37934026595183445, 'min_child_weight': 2, 'subsample':
0.8560882762127379, 'colsample bytree': 0.44023342436687823}. Best is
trial 84 with value: 0.9087746734778934.
Best trial: 86. Best value: 0.908871: 87%
[1:07:31<08:43, 40.28s/it]
[I 2025-10-25 12:38:14,633] Trial 86 finished with value:
0.9088711778297599 and parameters: {'max depth': 10, 'n estimators':
661, 'learning rate': 0.04302114057824755, 'reg_alpha':
0.16504739942571847, 'reg_lambda': 11.455438665423085, 'gamma': 0.33967694603075804, 'min_child_weight': 2, 'subsample':
0.8428257489422021, 'colsample_bytree': 0.423022115421517}. Best is
trial 86 with value: 0.9088711778297599.
Best trial: 86. Best value: 0.908871: 88%| 88/100
[1:08:09<07:54, 39.56s/it]
[I 2025-10-25 12:38:52.522] Trial 87 finished with value:
0.9079055332961257 and parameters: {'max depth': 10, 'n estimators':
685, 'learning rate': 0.044620069510953646, 'reg alpha':
0.19489851326186772, 'reg_lambda': 12.471481734089446, 'gamma': 0.33506690385579174, 'min_child_weight': 2, 'subsample': 0.8367360778304148, 'colsample_bytree': 0.4140379466337263}. Best is
trial 86 with value: 0.9088711778297599.
Best trial: 86. Best value: 0.908871: 89%
[1:08:39<06:43, 36.70s/it]
[I 2025-10-25 12:39:22,556] Trial 88 finished with value:
0.9085170763896244 and parameters: {'max depth': 10, 'n estimators':
664, 'learning rate': 0.053867742203066, 'reg alpha':
0.17987899479846345, 'reg lambda': 11.949082366344804, 'gamma':
```

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0.4154086627247045, 'min_child_weight': 2, 'subsample':
0.8460511430312017, 'colsample bytree': 0.42153612912397925}. Best is
trial 86 with value: 0.9088711778297599.
Best trial: 86. Best value: 0.908871: 90%| 90/100
[1:09:06<05:37, 33.70s/it]
[I 2025-10-25 12:39:49,258] Trial 89 finished with value:
0.9082917683124597 and parameters: {'max_depth': 10, 'n_estimators':
653, 'learning rate': 0.07536982114930335, 'reg_alpha':
0.17675356865701614, 'reg lambda': 12.020214022406742, 'gamma':
0.41591477579458486, 'min child weight': 2, 'subsample':
0.8440122305195209, 'colsample bytree': 0.4022785659117564}. Best is
trial 86 with value: 0.9088711778297599.
Best trial: 86. Best value: 0.908871: 91%| 91/100
[1:09:39<05:01, 33.55s/it]
[I 2025-10-25 12:40:22,448] Trial 90 finished with value:
0.908195191424304 and parameters: {'max depth': 10, 'n estimators':
698, 'learning rate': 0.05425266454923625, 'reg alpha':
0.19332535241846951, 'reg lambda': 11.382427770155529, 'gamma':
0.39240688776628024, 'min_child_weight': 2, 'subsample':
0.8050855002518399, 'colsample bytree': 0.42235543864822667}. Best is
trial 86 with value: 0.9088711778297599.
Best trial: 86. Best value: 0.908871: 92%| 92/100
[1:10:16<04:37, 34.74s/it]
[I 2025-10-25 12:40:59,979] Trial 91 finished with value:
0.9086137569010507 and parameters: {'max depth': 10, 'n estimators':
665, 'learning rate': 0.04237510773791768, 'reg alpha':
0.16476823259098186, 'reg lambda': 11.886226498926556, 'gamma':
0.3761679018497835, 'min_child_weight': 2, 'subsample':
0.8533209117645377, 'colsample bytree': 0.4384722480637415}. Best is
trial 86 with value: 0.9088711778297599.
Best trial: 86. Best value: 0.908871: 93%
[1:10:52<04:05, 35.10s/it]
[I 2025-10-25 12:41:35,915] Trial 92 finished with value:
0.9082917786747868 and parameters: {'max depth': 10, 'n estimators':
665, 'learning rate': 0.043048130246260356, 'reg alpha':
0.16596243851184309, 'reg_lambda': 11.816129252274207, 'gamma':
0.3798639181431983, 'min child weight': 2, 'subsample':
0.8214836589609484, 'colsample bytree': 0.40444116885301146}. Best is
trial 86 with value: 0.9088711778297599.
Best trial: 86. Best value: 0.908871: 94%
[1:11:23<03:22, 33.69s/it]
```

```
[I 2025-10-25 12:42:06,305] Trial 93 finished with value:
0.9088390028042529 and parameters: {'max depth': 10, 'n estimators':
670, 'learning rate': 0.05790580181033439, 'reg alpha':
0.1800291674654304, 'reg lambda': 12.293870004182317, 'gamma':
0.4119419156653194, 'min_child_weight': 2, 'subsample':
0.8450745580073581, 'colsample_bytree': 0.42392522543603833}. Best is
trial 86 with value: 0.9088711\overline{7}78297599.
Best trial: 86. Best value: 0.908871: 95%| 95/100
[1:11:55<02:46, 33.29s/it]
[I 2025-10-25 12:42:38,651] Trial 94 finished with value:
0.9084527056139565 and parameters: {'max depth': 10, 'n estimators':
675, 'learning rate': 0.05178370990011197, 'reg alpha':
0.17867094899261055, 'reg_lambda': 12.365636265683882, 'gamma': 0.40826347875490704, 'min_child_weight': 2, 'subsample': 0.8451850883245602, 'colsample_bytree': 0.4258603295500482}. Best is
trial 86 with value: 0.9088711778297599.
Best trial: 86. Best value: 0.908871: 96%
[1:12:21<02:04, 31.04s/it]
[I 2025-10-25 12:43:04,465] Trial 95 finished with value:
0.907615616109771 and parameters: {'max depth': 10, 'n estimators':
671, 'learning rate': 0.06730973906999785, 'reg alpha':
0.19936261949851203, 'reg lambda': 13.158363537407013, 'gamma':
0.4407279488861012, 'min child weight': 2, 'subsample':
0.8366771924301526, 'colsample_bytree': 0.41010174606673017}. Best is
trial 86 with value: 0.9088711778297599.
Best trial: 86. Best value: 0.908871: 97%
[1:12:50<01:31, 30.41s/it]
[I 2025-10-25 12:43:33,392] Trial 96 finished with value:
0.9081628920508724 and parameters: {'max depth': 10, 'n estimators':
657, 'learning rate': 0.05967809651555692, 'reg alpha':
0.17361538393985906, 'reg_lambda': 12.798375894460724, 'gamma':
0.4817911190996681, 'min_child_weight': 2, 'subsample':
0.8127863406451266, 'colsample_bytree': 0.43929263529029555}. Best is
trial 86 with value: 0.9088711778297599.
Best trial: 86. Best value: 0.908871: 98%| 98/100
[1:13:12<00:55, 27.96s/it]
[I 2025-10-25 12:43:55,633] Trial 97 finished with value:
0.907937635785343 and parameters: {'max depth': 10, 'n estimators':
682, 'learning rate': 0.08250734160537601, 'reg alpha':
0.16896976459042032, 'reg_lambda': 11.552634242290747, 'gamma': 0.40500012160103854, 'min_child_weight': 2, 'subsample':
0.8249795245018422, 'colsample bytree': 0.43787115141853744}. Best is
trial 86 with value: 0.9088711778297599.
```

```
Best trial: 86. Best value: 0.908871: 99%| 99/100
[1:13:54<00:32, 32.33s/it]
[I 2025-10-25 12:44:38,177] Trial 98 finished with value:
0.9084526848893024 and parameters: {'max depth': 10, 'n estimators':
664, 'learning rate': 0.03848213226199682, 'reg alpha':
0.19014612270615863, 'reg_lambda': 12.177248568593928, 'gamma':
0.3749119747547328, 'min child weight': 2, 'subsample':
0.8750075234842908, 'colsample bytree': 0.42781154424411116}. Best is
trial 86 with value: 0.9088711778297599.
Best trial: 86. Best value: 0.908871: 100%
[1:14:26<00:00, 44.67s/it]
[I 2025-10-25 12:45:10,022] Trial 99 finished with value:
0.908034202311172 and parameters: {'max depth': 10, 'n estimators':
651, 'learning_rate': 0.040675704342585474, 'reg_alpha':
0.21668973193947874, 'reg_lambda': 12.61310855326125, 'gamma':
0.6889753371519796, 'min_child_weight': 2, 'subsample': 0.8555059582924543, 'colsample_bytree': 0.4438311874429509}. Best is
trial 86 with value: 0.9088711778297599.
Best CV accuracy: 0.90887
Best hyperparameters found:
  max depth: 10
  n estimators: 661
  learning rate: 0.04302114057824755
  reg alpha: 0.16504739942571847
  reg lambda: 11.455438665423085
  gamma: 0.33967694603075804
  min child weight: 2
  subsample: 0.8428257489422021
  colsample bytree: 0.423022115421517
Training final model with best parameters...
Preparing and preprocessing test data...
Test data preprocessing complete.
Generating test predictions...
Submission saved to:
/home/iiitb/Desktop/IIITB/ML/project first half/xgb maxdepth8 optimize
d.csv
Sample Submission:
      id
               WeightCategory
  15533
             Obesity Type III
1 15534
           Overweight Level I
2 15535 Overweight Level II
```

```
Obesity_Type_II
  15536
                Normal Weight
4 15537
import pandas as pd
import numpy as np
import time
import warnings
from sklearn.preprocessing import StandardScaler, LabelEncoder
from sklearn.model selection import RepeatedStratifiedKFold
import xgboost as xgb
import optuna
optuna.logging.set verbosity(optuna.logging.INFO)
warnings.filterwarnings("ignore")
file path base = '/home/iiitb/Desktop/IIITB/ML/project first half/'
df train = pd.read csv(file path base + 'train.csv')
df test = pd.read csv(file path base + 'test.csv')
X = df train.drop(['id', 'WeightCategory'], axis=1)
y = df train['WeightCategory']
categorical cols = X.select dtypes(include=['object']).columns
X = pd.get dummies(X, columns=categorical cols, drop first=True)
num features =
['Age','Height','Weight','FCVC','NCP','CH20','FAF','TUE']
num features present train = [col for col in num features if col in
X.columns1
scaler = StandardScaler()
X[num_features_present_train] =
scaler.fit transform(X[num features present train])
le = LabelEncoder()
y enc = le.fit transform(y)
def objective(trial):
    params = {
        'max depth': trial.suggest int('max depth', 7, 13),
        'n estimators': trial.suggest int('n estimators', 550, 900),
        'learning rate': trial.suggest float('learning rate', 0.008,
0.035, log=True),
        'reg alpha': trial.suggest float('reg alpha', 0.10, 0.5,
log=True),
        'reg lambda': trial.suggest_float('reg_lambda', 5.0, 25.0,
log=True),
         gamma': trial.suggest float('gamma', 0.25, 1),
        'min_child_weight': trial.suggest_int('min_child_weight', 1,
6),
        'subsample': trial.suggest float('subsample', 0.60, 0.95),
```

```
'colsample bytree': trial.suggest float('colsample bytree',
0.35, 1),
        'objective': 'multi:softmax',
        'num class': len(le.classes ),
        'use label encoder': False,
        'eval_metric': 'mlogloss',
        'random state': 42,
        'verbosity': 0,
        'n jobs': -1,
        'early stopping rounds': 50,
        'tree method': 'hist'
    }
    cv = RepeatedStratifiedKFold(n splits=5, n repeats=2,
random state=42)
    scores = []
    for train idx, val idx in cv.split(X, y enc):
        X tr, X val = X.iloc[train idx], X.iloc[val idx]
        y tr, y val = y enc[train idx], y enc[val idx]
        model = xgb.XGBClassifier(**params)
        model.fit(X_tr, y_tr, eval_set=[(X_val, y_val)],
verbose=False)
        scores.append(model.score(X val, y val))
    return np.mean(scores)
study = optuna.create study(direction='maximize',
sampler=optuna.samplers.TPESampler(seed=42))
study.optimize(objective, n trials=100, show progress bar=True)
best params from optuna = study.best params.copy()
best params final = best params from optuna.copy()
best params final.update({
    'objective': 'multi:softmax',
    'num class': len(le.classes_),
    'use label encoder': False,
    'eval metric': 'mlogloss',
    'random state': 42,
    'verbosity': 0,
    'n jobs': -1,
    'tree method': 'hist'
})
final model = xqb.XGBClassifier(**best params final)
final_model.fit(X, y_enc, verbose=False)
X_test = df_test.drop('id', axis=1)
test ids = df test['id']
original train df = df train.drop(['id','WeightCategory'], axis=1)
```

```
for col in categorical cols:
    if col not in X test.columns:
        X test[col] = 'Missing'
    train categories =
original train df[col].astype('category').cat.categories
    X test[col] = pd.Categorical(X test[col],
categories=train categories)
X test = pd.get dummies(X test, columns=categorical cols,
drop first=True)
X test = X test.reindex(columns=X.columns, fill value=0)
num features present test = [col for col in num features if col in
X test.columns1
cols to scale test = [col for col in num features present train if col
in num features present test]
X test[cols to scale test] =
scaler.transform(X test[cols to scale test])
y test pred enc = final model.predict(X test)
y test pred = le.inverse transform(y test pred enc)
submission = pd.DataFrame({'id': test ids, 'WeightCategory':
y test pred})
submission.to csv(file path base + 'xgb refined peak search.csv',
index=False)
[I 2025-10-25 16:12:11,981] A new study created in memory with name:
no-name-216bc2ec-8b69-487c-9131-bb19a4e81daf
Loading Training and Test data...
Data loaded successfully.
Starting Data Preprocessing...
Preprocessing and Encoding complete.
Refined search around your peak performance (0.91267 output
accuracy)...
Starting REFINED OPTUNA search (tightened around peak performance)...
Best trial: 0. Best value: 0.906972: 1%| | 1/100
[01:44<2:51:46, 104.10s/it]
[I 2025-10-25 16:13:56,081] Trial 0 finished with value:
0.9069718876284385 and parameters: {'max depth': 9, 'n estimators':
883, 'learning rate': 0.02356579209151413, 'reg alpha':
0.26208630215377515, 'reg lambda': 6.427207987637222, 'gamma':
0.36699589025215196, 'min_child_weight': 1, 'subsample': 0.9031616510212273, 'colsample_bytree': 0.7407247576330858}. Best is
trial 0 with value: 0.9069718876284385.
```

```
Best trial: 1. Best value: 0.907133: 2%| | 2/100
[02:55<2:18:58, 85.09s/it]
[I 2025-10-25 16:15:07,854] Trial 1 finished with value:
0.9071328663792435 and parameters: {'max depth': 11, 'n estimators':
557, 'learning rate': 0.0334796508218928, 'reg alpha':
0.3818145165896869, 'reg_lambda': 7.037018186923738, 'gamma': 0.38636872540532546, 'min_child_weight': 2, 'subsample':
0.7064847850358382, 'colsample bytree': 0.6910916805609546}. Best is
trial 1 with value: 0.9071328663792435.
Best trial: 2. Best value: 0.907809: 3%|| | 3/100
[04:15<2:13:11, 82.39s/it]
[I 2025-10-25 16:16:27,032] Trial 2 finished with value:
0.907808832060045 and parameters: {'max depth': 10, 'n estimators':
652, 'learning_rate': 0.0197366875932559, 'reg_alpha':
0.12517051076140215, 'reg_lambda': 8.001480489146246, 'gamma':
0.5247713824702688, 'min_child_weight': 3, 'subsample': 0.8748115864875547, 'colsample_bytree': 0.4797879584029338}. Best is
trial 2 with value: 0.907808832060045.
[06:27<2:43:16, 102.05s/it]
[I 2025-10-25 16:18:39,230] Trial 3 finished with value:
0.9037529032649829 and parameters: {'max depth': 10, 'n estimators':
757, 'learning rate': 0.008567688777470862, 'reg alpha':
0.2658616083788978, 'reg_lambda': 6.579020684258305, 'gamma':
0.29878869473895964, 'min child weight': 6, 'subsample':
0.9379712115760958, 'colsample bytree': 0.8754582762756997}. Best is
trial 2 with value: 0.907808832060045.
Best trial: 4. Best value: 0.907938: 5%| | 5/100
[07:42<2:26:22, 92.45s/it]
[I 2025-10-25 16:19:54,639] Trial 4 finished with value:
0.9079375321620727 and parameters: {'max depth': 9, 'n estimators':
584, 'learning rate': 0.021961820777692202, 'reg alpha':
0.20307356380344246, 'reg lambda': 6.08514694186939, 'gamma':
0.6213826825834526, 'min_child_weight': 1, 'subsample': 0.9182621407275737, 'colsample_bytree': 0.518206988040011}. Best is
trial 4 with value: 0.9079375321620727.
[09:03<2:18:37, 88.48s/it]
[I 2025-10-25 16:21:15,418] Trial 5 finished with value:
0.9035596873147085 and parameters: {'max depth': 11, 'n estimators':
659, 'learning rate': 0.017236225881449775, 'reg alpha':
0.241064959021\overline{7}1612, 'reg_lambda': 6.732521111373228, 'gamma':
```

```
0.9771884708234189, 'min_child_weight': 5, 'subsample':
0.9288246295474661, 'colsample bytree': 0.9316377777779717}. Best is
trial 4 with value: 0.9079375321620727.
[11:37<2:50:09, 109.78s/it]
[I 2025-10-25 16:23:49,054] Trial 6 finished with value:
0.9047186410595603 and parameters: {'max_depth': 11, 'n_estimators':
873, 'learning rate': 0.009116155803152578, 'reg_alpha':
0.1370838023704289, 'reg_lambda': 5.377526047186648, 'gamma':
0.49399774807244823, 'min_child_weight': 3, 'subsample':
0.6949721611208636, 'colsample bytree': 0.8886793809487541}. Best is
trial 4 with value: 0.9079375321620727.
[13:04<2:37:24, 102.66s/it]
[I 2025-10-25 16:25:16,464] Trial 7 finished with value:
0.9065855282641797 and parameters: {'max depth': 9, 'n estimators':
648, 'learning rate': 0.017821582611824243, 'reg alpha':
0.12545899554294088, 'reg_lambda': 18.18367378104933, 'gamma': 0.30591298275982814, 'min_child_weight': 6, 'subsample':
0.8702856692538301, 'colsample bytree': 0.47916519299721205}. Best is
trial 4 with value: 0.9079375321620727.
[14:53<2:38:46, 104.69s/it]
[I 2025-10-25 16:27:05,619] Trial 8 finished with value:
0.9052014218770692 and parameters: {'max depth': 7, 'n estimators':
836, 'learning rate': 0.022707537186867518, 'reg alpha':
0.3232616187892149, 'reg lambda': 17.300746469980602, 'gamma':
0.30553348880056774, 'min_child_weight': 3, 'subsample': 0.6405541708337954, 'colsample_bytree': 0.9110172268191358}. Best is
trial 4 with value: 0.9079375321620727.
[16:35<2:35:54, 103.94s/it]
[I 2025-10-25 16:28:47,871] Trial 9 finished with value:
0.9047831050961717 and parameters: {'max_depth': 11, 'n_estimators':
666, 'learning rate': 0.008786774781018889, 'reg alpha':
0.16495569532358353, 'reg_lambda': 8.438463813264892, 'qamma':
0.7972046337535481, 'min child weight': 4, 'subsample':
0.9105244599017143, 'colsample bytree': 0.656939701355267}. Best is
trial 4 with value: 0.9079375321620727.
[17:54<2:22:38, 96.16s/it]
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[I 2025-10-25 16:30:06,409] Trial 10 finished with value:
0.9014997499570481 and parameters: {'max depth': 13, 'n estimators':
552, 'learning_rate': 0.01259839512063638, 'reg_alpha':
0.18372861301527957, 'reg lambda': 11.850856640416291, 'gamma':
0.7055352193360935, 'min_child_weight': 1, 'subsample': 0.8147467768264255, 'colsample_bytree': 0.3808050830766421}. Best is
trial 4 with value: 0.9079375321620727.
[19:13<2:13:17, 90.89s/it]
[I 2025-10-25 16:31:25,222] Trial 11 finished with value:
0.9078732546473482 and parameters: {'max depth': 8, 'n estimators':
609, 'learning rate': 0.02513671040084731, 'reg alpha':
0.10359848697743113, 'reg lambda': 10.0593984202376, 'gamma':
0.554487972334768, 'min_child_weight': 2, 'subsample':
0.8365500497745305, 'colsample bytree': 0.5280509317503476}. Best is
trial 4 with value: 0.9079375321620727.
[20:17<2:00:12, 82.90s/it]
[I 2025-10-25 16:32:29,739] Trial 12 finished with value:
0.9072294950790345 and parameters: {'max_depth': 7, 'n_estimators':
600, 'learning rate': 0.030733257640488496, 'reg alpha':
0.10185300223488546, 'reg lambda': 11.708116548908446, 'gamma':
0.6220309841272857, 'min child weight': 2, 'subsample':
0.8150754097662947, 'colsample_bytree': 0.5552337279361557}. Best is
trial 4 with value: 0.9079375321620727.
[21:35<1:56:36, 81.35s/it]
[I 2025-10-25 16:33:47,514] Trial 13 finished with value:
0.9066500855617345 and parameters: {'max depth': 8, 'n estimators':
730, 'learning rate': 0.025638005040063923, 'reg alpha':
0.49948720542730835, 'reg_lambda': 9.781102264750771, 'gamma':
0.7591716681521044, 'min child weight': 2, 'subsample':
0.8116105988209544, 'colsample bytree': 0.5923175054036708}. Best is
trial 4 with value: 0.9079375321620727.
Best trial: 4. Best value: 0.907938: 15%
[22:51<1:52:58, 79.75s/it]
[I 2025-10-25 16:35:03,550] Trial 14 finished with value:
0.899343080276326 and parameters: {'max depth': 8, 'n estimators':
605, 'learning rate': 0.012948942464580834, 'reg alpha':
0.19534104928944196, 'reg lambda': 24.7681214837072, 'gamma':
0.581696436491557, 'min_child_weight': 1, 'subsample':
0.7581347467571893, 'colsample_bytree': 0.3914360367489744}. Best is
trial 4 with value: 0.9079375321620727.
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[24:11<1:51:37, 79.74s/it]
[I 2025-10-25 16:36:23,263] Trial 15 finished with value:
0.9081629024131994 and parameters: {'max depth': 8, 'n estimators':
782, 'learning rate': 0.025848790624534594, 'reg alpha':
0.10178437486541467, 'reg_lambda': 5.35125213914853, 'gamma':
0.9070610077857587, 'min_child_weight': 2, 'subsample':
0.8646946696200327, 'colsample bytree': 0.49934500093537454}. Best is
trial 15 with value: 0.9081629024131994.
Best trial: 15. Best value: 0.908163: 17%
[25:31<1:50:22, 79.79s/it]
[I 2025-10-25 16:37:43,192] Trial 16 finished with value:
0.9064891378979109 and parameters: {'max depth': 9, 'n estimators':
785, 'learning rate': 0.02904069829932038, 'reg alpha':
0.15206197155881412, 'reg_lambda': 5.026251689922809, 'gamma':
0.9158308531271067, 'min_child_weight': 1, 'subsample': 0.7663799011321069, 'colsample_bytree': 0.7832481655500173}. Best is
trial 15 with value: 0.9081629024131994.
Best trial: 15. Best value: 0.908163: 18%
[27:08<1:56:02, 84.91s/it]
[I 2025-10-25 16:39:20,006] Trial 17 finished with value:
0.9058131514924549 and parameters: {'max depth': 7, 'n estimators':
801, 'learning rate': 0.014164921270315186, 'reg alpha':
0.214836862266134, 'reg_lambda': 5.59973309443534, 'gamma':
0.8469080827014867, 'min_child_weight': 4, 'subsample':
0.876756396377507, 'colsample_bytree': 0.4466365937146537}. Best is
trial 15 with value: 0.9081629024131994.
Best trial: 15. Best value: 0.908163: 19%
[28:42<1:58:19, 87.65s/it]
[I 2025-10-25 16:40:54,038] Trial 18 finished with value:
0.9071651761150019 and parameters: {'max depth': 13, 'n estimators':
713, 'learning rate': 0.020672539904208906, 'reg alpha':
0.32226248713477057, 'reg lambda': 8.24457262829307, 'gamma':
0.6936382532467943, 'min_child_weight': 2, 'subsample':
0.8543895238590868, 'colsample bytree': 0.6095415214072242}. Best is
trial 15 with value: 0.9081629024131994.
Best trial: 15. Best value: 0.908163: 20%
[30:18<2:00:33, 90.41s/it]
[I 2025-10-25 16:42:30,891] Trial 19 finished with value:
0.9059417686958658 and parameters: {'max depth': 9, 'n estimators':
700, 'learning rate': 0.014809727882262941, 'reg alpha':
0.4930296366738582, 'reg lambda': 14.442779175673301, 'qamma':
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0.45860243842985227, 'min child weight': 1, 'subsample':
0.9493440050174281, 'colsample bytree': 0.43325559688525617}. Best is
trial 15 with value: 0.9081629024131994.
Best trial: 15. Best value: 0.908163: 21%
[31:43<1:56:54, 88.79s/it]
[I 2025-10-25 16:43:55,911] Trial 20 finished with value:
0.9051048657135675 and parameters: {'max_depth': 8, 'n_estimators':
825, 'learning rate': 0.02801707642272915, 'reg alpha':
0.17014354361054632, 'reg lambda': 6.1347329909251975, 'gamma':
0.8789403000294367, 'min_child_weight': 3, 'subsample': 0.6040673365696436, 'colsample_bytree': 0.9959978102823107}. Best is
trial 15 with value: 0.9081629024131994.
Best trial: 15. Best value: 0.908163: 22%
[32:59<1:50:06, 84.70s/it]
[I 2025-10-25 16:45:11,055] Trial 21 finished with value:
0.907744409472742 and parameters: {'max depth': 8, 'n estimators':
597, 'learning rate': 0.025785897565355053, 'reg alpha':
0.11083576356497737, 'reg lambda': 9.988708139530647, 'gamma':
0.5683477891363193, 'min_child_weight': 2, 'subsample':
0.8418184536840471, 'colsample bytree': 0.531018175147829}. Best is
trial 15 with value: 0.9081629024131994.
Best trial: 15. Best value: 0.908163: 23%
[34:01<1:40:16, 78.13s/it]
[I 2025-10-25 16:46:13,865] Trial 22 finished with value:
0.9074869885440331 and parameters: {'max depth': 10, 'n estimators':
621, 'learning rate': 0.03489848162840784, 'reg alpha':
0.11012979722647415, 'reg_lambda': 7.4318647189927685, 'gamma':
0.6733015768268779, 'min_child_weight': 2, 'subsample':
0.906150232246568, 'colsample bytree': 0.5276351723801241}. Best is
trial 15 with value: 0.9081629024131994.
Best trial: 15. Best value: 0.908163: 24%
[35:08<1:34:37, 74.70s/it]
[I 2025-10-25 16:47:20,560] Trial 23 finished with value:
0.9059741198809327 and parameters: {'max_depth': 8, 'n_estimators':
581, 'learning rate': 0.02143888565632389, 'reg alpha':
0.12604650565878345, 'reg_lambda': 9.607538938478038, 'gamma':
0.9979895189612094, 'min child weight': 1, 'subsample':
0.8295162888794417, 'colsample_bytree': 0.6197503777514091}. Best is
trial 15 with value: 0.9081629024131994.
Best trial: 24. Best value: 0.908485: 25%
[36:37<1:38:39, 78.93s/it]
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[I 2025-10-25 16:48:49,352] Trial 24 finished with value:
0.9084848391901552 and parameters: {'max depth': 9, 'n estimators':
685, 'learning_rate': 0.02484778229702568, 'reg_alpha':
0.1403708164744048, 'reg_lambda': 5.954487209944323, 'gamma':
0.43359680106881915, 'min_child_weight': 2, 'subsample':
0.8984218325832057, 'colsample_bytree': 0.4928788845344612}. Best is
trial 24 with value: 0.9084848391901552.
Best trial: 24. Best value: 0.908485: 26%
[38:12<1:43:19, 83.78s/it]
[I 2025-10-25 16:50:24,462] Trial 25 finished with value:
0.906778796026089 and parameters: {'max depth': 9, 'n estimators':
744, 'learning rate': 0.019487814903471772, 'reg alpha':
0.1459133754528276, 'reg_lambda': 5.749312073401608, 'gamma': 0.43010724219161345, 'min_child_weight': 3, 'subsample':
0.8954639113169169, 'colsample_bytree': 0.35199828153403956}. Best is
trial 24 with value: 0.9084848391901552.
Best trial: 24. Best value: 0.908485: 27%
[39:25<1:38:06, 80.64s/it]
[I 2025-10-25 16:51:37,763] Trial 26 finished with value:
0.9078411003464956 and parameters: {'max depth': 10, 'n estimators':
689, 'learning_rate': 0.02904101865524286, 'reg_alpha':
0.21087904798553125, 'reg lambda': 5.2533912630229835, 'gamma':
0.7680763085093079, 'min child weight': 4, 'subsample':
0.7924650184058997, 'colsample_bytree': 0.45887644869079974}. Best is
trial 24 with value: 0.9084848391901552.
Best trial: 24. Best value: 0.908485: 28%
[41:06<1:43:59, 86.66s/it]
[I 2025-10-25 16:53:18,469] Trial 27 finished with value:
0.907969831535504 and parameters: {'max depth': 12, 'n estimators':
789, 'learning rate': 0.022926381560483242, 'reg alpha':
0.12118921644234974, 'reg_lambda': 5.004239694096067, 'gamma':
0.6167571136525614, 'min_child_weight': 1, 'subsample':
0.8854307884585234, 'colsample bytree': 0.5631784493551737}. Best is
trial 24 with value: 0.9084848391901552.
Best trial: 24. Best value: 0.908485: 29%| | 29/100
[43:13<1:56:52, 98.77s/it]
[I 2025-10-25 16:55:25,493] Trial 28 finished with value:
0.9079376668723244 and parameters: {'max depth': 13, 'n estimators':
768, 'learning rate': 0.01845316781806757, 'reg alpha':
0.1170196373589458, 'reg_lambda': 5.197805670506047, 'gamma': 0.3831355543541858, 'min_child_weight': 2, 'subsample':
0.7390693401513682, 'colsample bytree': 0.5698207715563748}. Best is
trial 24 with value: 0.9084848391901552.
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Best trial: 24. Best value: 0.908485: 30%
[45:15<2:03:26, 105.81s/it]
[I 2025-10-25 16:57:27,741] Trial 29 finished with value:
0.9060062430948044 and parameters: {'max depth': 12, 'n estimators':
891, 'learning_rate': 0.023924683528404145, 'reg alpha':
0.13836507242583268, 'reg_lambda': 6.107541638140267, 'gamma':
0.2593830034499103, 'min child weight': 1, 'subsample':
0.8939434596635647, 'colsample bytree': 0.7401567570884009}. Best is
trial 24 with value: 0.9084848391901552.
Best trial: 24. Best value: 0.908485: 31%
[46:54<1:59:07, 103.59s/it]
[I 2025-10-25 16:59:06,141] Trial 30 finished with value:
0.9075836068814972 and parameters: {'max depth': 12, 'n estimators':
851, 'learning_rate': 0.02721289817386966, 'reg_alpha':
0.15981797386593125, 'reg_lambda': 7.148283704587625, 'gamma':
0.4806203007203788, 'min_child_weight': 5, 'subsample':
0.868686700796856, 'colsample bytree': 0.6566495284167557}. Best is
trial 24 with value: 0.9084848391901552.
Best trial: 24. Best value: 0.908485: 32%
[49:08<2:08:00, 112.94s/it]
[I 2025-10-25 17:01:20,909] Trial 31 finished with value:
0.9081951706996498 and parameters: {'max depth': 12, 'n estimators':
773, 'learning rate': 0.016028921603986026, 'reg_alpha':
0.11688338632506916, 'reg lambda': 5.471356486600291, 'gamma':
0.40720239828559246, 'min child weight': 2, 'subsample':
0.7153144372545432, 'colsample_bytree': 0.5631977606671494}. Best is
trial 24 with value: 0.9084848391901552.
Best trial: 24. Best value: 0.908485: 33%| | 33/100
[51:15<2:10:43, 117.06s/it]
[I 2025-10-25 17:03:27,581] Trial 32 finished with value:
0.908098614536148 and parameters: {'max_depth': 12, 'n_estimators':
805, 'learning rate': 0.015875937881374046, 'reg alpha':
0.10085311969763104, 'req lambda': 5.0459622416165315, 'qamma':
0.4176100530581869, 'min_child_weight': 2, 'subsample':
0.7026146525740585, 'colsample bytree': 0.4943172351955355}. Best is
trial 24 with value: 0.9084848391901552.
Best trial: 24. Best value: 0.908485: 34%
[52:46<2:00:07, 109.20s/it]
[I 2025-10-25 17:04:58,455] Trial 33 finished with value:
0.9080985212752047 and parameters: {'max depth': 12, 'n estimators':
812, 'learning rate': 0.015536077553084334, 'reg alpha':
0.10116764280685461, 'reg_lambda': 5.9066300465106325, 'gamma':
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0.43785878572085724, 'min child weight': 2, 'subsample':
0.6852300940675702, 'colsample bytree': 0.41899888438930233}. Best is
trial 24 with value: 0.9084848391901552.
Best trial: 24. Best value: 0.908485: 35%| | 35/100
[53:49<1:43:22, 95.42s/it]
[I 2025-10-25 17:06:01,703] Trial 34 finished with value:
0.9067787338521267 and parameters: {'max_depth': 10, 'n_estimators':
764, 'learning rate': 0.010346400626700166, 'reg_alpha':
0.13569028016804033, 'reg lambda': 7.466416680302634, 'gamma':
0.3563674815431511, 'min_child_weight': 3, 'subsample': 0.7330265299780504, 'colsample_bytree': 0.5073516867562821}. Best is
trial 24 with value: 0.9084848391901552.
Best trial: 24. Best value: 0.908485: 36%
[54:53<1:31:30, 85.78s/it]
[I 2025-10-25 17:07:05,005] Trial 35 finished with value:
0.9068108985153064 and parameters: {'max depth': 12, 'n estimators':
739, 'learning rate': 0.01124744210153758, 'reg alpha':
0.11180944755069606, 'reg lambda': 6.3737084732327896, 'gamma':
0.3537979962996881, 'min_child_weight': 2, 'subsample':
0.6613805756465294, 'colsample bytree': 0.4867766300871167}. Best is
trial 24 with value: 0.9084848391901552.
Best trial: 24. Best value: 0.908485: 37%
[56:01<1:24:30, 80.48s/it]
[I 2025-10-25 17:08:13,108] Trial 36 finished with value:
0.9059096869313026 and parameters: {'max depth': 11, 'n estimators':
860, 'learning rate': 0.016954974366089462, 'reg alpha':
0.13073594722177978, 'reg lambda': 6.869891547765792, 'gamma':
0.5100289527034789, 'min child weight': 3, 'subsample':
0.7249218353778535, 'colsample bytree': 0.7356054888129702}. Best is
trial 24 with value: 0.9084848391901552.
Best trial: 24. Best value: 0.908485: 38%
[57:16<1:21:40, 79.04s/it]
[I 2025-10-25 17:09:28,791] Trial 37 finished with value:
0.9072938451300482 and parameters: {'max_depth': 13, 'n_estimators':
785, 'learning rate': 0.015356842026105455, 'reg alpha':
0.11863491411195193, 'reg_lambda': 5.592483195399902, 'gamma':
0.4219760068938336, 'min child weight': 2, 'subsample':
0.7105621418665209, 'colsample_bytree': 0.6907381466803653}. Best is
trial 24 with value: 0.9084848391901552.
Best trial: 24. Best value: 0.908485: 39%| 39/100
[58:07<1:11:41, 70.52s/it]
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[I 2025-10-25 17:10:19,437] Trial 38 finished with value:
0.9080663566120247 and parameters: {'max depth': 11, 'n estimators':
679, 'learning_rate': 0.019698607171433385, 'reg_alpha':
0.10706276436539375, 'reg lambda': 6.587672769342061, 'gamma':
0.4021869201555834, 'min_child_weight': 3, 'subsample': 0.6753529191256882, 'colsample_bytree': 0.48900817017277926}. Best is
trial 24 with value: 0.9084848391901552.
Best trial: 24. Best value: 0.908485: 40%
[58:52<1:02:50, 62.83s/it]
[I 2025-10-25 17:11:04,335] Trial 39 finished with value:
0.9075191635695399 and parameters: {'max depth': 9, 'n estimators':
709, 'learning rate': 0.03204939757065874, 'reg alpha':
0.14673728802595565, 'reg lambda': 7.869631993688132, 'gamma':
0.2515876284355687, 'min_child_weight': 2, 'subsample':
0.6396482483588831, 'colsample bytree': 0.6336746770869562}. Best is
trial 24 with value: 0.9084848391901552.
Best trial: 24. Best value: 0.908485: 41%
[1:00:00<1:03:29, 64.56s/it]
[I 2025-10-25 17:12:12,935] Trial 40 finished with value:
0.9080663669743518 and parameters: {'max depth': 10, 'n estimators':
826, 'learning_rate': 0.013227955692783223, 'reg_alpha':
0.2405813763847982, 'reg lambda': 5.535397764918698, 'gamma':
0.3466114558064923, 'min_child_weight': 3, 'subsample':
0.7816053860513906, 'colsample bytree': 0.5800433160140122}. Best is
trial 24 with value: 0.9084848391901552.
Best trial: 24. Best value: 0.908485: 42%
[1:01:01<1:01:13, 63.34s/it]
[I 2025-10-25 17:13:13,405] Trial 41 finished with value:
0.9081307688370008 and parameters: {'max depth': 12, 'n estimators':
807, 'learning rate': 0.015700394841631504, 'reg alpha':
0.10088757003323504, 'reg_lambda': 5.827138326556239, 'gamma':
0.4537131262835252, 'min_child_weight': 2, 'subsample':
0.6858013555127545, 'colsample_bytree': 0.4122845108604484}. Best is
trial 24 with value: 0.9084848391901552.
Best trial: 24. Best value: 0.908485: 43%| 43/100
[1:01:54<57:08, 60.14s/it]
[I 2025-10-25 17:14:06,091] Trial 42 finished with value:
0.9065855386265067 and parameters: {'max depth': 12, 'n estimators':
757, 'learning rate': 0.016152328034361795, 'reg alpha':
0.11405483378272611, 'reg lambda': 6.484762390944499, 'gamma':
0.5336934068648997, 'min_child_weight': 2, 'subsample':
0.6419234581258204, 'colsample bytree': 0.3986653642541822}. Best is
trial 24 with value: 0.9084848391901552.
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Best trial: 24. Best value: 0.908485: 44%
[1:02:57<57:05, 61.17s/it]
[I 2025-10-25 17:15:09,668] Trial 43 finished with value:
0.9078409552739167 and parameters: {'max depth': 12, 'n estimators':
800, 'learning rate': 0.014389158219015475, 'reg alpha':
0.10062826461636146, 'reg_lambda': 6.017373860379889, 'gamma':
0.4772817955837652, 'min child weight': 2, 'subsample':
0.7079033592679838, 'colsample bytree': 0.46596994017755894}. Best is
trial 24 with value: 0.9084848391901552.
Best trial: 24. Best value: 0.908485: 45%
[1:04:04<57:33, 62.80s/it]
[I 2025-10-25 17:16:16,258] Trial 44 finished with value:
0.9078088735093534 and parameters: {'max depth': 13, 'n estimators':
772, 'learning rate': 0.017608093038199314, 'reg alpha':
0.12217553591498549, 'reg_lambda': 5.500364273584554, 'gamma': 0.31481440622537527, 'min_child_weight': 1, 'subsample':
0.6691498496703852, 'colsample bytree': 0.41885681786163587}. Best is
trial 24 with value: 0.9084848391901552.
Best trial: 24. Best value: 0.908485: 46%
[1:05:10<57:31, 63.91s/it]
[I 2025-10-25 17:17:22,762] Trial 45 finished with value:
0.9060062741817856 and parameters: {'max depth': 11, 'n estimators':
844, 'learning rate': 0.011888173634809695, 'reg alpha':
0.10676691810248651, 'reg_lambda': 5.005028904262353, 'gamma':
0.4071718833633777, 'min_child_weight': 3, 'subsample': 0.7438621985910865, 'colsample_bytree': 0.36451297506946684}. Best is
trial 24 with value: 0.9084848391901552.
Best trial: 24. Best value: 0.908485: 47%
[1:06:09<55:07, 62.40s/it]
[I 2025-10-25 17:18:21,657] Trial 46 finished with value:
0.908034223035826 and parameters: {'max depth': 12, 'n estimators':
817, 'learning rate': 0.018599671081470877, 'reg alpha':
0.12956171461130836, 'reg lambda': 9.112787492183026, 'gamma':
0.45650784000043293, 'min_child_weight': 2, 'subsample':
0.9243431426261761, 'colsample bytree': 0.5028782008202823}. Best is
trial 24 with value: 0.9084848391901552.
Best trial: 24. Best value: 0.908485: 48%
[1:06:57<50:22, 58.13s/it]
[I 2025-10-25 17:19:09,796] Trial 47 finished with value:
0.9053625249757987 and parameters: {'max depth': 7, 'n estimators':
749, 'learning rate': 0.01377845940253632, 'reg alpha':
0.115026477778\overline{2}2325, 'reg_lambda': 6.943750393896713, 'gamma':
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0.5177839683548311, 'min_child_weight': 5, 'subsample':
0.6968069192343821, 'colsample bytree': 0.44141112646424596}. Best is
trial 24 with value: 0.9084848391901552.
Best trial: 24. Best value: 0.908485: 49%| 49/100
[1:07:37<44:41, 52.58s/it]
[I 2025-10-25 17:19:49,445] Trial 48 finished with value:
0.906167221845609 and parameters: {'max_depth': 9, 'n_estimators':
637, 'learning rate': 0.015841917274040503, 'reg alpha':
0.181246573804<del>0</del>3405, 'reg lambda': 12.689715198878346, 'gamma':
0.9481420959583894, 'min_child_weight': 1, 'subsample': 0.7196626836366852, 'colsample_bytree': 0.5429517275559007}. Best is
trial 24 with value: 0.9084848391901552.
Best trial: 24. Best value: 0.908485: 50%
[1:08:28<43:32, 52.26s/it]
[I 2025-10-25 17:20:40,937] Trial 49 finished with value:
0.9071971542562947 and parameters: {'max depth': 11, 'n estimators':
724, 'learning rate': 0.020333420493706456, 'reg alpha':
0.2872152763431139, 'reg_lambda': 8.77209418927049, 'gamma': 0.32650441497949245, 'min_child_weight': 2, 'subsample':
0.6548795164574156, 'colsample bytree': 0.40711789095455986}. Best is
trial 24 with value: 0.9084848391901552.
Best trial: 24. Best value: 0.908485: 51%
[1:09:39<47:14, 57.86s/it]
[I 2025-10-25 17:21:51,860] Trial 50 finished with value:
0.9052658859136805 and parameters: {'max depth': 11, 'n estimators':
805, 'learning rate': 0.016330304359504305, 'reg alpha':
0.10058788529114854, 'reg_lambda': 5.813251646063923, 'gamma':
0.38036731704426807, 'min_child_weight': 3, 'subsample':
0.756092438365324, 'colsample bytree': 0.8225581174693308}. Best is
trial 24 with value: 0.9084848391901552.
Best trial: 24. Best value: 0.908485: 52%
[1:10:45<48:08, 60.18s/it]
[I 2025-10-25 17:22:57,461] Trial 51 finished with value:
0.9083238811640044 and parameters: {'max_depth': 12, 'n_estimators':
866, 'learning rate': 0.015123221702444714, 'reg alpha':
0.10109941203335737, 'reg_lambda': 5.95989401499616, 'gamma': 0.4469578829209984, 'min_child_weight': 2, 'subsample':
0.686650242317674, 'colsample bytree': 0.47180420501155834}. Best is
trial 24 with value: 0.9084848391901552.
Best trial: 24. Best value: 0.908485: 53%| | 53/100
[1:11:47<47:37, 60.80s/it]
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[I 2025-10-25 17:23:59,725] Trial 52 finished with value:
0.9081629024131994 and parameters: {'max depth': 12, 'n estimators':
875, 'learning_rate': 0.014758466588177456, 'reg_alpha':
0.10702615500467058, 'reg lambda': 6.278864042286733, 'gamma':
0.5391074068396616, 'min_child_weight': 2, 'subsample': 0.6922001380402609, 'colsample_bytree': 0.4676614106898204}. Best is
trial 24 with value: 0.9084848391901552.
Best trial: 24. Best value: 0.908485: 54%
[1:12:51<47:20, 61.76s/it]
[I 2025-10-25 17:25:03,701] Trial 53 finished with value:
0.9071972267925842 and parameters: {'max depth': 13, 'n estimators':
875, 'learning rate': 0.012342302987000021, 'reg alpha':
0.13794993100980393, 'reg lambda': 6.285637361273634, 'gamma':
0.5461786523638315, 'min child weight': 2, 'subsample':
0.6174418100170301, 'colsample bytree': 0.46476273725104894}. Best is
trial 24 with value: 0.9084848391901552.
Best trial: 24. Best value: 0.908485: 55%
[1:13:53<46:24, 61.88s/it]
[I 2025-10-25 17:26:05,871] Trial 54 finished with value:
0.9069074339541542 and parameters: {'max depth': 11, 'n estimators':
862, 'learning_rate': 0.01452266362825779, 'reg alpha':
0.1082804784043711, 'reg lambda': 7.525474857188945, 'gamma':
0.4596008893762723, 'min child weight': 1, 'subsample':
0.6872218096482949, 'colsample bytree': 0.3841714887376978}. Best is
trial 24 with value: 0.9084848391901552.
Best trial: 24. Best value: 0.908485: 56%
[1:14:54<45:01, 61.39s/it]
[I 2025-10-25 17:27:06,128] Trial 55 finished with value:
0.9077443887480878 and parameters: {'max depth': 13, 'n estimators':
900, 'learning rate': 0.013482895664700687, 'reg alpha':
0.12249846454753328, 'reg_lambda': 5.371940895415502, 'gamma':
0.5992711323985043, 'min child weight': 6, 'subsample':
0.6870440436985608, 'colsample_bytree': 0.43553853891178446}. Best is
trial 24 with value: 0.9084848391901552.
Best trial: 24. Best value: 0.908485: 57%| 57/100
[1:15:48<42:24, 59.17s/it]
[I 2025-10-25 17:28:00,105] Trial 56 finished with value:
0.9079053364119117 and parameters: {'max depth': 12, 'n estimators':
838, 'learning rate': 0.0169701336714858, 'reg alpha':
0.10771668341345038, 'reg lambda': 6.706704858920836, 'gamma':
0.6657226213153957, 'min child weight': 2, 'subsample':
0.7187624586403022, 'colsample bytree': 0.5186014338431456}. Best is
trial 24 with value: 0.9084848391901552.
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Best trial: 24. Best value: 0.908485: 58%
[1:16:47<41:30, 59.30s/it]
[I 2025-10-25 17:28:59,706] Trial 57 finished with value:
0.9063603238102858 and parameters: {'max depth': 9, 'n estimators':
884, 'learning rate': 0.014905732051690653, 'reg alpha':
0.11558042887471212, 'reg_lambda': 23.516876806716215, 'gamma':
0.49752391623222825, 'min child weight': 1, 'subsample':
0.6451332315939416, 'colsample bytree': 0.5454673943631152}. Best is
trial 24 with value: 0.9084848391901552.
Best trial: 24. Best value: 0.908485: 59%
[1:17:36<38:19, 56.08s/it]
[I 2025-10-25 17:29:48,263] Trial 58 finished with value:
0.9068754972621695 and parameters: {'max depth': 8, 'n estimators':
864, 'learning rate': 0.024398823278534106, 'reg_alpha':
0.42453406309144365, 'reg_lambda': 6.212794791290743, 'gamma':
0.5730939749581193, 'min_child_weight': 3, 'subsample': 0.6254437588879187, 'colsample_bytree': 0.5957936430054003}. Best is
trial 24 with value: 0.9084848391901552.
Best trial: 24. Best value: 0.908485: 60%
[1:18:29<36:52, 55.30s/it]
[I 2025-10-25 17:30:41,768] Trial 59 finished with value:
0.9081308310109633 and parameters: {'max depth': 10, 'n estimators':
828, 'learning rate': 0.02162997127993332, 'reg_alpha':
0.12816261218437772, 'reg lambda': 5.796875837622251, 'gamma':
0.45971693806695213, 'min_child_weight': 2, 'subsample':
0.7995369755129551, 'colsample_bytree': 0.45392687936185194}. Best is
trial 24 with value: 0.9084848391901552.
Best trial: 24. Best value: 0.908485: 61%
[1:19:12<33:29, 51.54s/it]
[I 2025-10-25 17:31:24.515] Trial 60 finished with value:
0.9082917475878057 and parameters: {'max depth': 7, 'n estimators':
831, 'learning rate': 0.022062097338866896, 'reg alpha':
0.15347615548005047, 'reg lambda': 5.414591237435104, 'gamma':
0.8343763879871929, 'min_child_weight': 2, 'subsample':
0.7914211382073061, 'colsample bytree': 0.4695246599515813}. Best is
trial 24 with value: 0.9084848391901552.
Best trial: 24. Best value: 0.908485: 62%
[1:19:55<30:59, 48.94s/it]
[I 2025-10-25 17:32:07,392] Trial 61 finished with value:
0.9073904323805312 and parameters: {'max depth': 7, 'n estimators':
831, 'learning rate': 0.021932230006223234, 'reg alpha':
0.15700487917238343, 'reg lambda': 5.291247027042753, 'gamma':
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0.8396743923001353, 'min_child_weight': 2, 'subsample':
0.8037489702001521, 'colsample bytree': 0.4744089253770268}. Best is
trial 24 with value: 0.9084848391901552.
Best trial: 24. Best value: 0.908485: 63%| 63/100
[1:20:45<30:28, 49.42s/it]
[I 2025-10-25 17:32:57,921] Trial 62 finished with value:
0.9083883141136345 and parameters: {'max_depth': 8, 'n_estimators':
849, 'learning rate': 0.026692075712116358, 'reg_alpha':
0.14453816058936741, 'reg lambda': 5.633232252279819, 'gamma':
0.9230698721051153, 'min child weight': 2, 'subsample':
0.85240708857352, 'colsample bytree': 0.4575964392143463}. Best is
trial 24 with value: 0.9084848391901552.
Best trial: 24. Best value: 0.908485: 64%
[1:21:32<29:06, 48.51s/it]
[I 2025-10-25 17:33:44,327] Trial 63 finished with value:
0.9077766984838463 and parameters: {'max depth': 7, 'n estimators':
853, 'learning rate': 0.027202730042661213, 'reg alpha':
0.1694339816447177, 'reg_lambda': 5.463059984752852, 'gamma': 0.9188258477291088, 'min_child_weight': 2, 'subsample':
0.8309919249397391, 'colsample bytree': 0.5160572107020914}. Best is
trial 24 with value: 0.9084848391901552.
Best trial: 64. Best value: 0.908517: 65%
[1:22:15<27:18, 46.81s/it]
[I 2025-10-25 17:34:27,168] Trial 64 finished with value:
0.9085169934910079 and parameters: {'max depth': 8, 'n estimators':
870, 'learning rate': 0.030259113475918316, 'reg alpha':
0.14571612862304745, 'reg lambda': 7.173604236133437, 'gamma':
0.8803219813370706, 'min child weight': 2, 'subsample':
0.8554833747076822, 'colsample bytree': 0.4770577107051321}. Best is
trial 64 with value: 0.9085169934910079.
Best trial: 64. Best value: 0.908517: 66%
[1:22:59<26:04, 46.02s/it]
[I 2025-10-25 17:35:11,337] Trial 65 finished with value:
0.9079699144341206 and parameters: {'max depth': 8, 'n estimators':
845, 'learning rate': 0.030285963035066605, 'reg alpha':
0.14912920692161283, 'reg_lambda': 7.169061709849448, 'gamma':
0.8783774690418816, 'min child weight': 1, 'subsample':
0.8572492204044931, 'colsample bytree': 0.5488957316525407}. Best is
trial 64 with value: 0.9085169934910079.
Best trial: 64. Best value: 0.908517: 67%
[1:23:48<25:51, 47.02s/it]
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[I 2025-10-25 17:36:00,693] Trial 66 finished with value:
0.9077444820090314 and parameters: {'max depth': 8, 'n estimators':
891, 'learning_rate': 0.025362760198255112, 'reg_alpha':
0.18441560571254995, 'reg lambda': 7.951460440978937, 'gamma':
0.9584887292055018, 'min_child_weight': 2, 'subsample':
0.8498610544255563, 'colsample_bytree': 0.4928368074974443}. Best is
trial 64 with value: 0.9085169934910079.
Best trial: 64. Best value: 0.908517: 68%
[1:24:30<24:12, 45.38s/it]
[I 2025-10-25 17:36:42,230] Trial 67 finished with value:
0.90777661558523 and parameters: {'max depth': 7, 'n estimators': 793,
'learning rate': 0.033039739118615366, 'reg alpha':
0.1400212859207813, 'reg_lambda': 5.206930173635556, 'gamma': 0.8199798652277266, 'min_child_weight': 3, 'subsample':
0.8212550841241989, 'colsample bytree': 0.4257477746597457}. Best is
trial 64 with value: 0.9085169934910079.
Best trial: 64. Best value: 0.908517: 69%
[1:25:15<23:29, 45.46s/it]
[I 2025-10-25 17:37:27,882] Trial 68 finished with value:
0.9080664084236598 and parameters: {'max depth': 8, 'n estimators':
778, 'learning_rate': 0.02669421334000668, 'reg alpha':
0.15980010881349394, 'reg lambda': 5.645347638109829, 'gamma':
0.8997012163422008, 'min child weight': 2, 'subsample':
0.8812968751997228, 'colsample bytree': 0.5253962057542373}. Best is
trial 64 with value: 0.9085169934910079.
Best trial: 64. Best value: 0.908517: 70%
[1:25:55<21:47, 43.59s/it]
[I 2025-10-25 17:38:07,124] Trial 69 finished with value:
0.9079375736113807 and parameters: {'max depth': 8, 'n estimators':
672, 'learning rate': 0.028844268512471072, 'reg_alpha':
0.13311651321707968, 'reg_lambda': 5.949257161704015, 'gamma':
0.7853012986731176, 'min child weight': 1, 'subsample':
0.8626243158834652, 'colsample_bytree': 0.4490595770704811}. Best is
trial 64 with value: 0.9085169934910079.
Best trial: 64. Best value: 0.908517: 71%| | 71/100
[1:26:48<22:29, 46.55s/it]
[I 2025-10-25 17:39:00,559] Trial 70 finished with value:
0.9076158026316579 and parameters: {'max depth': 7, 'n estimators':
873, 'learning rate': 0.023705612632192126, 'reg alpha':
0.14155424908671443, 'reg lambda': 10.587940446016862, 'gamma':
0.7282266275295892, 'min_child_weight': 2, 'subsample':
0.9101585593662503, 'colsample bytree': 0.5726881312392791}. Best is
trial 64 with value: 0.9085169934910079.
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Best trial: 64. Best value: 0.908517: 72%| 72/100
[1:27:36<21:50, 46.82s/it]
[I 2025-10-25 17:39:48,017] Trial 71 finished with value:
0.9081952017866308 and parameters: {'max depth': 8, 'n estimators':
876, 'learning rate': 0.03133615319501425, 'reg alpha':
0.15220714980176164, 'reg_lambda': 6.381603677408007, 'gamma':
0.8691649248551014, 'min child weight': 2, 'subsample':
0.8963211086086561, 'colsample_bytree': 0.4807112830578609}. Best is
trial 64 with value: 0.9085169934910079.
Best trial: 64. Best value: 0.908517: 73%| 73/100
[1:28:18<20:32, 45.64s/it]
[I 2025-10-25 17:40:30,907] Trial 72 finished with value:
0.9080664395106413 and parameters: {'max depth': 8, 'n estimators':
854, 'learning_rate': 0.030805615439032617, 'reg_alpha':
0.1564366975359023, 'reg_lambda': 6.738529420533738, 'gamma': 0.8734281169736516, 'min_child_weight': 2, 'subsample':
0.8945182836847968, 'colsample bytree': 0.4794439578630928}. Best is
trial 64 with value: 0.9085169934910079.
Best trial: 64. Best value: 0.908517: 74%
[1:29:01<19:21, 44.67s/it]
[I 2025-10-25 17:41:13,324] Trial 73 finished with value:
0.9084526745269754 and parameters: {'max depth': 9, 'n estimators':
883, 'learning rate': 0.03403626405576363, 'reg alpha':
0.17124517613970092, 'reg lambda': 6.023183638981225, 'gamma':
0.9322489815242848, 'min_child_weight': 2, 'subsample':
0.9332577676463589, 'colsample_bytree': 0.5093338532873787}. Best is
trial 64 with value: 0.9085169934910079.
Best trial: 64. Best value: 0.908517: 75%| | 75/100
[1:29:43<18:15, 43.84s/it]
[I 2025-10-25 17:41:55.206] Trial 74 finished with value:
0.9078088942340076 and parameters: {'max depth': 9, 'n estimators':
882, 'learning rate': 0.0336173887492388, 'reg alpha':
0.17514871281882885, 'req lambda': 6.5001800275328625, 'qamma':
0.9443635837828503, 'min_child_weight': 2, 'subsample':
0.9445697128994706, 'colsample bytree': 0.5074498349131493}. Best is
trial 64 with value: 0.9085169934910079.
Best trial: 64. Best value: 0.908517: 76%
[1:30:23<17:09, 42.89s/it]
[I 2025-10-25 17:42:35,897] Trial 75 finished with value:
0.9078732753720024 and parameters: {'max depth': 9, 'n estimators':
865, 'learning rate': 0.030432103710139168, 'reg alpha':
0.15202353003867727, 'reg_lambda': 5.97641849281491, 'gamma':
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0.9767022471328402, 'min_child_weight': 4, 'subsample':
0.9213997864195553, 'colsample bytree': 0.5368092865762365}. Best is
trial 64 with value: 0.9085169934910079.
Best trial: 64. Best value: 0.908517: 77%| 77/100
[1:31:08<16:38, 43.41s/it]
[I 2025-10-25 17:43:20,503] Trial 76 finished with value:
0.9080664498729683 and parameters: {'max_depth': 9, 'n_estimators':
896, 'learning_rate': 0.03402648038552367, 'reg_alpha':
0.1924945217507263, 'reg lambda': 7.376723666761994, 'gamma':
0.9278007581506236, 'min child weight': 2, 'subsample':
0.9300988606023558, 'colsample_bytree': 0.43981013953190834}. Best is
trial 64 with value: 0.9085169934910079.
Best trial: 64. Best value: 0.908517: 78%
[1:31:49<15:40, 42.74s/it]
[I 2025-10-25 17:44:01,676] Trial 77 finished with value:
0.9075191635695397 and parameters: {'max depth': 8, 'n estimators':
818, 'learning rate': 0.03168320205060062, 'reg alpha':
0.17059473254510163, 'reg lambda': 14.504810081422889, 'gamma':
0.855007662092883, 'min child weight': 3, 'subsample':
0.8989074497184656, 'colsample bytree': 0.48103248260220033}. Best is
trial 64 with value: 0.9085169934910079.
Best trial: 64. Best value: 0.908517: 79%
[1:32:34<15:11, 43.39s/it]
[I 2025-10-25 17:44:46,609] Trial 78 finished with value:
0.9075192568304832 and parameters: {'max depth': 7, 'n estimators':
877, 'learning rate': 0.028104530856118016, 'reg alpha':
0.1452951298866071, 'reg lambda': 6.150945907296516, 'gamma':
0.8281949193509073, 'min child weight': 2, 'subsample':
0.9356900869193219, 'colsample bytree': 0.5600738093972268}. Best is
trial 64 with value: 0.9085169934910079.
Best trial: 64. Best value: 0.908517: 80%
[1:33:17<14:25, 43.26s/it]
[I 2025-10-25 17:45:29,538] Trial 79 finished with value:
0.9083239122509854 and parameters: {'max_depth': 10, 'n_estimators':
844, 'learning rate': 0.03177634058343102, 'reg alpha':
0.16172745760644847, 'reg_lambda': 5.661729635346918, 'gamma':
0.9050950122588716, 'min child weight': 1, 'subsample':
0.9165268065950656, 'colsample_bytree': 0.3807627958207295}. Best is
trial 64 with value: 0.9085169934910079.
Best trial: 64. Best value: 0.908517: 81%| 81/100
[1:34:02<13:51, 43.75s/it]
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[I 2025-10-25 17:46:14,429] Trial 80 finished with value:
0.9079375839737078 and parameters: {'max depth': 10, 'n estimators':
838, 'learning_rate': 0.03198863518425816, 'reg_alpha':
0.1966184262856684, 'req lambda': 7.10001832492738, 'qamma':
0.8973258478498811, 'min child weight': 1, 'subsample':
0.8853181891552921, 'colsample_bytree': 0.39302273125494414}. Best is
trial 64 with value: 0.9085169\overline{9}34910079.
Best trial: 64. Best value: 0.908517: 82%
[1:34:51<13:33, 45.19s/it]
[I 2025-10-25 17:47:02,978] Trial 81 finished with value:
0.9083238500770232 and parameters: {'max depth': 9, 'n estimators':
887, 'learning rate': 0.030140984849565438, 'reg alpha':
0.1527978480781973, 'reg_lambda': 5.585408059112602, 'gamma': 0.8602473864490824, 'min_child_weight': 1, 'subsample':
0.9143309700945932, 'colsample bytree': 0.4550840067681635}. Best is
trial 64 with value: 0.9085169934910079.
Best trial: 64. Best value: 0.908517: 83%
[1:35:39<13:06, 46.24s/it]
[I 2025-10-25 17:47:51,674] Trial 82 finished with value:
0.9078409863608978 and parameters: {'max depth': 9, 'n estimators':
887, 'learning_rate': 0.02884076260834662, 'reg_alpha': 0.16385585029803335, 'reg_lambda': 5.684210755992505, 'gamma':
0.809212522780557, 'min_child_weight': 1, 'subsample': 0.9032884307187228, 'colsample_bytree': 0.37092940543305225}. Best is
trial 64 with value: 0.9085169934910079.
Best trial: 83. Best value: 0.908549: 84%
[1:36:28<12:31, 46.98s/it]
[I 2025-10-25 17:48:40,389] Trial 83 finished with value:
0.908549230690477 and parameters: {'max depth': 9, 'n estimators':
867, 'learning rate': 0.026405310134944796, 'reg alpha':
0.15114994538590773, 'reg_lambda': 5.234212686026411, 'gamma':
0.8541527156797234, 'min child weight': 1, 'subsample':
0.9154685287768807, 'colsample bytree': 0.4597749493821352}. Best is
trial 83 with value: 0.908549230690477.
Best trial: 83. Best value: 0.908549: 85%| 85%| 85/100
[1:37:15<11:44, 46.99s/it]
[I 2025-10-25 17:49:27,382] Trial 84 finished with value:
0.9079376875969786 and parameters: {'max_depth': 9, 'n_estimators':
859, 'learning rate': 0.026132951497911217, 'reg alpha':
0.16537456894353494, 'reg lambda': 5.168651802614448, 'gamma':
0.977876014962281, 'min_child_weight': 1, 'subsample': 0.9183345720485788, 'colsample bytree': 0.4267551940257745}. Best is
trial 83 with value: 0.908549230690477.
```

```
Best trial: 83. Best value: 0.908549: 86%| 86%| 86/100
[1:38:07<11:17, 48.40s/it]
[I 2025-10-25 17:50:19,066] Trial 85 finished with value:
0.908517034940316 and parameters: {'max depth': 10, 'n estimators':
865, 'learning rate': 0.02464900558454704, 'reg alpha':
0.22839537773112714, 'reg_lambda': 5.436644552938625, 'gamma':
0.8922502664896693, 'min child weight': 1, 'subsample':
0.9127793375598292, 'colsample_bytree': 0.4548193831601266}. Best is
trial 83 with value: 0.908549230690477.
Best trial: 83. Best value: 0.908549: 87%| 87/100
[1:38:48<10:03, 46.39s/it]
[I 2025-10-25 17:51:00,770] Trial 86 finished with value:
0.9078089149586616 and parameters: {'max depth': 10, 'n estimators':
848, 'learning rate': 0.034883189514509895, 'reg alpha':
0.23039732664511198, 'reg_lambda': 5.720339592391384, 'gamma':
0.9264476464230557, 'min_child_weight': 1, 'subsample': 0.9127061041239034, 'colsample_bytree': 0.4502946025836068}. Best is
trial 83 with value: 0.908549230690477.
Best trial: 83. Best value: 0.908549: 88%
[1:39:31<09:03, 45.27s/it]
[I 2025-10-25 17:51:43,419] Trial 87 finished with value:
0.9076157300953687 and parameters: {'max depth': 10, 'n estimators':
653, 'learning rate': 0.024616305875860514, 'reg alpha':
0.2480084875487097, 'reg lambda': 5.239025635945466, 'gamma':
0.8537274692108678, 'min child weight': 1, 'subsample':
0.9388477649091822, 'colsample_bytree': 0.36403806840370734}. Best is
trial 83 with value: 0.908549230690477.
Best trial: 83. Best value: 0.908549: 89%| 89/100
[1:40:19<08:26, 46.06s/it]
[I 2025-10-25 17:52:31,343] Trial 88 finished with value:
0.907840996723225 and parameters: {'max depth': 10, 'n estimators':
868, 'learning rate': 0.029704507551474506, 'reg alpha':
0.17780682971467957, 'reg lambda': 6.033303624319953, 'gamma':
0.8903533145183651, 'min_child_weight': 1, 'subsample':
0.8757725288501241, 'colsample bytree': 0.4040989827074721}. Best is
trial 83 with value: 0.908549230690477.
Best trial: 83. Best value: 0.908549: 90%| 90/100
[1:41:06<07:44, 46.45s/it]
[I 2025-10-25 17:53:18,694] Trial 89 finished with value:
0.9082917890371138 and parameters: {'max depth': 9, 'n estimators':
890, 'learning rate': 0.02271113727957354, 'reg alpha':
0.22549579795340002, 'reg_lambda': 5.01624633921508, 'gamma':
```

```
0.9921466506306091, 'min_child_weight': 1, 'subsample':
0.9308312722104015, 'colsample bytree': 0.49913252283541415}. Best is
trial 83 with value: 0.908549230690477.
Best trial: 83. Best value: 0.908549: 91%| 91/100
[1:41:49<06:48, 45.40s/it]
[I 2025-10-25 17:54:01,646] Trial 90 finished with value:
0.9077445027336856 and parameters: {'max depth': 9, 'n estimators':
691, 'learning rate': 0.02704449582844297, 'reg_alpha':
0.27758832141218465, 'reg lambda': 5.569462374562386, 'gamma':
0.9481423834102599, 'min_child_weight': 1, 'subsample':
0.9461295199137875, 'colsample_bytree': 0.41439971693889743}. Best is
trial 83 with value: 0.908549230690477.
Best trial: 83. Best value: 0.908549: 92%| 92/100
[1:42:36<06:07, 45.88s/it]
[I 2025-10-25 17:54:48,652] Trial 91 finished with value:
0.9082918512110764 and parameters: {'max depth': 9, 'n estimators':
892, 'learning rate': 0.028000348251668472, 'reg alpha':
0.21137173922508365, 'reg lambda': 5.350547920204922, 'gamma':
0.9987947212397027, 'min_child_weight': 1, 'subsample':
0.9285887211297935, 'colsample bytree': 0.5013706660283594}. Best is
trial 83 with value: 0.908549230690477.
Best trial: 83. Best value: 0.908549: 93%| 93/100
[1:43:21<05:18, 45.46s/it]
[I 2025-10-25 17:55:33,119] Trial 92 finished with value:
0.9083239122509855 and parameters: {'max depth': 9, 'n estimators':
900, 'learning rate': 0.03279775204753406, 'reg alpha':
0.2115812902083221, 'reg lambda': 5.856700907503501, 'gamma':
0.9089959941883466, 'min child weight': 1, 'subsample':
0.9172015714662228, 'colsample bytree': 0.458560576194427}. Best is
trial 83 with value: 0.908549230690477.
Best trial: 83. Best value: 0.908549: 94%| 94/100
[1:44:01<04:23, 43.93s/it]
[I 2025-10-25 17:56:13,479] Trial 93 finished with value:
0.9079698315355043 and parameters: {'max depth': 9, 'n estimators':
883, 'learning rate': 0.032896166809139635, 'reg alpha':
0.18736712496961513, 'reg_lambda': 5.851469103127803, 'gamma':
0.9280280383671573, 'min child weight': 1, 'subsample':
0.9127432683404719, 'colsample_bytree': 0.454214594621293}. Best is
trial 83 with value: 0.908549230690477.
Best trial: 83. Best value: 0.908549: 95%| 95/100
[1:44:49<03:45, 45.17s/it]
```

```
[I 2025-10-25 17:57:01,554] Trial 94 finished with value:
0.9078411003464956 and parameters: {'max depth': 9, 'n estimators':
899, 'learning_rate': 0.02965784544748751, 'reg_alpha':
0.258040205192894, 'reg lambda': 6.160970102340118, 'gamma':
0.9096449188467247, 'min child weight': 1, 'subsample':
0.9155858337313796, 'colsample_bytree': 0.42980328306088245}. Best is
trial 83 with value: 0.9085492\overline{3}0690477.
Best trial: 83. Best value: 0.908549: 96%| 96/100
[1:45:41<03:09, 47.31s/it]
[I 2025-10-25 17:57:53,842] Trial 95 finished with value:
0.9080984798258964 and parameters: {'max depth': 10, 'n estimators':
868, 'learning rate': 0.027771467867481775, 'reg alpha':
0.2022779612609495, 'reg_lambda': 6.569381423788368, 'gamma': 0.8616443666413809, 'min_child_weight': 1, 'subsample':
0.8885224964507208, 'colsample bytree': 0.3804594625783567\. Best is
trial 83 with value: 0.908549230690477.
Best trial: 96. Best value: 0.908871: 97%| 97/100
[1:46:40<02:32, 50.77s/it]
[I 2025-10-25 17:58:52,684] Trial 96 finished with value:
0.9088710949311434 and parameters: {'max depth': 9, 'n estimators':
880, 'learning_rate': 0.03203491991265297, 'reg_alpha':
0.2183856109327376, 'req lambda': 5.620054034241966, 'qamma':
0.7488169818937879, 'min_child_weight': 1, 'subsample':
0.9077486227069126, 'colsample bytree': 0.4671920383353195}. Best is
trial 96 with value: 0.9088710949311434.
Best trial: 96. Best value: 0.908871: 98%
[1:47:51<01:53, 56.86s/it]
[I 2025-10-25 18:00:03,757] Trial 97 finished with value:
0.9078410899841686 and parameters: {'max depth': 10, 'n estimators':
857, 'learning rate': 0.032706089718196905, 'reg alpha':
0.2206724770101333, 'reg_lambda': 5.190498951058647, 'gamma':
0.8895097850600606, 'min_child weight': 1, 'subsample':
0.9045275711642359, 'colsample_bytree': 0.5145443134594655}. Best is
trial 96 with value: 0.9088710949311434.
Best trial: 96. Best value: 0.908871: 99%| 99/100
[1:49:33<01:10, 70.39s/it]
[I 2025-10-25 18:01:45,712] Trial 98 finished with value:
0.907969779723869 and parameters: {'max depth': 9, 'n estimators':
846, 'learning rate': 0.026219684644871434, 'reg alpha':
0.23906574590160554, 'reg lambda': 6.411260552664041, 'gamma':
0.7881243271834483, 'min_child_weight': 1, 'subsample':
0.8720487211916375, 'colsample bytree': 0.43993070946930296}. Best is
trial 96 with value: 0.9088710949311434.
```

```
Best trial: 96. Best value: 0.908871: 100%
[1:51:04<00:00, 66.65s/it]
[I 2025-10-25 18:03:16,501] Trial 99 finished with value:
0.9083560147402029 and parameters: {'max depth': 8, 'n estimators':
868, 'learning rate': 0.03428021741624994, 'reg alpha':
0.20923196298950916, 'reg_lambda': 5.929613969520971, 'gamma':
0.7415008225166015, 'min child weight': 1, 'subsample':
0.9238257672116901, 'colsample_bytree': 0.35396939407550343}. Best is
trial 96 with value: 0.9088710949311434.
Best CV accuracy: 0.90887
Best hyperparameters found:
  max depth: 9
  n estimators: 880
  learning rate: 0.03203491991265297
  reg alpha: 0.2183856109327376
  reg lambda: 5.620054034241966
  gamma: 0.7488169818937879
  min child weight: 1
  subsample: 0.9077486227069126
  colsample bytree: 0.4671920383353195
Training final model with best parameters...
Preparing and preprocessing test data...
Test data preprocessing complete.
Generating test predictions...
Submission saved to:
/home/iiitb/Desktop/IIITB/ML/project first half/xgb refined peak searc
h.csv
Sample Submission:
      id
               WeightCategory
  15533
             Obesity_Type_III
  15534
           Overweight Level I
  15535 Overweight_Level_II
3 15536
              Obesity Type II
4 15537
                Normal Weight
final model = xgb.XGBClassifier(**best params final)
final_model.fit(X, y_enc, verbose=False)
X test = df test.drop('id', axis=1)
test ids = df test['id']
for col in categorical cols:
    if col not in X test.columns:
        X_{\text{test[col]}} = 'Missing'
```

```
train categories = df train[col].astype('category').cat.categories
   X test[col] = pd.Categorical(X test[col],
categories=train categories)
X test = pd.get dummies(X test, columns=categorical cols,
drop first=True)
X test = X test.reindex(columns=X.columns, fill value=0)
num features present test = [col for col in num features if col in
X test.columns]
X test[num features present test] =
scaler.transform(X test[num features present test])
y test pred enc = final model.predict(X test)
y test pred = le.inverse_transform(y_test_pred_enc)
submission = pd.DataFrame({'id': test ids, 'WeightCategory':
y test pred})
submission.to_csv(file_path base + 'xgb fast optuna.csv', index=False)
[I 2025-10-25 21:35:07,832] A new study created in memory with name:
no-name-2d0af850-3dc1-4d72-af5d-57c14b1457f3
Starting Fast Optuna Search...
Best trial: 0. Best value: 0.907841: 1%| | 1/100
[00:46<1:17:22, 46.89s/it]
[I 2025-10-25 21:35:54,723] Trial 0 finished with value:
0.9078410588971872 and parameters: {'n_estimators': 675,
'learning_rate': 0.0568376349572445, 'max_depth': 9, 'subsample':
0.5598658484197037, 'colsample bytree': 0.4356018640442436,
'min_child_weight': 1, 'gamma': 0.5116167224336399, 'reg_alpha':
0.47323522915498706, 'reg lambda': 1.9814495152661715}. Best is trial
0 with value: 0.9078410588971872.
Best trial: 0. Best value: 0.907841: 2%| | 2/100
[01:53<1:35:04, 58.20s/it]
[I 2025-10-25 21:37:00,850] Trial 1 finished with value:
0.9075835861568429 and parameters: {'n_estimators': 742,
'learning rate': 0.020457440438566, 'max depth': 9, 'subsample':
0.5832442640800422, 'colsample_bytree': 0.4412339110678276,
'min_child_weight': 1, 'gamma': 0.5366809019706867, 'reg_alpha':
0.36084844859190757, 'reg_lambda': 1.8821833611219092}. Best is trial
0 with value: 0.9078410588971872.
Best trial: 2. Best value: 0.908259: 3%|
                                                   | 3/100
[02:45<1:29:41, 55.48s/it]
```

```
[I 2025-10-25 21:37:53,090] Trial 2 finished with value:
0.9082594689390282 and parameters: {'n estimators': 686,
'learning_rate': 0.027541120694067212, 'max_depth': 8, 'subsample': 0.5139493860652041, 'colsample_bytree': 0.4492144648535218,
'min child weight': 1, 'gamma': 0.5912139968434071, 'reg alpha':
0.45703519227860273, 'reg lambda': 1.4595759168058677}. Best is trial
2 with value: 0.9082594689390282.
Best trial: 2. Best value: 0.908259: 4%| | 4/100
[03:37<1:26:27, 54.03s/it]
[I 2025-10-25 21:38:44,909] Trial 3 finished with value:
0.9068752589286472 and parameters: {'n_estimators': 703,
'learning rate': 0.0383427782659158, 'max depth': 7, 'subsample':
0.5607544851901438, 'colsample_bytree': 0.43705241236872916, 
'min_child_weight': 1, 'gamma': 0.6897771074506667, 'reg_alpha':
0.49312640661491186, 'req lambda': 2.2509165525513994}. Best is trial
2 with value: 0.9082594689390282.
[04:32<1:26:06, 54.38s/it]
[I 2025-10-25 21:39:39,904] Trial 4 finished with value:
0.9084203958781979 and parameters: {'n_estimators': 661,
'learning_rate': 0.022265447964905188, 'max_depth': 9, 'subsample':
0.5440152493739601, 'colsample bytree': 0.4322038234844779,
'min_child_weight': 1, 'gamma': 0.5068777042230437, 'reg_alpha':
0.4818640804157564, 'reg lambda': 1.536413976080022}. Best is trial 4
with value: 0.9084203958781979.
[05:18<1:21:01, 51.72s/it]
[I 2025-10-25 21:40:26,463] Trial 5 finished with value:
0.9073903080326063 and parameters: {'n estimators': 733,
'learning_rate': 0.02816786790385486, 'max_depth': 8, 'subsample':
0.554671027934328, 'colsample_bytree': 0.4384854455525527,
'min child weight': 2, 'gamma': 0.6550265646722229, 'reg alpha':
0.4878997883128378, 'reg lambda': 2.363275555559433}. Best is trial 4
with value: 0.9084203958781979.
[05:58<1:14:02, 47.77s/it]
[I 2025-10-25 21:41:06,107] Trial 6 finished with value:
0.9073903909312229 and parameters: {'n estimators': 720,
'learning rate': 0.05506501823422673, 'max depth': 7, 'subsample':
0.5195982862419145, 'colsample bytree': 0.4245227288910538.
'min_child_weight': 1, 'gamma': 0.5777354579378964, 'reg_alpha':
0.35426980635477917, 'reg lambda': 2.277358761897508}. Best is trial 4
with value: 0.9084203958781979.
```

```
Best trial: 4. Best value: 0.90842: 8%| ■
                                                  | 8/100
[06:44<1:12:39, 47.39s/it]
[I 2025-10-25 21:41:52,665] Trial 7 finished with value:
0.907519028859288 and parameters: {'n estimators': 671,
'learning rate': 0.02723139071305203, 'max depth': 8, 'subsample':
0.5140924224974762, 'colsample bytree': 0.500219698075404,
'min child weight': 1, 'gamma': 0.6973773873201035, 'reg alpha':
0.4544489538593315, 'reg lambda': 1.458330385994424}. Best is trial 4
with value: 0.9084203958781979.
                                                  | 9/100
Best trial: 4. Best value: 0.90842: 9%|■
[07:23<1:07:39, 44.62s/it]
[I 2025-10-25 21:42:31,185] Trial 8 finished with value:
0.9077766984838463 and parameters: {'n estimators': 601,
'learning_rate': 0.048989609264119945, 'max_depth': 9, 'subsample':
0.5729007168040987, 'colsample_bytree': 0.4971270346685946,
'min_child_weight': 1, 'gamma': 0.5716931457088545, 'reg_alpha': 0.3231738119050259, 'reg_lambda': 2.3220344536382713}. Best is trial 4
with value: 0.9084203958781979.
Best trial: 4. Best value: 0.90842: 10%
                                                  | 10/100
[08:54<1:28:41, 59.12s/it]
[I 2025-10-25 21:44:02,797] Trial 9 finished with value:
0.9072293810934371 and parameters: {'n estimators': 725,
'learning_rate': 0.02876792093597593, 'max_depth': 7, 'subsample':
0.5310982321715663, 'colsample_bytree': 0.4525183322026747,
'min child weight': 2, 'gamma': 0.6275114942710426, 'reg alpha':
0.4774425485152653, 'reg_lambda': 1.813879402710534}. Best is trial 4
with value: 0.9084203958781979.
Best trial: 4. Best value: 0.90842: 11%
[10:14<1:37:09, 65.50s/it]
[I 2025-10-25 21:45:22.756] Trial 10 finished with value:
0.9077766570345382 and parameters: {'n_estimators': 789,
'learning rate': 0.02013220662649724, 'max depth': 9, 'subsample':
0.5377951549965202, 'colsample bytree': 0.47361885886977595,
'min_child_weight': 2, 'gamma': 0.5030253631826769, 'reg_alpha':
0.42271244390081464, 'reg lambda': 1.2616101661532841}. Best is trial
4 with value: 0.9084203958781979.
[11:09<1:31:21, 62.29s/it]
[I 2025-10-25 21:46:17,701] Trial 11 finished with value:
0.9082272731888674 and parameters: {'n estimators': 635,
'learning_rate': 0.024114404183710192, 'max_depth': 8, 'subsample':
0.5021965767671166, 'colsample bytree': 0.46343562717564113,
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'min_child_weight': 1, 'gamma': 0.5480133166277307, 'reg_alpha':
0.43517145701401755, 'reg lambda': 1.5561018635006951}. Best is trial
4 with value: 0.9084203958781979.
Best trial: 4. Best value: 0.90842: 13%
                                                  | 13/100
[11:48<1:19:55, 55.13s/it]
[I 2025-10-25 21:46:56,336] Trial 12 finished with value:
0.90796981081085 and parameters: {'n_estimators': 659,
'learning rate': 0.03533903995060137, 'max depth': 8, 'subsample':
0.5405312323857697, 'colsample bytree': 0.4806685008254879,
'min child weight': 1, 'gamma': 0.6153642749316125, 'reg alpha':
0.40\overline{2}01215\overline{0}3016097, 'reg_lambda': 1.5871979155311395\). Best is trial 4
with value: 0.9084203958781979.
[12:33<1:14:43, 52.13s/it]
[I 2025-10-25 21:47:41,536] Trial 13 finished with value:
0.9076479465701837 and parameters: {'n estimators': 633,
'learning rate': 0.02391978873038799, 'max depth': 8, 'subsample':
0.5999362438641551, 'colsample_bytree': 0.4203655265526033, 
'min_child_weight': 1, 'gamma': 0.5899642200291758, 'reg_alpha':
0.4494855346140843, 'req lambda': 1.2484941308722979}. Best is trial 4
with value: 0.9084203958781979.
[13:31<1:16:22, 53.91s/it]
[I 2025-10-25 21:48:39,582] Trial 14 finished with value:
0.9078088735093534 and parameters: {'n estimators': 767,
'learning_rate': 0.031055353447737806, 'max_depth': 9, 'subsample':
0.5010075037410588, 'colsample bytree': 0.4567378887220016,
'min_child_weight': 2, 'gamma': 0.538508025259816, 'reg_alpha': 0.400348584780444, 'reg_lambda': 1.7010445016758782}. Best is trial 4
with value: 0.9084203958781979.
Best trial: 4. Best value: 0.90842: 16%
[14:17<1:11:54, 51.37s/it]
[I 2025-10-25 21:49:25,044] Trial 15 finished with value:
0.9078088320600453 and parameters: {'n_estimators': 694,
'learning_rate': 0.02443926519916055, 'max_depth': 8, 'subsample':
0.5219390850067291, 'colsample_bytree': 0.4491518325342071,
'min child weight': 1, 'gamma': 0.6466997224328896, 'reg alpha':
0.49968149237698123, 'reg lambda': 1.4041528098525133}. Best is trial
4 with value: 0.9084203958781979.
Best trial: 4. Best value: 0.90842: 17%| ■ | 17/100
[14:57<1:06:31, 48.09s/it]
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[I 2025-10-25 21:50:05,531] Trial 16 finished with value:
0.907841017447879 and parameters: {'n_estimators': 642,
'learning_rate': 0.04037699486789134, 'max_depth': 8, 'subsample':
0.545575553156024, 'colsample bytree': 0.48293286660298157,
'min child weight': 1, 'gamma': 0.5619401559433432, 'reg alpha':
0.45511159059198514, 'reg lambda': 2.0664963311000344}. Best is trial
4 with value: 0.9084203958781979.
Best trial: 4. Best value: 0.90842: 18%
[15:44<1:05:24, 47.85s/it]
[I 2025-10-25 21:50:52,824] Trial 17 finished with value:
0.9070362584041064 and parameters: {'n estimators': 607,
'learning rate': 0.022284386112730287, 'max depth': 9, 'subsample':
0.5266323252011837, 'colsample_bytree': 0.4302459425334342, 
'min_child_weight': 2, 'gamma': 0.6083751615526403, 'reg_alpha':
0.4189484849380467, 'reg lambda': 1.6674049841500673}. Best is trial 4
with value: 0.9084203958781979.
Best trial: 4. Best value: 0.90842: 19%| ■ | 19/100
[16:46<1:10:02, 51.88s/it]
[I 2025-10-25 21:51:54,094] Trial 18 finished with value:
0.9078088735093534 and parameters: {'n estimators': 698,
'learning_rate': 0.030553105357816794, 'max_depth': 7, 'subsample':
0.511062381951883, 'colsample bytree': 0.44701306245259903,
'min_child_weight': 1, 'gamma': 0.5221446825955072, 'reg_alpha':
0.4631459674619086, 'reg lambda': 1.4137157588929812}. Best is trial 4
with value: 0.9084203958781979.
Best trial: 4. Best value: 0.90842: 20%
[17:46<1:12:27, 54.34s/it]
[I 2025-10-25 21:52:54,156] Trial 19 finished with value:
0.9080986663477836 and parameters: {'n_estimators': 656,
'learning_rate': 0.026176416981389512, 'max_depth': 8, 'subsample':
0.572863353136759, 'colsample_bytree': 0.4647519702128127,
'min child weight': 1, 'gamma': 0.6624432052865857, 'reg alpha':
0.3811840530818634, 'reg lambda': 1.7688443513857852}. Best is trial 4
with value: 0.9084203958781979.
Best trial: 4. Best value: 0.90842: 21%| | 21/100
[18:47<1:14:08, 56.31s/it]
[I 2025-10-25 21:53:55,054] Trial 20 finished with value:
0.9072615975682521 and parameters: {'n_estimators': 682,
'learning rate': 0.0323558768830078, 'max depth': 9, 'subsample':
0.5312064394429887, 'colsample bytree': 0.5128102868232872,
'min_child_weight': 2, 'gamma': 0.5913200930991319, 'reg_alpha':
0.3023241923255106, 'reg lambda': 2.491995620564621}. Best is trial 4
with value: 0.9084203958781979.
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Best trial: 4. Best value: 0.90842: 22%
                                                  | 22/100
[19:33<1:09:19, 53.33s/it]
[I 2025-10-25 21:54:41,432] Trial 21 finished with value:
0.9079053260495845 and parameters: {'n estimators': 629,
'learning_rate': 0.023218205887275016, 'max_depth': 8, 'subsample':
0.5027610649116776, 'colsample_bytree': 0.4625243211434026,
'min child weight': 1, 'gamma': 0.5485647585916936, 'reg alpha':
0.436963184298801, 'reg lambda': 1.548764442278793}. Best is trial 4
with value: 0.9084203958781979.
Best trial: 22. Best value: 0.908581: 23%
[20:20<1:06:01, 51.44s/it]
[I 2025-10-25 21:55:28,476] Trial 22 finished with value:
0.9085813435420216 and parameters: {'n estimators': 650,
'learning_rate': 0.02578581130395421, 'max_depth': 8, 'subsample':
0.5090888506879074, 'colsample_bytree': 0.4716306919726478,
'min_child_weight': 1, 'gamma': 0.5529147756269048, 'reg_alpha': 0.43522437190032626, 'reg_lambda': 1.5383053974575969}. Best is trial
22 with value: 0.9085813435420216.
Best trial: 22. Best value: 0.908581: 24%
[21:32<1:12:49, 57.49s/it]
[I 2025-10-25 21:56:40,070] Trial 23 finished with value:
0.9083560976388194 and parameters: {'n estimators': 656,
'learning_rate': 0.026017838821080046, 'max_depth': 8, 'subsample':
0.5118974698340633, 'colsample_bytree': 0.4791712344849933,
'min child weight': 1, 'gamma': 0.5239697021273682, 'reg alpha':
0.47535764263575375, 'reg lambda': 1.339510557260656}. Best is trial
22 with value: 0.9085813435420216.
Best trial: 22. Best value: 0.908581: 25%
[22:23<1:09:24, 55.53s/it]
[I 2025-10-25 21:57:31,037] Trial 24 finished with value:
0.9084203647912167 and parameters: {'n estimators': 653,
'learning rate': 0.021741314709098254, 'max depth': 8, 'subsample':
0.5128181175558134, 'colsample bytree': 0.49134588533699314,
'min_child_weight': 1, 'gamma': 0.5192919326882234, 'reg_alpha':
0.47060580599316565, 'reg lambda': 1.3281392976340498}. Best is trial
22 with value: 0.9085813435420216.
Best trial: 22. Best value: 0.908581: 26%
[23:08<1:04:47, 52.53s/it]
[I 2025-10-25 21:58:16,560] Trial 25 finished with value:
0.9073259268946117 and parameters: {'n estimators': 615,
'learning_rate': 0.021832672245957956, 'max_depth': 7, 'subsample':
0.5351278620220314, 'colsample bytree': 0.4927454568580796,
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'min_child_weight': 1, 'gamma': 0.5027795116358034, 'reg_alpha':
0.42895297322143666, 'reg lambda': 1.2043944972954095}. Best is trial
22 with value: 0.9085813435420216.
Best trial: 26. Best value: 0.908646: 27%
                                                   | 27/100
[23:52<1:00:53, 50.05s/it]
[I 2025-10-25 21:59:00,819] Trial 26 finished with value:
0.9086457350423436 and parameters: {'n_estimators': 647,
'learning_rate': 0.02135778951932844, 'max_depth': 9, 'subsample':
0.5484818900688542, 'colsample_bytree': 0.4888114091962343,
'min child weight': 1, 'gamma': 0.5247306106423945, 'reg alpha':
0.44431474858112235, 'reg lambda': 1.310080015536094}. Best is trial
26 with value: 0.9086457350423436.
Best trial: 26. Best value: 0.908646: 28%
[24:34<56:53, 47.41s/it]
[I 2025-10-25 21:59:42,087] Trial 27 finished with value:
0.9081950774387064 and parameters: {'n estimators': 619,
'learning_rate': 0.020389753271788748, 'max depth': 9, 'subsample':
0.5497081361911257, 'colsample bytree': 0.5106143009224111,
'min child_weight': 1, 'gamma': 0.5339500426201662, 'reg_alpha':
0.415618626945721, 'reg lambda': 1.644618724348398}. Best is trial 26
with value: 0.9086457350423436.
Best trial: 26. Best value: 0.908646: 29%
[25:13<53:13, 44.98s/it]
[I 2025-10-25 22:00:21,406] Trial 28 finished with value:
0.9080664705976224 and parameters: {'n estimators': 666,
'learning_rate': 0.025265316214930996, 'max_depth': 9, 'subsample':
0.5715294480257942, 'colsample bytree': 0.47129442625042467,
'min_child_weight': 1, 'gamma': 0.5600560105357463, 'reg_alpha':
0.4437353317962716, 'reg lambda': 1.5012312968899046}. Best is trial
26 with value: 0.9086457350423436.
Best trial: 26. Best value: 0.908646: 30%
[25:56<51:49, 44.43s/it]
[I 2025-10-25 22:01:04,527] Trial 29 finished with value:
0.9083882726643264 and parameters: {'n_estimators': 645,
'learning_rate': 0.02182595895218678, 'max_depth': 9, 'subsample':
0.5629553863747403, 'colsample_bytree': 0.5053702084294917,
'min child weight': 1, 'gamma': 0.5112345981645368, 'reg alpha':
0.3809716588406472, 'reg_lambda': 1.9803135141768018}. Best is trial
26 with value: 0.9086457350423436.
Best trial: 26. Best value: 0.908646: 31%
[26:27<46:19, 40.28s/it]
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[I 2025-10-25 22:01:35,138] Trial 30 finished with value:
0.9077444820090314 and parameters: {'n estimators': 676,
'learning_rate': 0.04397200172047998, 'max_depth': 9, 'subsample':
0.5436491959388067, 'colsample bytree': 0.4875984958782463.
'min child weight': 1, 'gamma': 0.5533494677640155, 'reg_alpha':
0.4794355056548615, 'reg lambda': 1.3327766780713328}. Best is trial
26 with value: 0.9086457350423436.
Best trial: 26. Best value: 0.908646: 32%
[27:06<45:19, 39.99s/it]
[I 2025-10-25 22:02:14,445] Trial 31 finished with value:
0.908227200652578 and parameters: {'n estimators': 647,
'learning rate': 0.022237311597359904, 'max depth': 8, 'subsample':
0.5541594156172124, 'colsample_bytree': 0.4902538782109293, 
'min_child_weight': 1, 'gamma': 0.5186781128397621, 'reg_alpha':
0.47042965114014085, 'reg lambda': 1.3375699175537465}. Best is trial
26 with value: 0.9086457350423436.
Best trial: 26. Best value: 0.908646: 33%
[27:47<44:53, 40.21s/it]
[I 2025-10-25 22:02:55,163] Trial 32 finished with value:
0.9080020065610113 and parameters: {'n estimators': 624,
'learning_rate': 0.021119171449964485, 'max_depth': 9, 'subsample':
0.5224287150834721, 'colsample bytree': 0.4864911141707048,
'min_child_weight': 1, 'gamma': 0.5309600205273494, 'reg_alpha':
0.46641575815324093, 'reg lambda': 1.3131135695280673}. Best is trial
26 with value: 0.9086457350423436.
Best trial: 26. Best value: 0.908646: 34%
[28:31<45:22, 41.25s/it]
[I 2025-10-25 22:03:38,860] Trial 33 finished with value:
0.9081629024131995 and parameters: {'n estimators': 664,
'learning rate': 0.023497181930487318, 'max depth': 8, 'subsample':
0.5077907995798814, 'colsample_bytree': 0.4756130375945638,
'min child weight': 1, 'gamma': 0.5013961902335302, 'reg alpha':
0.486743876538679, 'reg lambda': 1.4387686908125843}. Best is trial 26
with value: 0.9086457350423436.
Best trial: 26. Best value: 0.908646: 35%| | 35/100
[29:24<48:37, 44.89s/it]
[I 2025-10-25 22:04:32,236] Trial 34 finished with value:
0.9079053882235467 and parameters: {'n estimators': 649,
'learning rate': 0.02048683655753528, 'max depth': 9, 'subsample':
0.5949578643218633, 'colsample bytree': 0.4968860463915461.
'min_child_weight': 1, 'gamma': 0.5145941318231623, 'reg_alpha':
0.4435407856475014, 'reg lambda': 1.9195788460604775}. Best is trial
26 with value: 0.9086457350423436.
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Best trial: 26. Best value: 0.908646: 36%
[30:17<50:29, 47.34s/it]
[I 2025-10-25 22:05:25,293] Trial 35 finished with value:
0.9079698833471394 and parameters: {'n estimators': 710,
'learning rate': 0.022893833239092148, 'max depth': 8, 'subsample':
0.5662310962450503, 'colsample bytree': 0.50336543640576,
'min child weight': 1, 'gamma': 0.5387760976324443, 'reg alpha':
0.4107884821942376, 'reg lambda': 1.51672350640186}. Best is trial 26
with value: 0.9086457350423436.
Best trial: 26. Best value: 0.908646: 37%
[31:05<49:56, 47.56s/it]
[I 2025-10-25 22:06:13,363] Trial 36 finished with value:
0.9086457246800166 and parameters: {'n estimators': 682,
'learning rate': 0.027861180878566365, 'max_depth': 8, 'subsample':
0.5163184130166759, 'colsample bytree': 0.46651883665622335,
'min_child_weight': 1, 'gamma': 0.5290299995947705, 'reg_alpha': 0.4611186235326937, 'reg_lambda': 1.7397058459970678}. Best is trial
26 with value: 0.9086457350423436.
Best trial: 26. Best value: 0.908646: 38%
[32:00<51:31, 49.86s/it]
[I 2025-10-25 22:07:08,572] Trial 37 finished with value:
0.908066398061333 and parameters: {'n estimators': 682,
'learning_rate': 0.02897151865546144, 'max_depth': 9, 'subsample':
0.5546999220941732, 'colsample_bytree': 0.44207843172745015,
'min_child_weight': 1, 'gamma': 0.5669646187347603, 'reg_alpha': 0.460816300743621, 'reg_lambda': 1.7672761285974434}. Best is trial 26
with value: 0.9086457350423436.
Best trial: 26. Best value: 0.908646: 39%| | 39/100
[32:42<48:07, 47.33s/it]
[I 2025-10-25 22:07:50,012] Trial 38 finished with value:
0.9078410796218412 and parameters: {'n estimators': 690,
'learning rate': 0.026846990627451238, 'max depth': 7, 'subsample':
0.5808636043425881, 'colsample bytree': 0.5184434459974143,
'min_child_weight': 1, 'gamma': 0.544380531547993, 'reg_alpha':
0.49932641145883716, 'reg lambda': 1.6053590402517093}. Best is trial
26 with value: 0.9086457350423436.
Best trial: 26. Best value: 0.908646: 40%
[33:36<49:22, 49.37s/it]
[I 2025-10-25 22:08:44,147] Trial 39 finished with value:
0.9076156990083873 and parameters: {'n estimators': 708,
'learning_rate': 0.032901617337699786, 'max_depth': 8, 'subsample':
0.5268426824052267, 'colsample bytree': 0.42970757718779773,
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'min_child_weight': 1, 'gamma': 0.5314476671359004, 'reg_alpha':
0.4865390396089282, 'reg lambda': 2.1096300667432444}. Best is trial
26 with value: 0.9086457350423436.
Best trial: 26. Best value: 0.908646: 41%
[34:22<47:43, 48.54s/it]
[I 2025-10-25 22:09:30,737] Trial 40 finished with value:
0.9084527263386105 and parameters: {'n_estimators': 674,
'learning rate': 0.02527271402386223, 'max depth': 9, 'subsample':
0.5190346605177653, 'colsample_bytree': 0.4681158338984506,
'min_child_weight': 2, 'gamma': 0.5818622968829983, 'reg alpha':
0.38681031835316143, 'reg_lambda': 1.896515800583934}. Best is trial
26 with value: 0.9086457350423436.
Best trial: 26. Best value: 0.908646: 42%
[35:12<47:18, 48.95s/it]
[I 2025-10-25 22:10:20,648] Trial 41 finished with value:
0.9077767606578089 and parameters: {'n estimators': 671,
'learning rate': 0.02500441334730802, 'max depth': 9, 'subsample':
0.5169879922652264, 'colsample_bytree': 0.4676207958465193, 
'min_child_weight': 2, 'gamma': 0.5776364185726083, 'reg_alpha':
0.385540891814151, 'req lambda': 1.8840785808044824}. Best is trial 26
with value: 0.9086457350423436.
Best trial: 26. Best value: 0.908646: 43%
[35:59<45:54, 48.33s/it]
[I 2025-10-25 22:11:07,529] Trial 42 finished with value:
0.9081629542248347 and parameters: {'n estimators': 675,
'learning_rate': 0.02909535595727217, 'max_depth': 9, 'subsample':
0.5272521867559846, 'colsample bytree': 0.456693200078941,
'min child weight': 2, 'gamma': 0.5562308643457521, 'reg alpha':
0.35775986664969417, 'reg lambda': 1.7431558997557113}. Best is trial
26 with value: 0.9086457350423436.
Best trial: 26. Best value: 0.908646: 44%
[36:49<45:23, 48.63s/it]
[I 2025-10-25 22:11:56,847] Trial 43 finished with value:
0.908195191424304 and parameters: {'n_estimators': 752,
'learning_rate': 0.0272632605699647, 'max_depth': 9, 'subsample':
0.5188793777262666, 'colsample_bytree': 0.45947172671749265,
'min_child_weight': 2, 'gamma': 0.5761595521953012, 'reg_alpha':
0.36965070686266394, 'reg lambda': 1.8481842904874464}. Best is trial
26 with value: 0.9086457350423436.
Best trial: 26. Best value: 0.908646: 45%| 45/100
[37:54<49:07, 53.59s/it]
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[I 2025-10-25 22:13:02,016] Trial 44 finished with value:
0.9073903702065689 and parameters: {'n estimators': 640,
'learning_rate': 0.035858841091256886, 'max_depth': 9, 'subsample': 0.5073080408909103, 'colsample_bytree': 0.4762515327299858,
'min child weight': 2, 'gamma': 0.586502652374941, 'reg alpha':
0.4329211061532068, 'reg lambda': 2.1019701683548266}. Best is trial
26 with value: 0.9086457350423436.
Best trial: 26. Best value: 0.908646: 46%
[39:16<55:58, 62.19s/it]
[I 2025-10-25 22:14:24,281] Trial 45 finished with value:
0.9076157922693311 and parameters: {'n estimators': 669,
'learning rate': 0.025310806046445155, 'max depth': 8, 'subsample':
0.5342598812417759, 'colsample_bytree': 0.47030329478513144, 
'min_child_weight': 2, 'gamma': 0.5102139405359016, 'reg_alpha':
0.3906332726659051, 'req lambda': 1.969881847015458}. Best is trial 26
with value: 0.9086457350423436.
Best trial: 26. Best value: 0.908646: 47%
[39:46<46:30, 52.66s/it]
[I 2025-10-25 22:14:54,690] Trial 46 finished with value:
0.9077122758965434 and parameters: {'n estimators': 680,
'learning_rate': 0.055858196626611234, 'max_depth': 9, 'subsample':
0.5477449844390416, 'colsample bytree': 0.44173939696728054,
'min_child_weight': 1, 'gamma': 0.5249812021320089, 'reg_alpha':
0.334108476351373, 'reg lambda': 1.7085603339477196}. Best is trial 26
with value: 0.9086457350423436.
Best trial: 26. Best value: 0.908646: 48%
[40:42<46:23, 53.53s/it]
[I 2025-10-25 22:15:50,247] Trial 47 finished with value:
0.9077444716467044 and parameters: {'n_estimators': 715,
'learning rate': 0.023936866945307375, 'max depth': 9, 'subsample':
0.5408961790276164, 'colsample_bytree': 0.45464690921004497,
'min child weight': 1, 'gamma': 0.5430462473769331, 'reg alpha':
0.44767682493874145, 'reg lambda': 1.811146046666056}. Best is trial
26 with value: 0.9086457350423436.
Best trial: 26. Best value: 0.908646: 49%| 49/100
[41:43<47:19, 55.68s/it]
[I 2025-10-25 22:16:50,938] Trial 48 finished with value:
0.9077445234583397 and parameters: {'n estimators': 663,
'learning rate': 0.028376046883641304, 'max depth': 8, 'subsample':
0.5562435810283083, 'colsample bytree': 0.48364841201399694,
'min_child_weight': 2, 'gamma': 0.6051919935032882, 'reg_alpha':
0.36837867636382016, 'reg lambda': 1.6161687774915814}. Best is trial
26 with value: 0.9086457350423436.
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Best trial: 49. Best value: 0.908646: 50%
[42:34<45:17, 54.35s/it]
[I 2025-10-25 22:17:42,188] Trial 49 finished with value:
0.9086457557669976 and parameters: {'n estimators': 690,
'learning rate': 0.030037704814410867, 'max_depth': 9, 'subsample':
0.5052173648962378, 'colsample bytree': 0.43423356408714325,
'min child weight': 1, 'gamma': 0.6216534827037803, 'reg alpha':
0.4246693469203213, 'reg lambda': 1.507316053868602}. Best is trial 49
with value: 0.9086457557669976.
Best trial: 49. Best value: 0.908646: 51%
[43:20<42:16, 51.77s/it]
[I 2025-10-25 22:18:27,953] Trial 50 finished with value:
0.9082594067650659 and parameters: {'n estimators': 699,
'learning_rate': 0.029687550750951266, 'max_depth': 9, 'subsample':
0.5068595735572486, 'colsample bytree': 0.46734796726347055,
'min_child_weight': 1, 'gamma': 0.6220477643752467, 'reg_alpha': 0.42676335657837416, 'reg_lambda': 1.5058774114204745}. Best is trial
49 with value: 0.9086457557669976.
Best trial: 49. Best value: 0.908646: 52%
[43:58<38:18, 47.88s/it]
[I 2025-10-25 22:19:06,736] Trial 51 finished with value:
0.9080663669743518 and parameters: {'n estimators': 689,
'learning_rate': 0.027507124122712265, 'max_depth': 9, 'subsample':
0.5224435301896845, 'colsample_bytree': 0.43466246272957604,
'min child weight': 1, 'gamma': 0.6347967694981524, 'reg alpha':
0.40631410726656, 'reg lambda': 1.4109699272225176}. Best is trial 49
with value: 0.9086457557669976.
[44:47<37:33, 47.95s/it]
[I 2025-10-25 22:19:54,856] Trial 52 finished with value:
0.908195222511285 and parameters: {'n_estimators': 635,
'learning rate': 0.03200797807425661, 'max depth': 9, 'subsample':
0.5156531062426246, 'colsample bytree': 0.4207197237749554,
'min_child_weight': 1, 'gamma': 0.6320182721961446, 'reg_alpha':
0.45388497890776813, 'reg lambda': 1.4629740930505837}. Best is trial
49 with value: 0.9086457557669976.
Best trial: 53. Best value: 0.908871: 54%
[45:23<34:08, 44.54s/it]
[I 2025-10-25 22:20:31,432] Trial 53 finished with value:
0.9088711674674327 and parameters: {'n estimators': 727,
'learning_rate': 0.030483044346389376, 'max_depth': 9, 'subsample':
0.5004265648596836, 'colsample bytree': 0.44598996260357027,
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'min_child_weight': 1, 'gamma': 0.5982796330547343, 'reg_alpha':
0.4401589647055334, 'reg lambda': 1.6367059914120072}. Best is trial
53 with value: 0.9088711674674327.
Best trial: 53. Best value: 0.908871: 55%| | 55/100
[46:02<32:07, 42.82s/it]
[I 2025-10-25 22:21:10,265] Trial 54 finished with value:
0.9081307481123467 and parameters: {'n estimators': 739,
'learning_rate': 0.030111746283372015, 'max_depth': 9, 'subsample':
0.5044348444115111, 'colsample_bytree': 0.4472212213204126,
'min child weight': 1, 'gamma': 0.5969783583139726, 'reg alpha':
0.43792631539121857, 'reg_lambda': 1.6632472022623797}. Best is trial
53 with value: 0.90887116\overline{7}4674327.
Best trial: 53. Best value: 0.908871: 56%
[46:37<29:38, 40.42s/it]
[I 2025-10-25 22:21:45,065] Trial 55 finished with value:
0.9079375528867267 and parameters: {'n estimators': 723,
'learning rate': 0.0339757163114496, 'max depth': 8, 'subsample':
0.5004662168151666, 'colsample bytree': 0.4268061116934629,
'min child weight': 1, 'gamma': 0.6130269724646383, 'reg alpha':
0.39369351219866855, 'reg lambda': 1.5799268934714803}. Best is trial
53 with value: 0.9088711674674327.
Best trial: 53. Best value: 0.908871: 57%| 57/100
[47:21<29:53, 41.71s/it]
[I 2025-10-25 22:22:29,792] Trial 56 finished with value:
0.9078733479082919 and parameters: {'n estimators': 761,
'learning_rate': 0.03865896102587614, 'max_depth': 9, 'subsample':
0.5093012733896611, 'colsample bytree': 0.43900065896118584,
'min_child_weight': 1, 'gamma': 0.6721222282966456, 'reg_alpha':
0.42181632847600686, 'reg lambda': 1.7020046144919954}. Best is trial
53 with value: 0.9088711674674327.
Best trial: 53. Best value: 0.908871: 58%
[48:03<29:12, 41.73s/it]
[I 2025-10-25 22:23:11,583] Trial 57 finished with value:
0.9079698418978313 and parameters: {'n estimators': 734,
'learning rate': 0.031349451073230596, 'max depth': 8, 'subsample':
0.5043970077173849, 'colsample bytree': 0.4795077574690376,
'min child weight': 1, 'gamma': 0.5662974157374353, 'reg alpha':
0.41059845378295734, 'reg lambda': 1.8271985549333312}. Best is trial
53 with value: 0.9088711674674327.
Best trial: 53. Best value: 0.908871: 59%| 59/100
[48:46<28:45, 42.07s/it]
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[I 2025-10-25 22:23:54,434] Trial 58 finished with value:
0.9084525605413777 and parameters: {'n estimators': 707,
'learning_rate': 0.025999437372193315, 'max_depth': 7, 'subsample': 0.5134517437397399, 'colsample_bytree': 0.4500993559885126,
'min child weight': 1, 'gamma': 0.5836158053546611, 'reg alpha':
0.4391026647313721, 'reg_lambda': 1.9220661599276059}. Best is trial
53 with value: 0.9088711674674327.
Best trial: 53. Best value: 0.908871: 60%
[49:30<28:29, 42.74s/it]
[I 2025-10-25 22:24:38,754] Trial 59 finished with value:
0.90825942748972 and parameters: {'n estimators': 694,
'learning rate': 0.034209906954696605, 'max depth': 8, 'subsample':
0.5201357848818416, 'colsample_bytree': 0.46098633467817324, 
'min_child_weight': 1, 'gamma': 0.6004139977714537, 'reg_alpha':
0.4284355702844147, 'reg lambda': 1.7441978281732835}. Best is trial
53 with value: 0.9088711674674327.
Best trial: 60. Best value: 0.908968: 61%
[50:15<28:14, 43.44s/it]
[I 2025-10-25 22:25:23,791] Trial 60 finished with value:
0.9089677339932616 and parameters: {'n estimators': 729,
'learning rate': 0.028012343066800567, 'max depth': 9, 'subsample':
0.5003749074055666, 'colsample bytree': 0.4661611876848136,
'min_child_weight': 1, 'gamma': 0.6447719865208364, 'reg_alpha':
0.45652964466691026, 'reg lambda': 1.2670782078774123}. Best is trial
60 with value: 0.9089677339932616.
Best trial: 60. Best value: 0.908968: 62%
[51:06<28:48, 45.49s/it]
[I 2025-10-25 22:26:14,096] Trial 61 finished with value:
0.9082917372254787 and parameters: {'n estimators': 748,
'learning_rate': 0.028064880886148173, 'max depth': 9, 'subsample':
0.5007841724203529, 'colsample_bytree': 0.4653598789283857,
'min child weight': 1, 'gamma': 0.6495600881874664, 'reg alpha':
0.4526499479180536, 'reg lambda': 1.385796505619553}. Best is trial 60
with value: 0.9089677339932616.
Best trial: 60. Best value: 0.908968: 63%| 63/100
[51:47<27:16, 44.23s/it]
[I 2025-10-25 22:26:55,376] Trial 62 finished with value:
0.9083238604393502 and parameters: {'n estimators': 717,
'learning rate': 0.03006398340067283, 'max depth': 9, 'subsample':
0.5057670686966359, 'colsample bytree': 0.4728385876079455.
'min_child_weight': 1, 'gamma': 0.6414670160462936, 'reg_alpha':
0.4580088997201982, 'reg lambda': 1.4604884884157294}. Best is trial
60 with value: 0.9089677339932616.
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Best trial: 60. Best value: 0.908968: 64%
[52:34<27:03, 45.09s/it]
[I 2025-10-25 22:27:42,489] Trial 63 finished with value:
0.9086136532777802 and parameters: {'n estimators': 729,
'learning rate': 0.026450257608473783, 'max_depth': 9, 'subsample':
0.5098975128500058, 'colsample bytree': 0.45894537723515744,
'min child weight': 1, 'gamma': 0.6202715441774569, 'reg alpha':
0.44475864108624064, 'req lambda': 1.258795745074992}. Best is trial
60 with value: 0.9089677339932616.
Best trial: 60. Best value: 0.908968: 65%
[53:18<26:04, 44.69s/it]
[I 2025-10-25 22:28:26,228] Trial 64 finished with value:
0.9080343162967697 and parameters: {'n estimators': 730,
'learning_rate': 0.02658148414787226, 'max_depth': 9, 'subsample':
0.5119808467466027, 'colsample_bytree': 0.44389933475381554,
'min_child_weight': 1, 'gamma': 0.6192484441158096, 'reg_alpha': 0.4459904998329613, 'reg_lambda': 1.270092814048034}. Best is trial 60
with value: 0.9089677339932616.
Best trial: 60. Best value: 0.908968: 66%
[53:41<21:42, 38.32s/it]
[I 2025-10-25 22:28:49,679] Trial 65 finished with value:
0.9079698315355043 and parameters: {'n estimators': 745,
'learning_rate': 0.059913544715903384, 'max_depth': 9, 'subsample':
0.5108947855308464, 'colsample_bytree': 0.4583580347663187,
'min child weight': 1, 'gamma': 0.6599903520276029, 'reg alpha':
0.46673667712338457, 'reg_lambda': 1.2123539087414121}. Best is trial
60 with value: 0.9089677339932616.
Best trial: 60. Best value: 0.908968: 67%
[54:23<21:39, 39.39s/it]
[I 2025-10-25 22:29:31.577] Trial 66 finished with value:
0.90774445092205 and parameters: {'n estimators': 790,
'learning rate': 0.03105663686472816, 'max depth': 8, 'subsample':
0.5033386444723744, 'colsample bytree': 0.4506124254966358,
'min child weight': 1, 'gamma': 0.6765127980540131, 'reg alpha':
0.4331022193106324, 'reg lambda': 1.3583729441941024}. Best is trial
60 with value: 0.9089677339932616.
Best trial: 60. Best value: 0.908968: 68%
[55:30<25:26, 47.70s/it]
[I 2025-10-25 22:30:38,662] Trial 67 finished with value:
0.9084527781502458 and parameters: {'n estimators': 765.
'learning rate': 0.03305568928504436, 'max depth': 9, 'subsample':
0.5084126447109683, 'colsample bytree': 0.4359390520716674,
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'min_child_weight': 1, 'gamma': 0.6225860759416845, 'reg_alpha':
0.4409361839053979, 'reg lambda': 1.259840447911629}. Best is trial 60
with value: 0.9089677339932616.
Best trial: 60. Best value: 0.908968: 69%| 69/100
[56:06<22:45, 44.04s/it]
[I 2025-10-25 22:31:14,165] Trial 68 finished with value:
0.9081307791993278 and parameters: {'n_estimators': 777,
'learning_rate': 0.035923253428208984, 'max_depth': 8, 'subsample':
0.5151826764726257, 'colsample_bytree': 0.4532459951749072,
'min child weight': 1, 'gamma': 0.63994721596588, 'reg alpha':
0.44968651493082057, 'reg lambda': 1.2973999857195302}. Best is trial
60 with value: 0.9089677339932616.
Best trial: 60. Best value: 0.908968: 70%
[56:53<22:33, 45.13s/it]
[I 2025-10-25 22:32:01,836] Trial 69 finished with value:
0.9083882830266534 and parameters: {'n estimators': 730,
'learning_rate': 0.028045941839743016, 'max depth': 9, 'subsample':
0.5034661381030239, 'colsample bytree': 0.46420228495511984,
'min child_weight': 1, 'gamma': 0.630348031511651, 'reg_alpha':
0.41730002827167156, 'reg lambda': 1.2343301898691739}. Best is trial
60 with value: 0.9089677339932616.
Best trial: 60. Best value: 0.908968: 71%
[57:43<22:28, 46.50s/it]
[I 2025-10-25 22:32:51,525] Trial 70 finished with value:
0.9082273664498108 and parameters: {'n estimators': 703,
'learning_rate': 0.029298429474779415, 'max_depth': 9, 'subsample':
0.5001644696257844, 'colsample bytree': 0.44460622762949303,
'min_child_weight': 1, 'gamma': 0.6108232397262573, 'reg_alpha':
0.46011020703938693, 'reg lambda': 1.5474968326692684}. Best is trial
60 with value: 0.9089677339932616.
Best trial: 60. Best value: 0.908968: 72%
[58:43<23:36, 50.60s/it]
[I 2025-10-25 22:33:51,697] Trial 71 finished with value:
0.9085170660272972 and parameters: {'n_estimators': 799,
'learning_rate': 0.03293421972600636, 'max_depth': 9, 'subsample':
0.5089986154186801, 'colsample_bytree': 0.43419364880748684,
'min child weight': 1, 'gamma': 0.6237449237923022, 'reg alpha':
0.44101144358570393, 'reg lambda': 1.2790160298952389}. Best is trial
60 with value: 0.9089677339932616.
Best trial: 60. Best value: 0.908968: 73%| 73/100
[59:24<21:27, 47.67s/it]
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[I 2025-10-25 22:34:32,547] Trial 72 finished with value:
0.908774621666258 and parameters: {'n estimators': 800,
'learning_rate': 0.03193592026838488, 'max_depth': 9, 'subsample':
0.5093854110298408, 'colsample_bytree': 0.43060438542183466,
'min child weight': 1, 'gamma': 0.6269351321171152, 'reg alpha':
0.4268336482616307, 'reg_lambda': 1.3772154838183597}. Best is trial
60 with value: 0.9089677339932616.
Best trial: 60. Best value: 0.908968: 74%
[1:00:04<19:38, 45.31s/it]
[I 2025-10-25 22:35:12,335] Trial 73 finished with value:
0.9085814057159839 and parameters: {'n_estimators': 723,
'learning rate': 0.03128736341108016, 'max depth': 9, 'subsample':
0.505834144431345, 'colsample_bytree': 0.43962107651623256, 
'min_child_weight': 1, 'gamma': 0.6521076426899285, 'reg_alpha':
0.4256264249857798, 'req lambda': 1.371160707050673}. Best is trial 60
with value: 0.9089677339932616.
Best trial: 60. Best value: 0.908968: 75%
[1:00:45<18:17, 43.89s/it]
[I 2025-10-25 22:35:52,909] Trial 74 finished with value:
0.9080985938114938 and parameters: {'n estimators': 738,
'learning_rate': 0.03169490018717343, 'max_depth': 9, 'subsample':
0.5054468908204506, 'colsample bytree': 0.43843514816537205,
'min_child_weight': 1, 'gamma': 0.6549433953958227, 'reg_alpha':
0.422829813639814, 'reg lambda': 1.3910723600907386}. Best is trial 60
with value: 0.9089677339932616.
Best trial: 60. Best value: 0.908968: 76%
[1:01:27<17:24, 43.50s/it]
[I 2025-10-25 22:36:35,518] Trial 75 finished with value:
0.9075514007690089 and parameters: {'n estimators': 724,
'learning_rate': 0.03036331748189196, 'max_depth': 9, 'subsample':
0.5165852312815752, 'colsample_bytree': 0.4321545294017978,
'min child weight': 1, 'gamma': 0.6463249748008881, 'reg alpha':
0.4129064456543715, 'reg lambda': 1.3647675979950067}. Best is trial
60 with value: 0.9089677339932616.
Best trial: 60. Best value: 0.908968: 77%| 77/100
[1:02:04<15:54, 41.48s/it]
[I 2025-10-25 22:37:12,289] Trial 76 finished with value:
0.9082596243739343 and parameters: {'n estimators': 753,
'learning rate': 0.03487575415836668, 'max depth': 9, 'subsample':
0.5127894538881921, 'colsample bytree': 0.4261162728185798.
'min_child_weight': 1, 'gamma': 0.6691329539504541, 'reg_alpha':
0.4280271686693906, 'reg lambda': 1.4376242097651764}. Best is trial
60 with value: 0.9089677339932616.
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Best trial: 60. Best value: 0.908968: 78%
[1:02:45<15:07, 41.27s/it]
[I 2025-10-25 22:37:53,058] Trial 77 finished with value:
0.9075192050188481 and parameters: {'n estimators': 714,
'learning rate': 0.024477021736405532, 'max depth': 9, 'subsample':
0.5835339984624044, 'colsample bytree': 0.4286251396844537,
'min child weight': 1, 'gamma': 0.6417736781367636, 'reg alpha':
0.4034675272894454, 'reg lambda': 1.3126331169913952}. Best is trial
60 with value: 0.9089677339932616.
Best trial: 60. Best value: 0.908968: 79%
[1:03:21<13:53, 39.68s/it]
[I 2025-10-25 22:38:29,028] Trial 78 finished with value:
0.9079699247964477 and parameters: {'n estimators': 729,
'learning_rate': 0.05246607983345815, 'max_depth': 9, 'subsample':
0.5020391993106116, 'colsample bytree': 0.4465399293625719,
'min_child_weight': 1, 'gamma': 0.6168937732214402, 'reg_alpha': 0.4722556368159838, 'reg_lambda': 1.4756180717213432}. Best is trial
60 with value: 0.9089677339932616.
Best trial: 60. Best value: 0.908968: 80%| 80/100 | 80/100
[1:04:01<13:14, 39.73s/it]
[I 2025-10-25 22:39:08,877] Trial 79 finished with value:
0.9081951914243038 and parameters: {'n estimators': 779,
'learning_rate': 0.02880203921311604, 'max_depth': 9, 'subsample':
0.5240352771372684, 'colsample_bytree': 0.4238197773676324,
'min child weight': 1, 'gamma': 0.6821811795533939, 'reg alpha':
0.45049794486327244, 'reg lambda': 1.2231906232025505}. Best is trial
60 with value: 0.9089677339932616.
Best trial: 60. Best value: 0.908968: 81%| 81/100
[1:04:35<12:05, 38.20s/it]
[I 2025-10-25 22:39:43,498] Trial 80 finished with value:
0.9080020894596276 and parameters: {'n_estimators': 720,
'learning rate': 0.03700946715957194, 'max depth': 9, 'subsample':
0.5105059701911738, 'colsample bytree': 0.4396934626024846,
'min child weight': 1, 'gamma': 0.6358341665549075, 'reg alpha':
0.4331811318290038, 'reg lambda': 1.3700576466149217}. Best is trial
60 with value: 0.9089677339932616.
Best trial: 60. Best value: 0.908968: 82%| 82/100 | 82/100
[1:05:18<11:53, 39.62s/it]
[I 2025-10-25 22:40:26,433] Trial 81 finished with value:
0.9088388991809824 and parameters: {'n estimators': 696,
'learning_rate': 0.025974430813807733, 'max_depth': 9, 'subsample':
0.5066714212838316, 'colsample bytree': 0.47604995807857964,
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'min_child_weight': 1, 'gamma': 0.6536889715027272, 'reg_alpha':
0.4221493842321466, 'reg lambda': 1.6368404450406693}. Best is trial
60 with value: 0.9089677339932616.
Best trial: 60. Best value: 0.908968: 83%| 83/100
[1:06:01<11:27, 40.46s/it]
[I 2025-10-25 22:41:08,869] Trial 82 finished with value:
0.9083238915263315 and parameters: {'n_estimators': 703,
'learning rate': 0.02715436665388998, 'max depth': 9, 'subsample':
0.506555027348645, 'colsample_bytree': 0.4761192255791451,
'min child weight': 1, 'gamma': 0.6533845470499938, 'reg alpha':
0.4219169477577072, 'reg lambda': 1.6152601777140656}. Best is trial
60 with value: 0.9089677339932616.
Best trial: 60. Best value: 0.908968: 84%
[1:06:37<10:30, 39.38s/it]
[I 2025-10-25 22:41:45,727] Trial 83 finished with value:
0.9083238708016772 and parameters: {'n estimators': 685,
'learning_rate': 0.031059630063324267, 'max_depth': 9, 'subsample':
0.5035866191903011, 'colsample_bytree': 0.4609303797007522, 
'min_child_weight': 1, 'gamma': 0.6262397628423362, 'reg_alpha':
0.44491112818691775, 'reg lambda': 1.2909033851770522}. Best is trial
60 with value: 0.9089677339932616.
Best trial: 60. Best value: 0.908968: 85%| | 85/100
[1:07:18<09:56, 39.79s/it]
[I 2025-10-25 22:42:26,471] Trial 84 finished with value:
0.9084527470632645 and parameters: {'n estimators': 694,
'learning rate': 0.022774223453191064, 'max depth': 9, 'subsample':
0.514084876966977, 'colsample bytree': 0.43233115458104815,
'min_child_weight': 1, 'gamma': 0.662267754099006, 'reg_alpha':
0.4626722869282454, 'reg lambda': 1.4950744149054067}. Best is trial
60 with value: 0.9089677339932616.
Best trial: 60. Best value: 0.908968: 86%
[1:08:11<10:11, 43.67s/it]
[I 2025-10-25 22:43:19,213] Trial 85 finished with value:
0.9084204580521602 and parameters: {'n_estimators': 736,
'learning rate': 0.028053818628521934, 'max depth': 9, 'subsample':
0.5067119972752134, 'colsample_bytree': 0.49349518295650124,
'min child weight': 1, 'gamma': 0.6479904546256879, 'reg alpha':
0.4249699216968398, 'reg lambda': 1.425561823306235}. Best is trial 60
with value: 0.9089677339932616.
Best trial: 60. Best value: 0.908968: 87%| 87/100
[1:08:53<09:20, 43.13s/it]
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[I 2025-10-25 22:44:01,083] Trial 86 finished with value:
0.9082917475878057 and parameters: {'n estimators': 605,
'learning_rate': 0.02669608110956047, 'max_depth': 9, 'subsample':
0.5099896614789737, 'colsample_bytree': 0.4560795445967504,
'min_child_weight': 1, 'gamma': 0.6059296579167244, 'reg_alpha':
0.45664356590780697, 'reg lambda': 1.6537237179103448}. Best is trial
60 with value: 0.9089677339932616.
Best trial: 60. Best value: 0.908968: 88%| 88/100 | 88/100
[1:09:37<08:40, 43.36s/it]
[I 2025-10-25 22:44:44,965] Trial 87 finished with value:
0.9086457661293247 and parameters: {'n estimators': 711,
'learning_rate': 0.029217515148365226, 'max depth': 9, 'subsample':
0.5183608222433225, 'colsample_bytree': 0.4819509079565678, 
'min_child_weight': 1, 'gamma': 0.6589704255661333, 'reg_alpha':
0.43308736368392164, 'reg lambda': 1.346554142481206}. Best is trial
60 with value: 0.9089677339932616.
Best trial: 60. Best value: 0.908968: 89%| 89/100 | 89/100
[1:10:20<07:57, 43.40s/it]
[I 2025-10-25 22:45:28,457] Trial 88 finished with value:
0.908259614011607 and parameters: {'n estimators': 712,
'learning rate': 0.029556118591849385, 'max depth': 9, 'subsample':
0.5188170770655134, 'colsample bytree': 0.48172734076400375,
'min_child_weight': 1, 'gamma': 0.6671794787694574, 'reg_alpha':
0.4329322406233371, 'reg lambda': 1.3379556738746579}. Best is trial
60 with value: 0.9089677339932616.
Best trial: 60. Best value: 0.908968: 90%
[1:11:19<07:59, 47.97s/it]
[I 2025-10-25 22:46:27,091] Trial 89 finished with value:
0.9080663876990058 and parameters: {'n estimators': 706,
'learning_rate': 0.021096463129838205, 'max_depth': 9, 'subsample':
0.5024829546158213, 'colsample_bytree': 0.48848045806364504,
'min child weight': 1, 'gamma': 0.6299962956671482, 'reg alpha':
0.46579828169976906, 'reg lambda': 1.5874111948782343}. Best is trial
60 with value: 0.9089677339932616.
Best trial: 60. Best value: 0.908968: 91%| 91/100
[1:12:04<07:04, 47.21s/it]
[I 2025-10-25 22:47:12,537] Trial 90 finished with value:
0.9079376150606888 and parameters: {'n estimators': 690,
'learning_rate': 0.02449656324694884, 'max_depth': 9, 'subsample':
0.5306641202827759, 'colsample bytree': 0.4836299964914089.
'min_child_weight': 1, 'gamma': 0.6385692907977946, 'reg_alpha':
0.41676430112366447, 'reg lambda': 1.6917771666396313}. Best is trial
60 with value: 0.9089677339932616.
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Best trial: 60. Best value: 0.908968: 92%
[1:12:56<06:28, 48.53s/it]
[I 2025-10-25 22:48:04,126] Trial 91 finished with value:
0.9083882830266534 and parameters: {'n estimators': 727,
'learning_rate': 0.02858632402222474, 'max_depth': 9, 'subsample':
0.5169273412291121, 'colsample bytree': 0.48551789433097026,
'min child weight': 1, 'gamma': 0.6940707427778174, 'reg alpha':
0.43709798273950734, 'req lambda': 1.2535596280684644}. Best is trial
60 with value: 0.9089677339932616.
Best trial: 60. Best value: 0.908968: 93%
[1:13:46<05:42, 48.90s/it]
[I 2025-10-25 22:48:53,915] Trial 92 finished with value:
0.9081950878010332 and parameters: {'n estimators': 722,
'learning_rate': 0.027522387059566798, 'max_depth': 9, 'subsample':
0.505324450973277, 'colsample_bytree': 0.47576753391219423,
'min_child_weight': 1, 'gamma': 0.6447649827086108, 'reg_alpha':
0.42987066699761983, 'reg lambda': 1.299876974444279}. Best is trial
60 with value: 0.9089677339932616.
Best trial: 60. Best value: 0.908968: 94%
[1:14:22<04:31, 45.18s/it]
[I 2025-10-25 22:49:30,407] Trial 93 finished with value:
0.9078409863608978 and parameters: {'n estimators': 698,
'learning_rate': 0.032456260676073566, 'max_depth': 9, 'subsample':
0.5000202439585713, 'colsample_bytree': 0.4782125419979585,
'min child weight': 1, 'gamma': 0.6609006084687494, 'reg alpha':
0.4478668042869812, 'reg lambda': 1.40233376048445}. Best is trial 60
with value: 0.9089677339932616.
Best trial: 60. Best value: 0.908968: 95%| 95/100
[1:15:12<03:53, 46.64s/it]
[I 2025-10-25 22:50:20.452] Trial 94 finished with value:
0.9080663358873705 and parameters: {'n estimators': 743,
'learning rate': 0.030080275652736896, 'max depth': 9, 'subsample':
0.5084968947584984, 'colsample bytree': 0.4731563715210445,
'min child weight': 1, 'gamma': 0.652028571348963, 'reg alpha':
0.4085338753590346, 'reg lambda': 1.2390563899757039}. Best is trial
60 with value: 0.9089677339932616.
Best trial: 60. Best value: 0.908968: 96%| 96/100
[1:15:57<03:03, 45.98s/it]
[I 2025-10-25 22:51:04,881] Trial 95 finished with value:
0.9083883244759615 and parameters: {'n estimators': 716.
'learning_rate': 0.03064637771871952, 'max_depth': 9, 'subsample':
0.5114308357656866, 'colsample bytree': 0.4679140785585934,
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'min_child_weight': 1, 'gamma': 0.6837400792325293, 'reg_alpha':
0.39825551847659496, 'reg lambda': 1.526613556595299}. Best is trial
60 with value: 0.9089677339932616.
Best trial: 60. Best value: 0.908968: 97%| 97/100
[1:17:49<03:17, 65.83s/it]
[I 2025-10-25 22:52:57,039] Trial 96 finished with value:
0.9082273871744648 and parameters: {'n_estimators': 678,
'learning rate': 0.03343670214951915, 'max depth': 9, 'subsample':
0.513646808123811, 'colsample_bytree': 0.47083822246294244,
'min child weight': 1, 'gamma': 0.658527481486338, 'reg alpha':
0.44059186506290754, 'reg_lambda': 1.5715065237773713}. Best is trial
60 with value: 0.9089677339932616.
Best trial: 60. Best value: 0.908968: 98%| 98/100
[1:18:53<02:10, 65.22s/it]
[I 2025-10-25 22:54:00,843] Trial 97 finished with value:
0.9078732650096754 and parameters: {'n estimators': 684,
'learning_rate': 0.03193424211138466, 'max depth': 9, 'subsample':
0.5240375683616427, 'colsample bytree': 0.43262566690011134,
'min child weight': 1, 'gamma': 0.5277832923767415, 'reg alpha':
0.48060378331034737, 'reg lambda': 1.3521854447930128}. Best is trial
60 with value: 0.9089677339932616.
Best trial: 60. Best value: 0.908968: 99%| 99/100
[1:20:14<01:10, 70.19s/it]
[I 2025-10-25 22:55:22,625] Trial 98 finished with value:
0.9084204684144872 and parameters: {'n estimators': 709,
'learning_rate': 0.026121854858440638, 'max_depth': 9, 'subsample':
0.5024751830846097, 'colsample bytree': 0.4376866652652031,
'min_child_weight': 1, 'gamma': 0.6349197065414566, 'reg_alpha':
0.4194312397863926, 'reg lambda': 1.3261887988116219}. Best is trial
60 with value: 0.9089677339932616.
Best trial: 60. Best value: 0.908968: 100% | 100/100
[1:21:02<00:00, 48.62s/it]
[I 2025-10-25 22:56:10,092] Trial 99 finished with value:
0.9081306859383844 and parameters: {'n_estimators': 659,
'learning rate': 0.029557890752313756, 'max depth': 9, 'subsample':
0.5075321783609258, 'colsample_bytree': 0.46533390201416847,
'min child weight': 1, 'gamma': 0.6667195894248964, 'reg alpha':
0.4553316128370863, 'reg_lambda': 1.6302325082186848}. Best is trial
60 with value: 0.9089677339932616.
Best CV accuracy: 0.90897
Best parameters found:
```

```
n_estimators: 729
learning_rate: 0.028012343066800567
max_depth: 9
subsample: 0.5003749074055666
colsample_bytree: 0.4661611876848136
min_child_weight: 1
gamma: 0.6447719865208364
reg_alpha: 0.45652964466691026
reg_lambda: 1.2670782078774123
```

Best output accuracy: - .91322

```
import pandas as pd
import numpy as np
import time
import warnings
from sklearn.preprocessing import StandardScaler, LabelEncoder
from sklearn.model selection import train test split
import xgboost as xgb
from sklearn.metrics import accuracy score, classification report
warnings.filterwarnings("ignore")
file path base = '/content/drive/MyDrive/Obesity Data/'
df = pd.read csv(file path base + 'train.csv')
X = df.drop(['id', 'WeightCategory'], axis=1)
y = df['WeightCategory']
categorical cols = X.select dtypes(include=['object']).columns
X = pd.get dummies(X, columns=categorical cols, drop first=True)
numerical cols =
['Age','Height','Weight','FCVC','NCP','CH20','FAF','TUE']
numerical cols present = [col for col in numerical cols if col in
X.columns1
scaler = StandardScaler()
X[numerical cols present] =
scaler.fit transform(X[numerical cols present])
le = LabelEncoder()
y enc = le.fit_transform(y)
X_train, X_val, y_train_enc, y_val_enc = train_test split(
   X, y_enc, test_size=0.20, stratify=y_enc, random_state=42
)
_, _, y_train_orig, y_val_orig = train test split(
    X, y, test_size=0.20, stratify=y, random state=42
```

```
)
best optuna params = {
    'n estimators': 685,
    'learning rate': 0.031059630063324267,
    'max depth': 9,
    'subsample': 0.5035866191903011,
    'colsample bytree': 0.4609303797007522,
    'min_child_weight': 1,
    'gamma': 0.6262397628423362,
    'reg alpha': 0.44491112818691775,
    'reg lambda': 1.2909033851770522,
    'gamma' : 0.5,
    'objective': 'multi:softmax',
    'num class': len(le.classes ),
    'use_label_encoder': False,
    'eval metric': 'mlogloss',
    'random state': 42,
    'n jobs': -1,
    'verbosity': 0,
    'early stopping rounds': 50
}
model = xgb.XGBClassifier(**best optuna params)
start time = time.time()
model.fit(X train, y train enc, eval set=[(X val, y val enc)],
verbose=False)
end time = time.time()
best iteration = model.get booster().best iteration
y pred enc = model.predict(X val)
y pred = le.inverse transform(y pred enc)
accuracy = accuracy_score(y_val_orig, y_pred)
print(f"Validation Accuracy: {accuracy:.5f}")
print(classification report(y val orig, y pred, zero division=0))
print(f"Training time: {(end time - start time):.2f} seconds")
print(f"Best iteration: {best iteration}")
Validation Accuracy: 0.90731
                     precision recall f1-score
                                                      support
Insufficient Weight
                                     0.94
                                               0.94
                                                          374
                           0.93
      Normal Weight
                           0.89
                                     0.90
                                               0.89
                                                           469
                          0.89
                                     0.87
                                               0.88
                                                          441
     Obesity Type I
    Obesity_Type_II
                          0.96
                                     0.98
                                               0.97
                                                          481
   Obesity_Type_III
                          0.99
                                     1.00
                                               0.99
                                                          597
 Overweight Level I
                          0.81
                                     0.76
                                               0.78
                                                          369
```

Overweight_Level_II	0.81	0.84	0.82	376
accuracy macro avg weighted avg	0.90 0.91	0.90 0.91	0.91 0.90 0.91	3107 3107 3107

Training time: 61.47 seconds Best iteration: 683