

# DA6701 – Assignment 2

## Multi-Dimensional Return Forecasting and Portfolio Management – Report

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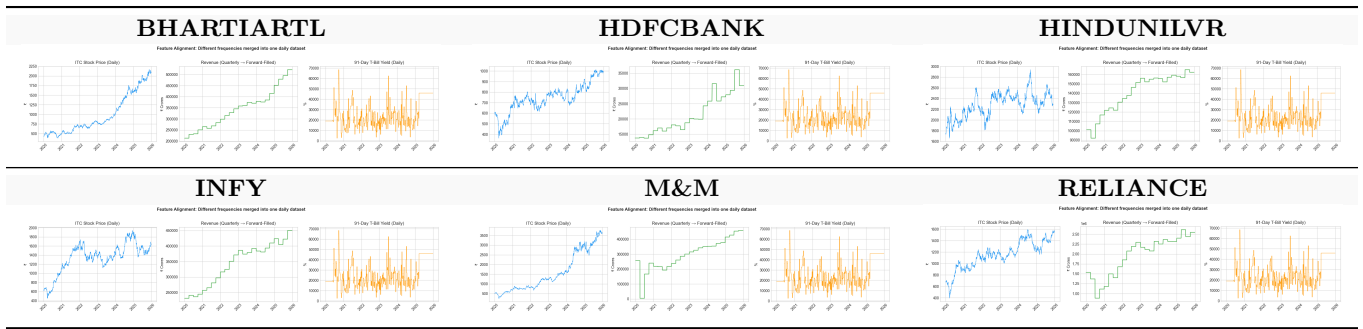
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## Data and Feature Engineering

- **Dataset Composition:** Pipeline utilizes approximately 1297 trading days worth data.
- **Feature Engineering:** Over 27 new features were created.
- **Preprocessing:**
  - **Winsorization:** Range is capped from 1st to 99th percentile to prevent wild outliers from affecting the data.
  - **Feature Selection:** Mutual Information scores are used to select the best features.
  - **Rolling Standardization:** Features are normalised using a rolling window to prevent look-ahead bias.



## Model Architecture and Validation

- We used LGBM and XGBoost.
- A chronological split (70% train, 15% Val, 15% Test) is used.
- Hyperparameter tuning is done with optuna.

## Feature Importance

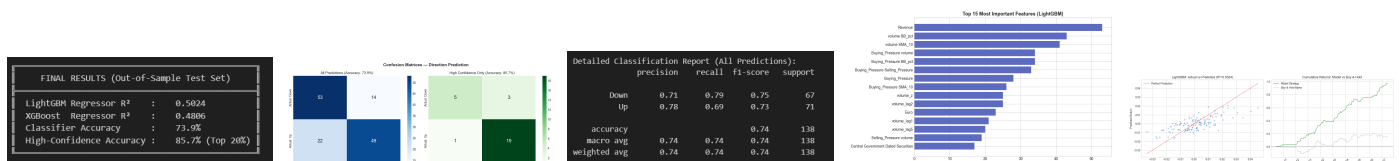
SMA, EMA, RSI, MACD, Bollinger Bands, Buying and Selling Pressure, Lagged Features were added.

## Model Results

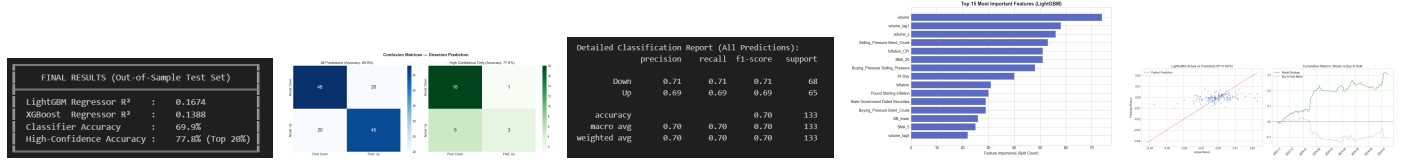
- **Classification:** Will the stock outperform the sector (1) or not (0)?
- A "bad" feature set gives ~50% accuracy while a well engineered feature set gives ~65-74% accuracy.
- When consistently applied across many trades, 65% is highly useful.

## Performance Metrics

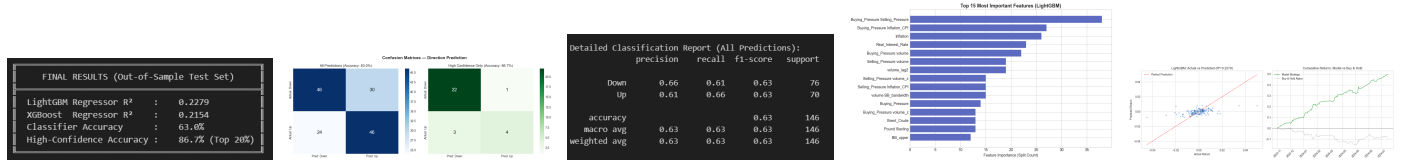
### BHARTIARTL



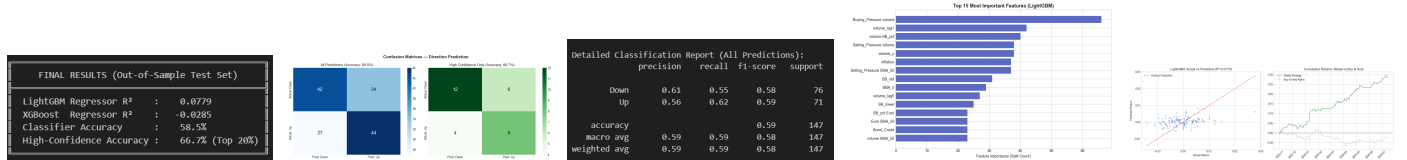
## HDFCBANK



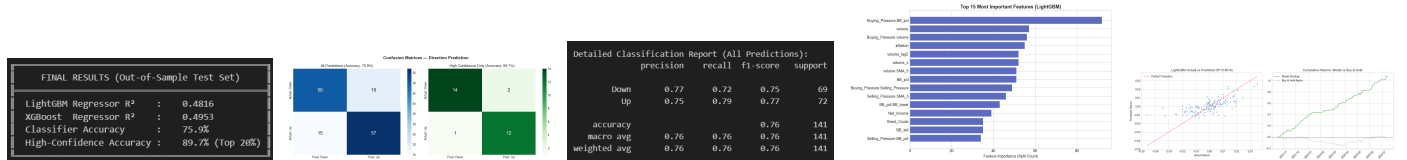
## HINDUNILVR



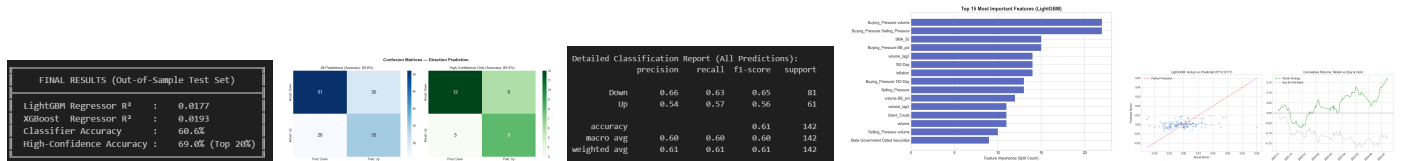
## INFY



## M&M



## RELIANCE



## Implementation Workflow: Production-Grade ML Pipeline

All six notebooks were standardized into a unified 27-cell architecture to ensure reproducibility and prevent look-ahead bias.

- **Target Formulation:**  $Target = \ln(Close_{t+1}/Close_t)$  (Next-day log returns).
- **Feature Space (40 total):** Lags (1–10d), Rolling Volatility/Mean (5–20d), RSI, MACD, Bollinger Bands, Momentum, Volume Ratios, and explicitly lagged (+1d) Macro/Fundamental data.
- **Walk-Forward Validation:** 5-fold expanding window cross-validation (Train: Jan 2020–Sep 2025; Test: Oct–Dec 2025).
- **Model Ensemble:** Competitive training between XGBoost, LightGBM, and a 2-layer LSTM (PyTorch). Best model selected per stock via minimal Walk-Forward RMSE.

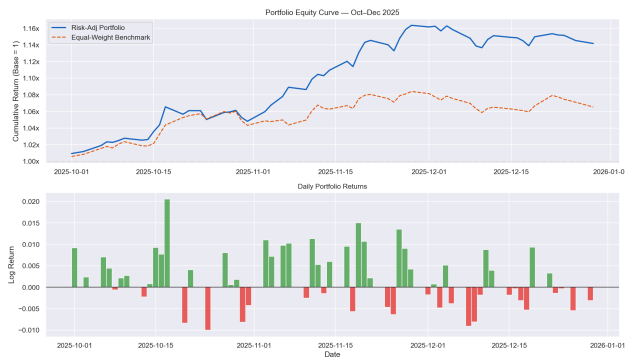
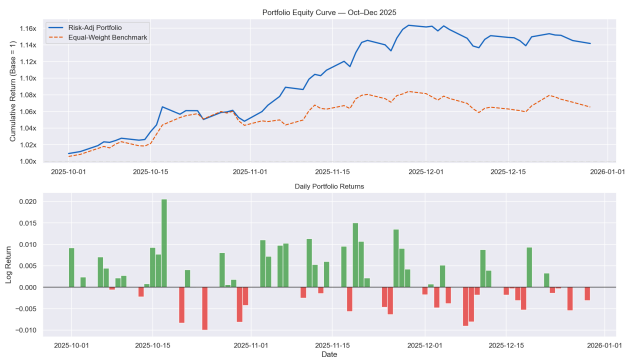
## Per-Stock Validation Results

Stock	Best Model	WF-RMSE	WF-Hit %	Test Hit %	Test Sharpe
RELIANCE	XGB	0.01442	51.9%	50.0%	-0.59
HDFCBANK	LGBM	0.01348	50.7%	40.0%	0.93
INFY	LGBM	0.01533	51.2%	45.0%	-3.43
M&M	XGB	0.01818	52.7%	62.5%	3.80
BHARTIARTL	LGBM	0.01420	52.4%	60.0%	2.49
HINDUNILVR	LGBM	0.01268	50.7%	47.5%	-1.92

## Portfolio Aggregation (Oct–Dec 2025)

Predictions were combined using a Risk-Adjusted weighting scheme:  $Weight_i = \frac{Return_{pred}}{\sigma_{hist}}$ , clipped to long-only and normalized.

Metric	Risk-Adj Portfolio	Equal-Weight
CAGR (ann.)	74.5%	30.4%
Sharpe Ratio	6.96	3.67
Max Drawdown	-2.32%	-2.34%



## Key Quality Controls

- **Leakage Prevention:** Strictly chronological splits; `RobustScaler` fit on training data only.
- **Stationarity:** Log-returns used for all targets to satisfy ML assumptions.
- **Reproducibility:** Global seeds fixed (42); LSTM utilizes Dropout (0.2) and L2 decay (1e-4).