

Report on Data Science Project

DINESH, VINAY, MAHAK, SUNIL(team 11)
Department of Information Technology,
National Institute of Technology Karnataka, Surathkal, India

Novel Contributions

1 Dynamic Graph Construction

- Sliding Window Correlations: Updates graphs over time.
- Multi-Level Hierarchical Learning: Combines intra- and inter-sector dependencies.

2 Integration of LSTM with HGAT

- Replaces GRU with LSTM for better temporal modeling.
- Fuses multi-level embeddings: LSTM + HGAT + Dynamic Graph.

3 Reinforcement Learning for Trading Optimization

- PPO for trading actions.
- Continuous learning from market feedback.

Proposed Methodology

1 Data Collection and Preprocessing

- Weekly aggregation of NIFTY 500 and Yahoo Finance data.
- Cleaning, normalization, wavelet/Fourier transforms.
- Data augmentation: Gaussian noise, bootstrapping.

2 Time-Series Forecasting Using LSTM

- Predicts short and long-term stock trends.
- Hybrid LSTM-CNN and Transformer-LSTM for better features.

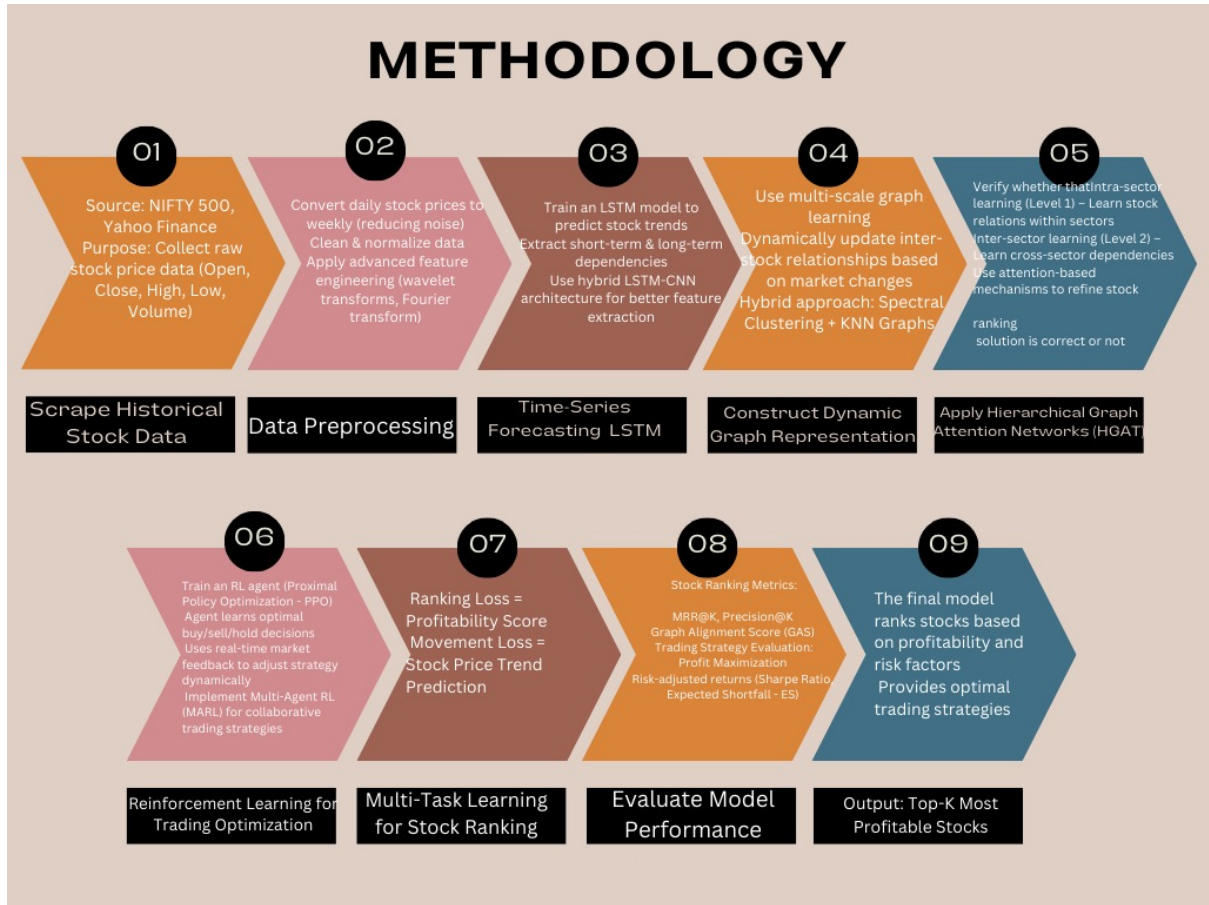


Figure 1: Methodology Diagram

3 Dynamic Graph Construction

- Multi-scale graph learning.
- Spectral clustering, KNN graphs.
- Dynamic edge weights and thresholds.

4 Hierarchical Graph Attention Network (HGAT)

- Learns intra- and inter-sector relationships.
- Self-supervised contrastive learning.
- Graph diffusion and BPR loss.

5 Reinforcement Learning for Trading

- PPO trains buy/hold/sell agents.
- Multi-agent RL adapts strategies.

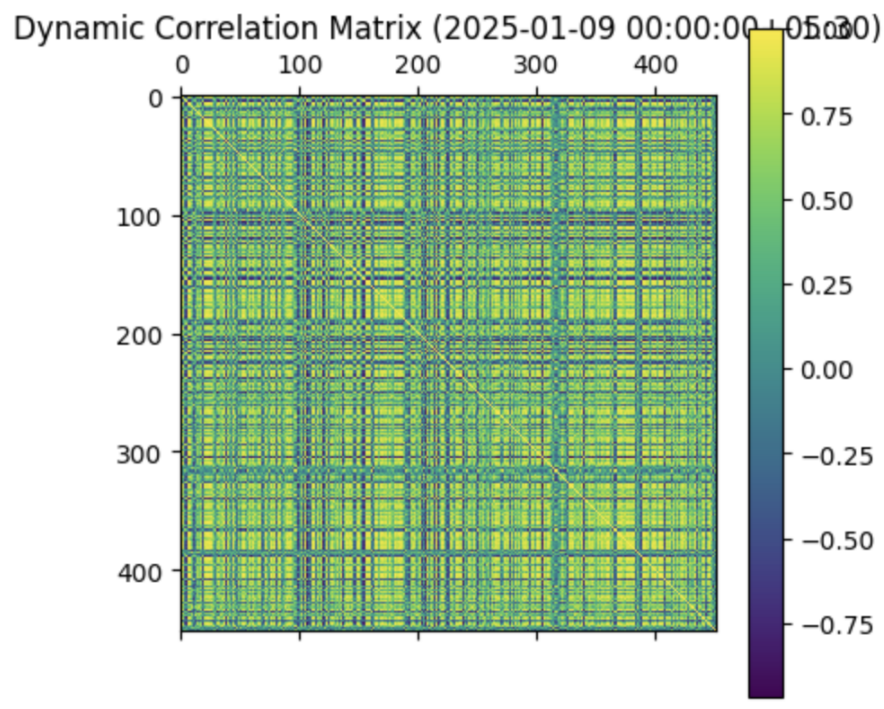


Figure 2: DYNAMIC CORRELATION GRAPH

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Epoch 1/10
8073/8073 ————— 33s 4ms/step - loss: nan - val_loss: nan
Epoch 2/10
8073/8073 ————— 32s 4ms/step - loss: nan - val_loss: nan
Epoch 3/10
8073/8073 ————— 32s 4ms/step - loss: nan - val_loss: nan
Epoch 4/10
8073/8073 ————— 32s 4ms/step - loss: nan - val_loss: nan
Epoch 5/10
8073/8073 ————— 32s 4ms/step - loss: nan - val_loss: nan
Epoch 6/10
8073/8073 ————— 32s 4ms/step - loss: nan - val_loss: nan
Epoch 7/10
8073/8073 ————— 32s 4ms/step - loss: nan - val_loss: nan
Epoch 8/10
8073/8073 ————— 31s 4ms/step - loss: nan - val_loss: nan
Epoch 9/10
8073/8073 ————— 31s 4ms/step - loss: nan - val_loss: nan
Epoch 10/10
8073/8073 ————— 31s 4ms/step - loss: nan - val_loss: nan
8073/8073 ————— 10s 1ms/step
1807/1807 ————— 2s 1ms/step
Train Embeddings Shape: (258332, 16)
Test Embeddings Shape: (57819, 16)
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Figure 3: LSTM

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Episode 33/50 - Epsilon: 0.8518
Episode 34/50 - Epsilon: 0.8475
Episode 35/50 - Epsilon: 0.8433
Episode 36/50 - Epsilon: 0.8391
Episode 37/50 - Epsilon: 0.8349
Episode 38/50 - Epsilon: 0.8307
Episode 39/50 - Epsilon: 0.8266
Episode 40/50 - Epsilon: 0.8224
Episode 41/50 - Epsilon: 0.8183
Episode 42/50 - Epsilon: 0.8142
Episode 43/50 - Epsilon: 0.8102
Episode 44/50 - Epsilon: 0.8061
Episode 45/50 - Epsilon: 0.8021
Episode 46/50 - Epsilon: 0.7981
Episode 47/50 - Epsilon: 0.7941
Episode 48/50 - Epsilon: 0.7901
Episode 49/50 - Epsilon: 0.7862
Episode 50/50 - Epsilon: 0.7822
Top K Stocks to Buy:
      Stock  return_ratio  Action
107714  ADANIENT      0.200394      1
169902  EASEMYTRIP    0.200157      1
113783   MMTC        0.200000      1
47153   ADANIGREEN    0.200000      0
294440  APARINDS      0.199972      1
195171  KIRLOSENG     0.199969      1
272548   FACT        0.199967      1
88615   ITI          0.199823      1
29057   TTML         0.199668      1
146296  GSFC         0.199531      0

```

Figure 4: R-LEARNING RANKINGS

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Precision: 0.0222
Accuracy: 0.6053
Mean Reciprocal Rank (MRR): 0.2728
Mean Absolute Error (MAE): 0.0009

```

Figure 5: K=5

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Precision: 0.0222
Accuracy: 0.6080
Mean Reciprocal Rank (MRR): 0.2774
Mean Absolute Error (MAE): 0.0008

```

Figure 6: K=10