DeepTrade AI: Multi-Model Stock Prediction with NLP & Automated Trading

Project Overview:

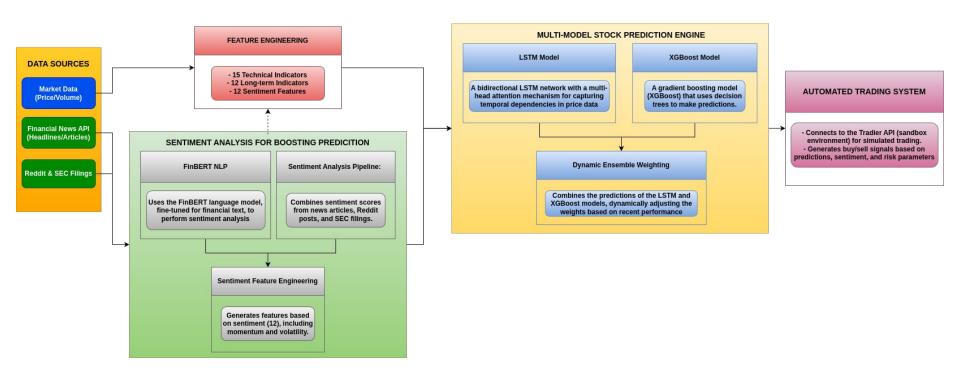
An end-to-end automated stock trading system that combines machine learning price prediction with NLP-based sentiment analysis. The system features a bidirectional LSTM with attention mechanism and XGBoost ensemble for multi-timeframe price forecasting, and integrates FinBERT for real-time sentiment analysis of financial news, Reddit posts, and SEC filings. The architecture employs dynamic model weighting, comprehensive risk management controls, and simulated execution through the Tradier API, achieving 55-65% directional accuracy and a 58.5% win rate in paper trading.

GitHub Repository: DeepTrade-Al

Key Technologies and Skills Used:

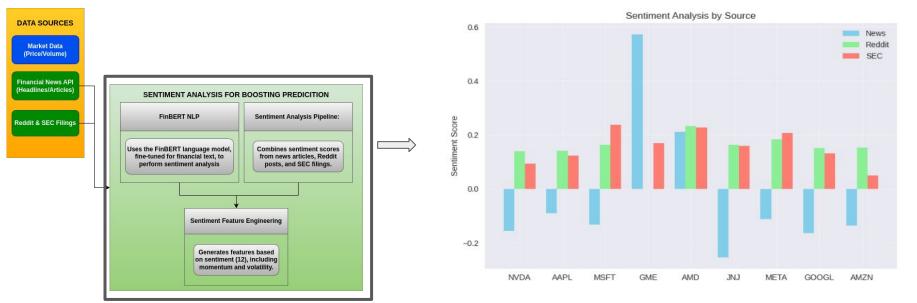
Languages & Frameworks: Python, PyTorch, TensorFlow, CUDA, Scikit-learn, Pandas, NumPy, Hugging Face Transformers Machine Learning: Gradient Boosting (XGBoost), Feature Engineering, Regression, Time Series Forecasting Deep Learning: LSTM Networks (Bidirectional, Attention), Model Ensembling, Model Training & Hyperparameter Optimization Cloud Computing: AWS SageMaker (for distributed model training and hyperparameter optimization)
Natural Language Processing (NLP): FinBERT, Sentiment Analysis, Text Processing, Financial Text Mining

Pipeline:



System Architecture and Data Flow

Multi-Source Sentiment Analysis:



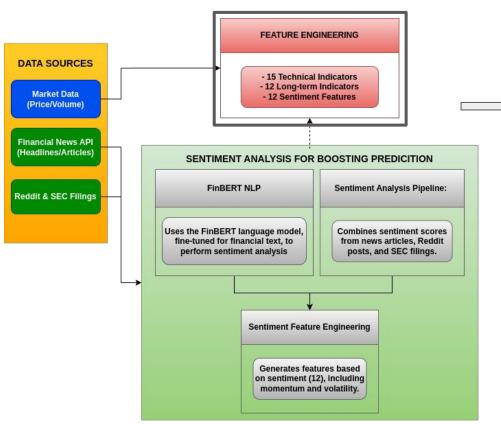
System Architecture and Data Flow

Sentiment scores for selected stocks; Positive values indicate positive sentiment, negative values indicate negative sentiment, and values near zero indicate neutral sentiment

Key Details:

- Tokenization: "Converts text data into numerical representations for the FinBERT model."
- 3-class Classification: "Classifies sentiment as positive, negative, or neutral."
- Multi-Source Weighted Integration: Financial News (40%)/Reddit(30%)/SEC(30%)

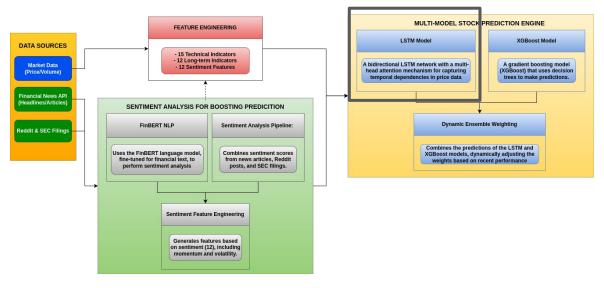
Feature Engineering:



The prediction models utilize a comprehensive set of 39 features, encompassing technical indicators, long-term trend indicators, and sentiment-derived features. This rich feature set provides a holistic view of market dynamics, enabling the models to capture complex relationships and patterns

System Architecture and Data Flow

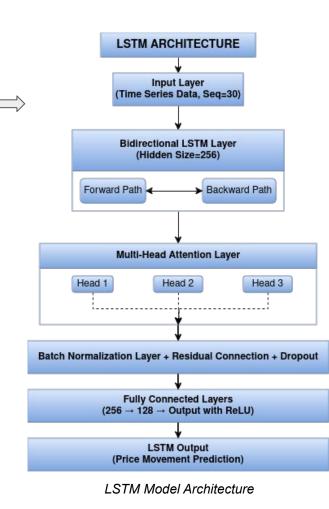
LSTM for Price Prediction



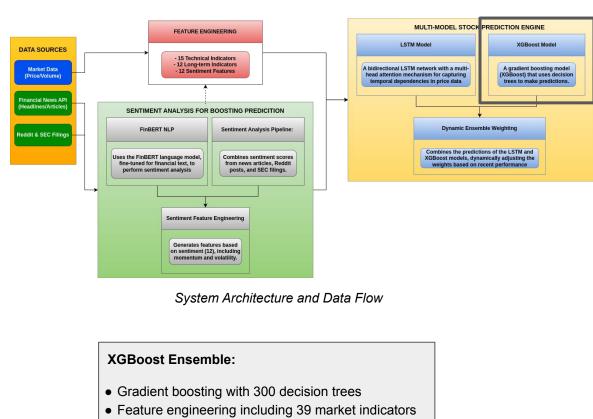
System Architecture and Data Flow

LSTM Neural Network:

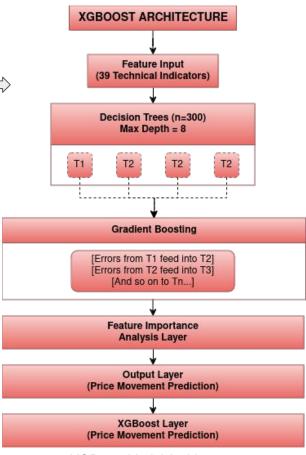
- Bidirectional LSTM: The network processes the input sequence in both forward and backward directions, allowing it to learn from past and future context.
- Multi-Head Attention: This mechanism allows the model to focus on different parts of the input sequence that are most relevant for prediction. The model uses 3 attention heads.
- Batch Normalization: Batch normalization layers are used after each LSTM layer to improve training stability and speed.



XGBoost for Price Prediction

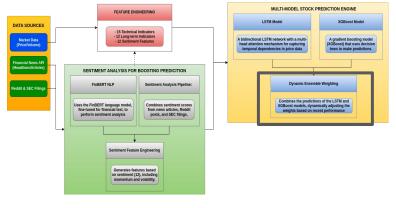


Min child weight:3, Subsample:0.8, Max depth:8



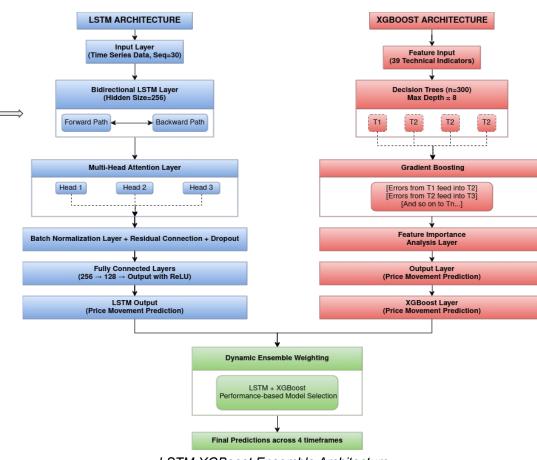
XGBoost Model Architecture

LSTM-XGBoost Ensemble for Price Prediction



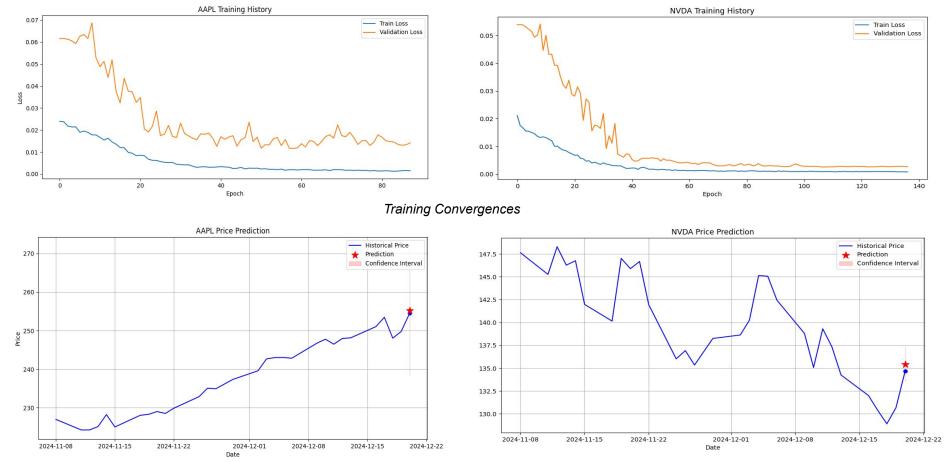
System Architecture and Data Flow

The trained ensemble model achieves a directional accuracy of 55-65% across multiple timeframes (5min, 15min, 30min, 1h) and a mean absolute error of 0.3-0.4% on normalized returns for multiple stocks.



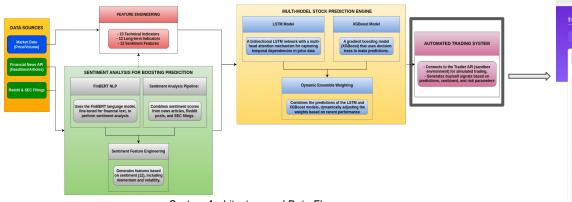
LSTM-XGBoost Ensemble Architecture

LSTM-XGBoost Ensemble Model Prediction Results with Examples



Price Predictions

Automated Trading System: Real-World Implementation



System Architecture and Data Flow

Trading Strategy Integration:

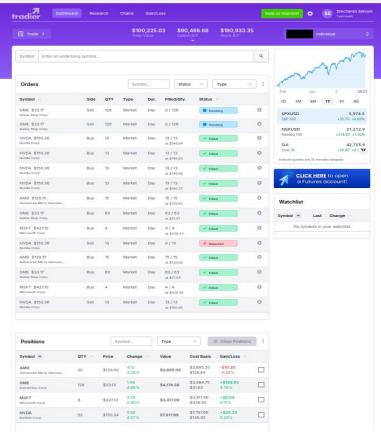
- Entry conditions: Positive directional accuracy in 5/15/30min timeframes.
- Exit conditions: Negative directional forecast

Risk Management Framework:

- Risk per trade: 1% of capital
- Stop-loss: 1.5%, Take-profit: 3%

System Validation:

- Win rate: 58.5% across 9 major stocks
- +0.32% net return



Screenshot of the Tradier paper trading interface, showing successful execution of trades (01/06/2025)