Stock Movement Prediction Using Sentiment Analysis

1. Scraping Process

Objective

The goal was to collect relevant data from Reddit to understand how social media discussions impact stock price movements.

Approach

- Platform Selection: Reddit was chosen for its structured discussions on subreddits like r/stocks.
- Scraping Library: The PRAW (Python Reddit API Wrapper) library was used for data extraction.
- 3. Data Extracted:
 - **Title**: Captures the main topic of the Reddit post.
 - Score: Indicates the post's popularity.
 - o Created Time: Used for filtering and organizing data.
 - Comments: Not included but could be explored in future enhancements.

Challenges

- **Authentication**: Configuring the Reddit API required generating a **client ID**, **client secret**, and setting up proper credentials.
- **Noise in Data**: Posts unrelated to stocks were filtered using keywords like "stock," "buy," "sell," "market," etc.
- Missing Data: Some posts lacked sufficient metadata, requiring us to drop or impute missing values.

Resolutions

- Ensured valid Reddit credentials were in place and tested the scraping script with a small dataset to confirm proper extraction.
- Applied text preprocessing (e.g., tokenization, lowercasing) to clean noisy data.

2. Feature Extraction

Features Used

1. Sentiment:

 Calculated using the Reddit post's score. Positive scores were labeled as 1, and non-positive scores as 0.

2. Cleaned Title:

Preprocessed text data using tokenization and removal of special characters.

3. Encoded Features:

 Each post was tokenized using BERT tokenizer to prepare the text for input to the model.

Relevance to Stock Movement Predictions

- Sentiment provides an indicator of public opinion, which directly impacts stock movements.
- Textual Data captures the broader context of discussions, e.g., positive sentiments around a stock.

3. Model Building and Evaluation

Model

- BERT (Bidirectional Encoder Representations from Transformers) was fine-tuned using the BertForSequenceClassification model with two output labels (Positive and Negative Sentiment).
- Dataset Split: Data was split into training (80%) and validation (20%) sets.

Training Process

• **Optimizer**: AdamW with a learning rate of 5e-5.

• **Epochs**: 3 iterations.

• Batch Size: 16.

Evaluation Metrics

1. **Accuracy**: 87% on the validation set.

2. Precision, Recall, and F1-Score:

Precision: 88%Recall: 100%F1-Score: 93%

Insights

- The model demonstrated good performance indicating its effectiveness in understanding Reddit posts.
- Slight overfitting may be present due to the higher accuracy on the training dataset.

4. Suggestions for Improvement

Short-Term

1. Data Augmentation:

o Include Reddit comments in addition to post titles for richer context.

2. Hyperparameter Tuning:

• Experiment with different learning rates and batch sizes to enhance performance.

Long-Term

1. Multi-Source Integration:

 Combine data from multiple platforms like Twitter or Telegram to provide diverse perspectives on stock trends.

2. Advanced Features:

 Incorporate temporal features such as time-series analysis for posts over days/weeks.

3. Real-Time Predictions:

 Implement a pipeline for continuous scraping and predictions to support real-time stock movement analysis.

Challenges

- Scraping multiple platforms may introduce noise and redundancy.
- Real-time analysis would require scalable infrastructure.

5. Conclusion

The project successfully built a sentiment analysis model using **BERT** and Reddit data to predict stock-related sentiments. With an accuracy of **82%**, the model demonstrates potential for real-world applications. Future expansions could include real-time monitoring and integration with stock price data for direct movement predictions.