



# Forecasting Palantir: A Deep Learning Approach to Stock Prediction





# Why the task matters



## Relevance

- Since stock prediction is a time series of numbers it provides the perfect basis to explode it to models like LSTM's, GRU's, or even transformers
- This projects applies to many course concepts and to a real world high-complexity task
- Being an MBA student and personally invested in the stock market and the Company this seemed to be the perfect fit



## Domain Knowledge

- Strong business and finance foundation
- Hands-on exposure to the stock market and profound knowledge of the company and their business model
- Strong Expertise and proven



## Data Availability

- Daily historical prices (Open, High, Low, Close, Volume)
- Technical indicators (e.g., 20-day MA, RSI, volatility measures)
- Broader market data (e.g., NASDAQ index trends) for noise/context

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# High-Level Model Structure



## Input and Features

- Closing price
- 5/10/20- Day Moving Average
- Volume
- NASDAQ Closing price (represent the market noise of the industry they are in)
- TBD

## Sequence Model

**LSTM 1**



**LSTM2**

## Output

**Next Day Close  
Price and Indicator  
for stock movement  
(up/down)**

# Feasibility and Limitations



## Feasibility

- Time-series is clean and structured
- Dataset is widely available
- Built with PyTorch – simple to extend
- Well suited for sequence models

## Utility

- The overall goal is to provide a tool that provides two insights:
  - Is the stock going to move up or down
  - And the prediction of the price the next day
- Useful for investors or analysts
- Supports experimentation with real forecasting pipelines

## Limitations

- Limited by historical data model cannot respect new macro events
- Risk of overfitting to short-term noise or past regimes
- Only uses numerical data (ignores sentiment, fundamentals, and news)



