1. What question(s) are you trying to answer in your project?

- 1. Which time series forecasting method performs best for NVIDIA & AMD stock price prediction?
 - Comparing LSTM, Transformer-based models (Time-Series Transformer & TFT), and Zero-shot forecasting (Lag-Llama) for stock price prediction.
 - Evaluating different methods using **MSE**, **MAE**, and other key metrics to determine the most effective approach.
- 2. Is Transformer actually suitable for financial time series forecasting?
 - Transformer models have proven powerful in NLP tasks, but can they outperform traditional methods (LSTM, Lag-Llama) in stock price forecasting?
 - Do Transformers struggle with learning price fluctuations? Are they more prone to predicting mean values instead of trends?
- 3. Can Zero-shot forecasting (Lag-Llama) replace traditional time series models?
 - Without training a model, can LLM-based Zero-shot forecasting generate competitive stock price predictions?
 - Does it effectively capture market patterns, or is it merely approximating an average trend?

2. Who are the key stakeholders for your project?

1 Investors & Financial Analysts

- They seek accurate stock price forecasts to optimize trading strategies.
- They care about **model interpretability**—understanding why a model predicts price movements rather than just getting a black-box output.

2 Machine Learning & Al Researchers

- Interested in whether **Transformer-based models can outperform traditional** methods in financial forecasting.
- Focused on model generalization, computational efficiency, and data requirements.

3 Quantitative Trading Firms & FinTech Companies

- They aim to enhance algorithmic trading accuracy while keeping computational costs manageable.
- Need models that balance accuracy, interpretability, and execution speed.

4 Retail Investors

• They might be curious about **Al-driven stock predictions**, but their main concern is whether these forecasts are actually reliable for making investment decisions.

Different stakeholders have **competing priorities**, which create conflicts in the following areas:

📌 1. Model Accuracy vs. Computational Cost

- Researchers & Trading Firms may prefer high-accuracy models (e.g., Transformer & TFT), even if they require expensive computations.
- Retail investors & smaller firms might prioritize faster, lower-cost models (e.g., LSTM) even if they sacrifice some accuracy.

2. Interpretability vs. Predictive Power

- **Investors & financial analysts** need interpretable models (e.g., LSTM, statistical models) to justify trading decisions.
- Transformers & Zero-shot models (Lag-Llama) might be more accurate but act as black-box models, making it difficult to explain their decisions.

*3. Training vs. Zero-shot Forecasting

- Traditional ML models require training on historical data but often yield more reliable results.
- Zero-shot forecasting (Lag-Llama) does not require training, making it computationally efficient—but is it accurate enough to replace traditional forecasting models?