# Fine-Tuning LLMs on Finance-Alpaca Dataset: A Comparative Study of LLMs for Financial Text Generation Tasks

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Capstone Project By Jasmine Coloma Professor Ying Mao Fall 2023

### **Abstract and Introduction**

#### **MOTIVATION:**

LLMs can encounter challenges when understanding complex financial language

#### **RESEARCH QUESTION:**

1. What is the comparative impact of fine tuning LLMs on a financial instruction dataset?

2. How do these models differ in their ability to comprehend and generate finance specific text?

#### **OBJECTIVE:**

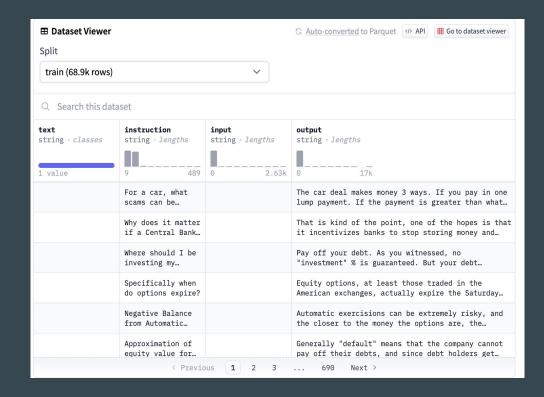
 To conduct a comparative analysis of LLMs on financial text dataset

See how well each LLM can adapt to financial language

### **Dataset**

- Alpaca Finance dataset:
  - 4 features ("text," "instruction,""input," "output").





# Methodology

Model Quantization	Base Model Evaluation	Fine Tune Model	Model Inference and Evaluation
Load models	Perplexity	Using QLoRA	Training loss
using 4-bit	Metric		comparison
precision		Task: text	
	Model Inference	generation	Perplexity comparison
		Trained for 100	
		steps and 1 epoch	Model inference comparison

# **Step 1 - Model Selection: LLM Comparisons**

Selection criteria: autoregressive models

Model	Llama 2	Falcon	Bloom	GPT2
Model size	7B	7.5B	3B	1.5B

## **Step 2 - Model Quantization**

- Motivation for 4 bit-precision
  - Challenge of running/training LLMs on consumer hardware
  - Bitsandbytes library

## **Step 3 - Base Model Evaluation**

- **Perplexity Metric:** evaluates how well a model predicts a sequence of words
  - Lower perplexity = better performance
  - Evaluated perplexity over 100 samples of the dataset

#### Model Inference

Asked the model - "Why should portfolios be diversified?"

## **Step 4 - Fine Tune Strategy**

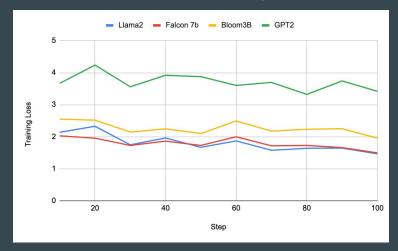
#### 1. SFTtrainer

- a. Training parameters:
  - i. 100 steps
  - ii. 1 epoch,
  - iii. learning rate = 2e-4
  - iv. batch size 4

#### 2. QLoRA

- a. Allows for efficient fine-tuning on limited resources
- b. Helps reduce memory without sacrificing performance

# **Step 5 - Experiment Results: Training Loss Comparison**



Steps	Llama 2	Falcon	Bloom	GPT2
10	2.14	2.0287	2.5505	3.6686
100	1.4675	1.4953	1.9554	3.4193
Total loss	0.6725	0.5334	0.5951	0.2

# **Step 5 - Experiment Results: Perplexity Comparison**

Model Name	Base Model	Fine Tuned Model	Performance Gain
Llama 2	2.58	3.62	+1.04
Falcon	1.73	1.69	-0.04
BLOOM	3.14	2.96	-0.18
GPT2	1.3125	1.016	-0.13

### Step 5 - Experiment Result: Model Inference

Analysis of responses to the query "Why should portfolios be diversified?"

#### Falcon Base Model Response: Falcon Fine Tuned Response: Portfolios should be diversified because it Diversification is a key principle of reduces the risk of the portfolio. investing. It helps to reduce the risk of a Diversification reduces the risk of the portfolio by spreading investments across portfolio by spreading the risk across different asset classes and sectors. This can different assets. Diversification also reduces help to reduce volatility and increase returns the risk of the portfolio by spreading the over time. Additionally, diversification can risk across different industries. help to smooth out fluctuations in the market, making it easier to manage risk. Diversification can also help to reduce the impact of individual investment losses. By spreading investments across different asset classes and sectors, a portfolio can be more ...

**Overall fine tuned responses:** more refined, coherent, and detailed

### **Result Conclusions**

• Llama 2 had **highest** training loss but worse perplexity

• Bloom exhibited **most improvement** in perplexity

• GPT2 & Falcon showed consistent improvement

All models exhibited more detailed yet clear explanations in model inference.

### **Limitations and Future Work**

• Evaluation limited to a select set of models; other models might be better

- Training for only 1 epoch
  - a. Potential data inconsistencies, biases, etc.

Exclusively concentrates on financial text generation

### **Conclusions**

• Research contributes to understanding LLM adaptability to finance domain

Can serve as a base analysis for further research on this topic