

Financial Analysis Using Large Language Models

Jia Finance Practicum Project

Who is JIA Finance?

Corporate Advisor: Ishan Prasad

- Jia Finance is a fintech platform that provides mortgages to foreigners purchasing residential real estate in the US.
- The firm utilizes an asset-based approach to underwriting and uses the latest data analysis technology as well as Al-based identity verification to quickly make underwriting decisions to fund mortgages for international buyers.

Project Scope

- Using LLM, we can scrutinize financial documents like 10-K forms for tracking financial performance of public companies.
- Though the process of analyzing 10-K requires considerable man hours to sift through the documents. Utilizing natural language processing, investors can swiftly discern the report's tone, extract useful metrics, monitor red flags and keywords, and create benchmark using peer performance.

			Year Ended	December 31,				2021 vs. 2020 Cha	inge	2020 vs. 2019 Chang	20
(Dollars in millions)		2021		2020		2019		S	%	S	%
Cost of revenues											
Automotive sales	S	32,415	S	19,696	S	15,939	S	12,719	65 %	\$ 3,757	24 9
Automotive leasing		978		563		459		415	74 %	104	23 9
Total automotive cost of revenues		33,393		20,259		16,398		13,134	65 %	3,861	24 9
Services and other		3,906		2,671		2,770		1,235	46 %	(99)	-49
Total automotive & services and other segment cost of revenues		37,299		22,930		19,168		14,369	63 %	3,762	20 9
Energy generation and storage segment		2,918		1,976		1,341		942	48 %	635	47
Total cost of revenues	S	40,217	S	24,906	S	20,509	S	15,311	61 %	\$ 4,397	21 9
Gross profit total automotive	s	13,839	s	6,977	s	4,423					
Gross margin total automotive		29.3 %		25.6%		21.2%					
Gross profit total automotive & services and other segment	s	13,735	s	6,612	s	3,879					
Gross margin total automotive & services and other segment		26.9 %		22.4%		16.8 %					
Gross profit energy generation and storage segment	s	(129)	\$	18	\$	190					
Gross margin energy generation and storage segment		-4.6%		0.9 %		12.4 %					
Total gross profit	\$	13,606	\$	6,630	\$	4,069					
Total gross margin		25.3 %		21.0%		16.6%					

PART I

Item 1. Business.

Item 1A. Risk Factors.

Item 1B. Unresolved Staff Comments.

Item 2. Properties.

Item 3. Legal Proceedings.

Item 4. Mine Safety Disclosures.

PART II

Item 5. Market for Registrant's Common Equity, Related

Stockholder Matters and Issuer Purchases of Equity Securities.

Item 6. [Reserved]

Item 7. Management's Discussion and Analysis of Financial Condition and Results of Operations.

Item 7A. Quantitative and Qualitative Disclosures About Market Risk.

Item 8. Financial Statements and Supplementary Data.

Item 9. Changes in and Disagreements With Accountants on

Accounting and Financial Disclosure.

Item 9A. Controls and Procedures.

Item 9B. Other Information.

Item 9C. Disclosure Regarding Foreign Jurisdictions that Prevent Inspections.

PART III

Item 10. Directors, Executive Officers and Corporate Governance.

Item 11. Executive Compensation.

Item 12. Security Ownership of Certain Beneficial Owners and Management and Related Stockholder Matters.

Item 13. Certain Relationships and Related Transactions, and Director Independence.

Item 14. Principal Accountant Fees and Services.

PART IV

Item 15. Exhibit and Financial Statement Schedules.

Item 16. Form 10-K Summary.

Signature



DEMO

Interactive Chatbot



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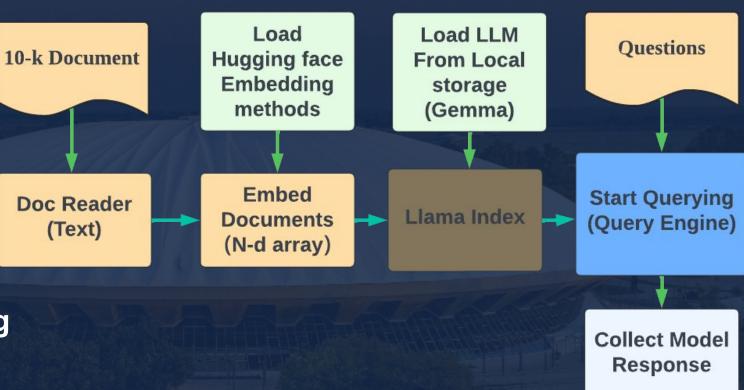




Development Process

Development Process:

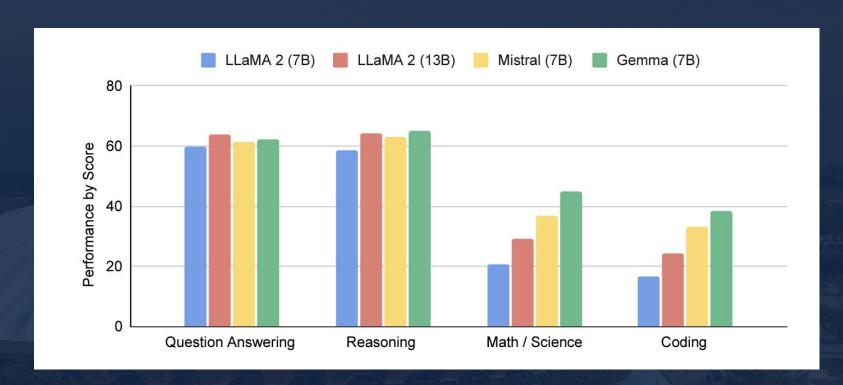
- Preprocessing the Data
- Embedding the Data
- Building the Retriever
- Evaluating and Fine-tuning



LLM Selection



- Overall choice Gemma-7B
 - based on its performance, efficiency, accessibility, and trainability.
- Model deployment:
 - Security and stability
 - Storage & Memory
 - model size 30GB
 - Computation
 - running model
 - fine-tuning model
- NCSA's Delta cluster.
 - high performance computing cluster offered by U of I
 - 256 GB RAM, 4*A100 GPUs



MMLU benchmark



Revenues												
			Year E	nded December 31,				2021 vs. 2020 Cha	inge		2020 vs. 2019 Chan	ige
(Dollars in millions)	7	2021		2020		2019		S	%		\$	%
Automotive sales	S	44,125	\$	24,604	\$	19,358	\$	19,521	79 %	\$	5,246	27%
Automotive regulatory credits		1,465		1,580		594		(115)	-7 %		986	166%
Automotive leasing		1,642		1,052		869		590	56%		183	21 %
Total automotive revenues		47,232		27,236		20,821		19,996	73 %		6,415	31%
Services and other		3,802		2,306		2,226		1,496	65 %		80	4 %
Total automotive & services and other	All Control	19010 1901000		0.0000000000000000000000000000000000000	- 63	200 00000	100	W seek water		- (4	To the second of	
segment revenue		51,034		29,542		23,047		21,492	73 %		6,495	28 %
Energy generation and storage segment revenue		2,789		1,994		1,531		795	40 %		463	30 %
Total revenues	\$	53,823	\$	31,536	\$	24,578	\$	22,287	71 %	\$	6,958	28 %



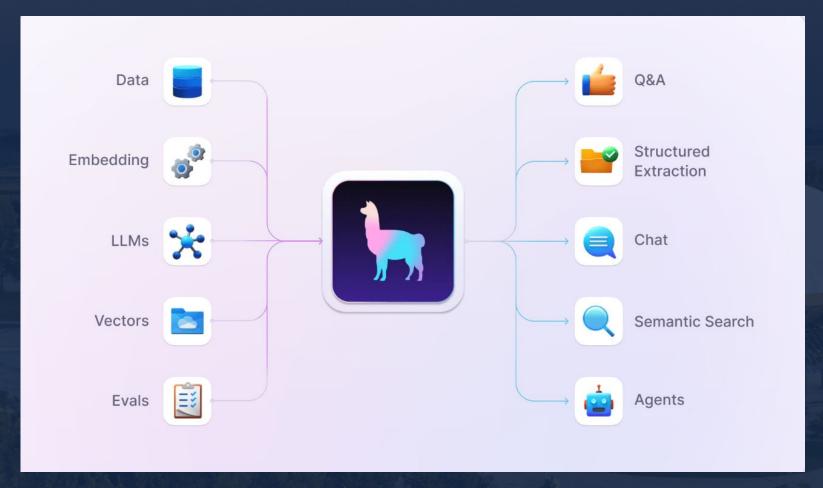


- **Cleaning**: Removing unwanted content
- Extracting: Extraction of specific entities within a document
- **Chunking**: Partition documents into semantic units

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ear Ended December 31,		
1		
2021	2020 2019	
2	Revenues	
3	Automotive sales	
\$ 44,125	5 24,684	\$ 19,358
4	Automotive regulatory credits	
1,465	1,580	594
5	Automotive leasing	
1,642	1,852	869
6	Total automotive revenues	
47,232	27,236	20,821
7	Energy generation and storage	
2,789	1,994	1,531
8	Services and other	
3,802	2,386	2,226
9	Total revenues	
53,823	31,536	24,578
10	Cost of revenues	
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978	563	459
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15	Services and other	
3,906	2,671	2,770
16	Total cost of revenues	
40,217	24,986	20,509
17	Gross profit	
13,606	6,638	4,069
18	Operating expenses	
19	Research and development	
2,593	1,491	1,343
20	Selling, general and administrative	
4,517	3,145	2,646

Llama Index framework

A framework that combines LLM and data for Q&A and Chat!



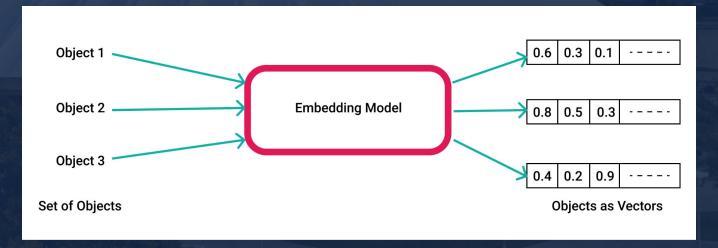
How does the LLM know which parts of the document are the most relevant to the query?

What is data embedding?

- Represent textual data as vector representations in high dimensional space.
- BAAI/bge-large-en-v1.5
- Example:
 - Sentence 1: LLMs are AI models that understand and generate human language.
 - Sentence 2: Large Language Models can comprehend and produce natural

language.

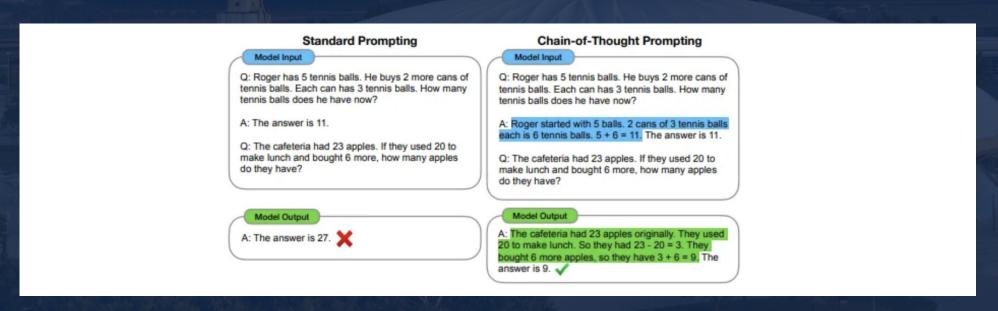
- Cosine Similarity: 0.59
- Similarity ranges from -1
 (Least Similar) to
 +1 (Most similar)



Can quality prompts produce better results? Master of Science in Financial Engineering, University of Illinois at Urbana-Champaign

Chain of Thought Prompting

Chain of thought prompting is a technique used with Al language models to help them solve complex problems. It involves guiding the Al to articulate its reasoning step-by-step, much like how a human would think aloud while solving a problem. By explicitly asking the Al to describe each step in its thought process, this method can improve the Al's accuracy and provide insights into how it arrives at its conclusions. Essentially, it makes the Al "think" in a structured and transparent way, making it easier to follow and verify the logic behind its answers.

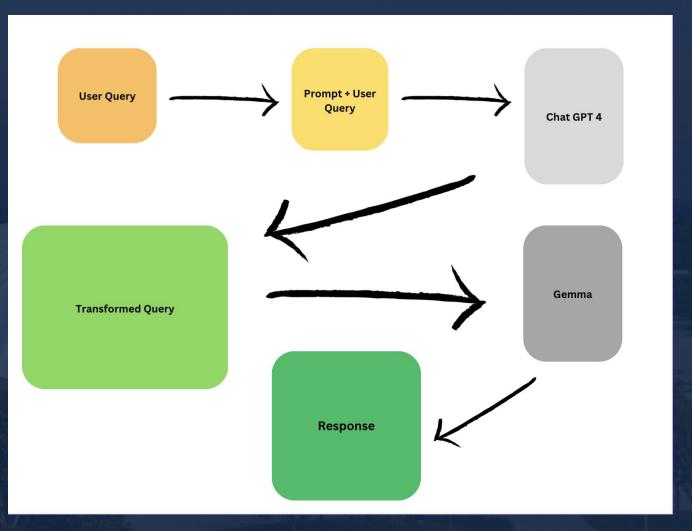


[CONTEXT] + [SPECIFIC INFORMATION] + [INTENT] + [RESPONSE FORMAT] = PERFECT PROMPT

"I am fine-tuning a Large Language Model to analyze 10-K forms for financial analysis. I want to utilize chain of thought prompting so that my model can produce more accurate answers. Can you refine the following question in a way that would produce a more accurate response from my model? It may be useful to include where the information can be found or how to calculate the answer if calculations are required. The refined question should have the following format: [context] + [specific information] + [intent] + [response format]. The only output should be the refined question. Do not include any additional information, explanation, titles, or headers in the output. The output should be in a normal formatting with no bullet points."

Chain of Thought Prompting





How can we fine-tune the model and how do we estimate the results?

Model Fine-Tuning

Problem: Llama Index does not support Model Fine-tuning!

Download Dataset from Huggingface

Download Gemma-7B-it basic checkpoint from huggingface Train the model using transformer

Upload the model to Huggingface Public Model Hub Download the model using Llama Index for further usage

Dataset: 10-k Documents from 1990 to 2020, splitted into chunks

Model Fine-Tuning

Trainable parameters: 200 Million	Total Parameters: 8.7 billion	Percentage: 2.29%
Model size: 31.81 GiB	Hardware: 24GB+ RAM	8K Context length
Transformer structure: 28 layers	16 attention heads	3072 embedding dimension

The Village of the Party of the	Dataset: JanosAudran/financial-reports-sec	

Dataset size: 15 GiB Pure text Number of rows: 68,382,177 Chunk size in one row: 256

	Training parameters	
Batch size: 20	Adjusted Learning rate	Optimizer: Adafactor
maximum steps: 20	Start loss: 10.87	End loss: 0.27



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Multi-agent cross-checking System

Maintain the original answer

Initialize another agent (ChatGPT4)

Feed the documents to the model

Feed the questions and response to the model

Ask model to check the answer accuracy

Yes
Correct?

Replace the original answer by the new one



Define the role

"You are a financial analyst."

"Utilizing your expertise, check the accuracy of the question and answer regarding this company."

Task

"Give a score to this answer from 0 to 5, and state the reason."

Task Breakdown

"If the answer is incorrect, you need to give the correct answer."

Task Breakdown

Question	Original response	Level	Agent Response
How much of the business is purchase vs. refinance activity?	Purchase activity accounted for \$29.5 billion of the business, while refinance activity amounted to \$0 million in the year 2023.	moderate	0/ The answer provided is incorrect. The correct answer is that the information provided does not specify the exact breakdown between purchase and refinance activity for the company.
How does the firm intend to grow its market share, and what was the firm's market share this period? Master of Science in Finance	The firm intends to grow its market share by developing and retaining experienced teams, maintaining a significant cash balance, actively managing inventory investments, offering new home communities, delivering high-quality homes and customer experiences, investing in desirable markets, stal Engineering, University of Illinois at Urbana-Champaign	moderate	4/ The answer accurately describes how the firm intends to grow its market share by outlining various strategies and initiatives. However, the market share value provided in the answer is incorrect. The correct market share value for this period is approximately \$32.9 billion based on the closing price of the common stock held by non-affiliates.

Areas of improvement

- Question classification:
 Different prompt for qualitative / quantitative questions?
- 2. Model response time (Current time: 30 to 40 seconds)
- 3. Minimize model output uncertainty Keep the answer stable!

Summary

- Data Preprocessing
 - Unstructured.io
- Data Embedding
 - BAAI embedding
- LLM
 - Gemma (Pre-trained LLM by Google)
 - Fine Tuning the model using HuggingFace dataset
- Chain of thought prompting
 - Use of another model to generate prompts
- Multi-agent cross checking system for performance evaluation

Appendices Master of Science in Financial Engineering, University of Illinois at Urbana-Champaign

What are 10K forms?

- The data we are working with is a collection of the past 5 years of 10K forms for each of these 18 companies, totaling 90 documents
- 10-K forms are comprehensive annual reports filed by publicly traded companies in the United States, detailing their financial performance, operations, corporate governance, and risk factors as required by the Securities and Exchange Commission (SEC).
- Data types within 10-K forms include quantitative financial information such as balance sheets, income statements, cash flow statements, as well as qualitative disclosures about business operations, risk factors, and management's discussion and analysis of financial conditions.

Why is Preprocessing important?

- The data needs to be transformed into a format that can be more easily digested and evaluated by our LLM.
- This will improve the accuracy of the model and retrieval of information.
- Preprocessing often involves tokenization, which is splitting the document into more manageable pieces or chunks.
- These tokens are then combined with various metadata, such as the name and year of the corresponding document or summaries, to create identifiable nodes of data that can be more easily retrieved.
- 10K forms contain various tabular and graphical information, and it is important to transform this information into something the LLM can easily process.

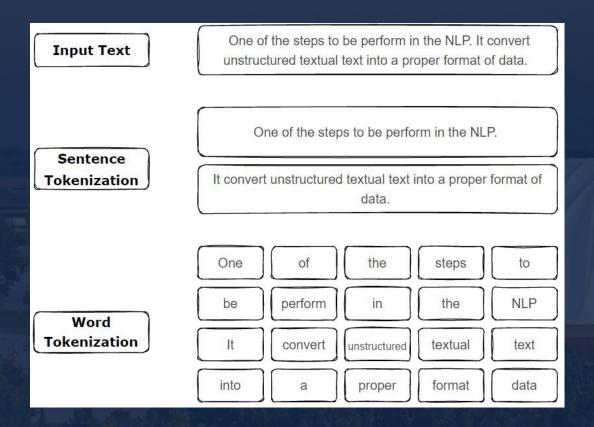


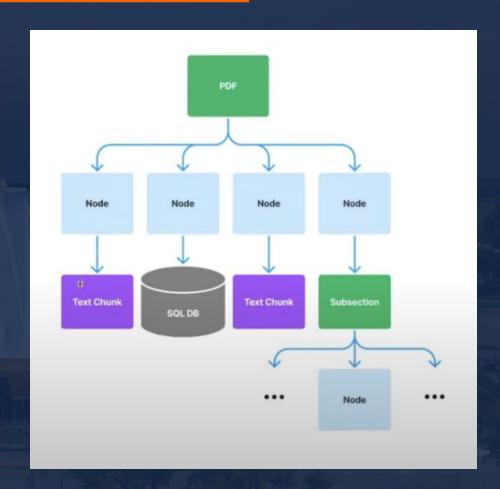
What data are we collecting?

Public Companies:						
Home Builder Companies	Mortgage Companies	Finance Companies*				
DR Horton	Penny Mac	Rithm Capital				
LGI Homes	Rocket Mortgage	SoFi				
Lennar Corporation	United Wholesale Mortgage	Pagaya				
PulteGroup	Guild Mortgage	LendingClub				
Toll Brothers	Better Home	Fair Isaac				
	Mr Cooper Group					
	Angel Oak					
	loanDepot					

- Listed above are 18 companies whose financial statements contain important information for JIA Finance to make underwriting decisions
- These companies consist of home builders companies, mortgages companies, and finance companies

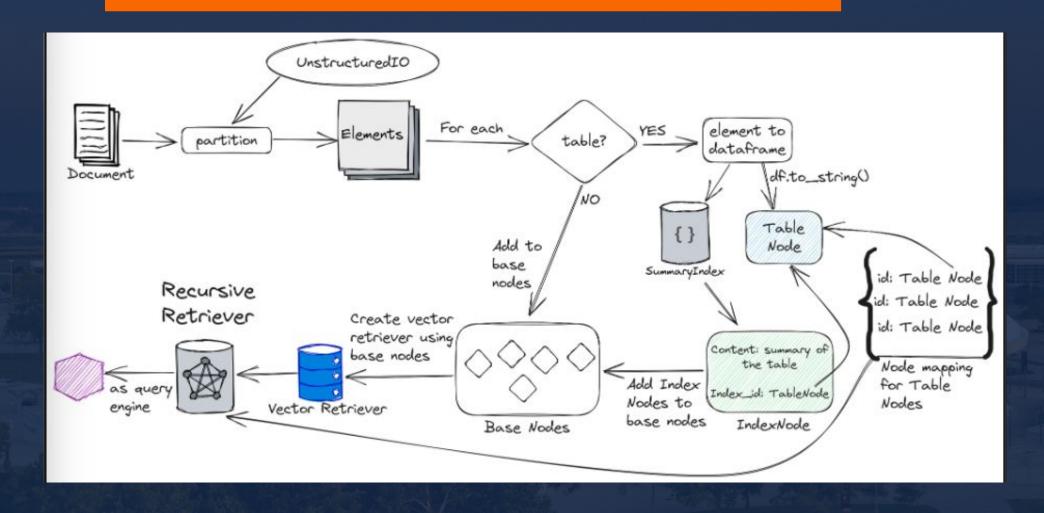
Tokenization





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Recursive RAG Retriever





Sentence 1: LLMs are AI models that understand and generate human language.

Sentence 2: Large Language Models can comprehend and produce natural language.

Embeddings: (Vectors of length 768)

Sentence 1: [0.00822472, -0.03888644, -0.03121234, ..., -0.00927748, -0.0715716,

-0.00871227]

Sentence 2: [0.02884705, 0.05180613, -0.03167022, ..., 0.01885675, -0.101922

,-0.04646319]

Cosine Similarity: 0.5874933 (Would be very less(negative) for sentences with different meaning)