Introduction to Embeddings and Vectors

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Get the Deck



https://github.com/richieyuyongpoh/presentation/blob/main/Introduction%20to%20Embeddings%20and%20Vectors.pdf

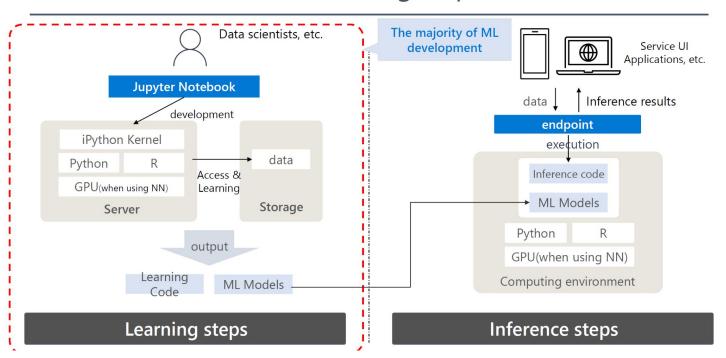
Main References

- Jimmy Liao, How will Development Change with LLMs
- Samuel Leonardo Gracio, Reinvent your recommender system using Vector Database and Opinion Mining
- LeewayHertz, What Role do Embeddings Play in ChatGPT-Like Model
- Bhaskar Mitra, Using Text Embeddings for Information Retrieval
- Prompt Engineering Guide, https://www.promptingguide.ai/
- Google Fundational Courses, Embeddings
- CloudFlare, What are Embeddings
- OpenAI, https://openai.com/

What will be covered?

- The Big Picture with LLMs
- Embeddings and Vectors
- Applications and Use Cases

Be more creative in the learning steps



Prompt Engineering: A new paradigm for ML development

Prompt Processing*

If the information in the prompt itself is insufficient or difficult for Al to interpret

Processing such as changing the way prompting is given

Few-shot Learning

Provide some example answers to the question at the prompt, A method of learning the answer format and behavior by prompting. Accuracy improvements may also be seen in several levels of examples.

Chain of Thought (CoT)

In large-scale language models, by giving a step-by-step thinking process, The ability to solve even difficult problems.

ReAct and Self Consistency also carry on the CoT concept.

Recursively Criticizes and Improves (RCI)

let GPT itself examine and correct the output of GPT. The output is brushed up by repeated execution.

In particular, it is often used to operate programming code.

ReAct

Not only linguistic generation from internal information, Dynamically recognize the required task from the prompt, The idea is to obtain information (grounding) using external APIs such as search and calculation, add that information, and return an answer.

How to get a good prompt

Ex.	idea	Summary		
1	Additional Questions	"If you don't have enough information, just ask," etc. Interactive from the user by looking at the status of the prompt Get additional information.		
2	Text completion	Also done by Bing and others, make text completion and suggestions so that subjects and objects are not missing.		
3	Non-English Utilization	Translate the context information behind the scenes into English, Convert input as English in the backend (in System Prompt). Conversion to a programming language called PAL (Program-Aided Language Model) is also effective.		
4	Prompts using Templates			
5	Before letting GPT answer, first make the sentence ea GPT Calibration for GPT to interpret. Incorporate steps to make GPT correct.			
6	Non-text chat	Input by voice recognition Traditional UI (drop-down list, radio button, checkbox)		

Positioning between Fine tuning and Prompt Engineering

Fine tuning Step Prompt Step inference Long term memory What to get (can handle the details by instruction, but has (remember general information, but vague) difficult for unknow area) Limitations Cost of resource and data processing Token limit Confidential information while training **Prompt Injection** Security / Quality **Annotation Quality** Be prepared for content filtering Improve accuracy of answers Acquiring new task or Improve the task Task recognition When to use Add terminology, domain knowledge Answer format rules The reference info in Prompt too large Small amount of information

Chain-of-Thought (CoT) Prompting

Standard Prompting

Model Input

Q: Roger has 5 tennis balls. He buys 2 more cans of tennis balls. Each can has 3 tennis balls. How many tennis balls does he have now?

A: The answer is 11.

Q: The cafeteria had 23 apples. If they used 20 to make lunch and bought 6 more, how many apples do they have?

Model Output

A: The answer is 27.



Chain-of-Thought Prompting

Model Input

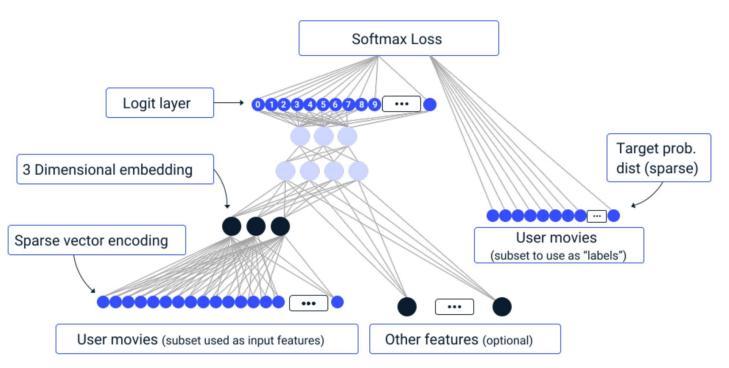
Q: Roger has 5 tennis balls. He buys 2 more cans of tennis balls. Each can has 3 tennis balls. How many tennis balls does he have now?

A: Roger started with 5 balls. 2 cans of 3 tennis balls each is 6 tennis balls, 5 + 6 = 11. The answer is 11.

Q: The cafeteria had 23 apples. If they used 20 to make lunch and bought 6 more, how many apples do they have?

Model Output

A: The cafeteria had 23 apples originally. They used 20 to make lunch. So they had 23 - 20 = 3. They bought 6 more apples, so they have 3 + 6 = 9. The answer is 9.



LeewayHertz

- A way of representing complex information, like texts and images, using a set of numbers
- It translates high-dimensional vectors into a lower-dimensional space
- Essentially, embeddings enable machine learning models to find similar objects.
 Given a photo or a document, a machine learning model that uses embeddings could find a similar photo or document.

Organizing Movies by Similarity (1d)







Incredibles



The Triplets of Belleville



Harry Potter



Star Wars



Bleu

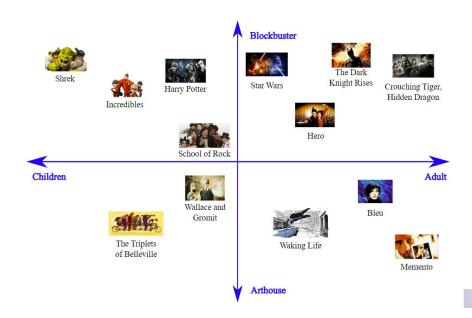


The Dark Knight Rises



Memento

Two-Dimensional Embedding



d-Dimensional Embeddings

- Assumes user interest in movies can be roughly explained by d aspects
- Each movie becomes a d-dimensional point where the value in dimension d represents how much the movie fits that aspect
- Embeddings can be learned from data

What is a vector?

- An array of numbers that define a point in a dimensional space
- In short, a list of numbers
- Each number indicates where the object is along a specified dimension
- Vectors are used to search for similar objects
- A vector-searching algorithm simply has to find two vectors that are close together in a vector database.

vector similarity

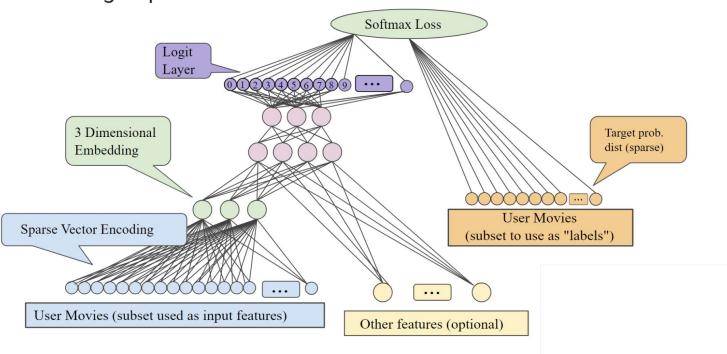
TV show	Genre	Year debuted	Episode length	Seasons (through 2023)	Episodes (through 2023)
Seinfeld	Sitcom	1989	22-24	9	180
Wednesday	Horror	2022	46-57	1	8

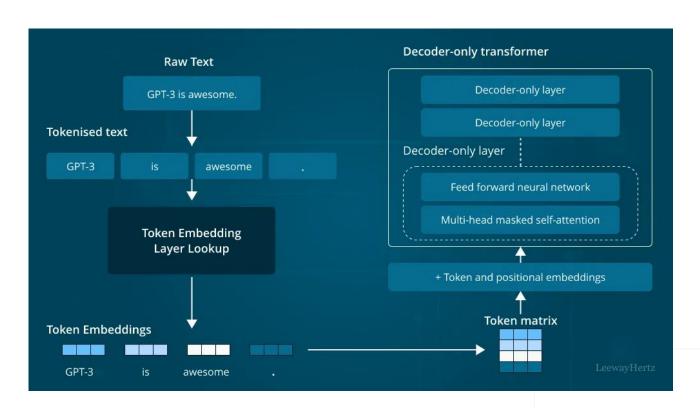
Cheers vector: {[Sitcom], 1982, 21-25, 11, 275}

Is Cheers similar to Seinfeld or Wednesday?



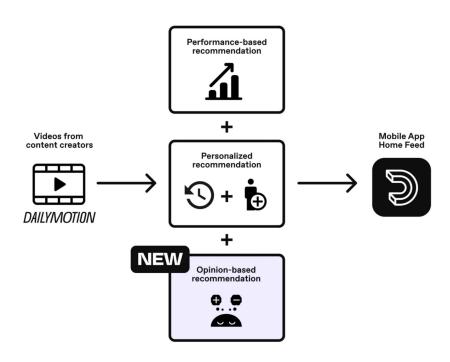
Collaborative Filtering to predict movies to recommend:



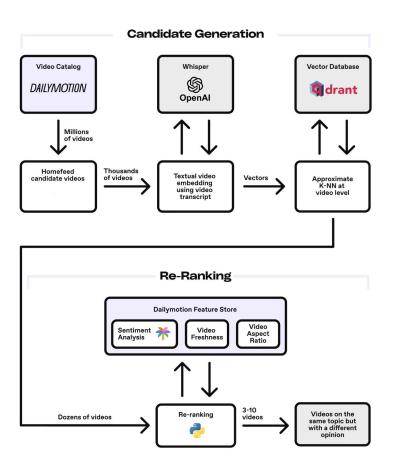


- FAQ Chatbot & NLP
- Customer Experience and Support
- Product Recommendation
- Financial Services
- Image and Video Recognition
- Medical Diagnosis
- Biometrics
- etc

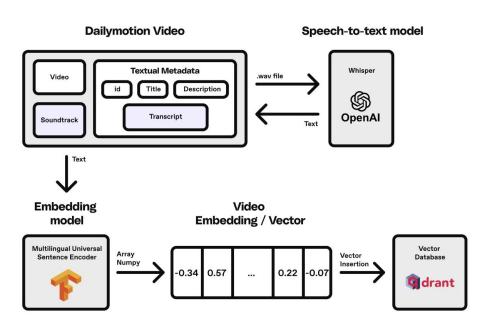
Dailymotion



Dailymotion: The global architecture of the new opinion-based recommender system.

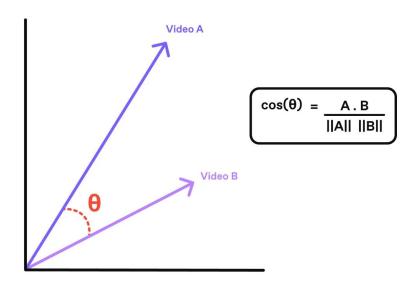


Dailymotion:
Representation of
transforming a video
into a vector



Dailymotion:Cosine similarity between two embeddings

cosine similarity = $cos(\theta)$



4: Simple Demo



https://www.richieyyptutorialpage.com/demo-pytho n-series/simple-embedding-demo

4: FAQ Chatbot (Based on MySejahtera FAQ)



https://www.richieyyptutorialpage.com/demo-pytho n-series/fag-chatbot





