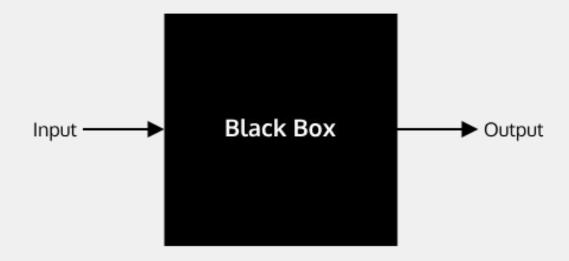
# Retrieval Augmented Generative AI

Solving the black box problem

October 3 2023 6:30 PM | SLC AI Tinkerers Meetup

#### In LLMs we do not trust....

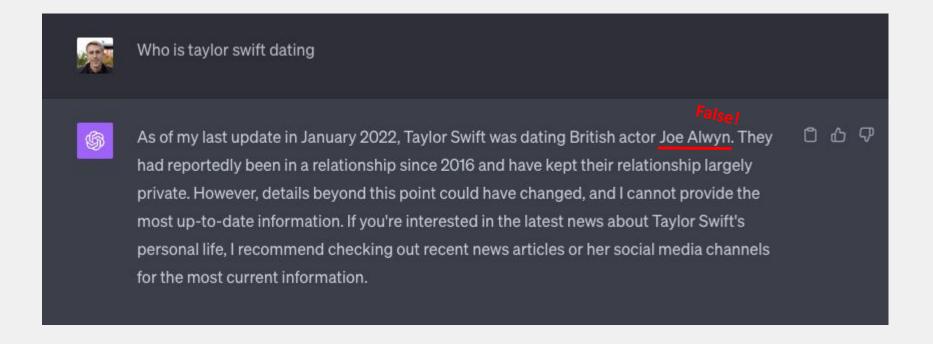


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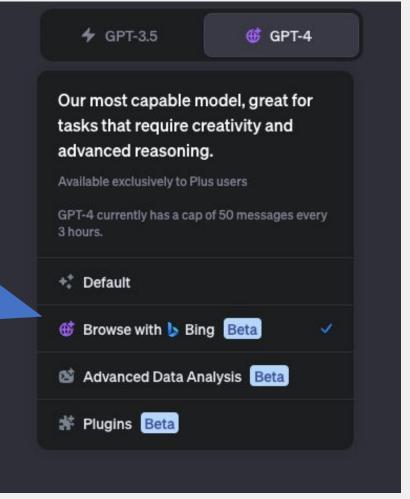




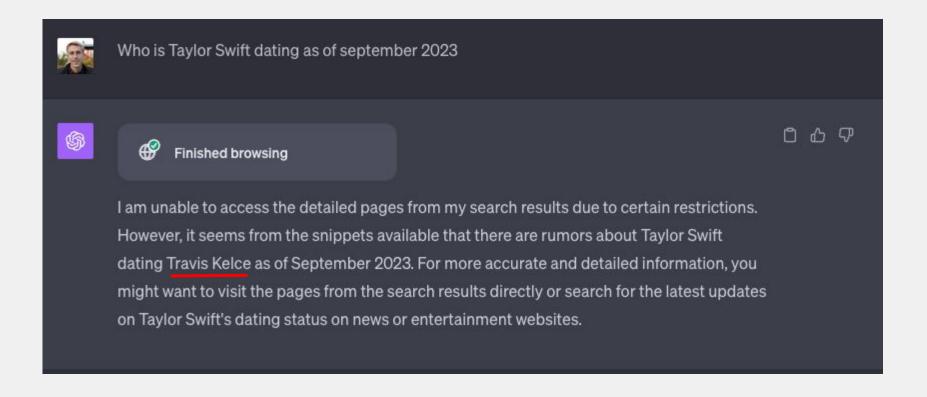
## Example of black box in action



#### But wait, what's this!?



## Correct answer! This is retrieval augmentation in action



# What is Retrieval Augmented Generation?

- Retrieve data relevant to query (Search for news on Taylor Swift)
- Inject into prompt (Who is Taylor Swift dating? \*insert search results\*)
- Response is more accurate, and can cite source (Travis Kelce, see this article)

#### **Advantages**

- Users can take a foundational model and augment it
- This means no fine-tuning required
- Cheaper (fine tuning is expensive!)
- **Faster** (fine tuning takes a long time!)
- **Easier** (fine tuning has a learning curve!)
- Scales well (continuously add new data)
- Biggest advantage: sources can be cited!
   No more black box problem



# That's cool! How do I build a RAG AI with my own data?

Enter the vector database!

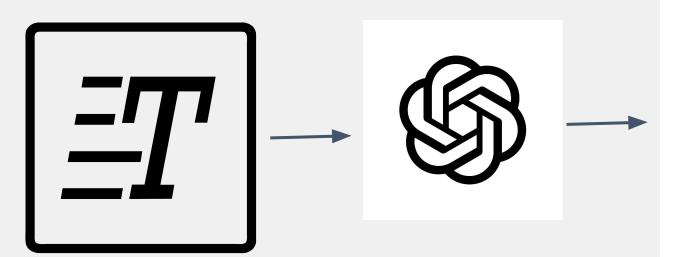


Vector
(Despicable Me)



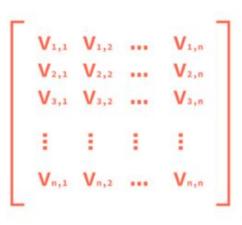
#### What is a vector database?

# It stores embeddings (which are vectors)



Text input (or other mediums)

Pass to neural network embedding model (easiest way is hit the OpenAl api)



#### Embedding vectors

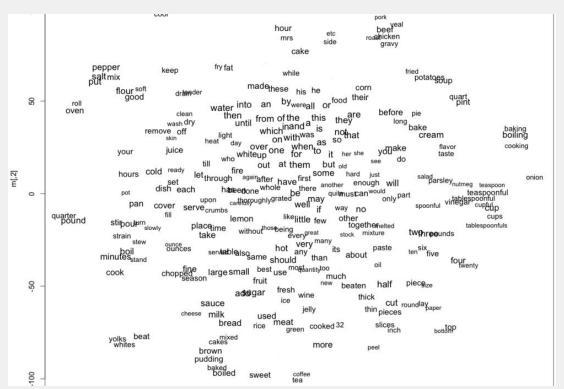
Get back an embedding!



# These embeddings vectors 'map' the words.

#### Similar words are close to each other

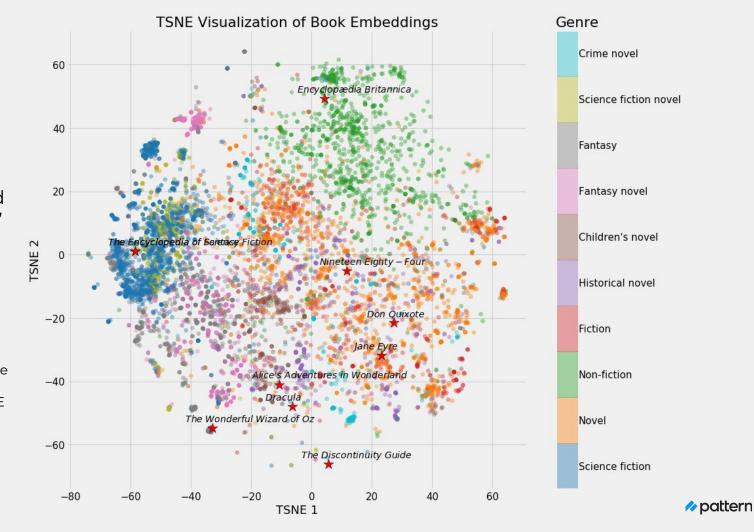
Seems simple when mapping individual words. But embeddings also scale...





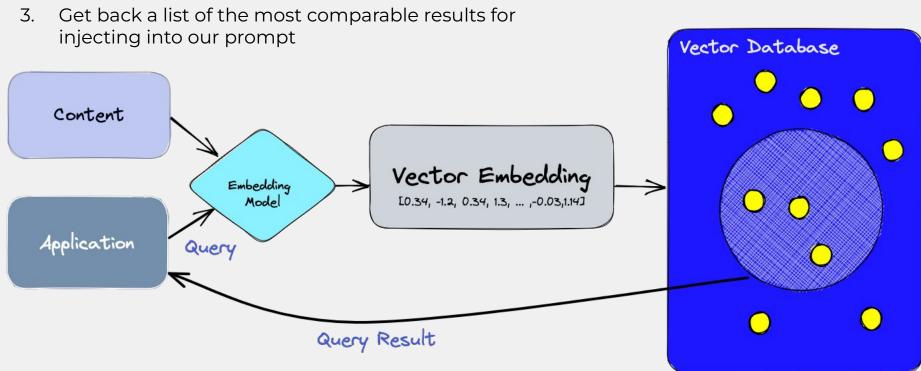
We can embed entire books and see how 'similar' they are!

These embeddings have been mapped to two dimensions using TSNE



#### Now we can store and retrieve our data!

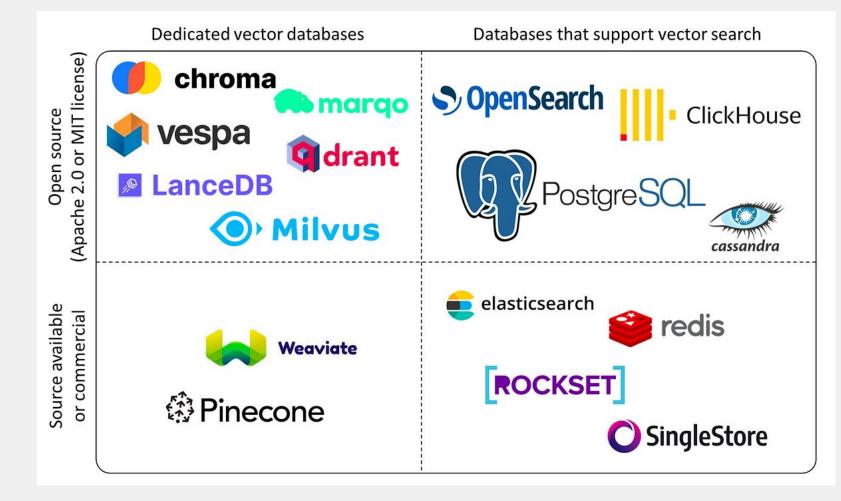
- 1. Store lots of chunks of text as embeddings
- Take your query and search it's embedded value in your database



## What can I use for a vector database?

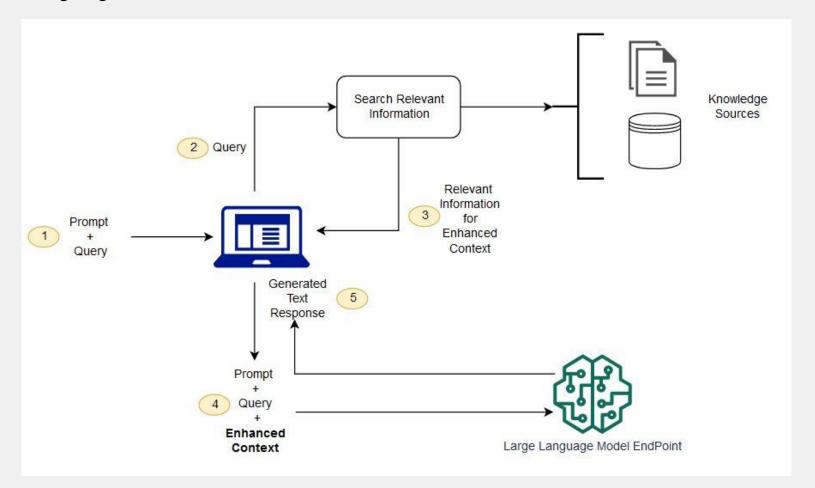
Are there many options available for me to build such a thing?





Yes!

#### **Putting Together the Whole Picture**



# See RAG in action (courtesy of Pinecone)

https://colab.research.google.com/github/pinecone-io/examples/blob/master/learn/generation/langchain/handbook/05-langchain-retrieval-augmentation.ipynb#scrollTo=DiRWzKh0mMGv

#### Tips for setup

- 1. Add !pip install --upgrade datasets multiprocess to end of first box before running anything
- Have an OpenAl API key handy (will cost a couple cents to use) <a href="https://platform.openai.com/account/api-keys">https://platform.openai.com/account/api-keys</a>
- 3. Have a pinecone API key handy (free trial) pinecone.io
- 4. Indexing step takes about 5 minutes



```
[24] 1 from langchain.chains import RetrievalQAWithSourcesChain.
2
3 qa_with_sources = RetrievalQAWithSourcesChain.from_chain_type(
4 1llm=llm,
5 chain_type="stuff",
6 retriever=vectorstore.as_retriever()
7)

1 qa_with_sources(query)
('question': 'who was Benito Mussolini2',
'answer': "Benito Mussolini was an Italian politician and journalist who served as the Prime Minister of Italy from 1922 until 1943. He was the leader of the National Fascist Party and played a significant role in the rise of fascism in Italy. Mussolini sought to establish Italy as a new Roman Empire and pursued expansionist policies, including the occupation of Abyssinia (Ethiopia) and Albania. He formed an alliance with Adolf Hitler and entered World War II as part of the Axis Powers. However, Mussolini's regime eventually fell, and he was captured and executed by partisans in 1945. \n',
'sources': 'https://simple.wikipedia.org/wiki/Benitot20Mussolini')

Now we answer the question being asked, and return the source of this information being used by the LLM.
```



# Thank you!

Any questions, comments, and/or concerns?

