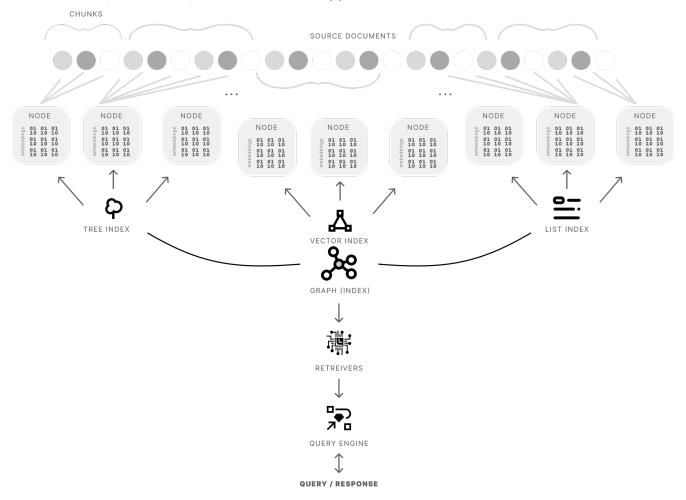
### Composable graphs with $\mathcal{MMR}$ queries over vector stores.

Composability allows you to to define lower-level indices for each document, and higher-order indices over a collection of documents. For e.g. imagine defining (1) a tree index for the text within each document, and (2) a list index over each tree index (per document) within your collection.[1]



Querying the index or a graph involves a three main components:

Retreivers $\rightarrow$	Response Synthesizer $\rightarrow$	Query Engine
A retriever class retrieves a set of Nodes from an index given a query.	This class takes in a set of Nodes and synthesizes an answer given a query.	This class takes in a query and returns a Response object. It can make use of Retrievers and Response Synthesizer modules under the hood.

For the query logic itself we will use maximum marginal relevance or  $\mathcal{MMR}$ . In this we iteratively find documents that are dissimilar to previous results. It has been shown to improve performance for LLM retrievals [2].

The maximum marginal relevance algorithm is as follows:

$$\text{MMR} = \arg \max_{d_i \in D \setminus R} [\lambda \cdot Sim_1(d_i, q) - (1 - \lambda) \cdot \max_{d_j \in R} Sim_2(d_i, d_j)]$$

Here, D is the set of all candidate documents, R is the set of already selected documents, q is the query,  $Sim_1$  is the similarity function between a document and the query, and  $Sim_2$  is the similarity function between two documents.  $d_i$  and  $d_j$  are documents in D and R respectively.

The parameter  $\lambda$  (mmr\_threshold) controls the trade-off between relevance (the first term) and diversity (the second term). If mmr\_threshold is close to 1, more emphasis is put on relevance, while a mmr\_threshold close to 0 puts more emphasis on diversity.

#### How to use

1. Rename .env.sample to .env and set environment variables OPENAI\_API\_KEY.

```
# within main folder
mv .env.example .env
echo "OPENAI_API_KEY=<your/key/here>" >> .env
```

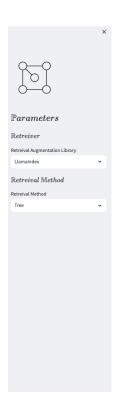
2. Run App

streamlit run main.py



## $\begin{tabular}{ll} $\mathbb{C}$ omposable $\mathbb{G}$ raphs with $\mathbb{V}$ ector indices for $\mathcal{LLM}$ queries. \end{tabular}$

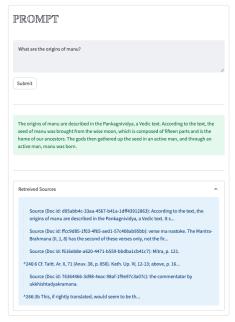
What are the origins of manu?	
Submit	,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,



#### 3. Technologies used

- 1. openai
- 2. llamaindex
- 3. langchain
- 4. streamlit

# $\begin{tabular}{ll} $\mathbb{C}$ omposable $\mathbb{G}$ raphs with $\mathbb{V}$ ector indices for $\mathcal{LLM}$ queries. \end{tabular}$



#### Indices

All example indices can be found in the **storage**/ folder for loading/reusing in other projects! They are primarily based on CC-commons Indic literature and history corpus from gutenberg.