Generating Answers

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INTRODUCTION

• This initiative is centered on producing responses to queries within a predefined context, demonstrating the combined use of the Cohere API for generating embeddings and the Annoy library for effective nearest neighbor searches. The main objective is to develop a system adept at comprehending and addressing questions relevant to a particular field.

DESIGN



Input text : whole information as Text or you can pass PDF using Langchain



Chunking: split the data into parts



Embedding: generate the Vector Space



Search Index: Search using index



Query: Pass the query or question from that text



Search: it searches using semantic search with Dense retrieval



Question: follow up Question or new question



Answers: Result

IMPLEMENTATION

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Input: Text and query

Setup the packages

Index Creation with Annoy: The Annoy library is employed to construct a search index from the embeddings, facilitating the quick retrieval of closely related items in the vector space.

Article Search **Function:** A function is developed to find articles matching specific queries using the search index and Cohere embeddings, providing pertinent context for answer formulation.

Answer
Generation
Process: A
specialized
function is
crafted to
produce
responses to
queries based
on the context
derived from
the search,
using Cohere
for response
generation.

Utility
Function
Developmen
t: Essential
utility
functions like
print_result
are created for
result
formatting and
display.

Non-Active Functions:
Some functions, such as search_wikiped ia_subset and generate_given _context, are included in the script but remain unused in the current implementation

TEST

CONCLUSION

The "Generating Answers" initiative highlights a dynamic integration of the Cohere API and the Annoy library, presenting an efficient method for interpreting and answering queries in natural language. The project skillfully combines data processing, the creation of embeddings, and the development of a search index, illustrating a comprehensive method for improving semantic comprehension. This integration of technologies enables the system to effectively produce articulate responses rooted in contextual insights, showcasing a solid foundation for constructing advanced question-answering systems.

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GITHUB:

https://github.com/Maryam-

 $Zubair/Machine Learning_Assignment/tree/main/ChatGP$

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