

Search Like a Human Building an Al Semantic Copilot with Elasticsearch



<u>Introduction</u>

- Problem: Why Traditional Search Falls Short.
- Content:
 Users expect natural language understanding:
 "Show me mind-bending sci-fi thrillers similar to Inception"
 "Find white sneakers with black stripes in size 9"
 "Suggest spicy Indian vegetarian dishes for dinner"

Keyword-based systems fail to understand meaning Need semantic search that understands intent

 Tools: Javascript , Microsoft Azure Open Al and Elasticsearch

How do search like Semantic?

Elasticsearch Store and search vectors using kNN

Azure OpenAl Generate text embeddings from queries (text-embedding-ada-002)

Node.js API Connect everything together

HTML/JS Frontend
Display results in real-time

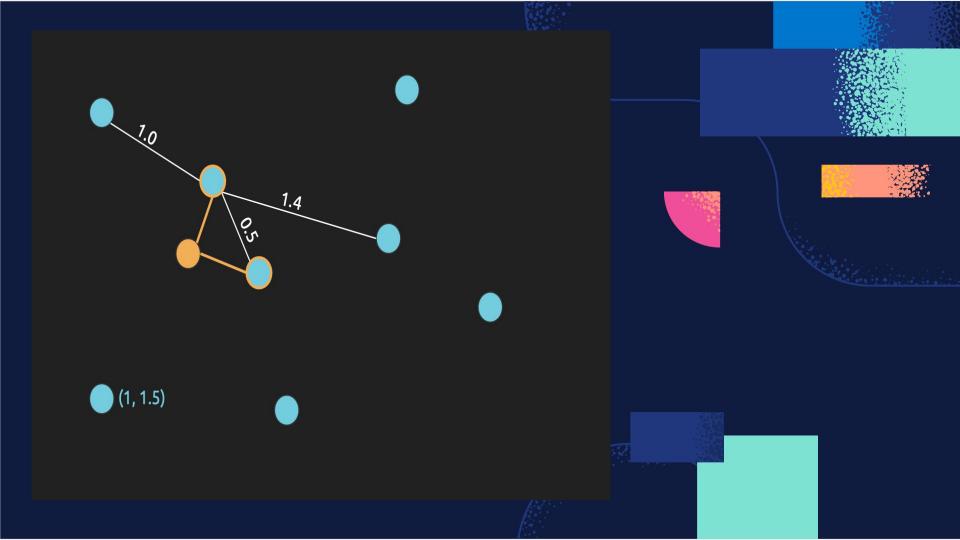
Benefits of Semantic Search

- Better user experience
- Accurate, meaningful results
- Scalable with Elasticsearch
- Future-ready with AI embeddings
- Easy to extend to products, articles, or documents

Elasticsearch Nearest Neighbor Search

The knn search option accepts a number of parameters that configure the search:

- field: the field in the index to search. The field must have a dense vector type.
- query_vector: the embedding to search for. This should be an embedding generated from the search text.
- num_candidates: the number of candidate documents to consider from each shard. Elasticsearch retrieves this many candidates from each shard, combines them into a single list and then finds the closest "k" to return as results.
- k: the number of results to return. This number has a direct effect on
 performance, so it should be kept as small as possible. The value passed in this
 option must be less than num_candidates.



Dataset Overview



Final Thoughts

The Future of Search Is Semantic

- Embeddings are the future of search
- Combining LLMs with vector databases unlocks powerful tools
- This project shows how easy it is to build intelligent search today

Questions?

Thank You

More about me



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