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Contact Info:

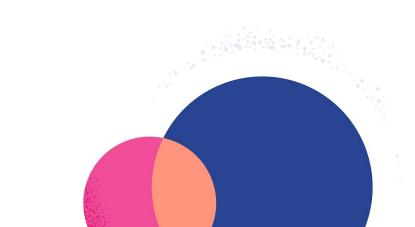
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Improving Retrieval Quality for Search



Agenda



Revisit Text Search (Lexical Search)

Vector Search

Semantic Search with dense vectors (Dense Vector Retrieval)

Semantic Search with sparse vectors (Learned Sparse Retrieval)

Hybrid Retrieval





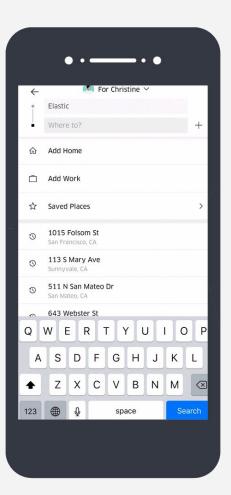
What is Elastic?



Elastic is a search company.

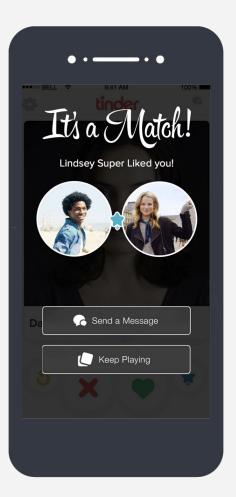






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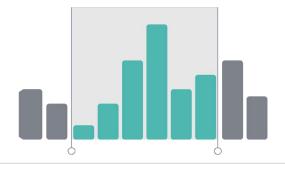


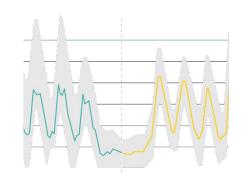


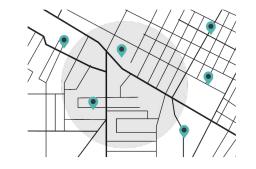


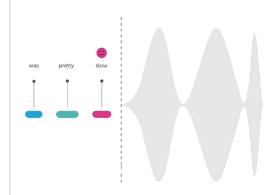






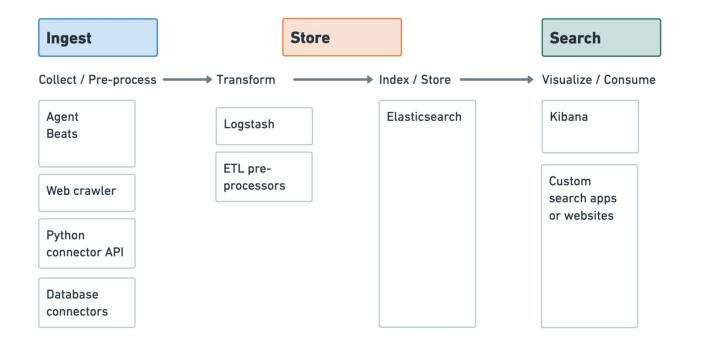






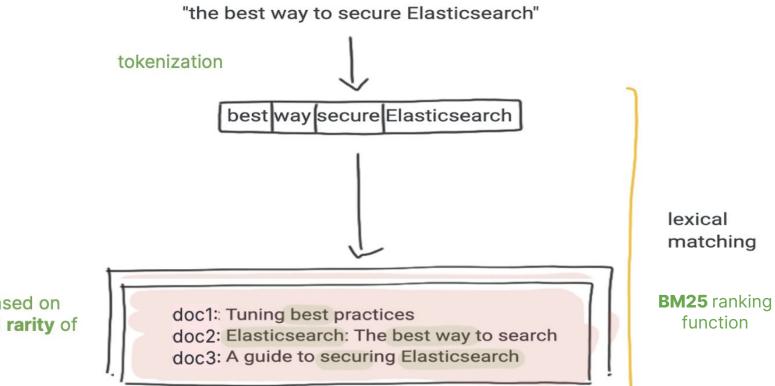


Typical Search Architecture



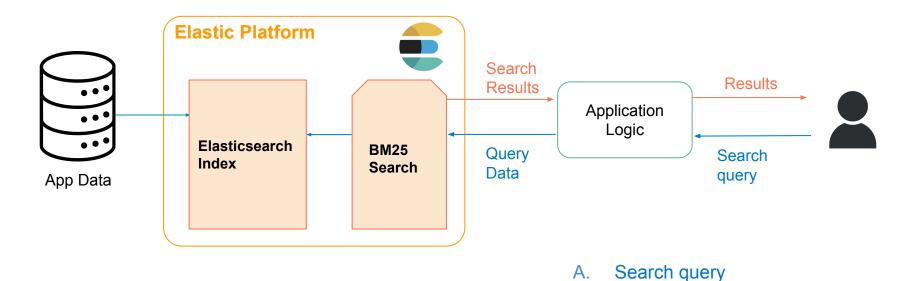


Text search revisited



relevance is based on **frequency** and **rarity** of these terms

Text search architecture revisited



Execution of BM25 Search

Β.



Text Search is useful for many use cases

	Where it works	Where it may fall short
Text Search (BM25)	Well understood	Vocabulary mismatch
	Interpretable	Context (semantic mismatch)



Users expect more from Search

In order to stream from our service you will need a high quality connection. The required connection speed for using the service will vary depending on the quality of t you wish

connection speed requirements

Q

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recommend at least...

Retrieve results based on intent and contextual meaning of search queries, not just terms



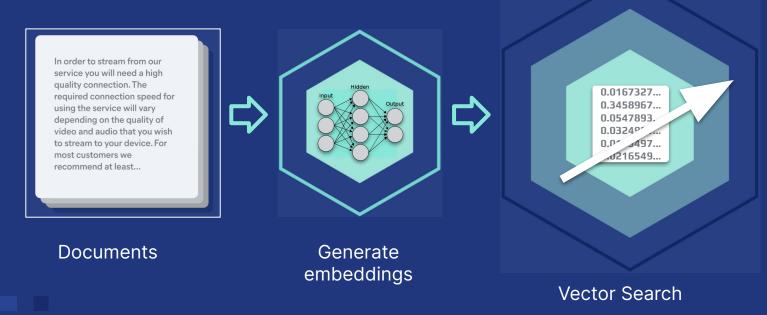
Vector Search enables Semantic Search





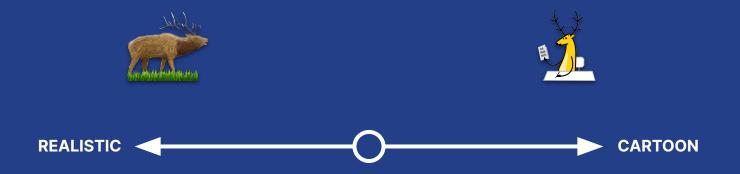
What is **Vector Search**?

Search based on Vector Representations (or "vector embeddings")



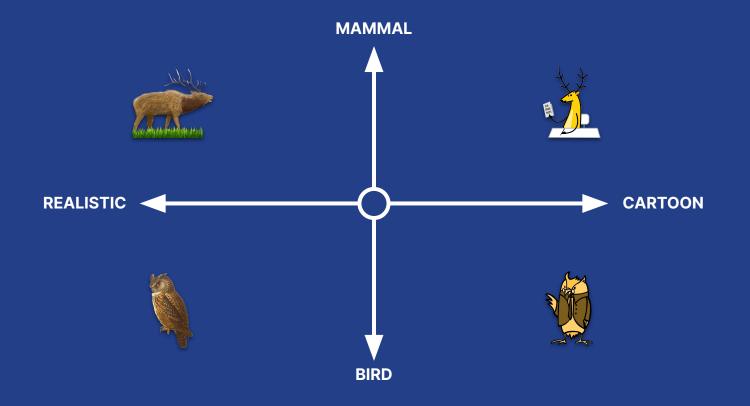


Embeddings represent your data Example: 1-dimensional vector



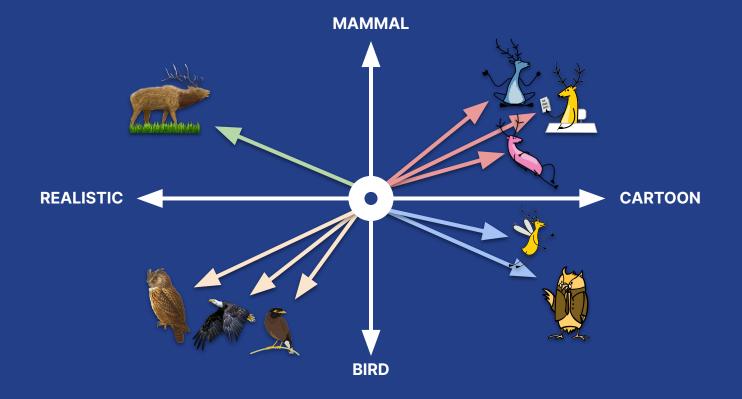


Multiple dimensions represent different aspects of data



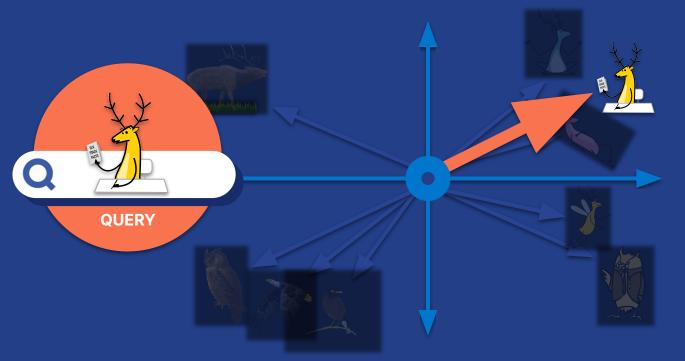


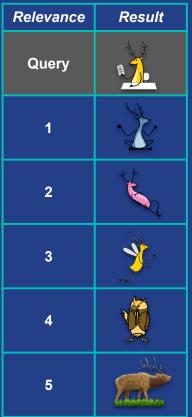
In the "embedding space", similar data are grouped together





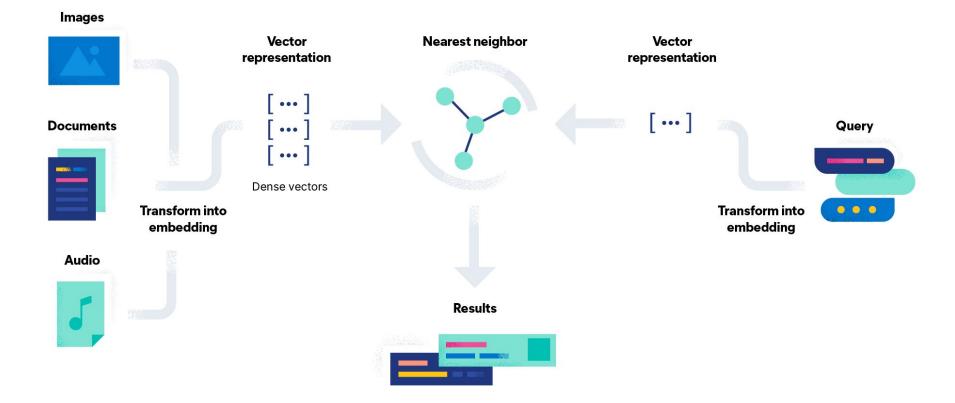
Vector search ranks objects by similarity (relevance) to the query







Vector search: conceptual architecture



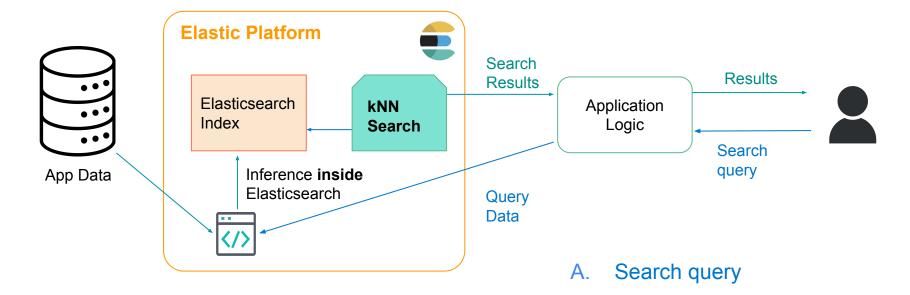
How to generate context aware text embeddings? Apply a Natural Language Processing (NLP) model!

(+)With Elastic \rightarrow Import and Deploy <u>proprietary</u> or <u>third party</u> NLP models.

Deploy and use the model **Select** the text embedding model **Import** into Elasticsearch Q Find apps, content, and more. 36/ Models 173 Filter by name new Full-text search 11 Sort: Trending odel Management Trained Models sentence-transformers/all-MiniLM-L6-v2 誤 Sentence Similarity + Updated Nov 7, 2022 + 生 2.43M + ♡ 938 \$ eland import hub model Trained TECHNICAL PREVIEW hkunlp/instructor-xl (Auto refresh Off Models 1 Sentence Similarity + Updated Jan 21 + ± 12.3k + ♥ 375 --11rl sentence-transformers/all-monet-base-v2 2 Sentence Similarity + Updated Jul 11, 2022 + ± 9.29M + ♥ 375 https://Elastic Cluster URL Total trained models: 8 thenlper/gte-large --hub-model-id bert model ﷺ Sentence Similarity + Updated about 1 month ago + ± 35.6k + ♡ 114 Q Search. --task-type text embedding thenlper/gte-base 1 Sentence Similarity + Updated Aug 14 + ± 79.4k + ♥ 38 ID 个 Description State Type bhadreshintfloat/multilingual-e5-base savani_dis W Sentence Similarity + Updated 16 days ago + ↓ 30.3k + ♥ 86 Model bhadresh-savani/distilbert-base tilbertpytorch uncased-emotion for task type started basetext_classifica thenlper/gte-small 'text_classification' ₩ Sentence Similarity + Updated Aug 14 + ± 22.1k + ♡ 35 uncasedotion . . **O** PyTorch eland Hugging Face Elastic Python Elasticsearch Cluster Open repository: Client huggingface.co

Generating dense embeddings with Elastic: Two options

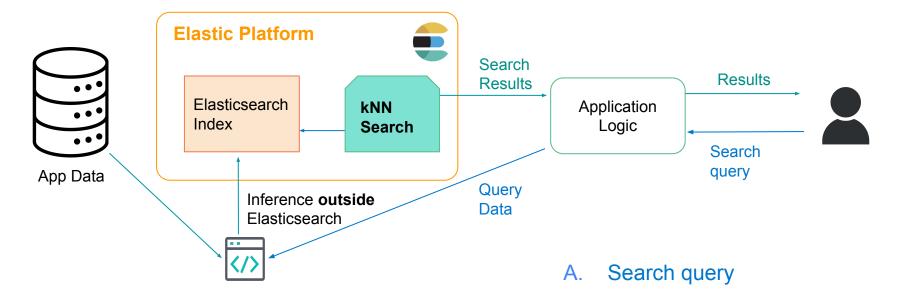
Generate embeddings *inside* Elasticsearch



B. Execution of kNN Search

Generating dense embeddings with Elastic: Two options

Generate embeddings **<u>outside</u>** Elasticsearch



B. Execution of kNN Search

Dense Vector retrieval performs well (conditions apply*)

	Where it works	Where it may fall short
Dense Vector	Can beat other approaches for semantic search	Domain adaptation*

Not easily interpretable (no exact match)



Learned sparse retrieval - an alternative approach for Semantic Search

Provides 'trade-offs' over dense retrieval and traditional sparse retrieval methods (BM25)



Term Expansion

By identifying contextual importance between terms, it utilizes that knowledge to improve sparse vector embeddings

Query: "Comfortable furniture for a large balcony"

Doc: "is a comfortable and stylish garden lounge set, including a sofa, chairs, and a side table for outdoor relaxation"



Term Expansion

By identifying contextual importance between terms, it utilizes that knowledge to improve sparse vector embeddings

Query: "Comfortable furniture for a large balcony"

without term expansion (lexical search)

Doc: "is a comfortable and stylish garden lounge set, including a sofa, chairs, and a side table for outdoor relaxation"



Term Expansion

By identifying contextual importance between terms, it utilizes that knowledge to improve sparse vector embeddings



Doc: "is a comfortable and stylish garden lounge set, including a sofa, chairs, and a side table for outdoor relaxation"



Elastic provides a 'built-in' option for this approach!

(2) Start the model deployment

(1) Download the model

Trained Models		🤔 Auto refresh Of	f	ි Refres
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Learn more 🖸

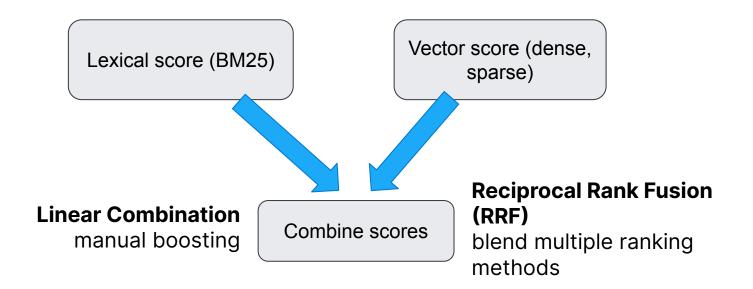
Learned Sparse retrieval is an improvement over text search

	Where it works	Where it may fall short	
Text Search (BM25)	Well understood	Vocabulary mismatch	
	Interpretable	Semantic mismatch	
Learned Sparse Retrieval (ELSER)	 Interpretable / Well understood (tokens) 	Larger index (terms/tokens)	
	+ Vocabulary/semantic matching - Perform Semantic Search	<u>Dense vector retrieval</u> can outperform learned sparse retrieval for <u>semantic search</u>	



Is a combined approach a better idea?

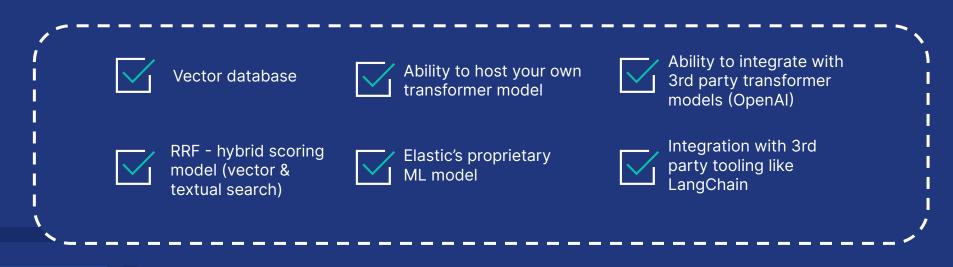
Hybrid retrieval:





Powered by

Elasticsearch Relevance Engine[™]





Join the Elastic community!







Learn More: elastic.co/ search-labs

Try free on Cloud: elastic.co/cloud

Connect on Slack: ela.st/slack



Thank you!

Questions and demo at the Elastic booth

Contact Info: E-mail: priscilla.parodi@elastic.co Linkedin: Priscilla Parodi



