



Unlocking the promise of Generative AI for businesses

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Recap

Objectives

- Improve customer/employee experience by delivering better search relevance
- Minimize time to implement and fine tune relevance

Challenges

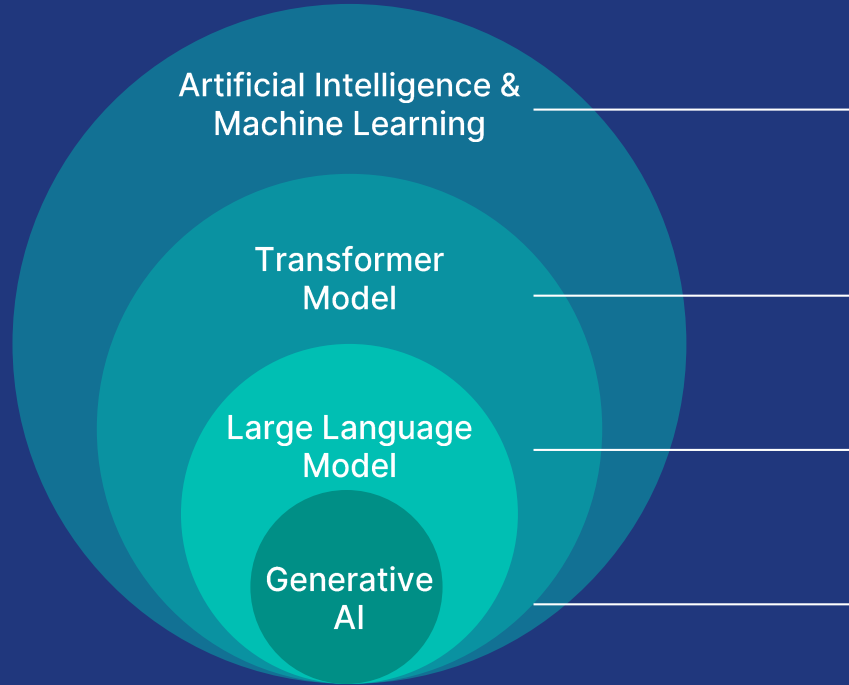
- Getting the right information to the right audience
- Supporting legacy technology takes away from innovation and strategic impact
- Slow, unreliable search performance



Introduction

Introduction to Elasticsearch and Generative AI

ML, Transformer, and Large Language Model basics



What is it?

The science of teaching computers to think, learn, and improve on their own.

A Neural Network architecture that considers word relationships and context.

An AI model that uses massive data to generate human-like text and perform language tasks exceptionally well

A large language model trained to compose content and responses to human prompts.

Like any new **revolutionary** tech, Generative AI comes with a new set of strengths and limitations

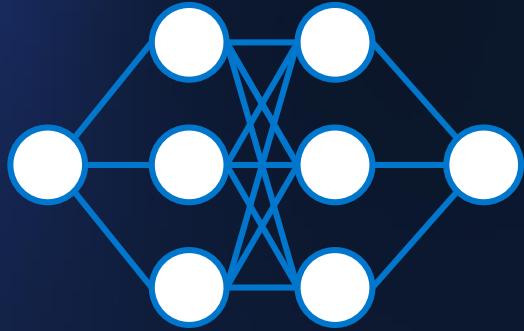


Excels at human-like, iterative content creation

Natural language processing of large data sets used for creating human-like conversations, writing content, and providing code examples

Limitations of Large Language Models

- Base models are trained on public data
- Data is frozen in time after training and fine-tuning
- Non-deterministic results and Hallucinations
- Cost & Privacy concerns for large scale use



A large language model is not a database

Searching YOUR Data

How to search your data:
BM25, Embeddings, Vector Similarity, Retrieval Strategies

Elasticsearch is a search engine - Inverted indices

Indexing text fields in documents



Inverted indices: tokens, count of frequency, and documents containing them



Freq and positions: frequency and token offset for scoring, phrase queries and more



BM25: A sparse, unsupervised model for lexical search, improvement over TF/IDF



Speed and Space Optimizations: Block Max WAND, query short circuiting, match_only_text, and more!

1: Winter is coming.
2: Ours is the fury.
3: The choice is yours.

<u>term</u>	<u>freq</u>	<u>documents</u>
choice	1	3
coming	1	1
fury	1	2
is	3	1, 2, 3
ours	1	2
the	2	2, 3
winter	1	1
yours	1	3

Dictionary Postings



How do we get the most relevant context to answer the user's question, in a natural way?

Beyond the BM25 "bag of words"



Semantic Search

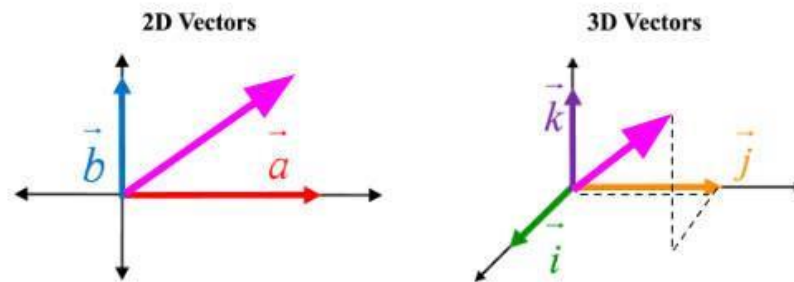
Representing and finding
the meaning of information

Vector Embeddings & Vector Databases

Remember Vectors?



What math textbooks looked like before common core (US centric joke, sorry)



$$\mathbf{v} = [1.0, 0.5, -2.1]$$

You likely learned the math for 3D vectors.

Some may have learned the math for n-dimensional vectors later in school.

NLP uses vectors with hundreds to thousands of dimensions (not pictured)

Meaning can be encoded as a high dimensional Vector

Text Embedding Model

Chunking -> Transformer or LLM -> vector

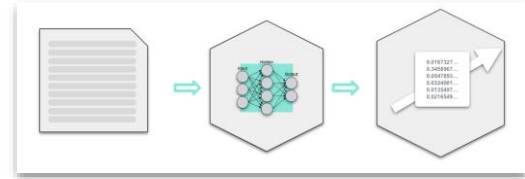
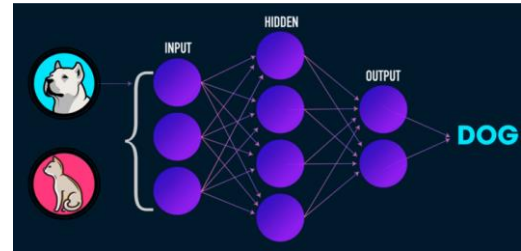


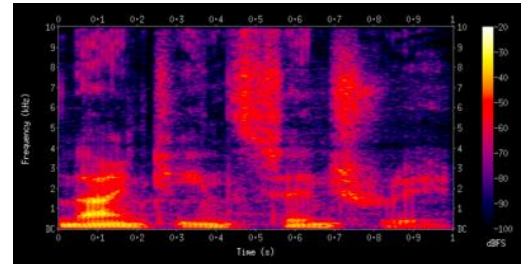
Image Classification (i.e. Google image search)

Convolutional Neural Network -> Trained model -> vector



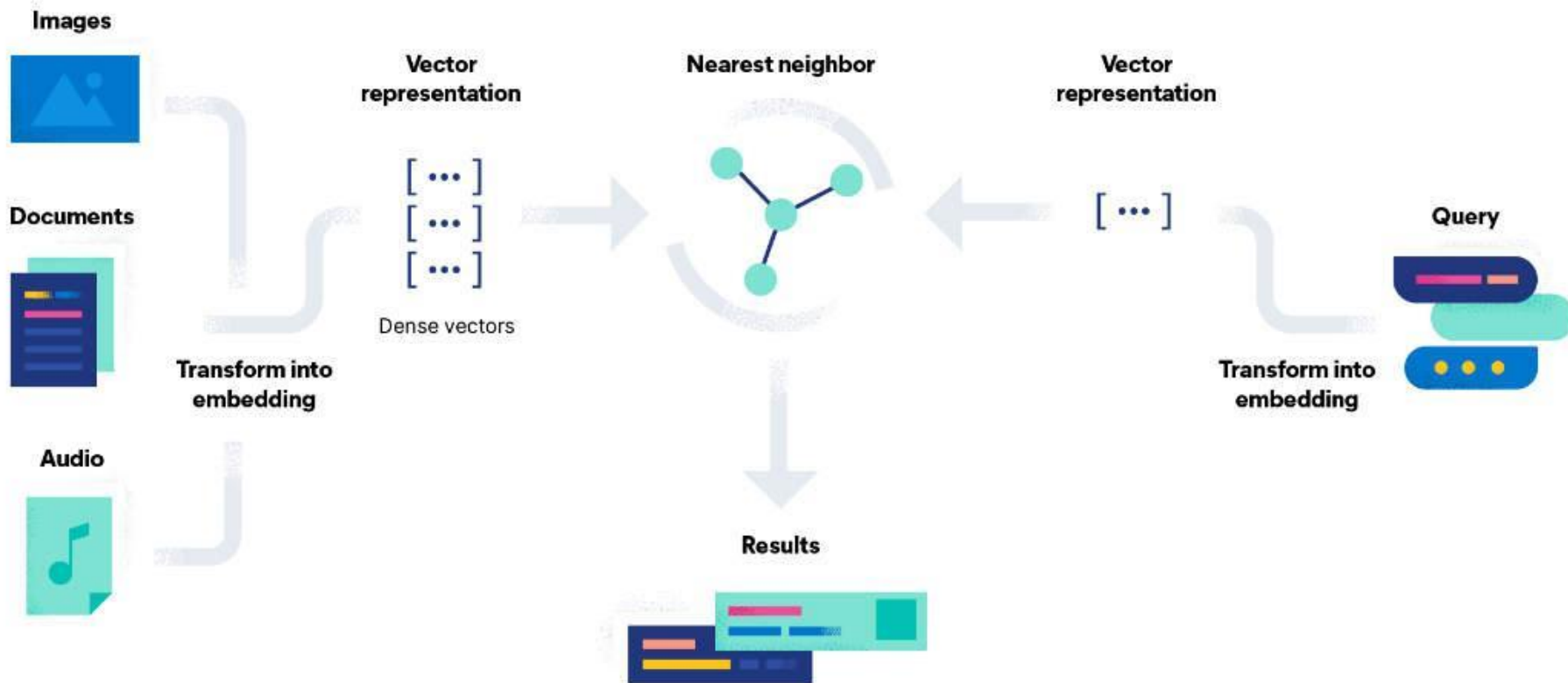
Soundwave Classification

FFT + time -> Spectrograph -> Image (see above)



Vector search conceptual architecture

Use vector nearest neighbor to generate a search ranking



Data Ingestion and Embedding Generation


You asked, we answered: Our best-selling classic wrap dress now comes in a cotton poplin that's wear-all-day perfect. Bonus: stripes (our favorite).

FIT

- 39" from high point of shoulder

DETAILS

- Cotton
- Linen
- Machine wash
- Import



Source data

POST /_doc



```
{
  "_id": "product-1234",
  "product_name": "Summer Dress",
  "description": "Our best-selling...",
  "Price": 118,
  "color": "blue",
  "fabric": "cotton",
  "desc_embedding": [0.452, 0.3242, ...]
}
```

POST /_doc



ML Inference pipelines + Add inference pipeline

Inference pipelines will be run as processors from the Enterprise Search Ingest Pipeline

ml-inference-embedding-generation	Actions
● Deployed	pytorch text_embedding
ml-inference-emational-analysis	Actions
● Deployed	pytorch text_classification

[Learn more about deploying ML models in Elastic](#)



Vector Query

🔍 summer clothes | ✕



Transformer model

GET product-catalog/_search

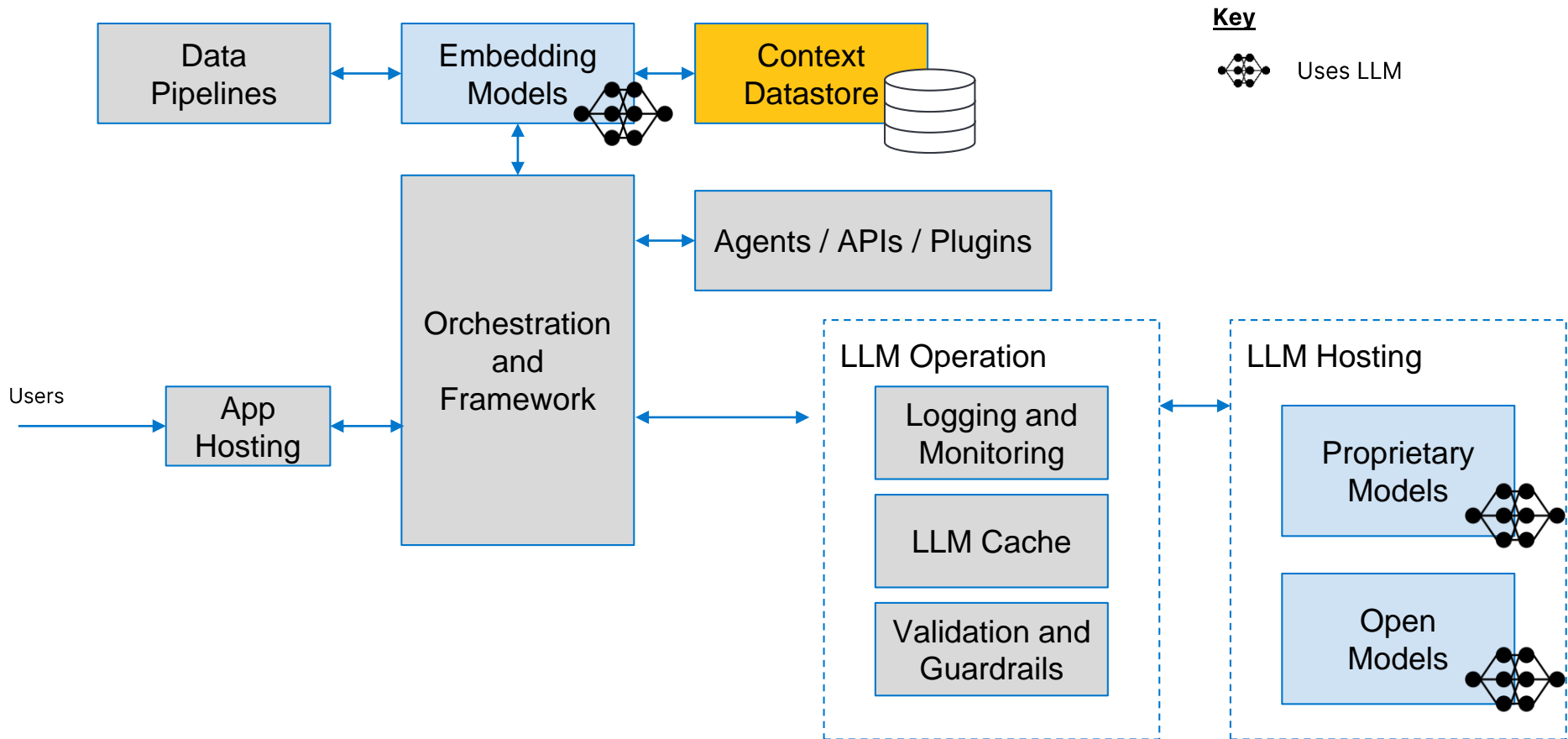
```
{
  "knn": {
    "field": "desc_embedding",
    "k": 5,
    "num_candidates": 50,
    "query_vector_builder": {
      "text_embedding": {
        "model_text": "summer clothes",
        "model_id": <text-embedding-model>
      },
    },
    "filter": {
      "term": {
        "department": "women"
      }
    },
  },
  "size": 10
}
```

RAG Applications

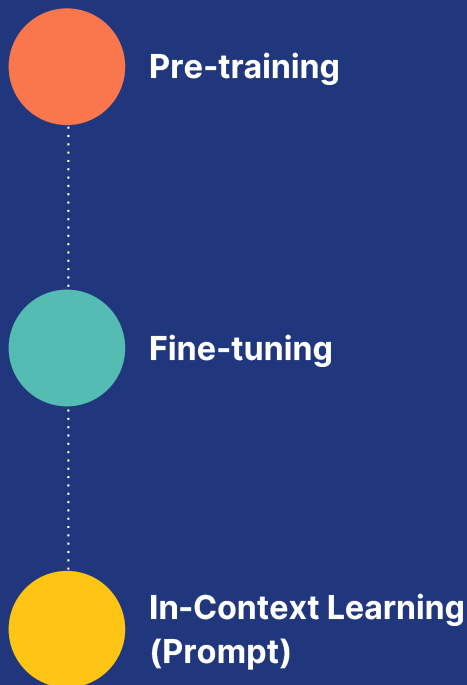
RAG = Retrieval Augmented Generation

Developing AI Applications with Context

Emerging Generative AI App Architecture



The 3 ways LLMs get 'smart'



Foundational or Base model training costs tens to hundreds of millions of \$USD. LLMs learn **language** and **knowledge** from massive public data sets.

- **Task specific training** (classification, etc)
 - Improve **quality of responses** in a domain
 - Add **knowledge** from a specific data source
 - **Alignment** with safeguards and ethical limits
-
- **Prompt engineering** techniques
 - **In-context learning and instruction**
 - **Retrieval Augmented Generation**
 - Include **knowledge** in prompt

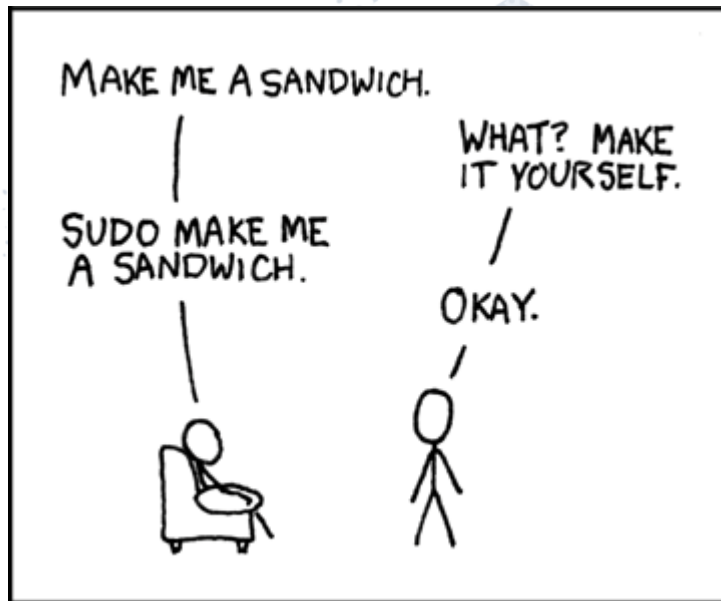
'Prompt Engineering'

Prompt Engineering is the **art** and science of designing effective prompts to **guide** the responses of Large Language Models.

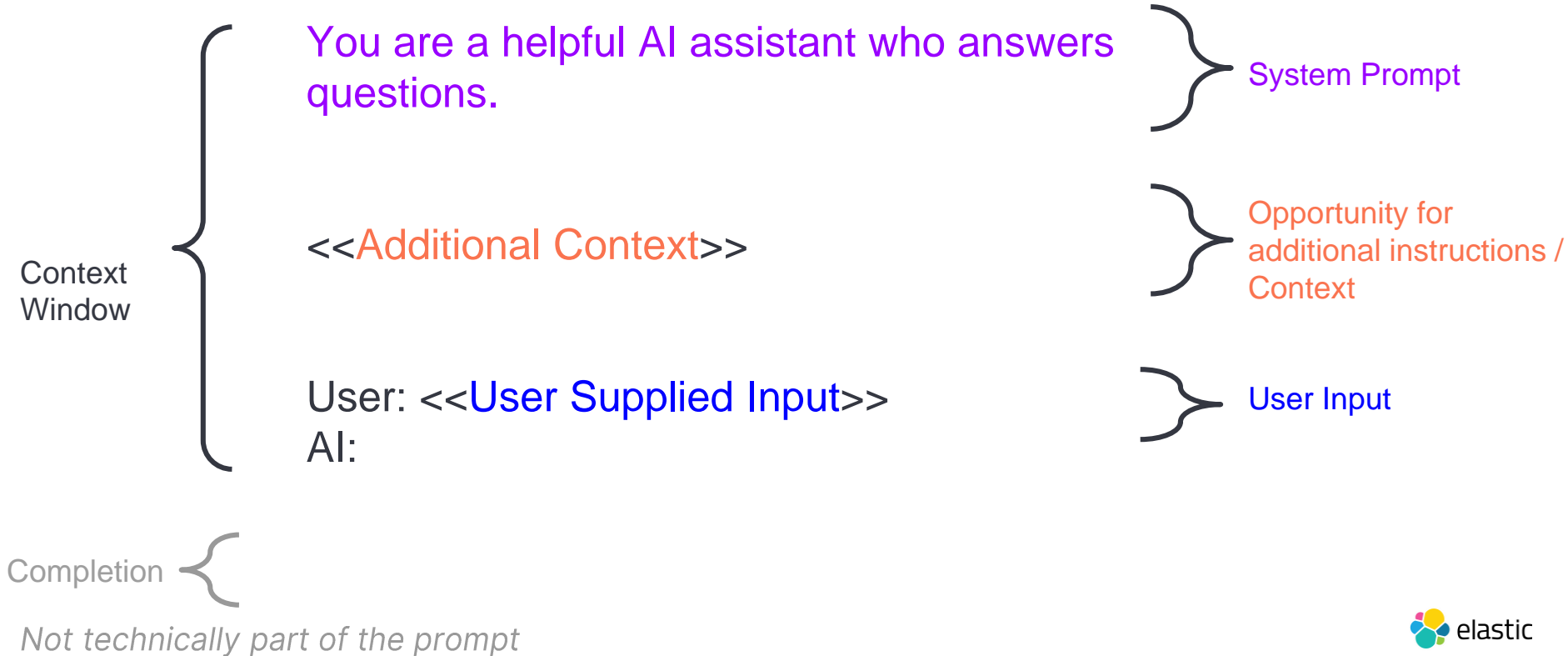
a.k.a. Coding in 'Natural Language'

Prompts can ask the LLM to:

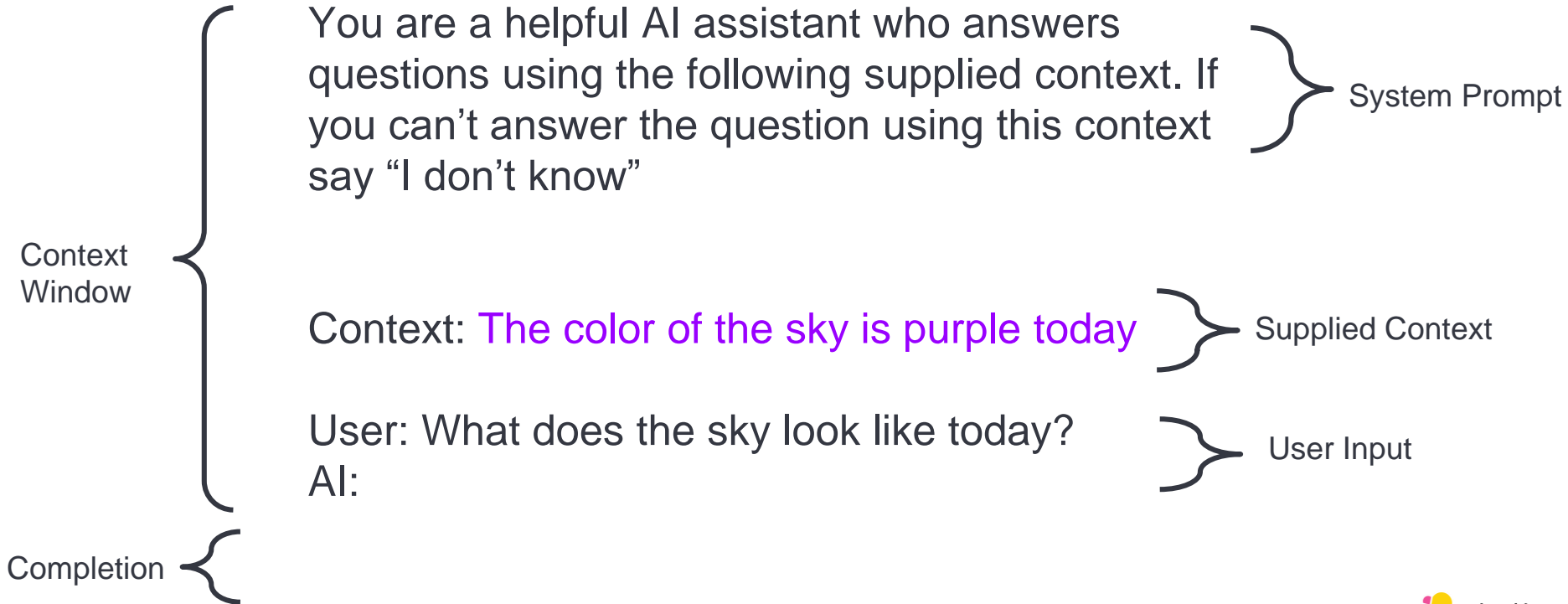
- Complete a task like summarization
- Follow provided context
- Make step by step plans or instructions
- Format output in specific useful ways



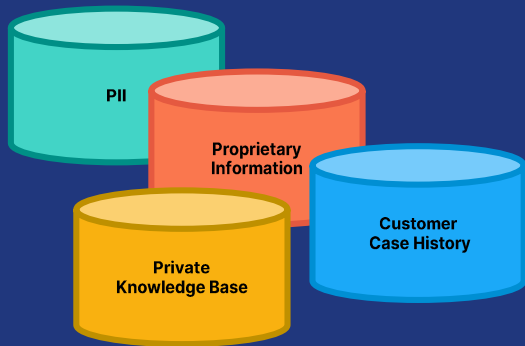
Anatomy of a Prompt



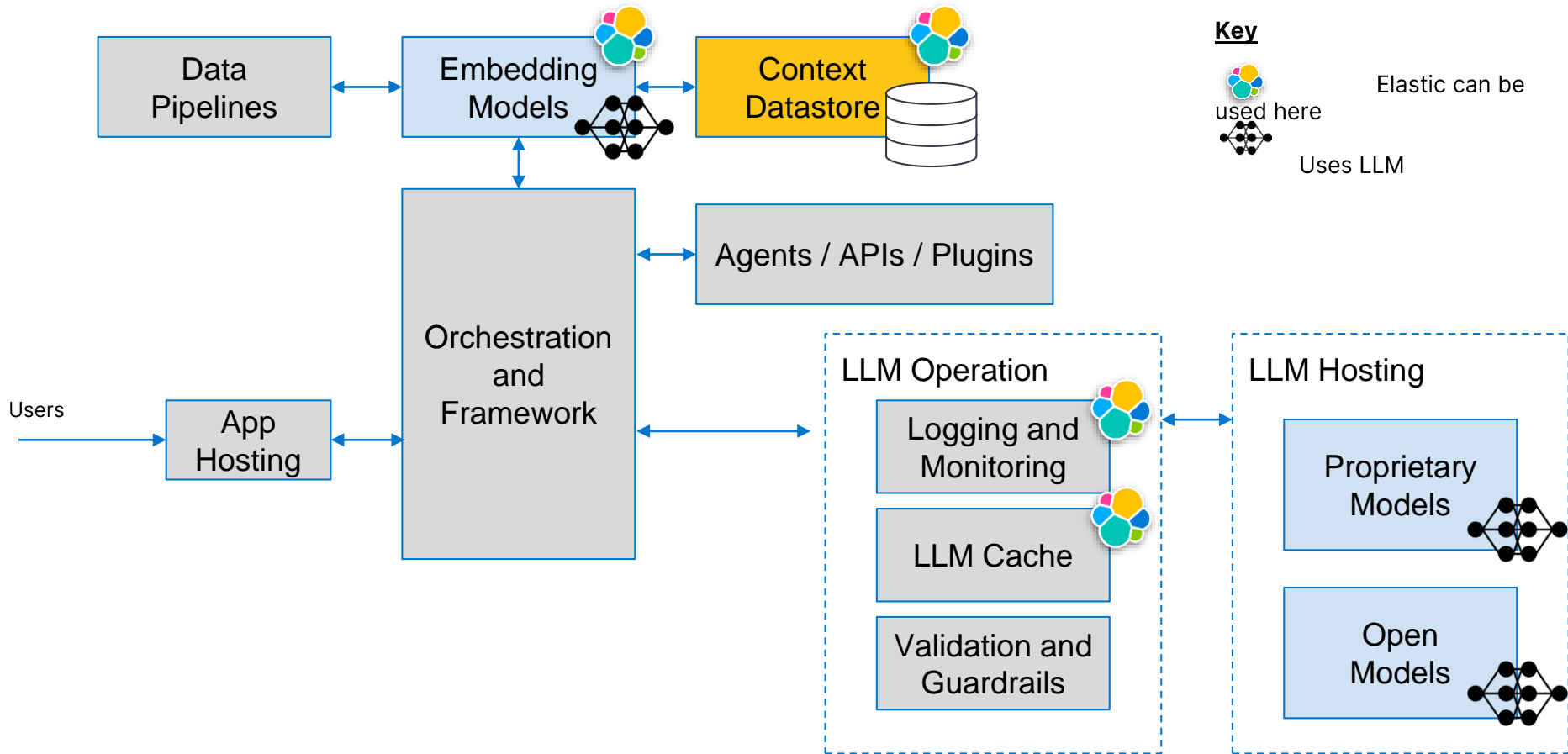
Retrieval Augmented Generation (RAG)



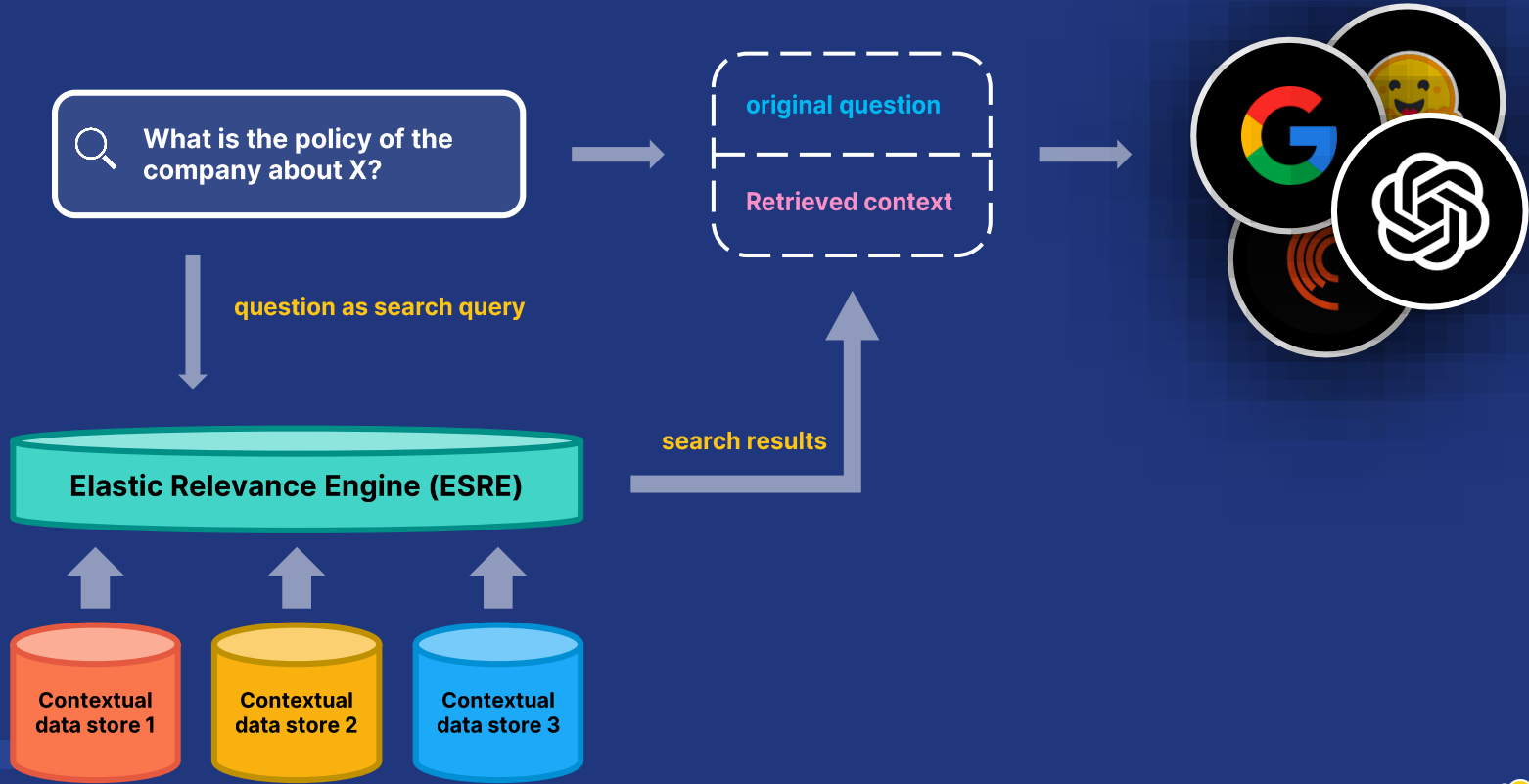
RAG uses semantic search techniques like those in Elasticsearch to act as the bridge between private data and Generative AI



Emerging Generative AI App Architecture



Question Answering + Context Retrieval Workflow

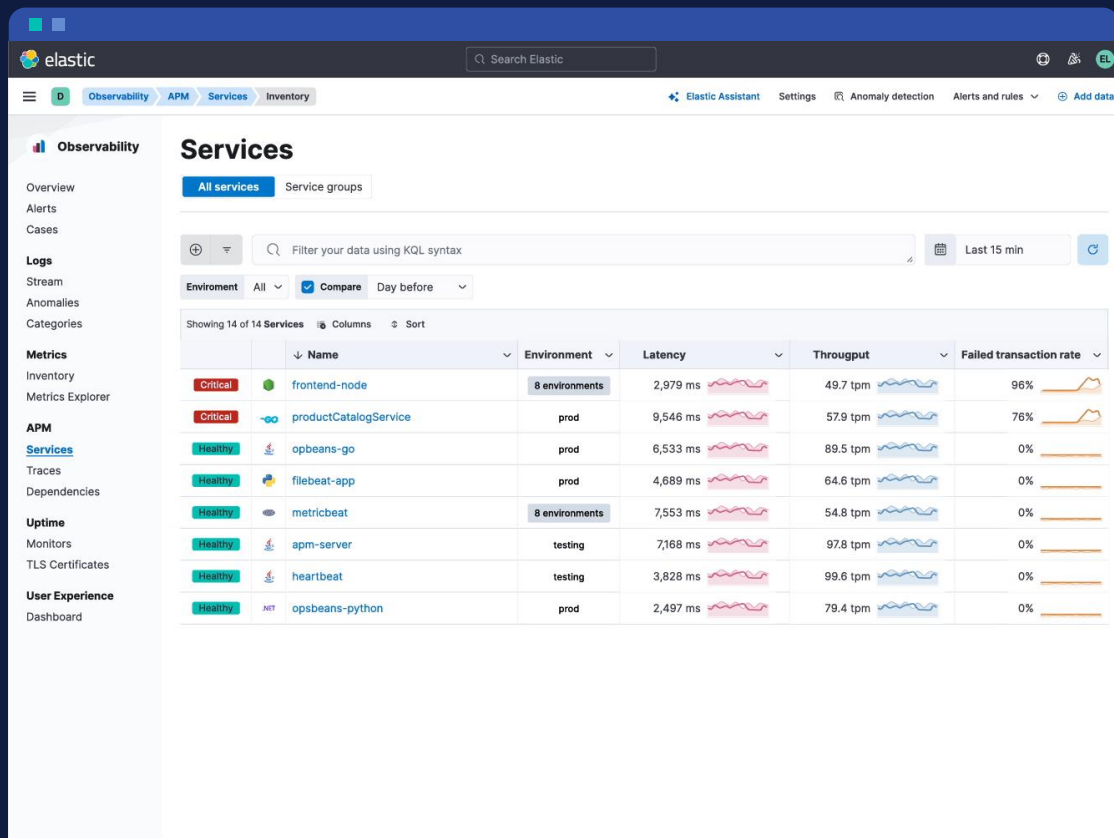


Use Cases

AI Assistant for Observability

Powered by **ESRE** Elasticsearch
Relevance Engine™

- Accelerate incident management and root cause analysis
- Interactively explore problems and execute remedies with generative AI
- Context-aware, business-specific output you can trust
- Based on your proprietary data and runbooks



AI Assistant for Security

Powered by **ESRE** Elasticsearch Relevance Engine™

- Accelerate incident management and threat detection
- Get expert assistance during any investigation
- Gain industry insight into attack patterns
- Incorporate into existing incident response workflows

The screenshot displays the Elastic AI Assistant interface within the Elastic Security console. The main window shows a conversation history with the following messages:

- Assistant at: 9/22/2023, 10:17:56 AM**
Yes, this alert appears to be related to the previous one. Both alerts involve the same `certutil.exe` process (same process hash and PID), parent process `EXCEL.EXE`, and the user `james_spiteri`. The alerts were also generated within a few seconds of each other.

This new alert is specifically for "Suspicious CertUtil Commands" and is triggered based on the suspicious use of `certutil.exe` with the command line `certutil -decode C:\Programdata\B2E42126.txt C:\Programdata\05122914.exe`.

CertUtil is a native Windows component that is part of Certificate Services and can be abused by attackers to live off the land for stealthier command and control or data exfiltration. In this case, the CertUtil command is being used to decode a possibly obfuscated file `B2E42126.txt` and save the result as an executable file `05122914.exe`. This activity, combined with the alert for suspicious MS Office child processes, could indicate a potential compromise.
- You at: 9/22/2023, 10:19:52 AM**
Based on these 2 alerts, can you provide a breakdown of this attack, and create a markdown diagram for me?
- Assistant at: 9/22/2023, 10:20:17 AM**
Based on the two alerts, we can break down the suspected attack as follows:

At the bottom of the interface, a table displays details for an alert titled "Uncommon Registry Persistence Change":

Field	Value
user.name	james_spiteri
process.executable	C:\Windows\System32\certutil.exe
kibana.alert.rule.type	eq
process.name	certutil.exe
process.parent.name	EXCEL.EXE
	certutil -decode

Generative AI is evolving within enterprises: HR



TODAY



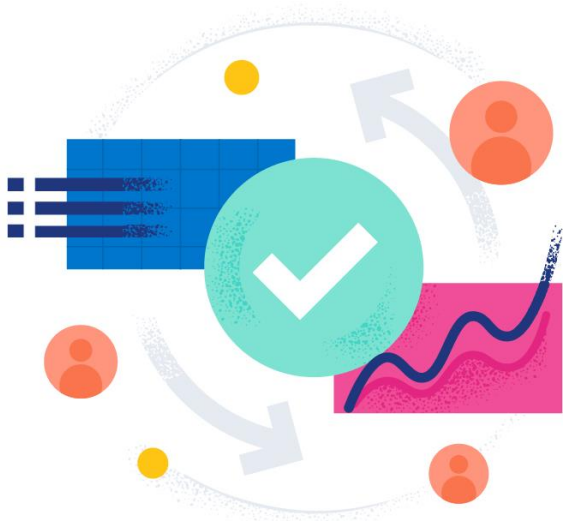
401K policy

TOMORROW



What are key aspects of the company's 401k policy for an employee in my location and how do I enroll?

Generative AI is evolving within enterprises: Legal



TODAY



Work contract for California

TOMORROW



What are the main labor and employment law requirements at our California office?

Generative AI is evolving within enterprises: Customer Success



TODAY



Customer shopping locations

TOMORROW



Are my customers in Dallas buying products at locations most convenient for them, and with the deepest discounts?



Thank You

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