# Large Language Models in Domino Applications

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### Serdar Basegmez

- Developer/Half-blooded Admin
- New(ish) Londoner Ex-Istanbulite
- Freelance Consultant at Developi UK
- Member Director at OpenNTF Board

- Notes/Domino since 1999
- IBM Champion Alumni (2011-2018)
- HCL Ambassador (2020-2024)

- Blog: LotusNotus.com / Twitter: @serdar\_basegmez
- Also tweets/writes/speaks/podcasts on scientific skepticism





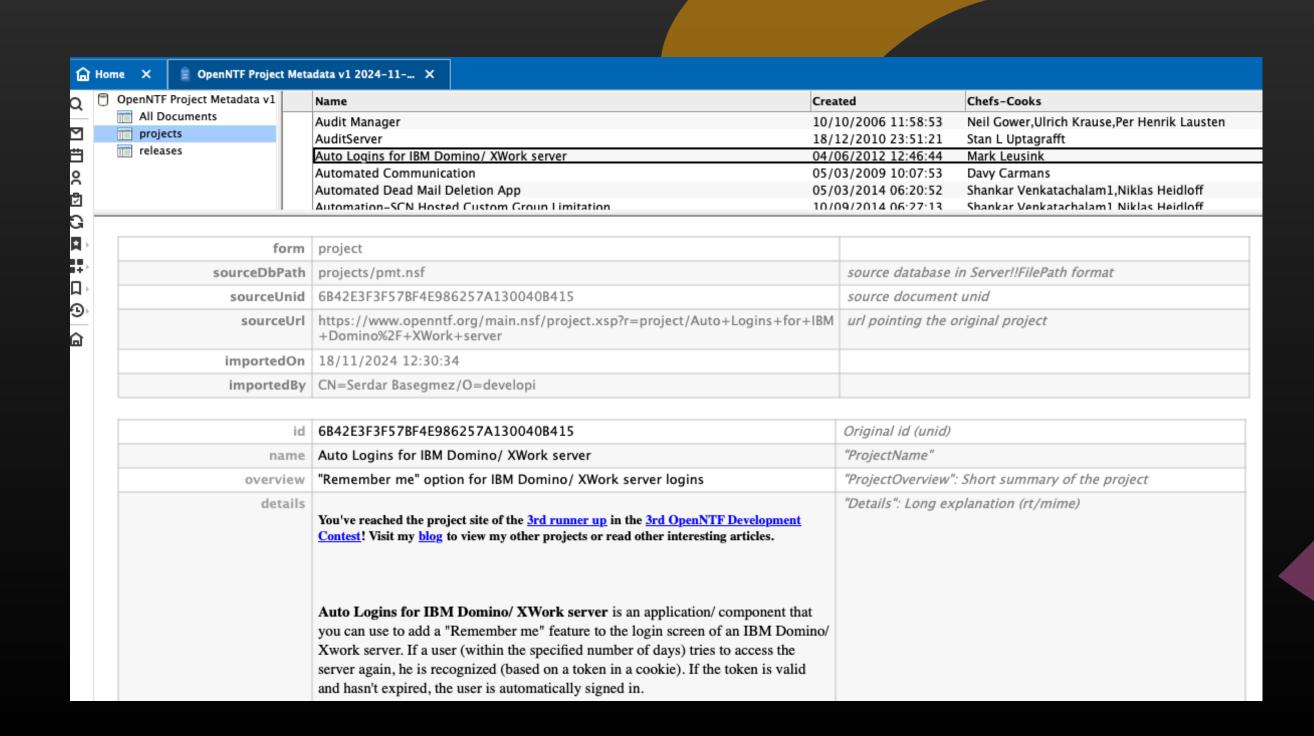


### **Everything Open Source!**

Demos, source codes, libraries, integrations, datasets...



https://github.com/sbasegmez



### Today...

Large Language Models

Glossary of Terms

Potential Applications

LLM Integration Methods

Assessing Our Toolbox

Conclusion



### Understanding the Impact

Large Language Models: Transformative or Overhyped?



### Game Changer?

- A new paradigm in programming?
  - Programming with prompts...
- New ways to interact
  - Conversation Chat or audio
  - Accessibility
- Ability to use "unusable" data
  - Extract value from documents, audio, images
  - Multilingual content
  - Cultural context, specialized knowledge...



### Or, Yet Another Big Hype?

- Safety, security, privacy, compliance
  - Ethical issues
  - Bias and Fairness
- "Glorified auto-complete"?
  - Lack of creativity and critical thinking
- Indeterministic behaviour
  - "Temperature" trade-off
  - Hallucinations
- Scalability and Efficiency



### Insanity Check...

- Nearly 80% of Al projects fail!
  - Double rate of other IT projects.

- Why?
  - Misunderstood problem definition
  - Complex problem
  - Data Quality and Availability
  - Technology-driven rather than solution-focused
  - Infrastructure is not sufficient



### Foundations and Evolution

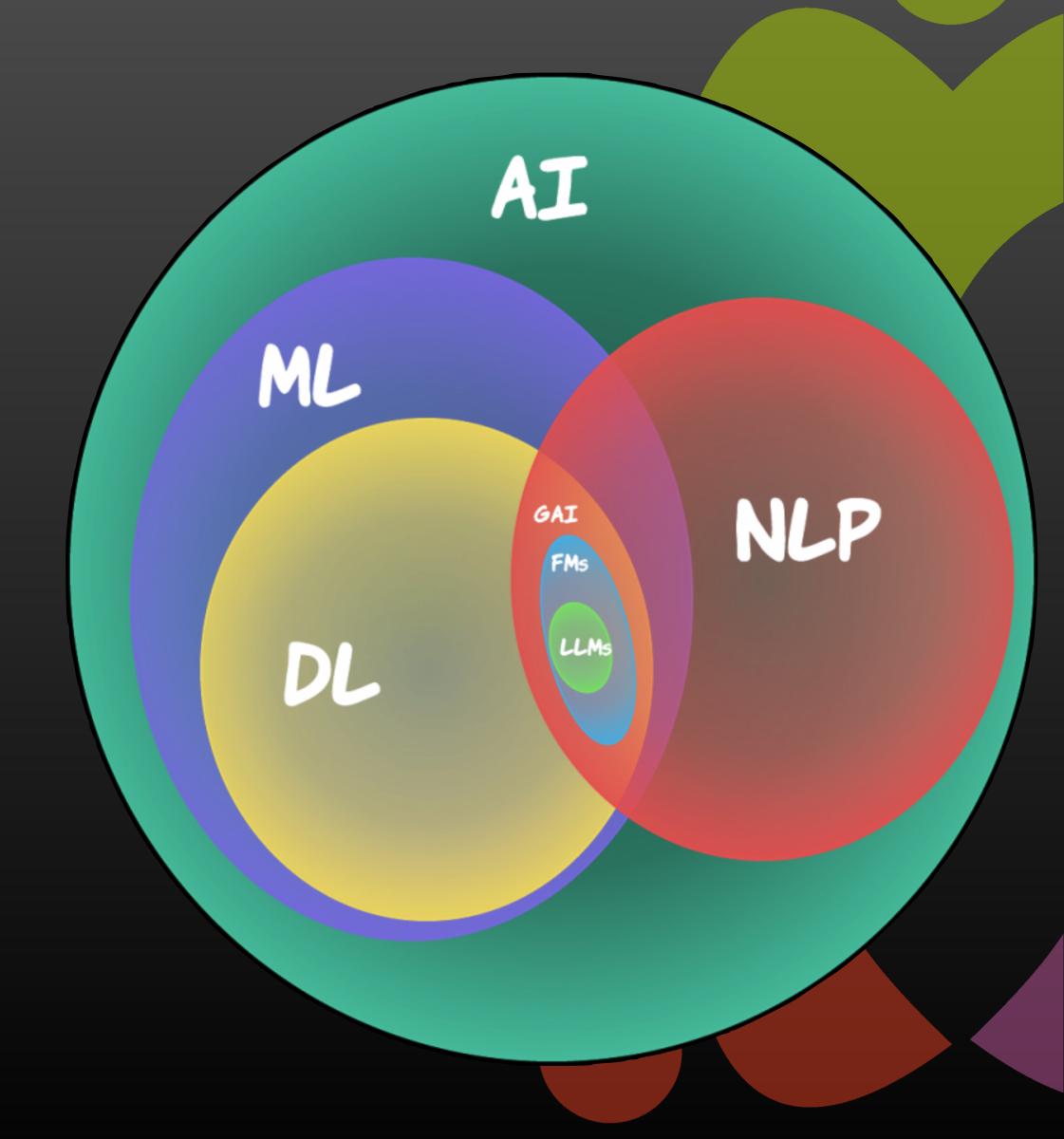
Key concepts and their progression



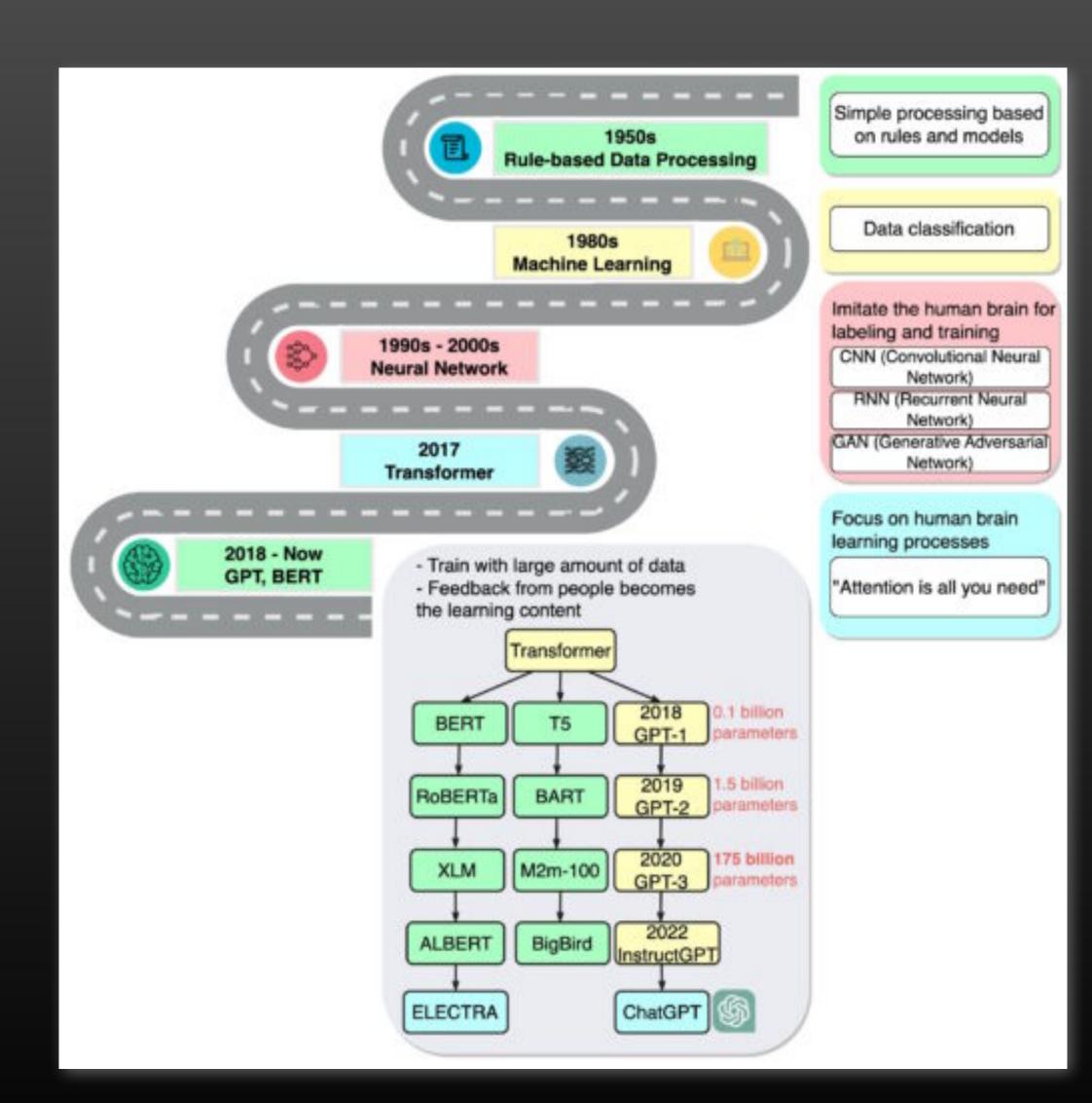
### Glossary of Terms

- Artificial Intelligence
- Machine Learning
- Deep Learning
- Natural Language Processing

- Generative Al
- Foundation Models
- Large Language Models



### **Short History**



#### **Attention Is All You Need**

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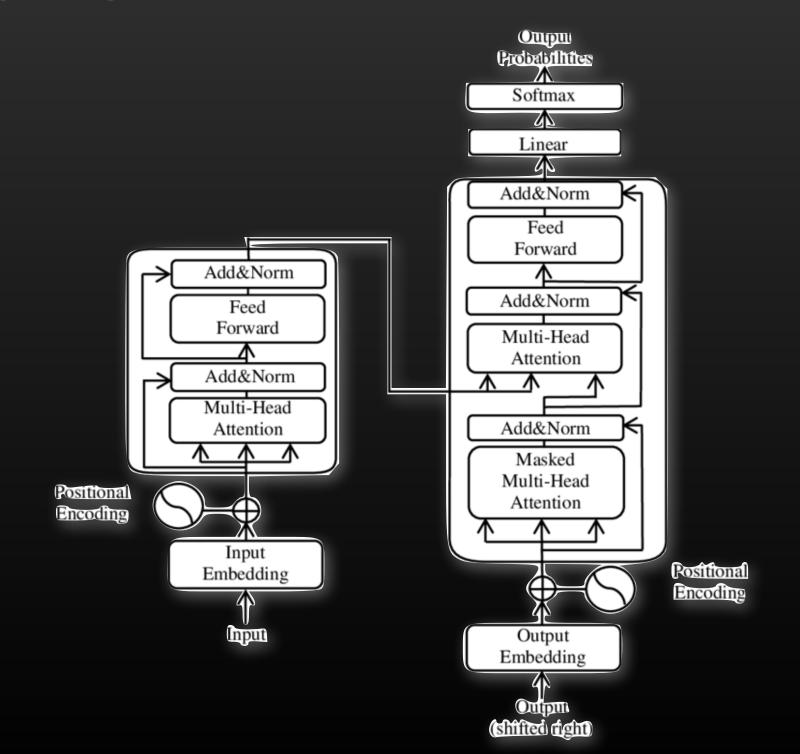
#### **Abstract**

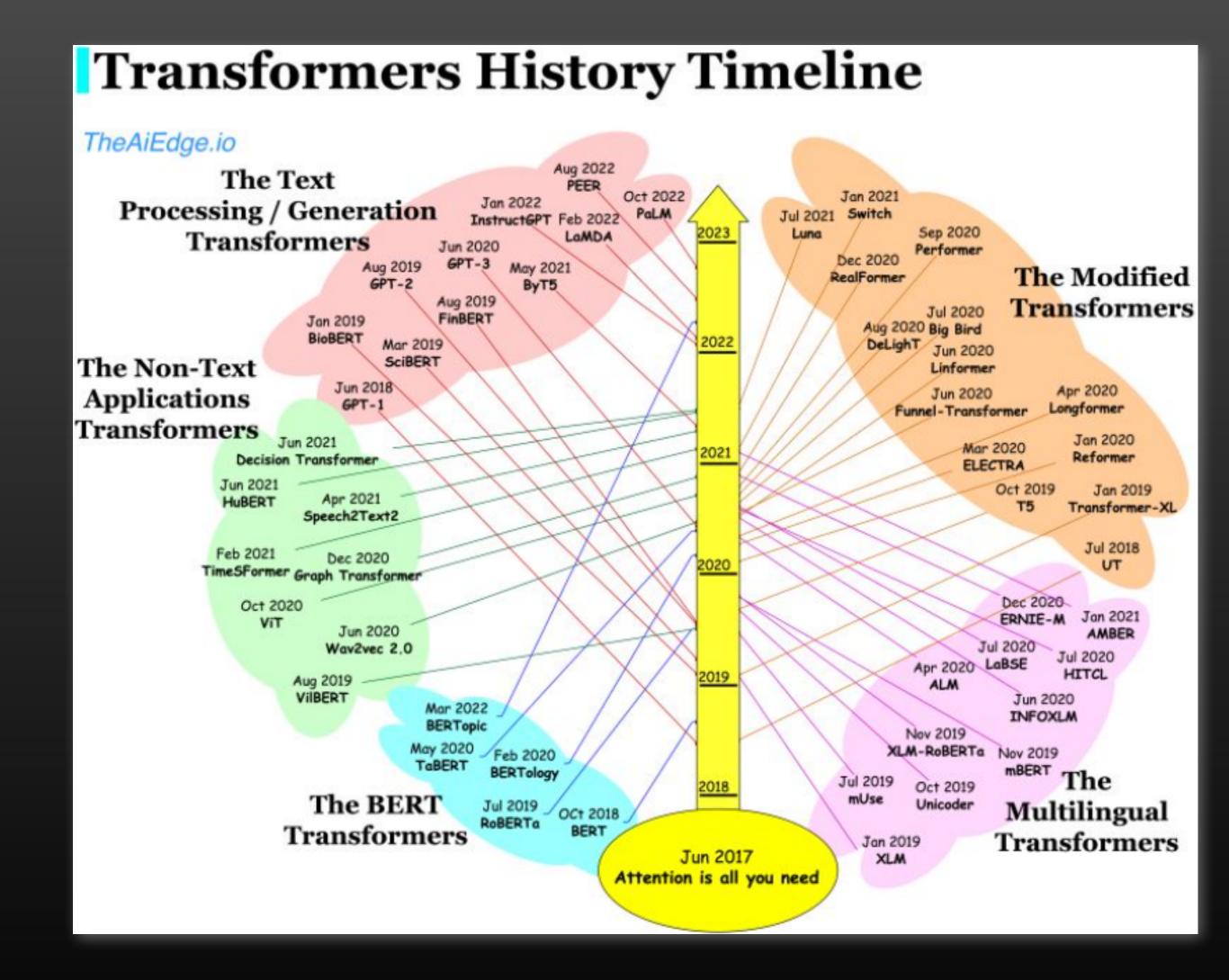
The dominant sequence transduction models are based on complex recurrent or convolutional neural networks that include an encoder and a decoder. The best performing models also connect the encoder and decoder through an attention mechanism. We propose a new simple network architecture, the Transformer, based solely on attention mechanisms, dispensing with recurrence and convolutions entirely. Experiments on two machine translation tasks show these models to be superior in quality while being more parallelizable and requiring significantly less time to train. Our model achieves 28.4 BLEU on the WMT 2014 English-

### Glossary of Terms

#### Transformers

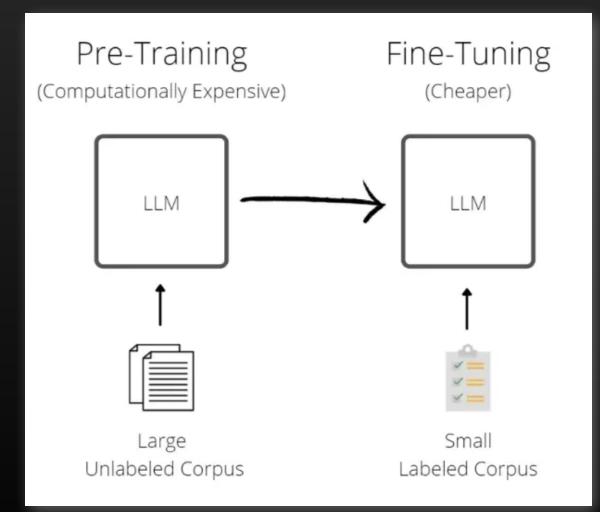
- BERT: Bidirectional Encoder
   Representations from Transformers
- GPT: Generative pre-trained transformer



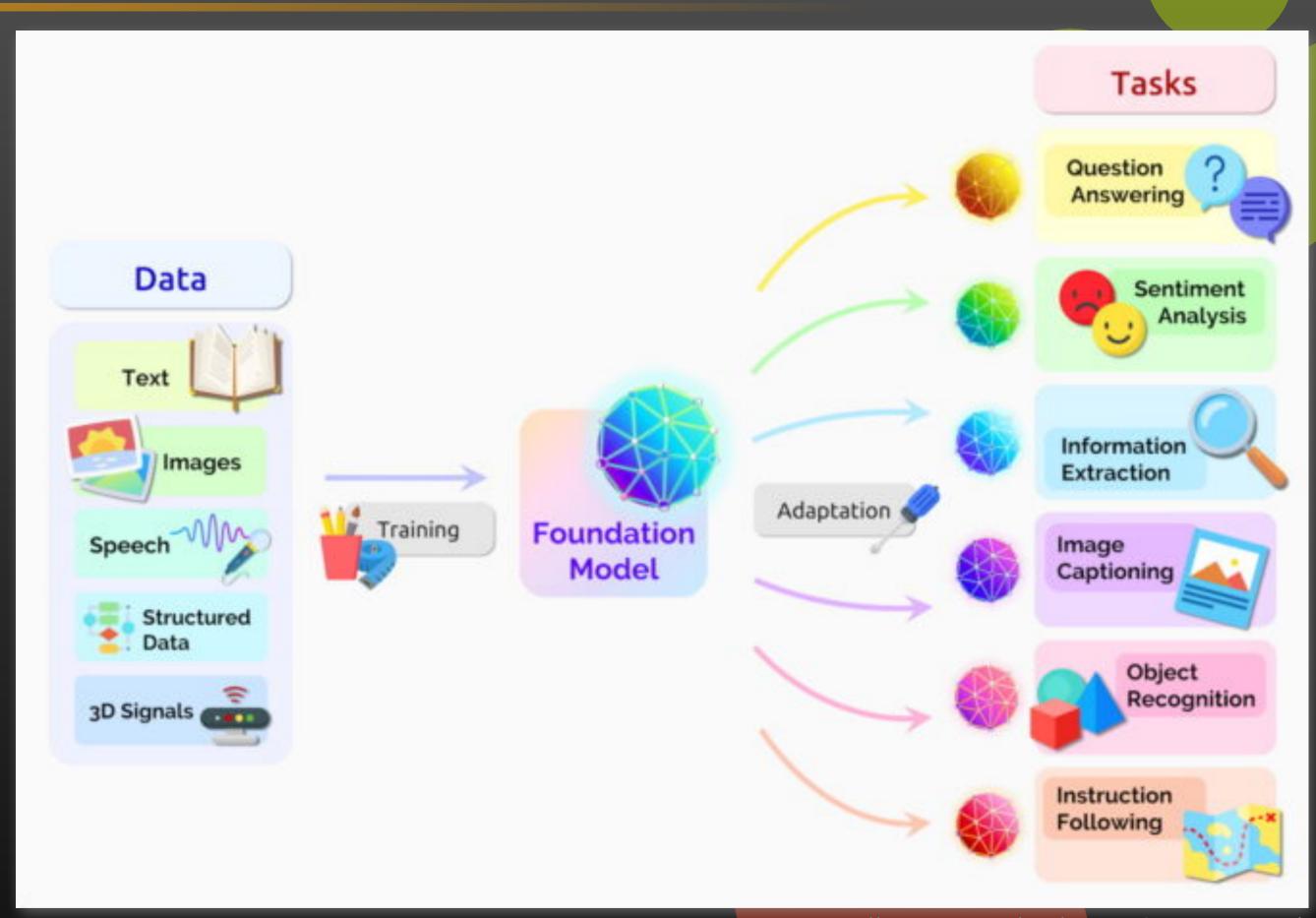


### Glossary of Terms - Models

- Large Language Models
- Base / Foundation Models
- Modalities
- Tasks
- Fine tuning



https://research.aimultiple.com/large-language-models/

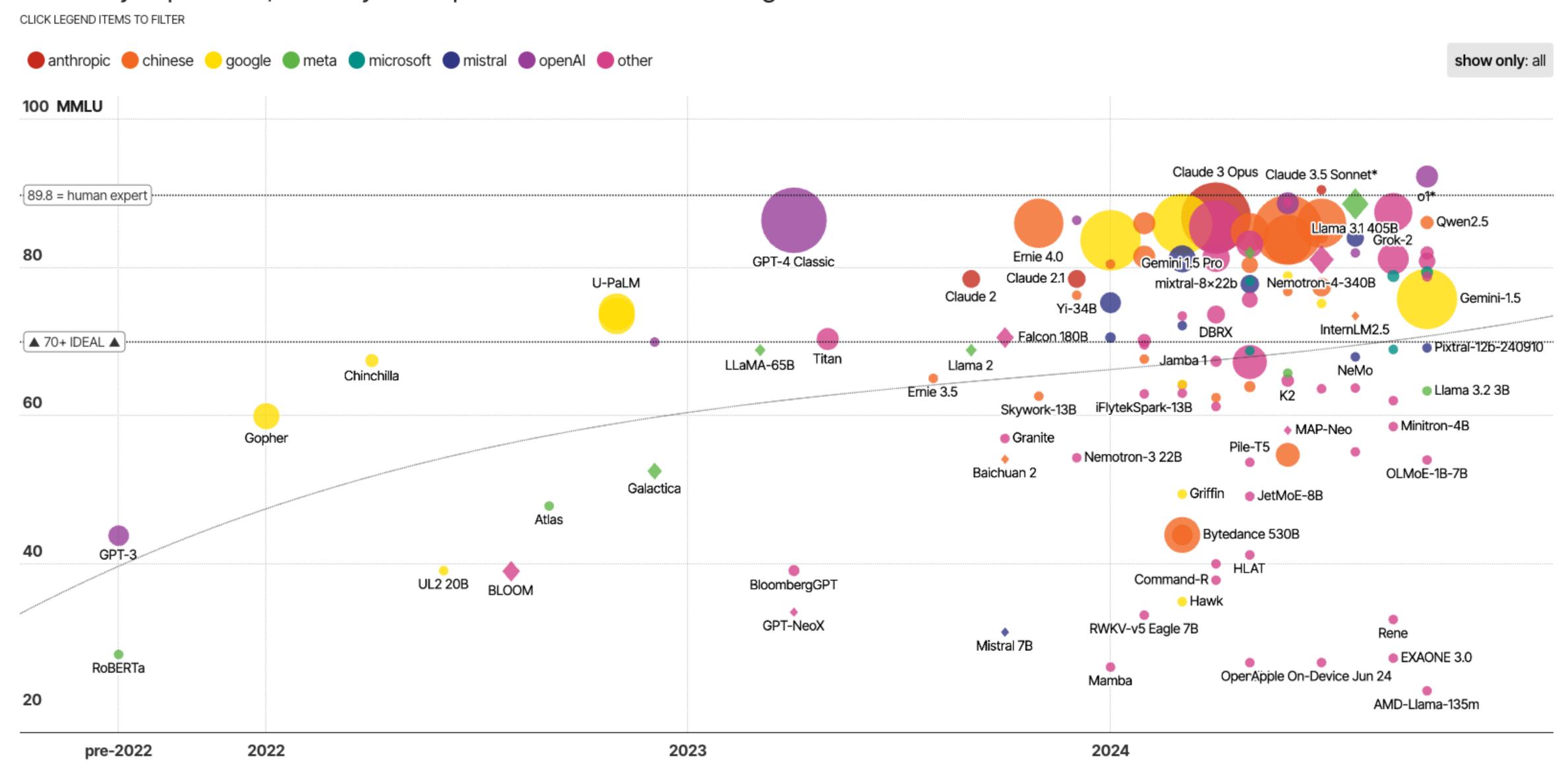


https://blogs.nvidia.com/blog/what-are-foundation-models/

### Major Large Language Models (LLMs)

Parameters (Bn) open access

ranked by capabilities, sized by billion parameters used for training



### What is it Good for: Large Language Model Tasks

- Text summarisation / Simplification
- Sentiment analysis
- Chatbots / Conversational Al
- Classification / Entity recognition
- Semantic Search
- Speech recognition
- Recommendation

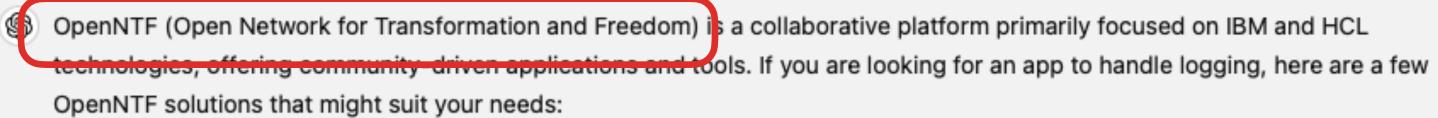
- Text/Image/Audio/Video Generation
- Text-to-speech synthesis
- Spell/Grammar correction
- Translation
- Fraud detection
- Code generation
- Al Agents

### Demo

Suggest an OpenNTF app for logging

### What ???

Suggest an OpenNTF app for logging



#### 1. XPages Log File Reader

#### XPages Log File Reader

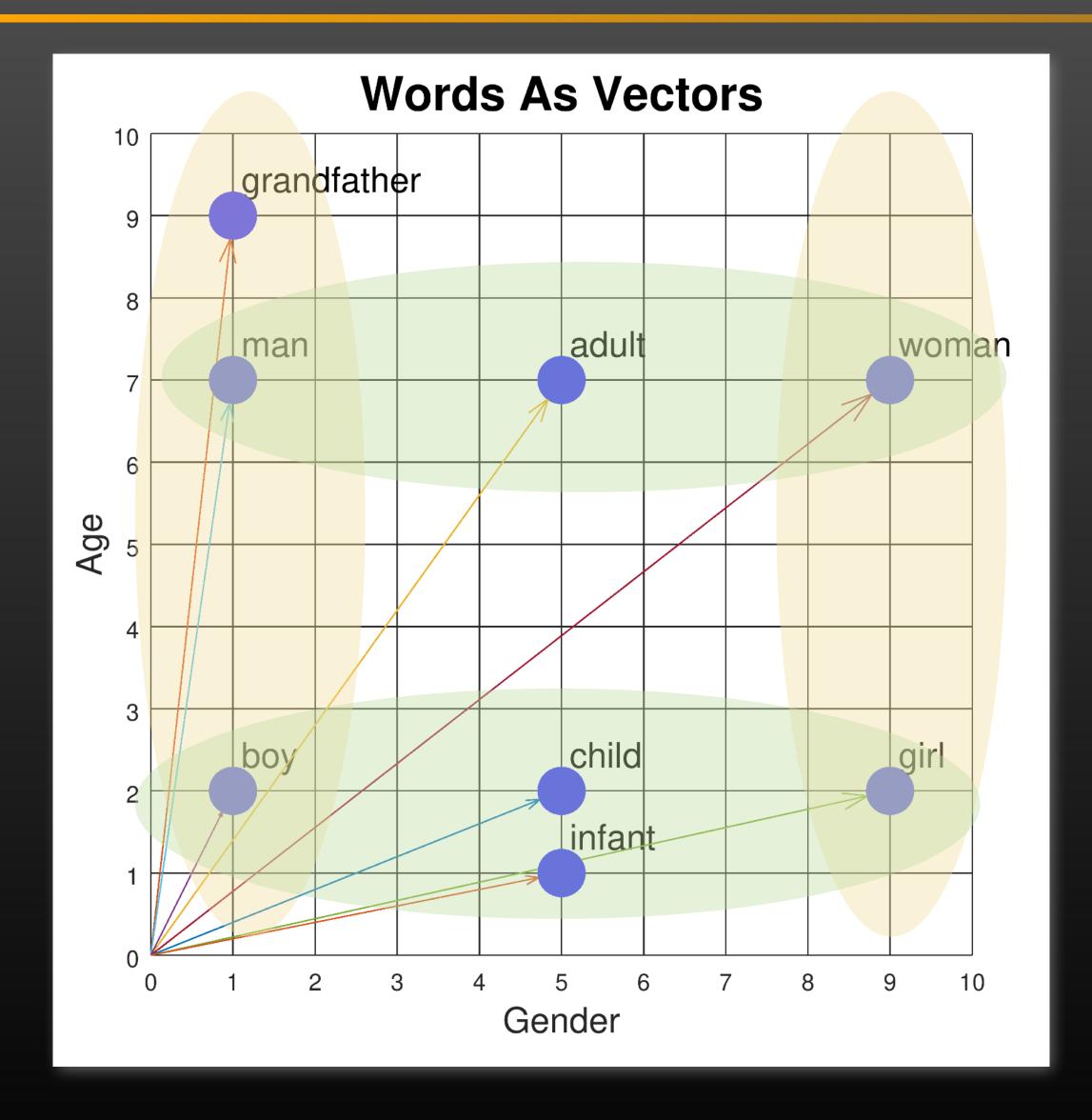
- Purpose: This application allows developers and administrators to read and analyse Domino server log files through an XPages interface.
- Features:
  - · Easy access to log data from Domino servers.
  - · Filter and search logs to identify issues quickly.

### Word Embeddings

Vectors and Vector Search



### Word Embeddings



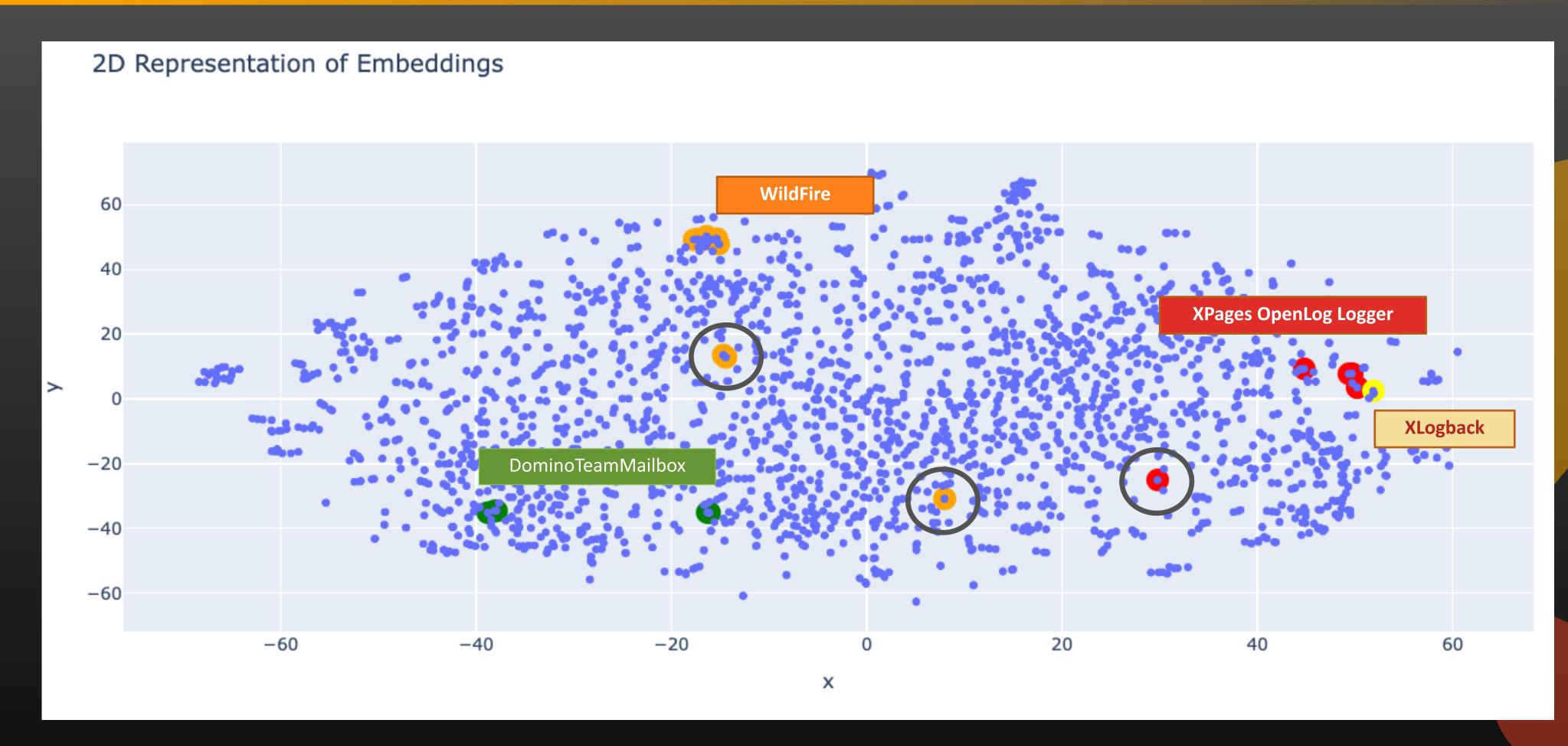
 Vector representation for words in multi-dimensional space

### Word Embeddings - Real Life



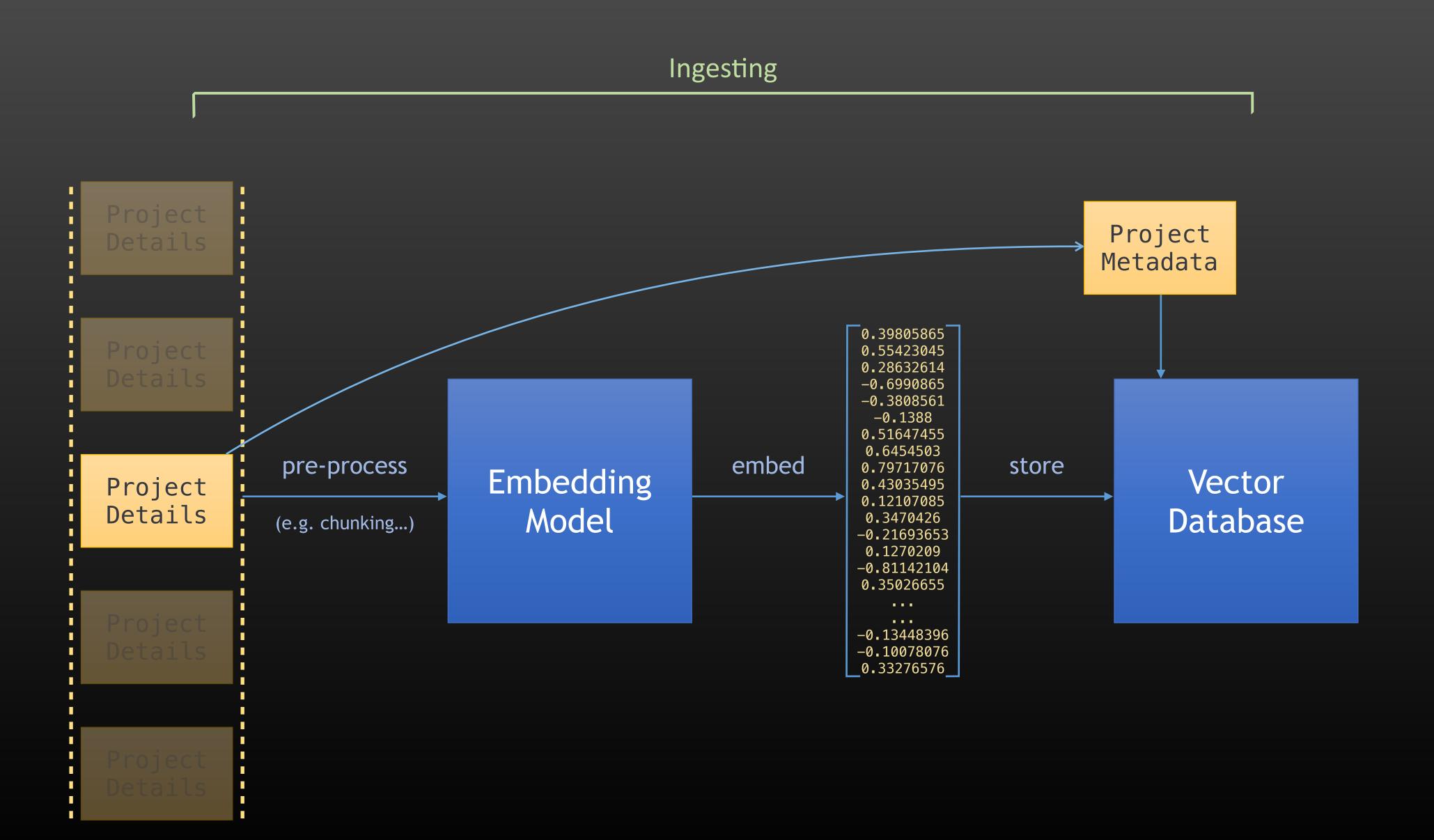
Vector space representation of project embeddings

### Word Embeddings - Real Life

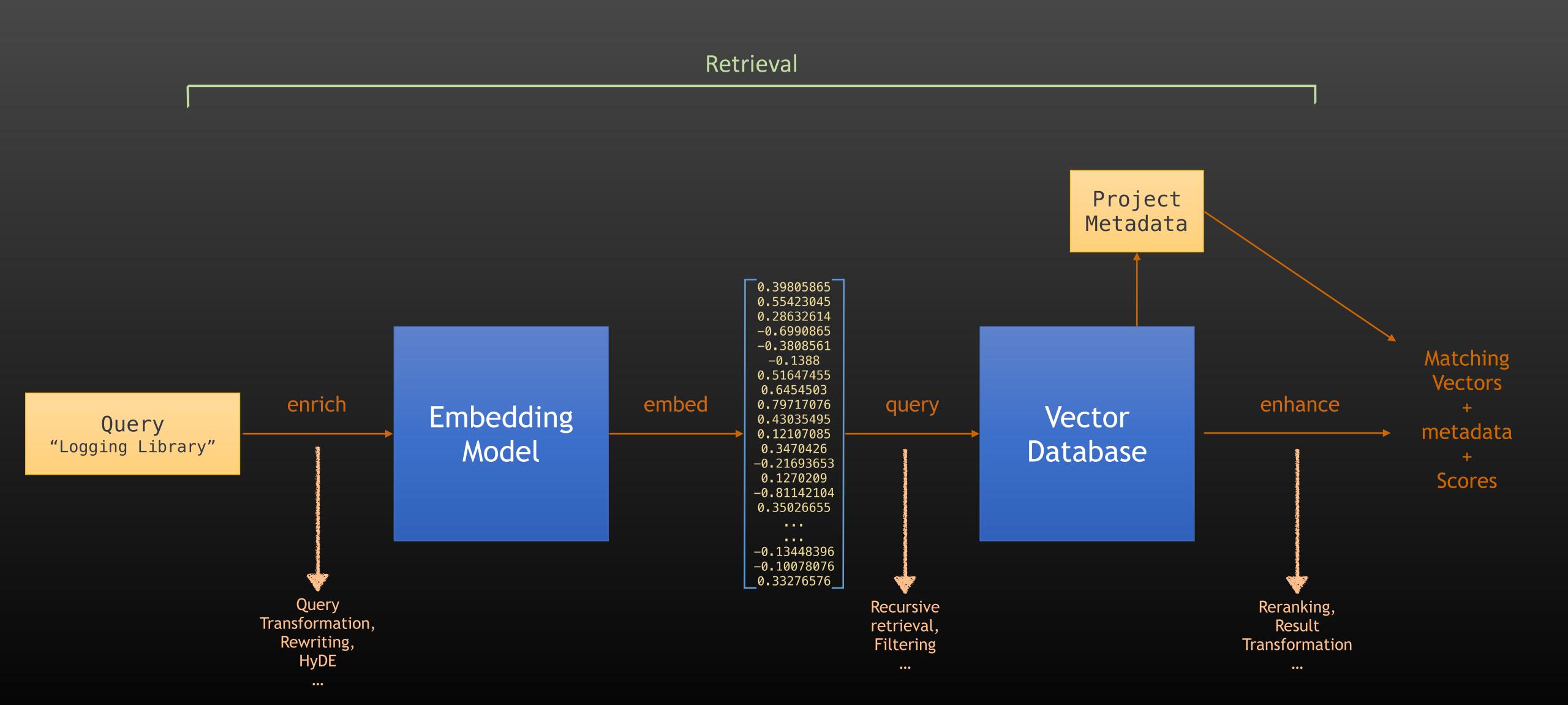


Vector space representation of project embeddings

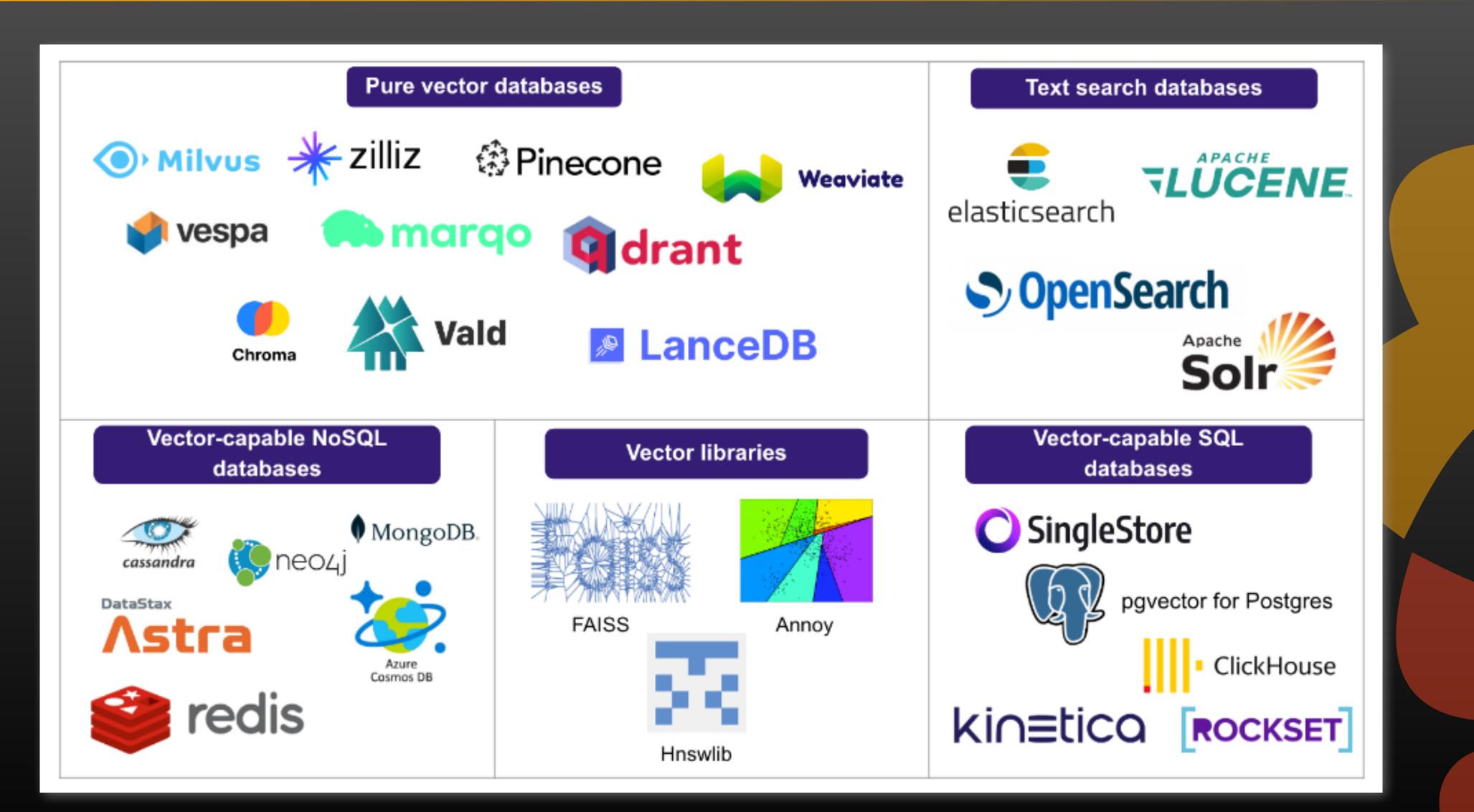
### Building a Vector Store



### Query a Vector Store



### Vector Database



### Picking the Right Model

Finding the right fit for the task



### Model Cards

#### DATA FOCUSED

- Data Sheets • • •
- Data Statements • •
- Data Nutrition Labels • •
- Data Cards for NLP
- Dataset Development Lifecycle Documentation Framework
- Data Cards • •

#### MODELS & METHODS **FOCUSED**

- Model Cards • • •
- Value Cards • •
- Method Cards
- Consumer Labels for Models . . . .

#### SYSTEMS FOCUSED

- System Cards • •
- FactSheets•
- ABOUT ML

#### SAMPLE OF POTENTIAL AUDIENCES

- ML Engineers
   Model Developers/Reviewers
   Students

- Ethicists
- Data Scientists/Business Analysts

- Policymakers
- Impacted Individuals

### Word Embeddings - Models

Large Language Model

### Local Models (e.g. Ollama, Onnx files...)

- ✓ Your data won't leave the server
- ✓ Most are free with permissive licenses
- ✓ No vendor lock-in
- ✓ No cost per operation
- Model files are huge.
- LLM tasks are resource-intensive
- Less capable models
- Programmability restrictions



### Word Embeddings - Models

Large Language Model

### Cloud Models (e.g. OpenAI, Vertex AI, etc.)

- Managed services
- ✓ Pay-per-use model
- √ Easy to use RESTful API and native SDKs
- ✓ Scalable / Available
- ✓ High performance / High quality
- ✓ Much better in complicated tasks
- Privacy and security concerns
- Network latency
- High costs for very busy systems
- Vendor lock-in

### Decide and Test the Model

```
from qdrant_client import QdrantClient
from qdrant_client.models import Distance, VectorParams
client = QdrantClient(url="http://localhost:6333")
def newCollection(collection_name, model, items):
    points_data = []
    for item in items:
        points_data.append({
            "id": points_data.__len__() + 1,
            "vector": model.encode(f"{item['name']}: {item['overview']} {item['content']}"),
            "payload": {
               "unid": item["unid"],
                "name": item["name"],
                "lastUpdated": item["lastUpdated"]
    client.delete_collection(collection_name)
    client.create_collection(
        collection_name=collection_name,
        vectors_config=VectorParams(size=model.get_sentence_embedding_dimension(), distance=Distance.COSINE)
    client.upsert(collection_name=collection_name, points=points_data)
newCollection("col1_mxbai", model_mxbai, data)
newCollection("col1_msmarco", model_msmarco, data)
```

### Suggestion: Learn Python!

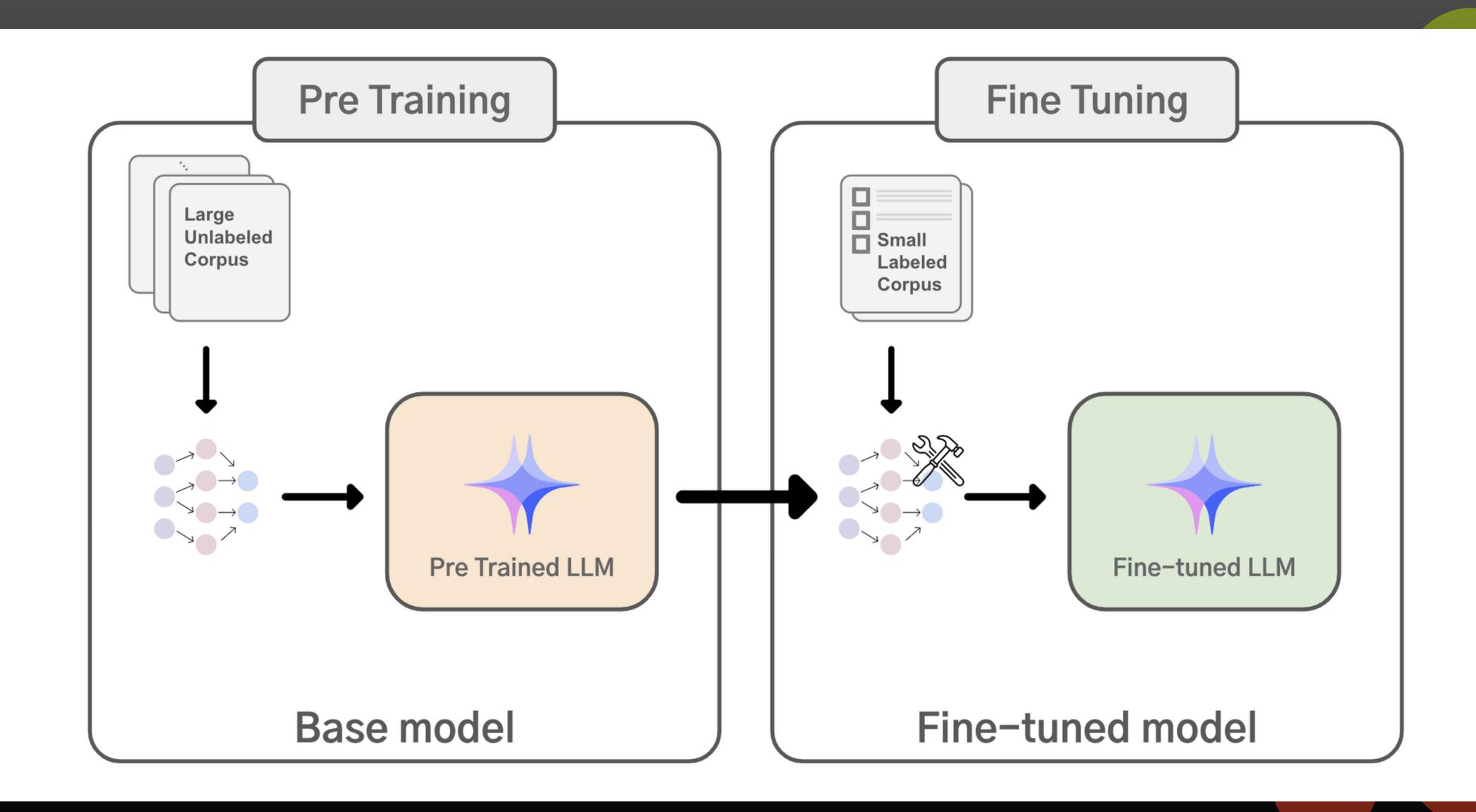
```
from qdrant_client import models
def searchCollection(collection_name, model, search_term):
    result = client.search(
        collection_name=collection_name,
        search_params=models.SearchParams(hnsw_ef=128, exact=False),
        query_vector=model.encode(search_term),
        limit=10,
    print(f"\nModel {collection_name} search for '{search_term}':")
    print(30*"-")
   # Incoming result:
   # ScoredPoint(id=412, version=0, score=0.73846227, payload={ [name/unid/lastupdated] }, vec
    for hit in result:
       print(f"{hit.payload['name']}: {hit.score:.3f}")
    print(30*"-")
    return result
def test_models(search_term):
    searchCollection("col1_mxbai", model_mxbai, search_term)
    searchCollection("col1_msmarco", model_msmarco, search_term)
test_models("logging library")
```

### Improving Models

Tweaking the Brain

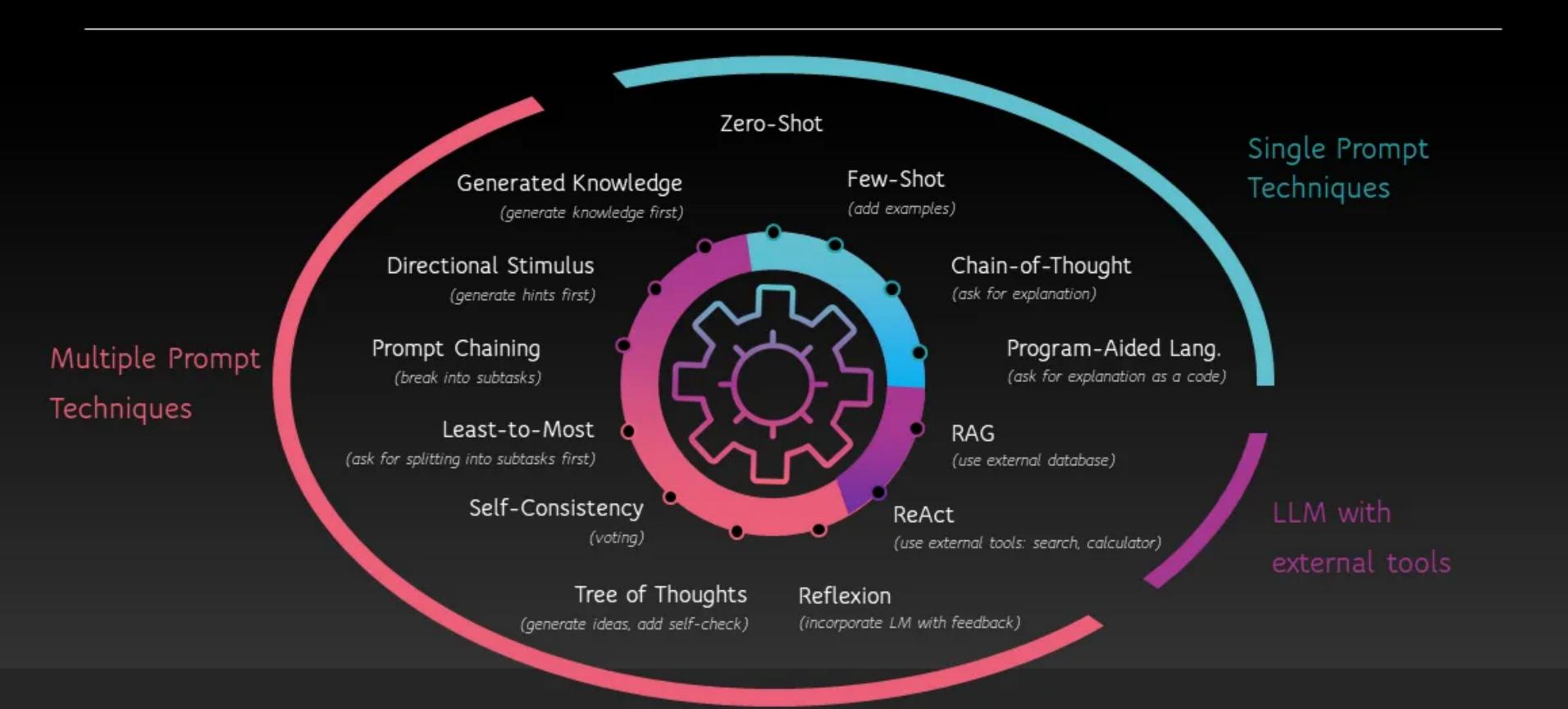


### Increase Knowledge: Fine Tune (Transfer Learning)



### Improve Behavior: Prompt Engineering

### Prompt Engineering Techniques

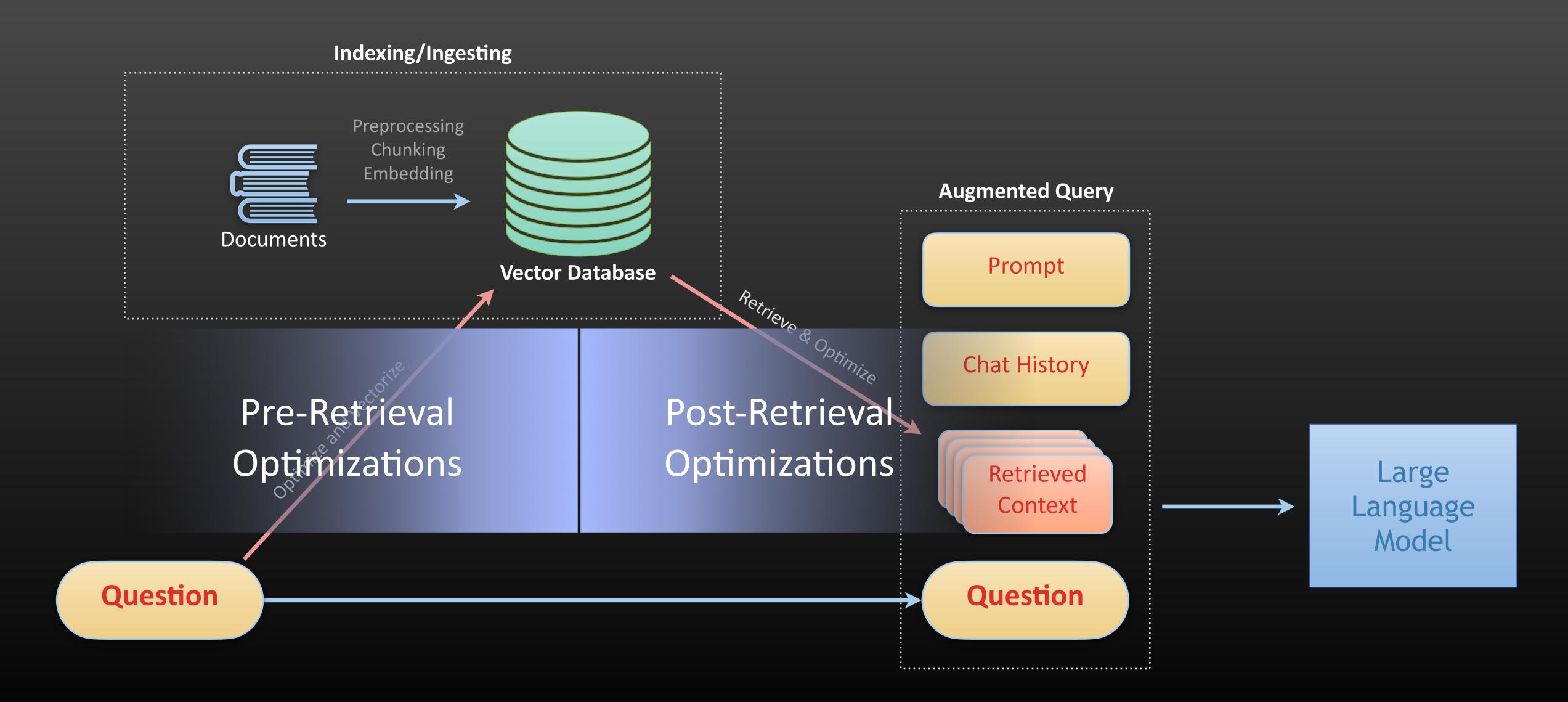


### Improve Prompts: Retrieval-augmented generation

#### Scenario

- Domain Knowledge in documents, databases, etc.
- LLM to respond questions aligned with domain knowledge

### Improve Prompts: Retrieval-augmented generation



### Demo

Prompts and Chat



## Working with LLMs for Domino Apps

LLM Integration is a simple REST API integration

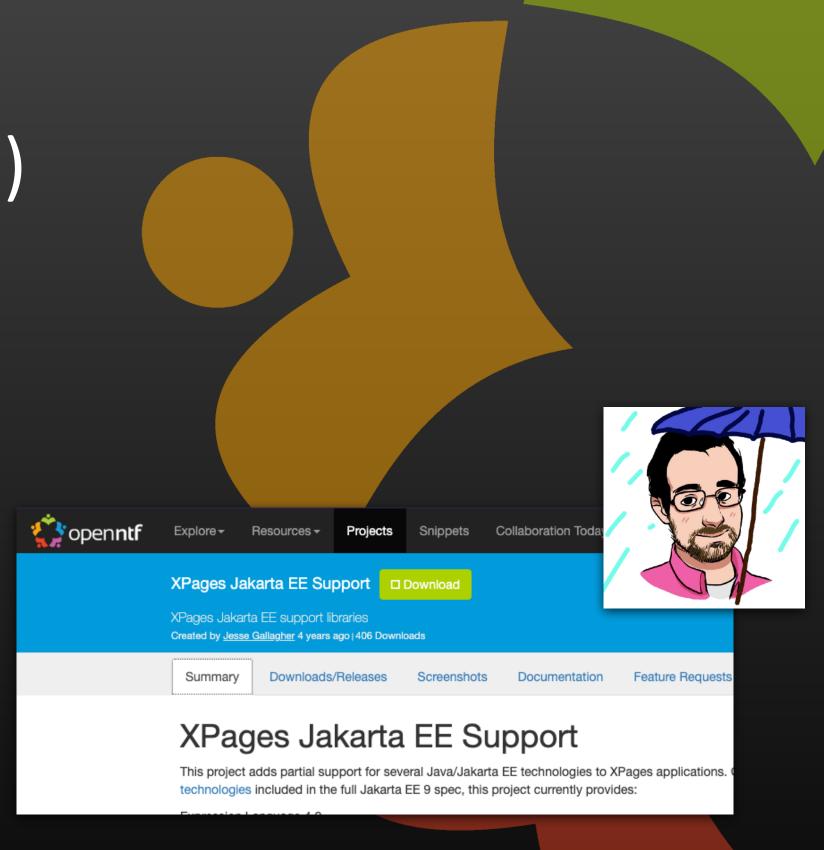


### Access LLMs using Java in Domino

- XPages
- OSGi Plugins
- RESTful API (OpenNTF JakartaEE project by Jesse)
- Java Agents (Notes Client or Server side)
- DOTS
- Java Addin









### For Java Developers

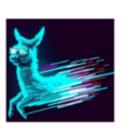
#### Current LangChain4j Integrations











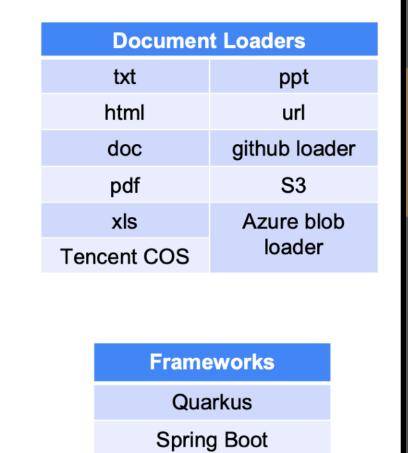
LLM Integrations		
Amazon Bedrock	Google Vertex AI PaLM 2	
Azure OpenAl	HuggingFace	
ChatGLM	LocalAl	
DashScope	Ollama	
Google Vertex Al Gemini	OpenAl	
	Mistral	



Current LangChain4j Integrations			
	Embedding Stores		
	Chroma	Astra DB	
	Elasticsearch	Cassandra	
	Milvus	Neo4j	
	Pinecone	OpenSearch	
	Vespa	PGVector	
	Weaviate	MongoDB	
	Redis	Qdrant	
	Code Execution Engines		

GraalVM Polyglot/Truffle

Judge0





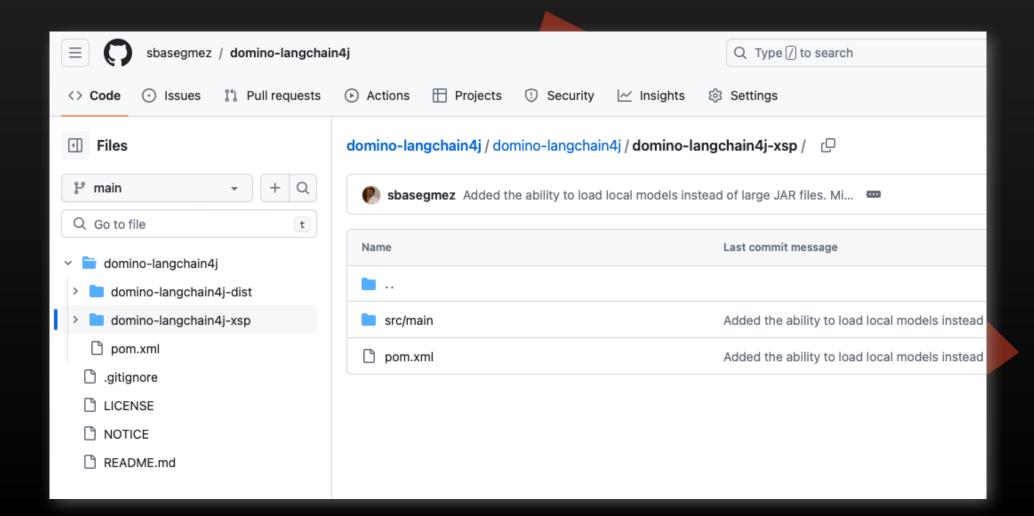
LangChain4j is very promising

### A New Project: Domino-LangChain4j

- Experimental phase
- Import langchain4j library into Domino
  - Utilise ChatModel w/ Local or Cloud LLM
  - Embedding
  - RAG
- Server and Designer plugins
- Add some utilities
  - Local Model Support
  - Managed beans
  - Configuration / Logging
  - RAG document loaders for Domino
- Looking into Java Agent and DOTS support
- Feedbacks are welcome!

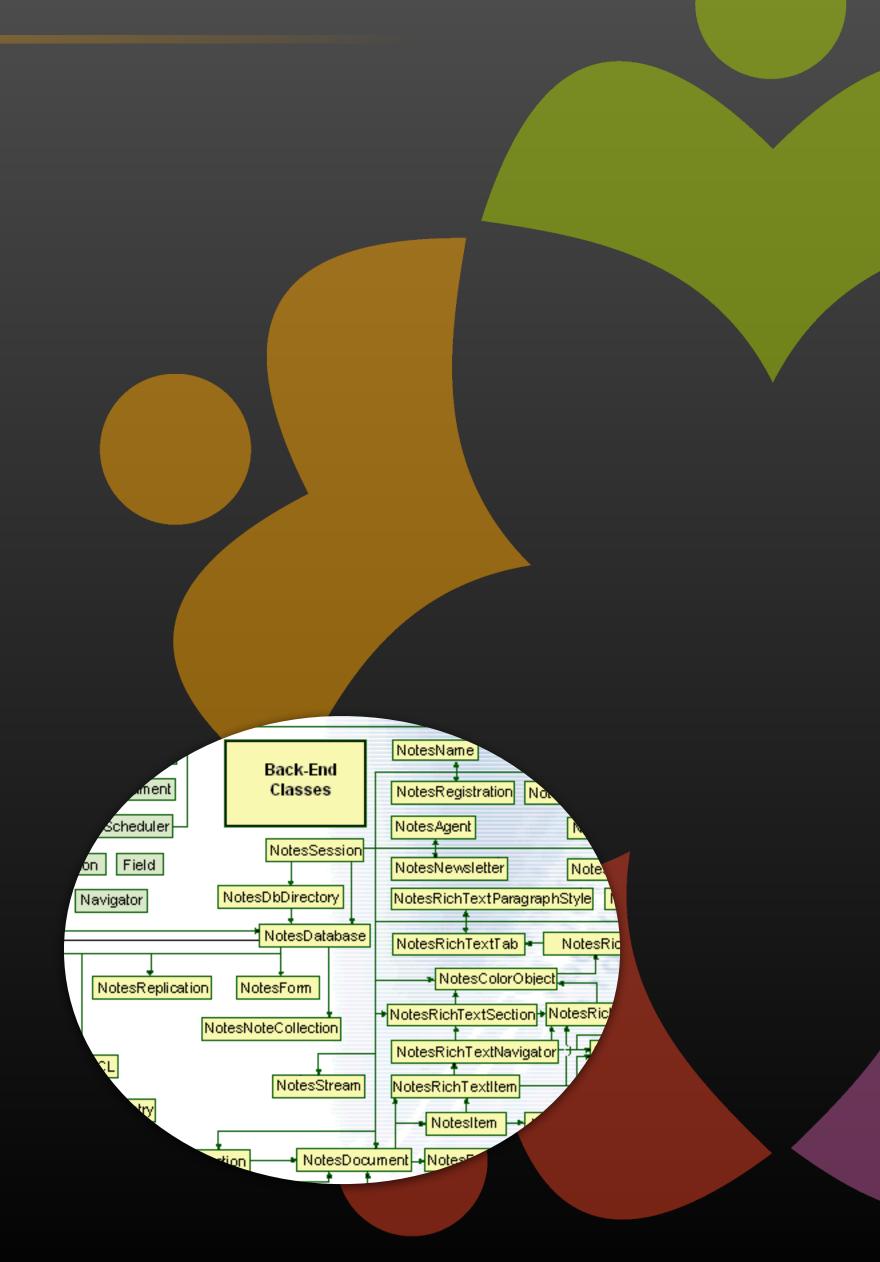






### Access LLMs using LotusScript in Domino

- For LotusScript, there are still options.
  - Use RESTful access using LotusScript
- Use Java Agent
  - LLM integration might be done with Java agents.
     LotusScript can call agents



### Other LLM Projects

- HCL Domino IQ (Future Product)
  - Uses Llama.cpp
  - Integrated to the server
- Open LLM Integrator on OpenNTF
  - Ollama integration with RAG and QDrant support
  - By Erik Schmalz
- ChatGPT APIs for Domino on OpenNTF
  - Credits: Ayhan Sahin & Christian Sadeghi



**Openntf** 

Summary

models offered.

Downloads/Releases

Open LLM Integrator

Screenshots

In order to use this application, you need to install Ollama Server application and select and set up the appropriate



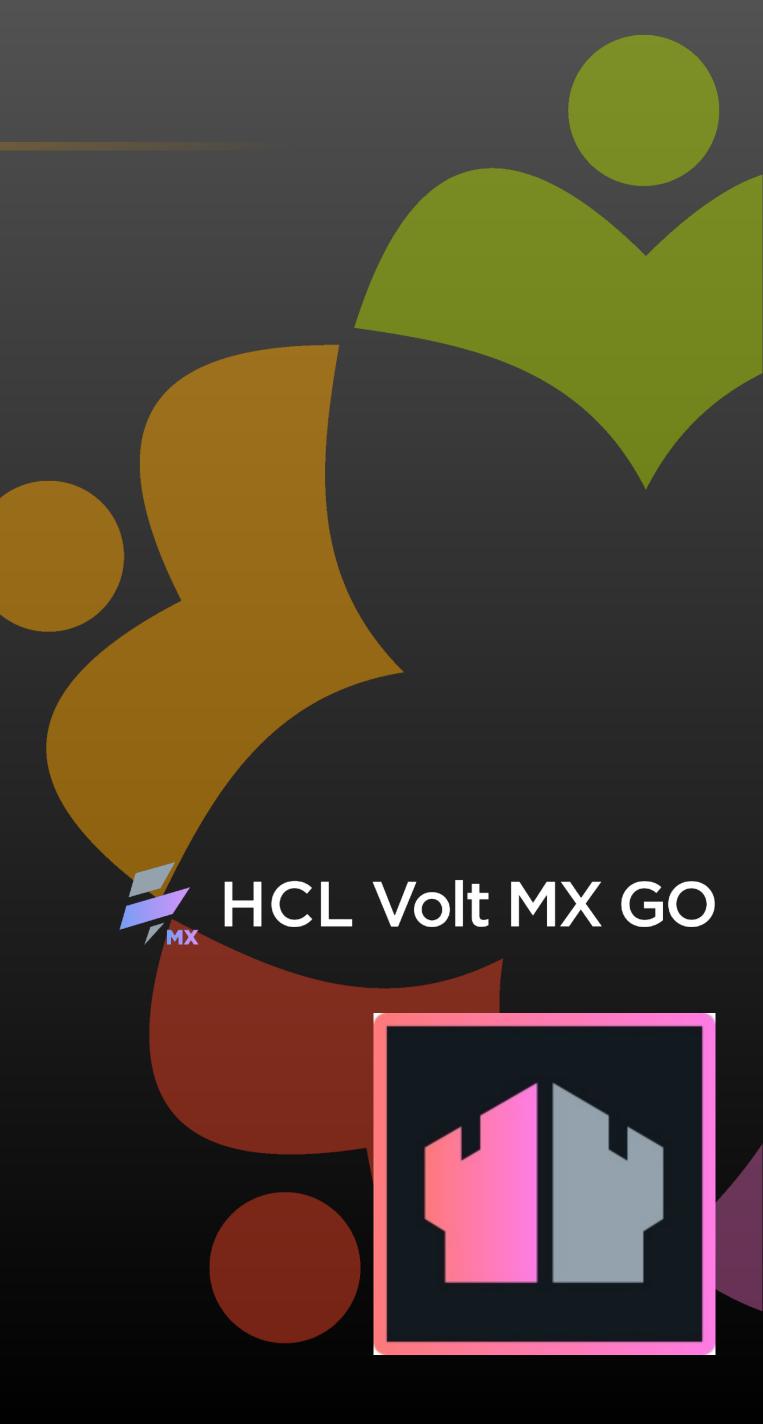
Documentation

Feature Requests

Defects

### Integration Outside of the Domino Server

- Implement LLM logic in your favorite platform
  - Volt MX
  - Python
  - Java
  - JavaScript
  - • •
- Access to Domino Data
  - Using Domino REST API
  - Implement your own services with the OpenNTF Jakarta EE Project



### Topics for Another Day...

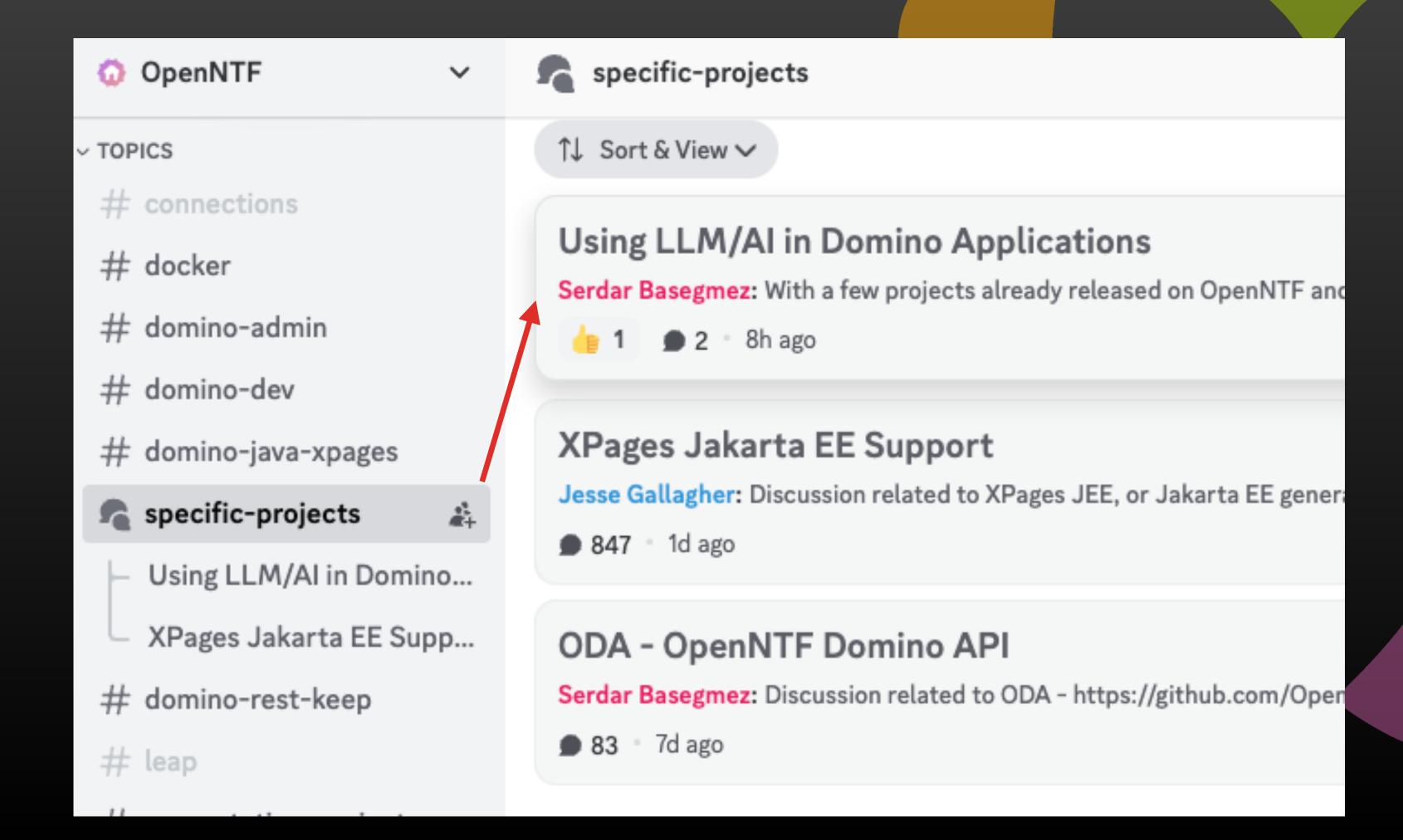
- Models deep-dive
- Prompt Engineering
- Development Methodology
  - Prototyping, validation, optimization, testing, lifecycle
- Safety and Security
  - Guardrails, moderation
  - Prompt Injections
  - Regular Compliance Audits
  - Al Accountability



### Feedbacks and Discussions

- OpenNTF Discord Server
  - Specific Projects —> Using LLM/AI in Domino Applications





### Resources

- All the demo materials:
  - https://github.com/sbasegmez/LLM-Demos

- OpenNTF Projects Metadata:
  - TBA

- Domino-Langchain4j experimental version:
  - https://github.com/sbasegmez/domino-langchain4j



### More Good Stuff: Odds and Ends

- Further reading...
  - Huggingface blogs
  - RAG Retrieval Augmented Generation
  - Multimodal approaches
  - Prompt Engineering
- Courses, guides
  - Quick Start Guide to Large Language Models (LLMs)
     Course by Sinan Ozdemir
  - Large Language Models: Application through Production Databricks
  - Large Language Model Ebooks
     NVidia

