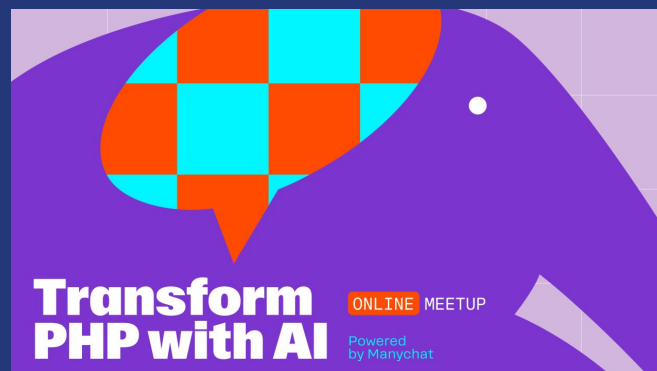




Build a Retrieval-Augmented Generation (RAG) in PHP

Enrico Zimuel, *Tech Lead & Principal Software Engineer*



Nov 27, 2024 - Barcelona

Agenda

- Large Language Model (LLM)
- Transformers architecture
- Top-k and temperature
- Retrieval Augmented Generation (RAG)
- Embedding and Vector Search
- LLPhant for PHP
- Llama 3.2
- Elasticsearch
- Demo

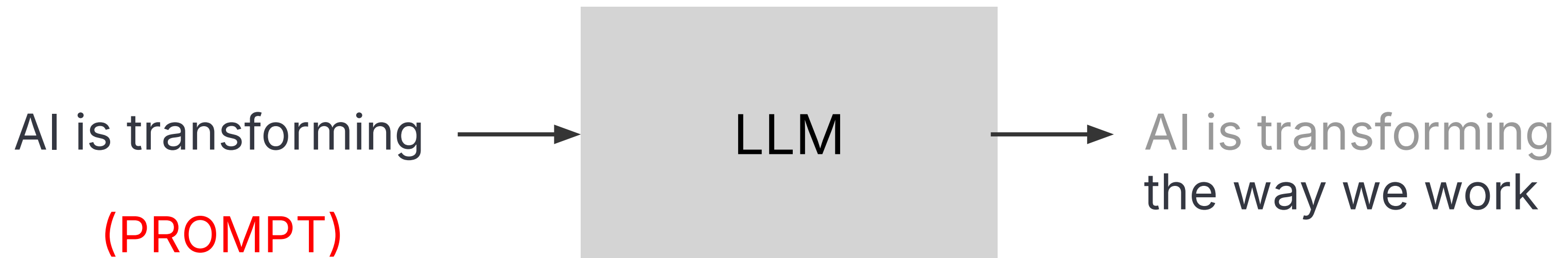


Image generated using dall-e-3

Large Language Model (LLM)

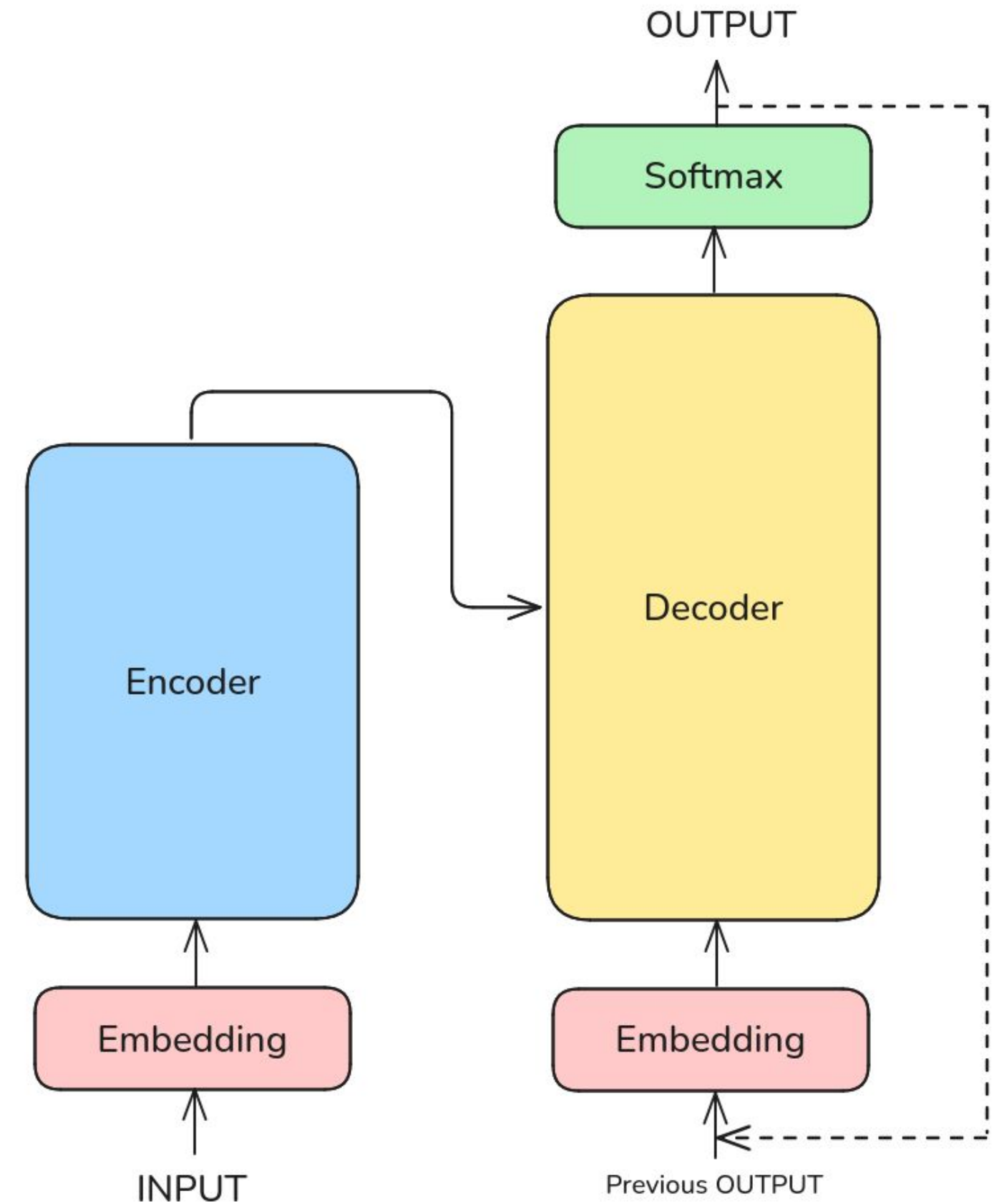
LLM

- **Large Language Model (LLM)** are probabilistic models that produce sentence in natural language
- These models work by completing sentences



Transformer architecture

- Introduced in [Attention is All You Need](#) paper in 2017
- Basement of all LLMs
- The sentences are analyzed using a **self-attention** mechanism: each part of a sentence is evaluated in relation to every other part to understand contextual relationships and assign appropriate weights

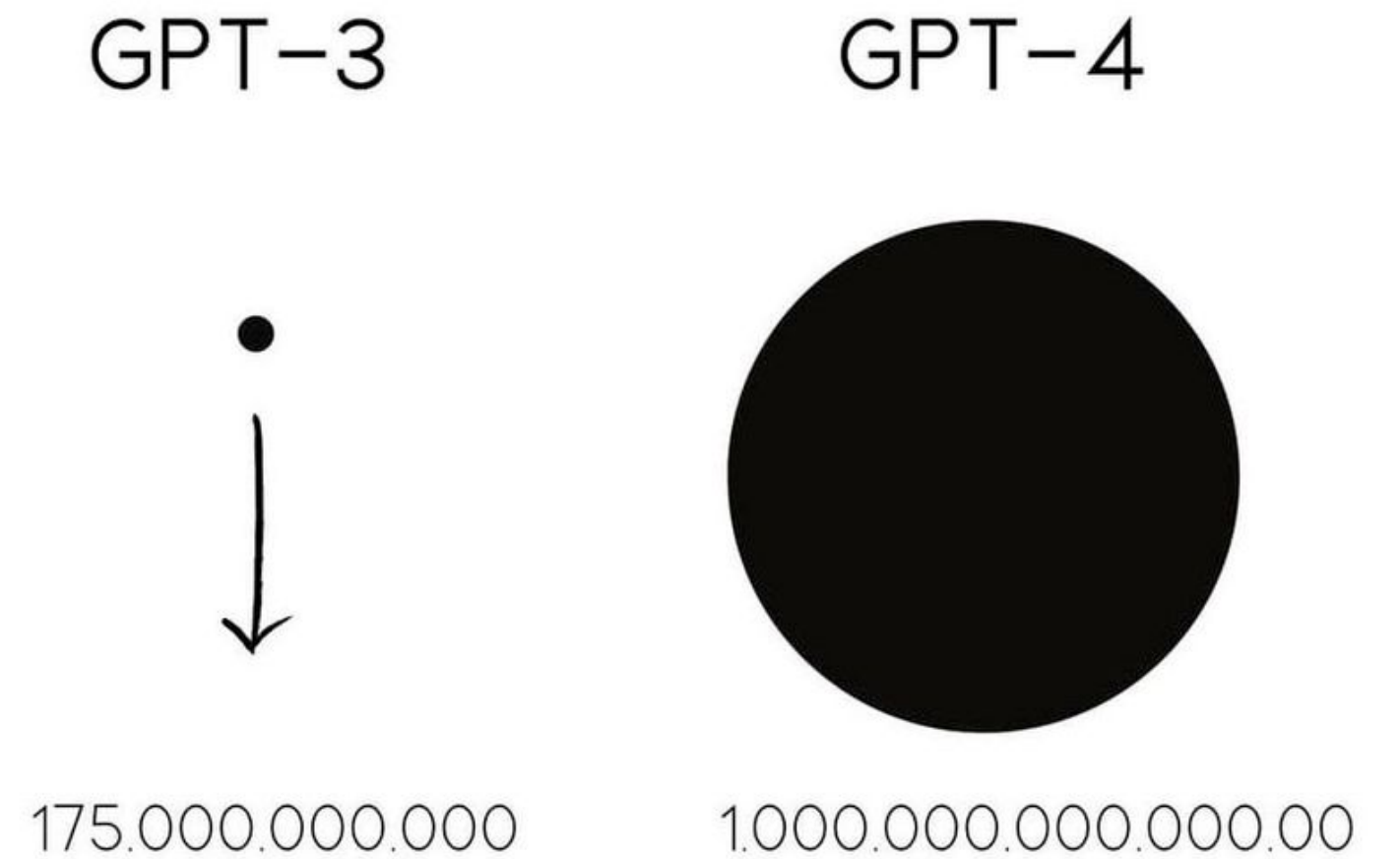


LLM

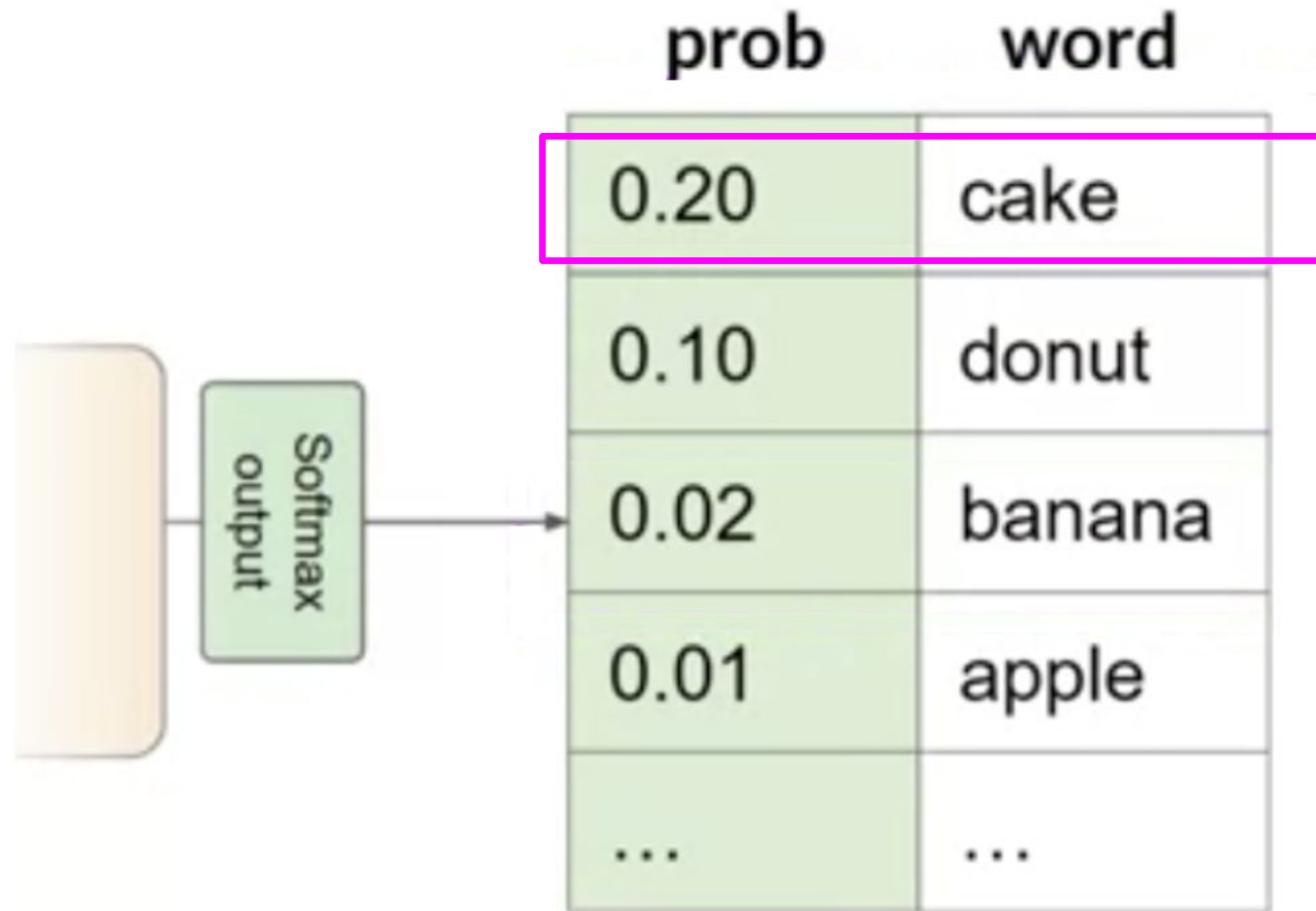
- **Large Language Model (LLM)** consisting of a neural network with many parameters (typically billions of weights or more), trained on large quantities of unlabelled text using self-supervised learning
- A message is splitted in **tokens**
- Each token is translated in a number using an operation called **embeddings**
- LLM **repeatedly predicting** the next token

Size of GPT-4

- Around **1.76 trillion** parameters
- Neural network with **120** layers
- Process up to **25,000** words at once
- Estimated training cost is \$200M using 10,000 [Nvidia A100 GPU](#) for 11 months

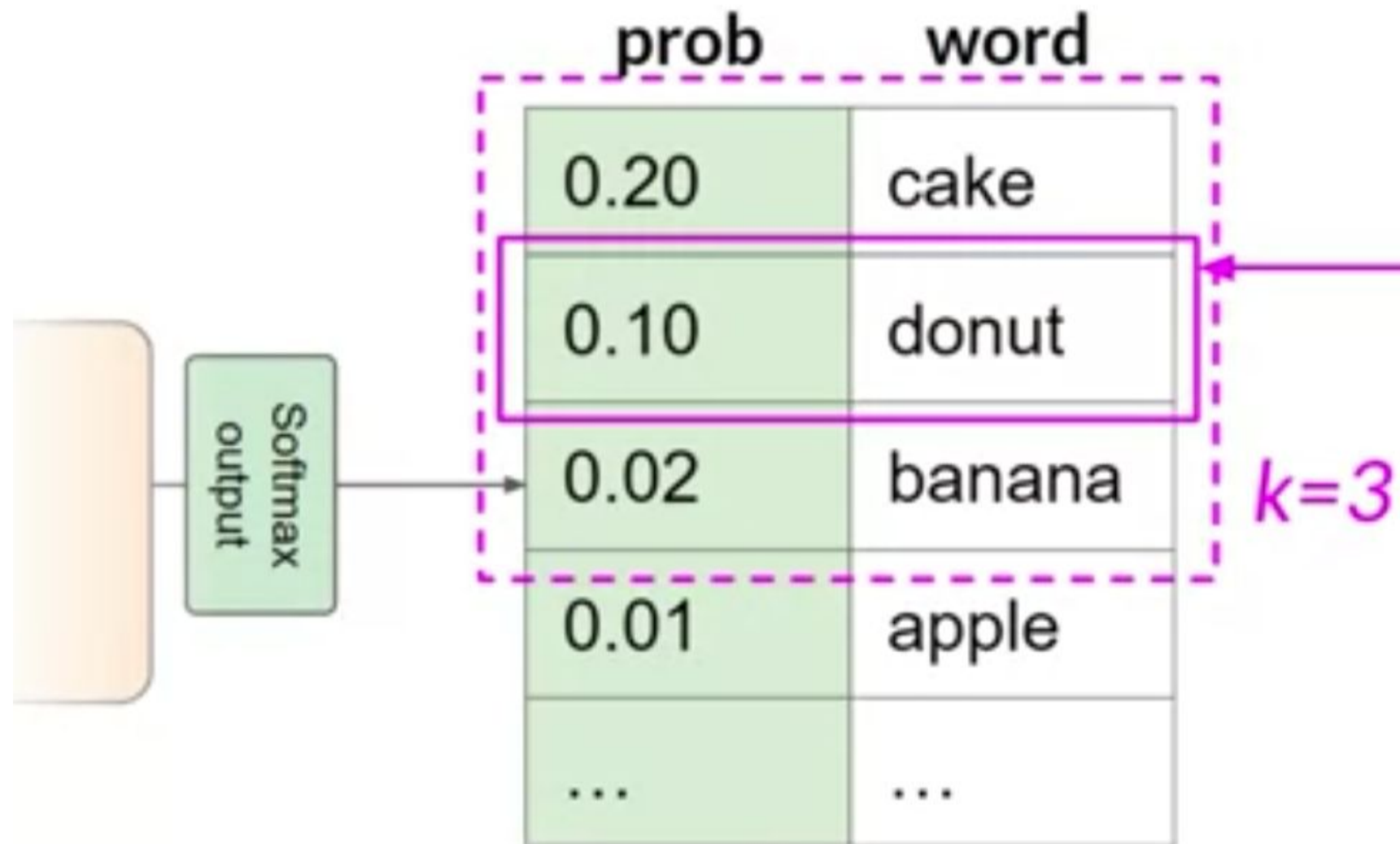


Predict the next word



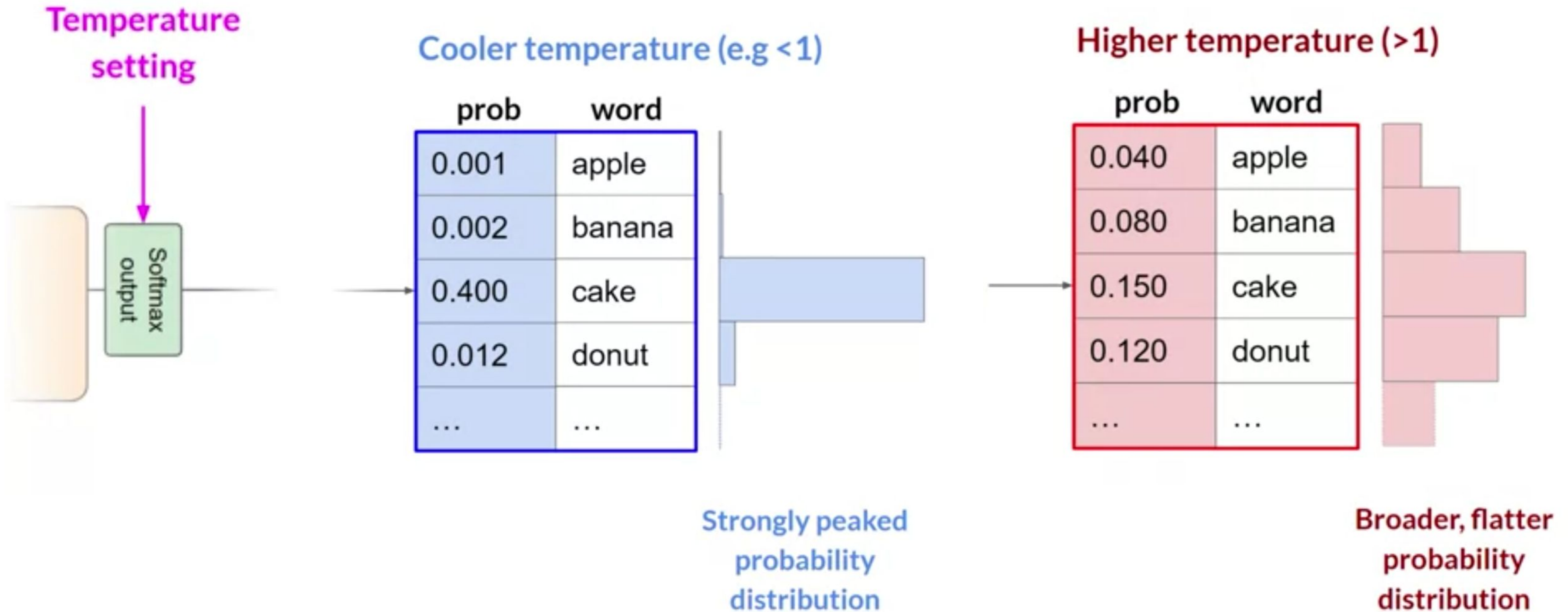
Choose the one with greatest probability (greedy algorithm)

Top-k



top-k: select an output from the top-k results after applying random-weighted strategy using the probabilities

Temperature



Prompt engineering

- You can encounter situations where the model doesn't produce the outcome that you want on the first try
- You may have to revisit the language several times to get a good answer
- The development and improvement of the prompt is known as **prompt engineering**
- One powerful strategy is to include examples of the task that you want the model to carry out inside the prompt
- This is called **In-Context Learning (ICL)**

ICL - zero shot inference



ICL - one shot inference

Prompt

Classify this review:

I loved this movie!

Sentiment:

Positive

Classify this review:

I don't like this chair.

Sentiment:



Completion

Classify this review:

I loved this movie!

Sentiment:

Positive

Classify this review:

I don't like this chair.

Sentiment:

Negative

ICL - few shot inference

Prompt

Classify this review:
I loved this movie!
Sentiment:
Positive
Classify this review:
I don't like this chair.
Sentiment:
Negative
Classify this review:
This is not great.
Sentiment:

LLM

Completion

Classify this review:
I loved this movie!
Sentiment:
Positive
Classify this review:
I don't like this chair.
Sentiment:
Negative
Classify this review:
This is not great.
Sentiment:
Negative

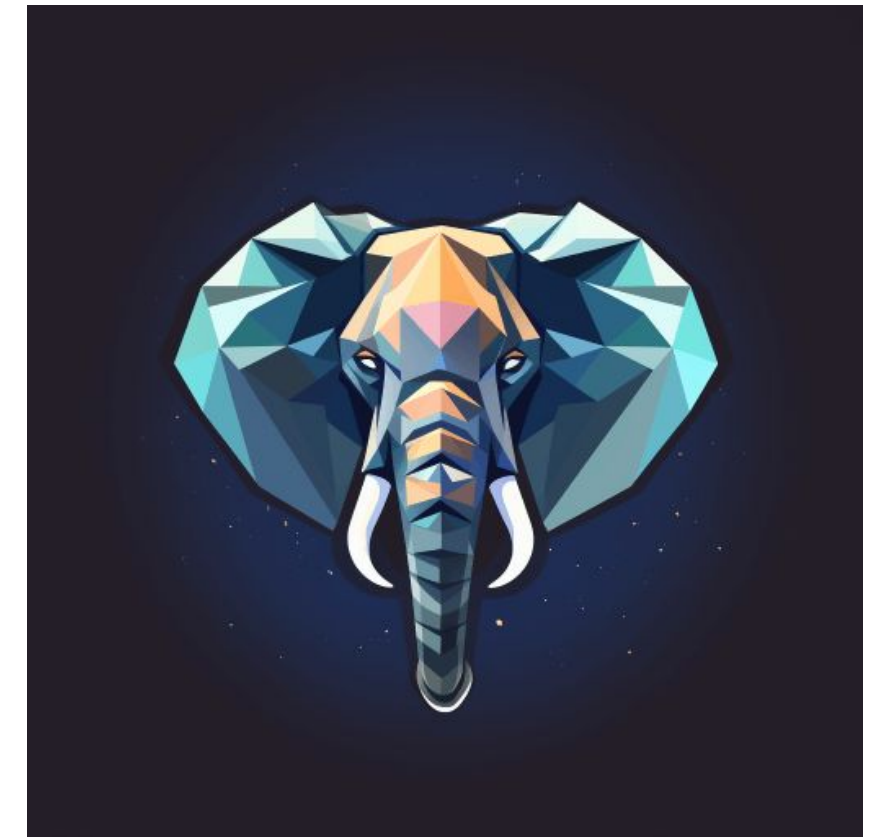
Ollama

- [Ollama](#) is a software for downloading and running LLMs locally
- Llama 3, Phi 3, Mistral, Gemma, and [other models](#)
- Simple command line tool:
 - `ollama pull llama3.2:3b`
 - `ollama run llama3.2:3b`



LLPhant

- [LLPhant](#) is a comprehensive open source Generative AI framework for PHP
- The goal is to offer an easy to use library to build GenAI applications in PHP
- LLM supported: OpenAI, Anthropic, Ollama, Mistral
- Vector databases: Elasticsearch, File, Memory, Milvus, Qdrant, Redis, Milvus, Chroma, etc
- Started by [Maxime Thoonsen](#)



Example: LLPhant with Llama3.2

```
use LLPhant\Chat\OllamaChat;  
use LLPhant\OllamaConfig;  
  
$config = new OllamaConfig();  
$config->model = 'llama3.2';  
$chat = new OllamaChat($config);  
  
$response = $chat->generateText('What is the capital of Italy?');  
// The capital city of Italy is Rome  
printf("%s\n", $response);
```

Retrieval-Augmented Generation (RAG)

Retrieval-Augmented Generation (RAG)

- **RAG** is a technique in natural language processing that combines information retrieval systems with **Large Language Models** (LLM) to generate more informed and accurate responses
- It is composed by the following parts:
 - **Retrieval-Augmented**
 - **Generation**

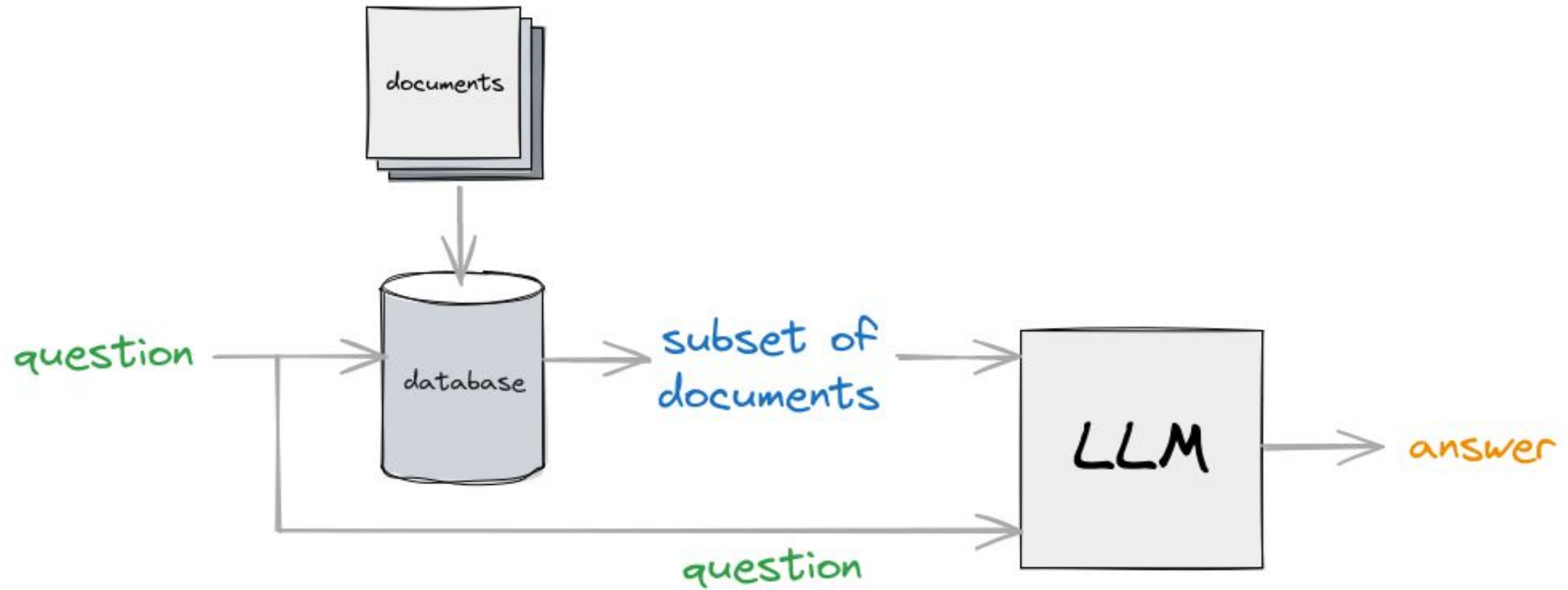
Generation

- LLMs are very powerful but have some limitations:
 - **No source** (potential hallucinations)
 - How can I verify the information coming from an LLM?
 - What sources has been used to generate the answer?
 - **Out of date**
 - An LLM is trained in a period of time
 - For update we need to retraining the model (very expensive)

Retrieval-Augmented

- We collect sets of private or public document
- We build a **retrieval system** (e.g. a database) to extract a subset of documents using a **question**
- Then we pass the **question + documents found** to an LLM as prompt with a context
- The LLM can give an answer using the updated documents

RAG architecture

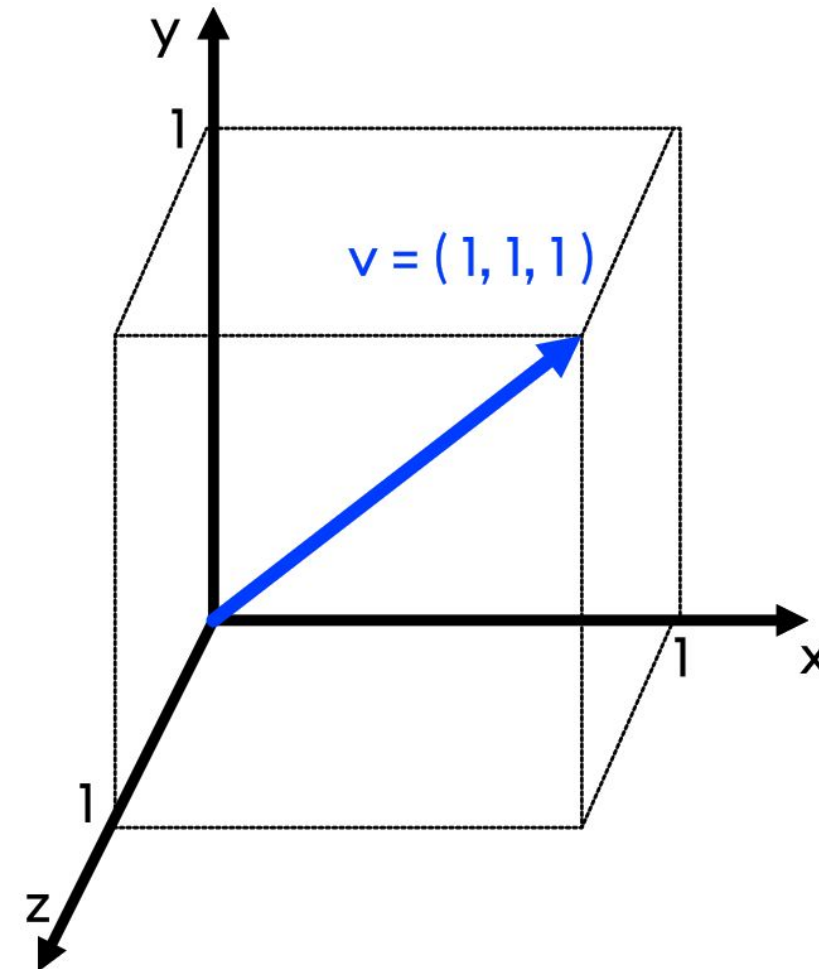
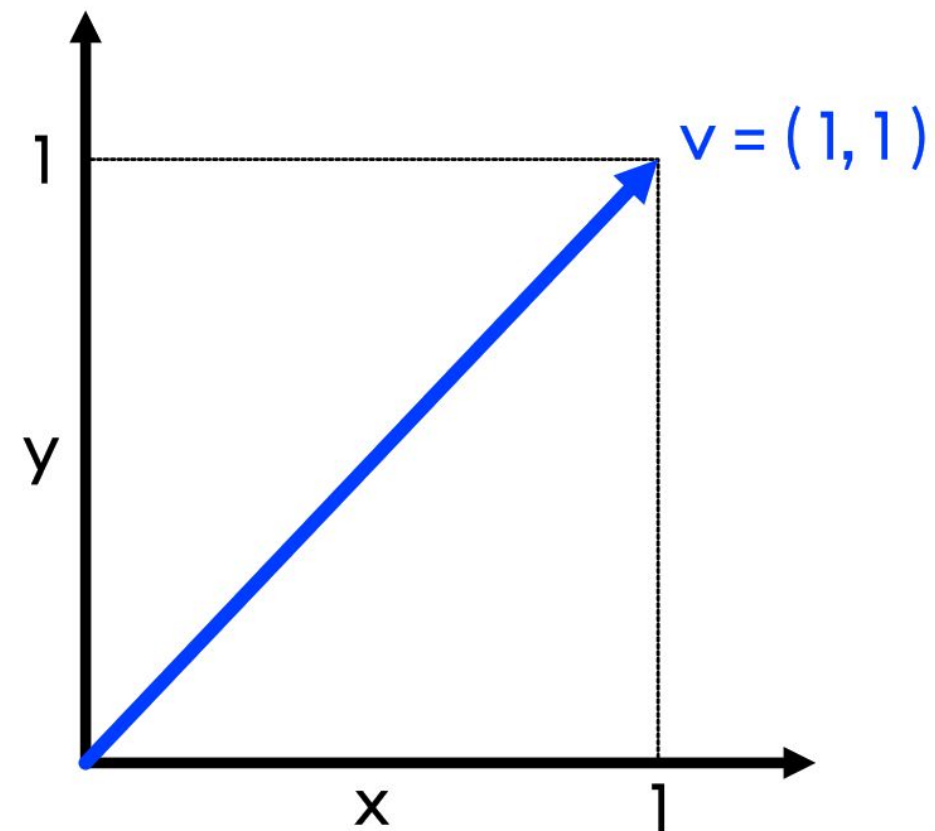


Retrieve documents from a question

- How we can retrieve documents in a database using a question?
- We need to use **semantic search**
- One solution is to use a **vector database**
- A vector database is a system that uses **vectors** (set of numbers) to retrieve information

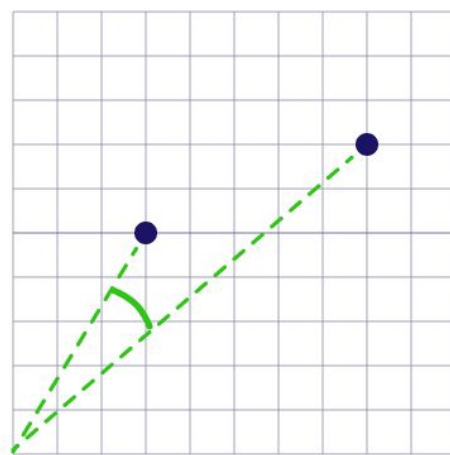
What is a vector?

- A vector is a set of numbers
- Example: a vector of 3 elements $[2, 5, -10]$
- A vector can be represented in a multi-dimensional space (eg. Llama3.2 uses 3072 dimensions)



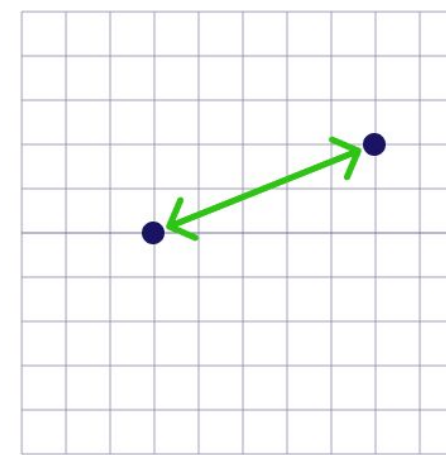
Similarity between two vectors

- Two vectors are (semantically) similar if they are close to each other
- We need to define a way to measure the similarity



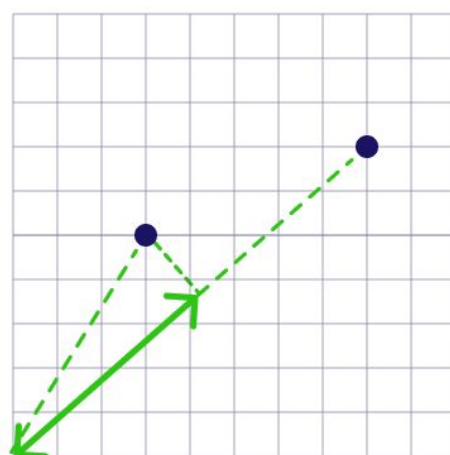
Cosine Distance

$$1 - \frac{A \cdot B}{\|A\| \|B\|}$$



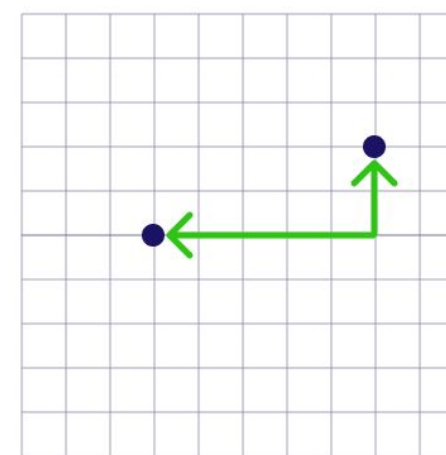
Squared Euclidean (L2 Squared)

$$\sum_{i=1}^n (x_i - y_i)^2$$



Dot Product

$$A \cdot B = \sum_{i=1}^n A_i B_i$$

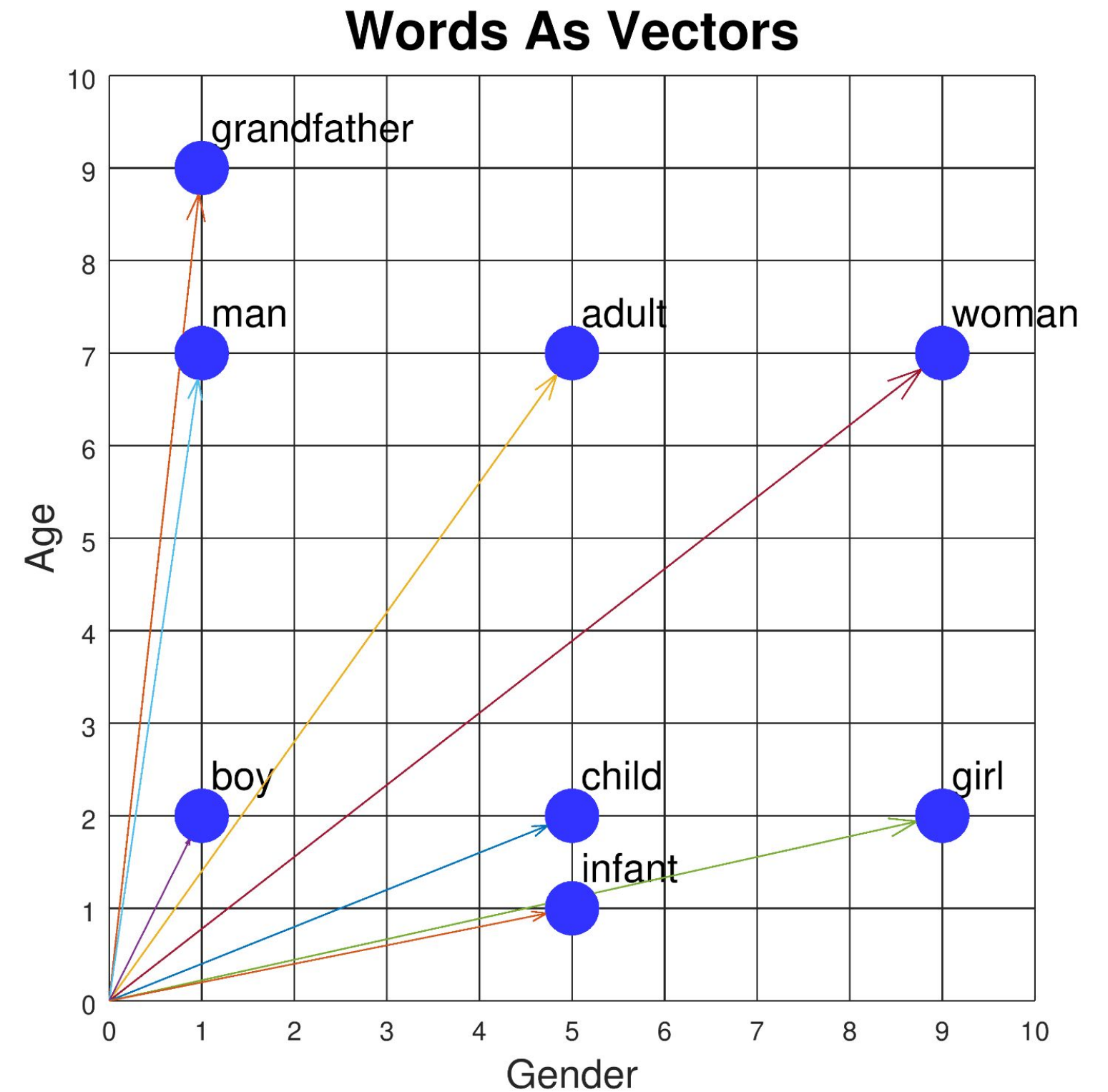


Manhattan (L1)

$$\sum_{i=1}^n |x_i - y_i|$$

Embedding

- Embedding is the translation of an input (document, image, sound, movie, etc) to a vector
- There are many techniques, using an LLM typically this is done by a neural network
- The goal is to group information that are semantically related to each other
- <https://projector.tensorflow.org/>



Vector database + LLM

- The search query (***question***) is in natural language
- We use semantic search to retrieve top-n relevant documents (***context***)
- We send the following prompt to the LLM (example):
 - *Given the following **{context}** answer to the following **{question}***

Split the documents in chunk

- We need to store data in the vector database using chunk of information
- We cannot use big documents since we need to pass it in the context part of the prompt for an LLM that typically has a token limit (e.g. Llama3.2 up to 128K)
- We need to split the documents in **chunk** (part of words)

Elasticsearch (vector database)

- [Elasticsearch](#) is Free and Open Source ([AGPL](#)), Distributed, RESTful Search Engine
- Distributed search and analytics engine, scalable data store and **vector database** optimized for speed and relevance on production-scale workloads.
- You can run it locally with a single command:
 - **`curl -fsSL https://elastic.co/start-local | sh`**

RAG demo: LLPhant + Llama 3.2 + Elasticsearch

Available on github: [ezimuel/llphant-llama-elasticsearch](https://github.com/ezimuel/llphant-llama-elasticsearch)



References

- [What is retrieval-augmented generation?](#) IBM research
- Ashish Vaswan et al., [Attention Is All You Need](#), Proceedings of 31st Conference on Neural Information Processing Systems (NIPS 2017)
- Albert Ziegler, John Berryman, [A developer's guide to prompt engineering and LLMs](#), Github blog post
- Sebastian Raschka, [Build a Large Language Model \(From Scratch\)](#), Manning, 2024
- [Elasticsearch as vector database](#), Elastic Search Labs
- [Elasticsearch search relevance](#), Elastic Search Labs
- E.Zimuel, [Generative AI and Large Language Model in PHP](#), phpDay 2024 conference, Verona (Italy)
- E.Zimuel, [Retrieval-Augmented Generation for talking with your private data using LLM](#), AI Heroes 2023 conference, Turin (Italy)

Thanks!

More information: www.elastic.co

Contact information: enrico.zimuel (at) elastic.co