## Open APIs for Open Minds

# Enhancing Tourism with Smart Technologies

Integrating FIWARE and Large Language Models

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- 1. Introduction to Generative Al and LLM
- 2. Retrieval Augmented Generation (RAG) and FIWARE
- 3. Implementation in FIWARE
- 4. DEMO: Naples Tourism use case
- 5. Summary



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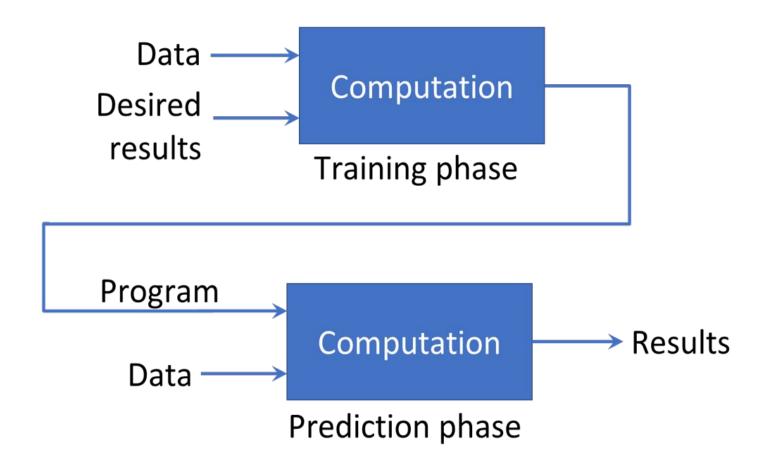
## Machine Learning





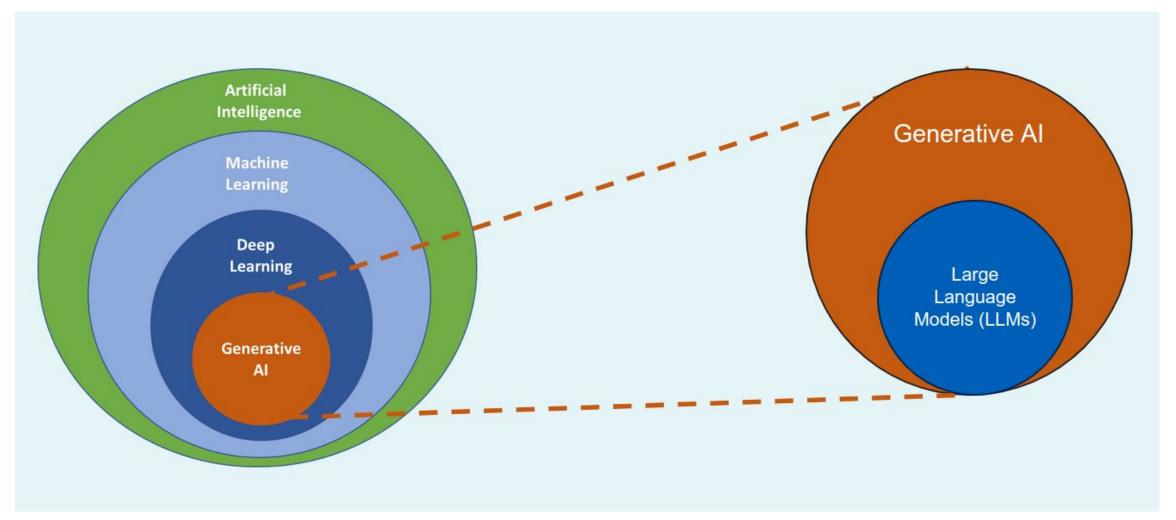


## Machine Learning





## **Generative AI**





## Large Language Models (LLM)

- From large amounts of text we train a model that learns to predict the following Word (in reality they predict tokens not words)
- Language models assume that the text can be predicted as a probabilistic system in which the next word depends on the previous ones





## Large Language Models (LLM)

Once we have this model, we can complete sentences using the model to obtain each word

When the dog saw the thief barked ...
When the dog saw the thief barked to ...
When the dog saw the thief barked to scare ...

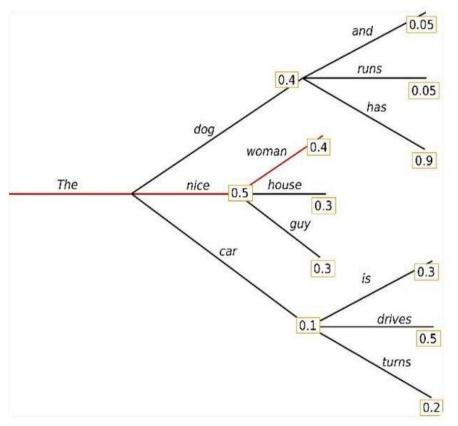


## Large Language Models (LLM)

Once we have this model, we can complete sentences using the model to obtain each word

As it is based on the probabilities of the next Word (token), there are a lot of possible paths depending on how the model decides:

- Path more probable -> deterministic (always the same result)
- Roll a die to choose the next Word (token) based on their probabilities

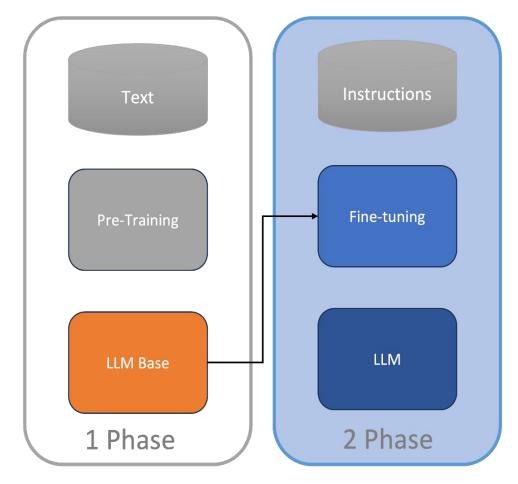




#### Chatbots

Additional training (i.e., fine-tuning)
 to learn how to solve instructions
 (Q-A)

 The data provided for this new training is pairs of instructions with your solution





### **Limitations: Context Window**

- Amount of information (tokens) that the model is able to process (remember) in a single session (conversation)
- LLMs are able to answer very complex questions because they have been trained with a lot of information
- But they are unable to remember what you said to them a few hours ago, or minutes ago...
- The Context Window Is Getting Bigger and Bigger (1 h of video, 700 k words)



#### **Limitations of LLMs**

- Hallucinations
- Traceability
- Information obsolescence
- Lack of expertise in some

areas

How many gold medals did Spain win in Paris '24?

< 2/2 >

Spain won 6 gold medals at the Paris 2024 Olympics. It was a strong showing for the Spanish team!

O O C O O O

How many gold medals did Spain win in Paris '24?

< 3/3 >

As of my last update, the 2024 Paris Olympics hadn't occurred yet, so I don't have the final count of gold medals won by Spain. For the most up-to-date information, I recommend checking the latest news or the official Olympic website. If there's anything else you're curious about or need help with, feel free to ask!



## Limitations: Context Window and Knowledge of the LLM

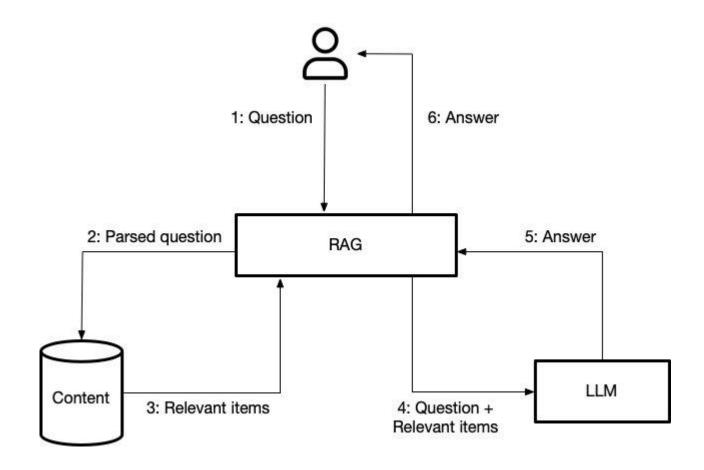
To solve these problems there are alternatives

 Fine-tuning: retrain-your model with additional data (not valid for real-time data, and not easy to obtain good results)

 Retrieval Augmented Generation architectures: the model receives the information from an external system and answers the question with that information received



## Retrieval Augmented Generation (RAG)





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## Retrieval Augmented Generation (RAG)

#### Example

How many gold medals did Spain win in Paris'24?

Country, Gold, Silver, Bronze, Total United States, 40, 44, 42, 126 Spain, 5, 4, 9, 18



Spain won 5 gold medals in the Paris 2024 Olympics.



## Retrieval Augmented Generation (RAG)

- Typically the information is stored in a vector database
  - Information is divided in chunks
  - 2. These chunks are stored in vector databases
  - When the user makes a question the most semantic similar chunks are retrieved
  - 4. The question + the chunks are passed to the LLM
  - The LLM answers the question using only the information from the chunks

But this is not the only option...

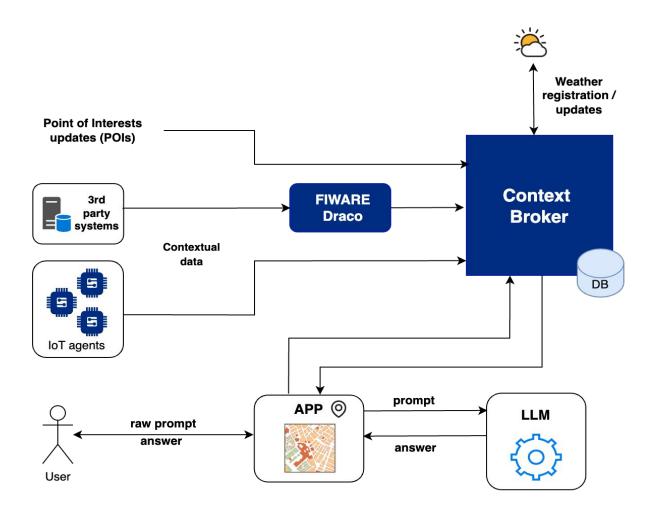


### FIWARE and RAG

- We can use the Context Broker as the source of content.
- Advantages:
  - Modeled in NGSI -> The LLMs have been trained with NGSI entities so they understand them
  - Up-to date information
  - Filter by geo-location, time, entity type, properties, etc.
  - Possibility to build NGSI queries using the LLM (not explored yet)



## FIWARE in Real—Time Spatial RAG

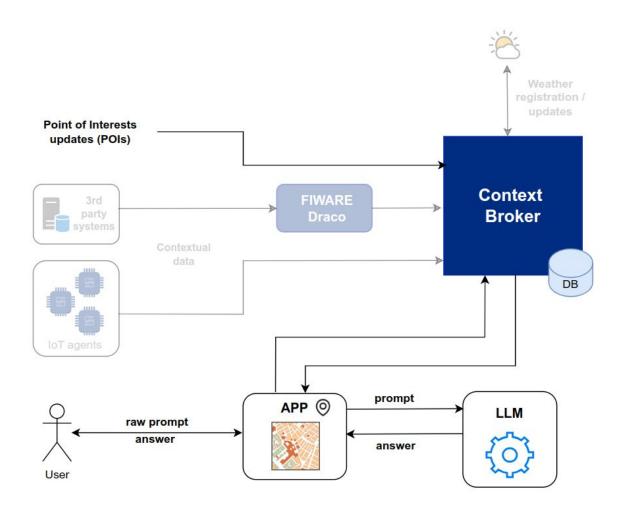




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## FIWARE in Real—Time Spatial RAG: TOURISM USE CASE





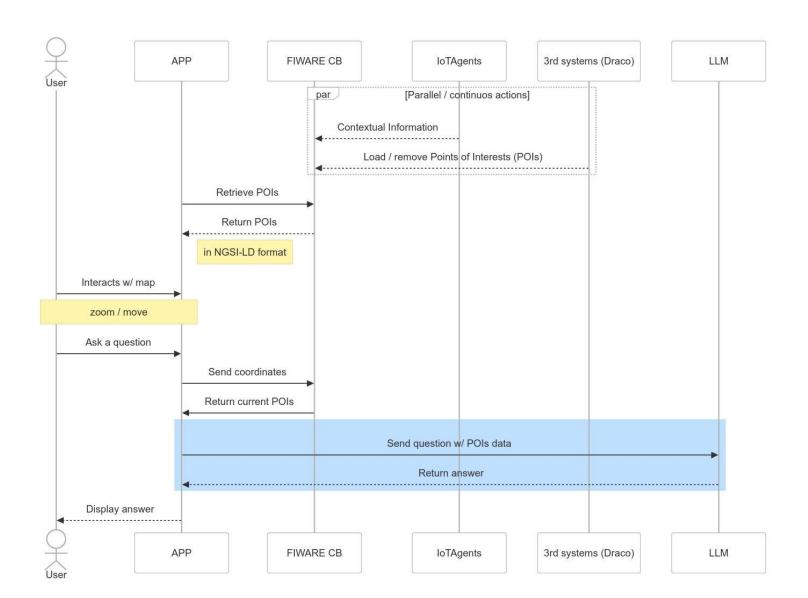
## FIWARE in Real—Time Spatial RAG: TOURISM USE CASE. Entities

```
"id": "urn:ngsi-ld:PoI:Naples:001",
"type": "PoI",
"title": {
 "type": "Property",
 "value": "Pompeii Archaeological Site"
"relevance": {
 "type": "Property",
  "value": 0
"image": {
 "type": "Property",
 "value": "pompei.jpg"
"location": {
  "type": "GeoProperty",
  "value": {
    "type": "Point",
    "coordinates": [
     14.487198,
      40.749351
```

```
"price": {
     "type": "Property",
     "value": "16€"
   "description": {
     "type": "Property",
     "value": "Pompeii is a vast archaeological site that once
was a thriving Roman city. It was buried under meters of ash
after the catastrophic eruption of Mount Vesuvius in 79 A.D."
   "capacity": {
     "type": "Property",
     "value": 30000
   "occupancy": {
     "type": "Property",
     "value": 12000
```



## FIWARE in Real—Time Spatial RAG: TOURISM USE CASE. Interactions





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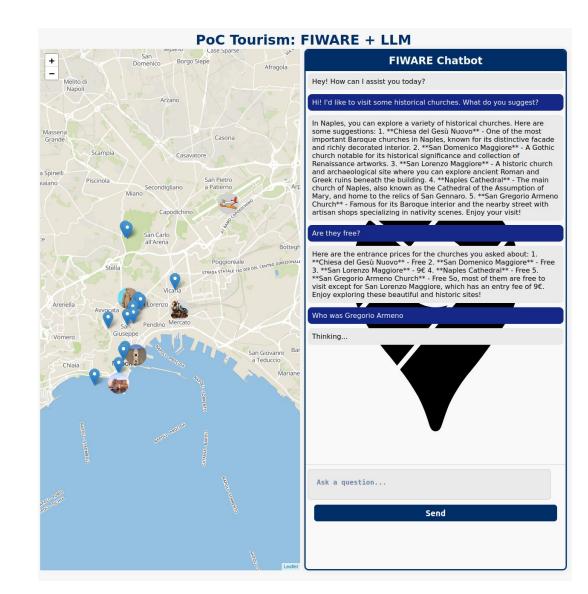
#### Demo

DEMO: during the demo we will:

- Ask info about current place
- Reformulate to refine the inquiry
- Drill down the question
- Explore options
- Iterate over possible options
- Change language to ask more information



#### Demo: Interactive part





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## Summary

- LLM ⊂ GenAl ⊂ Deep Learning ⊂ Machine Learning ⊂ Artificial Intelligence
- RAG (Retrieval-Augmented Generation) in LLMs refers to a technique that combines large language models with external retrieval systems to fetch relevant documents or data, used to enhance the model's response generation
- We have used FIWARE to perform the RAG and support the LLM with contextual data from a city
- The **CB** is used as a central piece that handles the contextual data
- This sets up the basis for a general architecture that could be used to enhance the experience in tourism and could be the ground for other types of powered-by-FIWARE solutions that could be improved to interact with users.

#### Resources

- Link to the code and slides
  - https://github.com/dncampo/Poc\_LLM\_CB\_FIWARE-Naples



## Thank you!

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