

Hands-On GenAI: LLMs, RAGs, and Agentic Systems for Beginners

Day 10

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Recap

LLMs

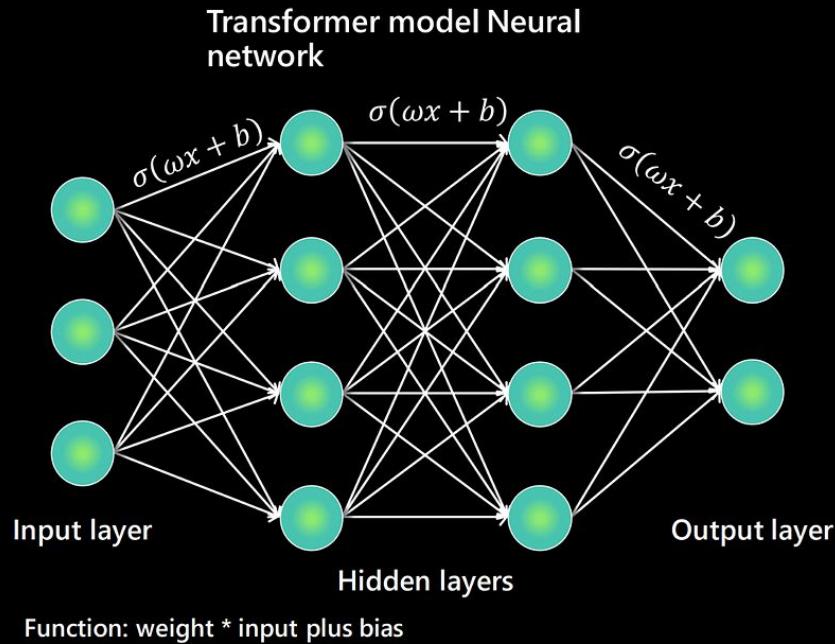
1. User input: is a (long) string
 2. Preprocess text: convert to lowercase, remove special characters, etc.
 3. Tokenize text: convert to separate tokens
 4. Convert to embeddings (numerical representations)
-
1. "What is the weather today?"
 2. "what is the weather today"
 3. ["what", "is", "the", "weather", "today"]
 4. [[0.78 0.98 0.86] ,
[0.76 0.54 0.38] ,
[0.73 0.58 0.84] ,
[0.28 0.23 0.88] ,
[0.18 0.17 0.59]
]

LLMs

1. ML algorithm is run on this (eg. attention) → generates a sequence of vectors
 2. Vectors are decoded into their alphabetical representation
1. relationship between two word vectors (Q , K , V) = $\text{softmax}(QK^T/\sqrt{d_k}) * V$
=> testing the model =>

```
[ [0.12 0.91 0.60] ,  
  [0.21 0.45 0.37] ,  
  [0.22 0.59 0.09] ,  
  [0.24 0.25 0.01]  
 ]
```
 2. "It is cloudy today"

How large are they?



BERT Large - 2018

345M

GPT2 - 2019

1.5B

GPT3 - 2020

175B

Turing Megatron NLG
2021

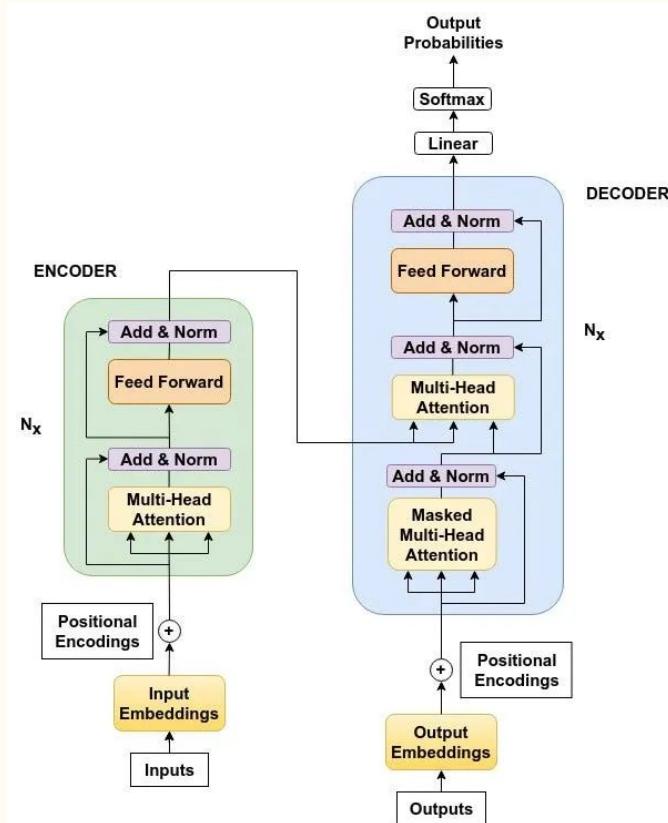
530B

GPT4 – 2023

1.4T (estimated)

Source:
<https://bobrupakroy.medium.com/what-are-the-parameters-in-lm-76da7040e607>

LLM Parameters



Transformer Architecture

Source:
<https://bobrupakroy.medium.com/what-are-the-parameters-in-lm-76da7040e607>

RAG

1. Facts are stored as vectors in a vector database.
 2. User asks a question/ prompt
 3. Prompt text processing, conversion to numerical (vector representation)
 4. Retrieval: of top k similar vectors, through cosine similarity \Rightarrow cosine similarity is computed for every fact_vector with query_vector, and top k most similar vectors are chosen.
-
1. Huge collection of fact_vector (one vector for each fact)
 2. question = "What is a smartphone battery made of?"
 3. Result: query_vector = [0.72 0.96 0.83 0.96 0.87 0.67 0.34 0.23 0.12 0.99]
 4. cosine_similarity(query_vector, fact_vector) = dot_product(query_vector, fact_vector)

RAG

1. **Retrieval:** of top k similar vectors, through cosine similarity.
2. Decoding these vectors into facts.

1. if k = 2, let's say the most similar vectors were
fact_vector_5 = [0.92 0.16
0.03 0.50 0.17 0.65 0.12 0.13
0.42 0.49], and fact_vector_8
= [0.22 0.66 0.73 0.13 0.91
0.65 0.39 0.13 0.01 0.02]
2. fact_5 = "Smartphones
typically feature lithium-ion
or lithium-polymer
batteries."
fact_8 = "Smartphones are
typically equipped with a
power button and volume
buttons."

RAG

1. **Augmentation:** of prompt to LLM, by providing context of facts retrieved.
 1. `context = fact_5 + fact_8`
`prompt = f"Using the`
`provided context:`
`{context}, answer this`
`question: {question}"`
 2. `generated_response = "The`
`answer is lithium-ion or`
`lithium-polymer batteries.`
`The answer: Lithium-ion`
`batteries are made of`
`lithium, a type of metal."`
2. **Generation:** of output by LLM.

AI Models vs AI Agents

AI models-

- require human intervention
- predefined constraints

AI agents-

- have more autonomy
- require limited human intervention
- have goal-driven behaviour
- have adaptability

source:

<https://www.ibm.com/think/topics/agentic-ai>

AI Models vs AI Agents

AI model example:

- User: "What is the capital of France?"
- AI Model: "The capital of France is Paris."

source:

https://dev.to/abhishekjaiswal_4896/the-difference-between-ai-agents-and-traditional-ai-models-1aj

AI agents example:

- User: "Research the latest trends in AI, summarize key points, and email me a report."
- AI Agent's Steps:
 - Searches online for the latest AI research papers and news articles.
 - Summarizes key trends and insights from multiple sources.
 - Generates a well-structured report.
 - Sends the report to the user's email automatically.

An Agent in a Project: Demo

Thank You!