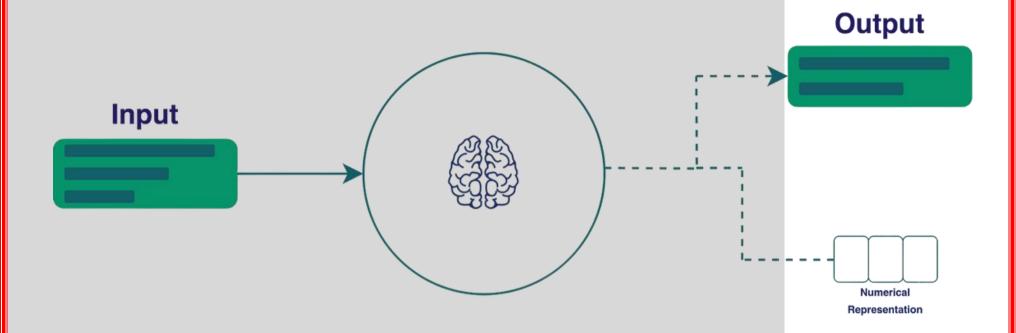
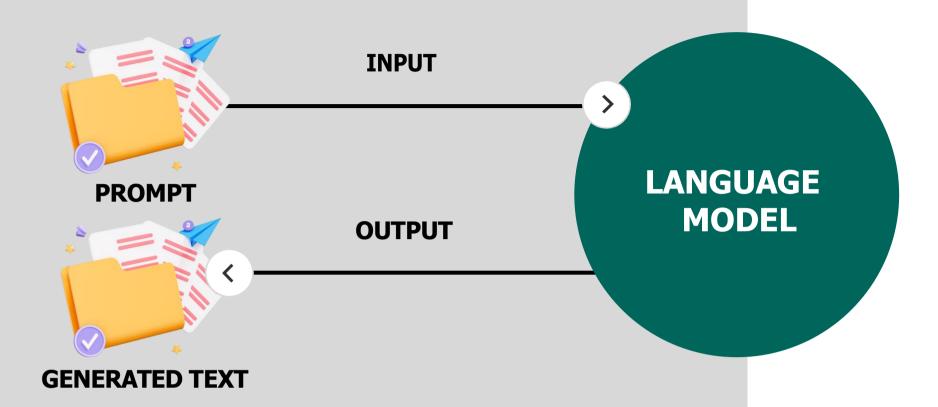
MASTERING GENERATIVE

LLM Basics



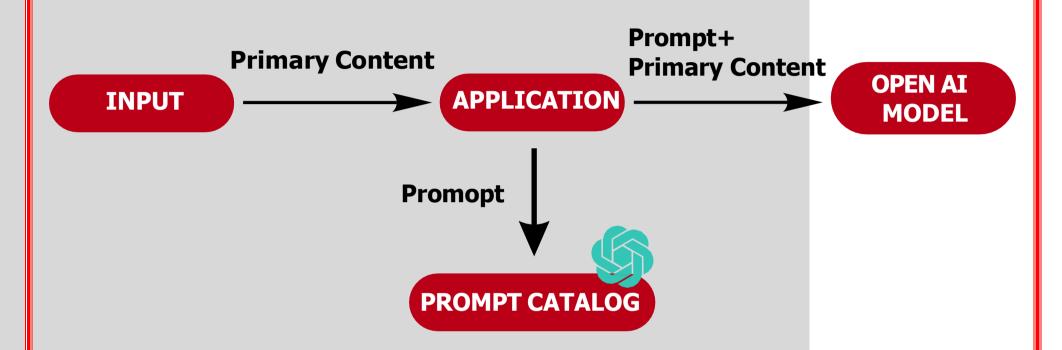
- Basics: Understand key terms, uses, issues, and frameworks.
- Data: Know the training data and potential biases.
- Scale: Be aware of LLM size and training costs.
- Training: Understand differences between LLM and ML training.
- Purpose: Define clear objectives (chatbot, Q&A, image generator).

Prompt Engineering



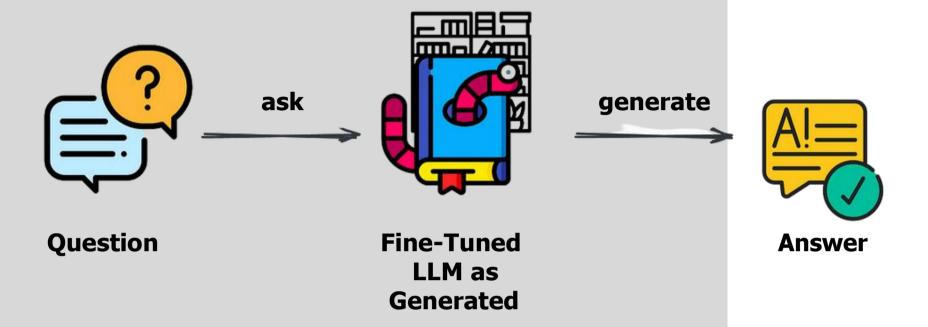
- What is it? Designing precise inputs for LLMs to get specific results.
- Example: Instead of "Create a Twitter post," use "Create a snappy Twitter post for millennials."
- Fine-tuning: Adjust prompts to achieve the exact output you need.
- Growing Field: Prompt engineering has become a specialized job.
- The Art of AI: It's all about mastering the nuances of communication with AI.

Prompt Engineering with OpenAl



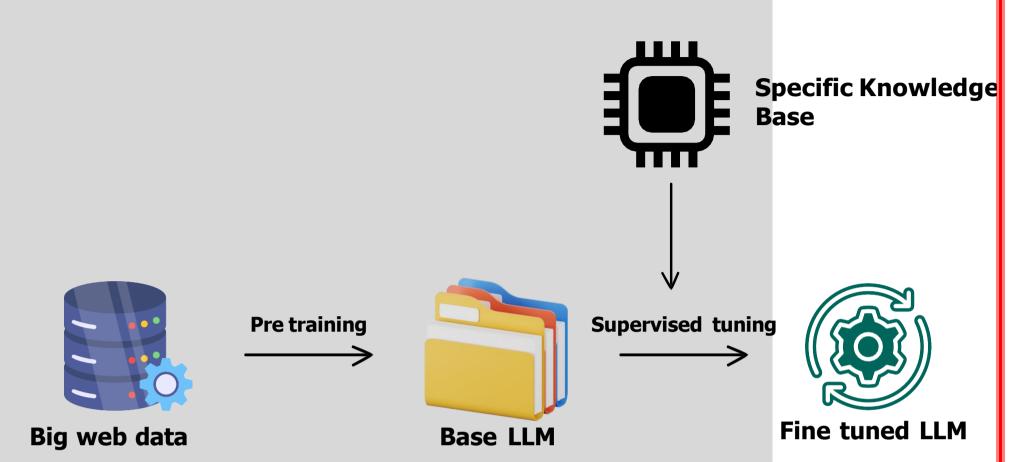
- Stay Updated: Use the latest OpenAI API version and relevant plugins.
- Follow Guidelines: Ensure you adhere to any specific instructions from tools like Microsoft's Azure OpenAI GPT models.
- Career Essential: Knowing prompt engineering with OpenAI is crucial for your career in LLMs and generative AI.

Question-Answering



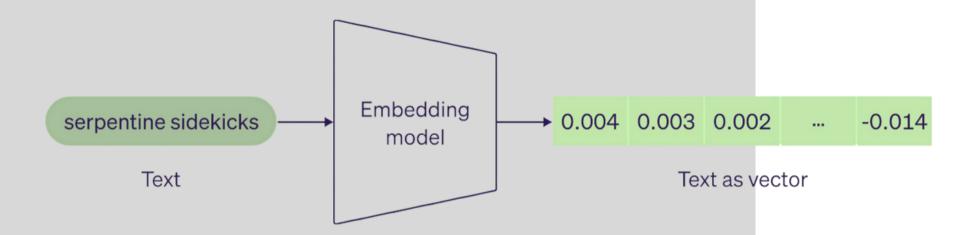
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Fine-Tuning



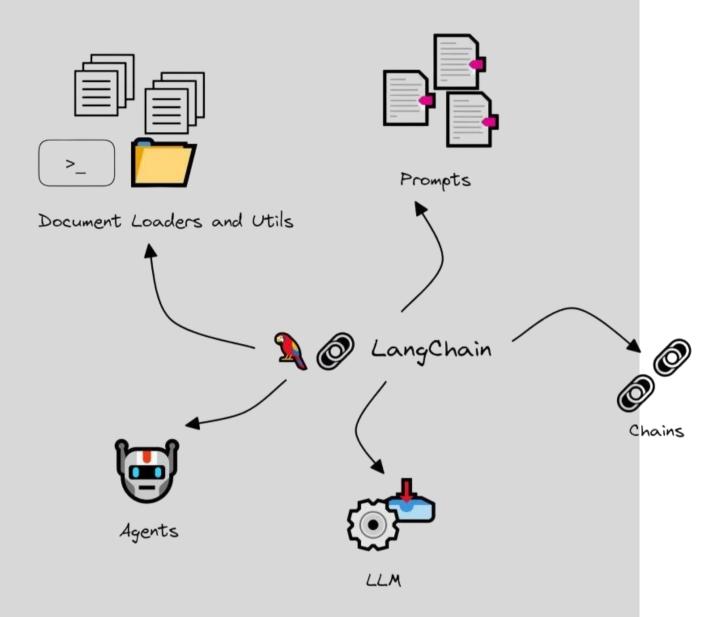
- Enhance LLM abilities in text generation, translation, summarization, and QA.
- Customize for specific applications like chatbots or medical systems.
- Various methods for fine-tuning, including supervised learning.
- Continuous refinement ensures adaptability to evolving tasks and domains.
- Leverage labeled datasets to train LLMs efficiently and effectively.

Embedding Models



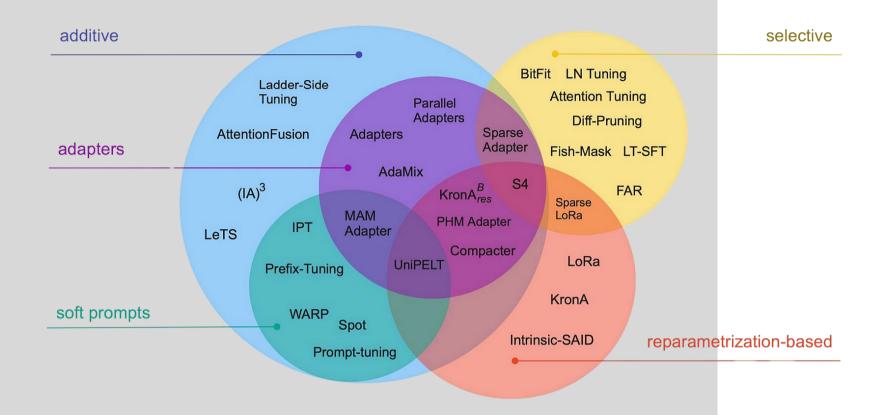
- Embedding models map natural language to vectors for downstream LLMs.
- Fine-tune pipelines with multiple models to capture data nuances effectively.
- LLMs benefit from pre-trained word embeddings, enhancing semantic understanding.
- Embedding models lay the groundwork for coherent and contextually relevant text generation.

LangChain



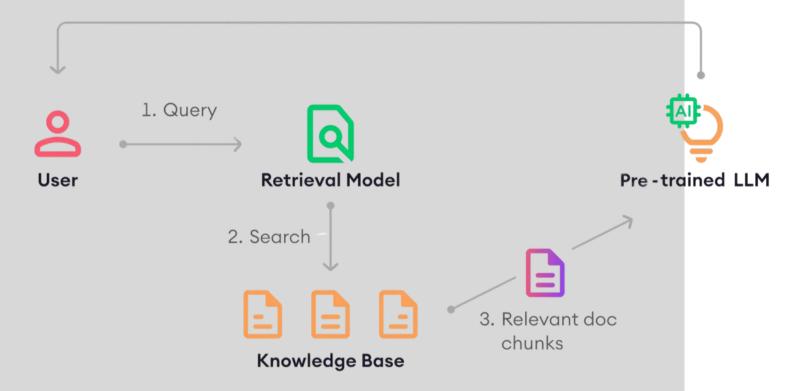
- Chain multiple models for complex tasks like classification, text, and code generation.
- Integrate with diverse systems for tasks like
 API calls, data science, and querying.
- Use Agents to interact with external systems, executing actions guided by LLMs.
- Agents empower LLMs to select and execute actions using a range of tools.

Parameter Efficiency/Tuning



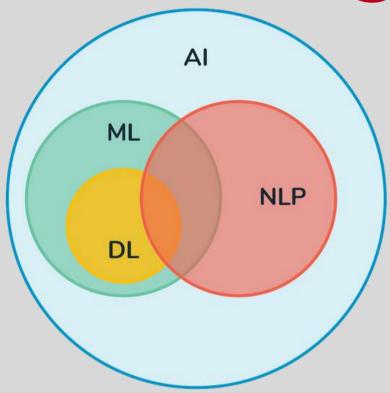
- Adapt large language models like GPT or BERT for specific tasks with minimal parameter overhead.
- Add compact task-specific "adapters" to pretrained models instead of fine-tuning the entire model.
- Reduce computational and memory requirements, making fine-tuning more feasible.

RAG



- Retrieval-based model: Retrieves relevant documents from a knowledge base based on input text.
- Linking: Connects retrieved documents with input text.
- Generative model: Uses linked input and documents to generate output text.
- Integration: Considers both input text and retrieved documents for enhanced text generation.

Natural Language Processing



- Foundation: LLMs and generative AI are rooted in NLP principles.
- Training Data: Massive datasets of text and code drive LLMs, leveraging NLP techniques.
- Meaning Understanding: NLP aids LLMs in comprehending data semantics and generating text.
- Effectiveness: Strong grasp of NLP is essential for leveraging LLMs efficiently.