

# LLM Lingo: Must-Know Terms

Around 70 of the most frequently used LLM terms, along with simple explanations, sorted by topics.

## Part 1

### Basic LLM Related Terms

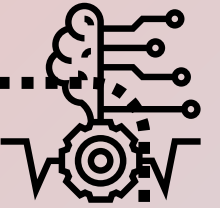
Basic terms that are in the LLM space like **foundation model**, **prompting**, **context-length** etc.



## Part 2

### Fine-tuning

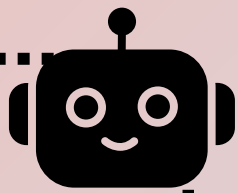
Fine-tuning related terms like **PEFT**, **quantization**, **LoRA** etc.



## Part 3

### RAG and Agents

Terms used in RAG like **vector databases**, **chunking** etc. and some basic agent terms like **function calling**



## Part 4

### Enterprise Ready LLMs

Terms that are commonly used while deploying LLMs like **compliance**, **GDPR**, **PII**, **ethics** etc.



## Part 5

### LLM Vulnerabilities & Attacks

Terms related to LLM attacks like **adversarial attacks**, **prompt injection**, **prompt leaking** etc.



## Part 6

### LLM Learning Paradigms

Different learning paradigms that LLM use during training like **unsupervised learning**, **reinforcement learning** etc

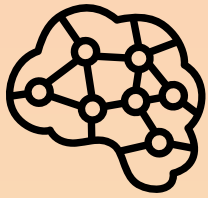


# LLM Lingo: Must-Know Terms

## Part 1

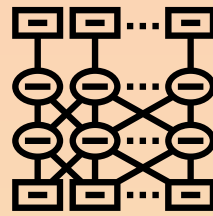
Created By: Aishwarya Naresh Reganti

### Foundation Model



LLM designed to generate and understand human-like text across a wide range of use-cases

### Transformer



A popular LLM design known for its attention mechanism and parallel processing abilities

### Prompting



Providing carefully crafted inputs to an LLM to generate desired outputs

### Context-Length



Maximum number of input words/tokens an LLM can consider when generating an output.

### Few-Shot Learning



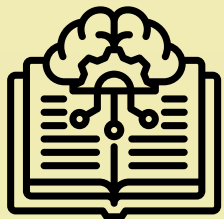
Providing very few examples to an LLM to assist it in performing a specific task.

### Zero-Shot Learning



Providing only task instructions to the LLM relying solely on its pre-existing knowledge

### RAG



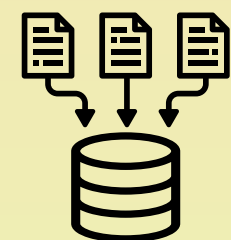
Retrieval-Augmented Generation. Appending retrieved information to improve LLM response

### Knowledge Base(KB)



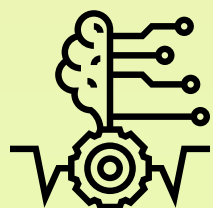
Collection of documents from which relevant information is retrieved in RAG

### Vector Database



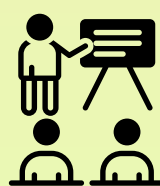
Stores vector representations of the KB, aiding the retrieval of relevant information in RAG.

### Fine-Tuning



Adapting an LLM to a specific task or domain by further training it on task-specific data.

### Instruction Tuning



Adjusting an LLM's behavior during fine-tuning by providing specific guidelines/directives

### Hallucination



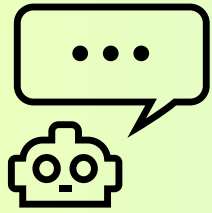
Tendency of LLMs to sometimes generate incorrect or non-factual information.

# LLM Lingo: Must-Know Terms

## Part 2: Fine-Tuning Edition

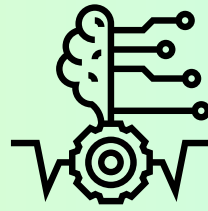
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### In-Context Learning



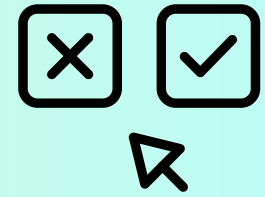
Integrating task examples into prompts, enabling LLMs to handle new tasks without fine-tuning.

### SFT



Supervised Fine-Tuning. Updating a pre-trained LLM with labeled data to perform a specific task.

### Contrastive Learning



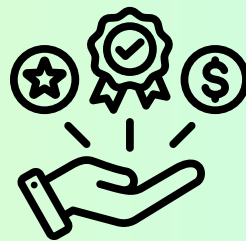
Fine-tuning method that improves LLM by teaching it to discern data similarity and differences.

### Transfer Learning



Applying pre-trained knowledge from large datasets to improve LLM performance on smaller, task specific data.

### Reward Modeling



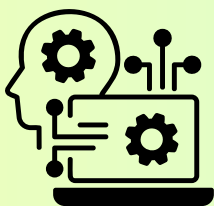
Designing objectives to reward LLM outputs during the reinforcement learning process.

### Reinforcement Learning



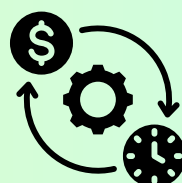
Training LLMs through trial and error, with rewards/penalties based on its generated outputs

### RLHF



Reinforcement Learning from Human Feedback. Human feedback is used as reward/penalty for LLM

### PEFT



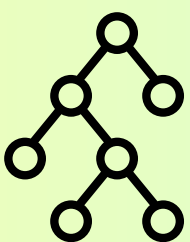
Parameter-Efficient Fine-Tuning updates only few parameters of LLMs and is hence both compute and cost efficient.

### Quantization



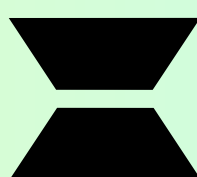
Reducing the precision of LLM parameters to save computational resources without sacrificing performance.

### Pruning



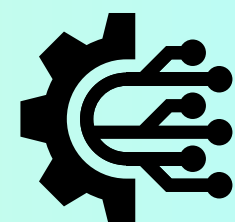
Trimming surplus connections or parameters to make LLMs smaller and faster yet performant

### LoRA



Low-Rank Adaption is a PEFT method that inserts a smaller set of new weights to the LLM & trains only those.

### Freeze Tuning



Fine-tune with most of the LLM's weights frozen, except for some layers, generally, the task specific layers



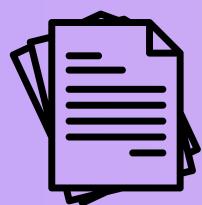
# LLM Lingo: Must-Know Terms

## Part 3: RAG + LLM Agents

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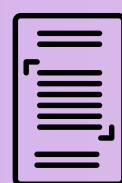
RAG or Retrieval-Augmented Generation in LLMs, combines regular language generation with information retrieval, enabling models to access external knowledge sources and improve the quality and relevance of their generated outputs. Below are some key terms:

### Knowledge Base(KB)



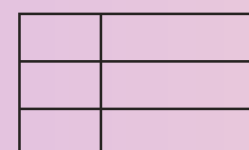
Collection of documents from which relevant information is retrieved in RAG

### Chunking



Breaking the KB into smaller pieces for efficient storage and retrieval during RAG

### Indexing



Organizing and storing KB chunks in a structured manner for efficient retrieval

### Embedding Model



An LLM that converts KB text chunks into numerical format called vectors/embeddings

### Vector Database



Database optimized for storing and retrieving vector representations generated from the KB

### Vector Search



Finding the most relevant KB chunks based on vector similarity scores for a given input query.

### Retrieval



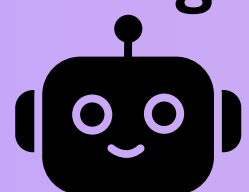
Approach used to rank and fetch KB chunks from the vector search. This will serve as additional context for the LLM.

### AGI



Artificial General Intelligence aims to create machines that can learn and reason like humans across various tasks.

### LLM Agent



LLM applications that can execute complex tasks by combining LLMs with modules like planning and memory

### Agent Memory



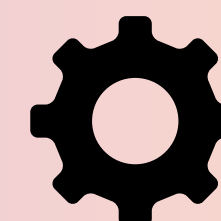
A module that stores the agent's past experiences and interactions with the user and environment.

### Agent Planning



Module that divides the agent's tasks into smaller steps to address the user's request efficiently.

### Function Calling



Ability of LLM agents to request information from external tools and APIs in order to execute a task

# LLM Lingo: Must-Know Terms

## Part 4: Enterprise Ready LLMs

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### LLM Bias



Systematic prejudices in the LLM's predictions, often stemming from training data

### XAI



Explainable AI. Making the model's outputs understandable and transparent to humans.

### Responsible AI



Ensuring ethical, fair, and transparent development and use of AI systems.

### AI Governance



Legal policies & frameworks that regulate the development & deployment of AI systems

### Compliance



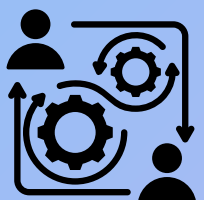
Ensuring adherence to legal requirements in the development & deployment of AI systems.

### GDPR



General Data Protection Regulation protecting individuals' privacy rights and governing data handling in the EU

### Alignment



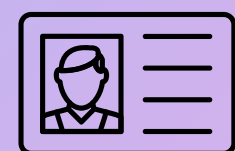
Ensuring that the outputs of LLMs are consistent with human values and intentions.

### Model Ethics



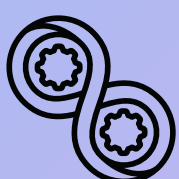
Ensuring ethical behavior (transparency, fairness, accountability etc.) when deploying public facing AI

### PII



Personally Identifiable Information. Should not be stored or used without proper processes and user consent.

### LLMOps



Managing and optimizing operations for LLM deployment and maintenance.

### Privacy-preserving AI



Methods to train and use LLMs while safeguarding sensitive data privacy.

### Adversarial Defense



Methods to prevent malicious attempts to manipulate LLMs, ensuring their security.



# LLM Lingo: Must-Know Terms

## Part 5: LLM Vulnerabilities and Attacks

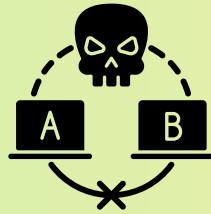
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### Adversarial Attacks



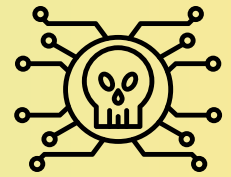
Deliberate attempts to trick LLMs with carefully crafted inputs, causing them to make mistakes.

### Black-Box Attacks



Trying to attack an LLM without knowing its internal workings or parameters.

### White-Box Attacks



Attacking an LLM with full knowledge of its internal architecture and parameters.

### Vulnerability



Weaknesses or flaws in LLMs that can be exploited for malicious purposes.

### Deep-fakes



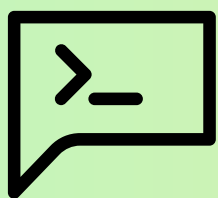
Synthetic media generated by LLMs, often used to create realistic but fake images or videos.

### Jailbreaking



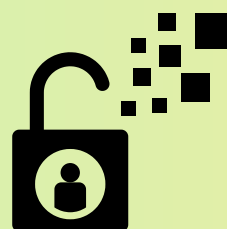
Attempting to bypass security measures around an LLM to make it produce unsafe outputs.

### Prompt Injection



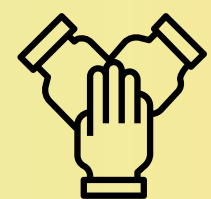
Hijacking the LLM's original prompts to make it perform unintended tasks

### Prompt Leaking



Tricking an LLM to reveal information from its training or inner workings.

### Red-Teaming



Assessing the security and robustness of LLMs through simulated adversarial attacks.

### Robustness



The ability of an LLM to perform accurately despite encountering adversarial inputs.

### Alignment



Ensuring that the behavior of an LLM is consistent with human values.

### Watermarking



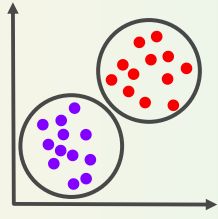
Embedding hidden markers into LLM-generated content to track its origin or authenticity.

# LLM Lingo: Must-Know Terms

## Part 6: LLM Learning Paradigms

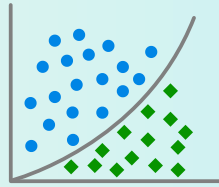
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### Unsupervised Learning



Learning patterns and structures from data without specific guidance or labels.

### Supervised Learning



Learning from labeled examples & associating inputs with correct outputs.

### Reinforcement Learning



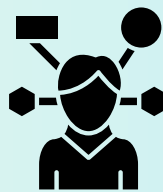
Learning through trial and error, with rewards or penalties based on generated outputs

### Meta-Learning



Learning to learn by extracting general knowledge from diverse tasks and applying it to new ones.

### Multi-task Learning



Learning to perform multiple tasks & sharing knowledge between related tasks for better performance.

### Zero-Shot Learning



Providing only task instructions to the LLM relying solely on its pre-existing knowledge

### Few-Shot Learning



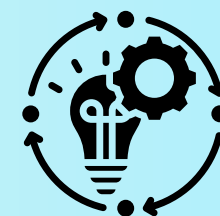
Learning from a small number of examples for new tasks and adapting quickly with minimal data.

### Online Learning



Continuously learning from incoming data streams and updating knowledge in real-time.

### Continual Learning



Learning sequentially from a stream of tasks or data without forgetting previously learned knowledge.

### Federated Learning



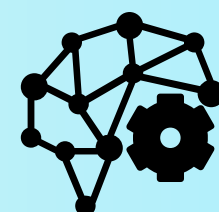
Training across multiple decentralized devices without sharing raw data, preserving user privacy.

### Adversarial Learning



Training against adversaries or competing models to improve robustness and performance.

### Active Learning



Interacting with humans or the environment to select and label the most useful data for training.