

# Generative Al Accelerator

# Where we are?

#### Intro to Generative Al

- Intro to Al
- History and evolution of Al
- Discriminative vs Generative Al
- Al Landscape
- Open Source vs Proprietary LMs
- Responsible AI Ethics, Bias etc

#### **LLM Foundations**

- Transformers Intuition
- Next Token Prediction
- LLM Training Phases
- Pre-training/ Post-training
- Model behaviour parameters
- Different types of LLMs
- Limitations of LLMs

#### **Prompt Engineering**

- Prompting Techniques
- CoT/Tool Calling
- Prompt Evaluation
- Prompt OptimizationContext Engineering
- Prompt Hacking/ Jailbreaks

#### **RAG Systems**

- RAG basics
- Different types of RAG
- RAG Evaluation
- Improving RAG performance

# Designing LLM Systems

- Choosing the right stack
- Evaluating LLM systems
- Design Tradeoffs: Latency/Cost
- Performance Optimization
- Security and Privacy
- Case Study Product Search

#### Al Agents

- Intro to Agents
- Tool Use and Memory
- Workflow vs Agents
- Agent orchestration patterns
- Agent Evaluation
- Model Context Protocol (MCP)

#### Capstone Project

- Project to apply your learnings
- Demo Day



How do we measure if Retrieval Augmented Generation really works?

# Why evaluate RAG?

### **LLMs** ≠ Truth Engines:

Large Language Models generate fluent text, not verified facts.

## **RAG** # Perfect Grounding:

Retrieval adds context, but irrelevant or missing documents can still mislead the model.

### **Real-World Impact:**

Inaccurate responses can cause misinformation in customer support, wrong recommendations, or compliance risks in enterprises.

# **Evaluation = Feedback Loop:**

It helps identify whether the problem lies in:

- Retrieval → wrong or missing documents
- Generation → hallucinated or incomplete answers
- Grounding → weak connection between retrieved context and final output

