

# AI Bootcamp to Accelerate your Generative AI Skills

## Who Should Attend?

### 1. The Forward-Thinking Data Analyst

- *Background:* Works with large datasets, performs data cleaning and reporting.
- *Goal:* Use NLP and generative AI to derive deeper insights, automate repetitive tasks, and enhance reporting with natural-language summaries.

### 2. The Aspiring AI Engineer

- *Background:* Familiar with Python and basic machine learning concepts.
- *Goal:* Master prompt engineering, fine-tune language models, and build end-to-end solutions that incorporate RAG for more accurate, context-aware responses.

### 3. The Tech-Savvy Consultant

- *Background:* Advises clients on technology adoption, change management, or operational efficiency.
- *Goal:* Gain hands-on practice in developing AI-driven case studies—enabling advanced client solutions, faster prototyping, and better strategic recommendations

## Key Takeaways

- **Practical Skills:** Develop, deploy, and refine AI-driven solutions from the ground up.
- **Industry-Relevant Projects:** Build portfolio-ready prototypes, like a RAG-powered chatbot.
- **Custom Use Cases:** Apply your newfound skills to challenges directly related to your organization or field.

## Overview of the Syllabus

| Week | Topic  | Lecture   | Lab   |
|------|--|---|---|
|      | <b>GenAI, LLMs, Prompts</b>                    | <b>Duration: 2 hours</b><br>Generative AI Overview and Applications<br>LLM Architectures, Capabilities, and Limitations<br>Engineering effective prompts and types of prompts   | <b>Duration: 3 hours</b><br>Use AI models.<br>Create, Test, Refine prompts.<br>Analyze LLM outputs.                       |
| 2    | <b>Document/Text Processing</b>                | <b>Duration: 2.5 hours</b><br>NLP concepts, Sentiment Analyses<br>Word Embedding principles (word2vec GloVe)<br>NLP+ LLM Integration for data processing  | <b>Duration: 2.5 hours</b><br>Explore NLTK spaCy Libraries<br>Explore embeddings with data<br>Build end-end NLP pipeline. |
| 3    | <b>RAGs</b>                                    | <b>Duration: 2.5 hours</b><br>RAG principles and Architectures<br>RAG + LLM for specificity and external knowledge<br>RAG applications in industry  | <b>Duration: 2.5 hours</b><br>Develop a RAG pipeline<br>Explore RAG+LLM integrations<br>Team discussions on RAGs          |
| 4    | <b>Case Study</b><br>(Sample/<br>Personalized) | <p><b>Objective:</b> Develop a prototype chatbot that leverages Retrieval Augmented Generation (RAG) to answer customer inquiries by retrieving relevant information from a curated set of support documents (e.g., FAQs, product manuals, and troubleshooting guides).</p> <p><b>Deliverable:</b> A functional prototype of a RAG-powered Q&amp;A chatbot.</p> <p><b>Design:</b></p> <ol style="list-style-type: none"> <li>1. Data Preparation: <ul style="list-style-type: none"> <li>○ Corpus: Provide a set of documents such as company FAQs, product documentation, and support guides.</li> <li>○ Preprocessing: Use NLP techniques to clean and tokenize text, preparing it for embedding generation.</li> </ul> </li> <li>2. Embedding and Retrieval: <ul style="list-style-type: none"> <li>○ Word Embeddings: Generate embeddings for the documents using pre-trained models (e.g., Word2Vec, GloVe, or transformer-based embeddings).</li> <li>○ Retrieval Mechanism: Implement a simple similarity search (e.g., cosine similarity) to identify the most relevant documents in response to a user query.</li> </ul> </li> <li>3. Prompt Engineering and Generation: <ul style="list-style-type: none"> <li>○ Contextual Prompt: Create a prompt template that incorporates the retrieved documents as context.</li> <li>○ Response Generation: Use a pre-trained large language model to generate an answer based on the prompt and context.</li> </ul> </li> <li>4. Integration and Evaluation: <ul style="list-style-type: none"> <li>○ Prototype: Develop a simple interface (command-line or web-based) where users can input queries and receive generated responses.</li> <li>○ Testing: Evaluate the system with various queries, refine prompt templates, and adjust retrieval parameters to improve accuracy and relevance.</li> </ul> </li> </ol> <p><b>Step-by-Step Implementation:</b></p> <ol style="list-style-type: none"> <li>1. Set up the environment, preprocess the document corpus, generate embeddings.</li> <li>2. Develop, test the retrieval system that selects relevant documents for a query.</li> <li>3. Integrate the retrieval output with a prompt for the language model and build the prototype interface. Conclude with group demonstrations and feedback</li> </ol> |   |