

Assignment 1: Gen AI Application

Problem Statement

You are required to build a **Retrieval-Augmented Generation (RAG)** application with **agentic behavior**, a **Streamlit UI**, and a **FastAPI backend**.

Your tasks:

1. Create a Knowledge Base

- Use any publicly available dataset (e.g., airline SOP documents, financial policies, retail product descriptions, etc.).
 - Convert documents into embeddings and store them locally (e.g., FAISS, ChromaDB).
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2. Build a RAG Pipeline (in FastAPI)

Your FastAPI application must expose the following endpoints:

1. /ask

- Input: { "query": "..."}
- Output: A GenAI response grounded in retrieved documents.

2. /agent-ask

- Implement a simple **task-solving agent** that:
 - Breaks a user query into smaller sub-queries
 - Retrieves relevant context
 - Produces a final synthesized answer
 - Example queries the agent should handle:
 - “Compare policies between Document A and B and summarize differences.”
 - “Find steps from all SOPs and create a combined checklist.”
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3. Streamlit Application

Build a frontend with:

- A text input box with chat interface

- A dropdown to select **RAG** or **Agent Mode**
 - Display retrieved context + final answer
 - Call your FastAPI backend
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4. Deployment

You must deploy the app so we can test it:

- Provide:
 - Public URL for testing
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5. Submission Requirements

- Submit a **ZIP file of your codebase**, containing:
 - FastAPI backend code
 - Streamlit UI
 - RAG pipeline
 - Agent logic
 - Requirements.txt
- A video explaining the working of the GenAI application

Assignment 2: ML Lifecycle

Problem Statement

Consider any open dataset -

Example: Telco Customer Churn dataset, Loan Default dataset, or any Classification dataset.

Your task:

1. Perform EDA

- Identify missing values, outliers
- Understand feature distributions
- Identify key drivers of the target variable

- Provide visualizations
 - Summarize insights in business terms
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2. Preprocessing & Feature Engineering

- Handle missing values
 - Encode categorical features
 - Scale numerical variables if needed
 - Engineer at least **2 new features**
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3. Build and Train Models

Train at least **three different models**, e.g.:

- Logistic Regression
 - Random Forest
 - XGBoost / LightGBM
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4. Evaluate Model Performance

Include:

- Accuracy
 - Precision, Recall, F1
 - ROC curve & AUC
 - Confusion matrix
 - Feature importance
 - Business interpretation of results
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5. Select the Best Model

Explain:

- Why this model performs best
- How it would be deployed

- What monitoring would look like in production
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6. Deliverables

Submit a ZIP containing:

- Jupyter notebook
- Dataset (or link)
- Model files
- A video explaining the solution & the full life cycle