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COMPUTER SCIENCE
INTRODUCTION TO AI
END OF SEMESTER EXAMINATION

PROJECT DOCUMENTATION

🔧 FEATURE USAGE GUIDE

1) Regression Analysis

- Navigate to the **Services** section.
- Select **Regression**.
- Upload your dataset in CSV format.
- Choose your **target column** and the **independent features**.
- (Optional) Apply preprocessing like missing value handling or normalization.
- Click **Run Regression** to train and evaluate the model.
- Visual feedback including plots and metrics will be displayed.
- If using a single feature, you can input a value to generate a prediction.

2) Clustering

- Go to the **Services** menu.
- Choose the **Clustering** feature.
- Upload your dataset as a CSV file.
- Specify the number of clusters you want.
- Click **Run Clustering** to execute.
- View the resulting clusters and corresponding visualizations.

3) Neural Network Classifier

- Head to the **Services** section.
- Click on **Neural Networks**.
- Upload your CSV file.
- Select the **target column** for classification.
- Adjust hyperparameters like **epochs** and **learning rate**.
- Train the model and observe the training progress.
- Optionally, upload test data or a pretrained model for predictions.
- Download results if desired.

4) LLM-Based Q&A (RAG System)

- From the **Services** tab, choose **Large Language Model**.
 - Type in your question related to Academic City's student policy.
 - The system retrieves relevant document snippets and uses an AI model to generate a detailed response.
 - Each answer is accompanied by a confidence score based on the semantic match.
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■ DATASETS & MODELS USED

Datasets

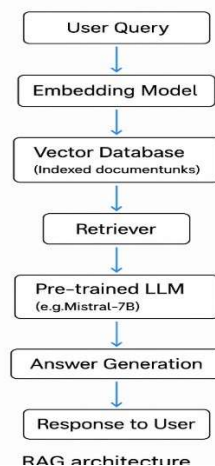
- The Q&A module uses the **Academic City Student Policy PDF**.
- **User-provided CSVs** power the Regression, Clustering, and Neural Network modules.

Machine Learning Models

- **Linear Regression** from Scikit-learn
 - **KMeans Clustering** using Scikit-learn
 - **Feedforward Neural Network** built with Keras
 - **Mistralai/Mistral-7B-Instruct-v0.1** accessed via HuggingFace for the LLM RAG component, combined with **FAISS** for semantic retrieval.
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□ ARCHITECTURE FOR THE LLM RAG SYSTEM DESIGN

- This system merges document chunk retrieval (via FAISS indexing) with a large language model (LLM) to ensure accurate, contextually grounded answers.
- Confidence scores are generated based on the semantic similarity between the query and document chunks.



□ METHODOLOGY

Document Preparation

- The Student Policy PDF is parsed and cleaned by removing digits, extra spaces, and then split into standalone sentences.

Vectorization & Indexing

- Each sentence is transformed into a TF-IDF vector.
- These vectors are indexed using FAISS for rapid similarity searches.

Handling User Queries

- When a user submits a query, it's also vectorized and compared against the index to find the top 3 most relevant sentences.
- These matches are used to calculate confidence scores using distance metrics.

Generating Answers

- A combined prompt of the question and relevant document context is passed to the **Mistral-7B model** through HuggingFace's API.
- The AI generates a response based on the specific context retrieved.

User Output

- The answer is shown along with the associated confidence level.
- A history of past questions and responses is maintained on the interface.

■ EVALUATION AND BENCHMARKING

LLM RAG System Performance

- This approach ensures the answers are policy-specific, leveraging the actual content of the document.
- Confidence scores help measure how well the retrieved context matches the user's query.

Compared with ChatGPT

- Unlike ChatGPT, which responds generally unless explicitly given the document, this system pulls directly from the student policy.

- The responses from LLM RAG are grounded in the document and provide more accurate, relevant answers for policy-related queries.
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✦ EXAMPLES

Query	ChatGPT Reply	LLM RAG Reply
<i>What is the dress code?</i>	“Usually includes modest clothing.”	“The policy clearly states formal attire must be worn on designated days.”
<i>Who to contact for absence?</i>	“Typically your professor or department head.”	“The policy advises contacting the Registrar’s office via email at acity.edu.gh.”

✓ **Conclusion:** LLM RAG offers more accurate and policy-specific responses, significantly outperforming generic models like ChatGPT when it comes to document-grounded Q&A.
