Transforming Tabular Data: Leveraging Prompt Engineering with LLMs for Scalable Feature Engineering

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Competition Category: Role of Prompt Engineering for Tabular Data Tasks

Challenge: Develop solutions that leverage the potential of large language models (LLMs) to transform how we interact with, analyse, and derive insights from tabular data.

Ideathon Platform: MachineHack (Nov 29 - Dec 22, 2024)

Project Name: Runnable Application for Feature Engineering Using Prompt Engineering

Here's a step-by-step guide to running the provided Python script:

Step 1) Prerequisites: Install required libraries using pip in the Anaconda Prompt

```
pip install openai pandas streamlit
```

Step 2) Save the Script:

2.1) Open a text editor (e.g., Notepad).

2.2) Copy the python code provided in the file and paste it into the editor.

2.3) Save the file with the name "feature_engineering_app.py".

Step 3) Set Up OpenAI API Key: Replace "your_openai_api_key" in the script with your actual OpenAI API key

```
openai.api_key = "your_openai_api_key"
```

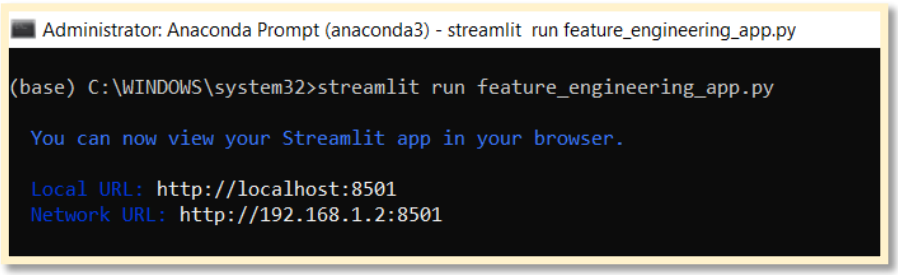
Step 4) Run the Application:

4.1) Open an Anaconda Prompt & Navigate to the directory where you saved the feature_engineering_app.py file.

4.2) Start the Streamlit App: Run the following command to start the app

```
streamlit run feature_engineering_app.py
```

4.3) Access the Application in Your Browser: After running the above command, Streamlit will display a local URL in your terminal, something like this:

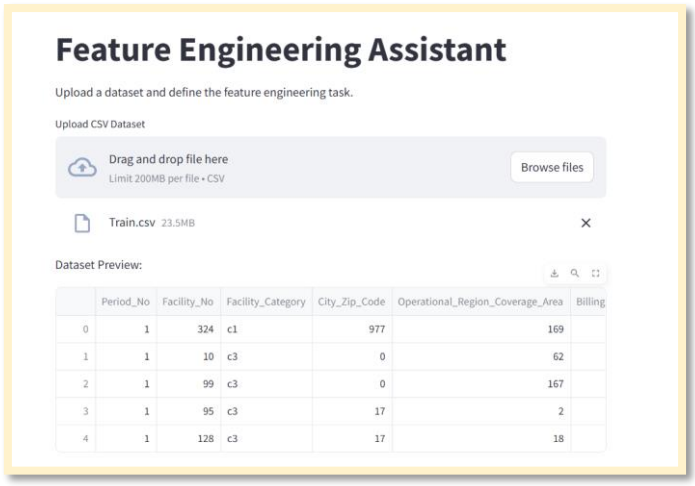


Open the Local URL in your browser to interact with the application.

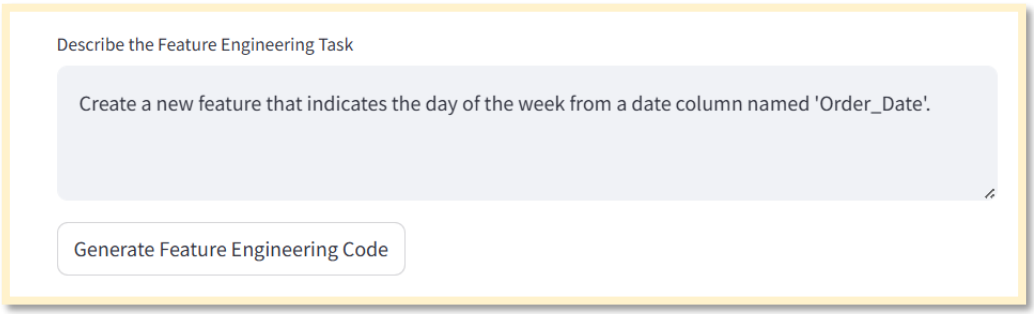
Step 5) Using the App:

a) **Upload Dataset:** Click on the "Browse files" button in the app to upload a CSV file.

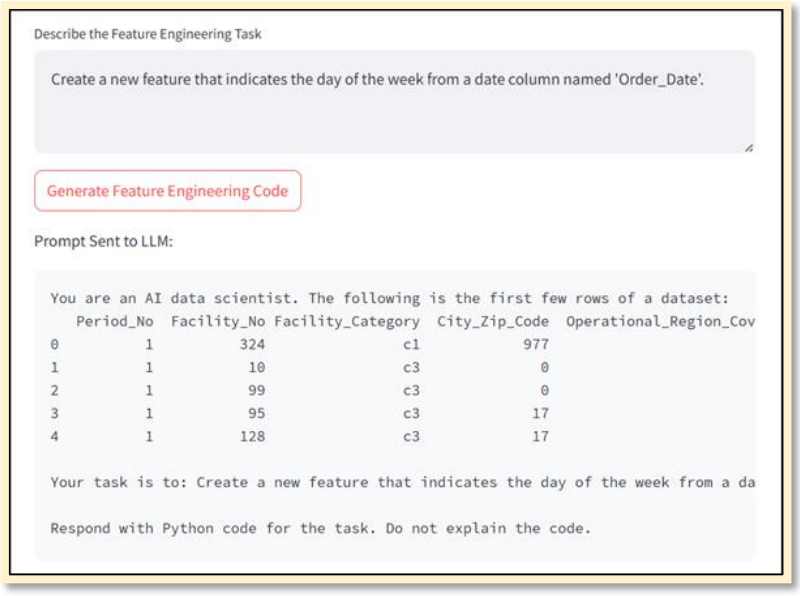
Example CSV:



b) **Describe the Feature Engineering Task:** Enter your desired feature engineering instruction in the text area, such as:



c) **Generate Code:** Click on the "Generate Feature Engineering Code" button.



The app will display the prompt sent to the LLM, the Python code generated, and (optionally) the transformed dataset.

Results: This application demonstrates the integration of LLMs with prompt engineering to automate feature engineering, highlighting its scalability across various tasks. It provides an accessible, user-friendly interface to enhance the productivity of data analysts and non-technical users.