### HCLTECH HACKATHON

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# Assistant chatbot

#### category

I first ensure that the model returns correct category given the complaint feedback from the user.

I use pydantic model to ensure return type of a json as shown.

<u>I use a smaller gpt4o-mini</u> model for it

```
> ~ # □ ···
categorization.py U X
complain_context_generator.py U
                                                              category_processor.py U
                                                                                            query_category.py U
categorization.py > ...
       def categorize_query(query: str) -> Union[QueryCategory, None]:
               llm_response = response.choices[0].message.content
               print(llm response)
               return process_category_response(llm_response)
           except Exception as e:
               print(f"Error calling LLM API: {str(e)}")
               return None
       def main():
           # examples taken from the spreadsheet data to test
           sample_queries = [
               "I am satisfied with the low minimum balance requirement for my Savings Account.",
               "I am satisfied with the security features of online banking.",
               "I am satisfied with the easy online account opening for my Savings Account.",
               "I am satisfied with the easy fund transfer from my Current Account."
           for query in sample queries:
               category = categorize_query(query)
               if category:
                   print(f"Quary, Javaryl")
                                                                                                           ∑ zsh ∧ + ∨ ∏ 前 ··· ^ ×
                      COMMENTS PORTS DEBUG CONSOLE TERMINAL
 ~/Work/hcltech-hackathon // main ● ? python3 categorization.py
                                                                                                                      7789 14:56:26
     "category": "Savings Account",
     "confidence": 0.95
Query: I am satisfied with the low minimum balance requirement for my Savings Account.
Category: Savings Account
 Confidence: 0.95
    "category": "Online Banking",
    "confidence": 0.85
 Query: I am satisfied with the security features of online banking.
Category: Online Banking Confidence: 0.85
    "category": "Savings Account",
    "confidence": 0.95
Query: I am satisfied with the easy online account opening for my Savings Account.
 Category: Savings Account
Confidence: 0.95
    "category": "Current Account",
     "confidence": 0.95
Query: I am satisfied with the easy fund transfer from my Current Account.
Category: Current Account
Confidence: 0.95
Execution time: 6.37s
```

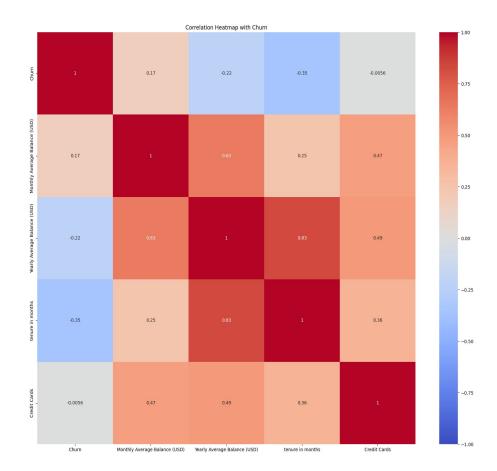
## Chatbot responses

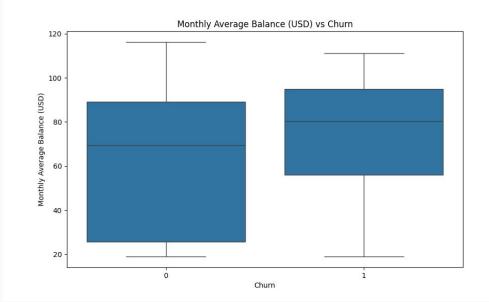
I use DYNAMIC PROMPT to select data points from the spreadsheet based on the user's query and then use the OpenAI gpt-4o for generating responses.

```
main.py ×
main.py > ..
       # Read the Excel file into a pandas DataFrame
       CUSTOMER_CHURN_DATA_XLSX = './given/customer_churn_data.xlsx'
       df = pd.read excel(CUSTOMER CHURN DATA XLSX)
       client = get_openai_client()
      # iterate
       for index, row in df.iterrows():
           customer_feedback = str(row['Customer_Feedback'])
           # Generate the prompt for the LLM
           final_prompt = generate_prompt(complaint=customer_feedback, row=row)
               response = client.chat.completions.create(
                       messages=[{
                           "role": "user".
                           "content": final prompt,
                       model="apt-40".
               final_response = response.choices[0].message.content
           except Exception as e:
               print(f"Error calling LLM API: {str(e)}")
           # Store the response in the 'Chatbot Response' column
           df.at[index, 'Chatbot Response'] = final_response
           print(f"{index} done")
      df.to_csv("given/new_data.csv")
                      COMMENTS PORTS
                                            DEBUG CONSOLE TERMINAL
85 done
86 done
87 done
91 done
92 done
93 done
94 done
95 done
98 done
99 done
100 done
101 done
102 done
103 done
104 done
```

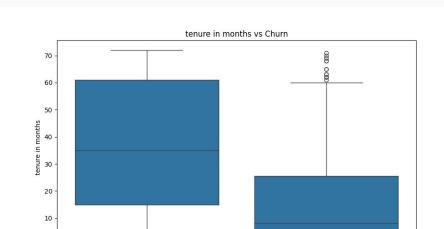
#### churn

Since churn is already given as a yes/no in the dataset.. So I decided to correlate it with the most relevant factors eg. credit cards, monthly & yearly average balances.





more plots



Thanks!