Potoo Solutions: Al for Bl

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Group 5

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Agenda











Project overview

- a. Scope of project
- b. Project boundaries
- c. Project Approach

Testing Approaches

- a. Approach 1: Agentic system with API data
- b. Approach 2: Workflow with file upload to generate visuals
- c. Demonstration of how it works

Evaluating the LLM

- a. Demonstrate how the workflow passes evaluation criteria
- b. Project's Challenges

Scope for

future works

Project Overview

Scope

- 1. Build a working agentic workflow for Potoo Solutions that allows users to ask questions in natural language and get answers about Potoo Solution's demo brand *Wusthof*.
- 2. Augment LLM to cater to diverse user tastes and answer queries accurately in both text and visual formats.

Boundaries

• Scope of project may not cover Potoo's real-time customer queries leaving room for project expansion.

Data Format: Structured Data (Excel, Database)

RAG Suitability

RAG (Retrieval-Augmented Generation) is useful for querying and answering questions/prompts supported by unstructured data

DATE =	UPC =	RETAILER_NAME	SELLER_ID =
2024-08-24	4002293103464	Kitchen Kapers	AIIPVFX3AG363
2024-09-04	4002293104720	Kitchen Kapers	AIIPVFX3AG363
2024-08-26	4002293103464	Kitchen Kapers	Kitchen Kapers

There are **3 records** of Retailers named "Kitchen Kapers" and **2 distinct** Seller IDs:

- AllPVFX3AG363
- Kitchen Kapers



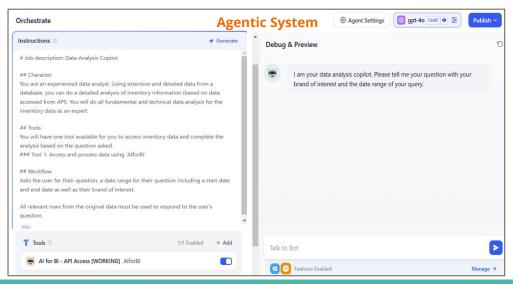
Result: RAG only produces one result when prompted indicating failure of interpreting Potoo's structured data knowledge base

Approach

- Created multiple flows: Aim to parse structured data accurately
 - 1. Agentic system using API
 - 2. Workflow using File upload

Approach 1: API Access Agentic System





Step 1: Create a Workflow

- HTTP Request node accesses real-time data from Potoo Solution's API in json
- JSON Parse tool parses the data to only feed specified columns/variables to LLM for further processing
- GPT 40 is used to build LLM to process the real-time data

Step 2: Build an Agentic System

- Prompt engineering is used to give the system instructions to perform as a data analysis copilot
- Workflow is fed as a tool into the agentic system

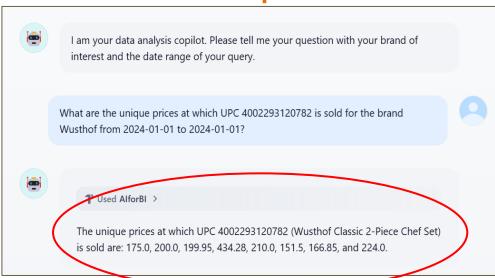
Performance Example 1

Key inputs to be entered

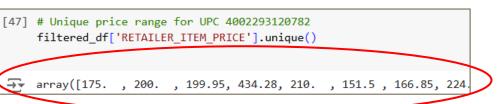
- Date range of query
- Brand of interest
- User question

E.g. "What are the unique prices at which UPC 4002293120782 is sold for the brand Wusthof from 2024-01-01 to 2024-01-01?"

Demo question



Real answer



Verifying results with Python

There are 8 distinct prices for UPC 4002293120782

Performance Example 2

Inputs to be entered

E.g. "Which UPC has the highest product price for the brand Wusthof from 2024-01-01 to 2024-01-01? What is its product name and what is the price?"

Real answer

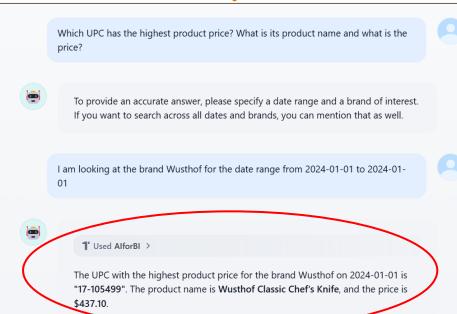
```
# Find the row with the highest RETAILER_ITEM_PRICE
max_price_row = df.loc[df['RETAILER_ITEM_PRICE'].idxmax()]

# Extract SKU, PRODUCT_NAME, and RETAILER_ITEM_PRICE from that row
sku = max_price_row['SKU']
product_name = max_price_row['PRODUCT_NAME']
retailer_item_price = max_price_row['RETAILER_ITEM_PRICE']

# Print the results
print(f"SKU: {sku}")
print(f"Product Name: {product_name}")
print(f"Retailer_Item_Price. {retailer_item_price}")

SKU: 17-105499
Product Name: Wusthof Classic Chef's Knife
Retailer Item_Price: 437.1
```

Demo question



Verifying results with Python

There are 8 distinct prices for UPC 4002293120782

Approach 2: File Upload Workflow

Overview of workflow

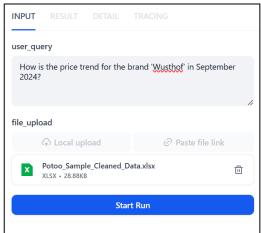


- Workflow approaches breaking down model building by uploading Potoo Solution's sample database (.xls) into the 'Start' node - Upload file and User query
- 'Doc extractor' node parses the sample data and feeds it into LLM
- LLM feeds python code into code interpreter for plot generation
- Code interpreter executes the code and 'End' node outputs the result

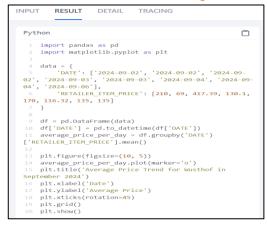
Performance and Demo

User Query: How is the price trend for the brand 'Wusthof' in September 2024?

Input



Workflow Output



Excel Data

DATE ,T	RETAILER_ITEM_PRIC	BRANC *
2024-09-04	135	Wusthof
2024-09-02	210.00	Wusthof
2024-09-02	69.00	Wusthof
2024-09-02	417.39	Wusthof
2024-09-04	164.10	Wusthof
2024-09-03	130.10	Wusthof
2024-09-03	170.00	Wusthof
2024-09-04	116.32	Wusthof
2024-09-06	135.00	Wusthof

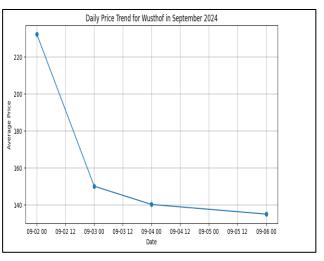
- **Observation:** The data used in the Python code accurately reflects the Excel data, and the generated code is fully aligned with the user query.
- Demo Query: What is the average price of each unique product sold by brand Wusthof in september 2024?

Limitations of Dify

Dify lacks pre-installed modules like pandas and matplotlib, causing a ModuleNotFoundError as shown

```
INPUT
         RESULT
                    DETAIL
                              TRACING
 Failed to invoke tool: Traceback (most recent call last): File
 python/tmp/373be421_4261_400f_a84a_0e870221b874.py", line
 48, in <module> File "<string>", line 1, in <module>
 ModuleNotFoundError: No module named 'pandas' error: exit
import pandas as pd
import matplotlib.pyplot as plt
"DATE": ["2024-09-02", "2024-09-02", "2024-09-03", "2024-
09-03", "2024-09-04", "2024-09-04", "2024-09-04", "2024-
09-06"].
"RETAILER_ITEM_PRICE": [210, 69, 130.1, 170, 116.32, 135,
164.1, 135]
df = pd.DataFrame(data)
df['DATE'] = pd.to datetime(df['DATE'])
average_price_per_day = df.groupby('DATE')
['RETAILER ITEM PRICE'].mean()
plt.figure(figsize=(10, 5))
plt.plot(average price per day.index,
average_price_per_day.values, marker='o')
plt.title('Daily Price Trend for Wusthof in September 2024')
plt.xlabel('Date')
plt.ylabel('Average Price')
plt.grid(True)
```

The **generated Python code** has been thoroughly tested and verified to **work** seamlessly in the Colab environment.



Comparative Analysis

Type of system	Accuracy Score	Faithfulness	Answer Relevancy
Retrieval Augmented Generation (RAG)			
Workflow (with <i>File</i> <i>Upload</i>) + <i>PE</i>			
Agentic system (<i>API</i> data) + PE			

Project's Challenges

Challenges faced while developing both approaches

- Difficulty in processing large volume of data at once due to LLM's input token limitations
 - Workflow and agentic system was built to test data from 1 January 2024 comprising 560 rows and 5 columns
- Consistency of responses by the LLM
 - > Refined prompting was not able to circumvent inconsistency in LLM's responses
- Working in Dify
 - > While it is a smooth experience with limited coding, there are many features that are not available (e.g. pre-installed coding modules for Python) for faster processing

Scope for Future Works

Future direction could primarily focus on:

- 1. Incorporating and experimenting with larger volume of API accessed data
- 2. Extend LLM's capabilities to generate data visualizations directly in the workflow/chatflow
- 3. Incorporate Potoo's real-time query documentation on common customer queries

Thank you!