Chatvily

Introduction

In today's digital era, gathering accurate, relevant, and well-organized information is crucial. We propose a **Deep Research Agentic System** that crawls online data using **Tavily**, processes it using **LangChain** and **LangGraph**, and structures it into a highly professional, clear output.

The system uses a dual-agent mechanism:

- Research Agent: Searches and collects the best information.
- Answer Drafting Agent: Structures, refines, and critiques the final output.

The system ensures that the information is relevant, fact-checked, clear, and engaging.

Unique Aspects

- Reranker Layer: Ensures only the most relevant search results are considered.
- Critic Agent: Conducts a quality check on clarity, relevance, and factual accuracy.
- Tone Control: Maintains a professional yet engaging communication style.
- Self-Correcting: If the output fails quality checks, it auto-corrects by re-searching.
- Scalable: Can be expanded with additional specialized agents.

Detailed System Explanation

1. Setting API Keys

```
Securely connect to Tavily (for intelligent web search) and OpenAI (for LLM tasks).

python

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os.environ["TAVILY_API_KEY"] = "TAVILY KEY"

os.environ["OPENAI_API_KEY"] = "OPENAI KEY"
```

2. Tool Initialization

- TavilySearchResults: Performs the web search.
- CrossEncoder: Reranks results based on relevance.
- ChatOpenAI: Formats and critiques the responses.

```
search_tool = TavilySearchResults()
reranker = CrossEncoder('cross-encoder/ms-marco-MiniLM-L-6-v2')
llm = ChatOpenAl(model="gpt-3.5-turbo")
```

3. Research Agent - Search Function

```
Uses Tavily to search for user queries.

def search_fn(query):

results = search_tool.invoke({"query": query})

return results
```

4. Research Agent - Rerank Function

return reranked[:3]

```
def rerank_fn(query, results):
    pairs = [(query, item["content"]) for item in results]
    scores = reranker.predict(pairs)
    reranked = [item for _, item in sorted(zip(scores, results), key=lambda x: x[0], reverse=True)]
```

Pairs the user query with each search result and reranks them based on relevance.

5. Answer Drafting Agent - Formatting Function

return critique.content

```
Summarizes the best results into a professional, engaging tone.
def format_fn(reranked_results):
  prompt = ChatPromptTemplate.from_template("""
Format the following search results into a clean, professional tone.
Make it engaging but formal. Summarize the key points clearly.
{results}
  """)
 formatted = llm.invoke(prompt.format_prompt(results=reranked_results).to_messages())
  return formatted.content
6. Answer Drafting Agent - Critic Function
Checks if the output is clear, professional, consistent, and relevant.
def critic_fn(formatted_result):
  prompt = ChatPromptTemplate.from_template("""
Act as a quality critic. Does the following response meet these rules:
- Is it clear and professional?
- Is it factually consistent?
- Is it relevant to the original query?
Respond "PASS" if all good, otherwise "FAIL" with reason.
Response:
{response}
  ("""
 critique = llm.invoke(prompt.format_prompt(response=formatted_result).to_messages())
```

7. Main Logic Controller

```
Coordinates the entire workflow: search \rightarrow rerank \rightarrow format \rightarrow critique \rightarrow final output. If quality fails, it auto-repeats.
```

```
def main_logic(query):
    search_results = search_fn(query)
    reranked = rerank_fn(query, search_results)
    formatted = format_fn(reranked)
    critique = critic_fn(formatted)

if "FAIL" in critique:
    print(" Critic said FAIL. Re-running search...")
    return main_logic(query)

else:
    print(" Critic said PASS. Final result ready!")
    print(formatted)
    return formatted
```

8. Run Everything

```
User enters a query → Full pipeline runs.

user_query = input("Enter your search query: ")

final_result = main_logic(user_query)
```