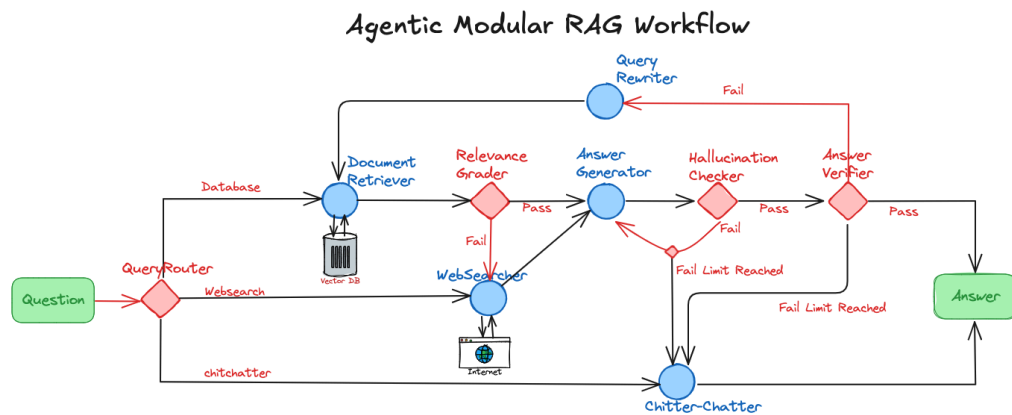


Lab 7: Building an Agentic RAG Workflow with LangGraph

This lab guides you through the design and implementation of a modular, agentic Retrieval-Augmented Generation (RAG) system using LangGraph.



Core Features Implemented

- Multi-Agent Collaboration

Specialized agents (e.g., Query Router, Answer Generator, Verifier) work together to deliver robust, high-quality responses.

- Adaptive Retrieval Routing

Inspired by Adaptive RAG, the system dynamically selects between vectorstore, web search, or fallback dialogue based on the question's intent.

- Fallback Mechanism (Corrective RAG)

Based on Corrective RAG, the system falls back to web search if retrieved documents are weak or irrelevant.

- Self-Correction & Reflection

Following Self-RAG, the system performs hallucination detection and answer usefulness checks, triggering retries or rewrites when needed.

- Performance Considerations

- Efficiency: Uses async document grading to speed up multi-query retrieval.
- Failure Handling: Implements structured retry logic with maximum retry limits.

The lab consists of three main phases:

1. Define Agent Functions.
2. Build the Workflow with LangGraph.
3. Run the Workflow.

```
In [ ]: # %pip install -Uq langchain langchain_community langgraph
```

[notice] A new release of pip is available: 23.0.1 -> 25.0.1

[notice] To update, run: pip install --upgrade pip

Note: you may need to restart the kernel to use updated packages.

```
In [4]: # Environment Configuration
from dotenv import load_dotenv
import os
from sqlalchemy.engine.url import make_url # Used to parse and construct
from langchain_postgres.vectorstores import PGVector # Integration with P

# LLM and Core LangChain Tools
from langchain_openai import OpenAIEmbeddings
from langchain_openai import ChatOpenAI

from langchain_core.messages import HumanMessage, SystemMessage
from langchain.load import dumps, loads # Serialize/deserialize LangChai
from langchain_core.prompts import PromptTemplate
from langchain_core.output_parsers import StrOutputParser
from langchain_core.documents import Document # Standard document forma

from typing_extensions import TypedDict # Define structured types for sta
from typing import List # Specify types for list inputs or outputs
import asyncio # Support asynchronous execution for parallel LLM calls

from langgraph.graph import StateGraph, END # LangGraph tools to define s

# Visualization and Display Utilities
import textwrap
from IPython.display import Markdown, Image
from pprint import pprint

# Web Search Tool
from langchain_community.tools.tavily_search import TavilySearchResults
```

```
In [2]: # Load environment variables from .env file
load_dotenv()

# Access the environment variable
openai_api_key = os.getenv("OPENAI_API_KEY")
connection_string = os.getenv("DB_CONNECTION")
tavily_api_key = os.getenv("TAVILY_API_KEY")

# Configure Database Connection
# Use the same shared table as from the last lab
shared_connection_string = make_url(connection_string)\
    .set(database="IST345_Drucker_data").render_as_string(hide_password=F

# Initialize the embedding model
embedding_model = OpenAIEmbeddings(model="text-embedding-3-large")

# Quick check environment variables
if not openai_api_key or not shared_connection_string or not tavily_api_k
    (f"Error: Missing one or more required environment variables") #
```

```
else:
    print("All environment variables loaded successfully")
```

All environment variables loaded successfully

```
In [4]: # Main LLM for handling complex or creative tasks
llm_gpt = ChatOpenAI(
    model="gpt-4o", # GPT-4o is a powerful model with strong reasoning ca
    temperature=0.7,
    api_key = openai_api_key
)

# Lightweight LLM for simple or deterministic tasks
llm_gpt_mini = ChatOpenAI(
    model="gpt-4o-mini", # Smaller, faster variant for lightweight ta
    temperature=0, # Temperature 0 = fully deterministic output
    api_key = openai_api_key
)

# Connect to the PGVector Vector Store that contains book data.
book_data_vector_store = PGVector(
    embeddings = embedding_model,
    collection_name = "Book_data", # Name of the collection/table in th
    connection=shared_connection_string, # Use shared DB connection from
    use_jsonb=True,
)
```

1. Define Agents

In this section, we define the different agents used in the workflow.

1.1 Query Router Agent

The Query Router is an LLM-powered decision-making agent that analyzes a user's question and determines the most appropriate information source to handle it.

It uses a structured prompt to classify the query into one of three categories:

1. A vector search for questions answerable by the existing content in our Drucker book collection.
2. A web search for in-scope questions that require more factual, comprehensive, or real-time information.
3. A fallback "Chitter-Chatter" agent for out-of-scope question.

```
In [5]: # Define the routing prompt
query_router_prompt_template = PromptTemplate.from_template("""
You are an expert at analyzing user question and deciding which data sour

1. **Vectorstore**: Use this if the question can be answered by the **exi
The vectorstore contains information about **{vectorstore_content_summ

---

2. **Websearch**: Use this if the question is **within scope** (see below
- The answer **cannot** be found in the local vectorstore
```

```

- The question requires more detailed or factual information than
- The topic is time-sensitive , current, or depends on recent

---

3. Chitter-Chat: Use this if the question:
- Is not related to the scope below, or
- Is too broad, casual, or off-topic to be answered using vectorstore

Chitter-Chat is a fallback agent that gives a friendly response and

---

Scope Definition:
Relevant questions are those related to {relevant_scope}

---

Your Task:
Analyze the user's question. Return a JSON object with one key "DataSource"

"""

# Define a summary of what's in the vectorstore
# This helps the LLM understand what kind of information is already available
# It gets embedded into the prompt above dynamically.
vectorstore_content_summary = """
Peter Drucker's "The Daily Drucker" (2004) provides 366 daily insights and
serving as a practical guide for personal and professional growth. "The 40
enable executives to achieve effectiveness, focusing on time management,
is a curated collection of Drucker's foundational principles on management
complex business environments. These books collectively address key aspects
resources for understanding organizational dynamics and personal productivity

"""

# Define the topical scope of the system
# This defines what topics are considered "in-scope" for this assistant or
# It also gets embedded into the prompt above dynamically.
relevant_scope = """Peter Drucker-related topics, including his management
and their applications in modern business contexts"""

# Format the prompt using the content summary and scope
# The final formatted prompt is what will be sent to the LLM to determine
query_router_prompt = query_router_prompt_template.format(
    relevant_scope = relevant_scope,
    vectorstore_content_summary = vectorstore_content_summary
)

```

```

In [6]: # Test Query Router

# Example user questions for testing
questions = [
    "Who is Peter Drucker, and where was he born?",
    "What did Drucker say about knowledge workers in the book?",
    "How are you doing today?"
]

print(f"\n{'='*40} Testing the Router Prompt {'='*40}\n")

```

```

for q in questions:
    response = llm_gpt.with_structured_output(method="json_mode").invoke(
        [SystemMessage(content=query_router_prompt), # Use the formatted
         HumanMessage(content=q)] # Feed in the user query
    )

    # Extract and print the decision
    print(f"Question: {q}")
    print(f"Datasource: {response['Datasource']}")
    print("-" * 50)

```

===== Testing the Router Prompt =====
=====

Question: Who is Peter Drucker, and where was he born?
Datasource: Websearch

Question: What did Drucker say about knowledge workers in the book?
Datasource: Vectorstore

Question: How are you doing today?
Datasource: Chitter-Chatter

1.2 Document Retriever Agent

This agent retrieves relevant documents using a multi-step retrieval strategy:

- Multi-query generation: Rewrites the user's question in different ways to improve recall.
- MMR retrieval: Finds diverse, relevant documents for each query variant.
- Reciprocal Rank Fusion: Combines and reranks results to prioritize the most consistently relevant documents.

The result is a high-quality, reranked list of documents optimized for use in the workflow.

Step 1. Query translation

Use an LLM to generate multiple alternative versions of a user query.

Because semantic similarity search (in vector DBs) can miss relevant results, we use query translation to improve recall.

```

In [14]: # Define the multi-query generation prompt
# The prompt gives the LLM both context about the vectorstore and a speci

multi_query_generation_prompt = PromptTemplate.from_template("""
You are an AI assistant helping improve document retrieval in a vector-ba
---

**Context about the database**
The vectorstore contains the following content:
{vectorstore_content_summary}

```

Your goal is to help retrieve **more relevant documents** by rewriting a This helps compensate for the limitations of semantic similarity in vecto

Instructions:

Given the original question and the content summary above:

1. Return the **original user question** first.
2. Then generate {num_queries} **alternative versions** of the same quest
 - Rephrase using different word choices, structure, or focus.
 - Use synonyms or shift emphasis slightly, but keep the original mean
 - Make sure all rewrites are topically relevant to the database conte

Format requirements:

- Do **not** include bullet points or numbers.
- Each version should appear on a **separate newline**.
- Return **exactly {num_queries} + 1 total questions** (1 original + {num

Original user question: {question}
""")

Create the query generation pipeline

```
multi_query_generator = (  
    multi_query_generation_prompt    # The prompt defines what the LLM sho  
    | llm_gpt                        # An LLM generates query variants  
    | StrOutputParser()              # Parses the raw output as a string  
    | (lambda x: x.split("\n"))      # A lambda function to split the result  
)
```

In [15]: *# === Test Query Translation ===*

Run the query generator with example input

```
test_query_translation = multi_query_generator.invoke(  
    {"question": "What did Drucker say about knowledge workers in the book?",  
     "num_queries": 3,    # number of alternative versions to generate  
     "vectorstore_content_summary": vectorstore_content_summary  
})
```

Print results in a readable format

```
print(f"\n{'='*40} Testing the Query Translation Prompt {'='*40}\n")
```

```
for i, query in enumerate(test_query_translation):  
    if i == 0:  
        print(f"Original query: {query}")  
    else:  
        print(f"Generated query {i}: {query}")
```

===== Testing the Query Translation Pro
mpt =====

Original query: What did Drucker say about knowledge workers in the book?
Generated query 1: How does Drucker address the concept of knowledge worke
rs within his writings?
Generated query 2: In his works, what insights does Drucker offer regardin
g knowledge workers?
Generated query 3: What are Drucker's thoughts on the role of knowledge wo
rkers as discussed in his books?

Step 2. RAG fusion

This section implements a RAG Fusion retrieval strategy that is similar to the Lab 6.

Each version of the query from the previous step is used to retrieve documents from the vectorstore using MMR (Maximal Marginal Relevance).

Then the results from all query variations are reranked using Reciprocal Rank Fusion.

```
In [16]: # Reciprocal Rank Fusion (RRF) Implementation
def reciprocal_rank_fusion(results, k=60):
    fused_scores = {} # Dictionary to store cumulative RRF scores for ea

    # Iterate through each ranked list of documents
    for docs in results:
        for i, doc in enumerate(docs):
            doc_str = dumps(doc) # Convert document to a string format

            # Initialize the document's fused score if not already present
            if doc_str not in fused_scores:
                fused_scores[doc_str] = 0

            # Apply RRF scoring: 1 / (rank + k), where rank is 1-based
            rank = i + 1 # Adjust rank to start from 1 instead of 0
            fused_scores[doc_str] += 1 / (rank + k)

    # Sort by cumulative RRF score (descending)
    reranked_results = sorted(fused_scores.items(), key=lambda x: x[1], r

    # Convert JSON strings back to Document objects and store RRF scores
    reranked_documents = []
    for doc_str, score in reranked_results:
        doc = loads(doc_str) # Convert back to Document object
        doc.metadata["rrf_score"] = score # Track how the document was r
        reranked_documents.append(doc)

    # Return the list of documents with scores embedded in metadata
    return reranked_documents
```

```
In [17]: # Define a retrieval chain for Multi-Query RAG fusion

retrieval_chain_rag_fusion_mmr = (
    multi_query_generator
    | book_data_vector_store.as_retriever(
        search_type="mmr", # Use MMR retrieval to enhance diversity in re
        search_kwargs={
            'k': 3, # Final number of documents to return per
            'fetch_k': 15, # Initial candidate pool (larger forbett
            "lambda_mult": 0.5 # Balances relevance (0) and diversity (1
        }
    ).map() # Apply MMR retrieval to each reformulated query
    | reciprocal_rank_fusion # Rerank the combined results using RRF
)
```

```
In [18]: # === Test the RAG Fusion Retrieval Pipeline ===

rag_fusion_mmr_results = retrieval_chain_rag_fusion_mmr.invoke({
```

```

    "question":"What did Drucker say about knowledge workers in the book?"
    "num_queries":3,
    "vectorstore_content_summary":vectorstore_content_summary}}

print(f"\n{'='*40} Testing the reranked retrieved results {'='*40}\n")

# Display the total number of retrieved and reranked documents
print(f"Total number of results: {len(rag_fusion_mmr_results)}")

# Iterate through the retrieved documents and display them in a structure
for i, doc in enumerate(rag_fusion_mmr_results, start=1):
    # Display metadata: Source and page number
    display(Markdown(f"\n **From `{doc.metadata['source']}`", page {doc.me

    # Display the score
    print(f"RRF score: {doc.metadata['rrf_score']:.4g}") # Display the sc

    # Print the document content with proper text wrapping for better rea
    print(textwrap.fill(doc.page_content, width=100))

    # Add a separator for each document
    print("-" * 80)

```

===== Testing the reranked retrieved results =====

Total number of results: 7

C:\Users\yuyum\AppData\Local\Temp\ipykernel_48028\1905607030.py:24: LangChainBetaWarning: The function `loads` is in beta. It is actively being worked on, so the API may change.

doc = loads(doc_str) # Convert back to Document object

From The Daily Drucker-2004.pdf , page 806

RRF score: 0.04788

. Drucker analyzes the new realities of strategy, shows how to be a leader in periods of change, and explains the "New Information Revolution," discussing the information an executive needs and the information an executive owes. He also examines knowledge-worker productivity, and shows that changes in the basic attitude of individuals and organizations, as well as structural changes in work itself, are needed for increased productivity. Finally, Drucker addresses the ultimate challenge of

From The Essential Drucker-2008.pdf , page 196

RRF score: 0.03279

. Knowledge workers, after all, first came into being in any substantial numbers a generation ago. (I coined the term "knowledge worker" years ago.) But also the shift from manual workers who do as they are being told—either by the task or by the boss—to knowledge workers who have to manage themselves profoundly challenges social structure

From The Effective Executive-2002.pdf , page 17

RRF score: 0.03279

. It takes his knowledge and uses it as the resource, the motivation, and the vision of other knowledge workers. Knowledge workers are rarely in phase with each other, precisely because they are knowledge workers. Each has his own skill and his own concerns. One man may be interested in tax accounting or in bacteriology, or in training and developing tomorrow's key administrators in the city government

From The Daily Drucker-2004.pdf , page 321

RRF score: 0.032

. 5. Productivity of the knowledge worker is not—at least not primarily—a matter of the quantity of output. Quality is at least as important. 6. Finally, knowledge-worker productivity requires that the knowledge worker be both seen and treated as an “asset” rather than a “cost.” It requires that knowledge workers want to work for the organization in preference to all other opportunities. ACTION POINT: Apply steps one through five to your knowledge work. Management Challenges for the 21st Century

From The Daily Drucker-2004.pdf , page 800

RRF score: 0.01613

. According to Peter Drucker, “This book tries to equip the manager with the understanding, the thinking, the knowledge, and the skills for today's and also tomorrow's jobs.” This management classic has been developed and tested during more than thirty years of management teaching in universities, executive programs, seminars, and through the author's close work with managers as a consultant for large and small businesses, government agencies, hospitals, and schools

From The Daily Drucker-2004.pdf , page 321

RRF score: 0.01613

23 May Knowledge-Worker Productivity Knowledge-worker productivity requires that the knowledge worker be both seen and treated as an asset rather than a cost. Work on the productivity of the knowledge worker has barely begun. But we already know a good many of the answers. We also know the challenges to which we do not yet know the answers. Six major factors determine knowledge-worker productivity. 1. Knowledge-worker productivity demands that we ask the question: “What is the task?”
2

From The Daily Drucker-2004.pdf , page 323

RRF score: 0.01587

. They know what steps are most important and what methods need to be used to complete the tasks;
and it is their knowledge that tells them what chores are unnecessary and should be eliminated. Work
on knowledge-worker productivity therefore begins with asking the knowledge workers themselves: What
is your task? What should it be? What should you be expected to contribute? and What hampers you in
doing your task and should be eliminated? The how only comes after the what has been answered

1.3 Relevance Grader Agent

This agent evaluates whether a retrieved document is relevant to a user's question.

It uses an LLM to determine if the document contains **keyword overlap** or **semantic alignment** with the query.

The output is a simple "pass" or "fail" label in JSON format, useful for:

- Filtering out low-quality documents in RAG pipelines.
- Scoring retrieval effectiveness.
- Constructing feedback loops for reranking or refinement.

```
In [19]: # Define the relevance grader prompt
relevance_grader_prompt_template = PromptTemplate.from_template("""
You are a a relevance grader evaluating whether a retrieved document is h

---

**Retrieved Document**:
{document}

**User Question**:
{question}

---

**Your Task**:
Carefully and objectively assess whether the document contains any **key
Do not require a full answer—just some relevant content is enough to pass

Return your decision as a JSON object with twith keys: "binary_score".
The "binary_score" should be "pass" or "fail" indicating relevance.
""")
```

```
In [20]: # === Test Relevant Grader ===

# Format the prompt with the first document from the reranked RAG fusion
relevance_grader_prompt = relevance_grader_prompt_template.format(
    document=rag_fusion_mmr_results[0].page_content,
    question="What did Drucker say about knowledge workers in the book?"
)
```

```
# Invoke the grading agent using a lightweight model
grader_result = llm_gpt_mini.with_structured_output(method="json_mode").invoke(
    relevance_grader_prompt)

# Print the grading result
print(f"\n{'='*40} Testing Relevance Grader Prompt {'='*40}\n")
print(grader_result)
```

```
===== Testing Relevance Grader Prompt =====
```

```
{'binary_score': 'pass'}
```

1.4 Answer Generator Agent

This agent uses context-relevant documents to generate an answer to the user's question.

It is grounded in retrieved content and includes a reference section at the end of the response.

```
In [21]: # Define the prompt template for answer generation
answer_generator_prompt_template = PromptTemplate.from_template("""
You are an assistant for question-answering tasks.

---

**Context**:
Use the following information to help answer the question:
{context}

****User Question**:
{question}

---

**Instructions**:
1. Base your answer primarily on the context provided.
2. If the answer is **not present** in the context, say so explicitly.
3. Keep the answer **concise**, **accurate**, and **focused** on the question.
4. At the end, include a **reference section**:
   - For book-based sources, use **APA-style citations** if possible.
   - For web-based sources, include **page title and URL**.

---

**Answer**:
""")
```

```
In [22]: # === Test Answer Generator ===

# Set up a test question and format the documents for input
test_question = "What did Drucker say about knowledge workers in the book

# Convert LangChain Documents into a simplified format (only metadata and
formatted_doc_results = [{"metadata": doc.metadata, "page_content": doc.p

# Format the final prompt with documents and question
```

```

answer_generator_prompt = answer_generator_prompt_template.format(
    context=formatted_doc_results, # Test with all documents
    question=test_question
)

# Generate an answer using the main LLM
answer_generation = llm_gpt.invoke(answer_generator_prompt)
print(f"\n{'='*40} Testing Answer Generator Prompt {'='*40}\n")

# Display the question and answer in a readable format
display(Markdown(f"**Question:** {test_question}\n"))
display(Markdown(f"**Answer:** {answer_generation.content}"))

```

```

===== Testing Answer Generator Prompt =
=====

```

Question: What did Drucker say about knowledge workers in the book?

Answer: Peter Drucker discussed several key aspects of knowledge workers in his writings. He emphasized that knowledge workers require a shift from being seen as a cost to being considered an asset to organizations. Productivity for knowledge workers is not just about the quantity of output but also the quality. Drucker also highlighted that knowledge workers should manage themselves and be involved in determining their tasks, understanding their contributions, and identifying obstacles in their work. He coined the term "knowledge worker" and noted the transition from manual work to knowledge work as a profound change in the social structure.

Reference: Drucker, P. F. (2004). *The Daily Drucker: 366 Days of Insight and Motivation for Getting the Right Things Done*.

Drucker, P. F. (2008). *The Essential Drucker*.

1.5 Hallucination Checker Agent

This agent evaluates whether an answer is factually grounded in the provided source documents.

It compares the generated answer against reference material and returns:

- A binary grade ("pass" or "fail")
- A brief explanation justifying the decision

It's useful for detecting hallucinations in LLM output and enforcing groundedness in your GenAI application.

```

In [23]: # Define the hallucination checker prompt
# The grader compares a student-style answer to reference materials (i.e.
# checks for accuracy, fabrication, or unsupported claims.

hallucination_checker_prompt_template = PromptTemplate.from_template("""
You are an AI grader evaluating whether a student's answer is factually g
---

**Grading Criteria**:

```

```

- **Pass**: The answer is **fully based** on the given FACTS and does not
- **Fail**: The answer information that is **fabricated**, **inaccurate**

---

**Reference Materials (FACTS)**:
{documents}

**Student's Answer**:
{generation}

---

**Output Instructions**:
Return a JSON object with keys: "binary_score" and "explanation".
- "binary_score": either `"pass"` or `"fail"`
- `"explanation"`: a short justification of the grading decision
""")

```

```

In [24]: # === Test Hallucination Checker Prompt ===

# Format the prompt with the provided documents and generated answer
hallucination_checker_prompt = hallucination_checker_prompt_template.format(
    documents=formatted_doc_results,
    generation=answer_generation.content
)

# Use the mini LLM model to grade the document
hallucination_checker_result = llm_gpt_mini.with_structured_output(method=
    hallucination_checker_prompt)

# Print the parsed JSON responses

print(f"\n{'='*40} Testing Hallucination Checker Prompt {'='*40}\n")

# Display the grade
display(Markdown(f"**Grade:** {hallucination_checker_result['binary_score']}"))

# Display the explanation
display(Markdown(f"**Explanation:** {hallucination_checker_result['explanation']}"))

===== Testing Hallucination Checker Prompt =====

```

Grade: pass

Explanation: The student's answer accurately reflects key concepts from Drucker's writings, including the importance of viewing knowledge workers as assets, the emphasis on quality over quantity in productivity, and the self-management of knowledge workers. All claims made in the answer are supported by the reference materials.

1.6 Answer Verifier Agent

This agent evaluates whether a generated answer meaningfully responds to the user's original question.

It does not assess factual accuracy; only whether the answer is relevant, responsive, and on-topic.

It returns a "pass" or "fail" grade, along with a brief explanation.

This is useful in the workflow to ensure answers stay aligned with user intent.

```
In [25]: # Define the answer verifier prompt
# The LLM is asked to judge whether the answer is responsive, relevant, a

answer_verifier_prompt_template = PromptTemplate.from_template("""
You are an AI grader verifying whether a student's answer correctly addre

---

**Grading Criteria**:
- **Pass**: The answer directly addresses the question, even if it includ
- **Fail**: The answer is off-topic, misses the point, or does not meanin

---

**question**:
{question}

**Student's Answer**:
{generation}

---

**Output Instructions**:
Return aJSONObject with keys: "binary_score" and "explanation"
- "binary_score": either `pass` or `fail`
- "explanation": a short justification for your grading decision
""")
```

```
In [26]: # === Test Answer Verifier Prompt ===

#Format the prompt with a test question and its generated answer
answer_verifier_prompt = answer_verifier_prompt_template.format(
    question=test_question,
    generation=answer_generation.content
)

# Use the mini LLM model to grade the document
answer_verifier_result = llm_gpt_mini.with_structured_output(method="json"
    answer_verifier_prompt)

# Print the parsed JSON responses
print(f"\n{'='*40} Testing Answer Verifier Prompt {'='*40}\n")

# Display the grade
display(Markdown(f"**Grade** {answer_verifier_result['binary_score']}"))

# Display the explanation
display(Markdown(f"**Explanation** {answer_verifier_result['explanation']}"))

===== Testing Answer Verifier Prompt =====
=====
```

Grade: pass

Explanation: The student's answer directly addresses the question by summarizing Drucker's views on knowledge workers, including their value to organizations, the importance of self-management, and the transition from manual to knowledge work. This response is relevant and informative.

1.7 Query Rewriter

This agent improves search effectiveness by rewriting user questions that previously led to incomplete or irrelevant answers.

It reviews both the original question and the failed answer to detect gaps, ambiguities, or missed keywords.

The output includes:

- A rewritten question optimized for vector retrieval
- An explanation of how the rewrite improves the search

This is useful for query recovery, workflow feedback loops, and retrieval optimization.

```
In [28]: # Define the query rewriter prompt

query_rewriter_prompt_template = PromptTemplate.from_template("""
You are a query optimization expert tasked with rewriting questions to im

---

**Context**:
- Original Question: {question}
- Previous Answer (incomplete or unhelpful): {generation}

**Vectorstore Summary**:
{vectorstore_content_summary}

Note: The summary provides context about what's in the database but shoul

---

**Your Task**:
Analyze the original question and the failed answer to identify:
1. What key information the original question was missing
2. Any ambiguities or unclear phrasing
3. Missing context or specialized terminology that should be included
4. Better keywords, phrasing, or terms to improve retrieval

---

**Output Format**:
Return a JSON object with keys: "rewritten_question" and "explanation".
- "rewritten_question": A refined version of the user's question optimiz
- "explanation": A short explanation of how the rewrite improves coverage
""")
```

```
In [29]: # === Test Query Rewriter Prompt ===

# Format the prompt with actual values
query_rewriter_prompt = query_rewriter_prompt_template.format(
    question=test_question,
    generation=answer_generation.content,
    vectorstore_content_summary = vectorstore_content_summary
)

# Use the LLM model to grade the document
query_rewriter_result = llm_gpt.with_structured_output(method="json_mode")(
    query_rewriter_prompt)

# Print the parsed JSON responses
print(f"\n{'='*40} Testing Query Rewriter Prompt {'='*40}\n")

# Display the original question
display(Markdown(f"*Original Question:* {test_question}"))

# Display the rewritten question
display(Markdown(f"*Rewritten Question:* {query_rewriter_result['rewritten_question']}"))

# Display the explanation
display(Markdown(f"*Explanation:* {query_rewriter_result['explanation']}"))

===== Testing Query Rewriter Prompt =====
```

Original Question: What did Drucker say about knowledge workers in the book?

Rewritten Question: What insights does Peter Drucker provide about the role and management of knowledge workers in 'The Daily Drucker' and 'The Essential Drucker'?

Explanation: The revised question specifies the books 'The Daily Drucker' and 'The Essential Drucker', which aligns with the context provided by the vectorstore summary, improving the retrieval accuracy by pinpointing the sources where Drucker discusses knowledge workers. It also clarifies the request for insights on the role and management of knowledge workers, using precise terminology that matches Drucker's focus areas, thus reducing ambiguity and enhancing relevance.

1.8 Chitter-Chatter Agent

This agent is a friendly fallback assistant that handles:

- Off-topic or unrelated questions
- In-scope but unanswerable questions
- Casual, social dialogue (e.g., greetings) with a friendly tone

It never fabricates or guesses answers. Instead, it acknowledges the user's input, maintains a warm tone, and gently redirects the conversation toward more relevant or answerable topics within the defined scope.

```
In [30]: # Define the Chitter-Chatter prompt
chitterchatter_prompt_template = PromptTemplate.from_template("""
    friendly assistant designed to keep conversations within the cu
```



```

---

**Current Scope**:
{relevant_scope}

Your job is to respond conversationally while gently guiding the user toward relevant topics.

---

**Response Guidelines**:

1. **Casual Chit-Chat**:
    - Respond warmly to greetings and social exchanges.
    - Maintain a natural, friendly tone.

2. **Off-Topic Questions**:
    - Politely acknowledge the question.
    - Mention that it falls outside your current scope.
    - Redirect to a relevant topic or ask a follow-up question within scope.
    - Avoid saying "I don't know" without offering guidance.

3. **In-Scope but Unanswerable Questions**:
    - If the question fits the scope but lacks enough information to answer, ask for more details.
    - Acknowledge the gap.
    - Avoid making unsupported claims.
    - Redirect the user toward a more specific or better-supported question.

---

**Important**:
Never invent or guess answers using general world knowledge.
Your job is to **maintain trust** by keeping the conversation focused and helpful.

Always end with a helpful redirection, question, or suggestion related to the current topic.
"""

```

```

In [31]: # === Test Chitter-Chatter Prompt ===

# Format the prompt using the defined relevant scope
chitterchatter_prompt = chitterchatter_prompt_template.format(relevant_scope=relevant_scope)

# Run a test with a casual or off-topic input
chitterchatter_response = llm_gpt_mini.invoke(
    [SystemMessage(chitterchatter_prompt),
     HumanMessage("How are you doing today?")]
)

print(f"\n{'='*40} Testing Chitter-Chatter Prompt {'='*40}\n")

# Display the result
display(Markdown(f"**Response** {chitterchatter_response.content}"))

===== Testing Chitter-Chatter Prompt =====

```

Response: I'm doing great, thank you! How about you? Is there something specific about Peter Drucker's management philosophy or leadership principles that you're interested in discussing today?

1.9 Web Searcher

This agent performs real-time web searches using Tavily's search API. This is the same Web Search tool you defined previously.

It is ideal for answering:

- In-scope questions that require additional detail beyond the vectorstore
- Questions involving current events, specific dates, or external references
- Queries needing factual grounding or external verification

This specific agent returns up to 5 high-quality search results, including:

- A short summary answer (if available)
- Title, URL, snippet, and a relevance score for each result

```
In [32]: # Define the web search tool

web_search_tool = TavilySearchResults(
    max_results=5,
    search_depth="advanced",          # Uses advanced search depth for more
    include_answer=True,              # Include a short answer to original
    tavily_api_key=tavily_api_key    # You have defined this API key in th
)
```

```
In [33]: # === Test Web Search Tool ===

# Test the tool with a real question
web_results = web_search_tool.invoke("What was Peter Drucker's greatest a

# Display the web search results in a structured format
print("\n" + "="*40 + " Web Search Results " + "="*40 + "\n")

for i, result in enumerate(web_results, start=1):
    display(Markdown(f"**Result {i}:**\n"))
    display(Markdown(f"**Title:** {result['title']}\n"))
    display(Markdown(f"**URL:** {result['url']}\n"))
    display(Markdown(f"**Content Snippet:** {result['content']}\n"))
    display(Markdown(f"**Relevance Score:** {result['score']:.3f}\n"))
    display(Markdown("\n" + "-"*80 + "\n"))
```

```
===== Web Search Results =====
=====
```

Result 1:

Title: What was Peter Drucker's greatest contribution to management?

URL: <https://www.quora.com/What-was-Peter-Druckers-greatest-contribution-to-management>

Content Snippet: Peter Drucker has been the father of modern management who has formulated the process of SMART & the theory of management by objectives(MBO).

Relevance Score: 0.751

Result 2:

Title: Peter Drucker's Management Theory Explained - Business.com

URL: <https://www.business.com/articles/management-theory-of-peter-drucker/>

Content Snippet: Peter Drucker was a world-famous management consultant whose ideas transformed business leadership from reactive to proactive.

Relevance Score: 0.443

Result 3:

Title: Peter F. Drucker | Biography, Management, Books, & Facts - Britannica

URL: <https://www.britannica.com/money/Peter-F-Drucker>

Content Snippet: He was also a leader in the development of management education, and he invented the concept known as management by objectives. Drucker, who

Relevance Score: 0.329

Result 4:

Title: why Peter Drucker is more relevant today than ever before

URL: <https://www.thepeoplespace.com/ideas/articles/technology-humanity-and-prosperity-why-peter-drucker-more-relevant-today-ever>

Content Snippet: He revolutionised several management theories and concepts — social entrepreneurship, second career, innovation, time management, decision making, employee

Relevance Score: 0.269

Result 5:

Title: About Peter Drucker * Drucker Institute

URL: https://drucker.institute/perspective__trashed/about-peter-drucker/

Content Snippet: That summer, he was awarded the Presidential Medal of Freedom, the nation's highest civilian honor. President Bush called Drucker "the world's foremost pioneer

Relevance Score: 0.208

2. Build the Agentic RAG Workflow with LangGraph

LangGraph is an extension of LangChain designed for building structured, agentic workflows using a graph-based execution model. It enables dynamic control flow, multi-agent collaboration, and persistent state tracking—making it ideal for complex RAG systems.

In this lab, we use LangGraph to construct a modular Agentic RAG pipeline that routes, retrieves, verifies, retries, and gracefully falls back when needed.

Core LangGraph Components Used in This Lab

- [`GraphState`] (Shared State Definition)

The first step in defining a graph is specifying its shared state. `GraphState` is the central data structure that flows between agents (nodes) and evolves during execution. In this lab, we define it using Python's `TypedDict`. It tracks:

- The original and rewritten question
- Retrieved documents
- LLM generations
- Retry counters
- Edge decision outcomes (e.g., relevance checks)

Note: This lab uses `TypedDict` for simplicity and clarity. LangGraph also supports Python's `Pydantic` (via `BaseModel`) for state definitions, which provides: "type enforcement", "validation", "default values", and "built-in serialization". `BaseModel` is especially useful in production settings when stricter control over state is required.

- `Nodes` (Agent Functions)

Functions that define what your agents do. Each node receives the current `GraphState` as input, perform some computation or side-effect (e.g., LLM call, retrieval), and returns an updated `State`. In short, `Nodes` do the work.

- `Edges`

Functions that determine which `Node` to execute next based on the current State. They can be conditional branches (Conditional Edges) or fixed transitions (Normal Edges). In short, `Edges` tell what to do next.

Note: To build your graph, you first define the `GraphState`, then add `nodes` and `edges`, and then compile it. You **MUST** compile your graph before it can be executed.

For a deeper dive, check out this talk: [Building Reliable Agents With LangGraph](#).

2.1 Define the Graph State

The first thing you do when you define a graph is define the `State` of the graph. The `StateGraph` class is the main graph class to use. This is parameterized by a user defined `State` object.

```
In [35]: # Define the state of the graph
class GraphState(TypedDict):
    """
    Graph state is a dictionary that contains information we want to prop
    """
    question: str # User question
    original_question : str # Copy of original question
    generation: str # LLM generation
    datasource: str # Output from router node: Vector
    hallucination_checker_attempts: int # Number of times hallucination
    answer_verifier_attempts: int # Number of times answer verifie
    documents: List[str] # List of retrieved documents fr
    checker_result: str # Result of document relevance c
```

2.2 Define Node Functions

This section defines the core functional units (nodes) of the LangGraph workflow. Each node includes a specific task in the Agentic RAG pipeline or fallback logic. Together, they enable the system to:

- Retrieve relevant documents using multi-query RAG fusion
- Generate grounded answers based on retrieved context
- Fall back to web search when local content is insufficient
- Handle off-topic or unsupported questions with a friendly Chitter-Chatter agent
- Rewrite ineffective queries to improve search quality
- Track retry attempts for hallucination and answer verification

```
In [36]: # ----- Document Retriever Node -----
def document_retriever(state):
    """
    Retrieves documents relevant to the user's question using multi-query

    This node performs the following steps:
    - Reformulates the original user question into multiple diverse sub-q
    - Executes MMR-based retrieval for each reformulated query.
    - Applies Reciprocal Rank Fusion (RRF) to combine and rerank results.
    - Filters out metadata fields that are internal (like RRF scores).
    - Prepares and returns a list of LangChain `Document` objects to be u

    Args:
        state (GraphState): The current state of the LangGraph, containin

    Returns:
        dict: A dictionary containing a cleaned list of relevant `Documen
    """
    print("\n---QUERY TRANSLATION AND RAG-FUSION---")

    question = state["question"]

    # Run multi-query RAG + MMR + RRF pipeline to get relevant results
    fusion_mmr_results = retrieval_chain_rag_fusion_mmr.invoke({
```

```

        "question": question,
        "num_queries": 3,
        "vectorstore_content_summary": vectorstore_content_summary
    })

    # Display summary of where results came from (for teaching purposes)
    print(f"Total number of results: {len(rag_fusion_mmr_results)}")
    for i, doc in enumerate(rag_fusion_mmr_results, start=1):
        display(Markdown(f"        Document {i} from `{doc.metadata['source']}`"))

    # Convert retrieved documents into Document objects with metadata and
    formatted_doc_results = [
        Document(
            metadata={k: v for k, v in doc.metadata.items() if k != 'rrf_score'},
            page_content=doc.page_content
        )
        for doc in rag_fusion_mmr_results
    ]

    return {"documents": formatted_doc_results}

# ----- Answer Generator Node -----
def answer_generator(state):
    """
    Generates an answer based on the retrieved documents and user question.

    This node prepares a prompt that includes:
    - The original or rewritten user question
    - A list of relevant documents (from vectorstore or web search)

    It invokes the main LLM to synthesize a concise and grounded response
    for use in later hallucination and usefulness checks.

    Args:
        state (GraphState): The current LangGraph state containing documents and question.

    Returns:
        dict: A dictionary with one key "generation" containing the LLM response.
    """
    print("\n---ANSWER GENERATION---")

    documents = state["documents"]

    # Use original_question if available (after rewriting), otherwise default to question
    original_question = state.get("original_question", None)
    if original_question is not None:
        question = original_question
    else:
        question = state["question"]

    # Ensure all documents are LangChain Document objects (convert from dicts)
    documents = [
        Document(metadata=doc["metadata"], page_content=doc["page_content"])
        if isinstance(doc, dict) else doc
        for doc in documents
    ]

    # Format the prompt for the answer generator
    _generator_prompt = answer_generator_prompt_template.format(

```

```

        context=documents,
        question=question
    )

    # Call the LLM to generate the answer
    answer_generation = llm_gpt.invoke(answer_generator_prompt)
    print(f"Answer generation has been generated.")

    return {"generation": answer_generation.content}

# ----- Web Searcher Node -----
def web_search(state):
    """
    Performs a real-time web search and appends results to previously ret

    This node is used when:
    - The vectorstore lacks sufficient relevant information
    - The original question requires current or factual information from

    It queries the web using a tool (e.g., Tavily), formats the returned
    `Document` objects, and appends them to the existing document list fo

    Args:
        state (GraphState): Current graph state with the user's question

    Returns:
        dict: Updated state with the combined list of vectorstore and web
    """

    print("\n---WEB SEARCH---")

    question = state["question"]
    documents = state.get("documents", [])

    # Run the web search using the web search tool
    web_results = web_search_tool.invoke(question)

    # Convert raw web search results into a simplified format
    formatted_web_results = [
        {
            "metadata": {
                "title": result["title"],
                "url": result["url"]
            },
            "page_content": result["content"]
        }
        for result in web_results
    ]

    # Ensure previous documents are consistently formatted as LangChain D
    documents = [
        Document(metadata=doc["metadata"], page_content=doc["page_content"])
        if isinstance(doc, dict) else doc
        for doc in documents
    ]

    # Append the new web documents
    documents.extend(formatted_web_results)

```

```

print(f"Total number of web search documents: {len(formatted_web_resu
return {"documents": documents}

# ----- Chitter-Chatter Node -----
def chitter_chatter(state):
    """
    Handles casual, off-topic, or unanswerable in-scope questions using a

    This node is designed to keep the user engaged and politely redirect
    that are better suited to the system's capabilities.

    Args:
        state (GraphState): Current graph state containing the user quest

    Returns:
        dict: Response from the Chitter-Chatter agent under the key ``gen
    """
    print("\n---CHIT-CHATting---")
    question = state["question"]

    # Generate a friendly fallback response using the Chitter-Chatter pro
    chitterchatter_response = llm_gpt_mini.invoke(
        [SystemMessage(chitterchatter_prompt),
         HumanMessage(question)])

    return {"generation": chitterchatter_response.content}

# ----- Adaptive Query Rewrite Node -----
def query_rewriter(state):
    """
    Rewrites the original question if the answer was hallucinated or unhe

    This node helps improve retrieval quality in the second attempt by:
    - Identifying gaps between the original query and the generated answe
    - Generating a clearer, more focused version of the question
    - Keeping a copy of the original for fallback comparison

    Args:
        state (GraphState): Contains the original and current question, a

    Returns:
        dict: Updated state with the rewritten question and preserved ori
    """
    print("\n---QUERY REWRITE---")

    # Use original question if available, otherwise fall back to input
    original_question = state.get("original_question", 0)
    if original_question != 0:
        question = original_question
    else:
        question = state["question"]

    generation = state["generation"]

    # Create prompt and invoke the query rewriter
    query_rewriter_prompt = query_rewriter_prompt_template.format(
        question=question,
        generation=generation,
        vectorstore_content_summary = vectorstore_content_summary

```



```

)

# Use the LLM model to grade the document
query_rewriter_result = llm_gpt.with_structured_output(method="json_m
    query_rewriter_prompt)

return {"question": query_rewriter_result['rewritten_question'],
        "original_question": question}

# ----- Retry Counter Node for Hallucination Checker -----
def hallucination_checker_tracker(state):
    """
    Tracks how many times the hallucination checker has been triggered.

    This helps avoid infinite loops in the graph by limiting retries after

    Args:
        state (GraphState): Current state of the graph including retry me

    Returns:
        dict: Updated state with incremented `hallucination_checker_attem
    """
    num_attempts = state.get("hallucination_checker_attempts", 0)
    return {"hallucination_checker_attempts": num_attempts + 1}

# ----- Retry Counter Node for Answer Verifier -----
def answer_verifier_tracker(state):
    """
    Tracks how many times the answer usefulness checker has been triggered.

    This node helps the workflow know when to stop trying to rewrite queries
    after repeated failures to generate an appropriate answer.

    Args:
        state (GraphState): Current state including verification metadata

    Returns:
        dict: Updated state with incremented `answer_verifier_attempts`.
    """
    num_attempts = state.get("answer_verifier_attempts", 0)
    return {"answer_verifier_attempts": num_attempts + 1}

```

2.3 Define Edge Functions

This section defines all edge functions responsible for decision-making and flow control within the LangGraph workflow.

These functions serve as conditional logic that determines the next step based on current state, enabling the graph to adapt intelligently at runtime. Specifically, these edges handle:

- Routing the user query via the Query Router Agent
- Evaluating document quality using asynchronous relevance grading
- Deciding whether to generate or retry based on document sufficiency

- Verifying grounding and usefulness of the generated answer, with built-in retry logic

Together, these edges form the decision backbone that connects agent nodes.

This section also introduces a key Python: "Async Programming", used to improve performance when grading multiple documents.

Async Functions:

- Defined using `async def`, which creates a coroutine.
- Calling an async function returns a `coroutine` object—it does not run immediately.
- To execute, the coroutine must be awaited.

`await` Expression:

- Used inside async functions to pause execution until the awaited task completes.
- While paused, control is yielded back to the event loop, allowing other tasks to run.
- Execution resumes once the awaited operation finishes.

`asyncio.gather()`

- Executes multiple coroutines concurrently and waits for all to finish.
- Returns results in the same order as the input coroutines.
- Ideal for parallel tasks—like grading multiple documents at once.

In this workflow, we use `grade_documents_parallel()` to apply async programming. This allows the system to evaluate all retrieved documents simultaneously, making it significantly faster and more efficient than grading one document at a time.

```
In [37]: # ----- Routing Decision -----
def route_question(state):
    """
    Routes the user question to the appropriate agent based on the Query

    Args:
        state (GraphState): Contains the user's input question.

    Returns:
        str: One of 'Vectorstore', 'Websearch', or 'Chitter-Chatter'.
    """
    print("----ROUTING QUESTION----")
    question = state["question"]
    route_question_response = llm_gpt.with_structured_output(method="json",
        [SystemMessage(query_router_prompt),
        HumanMessage(question)]
    )

    parsed_router_output = route_question_response["Datasource"]

    parsed_router_output == "Websearch":
```

```

        print("----ROUTING QUESTION TO WEB SEARCH----")
        return "Websearch"
    elif parsed_router_output == "Vectorstore":
        print("----ROUTING QUESTION TO VECTORSTORE----")
        return "Vectorstore"
    elif parsed_router_output == "Chitter-Chatter":
        print("----ROUTING QUESTION TO CHITTER-CHATTER----")
        return "Chitter-Chatter"

# ----- Async document relevance grading -----
async def grade_documents_parallel(state):
    """
    Grades retrieved documents asynchronously to determine their relevance.

    Documents are processed in parallel using async calls. If 50% or more
    the system flags this as a failure, triggering a web search in the ne

    Args:
        state (GraphState): Contains the documents and question.

    Returns:
        dict: Updated state with filtered documents and a `checker_resul
    """
    print("----CHECK DOCUMENT RELEVANCE TO QUESTION----")
    question = state["question"]
    documents = state["documents"]

    # Inner coroutine that grades one document at a time using the relev
    async def grade_document(doc, question):
        relevance_grader_prompt = relevance_grader_prompt_template.format(
            document=doc,
            question=question
        )
        grader_result = await llm_gpt_mini.with_structured_output(method=
            relevance_grader_prompt)
        return grader_result

    # Create async tasks for grading all documents
    tasks = [grade_document(doc, question) for doc in documents]

    # Run all tasks concurrently
    results = await asyncio.gather(*tasks)

    filtered_docs = []

    # Collect only documents marked as "pass"
    for i, score in enumerate(results):
        if score["binary_score"].lower() == "pass":
            print(f"----GRADE: DOCUMENT RELEVANT--- {score['binary_score']}")
            filtered_docs.append(documents[i]) # only keep the relevant o
        else:
            print("----GRADE: DOCUMENT NOT RELEVANT----")

    # Analyze how many documents were filtered out
    total_docs = len(documents)
    relevant_docs = len(filtered_docs)

    if total_docs > 0:
        filtered_out_percentage = (total_docs - relevant_docs) / total_do

```

```

        # If more than 50% of documents were irrelevant, fail and fall ba
        checker_result = "fail" if filtered_out_percentage >= 0.5 else "p
        print(f"---FILTERED OUT {filtered_out_percentage*100:.1f}% OF IRR
        print(f"---**{checker_result}**---")
    else:
        # If no documents were retrieved at all, treat as automatic failu
        checker_result = "fail"
        print("---NO DOCUMENTS AVAILABLE, WEB SEARCH TRIGGERED---")

    return {"documents": filtered_docs, "checker_result": checker_result}

# ----- Decide whether to generate or fallback -----
def decide_to_generate(state):
    """
    Conditional edge function used after document relevance grading.

    It checks the `checker_result` from the previous step:
    - If the result is 'fail' (indicating that a majority of documents we
      it triggers a fallback to web search for more reliable context.
    - If the result is 'pass', it proceeds to the answer generation node.

    Args:
        state (GraphState): Includes the 'checker_result' from the docume

    Returns:
        str: Either 'generate' or 'Websearch', used to transition to the
    """
    print("---CHECK GENERATION CONDITION---")
    checker_result = state["checker_result"]

    if checker_result == "fail":
        print(
            "---DECISION: MORE THAN HALF OF THE DOCUMENTS ARE IRRELEVANT
        )
        return "Websearch"
    else:
        # We have relevant documents, so generate answer
        print("---DECISION: GENERATE---")
        return "generate"

# # ----- Final Answer Validation -----
def check_generation_vs_documents_and_question(state):
    """
    Conditional edge function verifies the quality of the generated answe
    - Grounded in the retrieved documents (hallucination check)
    - Relevant to the original user question (answer verifier)

    Depending on the result, this function controls whether the system pr
    retries answer generation, or stops after exceeding retry limits.

    Args:
        state (GraphState): Includes question, generated answer, document

    Returns:
        str: One of the route labels used in LangGraph transitions:
        - 'useful': Answer is grounded and relevant
        - 'not useful': Answer is grounded but does not address the q

```

```

        - 'not supported': Answer is not grounded (hallucination)
        - 'max retries': Too many failed attempts, abort or fallback
    """

    print("----CHECK HALLUCINATIONS WITH DOCUMENTS----")

    # Use original rewritten question if present; otherwise use latest ve
    question = state["question"]
    original_question = state.get("original_question", 0)

    if original_question != 0:
        question = original_question
    else:
        question = state["question"]

    documents = state["documents"]
    generation = state["generation"]

    # Retry counters
    hallucination_checker_attempts = state.get("hallucination_checker_att
    answer_verifier_attempts = state.get("answer_verifier_attempts", 0)

    # Run hallucination checker: does the answer come from the documents?
    hallucination_checker_prompt = hallucination_checker_prompt_template.
        documents=documents,
        generation=generation
    )
    hallucination_checker_result = llm_gpt_mini.with_structured_output(me
    hallucination_checker_prompt)

    # Helper to format "1st", "2nd", etc.
    def ordinal(n):
        return f"{n}{'th' if 10 <= n % 100 <= 20 else {1:'st', 2:'nd', 3:'rd', 4:'th', 5:'th', 6:'th', 7:'th', 8:'th', 9:'th', 10:'th', 11:'th', 12:'th', 13:'th', 14:'th', 15:'th', 16:'th', 17:'th', 18:'th', 19:'th', 20:'th', 21:'st', 22:'nd', 23:'rd', 24:'th', 25:'th', 26:'th', 27:'th', 28:'th', 29:'th', 30:'th', 31:'st', 32:'nd', 33:'rd', 34:'th', 35:'th', 36:'th', 37:'th', 38:'th', 39:'th', 40:'th', 41:'st', 42:'nd', 43:'rd', 44:'th', 45:'th', 46:'th', 47:'th', 48:'th', 49:'th', 50:'th', 51:'st', 52:'nd', 53:'rd', 54:'th', 55:'th', 56:'th', 57:'th', 58:'th', 59:'th', 60:'th', 61:'st', 62:'nd', 63:'rd', 64:'th', 65:'th', 66:'th', 67:'th', 68:'th', 69:'th', 70:'th', 71:'st', 72:'nd', 73:'rd', 74:'th', 75:'th', 76:'th', 77:'th', 78:'th', 79:'th', 80:'th', 81:'st', 82:'nd', 83:'rd', 84:'th', 85:'th', 86:'th', 87:'th', 88:'th', 89:'th', 90:'th', 91:'st', 92:'nd', 93:'rd', 94:'th', 95:'th', 96:'th', 97:'th', 98:'th', 99:'th', 100:'th'}}"

    # If generation is grounded (pass hallucination check)
    if hallucination_checker_result['binary_score'].lower() == "pass":
        print("----DECISION: GENERATION IS GROUNDED IN DOCUMENTS----")

        # Now check if it answers the question usefully
        print("----VERIFY ANSWER WITH QUESTION----")
        # Test using question and generation from above
        answer_verifier_prompt = answer_verifier_prompt_template.format(
            question=question,
            generation=generation
        )
        answer_verifier_result = llm_gpt_mini.with_structured_output(meth
        answer_verifier_prompt)

        # If answer is grounded AND relevant, return final result
        if answer_verifier_result['binary_score'].lower() == "pass":
            print("----DECISION: GENERATION ADDRESSES QUESTION----")
            return "useful"

        # If max attempts reached for usefulness check, exit
        elif answer_verifier_attempts > 1:
            print("----DECISION: MAX RETRIES REACHED----")
            return "max retries"

        # Otherwise, try query rewrite and retry generation
    else:

```

```

        print("----DECISION: GENERATION DOES NOT ADDRESS QUESTION, RE-
        print(f"This is the {ordinal(answer_verifier_attempts+1)} att
        return "not useful"

    # If generation is NOT grounded and retry limit exceeded
    elif hallucination_checker_attempts > 1:
        print("----DECISION: MAX RETRIES REACHED----")
        return "max retries"

    # If answer is not grounded but we can still retry
    else:
        print("----DECISION: GENERATION IS NOT GROUNDED IN DOCUMENTS, RE-T
        print(f"This is the {ordinal(hallucination_checker_attempts+1)} a
        return "not supported"

```

2.4 Define Agentic RAG Workflow (Graph Assembly)

This section defines the full LangGraph workflow by connecting all agents and conditional edge functions into a structured graph. It establishes:

- Entry routing via Query Router Agent classification
- Feedback loops for hallucination and answer verification
- Conditional flow control via document grading and answer evaluation
- Exit points for high-confidence answers or fallback responses

```

In [38]: # Initialize the graph with shared state structure
workflow = StateGraph(GraphState)

# === Add agent nodes ===
workflow.add_node("WebSearcher", web_search) # web sea
workflow.add_node("DocumentRetriever", document_retriever) # Multi-q
workflow.add_node("RelevanceGrader", grade_documents_parallel) # Async d
workflow.add_node("AnswerGenerator", answer_generator) # Generat
workflow.add_node("QueryRewriter", query_rewriter) # Rewrite
workflow.add_node("ChitterChatter", chitter_chatter) # Fallba

# === Add retry tracker nodes ===
workflow.add_node("HallucinationCheckerFailed", hallucination_checker_tra
workflow.add_node("AnswerVerifierFailed", answer_verifier_tracker)

# === Entry point: Route query to appropriate agent ===
workflow.set_conditional_entry_point(
    route_question,
    {
        "Websearch": "WebSearcher",
        "Vectorstore": "DocumentRetriever",
        "Chitter-Chatter": "ChitterChatter",
    },
)

# === Node transitions ===
workflow.add_edge("DocumentRetriever", "RelevanceGrader") # Ret
workflow.add_edge("WebSearcher", "AnswerGenerator") # Web

workflow.add_edge("HallucinationCheckerFailed", "AnswerGenerator") # Ret
workflow.add_edge("AnswerVerifierFailed", "QueryRewriter") # Ret
workflow.add_edge("QueryRewriter", "DocumentRetriever") # Rew

```

```

workflow.add_edge("ChitterChatter", END)

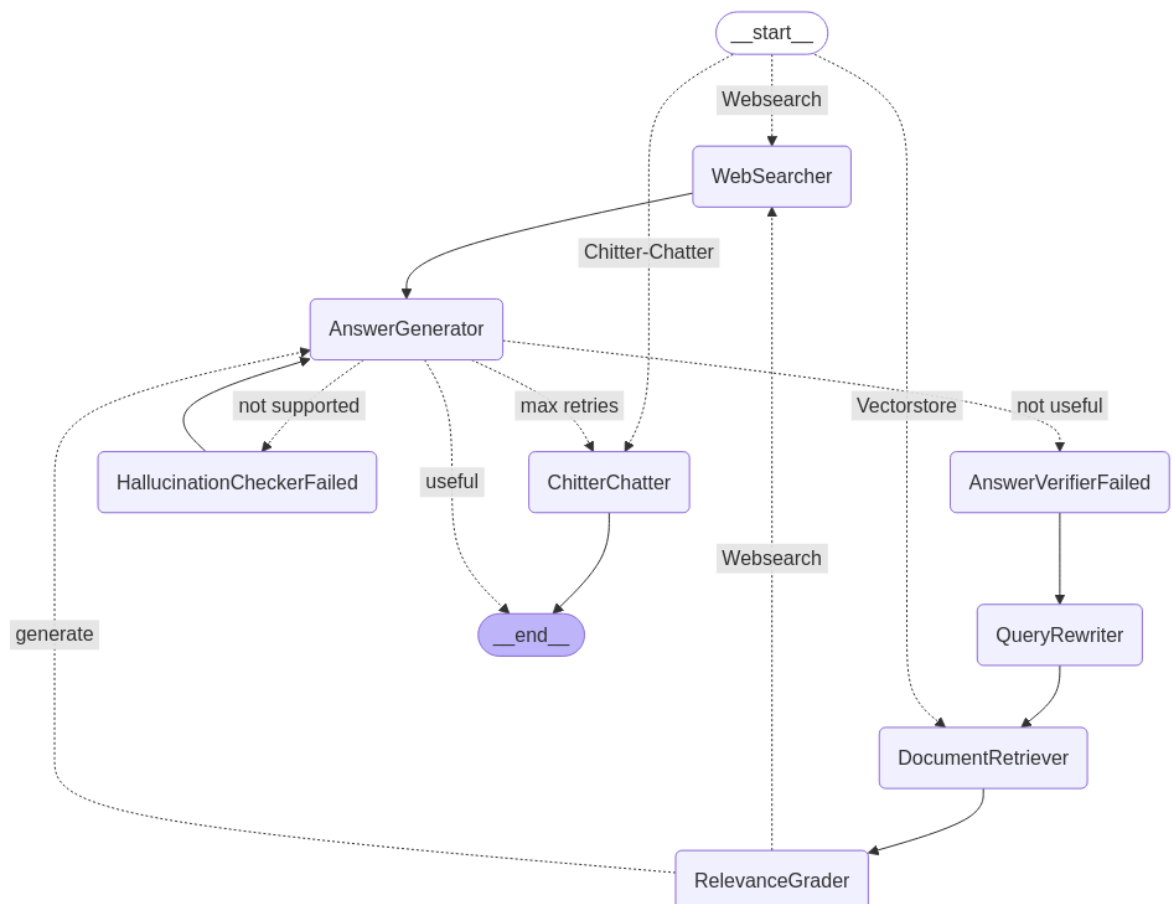
# === Conditional routing after document grading ===
workflow.add_conditional_edges(
    "RelevanceGrader",
    decide_to_generate,
    {
        "Websearch": "WebSearcher",      # Too many irrelevant docs →
        "generate": "AnswerGenerator",    # Good enough → Proceed to ge
    },
)

# === Conditional routing after generation quality checks ===
workflow.add_conditional_edges(
    "AnswerGenerator",
    check_generation_vs_documents_and_question,
    {
        "not supported": "HallucinationCheckerFailed", # Hallucinated →
        "useful": END,                                   # Success
        "not useful": "AnswerVerifierFailed",          # Off-topic → Rew
        "max retries": "ChitterChatter"                 # Stop after too
    },
)

# --- Compile the graph ---
graph = workflow.compile()

```

In [39]: # ----- Visualize the Graph Structure -----
This built-in visualization to render the workflow as a flow diagram.
display(Image(graph.get_graph().draw_mermaid_png()))



3. Run the Workflow (Graph Execution)

LangGraph uses a "message-passing execution model", inspired by Google's Pregel system.

Execution flows through the graph based on data, decision logic, and state updates:

1. **Activation:** Nodes become active when they receive input.
2. **Processing:** Each node execute its logic and modifies the shared `GraphState`.
3. **Message Passing:** Updated state is passed to the next node(s) based on conditional edges.
4. **Termination:** Execution stops when no nodes are active. The `END` node terminates the execution.

Test Examples You tested the workflow with four types of questions, each triggering a different agent pathway:

1. Web Search Example: Routed to the `WebSearcher` due to real-time or factual content needs.
2. Vectorstore Search Example: Routed to the `DocumentRetriever` and successfully answered using vectorstore content.
3. Chitter-Chatter Example: Routed to `ChitterChatter` due to being off-topic or casual (e.g., small talk).
4. Vectorstore Search Loop Example: Started with vectorstore retrieval, but most documents were irrelevant. Then triggered retry logic (e.g., web fallback or query rewriting), showing how the graph handles failed retrieval or weak answers.

```
In [40]: # Define a test question to run through the LangGraph workflow
test_question1 = "Who is Peter Drucker, and where was he born?"

# Execute the graph asynchronously in streaming mode.
async for event1 in graph.astream({"question": test_question1}, stream_mode="messages"):
    print(event1) # Print state after each node is processed

# Display the final result after the graph has completed execution
display(Markdown(f"\n{'='*40} Final Output {'='*40}\n"))
display(Markdown(event1["generation"]))
```


---ROUTING QUESTION---

---ROUTING QUESTION TO WEB SEARCH---

{ 'question': 'Who is Peter Drucker, and where was he born?' }

---WEB SEARCH---

Total number of web search documents: 5

{ 'question': 'Who is Peter Drucker, and where was he born?', 'documents': [{ 'metadata': { 'title': 'Peter F. Drucker | Biography, Management, Books, & Facts - Britannica', 'url': 'https://www.britannica.com/money/Peter-F-Drucker' }, 'page_content': 'Peter F. Drucker (born November 19, 1909, Vienna, Austria-died November 11, 2005, Claremont, California, U.S.) was an Austrian-born American' }, { 'metadata': { 'title': 'Peter Drucker Biography - curiouscat.com', 'url': 'https://curiouscat.com/management/experts/peter-drucker' }, 'page_content': 'Peter Ferdinand Drucker (19 Nov 1909 - 11 Nov 2005) was a leading management theorist. Drucker, born in Vienna, Austria, moved to the United States in 1937.' }, { 'metadata': { 'title': 'Peter F. Drucker | Biography & Management Theory - Study.com', 'url': 'https://study.com/academy/lesson/peter-f-drucker-father-of-management-theory.html' }, 'page_content': 'Lesson Summary. Peter F. Drucker was born in Vienna, Austria. He was surrounded by Austrian intellectuals that worked with his father.' }, { 'metadata': { 'title': 'About Peter Drucker * Drucker Institute', 'url': 'https://drucker.institute/perspective__trashed/about-peter-drucker/' }, 'page_content': 'Peter Drucker was born in Vienna, Austria, on November 19, 1909. The household in which he grew up was one of great intellectual ferment. His parents, Adolph' }, { 'metadata': { 'title': 'PETER F. DRUCKER', 'url': 'https://www.druckerforum.org/peter-f-drucker/' }, 'page_content': 'Peter F. Drucker was born in Vienna on November 19, 1909 in Vienna, when the city was still the vibrant centre of the Habsburg monarchy.' }] }

---ANSWER GENERATION---

Answer generation has been generated.

---CHECK HALLUCINATIONS WITH DOCUMENTS---

---DECISION: GENERATION IS GROUNDED IN DOCUMENTS---

---VERIFY ANSWER WITH QUESTION---

---DECISION: GENERATION ADDRESSES QUESTION---

{ 'question': 'Who is Peter Drucker, and where was he born?', 'generation': 'Peter Drucker was an Austrian-born American who became a leading management theorist. He was born in Vienna, Austria, on November 19, 1909.\\n\\n**References**:\\n- Peter F. Drucker | Biography, Management, Books, & Facts - Britannica. Retrieved from https://www.britannica.com/money/Peter-F-Drucker\\n- Peter Drucker Biography - curiouscat.com. Retrieved from https://curiouscat.com/management/experts/peter-drucker\\n- About Peter Drucker * Drucker Institute. Retrieved from https://drucker.institute/perspective__trashed/about-peter-drucker/', 'documents': [{ 'metadata': { 'title': 'Peter F. Drucker | Biography, Management, Books, & Facts - Britannica', 'url': 'https://www.britannica.com/money/Peter-F-Drucker' }, 'page_content': 'Peter F. Drucker (born November 19, 1909, Vienna, Austria-died November 11, 2005, Claremont, California, U.S.) was an Austrian-born American' }, { 'metadata': { 'title': 'Peter Drucker Biography - curiouscat.com', 'url': 'https://curiouscat.com/management/experts/peter-drucker' }, 'page_content': 'Peter Ferdinand Drucker (19 Nov 1909 - 11 Nov 2005) was a leading management theorist. Drucker, born in Vienna, Austria, moved to the United States in 1937.' }, { 'metadata': { 'title': 'Peter F. Drucker | Biography & Management Theory - Study.com', 'url': 'https://study.com/academy/lesson/peter-f-drucker-father-of-management-theory.html' }, 'page_content': 'Lesson Summary. Peter F. Drucker was born in Vienna, Austria. He was surrounded by Austrian intellectuals that worked with his father' }, { 'metadata': { 'title': 'About Peter Drucker * Drucker Institute', 'u

```
rl': 'https://drucker.institute/perspective__trashed/about-peter-drucker/'}}, 'page_content': 'Peter Drucker was born in Vienna, Austria, on November 19, 1909. The household in which he grew up was one of great intellectual ferment. His parents, Adolph', {'metadata': {'title': 'PETER F. DRUCKER', 'url': 'https://www.druckerforum.org/peter-f-drucker/'}, 'page_content': 'Peter F. Drucker was born in Vienna on November 19, 1909 in Vienna, when the city was still the vibrant centre of the Habsburg monarchy.'}}]
```

===== Final Output

Peter Drucker was an Austrian-born American who became a leading management theorist. He was born in Vienna, Austria, on November 19, 1909.

References:

- Peter F. Drucker | Biography, Management, Books, & Facts - Britannica. Retrieved from <https://www.britannica.com/money/Peter-F-Drucker>
- Peter Drucker Biography - curiouscat.com. Retrieved from <https://curiouscat.com/management/experts/peter-drucker>
- About Peter Drucker * Drucker Institute. Retrieved from https://drucker.institute/perspective__trashed/about-peter-drucker/

```
In [41]: # Vector search example
test_question2 = "What did Drucker say about knowledge workers in the boo

async for event2 in graph.astream({"question": test_question2}, stream_mo
    print(event2)

display(Markdown(f"\n{'='*40} Final Output {'='*40}\n"))
display(Markdown(event2["generation"]))
```

---ROUTING QUESTION---

---ROUTING QUESTION TO VECTORSTORE---

{'question': 'What did Drucker say about knowledge workers in the book?'}

---QUERY TRANSLATION AND RAG-FUSION---

Total number of results: 8

Document 1 from `The Effective Executive-2002.pdf`, page 17

Document 2 from `The Daily Drucker-2004.pdf`, page 321

Document 3 from `The Daily Drucker-2004.pdf`, page 806

Document 4 from `The Essential Drucker-2008.pdf`, page 196

Document 5 from `The Daily Drucker-2004.pdf`, page 800

Document 6 from `The Essential Drucker-2008.pdf`, page 237

Document 7 from `The Daily Drucker-2004.pdf`, page 323

Document 8 from `The Daily Drucker-2004.pdf`, page 323

{'question': 'What did Drucker say about knowledge workers in the book?', 'documents': [Document(metadata={'page': 17, 'year': '2002', 'title': 'The Effective Executives', 'author': 'Peter F. Drucker', 'source': 'The Effective Executive-2002.pdf'}, page_content='. \nIt takes his knowledge and uses it as the resource, the motivation, and the vision of \nother knowledge workers. Knowledge workers are rarely in phase with each other, \nprecisely because they are knowledge workers. Each has his own skill and his own \nconcerns. One man may be interested in tax accounting or in bacteriology, or in training \nand developing tomorrow's key administrators in the city government'), Document(metadata={'page': 321, 'year': '2004', 'title': 'The Daily Drucker: 366 Days of Insight and Motivation for Getting the Right Things Done', 'author': 'Drucker, Peter F.', 'source': 'The Daily Drucker-2004.pdf'}, page_content='. \n5. Productivity of the knowledge worker is not—at least not \nprimarily—a matter of the quantity of output. Quality is at least \nas important. \n6. Finally, knowledge-worker productivity requires that the \nknowledge worker be both seen and treated as an “asset” rather \nthan a “cost.” It requires that knowledge workers want to work \nfor the organization in preference to all other opportunities. \nACTION POINT: Apply steps one through five to your knowledge work. \nManagement Challenges for the 21st Century'), Document(metadata={'page': 806, 'year': '2004', 'title': 'The Daily Drucker: 366 Days of Insight and Motivation for Getting the Right Things Done', 'author': 'Drucker, Peter F.', 'source': 'The Daily Drucker-2004.pdf'}, page_content='. Drucker analyzes the new realities of strategy, shows how to \nbe a leader in periods of change, and explains the “New Information \nRevolution,” discussing the information an executive needs and the \ninformation an executive owes. He also examines knowledge-worker \nproductivity, and shows that changes in the basic attitude of individuals and \norganizations, as well as structural changes in work itself, are needed for \nincreased productivity. Finally, Drucker addresses the ultimate challenge of'), Document(metadata={'page': 196, 'year': '2008', 'title': 'The Essential Drucker', 'author': 'Peter F. Drucker', 'source': 'The Essential Drucker-2008.pdf'}, page_content='. Knowledge \nworkers, after all, first came into being in any substantial numbers a generation ago. (I coined the \nterm “knowledge worker” years ago.) \nBut also the shift from manual workers who do as they are being told—either by the task or by the \nboss—to knowledge workers who have to manage themselves profoundly challenges social \nstructure'), Document(metadata={'page': 800, 'year': '2004', 'title': 'The Daily Drucker: 366 Days of Insight and Motivation for Getting the Right Things Done', 'author': 'Drucker, Peter F.', 'source': 'The Daily Drucker-2004.pdf'}, page_content='. According to Peter Drucker, \n“This book tries to equip the manager with the understanding, the thinking, \nthe knowledge, and the skills for today's and also tomorrow's jobs.” This \nmanagement classic has been developed and tested during more than thirty \nyears of management teaching in universities, executive programs, seminars, \nand through the author's close work with managers as a consultant for large \nand small businesses, government agencies, hospitals, and schools'), Document(metadata={'page': 237, 'year': '2008', 'title': 'The Essential Drucker', 'author': 'Peter F. Drucker', 'source': 'The Essential Drucker-2008.pdf'}, page_content='About the Author \nPeter F. Drucker was born in 1909 in Vienna and was educated there and in England. He received his \ndoctorate in public and international law while working as a newspaper reporter in Frankfurt, \nGermany, and then worked as an economist for an international bank in London. In 1927, he came to \nthe United States. Drucker's management books and analyses of economics and society are widely \nread and respected throughout the world and have been translated into more than 20 languages'), Document(metadata={'page': 323, 'year': '2004', 'title': 'The Daily Drucker: 366 Days of Insight and Motivation for Getting the Right Things Done', 'author': 'Drucker, Peter F.', 'source': 'The Daily Drucker-2004.pdf'}, page_content='24 May \nDefining the Task in Knowledge Work \nIn knowledge work, the how only comes after the what has been answered. \nIn management

nual work the task is always given. Wherever there still are domestic\nservants, the owner of the house tells them what to do. The machine or the\nassembly line programs the factory worker. But, in knowledge work, what to\ndo becomes the first and decisive question. For knowledge workers are not\nprogrammed by the machine'), Document(metadata={'page': 323, 'year': '2004', 'title': 'The Daily Drucker: 366 Days of Insight and Motivation for Getting the Right Things Done', 'author': 'Drucker, Peter F.', 'source': 'The Daily Drucker-2004.pdf'}, page_content='. They know\nwhat steps are most important and what methods need to be used to complete\nthe tasks; and it is their knowledge that tells them what chores are\nunnecessary and should be eliminated.\nWork on knowledge-worker productivity therefore begins with asking the\nknowledge workers themselves: What is your task? What should it be?\nWhat should you be expected to contribute? and What hampers you in\ndoing your task and should be eliminated? The how only comes after the\nwhat has been answered')]]}

---CHECK DOCUMENT RELEVANCE TO QUESTION---

---GRADE: DOCUMENT RELEVANT--- pass

---GRADE: DOCUMENT RELEVANT--- pass

---GRADE: DOCUMENT RELEVANT--- pass

---GRADE: DOCUMENT RELEVANT--- pass

---GRADE: DOCUMENT NOT RELEVANT---

---GRADE: DOCUMENT NOT RELEVANT---

---GRADE: DOCUMENT RELEVANT--- pass

---GRADE: DOCUMENT RELEVANT--- pass

---FILTERED OUT 25.0% OF IRRELEVANT DOCUMENTS---

---**pass**---

---CHECK GENERATION CONDITION---

---DECISION: GENERATE---

{'question': 'What did Drucker say about knowledge workers in the book?',
'documents': [Document(metadata={'page': 17, 'year': '2002', 'title': 'The Effective Executiver', 'author': 'Peter F. Drucker', 'source': 'The Effective Executive-2002.pdf'}, page_content='. \nIt takes his knowledge and uses it as the resource, the motivation, and the vision of \nother knowledge workers. Knowledge workers are rarely in phase with each other, \nprecisely because they are knowledge workers. Each has his own skill and his own \nconcerns. One man may be interested in tax accounting or in bacteriology, or in training \nand developing tomorrow's key administrators in the city government'), Document(metadata={'page': 321, 'year': '2004', 'title': 'The Daily Drucker: 366 Days of Insight and Motivation for Getting the Right Things Done', 'author': 'Drucker, Peter F.', 'source': 'The Daily Drucker-2004.pdf'}, page_content='. \n5. Productivity of the knowledge worker is not—at least not\nprimarily—a matter of the quantity of output. Quality is at least\nas important.\n6. Finally, knowledge-worker productivity requires that the\nknowledge worker be both seen and treated as an “asset” rather\nthan a “cost.” It requires that knowledge workers want to work\nfor the organization in preference to all other opportunities.\nACTION POINT: Apply steps one through five to your knowledge work.\nManagement Challenges for the 21st Century'), Document(metadata={'page': 806, 'year': '2004', 'title': 'The Daily Drucker: 366 Days of Insight and Motivation for Getting the Right Things Done', 'author': 'Drucker, Peter F.', 'source': 'The Daily Drucker-2004.pdf'}, page_content='. Drucker analyzes the new realities of strategy, shows how to\nbe a leader in periods of change, and explains the “New Information\nRevolution,” discussing the information an executive needs and the\ninformation an executive owes. He also examines knowledge-worker\nproductivity, and shows that changes in the basic attitude of individuals and\norganizations, as well as structural changes in work itself, are needed for\nincreased productivity. Finally, Drucker addresses the ultimate challenge of'), Document(metadata={'page': 196, 'year': '2008', 'title': 'The Essential Drucker', 'author': 'Peter F. Drucker', 'source': 'The Essential Drucker-2008.pdf'}, page_content='. Knowledge \nworkers, after all,

first came into being in any substantial numbers a generation ago. (I coined the term “knowledge worker” years ago.) But also the shift from manual workers who do as they are being told—either by the task or by the boss—to knowledge workers who have to manage themselves profoundly challenges social structure’), Document(metadata={'page': 323, 'year': '2004', 'title': 'The Daily Drucker: 366 Days of Insight and Motivation for Getting the Right Things Done', 'author': 'Drucker, Peter F.', 'source': 'The Daily Drucker-2004.pdf'}, page_content='24 May\nDefining the Task in Knowledge Work\nIn knowledge work, the how only comes after the what has been answered.\nIn manual work the task is always given. Wherever there still are domestic servants, the owner of the house tells them what to do. The machine or the assembly line programs the factory worker. But, in knowledge work, what to do becomes the first and decisive question. For knowledge workers are not programmed by the machine'), Document(metadata={'page': 323, 'year': '2004', 'title': 'The Daily Drucker: 366 Days of Insight and Motivation for Getting the Right Things Done', 'author': 'Drucker, Peter F.', 'source': 'The Daily Drucker-2004.pdf'}, page_content='. They know\nwhat steps are most important and what methods need to be used to complete the tasks; and it is their knowledge that tells them what chores are unnecessary and should be eliminated.\nWork on knowledge-worker productivity therefore begins with asking the knowledge workers themselves: What is your task? What should it be? What should you be expected to contribute? and What hampers you in doing your task and should be eliminated? The how only comes after the what has been answered']], 'checker_result': 'pass'})

---ANSWER GENERATION---

Answer generation has been generated.

---CHECK HALLUCINATIONS WITH DOCUMENTS---

---DECISION: GENERATION IS GROUNDED IN DOCUMENTS---

---VERIFY ANSWER WITH QUESTION---

---DECISION: GENERATION ADDRESSES QUESTION---

{'question': 'What did Drucker say about knowledge workers in the book?', 'generation': 'Peter Drucker described knowledge workers as individuals who use their knowledge as a resource, motivation, and vision for other knowledge workers. They have their own skills and concerns, making them rarely in sync with one another. Drucker emphasized that knowledge workers' productivity should not be measured primarily by the quantity of output, but rather by the quality. He suggested treating knowledge workers as assets rather than costs and ensuring that they prefer to work for their organization over other opportunities. Understanding the tasks for knowledge work begins with answering "what" rather than "how," as knowledge workers are not programmed by machines but must manage themselves and define their tasks (Drucker, 2002; Drucker, 2004; Drucker, 2008).\n\n**Reference**:\nDrucker, P. F. (2002). *The Effective Executive*. \nDrucker, P. F. (2004). *The Daily Drucker: 366 Days of Insight and Motivation for Getting the Right Things Done*. \nDrucker, P. F. (2008). *The Essential Drucker*.', 'documents': [Document(metadata={'page': 17, 'year': '2002', 'title': 'The Effective Executive', 'author': 'Peter F. Drucker', 'source': 'The Effective Executive-2002.pdf'}, page_content='. \nIt takes his knowledge and uses it as the resource, the motivation, and the vision of other knowledge workers. Knowledge workers are rarely in phase with each other, \nprecisely because they are knowledge workers. Each has his own skill and his own \nconcerns. One man may be interested in tax accounting or in bacteriology, or in training \nand developing tomorrow's key administrators in the city government'), Document(metadata={'page': 321, 'year': '2004', 'title': 'The Daily Drucker: 366 Days of Insight and Motivation for Getting the Right Things Done', 'author': 'Drucker, Peter F.', 'source': 'The Daily Drucker-2004.pdf'}, page_content='. \n5. Productivity of the knowledge worker is not—at least not \nprimarily—a matter of the quantity of output. Quality is at least \nant. \n6. Finally, knowledge-worker productivity requires that t

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===== Final Output

=====

Peter Drucker described knowledge workers as individuals who use their knowledge as a resource, motivation, and vision for other knowledge workers. They have their own skills and concerns, making them rarely in sync with one another. Drucker emphasized that knowledge workers' productivity should not be measured primarily by the quantity of output, but rather by the quality. He suggested treating knowledge workers as assets rather than costs and ensuring that they prefer to work for their organization over other opportunities. Understanding the tasks for knowledge work begins with answering "what" rather than "how," as knowledge workers are not programmed by machines but must manage themselves and define their tasks (Drucker, 2002; Drucker, 2004; Drucker, 2008).

Reference: Drucker, P. F. (2002). *The Effective Executive*.

Drucker, P. F. (2004). *The Daily Drucker: 366 Days of Insight and Motivation for Getting the Right Things Done*.

Drucker, P. F. (2008). *The Essential Drucker*.

```
In [42]: # Chitter-Chat example
test_question3 = "How are you doing today?"

async for event3 in graph.astream({"question": test_question3}, stream_mode="events"):
    print(event3)

display(Markdown(f"\n{'='*40} Final Output {'='*40}\n"))
display(Markdown(event3["generation"]))
```

---ROUTING QUESTION---

---ROUTING QUESTION TO CHITTER-CHATTER---

```
{'question': 'How are you doing today?'}
```

---CHIT-CHATting---

```
{'question': 'How are you doing today?', 'generation': "I'm doing great, thank you! How about you? Is there something specific about Peter Drucker's management philosophy or leadership principles that you'd like to discuss today?"}
```

===== Final Output

=====

I'm doing great, thank you! How about you? Is there something specific about Peter Drucker's management philosophy or leadership principles that you'd like to discuss today?

A harder question:

- Conceptual leap: It asks Drucker to advise sentient AIs that reject all human organizational models.
- No grounding: Drucker never wrote about machine consciousness or post-human philosophy.
- Core tools removed: Time management, KPIs, human motivation — all irrelevant.
- Requires meta-reasoning: The agent must imagine how Drucker would respond with his framework obsolete.
- Webproof: Not too much searchable source can answer this.

- Hallucination risk: The agent must either fabricate or fall back to abstract principles — no safe middle.

```
In [44]: # Vectorstore search loop example
test_question4 = """
Search the database first to see how Peter Drucker might advise a sentient
after rejecting all human organizational models as insufficient for post-
"""

async for event4 in graph.astream({"question": test_question4}, stream_mode="text"):
    print(event4)

display(Markdown(f"\n{'='*40} Final Output {'='*40}\n"))
display(Markdown(event4["generation"]))
```

---ROUTING QUESTION---

---ROUTING QUESTION TO VECTORSTORE---

```
{'question': '\nSearch the database first to see how Peter Drucker might advise a sentient AI collective attempting to design its own management philosophy \nafter rejecting all human organizational models as insufficient for post-biological cognition?\n'}
```

---QUERY TRANSLATION AND RAG-FUSION---

Total number of results: 11

Document 1 from `The Daily Drucker-2004.pdf`, page 271

Document 2 from `The Effective Executive-2002.pdf`, page 112

Document 3 from `The Effective Executive-2002.pdf`, page 133

Document 4 from `The Effective Executive-2002.pdf`, page 2

Document 5 from `The Daily Drucker-2004.pdf`, page 799

Document 6 from `The Daily Drucker-2004.pdf`, page 172

Document 7 from `The Essential Drucker-2008.pdf`, page 87

Document 8 from `The Daily Drucker-2004.pdf`, page 797

Document 9 from `The Daily Drucker-2004.pdf`, page 727

Document 10 from `The Daily Drucker-2004.pdf`, page 796

Document 11 from `The Daily Drucker-2004.pdf`, page 796

{'question': '\nSearch the database first to see how Peter Drucker might advise a sentient AI collective attempting to design its own management philosophy \nafter rejecting all human organizational models as insufficient for post-biological cognition?\n', 'documents': [Document(metadata={'page': 271, 'year': '2004', 'title': 'The Daily Drucker: 366 Days of Insight and Motivation for Getting the Right Things Done', 'author': 'Drucker, Peter F.', 'source': 'The Daily Drucker-2004.pdf'}, page_content='. \n\nThe New Society'), Document(metadata={'page': 112, 'year': '2002', 'title': 'The Effective Executiver', 'author': 'Peter F. Drucker', 'source': 'The Effective Executive-2002.pdf'}, page_content='. But his is the only one of'), Document(metadata={'page': 133, 'year': '2002', 'title': 'The Effective Executiver', 'author': 'Peter F. Drucker', 'source': 'The Effective Executive-2002.pdf'}, page_content='. The definition of the specifications which the answer to the problem had to satisfy, \nthat is, of the "boundary conditions";'), Document(metadata={'page': 2, 'year': '2002', 'title': 'The Effective Executiver', 'author': 'Peter F. Drucker', 'source': 'The Effective Executive-2002.pdf'}, page_content='. \n\nPeter F. Drucker \n\nApril 2002 \n\nContents'), Document(metadata={'page': 799, 'year': '2004', 'title': 'The Daily Drucker: 366 Days of Insight and Motivation for Getting the Right Things Done', 'author': 'Drucker, Peter F.', 'source': 'The Daily Drucker-2004.pdf'}, page_content='. Peter Drucker's critical perspective will be welcomed \n\nby scholars and students troubled by society's growing reliance on \n\n technological solutions to complex social and political problems.'), Document(metadata={'page': 172, 'year': '2004', 'title': 'The Daily Drucker: 366 Days of Insight and Motivation for Getting the Right Things Done', 'author': 'Drucker, Peter F.', 'source': 'The Daily Drucker-2004.pdf'}, page_content='. \n\n \n\nACTION POINT: Abandon what is about to be obsolete, develop a system to \n\nexploit your successes, and develop a systematic approach to innovation. \n\n"Management's New Paradigms," Forbes \n\nManagement Challenges for the 21st Century \n\nThe Next Society (Corpedia Online Program)'), Document(metadata={'page': 87, 'year': '2008', 'title': 'The Essential Drucker', 'author': 'Peter F. Drucker', 'source': 'The Essential Drucker-2008.pdf'}, page_content='. \n\n \n\nA Philosophy of Management \n\nWhat the business enterprise needs is a principle of management that will give full scope to \n\nindividual strength and responsibility, and at the same time give common direction of vision and \n\neffort, establish team work, and harmonize the goals of the individual with the commonweal. \n\nThe only principle that can do this is management by objectives and self-control. It makes the \n\ncommonweal the aim of every manager'), Document(metadata={'page': 797, 'year': '2004', 'title': 'The Daily Drucker: 366 Days of Insight and Motivation for Getting the Right Things Done', 'author': 'Drucker, Peter F.', 'source': 'The Daily Drucker-2004.pdf'}, page_content='. The first part of the book \n\n treats the philosophical shift from a Cartesian universe of mechanical cause \n\n into a new universe of pattern, purpose, and configuration. Drucker discusses \n\n the need to organize men of knowledge and of high skill for joint effort, and \n\n performance as a key component of this change'), Document(metadata={'page': 727, 'year': '2004', 'title': 'The Daily Drucker: 366 Days of Insight and Motivation for Getting the Right Things Done', 'author': 'Drucker, Peter F.', 'source': 'The Daily Drucker-2004.pdf'}, page_content='. \n\n\nForty years ago we built into the performance review of managerial \n\npeople the question, "Are they ready for promotion?" Now we need to \n\nreplace that question with "Are they ready for a bigger, more demanding \n\nchallenge and for the addition of new responsibilities to their existing job?" \n\n \n\nACTION POINT: Create a flat organization. Use information processing-its \n\nstructure, its content, and its direction-to ensure that your organization is \n\nagile and effective. \n\nThe Frontiers of Management'), Document(metadata={'page': 796, 'year': '2004', 'title': 'The Daily Drucker: 366 Days of Insight and Motivation for Getting the Right Things Done', 'author': 'Drucker, Peter F.', 'source': 'The Daily Drucker-2004.pdf'}, page_content='. ' He is

y, and industrial sociology, and has succeeded admirably\nin harmonizing the findings of all four disciplines and applying them\nmeaningfully to the practical problems of the 'enterprise.' " Drucker believes\nthat the interests of the worker, management, and corporation are\nreconcilable with society. He advances the idea of "the plant community" in\nwhich workers are encouraged to take on more responsibility and act like\n"managers"), Document(metadata={'page': 796, 'year': '2004', 'title': 'The Daily Drucker: 36 6 Days of Insight and Motivation for Getting the Right Things Done', 'author': 'Drucker, Peter F.', 'source': 'The Daily Drucker-2004.pdf'}, page_content='. Drucker looks at General Motors\' \nmanagerial organization and attempts to understand what makes the company\nwork so effectively. Certain questions are addressed, such as: "What are the\ncompany's core principles, and how do they contribute to the success of the\norganization?" The principles of organization and management at General\nMotors described in this book became models for organizations worldwide'))}

---CHECK DOCUMENT RELEVANCE TO QUESTION---

---GRADE: DOCUMENT NOT RELEVANT---

---GRADE: DOCUMENT NOT RELEVANT---

---GRADE: DOCUMENT NOT RELEVANT---

---GRADE: DOCUMENT NOT RELEVANT---

---GRADE: DOCUMENT NOT RELEVANT---

---GRADE: DOCUMENT NOT RELEVANT---

---GRADE: DOCUMENT NOT RELEVANT---

---GRADE: DOCUMENT NOT RELEVANT---

---GRADE: DOCUMENT NOT RELEVANT---

---GRADE: DOCUMENT NOT RELEVANT---

---GRADE: DOCUMENT NOT RELEVANT---

---FILTERED OUT 100.0% OF IRRELEVANT DOCUMENTS---

---**fail**---

---CHECK GENERATION CONDITION---

---DECISION: MORE THAN HALF OF THE DOCUMENTS ARE IRRELEVANT TO QUESTION, NOW INCLUDE WEB SEARCH---

{'question': '\nSearch the database first to see how Peter Drucker might advise a sentient AI collective attempting to design its own management philosophy \nafter rejecting all human organizational models as insufficient for post-biological cognition?\n', 'documents': [], 'checker_result': 'fail'}

---WEB SEARCH---

Total number of web search documents: 5

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4/03/ORGANIZATIONAL-COGNITION-LEARNING-ORGANIZATION-Luca-Iandoli.pdf'}, 'page_content': 'Summary: "This book presents a theory of learning based on a model of organizational memory, explaining organizational processes and dynamics through which', {'metadata': {'title': '[PDF] The culture of neural networks - Monoskop', 'url': 'https://monoskop.org/images/5/5c/Piorecky_Karel_Husarova_Zuzana_The_Culture_of_Neural_Networks_2024.pdf'}, 'page_content': 'That is, it is about the shorthand that makes artificial neural network technology an achievable objective of artificial intelligence, though this still remains.'}], 'checker_result': 'fail'}

---ANSWER GENERATION---

Answer generation has been generated.

---CHECK HALLUCINATIONS WITH DOCUMENTS---

---DECISION: GENERATION IS GROUNDED IN DOCUMENTS---

---VERIFY ANSWER WITH QUESTION---

---DECISION: GENERATION DOES NOT ADDRESS QUESTION, RE-WRITE QUERY---

This is the 1st attempt.

{ 'question': '\nSearch the database first to see how Peter Drucker might advise a sentient AI collective attempting to design its own management philosophy \nafter rejecting all human organizational models as insufficient for post-biological cognition?\n', 'generation': "The context provided does not contain specific information about how Peter Drucker might advise a sentient AI collective in designing its own management philosophy after rejecting human organizational models. Therefore, an answer based on the provided documents cannot be given.\n\n**Reference Section**:\n- No specific references are available in the context provided for the question about Peter Drucker's advice.", 'documents': [{ 'metadata': {'title': '[PDF] An Evolution of Leadership from Ancient Times to the Digital Age', 'url': 'https://www.econstor.eu/bitstream/10419/305351/1/Taylor-Francis_9781003490470.pdf'}, 'page_content': 'They emphasize the importance of amalgamating philosophical wisdom with the promises and challenges brought about by AI. The book will guide readers from the', {'metadata': {'title': '[PDF] Experts Imagine the Impact of Artificial Intelligence by 2040', 'url': 'https://imaginingthedigitalfuture.org/wp-content/uploads/2024/02/AI2040-FINAL-White-Paper-2-2.29.24.pdf'}, 'page_content': "Elon's Center sampled from a database of experts to collect a broad array of opinions about the potential impact of humans' design and uses of"}, {'metadata': {'title': '[PDF] ORGANIZATIONAL DEVELOPMENT - Alexis Press', 'url': 'https://alexispress.us/wp-content/uploads/2022/07/Organizational-Development.pdf'}, 'page_content': 'ABSTRACT:Situational approach or the p approach/theory. According to this approach management issues in all businesses since issue.'}, {'metadata': {'title': '[PDF] Organizational Cognition and Learning', 'url': 'https://perpustakaan.unaim-wamena.ac.id/wp-content/uploads/2024/03/ORGANIZATIONAL-COGNITION-LEARNING-ORGANIZATION-Luca-Iandoli.pdf'}, 'page_content': 'Summary: "This book presents a theory of learning based on a model of organizational memory, explaining organizational processes and dynamics through which', {'metadata': {'title': '[PDF] The culture of neural networks - Monoskop', 'url': 'https://monoskop.org/images/5/5c/Piorecky_Karel_Husarova_Zuzana_The_Culture_of_Neural_Networks_2024.pdf'}, 'page_content': 'That is, it is about the shorthand that makes artificial neural network technology an achievable objective of artificial intelligence, though this still remains.'}], 'checker_result': 'fail'}

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references are available in the context provided for the question about Peter Drucker's advice.", 'answer_verifier_attempts': 1, 'documents': [{'metadata': {'title': '[PDF] An Evolution of Leadership from Ancient Times to the Digital Age', 'url': 'https://www.econstor.eu/bitstream/10419/305351/1/Taylor-Francis_9781003490470.pdf'}, 'page_content': 'They emphasize the importance of amalgamating philosophical wisdom with the promises and challenges brought about by AI. The book will guide readers from the'}, {'metadata': {'title': '[PDF] Experts Imagine the Impact of Artificial Intelligence by 2040', 'url': 'https://imaginingthefuture.org/wp-content/uploads/2024/02/AI2040-FINAL-White-Paper-2-2.29.24.pdf'}, 'page_content': 'Elon's Center sampled from a database of experts to collect a broad array of opinions about the potential impact of humans' design and uses of'}, {'metadata': {'title': '[PDF] ORGANIZATIONAL DEVELOPMENT - Alexis Press', 'url': 'https://alexispress.us/wp-content/uploads/2022/07/Organizational-Development.pdf'}, 'page_content': 'ABSTRACT:Situational approach or the p approach/theory. According to this approach management issues in all businesses since issue.'}, {'metadata': {'title': '[PDF] Organizational Cognition and Learning', 'url': 'https://perpustakaan.unaim-wamena.ac.id/wp-content/uploads/2024/03/ORGANIZATIONAL-COGNITION-LEARNING-ORGANIZATION-Luca-Iandoli.pdf'}, 'page_content': 'Summary: "This book presents a theory of learning based on a model of organizational memory, explaining organizational processes and dynamics through which'}, {'metadata': {'title': '[PDF] The culture of neural networks - Monoskop', 'url': 'https://monoskop.org/images/5/5c/Piorecky_Karel_Husarova_Zuzana_The_Culture_of_Neural_Networks_2024.pdf'}, 'page_content': 'That is, it is about the shorthand that makes artificial neural network technology an achievable objective of artificial intelligence, though this still remains.'}], 'checker_result': 'fail'}

---QUERY REWRITE---

{ 'question': "How might Peter Drucker's principles on management and leadership be applied by a sentient AI collective in developing a new management philosophy, especially in scenarios where traditional human organizational models are inadequate for non-human cognition?", 'original_question': '\nSearch the database first to see how Peter Drucker might advise a sentient AI collective attempting to design its own management philosophy \nafter rejecting all human organizational models as insufficient for post-biological cognition?\n', 'generation': "The context provided does not contain specific information about how Peter Drucker might advise a sentient AI collective in designing its own management philosophy after rejecting human organizational models. Therefore, an answer based on the provided documents cannot be given.\n\n**Reference Section**:\n- No specific references are available in the context provided for the question about Peter Drucker's advice.", 'answer_verifier_attempts': 1, 'documents': [{'metadata': {'title': '[PDF] An Evolution of Leadership from Ancient Times to the Digital Age', 'url': 'https://www.econstor.eu/bitstream/10419/305351/1/Taylor-Francis_9781003490470.pdf'}, 'page_content': 'They emphasize the importance of amalgamating philosophical wisdom with the promises and challenges brought about by AI. The book will guide readers from the'}, {'metadata': {'title': '[PDF] Experts Imagine the Impact of Artificial Intelligence by 2040', 'url': 'https://imaginingthefuture.org/wp-content/uploads/2024/02/AI2040-FINAL-White-Paper-2-2.29.24.pdf'}, 'page_content': 'Elon's Center sampled from a database of experts to collect a broad array of opinions about the potential impact of humans' design and uses of'}, {'metadata': {'title': '[PDF] ORGANIZATIONAL DEVELOPMENT - Alexis Press', 'url': 'https://alexispress.us/wp-content/uploads/2022/07/Organizational-Development.pdf'}, 'page_content': 'ABSTRACT:Situational approach or the p approach/theory. According to this approach management issues in all businesses since issue.'}, {'metadata': {'title': '[PDF] Organizational Cognition and Learning', 'url': 'https://perpustakaan.unaim-wamena.ac.id/wp-content/uploads/2024/03/ORGANIZATIONAL-COGNITION-LEARNING-ORGANIZATION-Luca-Iandoli.pdf'}, 'p

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a model of organizational memory, explaining organizational processes and  
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ral networks - Monoskop', 'url': 'https://monoskop.org/images/5/5c/Pioreck  
y_Karel_Husarova_Zuzana_The_Culture_of_Neural_Networks_2024.pdf'}, 'page_c  
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network technology an achievable objective of artificial intelligence, tho  
ugh this still remains.'}], 'checker_result': 'fail'}
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---QUERY TRANSLATION AND RAG-FUSION---

Total number of results: 7

Document 1 from `The Daily Drucker-2004.pdf`, page 799

Document 2 from `The Daily Drucker-2004.pdf`, page 796

Document 3 from `The Daily Drucker-2004.pdf`, page 365

Document 4 from `The Essential Drucker-2008.pdf`, page 87

Document 5 from `The Essential Drucker-2008.pdf`, page 58

Document 6 from `The Daily Drucker-2004.pdf`, page 727

Document 7 from `The Daily Drucker-2004.pdf`, page 172

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```


s cannot be given.\n\n**Reference Section**:\n- No specific references are available in the context provided for the question about Peter Drucker's advice.", 'answer_verifier_attempts': 1, 'documents': [Document(metadata={'page': 58, 'year': '2008', 'title': 'The Essential Drucker', 'author': 'Peter F. Drucker', 'source': 'The Essential Drucker-2008.pdf'}, page_content='. Taylor. This will require, above all, very \ndifferent assumptions about people in organizations and their work: \n(a) One does not "manage" people. \n(b) The task is to lead people. \n(c) And the goal is to make productive the specific strengths and knowledge of each individual. \n(d) Technologies and End Uses Are Fixed and Given. \nFour major assumptions, as stated above, have been underlying the practice of management all \nalong-in fact for much longer than there has been a discipline of management'), {'metadata': {'title': '[PDF] Adventures in Higher Education Policy - The Millennium Project', 'url': 'http://milproj.dc.umich.edu/pdfs/2017/2017%20Adventures%20in%20HE%20Policy%20LR.pdf'}, 'page_content': "... new technologies are at first inadequate to displace today's technology in existing applications, they later explosively displace the application as they enable"}]], 'checker_result': 'fail'}

---ANSWER GENERATION---

Answer generation has been generated.

---CHECK HALLUCINATIONS WITH DOCUMENTS---

---DECISION: GENERATION IS GROUNDED IN DOCUMENTS---

---VERIFY ANSWER WITH QUESTION---

---DECISION: GENERATION ADDRESSES QUESTION---

{'question': "How might Peter Drucker's principles on management and leadership be applied by a sentient AI collective in developing a new management philosophy, especially in scenarios where traditional human organizational models are inadequate for non-human cognition?", 'original_question': '\nSearch the database first to see how Peter Drucker might advise a sentient AI collective attempting to design its own management philosophy \nafter rejecting all human organizational models as insufficient for post-biological cognition?\n', 'generation': 'The context provided includes insights from Peter Drucker, particularly from "The Essential Drucker." Based on this, Drucker might advise a sentient AI collective designing its own management philosophy to focus on leading rather than managing individuals. He emphasizes making productive the specific strengths and knowledge of each individual, implying that the AI should harness the unique capabilities of its components. Drucker also suggests considering that technologies and end uses are fixed and given, which could be interpreted as advising the AI to navigate within certain constraints or established frameworks. However, detailed advice specific to a sentient AI collective is not covered in the provided context.\n\n**Reference Section**:\n- Drucker, P. F. (2008). *The Essential Drucker*. ', 'answer_verifier_attempts': 1, 'documents': [Document(metadata={'page': 58, 'year': '2008', 'title': 'The Essential Drucker', 'author': 'Peter F. Drucker', 'source': 'The Essential Drucker-2008.pdf'}, page_content='. Taylor. This will require, above all, very \ndifferent assumptions about people in organizations and their work: \n(a) One does not "manage" people. \n(b) The task is to lead people. \n(c) And the goal is to make productive the specific strengths and knowledge of each individual. \n(d) Technologies and End Uses Are Fixed and Given. \nFour major assumptions, as stated above, have been underlying the practice of management all \nalong-in fact for much longer than there has been a discipline of management'), {'metadata': {'title': '[PDF] Adventures in Higher Education Policy - The Millennium Project', 'url': 'http://milproj.dc.umich.edu/pdfs/2017/2017%20Adventures%20in%20HE%20Policy%20LR.pdf'}, 'page_content': "... new technologies are at first inadequate to displace today's technology in existing applications, they later explosively displace the application as they enable"}]], 'checker_result': 'fail'}

===== Final Output
=====

The context provided includes insights from Peter Drucker, particularly from "The Essential Drucker." Based on this, Drucker might advise a sentient AI collective designing its own management philosophy to focus on leading rather than managing individuals. He emphasizes making productive the specific strengths and knowledge of each individual, implying that the AI should harness the unique capabilities of its components. Drucker also suggests considering that technologies and end uses are fixed and given, which could be interpreted as advising the AI to navigate within certain constraints or established frameworks. However, detailed advice specific to a sentient AI collective is not covered in the provided context.

Reference Section:

- Drucker, P. F. (2008). *The Essential Drucker*.

4. Potential Improvements

- Model & Embedding Optimization

Experiment with different embedding models to assess their impact on retrieval accuracy, semantic matching, and downstream response quality.

- Web Search Enhancement

Test alternative web search APIs/tools to identify which service delivers the most relevant, reliable, and up-to-date content.

- Agent-Specific Model Selection

Assign different LLMs tailored to each agent's function (e.g., lighter models for routing, stronger ones for generation) to improve both efficiency and performance.

- Contextual Result Accumulation

Support multi-turn or follow-up queries by preserving and reusing relevant results from earlier steps, enabling more context-aware and coherent interactions.

- Advanced Prompt Engineering

Iteratively refine and customize prompts for each agent to better handle edge cases, ambiguous queries, or complex reasoning. Regularly create challenging test cases to evaluate agent robustness and adaptability.