



OpenAI Tutorial

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A computer monitor is shown in the foreground, displaying a code editor with Python code. The code includes comments in German and uses the OpenAI Python SDK to set up an API client. The monitor is on a desk, and a keyboard is partially visible in the bottom left corner. The background is slightly blurred, showing a bookshelf.

Setting API KEY

The Chat Completions API is the core interface for interacting with OpenAI's language models. It provides a flexible way to hold conversations with AI models like GPT-4o.

API Key Setup:

```
from openai import OpenAI  
client = OpenAI(api_key = "YOUR_API_KEY")
```

Chat Completions API

Basic Parameters:

- Model (request): Select different model like "gpt-4o" or "gpt-o1" model
- Messages (request): A list of messages that form the conversation, include "developer"(system), "assistant" and "user"
- Temperature: controls randomness in the output between 0 and 2. Higher values like 0.8 will make the output more random
- More Like "max_token", "stream"...

This is a basic chat completions api request structure:

```
response = client.chat.completions.create(  
    model="gpt-4o",  
    messages=[  
        {"role": "system", "content": "You are a helpful assistant."},  
        {"role": "user", "content": "Hello, how can AI help in education?"}  
    ]  
)
```

Conversation Management

```
# Building a conversation
conversation = [
    {"role": "system", "content": "You are a mathematics tutor helping with calculus."}
]

# First user question
conversation.append({"role": "user", "content": "What is a derivative?"})

response = client.chat.completions.create(
    model="gpt-4o",
    messages=conversation
)

# Add the assistant's response to the conversation
conversation.append({"role": "assistant", "content": response.choices[0].message.content})

# Second user question
conversation.append({"role": "user", "content": "Can you give me an example of using derivatives?"})

# Get next response with full conversation history
next_response = client.chat.completions.create(
    model="gpt-4o",
    messages=conversation
)

print(next_response.choices[0].message.content)
```

Function Calling

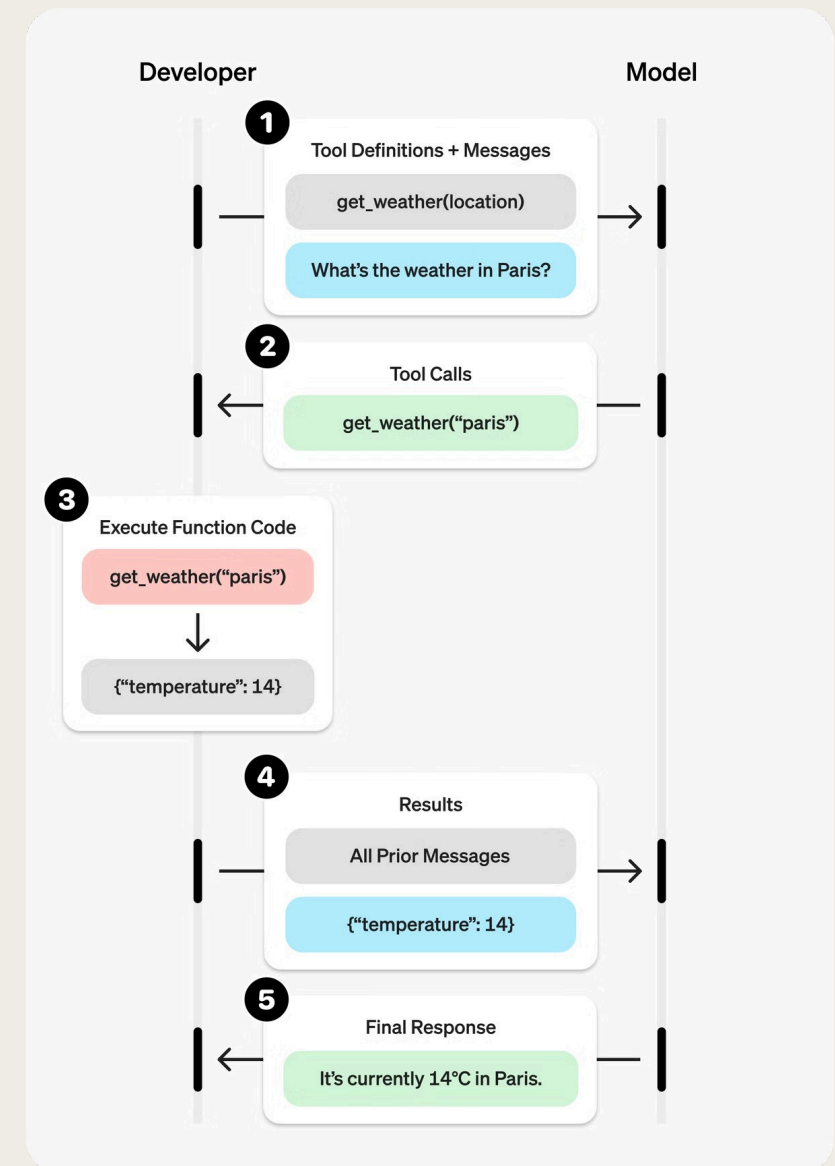
- 1 Call model with functions defined – along with your system and user messages
- 2 Model decides to call function(s) – model returns the name and input arguments
- 3 Execute function code – parse the model's response and handle function calls
- 4 Supply model with results – so it can incorporate them into its final response
- 5 Model responds – incorporating the result in its output

Function Calling Sample Code

```
tools = [{  
  "type": "function",  
  "function": {  
    "name": "get_weather",  
    "description": "Get current temperature for provided coordinates  
in celsius.",  
    "parameters": {  
      "type": "object",  
      "properties": {  
        "latitude": {"type": "number"},  
        "longitude": {"type": "number"}  
      },  
      "required": ["latitude", "longitude"],  
      "additionalProperties": False  
    },  
    "strict": True  
  }  
}]
```

```
messages = [{"role": "user", "content": "What's the weather like in Paris  
today?"}]
```

```
completion = client.chat.completions.create(  
  model="gpt-4o",  
  messages=messages,  
  tools=tools,  
)
```



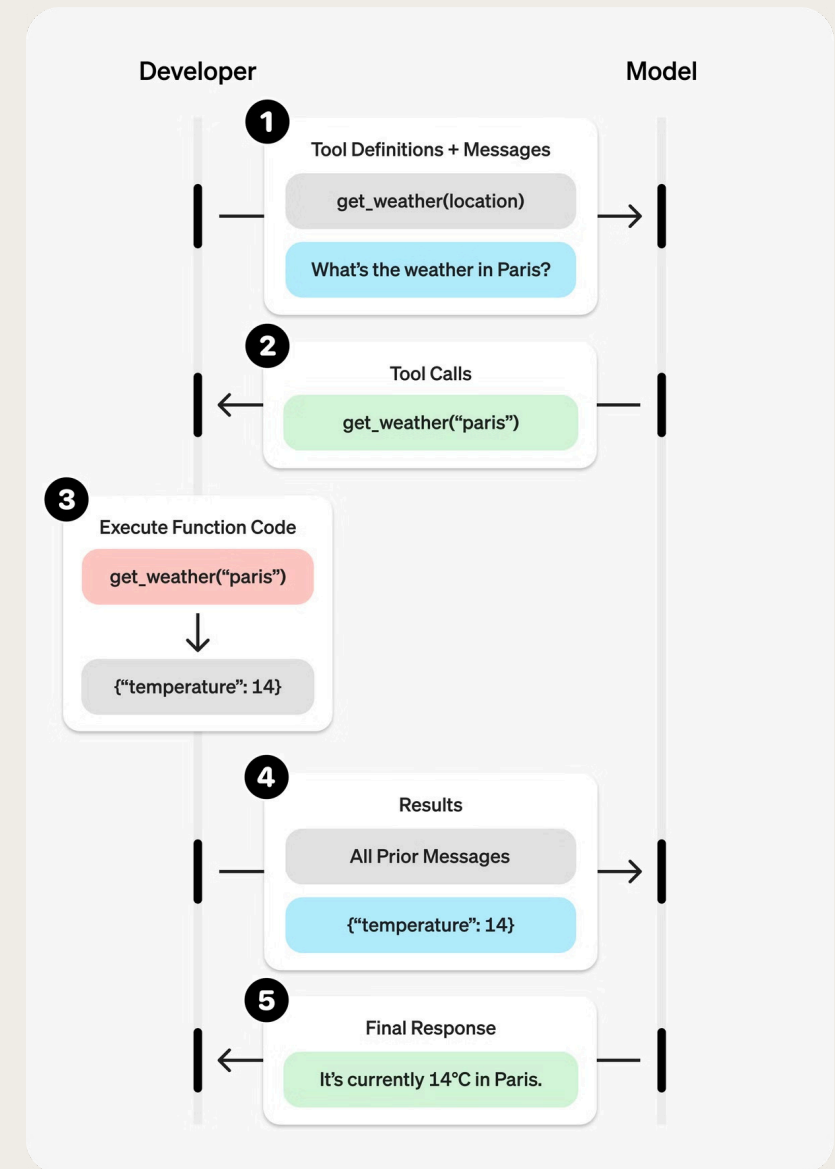
Function Calling Sample Code

```
# Step 3: Execute get_weather function
tool_call = completion.choices[0].message.tool_calls[0]
args = json.loads(tool_call.function.arguments)

result = get_weather(args["latitude"], args["longitude"])

# Step 4: Supply result and call model again
messages.append(completion.choices[0].message)
messages.append({
    "role": "tool",
    "tool_call_id": tool_call.id,
    "content": str(result)
})

completion_2 = client.chat.completions.create(
    model="gpt-4o",
    messages=messages,
    tools=tools,
)
```



Structure Output Json Mode

Structured Outputs is a feature that ensures the model will always generate responses that adhere to your supplied **JSON Schema**, so you don't need to worry about the model omitting a required key, or hallucinating an invalid enum value.

```
class CalendarEvent(BaseModel):
    name: str
    date: str
    participants: list[str]

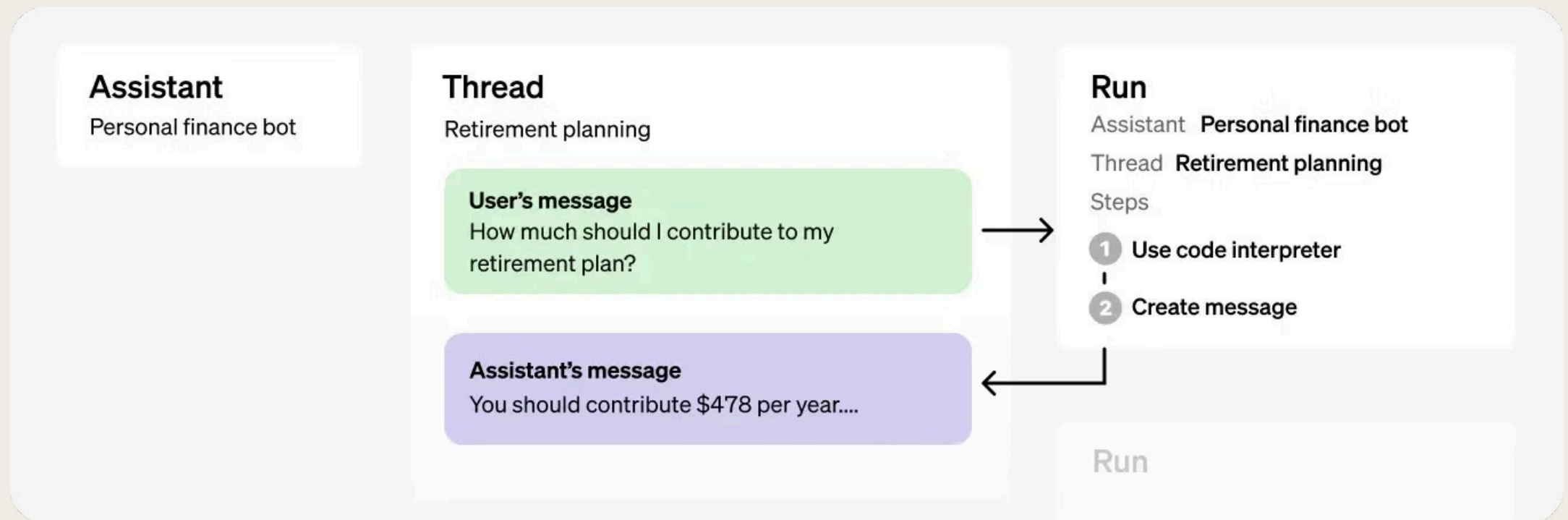
completion = client.beta.chat.completions.parse(
    model="gpt-4o-2024-08-06",
    messages=[
        {"role": "system", "content": "Extract the event information."},
        {"role": "user", "content": "Alice and Bob are going to a science fair on Friday."},
    ],
    response_format=CalendarEvent,
)

event = completion.choices[0].message.parsed
```

Vision Image Input

```
response = client.chat.completions.create(  
    model="gpt-4o-mini",  
    messages=[  
        {  
            "role": "user",  
            "content": [  
                {  
                    "type": "text",  
                    "text": "What are in these images? Is there any difference between them?",  
                },  
                {  
                    "type": "image_url",  
                    "image_url": {  
                        "url": "https://upload.wikimedia.org/wikipedia/.../2560px-Gfp-wisconsin-madison-the-nature-boardwalk.jpg",  
                    },  
                },  
                {  
                    "type": "image_url",  
                    "image_url": {"url": f"data:image/jpeg;base64,{base64_image}"},  
                },  
            ],  
        },  
    ],  
)
```

Assistants API Overview



1 Persistent Conversation Management

Assistants API automatically maintains conversation history and context across sessions, eliminating the need for developers to manually track and manage message history with each request.

2 Integrated Tool Ecosystem

Assistants API provides built-in tools like Code Interpreter, Retrieval, and File Processing that work seamlessly within the API, enabling complex capabilities without requiring separate infrastructure or custom implementations.

3 Structured Interaction Framework

Assistants API offers a more organized architecture through its Assistant, Thread, and Run components, creating clearer separation of concerns and more maintainable code for complex applications.

Assistant API Steps

- 1 Create an Assistant by defining its custom instructions and picking a model. If helpful, add files and enable tools like Code Interpreter, File Search, and Function calling.

An Assistant represents an entity that can be configured to respond to a user's messages using several parameters like model, instructions, and tools.

- 2 Create a Thread when a user starts a conversation.

A Thread represents a conversation between a user and one or many Assistants. You can create a Thread when a user (or your AI application) starts a conversation with your Assistant.

- 3 Add Messages to the Thread as the user asks questions.

The contents of the messages your users or applications create are added as Message objects to the Thread. Messages can contain both text and files. There is a limit of 100,000 Messages per Thread and we smartly truncate any context that does not fit into the model's context window.

- 4 Run the Assistant on the Thread to generate a response by calling the model and the tools.

Once all the user Messages have been added to the Thread, you can Run the Thread with any Assistant. Creating a Run uses the model and tools associated with the Assistant to generate a response. These responses are added to the Thread as assistant Messages.

Step 1: Create an Assistant

```
# Basic assistant creation
assistant = client.beta.assistants.create(
    name="Math Tutor",
    instructions="You are a helpful math tutor. Explain concepts clearly and provide step-by-step solutions.",
    model="gpt-4o"
)

print(f"Created assistant with ID: {assistant.id}")
```

Step 2: Create a Thread

```
# Create a new thread
thread = client.beta.threads.create()
print(f"Created thread with ID: {thread.id}")
```

Multiple:

```
# Create threads for different users or conversations
user1_thread = client.beta.threads.create()
user2_thread = client.beta.threads.create()
```

Step 3: Add a Message to the Thread

```
# Add a message to the thread
message = client.beta.threads.messages.create(
    thread_id=thread.id,
    role="user",
    content="I need to solve the equation `3x + 11 = 14`. Can you help me?"
)
```

Multiple:

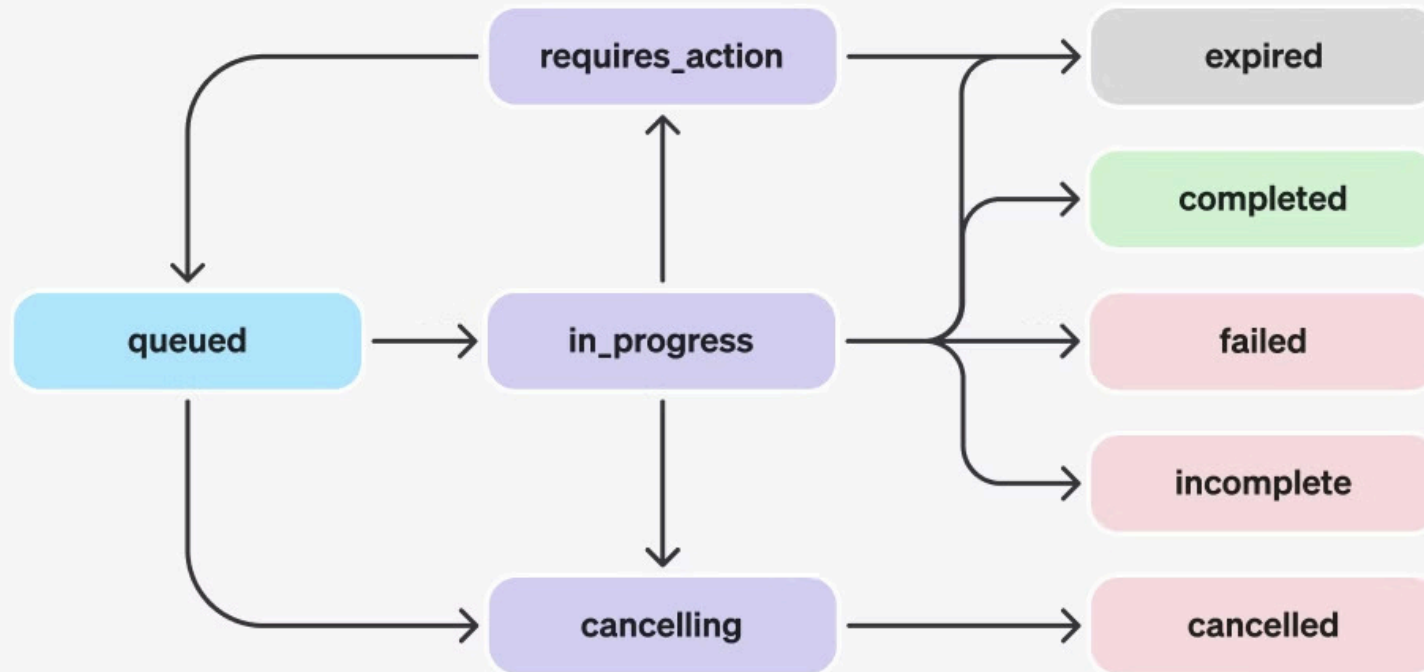
```
# Add messages to different threads
client.beta.threads.messages.create(
    thread_id=user1_thread.id,
    role="user",
    content="I need help with trigonometry"
)

client.beta.threads.messages.create(
    thread_id=user2_thread.id,
    role="user",
    content="Can you explain calculus to me?"
)
```

Step 4: Create a Run

```
# Run the assistant on the thread
run = client.beta.threads.runs.create(
    thread_id=thread.id,
    assistant_id=assistant.id
)
```

Runs and Run Steps



Assistants API Tools

1

File Search

Built-in RAG tool to process and search through files

2

Code Interpreter

Write and run python code, process files and diverse data

3

Function Calling

Use your own custom functions to interact with your application

File Search

```
# Create a vector store caled "Financial Statements"
vector_store = client.beta.vector_stores.create(name="Financial Statements")

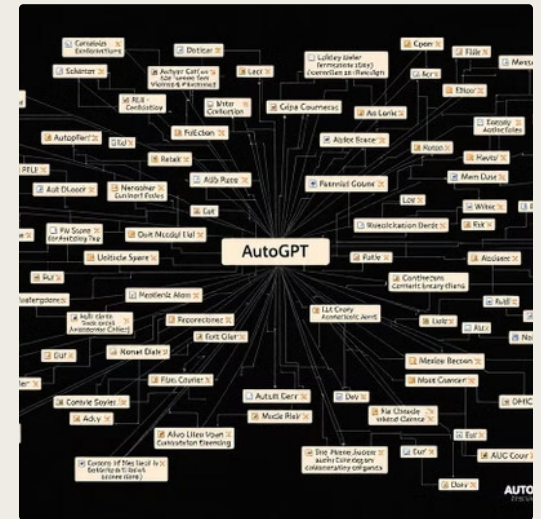
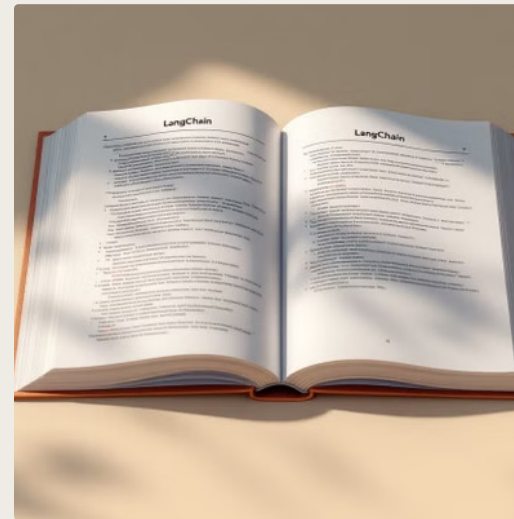
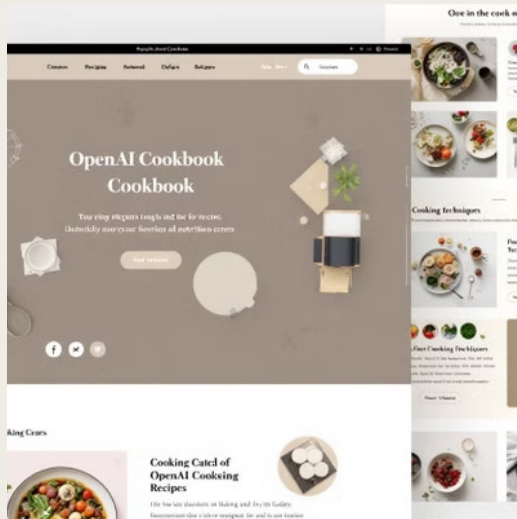
# Ready the files for upload to OpenAI
file_paths = ["edgar/goog-10k.pdf", "edgar/brka-10k.txt"]
file_streams = [open(path, "rb") for path in file_paths]

# Use the upload and poll SDK helper to upload the files, add them to the vector store,
# and poll the status of the file batch for completion.
file_batch = client.beta.vector_stores.file_batches.upload_and_poll(
    vector_store_id=vector_store.id, files=file_streams
)

# You can print the status and the file counts of the batch to see the result of this operation.
print(file_batch.status)
print(file_batch.file_counts)

# Update the assistant to use the new Vector Store
assistant = client.beta.assistants.update(
    assistant_id=assistant.id,
    tool_resources={"file_search": {"vector_store_ids": [vector_store.id]}},
)
```

Extended Reading



- OpenAI Document
- Prompt engineering
- LangChain
- AutoGPT