# **GEMINI TELEGRAM BOT**

# # Import necessary modules

```
import os # Provides functions for interacting with the operating system
import io # Handles I/O operations (e.g., working with in-memory streams)
import docx # Allows reading and writing Microsoft Word (.docx) files
import time # Provides time-related functions (e.g., delays, timestamps)
import PyPDF2 # Enables reading and manipulating PDF files
import asyncio # Supports asynchronous programming with coroutines
import logging # Handles logging for debugging and monitoring
import PIL.Image
# Allows image processing using the Python Imaging Library (Pillow)
from datetime import datetime # Deals with date and time operations
from textblob import TextBlob
# Performs NLP tasks like sentiment analysis
from pymongo import MongoClient # Connects to and interacts with MongoDB
from serpapi import GoogleSearch
# Fetches search results from Google using SerpAPI
from pyrogram import Client, filters
# Handles Telegram bot interactions using Pyrogram
from pyrogram.types import Message, ReplyKeyboardMarkup, KeyboardButton
Telegram-specific UI elements
from pyrogram.enums import ParseMode # Defines text formatting options
for Telegram (Markdown, HTML, etc.)
from pyrogram.errors import FloodWait
# Handles rate-limiting errors in Telegram bots
import google.generativeai as genai
# Provides access to Google's Generative AI models
from config import (
# Imports bot credentials and API keys from a configuration file
   API ID, API HASH, BOT TOKEN, GOOGLE API KEY, MODEL NAME, SERP API KEY
# Set up logging configuration to monitor bot activity
logging.basicConfig(level=logging.INFO) # Set logging level to INFO
logger = logging.getLogger(__name__) # Create a logger instance
# Initialize the Pyrogram Client for the Telegram bot
app = Client(
    "gemini session", # Name of the session
   api id=API ID,  # Telegram API ID (from config)
   api hash=API HASH, # Telegram API Hash (from config)
```

```
bot_token=BOT_TOKEN, # Telegram Bot Token (from config)
    parse_mode=ParseMode.MARKDOWN # Enables Markdown formatting in
messages
)

# Validate that the Google API key is set before proceeding

if not GOOGLE_API_KEY:
    logger.error("GOOGLE_API_KEY is missing! Please check your
config.") # Log an error if API key is missing
    exit(1) # Terminate the script

# Configure Google Generative AI with the provided API key
genai.configure(api_key=GOOGLE_API_KEY)

# Initialize the AI model with the specified model name

model = genai.GenerativeModel(MODEL_NAME)

# MongoDB Connection
```

# # MongoDB Connection MONGO\_URI = "your-mongodb-url" client = MongoClient(MONGO\_URI, serverSelectionTimeoutMS=5000) # Initialize MongoDB client with a 5-second timeout # Verify MongoDB connection try: client.admin.command('ping') # Sends a "ping" command to check if the database is reachable logger.info("Connected to MongoDB") # Logs successful connection except Exception as e: logger.error(f"MongoDB connection failed: {e}") # Logs error message if connection fails exit(1) # Exits the script if the database connection cannot be established

# Define MongoDB database and collections

```
db = client["telegram bot"] # Select the 'telegram bot' database
users collection = db["users"] # Collection for storing user details
chat history collection = db["chat history"]
# Collection for storing chat history
file metadata collection = db["file metadata"]
# Collection for storing metadata of uploaded files
websearch history collection = db["websearch history"]
# Collection for storing web search history
sentiment history = db["sentiment history"]
# Collection for storing sentiment analysis results
# Bot Commands Help Text
HELP TEXT = """
**Bot Commands: **
**/start** - Start the bot and register yourself.
**/text <prompt>** - Generate AI-powered responses for any query.
**/img** - Analyze and describe images using AI.
**/file** - Upload a document to get a summarized version.
**/sentiment <text>** - Analyze the sentiment of the given text.
**/websearch <query>** - Search the web and get top results
instantly.
i **/help** - View this command list anytime.
# Decorator function to append the help message after
executing a command
def append help(func):
   async def wrapper(client: Client, message: Message):
       await func(client, message) # Execute the original function
       await message.reply text(HELP TEXT)
     # Send the help text after executing the command
   return wrapper
# Function to analyze sentiment of a given text
def get sentiment(text):
   analysis = TextBlob(text) # Create a TextBlob object to analyze text
   polarity = analysis.sentiment.polarity # Extract the polarity score
(-1 to 1)
    # Determine sentiment based on polarity score
  if polarity > 0:
```

```
return "positive"
     # Positive sentiment if polarity is greater than 0
   elif polarity < 0:</pre>
       return "negative" # Negative sentiment if polarity is less than 0
   else:
       return "neutral" # Neutral sentiment if polarity is exactly 0
# Command handler for the /start command
@app.on message(filters.command("start"))
@append help # Adds the help message after executing this command
async def start handler(client: Client, message: Message):
       user id = message.from user.id # Extract the user ID
       user data = users collection.find one({"chat id": user id})
     # Check if user exists in the database
       if not user data:
     # If the user is not found, register them in the database
           users collection.insert one({
               "chat id": user id,
               "first name": message.from user.first name,
           # Store user's first name
               "last name": message.from user.last name,
              # Store user's last name
              "username": message.from user.username,
           # Store Telegram username
               "phone number": None
           # Placeholder for phone number (to be updated later)
           })
           # Create a keyboard button to request the user's phone number
           keyboard = ReplyKeyboardMarkup(
                [[KeyboardButton("Share Phone Number",
request contact=True)]], # Button for phone sharing
            resize keyboard=True
           # Resize the keyboard for better display
           # Send a welcome message prompting the user to share their
phone number
           await message.reply text("Welcome! Please share your phone
number.", reply markup=keyboard)
   else:
```

```
# If user is already registered, greet them
           await message.reply text("Welcome back!")
   except Exception as e:
        logger.error(f"Error in start handler: {e}") # Log any errors
that occur
# Handler for when user shares their contact (phonenumber)
@app.on message(filters.contact)
async def save phone number (client: Client, message: Message):
   try:
       user id = message.from user.id # Get the user ID
       contact number = message.contact.phone number # Extract phone
number from the contact message
        # Update the user's phone number in the database
       users collection.update one({"chat id": user id}, {"$set":
{"phone number": contact number}})
        # Confirm successful phone number storage
       await message.reply text(" < Your phone number has been saved
successfully!")
   except Exception as e:
       logger.error(f"Error in save phone number: {e}") # Log any errors
```

# # Handler for the /text command, which generates AI responses

```
@app.on_message(filters.command("text"))
@append_help # Adds help text after executing the command
async def gemi_handler(client: Client, message: Message):
    # Send a temporary loading message to inform the user
    loading_message = await message.reply_text("**Generating response,
please wait...**")

try:
    # Ensure the user provided a prompt
    if len(message.text.strip()) <= 5:
        await message.reply_text("**Provide a prompt after the
command.**")</pre>
```

```
return # Exit the function if no valid prompt is provided
        # Extract the user prompt (text after the command)
        prompt = message.text.split(maxsplit=1)[1]
        # Generate a response using the AI model
        response = model.generate content(prompt)
        # Extract the generated text response
        response text = response.text
        # Perform sentiment analysis on the user's prompt
        sentiment = get sentiment(prompt)
        # Store the conversation details in MongoDB
        chat history collection.insert one({
            "chat id": message.from user.id,  # Store user ID
            "user query": prompt, # Store the user's query
            "bot response": response text, # Store the generated response
            "sentiment": sentiment, # Store the analyzed sentiment
            "timestamp": datetime.utcnow()  # Store the timestamp (UTC
format)
        })
        # Customize the bot's response based on the sentiment of the
prompt
        sentiment responses = {
            "positive": " Great! Here's what I came up with:",
            "negative": " I sense that things might be tough. Here's
something that might help:",
            "neutral": " Here's the information you requested:"
        response text = f"{sentiment responses.get(sentiment,
'') } \n\n{response text}"
        # Ensure the message length does not exceed Telegram's 4096
character limit
        await message.reply text(response text if len(response text) <=</pre>
4000 else response text[:4000] + "...")
   except Exception as e:
        # Handle errors gracefully and inform the user
        await message.reply text(f"**An error occurred: {str(e)}**")
  finally:
```

```
# Delete the loading message once the response is generated
        await loading message.delete()
# Handler for the /img command, which analyzes an image
sent by the user
@app.on message(filters.command("img") & filters.photo)
@append help # Adds help text after executing the command
async def analyze image(client: Client, message: Message):
    # Send a temporary message to indicate that processing is happening
   processing message = await message.reply text(" **Analyzing image,
please wait...**")
   try:
       # Download the image file as in-memory data
        img data = await client.download media(message.photo,
in memory=True)
        # Open the image using PIL (Pillow)
        img = PIL.Image.open(io.BytesIO(img data.getbuffer()))
        # If the user provides a caption, use it as a prompt; otherwise,
use a default prompt
       prompt = message.caption or "Describe this image."
        # Generate a response from the AI model using both the prompt and
the image
       response = model.generate content([prompt, img])
        # Extract the AI-generated text response
        response text = response.text
        # Store metadata about the analyzed image in MongoDB
        file metadata collection.insert one({
            "chat id": message.from user.id,  # Store user ID
            "file name": "photo.jpg", # Use a default name as Telegram
doesn't provide file names for photos
            "file type": "image", # Indicate that it's an image file
            "description": response_text,  # Store the generated
description of the image
            "timestamp": datetime.utcnow()  # Store the timestamp (UTC
format)
       })
       # Send the AI-generated description of the image to the user
```

```
await message.reply text(f" **Image
Analysis:**\n{response text}")
   except Exception as e:
        # Handle errors gracefully and notify the user
       await message.reply text(f" X Error analyzing image: {str(e)}")
   finally:
       # Delete the loading message after processing is complete
       await processing message.delete()
# Handler for the /file command, which analyzes a document
file sent by the user
@app.on_message(filters.command("file") & filters.document)
@append help # Adds help text after executing the command
async def analyze file(client: Client, message: Message):
    # Send a temporary message to indicate that processing is happening
   processing message = await message.reply text(" **Analyzing file,
please wait...**")
   try:
       file = message.document # Get file metadata from the message
       file name = file.file name # Extract file name
       file type = file.mime type # Extract file type (MIME type)
        # If the user provides a caption, use it as a prompt; otherwise,
use a default prompt
       prompt = message.caption if message.caption else "Summarize the
contents of this file."
        # Download the file as in-memory data
        file data = await client.download media(file, in memory=True)
        # Initialize extracted text as an empty string
        extracted text = ""
        # Check the file type and extract text accordingly
        if file type == "application/pdf": # PDF Files
           pdf reader =
PyPDF2.PdfReader(io.BytesIO(file data.getbuffer()))
           extracted text = " ".join([page.extract text() for page in
pdf reader.pages if page.extract text()])
```

```
elif file type == "text/plain": # TXT Files
            extracted text = file data.getvalue().decode("utf-8")
        elif file type in ["application/vnd.openxmlformats-
officedocument.wordprocessingml.document", "application/msword"]: # DOCX
Files
            doc = docx.Document(io.BytesIO(file data.getbuffer()))
            extracted_text = "\n".join([para.text for para in
doc.paragraphs])
        else:
            # If the file format is unsupported, return an error message
            extracted text = " \( \) Unsupported file format. Only PDF, TXT,
and DOCX are supported."
        # Ensure that extracted text is not empty
        if not extracted text.strip():
            extracted text = " No readable text found in this file."
        # Generate AI response using the extracted text (limited to 4000
characters to avoid excessive length)
        response =
model.generate content(f"{prompt}\n\n{extracted text[:4000]}")
        response text = response.text
        # Store metadata about the analyzed file in MongoDB
        file metadata collection.insert one({
            "chat id": message.from user.id,  # Store user ID
            "file name": file name, # Store the file name
            "file type": file type, # Store the file type
            "description": response text, # Store the AI-generated
summary or analysis
            "timestamp": datetime.utcnow()  # Store the timestamp (UTC
format)
        })
        # Send the AI-generated summary of the file to the user (limited
to 4000 characters)
        await message.reply text(f" **File
Analysis:**\n{response text[:4000]}")
   except Exception as e:
        # Handle errors gracefully and notify the user
       await message.reply text(f" X Error analyzing file: {str(e)}")
```

```
finally:
    # Delete the loading message after processing is complete
    await processing_message.delete()
```

# # Handler for the /sentiment command, which analyzes the sentiment of a given text

```
@app.on message(filters.command("sentiment"))
@append help # Adds help text after executing the command
async def sentiment handler(client: Client, message: Message):
    # Check if the user provided text after the command
   if len(message.text.split()) < 2: # Ensures the command is followed
by text input
       await message.reply text(" **Usage:** /sentiment <text>") #
Prompt user to provide input
       return # Exit function if no text is provided
    # Extract the text input from the message, excluding the command
itself
   text = message.text.split(maxsplit=1)[1]
    # Perform sentiment analysis using TextBlob
   blob = TextBlob(text) # Create a TextBlob object with the given text
   sentiment = blob.sentiment.polarity # Extract polarity score (-1 to
1)
    # Determine sentiment category based on polarity score
   if sentiment > 0:
        sentiment result = "8 Positive" # Positive sentiment if
polarity is greater than 0
   elif sentiment < 0:</pre>
        sentiment result = " Negative" # Negative sentiment if
polarity is less than 0
   else:
        sentiment result = " Heutral" # Neutral sentiment if polarity
is exactly 0
    # Store sentiment analysis result in MongoDB for future reference
    sentiment history.insert one({
        "chat id": message.from user.id,  # Store user ID
        "user query": text, # Store the analyzed text input
        "sentiment": sentiment result, # Store sentiment category
(Positive/Negative/Neutral)
```

```
"timestamp": datetime.utcnow() # Store timestamp in UTC format
})

# Reply to the user with the sentiment analysis result
await message.reply_text(f" **Sentiment
Analysis:**\n{text}\n\nResult: {sentiment_result}")
```

# # Handler for the /websearch command, which performs a web search using the SERP API and generates a summary.

```
@app.on message(filters.command("websearch"))
@append help # Adds help text after executing the command
async def web search(client: Client, message: Message):
    # Check if the user provided a query after the command
   if len(message.text.split()) < 2: # Ensures the command is followed</pre>
by a query
        await message.reply text("Q **Usage:** /websearch
your query") # Prompt user to provide a query
        return # Exit function if no query is provided
   query = message.text.split(maxsplit=1)[1]
     # Extract the query from the message
   loading msg = await message.reply text(" Searching the web...")
      # Send a loading message while searching
    # Perform sentiment analysis on the search query using the
get sentiment function
    sentiment = get sentiment(query)
    try:
        # Initialize the SERP API client with the provided query, API key
        search = GoogleSearch({"q": query, "api_key": SERP_API_KEY})
        # Perform the search and retrieve top 3 organic results
        results = search.get dict().get("organic results", [])[:3]
     # Get top 3 search results
        # Check if results are found
        if not results:
            await message.reply text("X No results found.")
     # Inform the user if no results were found
            return # Exit the function if no results are found
```

```
# Construct the response text with titles and links for the top 3
results
        response text = "\n".join([f"♦)
[{r['title']}]({r['link']})\n{r['snippet']}" for r in results])
        # Generate a summary of the search results using Gemini AI model
        summary prompt = "Summarize the following web search results:\n" +
response text # Prepare prompt for AI summary
        ai summary = model.generate content([summary prompt]) # Generate
content using the AI model
       summary text = ai summary.text.strip() # Extract the generated
summary text
        # Construct the final response with search results and AI-
generated summary
        response = f" **Top Search
Results:**\n\n{response text}\n\n**Gemini AI Summary:**\n{summary text}"
        # Store the query, bot response, and sentiment in MongoDB for
future reference
        websearch history collection.insert one({
            "chat id": message.from user.id,  # Store user ID
            "user query": query, # Store the search query
            "bot response": response, # Store the bot's response (search
results and summary)
            "sentiment": sentiment, # Store sentiment of the query
            "timestamp": datetime.utcnow()  # Store the timestamp of the
query
        })
        # Customize the response based on sentiment analysis (positive,
negative, or neutral)
       sentiment responses = {
            "positive": " Great! Here are some useful links:", #
Positive sentiment response
           "negative": " It seems like you have concerns. These
results might help:",  # Negative sentiment response
           "neutral": "\mathbb{Q} Here are the search results you requested:" #
Neutral sentiment response
        }
        # Final response combining sentiment and search results
        response = f"{sentiment responses.get(sentiment,
'')}\n\n{response text}\n\n**Gemini AI Summary:**\n{summary text}"
```

```
# Send the final response to the user with search results and
summary
    await message.reply_text(response)

except Exception as e:
    # Handle any exceptions (e.g., API issues, connection errors)
    await message.reply_text(f"    Error: {str(e)}") # Inform the
user about the error

finally:
    # Delete the loading message after the processing is complete
    await loading_msg.delete()

# Run the bot if this script is executed directly

if __name__ == "__main__":
    logger.info("Starting the bot...") # Log message to indicate the bot
is starting
    app.run() # Start the Pyrogram bot
```

**Documenting How ChatGPT Was Used to Develop the Project** 

## 1. Idea Generation and Brainstorming

- ChatGPT was used to generate and refine ideas for various features and components of the Telegram bot. As the bot was designed to perform multiple tasks (such as text generation, sentiment analysis, image description, and file summarization), ChatGPT helped provide detailed concepts for each feature.
- **Feature Planning:** ChatGPT helped define and refine the bot's capabilities based on user input, providing insights on how to organize the command structure and what AI functionalities to prioritize.
- **Bot Command Design:** The chatbot commands (like /start, /text, /img, /file, /sentiment, /websearch) were brainstormed with the help of ChatGPT, which also suggested ways to structure the commands for better user experience.

### 2. Pseudocode and Algorithm Design

- Whenever complex features needed to be implemented, ChatGPT assisted in developing pseudocode and high-level algorithms.
- **Clear Structure:** ChatGPT provided clear, logical steps for implementing algorithms, which helped in breaking down the tasks into manageable chunks.
- Efficient Code Implementation: By suggesting how to structure loops, conditionals, and function calls, ChatGPT made it easier to implement the necessary code efficiently, reducing development time.

#### 3. Code Snippets and Examples

- ChatGPT was actively consulted for specific code snippets, such as how to implement text generation using the Gemini model, extract text from a PDF using PyPDF2, or perform sentiment analysis using TextBlob.
- **Code Writing:** ChatGPT suggested optimal ways to write functions and integrate third-party libraries (like PIL, TextBlob, and PyPDF2).
- **Debugging Help:** Whenever there was a bug or issue with a specific function, ChatGPT helped by offering debugging strategies and code fixes, which accelerated the process of making the project functional.

# 4. Documentation and Commenting

- ChatGPT was used to generate clear and concise documentation for the codebase, helping to explain the purpose of different functions, commands, and integrations.
- It provided documentation on how each feature works, such as the interaction flow for each bot command, the AI model's functionality, or the sentiment analysis pipeline.
- **Automated Documentation:** ChatGPT produced helpful comments and docstrings for functions, reducing the manual effort needed for documenting the code.

• **Consistency in Documentation:** By using a consistent format for commenting the code, ChatGPT helped make the project more maintainable and readable.

# 6. Testing and Edge Case Handling

- When it came to testing the features of the bot, ChatGPT helped identify possible edge cases and provided testing strategies for different functionalities (e.g., testing the sentiment analysis feature with various types of input, testing file upload capabilities with different formats, or handling long text inputs in queries).
- Additionally, ChatGPT helped suggest test cases for validating the AI model's responses and ensuring that the web search feature returned accurate and relevant results.
- Edge Case Identification: ChatGPT assisted in thinking through potential edge cases that could cause issues, such as malformed input, file types that could break the bot, or extreme edge cases in sentiment analysis.
- **Testing Framework:** ChatGPT provided guidance on setting up a basic testing framework for validating different functionalities, which ensured that the bot remained robust during development.

# **Summary of ChatGPT's Contributions:**

- 1. **Idea Generation & Brainstorming:** Assisted in generating the core ideas and features for the project, ensuring the bot was user-friendly and comprehensive.
- 2. **Pseudocode & Algorithm Design:** Helped with designing high-level algorithms, turning conceptual ideas into manageable steps for implementation.
- 3. **Code Snippets & Examples:** Provided practical code snippets for various integrations and features, speeding up development.
- 4. **Documentation & Commenting:** Streamlined the documentation process by generating useful comments and explanations for the codebase.
- 5. **Testing & Edge Case Handling:** Contributed to identifying potential edge cases and providing testing strategies to ensure the bot's robustness.
- 6. **Deployment Strategy & Best Practices:** Assisted in preparing the project for deployment and ensuring it followed industry best practices.

Overall, ChatGPT was an essential tool in accelerating development, streamlining workflows, and ensuring the project was both functional and user-friendly.

# **\* MAJOR HIGHLIGHTS**

# 1. Google Generative AI (Gemini Model)

#### **Purpose:**

• **Content Generation and Summarization:** The Gemini AI model is used throughout the project to generate content or summaries based on user queries, uploaded images, and files. It adds an intelligent, conversational element to the bot, allowing users to receive meaningful responses, summaries, and descriptions.

# **Use Cases in the Project:**

#### • Text Generation for Queries:

When users input a prompt (via /text), the Gemini AI model generates a relevant response, creating meaningful content from the provided input. For example, if a user asks a question or provides a prompt, Gemini generates a detailed response based on the query.

# • Image Analysis and Descriptions:

o The model generates descriptions for images shared by users. For example, in the /img command, when a user sends a photo, the AI analyzes the image and generates a response, describing what is visible in the image.

# • File Content Summarization:

o When users upload documents (via /file), the AI model is used to generate summaries of the document content. Whether it's a PDF, DOCX, or text file, Gemini processes the document and creates a concise summary for the user.

#### • Web Search Result Summarization:

o After performing a web search (via /websearch), the model generates a summary of the top search results. The AI provides a brief, understandable summary of web pages returned by the Google Search API.

#### **How It Contributes:**

- **Enhances User Experience:** The Gemini model's ability to generate human-like responses makes interactions with the bot more engaging and informative.
- **AI-Driven Analysis:** By incorporating generative AI, the bot can provide personalized, context-aware responses, even summarizing complex information like web search results and document contents.
- **Time Efficiency:** The use of AI ensures that users receive quick, accurate answers and summaries without having to manually read through long documents or search results.

# 2. TextBlob (Sentiment Analysis)

#### **Purpose:**

• **Sentiment Analysis:** TextBlob is used to perform sentiment analysis on user-provided text (queries or messages). It helps determine whether the sentiment of the input is positive, negative, or neutral.

# **Use Cases in the Project:**

#### • Sentiment Detection for Text Queries:

o In commands like /sentiment, TextBlob analyzes the sentiment of a given text. For example, if a user provides a statement or a piece of text, the bot analyzes it to determine whether the sentiment is positive, negative, or neutral.

## • Sentiment Analysis for Web Search Queries:

o When users perform a web search, TextBlob analyzes the sentiment of the search query. This sentiment analysis helps to customize the response by adding a personalized touch, such as providing a positive tone for a happy query or offering helpful links for a negative sentiment query.

#### • Sentiment Analysis of AI-Generated Content:

The bot also analyzes the sentiment of generated content to adjust the tone of responses. For example, it can adapt its responses to be more empathetic when dealing with negative sentiments, or offer a more enthusiastic tone for positive queries.

#### **How It Contributes:**

- **Personalized Interactions:** Sentiment analysis allows the bot to provide responses that align with the user's emotional state, making interactions feel more human and personalized.
- **Improved User Engagement:** By adjusting the tone of responses based on sentiment, the bot can ensure that the user experience is more appropriate, empathetic, and effective.
- **Better Content Moderation:** Sentiment analysis can help the bot filter or adjust content that may not be appropriate or may need a more cautious response.

# 3. SERP API (Google Search API)

#### **Purpose:**

• **Web Search Integration:** The SERP API is used to perform real-time searches on the web based on the user's query. It retrieves organic search results, providing the bot with links, titles, and snippets from the top web pages.

# **Use Cases in the Project:**

#### • Searching for Information:

o When a user sends a query using the /websearch command, the bot uses the SERP API to fetch relevant search results. The search results are then presented to the user, along with an AI-generated summary of the findings.

# **How It Contributes:**

- **Real-Time Data Retrieval:** The SERP API provides the bot with up-to-date web data, which makes the bot capable of responding to user queries with the most recent and relevant information available.
- Enhances the Scope of the Bot: By integrating web search functionality, the bot can provide answers to a wider range of questions, especially those that go beyond the bot's in-built knowledge.

# 4. PyPDF2 (PDF Text Extraction)

## **Purpose:**

• **Text Extraction from PDF Files:** PyPDF2 is used to extract text from PDF documents that are uploaded by users.

## **Use Cases in the Project:**

- PDF Document Processing:
  - When users upload PDF files through the /file command, PyPDF2 reads and extracts text from the file to allow the bot to summarize the content or perform further analysis.

#### **How It Contributes:**

- **File Handling Capabilities:** The bot can handle various file types, especially PDFs, which are common for documents. This allows users to interact with the bot without worrying about file formats, ensuring ease of use.
- **Content Extraction:** By extracting text from PDFs, the bot can analyze and summarize documents, providing valuable insights without requiring users to manually read lengthy content.

# 5. Pillow (Image Processing)

### **Purpose:**

• **Image Analysis and Handling:** Pillow (PIL) is used to process and manipulate images sent by users. The library helps in opening, resizing, and handling image files.

## **Use Cases in the Project:**

- Image Analysis:
  - The bot uses Pillow to handle images that users send in the /img command. The images are processed and analyzed to generate descriptions, summaries, or insights.

#### **How It Contributes:**

• **Image Recognition and Description:** By integrating image processing, the bot can provide detailed descriptions of images, making it more versatile in handling diverse input from users (e.g., text and images).

# **Summary of AI Tool Contributions to the Project:**

- **Generative AI (Gemini Model)** enhances the bot's ability to understand and generate human-like responses, summaries, and content across a variety of inputs (text, images, files, web searches).
- **TextBlob** ensures the bot can analyze the sentiment of both user queries and generated content, allowing it to respond appropriately to the user's emotional tone.
- **SERP API** empowers the bot to perform real-time web searches and deliver current, relevant information, enhancing its knowledge base.
- **PyPDF2** provides the ability to extract and analyze text from PDF files, allowing the bot to summarize and understand documents.
- **Pillow (PIL)** enables the bot to process and describe images, expanding its capabilities beyond just text interactions.

By combining these AI tools, the project leverages a powerful and flexible approach to delivering interactive, intelligent, and personalized user experiences.