# HANDS-ON MACHINE LEARNING (HOME)

Lecture/Lab – 3

An Introduction to Machine Learning Using LLMs as an Example – Part III

Goal: Extract Discrete Variables from TCGA Pathology Reports Using Local LLMs

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#### Overview

#### Running LLMs on Local Machines Using Ollama

- Python
  - Miniconda
  - VSCode
- Prompting and Retrieval Augmented Generation (RAG)
  - LangChain

- Graphical User Interface
  - Open WebUI

#### ollama

#### Windows

https://ollama.com/

- Download and install
- Running LLMs, locally
- http://localhost:11434

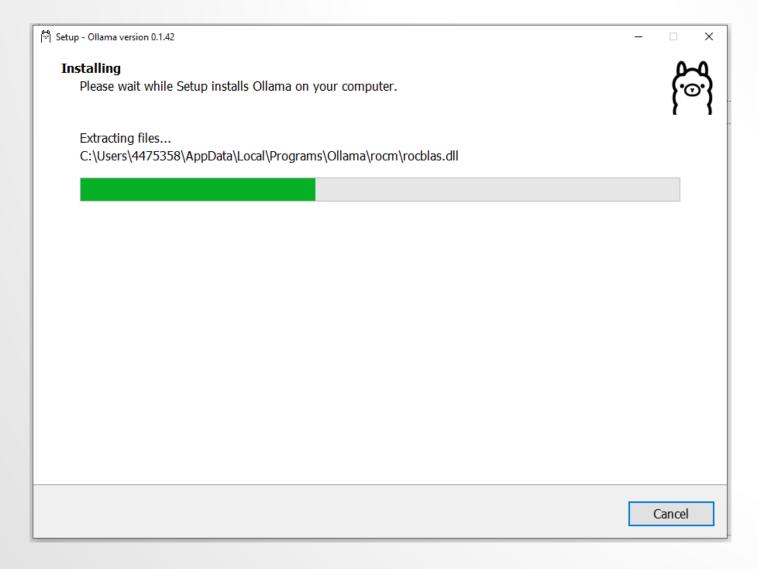
#### Linux

Download ollama binary

curl -L https://ollama.com/download/ollama-linux-amd64 -o ollama ollama server&

http://localhost:11434

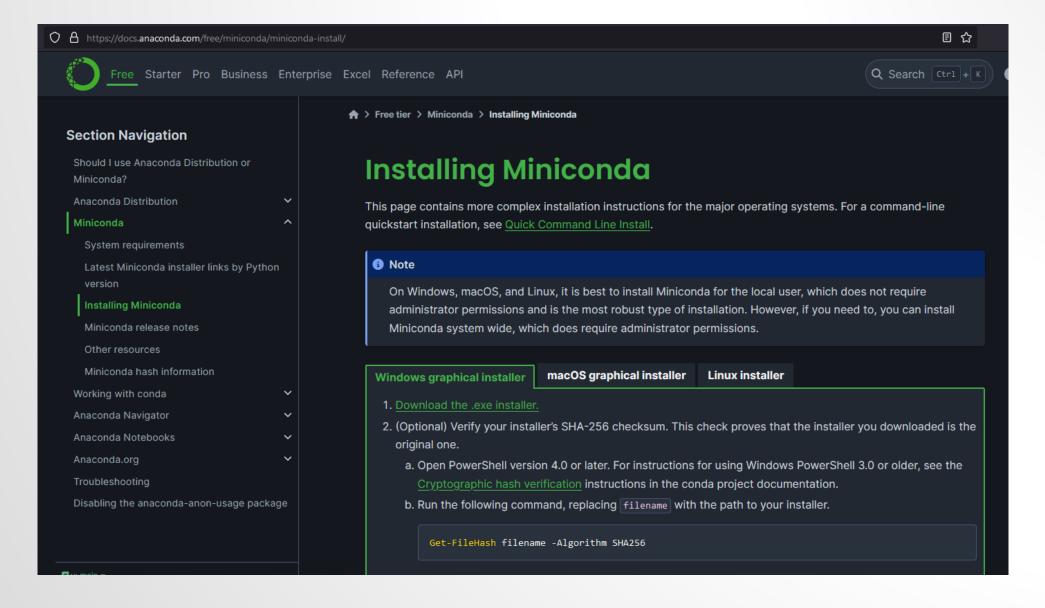
#### Download and install ollama



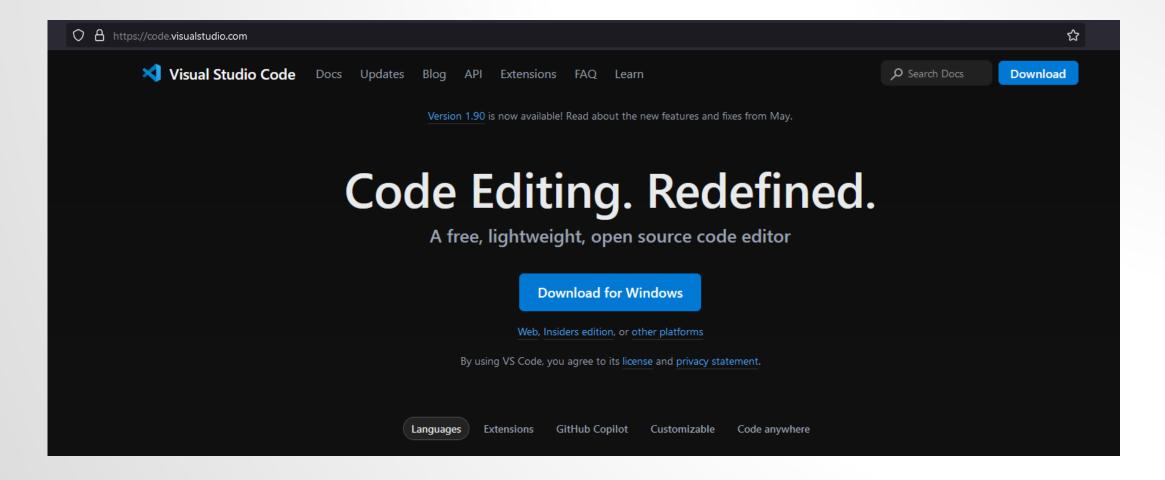
## Graphical User Interface - Open Webui

https://github.com/open-webui/open-webui

#### **Download and install Miniconda**



#### Install VSCode



#### Create a virtual environment and activate it

- Open VSCode
- Open a Terminal
- Create a virtual environment, activate it

conda create --name llm-path-reports python=3.10
conda activate llm-path-reports

## Clone GitHub Repo and Install requirments

(Ilm-path-reports) > git clone <a href="https://github.com/grasool/HOME">https://github.com/grasool/HOME</a>

(Ilm-path-reports) > pip install -r.\requirements.txt

#### Run Ollama

- · ollama
- Ollama list
- Ollama pull llama 3

```
C:\Users\4475358>ollama
 ollama [flags]
 ollama [command]
vailable Commands:
  serve
             Start ollama
             Create a model from a Modelfile
  create
             Show information for a model
             Run a model
             Pull a model from a registry
             Push a model to a registry
  list
             List models
             List running models
             Copy a model
  ср
             Remove a model
             Help about any command
 help
Flags:
 -h, --help
                 help for ollama
  -v, --version Show version information
Use "ollama [command] --help" for more information about a command.
C:\Users\4475358>
```

```
C:\Users\4475358>ollama list

NAME ID SIZE MODIFIED

llama3:latest 365c0bd3c000 4.7 GB About a minute ago

llama2:latest 78e26419b446 3.8 GB 3 months ago
```

```
C:\Users\4475358>ollama pull llama3
pulling manifest
pulling 6a0746a1ec1a... 100% 🗈

    4.7 GB

pulling 4fa551d4f938... 100% [
                                                                                                                                                 12 KB
pulling 8ab4849b038c... 100% E
                                                                                                                                                 254 B
pulling 577073ffcc6c... 100% [
                                                                                                                                                 110 B
pulling 3f8eb4da87fa... 100% 🗈
                                                                                                                                                 485 B
verifying sha256 digest
writing manifest
removing any unused layers
success
```

```
from langchain community.llms import Ollama
from langchain.prompts import ChatPromptTemplate
# Setup Model
llm_model = "llama3"
chat = Ollama(model=llm_model, temperature=0.0)
# Setup prompt template
template_string = """You are a helpful assistant. \
        Answer this question: {question}. """
prompt_template = ChatPromptTemplate.from_template(template_string)
# Build prompt
my_question = "Where is Moffitt Cancer Center?"
llm input prompt string = prompt template.format messages(question=my question)
print(llm input prompt string)
# Run model
llm answer = chat.invoke(llm input prompt string)
# Print answer
print(llm answer)
```

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# Print answer
print(llm answer)
```

```
(llm-path-reports) PS C:\Works\HOME> & C:/Users/4475358/.conda/envs/llm-path-reports/python.exe
c:/Works/HOME/Lecture-3/test-langchain-ollama.py
[HumanMessage(content='You are a helptul assistant.
                                                            Answer this question: Where is Moffi
tt Cancer Center?. ')]
The H. Lee Moffitt Cancer Center is located in Tampa, Florida. It's a comprehensive cancer center
r that provides advanced treatment options and innovative care to patients with various types of
 cancer.
To be more specific, the Moffitt Cancer Center is situated at:
12902 USF Pine Dr.
Tampa, FL 33612
If you're planning to visit or seek medical attention, I recommend checking their website for di
rections, parking information, and any updates on their services. Would you like me to provide m
ore details about Moffitt's facilities, treatment options, or anything else?
Execution time: 50.05071258544922 seconds
```

(11m-path-reports) PS C:\Works\HOME> & C:/Users/4475358/.conda/envs/11m-path-reports/python.exe c:/Works/HOME/Lecture-3/test-langchain-ollama.pv Answer this question: Where is Moffi [HumanMessage(content='You are a helpful assistant. tt Cancer Center?. ')] The H. Lee Moffitt Cancer Center is located in Tampa, Florida. It's a comprehensive cancer cente r that provides advanced treatment options and innovative care to patients with various types of cancer. To be more specific, the Moffitt Cancer Center is situated at: 12902 USF Pine Dr. Tampa, FL 33612 If you're planning to visit or seek medical attention, I recommend checking their website for di rections, parking information, and any updates on their services. Would you like me to provide m ore details about Moffitt's facilities, treatment options, or anything else? Execution time: 50.05071258544922 seconds

```
(11m-path-reports) PS C:\Works\HOME> & C:/Users/4475358/.conda/envs/11m-path-reports/python.exe
 c:/Works/HOME/Lecture-3/test-langchain-ollama.py
 [HumanMessage(content='You are a helpful assistant.
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## Path Report



COLLEGE CHALLES

No clinical history is given.

PRE-OP DIAGNOSIS: Right kidney mass.

POST-OP DIAGNOSIS: Same.

EDURE: Right nephrectomy.



#### FINAL DIAGNOSIS:

KIDNEY, RIGHT, NEPHRECTOMY -

- A. RENAL CELL CARCINOMA, CONVENTIONAL (CLEAR) CELL TYPE
- B. FUHRMAN'S NUCLEAR GRADE IS 2 OF 4
- C. THE GREATEST DIAMETER OF THE NEOPLASM IS 4 CM.
- D. THE NEOPLASM IS CONFINED WITHIN THE RENAL CAPSULE
- E. NO INVASION OF THE RENAL VEIN IS IDENTIFIED
- F. NO EVIDENCE OF ANGIOLYMPHATIC INVASION IS IDENTIFIED.
- G. ALL SURGICAL MARGINS ARE FREE OF THE NEOPLASM
- H. THE NON-NEOPLASTIC KIDNEY IS UNREMARKABLE
- THE ADRENAL GLAND IS UNREMARKABLE
- J. TNM STAGE: pT1a Nx MX.
- K. TNM HISTOLOGIC GRADE = G2.

#### COMMENT:

A panel of immunohistochemical stains has been performed, which showed the tumor cells to be strongly positive for CD10, and weakly positive for RCC and E-cadherin. Tumor cells are focally positive for cytokeratin 7 and negative for Ber-EP4. A colloidal iron stain was also negative. This staining pattern supports a diagnosis of conventional type renal cell carcinoma.

#### CASE SYNOPSIS:

SYNOPTIC DATA - PRIMARY KIDNEY TUMORS

SPECIMEN TYPE:

Radical nephrectomy

LATERALITY:

Right

TUMOR SITE: FOCALITY:

Middle Unifocal

TUMOR SIZE:

Greatest dimension: 4.0 cm Additional dimensions: 3.5 cm

Tumor limited to kidney

MACROSCOPIC EXTENT OF TUMOR:

Clear cell (conventional) renal carcinoma

HISTOLOGIC TYPE:

HISTOLOGIC GRADE (Fuhrman Nuclear Grade):

PATHOLOGIC STAGING (pTNM):

pT1a

DΝΧ

Number of regional lymph nodes examined: 0

MARGINS:

Margins uninvolved by invasive carcinoma

ADRENAL GLAND: Uninvolved by tumor

Absent VENOUS (LARGE VESSEL) INVASION (V): LYMPHATIC (SMALL VESSEL) INVASION (L): Absent

ADDITIONAL PATHOLOGIC FINDINGS:

Glomerular disease (type): Minimal global glomerusclerosis

# Stage – 0 (OCR)



No clinical history is given. PRE-OP DIAGNOSIS: Right kidney mass.

POST-OP DIAGNOSIS: Same CEDURE: Right nephrectomy.



#### FINAL DIAGNOSIS:

KIDNEY, RIGHT, NEPHRECTOMY -

- RENAL CELL CARCINOMA, CONVENTIONAL (CLEAR) CELL TYPE
- B. FUHRMAN' S NUCLEAR GRADE IS 2 OF 4
- THE GREATEST DIAMETER OF THE NEOPLASM IS 4 CM.
- THE NEOPLASM IS CONFINED WITHIN THE RENAL CAPSULE
- NO INVASION OF THE RENAL VEIN IS IDENTIFIED
- NO EVIDENCE OF ANGIOLYMPHATIC INVASION IS IDENTIFIED.
- ALL SURGICAL MARGINS ARE FREE OF THE NEOPLASM
- THE NON-NEOPLASTIC KIDNEY IS UNREMARKABLE
- THE ADRENAL GLAND IS UNREMARKABLE
- TNM STAGE: pT1a Nx MX.
- K. TNM HISTOLOGIC GRADE = G2.

#### COMMENT:

A panel of immunohistochemical stains has been performed, which showed the tumor cells to be strongly positive for CD10, and weakly positive for RCC and E-cadherin. Tumor cells are focally positive for cytokeratin 7 and negative for Ber-EP4. A colloidal iron stain was also negative. This staining pattern supports a diagnosis of conventional type renal cell carcinoma.

#### CASE SYNOPSIS:

SYNOPTIC DATA - PRIMARY KIDNEY TUMORS

SPECIMEN TYPE: Radical nephrectomy

LATERALITY: Right TUMOR SITE: Middle Unifocal FOCALITY:

Greatest dimension: 4.0 cm TUMOR SIZE:

Additional dimensions: 3.5 cm MACROSCOPIC EXTENT OF TUMOR: Tumor limited to kidney

Clear cell (conventional) renal carcinoma HISTOLOGIC TYPE:

HISTOLOGIC GRADE (Fuhrman Nuclear Grade):

PATHOLOGIC STAGING (pTNM): pT1a

Number of regional lymph nodes examined: 0

Margins uninvolved by invasive carcinoma MARGINS:

Uninvolved by tumor ADRENAL GLAND: VENOUS (LARGE VESSEL) INVASION (V): Absent

LYMPHATIC (SMALL VESSEL) INVASION (L): Absent

Glomerular disease (type): Minimal global glomerusclerosis ADDITIONAL PATHOLOGIC FINDINGS:

Input report after OCR: ...., .. .".1 .. """. 1"1. No clinical history is given. PRE-OP DIAGNOSIS: Right kidney mass. POST-OP DIAGNOSIS: Same. ~EDURE: Right nephrectomy. FINAL DIAGNOSIS: KIDNEY, RIGHT, NEPHRECTOMY-A. RENAL CELL CARCINOMA, CONVENTIONAL (CLEAR) CELL TYPE FUHRMAN'S NUCLEAR GRADE IS 2 OF 4 C. THE GREATEST DIAMETER OF THE NEOPLASM IS 4 CM. THE NEOPLASM IS CONFINED WITHIN THE RENAL CAPSULE NO THVASTON OF THE RENAL VETN IS TDENTIFIED NO EVIDENCE OF ANGIOLYMPHATIC INVASION IS IDENTIFIED. ALL SURGICAL MARGINS ARE FREE OF THE NEOPLASM THE NON-NEOPLASTIC KIDNEY IS UNREMARKABLE I. THE ADRENAL GLAND IS UNREMARKABLE J. TNM STAGE: pTle Nx MX. HISTOLOGIC. GRADE = G2. A panel of immunohistochemical stains has been perfonned, which showed the tumor cells to be strongly positive for COlO, and weakly positive for RCC and E-cadherln. Tumor cells are focally positive for cytokeratin 7 and negative for Ber EP4. A colloidal iron stain was also negative. This staining pattern supports a diagnosis of conventional type renal cell carcinoma. CASE SYNOPSIS: SYNOPTIC DATA -PRIMARY KIDNEY TUMORS SPECIMEN TYPE: Radical nephrectomy LATERALITY: Right TUMOR SITE: Middle FOCALITY: Unifocal TUMOR SIZE: Greatest dimension: 4.0 cm Additional dimensions: 3.5 cm MACROSCOPIC EXTENT OF TUMOR: Tumor limited to kidney HISTOLOGIC TYPE: Clear cell (conventional) renal carcinoma HISTOLOGIC GRADE (Fuhrman Nuclear Grade): G2 PATHOLOGIC STAGING (pTNM): pT1a Number of regional lymph nodes examined: 0 MARGINS: Margins uninvolved by invasive carcinoma ADRENAL GLAND: Uninvolved by tumor

ADDITIONAL PATHOLOGIC FINDINGS: Glomerular disease (type): Minimal global glomerusclerosis

VENOUS (LARGE VESSEL) INVASION (V): Absent LYMPHATIC (SMALL VESSEL) INVASION (L): Absent

## Stage – I (Unstructured Data Extraction)

LYMPHATIC (SMALL VESSEL) INVASION (L): Absent

ADDITIONAL PATHOLOGIC FINDINGS: Glomerular disease (type): Minimal global glomerusclerosis

First Stage Processing - LLM Extracted Data and Justification: Here is the extracted information: 1. "Site": Kidney Justification: The report mentions a right kidney mass and nephrectomy, indicating that the tumor is located in the kidney. 2. "Laterality": Right No clinical history is given. Justification: The report explicitly states that the tumor is located on the right side (LATERALITY: Right). 3. "Histology": Renal cell carcinoma, conventional (clear) cell type Justification: The report mentions a diagnosis of renal cell carcinoma, conventional (clear) cell type, based on histological 4. "Stage": pTle Nx MX Justification: The report provides the TNM staging information as pTle Nx MX, indicating that the tumor is limited to the kid ney (pTle), there is no lymph node involvement (pNx), and there is no distant metastasis (pMX). 5. "Grade": G2 Justification: The report mentions a histological grade of 2 out of 4 based on Fuhrman's nuclear grading system. 6. "Behavior": Not provided A panel of immunohistochemical stains has been perfonned, which showed th Justification: There is no information in the report that explicitly states the behavior of the tumor, so this category remai COlO, and weakly positive for RCC and E-cadherln. Tumor cells are focally EP4. A colloidal iron stain was also negative. This staining pattern supp ns blank. carcinoma. Note that some categories may have missing information if it is not provided in the report. HISTOLOGIC TYPE: Clear cell (conventional) renal carcinoma HISTOLOGIC GRADE (Fuhrman Nuclear Grade): G2 PATHOLOGIC STAGING (pTNM): pT1a Number of regional lymph nodes examined: 0 MARGINS: Margins uninvolved by invasive carcinoma ADRENAL GLAND: Uninvolved by tumor VENOUS (LARGE VESSEL) INVASION (V): Absent

22

## Stage-II Processing (JSON Object)

```
Second Stage Processing - Discrete Variables:

{
    "site": "Kidney",
    "laterality": "Right",
    "histology": "Renal cell carcinoma, conventional (clear) cell type",
    "stage": "pTle Nx MX",
    "grade": "G2",
    "behavior": null
}
```

Laptop CPU execution time: < 9 minutes</li>

Execution time: 523.7728755474091 seconds

#### **Imports**

```
# This is the main script for processing pathology reports in PDF format using LLM.
     # The script uses the langchain community library to interact with the LLMs.
     import os
     import glob
     import random
     import time
     from langchain_community.llms import Ollama
     from langchain.prompts import ChatPromptTemplate
11
12
     from langchain_community.document_loaders import PyPDFLoader
     import os
     from langchain community.llms import Ollama
     from langchain.prompts import ChatPromptTemplate
     from langchain core.output parsers import JsonOutputParser
     from langchain.prompts import PromptTemplate
     from langchain core.pydantic v1 import BaseModel, Field
     import json
     import subprocess
```

## Stage-I

```
def process_pdf(pdf_file_to_open, llm_model):
   chat = Ollama(model=llm model, temperature=0.0)
   template_string = """You are a helpful assistant with knowlede in surgical pathology. \
      Your task is to process the given surgical pathology report and extract specific information and justify \
      the extracted information in one sentence.
      The reports are related to various cancers and have been converted into text using OCR from PDF files. \
      Therefore, ignore any OCR errors and focus on the content of the report. \
      For each report, fill the following categories "Site", "Laterality (left or right)", "Histology", "Stage (TNM format)", \
      "Grade (Grade I (Low grade or well-differentiated), \
      Grade II (Intermediate grade or moderately differentiated), Grade III (High grade or poorly differentiated),
      and Grade IV (High grade or undifferentiated))", "Behavior".
     An example output is given here: \
     1. "Site": brain. \
      2. "Laterality": left. \
     3. "Histology": adenocarcinoma, as the report mentioned the histology of the tumor. \
     4. "Stage": T2N0Mx, as the tumor invaded the muscularis propria and the lymph nodes were not affected based on the report. \
     5. "Grade": III, as the tumor showed moderate differentiation based on the report. \
     6. "Behavior": malignant, as the tumor showed invasion of the surrounding tissues based on the report. \
     Here is the report {report}.
      Restrict your output to the six categories only that include "Site", "Laterality", "Histology", "Stage", "Grade", \
      and "Behavior" and one sentence for the justification of the choice. \ You, 1 second ago • Uncommitted changes
      For the missing information, say "not provided".
   prompt_template = ChatPromptTemplate.from_template(template_string)
   loader = PyPDFLoader(pdf file to open)
   pages = loader.load()
   report = ' '.join(page.page content for page in pages)
   print("-----")
   print("-----")
   print("Input report after OCR:")
   print("-----")
   print("-----")
   print(report)
   print("-----")
   print("-----")
   llm input report = prompt template.format messages(report=report)
   extracted_data = chat.invoke(llm_input_report)
   return extracted data
```

### Stage-II

```
Ghulam Rasool, 2 hours ago | 1 author (Ghulam Rasool)

class path_variables(BaseModel):

site: str = Field(description="site of the cancer as described in the pathology report")

laterality: str = Field(description="laterality of the cancer as described in the pathology report")

histology: str = Field(description="histology of the cancer as described in the pathology report")

stage: str = Field(description="stage of the cancer as described in the pathology report")

grade: str = Field(description="grade of the cancer as described in the pathology report")

behavior: str = Field(description="behavior of the cancer as described in the pathology report")
```

```
def extract json output(extracted report data, model):
   query string = """
       DO NOT MAKE UP ANY INFORMATION. THIS IS A RETRIEVAL TASK ONLY.
       Structure the information presented in a pathology report into JSON format. \
       The missing information should be represented as null. \
       DO NOT MAKE UP ANY INFORMATION. Here is the report \
   parser = JsonOutputParser(pydantic object=path variables)
   prompt = PromptTemplate(
        template="Answer the user query. \n{format_instructions}\n{query}\n{report}",
        input_variables=["query", "report"],
       partial_variables={"format_instructions": parser.get_format_instructions()},
   model = Ollama(model=llm_model, temperature=0.0)
    chain = prompt | model | parser
    try:
        json_variables = chain.invoke({"query":query_string, "report": extracted_report_data})
   except Exception as e:
       print(f"An error occurred: {e}")
        json variables = []
   return json_variables
```

#### **Main Function**

```
if name == " main ":
  pdf file = r'C:\Works\HOME\Lecture-3\kidney.pdf'
  print('Processing:', pdf_file)
  llm model = "llama3"
  start time = time.time()
  # Step 1: Process the PDF file using the LLM
  extracted_data = process_pdf(pdf_file, llm_model)
  print("First Stage Processing - LLM Extracted Data and Justification:")
  print("-----")
  print("-----")
  print(extracted data)
  print("-----")
  print("-----")
  # Step: Extract the structured variables from the LLM output
  json_variables = extract_json_output(extracted_data, llm_model)
  print("Second Stage Processing - Discrete Variables:")
  print("-----")
  print("-----")
  print(json.dumps(json_variables, indent=4))
  end time = time.time()
  subprocess.Popen([pdf_file],shell=True)
  print(f"Execution time: {end_time - start_time} seconds")
```

## Questions