

Project Report: GeminiDecode: Multi Language Document Extraction by Gemini Pro

1. Introduction

1.1 Project Title: ***GeminiDecode: Multi Language Document Extraction by Gemini Pro***

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2. Project Overview

GeminiDecode: Multilanguage Document Extraction by Gemini Pro is a cutting-edge generative AI project designed to extract and process data from documents in multiple languages. Leveraging advanced **Natural Language Processing (NLP)** and **Deep Learning** models, this tool allows global businesses and organizations to efficiently extract, process, and categorize information from documents in over 50 different languages.

This solution caters to multiple industries, including the legal, financial, and healthcare sectors, ensuring efficiency, accuracy, and ease of document handling. The project automates the extraction of key information and supports various document formats, enabling streamlined workflows, improved productivity, and enhanced decision-making.

3. Technical Skills Required

- **Python:** For implementing the back-end logic and interfacing with APIs.
- **Deep Learning:** For model handling and document extraction.
- **Streamlit:** For building the user interface (UI) and deploying the application.

4. Project Scenarios and Use Cases

Scenario 1: Legal Sector In the legal sector, **GeminiDecode** can swiftly extract and organize multilingual legal documents, including contracts and affidavits. This helps law firms reduce

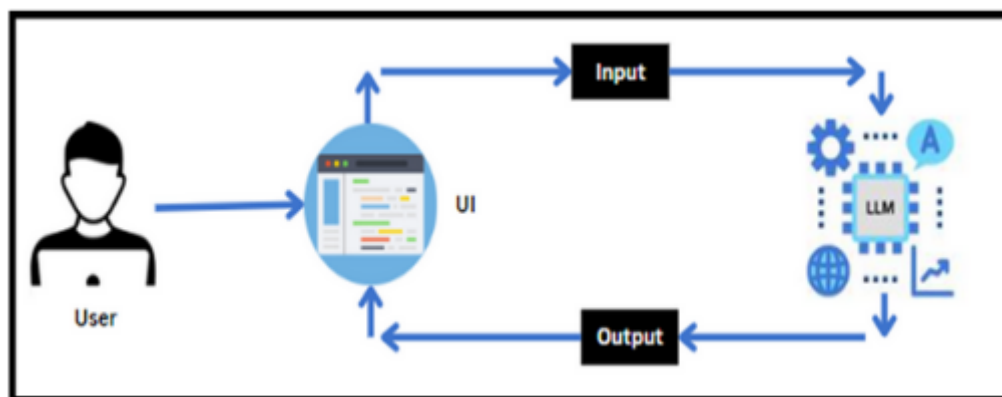
manual processing time, ensure compliance with various legal standards, and focus on client interaction and strategy.

Scenario 2: Financial Institutions Banks and financial institutions can use **GeminiDecode** to process loan applications and financial statements in multiple languages. Automating data extraction helps streamline loan approvals and ensure compliance with global financial regulations, improving customer service.

Scenario 3: Healthcare Industry Hospitals and clinics benefit from **GeminiDecode** by extracting and processing multilingual patient records, enabling healthcare providers to offer timely and accurate care. The system enhances the management of medical records and treatment plans, improving patient outcomes.

5. Technical Architecture

Below is the **technical architecture** used in this project:



The architecture consists of the following components:

- **User Interface (UI):** Users input data through a Streamlit-based front-end interface.
- **Input Processing:** The input is transmitted to the backend using a **Google API key**.
- **Gemini Pro Pre-trained Model:** Input is processed by Gemini Pro, which identifies, extracts, and categorizes information from documents.
- **Output:** The processed information is returned to the UI for presentation and analysis.

6. Project Flow

The project follows these key stages:

1. **User Input**
 - Users interact with the UI (Streamlit-based) to provide documents for extraction.
 - Input is transmitted to the backend via the **Google API key**.
 2. **Interfacing with Pre-trained Model**
 - The **Gemini Pro Pre-trained Model** processes the user-provided documents.
 - The model extracts and categorizes information using NLP and machine learning techniques.
 3. **Output**
 - The results of the document extraction are formatted and displayed on the front-end UI for review and further analysis.
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7. Implementation Steps

To execute the **GeminiDecode** project, follow these steps:

7.1 Requirements Specification

- **Create a `requirements.txt` file** to list the required libraries.
- Install all necessary dependencies, including:
 - `streamlit`
 - `transformers`
 - `requests`
 - Other libraries required for NLP and API interfacing.

7.2 Google API Key Initialization

- **Generate a Google API key** to access cloud-based models.
- Add the Google API key to a `.env` file for secure storage.

7.3 Interfacing with the Pre-trained Model

- Load the **Gemini Pro Pre-trained Model**.
- Implement functions to:
 - Handle user input (read PDF documents, etc.).

- Send input to the Gemini API for processing.
- Receive and display the processed output.

7.4 Deployment and Hosting

- Integrate the application with **Streamlit** for easy deployment.
- Deploy the project to a platform like **Heroku** or **Streamlit Cloud** for real-time document extraction and processing.

8. File Organization

Organize your project files into a well-structured directory for ease of management and deployment. Below is an example of the project folder structure:

bash

Copy code

GeminiDecode/

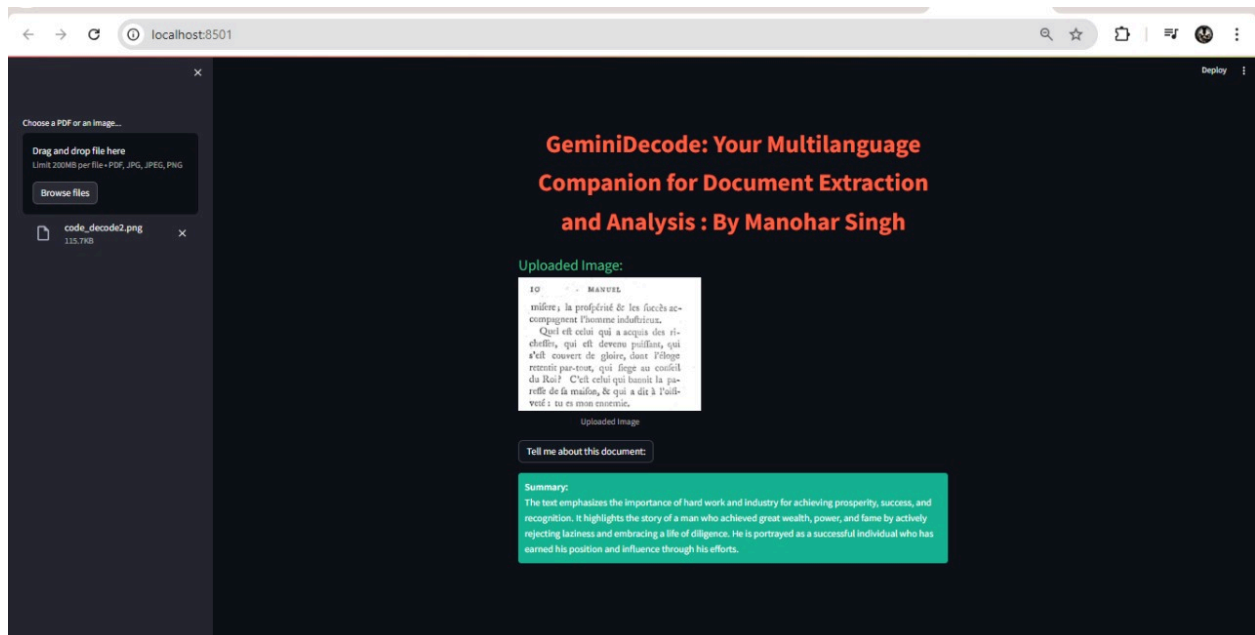
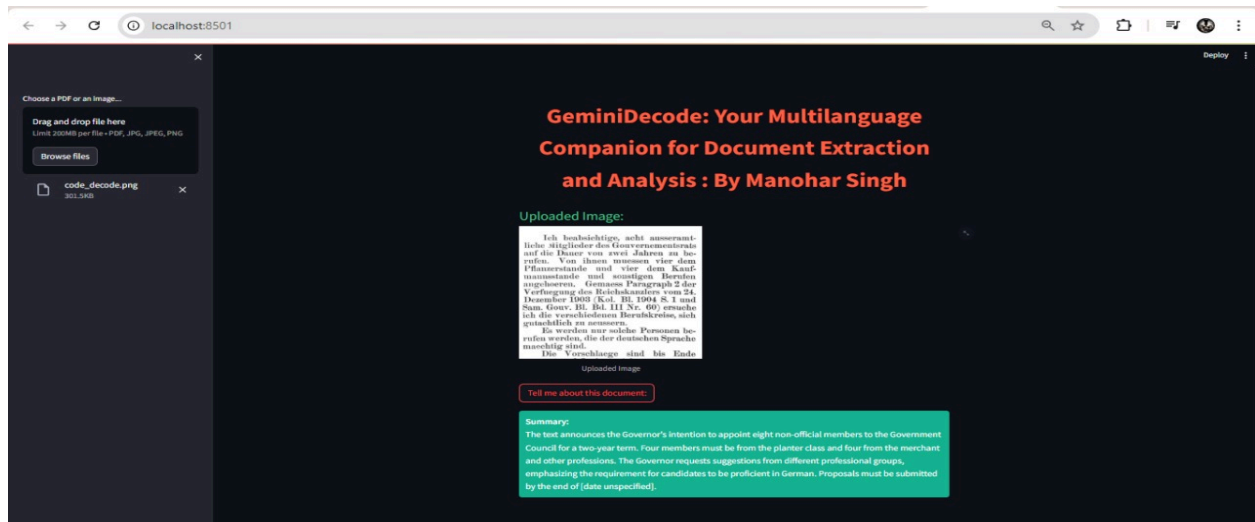
```
|
├── assets/          # Stores images used in the UI.
├── outputs_screenshots/ #screenshots of execution and User Interface
├── .env            # Contains Google API key for secure access.
├── app.py          # Main application file for Streamlit and model logic.
├── requirements.txt # Lists all necessary libraries and dependencies.
└── README.md       #Provides an overview of the project and instructions.
```

9. Code Execution Screenshots

The following screenshots showcase various stages of the project execution, including code snippets, model interactions, and the user interface (UI) on Streamlit. These screenshots serve as proof of concept and successful execution of the GeminiDecode project.

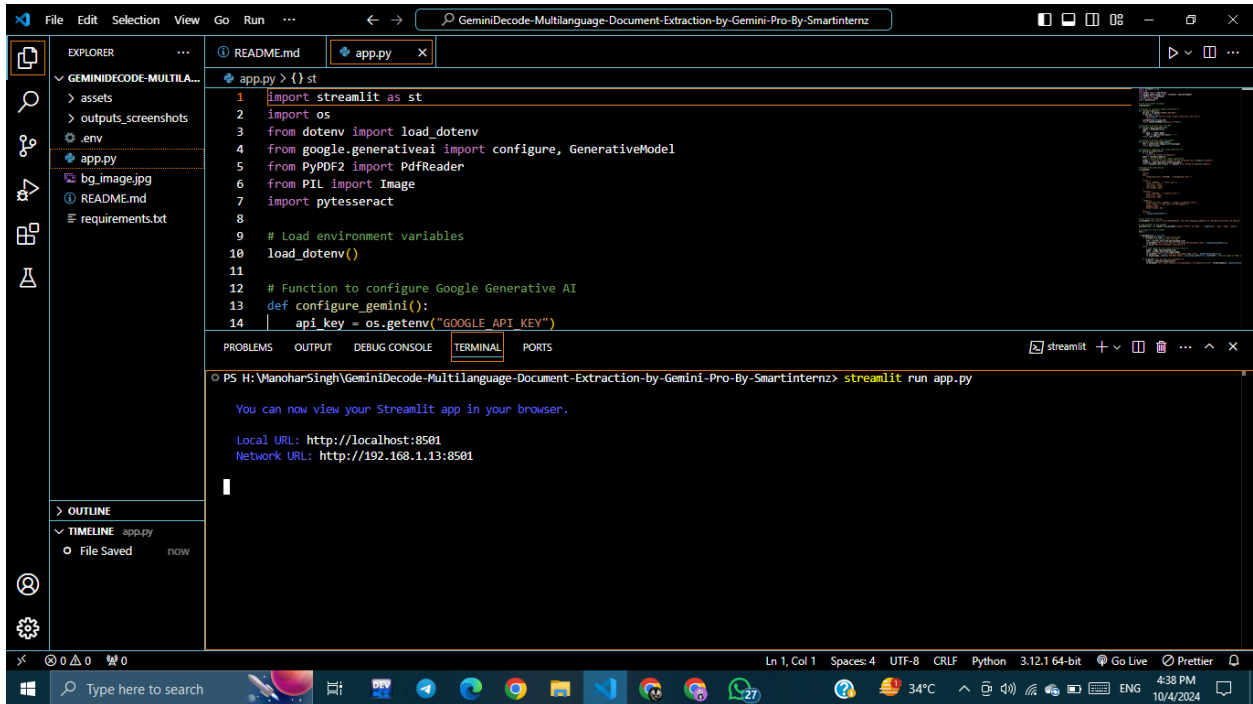
1. Streamlit UI Interface

A screenshot demonstrating the front-end interface for document input and output display.



2. Code Snippets

Screenshots of the key sections of the Python code, including loading the Gemini Pro Pre-trained model, input processing, and API interfacing.



The screenshot displays a Visual Studio Code editor window with a Python file named `app.py` open. The code is a Streamlit application for document extraction using the Gemini Pro model. The code includes imports for Streamlit, OS, dotenv, Google Generative AI, PyPDF2, PIL, and pytesseract. It defines a function `configure_gemini` to load the Gemini Pro model and a `main` function to process a document image. The terminal shows the command `streamlit run app.py` being executed, and the output indicates the application is running on `http://localhost:8501`.

```
1 import streamlit as st
2 import os
3 from dotenv import load_dotenv
4 from google.generativeai import configure, GenerativeModel
5 from PyPDF2 import PdfReader
6 from PIL import Image
7 import pytesseract
8
9 # Load environment variables
10 load_dotenv()
11
12 # Function to configure Google Generative AI
13 def configure_gemini():
14     api_key = os.getenv("GOOGLE_API_KEY")
```

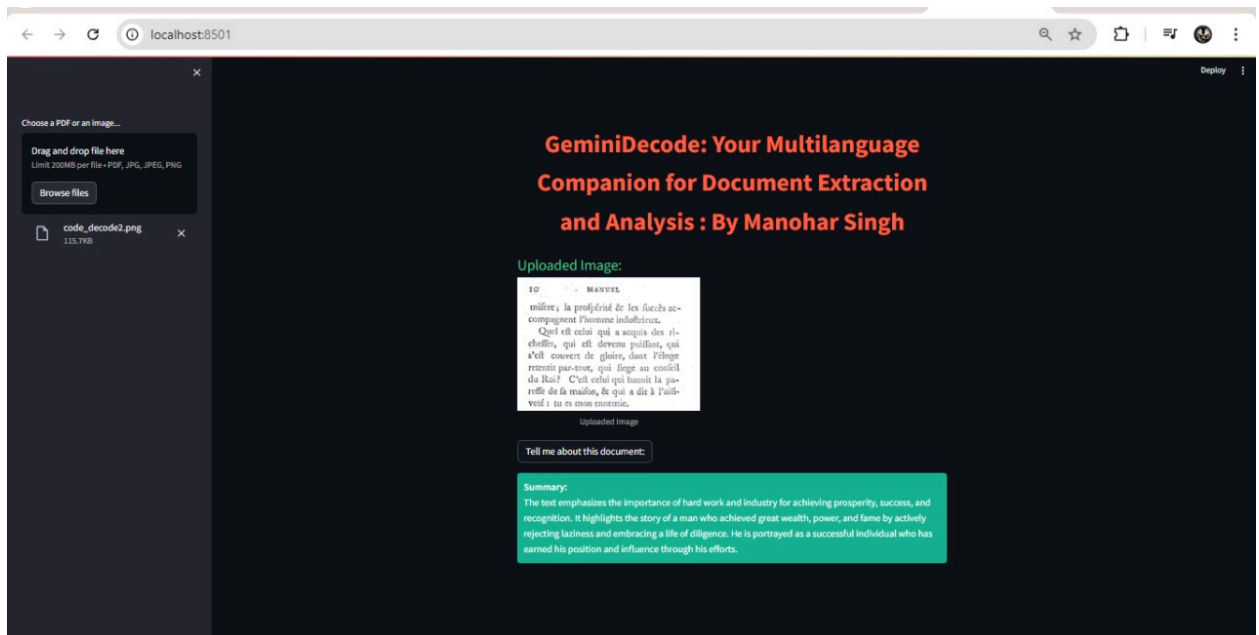
streamlit run app.py

You can now view your Streamlit app in your browser.

Local URL: <http://localhost:8501>
Network URL: <http://192.168.1.13:8501>

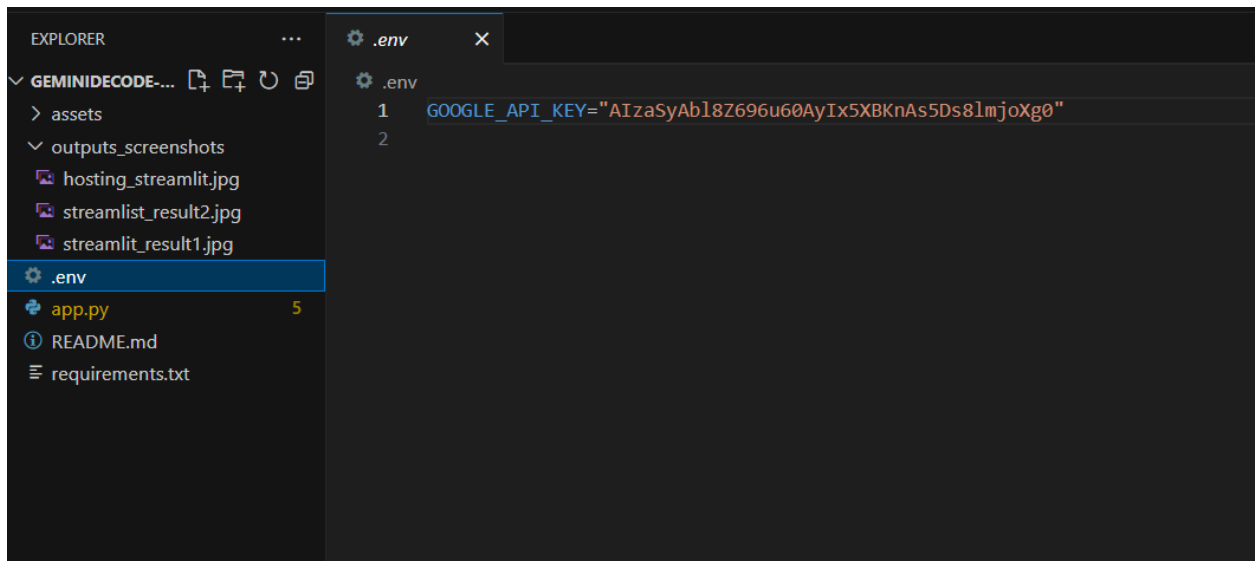
3. Model Output Example

A screenshot of the output displayed after processing a multilingual document.



4. Google API Key Initialization

Screenshot showing how the Google API key is initialized in the .env file for secure usage.



10. Challenges Faced

During the implementation of the **GeminiDecode** project, the following challenges were encountered:

- **Handling Multiple Languages:** Processing and extracting information from multiple languages required optimization in terms of NLP models and their deployment.
- **API Response Time:** Ensuring minimal delay in processing large document files.
- **Data Formatting:** Properly formatting the extracted information for presentation in the UI.

These challenges were addressed through model fine-tuning and API optimization techniques.

11. Conclusion

The **GeminiDecode: Multilanguage Document Extraction by Gemini Pro** project was successfully executed and tested. It has proven to be a powerful tool for extracting, processing, and categorizing multilingual documents in industries such as legal, financial, and healthcare. With the integration of advanced machine learning and NLP techniques, the project automates document management tasks, reduces manual effort, and improves decision-making processes across global organizations.

By utilizing a user-friendly interface built with Streamlit, the project is easily accessible and can be deployed for real-time document extraction tasks.

12. Future Work

Further improvements could include:

- Expanding the language support to additional languages.
- Enhancing the accuracy of document classification and extraction.
- Adding more customizability to the extraction process to allow users to specify sections of documents to extract.