SMARTSDL C Documentation

Overview

SMARTSDLC is an Al-powered software development lifecycle assistant built using Python, Gradio, PyTorch, and the Transformers library. It provides two main functionalities: Requirements Analysis and Code Generation. The application allows users to upload a PDF document or input text requirements, which are then analyzed to extract functional, non-functional, and technical requirements. Additionally, it can generate code in various programming languages based on user-specified requirements. This documentation provides an overview of the code structure, functionality, setup instructions, and usage quidelines.

Table of Contents

**Features** 

Dependencies

Code Structure

Functionality

Requirements Analysis

Code Generation

Setup Instructions

Usage

Limitations

Future Improvements

**Features** 

Requirements Analysis: Extracts and organizes software requirements from a PDF file or text input into:

**Functional Requirements** 

Non-Functional Requirements

**Technical Specifications** 

Code Generation: Generates code in multiple programming languages (Python, JavaScript, Java, C++, C#, PHP, Go, Rust) based on user-provided requirements.

User Interface: Built with Gradio, providing an intuitive web-based interface with two tabs:

Code Analysis: For uploading PDFs or entering requirements text.

Code Generation: For specifying code requirements and selecting a programming language.

Al Model: Powered by the ibm-granite/granite-3.2-2b-instruct model for natural language processing and code generation.

Dependencies

The following Python libraries are required to run the application:

gradio: For creating the web-based user interface.

torch: For handling the Al model and tensor computations.

transformers: For loading and using the pre-trained language model.

PyPDF2: For extracting text from PDF files.

io: For handling file streams.

These dependencies can be installed using the following commands:

pip install gradio torch transformers

pip install transformers torch gradio PyPDF2 -q

Code Structure

The code is structured as follows:

Imports:

Imports necessary libraries (gradio, torch, transformers, PyPDF2, io).

Model Setup:

Loads the ibm-granite/granite-3.2-2b-instruct model and tokenizer.

Configures the model to use GPU (if available) with torch.float16 precision or CPU with torch.float32.

Sets the padding token to the end-of-sequence token if not defined.

Core Functions:

generate\_response(prompt, max\_length): Generates a response using the Al model for a given prompt.

 $extract\_text\_from\_pdf(pdf\_file) : Extracts \ text \ from \ an \ uploaded \ PDF \ file.$ 

requirement\_analysis(pdf\_file, prompt\_text): Analyzes requirements from a PDF or text input and

organizes them.

code\_generation(prompt, language): Generates code in the specified programming language based on

user requirements.

Gradio Interface:

Creates a web interface with two tabs: "Code Analysis" and "Code Generation".

The interface includes file upload, text input, dropdowns, buttons, and output text boxes for user

interaction.

Functionality

Requirements Analysis

Input: Users can upload a PDF file containing software requirements or input requirements as text.

Processing:

If a PDF is uploaded, the extract\_text\_from\_pdf function extracts text using PyPDF2.

The extracted text or user-provided text is passed to the requirement\_analysis function.

A prompt is constructed to instruct the AI model to organize requirements into:

Functional Requirements: Features and functionalities the software must provide.

Non-Functional Requirements: Performance, usability, and other quality attributes.

Technical Specifications: Technical constraints or requirements.

Output: The Al model generates a structured response with the categorized requirements.

### Code Generation

Input: Users provide a description of the desired code and select a programming language from a dropdown menu.

Processing:

The code\_generation function constructs a prompt specifying the programming language and requirements.

The Al model generates code based on the prompt.

Output: The generated code is displayed in a text box.

Setup Instructions

Install Dependencies:Run the following commands in your Python environment: pip install gradio torch transformers pip install transformers torch gradio PyPDF2 -q

Run the Code:

Ensure you have a compatible Python environment (Python 3.7 or higher recommended). Copy the provided code into a .py file or a Jupyter Notebook (e.g., SMARTSDLC.ipynb). Execute the script or notebook. The Gradio interface will launch a web server.

#### Access the Interface:

If running locally, open the provided URL (e.g., <a href="http://127.0.0.1:7860">http://127.0.0.1:7860</a>) in a web browser. If running in a shared environment (e.g., Google Colab), use the provided public URL.

## Hardware Requirements:

GPU (Optional): For faster model inference, a CUDA-compatible GPU is recommended.

CPU: The application works on CPU but may be slower for large inputs.

RAM: At least 8GB of RAM is recommended for smooth operation.

Usage

### Requirements Analysis:

Navigate to the "Code Analysis" tab.

 $\label{thm:policy} \textbf{Upload a PDF file containing software requirements or enter requirements in the text box.}$ 

Click the "Analyze" button to view the categorized requirements in the output box.

### Code Generation:

Navigate to the "Code Generation" tab.

Enter the code requirements in the text box (e.g., "Create a function to calculate the factorial of a number").

Select the desired programming language from the dropdown (e.g., Python, JavaScript).

Click the "Generate Code" button to view the generated code.

## Example Inputs:

# Requirements Analysis:

PDF: Upload a document with requirements like "The system must allow user login with email and password."

Text: "The software should support user authentication, have a response time under 2 seconds, and use a MySQL database."

### Code Generation:

Prompt: "Write a function to sort an array of integers in ascending order."

Language: Python

Expected Output: A Python function implementing a sorting algorithm.

### Limitations

PDF Extraction: The PyPDF2 library may struggle with poorly formatted PDFs or scanned documents without OCR.

Model Accuracy: The ibm-granite/granite-3.2-2b-instruct model may occasionally produce incomplete or incorrect outputs, especially for complex requirements or code.

Language Support: Code generation is limited to the languages listed in the dropdown. Support for other languages may require modifications.

Performance: On CPU, processing large PDFs or generating complex code may be slow.

Error Handling: Limited error handling for invalid PDFs or malformed inputs.

## Future Improvements

Enhanced PDF Processing: Integrate OCR or advanced PDF parsing libraries for better text extraction.

Model Fine-Tuning: Fine-tune the Al model for specific software engineering tasks to improve accuracy.

Additional Languages: Expand the list of supported programming languages.

Code Validation: Add functionality to validate and test generated code.

Export Options: Allow users to download analysis results or generated code as files.

Improved Error Handling: Implement robust error handling for invalid inputs or processing errors.