Smart SDLC – Al-Enhanced Software Development Lifecycle

Generative AI with IBM - Project Report

Submitted by: Team Leader - Subbashini G

Team Members - Swathi S, Tamil selvi S, Swetha S, Swetha R

TABLE OF CONTENTS

- 1. ABSTRACT
- 2. OBJECTIVES
- 3. SYSTEM REQUIREMENTS
- 4. IMPLEMENTATION
- 5. PROJECT FILES
- 6. APPENDIX: SOURCE CODE
- 7. OUTPUT SCREENSHOTS
- 8. CONCLUSION

1.ABSTRACT

The Smart SDLC Platform is a smart tool designed to help automate different stages of software development. It uses advanced AI models like Granite-20B-Code-Instruct and LangChain to understand user inputs such as code, questions, prompts, and PDF files.

Based on these inputs, the platform can generate software requirements, code, test cases, and summaries.

The platform is built using FastAPI and Streamlit, making it fast and easy to use. It also includes a floating chatbot assistant that helps users by answering questions and giving coding support. Tools like PyMuPDF and Uvicorn, along with AI services from IBM Watsonx, are used to make the platform work efficiently.

Overall, SmartSDLC helps developers save time, reduce manual work, and speed up the software development process by turning natural language inputs into useful software components.

2. OBJECTIVES

- ✓ Automate and streamline the Software Development Life Cycle (SDLC) using Al-driven tools.
- ✓ Leverage advanced language models like Granite-20B-Code-Instruct via IBM Watsonx for intelligent code and text processing.
- ✓ Use LangChain to integrate LLM capabilities into the development pipeline.
- ✓ Accept diverse user inputs.
- ✓ Convert inputs into structured outputs
- ✓ Build with FastAPI and Streamlit for a responsive backend and easy-to-use frontend.
- ✓ Include a floating chatbot assistant.

3.SYSTEM REQUIREMENTS

Software Requirements:

- Programming Languages & Tools:
- Python 3.10
- Fast API For building APIs
- Streamlit For interactive front-end interface
- Uvicorn ASGI server for FastAPI
- PyMuPDF (fitz) For reading PDF documents
- AI & NLP Tools:
- IBM Watsonx AI & Granite Models For AI model inference
- LangChain To integrate LLMs and tools
- Git & GitHub -

Hardware Requirements:

- A system capable of running Python 3.10 (Windows/Linux/Mac)
- At least 8 GB RAM (recommended for Al processing)
- Stable internet connection (for API calls and model access)
- Optional: GPU (if running models locally or handling large documents).

4.PROJECT WORKFLOW

Environment Setup:

Python 3.10 with FastAPI, Streamlit, LangChain, PyMuPDF, and IBM Granite Al via Watsonx.

Input Module:

Handles PDFs, code snippets, and natural language prompts.

Requirement & User Story Extraction:

Al extracts requirements from PDFs and creates user stories.

Code Generation & Testing:

Converts prompts to code, generates unit tests, and fixes bugs automatically.

Code Summarization:

Provides simple explanations of code functionality.

Report & Interface:

Compiles outputs into downloadable reports; user-friendly web interface with chatbot support.

Feedback & Safety:

Collects user feedback; reminds users to verify Al-generated outputs.

5.PROJECT FILES

- env Stores API keys and configurations
- App.py Main application file with Gradio blocks
- requirements.txt Lists required dependencies

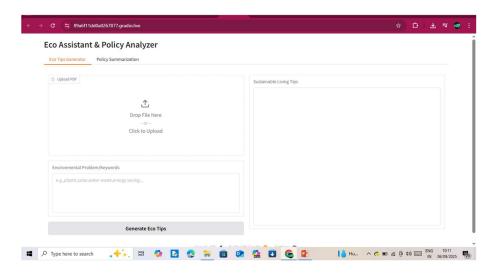
6. APPENDIX: SOURCE CODE

```
import gradio as gr
import torch
from transformers import AutoTokenizer, AutoModelForCausalLM
import PyPDF2
import io
# Load model and tokenizer
model name = "ibm-granite/granite-3.2-2b-instruct"
tokenizer = AutoTokenizer.from pretrained(model name)
model = AutoModelForCausalLM.from pretrained(
  model name.
  torch dtype=torch.float16 if torch.cuda.is available() else torch.float32,
  device map="auto" if torch.cuda.is available() else None
)
if tokenizer.pad token is None:
  tokenizer.pad token = tokenizer.eos token
def generate_response(prompt, max_length=1024):
  inputs = tokenizer(prompt, return tensors="pt", truncation=True, max length=512)
```

```
if torch.cuda.is available():
    inputs = {k: v.to(model.device) for k, v in inputs.items()}
  with torch.no grad():
    outputs = model.generate(
       **inputs,
       max length=max length,
       temperature=0.7,
       do sample=True,
       pad_token_id=tokenizer.eos_token_id
    )
  response = tokenizer.decode(outputs[0], skip_special_tokens=True)
  response = response.replace(prompt, "").strip()
  return response
def extract text from pdf(pdf file):
  if pdf file is None:
    return ""
  try:
    pdf reader = PyPDF2.PdfReader(pdf file)
    text = ""
    for page in pdf reader.pages:
       text += page.extract_text() + "\n"
    return text
  except Exception as e:
    return f"Error reading PDF: {str(e)}"
def requirement analysis(pdf file, prompt text):
  # Get text from PDF or prompt
  if pdf file is not None:
    content = extract text from pdf(pdf file)
     analysis prompt = f"Analyze the following document and extract key software requirements.
Organize them into functional requirements, non-functional requirements, and technical
specifications:\n\n{content}"
  else:
     analysis prompt = f"Analyze the following requirements and organize them into functional
requirements, non-functional requirements, and technical specifications:\n\n{prompt text}"
  return generate response(analysis prompt, max length=1200)
def code generation(prompt, language):
  code prompt = f"Generate {language} code for the following
requirement:\n\n{prompt}\n\nCode:"
  return generate response(code prompt, max length=1200)
# Create Gradio interface
with gr.Blocks() as app:
  gr.Markdown("# Al Code Analysis & Generator")
  with gr.Tabs():
    with gr.TabItem("Code Analysis"):
       with gr.Row():
          with gr.Column():
            pdf_upload = gr.File(label="Upload PDF", file_types=[".pdf"])
```

```
prompt input = gr.Textbox(
              label="Or write requirements here",
              placeholder="Describe your software requirements...",
              lines=5
            analyze btn = gr.Button("Analyze")
         with gr.Column():
            analysis output = gr.Textbox(label="Requirements Analysis", lines=20)
       analyze_btn.click(requirement_analysis, inputs=[pdf_upload, prompt_input],
outputs=analysis_output)
    with gr.TabItem("Code Generation"):
       with gr.Row():
         with gr.Column():
            code prompt = gr.Textbox(
              label="Code Requirements".
              placeholder="Describe what code you want to generate...",
              lines=5
            language dropdown = gr.Dropdown(
              choices=["Python", "JavaScript", "Java", "C++", "C#", "PHP", "Go", "Rust"],
              label="Programming Language",
              value="Python"
            generate_btn = gr.Button("Generate Code")
         with gr.Column():
            code output = gr.Textbox(label="Generated Code", lines=20)
       generate_btn.click(code_generation, inputs=[code_prompt, language_dropdown],
outputs=code output)
app.launch(share=True)
```

7. OUTPUT SCREENSHOT



Eco Assistant & Policy Analyzer is a user-friendly web tool that helps promote sustainable living and environmental awareness. It features:

- Eco Tips Generator Users can upload documents or enter keywords to get practical tips for eco-friendly living.
- Policy Summarization Simplifies complex environmental policies into easyto-understand summaries.

Built using **Gradio**, the tool makes environmental guidance and policy insights accessible to everyone.

8.CONCLUSION

SmartSDLC is an AI-powered platform designed to automate and enhance the entire Software Development Lifecycle (SDLC) — from requirements analysis and code generation to testing, bug fixing, and documentation. By leveraging cutting-edge technologies such as IBM Watsonx, FastAPI, LangChain, and Streamlit, it streamlines development processes, minimizes manual errors, and accelerates software delivery.

Featuring a modular architecture and an intuitive user interface, SmartSDLC empowers both technical and non-technical users to manage SDLC tasks effectively. Its key functionalities include AI-driven requirement classification, automated user story creation, code generation from natural language input, test case generation, intelligent bug resolution, and built-in chat assistance.

SmartSDLC significantly improves development productivity and quality, while setting the stage for future enhancements like CI/CD integration, team collaboration tools, version control, and cloud deployment—making it a robust solution for modern, agile development environments.