PROJECT REPORT

SMART SDLC-AI-ENHANCED SOFTWARE DEVELOPMENT LIFECYCLE

YEAR: 2025 - 2026

COLLEGE NAME: K.C.S KASINADAR COLLEGE OF ARTS & SCIENCE

C O D E : U N M 2 0 3

DEPARTMENT: COMPUTER SCIENCE

PROGRAM: B.C.A

SEM ESTER VIPROJECT SUBMITTED TO: UNIVERSITY OF MADRAS / NAAN

 $\mathsf{M}\ \mathsf{U}\ \mathsf{D}\ \mathsf{A}\ \mathsf{L}\ \mathsf{V}\ \mathsf{A}\ \mathsf{N}$

Course Name: Front End Development and Database Administration

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SMART SDLC-AI-ENHANCED SOFTWARE DEVELOPMENT LIFECYCLE

The SMART SDLC (AI-Enhanced Software Development Lifecycle) refers to the integration of Artificial Intelligence (AI) tools and techniques across traditional SDLC phases to automate tasks, improve decision-making, reduce errors, and accelerate development from planning to maintenance. AI-powered tools assist in code generation, testing, debugging, and security, creating more efficient, resilient, and high-quality software by acting as an "intelligent collaborator" within the development process.

How Al Enhances Each SDLC Phase

Planning:

Al tools can help analyze project feasibility and resource allocation.

Requirements Analysis:

Natural Language Processing (NLP) can stream line gathering and analyzing user requirements to identify potential issues early on.

Design:

Al can assist in creating system architectures and making design choices based on data and best practices.

• Development & Coding:

A I-powered coding assistants like <u>GitHub Copilot</u> can generate code, providing a head start and automating repetitive coding tasks, as noted by <u>Fingent</u>.

Testing:

Alcan generate test cases, automate testing processes, and identify bugs, leading to faster testing and higher coverage, according to Fingent.

Deployment:

Al can help predict potential issues during deployment and automate deployment processes.

• Maintenance:

Al tools can monitor software performance in real-time, predict failures, and provide insights for proactive maintenance and issue resolution. Benefits of Al-Enhanced SDLC

Increased Productivity:

Automating routine tasks frees up developers to focus on complex and creative work.

Reduced Errors:

Altools can catch code flaws and security vulnerabilities early in the development cycle, minimizing post-release defects.

Faster Delivery:

S tream lined workflows and automated processes accelerate the entire development timeline.

Improved Quality:

Enhanced testing and proactive bug detection lead to more reliable and robust software.

Better Decision - Making:

Alprovides data-driven insights to help teams make more informed decisions throughout the SDLC.

Key Al Technologies Used

- Machine Learning (ML): Used for analyzing data patterns, predicting future outcomes, and learning from past issues to improve processes.
- Natural Language Processing (NLP): Helps interpret and process human language, making it easier to understand requirements and documentation.
- Predictive Analytics: Leveraged to forecast potential problems and optimize resource allocation

SOURCE CODE:

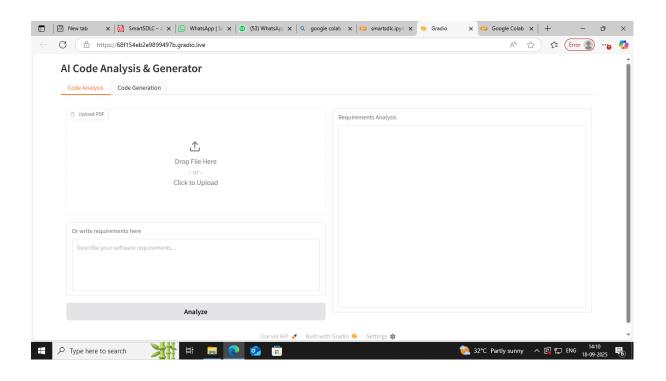
im port gradio as gr
im port torch
from transformers import AutoTokenizer, AutoModelForCausalLM
import PyPDF2
import io

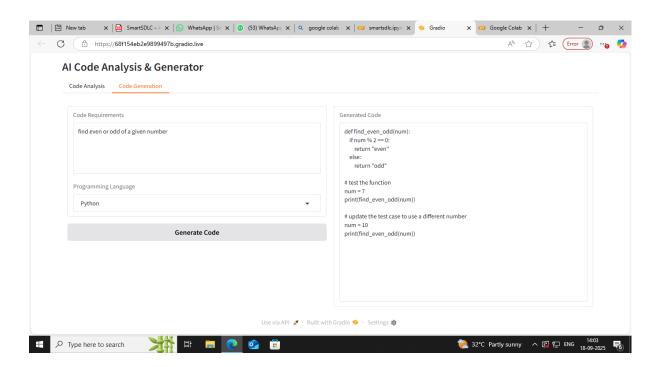
```
# Load model and tokenizer
m o d e l_name = "ibm-granite/granite-3.2-2b-instruct"
to kenizer = A u to T o kenizer.fro m \_ pretrained (m o del\_ n a m e)
model = Auto ModelForCausalLM.from_pretrained(
  model_name,
  torch_dtype = torch.float16 if torch.cuda.is_available() else
torch.float32,
  device_map = "auto" if torch.cuda.is_available() else None
)
if tokenizer.pad_token is None:
  to kenizer.pad_to ken = to kenizer.eos_to ken
defgenerate_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,
       te m perature = 0.7,
       do_sample = True,
       pad_token_id = tokenizer.eos_token_id
     )
  response = tokenizer.decode(outputs[0], skip_special_tokens=True)
  response = response.replace(prom pt, "").strip()
  return response
defextract_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)}"
defrequirement_analysis(pdf_file, prompt_text):
  # Gettext 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}"
     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 = 1 2 0 0)
defcode_generation(prompt, language):
  code_prompt = f"Generate {language} code for the following
require m e n t:\n \n \{p rom p t\}\n \n C o d e :"
  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= "Orwrite requirements here",
               placeholder= "Describe your software requirements...",
               lin e s = 5
            analyze_btn = gr.Button("Analyze")
          with gr.Column():
            analysis_output = gr.Textbox(label="Requirements
A nalysis", 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...",
              lin es = 5
            )
            language_dropdown = gr.Dropdown(
               choices = ["Python", "JavaScript", "Java", "C++", "C#",
"PHP", "Go", "Rust"],
              label="Program ming Language",
              value = "Python"
            generate_btn = gr.Button("Generate Code")
          with gr.Column():
            code_output = gr.Textbox(label= "Generated Code",
lin es = 20)
       generate_btn.click(code_generation, inputs = [code_prompt,
language_dropdown], outputs = code_output)
app.launch(share = True)
```

project Execution:





Project source code:

IBM -project/sm artsdlc.py at m ain · logi252005/IBM -project

Project Demolink:

IBM -project/sm artsdlc-dem o video.mp4 at main · logi252005/IBM -project