

# Sustainable Smart City Assistant Using IBM Granite LLM

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## Documentation

### 1. Introduction

**Project Title: Sustainable Smart City Assistant Using IBM granite LLM**

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### 2. Project Overview

The Sustainable Smart City Assistant leverages **IBM's Granite Large Language Model (LLM)** to provide intelligent, real-time support for urban planning, resource management, citizen engagement, and sustainability initiatives in smart cities. This AI-driven assistant helps city administrators, residents, and businesses make data-informed decisions that align with sustainability goals and improve quality of life.

#### • Features:

- **Personalized Sustainability Advisor**  
Delivers tailored eco-friendly tips and actionable recommendations based on user behavior, location, and city sustainability goals.
- **Intelligent Policy Digest**  
Transforms lengthy regulatory and environmental documents into concise, easy-to-understand summaries highlighting critical points and community impact.
- **Seamless Document Processing**  
Supports direct uploading of PDFs and other document formats, automatically extracting and analyzing text for efficient policy review and insights.

- **User-Friendly Interactive Dashboard**  
Built with Gradio, the interface enables effortless conversations, document uploads, and visualization of sustainability data through a clean and accessible design.
- **Advanced AI Powered by IBM Granite via Hugging Face**  
Integrates IBM Granite LLM through Hugging Face Transformers for high-accuracy natural language understanding, supporting multilingual queries and contextual responses.

### 3. Architecture

Frontend (Gradio):

Provides a tabbed interface with input fields, buttons, and output textboxes for both eco tips and policy summaries.

Backend (PyTorch + Transformers):

Handles text processing, PDF parsing, and interaction with the IBM Granite model.

LLM Integration (IBM Granite):

Granite-3.2-2b-instruct model from Hugging Face provides language understanding and generation.

### 4. Setup Instructions

**Prerequisites:**

- Python 3.9 or later
- pip and virtual environment tools
- Internet access to download Hugging Face models

**Installation Process:**

- Install dependencies: `pip install transformers torch gradio PyPDF2`
- Run the script: `python maja.txt`
- The Gradio app will launch locally with a shareable link.

## 5. Folder Structure

maja.txt – Main application file containing model loading, functions, and Gradio interface.

functions/ – (Optional future extension) Separate utility functions.

uploads/ – (Optional) Store uploaded PDF files.

## 6. Running the Application

- Run python code
- Open the link in your browser provided by Gradio.
- Navigate to Eco Tips Generator tab for sustainability suggestions.
- Navigate to Policy Summarization tab to analyze PDFs or pasted text.

## 7. User Flows

Eco Tips Generator: Input keywords → AI generates actionable tips.

Policy Summarization: Upload PDF or paste text → AI produces key points.

## 8. Authentication

The current version runs in an open environment for demonstration. Future enhancements may include user authentication and access control.

## 9. User Interface

- Tabbed layout with two main sections (Eco Tips Generator, Policy Summarization).
- File upload support for policy analysis.
- Large output textboxes for readability.
- Shareable Gradio link for easy collaboration.

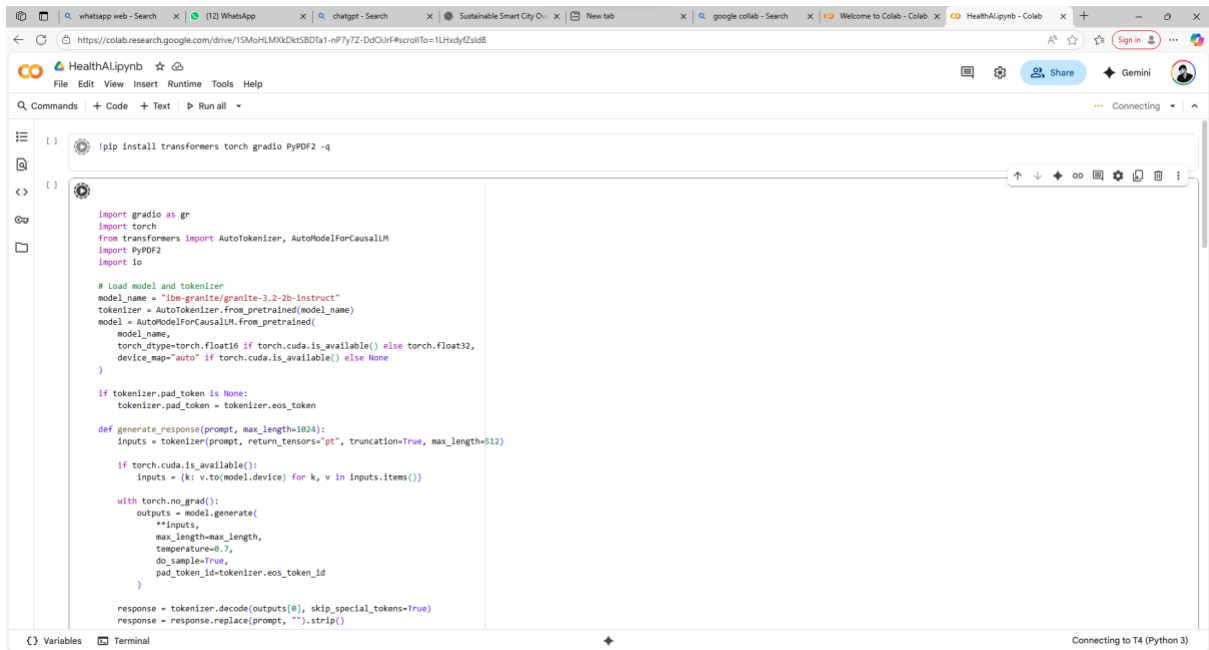
## 10. Testing

Unit Testing: Functions like `eco_tips_generator` and `policy_summarization` were tested individually.

Manual Testing: Gradio interface tested with sample inputs and PDFs.

Edge Case Handling: Empty input, corrupted PDFs handled with error messages.

## 11. Screenshots



```
!pip install transformers torch gradio PyPDF2 -q

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()
```

```
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 eco_tips_generator(problem_keywords):
    prompt = f"Generate practical and actionable eco-friendly tips for sustainable living related to: {problem_keywords}. Provide specific solutions and suggestions:"
    return generate_response(prompt, max_length=1000)

def policy_summarization(pdf_file, policy_text):
    # Get text from PDF or direct input
    if pdf_file is not None:
        content = extract_text_from_pdf(pdf_file)
        summary_prompt = f"Summarize the following policy document and extract the most important points, key provisions, and implications:\n\n{content}"
    else:
        summary_prompt = f"Summarize the following policy document and extract the most important points, key provisions, and implications:\n\n{policy_text}"

    return generate_response(summary_prompt, max_length=1200)

# Create Gradio Interface
with gr.Blocks() as app:
    gr.Markdown("# Eco Assistant & Policy Analyzer")

    with gr.Tabs():
        with gr.Tabitem("Eco Tips Generator"):
            with gr.Row():
                with gr.Column():
                    keywords_input = gr.Textbox(
                        label="Environmental Problem/Keywords",
                        placeholder="e.g., plastic, solar, water waste, energy saving...",
                    )
                    generate_tips_btn = gr.Button("Generate Eco Tips")

            with gr.Column():
                tips_output = gr.Textbox(label="Sustainable Living Tips", lines=15)

            generate_tips_btn.click(eco_tips_generator, inputs=keywords_input, outputs=tips_output)

        with gr.Tabitem("Policy Summarization"):
            with gr.Row():
                with gr.Column():
                    pdf_upload = gr.File(label="Upload Policy PDF", file_types=[".pdf"])
                    policy_text_input = gr.Textbox(
                        label="Or paste policy text here",
                        placeholder="Paste policy document text...",
                        lines=5
                    )
                    summarize_btn = gr.Button("Summarize Policy")

            with gr.Column():
                summary_output = gr.Textbox(label="Policy Summary & Key Points", lines=20)

            summarize_btn.click(policy_summarization, inputs=[pdf_upload, policy_text_input], outputs=summary_output)

app.launch(share=True)
```

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)
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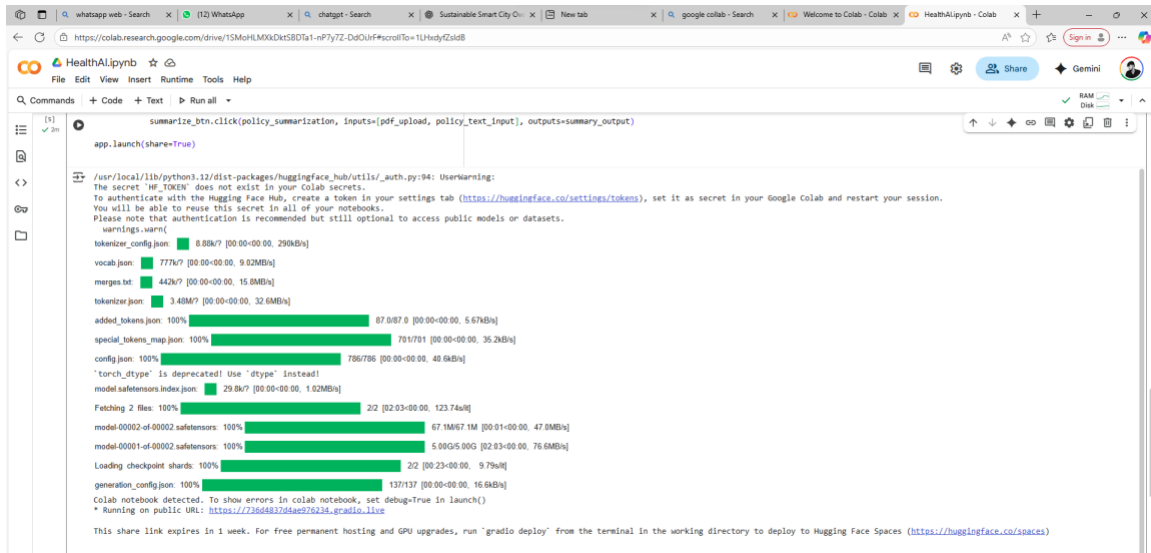
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```



## 12. Known Issues

- Large PDFs may slow down summarization.
- Mobile app performance optimization pending.
- Requires stable internet connection for model usage.
- Limited offline functionality.

## 13. Future Enhancements

- Add multi-language support.
- Provide visual insights such as charts or graphs.
- Detailed policy recommendations beyond summarization.
- Extend API endpoints for integration with other applications.

## 14. GitHub Link:

Code link:

<https://github.com/Prathivraj09/IBM-Project>