# Sustainable Smart City Assistant Using IBM Granite LLM

Category: Generative AI with IBM

Skills Required: Machine Learning, IBM Cloud, IBM Watson

#### 1. Introduction

Project Title: Sustainable Smart City Assistant Using IBM Granite LLM

Purpose: To empower cities and residents with AI-powered tools for urban sustainability, governance, and citizen engagement.

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## 2. Project Overview

The Sustainable Smart City Assistant is an AI-powered platform leveraging IBM Watsonx Granite LLM and modern data pipelines to enhance urban sustainability and citizen engagement. It integrates features like policy summarization, KPI forecasting, eco-tips, anomaly detection, and a conversational chat assistant through a modular FastAPI backend and Streamlit/Gradio dashboard.

## 3. Features

- Conversational Interface: Natural language interaction for citizens and officials.
- Policy Summarization: Converts lengthy policy documents into concise summaries.
- Resource Forecasting: Predicts energy, water, and waste usage trends.
- Eco-Tip Generator: Recommends eco-friendly lifestyle practices.
- Citizen Feedback Loop: Collects and processes citizen feedback.
- KPI Forecasting: Projects future consumption trends and KPIs.
- Anomaly Detection: Detects unusual data patterns and issues.
- Multimodal Input Support: Accepts text, PDFs, and CSVs for analysis.
- Streamlit/Gradio UI: Interactive dashboard for officials and citizens.

#### 4. Architecture

The platform uses a modular architecture with the following components: Frontend (Streamlit / Gradio): Interactive dashboards, chat, feedback forms, and report viewers. Backend (FastAPI): High-performance REST APIs to handle data and LLM requests. LLM Integration (IBM Watsonx Granite): Core natural language

understanding and generation. • Vector Search (Pinecone): Enables semantic search across documents. • ML Modules: Forecasting and anomaly detection using Scikit-learn and Pandas.

#### 5. Use Case Scenarios

- Policy Search & Summarization: Municipal planners can upload policy documents and instantly receive concise, citizen-friendly summaries.
- Citizen Feedback Reporting: Residents report issues like water leaks via the assistant, which logs them with category tagging.
- KPI Forecasting: Administrators upload KPI data, and the assistant forecasts future resource usage.
- Eco Advice & Sustainability: Generates personalized eco-friendly tips to encourage sustainable living.

## 6. Setup Instructions

Prerequisites: • Python 3.9 or later • IBM Watsonx and Pinecone API keys • Internet accessSteps to Run: 1. Clone the repository and install dependencies. 2. Create a .env file and configure credentials. 3. Launch FastAPI backend and Streamlit/Gradio frontend. 4. Upload data and interact with the assistant modules.

#### 7. API Documentation

Available API Endpoints: • POST /chat/ask  $\rightarrow$  Responds with AI-generated answers. • POST /upload-doc  $\rightarrow$  Upload and embed documents. • GET /search-docs  $\rightarrow$  Semantic policy search. • GET /get-eco-tips  $\rightarrow$  Provides sustainability tips. • POST /submit-feedback  $\rightarrow$  Collects citizen feedback.

## 8. Authentication & Security

The demo version is open, but secure deployments can integrate: • Token-based authentication (JWT/API Keys) • OAuth2 with IBM Cloud • Role-based access (Admin, Citizen, Researcher)

#### 9. User Interface Design

• Sidebar navigation• KPI visualizations and summaries• Tabs for chat, eco tips, and forecasting• PDF report downloads

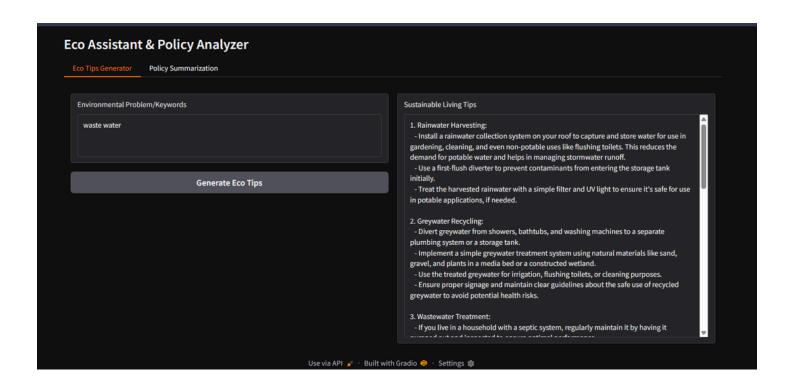
## 10. Testing & Validation

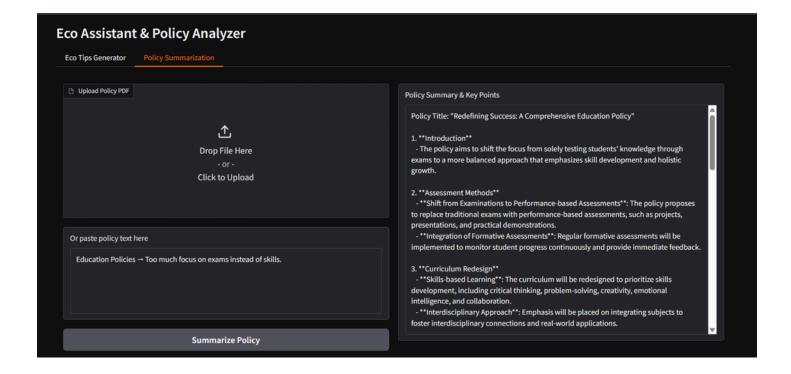
Testing includes: • Unit testing for prompt functions • API testing via Swagger/Postman • Manual testing for uploads and outputs • Edge-case handling for large files and invalid inputs

#### 11. Future Enhancements

• Integration with IoT sensor networks• Advanced analytics dashboards• Support for multilingual city policies• Mobile app integration

## 12. Screen shots





## 13. Source Code

!pip install transformers torch gradio PyPDF2 -q

import gradio as gr

import torch

from transformers import AutoTokenizer, AutoModelForCausalLM

import PyPDF2

```
# 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 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...",
            lines=3
          )
          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)
```

## 14. GitHub & Project Demo Link GitHub Repo:

https://github.com/mahasri2007/sub-smart-city.git