Sustainable Smart City Assistant using IBM Granite LLM

1.introduction:

- PROJECT TITLE: SUSTAINABLE SMART CITY ASSISTANT USING IBM GRANITE LLM.
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2.PROJECT OVERVIEW: The Sustainable Smart City Assistant is an Alpowered system designed to improve urban living by enabling intelligent, data-driven decision-making. It leverages IBM Granite Large Language Models (LLMs) to provide context-aware insights, natural language interactions, and sustainable solutions for city governance and citizen engagement.

<u>3. Role of IBM Granite:</u> LLM IBM Granite LLM provides advanced natural language understanding and generation capabilities. It powers conversational interfaces, decision support systems, and real-time policy simulations. Its enterprise-grade design ensures trust, transparency, and scalability for smart city applications.

4.AUTHENTICATION: Authentication

All endpoints require authentication via API key or OAuth 2.0. Use HTTPS for all requests.

Header (API Key):

Authorization: Bearer <API_KEY>

<u>OAuth 2.0:</u> Bearer tokens obtained via the token endpoint with client credentialgrant.

Rate limiting & Throttling

<u>Default rate limits:</u> • 60 requests per minute per API key (standard)

• 600 requests per minute (enterprise tier).

When limits are exceeded, API returns HTTP 429 with Retry-After header.

5. Use Cases • :

• **Mobility:** Al-driven traffic optimization and public transport planning .

- Energy: Smart grid demand prediction and renewable integration.
- Waste: Route optimization for collection and recycling .
- <u>Water</u>: Leak detection and efficient distribution management Governance: Policy simulation and automated reporting.
- Citizen Services: 24/7 multilingual digital assistant.

6. Sustainability Impact:

The assistant supports the United Nations Sustainable Development Goals (SDGs) by reducing carbon emissions, improving resource efficiency, and fostering inclusive governance. Predictive analytics enable cities to minimize energy waste, optimize water usage, and create greener mobility solutions.

7. Challenges & Future Scope: Key challenges include ensuring data privacy, achieving interoperability among diverse city systems, and managing scalability. Future advancements will involve deeper Al integration with edge computing, autonomous city operations, and continuous learning from citizen feedback.

8. Roles & Scopes:

Access control is managed via roles and scopes assigned to tokens.

Citizen Role:

- tips:read
- chat:read

Planner Role:

- policy:read
- analytics:read

Admin Role:

- * (all endpoints)

IoT Device Role:

- telemetry:write

6. Security Best Practices

- * Always use HTTPS (TLS 1.2+)
- * Rotate API keys every 90 days

if tokenizer.pad token is None:

- * Use short-lived OAuth tokens in production
- * Store secrets in vaults (e.g., AWS Secrets Manager, HashiCorp Vault)
- * Enable audit logging for all authentication events
- * Apply role-based access control (RBAC)
- * Monitor failed login attempts and anomalies

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
```

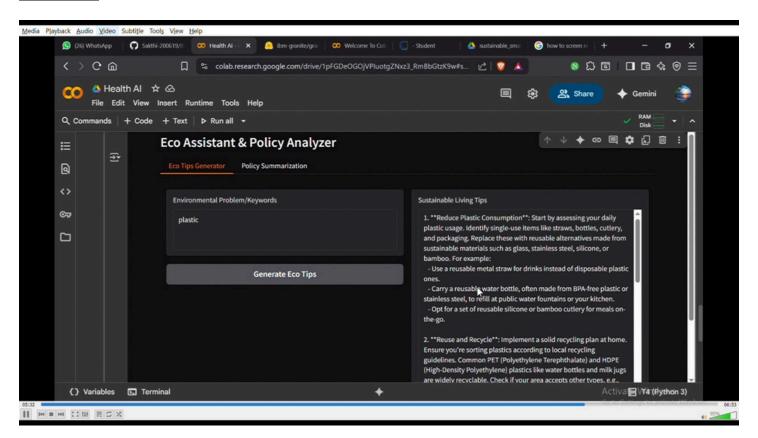
```
tokenizer.pad token=tokenizer.eos token
def generate_response(prompt, max_length=1024):
  inputs=tokenizer(prompt, return_tensors="pt", truncation=True,
max_length=512)
if torch.cuda.is_available():
  inputs={k: v.t(model.device) for k, v in inputs.items()}
with torch.no grad():
 outputs = model.generate(
   **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:
```

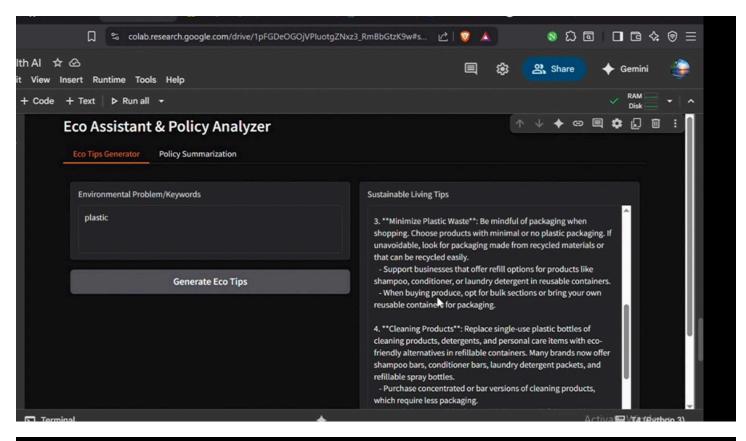
```
text+=page.extract_text() + "\n"
   return text
  except Exception as e:
return f"Error reading POF: {str(e)}"
def eco_tips_generator (problem_keywords):
  prompt=f"Generate practical and actionable eco-friendly tips for
sustainable living related to:{problem_keywords}.provide specific
solutions 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"):
```

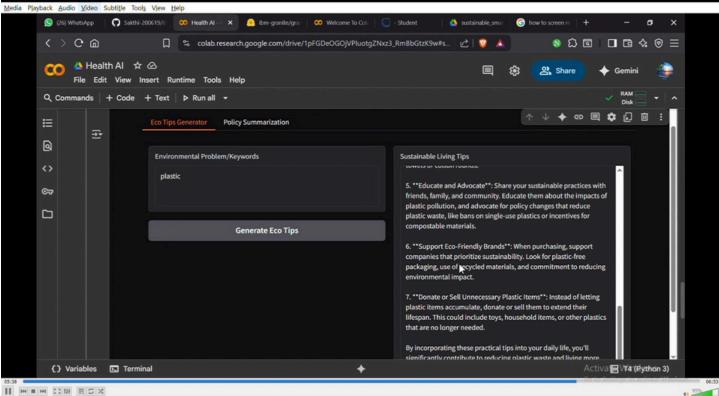
```
with gr.Row():
        with gr.Column():
         keywords_input=gr.Textbox(
            label="Environmental Problem/Keywords",
            placeholder="e.g., plastic, solar, water waste, energy
saving...",
            lines=3
         with gr.Column():
           keywords input=gr.Textbox(
            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():
```

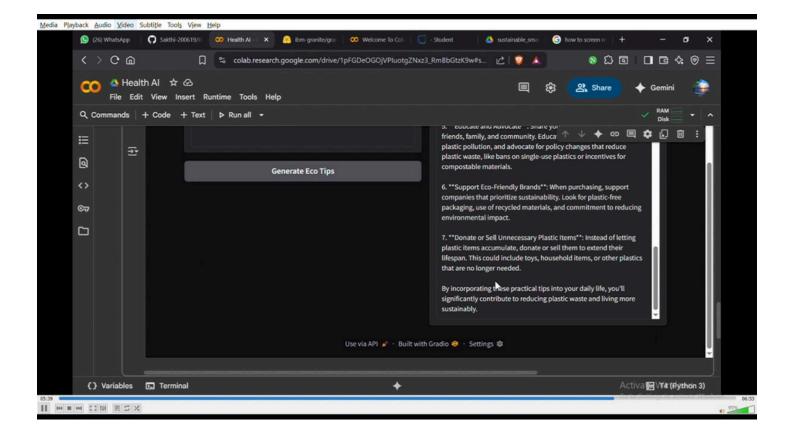
summarize_btn.click(policy_summarization, inputs=[pdf_upload,
policy_text_input], outputs=summary_output)
app.launch(share=True)

OUTPUT:









10. Conclusion:

By combining IBM Granite LLM with IoT and smart city platforms, the Sustainable Smart City Assistant provides a scalable, trustworthy, and citizen-focused solution to build more sustainable and livable urban environments.