Smart Traffic Management

"Multi-Agent System for Smart Traffic Management"

Context:

Urban traffic congestion leads to significant time loss, pollution, and fuel consumption. Static traffic signals and traditional rule-based systems fail to optimize traffic flow dynamically. Challenge:

Develop a multi-agent system where Al-powered traffic lights, smart vehicles, and drones collaborate to manage real-time traffic, reducing congestion and improving emergency vehicle response times.

Goals:

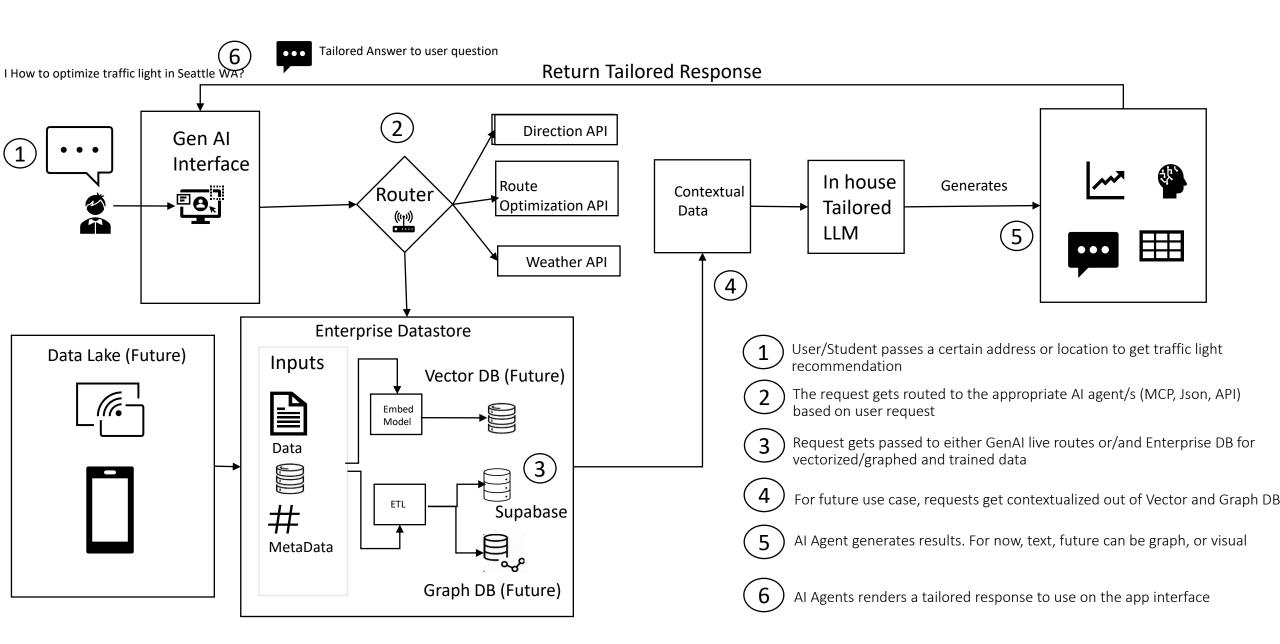
Enable Al agents to analyze live traffic data and adjust signals dynamically.

- Implement vehicle-to-infrastructure (V2I) communication for smarter routing.
- Simulate various urban traffic scenarios (rush hour, accidents, roadblocks).
- Ensure fairness in traffic distribution, preventing excessive delays in specific areas.

Bonus Challenges:

- Introduce an agent that prioritizes emergency vehicles while optimizing overall traffic flow.
- Develop an Al-based toll pricing system that dynamically adjusts based on congestion levels.

Smart Traffic Management High-level Process Overview



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Data Samples

```
"Seattle, WA": {
    "streets": [
            "name": "Main St",
            "vehicle_count": 120,
            "average_speed": 25,
            "incidents": ["Minor accident near 3rd Ave"],
            "pedestrian_activity": "High",
            "cyclist_activity": "Moderate",
            "time_of_day": "Peak Hours"
        },
            "name": "Pine St",
            "vehicle_count": 80,
            "average_speed": 35,
            "incidents": [],
            "pedestrian_activity": "Moderate",
            "cyclist_activity": "Low",
            "time_of_day": "Off-Peak Hours"
        },
            "name": "4th Ave",
```

Key Code Snippet - Leveraging MCP

Connecting to MCP server for traffic live data

```
from praisonaiagents import Agent, MCP
import gradio as gr
import requests
import json
# Define a function to analyze traffic and adjust signal lights
def analyze_traffic_MCP(query):
    # Connect to a hypothetical MCP server for live traffic data
    agent = Agent(
        instructions="""You are an AI traffic management assistant. Analyze live traffic data an
        recommendations for adjusting signal lights dynamically to optimize traffic flow."",
        11m="gpt-4o-mini",
        tools=MCP("npx -y @opentraffic/mcp-server-traffic --live-data")
    # Process the query through the agent
```

Key Code Snippet – Connecting using API calls

```
def analyze_traffic_api_agent(location):
    api_key = "AIzaSyB05CAgY-S2AqTlmNWJ0Q66rEbY5iWJd9A"
    url = f"https://maps.googleapis.com/maps/api/traffic/json?location={location}&key={api_key}'
    response = requests.get(url)
    if response.status code == 200:
       traffic data = response.json()
       # Process the traffic data into a readable format
       if "error" in traffic data:
            return f"Error: {traffic_data['error']['message']}"
        else:
            # Format the traffic data into Markdown
            formatted data = "### Traffic Data\n\n"
            for key, value in traffic data.items():
                formatted_data += f"- **{key}**: {value}\n"
            return formatted data
    else:
        return f"Error: Unable to fetch traffic data (HTTP {response.status_code})"
```

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Key Code Snippet – Rendering simple interface

```
# Create a Gradio interface

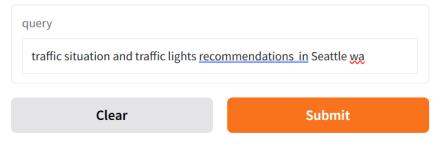
demo_mcp = gr.Interface(
    fn=analyze_traffic_MCP,
    inputs=gr.Textbox(placeholder="Describe the traffic situation or ask for signal adjustments.
    outputs=gr.Markdown(),
    title="AI Traffic Management Assistant",
    description="This app analyzes live traffic data and provides recommendations for adjusting)

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```

Interface Screenshots

Al Traffic Management Assistant

This app analyzes live traffic data and provides recommendations for adjusting signal lights dynamically.



Traffic Signal Adjustment Recommendations

To provide effective recommendations for adjusting traffic signals in Seattle, WA, I would typically analyze live traffic data such as vehicle counts, speeds, congestion levels, and pedestrian activity. However, without real-time data, I can offer general recommendations based on common traffic patterns and issues in urban environments like Seattle.

Common Traffic Patterns in Seattle:

Peak Hours: Morning (7 AM - 9 AM) and evening (4 PM - 6 PM) rush hours typically see increased traffic.

Major Routes: Highways like I-5 and I-90, as well as major arterial roads such as Aurora Ave N and Rainier Ave S, often experience congestion.

Public Transit: Areas with high bus traffic, especially near transit hubs, require careful signal management to facilitate public transport.

Pedestrian Activity: High foot traffic areas, especially near schools, parks, and shopping districts, need signals that prioritize pedestrian safety.

Recommendations for Signal Light Adjustments:

Adaptive Signal Control:

Implement adaptive traffic signal control systems that adjust signal timings based on real-time traffic conditions. This can help reduce congestion during peak hours.

Prioritize Public Transit:

Use transit signal priority (TSP) to give buses and light rail vehicles a green light when they approach intersections, reducing delays and improving service reliability.

Future Enhancements

Scaling and Optimization Options

- 1. Horizontal Scaling with load balancer and Elastic Pool technologies needed to scale out the application to handle millions of requests.
- 2. Vector and Graph DB to optimize AI agent customization per user.
- 3. Use of Data lake/Delta Lake/OneLake technologies for future advanced analytics and AI/ML model experimentation.
- 4. Embedding analytics visualization tools for more enhanced and self-service capabilities.
- 5. Security: for example, Using MFA and data encryption for PII
- 6. CDN: Zone/Region architecture for optimal availability