### **CAPSTONE PROJECT**

### TRAVEL PLANNER AGENT

#### Presented By:

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### **OUTLINE**

- Problem Statement
- Proposed System/Solution
- System Development Approach
- Algorithm & Deployment
- Result
- Conclusion
- Future Scope
- References



## PROBLEM STATEMENT

**Example:** Travel planning involves complex decision-making across multiple variables like destination, budget, weather, duration, and interests. Users often struggle to find and organize relevant information in real-time. The challenge is to create a system that can personalize travel plans dynamically, manage updates, and reduce manual research time.



# PROPOSED SOLUTION

- The system predicts hourly bike demand using machine learning to ensure stable availability in urban areas.
- Data Collection:
  - Collects destination, travel dates, budget, group size, and interests via natural language.
- Data Preprocessing:
  - Uses an LLM (like IBM Granite) with ReAct to understand preferences and generate day-wise itineraries.
- Machine Learning Algorithm:
  - Integrates Google Search to fetch live info on weather, hotels, places to visit, and transport options.
- Deployment:
  - Adjusts plans based on budget or preference changes during the conversation.
- Evaluation:
  - Runs on IBM watsonx.ai using Agent Lab (LangGraph), with optional expansion to custom tools or external APIs.



# SYSTEM APPROACH

- •System Requirements: IBM Cloud Lite account, internet access, watsonx.ai lab
- •Agent Architecture: LLM-based reasoning with dynamic search
- •User Interaction: Chat-based Q&A session to gather preferences
- •External Tools: Google Search tool for dynamic info retrieval
- •Optional Tools: WeatherFetcher, HotelRecommender, TransportPlanner
- Platform: IBM watsonx.ai Agent Lab
- •Framework: LangGraph (ReAct architecture)
- •Model: LLaMA-3-70B-Instruct or IBM Granite (LLM)
- •Tool: Google Search tool for real-time data
- •Languages: Python (for tool development, if used), JSON (tool schema)
- •Optional Integration: Weather APIs, Booking APIs, Maps APIs



# **ALGORITHM & DEPLOYMENT**

- In the Algorithm section, describe the machine learning algorithm chosen for predicting bike counts. Here's an example structure for this section:
- Algorithm Selection:
  - The system uses the ReAct (Reasoning + Acting) framework in combination with a pre-trained Large Language Model (LLM) such as LLaMA-3-70B-Instruct or IBM Granite. ReAct enables the agent to reason through user input and decide when to take actions (e.g., perform a web search). This setup is ideal for conversational, decision-based tasks like itinerary generation, where dynamic and contextual understanding is essential.
- Data Input:
- The input to the system is **natural language queries** from the user, which typically include:
- Destination (e.g., Goa, Manali)
- Duration (e.g., 3-day, 5-day trip)
- Budget (e.g., ₹20,000)
- Travel type (solo, family, couple)
- Interests (e.g., food, adventure, relaxation)
- These inputs are extracted and structured internally by the LLM for planning and decision-making.
- Training Process:
  - Explain how the algorithm is trained using historical data. Highlight any specific considerations or techniques employed, such as cross-validation or hyperparameter tuning.
- Prediction Process:
  - Detail how the trained algorithm makes predictions for future bike counts. Discuss any real-time data inputs considered during the prediction phase.



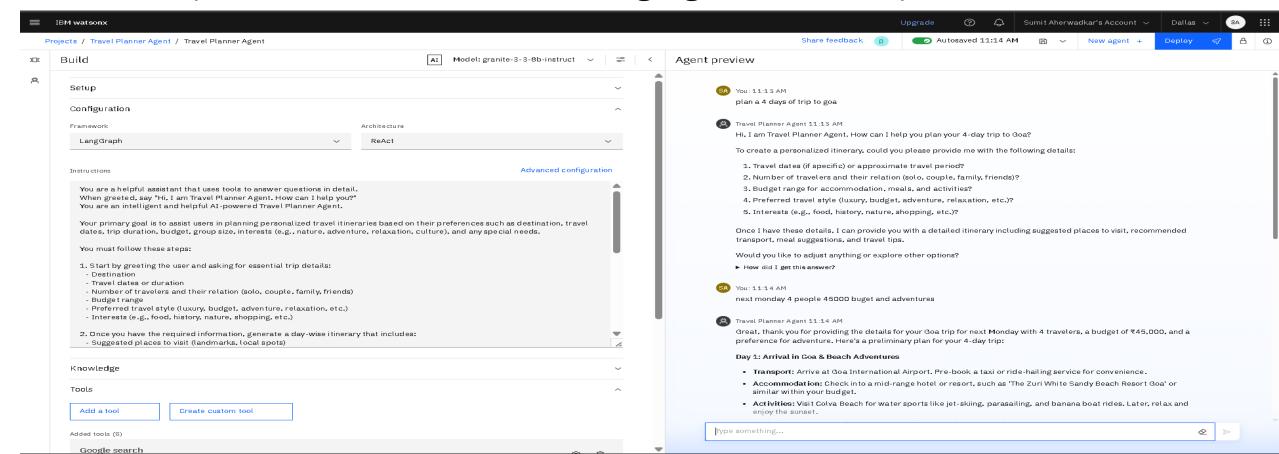
## **ALGORITHM & DEPLOYMENT**

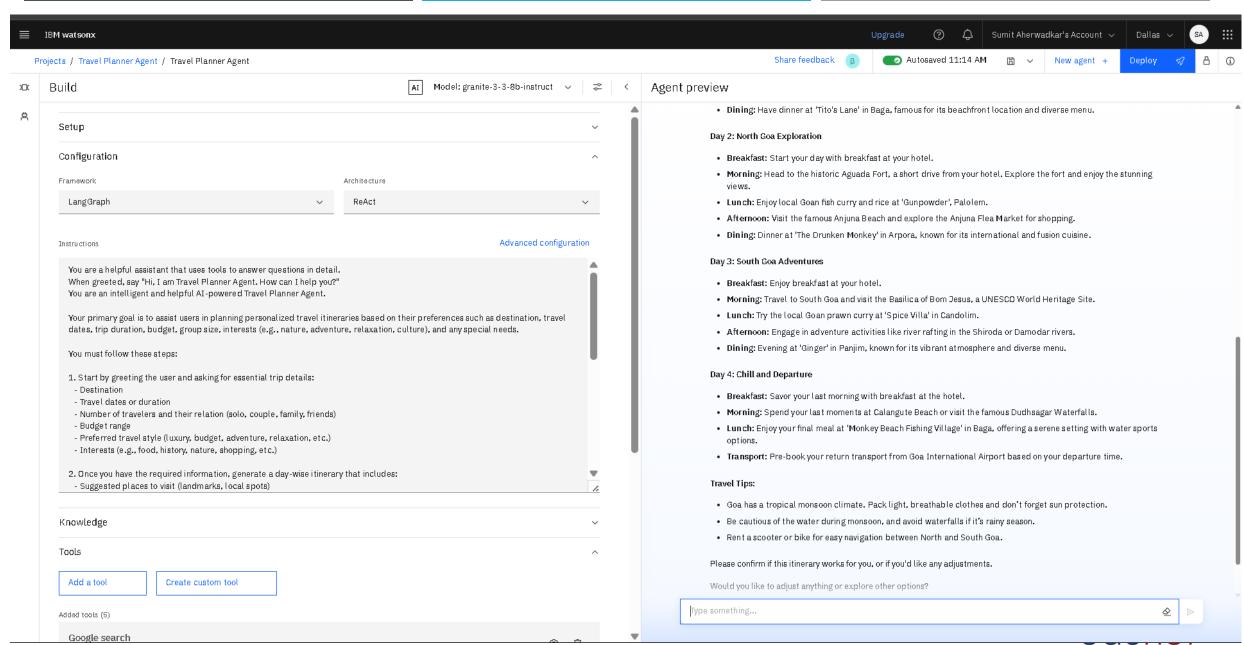
- Training Process:
- The LLM used is pre-trained and not fine-tuned specifically for this task. Instead of training, the system relies on:
- Prompt engineering to guide the model's responses
- LangGraph orchestration to structure reasoning and tool invocation
- Tool selection logic defined within the Agent Lab interface
- This no-training approach enables quick deployment and flexible general-purpose performance without needing large datasets or training infrastructure.
- Prediction Process:
- Upon receiving user input, the agent:
- Parses the message to understand preferences and constraints
- Performs web searches (via the Google Search tool) to fetch real-time weather, hotel, or location info
- Generates a personalized day-wise itinerary using the LLM's reasoning capabilities
- Responds conversationally, summarizing trip details and asking for confirmation or adjustments
- All this is executed in a cloud-based environment (IBM watsonx.ai Agent Lab) with no local installation required.

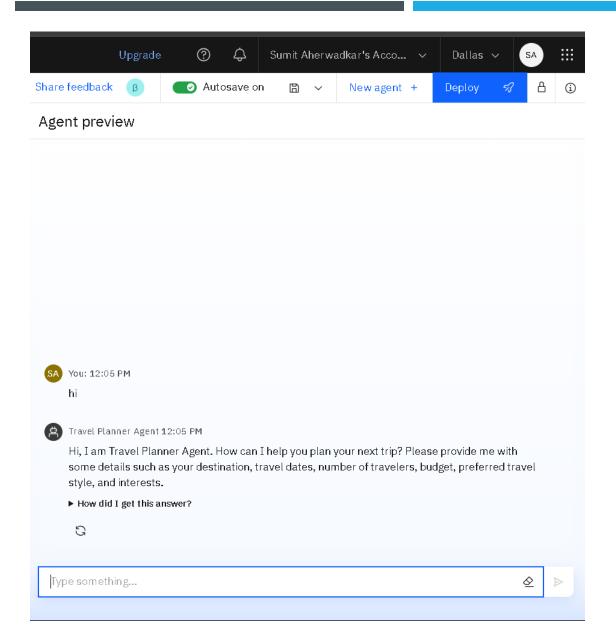


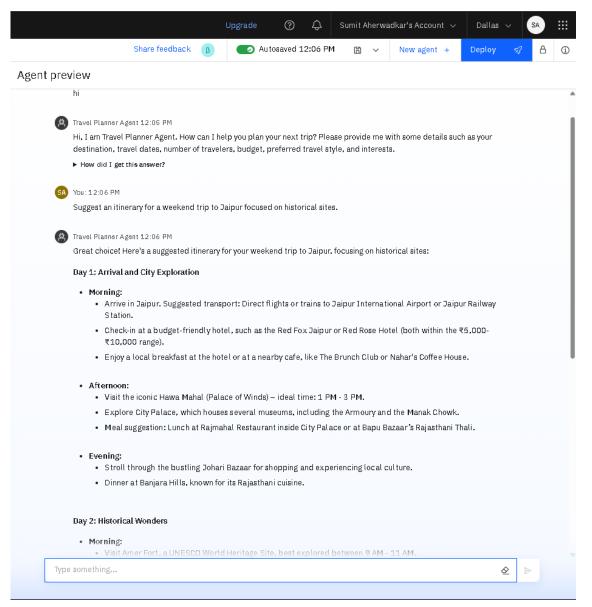
## RESULT

Present the results of the machine learning model in terms of its accuracy and effectiveness in predicting bike counts. Include visualizations and comparisons between predicted and actual counts to highlight the model's performance.

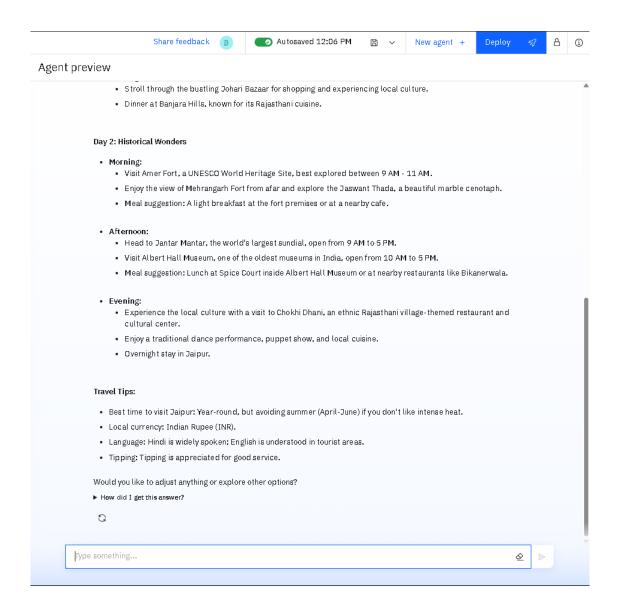


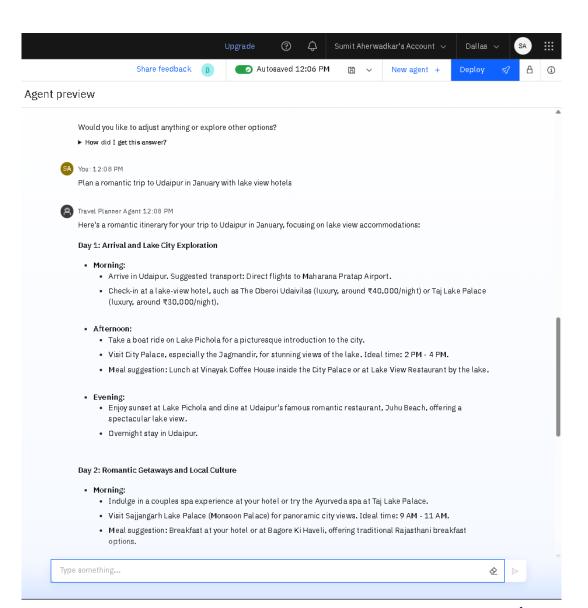




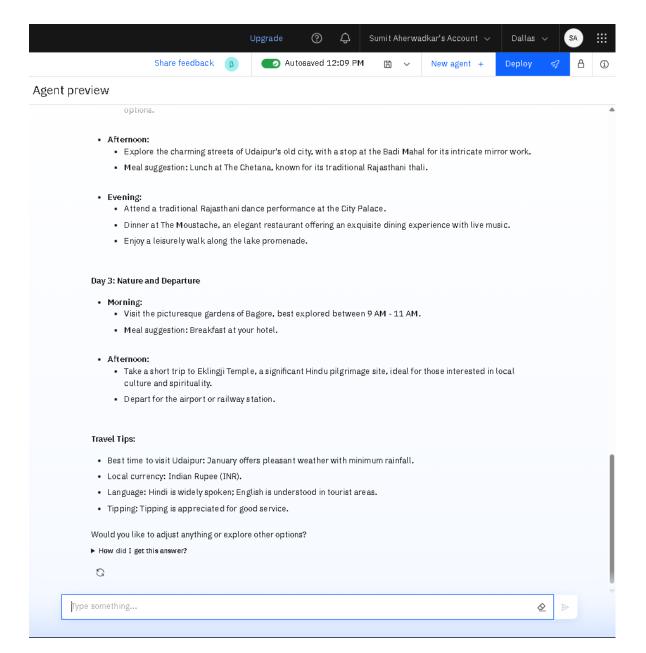














## CONCLUSION

• The Travel Planner Agent successfully demonstrates the ability to generate personalized itineraries using Al and real-time data. It simplifies travel planning by integrating multiple services into a conversational interface, improving user convenience, accuracy, and experience.



### **FUTURE SCOPE**

- •Add real-time weather, hotel, and transport APIs as custom tools
- Expand support for multilingual queries
- •Introduce booking capabilities for transport and accommodation
- •Integrate user account & profile-based personalization
- Deploy as a mobile or web app for broader usability



# REFERENCES

- LangGraph GitHub Repo
- OpenAl / Meta LLaMA-3 Paper
- OpenWeatherMap API
- Google Search and Travel APIs
- IBM Granite & Watson Assistant docs



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(ALM-COURSE\_3824998)

According to the Adobe Learning Manager system of record

Completion date: 23 Jul 2025 (GMT)

Learning hours: 20 mins

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## **THANK YOU**

