SMART INDIA HACKATHON 2024



TITLE PAGE

- Problem Statement ID 1647
- Problem Statement Title -

Development of AI-ML based models for predicting prices of agri-horticultural commodities such as pulses and vegetable

- Theme- Agriculture, FoodTech & Rural Development
- PS Category- Software
- Team ID-
- Team Name (Registered on portal)





IDEA TITLE - ANNAPURANAM



Proposed Solution

- Detailed explanation of the proposed solution Develop a LSTM based machine learning model for price prediction based on inputs:
 - State of production | Commodity type | Time till next harvest
 - Weather inputs like rain and temperature variance since sowing season
- How it addresses the problem -Show an interactive dashboard to monitor strong future fluctuation signals
- Innovation and uniqueness of the solution Allow simple query through platform integrated LLMs that can write/execute python code at the backend.



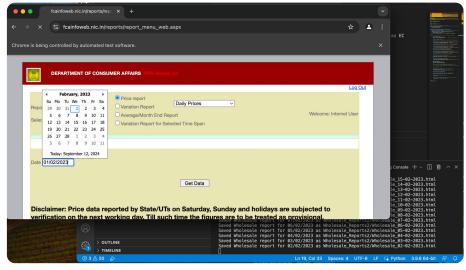
Where is tomato most expensive today?

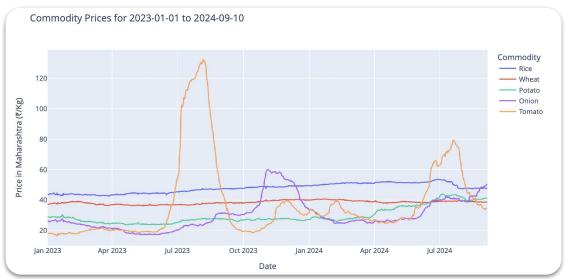




TECHNICAL APPROACH







Technologies to be used

Model Training: TensorFlow for LLM architecture

Visualization: Plotly Dash/Streamlit

LLM Integration: OpenAl

Backend Execution: Flask or FastAPI

Weather Data: OpenWeather API

Database Storage: MongoDB

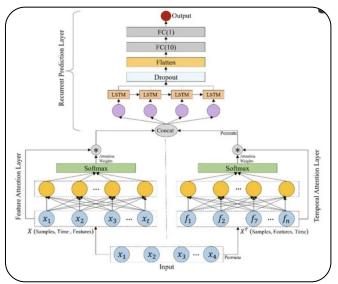
Methodology and process for implementation

- Collect and process relevant data
 (Took 3 hours for 2 year of daily records)
 - Automated scraping through public reports from ministry's website and processed in python.
- 2. Build LSTM model by considering various feature evaluation
- 3. Design interactive dashboard Visualize prediction results and other relevant data
- 4. Incorporate language models Enable user queries



FEASIBILITY AND VIABILITY





Analysis of the feasibility of the idea

- Technical Complexity: LSTM implementation manageable within timeframe. Dashboard creation achievable using Plotly Dash
- Scalability Works for initial prototype scale, using the actual dataset updates in real-time by Ministry.
- LLM Integration LLM API integration straightforward via OpenAI. Simple natural language query support possible

Potential challenges and risks

- O **Technical Complexity**: Fine-tuning the model for accurate predictions within the hackathon timeframe could be challenging.
- LLM Integration: Generating accurate Python code from user queries might not be perfect, leading to incorrect executions.

Strategies for overcoming these challenges

- Technical Complexity: Implement a simplified LSTM model initially, fine-tune after demo.
- LLM Integration: Predefined query templates/prompt to reduce complexity and avoid incorrect outputs.



IMPACT AND BENEFITS





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Potential impact on the target audience

- Policy Makers: Get real-time analytics and forecasting for more effective interventions in commodity markets.
- Consumers: Get stabilized prices and ensured better supply chain management.
- Farmers: Provides actionable insights into price trends, helping them make informed decisions on when to sell or hold crops.

Benefits of the solution

- Social: Reducing income inequality and promoting fair pricing.
- Economic: Improves market efficiency, reduces price volatility.
- Environmental: Encourages sustainable farming practices by optimizing harvest timing.
- Technological: Leads to greater adoption of AI and machine learning in farming.
- Consumer Protection: Helps stabilize commodity prices, potentially reducing inflation and improving food security.



RESEARCH AND REFERENCES



- Details / Links of the reference and research work
 - Prices Data: https://fcainfoweb.nic.in/reports/report_menu_web.aspx
 - Rainfall Data: https://mausam.imd.gov.in/responsive/rainfallinformation.php
 - Dashboard Samples: https://streamlit.io/gallery
 - Assistant based: https://streamly.streamlit.app/
 - Forecasting Agricultural Commodity Prices Using Dual Input Attention LSTM: https://www.mdpi.com/2077-0472/12/2/256
 - Icons and elements: <u>www.flaticon.com</u>