

# Farmer's Items Market Price Forecasting using Machine Learning

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## Abstract

Agriculture continues to be the foundation of numerous developing economies, particularly in nations like India, where a significant portion of the population depends on farming for their sustenance. A primary challenge confronting farmers today is the volatility of market prices for their agricultural products. Price variations resulting from seasonal trends, discrepancies in demand and supply, delays in transportation, natural disasters, and inflation can cause considerable financial setbacks for farmers, especially for small and marginal ones who do not have access to timely and precise price information. This leads to ineffective decision-making regarding the optimal timing and location for selling their produce, often compelling them to sell at low or unjust prices. To tackle this pressing issue, the project introduces a Machine Learning-based forecasting system designed to predict future market prices for various agricultural commodities. By examining historical price data alongside key factors such as crop seasonality, geographic location, climate conditions, and demand trends, the system develops predictive models that can estimate near-future price ranges with a fair degree of accuracy. Machine learning techniques, including Linear Regression, Random Forest, and LSTM (Long Short-Term Memory), are employed to identify patterns and correlations within the data. The primary aim of this system is to equip farmers with actionable insights that can assist them in making well-informed decisions—whether to sell immediately, wait for a price increase, or store their produce. The solution features an intuitive dashboard where farmers can choose crop types, regions, and forecast timeframes to visualize forthcoming price trends.

## Introduction

Agriculture is a crucial sector in the Indian economy, sustaining nearly 60% of the nation's population. A significant challenge that farmers encounter today is the high price volatility of agricultural commodities in the marketplace. These abrupt fluctuations are affected by various unpredictable factors, including weather conditions, crop diseases, transportation issues, seasonal changes, demand-supply imbalances, and evolving market trends. As a result of this uncertainty, farmers frequently struggle to

obtain fair prices for In numerous rural areas, farmers rely on middlemen or local agents to market their goods. These intermediaries often exploit the farmers' lack of price awareness, leading to exploitation and a loss of potential profits. To address these issues, a technology-driven approach is essential, equipping farmers with timely and precise information regarding market prices.

This initiative, "Farmers' Items Market Price Forecasting Using Machine Learning," seeks to develop an intelligent forecasting system that assists in predicting future prices of agricultural

commodities. By utilizing advanced machine learning algorithms on historical market data, the system can produce accurate price forecasts for various crops, including rice, wheat, onions, tomatoes, and more. It will consider factors such as location, season, crop type, weather data, and historical pricing trends.

The primary objective of this project is to aid farmers in making data-informed decisions about when and where to sell their products to maximize their returns. With reliable price predictions, farmers can mitigate uncertainty, avoid distress selling, enhance their storage and transportation strategies, and ultimately bolster their financial stability. their products, which directly impacts their income and livelihood.

## Literature Survey Existing System

In the present agricultural economy, farmers frequently find themselves vulnerable to erratic market prices. The conventional systems for forecasting market prices are either manual, relying on insufficient data, or dependent on government advisories that are often delayed and region-specific. These systems predominantly neglect to account for real-time factors such as variations in demand and supply, weather conditions, logistics, consumer trends, and global commodity influences. Numerous farmers continue to rely on intermediaries who manipulate prices, resulting in considerable income disparity and a lack of transparency in the price-setting process.

The current systems typically employ linear regression or basic trend analysis, which are

inadequate for managing non-linear patterns in time-series agricultural data. Additionally, there is a significant lack of integration among real-time data sources, including mandi prices, climate statistics, transportation delays, and economic indicators like fuel prices or export-import data. Moreover, most existing platforms are not user-friendly for farmers, particularly those in rural areas with limited access to technology or education.

Another drawback of the current systems is the lack of tailored predictions for various commodities across different regions. A uniform model does not effectively address the agricultural sector, where pricing trends differ considerably for each product based on location, soil conditions, local demand, seasonal cycles, and regional competition. These deficiencies highlight the need for a more sophisticated, data-driven approach to forecasting.

## Problem Statement

Agriculture continues to be the foundation of numerous developing economies, especially in nations such as India, where a considerable segment of the population relies on farming for their income. Nevertheless, farmers frequently encounter significant income volatility due to unpredictable changes in market prices for their agricultural products. These price fluctuations are affected by various factors, including climatic conditions, imbalances in supply and demand, transportation expenses, interference from middlemen, and government regulations. Consequently, farmers often find themselves uncertain about the optimal timing to sell their

crops for the highest profit. This uncertainty results in distress sales, post-harvest losses, and diminished bargaining power. There is an urgent requirement for a data-driven approach that can equip farmers with dependable price forecasts, enabling them to make well-informed selling choices and achieve improved financial results.

## Proposed System

The proposed system introduces a robust Machine Learning-based platform that forecasts market prices of various agricultural products (such as rice, wheat, vegetables, fruits, etc.) with higher accuracy by leveraging multi-dimensional datasets. Unlike conventional systems, this model is dynamic, scalable, and capable of adapting to changes in historical trends, real-time market fluctuations, and contextual variables like weather and location. This system incorporates supervised machine learning models like Random Forest, XGBoost, and Long Short-Term Memory (LSTM) for time-series forecasting. These models have been proven to handle large and complex datasets, learn patterns over time, and offer accurate future predictions. The data pipeline will include preprocessing steps like missing value imputation, outlier detection, seasonal decomposition, and normalization to prepare the data for model training. Furthermore, the system integrates real-time data from APMC markets, Agmarknet, weather APIs, and historical mandi price datasets, enhancing prediction precision. The model is trained and evaluated on these datasets, and it outputs region-specific, product-wise price trends. The system also aims to be accessible via a mobile app or web interface where farmers can select a commodity, input location, and receive future price

forecasts in a visual and simple format. This empowers them to make strategic decisions regarding crop selling time, storage, and market selection.

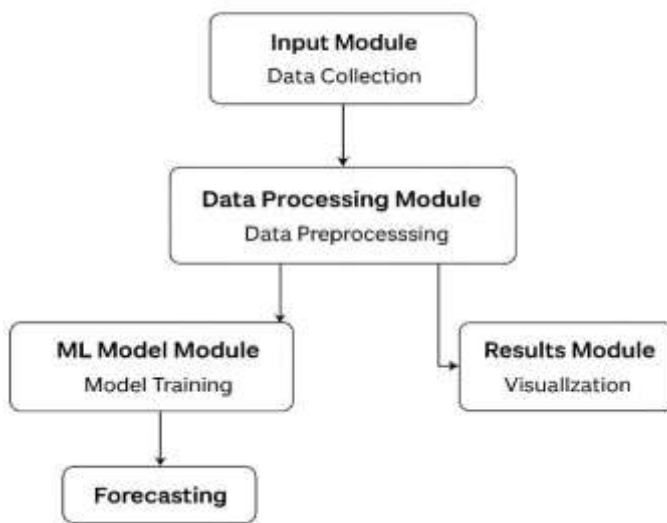
## System Requirements Specifications

### Functional Requirements

The system is required to gather and retain historical market prices for a variety of agricultural products.

- Users should have the capability to enter crop names, market locations, and preferred date ranges.
- The system is expected to preprocess the data by addressing missing entries, performing normalization, and applying encoding techniques.
- It must facilitate the training of machine learning models and forecasting based on seasonal trends and location-specific data.
- Users ought to be able to access future price forecasts through tables and interactive graphs.
- The system should enable the export of forecast data in downloadable formats such as PDF or CSV files.
- It is essential to keep a record of previous predictions for the purpose of analysis and informed decision-making.
- The administrative panel must offer tools for model monitoring, user management, and data validation.

## Architecture Diagram



## Architecture Review

### • Input Module

- Responsible for gathering raw data from diverse sources such as APMC markets, governmental records, and online commodity platforms.
- The data encompasses crop prices, weather conditions, seasons, locations, and demand trends.

### • Data Processing Module

- Prepares the collected data by rectifying missing values, addressing outliers, encoding categorical variables, and normalizing price ranges.
- Transforms raw data into a structured format that is appropriate for training machine learning models.

### • ML Model Module

- Utilizes the cleaned data to train machine learning models including Linear

Regression, Random Forest, or LSTM.

- Understands the correlation between various features and market prices to facilitate forecasting.

### • Forecasting

- Implements the trained model to estimate future prices of agricultural products.
- Assists farmers in making informed choices regarding the optimal timing for selling or storing their produce.

### • Results Module

- Displays the forecasted outcomes through charts, tables, and graphs.
- Enables users to choose specific crops, districts, and time frames for enhanced insights.

## Implementation



The recommendation system was implemented using Python, leveraging its extensive libraries for data processing and machine learning. Initially, the dataset containing user preferences and item details was collected and thoroughly preprocessed by removing duplicates, handling missing values, and

encoding categorical variables where necessary. Once the data was cleaned, feature extraction techniques were applied to identify meaningful patterns. For building the recommendation model, both content-based filtering and collaborative filtering approaches were explored. Content-based filtering analyzed the item attributes to suggest similar products, while collaborative filtering utilized user-item interaction data to predict preferences by identifying similarities between users or items. The model's performance was assessed using evaluation metrics such as Root Mean Square Error (RMSE) and Mean Absolute Error (MAE), ensuring its accuracy and reliability. Finally, the system was integrated into a simple user interface, allowing users to receive personalized and accurate recommendations based on their historical interactions and preferences.

## Conclusion

In this project, *Farmer's Items Market Price Forecasting using Machine Learning*, we addressed one of the most critical challenges faced by farmers: the unpredictability of market prices for agricultural commodities. By developing a robust machine learning-based forecasting system, we provided a practical solution that empowers farmers to make informed decisions regarding the optimal time and place to sell their produce, ultimately improving their financial stability and reducing dependency on intermediaries.

The system leverages advanced algorithms such as Random Forest, XGBoost, and LSTM to analyze multi-dimensional data including historical prices, weather conditions, seasonal trends, and geographic variations. Through effective data

preprocessing, real-time data integration, and accurate forecasting models, the system delivers precise price predictions tailored to specific crops and regions.

The user-friendly interface ensures accessibility for farmers, enabling them to easily retrieve forecast information, plan their sales strategies, and avoid losses due to unfavorable market conditions. Moreover, by minimizing the role of middlemen and enhancing market transparency, the system contributes to a more equitable agricultural economy.

Overall, this machine learning-based forecasting platform holds great promise for transforming the agricultural sector, promoting financial security for farmers, and encouraging the adoption of technology-driven solutions in rural communities.

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