

Walmart Project

Capstone Project 1

Table of Contents

- 1. Problem Statement
- 2. Project Objective
- 3. Data Description
- 4. Data Pre-processing Steps and Inspiration
- 5. Choosing the Algorithm for the Project
- 6. Motivation and Reasons for choosing the Algorithm
- 7. Assumptions
- 8. Model Evaluation and Techniques
- 9. Inferences from the same.
- 10. Future Possibilities of the Project
- 11. Conclusion
- 12. Reference

WALMART PROJECT GOKUL VENUGOPAL PAGE

Problem Statement

A Retail Company with multiple outlets stores is having poor revenue returns from retail store with most of them facing bankruptcy. This project undertakes to review sales records from the stores with a view to provide useful insights to the company and also to forecast sales outlook for the next 12-weeks.

Project Objective

A Retail Company with multiple outlets across the country are facing issues with inventory management. The task is to come up with useful insights to the company and also to forecast sales outlook for the next 12-weeks.

Data Description

The walmart.csv contains 6435 rows and 8 columns.

Feature Name	Description Store number	
Store		
Date	Week of Sales	
Weekly_Sales	Sales for the given store in that week	
Holiday_Flag	If it is a holiday week	
Temperature	Temperature on the day of the sale	
Fuel_Price	Cost of the fuel in the region	
CPI	Consumer Price Index	
Unemployment	Unemployment Rate	

From the given data set of the company, it is observed that the data consist of Six thousand four hundred and thirty- five (6345) records with seven features (captured weekly) as follows.

Stores: There are 45 stores and each store has 143 entries with below information.

- Date (Weekly)
- Total sales record for the week
- Holiday Flag (1 or 0)
- Temperature: Average Temperature recorded during the week.
- Fuel Price: Average price of Fuel for the week
- CPI: Consumer Price Index for the week.
- Unemployment: Rate of Unemployment for the week

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Data Preprocessing Steps and Inspiration

The Preprocessing steps included the following steps:

Step 1: Load Data

Step 2: Perform Exploratory Data Analysis

- a. Check number of records and its distribution
- b. Check Data types
- c. Check for missing data, invalid entries and duplicates
- d. Examine the correlation of the independent features with target (weekly sales) variable.
- e. Check for outliers that are known to distort

Step 3: Model Predictions, two approaches of Time series model

a. ARIMA

b. SARIMAX

Step 4: Forecast

Step 5: Compare Results from different model.

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Choosing the Algorithm for the Project

Model Selection

Examination of the plot of weekly sales shows continuously time varying data (as shown above). A Time series model (ARIMA, SARIMAX) will be employed for the prediction and forecast.

ARIMA

An Autoregressive Integrated moving average (ARIMA) model is a generalization of an autoregressive moving average (ARMA) model. To better comprehend the data or to forecast upcoming series points, both of these models are fitted to time series data.

Assumptions on which ARIMA model is based:

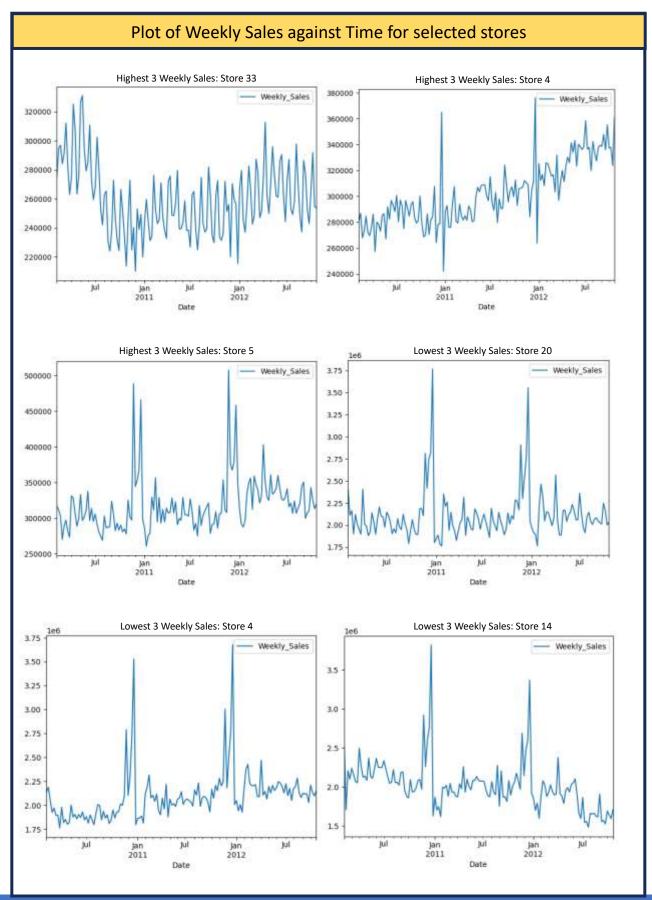
- 1. Data does not contain anomalies
- 2. Model Parameter and error term is constant
- 3. Historic timepoints dictate behaviour of present timepoints
- 4. Time series is stationary.

SARIMAX

SARIMAX (Seasonal Auto-Regressive Integrated Moving Average with exogenous factors) is an updated version of the ARIMA model. we can say SARIMAX is a seasonal equivalent model like SARIMA and Auto ARIMA. it can also deal with external effects.

WALMART PROJECT GOKUL VENUGOPAL PAGE

Model Technique and Evaluation

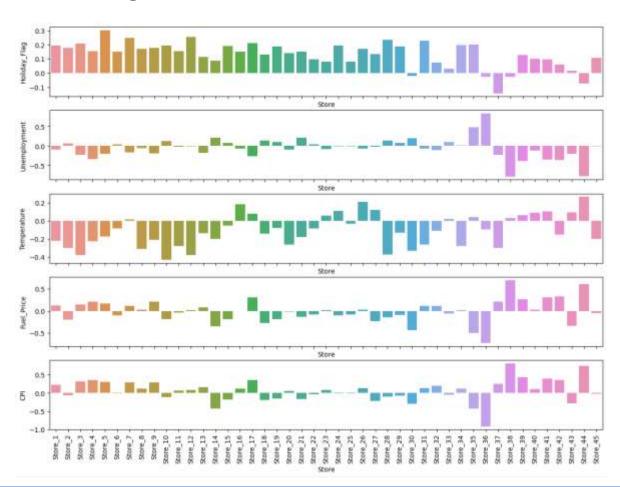


Model Evaluation

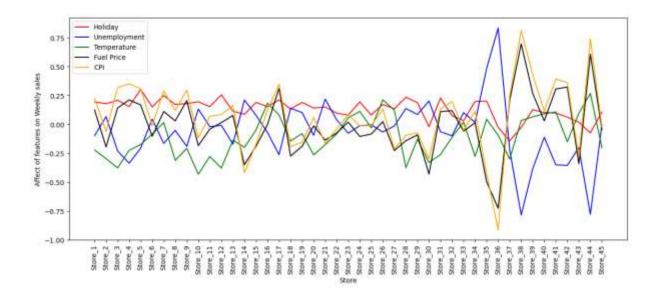
The following techniques and steps were involved in the evaluation of the model.

- 1. Load necessary Libraries
- 2. Load the dataset
- 3. Perform Exploratory Data Analysis (EDA) on the dataset.
 - a. Find the shape and size of the data
 - b. Check for invalid and null entries
 - c. Explore Data description
- d. Examine the correlation of independent variables to the target variable (Weekly sales)
- e. Line plot of the effects of independent variables to the target variable (Weekly sales)
- 4. Model Prediction
- 5. Forecast

Model Design



It was observed from the EDA that the effects of the Independent Features on the target variable. Some stores are affected greatly by unemployment, while some affected by Fuel price. This type of fluctuating data will be very hard to build a general model for all stores. As seen in above graph, only 4 features namely had good correlation (Threshold > +/- 0.2) for some of the stores. The 4 features are Unemployment, Fuel Price, CPI and Temperature. Also, CPI was observed to be closely following Fuel price which indicates same effect on the weekly sales by both the features. But both these are uncontrollable factor so cannot be taken into consideration for improving sales. So, 2 features i.e., Temperature and Holiday will be taken into consideration for insights in each store.



Based on the findings above, the decision was to take only 4 features i.e., Temperature, Unemployment, Fuel Price and Holiday will be taken into consideration for insights in each store.

For simplicity and ease of presentation, I have decided to take to limit my predictions from for Top 3 and bottom 3 stores based on weekly sales revenue. That notwithstanding, the model could be always be used to provide prediction for each store.

Model Approach

Timeseries Model, ARIMA

- a. First step is to check the stationarity of the dataset using Adfuller test and KPSS test. Also using Visual method to confirm the stationarity
- b. Next is to plot ACF and PACF graph for finding the orders (p, d, q) manually.
- c. To find the best ARIMA orders, we will be using AutoARIMA to get the best orders.
- d. SARIMAX will be used if seasonality is found the dataset of each respective stores.
- e. Predict using SARIMAX and ARIMA and based on lowest error select the best model for the respective store.
- f. Forecast 12-week sales using the best model.

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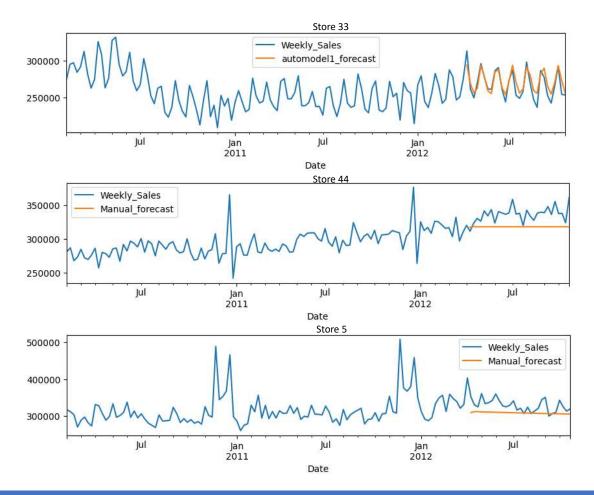
Inferences from the Project

1. ARIMA Model:

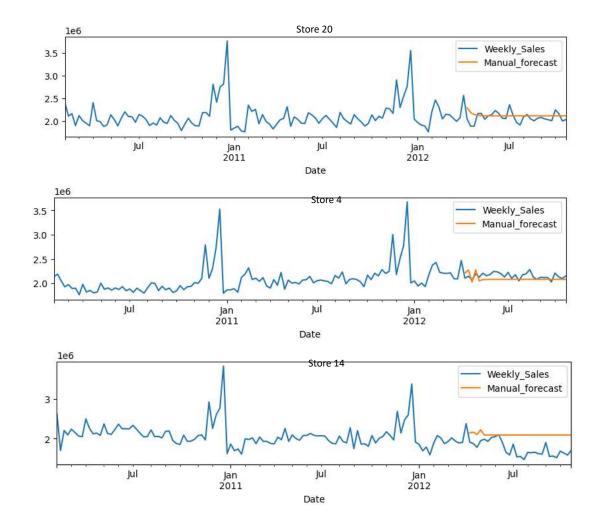
a. Predictions

Predictions were performed for 6 stores (Top 3 [33, 44, 5] and Bottom 3 [20, 4, 14]in order of decreasing Weekly sales) using ARIMA. The predictions are summarized in the Table and Graphs below.

Store No	ARIMA			
	MAE	MAPE	RMSE	
33	7592.86	0.0289	9434.7	
44	19003.03	0.055	21476.02	
5	27928.59	0.082	30236.67	
20	97959.87	0.0479	125459.2	
4	77797.49	0.0356	95336.36	
14	356805.3	0.216	396174.1	

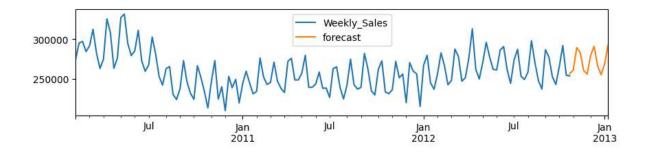


WALMART PROJECT GOKUL VENUGOPAL PAGE



ARIMA was performing poorly for most of stores expect for Store 33 showing that seasonality, trend and noise is affecting the actual forecast which cannot be handled by simple ARIMA model. Due to this SARIMAX was also used to make better fit model for the timeseries data.

b. ForecastForecasting for Store 33 using ARIMA model

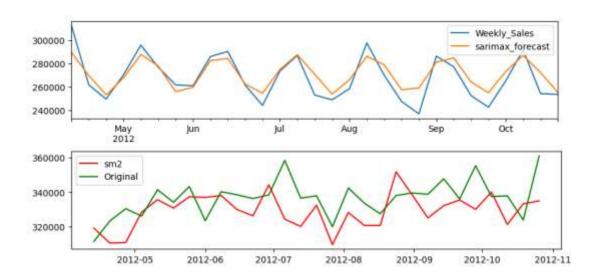


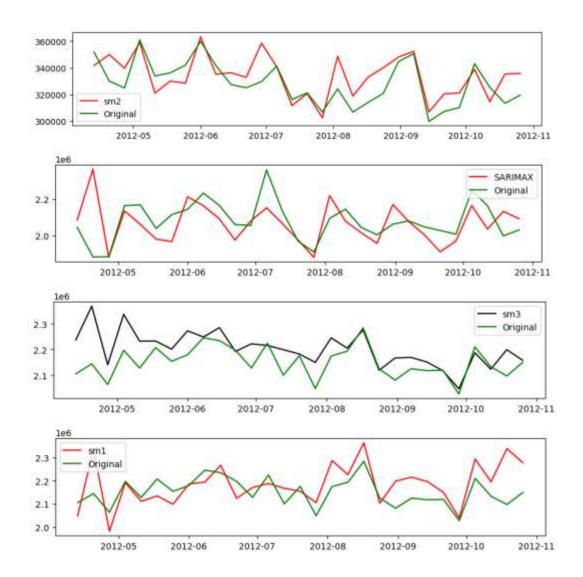
2. SARIMAX Model:

a. Predictions

Predictions were performed for 6 stores (Top 3 [33, 44, 5] and Bottom 3 [20, 4, 14]in order of decreasing Weekly sales) using SARIMAX. The predictions are summarized in the Table and Graphs below

Store No	SARIMAX			
	MAE	МАРЕ	RMSE	
33	7848.55	0.0296	9907.41	
44	11143.8	0.0327	13658.97	
5	11543.12	0.0356	13232.41	
20	84831.59	0.041	121991.9	
4	56267.09	0.02639	77335.1	
14	92897.16	0.043	107091.3	



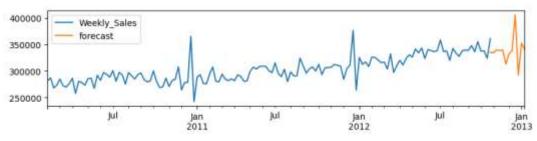


As seen the SARIMAX performs better as compared to ARIMA model and it is clearly evident using the different metrics like RMSE (Root Mean Squared Error), MAPE (Mean Absolute Percentage Error), MAE (Mean Absolute Error).

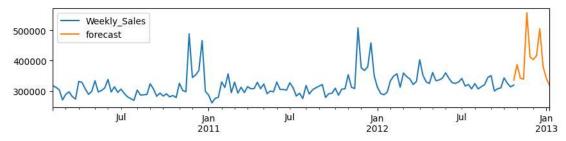
b. Forecast

Forecasting for the remaining 5 stores using SARIMAX model are given below:

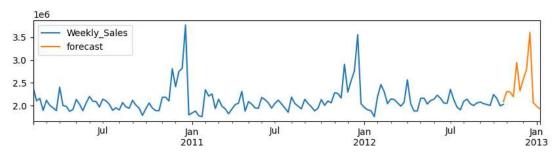
Forecasting for Store 44 using ARIMA model



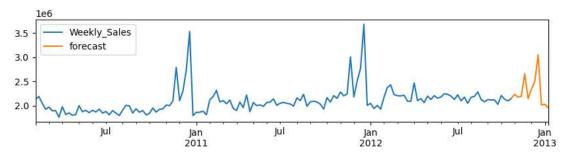
Forecasting for Store 5 using SARIMAX model



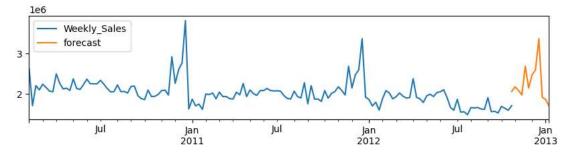
Forecasting for Store 20 using SARIMAX model



Forecasting for Store 4 using SARIMAX model

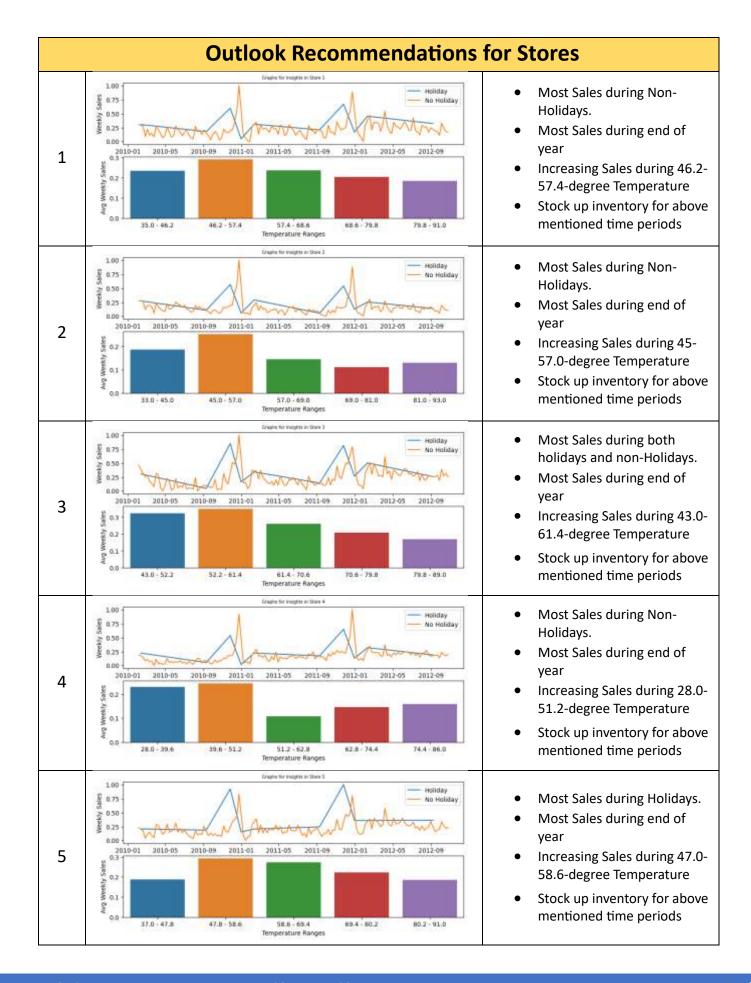


Forecasting for Store 14 using SARIMAX model

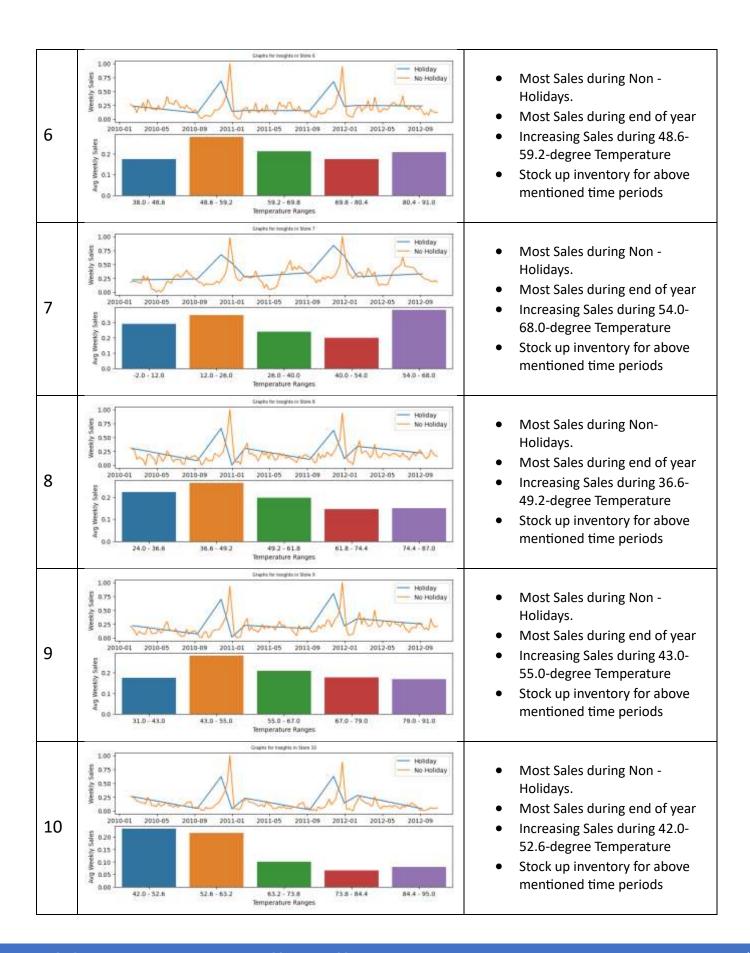


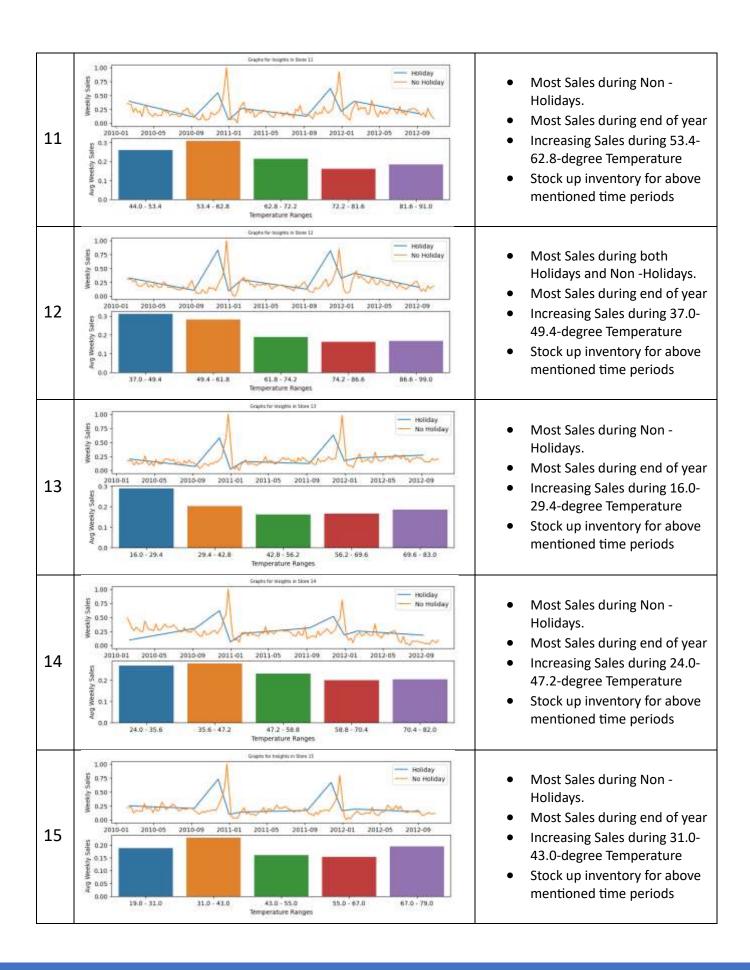
The forecast shows anticipated variabilities as observed in the Weekly sales history. However, overall current sales for stores studied so far will not be enough to save the company from bankruptcy.

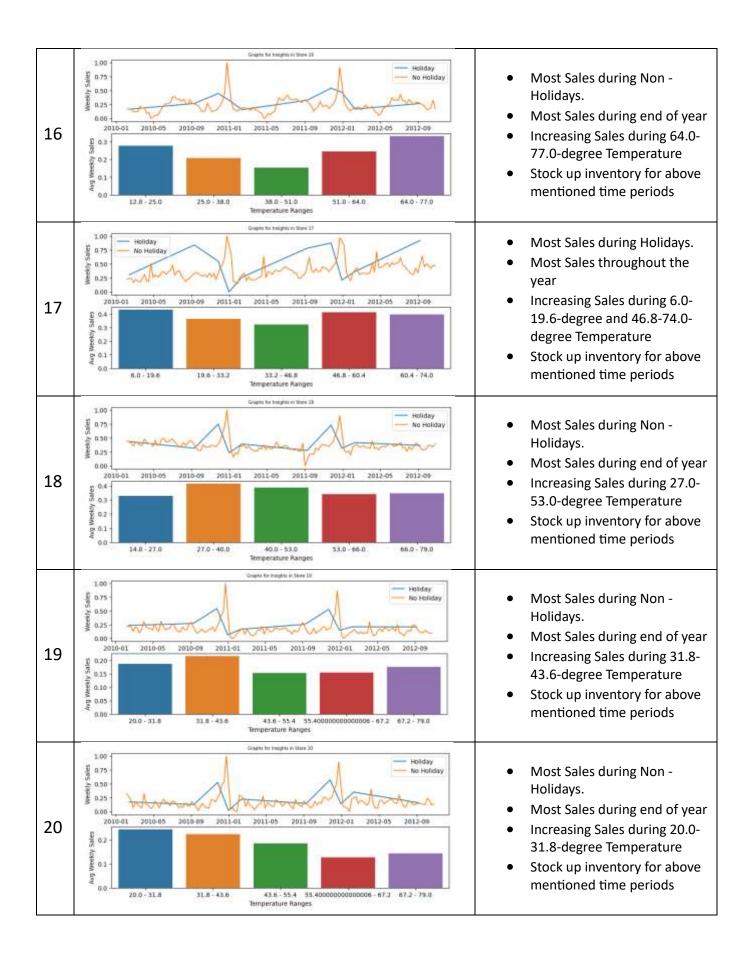
WALMART PROJECT GOKUL VENUGOPAL PAGE | 1

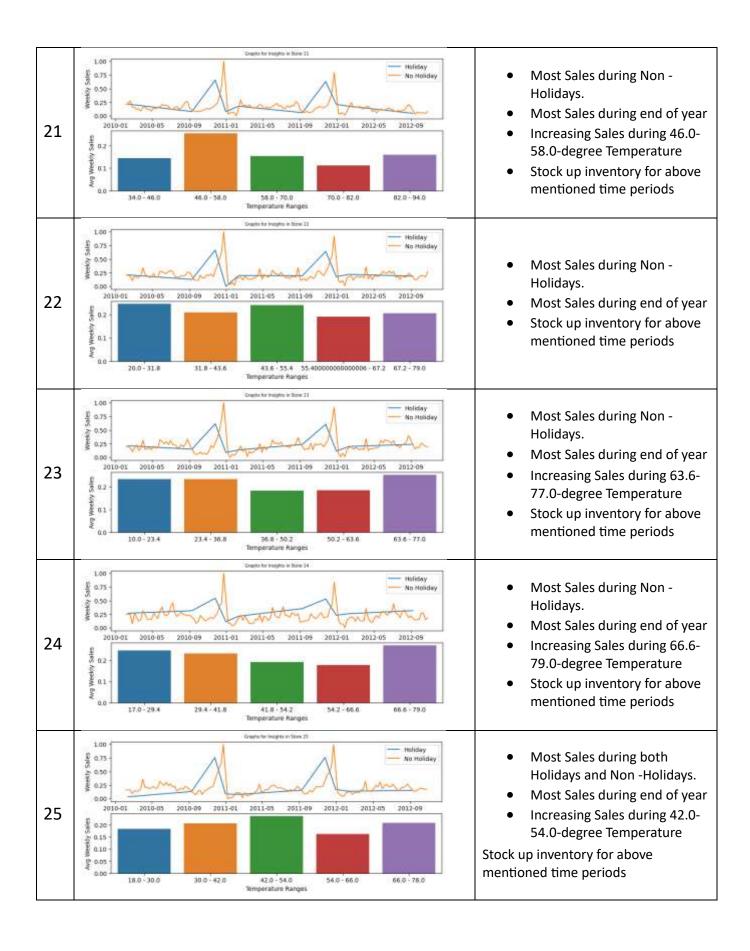


WALMART PROJECT GOKUL VENUGOPAL PAGE | 1

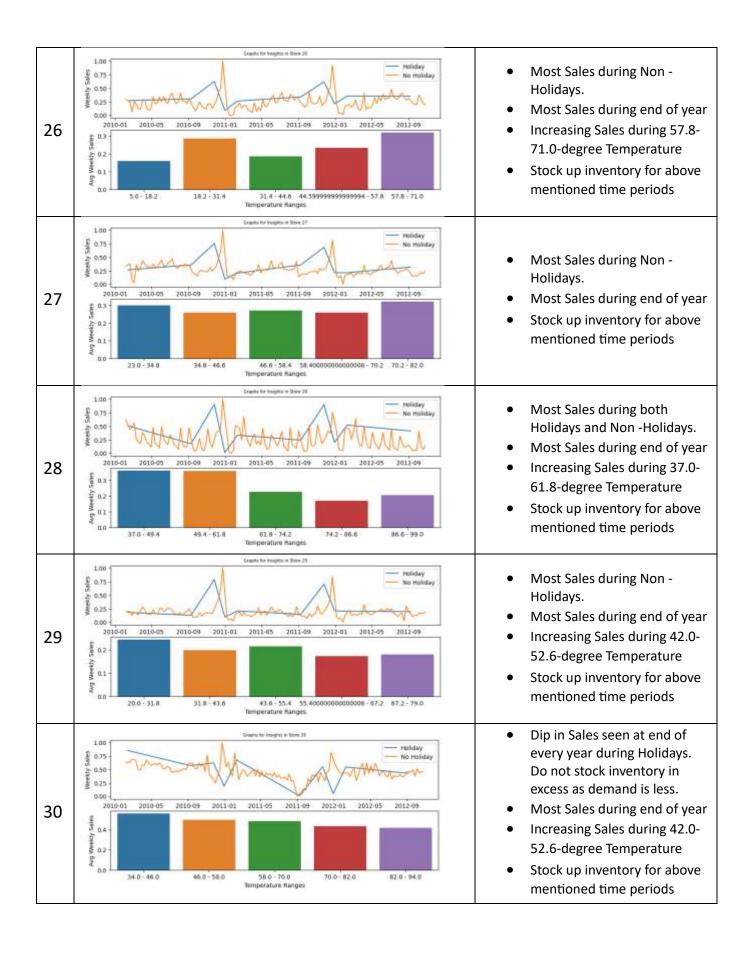


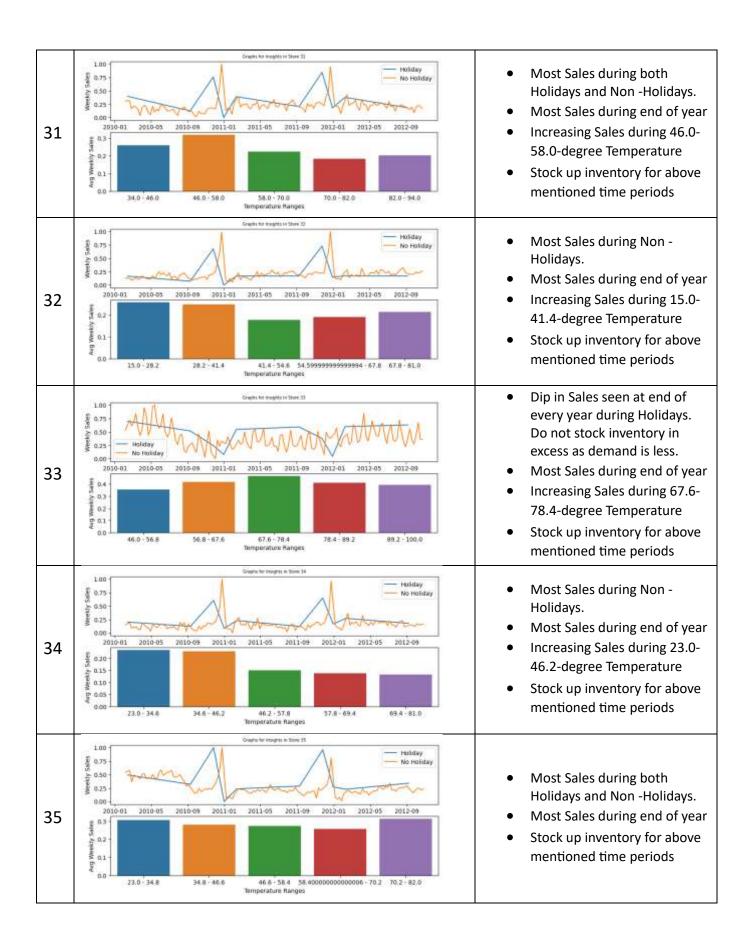


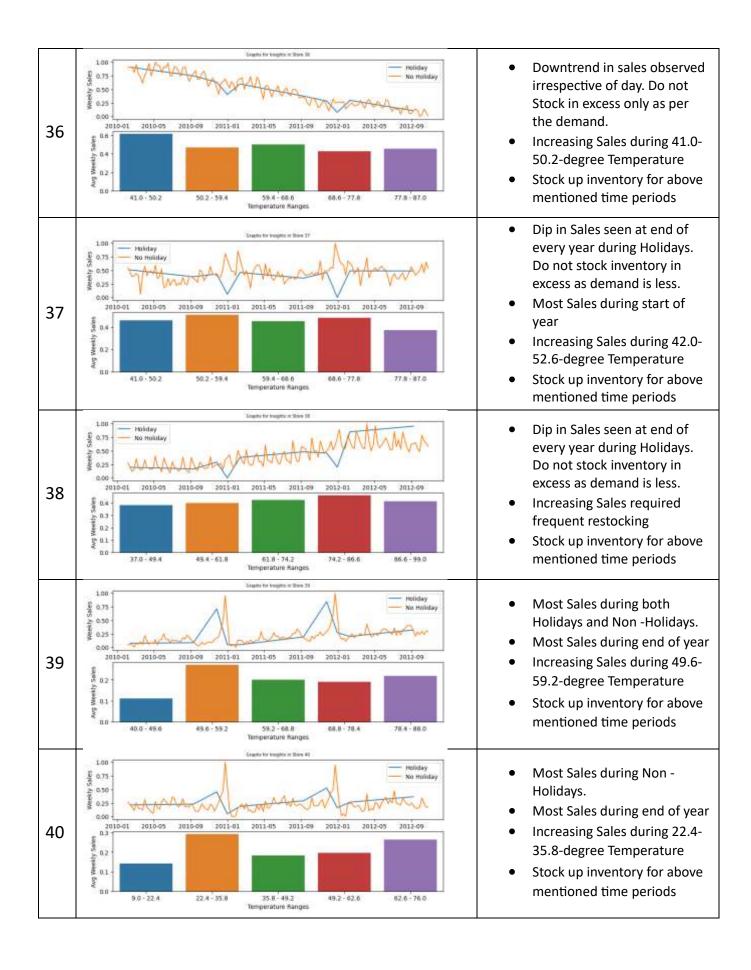


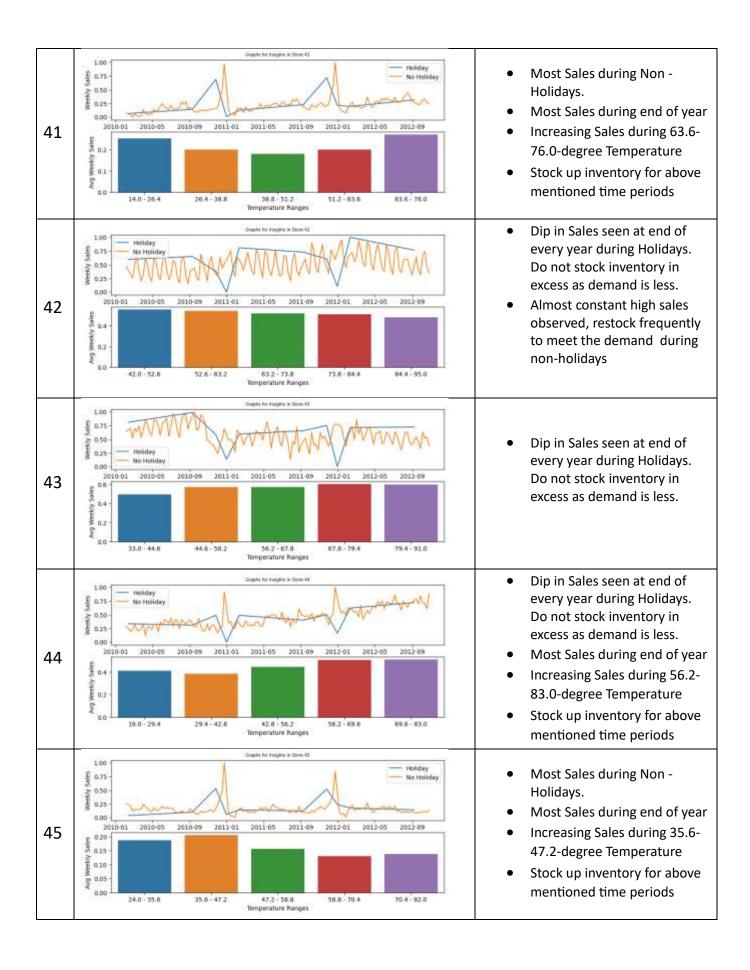


WALMART PROJECT GOKUL VENUGOPAL FAGE | 20









Future Possibilities

Machine learning has shown significant promise in the field of sales prediction and forecasting. By analysing historical data, identifying patterns, and making predictions, machine learning algorithms can help businesses optimize their sales strategies, inventory management, and overall decision-making process.

It's important to note that successful implementation of machine learning for sales prediction requires quality data, appropriate algorithm selection, and continuous refinement of models. Additionally, domain expertise is crucial to interpret the predictions accurately and make informed business decisions.

As technology continues to advance, machine learning's role in sales prediction is likely to become even more sophisticated and integral to sales strategy development.

Conclusion

The project undertook a study of retail company with 45 outlets store with 2 years' worth of data for each store. Some important findings form the report include the following.

- Sales revenue projection for the next 12 weeks down for most of the stores
- Some stores have very weak or sales activities during some period of the year.
- 3. To improve the sales revenue, the following steps are recommended:
 - a. Develop targeted marketing campaigns for each store's local audience, considering factors like local culture and preferences
 - b. Concerted efforts by the company to find out through local market surveys and past sales record what products are in high demand by the local population at any given period of year and make efforts to replenish stocks.
 - c. Offer competitive prices while maintaining profit margins.

 Implement dynamic pricing strategies to adjust prices based on demand and competition through commercial outreach, social media, can improve sales
 - d. Identify slow-moving products and implement strategies to reduce excess inventory and minimize losses.
 - e. Establish an online presence to cater to customers who prefer online shopping. Offer features like online ordering, in-store pickup, and seamless online-offline experiences.
- 4. Collect feedback from customers is key step to identify pain points and areas for improvement.
- 5. Each store only has 2-year data collecting even more data can help to study sales patter even better and plan effectively.
- 6. Some Stores have to be shutdown if the sales revenue does not improve.

WALMART PROJECT GOKUL VENUGOPAL PAGE | 2

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- 2. https://pandas.pydata.org/docs/reference/api/pandas.tseries.offsets.Dateoffset.html
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WALMART PROJECT GOKUL VENUGOPAL PAGE | 2