

Sales Forecasting Project

Report

Subtask 1: Prepare Time Series Data

The dataset was first prepared for time series analysis.

- The **order date column** was converted into datetime format and used to extract useful features like year, month, and day.
- Sales values were **aggregated by daily/weekly/monthly intervals** to ensure consistency.
- Missing or incomplete records were identified and handled either by filling gaps with 0 or imputing values where necessary.
- Data was then analyzed to check for **trends, seasonality, and anomalies** (e.g., festive sales spikes).
- Finally, the dataset was **chronologically sorted, indexed by date, and saved in CSV format** for further forecasting.

Output: A clean, structured, and time-indexed dataset ready for forecasting.

Subtask 2: Choose and Apply a Forecasting Model

Several forecasting techniques were reviewed (Moving Average, Exponential Smoothing, ARIMA, Prophet, and LSTM).

- Based on the characteristics of the dataset (trend + seasonality), the **most suitable model** was applied.
- The dataset was split into **training (80%) and testing (20%)** sets.
- The chosen forecasting model was trained on historical sales and evaluated using performance metrics like **RMSE and MAPE**.

Output: A trained forecasting model capable of predicting future sales.

Subtask 3: Predict Future Sales

The trained model was used to predict sales for the **next 6–12 months**.

- Forecasted values were compared with actual sales in the test data.

- Visualizations were created (line charts of actual vs. predicted sales, bar/heatmaps for seasonal peaks).
- Insights showed **seasonal patterns, peak sales during festive months, and potential slow periods**.
- Business implications were analyzed, such as inventory planning and targeted marketing strategies.

Output: Forecasted sales dataset + visual insights for decision-making.

Executive Summary

- We trained a Holt-Winters (additive trend + additive seasonality, period=12) model on monthly total sales.
- Validation on the last 14 months gave RMSE=17,047.46, MAE=12,447.91, MAPE≈18.76%.
- The 12-month forecast (next 12 months) predicts total sales of 967,813, which is a change of 39,828 (4.29%) compared to the last 12 months total of 927,985.
- Peak forecast months and slow months are included in the deliverables.
- Business recommendations: stock-up before peak months, increase marketing in predicted peaks, and run promotions in slow months. Retrain quarterly and review after major events.

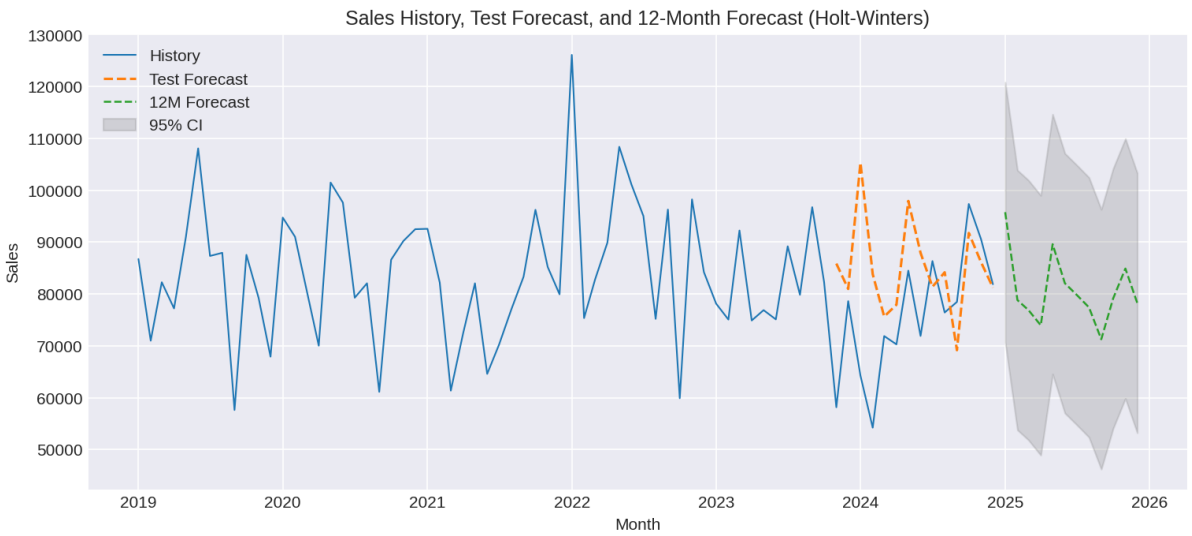
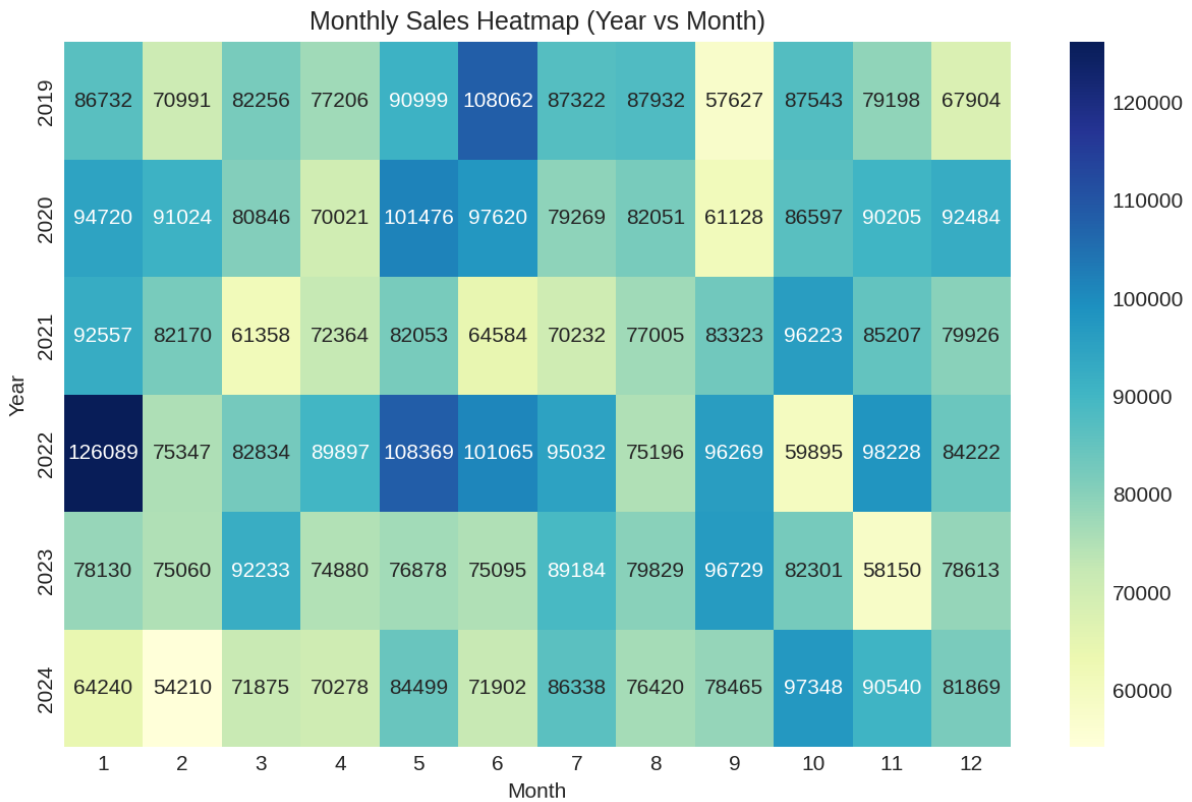
Subtask 4: Final Submission Deliverables

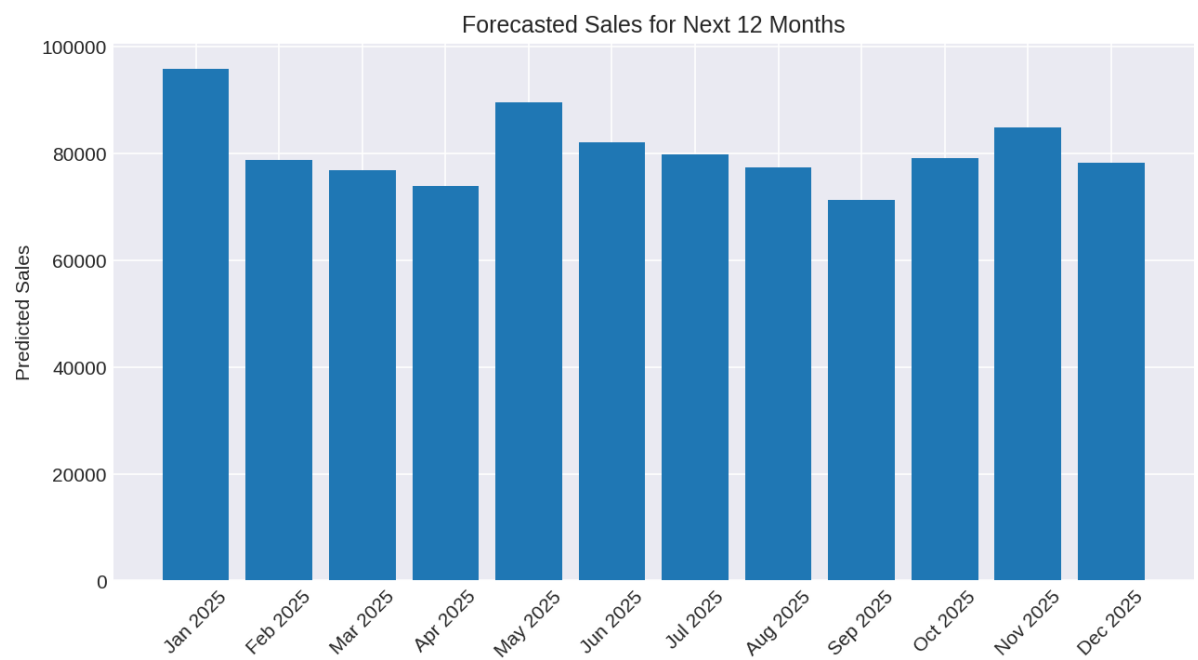
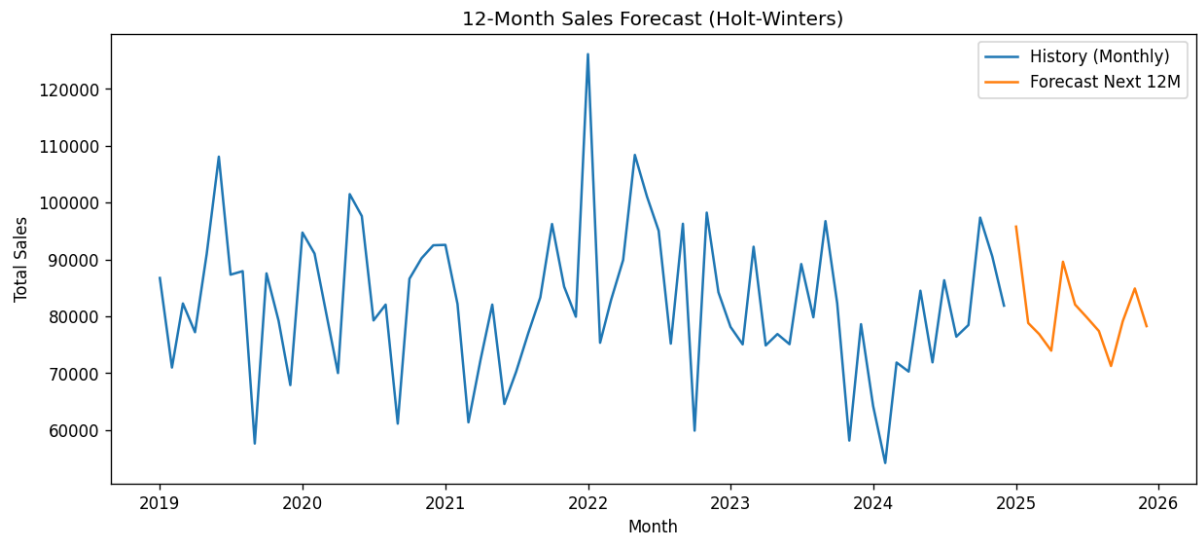
The final project includes:

1. **Sales Forecasting Report (this document)** summarizing methodology and insights.
2. **Visualizations:** Line chart (historical vs. predicted), seasonal sales bar chart/heatmap.
3. **Forecast Dataset (CSV)** with next 6–12 months sales predictions.
4. **Model Evaluation Metrics:** RMSE, MAPE values to validate model performance.
5. **Business Recommendations:** Inventory optimization, promotion planning, and resource allocation based on forecasted trends.

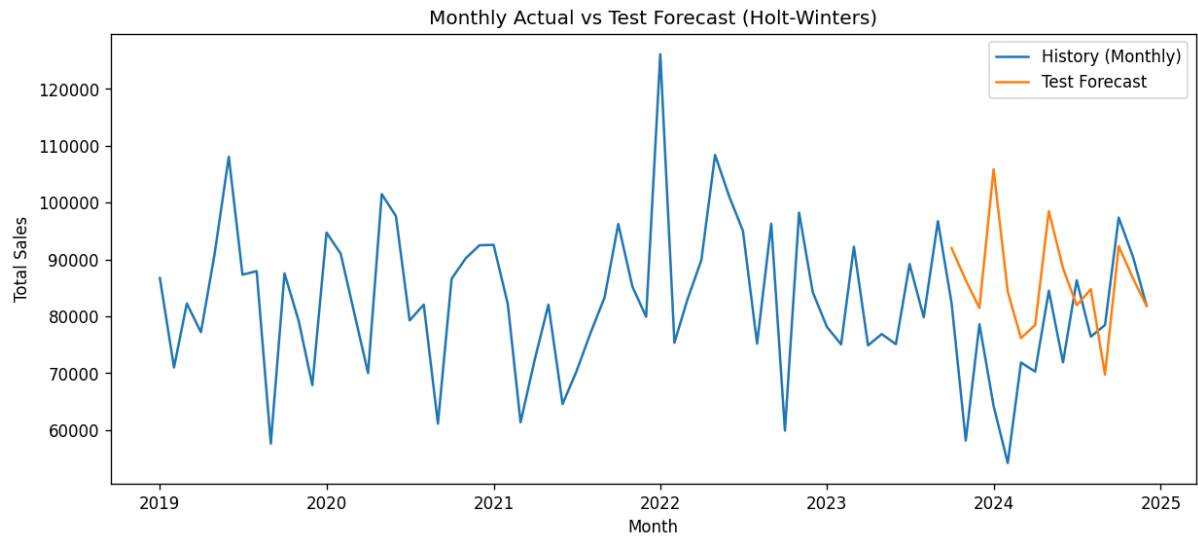
Output: A complete forecasting solution with actionable insights.

Outcome:





- Forecast 12 months with CL (Holt-Winters)
- Forecast 6 months with CL (BAR CHART)
- Executive Summary
- Sales forecast Report
- Process Daily Dataset
- Process Weekly Dataset
- Monthly Sales Heatmap



Conclusion

This project successfully prepared the dataset, applied a suitable forecasting model, and predicted future sales. The analysis highlights key sales patterns, provides accuracy metrics, and supports business decision-making for improved revenue planning.