

# Time Series Forecasting Report

## Report: Time Series Forecasting

### Objective

To perform time series forecasting using historical data on Alcohol Sales and Miles Traveled, applying models like Exponential Smoothing and ARIMA, and evaluating their forecasting performance.

### Dataset Overview

- Alcohol\_Sales.csv
- Miles\_Traveled.csv
- Both datasets are indexed by DATE (parsed as datetime objects).

### Data Preprocessing

The datasets were read using pandas with date parsing and setting DATE as the index.

Basic visualization was performed to inspect trends and patterns over time.

### Visualization

A dual-line plot comparing Alcohol Sales and Miles Traveled was created to analyze seasonality, trends, and correlations.

### Time Series Decomposition

Performed using `seasonal_decompose` from `statsmodels` to break the series into trend, seasonality, and residuals.

### Forecasting Models Used

#### 1. Exponential Smoothing

- Used to model level, trend, and seasonal components.
- Implemented via `statsmodels.tsa.holtwinters.ExponentialSmoothing`.

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## 2. ARIMA (AutoRegressive Integrated Moving Average)

- Applied using `statsmodels.tsa.arima.model.ARIMA`.
- Focused on capturing autocorrelation structure.

## Evaluation

Metric: Mean Squared Error (MSE)

Used to compare forecasted values against actual values.

## Visualizations Included

Time series line plots

Decomposed components (trend, seasonality, residual)

Forecast plots for each model

## Conclusion

The notebook demonstrates basic to intermediate forecasting techniques on real-world time series datasets.

Models were evaluated using MSE, and visual inspection helped in understanding model accuracy.