

# Assignment-4

## *Data Analysis and Interpretation*

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# 1 Parking Lot Problem

- **Part (a):**

- Mean Absolute Percentage Error (MAPE): 5.04%
- Mean Absolute Scaled Error (MASE): 0.80

- **Part (b):**

- Mean Absolute Percentage Error (MAPE): 1.85%
- Mean Absolute Scaled Error (MASE): 0.27

## 2 Analysis of Forecasting Metrics on a Real Dataset

### 2.1 Limitations of MAPE as a Metric

The Mean Absolute Percentage Error (MAPE) is often chosen as a metric for forecast accuracy. It measures the percentage discrepancy between actual and forecasted values:

$$\text{MAPE} = \frac{1}{n} \sum_{i=1}^n \left| \frac{A_i - F_i}{A_i} \right| \times 100 \quad (1)$$

where:

- $A_i$  denotes the actual observed values,
- $F_i$  represents the forecasted values, and
- $n$  is the count of observations.

#### 2.1.1 Challenges with MAPE

- **Instability with Low Values:** For data points with very low actual values, MAPE can produce extremely high percentages or even become undefined, distorting the error measurement and making it less reliable.
- **Skewed Emphasis on Small Values:** Since MAPE calculates error inversely proportional to actual values, it can unfairly magnify errors for lower values, causing unbalanced error representation across different demand levels.
- **Minimal Impact on Peak Periods:** High-demand periods, often critical for resource allocation, might not be adequately emphasized by MAPE since it does not inherently prioritize errors based on demand magnitude.

## 2.2 Alternative Metric: RMSE

To address MAPE's limitations, the Root Mean Squared Error (RMSE) offers an alternative by emphasizing larger errors. RMSE is defined as follows:

$$\text{RMSE} = \sqrt{\frac{1}{n} \sum_{i=1}^n (A_i - F_i)^2} \quad (2)$$

### 2.2.1 Advantages of RMSE

- **Focus on Significant Errors:** RMSE increases the weight on larger errors, making it more effective at reflecting significant deviations, which is beneficial when forecasting for resource-sensitive applications.
- **Suitability for Planning Scenarios:** By providing a comprehensive error measure, RMSE is useful in settings where larger forecast deviations could lead to substantial planning or resource allocation issues.

### 2.2.2 Example Calculation

Let's consider the following data points to compare MAPE and RMSE more concretely:

Month	Actual Count ( $A_i$ )	Forecasted Count ( $F_i$ )
Month 1	100	80
Month 2	5	10

Table 1: Sample Actual vs Forecasted Values

For MAPE:

$$\text{MAPE} = \frac{1}{2} \left( \frac{|100 - 80|}{100} \times 100 + \frac{|5 - 10|}{5} \times 100 \right) = 60\%$$

For RMSE:

$$\text{RMSE} = \sqrt{\frac{(100 - 80)^2 + (5 - 10)^2}{2}} = \sqrt{\frac{400 + 25}{2}} = \sqrt{212.5} \approx 14.58$$

In this case, RMSE provides a less exaggerated view of the errors, better aligning with practical needs.

## 2.3 Assessing Pre-COVID vs. Post-COVID Differences

To understand the potential shift in trends across different time periods, particularly around the COVID-19 period, we examine the first differenced series,  $\Delta Y$ , which is assumed to be weakly stationary.

### 2.3.1 Method Selection: Two-Sample t-Test

A two-sample t-test is appropriate to determine whether there's a statistically significant difference in mean values of  $\Delta Y$  across two periods: pre-COVID (before December 2019) and post-COVID (after January 2022).

### Hypothesis Formulation:

- **Null Hypothesis** ( $H_0$ ): The average of  $\Delta Y$  remains consistent across both periods, implying no significant difference.
- **Alternative Hypothesis** ( $H_a$ ): The averages of  $\Delta Y$  differ between the two periods, indicating a significant shift.

**Example of Calculation:** Assuming hypothetical values for pre- and post-COVID periods, the following table summarizes the differenced data:

Period	First Differences ( $\Delta Y$ )
Pre-COVID	5, 7, 8, 6, 5
Post-COVID	2, 3, 1, 4, 3

Table 2: Sample First Differences by Period

Calculated mean values:

- Pre-COVID Mean:  $\frac{5+7+8+6+5}{5} = 6.2$
- Post-COVID Mean:  $\frac{2+3+1+4+3}{5} = 2.6$

Using statistical software or t-table values, the test determines if there is a significant p-value indicating a shift in averages between these periods.