

## Scenario set question 1 Time series

### Scenario 1: Identifying Trend

You are given a dataset showing daily temperature for the past year. You notice the values generally increase over time. How do you confirm the presence of a trend?

1. Plot the time series to visually check if values rise over time.
2. Apply a moving average to smooth fluctuations.
3. Perform time series decomposition to view the trend component.
4. Compare rolling means across different periods.
5. Use a trend test (e.g., Mann-Kendall) to statistically confirm it.

### Scenario 2: Detecting Seasonality

You are analyzing monthly electricity usage and notice similar spikes every summer. How do you confirm seasonality?

1. Plot the time series to see repeating patterns at regular intervals.
2. Check seasonal decomposition to view the seasonality component.
3. Compare month-wise averages (e.g., each July vs other months).
4. Use autocorrelation plots (ACF) to see repeating peaks.
5. Examine rolling seasonal patterns across multiple years.

### Scenario 3: Making Data Stationary

Your time series model gives poor results. The data shows trend and seasonality. How do you prepare it for ARIMA?

1. Remove the trend using differencing (e.g., first difference).
2. Remove seasonality using seasonal differencing.
3. Apply transformations (log, sqrt) to stabilize variance if needed.
4. Re-check stationarity using ADF test after each step.
5. Use the transformed, stationary series as input for ARIMA.

### Scenario 4: Choosing Model for No Seasonality

Your time series data (e.g., daily website visits) has no visible seasonality but has a clear trend. Which model would you choose?

1. Use ARIMA since the data has trend but no seasonality.
2. Apply differencing to remove the trend before modeling.
3. Select AR and MA terms using ACF/PACF plots.
4. Train the ARIMA model on the stationary series.
5. Validate the model using residual diagnostics and forecasts.

### Scenario 5: Evaluating Forecast Performance

You forecast sales for 3 months and want to check how accurate it is. What steps do you follow?

1. Compare forecasted values with actual sales for the 3 months.
2. Calculate error metrics such as MAE, MSE, or RMSE.
3. Compute MAPE to understand percentage error.
4. Plot actual vs. predicted values to visually assess accuracy.
5. Check residuals to ensure no major patterns or bias remain.

### Scenario 6: Sudden Spike in Data

1. You're analyzing daily water consumption. One day, the value suddenly jumps very high. What will you do?

1. Check if the spike is a real event or a data entry error.
2. Compare the value with neighboring days to confirm abnormality.
3. Use statistical methods (like z-score or IQR) to detect outliers.
4. Decide whether to keep, correct, or remove the spike based on context.
5. If kept, apply smoothing or robust models to reduce its impact.

### Scenario 7: Sales Goes Up Every December

You're analyzing a store's monthly sales. Sales always increase in December. What does this indicate?

1. A repeating rise every December indicates seasonality in the data.
2. Specifically, it shows annual seasonal patterns tied to holidays/festivals.
3. This seasonal effect should be included in forecasting models like SARIMA.
4. Compare December sales across years to confirm consistency.
5. Plan seasonal adjustments or promotions based on this predictable pattern.

### Scenario 8: Forecasting with Missing Values

You are forecasting stock prices, but a few days of data are missing. What should you do?

1. Identify which dates have missing stock prices.
2. Decide if missing points are small enough to fill or too large to recover.
3. Use interpolation (linear/time-based) to fill small gaps.
4. For larger gaps, consider forward/backward fill or removing those days.
5. Re-check the cleaned dataset before training the forecasting model.