

Components of Time Series

> Long term trend

Long term trend is the overall general direction of the data, obtained ignoring any short-term effects such as seasonal variations or noise.

> Seasonality

Seasonality refers to periodic fluctuations that are repeated throughout the whole time series period.

> Stationarity

Stationarity is an important characteristic of time series. A time series is said to be stationary if its mean, variance, and covariance do not have significant changes over time. There are many transformations that can extract the stationary part of a non-stationary process.

> Noise

Every set of data has noise, that refers to random fluctuations or variations due to uncontrolled factors.

Autocorrelation

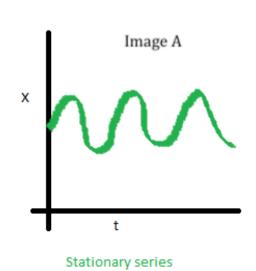
Autocorrelation is the correlation between the time series and a lagged version of itself and is used to identify seasonality and trend in time series data.

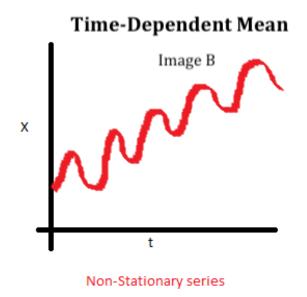
Metrics to Evaluate Time Series Forecast

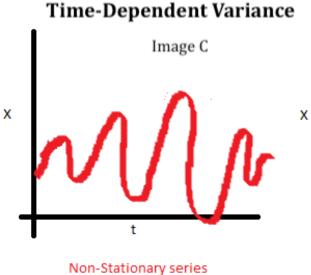
- MAE (Mean Absolute Error): Mean of absolute differences between actual value and forecasted value.
- RMSE (Root Mean Square Error): Root of mean of squares of difference between actual value and forecasted value.
- MAPE (Mean Absolute Percentage Error): Mean of absolute differences in percentage of actual value.
- MASE (Mean Absolute Scaled Error): Divide MAE of the forecasting model by MAE of a naive forecasting model (a simple model that always forecasts one previous value).

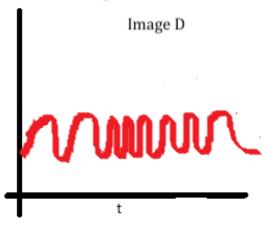
Stationarity

The Principles of Stationarity









Non-Stationary series

Time-Dependent Covariance