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# **HIS PROJECT - TSA**

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## **Task 04**

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# 1 Chapter 04

## 1.1 Setting up a test harness

### 1.1.1 Creating holdout (test) and validation dataset

As a standard practice, in machine learning, we set aside two parts of the dataset, name them validation data and test data, and don't use them at all to train the model.

- **validation set:** is used in the modeling process to assess the quality of the model. To select between different model classes, tune the hyperparameters, perform feature selection, and so on, we need a dataset.
- **holdout (test) set:** is like the final test of your chosen model. It tells you how well your model is doing in unseen data.

The best practice is to use the most recent part of the dataset as the test data. Additionally, it is advisable to have validation and test datasets of equal size to ensure that the decisions made during the modeling process, based on the validation data, are as applicable as possible to the test data.

### 1.1.2 Choosing an evaluation metric

In time series forecasting realm, there are scores of metrics with no real consensus on which ones to use. One of the reasons for this overwhelming number of metrics is that no one metric measures every characteristic of a forecast.

- **Mean Absolute Error (MAE):**
- **Mean Squared Error (MSE):**
- **Mean Absolute Scaled Error (MASE):**
- **Forecast Bias (FB):**

## 1.2 Generating strong baseline forecasts

Time series forecasting has been around since the early 1920s, and through the years, many brilliant people have come up with different models, some statistical and some heuristic-based.

Referred to as:

- Naïve forecast
- Moving average forecast
- Seasonal naive forecast
- Exponential smoothing (ETS)

- Simple exponential smoothing (SES)
- Double exponential smoothing (DES)
- Triple exponential smoothing or Holt -Winters (HW)
- The Autoregressive Integrated Moving Average (ARIMA)
- Theta Forecast
- Fast Fourier Transform forecast

After performing the aforementioned forecasting techniques it is important to remember that a comparison of their performance should be made. Not only with respect to the forecast bias of each technique but also the time elapsed to perform the techniques. When the 2-3 top candidates have finally been chosen, the forecasting algorithm can now be used on the validation and test data to assess which is the most adequate.

### **1.3 Assessing the forecastability of a time series**

### **1.4 Setting a Strong Baseline Forecast**

## **2 Chapter 05**

## **3 Chapter 06**

## **4 Chapter 07**