

# Differencing and smoothing

Many time series models are built under the assumption of stationarity. However, time series data often present non-stationary features such as trends or seasonality. Practitioners may consider techniques for detrending, deseasonalizing and smoothing that can be applied to the observed data to obtain a new time series that is consistent with the stationarity assumption. We briefly discuss two methods that are commonly used in practice for detrending and smoothing.

## Differencing

The first method is **differencing**, which is generally used to remove trends in time series data. The first difference of a time series is defined in terms of the so called difference operator denoted as  $D$ , that produces the transformation

$$Dy_t = y_t - y_{t-1}.$$

Higher order differences are obtained by successively applying the operator  $D$ . For example,

$$D^2y_t = D(Dy_t) = D(y_t - y_{t-1}) = y_t - 2y_{t-1} + y_{t-2}.$$

Differencing can also be written in terms of the so called backshift operator  $B$ , with

$$By_t = y_{t-1},$$

so that  $Dy_t = (1 - B)y_t$  and  $D^d y_t = (1 - B)^d y_t$ .

## Smoothing

The second method we discuss is moving averages, which is commonly used to “smooth” a time series by removing certain features (e.g., seasonality) to highlight other features (e.g., trends). A moving average is a weighted average of the time series around a particular time  $t$ . In general, if we have data  $y_{1:T}$ , we could obtain a new time series such that

$$z_t = \sum_{j=-q}^p a_j y_{t+j},$$

for  $t = (q+1) : (T-p)$ , with  $a_j \geq 0$  and  $\sum_{j=-q}^p a_j = 1$ . Often we work with moving averages in which  $p = q$  (centered) and  $a_j = a_{-j}$  (symmetric) for all  $j$ .

Assume we have periodic data with period  $d$ . Then, symmetric and centered moving averages can be used to remove such periodicity as follows:

- If  $d = 2q$  :

$$z_t = \frac{1}{d} \left( \frac{1}{2} y_{t-q} + y_{t-q+1} + \dots + y_{t+q-1} + \frac{1}{2} y_{t+q} \right).$$

- If  $d = 2q + 1$  :

$$z_t = \frac{1}{d} (y_{t-q} + y_{t-q+1} + \dots + y_{t+q-1} + y_{t+q}).$$

**Example:** To remove seasonality in monthly data (i.e., seasonality with a period of  $d = 12$  months), one can use a moving average with  $p = q = 6$ ,  $a_6 = a_{-6} = 1/24$ , and  $a_j = a_{-j} = 1/12$  for  $j = 0, \dots, 5$ , resulting in

$$z_t = \frac{1}{24} y_{t-6} + \frac{1}{12} y_{t-5} + \dots + \frac{1}{12} y_{t+5} + \frac{1}{24} y_{t+6}.$$