

## Time Series

A time series  $Y_t$  ( $t = 1, 2, \dots, n$ ) is a set of  $n$  observations recorded through time  $t$ , (e.g. days, weeks, months). The arithmetic mean of blocks of  $k$  successive values

$$\frac{Y_1 + Y_2 + \dots + Y_k}{k}, \frac{Y_2 + Y_3 + \dots + Y_{k+1}}{k}, \dots$$

is a **simple moving average** of order  $k$ , itself a time series which is *smoother* than  $Y_t$  and can be used to track, or estimate, the underlying level,  $\mu_t$ , of  $Y_t$ . If  $0 < \alpha < 1$  an **exponentially weighted moving average** (EWMA) at time  $t$  uses a discounted weighted average of current and past data to estimate  $\mu_t$  with

$$\hat{\mu}_t = \alpha Y_t + \alpha(1 - \alpha)Y_{t-1} + \alpha(1 - \alpha)^2 Y_{t-2} + \dots$$

This is equivalent to the recurrence relation

$$\hat{\mu}_t = \alpha Y_t + (1 - \alpha)\hat{\mu}_{t-1}$$

Moving averages are often plotted on the same graph as  $Y_t$ . If  $Y_t$  additionally contains trend,  $R_t$ , the rate of change of data per unit time, and  $\mu_t = \mu_{t-1} + R_{t-1}$ , then the recurrence relation is

$$\hat{\mu}_t = \alpha Y_t + (1 - \alpha)(\hat{\mu}_{t-1} + \hat{R}_{t-1})$$

If  $0 < \beta < 1$  the *trend smoothing equation* is

$$\hat{R}_t = \beta(\hat{\mu}_t - \hat{\mu}_{t-1}) + (1 - \beta)\hat{R}_{t-1}$$

known as *Holt's Linear Exponential Smoothing*.

If  $Y_t$  also contain *seasonality*,  $S_t$ , a smoothing constant  $\gamma$ , ( $0 < \gamma < 1$ ) is used in a *seasonal smoothing equation*,

$$\hat{S}_t = \gamma Y_t / \hat{\mu}_t + (1 - \gamma)\hat{S}_{t-k},$$

assuming the periodicity is  $k$ , with *multiplicative* seasonality. For monthly data  $k = 12$ .

**Forecasting from time  $n$  (now) to time  $n + h$  ( $h = 1, 2, \dots$ )**

*Level only*,  $\hat{Y}_{n+h} = \hat{\mu}_n$ , the latest EWMA.

*Level and constant trend*,  $\hat{Y}_{n+h} = a + b(n + h)$ , the simple linear regression trend line of  $Y_t$  against  $t$ .

*Level and changing trend*,  $\hat{Y}_{n+h} = \hat{\mu}_n + h\hat{R}_n$ .

*Level, changing trend and seasonality*,  $\hat{Y}_{n+h} = \hat{\mu}_n + h\hat{R}_n$ , where  $\hat{\mu}_n = \alpha Y_n / \hat{S}_{n-12} + (1 - \alpha)(\hat{\mu}_{n-1} + \hat{R}_{n-1})$ .