

### **Forecasting Trend Time Series**

**Incorporating Steps and Trends** 





## **Trend Fitting**

Where the time series has a trend, a <u>trend fitting and</u> <u>extrapolation</u> approach may be used for prediction

Trend fitting is a method where the time series is linked to **some function of a time index** 

$$Y_t = f \text{ (time)}$$

A <u>linear trend</u> equation is typically assumed although it depends on the trend observed

Prediction is based on extrapolation by <u>substitution of</u> the appropriate value for the time index



### **Trend Fitting (cont)**

Assuming a <u>linear</u> trend the equation is

$$\mathbf{Y}_{\mathsf{t}} = \mathbf{a} + \mathbf{b} * \mathbf{t}$$

where t = a time index and  $\alpha$  and  $\beta$  are constants

The values of  $\alpha$  and  $\beta$  typically estimated by regression of the time series (Y) against time index (t)

EXCEL has numerous alternative ways of estimating the above equation and/or trend fitting/extrapolation including the regression routine

# Holts Exponential Smoothing (HES)



The two general methods already studied (MA, SES) are useful when the time series is **predominantly horizontal** but will **not be good predictors** when the **time series has other systematic components** 

If the time series has a trend then MA and SES will be poor predictors

A simple **extension of the SES model (Holt's Model)** which incorporates a **trend component** can be used for better prediction

Like SES, <u>Holts Exponential Smoothing (HES)</u> uses a smoothing algorithm to **remove random influences** from the time series revealing the underlying systematic components.



### **HES Equations**

HES is characterised by three equations;

1. 
$$L_t = \alpha Y_t + (1 - \alpha) (L_{t-1} + T_{t-1})$$

2. 
$$T_t = \beta (L_t - L_{t-1}) + (1 - \beta)(T_{t-1})$$

$$3. \quad \mathbf{F}_{t+m} = \mathbf{L}_t + \mathbf{m} \mathbf{T}_t$$

The first equation is for <u>level</u>, the second for <u>trend</u> and the third is the <u>forecasting equation</u> for "<u>m" periods into</u> the <u>future</u>

#### **HES Equations (cont)**



 $L_t$ = Smoothed level at period (t)

 $Y_t$ = Actual time series value at period t

 $\alpha$  = Smoothing constant for level

T<sub>t</sub>= Trend estimate at period t

 $\beta$ = Smoothing constant for the trend (o <=  $\beta$  <= 1)

m = Number of periods ahead to be forecast

 $F_{t+m}$  = Holt's forecast value for period t + m



### HES (cont.)

The values of  $\alpha$ ,  $\beta$  are arbitrarily determined

Typically **between o and 1 inclusive** although some programs (eg MINITAB) ignore this restriction

Try different  $\alpha$ ,  $\beta$  to determine the "optimum" combination (as assessed by error criteria (MSE, MAE, MAPE)

**SOLVER in EXCEL** can also be used to find the optimum by minimising a chosen error criterion

Initialisation of the model requires initial estimates for  $L_t$  and  $T_t$ .  $L_t$  is usually the <u>initial time series</u> value  $(Y_1)$ 

 $T_t$  is usually the <u>average</u> of the <u>increase/decrease</u> in the first few periods (use either zero or  $(Y_2-Y_1)$  or  $((Y_3-Y_1)/2)$ )