### Seminar in Business Processes and Operations Management

**BUEC 865** Dr. C. Kydd

### **Exponential Smoothing** Model

Uses a combination of most recent data point and most recent forecast

Weight (smoothing constant) is assigned to data and forecast points
Weights can be varied between 0 and 1

Exponential Smoothing Model formulas

$$F_{t+1} = \alpha D_t + (1-\alpha) F_t$$

$$F_{t+1} = F_t + \alpha (D_t - F_t)$$

 $F_{t+1}$  = Forecast for the next period (Period t+1)

 $\alpha = \text{Smoothing constant} \quad (0 \le \alpha \le 1)$ 

 $D_t$  = Actual data for current period (Period t)

 $F_t$  = Forecast for current period made in last period

### Smoothing constant

 $\alpha$  = smoothing constant

$$0 \le \alpha \le 1$$
  
.1 $\le \alpha \le .4$  usually

$$\begin{array}{ll} Forecast \\ for \ next \ pd. \end{array} = \ \alpha \ \begin{bmatrix} current \\ demand \end{bmatrix} + \ (1-\alpha) \begin{bmatrix} previous \\ forecast \end{bmatrix}$$

Exponential Smoothing Model - formulas

$$F_{t+1} = \alpha D_t + (1-\alpha) F_t$$
or
$$F_{t+1} = F_t + \alpha (D_t - F_t)$$

 $F_{t+1}$  = Forecast for the next period (Period t+1)

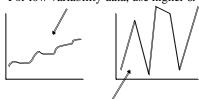
 $\alpha$  = Smoothing constant (0 <=  $\alpha$  <=1)

 $D_t$  = Actual data for current period (Period t)

 $F_t$  = Forecast for current period made in last period

### **Smoothing Constant**

For low variability data, use higher  $\alpha$ 



For high variability data, use lower  $\alpha$ 

#### **Exponential Smoothing Formulas**

$$F_{t+1} = \alpha D_t + (1-\alpha) F_t$$

$$F_t = \alpha D_{t-1} + (1-\alpha) F_{t-1}$$

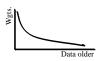
$$\begin{aligned} F_{t+1} &= \alpha \; (D_t) + (1\text{-}\alpha) \; [\; \alpha D_{t-1} + (1\text{-}\alpha) \; F_{\; t\text{-}1} \; ] \\ F_{t-1} &= \boxed{\alpha D_{\; t\text{-}2} + (1\text{-}\alpha) \; (F_{\; t\text{-}2})} \end{aligned}$$

# Exponential Smoothing Formulas (cont.)

$$\begin{split} F_{t+1} &= \\ &\alpha D_t + (1 \text{-}\alpha) \left[ \ \alpha D_{t\text{-}1} + (1 \text{-}\alpha) \left( \alpha D_{t\text{-}2} + (1 \text{-}\alpha) F_{t\text{-}2} \right) \right] \end{split}$$

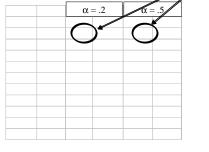
## Exponential Smoothing Formulas (cont.)

$$\begin{array}{cccc} \underline{Data} & \underline{Weight} & \underline{\mathit{example}} \\ D_t & \alpha & .3 \\ D_{t-1} & \alpha \, (1\text{-}\alpha) & .21 \\ D_{t-2} & \alpha \, (1\text{-}\alpha)^2 & .147 \end{array}$$









### Comparison of Exponential Smoothing Models

E.S. 
$$\alpha = .2$$

E.S. 
$$\alpha = .5$$

$$MAD = 9.31$$

$$MAD = 7.37$$

$$Bias = 8.39$$

Bias 
$$= 4.83$$

E.S. with  $\alpha = .5$  yields better MAD and Bias, so provides better model

# Exponential Smoothing model - tracking signals with $\alpha = .2$

$$T.S._1 = -1$$

$$T.S._6 = 4.5$$

$$T.S._2 = 1.3$$

$$T.S._7 = 5.7$$

$$T.S._3 = 1.4$$

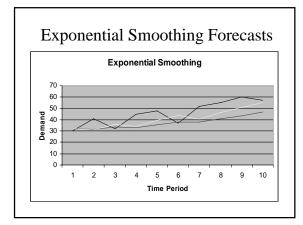
$$T.S._8 = 6.9$$

$$T.S._4 = 2.9$$

$$T.S._9 = 8.0$$

$$T.S._5 = 4.1$$

$$T.S._{10} = 9.0$$



# 

Pd.	$D_{t}$	$F_t'$	F <sub>t</sub> "	$F_t'-F_t''$	$b_t = F_t'' - F_t$
1	30	32	32	0	
2	41	31.6	32	-0.4	0
3	32	33.5	31.9	1.6	-0.1
4	45	33.2	32.2	1	0.3
5	48	35.6	32.4	3.2	0.2
6	37	38.1	33	5.1	0.6
7	52	37.9	34	3.9	1
8	55	40.7	34.8	5.9	0.8
9	60	43.6	36	7.6	1.2
10	57	46.9	37.5	9.4	1.5

### Double Smoothing (cont.)

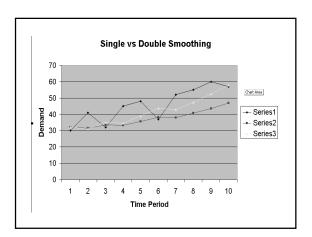
Final Forecast	_e_	64.3
		$MAD = \frac{64.3}{9} = 7.1$
31.2	9.8	$\frac{1111D}{9} = \frac{7.11}{9}$
35	-3	
34.5	10.5	
39	9	
43.8	-6.8	Dies 43.10 _ 4.70
42.8	9.2	Bias= $\frac{43.10}{9}$ = 4.79
47.4	7.6	9
52.4	7.6	
57.8	-0.8	

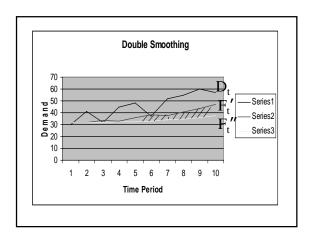
Comparison of Exponential Smoothing Models

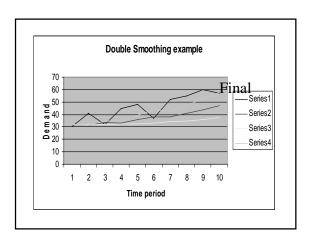
E.S.  $\alpha = .2$  E.S.  $\alpha = .2$  (double)

MAD = 9.31 MAD = 7.14 Bias = 8.39 Bias = 4.79

Double E.S. yields better MAD and Bias, so provides better model







### Problem 1: Forecasting

You are given sales data for 10 periods.

- 1. Apply the four time series models to forecast sales for periods 6-11.
- 2. Compare results and suggest which is best and why.

For these, use sales and time period only.

Problem 1 Data						
Period	Sales	Mkt price	GNP	CPI		
1	35725	57.63	115	161.2		
2	47180	78.50	125	170.5		
3	54965	62.88	132	181.5		
4	63220	53.75	140	195.4		
5	66315	50.00	150	217.4		
6	57730	45.00	163	246.8		
7	62700	38.50	178	272.4		
8	60025	62.38	195	289.1		
9	74590	74.38	206	298.4		
10	83900	78.38	213	311.1		

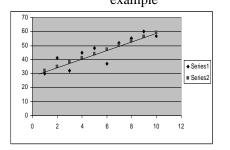
### Simple Linear Regression

Attempts to fit a straight line to set of data points

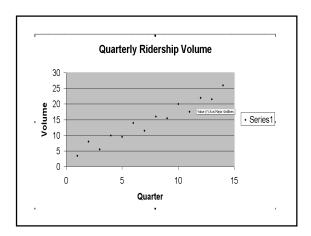
Uses least squares estimate to determine line (minimizes sum of squared deviations between points and line)

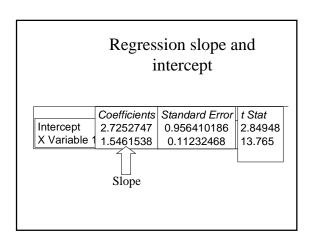
Uses equation Y = a + bX where Y is dependent variable, X is independent variable, a is Y-intercept, b is slope

# **Simple Linear Regression -** example



Simple Linear Regression  Mass Transit Quarterly Ridership Data					
Qtr., X  1 2 3 4 5 6 7 8 9 10	Volume, Y 3.5 8 5.5 10 9.5 14 11.5 16 15.5 20	$Y=a+bX$ or $T_x=a+bX$ Dependent var. Independent var. $a=Y-intercept$ $b=slope$			





### Simple Linear Regression

Quarterly Ridership Problem (cont'd.)

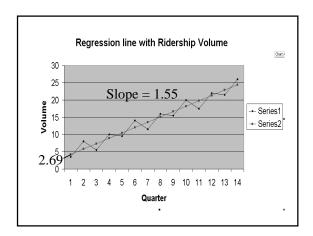
$$T_x = a + bX = 2.69 + 1.55X$$
  
Y-intercept  $\stackrel{\frown}{\Box}$  slope

To predict future values of T<sub>x</sub> (points on the regression line):

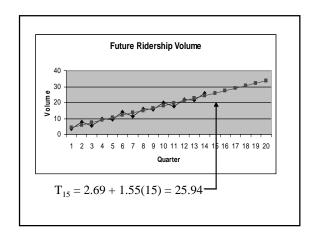
$$T_{15} = 2.69 + 1.55(15) = 25.94$$

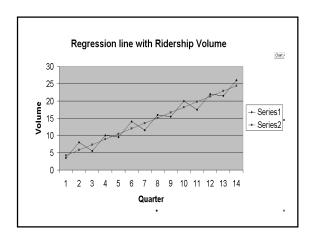
$$T_{16} = 2.69 + 1.55(16) = 27.49$$
  
 $T_{17} = 2.69 + 1.55(17) = 29.04$ 

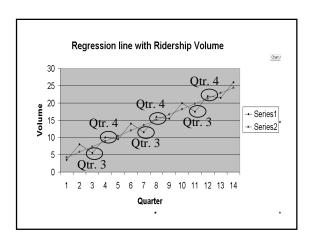
$$T_{17} = 2.69 + 1.55(17) = 29.04$$



SUMMARY OUTPUT			
Regression Statistics			
Multiple R	0.96976268		
R Square	0.94043966		
Adjusted R Square	0.9354763		
Standard Error	1.69420473		
Observations	14		







### **Incorporating Seasonality**

1. Compute measure of seasonality

$$S = \frac{Y}{T} = \frac{Actual\ volume}{Computed\ trend\ value}$$

- 2. Take average  $\overline{(S)}$  of all relevant S's
- 3. Apply S to appropriate values of T to forecast into the future

#### Finding Seasonal Factor for Quarter 4

1. Compute measure of seasonality

$$T_4 = 2.69 + 1.55 (4) = 8.89$$
 $T_8 = 2.69 + 1.55 (8) = 15.09$ 
 $T_{12} = 2.69 + 1.55 (12) = 21.29$ 
Computed

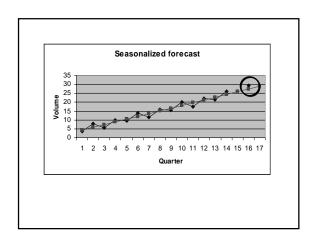
$$S_4 = 10/8.89 = 1.12$$
  
 $S_8 = 16/15.09 = 1.06$   
 $S_{12} = 22/21.2 = 1.03$ 
2. Average S's  $\overline{S}_4 = 1.07$ 

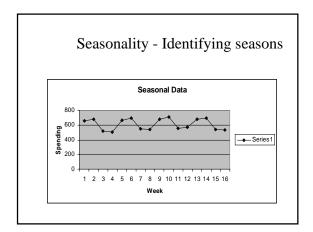
#### Finding Seasonal Factor for Quarter 4

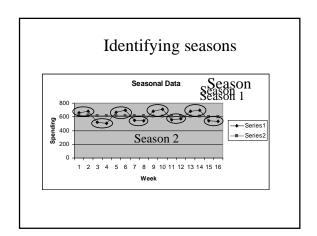
3. Apply  $\overline{S}$  to appropriate values of T

$$T_{16} = 2.69 + 1.55(16) = 27.49$$

Adjusted 
$$T_{16} = 27.49 (1.07) = 29.41$$







### Problem 1 (cont'd.)

Use simple regression to forecast values for sales, market price, CPI and GNP for periods 11-13.

Comment on how well each forecast fits the data.

In each case, time will be the independent variable.

Use POM for Windows/Forecasting/Time series module.

### Forecasting Quiz

Find forecast for April and May using

- 1) 2-month moving average
- 2) Exp. Smooth. with  $\alpha = .4$

	<u>Actual</u>	Forecast
Jan.	30 units	
Feb.	45 units	
Mar.	55 units	50
Apr.	65 units	
May	70 units	

### Forecasting Quiz

Find forecast for April and May using

3) Double smoothing model with  $\alpha = .4$  (again assume F"<sub>Mar</sub> = 50)

Which model works best here?

1	1
1	7

### Forecasting Quiz - Solution

Find forecast for April and May using

1) 2-month moving average

$$F(April) = (45 + 55)/2 = 50$$
  $e = 15$ 

$$F(May) = (55 + 65)/2 = 60$$
  $e = 10$ 

$$MAD = 25/2 = 12.5$$

Bias = 
$$25/2 = 12.5$$

#### Forecasting Quiz

Find forecast for April and May using

2) E.S. model with  $\alpha$ =.4

$$F(April) = .4(55) + .6(50) = 52$$
  $e = 13$ 

$$F(May) = .4(65) + .6(52) = 57.2 e = 12.8$$

$$MAD = 25.8/2 = 12.9$$

Bias = 
$$25.8/2 = 12.9$$

### Forecasting Quiz

Find forecast for April and May using

3) Double smoothing model with  $\alpha = .4$ 

(again assume F"<sub>Mar</sub> = 50)

$$F''(April) = .4(50) + .6(50) = 50$$

$$F''(May) = .4(52) + .6(50) = 50.8$$

### Forecasting Quiz