

Operations Management

Chapter 4 – Forecasting Demand

***PowerPoint presentation to accompany
Heizer/Render, Operations Management, 12 th Ed.***

What is Forecasting?

Predict the next number in the pattern:

a) 3.7, 3.7, 3.7, 3.7, 3.7, ?

b) 2.5, 4.5, 6.5, 8.5, 10.5, ?

c) 5.0, 7.5, 6.0, 4.5, 7.0, 9.5, 8.0, 6.5, ?



What is Forecasting?

a) 3.7, 3.7, 3.7, 3.7, 3.7, **3.7**

b) 2.5, 4.5, 6.5, 8.5, 10.5, **12.5**

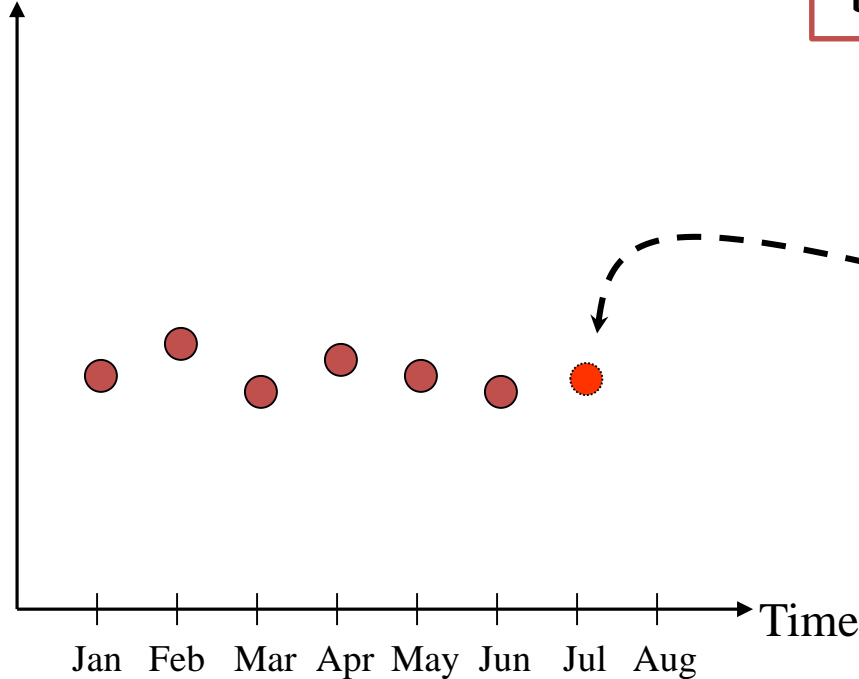
c) 5.0, 7.5, 6.0, 4.5, 7.0, 9.5, 8.0, 6.5, **9**

- Demand forecasting is a Process of predicting a future event
- Underlying basis of all business decisions:
Production, Inventory, Personnel, Facilities

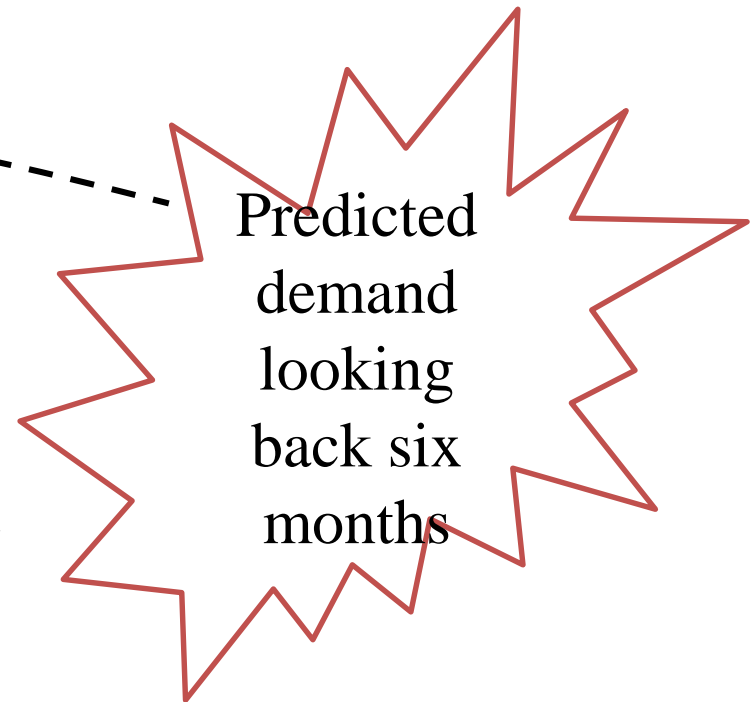
What is Forecasting?

We try to predict the future by looking back at the past

Demand for Mercedes E Class



- Actual demand (past sales)
- Predicted demand

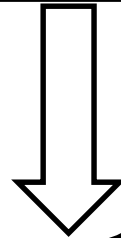


Realities of Forecasting

- Most forecasting methods assume that there is some underlying stability in the system
- Forecasts are more accurate for shorter time periods
- Every forecast should include an error estimate
- Forecasts are seldom perfect..... unpredictable outside factors may impact the forecast

Realities of Forecasting

If Forecasts are seldom perfect(almost always wrong), why do we need to forecast?



“Best” educated guesses about future are more valuable for purpose of Planning than no forecasts and hence no planning.

Importance of Forecasting in OM

- Departments throughout the organization depend on forecasts to formulate and execute their plans.
- Finance needs forecasts to project cash flows and capital requirements.
- Human resources need forecasts to anticipate hiring needs.

Production needs forecasts to plan production levels, workforce, material requirements, inventories, etc.

Forecasting Time Horizons

1. *Short-range forecast*

- ▶ Up to 1 year, generally less than 3 months
- ▶ Purchasing, job scheduling, workforce levels, job assignments, production levels

2. *Medium-range forecast*

- ▶ 3 months to 3 years
- ▶ Sales and production planning, budgeting

3. *Long-range forecast*

- ▶ 3+ years
- ▶ New product planning, facility location, research and development

Types of forecasting methods

Qualitative methods

Rely on subjective opinions from one or more experts.

Quantitative methods

Rely on data and analytical techniques.

Qualitative Methods

➤ Used when situation is vague and little data exist



New products

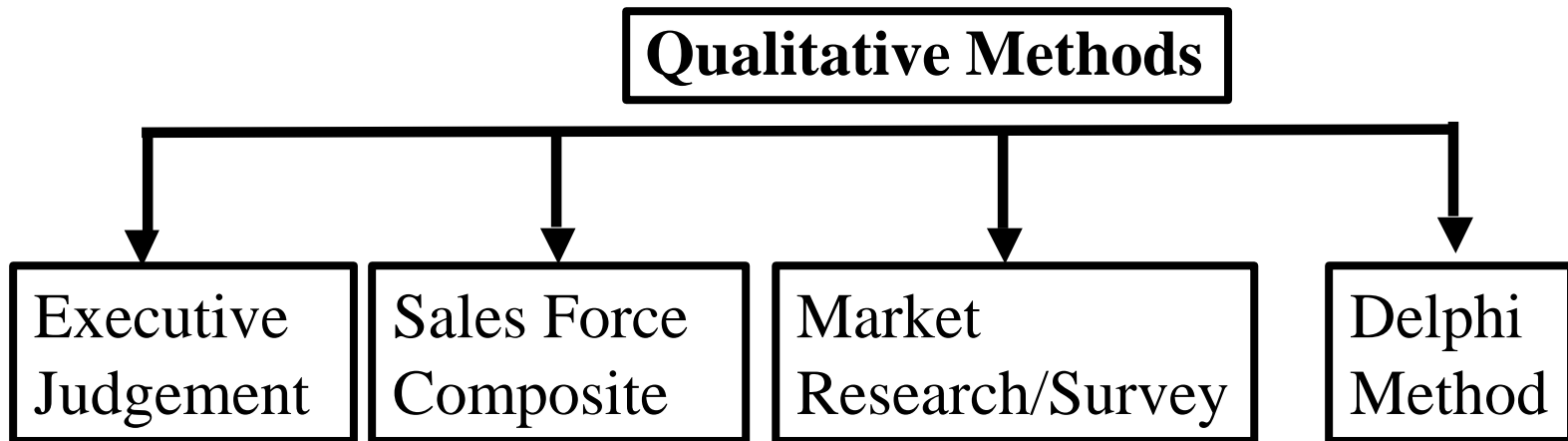
New technology

➤ Involves intuition, experience

Types of forecasting methods

Qualitative Methods

Briefly, the qualitative methods are:



Executive
Judgment

Opinion of a group of high level experts or managers is pooled

Types of forecasting methods

Qualitative Methods

Sales Force
Composite

Each regional salesperson provides his/her sales estimates. Those estimates are then reviewed to make sure they are realistic. All regional estimates are then pooled at national levels to obtain an overall estimates.

Market
Research/
Survey

Solicits input from customers pertaining to their future purchasing plans.

Delphi
Method

Typically, the procedure consists of the following steps:

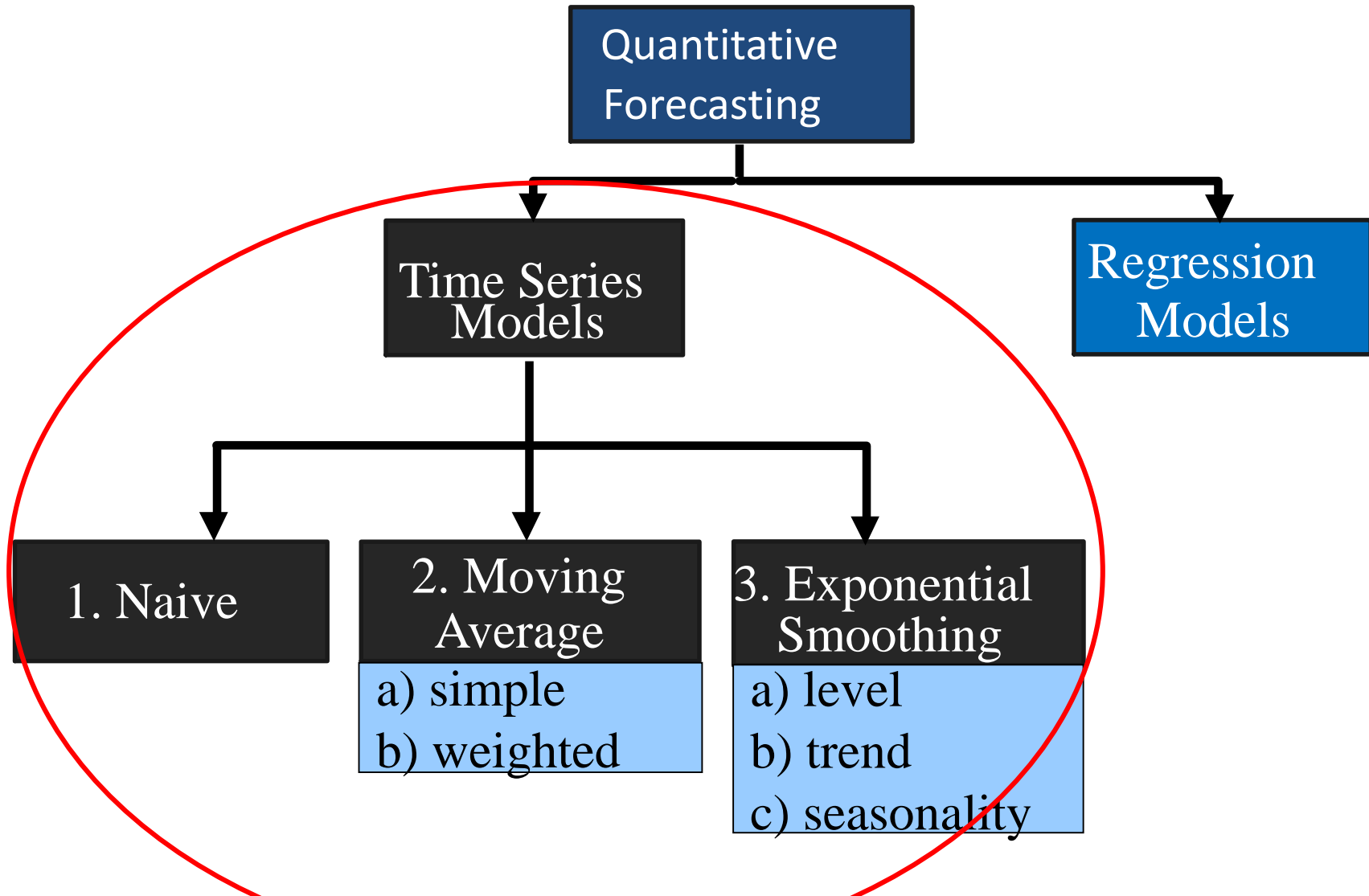
Types of forecasting methods

Qualitative Methods

- Each expert in the group makes his/her own forecasts in form of statements
- The coordinator collects all group statements and summarizes them
- The coordinator provides this summary and gives another set of questions to each group member including feedback as to the input of other experts
- The above steps are repeated until a consensus is reached.

Types of forecasting methods

Quantitative Methods



Types of forecasting methods

Quantitative Methods

- Used when situation is ‘stable’ and historical data exist



Existing products

Current technology

- Involves mathematical techniques

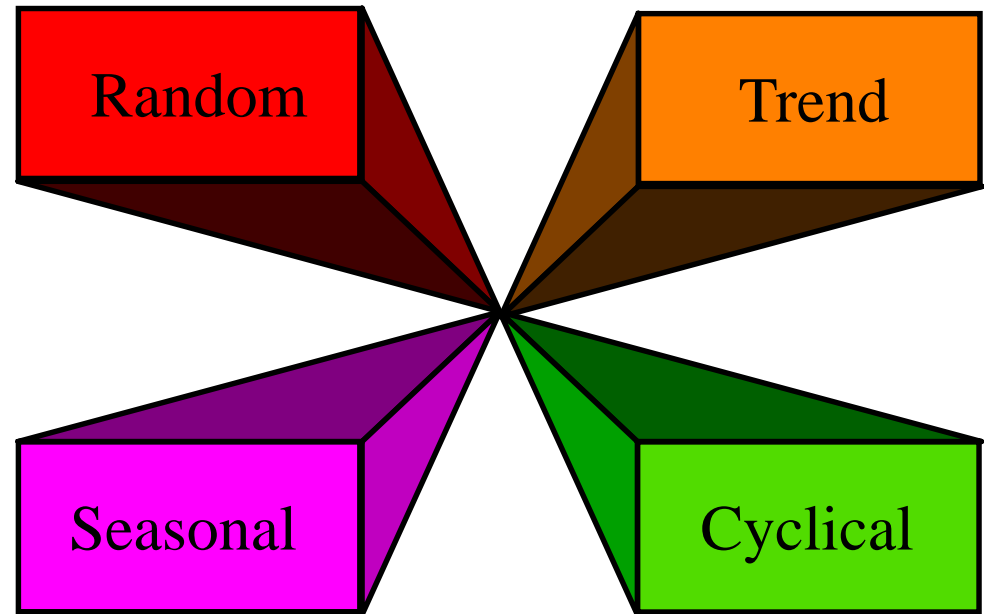
Types of forecasting methods

Quantitative Methods

Time Series Models

- Try to predict the future based on past data
- Assume that factors influencing the past will continue to influence the future

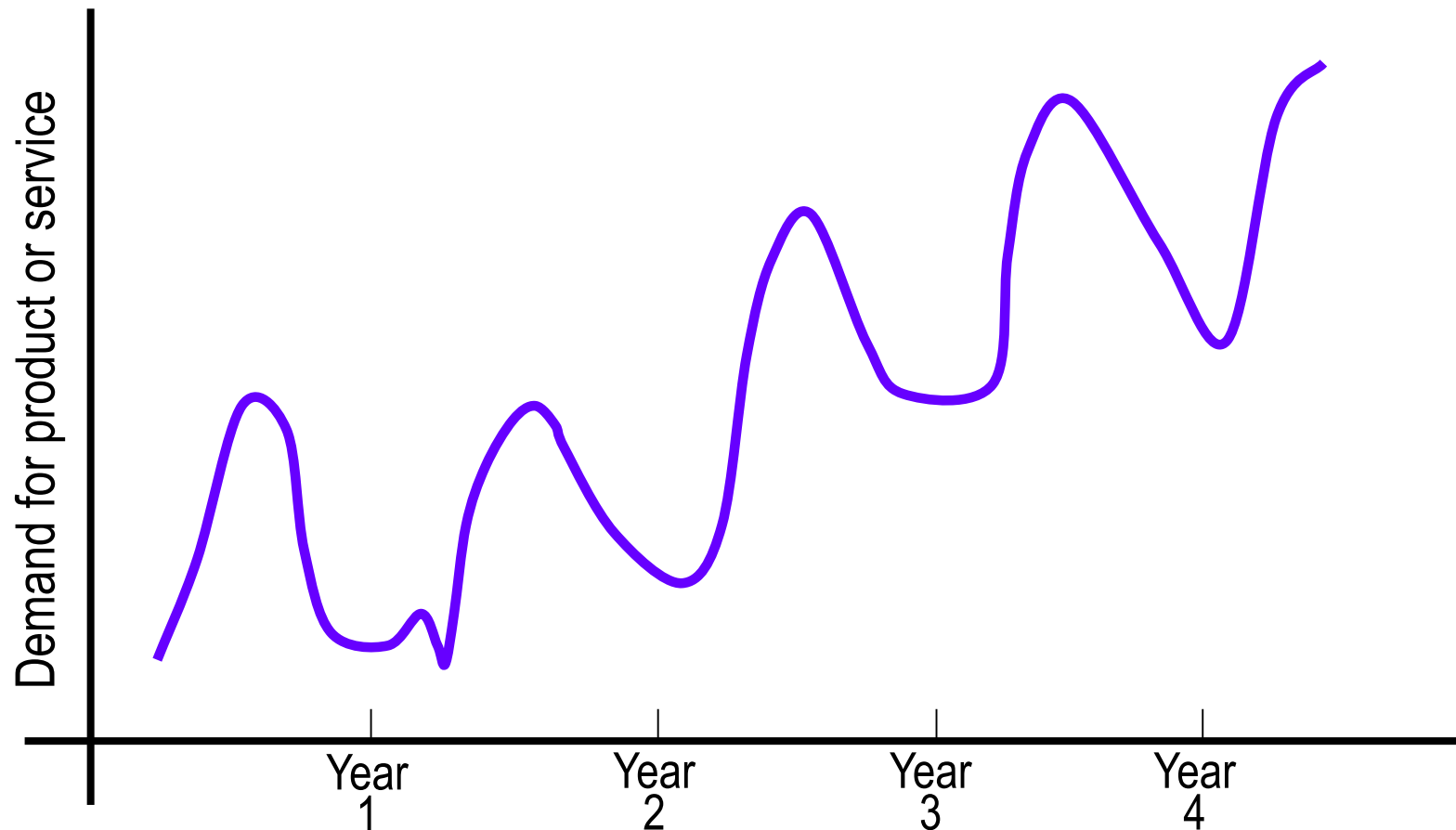
Time Series Models:
Components



Types of forecasting methods

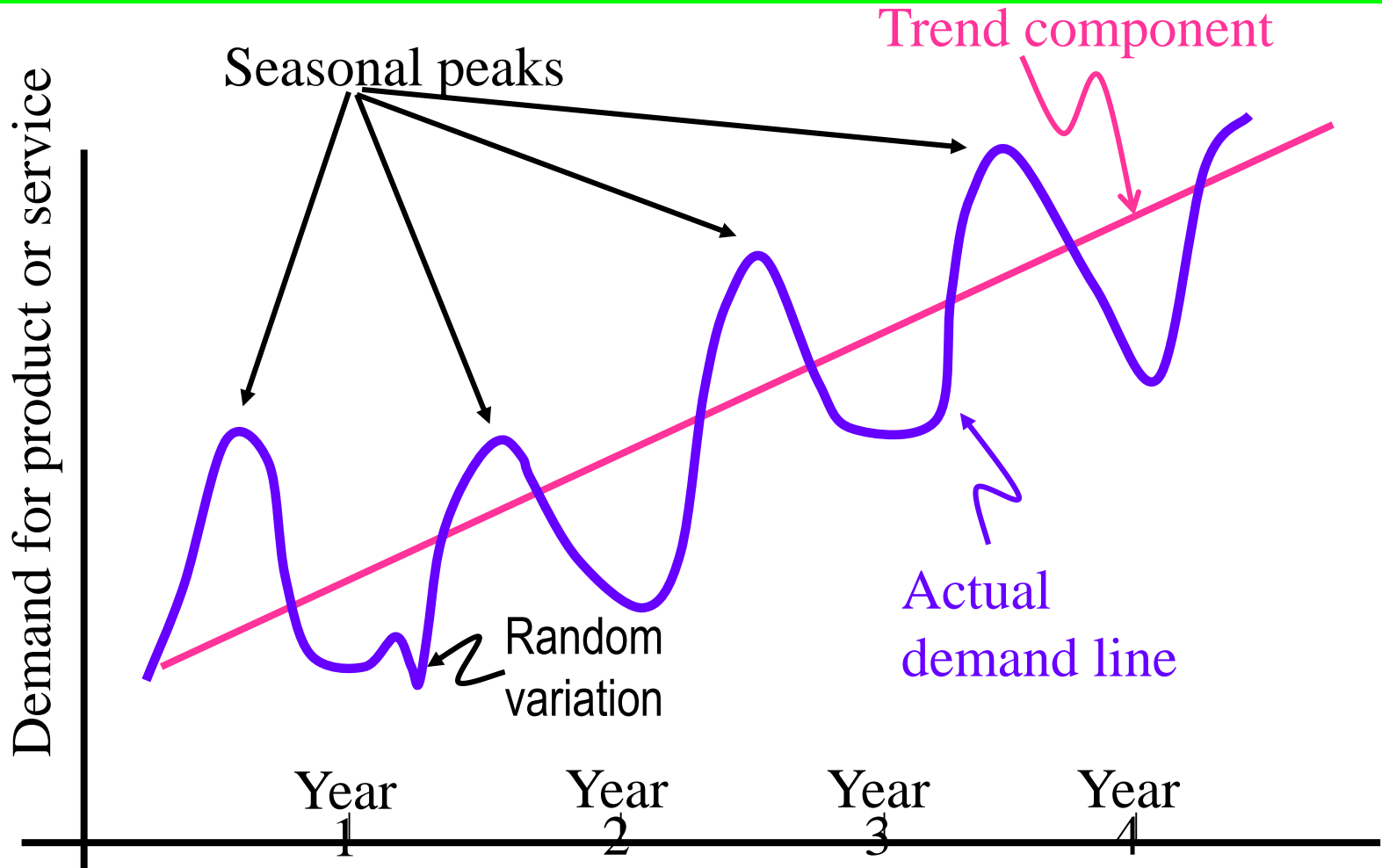
Quantitative Methods

Product Demand over Time



Types of forecasting methods

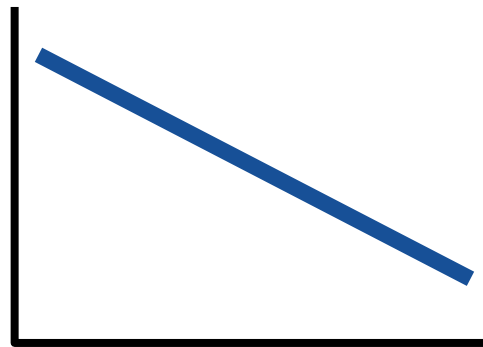
Quantitative Methods



Now let's look at some time series approaches to forecasting...

Trend Component

- Persistent, overall upward or downward pattern
- Changes due to population, technology, age, culture, etc.
- Typically several years duration



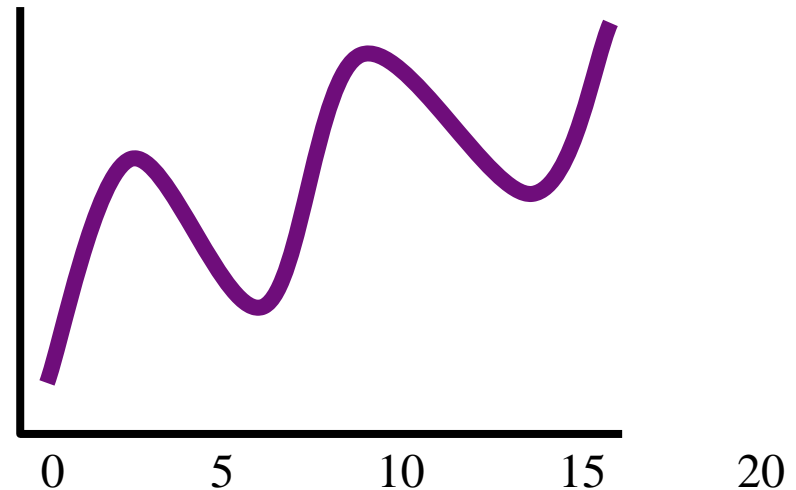
Seasonal Component

- Regular pattern of up and down fluctuations
- Due to weather, customs, etc.
- Occurs within a single year

| PERIOD LENGTH | "SEASON" LENGTH | NUMBER OF "SEASONS" IN PATTERN |
|---------------|-----------------|--------------------------------|
| Week | Day | 7 |
| Month | Week | 4 – 4.5 |
| Month | Day | 28 – 31 |
| Year | Quarter | 4 |
| Year | Month | 12 |
| Year | Week | 52 |

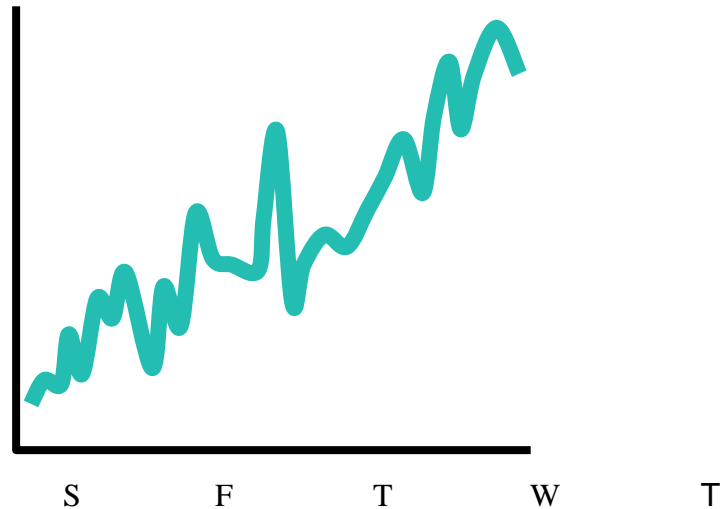
Cyclical Component

- Repeating up and down movements
- Affected by business cycle, political, and economic factors
- Multiple years duration



Random Component

- Erratic شاذ, unsystematic غير منتظم, 'residual' fluctuations
- Due to random variation or unforeseen events
- Short duration and nonrepeating



Types of forecasting methods

Quantitative Methods

1. Naive Approach

- Demand in *next* period is the same as demand in *most recent* period

✓ May sales = 48 → June forecast = 48

- Usually not good

Types of forecasting methods

Quantitative Methods

2a. Simple Moving Average

Assumes an average is a good estimator of future behavior

- Used if *little or no trend*
- Used for smoothing

The formula for the simple moving average is:

$$F_{t+1} = \frac{A_t + A_{t-1} + A_{t-2} + \dots + A_{t-n+1}}{n}$$

F_{t+1} = Forecast for the upcoming period, t+1

n = Number of periods to be averaged

A_t = Actual occurrence in period t

Types of forecasting methods

Quantitative Methods

Ex: 1

| Week | Demand |
|------|--------|
| 1 | 650 |
| 2 | 678 |
| 3 | 720 |
| 4 | 785 |
| 5 | 859 |
| 6 | 920 |
| 7 | 850 |
| 8 | 758 |
| 9 | 892 |
| 10 | 920 |
| 11 | 789 |
| 12 | 844 |

- *Question: What are the 3-week and 6-week moving average forecasts for demand?*
- Assume you only have 3 weeks and 6 weeks of actual demand data for the respective forecasts

Types of forecasting methods

Quantitative Methods

Calculating the moving averages gives us:

| Week | Demand | | | |
|------|--------|--------|--------|--|
| 1 | 650 | | | |
| 2 | 678 | | | |
| 3 | 720 | | | |
| 4 | 785 | 682.67 | | |
| 5 | 859 | 727.67 | | |
| 6 | 920 | 788.00 | | |
| 7 | 850 | 854.67 | 768.67 | |
| 8 | 758 | 876.33 | 802.00 | |
| 9 | 892 | 842.67 | 815.33 | |
| 10 | 920 | 833.33 | 844.00 | |
| 11 | 789 | 856.67 | 866.50 | |
| 12 | 844 | 867.00 | 854.83 | |

$$F_4 = (650 + 678 + 720) / 3$$

$$= 682.67$$

$$F_7 = (650 + 678 + 720 + 785 + 859 + 920) / 6$$

$$= 768.67$$

Types of forecasting methods

Quantitative Methods

Ex: 2

| Week | Demand |
|------|--------|
| 1 | 820 |
| 2 | 775 |
| 3 | 680 |
| 4 | 655 |
| 5 | 620 |
| 6 | 600 |
| 7 | 575 |

- Question: What is the 3 week moving average forecast for this data?
- Assume you only have 3 weeks and 5 weeks of actual demand data for the respective forecasts

Types of forecasting methods

Quantitative Methods

Solution

| Week | Demand | 3-Week | 5-Week |
|------|--------|--------|--------|
| 1 | 820 | | |
| 2 | 775 | | |
| 3 | 680 | | |
| 4 | 655 | 758.33 | |
| 5 | 620 | 703.33 | |
| 6 | 600 | 651.67 | 710.00 |
| 7 | 575 | 625.00 | 666.00 |

Types of forecasting methods

Quantitative Methods

Ex: 3

| MONTH | ACTUAL SHED SALES | 3-MONTH MOVING AVERAGE |
|-----------|-------------------|------------------------|
| January | 10 | |
| February | 12 | |
| March | 13 | |
| April | 16 | |
| May | 19 | |
| June | 23 | |
| July | 26 | |
| August | 30 | |
| September | 28 | |
| October | 18 | |
| November | 16 | |
| December | 14 | |

Types of forecasting methods

Quantitative Methods

Solution


| MONTH | ACTUAL SHED SALES | 3-MONTH MOVING AVERAGE |
|-----------|-------------------|-------------------------------------|
| January | 10 | |
| February | 12 | |
| March | 13 | |
| April | 16 | $(10 + 12 + 13)/3 = 11 \frac{2}{3}$ |
| May | 19 | $(12 + 13 + 16)/3 = 13 \frac{2}{3}$ |
| June | 23 | $(13 + 16 + 19)/3 = 16$ |
| July | 26 | $(16 + 19 + 23)/3 = 19 \frac{1}{3}$ |
| August | 30 | $(19 + 23 + 26)/3 = 22 \frac{2}{3}$ |
| September | 28 | $(23 + 26 + 30)/3 = 26 \frac{1}{3}$ |
| October | 18 | $(29 + 30 + 28)/3 = 28$ |
| November | 16 | $(30 + 28 + 18)/3 = 25 \frac{1}{3}$ |
| December | 14 | $(28 + 18 + 16)/3 = 20 \frac{2}{3}$ |

Types of forecasting methods

Quantitative Methods

Ex: 4

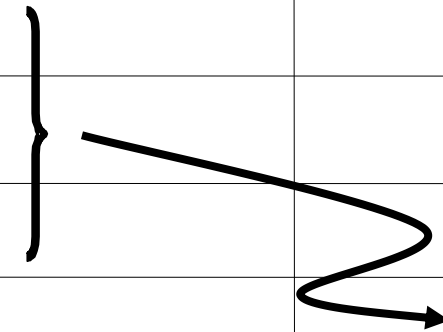
You're manager in Amazon's electronics department. You want to forecast ipod sales for months 4-6 using a 3-period moving average.

| Month | Sales (000) | |
|-------|-------------|--|
| 1 | 4 | $F_{t+1} = \frac{A_t + A_{t-1} + A_{t-2} + \dots + A_{t-n+1}}{n}$  |
| 2 | 6 | |
| 3 | 5 | |
| 4 | ? | |
| 5 | ? | |
| 6 | ? | |

Types of forecasting methods

Quantitative Methods

| Month | Sales (000) | Moving Average (n=3) |
|-------|----------------|-------------------------|
| 1 | 4 | NA |
| 2 | 6 | NA |
| 3 | 5 | NA |
| 4 | ? | (4+6+5)/3=5 |
| 5 | ? | |
| 6 | ? | |



$$F_{t+1} = \frac{A_t + A_{t-1} + A_{t-2} + \dots + A_{t-n+1}}{n}$$

Types of forecasting methods

Quantitative Methods

What if ipod sales were actually 3 in month 4?

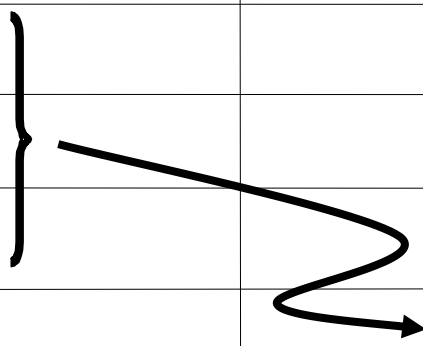
| Month | Sales (000) | Moving Average (n=3) |
|-------|----------------|-------------------------|
| 1 | 4 | NA |
| 2 | 6 | NA |
| 3 | 5 | NA |
| 4 | ? 3 | 5 |
| 5 | ? | |
| 6 | ? | |

Types of forecasting methods

Quantitative Methods

Forecast for Month 5?

| Month | Sales (000) | Moving Average (n=3) |
|-------|----------------|-------------------------|
| 1 | 4 | NA |
| 2 | 6 | NA |
| 3 | 5 | NA |
| 4 | 3 | 5 |
| 5 | ? | $(6+5+3)/3=4.667$ |
| 6 | ? | |



Types of forecasting methods

Quantitative Methods

2b. Weighted Moving Average

Simple moving average models weight all previous periods equally

Weighted Moving Average Gives more emphasis to recent data

The formula for the moving average is:

$$F_{t+1} = w_1 A_t + w_2 A_{t-1} + w_3 A_{t-2} + \dots + w_n A_{t-n+1}$$

Weights decrease for older data

$$\sum_{i=1}^n w_i = 1$$

Types of forecasting methods

Quantitative Methods

Ex: 1

Given the weekly demand and weights, what is the forecast for the 4th period or Week 4?

| Week | Demand |
|------|--------|
| 1 | 650 |
| 2 | 678 |
| 3 | 720 |
| 4 | |

Weights:

t-1 .5

t-2 .3

t-3 .2

Note that the weights place more emphasis on the most recent data, that is time period “t-1”.

Types of forecasting methods

Quantitative Methods

Solution

| Week | Demand | Forecast |
|------|--------|----------|
| 1 | 650 | |
| 2 | 678 | |
| 3 | 720 | |
| 4 | | 693.4 |

$$F_4 = 0.5(720) + 0.3(678) + 0.2(650) = 693.4$$

Types of forecasting methods

Quantitative Methods

Ex: 2

Question: Given the weekly demand information and weights, what is the weighted moving average forecast of the 5th period or week?

| Week | Demand |
|------|--------|
| 1 | 820 |
| 2 | 775 |
| 3 | 680 |
| 4 | 655 |

Weights:

t-1 .7

t-2 .2

t-3 .1

Types of forecasting methods

Quantitative Methods

Solution

| Week | Demand | Forecast |
|------|--------|----------|
| 1 | 820 | |
| 2 | 775 | |
| 3 | 680 | |
| 4 | 655 | |
| 5 | | 672 |

$$F_5 = (0.1)(755) + (0.2)(680) + (0.7)(655) = 672$$

Types of forecasting methods

Quantitative Methods

Ex: 3

| MONTH | ACTUAL SHED SALES | 3-MONTH WEIGHTED MOVING AVERAGE |
|-----------|--|---|
| January | 10 | |
| February | 12 | |
| March | 13 | |
| April | 16 | $[(3 \times 13) + (2 \times 12) + (10)]/6 = 12 \frac{1}{6}$ |
| May | WEIGHTS APPLIED | PERIOD |
| June | 3 | Last month |
| July | 2 | Two months ago |
| August | 1 | Three months ago |
| September | 6 | Sum of the weights |
| October | | Forecast for this month = |
| November | $3 \times \text{Sales last mo.} + 2 \times \text{Sales 2 mos. ago} + 1 \times \text{Sales 3 mos. ago}$ | |
| December | Sum of the weights | |

Weighted Moving Average

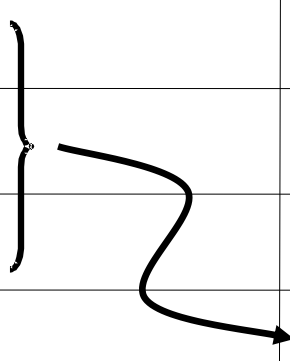
| MONTH | ACTUAL SHED SALES | 3-MONTH WEIGHTED MOVING AVERAGE |
|-----------|-------------------|---|
| January | 10 | |
| February | 12 | |
| March | 13 | |
| April | 16 | $[(3 \times 13) + (2 \times 12) + (10)]/6 = 12 \frac{1}{6}$ |
| May | 19 | $[(3 \times 16) + (2 \times 13) + (12)]/6 = 14 \frac{1}{3}$ |
| June | 23 | $[(3 \times 19) + (2 \times 16) + (13)]/6 = 17$ |
| July | 26 | $[(3 \times 23) + (2 \times 19) + (16)]/6 = 20 \frac{1}{2}$ |
| August | 30 | $[(3 \times 26) + (2 \times 23) + (19)]/6 = 23 \frac{5}{6}$ |
| September | 28 | $[(3 \times 30) + (2 \times 26) + (23)]/6 = 27 \frac{1}{2}$ |
| October | 18 | $[(3 \times 28) + (2 \times 30) + (26)]/6 = 28 \frac{1}{3}$ |
| November | 16 | $[(3 \times 18) + (2 \times 28) + (30)]/6 = 23 \frac{1}{3}$ |
| December | 14 | $[(3 \times 16) + (2 \times 18) + (28)]/6 = 18 \frac{2}{3}$ |

Types of forecasting methods

Quantitative Methods

Ex: 4 Weighted Moving Average: 3/6, 2/6, 1/6

| Month | Sales (000) | Weighted Moving Average |
|-------|----------------|-------------------------------|
| 1 | 4 | NA |
| 2 | 6 | NA |
| 3 | 5 | NA |
| 4 | ? | 31/6 = 5.167 |
| 5 | ? | |
| 6 | ? | |



$F_{t+1} = w_1 A_t + w_2 A_{t-1} + w_3 A_{t-2} + \dots + w_n A_{t-n+1}$

Types of forecasting methods

Quantitative Methods

$$F_{t+1} = w_1 A_t + w_2 A_{t-1} + w_3 A_{t-2} + \dots + w_n A_{t-n+1}$$

| Month | Sales (000) | Weighted Moving Average |
|-------|----------------|-------------------------------|
| 1 | 4 | NA |
| 2 | 6 | NA |
| 3 | 5 | NA |
| 4 | 3 | $31/6 = 5.167$ |
| 5 | 7 | $25/6 = 4.167$ |
| 6 | | $32/6 = 5.333$ |

Types of forecasting methods

Quantitative Methods

3a. Exponential Smoothing

- Assumes the most recent observations have the highest predictive value
- gives more weight to recent time periods

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

e_t

F_{t+1} = Forecast value for time $t+1$

A_t = Actual value at time t

α = Smoothing constant

Need initial forecast F_t to start.

Types of forecasting methods

Quantitative Methods

- Form of weighted moving average
 - Weights decline exponentially
 - Most recent data weighted most
- Requires smoothing constant (α)
 - Ranges from 0 to 1
 - Subjectively chosen
- Involves little record keeping of past data

Types of forecasting methods

Quantitative Methods

Ex: 4

| Week | Demand |
|------|--------|
| 1 | 820 |
| 2 | 775 |
| 3 | 680 |
| 4 | 655 |
| 5 | 750 |
| 6 | 802 |
| 7 | 798 |
| 8 | 689 |
| 9 | 775 |
| 10 | |

- Question: Given the weekly demand data, what are the exponential smoothing forecasts for periods 2-10 using $\alpha=0.10$ and $\alpha=0.60$?
- Assume $F_1=D_1$

Types of forecasting methods

Quantitative Methods

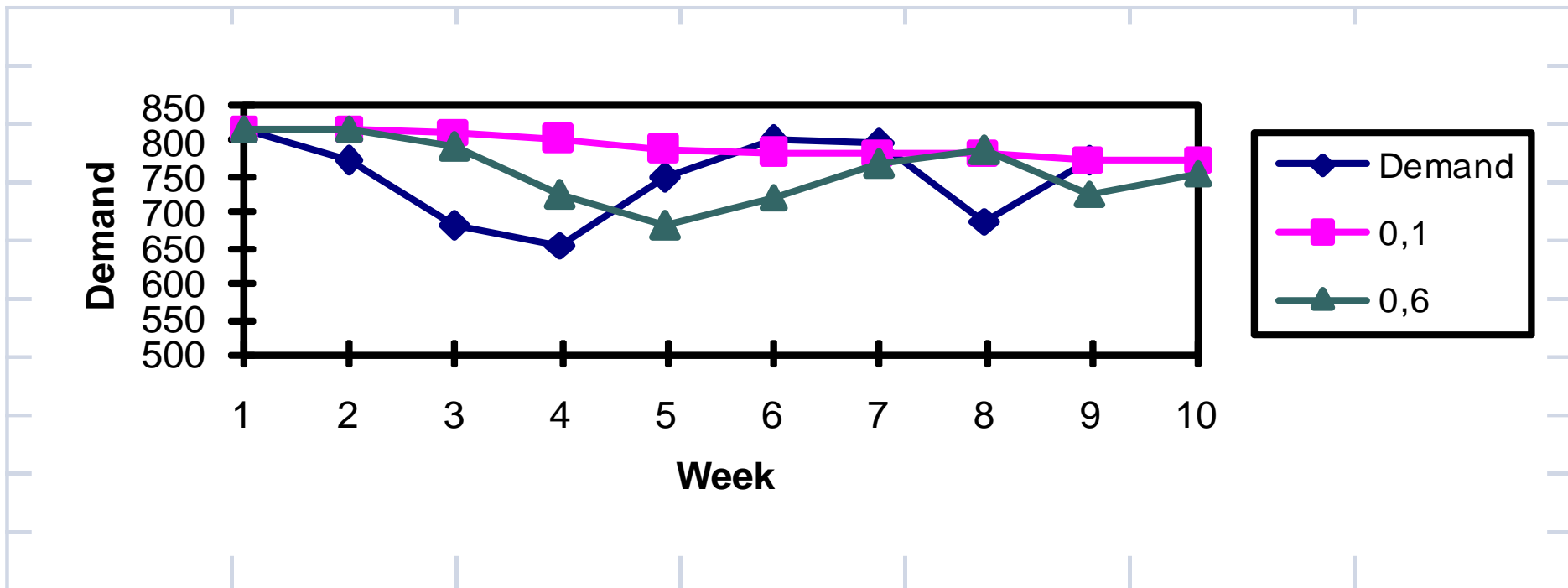
Solution

| Week | Demand | <i>0.1</i> | <i>0.6</i> |
|------|--------|------------|------------|
| 1 | 820 | 820.00 | 820.00 |
| 2 | 775 | 820.00 | 820.00 |
| 3 | 680 | 815.50 | 820.00 |
| 4 | 655 | 801.95 | 817.30 |
| 5 | 750 | 787.26 | 808.09 |
| 6 | 802 | 783.53 | 795.59 |
| 7 | 798 | 785.38 | 788.35 |
| 8 | 689 | 786.64 | 786.57 |
| 9 | 775 | 776.88 | 786.61 |
| 10 | | 776.69 | 780.77 |

Types of forecasting methods

Quantitative Methods

Exponential Smoothing Problem Plotting



Note how that the smaller alpha the smoother the line in this example.

Types of forecasting methods

Quantitative Methods

Ex: 4

| Week | Demand |
|------|--------|
| 1 | 820 |
| 2 | 775 |
| 3 | 680 |
| 4 | 655 |
| 5 | |

Question: What are the exponential smoothing forecasts for periods 2-5 using $\alpha = 0.5$?

Assume $F_1 = D_1$

Types of forecasting methods

Quantitative Methods

$$F_1 = 820 + (0.5)(820 - 820) = 820$$

$$F_3 = 820 + (0.5)(775 - 820) = 797.75$$

| Week | Demand | 0.5 |
|------|--------|--------|
| 1 | 820 | 820.00 |
| 2 | 775 | 820.00 |
| 3 | 680 | 797.50 |
| 4 | 655 | 738.75 |
| 5 | | 696.88 |

Exponential Smoothing Example

Ex: 4

Predicted demand = 142 Ford

Mustangs

Actual demand = 153

Smoothing constant $a = .20$


$$\text{New forecast} = 142 + .2(153 - 142)$$

$$= 142 + 2.2$$

$$= 144.2 \approx 144$$

cars

Effect of Smoothing Constants

- Smoothing constant generally $.05 \leq \alpha \leq .50$
- As α increases, older values become less significant

| WEIGHT ASSIGNED TO | | | | | |
|--------------------|------------------------------------|--|--|--|--|
| SMOOTHING CONSTANT | MOST RECENT PERIOD (α) | 2 ND MOST RECENT PERIOD $\alpha(1 - \alpha)$ | 3 RD MOST RECENT PERIOD $\alpha(1 - \alpha)^2$ | 4 th MOST RECENT PERIOD $\alpha(1 - \alpha)^3$ | 5 th MOST RECENT PERIOD $\alpha(1 - \alpha)^4$ |
| $\alpha = .1$ | .1 | .09 | .081 | .073 | .066 |
| $\alpha = .5$ | .5 | .25 | .125 | .063 | .031 |

Ex: 5

i

Ai

| Week | Demand |
|------|--------|
| 1 | 820 |
| 2 | 775 |
| 3 | 680 |
| 4 | 655 |
| 5 | 750 |
| 6 | 802 |
| 7 | 798 |
| 8 | 689 |
| 9 | 775 |
| 10 | |

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

Given the weekly demand data what are the exponential smoothing forecasts for periods 2-10 using $\alpha = 0.10$?

Assume $F_1 = D_1$

Types of forecasting methods

Quantitative Methods

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

| i | A _i | F _i |
|------|----------------|----------------|
| Week | Demand | $\alpha = 0.1$ |
| 1 | 820 | 820.00 |
| 2 | 775 | |
| 3 | 680 | |
| 4 | 655 | |
| 5 | 750 | |
| 6 | 802 | |
| 7 | 798 | |
| 8 | 689 | |
| 9 | 775 | |
| 10 | | |

$$F_2 = F_1 + \alpha(A_1 - F_1) = 820 + .1(820 - 820) = 820$$

Types of forecasting methods

Quantitative Methods

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

| i | Ai | Fi |
|------|--------|----------------|
| Week | Demand | $\alpha = 0.1$ |
| 1 | 820 | 820.00 |
| 2 | 775 | 820.00 |
| 3 | 680 | |
| 4 | 655 | |
| 5 | 750 | |
| 6 | 802 | |
| 7 | 798 | |
| 8 | 689 | |
| 9 | 775 | |
| 10 | | |

$$F_3 = F_2 + \alpha(A_2 - F_2) = 820 + .1(775 - 820) = 815.5$$

Types of forecasting methods

Quantitative Methods

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

| i | Ai | Fi |
|------|--------|----------------|
| Week | Demand | $\alpha = 0.1$ |
| 1 | 820 | 820.00 |
| 2 | 775 | 820.00 |
| 3 | 680 | 815.50 |
| 4 | 655 | |
| 5 | 750 | |
| 6 | 802 | |
| 7 | 798 | |
| 8 | 689 | |
| 9 | 775 | |
| 10 | | |

This process
continues
through week 10

Types of forecasting methods

Quantitative Methods

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

| i | Ai | Fi | $\alpha = 0.1$ | $\alpha = 0.6$ |
|------|--------|--------|----------------|----------------|
| Week | Demand | | | |
| 1 | 820 | 820.00 | 820.00 | 820.00 |
| 2 | 775 | 820.00 | 820.00 | 820.00 |
| 3 | 680 | 815.50 | 815.50 | 793.00 |
| 4 | 655 | 801.95 | 801.95 | 725.20 |
| 5 | 750 | 787.26 | 787.26 | 683.08 |
| 6 | 802 | 783.53 | 783.53 | 723.23 |
| 7 | 798 | 785.38 | 785.38 | 770.49 |
| 8 | 689 | 786.64 | 786.64 | 787.00 |
| 9 | 775 | 776.88 | 776.88 | 728.20 |
| 10 | | 776.69 | 776.69 | 756.28 |

What if the α constant equals 0.6

Types of forecasting methods

Quantitative Methods

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

| i | Ai | Fi | |
|-----------|--------|----------------|----------------|
| Month | Demand | $\alpha = 0.3$ | $\alpha = 0.6$ |
| January | 120 | 100.00 | 100.00 |
| February | 90 | 106.00 | 112.00 |
| March | 101 | 101.20 | 98.80 |
| April | 91 | 101.14 | 100.12 |
| May | 115 | 98.10 | 94.65 |
| June | 83 | 103.17 | 106.86 |
| July | | 97.12 | 92.54 |
| August | | | |
| September | | | |
| | | | |

What if the α constant equals 0.6

Ex: 6

Company A, a personal computer producer purchases generic parts and assembles them to final product. Even though most of the orders require customization, they have many common components. Thus, managers of Company A need a good forecast of demand so that they can purchase computer parts accordingly to minimize inventory cost while meeting acceptable service level. Demand data for its computers for the past 5 months is given in the following table.

Types of forecasting methods

Quantitative Methods

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

| i | A _i | F _i | |
|----------|----------------|----------------|----------------|
| Month | Demand | $\alpha = 0.3$ | $\alpha = 0.5$ |
| January | 80 | 84.00 | 84.00 |
| February | 84 | 82.80 | 82.00 |
| March | 82 | 83.16 | 83.00 |
| April | 85 | 82.81 | 82.50 |
| May | 89 | 83.47 | 83.75 |
| June | | 85.13 | 86.38 |
| July | | ?? | ?? |
| | | | |
| | | | |
| | | | |

What if the α constant equals 0.5

Types of forecasting methods

Quantitative Methods

4. Linear Regression Model

$$\hat{y} = a + bx$$

Fitting a trend line to historical data points to project into the medium to long-range

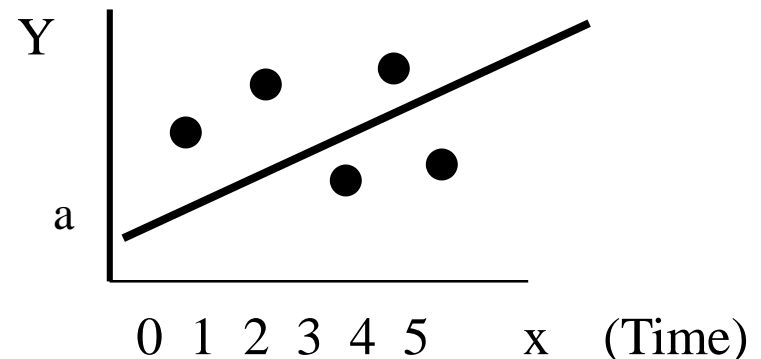
Linear trends can be found using the least squares technique

where y = computed value of the variable to be predicted
(dependent variable)

a = y -axis intercept

b = slope of the regression line

x = the independent variable



- **Y** is the regressed forecast value or dependent variable in the model,
- **a** is the intercept value of the the regression line, and
- **b** is similar to the slope of the regression line. However, since it is calculated with the variability of the data in mind, its formulation is not as straight forward as our usual notion of slope.

$$a = \bar{y} - b\bar{x}$$

$$b = \frac{\sum xy - n(\bar{y})(\bar{x})}{\sum x^2 - n(\bar{x})^2}$$

Question: Given the data below, what is the simple linear regression model that can be used to predict sales?

| Week | Sales |
|------|-------|
| 1 | 150 |
| 2 | 157 |
| 3 | 162 |
| 4 | 166 |
| 5 | 177 |

Answer: First, using the linear regression formulas, we can compute “a” and “b”.

| Week | Week*Week | Sales | Week*Sales |
|---------|-----------|---------|------------|
| 1 | 1 | 150 | 150 |
| 2 | 4 | 157 | 314 |
| 3 | 9 | 162 | 486 |
| 4 | 16 | 166 | 664 |
| 5 | 25 | 177 | 885 |
| 3 | 55 | 162.4 | 2499 |
| Average | Sum | Average | Sum |

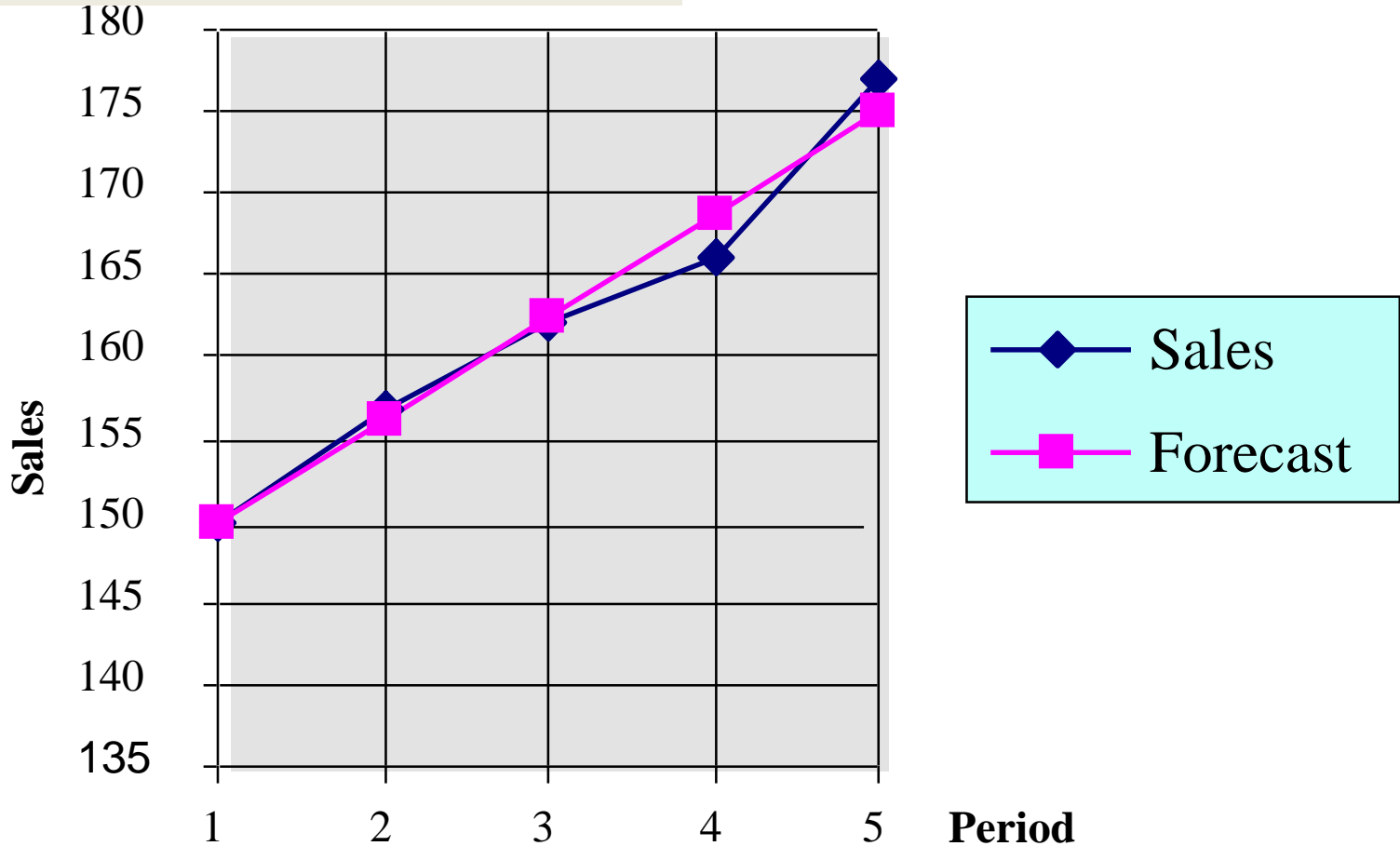
$$b = \frac{\sum xy - n(\bar{y})(\bar{x})}{\sum x^2 - n(\bar{x})^2} = \frac{2499 - 5(162.4)(3)}{55 - 5(9)} = \frac{63}{10} = \mathbf{6.3}$$

$$a = \bar{y} - b\bar{x} = 162.4 - (6.3)(3) = \mathbf{143.5}$$

The resulting regression model is:

$$Y_t = 143.5 + 6.3x$$

Now if we plot the regression generated forecasts against the actual sales we obtain the following chart:



EX 1

| YEAR | ELECTRICAL POWER DEMAND | YEAR | ELECTRICAL POWER DEMAND |
|------|-------------------------|------|-------------------------|
| 1 | 74 | 5 | 105 |
| 2 | 79 | 6 | 142 |
| 3 | 80 | 7 | 122 |
| 4 | 90 | | |

| YEAR (x) | ELECTRICAL POWER DEMAND (y) | x ² | xy |
|-----------------|-----------------------------|--------------------|---------------------|
| 1 | 74 | 1 | 74 |
| 2 | 79 | 4 | 158 |
| 3 | 80 | 9 | 240 |
| 4 | 90 | 16 | 360 |
| 5 | 105 | 25 | 525 |
| 6 | 142 | 36 | 852 |
| 7 | 122 | 49 | 854 |
| $\Sigma x = 28$ | $\Sigma y = 692$ | $\Sigma x^2 = 140$ | $\Sigma xy = 3,063$ |

$$\bar{x} = \frac{\sum x}{n} = \frac{28}{7} = 4 \quad \bar{y} = \frac{\sum y}{n} = \frac{692}{7} = 98.86$$

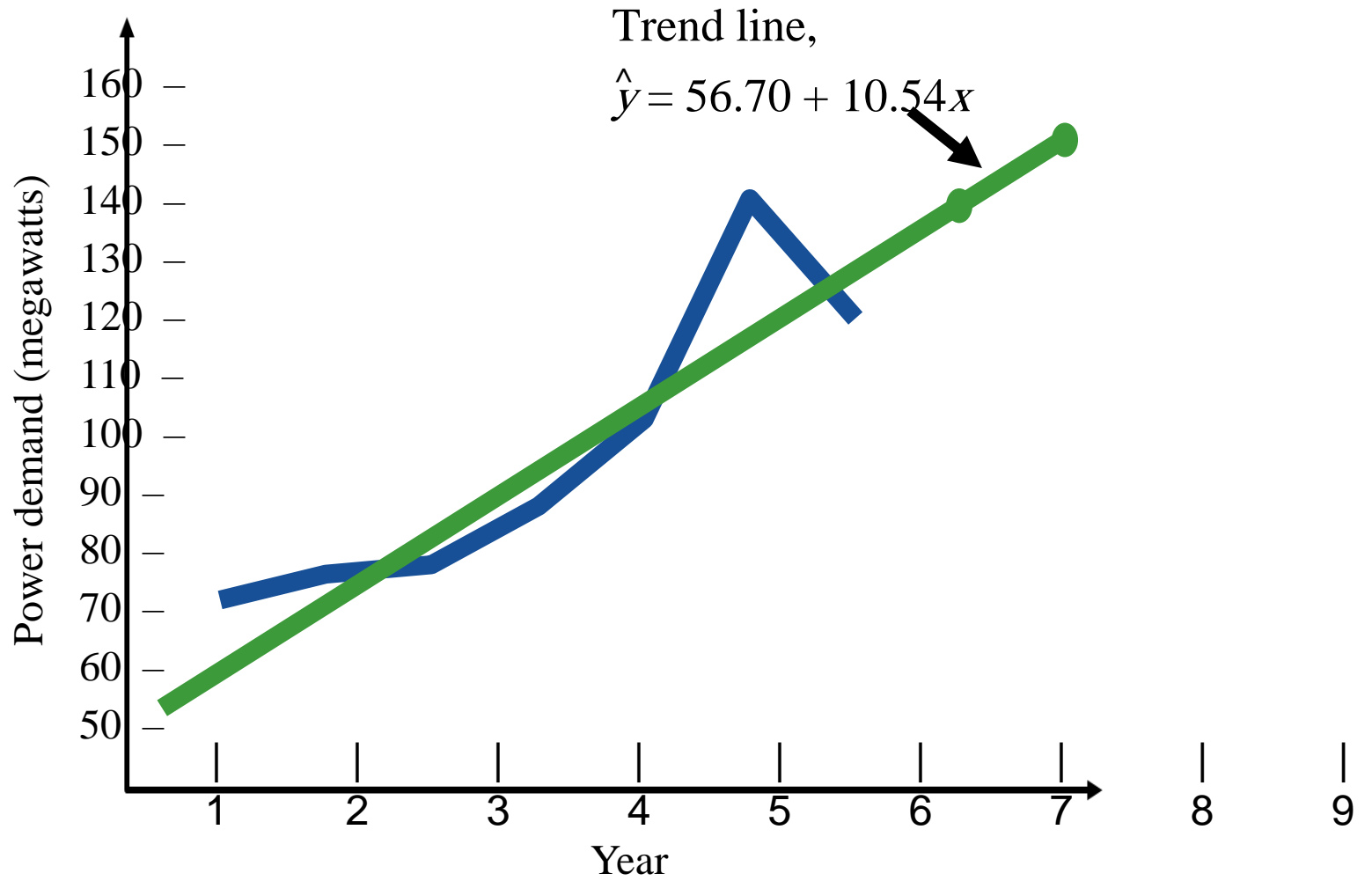
| YEAR (x) | ELECTRICAL POWER DEMAND (y) | x ² | xy |
|-----------------|-----------------------------|--------------------|---------------------|
| 7 | 122 | 49 | 854 |
| $\Sigma x = 28$ | $\Sigma y = 692$ | $\Sigma x^2 = 140$ | $\Sigma xy = 3,063$ |

$$b = \frac{\sum xy - n\bar{x}\bar{y}}{\sum x^2 - n\bar{x}^2} = \frac{3,063 - (7)(4)(98.86)}{140 - (7)(4^2)} = \frac{295}{28} = 10.54$$

$$a = \bar{y} - b\bar{x} = 98.86 - 10.54(4) = 56.70$$

$$\text{Thus, } \hat{y} = 56.70 + 10.54x$$

Demand in year 8 = $56.70 + 10.54(8)$
= 141.02, or 141 megawatts



To Use a Forecasting Method

- Collect historical data
- Select a model
 - Moving average methods
 - Select n (number of periods)
 - For weighted moving average: select weights
 - Exponential smoothing
 - Select α
- Selections should produce a good forecast
 - ...but what is a good forecast?

A Good Forecast

- ◆ Has a small error
 - ◆ $\text{Error} = \text{Demand} - \text{Forecast}$

The objective is to obtain the most accurate forecast no matter the technique

We generally do this by selecting the model that gives us the lowest forecast error

Ideal values =0 (i.e., no forecasting error)

Measures of Forecast Error

a. MAD = Mean Absolute Deviation

$$\text{MAD} = \frac{\sum_{t=1}^n \overbrace{|A_t - F_t|}^{e_t}}{n}$$

b. MSE = Mean Squared Error

$$\text{MSE} = \frac{\sum_{t=1}^n (A_t - F_t)^2}{n}$$

c. RMSE = Root Mean Squared Error

$$\text{RMSE} = \sqrt{\text{MSE}}$$

Common Measures of Error

Given the quarters demand data, what are the exponential smoothing forecasts for periods 2-9 using $\alpha = 0.10$? and $\alpha = 0.50$?

Assume $F_1 = D_1$

| QUARTER | ACTUAL TONNAGE UNLOADED |
|---------|-------------------------------|
| 1 | 180 |
| 2 | 168 |
| 3 | 159 |
| 4 | 175 |
| 5 | 190 |
| 6 | 205 |
| 7 | 180 |
| 8 | 182 |
| 9 | ? |

Common Measures of Error

$$F_{t+1} = F_t + a(A_t - F_t)$$

Mean Absolute Deviation (MAD)

$$MAD = \frac{\sum |\text{Actual} - \text{Forecast}|}{n}$$

| QUARTER | ACTUAL TONNAGE UNLOADED | FORECAST WITH $\alpha = .10$ | FORECAST WITH $\alpha = .50$ |
|---------|-------------------------|---------------------------------------|------------------------------|
| 1 | 180 | 175 | 175 |
| 2 | 168 | $175.50 = 175.00 + .10(180 - 175)$ | 177.50 |
| 3 | 159 | $174.75 = 175.50 + .10(168 - 175.50)$ | 172.75 |
| 4 | 175 | $173.18 = 174.75 + .10(159 - 174.75)$ | 165.88 |
| 5 | 190 | $173.36 = 173.18 + .10(175 - 173.18)$ | 170.44 |
| 6 | 205 | $175.02 = 173.36 + .10(190 - 173.36)$ | 180.22 |
| 7 | 180 | $178.02 = 175.02 + .10(205 - 175.02)$ | 192.61 |
| 8 | 182 | $178.22 = 178.02 + .10(180 - 178.02)$ | 186.30 |
| 9 | ? | $178.59 = 178.22 + .10(182 - 178.22)$ | 184.15 |

Determining the MAD

| QUARTER | ACTUAL TONNAGE UNLOADED | FORECAST WITH $\alpha = .10$ | ABSOLUTE DEVIATION FOR $\alpha = .10$ | FORECAST WITH $\alpha = .50$ | ABSOLUTE DEVIATION FOR $\alpha = .50$ |
|---|-------------------------|------------------------------|---------------------------------------|------------------------------|---------------------------------------|
| 1 | 180 | 175 | 5.00 | 175 | 5.00 |
| 2 | 168 | 175.50 | 7.50 | 177.50 | 9.50 |
| 3 | 159 | 174.75 | 15.75 | 172.75 | 13.75 |
| 4 | 175 | 173.18 | 1.82 | 165.88 | 9.12 |
| 5 | 190 | 173.36 | 16.64 | 170.44 | 19.56 |
| 6 | 205 | 175.02 | 29.98 | 180.22 | 24.78 |
| 7 | 180 | 178.02 | 1.98 | 192.61 | 12.61 |
| 8 | 182 | 178.22 | 3.78 | 186.30 | 4.30 |
| Sum of absolute deviations: | | | 82.45 | | 10.33 |
| MAD = $\frac{\Sigma \text{Deviations} }{n}$ | | | 10.31 | | 12.33 |

Determining the MSE

Mean Squared Error (MSE)

$$\text{MSE} = \frac{\sum_{t=1}^n (A_t - F_t)^2}{n}$$

| QUARTER | ACTUAL TONNAGE UNLOADED | FORECAST FOR $\alpha = .10$ | (ERROR) ² |
|---------|-------------------------|-----------------------------|-----------------------------------|
| 1 | 180 | 175 | 5 ² = 25 |
| 2 | 168 | 175.50 | (-7.5) ² = 56.25 |
| 3 | 159 | 174.75 | (-15.75) ² = 248.06 |
| 4 | 175 | 173.18 | (1.82) ² = 3.31 |
| 5 | 190 | 173.36 | (16.64) ² = 276.89 |
| 6 | 205 | 175.02 | (29.98) ² = 898.80 |
| 7 | 180 | 178.02 | (1.98) ² = 3.92 |
| 8 | 182 | 178.22 | (3.78) ² = 14.29 |
| | | | Sum=1526.52 |

$$\text{MSE} = \frac{\sum (\text{Forecast errors})^2}{n} = 1,526.52 / 8 = 190.8$$

Determining the MAPE

Mean Absolute
Percent Error

$$\text{MAPE} = \frac{\sum_{i=1}^n 100|\text{Actual}_i - \text{Forecast}_i| / \text{Actual}_i}{n}$$

| QUARTER | ACTUAL TONNAGE UNLOADED | FORECAST FOR $\alpha = .10$ | ABSOLUTE PERCENT ERROR 100(ERROR/ACTUAL) |
|---------|-------------------------|-----------------------------|---|
| 1 | 180 | 175.00 | 100(5/180) = 2.78% |
| 2 | 168 | 175.50 | 100(7.5/168) = 4.46% |
| 3 | 159 | 174.75 | 100(15.75/159) = 9.90% |
| 4 | 175 | 173.18 | 100(1.82/175) = 1.05% |
| 5 | 190 | 173.36 | 100(16.64/190) = 8.76% |
| 6 | 205 | 175.02 | 100(29.98/205) = 14.62% |
| 7 | 180 | 178.02 | 100(1.98/180) = 1.10% |
| 8 | 182 | 178.22 | 100(3.78/182) = 2.08% |
| | | | Sum of % errors = 44.75% |

$$\text{MAPE} = \frac{\sum \text{absolute percent error}}{n} = \frac{44.75\%}{8} = 5.59\%$$

Comparison of Forecast Error

| Quarter | Actual Tonnage Unloaded | Rounded Forecast with $\alpha = .10$ | Absolute Deviation for $\alpha = .10$ | Rounded Forecast with $\alpha = .50$ | Absolute Deviation for $\alpha = .50$ |
|---------|-------------------------|--------------------------------------|---------------------------------------|--------------------------------------|---------------------------------------|
| 1 | 180 | 175 | 5.00 | 175 | 5.00 |
| 2 | 168 | 175.5 | 7.50 | 177.50 | 9.50 |
| 3 | 159 | 174.75 | 15.75 | 172.75 | 13.75 |
| 4 | 175 | 173.18 | 1.82 | 165.88 | 9.12 |
| 5 | 190 | 173.36 | 16.64 | 170.44 | 19.56 |
| 6 | 205 | 175.02 | 29.98 | 180.22 | 24.78 |
| 7 | 180 | 178.02 | 1.98 | 192.61 | 12.61 |
| 8 | 182 | 178.22 | 3.78 | 186.30 | 4.30 |
| | | | <u>82.45</u> | | <u>98.62</u> |

MAD 10.31

MSE 190.82

MAPE 5.59%

12.33

195.24

6.76%