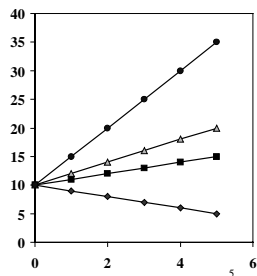


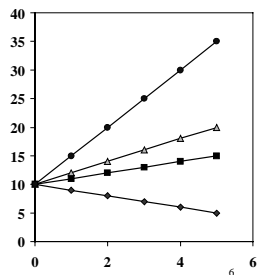
The Slope

- ⊕ The slope is how steep the line is
- ⊕ The slope is defined as the change in the Y axis value divided by the change in the X axis value
- ⊕ By just looking at the lines, which one has the steepest slope?



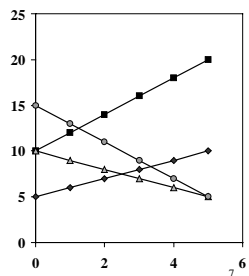
Slope

- ⊕ Look at the left-most two points
- ⊕ For the blue line the change in Y is $15 - 10 = 5$. The change in X is $1 - 0 = 1$. The slope is $5 / 1 = 5$
- ⊕ The slope of the green line is $(12 - 10) / (1 - 0) = 2$
- ⊕ Black's slope is 1
- ⊕ Red's slope is -1



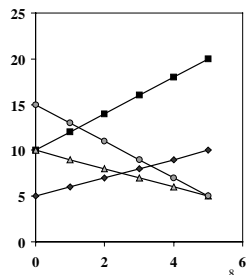
Intercept

- ⊕ The intercept is the Y axis value when X equals 0
- ⊕ It is where the line strikes the Y axis when $X = 0$
- ⊕ Blue's intercept is 15
- ⊕ Black and green's intercept is 10
- ⊕ Red's intercept is 5



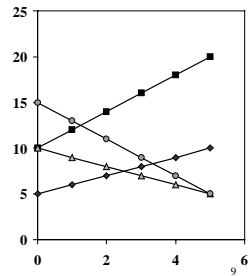
Equation of a Line

- ⊕ To determine the equation of the black line, first determine its slope and intercept
- ⊕ Slope = $(12-10)/(1-0) = 2$
- ⊕ Intercept = 10
- ⊕ $Y' = 2 * X + 10$



Equation of a Line

- ⊕ $Y' = 2 * X + 10$
- ⊕ What value of Y is predicted when the value of X = 5?
- ⊕ $Y' = 2 * 5 + 10 = 20$
- ⊕ Because the two variables are perfectly correlated, we can exactly predict the Y value given the X value



Regression When $|r| < 1.0$

- ⊕ When the two variables are not perfectly correlated with each other, the points in a scatterplot will not fall directly on a line
- ⊕ Thus, we will not be able to accurately predict the value of one variable given the value of the other variable
- ⊕ The closer $|r|$ is to 0, the less accurate our predictions will be

10

Determining Slope and Intercept when $|r| < 1.0$

- ⊕ How do we determine the equation of the line when the data points do not fall on a line?
- ⊕ We should try to find the line that does the best job of describing the data points
- ⊕ That line is called the *line of best fit*, the *regression line*, or the *least squares line*; all three terms are synonymous

11

Line of Best Fit

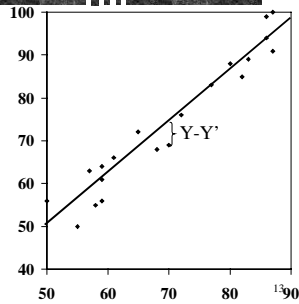
- ⊕ The line that we select as the regression line should minimize the errors that we make in our predictions
- ⊕ The error in our prediction is given by:

$$\sum (Y - Y')^2$$

12

$$\Sigma(Y - Y')^2$$

- ⊕ What does this formula say?
- ⊕ For each X, Y pair, calculate the predicted Y given X
- ⊕ Subtract the predicted from the observed
- ⊕ Square the difference
- ⊕ Sum the squared differences



Why Square Y - Y'?

- ⊕ You may wonder why we square the difference between the observed and predicted Y values
- ⊕ The regression line (the line containing all the Y' values) is similar to the mean
- ⊕ Recall that $\Sigma(X - \bar{X})^2$ was smaller than if we had substituted any other number for the mean
- ⊕ That is, the mean minimizes the sum 14

Why Square Y - Y'?

- ⊕ Thus, substituting Y' for the mean will make the squared errors smaller than if any other value was substituted

15

How To Determine the Slope

- ⊕ The slope of the regression line should be influenced by three factors:
 - ⊕ s_x
 - ⊕ s_y
 - ⊕ r

16

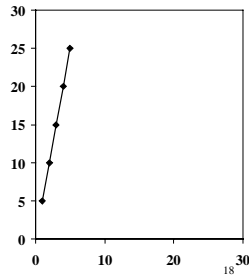
How To Determine the Slope

- ⊕ The two standard deviations basically serve to standardize the difference in the variations of the two distributions
- ⊕ The slope is proportional to the ratio:
 s_y / s_x
- ⊕ The next several slides assume that X and Y are perfectly correlated

17

How To Determine the Slope

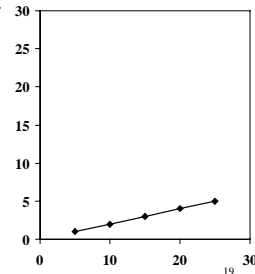
- ⊕ If the standard deviation of X is small relative to the standard deviation of Y, then a small change in X should lead to a larger change in Y



- ⊕ That is, the slope should be large (large ΔY / small ΔX)
- ⊕ $s_y / s_x = 7.07 / 1.41 = 5$

How To Determine the Slope

- ⊕ If the standard deviation of X is large compared to the standard deviation of Y, then a small change in X should lead to an even smaller change in Y



- ⊕ The slope should be small (smaller ΔY / small ΔX)
- ⊕ $s_y / s_x = 1.41 / 7.07 = 0.2$

How To Determine the Slope

- ⊕ The slope also depends on the correlation of the two variables
- ⊕ When the correlation is perfect, the slope is given by the ratio of the standard deviations
- ⊕ When no correlation exists, the best prediction is always the mean no matter what the value of X is
- ⊕ Thus, when $r = 0$, the slope should equal 0

20

How To Determine the Slope

- ⊕ When $|r|$ is between 0 and 1, the slope should be between 0 and s_y / s_x
- ⊕ The closer r is to 0, the closer the slope should be to 0
- ⊕ The closer $|r|$ is to 1, the closer the slope should be s_y / s_x
- ⊕ Thus, the slope is given by:
slope = $r * s_y / s_x$

21

Computational Formula for Slope

- ⊕ The computational formula for the slope of the regression line is:

$$\text{slope} = \frac{\sum XY - \frac{(\sum X)(\sum Y)}{N}}{\sum X^2 - \frac{(\sum X)^2}{N}}$$

22

How To Determine the Intercept

- ⊕ Given that $Y' = \text{slope} * X + \text{intercept}$, \bar{X} , \bar{Y} , and $r = 1$, with a little algebra, we can solve for the intercept
- ⊕ intercept = $\bar{Y} - \text{slope} * \bar{X}$

23

Types of Variation in Regression

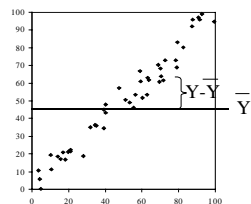
- ⊕ There are three types of variation that are often mentioned when regression is discussed:
 - ⊕ Total variation
 - ⊕ Explained variation
 - ⊕ Unexplained variation

24

Total Variation

- The total variation is identical to the variation of the variable being predicted

$$S^2 = \frac{\sum (Y - \bar{Y})^2}{N}$$

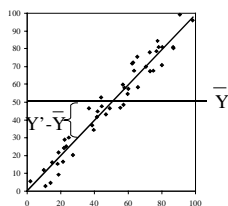


25

Explained Variation

- The explained variation is the variation in Y that is can be explained by the regression equation

$$\text{Explained } s^2 = \frac{\sum (Y' - \bar{Y})^2}{N}$$

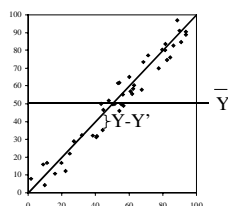


26

Unexplained Variation

- The unexplained variation is the variation in Y that cannot be explained by the regression equation

$$\text{Unexplained } s^2 = \frac{\sum (Y - Y')^2}{N}$$



27

Total Variation

- Total variation = explained variation + unexplained variation

$$\frac{\sum (Y - \bar{Y})^2}{N} = \frac{\sum (Y' - \bar{Y})^2}{N} + \frac{\sum (Y - Y')^2}{N}$$

28

Partitioning of the Variance

- ⊕ When we divide the total variance into two or more sub-totals, we are *partitioning the variance*
- ⊕ This concept of dividing the total variation into different categories becomes an essential aspect of one of the most important inferential statistics, the ANalysis Of VAriance (ANOVA)

29

Coefficient of Determination

- ⊕ The coefficient of determination, r^2 , was defined as the proportion of variation in the Y data that was explainable by variation in the X data
- ⊕ This can be given by the following formula

$$r^2 = \frac{\text{explained } S^2}{\text{total } S^2} = \frac{\frac{\sum(Y' - \bar{Y})^2}{N}}{\frac{\sum(Y - \bar{Y})^2}{N}}$$

30