## Economics 103 – Statistics for Economists

Francis J. DiTraglia

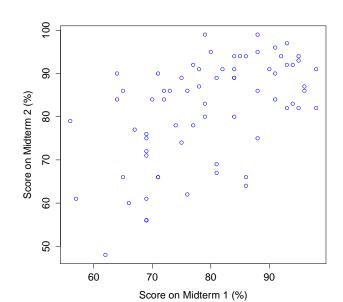
University of Pennsylvania

Lecture # 4

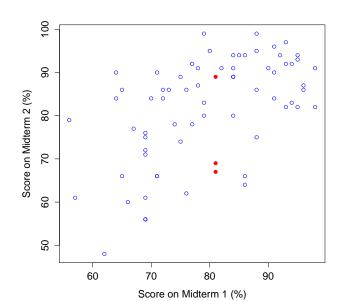
# Introduction to Regression

# Predict Second Midterm given 81 on First

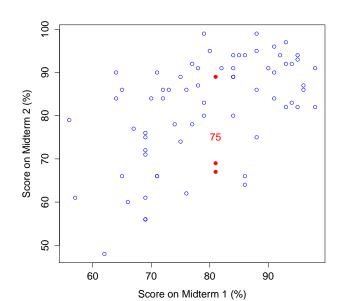




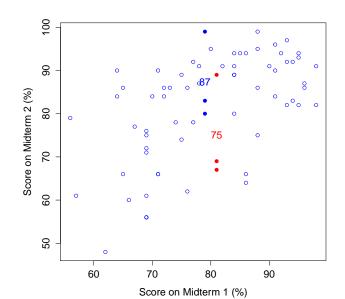
## Predict Second Midterm given 81 on First



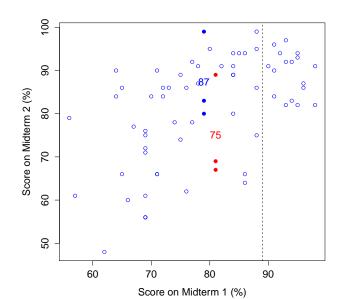
## Predict Second Midterm given 81 on First



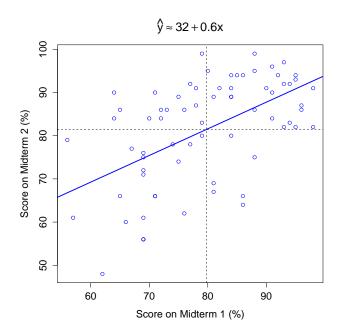
# But if they'd only gotten 79 we'd predict higher?!



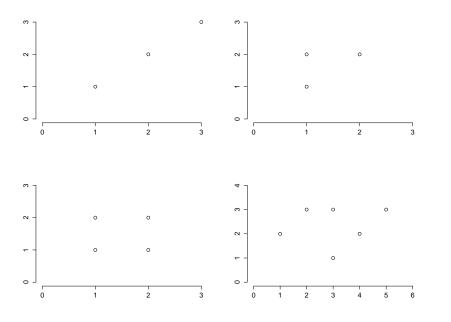
## No one who took both exams got 89 on the first!

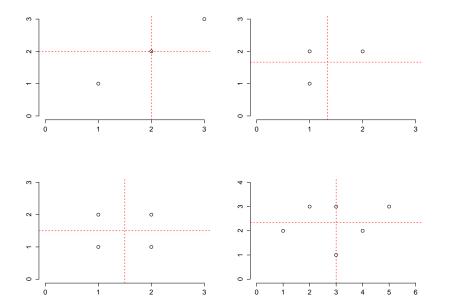


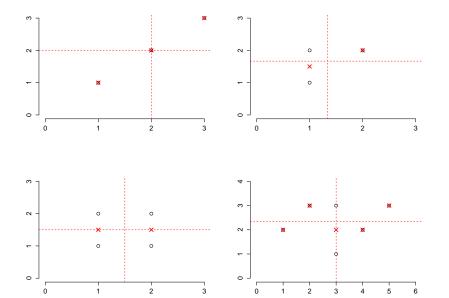
# Regression: "Best Fitting" Line Through Cloud of Points

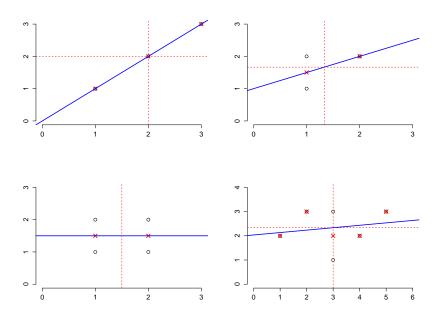


# Fitting a Line by Eye









# But How to Do this Formally?

## Least Squares Regression – Predict Using a Line

### The Prediction

Predict score  $\hat{y} = a + bx$  on 2nd midterm if you scored x on 1st

How to choose (a, b)?

Linear regression chooses the slope (b) and intercept (a) that minimize the sum of squared vertical deviations

$$\sum_{i=1}^{n} d_i^2 = \sum_{i=1}^{n} (y_i - a - bx_i)^2$$

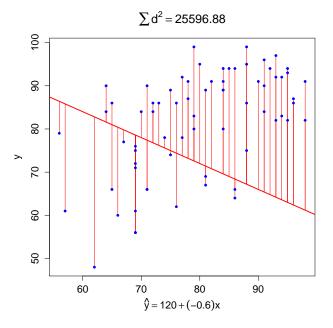
Why Squared Deviations?

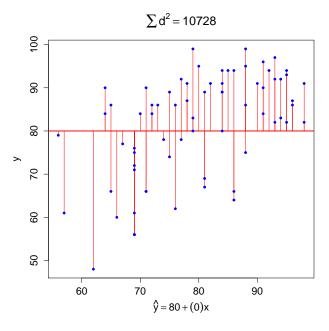
## Important Point About Notation

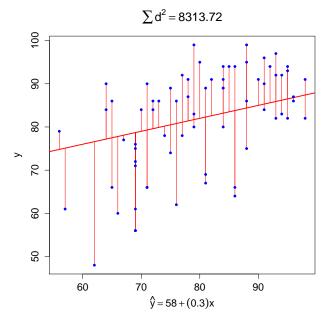
minimize 
$$\sum_{i=1}^{n} d_i^2 = \sum_{i=1}^{n} (y_i - a - bx_i)^2$$

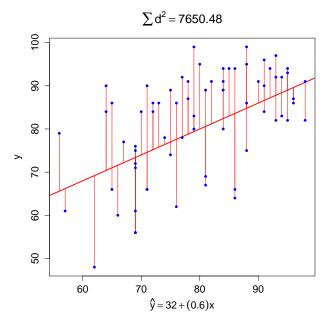
$$\hat{y} = a + bx$$

- $(x_i, y_i)_{i=1}^n$  are the observed data
- $\hat{y}$  is our prediction for a given value of x
- ▶ Neither x nor  $\hat{y}$  needs to be in out dataset!



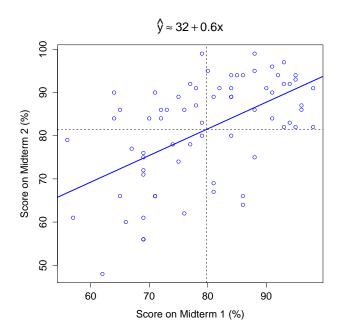




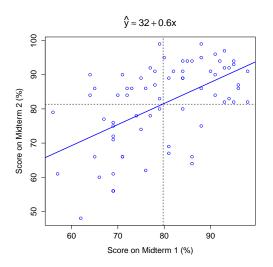


## Prediction given 89 on Midterm 1?





## Prediction given 89 on Midterm 1?



$$32 + 0.6 \times 89 = 32 + 53.4 = 85.4$$

## You Need to Know How To Derive This



## Minimize the sum of squared vertical deviations from the line:

$$\min_{a,b} \sum_{i=1}^{n} (y_i - a - bx_i)^2$$

How should we proceed?

- (a) Differentiate with respect to x
- (b) Differentiate with respect to y
- (c) Differentiate with respect to x, y
- (d) Differentiate with respect to a, b
- (e) Can't solve this with calculus.

## **Objective Function**

$$\min_{a,b} \sum_{i=1}^{n} (y_i - a - bx_i)^2$$

## FOC with respect to a

$$-2\sum_{i=1}^{n} (y_i - a - bx_i) = 0$$

$$\sum_{i=1}^{n} y_i - \sum_{i=1}^{n} a - b\sum_{i=1}^{n} x_i = 0$$

$$\frac{1}{n} \sum_{i=1}^{n} y_i - \frac{na}{n} - \frac{b}{n} \sum_{i=1}^{n} x_i = 0$$

$$\bar{y} - a - b\bar{x} = 0$$

# Regression Line Goes Through the Means!

$$ar{y} = a + bar{x}$$

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

$$\sum_{i=1}^{n} (y_i - a - bx_i)^2 = \sum_{i=1}^{n} (y_i - \bar{y} + b\bar{x} - bx_i)^2$$
$$= \sum_{i=1}^{n} [(y_i - \bar{y}) - b(x_i - \bar{x})]^2$$

#### FOC wrt b

$$-2\sum_{i=1}^{n} [(y_{i} - \bar{y}) - b(x_{i} - \bar{x})](x_{i} - \bar{x}) = 0$$

$$\sum_{i=1}^{n} (y_{i} - \bar{y})(x_{i} - \bar{x}) - b\sum_{i=1}^{n} (x_{i} - \bar{x})^{2} = 0$$

$$b = \frac{\sum_{i=1}^{n} (y_{i} - \bar{y})(x_{i} - \bar{x})}{\sum_{i=1}^{n} (x_{i} - \bar{x})^{2}}$$

# Simple Linear Regression

### **Problem**

$$\min_{a,b} \sum_{i=1}^{n} (y_i - a - bx_i)^2$$

### Solution

$$b = \frac{\sum_{i=1}^{n} (y_i - \bar{y}) (x_i - \bar{x})}{\sum_{i=1}^{n} (x_i - \bar{x})^2}$$
$$a = \bar{y} - b\bar{x}$$

# Relating Regression to Covariance and Correlation

$$b = \frac{\sum_{i=1}^{n} (y_i - \bar{y}) (x_i - \bar{x})}{\sum_{i=1}^{n} (x_i - \bar{x})^2} = \frac{\frac{1}{n-1} \sum_{i=1}^{n} (y_i - \bar{y}) (x_i - \bar{x})}{\frac{1}{n-1} \sum_{i=1}^{n} (x_i - \bar{x})^2} = \frac{s_{xy}}{s_x^2}$$

$$r = \frac{s_{xy}}{s_x s_y} = b \frac{s_x}{s_y}$$

## Comparing Regression, Correlation and Covariance

#### Units

Correlation is unitless, covariance and regression coefficients (a, b) are not. (What are the units of these?)

## Symmetry

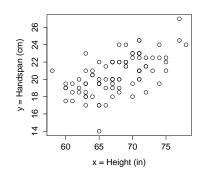
Correlation and covariance are symmetric, regression isn't. (Switching x and y axes changes the slope and intercept.)

#### On the Homework

Regression with z-scores rather than raw data gives  $a = 0, b = r_{xy}$ 

$$s_{xy} = 6$$
,  $s_x = 5$ ,  $s_y = 2$ ,  $\bar{x} = 68$ ,  $\bar{y} = 21$ 

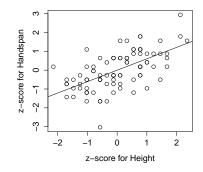
What is the sample correlation between height (x) and handspan (y)?





$$s_{xy} = 6$$
,  $s_x = 5$ ,  $s_y = 2$ ,  $\bar{x} = 68$ ,  $\bar{y} = 21$ 

What is the sample correlation between height (x) and handspan (y)?



$$r = \frac{s_{xy}}{s_x s_v} = \frac{6}{5 \times 2} = 0.6$$

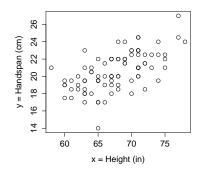


$$s_{xy} = 6$$
,  $s_x = 5$ ,  $s_y = 2$ ,  $\bar{x} = 68$ ,  $\bar{y} = 21$ 

What is the value of *b* for the regression:

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

where x is height and y is handspan?



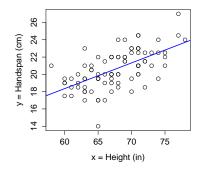


$$s_{xy} = 6$$
,  $s_x = 5$ ,  $s_y = 2$ ,  $\bar{x} = 68$ ,  $\bar{y} = 21$ 

What is the value of b for the regression:

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

where x is height and y is handspan?



$$b = \frac{s_{xy}}{s_{\star}^2} = \frac{6}{5^2} = 6/25 = 0.24$$

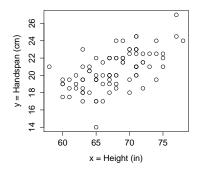


$$s_{xy} = 6$$
,  $s_x = 5$ ,  $s_y = 2$ ,  $\bar{x} = 68$ ,  $\bar{y} = 21$ 

What is the value of a for the regression:

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

where x is height and y is handspan? (prev. slide b = 0.24)



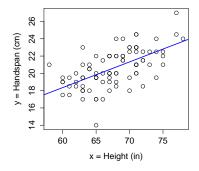


$$s_{xy} = 6$$
,  $s_x = 5$ ,  $s_y = 2$ ,  $\bar{x} = 68$ ,  $\bar{y} = 21$ 

What is the value of *a* for the regression:

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

where x is height and y is handspan? (prev. slide b = 0.24)



$$a = \bar{y} - b\bar{x} = 21 - 0.24 \times 68 = 4.68$$

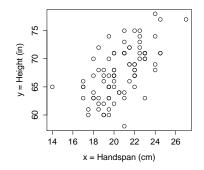


$$s_{xy} = 6$$
,  $s_y = 5$ ,  $s_x = 2$ ,  $\bar{y} = 68$ ,  $\bar{x} = 21$ 

What is the value of b for the regression:

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

where x is handspan and y is height?



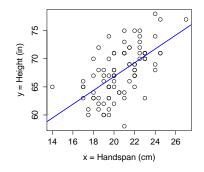


$$s_{xy} = 6$$
,  $s_y = 5$ ,  $s_x = 2$ ,  $\bar{y} = 68$ ,  $\bar{x} = 21$ 

What is the value of b for the regression:

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

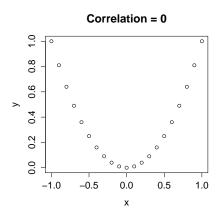
where x is handspan and y is height?



$$b = \frac{s_{xy}}{s_x^2} = 6/2^2 = 1.5$$

#### **EXTREMELY IMPORTANT**

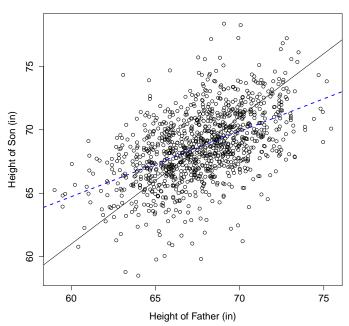
- ▶ Regression, Covariance and Correlation: linear association.
- ► Linear association ≠ causation.
- ▶ Linear is not the only kind of association!



# Regression to the Mean and the Regression Fallacy

Please read Chapter 17 of "Thinking Fast and Slow" by Daniel Kahnemann which I have posted on Piazza. This reading is fair game on an exam or quiz.

#### **Pearson Dataset**



## Regression to the Mean

Skill and Luck / Genes and Random Environmental Factors

Unless  $r_{xy} = 1$ , There Is Regression to the Mean

$$\frac{\hat{y} - \bar{y}}{s_y} = r_{xy} \frac{x - \bar{x}}{s_x}$$

Least-squares Prediction  $\hat{y}$  closer to  $\bar{y}$  than x is to  $\bar{x}$ 

You will derive the above formula in this week's homework.