# STA 674

Regression Analysis And Design Of Experiments

Fitting Simple Linear Regression Models – Lecture 4

### Fitting Simple Linear Regression Models

- Last time: finally got to the LS estimates of  $b_0$ ,  $b_1$ , and  $s_e^2$ .
- Before we get to the useful (interval) estimates of these, we need to do a cursory review of sampling distributions.

### Fitting Simple Linear Regression Models

#### Sampling distribution – what?

- Fact: If we repeat an experiment then we will get different estimates of the parameters because the errors will be different.
- For example:
  - If each of us repeated Hooker's (boiling temperature versus pressure) experiment then we'd all get slightly different estimates of  $b_0$  and  $b_1$ .
- Definition: The **sampling distribution** of a parameter estimate is the distribution of the values we would get from collecting many, many data sets and computing estimates for each

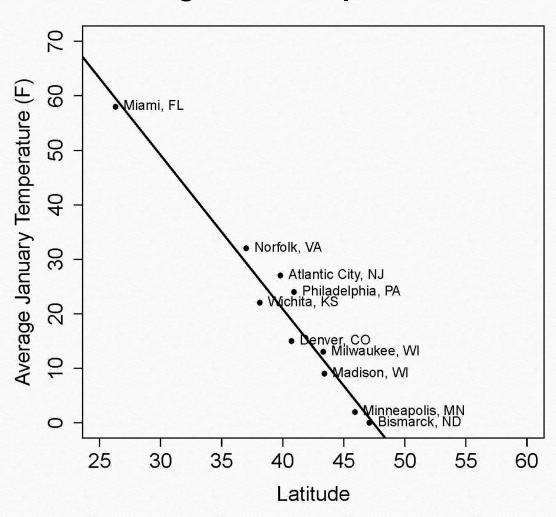
### Fitting Simple Linear Regression Models

### Sampling distribution – smaller example

- The original US temperature data contain 56 cities.
- The 10 that I chose formed one possible sample. There are over 3 billion possible samples of 10 cities that I could have chose.
- What would the results have been if I had chosen a different sample of 10 cities?

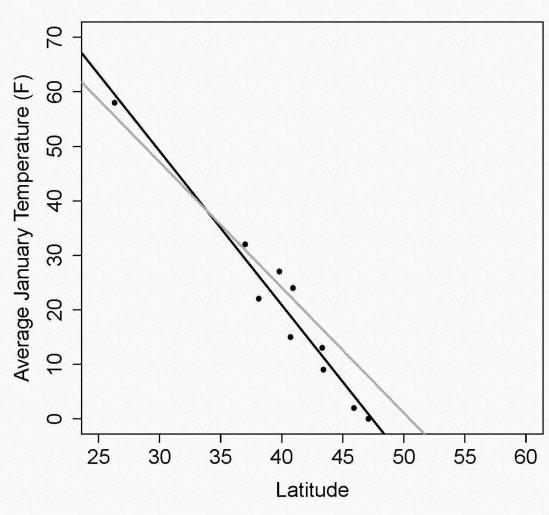
### Average Jan. Temp. vs Latitude

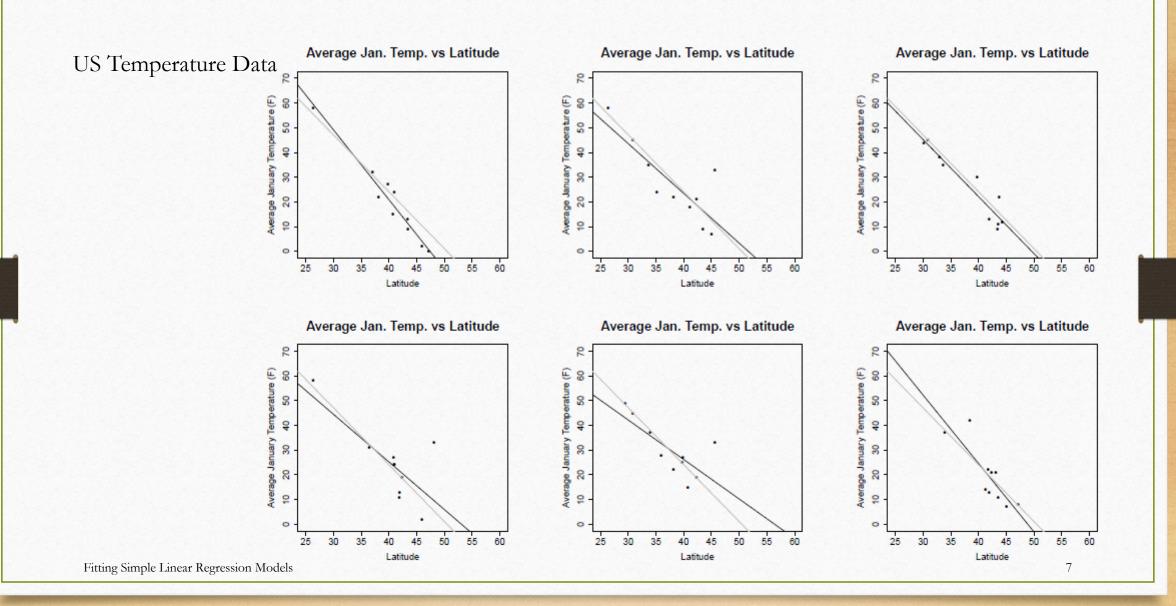
US Temperature Data



### Average Jan. Temp. vs Latitude

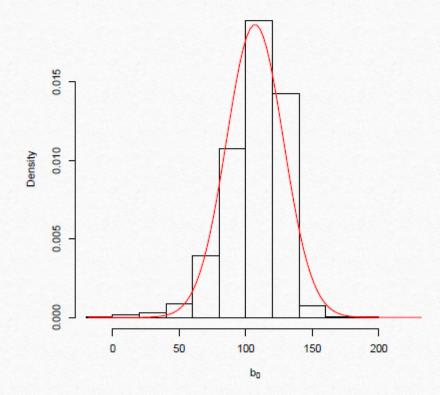
US Temperature Data



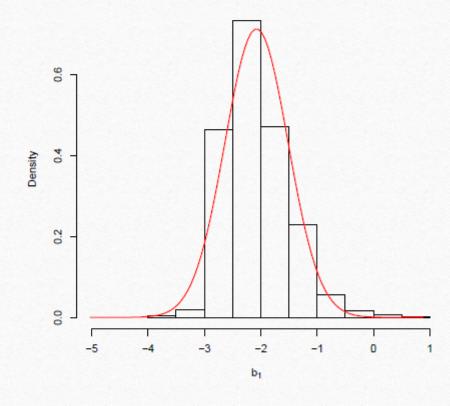


### US Temperature Data: Variation in LS estimates for 1000 samples of 10 cities

Histogram of b<sub>0</sub> from Repeated Samples



#### Histogram of b<sub>1</sub> from Repeated Samples



### Fitting Simple Linear Regression Models

### Least squares estimation - sampling distributions of the LS estimates

The assumptions of the regression model:

$$y_i = \beta_0 + \beta_1 x_i + e_i$$

Are that:

- 1. the expected value of the  $e_1, e_2, ..., e_n$  is 0.
- 2. the variance of the  $e_1, e_2, ..., e_n$  is  $\sigma_e^2$
- 3.  $e_1, e_2, ..., e_n$  are normally distributed
- 4.  $e_1, e_2, ..., e_n$  are independent

IF these assumptions are satisfied THEN ...

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IF these assumptions are satisfied THEN ...

$$b_0 \sim \text{Normal}\left(\beta_0, \sigma_e^2 \left(\frac{1}{n} + \frac{\bar{x}^2}{(n-1)s_x^2}\right)\right)$$

$$b_1 \sim \text{Normal}\left(\beta_1, \frac{\sigma_e^2}{(n-1)s_x^2}\right)$$