

Lecture 5  
Inference for Simple Linear Regression  
STAT 432, Spring 2021

A **parameter** is a numerical characteristic of a population (e.g., the population mean height  $\mu$  of all students at CSUEB)

**Statistical inference** refers to the process of using data collected from a sample to answer questions about population parameters.

- ▶ Point estimate: our best guess for the value of the population parameter (e.g., the sample mean height  $\bar{x}$  of  $n = 100$  randomly selected CSUEB students)
- ▶ Confidence interval: a plausible range of values for the population parameter
- ▶ Hypothesis test: is a specific value of the population parameter plausible?

## Simple linear regression model for the population:

$$y = \beta_0 + \beta_1 x + \epsilon$$

$\beta_0$  and  $\beta_1$  are the population parameters (fixed and unknown)

## Least squares line (estimated from the sample):

$$\hat{y} = \hat{\beta}_0 + \hat{\beta}_1 x$$

$\hat{\beta}_0$  and  $\hat{\beta}_1$  are the estimates (random, varies from sample to sample)



$1 - \alpha$  confidence interval for the slope  $\beta_1$ :

$$\hat{\beta}_1 \pm t_{\alpha/2; n-2} se(\hat{\beta}_1)$$

- ▶  $\hat{\beta}_1$  is the point estimate
- ▶  $t_{\alpha/2; n-2}$  is the t-critical value, with  $n - 2$  degrees of freedom
- ▶  $se(\hat{\beta}_1)$  is the standard error
- ▶  $1 - \alpha$  is the confidence level (e.g.,  $\alpha = 0.05$  for a 95% confidence interval)

Hypothesis test for whether or not the slope  $\beta_1$  is zero. We can also interpret this as a hypothesis test for whether or not there is a linear association between  $x$  and  $y$ .

$$H_0 : \beta_1 = 0$$

$$H_A : \beta_1 \neq 0$$

Test statistic:

$$t = \frac{\hat{\beta}_1}{se(\hat{\beta}_1)}; \quad df=n-2$$

Formulas for standard error computations (can use software for these computations):

- ▶ Residual standard error:

$$\hat{\sigma} = \sqrt{\frac{\sum_{i=1}^n \hat{e}_i^2}{n-2}} = \sqrt{\frac{\sum_{i=1}^n (y_i - \hat{y}_i)^2}{n-2}}$$

- ▶ Standard error of  $\hat{\beta}_1$ :

$$se(\hat{\beta}_1) = \frac{\hat{\sigma}}{\sqrt{\sum_{i=1}^n (x_i - \bar{x})^2}}$$

- ▶ Recall the conditions for simple linear regression:
  - ▶ **Linearity.** The data should follow a linear trend.
  - ▶ **Constant variability.** The variability of points around the least squares line remains roughly constant.
  - ▶ **Normality.** The residuals should be approximately normally distributed with mean 0.
  - ▶ **Independence.** Values of the response variable are independent of each other.
- ▶ When computing confidence intervals or conducting hypothesis tests for linear regression models it is important that the conditions are adequately statistified.
- ▶ Residuals plots are a useful way to check the conditions (especially linearity and constant variability). A histogram of the residuals can also be used to check normality.