

To Do

Read Sections 6.1 – 6.2.

**Assignment 4 is due Friday
November 25.**

Last Class

- (1) Distribution of $\tilde{\beta}$ the Maximum Likelihood Estimator of the Slope (with Proof)**
- (2) Distribution of S_e^2 the Estimator of σ^2 (no Proof)**

Today's Class

- (1) Confidence Interval for Slope β and Testing $H_0: \beta = \beta_0$
- (2) Test of No Relationship between Response and Explanatory Variates
- (3) Confidence interval for the mean response $\mu(x) = \alpha + \beta x$
- (4) Prediction Interval for an Individual Response Y

Theorem from Last Day

Since

$$\frac{\tilde{\beta} - \beta}{\sigma / \sqrt{S_{XX}}} \sim G(0,1) \quad \text{and} \quad \frac{(n-2)S_e^2}{\sigma^2} \sim \chi^2(n-2)$$

independently then

$$\frac{\tilde{\beta} - \beta}{S_e / \sqrt{S_{XX}}} \sim t(n-2)$$

This pivotal quantity can be used to construct confidence intervals and test hypotheses for β .

Inferences for the Slope β - Summary

A 100p% confidence interval for β is given by

$$\hat{\beta} \pm as_e / \sqrt{S_{XX}}$$

where $P(T \leq a) = (1 + p)/2$ and $T \sim t(n-2)$.

For testing $H_0: \beta = \beta_0$ the p -value is

$$p - value = 2 \left[1 - P \left(T \leq \frac{|\hat{\beta} - \beta_0|}{s_e / \sqrt{S_{XX}}} \right) \right]$$

where $T \sim t(n-2)$.

Hypothesis of No Relationship

Since $\mu(x) = \alpha + \beta x$, a test of

$$H_0: \beta = 0$$

is a test of the hypothesis that the mean $\mu(x)$ does not depend on x .

This hypothesis is usually referred to as “the hypothesis of no relationship” between the variates Y and x .

STAT 230 and 231 Final Grades

$$\bar{x} = 76.7333 \quad \bar{y} = 72.2333$$

$$S_{XX} = 5135.8667 \quad S_{XY} = 5106.8667 \quad S_{YY} = 7585.3667$$

$$\hat{\beta} = \frac{S_{XY}}{S_{XX}} = \frac{5106.8667}{5135.8667} = 0.9944$$

$$\hat{\alpha} = \bar{y} - \hat{\beta} \bar{x} = 72.2333 - \left(\frac{5106.8667}{5135.8667} \right) (76.7333) = 4.0667$$

$$s_e = \sqrt{\frac{1}{n-2} (S_{YY} - \hat{\beta} S_{XY})}$$

$$= \sqrt{\frac{1}{28} [7585.3667 - (0.9944)(5106.8667)]} = 9.4630$$

95% Confidence Interval for β

Since $P(T \leq 2.0484) = (1+0.95)/2 = 0.975$
where $T \sim t(28)$ a 95% confidence
interval for β is

$$\begin{aligned} & \hat{\beta} \pm 2.0484 s_e / \sqrt{S_{xx}} \\ &= 0.9944 \pm 2.0484(9.4630) / \sqrt{5135.8667} \end{aligned}$$

or **[0.7239, 1.2648]**.

95% Confidence Interval for β

Since the 95% confidence interval for β , $[0.7239, 1.2648]$ does not contain the value $\beta = 0$, the p -value for testing $H_0: \beta = 0$ is smaller than 0.05.

Therefore there is evidence against the hypothesis of no relationship between STAT 231 final grades and STAT 230 final grades.

***p*-value for Testing $H_0: \beta = 0$**

The actual *p*-value for testing $H_0: \beta = 0$ is

$$\begin{aligned} p\text{-value} &= 2 \left[1 - P \left(T \leq \frac{|\hat{\beta} - 0|}{s_e / \sqrt{S_{xx}}} \right) \right] \text{ where } T \sim t(28) \\ &= 2 \left[1 - P \left(T \leq \frac{0.9944}{9.4630 / \sqrt{5135.8667}} \right) \right] \\ &= 2[1 - P(T \leq 7.5304)] \approx 0 \end{aligned}$$

There is very strong evidence against the hypothesis of no relationship between STAT 230 final grades and STAT 231 final grades.

$$H_0: \beta = 1$$

What does the hypothesis

$H_0: \beta = 1$ represent?

$$H_0: \beta = 1$$

The parameter β represents the change in the mean STAT 231 final grade in the study population for a one mark increase in STAT 230 final grade.

The hypothesis $H_0: \beta = 1$ means that we are hypothesizing that in the study population for every one mark increase in STAT 230 final grade there is a one mark increase in the mean STAT 231 final grade.

$$H_0: \beta = 1$$

Since the 95% confidence interval,

$$[0.7239, 1.2648]$$

does contain the value $\beta = 1$, the p -value for testing $H_0: \beta = 1$ is larger than 0.05 and there is no evidence against the hypothesis $H_0: \beta = 1$.

$$H_0: \beta = 1$$

The actual p-value for testing $H_0: \beta = 1$ is

$$\begin{aligned} p\text{-value} &= 2 \left[1 - P \left(T \leq \frac{|\hat{\beta} - 1|}{s_e / \sqrt{S_{xx}}} \right) \right] \text{ where } T \sim t(28) \\ &= 2 \left[1 - P \left(T \leq \frac{|0.9944 - 1|}{9.4630 / \sqrt{5135.8667}} \right) \right] \\ &= 2[1 - P(T \leq 0.0428)] \end{aligned}$$

Since $P(T \leq 0.2558) = 0.6$, $p\text{-value} \geq 2(1 - 0.6) = 0.8$.

Confidence Interval for $\mu(x) = \alpha + \beta x$

Suppose we wanted a confidence interval for the mean STAT 231 final grade for students who obtained a final grade of 75 in STAT 230, that is, we want a confidence interval for $\mu(75) = \alpha + \beta(75)$.

More generally we are often interested in a confidence interval for the mean response $\mu(x) = \alpha + \beta x$ for a specified value of x .

Confidence Interval for $\mu(x) = \alpha + \beta x$

By the Invariance Property of Maximum Likelihood Estimates the maximum likelihood estimator of $\mu(x)$ is the random variable

$$\tilde{\mu}(x) = \tilde{\alpha} + \tilde{\beta}x = \bar{Y} + \tilde{\beta}(x - \bar{x})$$

We need the distribution of $\tilde{\mu}(x)$ to construct a confidence interval for the mean response $\mu(x) = \alpha + \beta x$.

Confidence Interval for $\mu(x) = \alpha + \beta x$

It can be shown that (with some effort!)

$$\begin{aligned}\tilde{\mu}(x) &= \tilde{\alpha} + \tilde{\beta}x = \bar{Y} + \tilde{\beta}(x - \bar{x}) \\ &= \sum_{i=1}^n \left[\frac{1}{n} + (x - \bar{x}) \frac{(x_i - \bar{x})}{S_{XX}} \right] Y_i\end{aligned}$$

where $Y_i \sim \mathbf{G}(\alpha + \beta x_i, \sigma)$ for $i=1,2,\dots,n$
independently

Distribution of $\tilde{\mu}(x)$

$$\tilde{\mu}(x) \sim G\left(\mu(x), \sigma \sqrt{\frac{1}{n} + \frac{(x - \bar{x})^2}{S_{xx}}}\right)$$

where

$$\tilde{\mu}(x) = \tilde{\alpha} + \tilde{\beta}x$$

and

$$\mu(x) = \alpha + \beta x$$

Distribution of $\tilde{\mu}(x)$

Equivalently

$$\frac{\tilde{\mu}(x) - \mu(x)}{\sigma \sqrt{\frac{1}{n} + \frac{(x - \bar{x})^2}{S_{xx}}}} \sim G(0,1)$$

where

$$\tilde{\mu}(x) = \tilde{\alpha} + \tilde{\beta}x$$

and

$$\mu(x) = \alpha + \beta x$$

Distribution of $\tilde{\mu}(x)$

Since we don't know σ we use

$$\frac{\tilde{\mu}(x) - \mu(x)}{S_e \sqrt{\frac{1}{n} + \frac{(x - \bar{x})^2}{S_{XX}}}} \sim t(n - 2)$$

to construct confidence intervals for $\mu(x) = \alpha + \beta x$.

Confidence Interval for $\mu(x) = \alpha + \beta x$

A $100p\%$ confidence interval for
 $\mu(x) = \alpha + \beta x$ = mean response at x is

$$\begin{aligned} \hat{\mu}(x) \pm as_e \sqrt{\frac{1}{n} + \frac{(x - \bar{x})^2}{S_{xx}}} \\ = \hat{\alpha} + \hat{\beta}x \pm as_e \sqrt{\frac{1}{n} + \frac{(x - \bar{x})^2}{S_{xx}}} \end{aligned}$$

where $P(T \leq a) = (1+p)/2$ and $T \sim t(n - 2)$.

Confidence Interval for the Intercept α

Since $\mu(0) = \alpha + \beta(0) = \alpha$, a 100p% confidence interval for α , is given by

$$\hat{\alpha} \pm as_e \sqrt{\frac{1}{n} + \frac{(\bar{x})^2}{S_{XX}}}$$

If \bar{x} is large in magnitude (which means the average x_i is large), then the confidence interval for α will be very wide.

This would be disturbing if the value $x = 0$ is a value of interest, but often it is not.

STAT 230/231 Example

Since $P(T \leq 2.0484) = 0.975$ where $T \sim t(28)$ a 95% confidence interval for the mean STAT 231 final grade for students who obtained a final grade of 75 in STAT 230 is

$$\begin{aligned} & \hat{\alpha} + \hat{\beta}x \pm as_e \sqrt{\frac{1}{n} + \frac{(x - \bar{x})^2}{S_{xx}}} \\ &= -4.0667 + 0.9944(75) \\ & \pm 2.0484(9.4630) \sqrt{\frac{1}{30} + \frac{(75 - 76.7333)^2}{5135.8667}} \\ &= 70.51 \pm 3.5699 \end{aligned}$$

or [66.9, 74.1]

Confidence Interval for an Individual Response Y at x

“It's all about me.”

Suppose we wanted an interval for Y = the STAT 231 mark for **one student who obtained a mark of $x = 75$ in STAT 230.**

Confidence Interval for an Individual Response Y at x

Let Y = potential observation for given value of x .

Since $Y \sim G(\alpha + \beta x, \sigma)$

and $\tilde{\mu}(x) \sim G\left(\alpha + \beta x, \sigma \sqrt{\frac{1}{n} + \frac{(x - \bar{x})^2}{S_{xx}}}\right)$

$$Y - \tilde{\mu}(x) \sim G\left(0, \sigma \sqrt{1 + \frac{1}{n} + \frac{(x - \bar{x})^2}{S_{xx}}}\right)$$

Confidence Interval for an Individual Response Y at x

Equivalently

$$\frac{Y - \tilde{\mu}(x)}{\sigma \sqrt{1 + \frac{1}{n} + \frac{(x - \bar{x})^2}{S_{XX}}}} \sim G(0,1)$$

Since we don't know σ we use

$$\frac{Y - \tilde{\mu}(x)}{S_e \sqrt{1 + \frac{1}{n} + \frac{(x - \bar{x})^2}{S_{XX}}}} \sim t(n-2)$$

100p% Prediction Interval for a Future Response Y

The corresponding interval is

$$\hat{\alpha} + \hat{\beta}x \pm s_e \sqrt{1 + \frac{1}{n} + \frac{(x - \bar{x})^2}{S_{xx}}}$$

where $P(T \leq a) = (1+p)/2$ and $T \sim t(n - 2)$.

The interval is called a **100p% prediction interval** instead of a confidence interval, since Y is not a parameter but a random variable.

STAT 230/231 Example

A 95% prediction interval for the STAT 231 mark for a randomly chosen student who obtained a mark of 75 in STAT 230 is

$$\begin{aligned} & \hat{\alpha} + \hat{\beta}x \pm as_e \sqrt{1 + \frac{1}{n} + \frac{(x - \bar{x})^2}{S_{xx}}} \\ &= -4.0667 + 0.9944(75) \\ & \quad \pm 2.0484(9.4630) \sqrt{1 + \frac{1}{30} + \frac{(75 - 76.7333)^2}{5135.8667}} \\ &= 70.51 \pm 19.7100 \end{aligned}$$

or **[50.8, 90.2]**

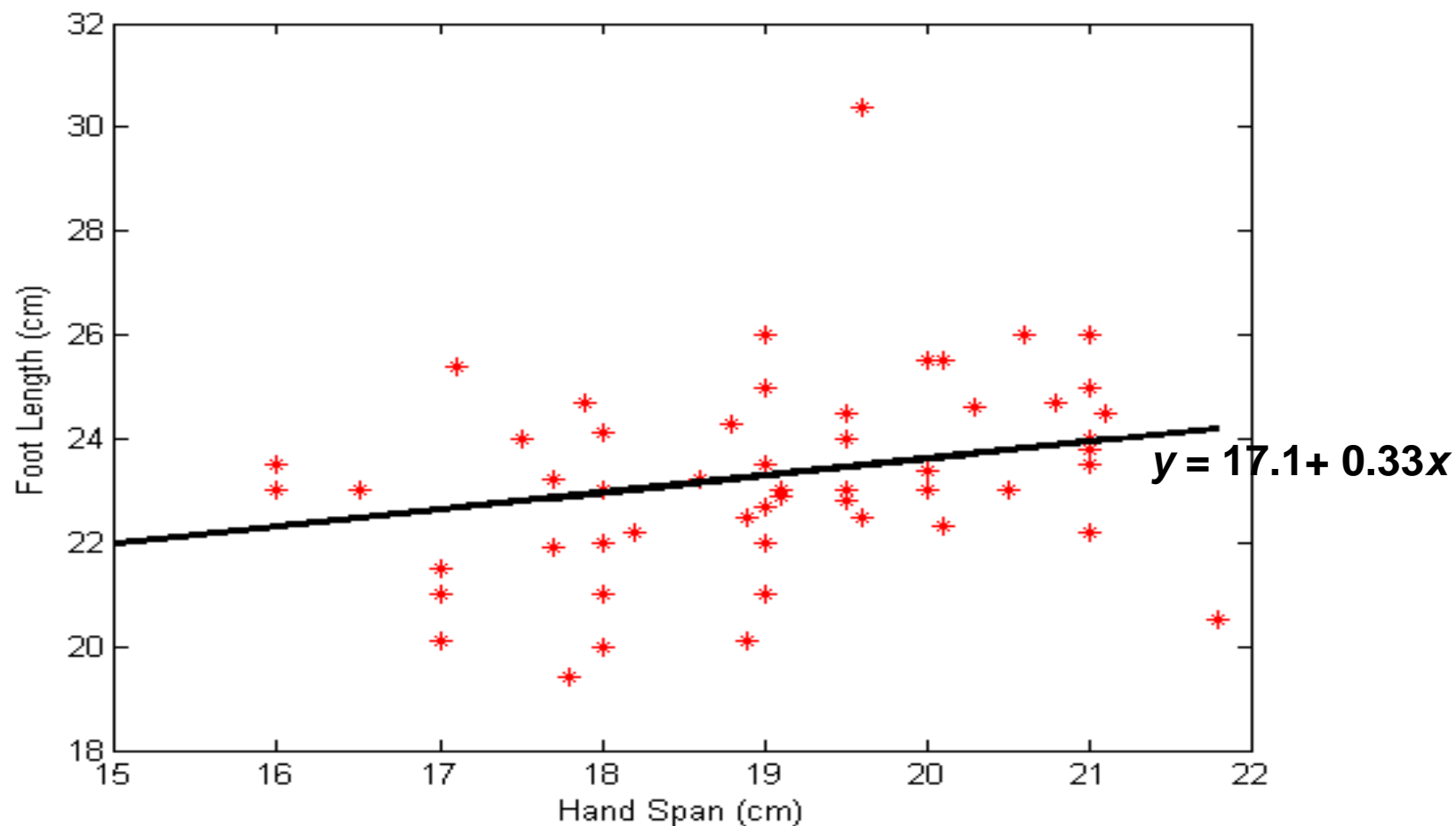
STAT 230/231 Example

A 95% prediction interval for the STAT 231 mark for a randomly chosen student who obtained a mark of 60 in STAT 230 is

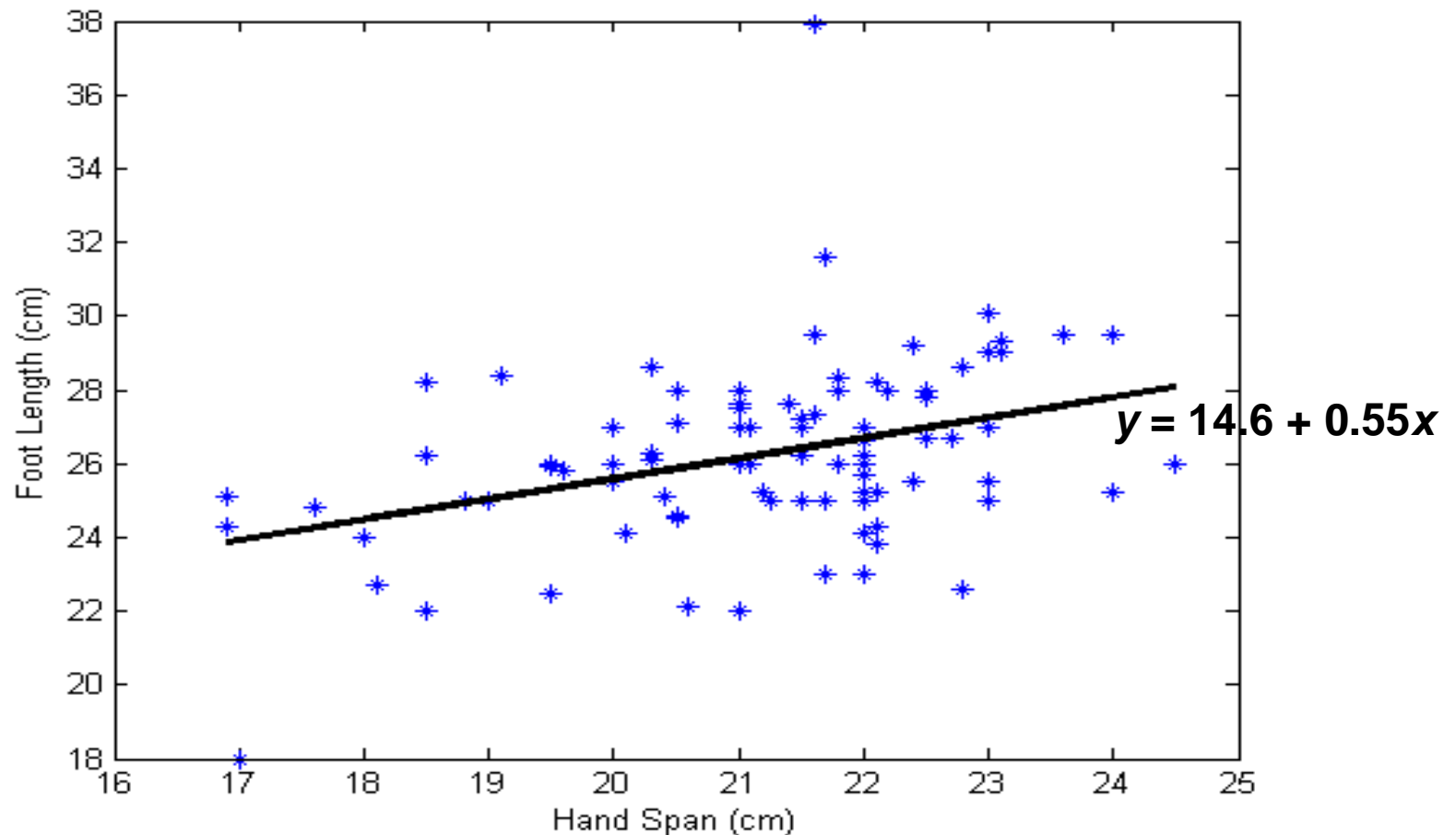
$$\begin{aligned} & \hat{\alpha} + \hat{\beta}x \pm as_e \sqrt{1 + \frac{1}{n} + \frac{(x - \bar{x})^2}{S_{xx}}} \\ &= -4.0667 + 0.9944(60) \\ & \quad \pm 2.0484(9.4630) \sqrt{1 + \frac{1}{30} + \frac{(60 - 76.7333)^2}{5135.8667}} \\ &= 55.59 \pm 20.22 \end{aligned}$$

or **[35.4, 75.8]**

Foot Length versus Hand Span - Females



Foot Length versus Hand Span - Males



Foot Length versus Hand Span – Males (Blue), Females (Red)

