

## Lecture 35: Chapter 13, Section 2 Two Quantitative Variables Interval Estimates

- PI for Individual Response, CI for Mean Response
- Explanatory Value Close to or Far from Mean
- Approximating Intervals by Hand
- Width of PI vs. CI
- Guidelines for Regression Inference

## Looking Back: Review

- **4 Stages of Statistics**
  - Data Production (discussed in Lectures 1-4)
  - Displaying and Summarizing (Lectures 5-12)
  - Probability (discussed in Lectures 13-20)
  - Statistical Inference
    - 1 categorical (discussed in Lectures 21-23)
    - 1 quantitative (discussed in Lectures 24-27)
    - cat and quan: paired, 2-sample, several-sample (Lectures 28-31)
    - 2 categorical (discussed in Lectures 32-33)
    - 2 quantitative

## Correlation and Regression (Review)

- Relationship between 2 quantitative variables
    - Display with **scatterplot**
    - Summarize:
      - **Form**: linear or curved
      - **Direction**: positive or negative
      - **Strength**: strong, moderate, weak
- If form is linear, **correlation**  $r$  tells direction and strength.  
 Also, equation of **least squares regression line** lets us predict a response  $\hat{y}$  for any explanatory value  $x$ .

## Population Model; Parameters and Estimates

Summarize linear relationship between **sampled**  $x$  and  $y$  values with line  $\hat{y} = b_0 + b_1x$  minimizing sum of squared residuals  $y_i - \hat{y}_i$ . Typical residual size is

$$s = \sqrt{\frac{(y_1 - \hat{y}_1)^2 + \dots + (y_n - \hat{y}_n)^2}{n-2}}$$

Model for **population** relationship is  $\mu_y = \beta_0 + \beta_1x$  and responses vary normally with standard deviation  $\sigma$

- Use  $b_0$  to estimate  $\beta_0$
- Use  $b_1$  to estimate  $\beta_1$
- Use  $s$  to estimate  $\sigma$

**Looking Back:** Our hypothesis test focused on slope.

## Regression Null Hypothesis (Review)

- $H_0 : \beta_1 = 0$   
→ no population relationship between  $x$  and  $y$
- Test statistic  $t = \frac{b_1 - 0}{SE_{b_1}}$
- $P$ -value is probability of  $t$  this extreme, if  $H_0$  true  
(where  $t$  has  $n-2$  df)

## Confidence Interval for Slope (Review)

Confidence interval for  $\beta_1$  is

$$b_1 \pm \text{multiplier}(SE_{b_1})$$

where *multiplier* is from  $t$  dist. with  $n-2$  df.

If  $n$  is large, 95% confidence interval is

$$b_1 \pm 2(SE_{b_1}).$$

If CI does not contain 0, reject  $H_0$ , conclude  $x$  and  $y$  are related.

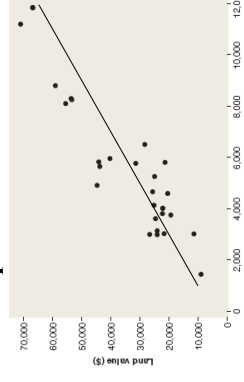
## Interval Estimates in Regression

Seek **P**rediction and **C**onfidence **I**ntervals for

- **Individual** response to given  $x$  value (**PI**)
  - For large  $n$ , approx. 95% **PI**:  $\hat{y} \pm 2s$
  - **Mean** response to subpopulation with given  $x$  value (**CI**)

- For large  $n$ , approx. 95% **CI**:  $\hat{y} \pm 2 \frac{s}{\sqrt{n}}$
- Both intervals centered at predicted  $y$ -value  $\hat{y}$ .

These approximations may be poor if  $n$  is small or if given  $x$  value is far from average  $x$  value.



**A Closer Look:** His lot is smaller than average but valued higher than average; some cause for concern because the relationship is strong and positive. But it's not perfect, so we seek statistical evidence of an unusually high value for the lot's size.

## Example: Reviewing Data in Scatterplot

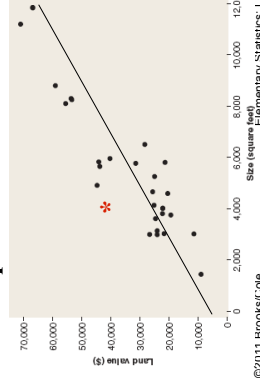
- **Background:** Property owner feels reassessed value \$40,000 of his 4,000 sq.ft. lot is too high. For random sample of 29 local lots, means are 5,619 sq.ft. for size, \$34,624 for value. Regression equation  $\hat{y} = 1,551 + 5.885x$ ,  $r = +0.927$ ,  $s = \$6,682$ .
- **Question:** Where would his property appear on scatterplot?
- **Response:**

## Example: An Interval Estimate

- **Background:** Property owner feels reassessed value \$40,000 of his 4,000 sq.ft. lot is too high. For random sample of 29 local lots, means are 5,619 sq.ft. for size, \$34,624 for value. Regression equation  $\hat{y} = 1,551 + 5.885x$ ,  $r = +0.927$ ,  $s = \$6,682$ .
- **Questions:** What range of values are within two standard errors of the predicted value for 4,000 sq.ft.? Does \$40,000 seem too high?
- **Responses:** Predict  $\hat{y} =$  \_\_\_\_\_
- **Approximate** range of plausible values for individual 4,000 sq.ft. lot is \_\_\_\_\_

## Example: Interval Estimate on Scatterplot

- **Background:** A homeowner's 4,000 sq.ft. lot is assessed at \$40,000. Predicted value is \$25,091 and predicted range of values is (\$11,727, \$38,455).
- **Question:** Where do the prediction and range of values appear on the scatterplot?
- **Response:**



## Prediction Interval vs. Confidence Interval

- Prediction interval corresponds to 68-95-99.7 Rule for **data**: where an **individual** is likely to be.
  - **PI is wider**: individuals vary a great deal
- Confidence interval is **inference** about **mean**: range of plausible values for **mean** of sub-population.
  - **CI is narrower**: can estimate mean with more precision
- Both PI and CI in regression **utilize info about x** to be more precise about **y** (PI) or mean **y** (CI).

## Example: Prediction or Confidence Interval

- **Background:** Property owner feels reassessed value \$40,000 of his 4,000 sq.ft. lot is too high. Based on a random sample of 29 local lots, software was used to produce interval estimates when size equals 4,000 sq.ft.

Predicted Values for New Observations			
New Obs	Fit	SE Fit	95.0% CI
1	25094	1446	( 22127, 28060)
Values of Predictors for New Observations			
New Obs	Size		
1	4000		

- **Questions:** What is the “Fit” value reporting? Which interval is relevant for the property owner’s purposes: CI or PI?
- **Responses:** *Fit* is \_\_\_\_\_  
The \_\_\_\_\_ is relevant: he wants to show that his individual lot is over-assessed.

## Examples: Series of Estimation Problems

- Based on sample of male weights, estimate
  - weight of individual male } *No regression*
  - mean weight of all males } *needed.*
- Based on sample of male **hts** and **weights**, estimate
  - weight of individual male, 71 inches tall
  - mean weight of all 71-inch-tall males
  - weight of individual male, 76 inches tall
  - mean weight of all 76-inch-tall males

Examples use data from sample of college males.

## Example: Estimate Individual Wt, No Ht Info

- **Background:** A sample of male weights have mean 170.8, standard deviation 33.1. Shape of distribution is close to normal.
- **Question:** What interval should contain the weight of an individual male?
- **Response:** Need to know distribution of weights is approximately normal to apply 68-95-99.7 Rule: Approx. 95% of individual male weights in interval \_\_\_\_\_

## Example: Estimate Mean Wt, No Ht Info

- **Background:** A sample of 162 male weights have mean 170.8, standard deviation 33.1.
- **Questions:**
  - What interval should contain the mean weight of all males?
  - How does it compare to this interval for an individual male's weight?  $170.8 \pm 2(33.1) = (104.6, 237.0)$
- **Responses:**
  - Need to know \_\_\_\_\_ to construct approximate 95% confidence interval for mean:
  - Interval for **mean** involves division by square root of  $n$  → \_\_\_\_\_ than interval for individual

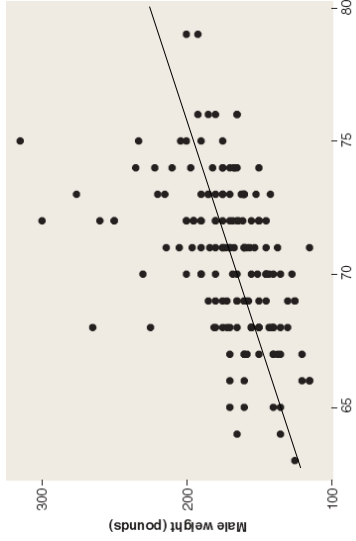
## Examples: Series of Estimation Problems

- Based on sample of male weights, estimate
  - weight of individual male
  - mean weight of all males
- Based on sample of male heights and weights, estimate
  - weight of individual male, 71 inches tall
  - mean weight of all 71-inch-tall males
  - weight of individual male, 76 inches tall
  - mean weight of all 76-inch-tall males

*Need regression*

## Examples: Series of Estimation Problems

The next 4 examples make use of regression on height to produce interval estimates for weight.



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## Example: Predict Individual Wt, Given Av. Ht

**Background:** Male hts: mean about 71 in. Wts: s.d. 33.1 lbs. Regression of wt on ht has  $r = +0.45$ ,  $p = 0.000$ . Regression line is  $\hat{y} = -188 + 5.08x$  and  $s = 29.6$  lbs.

**Questions:** How much heavier is a sampled male, for each additional inch in height? Why is  $s < s_y$ ? What interval should contain the weight of an individual 71-inch-tall male? (Got interval estimates for  $x = 71$ .)

New Obs	Fit	SE Fit	95.0% CI	95.0% PI
1	172.83	2.35	( 168.20, 177.47)	( 114.20, 231.47)

### Responses:

- For each additional inch, sampled male weighs \_\_\_ lbs more.
- $s < s_y$  because wts vary \_\_\_ about line than about mean.
- Look at \_\_\_ for  $x = 71$ : \_\_\_\_\_

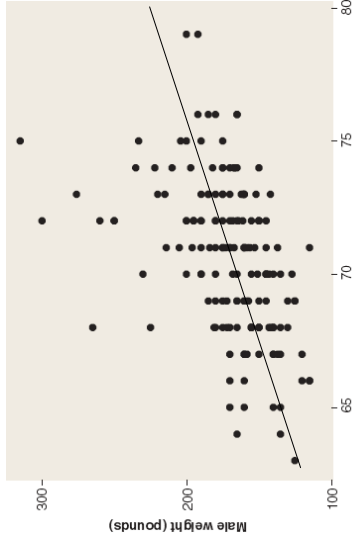
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## Example: Approx. Individual Wt, Given Av. Ht

The next 4 examples make use of regression on height to produce interval estimates for weight.



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## Example: Est Mean Wt, Given Average Ht

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**Questions:** What interval should contain mean weight of all 71-inch-tall males? How do we approximate the interval by hand? Is it close?

New Obs	Fit	SE Fit	95.0% CI	95.0% PI
1	172.83	2.35	( 168.20, 177.47)	( 114.20, 231.47)

### Responses:

- Software  $\rightarrow$  \_\_\_ for  $x = 71$
- Predict  $y$  for  $x = 71$ :  $\hat{y} = -188 + 5.08(71) = 172.7$  Approx. Close? \_\_\_\_\_

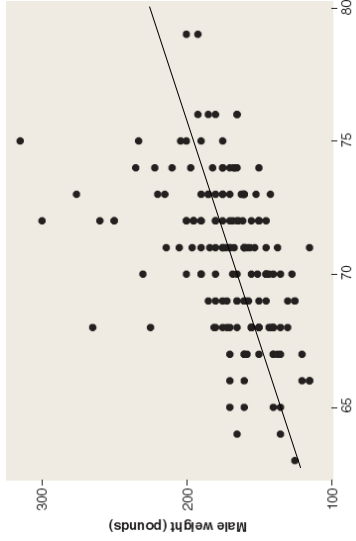
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## Example: Approx. Individual Wt, Given Av. Ht

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**Questions:** How much heavier is a sampled male, for each additional inch in height? Why is  $s < s_y$ ? What interval should contain the weight of an individual 71-inch-tall male? (Got interval estimates for  $x = 71$ .)

New Obs	Fit	SE Fit	95.0% CI	95.0% PI
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### Responses:

- For each additional inch, sampled male weighs \_\_\_ lbs more.
- $s < s_y$  because wts vary \_\_\_ about line than about mean.
- Look at \_\_\_ for  $x = 71$ : \_\_\_\_\_

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## Example: Est Mean Wt, Given Average Ht

**Background:** Male hts: mean about 71 in. Wts: s.d. 33.1 lbs. Regression of wt on ht has  $r = +0.45$ ,  $p = 0.000$ . Regression line is  $\hat{y} = -188 + 5.08x$  and  $s = 29.6$  lbs.

**Questions:** What interval should contain mean weight of all 71-inch-tall males? How do we approximate the interval by hand? Is it close?

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1	172.83	2.35	( 168.20, 177.47)	( 114.20, 231.47)

### Responses:

- Software  $\rightarrow$  \_\_\_ for  $x = 71$
- Predict  $y$  for  $x = 71$ :  $\hat{y} = -188 + 5.08(71) = 172.7$  Approx. Close? \_\_\_\_\_

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### Example: Estimate Wt, Given Tall vs. Av. Ht

- Background:** Regression of male wt on ht produced equation  $\hat{y} = -188 + 5.08x$ . For height 71 inches, estimated weight is  $\hat{y} = -188 + 5.08(71) = 172.7$
  - Question:** How much heavier will our estimate be for height 76 inches?
  - Response:** Since \_\_\_\_\_, predict \_\_\_\_\_ more lbs for each additional inch; \_\_\_\_\_ more lbs for 76, which is 5 additional inches: \_\_\_\_\_
- Instead of weight about 173, estimate weight about \_\_\_\_\_

### Example: Est Individual Wt, Given Tall Ht

- Background:** Regression of male weight on height has  $r = +0.45$ ,  $p = 0.000 \rightarrow$  strong evidence of moderate positive relationship. Reg. line  $\hat{y} = -188 + 5.08x$  and  $s = 29.6$  lbs. Got interval estimates for  $x = 76$ .  

New Obs	Fit	SE Fit	95.0% CI	95.0% PI
1	198.21	4.88	( 188.58, 207.84)	( 138.97, 257.45)
- Questions:** What interval should contain the weight of an individual male, 76 inches tall? How does the interval compare to the one for ht=71?
- Responses:**
  - \_\_\_\_\_ for  $x = 76$
  - Predicted wt (fit) about \_\_\_\_\_ lbs more for  $x = 76$  than for 71: \_\_\_\_\_ (5 more lbs per additional inch).

### Example: Approx. Individual Wt for Tall Ht

- Background:** Regression of male weight on height has  $r = +0.45$ ,  $p = 0.000 \rightarrow$  strong evidence of moderate positive relationship. Reg. line  $\hat{y} = -188 + 5.08x$  and  $s = 29.6$  lbs. Got interval estimates for  $x = 76$ .  

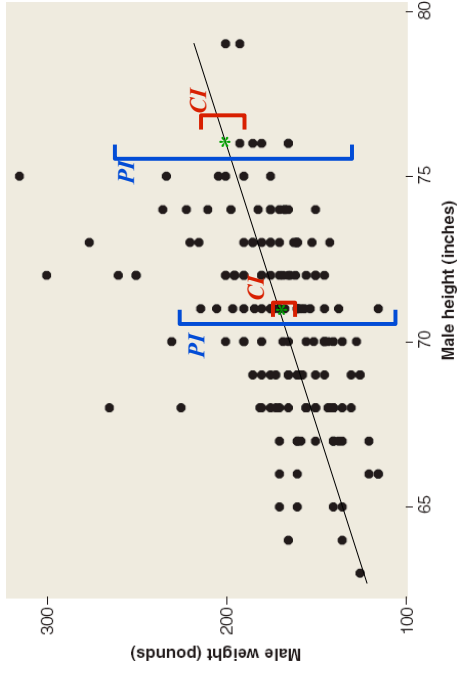
New Obs	Fit	SE Fit	95.0% CI	95.0% PI
1	198.21	4.88	( 188.58, 207.84)	( 138.97, 257.45)
- Questions:**
  - How do we *approximate* the prediction interval by hand?
  - Is it close to the true interval?
- Responses:**
  - Predict  $y$  for  $x = 76$ : \_\_\_\_\_
  - Approx. PI = \_\_\_\_\_
  - Close? \_\_\_\_\_

### Example: Est Mean Wt, Given Tall Ht

- Background:** Regression of 162 male wts on hts has  $r = +0.45$ ,  $p = 0.000 \rightarrow$  strong evidence of moderate positive relationship. Reg. line  $\hat{y} = -188 + 5.08x$  and  $s = 29.6$  lbs. Got interval estimates for  $x = 76$ .  

New Obs	Fit	SE Fit	95.0% CI	95.0% PI
1	198.21	4.88	( 188.58, 207.84)	( 138.97, 257.45)
- Questions:**
  - What interval should contain **mean** wt of all 76-in males?
  - How do we *approximate* the interval by hand? Is it close?
- Responses:**
  - Refer to \_\_\_\_\_
  - Predict  $y$  for  $x = 76$ :  $\hat{y} = -188 + 5.08(76) = 198.1$
- Close? \_\_\_\_\_

## Examples: PI and CI for Wt; Ht=71 or 76

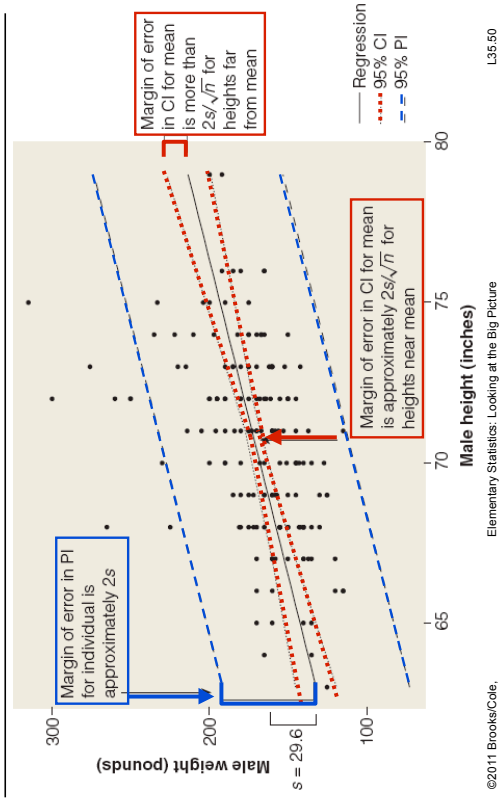


## Interval Estimates in Regression (Review)

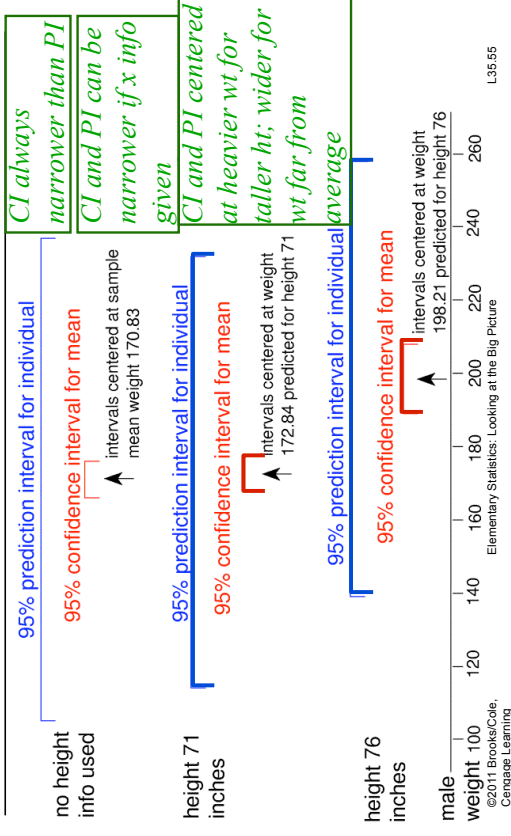
Seek interval estimates for

- Individual response to given  $x$  value (PI)
  - For large  $n$ , approx. 95% PI:  $\hat{y} \pm 2s$
- Mean response to subpopulation with given  $x$  value (CI)
  - For large  $n$ , approx. 95% CI:  $\hat{y} \pm 2\frac{s}{\sqrt{n}}$
- Intervals **approximately correct only for  $x$  values close to mean**; otherwise wider
  - Especially CI much wider for  $x$  far from mean

## PI and CI for $x$ Close to or Far From Mean



## Summary of Example Intervals



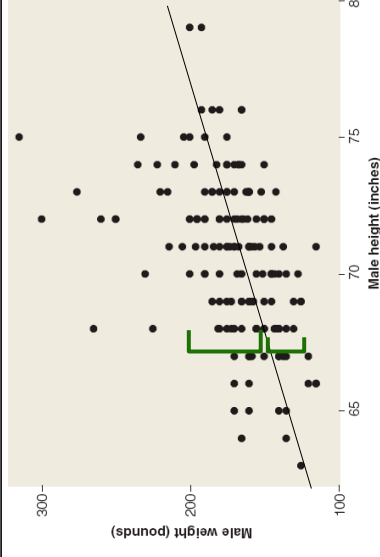
## Example: A Prediction Interval Application

- **Background:** A news report stated that Michael Jackson was a fairly healthy 50-year-old before he died of an overdose. “His 136 pounds were in the acceptable range for a 5-foot-9 man...”
- **Question:** Based on the regression equation  $\hat{y} = -188 + 5.08x$  and  $s = 29.6$  lbs, would we agree that 136 lbs. is not an unusually low weight?
- **Response:** For  $x = 69$ , predict  $y =$  \_\_\_\_\_; his weight 136 Our PI is \_\_\_\_\_

*A Closer Look: Our PI is a bit misleading because the distribution of weights is actually somewhat right-skewed, not normal. More of the spread reported in  $s = 29.6$  comes about from unusually heavy men, and less from unusually light men.*

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## Example: A Prediction Interval Application



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## Guidelines for Regression Inference

- Relationship must be linear
- Need random sample of independent observations
- Sample size must be large enough to offset non-normality
- Need population at least 10 times sample size
- Constant spread about regression line
- Outliers/influential observations may impact results
- Confounding variables should be separated out

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## Lecture Summary (Inference for $Quan \rightarrow Quan$ ; PI and CI)

- Interval estimates in regression: PI or CI
  - Non-regression PI (individual) and CI (mean)
  - Regression PI and CI for  $x$  value near mean or far
  - Approximating intervals by hand
  - Width of PI vs. CI
  - Guidelines for regression inference

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