

Ch 15 Inference for Regression Activity Intro

Activity 15 Ideal Proportions (pg. 888)

"The length of a man's outspread arms is equal to his height."

① x = distance between fingertips of outstretched arms
 y = height (in cm)

② create scatterplot

strong, moderate, weak association?

③ Find the linear regression line, r and r^2 .

Interpret meaning in context.

r = correlation; $[-1, 1]$ strong, mod, weak; +/-

r^2 = coefficient of determination; % of variation in y that can be explained by the least-squares line of y on x

④ Construct a residual plot

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

Calculator

Least-Squares line: Data $\Rightarrow L_1 \text{ ; } L_2$

STAT CALC Lin Reg: a+bx (L_1, L_2, Y_1)

$Y_1 \Rightarrow$ Vars | Y-vars | Function | Y_1

$r^2/r \Rightarrow$ Catalog Diagnostic ON

Graph zoomSTAT

Residual Plot: Clear $y =$

$L_3: Y_1(L_1)$ $L_4: L_2 - L_3 \Rightarrow$ Residuals in L_4

Stat Plot $x: L_2$ $y: L_4$

Zoom Stat

Least Squares Regression (LSR) line can be used to predict y -values for a given x

parameter		estimate
μ	mean	\bar{x}
σ	stand. dev.	s
p	proportion	\hat{p}
$\mu_y = \alpha + \beta x$	LSR	$\hat{y} = a + bx$

Example 15.1 Enter data into calculator (Cyring L, IQ L)
Graph Scatter plot / LSR

$\hat{y} = 91.268 + 1.493x$ Read through example independently

True Regression Line = Line of Means or Line of Averages

$\mu_y = \alpha + \beta x$

mean response of y moves along the line as the explanatory variable x takes different values

Graphic (pg 892) - All of the curves have the same σ , so the variability of y is the same for all values of x .

Conditions for Regression Inference : L I N E S S

L - linear; Scatterplot is roughly linear

I - independent; no item is measured twice

N - normal; boxplot of residuals is $\sim N$

S - SRS

S - Standard Deviation; s of y is the same for all values of x ; No fanning of the residual plot

15 cont'd

Example 15.2 Baby crying and IQ

$$\hat{y} = 91.268 + 1.493x$$

(Diff from text)

(A) Two babies had 9 crying peaks.

Find their predicted IQ's and residuals

Predicted IQ: $\hat{y} = 91.27 + 1.493(9)$
 $= 104.707$

Residuals: $103 - 104.707 = -1.707$

actual lies 1.707 points below LSR

$119 - 104.707 = 14.293$

actual lies 14.293 above LSR

(B) Determine all residuals

Calculator: Crying = L_1 , IQ = L_2

$L_3 = 91.268 + 1.493(L_1)$

$L_4 = L_2 - L_3$

Statplot
PS 994

1-VAR Stats L_4 $\Sigma = -.057$ sum is ~ 0

(Equal distribution +/-)

Residual Plot StatPlot: on: X: L_2 Y: L_4
 zoom Stat

Standard Error of the LSR

$$s = \sqrt{\frac{\sum (y - \hat{y})^2}{n - 2}} = \sqrt{\frac{\sum \text{residuals}^2}{n - 2}}$$

* used in
 computation for
 CI of slope

Calculator: $S = \sqrt{\frac{\sum \text{residuals}^2}{n-2}} \Rightarrow \hat{\sigma}^2 = \frac{\sum \text{residuals}^2}{n-2}$

L4 has all residuals from Ex. 15.2
 1 VAR Stat L4 $\Rightarrow \sum x^2 = 11023.389$
 $n = 38$

$$s^2 = \frac{11023.389}{38-2} = 306.20525 \quad s = \sqrt{306.20525} = 17.499$$

$S = 17.499$

Confidence Interval for Regression Slope $b \pm t^* SE_b$

$b =$ slope estimate $\hat{y} = a + bx$

$t^* =$ critical value @ CI% for $df = n-2$

$\text{invT}(\frac{1+CI}{2}, df)$ or Table C

$$SE_b = S_b = \frac{\sqrt{\frac{\sum (y - \hat{y})^2}{n-2}}}{\sqrt{\sum (x - \bar{x})^2}}$$

* on formula sheet
 * use list functions to complete the formula

Review

* Figure 15.5 Minitab output

* Table C t^* $df = n-2 = 38-2 = 36$ use $df = 30$

95% CI $\Rightarrow t^* = 2.042$

* Calculator $\text{invT}(.975, 36) = 2.20281$

HW 15.8

* Discuss 2 on scatterplot

* minitab output

15 cont'd

Steps for Constructing a Confidence Interval for the Regression Slope

① Procedure: t-confidence interval for the slope of the true regression line relating _____ and _____ for all _____.

② Check Conditions LINESS

③ Computation:

$$b_1 \pm t^* SE_{b_1}$$

$$SE_{b_1} = S_{b_1} = \frac{S}{\sqrt{\sum(x-\bar{x})^2}} = \sqrt{\frac{\frac{\sum(y-\hat{y})^2}{n-2}}{\sum(x-\bar{x})^2}}$$

on formula sheet

④ Conclusion:

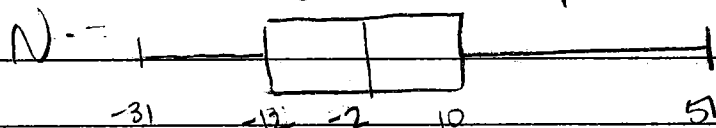
Based on this sample, I am _____% confident that the slope of the true LSR relating _____ and _____ for all _____ is between _____ and _____.

Example

Construct a 95% confidence interval for the slope of the true regression lines of crying babies and IQ.

① We will use a t -confidence interval for the slope of the true regression line relating crying events and ID for all infants.

- ② L - The scatterplot is roughly linear
 I - We assume each infant was only tested once and independence is satisfied.



The boxplot of residuals although skewed slightly to the right we will proceed as $\approx N$.

SRS is stated

The standard deviation of y is approximately the same for all x -values. The scatterplot of the residuals is $\approx N$ with no fanning.

③ $b_1 \pm t^* SE_{b_1}$ $b_1 = 1.493$ $df = 38 - 2 = 36$ $t^* = 2.028$

$1.493 \pm 2.028(.487)$

$(2.481, .505)$

$$SE_{b_1} = \frac{\sqrt{\frac{\sum(y - \hat{y})^2}{n - 2}}}{\sqrt{\sum(x - \bar{x})^2}}$$

$\bar{x} = 17.395$

$$= \frac{\sqrt{\frac{11023.3889}{38 - 2}}}{\sqrt{1291.079}} = \boxed{.487}$$

④ Based on the sample, I am 95% confident that the slope of the true LSR relating infant crying peaks and ID at age 3 for all infants snapped with a rubber band then monitored for crying for 20 minutes is between 2.481 and .505.

Ch. 15) cont'd

* A linear relationship may not exist between two variables, like phone number and SSN. However, it is unlikely $b_1 = 0$ and each sample will likely produce a different b_1 . Estimated slope is insignificant

t-Test for Slope

Determines if a nonzero correlation represents a true linear relationship with a nonzero slope and so knowing the value of x is helpful in predicting y

Significance test for Regression Slope

pg 904
Graphic

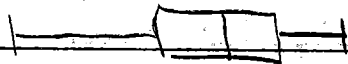
"Is that trend real, or could that happen by chance?"
"How far is b_1 from 0 (or some other H_0) in terms of standard error?"

Steps in t-Test for Slope

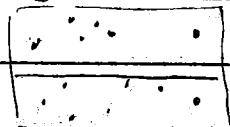
- ① Hypothesis: $H_0: \beta_1 = 0$ $H_a: \beta_1 \neq 0$
- ② Procedure: t-test for the slope of the true regression line (β_1) relating _____ ? _____
for all _____
- ③ Conditions: LINESS
- ④ Computation: $t = \frac{b_1 - \beta_1}{S_{b_1}}$ $df = n - 2$ $p\text{-value} = \underline{\hspace{2cm}}$
- ⑤ Interpret in context

Example! Temperature and cricket chirps (15 part slide)

- ① $H_0: \beta_0 = 0$ $H_a: \beta_0 \neq 0$
- ② t-test for the slope of the true regression line (β_1) relating the number of cricket chirps and the temperature for all one second intervals.
- ③ L - The scatterplot appears linear
Independent / SRS - We will consider this a random, independent sample of 1-sec. intervals
N_e - Box plot of residuals is $\sim N$.



S - The residual plot shows no fanning



$$\textcircled{4} t = \frac{b_1 - \beta}{s_{b_1}} \quad b_1 = 3.291 \quad s_{b_1} = \frac{\sqrt{\frac{140.547}{13}}}{\sqrt{40.557}} = .601$$
$$\beta = 0$$

$$t = \frac{3.291 - 0}{.601} = 5.475$$

$$df = 15 - 2 = 13 \quad p\text{-value} = 2P(T > 5.475) \approx 0$$

⑤ Since the p-value $0 < \alpha = .05$ we reject H_0 . There is significant evidence to conclude temperature can be predicted by the number of cricket chirps.

(minitab output slides 15II and pg 905)

Ch. 15 List Procedure for Regression Inference

L1	L2	L3	L4	L5
X	Y	Y(L1)	L2-L3	L1 - \bar{X}



LinReg (atbx)
L1, L2
Store Eq: Y1

Residuals $\bar{X} \Rightarrow$ 1 var stat L1
1 VAR Stat L4
 $\Sigma x^2 = \Sigma (y - \hat{y})^2$

1 VAR Stat L5
 $\Sigma x^2 = \Sigma (x - \bar{x})^2$

$$SE_{b_1} = S_{b_1} = \frac{\sqrt{\frac{\Sigma (y - \hat{y})^2}{n - 2}}}{\sqrt{\Sigma (x - \bar{x})^2}}$$

$y - \hat{y} \Rightarrow$ residuals
df = n - 2

Box Plot of Residuals - tests for Normality

StatPlot 1(on)

1-111 L4

Residual Plot - test for consistent σ

StatPlot 1(on)

no fanning

⋮ L2, L4

