

# Section 8.4

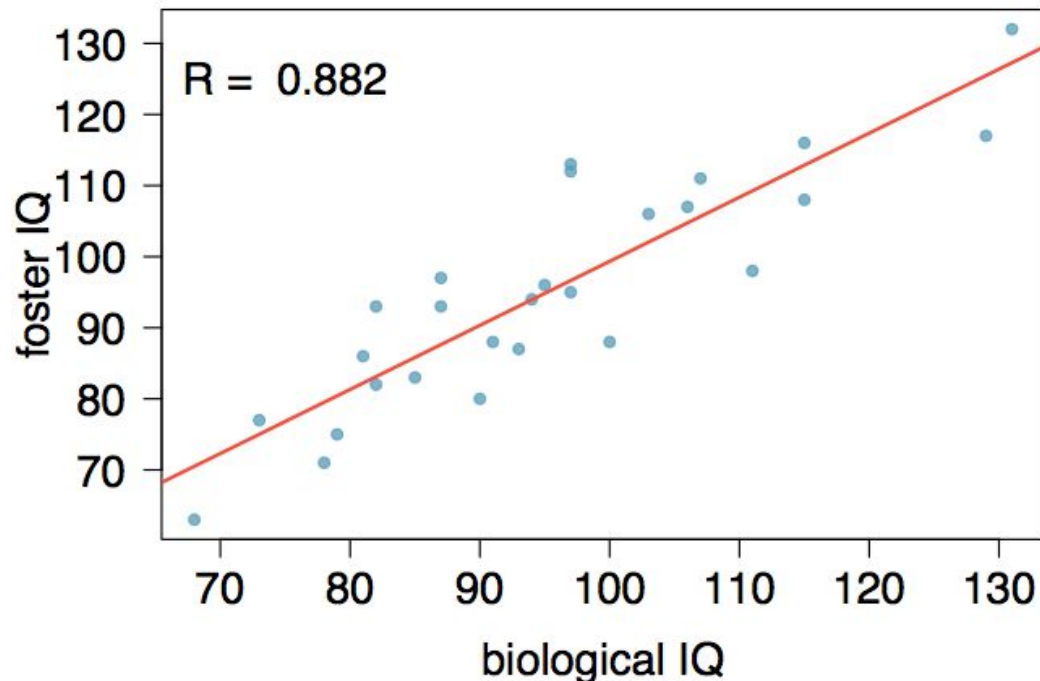
# Inference for Linear Regression

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Stats 7 Summer Session II 2022

# Nature or nurture?

In 1966 Cyril Burt published a paper called "The genetic determination of differences in intelligence: A study of monozygotic twins reared apart?" The data consist of IQ scores for [an assumed random sample of] 27 identical twins, one raised by foster parents, the other by the biological parents.



# Practice

Which of the following is false?

**Coefficients:**

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	9.20760	9.29990	0.990	0.332
bioIQ	0.90144	0.09633	9.358	1.2e-09

Residual standard error: 7.729 on 25 degrees of freedom  
Multiple R-squared: 0.7779, Adjusted R-squared: 0.769  
F-statistic: 87.56 on 1 and 25 DF, p-value: 1.204e-09

- (a) Additional 10 points in the biological twin's IQ is associated with additional 9 points in the foster twin's IQ, on average.
- (b) Roughly 78% of the foster twins' IQs can be accurately predicted by the model.
- (c) The linear model is  $\widehat{fosterIQ} = 9.2 + 0.9 \times bioIQ$
- (d) Foster twins with IQs higher than average IQs tend to have biological twins with higher than average IQs as well.

# Inference for the slope

- With a linear model our focuses are usually either:
  - Predicting values of the response for a given value of the explanatory/predictor variable
  - Studying the linear relationship between the response and the explanatory variables
    - Testing if the slope,  $\beta_1$ , is non zero (i.e. the two variables are linearly related)
    - Creating a confidence interval to estimate the slope
  - We could do the same investigation with the intercept,  $\beta_0$ , but the intercept is rarely of interest

## Testing for the slope

Assuming that these 27 twins comprise a representative sample of all twins separated at birth, we would like to test if these data provide convincing evidence that the IQ of the biological twin is a significant predictor of IQ of the foster twin. What are the appropriate hypotheses?

(a)  $H_0: b_0 = 0; H_A: b_0 \neq 0$

(b)  $H_0: \beta_0 = 0; H_A: \beta_0 \neq 0$

(c)  $H_0: b_1 = 0; H_A: b_1 \neq 0$

(d)  $H_0: \beta_1 = 0; H_A: \beta_1 \neq 0$

## Testing for the slope (cont.)

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	9.2076	9.2999	0.99	0.3316
biolQ	0.9014	0.0963	9.36	0.0000

- We always use a t-test in inference for regression.  
**Remember:** test statistic  $T = (\text{point estimate} - \text{null value}) / \text{SE}$
- Point estimate =  $b_1$  is the observed slope.
- $SE_{b_1}$  is the standard error associated with the slope.
- Degrees of freedom associated with the slope is  $df = n - 2$ , where  $n$  is the sample size.

## Testing for the slope (cont.)

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	9.2076	9.2999	0.99	0.3316
bioIQ	0.9014	0.0963	9.36	0.0000

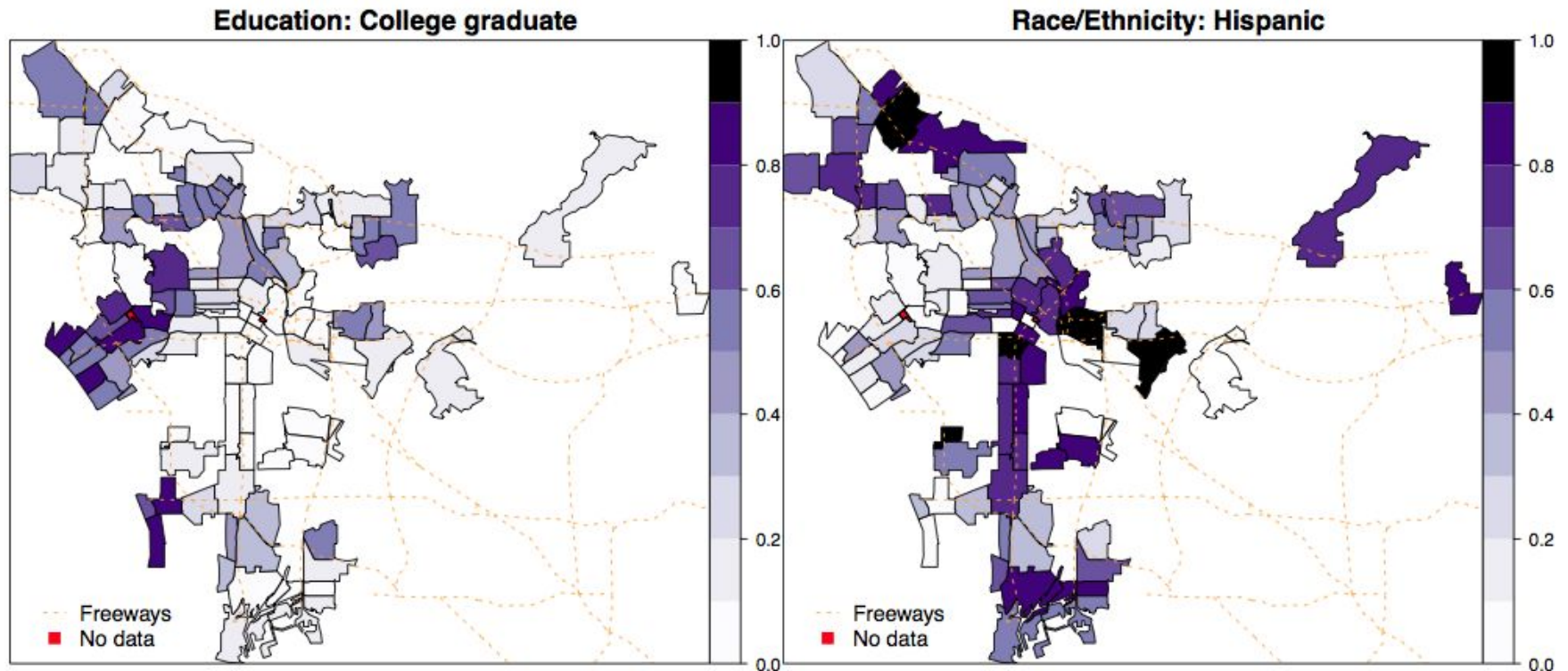
$$T = \frac{0.9014 - 0}{0.0963} = 9.36$$

$$df = 27 - 2 = 25$$

$$p\text{-value} = P(|T| > 9.36) < 0.01$$

# Percent college graduate vs. percent Hispanic in LA

What can you say about the relationship between % college graduate and % Hispanic in a sample of 100 zip code areas in LA?







## % college educated vs. % Hispanic in LA - linear model

Which of the below is the best interpretation of the slope?

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	0.7290	0.0308	23.68	0.0000
%Hispanic	-0.7527	0.0501	-15.01	0.0000

- (a) A 1% increase in Hispanic residents in a zip code area in LA is associated with a 75% decrease in % of college grads.
- (b) A 1% increase in Hispanic residents in a zip code area in LA is associated with a 0.75% decrease in % of college grads.
- (c) An additional 1% of Hispanic residents decreases the % of college graduates in a zip code area in LA by 0.75%.
- (d) In zip code areas with no Hispanic residents, % of college graduates is expected to be 75%.

## % college educated vs. % Hispanic in LA - linear model

Do these data provide convincing evidence that there is a statistically significant relationship between % Hispanic and % college graduates in zip code areas in LA?

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	0.7290	0.0308	23.68	0.0000
hispanic	-0.7527	0.0501	-15.01	0.0000

*Yes, the p-value for % Hispanic is low, indicating that the data provide convincing evidence that the slope parameter is different than 0.*

How reliable is this p-value if these zip code areas are not randomly selected?

*Not very...*

## Confidence interval for the slope

Remember that a confidence interval is calculated as *point estimate*  $\pm$  *ME* and the degrees of freedom associated with the slope in a simple linear regression is  $n - 2$ . Which of the below is the correct 95% confidence interval for the slope parameter? Note that the model is based on observations from 27 twins.

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	9.2076	9.2999	0.99	0.3316
bioIQ	0.9014	0.0963	9.36	0.0000

(a)  $9.2076 \pm 1.65 \times 9.2999$

(b)  $0.9014 \pm 2.06 \times 0.0963$

(c)  $0.9014 \pm 1.96 \times 0.0963$

(d)  $9.2076 \pm 1.96 \times 0.0963$

$$n = 27 \quad df = 27 - 2 = 25$$

$$> qt((1 - 0.95) / 2, df = 25)$$

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[1] -2.059539
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(0.7, 1.1)

# Recap

- Inference for the slope for a single-predictor linear regression model:

- Hypothesis test:
$$T = \frac{b_1 - \text{null value}}{SE_{b_1}} \quad df = n - 2$$

- Confidence interval:

$$b_1 \pm t_{df=n-2}^* SE_{b_1}$$

- The null value is often 0 since we are usually checking for *any* relationship between the explanatory and the response variable.
- The regression output gives  $b_1$ ,  $SE_{b_1}$ , and *two-tailed* p-value for the t-test for the slope where the null value is 0.
- We rarely do inference on the intercept, so we'll be focusing on the estimates and inference for the slope.

# Caution

- Always be aware of the type of data you're working with: random sample, non-random sample, or population.
- Statistical inference, and the resulting p-values, are meaningless when you already have population data.
- If you have a sample that is non-random (biased), inference on the results will be unreliable.
- The ultimate goal is to have independent observations.

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