# Multiple Linear Regression - Inference

#### **Research Objective**

**Research Question:** What determines a person's height?

Population: All BYU students.

**Parameter of Interest:** 

- Some number measuring the "relationship" between height and various other explanatory variables such as fathers height, mother's height, etc.
- For regression, these are the "slopes" or "effects" (e.g.  $\beta_1$ ) in the model.

Sample: A convenience sample of 1727 BYU students who are in Stat 121.

#### **Research Objective**

**Research Question:** Is the height of a student influenced by any of the explanatory variables?

$$\begin{split} \text{Height}_{i} &= \beta_{0} + \beta_{1}\text{MH}_{i} + \beta_{2}\text{FH}_{i} + \beta_{3}\text{Sports}_{i} + \beta_{4}\text{Sex}_{i} + \beta_{5}\text{Shoe}_{i} + \epsilon_{i} \\ \epsilon_{i} &\sim \mathcal{N}(0, \sigma^{2}) \end{split}$$

What would it mean if  $\beta_1 = \beta_2 = \cdots = \beta_5 = 0$ ?

- There is no relationship between the student's height (y) and ANY of the explanatory variables.
- This can be a useful hypothesis to test, particularly if you have a lot of explanatory variables.

# **Overall Hypothesis Testing in MLR**

**Research Question:** Is the height of a student influenced by any of the explanatory variables?

Steps of hypothesis testing:

- 1. Formulate null and alternative hypotheses.
- 2. Gather the data and see if our sample data matches (or doesn't match) the null hypothesis.
- 3. Draw a conclusion about  $H_0$ .

**Research Question:** Is the height of a student influenced by any of the explanatory variables?

Knowing what we did with other hypothesis tests, how would we write out our hypotheses?

 $egin{array}{c} H_0:\ H_a: \end{array}$ 

**Research Question:** Is the height of a student influenced by any of the explanatory variables?

Knowing what we did with other hypothesis tests, how would we write out our hypotheses?

$$egin{aligned} H_0:η_1=eta_2=eta_3=eta_4=eta_5=0\ H_a:& ext{At least one }eta ext{ is not zero} \end{aligned}$$

**Research Question:** Is the height of a student influenced by any of the explanatory variables?

Step 2 - Compare our data result with what we expect to see if the null hypothesis is true.

- How do we do this?
- $R^2$  will be key
- Recall that  $R^2$  is the percent of variability in the response explained by all the explanatory variables. So if  $R^2$  is close to 1 then the explanatory variables are doing a good job but  $R^2$  close to 0 means none of our explanatory variables are helpful.

**Research Question:** Is the height of a student influenced by any of the explanatory variables?

Step 2 - Compare our data result with what we expect to see if the null hypothesis is true.

$$F = rac{R^2/P}{(1-R^2)/(n-P-1)}$$

- If the null is true then Fpprox 0.
- We have F = 1443.941. Is this different enough from 0 to lead us to believe that  $H_0$  is false?

Theorem: The sampling distribution of F

If the LINE assumptions of the regression model are appropriate, then

$$F = rac{R^2/P}{(1-R^2)/(n-P-1)}$$

is a test statistic and its sampling distribution follows an F distribution with degrees of freedom P and n - P - 1.

- So...what does that theorem mean?
  - The F-distribution tells us what we *should* see in our sample if  $H_0$  is true
  - The LINE assumptions about the population need to hold.



Reminder, the LINE assumptions are:

- L Linear relationship between QUANTITATIVE  $x \, {\rm \dot{s}}$  and y
- I Independence (one obs. doesn't impact the other)
- N Normal residuals (distance from line is normal)
- E Equal variance of residuals (spread about the line is constant)
- How would we see if there is a linear relationship between *x*'s and *y*?
- Scatterplots work OK but can be deceiving because we have many x's
- Added variable plots!



Are the relationships approximately linear?

How would we see if there is independence? In other words, how can we "check" if one observation doesn't influence another?

- Critical Thinking!
- Does it "make sense" that one student's height would be related to another student's height?
- Maybe if there are relatives in the class but its likely a minimal influence.

How would we see if the residuals are normal?

- 1. Calculate the residuals as  $\epsilon_i = y_i (\hat{eta}_0 + \hat{eta}_1 x_{1i} + \dots + \hat{eta}_5 x_{5i})$  (don't worry the computer will do this for you)
- 2. Draw a histogram (or density plot) of residuals

How would we see if the residuals are normal?

- 1. Calculate the residuals as  $\epsilon_i = y_i (\hat{eta}_0 + \hat{eta}_1 x_{1i} + \dots + \hat{eta}_5 x_{5i})$  (don't worry the computer will do this for you)
- 2. Draw a histogram (or density plot) of residuals



Is this approximately normal?

• Skewness = 0.1193634

How would we see if there is "equal spread" of the residuals about the fitted line?

How would we see if there is "equal spread" of the residuals about the fitted line?

• Fitted values vs. residuals plot



Is this roughly "equal spread"?

• Yes except for a few outliers

# **Overall Hypothesis Tests in MLR**

**Research Question:** Is the height of a student influenced by any of the explanatory variables?

Step 2 - Measuring if our data is consistent with the null hypothesis:

• The LINE assumptions are met so we can use the *F*-distribution to do our overall hypothesis test (also called an omnibus test).

# **Overall Hypothesis Tests in MLR**

**Research Question:** Is the height of a student influenced by any of the explanatory variables?

Step 2 - Measuring if our data is consistent with the null hypothesis:

- 1. Test statistic: In our height example F = 1443.941 (it should be 0 if  $H_0$  is true).
- 2. p-value: probability of observing our sample result or "more extreme" (as stated by  $H_a$ ) if the null hypothesis is true. Our p-value is 0.

Step 3: Draw a conclusions about  $H_0: \beta_1 = \cdots = \beta_5 = 0$ . Using  $\alpha = 0.05$ , what do we conclude?

- Our data is NOT consistent with the null hypothesis so we conclude that at least 1 explanatory variable does have an effect on height.
- If we reject, this is a painfully vague conclusion. We need to get more specific.

Back to the Melanoma example...



2) Select Variables			
Please select up to 30 explanator	y variables to use. Each explanatory variable should "explain" wha	t happens to the response variable. NOTE: Only numeric variables will be given as op	tions
Select Response Variable: Mort Set the	e response variable		•
Select Explanatory Variable(s):			
Lat Ocean Long	Choose any explanatory variables – ple	ease READ questions carefully about	
Show 5 ventries	what explanatory variables to use	Search:	:
	Lat 🔶	Ocean 🔶	Long 🔶
1	33	1	87
2	34.5	0	112
3	35	0	92.5
4	37.5	1	119.5
5	39	0	105.5
Showing 1 to 5 of 49 entries		Previous 1 2 3 4	5 10 Next
Proceed to EDA			

Because there are many variables, we have to explore them one at a time





5) Regression Analysis						
Regression Analysis of: Mo Coefficient Table:	rt (Y) explained by Lat, Ocean, Long	(X's)				
Confidence Level for Slope and I	Intercept:					0.95 0.99
0.5 0.55	0.6	0.65 0.	.7 0.75	0.8	0.85 0.9	0.95 0.99
Show 5 ~ entries						
Test					F-statistic 🍦	p.value 🔶
1 F-test for all slop	pes are equal to zero				50.826	0
Showing 1 to 1 of 1 entries				This has	all the info for the F-test (o	Previous 1 Next
chules						veran testj
chow 5 Pennes	Estimate 🍦	Std. Error 🔶	t value 崇	p value 🍦	CI Lower Bound 🔶	CI Upper Bound 🔶
(Intercept)	Estimate 🔶 349.2369	<b>Std. Error</b> 27.0596	t value ∳ 12.9062	p value ∳ 0	CI Lower Bound 294.7361	CI Upper Bound 403.7377
(Intercept)	Estimate 🔶 349.2369 -5.495	<b>Std. Error</b> ♦ 27.0596 0.5289	t value ∳ 12.9062 -10.3898	p value ∳ 0 0	CI Lower Bound 294.7361 -6.5602	CI Upper Bound ∳ 403.7377 -4.4297
(Intercept) Lat Ocean	Estimate 349.2369 -5.495 21.7976	Std. Error           27.0596           0.5289           5.2263	t value 12.9062 -10.3898 4.1707	p value ♦ 0 0 0.0001	CI Lower Bound 294.7361 -6.5602 11.2712	CI Upper Bound ∳ 403.7377 -4.4297 32.324
(Intercept) Lat Ocean Long	Estimate	Std. Error ♦         27.0596         0.5289         5.2263         0.1732	t value ♦ 12.9062 -10.3898 4.1707 0.7037	p value ♦ 0 0 0.0001 0.4852	CI Lower Bound 294.7361 -6.5602 11.2712 -0.227	Cl Upper Bound ♦ 403.7377 -4.4297 32.324 0.4708
(Intercept) Lat Ocean Long Showing 1 to 4 of 4 entries	Estimate	Std. Error ♦         27.0596         0.5289         5.2263         0.1732	t value ♦ 12.9062 -10.3898 4.1707 0.7037	p value ♦ 0 0 0.0001 0.4852	CI Lower Bound 294.7361 -6.5602 11.2712 -0.227	Cl Upper Bound ♦ 403.7377 -4.4297 32.324 0.4708 Previous 1 Next
(Intercept) Lat Ocean Long Showing 1 to 4 of 4 entries R-squared: 0.7721 sigma-hat: 16.4806	Estimate 349.2369 -5.495 21.7976 0.1219	Std. Error         27.0596         0.5289         5.2263         0.1732	t value ♦ 12.9062 -10.3898 4.1707 0.7037	p value ♦ 0 0 0 0 0.0001 0.4852	CI Lower Bound 294.7361 -6.5602 11.2712 -0.227	Cl Upper Bound 403.7377 -4.4297 32.324 0.4708 Previous 1 Next

#### **Research Objective**

$$\begin{aligned} \text{Height}_{i} &= \beta_{0} + \beta_{1} \text{MH}_{i} + \beta_{2} \text{FH}_{i} + \beta_{3} \text{Sports}_{i} + \beta_{4} \text{Sex}_{i} + \beta_{5} \text{Shoe}_{i} + \epsilon_{i} \\ \epsilon_{i} &\sim \mathcal{N}(0, \sigma^{2}) \end{aligned}$$

**Research Question:** Is the height of a student influenced by whether they played sports in HS?

What would it mean if  $\beta_3 = 0$ ?

• There is no relationship between the height (y) and sports in HS (x).

#### **Population vs. Sample Slope**

$$\begin{split} \text{Height}_{i} &= \beta_{0} + \beta_{1}\text{MH}_{i} + \beta_{2}\text{FH}_{i} + \beta_{3}\text{Sports}_{i} + \beta_{4}\text{Sex}_{i} + \beta_{5}\text{Shoe}_{i} + \epsilon_{i} \\ \epsilon_{i} &\sim \mathcal{N}(0, \sigma^{2}) \end{split}$$

Our fitted model:

 $\hat{y} = 23.26 + 0.28 imes \mathrm{MH}_i + 0.21 imes \mathrm{FH}_i + 0.35 imes \mathrm{Sports}_i + 3.19 imes \mathrm{Sex}_i + 1.06 imes \mathrm{Shoe}_i$ 

So, doesn't this mean that  $eta_3 
eq 0$  because  $\hat{eta}_3 = 0.348?$ 

- Not necessarily!  $\beta_3 \neq \hat{\beta}_3$
- We need to do a test for  $\beta_3$

# Hypothesis Testing for a Single slope

**Research Question:** Does sports in HS impact height?

Steps of hypothesis testing:

- 1. Formulate null and alternative hypotheses.
- 2. Gather the data and see if our sample data matches (or doesn't match) the null hypothesis.
- 3. Draw a conclusion about  $H_0$ .

# Hypothesis Testing for a Single slope

Knowing what we did with other hypothesis tests, how would we write out our hypotheses?

 $H_0:$  $H_a:$ 

# Hypothesis Testing for a Single slope

Knowing what we did with other hypothesis tests, how would we write out our hypotheses?

$$egin{aligned} H_0:η_3=0\ H_a:η_3
eq 0 \end{aligned}$$

# Hypothesis Testing for a Single Slope

Step 2 - - gather the data and see if our sample data matches (or doesn't match) the null hypothesis (note: do this only if LINE assumptions are valid)

Measuring if our data is consistent with the null hypothesis:

- 1. Standardized test statistic: the number of standard errors away from the hypothesized value our data is. In our height example t = 3.2148829.
- 2. p-value: probability of observing our sample result or "more extreme" (as stated by  $H_a$ ) if the null hypothesis is true. Our p-value is 0.001.

Step 3: Draw a conclusions about  $H_0: \beta_3 = 0$ . Using  $\alpha = 0.05$ , what do we conclude about  $\beta_3$ ?

• Our data is NOT consistent with the null hypothesis so we conclude that the sports in HS does have an effect on height.

#### Vagueness of Hypothesis Tests

If we reject  $H_0: \beta_3 = 0$  and conclude  $H_A: \beta_3 \neq 0$  then we really haven't concluded anything other than there is an effect.

# CIs for a Single Slope

**Research Question:** Comparing individuals who played sports in HS to those who didn't, what is the difference in height?

#### Answer:

- A 95% confidence interval for  $\beta_3$  is calculated as (0.136,0.561).
- How do we interpret this interval?
- Holding all else constant, we are 95% confident that if a student played sports in HS vs not, we expect their height to be between 0.136 and 0.561 inches taller.
- Notice, that the interpretation says **expect** NOT will.

5) Regression Analysis						
Regression Analysis of: Mort (Y) Coefficient Table:	explained by Lat, Ocean, Long	g (X's)				
Confidence Level for Slope and Intercep	pt:					0.95 0.99
0.5 0.55 Show 5 ∨ entries	0.6	0.65	0.7	0.75	Set the confidence l	evel
Test				\$	F-statistic 🔶	p.value 🔶
1 F-test for all slopes are	equal to zero				50.826	0
Showing 1 to 1 of 1 entries						Previous 1 Next
	Estimate 🍦	Std. Error 🍦	t value 🍦	p value 🍦	CI Lower Bound 🍦	Cl Upper Bound 🔶
(Intercept)	<b>Estimate</b> 349.2369	<b>Std. Error</b> 27.0596	t value 12.9062	p value	CI Lower Bound	Cl Upper Bound 403.7377
(Intercept) Lat	Estimate   349.2369 -5.495	Std. Error           27.0596           0.5289	t value 12.9062 -10.3898	p value 🔶 0 0	CI Lower Bound 294.7361 -6.5602	Cl Upper Bound 403.7377 -4.4297
(Intercept) Lat Ocean	Estimate 349.2369 -5.495 21.7976	Std. Error           27.0596           0.5289           5.2263	t value 12.9062 -10.3898 4.1707	p value ♦ 0 0 0.0001	Cl Lower Bound 294.7361 -6.5602 11.2712	CI Upper Bound 403.7377 -4.4297 32.324
(Intercept) Lat Ocean Long	Estimate ↓ 349.2369 -5.495 21.7976 0.1219	Std. Error ↓           27.0596           0.5289           5.2263           0.1732	t value           12.9062           -10.3898           4.1707           0.7037	p value ♦ 0 0 0.0001 0.4852	CI Lower Bound 294.7361 -6.5602 11.2712 -0.227	CI Upper Bound 403.7377 -4.4297 32.324 0.4708
(Intercept) Lat Ocean Long Showing 1 to 4 of 4 entries	Estimate 349.2369 -5.495 21.7976 0.1219	Std. Error         27.0596         0.5289         5.2263         0.1732	t value 12.9062 -10.3898 4.1707 0.7037	p value ♦ 0 0 0.0001 0.4852	Cl Lower Bound 294.7361 -6.5602 11.2712 -0.227	Cl Upper Bound ↓ 403.7377 -4.4297 32.324 0.4708 Previous 1 Next
(Intercept) Lat Ocean Long Showing 1 to 4 of 4 entries R-squared: 0.7721 sigma-hat: 16.4806	Estimate 349.2369 -5.495 21.7976 0.1219	Std. Error         27.0596         0.5289         5.2263         0.1732	t value 12.9062 -10.3898 4.1707 0.7037	p value ♦ 0 0 0.0001 0.4852	CI Lower Bound 294.7361 294.7361 -6.5602 11.2712 -0.227 These are all the individual	CI Upper Bound       403.7377       -4.4297       32.324       0.4708       Previous     1       Next       (one slope) tests
(Intercept) Lat Ocean Long Showing 1 to 4 of 4 entries R-squared: 0.7721 sigma-hat: 16.4806 Proceed to Predictions	Estimate ↓ 349.2369 -5.495 21.7976 0.1219	Std. Error         27.0596         0.5289         5.2263         0.1732	t value (12.9062) -10.3898 4.1707 0.7037	p value ♦ 0 0 0.0001 0.4852	Cl Lower Bound 294.7361 294.7361 -6.5602 11.2712 -0.227 These are all the individual and corresponding confide usually don't look at the individual	CI Upper Bound 403.7377 -4.4297 32.324 0.4708 Previous 1 Next (one slope) tests nce intervals. We tercept line but you

#### **Nuances of MLR Inference**

Reminder that correlation is not causation:

- Just because you found a significant effect, does not mean that the explanatory variable causes and change in the response.
- Causation is established with experimentation

#### **Nuances of MLR Inference**

**Directionality:** MLR just exploits correlation even if the direction doesn't make sense. Does X lead to a change in Y or does Y lead to a change in X?

- 1. Does father's height lead to an increase in child's height or vice versa?
- 2. Does sports in high school lead to an increase in child's height or vice versa?

#### **Nuances of MLR Inference**

What do we do if the LINE assumptions aren't quite appropriate?

- Throw out outliers (not recommended)
- Ignore them and do inference anyway (but acknowledge that your inferences could be very wrong not recommended)
- Use more explanatory variables (we left a lot out).
- Consult a statistician (or better yet take more stats classes and we'll teach you)

## **Additional MLR Inference Practice:**

Measuring possum head size can be difficult. However, other characteristics of the possum which are easier to measure may be associated with head size. Use sex, age, total length and tail length as explanatory variables to explain head size and answer the following questions.

1. Do the LINE assumptions hold for the possum dataset? • Yes

- 2. Do any of sex, age, total length or tail length have an effect on head length? Use lpha=0.05.
- Yes the F statistic is 29.461 with a p-value of 0.

## **Additional MLR Inference Practice:**

Measuring possum head size can be difficult. However, other characteristics of the possum which are easier to measure may be associated with head size. Use sex, age, total length and tail length as explanatory variables to explain head size and answer the following questions.

- 3. Which of sex, age, total length and tail length have an effect on head length? Use lpha=0.05.
- All of them except tail length.
- 4. If the total length goes up by 1, how much do we expect the head length to change? Use 90% confidence level.
- Estimate of 0.6381 with a 90% interval of (0.5184, 0.7579).

# Key Terminology

- LINE Assumptions
- Overall Hypothesis tests

- Confidence intervals for  $\beta_1$
- Checking LINE assumptions
- Hypothesis tests for single slope