# Math 300 Lesson 11 Notes

# Simple Linear Regression - Continuous Predictor

#### YOUR NAME HERE

# June, 2022

# Contents

Objectives
Reading
Lesson
Documenting software

# **Objectives**

- 1. Use the skimr package to summarize multiple numerical variables in a data frame.
- 2. Build a scatterplot to describe the relationship between two continuous, numerical variables; use geom\_smooth() to visualize the best fit line.
- 3. Fit a linear regression model between two variables using the lm() function and interpret the output. This includes the interpretation of slope and the use of association and not causation.
- 4. Generate a table of observations, fitted values, and residuals from a linear regression object.

### Reading

Chapter 5 - 5.1

### Lesson

Work through the learning checks LC5.1 - LC5.3. Complete the code as necessary.

- Regression can be used for explanatory and predictive purposes. It falls on that line between traditional statistics/econometrics and machine learning. In this course we focus on its more traditional use to interpret the relationship between predictors and a response. Math 378 is our machine learning course and expands on linear regression in this framework.
- Note the many different terms for x and y in regression. These names come from different fields. For example, y is called the response, dependent variable, outcome, and output. Meanwhile, x is called input, predictor, independent variable, and explanatory variable. Also note that in linear regression, y is numerical while x can be numerical or categorical.

- We are using new packages. The tidyverse package is a wrapper and actually loads readr, dplyr, ggplot2, and tidyr.
- The interpretation of the slope has the key phrase **average**. For a one unit change in **x**, the average value of **y** changes by the value of the slope.

#### Setup

```
library(tidyverse)
library(moderndive)
library(skimr)
library(gapminder)
```

Create the data needed for the exercises.

```
# Select only the columns needed. Complete the code and remove comment labels
#evals_ch5 <- evals %>%
# select(ID, _____, bty_avg, _____)
```

Let's look at 5 random rows of data.

```
# Complete the code and remove comment labels
#set.seed(1234)
#evals_ch5 %>%
# sample_n(size = _____)
```

#### LC 5.1 (Objective 1)

(LC 5.1) Refer to the example in section 5.1.1. Conduct a new exploratory data analysis with the same outcome variable y being score but with age as the new explanatory variable x. Remember, this involves three things:

- Looking at the raw data values.
- Computing summary statistics.
- Creating data visualizations.

What can you say about the relationship between age and teaching scores based on this exploration?

### Solution:

• Looking at the raw data values:

```
# Complete the code and remove comment labels
#glimpse(_____)
```

• Computing summary statistics:

```
# Complete the code and remove comment labels
#evals_ch5 %>%
# select(______) %>%
# skim()
```

• Bivariate summary:

• Creating data visualizations:

```
# Create scatterplot. Complete the code and remove comment labels

#ggplot(______, aes(x = _____, y = score)) +

# geom_jitter(alpha=_____) +

# labs(

# x = "______", y = "Teaching Score",

# title = "_____") +

# geom_smooth(method = "lm", se = _____) +

# theme_classic()
```

# LC 5.2 (Objective 2)

(LC 5.2) Fit a new simple linear regression using lm(score ~ age, data = evals\_ch5) where age is the new explanatory variable x. Get information about the "best-fitting" line from the regression table by applying the get\_regression\_table() function. How do the regression results match up with the results from your earlier exploratory data analysis?

#### Solution:

```
# Complete the code and remove comment labels
# Fit regression model:
#score_age_model <- lm(______, data = _____)
# Get regression table:
#get_regression_table(score_age_model)</pre>
```

### LC 5.3 (Objective 3)

(LC 5.3) Generate a data frame of the residuals of the model where you used age as the explanatory x variable.

#### Solution:

# Documenting software

File creation date: 2022-06-21R version 4.1.3 (2022-03-10)

• tidyverse package version: 1.3.1

• skimr package version: 2.1.4

• gapminder package version: 0.3.0

• moderndive package version: 0.5.4