Simple Linear Regression - Tasks

M Douch based on Florian Oswald et al.

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Task 1

Task 1: Getting to know the data

- Load the data from [here] (https: //www.dropbox.com/s/wwp2cs9f0dubmhr/grade5.dta?dl=1) as grades.
- Need a function to import *.dta* files (Stata).

R Code

```
library(haven)
link <- "https://www.dropbox.com/s/wwp2cs9f0dubmhr/grade5.dta?dl=1"
grades <- read_dta(link)</pre>
```

Task 1: Getting to know the data

Alternative: load from the directory

R Code

- 1. Clear the environment (remove all objects) rm(list = ls())
- Check your current working directory getwd()
- 3. Set your working directory (optional, adjust path as needed) setwd("C:/Users/YourName/Path/To/Data")
- 4. Confirm the directory has the data file list.files()
- 5. Load the data grades <- read _dta("grade5.dta")

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- How many observations are there?

R Code: Number of Observations

nrow(grades)

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nrow(grades)

• What variables do we have?

R Code: Variable Names

names(grades)

*Note that if you view the data, under each column name you have the variable's label, which is very convenient.

• What do avgmath and avgverb correspond to?

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- Use skim() for summary statistics of classize, avgmath and avgverb.
- Use the 'skim' function from the 'skimr' package to obtain common summary statistics for the variables 'classize', 'avgmath' and 'avgverb'. (*Hint: use 'dplyr' to 'select' the variables and then simply pipe ('%>%') 'skim()'.*)

R Code: Summary Statistics with skimr

```
library(skimr)
library(tidyverse)
grades %>%
select(classize, avgmath, avgverb) %>%
skim()
```

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Note: Class sizes range from 5 to 44 (average \approx 30). Average math scores were slightly lower and more dispersed than average verb scores. No missing values.

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- What would you do to get a first insight? First Insight: A scatter plot would provide a first insight.

R Code: Scatter Plots

```
\label{eq:complete} \begin{array}{l} library(cowplot) \\ scatter \ verb <- \ grades \ \%>\% \\ ggplot(f) + aes(x = classize, y = avgverb) + geom\_point(f) + scale\_x\_continuous(limits = c(0, 45), breaks = seq(0,45,5)) + scale\_y\_continuous(limits = c(0, 100), breaks = seq(0, 100, 20)) + labs(x = "Class size", y = "Average reading score") \\ scatter\_math <- \ grades \ \%>\% \\ ggplot(f) + aes(x = classize, y = avgmath) + geom\_point(f) + scale\_x\_continuous(limits = c(0, 45), breaks = seq(0,45,5)) + scale\_y\_continuous(limits = c(0, 100), breaks = seq(0, 100, 20)) + labs(x = "Class size", y = "Average math score") \\ "Average math score") \\ plot\_grid(scatter\_verb, scatter\_math, labels = c("Reading", "Mathematics")) \\ \end{array}
```

- Compute the correlation between class size and math/verbal scores.
- Is the relationship positive/negative, strong/weak?

R Code: Correlation

```
grades \%>\% summarise(cor_verb = cor(classize, avgverb), cor_math = cor(classize, avgmath))
```

Task 2

Task 2: OLS Regression

Run the following code to aggregate the data at the class size level:

R Code: Data Aggregation

```
grades_avg_cs <- grades %>%
group_by(classize) %>%
summarise(avgmath_cs = mean(avgmath), avgverb_cs = mean(avgverb))
```

 Compute the OLS coefficients b₀ and b₁ of the regression of avgmath_cs on classize using the formulas:

$$b_1 = \frac{\mathsf{Cov}(x,y)}{\mathsf{Var}(x)}, \quad b_0 = \bar{y} - b_1 \bar{x}$$

(*Hint:* you need to use the 'cov', 'var', and 'mean' functions.)

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(*Hint:* you need to use the 'cov', 'var', and 'mean' functions.)

R Code: Manual OLS Calculation

```
cov_x_y = grades_avg_cs %>%
summarise(cov(classize, avgmath_cs))
var_x = var(grades_avg_cs$classize)
b_1 = cov_x_y / var_x
b_1

y_bar = mean(grades_avg_cs$avgmath_cs)
x_bar = mean(grades_avg_cs$classize)
b_0 = y_bar - b_1 * x_bar
```

- Regress average verbal score (avgverb_cs) on class size (classize).
- Interpret the coefficients b_0 and b_1 .

R Code: OLS Regression

lm(avgverb_cs ~classize, grades_avg_cs)

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- What is the predicted average verbal score when the class size is equal to 30 students?

- Is the slope coefficient for verbal score similar to the one for average math score? Was this expected?
- What is the predicted average verbal score when class size is equal to 0? **Answer**: $b_0 \approx 69.19$. This makes **no sense** in this context.
- What is the predicted average verbal score when the class size is equal to 30 students?

Task 3

Task 3: R^2 and Goodness of Fit

1. Regress avgmath_cs on classize. Assign to math_reg.

R Code: Math Regression

math_reg <- lm(avgmath_cs ~classize, grades_avg_cs)

2. Pass 'math_reg' in the 'summary()' function. What is the (multiple) R^2 for this regression? How can you interpret it?

R Code: Summary of Regression

summary(math_reg)

- 3. Compute the squared correlation between classize and avgmath_cs.
 - What does this tell you about the relationship between R^2 and correlation?

R Code: Squared Correlation

```
grades_avg_cs \%>\%
summarise(cor_sq = cor(classize, avgmath_cs)<sup>2</sup>)
```

- Use $R^2 = \frac{\text{SSE}}{\text{SST}}$, where SST = Var(y) and SSE = Var(\hat{y}) (predicted values).
- 4. Install and load the 'broom' package. Pass 'math_reg' in the 'augment()' function and assign it to a new object. Use the variance in 'avgmath_cs' (SST) and the variance in '.fitted' (predicted values; SSE) to find the R^2 using the formula on the previous slide.

R Code: R^2 from Variance

```
library(broom)
```

math_reg_aug <- augment(math_reg)</pre>

SST = var(grades_avg_cs\$avgmath_cs)

 $SSE = var(math_reg_aug\$.fitted)$

SSE/SST

- Repeat steps 1 and 2 for avgverb_cs.
- For which exam does class size explain more variance in students' scores?

R Code: Verbal Regression Summary

```
verb_reg <- lm(avgverb_cs ~classize, grades_avg_cs)
summary(verb reg)</pre>
```

- Repeat steps 1 and 2 for avgverb_cs.
- For which exam does class size explain more variance in students' scores?

R Code: Verbal Regression Summary

```
verb_reg <- lm(avgverb_cs ~classize, grades_avg_cs)
summary(verb reg)</pre>
```

Comparison: The R^2 is greater for maths (≈ 0.28) than for reading (≈ 0.046). Therefore, class size explains more of the variation in math scores than in reading scores.