

Math 455 Class 8 February 7

New homework is posted to Gradescope.

How to do this homework:

use the universal textbook problem-solving method:
imitate the examples in the textbook.

in this book there are often italicized bits of advice / rules of thumb that are very useful.

Example Ch 5, p. 59 on the interpretation of regression coefficients. (sentence interprets $\hat{\beta}_1$)

There is some art to this. First examples that are bad:

BAD: Eating 100 avocados per day caused patients to have blood pressure 1000 points lower.

Why this is bad: 3 reasons.

- implies a causal connection. Remember ice-cream sales and shark attacks.

Regression does not prove a causal connection.

We can use regression to give evidence for a causal connection (Hill Criteria, 5.7)

- Hard to imagine the change in explanatory variable, as it is impossible (?) to eat 100 avocados per day, every day, for the rest of your life.
- The change in dependent variable (response) is also beyond all comprehension.

Fixing these three things:

Better: Patients who ate 1 avocado per day tended to have blood pressure 10 points lower.

Notice that we have changed the numbers and "tended to have" does not imply a causal connection.

More subtle problems: see, e.g. "changes of scale", p. 103

If your regression coefficient is $\hat{\beta}_1 = 0.0000009134$

Hard to read and interpret!

Maybe you should change units: cm vs. km?

Q: What if there is no way to do this?

A: Fix it somehow! There is always a way, e.g. drop the variable.

Running a regression in R

Example p. 36

```
lmod <- lm ( Species ~ Area + Elevation, gata)
```

↑

linear model
object

: contains results. we get data out of this, data like regression coeff, p-values, using R functions, like summary.

lm is the "linear model" function.

"Species ~ Area + Elevation" is Wilkinson-Rogers notation for specifying the model.

in this case $Y = \text{Species}$ $X_1 = \text{Area}$
 $X_2 = \text{elevation}$.

the model specified is

$$Y = \beta_0 + \beta_1 \text{Species } X_1 + \beta_2 X_2 + \varepsilon.$$

There are lots of bells/whistles/conventions in this notation.

Some of these are hard to understand, e.g.

"Species ~ ."

this means "all variables in the data set OTHER THAN Species"

Other things that can be done: remove the intercept, change the regression so one ~~one~~ or more coefficient is specified in advance, etc.

More about the homework:

Textbook (Faraway) does NOT spend a lot of time explaining the theory. More info in Wackerly, et.al, Ch. 11.

In particular a regression has an F-statistic. this is used for the test of H_0 : all β_i are 0 (except β_0) vs. H_a : at least one of $\beta_1, \dots, \beta_p$ is not 0.

There is also a relative F-test, which can be used to compare 2 nested models.

Nested means there is a small model contained in a large one.

Model 1: $Y = \beta_0 + \beta_1 X_1 + \beta_2 X_2 + \varepsilon$

Model 2: $Y = \beta_0 + \beta_1 X_1 + \beta_2 X_2 + \beta_3 X_3 + \beta_4 X_4 + \varepsilon$

nested: Model 2 $\supset$ Model 1.

Models that are not nested:

Model 3: $Y = \beta_0 + \beta_1 X_1 + \beta_3 X_3 + \varepsilon$

Model 4: $Y = \beta_0 + \beta_1 X_1 + \beta_2 X_2 + \beta_4 X_4 + \varepsilon$.

Model 3 has X_3 , model 4 does NOT.

Even though model 4 is larger, it does NOT contain model 3.

Improving your regression: add/drop variables
transform variables.

"transform" might mean: replace Y with $\log(Y)$
or $\sqrt{Y}$ or ..., replace X with X^2 , or

add X^2 to the variable set etc.

Notice that this may:

- change coefficients

- change interpretation of coefficients.

- change significance of coefficients.

This allows us to consider

$$Y = C e^{\beta_1 X_1 + \beta_2 X_2 + \varepsilon} \quad \text{exponential.}$$

$$\log Y = \log C + \beta_1 X_1 + \beta_2 X_2 + \varepsilon$$

linear!

See problem 2 on HW.