

Recitation 2: Introduction to Research Methods for Politics

Dept. of Politics, NYU

POL-850

Spring 2020

Reminder

Homework 1 due this Wednesday (2.12), 5 pm, in my mailbox:

- ▶ Only typed work (pdf file generated by R Markdown), no handwriting.
- ▶ When answering a question, please include related R code.

How to use the template

1. Put your code in the R code chunk (gray chunk)

```
54 ▾ ## Question 1.1 (6 pts)
55
56 Load the data into R and check the dimensions of the data.
57 How many observations are
58 there? Using the function 'head()' and the argument 'n', list
59 all observations. What years are included in the dataframe? What type is each variable in the dataframe?
60
61 ▾ ## Answer 1.1
62
63 ▾ ```{r}
64 ## insert code here
65 ```
66
67 Insert written answer here
68
69 ▾ ## Question 1.2 (6 pts)
70
```

2. Comment your code using `#`

Your Friends to Work in R (1)

Coding can be exhausting, but not if you get help from some friends:

- ▶ The code I use during recitation. You can find it on NYU Classes, and use it as a baseline for your own code.
- ▶ Google: a query like "How to compute mean in R" will return dozens of results, with practical examples and code snippets. **Ideal when you do not know how to do something.**

Your Friends to Work in R (2)

Coding can be exhausting, but not if you get help from some friends:

- ▶ R, by using the question mark: `?mean`. **Ideal when you know a function exists, but you are not sure about the details.**
- ▶ The TAs!

Agenda

1. Review: Causal Effects
2. Relational Operators
3. Basic functions (ifelse, subset)

Causal Effects

Causal Effects

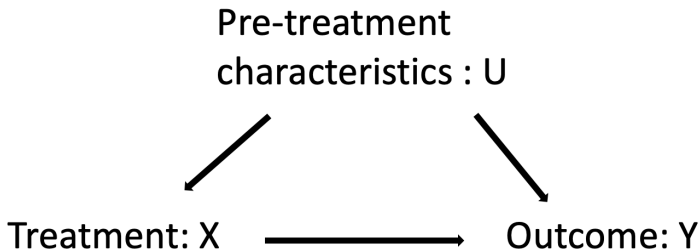
1. We are interested in : Causal Effect of X on Y
(e.g. Class size \rightarrow Student performance?
Democracy \rightarrow Economic development ?)
2. Potential outcome if individual i in the treatment group: $Y_i(1)$
Potential outcome if individual i in the control group: $Y_i(0)$
3. Individual causal effect: $\Delta Y_i = Y_i(1) - Y_i(0)$
4. Fundamental Problem of Causal Inference:
"To measure causal effects, we need to compare the factual outcome with the counterfactual outcome, but we can never observe the counterfactual outcome."

Causal Effects

1. Average causal effect: $\overline{\Delta Y} = \overline{Y(1)} - \overline{Y(0)}$
where $\overline{Y(1)}$ and $\overline{Y(0)}$ are the average outcomes for *all* observations.
2. We still face the missing data problem; Can you see why?
3. We can estimate it by: $\widehat{\Delta Y} = \widehat{\overline{Y(1)}} - \widehat{\overline{Y(0)}}$
where $\widehat{\overline{Y(1)}}$ and $\widehat{\overline{Y(0)}}$ are the average outcomes for *observed* units.
4. Are $\overline{Y(1)}$ and $\widehat{\overline{Y(1)}}$ the same? If not, when can we say $\widehat{\Delta Y}$ is a good estimate for $\overline{\Delta Y}$?

Causal Effects

1. How to define "a good estimate"? In the textbook, authors say $\widehat{\Delta Y}$ is unbiased.
2. How to guarantee "unbiased"?



Relational Operators

Why Bother?

Our ultimate goal as empiricists is to (credibly) compare units under treatment status ($T = 1$) and control status ($T = 0$).

- ▶ This requires splitting our data based on T !

This way, we will be able to:

- ▶ Compare outcomes: is $Y(1) = Y(0)$?
- ▶ Check that units are comparable based on observables: is $X(1) = X(0)$?
- ▶ And much more...

Behind Subsetting: Logical Operators

To subset, we need R to understand whether a given condition is satisfied (TRUE) or not (FALSE). TRUE and FALSE belong to a new class of objects in R, the one of *logicals*.

```
5 class(TRUE)
6
7 as.integer(TRUE)
8 as.integer(FALSE)
9
10 x <- c(TRUE, FALSE, TRUE) # a vector with logical values
11
12 mean(x) # proportion of TRUES
13 sum(x) # number of TRUES
14
15 FALSE & TRUE
16 TRUE & TRUE
17 TRUE | FALSE
18 FALSE | FALSE
19 TRUE & FALSE & TRUE
20
```

NOTE: Use "AND" ("&") and "OR" ("|") to evaluate multiple conditions.

Logical Operators to Evaluate Relationships

```
32 4 == 4
33 4 > 3
34
35 x <- c(3, 2, 1, -2, -1)
36 x == -1
37 x >= 2
38 x != 1
39
40 ## logical conjunction of two vectors with logical values
41 (x > 0) & (x <= 2)
42
43 ## logical disjunction of two vectors with logical values
44 (x > 2) | (x <= -1)
```

NOTE: The relational operator "**!=**" means "not equal to".

Practice with STAR dataset

```
49 star <- read.csv("STAR.csv")
50 head(star, n = 10) #### show first 10 rows
51
52 star$classtype == "small"
53
54 sum(star$classtype == "small")/nrow(star)
55 mean(star$classtype == "small")
56
57 tmp1 <- (star$classtype == "small") & (star$reading >= 600)
58 tmp2 <- (star$classtype == "regular") & (star$reading >= 600)
59
60 mean(tmp1)
61 mean(tmp2)
```

Basic Functions: ifelse and subset

Introduction: ifelse

Syntax: `ifelse(condition, return values if yes, return values if no)`

```
70 ifelse(star$classtype == "regular", 1, 0)
71
72 star$regular <- ifelse(star$classtype == "regular", 1, 0) ## create a new var
73 head(star, n=4)
74
75 star$treatment <- ifelse(star$classtype == "small", 1, 0) ## create a new var
76
77 mean(star$math)
78 mean(star$math[star$treatment == 1]) ## mean for treatment group
79
80 #### difference-in-means estimators
81 mean(star$math[star$treatment==1]) - mean(star$math[star$treatment==0])
82 mean(star$reading[star$treatment==1]) - mean(star$reading[star$treatment==0])
83 mean(star$graduated[star$treatment==1]) - mean(star$graduated[star$treatment==0])
```

Introduction: subset

Syntax: `subset(data.frame, condition)`

```
88 subset(star, classtype == "small")
89
90 ## sub dataset only contains small class
91 star_small <- subset(star, classtype == "small")
92 star_small
93
94 star_regular <- subset(star, classtype == "regular")
95
96 ### now, we can use this new data.frame
97 mean(star_small$math) - mean(star_regular$math)
98 mean(star_small$reading) - mean(star_regular$reading)
99 mean(star_small$graduated) - mean(star_regular$graduated)
```

Introduction: subset

We can specify multiple conditions to the function subset.

```
103 star_small_grad <- subset(star,  
104                           subset = (classtype == "small" & graduated == 1))  
105 star_small_nongrad <- subset(star,  
106                             subset = (classtype == "small" & graduated == 0))  
107  
108 mean(star_small_grad$reading) - mean(star_small_nongrad$reading)  
109 mean(star_small_grad$math) - mean(star_small_nongrad$math)  
110
```