Lecture 1

Quantitative Political Science

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Lecture Date: 2023/08/24

Slides Updated: 2023-08-21

Agenda

- 1. Meet the instructor
- 2. Why are you here?
- 3. Variables 101
- 4. Data and summarizing data
- 5. Review syllabus and expectations

Meet the instructor

- Education
 - PhD from NYU Politics in 2019
 - Postdocs at Princeton Niehaus & NYU CSMaP
- Published some things
 - Methods-ey: external validity 1, 2; measurement 3, 4
 - Substantive: economics & populism 1; Covid-19 & U.S. politics 2, 3; IPE 4; academic naval-gazing 5
 - Popular press: 1, 2, Podcasts
- Work
 - World Bank / IFC
 - MarketCast

Why are you here?

- I.e., why are you getting a PhD in political science?
 - You enjoy asking and answering questions about politics
 - What questions are you interested in?

How I can help

- Quantitative analysis is one of many tools to answer your questions
 - Based on **numerical** measurements
 - Interested in developing and testing **generalizable** theories
 - Measurements and analyses that are easily replicable by others
- Contrast to other two dominant paradigms in political science research
 - 1. Qualitative analyses
 - 2. Formal modeling
- Never ever fall into the petty trap that quantitative is somehow "better"
 - "A foolish consistency is the hobgoblin of little minds, adored by little statesmen and philosophers and divines."
- You are here because you do not have a little mind

Quantitative Analysis

- Political scientists work with quantitative data for three reasons:
 - 1. What can we say about the data **we have**?
 - 2. What can we say about the data we don't have?
 - 3. What can we say about the data we'd expect to see?
- Answering these questions requires three types of statistics
 - 1. Descriptive
 - 2. Inferential (from samples to populations)
 - 3. Prediction (from models to hypotheticals)
- All three of these approaches rely on a test statistic
 - A number that summarizes data

Variables 101

- We study **units**
 - Phenomena about which we wish to make statements
 - AKA **cases**: people, counties, nations, dyads, etc.
- Units have attributes
 - Characteristics of a **unit** that distinguish it from other units
- Variables are *logical* groupings of *mutually exclusive* attributes
 - o An important part of quantitative research is assigning a value to each attribute
 - The variable GDP per capita takes on the value \$2,256 for the unit India
 - The variable year takes on the value first for the unit you
- In quantitative analysis, we assign **scores** to each **value**

Levels of measurement

- Variables can be measured at four levels
 - 1. **Nominal**: cannot be ordered in any logical way
 - 2. **Ordinal**: can be ordered, but no meaning to differences
 - 3. Interval: ordinal variables whose differences can be compared
 - 4. Ratio: zero is meaningful -- nothing of the quantity measured
- Mathematical operations can be conducted on these levels
 - 1. **Nominal**: equality (=) only (do these take on the same value or not?)
 - 2. **Ordinal**: equality, greater than (>) or less than (<)
 - 3. **Interval**: addition (+), subtraction (), averages ($rac{1}{n}\sum_{i}x_{i}$)
 - 4. **Ratio**: multiplication (*) and division (/)

Tricky cases:

• Celsius? Latitude and longitude? Binary variables?

Data structures

- Data table (or data frame)
 - Rows are units
 - Columns are variables
 - Cells are scores
- **List**: Tree-like structure
 - Units are outermost node
 - Attributes are child nodes
 - Scores are children of attributes
- We will only use data table / data frames this semester
- But if you want to communicate with CS / engineers / data scientists, **lists** are their world!

```
## id state age GPA
## 1 PE MI 25 4.04
## 2 VF MT 24 3.73
## 3 EP CO 23 3.84
## 4 LD IA 25 3.90
## 5 OB IN 29 3.97
## 6 IG VA 27 3.84
```

Why not just present this table as is?

```
id state age
##
                    GPA
##
                25 3.90
                24 4.08
      VE
            DE
      EB
            MT
                24 3.79
                25 4.04
                24 4.03
                24 3.93
                26 3.90
                23 3.85
                28 3.86
## 10 DI
                24 3.96
            VC 20 / 11
```

- Fundamental tension in quantitative analysis: detail versus parsimony
- Use a frequency table?

```
## 1 23 1
## 2 27 1
## 3 26 2
## 4 28 2
## 5 29 3
## 6 30 4
## 7 25 6
## 8 24 7
```

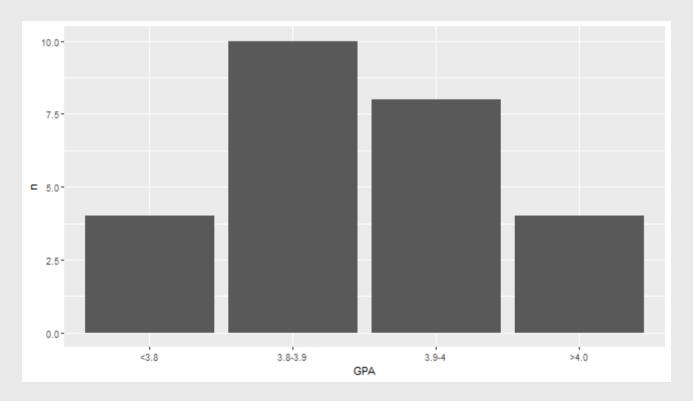
What about for GPA?

```
##
         GPA n
      3.7609 1
     3.7628 1
      3.7805 1
     3.7861 1
     3.8276 1
##
     3.8545 1
     3.8597 1
## 7
     3.8633 1
     3.8683 1
## 10 3.8808 1
## 11 3.8822 1
## 12 3.8826 1
## 13 3.8830 1
## 14 3.8847 1
## 15 3.8992 1
## 16 3.9007 1
  17 3.9258 1
## 18 3.9336 1
## 19 3.9550 1
```

• **Recode** data into categories, then use a frequency table

```
## GPA n
## 1 <3.8 4
## 2 3.8-3.9 10
## 3 3.9-4 8
## 4 >4.0 4
```

• Also can visualize with a **plot**



Summarizing data: central tendency

- Central Tendency: The typical value
 - Mode: most frequently observed value (which levels of measurement (LOM)?)
 - \circ **Median**: value of smallest observation for which the cumulative percentage is \geq 50 (which LOM?)
 - \circ **Mean**: average $ar{y} = rac{1}{N} \sum_{i=1}^N y_i$ (which LOM?)

Summarizing data: dispersion

- **Dispersion**: The *spread*
 - **Range**: difference between smallest and largest values (LOM?)
 - IQR: difference between 75th%ile and 25%ile (LOM?)
 - \circ Variance: $s^2 = rac{1}{N} \sum_{i=1}^N (y_i ar{y})^2$ (LOM?)

Summarizing data: qualitative description

- Frequency distribution may be "symmetric" or "skewed"
 - Median is typically better than mean if data is skewed
- May be "unimodal" or "bimodal"
- Qualitative descriptions can be quantified

$$\circ$$
 I.e., skew $g_1 = rac{1}{N*s^3} \sum_{i=1}^N (y_i - ar{y})^3$

- \circ If $g_1=0$, symmetric
- \circ If $g_1 < 0$, skewed left
- \circ If $g_1 > 0$, skewed right

Logistics

- Syllabus review
- TA office hours and labs