Week 2 Handout

Gov 50 Data Science for the Social Sciences

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Math Recap

Probabilities and expectations

	Discrete Variables	Continuous Variables
Distribution		
	41	
Probability	$\Pr(X = x) = \frac{n_x}{n}$	f(a < x < b) = F(b) - F(a)
Expectation	$\mathbb{E}[X] = \sum_{x} x \Pr(X = x)$	$\mathbb{E}[X] = \int_{-\infty}^{\infty} x f(x) dx$
Expectation	$\mathbf{E}[\mathbf{x}] = \mathbf{Z} \mathbf{x} \mathbf{x} \mathbf{r} \mathbf{r} (\mathbf{x} - \mathbf{x})$	$\mathbb{E}[X] = \int_{-\infty}^{\infty} x f(x) dx$

Q: What is the expectation of a six-sided die?

Let *X* be the number on the die,

$$\mathbb{E}[X] = 1 \cdot \Pr(X = 1) + 2 \cdot \Pr(X = 2) + \dots + 6 \cdot \Pr(X = 6)$$

$$= 1 \cdot \frac{1}{6} + 2 \cdot \frac{1}{6} + \dots + 6 \cdot \frac{1}{6}$$

$$= \frac{21}{6}$$

$$= 3.5$$

Conditional Probabilities and Conditional Expectation

Bayes Rule:

$$Pr(A \mid B) = \frac{Pr(A \cap B)}{Pr(B)} = \frac{Pr(B \mid A)Pr(A)}{Pr(B)}$$

Q: What is the expectation of a six-sided die, given that the number is even?

$$\mathbb{E}[X \mid \text{even}] = 2 \cdot \frac{1}{3} + 4 \cdot \frac{1}{3} + 6 \cdot \frac{1}{3}$$
$$= \frac{12}{3}$$
$$= 4$$

Potential Outcomes Framework and Causal Inference

Class practice

Introduction

Are democracies better for economic development than nondemocracies (autocracies)? Could having a democratic regime be related to economic growth? These questions have fascinated social scientists, policy-makers and pundits for many decades now. In a recent publication, Acemoglu, Naidu, Restrepo and Robinson (2019) describe the debate in the following terms:

"With the spectacular economic growth under nondemocracy in China, the eclipse of the Arab Spring, and the recent rise of populist politics in Europe and the United States, the view that democratic institutions are at best irrelevant and at worst a hindrance for economic growth has become increasingly popular in both academia and policy discourse. For example, the prominent New York Times columnist Tom Friedman (2009) argues that 'one-party nondemocracy certainly has its drawbacks. But when it is led by a reasonably enlightened group of people, as China is today, it can also have great advantages. That one party can just impose the politically difficult but critically important policies needed to move a society forward in the 21st century'. Robert Barro (1997, 1) states this view even more boldly: 'More political rights do not have an effect on growth.'"

In this exercise we'll take a stab at this debate by answering the following related question: are democracies richer than autocracies?

We will be working with the dataset from the paper by Acemoglu, Naidu, Restrepo and Robinson (2019). The dataset includes the following variables:

Name	Description	
country_name	Country name	
wbcode	World Bank country code	
year	Year	
gdppc	GDP per capita (constant 2000 US\$)	
region	Geographical region	
dem	Democracy measure (1 = Democracy; $0 = Autocracy$)	

Question 1: Loading the dataset

Before we can get started working with data, we first need to load the data into R. Datasets can come in many file types, but the most common is a CSV, which stands for "comma-separated values". Use the read.csv() function from the R package readr to read your data into R and call it anrr. You'll find the dataset under the folder data. This is the original data used in their study.

Question 2: Inspecting the dataset I

Use the head() function to view the first several rows of the data. What can you notice about the variable gdppc?

Answer 2

```
head(anrr)
## # A tibble: 6 x 7
##
      ...1 country_name wbcode year gdppc region
                                                    dem
                               <dbl> <dbl> <chr> <dbl>
##
     <dbl> <chr>
                        <chr>
## 1
         1 Afghanistan AFG
                                1960
                                        NA MNA
         2 Afghanistan AFG
                                                      0
## 2
                                1961
                                        NA MNA
## 3
         3 Afghanistan AFG
                                1962
                                        NA MNA
                                                      0
         4 Afghanistan AFG
                                1963
                                                      0
## 4
                                        NA MNA
## 5
        5 Afghanistan AFG
                                1964
                                        NA MNA
                                                      0
## 6
         6 Afghanistan AFG
                                1965
                                                      0
                                        NA MNA
```

There are 9384 country-years with missing values in the GDP per capita variable.

Question 3: Inspecting the dataset II

Use the function glimpse() to look at a summary of the dataset. What can you notice about the variable dem?

```
library(dplyr)
##
## Attaching package: 'dplyr'
## The following objects are masked from 'package:stats':
##
                                              filter, lag
##
## The following objects are masked from 'package:base':
##
                                              intersect, setdiff, setequal, union
##
glimpse(anrr)
## Rows: 9,384
## Columns: 7
## $ ...1
                                                                                                                    <dbl> 1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14, 15, 16, 17~
## $ country_name <chr> "Afghanistan", "Afghanistan", "Afghanistan", "Afghanistan"
                                                                                                                    <chr> "AFG", "AFG"
## $ wbcode
## $ year
                                                                                                                    <dbl> 1960, 1961, 1962, 1963, 1964, 1965, 1966, 1967, 1968, 196~
## $ gdppc
                                                                                                                    <chr> "MNA", "MNA", "MNA", "MNA", "MNA", "MNA", "MNA", "ANA", "ANA", "MNA", "MNA"
## $ region
```

The variable dem is numeric (a double).

Question 4: Measuring democracy

There are many potential ways to code if a country is democratic or not. Researchers came up with criteria to classify political regimes as a binary variable (if you are interested how, check Boix, Miller and Rosato 2012). In these measurements, democracies are often coded as a 1, and nondemocracies (autocracies) as a 0.

Use the function table to see how many observations are democracies and how many autocracies in the data. (Hint: to tabulate the values of a variable, pass as arguments to table something of the form dataframe\$variable).

Answer 4

```
table(anrr$dem)

##
## 0 1
## 4956 3777
```

Question 5

Now add as an argument to the function table the option useNA = "always". How many missing values does the variable dem has?

Answer 5

```
table(anrr$dem, useNA = "always")
##
## 0 1 <NA>
## 4956 3777 651
```

Question 6

When we create data visualizations, we sometimes want to make numeric variables like dema factor. In this way, we can acknowledge that, although imported as numeric, the variable represents two distinct categories: democracy and autocracy. Run the following code:

Since we are only comparing democracies and autocracies, we can also leave aside the NAs for visualization sake.

```
anrr <- anrr |>
filter(!is.na(dem))
```

Check the class of the new variable and corroborate that it shares the same number of democracies and autocracies, but that we have no NAs.

Answer 6

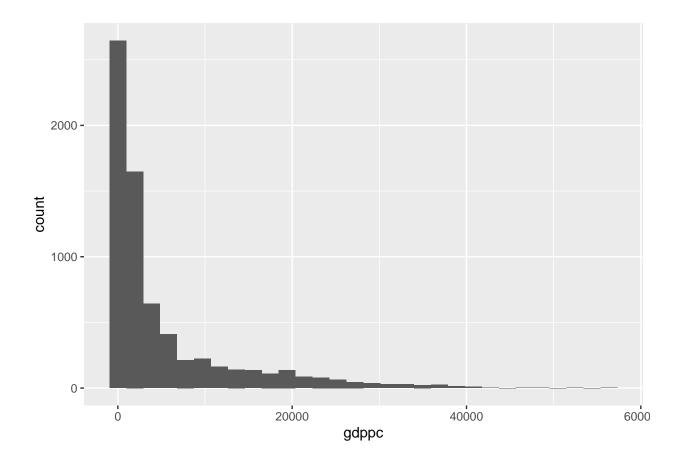
Question 7: Visualizing the income distribution

Now we can start comparing how rich (or poor) are democracies relative to autocracies.

Using ggplot, plot an histogram of the variable gdppc. What can you say about the distribution of the variable?

```
library(ggplot2)

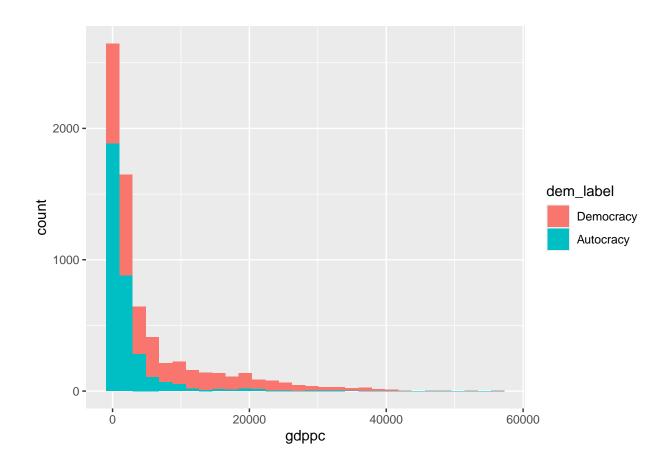
ggplot(anrr, aes(x=gdppc)) +
    geom_histogram()
```



Question 8: Comparing income by regime

The plot above is showing the distribution of income for both democracies and autocracies together. Pass the argument fill = dem_label to the aes() to see how the distributions differ by political regime. From this plot, can you tell if democracies are richer or poorer than autocracies?

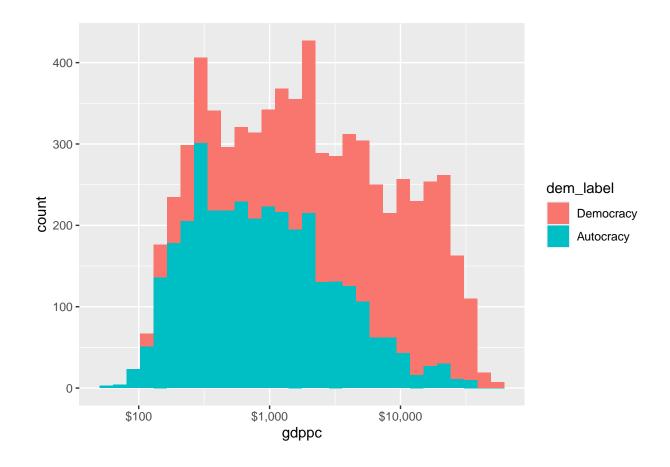
```
ggplot(anrr, aes(x=gdppc, fill = dem_label)) +
  geom_histogram()
```



Question 9: Log scale

When the distribution of our variable is highly skewed, we often transform the variable by a logarithmic scale. Make the same plot as in 2 above, but adding scale_x_log10(labels = scales::dollar) as an argument to the ggplot. Does this transformation makes clearer the income differences between democracies and autocracies?

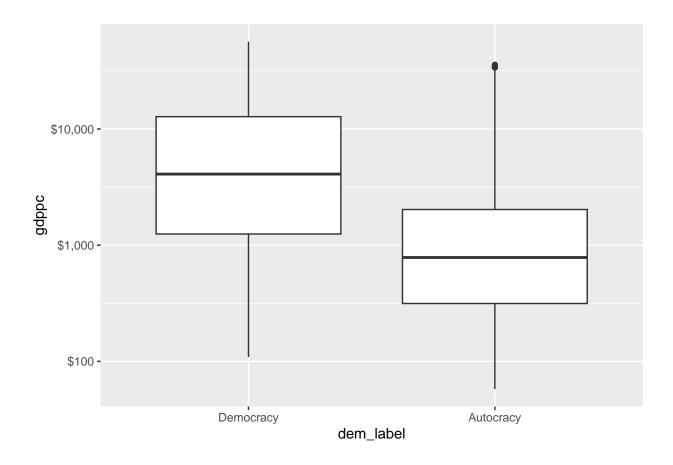
```
ggplot(anrr, aes(x=gdppc, fill = dem_label)) +
  geom_histogram() +
  scale_x_log10(labels = scales::dollar)
```



Question 10: Comparing with boxplot

An alternative way to get to the question is to compare the median income between political regimes. We can do this using geom_boxplot. What does this plot tells you about the distribution of income in democracies when compared to autocracies?

```
ggplot(anrr, aes(x=dem_label, y=gdppc)) +
  geom_boxplot() +
  scale_y_log10(labels = scales::dollar)
```



Question 11: Comparing by country groups

Finally, do these patterns vary by world region? Add facet_wrap(~region) to your ggplot to see the division by geographic/economic region. Region acronyms are AFR: Africa , EAP: East Asia and the Pacific, ECA: Europe and Central Asia, INL: OECD and high income countries, LAC: Latin America and Caribbean, MNA: Middle East and North Africa, SAS: South Asia. Can you find a region of the world where the median income of autocracies is higher than that of democracies?

```
ggplot(anrr, aes(x=gdppc, fill = dem_label)) +
  geom_boxplot() +
  scale_x_log10(labels = scales::dollar) +
  facet_wrap(~region)
```

