ECON 172 - Section 1: Introduction to R

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1 Welcome to Econ 172 Section

Hi! We are your GSIs for Econ 172 this semester, Roberto and Jed. Our goal is to make these sections as interactive as possible. The more you and your classmates participate, the more everyone will get out of section. We will mainly focus on reviewing the econometric methods discussed in lecture and applying them using R.

Feel free to attend any of the sections led by the GSI of the section you enrolled in.

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- Monday 2-3 PM (101)
- Tuesday 10-11 AM (102)
- Friday 9-10 AM (106)
- Office Hours: TBD

Jed jed_silver@berkeley.edu

- Wednesday 3-4 PM (103)
- Thursday 9-10 AM (104)
- Thursday 5-6 PM (105)
- Office Hours: Thursday 6-7 PM

All of the Zoom links for section are on bCourses. We will also post all materials on bCourses.

Please try to attend section regularly if you are in a time zone that allows you to do so and if not, please watch the recordings that we will post on bCourses. As in lecture, we'd appreciate if you keep your camera on for section, but understand that this may not always be possible for all students. We understand that things are still difficult this semester and we will try to be as accommodating as possible, so please communicate any issues to us and we will try our best to help.

Feel free to email with any questions you may have that can be answered over email, for example, administrative questions. Please include [ECON 172] in the subject line of any email communications. We will do do our best to respond to all emails within 48 hours, perhaps longer over weekends. For in-depth questions that will require a longer response, come by office hours instead. Often an in-person discussion can help clear things up more effectively than communication over email.

2 Welcome to R

R is a programming language and environment for statistical computing and graphics which is free to use and runs on Windows, Unix and MacOS. To do analysis in R, one can either write one's own commands or take advantage of the huge number of pre-existing packages written by R users and freely available. We are

not assuming any prior knowledge of R. If you are not familiar with programming, it can be a bit tricky at first but we are here to make sure you get the hang of it. R is used throughout the social sciences and in industry, so while this course will provide a basic overview, you may find it wortwhile to dive a bit deeper and really master it.

2.1 Getting Started

2.1.1 Using R on your local machine (optional)

R can be downloaded from https://cran.rstudio.com/. We also recommend downloading RStudio, a user interface which makes R easier to use. It can be downloaded from https://www.rstudio.com/products/rstudio/download/.

If you run R on your local computer, you will need to install any packages you want to use (just one time), then load them into R's library for the current session. After that, it's easy to access any of the commands in the package (which are also generally extensively documented in a sort of user guide for the package).

2.1.2 Using R on your UC Berkeley Datahub (recommended)

RStudio also runs on the UC Berkeley Datahub, so you don't need to install it locally on your computer. To start RStudio on Datahub, simply go to: https://r.datahub.berkeley.edu/

In our sections (and for the problem sets), we are going to show you this way of using RStudio. However, anything that we show you should also work on your local machine.

2.2 R Code

When we use R, we want to write and save our commands in a script. This will let us reproduce everything we do without having to retype our commands every time we work on a project. To do this, you can write and save a file as a .R file, which typically contains only code and code comments. You can find a separate .R file containing the code used to conduct the analysis below in the folder with the material for this section.

2.3 RMarkdown

What's even nicer about R is a file format called RMarkdown, such as the one you are currently working in. This file type allows you to combine text, mathematical equations (in a format called LaTeX), code as well as dynamic output from this code, and automatically compile / output everything into a pdf document. (In RMarkdown language, this process is called "knitting" the document.)

2.4 Setting up the Environment

First, we need to set up the R environment, including installing all the required packages, loading all the required packages. This is done using R-code. The way to let RMarkdown know that the following output is a 'code chunk' is by enclosing the code chunk in triple quotations, as below. The code chunk indicator is followed by curly brackets, and r (indicating this code chunk is in the R language), a name for the code chunk (here "setup"), and some additional parameters. The parameter "results='hide'" and "message=FALSE" lets RMarkdown know that we will not want to include the output of or any messages generated by this code chunk in the pdf document (but the code will still run).

In your code, you can add comments, using # (beginning of the line) and ## (end of the line). These comments are not run as code, but will help your later self (and others reading your code) to understand what you were doing.

```
knitr::opts_chunk$set(echo = TRUE) ## sets global option to include code chunks in-line
#install.packages("haven") ## only run once to install
library(haven) #this library allows you to load datasets in Stata format
#install.packages("tidyverse") ## only run once to install
```

```
library(tidyverse)
#install.packages("summarytools") ## only run once to install
library(summarytools)

## Warning: package 'summarytools' was built under R version 4.0.3
#install.packages("stargazer") ## only run once to install
library(stargazer) ##This package is great for making tables in .html, .tex and many other formats.

## Warning: package 'stargazer' was built under R version 4.0.3
#install.packages("broom") ## only run once to install
library(broom)
```

The code chunk specifies that, by default, code chunks are part of the pdf output. Next, it installs a few packages. You only need to run this installation once. Afterwards, the package will remain installed on your datahub. (Remove the # before the install packages commands to install each package for the first time, then "comment out" these lines using #, as above.) Next, the "library" command lets R know that you will want to use these packages in your code.

2.5 Loading Data

The first step in using R for statistical analysis is to store data in what R calls a "data frame," a matrix whose columns have different modes (like numeric, words, etc.) and each row is a unit of observation (i.e. a person or a country, etc.)

R can load data from many types of files, including .csv and .txt. R has its own type of data file, with a .rds or .Rdata extension, and can also load data from Stata, SPSS, and SAS (other statistics computing programs) with the appropriate commands. The .rds format allows for a single object to be saved at a time, like a single dataset. The .RData format allows for multiple objects to be saved at once. Either one is a great way to save your data in R format.

For this section, we will be using data from Robinson, Acemoglu and Johnson (2001). For your convenience, the following link imports this data directly into your Datahub repository: Datahub The folder should now appear as "ECON-172-Sp21" on your workspace (bottom right). Click on the folder, then "Section01" and then click on the .Rmd file to open the file that produced the PDF you're reading.

When we load a dataframe in R, we give it a name for reference, such as mydata or something more specific, like prop_rights. We also need to tell R where to find the data. We can do that by specifying a full file path to its exact location of the data. For the purposes of this section, we'll use the dataset prop_right" which is an example dataset that contains data on a property rights index (protection against expropriation risk), GDP per capita and other variables for countries that were European colonies in the past.

```
## Load in data in csv form, from the folder just loaded into your directory.
## Important: Paths are always relative to the RMarkdown file location
## Alternatively, specify your working directory using
## knitr::opts_chunk$set(root.dir = "xxx")
prop_rights <- read.csv("prop_rights.csv")</pre>
```

Next we can visualize the database we just imported, and take a look at what variables are included. To do so, it's easiest to click directly on the dataset in your Environment pane on the top right in RStudio.

```
## Let's take a look on the dataset.
prop_rights ## print the dataframe to screen
head(prop_rights) ## or print first few observations

## Let's take a look the variables in this dataset.
names(prop_rights) ## just print
```

```
prop_rights_variables <- names(prop_rights) ## store list, assigned "prop_rights_variables"
prop_rights_variables ## take a look at this list</pre>
```

3 Analysis in R

3.1 Summary Statistics

Once a dataframe is loaded in R, it's very easy to run basic summary statistics. Unlike many other statistics programs, R provides fairly minimal output, and it's possible to store the results of a command without ever displaying them. For example, we may be interested in summary statistics for the GDP per capita of countries in the sample, which we can get by typing {summary(prop_rights\\$gdppc). This command instructs R to look in the dataframe listed before the \$ symbol (in this case, prop_rights), and to summarize the variable that follows the \$ symbol (in this case, gdppc). Note, the results = 'markup' bit tells RMarkdown to display the output of the code exactly as displayed in the R console.

```
## Let's look at some summary statistics.
## Here is the min, 25%-ile, median, mean, 75%-ile, and max for the variable qdppc
summary(prop_rights$gdppc) ## just print
##
      Min. 1st Qu.
                    Median
                               Mean 3rd Qu.
                                               Max.
              1480
##
       450
                               5445
                                              27330
                      2835
                                       6968
prop_rights_sumstats <- summary(prop_rights$gdppc) ## store summary, "prop_rights_variables"
prop rights sumstats ## take a look at this summary
##
      Min. 1st Qu.
                    Median
                               Mean 3rd Qu.
##
       450
              1480
                      2835
                               5445
                                       6968
                                              27330
```

Or, we may be interested specifically in the **mean** value of GDP per capita for countries in the sample specifically, which we can get by typing mean(prop_rights\\$gdppc). As before, this command instructs R to look in the prop_rights dataframe and calculate the mean for the variable gdppc. We can store value resulting from this calculation, by using the <- symbol to store the results in a new object called meangdp, which can be called by typing meangdp into the console, or used in another command or context later on.

```
## Let's look specifically at mean GDP
mean(prop_rights$gdppc) ## just print

## [1] 5445.458

meangdp <- mean(prop_rights$gdppc) ## save as a variable, called "meangdp"
meangdp ## take a look at the value stored in of "meangdp"</pre>
```

[1] 5445.458

Similar commands exist for standard deviation (sd), variance (var), minimum (min), maximum (max), median (med), range (range) and quantile (quantile).

3.2 Regression

Linear regression is also very straightforward to apply in R. The lm command, which comes preloaded, can be used for univariate or multivariate regression:

$$Y_i = \alpha + \beta X_{1,i} + \gamma X_{2,i} + \dots + e_i$$

(Note: The above mathematical expression was written in a typesetting environment called LATEX, and can easily be included in RMarkdown. For examples, see the Script.) Examples of how to run a regression of log GDP per capita on protection against the property rights index (protection) and latitude are below:

```
## Univariate regression of log gdp per capita on property rights index.
lm(logGDP ~ protection, data=prop_rights) # display regression results
##
## Call:
## lm(formula = logGDP ~ protection, data = prop_rights)
##
## Coefficients:
## (Intercept)
                 protection
        4.6604
                     0.5221
##
## Multivariate, adding absolute latitude.
lm(logGDP ~ protection + lat_abst, data=prop_rights) # display regression results
##
## Call:
## lm(formula = logGDP ~ protection + lat_abst, data = prop_rights)
## Coefficients:
## (Intercept)
                                 lat abst
                 protection
        4.7281
                     0.4679
                                   1.5769
##
## Storing results
reg1 <- lm(logGDP ~ protection, data=prop_rights) # save results, assign to "req1"
reg2 <- lm(logGDP ~ protection + lat_abst, data=prop_rights) # save results, assign to "reg2"
In the last example, we store the univariate regression results in the object reg1 and the multivariate regression
results in the object reg2. If you try this in R, you'll notice the results don't display on the screen when you
store the results, but you can access them by naming the object.
## Now access regression results for the multivariate regression
summary(reg2) # access complete regression results
##
## Call:
## lm(formula = logGDP ~ protection + lat_abst, data = prop_rights)
##
## Residuals:
##
       Min
                1Q Median
                                 3Q
                                        Max
## -1.6845 -0.4233 0.1408 0.4584 1.1858
##
## Coefficients:
##
               Estimate Std. Error t value Pr(>|t|)
## (Intercept) 4.72808
                           0.39732 11.900 < 2e-16 ***
## protection
               0.46789
                            0.06416
                                     7.292 7.29e-10 ***
                1.57688
                           0.71031
                                      2.220 0.0301 *
## lat_abst
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.6917 on 61 degrees of freedom
## Multiple R-squared: 0.5745, Adjusted R-squared: 0.5605
## F-statistic: 41.18 on 2 and 61 DF, p-value: 4.805e-12
```

3.3 Tables

There are many packages in R that let one create very nice tables. We recommend **stargazer**. Take the two regressions we ran, stored as **reg1** and **reg2**, and say we want to export both to a table to be used in word or LaTeX.

To include the final table in the knitted pdf created by RMarkdown, you can use the following template. (Note, we want to use results='asis' and header=FALSE, in order for RMarkdown to compile the LaTeX table into a pretty format when knitting the pdf):

Table 1:	Property	Rights	and	Development

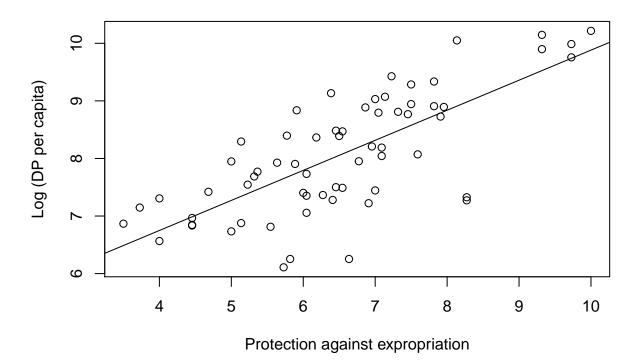
	Dependent	t variable:
	logG	FDP
	(1)	(2)
protection	0.522***	0.468***
	(0.061)	(0.064)
	t = 8.533	t = 7.292
lat_abst		1.577^{**}
		(0.710)
		t = 2.220
Constant	4.660***	4.728***
	(0.409)	(0.397)
	t = 11.408	t = 11.900
Observations	64	64
Note:	*p<0.1; **p<	0.05; ***p<0.01

The above command combines regression output saved in reg1 and reg2 into one table. Stargazer lets you specify title, alignment of coefficients inside columns and rows, which statistics to report (here, the variable names v, coefficients c, significance stars *, standard errors s and t-statistics t). Moreover, you can tell it to omit statistics such as the log-likelihood LL, the R^2 rsq, etc. You can also set the table placement in the final pdf (here, table.placement="H" means that the table will be placed directly after the code chunk). Stargazer has extensive documentation (try ??stargazer in R) and can also be used to make great summary statistics tables - try it out! Note that type="latex" produces table code that can be interpreted by LateX to produce nicely-formatted tables when you knit your RMarkdown file to PDF. Another alternative is to use the type="text" option to have a nicely formatted version of the table print to screen (for example, to later copy/paste into a word document).

3.4 Plotting

R also makes it easy to do basic plots. For example, one might want to create a scatterplot from two variables, and graph the regression line - or line of best fit - on the plot. This is easy to do with the plot command, which as a default plots the first variable listed on the horizontal axis, and the second on the vertical axis. Below we show a plot of the property rights index (protection against expropriation risk) and log of GDP per capita with titles and labels. We also use the abline command to add a regression line to the graph.

```
## A scatterplot with titles and line of best fit from univariate regression plot(prop_rights$protection,prop_rights$logGDP,
```



4 Resources for further study

- UCLA has a number of excellent resources to help you learn more about how to use R, available at: http://www.ats.ucla.edu/stat/r/.
- You can also find interactive lessons at Try R http://tryr.codeschool.com/ for practice writing code.
- We also recommend the text on the syllabus, Hanck, Christoph, Martin Arnold, Alexander Gerber and Martin Schmelzer. (2018). Introduction to Econometrics with R, https://www.econometrics-with-r.org/.
- You may also find https://www.r-bloggers.com/ to be a useful resource for specific questions.
- Finally, we include two useful resources that summarize useful R commands and common RMarkdown features.



Cheat Sheet

Getting Help

help.search('weighted mean') Get help of a particular function.

help(package = 'dplyr') Find help for a package.

Search the help files for a word or phrase.

str(iris)

Get a summary of an object's structure. Find the class an object belongs to. class(iris)

Selecting Vector Elements

By Position

Jsing Libraries

Download and install a package from CRAN. install.packages('dplyr')

Library(dplyr)

Load the package into the session, making all

dplyr::select

its functions available to use.

Use a particular function from a package.

data(iris)

Load a built-in dataset into the environment.

Working Directory

Find the current working directory (where inputs are found and outputs are sent). getwd()

than zero.

x[x < 0]

setwd('C://file/path')

Change the current working directory.

Use projects in RStudio to set the working directory to the folder you are working in.

Programming	While Loop	while (condition){	Do something }	Example	white (i < 5){ print(i)	i <- i + 1 }		Functions	<pre>function_name <- function(var){</pre>	Do something	return(new_variable)
Progra	For Loop	for (variable in sequence){	Do something }	Example	for (i in 1:4){ j <- i + 10	<pre>print(j) }</pre>		If Statements	<pre>if (condition) {</pre>	Do something } else {	Do something different
	rs	Join elements into a vector	An integer sequence	A complex sequence	Repeat a vector	Repeat elements of a vector	JS S		Return x reversed.	unique(x) Seemplone values	de values.
/ectors	Creating Vectors	2 4 6	23456	2.0 2.5 3.0	121212	111222	Vector Functions	rev(x)	Return x	unique(x)	רפת מו
>	Creat	c(2, 4, 6)	2:6	seq(2, 3, by=0.5) 2.0 2.5 3.0	rep(1:2, times=3)	rep(1:2, each=3)	Vecto	sort(x)	Return x sorted.	table(x)	See coulins of values.

square <- function(x) $\{$

Example

squared <- x*x

} else {

The fourth element.

x[4]

All but the fourth.

x[-4]

quared)

return(s	i a di
<u>~</u>	N. S. W.
	1
	1
	200
	п
	п
DI TIIC NO	п
N TIL	п
<u>~</u>	

Elements two to four.

x[2:4]

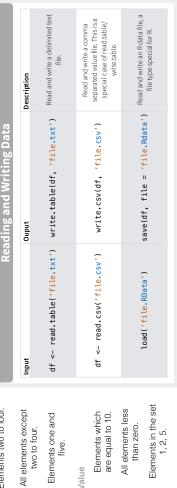
Elements one and

x[c(1, 5)]

x[-(2:4)]

By Value

x[x == 10]



Greater than or equal to	Less than or equal to	
a >= b	a <= b	
a > b Greater than	Less than	
a > b	a < b	
Are equal	Not equal	
q	q =	
a ===	<u></u> :	
Conditions		

Element with name 'apple'.

x['apple']

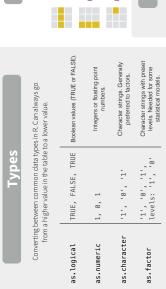
Named Vectors

x[x %in% c(1, 2, 5)]

Learn more at web page or vignette • package version • Updated:3/15

is.null(a) Is null is.na(a)

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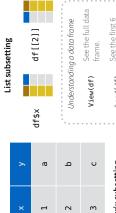
Maths Functions

	אומרווס ו מווכרוסווס		
log(x)	Natural log.	sum(x)	Sum.
exp(x)	Exponential.	mean(x)	Mean.
max(x)	Largest element.	median(x)	Median.
min(x)	Smallest element.	quantile(x)	Percentage quantiles.
round(x, n)	Round to n decimal places.	rank(x)	Rank of elements
signif(x, n)	Round to n significant figures.	var(x)	The variance.
cor(x, y)	Correlation.	sd(x)	The standard deviation.

Variable Assignment > a <- 'apple' [1] 'apple' ۸

□ %*% □ t(m) Transpose Matrix Multiplication solve(m, n) Find x in: $m^* x = n$ New list with only element l['y'] named y. $l <- list(x = 1:5, \ y = c('a', 'b'))$ A list is callection of elements which can be of different types. m <- matrix(x, nrow = 3, ncol = 3) Create a matrix from x. Element named 1\$x m[2, 3] - Select an element m[, 1] -Selecta column New list with only the first m[2,] -Selectarow 1[1] element. Second element of I. 1[[2]]

Data Frames	c– data. frame ($x = 1:3$, $y = c('a', 'b', 'c')$ A special case of a list where all elements are the same length.
Also see the dplyr library.	df <- data



See the full data frame.	See the first 6 rows.	cbind - Bind columns.	rbind - Bind rows.	†
View(df)	head(df)	nrow(df) Number of rows.	ncol(df) Number of columns.	dim(df) Number of columns and rows.
U	setting			
м	Matrix subsetting	df[, 2]	df[2,]	df[2, 2]

Strings	Also see the stringr library.
paste(x, y, sep = ')	Join multiple vectors together.
<pre>paste(x, collapse = ' ')</pre>	Join elements of a vector together.
grep(pattern, x)	Find regular expression matches in x.
gsub(pattern, replace, x)	Replace matches in x with a string.
toupper(x)	Convert to uppercase.
tolower(x)	Convert to lowercase.
nchar(x)	Number of characters in a string.

Statistics

cut(x, breaks = 4)
Turn a numeric vector into a factor but 'cutting' into

Turn a vector into a factor. Can set the levels of the factor and factor(x)

the order.

Factors

prop.test Testfora	almerence between proportions.	aov Analysis of variance.
<pre>t.test(x, y) Preform a t-test for</pre>	difference between means.	pairwise.t.test Preform a t-test for paired data.
<pre>lm(x ~ y, data=df) Linear model.</pre>	$glm(x \sim y, data=df)$ Generalised linear model.	summary Get more detailed information out a model.

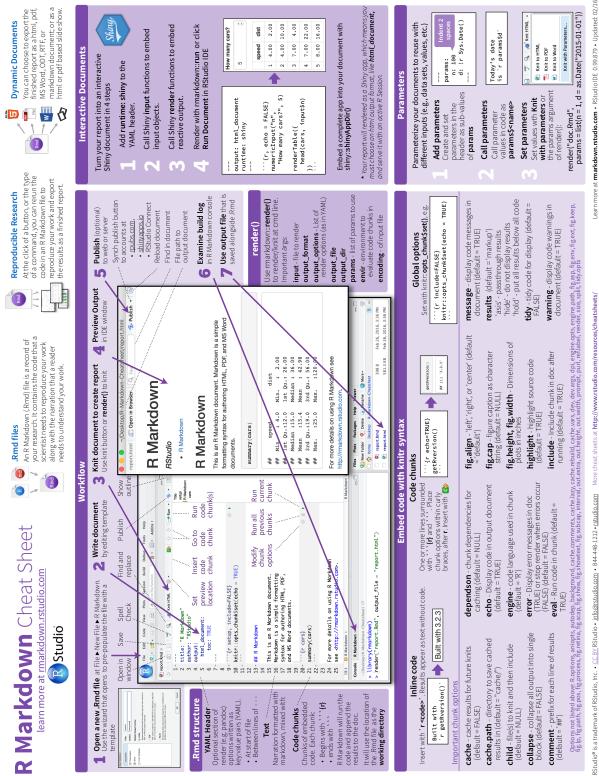
Distributions

Quantile	duorm	qpois	qbinom	qunif
Cumulative Distribution	pnorm	ppois	pbinom	punif
Density Function	dnorm	dpois	dbinom	dunif
Random Variates	rnorm	rpois	rbinom	runif
	Normal	Poison	Binomial	Uniform

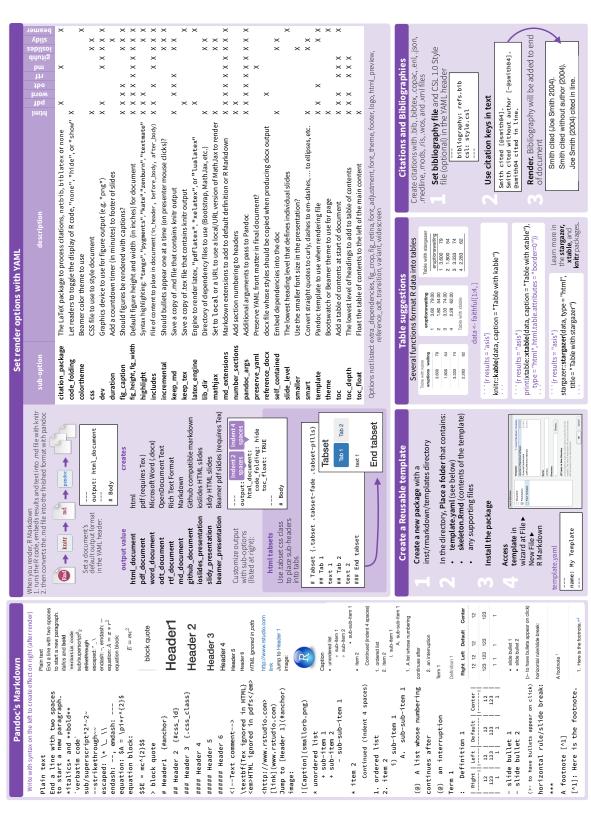
	5
2 library.	hist(x) Histogram of x.
he ggplot	
Also see the ggp	<pre>plot(x, y) Values of x against y.</pre>
lotting	plot(x) Values of x in order.

Histogram of X.	See the lubridate library.
Values of x against y.	See
ü	10
Values of x in order.	Dates

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