Data Science Workshop Session 1 SCRIPTS Data and Methodology Center

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Plan for this workshop series

This workshop series is geared toward learning basic data management in R. This includes tasks like manipulating variables, creating new variables, subsetting data, reshaping data, and merging. We will also cover some introductory regular expression applications. In this workshop series we will cover only basic visualization methods in R. Aspects like data analysis, web-scraping, or higher-level statistical programming are not covered.

Scheduled sessions:

- 1. **Introduction to R** (working directories, arithmetic, logical operators, basic indexing, data types, basic functions such as sum, mean, names, seq, rep, installing packages, reading and writing data, dealing with missing data, data frames, indexing on data frames, getting an overview of the data with multivariate numerical and graphical summaries).
- 2. Basic data management (dplyr,tidyr,stringr)

Getting started in R

R is a programming language for statistical computing and data visualization, that is a open source alternative to commercial statistical packages such as Stata or SPSS. R is maintained and developed by a vibrant community of programmers and statisticians and offers many user-written packages to extend basic functionality.

In this workshop, we will be using R together with the integrated development environment (IDE) **RStudio**. In addition to offering a 'cleaner' programming development than the basic R editor, RStudio offers a large number of added functionalities for integrating code into documents, built-in tools and web-development. To get started, please download the latest version of RStudio and R from this website:

https://www.rstudio.com/products/rstudio/download/

Getting Help

The key to learning R is: **Google!** This workshop will give you an overview over basic R functions, but to really learn R you will have to actively use it yourself, trouble shoot, ask questions, and google! The R mailing list and other help pages such as http://stackoverflow.com offer a rich archive of questions and answers by the R community. For example, if you google "recode data in r" you will find a variety of useful websites explaining how to do this on the first page of the search results. Also, don't be surprised if you find a variety of different ways to execute the same task.

RStudio also has a useful help menu. In addition, you can get information on any function or integrated data set in R through the console, for example:

```
?plot
```

In addition, there are a lot of free R comprehensive guides, such as Quick-R at http://www.statmethods.net or the R cookbook at http://www.cookbook-r.com.

Working Directories

R needs to know where to look for files if you want to read data and where to store files if you write data. This includes the code you will be typing in this session. The getwd() command returns the current working directory. We can change the working directory with setwd() (see below).

Think of your computer as a filing cabinet. During the course of this workshop, you will be writing a number of R scripts, that are essentially text files with commands for R. In order to execute these files, we need to tell R where to look for the list of commands we want to execute. Setting a working directory is analogous to telling R in which file in the filing cabinet we stored our document (code) and into which file in the filing cabinet to put new documents (such as graphs, new data frames, new code).

```
getwd() # Prints the current working directory
```

```
## [1] "/cloud/project"
setwd("/cloud/project")
```

Important for Windows users: In R, the backslash is an escape character (we will be talking more about

this in the session about stringr). Therefore, entering file paths is a little different in Windows than on a Mac. On a windows machine you would enter:

```
setwd("C:/Documents and Settings/Data")
```

OR

setwd("C:\\Documents and Settings\\Data")

Arithmetic in R

You can use R as a calculator!

	Operator	Example
Addition	+	2+4
Subtraction	_	2-4
Multiplication	*	2*4
Division	/	4/2
Exponentiation	^	2^4
Square Root	sqrt()	sqrt(144)
Absolute Value	abs()	abs(-4)

```
1±0
```

[1] 36

sqrt(144)

```
## [1] 12
```

Just like any regular calculator, you have to pay attention to the order of operations! Example:

```
6 * 8 - sqrt(7) + abs(-10) * (4/5)

## [1] 53.35425

6 * (8 - sqrt(7)) + abs(-10) * (4/5)

## [1] 40.12549
```

Logical operators

	Operator
Less than	<
Less than or equal to	<=
Greater than	>
Greater than or equal to	>=
Exactly equal to	==
Not equal to	! =
Not x	! x
x or y	хІу
x and y	х & у

Logical operators are incredibly helpful for any type of exploratory analysis, data cleaning and/or visualization task.

```
4 > 2

## [1] TRUE

4 <= 2

## [1] FALSE
```

Objects in R

Assigning values to objects

R stores information as an *object*. You can name objects whatever you like. Just remember to not use names that are reserved for build-in functions or functions in the packages you use, such as sum, mean, or abs. Most of the time, R will let you use these as names, but it leads to confusion in your code.

A few things to remember

- $\bullet\,$ Do not use special characters such as \$ or %. Common symbols that are used in variable names include . or .
- Remember that R is case sensitive.
- To assign values to objects, we use the assignment operator <-. Sometimes you will also see = as the assignment operator. This is a matter of preference and subject to debate among R programmers. Personally, I use <- to assign values to objects and = within functions.
- The # symbol is used for commenting and demarcation. Any code following # will not be executed.

Below, R stores the result of the calculation in an object named result. We can access the value by referring to the object name.

```
result <- 5/3
result # Implicitly printing the output

## [1] 1.666667
print(result) # Explicitly printing the output

## [1] 1.666667</pre>
```

If we assign a different value to the object, the value of the object will be changed.

```
result <- 5-3
result
```

[1] 2

Vectors

R can deal with a variety of data types, including vectors, scalars, matrices, data frames, factors, and lists. Today, we will focus on vectors.

A vector is one of the simplest type of data you can work with in R. "A vector or a one-dimensional array simply represents a collection of information stored in a specific order" (Imai 2017: 14). It is essentially a list of data of a single type (either numerical, character, or logical). To create a vector, we use the function c() ('concatenate'') to combine separate data points. The general format for creating a vector in R is as follows:name_of_vector <- c("what you want to put into the vector")' Suppose, we have data on the population in millions for the five most populous countries in 2016. The data come from the World Bank.

```
pop1 <- c(1379, 1324, 323, 261, 208)
pop1
```

```
## [1] 1379 1324 323 261 208
```

We can use the function c() to combine two vectors. Suppose we had data on 5 additional countries.

```
pop2 <- c(194, 187, 161, 142, 127)
pop <- c(pop1, pop2)
pop
```

```
## [1] 1379 1324 323 261 208 194 187 161 142 127
```

Variable types

There are four main variable types you should be familiar with:

- Numerical: Any number. Integer is a numerical variable without any decimals.
- Character: This is what Stata (and other programming languages such as Python) calls a string. We typically store any alphanumeric data that is not ordered as a character vector.
- Logical: A collection of TRUE and FALSE values.
- Factor: Think about it as an ordinal variable, i.e. an ordered categorical variable.

First, lets check which variable type our population data were stored in. The output below tells us that the object pop is of class numeric, and has the dimensions [1:10], that is 10 elements in one dimension.

```
str(pop)
```

```
## num [1:10] 1379 1324 323 261 208 ...
```

Suppose, we wanted to add information on the country names. We can enter these data in character format. To save time, we will only do this for the five most populous countries.

```
cname <- c("CHN", "IND", "USA", "IDN", "BRA")
str(cname)</pre>
```

```
## chr [1:5] "CHN" "IND" "USA" "IDN" "BRA"
```

Now, lets code a logical variable that shows whether the country is in Asia or not. Note that R recognizes both TRUE and T (and FALSE and F) as logical values.

```
asia <- c(TRUE, TRUE, F, T, F)
str(asia)
```

```
## logi [1:5] TRUE TRUE FALSE TRUE FALSE
```

Lastly, we define a factor variable for the regime type of a country in 2016. This variable can take on one of four values (based on data from the Economist Intelligence Unit): Full Democracy, Flawed Democracy, Hybrid Regimes, and Autocracy. Note that empirically, we don't have a "hybrid category" here. We could define an empty factor level, but we will skip this step here.

```
regime <- c("Autocracy", "FlawedDem", "FullDem", "FlawedDem", "FlawedDem")
regime <- as.factor(regime)
str(regime)</pre>
```

```
## Factor w/ 3 levels "Autocracy", "FlawedDem", ...: 1 2 3 2 2
```

Data types are important! R will not perform certain operations if you don't get the variable type right. The good news is that we can switch between data types. This can sometimes be tricky, especially when you are switching from a factor to a numerical type¹. We won't go into this too much here; just remember: Google is your friend!

Let's convert the factor variable regime into a character. Also, for practice, lets convert the asia variable to character and back to logical.

```
regime <- as.character(regime)
str(regime)

## chr [1:5] "Autocracy" "FlawedDem" "FullDem" "FlawedDem" "FlawedDem"
asia <- as.character(asia)
str(asia)

## chr [1:5] "TRUE" "TRUE" "FALSE" "TRUE" "FALSE"
asia <- as.logical(asia)
str(asia)</pre>
```

logi [1:5] TRUE TRUE FALSE TRUE FALSE

Exercise 1: Why won't R let us do the following?

no_good <- (a,b,c)

no_good_either <- c(one, two, three)</pre>

Exercise 2: What's the difference? (Bonus: What do you think is the class of the output vector?)

¹Sometimes you have to do a work around, like switching to a character first, and then converting the character to numeric. You can concatenate commands: myvar <- as.numeric(as.character(myvar)).

```
diff <-c(TRUE, "TRUE")
Exercise 3: What is the class of the following vector?</pre>
```

Vector operations

vec <- c("1", "2", "3")

You can do a variety of things like have R print out particular values or ranges of values in a vector, replace values, add additional values, etc. We will not get into all of these operations today, but be aware that (for all practical purposes) if you can think of a vector manipulation operation, R can probably do it.

We can do arithmatic operations on vectors! Let's use the vector of population counts we created earlier and double it.

```
pop1
## [1] 1379 1324 323 261 208
pop1_double <- pop1 * 2
pop1_double
## [1] 2758 2648 646 522 416

Exercise 4: What do you think this will do?
pop1 + pop2

Exercise 5: And this?
pop_c <- c(pop1, pop2)</pre>
```

Functions

There are a number of special functions that operate on vectors and allow us to compute measures of location and dispersion.

	Function
min()	Returns the minimum of the values or object.
max()	Returns the maximum of the values or object.
sum()	Returns the sum of the values or object.
<pre>length()</pre>	Returns the length of the values or object.
mean()	Returns the average of the values or object.
median()	Returns the median of the values or object.
<pre>var()</pre>	Returns the variance of the values or object.
sd()	Returns the variance of the values or object.

```
min(pop)

## [1] 127

max(pop)

## [1] 1379
```

```
mean(pop)
```

[1] 430.6

Exercise 6: Using functions in R, how else could we compute the mean population value?

[1] 430.6

Accessing elements of vectors

There are many ways to access elements that are stored in an object. Here, we will focus on a method called *indexing*, using square brackets as an operator.

Below, we use square brackets and the index 1 to access the first element of the top 5 population vector and the corresponding country name vector.

```
pop1[1]
```

```
## [1] 1379
```

cname[1]

[1] "CHN"

We can use indexing to access multiple elements of a vector. For example, below we use indexing to implicitly print the second and fifth elements of the population and the country name vectors, respectively.

```
pop[c(2,5)]
```

```
## [1] 1324 208
```

cname[c(2,5)]

```
## [1] "IND" "BRA"
```

We can assign the first element of the population vector to a new object called first.

```
first <- pop[1]
```

Below, we make a copy of the country name vector and delete the *last* element. Note, that we can use the length() function to achieve the highest level of *generalizability* in our code. Using length(), we do not need to know the index of the last element of out vector to drop the last element.

```
cname_copy <- cname
## Option 1: Dropping the 5th element
cname_copy[-5]</pre>
```

```
## [1] "CHN" "IND" "USA" "IDN"
```

```
## Option 2 (for generalizability): Getting the last element and dropping it.
length(cname_copy)
```

[1] 5

```
cname_copy[-length(cname_copy)]
```

```
## [1] "CHN" "IND" "USA" "IDN"
```

Indexing can be used to alter values in a vector. Suppose, we notice that we wrongly entered the fifth element of the regime type vector (or the regime type changed).

regime

```
## [1] "Autocracy" "FlawedDem" "FullDem" "FlawedDem" "FlawedDem"
regime[5] <- "FullDem"
regime</pre>
```

```
## [1] "Autocracy" "FlawedDem" "FullDem" "FlawedDem" "FullDem"
```

Exercise 7: We made even more mistakes when entering the data! We want to subtract 10 from the third and fifth element of the top 5 population vector. *How would you do it*?

More functions

The myriad of functions that are either built-in to base R or parts of user-written packages are the greatest stength of R. For most applications we encounter in our daily programming practice, R already has a function, or someone smart wrote one. Below, we introduce a few additional helpful functions from base R.

	Function
seq() rep() names() which()	Returns sequence from input1 to input2 by input3. Repeats input1 input2 number of times. Returns the names (labels) of objects. Returns the index of objects.

Let's create a vector of indices for our top 5 population data.

```
cindex <- seq(from = 1, to = length(pop1), by = 1)
cindex</pre>
```

```
## [1] 1 2 3 4 5
```

We can use the rep() function to repeat data.

```
rep(30, 5)
```

```
## [1] 30 30 30 30 30
```

Exercise 8 Use rep() and seq to print the following data.

```
## [1] 1.0 1.5 2.0 2.5 3.0 1.0 1.5 2.0 2.5 3.0
```

Suppose, we wanted to record whether we had completed the data collection process for the top 10 most populous countries. First, suppose we completed the process on every second country.

```
completed <- rep(c("yes","no"), 5)
completed</pre>
```

```
## [1] "yes" "no" "yes" "no" "yes" "no" "yes" "no" "yes" "no"
```

Now suppose that we have completed the data collection process for the first 5 countries, but not the latter 5 countries (we don't have their names, location, or regime type yet).

```
completed2 <- rep(c("yes","no"), each = 5)
completed2</pre>
```

```
## [1] "yes" "yes" "yes" "yes" "no" "no" "no" "no" "no" "no"
```

We can give our data informative labels. Let's use the country names vector as labels for our top 5 population vector.

```
names (pop1)
## NULL
cname
## [1] "CHN" "IND" "USA" "IDN" "BRA"
names(pop1) <- cname</pre>
names (pop1)
## [1] "CHN" "IND" "USA" "IDN" "BRA"
pop1
##
    CHN
         IND
               USA
                     IDN
                          BRA
## 1379 1324
               323
                     261
                          208
```

We can use labels to access data using indexing and logical operators. Suppose, we wanted to access the population count for Brazil in our top 5 population data.

```
pop1[names(pop1) == "BRA"]
```

```
## BRA
## 208
```

Exercise 9 Access all top 5 population ratings that are greater or equal than the mean value of population ratings.

```
## [1] 699
## CHN IND
## 1379 1324
```

Exercise 10 Access all top 5 population ratings that are less than the population of the most populous country, but not the US.

```
## IND IDN BRA
## 1324 261 208
```

[1] 4

Operating on multiple vectors simultaneously

We did not work with data frames yet, but remember that our data input is ordered. The first element of the pop1 vector corresponds with the first element of the cname, regime, and asia vectors. We can use this to run more sophisticated queries on our data.

Suppose, we wanted to know the regime type of Indonesia. Given that our vectors are ordered, we can use indexing to extract the data. First, lets see what happens if we run a simple logical query.

```
cname == "IDN"

## [1] FALSE FALSE TRUE FALSE
regime[cname == "IDN"]

## [1] "FlawedDem"

We can also use the which() function that returns the index of the vector element.
which(cname == "IDN")
```

```
regime[which(cname == "IDN")]
```

```
## [1] "FlawedDem"
```

Exercise 11 Print out the population count for all Asian countries within the top 5 most populous countries that are not autocracies.

```
## IND IDN
## 1324 261
```

Working with data frames in R

Importing data

Most data formats we commonly use are not native to R and need to be imported. Luckily, there is a variety of packages available to import just about any non-native format. One of the essential libraries is called foreign and includes functions to import .csv, .dta (Stata), .dat, ..sav (SPSS), etc. The foreign package is pre-installed. However, we need to tell R that we want to use functions from this package by calling it into the working environment using the library() command.

```
library(foreign)
```

In this example, we will use data from the Armed Conflict Location & Event Data Project (ACLED), which offers real-time data and analysis on political violence and protests around the world. This dataset includes riots and protests from January 2010 to December 2019.

When reading the data, remember that you need to either specify the complete file name or change your working directory using setwd() to the folder in which you saved the data.

```
mydata <- read.csv("ACLED_countries.csv")
class(mydata)</pre>
```

```
## [1] "data.frame"
```

Dimensions of a data frame

Let's find out what this data looks like. First, use the str() function to explore the variable names and which data class they are stored in. Note: int stands for integer and is a special case of the class numeric.

```
str(mydata)
```

If we are only interested in what the variables are called, we can use the names() function.

names (mydata)

As we saw in the last session, we can alter the names of vectors by using the names() function and indexing. Because data frames are essentially just combinations of vectors, we can do the same for variable names inside data frames. Suppose we want to change the variable nconflicts.

We can use the summary() function to get a first look at the data.

```
summary(mydata)
```

```
##
           country
                                            region
                                                       nconflict
##
    Tunisia
               : 2
                     Europe
                                               :15
                                                     Min.
                                                            :
                                                                  1.0
    Afghanistan: 1
                     Middle East
                                                     1st Qu.:
                                                               315.5
##
                                               :15
  Albania
                                               :15
##
               : 1
                     Western Africa
                                                     Median: 1250.0
  Algeria
                     Eastern Africa
                                               :13
                                                            : 6216.3
               : 1
                                                     Mean
  Angola
               : 1
                     Caucasus and Central Asia: 8
                                                     3rd Qu.: 5993.5
##
##
    Armenia
               : 1
                     Middle Africa
                                               : 8
                                                     Max.
                                                             :70734.0
##
   (Other)
               :96
                     (Other)
                                               :29
  nconflict_no_fatalities
                               fatalities
                                          0.0
##
   Min.
          :
                1
                            Min.
                                   :
##
   1st Qu.:
             289
                            1st Qu.:
                                         17.5
##
  Median: 1037
                            Median:
                                        236.0
  Mean
          : 4667
                            Mean
                                       9543.3
    3rd Qu.: 4536
                            3rd Qu.:
##
                                       7020.0
##
    Max.
           :63665
                            Max.
                                    :119973.0
##
```

A data frame has two dimensions: rows and columns.

```
nrow(mydata) # Number of rows

## [1] 103
ncol(mydata) # Number of columns

## [1] 5
dim(mydata) # Rows first then columns.

## [1] 103 5
```

Accessing elements of a data frame

As a rule, whenever we use two-dimensional indexing in R, the order is: [row,column]. To access the first row of the data frame, we specify the row we want to see and leave the column slot following the comma empty.

```
mydata[1,]

## country region nconflict nconflict_no_fatalities fatalities
## 1 Afghanistan Southern Asia 40765 23946 119973
```

We can use the concatenate function c() to access multiple rows (or columns) at once. Below we print out the first and second row of the dataframe.

mydata[c(1,2),]

```
## country region nconflict nconflict_no_fatalities fatalities
## 1 Afghanistan Southern Asia 40765 23946 119973
## 2 Albania Europe 582 581 1
```

We can also access a range of rows by separating the minimum and maximum value with a :. Below we print out the first five rows of the dataframe.

mydata[1:5,]

##		country			region	${\tt nconflict}$	<pre>nconflict_no_fatalities</pre>
##	1	${\tt Afghanistan}$		Souther	n Asia	40765	23946
##	2	Albania			Europe	582	581
##	3	Algeria		Northern	${\tt Africa}$	7362	5321
##	4	Angola		Middle	${\tt Africa}$	1108	728
##	5	Armenia	${\tt Caucasus}$	and Centra	al Asia	3118	3110
##		fatalities					
##	1	119973					
##	2	1					
##	3	8451					
##	4	13788					
##	5	8					

If we try to access a data point that is out of bounds, R returns the value NULL.

mydata[3,3]

[1] 7362

Exercise 1 Access the element of the data frame mydata that is stored in row 1, column 1.

- ## [1] Afghanistan
- ## 102 Levels: Afghanistan Albania Algeria Angola Armenia Azerbaijan ... Zimbabwe

Exercise 2 Access the element of the data frame mydata that is stored in column 2, row 3.

- ## [1] Northern Africa
- ## 10 Levels: Caucasus and Central Asia Eastern Africa Europe ... Western Africa

The \$ operator

The \$ operator in R is used to specify a variable within a data frame. This is an alternative to indexing.

mydata\$nconflict

```
##
     [1] 40765
                   582
                        7362
                                1108
                                      3118 12627
                                                    1861 16802
                                                                   293
                                                                          215
                                                                                 459
                                                                                         52
    [13]
                  2105
                        7760
                                1919
                                      2621
##
            765
                                             4625
                                                     912
                                                            411
                                                                   312 14519
                                                                                 112 10260
##
    [25]
             50
                   191
                          190
                                5122
                                        215
                                              197
                                                    1446
                                                            858
                                                                  1312
                                                                         1121
                                                                                 151 67561
                  5708 22354
##
    [37]
           3371
                                1580
                                      1838
                                              491
                                                     374
                                                           6427
                                                                   289
                                                                           39
                                                                                 319
                                                                                         41
##
    [49]
           2956
                    83
                         1294
                                8857
                                      1365
                                              512
                                                     384
                                                           3193
                                                                   598
                                                                          444
                                                                                 213
                                                                                      2177
           1250
                  9070
##
    [61]
                          682
                                5143
                                        971
                                            15011
                                                     275
                                                               9 54496
                                                                         6568
                                                                                9690
                                                                                          5
                         3734
                                 370
                                      6096
                                                    1275
                                                                                          2
##
    [73]
            243
                  1134
                                              963
                                                           1079 29667 11667
                                                                                6227
           3855 12528 70734
##
    [85]
                                  68
                                        782
                                             7420
                                                     321
                                                           5756
                                                                     1 10334
                                                                                  21
                                                                                      4866
    [97] 30131
                          102
                                 503 45298
                                             1148
                                                    5891
```

Exercise 3: How would you access all elements of the variable country using indexing rather than the \$ operator?

```
##
     [1] Afghanistan
                                      Albania
##
     [3] Algeria
                                      Angola
     [5] Armenia
##
                                      Azerbaijan
##
     [7] Bahrain
                                      Bangladesh
##
     [9] Belarus
                                      Benin
##
   [11] Bosnia and Herzegovina
                                      Botswana
   [13] Bulgaria
                                      Burkina Faso
   [15] Burundi
                                      Cambodia
##
   [17] Cameroon
                                      Central African Republic
##
  [19] Chad
                                      Croatia
  [21] Cyprus
                                      Democratic Republic of Congo
##
  [23] Djibouti
                                      Egypt
  [25] Equatorial Guinea
                                      Eritrea
## [27] eSwatini
                                      Ethiopia
## [29] Gabon
                                      Gambia
##
   [31] Georgia
                                      Ghana
##
  [33] Greece
                                      Guinea
##
  [35] Guinea-Bissau
                                      India
##
  [37] Indonesia
                                      Iran
## [39] Iraq
                                      Israel
##
  [41] Ivory Coast
                                      Jordan
  [43] Kazakhstan
                                      Kenya
## [45] Kosovo
                                      Kuwait
## [47] Kyrgyzstan
                                      Laos
## [49] Lebanon
                                      Lesotho
## [51] Liberia
                                      Libya
## [53] Madagascar
                                      Malawi
## [55] Malaysia
                                      Mali
##
  [57] Mauritania
                                      Moldova
## [59] Montenegro
                                      Morocco
## [61] Mozambique
                                      Myanmar
## [63] Namibia
                                      Nepal
##
  [65] Niger
                                      Nigeria
##
  [67] North Macedonia
                                      Oman
##
   [69] Pakistan
                                      Palestine
##
  [71] Philippines
                                      Qatar
  [73] Republic of Congo
##
                                      Romania
## [75] Russia
                                      Rwanda
   [77] Saudi Arabia
##
                                      Senegal
## [79] Serbia
                                      Sierra Leone
## [81] Somalia
                                      South Africa
## [83] South Sudan
                                      Spain
## [85] Sri Lanka
                                      Sudan
## [87] Syria
                                      Tajikistan
## [89] Tanzania
                                      Thailand
                                      Tunisia
## [91] Togo
   [93] Tunisia
                                      Turkey
##
  [95] Turkmenistan
                                      Uganda
                                      United Arab Emirates
  [97] Ukraine
## [99] Uzbekistan
                                      Vietnam
## [101] Yemen
                                      Zambia
## [103] Zimbabwe
## 102 Levels: Afghanistan Albania Algeria Angola Armenia Azerbaijan ... Zimbabwe
```

```
##
     [1] Afghanistan
                                      Albania
##
     [3] Algeria
                                      Angola
     [5] Armenia
##
                                      Azerbaijan
##
     [7] Bahrain
                                      Bangladesh
##
     [9] Belarus
                                      Benin
##
   [11] Bosnia and Herzegovina
                                      Botswana
   [13] Bulgaria
                                      Burkina Faso
  [15] Burundi
                                      Cambodia
##
   [17] Cameroon
                                      Central African Republic
##
  [19] Chad
                                      Croatia
  [21] Cyprus
                                      Democratic Republic of Congo
##
  [23] Djibouti
                                      Egypt
  [25] Equatorial Guinea
                                      Eritrea
## [27] eSwatini
                                      Ethiopia
## [29] Gabon
                                      Gambia
##
   [31] Georgia
                                      Ghana
##
  [33] Greece
                                      Guinea
## [35] Guinea-Bissau
                                      India
## [37] Indonesia
                                      Iran
## [39] Iraq
                                      Israel
##
  [41] Ivory Coast
                                      Jordan
## [43] Kazakhstan
                                      Kenya
## [45] Kosovo
                                      Kuwait
## [47] Kyrgyzstan
                                      Laos
## [49] Lebanon
                                      Lesotho
## [51] Liberia
                                      Libya
## [53] Madagascar
                                      Malawi
## [55] Malaysia
                                      Mali
## [57] Mauritania
                                      Moldova
## [59] Montenegro
                                      Morocco
## [61] Mozambique
                                      Myanmar
## [63] Namibia
                                      Nepal
##
  [65] Niger
                                      Nigeria
##
  [67] North Macedonia
                                      Oman
##
   [69] Pakistan
                                      Palestine
##
  [71] Philippines
                                      Qatar
  [73] Republic of Congo
##
                                      Romania
## [75] Russia
                                      Rwanda
   [77] Saudi Arabia
##
                                      Senegal
## [79] Serbia
                                      Sierra Leone
## [81] Somalia
                                      South Africa
## [83] South Sudan
                                      Spain
## [85] Sri Lanka
                                      Sudan
## [87] Syria
                                      Tajikistan
## [89] Tanzania
                                      Thailand
                                      Tunisia
## [91] Togo
   [93] Tunisia
                                      Turkey
##
  [95] Turkmenistan
                                      Uganda
                                      United Arab Emirates
  [97] Ukraine
## [99] Uzbekistan
                                      Vietnam
## [101] Yemen
                                      Zambia
## [103] Zimbabwe
```

Exercise 4: Print out every second element from the variable nconflict using indexing methods and the sequence function seq().

```
[1] 40765
                7362
                       3118
                             1861
                                     293
                                            459
                                                   765
                                                        7760
                                                               2621
                                                                       912
                                                                              312
                                                                                    112
                                                                              289
## [13]
            50
                 190
                        215
                              1446
                                     1312
                                            151
                                                  3371 22354
                                                               1838
                                                                       374
                                                                                    319
   [25]
          2956
                1294
                       1365
                               384
                                     598
                                            213
                                                  1250
                                                                971
                                                                       275 54496
                                                                                   9690
                                                          682
## [37]
           243
                3734
                       6096
                              1275 29667
                                           6227
                                                  3855 70734
                                                                782
                                                                       321
                                                                                      21
                 102 45298
## [49] 30131
                              5891
```

Exercise 5: What are two ways to find the mean value of the variable nconflict using indexing (i.e. not using the \$ operator)?

```
## Warning in mean.default(mydata[, 2]): argument is not numeric or logical:
## returning NA
## [1] NA
## [1] 6216.272
```

Exercise 6: How would you find the maximum value for number of conflicts using the \$ operator?

[1] 70734

Exercise 7: Print the country that corresponds to the maximum world population value using the \$ operator and indexing!

```
## [1] Syria
## 102 Levels: Afghanistan Albania Algeria Angola Armenia Azerbaijan ... Zimbabwe
```

NAs in R

NA is how R denotes missing values. For certain functions, NAs cause problems.

```
vec <- c(4, 1, 2, NA, 3)
mean(vec) #Result is NA!

## [1] NA

sum(vec) #Result is NA!

## [1] NA

We can tell R to remove the NA and execute the function on the remainder of the data.

mean(vec, na.rm = T)

## [1] 2.5</pre>
```

```
Adding observations
```

sum(vec, na.rm = T)

[1] 10

First, lets add another observation to the data. Suppose we wanted to add an observation for Germany, which will be a missing value. We can use the same operations we used for vectors to add data. Here, we will will use the rbind() function to do so. rbind() stands for "row bind." Save the output in a new data frame!

```
obs <- c("Germany", NA, NA)
mydata_new <- rbind(mydata, obs)
```

```
## Warning in `[<-.factor`(`*tmp*`, ri, value = "Germany"): invalid factor level,
## NA generated
dim(mydata_new)
## [1] 104 5</pre>
```

We can also create new variables that use information from the existing data. If we know the number of conflicts without fatalities by country, we can calculate the number of conflicts with fatalities to generate the variable nconflict_fatalities. By using the \$ operator, we can directly assign the new variable to the data frame mydata new.

```
mydata$nconflict_fatalities <- mydata$nconflict - mydata$nconflict_no_fatalities head(mydata, 3) #prints out the first 3 rows of the data frame
```

```
##
         country
                           region nconflict nconflict_no_fatalities fatalities
## 1 Afghanistan
                    Southern Asia
                                       40765
                                                                 23946
                                                                            119973
## 2
         Albania
                            Europe
                                         582
                                                                   581
                                                                                 1
## 3
                                        7362
                                                                  5321
                                                                              8451
         Algeria Northern Africa
##
     nconflict_fatalities
## 1
                     16819
## 2
## 3
                      2041
```

Saving data

Suppose we wanted to save this newly created data frame. We have multiple options to do so. If we wanted to save it as a native .RData format, we would run the following command.

```
# Make sure you specified the right working directory!
# save(mydata, file = "mydata_new.RData")
```

Most of the time, however, we would want to save our data in formats that can be read by other programs as well. .csv is an obvious choice. After saving the new file, check the folder that is your working directory to see whether the newly saved file shows up there!

```
# write.csv(mydata_new, file = "mydata_new.csv")
```

(Very basic) data visualization

Today, we will be covering some basics of data visualization in R using the native plotting functions. For more advanced data visualization functions, see the ggplot2 package and the related material for a two-session introductory workshop on data visualization https://github.com/thereseanders/Workshop-Intro-to-ggplot2.

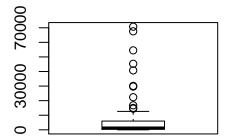
Basic graphical summaries of data

Type	Operator
Histogram	hist()
Stem and leaf plot	stem()
Boxplot	<pre>boxplot()</pre>
Kernel density plot	<pre>plot(density())</pre>
Basic scatterplot	plot()

Boxplot of population density

We can get an overview of the number of conflicts per country using the boxplot() function.

boxplot(mydata\$nconflict)



Suppose we wanted to know whether there are more conflicts with fatalities on average for countries with a higher overall number of conflicts. Let's first look at the distribution of number of conflicts.

summary(mydata\$nconflict)

```
## Min. 1st Qu. Median Mean 3rd Qu. Max.
## 1.0 315.5 1250.0 6216.3 5993.5 70734.0
```

Create a new dummy variable that codes whether a state has a relatively high number of conflicts (greater than the median).

```
median(mydata$fatalities)
```

```
## [1] 236
```

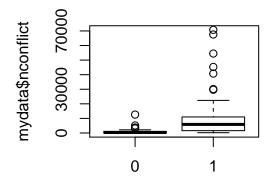
mydata\$fatalities_high <- ifelse(mydata\$fatalities > median(mydata\$fatalities), 1, 0)
head(mydata\$fatalities_high)

```
## [1] 1 0 1 1 0 0
```

table(mydata\$fatalities_high) # We split the observations exactly in half.

We can display two indicators in the same boxplot. We can use this feature to answer the question whether larger states are on average more or less dense than smaller states.

boxplot(mydata\$nconflict ~ mydata\$fatalities_high)

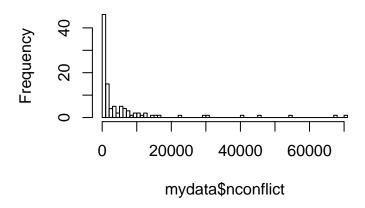


mydata\$fatalities_high

Histogram of number of conflicts

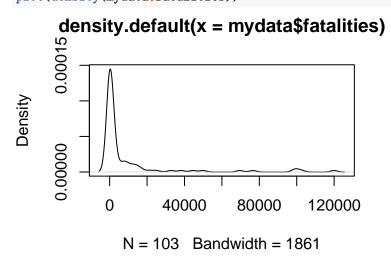
hist(mydata\$nconflict,breaks=100)

Histogram of mydata\$nconflict



Density plot of fatalities

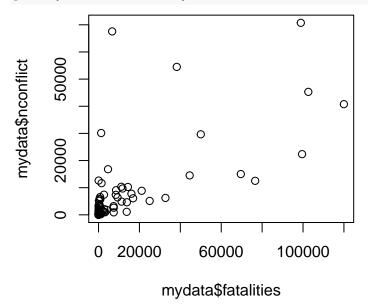
plot(density(mydata\$fatalities))



Basic scatter plots

How does population size vary with area? Are the correlated at all?

plot(mydata\$fatalities,mydata\$nconflict)

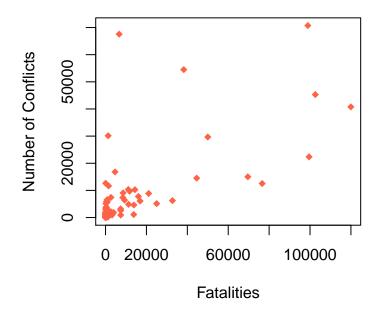


Basic graphic options

Aside from arguably not being very informative, the plot above is not very pretty. Lets give it titles, use color and shapes!

```
plot(mydata$fatalities,mydata$nconflict,
    main = "ACLED (2010-2019)", #Adding a main title.
    xlab = "Fatalities", #Adding a x-axis title.
    ylab = "Number of Conflicts", #Adding a y-axis title.
    col = "tomato", #Changing the color of the data points.
    pch = 18) #Changing the shape
```

ACLED (2010-2019)



Yeah, ok. Its not much prettier (especially the labeling on the axes), but you get the point...

A few additional notes on graphical options:

- R can display any color in the RBG or HEX system. However it also has a ton of colors that you can just refer to by name, see http://www.stat.columbia.edu/~tzheng/files/Rcolor.pdf.
- Same with the shapes and line types, see http://www.cookbook-r.com/Graphs/Shapes_and_line_types/.
- R colors in all their glory: http://www.stat.columbia.edu/~tzheng/files/Rcolor.pdf.

Sources

Economist Intelligence Unit (2017): Democracy Index. https://infographics.economist.com/2017/DemocracyIndex/.

Imai, Kosuke (2017): Quantitative Social Science. An Introduction. Princeton and Oxford: Princeton University Press.

Raleigh, Clionadh, Andrew Linke, Håvard Hegre and Joakim Karlsen. 2010. Introducing ACLED – Armed Conflict Location and Event Data. *Journal of Peace Research* 47(5), 651-660.

World Bank (2017): Population, total. https://data.worldbank.org/indicator/sp.pop.totl?end=2016&start=2015.