Introduction to R

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Insert Presentation Date

Outline

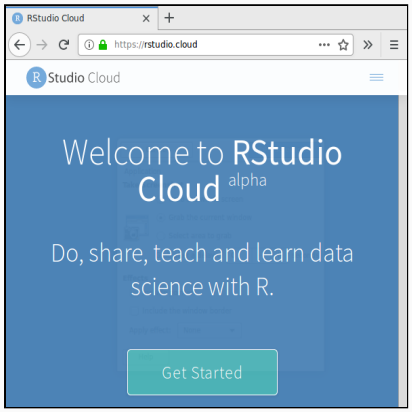
- What is `R`
- Setting up `RStudio` in your computer
- Examples
- Basics of `R`
- Some `R` Coding
- Importing a dataset in `RStudio`

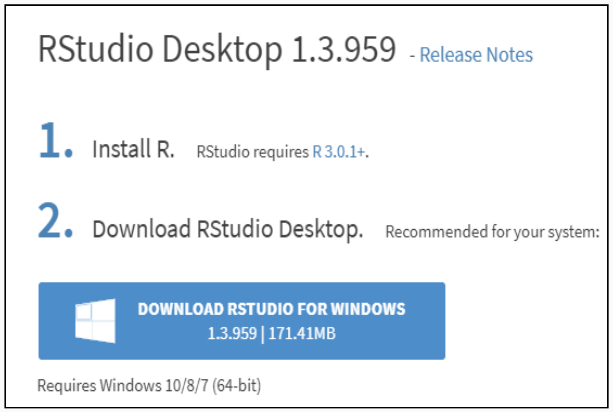
What is R?

- R is a programming language environment for statistical computing
- Widely used among statisticians, data scientists and is currently ranked 9th in the TIOBE index
- R includes a wide variety of statistical and graphical techniques
- Many free resources to learn `R`
- **It's free!**

Setting up R

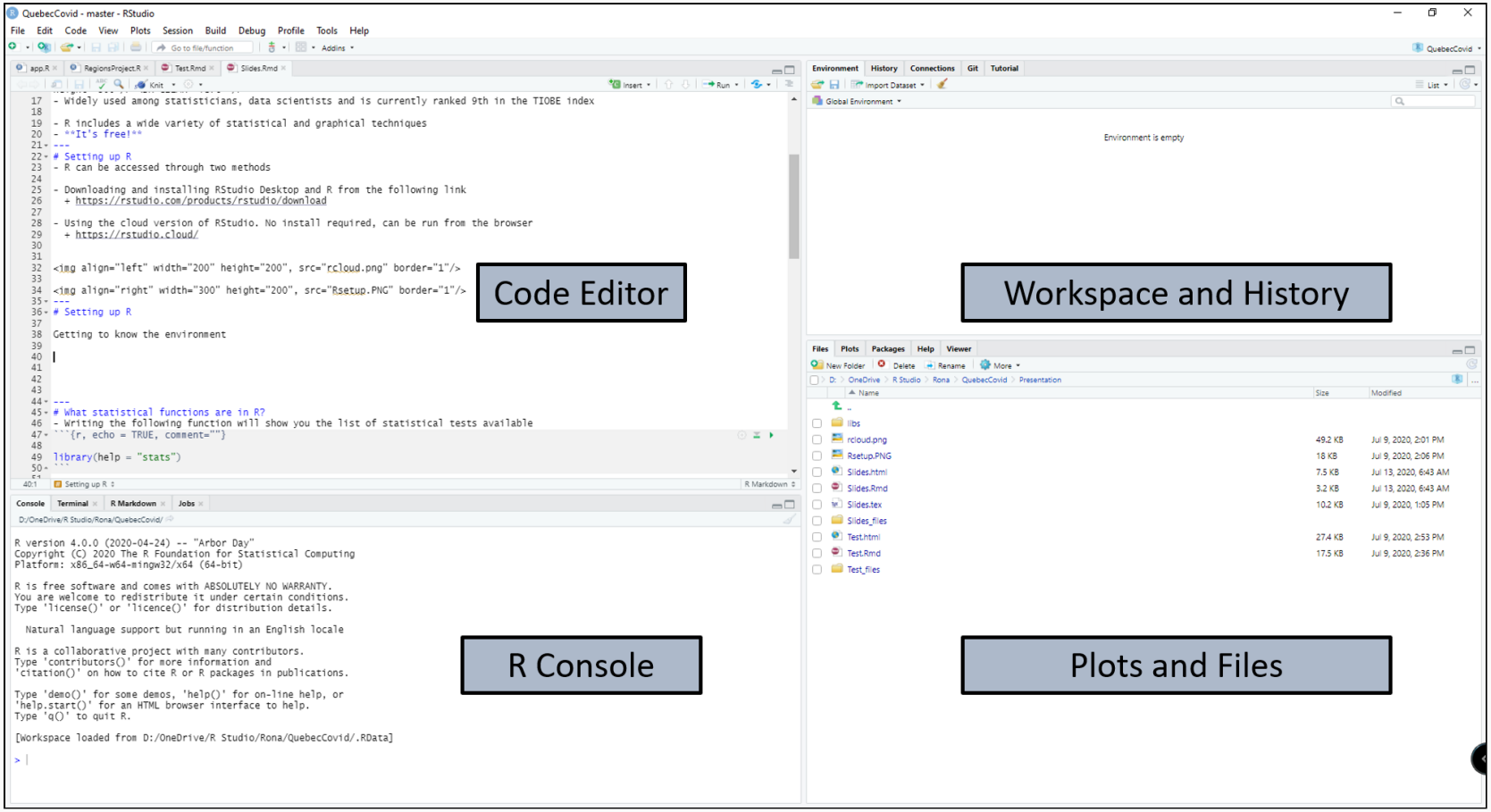
- R can be accessed through two methods
- Downloading and installing RStudio Desktop and R from the following link
  - https://rstudio.com/products/rstudio/download
- Using the cloud version of RStudio. No install required, can be run from the browser
  - https://rstudio.cloud/





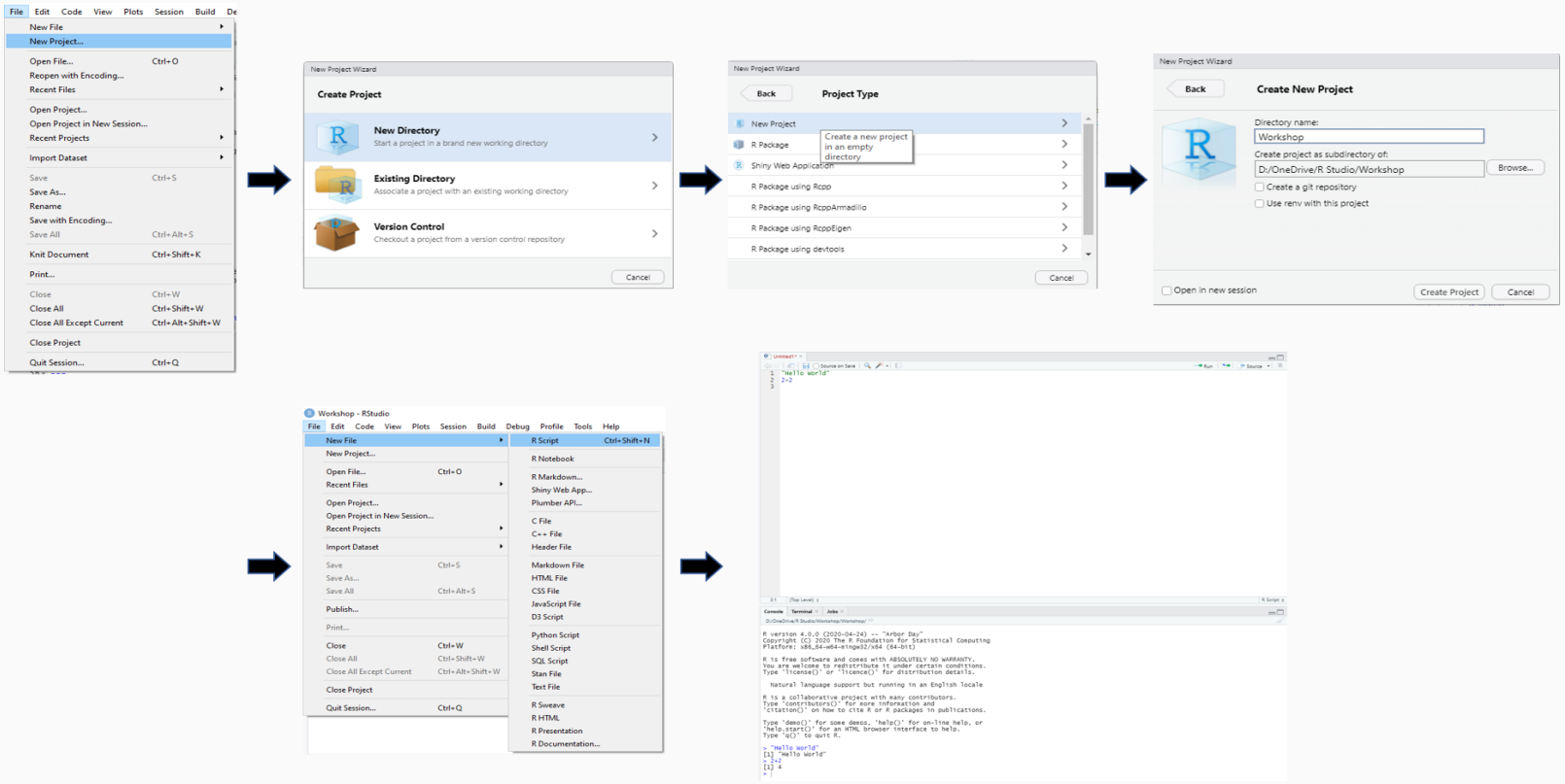
Working to RStudio

- Getting to know the environment



Working with RStudio

- Before starting any kind of work it is **very important** to have your environment set
- An environment is the space where all your variables, formulas, plots and everything related to your project will be saved



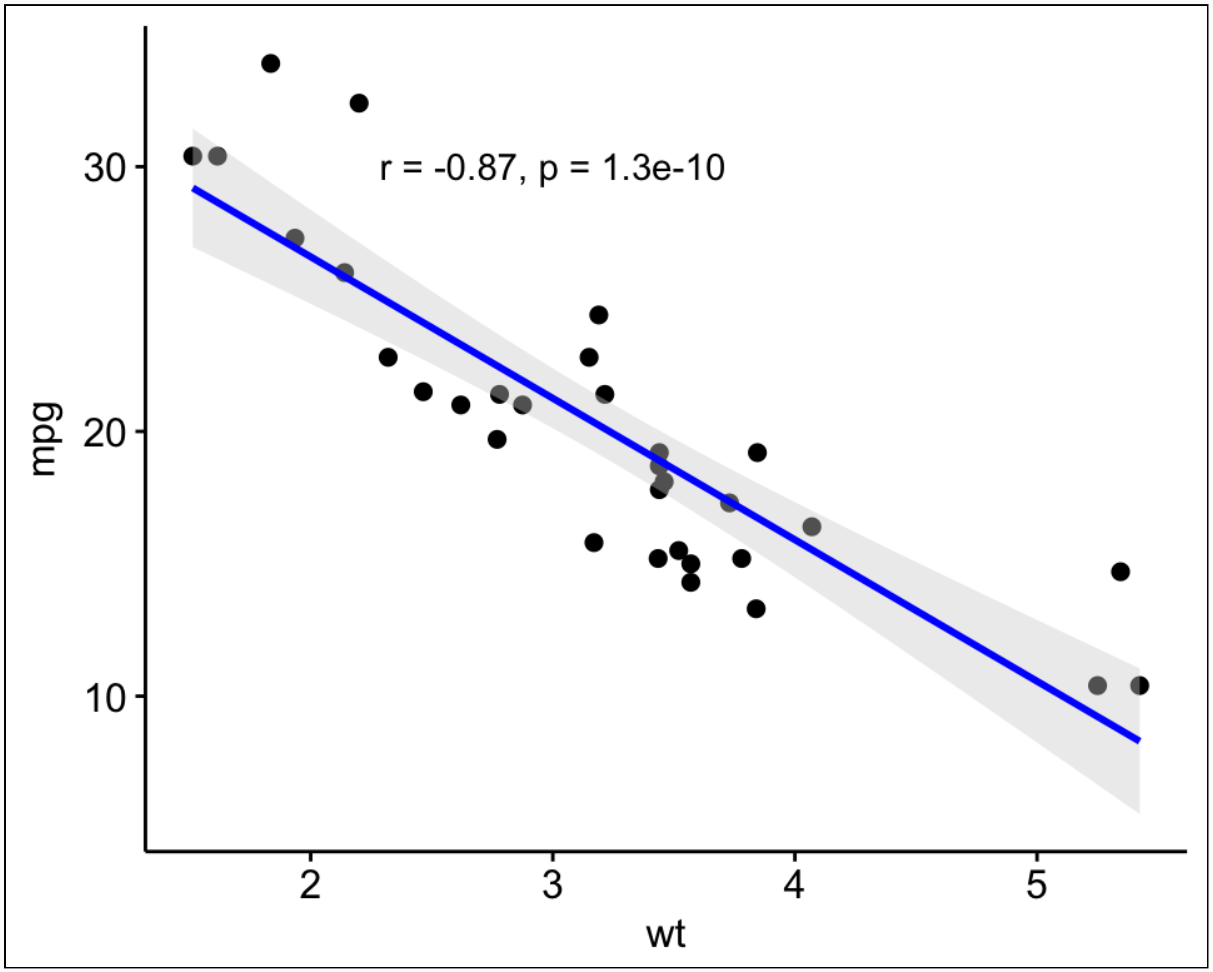
What statistical functions are in R?

- Writing the following function will show you the list of statistical tests available

```
library(help = "stats")
```

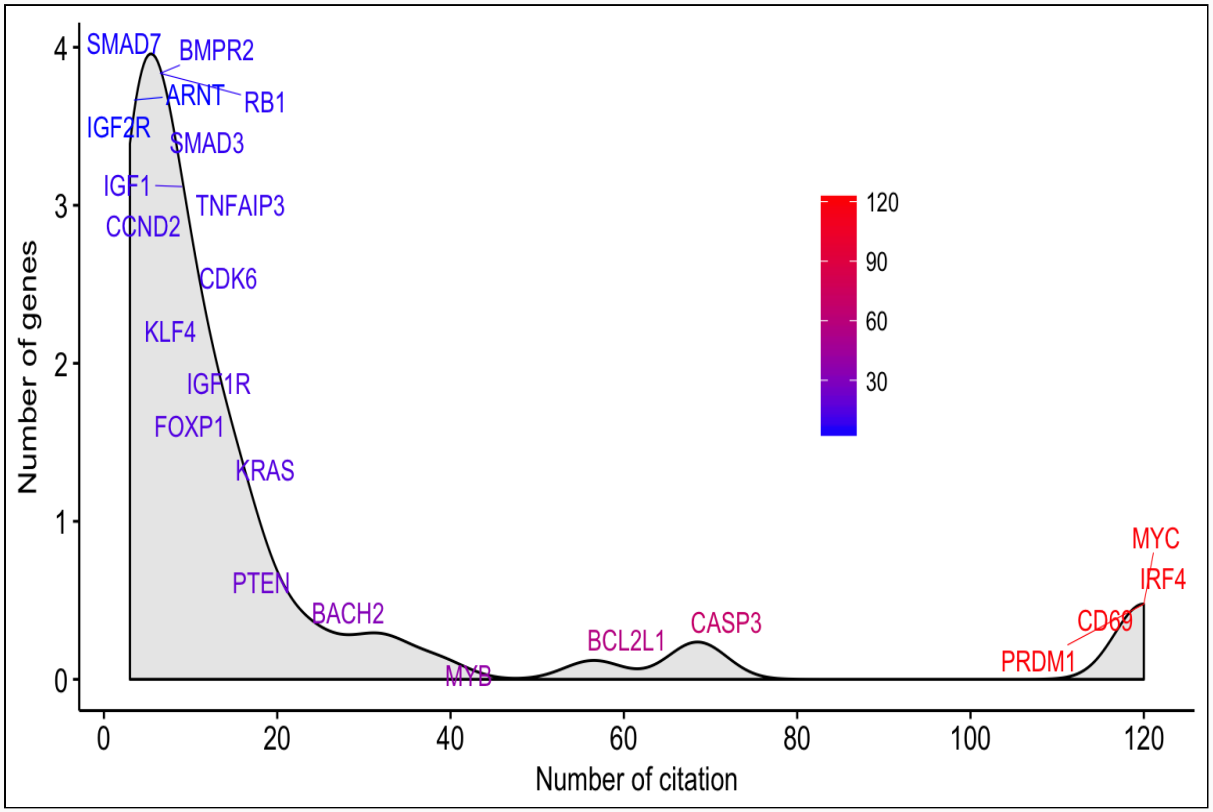
Examples of R Graphics

Scatterplot with statistical notation



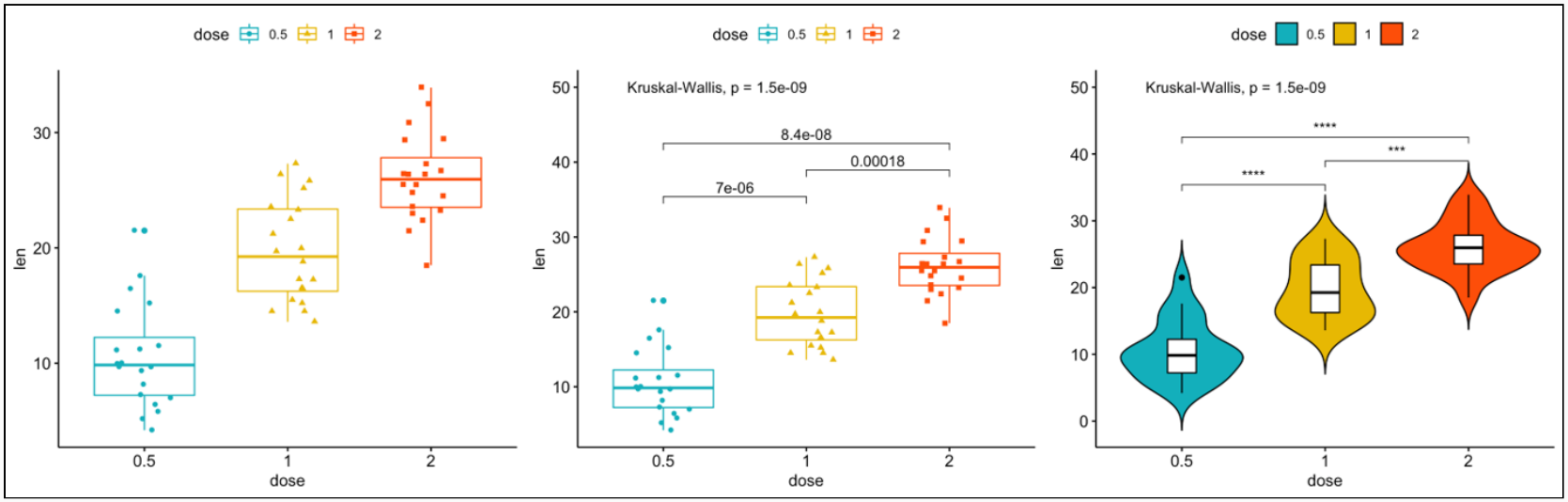
Examples of R Graphics

Histograms



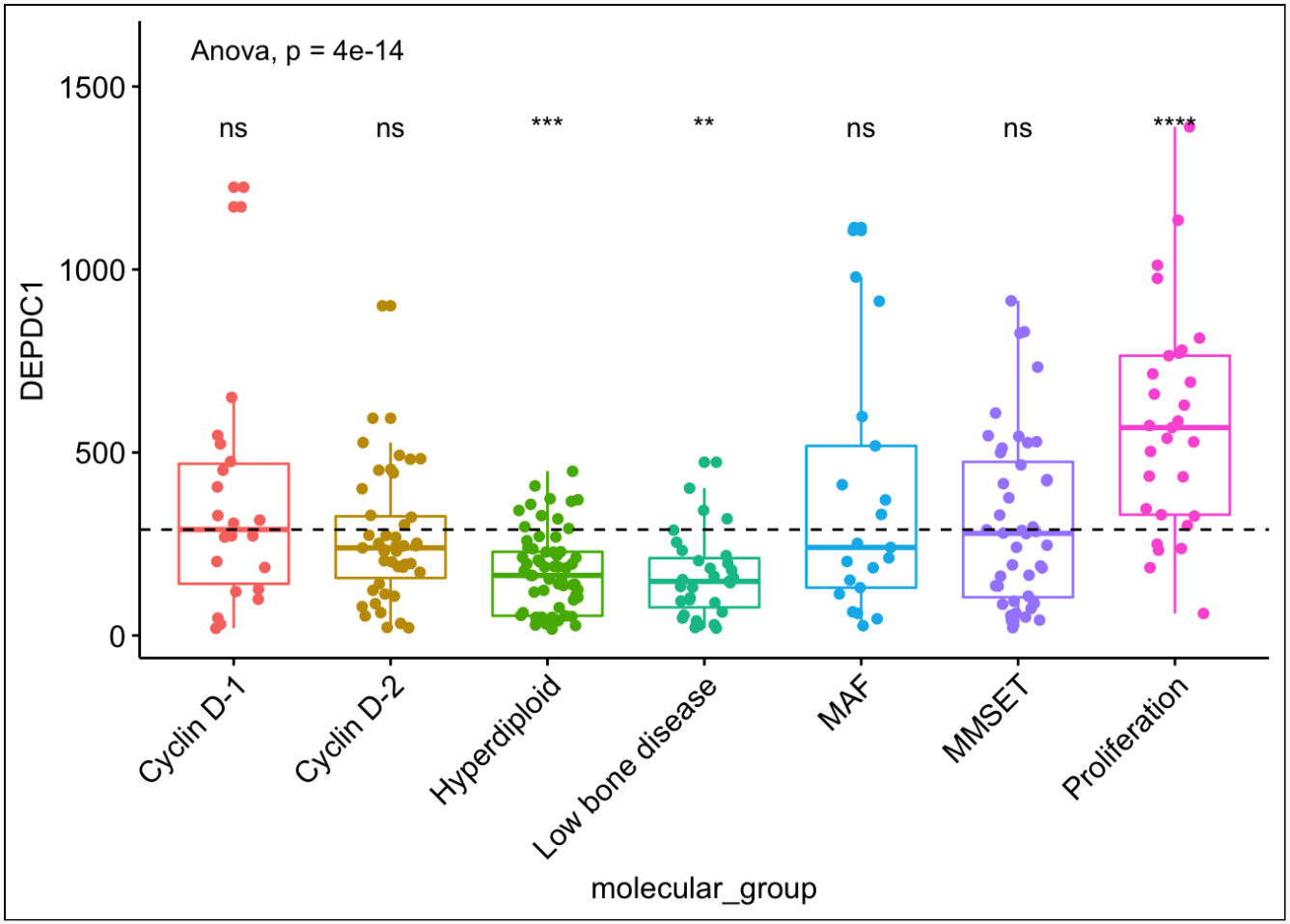
Examples of R Graphics

Boxplot



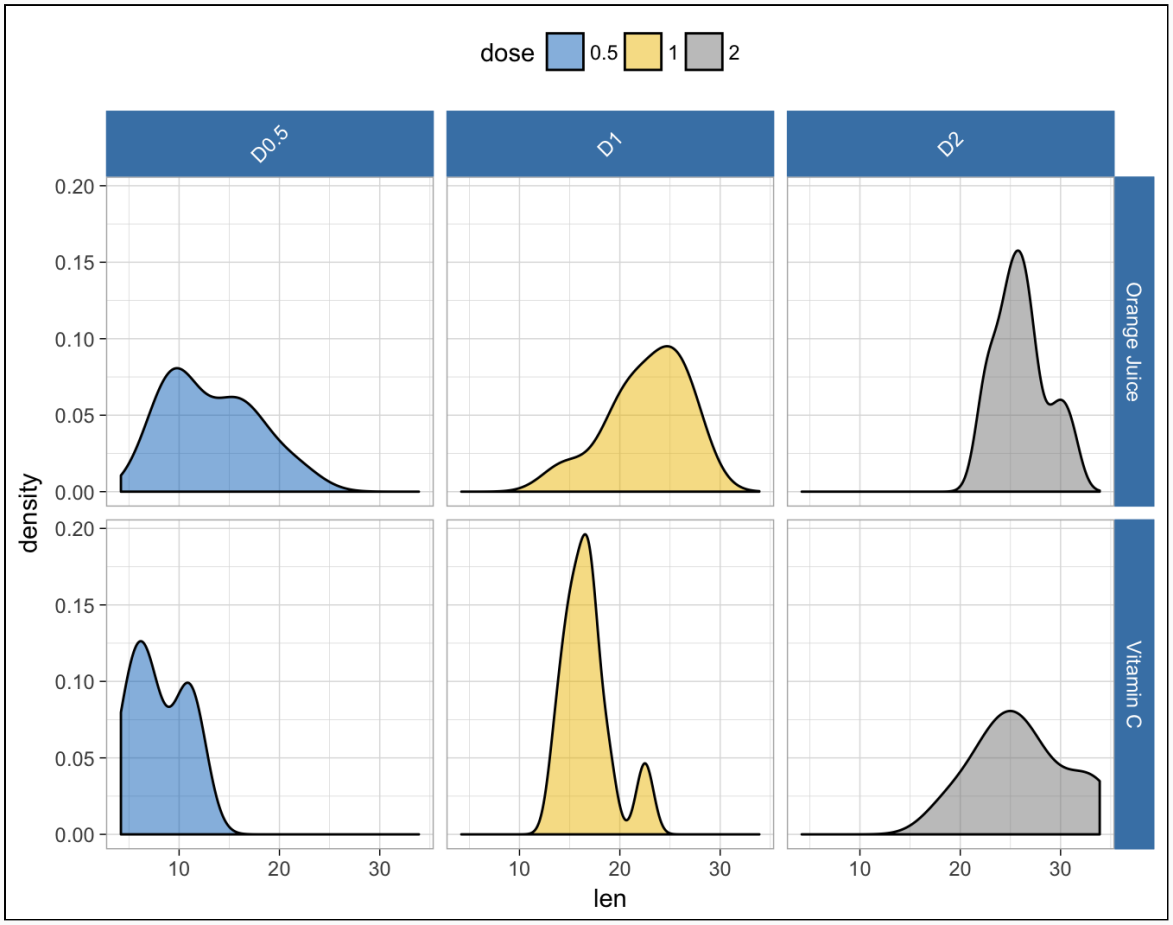
Examples of R Graphics

Boxplot



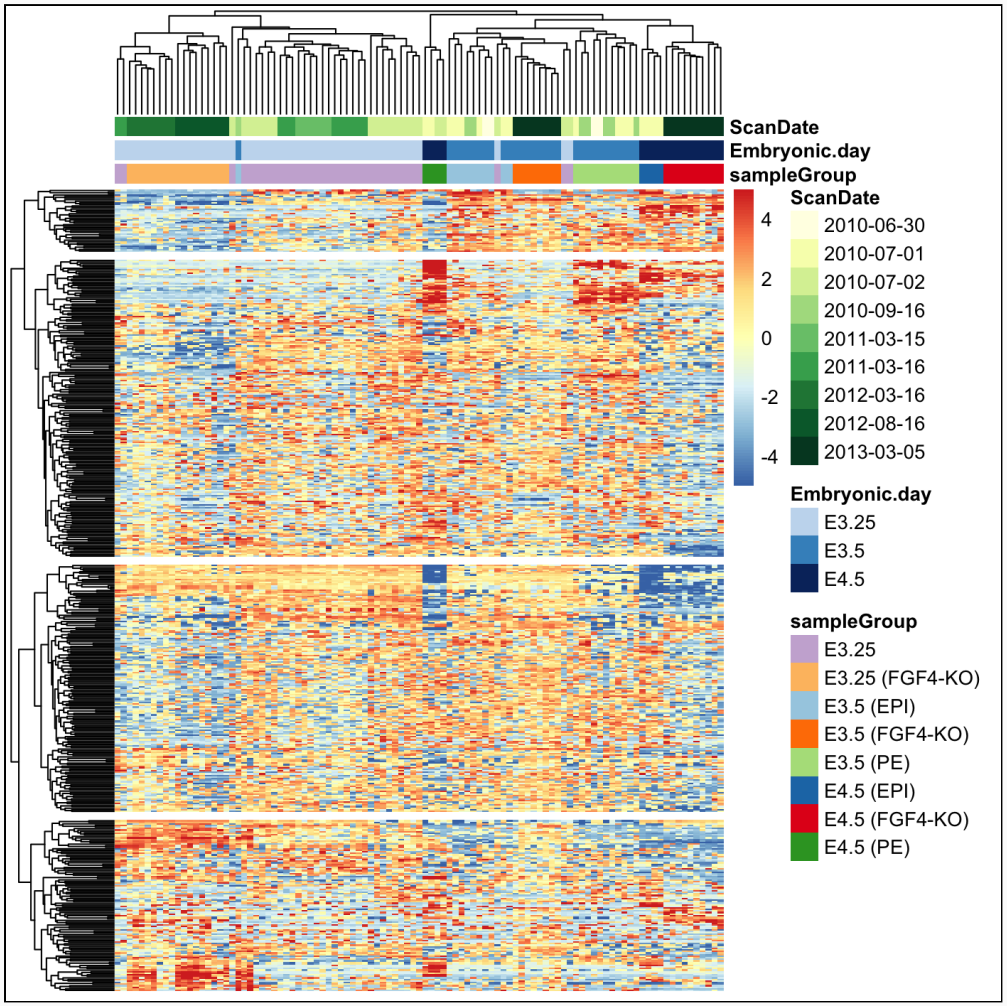
Examples of R Graphics

Faceting graphs



Examples of R Graphics

Heatmaps



Specialized tools for Bioinformatics

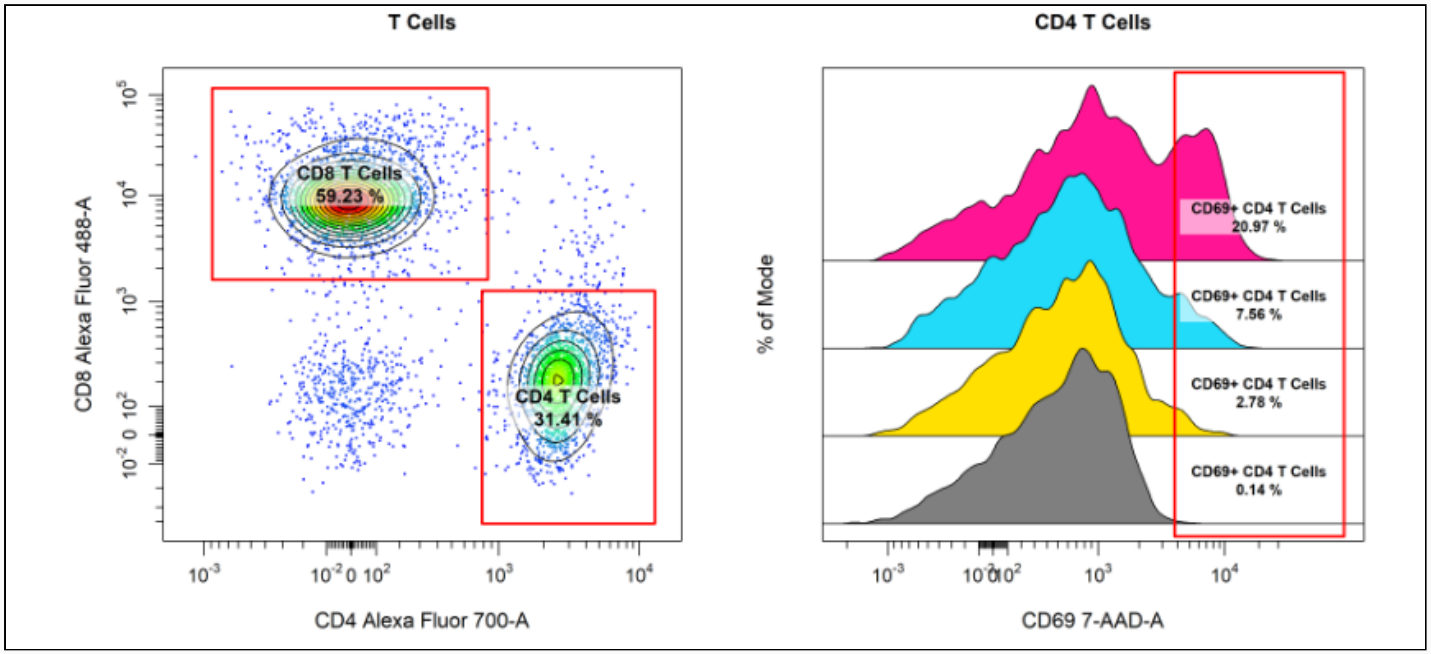
Bioconductor is a project that focuses in providing `R` tools for analysis of high-throughput biological assays. Some of the project's goals include:

- Provide widespread access to a borad range of powerful statistical and graphical methods for analysis of genomic data
- Facilitate he inclusion of biological metadata in the analysis of genomic data from different sources
- Provide a common software platform that enables rapid development of interoperable software
- Train researchers on computational and statistical methods

Specialized tools for Bioinformatics

CytoExploreR

- Specifically designed to integrate all existing cytometry analysis techniques (e.g. manual gating, automated gating and dimension reduction)



Specialized tools for Bioinformatics

CytoExploreR

Basic R Operations

```
1+1  #Simple Arithmetic
```

```
[1] 2
```

```
2 + 3 * 4 #PEMDAS also applies
```

```
[1] 14
```

```
3 ^ 2 #Exponentiation
```

```
[1] 9
```

```
exp(1) #Basic mathematical functions are available
```

```
[1] 2.718282
```

Basic R Operations

```
sqrt(10)
```

```
[1] 3.162278
```

```
pi #constants are also defined within R
```

```
[1] 3.141593
```

```
2+pi*43 #Some random operation
```

```
[1] 137.0885
```

Variables in R

- Variables are names for values. Created variables can be found in the workspace pane.
- Numeric
  - Store floating point values
- Boolean (True or False)
  - TRUE or FALSE are reserved values, can not be used to name variables
- Strings
  - Essentially text, sequences of characters
- A variable can be determined with the `←` operator

Variables in R

- Shortcut for the `←` operator, `Alt + -`
- `x ← 1` Assigns 1 to a variable called `x`
- `y ← 3` Assigns 3 to a variable called `y`
- `z ← 4` Assigns 4 to a variable called `z`
- `x*y*x` We can perform operations with variables, if allowed

```
[1] 12
```

Variables in R

- Variables under the same name can overwrite each other when called later

```
x ← 1
x ← 3

x == 1 #Is x still 1?
```

```
[1] FALSE
```

```
x == 3 #Is x now 3?
```

```
[1] TRUE
```

Variables in R

- Certain rules must be respected when naming variables
- `letters_123. ← 4` Letters, numbers, dots, and underscores are allowed
  - `letters.123_awesome`
  - Can start with a dot but it should not be followed by a number `.letters123`
- Variables can't start with numbers `1illegalvariable`
- The `%` character isn't allowed `another%badvariable`
- Starting with `_` isn't valid `_lastbadvariable`
- Variables can hold Numeric, Boolean and String values the same way
  - `x ← 23`
  - `y ← "I love this variable"`

Vectors

- In `R` a vector is a collection of numbers, individual numbers are elements of the vector

```
myvector ← c(10,20,30,40,50)

myvector + 1
```

```
[1] 11 21 31 41 51
```

```
myvector + myvector
```

```
[1]  20  40  60  80 100
```

```
length(myvector)
```

```
[1] 5
```

```
c(myvector, myvector)
```

```
 [1] 10 20 30 40 50 10 20 30 40 50
```

Vectors

- We can create and index vectors with `:`
- Indexing means accessing specific elements in the vector

```
autovector ← 1:10
autovector
```

```
 [1]  1  2  3  4  5  6  7  8  9 10
```

```
autovector[1]
```

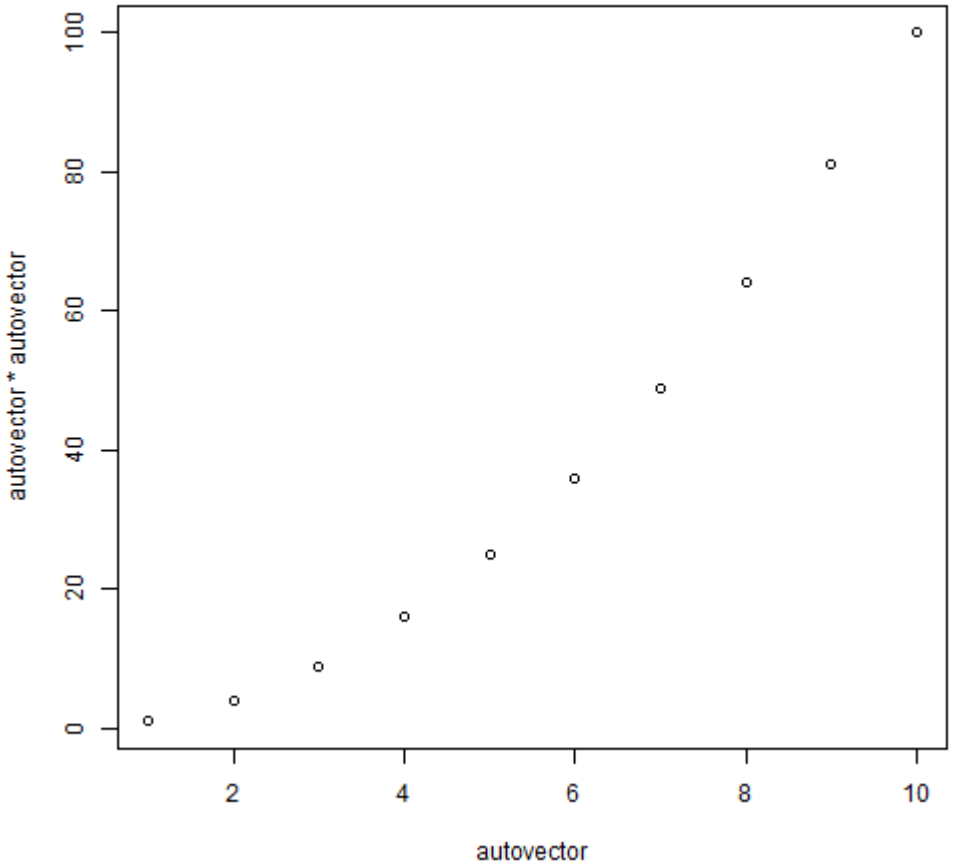
```
[1] 1
```

```
items ← c("spam", "eggs", "beans", "bacon", "sausage")
items[1:4]
```

```
[1] "spam"  "eggs"  "beans" "bacon"
```

Vectors

```
plot(autovector, autovector*autovector)
```



Functions

- Functions are things that `R` do for us: calculate, manipulate data, read and write to files, produce plots.
- `R` contains many functions and we can also build our own (more advanced stuff)

```
myvector
```

```
[1] 10 20 30 40 50
```

```
sum(myvector)
```

```
[1] 150
```

```
rep(42,10) #Repeats the number 42, 10 times
```

```
 [1] 42 42 42 42 42 42 42 42 42 42
```

```
rep(c(1,2,3),10) #Functions work from the inner brackets to the outer ones. In this ca
```

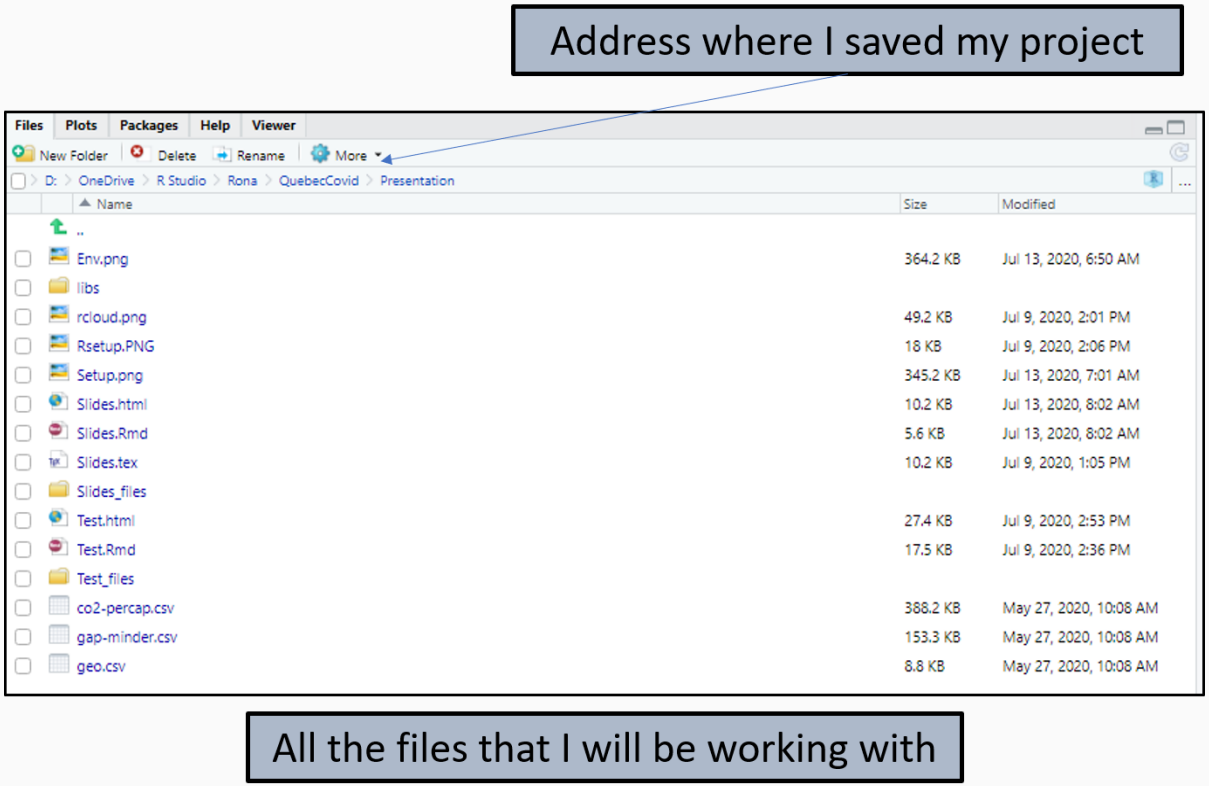
```
 [1] 1 2 3 1 2 3 1 2 3 1 2 3 1 2 3 1 2 3 1 2 3 1 2 3 1 2 3 1 2 3
```

Data frames and the Tidyverse

- Data frames is the `R` name for tabular data. Essentially tables.
- Generally we want each row in a data frame to represent a unit of observation and each column to contain a different type of information
- As a beginner, it is advisable to prepare your tabular data in Excel prior to working in `RStudio`
- We will work with data using a package called `Tidyverse` which contains many popular tools used for data analysis in `R`
- To install `Tidyverse` in your computer write `install.packages("tidyverse")`
- To load `Tidyverse` write `library(tidyverse)`
- Whenever you close `RStudio` make sure to run the `library(tidyverse)` function the first time you log back in

Opening a csv file

- Save the file you wish to work with in the directory where you saved your project
- This will ensure you don't have you search everywhere in your computer and makes it easy for `RStudio` to access them



Opening a csv file

```
geo ← readr::read_csv("geo.csv")
geo
```

```
# A tibble: 196 x 7
   name                region   oecd  g77     lat   long income2017
   <chr>               <chr>    <lgl> <lgl> <dbl>  <dbl> <chr>
 1 Afghanistan         asia     FALSE TRUE   33    66    low
 2 Albania             europe   FALSE FALSE  41    20    upper_mid
 3 Algeria             africa   FALSE TRUE   28     3    upper_mid
 4 Andorra             europe   FALSE FALSE  42.5   1.52 high
 5 Angola              africa   FALSE TRUE  -12.5  18.5  lower_mid
 6 Antigua and Barbuda americas FALSE TRUE   17.0 -61.8  high
 7 Argentina           americas FALSE TRUE  -34   -64    upper_mid
 8 Armenia             europe   FALSE FALSE  40.2  45    lower_mid
 9 Australia           asia     TRUE  FALSE -25   135    high
10 Austria             europe   TRUE  FALSE  47.3  13.3  high
# ... with 186 more rows
```

Exploring the data

- We can extract details from the dataset

```
nrow(geo)  #Number of observations
```

```
[1] 196
```

```
ncol(geo)  #Number of columns
```

```
[1] 7
```

```
colnames(geo)  #Names of the columns
```

```
[1] "name"       "region"     "oecd"       "g77"        "lat"
[6] "long"       "income2017"
```

Exploring the data

```
summary(geo)  #Descriptive statistics (if available) for each column
```

```
    name               region              oecd            g77
Length:196         Length:196         Mode :logical   Mode :logical
Class :character   Class :character   FALSE:165       FALSE:65
Mode  :character   Mode  :character   TRUE :31        TRUE :131


     lat               long           income2017
Min.   :-42.00   Min.   :-175.000   Length:196
1st Qu.:  4.00   1st Qu.:  -5.625   Class :character
Median : 17.42   Median :  21.875   Mode  :character
Mean   : 19.03   Mean   :  23.004
3rd Qu.: 39.82   3rd Qu.:  51.892
Max.   : 65.00   Max.   : 179.145
```

Indexing data frames

- The syntax to subset (selecting parts of a table) data frames is `[row,column]`

```
geo[4, 2]  #Here were are taking row 4, column 2.
```

```
# A tibble: 1 x 1
  region
  <chr>
1 europe
```

```
geo[4, "region"]  #We can also use the name of the column instead. Be aware that this
```

```
# A tibble: 1 x 1
  region
  <chr>
1 europe
```

Indexing data frames

```
geo[4, ]  #We can choose a row and all the columns that are associated with it by leav
```

```
# A tibble: 1 x 7
  name    region oecd  g77     lat  long income2017
  <chr>   <chr>  <lgl> <lgl> <dbl> <dbl> <chr>
1 Andorra europe FALSE FALSE  42.5  1.52 high
```

```
head(geo[, "region"])  #Similarly we can choose an entire column alone with all its ro
```

```
# A tibble: 6 x 1
  region
  <chr>
1 asia
2 europe
3 africa
4 europe
5 africa
6 americas
```

Working with Data

Indexing data frames

```
geo[c(1, 50, 180), ]  #We can index different rows in one line
```

```
# A tibble: 3 x 7
  name               region   oecd  g77     lat  long income2017
  <chr>              <chr>    <lgl> <lgl> <dbl> <dbl> <chr>
1 Afghanistan        asia     FALSE TRUE   33    66   low
2 Dominican Republic americas FALSE TRUE   19   -70.7 upper_mid
3 Turkmenistan       asia     FALSE TRUE   39.8  59.7 upper_mid
```

```
head(geo[2:8, ])  #Selecting a sequence of rows is called slicing
```

```
# A tibble: 6 x 7
  name                region   oecd  g77     lat   long income2017
  <chr>               <chr>    <lgl> <lgl> <dbl>  <dbl> <chr>
1 Albania             europe   FALSE FALSE  41    20    upper_mid
2 Algeria             africa   FALSE TRUE   28     3    upper_mid
3 Andorra             europe   FALSE FALSE  42.5   1.52 high
4 Angola              africa   FALSE TRUE  -12.5  18.5  lower_mid
5 Antigua and Barbuda americas FALSE TRUE   17.0 -61.8  high
6 Argentina           americas FALSE TRUE  -34   -64    upper_mid
```

Logical indexing

- Instead of specifying rows and columns we can use logic to determine `TRUE` values that fit our request
- The `tidyverse` package provides the function `filter()` which allows this to happen in a much easier way than using pure `R`
- Some operators available include
  - `x == y` Equal to
  - `x != y` Not equal to
  - `x < y` Less than
  - `x > y` Greater than
  - `x <= y` Less than or equal to
  - `x >= y` Greater than or equal to

Logical indexing

- We want countries that are positioned on latitudes that are negative, meaning southern

```
dplyr::filter(geo, lat < 0)  #Select the dataframe first, then the condition, negative
```

```
# A tibble: 40 x 7
   name             region   oecd  g77     lat  long income2017
   <chr>            <chr>    <lgl> <lgl> <dbl> <dbl> <chr>
 1 Angola           africa   FALSE TRUE  -12.5  18.5 lower_mid
 2 Argentina        americas FALSE TRUE  -34   -64   upper_mid
 3 Australia        asia     TRUE  FALSE -25   135   high
 4 Bolivia          americas FALSE TRUE  -17   -65   lower_mid
 5 Botswana         africa   FALSE TRUE  -22    24   upper_mid
 6 Brazil           americas FALSE TRUE  -10   -55   upper_mid
 7 Burundi          africa   FALSE TRUE   -3.5  30   low
 8 Chile            americas TRUE  TRUE  -33.5 -70.6 high
 9 Comoros          africa   FALSE TRUE  -12.2  44.4 low
10 Congo, Dem. Rep. africa   FALSE TRUE   -2.5  23.5 low
# ... with 30 more rows
```

Simple Graph Creation

- `Tidyverse` contains the package `ggplot2` which can assist in creating custom made plots
- Three things to consider
  - A dataframe
  - How the columns of the dataframe can be translated into positions, colors, size, and shapes of graphical elements. The "aesthetics"
  - The actual graphical element to display (scatterplot, lineplot, barplots)

```
## # A tibble: 6 x 11
##   name   year population gdp_percap life_exp region oecd  g77     lat   long
##   <chr> <dbl>      <dbl>      <dbl>    <dbl> <chr>  <lgl> <lgl> <dbl>  <dbl>
## 1 Afgh~  1800    3280000        603     28.2 asia   FALSE TRUE   33    66
## 2 Alba~  1800     410445        667     35.4 europe FALSE FALSE  41    20
## 3 Alge~  1800    2503218        715     28.8 africa FALSE TRUE   28     3
## 4 Ando~  1800       2654       1197     NA   europe FALSE FALSE  42.5   1.52
## 5 Ango~  1800    1567028        618     27.0 africa FALSE TRUE  -12.5  18.5
## 6 Anti~  1800      37000        757     33.5 ameri~ FALSE TRUE   17.0 -61.8
## # ... with 1 more variable: income2017 <chr>
```

Simple Graphs

```
ggplot2::ggplot(gap_geo, ggplot2::aes(x = year, y = life_exp)) + ggplot2::geom_point()
```

Simple Graphs

Let's trying adding more `Aesthetics(R)`

```
ggplot2::ggplot(gap_geo, ggplot2::aes(x = year, y = life_exp, color = region, size = p
    ggplot2::geom_point()
```

Simple Graphs

-Boxplots look cool in publications (not these ones though) -The year aesthetic separates the boxplots by year, otherwise it would be one single boxplot

```
ggplot2::ggplot(gap_geo, ggplot2::aes(x = year, y = life_exp, color = region, group =
    ggplot2::geom_boxplot()
```