

# Introduction to R - Young Researchers Fellowship Program

## Lecture 1 - The bare basics

Daniel Sánchez Pazmiño

Laboratorio de Investigación para el Desarrollo del Ecuador

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## Brief intro to R



### Tip

*R is, at its heart, an elegant and beautiful language, well tailored for data analysis and statistics.* Hadley Wickham, Chief Scientist at Posit (formerly RStudio), 2019.









# Why R for academic research?

- Scripting capabilities: complex analyses are the bread and butter of modern social science research.
  - R will allow you to automate these analyses, and to reproduce them easily.
  - Other click-and-point software (like SPSS, Minitab, etc.) are worse at this.
- Reproducibility: R allows you to write scripts that can be shared with others, and that can be run by others to reproduce your results.
  - This is a key aspect of the scientific method, and is becoming more and more important in academic research (journals require it).
  - Reproducibility allows for easier error detection and correction.
  - Other software make this very complicated, even those that allow you to write scripts (like Stata).



- Free: less inequality in access to software, and less dependence on the whims of your institution, or on the availability of pirated software.
- Flexibility and availability of packages: R has a large community of developers, and a large repository of packages that can be used for a wide variety of purposes.
  - Often cutting-edge methods are available in R before they are available in other software as statisticians directly contribute to R rather than waiting for software developers to implement them.
  - This is especially important for interdisciplinary research.

# The R ecosystem

- **R:** The base software, which you can download from CRAN.
- **IDEs:** Integrated Development Environments, which make it easier to write and run R scripts.
  - RStudio: The most popular IDE for R.
  - Jupyter: A notebook interface that allows you to write and run code in a more interactive way.
  - VS Code: A general-purpose code editor that can be used for R.
  - Many more...
- **Packages:** Libraries of functions that extend the capabilities of R.
  - CRAN: The main repository of R packages.
  - Bioconductor: A repository of packages for bioinformatics.
  - GitHub: where many developers share their packages before they are published in CRAN (if ever).

# The R ecosystem

- Graphical user interfaces (GUIs): Software that allows you to use R without writing code.
  - R Commander: A GUI for R.
  - Rattle: A GUI for data mining.
  - RKWard: A GUI for R that integrates with the KDE desktop environment.
  - Many more...
- Analytical software based on R: Software that uses R in the background, but that is not R itself.
  - JASP: A software package that uses R in the background.
  - Jamovi: A software package that uses R in the background.
  - Many more...

# The R ecosystem

- Database integrations and APIs: R can connect to many databases and APIs, allowing you to pull data directly into R for analysis.
- Document markup: R can be used to create reports, papers, and presentations.
  - R Markdown: A package that allows you to write reports, papers, and presentations in R.
  - Bookdown: A package that allows you to write books in R.
  - Quarto: A package that allows you to write reports, papers, and presentations in R.
  - Knitr: A package that allows you to create dynamic documents in R using  $\text{\LaTeX}$ .
- Business intelligence software integrations: R can be used to create reports and visualizations integrated with business intelligence software, such as Tableau, Power BI, and Qlik.

## The garden and the gardener: CRAN and package developers

- Expert programmers in R don't just use R, they also contribute to it.
  - They write packages that extend the capabilities of R.
  - They share these packages with the community by publishing them on CRAN, GitHub, or other repositories.
  - They maintain these packages, fixing bugs and adding features as needed.
- CRAN has very strict rules for package submission, which ensures that packages are of high quality.
  - This is one of the reasons why R is so popular in academia and industry.
  - It is also one of the reasons why R is so powerful and flexible.
- Other open-source languages (like Python) have similar ecosystems, but R's gardener, the R Core Team, is very active and has a lot of experience in maintaining a large and complex software project.
  - This is one of the reasons why R is so reliable and trustworthy.



Where R is being used in academia

- **Social science:** Paid solutions like Stata, SPSS, SAS and EViews are unfortunately still popular
  - Economics: Stata largely the leader, but R gaining traction very quickly.
  - Political science: code-based software like Stata still trying to overcome the click-and-point software, but some researchers starting to use open-source alternatives
  - Psychology: SPSS is still the leader, but R and Stata gain field

See more:

- 87% of social researchers say they want or need to learn R
- Why universities are switching to R for teaching social science



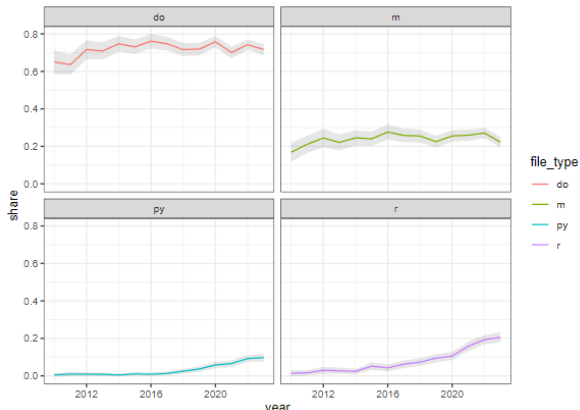






# Where R is strong: lessons from usage in Economics

- An analysis of replication files in Sebastian Kranz replication database finds that both R and Python files have increased in the last years, yet are still a minority compared to Stata and Matlab.



- The recent “data revolution” created the notion that R is in high demand in industry.
  - At best, this is a half-truth: R is in high demand in some industries, but not in all.
- R is very academic still, so industries that use it are those that are academia-adjacent
  - Health: R is used in large pharmaceutical companies, hospitals, and research institutions.
  - Insurance: R is used in some insurance companies, especially those that do actuarial work.

# Outside academia: R in industry

- Government and non-profits: R is used in some government agencies and non-profits, especially those that do research or data analysis.
  - Governments employ many academics, so it is not surprising that they use R.
- Economics consulting firms are starting to use R more, but are still largely using very antiquated paid software (i.e. SPSS, EViews).
  - Legacy code, inflexible leadership and the priority of speed over quality keep private industry from innovating in workflows.
- Some industries like health and insurance are in regulatory capture, which is why paid software is still the norm.
  - SAS is only slowly being phased out, as government regulation requires pharmaceutical companies to provide analyses in “approved” tools.



## Some thoughts for the future

- R is likely to overcome paid software in several fields in the next few years, but will probably do so slowly.
  - The main reason is that paid software is very entrenched in academia, and it is hard to change the status quo.
- Python is not so much a competitor to R for academic work, but it is a competitor for data science work.
  - Python is more versatile and better suited for production work.
  - Yet, it is not as good as R for data analysis and statistics.

## Some thoughts for the future

- Julia is a language that may prove to be the next great thing in academic research, as it is as good as R for data analysis and statistics, and offers some of the production capabilities of Python.
  - It is superior to R in reproducibility and speed.
  - Also offers better possibilities for numerical computing, which R lacks.
  - The tidyverse is being implemented in Julia! Check out Tidier.jl
- Posit (formerly RStudio), is slowly moving to a more general-purpose data science platform, with the introduction of Quarto and their new IDE, Positron.
  - Posit is a company after all, and academia is a limited market.
- Keep learning, and keep an open mind. The worst enemy of a good researcher is arrogance.



# Getting started with R



# RGui

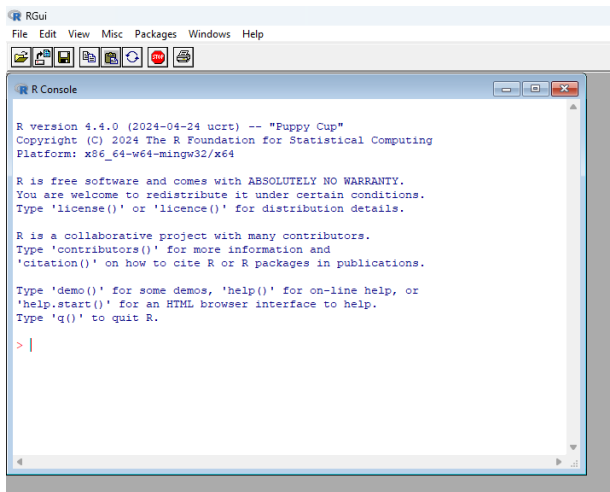


Figure 4: The RGui

# RGui and RStudio

- We typically don't use the RGui for anything other than updating R.
  - It isn't very user-friendly, and doesn't have many features.
- Rather, we will download an IDE, which makes it easier to write and run R scripts.
  - RStudio is the most popular IDE for R, and is what we will use in this course.
  - As you become more adept at R, you may want to try other IDEs, like VS Code.
- You CANNOT run RStudio without having R installed on your computer.
  - RStudio is just a front-end for R, and needs R to be installed in order to work.

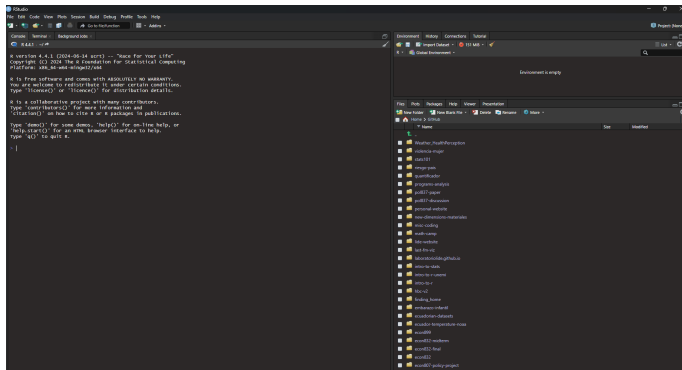


Figure 5: The RStudio IDE

You may see RStudio with a white background when you first open it. I have customized my RStudio to have a dark background. I suggest you do the same, as it is easier on the eyes.

# Advantages of RStudio

- **Script editor:** A place to write and run R scripts.
  - Syntax highlighting, code completion, and other features to make writing code easier.
  - You can run code line-by-line, or all at once.
  - The console is integrated, so you can see the output of your code as you run it.
- **R Projects:** helps for an easy organization of your work, especially file paths.
- **Environment:** A place to see the objects in your workspace, and to manage them.



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# RStudio Panes

## Source Pane

Edit and run scripts (e.g. Rmarkdown templates), and view datasets

Tip: Start new script

Tip: Run script

## Environment Pane

Overview of objects (datasets, parameters, lists, etc.) you have imported or created.

Tip: Zoom and export plots

## R Console Pane

R commands run are shown here, and non-graphic output and errors are displayed

## Plots, Packages, and Help Pane

Commonly used to view graphics, install packages, and view help

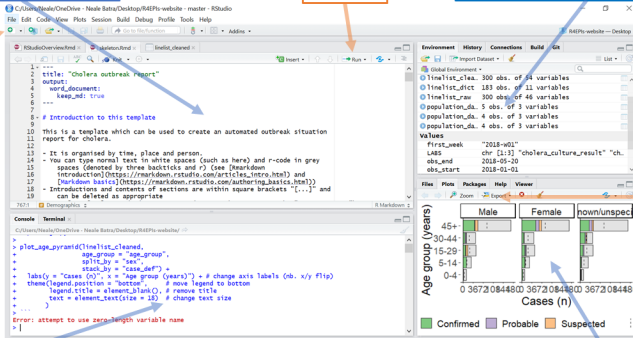


Figure 6: RStudio Panes

# Posit Cloud

- Posit Cloud is a cloud-based version of RStudio.
  - It is free to use, and is a good option if you don't want to install R and RStudio on your computer.
  - It is also a good option if you want to collaborate with others on R projects.
- You will need to create an account to use Posit Cloud.
  - You can sign up for an account at [cloud.rstudio.com](https://cloud.rstudio.com).
- You can also use Posit Cloud to run R scripts on a server.
  - This is free to a certain extent, but you may need to pay for more resources if you use it a lot.

# Datalab, formerly DataCamp Workspace

- Datalab is a cloud-based version of Jupyter Notebooks.
  - It is free to use to a certain extent, and is a good option if you don't want to install R and RStudio on your computer.
  - It is also a good option if you want to collaborate with others on R projects.
- Use a DataCamp account to access DataLab, and be sure to select R as the language you want to use (Python also available).
- Similar resources to run R online are available at [repl.it](https://repl.it).

- When you open RStudio, it is a good idea to store your work in an R Project.
  - This will help you keep your work organized, and will make it easier to share your work with others.
- To create an R Project, click on **File** in the top menu, then click on **New Project**.
- You will be asked to choose a directory for your project.
  - Choose a directory where you want to store your project, and give your project a name.

## Getting started: using the R Project

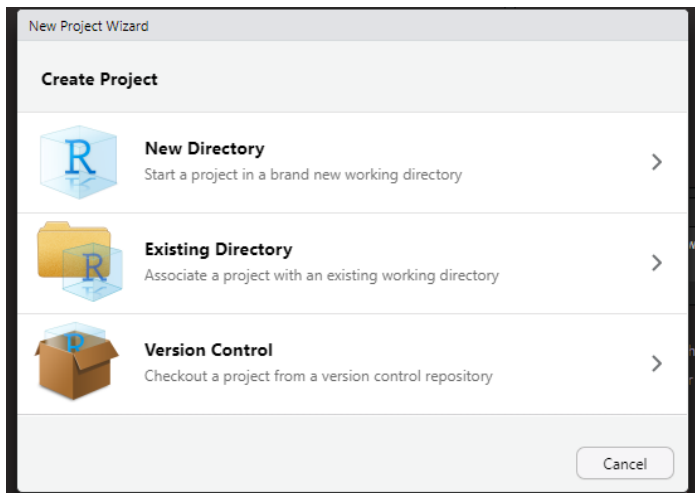


Figure 7: Creating an R Project

- R Projects let you store your work in a single directory, and keep all of your files and data in one place.
  - This makes it easier to share your work with others, and to keep track of your work over time.
- Creating a project creates an `.Rproj` file in the project directory.
  - This file tells RStudio that the directory is an R Project, and allows RStudio to manage the project.
- When thinking about file paths, everything is relative to the project directory.
  - This means that you can move your project directory to another location, and everything will still work.

## An analogy: RStudio as a kitchen

- The R console: the stove, where you cook your food.
  - The food needs to be cooked in the stove, there is no way around this.
  - You may cook the food in the stove, remembering steps by heart, or you may use a `**recipe/`
- The script editor: the recipe book, where you write down the steps to cook your food.
  - You can write down the steps to cook your food, and then follow them to cook the food.
  - You can also share the recipe with others, so that they can cook the food too.
- The environment: the kitchen table, where you put your ingredients and intermediate outputs from the recipe.
  - You can see what ingredients you have, and what ingredients you need to buy.
  - You can also see how much of each ingredient you have.
  - As you cook intermediate steps, you may store the results in the kitchen table.

## An analogy: RStudio as a kitchen

- The file viewer: the pantry, where you store your ingredients and recipes.
  - You can see what ingredients you have, and what ingredients you need to buy.
  - You can also see what recipes you have, and what recipes you need to buy.
  - As you cook, you take ingredients from the pantry (files) and put them in the kitchen table (environment).
- Libraries: kitchen utensils which you need to use to cook your food.
  - You may have a lot of kitchen utensils, but you only use the ones you need to cook your food.
  - In principle you can cook with just a stove, but it is much easier to cook with the right utensils.
  - If you see a recipe that requires a kitchen utensil you don't have, you can buy it or borrow it from someone else (install a package).





# Installing packages

- To install a package, you can use the `install.packages()` function.

```
install.packages("dplyr")  
install.packages("ggplot2")  
install.packages("babynames")
```

- You may also install from the Packages pane in the bottom right corner of RStudio.



# Error, warning and message messages

- When you run code in R, you may see three types of messages: errors, warnings, and messages.
- **Errors:** These are messages that tell you that something went wrong.
  - You need to fix the error before you can continue running the code.
  - Errors are usually in red text, and are followed by a message that tells you what went wrong.
- **Warnings:** These are messages that tell you that something might be wrong.
  - You can usually continue running the code, but you should be aware that something might be wrong.
  - Warnings are usually in yellow text, and are followed by a message that tells you what might be wrong.





# Dealing with file paths

- Whenever you work with files in R, you need to think about file paths.
  - A file path is the location of a file on your computer.
  - File paths can be absolute or relative.
  - Absolute file paths start at the root directory of your computer.
  - Relative file paths start at the working directory of your R session.
- Absolute file paths are those that start at the root directory of your computer.
  - They are usually very long, and are hard to read and write.
  - They are also hard to share with others, as they are specific to your computer.

`C:/Users/daniel/Documents/MyProject/data.csv`

- Relative file paths are those that start at a specific directory on your computer.
  - They are usually shorter, and are easier to read and write.
  - They are also easier to share with others, as they are not specific to your computer.

data.csv

- Relative paths may cause problems when trying to run code from other computers.
  - Though, we've seen that R Projects help with this!



- To set the working directory in R, you can use the `setwd()` function.
  - This becomes unnecessary when using R Projects, but it is good to know.
  - `setwd()` takes a single argument, which is the path to the directory you want to set as the working directory.
  - Similar to `cd` in the terminal.
- There are packages that can help you work with file paths in R.
  - The `here` package is a good option, as it allows you to work with file paths in a platform-independent way.





## Working with R in RStudio

- To create a variable or object in R, you can use the `<-` operator.
  - This is called the assignment operator, and is used to assign a value to a variable.
  - The variable name goes on the left side of the assignment operator, and the value goes on the right side.
  - The value can be a number, a character string, a logical value, or a vector.
  - The variable name can be any combination of letters, numbers, and underscores, but cannot start with a number.

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# Other assigners

- You can assign the other way around:

```
1 + 1 -> a
```

```
a
```

```
[1] 2
```

```
# Don't do this, it is less readable!!
```

```
b = 1 + 1
```

```
b
```

```
[1] 2
```

# Classes of objects in R

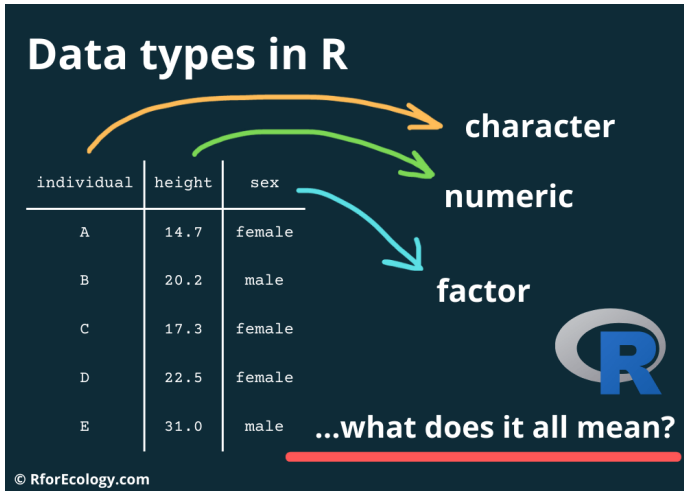


Figure 8: Classes of objects in R



# Types of structures in R

- **Vectors:** A single column of data.
  - Can be numeric, character, or logical, and can mix types.
  - Can be of any length.
  - Can be created with the `c()` function (means concatenate).
- **Matrices:** A two-dimensional array of data.
  - Can be numeric, character, or logical, but all elements must be of the same type.
  - Can be of any size.
  - Can be created with the `matrix()` function.

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- To see the structure of an object in R, you can use the `str()` function.
  - This function shows you the class of the object, and the structure of the object.
  - The structure of an object is the way that the object is organized in memory.
  - The structure of an object is important, as it determines how you can work with the object.

# Vectors

- Single dimension of data.
  - Can be numeric, character, or logical.
  - Can mix types.
  - Can be of any length.
- Can be created with the `c()` function, indexing or functions that return vectors as output.
- We may access to specific elements of the vector using the `[]` operator (indexing).

# Vector indexing

- To access an element of a vector, you can use the `[]` operator.
  - This operator takes a single argument, which is the index of the element you want to access.
  - The index can be a single number, a vector of numbers, or a logical vector.
  - The index can be positive, negative, or zero.
- Positive indices access elements from the beginning of the vector.
- Negative indices access elements from the end of the vector.
- Zero indices return an empty vector (R is 1-indexed, unlike Python).

# Logical operators

- **Comparison operators:** These operators compare two values, and return a logical value.
  - ==: Equal to.
  - !=: Not equal to.
  - >: Greater than.
  - <: Less than.
  - >=: Greater than or equal to.
  - <=: Less than or equal to.
- When running a comparison, R will return TRUE if the comparison is true, and FALSE if the comparison is false.
  - One can also write T and F instead of TRUE and FALSE.
  - 1 is often used as TRUE, and 0 as FALSE, R will convert these to TRUE and FALSE respectively when needed.

# Logical operators

- **Logical operators:** These operators combine two logical values, and return a logical value.
  - **&:** Logical AND.
  - **|:** Logical OR.
  - **!:** Logical NOT.
- The AND operator checks if both values are TRUE, and returns TRUE if they are.
- The OR operator checks if at least one value is TRUE, and returns TRUE if it is.
- The NOT operator checks if the value is TRUE, and returns FALSE if it is.

# Working with functions

- A function in R is a block of code that performs a specific task.
  - Functions take arguments, which are the inputs to the function.
  - Functions return values, which are the outputs of the function.
  - Functions can be built-in, or user-defined.
- When we “write” the code of a function to invoke it, we say that we are “calling” the function.
  - Hence, you will hear call and invoke used interchangeably.



# Working with functions

- The general form of a function in R is `'function_name(argument1, optional_argument1, optional_argument2, ...)`
  - The function name is the name of the function.
  - The arguments are the inputs to the function.
  - The optional arguments are the inputs to the function that are not required, as they have default or predefined values.

# Getting help with R

- To get help with R, you can use the `help()` function.
  - This function takes a single argument, which is the name of the function you want help with.
  - The help file for the function will be displayed in the Help pane in RStudio.
  - You can also use the `?` operator to get help with a function.
  - Typically, it will show you the arguments of the function, and some examples of how to use it, as well as predefined values for optional arguments.
- To get help with a package, you can use the `help(package = "package_name")` function.
- To get help with a topic, you can use the `help.search()` function.

# Getting help with R

- You can use many outside resources to get help with R.
  - The R documentation is a good place to start.
  - The RStudio Community is a good place to ask questions.
  - The Stack Overflow is a good place to ask questions.
  - The R-bloggers is a good place to find tutorials and examples.
- Reddit has plenty of R communities, such as r/rstats where basic questions are welcome (as Stack Overflow is a bit more strict).
- The RStudio Cheat Sheets are a good resource for learning R.
- Using Slack, Discord, or other chat platforms to ask questions is also a good idea
  - Never hesitate to ask questions, as it is the best way to learn.
  - Use the quote marks in Slack to write code, so that it is easier to read when asking questions. “”

# The LLM scene in R

- Of course, R can be learned by asking one of the new chatbots that are available.
  - ChatGPT codes well in R.
  - Gemini is a good option for R help, and recently added integration with RStudio.
  - Claude is a good option for R help, and is free to use.
  - Copilot is a good option for R help, and is free to use.
- GitHub Copilot is free for students and teachers, integrating amazingly with RStudio.
  - Check out the GitHub Copilot quickstart guide
  - Predicts code and comments, and can be used to write entire scripts.
- Learning R can be done by using AI, but it is better to learn the basics first, and then use AI to help with more complex tasks.

# Dataframes in R



# Creating dataframes

- To create a dataframe in R, you can use the `data.frame()` function.
  - This function takes a series of arguments, which are the columns of the dataframe.
  - The columns can be vectors, matrices, or other dataframes.
  - The columns must be the same length, or R will recycle the shorter columns to make them the same length.
- You can also create a dataframe by importing data from a file.
  - This is the most common way to create a dataframe in R, as it allows you to work with real data.





# Working with dataframes

- There are some key functions for working with dataframes in R.
  - `head()`: Shows the first few rows of the dataframe.
  - `tail()`: Shows the last few rows of the dataframe.
  - `str()`: Shows the structure of the dataframe.
  - `summary()`: Shows a summary of the dataframe.
- For accesing information within the dataframe, we may work with the following functions:
  - `nrow()`: Shows the number of rows in the dataframe.
  - `ncol()`: Shows the number of columns in the dataframe.
  - `dim()`: Shows the dimensions of the dataframe.
  - `names()`: Shows the names of the columns in the dataframe.
  - `colnames()`: Shows the names of the columns in the dataframe.
  - `rownames()`: Shows the names of the rows in the dataframe.
  - `unique()`: Shows the unique values in a column of the dataframe.
  - `table()`: Shows the frequency of values in a column of the dataframe.

# Working with dataframes

- Indexing also works for dataframes, we need to maintain the two-dimensional structure of the dataframe when using indices.
- For example:
  - `df[1, 2]` will return the value in the first row and second column of the dataframe.
  - `df[1, ]` will return the first row of the dataframe.
  - `df[, 2]` will return the second column of the dataframe.
  - `df[1:3, ]` will return the first three rows of the dataframe.
  - `df[, c(1, 3)]` will return the first and third columns of the dataframe.

## Working with dataframes

- We may also use logical operators to subset the dataframe.
  - `df[df$age > 30, ]` will return the rows of the dataframe where the age is greater than 30.
  - `df[df$age > 30 & df$height > 1.80, ]` will return the rows of the dataframe where the age is greater than 30 and the height is greater than 1.80.

# Working with dataframes

- If we need to access a column by name, we can use the `$` operator.
  - `df$name` will return the column of the dataframe with the name `name`.
  - `df$age` will return the column of the dataframe with the name `age`.
  - `df$height` will return the column of the dataframe with the name `height`.
  - This is simply a shortcut for `df[, "name"]`, `df[, "age"]`, and `df[, "height"]`.
- We can also use the `subset()` function to subset the dataframe.
  - `subset(df, age > 30)` will return the rows of the dataframe where the age is greater than 30.
  - `subset(df, age > 30 & height > 1.80)` will return the rows of the dataframe where the age is greater than 30 and the height is greater than 1.80.
  - Again, this is a base R shorthand for `df[df$age > 30, ]` and `df[df$age > 30 & df$height > 1.80, ]`.

# Working with dataframes

- R does not allow for numerically named columns, so we need to use the apostrophe character to access them.
  - `df$3` will return the column of the dataframe with the name 3'.
- In general, any “invalid” name (e.g. starting with a number, having spaces, etc.) can be accessed with the ' character.
  - Good practice is to avoid these names, as they can be hard to work with.
  - The best names are lowercase, with underscores separating words, and no spaces or special characters.

# Working with dataframes

- The `attach()` function can be used to attach the dataframe to the search path.
  - This means that you can access the columns of the dataframe without using the `$` operator.
  - This is not recommended, as it can lead to confusion and errors.
  - It is better to use the `$` operator, as it is more explicit and less error-prone.
- The `detach()` function can be used to detach the dataframe from the search path.
- `with()` is a better alternative to `attach()`, as it allows you to access the columns of the dataframe without using the `$` operator, but only within the `with()` function.
  - Generally, it is better to just stick to the `$` operator.
  - We will later use tidyverse functions to access columns without the `$` operator.



# The mtcars dataset

- The variables are:
  - mpg: Miles per gallon.
  - cyl: Number of cylinders.
  - disp: Displacement (cubic inches).
  - hp: Gross horsepower.
  - drat: Rear axle ratio.
  - wt: Weight (1000 lbs).
  - qsec: 1/4 mile time.
  - vs: V/S.
  - am: Transmission (0 = automatic, 1 = manual).
  - gear: Number of forward gears.
  - carb: Number of carburetors.
- In Stata, the auto dataset is often used in the same way as the mtcars dataset in R.



# The mtcars dataset

- Access the mtcars dataset by typing mtcars in the console. No need to load it, as it is built-in.

```
head(mtcars)
```

	mpg	cyl	disp	hp	drat	wt	qsec	vs	am	gear
Mazda RX4	21.0	6	160	110	3.90	2.620	16.46	0	1	4
Mazda RX4 Wag	21.0	6	160	110	3.90	2.875	17.02	0	1	4
Datsun 710	22.8	4	108	93	3.85	2.320	18.61	1	1	4
Hornet 4 Drive	21.4	6	258	110	3.08	3.215	19.44	1	0	3
Hornet Sportabout	18.7	8	360	175	3.15	3.440	17.02	0	0	3
Valiant	18.1	6	225	105	2.76	3.460	20.22	1	0	3

## Importing data from a file

- We will later learn the specifics about importing files into R as dataframes.
  - This is the most common way to work with dataframes in R.
- We can use base R functions to load the used car sales data in the data directory of the GitHub repo.

```
# Load the data using read.csv, the base R function for loading
cars <- read.csv("data/used_cars_ecuador.csv")

str(cars)
```



---

# Creating lists

```
name <- c("Alice", "Bob", "Charlie", "David", "Eve")

age <- matrix(c(25, 30, 35, 40, 45), nrow = 5, ncol = 1)

height <- data.frame(height = c(1.60, 1.70, 1.80, 1.90, 2.00))

# Create list

lst <- list(name, age, height)

str(lst)
```

List of 3

```
$ : chr [1:5] "Alice" "Bob" "Charlie" "David" ...
$ : num [1:5, 1] 25 30 35 40 45
$ : 'data.frame': 5 obs. of 1 variable:
..$ height: num [1:5] 1.6 1.7 1.8 1.9 2
```

# Working with lists

- There are some key functions for working with lists in R.
  - `length()`: Shows the number of objects in the list.
  - `names()`: Shows the names of the objects in the list.
  - `str()`: Shows the structure of the list.
  - `summary()`: Shows a summary of the list.
- For accessing information within the list, we may work with `[[`.
  - Double brackets `[[` are used to access the elements of the list.
- Indexing by list works similarly to indexing by dataframe, but we need to use the `[[` operator to access elements of the list.

# Working with lists

- For example:
  - `lst[[1]]` will return the first element of the list.
  - `lst[[2]]` will return the second element of the list.
  - `lst[[3]]` will return the third element of the list.
  - `lst[[1]][1]` will return the first element of the first element of the list.
  - `lst[[2]][1, 1]` will return the first element of the first element of the second element of the list.
  - `lst[[3]][1, 1]` will return the first element of the first element of the third element of the list.

# Working with lists

- We can also use the `$` operator to access elements of the list.
  - `lst$name` will return the element of the list with the name `name`.
  - `lst$age` will return the element of the list with the name `age`.
  - `lst$height` will return the element of the list with the name `height`.
  - This is simply a shortcut for `lst[[1]]`, `lst[[2]]`, and `lst[[3]]`.



# Working with lists

```
lst[[1]]
```

```
[1] "Alice"    "Bob"      "Charlie"  "David"    "Eve"
```

```
lst[[2]]
```

```
      [,1]  
[1,]  25  
[2,]  30  
[3,]  35  
[4,]  40  
[5,]  45
```

# Working with lists

```
lst[[3]]
```

```
      height
1      1.6
2      1.7
3      1.8
4      1.9
5      2.0
```

```
lst[[3]]$height
```

```
[1] 1.6 1.7 1.8 1.9 2.0
```