

# Introduction to R

# Session 1

Department of Sociology | University of Oxford

Casey Breen

2023-10-05

# Welcome to “Intro to R”

- Two sessions:
  - Thursday, 1pm - 4pm
  - Friday, 9:30am - 12:30pm
- Course materials available from:
  - [www.github.com/caseybreen/intro\\_r](http://www.github.com/caseybreen/intro_r)

# Course goals

- Why **R** is a powerful tool for social science research
- Install **R** and **RStudio**
- Introduction to **R** syntax and data types
- Basic understanding of data manipulation + visualization

# Course agenda

- Session 1

- Introduction + installing [R](#) and [RStudio](#)
- Overview of [RStudio](#) interface + [R](#) scripts, notebooks, quarto
- Basic syntax and data types
- Data import and export

- Session 2

- Data manipulation ([dplyr](#))
- Data visualization ([ggplot2](#))
- Best practices: coding style, commenting, and documentation
- Resources for self-teaching

# R and RStudio

- R is a statistical programming language
  - Download: <https://cloud.r-project.org>
- RStudio is an integrated development environment (IDE) for **R** programming
  - Download: <http://www.rstudio.com/download>

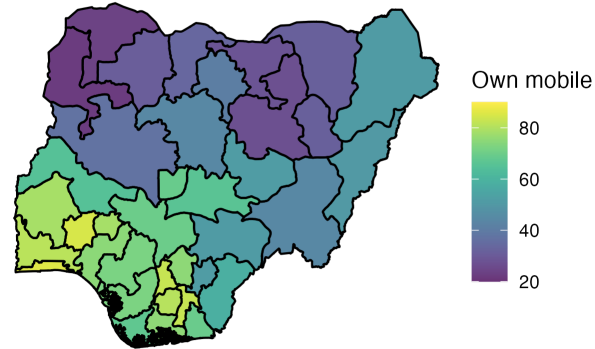
# Why R?

- Free, open source — great for reproducibility and open science
- Powerful language for data manipulation, statistical analysis
- Publication-ready data visualizations
- Well supported, excellent community

# Data visualization

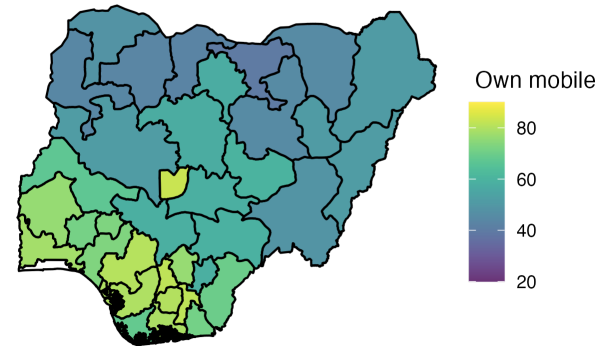
**A**

Observed



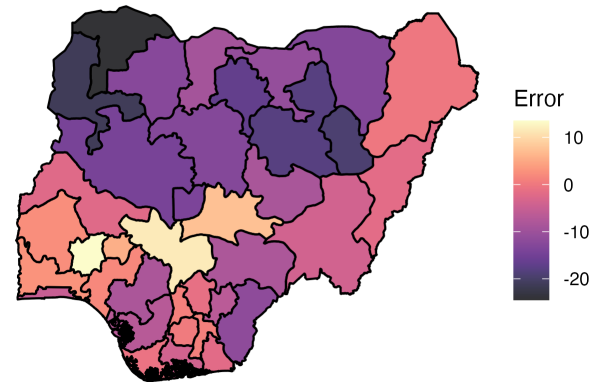
**B**

Predicted



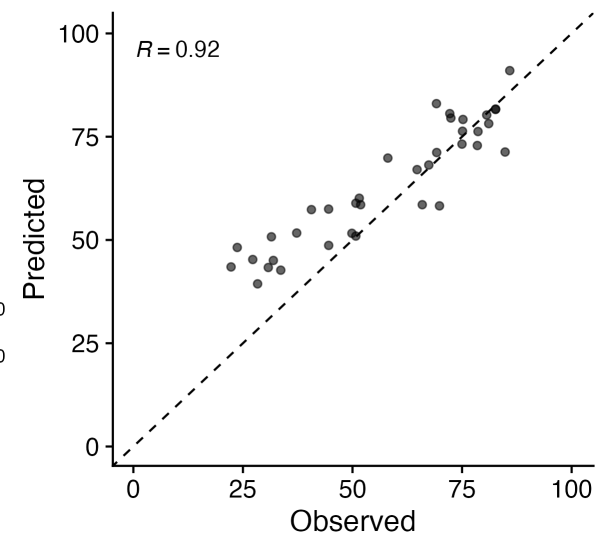
**C**

Error



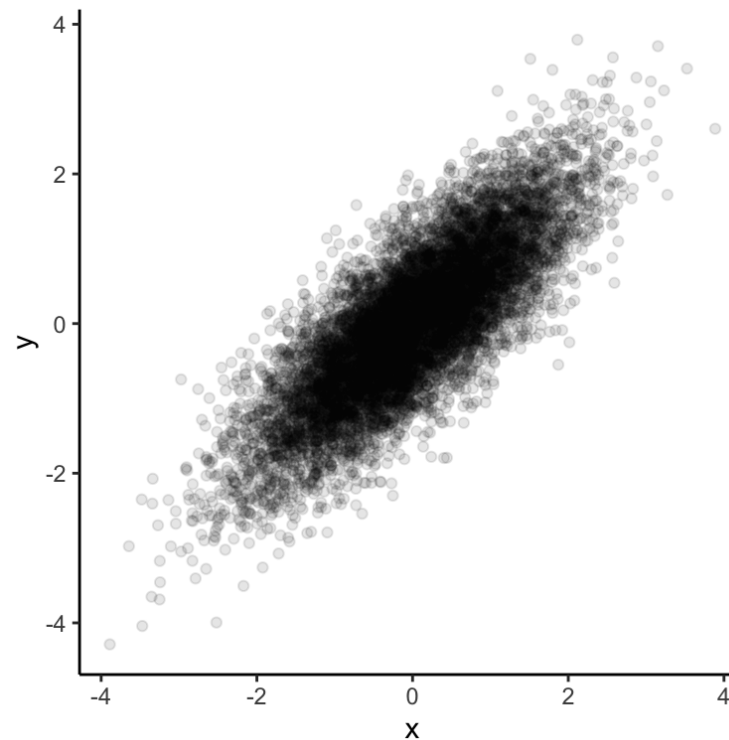
**D**

Mobile Usage, Women (Error)



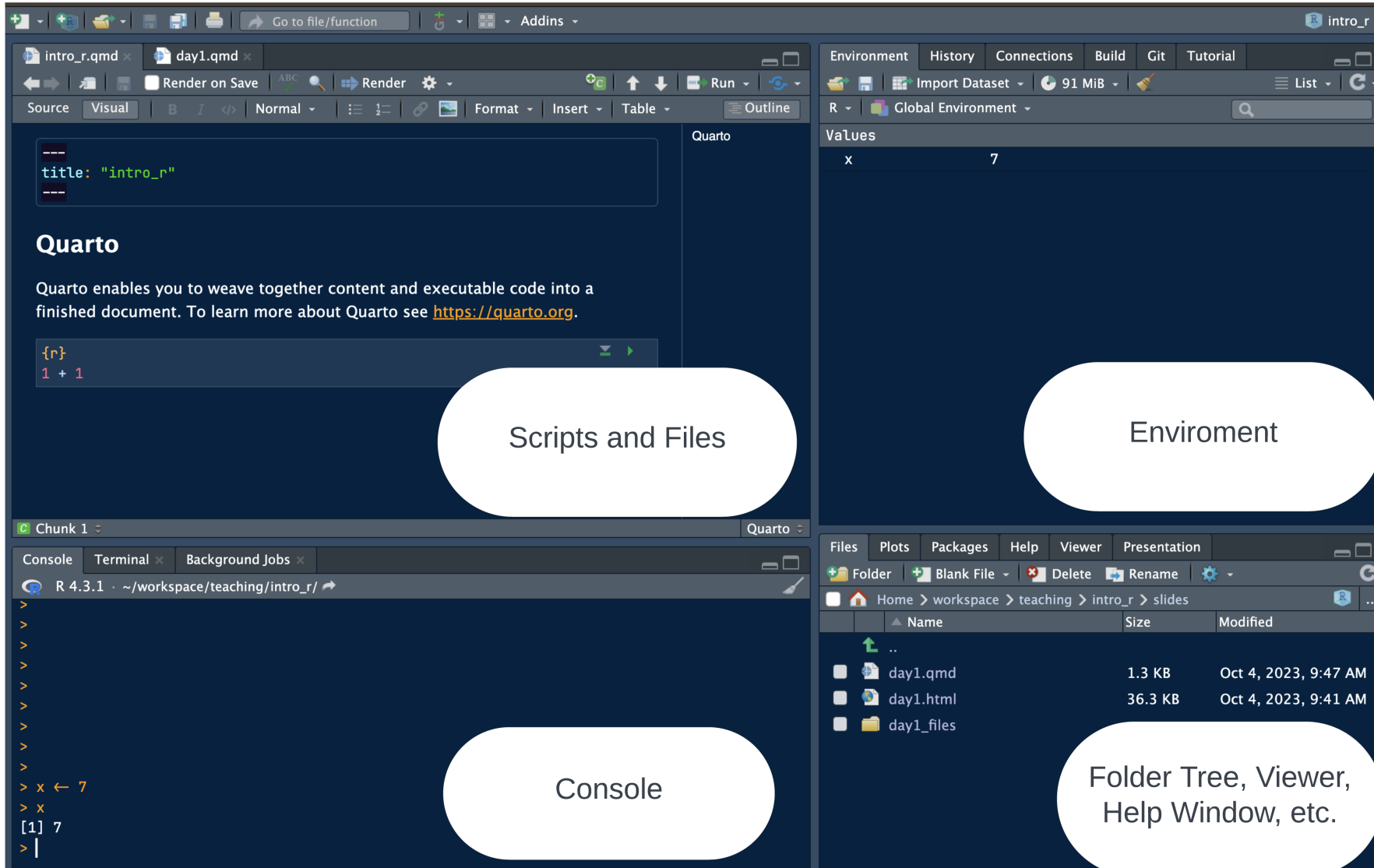
# Easy to simulate + plot data

```
1 # Generate random data for x
2 x <- rnorm(n = 10000)
3 y <- 0.8 * x + rnorm(10000, 0, sqrt(1 - 0.8^2))
4 # Create data.frame
5 data_df <- data.frame(x = x, y = y)
6 # Generate df
7 data_df %>%
8   ggplot(aes(x = x, y = y)) +
9   geom_point(alpha = 0.1) +
10  theme_classic()
```





# RStudio Panes

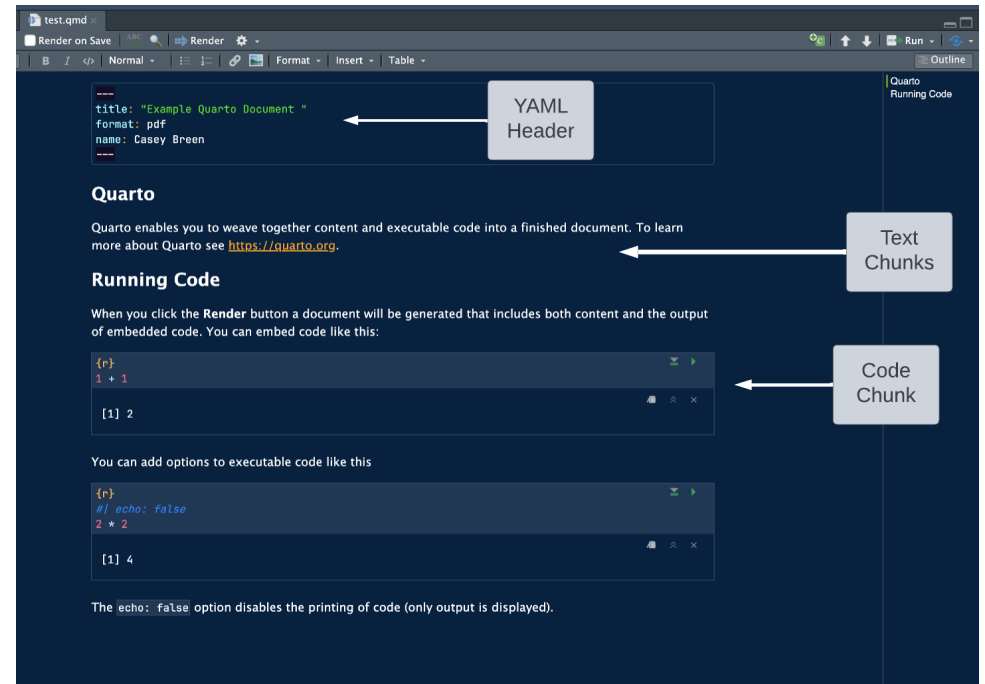


# R Scripts, R-Notebooks

- **Scripts:**
  - Just code
  - Ideal for simple tasks (and multi-script pipelines)
- **Notebooks** (Quarto, R Notebook):
  - Integrated code, text, and outputs (great for documentation!)
  - Interactive
  - We will focus on notebooks

# Quarto Document

- Notebook-Style Layout: Supports interactive code and text chunks.
  - Code Chunks: Segments for code execution
  - Text Chunks: Annotations or explanations in Markdown format.
- Inline Output: Figures and code output display directly below the corresponding code chunk



# Live Coding Session 1: Creating new Quarto file

- I'll demo first; please pay attention

# You turn

- Please create a quarto document
- You can use this document for the rest of this session
- Add a new code chunk
  - click + point: **Insert** → **Executable cell** → **R**
  - macOS: **Cmd** + **Option** + **I**
  - Windows/Linux: **Ctrl** + **Alt** + **I**

# Objects

- Everything in R is an object
  - **Vectors:** Ordered collection of same type.
  - **Data Frames:** Table of columns and rows.
  - **Function:** Reusable code block.
  - **List:** Ordered collection of objects.

```
1  ## Objects in R
2
3  ## Numeric like `1`, `2.5`
4  x <- 2.5
5
6  ## Character: Text strings like `"hello"`
7  y <- "hello"
8
9  ## Boolean: `TRUE`, `FALSE`
10 z <- TRUE
11
12 ## Vectors
13 vec1 <- c(1, 2, 3)
14 vec2 <- c("a", "b", "c")
15
16 ## data.frames
17 df <- data.frame(vec1, vec2)
```

# Functions

- Built-in “base” functions

```
1 ## Functions in R
2 result_sqrt <- sqrt(25)
3 result_sqrt
```

```
[1] 5
```

- Custom, user-defined functions

```
1 # User-Defined Functions: Custom functions
2 my_function <- function(a, b) {
3   return(a^2 + b)
4 }
5
6 my_function(2, 3)
```

```
[1] 7
```

- Functions from packages

# Installing packages

- Packages: pre-built code and functions.
- Packages are generally installed from the Comprehensive R Archive Network (CRAN)

## Install: new packages

```
1 install.packages("tidyverse")
```

## Library: load installed packages

```
1 library(tidyverse)
```

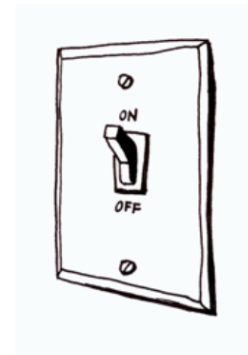
### Installing a package

```
install.packages('my.package')
```



### Loading a package

```
library('mypackage')
```



YaRrr! The Pirates Guide to R. Nathaniel D. Phillips, 2018.



# Running code

- Run code in a quarto document (or script, or R notebook)
  - Exception: install packages, quick checks in console
- To run a single line of code
  - Cursor over line, **Ctrl + Enter** (Windows/Linux) or **Cmd + Enter** (Mac).
- To run a full code chunk (or script)
  - **Ctrl + Shift + Enter** (Windows/Linux) or **Cmd + Shift + Enter** (Mac).

# Your Turn

- Create a new code cell in [quarto](#) document
  - ([insert](#) -> [executable cell](#) -> [R](#))
- Run each line one at a time
- Run the full code chunk

```
1 x <- 1
2
3 y <- (x + 1)^3
4
5 cat("Thank you for attending R session number", x, "!")
```

Thank you for attending R session number 1 !

# Break

10 minute break

# Basic Syntax: assignment

- Use `<-` or `=` for assignment
  - `<-` is preferred and advised for readability
- Formally, assignment means “assign the result of the operation on the right to object on the left”

```
1  ## Add comments
2
3  x <- 7 # assigns 7 to x
4
5  ## Quesiton: what does this do?
6  y <- x <- 25
```

# Basic Syntax: comments

- Use `#` to start a single-line comment
- Include lots of comments when you're writing code

```
1 ## Add comments
2
3 x <- 7 # assigns 1 to x
4 x <- 12
```

# Basic syntax: operators

```
1 ## R as a calculator (# adds a comment)  
2 3 * 3
```

```
[1] 9
```

```
1 ## Division  
2 12/4
```

```
[1] 3
```

```
1 ## Subtraction  
2 100-12
```

```
[1] 88
```

```
1 ## Exponents (10^2)  
2 10 ** 2
```

```
[1] 100
```

# Basic syntax: comparisons

Operator	Symbol
AND	&
OR	
NOT	!
Equal	==
Not Equal	!=
Greater/Less Than	> or <
Greater/Less Than or Equal	>= or <=
Element-wise In	%in%

```
1 ## Logical operators
2
3 10 == 10
```

```
[1] TRUE
```

```
1 9 == 10
```

```
[1] FALSE
```

```
1 9 < 10
```

```
[1] TRUE
```

```
1 "apple" %in% c("bananas", "oranges")
```

```
[1] FALSE
```

```
1 "apple" %in% "bananas" | "apple" %in% "apple"
```

```
[1] TRUE
```

```
1 "apple" %in% "bananas" & "apple" %in% "apple"
```

```
[1] FALSE
```

# Data structures

- There are lots of data structures; we'll focus on **vectors** and **data frames**.
  - **Vectors**: One-dimensional arrays that hold elements of a single data type (e.g., all numeric or all character).
  - **Data Frames**: Two-dimensional tables where each column can have a different data type; essentially a list of vectors of equal length.



# Vectors and data frames

- Vector example

```
1 ## Vector Example
2 vec_example <- c(1, 2, 3, 4, 5)
3
4 print(vec_example)
```

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

- Data frame example

```
1 # Data.frame example
2 example_df <- data.frame(
3   ID = c(1, 2, 3, 4),
4   Name = c("Alice", "Bob", "Charlie", "David"),
5   Age = c(25, 30, 35, 40),
6   Score = c(90, 85, 88, 76)
7 )
```

ID	Name	Age	Score
1	Alice	25	90
2	Bob	30	85
3	Charlie	35	88

# Data types

- Each **vector** or **data frame** column can only contain one data type:
  - **Numeric**: Used for numerical values like integers or decimals.
  - **Character**: Holds text and alphanumeric characters.
  - **Logical**: Represents binary values - TRUE or FALSE.

```
1 ## generate vectors
2 vec <- c(1, 2, 3)
3 vec1 <- c("a", "b", "c")
4
5 ## check type
6 class(vec1)
```

```
[1] "character"
```

```
1 class(vec2)
```

```
[1] "character"
```

# Generating Sequences in R

- Manually write out sequence using `c()`

```
1 ## Basic
2 c(1, 2, 3, 4, 5, 6, 7, 8, 9, 10)
```

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

- Colon operator (`:`), creates sequences with increments of 1

```
1 c(1:10)
```

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

- `seq()` Function: More flexible and allows you to specify the **start**, **end**, and **by** parameters.

```
1 ## seq 1-10, by = 2
2 seq(1, 10, by = 2)
```

```
[1] 1 3 5 7 9
```

# Basic functions

- Function: Input arguments, performs operations on them, and returns a result
- For each of the below functions, what are the:
  - Input arguments?
  - Operations performed?
  - Results?

```
1 ## generate random draws from a standard normal distribution
2 random_draws <- rnorm(n = 5,
3     mean = 0,
4     sd = 1)
5
6 ## find the mean
7 mean(random_draws)
```

```
[1] 0.4747815
```

```
1 ## find the median
2 median(random_draws)
```

```
[1] 0.3997984
```

```
1 ## find the standard deviation
2 sd(random_draws)
```

```
[1] 0.5644327
```

# In-class exercise 1

1. Assign `x` and `y` to take values 3 and 4.
2. Create a new variable `z` as the product of variables `x` and `y`.
3. Write code to calculate the square of 3. Assign this to a variable `three_squared`.
4. Write a logical expression on whether `x` is greater than 10. When might you need to filter data based on a condition?
5. Write a logical expression testing whether `x` is *not* greater than 10.

# Exercise 1 solutions

1. Assign `x` and `y` to take values 3 and 4.

```
1 x <- 3
2 y <- 4
```

2. Create a new variable `z` as the product of `x` and `y`.

```
1 z <- x * y
```

3. Calculate the square of 3 and assign it to a variable called `three_squared`.

```
1 three_squared <- 3^2
```

4. Write a logical expression to check if `three_squared` is greater than 10.

```
1 three_squared > 10
```

```
[1] FALSE
```

5. Write a logical expression to test whether `three_squared` is not greater than 10. Use the `negate` symbol.

```
1 !(three_squared <= 10)
```

```
[1] FALSE
```

# In-class exercise 2

1. Generate vectors containing the numbers 100, 101, 102, 103, 104, and 105 using 3 different methods (e.g., `c()`, `seq()`, `:`). In what scenarios might each method be most convenient?
2. Generate a sequences of all **even** numbers between 0 and 100.
3. Create a descending sequence of numbers from 100 to 1, and assign it to a variable.

# Exercise 2 solutions

1. Generate vectors containing the numbers 100 to 105 using three different methods (`c()`, `seq()`, `:`). Discuss the convenience of each method.

```
1 # Generate a vector using c() method
2 vector_c <- c(100, 101, 102, 103, 104, 105)
3
4 # Generate a vector using seq() method
5 vector_seq <- seq(100, 105, by = 1)
6
7 # Generate a vector using : operator
8 vector_colon <- c(100:105)
```

2. Generate a sequence of all even numbers between 0 and 100.

```
1 # Generate a sequence of all even numbers between 0 and 100
2 even_seq <- seq(0, 100, by = 2)
```

3. Create a descending sequence of numbers from 100 to 1, and assign it to a variable.



# Break

- 10 minutes
- Tea + cake

# Indexing vectors

- Basic indexing, specify position

```
1 vec <- c(1, 2, 3, 4, 5)
2 first_element <- vec[1]
3 third_element <- vec[3]
```

- Conditional indexing, specify position

```
1 vec <- seq(5, 33, by = 2)
2 vec[vec > 25]
```

```
[1] 27 29 31 33
```

# Working with data frames

- Data frames are the most common and versatile data structure in R
- Data frames are structured as rows (observations) and columns (variables)

```
1 test_scores <- data.frame(  
2   id = c(1, 2, 3, 4, 5),  
3   name = c("Alice", "Bob", "Carol", "Dave", "Emily"),  
4   age = c(25, 30, 22, 28, 24),  
5   gender = c("F", "M", "F", "M", "F"),  
6   score = c(90, 85, 88, 92, 89)  
7 )  
8  
9 knitr::kable(test_scores)
```

id	name	age	gender	score
1	Alice	25	F	90
2	Bob	30	M	85
3	Carol	22	F	88
4	Dave	28	M	92
5	Emily	24	F	89

# Working with data frames

- `head()` - looks at top rows of the data frame
- `$` operator - access a column as a vector

```
1 ## print first two rows first row
2 head(test_scores, 2)
```

	id	name	age	gender	score
1	1	Alice	25	F	90
2	2	Bob	30	M	85

```
1 ## access name column
2 test_scores$name
```

```
[1] "Alice" "Bob"    "Carol" "Dave"   "Emily"
```

```
1 ## all rows, columns 1-3
2 test_scores[,1:3]
```

	id	name	age
1	1	Alice	25
2	2	Bob	30
3	3	Carol	22
4	4	Dave	28
5	5	Emily	24

```
1 ## all columns, rows 4-5
2 test_scores[4:5,]
```

	id	name	age	gender	score
4	4	Dave	28	M	92
5	5	Emily	24	F	89

# Subsetting data frames

- Methods:
  - `$`: Single column by name.
  - `df[i, j]`: Row `i` and column `j`.
  - `df[i:j, k:l]`: Rows `i` to `j` and columns `k` to `l`.
- Conditional Subsetting: `df[df$age > 25, ]`.

# Quiz

Which rows and will this return?

```
1 test_scores[1:3, ]
```

- Which rows and which columns will this return?

```
1 test_scores[test_scores$score >= 90, ]
```

# Answers

```
1 test_scores[test_scores$score >= 90, ]
```

	id	name	age	gender	score
1	1	Alice	25	F	90
4	4	Dave	28	M	92

```
1 test_scores[test_scores$score >= 90, ]
```

	id	name	age	gender	score
1	1	Alice	25	F	90
4	4	Dave	28	M	92

# Explore **data frame** characteristics

## Check number of rows

```
1 ## check number of rows (observations)
2 nrow(test_scores)
```

```
[1] 5
```

## Check number of columns

```
1 ## check number of columns (variables)
2 ncol(test_scores)
```

```
[1] 5
```

## Check column names

```
1 names(test_scores)
```

```
[1] "id"      "name"    "age"     "gender"  "score"
```



# Reading in data

## Common Formats

- CSV, Excel, TXT

## Key Functions

- **`read.csv()`**: Read CSV files
  - Faster alternatives: `read_csv` from `tidyverse` and `fread()` from `data.table`
- **`read.table()`**: Read text files
- **`readxl::read_excel()`**: Read Excel files

```
1 ## read in CSV file
2 df <- read.csv("/path/to/your/data.csv")
3 df <- read_csv("/path/to/your/data.csv") ## faster
4
5 ## read in stata file
6 library(haven)
7 data <- read_dta("path/to/file.dta")
```

# In-class exercise 3

1. Generate a sample of 100 observations drawn from a normal distribution with a mean of 10 and a standard deviation of 2. How is this type of random sampling useful in statistical analysis?
2. Calculate the mean of this generated sample. How does this sample mean relate to the population mean (hint: population mean = 10) of the distribution?
3. Calculate the difference between the sample mean and the population mean. Why the discrepancy?
4. Repeat steps 1--3 with a sample of 10,000. Did the difference between the sample mean and the population mean decrease? Will this always be the case?

# Exercise 3 solutions

```
1 # Generate a sample of 1,000 draws from a normal distribution with mean = 10 and sd = 2
2 sample_data_100 <- rnorm(100,
3                           mean = 10,
4                           sd = 2)
5
6 # Calculate the mean of this sample
7 sample_mean_100 <- mean(sample_data_100)
8
9 # Calculate the difference between the mean of the sample and the expected value of the mean
10 difference_100 <- abs(sample_mean_100 - 10)
11
12 difference_100
```

```
[1] 0.09917549
```

```
1 # Calculate the Z-score for the sample mean
2 sample_data_10000 <- rnorm(10000,
3                             mean = 10,
4                             sd = 2)
5
6 # Calculate the mean of this sample
7 sample_data_10000 <- mean(sample_data_10000)
8
9 # Calculate the difference between the mean of the sample and the expected value of the mean
10 sample_data_10000 <- abs(sample_data_10000 - 10)
11
12 sample_data_10000
```

```
[1] 0.0298043
```

# Thank you

- Session tomorrow: 9:30am – 12:30pm
- Homework: please complete all exercises in problem set 1 independently
- Questions: [casey.breen@sociology.ox.ac.uk]  
(casey.breen@sociology.ox.ac.uk)

