Introduction to R Day 1

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Pathway

Intro

Intro to Data and Tools, Overview of R

Foundations

Commenting, Types, Variables, Statements, Data Structures, Packages

RStudio

Desktop/Server Version, Interface, Customisation, R Scripts, Hints/Tips

Workflow

Overview (Collect, Explore, Wrangle, Viz, Outputs), Git(Hub/ea), RMarkdown, RAP, Templates, Style Guide

Wrangle

Tidyverse (dplyr/magrittr), Pipes, Functions (Filter, Mutate, Arrange, etc.), PHS Methods

Explore

Mean, Median, Summary Function, Frequencies/Cross-Tabs

Data Flow

Directories/File Paths, CSVs, SPSS (haven), SMRA/Other Databases

Visualise

Intro to ggplot2, Line Graphs, Bar Plots, Scatterplots, Customisation

Output

Overview of RMarkdown, Shiny, etc.

Review

Overview, Next Steps, Q&A



Introduction – the why

R

- is a **programming language** widely used for *data analysis*, statistics, and *graphics*;
- is open source, available on all major operating systems;
- has the functionality to go from raw data to interactive reports and web apps;
- and it's part of the PHS analytical strategy.





Foundations – the what

Building Blocks

- Basic Data Types how is fundamental data, like numbers and text, represented? This is then the foundations of more complex, composite data types, e.g. tables.
- Variables named storage to track "objects" across our program
- Statements a complete line of code, made of expressions and operators
- Control Flow branching (if statements) and iteration (loops)
- Functions reusable code that *can* take inputs and give outputs

Basic Data Types

- Character (String) e.g. "Hello World!"
- Numeric (Float/Real) e.g. 123.5
- Logical (Boolean) e.g. TRUE

```
typeof("Hello World")
is.numeric(123.5)
print(typeof(TRUE))
```

```
> [1] "character"
> [1] TRUE
> [1] "logical"
```



Type Conversion

```
as.<data type>()
```

- as.character () conversions tend to succeed without fault
- as.numeric () TRUE and FALSE become 1 and 0, character types needs to be formatted correctly
- as.logical () everything except O becomes TRUE for numeric conversions, character can be upper, lower, or proper case versions

```
as.character(123.5)
as.numeric("123.5")
as.logical("False")
```

```
> [1] "123.5"
> [1] 123.5
> [1] FALSE
```



Variables

- Naming letters, numbers, dots `.` or underscores `_` are all okay. However, you can't start with an underscore or number, or a dot then a number. Any existing terminology is also reserved from being used as a variable. Following style guidance is also important.
- Assignment variables are assigned mainly with `<-` but you may also see `=` being used.

```
# Good
Totals
sumOfPatientsUnder60

# Bad
60YearOldPatients
_template
TRUE
```



Operators

Precedence	Operator	Description
1	^	Exponentiation (right to left evaluation)
2	88	Modulus
3	* /	Multiplication, Division
4	+ -	Addition, Subtraction
5	< > <= >= == !=	Comparison Operators (Less Than, More Than, Less Than or Equal To, More Than or Equal To, Equal To, Not Equal To)
6	!	Logical NOT
7	& & &	Logical AND
8	1 11	Logical OR



Knowledge Check

We're going to use an app for a lot of our interactive work, especially today. (Hint: some questions have hint buttons).

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Foundations





Anatomy of a Program

```
# Example 1
hello_world <- "Hello World"
print(hello_world)</pre>
```

> [1] "Hello World"

```
• # Example 1 - comment
```

- hello world variable
- <- assignment operator (alt + -)</pre>
- "Hello World" character (" or ')
- print() function



Style Guide

• Naming – variables and filenames should have meaningful names in snake_case format, preferring all lower case.

Structure

- Space after a comma
- No spaces before or after parenthesis
- Comments to explain code and create sections within the code
- Prefer " over ' for characters

```
# Bad
pts <-c ( 'Al', 'Bert' )

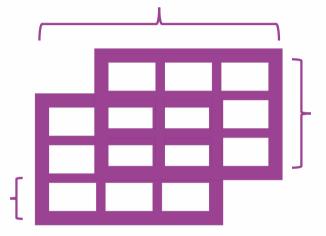
# Good
patients <- c("Al", "Bert")</pre>
```



Foundations – the what

Data Structures

- Vectors
- Lists
- Factors
- Matrices
- Data Frame
- str() provides an overview and description of the data structure





Vectors

contain multiple objects of the *same* basic class

- Create: c(...) or
 vector(<type>, <length>)
- Access: <vector>[<index>]

```
c("a", "c", "f", "b")[1]
c(2, 5, 1, "abc")[3:4]
vector("logical", 4)
```

```
> [1] "a"
> [1] "1" "abc"
> [1] FALSE FALSE FALSE
```



Knowledge Check

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• Data Structures - Vectors





Lists

are a special type of vector that can contain objects of *different* classes, including other lists

```
• Create: list(...)
```

• Sub-list: <list>[<index>]

• Access: <list>[[<index>]]

```
list("abc", 4, FALSE)[1:2]
list(list(2, 3), "abc")[[2]]
> [[1]] > [1] "abc"
```

> [1] "abc"

> [[2]]

> [1] 4



Naming Elements

can be done on vectors and lists during or after creation.

• At creation:

```
c("<name>", = <item>) or
list("<name>", = <item>)
```

• After:

```
names(<object>) <-
c("<name>")
```

```
x <- list("Ch" = "a", "Nm" = 2)
names(x) <- c("Char", "Num")
x$Char</pre>
```

```
> [1] "a"
```



Factors

are used to represent categorical data, with both ordered and unordered variations.

```
• Create: factor(c(...), ordered = <TRUE/FALSE>)
```

```
• Levels: factor(c(...), levels = c(...))
```

```
x <- factor(c("M", "F", "M"),
    levels = c("F", "M"))
x</pre>
```

```
> [1] M F M
Levels: F M
```



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• Data Structures - Factors





Matrices

expand our dimensions with a nrow and ncol arguments, constructed column-wise

- Create: matrix (<data>, nrow =
 <int>, ncol = <int>)
- Access:

```
<matrix>[<row>, <col>]
```

```
x <- matrix(1:6, 2, 3)
x
x[2, 3]
```

```
> [,1][,2][,3] > [1] 6
> [1,] 1 3 5
> [2,] 2 4 6
```



Data Frames

are used to store tabular data, each column contains one variable, each row contains an observation

• Subset: []

• Access: [[]] or \$

```
> name score
> 1 Harry 62
> 2 Sarah 91
```



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• Data Structures - Data Frames





Tibbles

data frames that attempt to make our lives a bit easier. First, load the package: library(tibble)

- Create: tibble ("<name>" = <element(s)>) or coerce an existing data frame with as_tibble (<object>)
- Subset and Access: [] [[]] or \$

```
> name score
> 1 Harry 62
> 2 Sarah 91
```



Foundations – the what

Anatomy of a Function

Functions allow us to bundle code for reuse, taking inputs, doing something and, optionally, providing outputs.

```
mult_2 <- function(x) {
    x <- x * 2
    return(x)
}</pre>
```

```
mult_2(4) > [1] 8
```



Packages

are used to expand the functionality of R with more functions

- Install: install.packages ("<package>")
- Load: library (<package>) or require (<package>) if loading packages as part of functions as it returns a logical value and a warning if the package isn't installed.

install.packages("tidyverse")
library(tidyverse)

- output varies by package, warnings (not errors) are normal. An example would be where a function 'masks' that of another R function.



Control Flow - if statements

Package: dplyr

Load package, library (dplyr)

• if_else (<condition>,
 <true>, <false>)

```
library(dplyr)
x <- 5
if_else(x > 10, TRUE, FALSE)
```

> [1] FALSE



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• Control Flow - If





Control Flow - case statements

Package: dplyr

Load package, library (dplyr)

• case_when (<condition> ~
 <result>)

```
> [1] "LT3" "LT3" NA "Even" NA
```



Knowledge Check

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• Control Flow - Case





Iteration – for loop

allowing us to do the same thing repeatedly with different inputs.

for(<value> in
<sequence>)
{<statement>}



Iteration – loop with purrr

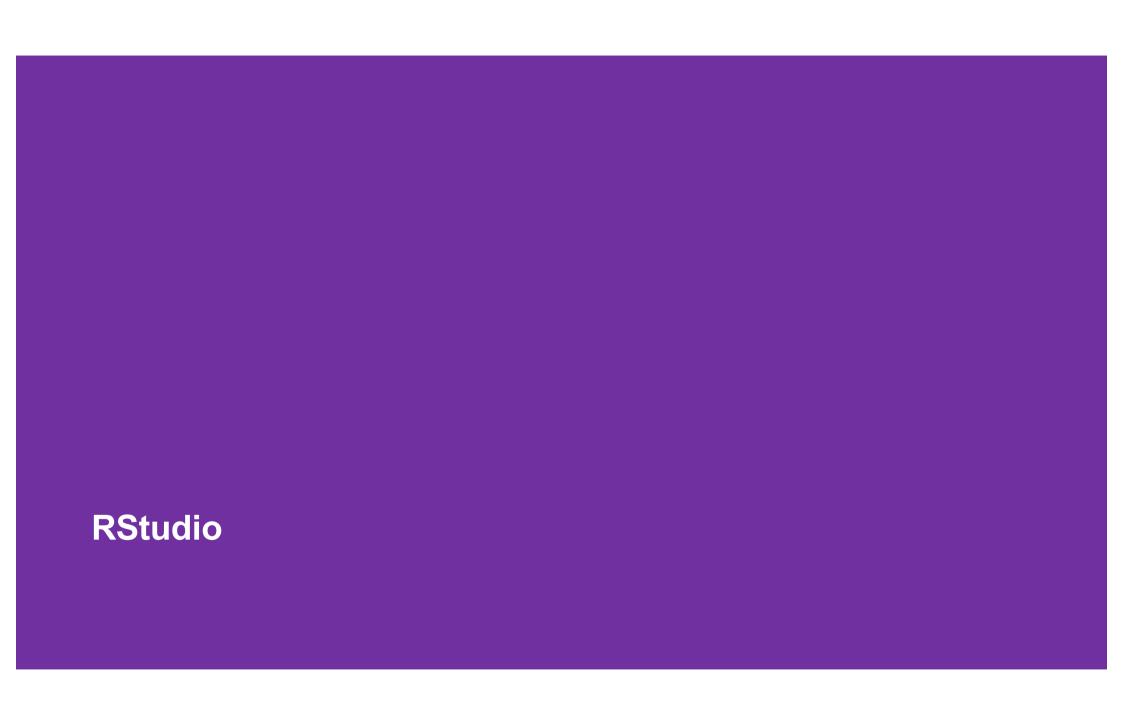
Package: purrr

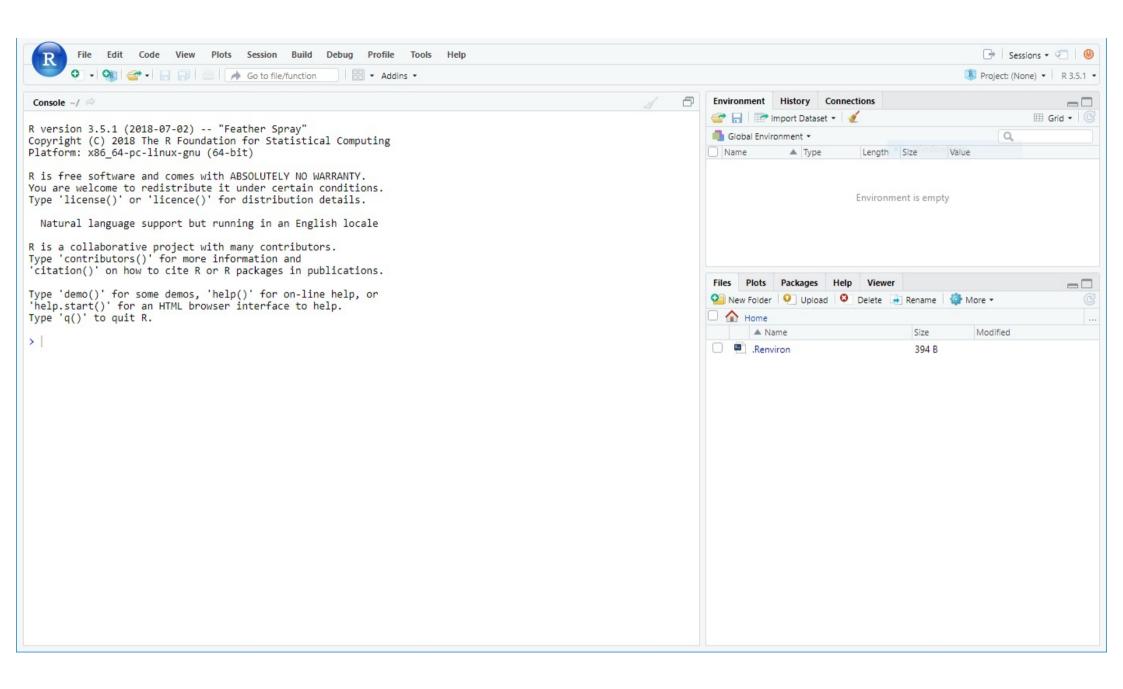
Load: library (purrr)

• map(<object>,
 <function>)

```
library(purrr)
files <- list.files(pattern = ".csv")
all_files <- map(files, read_csv)</pre>
```





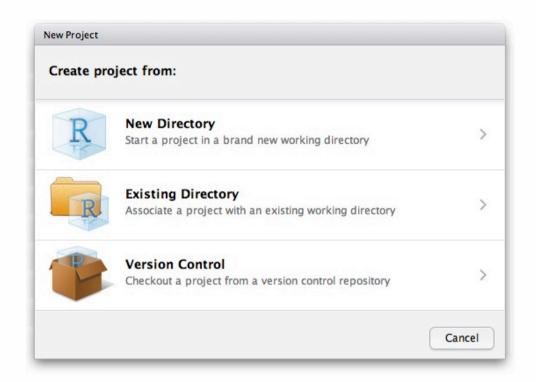


R Projects

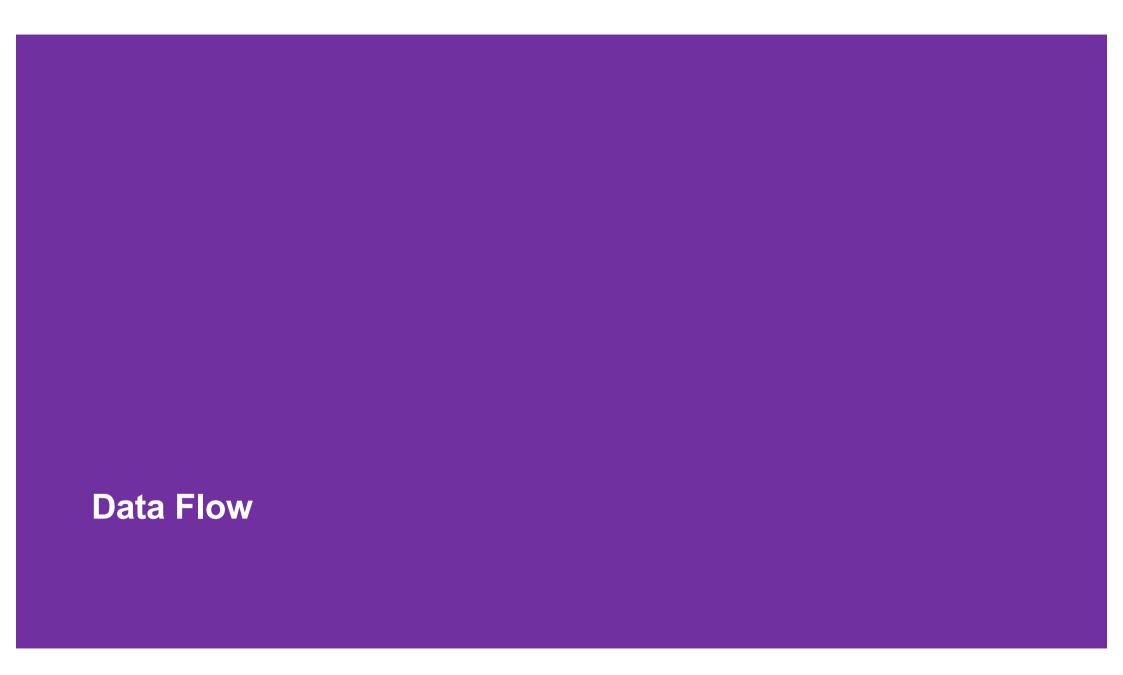
keeps work separate, giving a project its own working directory, workspace, and history.

Opening an .Rproj file will:

- Start a new R session
- Load project specifics and settings
- The project directory is set as the current working directory.
- Create: available in the Projects menu or global toolbar.







Working Directory

- Current: getwd()
- Set new: setwd (<filepath>)
- We can also use the here package, with here ()

RStudio also provides options through the user interface for navigating files and directories.

```
setwd("/home/learnr01/intro_R")
getwd()
here()
```

- > [1] "/home/learnr01/intro R"
- > [1] "/home/learnr01/intro_R"



Read CSV

Package: readr

- Load package, library (readr)
- 2. read csv (<filepath>)
- 3. Check output

```
library(readr)
borders_csv <-
     read_csv("data/Borders.csv")
View(borders_csv)</pre>
```



RDS

Package: readr

Load package, library (readr)

Read

• read rds (<filepath>)

Write

•write_rds (<object>,
 <filepath>)

```
library(readr)

borders_RDS <-
    read_rds("data/borders.rds")

write_rds(borders,
    "data/borders.rds")</pre>
```



Read SPSS

Package: haven

- Load package, library (haven)
- 2. read sav (<filepath>)
- 3. Check output

```
library(haven)
borders_spss <-
     read_sav("data/Borders.sav")
View(borders_spss)</pre>
```



Read Web

The packages/functions used will vary depending on the structure of the data. This example uses a CSV so the process to follow is the same as before.

Package: readr

- Load package, library (readr)
- 2. read_csv (<filepath>)
- 3. Check output

```
library(readr)
hospital_codes <- read_csv("
    https://www.opendata.nhs.sco
t/dataset/[...].csv")
View(hospital_codes)</pre>
```



Open Data – CKAN API

Package: ckanr

- Load package, library (ckanr)
- 2. Set up connection and resource ID
- 3. dplyr::tbl(<src>, <res>)
 %>%
 as tibble()
- 4. Check output

```
library(ckanr)
ckan <-
          src_ckan("https://www.openda
          ta.nhs.scot")
res_id <- "<ID>"
resource <- dplyr::tbl(src =
ckan$con, from = res_id) %>%
          as_tibble()
```



Database (SMRA)

Package: odbc

- Load package, library (odbc)

- 4. Check output



Write CSV

Package: readr

- Load package, library (readr)
- 3. Check output

- the write functions expect a dataframe as the object.





Mean/Median & Summary

• mean () and median () are passed arrays of values (usually from a data frame) to return the mean and median value

```
mean(borders[["LengthOfStay"]])
summary(borders$LengthOfStay)
```

• summary () returns all summary statistics based on a given array (usually from a data frame)

```
> [1] 4.297008
> Min. 1st Qu. Median Mean 3rd Qu. Max
0.000 0.000 1.000 4.297 4.000 458
```

• We access columns using [[]] or \$



Frequencies & Crosstabs

- Frequency: table (<df>\$<col>)
- Crosstab: table (<df>\$<col1>, <df>\$<col2>)
- To add column and row totals, the function addmargins () can be used

addmargins(table(borders\$Hospita
lCode, borders\$Sex))

	1	2	3	Sum	
A210H	1	0	0	1	
в102н	56	100	0	156	
в103н	50	108	0	158	
Sum	11947	13340	2	25289	



Exercise 1

- Read in "Borders.csv" (giving the data frame an appropriate name)
- 2. What are the mean, median, and max values from the LengthOfStay variable? Can you do this in one step?
- 3. Produce a **frequency table** to check the sex variable, save this as an object with an appropriate name
- 4. Export the frequency table as a csv file. You'll need to use as .data.frame()







Project

- 1. Get familiar with RStudio and change some settings to your preference.
- 2. Create a new script and import some open data (using the API or .csv).
 - e.g. COVID-19 Daily Case Trends
- 3. Produce some summary statistics and a frequency table.





Getting Help

- Vignettes (Help) / `?<function>`
- Google / Stack Overflow tag queries "[r] & [tidyverse]"
- R User Group Teams Technical Queries
- Transforming Publishing



