Introduction to R_{v2.0}

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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

Data





R

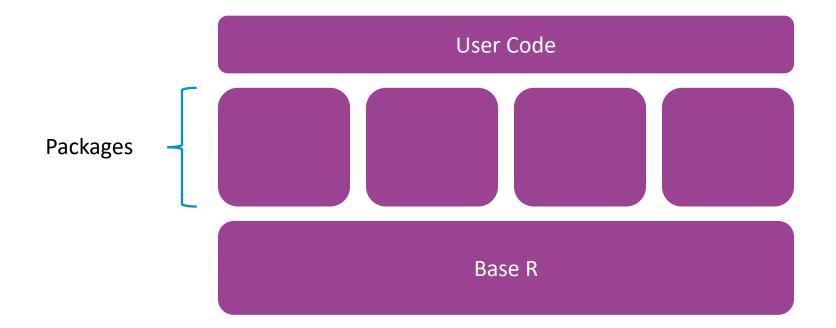
- is a **programming language** widely used for *data analysis*, statistics, and *graphics*;
- is open source, available on all major operating systems;
- has the functionality to go from raw data to interactive reports and web apps;
- and it's part of the PHS analytical strategy.





Foundations – the what

R Structure





Input

```
# 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



Basic Data Types

- Character e.g. "Hello World!"
- Numeric (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

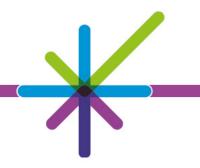
```
as.character(123.5)
as.numeric("123.5")
as.logical("False")
```

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



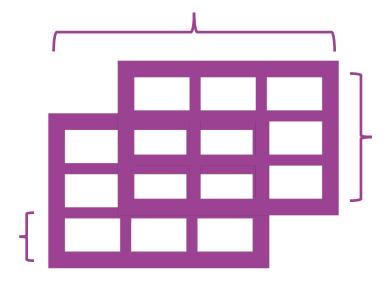
Operators

Precedence	Operator	Description		
1	^	Exponentiation (right to left evaluation)		
2	%%	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		



Data Structures

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





Vectors

- contain objects of the *same* basic class
- Create: c(...) or vector(<type>, <length>)
- Access: <vector>[<index>]

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

```
> [1] FALSE FALSE FALSE FALSE
> [1] "a"
> [1] "5"
```



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"
> [[2]]
> [1] 4
```

```
> [1] "abc"
```



Naming List Elements

• can be done during or after creation.

• At creation:

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

• After:

```
names(<list>) <-
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

```
• Create: factor(c(...))
```

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

```
x <- factor(c(1, 2, 1, 2, 2, 1),
    levels = c(2, 1))
x</pre>
```

```
> [1] 1, 2, 1, 2, 2, 1
Levels: 2 1
```



Matrices

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

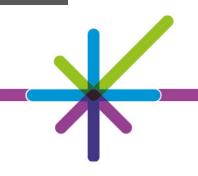
```
• Create: matrix (<data>, nrow =
<int>, ncol = <int>)
```

Access:

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

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

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



Data Frames

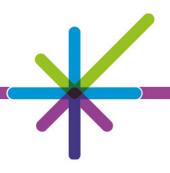
 are used to store tabular data, each column contains one variable, each row contains an observation

```
• Create: data.frame("<name>" = <element(s)>)
```

• Subset: []

• Access: [[]] or \$

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



Packages

 are used to expand the functionality of R with additional functions

• Install:

install.packages("<package>")

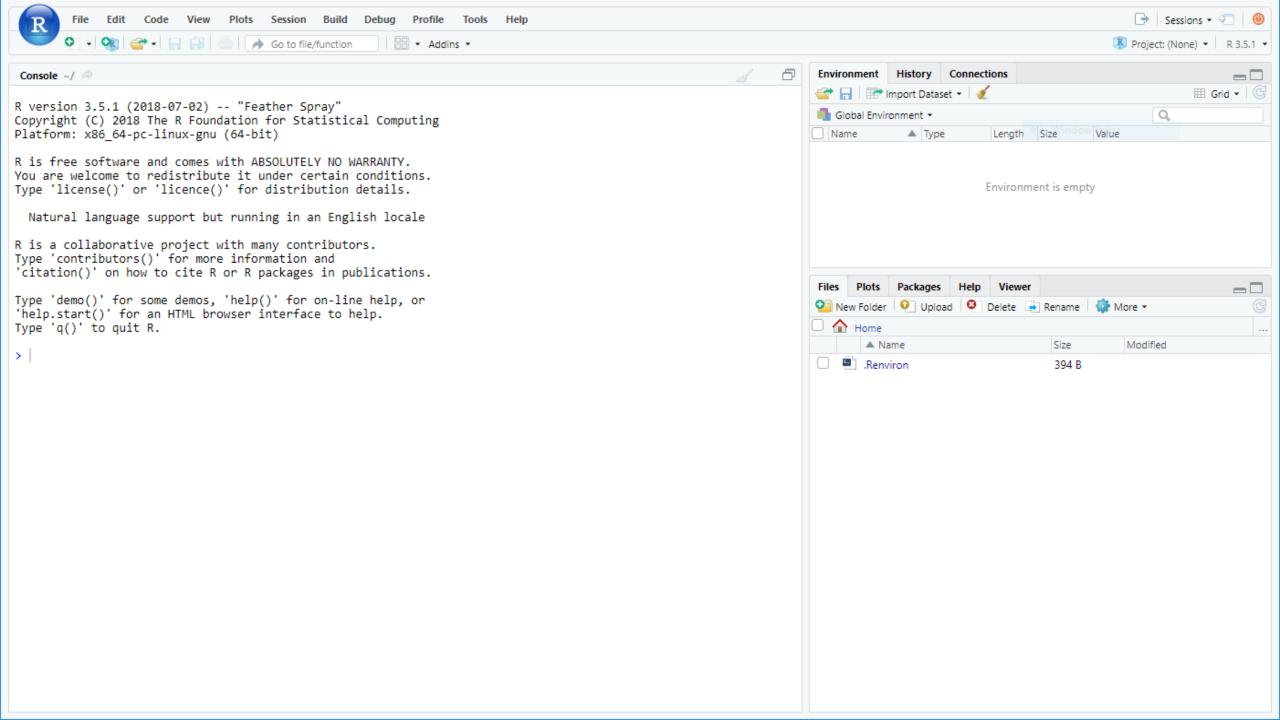
• Load: library(<package>) or require(<package>)

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

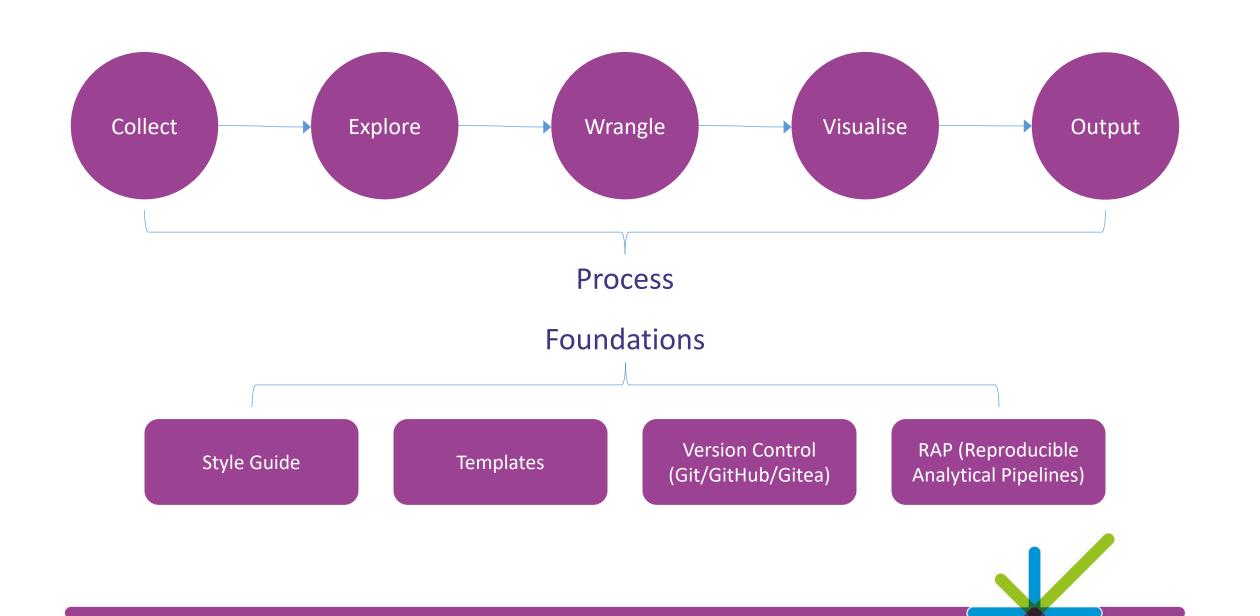
- output varies by package, warnings (not errors) are normal. An example would be where a function 'masks' that of another R function.



RStudio



Workflow



Data Flow

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 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>
```

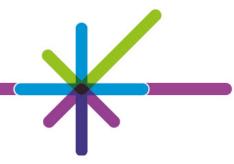


Read SPSS

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



Database (SMRA)

Package: odbc

- Load package, library (odbc)

- 4. Check output



RDS

Read

- 1. readRDS (<filepath>)
- 2. Check output

Write

- 2. Check output

```
borders_RDS <-
    readRDS("data/borders.rds")

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



Write CSV

Package: readr

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



Explore

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

- 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?
- 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





Wrangle

Tidyverse

- is a suite of packages for data exploration, manipulation, and visualisation; it's best practice to utilise these where possible;
- functions have a consistent format, i.e. function (data, task);
- gives us the package dplyr





dplyr

library(dplyr)

- is a grammar of data manipulation, providing a set of "verbs" to help solve most data manipulation challenges
- filter()
- mutate()
- arrange()
- select()
- group_by()

- summarise()
- count()
- rename()
- recode()





Pipe Operator

- %>% is used to link functions together, passing the previous to the next
- Using the pipe operator makes
 R code more readable and
 prevents extensive parenthesis
 building up with multiple
 function calls
- Readable as "and then"
- Shortcut: (ctrl + shift + M)

arrange (Dateofbirth)



Filter

```
filter(<data>, <logical
expression>)
```

• picks cases based on their values



Mutate

```
mutate(<data>, <new_col> =
<expression>)
```

 adds new variables that are functions of existing variables



Arrange

```
arrange(<data>,
<variables>)
```

- changes the ordering of rows
- desc() to sort in descending order

```
# sort by Hospital Code
borders %>%
    arrange(HospitalCode)
```



Select

```
select(<data>,
<expression>)
```

- picks variables based on their names
- prepend "-" to a variable to remove

```
# remove Postcode
borders %>%
    select(-Postcode)
```



Exercise 2

- 1. Read in "Borders.csv" (giving the data frame an appropriate name)
- 2. What patients had a LengthOfStay of between 2 and 10 days?
- 3. Which of these patients were under Specialty E12 or C8?
- 4. Remove all columns other than URI, Specialty, and LengthOfStay
- 5. Store this data in an appropriately named data frame ordered by LengthOfStay





Group By

```
group_by(<data>,
<col_name>)
```

- groups variables to perform operations
- This doesn't visibly affect the data, but we can see the output shows the grouping. We can then perform other operations on the groups.

```
# sort by Hospital Code
borders %>%
    group_by(HospitalCode)
```

```
> ...
# Groups: HospitalCode [48]
...
```



Summarise

```
summarise(<data>, <name> =
<expression>)
```

 reduced multiple values down to a single summary

```
# avg length of stay by hospital
borders %>%
    group_by(HospitalCode) %>%
    summarise(mean_los =
        mean(LengthOfStay))
```



Count

```
count(<data>, <variables>)
```

 useful for running frequencies, this calls group_by() and produces counts for a specified column

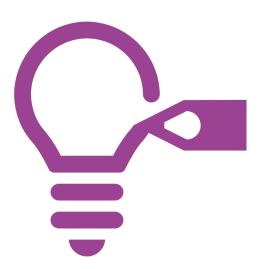
• sort by descending order using sort = TRUE as an argument

```
# counts of specialty
borders %>%
    count(Specialty, sort = TRUE)
```



Exercise 3

- 1. Read in "Borders.csv" (giving the data frame an appropriate name)
- 2. What is the earliest admission date by specialty?
- 3. What is the latest discharge date by specialty?
- 4. What are the number of admissions per hospital, per specialty?





Rename

```
rename(<data>, <new_name> =
<existing_name>)
```

• renaming specific columns in a data frame



Recode

```
mutate(<col> = recode(<col>,
  <existing_code> =
  <new_code>))
```

- for changing values within a column
- works best when used with mutate()



Exercise 4

- 1. Select the URI, Specialty, and Dateofbirth columns from the borders data and save to a new data frame.
- 2. Arrange this new data in ascending order by Specialty and check the results.
- 3. Extract the records with a missing Dateofbirth (hint: ?filter)
- 4. Finally, recode Specialty "A1" to be "General Medicine"





Joining Tables

```
<type>_join(<data1>,
<data2>, by =
<common_variable>)
```

 for merging data by matching together using common variable(s)

```
# merge baby data
baby5 <- read_csv("data/Baby5.csv")
baby6 <- read_csv("data/Baby6.csv")
baby_joined %>%
    left_join(baby5, baby6, by =
        c("FAMILYID", "DOB"))
```

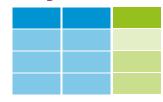


Join Types

left_join()

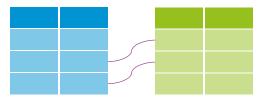
all rows from the 'left', any
 matches from the 'right'





inner_join()

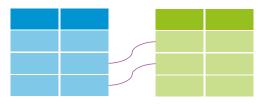
rows of matched fields from data sets





right_join()

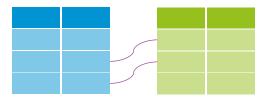
all rows from the 'right', any
 matches from the 'left'

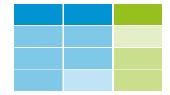




full_join()

all rows, na for non-matched fields





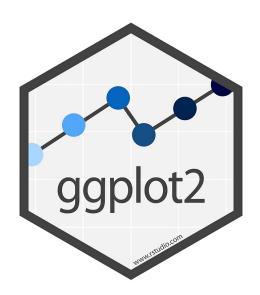


Visualise

ggplot2

library(ggplot2)

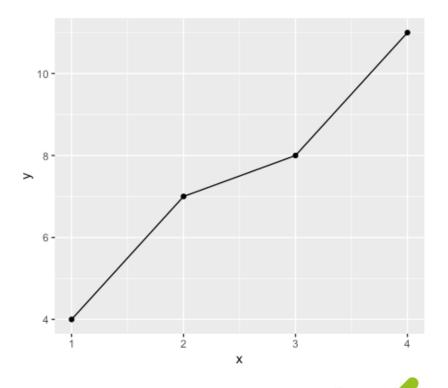
- is a system for creating graphics to visualise data
- Begin with a call to ggplot(), supplying the data and aesthetic mappings using aes()
- Then add layers, scales, coords, and facets with +



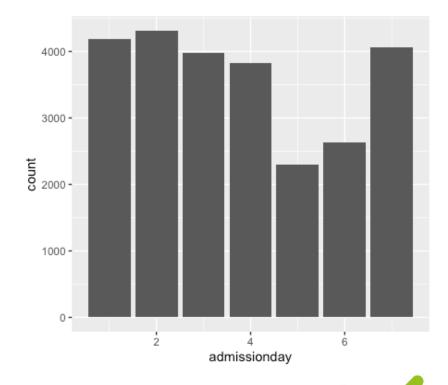


Line Graph

```
# Define data
data <- data.frame(</pre>
          x = c(1, 3, 2, 4),
          y = c(4, 8, 7, 11))
# Plot the data
line_plot <- ggplot(data,</pre>
                  aes(x, y)) +
     geom_line() +
     geom_point()
```

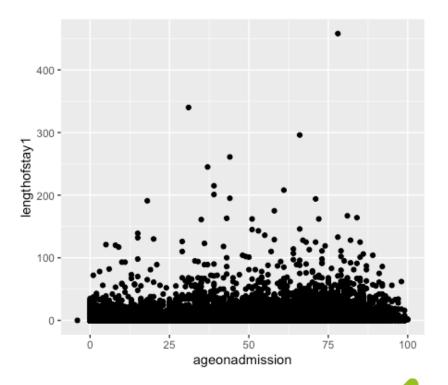


Bar Plot



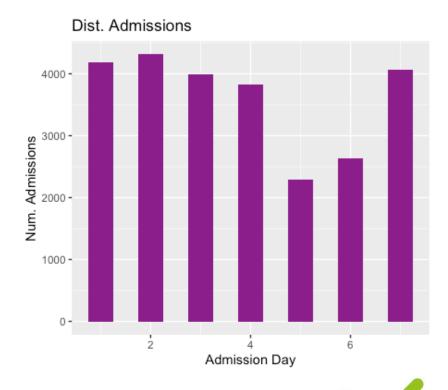
Scatter Plot

```
# Define data
data <- read csv("</pre>
data/BORDERS (inc Age).csv")
# Plot the data
scatter plot <-</pre>
ggplot(data, aes(x =
     ageonadmission,
    y = lengthofstay1)) +
  geom_point()
```



Customisation

```
# bar plot colour and width
bar plot <- ggplot(data) +</pre>
     geom\ bar(aes(x =
      admissionday),
      fill = "magenta4",
      width = 0.5) +
# add titles
     xlab("Admission Day") +
     ylab("Num. Admissions") +
     ggtitle("Dist. Admissions")
```



Exercise 5

- 1. Read in "Borders.csv" (giving the data frame an appropriate name)
- 2. Filter data to show only records from HospitalCode "B109H"
- 3. Create a histogram for patient's LengthOfStay, check the output
- 4. Add axis labels and a title to your histogram
- 5. Save the plot (as a PNG) to the plots folder in your working directory





Output

rmarkdown

- allows for the production of fully reproducible documents, with R (and other language) 'chunks' included
- supports many static and dynamic outputs, including PDF, MS Word, HTML, books, presentations, etc.
- is used in RAP publications, infographics, and more within PHS





shiny

- is for building interactive web apps through R;
- it provides an interactive approach to sharing data and analyses and can be hosted online or embedded in RMarkdown documents
- is used for some PHS publications and in development for more





Review

Getting Help

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





Next Steps

- Embed your new knowledge and skills!
- Expand your knowledge and skills with related technologies (e.g. git)
- Take R Further look at other training opportunities (phsmethods)



