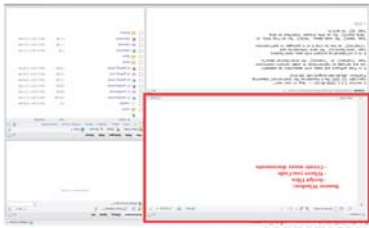


Ways to Use R

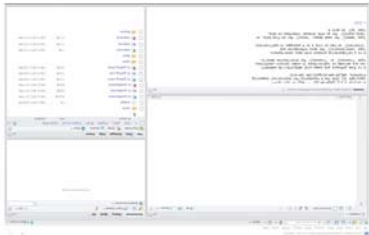


RStudio

<https://rshin.github.io/rshin/2015/01/01/rshin-2015-01-01/>

Adam J Sullivan

Intro to R Programming for Biostatistics
Day 1 - Getting Started with R



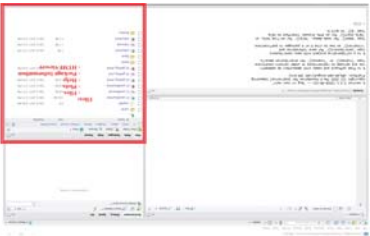
RStudio



Introduction to R
Presented by:

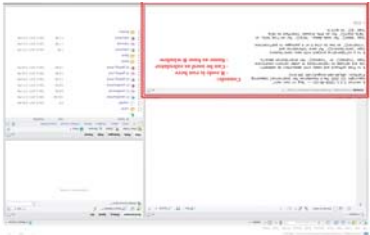


Base R

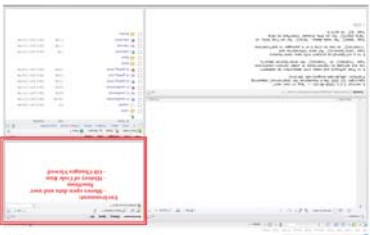


The most simple procedures that we can do in R is using R as a calculator. For example:

# Addition	5+4	## [1] 9
# Subtraction	124 - 26.82	## [1] 97.18



Operator	Description
+	Addition
-	Subtraction
*	Multiplication
/	Division
^	Exponentiation



R as a Calculator

	\$CC^{\circ}\$ [T] ##
	8/SE UNOVLST #
	0C [T] ##
	DxS UNLPLDICATION #

as well.

R works simply as a calculator but also can be used for more advanced operations

More Math in R

# Exponential	exp(1.5)	# [1] 4.481689
# Exponential	exp(1.5)	
# [1] 1.732851		
# Exponential	exp(1.7)	

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More Math in R

ε [1] ==
(0000100101 01 base 10) a
5'1 [1] ==
(000100101 0 base 10) a

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Logical Operators

Once we have used some basic arithmetic operators we move into logic.

Operator	Description
<	Less Than
>	Greater Than
<=	Less Than or Equal To
>=	Greater Than or Equal To
=	Exactly Equal To
!=	Not Equal To
!a	Not a
abb	a AND b

We can then see an example of this:

Logic in R Example

## <- c(1:12)	
# save off the data	
# follows us 1 2 3 10 11 12	
a	
removing R do this	
## [1] 1 2 3 4 5 6 7 8 9 10 11 12	
	309
## [1] FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE	TRUE
## [1] TRUE	

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Logic in R Example

a0
[1]
TRUE FALSE FALSE FALSE FALSE FALSE FALSE TRUE TRUE TRUE
a9 a8
[1]
FALSE
a9 a8 a7
[1]
TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE
a[0:9] a6
[1]
1 2 3 10 11 12

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```
help("<")
help("<")
# Help for words (same use of quotations above work
?"<")
```

Further Help Function Use

We can also get help on characters and words by placing them in quotations

```
help(seq)
# help() function with seq as argument
# Shortcut for help() is ?
?seq
```

Help Function Example

Description	R Symbol
Square Root	sqrt
Floor	floor
Ceiling	ceiling
Log	log
Exponential function, e^x	exp
Factorial, !	factorial

We also have access to a wide variety of mathematical functions that are already built into R.

Math Functions in R

Description	R Symbol	Example
Comment	#	# This is a comment
Assignment	<-	x <- 5
Assignment	=	x = 5
Concatenation operator	c()	c(1,2,4)
Modulus	%/%	seq(a,b) %/% 6
Sequence from a to b by h	seq	0:3
Sequence Operator	:	

Some other operators we may want to use are listed below

Further Operators

The help() Function

- To get online help within an R session we use the help() function.
- For example if we want to create a sequence and know that we can use the function seq() but are unsure of the arguments

```
help.search("normal")
```

The other help items are great if you know what function you are looking for. Many times we do not know exactly what we are looking for and need to do a more comprehensive search to find a proper function.

The help.search() Function

```
example(perp)
```

- We can then see numerous examples that R has run for us.

- The benefit of this command comes when you are interested in seeing examples of graphics, where just seeing the command and not the final product may not be as intuitive for us

The example() Function

```
example(seq)
```

- Many times we just need to see some examples rather than read the entire documentation of a function or command.

- In this situation we would use the example() function

The example() Function

- R has a wealth of data built in.
- We can use data() function to find it

Built in Data

- Built in Data
- .txt, .xls,
- SPSS, SAS, Stata
- Web Scraping
- Databases

We use

Ways to get Data into R?

Importing Data into R

data()
- Specific packages data
data(package="tidy")

- List all Datasets

Built in Data

- Many files are separated by delimiters.
- Common Ones are
 - comma (,)
 - semi-colon (;)
 - tab (\t)
- We can use various functions to read these files in.

- There are many packages out there which handle all of these things.
- We will stick to using the tidyverse packages.
- This will provide consistency with all we do.

Delimited Files

- In the third session we will use the following functions in practice:
 - read.csv()
 - read.delim()

Reading Delimited Files

- readr is a collection of many functions
 - read_csv(): comma separated (CSV) files
 - read_tsv(): tab separated files
 - read_delim(): general delimited files
 - read_fwf(): fixed width files
 - read_table(): tabular files where columns are separated by white-space.
 - read_log(): web log files
 - readxl reads in Excel files.

readr in Tidyverse

- R can read files from many other software types.

Importing From Other Software

Enter Haven Package

- Haven is part of tidyverse.
- It contains the functions to read many different files.
- It can also write to those same data types.

```
read_sav(file, user_na = FALSE)
read_por(file, user_na = FALSE)
write_sav(data, path)
read_spss(file, user_na = FALSE)
```

For SPSS

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For SAS

```
read_sas(data.file, catalog.file = NULL, encoding = NULL)
write_sas(data, path)
```

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Data Objects in R

For Stata

```
read_dta(file, encoding = NULL)
read_stata(file, encoding = NULL)
write_dta(data, path, version = 14)
```

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Now That we Can Import Data

- We begin with a look at different kinds of data
- **Booleans:** Direct binary values: TRUE or FALSE in R.
 - **Integers:** Whole numbers or number that can be written without fractional component, represented by a fixed-length block of bits.

- Data Objects**
- **Characters:** fixed length block of bits with special coding.
 - **Strings:** Sequence of characters.
 - **Floating Point Numbers:** a fraction times an exponent, like 1.34×10^7 , however in R you would see 1.34e7.
 - **Missing:** Na, NaN, ...

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Examples of Type

is.na(7)	
## [1] FALSE	
is.na(7/0)	
## [1] FALSE	

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- Finding Type of Data**
- With types of data, R, has a built in way to help one determine the type that a certain piece of data is stored as, these consist of the following functions:
- **typeof()** this function returns the type
 - **is.type()** functions return Booleans for whether the argument is of the type **type**
 - **as.type()** functions try to change the argument to type **type**

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Examples of Type

typeof(7)	
## [1] "double"	
is.numeric(7)	
## [1] TRUE	

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Examples of Type

is.character(7)	
## [1] FALSE	
is.character("7")	
## [1] TRUE	

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Coercing Data Types

as.character(5/6)
[1] "0.8333333333333333"
as.numeric(as.character(5/6))
[1] 0.8333333

Equality of Data

6*as.numeric(as.character(5/6))
[1] 5
5/6 == as.numeric(as.character(5/6))
[1] FALSE

What Happened?

- What we can see happening here is a problem in the precision of what R has stored for a number.
- This can also occur when performing arithmetic operations on values as well.

all.equal() Function

When comparing numbers that we have performed operations on it is better to use the all.equal() function.

Further Tries at Equality

0.45 == 3*0.15
[1] FALSE
0.45-3*0.15
[1] 5.551115e-17
0.4 - 0.1 == 0.5 - 0.2
[1] FALSE

A Closer Look

5/6 - as.numeric(as.character(5/6))
[1] 3.330669e-16

[1] TRUE
a11.equal(0.45, 3*0.15)
[1] TRUE
a11.equal(0.4-0.1, 0.5-0.2)
[1] TRUE

Example

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x <- c(1,5,2, 6)
[1] 1 5 2 6
is.vector(x)
[1] TRUE

What we have used here is the concatenation operator which takes the arguments and places them in a vector in the order in which they were entered.

Creating Vectors

y <- c(1,6,4,8)
[1] 2 11 6 14

- We can do arithmetic with vectors in a similar manner as we have with integers.
- When we use operators we are doing something element by element or "elementwise."

Vector Arithmetic

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Vectors in R

What is a Vector?

- The fundamental data type in R is the vector.
- A vector is a sequence of data elements of the same type.

x*y
[1] 1 30 8 48
x/y
[1] 1.0000000 0.8333333 0.5000000 0.7500000
x %>% y
[1] 0 5 2 6

It is important to remember what happens when we consider an "elementwise" operation

Elementwise

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- Intuition would make us think that we could not perform this operation when the length of both vectors are not the same.
- However what R does is it rewrites x such that we have $x <- c(1, 5, 2, 6, 1, 5)$.
- This is called recycling. When R makes the shorter vector longer by repeating elements in the order they are listed in.

Recycling

<pre>## [1] 2 7 8 14 10 15</pre>
<pre>## Warning in x + z: longer object length is not a multiple of shorter object length</pre>
<pre>x+z</pre>

Recycling Example

- We do have to be careful when performing arithmetic operations on vectors.
- There is a concept called recycling and this happens when 2 vectors do not have the same length

Recycling

<pre>## [1] 4</pre>
<pre>length(x)</pre>
<pre>## [1] 4</pre>
<pre>length(y)</pre>
<pre>## [1] 4</pre>
<pre>length(z)</pre>
<pre>## [1] 6</pre>

Functions on Vectors

- There are various functions that we can run over a vector and as we continue on we will learn more about these functions.
- One of the simplest functions can help us with knowing information about Recycling that we encountered before. This is the `length()` function.

Functions on Vectors

<pre>x+z</pre>
<pre>## Warning in x + z: longer object length is not a multiple of shorter object length</pre>
<pre>## [1] 2 7 8 14 10 15</pre>
<pre>c(1, 5, 2, 6, 1, 5) + z</pre>
<pre>## [1] 2 7 8 14 10 15</pre>
<pre>## [1] 2 7 8 14 10 15</pre>

Recycling

- ***sd*** (***sd***) and ***var*** find the standard deviation and variance of a vector respectively.
 - ***median*** finds the median of a vector.
 - ***mean*** finds the arithmetic mean of a vector.
- in other classes.
- There are various other functions that can be run on vectors some you have seen

Built in Functions

<code>any(x>3)</code>
<code>## [1] TRUE</code>
<code>all(x>3)</code>
<code>## [1] FALSE</code>

Functions on Vectors

- We can use ***any()*** and ***all()*** in order to answer logical questions about elements
- Then length vector is very important with the writing of functions which we will get to in a later unit.

Functions on Vectors

- ***min()*** and ***max()*** find the minimum and maximum of a vector respectively.
- ***summary()*** returns a 5 number summary of the numbers in a vector.

Built in Functions

- ***x[]*** is a way to call up a specific element of a vector.
 - ***x[1]*** is the first element.
 - ***x[3]*** is the third element.
 - ***x[-3]*** is a vector with everything but the third element.
- We can call specific elements of a vector by using the following:

Vector Indexing

<code>which(x>3)</code>
<code>## [1] 2 4</code>

- The ***which()*** function will tell us the elements that are.
- For example, above we asked if any elements in vector ***x*** were greater than 3.
- In,
- Some functions help us work with the data more to return values in which we are interested

which() Function

Working with Vectors

first elements to make sure we have what we need	
15()	
## [1] "x" "y" "z"	
x[3]	
## [1] z	
x[-3]	
## [1] 1 5 6	

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Replacing Values

- We have seen how to subtract an element from a vector but we can use the same information to place it back in.
- We start with the same vector x that we started with.

x	
## [1] 1 5 2 6	
x[-3]	
x	
## [1] 1 5 6	

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Inserting Values

We can then add the original element back in

x	
x <- c(x[1:2], z, x[3])	
## [1] 1 5 2 6	

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Creating Vectors

Within R, we have not defined any y yet so it will not create a vector in this manner. There are multiple ways of creating vectors.

y1	
y1[1] <- 3	
y1[2] <- 15	
y1	
## [1] 3 15	

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Creating Vectors

There are multiple ways we can create a vector but we must let R know what we are doing

y[1]	
y[2] <- 15	

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Indexing with Booleans

- Before we used any(x > 3) and which(x > 3).
- Now we can see not only their position in the vector, but indexing allows us to return their values.

x[x > 3]	
## [1] 5 6	

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- With vectors it can be important to assign names to the values.
- Then when doing plots or considering maximum and minimums, instead of being given a numerical place within the vector we can be given a specific name or what that value represents.
- For example say that vector x represents the number of medications of 4 unique patients. We could then use the *names()* function to assign names to the values.

Naming Vector Elements

y7 <- rep(c(1,2,3),3)
[1] 1 2 3 1 2 3 1 2 3

Creating Vectors

y5 <- 3:10
y6

Aside from these ways to create the specific vector of (3,15) we can create vectors a couple more ways as well

Creating Vectors

y2 <- c(3,15)
[1] 3 15
y3 <- seq(from=3, to=15, length=2)
[1] 3 15
y4 <- seq(from=3, to=15, by=12)
[1] 3 15
---class #fid

Creating Vectors

x
[1] 1 5 2 6
names(x)
M.L.
names(x) <- c("Patient A", "Patient B", "Patient C", "Patient D")
x
Patient A Patient B Patient C Patient D
1 5 2 6