

R: A Hitchhikers Guide to Reproducible Research

- We built this software on base R code

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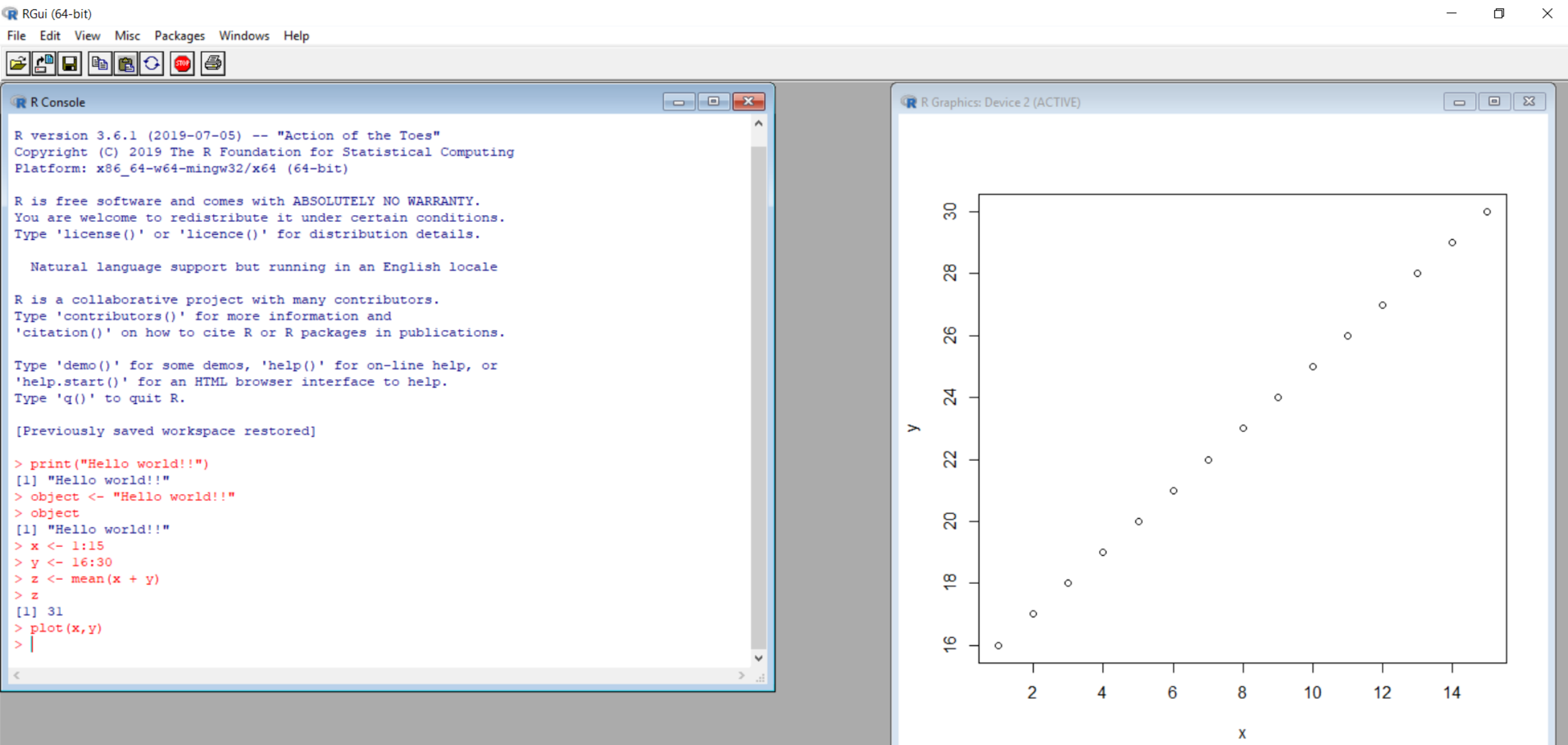
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To understand R, remember the following

- Everything that exists is an object
- Everything that happens is a function

R user interface versus RStudio



```

22 gather(sample, expression, GO.05:U0.3) %>%
23 separate(sample, c("nutrient", "rate"), sep = 1, convert = TRUE) %>%
24 mutate(nutrient = plyr::revalue(nutrient, nutrient_names)) %>%
25 filter(!is.na(expression), systematic_name != "")
26
27 # Plot the clean data
28
29 leu_plot <- cleaned_genes_tbl %>%
30 filter(BP == "leucine biosynthesis") %>%
31 ggplot(mapping = aes(x = rate, y = expression, color = nutrient))
32 ) +
33 geom_point() +
34 geom_smooth(method = "lm", se = FALSE) +
35 facet_wrap(~ name)
36
37 leu_plot

```

Code editor

```

37. Does this work on your system?
/R_Users_Workshop/T_L_workshop/reproducible-workflows/reproducible-workflows_2019/
cleaned_genes_tbl <- cleaned_genes_tbl %>%
+ separate(NAME, c("name", "BP", "MF", "systematic_name", "number"),
+           sep = "\\|\\|\\|\\|") %>%
+ mutate_at(vars(name:systematic_name), funs(trimws)) %>%
+ select(-number, -GID, -YORF, -GWEIGHT) %>%
+ gather(sample, expression, GO.05:U0.3) %>%
+ separate(sample, c("nutrient", "rate"), sep = 1, convert = TRUE) %>%
+ mutate(nutrient = plyr::revalue(nutrient, nutrient_names)) %>%
+ filter(!is.na(expression), systematic_name != "")
> leu_plot <- cleaned_genes_tbl %>%
+ filter(BP == "leucine biosynthesis") %>%
+ ggplot(mapping = aes(x = rate, y = expression, color = nutrient))
+ ) +
+ geom_point() +
+ geom_smooth(method = "lm", se = FALSE) +
+ facet_wrap(~ name)
> leu_plot
>

```

R console

Environment History Connections Git

Global Environment

Name	Type	Length	Size	Value
cleaned_ge...	tbl_df	7	11.3 ...	198430 obs. of 7 ...
leu_plot	gg	9	17.7 ...	List of 9
nutrient_n...	character	6	984 B	Named chr [1:6] "Gl...

Workspace



Files/Plots/Help

Base R

Cheat Sheet

Getting Help

Accessing the help files

?mean

Get help of a particular function.

help.search('weighted mean')

Search the help files for a word or phrase.

help(package = 'dplyr')

Find help for a package.

More about an object

str(iris)

Get a summary of an object's structure.

class(iris)

Find the class an object belongs to.

Using Packages

install.packages('dplyr')

Download and install a package from CRAN.

library(dplyr)

Load the package into the session, making all its functions available to use.

dplyr::select

Use a particular function from a package.

data(iris)

Load a built-in dataset into the environment.

Working Directory

getwd()

Find the current working directory (where inputs are found and outputs are sent).

setwd('C://file/path')

Change the current working directory.

Use projects in RStudio to set the working directory to the folder you are working in.

Vectors

Creating Vectors

c(2, 4, 6)	2 4 6	Join elements into a vector
2:6	2 3 4 5 6	An integer sequence
seq(2, 3, by=0.5)	2.0 2.5 3.0	A complex sequence
rep(1:2, times=3)	1 2 1 2 1 2	Repeat a vector
rep(1:2, each=3)	1 1 1 2 2 2	Repeat elements of a vector

Vector Functions

sort(x)	rev(x)
Return x sorted.	Return x reversed.
table(x)	unique(x)
See counts of values.	See unique values.

Selecting Vector Elements

By Position

x[4]	The fourth element.
x[-4]	All but the fourth.
x[2:4]	Elements two to four.
x[-(2:4)]	All elements except two to four.
x[c(1, 5)]	Elements one and five.

By Value

x[x == 10]	Elements which are equal to 10.
x[x < 0]	All elements less than zero.
x[x %in% c(1, 2, 5)]	Elements in the set 1, 2, 5.

Named Vectors

x['apple']	Element with name 'apple'.
-------------------	----------------------------

Programming

For Loop

```
for (variable in sequence){  
  Do something  
}
```

Example

```
for (i in 1:4){  
  j <- i + 10  
  print(j)  
}
```

If Statements

```
if (condition){  
  Do something  
} else {  
  Do something different  
}
```

Example

```
if (i > 3){  
  print('Yes')  
} else {  
  print('No')  
}
```

While Loop

```
while (condition){  
  Do something  
}
```

Example

```
while (i < 5){  
  print(i)  
  i <- i + 1  
}
```

Functions

```
function_name <- function(var){  
  Do something  
  return(new_variable)  
}
```

Example

```
square <- function(x){  
  squared <- x*x  
  return(squared)  
}
```

Reading and Writing Data

Also see the **readr** package.

Input	Output	Description
df <- read.table('file.txt')	write.table(df, 'file.txt')	Read and write a delimited text file.
df <- read.csv('file.csv')	write.csv(df, 'file.csv')	Read and write a comma separated value file. This is a special case of read.table/write.table.
load('file.RData')	save(df, file = 'file.Rdata')	Read and write an R data file, a file type special for R.

Conditions	a == b	Are equal	a > b	Greater than	a >= b	Greater than or equal to	is.na(a)	Is missing
	a != b	Not equal	a < b	Less than	a <= b	Less than or equal to	is.null(a)	Is null

Types

Converting between common data types in R. Can always go from a higher value in the table to a lower value.

<code>as.logical</code>	TRUE, FALSE, TRUE	Boolean values (TRUE or FALSE).
<code>as.numeric</code>	1, 0, 1	Integers or floating point numbers.
<code>as.character</code>	'1', '0', '1'	Character strings. Generally preferred to factors.
<code>as.factor</code>	'1', '0', '1', levels: '1', '0'	Character strings with preset levels. Needed for some statistical models.

Maths Functions

<code>log(x)</code>	Natural log.	<code>sum(x)</code>	Sum.
<code>exp(x)</code>	Exponential.	<code>mean(x)</code>	Mean.
<code>max(x)</code>	Largest element.	<code>median(x)</code>	Median.
<code>min(x)</code>	Smallest element.	<code>quantile(x)</code>	Percentage quantiles.
<code>round(x, n)</code>	Round to n decimal places.	<code>rank(x)</code>	Rank of elements.
<code>signif(x, n)</code>	Round to n significant figures.	<code>var(x)</code>	The variance.
<code>cor(x, y)</code>	Correlation.	<code>sd(x)</code>	The standard deviation.

Variable Assignment

```
> a <- 'apple'
> a
[1] 'apple'
```




The Environment

<code>ls()</code>	List all variables in the environment.
<code>rm(x)</code>	Remove x from the environment.
<code>rm(list = ls())</code>	Remove all variables from the environment.

You can use the environment panel in RStudio to browse variables in your environment.

Matrices

```
m <- matrix(x, nrow = 3, ncol = 3)
Create a matrix from x.
```

 <code>m[2,]</code>	- Select a row	<code>t(m)</code> Transpose
 <code>m[, 1]</code>	- Select a column	<code>m %*% n</code> Matrix Multiplication
 <code>m[2, 3]</code>	- Select an element	<code>solve(m, n)</code> Find x in: $m * x = n$

Lists

```
l <- list(x = 1:5, y = c('a', 'b'))
A list is a collection of elements which can be of different types.
```

<code>l[[2]]</code>	<code>l[1]</code>	<code>l\$x</code>	<code>l['y']</code>
Second element of l.	New list with only the first element.	Element named x.	New list with only element named y.




Also see the **dplyr** package.

Data Frames



```
df <- data.frame(x = 1:3, y = c('a', 'b', 'c'))
A special case of a list where all elements are the same length.
```

x	y
1	a
2	b
3	c

Matrix subsetting

<code>df[, 2]</code>	
<code>df[2,]</code>	
<code>df[2, 2]</code>	

List subsetting

<code>df\$x</code>		<code>df[[2]]</code>	
<i>Understanding a data frame</i>			
<code>View(df)</code>	See the full data frame.		
<code>head(df)</code>	See the first 6 rows.		

`nrow(df)`
Number of rows.

`ncol(df)`
Number of columns.

`dim(df)`
Number of columns and rows.

`cbind` - Bind columns.



`rbind` - Bind rows.



Strings

Also see the **stringr** package.

<code>paste(x, y, sep = ' ')</code>	Join multiple vectors together.
<code>paste(x, collapse = ' ')</code>	Join elements of a vector together.
<code>grep(pattern, x)</code>	Find regular expression matches in x.
<code>gsub(pattern, replace, x)</code>	Replace matches in x with a string.
<code>toupper(x)</code>	Convert to uppercase.
<code>tolower(x)</code>	Convert to lowercase.
<code>nchar(x)</code>	Number of characters in a string.

Factors

<code>factor(x)</code>	Turn a vector into a factor. Can set the levels of the factor and the order.
<code>cut(x, breaks = 4)</code>	Turn a numeric vector into a factor by 'cutting' into sections.

Statistics

<code>lm(y ~ x, data=df)</code> Linear model.	<code>t.test(x, y)</code> Perform a t-test for difference between means.	<code>prop.test</code> Test for a difference between proportions.
<code>glm(y ~ x, data=df)</code> Generalised linear model.	<code>pairwise.t.test</code> Perform a t-test for paired data.	<code>aov</code> Analysis of variance.
<code>summary</code> Get more detailed information out a model.		

Distributions

	Random Variates	Density Function	Cumulative Distribution	Quantile
Normal	<code>rnorm</code>	<code>dnorm</code>	<code>pnorm</code>	<code>qnorm</code>
Poisson	<code>rpois</code>	<code>dpois</code>	<code>ppois</code>	<code>qpois</code>
Binomial	<code>rbinom</code>	<code>dbinom</code>	<code>pbinom</code>	<code>qbinom</code>
Uniform	<code>runif</code>	<code>dunif</code>	<code>punif</code>	<code>qunif</code>

Plotting

Also see the **ggplot2** package.

 <code>plot(x)</code> Values of x in order.	 <code>plot(x, y)</code> Values of x against y.	 <code>hist(x)</code> Histogram of x.
---	---	---

Dates

See the **lubridate** package.

Basics of R code

Symbol	What it does	Example 1	Example 2
<code><-</code>	Assign operator Creates new objects	<pre>> x <- 5 > x [1] 5</pre>	<pre>> y <- "This" > y [1] "This"</pre>
<code>c()</code>	Helps create objects with more than one element	<pre>> v <- c(5,6,7,8) > v [1] 5 6 7 8</pre>	<pre>> w <- c("This", "is", "easy! ") > w [1] "This" "is" "easy!"</pre>
<code>#</code>	Computer ignores what is written. Used for adding notes to code	<pre>> # print("hello") ></pre>	<pre>> print("hello") [1] "hello"</pre>
<code>%>%</code>	Literally translates as "then do this"	<pre>> data %>% do_something_to(data)</pre>	<pre>> data %>% do_something_to(data) %>% do_something_else_to(data)</pre>
<code>%in%</code>	returns a logical vector indicating if there is a match	<pre>> "x" %in% c("x", "y", "z") [1] TRUE</pre>	<pre>> c("x", "y", "z") %in% "x" [1] TRUE FALSE FALSE</pre>
<code>?</code>	Access help	<pre>> ?mean()</pre>	<pre>> ?geom_point()</pre>

FYI: R is case sensitive!! Name.of.data ≠ name.of.data

Creating objects

For most of us, R is simply the creation of and manipulation of objects:

```
new_object <- c(1, 2, 3)
```

- the objects are then fed into functions to create amazing new objects

```
amazing_new_object <- function(new_object)
```

Broadly speaking the following is true in R:

- information

```
> data_frame <- function(information)
```

```
> plot <- function(data frame)
```

```
> model <- function(data frame)
```


Naming objects

There are a few simple rules to follow initially:

- Object names must start with a letter and can only contain letters, numbers, `'_'` and `'.'`
- Certain characters should not be used, e.g:
 - `c` is the concatenate function `'c()'`
 - `T` is used as shorthand for `TRUE`
 - `F` is used as shorthand for `FALSE`
- In this course I'll always use `x`, `y` and `z` as object names when demonstrating quick examples

Types of data structure

The main data types are;

```
# double (for double precision floating point numbers)
typeof(1.23)
```

```
# character
typeof("string")
```

```
# logical
typeof(FALSE)
```

```
# missing values are represented by NA
example <- c(1, 2, NA, 4)
```

Other examples include integers and complex numbers

Types of data structure

- Vectors come in two forms

A: Atomic vectors contain exactly one type of data

```
all_numbers      <- c(1, 2, 0.5, -0.5, 3.4)
all_characters   <- c("One", "too", "3")
all_logical      <- c(TRUE, FALSE) # NOTE: Type it out
```

B: Lists allow combinations of different types of data

```
this_is_a_list   <- list(1, TRUE, "Three", "4")
typeof(this_is_a_list)
[1] "list"
```

```
this_is_also_a_list <- list(all_numbers, all_characters)
```

Types of data structure

Matrices/Arrays:

- You can have a matrix of two or more dimensions

```
a_matrix <- matrix(1:9, 3, 3)
```

- Vectors and matrices can only contain **one** type of data
- **VERY VERY VERY NB:** If you try to create a vector with more than one data type, then it will undergo coercion to the least common denominator
- The coercion rule goes:
logical -> integer -> numeric -> complex -> character
- You can perform coercions yourself on vectors

Walkthrough

- `01_baseR_introduction.R`
 - Basic Code entry

Types of data structures

Dataframes:

- These are a special type of list
- Observations are in rows
- Variables are in columns
- Labels or other metadata may also be present

```
> a_data_frame <- data.frame(number = 1:10,  
                             char = sample(letters, 10),  
                             this_really_a_col_name = rep(c(TRUE, FALSE), 5))
```

- In the tidyverse dataframes are called 'tibbles'
- Some older functions don't work with tibbles
- We'll go through this in more detail later

Indexing

- Indexing can occur in one or two dimensions
- One dimension:

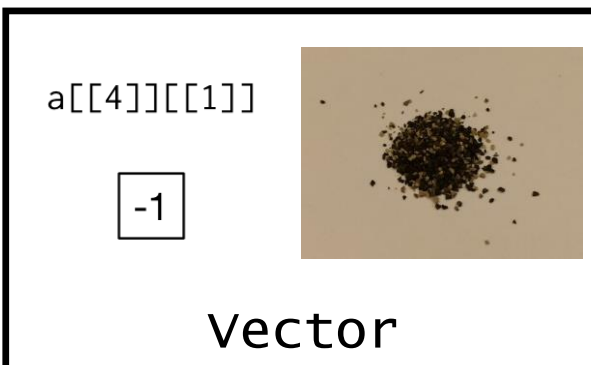
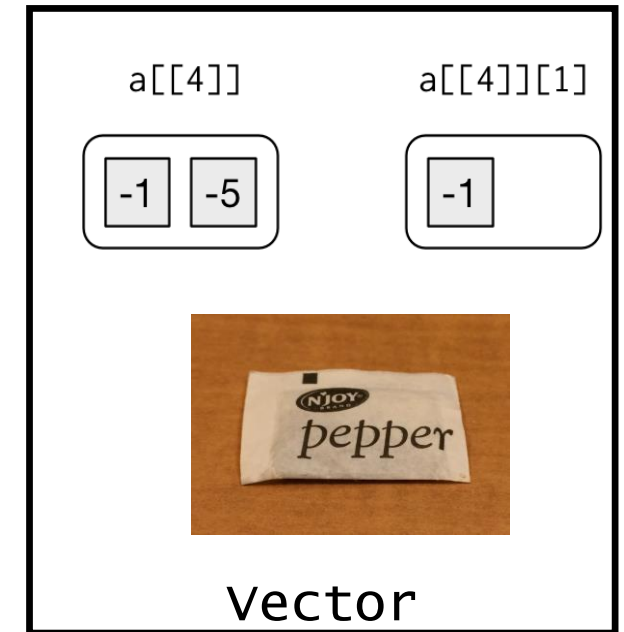
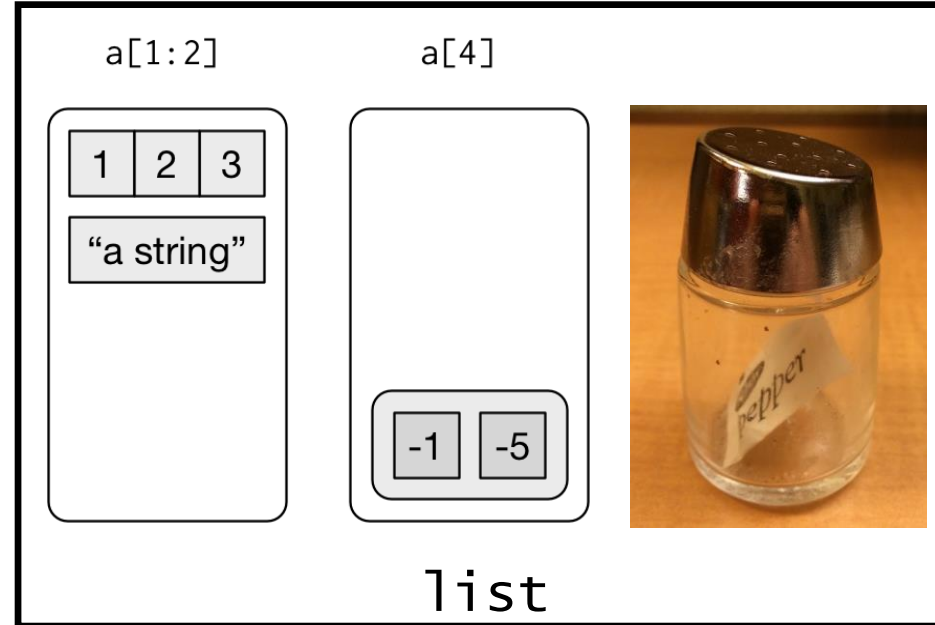
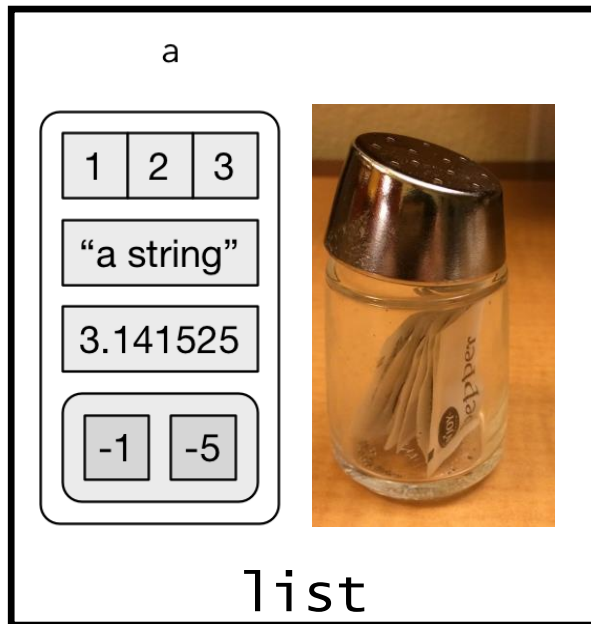
```
new_object <- c(1, 2, 3)
new_object[1]
[1] 1
```
- Two dimensions

```
a_data_frame[1, 1]
a_data_frame$number[1]
```
- In the tidyverse we don't use '[' much as `dplyr::filter()` and `dplyr::select()` allow you to solve the same problems
- However, given so much of the R has been written using these, it's worth recognising and understanding them

Indexing

```
# Recall
```

```
- this_is_also_a_list <- list(all_numbers, all_characters, all_logical)
```



```
# Important
```

- `[` extracts a sublist, results will be a list
- `[[` extracts a single component

Walkthrough

- `01_baseR_introduction.R`
 - Indexing dataframes and lists

Types of data structures

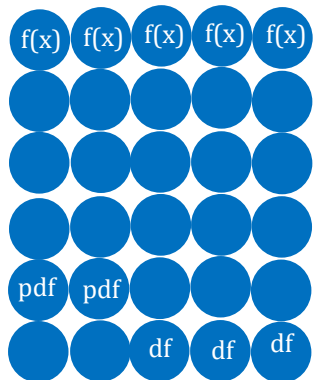
Factors:

- In R, factors are used to work with categorical variables
- Historically they were easier to work with than characters, hence many baseR functions automatically convert characters to factors
- This does not happen in the tidyverse
- One of the most important uses of factors is in statistical modeling;
 - since categorical variables are entered into statistical models differently than continuous variables, storing data as factors insures that the modeling functions will treat such data correctly

Walkthrough

- `01_baseR_introduction.R`
 - Factors example

Package contents



Base R:
Comes
pre-
loaded

Functions

```
> base::|
```

max.col	{base}
mean	{base}
mean.Date	{base}
mean.default	{base}
mean.difftime	{base}
mean.POSIXct	{base}
mean.POSIXlt	{base}
mem.limits	{base}
memCompress	{base}

mean(x, ...)

Generic function for the (trimmed) arithmetic mean.

Press F1 for additional help

Data sets

```
> data()
```

faithful
freeny
infert
iris
iris3
islands
lh
longley

iris

This famous (Fisher's or Anderson's) iris data set gives the measurements in centimeters of the variables sepal length and width and petal length and width, respectively, for 50 flowers from each of 3 species of iris. The species are *Iris setosa*, *versicolor*, and *virginica*.

Press F1 for additional help

Conflicts

```
> filter|
```

filter	{dplyr}
filter_	{dplyr}
filter_all	{dplyr}
filter_at	{dplyr}
filter_if	{dplyr}
Filter	{base}
Filters	

filter(x, filter, method = c("convolution", "recursive"), sides = 2L, circular = FALSE, init = NULL)

Applies linear filtering to a univariate time series or to each series separately of a multivariate time series.

Press F1 for additional help

Walkthrough

- `02_navigating_R_packages.R`

Worksheet

- 03_practice_worksheet.R