



IRRI

Fundamentals of Genomic Prediction and Data-Driven Crop Breeding (November, 2025)

Introduction and Learning R Software

**Module 1
November 24, 2025**

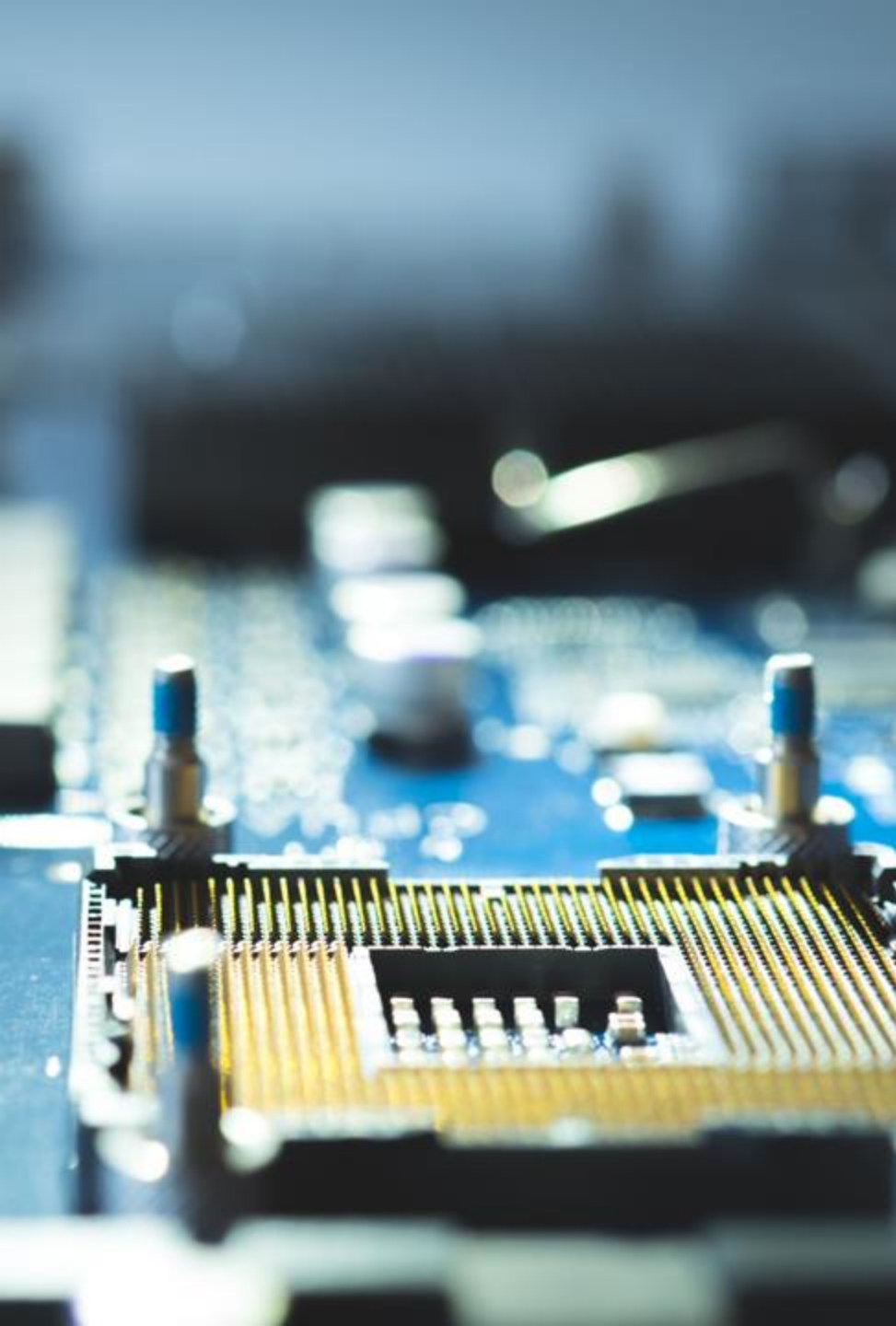
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Margaret Catolos**
Rice Breeding Innovations Platform
IRRI

Why should I learn R?

- R is a free open-source software and programming language.
 - Summarize, explore and model the data
- Reproducible research (code +text).
- Huge learning Resources and Community.
 - ❖ <https://bookdown.org/>.
 - ❖ <https://www.r-bloggers.com/>
 - ❖ <https://gkhajduk.github.io/R-resources/>
 - ❖ <https://www.computerworld.com/article/2497464/top-r-language-resources-to-improve-your-data-skills.html>
- Popular graphical capabilities.
- Dominant and useful variety of scientific disciplines.

In R Base functionality is extended through packages





What is R Studio

RStudio is an integrated development environment (IDE) for R

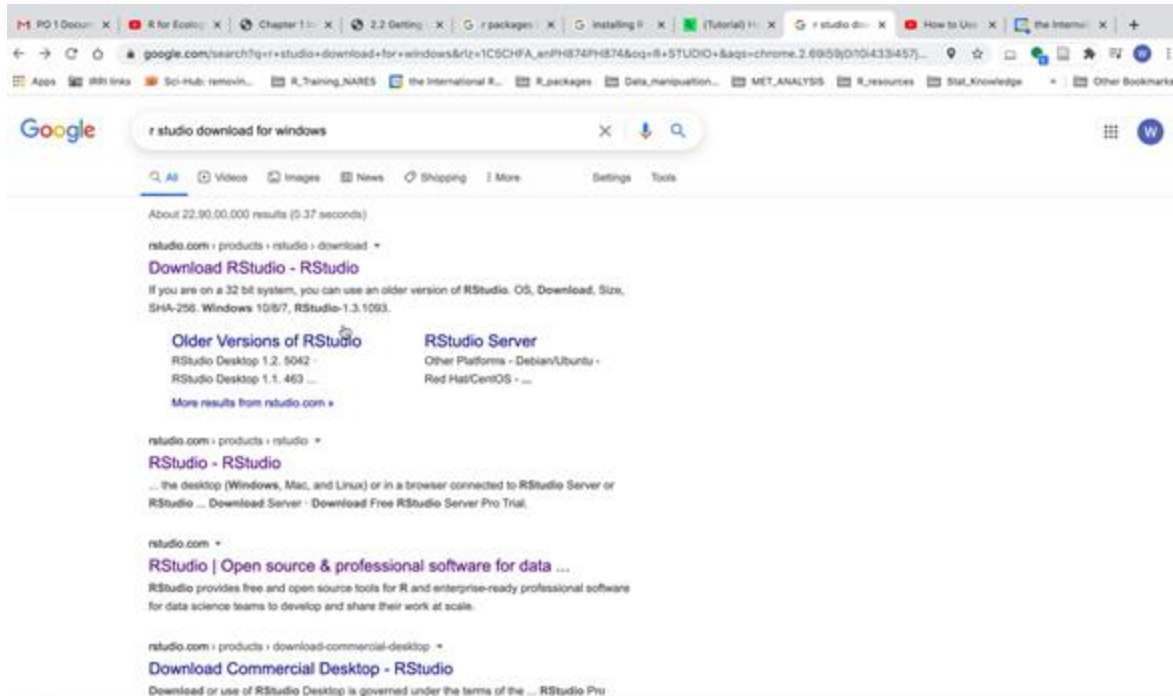
- Easy to control and manage the R scripts (point and click)
- View and interact with the objects in single environment.
- Easy to set your working directory and access files on your computer
- Graphics more accessible.

More features see the link:

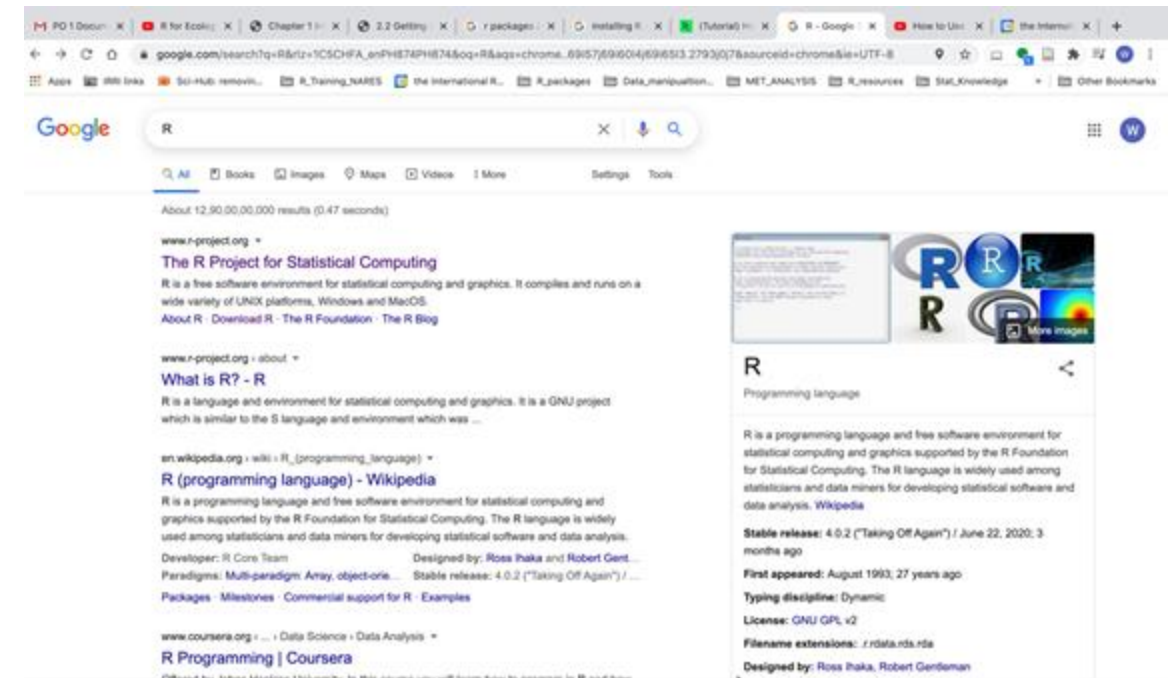
<https://rstudio.com/products/rstudio/features/>

Installing R and R studio

<https://rstudio.com/>

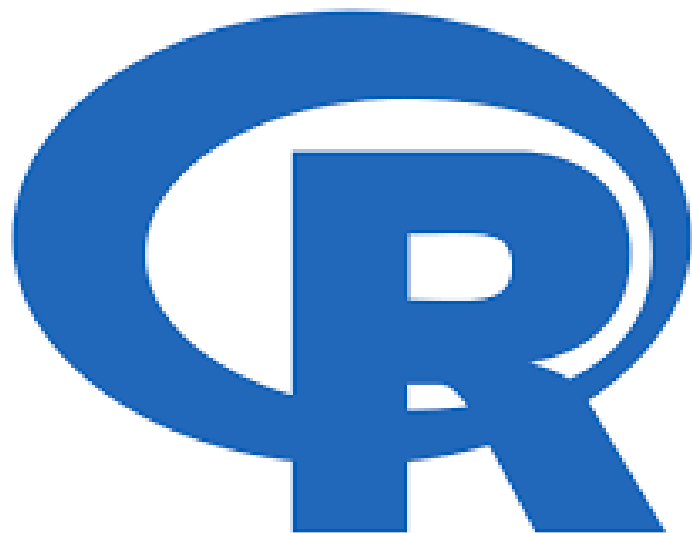


<https://www.r-project.org/>



<https://www.datacamp.com/community/tutorials/installing-R-windows-mac-ubuntu>

Installing and Loading R packages



Installing and Loading R packages

What is R package?

- Bundles of codes build by the people to perform certain tasks
- Maintained at Comprehensive R Archive Network (CRAN)/Bioconductor/GitHub

➤ Install from CRAN

```
install.package("ggplot2")
```

➤ Install from Bioconductor- Packages for life sciences related data

BiocManager handles all of the packages hosted on Bioconductor

```
install.packages("BiocManager")
```

```
BiocManager::install("SNPRelate")
```

```
BiocManager::install("phyloseq")
```

➤ Install from GitHub-


```
install.packages("devtools")
```

```
devtools::install_github("tidyr")
```

Resources

<http://www.sthda.com/english/wiki/installing-and-using-r-packages>

https://astrobiomike.github.io/R/installing_packages



R Essentials, Data types and Structures



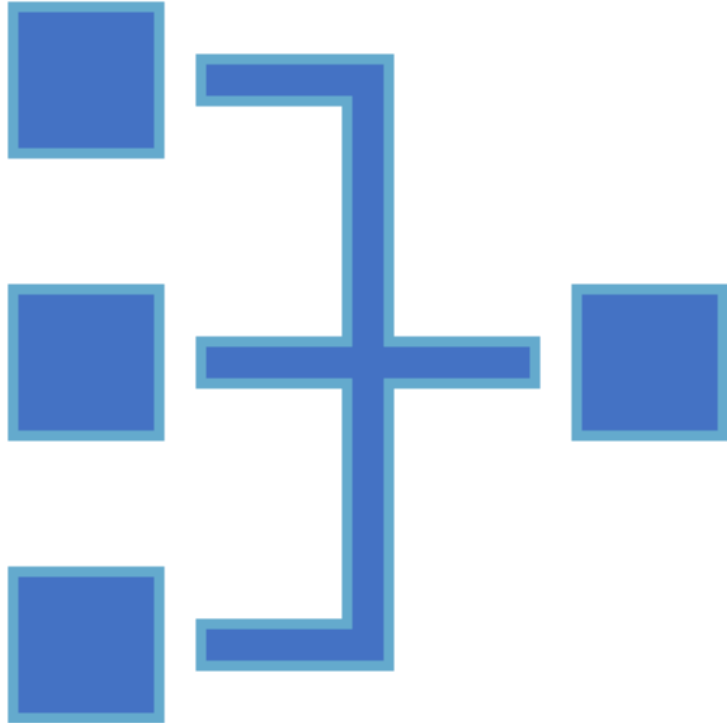
R essentials

R works on expression and objects

- **Users enters expression** (for example 2+2)
 - Expression involves operators or function calls.
 - Expression work on Objects
- **R evaluates it**
- **And Print the Results , 4**



What are Function Calls



- Calling a function which involves one or more variables.
For example *sum(x)* or *plot(x)*. Here sum and plot are function calls.
- Function format is followed by a set of parentheses containing one or more arguments. *function()*,
Example: *plot(height, weight)*, Height and weight are the arguments
- More arguments on function *plot()*
Plot(height, weight, pch=2, color="red"....)
- Positional matching
Plot(x=height, y=weight, pch=2...)

Operators in R

- Operator is a symbol that tells the compiler to perform specific mathematical or logical manipulations.
- R language is rich in built-in operators and provides following types of operators.

Arithmetic Operators in R

Operator	Description
+	Addition
-	Subtraction
*	Multiplication
/	Division
^	Exponent
%%	Modulus (Remainder from division)
%/%	Integer Division

Relational Operators in R

Operator	Description
<	Less than
>	Greater than
<=	Less than or equal to
>=	Greater than or equal to
==	Equal to
!=	Not equal to

Logical Operators in R

Operator	Description
!	Logical NOT
&	Element-wise logical AND
&&	Logical AND
	Element-wise logical OR
	Logical OR

Arithmetic Operators in R

Operator	Description
+	Addition
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Relational Operators in R

Operator	Description
<	Less than
>	Greater than
<=	Less than or equal to
>=	Greater than or equal to
==	Equal to
!=	Not equal to

Data types in R

Six data types are in R:

- **Character:** “Block”, “Replication”
- **Numeric** (real or decimal)- 2.4, 2, 10
- **Integer:** 2L, 3L
- **Logical:** TRUE, FALSE
- **Complex:** 1+6i

In R

class() - what kind of object is it

length()- how long it is



Data Structures in R

Vectors

- Vector is a basic data structure in R which contains a list of same elements
- Vectors are created using *function* `c()`
- `c()` concatenate the elements
- Examples
 - `X<-c(1,2,3,4,5)` # five components
 - Assignment operator in R is `<-`



Data Structures in R

Scalars

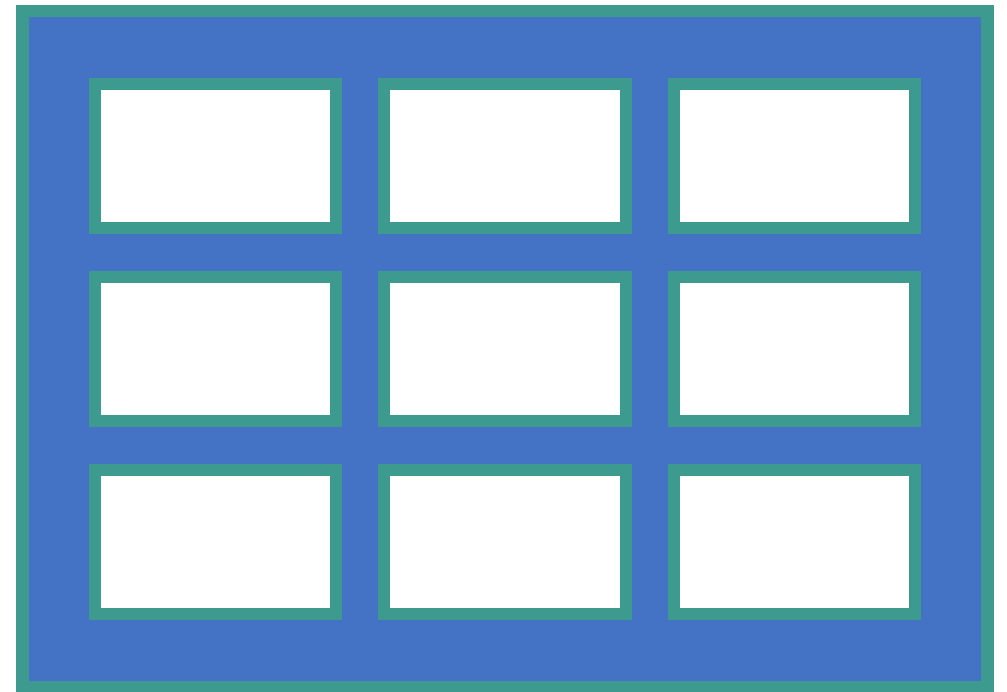
A scalar object is just a single value like a number or a name.

For example,

```
a <- 100
```

```
X <- "name"
```

Scalars don't have to be numeric, they can also be **characters** (also known as strings)



Data Structures in R

Matrices

- Matrices are numeric array of rows and columns.

Think as Stacked version of vectors where each row and column is basically a vector.

Combination of n vectors

matrix(data, nrow, ncol, byrow = FALSE)

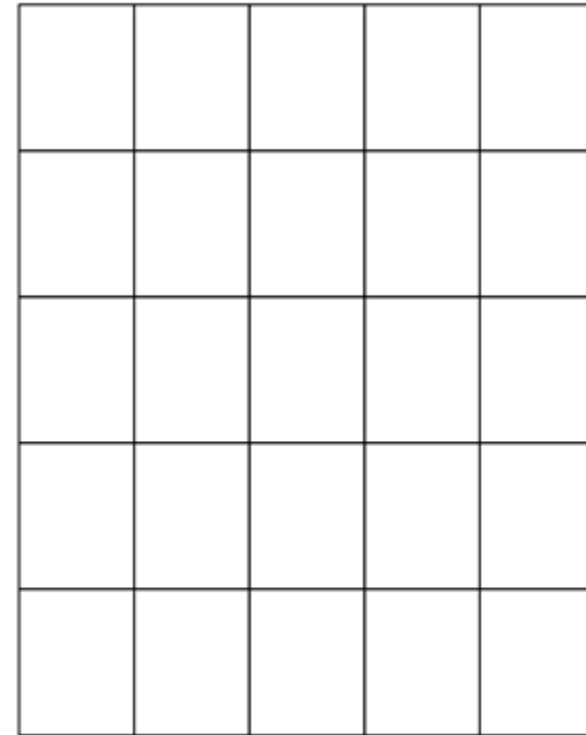
See demo how to create and deal with matrices



scalar



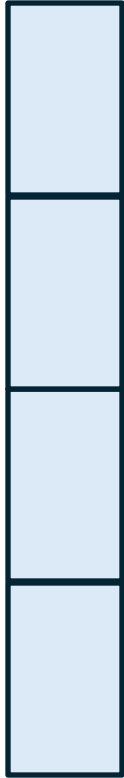
Vector



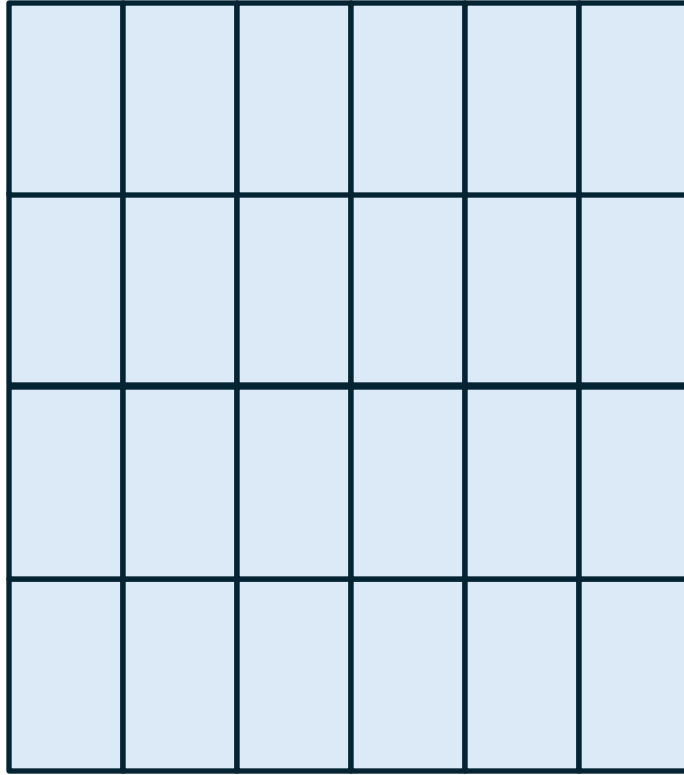
Matrix / Data Frame



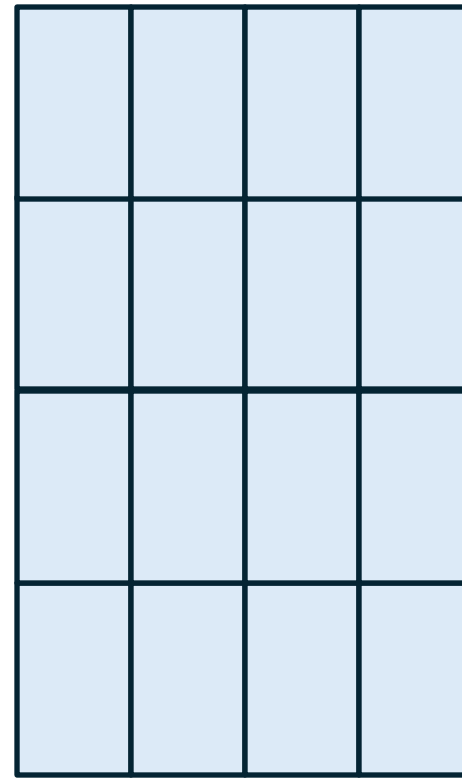
Scalar



Vector



Data frame



Matrix

Data Structures in R

LISTS

List is a data structure having components of mixed data types.

- To create a list we use **function** `list()`
- Demo in r



Data Structures in R

Data Frames

- More general than matrix, which has columns with different modes (numeric, character, factor).
- A data frame can be constructed by the *data.frame()*

- Demo in r

	Genotypes	Replication	Block	Yield
1	Genotyp1	1	Block1	2500
2	Genotype1	1	Block2	3500
3	Genotype2	2	Block1	3200
4	Genotype2	2	Block2	4500

Data Structures in R

FACTORS

- Factors are categorical variables with different levels or subdivisions.
- For example in plant Breeding Replications can be treated as factors with number of replications as factor levels
- RCBD design with 3 replications
- Replication has 3 levels
- *levels()* in R is used to determine number of levels





User defined functions

Sply, lapply...

Data manipulations, Control structures and Functions

Control Structure s

Allow users to control the flow of execution of a series of R expressions.

Commonly used control structures are:

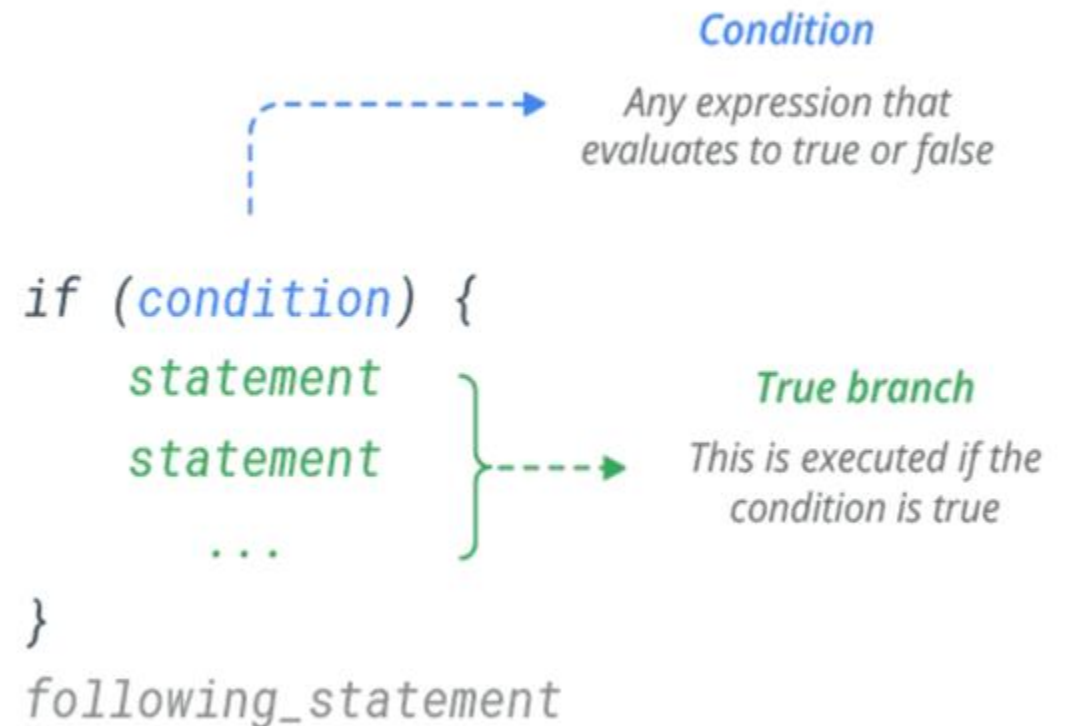
- ❖ ***if and else***: testing a condition and acting on it
- ❖ ***for***: execute a loop a fixed number of times
- ❖ ***while***: execute a loop *while* a condition is true
- ❖ ***repeat***: execute an infinite loop (must break out of it to stop)
- ❖ ***break***: break the execution of a loop
- ❖ ***next***: skip an iteration of a loop



If statement

Execute a block of code, if a specified condition is true

Demo in R



Adopted from: <https://www.learnbyexample.org/r-if-else-elseif-statement/>

If and else statement

Execute a block of code, if the condition is false

Demo in R

Syntax

```
if (condition) {  
  statement  
  statement  
  ...  
} else {  
  statement  
  statement  
  ...  
}  
following_statement
```

True branch
This is executed if the condition is true

False branch
This is executed if the condition is false

Adopted from: <https://www.learnbyexample.org/r-if-else-elseif-statement/>

Else If statement

Specify a new condition to test, if the first condition is false.

Demo in R

Syntax

```
if (condition) {  
    statement  
    statement  
    ...  
} else if (condition) {  
    statement  
    statement  
    ...  
} else {  
    statement  
    statement  
    ...  
}  
following_statement
```

First condition
This is executed if the first condition is true

New condition
A new condition to test if previous condition isn't true

False branch
This is executed if none of the conditions are true

<https://www.learnbyexample.org/r-if-else-elseif-statement/>

Multiple Condition Statements

Join two or more conditions into a single if statement

Logical operators: && (and), || (or) and ! (not). && (and) expression is True, if all the conditions are true.

Conditional statements act on single element!

Ifelse() function

function checks the condition for every element of a vector and selects elements from the specified vector depending upon the result.

Syntax

```
ifelse (condition, TrueVector, FalseVector)
```

Condition

Condition is checked for every element of a vector

True branch

Select element from this if the condition is true

False branch

Select element from this if the condition is false

Demo in R

Adopted from: <https://www.learnbyexample.org/r-if-else-elseif-statement/>

Apply functions

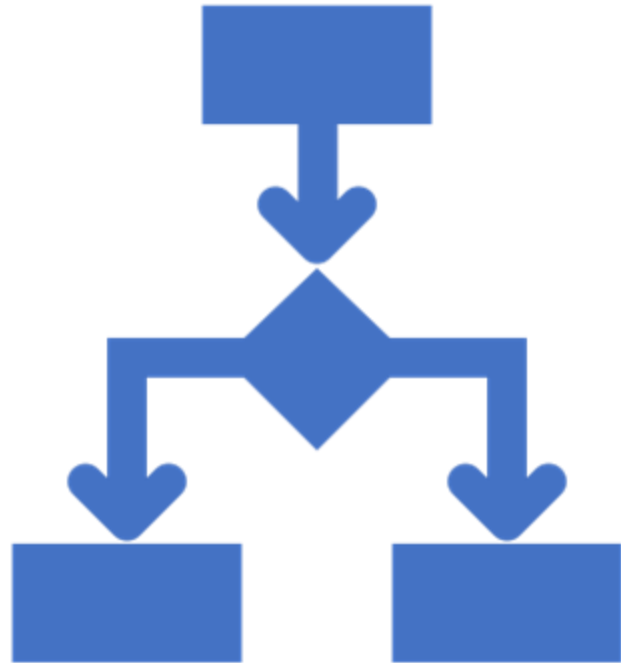
Repetitively perform an action on multiple chunks of data.

Runs faster than loop and requires less coding

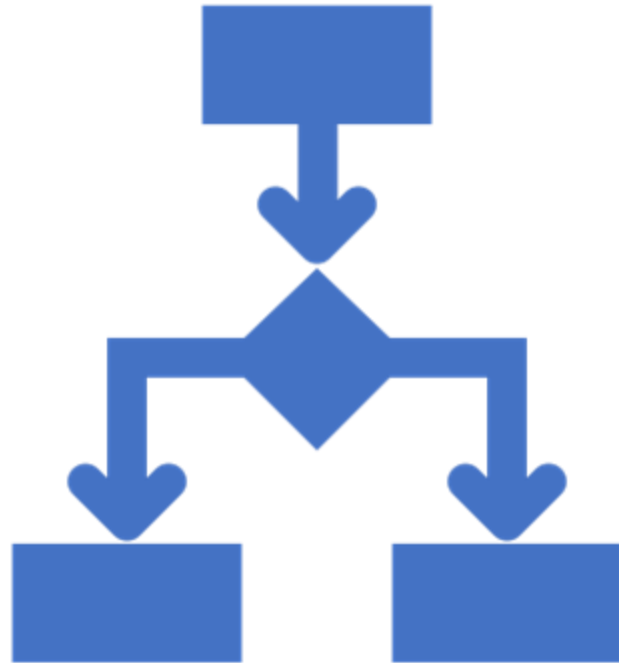
Basic function

`apply(X, MARGIN, FUN).`

- ❖ X is an array or matrix.
- ❖ Margin specifies whether you want to apply the function across rows (1) or columns (2)
- ❖ FUN is the function you want to use



1. *lapply* functions



lapply() operates on list and always returns a list, 'l' in *lapply()* refers to 'list'

lapply(X, FUN, ...)

X is a list

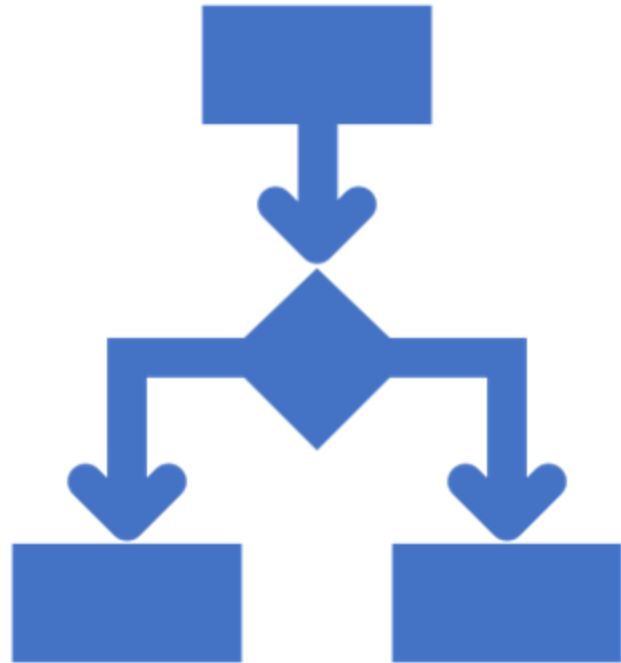
Fun, function to be applied

.... Additional arguments passed to function

- ❖ *lapply()* always returns a list whereas *apply()* can return a vector, list, matrix or array.
- ❖ No scope of MARGIN in *lapply()*, always to columns

2. *apply* functions

syntax for apply() is as follows:
apply(x, fun,.....)



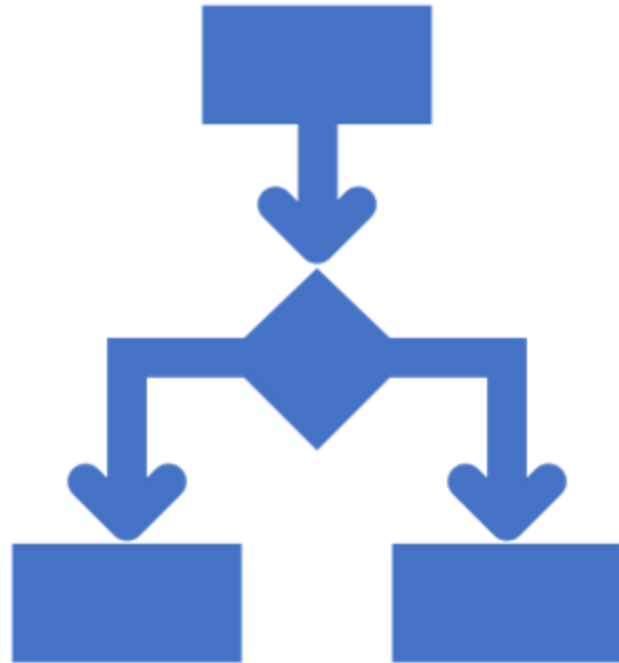
- ❖ `apply()` and `lapply()` work basically the same.
- ❖ The only difference is that `lapply()` always returns a list, whereas `apply()` tries to simplify the result into a vector or matrix.
- ❖ Additional argument if `simplify = F` then `apply()` returns a list similar to `lapply()`

3. *tapply* functions

tapply() function breaks the data set up into groups and applies a function to each group.

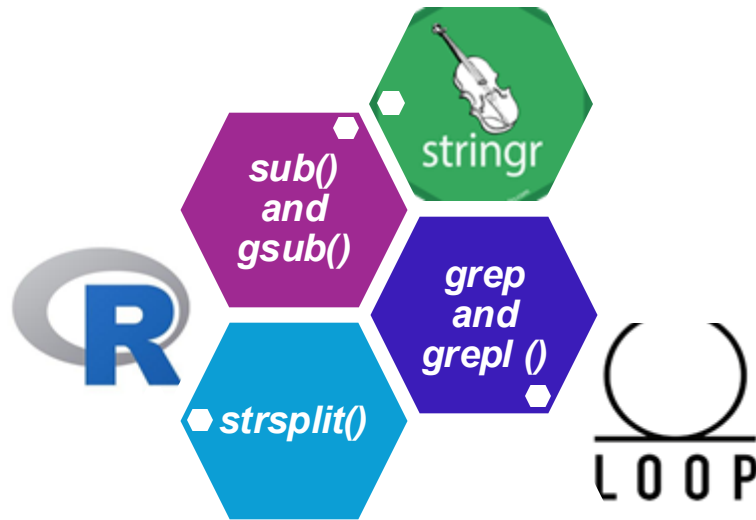
The syntax:

tapply(x, INDEX, FUN, ..., simplify)



- *x* is required vector
- A grouping factor or a list of factors
- The function to be applied
- Additional arguments
- Simplify return simplified results

Loops and String Manipulations in R



Loops in R

(Cycling or iterating)

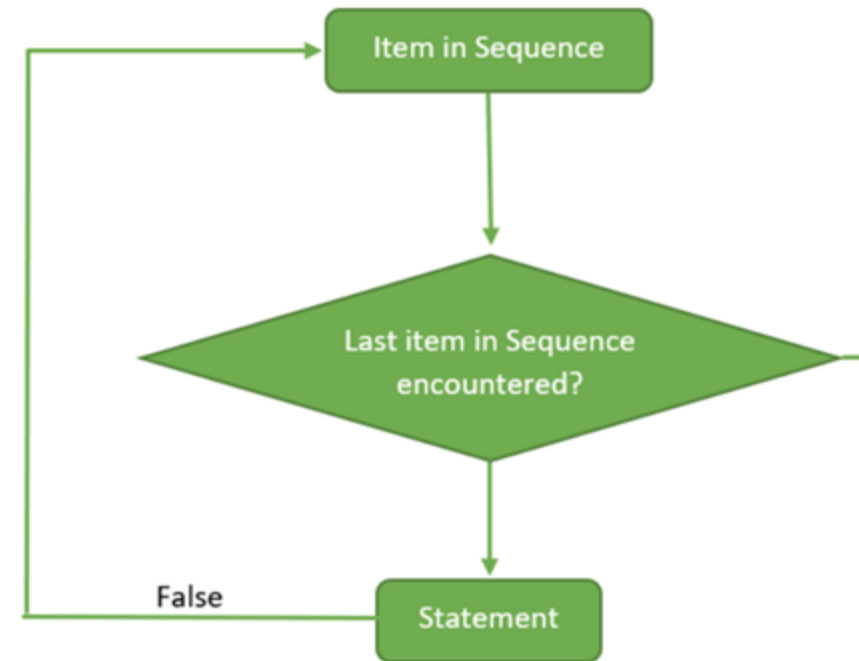
Control statement that allows multiple executions of a statement or a set of statements

For loop

- Loops over texts, data frames etc.
- Loops repeatedly depending upon the number of elements

Syntax

```
for (var in vector) {  
  statement(s)  
}
```



Loops in R

(Cycling or iterating)

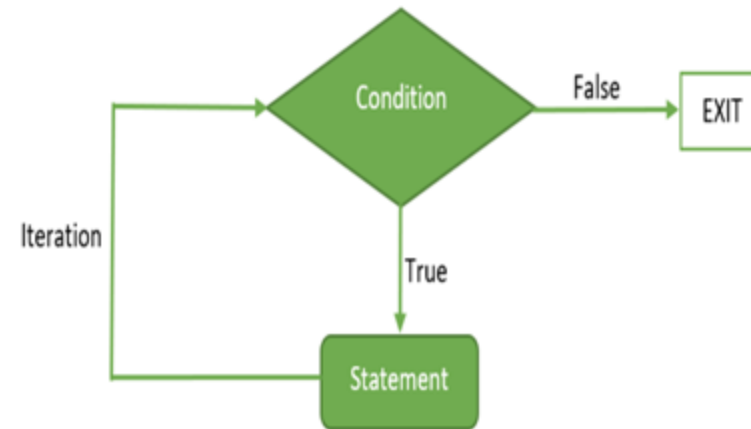
Control statement that allows multiple executions of a statement or a set of statements.

while loop

- Runs a statement or a set of statements repeatedly unless the given condition becomes false.
- Entry controlled loop.

Syntax

```
while ( condition )  
  { statement }
```



Loops in R

(Cycling or iterating)

Control statement that allows multiple executions of a statement or a set of statements.

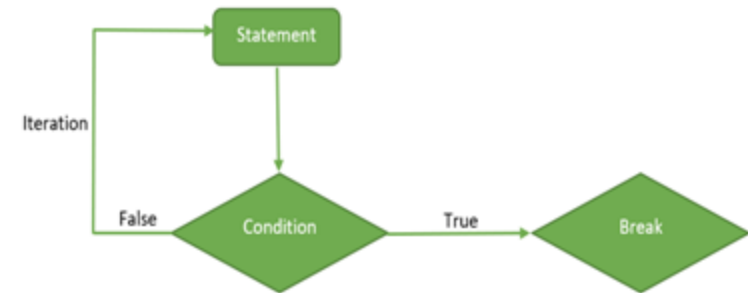
repeat loop

- run the same statement or a group of statements repeatedly until the stop condition has been encountered.
- Iterate infinitely if no condition given

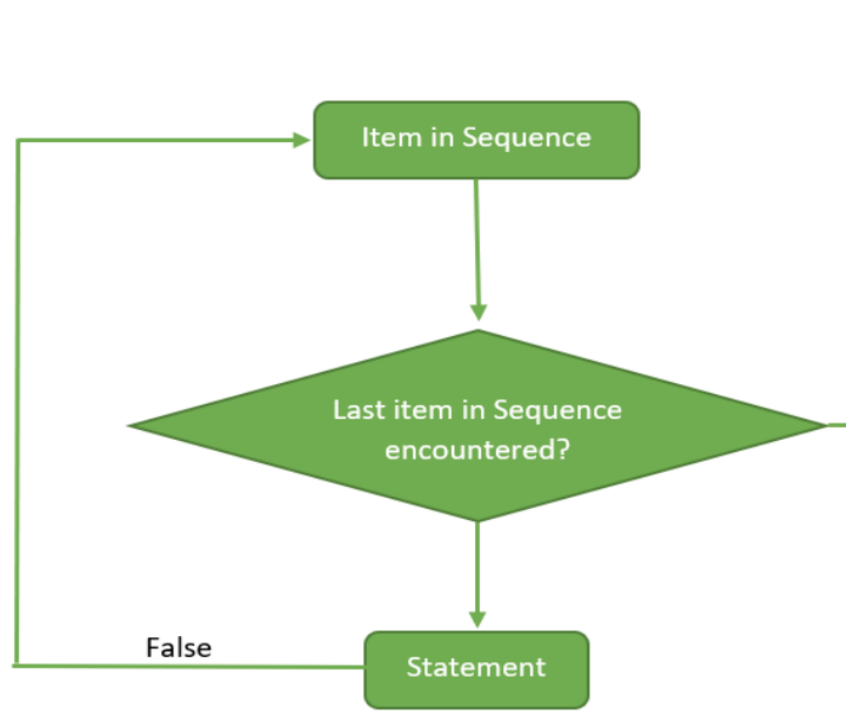
Syntax

```
repeat  
{ statement  
  if( condition )  
  { break }  
}
```

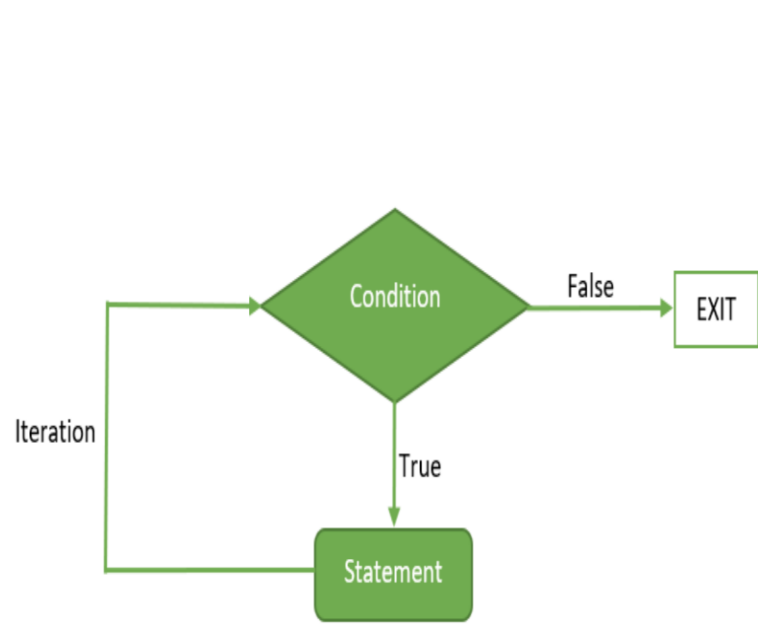
Flow Diagram:



<https://www.geeksforgeeks.org/loops-in-r-for-while-repeat/>

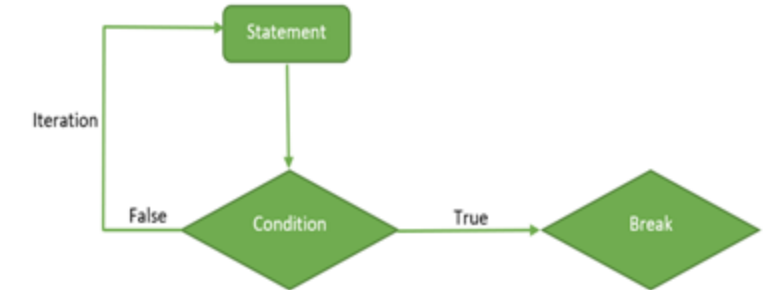


For loop



while loop

Flow Diagram:



repeat loop

String Manipulations in R

Basic String Manipulations

- *nchar()* number of characters
- *tolower()* convert to lower case
- *toupper()* convert to upper case
- *casefold()* case folding
- *chartr()* character translation
- *abbreviate()* abbreviation
- *substr()* substrings of a character vector



Set Operations

- ***union()*** set union
- ***intersect()*** intersection
- ***setdiff()*** set difference
- ***setequal()*** equal sets identical() exact equality
- ***is.element()*** is element
- ***%in%*** contains
- ***sort()*** sorting
- ***rep()*** repetition



Other string functions

- ***paste()*** concatenates several characters
- ***paste(..., sep = " ", collapse = NULL)***
- ***print()*** generic printing
- ***cat()*** concatenation

Your assignment what they does?

Working Directories

R is always pointed at a directory on your computer.

- Check current directory

getwd()

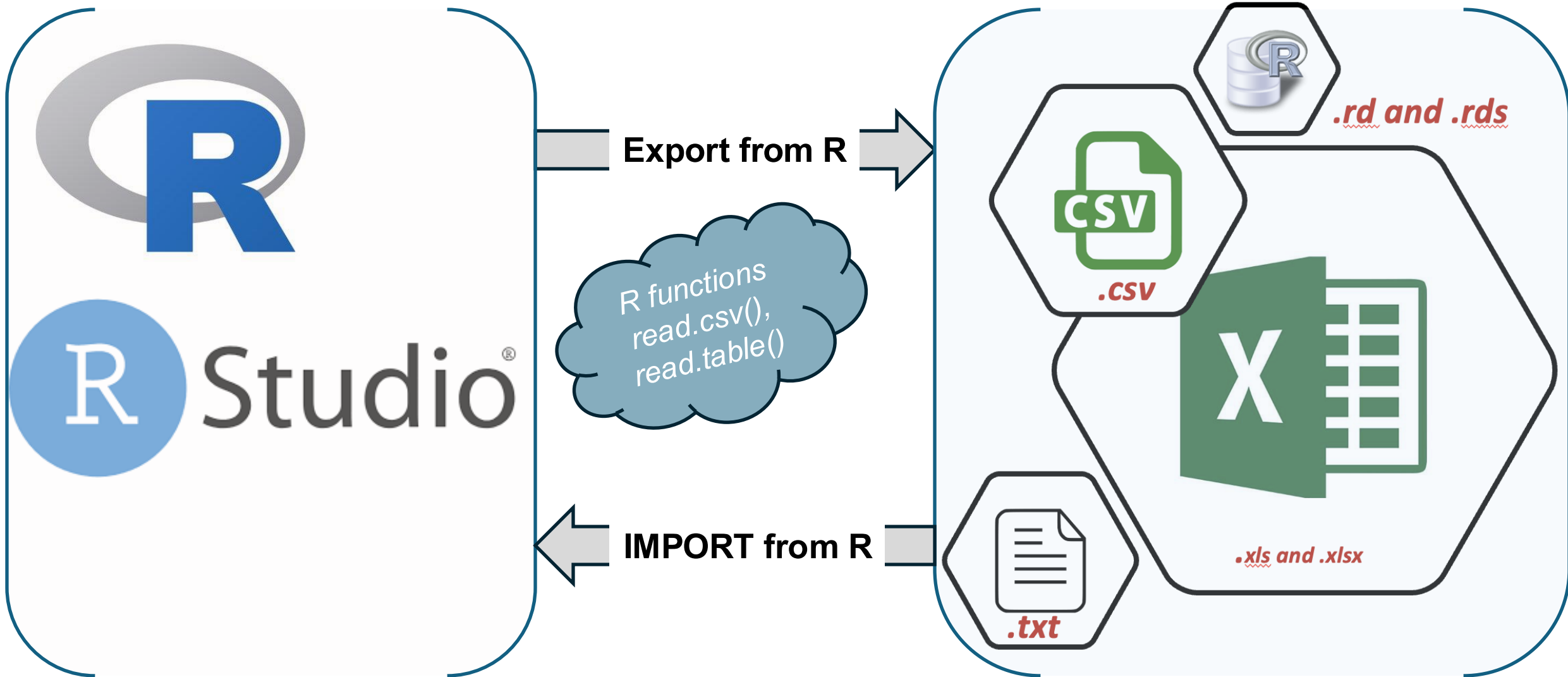
- Set working directory

setwd()

- Create working directory

dir.create()

Import and Export of Data in R





Importing Data from .csv and .txt files

Depending upon format, several variants are available

- **`read.csv()`**: for reading “**comma separated value**” files (“.csv”).
- **`read.table()`**: for reading “text files” (“.txt”)

Syntax:

`read.table(filename or path, header = FALSE, sep = “”)`

`Read.csv(filename or path, header=TRUE, sep=“”)`

Importing Data from excel file

Reading from Excel files

readxl package comes with the function **read_excel()** to read xls and xlsx files

```
my_data <- read_excel("my_file.xlsx", sheet = "data")
```

```
my_data <- read_excel("my_file.xlsx", sheet = 2)
```

```
my_data <- read_excel("my_file.xlsx", na = "---")
```

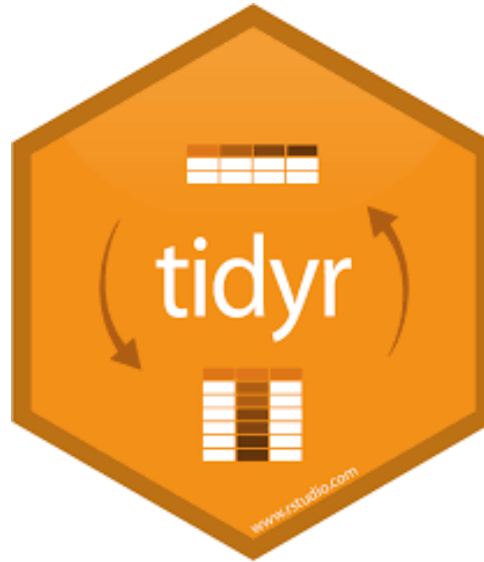
<http://www.sthda.com/english/wiki/reading-data-from-excel-files-xls-xlsx-into-r>

Exporting Data files

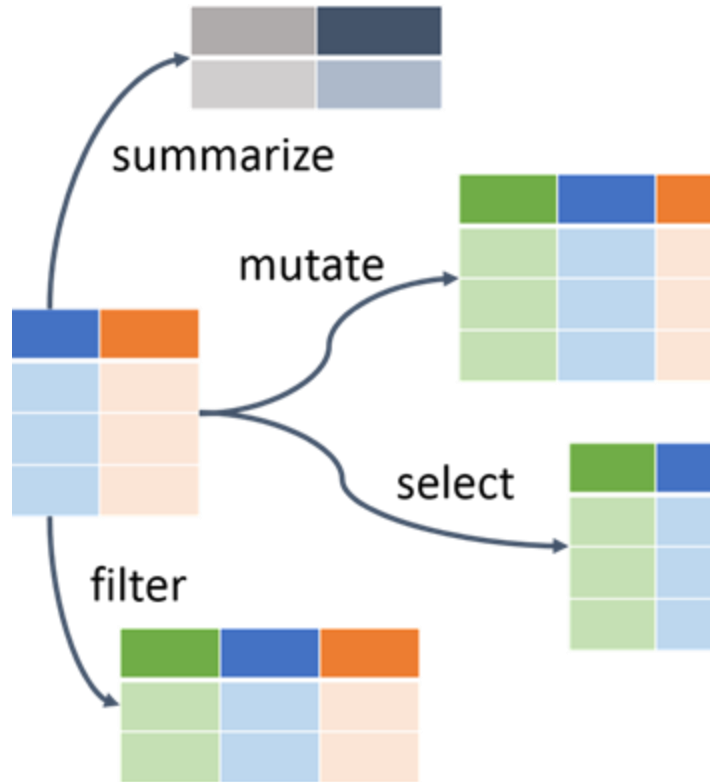
```
write.table(x, file = "",  
append = FALSE, quote =  
TRUE, sep = " ", eol = "\n", na  
= "NA", dec = ".", row.names  
= TRUE, col.names = TRUE....)
```

```
write.csv(x, file = "", append  
= FALSE, quote = TRUE, sep =  
" ", eol = "\n", na = "NA", dec  
= ".", row.names = TRUE,  
col.names = TRUE....)
```

Data Wrangling and Manipulations using R packages



dplyr R Package



- ***filter()*** chooses rows based on column values.
- ***slice()*** chooses rows based on location.
- ***arrange()*** changes the order of the rows.
- ***select()*** changes whether or not a column is included.
- ***rename()*** changes the name of columns.
- ***mutate()*** changes the values of columns and creates new columns.
- ***relocate()*** changes the order of the columns.
- ***summarise()*** collapses a group into a single row.

Pipe the functions using %>%

dplyr cheat sheet

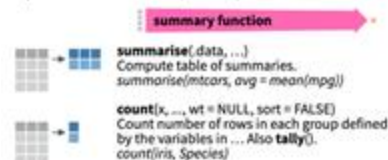
Data Transformation with dplyr : : CHEAT SHEET

dplyr functions work with pipes and expect tidy data. In tidy data:



Summarise Cases

These apply **summary functions** to columns to create a new table of summary statistics. Summary functions take vectors as input and return one value (see back).

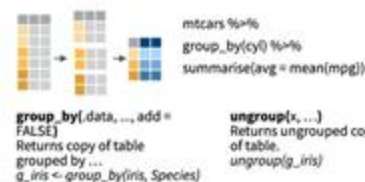


VARIATIONS

summarise_all() - Apply funs to every column.
summarise_at() - Apply funs to specific columns.
summarise_if() - Apply funs to all cols of one type.

Group Cases

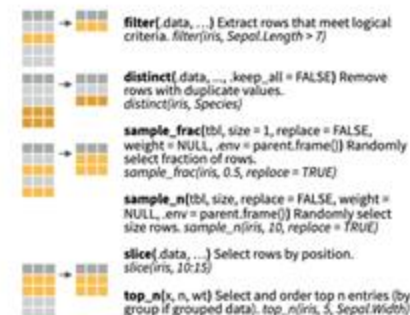
Use **group_by()** to create a "grouped" copy of a table. dplyr functions will manipulate each "group" separately and then combine the results.



Manipulate Cases

EXTRACT CASES

Row functions return a subset of rows as a new table.



Logical and boolean operators to use with filter()

`<` `<=` `is.na()` `%in%` `|` `xor()`
`>` `>=` `is.na()` `!` `&`

See `?base::Logic` and `?Comparison` for help.

ARRANGE CASES

arrange(data, ...) Order rows by values of a column or columns (low to high), use with **desc()** to order from high to low.
`arrange(mtcars, mpg)`
`arrange(mtcars, desc(mpg))`

ADD CASES

add_row(data, ..., before = NULL, after = NULL)
Add one or more rows to a table.
`add_row(faithful, eruptions = 1, waiting = 1)`

Manipulate Variables

EXTRACT VARIABLES

Column functions return a set of columns as a

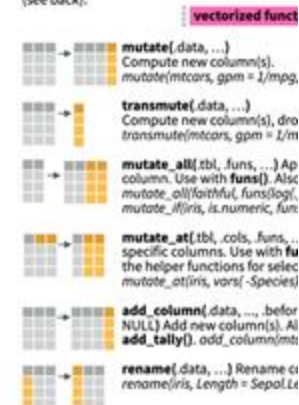


Use these helpers with **select()**, e.g. `select(iris, starts_with("Sepal"))`

contains(match) **num_range(prefix, range, ends_with(match))** **one_of(...)** **starts_with(match)** **matches(match)**

MAKE NEW VARIABLES

These apply **vectorized functions** to column vectors as input and return vectors of the same (see back).





dplyr cheat sheet

Vector Functions

TO USE WITH MUTATE ()

mutate() and **transmute()** apply vectorized functions to columns to create new columns. Vectorized functions take vectors as input and return vectors of the same length as output.

vectorized function

OFFSETS

dplyr::lag() - Offset elements by 1
dplyr::lead() - Offset elements by -1

CUMULATIVE AGGREGATES

dplyr::cumall() - Cumulative all()
dplyr::cumany() - Cumulative any()
cummax() - Cumulative max()
dplyr::cummean() - Cumulative mean()
cummin() - Cumulative min()
cumprod() - Cumulative prod()
cumsum() - Cumulative sum()

RANKINGS

dplyr::cume_dist() - Proportion of all values <=
dplyr::dense_rank() - rank w ties = min, no gaps
dplyr::min_rank() - rank with ties = min
dplyr::ntile() - bins into n bins
dplyr::percent_rank() - min_rank scaled to [0,1]
dplyr::row_number() - rank with ties = "first"

MATH

+, -, *, /, ^, %/%, %% - arithmetic ops
log(), log2(), log10() - logs
<, <=, >, >=, !=, == - logical comparisons
dplyr::between() - x >= left & x <= right
dplyr::near() - safe == for floating point numbers

MISC

dplyr::case_when() - multi-case if_else()

```
iris %>% mutate(Species = case_when(
  Species == "versicolor" ~ "versi",
  Species == "virginica" ~ "virgi",
  TRUE ~ Species))
```


dplyr::coalesce() - first non-NA values by element across a set of vectors
dplyr::if_else() - element-wise if() + else()
dplyr::na_if() - replace specific values with NA
pmax() - element-wise max()
pmin() - element-wise min()
dplyr::recode() - Vectorized switch()
dplyr::recode_factor() - Vectorized switch() for factors

Summary Functions

TO USE WITH SUMMARISE ()

summarise() applies summary functions to columns to create a new table. Summary functions take vectors as input and return single values as output.

summary function

COUNTS

dplyr::n() - number of values/rows
dplyr::n_distinct() - # of uniques
sum(!is.na()) - # of non-NA's

LOCATION

mean() - mean, also **mean(is.na())**
median() - median

LOGICALS

mean() - Proportion of TRUE's
sum() - # of TRUE's

POSITION/ORDER

dplyr::first() - first value
dplyr::last() - last value
dplyr::nth() - value in nth location of vector

RANK

quantile() - nth quantile
min() - minimum value
max() - maximum value

SPREAD

IQR() - Inter-Quartile Range
mad() - median absolute deviation
sd() - standard deviation
var() - variance

Row Names

Tidy data does not use rownames, which store a variable outside of the columns. To work with the rownames, first move them into a column.

rownames_to_column()
Move row names into col.

```
a <- rownames_to_column(iris,
var = "C")
```

column_to_rownames()
Move col in row names.

```
column_to_rownames(a, var = "C")
```

Also has **has_rownames()**, **remove_rownames()**

Combine Tables

COMBINE VARIABLES

x + **y** = **bind_cols(x, y)**

Use **bind_cols()** to paste tables beside each other as they are.

bind_cols(...) Returns tables placed side by side as a single table.
BE SURE THAT ROWS ALIGN.

Use a "Mutating Join" to join one table to columns from another, matching values with the rows that they correspond to. Each join retains a different combination of values from the tables.

left_join(x, y, by = NULL, copy = FALSE, suffix=c("x","y"),...)
Join matching values from y to x.

right_join(x, y, by = NULL, copy = FALSE, suffix=c("x","y"),...)
Join matching values from x to y.

inner_join(x, y, by = NULL, copy = FALSE, suffix=c("x","y"),...)
Join data. Retain only rows with matches.

full_join(x, y, by = NULL, copy = FALSE, suffix=c("x","y"),...)
Join data. Retain all values, all rows.

Use **by = c("col1", "col2", ...)** to specify one or more common columns to match on.
left_join(x, y, by = "A")

Use a named vector, **by = c("col1" = "col2")**, to match on columns that have different names in each table.
left_join(x, y, by = c("C" = "D"))

Use **suffix** to specify the suffix to give to unmatched columns that have the same name in both tables.
left_join(x, y, by = c("C" = "D"), suffix = c("1", "2"))

COMBINE CASES

x + **y** = **bind_rows(x, y)**

Use **bind_rows()** to paste tables below each other as they are.

bind_rows(..., .id = NULL)
Returns tables one on top of the other as a single table. Set .id to a column name to add a column of the original table names (as pictured)

intersect(x, y, ...)
Rows that appear in both x and y.

setdiff(x, y, ...)
Rows that appear in x but not y.

union(x, y, ...)
Rows that appear in x or y. (Duplicates removed). **union_all()** retains duplicates.

Use **setequal()** to test whether two data sets contain the exact same rows (in any order).

EXTRACT ROWS

x + **y** = **filter(x, ...)**

Use a "Filtering Join" to filter one table against the rows of another.

semi_join(x, y, by = NULL, ...)
Return rows of x that have a match in y. USEFUL TO SEE WHAT WILL BE JOINED.

anti_join(x, y, by = NULL, ...)
Return rows of x that do not have a match in y. USEFUL TO SEE WHAT WILL NOT BE JOINED.



Reshape Package

- reshape2 is based around two key functions: melt and cast:
- melt takes wide-format data and melts it into long-format data.
- cast takes long-format data and casts it into wide-format data.
- Think of working with metal: if you melt metal, it drips and becomes long. If you cast it into a mould, it becomes wide.

<https://uc-r.github.io/tidyr>

<https://ademos.people.uic.edu/Chapter9.html>

Reshaping Data with Tidyr

tidyr is a one such package which was built for the sole purpose of simplifying the process of creating tidy data.

- gather() makes “wide” data longer
- spread() makes “long” data wider
- separate() splits a single column into multiple columns
- unite() combines multiple columns into a single column

Go through this PPT

https://rpubs.com/bradleyboehmke/data_processing

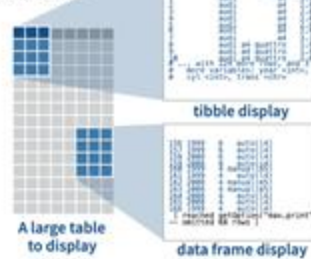
https://hbctraining.github.io/Intro-to-R/lessons/08_intro_tidyverse.html

tidyr cheat sheet

Tibbles - an enhanced data frame

The **tibble** package provides a new S3 class for storing tabular data, the tibble. Tibbles inherit the data frame class, but improve three behaviors:

- **Subsetting** - `[` always returns a new tibble, `[[` and `$` always return a vector.
- **No partial matching** - You must use full column names when subsetting
- **Display** - When you print a tibble, R provides a concise view of the data that fits on one screen



- Control the default appearance with options:
`options(tibble.print_max = n,
 tibble.print_min = m, tibble.width = Inf)`
- View full data set with **View()** or **glimpse()**
- Revert to data frame with **as.data.frame()**

CONSTRUCT A TIBBLE IN TWO WAYS

tibble(...)
Construct by columns.
`tibble(x = 1:3, y = c("a", "b", "c"))`

tribble(...)
Construct by rows.
`tribble(~x, ~y,
 1, "a",
 2, "b",
 3, "c")`

as_tibble(x, ...) Convert data frame to tibble.

enframe(x, name = "name", value = "value")

Tidy Data with tidyr

Tidy data is a way to organize tabular data. It provides a consistent data structure across packages.

A table is tidy if:



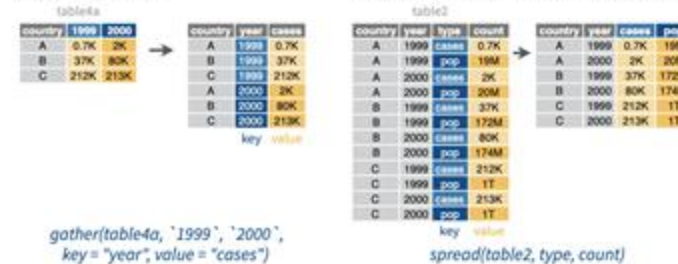
Reshape Data - change the layout of values in a table

Use **gather()** and **spread()** to reorganize the values of a table into a new layout.

gather(data, key, value, ..., na.rm = FALSE, convert = FALSE, factor_key = FALSE)
spread(data, key, value, fill = NA, convert = FALSE, drop = TRUE, sep = NULL)

gather() moves column names into a **key** column, gathering the column values into a single value column.

spread() moves the unique values of a **key** column into the column names, spreading the values of a **value** column across the new columns.



Handle Missing Values

drop_na(data, ...)
Drop rows containing NA's in ... columns.

fill(data, ..., direction = c("down", "up"))
Fill in NA's in ... columns with most recent non-NA values.

replace_na(data, replace = list(), ...)
Replace NA's by column.

Expand Tables - quickly create tables with combinations of values

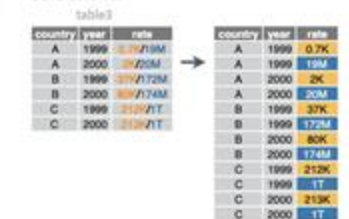
Split Cells

Use these functions to split or combine cells into individual, isolated values.

separate(data, col, into, sep = "[^:alnum:]")
`+, remove = TRUE, convert = FALSE, extra = "warn", fill = "warn", ...)`
 Separate each cell in a column to make several columns.



separate_rows(data, ..., sep = "[^:alnum:]")
`+, convert = FALSE)`
 Separate each cell in a column to make several rows.



unite(data, col, ..., sep = "_", remove = TRUE)
 Collapse cells across several columns to make a single column.



Additional Useful Resources

- ❖ <https://dplyr.tidyverse.org/>
- ❖ <https://uc-r.github.io/tidyr>
- ❖ <https://bookdown.org/mikemahoney218/IDEAR/data-wrangling.html>
- ❖ <https://exeter-data-analytics.github.io/AdVis/data-wrangling.html>
- ❖ <https://www.tidyverse.org/packages/>
- ❖ <https://atrebas.github.io/post/2019-03-03-datatable-dplyr/>



Regular Expressions

(Pattern matching and substitution)



A regular expression (a.k.a. regex) is a special text string for describing a certain amount of text.



Regular expression is a pattern that describes a set of strings.



For example, searching word "programming" in a large text document.

Regular Expressions

(Pattern matching and substitution)

grep: match a pattern

grepl: similar to grep, output as logical

regexpr: similar to grepl, output different and detailed

gregexpr: similar to regexpr, output as list

sub(): replacing one pattern with another one

gsub(): replacing one pattern with another one (all occurrences)

Writing R functions

Pieces of code that perform a desired operation on given input(s) and return the output back to the user.

Syntax

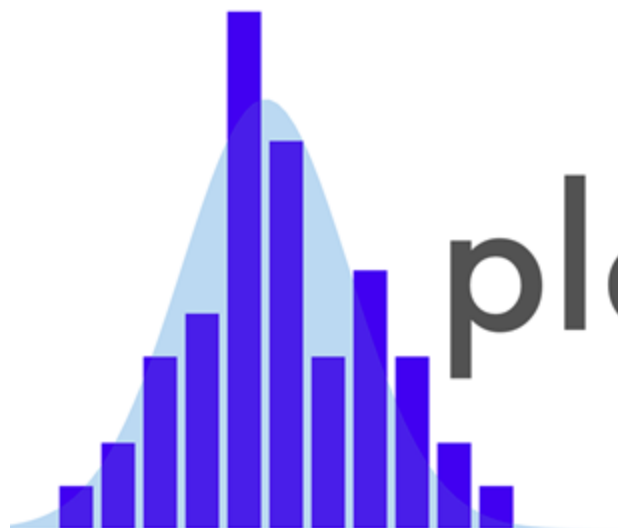
```
functionName <- function(argument1, argumen2...) {  
    #function Body  
    return(varc)  
}
```

- **Function Name** – Name of the function.
- **Arguments** – An argument is a placeholder. When a function is invoked, you pass a value to the argument.
- **Function Body** – Collection of statements that defines what the function does.
- **Return Value** – The return value of a function is the last expression in the function body to be evaluated.

Demo in R



Generating Reproducible Reports and Interactive Visualizations in R



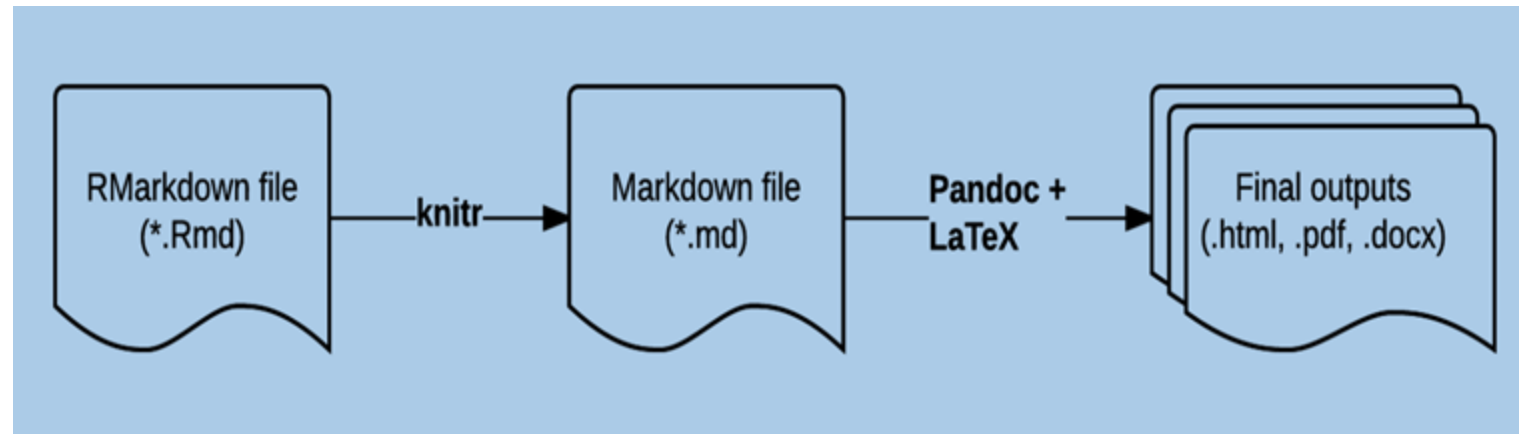
plotly





- Save the codes, execute them, and
- Generate high quality reproducible reports:
 - Edit any time
 - Seamless Visualization
 - Easy sharing.
- Knitr combines elements of R code and Markdown
- Convert the Analysis into Word, PDF, HTML etc.

R Markdown



<https://rmarkdown.rstudio.com/index.html>

https://rmarkdown.rstudio.com/articles_intro.html

<https://rstudio.com/wp-content/uploads/2015/02/rmarkdown-cheatsheet.pdf>

Workflow of Generating Reports



Open the File (.rmd)



Write the Code



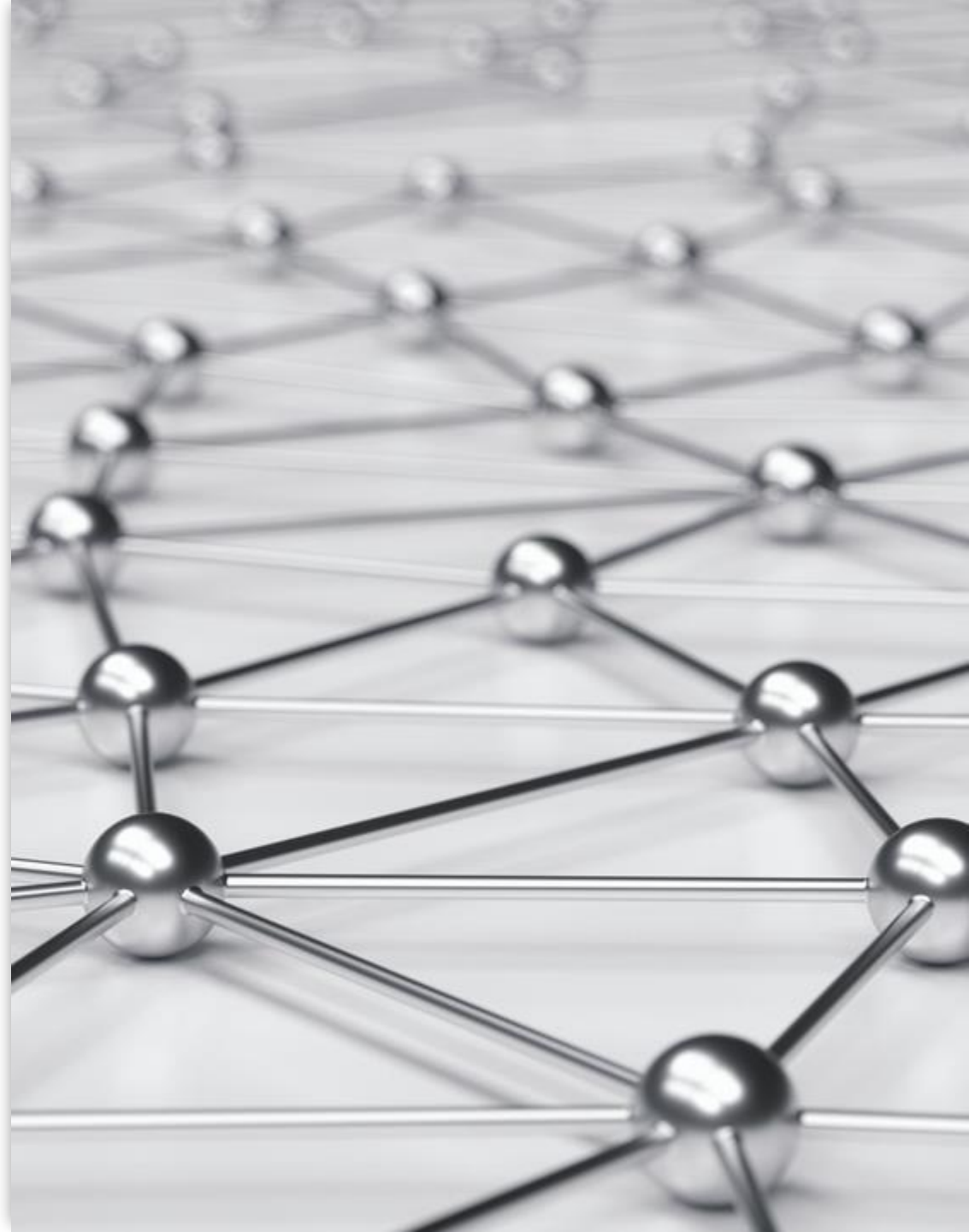
Embed the Code and Text



Render to generate the report

Useful resources to Learn R markdown

- <https://bookdown.org/yihui/bookdown/figures.html>
- <https://bookdown.org/yihui/bookdown/r-code.html>
- <https://bioconnector.github.io/workshops/r-rmarkdown.html>
- <http://bioconnector.github.io/markdown/#!/rmarkdown.md>
- <https://holtzy.github.io/Pimp-my-rmd/>
- <http://jianghao.wang/post/2017-12-08-rmarkdown-templates/>



The background of the slide is a dense, repeating pattern of a blue, textured surface. On this surface, there is a grid of circular indentations. Each indentation contains a clear glass dome. Inside each dome, a small, white, woolly sheep is visible, standing on a light pinkish-purple liquid. The sheep are positioned in various orientations within their domes. The overall aesthetic is futuristic and somewhat surreal.

Demo in R