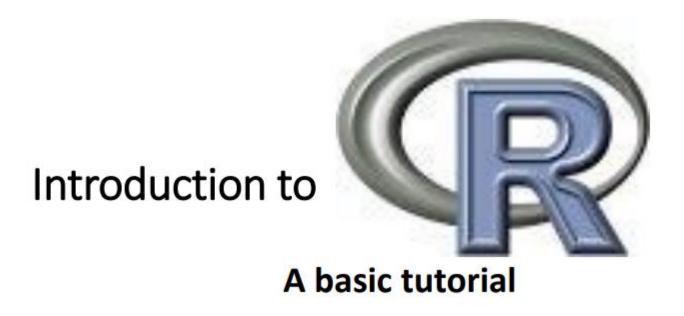
Bioinformatics softwares (R/PLINK) GBIO0002 Archana Bhardwaj

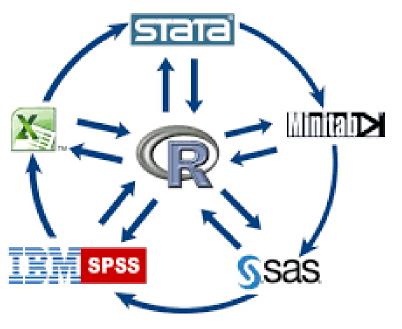
Outline

- R
- Features of R
- Simple Data Visualization
- Advance Data Visualization
- Data Transformation
- PLINK

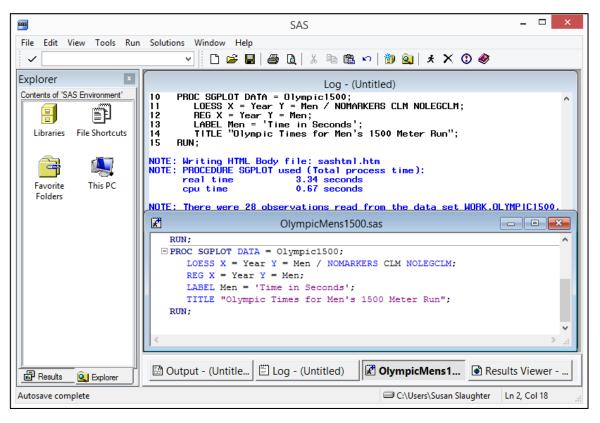


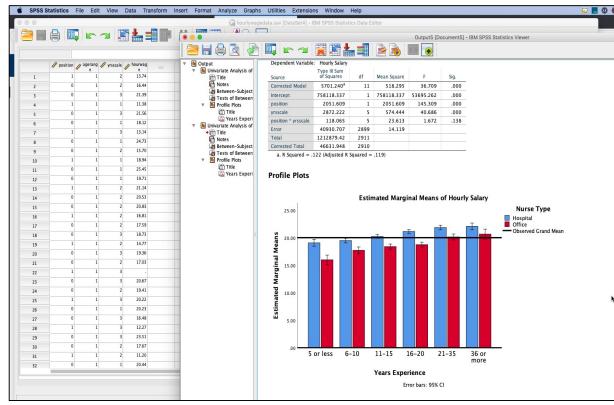
Definition of R

- R is a free software environment for statistical computing and grpahics
- R is considered to be one of the most widely used languages amongst statisticians, data miners, bioinfomaticians and others.
- Other commercial statistical packages are SPASS, SAS, Matlab



Statistical language GUI





SAS

Differences between Matlab and R

1. Open source:

R is an open source while Matlab is not open source

2. Speed:

R is slower than Matlab

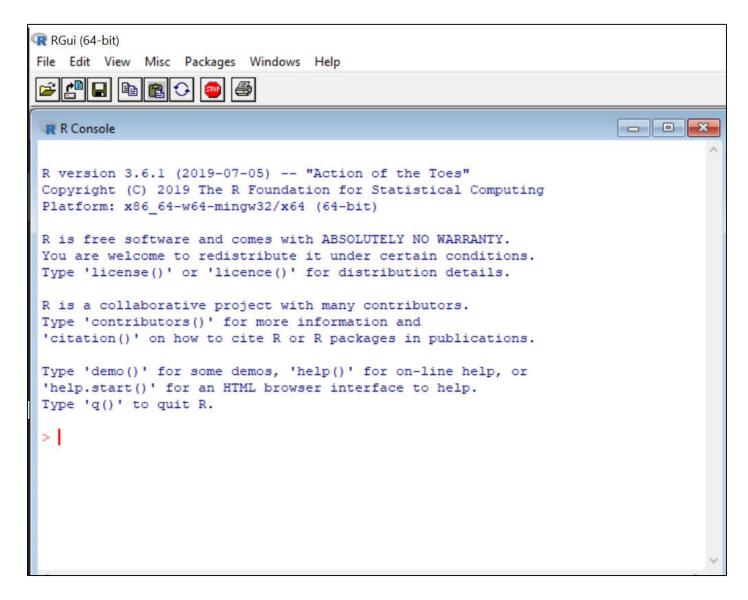
3. Functionality:

- R is mainly used for statistical analysis.
- Matlab is used for performing various engineering applications like image processing.

4. Ease of use:

- R follows programming lagugage syntax.
- Matlab is easy to program as multiple toolbox are available

R GUI



Less fancy and no frills, but free!

Why to learn R?

- Since it is free and open-source.
- R consist of wide selection of additional libraries.
- Main library repositories CRAN and Bioconductor.
- A massive set of packages for statistical modelling, machine learning, visualization, and importing and manipulating data.
- Powerful tools for communicating your results.
- Deep-seated language support for data analysis. This includes features like missing values, data frames, and vectorization.

Why to learn R?

- Cutting edge tools. Researchers in statistics and machine learning will often publish an R package to accompany their articles. This means immediate access to the very latest statistical techniques and implementations.
- RMarkdown makes it easy to turn your results into HTML files, PDFs, Word documents, PowerPoint presentations, dashboards and more.
- Shiny allows you to make beautiful interactive apps without any knowledge of HTML or javascript.
- RStudio, the IDE, provides an integrated development environment, tailored to the needs of data science, interactive data analysis, and statistical programming.
- R can connect easily to high-performance programming languages like C, Fortran, and C++.

Why to learn R?

- R's metaprogramming capabilities allow you to write magically succinct and concise functions.
- Excellent environment to handle ggplot2, dplyr, data.table, and more.
- Compared to other programming languages, the R community is more focused on results than processes.

Install R

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

and do the following (assuming you work on a windows computer)



Download

CRAN

R Project

About R
Logo
Contributors
What's New?
Reporting Bugs
Conferences
Search
Get Involved: Mailing Lists
Developer Pages
R Blog

R Foundation

The R Project for Statistical Computing

Getting Started

R is a free software environment for statistical computing and graphics. It compiles and runs on a wide variety of UNIX platforms, Windows and MacOS. To **download R**, please choose your preferred CRAN mirror.

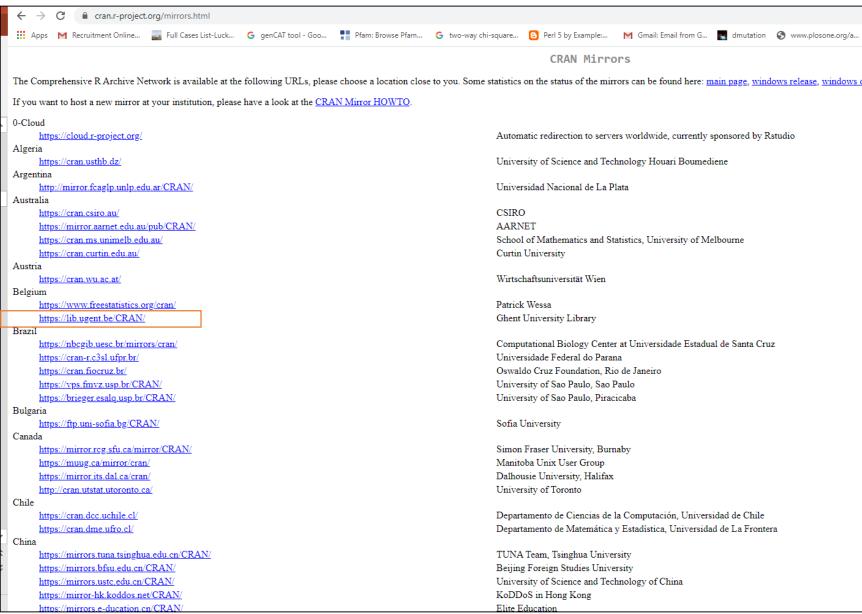
If you have questions about R like how to download and install the software, or what the license terms are, please read our answers to frequently asked questions before you send an email.

News

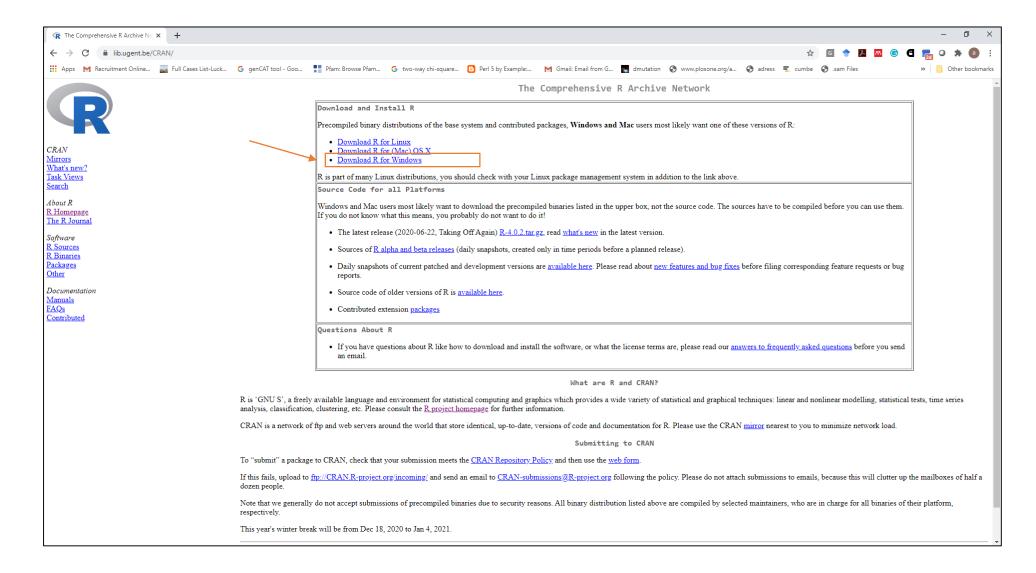
- R version 4.0.2 (Taking Off Again) has been released on 2020-06-22.
- useR! 2020 in Saint Louis has been cancelled. The European hub planned in Munich will not be an inperson conference. Both organizing committees are working on the best course of action.
- R version 3.6.3 (Holding the Windsock) has been released on 2020-02-29.
- . You can support the R Foundation with a renewable subscription as a supporting member

News via Twitter

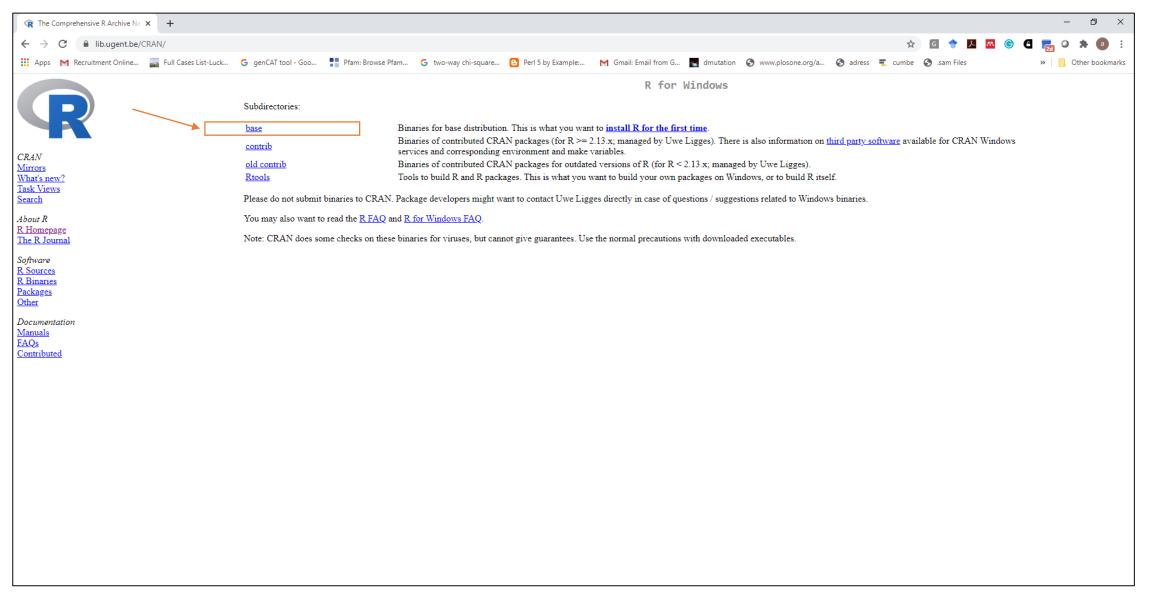
Choose download site



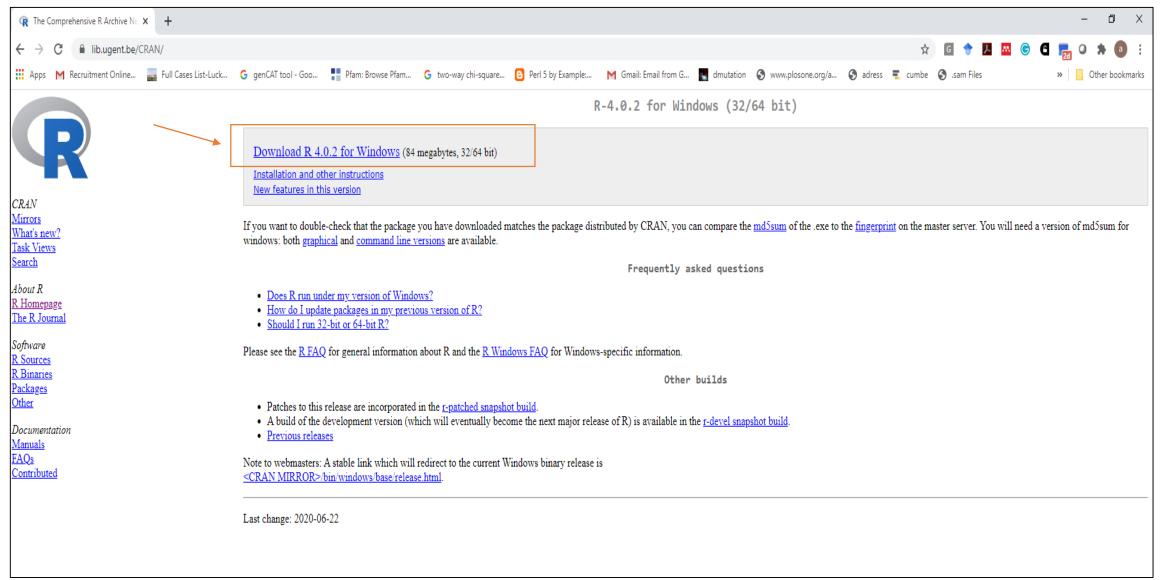
Choose Windows as target operation system



Click base



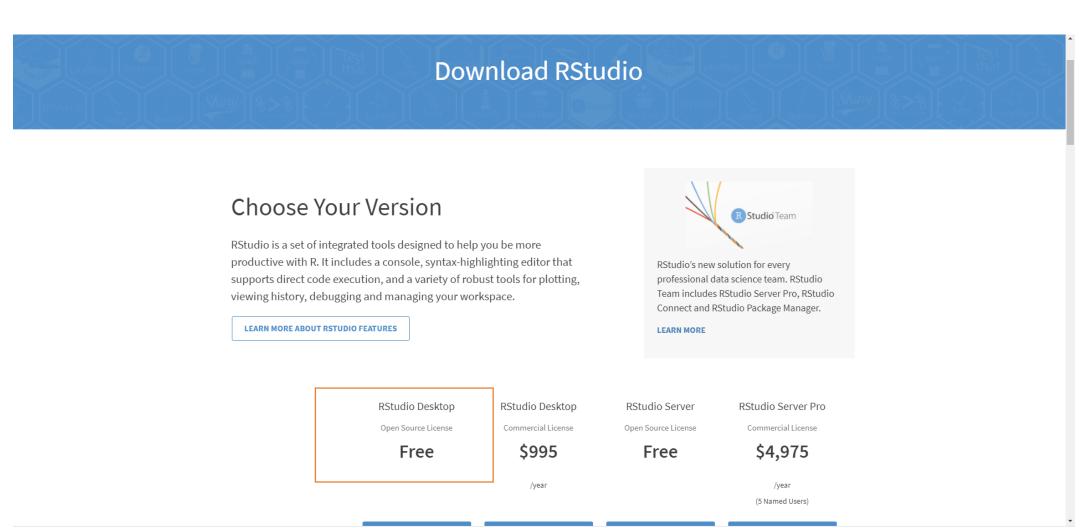
Click Download R 4.0.2



Download the .exe file and runeit (choose default answers for all questions) 15

Install Rstudio

https://rstudio.com/products/rstudio/download/



RStudio Desktop RStudio Desktop RStudio Server RStudio Server Pro Commercial License Commercial License Open Source License Open Source License \$4,975 \$995 Free Free /year /year (5 Named Users) **DOWNLOAD BUY DOWNLOAD** BUY

RStudio Desktop 1.3.1073 - Release Notes

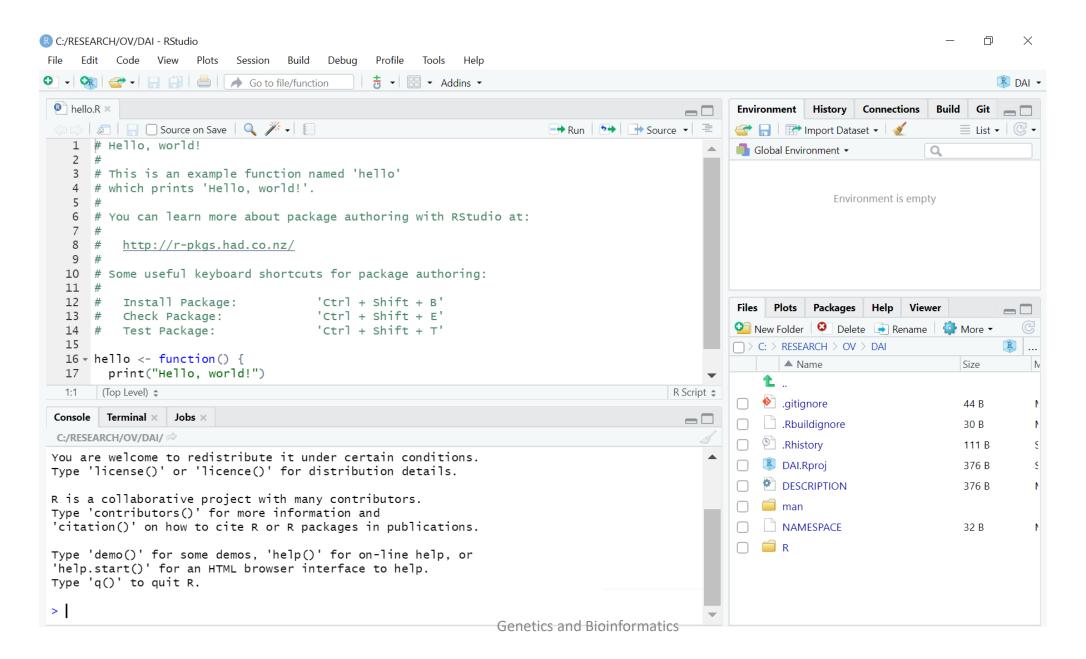
- 1. Install R. RStudio requires R 3.0.1+.
- 2. Download RStudio Desktop. Recommended for your system:



Requires Windows 10/8/7 (64-bit)



Rstudio layout



RStudio

- Bottom left: console window. Here you can type simple commands after the
 > prompt and R will execute your command. This is the most important window,
 because this is where R actually does stuff.
- Top left: editor window (also called script window). Collections of commands (scripts) can be edited and saved.
- Top right: workspace/history window. In the workspace window, you can see which data and values R has in its memory. You can view and edit the values by clicking on them. This history window shows what has been typed before.
- Bottom right: files/plots/packages/help window. Here you can open files, view plots
 (also previous plots), install and load packages and use the help function.

Use R or RStduio Lets use it!

Operators and DATA TYPES

Variables/Operators

Variables store one element

$$x < -25$$

Here x variable is assigned value 25

Check value assigned to the variable x

- Basic mathematical operators that could be applied to variables: (+), (-), (-),(*)
- Use parenthesis to obtain desired sequence of mathematical operations

Calculator

R can be used as a calculator. You can just type your equation in the command windows after the >:

Workspace

You can also give numbers a name. By doing so, they become so-called variables which can be used later. For example, you can type in the command window:

$$> a = 4$$

You can also ask R what a is (just type a ENTER in the command window):

> a

Or do calculations with a :

■ To remove all variables from R's memory, type

Concatenation function

c(),

Vectors

 Vectors have only 1 dimension and represent enumerated sequence of data. They can also store variables

■ The elements of a vector are specified/modified with braces (e.g. [number])

Logical operators

- These operators mostly work on vectors, matrices and other data types.
- Type of data is not important, the same operators are used for numeric and character data type

Operator	Description
<	less than
<=	less than or equal to
>	greater than
>=	greater than or equal to
==	exactly equal to
!=	not equal to

R workspace

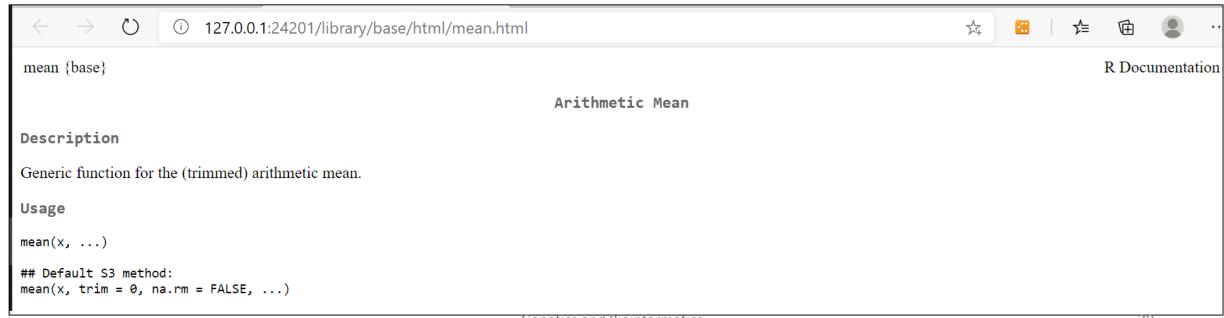
Display all workplace objects (variables, vectors etc.) via ls():

Useful tip: to save workplace and restore from a file use:

```
>save.image(file="workspace.rda")
>load(file="workspace.rda")
```

Information of function

- Any function in R has help information
- To invoke help use ? Sign and help():
 - >? mean
 - > help(mean)



Data types

- Data could be of 3 basic data types:
- numeric
- Character
- logical

Numeric variable type:

logical variable type:

Character variable type:

Data objects

- The main data objects in R:
 - Matrices (single data type)
 - Data frames (supports various data types)
 - Lists (contains set of vectors)
- Matrices are 2D objects (rows/columns) :

```
> m <- matrix(0,2,3)
> m
[,1] [,2] [,3]
[1,] 0 0 0
[2,] 0 0 0
```

List

List contain various vectors. Each vectors in the list can be acessed by double braces [[number]]

```
> x <- c(1,2,3,4)
> y <- c(2,3,4)
> L1 <- list(x,y)
> L1

[[1]]
        [1] 1 2 3 4
        [[2]]
        [1] 2 3 4
```

Data frames

Data frames are similar to matrices but can contain various data types

RStudio: Advance Feature

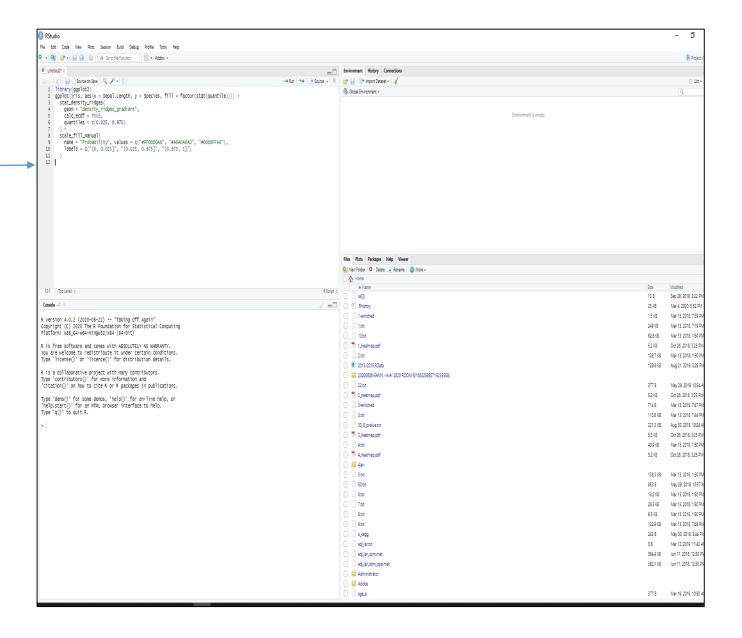
R markdown

- Markdown is a simple formatting language designed to make authoring content easy for everyone.
- Rather than write in complex markup code (e.g. HTML or LaTex), you write in plain text with formatting cues.

install.packages("rmarkdown")

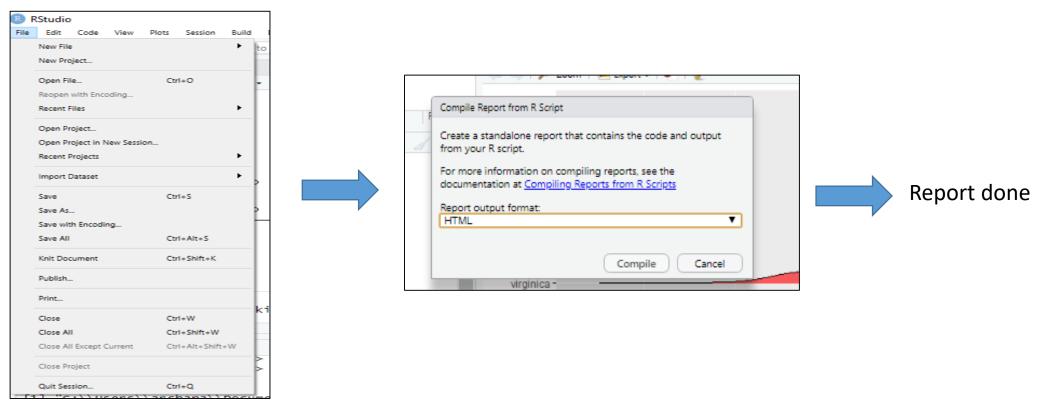
```
install.packages("ggplot2")
install.packages("ggridges")
```

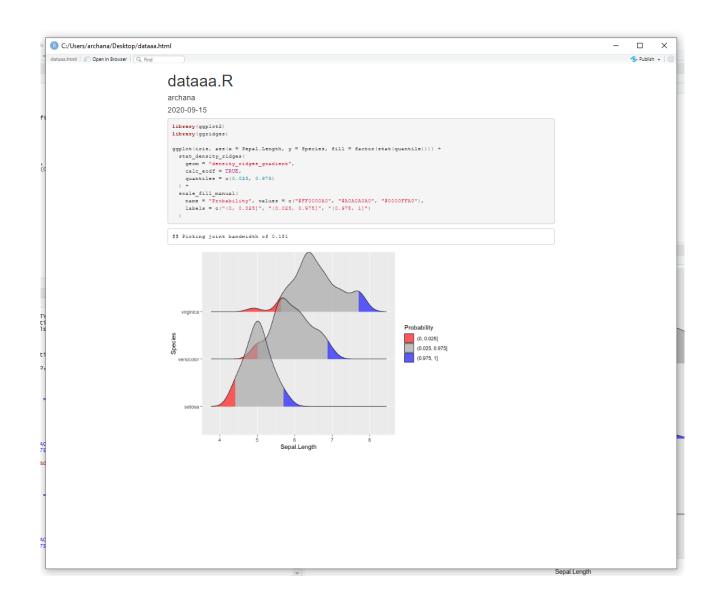
```
library(ggplot2)
library(ggridges)
ggplot(iris, aes(x = Sepal.Length, y =
Species, fill = factor(stat(quantile)))) +
 stat density ridges(
  geom = "density ridges gradient",
  calc_ecdf = TRUE,
  quantiles = c(0.025, 0.975)
 scale fill manual(
  name = "Probability", values =
c("#FF0000A0", "#A0A0A0A0",
"#0000FFA0"),
  labels = c("(0, 0.025]", "(0.025, 0.975]",
"(0.975, 1]")
```



Run code

Click option File -> knit document





Inbuilt functions in R

duplicated()function in R

Duplicate data can be removed during analysis.

It returns a logical vector that tells you whether the specified value is a duplicate

of a previous value.

> duplicated(c(1,2,1,3,1,4))

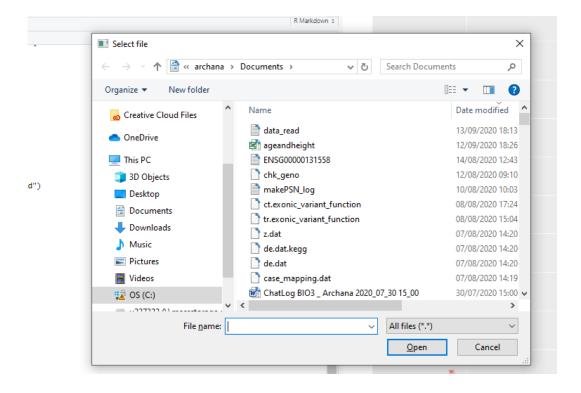
For all those values which are duplicate in the sample, true is returned.

```
File Edit View Misc Packages Windows Help
                                                                       - - X
R Console
Platform: x86 64-w64-mingw32/x64 (64-bit)
R is free software and comes with ABSOLUTELY NO WARRANTY.
You are welcome to redistribute it under certain conditions.
Type 'license()' or 'licence()' for distribution details.
R is a collaborative project with many contributors.
Type 'contributors()' for more information and
'citation()' on how to cite R or R packages in publications.
Type 'demo()' for some demos, 'help()' for on-line help, or
'help.start()' for an HTML browser interface to help.
Type 'q()' to quit R.
> data(iris)
> iris[1:5, ]
  Sepal.Length Sepal.Width Petal.Length Petal.Width Species
                       3.1
                                                0.2 setosa
 [1] FALSE FALSE TRUE FALSE TRUE FALSE
```

Read data in R

- file.choose() to walk through your directory to select a file and load.
- In case you don't want header then set header=FALSE.

> data <- file.choose()



- > data_R <- read.table(data)
- > data_R

Write data in R

- In R, we can write data frames easily to a file, using the write. table() command.
- The first argument refers to the data frame to be written to the output file, the second is the name of the output file.
- By default R will surround each entry in the output file by quotes, so we use quote=F

> write.table(x, file, sep = " ", row.names = TRUE, col.names = TRUE)

na.omit() function in R

- Rows which have NA values can be removed using the na.omit() function as below:
- Download data from website

```
> read <- read.table(file="data_read2.txt")
> read
   sample1 sample2 sample3 sample4 sample5 sample6
gene1
                            3
gene2
        NA 7
gene3
gene4
                                 3
gene5
        NA 2
gene6
gene7
        NA
gene8
> na.omit(read)
   sample1 sample2 sample3 sample4 sample5 sample6
         3
                            3
gene1
gene2
gene4
gene5
gene7
```

table() function in R

To count the number of observations in each level of factor, we can use the R table() command as below:

- > data(iris)
- > head(iris)
- > table(iris\$Species)

```
data(iris)
> head(iris)
  Sepal.Length Sepal.Width Petal.Length Petal.Width Species
           5.1
                        3.5
                                     1.4
                                                       setosa
           4.9
                        3.0
                                     1.4
                                                       setosa
           4.7
3
                        3.2
                                     1.3
                                                       setosa
           4.6
                                     1.5
4
                        3.1
                                                       setosa
           5.0
                        3.6
                                     1.4
                                                  0.2 setosa
           5.4
                        3.9
                                     1.7
                                                       setosa
 table(iris$Species)
    setosa versicolor
                        virginica
        50
                    50
                               50
```

merge() function in R

If you want to combine data from different sources in $\underline{\mathbb{R}}$, you can combine different sets of data in three ways:

```
> data(cars)
> data(iris)
> str(cars)
               50 obs. of 2 variables:
$ speed: num 4 4 7 7 8 9 10 10 10 11 ...
 $ dist : num 2 10 4 22 16 10 18 26 34 17 ...
> str(iris)
'data.frame': 150 obs. of 5 variables:
$ Sepal.Length: num 5.1 4.9 4.7 4.6 5 5.4 4.6 5 4.4 4.9 ...
$ Sepal.Width : num 3.5 3 3.2 3.1 3.6 3.9 3.4 3.4 2.9 3.1 ...
 $ Petal.Length: num 1.4 1.4 1.3 1.5 1.4 1.7 1.4 1.5 1.4 1.5 ...
 $ Petal.Width : num 0.2 0.2 0.2 0.2 0.2 0.4 0.3 0.2 0.2 0.1 ...
               : Factor w/ 3 levels "setosa", "versicolor", ...: 1 1 1 1 1 1 1 1 1 1 1 ...
> merged <- merge(iris,cars)
> str(merged)
                7500 obs. of 7 variables:
$ Sepal.Length: num 5.1 4.9 4.7 4.6 5 5.4 4.6 5 4.4 4.9 ...
 $ Sepal.Width : num 3.5 3 3.2 3.1 3.6 3.9 3.4 3.4 2.9 3.1 ...
 $ Petal.Length: num 1.4 1.4 1.3 1.5 1.4 1.7 1.4 1.5 1.4 1.5 ...
 $ Petal.Width : num 0.2 0.2 0.2 0.2 0.2 0.4 0.3 0.2 0.2 0.1 ...
               : Factor w/ 3 levels "setosa", "versicolor", ...: 1 1 1 1 1 1 1 1 1 1 1 ...
 $ Species
 $ speed
               : num 4 4 4 4 4 4 4 4 4 ...
 $ dist
               : num 2 2 2 2 2 2 2 2 2 2 ...
```

The *str()* command is designed to help you examine the structure of a data object rather than providing a statistical summary.

Input/output

Input from keyboard

You can use readline() for inputing a line from the keyboard in the form of a string:

```
> w <- readline()
bioinfo
> w
[1] "bioinfo"
```

■ In interactive mode, one can print the value of that variable by just typing the variable name or expression. In batch mode, one can use the print() function.

Input/output

- The function read.table() is used usually.
- The default value of a header is 'FALSE' and hence when you do not have a header, you need not say such.
- Basically, the character strings are considered as R factors. For turning this "feature" off, you can include the argument as.is=T in your call to read.table().
- When you have a spreadsheet export file, i.e. having a type .csv where the fields are divided by commas in place of spaces, use read.csv() in place of read.table().
- You can also use read.xls for reading core spreadsheet files.

Exercise

- **a** <- c(1,2,3,4,5,6,7,8)
- **b** <- c(3,4,5,6,7,8,10)
- Calculate mean of a and b
- Check if mean(a) > mean(b) or not ?
- Download data from website
- Write a program to read data (data_read3.txt) and display content.

summary() function in R

The *summary()* command will provide you with a statistical summary of your data.

> summary(iris)

```
summary(iris)
 Sepal.Length
                Sepal.Width
                                Petal.Length
                                                Petal.Width
                                                                     Species
                      :2.000
                                                      :0.100
                                               Min.
                                                               setosa
1st Ou.:5.100
              1st Qu.:2.800
                               1st Ou.:1.600
                                              1st Ou.:0.300
                                                              versicolor:50
Median:5.800
             Median :3.000
                               Median:4.350
                                              Median :1.300
                                                              virginica:50
      :5.843
                    :3.057
                                    :3.758
                                                      :1.199
             Mean
                               Mean
                                               Mean
               3rd Qu.:3.300
                               3rd Qu.:5.100
                                               3rd Ou.:1.800
3rd Ou.:6.400
Max.
       :7.900
               Max.
                      :4.400
                               Max.
                                      :6.900
                                                      :2.500
                                               Max.
```

The summary command is, therefore, more useful as we can see minimum, maximum, mean, etc values. The summary() command works for both matrix and data frame objects by summarizing the columns rather than the rows.

Name Commands in R

- Name command and its variants are used to find or add names to rows and columns of data structures.
- Below specified are few of the commands and their explanation:
- names() It works on matrix or data frame objects.
- rownames() It works on matrix or data frame objects and is used to give names to rows.
- colnames() It works on matrix or data frame objects and is used to give names to columns.
- dimnames() Gets row and column names for matrix or data frame objects, that is,
 it is used to see dimensions of the data frame.

Exercise

- Write a program to read data (data_read3.txt) and display content.
- Display rownames() and colnames()

Creating Subsets of Data in R

- The process of creating samples is called subsetting.
- As we know, data size is increasing exponentially.
- So, the data is divided into small-sized samples and analysis of samples is done.
- Different methods of subsetting in R are:
- **-** \$
- The dollar sign operator selects a single element of data. The result of this operator is always a vector when we use it with a data-frame.
- [|

 To retrieve 5 rows and all columns of already built-in dataset iris, the below command, is used:

- > data(iris)
- > iris[1:5,]

```
R Console
R version 4.0.2 (2020-06-22) -- "Taking Off Again"
Copyright (C) 2020 The R Foundation for Statistical Computing
Platform: x86 64-w64-mingw32/x64 (64-bit)
R is free software and comes with ABSOLUTELY NO WARRANTY.
You are welcome to redistribute it under certain conditions.
Type 'license()' or 'licence()' for distribution details.
R is a collaborative project with many contributors.
Type 'contributors()' for more information and
'citation()' on how to cite R or R packages in publications.
Type 'demo()' for some demos, 'help()' for on-line help, or
'help.start()' for an HTML browser interface to help.
Type 'q()' to quit R.
> data(iris)
> iris[1:5, ]
  Sepal.Length Sepal.Width Petal.Length Petal.Width Species
          5.1
                      3.5
                                              0.2 setosa
          4.9
                      3.0
                                  1.4
                                              0.2 setosa
          4.7
                      3.2
                                  1.3
                                              0.2 setosa
          4.6
                     3.1
                                 1.5
                                              0.2 setosa
          5.0
                      3.6
                                  1.4
                                              0.2 setosa
```

Data Visualization in R

Data Visualization in R

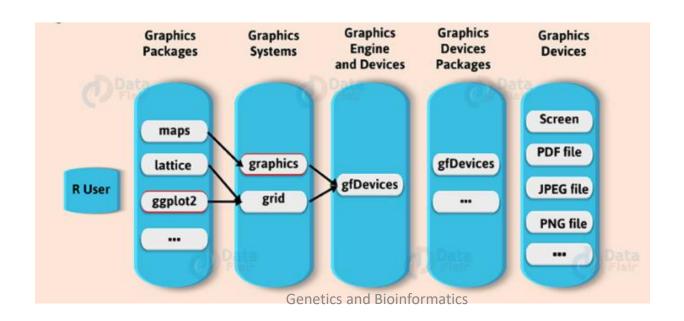
 R Programming helps us to learn this art by offering a set of inbuilt functions and also libraries to build visualizations and present data.

 Before we move forward for the technical implementation of the visualization, let's see first how to select the right chart type.

Data Visualization in R - Simple

Plotting in R

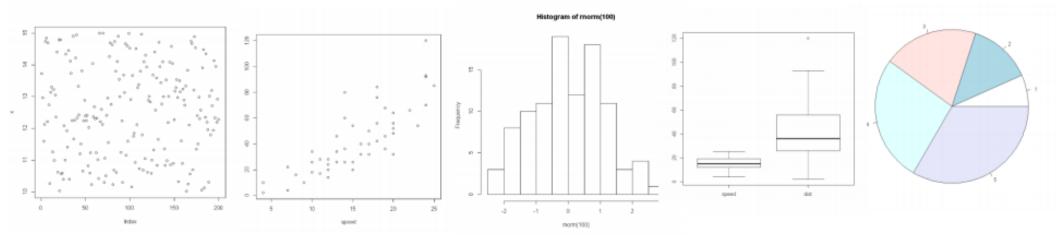
- R provides very rich set of plotting possibilities
- The basic command is plot()
- Each library has its own version of plot() function
- When R plots graphics, it opens 'grpahical device' that could be either a window or file



Plotting functions

R standard graphics available through package graphics, include several functions that provide statistical plots, like:

Function	Description		
plot(x)	plot of the values of x variable on the y axis		
	bi-variable plot of x and y values (both axis scaled based		
plot(x,y)	on values of x and y variables)		
pie(y)	circular pie-char		
boxplot(x)	Plots a box plot showing variables via their quantiles		
hist(x)	Plots a histogram(bar plot)		



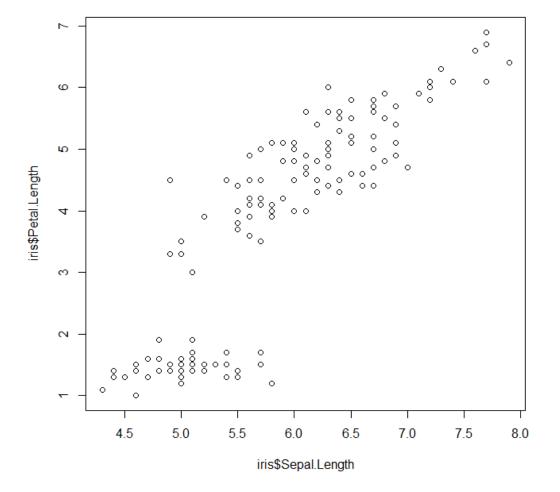
Plot function

- > data(iris)
- > head(iris)

Sepal.Length Sepal.Width Petal.Length Petal.Width Species

1	5.1	3.5	1.4	0.2 setosa
2	4.9	3.0	1.4	0.2 setosa
3	4.7	3.2	1.3	0.2 setosa
4	4.6	3.1	1.5	0.2 setosa
5	5.0	3.6	1.4	0.2 setosa
6	5.4	3.9	1.7	0.4 setosa

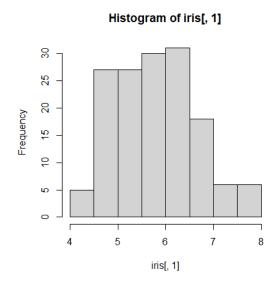
> plot(iris\$Sepal.Length,iris\$Petal.Length)

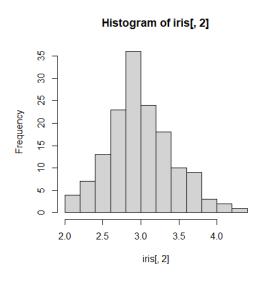


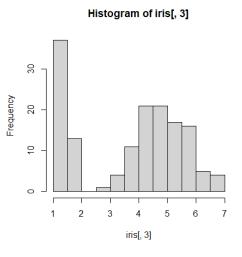
Hist function

- Histogram
- A histogram is used to plot a continuous variable. Also, It helps to break the data into bins and shows the frequency distribution of these bins. Thus, we can always change the bin size and see the effect it has on visualization.

- > hist(iris[,1])
- > hist(iris[,2])
- > hist(iris[,3])

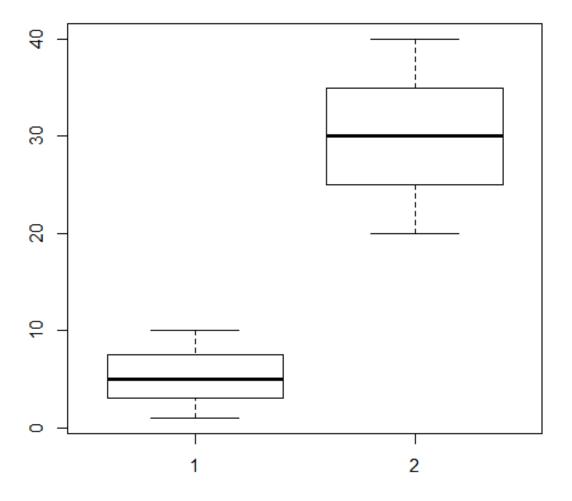






Boxplot function

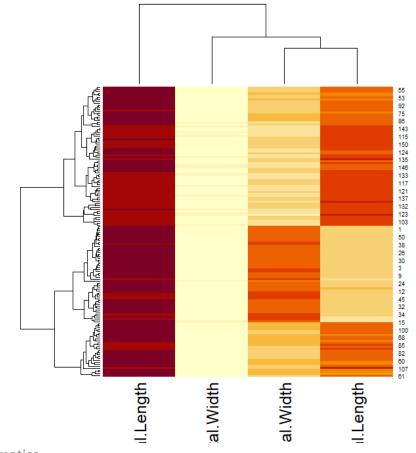
- > x <- c(1,5,10)
- > y <- c(20,30,40)
- > boxplot(x,y)



Heatmap in R

We use it for the intensity of colours. It is also used to display a relationship between two or three or many variables in a two-dimensional image. Thus, it allows us to explore two dimensions of the axis and the third dimension by an intensity of colour.

- > iris_filtered <- iris[,1:4]</pre>
- > heatmap(as.matrix(iris_filtered))

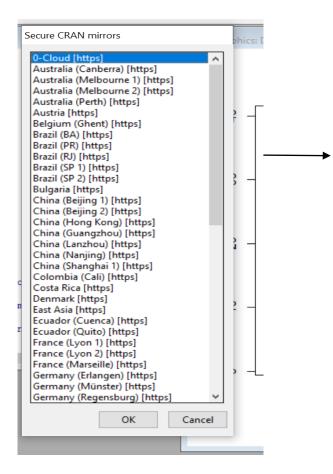


Data Visualization in R - Advanced

Library installation

install.packages

> install.packages("ggplot2")



ggplot2 installed successfully!

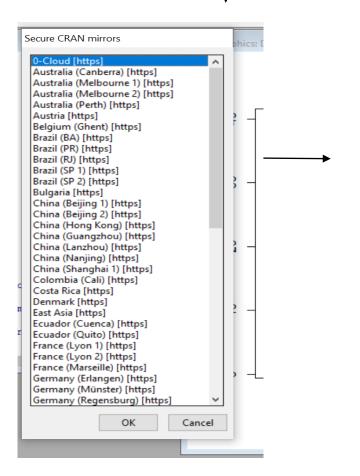
Method to Save Graphs to Files in R

- In order to save graphics to an image file, there are three steps in R:
- You can create a graphics device of PNG format using png(), JPG format using jpg() and PDF format using pdf().
- Plot your data.
- Closing the graphics device and saving the image using dev.off.

Corrgram Library installation

install.packages

> install.packages("Correlograms")

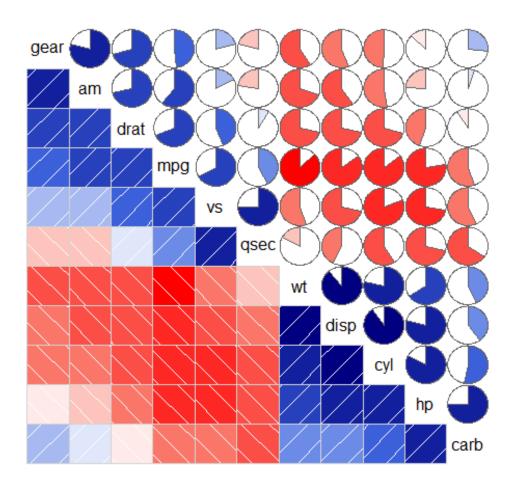


Correlograms installed successfully!

Corrgram in R

- We use it to test the level of correlation and also among the variable available in the dataset.
- Thus, the cells of the matrix can be shaded or coloured to show the co-relation value.
 - > library(corrgram)
 - > data(mtcars)
 - > corrgram(mtcars, order=TRUE, lower.panel=panel.shade, upper.panel=panel.pie, text.panel=panel.txt, main="Car Milage Data in PC2/PC1 Order")

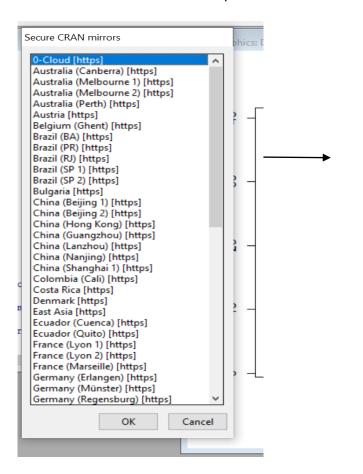
Car Milage Data in PC2/PC1 Order



circlize Library installation

install.packages

> install.packages("circlize")

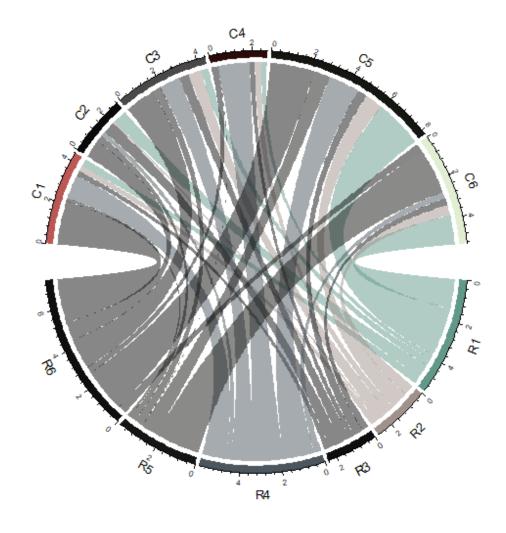


circlize installed successfully!

> library("circlize")
> mat = matrix(rnorm(36), 6, 6)
> rownames(mat) = paste0("R", 1:6)
> colnames(mat) = paste0("C", 1:6)

```
> mat = matrix(rnorm(36), 6, 6)
> rownames(mat) = paste0("R", 1:6)
 colnames(mat) = paste0("C", 1:6)
> mat
  -0.3726024 0.6870889
                         0.3497039 0.24123038 -2.3779166
             0.1061324 -0.6347521
                                    0.34156695
   0.2520366
              0.6003891
                         0.5031263 -0.25550267 0.4714030
                                                           0.3368118
R4 -1.1833159 -0.2451636
                        1.0311421 -1.55260575 1.5363097 0.3335470
   0.1313151
              0.3054917
                        0.4118880 -0.02681157 0.8029791
                                                           2.2468392
              0.8985047
                        1.4857436 -0.30758183 -2.1677454 0.3984287
   2.1757592
```

> chordDiagram(mat)

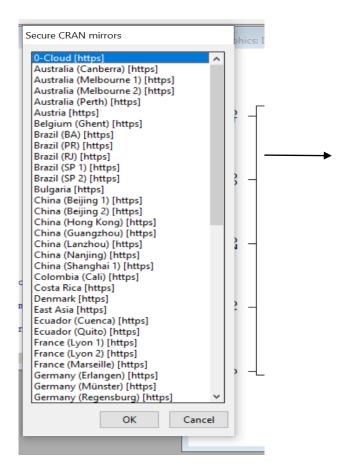


Links of two matrix data in R

ggplot2 Library installation

install.packages

> install.packages("igraph")



igraph installed successfully!

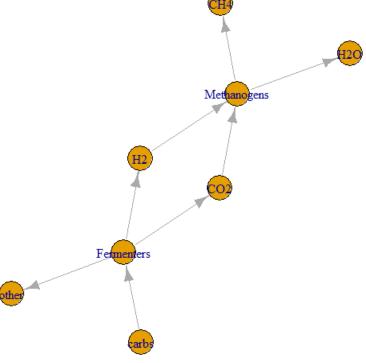
library(igraph)

A graph in made up of vertices (also called nodes or points) which are connected by edges (also called links or lines).

```
#Define Nodes
nodes=cbind('id'=c('Fermenters','Methanogens','carbs','CO2','H2','other','CH4','H2O'),
      'type'=c(rep('Microbe',2),rep('nonBio',6)))
nodes
#Define Links
links=cbind('from'=c('carbs',rep('Fermenters',3),rep('Methanogens',2),'CO2','H2'),
      'to'=c('Fermenters','other','CO2','H2','CH4','H2O',rep('Methanogens',2)),
      'type'=c('uptake',rep('output',5),rep('uptake',2)),
      'weight'=rep(1,8))
```

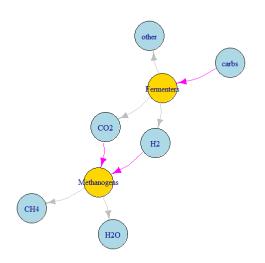
links

#Make the network
library(igraph)
net = graph_from_data_frame(links,vertices = nodes,directed = T)
plot(net)

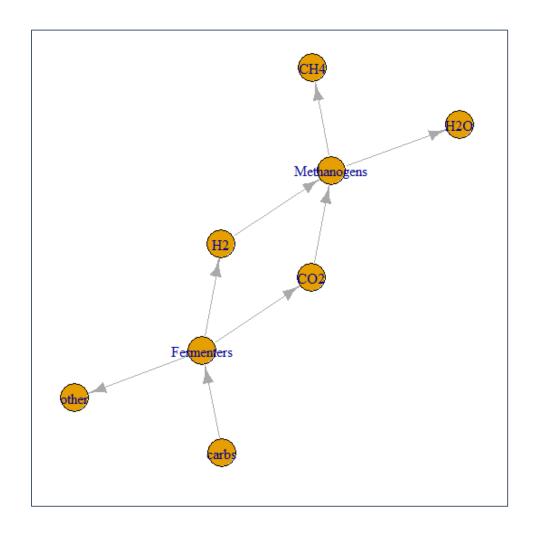


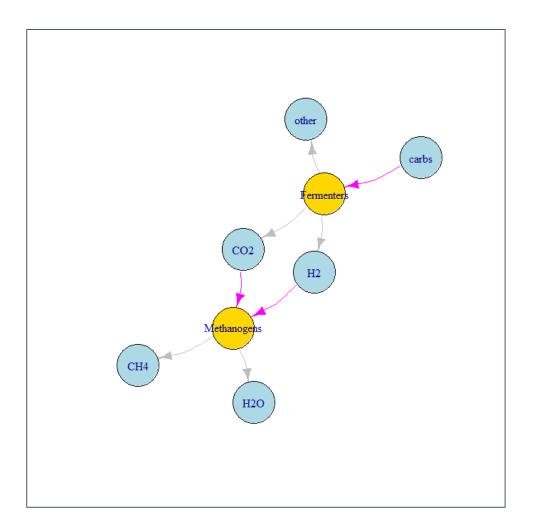
Change Appearance

- > colrs.v = c(nonBio = "lightblue", Microbe = "gold") #node colours
- > V(net)\$color = colrs.v[V(net)\$type]
- > colrs.e = c(output = "grey", uptake = "magenta") #edge colours
- > E(net)\$color = colrs.e[E(net)\$type]
- > plot(net, edge.curved=0.2, vertex.size=30) #make nodes bigger, curve arrows



Comparison of both plots





Density ridgeline plots

library(ggridges)

 The geom geom_density_ridges calculates density estimates from the provided data and then plots those, using the ridgeline visualization.

```
>ggplot(iris, aes(x = Sepal.Length, y = Species)) + geom_density_ridges()
```

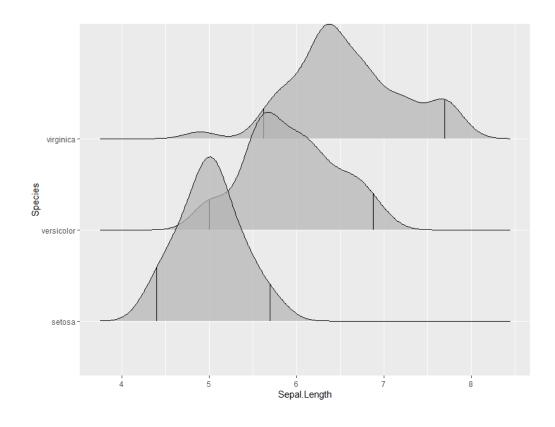
- The extent to which the different densities overlap can be controlled with the scale parameter.
- A setting of scale=1 means the tallest density curve just touches the baseline of the next higher one.
- Smaller values create a separation between the curves, and larger values create more overlap.

```
>ggplot(iris, aes(x = Sepal.Length, y = Species)) + geom_density_ridges(scale = 0.9)
```

Density ridgeline plots

We can also specify quantiles by cut points rather than number. E.g., we can indicate the
 2.5% and 97.5% tails.

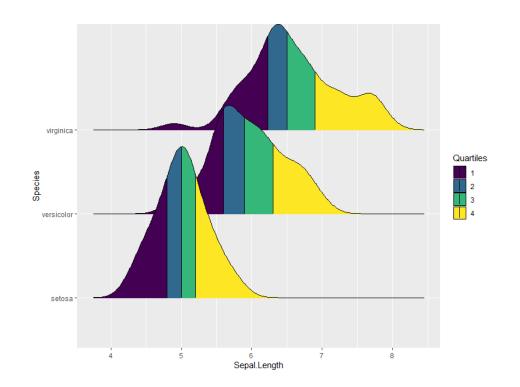
> ggplot(iris, aes(x = Sepal.Length, y = Species)) + stat_density_ridges(quantile_lines = TRUE, quantiles = c(0.025, 0.975), alpha = 0.7)



Quantile plots

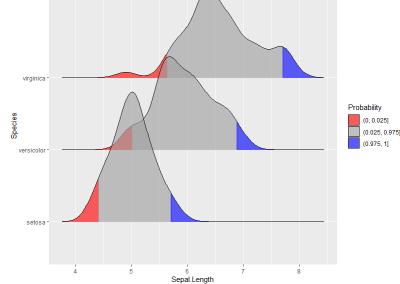
Using the geom geom_density_ridges_gradient we can also color by quantile, via the calculated stat(quantile) aesthetic. Note that this aesthetic is only calculated if calc_ecdf = TRUE.

```
> ggplot(iris, aes(x=Sepal.Length, y=Species, fill = factor(stat(quantile)))) + stat_density_ridges( geom = "density_ridges_gradient", calc_ecdf = TRUE, quantiles = 4, quantile_lines = TRUE ) + scale_fill_viridis_d(name = "Quartiles")
```



We can use the same approach to highlight the tails of the distributions.

```
> ggplot(iris, aes(x = Sepal.Length, y = Species, fill = factor(stat(quantile)))) +
    stat_density_ridges(
    geom = "density_ridges_gradient",
    calc_ecdf = TRUE,
    quantiles = c(0.025, 0.975)
) +
    scale_fill_manual(
    name = "Probability", values = c("#FF0000A0", "#A0A0A0A0", "#0000FFA0"),
    labels = c("(0, 0.025]", "(0.025, 0.975]", "(0.975, 1]")
)
```



Data Transformation

Dplyr pacakge

- dplyr is a grammar of data manipulation, providing a consistent set of verbs that help you solve the most common data manipulation challenges:
- The easiest way to get dplyr is to install the whole tidyverse:

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

• Alternatively, install just dplyr:

```
>install.packages("dplyr")
```

- mutate() adds new variables that are functions of existing variables
- select() picks variables based on their names.
- filter() picks cases based on their values.
- summarise() reduces multiple values down to a single summary.
- arrange() changes the ordering of the rows.

Dplyr pacakge

```
starwars %>% filter(species == "Droid")
```

- Principal Component Analysis (PCA) is a useful technique for exploratory data analysis, allowing you to better visualize the variation present in a dataset with many variables.
- It is particularly helpful in the case of "wide" datasets, where you have many variables for each sample.
- Turning your original variables into a smaller number of "Principal Components"

Principal Component Analysis (PCA)

- PCA is a type of linear transformation on a given data set that has values for a certain number of variables (coordinates) for a certain amount of spaces.
- This linear transformation fits this dataset to a new coordinate system in such a way that the most significant variance is found on the first coordinate.
- In this way, you transform a set of x correlated variables over y samples to a set of p uncorrelated principal components over the same samples.
- Where many variables correlate with one another, they will all contribute strongly to the same principal component.

Principal Component Analysis (PCA)

- Let us use the mtcars dataset, which is built into R.
- This dataset consists of data on 32 models of car, taken from an American motoring magazine (1974 Motor Trend magazine). For each car, you have 11 features, expressed in varying units (US units).
- Compute the Principal Components
- Because PCA works best with numerical data, you'll exclude the two categorical variables (vs and am).

Principal Component Analysis (PCA)

You are left with a matrix of 9 columns and 32 rows, which you pass to the prcomp() function, assigning your output to mtcars.pca.

```
> mtcars.pca <- prcomp(mtcars[,c(1:7,10,11)], center = TRUE,scale. = TRUE)
```

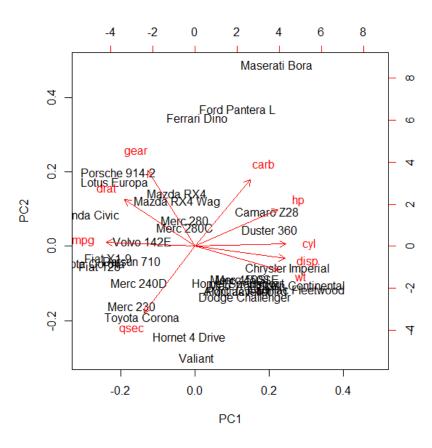
You will also set two arguments, center and scale, to be TRUE. Then you can have a peek at your PCA object with summary().

```
> summary(mtcars.pca)
```

>str(mtcars.pca)

Let us draw biplot

>biplot(mtcars.pca)



Exercise

- install library ggfortify
- Load data iris
 - >data(iris)
- Perform the PCA analysis
- Develop biplot
- List highly correlated variables
- Display graph by use of following command:

```
autoplot(iris.pca, data = iris, colour = 'Species')
autoplot(iris.pca, data = iris, colour = 'Species', label = TRUE, label.size = 3)
```

R Linear Regression

What is Linear Regression?

Regression analysis is a statistical technique for determining the relationship between two or more than two variables.

There are two types of variables in regression analysis –

independent variable and dependent variable.

- Linear Regression is of the following two types:
- Simple Linear Regression Based on the value of the single explanatory variable, the value of the response variable changes.
- Multiple Linear Regression The value is dependent upon more than one explanatory variables in case of multiple linear regression.

R Linear Regression

- Linear regression is one of the most basic statistical models out there.
- A linear regression can be calculated in R with the command lm.
- In the next example, use this command to calculate the height based on the age of the child.

- In this case, "a" and "b" are called the intercept and the slope respectively.
- With the same example, "a" or the intercept, is the value from where you start measuring.
- The slope measures the change of height with respect to the age in months. In general, for every month older the child is, his or her height will increase with "b".

Extracting information from a linear regression

- x <- lm(y~x, data=z)</p>
- summary(x) comprehensive summary of results
- print(x) precise version of the object
- deviance(x) residuals
- plot(x) returns plots: residuals, fitted values and some diagnostics
- coef(x) extract regression coefficients

R Linear Regression: example

Let us download the data to an object called ageandheight and then create the linear regression in the third line. The lm command takes the variables in the format:

Im([target variable] ~ [predictor variables], data = [data source])

- > library(readxl)
- > ageandheight <- read_excel("ageandheight.xls",
 sheet = "Hoja2")</pre>
- > ImHeight = Im(height~age, data = ageandheight)
- > summary(ImHeight) #Review the results

```
> ageandheight <- read excel("ageandheight.xls", sheet = "Hoja2")</pre>
> head(ageandheight)
# A tibble: 6 x 3
    age height no siblings
        <db1>
         76.1
    19
        78.1
    21 78.2
         78.8
> lmHeight = lm(height~age, data = ageandheight)
> summary(lmHeight)
Call:
lm(formula = height ~ age, data = ageandheight)
Residuals:
-66.059 5.193 5.810
Coefficients:
            Estimate Std. Error t value Pr(>|t|)
(Intercept) 64.8445
                        43.3815
                                           0.166
                                  1.495
age
              0.3835
                         1.8264
                                  0.210
                                           0.838
Residual standard error: 21.84 on 10 degrees of freedom
Multiple R-squared: 0.00439, Adjusted R-squared: -0.09517
F-statistic: 0.0441 on 1 and 10 DF, p-value: 0.8379
```

```
Call:
lm(formula = height ~ age, data = ageandheight)
Residuals:
   Min
        10 Median 30 Max
-66.059 5.193 5.810 6.617 7.533
Coefficients:
          Estimate Std. Error t value Pr(>|t|)
(Intercept) 64.8445
                   43.3815 1.495 0.166
           0.3835
                    1.8264 0.210 0.838
age
Residual standard error: 21.84 on 10 degrees of freedom
Multiple R-squared: 0.00439, Adjusted R-squared: -0.09517
F-statistic: 0.0441 on 1 and 10 DF, p-value: 0.8379
```

Coefficients.

In the orange square, you can see the values of the intercept ("a" value) and the slope ("b" value) for the age. These "a" and "b" values plot a line between all the points of the data.

Exercise

- 1) Open Rstudio
- 2) Run small code (given in exercise)
- 3) Select option: html, pdf or doc
- 4) Save file
- 5) Show your reports

PLINK: GWAS DATA

PLINK: Introduction

- PLINK is a free, open-source designed to perform a range of basic, large-scale analyses in a computationally efficient manner.
- PLINK is whole genome association analysis tool.
- PLINK has a well documented manual.
- Available for linux, MAC ansd MAC-DOS.
- Command line version is faster than graphical PLINK.

PLINK: Multi-feature tool

- Merge two or more files
- Extracts subsets (SNPs or individuals)
- Compress data in a binary file format
- PLINK has numerous useful features for managing and analyzing genetic data
- Read data in a variety of formats
- Recode and reorder files

Input Files

- Genotype data is a text file
- Pedigree file (.ped)
- Map file (.map)
- Genotype data is a compressed binary file
- Fam File (.fam)
- Bim file (.bim)
- Bed file (.bed)

PED Input File

Pedigree File - the first six columns are mandatory:

- Family ID
- Individual ID
- Paternal ID
- Maternal ID
- Sex (1=male; 2=female; other=unknown)
- Phenotype

MAP Input File

MAP File has 4 columns:

- chromosome (1-22, X, Y or 0 if unplaced)
- rs# or snp identifier
- Genetic distance (morgans)
- Base-pair position (bp units)

```
Column1 Column2 Column3 Column4
```

Others Input File

*.ped

FID	IID	PID	MID	Sex	Р	rs1	rs2	rs3
2	2	0	0 0	1	0	CC	AA	AC
3	3	0	0	1	1	CC	AA	AC

*.map

Chr	SNP	GD	BPP
1	rs1	0	870000
1	rs2	0	880000
1	rs3	0	890000

*.fam

FID	IID	PID	MID	Sex	P
1	1	0	0	2	1
2	2	0	0	1	0
3	3	0	0	1	1

*.bed

Contains binary version of the SNP info of the *.ped file. (not in a format readable for humans)

*.bim

Chr	SNP	GD	BPP	Allele 1	Allele 2
1	rs1	0	870000	C	T
1	rs2	0	880000	Α	G
1	rs3	0	890000	A	С

Covariate file

FID	IID	C1	C2	C3	_
1	1	0.00812835	0.00606235	-0.000871105	
2	2	-0.0600943	0.0318994	-0.0827743	
3	3	-0.0431903	0.00133068	-0.000276131	

Legend				
FID	Family ID	rs{x}	Alleles per subject per SNP	
IID	Individual ID	Chr	Chromosome	
PID	Paternal ID	SNP	SNP name	
MID	Maternal ID	GD	Genetic distance (morgans)	
Sex	Sex of subject	BPP	Base-pair position (bp units)	
P	Phenotype	C{x}	Covariates (e.g., Multidimensional Scaling (MDS) components)	

Type command : plink --file test –noweb

Provide you detail information of SNPs count, individuals count and check for the missingness and frequency test

```
Options in effect:
        --file test
        --noweb
 (of 2) markers to be included from [ test.map ]
6 individuals read from [ test.ped ]
 individuals with nonmissing phenotypes
Assuming a disease phenotype (1=unaff, 2=aff, 0=miss)
Missing phenotype value is also -9
3 cases, 3 controls and 0 missing
6 males, 0 females, and 0 of unspecified sex
Before frequency and genotyping pruning, there are 2 SNPs
6 founders and 0 non-founders found
Total genotyping rate in remaining individuals is 1
0 SNPs failed missingness test ( GENO > 1 )
0 SNPs failed frequency test ( MAF < 0 )</pre>
After frequency and genotyping pruning, there are 2 SNPs
After filtering, 3 cases, 3 controls and 0 missing
After filtering, 6 males, 0 females, and 0 of unspecified sex
```

Data Transformation

- Convert map and ped into binary format
 - plink.exe --file test --make-bed --out test --noweb
- Read binary file in plink

plink.exe –bfile test

```
Reading map (extended format) from [ test.bim ]
 markers to be included from [ test.bim ]
Reading pedigree information from [ test.fam ]
 individuals read from [ test.fam ]
 individuals with nonmissing phenotypes
Assuming a disease phenotype (1=unaff, 2=aff, 0=miss)
Missing phenotype value is also -9
3 cases, 3 controls and 0 missing
6 males, 0 females, and 0 of unspecified sex
Reading genotype bitfile from [ test.bed ]
Detected that binary PED file is v1.00 SNP-major mode
Before frequency and genotyping pruning, there are 2 SNPs
6 founders and 0 non-founders found
Total genotyping rate in remaining individuals is 1
0 SNPs failed missingness test ( GENO > 1 )
SNPs failed frequency test ( MAF < 0 )
After frequency and genotyping pruning, there are 2 SNPs
After filtering, 3 cases, 3 controls and 0 missing
After filtering, 6 males, 0 females, and 0 of unspecified sex
Analysis finished: Sun Oct 20 18:49:10 2019
```

Comparision

```
2 (of 2) markers to be included from [ test.map ]
6 individuals read from [ test.ped ]
6 individuals with nonmissing phenotypes
Assuming a disease phenotype (1=unaff, 2=aff, 0=miss)
Missing phenotype value is also -9
3 cases, 3 controls and 0 missing
6 males, 0 females, and 0 of unspecified sex
Before frequency and genotyping pruning, there are 2 SNPs
6 founders and 0 non-founders found
Total genotyping rate in remaining individuals is 1
0 SNPs failed missingness test ( GENO > 1 )
0 SNPs failed frequency test ( MAF < 0 )
After frequency and genotyping pruning, there are 2 SNPs
After filtering, 3 cases, 3 controls and 0 missing
After filtering, 6 males, 0 females, and 0 of unspecified sex
```

TEXT output

```
Reading map (extended format) from [ test.bim ]
2 markers to be included from [ test.bim ]
Reading pedigree information from [ test.fam ]
6 individuals read from [ test.fam ]
6 individuals with nonmissing phenotypes
Assuming a disease phenotype (1=unaff, 2=aff, 0=miss)
Missing phenotype value is also -9
3 cases, 3 controls and 0 missing
6 males, 0 females, and 0 of unspecified sex
Reading genotype bitfile from [ test.bed ]
Detected that binary PED file is v1.00 SNP-major mode
Before frequency and genotyping pruning, there are 2 SNPs
6 founders and 0 non-founders found
Total genotyping rate in remaining individuals is 1
0 SNPs failed missingness test ( GENO > 1 )
0 SNPs failed frequency test ( MAF < 0 )
After frequency and genotyping pruning, there are 2 SNPs
After filtering, 3 cases, 3 controls and 0 missing
After filtering, 6 males, 0 females, and 0 of unspecified sex
Analysis finished: Sun Oct 20 18:49:10 2019
```

BED OUTPUT

QC of genetic DATA

- A vital step that should be part of any GWAS is the use of appropriate QC.
- Without extensive QC, GWAS will not generate reliable results because raw genotype data are inherently imperfect.
- Errors in the data can arise for numerous reasons, for example, due to poor quality of DNA samples, poor DNA hybridization to the array, poorly performing genotype probes, and sample mix-ups or contamination.

QC of genetic DATA

The seven QC steps consist of filtering out of SNPs and individuals based on the following:

- (1)individual and SNP missingness,
- (2) inconsistencies in assigned and genetic sex of subjects (see sex discrepancy),
- (3) minor allele frequency (MAF),
- (4) deviations from Hardy–Weinberg equilibrium (HWE),
- (5) heterozygosity rate,
- (6) relatedness, and
- (7) ethnic outliers (see population stratification).

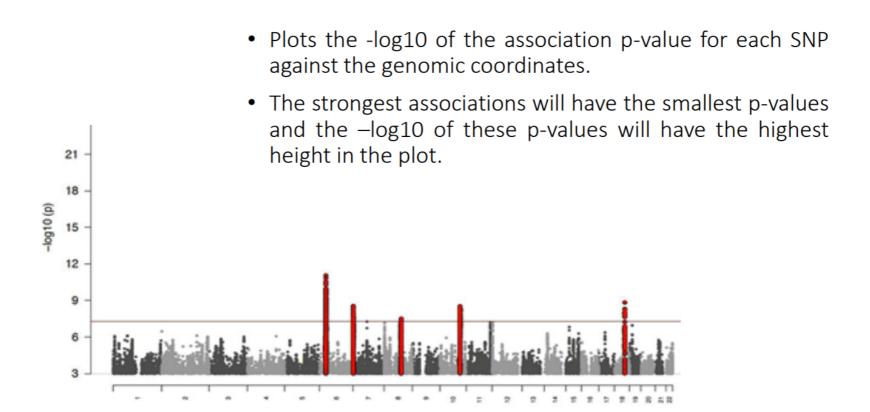
Important Commands

Step	Command	Function
1: Missingness of SNPs and individuals	geno	Excludes SNPs that are missing in a large proportion of the subjects. In this step, SNPs with low genotype calls are removed.
	mind	Excludes individuals who have high rates of genotype missingness. In this step, individual with low genotype calls are removed.
2: Sex discrepancy	check-sex	Checks for discrepancies between sex of the individuals recorded in the dataset and their sex based on X chromosome heterozygosity/homozygosity rates.
3: Minor allele frequency (MAF)	maf	Includes only SNPs above the set MAF threshold.
4: Hardy–Weinberg equilibrium (HWE)	hwe	Excludes markers which deviate from Hardy–Weinberg equilibrium.
5: Heterozygosity		Excludes individuals with high or low heterozygosity rates
	genome	Calculates identity by descent (IBD) of all sample pairs.
6: Relatedness	min	Sets threshold and creates a list of individuals with relatedness above the chosen threshold. Meaning that subjects who are related at, for example, pi-hat >0.2 (i.e., second degree relatives) can be detected.
7: Population	genome	Calculates identity by descent (IBD) of all sample pairs.
stratification	clustermds-plot k Genetics an	Produces a k-dimensional representation of any substructure in the disjoinformatics and data, based on IBS.

- -- assoc
- It works with case/control and continuous phenotypes.
- Case-control (1df chi-square test, outputs assoc)
- PLINK will recognise this is a case/control analysis because the phenotype just has:
- 1 (for controls),
- 2 (for cases), and
- 0/-9/non-numeric (for missing).

 Given a quantitative phenotype, --assoc writes regression statistics and Wald test results to <u>plink.qassoc</u>.

Manhattan plot



Useful Link

https://jokergoo.github.io/circlize_book/book/

https://igraph.org/r/doc/aaa-igraph-package.html

https://www.statmethods.net/advgraphs/correlograms.html