

Introduction to R and Rstudio

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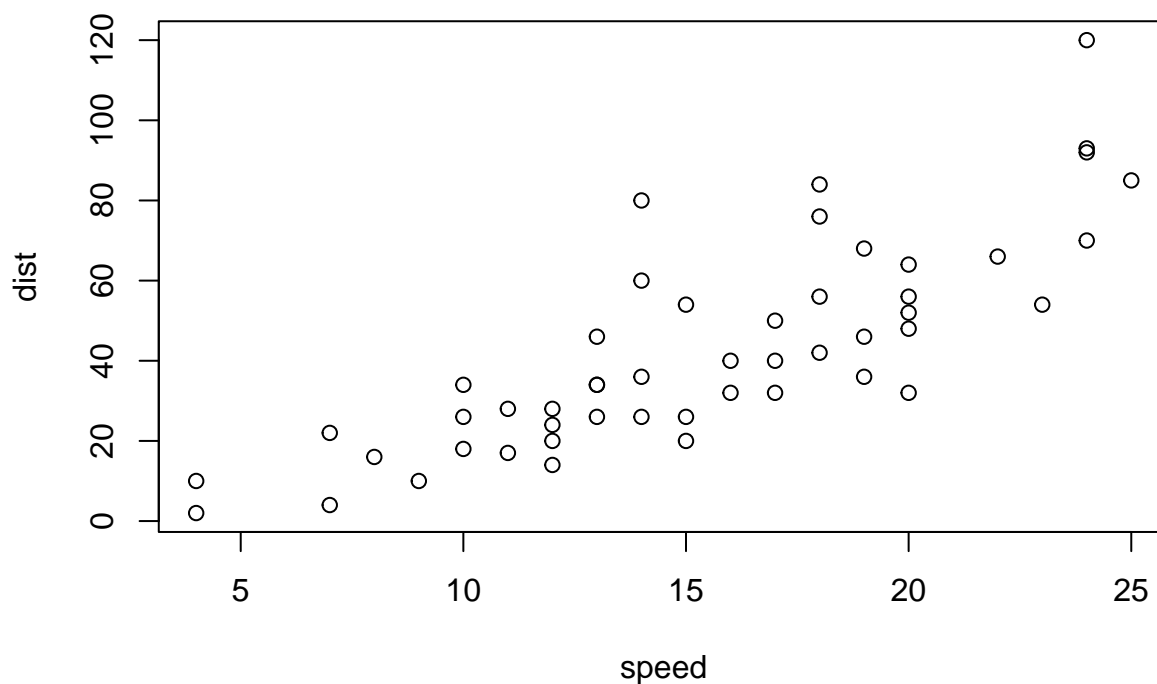
This is an R Markdown Notebook. Markdown is a lightweight markup language for creating formatted text using a plain-text editor. For more information see the book *R Markdown: The Definitive Guide*.

When you execute code within the notebook, the results appear beneath the code.

Try executing this chunk by clicking the *Run* button within the chunk or by placing your cursor inside it and pressing *Cmd+Shift+Enter*.

Example:

```
plot(cars)
```



Add a new chunk by clicking the *Insert Chunk* button on the toolbar or by pressing *Cmd+Option+I* (macOS) or *Ctrl+Alt+I* (Windows).

When you save the notebook, an HTML file containing the code and output will be saved alongside it (click the *Preview* button or press *Cmd+Shift+K* to preview the HTML file).

The preview shows you a rendered HTML copy of the contents of the editor. Consequently, unlike *Knit*, *Preview* does not run any R code chunks. Instead, the output of the chunk when it was last run in the editor is displayed.

Basic operations

```
6+6 # sums two numbers. The answer will appear underneath:
```

```
## [1] 12
```

```
6-6
```

```
## [1] 0
```

```
6*6
```

```
## [1] 36
```

```
6/6
```

```
## [1] 1
```

```
log(6)
```

```
## [1] 1.791759
```

```
log2(6)
```

```
## [1] 2.584963
```

```
log10(6)
```

```
## [1] 0.7781513
```

```
log(6,7)
```

```
## [1] 0.9207822
```

```
log(6,3)
```

```
## [1] 1.63093
```

```
6^6
```

```
## [1] 46656
```

```
sin(pi/2)
```

```
## [1] 1
```

```
cos(pi/2)
```

```
## [1] 6.123234e-17
```

```
# ETC
```

The hashtag # can be used inside chunks for commenting.

Define a variable with <- You can also use an equal sign, but this is not recommended. The “arrow” makes the code easier to read.

I can write **things!**

```
mysum <- 6+6 #sum two integers and assign it to a variable
```

```
mysum = 6+6 #same as above
```

```
mysum #check variable content by executing "mysum"
```

```
## [1] 12
```

Check the list of variables defined so far

```
ls()
```

```
## [1] "mysum"
```

The same variables are also listed in the *Environment* panel of RStudio

Remove a variable:

```
rm(mysum)
```

```
rm(list=ls()) # NB will remove all variables!
```

Objects in R:

R consists of a number of different data objects. Some of the most important are vectors, lists, matrices, and data frames. **Vectors** are one-dimensional and contain values of the same type (logical, integer, character, numeric etc.). **Lists** are data objects of R that contain various types of elements including strings, numbers, vectors, and a nested list inside it. It can also consist of matrices or functions as elements. **Matrices** are two-dimensional layout data with elements of the same data type. They usually contain numeric values in order to perform mathematical operations.

Data frame is a 2-dimensional data structure wherein each column consists of the value of one variable and each row consists of a value set from each column. Each column can be of separate types.

Data types:

Numeric: numbers with decimals

```
mynumber<-66.6
```

```
print(mynumber)
```

```
## [1] 66.6
```

```
mynumber
```

```
## [1] 66.6
```

```
class(mynumber) # check the type
```

```
## [1] "numeric"
```

```
#Integer: numbers with no decimals
```

```
mynumber.int<-as.integer(mynumber)
```

```
class(mynumber.int)
```

```
## [1] "integer"
```

```
mynumber.int
```

```
## [1] 66
```

Character: can be a letter or a combination of letters enclosed by quotes

```
mychar<-"bioinfo course"
```

```
mychar
```

```
## [1] "bioinfo course"
```

```
str(mychar)
```

```
## chr "bioinfo course"
```

```
class(mychar)
```

```
## [1] "character"
```

Logical: a variable that can be TRUE or FALSE (boolean)

```
im.true<-TRUE  
im.true<-FALSE  
im.true
```

```
## [1] FALSE
```

```
class(im.true)
```

```
## [1] "logical"
```

Factor: used to refer to a qualitative relationship. to generate a factor, we'll use a vector defined with the function `c()`

```
myfactor<-factor(c("good", "bad", "ugly", "good", "good", "bad", "ugly", "stupid"))  
myfactor
```

```
## [1] good  bad   ugly  good  good  bad   ugly  stupid  
## Levels: bad good stupid ugly
```

```
class(myfactor)
```

```
## [1] "factor"
```

```
levels(myfactor) # this can be used to check the levels of a factor
```

```
## [1] "bad"    "good"    "stupid"  "ugly"
```

```
nlevels(myfactor)
```

```
## [1] 4
```

```
class(levels(myfactor))
```

```
## [1] "character"
```

```
?nlevels
```

Lists It can contain elements of various data types (e.g.vectors,functions,matrices,another list) Example with three different data types in one list

```
list1<-c(1:15) # integer vector  
list2<-factor(1:5) # factor vector  
list3<-letters[1:5]  
grouped.lists<-list(list1,list2,list3)  
grouped.lists
```

```
## [[1]]  
## [1]  1  2  3  4  5  6  7  8  9 10 11 12 13 14 15  
##  
## [[2]]  
## [1] 1 2 3 4 5  
## Levels: 1 2 3 4 5  
##  
## [[3]]  
## [1] "a" "b" "c" "d" "e"
```

```
str(grouped.lists)
```

```
## List of 3
## $ : int [1:15] 1 2 3 4 5 6 7 8 9 10 ...
## $ : Factor w/ 5 levels "1","2","3","4",...: 1 2 3 4 5
## $ : chr [1:5] "a" "b" "c" "d" ...
```

Accessing elements of a list

```
grouped.lists[[2]] # accessing the first vector
```

```
## [1] 1 2 3 4 5
## Levels: 1 2 3 4 5
```

```
grouped.lists[[3]][5] # accessing the 5th element from the third vector
```

```
## [1] "e"
```

Ungroup the list

```
ungrouped.list<-unlist(grouped.lists)
class(ungrouped.list) # NB: the list becomes a character datatype. WHY?
```

```
## [1] "character"
```

```
length(ungrouped.list)
```

```
## [1] 25
```

```
length(grouped.lists)
```

```
## [1] 3
```

Vectors Objects that are used to store values or other information of the same data type They are created with the function “c()” that will generate a 1D array

```
species<-c(123,434,655,877,986) # we create a numeric vector
class(species)
```

```
## [1] "numeric"
```

```
length(species) # number of elements in the vector
```

```
## [1] 5
```

```
species[1] # accessing the fifth element in the vector
```

```
## [1] 123
```

```
species[1:3]
```

```
## [1] 123 434 655
```

```
sum(species) # sum the values in the vector
```

```
## [1] 3075
```

```
species.names<-c("dog","lion","human","pig","cow") # we create a character vector
class(species.names)
```

```
## [1] "character"
```

Create a sequence of numbers:

```
seq.num<-c(1:100)
seq.num
```

```
## [1] 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18
## [19] 19 20 21 22 23 24 25 26 27 28 29 30 31 32 33 34 35 36
## [37] 37 38 39 40 41 42 43 44 45 46 47 48 49 50 51 52 53 54
## [55] 55 56 57 58 59 60 61 62 63 64 65 66 67 68 69 70 71 72
## [73] 73 74 75 76 77 78 79 80 81 82 83 84 85 86 87 88 89 90
## [91] 91 92 93 94 95 96 97 98 99 100
```

```
sum(seq.num)
```

```
## [1] 5050
```

```
seq.num.by2 <- seq(1,100, 2)
```

we can access the 5th to the 10th element using the :

```
seq.num.by2[5:10]
```

```
## [1] 9 11 13 15 17 19
```

Matrix

Like a vector, a matrix stores information of the same data type, but it has 2 dimensions.

syntax for generating a matrix:

```
mymatrix <- matrix(vector, nrow=r, ncol=c, byrow=FALSE, dimnames=list(char_vector_rownames,
char_vector_colnames))
```

byrow=F indicates that the matrix should be filled by columns

```
mymatrix <- matrix(seq(1:100), nrow=10, ncol=10, byrow=FALSE, dimnames=list(c(3:12), letters[1:10]))
mymatrix
```

```
## a b c d e f g h i j
## 3 1 11 21 31 41 51 61 71 81 91
## 4 2 12 22 32 42 52 62 72 82 92
## 5 3 13 23 33 43 53 63 73 83 93
## 6 4 14 24 34 44 54 64 74 84 94
## 7 5 15 25 35 45 55 65 75 85 95
## 8 6 16 26 36 46 56 66 76 86 96
## 9 7 17 27 37 47 57 67 77 87 97
## 10 8 18 28 38 48 58 68 78 88 98
## 11 9 19 29 39 49 59 69 79 89 99
## 12 10 20 30 40 50 60 70 80 90 100
```

A matrix with random numbers:

```
mymatrix.rand <- matrix(sample (seq(1:100),100), nrow=10, ncol=10, byrow=FALSE, dimnames=list(c(1:10), 1:10))
mymatrix.rand
```

```
## a b c d e f g h i j
## 1 39 57 96 94 78 79 14 16 36 35
## 2 59 47 75 68 11 100 84 21 19 72
## 3 76 74 70 89 66 45 23 63 32 43
## 4 44 92 34 2 10 17 22 48 24 27
## 5 56 53 46 20 97 54 51 71 69 1
## 6 5 73 80 90 86 4 6 82 52 3
## 7 37 15 33 99 65 77 31 7 38 8
```

```
## 8  93 41 28 67 81  12 98 55 85 60
## 9  91 58  9 95 18  30 61 64 42 87
## 10 40 26 29 83 50  13 88 49 25 62
```

```
mymatrix[3:6,c(4,7)] # We select what sections of the matrix we want to look at
```

```
##      d  g
## 5 33 63
## 6 34 64
## 7 35 65
## 8 36 66
```

```
mymatrix[3:6,c("b","c")]
```

```
##      b  c
## 5 13 23
## 6 14 24
## 7 15 25
## 8 16 26
```

Dataframes are more general than a matrix and can contain different data types.

Variables or features are in columns, while observations are in rows.

=>NB: this is one of the most common objects in metabarcoding analyses<= Generated with the `data.frame()` function

```
my.data.frame<-data.frame(
  Name=c("Game of Thrones","MrRobot","WestWorld"),
  Budget=c(344,59,122),
  Seasons=c(8,4,3),
  Audience=c(300,14,80),
  Actors=c(221,56, 90)
)
print(my.data.frame)
```

```
##              Name Budget Seasons Audience Actors
## 1 Game of Thrones   344      8     300     221
## 2      MrRobot     59      4      14      56
## 3    WestWorld    122      3      80      90
```

```
row.names(my.data.frame)<-my.data.frame[,1] # Assign to the row names the names in the first column
my.data.frame<-my.data.frame[,-1] # Remove the first column
my.data.frame # By clicking this object in the "Environment" panel on the right, you'll see a window wi
```

```
##              Budget Seasons Audience Actors
## Game of Thrones   344      8     300     221
## MrRobot          59      4      14      56
## WestWorld        122      3      80      90
```

```
class(my.data.frame)
```

```
## [1] "data.frame"
```

```
ncol(my.data.frame) # Number of columns
```

```
## [1] 4
```

```
nrow(my.data.frame) # Number of rows
```

```
## [1] 3
```

```
colnames(my.data.frame) # Column names
```

```
## [1] "Budget" "Seasons" "Audience" "Actors"
```

```
rownames(my.data.frame) # Name of rows
```

```
## [1] "Game of Thrones" "MrRobot" "WestWorld"
```

```
colSums(my.data.frame) # Sum values in columns
```

```
## Budget Seasons Audience Actors  
## 525 15 394 367
```

```
rowSums(my.data.frame) # We sum the values, even if they make no sense in the example
```

```
## Game of Thrones MrRobot WestWorld  
## 873 133 295
```

```
rbind(my.data.frame,my.data.frame) # appends dataframes one below the other (column names identical)
```

```
## Budget Seasons Audience Actors  
## Game of Thrones 344 8 300 221  
## MrRobot 59 4 14 56  
## WestWorld 122 3 80 90  
## Game of Thrones1 344 8 300 221  
## MrRobot1 59 4 14 56  
## WestWorld1 122 3 80 90
```

```
cbind(my.data.frame,my.data.frame) # appends dataframes one next to the other (row names identical)
```

```
## Budget Seasons Audience Actors Budget Seasons Audience Actors  
## Game of Thrones 344 8 300 221 344 8 300 221  
## MrRobot 59 4 14 56 59 4 14 56  
## WestWorld 122 3 80 90 122 3 80 90
```

```
head(my.data.frame, 2) # Useful to have a look to the beginning of the dataframe (specially useful in b
```

```
## Budget Seasons Audience Actors  
## Game of Thrones 344 8 300 221  
## MrRobot 59 4 14 56
```

```
my.data.frame[1:2,2:4] # Useful to look at specific sections of the dataframe
```

```
## Seasons Audience Actors  
## Game of Thrones 8 300 221  
## MrRobot 4 14 56
```

Let's generate a dataframe with different data types

```
my.data.frame.2<-data.frame(  
  Name=c("Game of Thrones","MrRobot","WestWorld", "Chernobyl"),  
  Rating=c("Excellent","Very Good","Excellent", "Very Good"),  
  Audience.Restriction=c(TRUE,FALSE,TRUE, FALSE)  
)  
my.data.frame.2
```

```
## Name Rating Audience.Restriction  
## 1 Game of Thrones Excellent TRUE  
## 2 MrRobot Very Good FALSE  
## 3 WestWorld Excellent TRUE
```



```
## 4          Chernobyl Very Good          FALSE
```

Rename the row names

```
row.names(my.data.frame.2)<-my.data.frame.2[,1]
my.data.frame.2
```

```
##              Name      Rating Audience.Restriction
## Game of Thrones Game of Thrones Excellent          TRUE
## MrRobot          MrRobot Very Good          FALSE
## WestWorld        WestWorld Excellent          TRUE
## Chernobyl        Chernobyl Very Good          FALSE
```

```
my.data.frame.2<-my.data.frame.2[,-1] # Remove redundant column 1
my.data.frame.2
```

```
##              Rating Audience.Restriction
## Game of Thrones Excellent          TRUE
## MrRobot          Very Good          FALSE
## WestWorld        Excellent          TRUE
## Chernobyl        Very Good          FALSE
```

```
str(my.data.frame.2) # Let's look at the data types within this dataframe
```

```
## 'data.frame':    4 obs. of  2 variables:
## $ Rating          : chr  "Excellent" "Very Good" "Excellent" "Very Good"
## $ Audience.Restriction: logi  TRUE FALSE TRUE FALSE
```

Variables in this case are characters and logical (TRUE/FALSE)

#Merge two dataframes based in a pattern # We will use the series names to merge these dataframes as this is what they have in common

```
data.frame.large<-merge(my.data.frame, my.data.frame.2, by="row.names") # "by" indicates the column use
```

#Useful commands to work with tables or dataframes

```
getwd() # get working directory
```

```
## [1] "/Users/anderkkr/Dropbox/Projects/00_Master_projects/21_undervisning/2021/UNIS_AB332_prep"
```

You can change the working directory:

```
#setwd("path/to/my/directory") # set working directory
```

```
str(data.frame.large)
```

```
## 'data.frame':    3 obs. of  7 variables:
## $ Row.names      : 'AsIs' chr  "Game of Thrones" "MrRobot" "WestWorld"
## $ Budget         : num  344 59 122
## $ Seasons        : num  8 4 3
## $ Audience       : num  300 14 80
## $ Actors         : num  221 56 90
## $ Rating         : chr  "Excellent" "Very Good" "Excellent"
## $ Audience.Restriction: logi  TRUE FALSE TRUE
```

```
dim(data.frame.large) # Table dimensions
```

```
## [1] 3 7
```

```
nrow(data.frame.large) # Number of rows

## [1] 3

ncol(data.frame.large) # Number of columns

## [1] 7

colnames(data.frame.large) # Name of columns

## [1] "Row.names"          "Budget"              "Seasons"
## [4] "Audience"          "Actors"              "Rating"
## [7] "Audience.Restriction"

rownames(data.frame.large) # Name of rows

## [1] "1" "2" "3"

#colSums(data.frame.large[]) # Sum of numeric values in columns
#rowSums(data.frame.large) # Sum of numeric values in rows
head(data.frame.large) # See table header

##      Row.names Budget Seasons Audience Actors   Rating Audience.Restriction
## 1 Game of Thrones   344      8     300    221 Excellent                TRUE
## 2      MrRobot     59      4      14     56 Very Good                FALSE
## 3    WestWorld    122      3      80     90 Excellent                TRUE

t(data.frame.large) # Transpose table

##      [,1]      [,2]      [,3]
## Row.names "Game of Thrones" "MrRobot" "WestWorld"
## Budget    "344"          " 59"    "122"
## Seasons    "8"           "4"      "3"
## Audience   "300"          " 14"    " 80"
## Actors     "221"          " 56"    " 90"
## Rating     "Excellent"    "Very Good" "Excellent"
## Audience.Restriction "TRUE"      "FALSE"   "TRUE"
```

Table subsetting Format: `my.table[row, column]`

Replace *my.table* with a data frame in the following and see if you understand the different operations:

```
my.table[1,2] #Get value from row 1, column 2
my.table[1,]  #Get values from row 1 across all columns
my.table$column.name<-NULL #Remove column
my.table[-5,-2] # Remove row 5 and column 2
my.table[-(5:10),] # Remove rows 5 to 10, keep all columns
my.table[,-(which(colSums(my.table)==0))] # Remove columns that sum 0
```

Installing packages

R has a large repository of packages for different applications

```
install.packages("vegan") # Installs the community ecology package Vegan with hundreds of functions
library("vegan") # load Vegan
```

#Plotting

```
data("mtcars") # We load a dataset that comes with R
mtcars
```

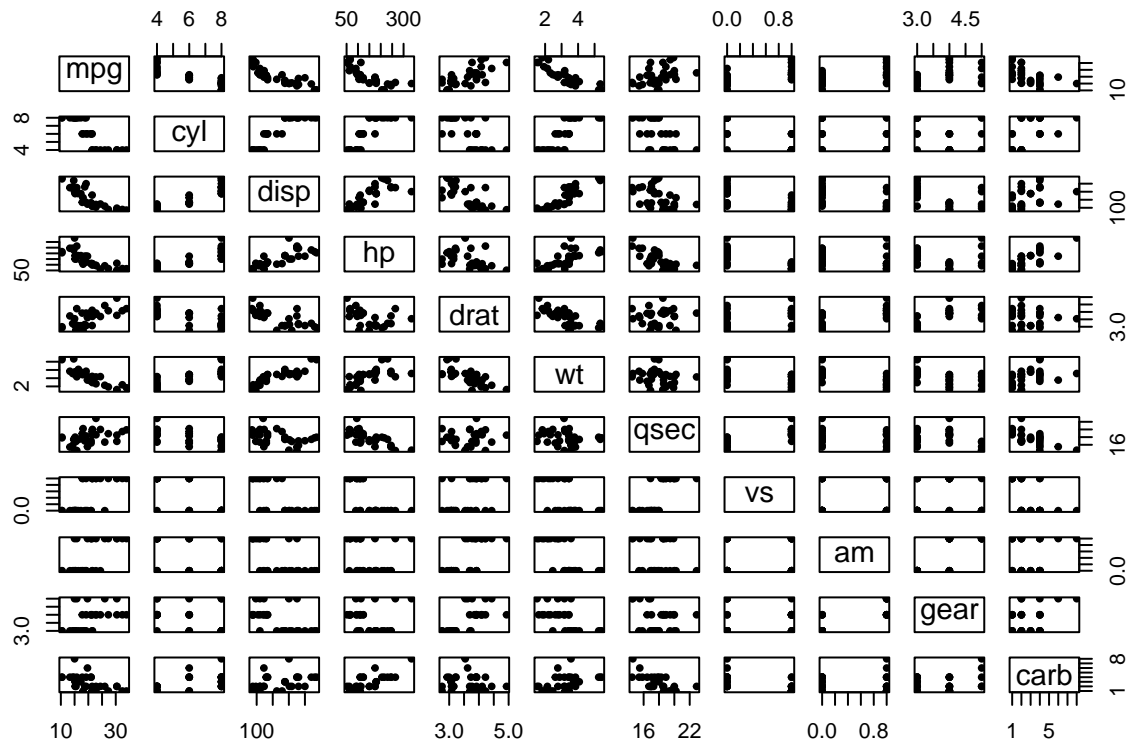
##	mpg	cyl	disp	hp	drat	wt	qsec	vs	am	gear	carb
## Mazda RX4	21.0	6	160.0	110	3.90	2.620	16.46	0	1	4	4
## Mazda RX4 Wag	21.0	6	160.0	110	3.90	2.875	17.02	0	1	4	4
## Datsun 710	22.8	4	108.0	93	3.85	2.320	18.61	1	1	4	1
## Hornet 4 Drive	21.4	6	258.0	110	3.08	3.215	19.44	1	0	3	1
## Hornet Sportabout	18.7	8	360.0	175	3.15	3.440	17.02	0	0	3	2
## Valiant	18.1	6	225.0	105	2.76	3.460	20.22	1	0	3	1
## Duster 360	14.3	8	360.0	245	3.21	3.570	15.84	0	0	3	4
## Merc 240D	24.4	4	146.7	62	3.69	3.190	20.00	1	0	4	2
## Merc 230	22.8	4	140.8	95	3.92	3.150	22.90	1	0	4	2
## Merc 280	19.2	6	167.6	123	3.92	3.440	18.30	1	0	4	4
## Merc 280C	17.8	6	167.6	123	3.92	3.440	18.90	1	0	4	4
## Merc 450SE	16.4	8	275.8	180	3.07	4.070	17.40	0	0	3	3
## Merc 450SL	17.3	8	275.8	180	3.07	3.730	17.60	0	0	3	3
## Merc 450SLC	15.2	8	275.8	180	3.07	3.780	18.00	0	0	3	3
## Cadillac Fleetwood	10.4	8	472.0	205	2.93	5.250	17.98	0	0	3	4
## Lincoln Continental	10.4	8	460.0	215	3.00	5.424	17.82	0	0	3	4
## Chrysler Imperial	14.7	8	440.0	230	3.23	5.345	17.42	0	0	3	4
## Fiat 128	32.4	4	78.7	66	4.08	2.200	19.47	1	1	4	1
## Honda Civic	30.4	4	75.7	52	4.93	1.615	18.52	1	1	4	2
## Toyota Corolla	33.9	4	71.1	65	4.22	1.835	19.90	1	1	4	1
## Toyota Corona	21.5	4	120.1	97	3.70	2.465	20.01	1	0	3	1
## Dodge Challenger	15.5	8	318.0	150	2.76	3.520	16.87	0	0	3	2
## AMC Javelin	15.2	8	304.0	150	3.15	3.435	17.30	0	0	3	2
## Camaro Z28	13.3	8	350.0	245	3.73	3.840	15.41	0	0	3	4
## Pontiac Firebird	19.2	8	400.0	175	3.08	3.845	17.05	0	0	3	2
## Fiat X1-9	27.3	4	79.0	66	4.08	1.935	18.90	1	1	4	1
## Porsche 914-2	26.0	4	120.3	91	4.43	2.140	16.70	0	1	5	2
## Lotus Europa	30.4	4	95.1	113	3.77	1.513	16.90	1	1	5	2
## Ford Pantera L	15.8	8	351.0	264	4.22	3.170	14.50	0	1	5	4
## Ferrari Dino	19.7	6	145.0	175	3.62	2.770	15.50	0	1	5	6
## Maserati Bora	15.0	8	301.0	335	3.54	3.570	14.60	0	1	5	8
## Volvo 142E	21.4	4	121.0	109	4.11	2.780	18.60	1	1	4	2

The data was extracted from the 1974 Motor Trend US magazine, and comprises fuel consumption and 10 aspects of automobile design and performance for 32 automobiles (1973–74 models).

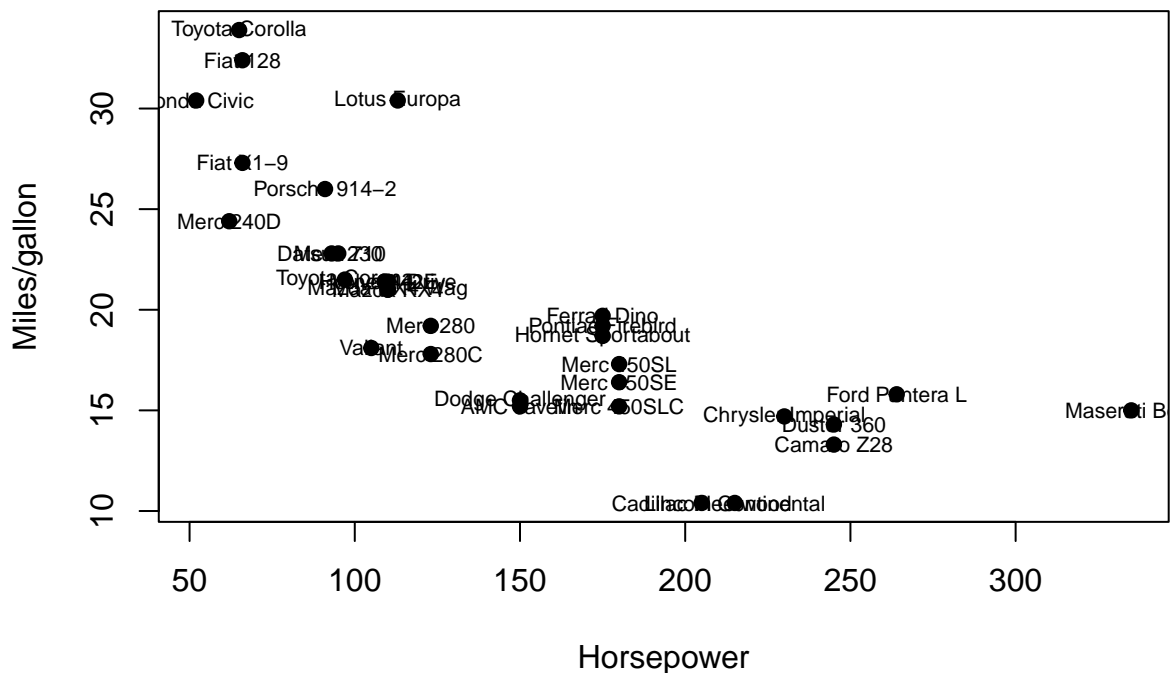
```
# [, 1] mpg Miles/(US) gallon
# [, 2] cyl Number of cylinders
# [, 3] disp Displacement (cu.in.)
# [, 4] hp Gross horsepower
# [, 5] drat Rear axle ratio
# [, 6] wt Weight (1000 lbs)
# [, 7] qsec 1/4 mile time
# [, 8] vs Engine (0 = V-shaped, 1 = straight)
# [, 9] am Transmission (0 = automatic, 1 = manual)
# [,10] gear Number of forward gears
# [,11] carb Number of carburetors

#boxplot(mtcars) # make a boxplot of variables across car models

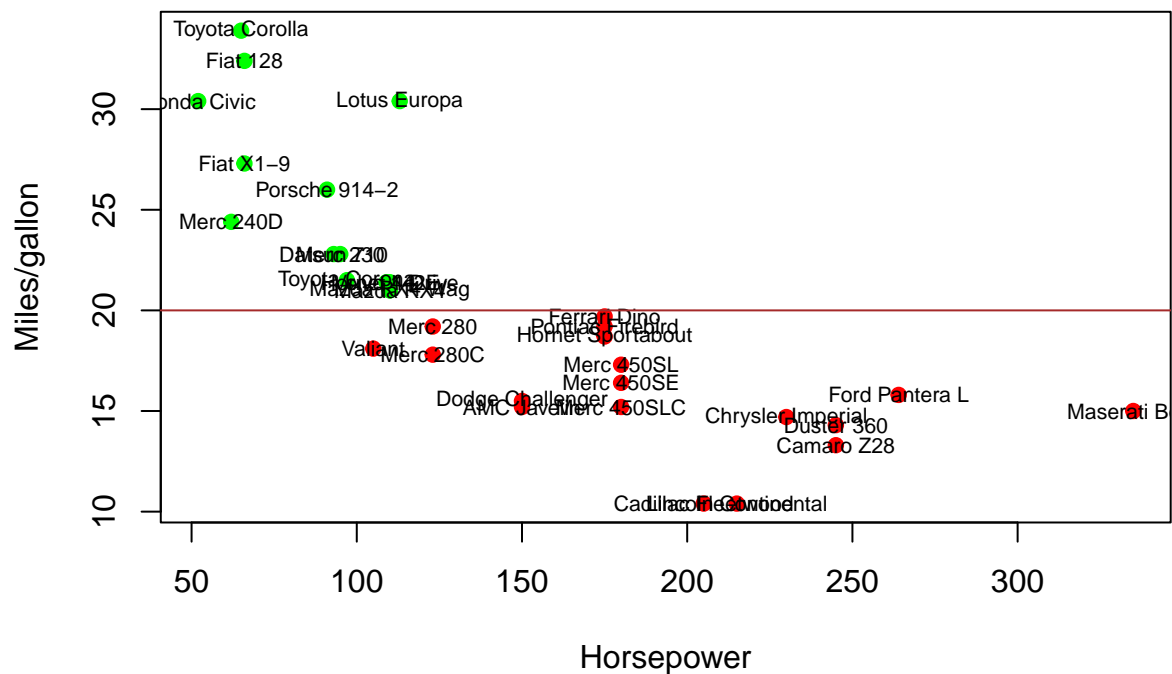
plot(mtcars, pch=19, cex=0.6) # make x-y plots for all variables
```



```
plot(mtcars$hp, mtcars$mpg, xlab="Horsepower", ylab="Miles/gallon", pch=19) # we plot horsepower vs. miles per gallon
text(mtcars$hp, mtcars$mpg, row.names(mtcars), cex=0.7) # we add the car model
```

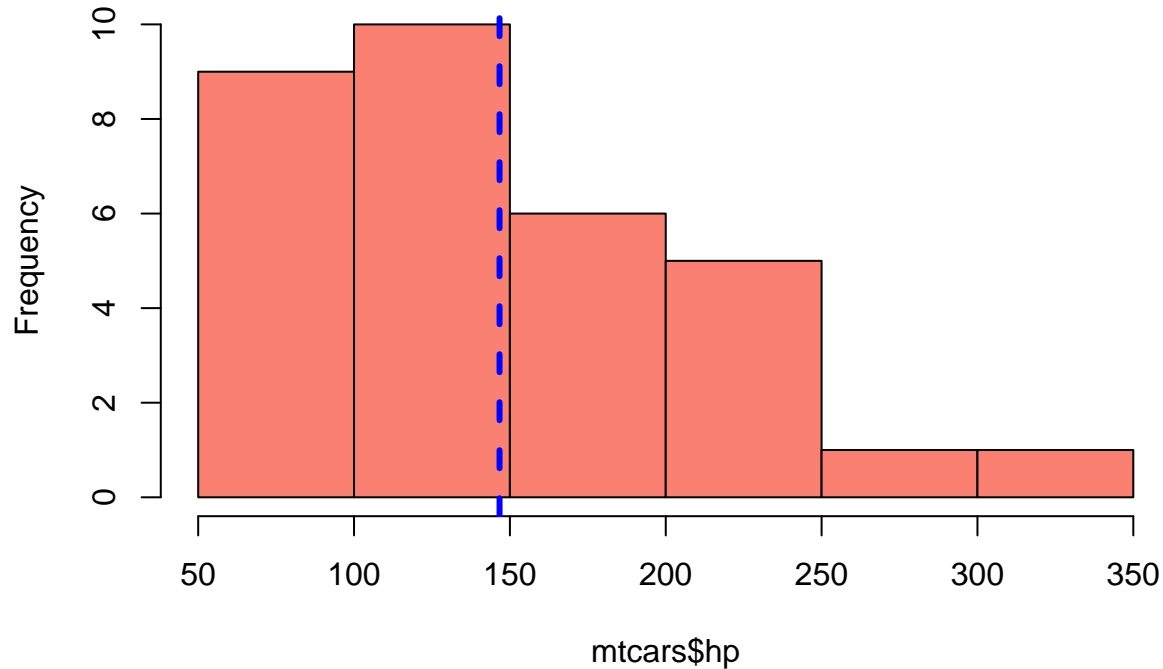


```
plot(mtcars$hp, mtcars$mpg, xlab="Horsepower", ylab="Miles/gallon", pch=19, col=ifelse(mtcars$mpg<20,"red", "green"))
text(mtcars$hp, mtcars$mpg, row.names(mtcars), cex=0.7) # we add the car model
abline(h=20, col="brown") # add a line to the plot
```



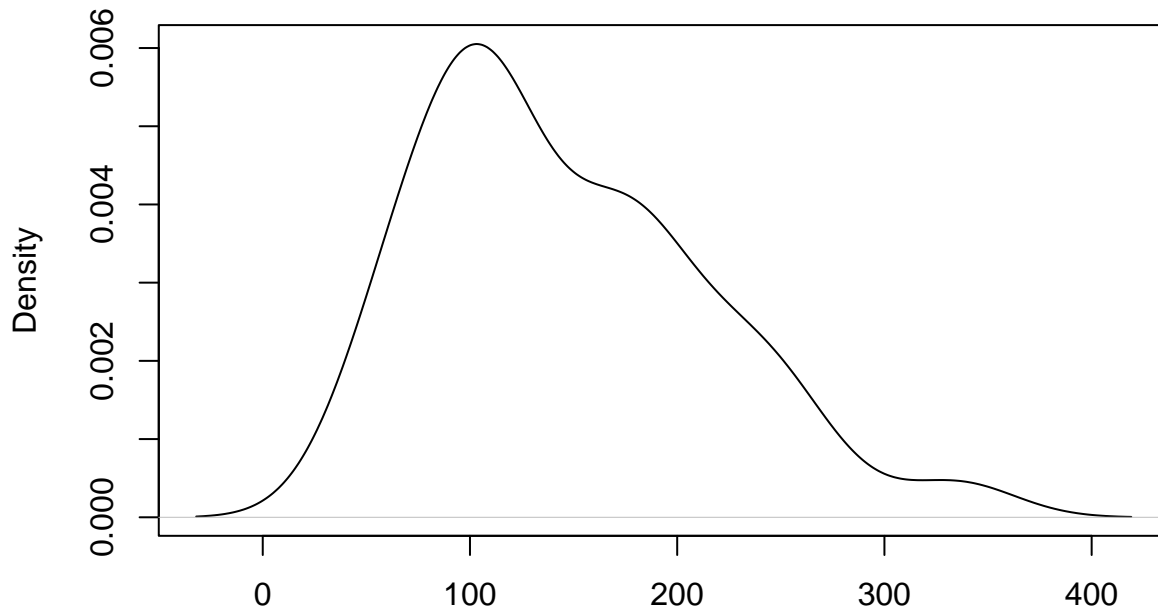
```
hist(mtcars$hp, col="salmon")
abline(v=mean(mtcars$hp), col="blue", lwd=3, lty=2)
```

Histogram of mtcars\$hp



```
plot(density(mtcars$hp))
```

density.default(x = mtcars\$hp)

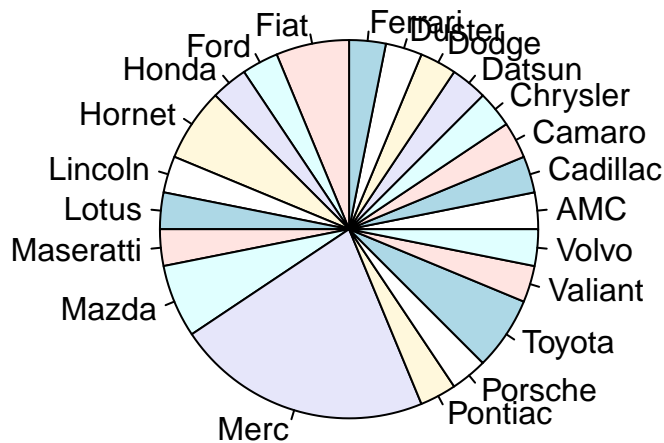


N = 32 Bandwidth = 28.04

```
brands<-c("Mazda", "Mazda", "Datsun", "Hornet", "Hornet", "Valiant", "Duster", "Merc", "Merc", "Merc", "I
        "Honda", "Toyota", "Toyota", "Dodge", "AMC", "Camaro", "Pontiac", "Fiat", "Porsche", "Lotus", "For
mtcars$brand<-brands # we add an extra column with brands
mtcars[1:5,] # let's check
```

```
##          mpg  cyl  disp  hp  drat   wt  qsec vs  am  gear  carb  brand
## Mazda RX4      21.0   6  160  110 3.90 2.620 16.46 0  1    4    4  Mazda
## Mazda RX4 Wag  21.0   6  160  110 3.90 2.875 17.02 0  1    4    4  Mazda
## Datsun 710     22.8   4  108   93 3.85 2.320 18.61 1  1    4    1  Datsun
## Hornet 4 Drive  21.4   6  258  110 3.08 3.215 19.44 1  0    3    1  Hornet
## Hornet Sportabout 18.7   8  360  175 3.15 3.440 17.02 0  0    3    2  Hornet
```

```
pie(table(mtcars$brand)) # we make a piechart of the brands
```



For plotting the package ggplot2 is very becoming increasingly popular: It is already included in the package tidyverse, but can also be installed by itself:

```
#install.packages("ggplot2")
#library(ggplot2)
```

Some tips and hits

Use “?” before any function to get the help page.

```
?sum
```

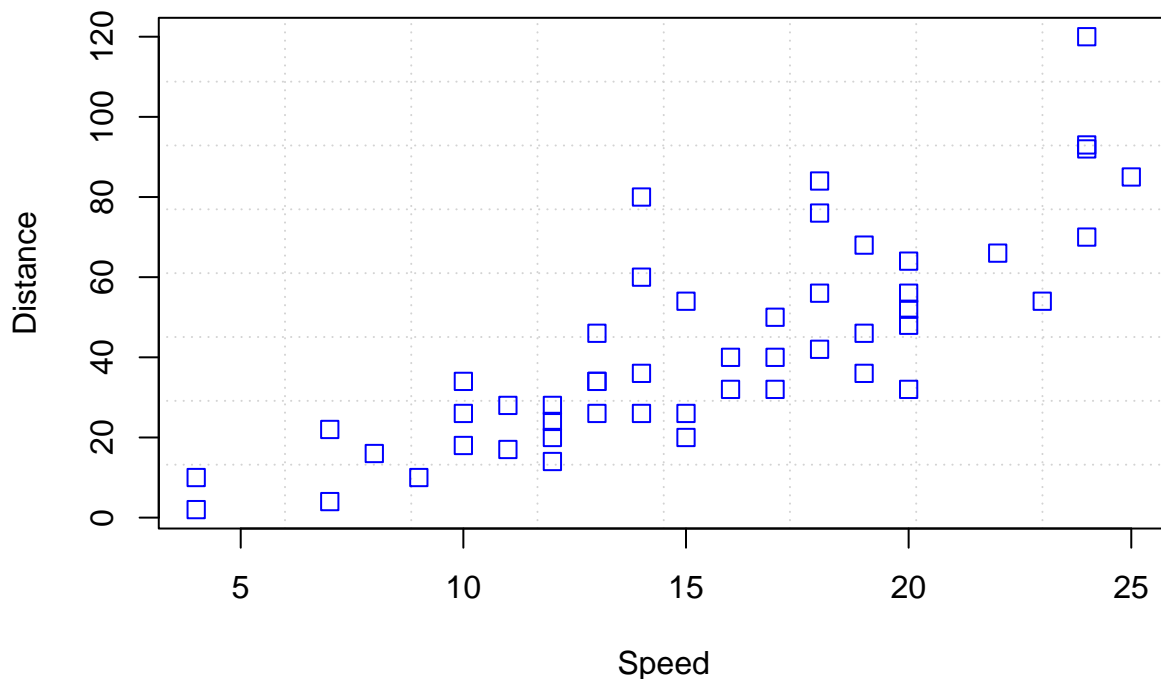
If you don't know where a function is from use “??”:

```
??rarefy
```

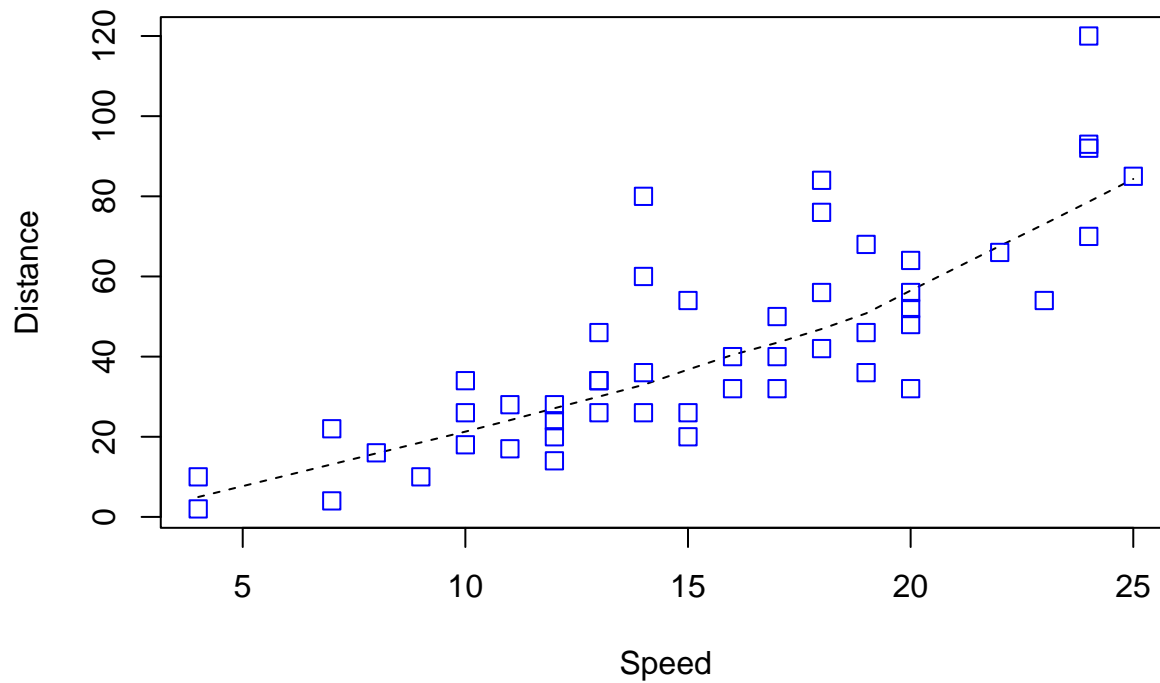
You can ask for an example for a function or package with `example()`. This will print both the code and the result (not all packages provides examples, though)

```
example(plot)
```

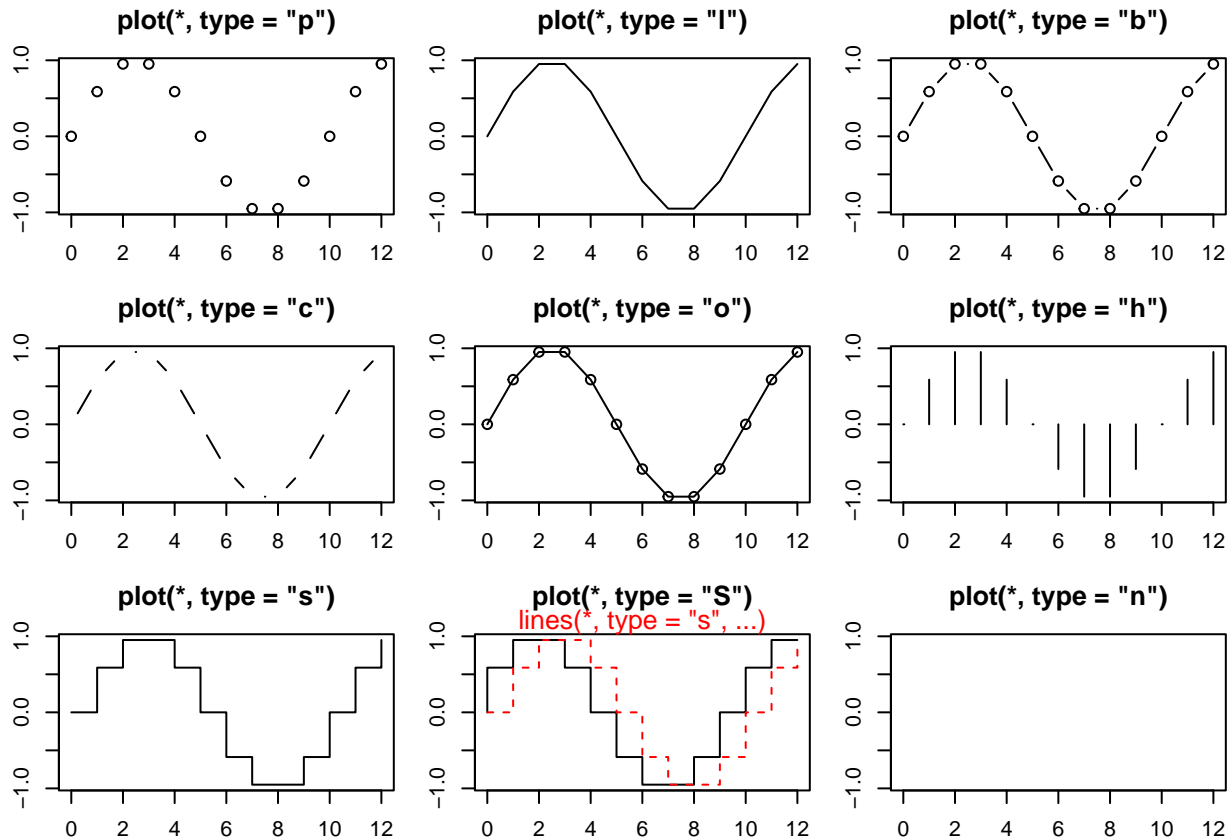
```
##
## plot> Speed <- cars$speed
##
## plot> Distance <- cars$dist
##
## plot> plot(Speed, Distance, panel.first = grid(8, 8),
## plot+      pch = 0, cex = 1.2, col = "blue")
```



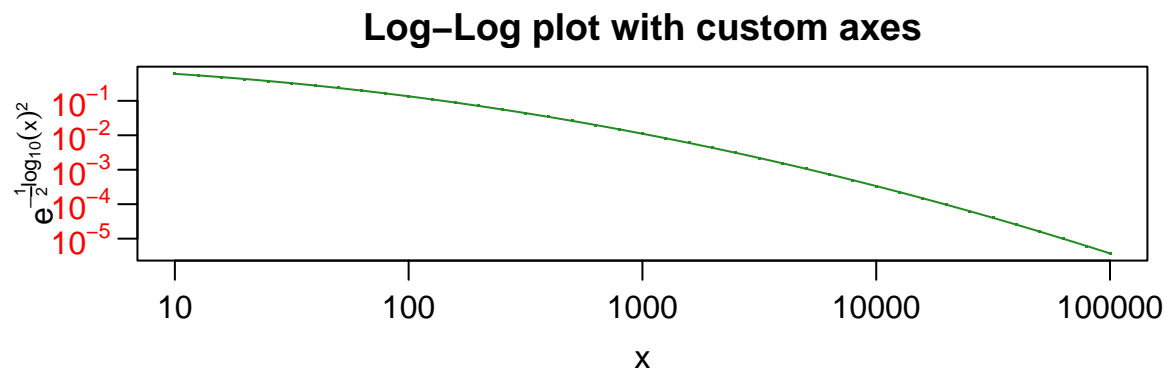
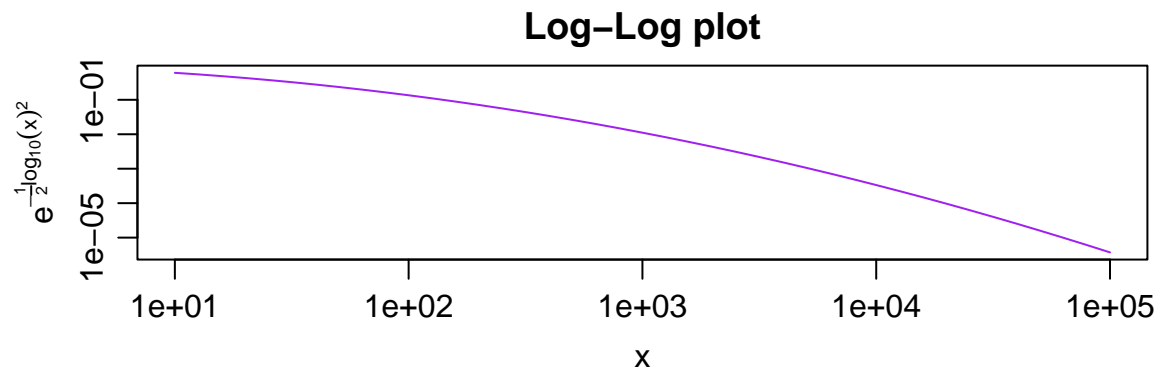
```
##
## plot> plot(Speed, Distance,
## plot+      panel.first = lines(stats::lowess(Speed, Distance), lty = "dashed"),
## plot+      pch = 0, cex = 1.2, col = "blue")
```



```
##
## plot> ## Show the different plot types
## plot> x <- 0:12
##
## plot> y <- sin(pi/5 * x)
##
## plot> op <- par(mfrow = c(3,3), mar = .1+ c(2,2,3,1))
##
## plot> for (tp in c("p","l","b", "c","o","h", "s","S","n")) {
## plot+   plot(y ~ x, type = tp, main = paste0("plot(*, type = \"", tp, "\"))")
## plot+   if(tp == "S") {
## plot+     lines(x, y, type = "s", col = "red", lty = 2)
## plot+     mtext("lines(*, type = \"s\\", ...) ", col = "red", cex = 0.8)
## plot+   }
## plot+ }
```

```
##
## plot> par(op)
##
## plot> ##--- Log-Log Plot with custom axes
## plot> lx <- seq(1, 5, length.out = 41)
##
## plot> yl <- expression(e^{-frac(1,2) * {log[10](x)}^2})
##
## plot> y <- exp(-.5*lx^2)
##
## plot> op <- par(mfrow = c(2,1), mar = par("mar")-c(1,0,2,0), mgp = c(2, .7, 0))
##
## plot> plot(10^lx, y, log = "xy", type = "l", col = "purple",
## plot+      main = "Log-Log plot", ylab = yl, xlab = "x")
##
## plot> plot(10^lx, y, log = "xy", type = "o", pch = ".", col = "forestgreen",
## plot+      main = "Log-Log plot with custom axes", ylab = yl, xlab = "x",
## plot+      axes = FALSE, frame.plot = TRUE)
```



```
##
## plot> my.at <- 10^(1:5)
##
## plot> axis(1, at = my.at, labels = formatC(my.at, format = "fg"))
##
## plot> e.y <- -5:-1 ; at.y <- 10^e.y
##
## plot> axis(2, at = at.y, col.axis = "red", las = 1,
## plot+   labels = as.expression(lapply(e.y, function(E) bquote(10^(E)))))
##
## plot> par(op)
```

End