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- (home & transport 0 at meter)
- 44 solar panels (connection!)
- removing gas

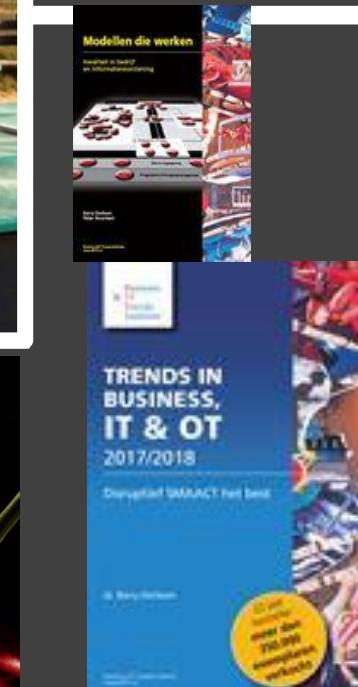


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on Business &
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TRENDS IN
BUSINESS,
IT & OT
2017/2018

Gepubliceerd SAMACT Nederland



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IRIS Case in R

- IRIS case is een echte doceer case voor R gebruik
- De dataset komt uit 1936 (The use of multiple measurements in taxonomic problems)
- 3 plantsoorten: Setosa, Virginica, Versicolor
- 4 meetpunten per plantsoort...in centimeters

IRIS case

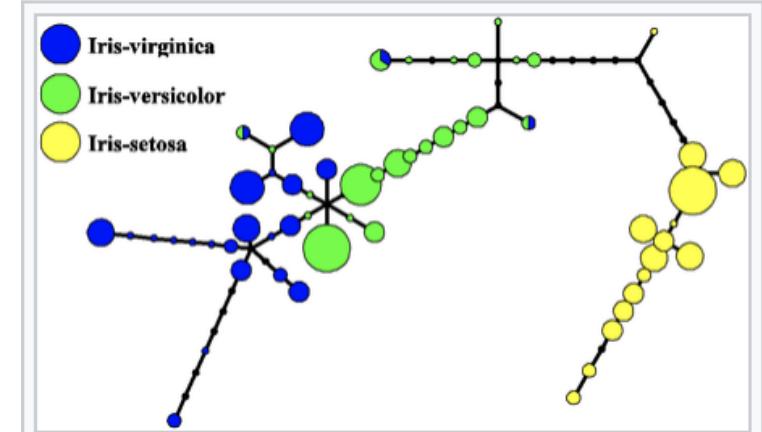
- Species: Versicolor, Setosa, Virginica
- Variabelen: sepal.length , sepal.width , petal.length & petal.width
- Dataset: 50 voorbeelden van species
- Mogelijkheden: Linear discriminant model (species). Classificatie, clustering en algorithms.



Iris Versicolor

Iris Setosa

Iris Virginica



An example of the so-called "metro map" for the *Iris* data set.^[4] Only a small fraction of *Iris-virginica* is mixed with *Iris-versicolor*. All other samples of the different *Iris* species belong to the different nodes.

Intro

- RStudio allows the user to run R in a more user-friendly environment. It is open-source (i.e. free) and available at <http://www.rstudio.com/>
- For R related tutorials and/or resources see the following links:
- <http://dss.princeton.edu/training/> <http://libguides.princeton.edu/dss>



Over R

> 10.000 add-on packages

>100.000
LinkIn R group

The screenshot shows the RStudio interface with three main panes:

- Console** (left): Displays the R startup message and a series of commands and their outputs. The commands include `getwd()`, matrix creation (A and B), and file listing.
- Workspace** (top right): Shows the current workspace with objects A and B defined as 4x2 double matrices.
- Files** (bottom right): Shows the directory structure under H:/MyData/RFiles, including a file named .Rhistory.

The console is where you can type commands and see output

The **workspace** tab shows all the active objects (see next slide). The **history** tab shows a list of commands used so far.

The **files** tab shows all the files and folders in your default workspace as if you were on a PC/Mac window. The **plots** tab will show all your graphs. The **packages** tab will list a series of packages or add-ons needed to run certain processes. For additional info see the **help** tab

Workspace tab (1)

The workspace tab stores any object, value, function or anything you create during your R session. In the example below, if you click on the dotted squares you can see the data on a screen to the left.

The screenshot shows the RStudio interface with three panes. The top-left pane is the code editor with an R script containing the following code:

```
1 betwd()
2 setwd("H:/MyData/RFiles")
3 getwd()
4 5*5
5 A <- matrix(c(1,2,3,4,5,6,7,8), nrow=4, ncol=2)
6 A
7 B <- matrix(c(1,2,3,4,5,6,7,8), nrow=4, ncol=2, byrow=TRUE)
8 B
```

A red arrow points from the text "Showing here matrix B. To see matrix A click on the respective tab." to the matrix B table in the bottom-left pane. The bottom-left pane displays a 4x2 matrix:

	V1 V2
1	1 2
2	3 4
3	5 6
4	7 8

The bottom-right pane is the "Workspace" tab of the Environment pane, showing the objects defined in the workspace:

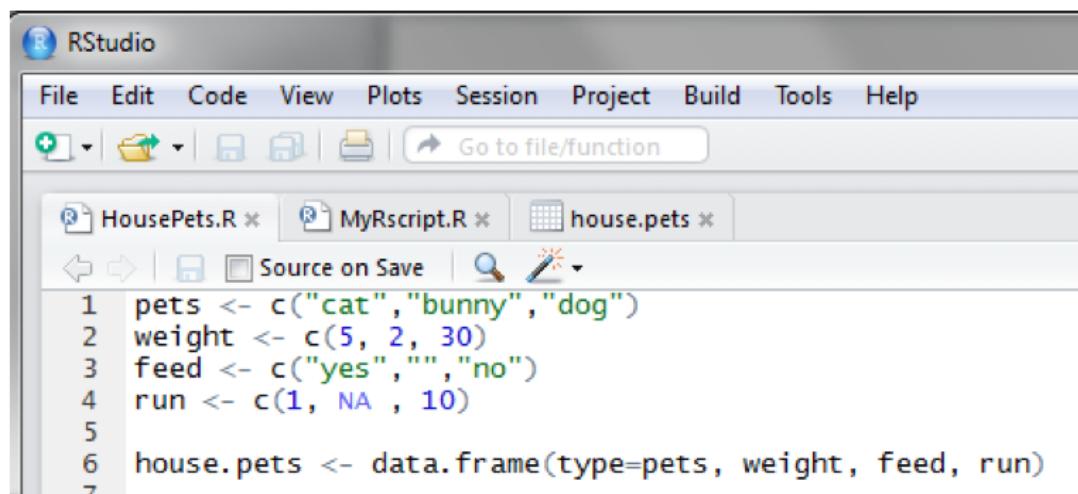
Data	Description
A	4x2 double matrix
B	4x2 double matrix
house.pets	3 obs. of 4 variables
Values	
feed	character[3]
pets	character[3]
run	numeric[3]
weight	numeric[3]

Red arrows point from the text "Showing here matrix B. To see matrix A click on the respective tab." to the matrix B table in the bottom-left pane, and from the text "click on the respective tab." to the A entry in the workspace table.

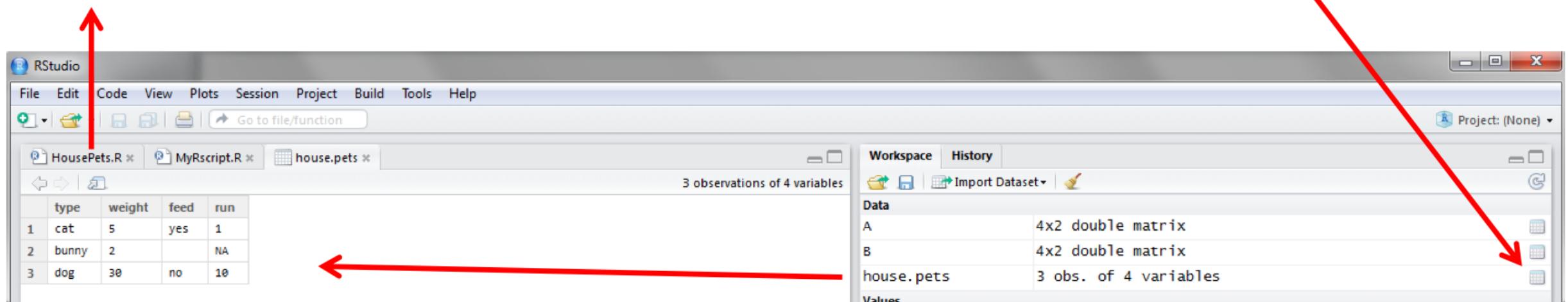
Showing here matrix B. To see matrix A click on the respective tab.

Workspace tab (2)

Here is another example on how the workspace looks like when more objects are added. Notice that the data frame `house.pets` is formed from different individual values or vectors.



```
1 pets <- c("cat", "bunny", "dog")
2 weight <- c(5, 2, 30)
3 feed <- c("yes", "", "no")
4 run <- c(1, NA, 10)
5
6 house.pets <- data.frame(type=pets, weight, feed, run)
7
```



Click on the dotted square to look at the dataset in a spreadsheet form.

The screenshot shows the RStudio interface with the workspace tab selected. A red arrow points to the dotted square icon in the toolbar of the workspace panel, which is used to view datasets in a spreadsheet format. Another red arrow points to the table representation of the `house.pets` data frame, which contains 3 observations of 4 variables:

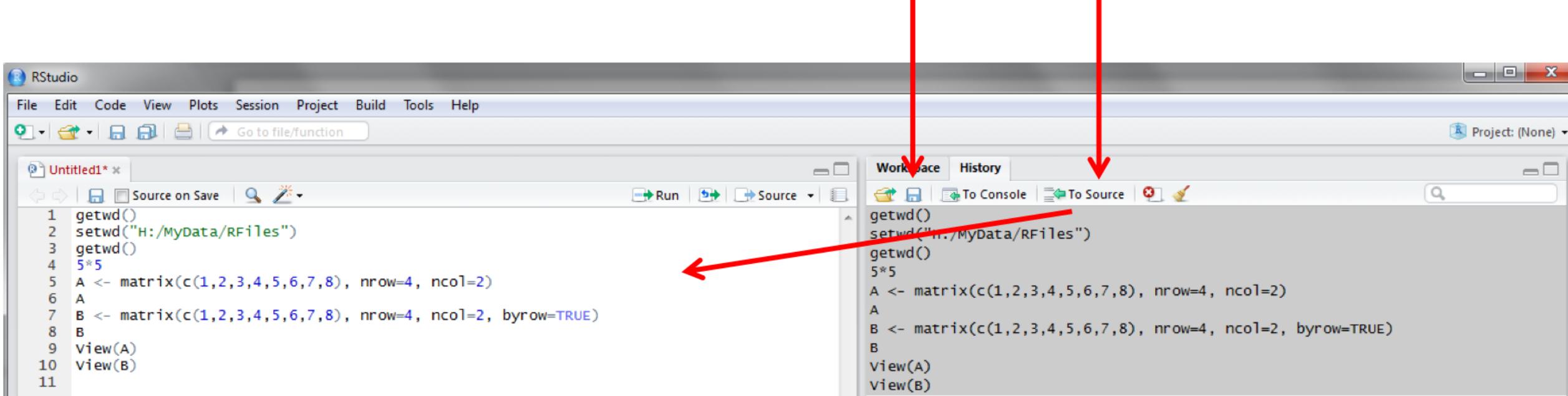
	type	weight	feed	run
1	cat	5	yes	1
2	bunny	2		NA
3	dog	30	no	10

The workspace panel also lists other objects: A (4x2 double matrix), B (4x2 double matrix), and house.pets (3 obs. of 4 variables). There is a red arrow pointing to the dotted square icon in the workspace panel's toolbar.

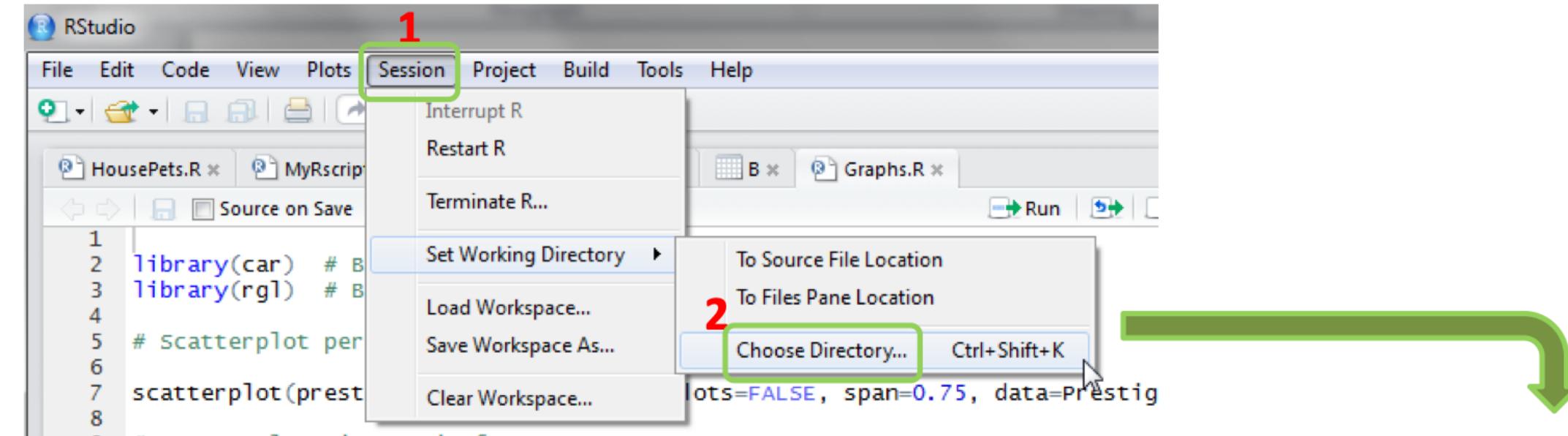
History tab

The history tab keeps a record of all previous commands. It helps when testing and running processes. Here you can either **save** the whole list or you can **select** the commands you want and send them to an R script to keep track of your work.

In this example, we select all and click on the “To Source” icon, a window on the left will open with the list of commands. Make sure to save the ‘untitled1’ file as an *.R script.



Changing the working directory

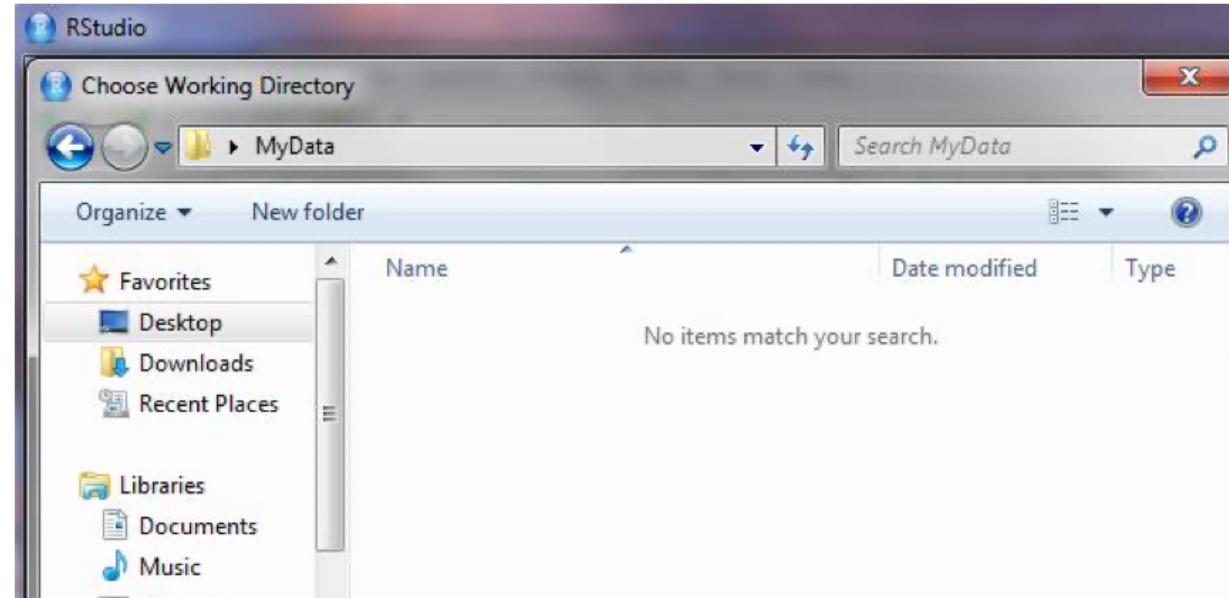


If you have different projects you can change the working directory for that session, see above. Or you can type:

```
# Shows the working directory (wd)
```

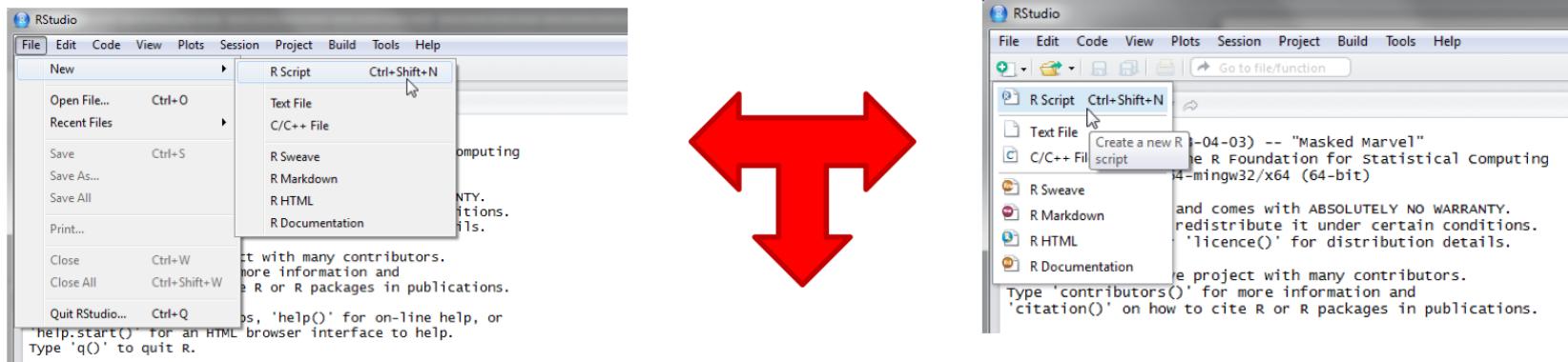
```
getwd()
```

```
# Changes the wd
```



R script (2)

To create a new R script you can either go to File -> New -> R Script, or click on the icon with the “+” sign and select “R Script”, or simply press Ctrl+Shift+N. Make sure to save the script.



The image shows the RStudio interface with the 'Untitled1' editor window open. The window contains a single line of R code: '1 |'. To the right of the editor, a toolbar features a 'Run' button, which is highlighted with a red arrow. Below the editor is the 'Console' tab, showing the path 'H:/MyData/RFiles/'. The 'Workspace' and 'Data' panes are visible on the right side of the interface.

Here you can type R commands and run them. Just leave the cursor anywhere on the line where the command is and press Ctrl-R or click on the ‘Run’ icon above. Output will appear in the console below.

R install

- Download R en voor de installatie uit voor het gewenste OS :
- <https://lib.ugent.be/CRAN/>
- Download RStudio for Desktop en voor de installatie uit:
- <https://www.rstudio.com/products/rstudio/#Desktop>

Scripts uitvoeren (dit zijn functionaliteiten)

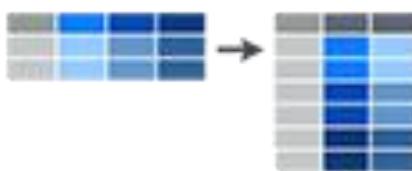
```
install.packages("caret", repos = 'https://lib.ugent.be/CRAN/')
install.packages("tidyr", repos = 'https://lib.ugent.be/CRAN/')
install.packages("ggthemes", repos = 'https://lib.ugent.be/CRAN/')
install.packages("MASS", repos = 'https://lib.ugent.be/CRAN/')
install.packages("e1071", repos = 'https://lib.ugent.be/CRAN/')
install.packages("randomForest", repos = 'https://lib.ugent.be/CRAN/')
install.packages("gbm", repos = 'https://lib.ugent.be/CRAN/')
install.packages("lida", repos = 'https://lib.ugent.be/CRAN/')
```

Voorbeeld: Caret Package

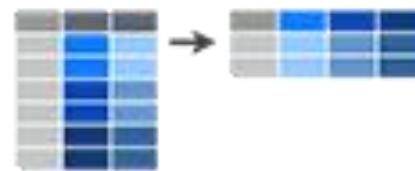
- The [caret](#) package (short for Classification And REgression Training) is a set of functions that attempt to streamline the process for creating predictive models. The package contains tools for:
- data splitting
- pre-processing
- feature selection
- model tuning using resampling
- variable importance estimation
- as well as other functionality

Organize Your Data for Easier Analyses in R

`gather()`



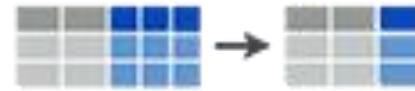
`spread()`



`separate()`



`unite()`



- `gather()`: collapse multiple columns into key-pair values
- `spread()`: reverse of gather. Separate one column into multiple
- `separate()`: separate one column into multiple
- `unite()`: unite multiple columns into one

tidyR package

Voorbeeld: Tidyr package

Observations

Tidy data

Variables

	A	B	C	D	E
1	State	Murder	Assault	UrbanPop	Rape
2	Alabama	13.2	236	58	21.2
3	Alaska	10	263	48	44.5
4	Arizona	8.1	294	80	31
5	Arkansas	8.8	190	50	19.5
6	California	9	276	91	40.6
7	Colorado	7.9	204	78	38.7
8	Connecticut	3.3	110	77	11.1
9	Delaware	5.9	238	72	15.8
10	Florida	15.4	335	80	31.9
11	Georgia	17.4	211	60	25.8
12	Hawaii	5.3	46	83	20.2
13	Idaho	2.6	120	54	14.2
14	Illinois	10.4	249	83	24
15	Indiana	7.2	113	65	21
16	Iowa	2.2	56	57	11.3
17	Kansas	6	115	66	18
18	Massachusetts	12.9	205	82	32.9

Test datasets

library(caret)

library(tidyr)

library(ggthemes)

library(MASS)

library(e1071)

library(randomForest)

library(gbm)

library(Ida)

Laten we naar
de data kijken
(5 rijen)

Verkrijgen eerste 5 rijen van elke subset

subset(iris, Species == "setosa")[1:5,]
CHECK result

subset(iris, Species == "versicolor")[1:5,]
CHECK result

subset(iris, Species == "virginica")[1:5,]

check result

Analyseer de 3 keer 5 rijen

- Waar zie je al onderscheid?
- Schrijf je eerste bevindingen op om de 3 soorten te onderscheiden

Exploratief data analyse



Snel is te zien dat petal.length van sort Setosa korter is dan petal.length van andere soorten. Is dit waar?



Get column "Species" for all lines where Petal.Length < 2



```
subset(iris, Petal.Length < 2)[, "Species"]
```



Je hebt nu een eerste selectie dat een deel van de data verklaart

lets meer van
de data leren



SUMMARY(IRIS)



WAT LEES JE HIER?

summary(iris)

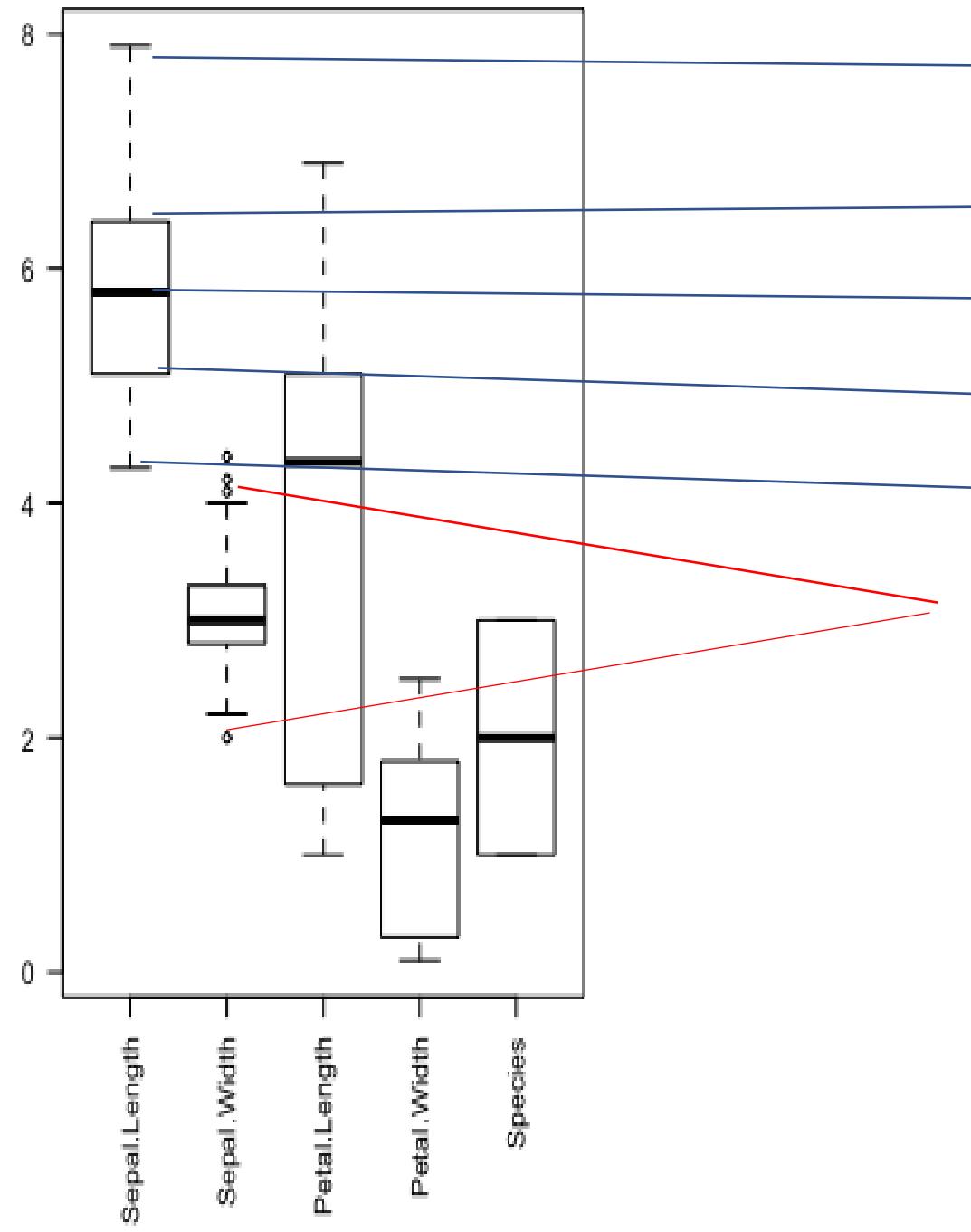
Sepal.Length	Sepal.Width
Min. :4.300	Min. :2.000
1st Qu.:5.100	1st Qu.:2.800
Median :5.800	Median :3.000
Mean :5.843	Mean :3.057
3rd Qu.:6.400	3rd Qu.:3.300
Max. :7.900	Max. :4.400
Petal.Length	Petal.Width
Min. :1.000	Min. :0.100
1st Qu.:1.600	1st Qu.:0.300
Median :4.350	Median :1.300
Mean :3.758	Mean :1.199
3rd Qu.:5.100	3rd Qu.:1.800
Max. :6.900	Max. :2.500
Species	
setosa :50	
versicolor:50	
virginica :50	

Visualeren?
→ boxplot

```
par(mar=c(7,5,1,1))
```

```
# more space to labels
```

```
boxplot(iris,las=2)
```



Maximum value (excl. outliers)

Upper Quartile: 25% of values are higher than this

Median: 50% of values are higher / 50% lower

Lower Quartile: 25% of values are Lower than this

Lower Quartile: 25% of values are lower than this

Outliers: values above or below 1,5 times the
interquartile range

Een scherper
beeld van elke
soort
verkrijgen

```
irisVer <- subset(iris, Species == "versicolor")
```

```
irisSet <- subset(iris, Species == "setosa")
```

```
irisVir <- subset(iris, Species == "virginica")
```

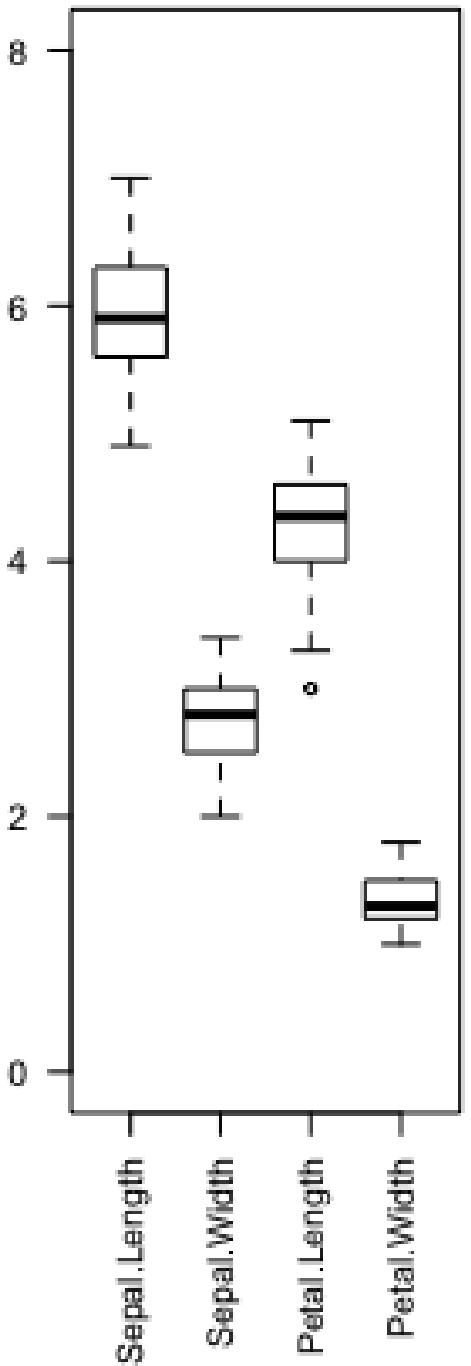
```
par(mfrow=c(1,3),mar=c(6,3,2,1))
```

```
boxplot(irisVer[,1:4], main="Versicolor",ylim = c(0,8),las=2)
```

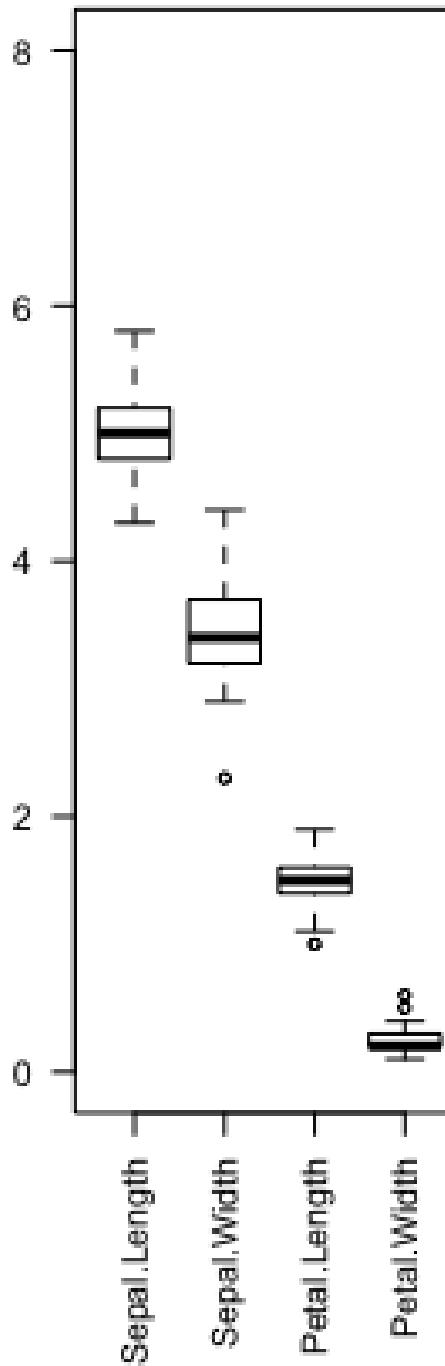
```
boxplot(irisSet[,1:4], main="Setosa",ylim = c(0,8),las=2)
```

```
boxplot(irisVir[,1:4], main="Virginica",ylim = c(0,8),las=2)
```

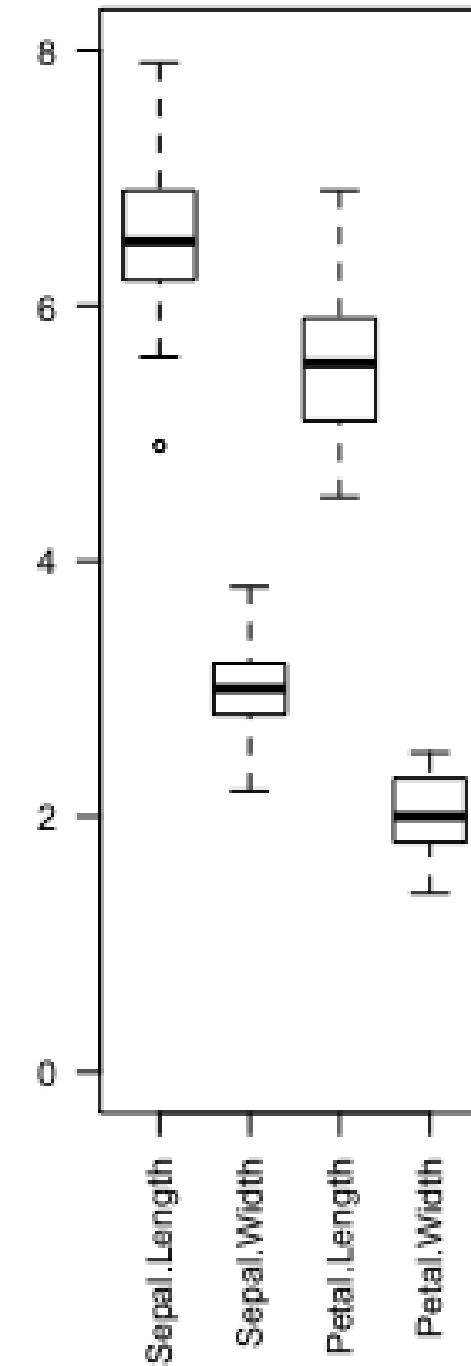
Versicolor



Setosa



Virginica



Histogram (te calculeren per attribute)

```
hist(iris$Petal.Length)
```

#histogram voor 1 attribute maar per soort

```
par(mfrow=c(1,3))
```

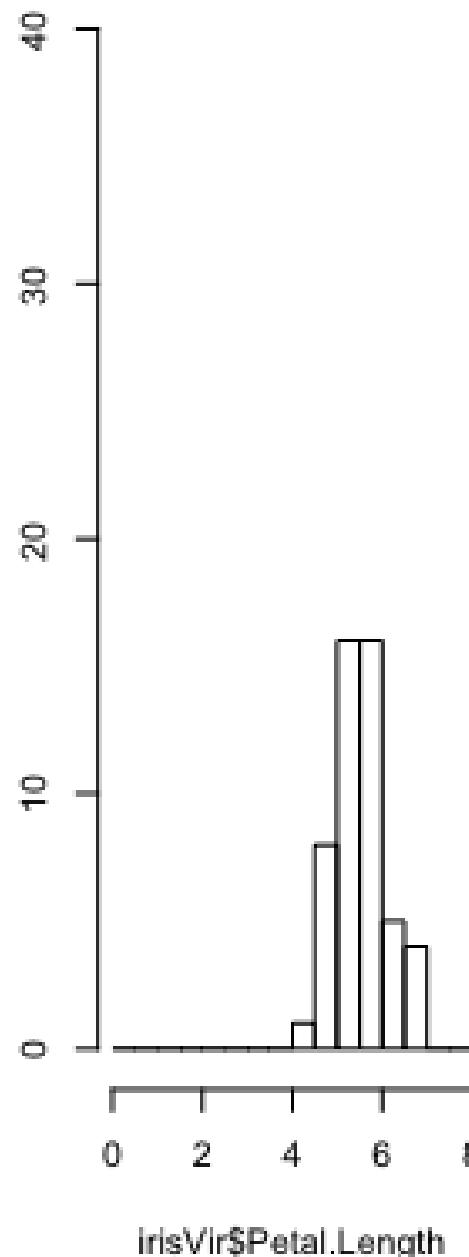
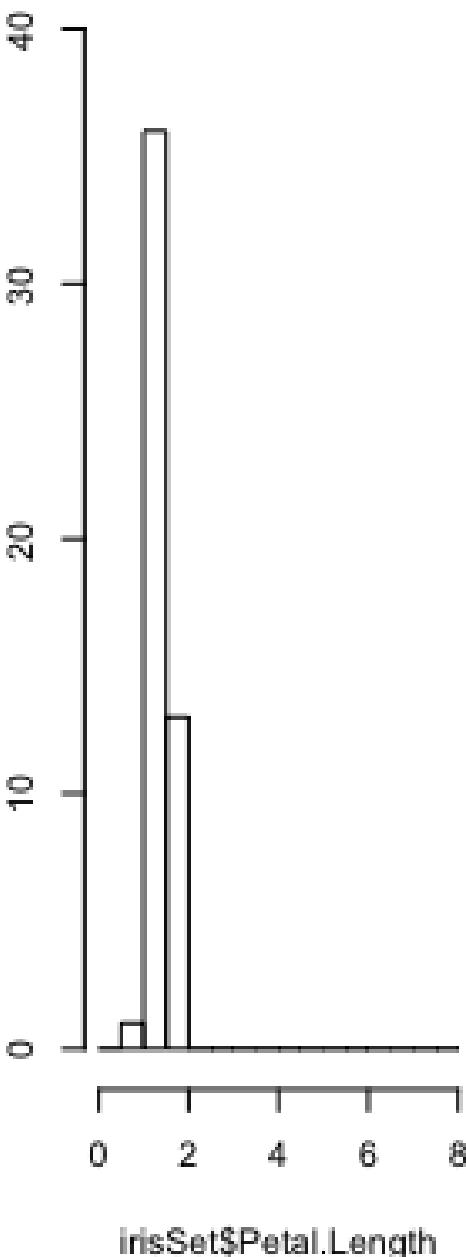
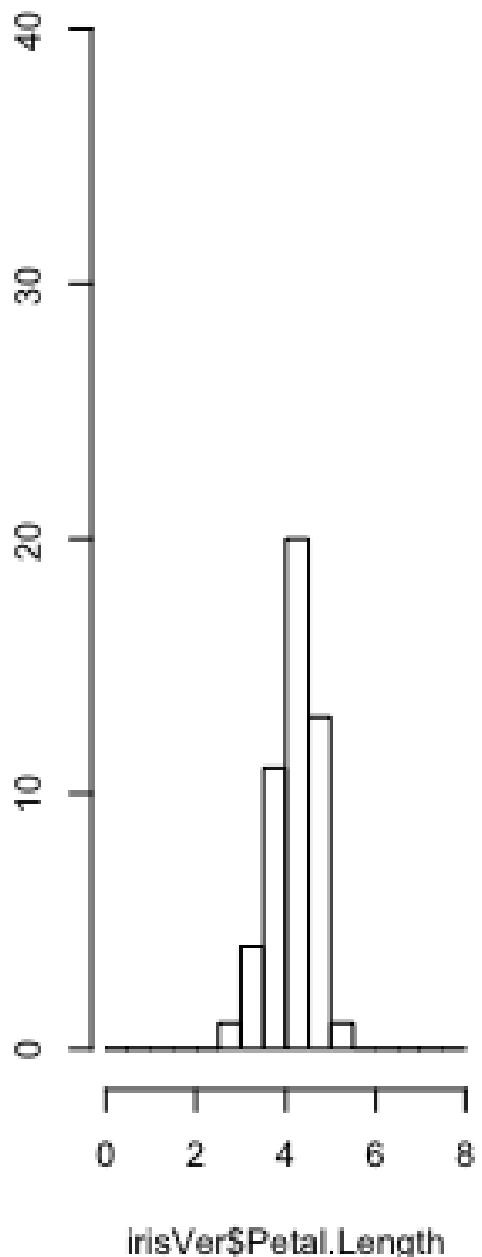
```
hist(irisVer$Petal.Length, breaks=seq(0,8,l=17), xlim=c(0,8), ylim=c(0,40))
```

```
hist(irisSet$Petal.Length, breaks=seq(0,8,l=17), xlim=c(0,8), ylim=c(0,40))
```

```
hist(irisVir$Petal.Length, breaks=seq(0,8,l=17), xlim=c(0,8), ylim=c(0,40))
```

Je ziet de distributie van de waarde van petal.length verschillend zijn per soort

histogram of irisVer\$Petal.Length histogram of irisSet\$Petal.Length histogram of irisVir\$Petal.Length



Violin plots tonen statistiek en data distributie

- **library(vioplot)**

Als het goed zie je dit:

- ## Loading required package: sm
- ## Package 'sm', version 2.2-5.4: type help(sm) for summary information

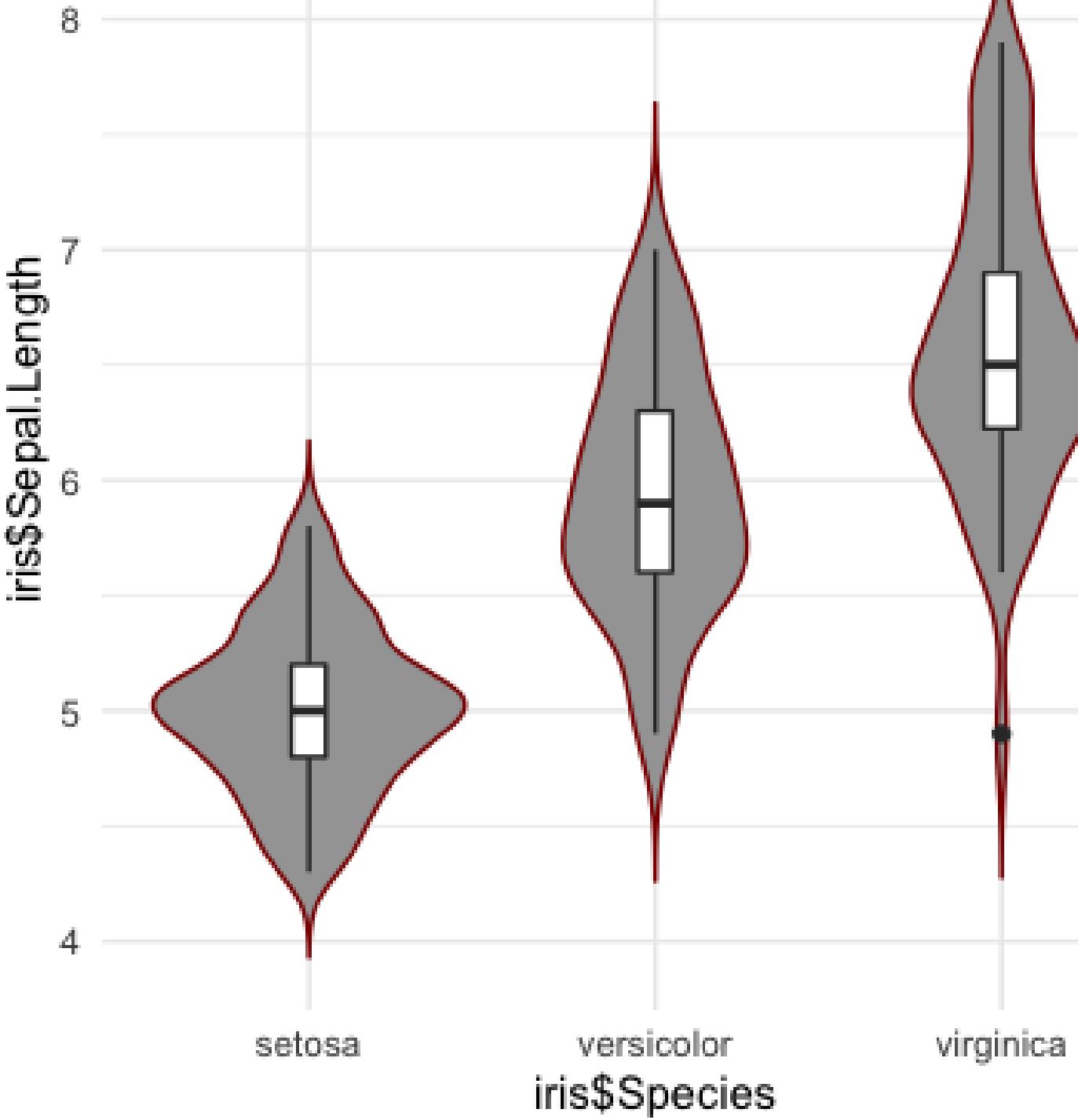
Code:

```
vioplot(iris$Sepal.Length,iris$Sepal.Width,iris$Petal.Length,  
iris$Petal.Width,  
names=c("Sep.Len","Sep.Wid","Pet.Len","Pet.Wid"),  
col="green")
```

MAAR.....HET KAN ZOMAAR MIS GAAN EN NU?

ZELF PROBEREN UIT TE VOGELEN, HINT:

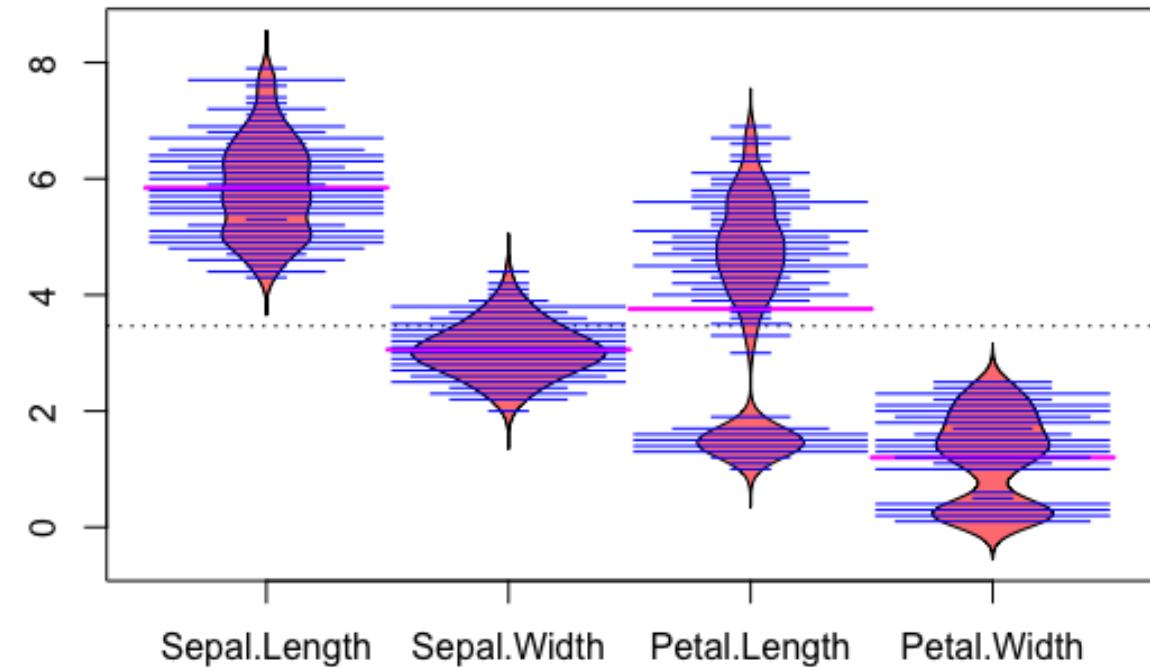
- <http://www.sthda.com/english/wiki/ggplot2-violin-plot-quick-start-guide-r-software-and-data-visualization>
- Hoe ver kom jij in 30 minuten?



Of:

```
> library(beanplot)
> xiris <- iris
> xiris$Species <- NULL
> beanplot(xiris, main = "Iris
flowers", col=c('#ff8080','#000
0FF','#0000FF','#FF00FF'),
border = "#000000")
```

Iris flowers



Correlaties tussen variabelen



```
corr <- cor(iris[,1:4])
```



```
round(corr,3)
```



Hoe lezen?

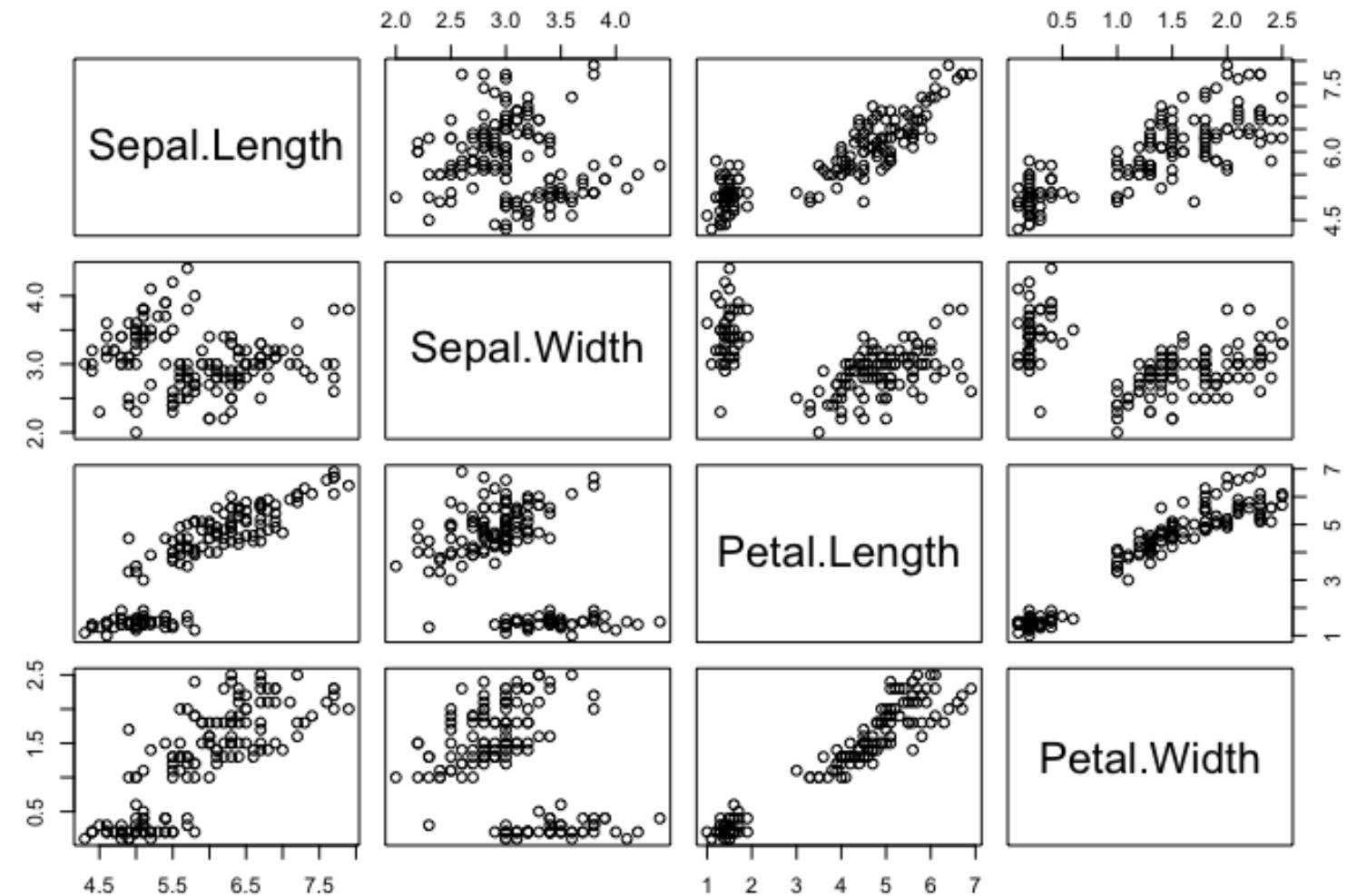
Correlatie tussen variabelen

- Sepal.Length Sepal.Width
- Sepal.Length 1.000 -0.118
- Sepal.Width -0.118 1.000
- Petal.Length 0.872 -0.428
- Petal.Width 0.818 -0.366
- Petal.Length Petal.Width
- Sepal.Length 0.872 0.818
- Sepal.Width -0.428 -0.366
- Petal.Length 1.000 0.963
- Petal.Width 0.963 1.000

Variabelen correleren volledig

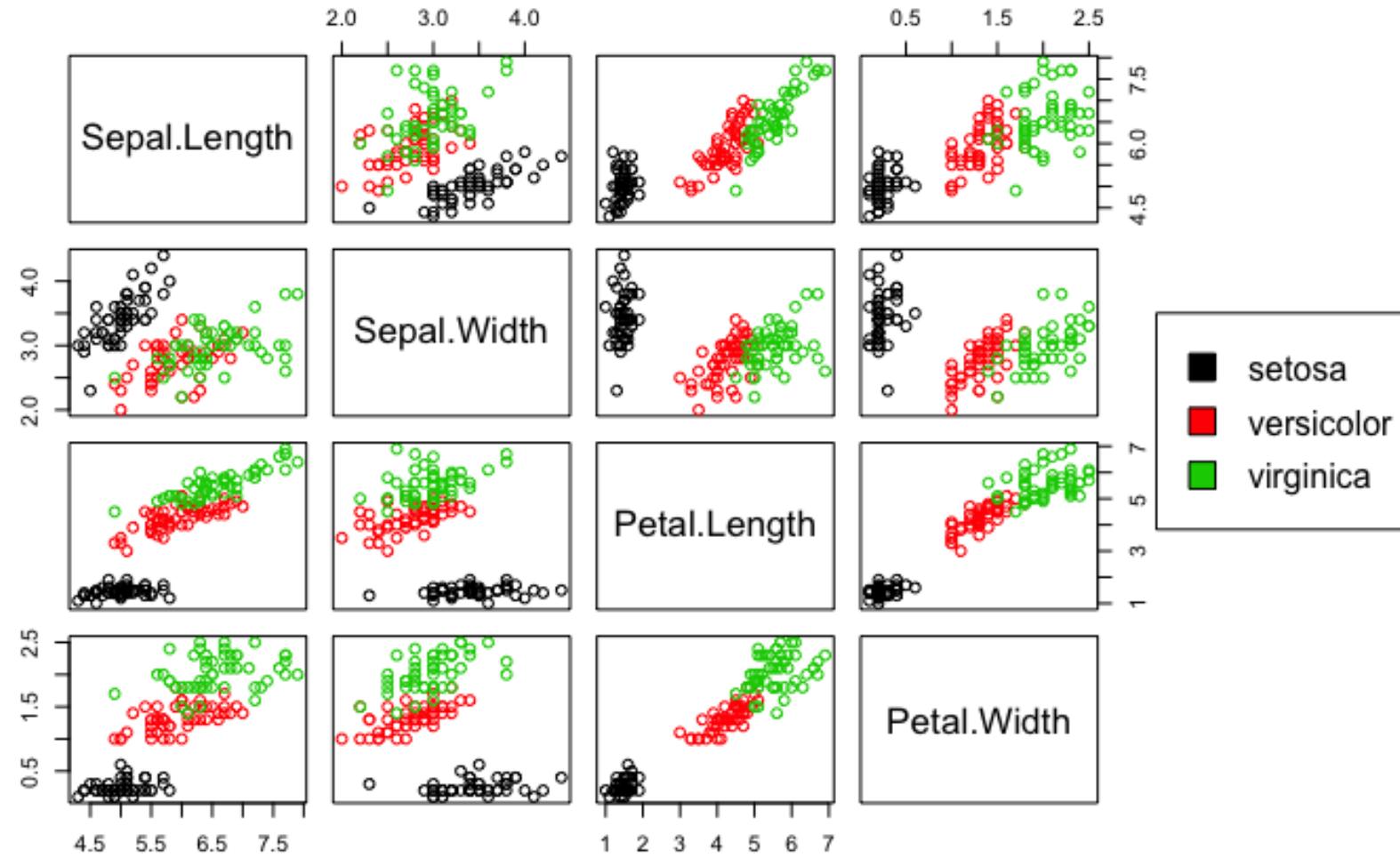
Scatterplot matrix

- `pairs(iris[,1:4])`
- Visuele bevestiging van vorige opdracht!

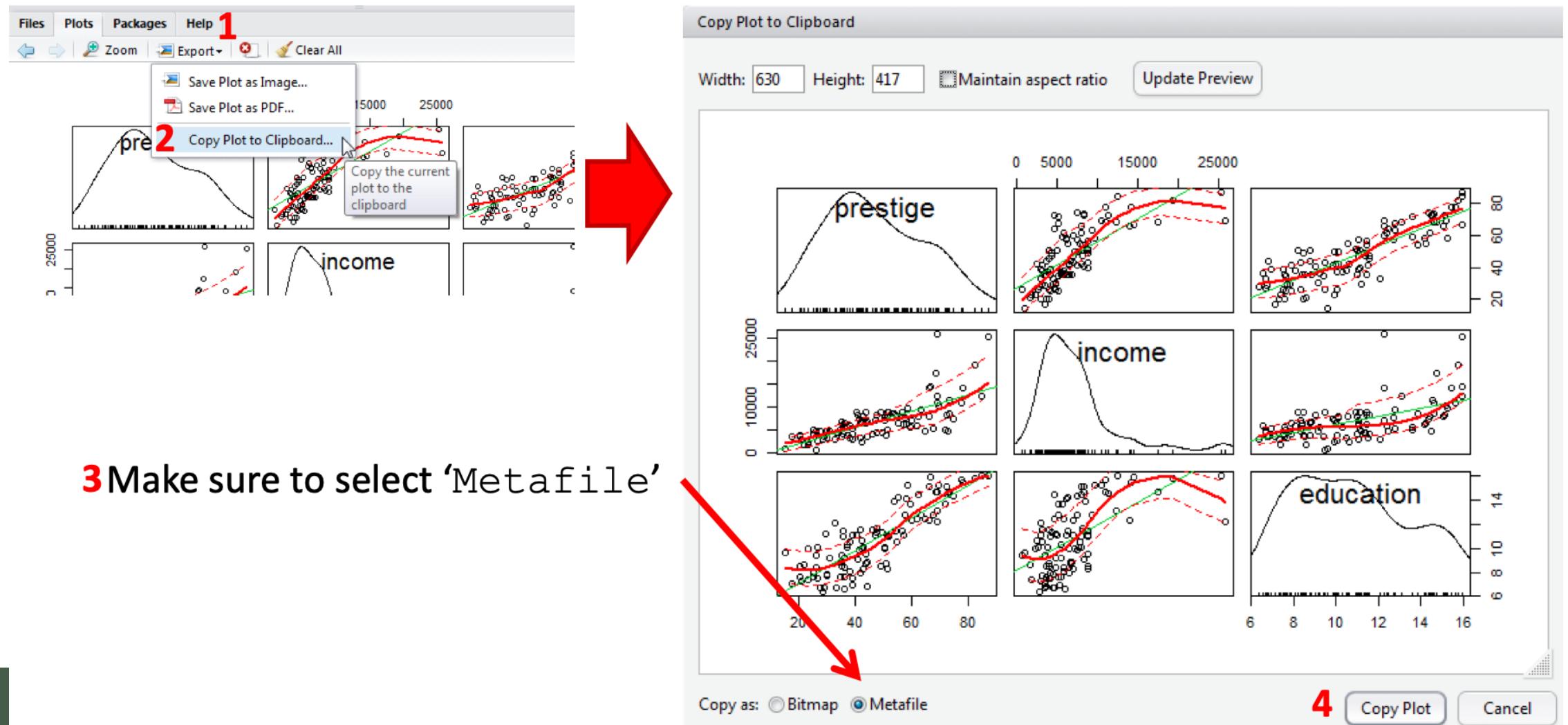


Dit kunnen we
ook doen per
soort

```
pairs(iris[,1:4],col=iris[,5],oma=c(4,4,6,12))  
par(xpd=TRUE)  
legend(0.85,0.6,  
as.vector(unique(iris$Species))),fill=c(1,2,3))
```



To extract the graph, click on “Export” where you can save the file as an image (PNG, JPG, etc.) or as PDF, these options are useful when you only want to share the graph or use it in a LaTeX document. Probably, the easiest way to export a graph is by copying it to the clipboard and then paste it directly into your Word document.



Een andere
manier is
parallel
coordinate
plot



```
library(MASS)
```



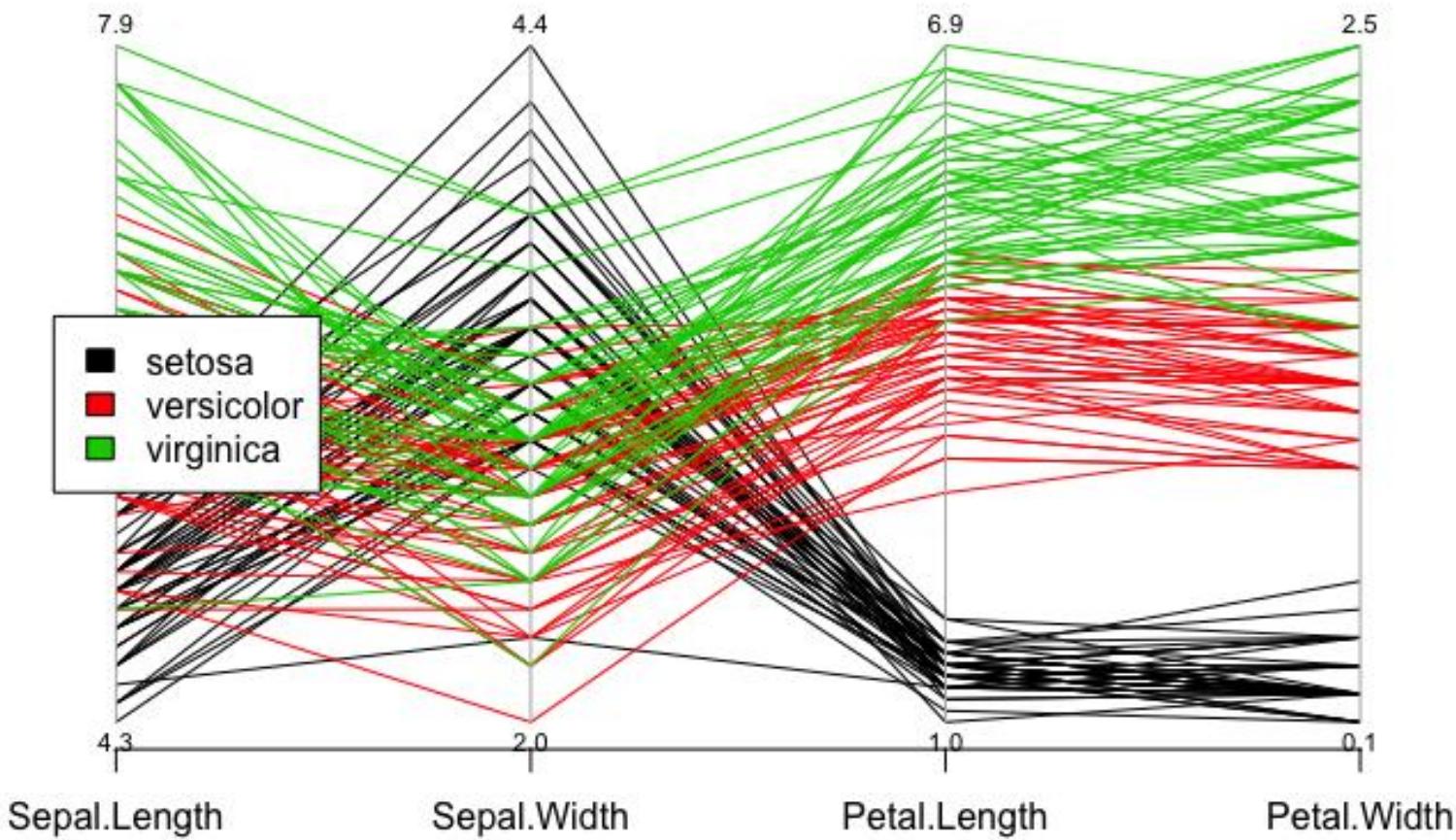
```
parcoord(iris[,1:4],  
col=iris[,5],var.label=TRUE,oma=c(4,4,6,12))
```



```
par(xpd=TRUE)
```



```
legend(0.85,0.6,  
as.vector(unique(iris$Species)),fill=c(1,2,3))
```

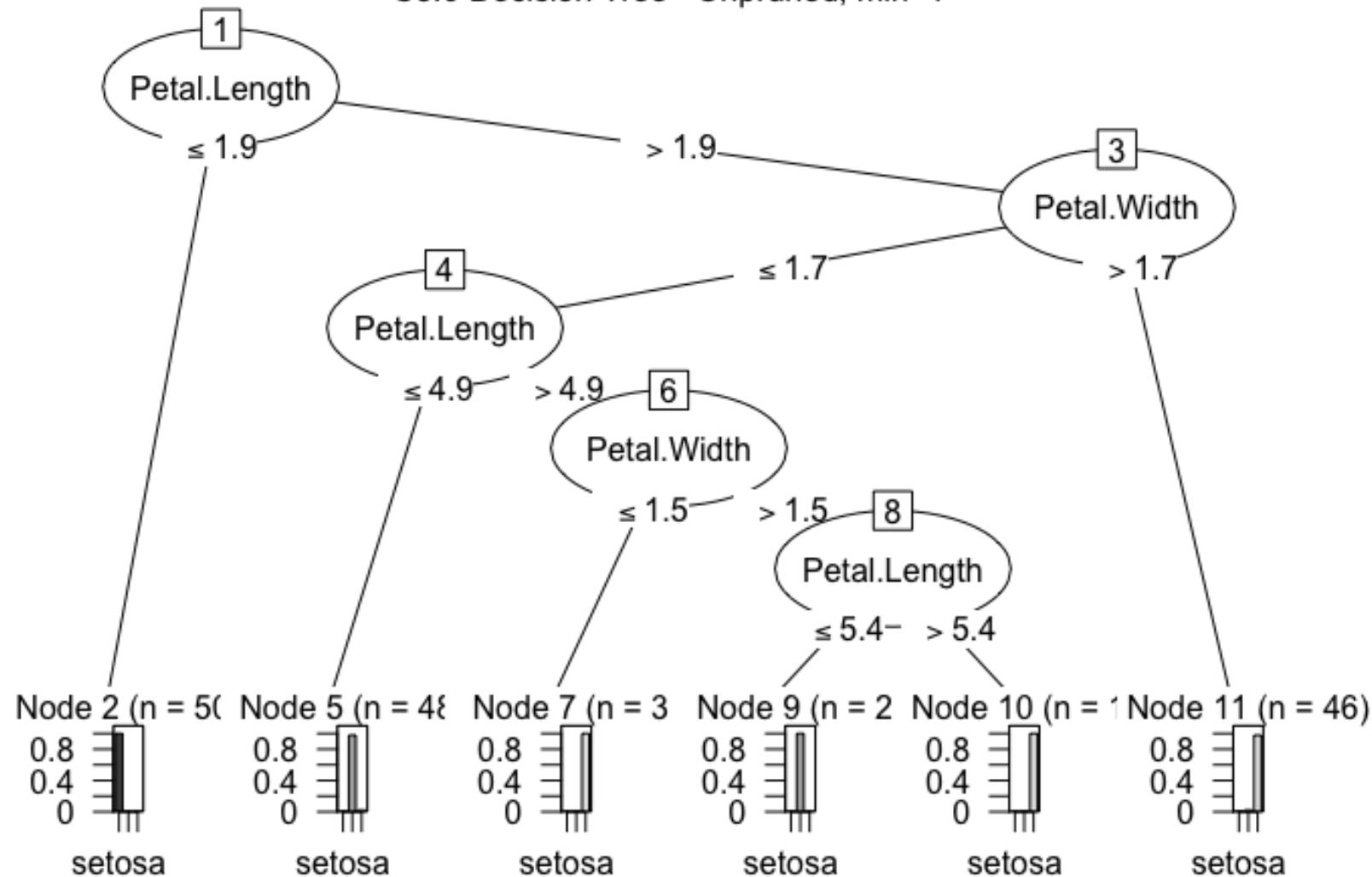


Laten we een besluitboom maken (classificatie)

- We weten dat er classes voor 150 instances van 'Irises'. Interessant is of er een predictive model is voor de soorten gebaseerd op petal en sepal width en length. Hiervoor maken ween besluitboom

```
library(C50)
input <- iris[,1:4]
output <- iris[,5]
model1 <- C5.0(input, output, control =
C5.0Control(noGlobalPruning = TRUE,minCases=1))
plot(model1, main="C5.0 Decision Tree - Unpruned,
min=1")
```

C5.0 Decision Tree - Unpruned, min=1



Eenvoudiger model maken



```
model2 <- C5.0(input, output, control =  
C5.0Control(noGlobalPruning = FALSE))
```



```
plot(model2, main="C5.0 Decision Tree -  
Pruned")
```

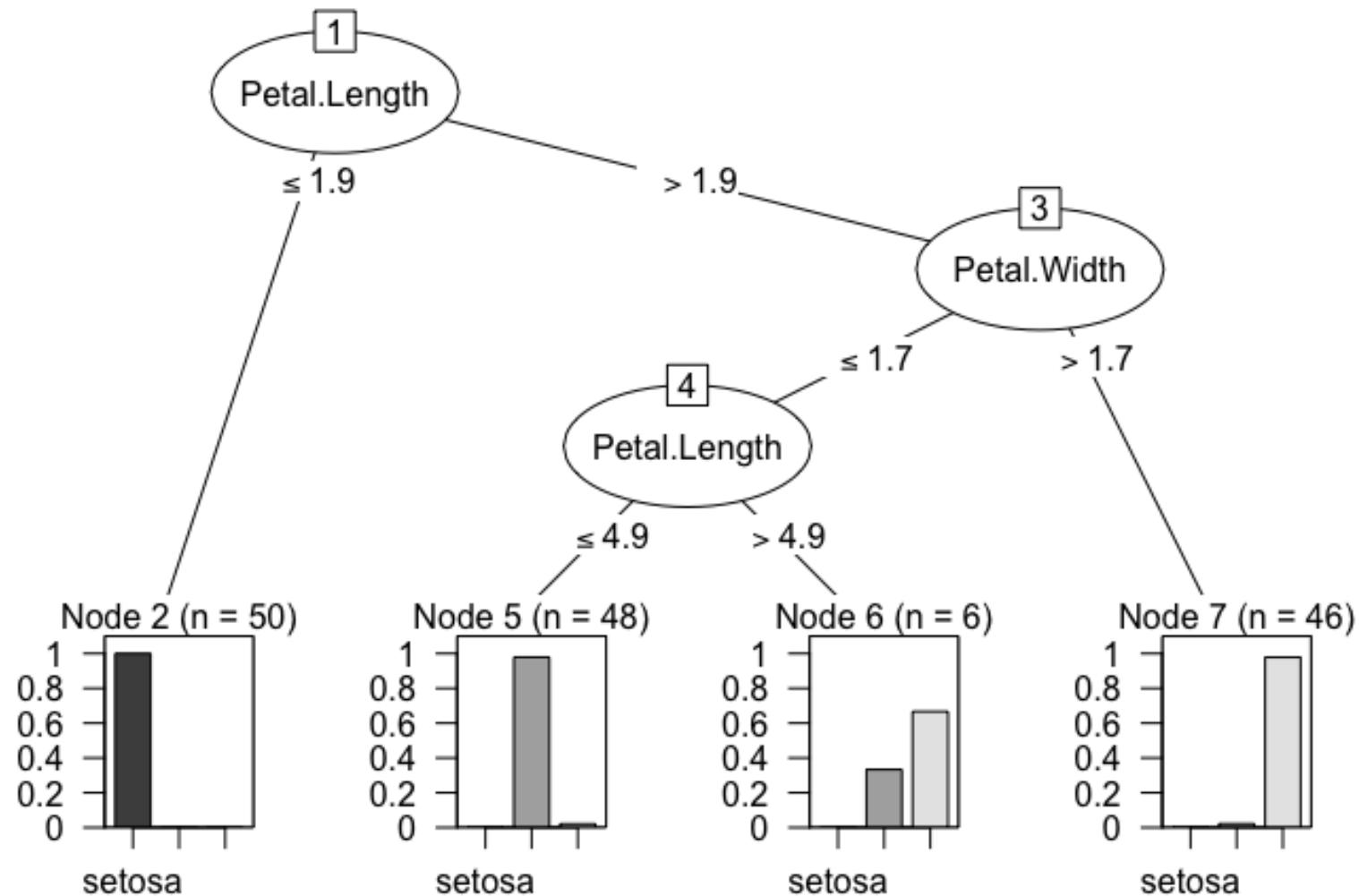


NA UITVOERING bovenstaande:



```
summary(model2)
```

C5.0 Decision Tree - Pruned



inzoomen



C5imp(model2,metric='usage')



Prediction gebaseerd op numerieke variabelen:



`newcases <- iris[c(1:3,51:53,101:103),]`



`newcases`

Voorspelling maken (bijv. Voor zonder species)



```
predicted <- predict(model2, newcases,  
type="class")
```



Predicted



Verrijken van je model:



```
predicted <- predict(model2, iris, type="class")
```



predicted

Vergelijk (als
in database
en voorspeld)

```
iris$predictedC501  
<- predicted
```

```
iris[iris$Species !=  
iris$predictedC501,]
```