## introduction

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## How to keep your code clean

#### Coding convention

- Pick a naming convention and stick to it
- camelCase = "this is a nice style"
- snake\_case = "this is ok too"
- Comment your code
- Look at the google style book to make sure your code is easilly readable by anyone
- https://google.github.io/styleguide/Rguide.xml
- they advice to use only "<-" and not "=" but i personally think it is pointless

#### Storage

Keep a README.md file at the root of your folder explaining where everything is, helping someone that knows nothing about your data to navigate your work. Keeping your work in the cloud, through services like dropbox, icloud, or google drive. The best would be github but it is not easy in the beginning.

#### **Folders**

Keep your folder clean, with clear names in minuscules separated by "\_":

- data
  - raw
  - preprocessed
  - analysis
    - \* analysis one ...
- scripts
  - preprocessing: scripts that transforms the raw data in processed data
  - analysis: scripts that use preprocessed data and performs analysis on it
  - markdown: your markdown files
  - r files: other R files, like utility functions
- media: here should go any ressources, presentations, images you produced or needed etc...
  - presentations
  - graphics
  - text
  - notes
- backups: you might need a backup folder when in doubt
  - data
  - script
  - media

#### Variables

R infer on its own the type of the variable that you want to create based on the input you give. All variables are at minimum a vector.

#### Atomic vector data type

```
# Character
a <- "This is a character vector"

# Numeric (integer)
a = 12
a <- 12

# Numeric (Float)
a <- 12.2

# Logical
a <- TRUE
a <- FALSE

print(a)</pre>
```

#### Combine atomic elements

## [1] FALSE

#### Vectors needs to be of same type

```
a <- c(1,2,3)
print(a)

## [1] 1 2 3

a <- c(1,"a")
print(a)

## [1] "1" "a"</pre>
```

#### List can mix types

```
a <- list(1,2,3)
print(a)

## [[1]]
## [1] 1
##
## [[2]]
## [1] 2
##
## [[3]]
## [1] 3
```

```
a <- list(1,"a")
print(a)

## [[1]]
## [1] 1
##
## [[2]]
## [1] "a"</pre>
```

#### Matrix 2\*2 Needs to be of same type

```
a <- matrix( c('a','a','b','c','b',2), nrow = 2, ncol = 3, byrow = T)
print(a)

## [,1] [,2] [,3]
## [1,] "a" "a" "b"
## [2,] "c" "b" "2"

a <- matrix( c('a','a','b','c','b',2), nrow = 2, ncol = 3, byrow = F)
print(a)

## [,1] [,2] [,3]
## [1,] "a" "b" "b"
## [2,] "a" "c" "2"</pre>
```

#### Array N\*N Needs to be of same type

```
a <- array(c('green', 'yellow'), dim = c(3,3,2))</pre>
print(a)
## , , 1
##
##
                 [,2]
                           [,3]
        [,1]
## [1,] "green" "yellow" "green"
## [2,] "yellow" "green" "yellow"
## [3,] "green" "yellow" "green"
##
## , , 2
##
        [,1]
                 [,2]
                           [,3]
## [1,] "yellow" "green" "yellow"
## [2,] "green" "yellow" "green"
## [3,] "yellow" "green" "yellow"
```

#### Factor

For categorical variables

```
# Create a vector.
apple_colors <- c('green', 'green', 'yellow', 'red', 'red', 'green')
# Create a factor object.</pre>
```

```
factor_apple <- factor(apple_colors)</pre>
# Print the factor.
print(factor_apple)
## [1] green green yellow red
                                   red
                                          red
                                                 green
## Levels: green red yellow
print(nlevels(factor_apple))
## [1] 3
# Change names of the factors
levels(factor_apple) <- c("Kindof Green", "Kindof Red", "Kindof Yellow")</pre>
print(factor_apple)
## [1] Kindof Green Kindof Green Kindof Yellow Kindof Red Kindof Red
## [6] Kindof Red
                     Kindof Green
## Levels: Kindof Green Kindof Red Kindof Yellow
Dataframe ++
first_names = c("Melissa", "Sibylle", "Zoe", "Maria")
ages = c(23, 22, 24, 25)
df <- data.frame(first_name = first_names,</pre>
                 age = ages,
                 subject = as.character(c("Activity", "Motivation", "Fluid Intelligence", NA)))
print(df)
## first name age
                               subject
## 1
       Melissa 23
                             Activity
## 2
       Sibylle 22
                           Motivation
## 3
           Zoe 24 Fluid Intelligence
## 4
         Maria 25
print(df$age)
## [1] 23 22 24 25
Missing values are "NOT ASSIGNED" or "NA"
print(df$subject)
## [1] Activity
                         Motivation
                                             Fluid Intelligence
## [4] <NA>
## Levels: Activity Fluid Intelligence Motivation
print(is.na(df$subject))
## [1] FALSE FALSE FALSE TRUE
```

#### I dont want a factor, I want characters!

Sometimes you have ot use function such as apply or sapply, that performs simple loops on your data.

```
print(sapply(df[, 3], as.character))
                            "Motivation"
## [1] "Activity"
                                                  "Fluid Intelligence"
## [4] NA
df[, 3] <- sapply(df[, 3], as.character)</pre>
print(df)
##
     first_name age
                               subject
## 1
                              Activity
        Melissa 23
## 2
        Sibylle 22
                            Motivation
            Zoe 24 Fluid Intelligence
## 3
## 4
          Maria 25
```

## **Operators**

#### Relational operators

```
print(12>23)

## [1] FALSE
a = 12
print(a == 12)

## [1] TRUE
print(a != 32)

## [1] TRUE
print(a >= 11)

## [1] TRUE
a = 32
```

#### Tests

```
if (a == 432) {
  print("a est egale a 12 !")
} else {
  print("pas egale a 12")
}
```

```
## [1] "pas egale a 12"
```

#### Logical operators

```
print((12>23)&&(12<23))

## [1] FALSE
a <- 12
print((a>20)||(a==12))

## [1] TRUE
print(!(a == 32))

## [1] TRUE
!is.na(a)

## [1] TRUE
```

#### Element wise logic

When a vector is tested against a vector of same length

```
a <- c(F, T, T)
b <- c(T, F, T)
print(a&b)

## [1] FALSE FALSE TRUE
print(a|b)</pre>
```

## [1] TRUE TRUE TRUE

#### Others

```
a <- 1:8
print(a)

## [1] 1 2 3 4 5 6 7 8

a <- rep("ce qui est repete", 4)
print(a)

## [1] "ce qui est repete" "ce qui est repete"
## [4] "ce qui est repete"</pre>
```

#### Flow control statements

Flow controls statements are all the statement of a language that will redirect the flow of execution of a program.

#### Conditional control

Sometime you want to execute something only if a condition is true. The most used is the "if/else if/else" statement.

```
a = c(F,F,F,T)

if (a[1]) {
    print("first")
} else if (a[2]) {
    print("second")
} else if (a[3]) {
    print("third")
} else if (a[4]) {
    print("quatrieme")
} else {
    print("invalid")
}
```

## [1] "quatrieme"

```
b <- c(F,F,F,T)

# TODO: write a statement that checks if b has any of its value equal to TRUE.

# If it does return all the indices of the TRUE values

# If not, say that you did not find a True value in any of the %SIZE% elements of b

# (hint:: ?any and ?which)</pre>
```

When you only want to check the value of *ONE* variable. Another way is to use the *switch* statement. It test the value of a variable against several possibilities, like so:

## [1] "Run you fools !"

#### Loops

You often need to repeat some statement. That's what for and while are here for !

```
for (i in 1:NROW(df)) {
  person = df[i,]

  print(paste0(person$first_name, person$age))
}
```

```
## [1] "Melissa23"
```

```
## [1] "Sibylle22"
## [1] "Zoe24"
## [1] "Maria25"
i = 0
while(i<10) {
  print(i)
  i = i + 1
}
## [1] 0
## [1] 1
## [1] 2
## [1] 3
## [1] 4
## [1] 5
## [1] 6
## [1] 7
## [1] 8
## [1] 9
```

#### **Functions**

Functions are the central concept to programming. You can think of it as a box containing a series of instructions. Usually they take input, perform change on it, and returns a value. Note that they do not always take input or return a value and act only a series on instructions that do not takes input and do not return anything, but for example will read or write some information on the disk, or setting some parameter.

#### A simple sum

```
sommeFunction = function (x, y, z = 1) {
    # Do some action on the parameters
    sum = x+y+z

# Return to sender the result of your computation
    return(sum)
}
somme = sommeFunction(1,2,3)
```

But of course R has a better function already built in!

```
sum(1,2,3,NA, na.rm = TRUE)
## [1] 6
```

#### A function that generates participant ids

What if you want to set some default parameters? Here is a more complex function that plays with strings to create random IDs for your subjects.

```
getRandomId = function(numberOfIds = 1, lenght=12, allowedCharacters = c(0:9, letters, LETTERS))
{
    # initialize vector
```

```
randomStrings = c(1:numberOfIds)
    # start the generation loop
   for (i in 1:numberOfIds)
        randomStrings[i] <- paste(sample(allowedCharacters, lenght, replace=TRUE),</pre>
                                  collapse="")
   }
    # return the strings
   return(randomStrings)
}
# TODO Now generate 650 ids !
ids = getRandomId(650)
# letters and LETTERS are variables declared by default in R containing the minuscule and capital lette
print(c(0:9, letters, LETTERS))
## [1] "0" "1" "2" "3" "4" "5" "6" "7" "8" "9" "a" "b" "c" "d" "e" "f" "g"
## [18] "h" "i" "j" "k" "l" "m" "n" "o" "p" "q" "r" "s" "t" "u" "v" "w" "x"
## [35] "y" "z" "A" "B" "C" "D" "E" "F" "G" "H" "I" "J" "K" "L" "M" "N" "O"
## [52] "P" "Q" "R" "S" "T" "U" "V" "W" "X" "Y" "Z"
# Add the ids to the data
newDataFrame = data.frame(id=ids, sertARien=1:650)
```

#### Create your own function

Choose one between those three possible function, and create them: \* A function that returns the product of two numbers such that aTimesb = product(a,b) \* A function that adds a prefix to a string, such that prefixedString = prefix(prefixString, string) \* A function that takes out the mean of each column of a data frame, and divides by the standard deviation (process called normalization)

#### Libraries

#### Install libraries

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

#### Load libraries

```
library(ggplot2)
library(psych)

##
## Attaching package: 'psych'
```

```
## The following objects are masked from 'package:ggplot2':
##
## %+%, alpha
help(package= "ggplot2")
#vignette("ggplot2-specs", package = "ggplot2")
```

#### Explore your data

#### Load data

Usually you will load three type of data: \* Excel files: .xslx \* Comma separated values: .csv \* R data file: .RData

You load them differently

```
# Load a RData
setwd("~/Google Drive/Master Students/courses/introduction_a_r")
load("data/raw/data.RData")
pasdenom = read.csv("~/Google Drive/Master Students/courses/introduction_a_r/data/raw/data.csv")
class(pasdenom)
## [1] "data.frame"
#or if you want to rename your data
renamedData = get(load("data/raw/data.RData"))
```

#### When in doubt, google it!

```
# But.. how to load EXCEL FILES ?
# TODO Check stackoverflow / Google and load excel and csv file
# "load xlsx file R"
# "load csv file in R"
# can you copy paste ?? read.clipboard
```

#### Explore your data

## isOlder\*

## IQ

Looking at your data before starting asking question is important to detect errors you might have made, wrong IDs, numbers of NA, wacky values... https://cran.r-project.org/web/packages/psych/vignettes/overview.pdf

```
# View(data)
# Psych has a lot of tools for exploratory analysis
library(psych)
describe(data)
##
                                      sd median trimmed
                vars
                      n
                            mean
                                                            mad
                                                                    min
                           41.06 14.42
                                           41.00
                                                   41.01
                                                          19.27
                                                                  18.00
## age
                   1 650
```

NA

97.49

 ${\tt NaN}$ 

98.64

NA

20.02

42.75

NA

 $\mathtt{NaN}$ 

99.62 19.65

2 650

3 650

```
## id*
                    6 650
                               NaN
                                       NA
                                                NA
                                                       NaN
                                                                NA
                                                                        Inf
                                              1.00
## university
                    7 650
                              0.63
                                     0.48
                                                       0.67
                                                              0.00
                                                                       0.00
                                   skew kurtosis
##
                     max
                           range
                                                     se
## age
                   65.00
                            47.00
                                   0.01
                                            -1.30
                                                   0.57
## isOlder*
                    -Inf
                             -Inf
                                     NA
                                               NA
## IQ
                  167.33
                          124.58
                                   0.43
                                             0.15
                                                   0.77
## responseTime 3376.00 2273.00
                                             0.50 16.14
                                  0.78
## performance
                    1.00
                             1.00 -0.01
                                            -0.26
                                                   0.01
## id*
                    -Inf
                             -Inf
                                     NA
                                               NA
                                                     NA
## university
                    1.00
                             1.00 -0.55
                                            -1.70 0.02
describeBy(data, group = "isOlder")
## $`FALSE`
##
                                            median trimmed
                                                               mad
                                                                        min
                 vars
                        n
                              mean
                                       sd
## age
                                    12.84
                                             39.00
                                                     38.18
                                                                      18.00
                    1 576
                             38.24
                                                             17.79
## isOlder*
                    2 576
                               NaN
                                       NA
                                                NA
                                                       NaN
                                                                NA
                                                                        Inf
## IQ
                    3 576
                           101.83
                                    19.39
                                            100.25
                                                    100.96
                                                             19.11
                                                                      42.75
                    4 576 1867.61 334.68 1826.00 1848.75 345.45 1103.00
## responseTime
                                              0.52
                                                                       0.00
                    5 576
                              0.51
                                     0.18
                                                       0.51
                                                              0.17
## performance
## id*
                    6 576
                               NaN
                                       NA
                                                NA
                                                       NaN
                                                                NA
                                                                        Inf
## university
                    7 576
                              0.64
                                     0.48
                                              1.00
                                                       0.67
                                                              0.00
                                                                       0.00
##
                     max
                           range
                                   skew kurtosis
                                                     se
## age
                   60.00
                            42.00
                                   0.04
                                            -1.31
                                                   0.54
## isOlder*
                    -Inf
                             -Inf
                                     NA
                                               NA
                                                     NA
                  167.33
                                             0.20
## IQ
                          124.58
                                   0.36
                                                   0.81
## responseTime 2927.00 1824.00
                                   0.49
                                            -0.23 13.94
## performance
                    1.00
                             1.00 -0.02
                                            -0.28
                                                  0.01
## id*
                    -Inf
                             -Inf
                                     NA
                                               NA
                                                     NA
## university
                    1.00
                             1.00 -0.56
                                            -1.69 0.02
##
## $`TRUE`
##
                                          median trimmed
                            mean
                                      sd
                                                              mad
                                                                       min
                                                                               max
                 vars
                      n
## age
                    1 74
                            62.95
                                    1.38
                                            63.00
                                                    62.93
                                                             1.48
                                                                     61.00
                                                                             65.00
## isOlder*
                    2 74
                              NaN
                                      NA
                                               NA
                                                       NaN
                                                               NA
                                                                       Inf
                                                                              -Inf
                    3 74
                                            81.73
                                                    81.99
                                                             9.98
## IQ
                            82.44
                                   11.45
                                                                     58.26
                                                                            111.36
                    4 74 2607.04 361.98 2533.00 2583.37 394.37 1969.00 3376.00
## responseTime
                    5 74
                                                     0.41
                                                                      0.06
## performance
                             0.40
                                    0.15
                                             0.42
                                                             0.15
                                                                              0.72
                                                                              -Inf
## id*
                    6 74
                              NaN
                                      NA
                                               NA
                                                      NaN
                                                                       Inf
                                                               NA
                    7 74
                                                     0.65
                                                             0.00
                                                                      0.00
                                                                              1.00
## university
                             0.62
                                    0.49
                                             1.00
##
                   range
                          skew kurtosis
                                             se
## age
                    4.00
                          0.00
                                   -1.39
                                          0.16
## isOlder*
                    -Inf
                            NA
                                      NA
                                             NΑ
## IQ
                   53.10 0.33
                                   -0.20 1.33
## responseTime 1407.00 0.52
                                   -0.67 42.08
                    0.65 -0.25
                                   -0.54 0.02
## performance
## id*
                    -Inf
                             NA
                                      NA
                                             NA
## university
                    1.00 -0.49
                                   -1.78 0.06
## attr(,"call")
## by.data.frame(data = x, INDICES = group, FUN = describe, type = type)
describeData(data)
```

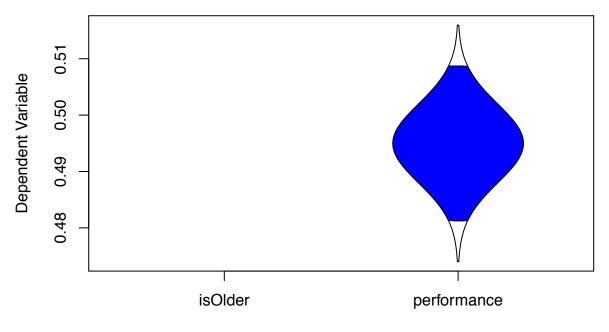
Number of variables = 7 of which all are numeri

are complete cases.

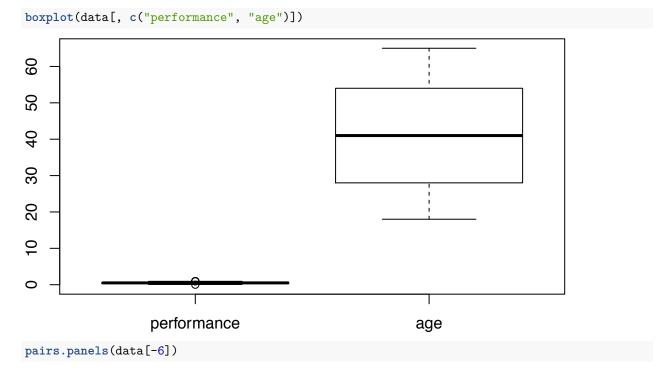
## n.obs = 650 of which 650

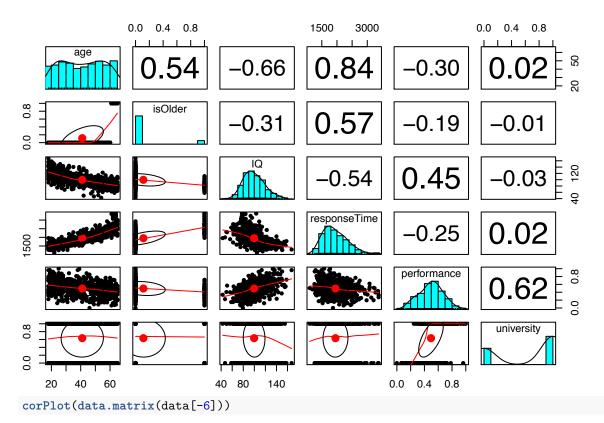
```
variable # n.obs type
                                           H1
                                                         H2
                                                                     НЗ
##
                               1
                                             57
                                                         20
## age
                       1
                           650
                                                                     41
                           650
                                          FALSE
## isOlder*
                                                      FALSE
                                                                   FALSE
## IQ
                           650
                                      82.47043
                                                   83.22762
                                                                80.90943
                                 1
## responseTime
                       4
                           650
                                1
                                           2176
                                                       1682
                                                                    2157
## performance
                       5
                           650
                               1 0.5669042
                                                  0.5501323
                                                               0.4339533
## id*
                           650
                                 3 846TVIi7vy6m 8TGgFovZhpwc M461gZCo2WCP
## university
                       7
                           650
                                              1
                                                          1
##
                        H4
                                    T1
                                                 T2
                                                             T3
## age
                        42
                                    60
                                                 65
                                                             24
## isOlder*
                    FALSE
                                 FALSE
                                              TRUE
                                                         FALSE
## IQ
                 104.19482
                               94.63783
                                           71.00707
                                                      117.35119
## responseTime
                      2122
                                   2574
                                               2211
                                                           1430
                              0.2779218
                                          0.3950018
## performance
                 0.5703210
                                                      0.4777132
              DeWYFMnTHzLv TRCgXxyJdnB9 fHTuBCMEv6XG xLWjZTfxYEit
## university
                        1
                                     0
                                                  1
##
                        Т4
## age
                        28
## isOlder*
                     FALSE
## IQ
                 110.41197
## responseTime
                      1567
## performance
                 0.6450491
## id*
               VVQZ108mdk4k
## university
head(data)
    age isOlder IQ responseTime performance
                                                         id university
                                2176 0.5669042 846TVIi7vy6m
## 1 57
          FALSE 82.47043
          FALSE 83.22762
                                       0.5501323 8TGgFovZhpwc
## 2 20
                                1682
## 3 41
          FALSE 80.90943
                                2157
                                       0.4339533 M461gZCo2WCP
## 4 42 FALSE 104.19482
                                2122 0.5703210 DeWYFMnTHzLv
                                                                      1
## 5 44 FALSE 92.82097
                                2300 0.5969814 QZAzwKbsEKVl
## 6 21 FALSE 119.71190
                               1886 0.4573477 Wz61IDoR5Vud
# Some quick plots
error.bars(data[, c("isOlder", "performance")])
```

## 95% confidence limits

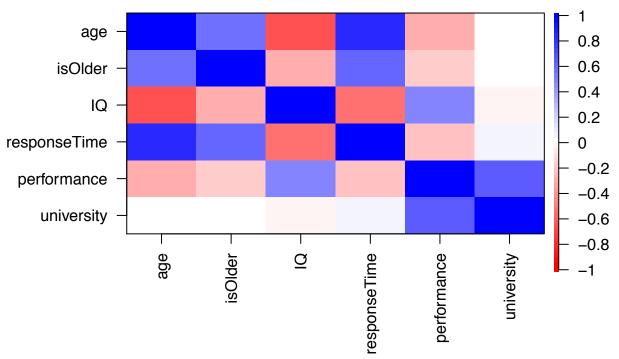


Independent Variable



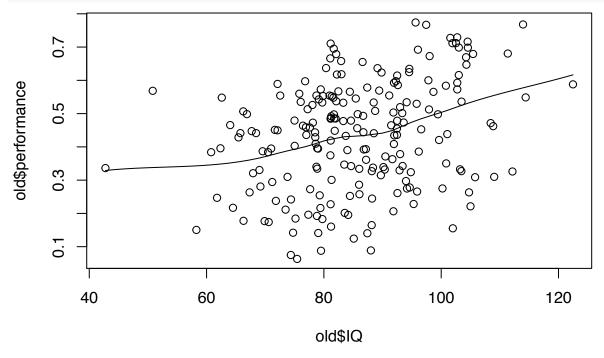


## **Correlation plot**



```
# Look for outliers
# outlier(data[, c("performance", "IQ", "responseTime")],cex=.8)
# Filter your data
```

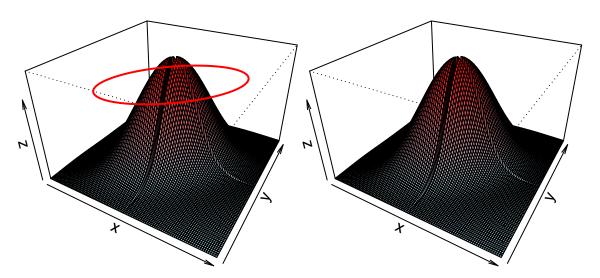
```
dataScrubbed = scrub(data,3:4,min=c(88,1300), max=c(115, 1600), newvalue=NA)
old = subset(data,data$age>50)
scatter.smooth(old$IQ, old$performance)
```



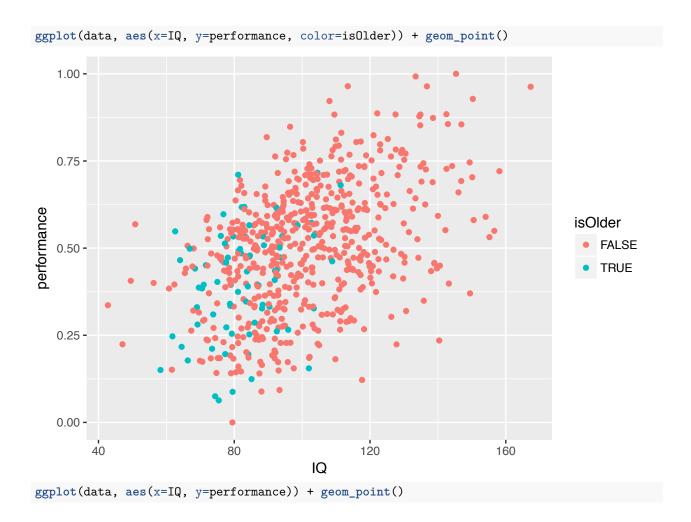
#### Correlations

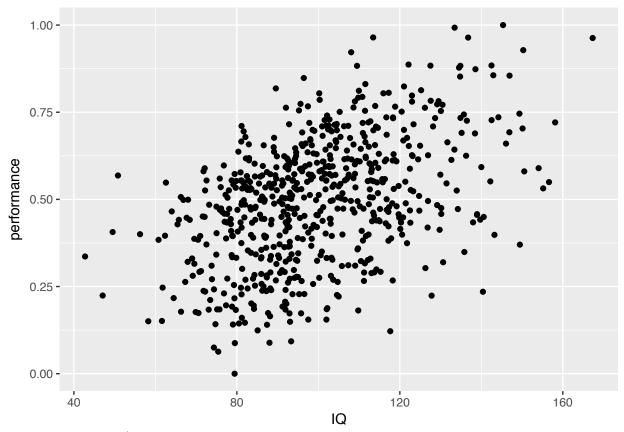
```
# What is a correlation ?
draw.cor(expand=20,cuts=c(0,0),r = 0.57)
```

## Bivariate density rho = 0.57 Bivariate density rho = 0.57



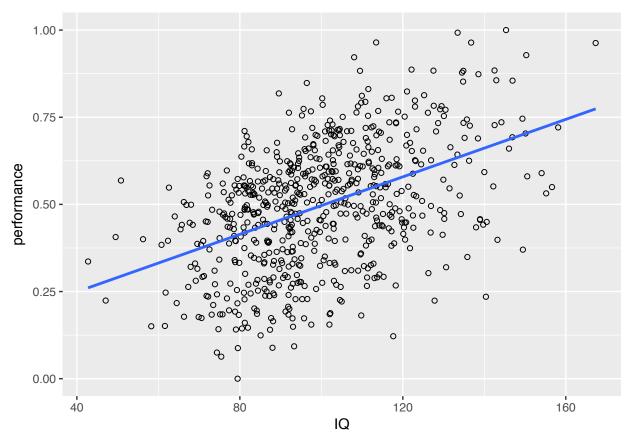
Plot Performance as a function of IQ. Performance = f(IQ).





performance = beta\*IQ + intercept + bruit

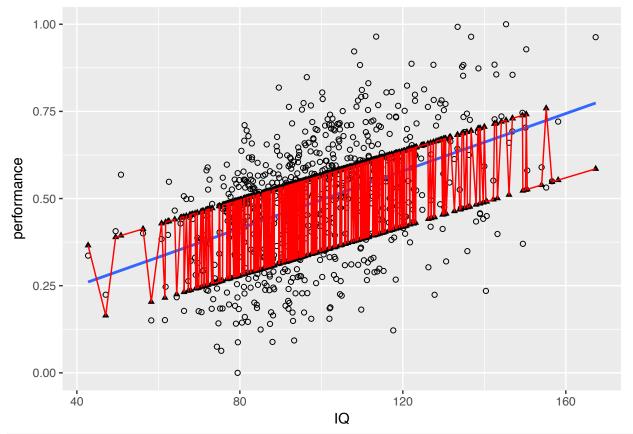
Now, what kind of relation do you think exist between the two?

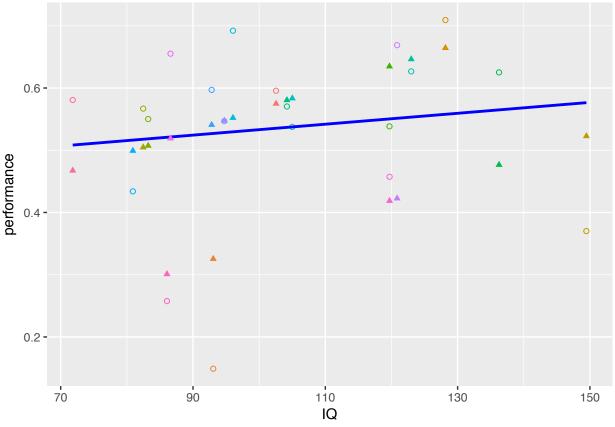


Usually you want to know the interaction between multiple variables. For example the performance might be linked to the IQ, the Age, and/or the university!

```
{\tt performance} = {\tt beta} IQ + age + university + IQ {\tt age} + {\tt intercept} + {\tt bruit}
```

```
# regression
regression = lm(formula = performance ~ IQ + age + university + IQ:age, data = data, na.action = na.omi
summary(regression)
##
##
  lm(formula = performance ~ IQ + age + university + IQ:age, data = data,
##
       na.action = na.omit)
##
## Residuals:
##
                  1Q
                       Median
                                             Max
##
  -0.30911 -0.06891 -0.00063 0.07283
                                        0.32474
##
## Coefficients:
##
                 Estimate Std. Error t value Pr(>|t|)
## (Intercept) -8.582e-02 7.511e-02
                                      -1.143
                                                 0.254
## IQ
                4.330e-03
                           6.731e-04
                                        6.433 2.43e-10 ***
                5.046e-05
                           1.695e-03
                                        0.030
                                                 0.976
## age
                2.357e-01
                           9.092e-03
                                       25.920
                                               < 2e-16 ***
## university
               -5.195e-07
                           1.705e-05
                                       -0.030
                                                 0.976
## IQ:age
                   0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## Signif. codes:
```





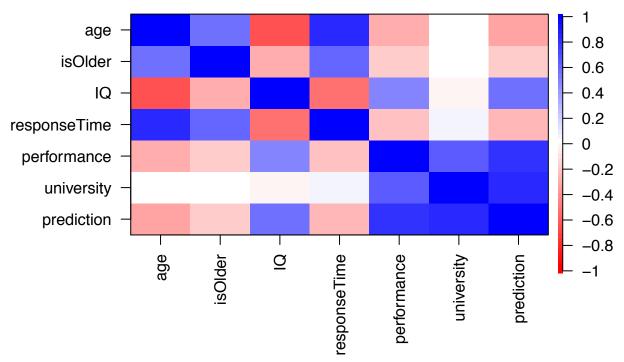
# But is this model the good one ?
regression2 = lm(formula = performance ~ IQ , data = data, na.action = na.omit)
summary(regression2)

```
##
## Call:
## lm(formula = performance ~ IQ, data = data, na.action = na.omit)
##
## Residuals:
       Min
                1Q
                     Median
                                 3Q
                                         Max
## -0.44747 -0.12092 0.02391 0.11823 0.41245
##
## Coefficients:
              Estimate Std. Error t value Pr(>|t|)
##
## (Intercept) 0.0844036 0.0322395
                                   2.618 0.00905 **
## IQ
             ## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.1589 on 648 degrees of freedom
## Multiple R-squared: 0.2064, Adjusted R-squared: 0.2051
## F-statistic: 168.5 on 1 and 648 DF, p-value: < 2.2e-16
# Anova(model1, model2, test="Chisq") allows to compare models
anova(regression, regression2, test="Chisq")
```

```
## Analysis of Variance Table
##
```

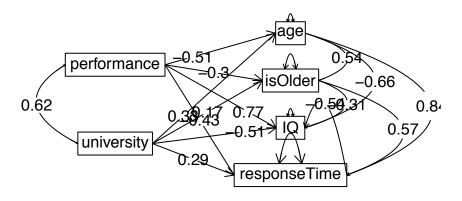
```
## Model 1: performance ~ IQ + age + university + IQ:age
## Model 2: performance ~ IQ
     Res.Df
                RSS Df Sum of Sq Pr(>Chi)
## 1
        645 7.9968
## 2
        648 16.3644 -3
                         -8.3676 < 2.2e-16 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
# Other way to analyses correlation (usually through covariance matrix)
# correlation and analysis
corr.test(data.matrix(data[-6]))
## Call:corr.test(x = data.matrix(data[-6]))
## Correlation matrix
##
                  age isOlder
                                  IQ responseTime performance university
                 1.00
                         0.54 - 0.66
                                             0.84
                                                        -0.30
## age
                 0.54
                         1.00 -0.31
                                             0.57
                                                        -0.19
                                                                   -0.01
## isOlder
## IQ
                -0.66
                        -0.31 1.00
                                            -0.54
                                                         0.45
                                                                   -0.03
                         0.57 -0.54
## responseTime 0.84
                                             1.00
                                                        -0.25
                                                                    0.02
## performance -0.30
                        -0.19 0.45
                                            -0.25
                                                         1.00
                                                                    0.62
                        -0.01 -0.03
                                                                    1.00
## university
                 0.02
                                             0.02
                                                         0.62
                        -0.18 0.53
                                                         0.78
                                                                    0.83
## prediction
                -0.35
                                            -0.28
##
                prediction
## age
                     -0.35
## isOlder
                     -0.18
## IQ
                      0.53
## responseTime
                     -0.28
## performance
                      0.78
## university
                      0.83
## prediction
                      1.00
## Sample Size
## [1] 650
## Probability values (Entries above the diagonal are adjusted for multiple tests.)
                 age isOlder IQ responseTime performance university
##
                        0.00 0.0
## age
                0.00
                                          0.00
                        0.00 0.0
## isOlder
                0.00
                                          0.00
                                                         0
                                                                    1
## IQ
                0.00
                        0.00 0.0
                                          0.00
                                                         0
                                                                    1
## responseTime 0.00
                                          0.00
                                                         0
                        0.00 0.0
                                                                     1
## performance 0.00
                                          0.00
                                                         0
                                                                    0
                        0.00 0.0
                0.66
                                                         0
                                                                    0
## university
                        0.82 0.4
                                          0.56
## prediction
                0.00
                        0.00 0.0
                                          0.00
                                                                    0
##
                prediction
## age
                         0
## isOlder
                         0
## IQ
                         0
## responseTime
                         0
## performance
                         0
## university
                         0
## prediction
                         0
## To see confidence intervals of the correlations, print with the short=FALSE option
corPlot(data.matrix(data[-6]))
```

## **Correlation plot**



# Multivariate correlation predict y columns with x
setCor(y = 1:4,x=c(5,7),data=data)

## **Regression Models**



## unweighted matrix correlation = -0.19

```
## Call: setCor(y = 1:4, x = c(5, 7), data = data)
##
## Multiple Regression from raw data
##
## Beta weights
## age isOlder IQ responseTime
## performance -0.51 -0.30 0.77 -0.43
```

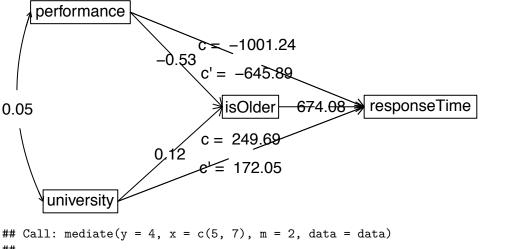
```
## university 0.33 0.17 -0.51
                                         0.29
##
## Multiple R
##
                    is0lder
                                    IQ responseTime
           age
##
          0.40
                       0.23
                                   0.61
                                                0.34
## multiple R2
                    isOlder
                                      IQ responseTime
           age
                      0.054
##
         0.158
                                   0.368
                                               0.116
##
##
  Unweighted multiple R
##
           age
                   isOlder
                                    IQ responseTime
          0.18
                       0.11
                                                0.15
##
                                   0.27
  Unweighted multiple R2
##
##
                   is0lder
                                      IQ responseTime
           age
##
          0.03
                       0.01
                                   0.07
                                                0.02
##
##
  SE of Beta weights
              age isOlder IQ responseTime
##
## performance 0.05
                    0.05 0.04
                                   0.05
## university 0.05
                      0.05 0.04
                                       0.05
##
  t of Beta Weights
##
                 age isOlder
                                 IQ responseTime
## performance -11.02 -6.05 19.40
                                        -9.19
## university 7.23 3.58 -12.89
                                           6.20
## Probability of t <
                  age isOlder IQ responseTime
## performance 0.0e+00 2.4e-09 0
                                      0e+00
## university 1.4e-12 3.7e-04 0
                                       1e-09
##
##
   Shrunken R2
##
                    is0lder
           age
                                      IQ responseTime
##
         0.156
                      0.051
                                  0.367
                                               0.113
##
## Standard Error of R2
##
          age
                isOlder
                                      IQ responseTime
##
         0.026
                     0.017
                                 0.030
                                               0.023
##
## F
##
                    is0lder
                                     IQ responseTime
          age
                      18.35
                                 188.74
##
         60.85
                                               42.44
## Probability of F <
                    is0lder
                                      IQ responseTime
           age
                   1.77e-08
                                            0.00e+00
##
      0.00e+00
                                0.00e+00
  degrees of freedom of regression
## [1] 2 647
##
## Various estimates of between set correlations
## Squared Canonical Correlations
## [1] 0.371 0.001
## Chisq of canonical correlations
```

```
## [1] 299.24  0.66
##
## Average squared canonical correlation = 0.19
## Cohen's Set Correlation R2 = 0.37
## Shrunken Set Correlation R2 = 0.36
## F and df of Cohen's Set Correlation 41.84 8 1280
## Unweighted correlation between the two sets = -0.19
```

#### Mediation analysis

```
mediate(y = 4, x = c(5,7), m = 2, data = data)
```

#### **Mediation model**

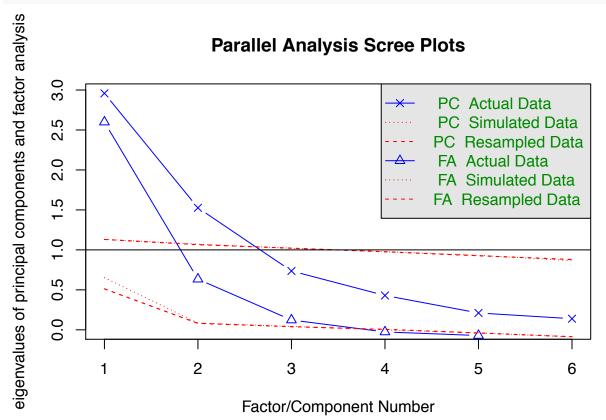


```
## The DV (Y) was responseTime . The IV (X) was performance university . The mediating variable(s) =
## Total Direct effect(c) of performance on responseTime = -1001.24
                                                                       S.E. = 108.92 t direct =
## Direct effect (c') of performance on responseTime removing isOlder = -645.89
## Indirect effect (ab) of performance on responseTime through is01der = -355.35
## Mean bootstrapped indirect effect = -354.54 with standard error = 58.72 Lower CI = -476.11
## Total Direct effect(c) of university on responseTime = 249.69
                                                                    S.E. = 40.27 t direct = 6.2
## Direct effect (c') of university on NA removing isOlder = 172.05
                                                                        S.E. = 34.28 t direct =
## Indirect effect (ab) of university on responseTime through isOlder
## Mean bootstrapped indirect effect = -354.54 with standard error = 58.72 Lower CI = 36.8
## R2 of model = 0.37
  To see the longer output, specify short = FALSE in the print statement
##
  Full output
##
## Total effect estimates (c)
              responseTime
                                    t Prob
                              se
## performance
                 -1001.24 108.92 -9.19 0e+00
## university
                   249.69 40.27 6.20 1e-09
```

```
##
## Direct effect estimates
                                (c')
               responseTime
                                       t
                    -645.89 94.39 -6.84 1.80e-11
##
  performance
##
  university
                     172.05 34.28 5.02 6.73e-07
##
         effect estimates
               is0lder
##
                         se
                                 t
  performance
                 -0.53 0.09 -6.05 2.40e-09
                  0.12 0.03 3.58 3.73e-04
  university
##
##
         effect estimates
##
           responseTime
                                   t Prob
                           se
##
  is0lder
                 674.08 41.45 16.26
##
##
    'ab' effect estimates
##
               responseTime
                                             lower
                                boot
                                        sd
## performance
                    -355.35 -354.54 58.72 -476.11 -244.53
## university
                      77.64
                              77.81 21.54
                                             36.80 122.48
```

## Dimensionality reduction

```
# How many components should you expect ?
dataWithoutId = data[-c(6,8)]
fa.parallel(dataWithoutId)
```

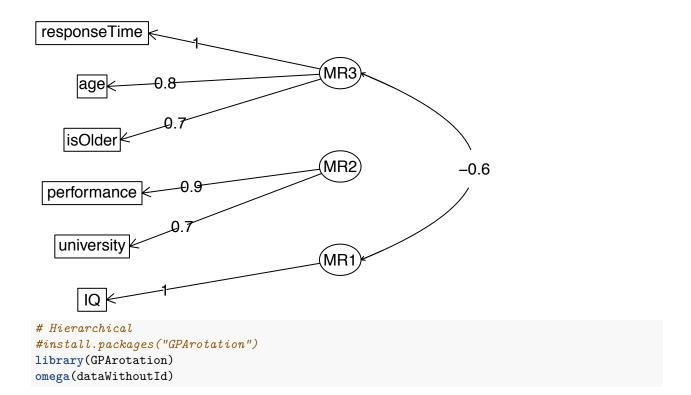


```
## Parallel analysis suggests that the number of factors = 3 and the number of components = 2
# Data loads on variable
principal(dataWithoutId, nfactors = 2)
## Principal Components Analysis
## Call: principal(r = dataWithoutId, nfactors = 2)
## Standardized loadings (pattern matrix) based upon correlation matrix
##
                 RC1
                       RC2
                             h2
                                  u2 com
                 0.93 -0.06 0.86 0.14 1.0
## age
                0.71 -0.01 0.50 0.50 1.0
## isOlder
                -0.75 0.20 0.60 0.40 1.1
## responseTime 0.90 -0.02 0.81 0.19 1.0
## performance -0.31 0.88 0.86 0.14 1.2
## university
                0.12 0.91 0.84 0.16 1.0
##
##
                          RC1 RC2
## SS loadings
                         2.84 1.64
## Proportion Var
                         0.47 0.27
## Cumulative Var
                         0.47 0.75
## Proportion Explained 0.63 0.37
## Cumulative Proportion 0.63 1.00
## Mean item complexity = 1.1
## Test of the hypothesis that 2 components are sufficient.
## The root mean square of the residuals (RMSR) is 0.1
## with the empirical chi square 186.72 with prob < 2.7e-39
## Fit based upon off diagonal values = 0.95
# Latent variable loads on data (age, IQ, university)
factanal(dataWithoutId, factors = 3)
##
## Call:
## factanal(x = dataWithoutId, factors = 3)
## Uniquenesses:
                     is0lder
##
            age
                                       IQ responseTime performance
##
          0.158
                       0.621
                                    0.005
                                                 0.129
                                                              0.005
##
    university
##
         0.484
##
## Loadings:
                Factor1 Factor2 Factor3
##
## age
                 0.842
                                -0.364
## isOlder
                 0.607
                -0.378
                                 0.919
## responseTime 0.909
                                -0.210
## performance -0.171
                         0.924
                                 0.335
## university
                         0.712
##
##
                 Factor1 Factor2 Factor3
## SS loadings
                   2.076 1.373
```

```
## Proportion Var 0.346 0.229 0.192
## Cumulative Var 0.346 0.575 0.766
##
## The degrees of freedom for the model is 0 and the fit was 0.0016
fa.diagram(fa(dataWithoutId, nfactors = 3))
```

## Loading required namespace: GPArotation

## **Factor Analysis**

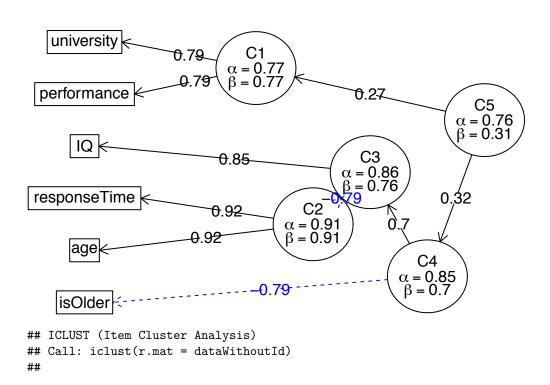


## **Omega**

```
responseTime-
                                isOlder
                             performance
                              university
                        0.9
                                   IQ
## Omega
## Call: omega(m = dataWithoutId)
## Alpha:
                          0.69
                          0.83
## G.6:
## Omega Hierarchical:
                          0.61
## Omega H asymptotic:
                          0.68
## Omega Total
                          0.9
##
## Schmid Leiman Factor loadings greater than 0.2
                     g
                         F1*
                               F2*
                                     F3*
                                           h2
                                               u2
                  0.67 -0.63
## age-
                                         0.84 0.16 0.53
## isOlder-
                  0.34 -0.51
                                         0.38 0.62 0.31
## IQ
                  0.91
                                    0.41 1.00 0.00 0.83
                                         0.87 0.13 0.38
## responseTime-
                  0.57 - 0.74
## performance
                  0.43
                              0.90
                                          1.00 0.00 0.19
## university-
                             -0.71
                                         0.52 0.48 0.00
##
## With eigenvalues of:
      g F1* F2* F3*
## 1.91 1.19 1.31 0.19
##
## general/max 1.46 max/min =
                                   6.86
## mean percent general = 0.37
                                   with sd = 0.29 and cv of 0.77
## Explained Common Variance of the general factor = 0.41
## The degrees of freedom are 0 and the fit is 0
## The number of observations was \, 650 \, with Chi Square = \, 1.05 \, with prob < NA
```

```
## The root mean square of the residuals is 0
## The df corrected root mean square of the residuals is \, NA
## Compare this with the adequacy of just a general factor and no group factors
## The degrees of freedom for just the general factor are 9 and the fit is 1.75
## The number of observations was 650 with Chi Square = 1128.6 with prob < 3.1e-237
## The root mean square of the residuals is 0.24
## The df corrected root mean square of the residuals is 0.31
## RMSEA index = 0.192 and the 90 % confidence intervals are 0.192 0.459
## BIC = 1070.31
## Measures of factor score adequacy
##
                                                   g F1* F2*
                                                                  F3*
## Correlation of scores with factors
                                                 0.91 0.88 0.99
                                                                0.52
## Multiple R square of scores with factors
                                                 0.84 0.78 0.98
## Minimum correlation of factor score estimates 0.67\ 0.56\ 0.95\ -0.45
##
##
   Total, General and Subset omega for each subset
                                                   g F1* F2* F3*
##
                                                0.90 0.87 0.31 0.99
## Omega total for total scores and subscales
## Omega general for total scores and subscales 0.61 0.36 0.27 0.82
## Omega group for total scores and subscales
                                                 0.26 0.51 0.05 0.17
#Clusters
iclust(dataWithoutId)
```

#### **ICLUST**



```
## Purified Alpha:
## [1] 0.76
##
## G6* reliability:
## [1] 0.63
##
## Original Beta:
## [1] 0.31
##
## Cluster size:
## [1] 6
## Item by Cluster Structure matrix:
##
                [,1]
## age
                -0.83
## isOlder
                -0.53
## IQ
                 0.67
## responseTime -0.78
## performance
                 0.65
## university
                 0.29
##
## With eigenvalues of:
## [1] 2.5
## Purified scale intercorrelations
## reliabilities on diagonal
## correlations corrected for attenuation above diagonal:
##
        [,1]
## [1,] 0.76
##
## Cluster fit = 0.71 Pattern fit = 0.9 RMSR = 0.2
# structural equatiom modelin
sem = esem(r = cor(dataWithoutId), varsX = c(5,6), varsY = 1:4, nfX = 2, nfY = 1,
n.obs = 650, plot = FALSE)
## The estimated weights for the factor scores are probably incorrect. Try a different factor extracti
print(sem)
## Exploratory Structural Equation Modeling Analysis using method = minres
## Call: esem(r = cor(dataWithoutId), varsX = c(5, 6), varsY = 1:4, nfX = 2,
       nfY = 1, n.obs = 650, plot = FALSE)
##
## For the 'X' set:
##
                  MR1 MR2
                -0.88 0.48
## responseTime 0.88 0.48
## For the 'Y' set:
##
                 MR1
## performance 1.00
## university
               0.62
## age
               -0.30
```

## isOlder

-0.19

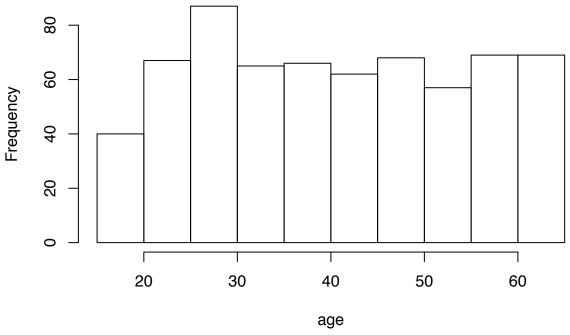
```
## Correlations between the X and Y sets.
        X1
            Х2
## X1 1.00 0.00 -0.49
## X2 0.00 1.00 0.07
## Y1 -0.49 0.07 1.00
## The degrees of freedom for the null model are 30 and the empirical chi square function was
## The degrees of freedom for the model are 0 and the empirical chi square function was 42.29
    with prob < NA
## The root mean square of the residuals (RMSR) is 0.05
## The df corrected root mean square of the residuals is NA
## with the empirical chi square 42.29 with prob < NA
\#\# The total number of observations was 650 with fitted Chi Square = 498.78 with prob < NA
## Empirical BIC = NA
## ESABIC = NA
## Fit based upon off diagonal values = 0.99
## To see the item loadings for the X and Y sets combined, and the associated fa output, print with sh
esem.diagram(sem)
```

### How to generate fake data

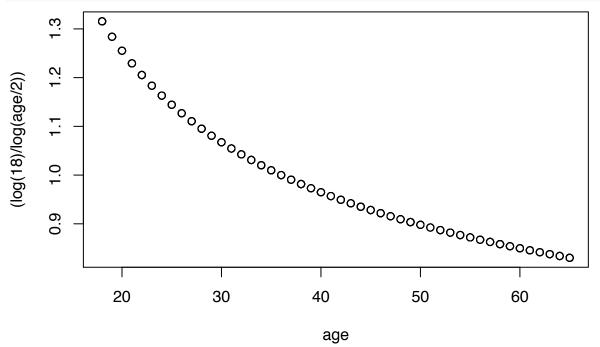
```
# Number of subjects
N = 650

# Sample from a list
age = sample(18:65, N, replace = T)
hist(age)
```

## Histogram of age



```
# Sample from gaussian distribution to generate IQ
baseIq = rnorm(N, 100, 15)
agedIq = baseIq * (log(18)/log(age/2))
data = data.frame(age = age, isOlder = age>60, IQ = agedIq)
plot(age, (log(18)/log(age/2)))
```



#### plot(age,agedIq)

```
0
      160
      140
      120
agedlq
      100
      80
                       0
      9
                                                                               000
                 20
                                  30
                                                  40
                                                                   50
                                                                                    60
                                                    age
```

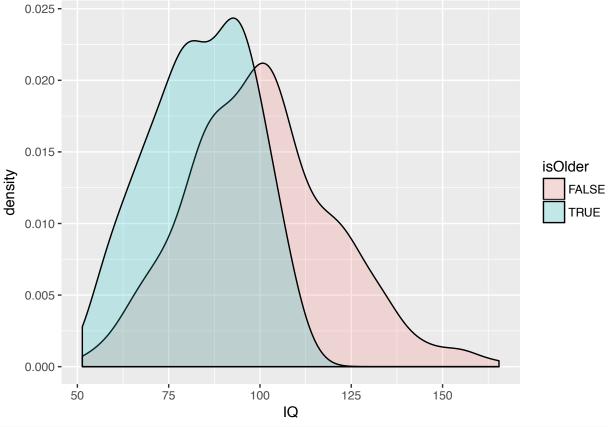
```
# A slightly more beatifull way of plotting
# install.packages("ggplot")
library(ggplot2)

# Line plot + prediction error
ggplot(data, aes(x = age, y = IQ, color=age)) + geom_point() + geom_smooth()
```

## `geom\_smooth()` using method = 'loess'



# # Histograms ggplot(data, aes(IQ, ..density.., fill = isOlder)) + geom\_density(alpha=0.2)



```
# Sample from a custom distribution

# For example the Exponential

# P(x) = lambda * exp (-lambda*x)

# lambda = 1 / mean

# CDF = 1 - exp(-lambda*x) ==> x = - ln (1-CDF) / lambda

mean = 1500

lambda = 1 / mean

CDF = sample(0:100, N, replace = T) / 100

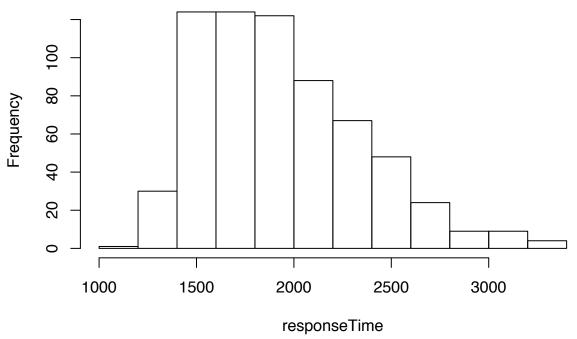
x = -log(1-CDF) / lambda
```

#### Complex distribution

Sample from from complex distribution such as Diffusion Model for Response Time Lets image you are slower with age !

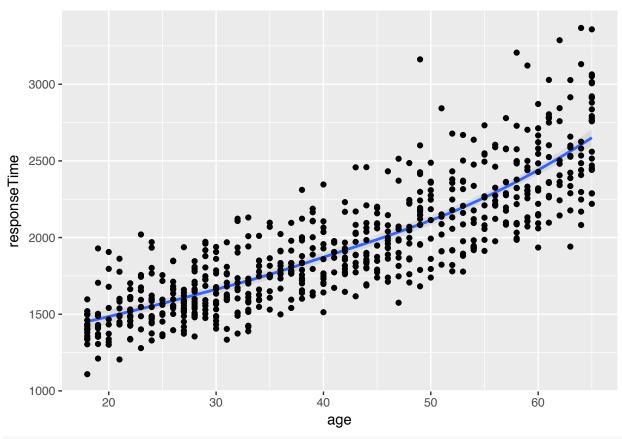
```
# Here bias depends on age
responseTime = rep(NA, N)
currentAccumulators = rep(0,N)
bias = sample(c(-1,1), N, T)+rnorm(N,0,0.1)
threshold = 1000
for(i in 1:5000) {
   currentAccumulators = currentAccumulators + bias * abs(rnorm(N, 0,1-(0.008*age)))
   indices = which(abs(currentAccumulators) > threshold)
   newIndices = indices[!indices%in%which(!is.na(responseTime))]
   responseTime[newIndices] = i
}
```

## Histogram of responseTime



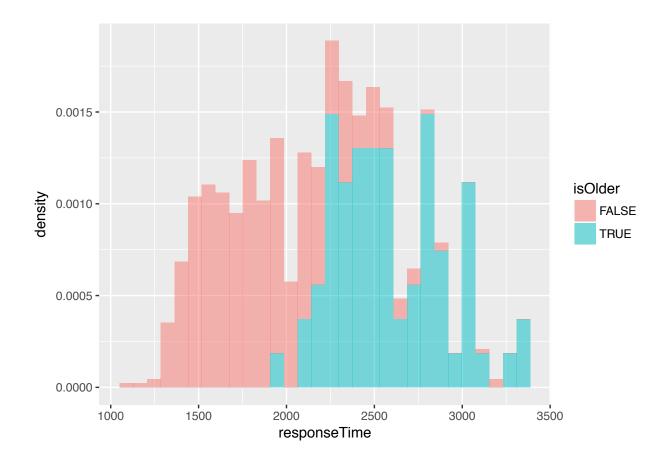
```
data$responseTime = responseTime
ggplot(data, aes(age, responseTime)) + geom_smooth(alpha=0.2) + geom_point()
```

## `geom\_smooth()` using method = 'loess'



ggplot(data, aes(responseTime, ..density.., fill=isOlder)) + geom\_histogram(alpha=0.5)

## `stat\_bin()` using `bins = 30`. Pick better value with `binwidth`.



#### Create fake correlation in the data

For example imagine IQ is linked to performance and wether or not they went to university

```
# University
university = rep(0,N)
university[which(rnorm(N)>-0.25)] = 1

# Generate fake performance linked to IQ
performance = rnorm(N,60,15)*agedIq/100 + 30*university

# Scale between 0 and 1
performance = performance - min(performance)
performance = performance / max(performance)

data$performance = performance
data$id = getRandomId(N)
data$age = age
data$university = university
```

## Save your data

```
# Set your working directory - your reference for the file system
setwd("~/Google Drive/Master Students/courses/introduction_a_r")
```

```
# Save into your data/raw
save(data, file = "data/raw/data.RData")

# CSV format
write.csv(data, file = "data/raw/data.csv", row.names = FALSE)

# XLS - needs a library
library(WriteXLS)
WriteXLS(data, "data/raw/data.xls")
```

#### Generate fake questionnaire data

```
questionaire_data = data.frame(like_sushi=sample(c(1:5),650, TRUE), like_risoto=sample(c(1:5), 650, TRUE)
WriteXLS(questionaire_data, "data/raw/questionnaire_data.xls")
## The Perl script 'WriteXLS.pl' failed to run successfully.
```

## **Example: Scaling function**

```
scale_by <- function (data, by = "minmax") {</pre>
        copied_data = data.frame(data)
        columns = names(data)
        switch(by,
               meanvar = {
                      center <- mean
                       spread <- sd
               },
               medianvar = {
                       center <- median
                       spread <- sd
               },
               minmax = {
                       center <- min
                       spread <- max
               })
        for (column in columns) {
                center_by = center(copied_data[[column]], na.rm = T)
                reduced_by = spread(copied_data[[column]], na.rm = T)
                copied_data[[column]] = (copied_data[[column]] - center_by) / reduced_by
        return(copied_data)
```

## Debugging

A good rule of thumb is that you will introduce a bug in your code every ten new lines. Besides real design issue in your code, bugs are usually due to overlooking certain extreme "use case" that you did not plan for, or because you did not pay attention to the values of your variables and content of your data.

Quick fix bugs: - Wrong variable name - Forgot a comma or parenthesis

Harder bugs: - Wrong format + Using a factor as a string or numeric + Some NA in the data + Some misformated strings or outlier values - Wrong data size - Silent bugs - return incorect values (+/- design issues)

Three steps to correct a bug: - Reproduce - Isolate : harder part - Correct

```
# Break on error
# browser()
# breakpoints
# print
```