Vall d'Hebron
Institut de Recerca

# Reproducible Research Using RMarkdown

## Basic Statistics with R for Biomedical Research

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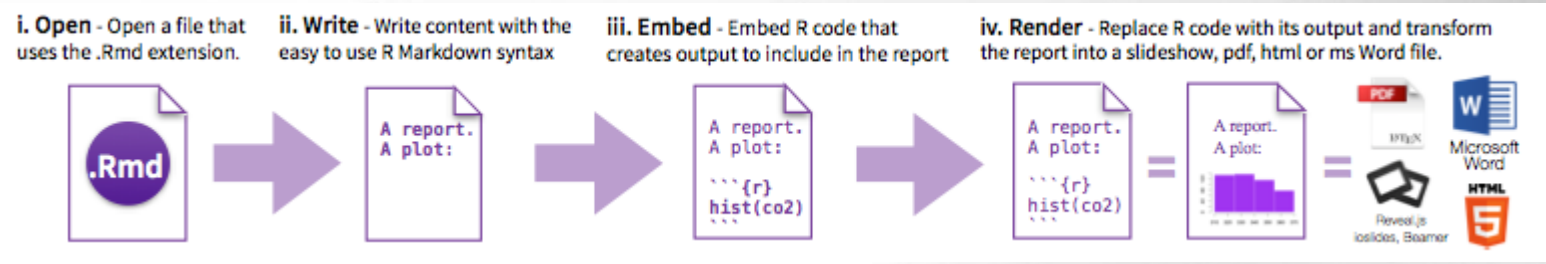
GRBIO

# Why Rmarkdown?

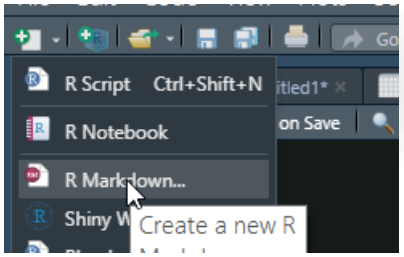


- Reproducbility (With one click, literally)
- Organization
- Share-ability (easily share or publish knitted files to HTML/RPubs)
- Annotate-ability (now I'm just making these up… Markdown makes it easy to annotate your code)
- Insert chunks of Python, Bash, SQL code

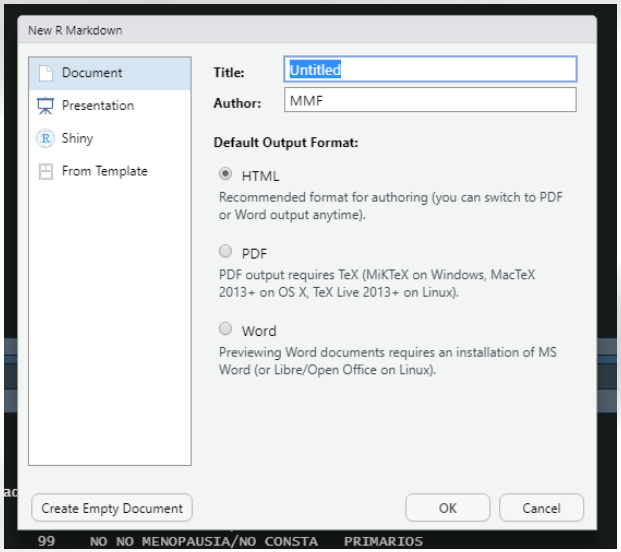
# Workflow




## Open an Rmd File




## Choose output



# Rmarkdown Example

```
---
title: "Example Rmarkdown"
author: "MMF"
date: "29/9/2020"
output: html_document
---

```{r setup, include=FALSE}
knitr::opts_chunk$set(echo = TRUE)
```

## R Markdown

This is an R Markdown document. Markdown is a simple formatting synt
<http://rmarkdown.rstudio.com>.

When you click the **Knit** button a document will be generated that
document. You can embed an R code chunk like this:

```{r cars}
summary(cars)
```

## Including Plots

You can also embed plots, for example:

```{r pressure, echo=FALSE}
plot(pressure)
```

Note that the `echo = FALSE` parameter was added to the code chunk t
```

# Example Rmarkdown

MMF

29/9/2020

## R Markdown

This is an R Markdown document. Markdown is a simple formatting syntax for authoring HTML, PDF, and MS Word documents. For more details on using R Markdown see http://rmarkdown.rstudio.com.

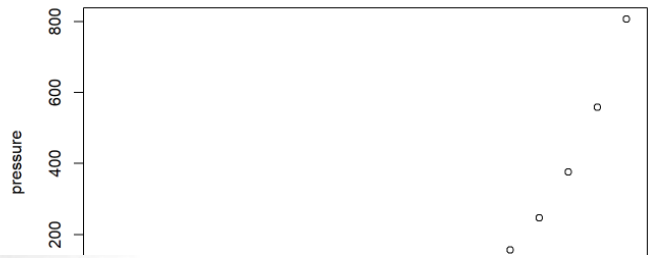
When you click the **Knit** button a document will be generated that includes both content as well as the output of any embedded R code chunks within the document. You can embed an R code chunk like this:

```
summary(cars)
```

```
##      speed           dist
##  Min.   : 4.0   Min.   :  2.00
##  1st Qu.:12.0   1st Qu.: 26.00
##  Median :15.0   Median : 36.00
##  Mean   :15.4   Mean   : 42.98
##  3rd Qu.:19.0   3rd Qu.: 56.00
##  Max.   :25.0   Max.   :120.00
```

## Including Plots

You can also embed plots, for example:



Vall d'Hebron
Institut de Recerca

# Descriptive Statistics: Summaries and Graphs

## Basic Statistics with R for Biomedical Research

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# Index

# GENERAL CONCEPTS



- **Population**: It represents the largest group of individuals who want to study and generally usually inaccessible.
- **Sample:** Subset of the population in which measurements are done. This sample should be representative of the original population (any individual has equal opportunity to be selected).
- **Variable**: Feature measurable and observable that represents a concept of study
- **Measure:** Procedure for assign quantitative or qualitative values to the characteristics of objects, people or events. If these procedures are not well measured the validity of the results is not guaranteed.

# STEPS IN A STATISTICAL STUDY ANALSYIS

1. **Make hypothesis about a population**
2. **Decide which data collect (Experimental design)**
   - Which individuals will be part of the study (samples)
   - Which data must be collected for each individual (variables)
3. **Collect Data**
4. **Describe (summarize) collected data**
   - Summary measures and graphs
   - Point estimations and confidence intervals
5. **Establish relations between two variables**
   - Set up Statistical Hypothesis test
   - Check application conditions
   - Calculate intensity relationship measures
6. **Multivariable analysis . Modelling**
   - Consider effects of several variables on an outcome
   - Regression models
   - More complex models

# Index

# TYPES OF VARIABLES


## QUALITATIVE

### NOMINAL

Measure qualities of an individual

**Examples: Sex, Treatment, Disease**

### ORDINAL

Measure qualities but they are ordered

**Examples: Educational level, Stage, Severity**

## QUANTITATIVE

### DISCRETE

Take only a finite possible values

**Examples: Nº of admissions, Nº of programmed visits**

### CONTINOUS

Can take an infinite number of values. Between two measures always can be another

**Example: Stay time, Age, Cholesterol level**

# Variable classification in a Study

- **Response, dependent or outcome variable**
  - One that answer the research question
- **Explain, independent or exposure variables**
  - They are those that are related to the causes of the events we want to study
- **Confounding or effect modifier variables**
  - Are those that can affect the relation between exposure and outcome variables
- **Universal variables**
  - Are those that can be exposures or confounders that always have to be considered. For example: sex, age, residence location, ethnic, etc.

# Descriptive analysis

- Data have to be **organized** to be useful (frequency or contingency tables)
- **Graph** data before calculating summary measures
- This actions can help to:
  - Select the **best summary** measure
  - **Transform** variables
  - Detect **outliers**

# Index

# Quantitative Variables

- We have a new variable (i.e. a biomarker and we want to summarize information)
  - Around which values is the variable ?
  - Values vary greatly between different individuals
  - Data are grouped or not

# Summary Measures

- Location
  - **Mean**
  - **Median**
  - Mode
- Dispersion
  - Range (Maximum-Minimum)
  - **Variance**
  - **Standard Deviation**
  - Variation Coefficient
  - Percentile
  - **Interquartile range (IQR)** or Interquartile interval
- Shape
  - Asymmetry
  - Kurtosis

# Data

```
days19 <-  c(3, 4, 6, 9, 12 )

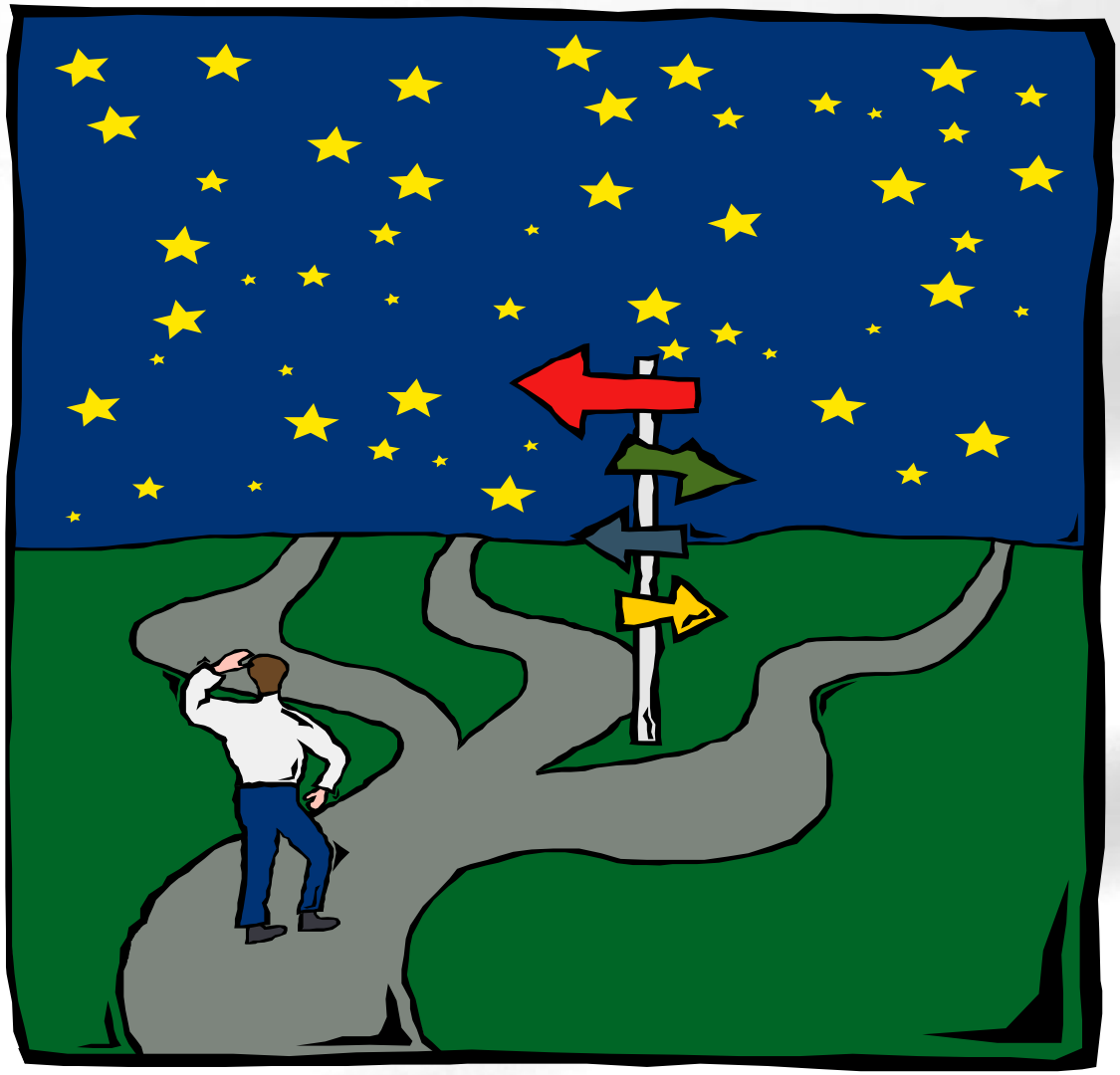
days19

days20 <-  c(3, 4, 6, 9, 20 )

days20
```

```
> days19 <-  c(3, 4, 6, 9, 12 )
> days19
[1]  3  4  6  9 12
> days20 <-  c(3, 4, 6, 9, 20 )
> days20
[1]  3  4  6  9 20
```

# Location measures



Mean

Median

Mode

# Mean


- Useful to locate data .
- Is the **sum** of observed values **over sample size**
- Can be altered by extreme values

$$\bar{x} = \frac{1}{n}\sum_{i=1}^{n} x_i$$

**Example Stay days**

**3, 4, 6, 9, 12** **Mean=6,8**

**3, 4, 6, 9, 20** **Mean=8,4**

# Median

- Is the point that divied in **two parts** the observations
- Observations are ordered from lowest to highest and median is the **central point**
- It is not altered by extreme observations


**Example Stay days**

**3, 4, 6, 9, 12** **Median=6**

**3, 4, 6, 9, 20** **Median=6**

# Example: Location measures

CODE:

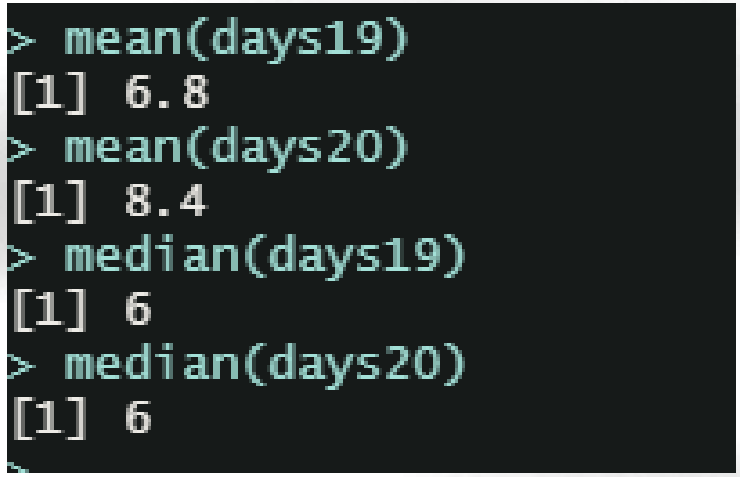
```
mean(days19)

mean(days20)

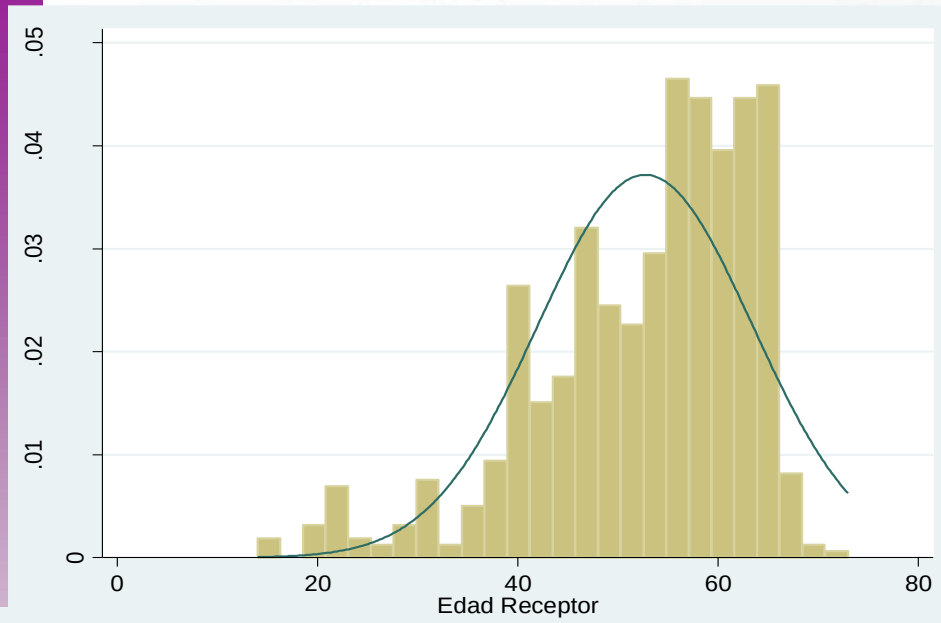
median(days19)

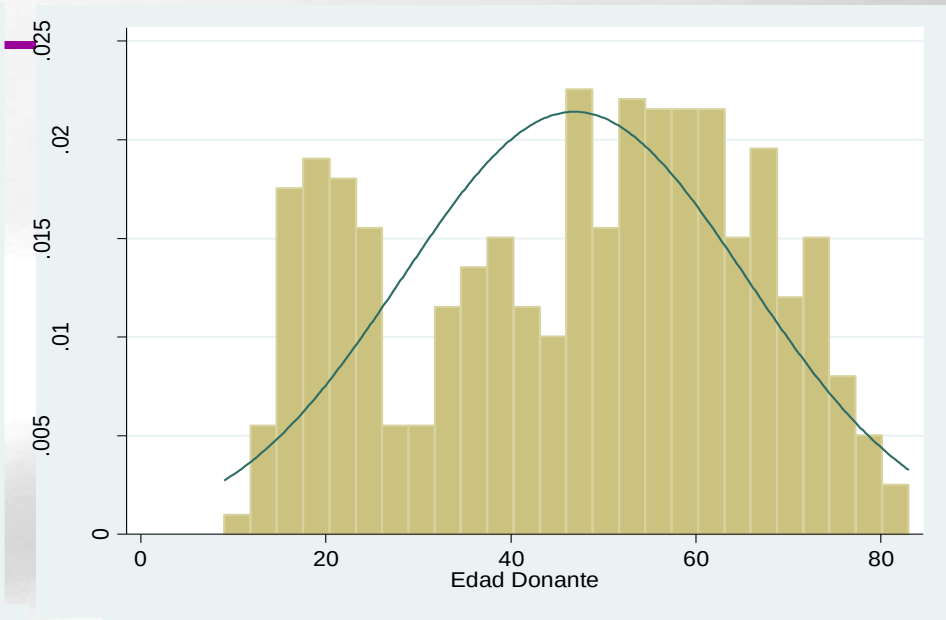
median(days20)
```

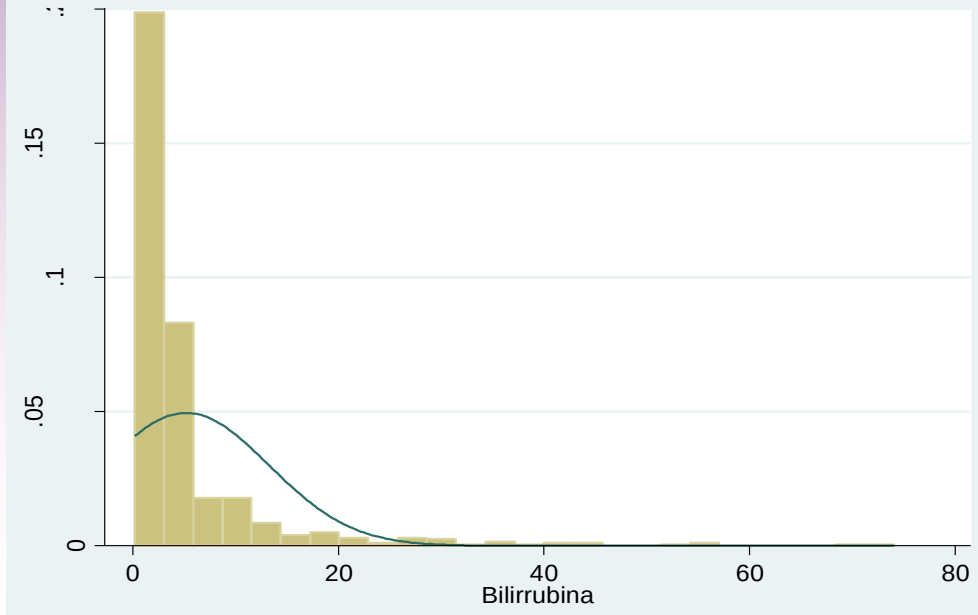
RESULT:

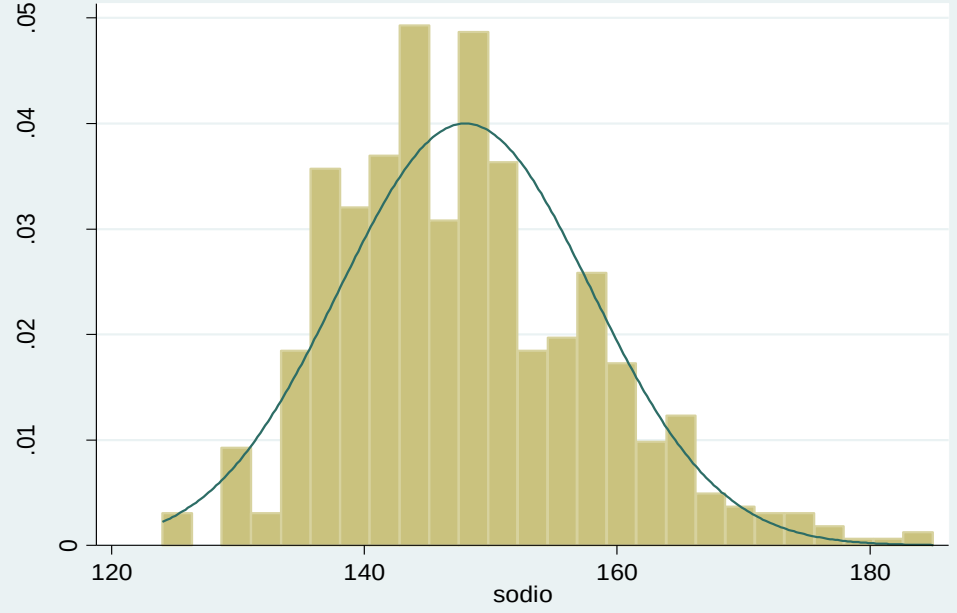


# Transplant study: Mean or Median



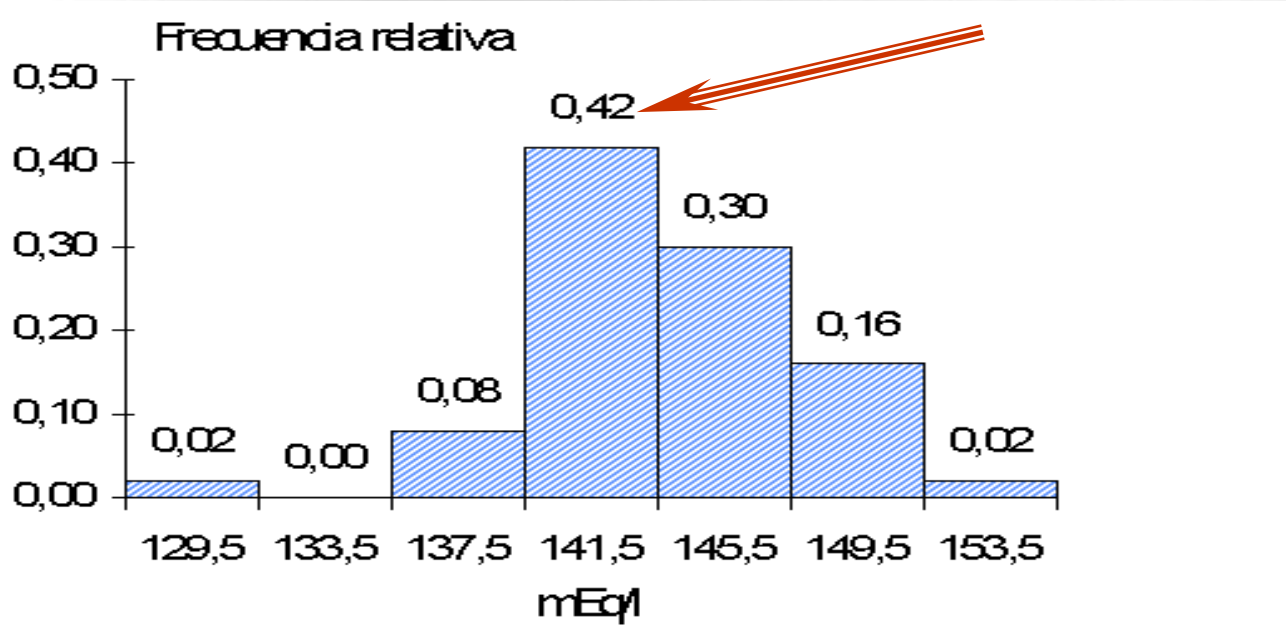



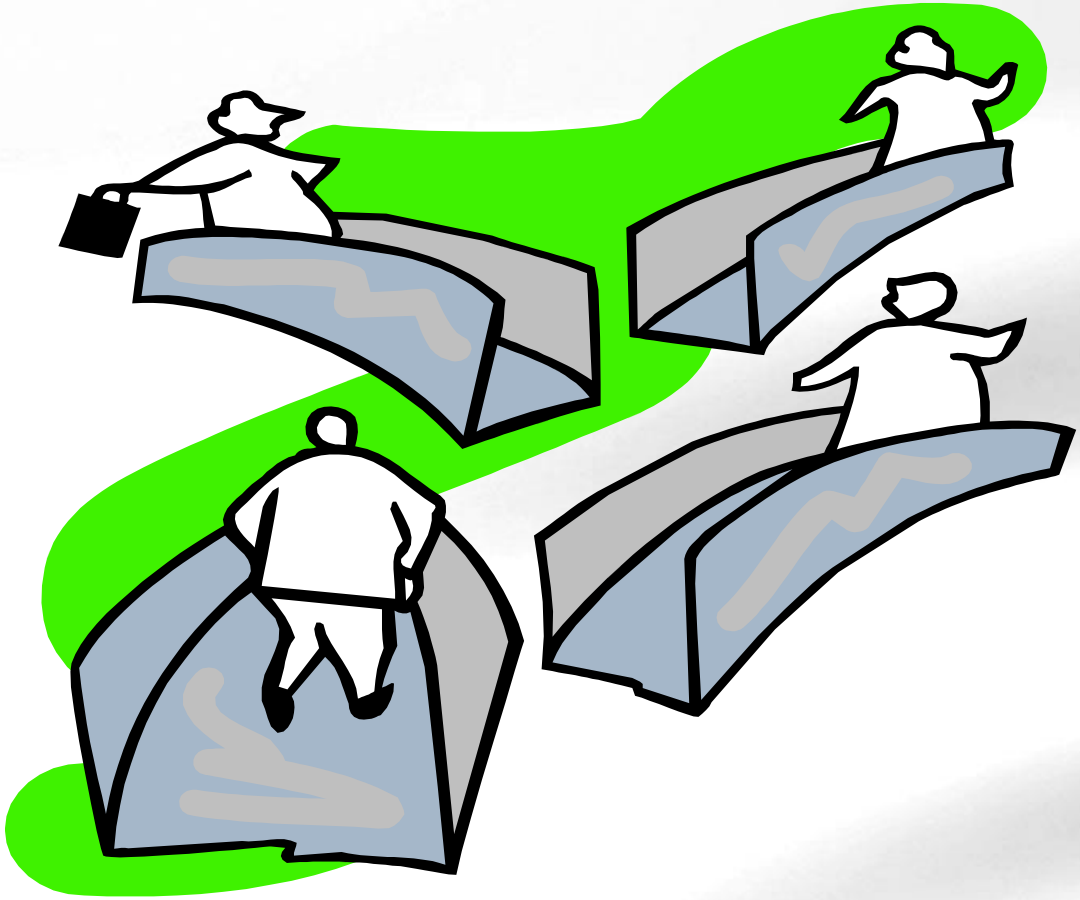
# Mode

- The most frequent value
- May be not unique
- In a quantitative variable is the maximum values of an histogram




Determinaciones de sodio

# Dispersion or variability measures

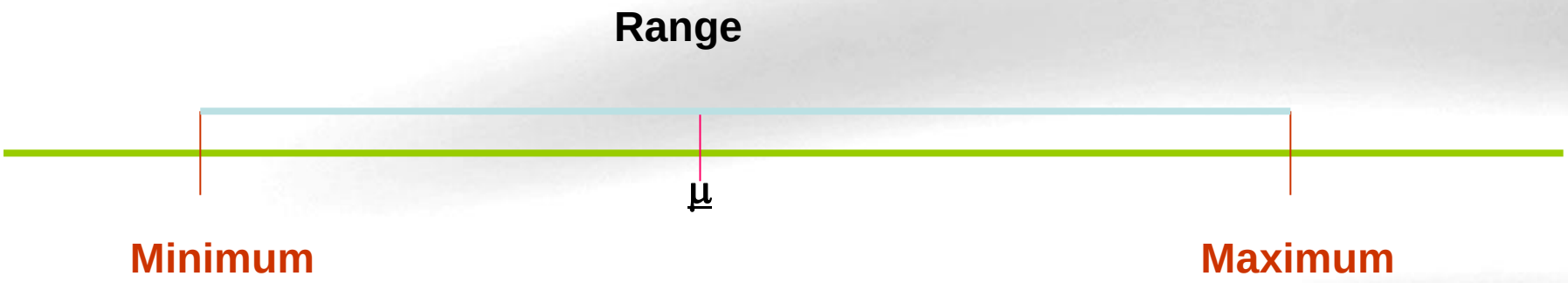


- Range ( Maximum-Minimum)
- Variance
- Standard Deviation
- Variation Coefficient
- Percentile
- Interquartile Range (IQR)

# Range



- Simplest measure of dispersion
- Is the difference between maximum and minimum value of the observations




**Example Stay days**

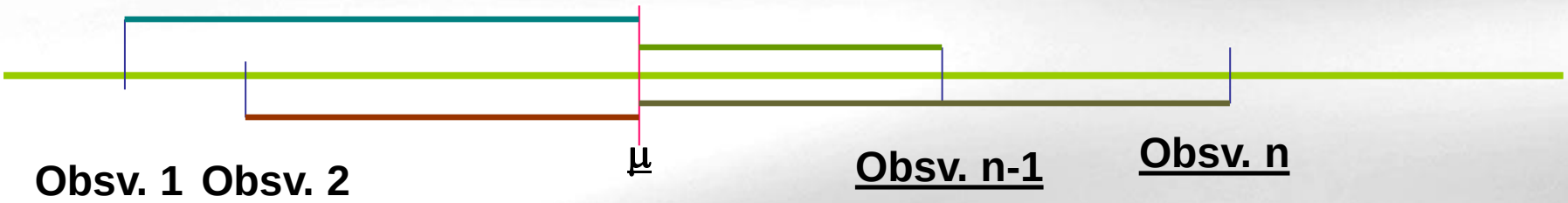
**3, 4, 6, 9, 12** **Range=12-3=9**

**3, 4, 6, 9, 20** **Range=20-3=17**

# Variance



- Mean difference of observations from mean in squared scale




| | d | d-mean | (d-mean)^2 | d | d-mean | (d-mean)^2 |
|---|---|---|---|---|---|---|
| | 3 | -3,8 | 14,44 | 3 | -5,4 | 29,16 |
| | 4 | -2,8 | 7,84 | 4 | -4,4 | 19,36 |
| | 6 | -0,8 | 0,64 | 6 | -2,4 | 5,76 |
| | 9 | 2,2 | 4,84 | 9 | 0,6 | 0,36 |
| | 12 | 5,2 | 27,04 | 20 | 11,6 | 134,56 |
| Sum | | 0 | 54,8 | | 0 | 189,2 |
| Sum/5 | 6,8 | 0 | 10,96 | 8,4 | 0 | 37,84 |

# Standard Deviation

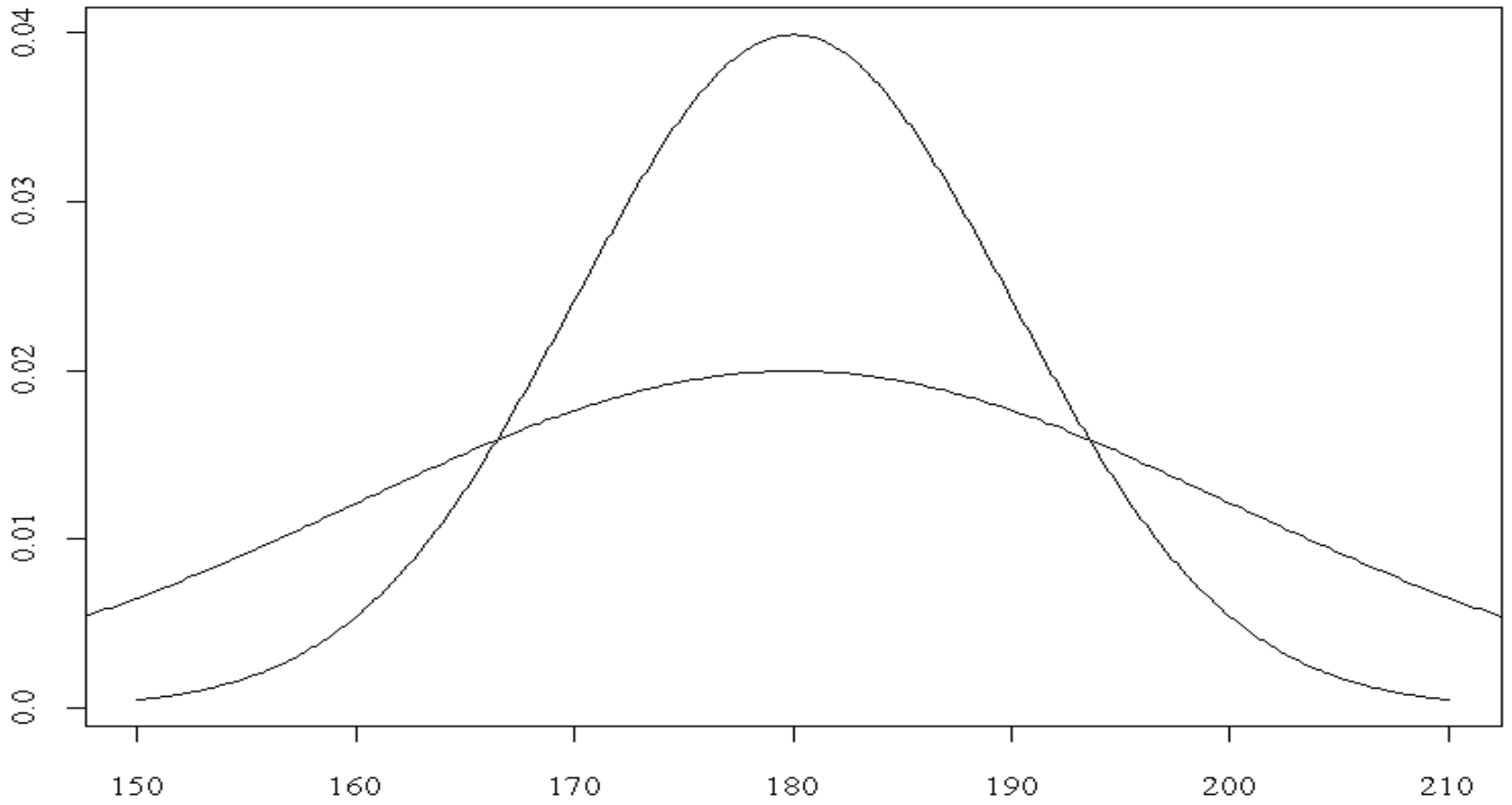


- Squared root of the variance
- It is measured in the same units than the variable

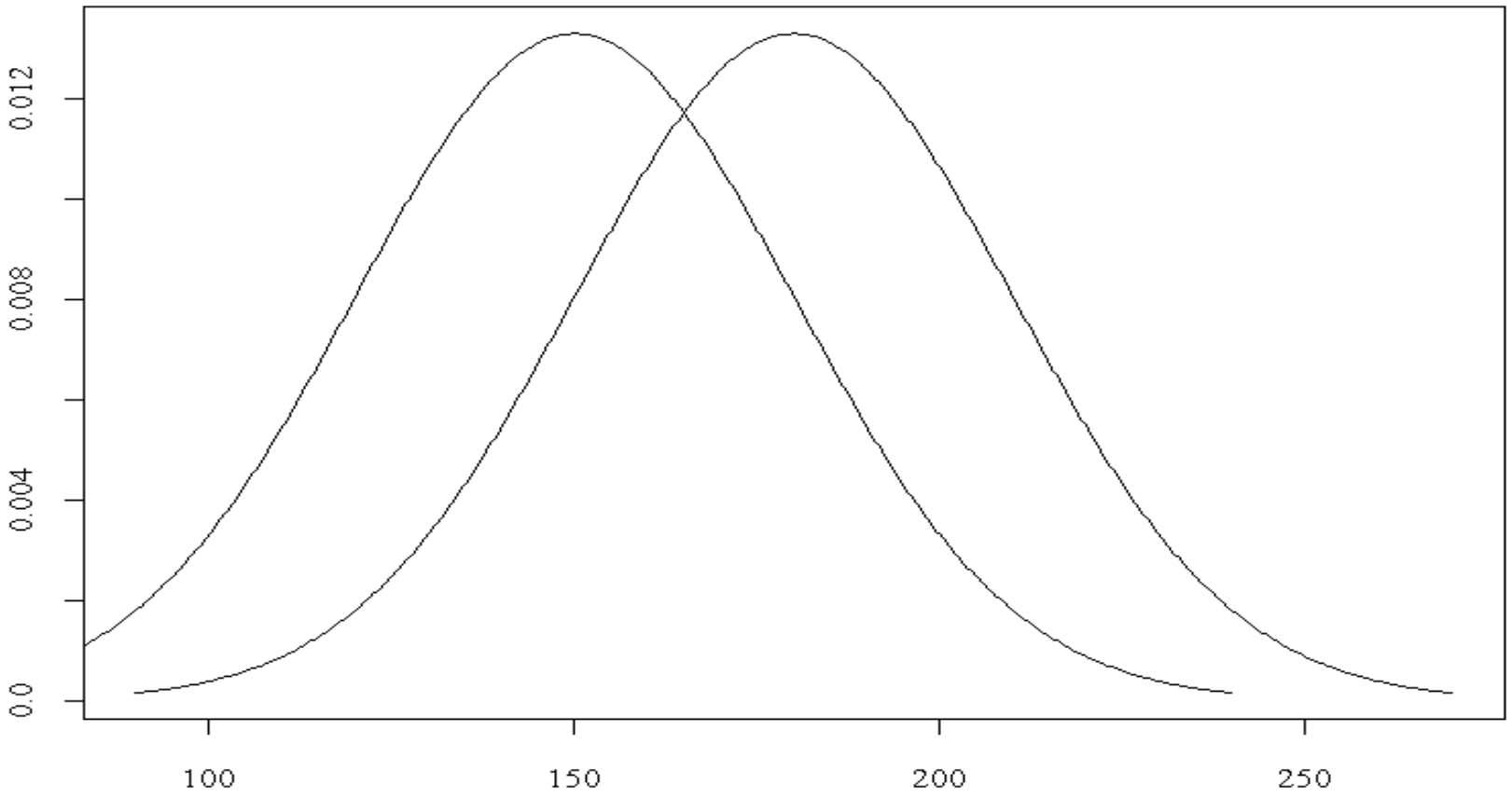
**Example Stay days**

| | |
|---|---|
| **3, 4, 6, 9, 12** | **Variance =10.96** |
| | **Std. Dev.=3.31** |
| **3, 4, 6, 9, 20** | **Variance= 37.84** |
| | **Std. Dev.= 6.15** |

# Same mean , different variances



# Same variances, different means



# Coefficient of Variation (CV)

- It is the ratio between standard deviation and mean
- Allows to compare the variability of variables measured in different scales

**Example Stay Days**

| | |
|---|---|
| **3, 4, 6, 9, 12** | **Std. Dev.= 3.31**<br>**Mean = 6.8**<br>**Variation Coef.= 0.49** |
| **3, 4, 6, 9, 20** | **Std. Dev. = 6.15**<br>**Mean = 8.4**<br>**Variation Coef.= 0.73** |

# Percentiles



- Observations are ranked from minimum to maximum and the point that leaves below p% of observations is selected
- There are some special percentiles
  - Deciles are percentiles 10, 20, 30, 40, 50, 60, 70, 80, 90
  - Quartiles are  percentiles 25, 50, 75
  - Quintiles are percentiles 20,40, 60,80
  - They are not affected by extreme observations
  - Interquartile range is difference between 25 and 75 percentile

# Example: Variability measures

CODE:

```
var(days19)

var(days20)

sd(days19)

sd(days20)

sd(days19)/mean(days19)

sd(days20)/mean(days20)

quantile(days19)

quantile(days20)
```

RESULT:

```
> var(days20)
[1] 47.3
> sd(days19)
[1] 3.701351
> sd(days20)
[1] 6.8775
> sd(days19)/mean(days19)
[1] 0.5443163
> sd(days20)/mean(days20)
[1] 0.8187499
> quantile(days19)
  0%  25%  50%  75% 100%
   3    4    6    9   12
> quantile(days20)
  0%  25%  50%  75% 100%
   3    4    6    9   20
```

# Summary measures in R

CODE:

```
summary(days19)

summary(days20)
```

RESULT:

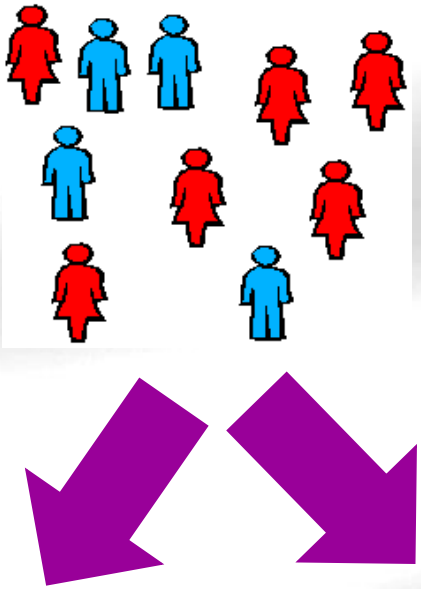
```
> summary(days19)
   Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
    3.0     4.0     6.0     6.8     9.0    12.0
> summary(days20)
   Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
    3.0     4.0     6.0     8.4     9.0    20.0
>
```

# Syllabus

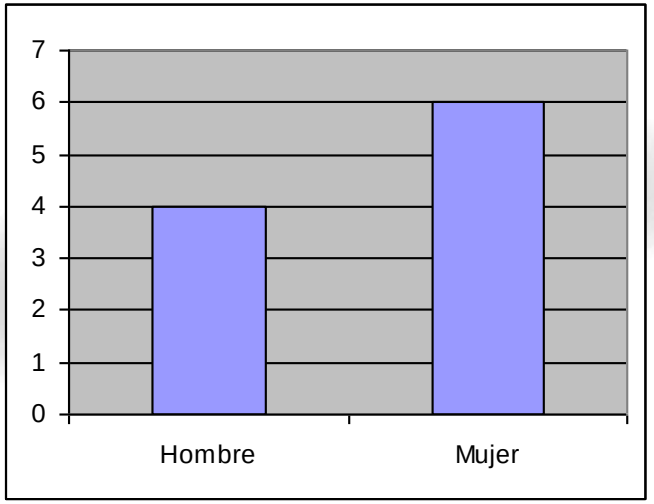
1. INTRODUCTION. ANALYSIS STRATEGY
2. VARIABLES CLASSIFICATION
3. SUMMARY MEASURES
    1. Measures of location/central tendency
    2. Measures of variability/dispersion
4. SUMMARY OF VARIABLES
    1. Contingency Tables
    2. Graphs
5. Examples & exercises

# Summary of variables

Frequency tables and graphs are two equivalent ways to present information. Both expose in an ordered way the collected data.a.



| Género | Frec. |
|---|---|
| Hombre | 4 |
| Mujer | 6 |



# Frequency table



## NÚMERO DE HIJOS ENTRE MUJERES DE 20 Y 30 AÑOS

| NÚMERO DE HIJOS | Frecuencia Absoluta ($f_i$) | Frecuencia Relativa ($fr_i$) | Frecuencia Acumulada ($F_i$) | Frecuencia Relativa Acumulada ($Fr_i$) |
|---|---|---|---|---|
| 0 | 175 | 0'35 | 175 | 0'35 |
| 1 | 225 | 0'45 | 400 | 0'80 |
| 2 | 75 | 0'15 | 475 | 0'95 |
| 3 o más | 25 | 0'05 | 500 | 1'00 |
| TOTAL | 500 | 1'00 | 500 | 1'00 |

# Frequency table



Cate go ries

## NÚMERO DE HIJOS ENTRE MUJERES DE 20 Y 30 AÑOS

| NÚMERO DE HIJOS | Frecuencia Absoluta ($f_i$) | Frecuencia Relativa ($fr_i$) | Frecuencia Acumulada ($F_i$) | Frecuencia Relativa Acumulada ($Fr_i$) |
|---|---|---|---|---|
| 0 | 175 | 0'35 | 175 | 0'35 |
| 1 | 225 | 0'45 | 400 | 0'80 |
| 2 | 75 | 0'15 | 475 | 0'95 |
| 3 o más | 25 | 0'05 | 500 | 1'00 |
| TOTAL | 500 | 1'00 | 500 | 1'00 |

# Taula de Freqüencia



Cate go ries

Nº of subjects by category

## NÚ[illegible] ENTRE MUJERES DE 20 Y 30 AÑOS

| NÚMERO DE HIJOS | Frecuencia Absoluta ($f_i$) | Frecuencia Relativa ($fr_i$) | Frecuencia Acumulada ($F_i$) | Frecuencia Relativa Acumulada ($Fr_i$) |
|---|---|---|---|---|
| 0 | 175 | 0'35 | 175 | 0'35 |
| 1 | 225 | 0'45 | 400 | 0'80 |
| 2 | 75 | 0'15 | 475 | 0'95 |
| 3 o más | 25 | 0'05 | 500 | 1'00 |
| TOTAL | 500 | 1'00 | 500 | 1'00 |

# Frequency table



Percentage of subjects
Freq /Total

Nº of subjects by category

Cate go ries

## NÚ[illegible] ENTRE MUJERES DE 20 Y 30 AÑOS

| NÚMERO DE HIJOS | Frecuencia Absoluta ($f_i$) | Frecuencia Relativa ($fr_i$) | Frecuencia Acumulada ($F_i$) | Frecuencia Relativa Acumulada ($Fr_i$) |
|---|---|---|---|---|
| 0 | 175 | 0'35 | 175 | 0'35 |
| 1 | 225 | 0'45 | 400 | 0'80 |
| 2 | 75 | 0'15 | 475 | 0'95 |
| 3 o más | 25 | 0'05 | 500 | 1'00 |
| TOTAL | 500 | 1'00 | 500 | 1'00 |

# Frequency table



Percentage of subjects
Freq /Total

Nº of subjects by category

Cate go ries

NÚMERO DE HIJOS ENTRE MUJERES DE 20 Y 30 AÑOS

| NÚMERO DE HIJOS | Frecuencia Absoluta ($f_i$) | Frecuencia Relativa ($fr_i$) | Frecuencia Acumulada ($F_i$) | Frecuencia Relativa Acumulada ($Fr_i$) |
|---|---|---|---|---|
| 0 | 175 | 0'35 | 175 | 0'35 |
| 1 | 225 | 0'45 | 400 | 0'80 |
| 2 | 75 | 0'15 | 475 | 0'95 |
| 3 o más | 25 | 0'05 | 500 | 1'00 |
| TOTAL | 500 | 1'00 | 500 | 1'00 |

Nº accumulated subects up to category
(Only ordinal or discrete variables)

# Frequency table



Percentage of subjects
Freq /Total

Nº of subjects by category

Cate go ries

NÚMERO DE HIJOS ENTRE MUJERES DE 20 Y 30 AÑOS

| NÚMERO DE HIJOS | Frecuencia Absoluta ($f_i$) | Frecuencia Relativa ($fr_i$) | Frecuencia Acumulada ($F_i$) | Frecuencia Relativa Acumulada ($Fr_i$) |
|---|---|---|---|---|
| 0 | 175 | 0'35 | 175 | 0'35 |
| 1 | 225 | 0'45 | 400 | 0'80 |
| 2 | 75 | 0'15 | 475 | 0'95 |
| 3 o más | 25 | 0'05 | 500 | 1 |
| TOTAL | 500 | 1 | 500 | |

Nº accumulated subects up to category
(Only ordinal or discrete variables)

Accumulated Frequency up to category
Freq Abs/Total

# Example: Frequency table

CODE:

```
grupo <- factor(c("A",

"A","B","B","A","C","C",

"A", "B"))

table(grupo)

library(gmodels)

CrossTable(grupo)
```

RESULT:

```
> table(grupo)
grupo
A B C
4 3 2
> library(gmodels)
> CrossTable(grupo)

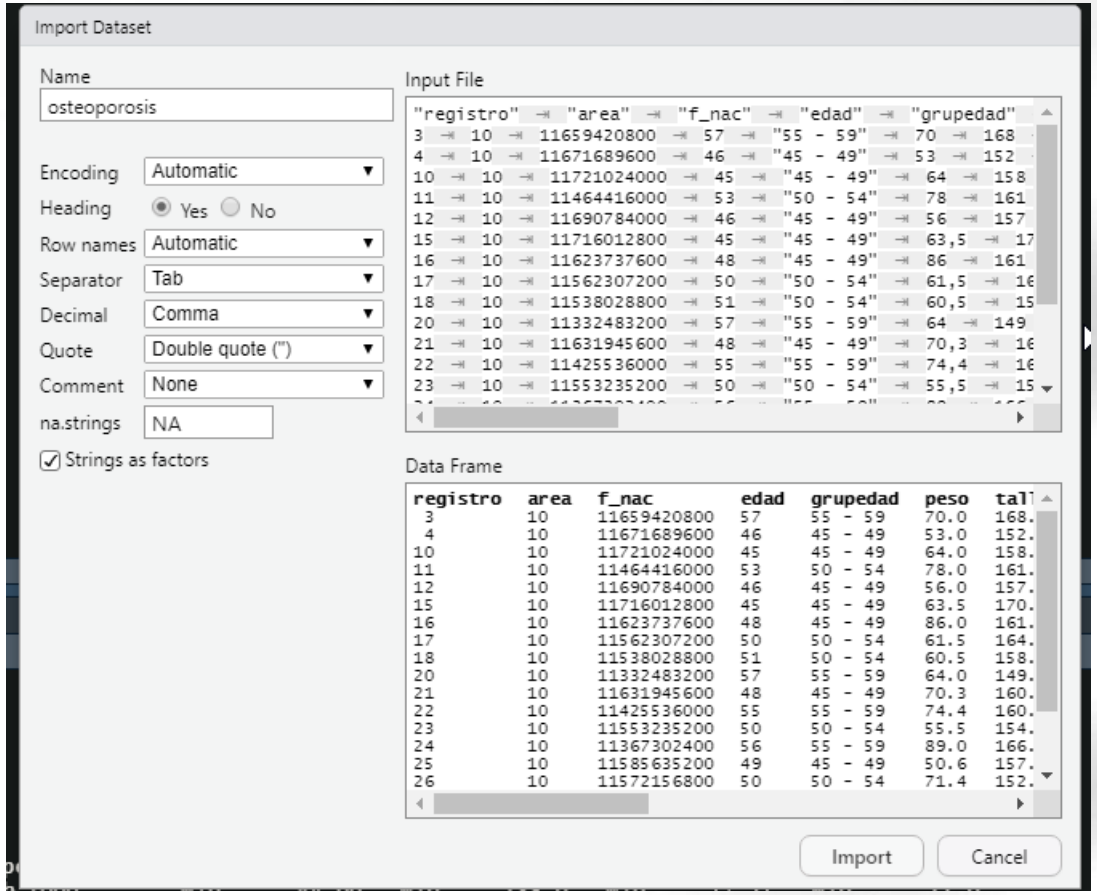

   Cell Contents
|-------------------------|
|                       N |
|         N / Table Total |
|-------------------------|


Total Observations in Table:  9


          |         A |         B |         C |
          |-----------|-----------|-----------|
          |         4 |         3 |         2 |
          |     0.444 |     0.333 |     0.222 |
          |-----------|-----------|-----------|
```

# Dataset osteoporosis

```
osteo <- read.delim2("/osteoporosis.csv")
```




```
#All dataset
summary(osteo)

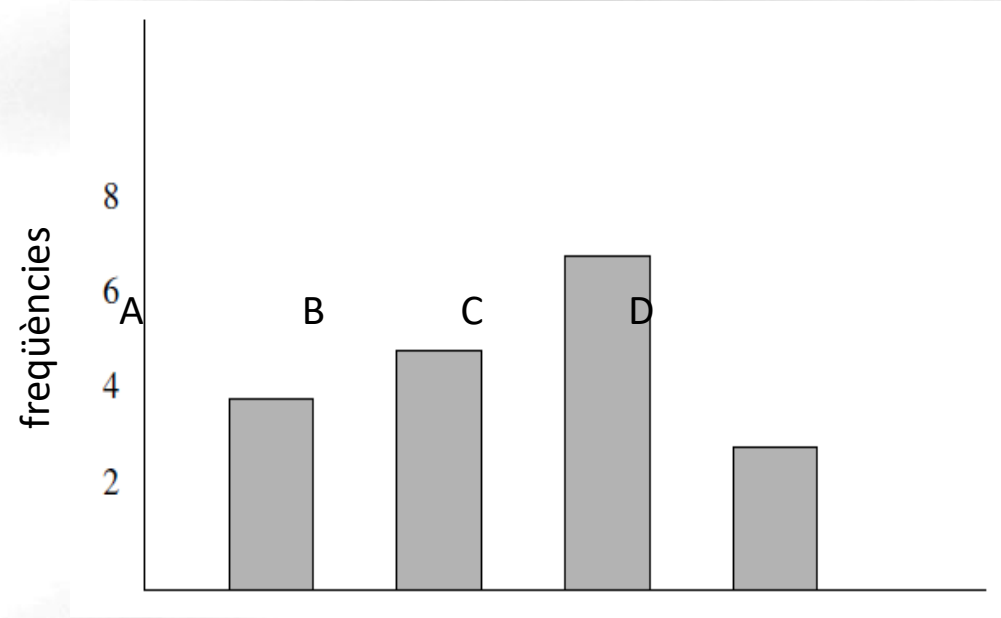
# quantitative variable
mean(osteo$edad)
sd(osteo$edad)
median(osteo$edad)
IQR(osteo$edad)

# qualitative variable
library(gmodels)
CrossTable(osteo$grupedad)
```

# Bar Graph

- Categorys are representened in X axis and frequencies in Y axis

- For comparing two population better use relative frequencies



# Example. Bar Graph

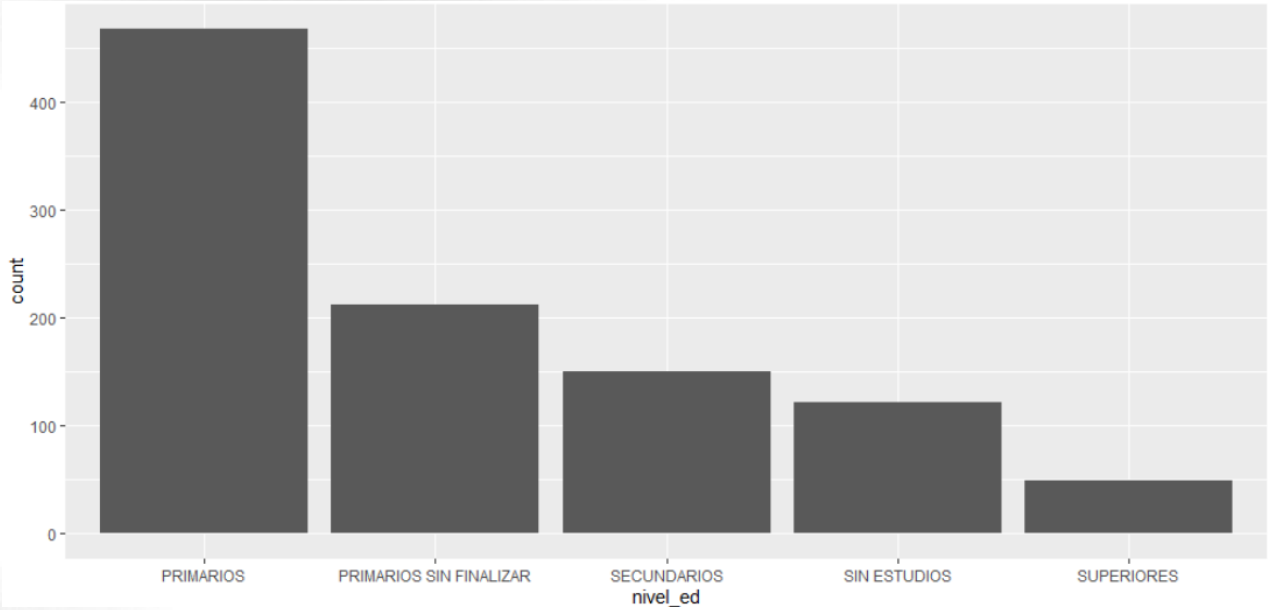
CODE:

```
require(ggplot2)

ggplot(data = osteo) +

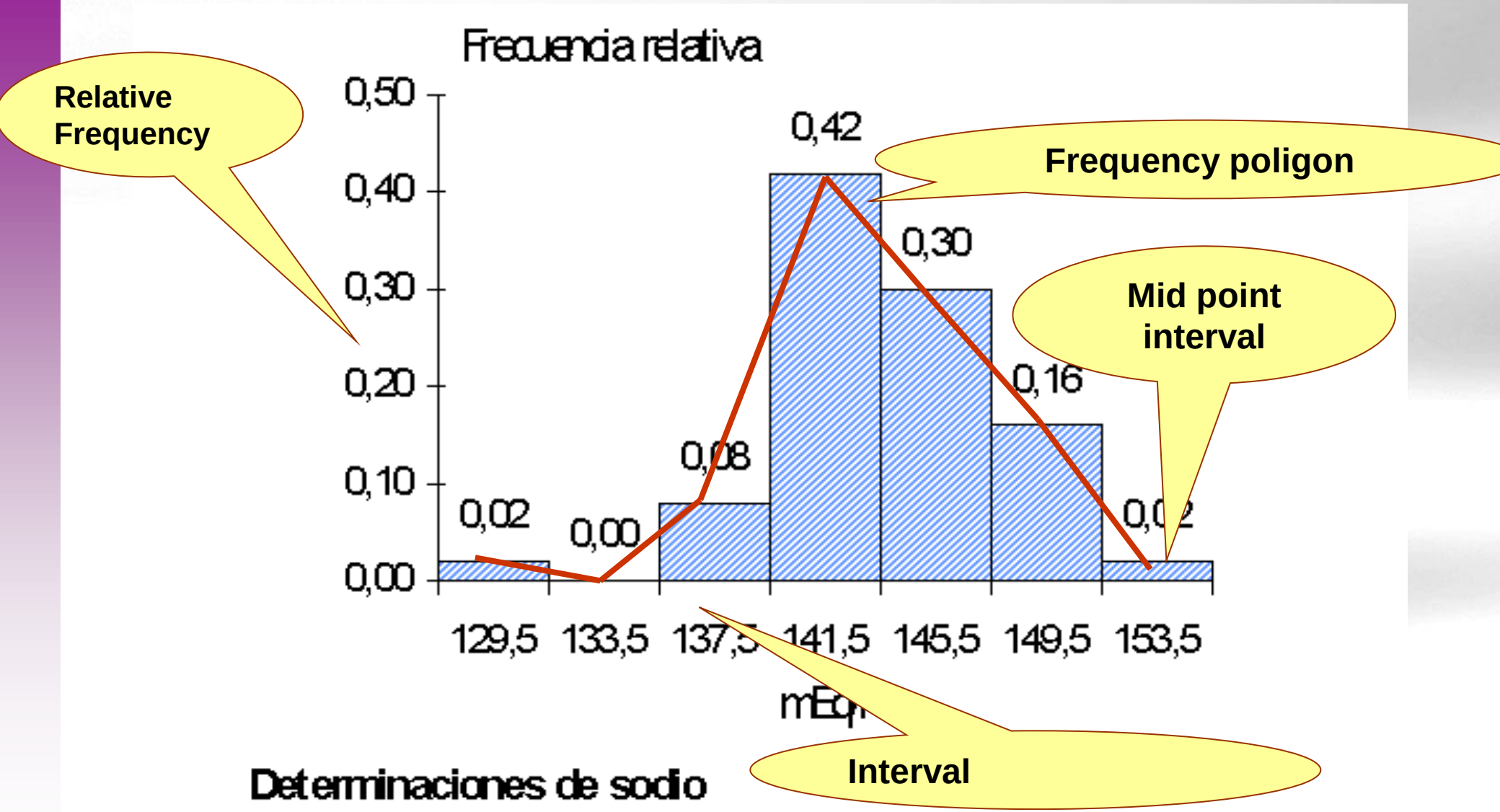
  geom_bar(mapping = aes(x = nivel_ed))
```

RESULT:



# Histogram



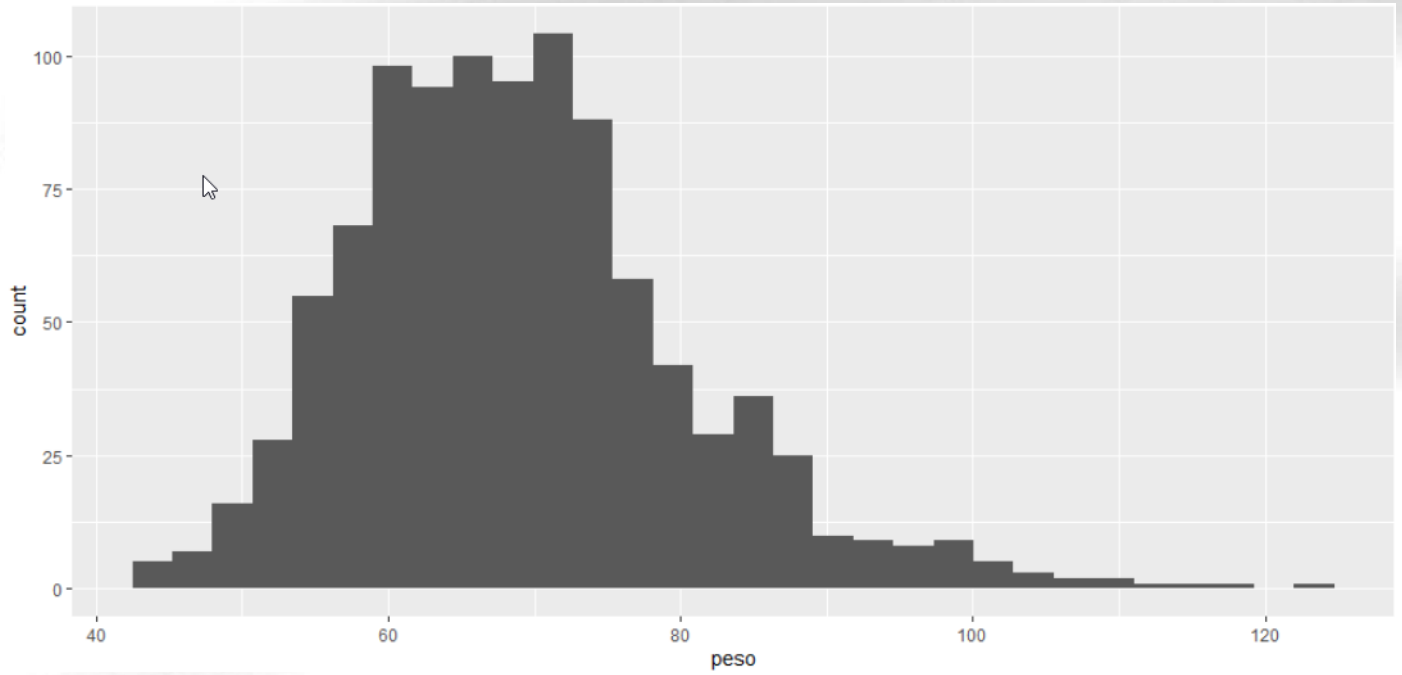

Relative Frequency
Frecuencia relativa
0,50
0,40
0,30
0,20
0,10
0,00
0,02
0,00
0,08
0,42
0,30
0,16
0,02
129,5
133,5
137,5
141,5
145,5
149,5
153,5
mEq
Frequency poligon
Mid point interval
Interval
Determinaciones de sodio

CODE:

```
ggplot(data = osteo) +

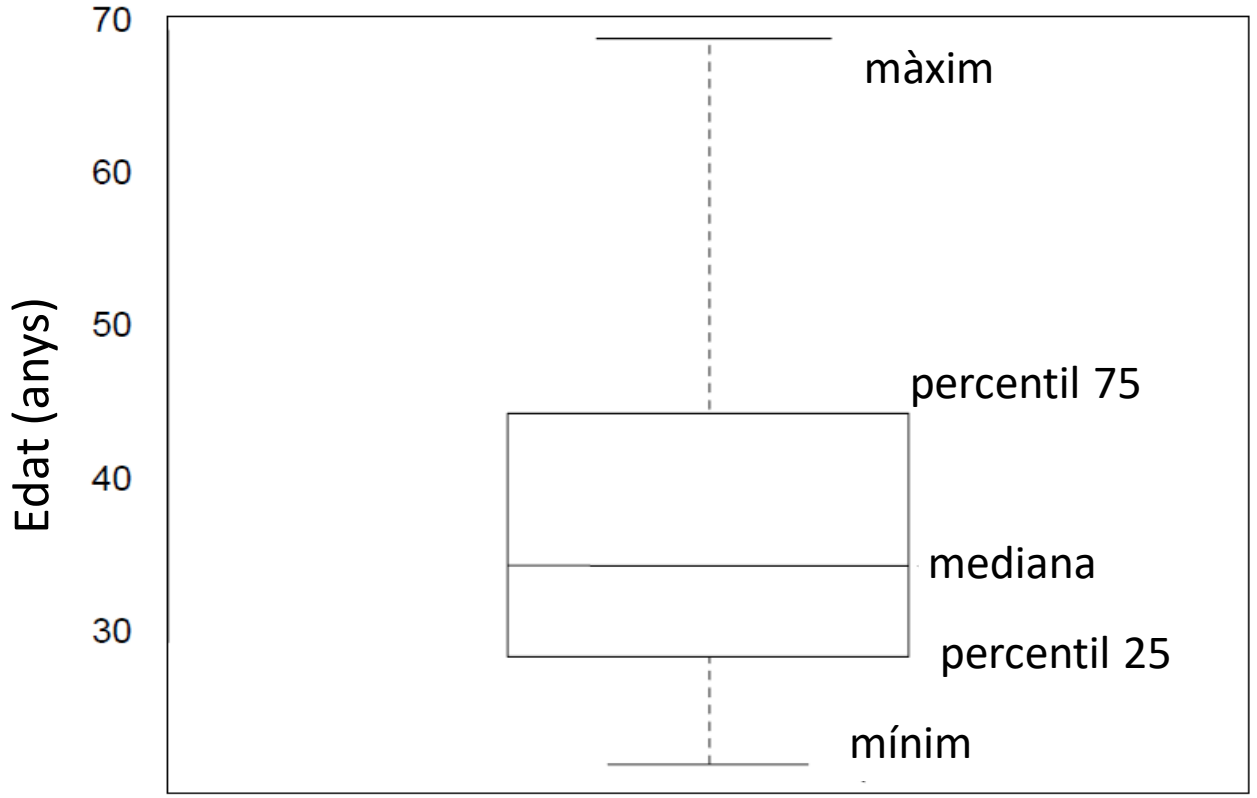
  geom_histogram(mapping = aes(x = peso))
```

RESULT:



# Boxplot

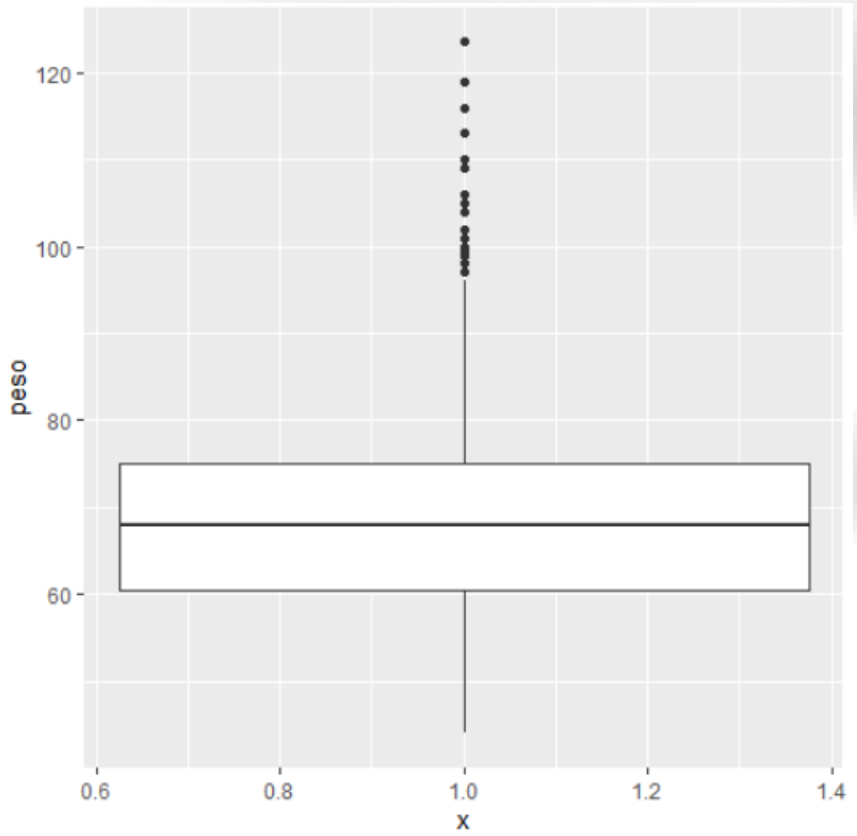
It is graphically represented the “five numbers": box are 25th and 75th percentiles, the middle line is the median (50th percentile) and the ends are the minimum and maximum values.



CODE:

```
ggplot(osteo,aes(y=peso,x=1))+
        geom_boxplot()
```

RESULT:



# Index

# Exercise



- Import diabetes.sav

BMI (Body mass index) of 149 patients →

- $MEAN_{BMI}$ =
- $SD_{BMI}$ =

SPB(Diastolic Pression Blood) of 149 patients →

- $MEAN_{SPB}$ =
- $SD_{SPB}$ =

Which variable have more variation? BMI or SPB

Calculate CV

- $CV_{BMI}$ = %
- $CV_{SPB}$ = %

- Create a boxplot with BMI

- Resume "MORT" variable. Create a frequency table and a barplot.