

# MS-A0503 First course in probability and statistics

## 3B Statistical datasets

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**Introduction**

Descriptive statistics

Empirical distribution

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Two-variable data (bivariate data, paired data)

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From data to population

# What is statistics (as a science)?

- Applying and developing methods for studying **random or uncertain** phenomena in the *real world*.
  - The methods are based on the mathematical laws of probability.
  - **Sources of uncertainty** are many: physical randomness; unknown properties of the real world phenomenon; random sampling on purpose; measurement errors; missing data . . .
  - . . . generally, the same math applies.
- Roughly:
  - Probability theory tells: How a certain process **produces** data.
  - Statistics tells: What **was** the process that produced the data.
- Statistics is applicable whenever you have data; especially if there is any kind of uncertainty or randomness.
  - Most fields of engineering and business have data, so they can (and do) use statistics.

# Two basic approaches of statistics

## Descriptive statistics

**Present** and **describe** the data “as it is”, either fully, or in a summary way.

- Tables (“the raw data”)
- Graphs (visualization)
- Numerical summaries or “statistics” (e.g. average, minimum, maximum)

## Statistical inference

**Infer** facts about the real phenomenon that lies “behind” the data.  
**Generalize**, e.g. sample  $\rightarrow$  population; or measurements  $\rightarrow$  universal physical law.

- Stochastic models
- Parameter estimation
- Hypothesis testing

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# Statistical data

Typically (not always), statistical data is in a table, the **data frame**, where

- rows correspond to **units**, e.g. people
- columns are the **variables** observed for each unit, e.g. height

**Caveat:** Different fields of science/engineering use different words. E.g. “units” may be “objects”, “items”, “data points” (geometrically thinking), “records” (in databases)

Depending on number of variables, we may call our data *univariate*, *bivariate*, *multivariate*.

## Levels of measurement = What kind of values

- **nominal scale**  $\approx$  **categorical**: just distinct classes  
gender: {male, female}  
piece of DNA: {Adenine, Cytosine, Guanine, Thymine}
- **ordinal scale**: classes have meaningful order  
cloth size: { XS < S < M < L < XL }  
Likert scale: { str. disagree < disagree < neutral < agree < str. agree }
- **numerical**: values have arithmetic meaning
  - **interval scale**: differences  $x - y$  are meaningful  
calendar dates, Celsius temperature
  - **ratio scale**: also quotients  $x/y$  are meaningful  
length, weight, distance, Kelvin temperature

### Notes:

- all can be *represented* as numbers, e.g. adenine=1, cytosine=2, guanine=3, but arithmetic might not make sense.
- nominal sometimes called “qualitative”, but other meanings
- this is *not* the discrete/continuous distinction. Numerical data can be well discrete; e.g. **counts** (frequencies)

## Data set (Data frame)

- **data set** = sequence of elements (units) of the same type, e.g. numbers, identifiers, or lists of values (one for each variable)
- Often arranged in a table; (R terminology) **"data frame"**
- Order of units often not meaningful, so we could treat it as a set (or *multiset*, if many identical observations possible)

E.g. course feedback: ((12345A, 5, 1, 5), (98759K, 1, 5, 2), (33312K, 4, 4, 3), (23453B, 4, 4, 3), (21453U, 3, 3, 3))

One string variable (student id), three numerical variables (general satisfaction, workload, usefulness)

Student ID	General	Workload	Usefulness
12345A	5	1	5
98759K	1	5	2
33312K	4	4	3
23453B	4	4	3
21453U	3	3	5

5 units, 4 variables



## Average and standard deviation

If we have univariate numerical data:  $\vec{x} = (x_1, \dots, x_n)$

Average (sample mean)  $m(\vec{x}) = \frac{1}{n} \sum_{i=1}^n x_i$

Variance  $\text{var}(\vec{x}) = \frac{1}{n} \sum_{i=1}^n (x_i - m(\vec{x}))^2$

Standard deviation  $\text{sd}(\vec{x}) = \sqrt{\text{var}(\vec{x})}$

Eg.  $\vec{y} = (0, 0, 1, 1, 2, 2)$

$$m(\vec{y}) = \frac{1}{6} (0 + 0 + 1 + 1 + 2 + 2) = 1$$

$$\text{var}(\vec{y}) = \frac{1}{6} \left( (0-1)^2 + (0-1)^2 + (1-1)^2 + (1-1)^2 + (2-1)^2 + (2-1)^2 \right) = \frac{2}{3}$$

$$\text{sd}(\vec{y}) = \sqrt{\frac{2}{3}} \approx 0.8165$$

**Caveat:** Sometimes  $n - 1$  used as divisor for variance and sd, for technical reasons; more about this later (in parameter estimation).

## Example

Calculate sample mean and standard deviation for the following data sets

$$\vec{x} = (1, 1, 1, 1, 1),$$

$$\vec{y} = (0, 0, 1, 1, 2, 2),$$

$$\vec{z} = (0, 2, 0, 2, 0, 2, 0, 2, 0, 2),$$

$$\vec{w} = (\underbrace{0, 0, 0, 0, \dots, 0, 0, 0, 0}_{666666 \text{ times}}, 1000000, \underbrace{0, 0, \dots, 0, 0}_{333333 \text{ times}}).$$

Dataset	Mean	SD
$\vec{x}$	1	0.0000
$\vec{y}$	1	0.8165
$\vec{z}$	1	1.0000
$\vec{w}$	1	999.9995

Average and standard deviation are *summaries*, they do not tell everything about the data. (Just like in probability distributions.)

# Computing the summary statistics *from data*

Notation	Name	R	Matlab	Excel
$m(\bar{x})$	Average	mean()	mean()	AVERAGE()
$sd(\bar{x})$	Standard deviation	$\sqrt{(1-1/n)*sd()^2}$	std(,1)	STDEV.P()
$sd_s(\bar{x})$	(Corrected) std.dev.	sd()	std()	STDEV.S()
$var(\bar{x})$	Variance	$(1-1/n)*var()$	var(,1)	VAR.P()
$var_s(\bar{x})$	(Corrected) variance	var()	var()	VAR.S()
$cov(\bar{x}, \bar{y})$	Covariance	$(1-1/n)*cov()$	cov(,1)	COVARIANCE.P()
$cov_s(\bar{x}, \bar{y})$	(Corrected) covariance	cov()	cov(1)	COVARIANCE.S()
$cor(\bar{x}, \bar{y})$	Correlation	cor()	corrcoef()	CORREL()
$q_{0.5}(\bar{x})$	Median	median()	median()	MEDIAN()
$q_{0.25}(\bar{x})$	Lower quartile	quantile(, .25)	quantile(, .25)	PERCENTILE.INC(, .25)
$q_{0.75}(\bar{x})$	Upper quartile	quantile(, .75)	quantile(, .75)	PERCENTILE.INC(, .75)

**Caveat.** Terminology and notation varies across sources. For technical reasons, many computer programs offer the so-called “unbiased” or “Bessel-corrected” *sample variance* and *sample standard deviation*, where the divisor is  $n - 1$  instead of  $n$ . (Don't worry now – we will talk about this in parameter estimation.)

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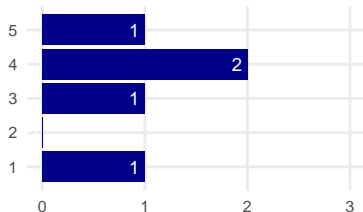
## Counts = Frequencies of values

The **count**, or (absolute) **frequency** of a value  $x$ , in the univariate dataset  $\vec{x}$ , is

$$n_{\vec{x}}(x) = \#\{i : x_i = x\}$$

Course feedback, pick one variable “General” → univariate data (5, 1, 4, 4, 3). Frequency as a table and a (horizontal) bar chart:

$x$	1	2	3	4	5
$n_{\vec{x}}(x)$	1	0	1	2	1



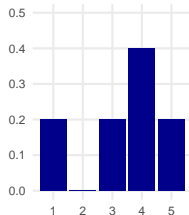
# Relative frequencies

The **proportion**, or **relative frequency** of value  $x$  in dataset  $\vec{x}$  is

$$f_{\vec{x}}(x) = \frac{n_{\vec{x}}(x)}{n} = \frac{\#\{j : x_j = x\}}{n}$$

Course feedback, pick “General”, dataset (5, 1, 4, 4, 3), relative frequencies as a table and (vertical) bar chart

$x$	1	2	3	4	5
$f_{\vec{x}}(x)$	$\frac{1}{5}$	0	$\frac{1}{5}$	$\frac{2}{5}$	$\frac{1}{5}$



Observation:  $\sum_x f_{\vec{x}}(x) = 1$ , thus  $f_{\vec{x}}(x)$  is a probability distribution!  
It is the **empirical distribution** of the dataset  $\vec{x}$ .

# Empirical distribution

## Proposition

If an element  $X$  is chosen uniformly at random, from the dataset  $\vec{x} = (x_1, \dots, x_n)$ , then  $X$  is a discrete random variable, whose density corresponds to the empirical distribution:  $f_X(x) = f_{\vec{x}}(x)$ . Furthermore,

$$\mathbb{E}(X) = m(\vec{x}), \quad (1)$$

$$\text{SD}(X) = \text{sd}(\vec{x}), \quad (2)$$

$$\text{Var}(X) = \text{var}(\vec{x}). \quad (3)$$

Also, for any function  $g$ , we have

$$\mathbb{E}[g(X)] = \frac{1}{n} \sum_{i=1}^n g(x_i). \quad (4)$$

## Example

For the dataset  $\vec{y} = (0, 0, 1, 1, 2, 2)$ , determine the empirical distribution, and its mean and standard deviation.

The relative frequencies are

$y$	0	1	2
$f_{\vec{y}}(y)$	$\frac{1}{3}$	$\frac{1}{3}$	$\frac{1}{3}$

If random variable  $Y$  has density  $f_{\vec{y}}(y)$ , then

$$\mathbb{E}(Y) = \sum_{y=0}^2 y f_{\vec{y}}(y) = 0 \times \frac{1}{3} + 1 \times \frac{1}{3} + 2 \times \frac{1}{3} = 1,$$

$$\text{Var}(Y) = \sum_{y=0}^2 (y-1)^2 f_{\vec{y}}(y) = (0-1)^2 \times \frac{1}{3} + (1-1)^2 \times \frac{1}{3} + (2-1)^2 \times \frac{1}{3} = \frac{2}{3}$$

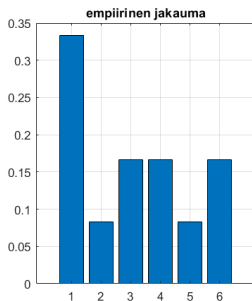
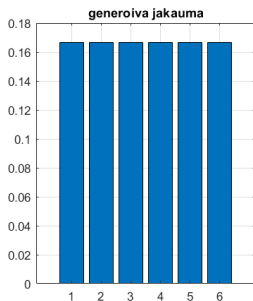
$$\implies m(\vec{y}) = \mathbb{E}(Y) = 1$$

$$\implies \text{sd}(\vec{y}) = \sqrt{\text{var}(\vec{y})} = \sqrt{\text{Var}(Y)} = \sqrt{\frac{2}{3}} \approx 0.8165$$



## Generating vs. empirical distribution

Rolling a fair die. Generating distribution (stochastic model) versus empirical distribution from 12 rolls that were (5,1,6,4,3,1,1,6,2,4,1,3).



- The empirical distribution *is* a probability distribution and you can use all tools of probability distributions.
- But *different distribution* from the generating one.
- For large  $n$  they are reasonably close (by LLN).

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## Binning (Grouping)

Eg. ages of all Finns 31.12.2015.

$n = 5\,487\,308$  units (data points)

Not probably good idea to draw as individual points (especially if ages are expressed in 1-day precision)

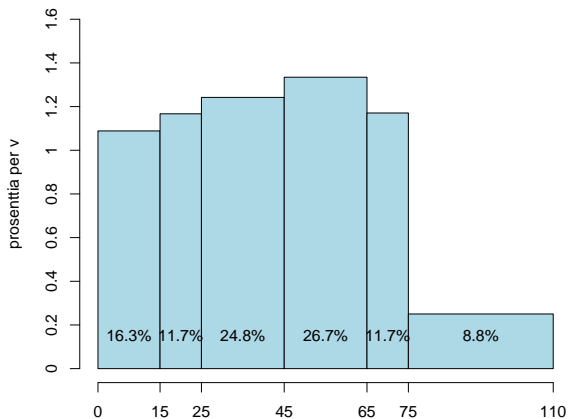
Let us **group** the data into **bins** at predefined boundaries.

Age (yr)	Frequency
0–14	896 023
15–24	640 387
25–44	1 363 155
45–64	1 464 640
65–74	642 428
75–	480 675

In fact we have transformed our real-valued variable (age) into a discrete one (group or bin index).

## Example: Histogram with unequal widths

Finnish age distribution 31.12.2015 [Source: Tilastokeskus]



Age (yr)	Frequency
0-14	896 023
15-24	640 387
25-44	1 363 155
45-64	1 464 640
65-74	642 428
75-	480 675

The bars are an *approximation* of the true density function.  
Could you find the proportion of Finns in the 1-year interval  $[13, 14)$ ?  
What about the the interval  $[109, 110)$  years? Would it be accurate?

## How to draw a histogram (allowing unequal bin widths)

- One bar for each bin (interval of possible values)
- Bar width = width of the interval
- Bar height = relative frequency *divided* by width

Example: Age distribution, first bar:

- Represents Finns with ages in interval  $[0, 15)$   
Note: values *strictly* smaller than 15; age in whole years 0–14
- Bar width = 15 years
- Frequency = 896023, relative frequency  
 $896023/5487308 \approx 16.3\%$
- Bar height =  $16.3/15 \approx 1.09$  (unit: % per year).
- Then bar *area* is the relative frequency.

(Typically, we use equal-width intervals, but not always.)

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## Bivariate data

Bivariate data = sequence (or multiset) of *pairs*

$$\vec{x}\vec{y} = ((x_1, y_1), \dots, (x_n, y_n)).$$

Alternatively, a pair  $(\vec{x}, \vec{y})$ , where  $\vec{x} = (x_1, \dots, x_n)$  and  $\vec{y} = (y_1, \dots, y_n)$  are univariate data (note: as ordered sequences, so we know which  $x_i$  and  $y_i$  belong together).

Course feedback: Two variables “General” and “Usefulness” composed as a bivariate dataset  $((5,5), (1,2), (4,3), (4,3), (3,3))$

Univariate statistics  $m(\vec{x}), m(\vec{y}), sd(\vec{x}), sd(\vec{y})$  are surely useful, but they tell nothing about the dependence between variables. Covariance and correlation tell (some aspects of) dependence.

$$\text{cov}(\vec{x}, \vec{y}) = \frac{1}{n} \sum_{i=1}^n (x_i - m(\vec{x}))(y_i - m(\vec{y}))$$

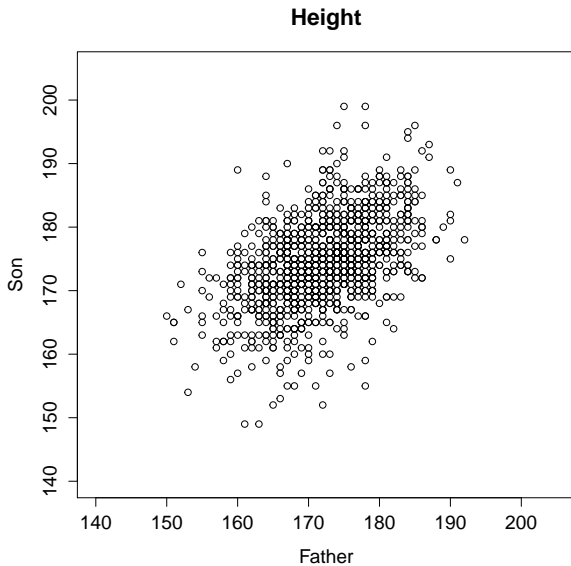
$$\text{cor}(\vec{x}, \vec{y}) = \frac{\text{cov}(\vec{x}, \vec{y})}{sd(\vec{x}) sd(\vec{y})}$$

# Example: Heights of father-son pairs

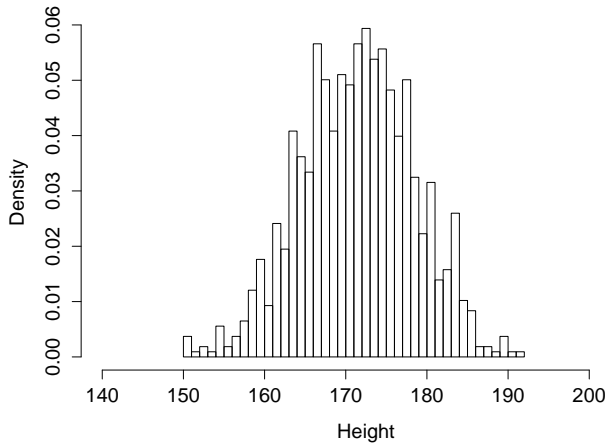
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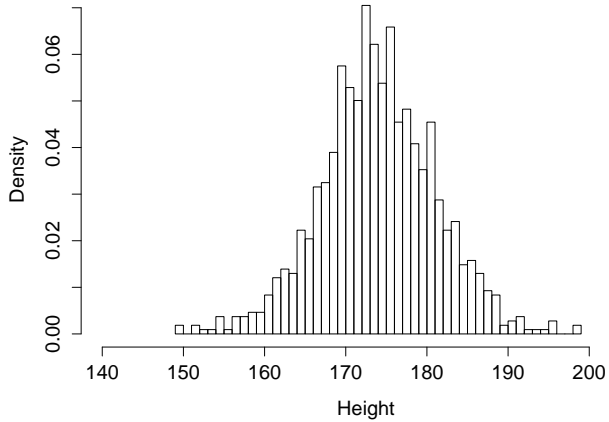
# Scatterplot (scatter diagram)



## Histogram of Fathers



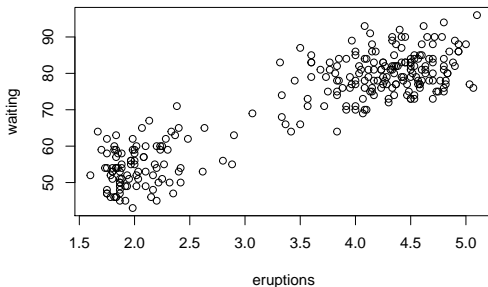
## Histogram of Sons



## Example: Eruptions of Old Faithful geysir

Scatterplot of 272 eruptions of *Old Faithful* (Yellowstone).

Two variables: eruption length and waiting time to next eruption.



Already a visual inspection (“eyeballing”) reveals interesting patterns (that are not captured by mean and standard deviation).

You can find the data in R, try `faithful` and `help("faithful")`

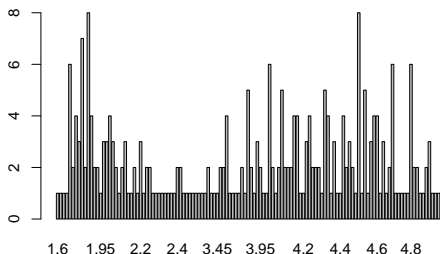
## Old Faithful: bar chart of one variable...

We could try listing *all different values* of eruption length, and collect their frequencies (within  $n = 272$ )

$x$	1.6	1.667	1.7	1.733	1.75	...	5.1
$n_{\bar{x}}(x)$	1	1	1	1	6	...	1

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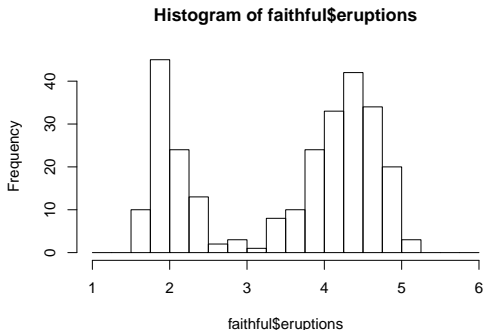
and draw a bar chart



Neither the table or the bar chart seems very informative.

## Old Faithful: histogram

But if we group the data into 0.25-minute intervals such as  $[2.00, 2.25)$ , and plot the counts, we have a better picture of the distribution.



Try this on your own! What happens if you use more (finer) intervals? What if you use less (coarser) intervals?

## Cross-tabulation (Contingency table)

The frequency of the pair  $(x, y)$ , in the data set, is

$$n_{\bar{x}y}(x, y) = \#\{i : x_i = x \text{ and } y_i = y\}$$

Course feedback: “General” and “Usefulness” as bivariate dataset ((5,5), (1,2), (4,3), (4,3), (3,3)) has this contingency table:

	y					
x	1	2	3	4	5	Sum
1	0	1	0	0	0	1
2	0	0	0	0	0	0
3	0	0	1	0	0	1
4	0	0	2	0	0	2
5	0	0	0	0	1	1
Sum	0	1	3	0	1	

## Cross-tabulation of relative frequencies

The **relative frequency** of the pair  $(x, y)$  is

$$f_{\vec{x}\vec{y}}(x, y) = \frac{\#\{i : x_i = x \text{ and } y_i = y\}}{n}$$

x	y					Sum
	1	2	3	4	5	
1	0	$\frac{1}{5}$	0	0	0	$\frac{1}{5}$
2	0	0	0	0	0	0
3	0	0	$\frac{1}{5}$	0	0	$\frac{1}{5}$
4	0	0	$\frac{2}{5}$	0	0	$\frac{2}{5}$
5	0	0	0	0	$\frac{1}{5}$	$\frac{1}{5}$
Sum	0	$\frac{1}{5}$	$\frac{3}{5}$	0	$\frac{1}{5}$	

$\sum_{x,y} f_{\vec{x}\vec{y}}(x, y) = 1$ , thus  $f_{\vec{x}\vec{y}}(x, y)$  is a probability distribution.  
 $f_{\vec{x}\vec{y}}(x, y)$  is the **empirical joint distribution** of the dataset  $\vec{x}\vec{y}$ .



## Empirical joint distribution

### Proposition

If a pair  $(X, Y)$  is chosen uniformly at random, from the dataset  $\vec{x}\vec{y} = ((x_1, y_1), \dots, (x_n, y_n))$ , it is a discrete random variable whose (joint) density is the empirical distribution  $f_{X,Y}(x, y) = f_{\vec{x}\vec{y}}(x, y)$  and also

$$\begin{aligned}\mathbb{E}(X) &= m(\vec{x}), & \mathbb{E}(Y) &= m(\vec{y}), \\ \text{SD}(X) &= \text{sd}(\vec{x}), & \text{SD}(Y) &= \text{sd}(\vec{y}), \\ \text{Var}(X) &= \text{var}(\vec{x}), & \text{Var}(Y) &= \text{var}(\vec{y}),\end{aligned}\tag{5}$$

and

$$\text{Cor}(X, Y) = \text{cor}(\vec{x}, \vec{y}),\tag{6}$$

$$\text{Cov}(X, Y) = \text{cov}(\vec{x}, \vec{y}).\tag{7}$$

Also, for any two-argument function  $g$ , we have

$$\mathbb{E}[g(X, Y)] = \frac{1}{n} \sum_{i=1}^n g(x_i, y_i).\tag{8}$$

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**Quantiles / Percentiles**

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## Quantiles/Percentiles of data

Suppose data can be ordered from smallest to largest.

(OK for numerical or ordinal data; not for nominal data.)

If  $0 < p < 1$ , then the  $p$ -quantile (or  $100p$ -percentile)  $Q(p)$  is roughly the point  $x$  such that *proportion*  $p$  of the data is smaller than  $x$ , and  $1 - p$  is greater.

- $Q(0.25)$  is **lower (first) quartile**; 25% of data is below
- $Q(0.5)$  is **median** or second quartile; 50% of data is below
- $Q(0.75)$  is **upper (third) quartile**; 75% of data is below

Note that half of data is between lower and upper quartiles.

R: `quantile(x,p)`, `summary(x)`, `median(x)`

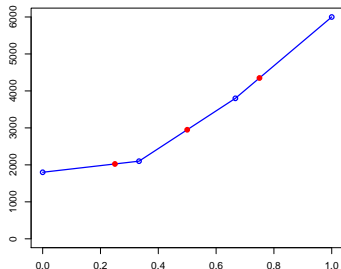
The “roughly” is because in finite data, you may not find exact quarters. There are some (varying) conventions for this.

## Quantile function

One way to define the **quantile function** of dataset  $(x_1, \dots, x_n)$ :

- Order the data as  $x_{(1)} \leq x_{(2)} \leq \dots \leq x_{(n)}$
- Divide the horizontal unit interval  $[0, 1]$  into equal parts, at points  $p_k = (k - 1)/(n - 1)$ ,  $k = 1, \dots, n$
- Plot the points  $(p_k, x_{(k)})$  and connect with lines

Example. Four **salaries** (eur/month): 3800, 1800, 2100, 6000



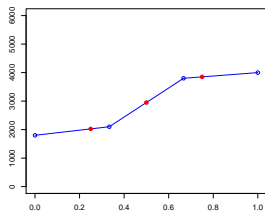
**Quartiles** = Evaluate the quantile function at 0.25, 0.50, 0.75

## Example: Three small datasets

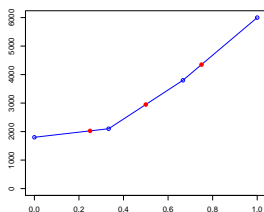
$$\vec{x} = (1800, 2100, 3800, 4000)$$

$$\vec{y} = (1800, 2100, 3800, 6000)$$

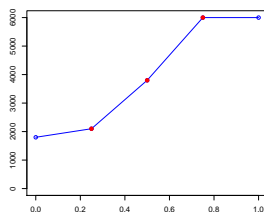
$$\vec{z} = (1800, 2100, 3800, 6000, 6000) \quad (n \text{ not divisible by four})$$



$$Q_x(0.5) = 2950,$$
$$m(x) = 2925$$



$$Q_y(0.5) = 2950,$$
$$m(x) = 3425$$



$$Q_z(0.5) = 3800,$$
$$m(x) = 3940$$

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**From data to population**

## Sample, population and “population”

The finite dataset you have, the “sample”, is often thought to “represent” qualities of a larger “population”.

sample	population
Pearson's 1000 fathers and sons	All father-son pairs in ... ?
1000 poll responses	Opinions of 5 million Finns now
272 eruptions of Old Faithful	All its eruptions (past? future? potential?)
Drug effect on 30 patients	Drug effect on future patients
100 rolls of a loaded die	Potential infinite sequence of rolls

**Population** is statistical jargon for

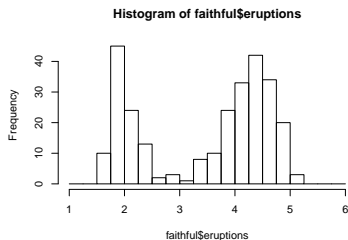
- where your data came from  
(**data-generating mechanism; data source**)
- what you are trying to understand by looking at the data

Hence, terms such as *sample mean* and *population mean*.

The “population” may be quite concrete, or a figure of speech.

# Old Faithful, once again

We have a *sample* of 272 eruption lengths. The real physical mechanism may be complicated, but perhaps we can think the lengths are **as if** they come from one particular **distribution**  $f$ . But what distribution?



One eruption length is a **random variable**  $X$  from some **generating distribution** or **underlying distribution** or **true distribution**. (“Population” if you want.)

Empirical distribution (shown above) approximates the generating distribution. Why? Think of the event  $\{2.0 \leq X < 2.25\}$ . We know by LLN that its *relative frequency* (long-run average)  $\approx$  its *probability* (in generating distribution).



Next lecture is about inference.

From the data, we will estimate some parameters concerning the reality “behind” the data.