

Why are data collected?



What kind of data are collected?



What kind of data are collected?

- Medical & pharmaceutical sciences...



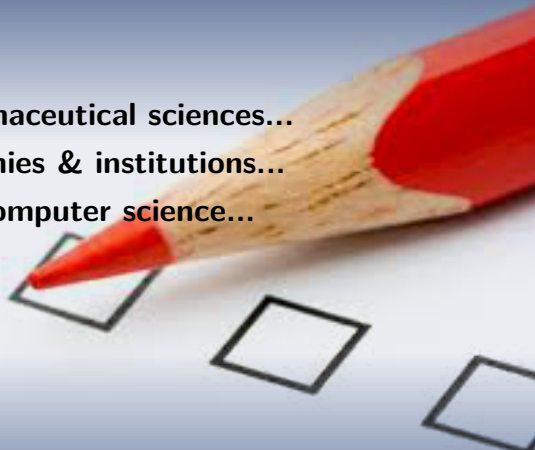
What kind of data are collected?

- ❑ Medical & pharmaceutical sciences...
- ❑ Financial companies & institutions...



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- ❑ Medical & pharmaceutical sciences...
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- ☐ Education & research...
- ☐ Sport, entertainment & fun...
- ☐ etc.

How to do it in a proper way?

Syllabus

❑ Introduction & Descriptive Statistics	<i>chapter 6</i>
❑ Introduction to Probability	<i>chapter 1</i>
❑ Random Variables	<i>chapter 2</i>
❑ Discrete and Continuous Probability Distributions	<i>chapters 3,4</i>
❑ The Normal Probability Distribution	<i>chapter 5</i>
❑ Sampling Distributions, Random Sample	<i>chapter 7</i>
❑ Inferences on a Population Mean	<i>chapter 8</i>
❑ Comparing Two Population Means	<i>chapter 9</i>
❑ Simple Linear Regression and Correlation	<i>chapter 12</i>
❑ Inferences on a Population Proportion	<i>chapter 10</i>
❑ The Analysis of Variance	<i>chapter 11</i>

How to do it in a proper way?

How to do it in a proper way?

Probability Statistics

- Random mechanism → produces random outcomes;

How to do it in a proper way?

Probability Statistics

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- A set of random outcomes (results) → drawing conclusions;

How to do it in a proper way?

Probability Statistics

- ❑ Random mechanism → produces random outcomes;
- ❑ A set of random outcomes (results) → drawing conclusions;
- ❑ **Probability** - theoretical background for a random mechanism;
(*describes how the mechanism works - but the principle is unknown*)

How to do it in a proper way?

Probability Statistics

- ❑ Random mechanism → produces random outcomes;
- ❑ A set of random outcomes (results) → drawing conclusions;
- ❑ **Probability** - theoretical background for a random mechanism;
(*describes how the mechanism works - but the principle is unknown*)
- ❑ **Statistics** - uses random outcomes to draw conclusions;
(*using statistics we can very precisely describe the mechanism behind*)

Statistics in Real Life...

- ❑ In probability, we could have a precise description of the random mechanism behind which would be accurate, but it is unknown!
- ❑ In statistics, we have data (measurements) that are quite often not precise, but the final conclusions drawn from such data are very accurate!

Statistics in Real Life...

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- ❑ In statistics, we have data (measurements) that are quite often not precise, but the final conclusions drawn from such data are very accurate!
- ❑ **"I only believe in statistics that I doctored myself."**
Winston S. Churchill (1874 – 1965)

Population vs. Sample



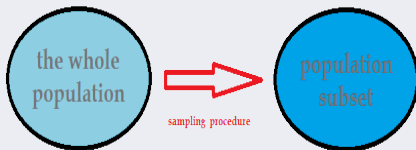
Population vs. Sample

Probability Theory
Random Distribution



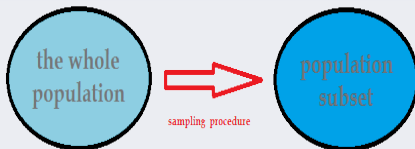
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Population vs. Sample

Probability Theory Finite Sample
Random Distribution Empirical Distribution



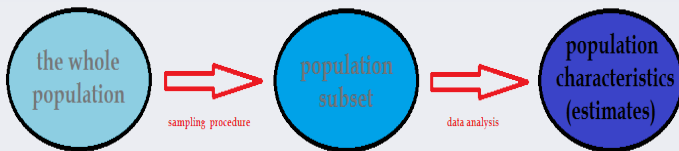
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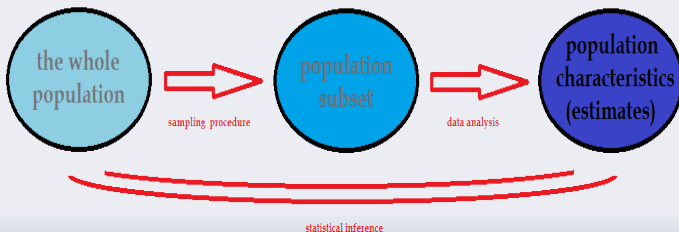
Population vs. Sample

Probability Theory Finite Sample Sample Statistics
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Population vs. Sample

Probability Theory Finite Sample Sample Statistics
Random Distribution Empirical Distribution Statistical Inference



Different Data types

■ Categorical variables

Different Data types

■ **Categorical variables**

■ **Numerical variables**

Different Data types

■ Categorical variables

■ Nominal categories

- categories with no ordering;

What kind of transportation do you use to get to work?

What kind of material do you mostly prefer?

■ Numerical variables

Different Data types

■ Categorical variables

■ Nominal categories

- categories with no ordering;
What kind of transportation do you use to get to work?
What kind of material do you mostly prefer?

■ Ordinal categories

- categories with possible ordering;
What was your most frequent grade last semester?
What energy category does the machine belong to?

■ Numerical variables

Different Data types

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■ **Integer values - counts**

- an apriori assumption of equidistant differences;
How many people visit this museum a day?
How many call a day is addressed to 911?

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How many people visit this museum a day?
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■ **Real values**

- any real value is possible...
length, weight, height, temperature, etc.

Data exploration

- ☐ What is the nature of data?
- ☐ What are the limitations of the experiment behind?
- ☐ What is the main question of interest?
- ☐ Is it possible to use data available to answer it?
- ☐ Is it important to visualize the data?
- ☐ How can one get an insight in the data?

Binary Data

- ❑ the simplest data type...
- ❑ different coding possible...
 - ❑ **logical**: TRUE | FALSE; 1|0; YES|NO;
 - ❑ **categorical**: A|B; 1|2; HOME|ABROAD;
- ❑ only a few options on how to represent such data;

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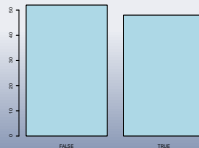
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Graphical view:



Categorical Data

- **Nominal categories...**

- **Machine Breakdowns:** what is a machine breakdown cause?

Categorical Data

■ Nominal categories...

- **Machine Breakdowns:** what is a machine breakdown cause?
elektrical | mechanical | misuse

Categorical Data

- **Nominal categories...**

- **Machine Breakdowns:** what is a machine breakdown cause?
elektrical | *mechanical* | *misuse*

- Data: a sequence of words '*elektrical*', '*mechanical*' or '*misuse*';

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- Data: a sequence of words 'electrical', 'mechanical' or 'misuse';

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Electrical	9
Mechanical	24
Misuse	13

Categorical Data

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Frequency summary:

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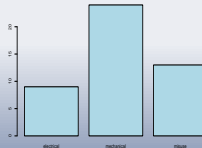
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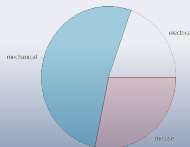
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■ Ordinal categories...

- **Number of Breakdowns:** How many times did a machine break down?

Categorical Data

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0 times / once / two times and more

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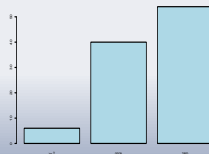
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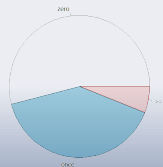
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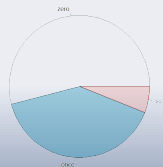
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- What is the main difference between nominal and ordinal categories?

Numerical Data

❑ Integer values - counts...

- ❑ How many hours can a machine work until it breaks down?

Numerical Data

■ Integer values - counts...

- How many hours can a machine work until it breaks down?
0, 1, 2, 3, ... etc.

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❑ Integer values - counts...

- ❑ How many hours can a machine work until it breaks down?
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0, 1, 2, 3, ... etc.
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1, 2, 3, etc. ≤ 5

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- How many people a day will use the lift?
0, 1, 2, 3, ... etc.
- How many passengers is in a driving car?
1, 2, 3, etc. ≤ 30

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- How many hours can a machine work until it breaks down?
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1, 2, 3, etc. ≤ 100

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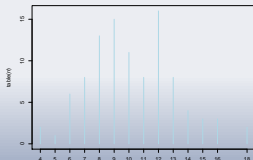
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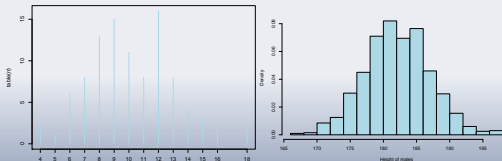


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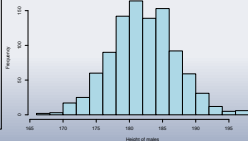
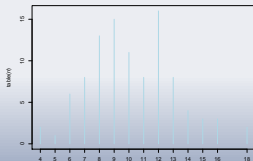


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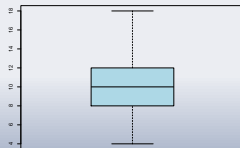
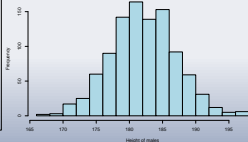
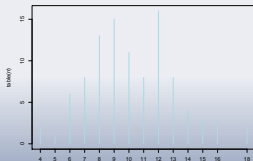


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■ Real values...

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data \rightarrow *positive real values...*

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❑ Data: a sequence of real valued observations...

- ❑ infinitely many different outcomes to be considered;
- ❑ the most common data presentation using histograms;

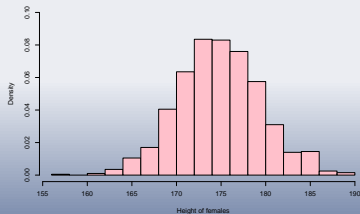
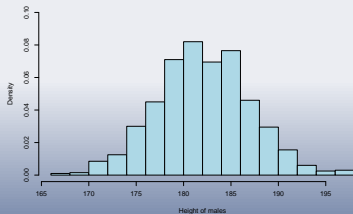
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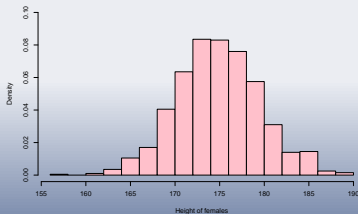
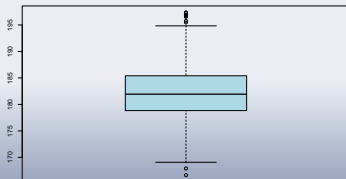
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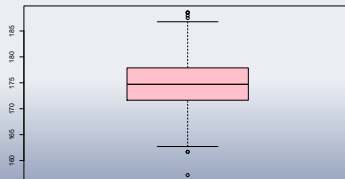
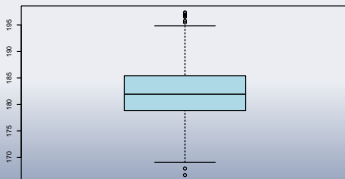
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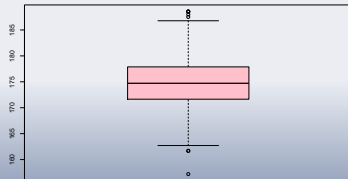
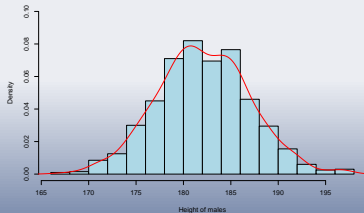
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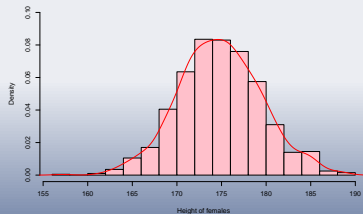
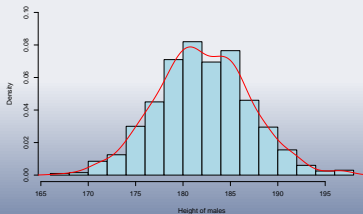
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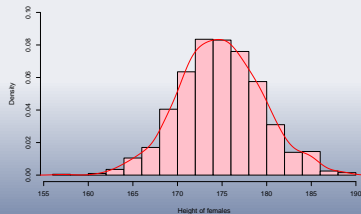
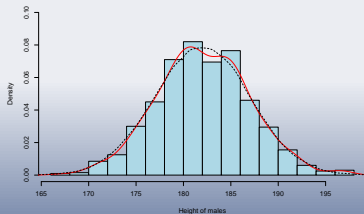
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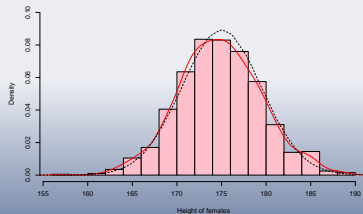
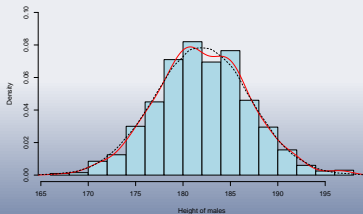
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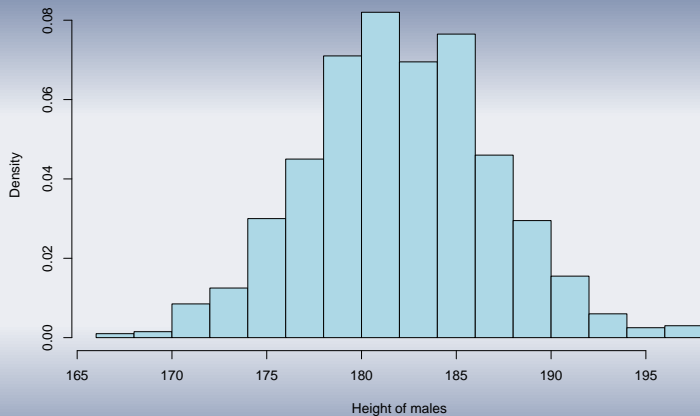
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Data: a sequence of real valued observations...

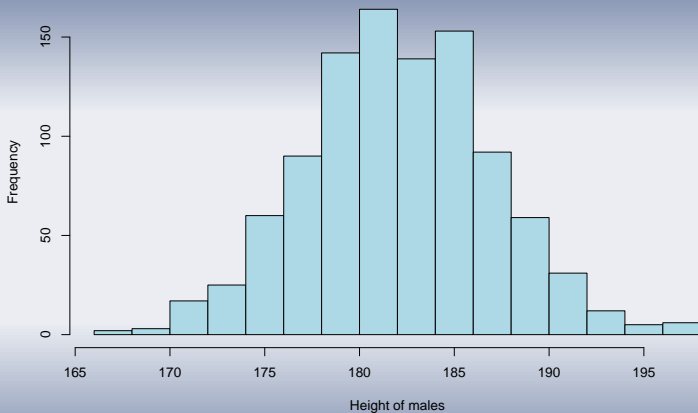
- infinitely many different outcomes to be considered;
- the most common data presentation using histograms;



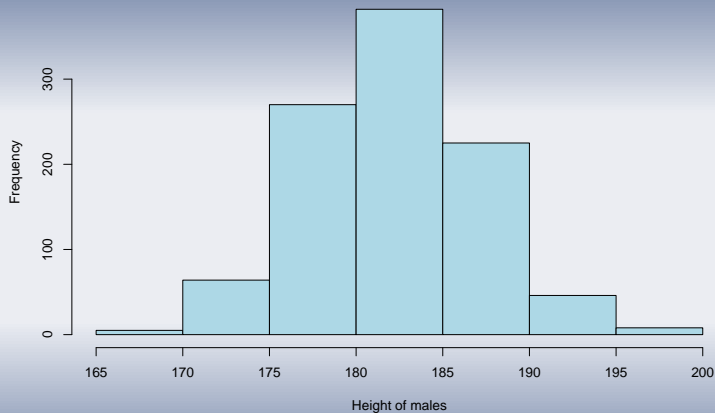
Histogram...



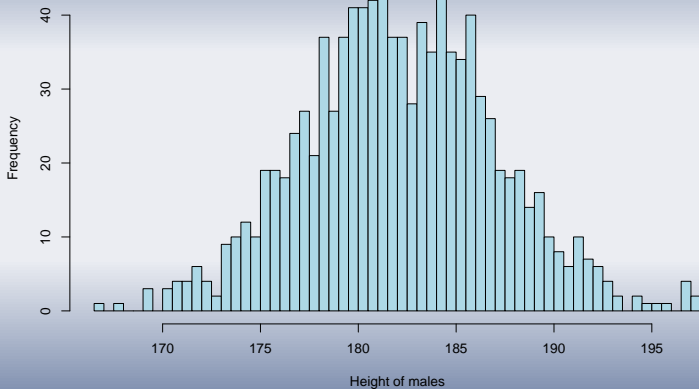
Histogram...



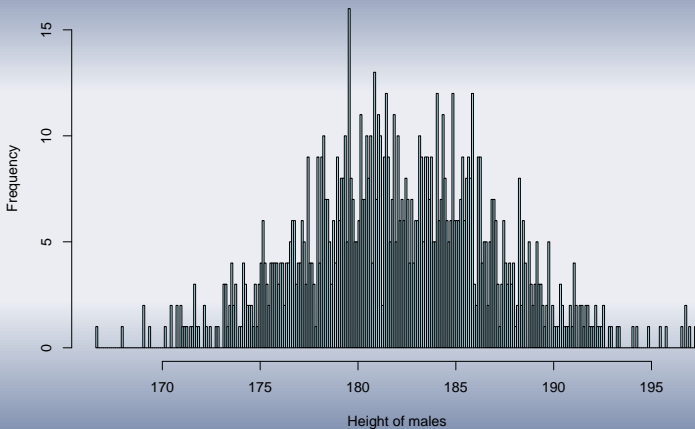
Histogram...



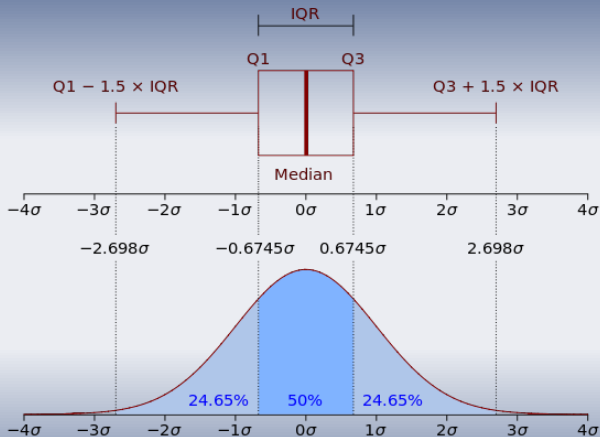
Histogram...



Histogram...



Boxplot figure...



Some nice properties

What can be directly observed from histogram (boxplot)?

- symmetric/non-symmetric data distribution;
- unimodal or multimodal data distribution;
- skewness of data distribution;
- indications of outlying observations;
- ...

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It is not so much important now when we only focus on data exploration part however, it will become really crucial when considering more sophisticated statistical modeling approaches.

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- ❑ How reliable such data samples are?
- ❑ Could we expect more reliability and confidence if more data is available?

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- ❑ The more data we obtain the more we have to pay for...
- ❑ How to find a reasonable balance?

From counts to ordinal categories

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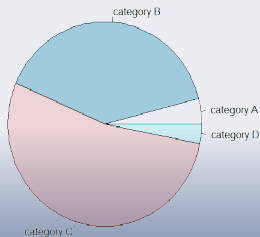
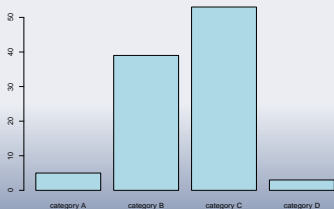
❑ What is the data sample now?

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What is the data sample now?



Higher dimensional data

- ❑ two categorical variables;
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- ❑ two numerical (continuous) variables;

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- ❑ two categorical variables;
- ❑ one categorical and the other numerical variable;
- ❑ two numerical (continuous) variables;
- ❑ more dimensional data \Rightarrow it gets even difficult to plot;
- ❑ multivariate statistical methods were proposed to be used instead;

Titanic survivals

	Sex	
Class	Male	Female
1st	0	0
2nd	0	0
3rd	35	17
Crew	0	0

,, Age = Child, Survived = No

	Sex	
Class	Male	Female
1st	5	1
2nd	11	13
3rd	13	14
Crew	0	0

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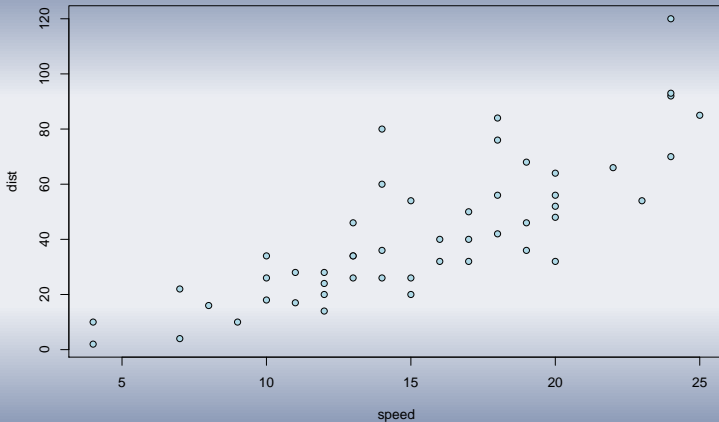
	Sex	
Class	Male	Female
1st	118	4
2nd	154	13
3rd	387	89
Crew	670	3

,, Age = Adult, Survived = No

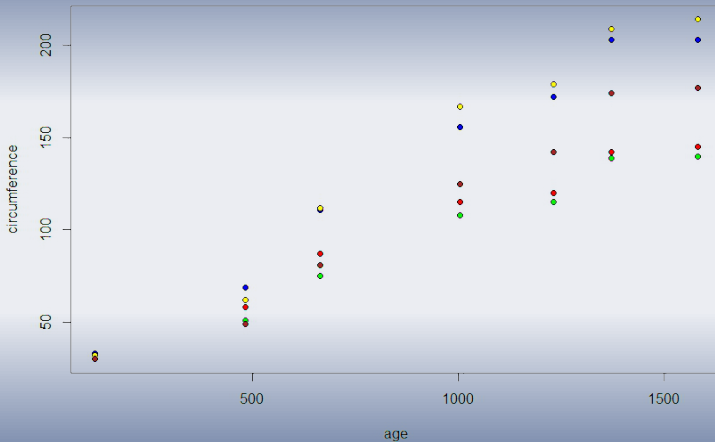
	Sex	
Class	Male	Female
1st	57	140
2nd	14	80
3rd	75	76
Crew	192	20

,, Age = Adult, Survived = Yes

Car Breaking Distance



Orange Trees



From Population to its Sample

population \Rightarrow population sample \Rightarrow statistical inference

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Sampling procedure:

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Sampling procedure:

- ❑ ideally we would like to obtain a **random sample**;
- ❑ **random sample** \rightarrow **independent and identically distributed observations**;

Sample Statistics

Two different worlds

...

Sample Mean

□ sample mean \equiv average

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- arithmetic average of all given observations;
- notation: $\bar{x}_n = \frac{\sum_{i=1}^n x_i}{n}$, where the actual observations are $\{x_1, x_2, \dots, x_n\}$;

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- ❑ **Advantage:** it is sensitive with respect to outlying observations;
- ❑ **Disadvantage:** it is sensitive with respect to outlying observations;
- ❑ Some other proposals: sample trimmed mean, weighted mean, etc.;

Sample Median

- middle observation from all ordered observations;
- $\tilde{x}_n = x_{(\frac{n+1}{2})}$ for n even and $\tilde{x}_n = [x_{(n/2)} + x_{(\frac{n+1}{2})}]/2$ for n odd;

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- ❑ Some other proposals as well ...;

Some other sample characteristics

- ❑ **sample mode (for categorical variables mostly);**
What is the relation between sample mean, median and mode?
- ❑ **sample variance & sample standard error;**
mostly it is difficult to imagine → sample range used “instead”
- ❑ **sample quantiles (quartiles, percentiles, etc.);**
some of them are more important than others...
- ❑ **coefficient of variation;**
spread of data relative to its middle value

To be continued...

- ☐ Probability, probability concepts;
- ☐ Random events, combination of events;
- ☐ Conditional probability;
- ☐ Independence principle;
- ☐ Law of Total Probability;
- ☐ Bayes Theorem in Probability;