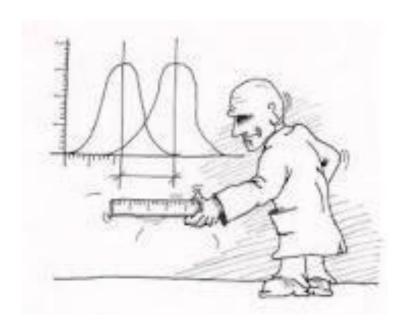
Biostatistics

Formalia

Lectures: Tuesdays, 10:10-11:45, HG E21, Beate Sick

Exercises: Tuesdays, 16:15-17:00, HG D7.1, Lisa Herzog

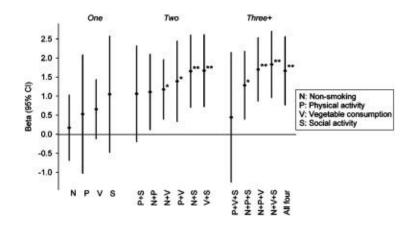
➤ Material: https://bsick.github.io/Biostatistics-Fall-2018/

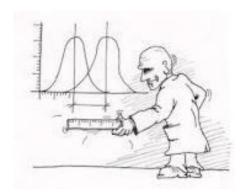


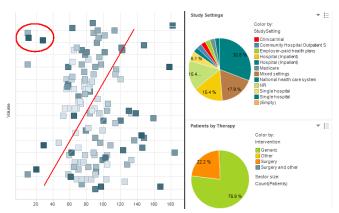
Goal of the module "Biostatistics"

Goal is to get more confident...

- in the most widely used statistical methods
- in reading data analysis sections in in scientific articles, especially in medical or biological journals
- Visualizing and analyzing own data







Biostatistics for Medical Physicists

Topics

Exam is on these topics, written, 45 minutes

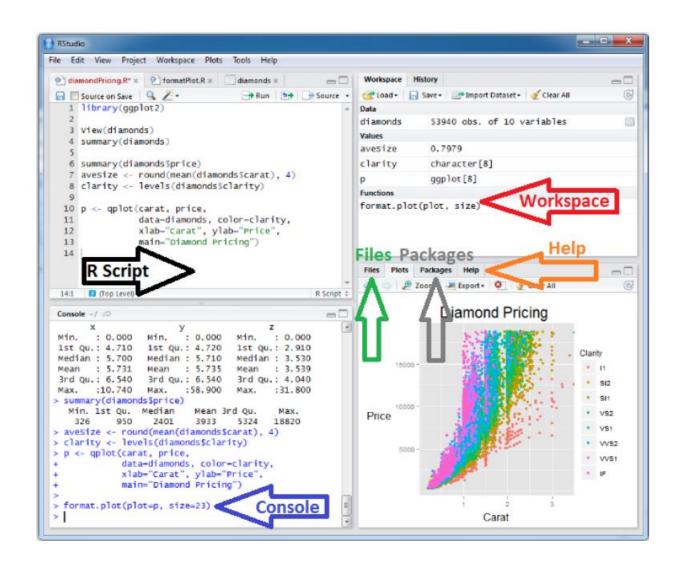
- data visualization
- basic terms and summary statistics
- study types, confounding
- diagnostic tests
- models/distribution-types
- parameter estimation
- testing, confidence intervals, p-values
- linear regression
- reliability analysis
- logistic regression
- outlook on more advanced or modern regression methods

We use R for performing statistical data analysis Recommended environment: RStudio

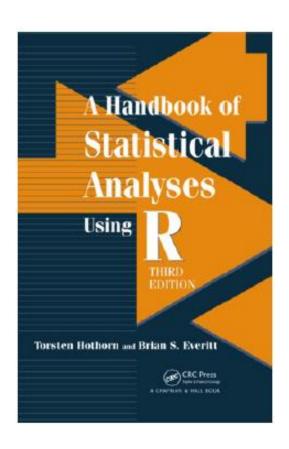


Main reasons:

- open source
- powerful
- wide spread
- reproducible
- transparent



Literature: no book needs to be purchased



Selected chapters from "A Handbook of Statistical Analysis Using R" by Torsten Hothorn, UZH, serve as recommended readings for this course and are provided on the course website. In addition some selected method articles will be recommended.

The R package HSAUR3 provides the selected chapters as PDF besides all data sets, examples and R code

http://CRAN.R-project.org/package=HSAUR3

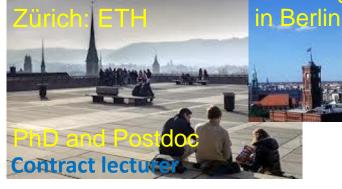
Torsten.Hothorn@R-project.org

My background: Important stations





Head of bioinformatics & biostatistics at DNA Array Facility UNIL





Prof. for applied statistics & Scientist

Providing trainings for life science researcher





Scientific collaborator, consultant, and lecturer in the field of biostatistics and medical research

Biostatistics for Medical Physicists Week 1

Topics this week:

- > Goals of this module
- Data types
- > methods for uni-variate data visualization
- > terms and key numbers

Research Article Example: Paper on hyperactivity form McCann et al.

Food additives and hyperactive behaviour in 3-year-old and 8/9-year-old children in the community: a randomised, double-blinded, placebo-controlled trial

Donna McCann, Angelina Barrett, Alison Cooper, Debbie Crumpler, Lindy Dalen, Kate Grimshaw, Elizabeth Kitchin Kris Lok, Lucy Porteous, Emily Prince, Edmund Sonuga-Barres, John O Warner, Jim Stevenson

Summary

Lancet 2007; 370: 1560-67

Published Online September 6, 2007 DOI:10.1016/S0140-6736(07)61306-3

See Comment page 1524

See Department of Error

page 1542

School of Psychology

Background We undertook a randomised, double-blinded, placebo-controlled, crossover trial to test whether intake of artificial food colour and additives (AFCA) affected childhood behaviour.

Methods 153 3-year-old and 144 8/9-year-old children were included in the study. The challenge drink contained sodium benzoate and one of two AFCA mixes (A or B) or a placebo mix. The main outcome measure was a global hyperactivity aggregate (GHA), based on aggregated z-scores of observed behaviours and ratings by teachers and parents, plus, for 8/9-year-old children, a computerised test of attention. This clinical trial is registered with Current Controlled Trials (registration number ISRCTN74481308). Analysis was per protocol.

What does it mean?

Typical "table 1" in a medical research article

Table. Patient Clinical Characteristics

	AII, n=68 (%)	No Recurrent Event Observed, n=54 (%)	Recurrent Event Observed, n=14 (%)	<i>P</i> Value
Demographic data				
Age, y (range)	65 (30–90)	64.7(30-88)	65.5 (47–90)	0.96
Men	47 (69)	38 (70.4)	9 (64.3)	0.75
Type of event				
TIA	5 (7.4)	4 (7.4)	1 (7.1)	0.6
Retinal ischemia	5 (7.4)	3 (5.6)	2 (14.3)	0.2
Stroke	58 (85.3)	47 (87)	11 (78.6)	0.2
Medical history, n (%)				
Smoking	29 (43)	20 (37)	9 (64.3)	0.21
Hypertension	49 (72)	37 (68.5)	12 (85.7)	0.32
Diabetes mellitus	13 (19)	7 (13.0)	6 (42.9)	0.02*
•••				

Clinical characteristics of all 68 patients (all) and patient groups without or with ipsilateral recurrent ischemic event. Number (n) and percentage or median and IQR are shown. CAD indicates coronary artery disease; IQR, interquartile range; mRS, modified Rankin Scale; NIHSS, National Institute of Health Stroke Scale; pAOD, peripheral artery occlusive disease; TIA, transient ischemic attack; and TOAST, Trial of ORG 10172 in Acute Stroke Treatment classification scheme for stroke etiology.

*P values <0.05 in Mann–Whitney U test or Fisher exact test.

Research Articel Example: Paper on hyperactivity form McCann et al.

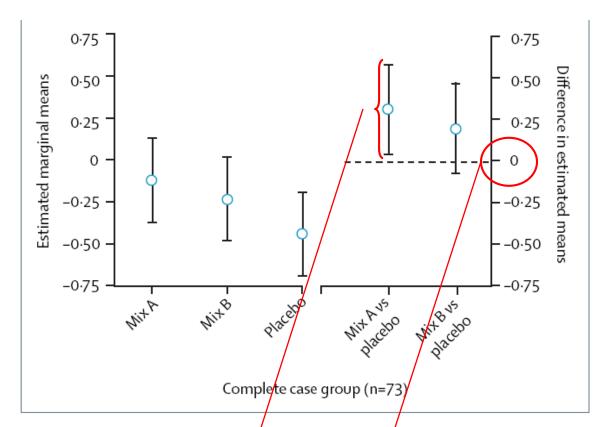


Figure 3: Estimated marginal means by challenge type and difference in estimated means in GHA under model 2 for 3-year-old children

What do bars indicate? ✓ What is special with the 0-line?

Research Article Example: Paper on hyperactivity form McCann et al.

		Entire sample (n=140)	Group with ≥85% consumption (n=130)	
	Model 1 Unadjus	ted		
	Intercept	-0·31 (-0·49 to -0·13)*	-0·33 (-0·53 to -0·13)†	
	Challenge type			□ What does
	Mix A vs placebo	0·20 (0·01 to 0·40)‡	0·24 (0·02 to 0·47)‡	
	Mix B vs placebo	0·16 (-0·04 to 0·35)	0·16 (-0·07 to 0·38)	"adjusted"
	Model 2 Adjusted	d		mean?
	Intercept	-0·54 (-0·89 to -0·18)*	-0·51 (-0·92 to -0·11)	
	Challenge type			How is it
	Mix A vs placebo	0·20 (0·01 to 0·39)‡	0·28 (0·05 to 0·51)‡	done?
	Mix B vs placebo	0·17 (-0·03 to 0·36)	0·19 (-0·04 to 0·41)	
			In model 2, in	
addi	tion to challenge	type, the effects of		What does *
		for: week during st	~	
	,	nber of additives i	*	mean?
		evel, and social clas	1	

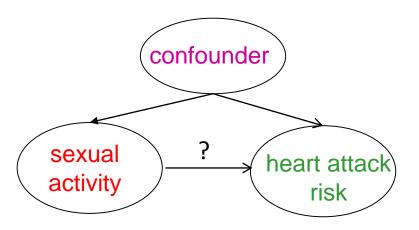
Example: Study in Caerphilly (Wales), 1979-2003

914 healthy men, between 45 and 95 years old, were chosen at random and followed over 10 years where they were interviewed e.g. about their sexual live. Moreover it was followed who suffered a heart attack in this period.

Result:

group	# men	# sexual active men	# sexual inactive men
all men	914	231	197
men suffering heart attack	11%105	8% 19	17% 33

What can we conclude?



12

Why do we need statistics?

Data vary!

Samples are random!

"We need statistics to draw intelligent decisions in the presence of uncertainty."

Numbers and Data

Numbers in mathematics	Number in data
exact	imprecise: random errors
certain	uncertain: "biased", faked, accidental coarse error
Just a number	Need for interpretation
Two number are either equal or different	Two observations are normally not exactly equal, but are they significantly different?

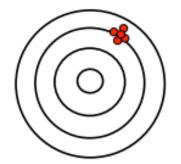
What kind of errors do we usually see?

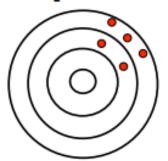
Variance: by random errors

Small variance Precise

Large variance Imprecise

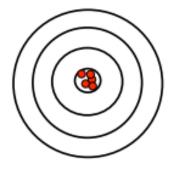


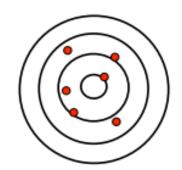




Bias







Systematic error or bias, accidental coarse error

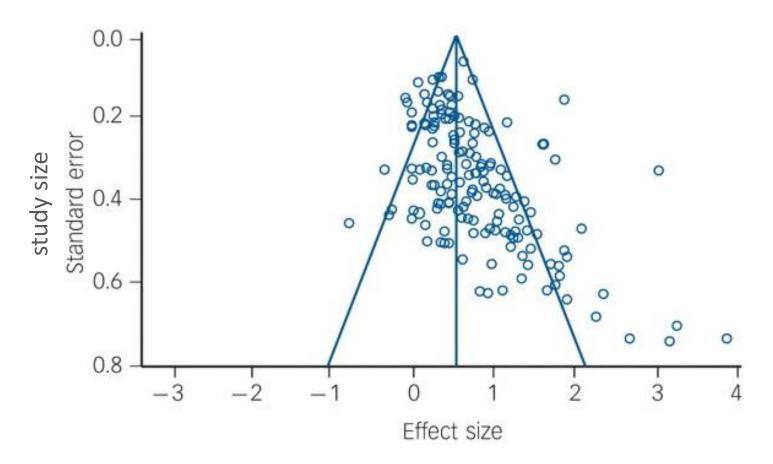


Wrong basic calibration



Accidently shifted comma suggested wrongly high iron content in spinach

What is a publication bias?



Publication bias is the tendency to more likely publish results that are positive (i.e. showing a significant finding) than negative results, leading to a misleading bias in the overall published literature. This is often visible in a funnel plot, where smaller studies with higher standard error tend to report more often positive results than large studies which are published irrespectively of the result.

Willful error or fake: the case of Henrik Schön

- 2001 each week a paper, many in Nature or Scienceach
 - Proposed as youngest Max Plank direktor in Stuttgart
 - Candidate for Nobelpreis

..

- 2002 Beasley Kommision...
- Withdrawl of PhD title because of dishonorable behaviour, but he did not fake during his PhD time in Konstanz



Der internationale Braunschweig-Preis geht an die Forschergruppe mit (von links) Jan Hendrik Schön, Christian Kloc (beide Bell Labs, Lucent Technologies, New Jerseey) und Professor Bertram Batlogg (ETH Zürich) als Gruppenleiter.

picture: ETH-Zürich

Fake: same noise implies same data -> data substitution

II. Data Substitution: Ambipolar triode characteristics

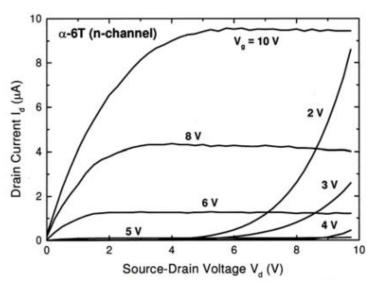


Figure 7. Triode characteristic from "LightEmitting" Paper (V), Fig. 1.: "alpha-sexithiophene (α -6T)"

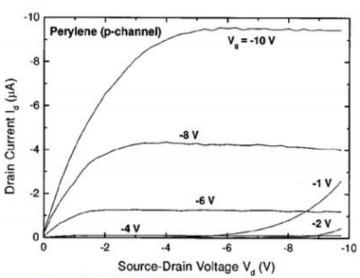


Figure 8. Triode characteristic from "Perylene" (VI), Fig. 2: "perylene". Note the sign change from Figure 7. One curve is missing.

Same data, different effects, different publications

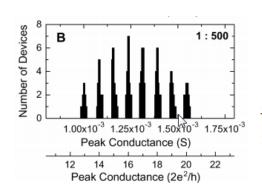
(V) "A Light-Emitting Field-Effect Transistor," J. H. Schön, A. Dodabalapur, Ch. Kloc, and B. Batlogg, Science 290, 963 (November 3, 2000)

(VI) "Perylene: A promising organic field-effect transistor material," J. H. Schön Ch.

Kloc, and B. Batlogg, Appl. Phys. Lett. 77, 3776 (December 4, 2000)

Pitfall when faking: precision

Presented data



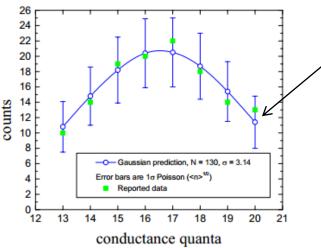


Figure 47. Comparison of reported number of devices in each conductance quantum bin with a fit to a Gaussian distribution. The reported data are extracted from original plotting data for Figure 46 from an electronic draft. The agreement exceeds the expected variation for such a small number of devices.

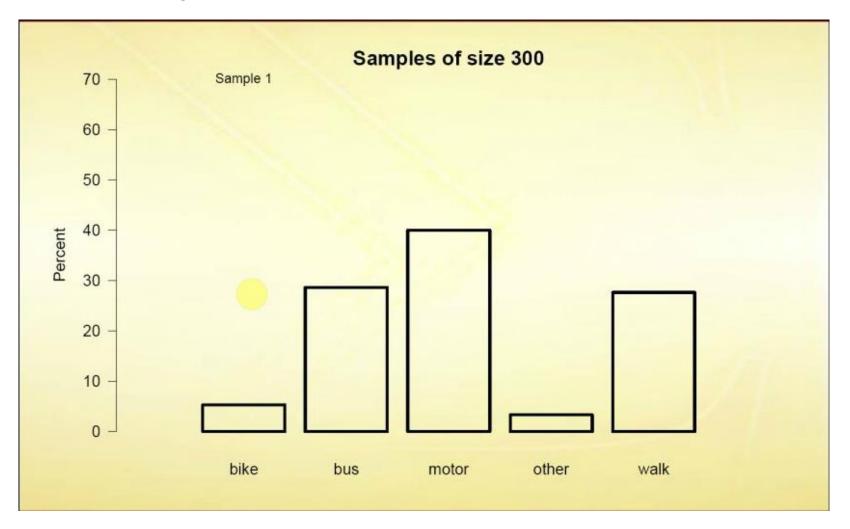
Blue: Theory

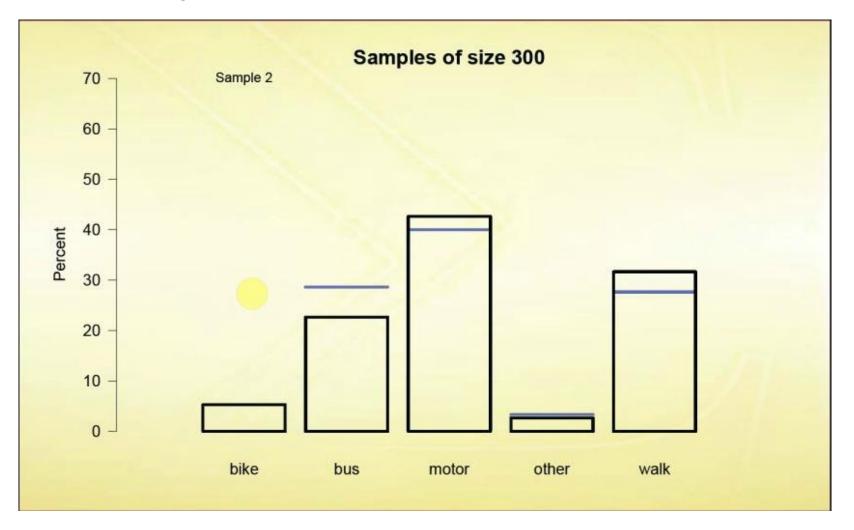
green: «observations»

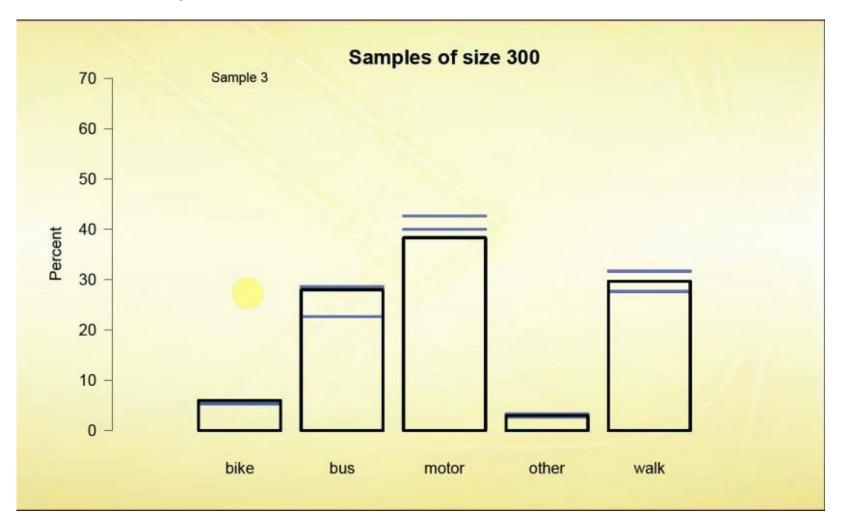
Blue standard error bar: 1/3 of all observations should outside of the bars.

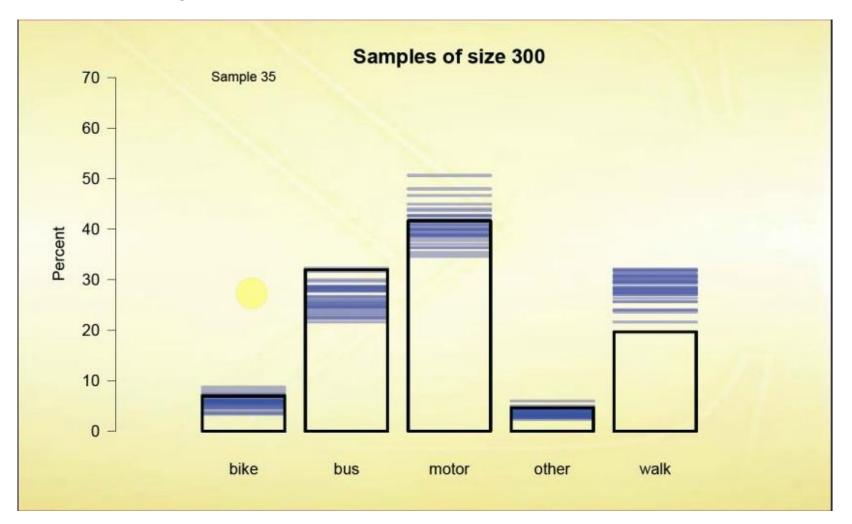
Probability to see such data by chance: 0.0012:

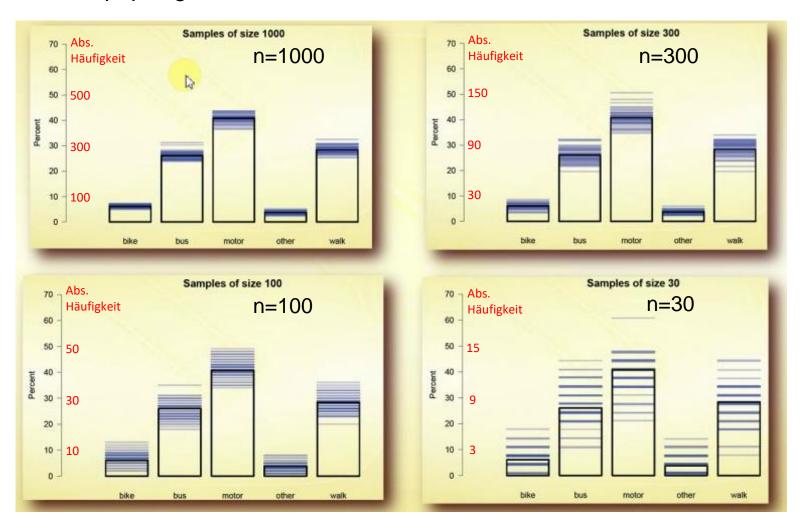
"Field-Effect Modulation of the Conductance of Single Molecules," Jan Hendrik Schön, Hong Meng, Zhenan Bao, Science 294, 2140 (December 7, 2001).

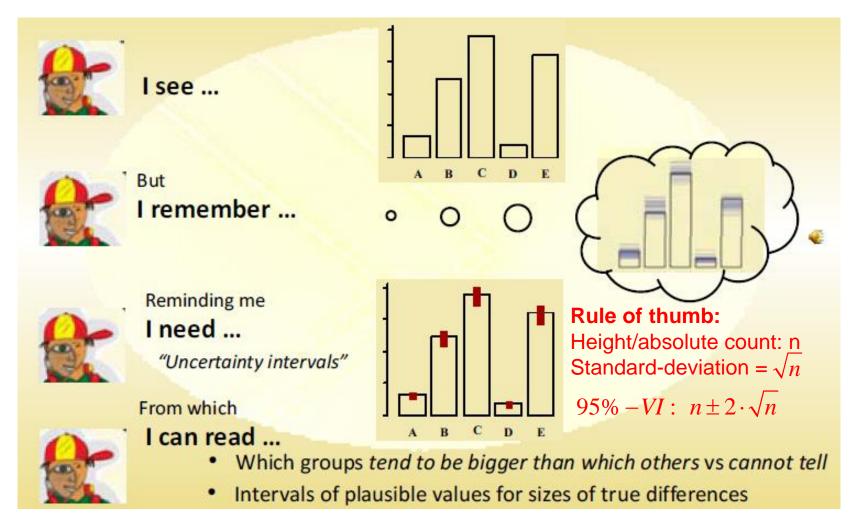












Fundamental terms

Feature or variable:

Properties which can take different values

- Age of a patient
- Sex of a student
- braking distance from 100 km/h for different car types

Population:

The complete set of all items that are relevant for the investigation

- All persons suffering a heart attack
- All at ETH inscribed students
- All uranium atoms

Sample:

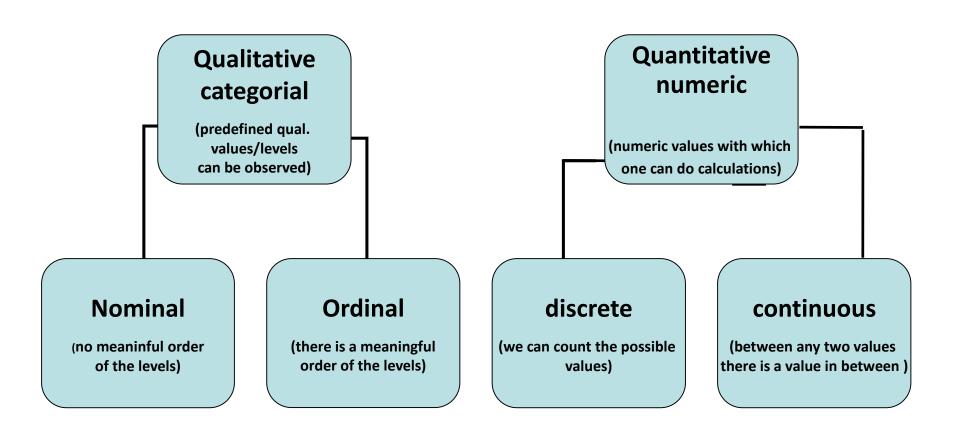
A subgroup of the entire population which have been selected in a certain manner (systematically, arbitrary, at random, stratified)

We can collect some data from each student

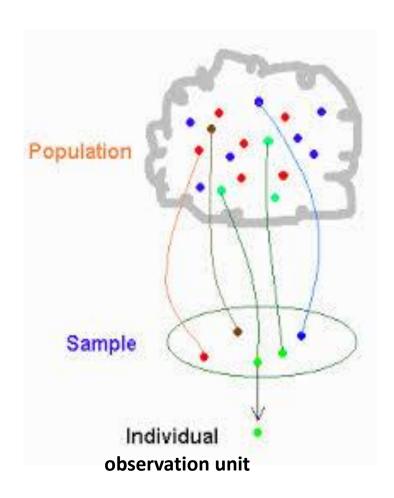
- 1) Age
- 2) Sex
- 3) Nationality
- 4) Height
- 5) Arm span
- 6) Number of siblings
- 7) Handedness
- 8) Rate the quality of the ETH Mensa food (0, 1, ..., 9)

How do the type of collected data differ?
How could we visualize the answers from the class for each question?

There are different types of data



We use a sample to learn about the population



Results from statistical inference are only correct, if the sample was representative.

A sample is representative if it does not systematically differ from the population (e.g. the percentages of male and female are similar in the sample than in the population).

In-class-exercise 1 Topic: When is a sample representative?

In the following examples we have a short description of research questions (RQ). Please answer for each example the 2 following questions (ignore the issue of sample sizes for the moment):

	a)	What is the popula	ation for which the	results should be g	eneralized?
	b)	Is the sample approp	oriate to answer the R0	Q for this population?	
1)		local anesthe	is the pain during to tics?" who decided to tak		
		anesthetic □ good	cs. ⊐ reasonable	□ bad	
2)		Sample: 100 Micro	e reading correct in pipettes picked at ra reasonable	•	Micropipettes?
4)		**	e average hemoglob who donate bloo		
		□ good	□ reasonable	□ bad	

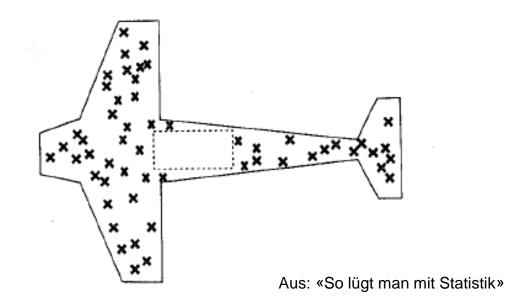
Census or Sample?



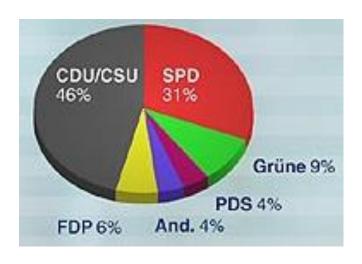
- In 2000 we had the last census (Vollerhebung) in Switzerland
- sampling is some times better, because of practical and data quality reasons

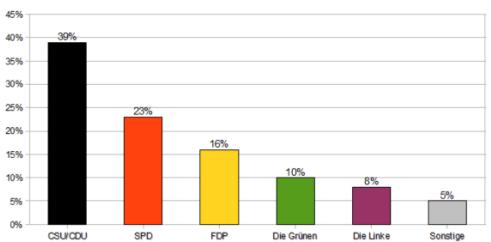
Representative Sample?

- Question: Which parts of bombers were most often hit? Which do we need to reinforce?
- Population: all bomber which came into operation
- Sample: 100 returned bombers . representative?

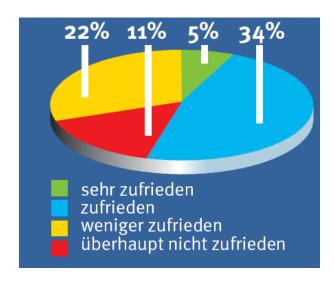


Visualizing categorical data by Bar-Chart or Pie-Chart These charts are simple - is there room form manipulation?



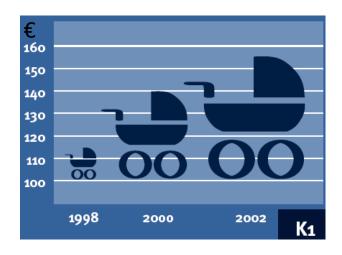


Half of all reader are satisfied with Klinsmann - true?



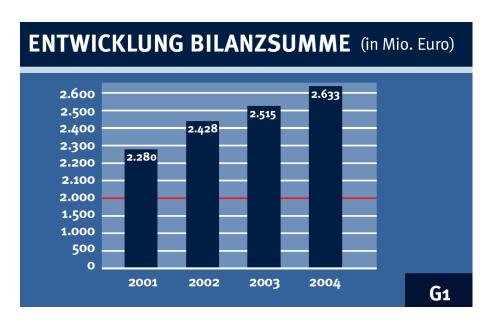
Pie-Chart from the German newspaper "Bild". Reader were asked how satisfied they are with soccer trainer Klinsmann.

Generous increase of child allowance - true?



Graph from government statement in the German red-green agenda 2010.

Good business development - true?



Bar Chart in the business report of a german bank (psd-Bank Rhein-Ruhr 2004)

How can we visualize continuous data?

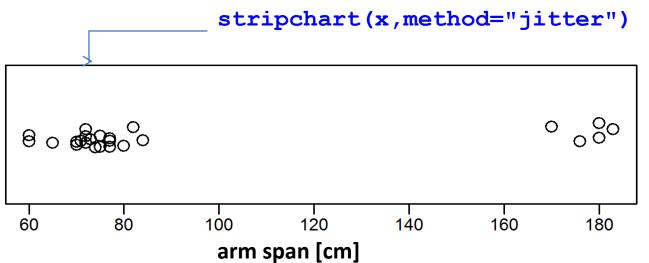
- 1) Age
- 2) Sex
- 3) Nationality
- 4) Height
- 5) Arm span
- 6) Number of siblings
- 7) Handedness
- 8) Rate the quality of the ETH Mensa food (0, 1, ..., 9)

How to summarize continuous data - e.g. arm span?

X: arm span	frequenc y
[60, 80)	17
[80,100)	2
[100,120)	0
[120,140)	0
[140,160)	0
[160,180)	4
[180,200)	1

- define non-overlapping classes/bins
- count number of observation per class
- draw histogram (no gaps between bars)





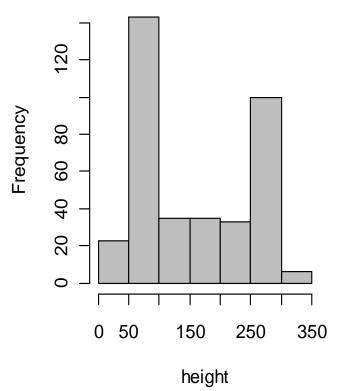
How to visualize continuous data?

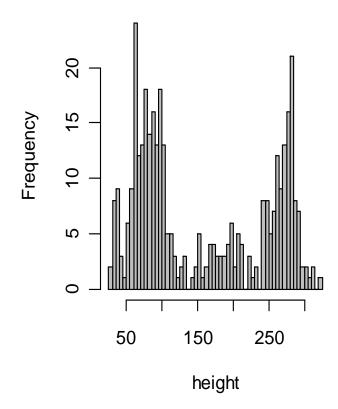
The height (cm) of 376 plants were measured.

head(dat\$height)

G	
height	
	57.9
	62.1
	55.8
	61.5
	68
	52.8
	70.5
	60.4
	75.2
	77.1
	70.4
	70.1
	27.6
	35
	•
	•

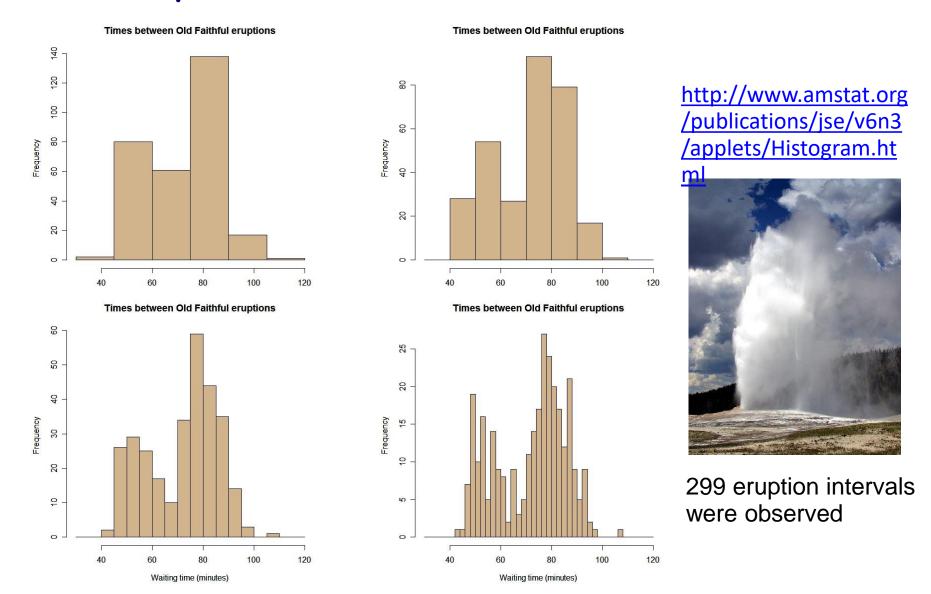
hist, few classes, big bin width
hist(dat\$height, nclass=7)
hist, many classes, small bin width
hist(dat\$height, nclass=100)





Are there subgroups? If yes – how many? How reliable is the height of a bar?

How many classes do we need?



Shape of the histogram may depend on the class choices

Rules for histograms

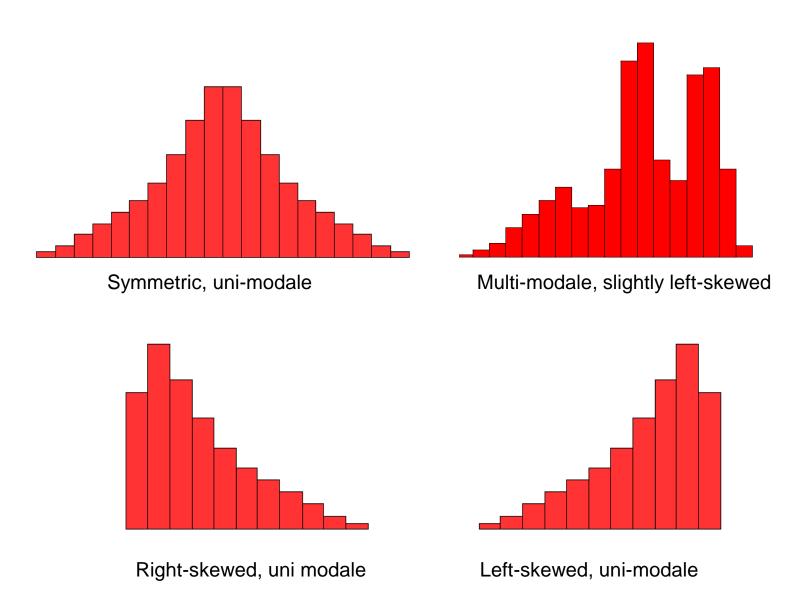


- Avoid classes with different width! (shape will change)
- How many classes: \sqrt{n} classes for n observations.

The shape can depend on the number classes and the class limits.

Attention: in a scaled histogram the *area* of the bar indicates the relative frequency, whereas in a unscaled histogram the *height* of the bar indicates the absolute frequency -> in case of unequal bin-widths the shape of the unscaled and scaled histograms can differ substantially.

Shapes of distributions



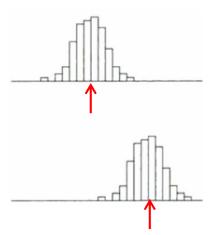
Measures for the location and variation

Data can be summarized by summary statistics. Most important key figure describe the center and the width of a distribution..

Measures for the location

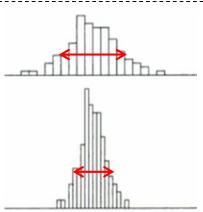
Where is the center?

What is a typical value?



Measures for the variation

A number which quantifies the width of the distribution.



Is the mean salary a «typical salary»?

The mean salary for Novartis employees was in 2009 around 220'000 CHF.

Willkommen bei Novartis Schweiz



Schweizer Arbeitsplätze



Grafik vergrössern

Novartis beschäftigt weltweit zurzeit rund

99.800 Mitarbeitende.

12.000 in der Schweiz

verteilt auf die acht

Standorte in Basel BS/BL, Stein AG, Embrach

ZH, Cham ZG, Bern BE, St-Aubin FR, Nyon

VD und Locarno TI. Eine kürzlich

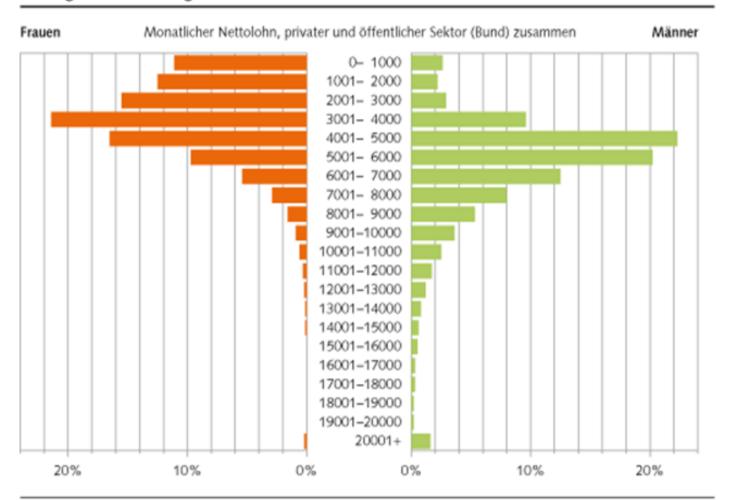
veröffentlichte Studie hat ergeben, dass für jeden direkten Arbeitsplatz bei Novartis in der Schweiz indirekt 2,5 weitere Arbeitsplätze geschaffen werden.

Die Gesamtsumme der Lohn- und Sozialleistungen für Mitarbeitende von Novartis in der Schweiz betrug im Jahr 2009 rund 2,6 Milliarden Franken.

mean.salary $\approx \frac{2.6Mrd.}{12000}$ $\approx 220000 CHF$

Distribution of salaries in Switzerland

Häufigkeitsverteilung der Arbeitnehmenden nach Lohnhöhenklassen 2008



Quelle: Schweizerische Lohnstrukturerhebung



