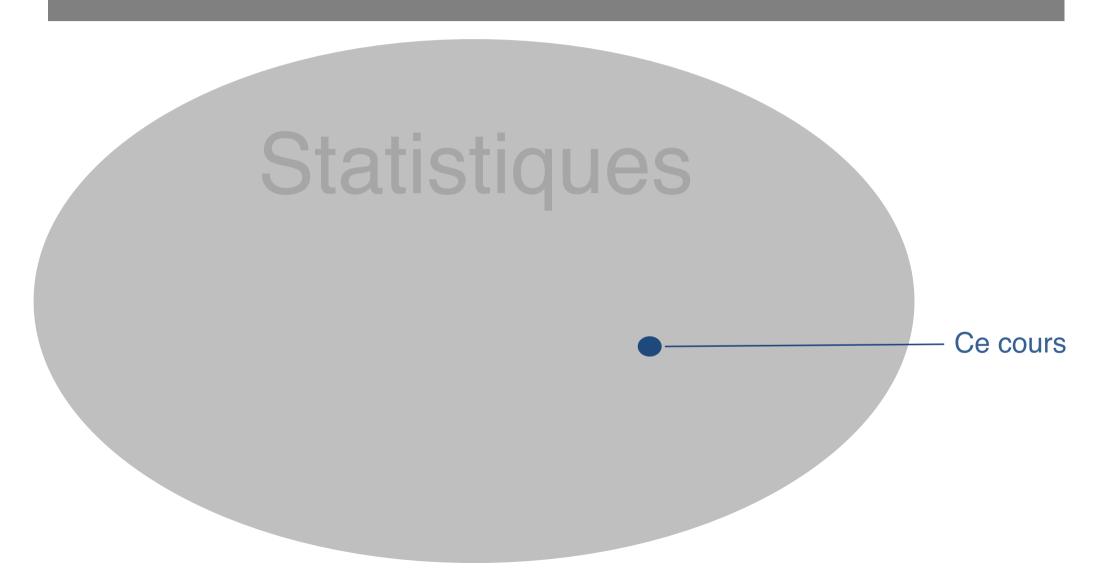
INTRODUCTION AUX NOUVELLES STATISTIQUES POUR L'IHM

Pierre Dragicevic



CONTENU DE CE COURS



OBJECTIFS

- Acquérir les intuitions et la terminologie de base sur les stats
- Première exposition à R
- Accent sur les aspects haut-niveau
- Sensibilisation aux abus des statistiques
- Accent sur les "nouvelles statistiques"

ORGANISATION

- Partie I notions élémentaires de stats
- Partie II analyses préliminaires en R
- Partie III bien utiliser les stats en IHM

A DEFINITION

• Statistics is the study of the collection, analysis, interpretation, presentation and organization of data.

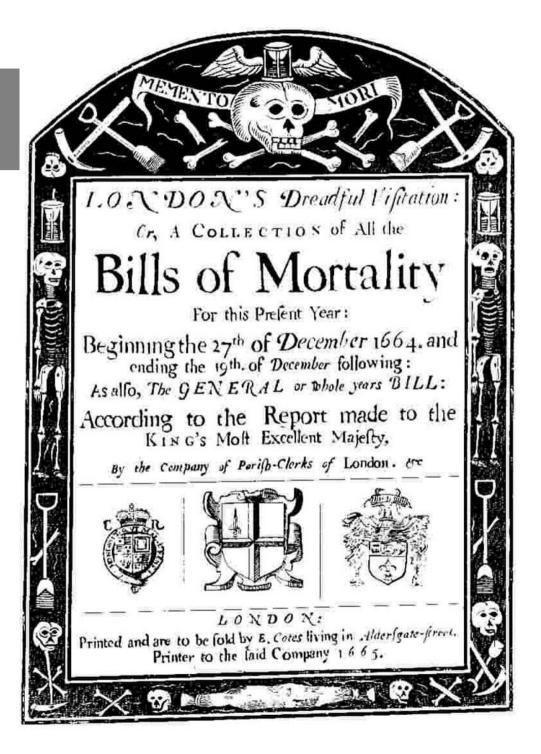
Dodge, Y. (2006) The Oxford Dictionary of Statistical Terms, OUP.

- 1750s German "Statistik" "analysis of data about the state"
- Quickly adopted in England (previously called "political arithmetics")

• John Graunt, 1662

Observations on the bills of mortality





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The Years of our Lord	1647	1648	1649	1650	1651	1652	165	31654	165	1656	1657.1	658	1650	1660	1629	1630	1631	1632	163	1634	1635	1636	1621	1625	1640	1052	1050	1649	Years.
Abortive, and flilborn Aged	335	329	327	351	380	181	384	433	482	410	463	467	421	544	400	439	410	445	500	475	507	522	1702	2005				1247	8559
Ague, and Fever: Apoplex, and fodainly	1260	884	751	970	1039	1212	1 2 82	1371	689	875	999	1800	2303	2148	950	1091	1115	1100	953	1279	1622	2360	4418	6235	3330	4903	4363	2377 4010 177	15757 23784
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Bleeding Bloudy Flux, Scouring, and Flux Burnt, and Scalded	155	176	802	289	833		200	386	168	368	362	233	346	251	449	438	352	348	278	512	346	330	1587	1466	1422			1597	78:8
Calenture Cancer, Gangrene, and Fiftula	26	29	31	10		53	36	37	73	31	3 24	35	63	52	20	14		28	27	30	1	30		4	2 105	31 4	150	114	13
Wolf Canker, Sore-mouth, and Thruth	66			42	68	51	5'3	72	44	81	19			68	6	4	4	1	132	143	5	74	15	70	190	244	161	133	689
Childbed Chrisomes, and Infants Colick, and Wind	1369	1254	1065	990	1237	1280	1050	1343	1 080	1393	1161	179	858	167	48	57	2035	2258	2130	2315	37	18955	105	8453	4678	4910	839 4788 497	490 4519 147	3364 32106 1389
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Goot	12	5	11	9	7	7	5	- /		7 13	10	13		4	18	1			14	17	7 5	8	71	50	35	25	36 45	28	134
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ivergrown, Spleen, and Rickets	53	46	56	.59	65	72	47	65	52	50	38	51	14	15	94			87	82	77	98	99	392 28	356	213	269	191	158	1421

Cancer, Gangrene, and Fiftula	26	29	31	19	31	53	36	37	73	31
Wolf	66	28		42	68		-			Sz
Canker, Sore-mouth, and Thrush	161	ATTENDED A	114	42				72	44	
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Chrisomes, and Infants	1369		1065	990					200	1393 1
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Cold, and Cough			00	00		10.5	41	36	21	
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Cramp	139		1				146	1		
Cut of the Stone	The same	2	1	3	10,0	- 1	1	2	4	1
Droply, and Tympiny	185	434	421	508	444	550	617	704	660	
Drosyned	47	40	130	27	49	50	13	30	43	- 49
Excellive drinking		7	2		920	3				
Executed	8	17	29	43	24	12	. 19	21	19	22
Fainted in a Path					ī	-	11/9	1000	10000	
Falling-Sickness	3	2	2	3		3	4	1	4	3
Flox, and fmall Pox	139	400	1100	184	525	1279	139	812	1294	823
Found dead in the Streets	6	6	9	8	7	9	14	4		4
French-Pox	18	29	15	18	21	20	10	20	20	23
Frighted	4	4	1		,		2		I	1
Goot	0	4	12	0	7	7	5	6	8	7
Grief	13	11	16	17	10	14	11	12	10	13
Jones I and and a second	-	4	1	1	.1			.,		16

- John Graunt, 1662
 Observations on the bills of mortality
 - First "life tables"
 - Dispelled several myths about the plague
 - First analysis of sex ratio
 - First realistic estimate of the population in London

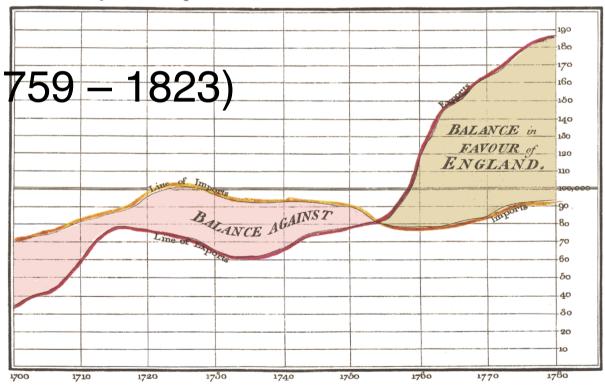
- Prompted collection of more data
- Parallel developments in probability theory
- Statistics then developed into a more rigorous discipline and was applied to:
 - Business & industry
 - Medicine
 - Science

— ...

STATS & VISUALIZATION

- Statistical Charts
 - William Playfair (1759 1823)

Exports and Imports to and from DENMARK & NORWAY from 1700 to 1780.



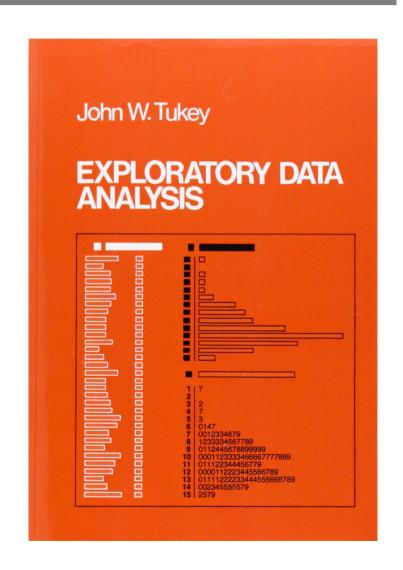
The Bottom line is divided into Years, the Right hand line into L10,000 each.

Published as the Act direct, 14 May 1786 by W. Playfair

Neel earlier 302 Strand, London.

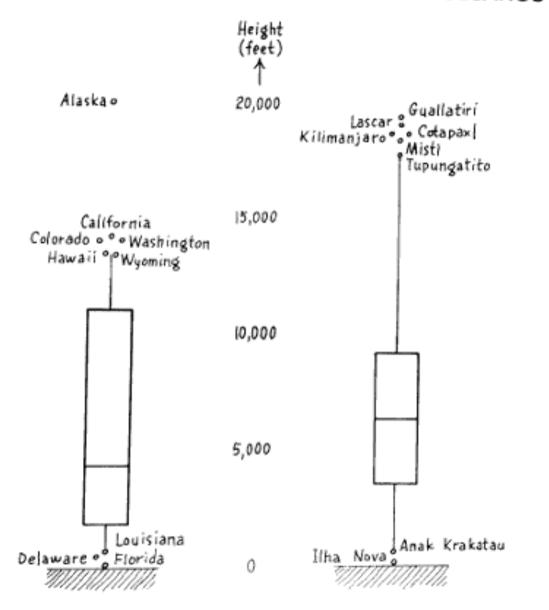
STATS & VISUALIZATION

- Exploratory Data Analysis
 - Tukey, 1977



Box-and-whisker plots with end values identified

- A) HEIGHTS of 50 STATES
- B) HEIGHTS of 219 VOLCANOS



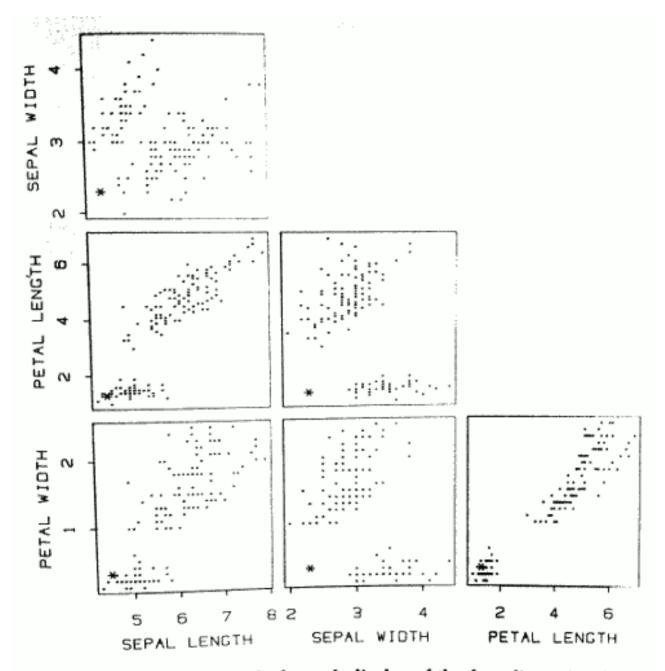
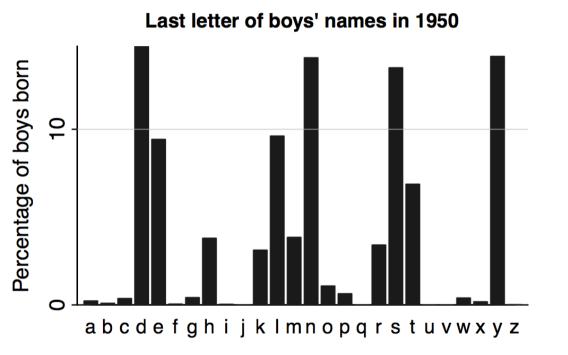
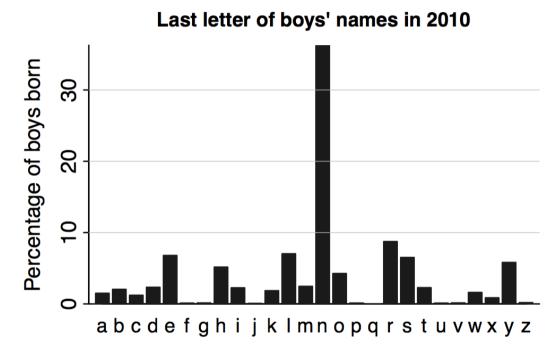
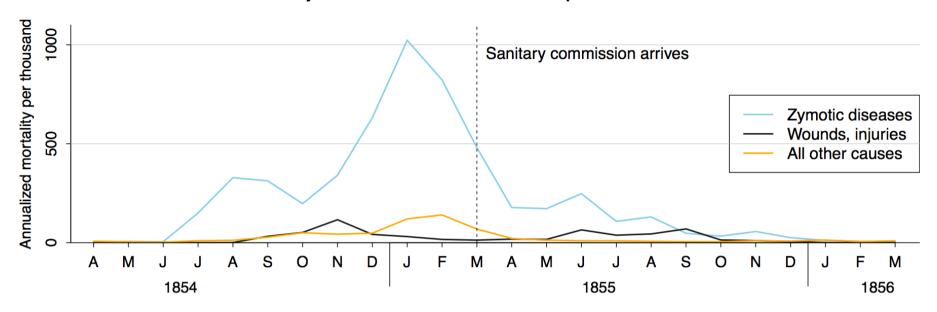


Figure 5.14 Generalized draftsman's display of the four-dimensional iris data (like Figure 5.11), with one flower plotted as an asterisk.

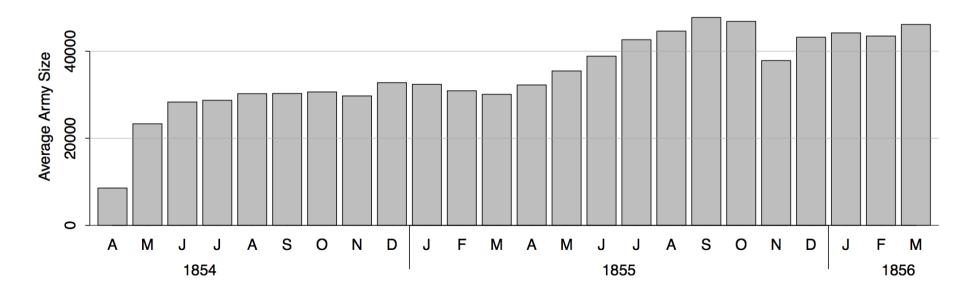




Mortality rates in the Crimean War from April 1854 to March 1856



British Army Size in the Crimean War from April 1854 to March 1856



46 64 54 77 67 68 62 56 38 Population N = 9

Random

Sample 38 62 67 62

n = 4

$$\overline{X} = \frac{\sum X}{n} = \frac{229}{4} = 57.25$$

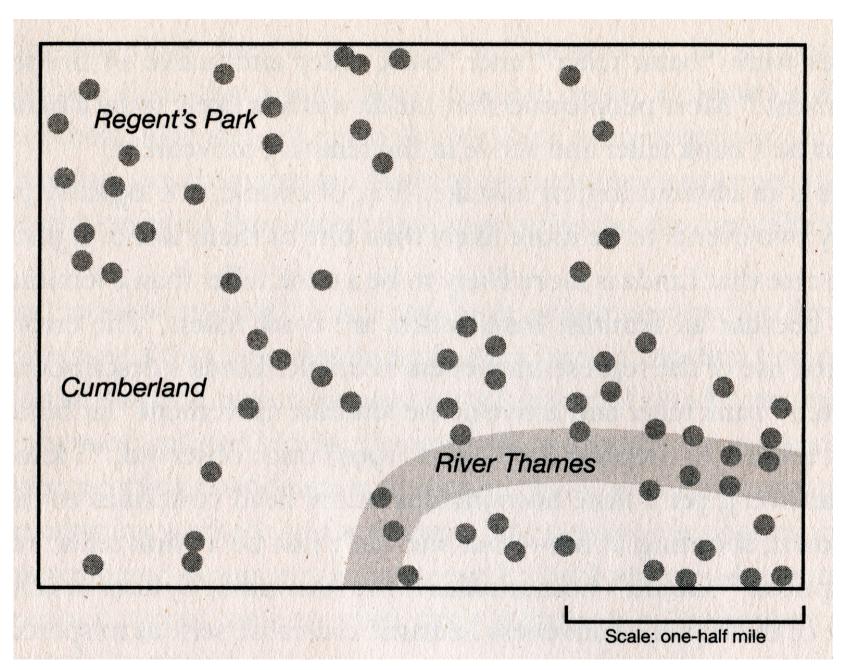
The mean of this Random Sample equals 57.25 (i.e. $\overline{\chi} = 57.25$)

 $\mu_{x} = \frac{\sum x}{N} = \frac{532}{9} = 59.11$

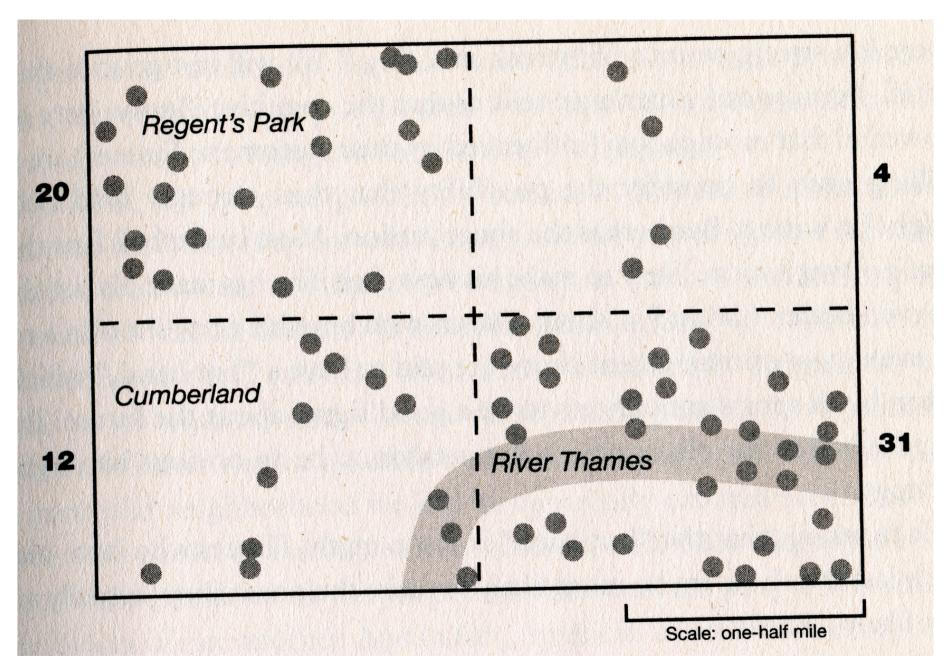
The Mean of this Population (μ_{x}) equals 59.11 (i.e. μ_{x} = 59.11)

The Central Limit Theorem tells us that $\bar{\chi}$ is an unbiased estimate of μ_{X} . (i.e. $\bar{\chi} \longrightarrow \mu_{X}$)

In short, with only one random sample to go on, the mean of the sample ($\bar{\chi}$ = 57.25) is our best estimate of the population mean (μ_{χ})



German bombings in London during WWII



German bombings in London during WWII

STATS & VISUALIZATION

- Confirmatory Analysis
 - Testing hypotheses
 - Example: is this new drug effective?
 - Strong focus on automatic procedures, computation and objectivity
 - Looking at data can impair objectivity:
 - Data dredging, snooping, fishing, mining

STATS & VISUALIZATION

Exploratory data analysis is sometimes compared to detective work: it is the process of gathering evidence.

Confirmatory data analysis is comparable to a court trial: it is the process of evaluating evidence.

Exploratory analysis and confirmatory analysis "can—and should—proceed side by side" (Tukey; 1977).

WHAT ARE STATS?

- A set of tools and methods
- Old tradition:
 - Origins in demographics
 - Draws from mathematics & probability theory
 - Visual representations are also important
 - A (generally) strong focus on (computationally cheap) numerical calculations

STATISTICAL TOOLS

DESCRIPTIVE STATISTICS



AN EXAMPLE

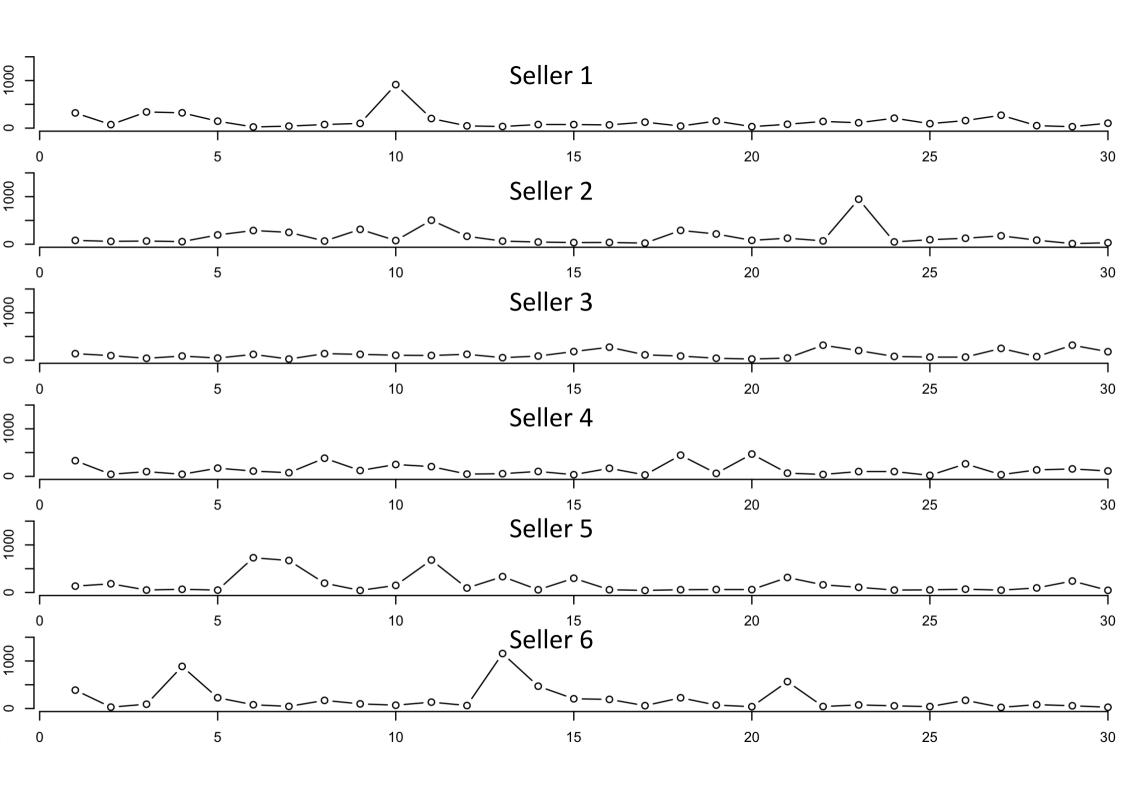
Selling encyclopedias

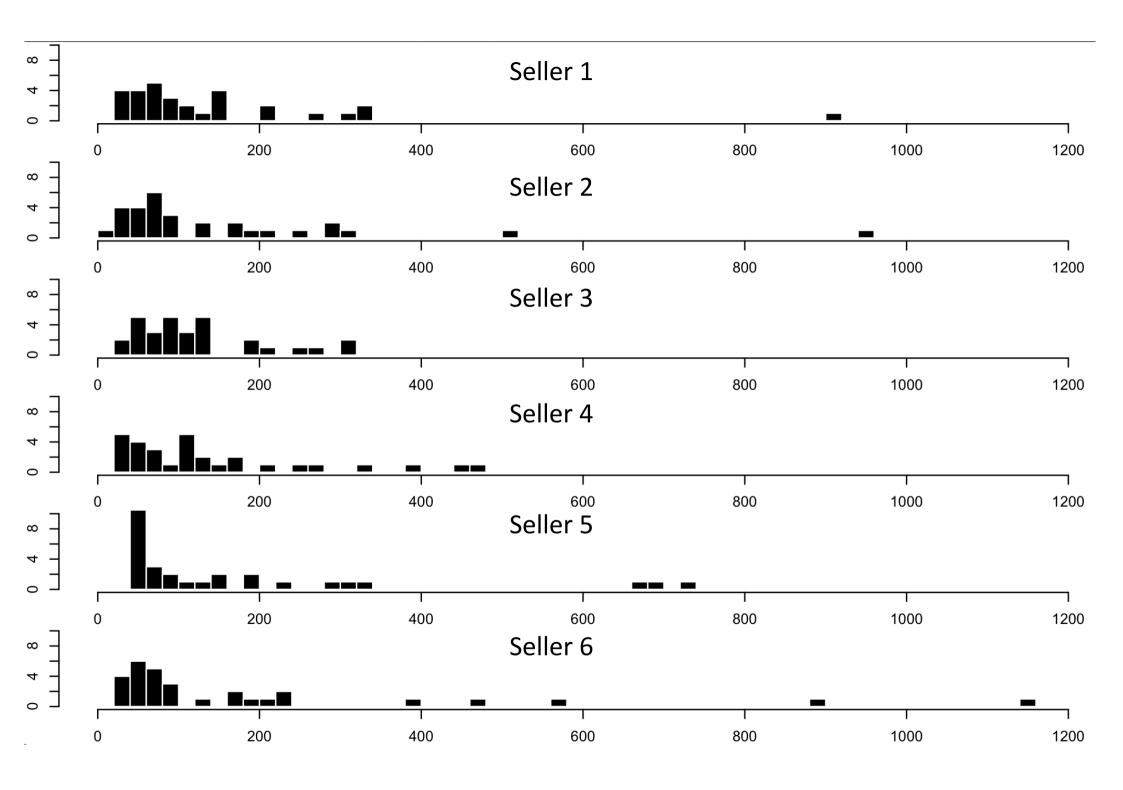




day	Seller 1	Seller 2	Seller 3	Seller 4	Seller 5	Seller 6
1	€320	€80	€139	€330	€133	€387
2	€74	€60	€98	€44	€182	€29
3	€340	€67	€42	€100	€51	€91
4	€322	€54	€89	€44	€67	€886
5	€146	€195	€47	€173	€49	€227
6	€24	€288	€124	€111	€730	€79
7	€42	€249	€26	€77	€672	€45
8	€76	€67	€140	€382	€195	€171
9	€99	€312	€125	€123	€43	€98
10	€915	€77	€106	€250	€149	€70
11	€202	€504	€101	€205	€682	€134
12	€47	€167	€126	€48	€93	€63
13	€34	€65	€55	€56	€333	€1,157
14	€76	€46	€89	€104	€56	€470
15	€75	€34	€184	€35	€299	€205
16	€68	€37	€275	€170	€57	€192

day	Seller 1	Seller 2	Seller 3	Seller 4	Seller 5	Seller 6
1	€320	€80	€139	€330	€133	€387
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14	€76	€46	€89	€104	€56	€470
15	€75	€34	€184	€35	€299	€205
16	€68	€37	€275	€170	€57	€192
17	€126	€23	€114	€30	€43	€60
18	€43	€290	€89	€446	€57	€226
19	€149	€215	€43	€63	€62	€72
20	€31	€81	€26	€469	€60	€39
21	€81	€127	€47	€68	€315	€566
22	€141	€70	€317	€40	€160	€42
23	€113	€947	€203	€102	€108	€76
24	€209	€48	€81	€102	€50	€56
25	€94	€95	€67	€21	€54	€41
26	€159	€125	€67	€263	€69	€173
27	€271	€176	€250	€35	€48	€24
28	€52	€85	€77	€136	€95	€82
29	€30	€12	€317	€157	€240	€58
30	€104	€31	€181	€113	€45	€27

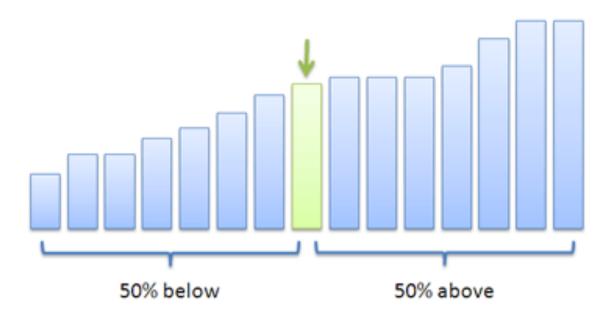




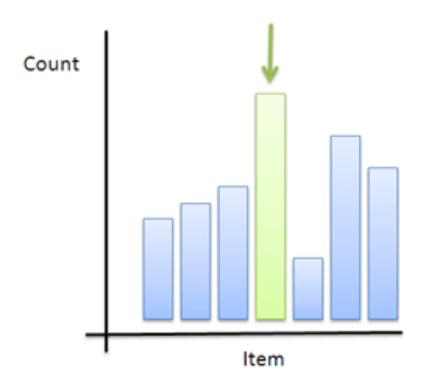
From Kalid Azad

Name & Meaning	Formula / Example	Used for
Arithmetic Mean [average]	$\frac{sum}{size} = \frac{a+b+c}{3}$	Most situations ("average item")
Median [middle value]	Middle of sorted list (2 middles? Average 'em)	Wildly varying samples (houses, incomes)
Mode [most popular]	Most popular value	No compromises (winner takes all)
Geometric Mean [average factor]	$\sqrt[3]{abc}$	Investments, growth, area, volume
Harmonic Mean [average rate]	$\frac{3}{\frac{1}{a} + \frac{1}{b} + \frac{1}{c}}$	Speed, production, cost

Median



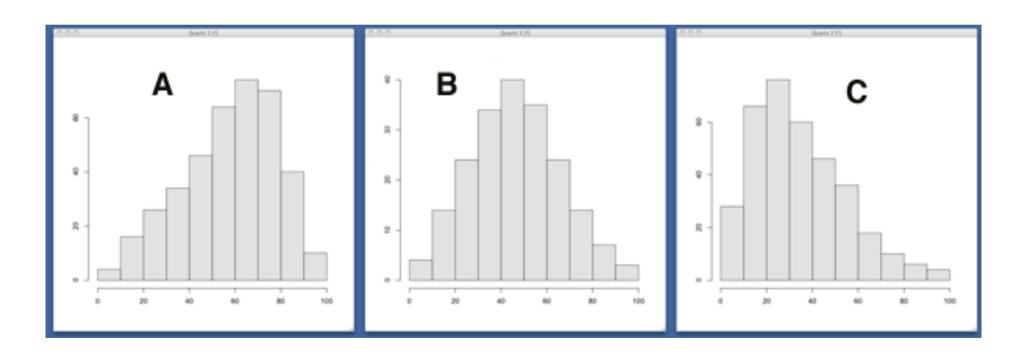
Mode (Most Popular)



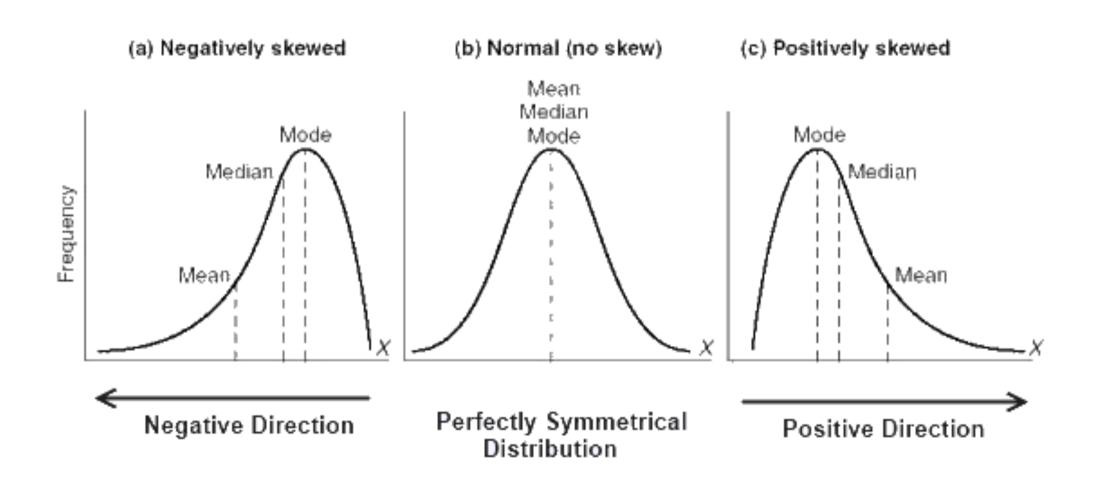
negative skew

symmetric

positive skew



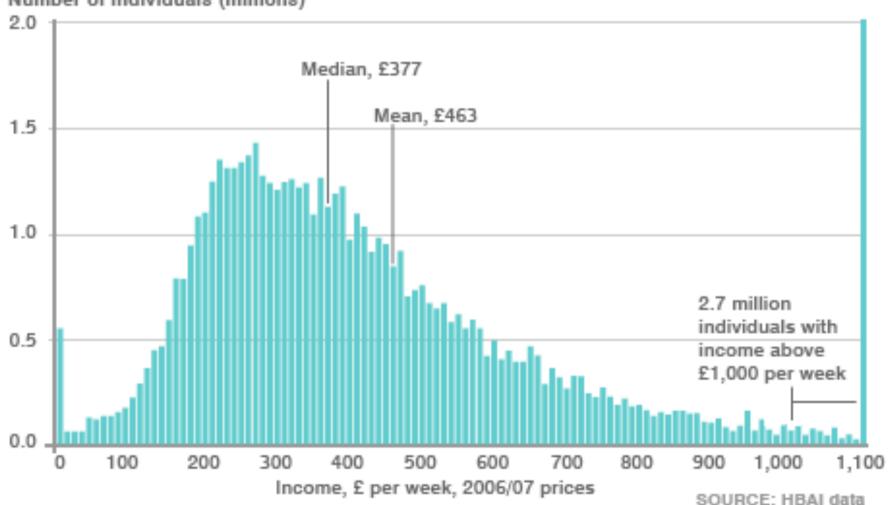




CENTRALTENDENCY

THE UK INCOME DISTRIBUTION IN 2006 / 7

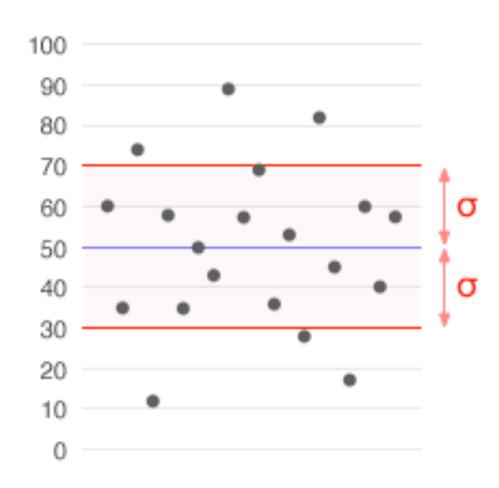
Number of individuals (millions)



DISPERSION

Standard Deviation

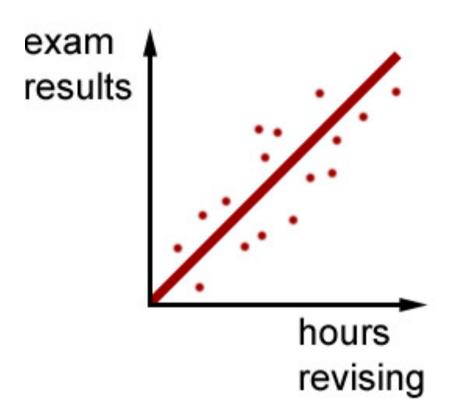
$$\sigma = \sqrt{\frac{1}{N}} \sum_{i=1}^{N} (x_i - \mu)^2$$



Source unknown

DEPENDENCE

Correlation

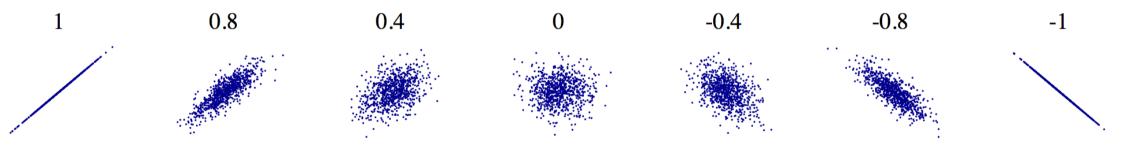


POSITIVE CORRELATION

 people who do more revision get higher exam results.

DEPENDENCE

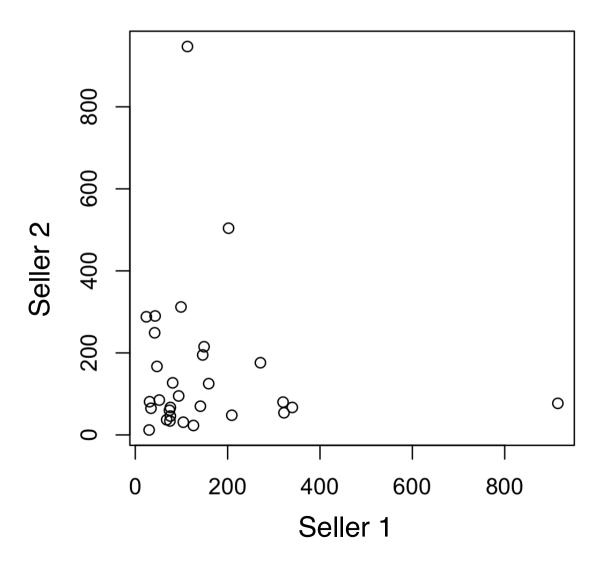
Correlation

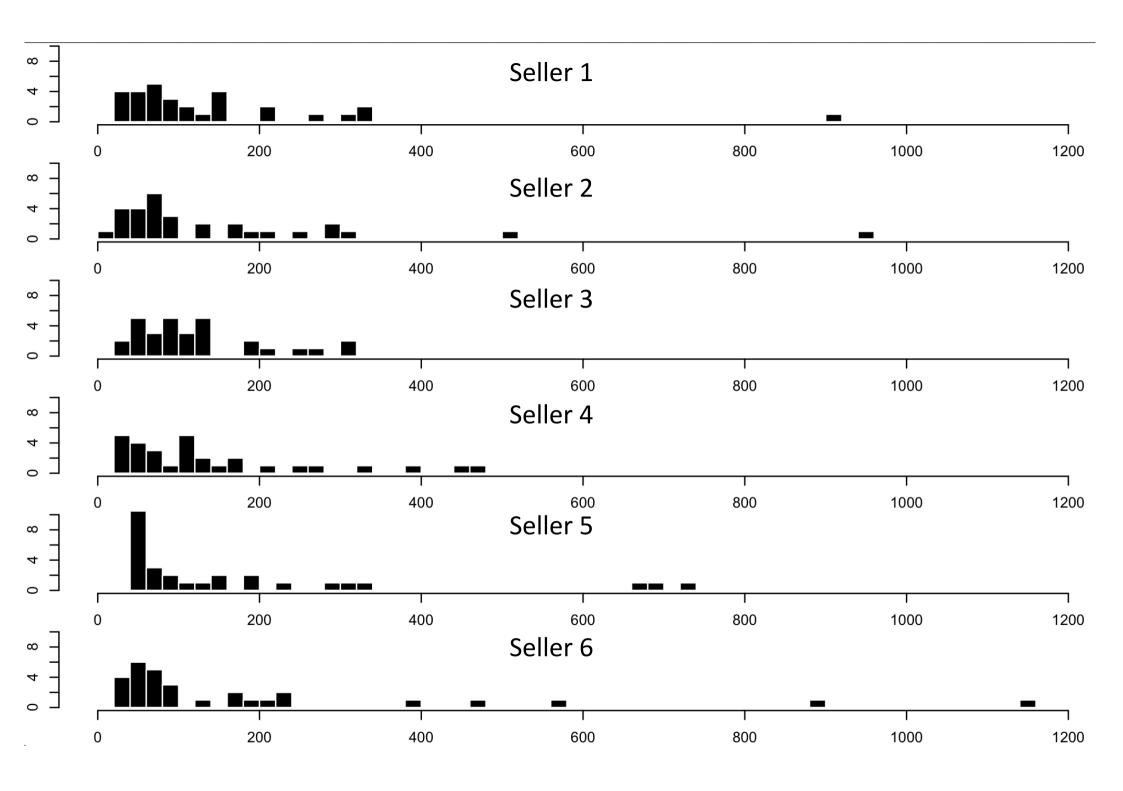


DEPENDENCE

Correlation

$$r = -0.08$$



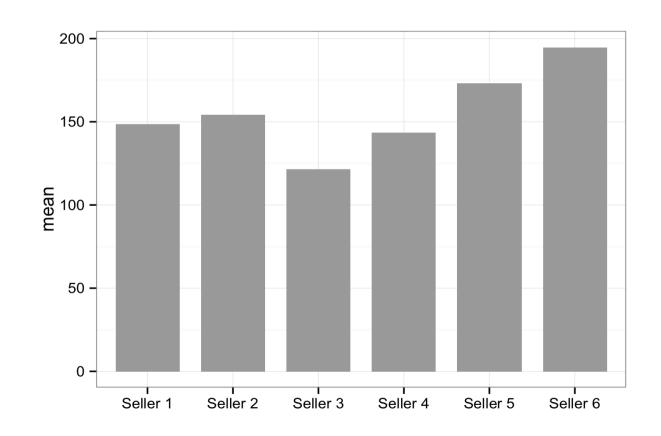


Average Sales

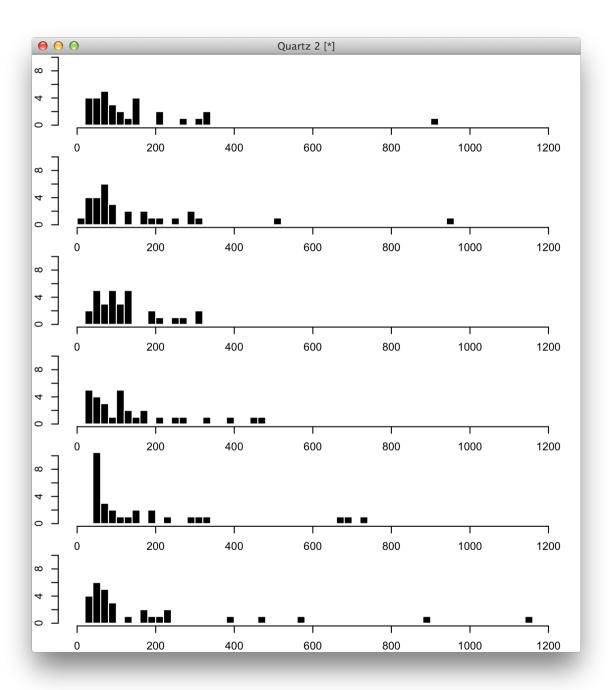
Seller 1	Seller 2	Seller 3	Seller 4	Seller 5	Seller 6
€149	€154	€122	€143	€173	€195

Average Sales

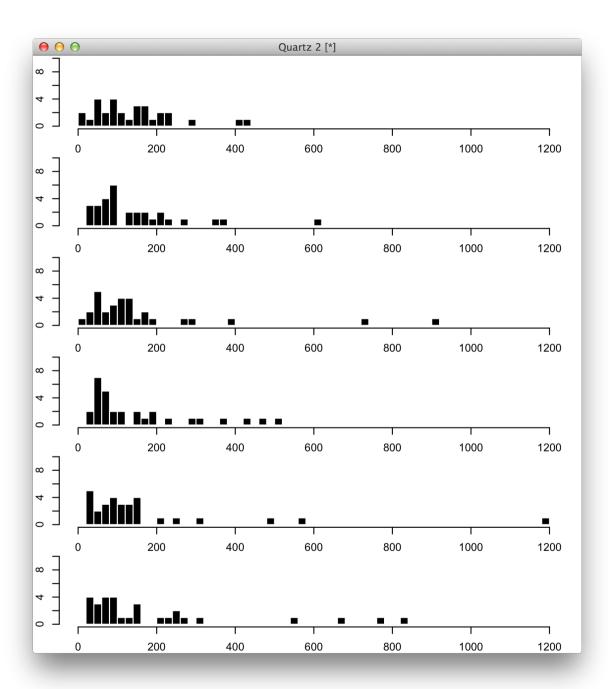
Seller 1	Seller 2	Seller 3	Seller 4	Seller 5	Seller 6
€149	€154	€122	€143	€173	€195



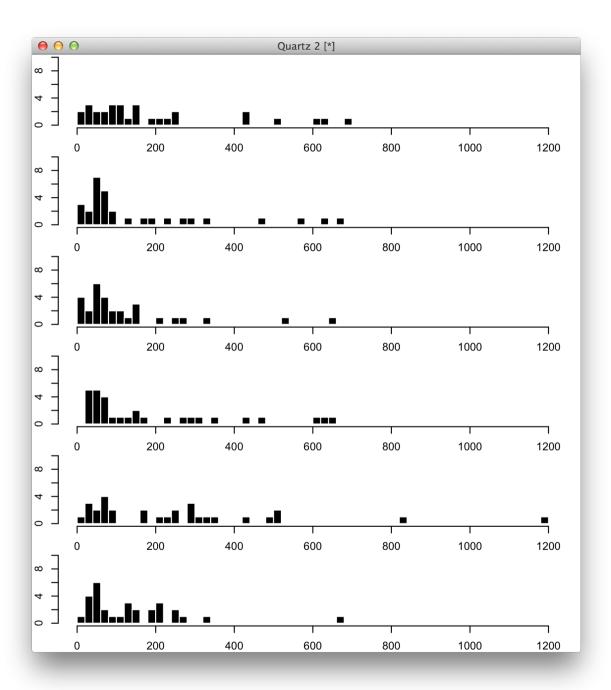
COLLECTING MORE DATA



September 2014



October 2014

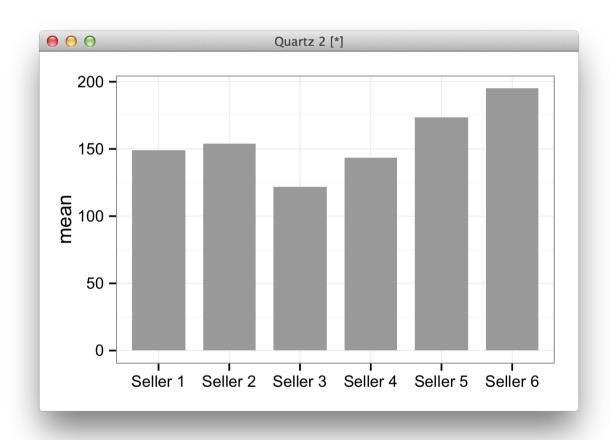


November 2014

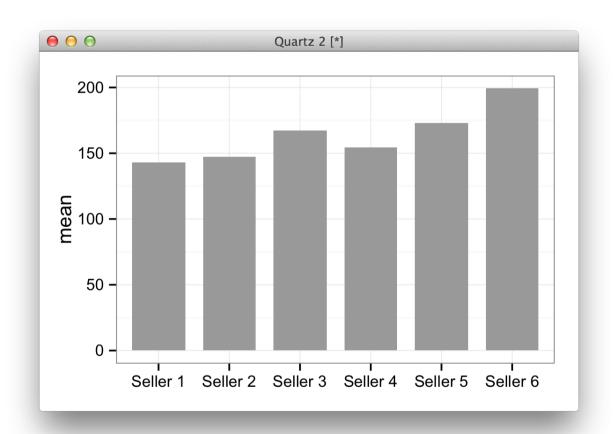
Quartz 2 [*]

december 2014

September 2014



October 2014



November 2014



December 2014



September 2014

day	Seller 1	Seller 2	Seller 3	Seller 4	Seller 5	Seller 6
1	€320	€80	€139	€330	€133	€387
2	€74	€60	€98	€44	€182	€29
3	€340	€67	€42	€100	€51	€91
4	€322	€54	€89	€44	€67	€886
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8	€76	€67	€140	€382	€195	€171
9	€99	€312	€125	€123	€43	€98
10	€915	€77	€106	€250	€149	€70
11	€202	€504	€101	€205	€682	€134
12	€47	€167	€126	€48	€93	€63
13	€34	€65	€55	€56	€333	€1,157
14	€76	€46	€89	€104	€56	€470
15	€75	€34	€184	€35	€299	€205
16	€68	€37	€275	€170	€57	€192

September 2014

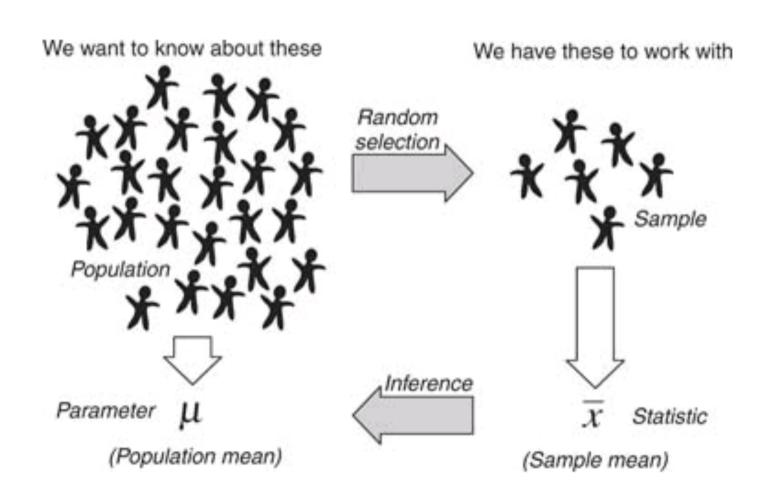


How much can we trust this chart?

STATISTICAL TOOLS

INFERENTIAL STATISTICS





Terminology:

- Population vs. sample
- Sample **statistic** (mean, median, etc.)
- Population parameter (mean, median, etc.)

Unit of statistical analysis

day	Seller 1	Seller 2	Seller 3	Seller 4	Seller 5	Seller 6
1	€320	€80	€139	€330	€133	€387
2	€74	€60	€98	€44	€182	€29
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11	€202	€504	€101	€205	€682	€134

Unit of statistical analysis

day	Seller 1
1	€320
2	€74
3	€340
4	€322
5	€146
6	€24
7	€42
8	€76
9	€99
10	€915

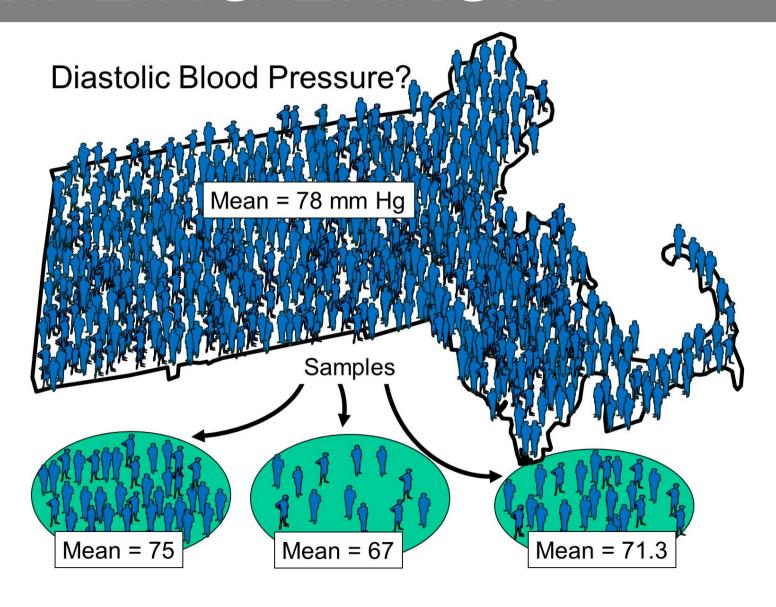
Unit of statistical analysis

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8	€76	€67	€140	€382	€195	€171
9	€99	€312	€125	€123	€43	€98
10	€915	€77	€106	€250	€149	€70
11	€202	€504	€101	€205	€682	€134

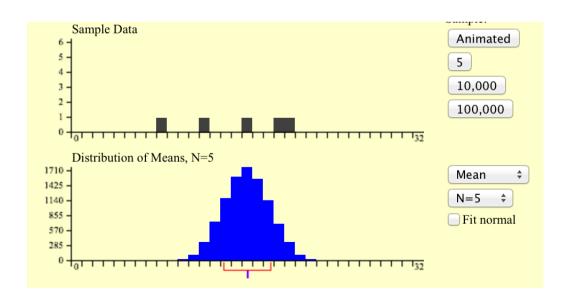
Unit of statistical analysis

Average Sales

Seller 1	Seller 2	Seller 3	Seller 4	Seller 5	Seller 6
€149	€154	€122	€143	€173	€195

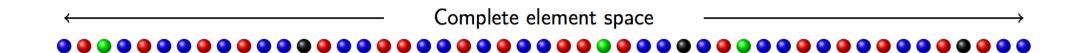


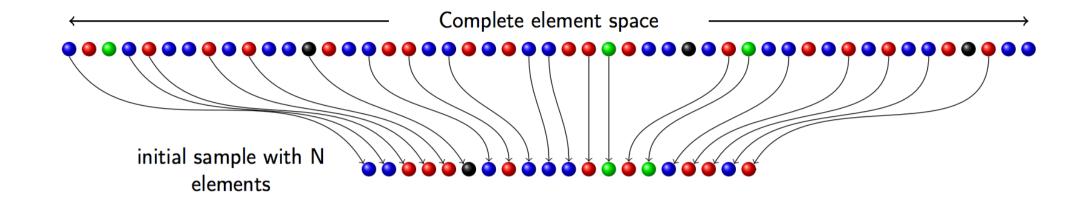
- Sampling distribution of a statistic
 - Demo

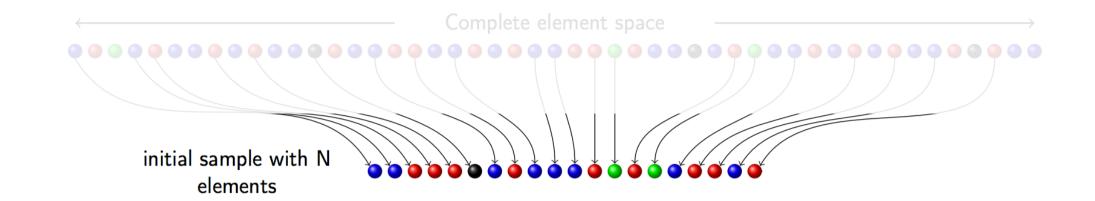


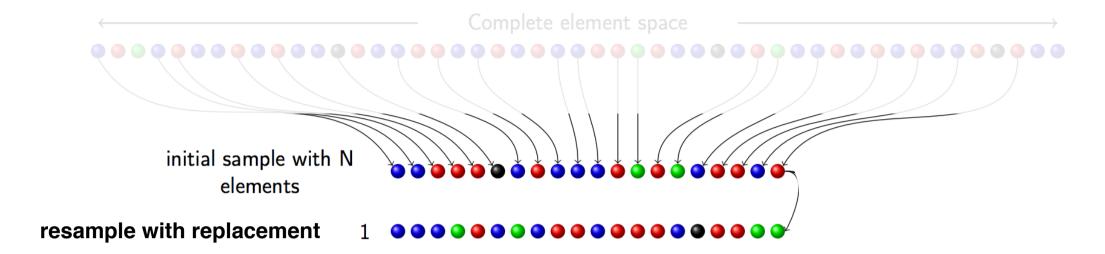
- Resampling techniques
 - Bootstrapping

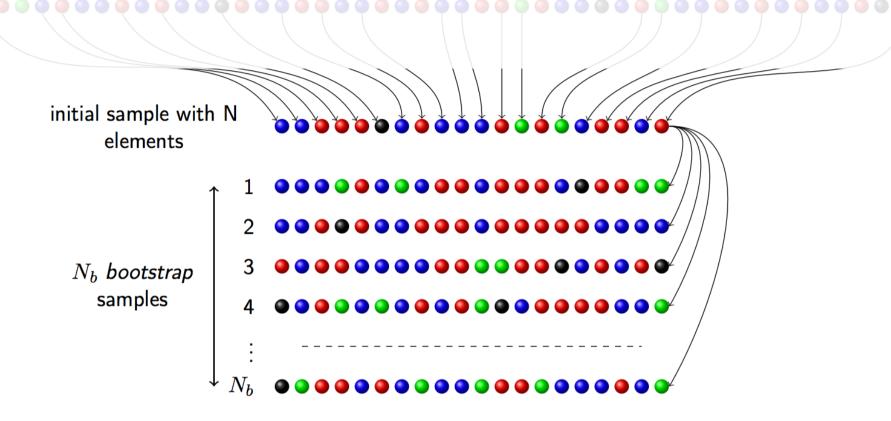


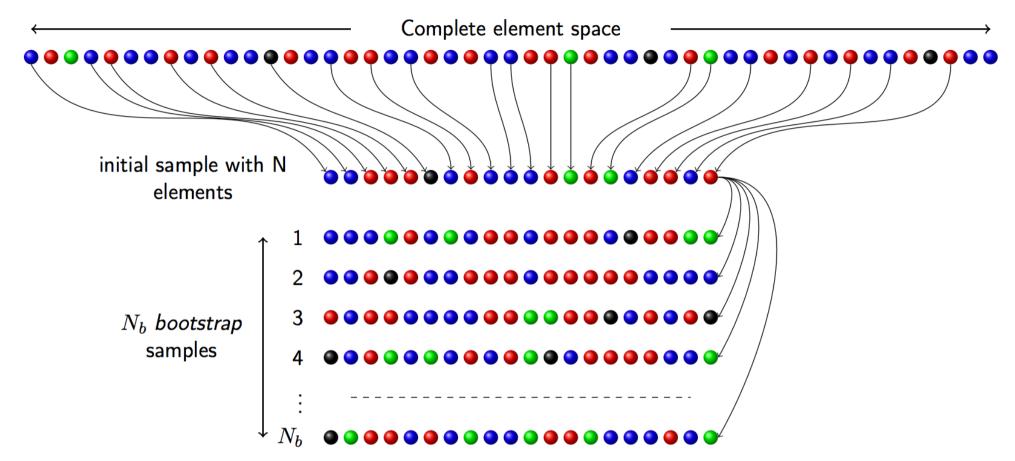








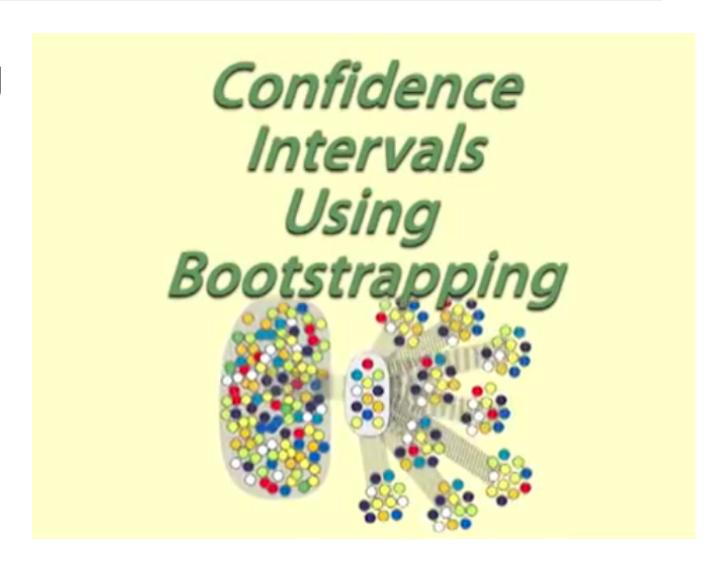




Theorem (B. Efron, Ann. Statist. 1979)

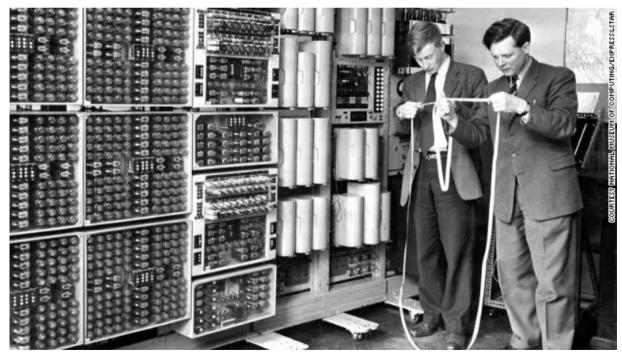
When N tend to infinity, the distribution of average values computed from bootstrap samples is equal to the distribution of average values obtained from ALL samples with N elements which can be constructed from the complete space. Thus the width of the distribution gives an evaluation of the sample quality.

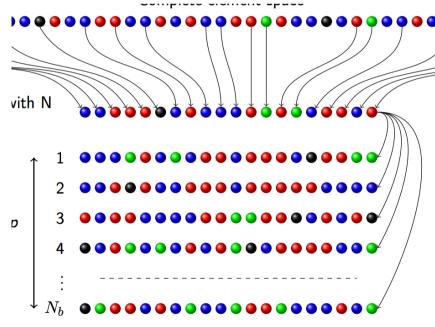
 Bootstrapping video

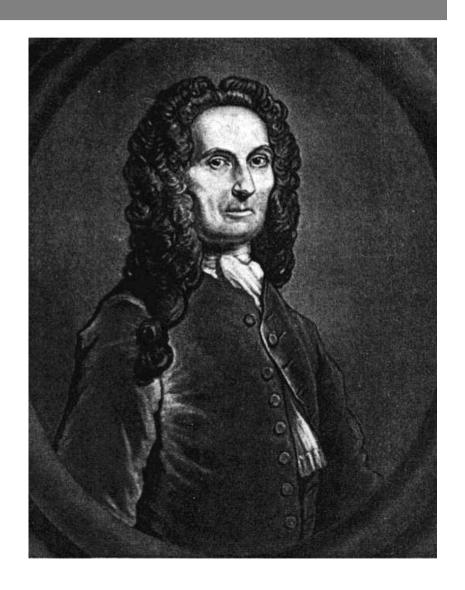


SAMPLING ERROR

How did people do before computers?

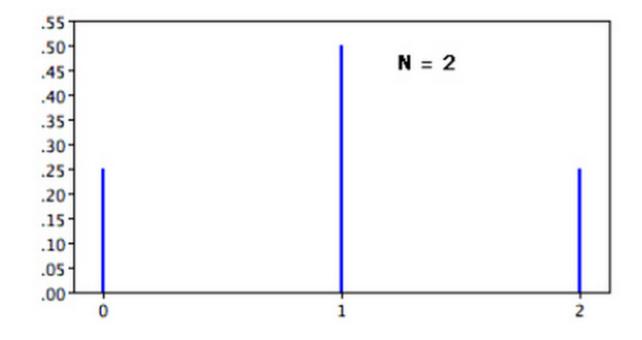




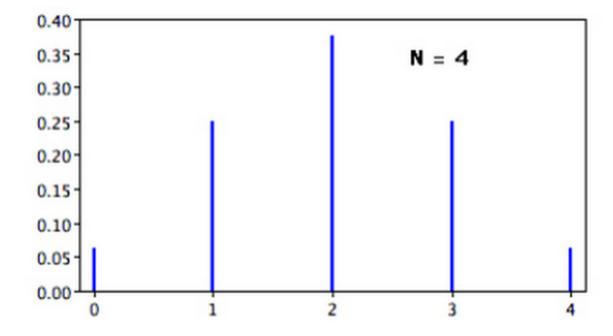




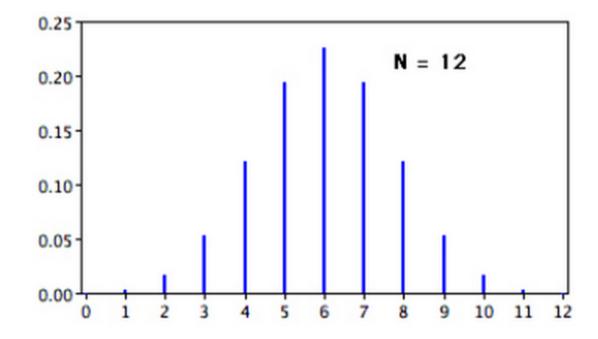




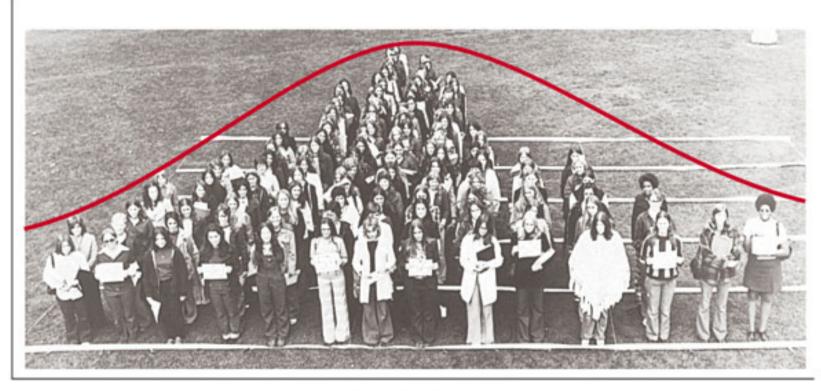




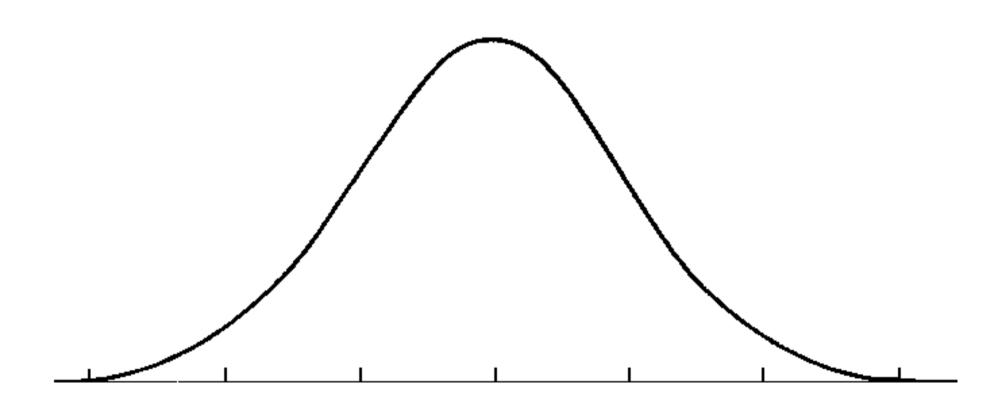




Number of individuals

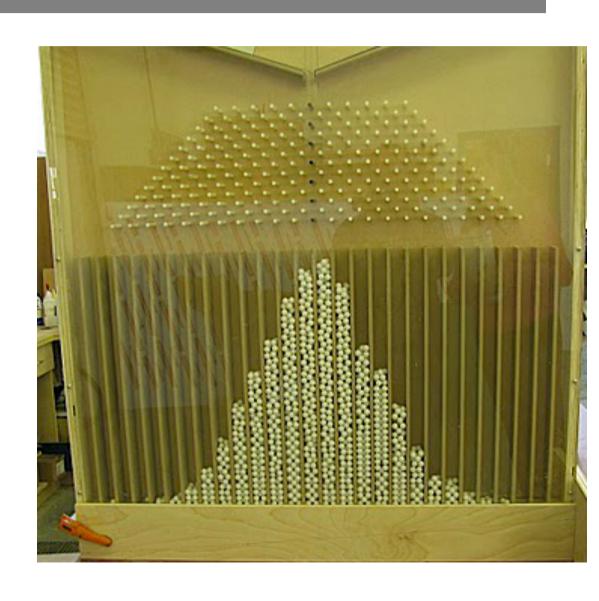


Height in inches



Sir Francis Galton
 1822 – 1911

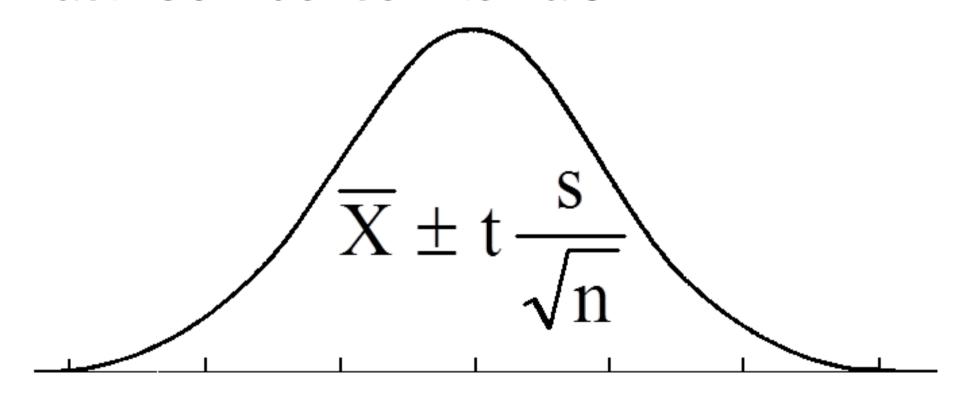
Bean Machine or Galton Board:



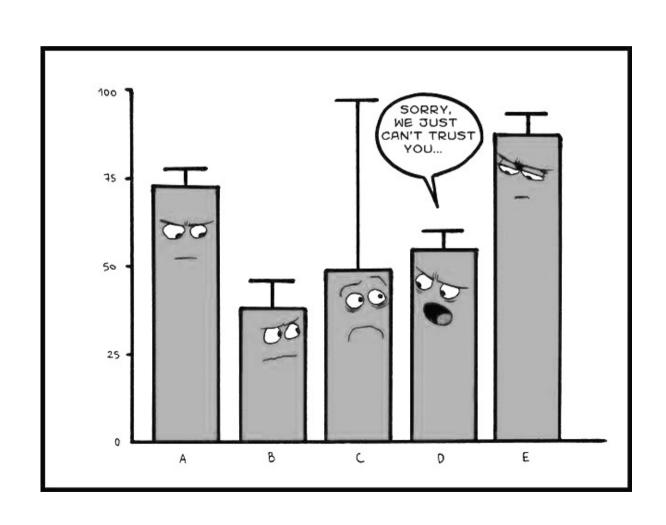
Central Limit Theorem

Given certain conditions, the arithmetic mean of a sufficiently large number of iterates of independent random variables, each with a well-defined expected value and well-defined variance, will be approximately normally distributed

"Exact" Confidence Intervals



t ~ 1.96 for large samples



- Several interpretations
- « a range of plausible values for μ. Values outside the Cl are relatively implausible. » (Cumming and Finch, 2005)
- Examples of presentation formats:

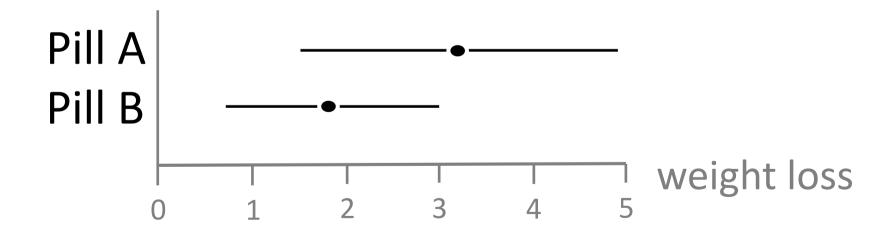
```
2.2m, 95% CI [1.6m, 2.8m]
```

2.2m +/- 0.6m

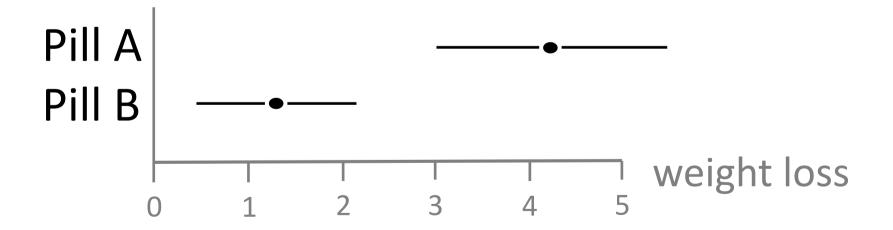
from 1.6m to 2.8m



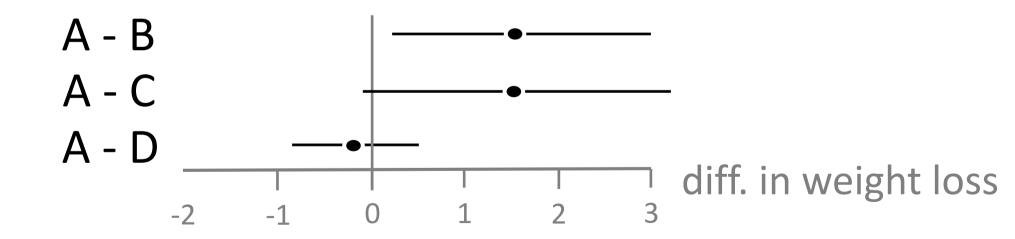
 « a range of plausible values for μ. Values outside the CI are relatively implausible. » (Cumming and Finch, 2005)



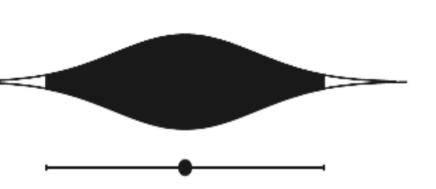
 « a range of plausible values for μ. Values outside the CI are relatively implausible. » (Cumming and Finch, 2005)



 « a range of plausible values for μ. Values outside the CI are relatively implausible. » (Cumming and Finch, 2005)



 "values close to our M are the best bet for μ, and values closer to the limits of our CI are successively less good bets."



(Cumming, 2013)

BACK TO OUR EXAMPLE

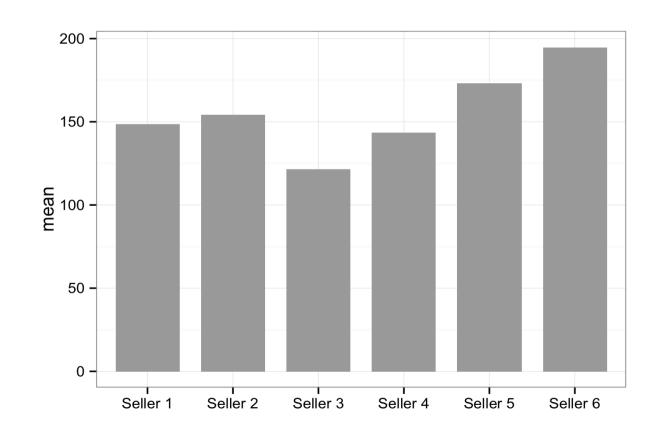
Selling encyclopedias

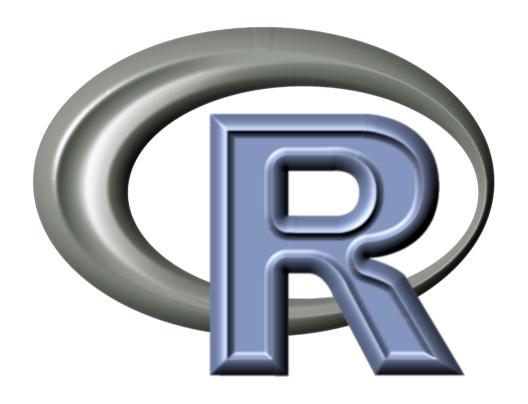




Average Sales

Seller 1	Seller 2	Seller 3	Seller 4	Seller 5	Seller 6
€149	€154	€122	€143	€173	€195





https://www.lri.fr/~dragice/stats-rjc.zip

Doing the New Statistics

Pierre Dragicevic

Statistics

In the context of HCI user studies

- 3 things to keep in mind:
 - Not all papers need a user study
 - Not all user studies are experiments
 - Doing experiments is not all about statistics

but doing experiments require doing some statistics

Statistics

We gave a data retrieval task to 12 subjects. Half of them used a bar chart and the other half used a line chart.

The measured accuracies were (12.1%, 11.6%, 18.3%, 19.2%, 11.1%, 7.0%) for bar charts, and (13.0%, 13.9%, 12.1%, 13.5%, 21.9%, 12.4%) for line charts.

Statistics

We gave a data retrieval task to 12 subjects. Half of them used a bar chart and the other half used a line chart.

Average accuracy was **9.2%** for bar charts, and **13.2%** for line charts.

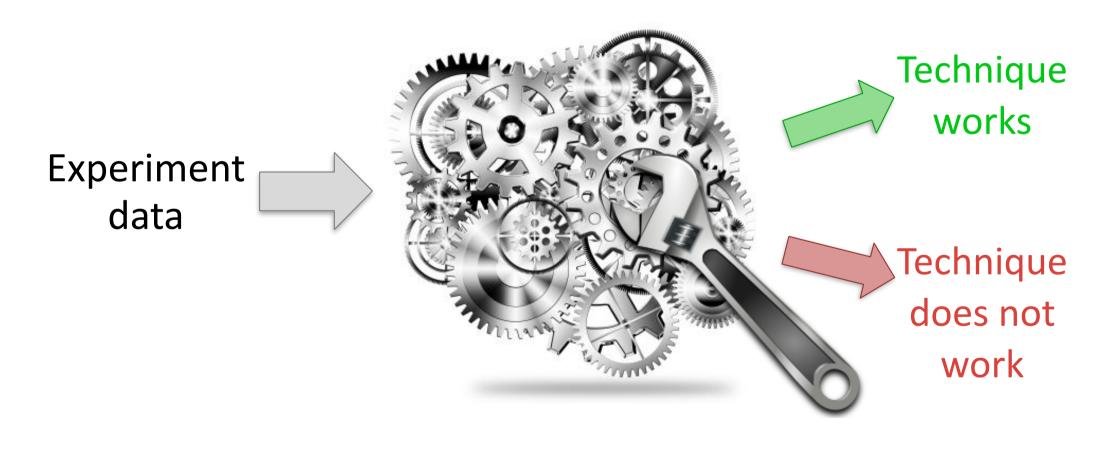
Bad HCI Statistics

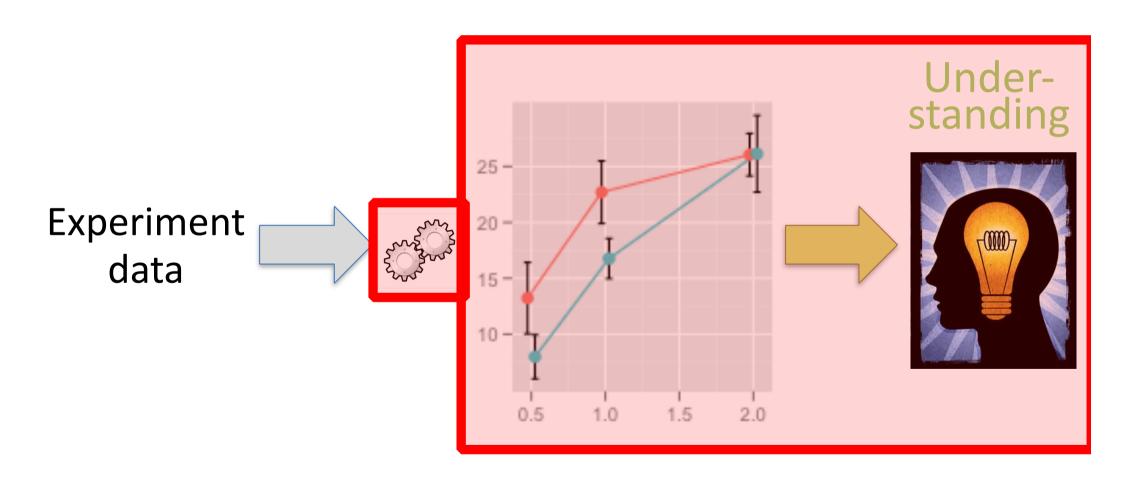


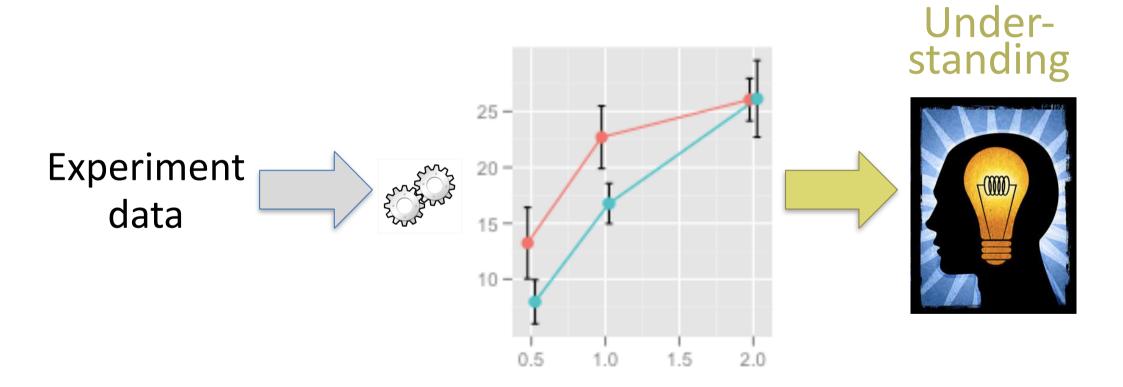
Bad HCI Statistics



Bad HCI Statistics





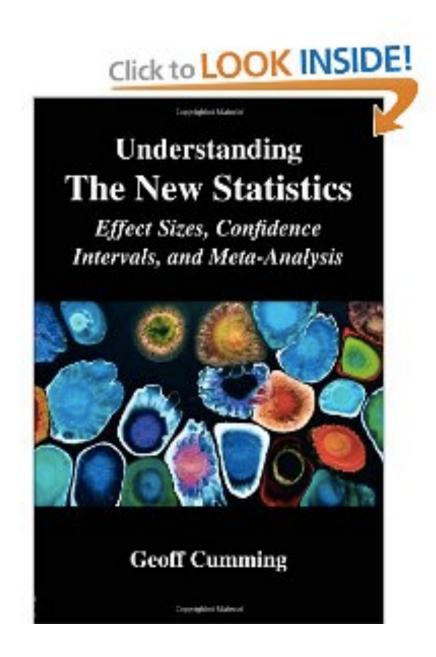






Understanding Understanding 25 -70007 20 -15 -10 -**Publication** Investigator Peers

The New Statistics



The New Statistics

« The techniques are not new, but adopting them widely would be new for many researchers, as well as highly beneficial. »

(Cumming, 2013)

The New Statistics

- The « old » statistics:
 - Null hypothesis significance testing (NHST)
 - p-values
 - Example: tech A is faster than tech B, p = .032
- The « new » statistics:
 - Estimations instead of tests
 - Effect sizes + confidence intervals (CIs)
 - Example: tech A is faster than tech B by 1.3 seconds, 95%
 CI [0.3s, 1.6s].

FAQ

- What's an effect size?
- What's a CI?
- Why switch to CIs? Who says that?
- Is reporting p-values + CIs OK?
- How to compute Cls?
- How to graph Cls?
- How to interpret Cls?
- Will my paper be rejected?

References

- (Keene, 1995) The log transform is special
- (Schmidt and Hunter, 1997) Eight common but false objections to the discontinuation of significance testing in the analysis of research data.
- (Wilkinson et al, 1999) Statistical Methods in Psychology Journals.
- (Cumming and Finch, 2005) Inference by Eye: Confidence Intervals and How to Read Pictures of Data.
- (Baguley, 2009). Standardized or simple effect size: What should be reported?
- (Sauro and Lewis, 2010). Average task times in usability tests: what to report?
- (Cumming, 2011). Cumming, G. Understanding the New Statistics: Effect Sizes, Confidence Intervals, and Meta-Analysis.
- (Dragicevic, 2012). My Technique is 20% Faster: Problems with Reports of Speed Improvements in HCI.
- (Kirby and Gerlanc, 2012). BootES: An R Package for Bootstrap Confidence Intervals on Effect Sizes
- (Cumming, 2013) The New Statistics: Why and How.

- Taken broadly, « the amount of something that might be of interest » (Cumming, 2011)
- E.g., writing « tech A is faster than tech B by
 1.3 seconds » is reporting an effect size
- Things like Cohen's d are standardized effect sizes
- Many recommend reporting simple (unstandardized) effect sizes

"Only rarely will uncorrected standardized effect size be more useful than simple effect size. It is usually far better to report simple effect size [...]"

(Baguley, 2009)

"If the units of measurement are meaningful on a practical level (e.g., number of cigarettes smoked per day), then we usually prefer an unstandardized measure (regression coefficient or mean difference) to a standardized measure (r or d)."

(Wilkinson et al., 1999)

"(i) a preference for simple effect size over standardized effect size, and (ii) the use of confidence intervals to indicate a plausible range of values the effect might take."

(Baguley, 2009)

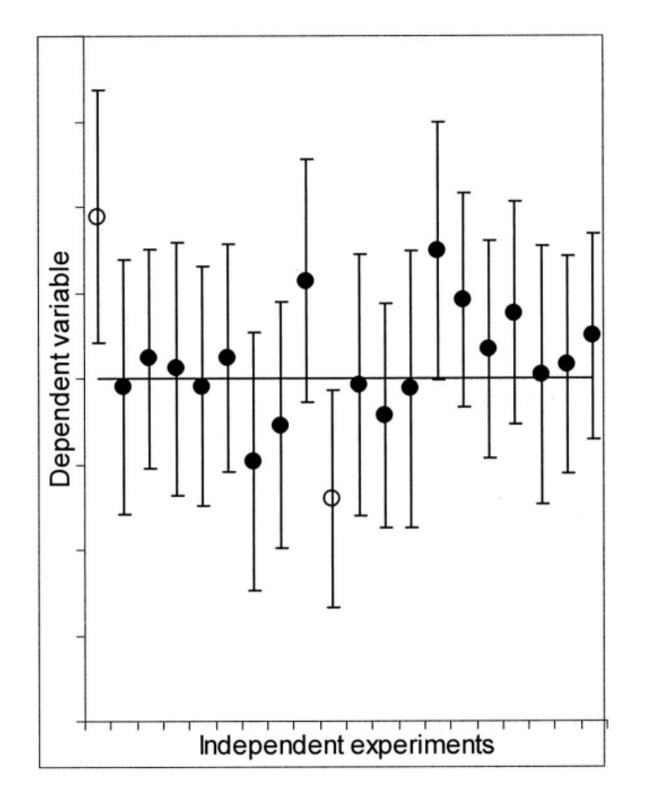
What's a confidence interval?

- Cumming gives several interpretations
- « a range of plausible values for μ . Values outside the CI are relatively implausible. » (Cumming and Finch, 2005)
- Examples of presentation formats:
 - 2.2 sec, 95% CI [1.6, 2.8]
 - $2.2 \sec +/- 0.6$
 - from 1.6 to 2.8 sec



What's a confidence interval?

- Cumming's favorite interpretation
- « our Cl is just one from an infinite sequence » (Cumming and Finch, 2005)



Make sure you check <u>the dance</u>
 <u>of p-values</u> on youtube

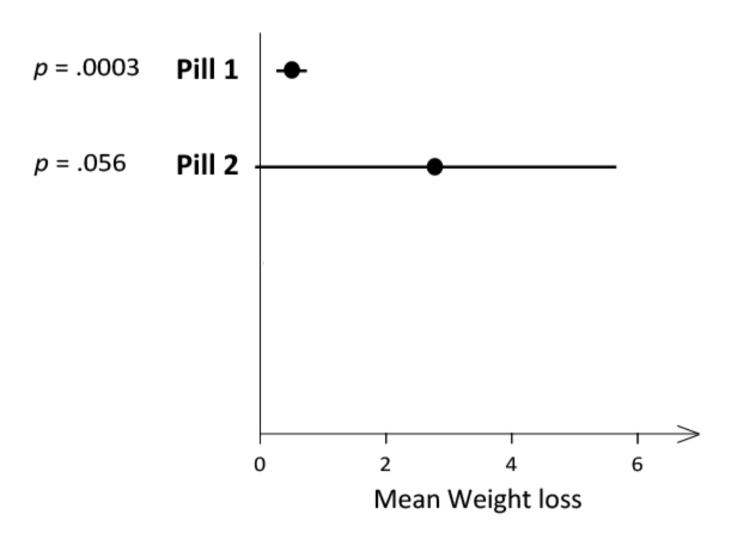
Why Switch to Cls? Who says that?

- 300+ articles by renown methodologists have been questioning NHST since the 1950s
- Many recommend switching to estimation
- Researchers have been mostly ignoring them, but now things seem to be changing
- The problem with NHST is mostly a human factor problem, so we should know more!

More at <u>www.aviz.fr/badstats</u>

Which weight-loss pill would you recommend?

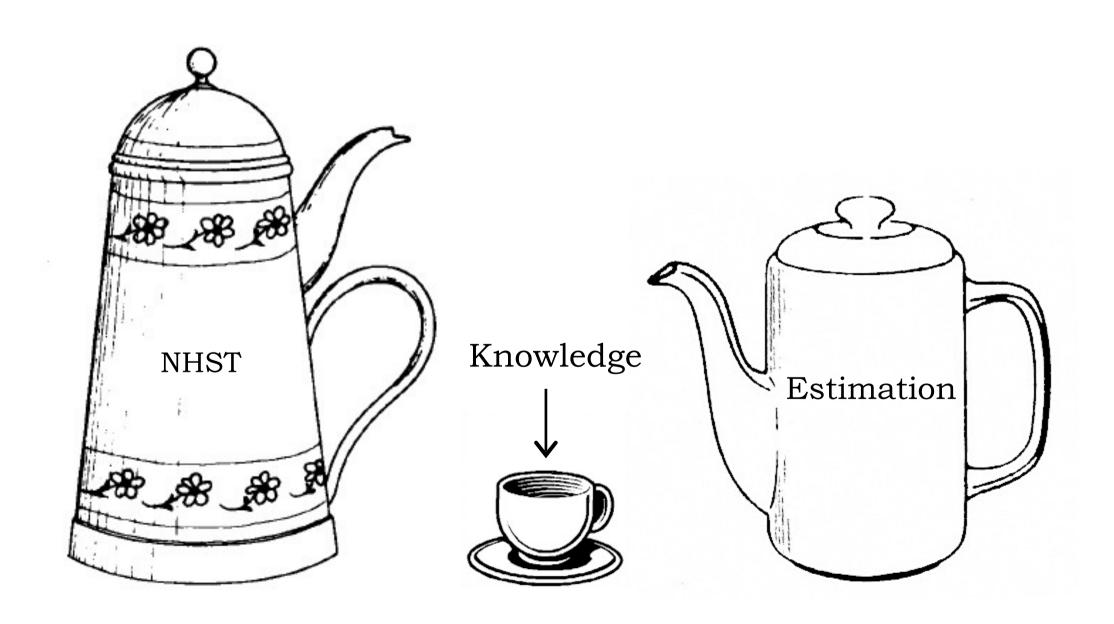




Error bars are 95% CIs p-values are based on a null hypothesis of no effect

Is reporting *p*-values + CIs OK?

Is reporting *p*-values + CIs OK?



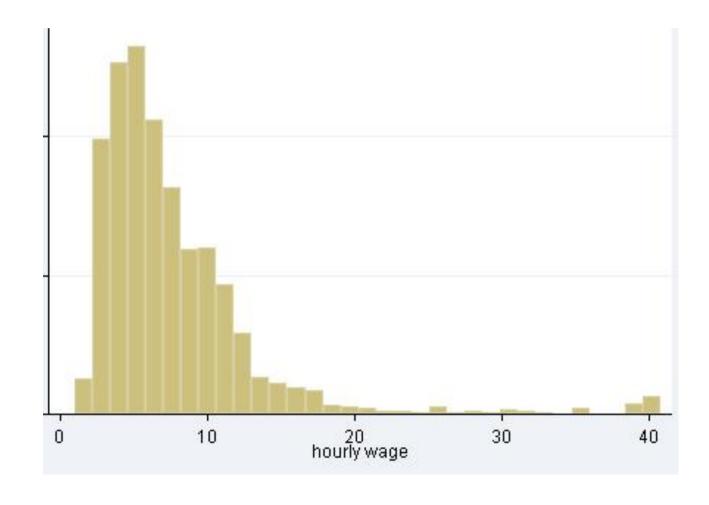
Is reporting *p*-values + CIs OK?



How to Compute Cls?

- Relatively easy using the R package
- A good place to start: <u>tinyurl.com/r-ci-tutorial</u>
- Aggregate your data first!!
- Less resources on:
 - Non-normal distributions
 - Complex designs
 - i.e., anything else than a between-subjects design with one factor and two levels

Skewed distributions



" For continuous positive data measured on an interval scale, a log transformed analysis should frequently be preferred to an untransformed analysis. No special justification beyond that sufficient to support an untransformed analysis should be required from the data obtained."

- Log transformation
 - Transform all your raw time measurements into logs
 - Do all your stats
 - Transform back when presenting your results

- Two important things:
 - Arithmetic means become geometric means
 - Differences between means become ratios between geometric means

Logarithmic identities

$$\log_b(xy) = \log_b(x) + \log_b(y)$$

$$\log_b(x^d) = d\log_b(x)$$

```
data = (a, b, c)
logdata = (log(a), log(b), log(c))
mean(logdata) = (log(a) + log(b) + log(c)) / 3
antilog(mean(logdata))
= exp [ (log(a) + log(b) + log(c)) / 3 ]
= exp [ log(abc) / 3 ]
= exp [ (1/3) * log(abc)]
= exp [ log((abc)<sup>1/3</sup>) ]
= abc<sup>1/3</sup>
```

- antilog(mean(logdata))
 - $= n^{th}$ root of the product of all measurements

- antilog(mean(logdata))
 - $= n^{th}$ root of the product of all measurements

• "The **geometric mean** is defined as the nth root of the product of n numbers." Wikipedia

$$\left(\prod_{i=1}^n a_i\right)^{1/n} = \sqrt[n]{a_1 a_2 \cdots a_n}.$$

"When providing an estimate of the average task time for small sample studies (n<25), the **geometric mean** is the best estimate of the center of the population (the median)."

(Sauro and Lewis, 2010)

"To find the **geometric mean**, convert the raw times using **a log-transformation**, find the mean of the transformed data, then transform back to the original scale by exponentiating."

(Sauro and Lewis, 2010)

- Log transformation
 - Transform all your raw time measurements into logs
 - Do all your stats
 - Transform back when presenting your results
- Two important things:
 - Arithmetic means become geometric means
 - Differences between means become ratios between geometric means

- Other distributions
 - Exponential
 - With both lower and upper bound

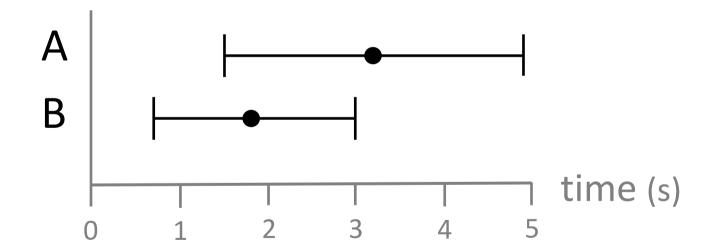
- Use bootstrapping (resampling)
 - R package boot
 - Simple and works with about any distribution (Kirby and Gerlanc, 2012)

Cls on Differences

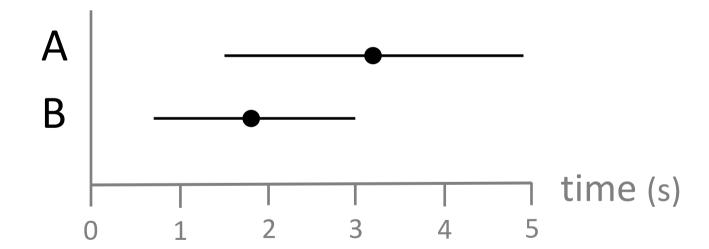
- Between-subject designs
 - different formula than sample mean
- Within-subject designs
 - just compute the difference on each pair of data points

- Multiple factors
- Multiple levels
- We'll see these later

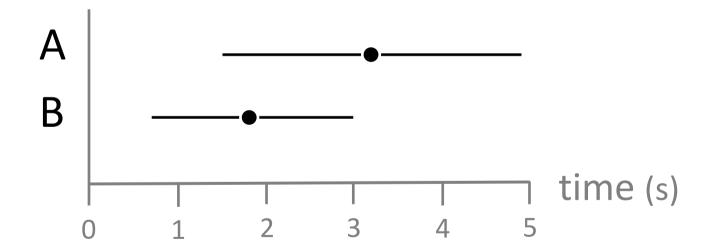
As error bars



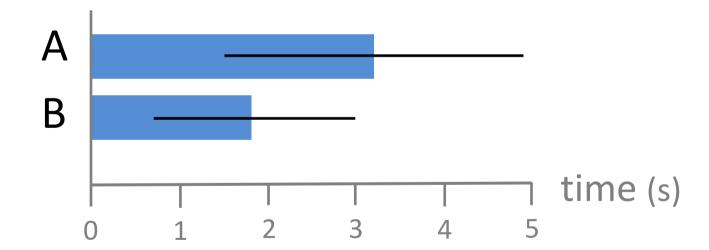
- As error bars
 - Better way:



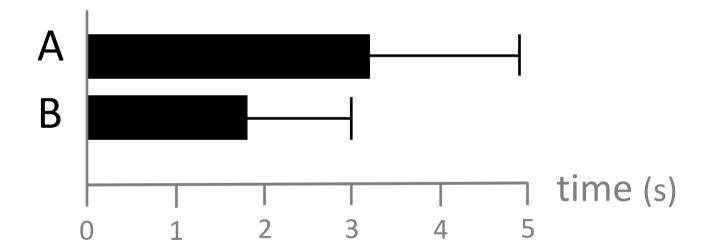
- As error bars
 - Slightly nicer:



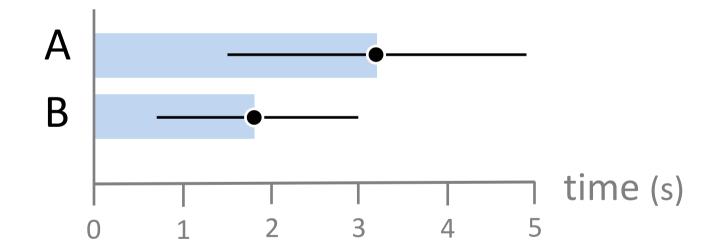
- As error bars
 - With bar charts:



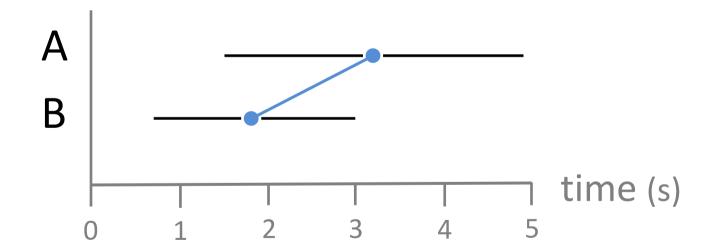
- As error bars
 - Dynamite plots:

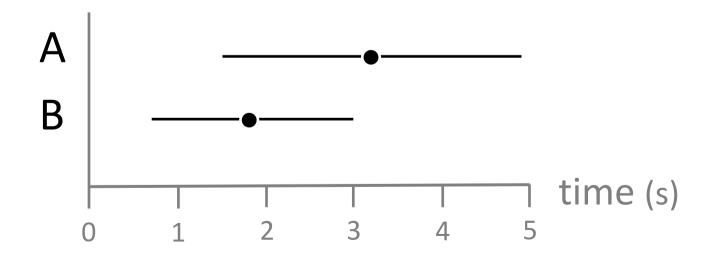


- As error bars
 - Perhaps a better approach:

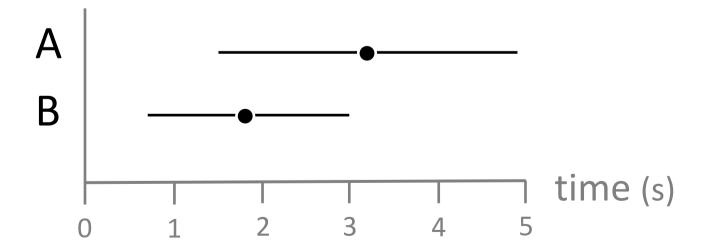


- As error bars
 - With line charts:

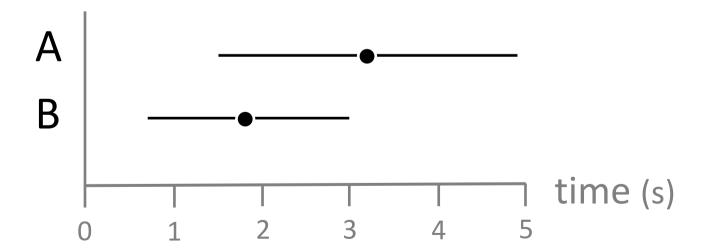




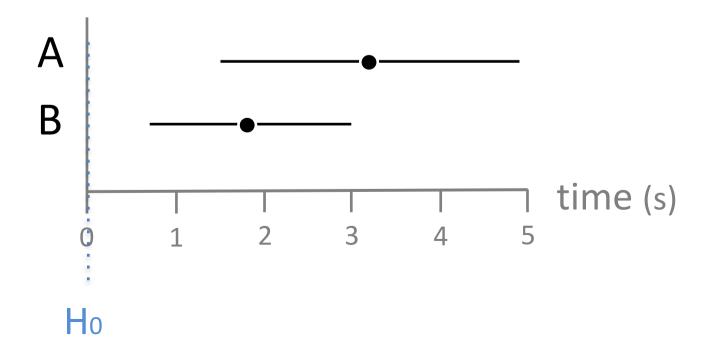
- Error bars could be anything
 - Standard Error (SE), Variance, various Cls, etc.
 - Use 95% CIs and specify in the legend



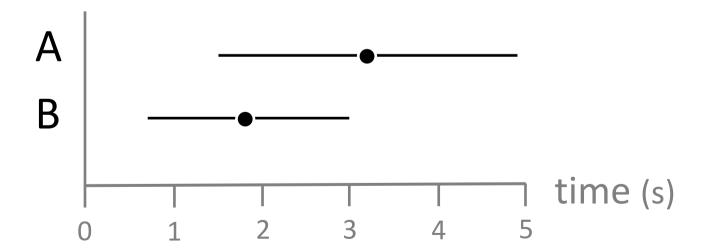
• Null hypothesis: H₀ = Os



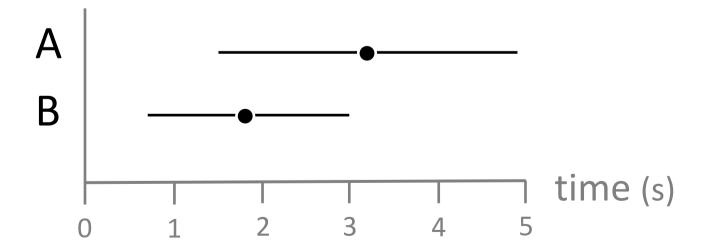
- Null hypothesis: H₀ = Os
 - For A, time is significantly different from 0s, p < .05
 - For B, time is significantly different from 0s, p < .05



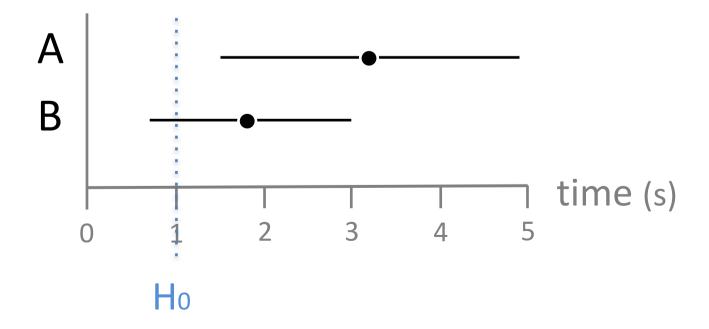
• Null hypothesis: H₀ = 1s



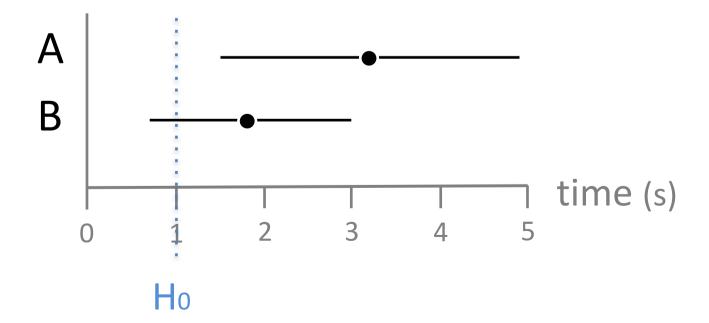
- Null hypothesis: H₀ = 1s
 - For A, time is significantly different from 1s, p < .05
 - For B, time is not sig. different from 1s, (p > .05)



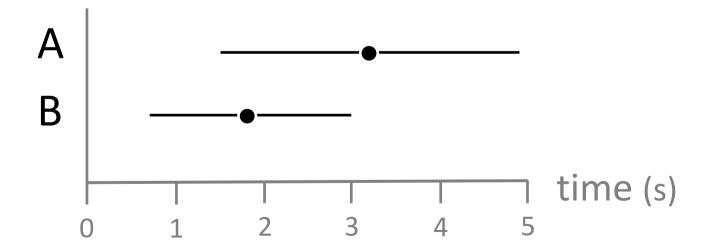
- Null hypothesis: H₀ = 1s
 - For A, time is significantly different from 1s, p < .05
 - For B, time is not sig. different from 1s, (p > .05)



« This is my least preferred way to interpret a
 CI: I earlier cited evidence that CIs can prompt
 better interpretation if NHST is avoided. »
 (Cumming and Finch, 2005)

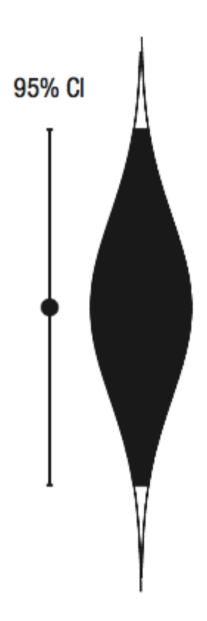


• « a range of plausible values for μ . Values outside the CI are relatively implausible. » (Cumming and Finch, 2005)

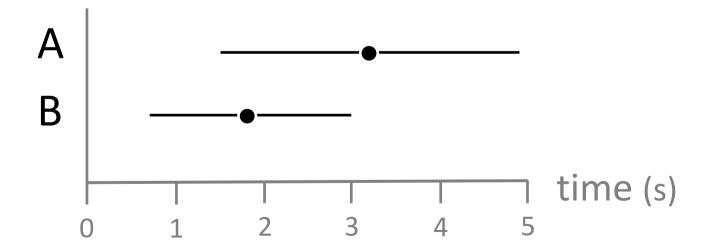


 "values close to our M are the best bet for μ, and values closer to the limits of our CI are successively less good bets."

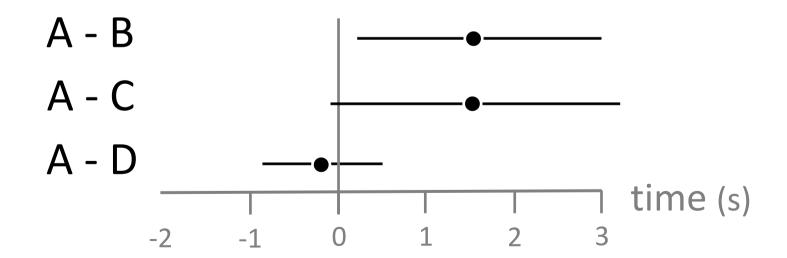
(Cumming, 2013)

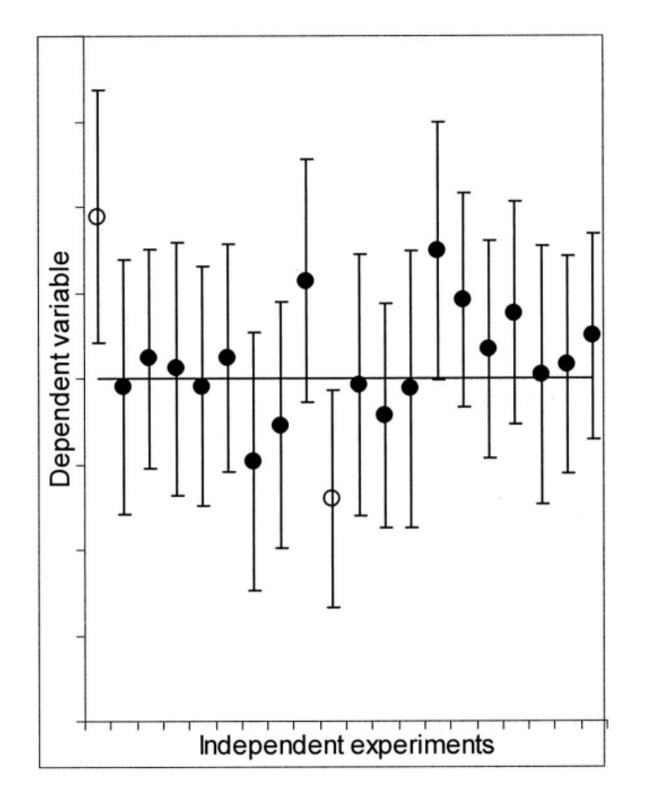


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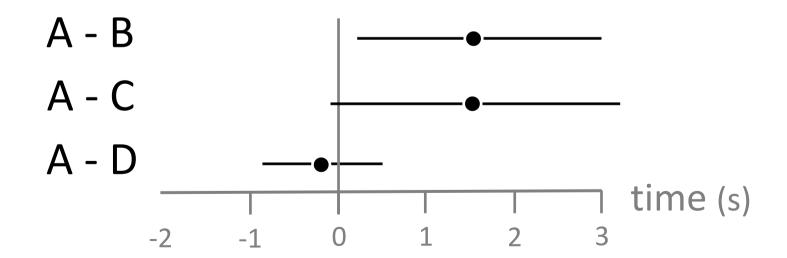
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Make sure you check <u>the dance</u>
 <u>of p-values</u> on youtube

• « a range of plausible values for μ . Values outside the CI are relatively implausible. » (Cumming and Finch, 2005)



"It seems clear that no confidence interval should be interpreted as a a significance test."

(Schmidt and Hunter, 1997)

Very hard!

- We believe that a user study should provide yes/no answers
- We believe that we need an objective procedure for deciding
- We've been brainwashed!

"It is best for individual researchers to present point estimates and confidence intervals and **refrain from attempting to draw final conclusions** about research hypotheses."

Schmidt and Hunter (1997)

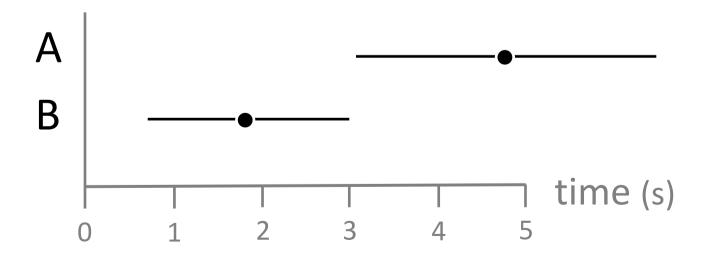
"We have the duty of [...] communicating our conclusions in intelligible form, in recognition of **the right of other free minds** to utilize them in **making their own decisions**."

Fisher (1955)

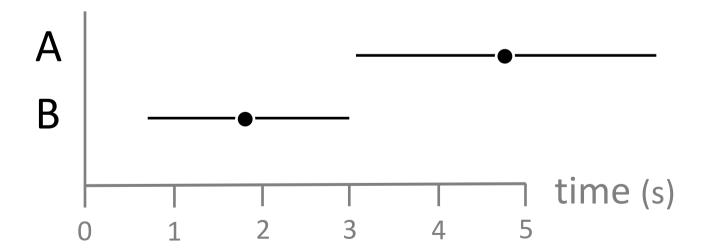
"[...] (Sciences) can only be successfully conducted by responsible and independent thinkers [...] The idea that this responsibility can be delegated to a giant computer programmed with Decision Functions belongs to the phantasy of circles rather remote from scientific research."

Fisher (1973), quoted by Smith et al. (2002)

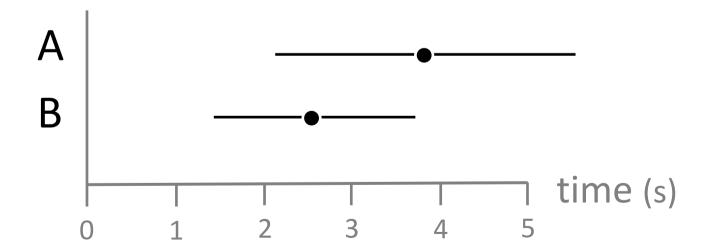
Overlap between Cls



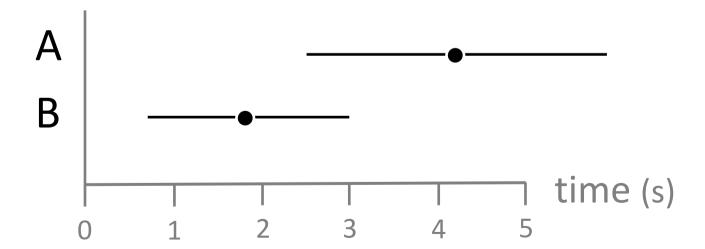
- Overlap between Cls
 - Case of between-subjects design
 - Is the difference statistically significant?



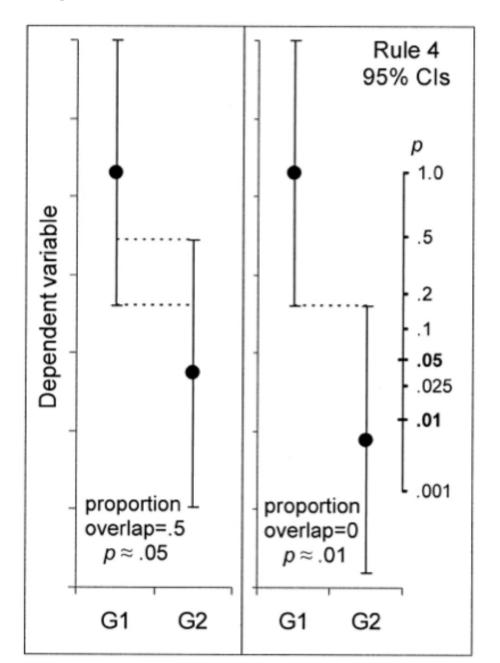
- Overlap between Cls
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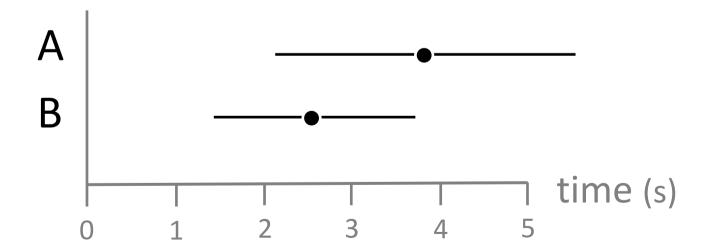
- Overlap between Cls
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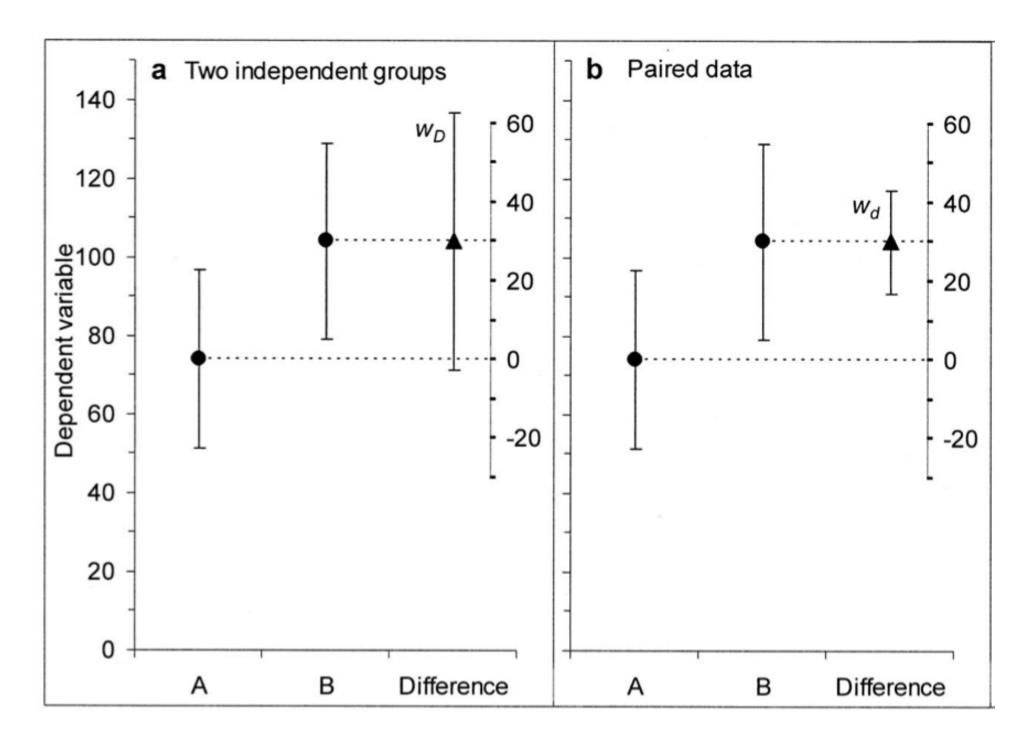


Overlap between Cls
 Cumming and Finch's
 Rule of Eye
 (Cumming and Finch, 2005)



- Overlap between Cls
 - Case of within-subject design
 - Is the difference statistically significant?





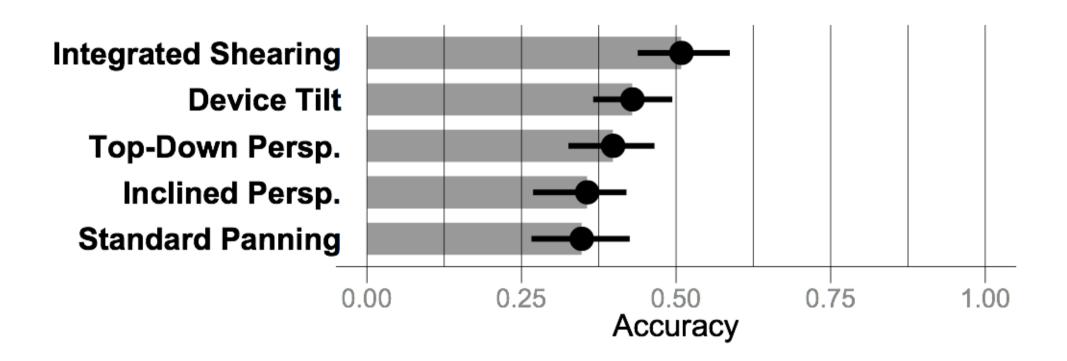
(Cumming and Finch, 2005)

- What if there are several factors/levels?
 - Inferential (or ANOVA) confidence intervals
 - Corrections for multiple comparisons
 - Complicated to compute AND to interpret

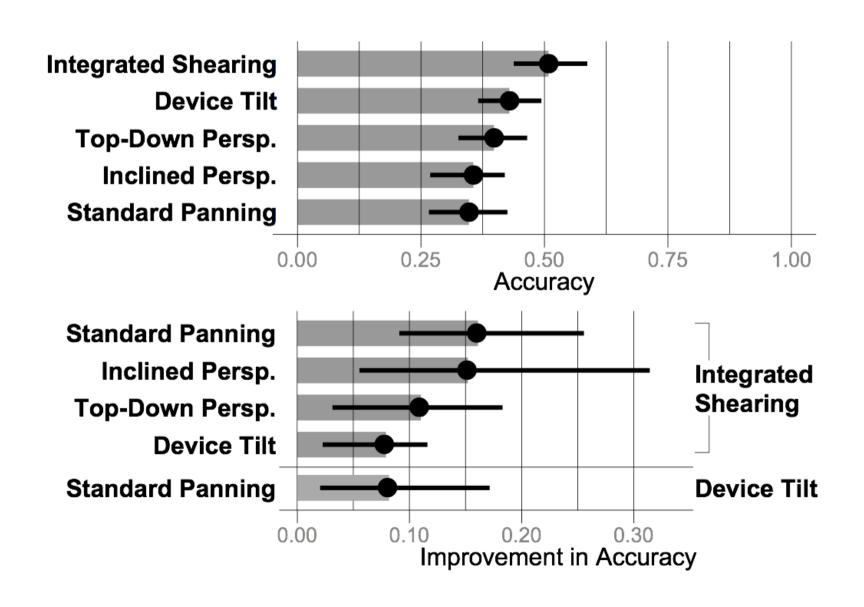
- What if there are several factors/levels?
 - Choose a simple experiment design
 - Pre-specify your research questions in advance
 - Only show and interpret the effects of interest
 - Don't correct for multiple comparisons
 - Do all your analyses on pilot data FIRST

Some Real Examples

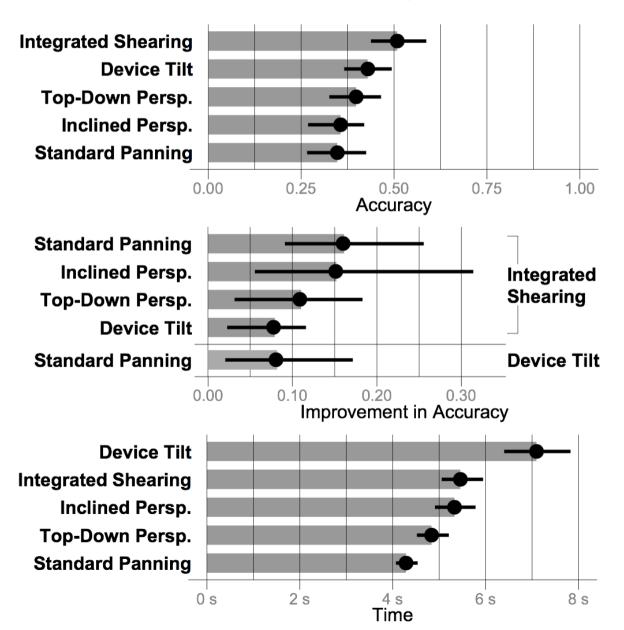
1 within-subject factor: *technique* (5 levels) 2 measures: *accuracy* and *time*



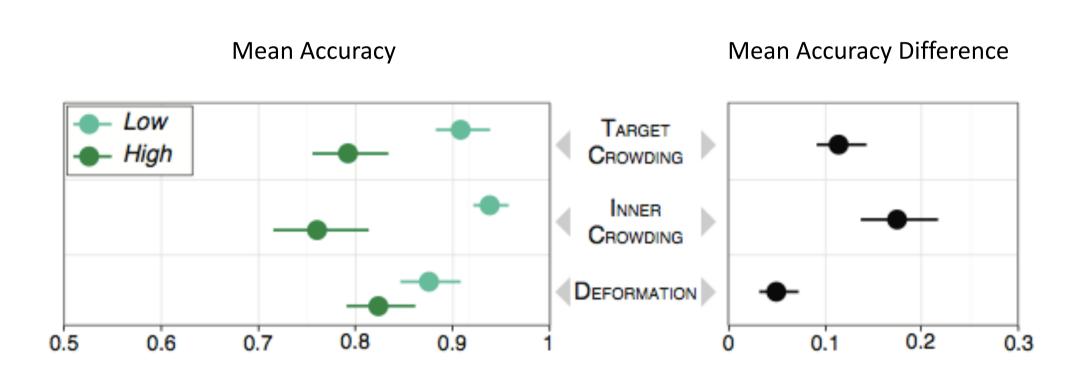
1 within-subject factor: *technique* (5 levels) 2 measures: *accuracy* and *time*



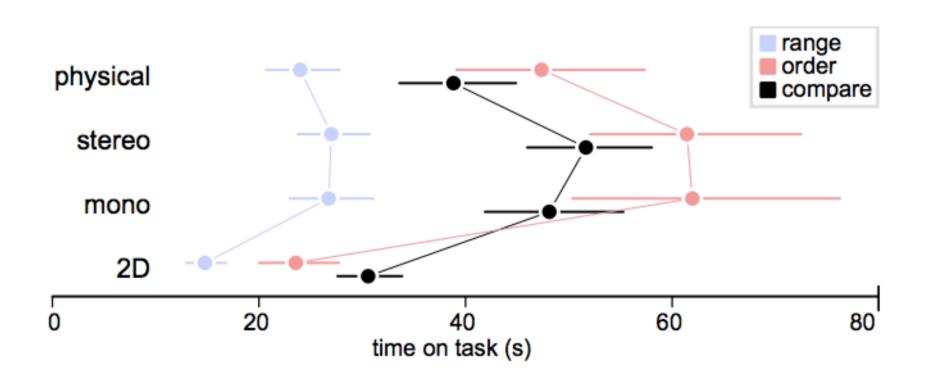
1 within-subject factor: *technique* (5 levels) 2 measures: *accuracy* and *time*



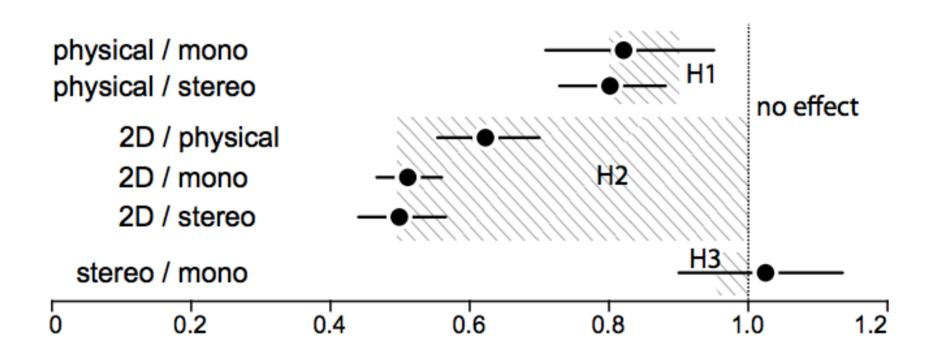
3 within-subject factors: target crowding, inner crowding, deformation (2 levels each) 1 measure: accuracy



2 within-subject factors: technique (4 levels) and task (3 levels) 1 measure: time

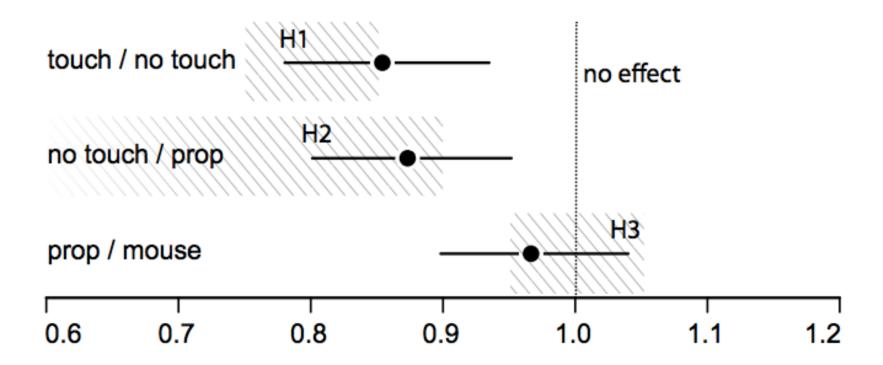


2 within-subject factors: technique (4 levels) and task (3 levels) 1 measure: time



1 within-subject factor: *technique* (4 levels) 1 measure: *time*

 $mouse \xrightarrow{rotation} prop \xrightarrow{realism} no \ touch \xrightarrow{+ \ touch} touch$



Will My Paper be Rejected?

- No (most likely)
 - If you don't over-interpret the patterns in your Cls
 - If you properly justify your approach

Due to growing concerns in various research fields over the limits of null hypothesis significance testing for reporting and interpreting experimental results [12], we base all our analyses and discussions on estimation, i.e., effect sizes with confidence intervals [13]. This approach also aligns with the latest recommendations from the APA [3].

To Go Further

Geoff Cumming

- Youtube channel
- Book: "The New Statistics"

Allen Downey

- Book "Think Stats Probability and Statistics for Programmers" (also a lecture)
- www.aviz.fr/badstats
 - Reading list on the p-value controversy
 - Examples of HCl papers without p-values