

Pere-Pau Vázquez

IDI – Quantitative and Qualitative Methods for Human-Subject Experiments

Outline

- Motivation
- Validity of experiments
- Experiment design
- Data analysis

Outline

- **Motivation**
- Validity of experiments
- Experiment design
- Data analysis

Motivation

- Measuring the response of humans to different experiments is the only way to:
 - Evaluate how humans perceive, manipulate, reason with applications or webpages
 - Measure utility of applications and webpages
- Key issue in software development
- It is important to do **before** launching any product!

Outline

- *Motivation*
- **Validity of experiments**
- Experiment design
- Data analysis

Validity of experiments

- Empirical method:
 - Develop a hypothesis, perhaps based on a theory
 - Make the hypothesis testable
 - Develop an empirical experiment
 - Collect and analyse data
 - Accept or refute the hypothesis
 - Relate the results back to the theory
 - If worthy, communicate the results to scientific community

Validity of experiments

- Experimental Validity
 - Does experiment really measure what we want it to measure?
 - Do our results really mean what we think (and hope) they mean?
 - Are our results reliable?
 - If we run the experiment again, will we get the same results?
 - Will others get the same results?

Validity of experiments

- Experimental variables:
 - Independent Variables
 - What the experiment is studying
 - Occur at different levels
 - Example: stereopsis, at the levels of stereo, mono
 - Systematically varied by experiment

Validity of experiments

- Experimental variables:
 - Dependent Variables
 - What the experiment measures
 - Assume dependent variables will be affected by independent variables
 - Must be measurable quantities
 - Time, task completion counts, error counts, survey answers, scores, etc.
 - Example: VR navigation performance, in total time

Validity of experiments

- Experimental variables:
 - Independent variables can vary in two ways
 - Between-subjects: each subject sees a different level of the variable
 - Example: 1/2 of subjects see stereo, 1/2 see mono
 - Within-subjects: each subject sees all levels of the variable
 - Example: each subject sees both stereo and mono

Validity of experiments

- Experimental variables:
 - Confounding factors (or confounding variables)
 - Factors that are not being studied, but will still affect experiment
 - Example: stereo condition less bright than mono condition
 - Important to predict and control confounding factors, or experimental validity will suffer
 - E. g.: Mono vs stereo and brightness

Outline

- *Motivation*
- *Validity of experiments*
- **Experiment design**
- Data analysis

Experiment design

- To avoid skewing effects, experiments must be designed carefully
 - E. g.: Learning a technique
 - After N repetitions of the same experiment, the user will go fast to solve the same problem
 - E. g.: Suffering fatigue
 - After N repetitions, if the task requires physical effort, the performance may suffer

Experiment design

- Counterbalancing design:
 - Avoid learning/fatigue effects by randomizing the tasks
 - Randomizing does not necessarily mean random, but sorting adequately users and conditions (systematic variation)

Experiment design

- Let's imagine we have 10 subjects and we want to test solving the same task (e. g. buying a book) using two different websites:

Subjects	First shopping	Second shopping
1, 3, 5, 7, 9	Website A	Website B
2, 4, 6, 8, 10	Website B	Website A

Experiment design

- Let's imagine we want to test solving the same task (e. g. buying a book) using three different devices (desktop, tablet, and mobile).
 - We will have the following conditions:

Device	Website	
	Website A	Website B
Smartphone		
Tablet		
Desktop		

Experiment design

- Say that we want each user to perform each task 4 times
 - We will have 3 (devices) x 2 (websites) x 4 (repetitions) = 24 tasks
 - Note that this grows with a factorial explosion!!!
- To ensure reliability, those tests must be performed in the adequate order
 - Different for each subject

Experiment design

- Latin squares (Not what you find on Wikipedia!):
 - Tabular expression of systematic variations
 - Can be used to adequately sort experimental tasks
 - Counterbalances to avoid confounding factors
 - Within-subjects variables: control fatigue and learning effects
 - Between-subjects variables: control other factors that change with time (e. g. network speed, cache contents)

Experiment design

- Latin squares. Properties:
 - Every level appears in the every position the same number of times
 - Every level is followed by every other level
 - Every level is preceded by every other level

Experiment design

- Latin squares. Examples:

2x2

1	2
2	1

6x3

1	2	3
2	3	1
3	1	2

1	3	2
2	1	3
3	2	2

4x4

1	2	3	4
2	4	1	3
3	1	4	2
4	3	2	1

Experiment design

- Studying the previous example (3 devices) x 2 websites with 4 repetitions:
 - Form a Cartesian product of latin squares:
 - 6x3 (devices) x 2x2 (conditions)
 - This will counterbalance properly a group of 12 subjects

Experiment design

Subject	Presentation order
1	1A, 1B, 2A, 2B, 3A, 3B
2	1B, 1A, 2B, 2A, 3B, 3A
3	2A, 2B, 3A, 3B, 1A, 1B
4	2B, 2A, 3B, 3A, 1B, 1A
5	3A, 3B, 1A, 1B, 2A, 2B
6	3B, 3A, 1B, 1A, 2B, 2A
7	1A, 1B, 3A, 3B, 2A, 2B
8	1B, 1A, 3B, 3A, 2B, 2A
9	2A, 2B, 1A, 1B, 3A, 3B
10	2B, 2A, 1B, 1A, 3B, 3A
11	3A, 3B, 2A, 2B, 1A, 1B
12	3B, 3A, 2B, 2A, 1B, 1A

Experiment design

- More examples in the PDF.

Outline

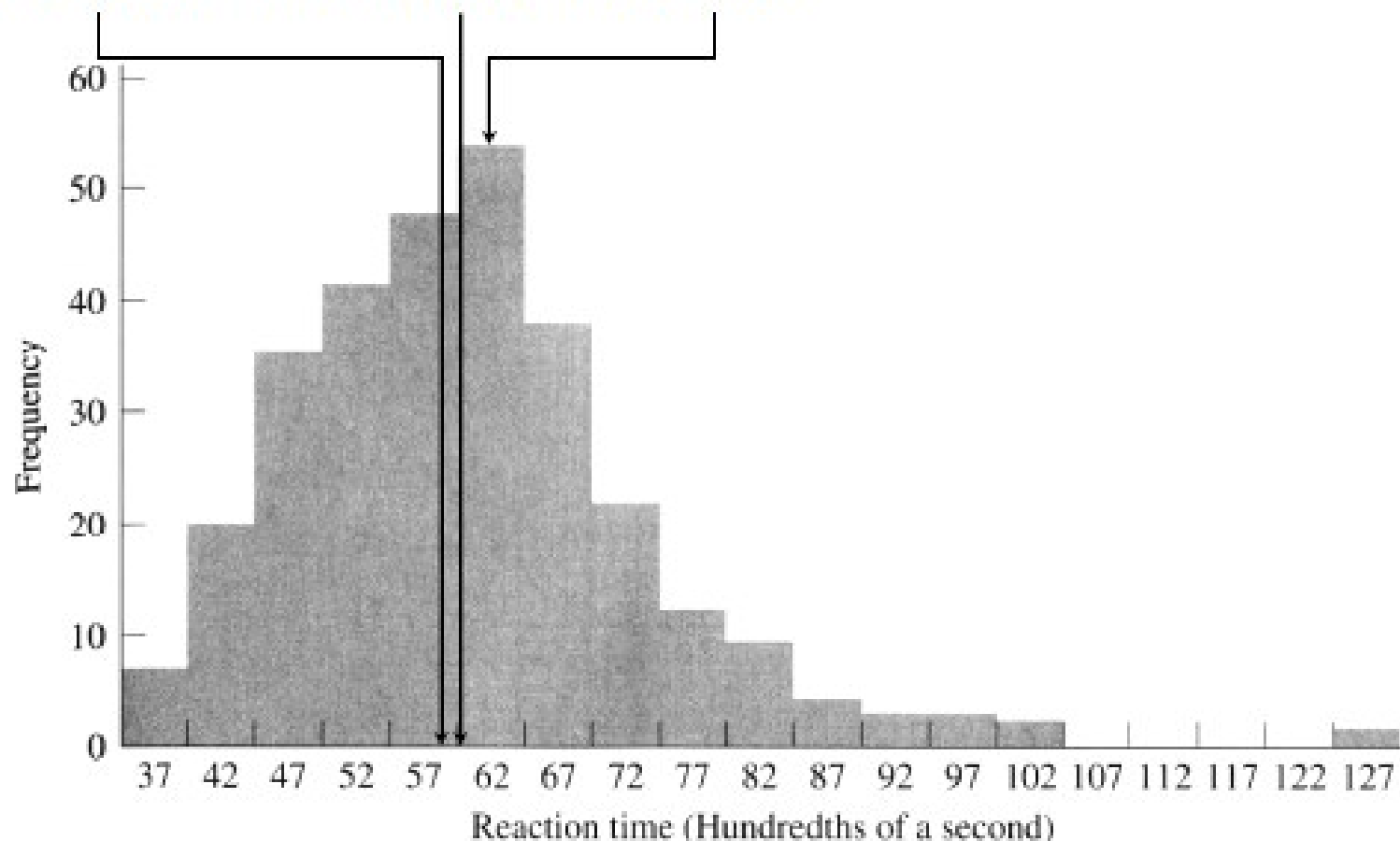
- *Motivation*
- *Validity of experiments*
- *Experiment design*
- **Data analysis**

Data analysis

- Descriptive statistics:
 - Describe and explore data
 - All types of graphs, histograms...
 - Understand data distribution
 - Start to think of significance tests
- Inferential statistics:
 - Detect relationships in data
 - Significance tests
 - Infer population characteristics from sample characteristics

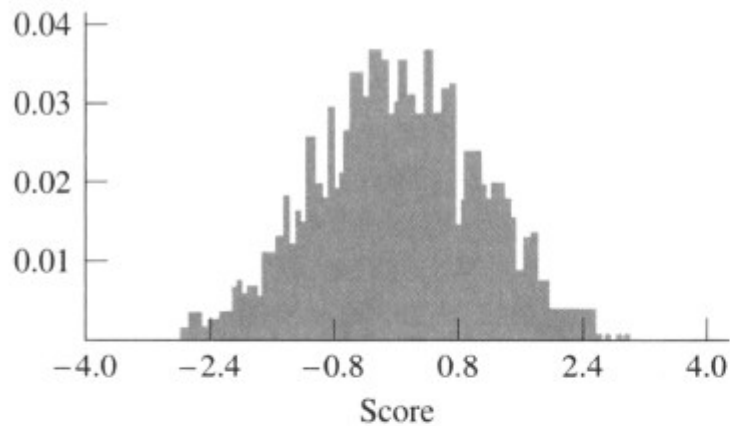
Data analysis

median = 59.5 mean = 60.26 mode = 62

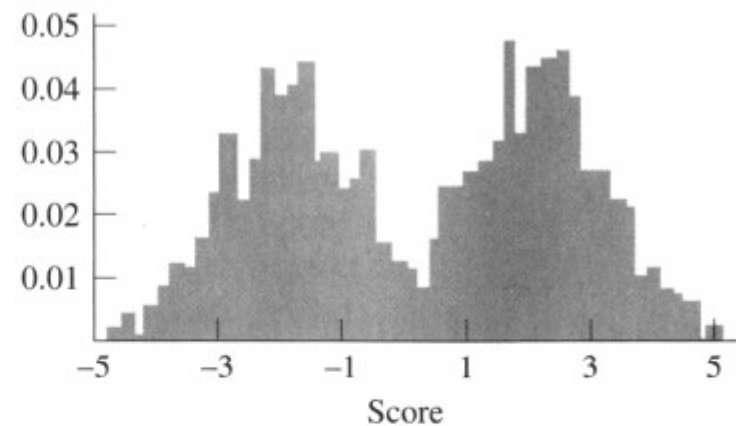


From [Howell 02] p 21

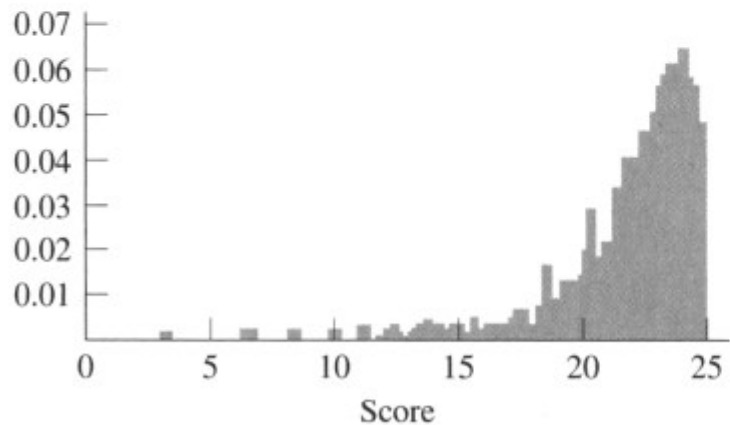
Data analysis



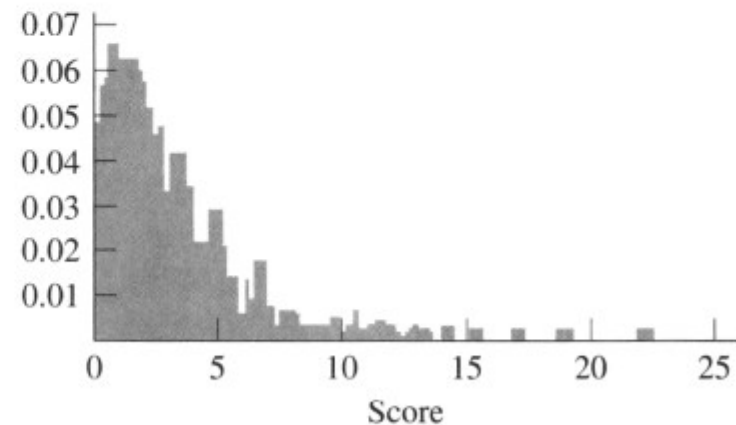
(a) Normal



(b) Bimodal



(c) Negatively skewed



(d) Positively skewed

Data analysis

From [Howell 02] p 21, 23

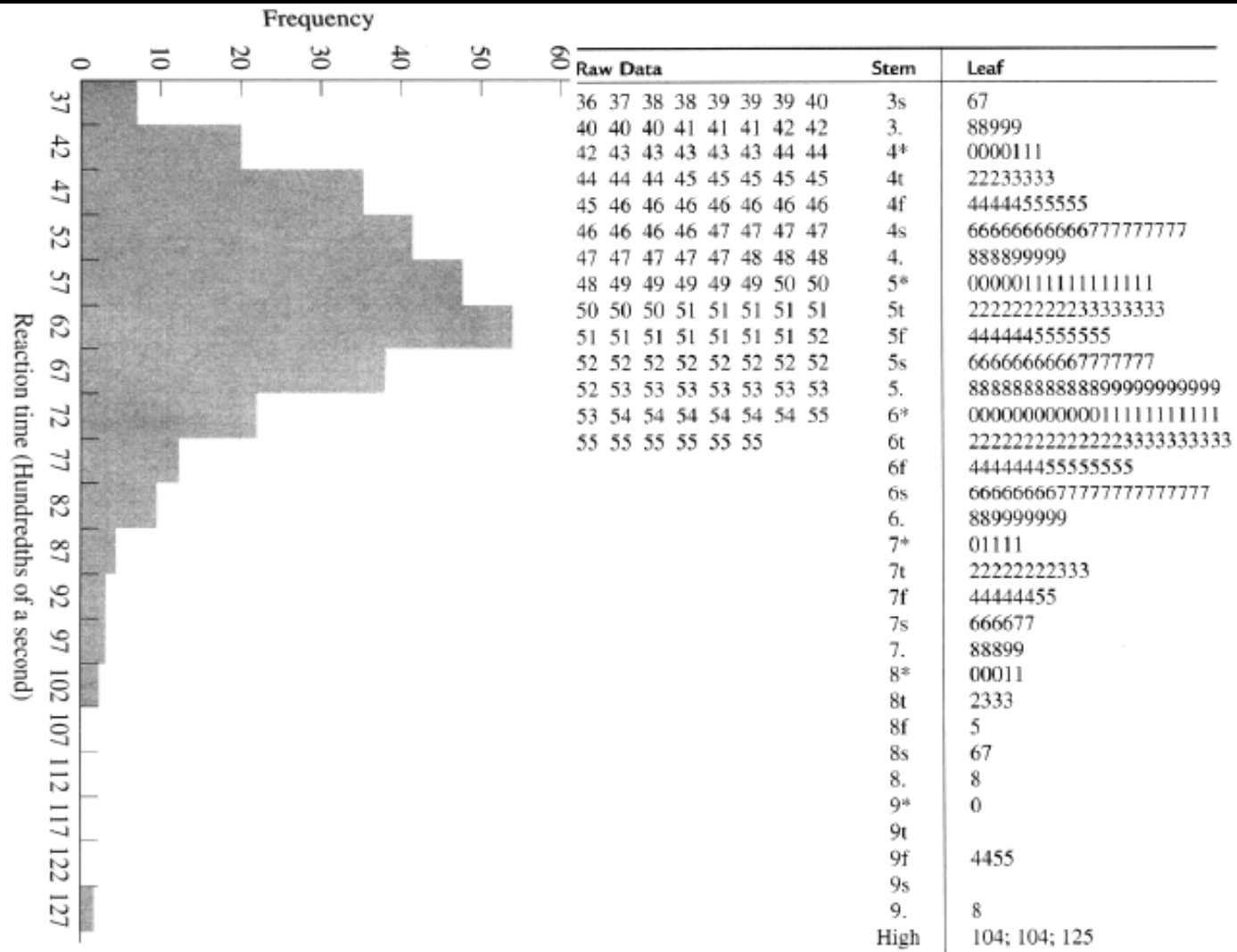
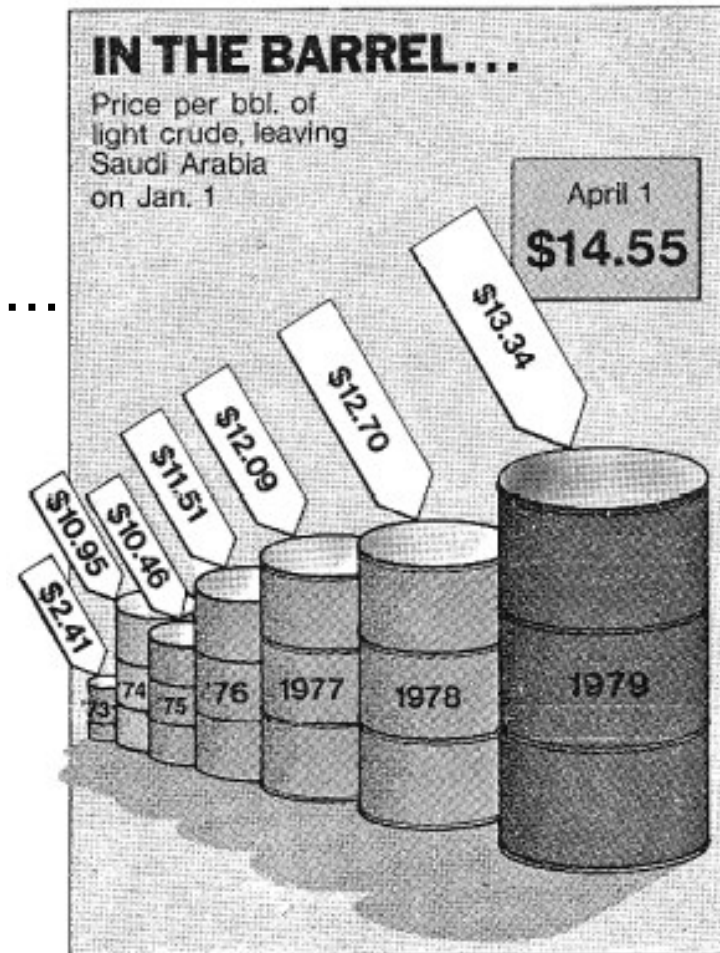


FIGURE 2.4 Stem-and-leaf display for reaction time data

Data analysis

- Not all examples are good
 - Be fair!!!
 - Tufte has plenty of examples...



Data analysis

- Reducing the numbers:
 - Central tendency:
 - Mean, median and mode
 - Variability/dispersion:
 - Mean absolute deviation, variance, standard deviation

Validity of experiments

Mean:

$$\bar{X} = \frac{\sum X}{N}$$

Mean absolute deviation:

$$\text{m.a.d.} = \frac{\sum |X - \bar{X}|}{N}$$

Variance:

$$s^2 = \frac{\sum (X - \bar{X})^2}{N - 1}$$

Standard deviation:

$$s = \sqrt{\frac{\sum (X - \bar{X})^2}{N - 1}}$$

$$\sigma^2 = \frac{\sum (X - \mu)^2}{N}$$

- Standard deviation uses same units as samples and mean.
- Calculation of population variance σ^2 is theoretical, because μ almost never known and the population size N would be very large (perhaps infinity).

Validity of experiments

- Hypothesis testing, analysis of variance
 - Read in the PDF, your notes of statistics previous courses...

Pere-Pau Vázquez

IDI – Quantitative and Qualitative Methods for Human-Subject Experiments