

# Power Analysis

Prof. Dr. Teena Hassan teena.hassan@h-brs.de

#### **Department of Computer Science**

Hochschule Bonn-Rhein-Sieg Sankt Augustin

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# Learning Goals







- At the end of today's lecture, you will be able to:
  - 1. Identify the factors that influence the **sample size** of participants required for an experiment.
  - 2. Apply **power analysis** to determine sample size for validating hypotheses in HRI experiments.



# Primary Reference







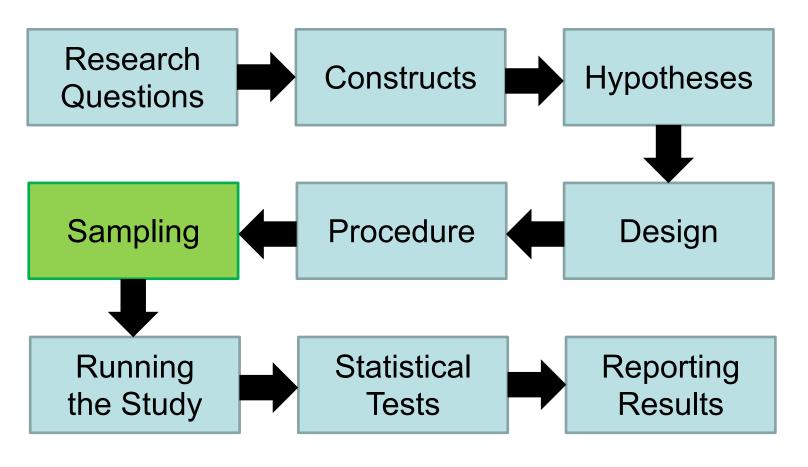
 Guy Hoffman and Xuan Zhao. 2020. A Primer for Conducting Experiments in Human Robot Interaction ACM Trans. Hum. Robot Interact . 10, 1, Article 6 (October 2020), 31 pages <a href="https://doi.org/10.1145/3412374">https://doi.org/10.1145/3412374</a>

Your reading assignment for last week!



### Hypothesis-Driven Experimental Research





(Hoffman & Zhao, 2020)



#### Make a Guess...







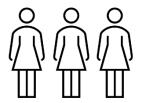
➤ Which design would require more samples to achieve the same statistical power?

A. Within-subjects design



B. Between-subjects design







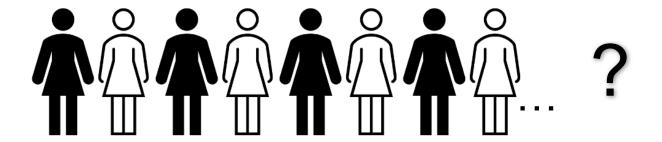
# How Many Samples are Enough?







- Very crucial question in empirical research.
- How many samples are needed to draw scientifically sound conclusions?







# Power Analysis

Simply put, a **statistically sound** method to determine **the sample size** and **power** of an experiment.







# Why is Power Analysis Important in HRI?







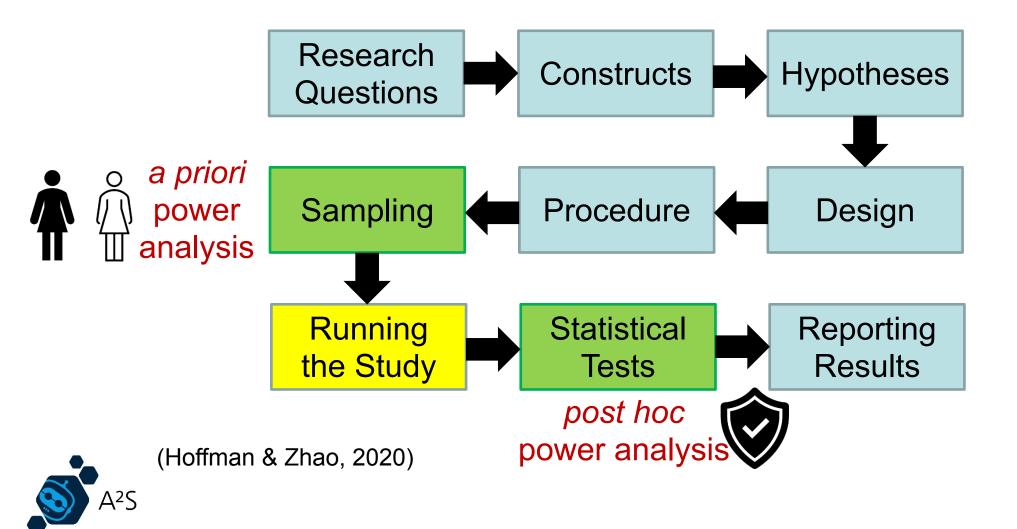
- Humans are unpredictable and exhibit variability in their behavior.
  - Inter- and intrapersonal variability
- When we conduct an experiment:
  - How can we be sure that the effects that we observed are meaningful and not noise?
- We should perform a priori and post hoc power analyses.

(Bartlett et al., 2022)



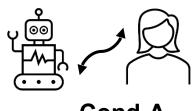
### Hypothesis-driven Experimental Research







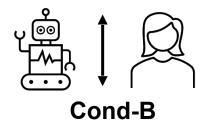
#### Level of users' trust in:





Interactive

#### Level of users' trust in:

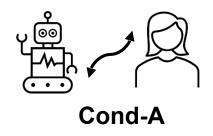


Straight Line

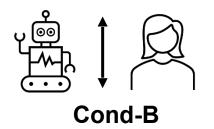




#### Level of users' trust in:







>H0 posits there is no difference between users' level of trust in cond-A and cond-B.

Null Hypothesis

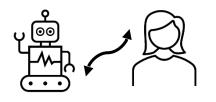


# Objective of the Experiment



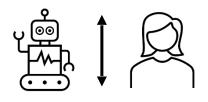
### Can we reject H0:

#### Level of users' trust in:





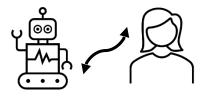
#### Level of users' trust in:



Cond-B

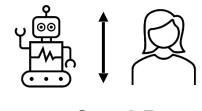
### • And accept H1:

#### Level of users' trust in:



Cond-A

#### Level of users' trust in:



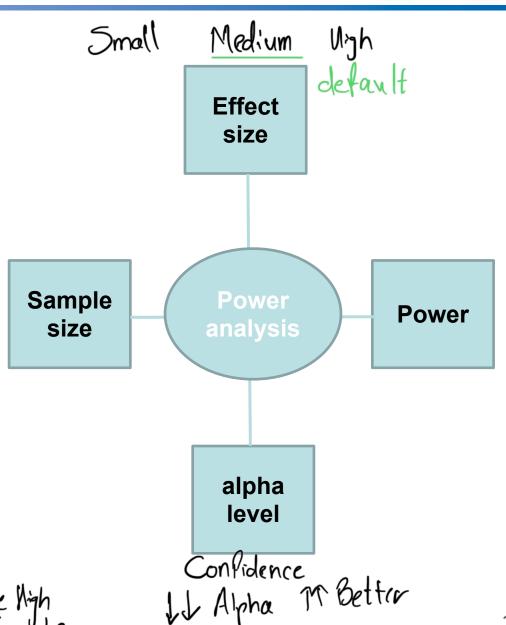
Cond-B



## Power Analysis

b-it O

- Examines the relationship between four parameters:
  - 1. alpha level
  - 2. Power
  - 3. Effect size
  - 4. Sample size
- Given any three of these parameters, we can predict the fourth.





Too small alpha contestigh Samples

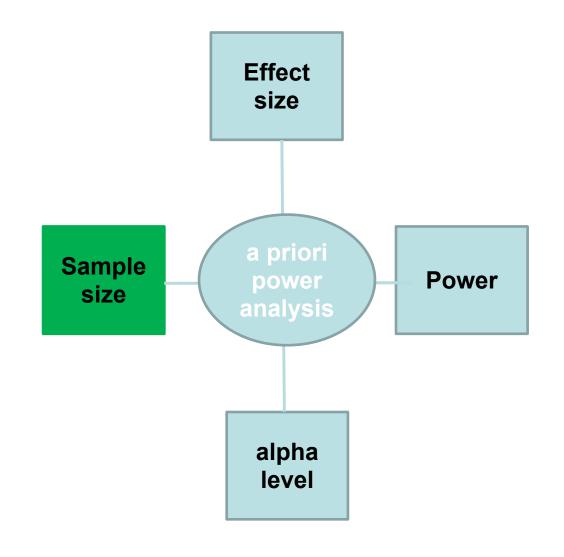
### a priori Power Analysis







- Given:
  - Expected or desired values for:
    - 1. alpha level
    - 2. Power
    - 3. Effect size
- Predict, sample size needed for the experiment.

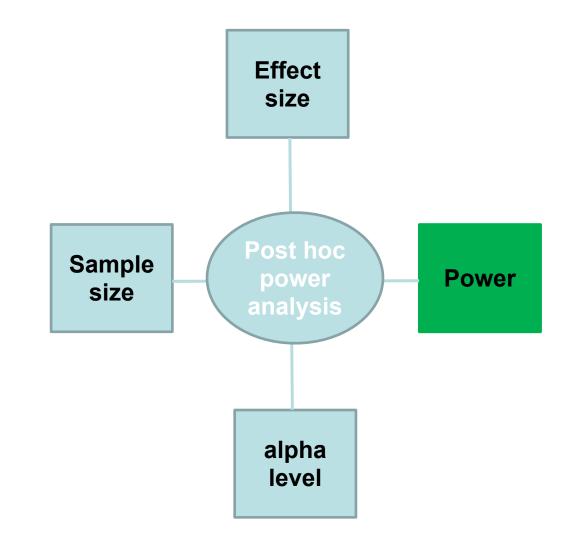




### post hoc Power Analysis

b-it O

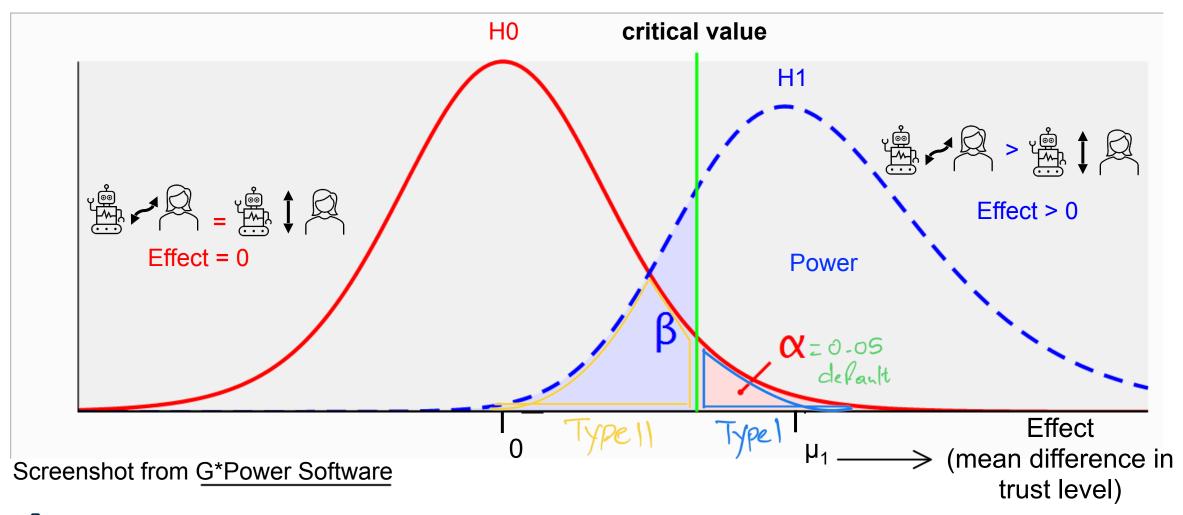
- Done after the experiment.
- Given:
  - Values computed from experimental data:
    - 1. Effect size
    - 2. Sample size
    - 3. alpha level
- Predict statistical power of the experiment.





### Alpha Level, Power, Effect Size







To get the same power after reducing alpha, we should increase the sample size.

### Errors Induced by Sampling – Type I, Type II

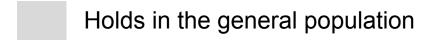






Test Results.

		Reject H0 Accept H1	Accept H0 Reject H1
	H0 is True H1 is False	α False positive rate (Type I error)	1- α True negative rate
r	H0 is False H1 is True	1- β True positive rate (POWER)	β False negative rate (Type II error)





Holds in the sampled population

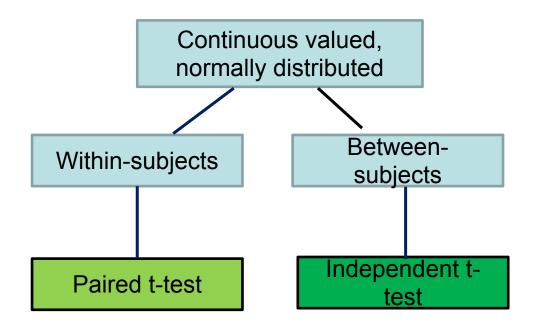
# **Estimating Effect Size**







- Effect size is the difference between outcomes measured in different conditions.
- ➤ Effect size computation depends on the method chosen for the statistical test.





# Computing Effects Using t-tests







Independent t-test

Paired t-test

- Between subject
  - Two separate groups of subjects A and B
  - Each group exposed to only one condition
  - Two sampling distributions
- Computes means of dependent variable in either distribution: μ<sub>A</sub>, μ<sub>B</sub>
- Difference between means:  $\Delta = \mu_A \mu_B$
- Effect: Compares Δ to zero.
- Effect size (Cohen's d): Δ / σpooled

- Within subject
  - One group of subjects
  - Each subject exposed to both conditions
     A and B
- One sampling distribution S based on difference  $\Delta_s$  in response of each subject s in condition A and condition B.

$$\Delta_s = a_s - b_s$$

- $\mu_S$ = Mean  $\Delta_s$  over all s
- Effect: Compares μ<sub>S</sub> to zero.
- Effect size (Cohen's d):  $\mu_S / \sigma_S$



$$\sigma_{pooled} = \sqrt{(\sum (a - \mu_A)^2 + \sum (b - \mu_B)^2) / (n_A + n_B - 2)}$$

# a priori Power Analysis







- Commonly used values:
  - alpha: 0.05 or 0.01
  - power: 0.8 (HRI), 0.95 (clinical)
  - effect-size (for t-tests): 0.5

	Small effect size	Medium effect size	Large effect size
Cohen's d	0.2	0.5	8.0



Jacob Cohen. 1992. A power primer. Psychol. Bull. 112, 1 (1992).

#### Conclusion







- In this lecture, you learned to:
  - 1. Identify the factors that influence the **sample size** of participants required for an experiment.
  - 2. Apply **power analysis** to determine sample size for validating hypotheses in HRI experiments.



#### References







- Guy Hoffman and Xuan Zhao. 2020. A Primer for Conducting Experiments in Human–Robot Interaction. ACM Trans. Hum.-Robot Interact. 10, 1, Article 6 (October 2020), 31 pages. <a href="https://doi.org/10.1145/3412374">https://doi.org/10.1145/3412374</a>
- Madeleine E. Bartlett, C. E. R. Edmunds, Tony Belpaeme, and Serge Thill. 2022.
   Have I Got the Power? Analysing and Reporting Statistical Power in HRI. J. Hum.-Robot Interact. 11, 2, Article 16 (June 2022), 16 pages. <a href="https://doi.org/10.1145/3495246">https://doi.org/10.1145/3495246</a>



#### **Next Week**







- Lecture on "Statistical Tests"
  - Thursday, 4.07.2024 at 9 am in B060, Sankt Augustin

