Biostatistics BootCamp Lecture 3:

Understanding of the Statistical Power and Sample Size in Protocol Development

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Lecture 1. Uncertainty & Hypothesis Tests:

- An ideal approach for a scientific research is to investigate the total population.
 - Since it is not usually feasible, we do not have population characteristics.
 - ➡ The need for Statistics.
- Statistics uses sample data to determine population parameters. ➡ Statistical inference and uncertainty.
- We need to perform the hypothesis testing procedure to quantify the uncertainty.



Open learning Initiative at Carnegie Mellon University, https://oli.cmu.edu/courses/concepts-of-statistics/

• A good design through hypothesis tests is related to the statistical uncertainty, the two types of errors, α and β .

Lecture 2. Control of α up to 5%

• When do we encounter the multiplicity?

When multiplicity is present, the usual statistical approach may necessitate an adjustment to the type 1 error. Multiplicity may arise, for example, from multiple primary variables, multiple comparisons of treatments, repeated evaluation over time, and/or interim analyses – ICH E9

• The multiplicity (multiple tests) leads to the inflation of type 1 error rate.

→ Inflated type 1 error (α): α = 0.098 for k=2, α = 0.19 for k=4.

- How can we overcome the multiplicity?
 - **\Rightarrow** Bonferroni method: α/k
 - Other methods like Holm method, Hochberg method : α/k



Power 1: What's Power? graphical understanding



Source: Statistical power from Towards data science (AY Yao)



Power 2: what does it mean?

- In hypothesis tests, the <u>H₀ always means that no scientific findings exist</u> and the <u>H₁ says that the study group shows better over the control group</u>.
 - The investigators/researchers typically hope to prove the alternative hypothesis. That is, they hope to reject the null hypothesis (with p<0.05).</p>
- The power of a test tells us how likely we are to find a significant difference given that the H_1 is true \Rightarrow The true mean (μ) is different from μ_0 .
- If the power is too low, then we have little chance of finding a significant difference even if the true mean is not equal to $\mu_0 \Rightarrow$ The detection power
- During the planning stage, we need how many subjects are needed in order to have a desired power of detecting a clinical meaningful difference.

Power 3: Elements affecting the power

- The sample size (N): As N increases, the power increases.
- The effect size (ES): As the ES increases, the power also increases with a fixed sample size
 Clinically/scientifically meaningful mean difference between control and treatment groups.
- Two types of errors (α and β):
 - As β increases, the power decreases (power = 1β)
- Precision of the measurements:
 - As the standard deviation of the measurement decreases, the power increases.



Sample Size (N) & ES .vs. Power (1- β)







Error (α) and Precision (σ) .vs. Power (1- β)





Sample size: General formula

$$n_i = 2 \left(\frac{Z_{1-\alpha/2} + Z_{1-\beta}}{ES} \right)^2$$

- This is a sample size formula for two independent samples with continuous outcome where n_i is the sample size for each group, α and β are two types of errors, and ES is the effect size.
- Hypothesis: $H_0(\mu_1 = \mu_1)$ versus $H_a(\mu_1 \neq \mu_1)$
- As the sample size increases, the power increases.



Effect size 1: What it is?

- The "effect size (ES)" means the degree to which the phenomenon is present in the population
 - Clinically meaningful difference in means, proportions, and odds ratios of the primary endpoint.
 - The larger this value, the greater the degree to which the phenomenon under study is manifested.
- The effects sizes are determined by the study design: primary endpoints (PEs), study purpose and design, and repeated measurements.
- In hypothesis testing, the H_0 always means that the effect size is zero.
- As the effect size increases ➡ The power also increases with a fixed sample size.

Effect size 2: α, β, and ES



Figure 1. Distribution under the null hypothesis (blue), distribution under the alternative hypothesis (red), and the blue area: 2.5% of alpha level. Figure 2. Distribution under the null hypothesis (blue), distribution under the alternative hypothesis (red), and the red area: beta level.



Source: Verhulst B (2022). Hypothesis Testing, effect size and statistical power.





Effect size 3: Cohen's d

- Be careful with interpreting Effect Sizes. How can we interpret the ES?
- Cohen's (1988) Standards for Interpreting Effect sizes
 - Cohens d is a <u>standardized effect size</u> for measuring the difference between two group means.

• How to calculate?
$$d = \frac{\bar{X}_1 - \bar{X}_2}{S_p}$$

- Cohen's d interpretation:
 - ➡ d=0.2 ES is small
 - ➡ d=0.5 ES is medium
 - ➡ d=0.8 ES is large





Effect size 4: Examples

A retrospective study was performed to test whether a new biomarker is more effective to cognitive impairment over an existing biomarker. Data were extracted from the existing source (EMR or clinical charts). PE (test scores) were measured and compared between two groups (higher score is the better). What is an ES for this study?

A pilot study was performed to obtain preliminary data for a promising treatment on PD. Investigators could have at most 35 PD patients from the clinics. The PE would be measured at baseline and remeasured in 6 month after the treatment. This data will be used for a formal CT. What is an ES for this study?

A clinical trial was performed to test whether an experimental drug would show strong efficacy over the existing standard drug treatment. This was a phase 2 two-arm trial in Alzheimer's disease (AD). Patients would be randomized after registration into either group (treatment and control group). The PE of a continuous biomarker would be repeatedly measured at baseline and in 6 month after the treatment. What is an ES for this study?

Software:

G*Power: a tool to compute statistical power analyses for many different *t* tests, *F* tests, χ2 tests, *z* tests and some exact tests. G*Power can also be used to compute effect sizes and to display graphically the results of power
 ➡ Manual and tutorial.

➡ Download: <u>G*Power 3.1.9.7 for Windows</u> and <u>G*Power 3.1.9.6 for Mac OS X</u>

- R and Python: a programming language for statistical computing and graphics.
 → Manual and tutorial.
- GraphPad: It provides sample size calculation with power analysis. Introductory video is at <u>https://www.graphpad.com/series/how-to-calculate-sample-size-using-power-analysis</u>.
- <u>Others</u>: SAS, STATA, SPSS, etc



G*Power:

🚡 G*Power 3.1.9.2			—		×
File Edit View Tests Calculato	r Help				
Central and noncentral distribution	Protocol of po	wer analyses			
Test family Statistical test					
t tests V Correlation: Point biserial model V					
Type of power analysis					
A priori: Compute required sample size – given $\alpha,$ power, and effect size $\qquad \lor$					
Input Parameters		Output Parameters			
Tail(s)	One 🗸	Noncentrality parameter δ			?
Determine => Effect size p	0.3	Critical t			?
α err prob	0.05	Df			?
Power (1-β err prob)	0.95	Total sample size			?
		Actual power			?
		X-Y plot for a range of values	- F	Calcul	ate



Practical example 1:

Let us assume that researchers are planning a study to investigate the analgesic efficacy of 2 drugs. Drug A has traditionally been used for postoperative pain control and drug B is a newly developed drug. A pain score using a Visual Analog Scale (VAS) will be measured at 6 hours post-operation. The researchers want to determine the sample size for the null hypothesis to be rejected with a 2-tailed test, α =0.05, and β =0.2.

- What is the primary endpoint?
- What analysis method should be applied?
- What are two errors?
- How can we determine the effect
 size? ➡ Three cases ...





- [Step 1] Analysis method? From the top menu, click ... Tests Means Two independent groups
- [Step 2] Input parameters from "Determine=>": Two tails, ES=0.5, alpha=0.05, power=0.8, ratio=1
- [Step 3] Click "Calculate". Then you can get the left window result.



Practical example 2:

Assume that a study <u>examined the effects of 2 treatments</u>, for which the measure of the effect is a **proportion**. Treatment A has been traditionally used for the prevention of post-herpetic neuralgia, and treatment B is a newly developed treatment. The researchers wanted to **determine the sample size** for the null hypothesis to be rejected using 2-tailed testing, α =0.05, and β =0.2. The number of patients was equal in both groups. Suppose the researchers conducted a pilot study examining the effects of treatments A and B. In the pilot study, the proportions of post-herpetic neuralgia development were 0.3 and 0.1, respectively.

- What is the primary endpoint?
- What analysis method should be applied?
- What are two errors?
- How can we determine the effect size? → Three cases ...



• [Step 1] Analysis method? From the top menu, click ... Tests - Proportions – Two independent groups: inequality, z-test.

[Step 2] Input parameters from "Determine=>": Two tails, p2=0.3, p1=0.1, alpha=0.05, power=0.8, ratio=1
[Step 3] Click "Calculate". Then you can get the left window result.







Button KS et al. (2013): Power failure due to small sample size

nature reviews neuroscience

ANALYSIS

Power failure: why small sample size undermines the reliability of neuroscience

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