

MODULE 4: STUDY DESIGN SAMPLE SIZE JUSTIFICATION

Bios6624: Advanced Statistical Methods and Analysis
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SAMPLE SIZE JUSTIFICATION

- The most common reason to contact a statistician
- Sample size is contingent on design, analysis plan and outcome
- With the wrong sample size
 - May not be able to make conclusions because the study is ‘underpowered’ i.e. Is result non- significant because treatment doesn’t work OR is sample size too small to detect difference
 - Waste time and money because your study is larger than required to answer the question of interest –may find statistical significance for a difference that is not clinically relevant

REVIEW: STATISTICAL DECISION MAKING

		True Difference	
		Present (H_a)	Absent (H_0)
Conclusion of Statistical Test	Significant	Correct (Power)	Type I (α) error
	Not Significant	Type II (β) error	Correct

POWER IS DEPENDENT ON

- Study design
- Distribution of data
- Type-I error rate (α)
- Two-sided test (usual) or One-sided Test
- Size of difference to be detected
- Variability of data
- Sample size

LET'S LOOK AT POWER FOR ONE SAMPLE Z-TEST

- You have continuous data as your outcome.
- You are interested in testing of the mean of the data differs from zero (Null hypothesis).
- You decided a priori the Z-test is the correct statistical analysis (you have a larger sample size).
- We want to know what the power of this test is for different alternative hypotheses.
- Recall, $\text{power} = p(\text{reject the null hypothesis} \mid \text{under the alternative})$
- So, power is the probability you reject the null when you should

POWER FOR ONE SAMPLE T-TEST

$$P\left(|Z^*| > Z_{1-\alpha/2} \mid \mu = \mu_{alt}, \sigma_{\bar{X}}^2\right)$$

$$Z^* = \frac{(\bar{X} - 0)}{\sqrt{\sigma_{\bar{X}}^2/n}}$$

Under the alternative: $\bar{X} \sim N(\mu_{alt}, \sigma_{\bar{X}}^2/n)$

$$P\left(\bar{X} > \sqrt{\sigma_{\bar{X}}^2/n} * Z_{1-\alpha/2} \mid \mu = \mu_{alt}, \sigma_{\bar{X}}^2\right) + P\left(\bar{X} < -\sqrt{\sigma_{\bar{X}}^2/n} * Z_{1-\alpha/2} \mid \mu = \mu_{alt}, \sigma_{\bar{X}}^2\right)$$

$$P\left(\bar{X} - \mu_{alt} > \sqrt{\frac{\sigma_{\bar{X}}^2}{n}} * Z_{1-\alpha/2} - \mu_{alt} \mid \mu = \mu_{alt}, \sigma_{\bar{X}}^2\right) + P\left(\bar{X} - \mu_{alt} < -\sqrt{\frac{\sigma_{\bar{X}}^2}{n}} * Z_{1-\alpha/2} - \mu_{alt} \mid \mu = \mu_{alt}, \sigma_{\bar{X}}^2\right)$$

$$P\left(\frac{\bar{X} - \mu_{alt}}{\sqrt{\frac{\sigma_{\bar{X}}^2}{n}}} > Z_{1-\alpha/2} - \frac{\mu_{alt}}{\sqrt{\frac{\sigma_{\bar{X}}^2}{n}}} \mid \mu = \mu_{alt}, \sigma_{\bar{X}}^2\right) + P\left(\frac{\bar{X} - \mu_{alt}}{\sqrt{\frac{\sigma_{\bar{X}}^2}{n}}} > -Z_{1-\alpha/2} - \frac{\mu_{alt}}{\sqrt{\frac{\sigma_{\bar{X}}^2}{n}}} \mid \mu = \mu_{alt}, \sigma_{\bar{X}}^2\right)$$

Which are normal probabilities we can compute.

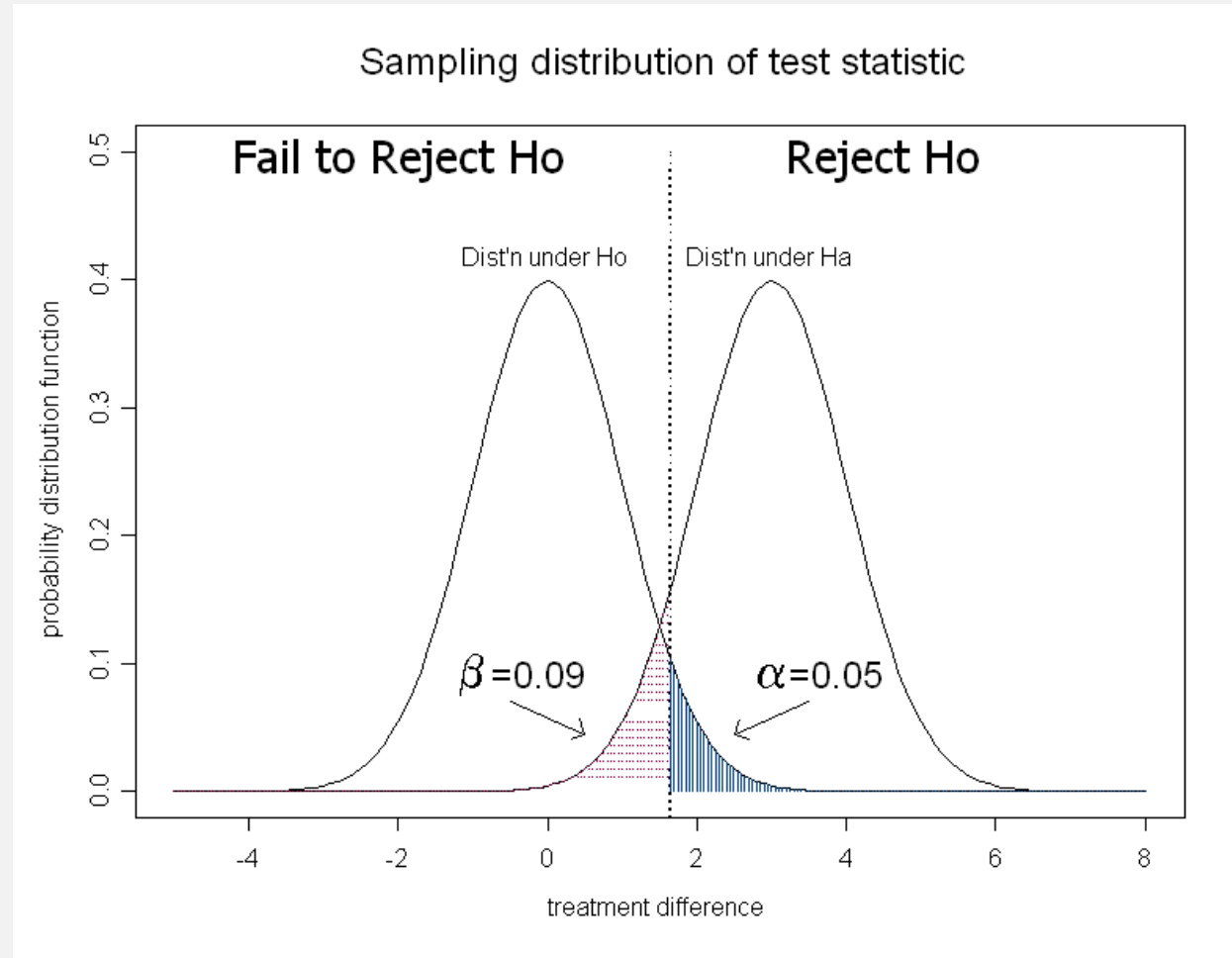
PICTURE OF POWER FOR SAMPLING DISTRIBUTIONS UNDER H_0 AND H_A

ASSUMES NORMAL DISTRIBUTION OR LARGE SAMPLE, ONE TAIL TEST OF MEAN

$H_0: N(0, 1)$

$H_A: N(3, 1)$

Beta = 1 - power
= Type II error

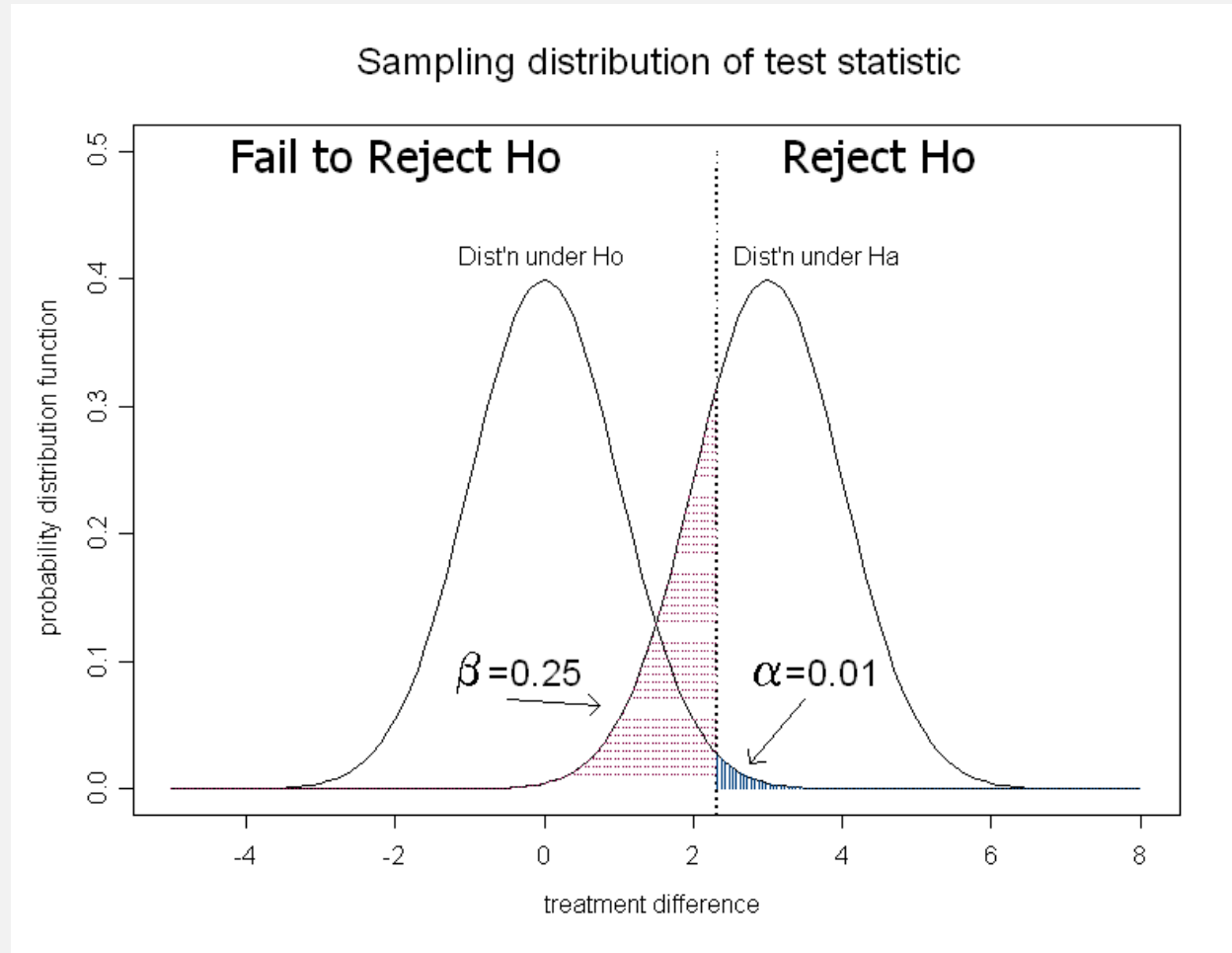


Power: $\Pr(\text{reject null hypothesis} \mid \text{alternative hypothesis is true})$

POWER INCREASES AS:

- Type-I error rate increases
 - Often fix α at 0.05 (other levels may be appropriate)
- Difference between H_0 and H_A increases (represented by a mean farther from zero in our example)
- Standard error decreases
 - Increase sample size
 - Decrease between subject variability

TRADE-OFF BETWEEN α AND β



POWER INCREASES AS:

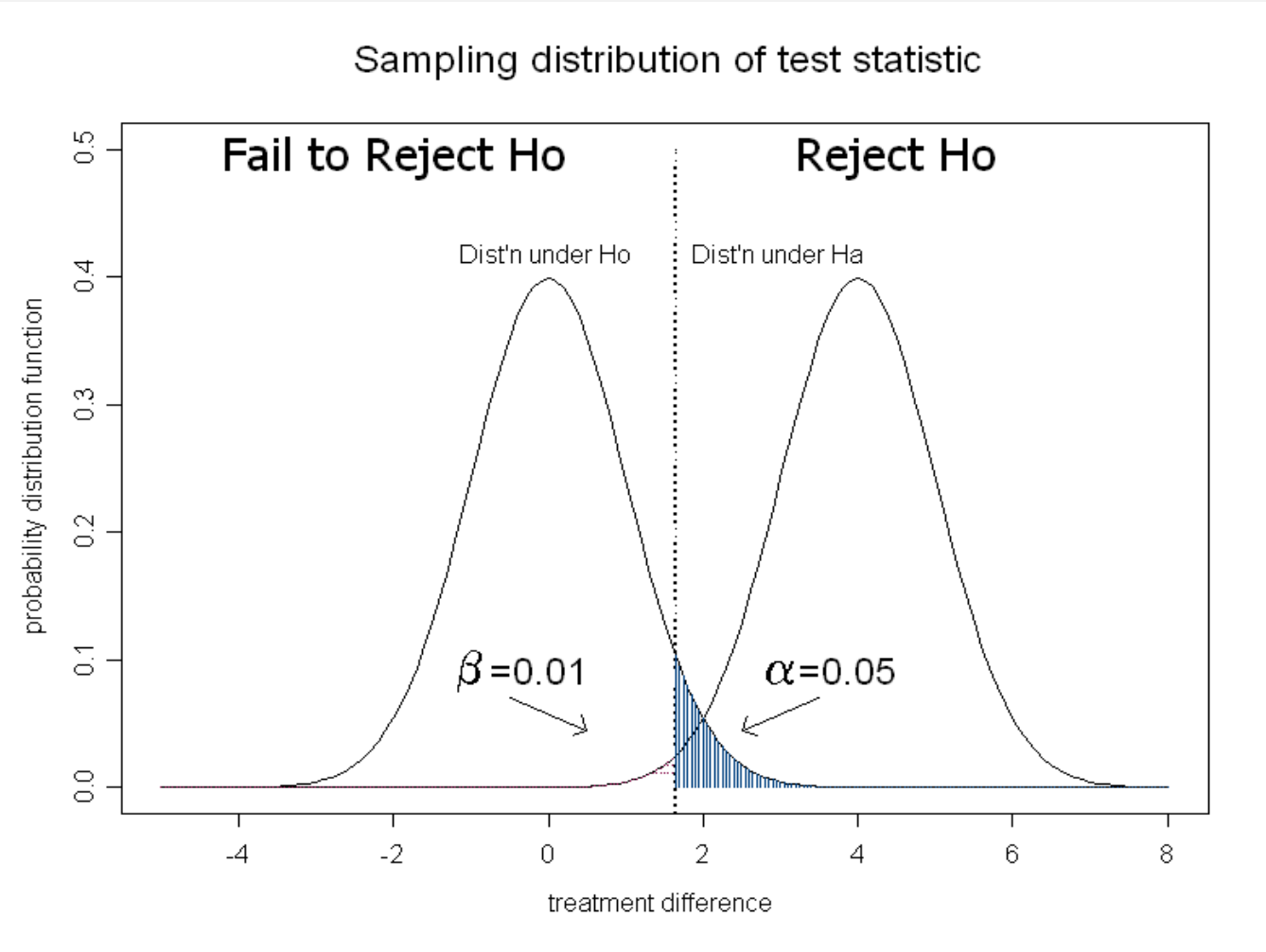
- Type-I error rate increases
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- Standard error decreases
 - Increase sample size
 - Decrease between subject variability
 - Decrease variability of assay (within subject)

POWER INCREASES AS DIFFERENCE BETWEEN H_0 AND H_A INCREASES

$H_0: \mu=0$

$H_A: \mu=4$ here

Much smaller red region.



POWER INCREASES AS:

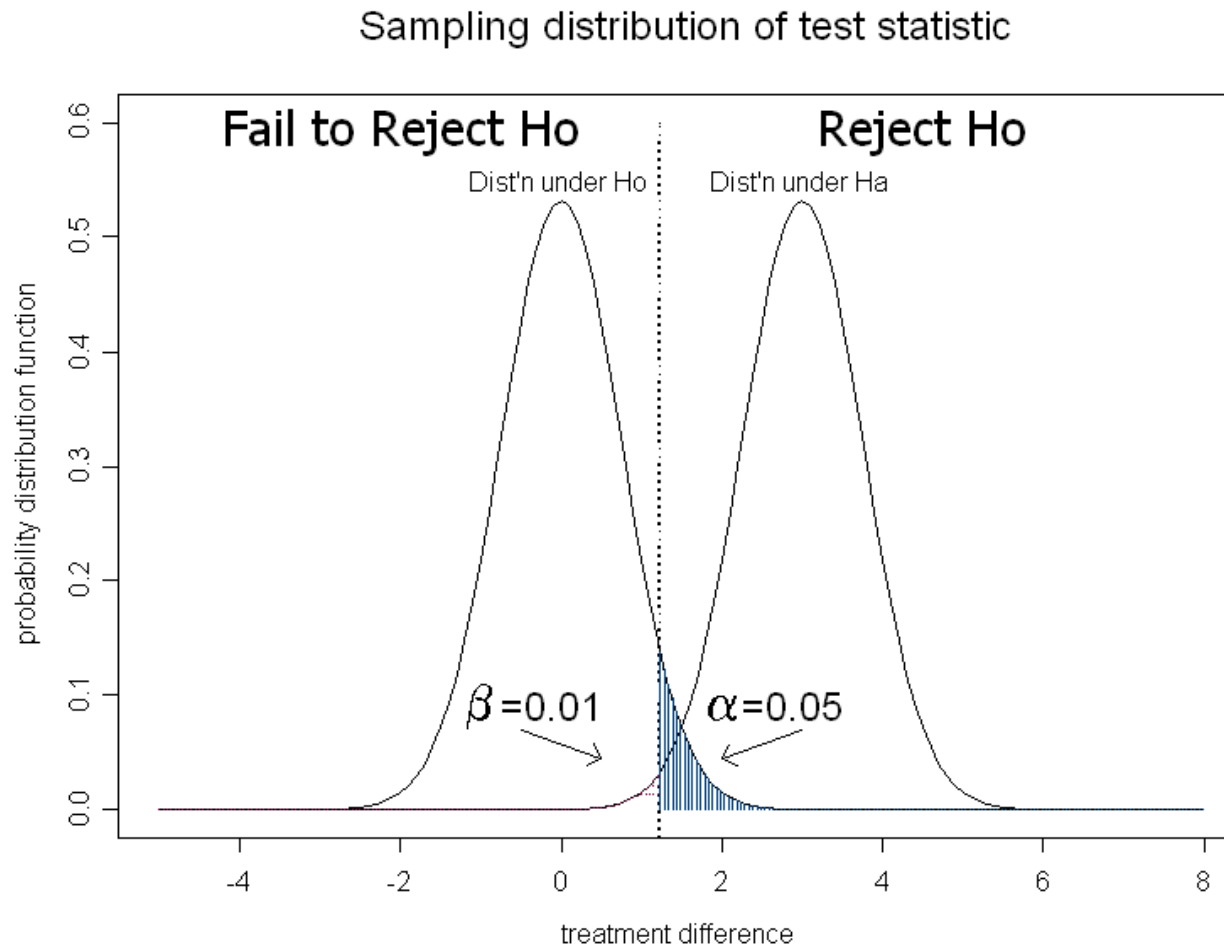
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POWER INCREASES AS STANDARD ERROR DECREASES

$H_0: \mu=0,$
 $SE=0.75$

$H_A: \mu=3,$
 $SE=0.75$

Notice the red region in much smaller.



GENERALIZATIONS

- Can generalize these calculations to:
- Two sample t-test.
- Paired t-test (useful for pre-post studies)
- Difference of differences (useful for pre-post studies on two groups)
- Difference of proportions (useful for 0/1 variables on two groups)
- Regression coefficients (these are t-tests)
- Correlations (this is a z-test)
- Many other things available in software.

Sample size ALWAYS requires the investigator to make some assumptions

- What difference do you expect between the groups?
 - How much variability do we expect in the measurements?
 - What is a clinically relevant difference
-
- Statistician cannot give you these estimates (unless you provide preliminary data)
 - It is the responsibility of clinical investigator to define these parameters (and the statistician to ask for them)

POWER ANALYSIS (SAMPLE SIZE/SIZE OF DIFFERENCE/POWER)

- First determine:
 - Hypothesis, study design, and analysis plan
 - Obtain from the investigator, variability in the outcome measure (usually a standard deviation)
 - Obtain from the investigator, clinically meaningful difference.
 - Acceptable Type I error
- Then solve for one of the following (after setting the values of the other two):
 - Sample Size
 - Difference to be detected
 - Power
- Write-up in grant should clearly describe ALL of the above

POWER

The power analysis section should include treatment of (at least) the following topics:

- Assumptions you made to estimate power
 - Simplifying assumptions (e.g. most sophisticated model will control for covariates, but you use a t-test to estimate power for primary test)
 - Details of the parameters you assumed (e.g. estimate of effect, alpha-level, sample size)
 - Where you got estimates of effect and variability in measures
- Power calculations
 - Estimate of power to detect effect or range of effect sizes for which you have power at X% based on your sample size
 - Cite what program you used (if you used one) or how you estimated power
 - Summary of power calculations if you have several

A FEW THINGS TO REMEMBER

- It is just an estimated sample size or power.
- Want to balance statistical and practical considerations and justify your choices.
- Choose sample size and power so that statistical significance will coincide with clinical significance.
- The sample size is calculated for the number of subjects who finish the study, not the number that start the study.
- The power section of your grant should be consistent with the analysis section of the grant.
 - Often make simplifying assumptions to estimate power.
 - Those assumptions and their presumed impact on power/sample size estimates should be very clearly described.

THANKS TO THESE INDIVIDUALS FOR SOME SLIDES AND/OR INFORMATION FOR THE POWER SECTION

- Tasha Fingerlin, PhD
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- NJC Short course notes (1998)
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 - Lynn Ackerson, PhD