

# Family Studies Center Methods Workshop

## Statistical Power Analysis

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## Overview

- Understand the role of statistical power analysis in family studies research
- Introduce concept of statistical power
- Develop intuitions about factors affecting statistical power
- Learn applications of power analysis when sample size is fixed

## Motivating Problem

- Research is *difficult*, *time-consuming*, and *expensive* to conduct
- *Before* we conduct a study, we want to be assured that we have a reasonable chance of finding an effect if, in fact, one exists
- We must recruit *sufficient numbers* of subjects into our study
- We must *also* consider efforts (and potential for risk) of study participants
- Nearly all studies entail at least some *risk* for participants (even after data are collected!)
- We must not recruit *too many* research subjects into our study

## Motivating Problem

- When number of potential subjects is *limited*, need to *identify design* that gives us the best chance of answering our question
- When number of subjects is *fixed* in advance, need to know *how big an effect* we can detect in our data with desired probability

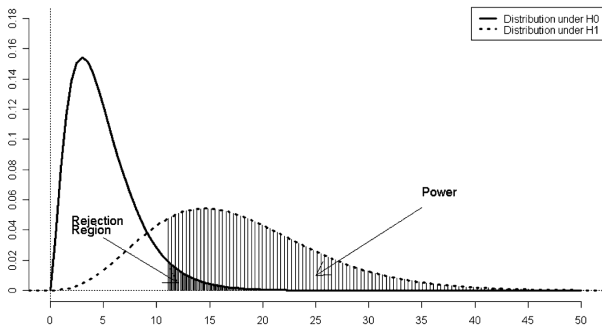
## Ways to be Wrong in Hypothesis Testing

Decision	True State of Affairs	
	$H_0$ True	$H_0$ False
Accept $H_0$	Correct $(1 - \alpha)$	$\beta$
Reject $H_0$	$\alpha$	Correct $(1 - \beta)$

## Ways to be Wrong in Hypothesis Testing

Decision	True State of Affairs	
	No Effect	Effect
No Effect	Correct( $1 - \alpha$ )	Type II Error
Effect	Type I Error	Correct( $1 - \beta$ )

## Central and Non-Central Distributions



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 $\alpha$ 

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## Central and Non-Central Distributions

- *Central* distributions apply when the null hypothesis ( $H_0$ ) is *true*
- They are *standardized*
- *Non-Central* distributions apply when  $H_0$  is *false*
- They are *not standardized*
- Non-Centrality Parameter ( $NCP, \lambda$ ) reflects *degree* to which ( $H_0$ ) is false
- Non-centrality parameter can affect both *location* and *shape* of distribution.



## Central and Non-Central Distributions

- Central  $\chi^2$  distribution with  $df$  degrees of freedom can be generated by squaring and summing  $df$  different random normal variates with means of 0 and variances of 1
- Non-central  $\chi^2$  distribution with  $df$  degrees of freedom and  $NCP = \lambda$  can be generated by squaring and summing  $df$  different random normal variates with means of

$$\mu = \sqrt{\frac{\lambda}{df}}$$

- Four variables are important for power analysis
- $\alpha$
- Power,  $(1 - \beta)$
- $N$
- Effect Size,  $(ES, \lambda)$
- Knowing any 3, solve for fourth
- Two other factors include choice of  $H_0$  and  $Pr(H_0 \text{ is false})$

- Conventions
- $\alpha = .05$
- Power  $\geq .80$
- $N$  (Some applications, may define minimum acceptable standards or heuristics for overall sample size, distinct from power conventions)
- Effect Size
- Power analyses are invaluable *a priori*, not so useful *a posteriori*  
(<http://www.stat.uiowa.edu/files/stat/techrep/tr378.pdf>)

# Visualizing Statistical Power

Statistical  
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Analysis

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$\alpha$

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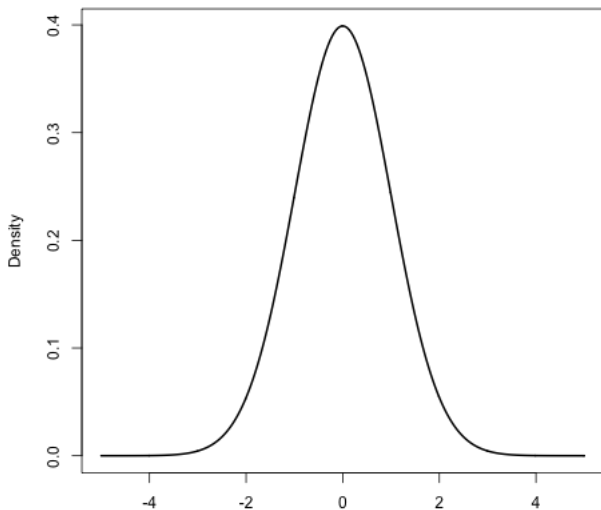
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**alpha = 0.05**



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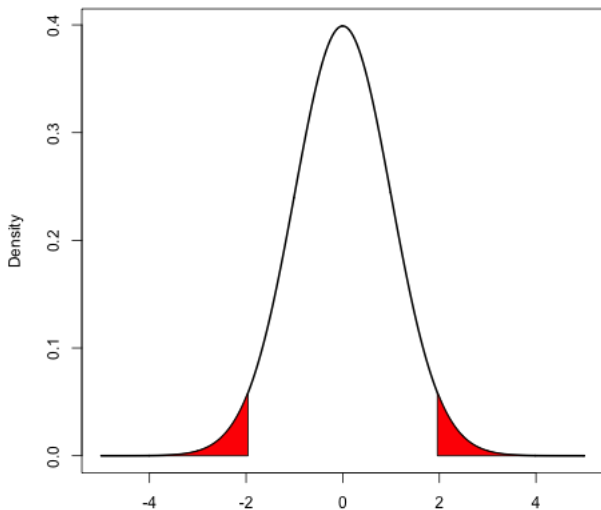
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**alpha = 0.05**





## Visualizing Statistical Power

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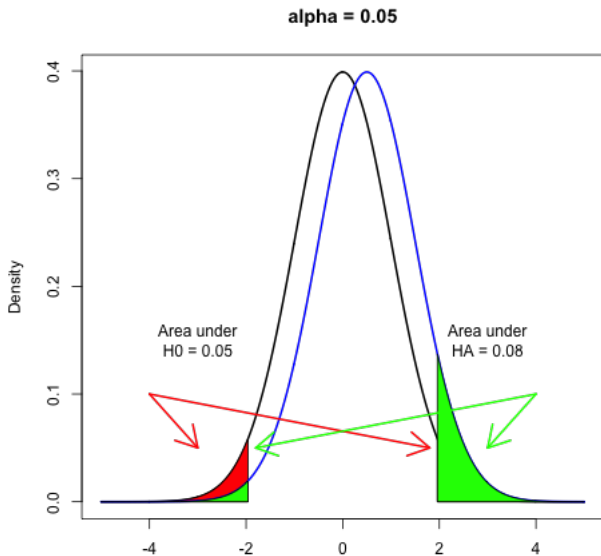
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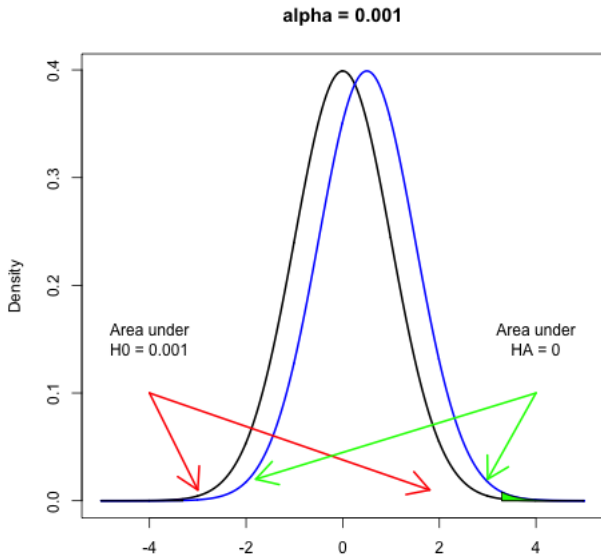
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## Effects of Increasing Alpha





## Effects of Increasing Alpha

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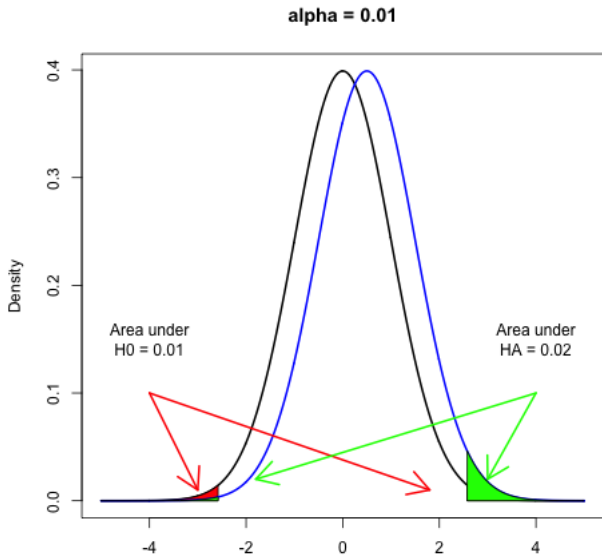
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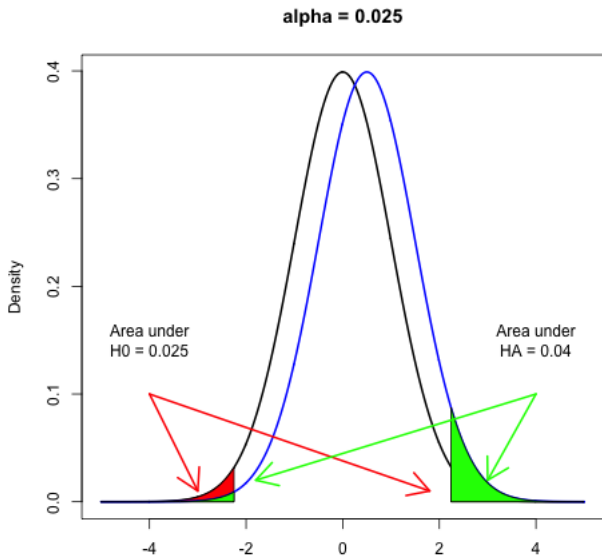
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## Effects of Increasing Alpha

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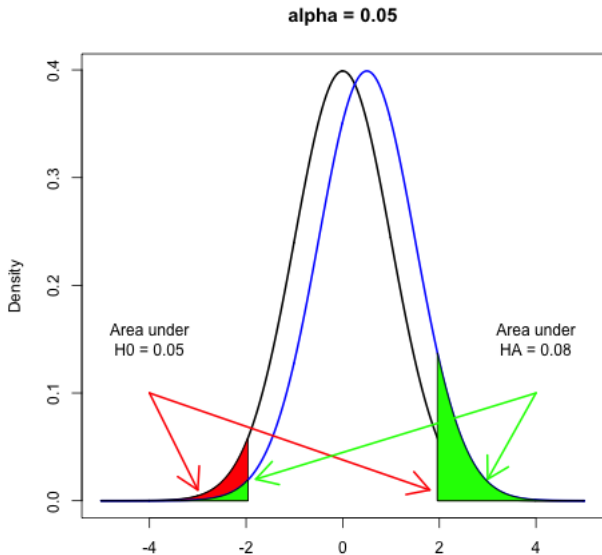
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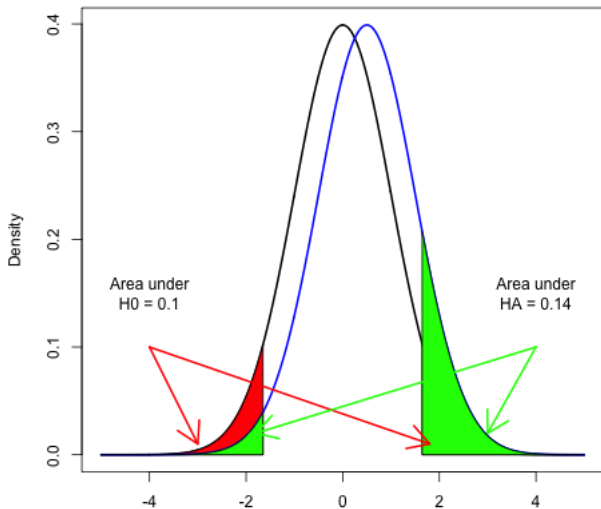
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## Effects of Increasing Alpha

**alpha = 0.1**

## Effects of Increasing Alpha

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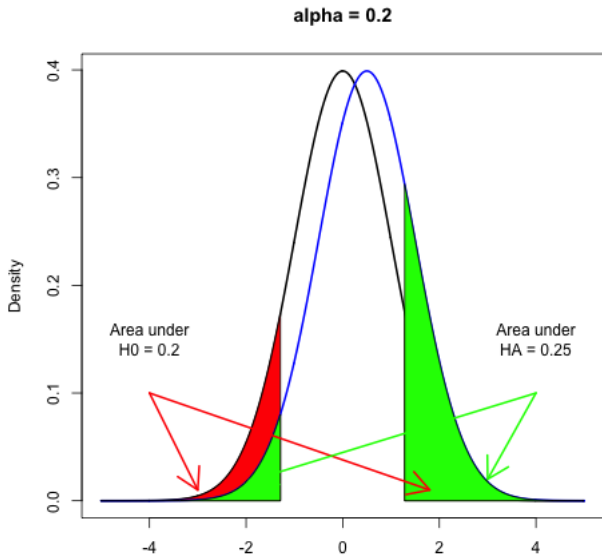
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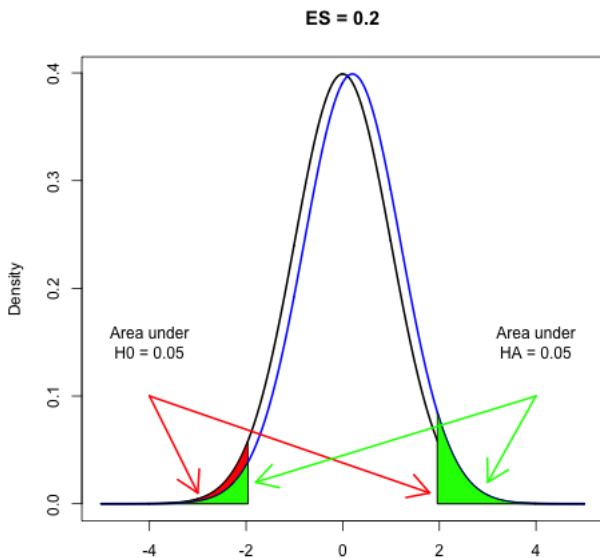
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## Effects of Increasing Effect Size



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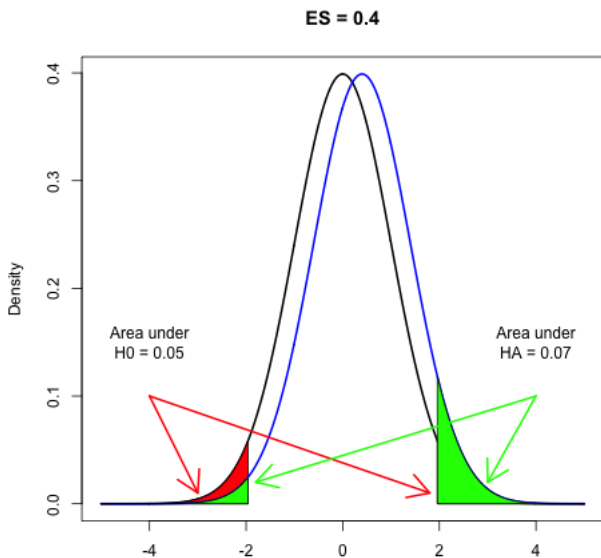
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## Effects of Increasing Effect Size



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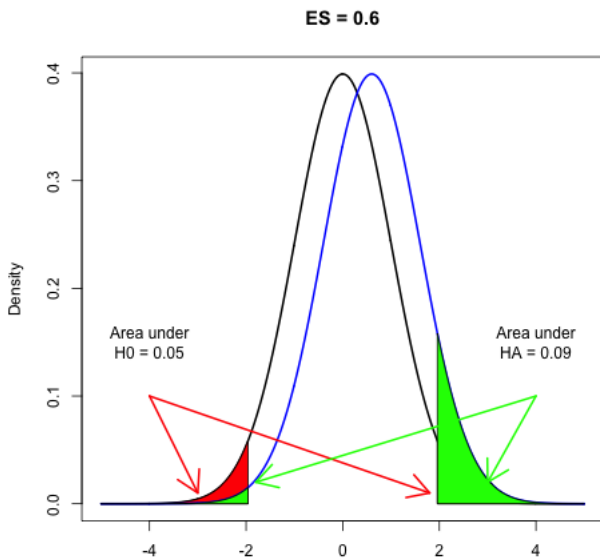
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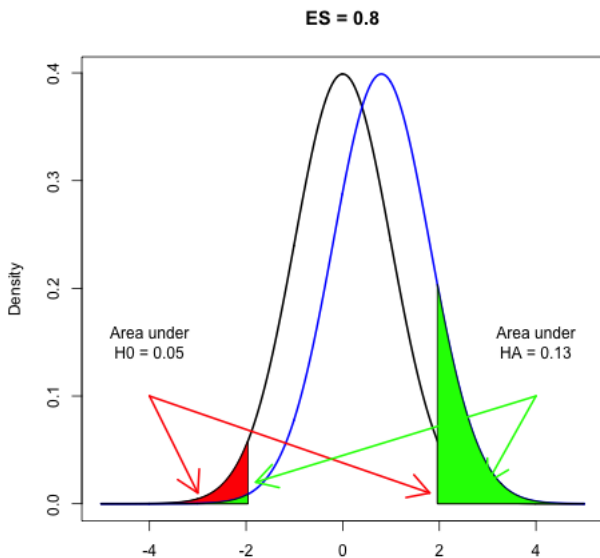
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## Effects of Increasing Effect Size



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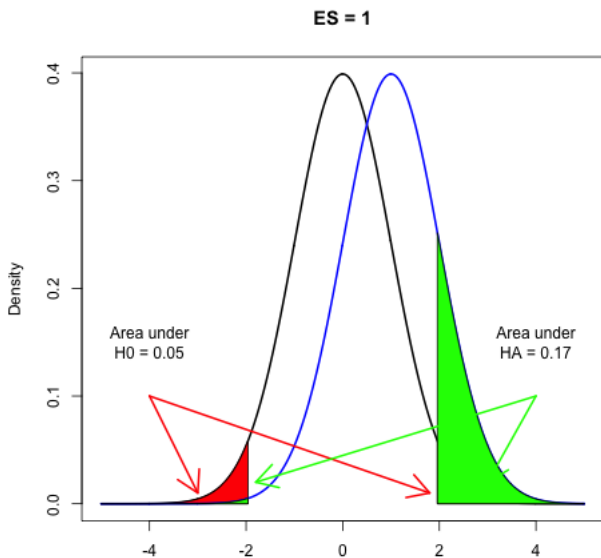
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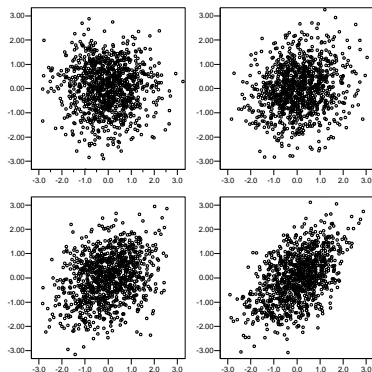
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## Effects of Increasing Effect Size



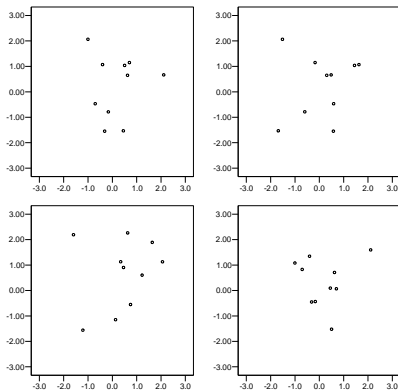
## Discerning Patterns: Large $N$

- Clockwise: (None, Small, Large, Moderate,  $N = 1000$ )



## Discerning Patterns: Small $N$

- Clockwise: (None, Small, Large, Moderate,  $N = 10$ )



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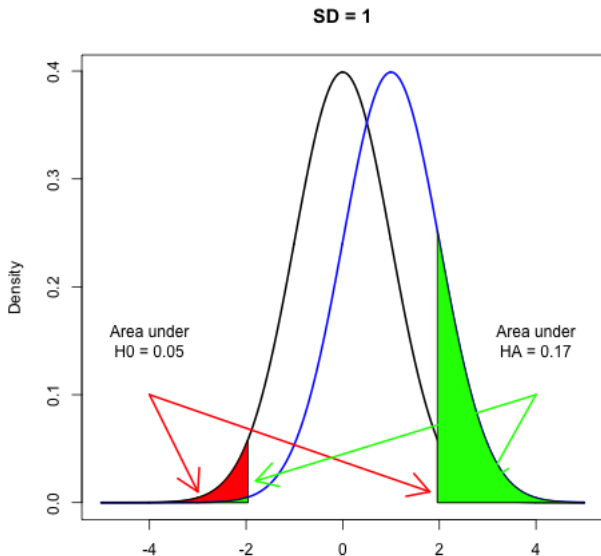
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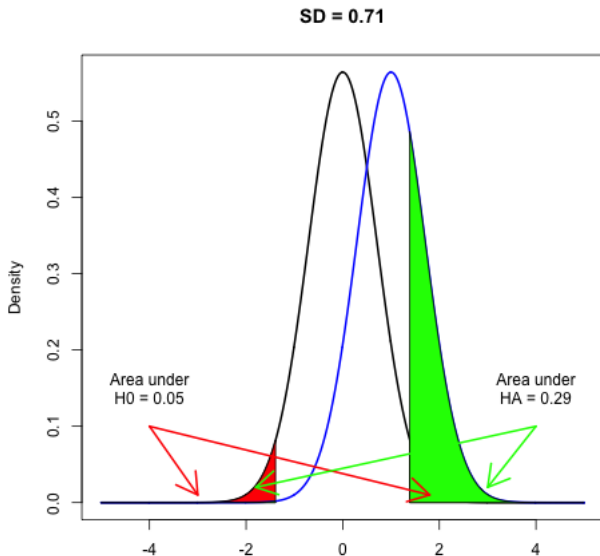
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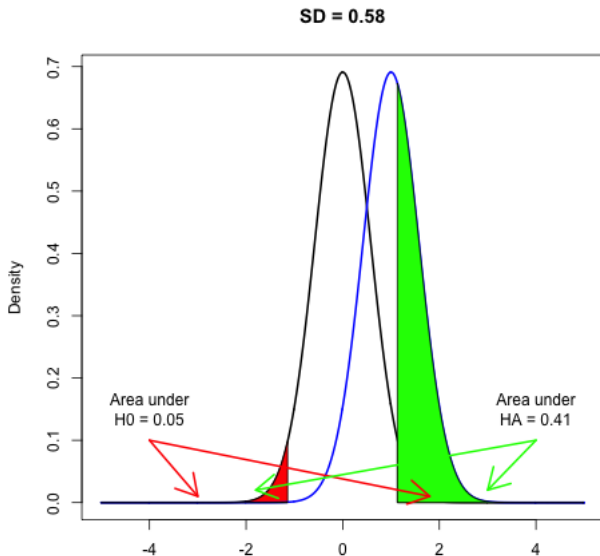
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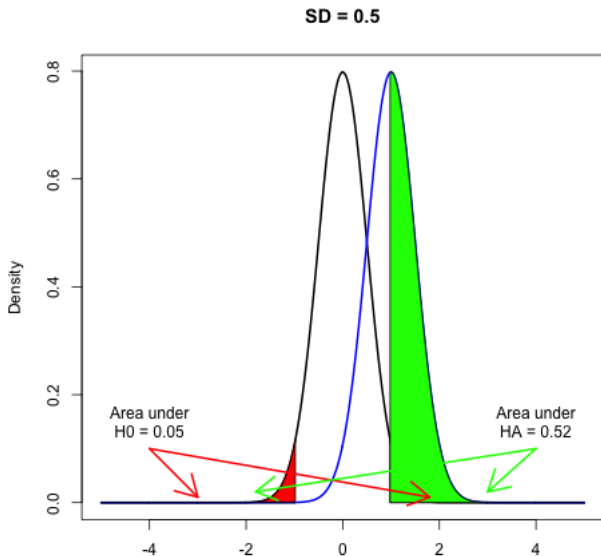
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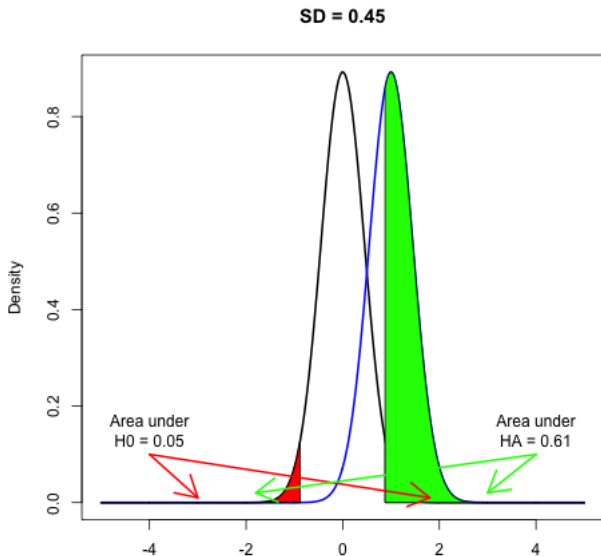
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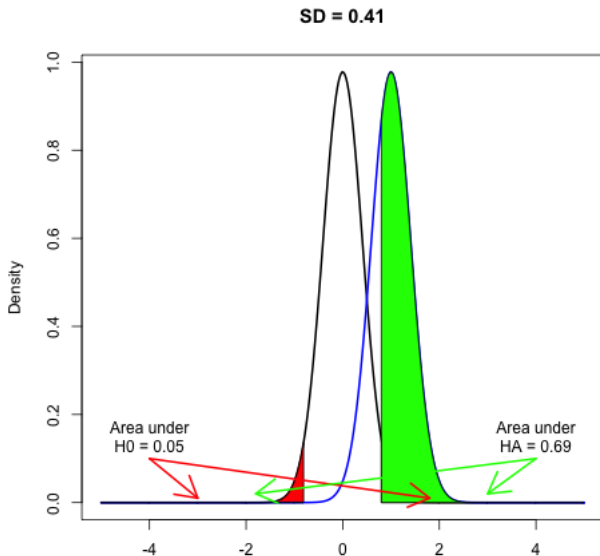
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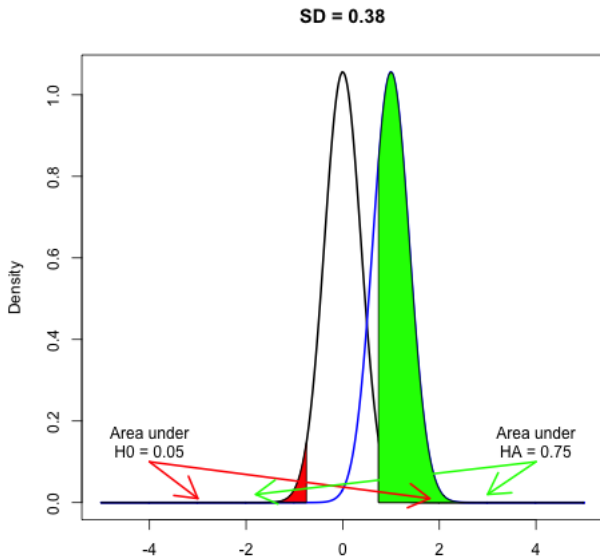
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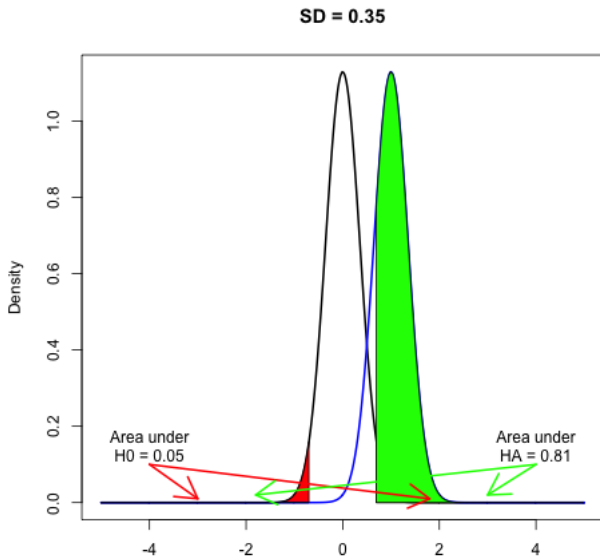
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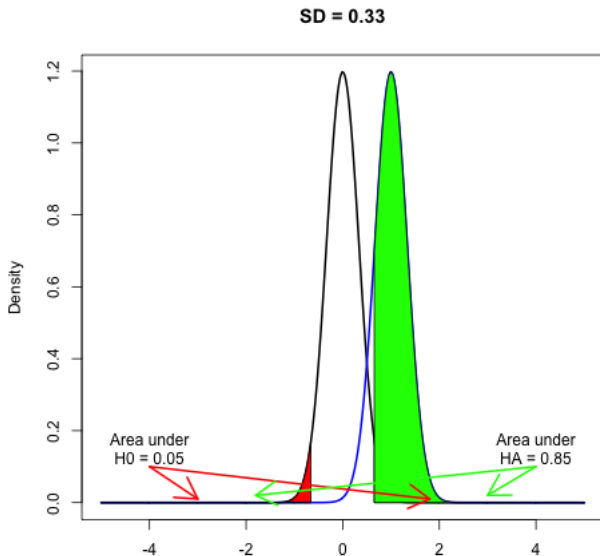
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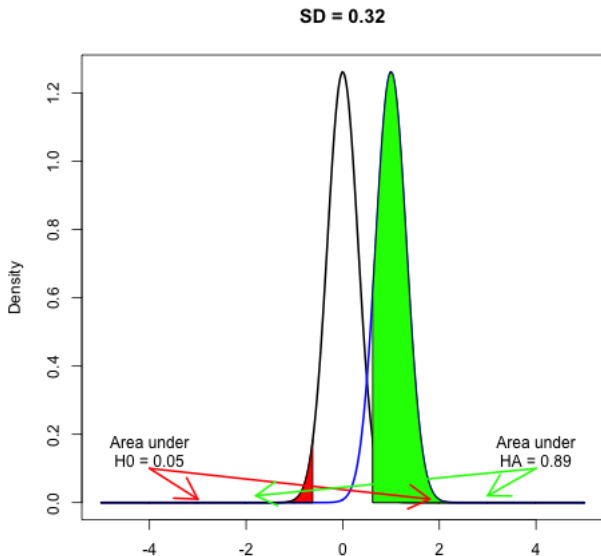
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## Summing Up

- Power of hypothesis test with significance level  $\alpha$  is probability we reject null hypothesis when the alternative is true
- Power is probability that data gathered will be sufficient to reject null hypothesis when it is false
- Power is of *critical importance*

## Summing Up

- Uses of power
- *A priori*: When designing study, select a sample size large enough to detect and effect of magnitude you believe is meaningful
- *A posteriori*: When test finds no significant difference/association, was there enough power to detect effect of meaningful magnitude?
- (Too little, too late. Can still be used to properly power next study.)
- See [http://www.ats.ucla.edu/stat/seminars/Intro\\_power/](http://www.ats.ucla.edu/stat/seminars/Intro_power/) for more.



## Statistical Power of a Test

- Significance testing is a balancing act
  - Chance  $\alpha$  of making Type I error
  - Chance  $\beta$  of making Type II error
  - Reducing  $\alpha$  increases  $\beta$ , and thus reduces the power of a test. It might be tempting to emphasize greater power (the more the better)
- With “too much power” statistical significance may be clinically inconsequential
- A Type II error is not definitive since a failure to reject the null hypothesis does not imply that the null hypothesis is correct
- Since  $H_0$  is either always true or false, we are only in danger of making one kind of error or the other (but we have no idea which one)

## Factors Affecting Power

- Size of effect an important factor in determining power. Higher probability of detecting larger effects
- More conservative significance levels (lower  $\alpha$ ) yield lower power. Less power with  $\alpha = .01$  than with  $\alpha = .05$ .
- Increasing the sample size decreases the spread of the *sampling distribution* and increases power, but there is a trade-off between gain in power and the time/expense of testing a larger sample
- Larger variance ( $\sigma^2$ ) implies a larger spread of the sampling distribution, ( $\sigma/\sqrt{N}$ ). The larger the variance, the lower the power.
- Variance is partly a property of the population, but can be reduced through careful study design.

## Power with Fixed Sample Size

- Many times,  $N$  is fixed, either by resource constraints or with secondary data analysis
- In this context, power analysis serves a different function
- Minimum detectable effect (MDE)
- What is the smallest effect size I can detect with power  $= (1 - \beta)$ , sample size  $= N$ , and alpha  $= \alpha$ ?
- (Stata Users: `db power`)

## Power with Fixed Sample Size

- Accuracy in parameter estimation (AIPE: <http://www.ats.ucla.edu/stat/stata/dae/aipe.htm>)
- Bracketing effect sizes (half-width,  $w$ ). For sample size  $N$ , find range that give  $p\%$  chance that the estimated interval will be  $\leq 2 \times w$
- The AIPE paradigm is a framework for managing width of confidence interval, independent of effect size)
- (Stata Users: `findit aipe`)

## Additional Resources for Power (Books)

- (Too) Simple  
Cohen, J. (1992). A power primer. *Psychological bulletin*, 112(1), 155.  
<http://classes.deonandan.com/hss4303/2010/cohen%201992%20sample%20size.pdf>
- Just Sufficient  
Cohen, J. (2013). *Statistical power analysis for the behavioral sciences*. Routledge Academic.  
[http://www.lrdc.pitt.edu/schneider/P2465/Readings/Cohen,%201988%20\(Statistical%20Power,%20273-406\).pdf](http://www.lrdc.pitt.edu/schneider/P2465/Readings/Cohen,%201988%20(Statistical%20Power,%20273-406).pdf)

## Additional Resources for Power (Books)

- More Contemporary

Murphy, K. R., Myers, B., & Wolach, A. H. (2009). *Statistical power analysis: A simple and general model for traditional and modern hypothesis tests*. Routledge.

- Extensions

Davey, A. (2009). *Statistical power analysis with missing data: A structural equation modeling approach*. Routledge.

Muthén, L. K., & Muthén, B. O. (2002). How to use a Monte Carlo study to decide on sample size and determine power. *Structural Equation Modeling*, 9(4), 599-620.  
[http://www.statmodel.com/bmuthen/ED231e/RelatedArticles/Article\\_097.pdf](http://www.statmodel.com/bmuthen/ED231e/RelatedArticles/Article_097.pdf)

## Additional Resources for Power (Software)

- G\*Power  
<http://www.gpower.hhu.de/en.html>
- R Package pwr  
<http://www.statmethods.net/stats/power.html>  
<http://cran.r-project.org/web/packages/pwr/pwr.pdf>
- R Package powerMediation  
<http://cran.r-project.org/web/packages/powerMediation/powerMediation.pdf>