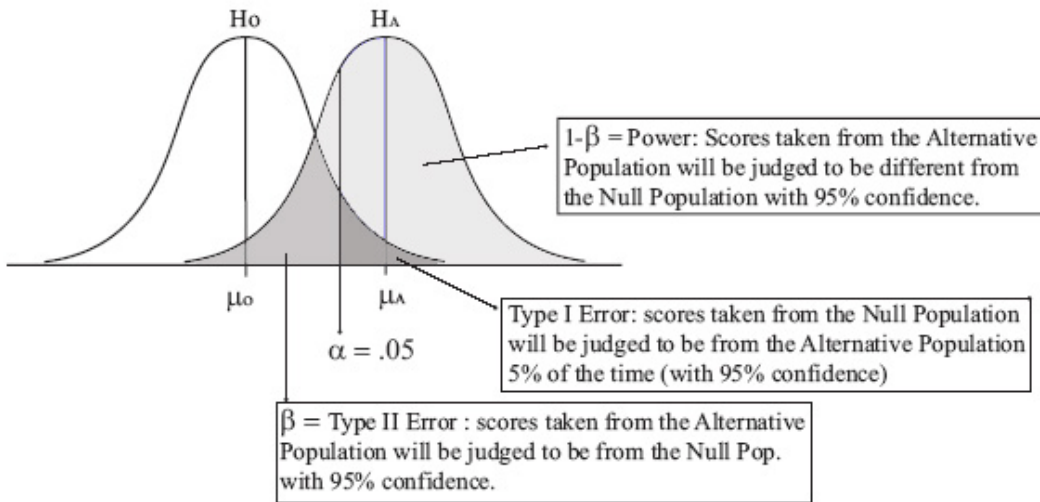


Power = Effect Size, Sample Size, & Alpha
Handout

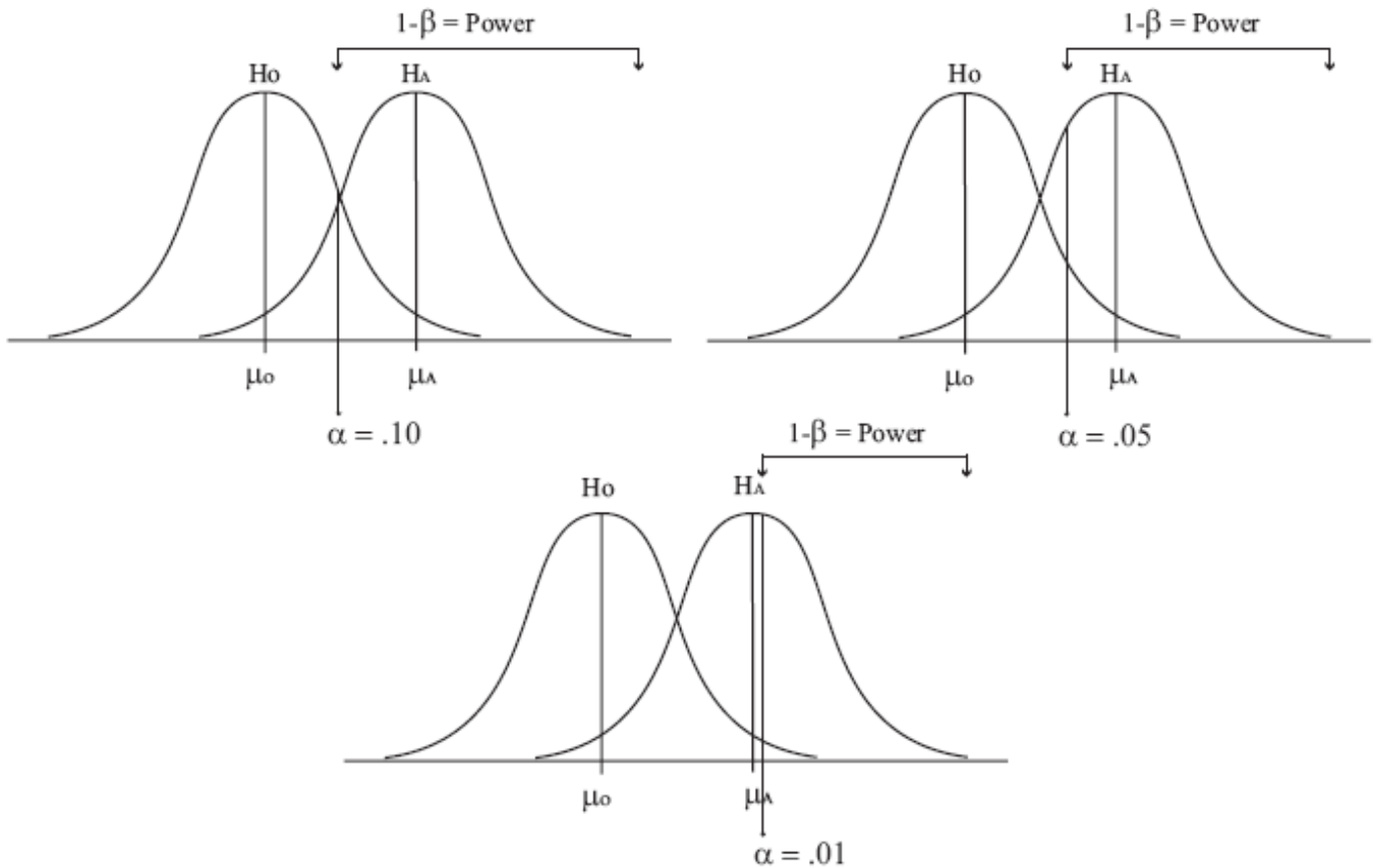
I. $Power = 1 - \beta$

II. Power is the probability that you will identify a significant relationship between 2 variables when in fact a significant relationship exists.



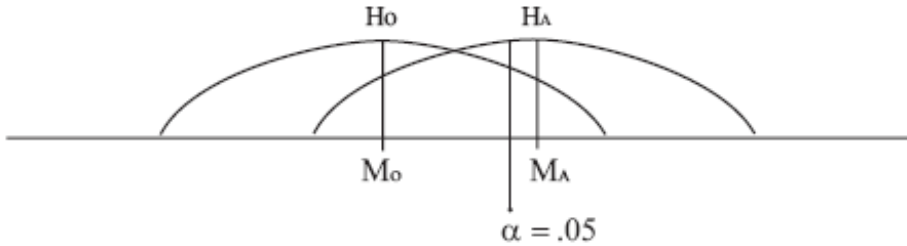
III. Power Determined by:

1. Alpha Level (Probability of Type II Error)

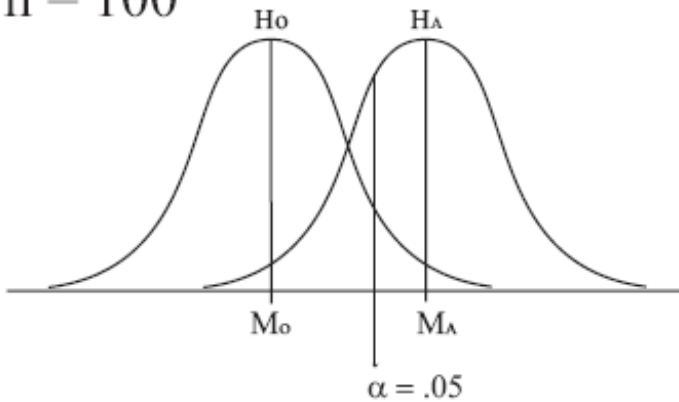


2. Sample Size

$$n = 10$$



$$n = 100$$



3. Effect Size (d , r^2 , R^2 , η^2 , ω^2 (omega), ϕ^2)

- d is the standardized distance between two means (Howell, 2002).
- r^2 , R^2 , η^2 & ϕ^2 are all essentially squared Correlation Coefficients.
- ω^2 (omega) is an adjusted version of a squared Pearson Correlation Coefficient.
- All of these can be converted to d

IV. Effect Size: Amount of Change in DV attributable to changes in the IV

1. d conceptualizes effect size as the standardized distance between two means

$$d = \left[\frac{\mu_1 - \mu_2}{\sigma} \right]$$

Single Sample and repeated measures t :

$$d = \frac{t}{\sqrt{df}}$$

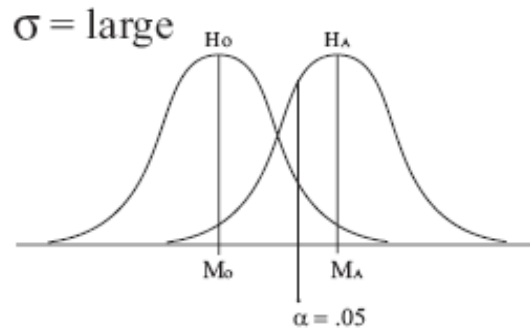
Independent sample t :

$$d = \frac{2t}{\sqrt{df}} \text{ When sample sizes are equal (or must be assumed to be)}$$

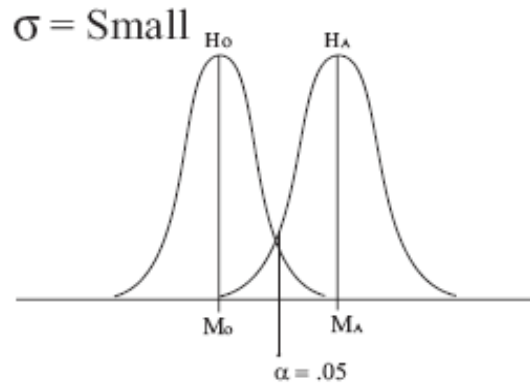
$$d = \frac{t(n_1 + n_2)}{\sqrt{df} \sqrt{n_1 n_2}} \text{ For unequal sample sizes}$$

2. As Variance Decreases - Effect Sizes Increase and Power Increases

- More variance spreads out the distribution and results in more overlap between the distributions.

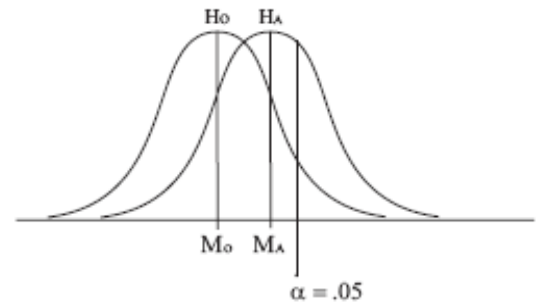


- Less variance makes the distribution more peaked and reduces the overlap between distributions and therefore results in more power.

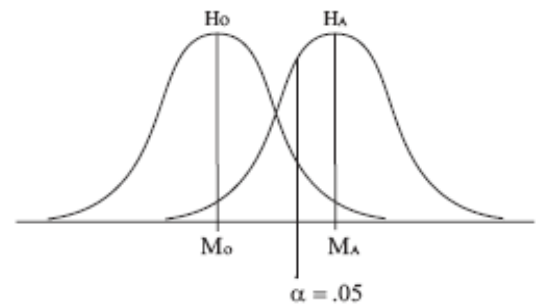


3. As Mean Differences Increase - Effect Size Increases and Power Increases

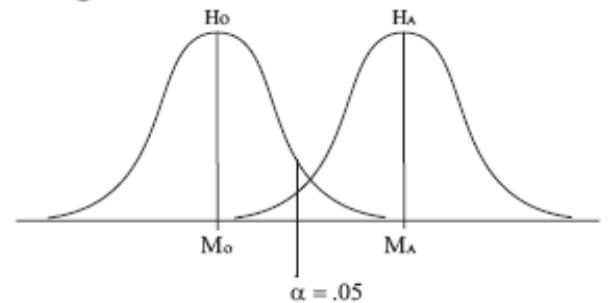
Small Mean Difference



Medium Mean Difference



Large Mean Difference



4. Cohen (1988, 1992) suggests the following framework to interpret effect sizes.

Effect Size	d	% overlap of Dists.
Small	.20	85%
Medium	.50	67%
Large	.80	53%

V. Calculating Power

- Power can be estimated using power table like the one Howell (2002) provides
- This table is based on δ (lower case delta), which is a Non-Centrality Parameter that expresses how wrong the Null Hypothesis is (see Howell, 2002, p. 231 for a full discussion).

1. Power for t -Tests

a. Single Sample t and Repeated Measures/Matched Sample t

$$\delta = d\sqrt{N} \quad \text{where } N = \text{number of sample observations}$$

Independent Sample t with equal group sizes assumed

$$\delta = d\sqrt{\frac{N}{4}} \quad \text{where } N = n_1 + n_2$$

For Unequal Group sizes use the following -

$$\delta = d\sqrt{\frac{n_h}{2}} = \text{where } n_h = \frac{2n_1n_2}{n_1 + n_2} \quad \text{which is the Harmonic mean.}$$

VI. Using Power to Calculate the Sample Size Need to Achieve a Desired Level of Power

$$N \text{ desired for Single Sample/Repeated Measures } t: N_{desired} = \left(\frac{\delta}{d}\right)^2$$

$$\text{When finding total } N \text{ desired for Independent Sample } t : N_{desired} = 4\left(\frac{\delta}{d}\right)^2$$

$$n_i = (N/2)$$

Following Cohen's (1988) framework, we would expect:

Given Power of .80 @ $\alpha=.05 = \delta = 2.8$

<u>Effect Size</u>	<u>d</u>	N for <u>1 sample t</u>	N for <u>2 sample t</u>
Small	.20	196	784
Medium	.50	32	128
Large	.80	13	52

VII. More Effect Sizes

1. Effect Size for Correlations:

r^2 = Coefficient of multiple determination

$$r^2 = \frac{SS_{\text{betweengroups}}}{SS_{\text{total}}} = r^2 = \left(\frac{\sum (X_x - \bar{X}_x)(X_y - \bar{X}_y)}{\sqrt{\sum (X_x - \bar{X}_x)^2} \sqrt{\sum (X_y - \bar{X}_y)^2}} \right)^2$$

Obtaining r^2 from d : $r^2 = \frac{d^2}{d^2 + 4}$

Obtaining d from r : $d = \frac{2r}{\sqrt{1 - r^2}}$

Noncentrality Parameter for r^2 , convert to d first, then: $\delta = d\sqrt{\frac{N}{4}}$

N desired: $N_{\text{desired}} = 4\left(\frac{\delta}{d}\right)^2$

Converting R^2 and r^2 to ω^2 (corrects for tendency of R^2 and r^2 to overestimate the population parameter):

$$\omega^2 = 1 - \frac{N - 1}{N - a}(1 - r^2)$$

2. Cohen's 1992 standards for effect sizes with correlation

Effect Size	r	r^2
Small	.10	.01
Medium	.30	.09
Large	.50	.25

3. Effect Size for X^2 (Chi-Square)

2x2 Matrix: $\phi^2 = \frac{X^2}{N}$ Noncentrality Parameter for 2x2 X^2 : $\delta = \left(\frac{2\phi}{\sqrt{1 - \phi^2}} \right) \sqrt{\frac{N}{4}}$

Larger than 2x2 Matrix: Cramer's $\phi^2 = \sqrt{\frac{X^2}{N(K - 1)}}$

where K = the smaller of the # of rows or # of columns

Noncentrality Parameter for X^2 larger than 2 x 2 = ?

4. Effect Size for Anova = R^2 (eta squared)

$$d = \frac{2\sqrt{F}}{\sqrt{df_{within}}} \quad \text{Because } F = t^2 \text{ (if you have two groups with equal group sizes)}$$

$$d = \frac{\sqrt{F}(n_1 + n_2)}{\sqrt{df_{within}} \sqrt{n_1 n_2}} \quad \text{For two groups with unequal group sizes.}$$

For more than 2 groups

$$R^2 = \frac{SS_{between\ groups}}{SS_{total}} \quad \text{Taken from Anova Tables.}$$

Or

$$R^2 = \frac{F(df_{between})}{(F(df_{between})) + (df_{within})}$$

$$\text{Noncentrality Parameter for } R^2 \text{ for ANOVA with 2 equal sized Groups: } \delta = \left(\frac{2R}{\sqrt{1-R^2}} \right) \sqrt{\frac{N}{4}}$$

Desired Sample Size for ANOVA for ANOVA with 2 Groups

$$N_{desired} = 4 \left(\frac{\delta}{d} \right)^2 \quad \text{Where } k \text{ is the number of groups.}$$

$$n_1 = \text{sample size for each group} = N_{desired} / k$$

Ancillary Formulas

Z test

$$z = \frac{\bar{X} - \mu}{\sigma}$$

Single Sample t

$$t = \frac{\bar{X} - \mu}{\frac{s}{\sqrt{n}}} \quad \text{where } df = n - 1$$

Independent Sample t-test

$$t = \frac{\bar{X}_1 - \bar{X}_2}{\sqrt{\left(\frac{(n_1 - 1)s_1^2 + (n_2 - 1)s_2^2}{n_1 + n_2 - 2} \right) \left(\frac{1}{n_1} + \frac{1}{n_2} \right)}} \quad \text{Where } df = n_1 + n_2 - 2$$

Repeated Measures t -test (match sample, correlated sample, difference t)

$$t = \frac{(\sum (X_1 - X_2))/n}{\sqrt{\left(\frac{\sum ((X_1 - X_2) - \sum (X_1 - X_2)/n)^2}{n - 1}\right)\left(\frac{1}{n}\right)}}$$

Where $df = n - 1$

Pearsons Product Moment Correlation

$$r = \frac{\sum XY^2 - \frac{(\sum X)(\sum Y)}{n}}{\sqrt{\sum X^2 - \frac{(\sum X)^2}{n}} \sqrt{\sum Y^2 - \frac{(\sum Y)^2}{n}}}$$

or

$$r = \frac{\sum (X - \bar{X})(Y - \bar{Y})}{\sqrt{\sum (X - \bar{X})^2} \sqrt{\sum (Y - \bar{Y})^2}}$$

Where $df = n - 2$

One Way ANOVA Summary Table

<i>Source</i>	<i>SS</i>	<i>df</i>	<i>MS</i>	<i>F</i>	<i>p</i>
Between	?	$k - 1$	$SS_{\text{btw}}/df_{\text{btw}}$	$MS_{\text{btw}}/MS_{\text{within}}$?
Within (error)	?	$n - k - 1$	$SS_{\text{within}}/df_{\text{within}}$		
TOTAL	$SS_{\text{btw}} + SS_{\text{within}}$	$n - 1$			