

## Effect Sizes: Definitional Formulae

See:

Rosenthal, R. (1994). Parametric measures of effect size. In H. Cooper & L. V. Hedges (Eds.), *The handbook of research synthesis* (pp. 231-244). New York: Russell Sage Foundation.

Rosenthal, R. Rosnow, R. L., & Rubin, D. B. (2000) Contrasts and effect sizes in behavioral research: A correlational approach. Cambridge, UK: Cambridge University Press.

Please feel free to contact me if any of these formulae seem incorrect – it is possible that typographical errors may have been made.

### Effect sizes based on correlations

$$r = \Phi = r_{pb} = \frac{\sum Z_x Z_y}{n} = \frac{\sum \left( \frac{X - \bar{X}}{\sqrt{\frac{\sum (X - \bar{X})^2}{n}}} \right) \left( \frac{Y - \bar{Y}}{\sqrt{\frac{\sum (Y - \bar{Y})^2}{n}}} \right)}{n} = \frac{\sum (X - \bar{X})(Y - \bar{Y})}{n s_x s_y} = \frac{\text{covariance}_{xy}}{s_x s_y}$$

*NOTE: These formulae are for the population correlation between x and y. You may also see these formulae with n-1 rather than n, which provides the sample estimate of the population correlation. Whichever form is chosen, be sure to use either n or n-1 in all parts of the formula.*

$$z_r = r' = \frac{1}{2} \log_e \left[ \frac{1+r}{1-r} \right] = \frac{1}{2} [\log_e (1+r) - \log_e (1-r)] \quad \text{where } \log_e \text{ is the natural log function}$$

(LN on some calculators)

To transform back from  $z_r$  ( $r'$ ) metric to  $r$  metric

$$r = \frac{e^{2z_r} - 1}{e^{2z_r} + 1}, \quad \text{where } e \text{ is the exponent function } (e^x \text{ on some calculators})$$

$$\text{Cohen's } q = Z_{r1} - Z_{r2}$$

Effect sizes a la d

$$\text{Cohen's } d = \frac{\bar{X}_1 - \bar{X}_2}{\sigma_{\text{pooled}}}$$

$$\text{Hedges' } g = \frac{\bar{X}_1 - \bar{X}_2}{s_{\text{pooled}}}$$

$$\text{Glass' } \Delta = \frac{\bar{X}_1 - \bar{X}_2}{s_{\text{control group}}}$$

Where

$$\sigma_{\text{pooled}} = \sqrt{\frac{(n_1) \sigma_1^2 + (n_2) \sigma_2^2}{n_1 + n_2}} \quad \text{and} \quad s_{\text{pooled}} = \sqrt{\frac{(n_1 - 1) s_1^2 + (n_2 - 1) s_2^2}{n_1 + n_2 - 2}}$$

$$\text{and} \quad \sigma_{\text{pooled}} = s_{\text{pooled}} \sqrt{\frac{n_1 + n_2 - 2}{n_1 + n_2}}$$

*You may also see  $\sigma_{\text{pooled}}$  referred to as  $\sigma_{\text{within}}$ , and  $s_{\text{pooled}}$  referred to as  $s_{\text{within}}$  or as  $\sqrt{MS_{\text{within}}}$*

### Effect sizes for proportions

$$\text{Cohen's } g = p - .50$$

*where  $p$  estimates a population proportion*

$$d' = p_1 - p_2$$

*where  $p_1$  and  $p_2$  are estimates of the population proportions*

$$\text{Cohen's } h = \arcsin p_1 - \arcsin p_2$$

$$\text{Probit } d' = Z_{p_1} - Z_{p_2}$$

*where  $Z_{p_1}$  and  $Z_{p_2}$  are standard normal deviate transformed estimates of population proportions*

$$\text{Logit } d' = \log_e \left[ \frac{p_1}{1 - p_1} \right] - \log_e \left[ \frac{p_2}{1 - p_2} \right]$$

## Transforming between Effect Sizes

### Computing $r$

$$r = \sqrt{\frac{d^2}{d^2 + 4}} \quad \text{-- if you assume that the two populations are equally numerous}$$

$$r = \sqrt{\frac{d^2}{d^2 + \frac{1}{PQ}}} \quad \text{-- if the populations are clearly different in size}$$

where  $P$  is the proportion of Group 1 individuals in the combined Group 1 + Group 2 population and  $Q = 1 - P$ .

$$r = \sqrt{\frac{g^2 n_1 n_2}{g^2 n_1 n_2 + (n_1 + n_2) df}}$$

where  $df = n_1 + n_2 - 2$  in the 2 - group case and  $n - 1$  in the 1 - group case

### Computing $d$

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

$$d = g \sqrt{\frac{n_1 + n_2}{df}}$$

### Computing $g$

$$g = \frac{r}{\sqrt{1 - r^2}} \sqrt{\frac{df(n_1 + n_2)}{n_1 n_2}}$$

$$g = \frac{d}{\sqrt{\frac{n_1 + n_2}{df}}}$$

## Transforming between Effect Sizes (continued)

transform between eta squared ( $\eta^2$ ) and omega squared ( $\omega^2$ )

$$\eta^2 \text{ for an effect} = \frac{\left( df_{\text{effect}} + \frac{n_i n_j \omega^2}{1 - \omega^2} \right) \left( \frac{1}{df_{\text{error}}} \right)}{\left( df_{\text{effect}} + \frac{n_i n_j \omega^2}{1 - \omega^2} \right) \left( \frac{1}{df_{\text{error}}} \right) + 1}$$

$$\omega^2 \text{ for an effect} = \frac{\frac{df_{\text{error}} \eta^2}{1 - \eta^2} - df_{\text{effect}}}{\left( \frac{df_{\text{error}} \eta^2}{1 - \eta^2} - df_{\text{effect}} \right) + n_i n_j}$$

## Computing significance tests from effect sizes

Recall, Inferential test statistic = Effect size X Size of Study

$$\chi^2_{(1)} = Z^2 = r^2 N$$

$$Z = \sqrt{\chi^2_{(1)}} = r\sqrt{N}$$

$$t = \frac{r}{\sqrt{1-r^2}} \sqrt{df}$$

$$t = d \left( \frac{\sqrt{n_1 n_2}}{(n_1 + n_2)} \right) \sqrt{df} \quad \text{for the independent sample } t \text{ - test and unequal sample sizes}$$

$$t = d \frac{\sqrt{df}}{2} \quad \text{for the independent sample } t \text{ - test and equal sample sizes}$$

$$t = d \sqrt{df} = \frac{\bar{X}_{\text{sample}} - \mu_{\text{population}}}{\sqrt{\frac{s^2_{\text{sample}}}{n_{\text{sample}}}}} \quad \text{for the one sample } t \text{ - test}$$

$$t = g \left( \frac{1}{\sqrt{\frac{1}{n_1} + \frac{1}{n_2}}} \right) = g \left( \sqrt{\frac{n_1 n_2}{n_1 + n_2}} \right) = \frac{\bar{X}_1 - \bar{X}_2}{\sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}}} \text{ or } \frac{\bar{X}_1 - \bar{X}_2}{\sqrt{\left( \frac{1}{n_1} + \frac{1}{n_2} \right) s_{\text{pooled}}^2}} \quad \text{for the independent sample } t \text{ - test}$$

$$F = t^2 = \frac{r^2}{1-r^2} (df_{\text{error}}) = g^2 \left( \frac{n_1 n_2}{n_1 + n_2} \right) = d^2 \left( \frac{df_{\text{error}}}{4} \right) \quad \text{for an } F \text{ test with numerator } df = 1$$

$$F = \frac{s^2_{\text{betweenmeans}}}{s^2_{\text{withinmeans}}} (n) \quad \text{assuming equal } n \text{'s - for } F \text{ tests with } \geq 1 \text{ numerator } df$$

$$F = \frac{\eta^2}{1-\eta^2} \left( \frac{df_{\text{error}}}{df_{\text{means}}} \right) \quad \text{for } F \text{ tests with } \geq 1 \text{ numerator } df$$

## Computing Effect Sizes from significance tests

$$r = \Phi = r_{pb} = \sqrt{\frac{\chi^2_{(1)}}{n}} = \frac{Z}{\sqrt{n}}$$

$$r = \sqrt{\frac{t^2}{t^2 + df}} = \sqrt{\frac{F}{F + df_{\text{error}}}} \quad \text{for } F \text{ tests with numerator } df = 1$$

$$d = t \left( \frac{n_1 + n_2}{\sqrt{df} \sqrt{n_1 n_2}} \right) \quad \text{which simplifies to } d = \frac{2t}{\sqrt{df}} \text{ if the groups have equal } n$$

$$d = \sqrt{F} \left( \frac{n_1 + n_2}{\sqrt{df} \sqrt{n_1 n_2}} \right) \quad \text{which simplifies to } d = \frac{2\sqrt{F}}{\sqrt{df}} \text{ if the groups have equal } n.$$

(For F tests with numerator  $df = 1$ )

$$g = t \left( \frac{\sqrt{n_1 + n_2}}{\sqrt{n_1 n_2}} \right) \quad \text{which simplifies to } g = \frac{2t}{\sqrt{n}} \text{ if the groups have equal } n$$

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(For F tests with numerator  $df = 1$ )

For F tests with any numerator degrees of freedom:

$$\eta^2 \text{ for an effect} = \eta^2 = \frac{\frac{F_{\text{effect}}(df_{\text{effect}})}{df_{\text{error}}}}{\frac{F_{\text{effect}}(df_{\text{effect}})}{df_{\text{error}}} + 1}$$

$$\omega^2 \text{ for an effect} = \omega^2 = \frac{(F_{\text{effect}} - 1)(df_{\text{effect}})}{(F_{\text{effect}} - 1)(df_{\text{effect}}) + n_i n_j}$$