# Meta-Analysis of Tests for the Odds Ratio of Two Proportions using a Random-Effects Model 

## Introduction

This procedure calculates power and number of studies for fixed-effect meta-analyses of parallel group studies that are summarized by the odds ratios in two group proportions. The sample size formulas used by this procedure were published in Hedges and Pigott (2001) and Harrer et al. (2022).

## Meta-Analysis

Meta-analysis refers to the systematic assessment of a group of two or more individual studies (experiments or clinical trials). The studies must be similar so that their results may be numerically compared. For example, in this procedure all studies must have as their outcome two group proportions whose odds ratio is used to compare their values.

By combining the results of several studies, their individual results may be summarized and assessed. The power of the combined test will be increased since it represents a larger sample.

## Fixed-Effects versus Random-Effects Models

The fixed-effects model, which is assumed by another procedure, assumes that all studies have the same true effect size. That is, it assumes that all factors that might influence the effect size are the same so that the true effect size for all studies is the same. Therefore, only the within study variability needs to be considered since the effect size variation occurs because of random error.

The random-effects model, which is assumed by this procedure, assumes that the studies in the analysis are similar, but not to the extent that they all have the same true effect size. Instead, it is assumed that each study has its own true effect size and that the particular outcome of the study is a sample from a distribution with that proportion. The random-effects model must not only account for the within study variability, but it must also account for the between study variability.

## Technical Details

## The Meta-Analysis of the Odds Ratio of Two Proportions

The sample size formulas used by this procedure were published in Harrer et al. (2022).
A series of $K$ studies are assumed to each have an effect size $O R$ which is the odds ratio between two group proportions. The results of these individual studies may be combined into one overall statistical test that draws a more precise conclusion.

## Test Statistic

Suppose each study has two populations from which dichotomous (binary) responses are recorded. The probability (or risk) of obtaining the event of interest in population 1 (the treatment group) is $p_{1}$ and in population 2 (the control group) is $p_{2}$. The corresponding failure proportions are given by $q_{1}=1-p_{1}$ and $q_{2}=1-p_{2}$.
An assumption is made that the responses from each group follow a binomial distribution. This means that the event probability, $p_{i}$, is the same for all subjects within the group and that the response from one subject is independent of that of any other subject.

Random samples of $m$ and $n$ individuals are obtained from these two populations. The data from these samples can be displayed in a 2-by-2 contingency table as follows

| Group | Success | Failure | Total |
| :--- | ---: | ---: | ---: |
| Treatment | $a$ | $c$ | $m$ |
| Control | $b$ | $d$ | $n$ |
| Total | $s$ | $f$ | $N$ |

The following alternative notation is sometimes used.

| Group | Success | Failure | Total |
| :--- | ---: | ---: | ---: |
| Treatment | $x_{11}$ | $x_{12}$ | $n_{1}$ |
| Control | $x_{21}$ | $x_{22}$ | $n_{2}$ |
| Total | $m_{1}$ | $m_{2}$ | $N$ |

The binomial proportions, $p_{1}$ and $p_{2}$, are estimated from these data using the formulae

$$
\hat{p}_{1}=\frac{a}{m}=\frac{x_{11}}{n_{1}} \quad \text { and } \quad \hat{p}_{2}=\frac{b}{n}=\frac{x_{21}}{n_{2}}
$$

## Odds Ratio

Chances are usually communicated as long-term proportions or probabilities. In betting, chances are often given as odds. For example, the odds of a horse winning a race might be set at 10-to-1 or 3-to-2. Odds can easily be translated into probability. An odds of 3-to-2 means that the event is expected to occur three out of five times. That is, an odds of 3-to-2 (1.5) translates to a probability of winning of 0.60.

The odds of an event are calculated by dividing the event risk by the non-event risk. Thus, in our case of two populations, the odds are

$$
o_{1}=\frac{p_{1}}{1-p_{1}} \text { and } o_{2}=\frac{p_{2}}{1-p_{2}}
$$

For example, if $p_{1}$ is 0.60 , the odds are $0.60 / 0.40=1.5$. Rather than represent the odds as a decimal amount, it is re-scaled into whole numbers. Instead of saying the odds are 1.5 -to- 1 , we say they are 3 -to- 2 .
Another way to compare proportions is to compute the ratio of their odds. The odds ratio of two events is

$$
\psi=\frac{o_{1}}{o_{2}}=\frac{p_{1} /\left(1-p_{1}\right)}{p_{2} /\left(1-p_{2}\right)}=\frac{p_{1} q_{2}}{p_{2} q_{1}}
$$

Although the odds ratio is more complicated to interpret than the risk ratio or risk difference, it is often the parameter of choice in meta-analyses of proportions because the log odds ratio is normally distributed for moderately sized samples.

## Meta-Analysis of Proportions

The formulas presented in this section are given Harrer et al. (2022) pages 404 and 446.
Suppose the individual tables are summarized into one 2-by-2 table whose individual cells now contain the corresponding cell averages which are still named $a, b, c$, and $d$. These counts can be reparametrized using the two group proportions $P 1$ and $P 2$ and the two average-group sample sizes $N 1$ and $N 2$. The average cell counts are then given by

$$
\begin{aligned}
& a=(P 1)(N 1) \\
& b=(P 2)(N 2) \\
& c=(1-P 1)(N 1) \\
& d=(1-P 2)(N 2)
\end{aligned}
$$

Using these summary statistics, an effect size called the log odds ratio can be calculated using

$$
\log (O R)=\log \left(\frac{a d}{b c}\right)
$$

along with an estimate of the within-group variance

$$
V_{W}=\frac{1}{a}+\frac{1}{b}+\frac{1}{c}+\frac{1}{d}
$$

For planning purposes, the between-group variance is estimated as a multiple of the within-group variance as follows

$$
V_{B}=R V_{W}
$$

Hedges and Pigott (2001) recommend using 0.333 for a small value, 0.667 for a medium value, and 1.0 for a large value of $R$. Note that the fixed-effect results is achieved when $R$ is set to 0 .

## I ${ }^{2}$ - Heterogeneity Index

Some authors measure the amount of heterogeneity in the meta-analysis using the heterogeneity index $l^{2}$ (also called the degree of heterogeneity or inconsistency) instead of $R$. This value is related to $R$ as follows. Note that often this value is presented as a percentage.

$$
I^{2}=\frac{V_{B}}{V_{B}+V_{W}}
$$

This value is related to $R$ using the relationship

$$
R=\frac{I^{2}}{1-I^{2}}
$$

## Random-Effects Standard Error

The final random-effects standard error used in the power calculation is

$$
S E_{\log O R}=\sqrt{\frac{V_{W}+V_{B}}{K}}
$$

The meta-analysis z-test is then calculated as

$$
Z=\frac{\log (O R)-\log \left(O R_{0}\right)}{S E_{\log O R}}
$$

which is based on the six parameters $P 1, P 2, N 1, N 2, R$, and $O R_{0}$.
We assume that $Z$ follows the standard normal distribution and so it can be used to test the null hypothesis $H_{0}: O R=O R_{0}$ at significance level of $\alpha$.

## Hypotheses

Three statistical hypotheses commonly used are

$$
\begin{array}{lll}
H_{0}: O R \leq O R_{0} & \text { vs. } & H_{1}: O R>O R_{0} \\
H_{0}: O R \geq O R_{0} & \text { vs. } & H_{1}: O R<O R_{0} \\
H_{0}: O R=O R_{0} & \text { vs. } & H_{1}: O R \neq O R_{0}
\end{array}
$$

## Power Calculation

The power of a one-tailed test that $O R=O R_{0}$ is given by

$$
\text { Power }=1-\Phi\left(c_{\alpha}-\lambda\right)
$$

where $O R$ is set by the alternative hypothesis, $c_{\alpha}$ is the $1-\alpha$ quantile of the standard normal distribution, and

$$
\lambda=\frac{\log (O R)-\log \left(O R_{0}\right)}{S E_{\log O R}}
$$

The power of a two-tailed test that $O R=O R_{0}$ is given by

$$
\text { Power }=1-\Phi\left(c_{\alpha / 2}-\lambda\right)+\Phi\left(-c_{\alpha / 2}-\lambda\right)
$$

These formulas can be used in search algorithms to determine the necessary number of studies required to achieve a desired level of power.

Note that the values of P1.0 (P1 assuming H0) and P1.1 (P1 assuming H1) can be obtained from the odds ratios using

$$
P 1.0=\frac{O R_{0}(P 1 /(1-P 1))}{1+O R_{0}(P 1 /(1-P 1))}
$$

and

$$
P 1.1=\frac{O R_{1}(P 1 /(1-P 1))}{1+O R_{1}(P 1 /(1-P 1))}
$$

where $O R_{1}$ is the odds ratio at which the power is calculated.

## Example 1 - Calculating Sample Size

Researchers wish to determine a reasonable number of studies for a meta-analysis with the following characteristics: two-group, equal-allocation trials, two-sided significance level of 0.05 , a power of 0.9 , average group size of all studies is $25, P 2=0.4, R=0.667, O R_{1}=1.5,1.75,2.0$, and $O R_{0}=1.0$.

## Setup

If the procedure window is not already open, use the PASS Home window to open it. The parameters for this example are listed below and are stored in the Example 1 settings file. To load these settings to the procedure window, click Open Example Settings File in the Help Center or File menu.

| Design Tab |
| :---: |
| Solve For .........................................................Number of Studies |
| Alternative Hypothesis .......................................Two-Sided (H1: OR $=$ OR0) |
| Power.............................................................0.90 |
| Alpha.............................................................0.05 |
| N1 (Average Sample Size of Group 1).................. 25 |
| N2 (Average Sample Size of Group 2)..................N1 |
| OR0 (Odds Ratio\|H0)........................................1.0 |
| OR1 (Odds Ratio\|H1)...........................................1.51.75 2 |
| P2 (Group 2 Proportion)....................................0.4 |
| Calculate Between-Study Variance From ...............R (Between-Study Var I Within-Study Var) |
| R (Between-Study Var / Within-Study Var) ............0.667 |

## Output

Click the Calculate button to perform the calculations and generate the following output.

## Numeric Reports



## References

Borenstein, M., Hedges, L.V., Higgins, J.P.T., and Rothstein, H.R. 2021. Introduction to Meta-Analysis, Second Edition. John Wiley and Sons.
Cooper, J., Hedges, L.V., Valentine, J.C. 2019. The Handbook of Research Synthesis and Meta-Analysis, Third Edition. Russell Sage Foundation.
Griffin, J.W. 2021. 'Calculating statistical power for meta-analysis using metapower', The Quantitative Methods for Psychology, Vol. 17, No. 1, pages 24-39.
Harrer, M., Cuijpers, P., Furukawa, T.A., and Ebert, D.D. 2022. Doing Meta-Analysis with R: A Hands-On Guide. CRC Press.
Hedges, L.V. and Pigott, T.D. 2001. 'The Power of Statistical Tests in Meta-Analysis', Psychological Methods, Volume 6, No. 3, pages 203-217.
Valentine, J.C., Pigott, T.D., and Rothstein, H.R. 2010. 'How Many Studies Do You Need? A Primer on Statistical Power for Meta-Analysis', Journal of Education and Behavioral Statistics, Volume 35, No. 2, pages 215-247.

This report shows the sample sizes for the indicated scenarios.

## Plots Section

## Plots



This plot represents the required number of studies for various values of OR1.

## Example 2 - Validation using Hand Calculations

We could not find a validation example for this procedure in the literature so we will validate the procedure using hand calculations.

Suppose a meta-analysis is contemplated with the following characteristics: two-group, equal-allocation trials, two-sided significance level of $0.05, K=10$, average group size of all studies is $10, P 2=0.5, R=1$, $O R_{1}=1.5$, and $O R_{0}=1.0$. Calculate the power.

Step 1: Compute the value of $P 1$.

$$
P 1=\frac{(O R 1)\left(\frac{P 2}{1-P 2}\right)}{1+(O R 1)\left(\frac{P 2}{1-P 2}\right)}=\frac{1.5}{1+1.5}=0.6
$$

Step 2: Compute the average cell counts.

$$
\begin{aligned}
& a=(P 1)(N 1)=0.6(10)=6 \\
& b=(P 2)(N 2)=0.5(10)=5 \\
& c=(1-P 1)(N 1)=0.4(10)=4 \\
& d=(1-P 2)(N 2)=0.5(10)=5
\end{aligned}
$$

Step 3: Compute the effect size.

$$
\log (O R 1)=\log (1.5)=0.4054651
$$

Step 4: Compute the within-study variance component.

$$
V_{W}=\frac{1}{a}+\frac{1}{b}+\frac{1}{c}+\frac{1}{d}=\frac{1}{6}+\frac{1}{5}+\frac{1}{4}+\frac{1}{5}=0.8166667
$$

Step 5: Compute the between-study variance component.

$$
V_{B}=R\left(V_{W}\right)=0.8166667
$$

Step 6: Compute the standard error of the effect size.

$$
S E_{\log O R}=\sqrt{\frac{V_{W}+V_{B}}{K}}=\sqrt{\frac{0.8166667+0.8166667}{10}}=0.4041452
$$

Step 7: Compute $\lambda$.

$$
\lambda=\frac{\log (O R)-\log \left(O R_{0}\right)}{S E_{\log O R}}=\frac{0.4054651-0}{0.4041452}=1.0032660
$$

Step 8: Compute $c_{\alpha / 2}$.

$$
c_{\alpha / 2}=\Phi(1-0.025)=1.95996398
$$

Step 9: Compute the power.

$$
\begin{aligned}
\text { Power } & =1-\Phi\left(c_{\alpha / 2}-\lambda\right)+\Phi\left(-c_{\alpha / 2}-\lambda\right) \\
& =1-\Phi(1.95996398-1.0032660)+\Phi(-1.95996398-1.0032660) \\
& =1-0.830640146+0.0015221455 \\
& =0.1708820
\end{aligned}
$$

## Setup

If the procedure window is not already open, use the PASS Home window to open it. The parameters for this example are listed below and are stored in the Example 2 settings file. To load these settings to the procedure window, click Open Example Settings File in the Help Center or File menu.

| Design Tab |
| :---: |
| Solve For ..........................................................Power |
| Alternative Hypothesis .......................................Two-Sided (H1: OR $=$ OR0) |
| Alpha.............................................................. 0.05 |
| K (Number of Studies) ...................................... 10 |
| N1 (Average Sample Size of Group 1).................. 10 |
| N2 (Average Sample Size of Group 2)..................N1 |
| OR0 (Odds Ratio\|H0)........................................1.0 |
| OR1 (Odds Ratio\|H1)........................................1.5 |
| P2 (Group 2 Proportion).....................................0.5 |
| Calculate Between-Study Variance From ..............R (Between-Study Var I Within-Study Var) |
| R (Between-Study Var / Within-Study Var) ............ 1 |

## Output

Click the Calculate button to perform the calculations and generate the following output.

| Solve For: <br> Hypotheses: <br> Between-Study Variance Input: |  |  | $\begin{aligned} & \text { Power } \\ & \text { H0: OR = ORO vs. H1: OR }=\text { OR0 } \\ & \text { R (Between-Study / Within-Study Variances) } \end{aligned}$ |  |  |  |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | Average Sample Size Per Study |  |  | Number of |  | Odds Ratio |  | Proportion |  |  | Between-Study Variance Calculation |  | Alpha |
| Power | Group 1 <br> N1 | Group 2 N2 | Total N | Studies K | Subjects KN | $\begin{array}{r} \text { OR\|HO } \\ \text { ORO } \end{array}$ | OR\|H1 OR1 | $\begin{array}{r} \mathrm{P} 1 \mid \mathrm{H} 0 \\ \mathrm{P} 1.0 \end{array}$ | $\begin{array}{r} \mathrm{P} 1 \mid \mathrm{H} 1 \\ \mathrm{P} 1.1 \end{array}$ | Control P2 | R | $\mathrm{I}^{2}$ |  |
| 0.17088 | 10 | 10 | 20 | 10 | 200 | 1 | 1.5 | 0.5 | 0.6 | 0.5 | 1 | 0.5 | 0.05 |

PASS also calculates a power of 0.17088 which validates the procedure.

