

## Chapter 310

# Non-Inferiority Tests for Two Total Variances in a Replicated Design

## Introduction

This procedure calculates power and sample size of tests of non-inferiority total variance (between + within) from a parallel (two-group) design with replicates (repeated measures) for the case when the ratio assumed by the null hypothesis is not necessarily one. This routine deals with the case in which the statistical hypotheses are expressed in terms of the ratio of the total variances.

A parallel design is used to compare two treatment groups by comparing subjects receiving each treatment. In this replicated design, each subject is measured  $M$  times where  $M$  is at least two. To be clear, each subject receives only one treatment, but is measured repeatedly.

Replicated parallel designs such as this are popular because they allow the assessment of total variances, between-subject variances, and within-subject variances.

It is assumed that either there is no carry-over from one measurement to the next, or there is an ample washout period between measurements.

## Technical Details

This procedure uses the formulation given in Chow, Shao, Wang, and Lokhnygina (2018), pages 221 - 224.

Suppose  $x_{ijk}$  is the response of the  $i$ th treatment ( $i = T, C$ ),  $j$ th subject ( $j = 1, \dots, N_i$ ), and  $k$ th replicate ( $k = 1, \dots, M$ ). The model analyzed in this procedure is

$$x_{ijk} = \mu_i + S_{ij} + e_{ijk}$$

where  $\mu_i$  is the treatment effect,  $S_{ij}$  is the random effect of the  $j$ th subject in the  $i$ th treatment, and  $e_{ijk}$  is the within-subject error term which is normally distributed with mean 0 and variance  $V_i = \sigma_{Wi}^2$ .

Unbiased estimates of these variances are given by

$$s_{Wi}^2 = \frac{1}{N_i(M-1)} \sum_{j=1}^{N_i} \sum_{k=1}^M (x_{ijk} - \bar{x}_{ij\cdot})^2, \quad i = T, C$$

where

$$\bar{x}_{ij\cdot} = \frac{1}{M} \sum_{k=1}^M x_{ijk}$$

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Similarly, the between-subject variances are estimated as

$$s_{Bi}^2 = \frac{1}{N_i - 1} \sum_{j=1}^{N_i} (\bar{x}_{ij} - \bar{x}_{i..})^2$$

where

$$\bar{x}_{i..} = \frac{1}{N_i} \sum_{j=1}^{N_i} \bar{x}_{ij}.$$

Now, estimators for the total variance are given by

$$\hat{\sigma}_{Ti}^2 = s_{Bi}^2 + \frac{(M - 1)}{M} s_{Wi}^2$$

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## Testing Variance Non-Inferiority

The following statistical hypotheses are used to test for total variance non-inferiority.

$$H_0: \frac{\sigma_{TT}^2}{\sigma_{TC}^2} \geq R_0 \quad \text{versus} \quad H_1: \frac{\sigma_{TT}^2}{\sigma_{TC}^2} < R_0,$$

where  $R_0$  is the non-inferiority limit.

Let  $\eta = \sigma_{TT}^2 - R_0(\sigma_{TC}^2)$  be the parameter of interest. The test statistic is  $\hat{\eta} = \hat{\sigma}_{TT}^2 - R_0(\hat{\sigma}_{TC}^2)$ .

### Non-Inferiority Test

For the non-inferiority test, compute the limit  $\hat{\eta}_U$  using  $\hat{\eta}_U = \hat{\eta} + \sqrt{\Delta_U}$

Reject the null hypothesis if  $\hat{\eta}_U < 0$ .

The  $\Delta$  is given by

$$\begin{aligned} \Delta_U = & h(1 - \alpha, N_T - 1) s_{BT}^4 + h(\alpha, N_C - 1) R_0^2 s_{BC}^4 + h(\alpha, N_T(M - 1)) \left[ \frac{(M - 1) s_{WT}^2}{M} \right]^2 \\ & + h(1 - \alpha, N_C(M - 1)) \left[ \frac{(M - 1) R_0 s_{WC}^2}{M} \right]^2 \end{aligned}$$

where

$$h(A, B) = \left( 1 - \frac{B}{\chi_{A,B}^2} \right)^2$$

and  $\chi_{A,B}^2$  is the upper quantile of the chi-square distribution with  $B$  degrees of freedom.

## Power

### Non-Inferiority Test

The power of the non-inferiority test is given by

$$\text{Power} = \Phi \left( z_{\alpha} - \frac{(R_1 - R_0)\sigma_{TC}^2}{\sqrt{\sigma^{*2}/n}} \right)$$

where

$$R_1 = \frac{\sigma_{TT}^2}{\sigma_{TC}^2}$$

$$\sigma_{TT}^2 = R_1 \sigma_{TC}^2$$

$$\sigma^{*2} = 2 \left[ \left( \sigma_{BT}^2 + \frac{\sigma_{WT}^2}{M} \right)^2 + R_0^2 \left( \sigma_{BC}^2 + \frac{\sigma_{WC}^2}{M} \right)^2 + \frac{(M-1)\sigma_{WT}^4}{M^2} + \frac{(M-1)R_0^2\sigma_{WC}^4}{M^2} \right]$$

where  $R_1$  is the value of the variance ratio stated by the alternative hypothesis and  $\Phi(x)$  is the standard normal CDF.

A simple binary search algorithm can be applied to the power function to obtain an estimate of the necessary sample size.

## Example 1 – Finding Sample Size

A company has developed a generic drug for treating rheumatism and wants to show that it is non-inferior to the standard drug in terms of total variability. A two-group, parallel design will be used to test the non-inferiority.

Company researchers set the non-inferiority limit to 1.5, the significance level to 0.05, the power to 0.90, M to 2, and the actual variance ratio values between 0.8 and 1.3. They also set  $\sigma^2_{\tau c} = 0.8$ ,  $\sigma^2_{w\tau} = 0.2$ , and  $\sigma^2_{wc} = 0.3$ . They want to investigate the range of required sample size values assuming that the two group sample sizes are equal.

### 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 .....	<b>Sample Size</b>
Power.....	<b>0.90</b>
Alpha.....	<b>0.05</b>
M (Measurements Per Subject) .....	<b>2</b>
R0 (Non-Inferiority Variance Ratio) .....	<b>1.5</b>
R1 (Actual Variance Ratio) .....	<b>0.8 0.9 1 1.1 1.2 1.3</b>
$\sigma^2_{\tau c}$ (Control Variance).....	<b>0.8</b>
$\sigma^2_{w\tau}$ (Treatment Variance) .....	<b>0.2</b>
$\sigma^2_{wc}$ (Control Variance).....	<b>0.3</b>

## Output

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

### Numeric Reports

#### Numeric Results

Solve For: [Sample Size](#)  
 Hypotheses:  $H_0: \sigma^2_{TT}/\sigma^2_{TC} \geq R_0$  vs.  $H_1: \sigma^2_{TT}/\sigma^2_{TC} < R_0$

Power		Sample Size			Measurements per Subject M	Total Variance			Within-Subject Variance		Alpha
						Ratio		Control $\sigma^2_{TC}$	Treatment $\sigma^2_{WT}$	Control $\sigma^2_{WC}$	
Target	Actual	Treatment N <sub>T</sub>	Control N <sub>C</sub>	Total N	Non-Inferiority R <sub>0</sub>	Actual R <sub>1</sub>	Control $\sigma^2_{TC}$				Treatment $\sigma^2_{WT}$
0.9	0.9030	72	72	144	2	1.5	0.8	0.8	0.2	0.3	0.05
0.9	0.9006	104	104	208	2	1.5	0.9	0.8	0.2	0.3	0.05
0.9	0.9005	161	161	322	2	1.5	1.0	0.8	0.2	0.3	0.05
0.9	0.9001	271	271	542	2	1.5	1.1	0.8	0.2	0.3	0.05
0.9	0.9002	521	521	1042	2	1.5	1.2	0.8	0.2	0.3	0.05
0.9	0.9001	1268	1268	2536	2	1.5	1.3	0.8	0.2	0.3	0.05

- Target Power The desired power value entered in the procedure. Power is the probability of rejecting a false null hypothesis.
- Actual Power The actual power achieved. Because N<sub>T</sub> and N<sub>C</sub> are discrete, this value is usually slightly larger than the target power.
- N<sub>T</sub> The number of subjects in the treatment group.
- N<sub>C</sub> The number of subjects in the control group.
- N The total number of subjects. N = N<sub>T</sub> + N<sub>C</sub>.
- M The number of replicates. That is, it is the number of times a treatment measurement is repeated on a subject.
- R<sub>0</sub> The non-inferiority limit for the total variance ratio.
- R<sub>1</sub> The value of the total variance ratio at which the power is calculated.
- $\sigma^2_{TT}$  The total variance of measurements in the treatment group. Note that  $\sigma^2_{TT} = \sigma^2_{BT} + \sigma^2_{WT}$ .
- $\sigma^2_{TC}$  The total variance of measurements in the control group. Note that  $\sigma^2_{TC} = \sigma^2_{BC} + \sigma^2_{WC}$ .
- $\sigma^2_{WT}$  The within-subject variance of measurements in the treatment group.
- $\sigma^2_{WC}$  The within-subject variance of measurements in the control group.
- Alpha The probability of rejecting a true null hypothesis.

#### Summary Statements

A parallel two-group replicated design will be used to test whether the total variance of the treatment ( $\sigma^2_{TT}$ ) is non-inferior to the total variance of the control ( $\sigma^2_{TC}$ ) by testing the total variance ratio ( $\sigma^2_{TT} / \sigma^2_{TC}$ ) against the non-inferiority ratio 1.5 ( $H_0: \sigma^2_{TT} / \sigma^2_{TC} \geq 1.5$  versus  $H_1: \sigma^2_{TT} / \sigma^2_{TC} < 1.5$ ). The comparison will be made using a one-sided, variance-difference test (treatment minus control) as described in Chow, Shao, Wang, and Lohknygina (2018), with a Type I error rate ( $\alpha$ ) of 0.05. Each subject will be measured 2 times. For the control group, the total variance ( $\sigma^2_{TC}$ ) is assumed to be 0.8, and the within-subject variance is assumed to be 0.3. The within-subject variance of the treatment group is assumed to be 0.2. To detect a total variance ratio ( $\sigma^2_{TT} / \sigma^2_{TC}$ ) of 0.8 with 90% power, the number of subjects needed will be 72 in the treatment group, and 72 in the control group.

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## Dropout-Inflated Sample Size

Dropout Rate	Sample Size			Dropout-Inflated Enrollment Sample Size			Expected Number of Dropouts		
	N <sub>T</sub>	N <sub>c</sub>	N	N <sub>T</sub> '	N <sub>c</sub> '	N'	D <sub>T</sub>	D <sub>c</sub>	D
20%	72	72	144	90	90	180	18	18	36
20%	104	104	208	130	130	260	26	26	52
20%	161	161	322	202	202	404	41	41	82
20%	271	271	542	339	339	678	68	68	136
20%	521	521	1042	652	652	1304	131	131	262
20%	1268	1268	2536	1585	1585	3170	317	317	634

Dropout Rate	The percentage of subjects (or items) that are expected to be lost at random during the course of the study and for whom no response data will be collected (i.e., will be treated as "missing"). Abbreviated as DR.
N <sub>T</sub> , N <sub>c</sub> , and N	The evaluable sample sizes at which power is computed. If N <sub>T</sub> and N <sub>c</sub> subjects are evaluated out of the N <sub>T</sub> ' and N <sub>c</sub> ' subjects that are enrolled in the study, the design will achieve the stated power.
N <sub>T</sub> ', N <sub>c</sub> ', and N'	The number of subjects that should be enrolled in the study in order to obtain N <sub>T</sub> , N <sub>c</sub> , and N evaluable subjects, based on the assumed dropout rate. After solving for N <sub>T</sub> and N <sub>c</sub> , N <sub>T</sub> ' and N <sub>c</sub> ' are calculated by inflating N <sub>T</sub> and N <sub>c</sub> using the formulas $N_{T'} = N_T / (1 - DR)$ and $N_{c'} = N_c / (1 - DR)$ , with N <sub>T</sub> ' and N <sub>c</sub> ' always rounded up. (See Julious, S.A. (2010) pages 52-53, or Chow, S.C., Shao, J., Wang, H., and Lohknygina, Y. (2018) pages 32-33.)
D <sub>T</sub> , D <sub>c</sub> , and D	The expected number of dropouts. $D_T = N_{T'} - N_T$ , $D_c = N_{c'} - N_c$ , and $D = D_T + D_c$ .

## Dropout Summary Statements

Anticipating a 20% dropout rate, 90 subjects should be enrolled in Group 1, and 90 in Group 2, to obtain final group sample sizes of 72 and 72, respectively.

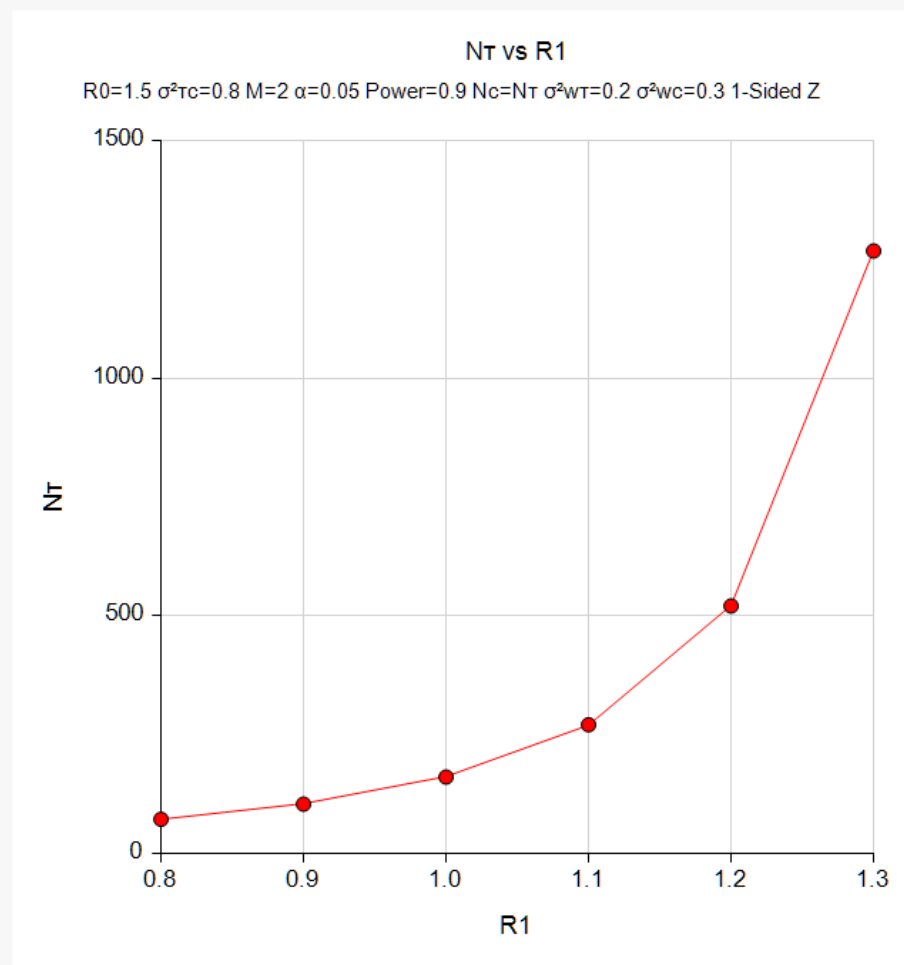
## References

- Chow, S.C., Shao, J., Wang, H., and Lohknygina, Y. 2018. Sample Size Calculations in Clinical Research, Third Edition. Taylor & Francis/CRC. Boca Raton, Florida.
- Chow, S.C., and Liu, J.P. 2014. Design and Analysis of Clinical Trials, Third Edition. John Wiley & Sons. Hoboken, New Jersey.

This report gives the sample sizes for the indicated scenarios.

## Plots Section

## Plots



This plot shows the relationship between sample size and R<sub>1</sub>.

## Example 2 – Validation using Chow et al. (2018)

We will use an example from Chow *et al.* (2018) pages 223-224 to validate this procedure.

In this example,  $R_0 = 1.21$ , significance level = 0.05,  $M = 3$ ,  $R_1 = 0.52$ ,  $\sigma^2_{\tau c} = 0.25$ ,  $\sigma^2_{\tau t} = 0.04$ ,  $\sigma^2_{w c} = 0.09$ . The problem is to find the sample size for the lower, one-sided test (note that this is a non-inferiority test). They find the per group sample size to be 28.

### 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 .....	<b>Sample Size</b>
Power.....	<b>0.80</b>
Alpha.....	<b>0.05</b>
M (Measurements Per Subject) .....	<b>3</b>
R0 (Non-Inferiority Variance Ratio).....	<b>1.21</b>
R1 (Actual Variance Ratio) .....	<b>0.52</b>
$\sigma^2_{\tau c}$ (Control Variance).....	<b>0.25</b>
$\sigma^2_{\tau t}$ (Treatment Variance) .....	<b>0.04</b>
$\sigma^2_{w c}$ (Control Variance).....	<b>0.09</b>

### Output

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

Numeric Results												
Solve For: <a href="#">Sample Size</a>												
Hypotheses: $H_0: \sigma^2_{\tau t} / \sigma^2_{\tau c} \geq R_0$ vs. $H_1: \sigma^2_{\tau t} / \sigma^2_{\tau c} < R_0$												
Power		Sample Size			Measurements per Subject M	Total Variance Ratio			Within-Subject Variance		Alpha	
Target	Actual	Treatment N <sub>t</sub>	Control N <sub>c</sub>	Total N		Non-Inferiority R <sub>0</sub>	Actual R <sub>1</sub>	Control $\sigma^2_{\tau c}$	Treatment $\sigma^2_{\tau t}$	Control $\sigma^2_{w c}$		
0.8	0.8044	28	28	56	3	1.21	0.52	0.25	0.04	0.09	0.05	

The sample size of 28 per group matches Chow et al. (2018).