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Chapter 478

Group-Sequential Non-Inferiority Tests for Two Means (Simulation) (Legacy)

This procedure uses simulation for the calculation of the boundaries as well as for calculation of power (and sample size). Futility boundaries are limited. Two test statistics and a variety of simulated distributions are available.

Introduction

This procedure can be used to determine power, sample size and/or boundaries for group sequential non-inferiority tests comparing the means of two groups. The common two-sample T-Test and the Mann-Whitney U test can be simulated in this procedure. Significance and futility boundaries can be produced. The spacing of the looks can be equal or custom specified. Boundaries can be computed based on popular alpha- and beta-spending functions (O'Brien-Fleming, Pocock, Hwang-Shih-DeCani Gamma family, linear) or custom spending functions. Boundaries can also be input directly to verify alpha- and/or beta-spending properties. Futility boundaries can be binding or non-binding. Maximum and average (expected) sample sizes are reported as well as the alpha and/or beta spent and incremental power at each look. Corresponding P-Value boundaries are also given for each boundary statistic. Plots of boundaries are also produced.

Technical Details

This section outlines many of the technical details of the techniques used in this procedure including the simulation summary, the test statistic details, and the use of spending functions.

An excellent text for the background and details of many group-sequential methods is Jennison and Turnbull (2000).

Simulation Procedure

In this procedure, a large number of simulations are used to calculate boundaries and power using the following steps:

- 1. Based on the specified distributions, random samples of size N1 and N2 are generated under the null distribution and under the alternative distribution. These are simulated samples as though the final look is reached.
- 2. For each sample, test statistics for each look are produced. For example, if N1 and N2 are 100 and there are 5 equally spaced looks, test statistics are generated from the random samples at N1 = N2 = 20, N1 = N2 = 40, N1 = N2 = 60, N1 = N2 = 80, and N1 = N2 = 100 for both null and alternative samples.

- 3. To generate the first significance boundary, the null distribution statistics of the first look (e.g., at N1 = N2 = 20) are ordered and the percent of alpha to be spent at the first look is determined (using either the alpha-spending function or the input value). The statistic for which the percent of statistics above (or below, as the case may be) that value is equal to the percent of alpha to be spent at the first look is the boundary statistic. It is seen here how important a large number of simulations is to the precision of the boundary estimates.
- 4. All null distribution samples that are outside the first significance boundary at the first look are removed from consideration for the second look. If binding futility boundaries are also being computed, all null distribution samples with statistics that are outside the first futility boundary are also removed from consideration for the second look. If non-binding futility boundaries are being computed, null distribution samples with statistics outside the first futility boundary are not removed.
- 5. To generate the second significance boundary, the remaining null distribution statistics of the second look (e.g., at N1 = N2 = 40) are ordered and the percent of alpha to be spent at the second look is determined (again, using either the alpha-spending function or the input value). The percent of alpha to be spent at the second look is multiplied by the total number of simulations to determine the number of the statistic that is to be the second boundary statistic. The statistic for which that number of statistics is above it (or below, as the case may be) is the second boundary statistic. For example, suppose there are initially 1000 simulated samples, with 10 removed at the first look (from, say, alpha spent at Look 1 equal to 0.01), leaving 990 samples considered for the second look. Suppose further that the alpha to be spent at the second look is 0.02. This is multiplied by 1000 to give 20. The 990 still-considered statistics are ordered and the 970th (20 in from 990) statistic is the second boundary.
- 6. All null distribution samples that are outside the second significance boundary and the second futility boundary, if binding, at the second look are removed from consideration for the third look (e.g., leaving 970 statistics computed at N1 = N2 = 60 to be considered at the third look). Steps 4 and 5 are repeated until the final look is reached.

Futility boundaries are computed in a similar manner using the desired beta-spending function or custom beta-spending values and the alternative hypothesis simulated statistics at each look. For both binding and non-binding futility boundaries, samples for which alternative hypothesis statistics are outside either the significance or futility boundaries of the previous look are excluded from current and future looks.

Because the final futility and significance boundaries are required to be the same, futility boundaries are computed beginning at a small value of beta (e.g., 0.0001) and incrementing beta by that amount until the futility and significance boundaries meet.

When boundaries are entered directly, this procedure uses the null hypothesis and alternative hypothesis simulations to determine the number of test statistics that are outside the boundaries at each look. The cumulative proportion of alternative hypothesis statistics that are outside the significance boundaries is the overall power of the study.

Generating Random Distributions

Two methods are available in **PASS** to simulate random samples. The first method generates the random variates directly, one value at a time. The second method generates a large pool (over 10,000) of random values and then draws the random numbers from this pool. This second method can cut the running time of the simulation by 70%.

As mentioned above, the second method begins by generating a large pool of random numbers from the specified distributions. Each of these pools is evaluated to determine if its mean is within a small relative tolerance (0.0001) of the target mean. If the actual mean is not within the tolerance of the target mean, individual members of the population are replaced with new random numbers if the new random number moves the mean towards its target. Only a few hundred such swaps are required to bring the actual mean to within tolerance of the target mean. This population is then sampled with replacement using the uniform distribution. We have found that this method works well as long as the size of the pool is the maximum of twice the number of simulated samples desired and 10,000.

The Statistical Hypotheses

This section will review the specifics of non-inferiority testing.

Remember that in the usual t-test setting, the null (H_0) and alternative (H_1) hypotheses for one-sided tests are defined as

$$H_0: \mu_1 - \mu_2 \le D$$
 versus $H_1: \mu_1 - \mu_2 > D$

Rejecting this test implies that the mean difference is larger than the value *D*. This test is called an *upper-tailed test* because it is rejected in samples in which the difference between the sample means is larger than *D*.

Following is an example of a *lower-tailed test*.

$$H_0$$
: $\mu_1 - \mu_2 \ge D$ versus H_1 : $\mu_1 - \mu_2 < D$

Non-inferiority tests are special cases of the above directional tests. It will be convenient to adopt the following specialized notation for the discussion of these tests.

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Group-Sequential Non-Inferiority Tests for Two Means (Simulation) (Legacy)

<u>Parameter</u>	PASS Input/Output	<u>Interpretation</u>
μ_1	Mean1	<i>Mean</i> of population 1. Population 1 is assumed to consist of those who have received the new treatment.
μ_2	Mean2	<i>Mean</i> of population 2. Population 2 is assumed to consist of those who have received the reference treatment.
M_{NI}	NIM	Margin of non-inferiority. This is a tolerance value that defines the magnitude of the amount that is not of practical importance. This may be thought of as the largest change from the baseline that is considered to be trivial. The absolute value is shown to emphasize that this is a magnitude. The sign of the value will be determined by the specific design that is being used.
δ	D	<i>True difference</i> . This is the value of $\mu_1 - \mu_2$, the difference between the means. This is the value at which the power is calculated.

Note that the actual values of μ_1 and μ_2 are not needed. Only their difference is needed for power and sample size calculations.

Non-Inferiority Tests

A *non-inferiority test* tests that the treatment mean is not worse than the reference mean by more than the equivalence margin. The actual direction of the hypothesis depends on the response variable being studied.

Case 1: High Values Good, Non-Inferiority Test

In this case, higher values are better. The hypotheses are arranged so that rejecting the null hypothesis implies that the treatment mean is no less than a small amount below the reference mean. The value of δ is often set to zero. The following are equivalent sets of hypotheses.

$$H_0: \mu_1 \ge \mu_2 + |M_{NI}|$$
 versus $H_1: \mu_1 < \mu_2 + |M_{NI}|$ $H_0: \mu_1 - \mu_2 \ge |M_{NI}|$ versus $H_1: \mu_1 - \mu_2 < |M_{NI}|$ $H_0: \delta \ge |M_{NI}|$ versus $H_1: \delta < |M_{NI}|$

Case 2: High Values Bad, Non-Inferiority Test

In this case, lower values are better. The hypotheses are arranged so that rejecting the null hypothesis implies that the treatment mean is no more than a small amount above the reference mean. The value of δ is often set to zero. The following are equivalent sets of hypotheses.

$$\begin{split} &H_0 \colon \mu_1 \leq \mu_2 - |M_{NI}| & \text{versus} & H_1 \colon \mu_1 > \mu_2 - |M_{NI}| \\ &H_0 \colon \mu_1 - \mu_2 \leq -|M_{NI}| & \text{versus} & H_1 \colon \mu_1 - \mu_2 > -|M_{NI}| \\ &H_0 \colon \delta \leq -|M_{NI}| & \text{versus} & H_1 \colon \delta > -|M_{NI}| \end{split}$$

Test Statistics

This section describes the test statistics that are available in this procedure.

Two-Sample T-Test

The two-sample t-test assumes that the data are a simple random sample from a population of normally distributed values that all have the same mean and variance. This assumption implies that the data are continuous, and their distribution is symmetric. The calculation of the t statistic is as follows

$$t_{df} = \frac{(\bar{X}_1 - \bar{X}_2) - (\mu_1 - \mu_2)}{s_{\bar{X}_1 - \bar{X}_2}}$$

where

$$\bar{X}_k = \frac{\sum_{i=1}^{N_k} X_{ki}}{N_k}$$

$$s_{\bar{X}_1 - \bar{X}_2} = \sqrt{\frac{\sum_{i=1}^{N_1} (X_{1i} - \bar{X}_1)^2 + \sum_{i=1}^{N_2} (X_{2i} - \bar{X}_2)^2}{N_1 + N_2 - 2}} \left(\frac{1}{N_1} + \frac{1}{N_2}\right)$$

$$df = N_1 + N_2 - 2$$

The significance of the test statistic is determined by computing the p-value based on the t distribution with degrees of freedom df. If this p-value is less than a specified level (often 0.05), the null hypothesis is rejected. Otherwise, no conclusion can be reached.

Mann-Whitney U Test

This test is the nonparametric substitute for the equal-variance t-test. Two key assumptions for this test are that the distributions are at least ordinal and that they are identical under H0. This implies that ties (repeated values) are not acceptable. When ties are present, the approximation provided can be used, but know that the theoretic results no longer hold.

The Mann-Whitney test statistic is defined as follows in Gibbons (1985).

$$z = \frac{W_1 - \frac{N_1(N_1 + N_2 + 1)}{2} + C}{s_W}$$

where

$$W_1 = \sum_{k=1}^{N_1} Rank(X_{1k})$$

The ranks are determined after combining the two samples. The standard deviation is calculated as

$$s_W = \sqrt{\frac{N_1 N_2 (N_1 + N_2 + 1)}{12} - \frac{N_1 N_2 \sum_{i=1} (t_i^3 - t_i)}{12 (N_1 + N_2)(N_1 + N_2 - 1)}}$$

where t_1 is the number of observations tied at value one, t_2 is the number of observations tied at some value two, and so forth.

The correction factor, C, is 0.5 if the rest of the numerator of Z is negative or -0.5 otherwise. The value of Z is then compared to the standard normal distribution.

Spending Functions

Spending functions can be used in this procedure to specify the proportion of alpha or beta that is spent at each look without having to specify the proportion directly.

Spending functions have the characteristics that they are increasing and that

$$\alpha(0) = 0$$

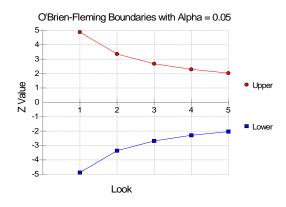
$$\alpha(1) = \alpha$$

The last characteristic guarantees a fixed α level when the trial is complete. This methodology is very flexible since neither the times nor the number of analyses must be specified in advance. Only the functional form of $\alpha(\tau)$ must be specified.

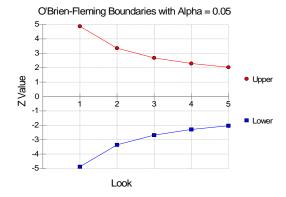
PASS provides several popular spending functions plus the ability to enter and analyze your own percents of alpha or beta spent. These are calculated as follows (beta may be substituted for alpha for beta-spending functions):

Hwang-Shih-DeCani (gamma family) $\alpha \left[\frac{1-e^{-\gamma t}}{1-e^{-\gamma}}\right], \gamma \neq 0; \quad \alpha t, \gamma = 0$

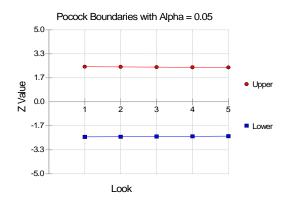
$$\alpha\left[\frac{1-e^{-\gamma t}}{1-e^{-\gamma}}\right], \gamma\neq 0; \ \alpha t, \gamma=0$$



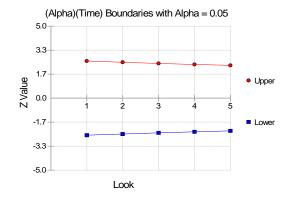
2. O'Brien-Fleming Analog $2-2\Phi\left(\frac{Z_{\alpha/2}}{\sqrt{t}}\right)$



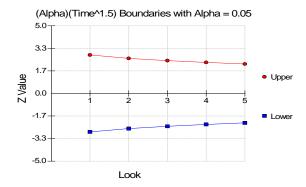
3. Pocock Analog $\alpha \cdot \ln(1 + (e-1)t)$



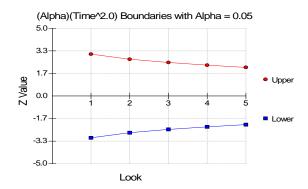
4. Alpha * time $\alpha \cdot t$



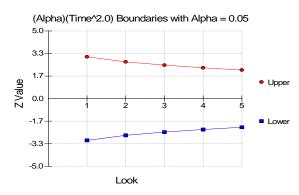
5. Alpha * time^1.5 $\alpha \cdot t^{3/2}$



6. Alpha * time^2 $\alpha \cdot t^2$



7. Alpha * time^C $\alpha \cdot t^C$



8. User Supplied Percents

A custom set of percents of alpha to be spent at each look may be input directly.

The O'Brien-Fleming Analog spends very little alpha or beta at the beginning and much more at the final looks. The Pocock Analog and (Alpha or Beta)(Time) spending functions spend alpha or beta more evenly across the looks. The Hwang-Shih-DeCani (C) (gamma family) spending functions and (Alpha or Beta)(Time^C) spending functions are flexible spending functions that can be used to spend more alpha or beta early or late or evenly, depending on the choice of C.

Example 1 – Power and Output

A clinical trial is to be conducted over a two-year period to compare the mean response of a new treatment to that of the current treatment. The current response mean is 108. The researchers would like to determine if the new treatment is not inferior to the current treatment, as it has fewer side effects and is less expensive. The new treatment is considered non-inferior if it has a mean greater than 103. Although the researchers do not know the true mean of the new treatment, they would like to examine the power that is achieved if the mean of the new treatment is also 108. The standard deviation for both groups is assumed to be 20. The sample size at the final look is to be 200 per group. Testing will be done at the 0.05 significance level. A total of five tests are going to be performed on the data as they are obtained. The O'Brien-Fleming (Analog) boundaries will be used

Find the power and test boundaries assuming equal sample sizes per arm.

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.

Solve For	Power
Test Type	T-Test
Higher Means Are	Better
Simulations	20000
Random Seed	3662304 (for Reproducibility)
Alpha	0.05
Group Allocation	Equal (N1 = N2)
Sample Size Per Group	200
Mean1 (Mean of Group 1, Control)	108
NIM (Non-Inferiority Margin)	5
% (NIM as Percent of Mean1)	Unchecked
Mean2 (Mean of Group 2, Treatment)	108
Standard Deviation	20
Looks & Boundaries Tab	
Specification of Looks and Boundaries	Simple
Number of Equally Spaced Looks	•
Alpha Spending Function	
Type of Futility Boundary	

Output

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

Scenario 1 Numeric Results for Group Sequential Test of Non-Inferiority

Solve For: Power Higher Means Are: Better

Hypotheses: H0: Mean1 - Mean2 = NIM; H1: Mean1 - Mean2 < NIM

Test Statistic: T-Test

Alpha-Spending Function: O'Brien-Fleming Analog

Beta-Spending Function: None Futility Boundary Type: None Number of Looks: 5 Simulations: 20000 Pool Size: 40000

Random Seed: 3662304 (User-Entered)

Numeric Summary for Scenario 1

	Power				Alpha		
Value	95% LCL	95% UCL	Target	Actual	95% LCL	95% UCL	Beta
0.7982	0.7926	0.8038	0.05	0.05	0.04698	0.05302	0.2018

		A	Average S	ample Siz	:e					
		Give	n H0	Give	en H1					
N1	N2	Grp1	Grp2	Grp1	Grp2	Non-Inf. Margin	Mean1	Mean2	Std Dev	
200	200	198	198	156	156	5	108	108	20	
Power			pro			a false null hypo ypothesis simul				
Power	95% LCL	and UCL			upper confid	ence limits for tl imulations.	ne power estin	mate. The wid	th of the inte	erval is
arget	Alpha		The	user-specif	ied probabil	ty of rejecting a	true null hypo	thesis. It is th	e total alpha	a spent.
Actual	Alpha					tually achieved tions that are o				ion of
Alpha 9	95% LCL	and UCL				ence limits for thur the second to the second to the second the second to the second the		a estimate. Th	ne width of the	he
Beta				. ,		a false null hyp at do not cross t				native
N1 and	l N2		The	sample size	es of each g	roup if the study	reaches the	final look.		
Averag	je Sample	e Size Given	pro	-	•	ample sizes of e sis simulations	0 1			
Averag	je Sample	e Size Given	pro		alternative h	ample sizes of e ypothesis simul				
	feriority M , Mean2,	largin and Std Dev	' The			ol mean that is et by the user to				on

Summary Statements

A group sequential trial with group sample sizes of 200 and 200 at the final look achieves 80% power to detect a difference of 5 at the 0.05 significance level (alpha) using a one-sided T-Test.

Accumulated Information Details for Scenario 1

	Accumulated Information	Accumu	Accumulated Sample Size		
Look	Percent	Group 1	Group 2	Total	
1	20	40	40	80	
2	40	80	80	160	
3	60	120	120	240	
4	80	160	160	320	
5	100	200	200	400	

Look

Accumulated Information Percent Accumulated Sample Size Group 1

Accumulated Sample Size Group 2 Accumulated Sample Size Total The number of the look.

The percent of the sample size accumulated up to the corresponding look.

The total number of individuals in group 1 at the corresponding look.

The total number of individuals in group 2 at the corresponding look.

The total number of individuals in the study (group 1 + group 2) at the corresponding look.

Boundaries for Scenario 1

	Significanc	e Boundary
Look	T-Value Scale	P-Value Scale
1	-4.02184	0.00007
2	-2.86704	0.00235
3	-2.33275	0.01025
4	-1.95629	0.02565
5	-1.73540	0.04172

Look

Significance Boundary T-Value Scale

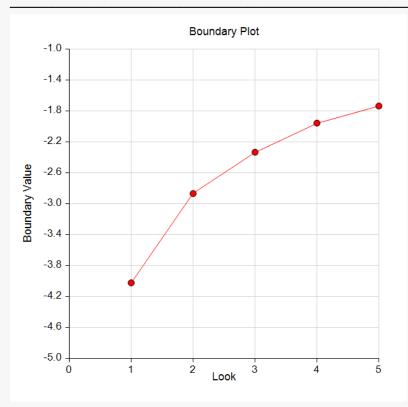
The number of the look.

The value such that statistics outside this boundary at the corresponding look indicate termination of the study and rejection of the null hypothesis. They are sometimes called efficacy boundaries.

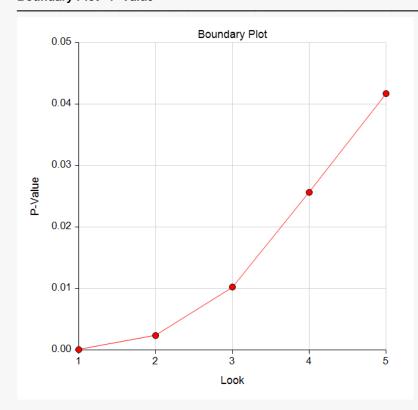
Significance Boundary P-Value Scale

The value such that P-Values outside this boundary at the corresponding look indicate termination of the study and rejection of the null hypothesis. This P-Value corresponds to the T-Value Boundary and is sometimes called the nominal alpha.

Boundary Plot



Boundary Plot - P-Value



Significance Boundaries with 95% Simulation Confidence Intervals for Scenario 1

	T-\	Value Bounda	ary	P	-Value Bound	ary	
Look	Value	95% LCL	95% UCL	Value	95% LCL	95% UCL	
1	-4.02184			0.00007			
2	-2.86704	-2.96536	-2.79875	0.00235	0.00175	0.00288	
3	-2.33275	-2.40599	-2.29943	0.01025	0.00845	0.01117	
4	-1.95629	-1.99095	-1.93026	0.02565	0.02367	0.02723	
5	-1.73540	-1.76698	-1.70882	0.04172	0.03900	0.04413	
Look		The number	of the look.				
T-Value	Boundary Value						indicate terminatior efficacy boundaries
P-Value	Boundary Value	of the stud		the null hypoth	nesis. Ťhis P-Vali	rresponding look ue corresponds to	indicate termination the T-Value

95% LCL and UCL The lower and

The lower and upper confidence limits for the boundary at the given look. The width of the interval is based on the number of simulations.

Alpha-Spending and Null Hypothesis Simulation Details for Scenario 1

			Tar	get	Ac	tual		
Look	Signif. B T-Value Scale	P-Value Scale	Spending Function Alpha	Cum. Spending Function Alpha	Alpha Spent	Cum. Alpha Spent	Proportion H1 Sims Outside Signif. Boundary	Cum. H1 Sims Outside Signif. Boundary
1	-4.02184	0.00007	0.00001	0.00001	0.00000	0.00000	0.0026	0.0026
2	-2.86704	0.00235	0.00193	0.00194	0.00195	0.00195	0.1052	0.1078
3	-2.33275	0.01025	0.00945	0.01140	0.00945	0.01140	0.2496	0.3573
4	-1.95629	0.02565	0.01703	0.02843	0.01705	0.02845	0.2663	0.6236
5	-1.73540	0.04172	0.02157	0.05000	0.02155	0.05000	0.1746	0.7982
Significa Spendir Cumula Alpha S Cumula	tive Alpha Spe	y P-Value Sca pha Function Alph ent	le	The value correspond the null head the null head the alpha the sum look. The propond statistics thought of the propond statistics thought of the propond signification of the propond statistics thought of the sum the propond signification of the propond signification of the sum the propond signification of the sum the sum the propond signification of the sum the number of the sum the number of the sum the sum the number of the sum the number of th	nding look ind hypothesis. The such that P-V nding look independent of the such as a spending furtion of the Spending furtion of the number Significant of the Alpha Station of the Alpha Station of the such as the incredition of the althoutside the Spending of the Alpha Station of the althoutside the Spending of the Alpha Station of the althoutside the Spending of the althoutside the Spending of the althout Spending Spendin	tistics outside dicate termina ney are somet 'alues outside dicate termina nis P-Value cod is sometimes allocate nction. It did alpha allocate nction. It hypothesis so be Boundary a lill hypothesis so termination to be remental power termative hypothesis termative hypothesis termative hypothesis termative hypothesis sometimes allocate not the secondary between the power termative hypothesis termative hypothesis termative hypothesis sometimes allocate not the secondary between the power termination to the termative hypothesis sometimes allocate not the secondary between the secondary be	simulations result up to and includin e corresponding I thesis simulations oundary at this lo	and rejection of acy boundaries. the and rejection of Significance hal alpha. It is rresponding and in statistics are given in general sook. It is ook. It is ook. It may be a resulting in general sook. It may be a resulting in general sook. It may be a resulting in general sook. It is ook. It is

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Group-Sequential Non-Inferiority Tests for Two Means (Simulation) (Legacy)

Scenario 1 Numeric Results for Group Sequential Test of Non-Inferiority

Solve For: Power Higher Means Are: Better

Hypotheses: H0: Mean1 - Mean2 = NIM; H1: Mean1 - Mean2 < NIM

Test Statistic: T-Test

Alpha-Spending Function: O'Brien-Fleming Analog

Beta-Spending Function: None Futility Boundary Type: None Number of Looks: 5 Simulations: 20000 Pool Size: 40000

Random Seed: 3662304 (User-Entered)

Numeric Summary of Scenarios

					Non-Inf.			Std	
Scenario	Power	N1	N2	Alpha	Margin	Mean1	Mean2	Dev	
1	0.7982	200	200	0.05	5	108	108	20	
Power			•	, , .	a false null hypesimulations that				
Alpha			•		ctually achieved as that are outsi	, ,			ion of th
N1 and N2		The	sample si	zes of each g	group if the stud	y reaches the	final look.		
Non-inferiorit	y margin	The	distance f	rom the conti	rol mean that is	still considere	ed non-inferior		
Meant Mea	n2 and Std Da	av The	naramete	re that ware c	eat by the user t	o define the n	ull and alterna	ativa cimulati	ion

The parameters that were set by the user to define the null and alternative simulation Mean1, Mean2, and Std Dev distributions.

Power and Alpha Summary

		Power				Alpha		
Scenario	Value	95% LCL	95% UCL	Target	Actual	95% LCL	95% UCL	Beta
1	0.7982	0.7926	0.8038	0.05	0.05	0.04698	0.05302	0.2018
Power						one of the looks.		portion of
Power 95% L	CL and UCL		nd upper confid		or the power of	estimate. The wi	dth of the interva	al is based
Target Alpha		The user-sp	ecified probabili	ty of rejecting	g a true null h	hypothesis. It is t	the total alpha sp	ent.
Alpha or Actu	ual Alpha			,	, ,	periment. It is the ance boundaries		of the null
Alpha 95% L	CL and UCL		nd upper confid the number of s		or the actual a	alpha estimate.	The width of the	interval is
Beta						is the total prop cance boundarie		tive

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Group-Sequential Non-Inferiority Tests for Two Means (Simulation) (Legacy)

Sample Size Summary

					,	Average S	ample Siz	e	
					Give	en H0	Give	en H1	
Scenario	Power	Alpha	N1	N2	Grp1	Grp2	Grp1	Grp2	
1	0.7982	0.05	200	200	198	198	156	156	
Power Alpha		pr bo	oportion of oundaries.	f alternativ	e hypothes	is simulatior	ns that are o	outside the	It is the total significance total proportion of
•		the	e null hype	othesis sin	nulations tha	at are outsid	le the signifi	cance bou	ındaries .
				. ,			-		indanes.
N1 and N2 Average Sam	nple Size Give	The n H0 The pr	sample si average of	or expecte of null hypo		ne study rea zes of each	ches the fin group if H0	al look. is true. Th	nese are based on the or futility boundaries

Run Time: 12.19 seconds.

References

Jennison, C., Turnbull, B.W. 2000. Group Sequential Methods with Applications to Clinical Trials. Chapman & Hall. Boca Raton, FL.

Devroye, Luc. 1986. Non-Uniform Random Variate Generation. Springer-Verlag. New York.

Matsumoto, M. and Nishimura, T. 1998. 'Mersenne twister: A 623-dimensionally equidistributed uniform pseudorandom number generator.' ACM Trans. On Modeling and Computer Simulations.

Zar, Jerrold H. 1984. Biostatistical Analysis (Second Edition). Prentice-Hall. Englewood Cliffs, New Jersey.

The values obtained from any given run of this example will vary slightly due to the variation in simulations.

Example 2 – Power for One-Sided Test with Futility Boundaries

Suppose researchers would like to compare two treatments with a non-inferiority test at each look. Further, suppose they would like to terminate the study early when it can be deemed highly unlikely that the new treatment is non-inferior to the standard. Suppose the control group mean is 108. The researchers wish to know the power of the test if the treatment group mean is also 108. The sample size at the final look is to be 200 per group. Testing will be done at the 0.05 significance level. A total of five tests are going to be performed on the data as they are obtained. The O'Brien-Fleming (Analog) boundaries will be used for both significance and futility boundaries.

Find the power and test boundaries assuming equal sample sizes per arm.

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.

Solve For	Power
Test Type	T-Test
Higher Means Are	Better
Simulations	20000
Random Seed	3730522 (for Reproducibility)
Alpha	0.05
Group Allocation	Equal (N1 = N2)
Sample Size Per Group	200
Mean1 (Mean of Group 1, Control)	108
NIM (Non-Inferiority Margin)	5
% (NIM as Percent of Mean1)	Unchecked
Mean2 (Mean of Group 2, Treatment)	108
Standard Deviation	20
Looks & Boundaries Tab	
Specification of Looks and Boundaries	Simple
Number of Equally Spaced Looks	5
Alpha Spending Function	O'Brien-Fleming Analog
Type of Futility Boundary	Non-binding
Number of Skipped Futility Looks	0
Beta Spending Function	O'Brien-Fleming Analog

Output

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

Scenario 1 Numeric Results for Group Sequential Test of Non-Inferiority

Solve For: Power Higher Means Are: Better

Hypotheses: H0: Mean1 - Mean2 = NIM; H1: Mean1 - Mean2 < NIM

Test Statistic: T-Test

Alpha-Spending Function: O'Brien-Fleming Analog Beta-Spending Function: O'Brien-Fleming Analog

Futility Boundary Type: Non-Binding

Number of Looks: 5 Simulations: 20000 Pool Size: 40000

Random Seed: 3730522 (User-Entered)

Numeric Summary for Scenario 1

	Power				Alpha		
Value	95% LCL	95% UCL	Target	Actual	95% LCL	95% UCL	Beta
0.723	0.7168	0.7292	0.05	0.0409	0.03816	0.04364	0.277

		1	Average S	ample Siz	е				
		Give	en H0	Give	en H1	Non-Inf.			Std
N1	N2	Grp1	Grp2	Grp1	Grp2	Margin	Mean1	Mean2	Dev
200	200	100	100	139	139	5	108	108	20

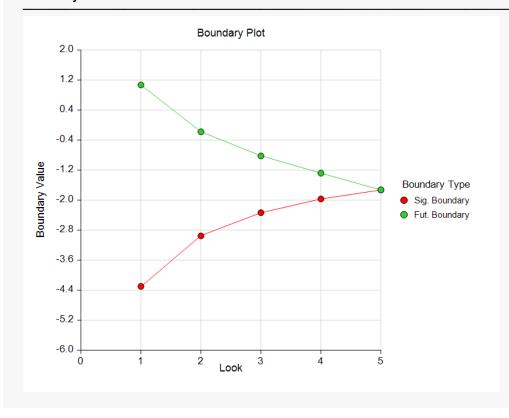
Accumulated Information Details for Scenario 1

	Accumulated Information	Accumu	ulated Sample	Size
Look	Percent	Group 1	Group 2	Total
1	20	40	40	80
2	40	80	80	160
3	60	120	120	240
4	80	160	160	320
5	100	200	200	400

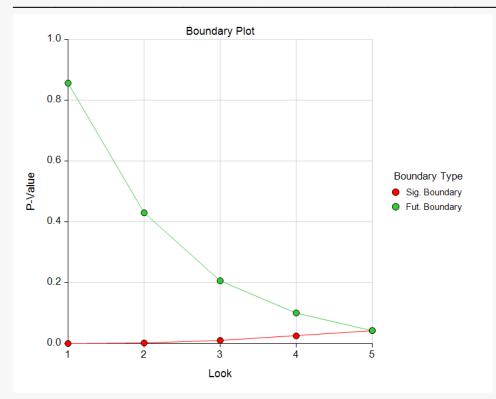
Boundaries for Scenario 1

	Significance	e Boundary	Futility B	oundary
Look	T-Value Scale	P-Value Scale	T-Value Scale	P-Value Scale
1	-4.29731	0.00002	1.07113	0.85629
2	-2.95183	0.00182	-0.17706	0.42984
3	-2.33320	0.01024	-0.82082	0.20628
4	-1.96622	0.02507	-1.28003	0.10073
5	-1.72660	0.04251	-1.72660	0.04251

Boundary Plot



Boundary Plot - P-Value



Significance Boundaries with 95% Simulation Confidence Intervals for Scenario 1

	T.	-Value Bounda	ary	P	-Value Bound	ary
Look	Value	95% LCL	95% UCL	Value	95% LCL	95% UCL
1	-4.29731			0.00002		
2	-2.95183	-3.10440	-2.87850	0.00182	0.00113	0.00227
3	-2.33320	-2.38824	-2.28687	0.01024	0.00886	0.01154
4	-1.96622	-2.00044	-1.93076	0.02507	0.02315	0.02720
5	-1.72660	-1.75402	-1.69822	0.04251	0.04010	0.04512

Futility Boundaries with 95% Simulation Confidence Intervals for Scenario 1

	T-	-Value Bounda	ary	P	-Value Bound	ary
Look	Value	95% LCL	95% UCL	Value	95% LCL	95% UCL
1	1.07113	1.03263	1.11477	0.85629		
2	-0.17706	-0.20336	-0.14771	0.42984	0.41956	0.44138
3	-0.82082	-0.84118	-0.79968	0.20628	0.20055	0.21235
4	-1.28003	-1.30311	-1.25551	0.10073	0.09674	0.10511
5	-1.72660	-1.74948	-1.69655	0.04251	0.04049	0.04528

Alpha-Spending and Null Hypothesis Simulation Details for Scenario 1

			Tai	rget	Ac	tual	Duamantian	C
	Signif. B	oundary	Spending	Cum. Spending		Cum.	Proportion H0 Sims Outside	Cum. H0 Sims Outside
Look	T-Value Scale	P-Value Scale	Function Alpha	Function Alpha	Alpha Spent	Alpha Spent	Futility Boundary	Futility Boundary
1	-4.29731	0.00002	0.00001	0.00001	0.00000	0.00000	0.14780	0.14780
2	-2.95183	0.00182	0.00193	0.00194	0.00195	0.00195	0.43225	0.58005
3	-2.33320	0.01024	0.00945	0.01140	0.00945	0.01140	0.23325	0.81330
4	-1.96622	0.02507	0.01703	0.02843	0.01620	0.02760	0.10290	0.91620
5	-1.72660	0.04251	0.02157	0.05000	0.01330	0.04090	0.04290	0.95910

Beta-Spending and Alternative Hypothesis Simulation Details for Scenario 1

			Tar	get	Ac	tual		
	Futility B	Soundary	Spending	Cum. Spending		Cum.	Proportion H1 Sims Outside	Cum. H1 Sims Outside
Look	T-Value Scale	P-Value Scale	Function Beta	Function Beta	Beta Spent	Beta Spent	Signif. Boundary	Signif. Boundary
1	1.07113	0.85629	0.0151	0.0151	0.0151	0.0151	0.0011	0.0011
2	-0.17706	0.42984	0.0706	0.0856	0.0706	0.0857	0.0887	0.0898
3	-0.82082	0.20628	0.0748	0.1605	0.0749	0.1605	0.2596	0.3494
4	-1.28003	0.10073	0.0637	0.2242	0.0637	0.2242	0.2486	0.5979
5	-1.72660	0.04251	0.0528	0.2770	0.0528	0.2770	0.1251	0.7230

The values obtained from any given run of this example will vary slightly due to the variation in simulations.

Example 3 - Enter Boundaries

With a set-up similar to Example 2, suppose we wish to investigate the properties of a set of significance (-3, -3, -2, -1) and futility (2, 1, 0, 0, -1) boundaries.

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 3** settings file. To load these settings to the procedure window, click **Open Example Settings File** in the Help Center or File menu.

Solve For	Alpha and Power (Enter Boundaries)
Test Type	T-Test
Higher Means Are	Better
Simulations	20000
Random Seed	3758223 (for Reproducibility)
Group Allocation	Equal (N1 = N2)
Sample Size Per Group	200
Mean1 (Mean of Group 1, Control)	108
NIM (Non-Inferiority Margin)	5
% (NIM as Percent of Mean1)	Unchecked
Mean2 (Mean of Group 2, Treatment)	108
Standard Deviation	20
Looks & Boundaries Tab	
Number of Looks	5
Equally Spaced	Checked
	Significance and Futility Boundaries
Significance Boundary	3 -3 -3 -2 -1 (for looks 1 through 5)
Futility Boundary	2 1 0 0 -1 (for looks 1 through 5)

Output

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

Scenario 1 Numeric Results for Group Sequential Test of Non-Inferiority

Solve For: Alpha and Power (Enter Boundaries)

Higher Means Are: Better

Hypotheses: H0: Mean1 - Mean2 = NIM; H1: Mean1 - Mean2 < NIM

Test Statistic: T-Test

Type of Boundaries: Significance and Futility Boundaries

Number of Looks: 5 Simulations: 20000 Pool Size: 40000

Random Seed: 3758223 (User-Entered)

Numeric Summary for Scenario 1

	Power			Alpha		
Value	95% LCL	95% UCL	Value	95% LCL	95% UCL	Beta
0.9243	0.9206	0.928	0.1532	0.14821	0.15819	0.0757

			Average S	ample Siz	e				
		Give	en H0	Give	en H1	Non-Inf.			Std
N1	N2	Grp1	Grp2	Grp1	Grp2	Margin	Mean1	Mean2	Dev
200	200	147	147	161	161	5	108	108	20

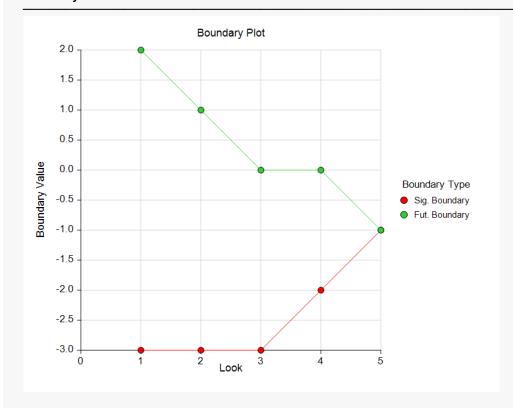
Accumulated Information Details for Scenario 1

	Accumulated Information	Accumu	ulated Sample	Size
Look	Percent	Group 1	Group 2	Total
1	20	40	40	80
2	40	80	80	160
3	60	120	120	240
4	80	160	160	320
5	100	200	200	400

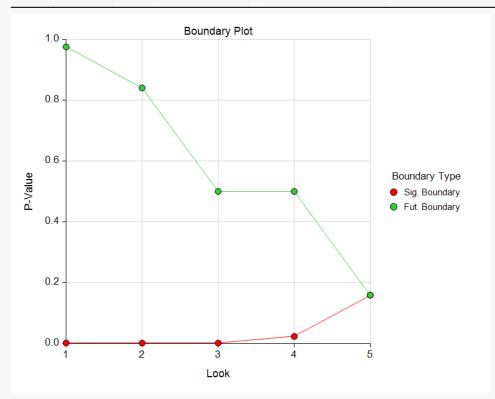
Boundaries for Scenario 1

	Significand	e Boundary	Futility Boundary	
Look	T-Value Scale	P-Value Scale	T-Value Scale	P-Value Scale
1	-3	0.00181	2	0.97551
2	-3	0.00157	1	0.84058
3	-3	0.00149	0	0.50000
4	-2	0.02318	0	0.50000
5	-1	0.15896	-1	0.15896

Boundary Plot



Boundary Plot - P-Value



Alpha-Spending and Null Hypothesis Simulation Details for Scenario 1

	Signif. E	Boundary		Cum.	Proportion H0 Sims Outside	Cum. H0 Sims Outside Futility Boundary	
Look	T-Value Scale	P-Value Scale	Alpha Spent	Alpha Spent	Futility Boundary		
1	-3	0.00181	0.00180	0.00180	0.02610	0.02610	
2	-3	0.00157	0.00105	0.00285	0.14180	0.16790	
3	-3	0.00149	0.00105	0.00390	0.34615	0.51405	
4	-2	0.02318	0.02095	0.02485	0.07680	0.59085	
5	-1	0.15896	0.12835	0.15320	0.25595	0.84680	

Beta-Spending and Alternative Hypothesis Simulation Details for Scenario 1

	Futility E	Boundary		Cum.	Proportion H1 Sims Outside	Cum. H1 Sims Outside	
Look	T-Value Scale	P-Value Scale	Beta Spent	Beta Spent	Signif. Boundary	Signif. Boundary	
1	2	0.97551	0.0012	0.0012	0.0359	0.0359	
2	1	0.84058	0.0043	0.0054	0.0600	0.0959	
3	0	0.50000	0.0226	0.0280	0.0826	0.1785	
4	0	0.50000	0.0044	0.0323	0.4225	0.6010	
5	-1	0.15896	0.0434	0.0757	0.3234	0.9243	

The values obtained from any given run of this example will vary slightly due to the variation in simulations.

Example 4 - Validation Using Simulation

With a set-up similar to Example 1, we examine the power and alpha generated by the set of two-sided significance boundaries (-4.6117, -2.9561, -2.3362, -1.9639, -1.7773).

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 4** settings file. To load these settings to the procedure window, click **Open Example Settings File** in the Help Center or File menu.

Solve For	Alpha and Power (Enter Boundaries)
Test Type	T-Test
Higher Means Are	Better
Simulations	20000
Random Seed	3774903 (for Reproducibility)
Group Allocation	Equal (N1 = N2)
Sample Size Per Group	200
Mean1 (Mean of Group 1, Control)	108
NIM (Non-Inferiority Margin)	5
% (NIM as Percent of Mean1)	Unchecked
Mean2 (Mean of Group 2, Treatment)	108
Standard Deviation	20
Looks & Boundaries Tab	
Number of Looks	5
Equally Spaced	Checked
Significance Boundary	4.6117, -2.9561, -2.3362, -1.9639, -1.7773

Output

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

Scenario 1 Numeric Results for Group Sequential Test of Non-Inferiority

Solve For: Alpha and Power (Enter Boundaries)

Higher Means Are: Better

Hypotheses: H0: Mean1 - Mean2 = NIM; H1: Mean1 - Mean2 < NIM

Test Statistic: T-Test

Type of Boundaries: Significance Boundaries Only

Number of Looks: 5 Simulations: 20000 Pool Size: 40000

Random Seed: 3774903 (User-Entered)

Numeric Summary for Scenario 1

	Power			Alpha			
Value	95% LCL	95% UCL	Value	95% LCL	95% UCL	Beta	
0.7839	0.7782	0.7896	0.04475	0.04188	0.04762	0.2161	

		Average Sample Size							
		Given H0		Given H1		Non-Inf.			Std
N1	N2	Grp1	Grp2	Grp1	Grp2	Margin	Mean1	Mean2	Dev
200	200	198	198	158	158	5	108	108	20

The values obtained from any given run of this example will vary slightly due to the variation in simulations. The power and alpha generated with these boundaries are very close to the values of Example 1.