Chapter 223

Group-Sequential Superiority by a Margin Tests for the Ratio of Two Proportions (Simulation)

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

Introduction

This procedure can be used to determine power, sample size and/or boundaries for group sequential superiority by a margin tests comparing the ratio of proportions from two groups. These tests are sometimes referred to as non-zero (or non-unity, for ratios and odds ratios) null tests. The tests that can be simulated in this procedure are the common two-sample Z-test with or without pooled standard error and with or without continuity correction, the T-test, and three score tests. 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 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 proportions, 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.
- 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 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.

Small Sample Considerations

When the sample size is small, say 200 or fewer per group, the discrete nature of the number of possible differences in proportions in the sampling distribution comes into play. This has led to a large number of proposed tests for comparing two proportions (or testing the 2 by 2 table of counts). For example, Upton (1982) considers twenty-two alternative tests for comparing two proportions. Sweeping statements about the power of one test over another are impossible to make, because the size of the Type I error depends upon the proportions used. At some proportions, some tests are overly conservative while others are not, while at other proportions the reverse may be true.

This simulation procedure, however, is based primarily on the ordering of the sample statistics in the simulation. The boundaries are determined by the spending function alphas. Thus, if a test used happens to be conservative in the single-look traditional sense, the boundaries chosen in the simulation results of this procedure will generally remove the conservative nature of the test. This makes comparisons to the one-look case surprising in many cases.

Definitions

Suppose you have two populations from which dichotomous (binary) responses will be 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 $q_2 = 1 - p_2$.

The 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	а	С	т
Control	b	d	n
Total	S	f	Ν

The following alternative notation is also used.

Group	Success	Failure	Total
Treatment	<i>x</i> ₁₁	<i>x</i> ₁₂	n_1
Control	<i>x</i> ₂₁	<i>x</i> ₂₂	n_2
Total	m_1	m_2	Ν

The binomial proportions p_1 and p_2 are estimated from these data using the formulae

$$\hat{p}_1 = \frac{a}{m} = \frac{x_{11}}{n_1}$$
 and $\hat{p}_2 = \frac{b}{n} = \frac{x_{21}}{n_2}$

Comparing Two Proportions

Let $p_{1.0}$ represent the group 1 proportion tested by the null hypothesis, H_0 . The power of a test is computed at a specific value of the proportion which we will call $p_{1.1}$. Let δ represent the smallest difference (margin of equivalence) between the two proportions that still results in the conclusion that the new treatment is not inferior to the current treatment. For a superiority test, $\delta > 0$. The set of statistical hypotheses that are tested is

$$H_0: p_{1.0} - p_2 \le \delta$$
 versus $H_1: p_{1.0} - p_2 > \delta$

which can be rearranged to give

$$H_0: p_{1,0} \le p_2 + \delta$$
 versus $H_1: p_{1,0} > p_2 + \delta$

There are multiple methods of specifying the margin of superiority. The most direct is to simply give values for p_2 and $p_{1.0}$. However, it is often more meaningful to give p_2 and then specify $p_{1.0}$ implicitly by specifying the difference, ratio, or odds ratio. Mathematically, assuming higher proportions are better, the definitions of these parameterizations are

<u>Parameter</u>	<u>Computation</u>	<u>Hypotheses</u>
Difference	$\delta = p_{1.0} - p_2$	$H_0: p_{1.0} - p_2 \le \delta_0 \text{ vs. } H_1: p_{1.0} - p_2 > \delta_0, \ \delta_0 < 0$
Ratio	$\phi = p_{1.0} / p_2$	$H_0: p_1 / p_2 \le \phi_0 \text{ vs. } H_1: p_1 / p_2 > \phi_0, \ \phi_0 < 1$
Odds Ratio	$\psi = Odds_{1.0} / Odds_2$	$H_0: o_{1.0} / o_2 \le \psi_0 \text{ vs. } H_1: o_{1.0} / o_2 > \psi_0, \ \psi_0 < 1$

Ratio

The ratio, $\phi = p_{1.0} / p_2$, gives the relative change in the probability of the response. Testing superiority uses the formulation

$$H_0: p_{1.0} / p_2 \le \phi_0$$
 versus $H_1: p_{1.0} / p_2 > \phi_0$

when higher proportions are better.

The following example is used to convey the concept of *superiority* as defined by the ratio. Suppose that 60% of patients ($p_2 = 0.60$) respond to the current treatment method. If a new treatment increases the response rate by no less than 10% ($\phi_0 = 1.10$), it will be considered non-inferior to the standard treatment. Substituting these figures into the statistical hypotheses gives

$$H_0: \phi \leq 1.10$$
 versus $H_1: \phi > 1.10$

In this example, when the null hypothesis is rejected, the concluded alternative is that the response rate is at least 66%. That is, the conclusion of superiority is that the new treatment's response rate is no worse than 10% more than that of the standard treatment.

Test Statistics

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

Miettinen and Nurminen's Likelihood Score Test

Miettinen and Nurminen (1985) proposed a test statistic for testing whether the ratio is equal to a specified value ϕ_0 . The regular MLE's, \hat{p}_1 and \hat{p}_2 , are used in the numerator of the score statistic while MLE's \tilde{p}_1 and \tilde{p}_2 , constrained so that $\tilde{p}_1 / \tilde{p}_2 = \phi_0$, are used in the denominator. A correction factor of N/(N-1) is applied to make the variance estimate less biased. The significance level of the test statistic is based on the asymptotic normality of the score statistic.

The formula for computing the test statistic is

$$z_{MNR} = \frac{\hat{p}_1 / \hat{p}_2 - \phi_0}{\sqrt{\left(\frac{\tilde{p}_1 \tilde{q}_1}{n_1} + \phi_0^2 \frac{\tilde{p}_2 \tilde{q}_2}{n_2}\right) \left(\frac{N}{N-1}\right)}}$$

where

$$\tilde{p}_1 = \tilde{p}_2 \phi_0$$

$$\tilde{p}_2 = \frac{-B - \sqrt{B^2 - 4AC}}{2A}$$

$$A = N\phi_0$$

$$B = -[n_1\phi_0 + x_{11} + n_2 + x_{21}\phi_0]$$

$$C = m_1$$

Farrington and Manning's Likelihood Score Test

Farrington and Manning (1990) proposed a test statistic for testing whether the ratio is equal to a specified value ϕ_0 . The regular MLE's, \hat{p}_1 and \hat{p}_2 , are used in the numerator of the score statistic while MLE's \tilde{p}_1 and \tilde{p}_2 , constrained so that $\tilde{p}_1 / \tilde{p}_2 = \phi_0$, are used in the denominator. A correction factor of N/(N-1) is applied to increase the variance estimate. The significance level of the test statistic is based on the asymptotic normality of the score statistic.

The formula for computing the test statistic is

$$z_{FMR} = \frac{\hat{p}_1 / \hat{p}_2 - \phi_0}{\sqrt{\left(\frac{\tilde{p}_1 \tilde{q}_1}{n_1} + \phi_0^2 \frac{\tilde{p}_2 \tilde{q}_2}{n_2}\right)}}$$

where the estimates \tilde{p}_1 and \tilde{p}_2 are computed as in the corresponding test of Miettinen and Nurminen (1985) given above.

Gart and Nam's Likelihood Score Test

Gart and Nam (1988), page 329, proposed a modification to the Farrington and Manning (1988) ratio test that corrects for skewness. Let $z_{FMR}(\phi)$ stand for the Farrington and Manning ratio test statistic described above. The skewness corrected test statistic, z_{GNR} , is the appropriate solution to the quadratic equation

 $(-\tilde{\varphi})z_{GNR}^2 + (-1)z_{GNR} + (z_{FMR}(\phi) + \tilde{\varphi}) = 0$

where

$$\tilde{\varphi} = \frac{1}{6\tilde{u}^{3/2}} \left(\frac{\tilde{q}_1(\tilde{q}_1 - \tilde{p}_1)}{n_1^2 \tilde{p}_1^2} - \frac{\tilde{q}_2(\tilde{q}_2 - \tilde{p}_2)}{n_2^2 \tilde{p}_2^2} \right)$$

$$\tilde{u} = \frac{q_1}{n_1 \tilde{p}_1} + \frac{q_2}{n_2 \tilde{p}_2}$$

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.

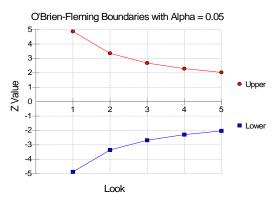
Spending functions have the characteristics that they are increasing and that

$$\alpha(0) = 0$$
$$\alpha(1) = \alpha$$

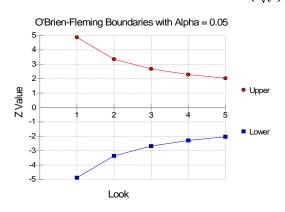
The last characteristic guarantees a fixed α level when the trial is complete. This methodology is very flexible since neither the times nor the number of analyses must be specified in advance. Only the functional form of $\alpha(\tau)$ must be specified.

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):

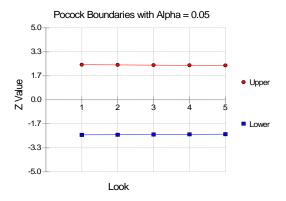
1. Hwang-Shih-DeCani (gamma family) $\alpha \left[\frac{1-e^{-\gamma t}}{1-e^{-\gamma}}\right], \gamma \neq 0; \quad \alpha t, \gamma = 0$



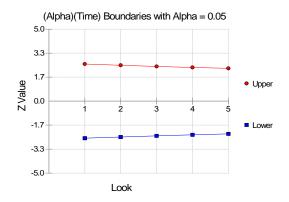
2. O'Brien-Fleming Analog $2 - 2\Phi\left(\frac{Z_{\alpha/2}}{\sqrt{t}}\right)$



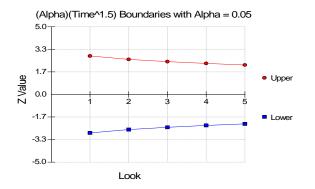
3. Pocock Analog $\alpha \cdot \ln(1 + (e - 1)t)$



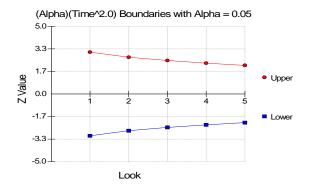
4. Alpha * time $\alpha \cdot t$



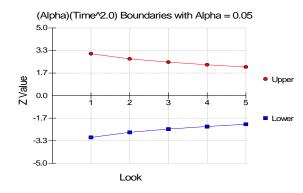
5. Alpha * time^1.5 $\alpha \cdot t^{3/2}$



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 proportion response of a new treatment to that of the current treatment. The current response proportion is 0.58. The researchers would like to show that the new treatment is at least 9% better than the standard treatment. This corresponds to a rate ratio of 1.09. Although the researchers do not know the true proportion of patients that will survive with the new treatment, they would like to examine the power that is achieved if the proportion under the new treatment is 0.68 (rate ratio of 1.1724). The sample size at the final look is to be 1000 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	Likelihood Score (Farr. & Mann.)
Higher Proportions Are	Better
Simulations	
Random Seed	
Alpha	0.05
Group Allocation	Equal (N1 = N2)
Sample Size Per Group	
R0 (Ratio H0 = P1.0/P2)	1.09
R1 (Ratio H1 = P1.1/P2)	1.1724
P2 (Group 2 Proportion)	0.58
Looks & Boundaries Tab	
Specification of Looks and Boundari	esSimple
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 Testing Proportion Ratio = R0

Solve For:	Power
Hypotheses:	H0: Proportion 1 / Proportion 2 = R0; H1: Proportion 1 / Proportion 2 > R0
Test Statistic:	Likelihood Score (Farrington and Manning) Test
Zero Adjustment Method:	None
Alpha-Spending Function:	O'Brien-Fleming Analog
Beta-Spending Function:	None
Futility Boundary Type:	None
Number of Looks:	5
Simulations:	100000
Random Seed:	3393736 (User-Entered)

Numeric Summary for Scenario 1

	Power		Alpha				
Value	95% LCL	95% UCL	Target	Actual	95% LCL	95% UCL	Beta
0.665	0.662	0.668	0.05	0.05	0.049	0.051	0.335

Power 95% LCL and UCL	The lower and upper confidence limits for the power estimate. The width of the interval is based on the number of simulations.
Target Alpha	The user-specified probability of rejecting a true null hypothesis. It is the total alpha spent.
Alpha or Actual Alpha	The alpha level that was actually achieved by the experiment. It is the total proportion of the null hypothesis simulations that are outside the significance boundaries.
Alpha 95% LCL and UCL	The lower and upper confidence limits for the actual alpha estimate. The width of the interval is based on the number of simulations.
Beta	The probability of accepting a false null hypothesis. It is the total proportion of alternative hypothesis simulations that do not cross the significance boundaries.
N1 and N2	The sample sizes of each group if the study reaches the final look.
Average Sample Size Given H0	The average or expected sample sizes of each group if H0 is true. These are based on the proportion of null hypothesis simulations that cross the significance or futility boundaries at each look.
Average Sample Size Given H1	The average or expected sample sizes of each group if H1 is true. These are based on the proportion of alternative hypothesis simulations that cross the significance or futility boundaries at each look.
R0	The superiority ratio is the proportion ratio between groups (Grp1 / Grp2) assuming the null hypothesis, H0.
R1	The proportion ratio between groups (Grp1 / Grp2) assuming the alternative hypothesis, H1.
P1.0	The proportion used in the simulations for Group 1 under H0.
P1.1	The proportion used in the simulations for Group 1 under H1.
P2	The proportion used in the simulations for Group 2 under H0 and H1.

Summary Statements

A group sequential trial with sample sizes of 1000 and 1000 at the final look achieve 66% power to detect a ratio of 1.17 of the treatment group proportion to a control group proportion of 0.58 with a superiority ratio of 1.09 with an overall Type I error rate (α) of 0.05 using a one-sided Likelihood Score (Farrington and Manning) Test.

Accumulated Information Details for Scenario 1

	Accumulated Information	Accumulated Sample Size					
Look	Percent	Group 1	Group 2	Total			
1	20	200	200	400			
2	40	400	400	800			
3	60	600	600	1200			
4	80	800	800	1600			
5	100	1000	1000	2000			

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

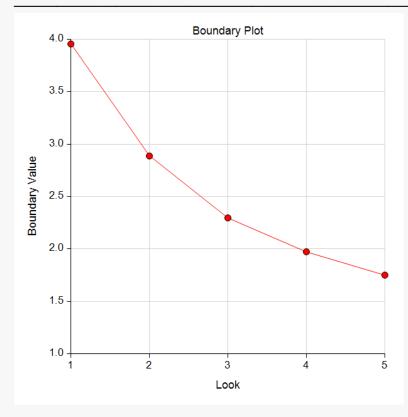
	Significand	e Boundary	
Look	Z-Value Scale	P-Value Scale	
1	3.95552	0.00004	
2	2.88733	0.00194	
3	2.29672	0.01082	
4	1.97325	0.02423	
5	1.75093	0.03998	
Look Significa	ance Boundary	Z-Value Scale	The number of the look. The value such that statistics outside this boundary at the corresponding look ind

Significance Boundary P-Value Scale

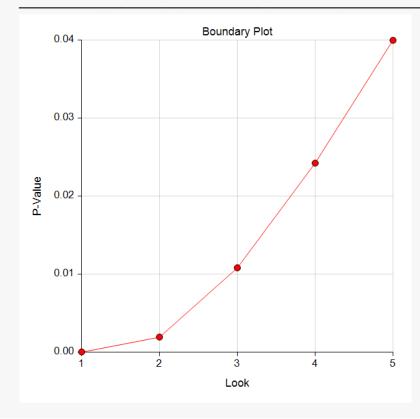
called efficacy boundaries. The value such that P-Values outside this boundary at the corresponding look indicate termination of the study and rejection of the null hypothesis. This P-Value corresponds to the Z-Value Boundary and is sometimes called the nominal alpha.

termination of the study and rejection of the null hypothesis. They are sometimes





Boundary Plot - P-Value



	Z	-Value Bound	ary	P	P-Value Boundary			
Look	Value	95% LCL	95% UCL	Value	95% LCL	95% UCL		
1	3.95552			0.00004				
2	2.88733	2.83778	2.91171	0.00194	0.00180	0.00227		
3	2.29672	2.28150	2.32325	0.01082	0.01008	0.01126		
4	1.97325	1.95988	1.98677	0.02423	0.02347	0.02500		
5	1.75093	1.73781	1.76507	0.03998	0.03878	0.04112		

Significance Boundaries with 95% Simulation Confidence Intervals for Scenario 1

The number of the look. Z-Value Boundary Value

P-Value Boundary Value

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. 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 Z-Value Boundary and is sometimes called the nominal alpha.

95% LCL and UCL

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

			Tai	rget Actual				
	Signif. E	Boundary	Spending	Cum. Spending		Cum.	Proportion H1 Sims Outside	Cum. H1 Sims Outside
Look	Z-Value Scale	P-Value Scale	Function Alpha	Function Alpha	Alpha Spent	Alpha Spent	Signif. Boundary	Signif. Boundary
1	3.95552	0.00004	0.000	0.000	0.000	0.000	0.001	0.001
2	2.88733	0.00194	0.002	0.002	0.002	0.002	0.058	0.060
3	2.29672	0.01082	0.009	0.011	0.009	0.011	0.200	0.259
4	1.97325	0.02423	0.017	0.028	0.017	0.028	0.224	0.483
5	1.75093	0.03998	0.022	0.050	0.022	0.050	0.181	0.665
4	1.97325	0.02423	0.017	0.028	0.017	0.028	0	.224

Look

Significance Boundary Z-Value Scale

Significance Boundary P-Value Scale

Spending Function Alpha

Cumulative Spending Function Alpha

Alpha Spent

Cumulative Alpha Spent

Proportion H1 Sims Outside Significance Boundary

Cumulative H1 Sims Outside Significance Boundary

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. 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 Significance Z-Value Boundary and is sometimes called the nominal alpha.

The intended portion of alpha allocated to the particular look based on the alpha-spending function.

The intended accumulated alpha allocated to the particular look. It is the sum of the Spending Function Alpha up to the corresponding look.

The proportion of the null hypothesis simulations resulting in statistics outside the Significance Boundary at this look.

The proportion of the null hypothesis simulations resulting in Significance Boundary termination up to and including this look. It is the sum of the Alpha Spent up to the corresponding look.

The proportion of the alternative hypothesis simulations resulting in statistics outside the Significance Boundary at this look. It may be thought of as the incremental power.

The proportion of the alternative hypothesis simulations resulting in Significance Boundary termination up to and including this look. It is the sum of the Proportion H1 Sims Outside Significance Boundary up to the corresponding look.

Solve For:	Power
Hypotheses:	H0: Proportion 1 / Proportion 2 = R0; H1: Proportion 1 / Proportion 2 > R0
Test Statistic:	Likelihood Score (Farrington and Manning) Test
Zero Adjustment Method:	None
Alpha-Spending Function:	O'Brien-Fleming Analog
Beta-Spending Function:	None
Futility Boundary Type:	None
Number of Looks:	5
Simulations:	100000
Random Seed:	3393736 (User-Entered)

Numeric Summary of Scenarios

Scenario	Power	N1	N2	Alpha	R0	R1	P1.0	P1.1	P2
1	0.665	1000	1000	0.05	1.09	1.17	0.63	0.68	0.58
Power	The probabil hypothesis			ull hypothesi				al proportic	on of alter
Alpha	The alpha le simulations		,	hieved by th gnificance bo		ent. It is the	total propo	ortion of the	e null hypo
N1 and N2	The sample	sizes of eac	h group if th	ne study read	ches the fin	al look.			
R0	The superior	ity ratio is t	ne proportio	n ratio betwe	en groups	(Grp1 / Gr	p2) assumi	ng the null	hypothes
R1	The proportion	on ratio bet	ween group	s (Grp1 / Grp	2) assumii	ng the alter	mative hypo	othesis, H1	
P1.0	The proportion	on used in t	he simulatic	ns for Group	1 under H	10.			
P1.1	The proportion	on used in t	he simulatio	ns for Group	1 under H	11.			
P2	The proportion	on used in t	he simulatio	ns for Group	2 under H	I0 and H1.			

Power and Alpha Summary

		Power				Alpha		
Scenario	Value	95% LCL	95% UCL	Target	Actual	95% LCL	95% UCL	Beta
1	0.665	0.662	0.668	0.05	0.05	0.049	0.051	0.335
Power							It is the total prop	portion of
Power 95% L	CL and UCL	The lower a		ence limits fo		ne significance b stimate. The wid	Ith of the interval	is based
Target Alpha		The user-sp	pecified probabili	ty of rejecting	a true null h	ypothesis. It is th	ne total alpha spe	ent.
Alpha or Actu	ual Alpha		evel that was ac is simulations that	,	, ,		total proportion of	of the null
Alpha 95% L	CL and UCL	The lower a		ence limits fo			he width of the ir	terval is
Beta		The probab	ility of accepting	a false null h		is the total propo ance boundaries	ortion of alternativ	/e

						Average S	ample Siz	е	
					Give	n H0	Give	en H1	
Scenario	Power	Alpha	N1	N2	Grp1	Grp2	Grp1	Grp2	
1	0.665	0.05	1000	1000	992	992	839	839	
Power		рі			a false null hypothesis				
Alpha		The	alpha leve				experiment.		otal proportion of
Арна			e null hypo	thesis simul	lations that a	are outside t	the significa	nce bound	laries.
N1 and N2		th	21		lations that a group if the		0		laries.
N1 and N2	nple Size Give	th The n H0 The pi	sample siz	es of each	group if the sample size:	study reach s of each gro	es the final oup if H0 is	look. true. Thes	daries. The are based on the futility boundaries

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

Example 2 – Power with Futility Boundaries

Continuing with Example 1, suppose that the researchers would also like to terminate the study early if there is indication that the treatment is not better than the standard by 0.05.

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
Test Type	Likelihood Score (Farr. & Mann.)
Higher Proportions Are	Better
Simulations	
Random Seed	Blank or Random
Alpha	0.05
Group Allocation	Equal (N1 = N2)
Sample Size Per Group	
R0 (Ratio H0 = P1.0/P2)	
R1 (Ratio H1 = P1.1/P2)	
P2 (Group 2 Proportion)	0.58

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 Testing Proportion Ratio = R0

Solve For:	Power
Hypotheses:	H0: Proportion 1 / Proportion 2 = R0; H1: Proportion 1 / Proportion 2 > R0
Test Statistic:	Likelihood Score (Farrington and Manning) Test
Zero Adjustment Method:	None
Alpha-Spending Function:	O'Brien-Fleming Analog
Beta-Spending Function:	O'Brien-Fleming Analog
Futility Boundary Type:	Non-Binding
Number of Looks:	5
Simulations:	100000
Random Seed:	5156712 (Computer-Generated)

Numeric Summary for Scenario 1

	Power				Alpha		
Value	95% LCL	95% UCL	Target	Actual	95% LCL	95% UCL	Beta
0.552	0.549	0.555	0.05	0.036	0.035	0.037	0.448

			Average S	ample Siz	e					
		Give	en H0	Give	en H1					
N1	N2	Grp1	Grp2	Grp1	Grp2	R0	R1	P1.0	P1.1	P2
1000	1000	419	419	636	636	1.09	1.17	0.63	0.68	0.58

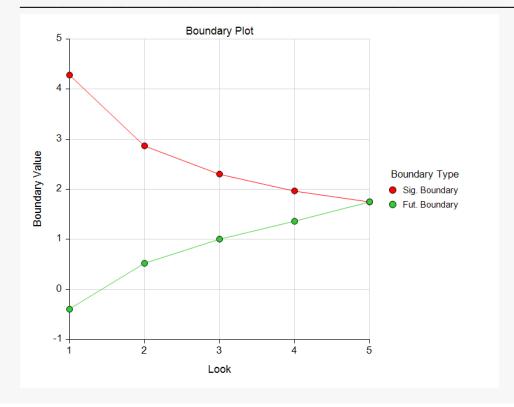
Accumulated Information Details for Scenario 1

	Accumulated Information	Accum	Accumulated Sample Size			
Look	Percent	Group 1	Group 2	Total		
1	20	200	200	400		
2	40	400	400	800		
3	60	600	600	1200		
4	80	800	800	1600		
5	100	1000	1000	2000		

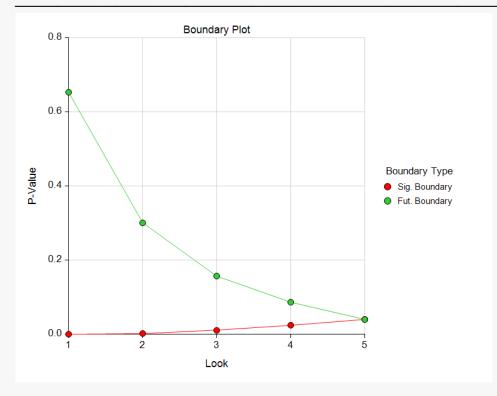
Boundaries for Scenario 1

	Significanc	e Boundary	Futility B	Boundary
Look	Z-Value Scale	P-Value Scale	Z-Value Scale	P-Value Scale
1	4.28236	0.00001	-0.39349	0.65302
2	2.86855	0.00206	0.52323	0.30041
3	2.30573	0.01056	1.00656	0.15707
4	1.96989	0.02443	1.36235	0.08654
5	1.75093	0.03998	1.75093	0.03998

Boundary Plot



Boundary Plot - P-Value



Significance Boundaries with 95% Simulation Confidence Intervals for Scenario 1

	z	-Value Bound	ary	P	-Value Bound	ary
Look	Value	95% LCL	95% UCL	Value	95% LCL	95% UCL
1	4.28236			0.00001		
2	2.86855	2.82882	2.91171	0.00206	0.00180	0.00234
3	2.30573	2.28178	2.32717	0.01056	0.00998	0.01125
4	1.96989	1.95480	1.98666	0.02443	0.02348	0.02530
5	1.75093	1.73663	1.76507	0.03998	0.03878	0.04123

Futility Boundaries with 95% Simulation Confidence Intervals for Scenario 1

	Z	-Value Bounda	ary	P	-Value Bound	ary
Look	Value	95% LCL	95% UCL	Value	95% LCL	95% UCL
1	-0.39349	-0.40348	-0.38359	0.65302		
2	0.52323	0.51470	0.53140	0.30041	0.29757	0.30338
3	1.00656	1.00137	1.01510	0.15707	0.15503	0.15832
4	1.36235	1.35228	1.36982	0.08654	0.08537	0.08814
5	1.75093	1.74325	1.76330	0.03998	0.03892	0.04064

			Tar	get	Ac	tual		
	Signif. B	oundary	Spending	Cum. Spending		Cum.	Proportion H0 Sims Outside	Cum. H0 Sims Outside
Look	Z-Value Scale	P-Value Scale	Function Alpha	Function Alpha	Alpha Spent	Alpha Spent	Futility Boundary	Futility Boundary
1	4.28236	0.00001	0.000	0.000	0.000	0.000	0.346	0.346
2	2.86855	0.00206	0.002	0.002	0.002	0.002	0.369	0.715
3	2.30573	0.01056	0.009	0.011	0.009	0.011	0.155	0.870
4	1.96989	0.02443	0.017	0.028	0.014	0.026	0.066	0.936
5	1.75093	0.03998	0.022	0.050	0.011	0.036	0.028	0.964

Beta-Spending and Alternative Hypothesis Simulation Details for Scenario 1

			Tar	rget	Ac	tual		-
	Futility B	Boundary	Spending	Cum. Spending		Cum.	Proportion H1 Sims Outside	Cum. H1 Sims Outside
Look	Z-Value Scale	P-Value Scale	Function Beta	Function Beta	Beta Spent	Beta Spent	Signif. Boundary	Signif. Boundary
1	-0.39349	0.65302	0.091	0.091	0.090	0.090	0.000	0.000
2	0.52323	0.30041	0.141	0.232	0.141	0.231	0.063	0.064
3	1.00656	0.15707	0.097	0.329	0.097	0.328	0.192	0.256
4	1.36235	0.08654	0.069	0.398	0.069	0.397	0.199	0.455
5	1.75093	0.03998	0.052	0.450	0.051	0.448	0.096	0.552

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, 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	Likelihood Score (Farr. & Mann.)
Higher Proportions Are	Better
Simulations	
Random Seed	Blank or Random
Group Allocation	Equal (N1 = N2)
Sample Size Per Group	
R0 (Ratio H0 = P1.0/P2)	1.09
R1 (Ratio H1 = P1.1/P2)	1.1724
P2 (Group 2 Proportion)	0.58

Looks & Boundaries Tab

Number of Looks	5
Equally Spaced	Checked
Types of Boundaries	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 Testing Proportion Ratio = R0

Solve For:	Alpha and Power (Enter Boundaries)
Hypotheses:	H0: Proportion 1 / Proportion 2 = R0; H1: Proportion 1 / Proportion 2 > R0
Test Statistic:	Likelihood Score (Farrington and Manning) Test
Zero Adjustment Method:	None
Type of Boundaries:	Significance and Futility Boundaries
Number of Looks:	5
Simulations:	100000
Random Seed:	5196644 (Computer-Generated)

Numeric Summary for Scenario 1

	Power			Alpha		
Value	95% LCL	95% UCL	Value	95% LCL	95% LCL	Beta
0.853	0.851	0.855	0.153	0.151	0.155	0.147

			Average S	ample Siz	е					
		Give	en H0	Give	en H1					
N1	N2	Grp1	Grp2	Grp1	Grp2	R0	R1	P1.0	P1.1	P2
1000	1000	738	738	844	844	1.09	1.17	0.63	0.68	0.58

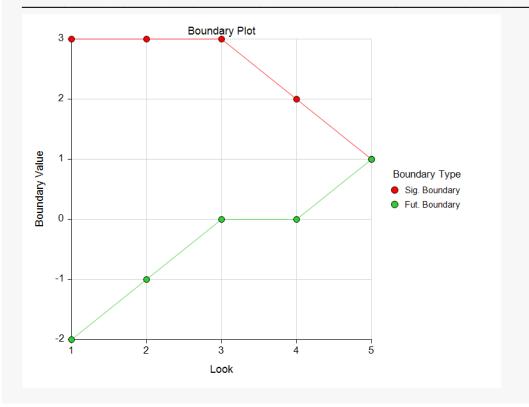
Accumulated Information Details for Scenario 1

Accumulated	Accumu	lated Sample	e Size
Percent	Group 1	Group 2	Total
20	200	200	400
40	400	400	800
60	600	600	1200
80	800	800	1600
100	1000	1000	2000
	Information Percent 20 40 60 80	Information Group 1 20 200 40 400 60 600 80 800	Information Group 1 Group 2 20 200 200 40 400 400 60 600 600 80 800 800

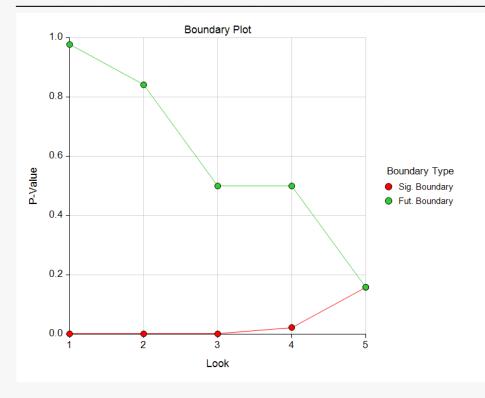
Boundaries for Scenario 1

	Significanc	e Boundary	Futility Boundary		
Look	Z-Value Scale	P-Value Scale	Z-Value Scale	P-Value Scale	
1	3	0.00135	-2	0.97725	
2	3	0.00135	-1	0.84134	
3	3	0.00135	0	0.50000	
4	2	0.02275	0	0.50000	
5	1	0.15866	1	0.15866	

Boundary Plot



Boundary Plot - P-Value



Alpha-Spending and Null Hypothesis Simulation Details for Scenario 1

	Signif. E	Boundary		Cum	Proportion H0 Sims	Cum. H0 Sims	
Look	Z-Value Scale	P-Value Scale	Alpha Spent	Cum. Alpha Spent	Outside Futility Boundary	Outside Futility Boundary	
1	3	0.00135	0.001	0.001	0.023	0.023	
2	3	0.00135	0.001	0.003	0.141	0.164	
3	3	0.00135	0.001	0.004	0.340	0.503	
4	2	0.02275	0.020	0.024	0.083	0.587	
5	1	0.15866	0.129	0.153	0.260	0.847	

Beta-Spending and Alternative Hypothesis Simulation Details for Scenario 1

	Futility E	Boundary			Proportion H1 Sims	Cum. H1 Sims Outside Signif. Boundary	
Look	Z-Value Scale	P-Value Scale	Beta Spent	Cum. Beta Spent	Outside Signif. Boundary		
1	-2	0.97725	0.002	0.002	0.021	0.021	
2	-1	0.84134	0.009	0.011	0.039	0.060	
3	0	0.50000	0.042	0.053	0.052	0.112	
4	0	0.50000	0.008	0.061	0.351	0.462	
5	1	0.15866	0.086	0.147	0.390	0.853	

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 2, we examine the power and alpha generated by the set of significance (3.85258, 2.90290, 2.29873, 1.96529, 1.73663) and futility (-0.40348, 0.51756, 0.99732, 1.35899, 1.73663) boundaries.

For reproducibility, we'll use a random seed of 4811181.

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.

Design Tab	
Solve For	Alpha and Power (Enter Boundaries)
Test Type	Likelihood Score (Farr. & Mann.)
Higher Proportions Are	Better
Simulations	
Random Seed	
Group Allocation	Equal (N1 = N2)
Sample Size Per Group	
R0 (Ratio H0 = P1.0/P2)	
R1 (Ratio H1 = P1.1/P2)	1.1724
P2 (Group 2 Proportion)	0.58
Looks & Boundaries Tab	
Number of Looks	5
Equally Spaced	Checked
Types of Boundaries	Significance and Futility Boundaries
Significance Boundary	

Output

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

Scenario 1 Numeric Results for Group Sequential Testing Proportion Ratio = R0

Solve For:	Alpha and Power (Enter Boundaries)
Hypotheses:	H0: Proportion 1 / Proportion 2 = $R0$; H1: Proportion 1 / Proportion 2 > $R0$
Test Statistic:	Likelihood Score (Farrington and Manning) Test
Zero Adjustment Method:	None
Type of Boundaries:	Significance and Futility Boundaries
Number of Looks:	5
Simulations:	100000
Random Seed:	4811181 (User-Entered)

Numeric Summary for Scenario 1

	Power			Alpha				
Value	95% LCL	95% UCL	Value	95% LCL	95% LCL	Beta		
0.554	0.551	0.557	0.037	0.036	0.038	0.446		

	N2	1	Average S							
N1		Given H0		Given H1						
		Grp1	Grp2	Grp1	Grp2	R0	R1	P1.0	P1.1	P2
1000	1000	419	419	637	637	1.09	1.17	0.63	0.68	0.58

The power and alpha generated with these boundaries are very close to the values of Example 2.