Chapter 401

Conditional Power and Sample Size Reestimation of Tests for Two Means in a 2x2 Cross-Over Design

Introduction

In sequential designs, one or more intermediate analyses of the emerging data are conducted to evaluate whether the experiment should be continued. This may be done to conserve resources or to allow a data monitoring board to evaluate safety and efficacy when subjects are entered in a staggered fashion over a long period of time. *Conditional power* (a frequentist concept) is the probability that the final result will be significant, given the data obtained up to the time of the interim look. *Predictive power* (a Bayesian concept) is the result of averaging the conditional power over the posterior distribution of effect size. Both of these methods fall under the heading of *stochastic curtailment* techniques. Further reading about the theory of these methods can be found in Jennison and Turnbull (2000), Chow and Chang (2007), Chang (2008), Proschan et. al (2006), and Dmitrienko et. al (2005).

This program module computes conditional and predicted power for the case when a *t*-test, computed from data obtained from a 2x2 cross-over design, is used to test whether two population means are different. It also provides *sample size reestimation* to achieve a specified conditional power value.

Technical Details

The 2x2 cross-over design may be described as follows. Randomly assign the subjects to one of two sequence groups so that there are n_1 subjects in sequence one and n_2 subjects in sequence two. In order to achieve design balance, the sample sizes n_1 and n_2 are assumed to be equal so that $n_1 = n_2 = N/2$.

Sequence one is given treatment A followed by treatment B. Sequence two is given treatment B followed by treatment A.

The data may be analyzed by a *t*-test. See one of the cross-over chapters for further details.

Conditional Power

The power of an experiment indicates whether a study is likely to result in useful results, given the sample size. Low power means that the study is *futile*: little chance of statistical significance even though the alternative hypothesis is true. A study that is futile should not be started. However, futility may be determined only after the study has started. When this happens, the study is *curtailed*.

The futility of a study that is underway can be determined by calculating its *conditional power*: the probability of statistical significance at the completion of the study given the data obtained so far.

It is important to note that conditional power at the beginning of the study before any data are collected is equal to the unconditional power. So, conditional power will be high even if early results are negative. Hence, conditional power will seldom result in study curtailment very early in the study.

From Jennison and Turnbull (2000) pages 205 to 208, the general upper one-sided conditional power at stage k for rejecting a null hypothesis about a parameter θ at the end of the study, given the observed test statistic, Z_k , is computed as

$$P_{uk}(\theta) = \Phi\left(\frac{Z_k\sqrt{I_k} - Z_{1-\alpha}\sqrt{I_K} + \theta(I_K - I_k)}{\sqrt{I_K - I_k}}\right),$$

the general lower one-sided conditional power at stage k is computed as

$$P_{lk}(\theta) = \Phi\left(\frac{-Z_k\sqrt{I_k} - z_{1-\alpha}\sqrt{I_K} - \theta(I_K - I_k)}{\sqrt{I_K - I_k}}\right),$$

and the general two-sided conditional power at stage *k* is computed as

$$P_{k}(\theta) = \Phi\left(\frac{Z_{k}\sqrt{I_{k}} - z_{1-\alpha/2}\sqrt{I_{K}} + \theta(I_{K} - I_{k})}{\sqrt{I_{K} - I_{k}}}\right) + \Phi\left(\frac{-Z_{k}\sqrt{I_{k}} - z_{1-\alpha/2}\sqrt{I_{K}} - \theta(I_{K} - I_{k})}{\sqrt{I_{K} - I_{k}}}\right),$$

where

 θ = the parameter being tested by the hypothesis

k =an interim stage at which the conditional power is computed (k = 1, ..., K - 1)

K = the stage at which the study is terminated, and the final test computed

 Z_k = the test statistic calculated from the observed data that has been collected up to stage k

 I_k = the information level at stage k

 I_K = the information level at the end of the study

 $z_{1-\alpha}$ = the standard normal value for the test with a type I error rate of α .

For a 2x2 cross-over design t-test of two means with null hypothesis H_0 : $\delta = \delta 0$, where $\delta 0$ is the hypothesized mean difference under the null hypothesis, and if $\delta 1$ is the actual mean difference under the alternative hypothesis, these components are computed in Chang (2008) page 71 as

 $heta=\delta 1-\delta 0$ (the expected difference of mean differences under the alternative hypothesis)

$$Z_k = \left(\frac{\overline{d}_{Y_1k} + \overline{d}_{Y_2k}}{2} - \delta 0\right) \sqrt{\hat{I}_k}$$
 (the *t*-statistic computed from the observed data)

$$I_k = \frac{n_k}{\sigma_d^2}$$
 (the interim information level)

$$I_K = \frac{N}{\sigma_d^2}$$
 (the final information level)

Conditional Power and Sample Size Reestimation of Tests for Two Means in a 2x2 Cross-Over Design

where

 \bar{d}_{Yik} the sample mean difference at stage k for those subjects receiving sequence i

 \hat{l}_k is the estimated information from the sample at stage k

 n_k is the sample size at stage k

N is the total sample size

 σ_d^2 is the variance of the subject differences

Computing conditional power requires you to set σ_d^2 , $\delta 0$, and $\delta 1$. Their values can come from the values used during the planning of the study, from similar studies, or from estimates made from the data that has emerged.

The Variance of the Pair Differences (σ_d^2)

If you have an estimate of σ_d^2 from the differences, you may enter the standard deviation of the pair differences (σ_d) directly. It is assumed that the variances of the two sequences are the same. A reasonable (but somewhat rough) estimate of σ_d^2 may be obtained using the range of pair differences as

$$\sigma_d^2 = \left(\frac{Range}{4}\right)^2$$

such that

$$\sigma_d = \frac{Range}{4}$$
.

If you have estimates of the variances of the original pair variables (σ_1^2 and σ_2^2) and the Pearson correlation between the pair variables (ρ), the variance of the pair differences (σ_d^2) may be calculated using the equation

$$\sigma_d^2 = \sigma_1^2 + \sigma_2^2 - 2\rho\sigma_1\sigma_2$$

such that

$$\sigma_d = \sqrt{\sigma_1^2 + \sigma_2^2 - 2\rho\sigma_1\sigma_2} \; .$$

If $\sigma_1^2=\sigma_2^2=\sigma_{_{\mathcal{X}}}^2$, then this formula reduces to

$$\sigma_d^2 = 2\sigma_x^2(1-\rho)$$

such that

$$\sigma_d = \sqrt{2\sigma_x^2(1-\rho)} \,.$$

Conditional Power and Sample Size Reestimation of Tests for Two Means in a 2x2 Cross-Over Design

If you have an estimate of the within-subject population variance (σ_w^2), then σ_d^2 may be calculated using the equation

$$\sigma_d^2 = 2\sigma_w^2$$

such that

$$\sigma_d = \sqrt{2\sigma_w^2}$$
.

Often, cross-over designs are analyzed using repeated measures ANOVA. σ_w^2 is estimated by the within mean square error (WMSE) from the repeated measures ANOVA.

Converting from a T-Statistic to a Z-Statistic

A common problem is that the procedure requires a Z-statistic, but the results from an analysis usually provide a T-statistic with a given degrees of freedom. So, the T-statistic must be transformed to a Z-statistic. One way to do this is to use the associated p-value. This is accomplished using the following steps:

Step 1. Find the p-value associated with the t-statistic

For example, suppose we have a t-statistic of 2.33 with 30 degrees of freedom. Using the **PASS** Probability Calculator, the probability for Student's T distribution is $Prob(t \ge T) = 0.0133616$.

Step 2. Convert this p-value to a z-statistic

Continuing the example, we can use the Normal distribution in the **PASS** Probability Calculator to determine that the z-score associated with a p-value of 0.0133616 is -2.215537403212. Since the t-statistic was positive, we use z = 2.215537403212.

Step 3. Enter the z-statistic for Zk (Current Test Statistic)

Continuing the example, we would enter 2.2155374 for Zk (Current Test Statistic).

Futility Index

The *futility index* is $1 - P_k(\theta)|H_1$. The study may be stopped if this index is above 0.8 or 0.9 (that is, if conditional power falls below 0.2 or 0.1).

Predictive Power

Predictive power (a Bayesian concept) is the result of averaging the conditional power over the posterior distribution of effect size. From Jennison and Turnbull (2000) pages 210 to 213, the general upper one-sided predictive power at stage k is given by

$$P_{uk} = \Phi\left(\frac{Z_k\sqrt{I_K} - Z_{1-\alpha}\sqrt{I_k}}{\sqrt{I_K - I_k}}\right).$$

the general lower one-sided predictive power at stage k is given by

$$P_{lk} = \Phi\left(\frac{-Z_k\sqrt{I_K} - z_{1-\alpha}\sqrt{I_k}}{\sqrt{I_K - I_k}}\right),$$

the general two-sided predictive power at stage k is given by

$$P_{k} = \Phi\left(\frac{|Z_{k}|\sqrt{I_{K}} - z_{1-\alpha/2}\sqrt{I_{k}}}{\sqrt{I_{K} - I_{k}}}\right) + \Phi\left(\frac{-|Z_{k}|\sqrt{I_{K}} - z_{1-\alpha/2}\sqrt{I_{k}}}{\sqrt{I_{K} - I_{k}}}\right),$$

with all terms defined as in the equations for conditional power.

Sample Size Reestimation

As Chang (2014) points out, after an interim analysis, it is often desirable to recalculate the target sample size using updated values for various nuisance parameters such as the variance. This process is known as *sample size reestimation*.

One method of calculating an adjusted sample size estimate is to search for the sample size that results in a predetermined value of conditional power. **PASS** conducts a binary search using the conditional power as the criterion. The result is called the *target sample size*.

Example 1 - Computing Conditional Power

Suppose a 2x2 cross-over study has been planned to detect a mean change of 1 at an alpha of 0.025 using a one-sided *t*-test. The total sample size is 52. The standard deviation is expected to be about 1.8. An interim analysis is run after half the data have been collected. This analysis yields a *t*-test value of 2.12.

The data monitoring board would also like to have the conditional power calculated for mean changes of 0.2, 0.4, 0.6, and 0.8.

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	Conditional Power
Alternative Hypothesis	H1: δ > δ0 (One-Sided)
Alpha	0.025
N (Target Sample Size)	52
nk (Sample Size at Look k)	26
δ0 (Null Difference)	0
δ1 (Actual Difference to Detect)	0.2 0.4 0.6 0.8 1
Standard Deviation Input Type	Enter the SD of Pair Differences
σd (SD of Pair Differences)	1.8
Test Statistic Input Type	Tk
Tk (Current Test Statistic)	2.12

Output

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

Numeric Reports

Numeric Results

Solve For: **Conditional Power** Test Type: 2x2 Cross-Over T-Test Hypotheses: $H0: \delta \le \delta 0$ vs. $H1: \delta > \delta 0$

D	_	Samp	le Size	Mean I	Difference		04	T4		
Conditional	Predictive	Target N	Look k	Null δ0	Actual δ1	δ1 - δ0	Standard Deviation σd	Test Statistic Tk	Alpha	Futility
0.42380	0.81244	52	26	0	0.2	0.2	1.8	2.12	0.025	0.57620
0.64594	0.81244	52	26	0	0.4	0.4	1.8	2.12	0.025	0.35406
0.82663	0.81244	52	26	0	0.6	0.6	1.8	2.12	0.025	0.17337
0.93416 0.98096	0.81244 0.81244	52 52	26 26	0 0	0.8 1.0	0.8 1.0	1.8 1.8	2.12 2.12	0.025 0.025	0.06584 0.01904

Conditional Power	The probability of rejecting a false null hypothesis at the end of the study given the data that have
	emerged so far.
Predictive Power	The result of averaging the conditional power over the posterior distribution of the effect size.
N	The target sample size from both sequences combined.
nk	The total sample size obtained through stage k from both sequences combined.
δ0	The mean difference assuming the null hypothesis.
δ1	The actual difference to detect under the alternative hypothesis at which conditional power is calculated.
δ1 - δ0	The difference of the mean differences. This is the minimum value that is to be detected by the

completed study. σd The standard deviation of the paired differences of subjects within the same sequence. It is assumed that the standard deviations of the two sequences are the same.

Tk The value of the test statistic from the observed data at stage k.

The probability of rejecting a true null hypothesis. Alpha

Equal to one minus the conditional power. A value greater than 0.9 or 0.8 indicates the study should be **Futility**

stopped because there is little chance of achieving statistical significance.

Summary Statements

A 2×2 cross-over design will be used to test whether the mean paired difference (δ) is greater than 0 (H0: $\delta \leq 0$ versus H1: $\delta > 0$). The comparison will be made using a one-sided, two-sample t-test, with a Type I error rate (α) of 0.025. The standard deviation of the paired differences of subjects within the same sequence is assumed to be 1.8 for both sequences. The desired difference to detect is $\delta 1 = 0.2$ (or $\delta 1 - \delta 0 = 0.2$). With a current total sample size of nk = 26 out of a target sample size of 52 (assuming half in Sequence 1 and half in Sequence 2), and with a current t-test statistic of 2.12, the conditional power is 0.4238. The predictive power is 0.81244, and the futility index is 0.5762.

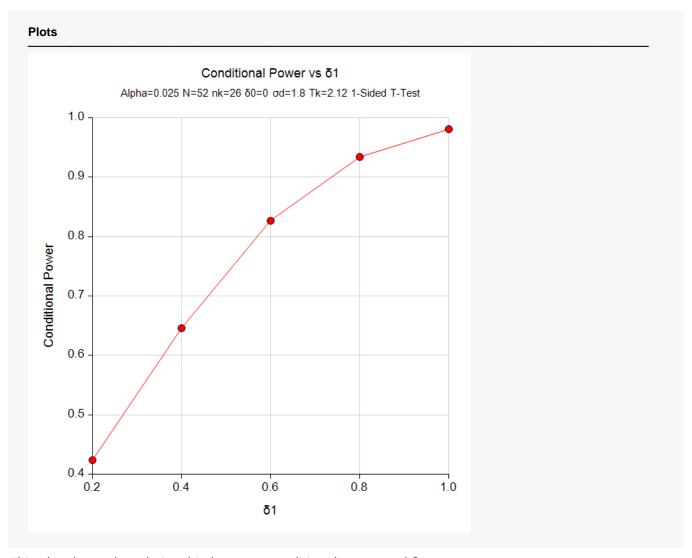
References

Jennison, C., and Turnbull, B.W. 2000. Group Sequential Methods with Applications to Clinical Trials. Chapman & Hall/CRC. New York.

Proschan, M., Lan, K.K.G., Wittes, J.T. 2006. Statistical Monitoring of Clinical Trials. Springer. New York. Chang, Mark. 2008. Classical and Adaptive Clinical Trial Designs. John Wiley & Sons. Hoboken, New Jersey. Chang, Mark. 2014. Adaptive Design Theory and Implementation Using SAS and R. CRC Press. New York.

This report shows the values of each of the parameters, one scenario per row. The definitions of each column are given in the Report Definitions section.

Plots Section



This plot shows the relationship between conditional power and $\delta 1$.

Example 2 - Validation

We could not find an example of a conditional power calculation for a 2x2 cross-over t-test in the literature. Since the calculations are relatively simple, as a validation we will perform the calculation of the third scenario (δ 1 = 0.6) of Example 1 by hand, except that we will use $Z_k = 2.12$ instead of $t_k = 2.12$.

In this case

$$I_k = \frac{n_k}{\sigma_d^2}$$
 $I_K = \frac{N}{\sigma_d^2}$ $I_K = \frac{52}{1.8^2}$ $I_K = \frac$

$$P_{uk}(\theta) = \Phi\left(\frac{Z_k\sqrt{I_k} - Z_{1-\alpha}\sqrt{I_K} + (\delta_1 - \delta_0)(I_K - I_k)}{\sqrt{I_K - I_k}}\right)$$

$$= \Phi\left(\frac{2.12\sqrt{8.0246914} - (1.959964)\sqrt{16.0493827} + (0.6 - 0)(16.0493827 - 8.0246914)}{\sqrt{16.0493827 - 8.0246914}}\right)$$

$$= \Phi\left(\frac{2.9683814}{2.8327886}\right)$$

$$= \Phi(1.0478655)$$

$$= 0.8526497$$

Conditional Power and Sample Size Reestimation of Tests for Two Means in a 2x2 Cross-Over Design

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	Conditional Power
Alternative Hypothesis	H1: δ > δ0 (One-Sided)
Alpha	0.025
N (Target Sample Size)	52
nk (Sample Size at Look k)	26
δ0 (Null Difference)	0
δ1 (Actual Difference to Detect)	0.6
Standard Deviation Input Type	Enter the SD of Pair Differences
σd (SD of Pair Differences)	1.8
Test Statistic Input Type	Zk
Zk (Current Test Statistic)	2.12

Output

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

Solve I ol.	blve For: Conditional Power										
Test Type:	2x2 Cross-Over	T-Test									
Hypotheses:	H0: $\delta \le \delta 0$ vs.	H1: $\delta > \delta 0$									
Power —		Samp	ole Size Mean Difference								
		Target	Look k	Null	Actual		Standard Deviation	Test Statistic			
Conditional	Predictive	N	nk	δ0	δ1	δ1 - δ0	σd	Zk	Alpha	Futility	
						0.6	1.8	2.12	0.025	0.14735	

The conditional power of 0.85265 matches the value calculated by hand.

Example 3 - Sample Size Reestimation

Suppose a 2x2 cross-over study has started. It is being conducted to detect a mean change of 1.00 at an alpha of 0.025 using a one-sided *t*-test and a null difference of 0.0. The target sample size is 52. The standard deviation is expected to be about 1.80. An interim analysis was run after half the data were collected. This analysis yields a z-test value of 2.12. This test value was the result of transforming the p-value of the t-test with the inverse, standard normal distribution. The sample standard deviation was 4.28. The data monitoring board would like to have sample size reestimated to obtain a conditional power of 0.8.

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	Sample Size Reestimation
Alternative Hypothesis	H1: δ > δ0 (One-Sided)
Conditional Power	0.8
Alpha	0.025
nk (Sample Size at Look k)	26
δ0 (Null Difference)	0
δ1 (Actual Difference to Detect)	1
Standard Deviation Input Type	Enter the SD of Pair Differences
σd (SD of Pair Differences)	4.28
Test Statistic Input Type	Zk
Zk (Current Test Statistic)	2.12

Output

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

Test Type:	Sample Size Reestimation 2x2 Cross-Over T-Test : $H0: \delta \leq \delta 0$ vs. $H1: \delta > \delta 0$									
Power		Samp	Sample Size		Mean Difference		Standard	Tool		
Powe	er .	Target	Look k	Null	Actual		Deviation	Test Statistic		
						T4 T0	σd	Zk	Alpha	F. 4:1:4.
Conditional	Predictive	N	nk	δ0	δ1	δ1 - δ0	oa	ZR.	Aipiia	Futility

Notice that the target sample size has increased from 52 to 85.