Chapter 443

Conditional Power of Two-Sample T-Tests for Non-Inferiority

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 predictive power for the case when a two-sample t-test is used to test non-inferiority between two means.

Technical Details

All details and assumptions usually made when using a two-sample non-inferiority t-test continue to be in force here.

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 $\theta$ at the end of the study, given the observed test statistic, $Z_k$, is computed as

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

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

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

where

- $\theta$ = the parameter being tested by the hypothesis
- $k$ = an interim stage at which the conditional power is computed ($k = 1, \ldots, K - 1$)
- $K$ = 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 $\alpha$.

Let $\mu_1$ and $\mu_2$ be the population means in groups 1 and 2, respectively. If we define $\delta = \mu_2 - \mu_1$, such that $\delta_0$ is the non-inferiority difference boundary, $\delta_1$ is the true population difference under the alternative hypothesis, and $\delta_k = \bar{x}_{2k} - \bar{x}_{1k}$ is the estimated mean difference from the observed data at stage $k$, then the parameter $\theta$ to test the one-sided non-inferiority alternative hypotheses of $H_1: \delta > \delta_0$ (higher means better) or $H_1: \delta < \delta_0$ (higher means worse) and other conditional power calculation components as outlined in Chang (2008) page 70 are

$$\theta = \delta_1 - \delta_0 \quad \text{(the expected difference under the alternative hypothesis)}$$

$$Z_k = (\delta_k - \delta_0) \sqrt{I_k} \quad \text{(the non-inferiority t-statistic computed from the observed data)}$$

$$I_k = \left(\frac{\sigma_1^2}{n_{1k}} + \frac{\sigma_2^2}{n_{2k}}\right)^{-1} \quad \text{(the interim information level)}$$

$$I_K = \left(\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}\right)^{-1} \quad \text{(the final information level)}$$

where

- $\hat{I}_k$ is the estimated information from the sample at stage $k$
- $n_{jk}$ is the sample size in group $j$ at stage $k$
- $n_j$ is the final sample size in group $j$
- $\sigma_j^2$ is the variance of group $j$. 

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Computing conditional power requires you to set δ₀, δ₁, σ₁, and σ₂, in addition to the current test statistic value, Zₖ. 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.

**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

\[
P_{uk} = \Phi \left( \frac{Z_k \sqrt{I_k} - z_{1-\alpha} \sqrt{I_k}}{\sqrt{I_k} - I_k} \right),
\]

and the general lower one-sided predictive power at stage \(k\) is given by

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

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

**Procedure Options**

This section describes the options that are specific to this procedure. These are located on the Design tab. For more information about the options of other tabs, go to the Procedure Window chapter.

**Design Tab**

The Design tab contains most of the parameters and options that you will be concerned with.

**Test**

**Higher Means Are**

Use this option to specify the direction of the non-inferiority test. If Higher Means are “Better”, the alternative hypothesis is \(H_1: \delta > \delta_0\), with \(\delta_0 < 0\). If Higher Means are “Worse”, the alternative hypothesis is \(H_1: \delta < \delta_0\), with \(\delta_0 > 0\).

**Alpha**

**Alpha**

This option specifies one or more values for the probability of a type-I error at the end of the study. A type-I error occurs when a true null hypothesis is rejected.

Values must be between zero and one. Historically, the value of 0.05 has been used for two-sided tests and 0.025 for one-sided tests. You may enter a range of values such as 0.01 0.05 0.10 or 0.01 to 0.10 by 0.01.
Sample Size

N1 (Group 1 Target Sample Size)
Enter one or more values of the target sample size of group 1, the total number of subjects planned for this group. This value must be an integer greater than one. Note that you may enter a list of values using the syntax 50,100,150,200,250 or 50 to 250 by 50.

N2 (Group 2 Target Sample Size)
Enter a value (or range of values) for the target sample size of group 2 or enter Use R to calculate N2 from N1. You may enter a range of values such as 10 to 100 by 10.

- **Use R**
  
  If Use R is entered here, N2 is calculated using the formula
  
  \[ N2 = \lceil R(N1) \rceil \]
  
  where R is the Sample Allocation Ratio and the operator \( \lceil \) is the first integer greater than or equal to Y. For example, if you want \( N1 = N2 \), select Use R and set \( R = 1 \).

R (Sample Allocation Ratio)
Enter a value (or range of values) for R, the allocation ratio between samples. This value is only used when N2 is set to Use R.

When used, N2 is calculated from N1 using the formula: \( N2 = \lceil R(N1) \rceil \) where \( \lceil \) is the next integer greater than or equal to Y. Note that setting \( R = 1.0 \) forces \( N2 = N1 \).

n1k (Group 1 Sample Size at Look k)
Enter the group 1 sample size obtained through look k. If this value is greater than N1, the value of N1 is increased to this amount.

n2k (Group 2 Sample Size at Look k)
Enter the group 2 sample size obtained through look k. If this value is greater than N2, the value of N2 is increased to this amount.

Equal Sample Sizes
If n1k is entered here, n2k is set equal to n1k.

Effect Size – Mean Difference

\( \delta_0 \) (Non-Inferiority Difference)
This is the difference bound of the non-inferiority test hypothesis. When Higher Means are “Better”, the alternative hypothesis is \( H1: \delta > \delta_0 \) and you can enter values in the range \( \delta_0 < 0 \) and \( \delta_0 < \delta_1 \). When Higher Means are “Worse”, the alternative hypothesis is \( H1: \delta < \delta_0 \) and you can enter values in the range \( \delta_0 > 0 \) and \( \delta_0 > \delta_1 \).

\( \delta_1 \) (Actual Difference to Detect)
This is the actual difference between the treatment mean and the reference mean at which power is calculated.

For non-inferiority tests, this value is often set to zero. When this value is non-zero, care should be taken that this value is consistent with the alternative hypothesis.

When Higher Means are “Better”, \( \delta_1 > \delta_0 \). When Higher Means are “Worse”, \( \delta_1 < \delta_0 \).
Effect Size – Standard Deviations

σ₁ (Standard Deviation of Group 1)
Enter a value for σ₁, the standard deviation of group 1. This value is controversial. Some think you should enter the value used in planning the study. Others think you should use the value estimated from the data obtained so far. Still others think you should use a confidence limit for σ₁ created from the current sample.

You can enter a range of values such as 1 2 3 or 1 to 5 by 1.

Press the σ button to the right to load the Standard Deviation Estimator window.

σ₂ (Standard Deviation of Group 2)
Enter a value for σ₂, the standard deviation of group 2. This value is controversial. Some think you should enter the value used in planning the study. Others think you should use the value estimated from the data obtained so far. Still others think you should use a confidence limit for σ₁ created from the current sample.

You can enter a range of values such as 1 2 3 or 1 to 5 by 1.

Equal Variance
Enter ‘σ₁’ to set σ₂ = σ₁.

Effect Size – Current Test Statistic

Zk (Current Test Statistic)
Enter the value of the T-statistic calculated from the data that have been obtained so far. The test statistic is calculated as

\[ Z_k = (\bar{x}_{2k} - \bar{x}_{1k} - \delta_0) \times \sqrt{\hat{\sigma}_k} \]

where \( \bar{x}_{1k} \) and \( \bar{x}_{2k} \) are the group 1 and group 2 sample means observed at stage k, respectively, and \( \hat{\sigma}_k \) is the information estimate at stage k.

Zk may be positive or negative. Usually, Zk ranges between -5 and 5.

The conditional power calculations used here are for large samples which are at least 20.
**Example 1 – Computing Conditional Power**

Suppose a study has been planned and is to be analyzed using a one-sided non-inferiority $t$-test against a lower difference bound of $\delta_0 = -1$ at an alpha of 0.025. The sample size is 60. The standard deviations are expected to be about 4. 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 like to have the conditional power calculated for mean changes, $\delta_1 = \mu_2 - \mu_1$, of -0.5, 0, 0.5, 1, and 1.5.

**Setup**

This section presents the values of each of the parameters needed to run this example. First, from the PASS Home window, load the **Conditional Power of Two-Sample T-Tests for Non-Inferiority** procedure window by expanding **Means**, then **Two Independent Means**, then clicking on **Conditional Power**, and then clicking on **Conditional Power of Two-Sample T-Tests for Non-Inferiority**. You may make the appropriate entries as listed below, or open **Example 1** by going to the **File** menu and choosing **Open Example Template**.

<table>
<thead>
<tr>
<th>Option</th>
<th>Value</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Design Tab</strong></td>
<td></td>
</tr>
<tr>
<td>Higher Means Are</td>
<td>Better (H1: $\delta &gt; \delta_0$)</td>
</tr>
<tr>
<td>Alpha</td>
<td>0.025</td>
</tr>
<tr>
<td>N1 (Group 1 Target Sample Size)</td>
<td>60</td>
</tr>
<tr>
<td>N2 (Group 2 Target Sample Size)</td>
<td>Use R</td>
</tr>
<tr>
<td>R (Sample Allocation Ratio)</td>
<td>1.0</td>
</tr>
<tr>
<td>n1k (Group 1 Sample Size at Look k)</td>
<td>30</td>
</tr>
<tr>
<td>n2k (Group 2 Sample Size at Look k)</td>
<td>$n_{1k}$</td>
</tr>
<tr>
<td>$\delta_0$ (Non-Inferiority Difference)</td>
<td>-1</td>
</tr>
<tr>
<td>$\delta_1$ (Actual Difference to Detect)</td>
<td>-0.5 0 0.5 1 1.5</td>
</tr>
<tr>
<td>$\sigma_1$ (Standard Deviation of Group 1)</td>
<td>4</td>
</tr>
<tr>
<td>$\sigma_2$ (Standard Deviation of Group 2)</td>
<td>$\sigma_1$</td>
</tr>
<tr>
<td>Zk (Current Test Statistic)</td>
<td>2.12</td>
</tr>
</tbody>
</table>

**Annotated Output**

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

**Numeric Results**

<table>
<thead>
<tr>
<th>Cond Power</th>
<th>Pred Power</th>
<th>Total Sample Sizes</th>
<th>Current Sample Sizes</th>
<th>Non-Infer</th>
<th>Actual Diff</th>
<th>Std Dev</th>
<th>Test Stat</th>
<th>Alpha</th>
<th>Futility</th>
</tr>
</thead>
<tbody>
<tr>
<td>0.43342</td>
<td>0.85040</td>
<td>60/60</td>
<td>30/30</td>
<td>-1.0</td>
<td>-0.5</td>
<td>4.0/4.0</td>
<td>2.120</td>
<td>0.025</td>
<td>0.56658</td>
</tr>
<tr>
<td>0.62417</td>
<td>0.85040</td>
<td>60/60</td>
<td>30/30</td>
<td>-1.0</td>
<td>0.0</td>
<td>4.0/4.0</td>
<td>2.120</td>
<td>0.025</td>
<td>0.37583</td>
</tr>
<tr>
<td>0.78831</td>
<td>0.85040</td>
<td>60/60</td>
<td>30/30</td>
<td>-1.0</td>
<td>0.5</td>
<td>4.0/4.0</td>
<td>2.120</td>
<td>0.025</td>
<td>0.21169</td>
</tr>
<tr>
<td>0.90055</td>
<td>0.85040</td>
<td>60/60</td>
<td>30/30</td>
<td>-1.0</td>
<td>1.0</td>
<td>4.0/4.0</td>
<td>2.120</td>
<td>0.025</td>
<td>0.09945</td>
</tr>
<tr>
<td>0.96154</td>
<td>0.85040</td>
<td>60/60</td>
<td>30/30</td>
<td>-1.0</td>
<td>1.5</td>
<td>4.0/4.0</td>
<td>2.120</td>
<td>0.025</td>
<td>0.03846</td>
</tr>
</tbody>
</table>

**References**

Report Definitions
Conditional Power is the probability of rejecting a false null hypothesis at the end of the study given the data that have emerged so far.
Predictive Power is the result of averaging the conditional power over the posterior distribution of the effect size.
N1|N2 are the anticipated total sample sizes of groups 1 and 2, respectively.
n1|n2 are the sample sizes of groups 1 and 2, respectively, obtained through stage k.
δ = μ2 - μ1 is the mean difference.
δ0 is the non-inferiority difference used to construct the hypotheses.
δ1 is the actual difference to detect under the alternative hypothesis at which conditional power is calculated.
σ1|σ2 are the standard deviations of groups 1 and 2, respectively.
Zk is the value of the test statistic from the observed data at stage k.
Alpha is the probability of rejecting a true null hypothesis.
Futility is one minus the conditional power. A value greater than 0.9 or 0.8 indicates the study should be stopped because there is little chance of achieving statistical significance.

Summary Statements
The first 30 of 60 subjects in group 1 and 30 of 60 subjects in group 2 achieve 43.342% conditional power to detect a difference of δ1 = -0.5 using a T-test for non-inferiority with a difference bound of δ0 = -1.0, group 1 standard deviation of σ1 = 4.0, group 2 standard deviation of σ2 =4.0, and a significance level of alpha = 0.025. The value of the test statistic from data that have emerged so far is Zk = 2.120. The futility index is 0.56658.

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 δ1.
Example 2 – Validation

We could not find an example of a conditional power calculation for a one-sided non-inferiority $t$-test in the literature. Since the calculations are relatively simple, we will validate the calculation of the first scenario ($\theta = \delta_1 - \delta_0 = 0.5$) of Example 1 by hand.

In this case

$$I_k = \left( \frac{\sigma^2_1}{n_{1k}} + \frac{\sigma^2_2}{n_{2k}} \right)^{-1}$$

$$= \left( \frac{16}{30} + \frac{16}{30} \right)^{-1}$$

$$= \frac{30 + 30}{16}$$

$$= 0.9375$$

$$I_K = \left( \frac{\sigma^2_1}{n_1} + \frac{\sigma^2_2}{n_2} \right)^{-1}$$

$$= \left( \frac{16}{60} + \frac{16}{60} \right)^{-1}$$

$$= \frac{60 + 60}{16}$$

$$= 1.875$$

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

$$= \Phi \left( \frac{2.12 \sqrt{0.9375} - 1.9599640 \sqrt{1.875} + 0.5 (1.875 - 0.9375)}{\sqrt{1.875 - 0.9375}} \right)$$

$$= \Phi(-0.1676848)$$

$$= 0.4334156$$

This value matches the first line of the report in Example 1.