Tests for Paired Means (Simulation)

Introduction

This procedure allows you to study the power and sample size of several statistical tests of the null hypothesis that the difference between two correlated means is equal to a specific value versus the alternative that it is greater than, less than, or not equal to that value. The paired t-test is commonly used in this situation. Other tests have been developed for the case when the data are not normally distributed. These additional tests include the Wilcoxon signed-ranks test, the sign test, and the computer-intensive bootstrap test.

Paired data may occur because two measurements are made on the same subject or because measurements are made on two subjects that have been matched according to other, often demographic, variables. Hypothesis tests on paired data can be analyzed by considering the differences between the paired items. The distribution of differences is usually symmetric. In fact, the distribution must be symmetric if the individual distributions of the two items are identical. Hence, the paired t-test and the Wilcoxon signed-rank test are appropriate for paired data even when the distributions of the individual items are not normal.

The details of the power analysis of the paired t-test using analytic techniques are presented in another **PASS** chapter and they won't be duplicated here. This chapter will only consider power analysis using computer simulation.

Technical Details

Computer simulation allows one to estimate the power and significance level that is actually achieved by a test procedure in situations that are not mathematically tractable. Computer simulation was once limited to mainframe computers. Currently, due to increased computer speeds, simulation studies can be completed on desktop and laptop computers in a reasonable period of time.

The steps to a simulation study are as follows:

- 1. Specify the method by which the test is to be carried out. This includes specifying how the test statistic is calculated and how the significance level is specified.
- 2. Generate a random sample, $X_1, X_2, ..., X_n$, from the distribution specified by the <u>alternative</u> hypothesis. Calculate the test statistic from the simulated data and determine if the null hypothesis is accepted or rejected. Each of these samples is used to calculate the power of the test.
- 3. Generate a random sample, Y_1, Y_2, \ldots, Y_n , from the distribution specified by the <u>null</u> hypothesis. Calculate the test statistic from the simulated data and determine if the null hypothesis is accepted or rejected. Each of these samples is used to calculate the significance level of the test.
- 4. Repeat steps 2 and 3 several thousand times, tabulating the number of times the simulated data lead to a rejection of the null hypothesis. The power is the proportion of simulation samples in step 2 that lead to rejection. The significance level is the proportion of simulated samples in step 3 that lead to rejection.

Data Distributions

A wide variety of distributions may be studied. These distributions can vary in skewness, elongation, or other features such as bimodality. A detailed discussion of the distributions that may be used in the simulation is provided in the chapter 'Data Simulator'.

Test Statistics

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

Paired T-Test

The paired *t*-test assumes that the paired differences, X_i , are a simple random sample from a population of normally-distributed difference 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*-test proceeds as follows

$$t_{n-1} = \frac{\bar{X} - D0}{s_{\bar{X}} / \sqrt{n}}$$

where

$$\bar{X} = \frac{\sum_{i=1}^{n} X_i}{n},$$

$$s_{\bar{X}} = \sqrt{\frac{\sum_{i=1}^{n} (X_i - \bar{X})^2}{n - 1}},$$

The significance of the test statistic is determined by computing the p-value. If this p-value is less than a specified level (usually 0.05), the hypothesis is rejected. Otherwise, no conclusion can be reached.

Wilcoxon Signed-Rank Test

The Wilcoxon signed-rank test is a popular, nonparametric substitute for the *t*-test. It assumes that the data follow a symmetric distribution. The test is computed using the following steps.

- 1. Subtract the hypothesized difference, *D0*, from each data value. Rank the values according to their absolute values.
- 2. Compute the sum of the positive ranks Sp and the sum of the negative ranks Sn. The test statistic, W_R , is the minimum of Sp and Sn.

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3. Compute the mean and standard deviation of W_R using the formulas

$$\mu_{W_R} = \frac{n(n+1)}{4}$$
$$\sigma_{W_R} = \sqrt{\frac{n(n+1)(2n+1)}{24} - \frac{\sum t^3 - \sum t}{48}}$$

where *t* represents the number of times the *i*th value occurs.

4. Compute the *z*-value using

$$z_W = \frac{W_R - \mu_{W_R}}{\sigma_{W_R}}$$

For cases when *n* is less than 38, the significance level is found from a table of exact probabilities for the Wilcoxon test. When *n* is greater than or equal to 38, the significance of the test statistic is determined by comparing the *z* value to a normal probability table. If this p-value is less than a specified level (often 0.05), the null hypothesis is rejected. Otherwise, no conclusion can be reached.

Sign Test

The sign test is popular because it is simple to compute. This test assumes that the data all follow the same distribution. The test is computed using the following steps.

- 1. Count the number of values strictly greater than *D*0. Call this value *X*.
- 2. Count the number of values strictly less than D0. Call this value Y.
- 3. Set m = X + Y.
- 4. Under the null hypothesis, *X* is distributed as a binomial random variable with a proportion of 0.5 and sample size of *m*.

The significance of *X* is calculated using binomial probabilities.

Bootstrap Test

The one-sample bootstrap procedure for testing whether the mean is equal to a specific value is given in Efron & Tibshirani (1993), pages 224-227. The bootstrap procedure is as follows.

- 1. Compute the mean of the sample. Call it \overline{X} .
- 2. Compute the t-value using the standard t-test. The formula for this computation is

$$t_X = \frac{\bar{X} - D0}{s_{\bar{X}} / \sqrt{n}}$$

where D0 is the hypothesized difference.

3. Draw a random, with-replacement sample of size *n* from the original *X* values. Call this sample Y_1, Y_2, \dots, Y_n .

4. Compute the t-value of this bootstrap sample using the formula

$$t_Y = \frac{\bar{Y} - \bar{X}}{s_{\bar{Y}} / \sqrt{n}}$$

- 5. For a two-tailed test, if $|t_Y| > |t_X|$ then add one to a counter variable, *A*.
- 6. Repeat steps 3 5 *B* times. *B* may be anywhere from 100 to 10,000.
- 7. Compute the *p*-value of the bootstrap test as (A + 1) / (B + 1)
- 8. Steps 1 7 complete one simulation iteration. Repeat these steps *M* times, where *M* is the number of simulations. The power and significance level are equal to the percent of the time the *p*-value is less than the nominal alpha of the test in their respective simulations.

Note that the bootstrap test is a time-consuming test to analyze, especially if you set *B* to a value much larger than 100.

Standard Deviations

Care must be used when either the null or alternative distribution is not normal. In these cases, the standard deviation is usually not specified directly. For example, you might use a gamma distribution with a shape parameter of 1.5 and a mean of 4 as the null distribution and a gamma distribution with the same shape parameter and a mean of 5 as the alternative distribution. This allows you to compare the two means. However, note that although the shape parameters are constant, the standard deviations are not. In cases such as this, the null and alternatives not only have different means, but different standard deviations!

The Standard Deviation of Paired Differences (σ) when using Simple Input

If you have results from a previous (or pilot) study, use the estimate of the standard deviation of paired differences, σ , from the study. Another reasonable (but somewhat rough) estimate of σ may be obtained using the range of paired differences as

$$\sigma = \frac{Range}{4}$$

If you have estimates of the expected standard deviations of the paired variables (σ_1 and σ_2) and the Pearson correlation between the paired variables (ρ), the standard deviation of paired differences (σ) may be calculated using the equation

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

such that

$$\sigma = \sqrt{\sigma_1^2 + \sigma_2^2 - 2\rho\sigma_1\sigma_2} \,.$$

If $\sigma_1 = \sigma_2 = \sigma_x$, then this formula reduces to

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

such that

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

If you have an estimate of the within-subject population standard deviation (σ_w), then σ may be calculated using the equation

$$\sigma^2 = 2\sigma_w^2$$

such that

$$\sigma = \sqrt{2\sigma_w^2} \,.$$

 σ_w is often estimated by the square root of the within mean square error (WMSE) from a repeated measures ANOVA.

Example 1 – Power at Various Sample Sizes

Researchers are planning a pre-post experiment to test whether the difference in response to a certain drug is different from zero. The researchers will use a paired t-test with an alpha level of 0.05. They want to compare the power at sample sizes of 50, 100, and 150 when the shift in the means is 0.6 from pre-test to post-test. They assume that the data are normally distributed with a standard deviation of paired differences of 2.53. Since this is an exploratory analysis, they set the number of simulation iterations to 2000.

A researcher is planning an experiment to test whether the mean response level to a certain drug is significantly different from zero. The researcher wants to use a t-test with an alpha level of 0.05. He wants to compute the power at various sample sizes from 5 to 40, assuming the true mean is one. He assumes that the data are normally distributed with a standard deviation of 2.

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 (a or b)** 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
Alternative Hypothesis	Two-Sided (H1: δ ≠ δ0)
Test Type	T-Test
Simulations	2000
Random Seed	5379518 (for Reproducibility)
Alpha	0.05
N (Sample Size)	50 100 150
Input Type	Simple
Distribution to Simulate	Normal
δ0 (Null Mean of Paired Diffs)	0
δ1 (Actual Mean of Paired Diffs)	0.6
Standard Deviation Input Type	Enter the SD of Paired Differences
σ (SD of Paired Differences)	2.53

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

Numeric Results

Numeric Res	ults							
Solve For: Hypotheses: Test Statistic: Simulated Dis	stribution:	Power H0: δ = Paired Normal	⊧δ0 vs. H T-Test I	1: δ ≠ δ0				
C.	ample	Mea	n of Paired	Differences	Standard	Effect	Al	nha
Bower	Size	Null	Actual	Difference	Deviation	Size		
			01	01-00		0.00745		
(0.02148) [0.3	50 38002 0.42	0 2298]	0.6	0.6	2.53	0.23715	0.05	0.048 (0.009) [0.038 0.057
0.6365	100	0	0.6	0.6	2.53	0.23715	0.05	0.046
(0.02108) [0.6	61542 0.65	5758]						(0.009) [0.036 0.055
0.8290 (0.0165) [0.81	150 25 0.8455	0	0.6	0.6	2.53	0.23715	0.05	0.054 (0.01) [0.044 0.064]

Simulations: 2000. Run Time: 2.55 seconds. User-Entered Random Seed: 5379518

Power	The probability of rejecting a false null hypothesis when the alternative hypothesis is true. The second row provides the precision and a 95% confidence interval for Power, (Power Precision) [95% LCL and UCL], based on the size of the simulation.
N	The size of the sample drawn from the population.
δ0	The value of the population mean of paired differences under the null hypothesis.
δ1	The actual value of the population mean of paired differences at which power and sample size are calculated.
δ1 - δ0	The difference between the actual and null mean of paired differences.
σ	The standard deviation of paired differences for the population. It measures the variability in the population.
δ1 - δ0 / σ	The Effect Size, which is the relative magnitude of the effect.
Target Alpha	The probability of rejecting a true null hypothesis. It is set by the user.
Actual Alpha	The alpha level that was actually achieved by the experiment. The second row provides the precision and a 95% confidence interval for Alpha, (Alpha Precision) [95% LCL and UCL], based on the size of the simulation.

Summary Statements

A paired design will be used to test whether the paired mean difference (δ) is different from 0 (H0: δ = 0 versus H1: $\delta \neq 0$). The comparison will be made using a two-sided, paired-sample t-test of the paired differences, with a Type I error rate (α) of 0.05. The standard deviation of the paired differences is assumed to be 2.53. To detect a paired mean difference of 0.6 with a sample size of 50 pairs, the power is 0.4015. These results are based on 2000 simulations (Monte Carlo samples) from the Normal distribution.

Dropout-Inflated Sample Size

Dropout Rate	Sample Size	Dropout- Inflated Enrollment Sample Size N'	Expected Number of Dropouts D	
20%	50	63	13	
20%	100	125	25	
20%	150	188	38	
Dropout Rate	The percentage of subj and for whom no resp	ects (or items) that a ponse data will be co	re expected to be llected (i.e., will b	e lost at random during the course of the study e treated as "missing"). Abbreviated as DR.
Ν	The evaluable sample sout of the N' subjects	size at which power i that are enrolled in t	s computed (as e	ntered by the user). If N subjects are evaluated
N'	The total number of sub based on the assume N' always rounded up	bjects that should be ad dropout rate. N' is b. (See Julious, S.A.	enrolled in the st calculated by infl (2010) pages 52-	udy in order to obtain N evaluable subjects, ating N using the formula N' = N / (1 - DR), with 53, or Chow, S.C., Shao, J., Wang, H., and

Dropout Summary Statements

Lokhnygina, Y. (2018) pages 32-33.)

The expected number of dropouts. D = N' - N.

Anticipating a 20% dropout rate, 63 subjects should be enrolled to obtain a final sample size of 50 subjects.

References

D

Chow, S.C., Shao, J., Wang, H., and Lokhnygina, Y. 2018. Sample Size Calculations in Clinical Research, Third Edition. Taylor & Francis/CRC. Boca Raton, Florida.

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

Machin, D., Campbell, M., Fayers, P., and Pinol, A. 1997. Sample Size Tables for Clinical Studies, 2nd Edition. Blackwell Science. Malden, MA.

This report shows the estimated power for each scenario. The first row shows the parameter settings and the estimated power and significance level (Actual Alpha). Note that because these are results of a simulation study, the computed power and alpha will vary from run to run. Thus, another report obtained using the same input parameters will be slightly different than the one above.

The second row shows two 95% confidence intervals in brackets: the first for the power and the second for the significance level. Half the width of each confidence interval is given in parentheses as a fundamental measure of the accuracy of the simulation. As the number of simulations is increased, the width of the confidence interval will decrease.

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Plots Section



This plot shows the relationship between sample size and power.

Example 2 – Finding the Sample Size

Continuing with Example 1, the researchers want to determine how large a sample is needed to obtain a power of 0.90?

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	Sample Size
Alternative Hypothesis	Two-Sided (H1: δ ≠ δ0)
Test Type	T-Test
Simulations	2000
Random Seed	2919111 (for Reproducibility)
Power	0.90
Alpha	0.05
Input Type	Simple
Distribution to Simulate	Normal
δ0 (Null Mean of Paired Diffs)	0
δ1 (Actual Mean of Paired Diffs)	0.6
Standard Deviation Input Type	Enter the SD of Paired Differences
σ (SD of Paired Differences)	2.53

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

Solve Fo Hypothes Test Stat Simulate	r: ses: tistic: d Distribution:	Sampl H0: δ = Paired Norma	e Size = δ0 vs. H T-Test I	1: δ ≠ δ0					
	Sample	Mea	n of Paired	Differences	Standard	Effect	۵۱	nha	
Power	Size	Null δ0	Actual δ1	Difference δ1 - δ0	Deviation σ	Size δ1 - δ0 / σ	 Target	Actual	
0.891	187) [0 87734 0 9	0	0.6	0.6	2.53	0.23715	0.05	0.046	[0.037 0.055]

User-Entered Random Seed: 2919111

The required sample size of 187 achieved a power of 0.891. The power of 0.891 is less than the target value of 0.900 because the sample size search algorithm re-simulates the power for the final sample size. Thus, it is possible for the search algorithm to converge to a sample size which exhibits the desired power, but then on a succeeding simulation to achieve a power that is slightly less than the target. To achieve more accuracy, a reasonable strategy would be to run simulations to obtain the powers using N's from 180 to 200 using a simulation size of 5000 or greater.

Example 3 – Comparative Results

Continuing with Example 2, the researchers want to study the characteristics of alternative test statistics.

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.

Design Tab	
Solve For	Power
Alternative Hypothesis	Two-Sided (H1: δ ≠ δ0)
Test Type	T-Test
Simulations	2000
Random Seed	4985023 (for Reproducibility)
Alpha	0.05
N (Sample Size)	50 100 150 200
Input Type	Simple
Distribution to Simulate	Normal
δ0 (Null Mean of Paired Diffs)	0
δ1 (Actual Mean of Paired Diffs)	0.6
Standard Deviation Input Type	. Enter the SD of Paired Differences
σ (SD of Paired Differences)	2.53
Reports Tab	
Show Comparative Reports	Checked
Include T-Test Results	Checked
Include Wilcoxon & Sign Test Results	Checked
Comparative Plots Tab	
Show Comparative Plots	Checked

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

-	~	
Power	Com	parison

Hypotheses: Simulated Distribution:	H0: δ = δ0 vs Normal	s. H1:δ≠δ0		
Mea Paired Dif Sample	n of iferences	Standard	F	Power

Sizo	NUU	Actual	Dovistion	Torget	i owei			
N	δ0	δ1	σ	Alpha	T-Test	Wilcoxon	Sign	
50	0	0.6	2.53	0.05	0.3505	0.3310	0.1930	
100	0	0.6	2.53	0.05	0.6495	0.6310	0.4130	
150	0	0.6	2.53	0.05	0.8215	0.8030	0.6015	
200	0	0.6	2.53	0.05	0.9080	0.8935	0.7185	

Simulations: 2000. Run Time: 4.15 seconds. User-Entered Random Seed: 4985023

Alpha Comparison

Hypotheses:	H0: δ = δ0	vs.	H1: δ ≠ δ0	
Simulated Distribution:	Normal			

Sample	Mean of Paired Differences		Standard	Alpha				
Size N	Null δ0	Actual δ1	Deviation σ	Target	T-Test	Wilcoxon	Sign	
50	0	0.6	2.53	0.05	0.053	0.051	0.035	
100	0	0.6	2.53	0.05	0.047	0.046	0.035	
150	0	0.6	2.53	0.05	0.053	0.053	0.048	
200	0	0.6	2.53	0.05	0.058	0.054	0.040	

Simulations: 2000. Run Time: 4.15 seconds. User-Entered Random Seed: 4985023

Comparative Plots



These results show that for paired data, the t-test and Wilcoxon test have very similar power and alpha values. The sign test is both less accurate and less powerful.

Example 4 – Validation using Zar (1984)

Zar (1984), pages 111-112, presents an example in which $\delta 0 = 0.0$, $\delta 1 = 1.0$, $\sigma = 1.25$, alpha = 0.05, and N = 12. Zar obtains an approximate power of 0.72. We will validate this procedure by running this example. To make certain that the results are very accurate, the number of simulations will be set to 10,000.

For reproducibility, we'll use a random seed of 6015683.

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.

Power
.Two-Sided (H1: δ ≠ δ0)
.T-Test
.10000
.6015683 (for Reproducibility)
.0.05
.12
Simple
Normal
.0
.1
Enter the SD of Paired Differences
.1.25

Output

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

Solve For: Hypotheses: Test Statistic: Simulated Distribution:		Power H0: δ = Paired Norma	⊧δ0 vs. H T-Test I	1: δ ≠ δ0					
	Sampla	Mea	n of Paired	Differences	Standard	Effoot	A 1.	aha	
Sample	Sample	Null	Actual	Difference	Deviation	Size	Aij		
Power	Ν	δ0	δ1	δ1 - δ0	σ	δ1 - δ0 / σ	Target	Actual	
0.7162 (0.00884) [12 0.70736 0.72	0 2504]	1	1	1.25	0.8	0.05	0.049 (0.004) [0.045 0.053]

This simulation obtained a power of 0.7162 which rounds to the 0.72 computed by Zar.

Example 5 – Validation using Machin (1997)

Machin, et. al. (1997), page 37, presents an example in which μ 0 = 0.0, μ 1 = 0.2, σ = 1.0, alpha = 0.05, and power = 0.80. They obtain a sample size of 199.

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 5** 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	Sample Size
Alternative Hypothesis	Two-Sided (H1: δ ≠ δ0)
Test Type	T-Test
Simulations	2000
Random Seed	6030438 (for Reproducibility)
Power	0.80
Alpha	0.05
Input Type	Simple
Distribution to Simulate	Normal
δ0 (Null Mean of Paired Diffs)	0
δ1 (Actual Mean of Paired Diffs)	0.2
Standard Deviation Input Type	Enter the SD of Paired Differences
σ (SD of Paired Differences)	1

Output

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

Solve For: Hypotheses: Test Statistic: Simulated Distribution:		Sampl H0: δ = Paired Norma	e Size = δ0 vs. H I T-Test al	1: δ ≠ δ0					
Sample		Mea	n of Paired	Differences	Standard Effect	Al			
Power	Size N	Null δ0	Actual δ1	Difference δ1 - δ0	Deviation σ	Size δ1 - δ0 / σ	Target	Actual	
0.815	199 0.79798 0.8	0 332021	0.2	0.2	1	0.2	0.05	0.046 (0.009)	[0.036 0.055]

Simulations: 2000. Run Time: 13.01 seconds. User-Entered Random Seed: 6030438

The sample size result matches the value of Machin (1997). If you run the simulation multiple times, you'll come up with values right around 199.

Example 6 – Non-Inferiority Test

A non-inferiority test is appropriate when you want to show that a new treatment is no worse than the standard. For example, suppose that a standard diagnostic test has an average score of 70. Unfortunately, this diagnostic test is expensive. A promising new diagnostic test must be compared to the standard. Researchers want to show that it is no worse than the standard.

Because of many benefits from the new test, clinicians are willing to adopt it even if it is slightly less accurate than the current test. How much less can the score of the new treatment be and still be adopted? Should it be adopted if the difference is -1? -2? -5? -10? There is an amount below 0 at which the difference between the two treatments is no longer considered ignorable. After thoughtful discussion with several clinicians, the *margin of non-inferiority* is set to -5.

The developers decided to use a paired t-test. They must design an experiment to test the hypothesis that the average difference between the two tests is greater than -5. The statistical hypothesis to be tested is

 $H_0: A - B \le -5$ versus $H_1: A - B > -5$

where A represents the mean of the new test and B represents the mean of the standard test. Notice that when the null hypothesis is rejected, the conclusion is that the average difference is greater than -5.

Experience has shown that the standard deviation of paired differences is 6.32. Following proper procedure, the researchers decide to use a significance level of 0.025 for this one-sided test to keep it comparable to the usual value of 0.05 for a two-sided test. They decide to look at the power for sample sizes of 5, 10, 15, 20, and 25 subjects. They decide to compute the power for the case when the two tests are equal.

Setup

If the procedure window is not already open, use the PASS Home window to open it. The parameters for this example are listed below and are stored in the **Example 6** 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
Alternative Hypothesis	Οne-Sided (H1: δ > δ0)
Test Type	T-Test
Simulations	2000
Random Seed	7466448 (for Reproducibility)
Alpha	0.025
N (Sample Size)	5 10 15 20 25
Input Type	Simple
Distribution to Simulate	Normal
δ0 (Null Mean of Paired Diffs)	5
δ1 (Actual Mean of Paired Diffs)	0
σ (Std Dev of Paired Diffs)	6.32

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

Numeric Results

Test Statistic: Paired T-Test Simulated Distribution: Normal	Solve For: Hypotheses: Test Statistic: Simulated Distribution:	Power H0: δ ≤ δ0 vs. H1: δ > δ0 Paired T-Test Normal
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		Mea	n of Paired	Differences					
	Sample Size	Null	Actual	Difference	Standard Deviation	Effect		oha	
Power N	N	δ0	δ1	δ1 - δ0	σ	σ δ1 - δ0 / σ	Target	Actual	
0.2880 (0.01985)	5 [0.26815 0.3	-5 0785]	0	5	6.32	0.79114	0.025	0.026 (0.007) [0.019 0.033]
0.6115 (0.02136)	10 [0.59014 0.6	-5 3286]	0	5	6.32	0.79114	0.025	0.020 (0.006) [0.013 0.026]
0.7985 (0.01758)	15 [0.78092 0.8	-5 1608]	0	5	6.32	0.79114	0.025	0.029 (0.007) [0.022 0.036]
0.9210 (0.01182)	20 [0.90918 0.9	-5 3282]	0	5	6.32	0.79114	0.025	0.027 (0.007) [0.02 0.034]
0.9675 (0.00777)	25 [0.95973 0.9	-5 17527]	0	5	6.32	0.79114	0.025	0.029 (0.007) [0.022 0.036]

Simulations: 2000. Run Time: 1.50 seconds. User-Entered Random Seed: 7466448

Plots



We see that a power of 0.8 is achieved at about 15 subjects, while a power of 0.9 requires about 20 subjects.