# Tests for One-Sample Sensitivity and Specificity 

## Introduction

The power analysis of a diagnostic test is often based on the sensitivity and specificity of the test. In such a test, the outcome of the diagnostic screening test is compared to the gold standard. In the common casecontrol study, the gold standard must be known before. In a prospective study, the gold standard is determined subsequent to the study, so the case-control framework is not appropriate. In this procedure, the power analysis and sample size requirements of such a design are considered.
In a prospective study, a group of $n$ subjects is obtained. Some of the subjects have the disease (condition of interest) and some do not. Each subject is given the diagnostic test for the disease. Subsequently, a gold standard test is used to obtain the true presence or absence of the disease. The gold standard may be a more expensive test, or it may be following the subject to determine if the disease status becomes more apparent.

The measures of diagnostic accuracy are sensitivity and specificity. Sensitivity (Se) is the probability that the diagnostic test is positive for the disease, given that the subject actually has the disease. Specificity ( Sp ) is the probability that the diagnostic test is negative, given that the subject does not have the disease.
Mathematically,

$$
\begin{aligned}
& \text { Sensitivity }=\operatorname{Pr}(+ \text { test } \mid \text { disease }) \\
& \text { Specificity }=\operatorname{Pr}(- \text { test } \mid \text { no disease })
\end{aligned}
$$

Li and Fine (2004) present sample size methodology for testing sensitivity and specificity using a prospective design. Their methodology will be used here. Other useful references are Obuchowski and Zhou (2002), Machin, Campbell, Tan, and Tan (2009), and Zhou, Obuchowski, and McClish (2002).

## Prospective Study Design

In a prospective study, a group of $n$ subjects is split into two groups: those with the disease of interest and those without it. Suppose a particular sample has $n_{1}$ with the disease and $n_{2}$ without the disease. A diagnostic test is administered to each subject (usually before the disease status is determined) and its output is recorded. The diagnostic test outcome is either positive or negative for the disease. Suppose that in the $n_{1}$ subjects with the disease, $s_{1}$ have a positive test outcome and $s_{2}$ have a negative outcome. Similarly, in the $n_{2}$ subjects without the disease $r_{1}$ have positive outcomes and $r_{2}$ have negative outcomes. Sensitivity is estimated by $s_{1} / n_{1}$ and specificity is estimated by $r_{2} / n_{2}$. A useful diagnostic test has high values of both $S e$ and $S p$.

Conditional on the values of $n_{1}$ and $n_{2}, \mathrm{~s}_{1}$ is $\operatorname{Binomial}\left(n_{1}, S e\right)$. Thus, a one-sided test of the statistical hypothesis $H_{0}: S e=S e_{0}$ versus $H_{1}: S e=S e_{1}>S e_{0}$ can be carried out using a binomial test. Hence, the power analysis is based on the binomial distribution conditional on the value of $n_{1}$. Similarly, the hypothesis test $H_{0}: S p=S p_{0}$ versus $H_{1}: S p=S p_{1}>S p_{0}$ is also based on the binomial distribution.

## Binomial Model

A binomial variable should exhibit the following four properties:

1. The variable is binary --- it can take on one of two possible values.
2. The variable is observed a known number of times. Each observation or replication is called a Bernoulli trial. The number of replications is $n$. The number of times that the outcome of interest is observed is $r$. Thus, $r$ takes on the possible values $0,1,2, \ldots, n$.
3. The probability, $P$, that the outcome of interest occurs is constant for each trial.
4. The trials are independent. The outcome of one trial does not influence the outcome of the any other trial.

The binomial probability is calculated using the formula

$$
b(x \mid n, \pi)=\binom{n}{x} \pi^{x}(1-\pi)^{n-x} \text { where }\binom{n}{r}=\frac{n!}{r!(n-r)!} .
$$

Using the binomial probability formula, the sample size necessary to meet both a significance level and a power requirement may be found by solving the following to equations simultaneously:

$$
\begin{gathered}
\text { Significance Level } B\left(s_{1}>s_{\alpha} \mid n_{1}, S e_{0}\right)=\alpha \\
\qquad B\left(s_{1}>s_{\alpha} \mid n_{1}, S e_{1}\right)=1-\beta
\end{gathered}
$$

where

$$
B\left(s_{1}>s_{\alpha} \mid n_{1}, \pi\right)=\sum_{x=s_{\alpha}+1}^{n_{1}} b\left(x \mid n_{1}, S e_{0}\right)
$$

A similar calculation can be made for the specificity. Further details of this procedure are given in the Tests for One Proportion chapter.

Note that these formulas give $n_{1}$, not $n$. To obtain $n, n_{1}$ is inflated by the disease prevalence $P$ to obtain $n=$ $n_{1} / P$. This is called Method 0 in the paper by Li and Fine (2004).

## Example 1 - Finding the Power

Suppose that diagnosing a certain type of cancer has required expensive and invasive test procedure. The sensitivity of this procedure is $71 \%$ and the specificity is $82 \%$. A new diagnostic test has been developed that is much less expensive and invasive. The developers of the test want to design a prospective study to compare the old and new tests using a two-sided binomial test with a significance level of 0.05 . They want to consider changes in sensitivity of $10 \%, 15 \%, 20 \%$, and $25 \%$. These changes translate to sensitivities of $79.20 \%, 81.65 \%, 85.20 \%$, and $88.75 \%$. The prevalence of the disease in the population of interest is $6 \%$. The power will be determined for trials with sample sizes between 300 and 3000 incremented by 300 . They want to consider a $10 \%$ increase in specificity which is $90.2 \%$.

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


## Output

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

## Numeric Reports

## Numeric Results

| Solve For: | Power |
| :--- | :--- |
| Sensitivity Hypotheses: | $\mathrm{HO}: \mathrm{Se}=\mathrm{Se} 0$ vs. $\mathrm{H} 1: \mathrm{Se} \neq \mathrm{Se0}$ |
| Specificity Hypotheses: | $\mathrm{HO}: \mathrm{Sp}=\mathrm{Sp0}$ vs. $\mathrm{H} 1: \mathrm{Sp} \neq \mathrm{Sp0}$ |
| Test Statistic: | Binomial Test |


| Power |  | Sample Size N1 and N | Sensitivity |  | Specificity |  | Alpha |  |  | PrevaIence P |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  | H0 | H1 | H0 | H1 |  | Se |  |  |
| Sens. | Spec. |  | Se0 | Se1 | Sp0 | Sp1 | Target | Actual | Actual |  |
| 0.0874 | 0.9719 |  | 18300 | 0.71 | 0.7920 | 0.82 | 0.902 | 0.05 | 0.0340 | 0.0360 | 0.06 |
| 0.1051 | 0.9999 | 36600 | 0.71 | 0.7920 | 0.82 | 0.902 | 0.05 | 0.0256 | 0.0425 | 0.06 |
| 0.1813 | 1.0000 | 54900 | 0.71 | 0.7920 | 0.82 | 0.902 | 0.05 | 0.0349 | 0.0488 | 0.06 |
| 0.3429 | 1.0000 | 721200 | 0.71 | 0.7920 | 0.82 | 0.902 | 0.05 | 0.0385 | 0.0439 | 0.06 |
| 0.3846 | 1.0000 | 901500 | 0.71 | 0.7920 | 0.82 | 0.902 | 0.05 | 0.0362 | 0.0481 | 0.06 |
| 0.4183 | 1.0000 | 1081800 | 0.71 | 0.7920 | 0.82 | 0.902 | 0.05 | 0.0333 | 0.0462 | 0.06 |
| 0.5340 | 1.0000 | 1262100 | 0.71 | 0.7920 | 0.82 | 0.902 | 0.05 | 0.0392 | 0.0463 | 0.06 |
| 0.5526 | 1.0000 | 1442400 | 0.71 | 0.7920 | 0.82 | 0.902 | 0.05 | 0.0342 | 0.0485 | 0.06 |
| 0.6427 | 1.0000 | 1622700 | 0.71 | 0.7920 | 0.82 | 0.902 | 0.05 | 0.0460 | 0.0466 | 0.06 |
| 0.6530 | 1.0000 | 1803000 | 0.71 | 0.7920 | 0.82 | 0.902 | 0.05 | 0.0398 | 0.0471 | 0.06 |
| 0.1317 | 0.9719 | 18300 | 0.71 | 0.8165 | 0.82 | 0.902 | 0.05 | 0.0340 | 0.0360 | 0.06 |
| 0.1843 | 0.9999 | 36600 | 0.71 | 0.8165 | 0.82 | 0.902 | 0.05 | 0.0256 | 0.0425 | 0.06 |
| 0.3207 | 1.0000 | 54900 | 0.71 | 0.8165 | 0.82 | 0.902 | 0.05 | 0.0349 | 0.0488 | 0.06 |
| 0.5476 | 1.0000 | 721200 | 0.71 | 0.8165 | 0.82 | 0.902 | 0.05 | 0.0385 | 0.0439 | 0.06 |
| 0.6159 | 1.0000 | 901500 | 0.71 | 0.8165 | 0.82 | 0.902 | 0.05 | 0.0362 | 0.0481 | 0.06 |
| 0.6699 | 1.0000 | 1081800 | 0.71 | 0.8165 | 0.82 | 0.902 | 0.05 | 0.0333 | 0.0462 | 0.06 |
| 0.7845 | 1.0000 | 1262100 | 0.71 | 0.8165 | 0.82 | 0.902 | 0.05 | 0.0392 | 0.0463 | 0.06 |
| 0.8114 | 1.0000 | 1442400 | 0.71 | 0.8165 | 0.82 | 0.902 | 0.05 | 0.0342 | 0.0485 | 0.06 |
| 0.8780 | 1.0000 | 1622700 | 0.71 | 0.8165 | 0.82 | 0.902 | 0.05 | 0.0460 | 0.0466 | 0.06 |
| 0.8918 | 1.0000 | 1803000 | 0.71 | 0.8165 | 0.82 | 0.902 | 0.05 | 0.0398 | 0.0471 | 0.06 |
| 0.2310 | 0.9719 | 18300 | 0.71 | 0.8520 | 0.82 | 0.902 | 0.05 | 0.0340 | 0.0360 | 0.06 |
| 0.3675 | 0.9999 | 36600 | 0.71 | 0.8520 | 0.82 | 0.902 | 0.05 | 0.0256 | 0.0425 | 0.06 |
| 0.5941 | 1.0000 | 54900 | 0.71 | 0.8520 | 0.82 | 0.902 | 0.05 | 0.0349 | 0.0488 | 0.06 |
| 0.8289 | 1.0000 | 721200 | 0.71 | 0.8520 | 0.82 | 0.902 | 0.05 | 0.0385 | 0.0439 | 0.06 |
| 0.8900 | 1.0000 | 901500 | 0.71 | 0.8520 | 0.82 | 0.902 | 0.05 | 0.0362 | 0.0481 | 0.06 |
| 0.9282 | 1.0000 | 1081800 | 0.71 | 0.8520 | 0.82 | 0.902 | 0.05 | 0.0333 | 0.0462 | 0.06 |
| 0.9713 | 1.0000 | 1262100 | 0.71 | 0.8520 | 0.82 | 0.902 | 0.05 | 0.0392 | 0.0463 | 0.06 |
| 0.9809 | 1.0000 | 1442400 | 0.71 | 0.8520 | 0.82 | 0.902 | 0.05 | 0.0342 | 0.0485 | 0.06 |
| 0.9925 | 1.0000 | 1622700 | 0.71 | 0.8520 | 0.82 | 0.902 | 0.05 | 0.0460 | 0.0466 | 0.06 |
| 0.9950 | 1.0000 | 1803000 | 0.71 | 0.8520 | 0.82 | 0.902 | 0.05 | 0.0398 | 0.0471 | 0.06 |
| 0.3829 | 0.9719 | 18300 | 0.71 | 0.8875 | 0.82 | 0.902 | 0.05 | 0.0340 | 0.0360 | 0.06 |
| 0.6187 | 0.9999 | 36600 | 0.71 | 0.8875 | 0.82 | 0.902 | 0.05 | 0.0256 | 0.0425 | 0.06 |
| 0.8519 | 1.0000 | 54900 | 0.71 | 0.8875 | 0.82 | 0.902 | 0.05 | 0.0349 | 0.0488 | 0.06 |
| 0.9715 | 1.0000 | 721200 | 0.71 | 0.8875 | 0.82 | 0.902 | 0.05 | 0.0385 | 0.0439 | 0.06 |
| 0.9892 | 1.0000 | 901500 | 0.71 | 0.8875 | 0.82 | 0.902 | 0.05 | 0.0362 | 0.0481 | 0.06 |
| 0.9958 | 1.0000 | 1081800 | 0.71 | 0.8875 | 0.82 | 0.902 | 0.05 | 0.0333 | 0.0462 | 0.06 |
| 0.9993 | 1.0000 | 1262100 | 0.71 | 0.8875 | 0.82 | 0.902 | 0.05 | 0.0392 | 0.0463 | 0.06 |
| 0.9997 | 1.0000 | 1442400 | 0.71 | 0.8875 | 0.82 | 0.902 | 0.05 | 0.0342 | 0.0485 | 0.06 |
| 1.0000 | 1.0000 | 1622700 | 0.71 | 0.8875 | 0.82 | 0.902 | 0.05 | 0.0460 | 0.0466 | 0.06 |
| 1.0000 | 1.0000 | 1803000 | 0.71 | 0.8875 | 0.82 | 0.902 | 0.05 | 0.0398 | 0.0471 | 0.06 |
| Sensitivity Power Specificity Power N |  | The power of the sensitivity test. It is based on the N1 subjects that have the disease. The power of the specificity test. It is based on the N2 subjects that do not have the disease. The total sample size of the study. It is equal to N1 + N2, where N1 = NP and N2 = N(1-P). |  |  |  |  |  |  |  |  |

## Tests for One-Sample Sensitivity and Specificity

| Se0 | The sensitivity under HO . The sensitivity is the proportion of diseased subjects that yield a positive test result. |
| :---: | :---: |
| Se1 | The sensitivity under H 1 . The sensitivity is the proportion of diseased subjects that yield a positive test result. |
| Sp0 | The specificity under HO . The specificity is the proportion of non-diseased subjects that yield a negative test result. |
| Sp1 | The specificity under H1. The specificity is the proportion of non-diseased subjects that yield a negative test result. |
| Target Alpha | The alpha (probability of rejecting HO when HO is true) that was desired. |
| Actual Sensitivity Alpha | The alpha that was actually achieved by the sensitivity test, calculated from the binomial distribution. |
| Actual Specificity Alpha | The alpha that was actually achieved by the specificity test, calculated from the binomial distribution. |
| P | The prevalence is the proportion of the population that actually has the condition (disease) of interest. |

## Summary Statements

A total sample size of 300 (which includes 18 subjects with the disease) achieves $9 \%$ power using a two-sided binomial test comparing sensitivities of 0.71 and 0.792 under the null and alternative hypotheses, respectively. This sample size also achieves $97 \%$ power using a two-sided binomial test comparing specificities of 0.82 and 0.902 under the null and alternative hypotheses, respectively. The target significance level is 0.05 for both tests. The actual significance level achieved by the sensitivity test is 0.034 and achieved by the specificity test is 0.036 . The prevalence of the disease is 0.06 .

## Dropout-Inflated Sample Size



## Dropout Summary Statements

Anticipating a $20 \%$ dropout rate, 375 subjects should be enrolled to obtain a final sample size of 300 subjects.

## References

Obuchowski, N.A., Zhou, X.H. 2002. 'Prospective studies of diagnostic test accuracy when disease prevalence is low,' Biostatistics, Volume 3, No. 4, pages 477-492.
Li, J., Fine, J. 2004. 'On sample size for sensitivity and specificity in prospective diagnostic accuracy studies,' Statistics in Medicine, Volume 23, pages 2537-2550.
Machin, D., Campbell, M.J., Tan, S.B., Tan, S.H. 2009. Sample Size Tables for Clinical Studies, Third Edition. Wiley-Blackwell, Chichester, United Kingdom.
Zhou, X.H., Obuchowski, N.A., McClish, D.K. 2002. Statistical Methods in Diagnostic Medicine. Wiley-Interscience, New York.

This report shows the values of each of the parameters, one scenario per row. Because of the discrete nature of the binomial distribution, the stated (Target) alpha is usually greater than the actual alpha. Hence, we also show the Actual Alpha along with the rejection region.

## Sens. Power

This is the power of the sensitivity test. It is calculated from the binomial distribution using the N1 observations of the diseased subjects.

## Spec. Power

This is the power of the specificity test. It is calculated from the binomial distribution using the N-N1 observations of the non-diseased subjects.

## N

This is the total sample size of the study, n . It is equal to $\mathrm{N} 1+\mathrm{N} 2$. The number of diseased subjects is $\mathrm{N} 1=$ $N P$. The number of non-diseased subjects is $\mathrm{N} 2=\mathrm{N}(1-\mathrm{P})$.

## Se0

This is the sensitivity under HO . The sensitivity is the proportion of diseased subjects that yield a positive test result.

## Se1

This is the sensitivity under H 1 . The difference between Se 1 and Se 0 is the difference that is detected by the study.

## Sp0

is the specificity under HO . The specificity is the proportion of non-diseased subjects that yield a negative test result.

## Sp1

This is the specificity under H 1 . The difference between Sp 1 and Sp 0 is the difference that is detected by the study.

## Target Alpha

This is the alpha (probability of rejecting HO when HO is true) that was desired. Because the binomial is a discrete distribution, the target value is seldom obtained. Rather, the actual value is lower than alpha.

## Actual Sens. Alpha

This is the alpha that was actually achieved by the sensitivity test, calculated from the binomial distribution.

## Actual Spec. Alpha

This is the alpha that was actually achieved by the specificity test, calculated from the binomial distribution.

## P

This is proportion of the population that actually has the condition (disease) of interest, called the prevalence.

## Sensitivity Plots Section

Sensitivity Plots



These plots show the relationship between sensitivity power, sample size, and Se 1 in this example.

## Specificity Plots Section

Specificity Plots



These plots show the relationship between specificity power, sample size, and Se1 in this example.

## Example 2 - Finding the Sample Size

Continuing with Example 1, suppose you want to study the impact of various choices for Se 1 on sample size. Using a significance level of 0.05 and $90 \%$ power, find the sample size when Se 1 is $79.20 \%, 81.65 \%, 85.20 \%$, and $88.75 \%$. Assume that a two-tailed binomial test will be used.

## 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 $\mathbf{2}$ 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 (Sensitivity) |
| H1 (Alternative Hypothesis) | H1: Se $\ddagger$ Se0, H1: Sp $\neq$ Sp0 |
| Power. | 0.90 |
| Alpha. | 0.05 |
| P (Prevalence) | 0.06 |
| Se0 (Null Sensitivity) | . 0.71 |
| Se1 (Alternative Sensitivity) | .. 0.7920 .8640 .8280 .8875 |
| Sp0 (Null Specificity).......... | . 0.82 |
| Sp1 (Alternative Specificity) | .. 0.902 |

## Output

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

## Numeric Results

| Solve For: | Sample Size (Sensitivity) |
| :--- | :--- |
| Sensitivity Hypotheses: | H0: $\mathrm{Se}=\mathrm{Se} 0$ vs. $\mathrm{H} 1: \mathrm{Se} \neq \mathrm{Se0}$ |
| Specificity Hypotheses: | H0: $\mathrm{Sp}=\mathrm{Sp0}$ vs. H1: $\mathrm{Sp} \neq \mathrm{Sp0}$ |
| Test Statistic: | Binomial Test |


| Power |  | Sample Size N1 and N | Sensitivity |  | Specificity |  | Alpha |  |  | Preva lence P |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  | $\begin{array}{r} \text { H0 } \\ \text { Se0 } \end{array}$ | $\begin{array}{r} \mathrm{H} 1 \\ \mathrm{Se} 1 \end{array}$ |  |  |  |  |  |  |
| Sens. | Spec. |  |  | Sp0 | Sp1 | Target | Actual | Actual |  |
| 0.9060 | 1 | 2994983 | 0.71 | 0.7920 | 0.82 | 0.902 | 0.05 | 0.0480 | 0.0480 | 0.06 |
| 0.9054 | 1 | 1732883 | 0.71 | 0.8165 | 0.82 | 0.902 | 0.05 | 0.0441 | 0.0482 | 0.06 |
| 0.9128 | 1 | 931550 | 0.71 | 0.8520 | 0.82 | 0.902 | 0.05 | 0.0399 | 0.0479 | 0.06 |
| 0.9153 | 1 | 55917 | 0.71 | 0.8875 | 0.82 | 0.902 | 0.05 | 0.0379 | 0.0413 | 0.06 |

This report shows the sample size needed to achieve 90\% power for each value of Se1.

## Example 3 - Validation using Li and Fine (2004)

Li and Fine (2004) page 2545 give the results of a power analysis indicate that if $\mathrm{Se} 0=0.5, \mathrm{Se} 1=0.9, \mathrm{P}=0.01$, alpha $=0.05$ (one-sided), and power $=0.90$, that $n=1100$.

## 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 $\mathbf{3}$ 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 .... | Sample Size (Sensitivity) |
| H1 (Alternative Hypothesis) | H1: Se > Se0, H1: Sp > Sp0 |
| Power.. | . 0.90 |
| Alpha........... | .. 0.05 |
| P (Prevalence). | 0.01 |
| Se0 (Null Sensitivity)... | . 0.5 |
| Se1 (Alternative Sensitivity) | 0.9 |
| Sp0 (Null Specificity)......... | . 0.5 |
| Sp1 (Alternative Specificity) | . 0.9 |

## Output

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

## Numeric Results



PASS has also obtained an $n$ of 1100 .
It is interesting to note that a sample size of 1050 will also result in the identical power. This is because the amount of interest is N1 which is 11. If $n$ is $1050, \mathrm{~N} 1=n \mathrm{P}=1050 \times 0.01=10.5 \sim 11$. Thus, all values of $n$ between 1050 and 1149 will result in the same power.
Also, the value of the power in the article is 0.904 while PASS has obtained 0.910 . This difference arises because 0.904 is the unconditional power while 0.910 is the conditional power.

