

# Power Calculations 1

```
signpow <- function(theta,n) # Power of the sign test
{
  L <- sqrt(n)*(.5-theta)/sqrt(theta*(1-theta))
  R <- 1.96/(2*sqrt(theta*(1-theta)))
  signpow <- 1 - pnorm(L+R) + pnorm(L-R)
  signpow
} # End of function signpow
```

```
> signpow(.5,100)
[1] 0.04999579
> signpow(.5,1000)
[1] 0.04999579
> signpow(.51,1000)
[1] 0.09687793
> signpow(.51,10000)
[1] 0.515994
> signpow(.51,20000)
[1] 0.8074681
```

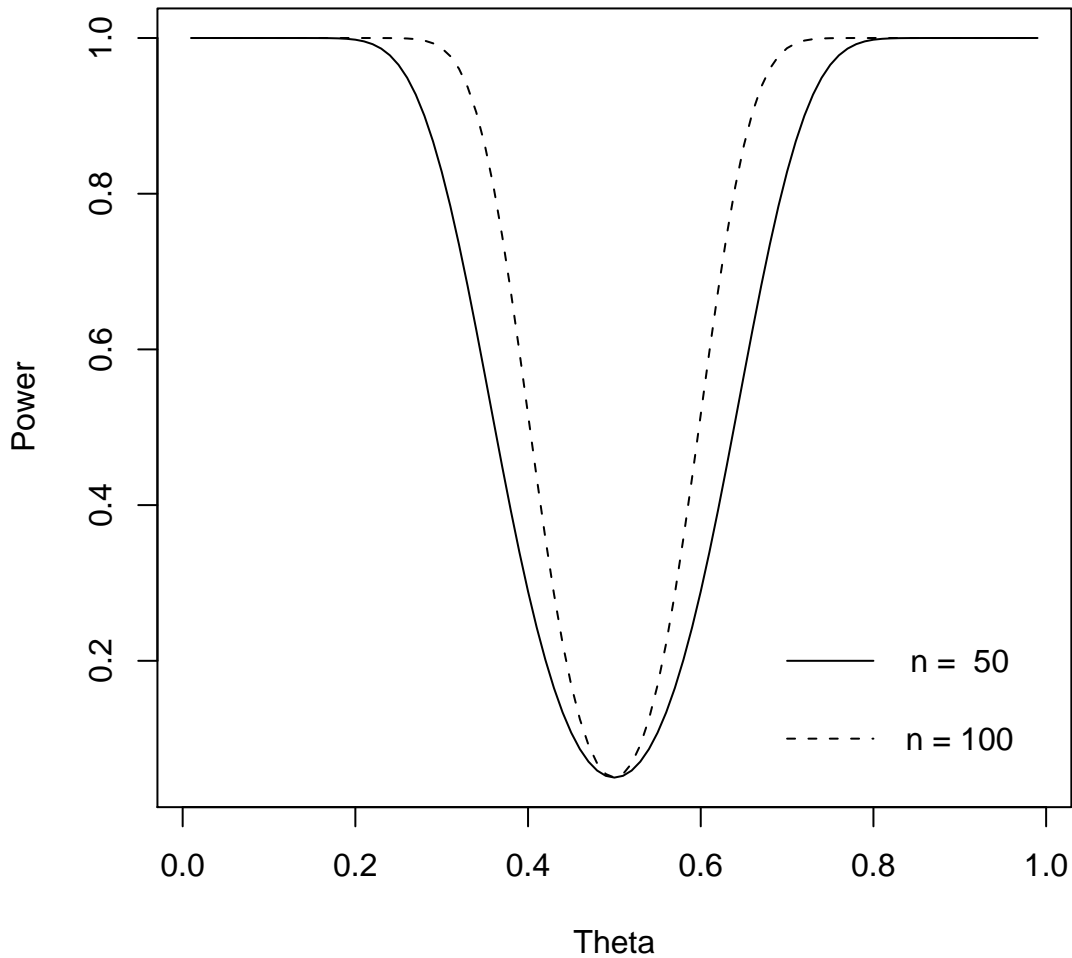
```
#####
# powplot.R - Plot Power of sign test as a function of true Theta #
#                                                                    #
# R --vanilla < powplot.R                                          #
#####
```

```
signpow <- function(theta,n) # Power of the sign test
{
  L <- sqrt(n)*(.5-theta)/sqrt(theta*(1-theta))
  R <- 1.96/(2*sqrt(theta*(1-theta)))
  signpow <- 1 - pnorm(L+R) + pnorm(L-R)
  signpow
} # End of function signpow
```

```
Theta <- seq(from=.01,to=.99,by=.01)
Power <- signpow(Theta,50)
p100 <- signpow(Theta,100)
system("rm powplot.pdf")
pdf("powplot.pdf")
```

```
plot(Theta,Power,type="l")
lines(Theta,p100,lty=2)
title("Power of the Sign Test")
x1 <- c(.70,.80) ; y1 <- c(.2,.2) ; lines(x1,y1,lty=1)
text(.9,.2,"n = 50")
x2 <- c(.70,.80) ; y2 <- c(.1,.1) ; lines(x2,y2,lty=2)
text(.9,.1,"n = 100")
```

# Power of the Sign Test



```
#####
# signpower.R - Find sample size for sign test to get required #
#               power. Do source("signpower.R") and then use #
#               the function size1 interactively. #
#####

signpow <- function(theta,n) # Power of the sign test
{
  L <- sqrt(n)*(.5-theta)/sqrt(theta*(1-theta))
  R <- 1.96/(2*sqrt(theta*(1-theta)))
  signpow <- 1 - pnorm(L+R) + pnorm(L-R)
  signpow
} # End of function signpow

size1 <- function(truet,needpow=0.80,nstart=1,nend=1000000)
{
  nn <- nstart ; pow <- 0
  while(pow<needpow)
  {
    pow <- signpow(truet,nn)
    nn <- nn+1
    if(nn>nend) stop("Too many iterations!")
  }
  cat("\n")
  cat("For true Theta of ",truet,", sign test requires a sample \n")
  cat("      size of ",nn-1," to have power of ",pow,"\n")
  cat("\n")
} # End function size1
```

---

```
> source("signpower.R")
> size1(.75)
```

```
For true Theta of 0.75 , sign test requires a sample
      size of 29 to have power of 0.8011995
```

```
> signpow(.75,28) # Just checking
[1] 0.7857723
>
> # Want power of 0.99 when theta = .51
> size1(0.51,0.99)
```

```
For true Theta of 0.51 , sign test requires a sample
      size of 45922 to have power of 0.99
```

```
>  
> # How about 60% chance of detecting gaze?  
> size1(0.60,0.99)
```

For true Theta of 0.6 , sign test requires a sample  
size of 450 to have power of 0.9900893

```
> # 75% chance of detecting gaze?  
> size1(0.75,0.99)
```

For true Theta of 0.75 , sign test requires a sample  
size of 64 to have power of 0.9907533

---

$$\Sigma = \begin{bmatrix} 1.42 & 0.42 & 0.42 & 0.00 & 0.00 \\ 0.42 & 1.42 & 0.42 & 0.00 & 0.00 \\ 0.42 & 0.42 & 1.42 & 0.00 & 0.00 \\ 0.00 & 0.00 & 0.00 & 1.18 & 0.18 \\ 0.00 & 0.00 & 0.00 & 0.18 & 1.18 \end{bmatrix}$$

Denoting the sample variance-covariance matrix by  $\mathbf{S}$ , and the  $j$ th sample variance (diagonal element of  $\mathbf{S}$ ) by  $s_j^2$ , the large-sample likelihood ratio test statistic for  $k$  variables may be written

$$G = n \left( \sum_{j=1}^k \log s_j^2 - \log |\mathbf{S}| \right),$$

where  $|\mathbf{S}|$  refers to the determinant of  $\mathbf{S}$ . Under the null hypothesis that  $\Sigma$  is diagonal,  $G$  has a chisquare distribution with  $\frac{1}{2}k(k-1)$  degrees of freedom (the number of unique off-diagonal elements). Here, the degrees of freedom equal 10.

$$\hat{P} \pm z_{1-\frac{\alpha}{2}} \sqrt{\frac{\hat{P}(1-\hat{P})}{M}}$$

This formula is implemented in the S function `merror` for “margin of error.”

```
merror <- function(phat,m,alpha) # (1-alpha)*100% merror for a proportion
{
  z <- qnorm(1-alpha/2)
  merror <- z * sqrt(phat*(1-phat)/m) # m is (Monte Carlo) sample size
  merror
}
```

Table 1: Monte Carlo Sample Size Required to Estimate Power with a Specified 99% Margin of Error

Margin of Error	Power Being Estimated					
	0.70	0.75	0.80	0.85	0.90	0.99
0.10	140	125	107	85	60	7
0.05	558	498	425	339	239	27
0.01	13,934	12,441	10,616	8,460	5,972	657
0.005	55,734	49,762	42,464	33,838	23,886	2,628
0.001	1,393,329	1,244,044	1,061,584	845,950	59,7141	65,686

```

# cvm.R
# Power for test of zero correlation for an entire matrix
# (Large Sample LR Test)
# Execute with source("cvm.R")
#
M <- 10000
sim <- numeric(M)
set.seed(32448)
n <- 50 ; v1 <- .42 ; v2 <- .18
s1 <- sqrt(v1) ; s2 <- sqrt(v2)
G <- function(datamat)
{
  nn <- dim(datamat)[1] ; kk <- dim(datamat)[2] ; df <- kk*(kk-1)/2
  G <- numeric(3)
  names(G) <- c("Chisq","df","P-value")
  S <- var(datamat)
  G[1] <- nn * ( sum(log(diag(S))) - sum(log(eigen(S)$values)) ) # $
  G[2] <- df
  G[3] <- 1 - pchisq(G[1],df)
  G
} # End function G
merror <- function(phat,m,alpha=0.01) # (1-alpha)*100% merror for a proportion
{
  z <- qnorm(1-alpha/2)
  merror <- z * sqrt(phat*(1-phat)/m) # m is (Monte Carlo) sample size
  merror
}

for(j in 1:M)
{

e1 <- rnorm(n,0,s1) ; e2 <- rnorm(n,0,s2)
x1 <- rnorm(n)+e1 ; x2 <- rnorm(n)+e1 ; x3 <- rnorm(n)+e1 ;
x4 <- rnorm(n)+e2 ; x5 <- rnorm(n)+e2
dat <- cbind(x1,x2,x3,x4,x5)
# print(G(dat))
print(j)
sim[j] <- G(dat)[3] < .05

}

poww <- length(sim[sim==1])/M
cat("Power = ", poww , "\n")
cat("Plus or Minus 99% Margin of error: ",merror(poww,M) , "\n")

```

Here's the output.

```

Power = 0.6987
Plus or Minus 99% Margin of error: 0.01181849

```

Here is the model.

$$\mathbf{Y} = \mathbf{X}\boldsymbol{\beta} + \boldsymbol{\epsilon},$$

where  $\mathbf{X}$  is an  $n \times r$  matrix of known constants,  $\boldsymbol{\beta}$  is a  $r \times 1$  vector of unknown constants, and  $\boldsymbol{\epsilon}$  is multivariate normal with mean zero and covariance matrix  $\sigma^2 \mathbf{I}_n$ , and  $\sigma^2 > 0$  is an unknown constant.

The null hypothesis is  $H_0 : \mathbf{C}\boldsymbol{\beta} = \mathbf{h}$  The  $F$  statistic for testing this null hypothesis is

$$F^* = \frac{(\mathbf{C}\hat{\boldsymbol{\beta}} - \mathbf{h})'(\mathbf{C}(\mathbf{X}'\mathbf{X})^{-1}\mathbf{C}')^{-1}(\mathbf{C}\hat{\boldsymbol{\beta}} - \mathbf{h})}{q \text{MSE}}$$

When  $H_0$  is false,  $F^*$  has a *noncentral F* distribution with parameters  $q$ ,  $n - r$  and  $\phi$ . A useful formula for  $\phi$  is

$$\phi = \frac{(\mathbf{C}\boldsymbol{\beta} - \mathbf{h})'(\mathbf{C}(\mathbf{X}'\mathbf{X})^{-1}\mathbf{C}')^{-1}(\mathbf{C}\boldsymbol{\beta} - \mathbf{h})}{\sigma^2}$$



## Comparing two means

```
> n <- seq(from=120,to=140,by=2) ; phi <- n/16 ; ddf <- n-2
> cbind(n,pf(qf(.95,1,ddf),1,ddf,phi,FALSE))
      n
[1,] 120 0.7752659
[2,] 122 0.7820745
[3,] 124 0.7887077
[4,] 126 0.7951683
[5,] 128 0.8014596
[6,] 130 0.8075844
[7,] 132 0.8135460
[8,] 134 0.8193475
[9,] 136 0.8249920
[10,] 138 0.8304825
[11,] 140 0.8358223
```

## Comparing $r$ means

```
fpow2 <- function(r,q,effsize,wantpow=0.80,alpha=0.05)
#####
# Power for the general multiple regression model, testing H0: C Beta = h #
#   r       is the number of beta parameters                            #
#   q       Number rows in the C matrix = numerator df                #
#   effsize is ncp/n, a squared distance between C Beta and h         #
#   wantpow is the desired power, default = 0.80                      #
#   alpha   is the significance level, default = 0.05                 #
#####
{
  pow <- 0 ; nn <- r+1 ; oneminus <- 1 - alpha
  while(pow < wantpow)
  {
    nn <- nn+1
    phi <- nn * effsize
    ddf <- nn-r
    pow <- 1 - pf(qf(oneminus,q,ddf),q,ddf,phi)
  }#End while
  fpow2 <- nn
  fpow2 # Returns needed n
}      # End of function fpow2

> 3 * var(c(0,.25,.5,.75)) / 4
[1] 0.078125

> source("fpow2.R")
> fpow2(r=4,q=3,effsize=0.078125)
[1] 144
```

```
> # Signal to noise ratio
> source("fpow2.R")
> fpow2(r=6,q=5,effsize=0.10)
[1] 134
```

Interaction example: Effect size is 0.01388889

```
> source("fpow2.R")
> fpow2(r=6,q=2,effsize=0.01388889,wantpow=0.80,alpha=0.05)
[1] 697
```