

Some Power Examples

Consider the usual linear model $\mathbf{Y} = \mathbf{X}\boldsymbol{\beta} + \boldsymbol{\varepsilon}$, where \mathbf{X} is $n \times r$ and $\boldsymbol{\beta}$ is $r \times 1$.

When $H_0: \mathbf{C}\boldsymbol{\beta} = \mathbf{h}$ is false (\mathbf{C} is $q \times r$ with linearly independent rows), the F statistic has a non-central F distribution with q and $n-r$ degrees of freedom, and non-centrality parameter

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

For a one-factor design, this may be written as

$$\phi = n \frac{\sum_{k=1}^r \frac{n_k}{n} (\mu_k - \mu_{\cdot})^2}{\sigma^2} = n \frac{\sum_{k=1}^r f_k (\mu_k - \mu_{\cdot})^2}{\sigma^2}$$

Suppose we have four treatments, and that the four population treatment means are equally spaced, one-quarter of a standard deviation apart. We'd like to be able to detect the differences among treatment means with probability 0.80, using the conventional significance level of $\alpha=0.05$. We'll use equal sample sizes.

Let

Non-centrality parameter = Sample size x Effect size

Without loss of generality, we'll let the four population treatment means be 0, 0.25, 0.50 and 0.75. Using R as a calculator, and remembering that the `var` function divides by the number of observations minus one, we'll calculate the effect size as

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

The program `fpow2.sas` uses `put` statements to write on the log file.

```

***** fpow2.sas *****/
options linesize = 79 pagesize = 100 noovp formdlim='-'; /* */
data fpower; /* Replace alpha, q, r, effsize and wantpow below */
  alpha = 0.05; /* Signif. level for testing H0: C Beta = h */
  q = 3; /* Numerator df = # rows in C matrix */
  r = 4; /* There are r beta parameters */
  effsize = 0.078125; /* Effect size is ncp/n */
  wantpow = .80; /* Find n to yield this power */
  power = 0; n = r+2; oneminus = 1-alpha; /* Initializing ... */
/*****/
do until (power >= wantpow);
  n = n+1 ;
  ncp = n * effsize;
  df2 = n-r;
  power = 1-probf(finv(oneminus,q,df2),q,df2,ncp);
end;
put ' ';
put ' *****';
put ' With ' r ' beta parameters, testing H0 of ' q ' linear';
put ' restrictions on the betas and an effect size of ' effsize ',';
put ' A sample size of ' n 'is needed to have probability ' ;
put ' ' wantpow ' of rejecting H0 at significance level alpha = ' alpha;
put ' *****';
put ' ';
put ' ';

```

The log file includes

```

*****
  With 4 beta parameters, testing H0 of 3 linear.
  restrictions on the betas and an effect size of 0.078125 ,
  A sample size of 144 is needed to have probability
  0.8 of rejecting H0 at significance level alpha = 0.05
*****

```

Now do the same thing with R

```

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

```

One may paste this function definition into the R window. Then,

```

> fpow2(r=4,q=3,effsize=0.078125)
[1] 144

```