Power Validation

```
library(knitr)
opts_chunk$set(tidy.opts=list(width.cutoff=60),tidy=TRUE)
```

Power Validation

##

[1] 108

Using the TrialSize package to calculate sample sizes base on examples from Chow SC, Shao J, Wang H. Sample Size Calculation in Clinical Research. New York: Marcel Dekker, 2008.

```
library(TrialSize)
library(TOSTER)
# Power for one-sample TOST based on raw scores
Example.3.1.4 <- OneSampleMean.Equivalence(alpha = 0.05, beta = 0.2,
    sigma = 0.1, delta = 0.05, margin = 0)
Example.3.1.4 # 35
## [1] 34.25539
powerTOSTone.raw(alpha = 0.05, statistical_power = 0.8, sd = 0.1,
   low_eqbound = -0.05, high_eqbound = 0.05) #35
## The required sample size to achieve 80 \% power with equivalence bounds of -0.05 and 0.05 is 35
## [1] 35
# Power for two-sample independent t-test based on raw scores
Example.3.2.4 <- TwoSampleMean.Equivalence(alpha = 0.1, beta = 0.1,
    sigma = 0.1, k = 1, delta = 0.05, margin = 0.01)
Example.3.2.4 #107 (but ceiling rounded 108)
## [1] 107.0481
# Margin 0.01 is identical to using bounds - margin, so 0.04
powerTOSTtwo.raw(alpha = 0.1, statistical_power = 0.9, low_eqbound = -0.04,
   high_eqbound = 0.04, sdpooled = 0.1) #108
## The required sample size to achieve 90 % power with equivalence bounds of -0.04 and 0.04 is 108 per
```

Table 1 Julious (2004) provides the sample sizes for *superiority* tests for a independent-samples t-test standardized effect sizes from 0.05 to 1.5, based equal sample sizes (in column 1). We can use the powerTOSTtwo function to calculate these values. Compared to an equivalence test, we can use the powerTOSTtwo function, when we double the alpha, and halve the Type 2 error. For alpha = 0.5 and a statistical power of 0.9, this means we will fill in an alpha of 0.1 and a statistical power of 0.95.

Table 1: bound N 0.05 8407 0.10 2103 0.15 935 0.20 527 0.25 338 0.30 235 0.35 173 0.40 133 0.45 105 0.50 86 0.55 71 0.60 60 0.65 51 0.70 44 0.75 39 0.80 34 0.85 31 0.90 27 0.95 25 1.00 23 1.05 21 1.10 19 1.15 17 1.20 16 1.25 15 1.30 14 1.35 13 1.40 12 1.45 12 1.50 11

```
require(TOSTER)
dlist \leftarrow seq(0.05, 1.5, 0.05)
samplesize <- numeric(length(dlist))</pre>
for (i in 1:length(dlist)) {
    samplesize[i] <- powerTOSTtwo(alpha = 0.1, statistical_power = 0.95,</pre>
        low_eqbound_d = -(dlist[i]), high_eqbound_d = (dlist[i]))
}
## The required sample size to achieve 95 % power with equivalence bounds of -0.05 and 0.05 is 8406 per
## The required sample size to achieve 95 % power with equivalence bounds of -0.1 and 0.1 is 2102 per g
##
## The required sample size to achieve 95 % power with equivalence bounds of -0.15 and 0.15 is 934 per
##
## The required sample size to achieve 95 % power with equivalence bounds of -0.2 and 0.2 is 526 per gr
##
## The required sample size to achieve 95 % power with equivalence bounds of -0.25 and 0.25 is 337 per
##
## The required sample size to achieve 95 % power with equivalence bounds of -0.3 and 0.3 is 234 per gr
##
## The required sample size to achieve 95 % power with equivalence bounds of -0.35 and 0.35 is 172 per
##
## The required sample size to achieve 95 \% power with equivalence bounds of -0.4 and 0.4 is 132 per gr
##
## The required sample size to achieve 95 % power with equivalence bounds of -0.45 and 0.45 is 104 per
##
## The required sample size to achieve 95 % power with equivalence bounds of -0.5 and 0.5 is 85 per gro
##
## The required sample size to achieve 95 % power with equivalence bounds of -0.55 and 0.55 is 70 per g
## The required sample size to achieve 95 % power with equivalence bounds of -0.6 and 0.6 is 59 per gro
##
## The required sample size to achieve 95 % power with equivalence bounds of -0.65 and 0.65 is 50 per g
##
## The required sample size to achieve 95 % power with equivalence bounds of -0.7 and 0.7 is 43 per gro
##
## The required sample size to achieve 95 % power with equivalence bounds of -0.75 and 0.75 is 38 per g
```

##

```
## The required sample size to achieve 95 % power with equivalence bounds of -0.8 and 0.8 is 33 per gro
##
## The required sample size to achieve 95 % power with equivalence bounds of -0.85 and 0.85 is 30 per g
##
## The required sample size to achieve 95 % power with equivalence bounds of -0.9 and 0.9 is 26 per gro
##
## The required sample size to achieve 95 % power with equivalence bounds of -0.95 and 0.95 is 24 per g
##
## The required sample size to achieve 95 % power with equivalence bounds of -1 and 1 is 22 per group,
##
  The required sample size to achieve 95 % power with equivalence bounds of -1.05 and 1.05 is 20 per g
##
## The required sample size to achieve 95 % power with equivalence bounds of -1.1 and 1.1 is 18 per gro
##
## The required sample size to achieve 95 % power with equivalence bounds of -1.15 and 1.15 is 16 per g
##
## The required sample size to achieve 95 % power with equivalence bounds of -1.2 and 1.2 is 15 per gro
##
## The required sample size to achieve 95 % power with equivalence bounds of -1.25 and 1.25 is 14 per g
##
## The required sample size to achieve 95 % power with equivalence bounds of -1.3 and 1.3 is 13 per gro
##
## The required sample size to achieve 95 % power with equivalence bounds of -1.35 and 1.35 is 12 per g
##
## The required sample size to achieve 95 % power with equivalence bounds of -1.4 and 1.4 is 11 per gro
##
## The required sample size to achieve 95 \% power with equivalence bounds of -1.45 and 1.45 is 10 per g
##
## The required sample size to achieve 95 % power with equivalence bounds of -1.5 and 1.5 is 10 per gro
##
samplesize
```

Calculations differ only one or 2 participants max, due to the approximation that is used.

234

22

172

20

526

26

934

30

33

10

337

24

[1] 8406 2102

38

10

[15]

[29]

We can also recreate Table 3, column 1, which includes sample sizes for 90% power and alpha = 0.025, assuming equal sample sizes and a try effect size of 0.

132

18

104

16

85

15

70

14

59

13

50

12

43

11

```
require(TOSTER)
dlist \leftarrow seq(0.05, 1.5, 0.05)
samplesize <- numeric(length(dlist))</pre>
for (i in 1:length(dlist)) {
    samplesize[i] <- powerTOSTtwo(alpha = 0.025, statistical_power = 0.9,</pre>
        low_eqbound_d = -(dlist[i]), high_eqbound_d = (dlist[i]))
}
## The required sample size to achieve 90 % power with equivalence bounds of -0.05 and 0.05 is 10396 pe
## The required sample size to achieve 90 % power with equivalence bounds of -0.1 and 0.1 is 2599 per g
##
## The required sample size to achieve 90 % power with equivalence bounds of -0.15 and 0.15 is 1156 per
##
## The required sample size to achieve 90 % power with equivalence bounds of -0.2 and 0.2 is 650 per gr
##
## The required sample size to achieve 90 % power with equivalence bounds of -0.25 and 0.25 is 416 per
##
## The required sample size to achieve 90 % power with equivalence bounds of -0.3 and 0.3 is 289 per gr
##
## The required sample size to achieve 90 % power with equivalence bounds of -0.35 and 0.35 is 213 per
##
## The required sample size to achieve 90 \% power with equivalence bounds of -0.4 and 0.4 is 163 per gr
##
## The required sample size to achieve 90 % power with equivalence bounds of -0.45 and 0.45 is 129 per
##
## The required sample size to achieve 90 % power with equivalence bounds of -0.5 and 0.5 is 104 per gr
##
## The required sample size to achieve 90 % power with equivalence bounds of -0.55 and 0.55 is 86 per g
## The required sample size to achieve 90 % power with equivalence bounds of -0.6 and 0.6 is 73 per gro
##
## The required sample size to achieve 90 % power with equivalence bounds of -0.65 and 0.65 is 62 per g
##
## The required sample size to achieve 90 % power with equivalence bounds of -0.7 and 0.7 is 54 per gro
##
## The required sample size to achieve 90 % power with equivalence bounds of -0.75 and 0.75 is 47 per g
```

##

```
##
## The required sample size to achieve 90 % power with equivalence bounds of -0.85 and 0.85 is 36 per g
##
## The required sample size to achieve 90 % power with equivalence bounds of -0.9 and 0.9 is 33 per gro
## The required sample size to achieve 90 % power with equivalence bounds of -0.95 and 0.95 is 29 per g
##
## The required sample size to achieve 90 % power with equivalence bounds of -1 and 1 is 26 per group,
##
## The required sample size to achieve 90 % power with equivalence bounds of -1.05 and 1.05 is 24 per g
##
## The required sample size to achieve 90 % power with equivalence bounds of -1.1 and 1.1 is 22 per gro
## The required sample size to achieve 90 % power with equivalence bounds of -1.15 and 1.15 is 20 per g
##
## The required sample size to achieve 90 % power with equivalence bounds of -1.2 and 1.2 is 19 per gro
##
## The required sample size to achieve 90 % power with equivalence bounds of -1.25 and 1.25 is 17 per g
##
## The required sample size to achieve 90 % power with equivalence bounds of -1.3 and 1.3 is 16 per gro
##
## The required sample size to achieve 90 % power with equivalence bounds of -1.35 and 1.35 is 15 per g
##
## The required sample size to achieve 90 % power with equivalence bounds of -1.4 and 1.4 is 14 per gro
##
## The required sample size to achieve 90 % power with equivalence bounds of -1.45 and 1.45 is 13 per g
##
## The required sample size to achieve 90 % power with equivalence bounds of -1.5 and 1.5 is 12 per gro
##
samplesize
   [1] 10396
               2599
                     1156
                            650
                                  416
                                        289
                                              213
                                                     163
                                                           129
                                                                 104
                                                                        86
```

The required sample size to achieve 90 % power with equivalence bounds of -0.8 and 0.8 is 41 per gro

We again see a difference in sample size of 1 at most due to the approximation using here.

36

14

33

13

29

12

26

24

22

41

15

[12]

[23]

73

20

62

19

54

17

47

16

Validation through simulation

We can calculate the sample size needed to achieve 80% power in a paired samples equivalence test:

```
require(TOSTER)
powerTOSTpaired(alpha = 0.05, statistical_power = 0.8, low_eqbound_dz = -0.1,
    high_eqbound_dz = 0.1) #Calculate n pairs

## The required sample size to achieve 80 % power with equivalence bounds of -0.1 and 0.1 is 857 pairs

## [1] 857

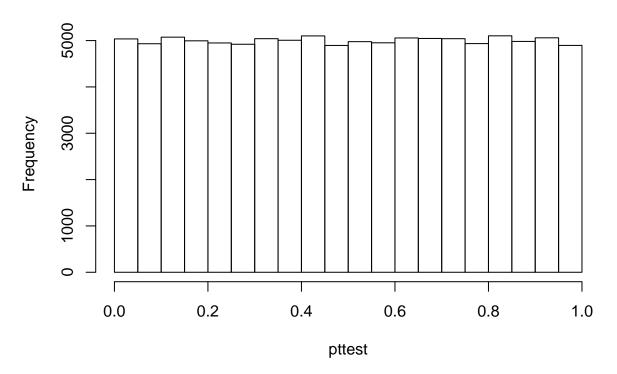
# Required N: 857 pairs
```

Subsequently, we can simulate paired samples TOST tests with given sample size, which gives 80% power.

```
library(mvtnorm) #to simulate correlated data
n <- 857 #set sample size simulation based on pwer
r <- 0 #set correlation
low_eqbound_dz = -0.1
high_eqbound_dz = 0.1
sd <- 4
alpha <- 0.05
nSims <- 1e+05 #number of simulated experiments
pttest <- numeric(nSims) #set up empty container for all simulated p-values
ptost <- numeric(nSims) #set up empty container for all simulated p-values
for (i in 1:nSims) {
           # for each simulated experiment
           a \leftarrow rmvnorm(n = n, mean = c(0, 0), sigma = matrix(c(sd, 0), sigma = m
                      r, r, sd), 2, 2))
           x \leftarrow a[, 1]
           y <- a[, 2]
           m1 \leftarrow mean(x)
           m2 \leftarrow mean(y)
           sd1 \leftarrow sd(x)
           sd2 \leftarrow sd(y)
           r12 \leftarrow cor(x, y)
           sdif \leftarrow sqrt(sd1^2 + sd2^2 - 2 * r12 * sd1 * sd2)
           se <- sdif/sqrt(n)</pre>
           low_eqbound <- low_eqbound_dz * sdif</pre>
           high_eqbound <- high_eqbound_dz * sdif
           t \leftarrow (m1 - m2)/se
           degree_f <- n - 1
           pttest[i] <- 2 * pt(abs(t), degree_f, lower = FALSE)</pre>
           t1 \leftarrow ((m1 - m2) + (low_eqbound))/se
           p1 <- 1 - pt(t1, degree_f, lower = FALSE)</pre>
           t2 \leftarrow ((m1 - m2) + (high\_eqbound))/se
           p2 <- pt(t2, degree_f, lower = FALSE)</pre>
           ttost \leftarrow ifelse(abs(t1) < abs(t2), t1, t2)
           LL90 \leftarrow ((m1 - m2) - qt(1 - alpha, degree_f) * se)
           UL90 <- ((m1 - m2) + qt(1 - alpha, degree f) * se)
           ptost[i] <- max(p1, p2)</pre>
}
```

```
# P-value distribution normal t-test (uniform because null is
# true)
hist(pttest, breaks = 20, xlim = c(0, 1))
```

Histogram of pttest

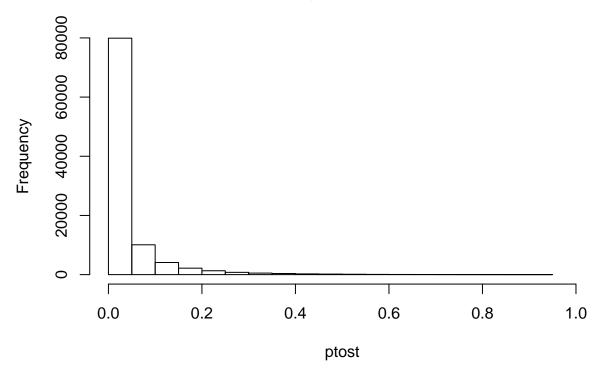


```
sum(pttest < 0.05) #Type 1 error rate for NHST

## [1] 5036

# P-value distribution TOST
hist(ptost, breaks = 20, xlim = c(0, 1))</pre>
```

Histogram of ptost



sum(ptost < 0.05) #Power for TOST</pre>

[1] 79913