Exact Binary Sequential Designs

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1 Example

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We start with a simple example of use of the package. Suppose one wanted to test

 $H_0: \quad \theta < 0.4$ vs. $H_1: \quad \theta \ge 0.4$

For example if historical treatments typically had less than 40% success rate, and you wanted to test that a new treatment had better rates. Suppose you wanted to do a study with no more than 50 subjects, that had a power of at least 80% to reject the null hypothesis at the one-sided 0.025 level if the new treatment had a success rate of 80%. We pick a large effect so that the design will be small. You also want to be able to detect if the new treatment has less than 40% success rate with a one-sided 0.05 level. To start we load the binseqtest package, create a binary sequential design, and give a table of with the design:

```
> library(binseqtest)
> b0 < -designOBF power (100, theta0 = .4, theta1 = .8, power = 0.8, tsalpha = c(0.05, 0.025))
> stopTable(b0)
H0: theta0= 0.4
confidence level= 0.925
alternative= two.sided : lowerError= 0.05 , upperError= 0.025
Boundaries binding= both
  N S estimate lower upper pLower pUpper pTwoSided
  6 6
         0.945 0.607 1.000 1.00000 0.00410
                                               0.00819
  7 6
         0.831 0.479 0.996 0.99590 0.01884
                                               0.03768
  9 7
         0.739 0.436 0.963 0.98116 0.03123
                                               0.06246
 11 0
         0.031 0.000 0.285 0.00363 1.00000
                                               0.00726
 11 1
         0.105 0.005 0.413 0.03023 0.99637
                                               0.06047
 11 2
         0.192 0.033 0.518 0.11892 0.96977
                                               0.23783
 11 3
         0.280 0.079 0.610 0.29628 0.88108
                                               0.59257
 11 4
         0.368 0.135 0.692 0.53277 0.70372
                                               1.00000
         0.456 0.200 0.766 0.75350 0.46723
 11 5
                                               0.93445
         0.543 0.271 0.832 0.89842 0.24650
 11 6
                                               0.49300
 11 7
         0.625 0.348 0.886 0.95787 0.10158
                                               0.20316
```

We can plot the design using the command plot(b0) giving Figure 1.

0.684 0.412 0.923 0.96877 0.04213

0.08426

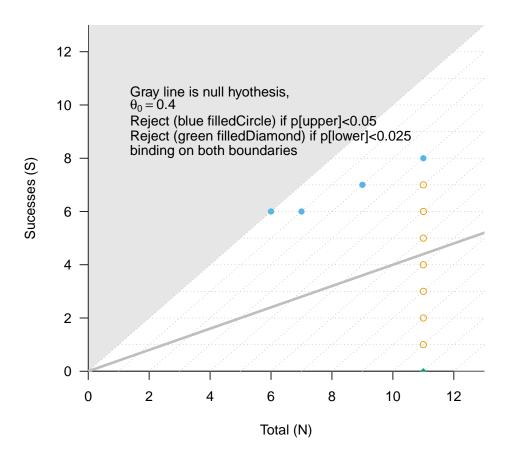


Figure 1: Binary sequential boundary for testing $H_0: \theta < 0.4$. Filled blue points are rejections and open orange points are failure to reject.

2 Simon's Two-Stage Designs

We can calculate statistics on Simon's Two-stage designs (Simon, 1989). The designs were developed for screening new treatments where the response is positive (e.g., treatment is working). The action at the end of the study is either (1) do not pursue further studies of the new treatment since it does not appear to be sufficiently promising, or (2) continue with further studies of the new treatment. For such as design, at the first stage, we sample n_1 observations and stop the study and do not pursue the new treatment further if the number of successes is less than or equal to r_1 . If the number of successes at the first stage is greater than r_1 , then we continue sampling until there are n total observations, and at the end we only conclude that the new treatment is sufficiently promising if the number of successes is greater than r. We can think of these types of designs as testing:

```
\frac{\text{Hypothesis}}{H_0: \theta \le \theta_0} \quad \frac{\text{Action}}{\text{No further study}}H_1: \theta > \theta_0 \quad \text{Further Study}
```

We set the type I error of the hypothesis at α and set the power under θ_1 at $1 - \beta$.

Suppose that $\theta_0 = .2$ and $\theta_1 = .4$ and suppose that $\alpha = .05$ and $\beta = .2$, then Simon's optimal design (see Table 1 of Simon, 1989) has $r_1 = 3$, $n_1 = 13$, r = 12 and n = 43. Describing that design using the binseqtest package is as follows:

```
> B<-designAb(Nk=c(13,43),a=c(3),theta0=.2,conf.level=.95,alternative="greater")
```

The alternative="greater" means that we only continue with further studies if we can show that $\theta > \theta_0$ using a nominal significance level of α , and the associated confidence intervals are at the $1-\alpha$ level.

We can plot this boundary using plotBoundEst(B) to get Figure 2.

We can double-check the power,

```
> powerBsb(B,theta=.4)
```

```
[1] 0.8002144
```

Suppose the study ends at the second stage with S = 20 and N = 43. Then the default statistics are given using the following command:

```
> stopTable(B,S=20,N=43)
H0: theta0= 0.2
confidence level= 0.95
alternative= greater : lowerError= 0.05 , upperError= 0
Boundaries binding= both
   N S estimate lower upper pLower pUpper pTwoSided
43 20   0.467 0.335   1 0.99998 0.00008   0.00016
```

Note the default is to use the one-sided p-values and confidence intervals since that is how the study was designed. We can change the statistics to calculate the two-sided confidence intervals and p-values using

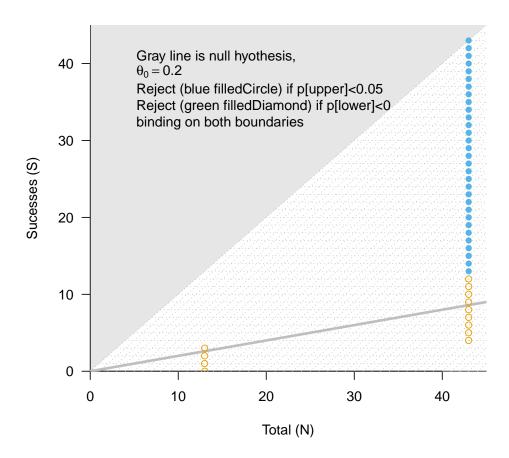


Figure 2: Binary sequential boundary for testing $H_0: \theta \leq 0.2$ using Simon's two-stage design with $r_1=3$ $n_1=13$, r=12 and n=43.

```
> B2<-modify(B,conf.level=.95,alternative="two.sided")
> stopTable(B2,S=20,N=43)

H0: theta0= 0.2
confidence level= 0.95
alternative= two.sided : lowerError= 0.025 , upperError= 0.025
Boundaries binding= both
   N S estimate lower upper pLower pUpper pTwoSided
43 20    0.467 0.313 0.627 0.99998 0.00008    0.00016
```

If we miss stopping the first stage at $n_1 = 13$ and instead stop at $n_1 = 19$ (i.e., we miss stopping at 13 through 18) we can modify the boundary as follows

```
> B3<-modify(B,missN=13:18)
> stopTable(B3,S=20,N=43)

H0: theta0= 0.2
confidence level= 0.95
alternative= greater : lowerError= 0.05 , upperError= 0
Boundaries binding= both
   N S estimate lower upper pLower pUpper pTwoSided
43 20   0.465 0.334   1 0.99998 0.00008   0.00016
```

and the new design is plotted in Figure 3.

3 Plotting Binary Sequential Boundaries

There are several options for plotting a binary sequential boundary using the bplottype option. First we create a boundary with mostly default settings

```
> b<-designOBF(50,k=Inf,theta0=.3)
```

By default the boundary comes with 95% confidence intervals, with binding stopping on both boundaries. In Figure 4 we plot four types of plots to represent this boundary. In the plots, the sky-blue points represent rejecting the null hypothesis and concluding that $\theta > 0.3$, while the green-blue points represent rejecting and concluding that $\theta < 0.3$.

References

Simon, R (1989). Optimal Two-Stage Designs for Phase II Clinical Trials. Controlled Clinical Trials. 10:1-10.

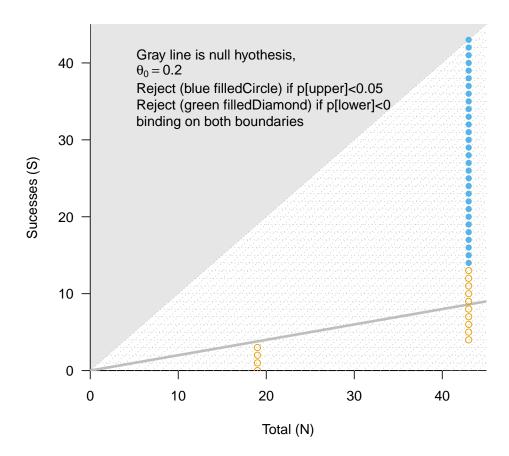


Figure 3: Binary sequential boundary for testing H_0 : $\theta \le 0.2$ using Simon's two-stage design originally with $r_1 = 3$ $n_1 = 13$, r = 12 and n = 43, but then modified to stop at the first stage at $n_1 = 19$.

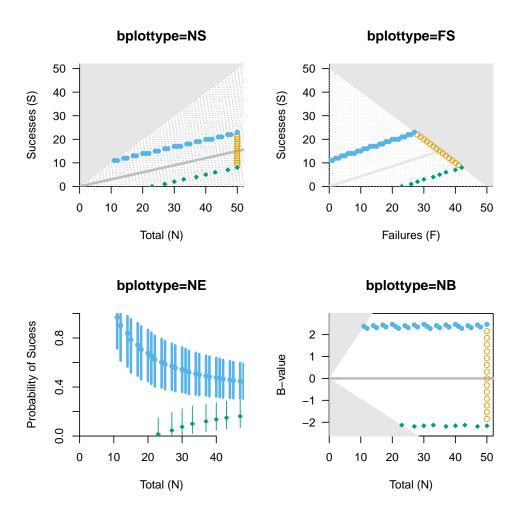


Figure 4: Binary sequential boundary for testing H_0 : $\theta = 0.3$ using an O-Brien-Fleming type boundary. The 'NE' type gives estimates and confidence intervals associated with the points.