# Analysis of Critical Values of Response Rate in Single-Arm Two-Stage Phase II Clinical Trials BIOSTAT 907 Term Project (Fall 2021)

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#### Overview

#### Single-arm phase II clinical trial

- A historical therapy vs. An experimental therapy
- Phase II: requires small sample size, test for safety and efficacy
- ►  $H_0: p = p_0$  vs.  $H_1: p = p_1$ , test the response rate (RR) of different therapies  $(p_1 = p_0 + \delta)$ , a clinically meaningful increasing)
  - ▶ Under  $H_0$ :  $X \sim Bin(n, p_0)$
  - Single-stage design: Given  $(\alpha^*, 1 \beta^*, p_0, p_1)$ , find a pair (n, a) to specify the sample size and rejection boundary
  - Start from  $n \ge n_0$ , find the smallest integer a such that based on level  $\alpha^*$ , we reject  $H_0$  if X > a and otherwise fail to reject. Then if

$$1 - \beta = P_1(X > a) \ge 1 - \beta^*$$

the pair (n, a) is the best design. If not, continue to the next n

► It is to *minimize the sample size* 

### Large Sample Approximation

Note that in a single-arm single-stage trial, given an (large) n, we can find the approximate value of a using large sample approximation

Denote  $\bar{X} = X/n$ 

Under  $H_0$ :

$$\sqrt{n}(\bar{X}-p_0) \xrightarrow{d} N(0, p_0(1-p_0))$$

Under  $H_1$ :

$$\sqrt{n}(\bar{X} - p_1) \xrightarrow{d} N(0, p_1(1 - p_1))$$

## Large Sample Approximation

So,  $P_0(X > a) \le \alpha^*$  is approximately equivalent to

$$\frac{\frac{a}{n} - p_0}{\sqrt{\frac{p_0(1 - p_0)}{n}}} \ge z_{1 - \alpha^*}$$

and  $P_1(X > a) \ge 1 - \beta^*$  is approximately equivalent to

$$\frac{\frac{a}{n} - p_1}{\sqrt{\frac{p_1(1 - p_1)}{n}}} \le z_{\beta^*}$$

where  $z_{\gamma}$  is the  $\gamma$ -quantile of N(0,1) distribution

## Large Sample Approximation

As such,

$$\sqrt{\frac{p_0(1-p_0)}{n}}z_{1-\alpha^*} + p_0 \leq \frac{a}{n} \leq \sqrt{\frac{p_1(1-p_1)}{n}}z_{\beta^*} + p_1$$

Then we approximate a/n by the average of the LHS and RHS

$$\frac{a}{n} \approx \frac{1}{2} \left( \underbrace{\sqrt{\frac{p_0(1-p_0)}{n}}}_{\approx 0 \text{ as } n \text{ becomes large}} z_{\beta^*} + \underbrace{p_0 + p_1}_{n} \right)$$

Therefore

$$a \approx \frac{n}{2}(p_0 + p_1)$$

(Note *a* is related to *n*)

### Two-Stage Design

Now, consider the single-arm two-stage design for phase II trials

Why two-stage?

- ► Economical and ethical consideration
- Stop early if no efficacy tested
- Data collected can be more informative than single-stage if we can proceed to phase III

## Best Two-Stage Design

 $H_0: p = p_0$  vs.  $H_1: p = p_1$ .  $n = n_1 + n_2$  for two stages respectively

- ►  $X_1 \sim Bin(n_1, p_i)$  is independent to  $X_2 \sim Bin(n_2, p_i)$  under  $H_i$ , i = 0, 1
- $\alpha = P_0(X_1 > a_1, X_1 + X_2 > a)$
- $1 \beta = P_1(X_1 > a_1, X_1 + X_2 > a)$
- ▶ Probability of early termination under  $H_0$ : PET =  $P_0(X_1 \le a_1)$
- Expected sample size under  $H_0$ : EN =  $n_1 \times PET + n \times (1 PET)$

Among designs with  $\alpha \leq \alpha^*$  and  $1 - \beta \geq 1 - \beta^*$ 

- ▶ **Optimal** design:  $(a_1, n_1, a, n)$  minimizes EN
- ► **Minimax** design:  $(a_1, n_1, a, n)$  minmizes  $n = n_1 + n_2$

This is the case of without considering *superiority (upper) stopping* of stage 1 (or, only *futility (lower) stopping*)

## Best Two-Stage Design

When we consider *both futility and superiority stopping*, there will be one more boundary value  $(b_1)$  to let us stop the trial immediately after stage 1:

$$1 - \alpha = P_0(X_1 \le a_1) + P_0(a_1 < X_1 < b_1, X_1 + X_2 \le a)$$

$$\beta = P_1(X_1 \le a_1) + P_1(a_1 < X_1 < b_1, X_1 + X_2 \le a)$$

- ► Probability of early termination under  $H_i$ : PET $(p_i)$  =  $P_i(X_1 \le a_1) + P_i(X_1 \ge b_1), i = 0, 1$
- Expected sample size:  $EN(p_i) = n_1 \times PET(p_i) + n \times (1 - PET(p_i)), i = 0, 1$  $EN(p_i) + EN(p_i)$

$$EN = \frac{EN(p_0) + EN(p_1)}{2}$$

Among designs with  $\alpha \leq \alpha^*$  and  $1 - \beta \geq 1 - \beta^*$ 

- ▶ **Optimal** design:  $(a_1, b_1, n_1, a, n)$  minimizes EN
- Minimax design:  $(a_1, b_1, n_1, a, n)$  minmizes  $n = n_1 + n_2$

## What are we looking for?

Recall that  $a \approx \frac{n}{2}(p_0 + p_1)$  in single-stage trials

Is there any similar pattern(s) in the critical values of hypothesis test in two-stage trials?

- With only futility stopping, is there any pattern(s) in  $a_1/n_1$  and a/n over different choices of  $(\alpha, 1 \beta, p_0, p_1)$ ?
- ▶ With both futility and superiority stopping considered, is there any pattern(s) in  $a_1/n_1$ ,  $b_1/n_1$  and a/n over different choices of  $(\alpha, 1 \beta, p_0, p_1)$ ?
- Our hunch
  - $ightharpoonup a_1/n_1 \approx p_0$
  - $b_1/n_1 \approx p_1$
  - ►  $a/n \approx (p_0 + p_1)/2$
- Anything else?

## Experiment

- We did some numerical experiments to find a sense
- Consider the following choices of the parameters
  - ▶  $p_0 \in [0.05, 0.7]$  with increment of 0.005
  - $p_1 = p_0 + 0.2$  and  $p_0 + 0.25$
  - level of test =  $\alpha^* = 0.05, 0.1$
  - power =  $1 \beta^* = 0.8, 0.85, 0.9$

## Experiment

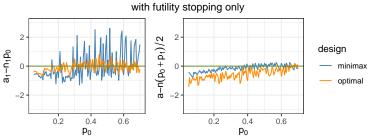
- ► For only futility stopping, plot the figures of
  - $a_1 n_1 p_0$
  - $a n(p_1 + p_0)/2$
  - ▶ and corresponding ratio figures, such as  $a_1/n_1 p_0$
- ► For considering both two stoppings, plot the figures of
  - $a_1 n_1 p_0$
  - $b_1 n_1 p_1$
  - $a n(p_1 + p_0)/2$
  - and corresponding ratio figures
- ▶ Investigate bias and variability over  $p_0$

Because of the similar results we get, we just present the case of frequency trends under

- $(\alpha^*, 1 \beta^*) = (0.1, 0.8), (0.05, 0.8), (0.1, 0.9), (0.05, 0.9)$
- $p_1 = p_0 + 0.2$ 
  - futility stopping only
  - both futility and superiority stopping

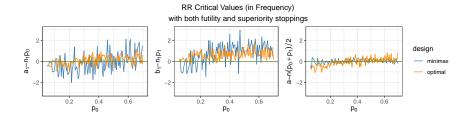
$$(\alpha^*, 1 - \beta^*) = (0.1, 0.8), p_1 - p_0 = 0.2$$





Stage	Design	Median	Mean	IQR	SD
Stage 1	Optimal	-0.080	-0.067	0.473	0.310
	Minimax	-0.420	-0.051	0.807	0.931
Stage 2	Optimal	-0.555	-0.544	0.530	0.350
	Minimax	-0.125	-0.143	0.317	0.225

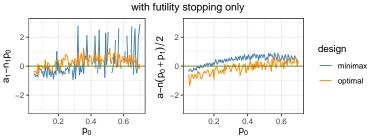
$$(\alpha^*, 1 - \beta^*) = (0.1, 0.8), p_1 - p_0 = 0.2$$



Stage	Design	Median	Mean	IQR	SD
Stage 1 Lower	Optimal	0.135	0.102	0.680	0.438
	Minimax	-0.160	-0.117	1.068	0.877
Stage 1 Upper	Optimal	0.550	0.593	0.600	0.409
	Minimax	0.945	0.756	1.215	0.946
Stage 2 Lower	Optimal	-0.050	-0.064	0.528	0.386
	Minimax	0.020	0.012	0.305	0.202

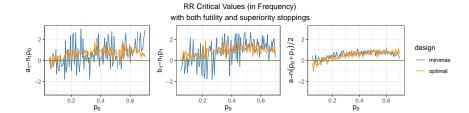
$$(\alpha^*, 1 - \beta^*) = (0.05, 0.8), p_1 - p_0 = 0.2$$





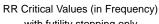
Stage	Design	Median	Mean	IQR	SD
Stage 1	Optimal	0.345	0.315	0.530	0.375
	Minimax	-0.040	0.603	1.437	1.427
Stage 2	Optimal	-0.100	-0.124	0.545	0.393
	Minimax	0.440	0.404	0.437	0.316

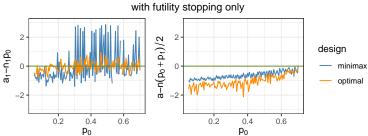
$$(\alpha^*, 1 - \beta^*) = (0.05, 0.8), p_1 - p_0 = 0.2$$



Stage	Design	Median	Mean	IQR	SD
Stage 1 Lower	Optimal	0.650	0.628	0.565	0.446
	Minimax	0.500	0.664	1.467	1.188
Stage 1 Upper	Optimal	0.900	0.882	0.557	0.424
	Minimax	1.000	0.844	1.318	1.082
Stage 2 Lower	Optimal	0.500	0.440	0.527	0.422
	Minimax	0.600	0.545	0.367	0.294

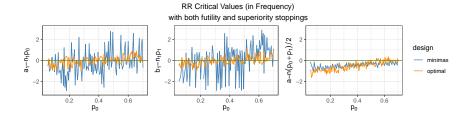
$$(\alpha^*, 1 - \beta^*) = (0.1, 0.9), p_1 - p_0 = 0.2$$





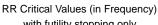
Stage	Design	Median	Mean	IQR	SD
Stage 1	Optimal	-0.040	-0.037	0.440	0.343
	Minimax	-0.450	0.078	0.930	1.254
Stage 2	Optimal	-1.120	-1.065	0.525	0.398
	Minimax	-0.690	-0.645	0.398	0.253

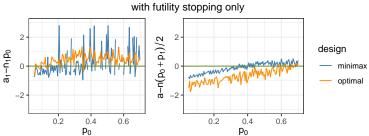
$$(\alpha^*, 1 - \beta^*) = (0.1, 0.9), p_1 - p_0 = 0.2$$



Stage	Design	Median	Mean	IQR	SD
Stage 1 Lower	Optimal	0.055	-0.012	0.560	0.401
	Minimax	-0.310	-0.221	1.215	1.134
Stage 1 Upper	Optimal	-0.075	-0.044	0.525	0.393
	Minimax	0.140	-0.083	1.700	1.394
Stage 2 Lower	Optimal	-0.550	-0.546	0.652	0.429
	Minimax	-0.465	-0.469	0.333	0.235

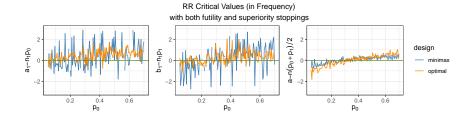
$$(\alpha^*, 1 - \beta^*) = (0.05, 0.9), p_1 - p_0 = 0.2$$





Stage	Design	Median	Mean	IQR	SD
Stage 1	Optimal	0.445	0.450	0.465	0.374
	Minimax	0.060	0.745	1.757	1.611
Stage 2	Optimal	-0.710	-0.677	0.697	0.451
	Minimax	-0.060	-0.107	0.535	0.341

$$(\alpha^*, 1 - \beta^*) = (0.05, 0.9), p_1 - p_0 = 0.2$$



Stage	Design	Median	Mean	IQR	SD
Stage 1 Lower	Optimal	0.480	0.478	0.658	0.499
	Minimax	0.300	0.633	1.730	1.378
Stage 1 Upper	Optimal	0.250	0.287	0.688	0.553
	Minimax	0.230	-0.023	1.630	1.435
Stage 2 Lower	Optimal	0.100	-0.001	0.728	0.519
	Minimax	0.090	0.053	0.395	0.311

### **Summary**

Motivated by the relationship between (a, n) and  $(p_0, p_1)$  in single-stage phase II trials, we investigated the trend of several critical values of testing RR in two-stage phase II trials with two kinds of stopping assumption

- ► Indeed, for both optimal and minimax design
  - $a_1 \approx n_1 p_0$
  - $b_1 \approx n_1 p_1$
  - ►  $a \approx n(p_0 + p_1)/2$
- In many cases, there is no too much difference, or no special pattern in biases (of  $a_1 n_1p_0$ ,  $b_1 n_1p_1$  and  $a n(p_0 + p_1)/2$ ) between optimal and minimax design
- ▶ Optimal design has smaller **variations** (by variance, IQR) in  $a_1 n_1p_0$  (and  $b_1 n_1p_1$ ), where minimax design has smaller **variations** in  $a n(p_0 + p_1)/2$

### Acknowledgement

The credit of the idea of this term project should be given to the instructor of this course, Dr. Sin-Ho Jung, and thanks for his two CTD softwares to help me to confirm the programming results.

#### References

► S-H Jung (2013). Randomized Phase II Cancer Clinical Trials (1st ed.). *Chapman and Hall/CRC*. (Chapter 2 and 3)

# Thank you!

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