



Sample Size Calculation

a closer look at 'hybrid' methods

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2020/11/04 (updated: 2020-11-04)



How many?

42 will do!



Preliminaries

- simple one-arm Wald test with test statistic

$$Z_n := \frac{\hat{\theta}_n - \theta_0}{\text{SE}[\hat{\theta}_n]}$$

- w.l.o.g. $\theta_0 = 0$, $\mathcal{H}_0 = \theta \leq 0$, and effect size standardized
- assume that $\text{SE}[\hat{\theta}_n]$ is deterministic function of n
(think: σ known in the case of θ being the mean of a normal distribution)
- maximal one-sided type one error rate is externally given as $\alpha = 0.025$
(i.e. critical value is 1.96, reject $\mathcal{H}_0 \Leftrightarrow Z_n > 1.96$)
- everything generalizes to
 - multi-arm comparisons
 - asymptotically normal test statistics (hazard ratios!)
 - etc.

Standard Approach

- probability to reject [1], $\Pr_{\theta}[Z_n > 1.96]$, is a function of θ and n
- pick point alternative $\theta_{alt} > 0$ and acceptable type-two error rate β
- chose smallest n such that $\Pr_{\theta_{alt}}[Z_n > 1.96] = 1 - \beta$
- n is monotone function of θ_{alt} : if $\theta_{alt} \nearrow$ then $n \searrow$
- in the following: $\beta = 0.2$, i.e. we aim for 80% power
- for $n = 42$ this works if $\theta_{alt} \approx 0.43$

wait, where does the point alternative come from?

[1] determining the sample size via power arguments is common but not the only criterion, could also use width of confidence interval, utility, ...

Choice of Point Alternative

MCID

- Minimal Clinically Important Difference (MCID)
- externally given
- $\theta \geq \theta_{\text{MCID}} : \Pr_{\theta}[Z_n > 1.96] \geq 1 - \beta$
i.e. all relevant effects are detected with minimum power (due to monotonicity)
- expensive: $n \rightarrow \infty$ as $\theta_{\text{MCID}} \rightarrow 0$

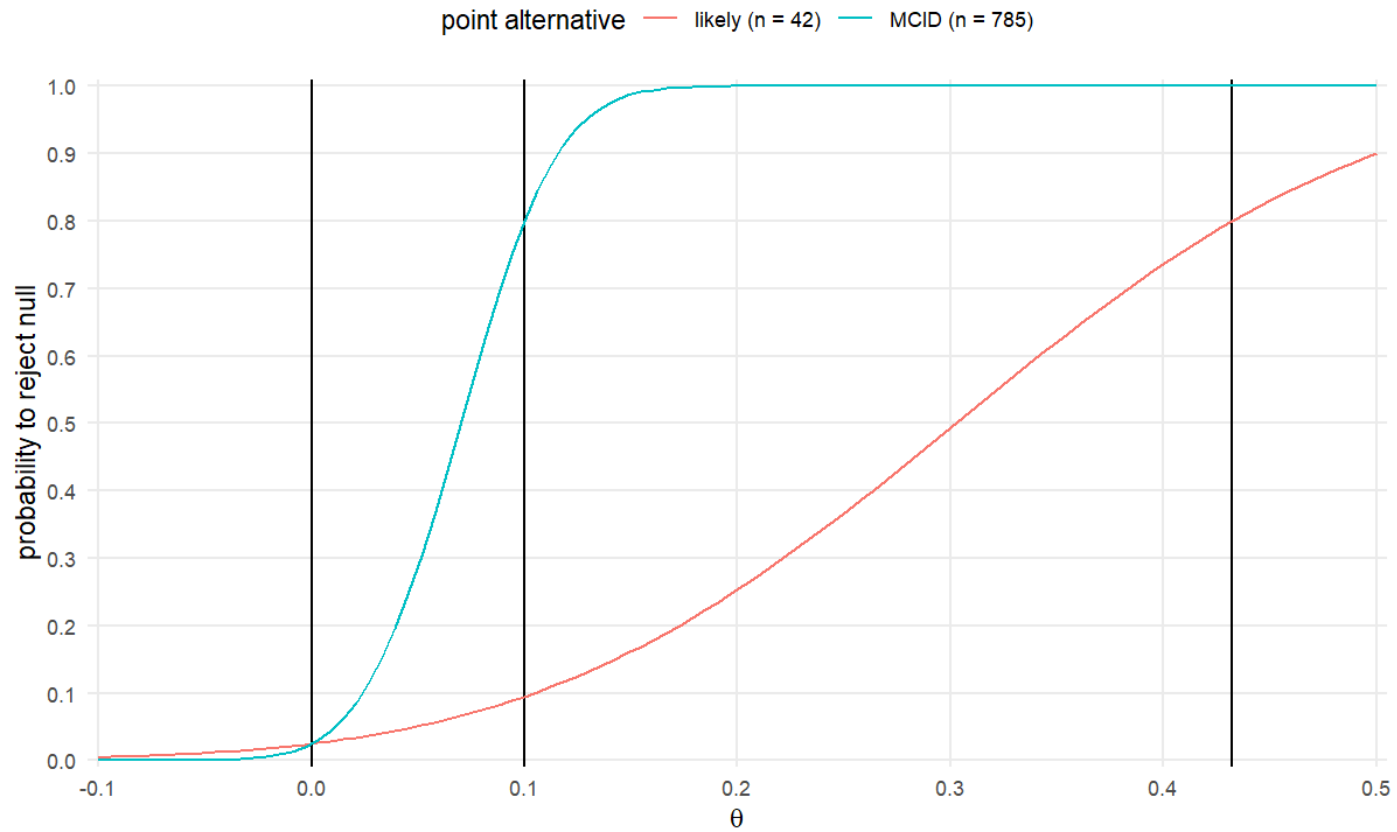
'likely' alternative

- sometimes it is 'very likely' that

$$\theta \gg \theta_{\text{MCID}}$$

- deriving n from θ_{MCID} is ineffective
- \rightsquigarrow consider all $\theta \geq \theta_{\text{MCID}}$ and their 'relative likelihood'
- what is 'relative likelihood'?

'Power Plays'



'Hybrid' Approaches

- 'hybrid' between **Bayesian** and **frequentist** methods¹
- model unknown parameter as random variable Θ
- uses prior distribution to quantify planning uncertainty about Θ

$$\Pr[\Theta \leq x] = \int^x \underbrace{\varphi(\theta)}_{\text{prior density}} d\theta$$

- $\varphi(\theta)$ is **weight function** encoding 'relative likelihood' values of θ
- final analysis remains entirely frequentist!

[1] not really Bayesian since Bayes Theorem is never invoked!

probability of study success
extended Bayesian expected power 1
average power
predictive power strength
probability of success
average probability of success
prior-adjusted power
predictive probability of success
probability of statistical success
assurance
predictive frequentist power
hybrid Neyman-Pearson-Bayesian probability

- see [<https://arxiv.org/abs/2006.15715>] for extensive literature review

'Bayesian' Point Alternative

Justify Point Alternative

- "[...] a somewhat arbitrarily chosen location parameter of the [prior] distribution (for example the mean, the median or the 70th percentile) [...]" could be used to determine n [1]
- simply Bayesian justification of the choice of point alternative θ_{alt}
- naive: $\theta_{alt} = E[\Theta]$
- problems:

1. probability to reject is non-linear function of effect size!

$$\underbrace{\Pr_{E[\Theta]}[Z_n > 1.96]}_{\text{probability to reject at expected effect}} \neq \underbrace{E[\Pr_{\Theta}[Z_n > 1.96]]}_{\text{expected probability to reject}}$$

2. $E[\Theta]$ may lie in $\mathcal{H}_0 \rightsquigarrow$ required sample size undefined

[1] D J Spiegelhalter and L S Freedman. A predictive approach to selecting the size of a clinical trial, based on subjective clinical opinion. *Statistics in Medicine*, 5(1):1–13, 1986.

Prior-Quantile Approach (1)

- use $(1 - \gamma)$ -quantile of the prior distribution *conditional* on a relevant effect:

$$Q_{1-\gamma}[\Theta \mid \Theta \geq \theta_{\text{MCID}}] := \inf_x \Pr[\Theta \geq x \mid \Theta \geq \theta_{\text{MCID}}] \geq \gamma$$

- " $Q_{1-\gamma}[\Theta \mid \Theta \geq \theta_{\text{MCID}}]$ is the θ such that $\Pr[\Theta \geq \theta] = \gamma$ "
- chose n such that power at $\theta_{alt} = Q_{1-\gamma}[\Theta \mid \Theta \geq \theta_{\text{MCID}}]$ is 80%
- $Q_{1-\gamma}[\Theta \mid \Theta \geq \theta_{\text{MCID}}] \geq \theta_{\text{MCID}} \rightsquigarrow$ sample size always well-defined

\rightsquigarrow nothing but a more formal way of justifying $\theta_{alt} \geq \theta_{\text{MCID}}$

\rightsquigarrow compatible with all sample size formulas!

Prior-Quantile Approach (2)

- how is that better than using $\theta_{alt} = E[\Theta \mid \Theta \geq \theta_{MCID}]$?
- recall that

$$\underbrace{\Pr_{E[\Theta]}[Z_n > 1.96]}_{\text{probability to reject at expected effect}} \neq \underbrace{E[\Pr_{\Theta}[Z_n > 1.96]]}_{\text{expected probability to reject}}$$

- $\theta \mapsto \Pr_{\theta}[Z_n > 1.96]$ is monotononously increasing, thus

$$\Pr[\Theta \geq \theta^* \mid \Theta \geq \theta_{MCID}] = \gamma \quad \wedge \quad \Pr_{\theta^*}[Z_n > 1.96] \geq 1 - \beta$$

$$\Rightarrow \Pr \left[\underbrace{\Pr_{\Theta}[Z_n > 1.96]}_{\text{'random power'}} \geq 1 - \beta \mid \Theta \geq \theta_{MCID} \right] = \gamma$$

- in words: *"When calculating sample size based on the $1 - \gamma$ quantile of the prior conditional on a relevant effect such that the power at this point alternative is $1 - \beta$, the probability that 'random power' exceeds $1 - \beta$ is γ ."*
- requires additional parameter γ , how to justify? Is median ($\gamma = 0.5$) ok?

Probability of Success & Expected Power

Probability of Success (1)

- what constitutes a 'success' ?
- surprisingly vague in the literature, often needs to be reverse-engineered from formula for 'probability of success'
- common: 'success' := 'reject \mathcal{H}_0 ', i.e.

$$\begin{aligned}\text{PoS}'(n) &:= \Pr[Z_n > 1.96] \\ &= \int \Pr_{\theta}[Z_n > 1.96] \varphi(\theta) \, \mathrm{d} \theta\end{aligned}$$

- integration includes the null hypothesis - **type-one errors are successes!**

Greedy!



Probability of Success (2)

'greedy' definition:

'success' := 'reject \mathcal{H}_0 '

$$\begin{aligned}\text{PoS}'(n) &:= \Pr[Z_n > 1.96] \\ &= \int \Pr_{\theta}[Z_n > 1.96] \varphi(\theta) \, d\theta\end{aligned}$$

long-term definition:

'success' := 'reject \mathcal{H}_0 ' and $\Theta \geq \theta_{\text{MCID}}$

$$\begin{aligned}\text{PoS}(n) &:= \Pr[Z_n > 1.96, \Theta \geq \theta_{\text{MCID}}] \\ &= \int_{\theta_{\text{MCID}}} \Pr_{\theta}[Z_n > 1.96] \varphi(\theta) \, d\theta\end{aligned}$$

$$\text{PoS}'(n) = \underbrace{\Pr[Z_n > 1.96, \Theta \leq 0]}_{\text{probability of type-one error}} + \underbrace{\Pr[Z_n > 1.96, 0 < \Theta < \theta_{\text{MCID}}]}_{\text{probability of irrelevant rejection}} + \text{PoS}(n)$$

in practice, $\Pr[Z_n > 1.96, \Theta \leq 0]$ small due to strict type-one error rate control at α and most prior mass in \mathcal{H}_0 concentrated on θ_0 .

Probability of Success (3)

- critically depends on definition of 'success'
- numerical difference between $\text{PoS}(n)$ and $\text{PoS}'(n)$ mostly negligible in practice (but not always) [1]
- remember: if you go with $\text{PoS}'(n)$, type-one errors are successes!
- pick carefully, be explicit, justify!

[1] <https://arxiv.org/abs/2006.15715>

Sample Size Calculation with PoS

- idea: determine n such that $\text{PoS}(n) = 1 - \beta$
- note that

$$\begin{aligned}\text{PoS}(n) &= \Pr[Z_n > 1.96, \Theta \geq \theta_{\text{MCID}}] \\ &= \Pr[Z_n > 1.96 \mid \Theta \geq \theta_{\text{MCID}}] \Pr[\Theta \geq \theta_{\text{MCID}}] \\ &\leq \Pr[\Theta \geq \theta_{\text{MCID}}]\end{aligned}$$

- **problem:** what if $1 - \beta > \Pr[\Theta \geq \theta_{\text{MCID}}] \rightsquigarrow n = ?$ ⚡
- **solutions:**
 1. either re-scale usual $1 - \beta$ to make it problem-specific (how?) ...
 2. ... or re-scale $\text{PoS}(n)$ to $[0, 1]$

Enter: Expected Power

- luckily, there is a natural way to re-scale $\text{PoS}(n)$ to $[0, 1]$

$$\text{PoS}(n) = \Pr[Z_n > 1.96 \mid \Theta \geq \theta_{\text{MCID}}] \Pr[\Theta \geq \theta_{\text{MCID}}]$$

$$\begin{aligned} \Leftrightarrow \underbrace{\text{PoS}(n) / \Pr[\Theta \geq \theta_{\text{MCID}}]}_{\in [0,1]} &= \Pr[Z_n > 1.96 \mid \Theta \geq \theta_{\text{MCID}}] \\ &= \mathbb{E} \left[\underbrace{\Pr_{\Theta}[Z_n > 1.96] \mid \Theta \geq \theta_{\text{MCID}}}_{\text{'random power'}} \right] \\ &=: \text{EP}(n) \end{aligned}$$

- $\text{EP}(n)$ is 'expected power' and $\text{PoS}(n) = \Pr[\Theta \geq \theta_{\text{MCID}}] \text{EP}(n)$
- joint probability of rejection and relevant effect $\text{PoS}(n)$ vs. conditional probability of rejection given relevant effect $\text{EP}(n)$

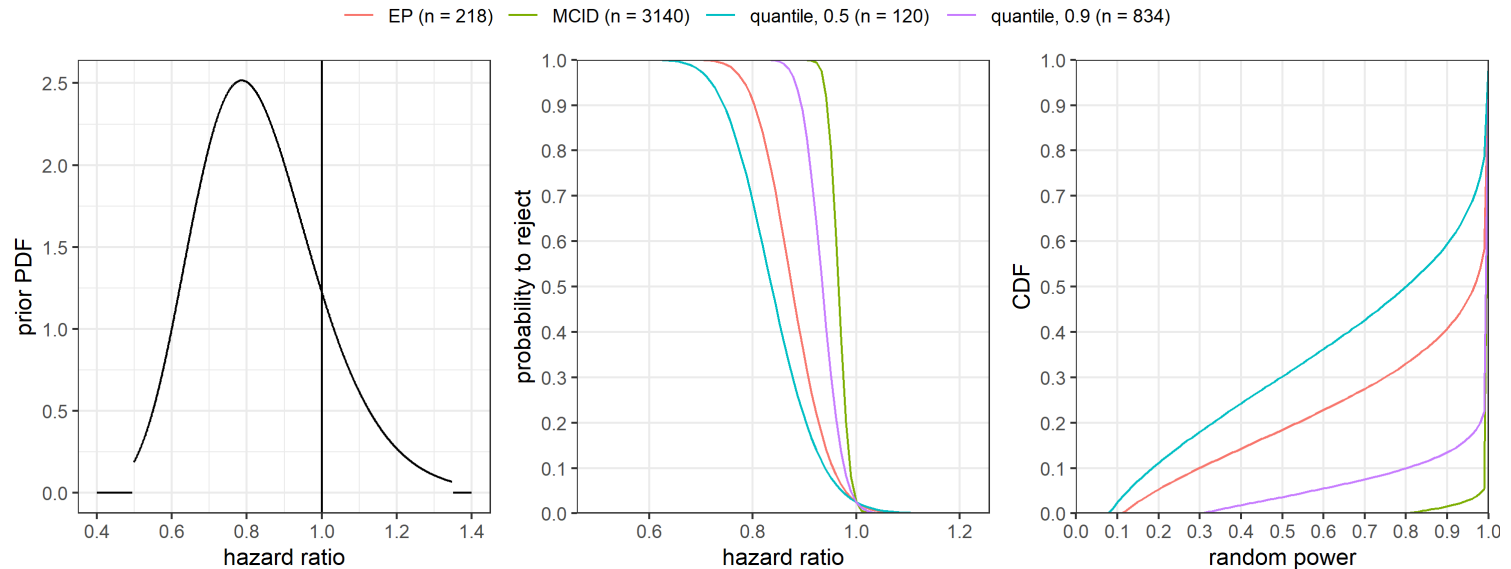
Just a Different Weight Function...

$$\begin{aligned} & \Pr[\Theta \geq \theta_{\text{MCID}}] \text{ EP}(n) \\ &= \Pr[\Theta \geq \theta_{\text{MCID}}] \int_{\theta_{\text{MCID}}} \Pr_{\theta}[Z_n > 1.96] \varphi(\theta | \Theta \geq \theta_{\text{MCID}}) \mathrm{d} \theta \\ &= \int_{\theta_{\text{MCID}}} \Pr_{\theta}[Z_n > 1.96] \underbrace{\Pr[\Theta \geq \theta_{\text{MCID}}] \varphi(\theta | \Theta \geq \theta_{\text{MCID}})}_{\varphi(\theta)} \mathrm{d} \theta \\ &= \text{PoS}(n) \end{aligned}$$

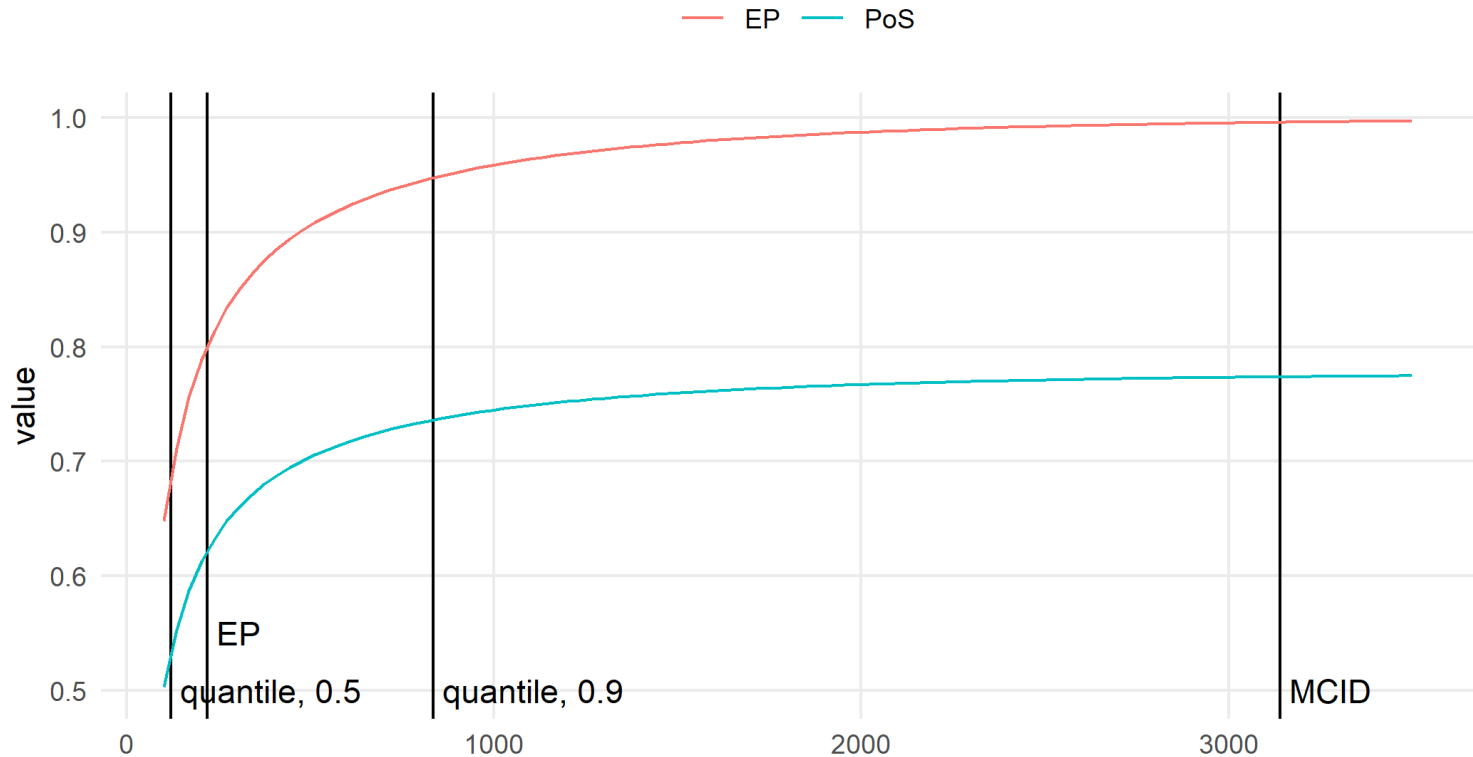
Example

Single-Arm, Survival Endpoint

- effect on hazard ratio scale $HR = \exp(-\theta)$
- $HR_{MCID} = 0.95, HR_0 = 1$
- $\alpha = 0.025, \beta = 0.2$



Expected Power and Pos vs n




Shiny/Code/Pre-Print

Thanks to Google Cloud (<https://cloud.google.com/>), OVH (<https://www.ovh.com/>), GESIS Notebooks (<https://notebooks.gesis.org>) and the Turing Institute (<https://turing.ac.uk>) for supporting us! 🐼



Starting repository: `kkmann/sample-size-calculation-under-uncertaintyv/master`

- pre-print:
[<https://arxiv.org/abs/2006.15715>]
- code:
[<https://github.com/kkmann/sample-size-calculation-under-uncertainty>]
-  Shiny latest
- follow-up: predictive power and sample size recalculation
[<https://arxiv.org/abs/2010.06567>]

Summary

- 'Bayesian' choice of θ_{alt} not straight-forward: power function is non-linear!
- 'quantile approach' is principled and simple to implement but requires additional parameter γ , confusing!
- definition of 'probability of success' is treacherous, not suitable for sample size calculation!
- 'expected power' is natural extension of power calculation to situation with prior information
- 'hybrid' methods make distinction between 'likely' & 'relevant' effects transparent
- Bayes Theorem not used - relevant when looking at **interim adaptations** [1]
- **no magic bullet** but consistent framework for incorporating uncertainty!

[1] <https://arxiv.org/abs/2010.06567>

Thanks!



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Slides created with the R package **xaringan**.