

Navigating Challenges in RCT Conduct: A Novel Bayesian Adaptive Semiparametric Approach Handling Primary and Secondary Endpoints in Pediatric Trial Design

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Background: Challenges in Pediatric RCT

- Conducting RCTs in pediatric settings presents several challenges.
 - Limited sample sizes (Huff, 2017),
 - Ethical considerations (Wightman, 2023),
 - Discordances in expert opinion about treatment effect (Linney, 2019),
 - Need to address multiple endpoints, i.e. safety secondary outcomes (Gkiourtzis, 2023).



Oral steroids for reducing kidney scarring in young children with febrile urinary tract infections: the contribution of Bayesian analysis to a randomized trial not reaching its intended sample size

Motivating Example, Rescue Trial

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- **RESCUE (REnal SCarring Urinary infEction)** trial is a randomized controlled double-blind trial
- The **study aims** to evaluate the effect of adjunctive oral steroids to prevent renal scarring in young children and infants with febrile urinary tract infections.
- **Extensive scarring** may progress to further renal injury with subsequent hypertension, decreased renal function, proteinuria, and sometimes end-stage renal disease (Peters, 2010).
- **Primary outcome** is the difference in scarring proportion between amoxicillin standard antibiotic therapy versus standard therapy + corticosteroids therapy.
- **Secondary outcome** acceptability of adjuvant steroid treatment in terms of the rate of discontinuation of treatment and the reported side effects.

Issues → Proposals

Issues in the Study Design in Pediatric RCT: The lesson learned

- Patient retention
- Highly informative Priors arising from the literature

- Advanced RCT and Bayesian approaches (Laptok, 2017)
- Power Prior Approaches and discounting factors (Ibrahim, 2015)

- Issues in Incorporating Expert opinion
- Secondary safety endpoints
- Prior data conflict
- Communication issues

- Semiparametric B-Spline priors (Azzolina, 2022)
- Two endpoints Bayesian sequential design (Gayewsky, 2023)



Two endpoint Bayesian sequential design

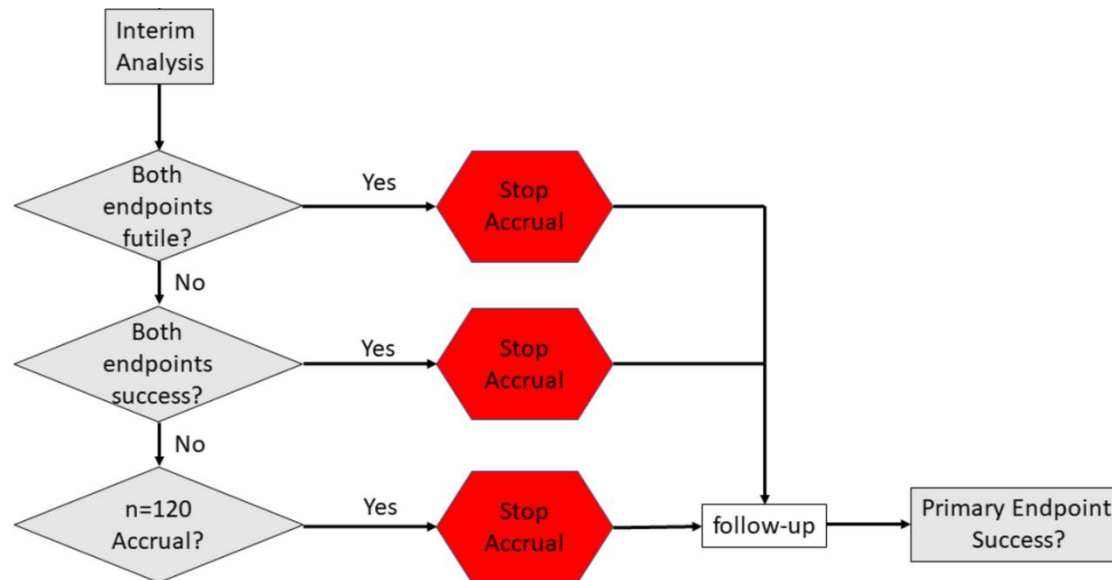
METHODOLOGY

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A novel Bayesian adaptive design incorporating both primary and secondary endpoints for randomized IIB chemoprevention study of women at increased risk for breast cancer

Byron J. Gajewski^{1,2*}, Bruce F. Kimler^{2,3}, Devin C. Koestler^{1,2}, Dinesh Pal Mudaranthakam^{1,2}, Kate Young^{1,2} and Carol J. Fabian^{2,4}



Only Parametric Priors?

Semiparametric Priors

1. Assuming to have p elicited quantiles $y_{\alpha_1}, \dots, y_{\alpha_p}$ modeled by a linear combination of B-spline, the prior distribution may be determined optimizing this objective function:

$$f(\theta, m, S, \phi, y) = \min_{F_{-m}, \dots, F_S} \left\{ \sum_{i=1}^p (\alpha_i - F(y_{\alpha_i}))^2 + \phi \int_{y_0}^{y_1} f(y)^2 dy \right\}$$

$$F_i \leq F_{i+1} \text{ for } i = -m, \dots, S - 1$$

$$\text{and } F_{-m} = 0, F_S = 1$$

2. F is a spline having m degree with a sequence of S inner knot $\lambda = (\lambda_{-m}, \dots, \lambda_{S+m+1})^T$.

3. $\phi > 0$ is a balancing factor penalizing the distance between the functions $F(y_{\alpha_i})$ adapted to the expert quantiles and the Uniform uninformative distribution in the domain $[y_0, y_1]$.

Design proposal

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Semiparametric Priors

Simulate the design propriety even if prior data conflict arise

Rescue Trial motivating example

Rescue Questions posed to the experts

“Based on your experience, what is the probability that a patient aged 0 to 2, with a value of procalcitonin $>1 \mu\text{g/L}$, treated with the recommended antibiotic regimen, has evidenced the presence of a renal scar event 6 months after the acute episode?”

“Based on your experience, what is the probability that a patient aged 0 to 2, with a value of procalcitonin $>1 \mu\text{g/L}$, treated with the recommended antibiotic regimen+dexametasone, has evidenced the presence of a renal scar event 6 months after the acute episode?”

Expert Opinions in Rescue Trial

| Expert | Opinion Control | Opinion Treatment |
|------------|-----------------|-------------------|
| 1 | 0.3 | 0.5 |
| 2 | 0.25 | 0.25 |
| 3 | 0.15 | 0.3 |
| 4 | 0.4 | 0.5 |
| 5 | 0.3 | |
| 6 | 0.2 | |
| 7 | 0.2 | 0.3 |
| 8 | 0.3 | 0.25 |
| | | |
| μ | 0.26 | 0.35 |
| σ | 0.08 | 0.12 |
| α_0 | 8 | 5 |
| β_0 | 22 | 10 |

Parametric Beta Priors

Beta(α, β)

$$\alpha = \alpha_0 d_0 + 1$$

$$\beta = \beta_0 d_0 + 1$$

$$\alpha_0 = \left[\left(\frac{1 - \mu}{\sigma^2} - \frac{1}{\mu} \right) \mu^2 \right] - 1$$

$$\beta_0 = \left[\alpha \left(\frac{1}{\mu} - 1 \right) \right] - 1$$



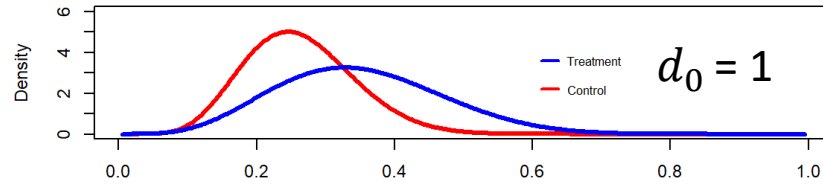
- $d_0 = 1$ Informative
- $d_0 = 0.5$ Low Informative
- $d_0 = 0$ Uninformative

Possible Prior-Data conflict

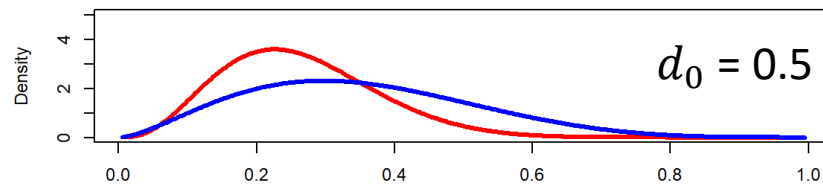
Elicited Priors

Parametric Beta

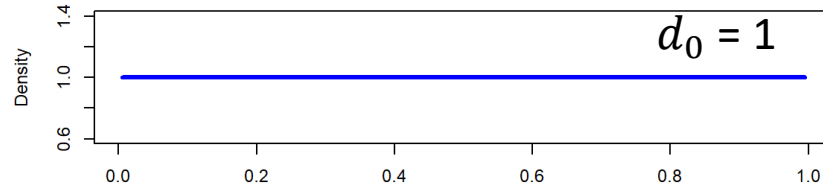
Informative prior



Low Informative prior

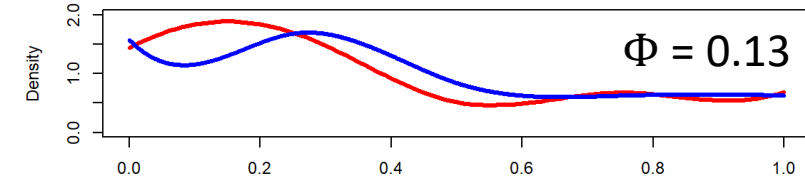


Uninformative prior

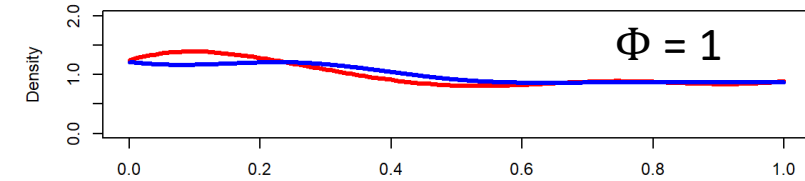


Semiparametric B Spline

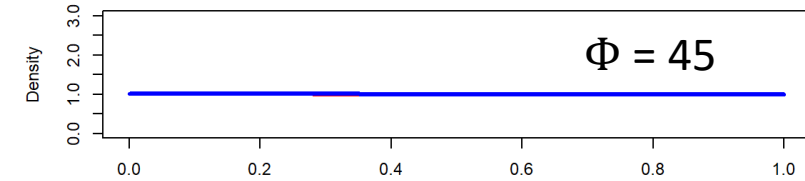
Informative prior



Low Informative prior



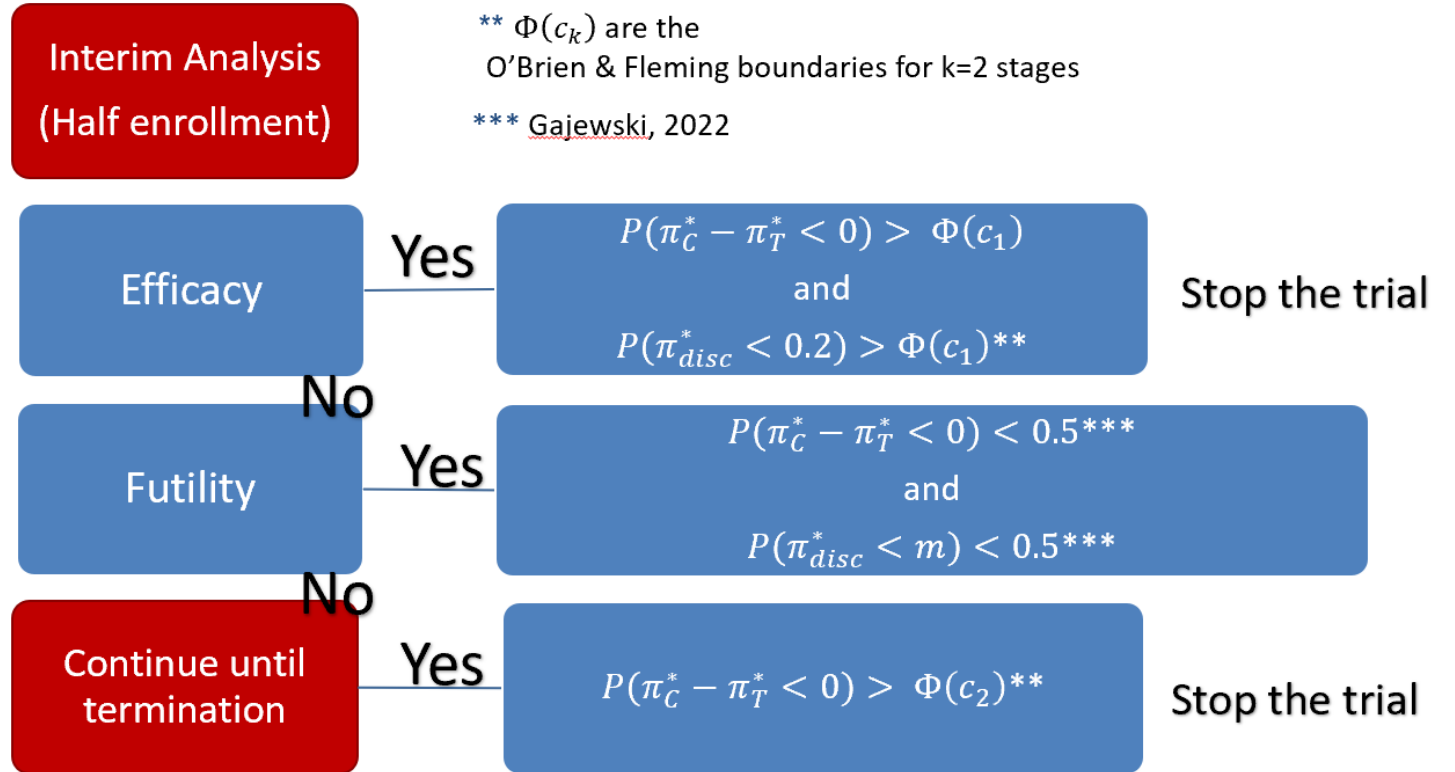
Low Informative prior



Trial design flowchart

** $pp_k = \Phi(c_k)$
 c_k are the O'Brien & Fleming boundaries
 Two-sided type I error of 5%.

*** Gajewski, 2022



** $\Phi(c_k)$ are the O'Brien & Fleming boundaries for k=2 stages

*** Gajewski, 2022

Simulation Plan

Data generation mechanism

Mimicking Rescue trial

$n = 40, \dots, 300$ per arm

$\Pi_{control} = 0.4$

$\Pi_{control} - \Pi_{treat} = 0; 0.18; 0.2$

$\Pi_{disc} = 0.18, \dots 0.22$



10,000 simulated data



Analysis

Two endpoints Bayesian Sequential design with

- Semiparametric B-Spline
- Parametric Beta priors



MCMC resampling:

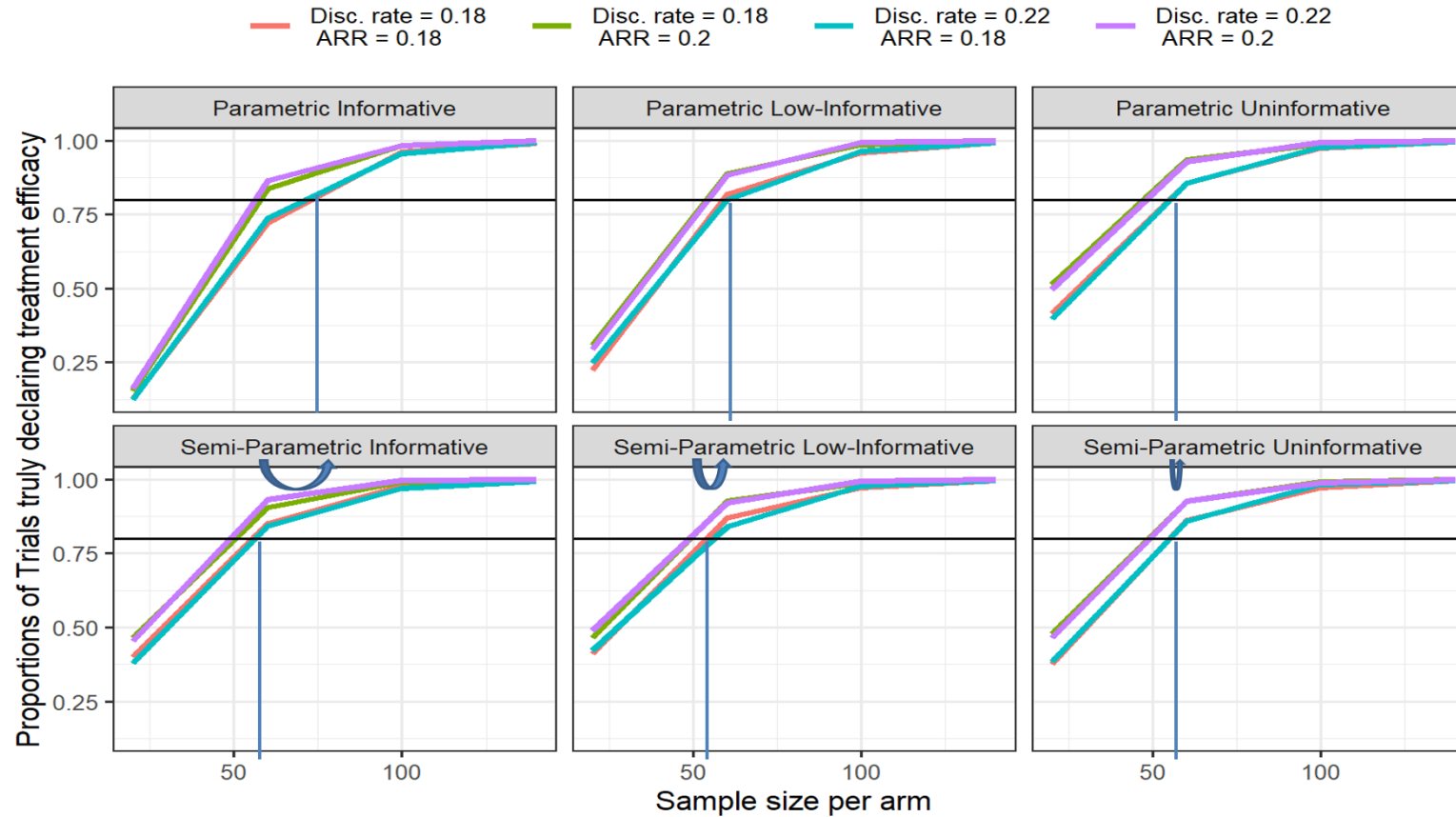
1. $\pi_{control}^*$ from $\pi_{control} | X_{control}$,
2. $\pi_{treatment}^*$ from $\pi_{treatment} | X_{treatment}$,
3. π_{disc}^* from $\pi_{disc} | X_{disc}$,
4. $ARR = \pi_{control}^* - \pi_{treatment}^*$



Design Proprieties indicators

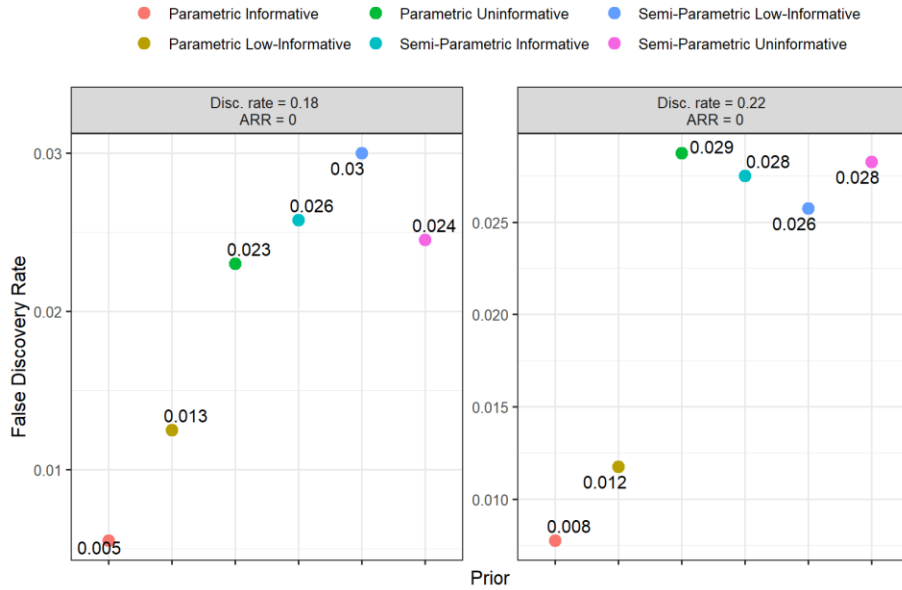
1. Percentage of trials truly declaring treatment efficacy
2. Percentage of trials declaring the treatment efficacy if the treatment does not work
3. Percentage of futility trials if the treatment is not effective

Results: Empirical Power

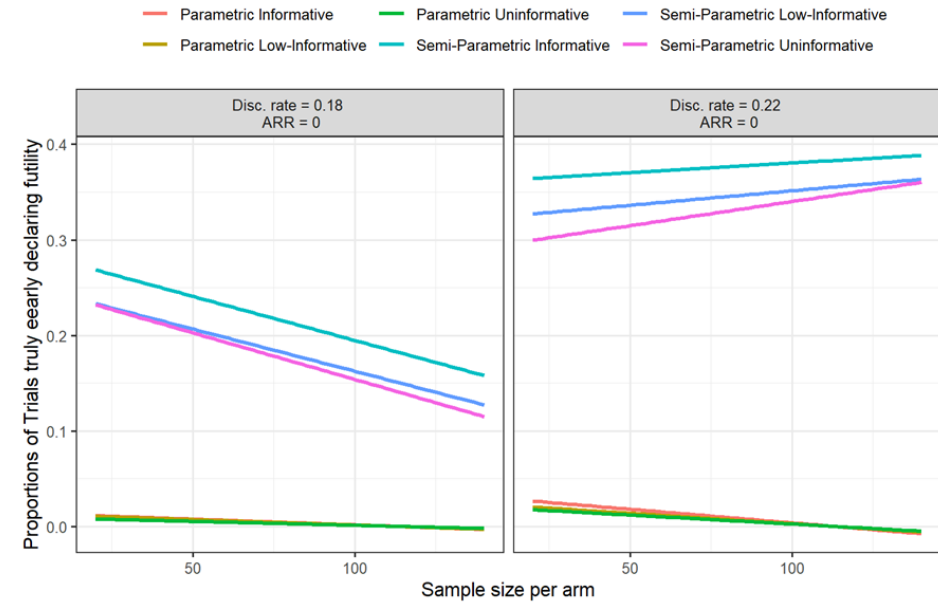


Proportions of simulated trials declaring the treatment effect, ad interim or at the end of the study, according to the sample size, simulation scenarios, and Prior Distributions

Results: False discovery rate and Correct Futility Rate



Average False Discovery Rate (FDR) over the sample size per simulation scenarios, and Prior Distributions



Proportions of simulated trials truly early declaring the futility ad interim, according to the sample size, simulation scenarios, and Prior Distributions

Implications

- **Enhanced Safety Monitoring:** Bayesian Sequential design aids in comprehensive evaluation of secondary safety endpoints, prioritizing pediatric patient welfare.
- **Improved Sensitivity:** Semiparametric priors outperform parametric priors, enabling precise identification of treatment effects in pediatric populations.
- **Strict Control of False Discoveries:** Maintains a nominal false discovery rate below 5%, ensuring reliable and trustworthy pediatric trial results.
- **Efficient Resource Allocation:** Allows early stopping for futility, optimizing resource utilization and expediting the development of effective pediatric treatments.

What's next?

Enhancing Advanced Design Communication and Applicability via Web Applications



Priors

Parametric (Beta)
 Semiparametric (B-Spline)

Expert elicitation

Control Opinions

Number of Experts for Control:

Control Event rate for Expert 1:

Control Event rate for Expert 2:

Control Event rate for Expert 3:

Treatment Opinions

Number of Experts for Treatment:

Control Event rate for Expert 1:

Control Event rate for Expert 2:

Control Event rate for Expert 3:

Control Event rate for Expert 4:

Design parameters

Sample size per arm (n):

Event rate in control arm:

Absolute Risk Reduction (ARR):

Secondary endpoint rate:

Acceptability rate:

Number of Simulations:

Prior Informativeness parameter:

