

Recent advances in methodology for clinical trials in small populations: the InSPiRe project

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Outline

The InSPiRe project

The challenge

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Motivation

Decision-theoretic model

Optimal sample size

Example



The InSPiRe project

The challenge

EU FP7 HEALTH2013.4.2-3 call:

New methodologies for clinical trials for small population groups

“to develop new or improved statistical design methodologies for clinical trials aiming at the efficient assessment of . . . a treatment for small population groups in particular for rare diseases or personalised . . . medicine”

Project funded February 2014 - May 2017



Progress and plans

WP1: Early phase dose-finding in small populations

Lead: Sarah Zohar (Paris)

- *develop and evaluate innovative designs for early-phase dose-finding trials*
- *develop efficient model-based designs using PK/PD data*

WP2: Decision-theoretic designs for small population clinical trials

Lead: Nigel Stallard (Warwick)


- *optimise trial design allowing for population size*
- *value-of-information approach*
- *determine appropriate levels of evidence*



WP3: Confirmatory trials for small populations and personalized medicines Lead: Martin Posch (Vienna)

- *develop methods for identification and confirmation of subgroups*
- *develop optimized adaptive enrichment designs*

WP4: Evidence synthesis in planning & interpretation of small population trials Lead: Tim Friede (Goettingen)

- *assess evidence synthesis methods in small populations*
 - *apply generalized evidence synthesis methods in paediatric studies*
- 

Sample size calculation in a small population

Motivation


Conventional method for definitive trial

- *fix error rates, α*
- *fix power, $1 - \beta$, to detect specified effect*

Concerned about consequences of incorrect conclusion

Idea:

model decision-making at end of trial explicitly
allow for population size and obtain sample size
accordingly



Decision-theoretic model

Total population size N

Two-arm trial: n_i patients in arm $i, i = 1, 2$

Observations Y_{i1}, \dots, Y_{in_i} with mean μ_i

Prior distribution with density $\pi(\mu_1, \mu_2)$

Gains:

patients in trial receiving treatment i : $n_i h(\mu_i)$

future patients receiving treatment i : $(N - n_1 - n_2)g(\mu_i)$
(if treatment i is recommended following trial)



Optimal decision at end of trial

Following observation of data Y ,
choose treatment $\arg \max E(g(\mu_i) \mid Y)$

Optimal trial sample size

Choose n_1 and n_2 to maximise

$$\sum_{i=1,2} \{n_i E(h(\mu_i))\} + (N - n_1 - n_2) E(\max E(g(\mu_i) \mid Y))$$


Optimal sample size

For small N , can obtain optimal n_1, n_2 directly

For Y with exponential family form distribution with conjugate prior and g satisfying δ -method
 N large $\implies n_i$ large $\implies g(\mu_i) \mid Y \sim \text{Normal}$

Optimum

$$n_i = \sqrt{\frac{N \int v(\mu) \pi(\mu, \mu) d\mu}{2 (E(\max_j \mu_j)) - E(\mu_i)}}$$

where $v(\mu_i) = \text{var}(Y_{ij})$

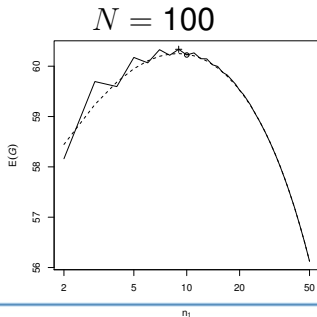
Note $n_i \sim N^{1/2}$ (extends result of Chen et al., 2003)



Example: single arm binary trial

Prior: $\mu_1 \sim \text{Beta}(1, 1)$ (prob. success for single arm)
 $\mu_2 = 0.5$ (prob. success for control)

$h(\mu) = g(\mu) = \mu$ (max. E(total no. successes))



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