## Bayesian clinical trials design and evaluation: a decision-theoretic view

Silvia Calderazzo, Manuel Wiesenfarth & Annette Kopp-Schneider

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## Why a decision-theoretic approach

A Bayesian decision-theoretic framework to clinical trial design and evaluation allows...

- providing a rationale for type I error inflation
  - by exploiting the relationship between error costs and prior probabilities on test decisions
  - by controlling a weighted sum of errors (see Grieve, 2015; Pericchi and Pereira, 2016)
- incorporating estimation and sampling costs

Andrew P. Grieve. How to test hypotheses if you must. *Pharmaceutical Statistics*, 14(2):139-150, 2015.

Luis Pericchi and Carlos Pereira. Adaptive significance levels using optimal decision rules: Balancing by weighting the error probabilities.

Brazilian Journal of Probability and Statistics. 30(1):70-90. 02 2016.

## Integrated risk

Test of  $H_0: \theta \leq \theta_0$  versus  $H_1: \theta > \theta_0$ Interest lies in minimizing the integrated risk

$$\begin{split} r(\pi,d) &= \int_{\Theta} \{c_1 \overbrace{I(\theta \leq \theta_0) P^f[\frac{d_{c_0,c_1}(\textbf{\textit{y}})}{(\textbf{\textit{g}})} = 1|\theta]}^{\text{type I error}} + c_0 \overbrace{I(\theta > \theta_0) P^f[\frac{d_{c_0,c_1}(\textbf{\textit{y}})}{(\theta > \theta_0)}]}^{\text{type II error}} \} \pi(\theta) d\theta \\ &+ \int_{\Theta} c_q E^f[\theta - d_q(\textbf{\textit{y}})]^2 \pi(\theta) d\theta + c_n n \end{split}$$

Optimal decisions:

for testing,

$$\mathbf{d}_{c_0,c_1}^{\pi}(\mathbf{y}) = \begin{cases} 1 & \text{if } P^{\pi}[\theta \leq \theta_0|\mathbf{y}] < c_0/(c_0 + c_1) & \text{(reject } H_0) \\ 0 & \text{otherwise} & \text{(keep } H_0), \end{cases}$$

- for estimation,  $d_q^{\pi}(y) = E^{\pi}[\theta|y]$
- for the sample size *n*, generally requires numerical procedures

## How can we exploit this machinery?

- Sensitivity analyses can be performed in the spirit of e.g. Sahu and Smith (2006) through the dichotomy between
  - Sampling (or design) prior  $\pi_s$ : generates the observed data and represents the 'truth'; induces **optimal decisions**  $d^{\pi_s}$
  - Analysis (or fitting) prior  $\pi_a$ : used to obtain the posterior distribution on which the **actual trial decisions**  $d^{\pi_a}$  are taken
- Such analyses can provide further insights into "robust" priors
- Sample size elicitation can be performed
  - through full risk optimization
  - to reach specific goals in testing and estimation ("goal sampling")

S.K. Sahu and T.M.F. Smith. A bayesian method of sample size determination with practical applications. *Journal of the Royal Statistical Society: Series A (Statistics in Society)*. 169(2):235-253. 2006.