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ISCB
Birmingham

Speaker: Moreno Ursino, PhD CRC, INSERM UMR 1138

Co-Authors: InSPiRe WP1

A Bayesian weighted quasi-likelihood design for Phase I/II clinical trial with repeated dose administration in preterm newborns







WP1 - InSPiRe

Innovative methodology for small populations research



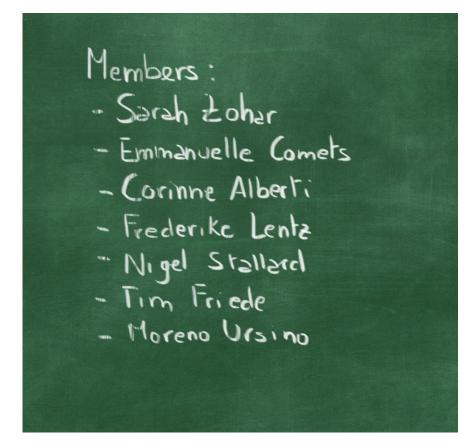
WP1 AIM

To develop novel methodology for improving **dose-finding** in early phase clinical trials.

Levneonat Clinical trial NCT02229123: a phase I/II trial aiming at finding the recommended dose of Levetiracetam for treating neonate's seizures was planned with a maximum sample size of 50.

Collaboration with:

- Dr Ying Yuan (MD Anderson Cancer Center, Houston, USA)
- Dr Geraldine Fevrais (Neonatal and pediatric intensive care unit, CHRU de Tours, Tours, France)



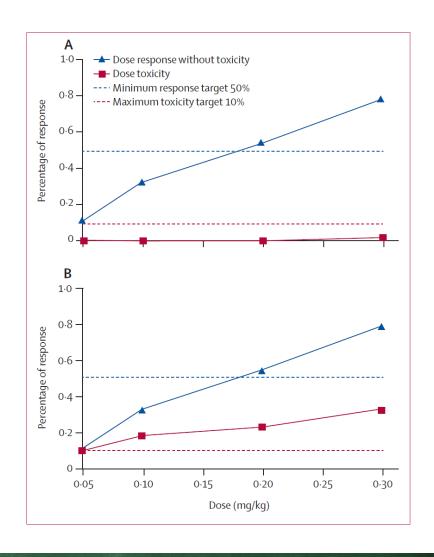
NEMO: NEonatal Seizure Using Medication Off-patent (NCT01434225)

Bumetanide for the treatment of seizures in newborn babies with hypoxic ischaemic encephalopathy (NEMO): an open-label, dose finding, and feasibility phase 1/2 trial

Ronit M Pressler*, Geraldine B Boylan*, Neil Marlow, Mats Blennow, Catherine Chiron, J Helen Cross, Linda S de Vries, Boubou Hallberg, Lena Hellström-Westas, Vincent Jullien, Vicki Livingstone, Barry Mangum, Brendan Murphy, Deirdre Murray, Gerard Pons, Janet Rennie, Renate Swarte, Mona C Toet, Sampsa Vanhatalo, Sarah Zohar, for the NEonatal seizure treatment with Medication Off-patent (NEMO) consortium†

A phase I/II dose-finding design with dual binary efficacy and safety endpoints.

During the trial no major adverse event were observed according to the safety composite endpoint defined in the protocol. However, after the inclusion of 14 patients unexpected safety event was measured, that is, hearing loss observed in three neonates at different doses.

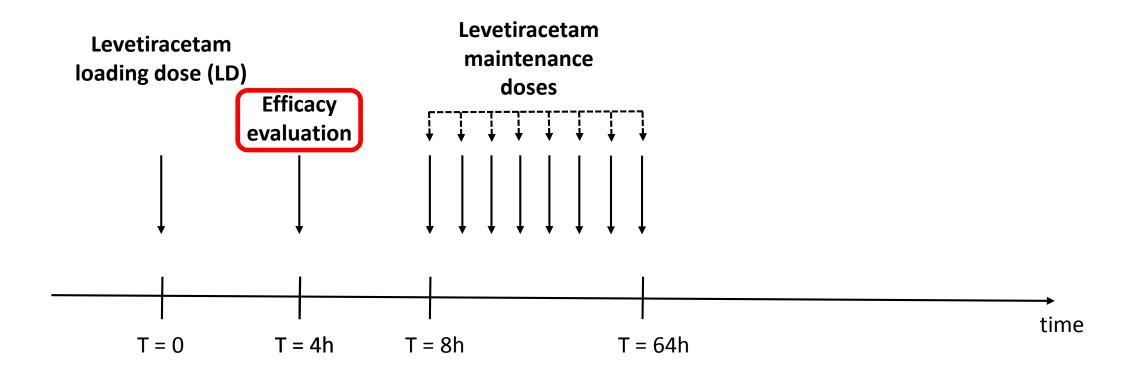


Levetiracetam loading dose (LD)



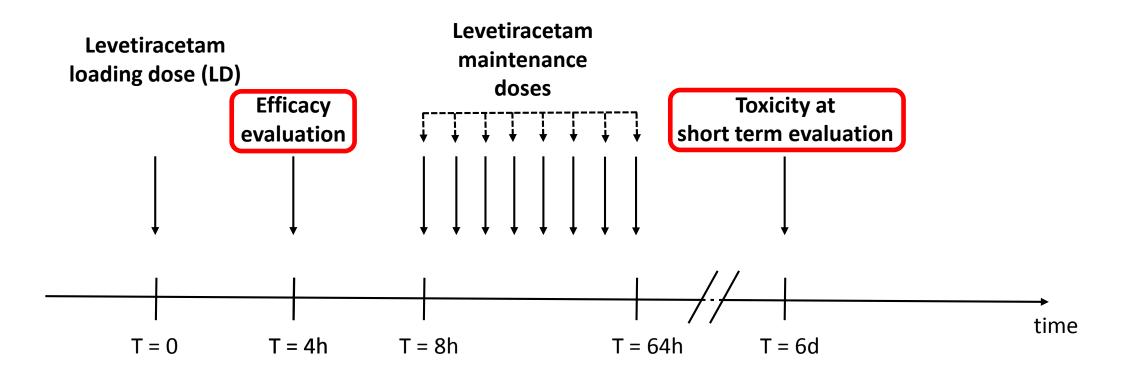


Efficacy: Reduction by 80% of seizures



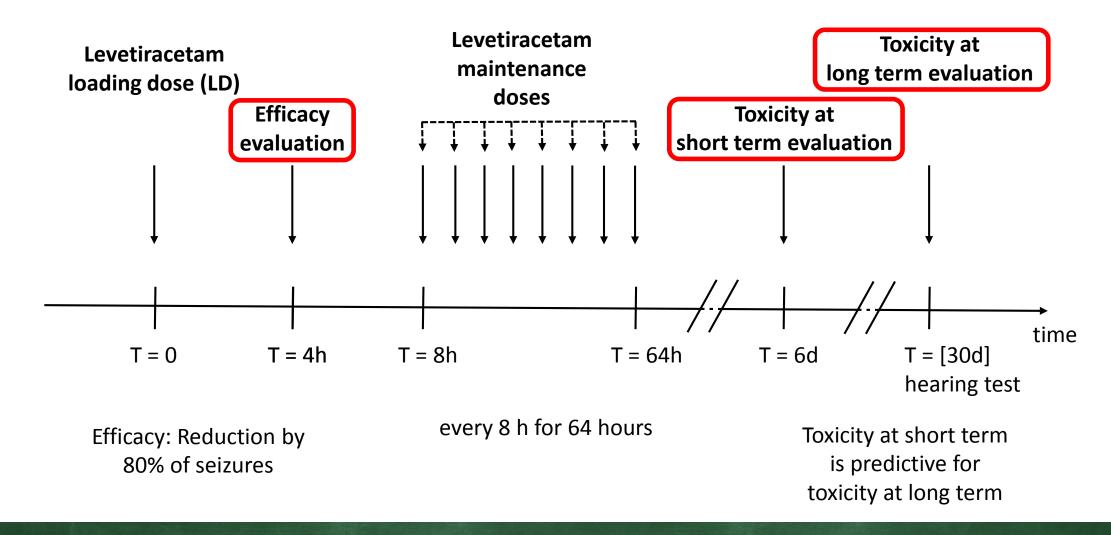
Efficacy: Reduction by 80% of seizures

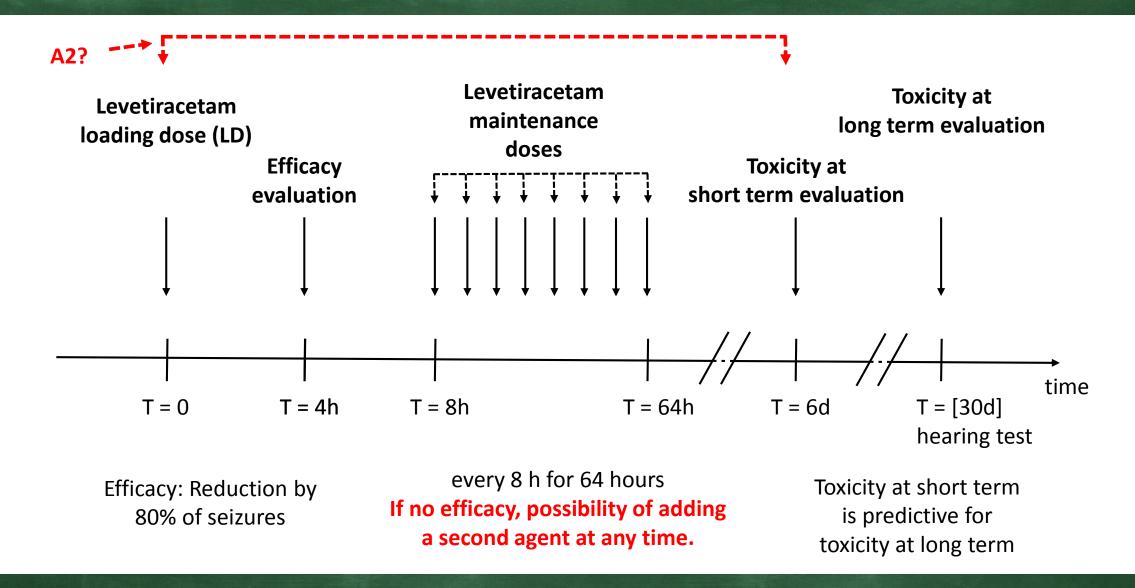
every 8 h for 64 hours



Efficacy: Reduction by 80% of seizures

every 8 h for 64 hours





Efficacy model

Bayesian logistic model (dose monotonicity)

Toxicity at short term (1)

$$WQd_{1}(y_{1}|y_{1},d_{1},\xi) = \frac{N_{1}}{|I|} \left(\left[1 - \left(1 - \frac{Mu}{I_{max}} \right) \right] P_{T_{1}} \right) \cdot \left(1 - \left[1 - \left(1 - \frac{Mu}{I_{max}} \right) \right] P_{T_{2}} \right) \cdot \left(1 - \left[1 - \left(1 - \frac{Mu}{I_{max}} \right) \right] P_{T_{2}} \right)$$

Toxicity at short term (2)

$$WQL_{1}(y_{1}|y_{1},\lambda_{1},\xi) = \frac{N_{1}}{\prod_{i=1}^{N_{1}} \left(\left[1 - \left(1 - \frac{M_{1}}{\prod_{max}} \right)^{\xi} \right] P_{T_{1}} \right)^{w_{i}} \cdot \left(1 - \left[1 - \left(1 - \frac{M_{1}}{\prod_{max}} \right)^{\xi} \right] P_{T_{2}} \right)^{w_{i}} \cdot \left(1 - \left[1 - \left(1 - \frac{M_{1}}{\prod_{max}} \right)^{\xi} \right] P_{T_{2}} \right)^{w_{i}} \cdot \left(1 - \left[1 - \left(1 - \frac{M_{1}}{\prod_{max}} \right)^{\xi} \right] P_{T_{2}} \right)^{w_{i}} \cdot \left(1 - \left[1 - \frac{M_{1}}{\prod_{max}} \right]^{\xi} \right) P_{T_{2}}$$

Probability of toxicity at short term

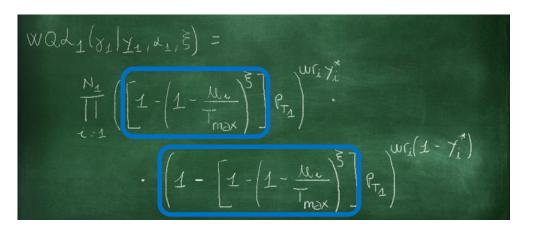
$$logit(R_1) = d_2 + e^{\delta_2} \times$$

$$\times \in \{\overline{d_1, ..., \overline{d_K}}\}$$

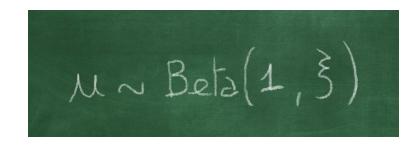
$$d_2 \quad \text{fixed}$$

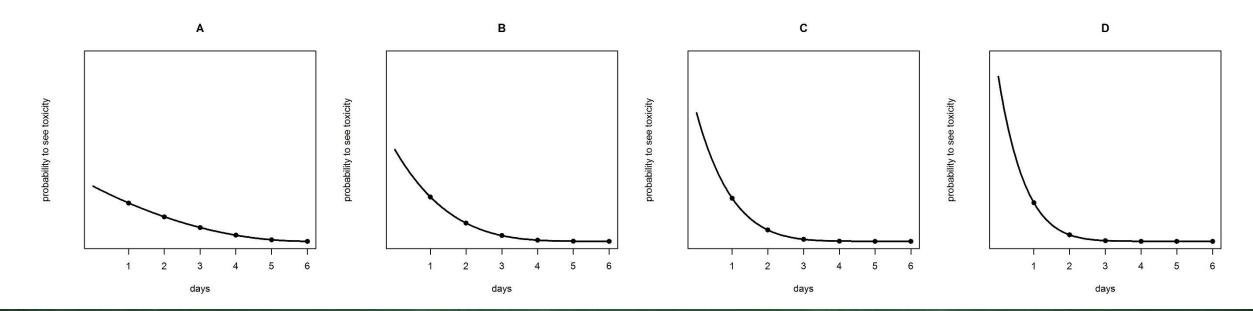
$$\delta_1 \sim \mathcal{N}(0, 1.34)$$

Toxicity at short term (3)

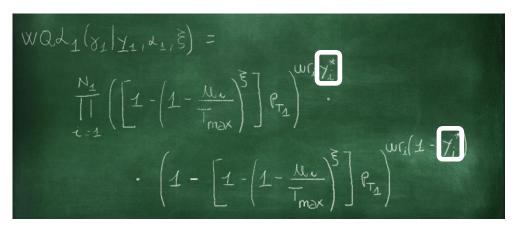


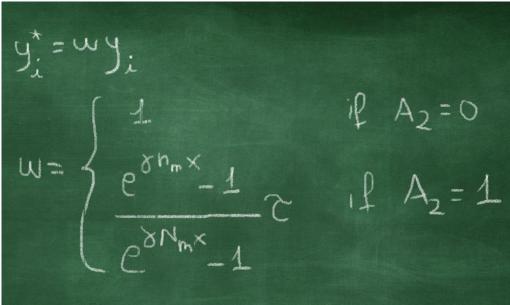
"Time-to-event"

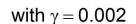


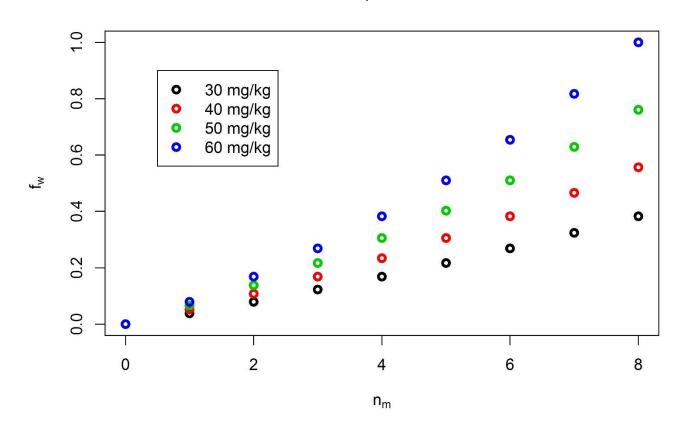


Toxicity at short term (4)



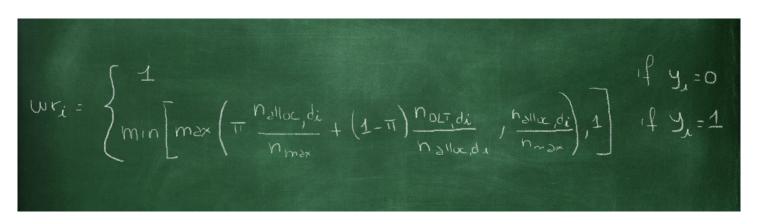


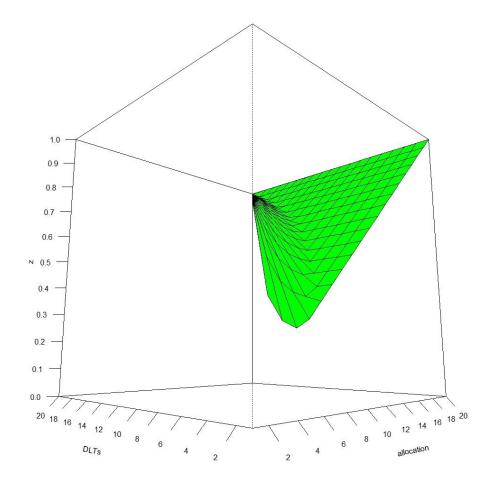




Toxicity at short term (5)

$$WQd_{1}(y_{1}|y_{1},d_{1},\xi) = \frac{N_{1}}{\prod_{i=1}^{N_{1}} \left(\left[1 - \left(1 - \frac{M_{1}}{\prod_{max}} \right)^{\xi} \right] P_{T_{1}} \right)}{\left(1 - \left[1 - \left(1 - \frac{M_{1}}{\prod_{max}} \right)^{\xi} \right] P_{T_{2}} \right)} \cdot \left(1 - \left[1 - \left(1 - \frac{M_{1}}{\prod_{max}} \right)^{\xi} \right] P_{T_{2}} \right)$$





Toxicity at long term

Conditional probability

$$\begin{aligned} & \text{logit}(P_{T_2}|Y_{T_1}) = J_3 + e^{S_1} \times + e^{S_2} y_1^* \\ & \times \epsilon \{d_1, ..., d_k\} \\ & J_3 \text{ fixed} \\ & S_1 \sim \mathcal{N}(0, 1.34) \\ & S_2 \sim \mathcal{N}(0, 1.34) \end{aligned}$$

Trial settings

Cohort: 2 neonates

Stopping rules:

$$P(p_{T1} > 0.1|d_1) > 0.9$$

$$P(p_{T2} > 0.1|d_1) > 0.9$$

$$P(p_E < 0.6|d_K) > 0.9$$

Dose allocation rules:

$$P(p_{E} < \tau_{e} - \epsilon_{e}) < g(N_{e})^{1_{N_{e}} > 11}$$

$$P(p_{T1} > \tau_{p1} + \epsilon_{1}) < g(N_{1})$$

$$P(p_{T2} > \tau_{p2} + \epsilon_{2}) < g(N_{2})^{1_{N_{2}} > 1}$$

$$g(N) = \max\left(0.5, 0.9 \frac{1}{1 + 0.04 * N}\right)$$

Dose selected: the one which has the highest efficacy among them selected through the previous constraints

Dose selection rules:

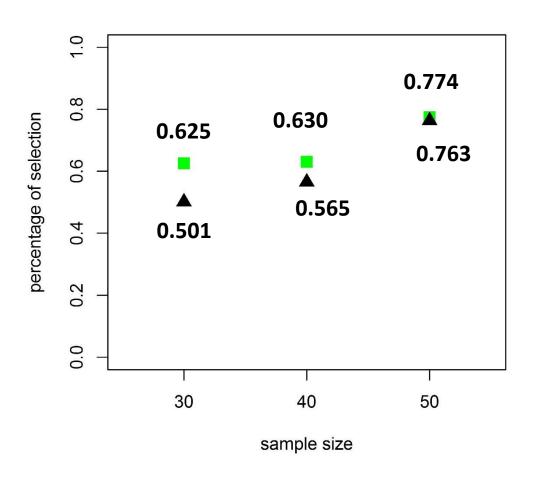
$$d_{e,\min} = \arg\min_{d \in D} |d - \tau_e|$$

$$d_{t,\max} = \min(\arg\min_{d \in D} |d - \tau_{p1}|,$$

$$\arg\min_{d \in D} |d - \tau_{p2}|)$$

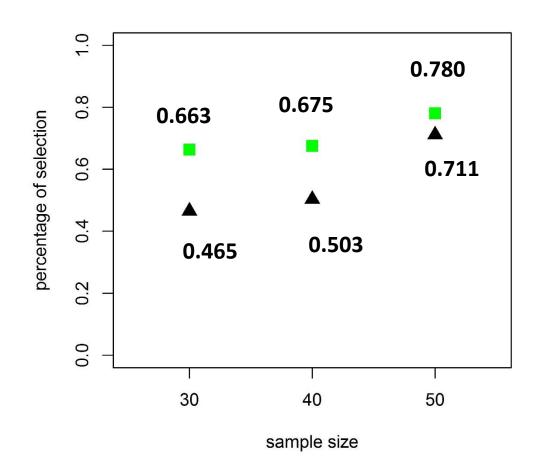
Simulations and Results (1)

	d ₁	d ₂	d_3	d ₄
p _E	0.001	0.01	0.1	0.2
p _{T1}	0.001	0.01	0.1	0.2
p _{T2}	0.6	0.7	0.8	0.9
P add A2	0			
p _{T1 A2}				
p _{T2 A2}				



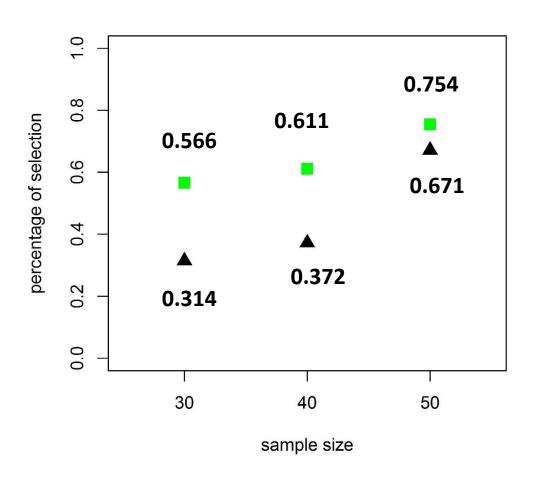
Simulations and Results (2)

	d ₁	d ₂	d_3	d ₄
p _E	0.001	0.01	0.1	0.2
p _{T1}	0.001	0.01	0.1	0.2
p _{T2}	0.6	0.7	0.8	0.9
P add A2	0.5			
p _{T1 A2}	0.005	0.05	0.15	0.25
p _{T2 A2}	0.005	0.05	0.15	0.25



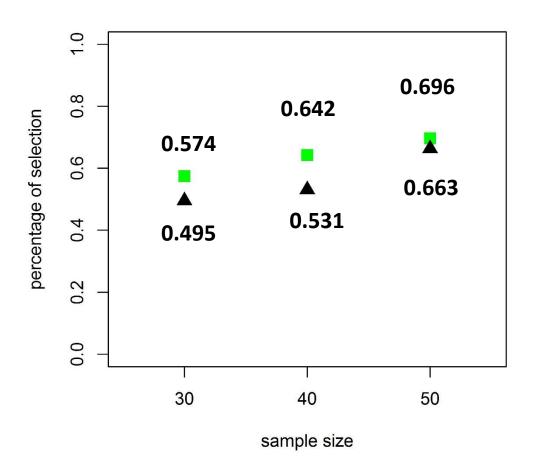
Simulations and Results (3)

	d ₁	d_2	d ₃	d ₄
p _E	0.01	0.1	0.25	0.35
p _{T1}	0.009	0.1	0.18	0.26
p _{T2}	0.6	0.7	0.8	0.9
P add A2	0.5			
p _{T1 A2}	0.01	0.01	0.25	0.35
p _{T2 A2}	0.01	0.01	0.25	0.35



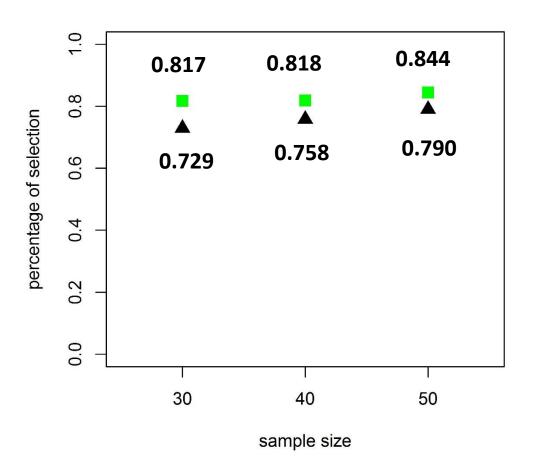
Simulations and Results (4)

	d ₁	d_2	d ₃	d ₄
p _E	0.001	0.01	0.1	0.2
p _{T1}	0.01	0.1	0.2	0.3
p _{T2}	0.6	0.7	0.8	0.9
P add A2	0.5			
p _{T1 A2}	0.005	0.05	0.15	0.25
p _{T2 A2}	0.005	0.05	0.15	0.25



Simulations and Results (5)

	d_1	d ₂	d ₃	d_4
p _E	0.001	0.005	0.01	0.05
p _{T1}	0.001	0.007	0.015	0.05
p _{T2}	0.3	0.4	0.5	0.6
P add A2	0.5			
p _{T1 A2}	0.005	0.009	0.012	0.06
p _{T2 A2}	0.005	0.009	0.012	0.06



Conclusion

This model could be a good trade-off for this clinical trial in which we need to deal with small sample size, tail probability estimation and, of course, safety of neonates.

Relevance weight can help at the beginning of the dose allocation to avoid to be stuck.

We improve the percentage of right dose selection without increasing a lot the dose limiting toxicities.

Improvements:

- Continuous analysis for efficacy at the end of the trial:
 - Bayesian beta regression
- Pharmacokinetics analysis including covariables in order to try to adjust the dose selected for each neonate subgroups

Aknowledgement



Sarah Zohar

Emmanuelle Comets

Frederike Lents

Corinne Alberti

Nigel Stallard

Tim Friede



Dr Ying Yuan



Dr Geraldine Fevrais (Neonatal and pediatric intensive care unit, CHRU de Tours) and all the statisticians and physicians of the group.