

# Discussion Meeting for MCP-Mod Qualification Opinion Request

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Novartis

10 July 2013

EMA, London, UK

# Attendees

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## Face to face:

- Dr. Frank Bretz    Global Statistical Methodology Head, Novartis
- Dr. Björn Bornkamp            Expert Statistical Methodologist, Novartis
- Dr. Geneviève Le Visage    Regulatory Intelligence Head, Novartis

## By telephone:

- Dr. José Pinheiro Senior Director, Janssen Research & Development

# Agenda

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## 1. Introduction:

- Qualification request
- Brief introduction to MCP-Mod
- In-scope, out-scope

## 2. Answers to Issues 5 – 11 raised by the SAWP

# Qualification request

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- Novartis is seeking qualification of MCP-Mod as an

*Efficient statistical methodology for model-based design and analysis of Phase II dose finding studies under model uncertainty.*

- The data supportive of this request consists of the following elements:
  - Worked examples, extensive simulations and real-life case studies to describe and quantify the performance
  - References from medical and statistical literature to illustrate applicability

# Background on MCP-Mod methodology

- MCP-Mod stands for:  
**M**ultiple **C**omparisons & **M**odelling

- Combines testing and estimation

- Design stage

- Pre-specification of candidate dose-response models

- Analysis stage: MCP-step

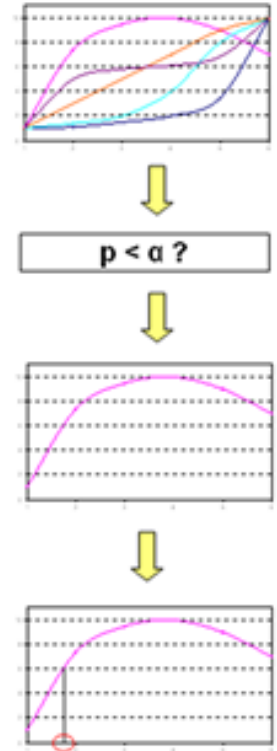
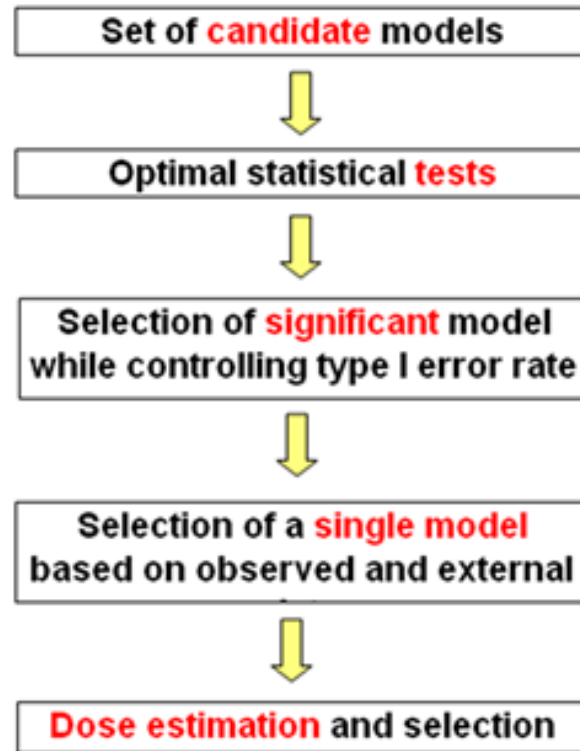
- Statistical test for dose-response signal. Model-selection based on significant dose-response models

- Analysis stage: Mod-step

- Dose-response and target dose estimation based on dose-response modelling

- Difference to traditional pairwise comparisons

- Use of dose-response modelling
- But, taking model uncertainty into account at *design* and *analysis* stage



# In-scope

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- Drug development stage
  - Phase II dose finding studies to support dose selection for Phase III
  
- Response
  - Univariate (efficacy or safety) variable (could be a binary, count, continuous or time-to-event endpoint). Observations could be cross-sectional (i.e. from a single time point) or longitudinal.
  
- Dose
  - Could be any univariate, continuous, quantitative measurement, as long as an ordering of the measurements is possible and the differences between measurements are interpretable
  
- Number of doses
  - For the MCP-step at least two distinct active doses are required
  - For the Mod-step, a minimum of three active doses required

# Out-of-scope or limited experience

- Predictions from a surrogate / biomarker or short term readout to a clinical Phase III endpoint.
- Titration designs and dose escalation studies (e.g. to estimate the maximum tolerable doses using continual reassessment methods).
- Exposure-response analyses or PK-PD models are not the purpose of this request, per se.
- Regimen finding for biologics where there is no steady state.
- Application of MCP-Mod in confirmatory studies.
- Multivariate problems, e.g., joint modeling of efficacy and toxicity, the presence of two primary endpoints, or drug combination trials.

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# Answers to Issues 5 – 11



# Issue 5

## *Selection of dose-range, number of doses and spaces of doses*

- Can the procedure itself directly help with these choices?
  - Maximum dose: Based on information from previous trials
  - Optimal design theory and clinical trial simulations
    - Input: Anticipated dose-response shapes & trial objective(s)
    - Output: Number and location of doses and allocation ratios to the doses
  - In practice one might deviate from optimal designs
    - Logistical/manufacturing constraints, considerations beyond primary efficacy endpoint
- ... guidance for an optimal strategy for these pre-selection exercises?
  - Candidate models: Honest reflection of potential dose-response curves
    - Not too many shapes (decrease in efficiency), too few shapes (risk of biased results)
    - Often 3-7 dose-response models/shapes seem sufficient
  - Dose-range, number of doses, location of doses case-specific; rules of thumb:
    - >10-fold dose-range, 4-7 active doses, logarithmic dose-spacing

# Issue 6

## *Considerations to optimise the choice of sample size*

- Is this optimally based on the precision with which the dose-response curve can be characterised, which would also need to consider dose-spacing and number of doses?
  - Sample size calculations should reflect the study objectives
  - Estimating dose response (DR) is considerably harder than testing it
  - Sample sizes for dose finding studies, based on power to detect DR signal, are inappropriate for dose selection and DR estimation
- Is there a minimum level of information below which the relative benefits of an MCP-Mod approach are lost compared to a 'traditional' approach?
  - Particularly in situations with small sample sizes, borrowing strength through modelling is beneficial, although validation of assumptions becomes difficult
  - MCP-Mod requires at least two (three) active doses for the MCP (Mod) step
    - Traditional approaches don't perform well either for a small number of doses

# Issue 7

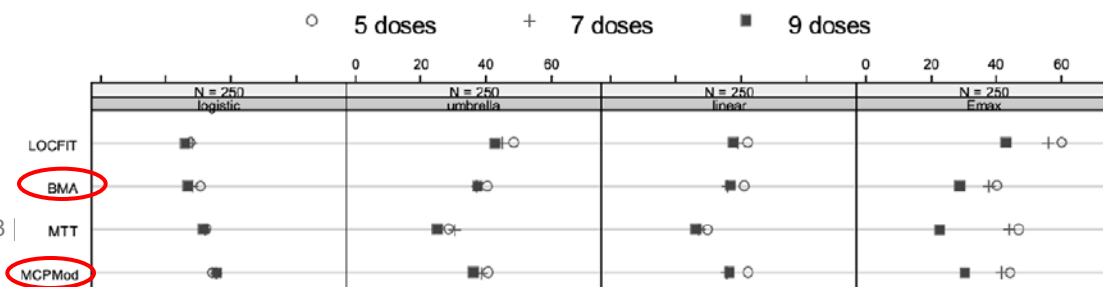
## *Rationale and the choice of nominal significance level*

- Is control at the traditional 5% level optimal from a sponsor point of view?
  - Depends on the specific trial and context
    - Understanding the false positive rate is important for any decision procedure
    - What certainty does the company need in the decision to move forward?
    - What is being tested
      - dose response signal detection relative to placebo? active control?
  - One major focus of MCP-Mod is estimation of the dose-response curve
    - if sample size was calculated for estimation, power for signal detection will be high
- Under what circumstances might the data exhibit a dose-response of interest but the procedure fails to identify this?
  - Idea of MCP-Mod: Define „dose-responses of interest“ at the design stage
    - Design of the study (doses, sample size) can be chosen to be able to identify these
  - When a dose-response signal cannot be identified with MCP-Mod, the effect size of the drug is most likely smaller than anticipated

# Issue 8a

## Model selection: Using more than one model

- Is it plausible to select more than one model with which to continue development?
  - It is likely that model uncertainty will remain after completing Phase II
  - If uncertainty remains, more than one model might be kept for future use (especially if MCP-Mod used with model averaging)
- ... how a model averaging approach can improve over the use of a single model when multiple pre-selected models are found to be of interest?
  - Difference in interpretation
    - "a" single model vs. "weighted average" of >1 model
  - Average performance is rather similar; see e.g. plot of correct target dose interval probabilities from Bornkamp et al. (2007)



# Issue 8b

## *Model selection: Challenges*

- ... find that a model shape not included in the initial set of potential models actually perform better than a model in the initial set. Is it a realistic possibility? How can such situation be handled pragmatically in the framework of MCP-Mod?
  - For a reasonably broad candidate set often one model will be a good approximation
  - MCP-Mod just one component for the decision making in view of Phase III
- Describe the properties in situations when the selection of trial doses turn out to be flawed such that model selection is driven by a set of doses with zero or maximal effect?
  - Estimation of the increasing part of the dose-response curve (and target dose) challenging, inferences driven by the model assumptions
  - Important to quantify uncertainty (parameter estimates and models)
  - Response-adaptive designs may offer the opportunity to react accordingly

# Issue 9

*'interpolation' between doses and 'extrapolation' outside the dose range*

- Discuss to what extent the procedure can support selection of a dose that has not been directly studied
  - Interpolation between doses is possible and encouraged
  - Extrapolation outside the dose range is discouraged
- Is inference restricted to the discrete set of doses used in the trial?
  - Traditional methods based on pairwise comparisons are not designed for extrapolation of information beyond the observed dose levels
  - MCP-Mod allows interpolation between doses under investigation
    - Recommend to always report uncertainty, e.g. on the "y-axis" (= effect estimates) or on the "x-axis" (= dose estimate)
    - Possibly accounting for multiplicity, e.g. use simultaneous confidence bands around dose response estimate instead of marginal confidence intervals at each dose

# Issue 10

## *Increase in efficiency compared to traditional pairwise comparisons*

- Does any increase in efficiency compared to traditional pairwise comparisons come at any cost to the developer, perhaps in terms of having less evidence to support for a particular dose level to take forward to Phase III?
  - Increase in efficiency by using modelling assumptions (i.e. prior information)
    - Testing and estimation gets optimized for realistic alternatives
    - Trade-off for unrealistic scenarios (e.g., zig-zag dose-response curve)
  - The dose to take forward to Phase III
    - Smoothing of dose-response estimates helps to safeguard against random highs (and lows), leading to a more robust planning for Phase III

# Issue 11

*Applicability ... without regard to therapeutic area or class of compound*

- MCP-Mod is applicable in any therapeutic area, since it essentially uses empirical dose-response models
- Discuss application in the context of dose selection that needs to consider both safety and efficacy
  - Any dose selection for Phase III requires safety / efficacy considerations
  - Need to understand safety / efficacy dose response relationships to estimate MED / MSD and thus the therapeutic window
  - Safety dose-response modelling less common, but MCP-Mod could be applied equally well
- Is there any quantitative approach to the synthesis of two univariate models, one for a key efficacy marker or parameter and one for safety?
  - One possibility is to derive a clinical utility index (CUI) that combines safety and efficacy information in one variable
  - In practice, derivation of CUI is quite hard
  - Limited experience at Novartis, but in principle MCP-Mod could be applied (unimodal shapes!)



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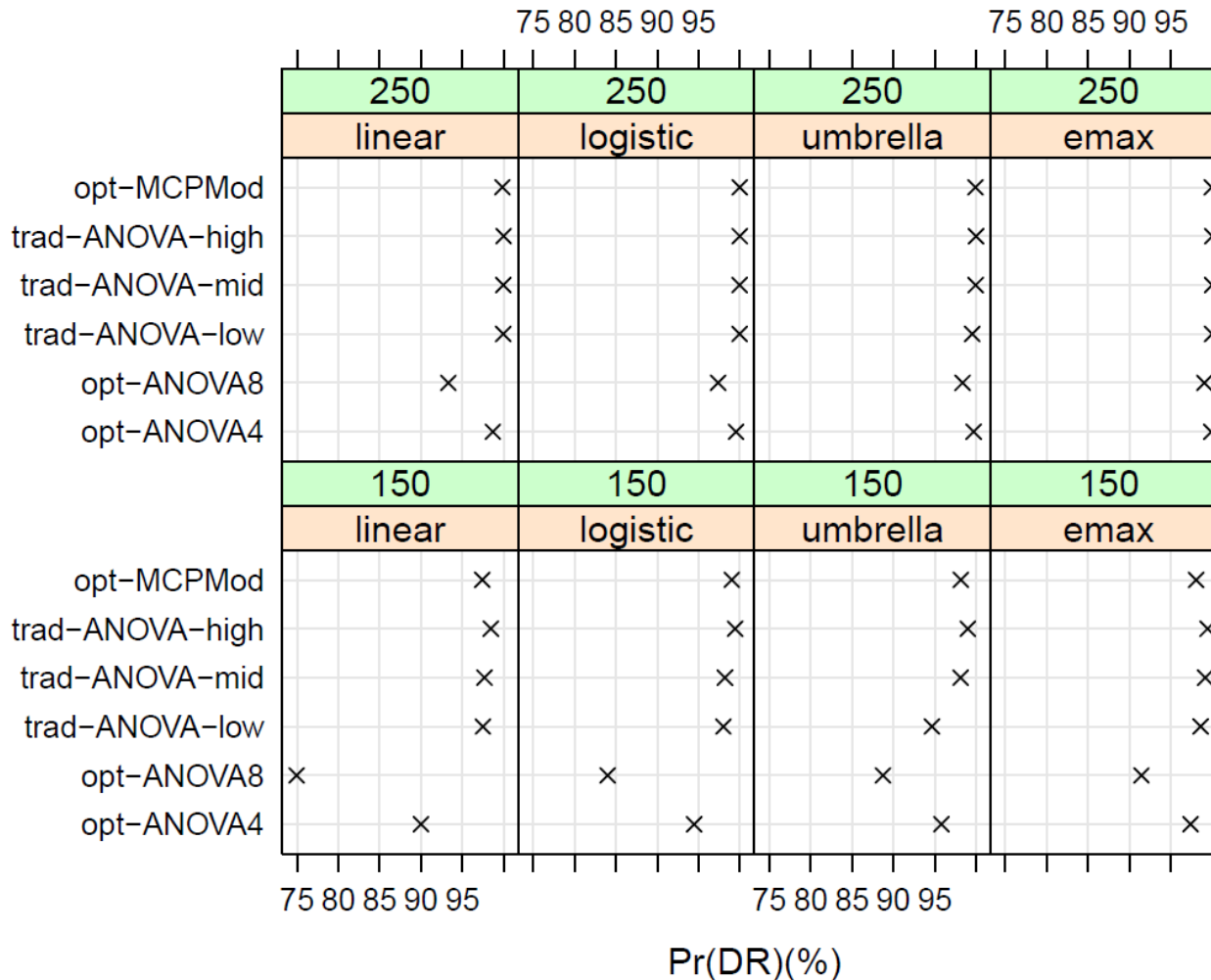
# Backup slides

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# Simulation Results Issue 1

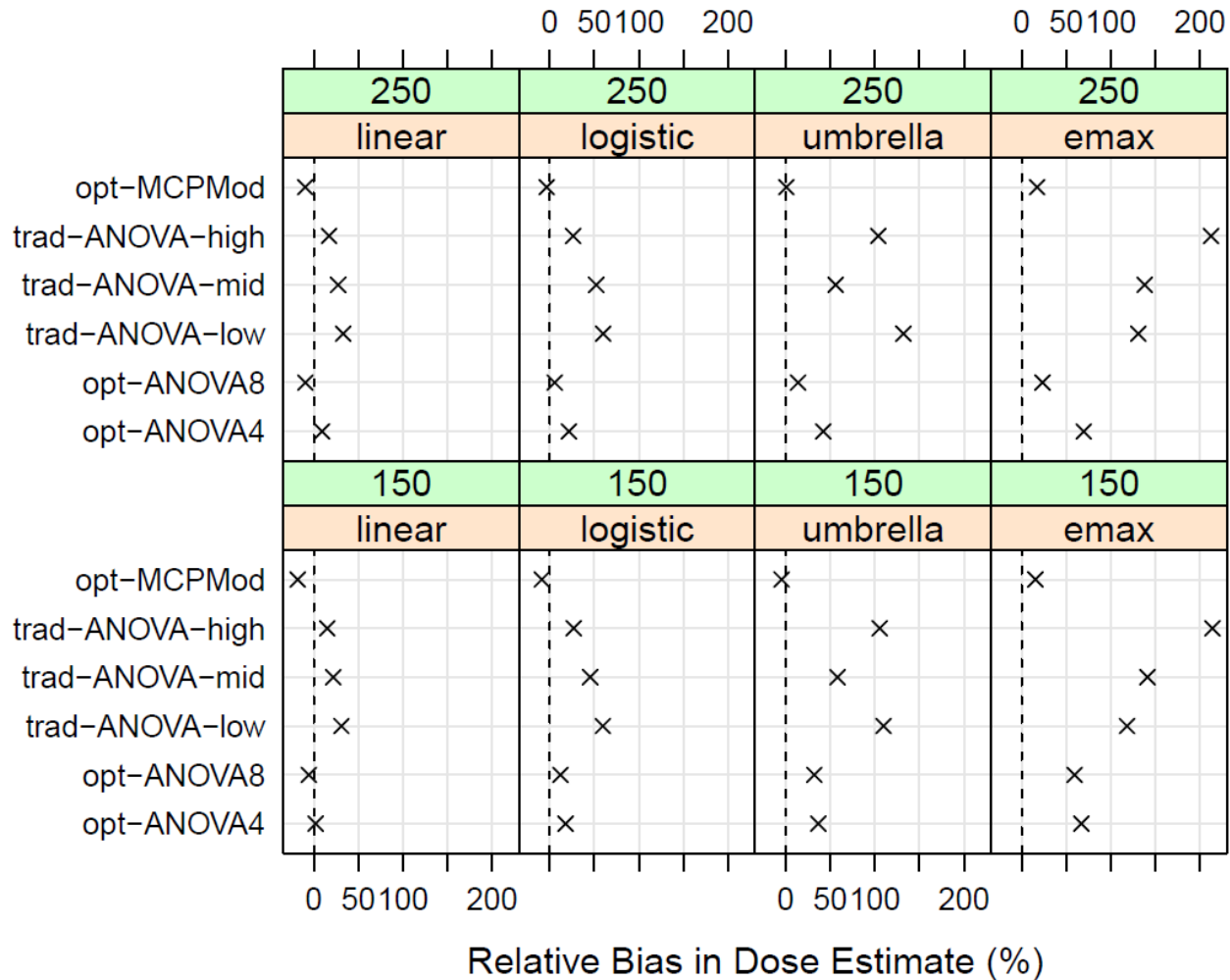
# Simulation Results

## Power to detect dose-response



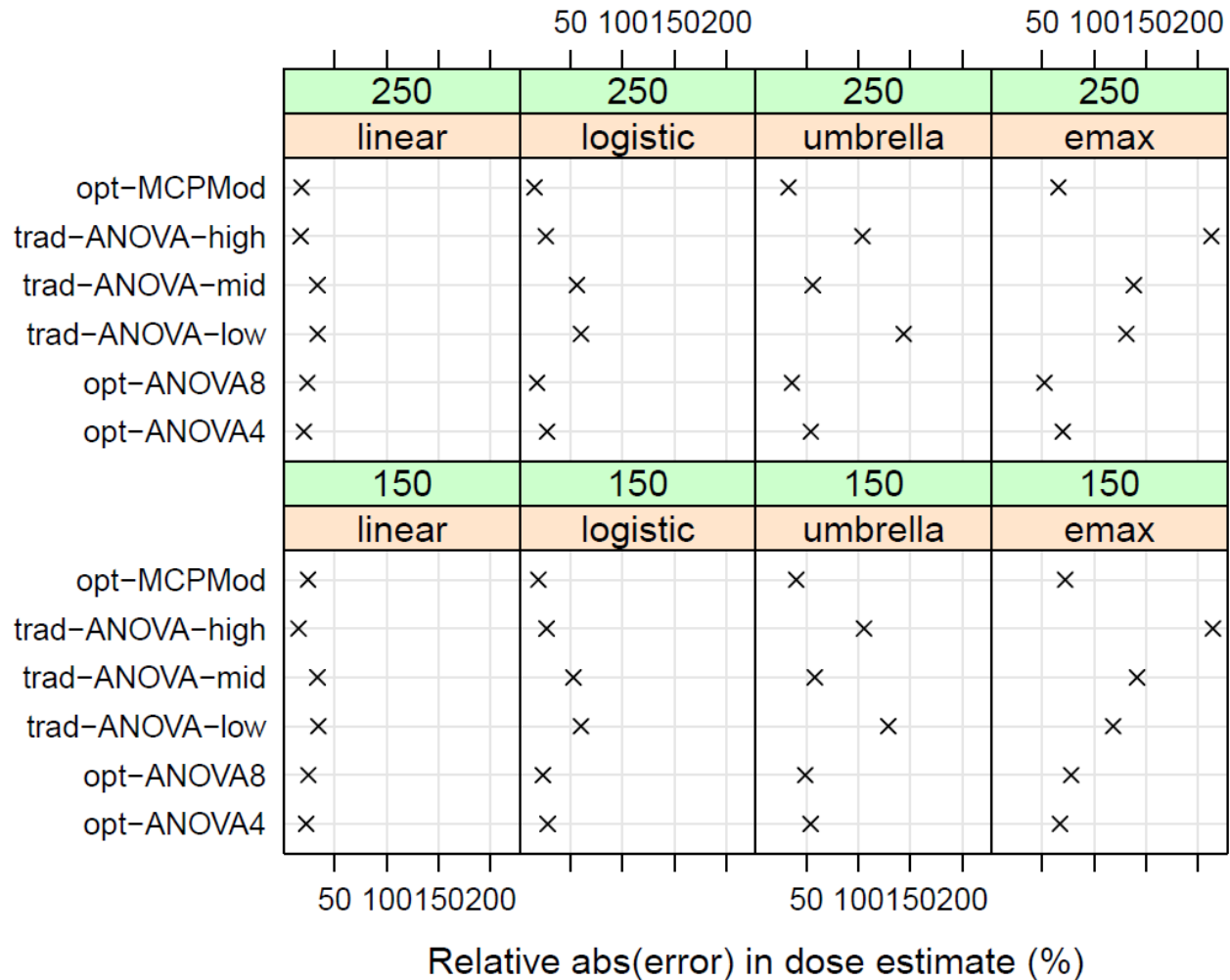
# Simulation Results

## Relative Bias in dose estimate



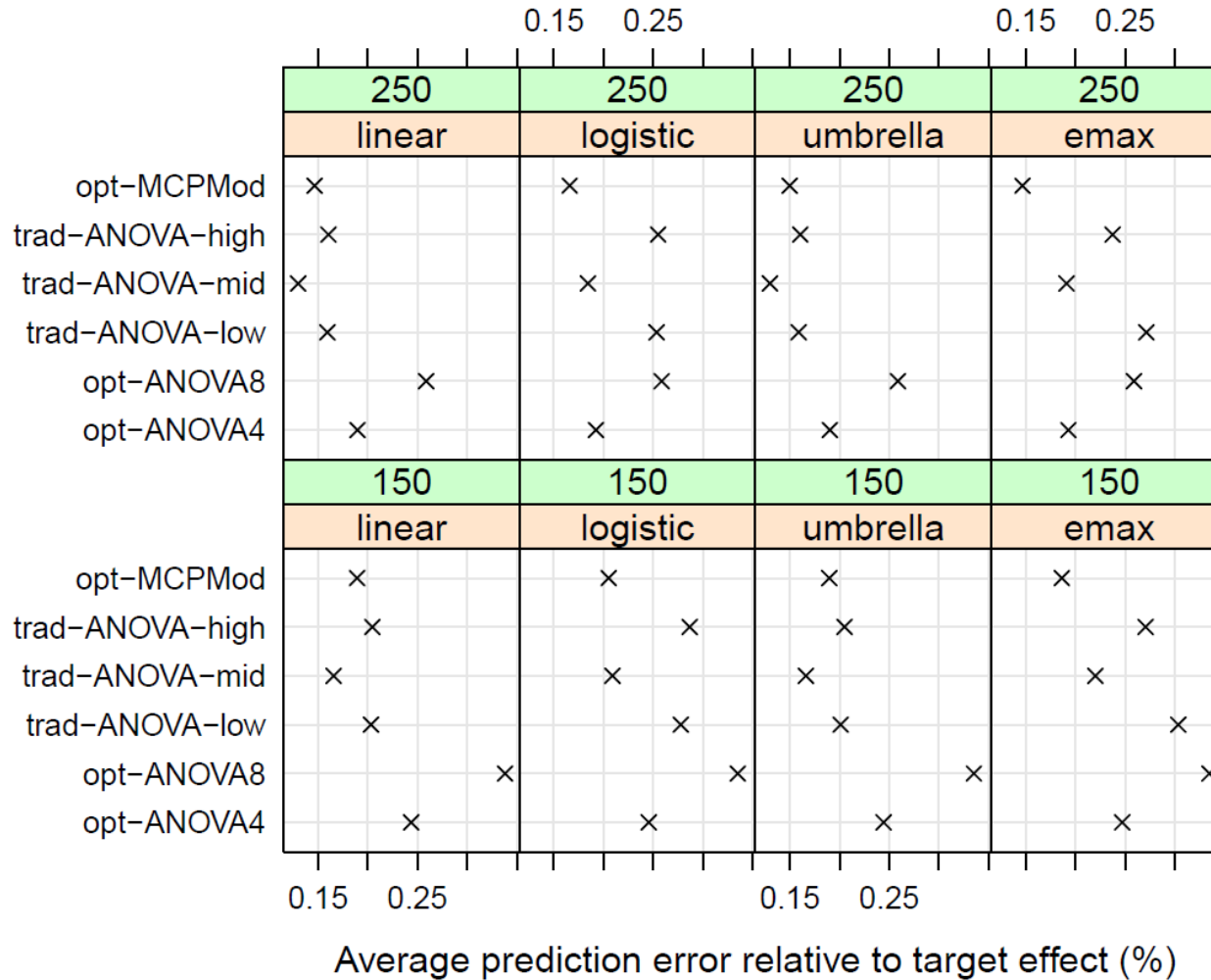
# Simulation Results

## Relative absolute error in dose estimate



# Simulation Results

*Average prediction error in estimating the dose-response function*



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# Case Example

# Example

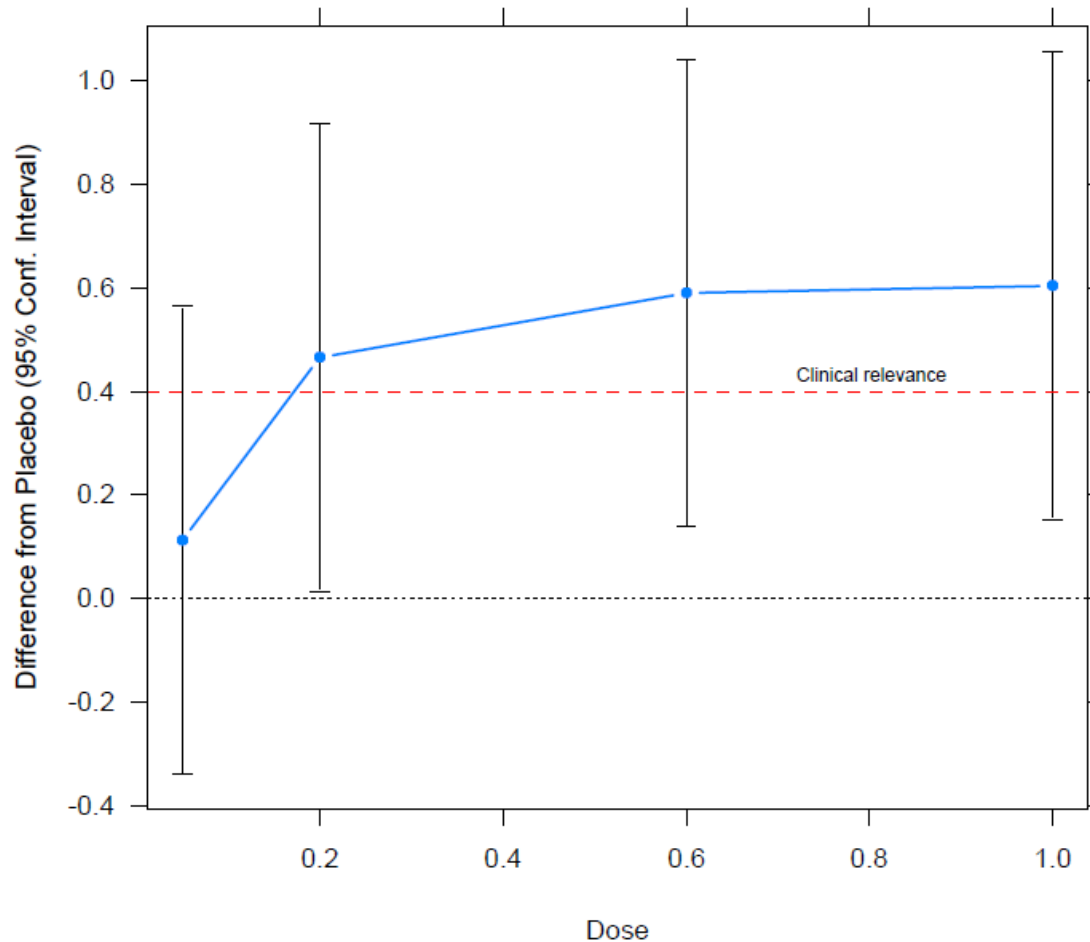
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- Randomized, double-blind parallel group Ph II trial with 100 patients equally allocated to placebo or one of four active doses: 0.05, 0.2, 0.6, or 1
- Normally distributed, homoscedastic primary endpoint
- Planned analysis: Fixed sequence test that preserves type I error at 5% two-sided level
- Conclusion: Top three doses are significantly better than placebo.



# Example

Which dose should be considered the MED?



# Example

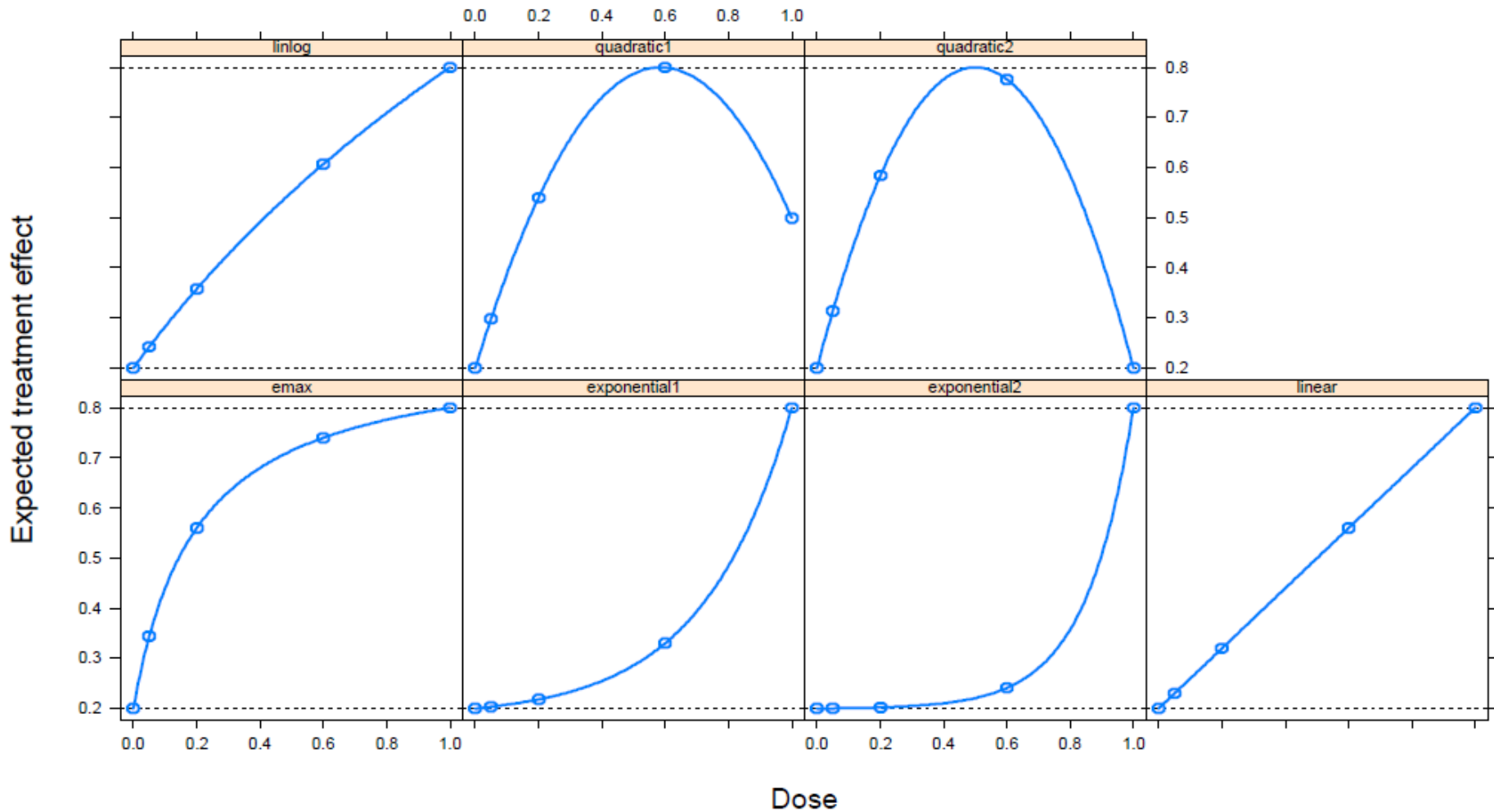
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- First step of MCP-Mod:

Set of **candidate** models

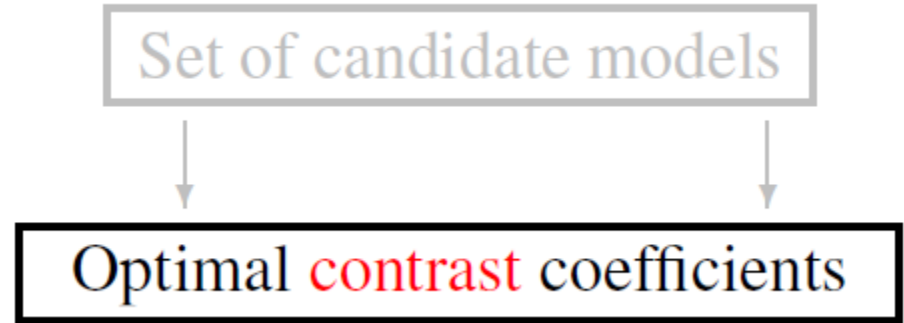
- Propose M dose-response models at planning stage to describe potential outcomes
- Model uncertainty directly acknowledged
- Requires strong collaboration with clinical team
  - Input based on available information (PK data, historical data)

# Example



# Example

- Second step of MCP-Mod:



- Each model will be tested using a contrast test with optimally chosen weights
- For each dose response model, contrast weights are chosen to maximize power in detecting that model if it is true

# Example

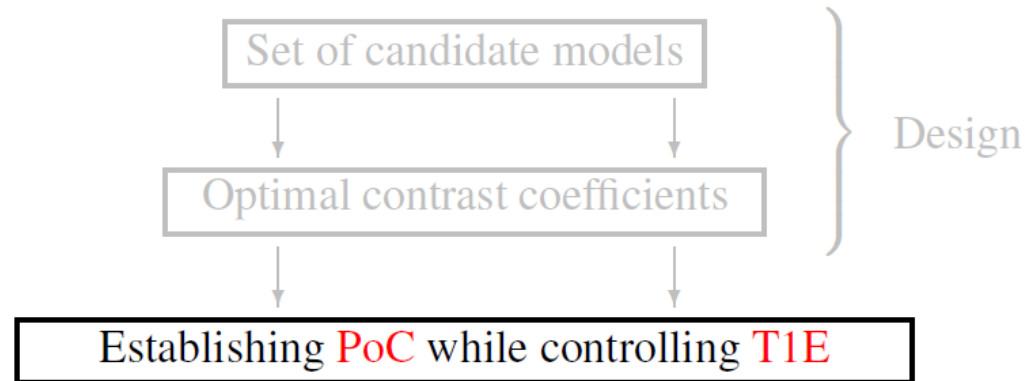
Optimal contrasts:

## Candidate Models

Dose	Linear	E <sub>max</sub>	Linlog	Exp 1	Exp 2	Quad 1	Quad 2
0	-0.44	-0.64	-0.47	-0.29	-0.24	-0.57	-0.42
0.05	-0.38	-0.36	-0.39	-0.29	-0.24	-0.36	-0.20
0.2	-0.20	0.06	-0.16	-0.26	-0.24	0.16	0.33
0.6	0.27	0.41	0.32	-0.04	-0.17	0.71	0.71
1	0.74	0.53	0.70	0.87	0.89	0.07	-0.42

# Example

- Third step of MCP-Mod:



- Each model will be tested using a contrast test with optimally chosen weights
- For each dose response model, contrast weights are chosen to maximize power in detecting that model if it is true

# Example

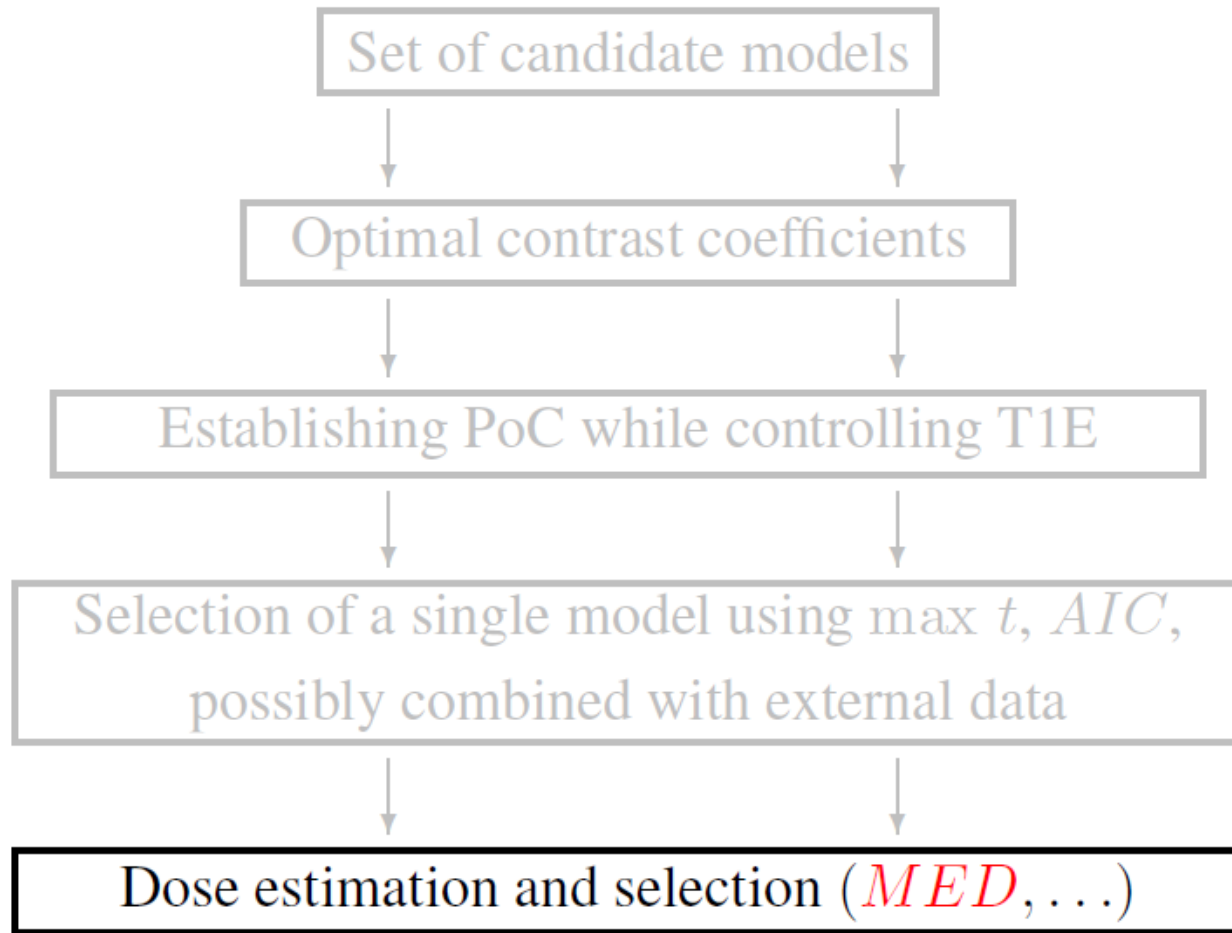
- Significant result is established if the maximum contrast test statistics (across all models) is larger than the critical value, i.e.
  - $\max T_m > \text{crit}_{1-\alpha}$
- All models with  $T_m > \text{crit}_{1-\alpha}$  are kept for possible use in dose-response modeling
- If  $\max T_m < \text{crit}_{1-\alpha}$  no significant dose-response
- Here  $\text{crit}_{0.95} = 2.15$

# Example

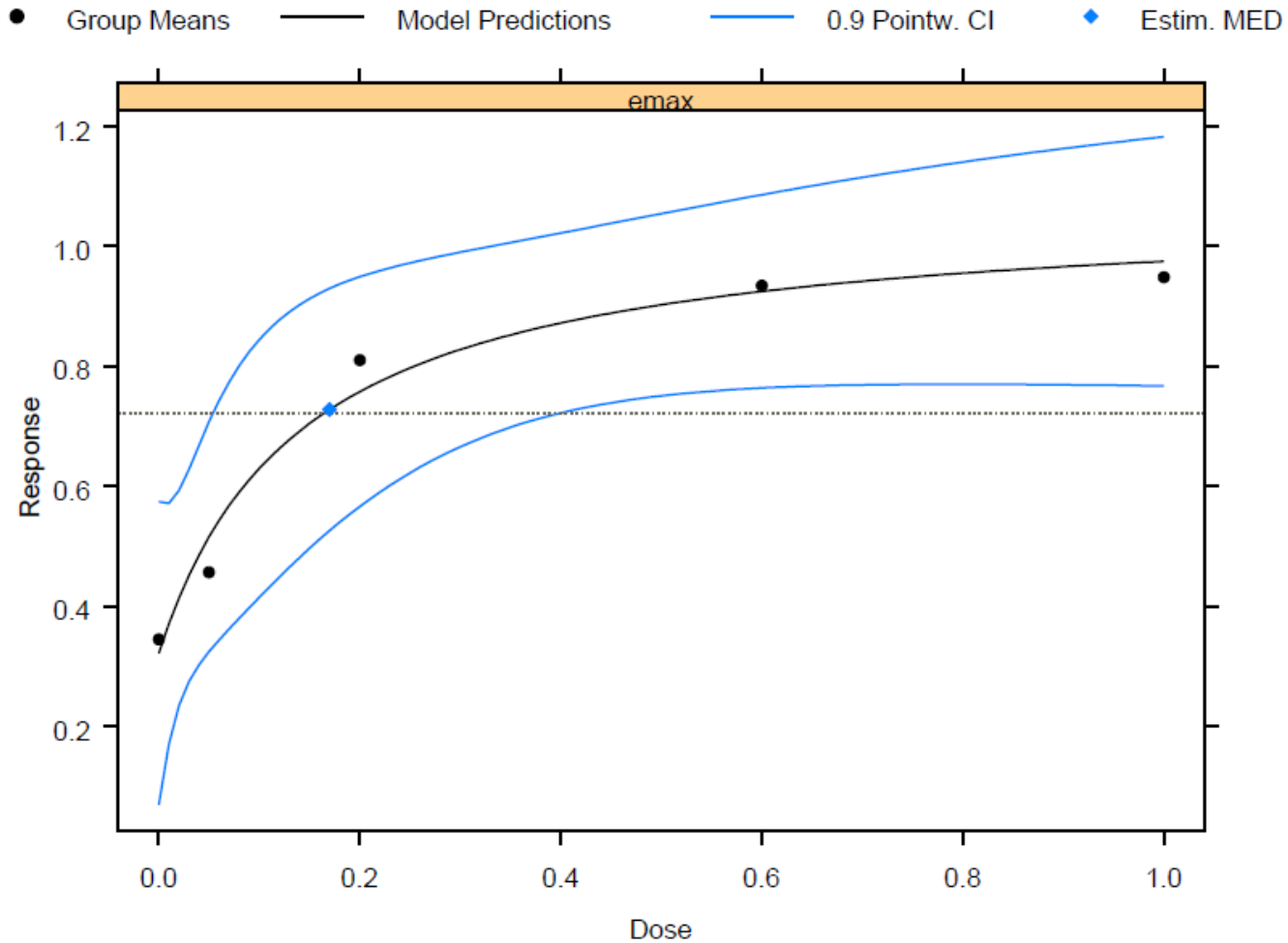
Contrast	est.	s.e.	$t$ -value ( $T_m$ )	$P$ -value	adj. $P$ -value
E <sub>max</sub>	0.55	0.159	3.46	0.0004	0.001
Linlog	0.49	0.159	3.11	0.0012	0.004
Quad 1	0.49	0.159	3.10	0.0013	0.004
Linear	0.47	0.159	2.97	0.0019	0.006
Exp 1	0.35	0.159	2.22	0.0145	0.044
Exp 2	0.30	0.159	1.90	0.0304	0.086
Quad 2	0.29	0.159	1.85	0.0337	0.094



# Example



# Example



# Example

