

Comparison of initial value routines for dose-response models

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Motivation

- Dose-response models are routinely used in different areas and often run in an automated manner.
- Dose-response relations are often described by sigmoid models
- The non-linearity of the sigmoid model requires an iterative parameter estimation procedure.
- The convergence behavior of the iteration procedure depends on the choice of starting values.

Dose-response model

$$f(x) = y = \text{min.resp} + (\text{max.resp} - \text{min.resp}) \cdot g(x, (b, ED_{50}))$$

- for the basic types of sigmoid models (e.g. 4 parameter logistic, Log-normal, Weibull)

$$g(x, (b, ED_{50})) = \tilde{g}(b \cdot \log(x) - b \cdot \log(ED_{50})).$$

y : measured response

x : corresponding dose

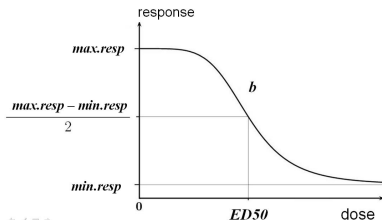
min.resp : lower horizontal asymptote

max.resp : upper horizontal asymptote

$g()$: sigmoid curve

ED_{50} : dose of the middle response

b : slope related



e.g., 4 parameter logistic model: $y = \text{min.resp} + \frac{(\text{max.resp} - \text{min.resp})}{1 + \left(\frac{x}{ED_{50}}\right)^b}$

Starting value routines

- Regression approach (Ritz and Streibig 2005)
 - Normolle (1993)
 - New idea
-
- For all routines, the starting values for the asymptotes are:

$$\begin{aligned} \mathit{max.resp}_{st} &= \mathit{max}_i(y_i) + 0.01 \cdot \left(\mathit{max}_i(y_i) - \mathit{min}_i(y_i) \right), \\ \mathit{min.resp}_{st} &= \mathit{min}_i(y_i) - 0.01 \cdot \left(\mathit{max}_i(y_i) - \mathit{min}_i(y_i) \right), \\ & i = 1, \dots, n \end{aligned}$$

Starting value routines - Regression approach

- Transform the model to get an equation with a linear term in $\log(x)$ on one side.

$$y = \text{min.resp}_{st} + (\text{max.resp} - \text{min.resp}) \cdot \tilde{g}(b \cdot \log(x) - b \cdot \log(ED_{50})).$$

$$\tilde{g}^{-1} \left(\frac{y - \text{min.resp}}{\text{max.resp} - \text{min.resp}} \right) = b \cdot \log(x) - b \cdot \log(ED_{50}).$$

- Insert starting values for min.resp and max.resp .
- Parameter estimation of the regression model.

Starting value routines - Normolle

- Starting value for ED_{50} : Middle of dose groups.

$$ED_{50\ st} = \min_i(x_i) + 0.5(\max_i(x_i) - \min_i(x_i)).$$

- Solve the model equation by b and calculate b_i for each observation.

$$b_i = \frac{\tilde{g}^{-1}\left(\frac{y - \min.\text{resp}_{st}}{\max.\text{resp}_{st} - \min.\text{resp}_{st}}\right)}{\log(x_i) - \log(ED_{50\ st})};$$

$$b_{st} = \text{median}(b_1, \dots, b_n); \quad i = 1, \dots, n \quad \text{and}$$

$$ED_{50\ i} = \exp\left(\frac{b \log(x) - \tilde{g}^{-1}\left(\frac{y - \min.\text{resp}_{st}}{\max.\text{resp}_{st} - \min.\text{resp}_{st}}\right)}{b_{st}}\right);$$

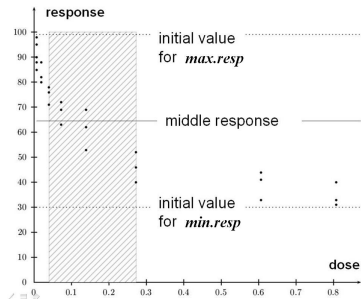
$$ED_{50\ st} = \text{median}(ED_{50\ 1}, \dots, ED_{50\ n}); \quad i = 1, \dots, n$$

Starting value routines - New approach

Starting value for ED_{50}

Dose interval für the starting value of ED_{50}

- Middle response:
$$mr = \frac{\text{max.resp} - \text{min.resp}}{2}$$
- Dose interval = [biggest dose with all observations above mr ; smallest dose with all observations below mr].
- $ED_{50\ st}$: weighted mean of the observations of the interval. Weight depends on the distance to the middle response.



Starting value routines - New approach

Starting value for b

- the first derivative at $x = ED_{50}$ is used for the starting value for b .

$$y' = f'(x) = (max.resp - min.resp) \cdot g'(x, (b, ED_{50}))$$

- Solve $f'(x)$ by b .
- $x = ED_{50}$; $f'(ED_{50}) = \text{slope of the lin. reg. of the dose interval}$

e.g., 4 parameter logistic model:

$$f'(ED_{50}) = \frac{-(max.resp - min.resp) \cdot b}{4 \cdot ED_{50}}$$
$$b = \frac{-4 \cdot ED_{50\ st} \cdot f'(ED_{50\ st})}{max.resp_{st} - min.resp_{st}}$$

Simulation

- Model: four-parameter logistic regression model by varying the parameters ED_{50} and b and the variance of the simulated data and $min.resp = 10$ and $max.resp = 200$.

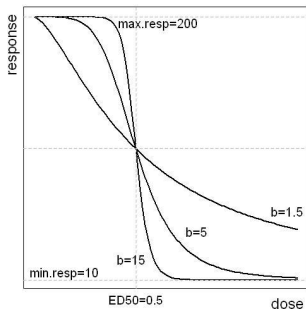
$$x = \{0.0, 0.025, 0.05, 0.1, 0.2, 0.4, 0.8\}$$

$$ED_{50} \in \{0.01, 0.09, 0.5, 1\}$$

$$b \in \{0.5, 1.5, 5, 15\}$$

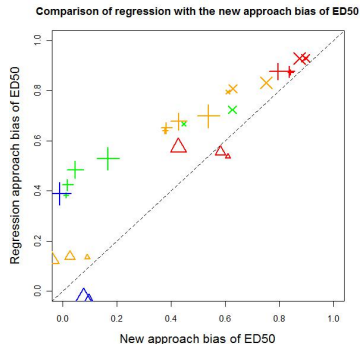
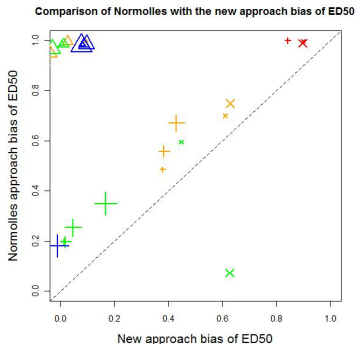
$$\sigma \in \{5, 10, 20, 30\}$$

- $N = 10000$ runs



Results

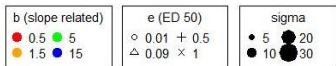
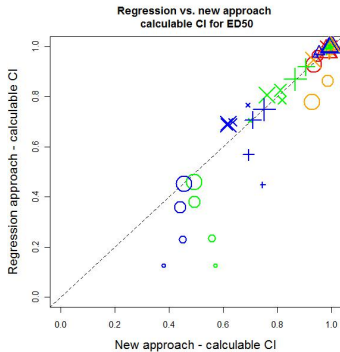
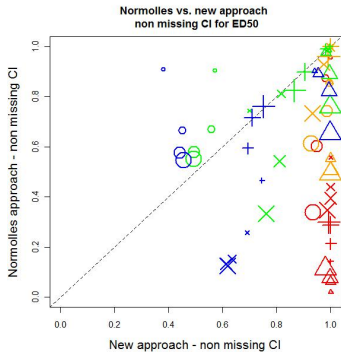
- Bias of ED_{50} .
- Comparison of Normolle vs. New and regression vs. New.



b (slope related)	e (ED_{50})	sigma
● 0.5	○ 0.01 + 0.5	● 5
● 5	△ 0.09 × 1	● 20
● 1.5		● 10
● 15		● 30

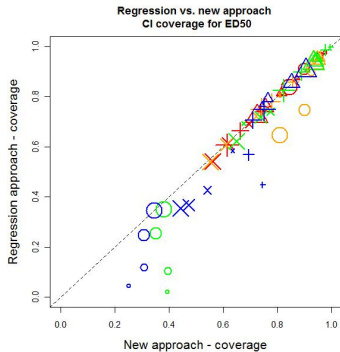
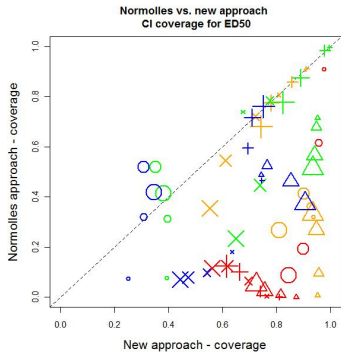
Results

- Non missing confidence intervals for ED_{50} .
- Comparison of Normolle vs. New and regression vs. New.



Results

- Coverage of the confidence interval of ED_{50} .
- Comparison of Normolles vs. New and regression vs. New.



b (slope related)	e (ED_{50})	sigma
● 0.5	○ 0.01 + 0.5	● 5
● 5	△ 0.09 × 1	● 10
● 1.5		● 20
● 15		● 30

Conclusion

In general

- All starting value routines perform best for the ideal design (i.e. curve with moderate slope and ED_{50} in the middle of dose range).
- Variation in the data has very low impact on iteration performance compared to the design.

Routine comparison

- The New approach is superior to Normolle in all cases except for the ideal design.
- The New approach performs better or at least not worse than regression approach in all designs.
- The New approach is especially preferable if reduction of bias is most relevant or if ED_{50} is suspected at either end of the dose range.

References

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