

# The algae growth inhibition test – robust initial values for parameter estimation

Anke Schulz

Bayer Schering Pharma AG  
Global Drug Discovery Statistics  
[anke.schulz@bayerhealthcare.com](mailto:anke.schulz@bayerhealthcare.com)

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# Overview

- 1 The algae growth inhibition test
- 2 Four-parameter logistic regression model
- 3 Initial values for iterative estimation procedure
- 4 Application

## 1 The algae growth inhibition test

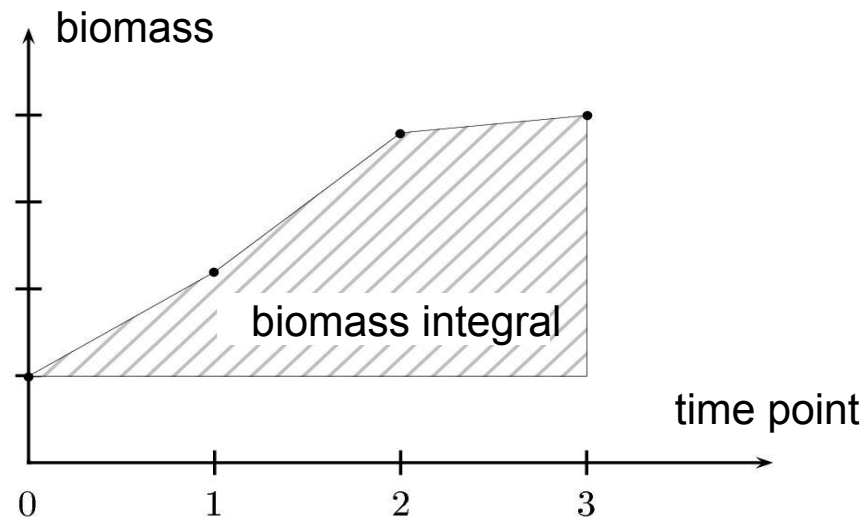
# Design

- Biological test to investigate the growth inhibition of algae under addition of a substance.
- Test routinely used in toxicology.
- Design:
  - One control group (usually six replicates).
  - Several dose groups of the same substance (usually five dose groups in dilution series with three replicates).
  - Equidistant time points starting at 0h (usually 0, 24, 48 and 72).

## 1 The algae growth inhibition test

# Biomass integral

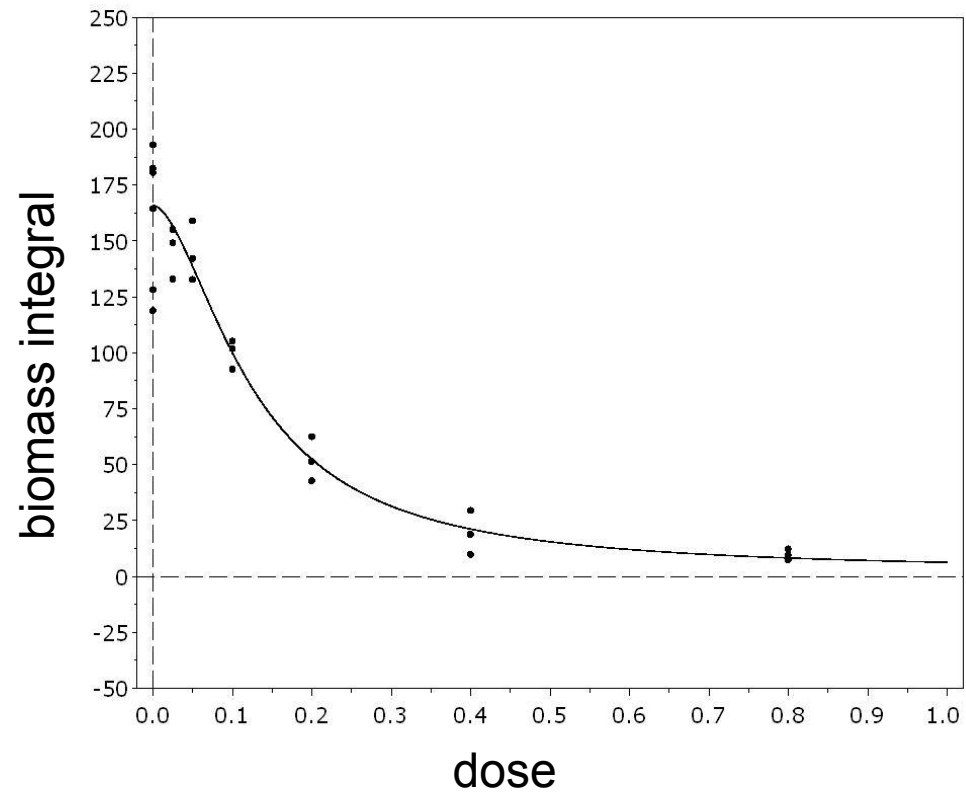
- Algae biomass is a function over time for each replicate.
- The growth is measured by the integral of the biomass.  
→ This integral will be the variable of interest.



# 1 The algae growth inhibition test

## Data

- Integral of biomass decreases by dose (in dependence on toxicity).



## 2 Four-parameter logistic regression model

# Logistic regression model

- Sigmoid model:
- Four-parameter logistic regression model:

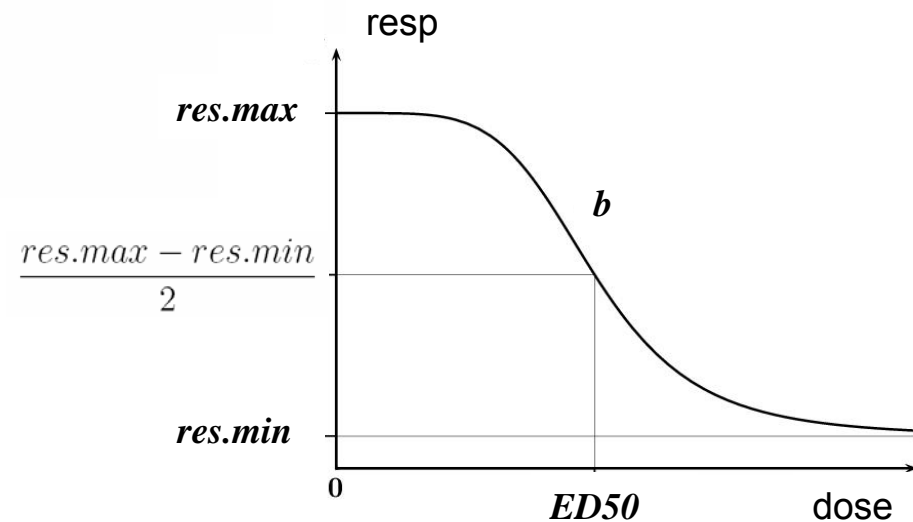
$$\text{resp}(\text{dose}) = \text{res.min} + \frac{\text{res.max} - \text{res.min}}{1 + \left(\frac{\text{dose}}{\text{ED50}}\right)^b} + \epsilon;$$

$$\epsilon \sim N(0, \sigma^2)$$

$$\text{dose} \in \mathbb{R}_0^+$$

$$\text{res.max}, \text{res.min}, b \in \mathbb{R}$$

$$\text{ED50} \in \mathbb{R}^+$$



## 2 Four-parameter logistic regression model

# Iterative parameter estimation

- The non-linearity of the model requires an iterative parameter estimation procedure and initial values.
- The convergence behavior depends on the choice of initial values.
- For **small variation**, procedures to obtain initial values developed by (Normolle, 1993) and (Ritz and Streibig, 2005) work very well.
- For **large variation** as found in the data of the algae test:
  - Robust M-estimators proposed by Huber (1964).
  - Our new robust procedure (2006) including an automated determination of initial values.

### 3 Initial values for iterative estimation procedure

## Parameters $res.min$ and $res.max$

- The initial values for the parameters  $res.min$  (minimum) and  $res.max$  (maximum) will be estimated as described by Ritz and Streibig (2005):

$$res.min_{ini} = \text{minimum observation}$$
$$res.max_{ini} = \text{maximum observation}$$

- However, initial values for  $res.min_{ini}$  and  $res.max_{ini}$  have only a limited impact on the convergence.



### 3 Initial values for iterative estimation procedure

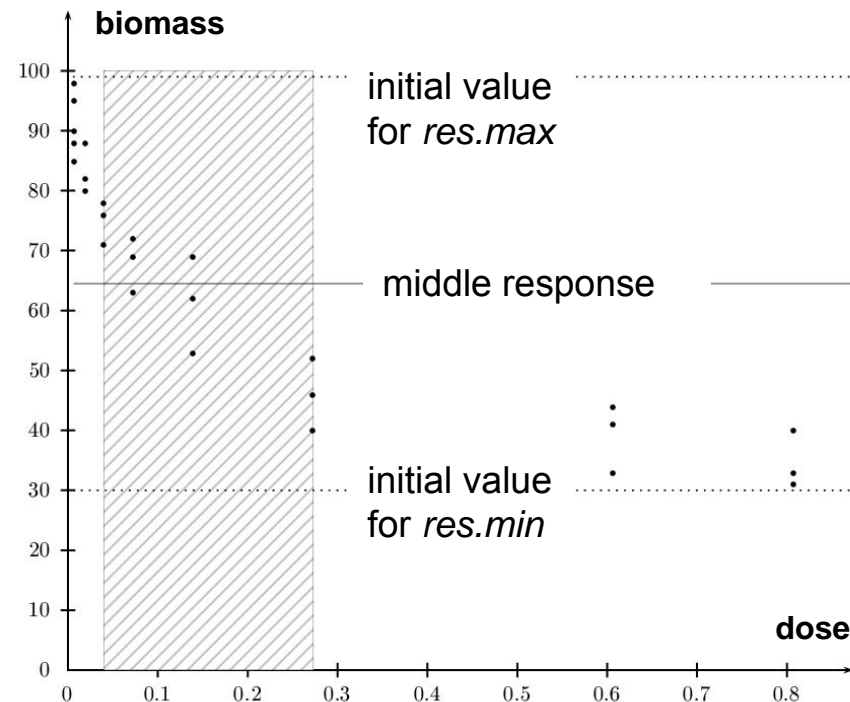
## Parameters $b$ and $ED50$

- Estimators for parameters  $b$  (slope) and  $ED50$  do have a functional relationship.
- If initial value for  $ED50$  is inaccurate, i.e.  $|ED50_{ini} - ED50| \gg 0$ , parameter estimates for  $b$  and  $ED50$  are unreliable.  
Then, the sign of the slope changes.
- Consequently, a reasonable dose interval for  $ED50_{ini}$  has to be found.

### 3 Initial values for iterative estimation procedure

## Parameters $b$ and $ED50$

- Middle response =  $\frac{res.max - res.min}{2}$ .
- Dose interval =  
[largest dose with all observations above the middle response ; smallest dose with all observations below the middle response].
- $ED50_{ini}$  : weighted mean of the dose of the observations of the interval.
  - Weight depends on the distance to the middle response.



### 3 Initial values for iterative estimation procedure

## Parameters $b$ and $ED50$

- Based on the initial values for  $res.max$ ,  $res.min$  and  $ED50$ , initial value for parameter  $b$  is calculated for each observation (Normolle, 1993):

$$b_j = \frac{\log\left(\frac{res.max_{ini} - resp_j}{resp_j - res.min_{ini}}\right)}{\log\left(\frac{dose_j}{ED50_{ini}}\right)}; \quad b_{ini} = \text{median}(b_1, \dots, b_n)$$

- For a first quality check of initial values, the sign of  $b_{ini}$  is compared to the sign of the slope parameter of a simple linear regression (negative sign of  $b$  corresponds to a positive slope).

### 3 Initial values for iterative estimation procedure

## Parameters $b$ and $ED50$

- If the sign of  $b_{ini}$  does not match the direction of the slope:
  - The sign of  $b_{ini}$  is changed.
  - The parameter  $ED50_{ini}$  is calculated for each observation based on the initial values for  $res.max$ ,  $res.min$  and  $b$ .
  - Median of all  $ED50_j$  is the new initial value for  $ED50$ .

$$ED50_j = dose_j \left( \frac{res.max_{ini} - resp_j}{resp_j - res.min_{ini}} \right)^{-\frac{1}{b_{ini}}}$$

$$ED50_{ini} = \text{median}(ED50_1, \dots, ED50_n)$$

- $b_{ini}$  is calculated again.

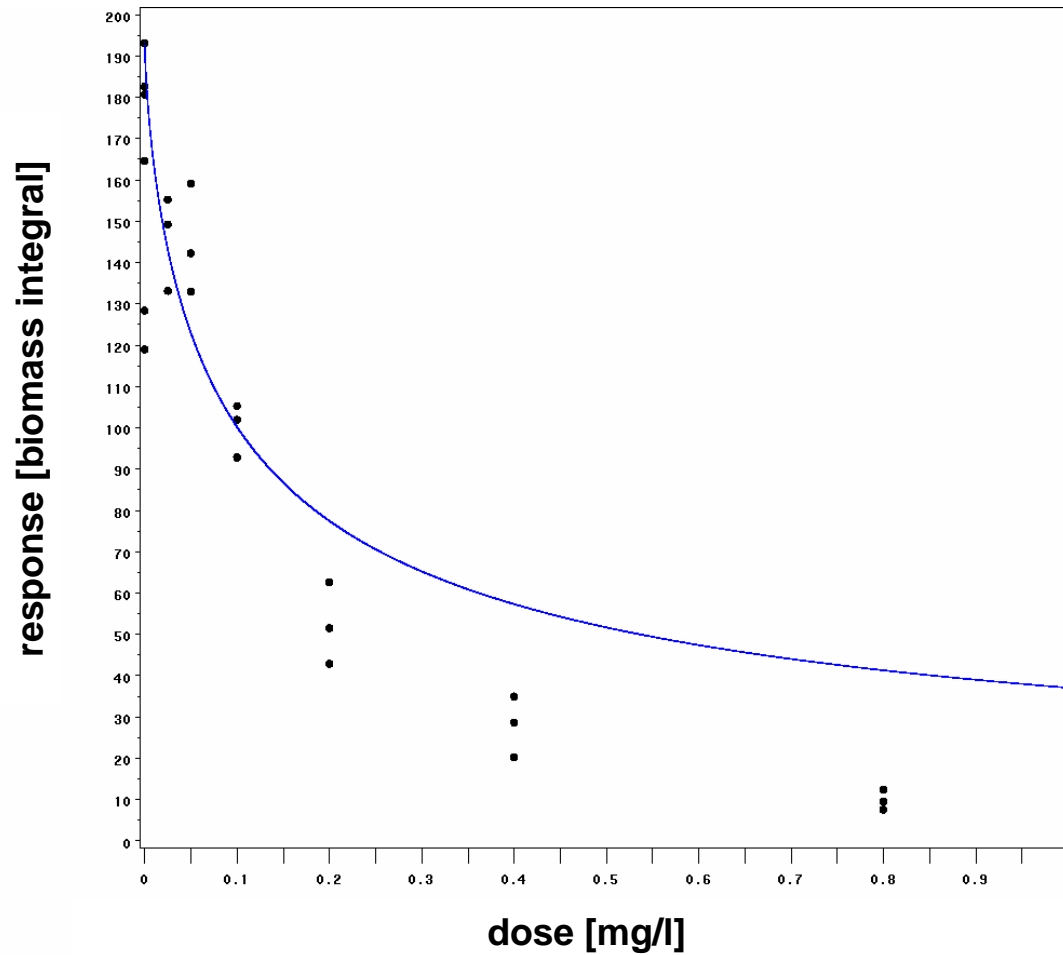
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## 4 Application

# Practical experience

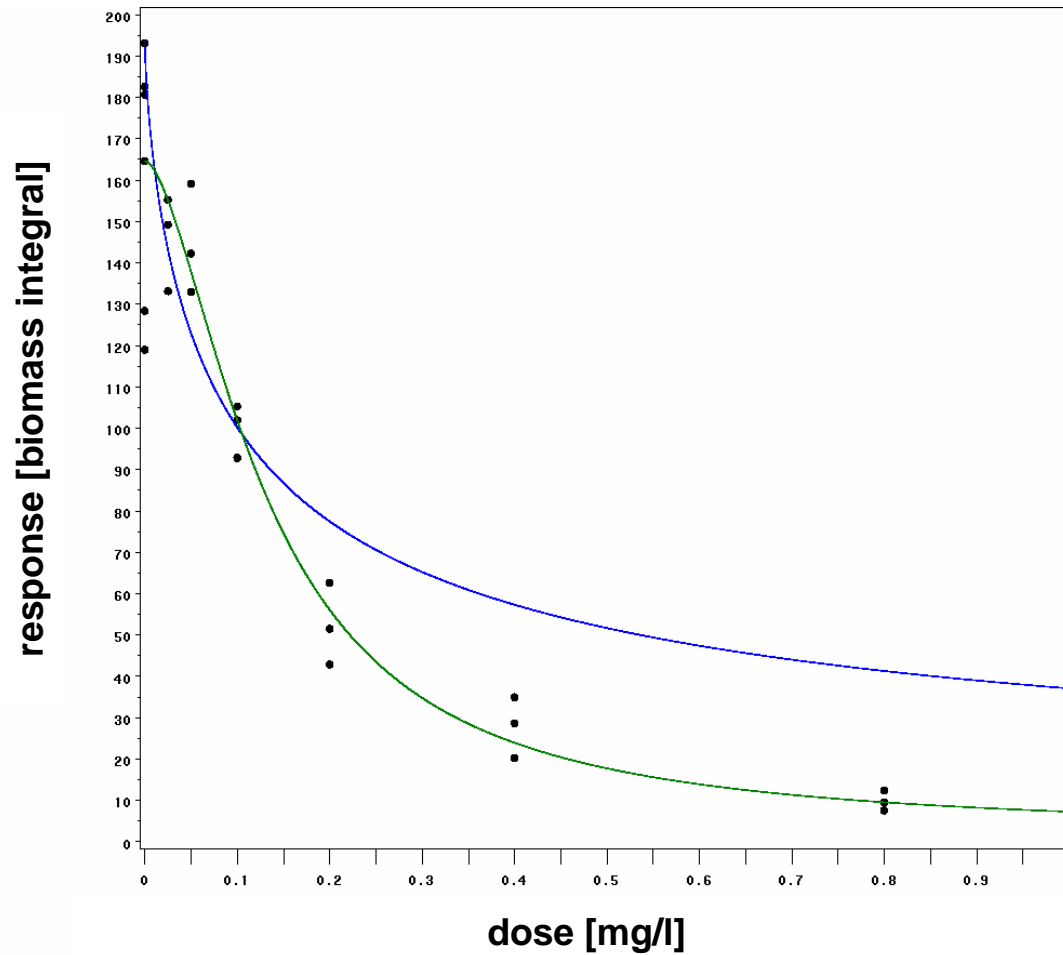
- Our proposed approach is tested on 42 real data sets.
- Common procedures lead to poor results, only 70% of all estimates are reasonable.
- Instead, our initial values yield sensible parameter estimates in all cases.

# 4 Application Example



Blue: Model of initial values

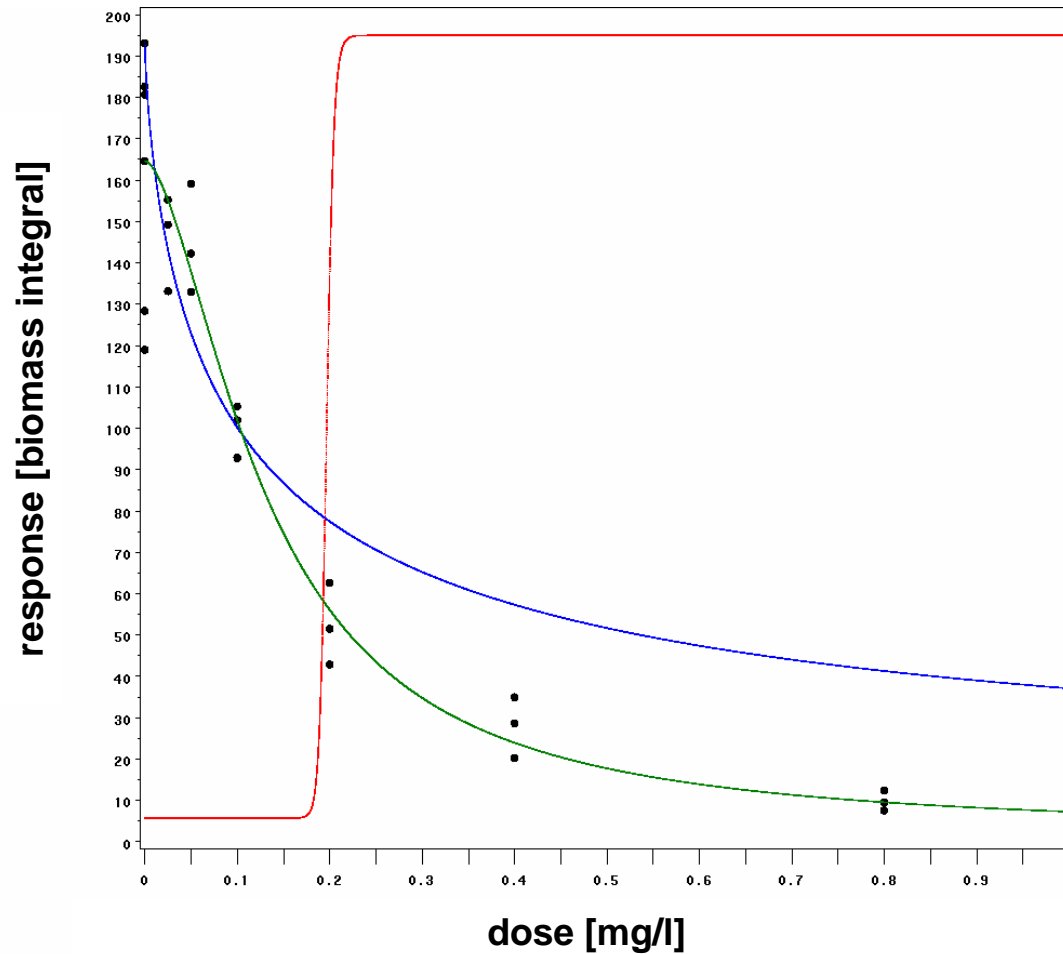
# 4 Application Example



Blue: Model of initial values

Green: Estimated model

# 4 Application Example



Blue: Model of initial values

Green: Estimated model

Red: Model of initial values  
(Normolle, 1993)



# References

- Huber, P.J. (1964). 'Robust estimation of local parameter'. *Annals of Mathematical Statistics* 35: 73-101 .
- Normolle, D.P. (1993). 'An Algorithm for Robust non-linear Analysis of Radioimmunoassays and other Bioassays'. *Statistics in Medicine* 12: 2025-2042
- Ritz, C. & Streibig, J.C. (2005). 'Bioassay Analysis using R'. *Journal of Statistical Software* 12, Issue 5.
- Schulz, A (2006). 'Das vierparametrische logistische Modell und seine Anwendung bei einem ökotoxikologischen Problem'. Master Thesis. Humboldt-Universität zu Berlin, Germany.