



EFSA Guidance for BMD analysis

Fitting Models & Goodness of Fit

1st March 2017

OUTLINE

- **General** principles of model fitting & goodness of fit
- **Continuous** dose-response data
 - Cell proliferation (CP) data
 - Candidate models
 - Illustration
- **Quantal** response data
 - Thyroid epithelial cell vacuolisation (TECV) data
 - Candidate models
 - Illustration

GENERAL PRINCIPLES

- Data vary according to a **distribution**
 - Continuous: log-normal with mean and variance
 - Quantal: binomial with probability parameter
- Model the **mean** parameter as **function of dose**
- **Fit** the model to data and get estimates (ML)
- **Compare** different **candidate models** using a goodness-of-fit criterion (AIC)

CONTINUOUS RESPONSE DATA

Cell Proliferation Example

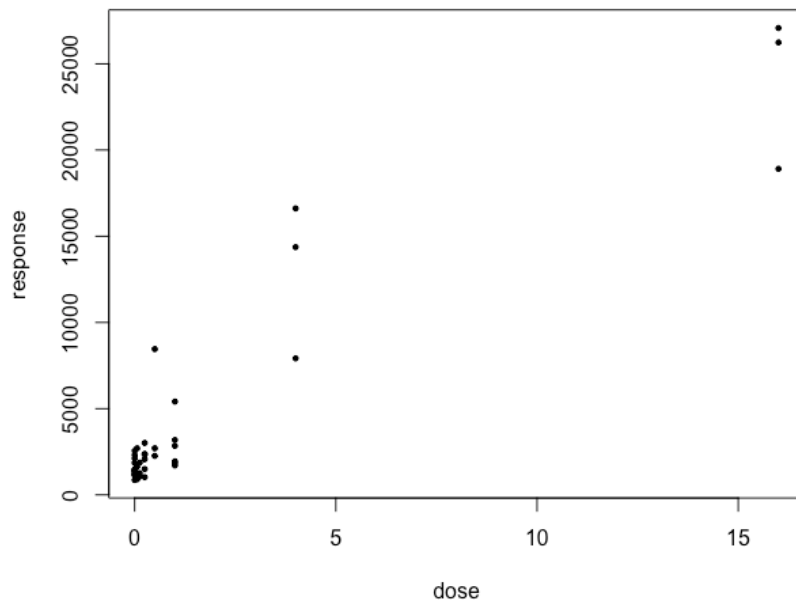
CELL PROLIFERATION

- Local Lymph Node Assay (LLNA) test
- Cell proliferation as an indicator of sensitization
- Related to 15 rubber chemicals, focus on one chemical
- Dose levels and number of observations per level

dose							
0	0.06	0.125	0.25	0.5	1	4	16
12	6	3	6	3	6	3	3

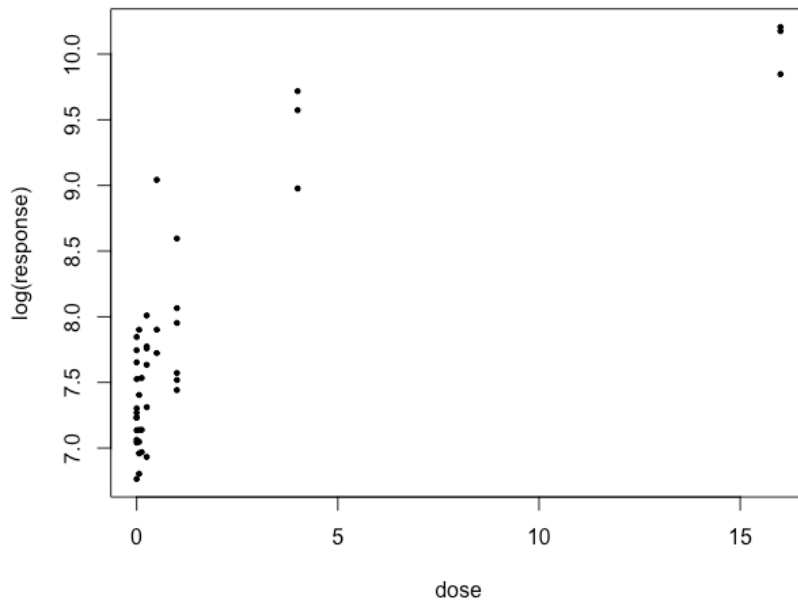
CELL PROLIFERATION

- Data of dose and cell proliferation on original scale



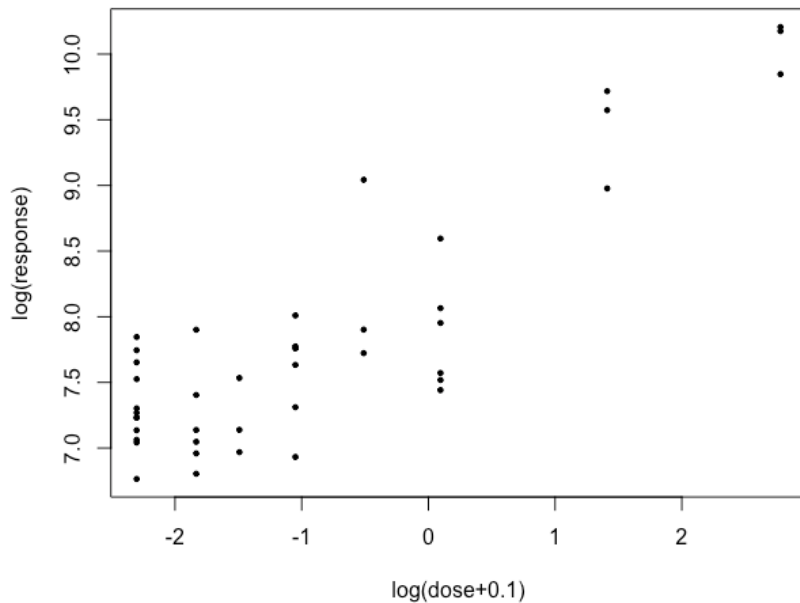
CELL PROLIFERATION

- Data of cell proliferation on log scale



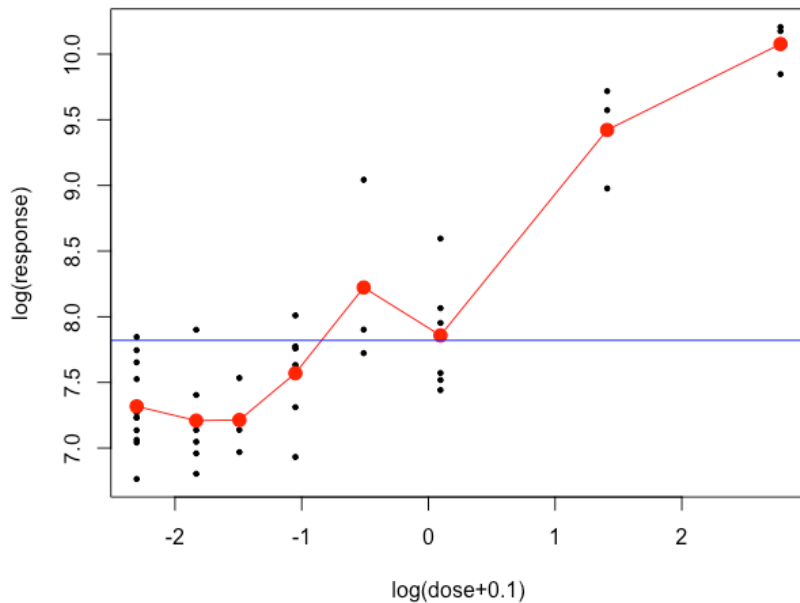
CELL PROLIFERATION

- Data of dose and cell proliferation both on log scale



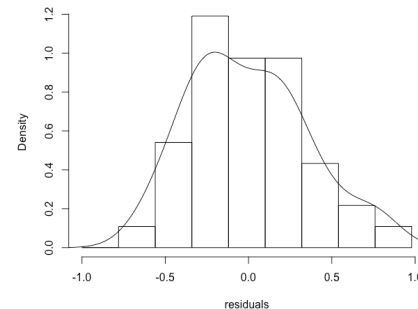
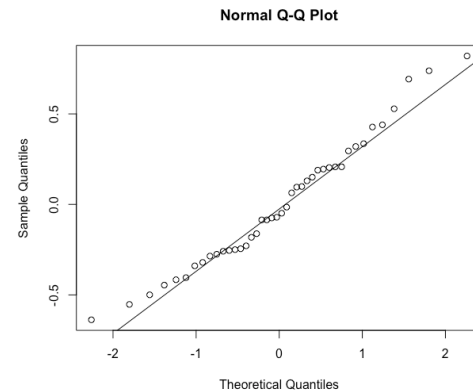
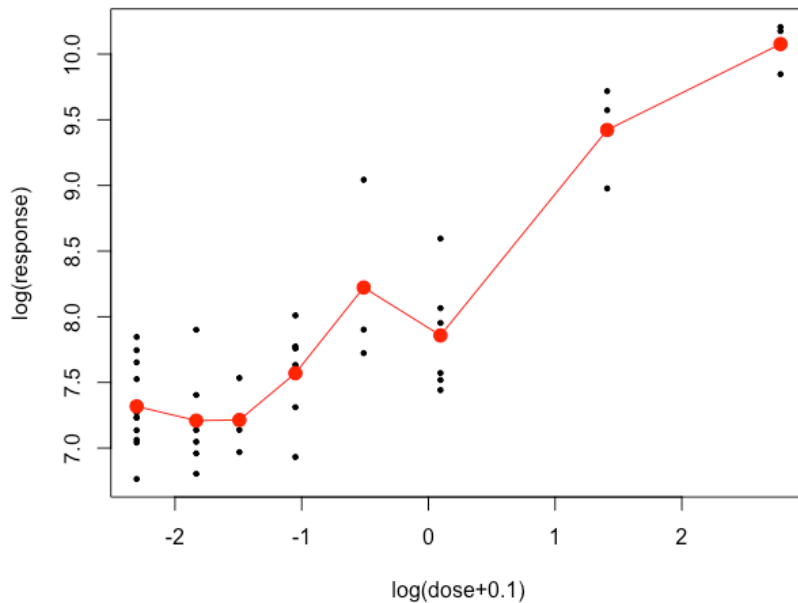
CELL PROLIFERATION

- **Null** model of **no effect**
- **Full** model: each dose level its own mean



CELL PROLIFERATION

- Equal variances: Levene's test ($p=0.89$)
- Normal distribution: Shapiro-Wilk ($p=0.42$)



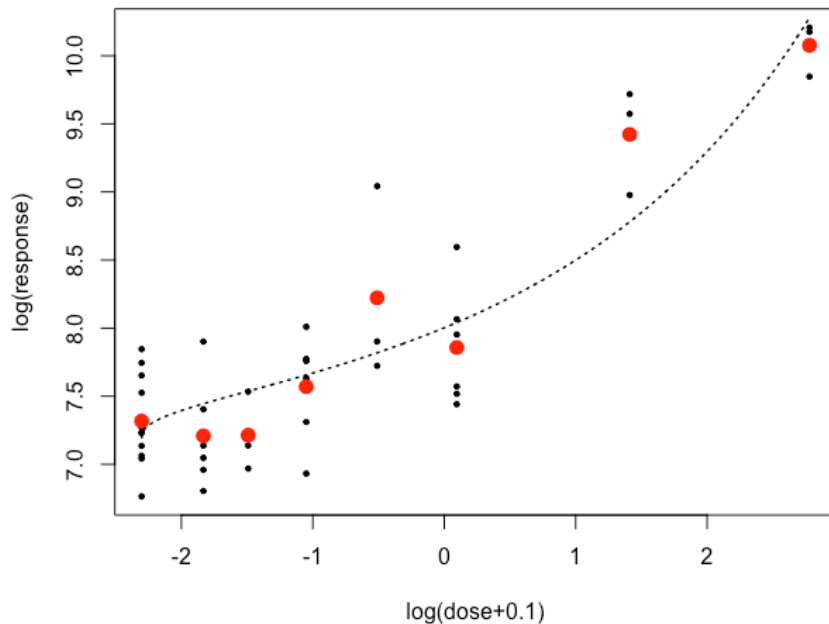
CELL PROLIFERATION

Candidate models for continuous data are models for the **mean** of the normal distribution on log-scale

<u><i>Continuous data</i></u>		
Exponential family		
3-parameter model ⁽ⁱⁱⁱ⁾	3	$y = a \exp(bx^d)$
4-parameter model ^(iv)	4	$y = a [c - (c-1)\exp(-bx^d)]$
Hill family		
3-parameter model ⁽ⁱⁱⁱ⁾	3	$y = a [1 - x^d / (b^d + x^d)]$
4-parameter model ^(iv)	4	$y = a [1 + (c-1)x^d / (b^d + x^d)]$

CELL PROLIFERATION

- 3 parameter Exponential model $y = a \exp(bx^d)$



Coefficients:

	Estimate	Std. Error
a	7.1959136	0.11088539
b	0.1110707	0.02044363
d	0.4206736	0.05800638
sigma	0.4392835	0.04792905

'log Lik.' -25.04554

MAXIMUM LIKELIHOOD ESTIMATION & INFERENCE

$$y = a \exp(bx^d)$$

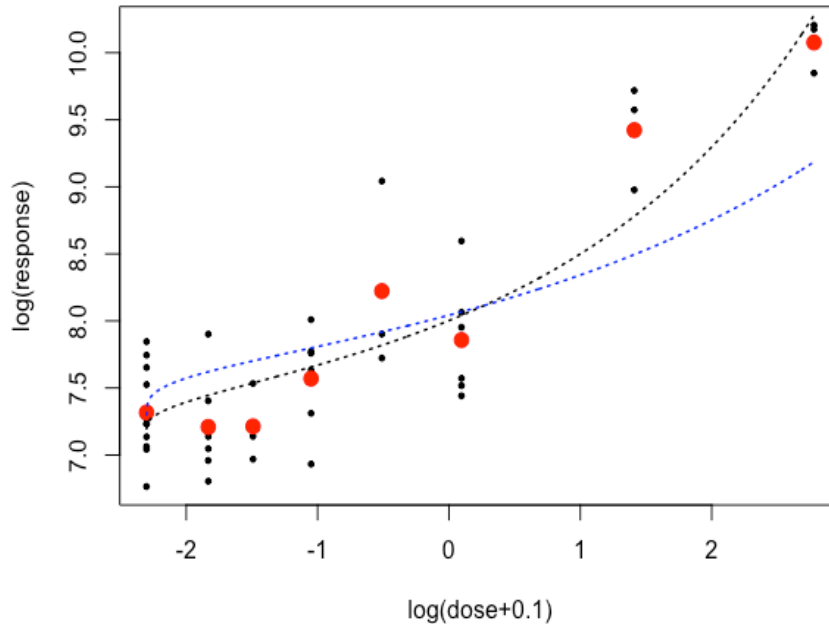
Maximum Likelihood (ML) estimation & inference

- Values for a , b and d that maximize the likelihood
- The likelihood of the data you observed
- Maximize likelihood = maximize log-likelihood

CELL PROLIFERATION

■ Maximum Likelihood (ML) estimation & inference

ML (most likely) values for the parameters



Coefficients:

	Estimate	Std. Error
a	7.1959136	0.11088539
b	0.1110707	0.02044363
d	0.4206736	0.05800638
sigma	0.4392835	0.04792905

'log Lik.' -25.04554

Less likely values for the parameters

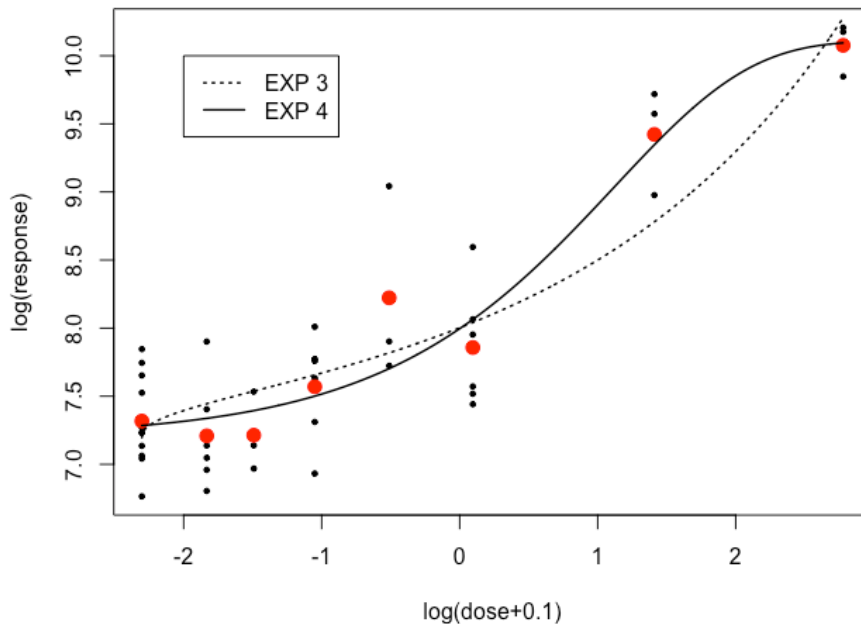
a = 7.3
b = 0.1
d = 0.3
log Lik = -37.099

MAXIMUM LIKELIHOOD ESTIMATION & INFERENCE

- ML is identical to **least squares estimation** in case of the **normal** distribution (possibly on log-scale)
- Difficulty for **non-linear models**: multiple local maxima and **convergence issues**

CELL PROLIFERATION

- 4 parameter Exponential model $y = a [c - (c-1)\exp(-bx^d)]$



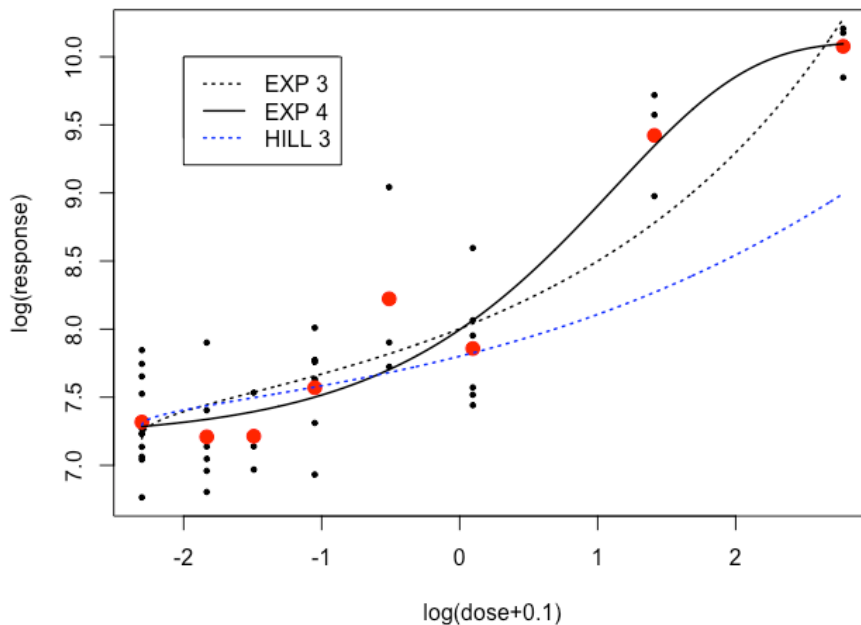
Coefficients:

	Estimate	Std. Error
a	7.2850093	0.09654781
b	0.3227275	0.08541427
c	1.3875325	0.03960234
d	1.0104150	0.22725460
sigma	0.3935590	0.04293928

'log Lik.' -20.42961

CELL PROLIFERATION

- 3 parameter Hill model $y = a [1 - x^d / (b^d + x^d)]$



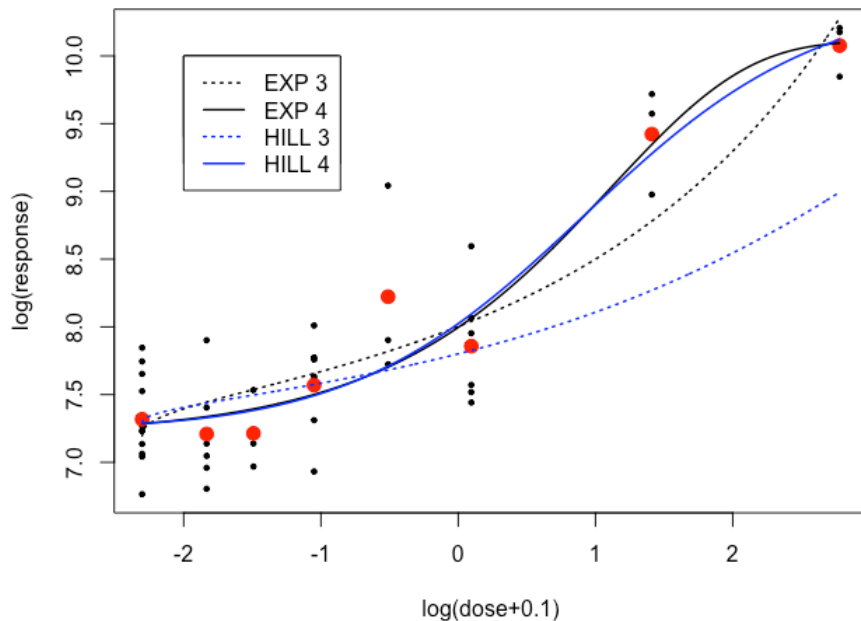
Coefficients:

	Estimate	Std. Error
a	7.2953731	0.11004240
b	183.7769316	57.44395880
d	0.4884650	0.06802027
sigma	0.4643575	0.05335079

'log Lik.' -27.06396

CELL PROLIFERATION

- 4 parameter Hill model $y = a [1 + (c-1)x^d/(b^d + x^d)]$



Coefficients:

	Estimate	Std. Error
a	7.2881474	0.09865037
b	2.5468752	1.00250825
c	1.4354407	0.07563080
d	1.1524649	0.33549528
sigma	0.3963275	0.04324196

'log Lik.' -20.7236

AKAIKE INFORMATION CRITERION

- **AIC** stems from information theory
- Measures the **information lost** by replacing the true unknown data generating model by the model used
- Estimates the **Kullback-Leibler distance**
- The **smaller AIC** the better the model
- Defined as

$$\text{AIC} = -2 \log\text{likelihood} + 2 \text{ \#par}$$

- The lower -2 loglikelihood the closer the fit to the data
- **Penalization** for overfitting, for too complex models
- AIC balances accuracy with complexity

CELL PROLIFERATION

Model	Log-likelihood	# par
Null	-56.006	2
Full	-15.659	9
EXP 3	-25.046	4
EXP 4	-20.430	5
HILL 3	-27.064	4
HILL 4	-20.724	5

par = number of parameters in mean model + variance parameter

CELL PROLIFERATION

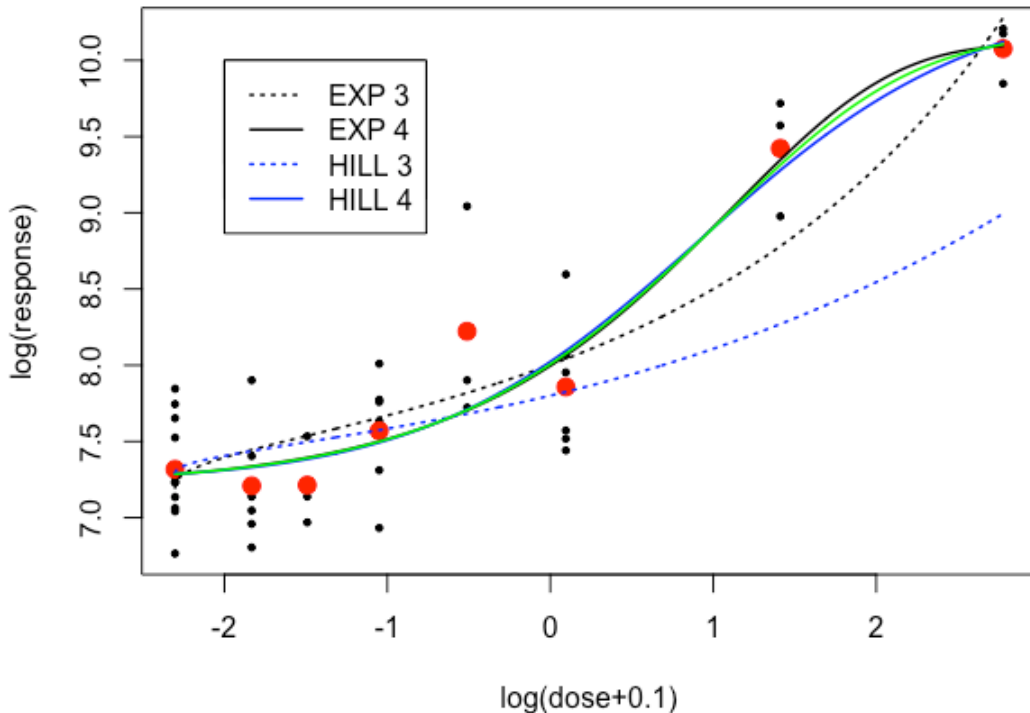
Model	Log-likelihood	# par	AIC
Null	-56.006	2	116.012
Full	-15.659	9	49.319
EXP 3	-25.046	4	58.091
EXP 4	-20.430	5	50.859
HILL 3	-27.064	4	62.128
HILL 4	-20.724	5	51.447

$$AIC_{\text{Min}} = 50.859 < AIC_{\text{FULL}} + 2 = 51.319$$

$$\text{All } AIC < AIC_{\text{NULL}} - 2 = 114.012$$

CELL PROLIFERATION

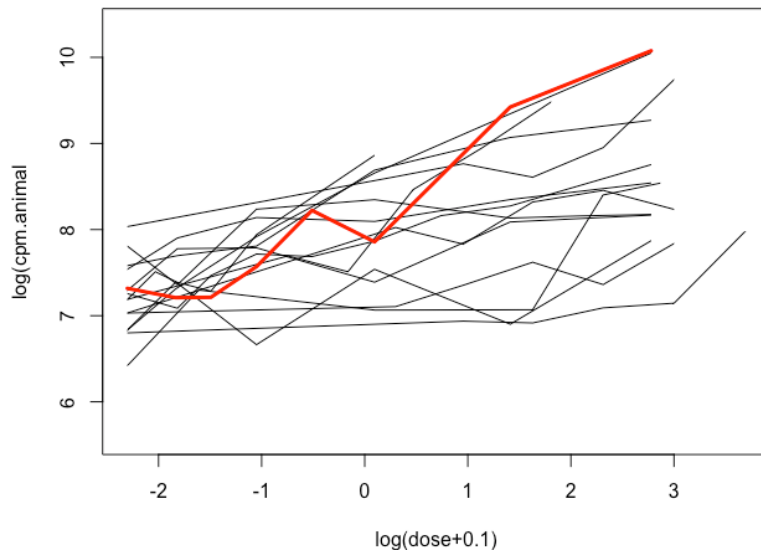
Model averaged fitted curve



Model	AIC	Weight
EXP 3	58.091	0.0056
EXP 4	50.859	0.5693
HILL 3	62.128	0.0007
HILL 4	51.447	0.4243

CELL PROLIFERATION

Just **one compound**, but there are 15 compounds



Approach: include compound as **covariate**, leading to a 4 parameter exponential model, with compound specific values for a and b and common values for c and d

QUANTAL RESPONSE DATA

thyroid epithelial cell vacuolisation

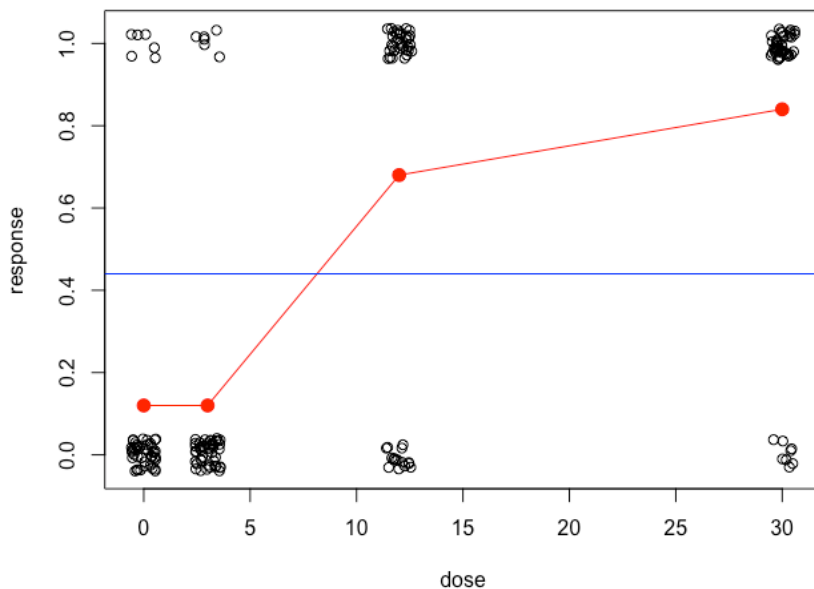
THYROID EPITHELIAL CELL VACUOLISATION

- 2 year study in rats
- Three doses of a substance
- Changes in thyroid epithelial cell vacuolisation

Dose (mg/kg day)	No of animals with thyroid epithelial vacuolisation	No of animals in dose group
0	6	50
3	6	50
12	34	50
30	42	50

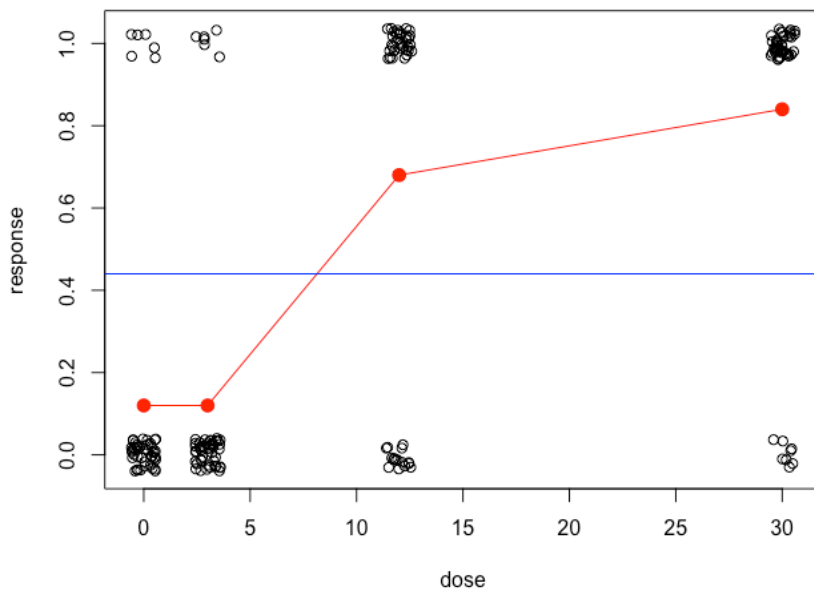
THYROID EPITHELIAL CELL VACUOLISATION

- **Null** model of **no effect**
- **Full** model: each dose level its own probability



THYROID EPITHELIAL CELL VACUOLISATION

- **Binomial** distribution essentially the only choice
- Extensions: beta-binomial model & hierarchical data models



THYROID EPITHELIAL CELL VACUOLISATION

Candidate models for quantal data

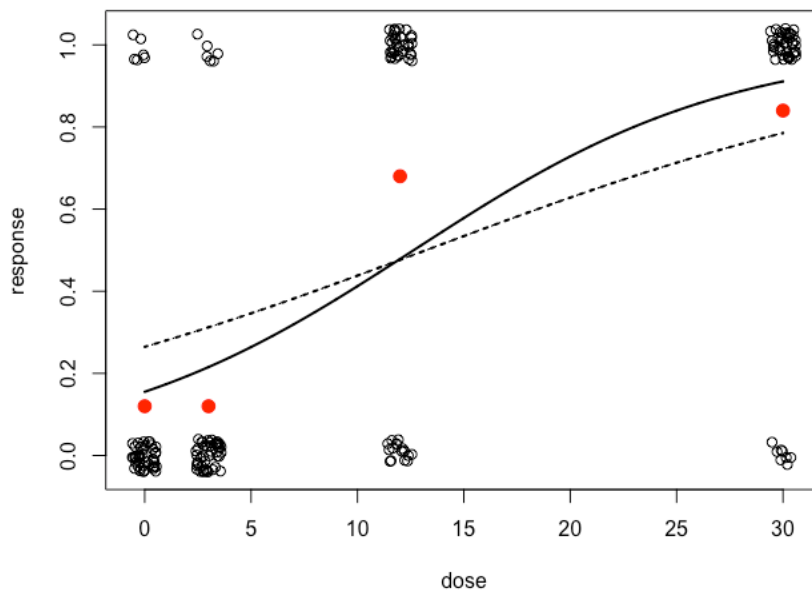
Quantal data

Logistic	2	$y = 1/(1 + \exp(-a - bx))$
Probit	2	$y = \text{CumNorm}(a + bx)$
Log-logistic	3	$y = a + (1 - a)/(1 + \exp(-\log(x/b)/c))$
Log-probit	3	$y = a + (1 - a) \text{CumNorm}(\log(x/b)/c)$
Weibull	3	$y = a + (1 - a) \exp((x/b)^c)$
Gamma	3	$y = a + (1 - a) \text{CumGam}(bx^c)$
LMS (two-stage) model	3	$y = a + (1 - a)(1 - \exp(-bx - cx^2))$

- ```
'log Lik.' -97.23388
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# THYROID EPITHELIAL CELL VACUOLISATION

- Probit model  $y = \text{CumNorm}(a + bx)$



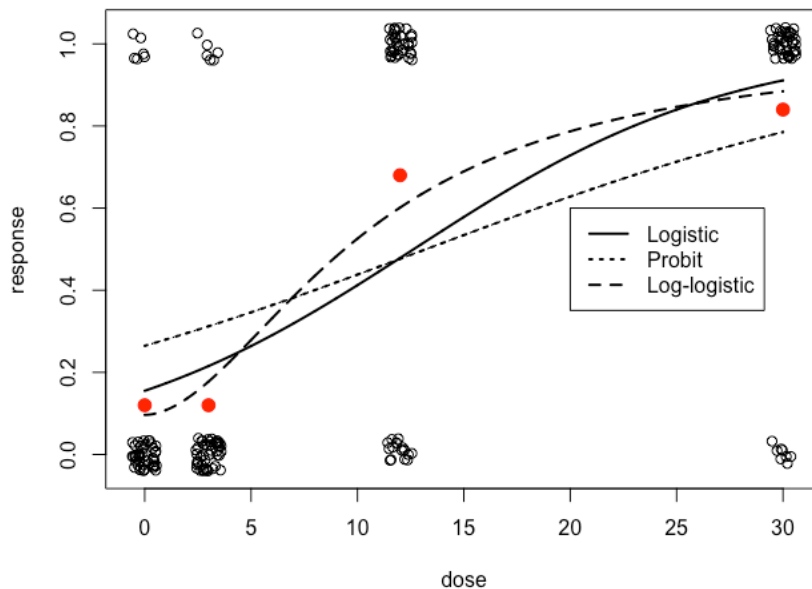
Coefficients:

|   | Estimate    | Std. Error  |
|---|-------------|-------------|
| a | -1.02188175 | 0.145057689 |
| b | 0.07728999  | 0.009641039 |

'log Lik.' -97.5356

# THYROID EPITHELIAL CELL VACUOLISATION

- Log-logistic model  $y = a + (1-a)/(1 + \exp(-\log(x/b)/c))$



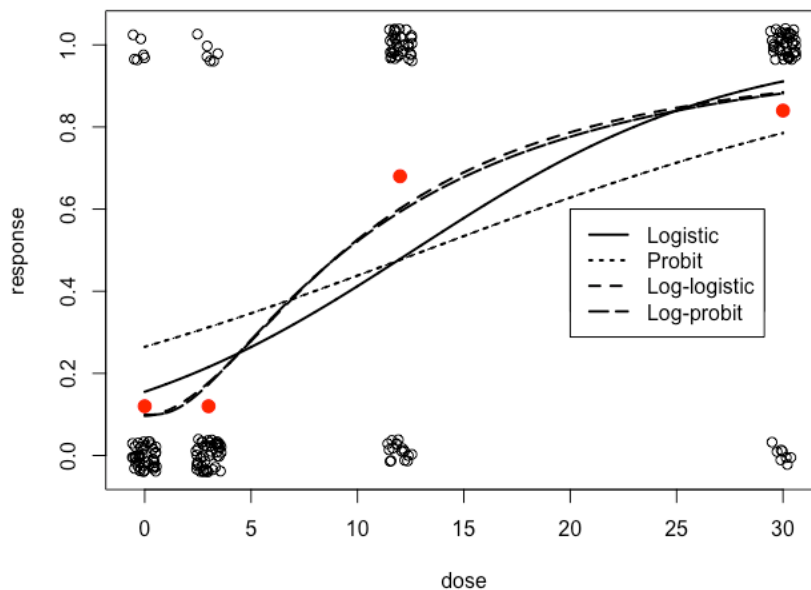
Coefficients:

|   | Estimate   | Std. Error |
|---|------------|------------|
| a | 0.0964394  | 0.03517616 |
| b | 10.5572758 | 1.52570441 |
| c | 0.5433158  | 0.10007674 |

'log Lik.' -91.90305

# THYROID EPITHELIAL CELL VACUOLISATION

- Log-probit model  $y = a + (1-a) \text{CumNorm}(\log(x/b)/c)$



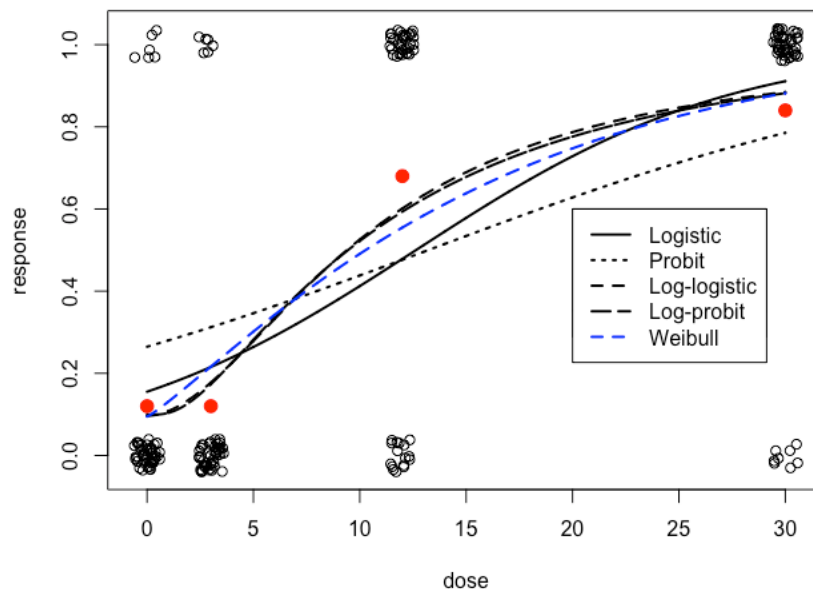
Coefficients:

|   | Estimate    | Std. Error |
|---|-------------|------------|
| a | 0.09922882  | 0.03617311 |
| b | 10.72949432 | 1.54589431 |
| c | 0.91708158  | 0.15473914 |

'log Lik.' -91.86382

# THYROID EPITHELIAL CELL VACUOLISATION

- Weibull model  $y = a + (1-a) \exp(-(x/b)^c)$



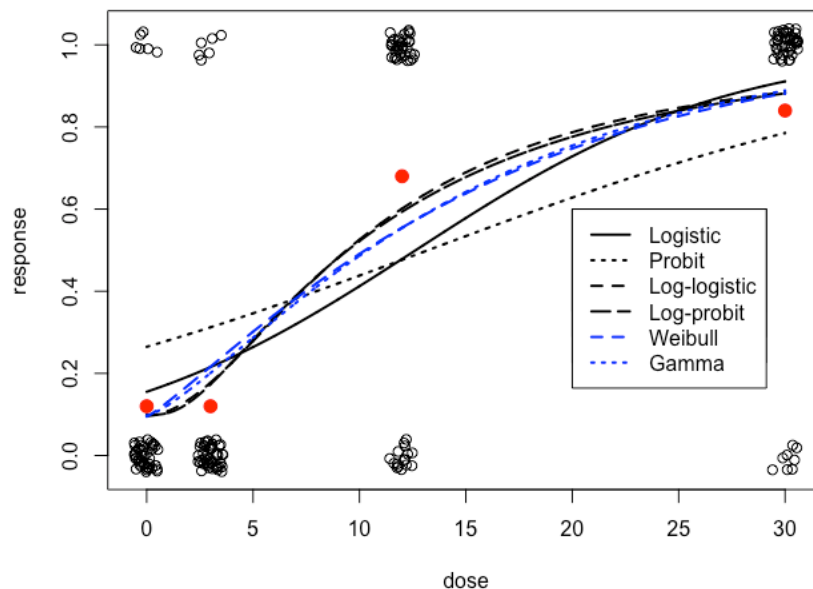
Coefficients:

|   | Estimate   | Std. Error |
|---|------------|------------|
| a | 0.09475043 | 0.03550996 |
| b | 0.04072331 | 0.02383889 |
| c | 1.15006741 | 0.19314240 |

'log Lik.' -93.77242

# THYROID EPITHELIAL CELL VACUOLISATION

- Gamma model  $y = a + (1-a) \text{CumGam}(bx^c)$



Coefficients:

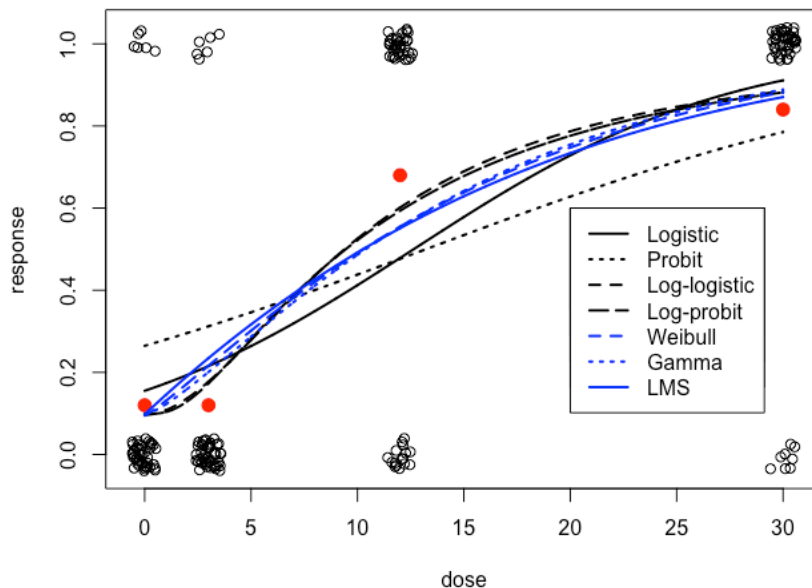
|   | Estimate   | Std. Error |
|---|------------|------------|
| a | 0.09758502 | 0.03628932 |
| b | 0.09005540 | 0.02978308 |
| c | 1.38160881 | 0.40025088 |

'log Lik.' -93.49419

# THYROID EPITHELIAL CELL VACUOLISATION

## ■ LMS model

$$y = a + (1-a)(1-\exp(-bx-cx^2))$$



Coefficients:

|   | Estimate     | Std. Error   |
|---|--------------|--------------|
| a | 0.0977349769 | 0.0363010174 |
| b | 0.0537970136 | 0.0137209538 |
| c | 0.0003621077 | 0.0004136556 |

'log Lik.' -94.19809

# THYROID EPITHELIAL CELL VACUOLISATION

| Model        | Log-likelihood | # par | AIC     |
|--------------|----------------|-------|---------|
| Null         | -137.186       | 1     | 276.372 |
| Full         | -90.020        | 4     | 188.039 |
| Logistic     | -97.234        | 2     | 198.468 |
| Probit       | -97.536        | 2     | 199.071 |
| Log-logistic | -91.903        | 3     | 189.806 |
| Log-probit   | -91.864        | 3     | 189.728 |
| Weibull      | -93.772        | 3     | 193.545 |
| Gamma        | -93.494        | 3     | 192.988 |
| LMS          | -94.198        | 3     | 194.396 |

$$AIC_{\text{Min}} = 189.728 < AIC_{\text{FULL}} + 2 = 190.039$$

$$\text{All } AIC < AIC_{\text{NULL}} - 2 = 274.372$$

# THYROID EPITHELIAL CELL VACUOLISATION

**Next presentation**

**Model Averaging and BMD estimation**

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