

Dose-response curves in Industrial Biotechnology

NCS-2012
Thomas A Poulsen
Novozymes A/S

3 08/11/2012 NOVOZYMES PRESENTATION

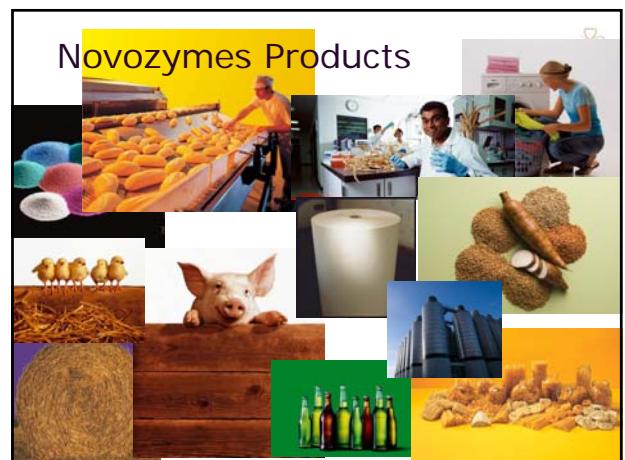
Novozymes

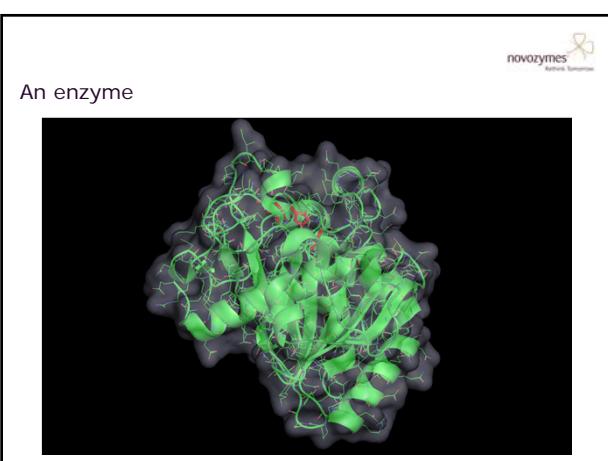
- Develops, produces and sells industrial enzymes
- Spun out of NovoNordisk in 2000
- 5800 Employees world wide
- 700 products in 130 countries within more than 30 different industries
- Sales ~ 1E10DKK (1.3E9 EUR)

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- Novozymes
- Laundry
- pKa-values / Enzyme kinetics

AGENDA





Washing is complicated

$$\frac{d\text{Fat}}{dt} = f(\text{fat, fat-type, variant, dose, detergent, detergent cone, pH, temperature, swatch, ...})$$

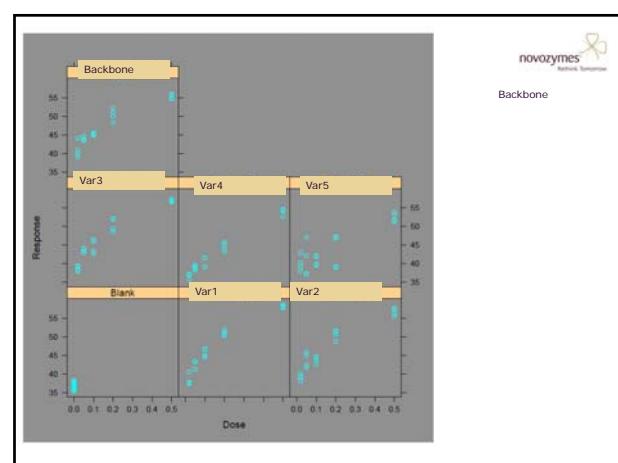
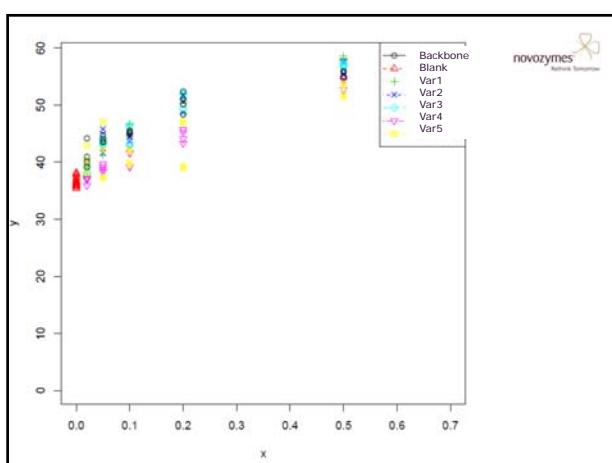
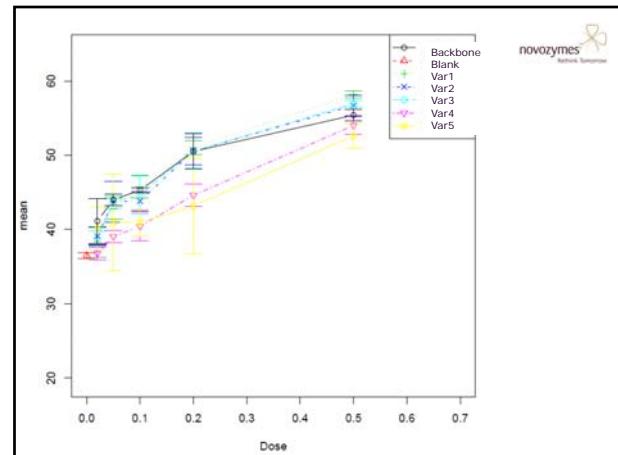
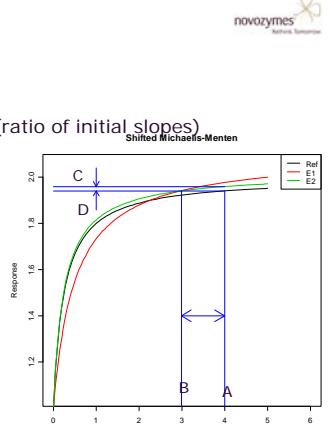
Could we study dependence on some of the parameters separately?

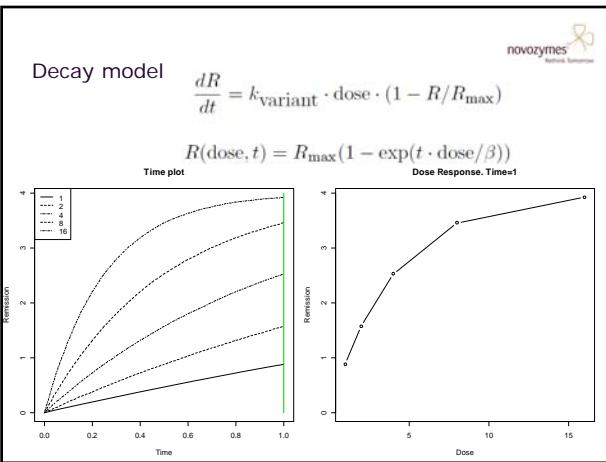
Parameters of interest

- RP: relative performance
- IF: Improvement factor (ratio of initial slopes)
- SF: Substitution Factor

$$SF = A/B$$

$$RP = C/D$$





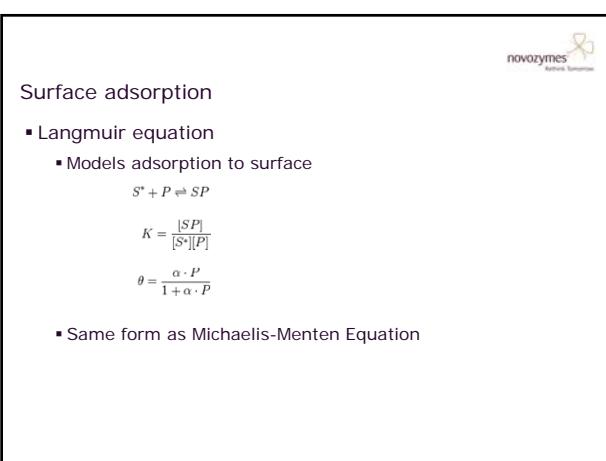
Estimates from drc

```
> summary(dd.drc)
A 'shifted Michaelis-Menten' model was fitted.

Parameter estimates:
```

	Estimate	Std. Error	t-value	p-value
Vm: Var1	32.615745	3.282793	9.935365	7.318e-18
Vm: Var2	29.550114	3.122574	9.463382	2.153e-16
Vm: Var3	29.302216	2.949913	9.933246	7.406e-18
Vm: Var4	114.722392	88.791193	1.292047	0.1987
Vm: Var5	72.908070	132.299997	0.551081	0.5825
Vm: Backbone	23.184242	1.974928	11.739298	2.643e-22
K: Var1	0.272534	0.059043	4.615852	9.447e-06
K: Var2	0.254111	0.059345	4.281943	3.622e-05
K: Var3	0.240404	0.054443	4.415731	2.131e-05
K: Var4	2.915903	2.608818	1.117710	0.2658
K: Var5	1.868360	4.201424	0.444697	0.6573
K: Backbone	0.143834	0.032735	4.393952	2.324e-05
y0:(Intercept)	37.095820	0.352603	105.205718	1.068e-125

Estimated residual variance: 2.987747



First order model

```
> summary(nlsList(Response ~ y0 * a*(1-2*(-Dose*T)) | Enzyme, data=d,
subset=d$Enzyme!="Blank", start=list(y0=50, a=5, T=2)))
Call:
Model: Response ~ y0 * a * (1 - 2*(-Dose * T)) | Enzyme
Data: d
```

	y0	Estimate Std. Error	a	Estimate Std. Error	T	Estimate Std. Error		
Var1	50.39845	0.2671837	Var1	3.827892	0.252251	Var1	5.740147	1.1720258
Var2	51.93990	0.1986147	Var2	3.676799	0.1982208	Var2	4.818542	0.8899828
Var3	51.35220	0.4210537	Var3	4.025224	0.3926118	Var3	6.842588	1.8143273
Var4	51.96671	0.2707380	Var3	4.119330	0.2547631	Var4	5.871807	1.1074860
Var5	50.24382	0.2662588	Var4	5.243593	0.2516271	Var5	5.705709	0.8518683
Backbone	49.70936	0.3759703	Var5	4.249818	0.3554718	Backbone	5.690506	1.4835772
Backbone	50.05580	0.2384233	Backbone	4.385852	0.2436905	Backbone	4.623022	0.8937416
Backbone	50.33227	0.2762553	Backbone	4.029864	0.2589696	Backbone	6.074128	1.1616943

Michaelis Menten Model

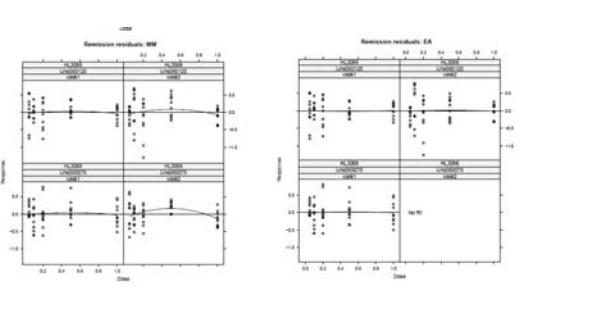
```
> summary( nlsList(Response ~ y0 + a*Dose/(c+Dose) | Enzyme, data=d,
subset=d$Enzyme!="Blank",start=list(y0=50, a=5, c=.5)))
Call:
Model: Response ~ y0 + a * Dose/(c + Dose) | Enzyme
Data: d

y0          a          c
Estimate Std. Error   Estimate Std. Error   Estimate Std. Error
Var1 50.04645 0.4437219 4.99783 0.3267290 0.2007349 0.0758759
Var2 51.71227 0.2983783 Var1 4.835225 0.2730711 0.2698033 0.0875264
Var2 50.76364 0.8180342 Var2 5.301707 0.6103626 0.1426677 0.0749920
Var3 51.60433 0.4475471 Var3 5.357709 0.3284988 0.1983283 0.0698840
Var4 49.77212 0.4396434 Var4 6.838845 0.3249970 0.2033743 0.0561070
Var5 49.21679 0.6584128 Var5 5.583683 0.4784864 0.1878077 0.0900356
Backbone 49.67558 0.3785044 Backbone 5.773523 0.3251463 Backbone 0.2548282 0.0841748
Backbone 49.99411 0.4554571 Backbone 5.221463 0.3325169 Backbone 0.1935510 0.0700469
```

Conclusions

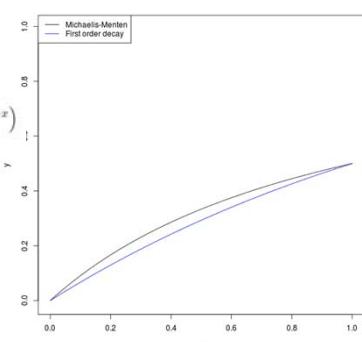
- Model choice not always obvious
- Both models are plausible
- No significant difference between models
- Needs more experiments
- But the project is closed

Residuals



$$M(x, v, k) = \frac{v x}{k + x}$$

$$E(x, a, b) = a \left(1 - 2^{-bx}\right)$$



2.1 Shifted Michaelis-Menten model

The (shifted) Michaelis-Menten model (4) has a number of nice features.

$$y = B + \frac{M \cdot x}{K + x} \quad (4)$$

where y is the cleaning, B is the cleaning by detergent alone, M is the asymptotic maximum, and K is the dose giving half of $M - B$. The enzyme dose is measured in x .

If we consider two enzymes (1 and 2) at the same dose, the relative performance is given in eq (5)

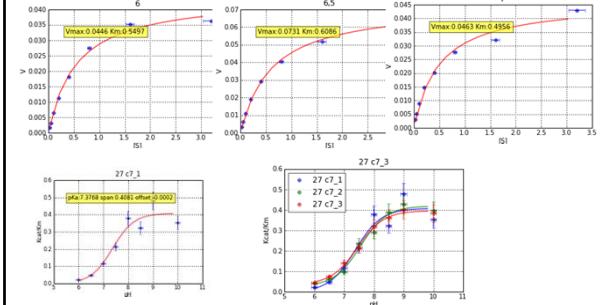
$$\text{RP} = \frac{y_1 - B}{y_2 - B} = \frac{M_1 \cdot x}{K_1 + x} \cdot \frac{k_2 + x}{M_2 \cdot k_1 + x} = \frac{M_1}{M_2} \cdot \frac{k_2 + x}{K_1 + x} \quad (5)$$

Asymptotically this is

$$\text{RP} \xrightarrow{x \rightarrow \infty} \frac{M_1}{M_2}, \quad \text{RP} \xrightarrow{x \rightarrow 0} \frac{M_1}{M_2} \cdot \frac{K_2}{K_1} = \text{IF} \quad (6)$$



Measuring pKa-values



Enzyme pKa values

Titration curves are of the general form (1), as illustrated in figure 1.

$$f(x) = \frac{1}{1 + a^{(b-x)}} \quad (1)$$

$$f'(x) = \frac{a^{(b-x)} \log(a)}{(1 + a^{b-x})^2} \quad (2)$$

In chemistry, b is the pKa value, x the pH, and $a = 10$. From the general form, we see that the slope at the pKa is $\log(10)/4 = 0.58$.

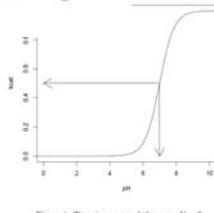
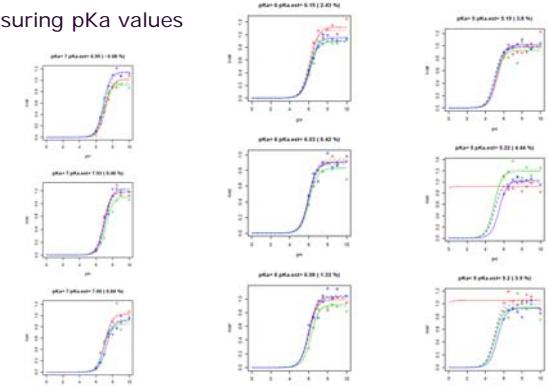


Figure 1: Titration curve of the case pKa=7



Measuring pKa values

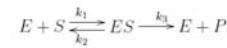


Labour effort (50 variants)



- Traditional MM:
 - 8 pH values
 - 8 substrate concentrations
 - 3 replicates
 - = 192 kinetic curves per variant

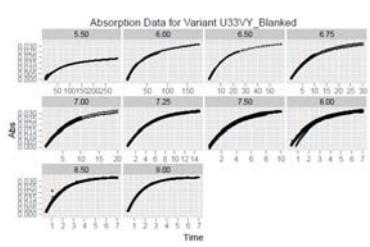
Enzyme kinetics



$$\begin{aligned}\dot{c} &= k_1 e s - (k_2 + k_3) c \\ \dot{p} &= k_3 c\end{aligned}$$

Reagent	Timedep. conc.	Initial conc.
Enzyme	e	e_0
Substrate	s	s_0
Enzyme-substrate complex	c	0
Product	p	0

Reader curves



Enzyme Kinetics Michaelis-Menten



$$\begin{aligned}\dot{c} &= k_1 e s - (k_2 + k_3) c \\ \dot{p} &= k_3 c\end{aligned}$$

$$e_0 = e + c$$

$$\dot{c} = 0$$

$$c = \frac{e_0 s}{\frac{k_2+k_3}{k_1} + s} = \frac{e_0 s}{k_M + s}$$

$$V = \dot{p} = k_3 c = k_3 e_0 \frac{s}{K_M + s} = V_{\max} \frac{s}{K_M + s}$$

Enzyme Kinetics: Substrate depletion

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$$V = \dot{p} = k_3 c = k_3 e_0 \frac{s}{K_M + s} = V_{\max} \frac{s}{K_M + s}$$

$$\dot{p} = -\dot{s}$$

$$\dot{s} = -k_3 e_0 \frac{s}{K_M + s}$$

$$\dot{s} = -k_3 e_0 \frac{s}{K_M + s} = -\frac{k_3 e_0 s}{K_M} + \frac{k_3 e_0 s^2}{K_M^2} - \frac{k_3 e_0 s^3}{K_M^3} + \dots$$

$$\dot{s} = -\frac{k_3 e_0}{K_M} s \quad s(t) = s_0 \exp\left(-\frac{k_3}{K_M} e_0 t\right)$$

$$p = s_0 \left(1 - \left(1 - \frac{e_0}{K_M}\right) e^{(-D t)}\right)$$

pKa values. Substrate depletion results

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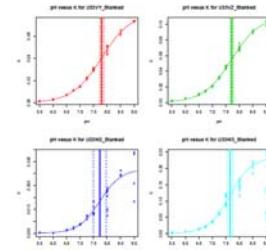


Figure 1: Models from cleaned data, *excluding* points marked by circles. Solid lines indicate estimated pK_a and the dashed lines are 19.6 standard errors.

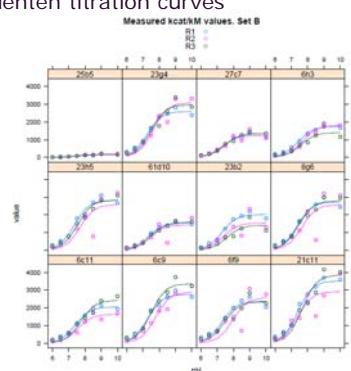
Labour effort (50 variants)

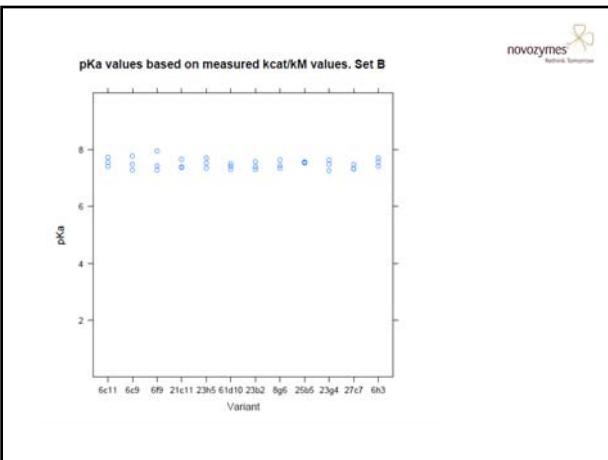
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- Traditional MM:
 - 8 pH values
 - 8 substrate concentrations
 - 3 replicates
 - = 192 kinetic curves per variant
- Substrate depletion:
 - 8 pH values
 - 3 substrate concentrations
 - 3 replicates
 - = 48 kinetic curves per variant

Michaelis-Menten titration curves

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Thanks to

- Jens E Nielsen, UCD / Novozymes
- Helen Webb, UCD
- Damien Farrell, UCD
- Esben P Friis, Novozymes
- Benjamin Høyer, DTU, Novozymes

▪ Ref (pKa): *Biochemistry*, 2012, 51 (26), pp 5339–5347

▪ [Animation package](#): Yihui Xie

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Conclusions

- pKa-values are difficult to measure
 - And even harder to change
- It is important to consider analysis as part of experiment design
- Non-linear analysis can save ¾ of the work

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