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INNOVATION LECTURES (I.N.N.O.I.E.C.) www.muni.cz

**Binding and Kinetics for Experimental Biologists**  
Lecture 1  
**Numerical Models for Biomolecular Interactions**

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Tento projekt je spolufinancován Evropským sociálním fondem a státním rozpočtem České republiky.

EVROPSKÁ UNIE **esf** MINISTERSTVO ŠKOLSTVÍ, MLÁDEŽE A TĚLOVÝCHOVY OP Vzdělávání pro konkurenceschopnost

INVESTICE DO ROZVOJE VZDĚLÁVÁNÍ

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**Lecture outline**

- The problem:**  
Traditional equations for fitting biomolecular binding data restrict the experimental design. Typically, at least one component must be present in very large excess.
- The solution:**  
Abandon algebraic equations entirely. Use iterative numerical models, which can be derived automatically by the computer.
- An implementation:**  
Software *DynaFit*.
- An example:**  
Kinetics of forked DNA binding to the protein-protein complex formed by DNA-polymerase sliding clamp (gp45) and clamp loader (gp44/62).

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**Traditional approach is based on algebraic models**

**Binding and Kinetics for Molecular Biologists**  
A typical "cookbook" for experimental biologists  
 $A + B \rightleftharpoons AB$   
molecular mechanism  
algebraic model  
James A. Goodrich  
Jennifer F. Kugel  
Cold Spring Harbor Laboratory Press  
Cold Spring Harbor, NY, 2007  
ISBN-10: 0879697369

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**Algebraic models restrict experiment design**

Goodrich & Kugel (2006) *Binding and Kinetics for Molecular Biologists*

**EXAMPLE:** Determine the association rate constant for  $A + B \rightarrow AB$

**Summary: How to experimentally measure  $k_1$**

- Develop an assay to monitor AB.
- Combine A and B at  $t = 0$  ( $[B]_0 \gg [AB]_{max}$ ).
- Measure accumulation of AB over time.
- Plot data and determine  $k_{obs}$ .
- Repeat Steps 2-4 at multiple different  $[B]_0$ .
- Plot  $k_{obs}$  versus  $[B]_0$  and determine  $k_1$ .
- Consider whether  $k_1$  is diffusion limited.

$AB_t = AB_{max}(1 - e^{-k_{obs}t})$   
 $k_{obs} = k_1[B]_0 + k_{-1}$

Typically, this condition is met by setting the initial concentration of B to be much greater than the initial concentration of A (e.g., 100-fold).

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**Experimental handbooks are full of restrictions**

Goodrich & Kugel (2007) *Binding and Kinetics for Molecular Biologists*

Under conditions where the concentration of B is much greater than  $K_{D(AB)}$  ( $[B] \gg K_{D(AB)}$ ) the following equation can be used p. 34

In using this technique, it is important to make sure that the concentration of unlabeled A is greater than the concentration of B in the reactions. p. 79

Third, reactions should be performed under conditions where the initial concentration of B ( $[B]_0$ ) is much greater than the sum of the final concentrations of AB and  $AB^*$  p. 120

Using the abortive initiation assay,  $k_{obs}$  values were measured at multiple different RNA polymerase concentrations (always greater than the concentration of DNA) at two different promoters. p. 126

etc.

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**Numerical vs. algebraic mathematical models**

FROM A VARIETY OF ALGEBRAIC EQUATIONS TO A UNIFORM SYSTEM OF DIFFERENTIAL EQUATIONS

**EXAMPLE:** Determine the rate constant  $k_1$  and  $k_{-1}$  for  $A + B \xrightleftharpoons[k_{-1}]{k_1} AB$

| ALGEBRAIC EQUATIONS                  | DIFFERENTIAL EQUATIONS               |
|--------------------------------------|--------------------------------------|
| $AB_t = AB_{max}(1 - e^{-k_{obs}t})$ | $d[A]/dt = -k_1[A][B] + k_{-1}[AB]$  |
| $k_{obs} = k_1[B]_0 + k_{-1}$        | $d[B]/dt = -k_1[A][B] + k_{-1}[AB]$  |
|                                      | $d[AB]/dt = +k_1[A][B] - k_{-1}[AB]$ |
| Applies only when $[B] \gg [A]$      | Applies under all conditions         |

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### Advantages and disadvantages of numerical models

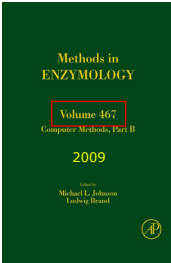
THERE IS NO SUCH THING AS A FREE LUNCH

| ADVANTAGE  | ALGEBRAIC MODEL | DIFFERENTIAL MODEL |
|--|-----------------|--------------------|
| can be derived for any molecular mechanism       | -               | +                  |
| can be derived automatically by computer         | -               | +                  |
| can be applied under any experimental conditions | -               | +                  |
| can be evaluated without specialized software    | +               | -                  |
| requires very little computation time            | +               | -                  |
| does not require an initial estimate             | +               | -                  |
| is resistant to truncation and round-off errors  | +               | -                  |
| has a long tradition: many papers published      | +               | -                  |

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### Specialized numerical software: DynaFit

MORE THAN 600 PAPERS PUBLISHED WITH IT (1996 - 2009)



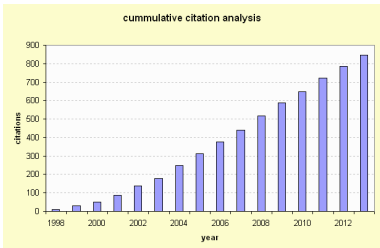
CHAPTER TEN  
**DYNAFIT—A SOFTWARE PACKAGE FOR ENZYMOLOGY**  
Petr Kuzmíč

DOWNLOAD <http://www.biokin.com/dynafit>  
**FREE** for academic users

Kuzmíč (2009) *Meth. Enzymol.*, **467**, 247-280  
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### DynaFit software: Citation analysis

APPROXIMATELY 850 JOURNAL ARTICLES PUBLISHED SINCE 1998



Kuzmíč, P. (1996) "Program DYNAFIT for the analysis of enzyme kinetic data: Application to HIV proteinase" *Anal. Biochem.* **237**, 260-273.  
Kuzmíč, P. (2009) "DynaFit - A software package for enzymology" *Meth. Enzymol.* **467**, 247-280.

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### Theoretical foundations: Mass Action Law

RATE IS PROPORTIONAL TO CONCENTRATION(S)

MONOMOLECULAR REACTIONS  
 $A \rightarrow \text{products}$   
rate is proportional to  $[A]$   
 $-d[A]/dt = k[A]$   
monomolecular rate constant  $1/\text{time}$

BIMOLECULAR REACTIONS  
 $A + B \rightarrow \text{products}$   
rate is proportional to  $[A] \times [B]$   
 $-d[A]/dt = -d[B]/dt = k[A] \times [B]$   
bimolecular rate constant  $1/(\text{concentration} \times \text{time})$

thought bubble: "rate" ... "derivative"

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### Theoretical foundations: Mass Conservation Law

PRODUCTS ARE FORMED WITH THE SAME RATE AS REACTANTS DISAPPEAR

EXAMPLE  
 $A \rightarrow P + Q$   
 $-d[A]/dt = +d[P]/dt = +d[Q]/dt$

COMPOSITION RULE ADDITIVITY OF TERMS FROM SEPARATE REACTIONS

mechanism:  
 $A \xrightarrow{k_1} B$   
 $B \xrightarrow{k_2} C$   
 $d[B]/dt = +k_1[A] - k_2[B]$

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### Composition Rule: Example

EXAMPLE MECHANISM

$$\begin{aligned}
 E + A &\xrightleftharpoons[k_{-1}]{k_{+1}} EA \\
 EA + B &\xrightleftharpoons[k_{-2}]{k_{+2}} EAB \\
 E + B &\xrightleftharpoons[k_{-3}]{k_{+3}} EB \\
 EB + A &\xrightleftharpoons[k_{-4}]{k_{+4}} EAB \\
 EAB &\xrightarrow{k_{+5}} E + P + Q
 \end{aligned}$$

RATE EQUATIONS

$$\begin{aligned}
 d[P]/dt &= +k_{+5}[EAB] \\
 d[EAB]/dt &= +k_{+2}[EA][B] - k_{-2}[EAB] + k_{+4}[EB][A] - k_{-4}[EAB] - k_{+5}[EAB]
 \end{aligned}$$

Similarly for other species...

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### A "Kinetic Compiler"

HOW DYNAFIT PROCESSES YOUR BIOCHEMICAL EQUATIONS

$$E + S \xrightleftharpoons[k_2]{k_1} E \cdot S \xrightarrow{k_3} E + P$$

| Input (plain text file):   | Rate terms:   | Rate equations:  |
|--|---|--|
| <pre>E + S ---&gt; ES : k1 ES ---&gt; E + S : k2 ES ---&gt; E + P : k3</pre> | $k_1 \times [E] \times [S]$<br>$k_2 \times [ES]$<br>$k_3 \times [ES]$ | $\frac{d[E]}{dt} = -k_1 \times [E] \times [S] + k_2 \times [ES] + k_3 \times [ES]$ $\frac{d[ES]}{dt} = +k_1 \times [E] \times [S] - k_2 \times [ES] - k_3 \times [ES]$ <p>Similarly for other species...</p> |

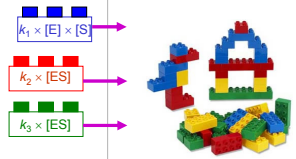

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### System of Simple, Simultaneous Equations

HOW DYNAFIT PROCESSES YOUR BIOCHEMICAL EQUATIONS

$$E + S \xrightleftharpoons[k_2]{k_1} E \cdot S \xrightarrow{k_3} E + P$$

"The LEGO method" of deriving rate equations

| Input (plain text file):   | Rate terms:   | Rate equations:   |
|--|---|---|
| <pre>E + S ---&gt; ES : k1 ES ---&gt; E + S : k2 ES ---&gt; E + P : k3</pre> |  |  |

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### DynaFit can analyze many types of experiments

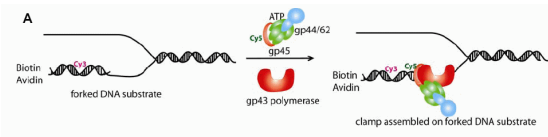
MASS ACTION LAW AND MASS CONSERVATION LAW IS APPLIED IN THE SAME WAY

|                                       | EXPERIMENT                    | DYNAFIT DERIVES A SYSTEM OF ...       |
|---------------------------------------|-------------------------------|---------------------------------------|
| chemistry<br>biophysics<br>enzymology | <b>Kinetics (time-course)</b> | Ordinary differential equations (ODE) |
|                                       | <b>Equilibrium binding</b>    | Nonlinear algebraic equations         |
|                                       | <b>Initial reaction rates</b> | Nonlinear algebraic equations         |

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### Example: DNA + clamp / clamp loader complex

DETERMINE ASSOCIATION AND DISSOCIATION RATE CONSTANT IN AN  $A + B \rightleftharpoons AB$  SYSTEM



Schematic representation of the clamp loading onto forked-DNA substrate. In the forked DNA-substrate, the primer carries a Cy3 fluorescent donor and the gp45 clamp contains an acceptor Cy5 dye.

**Typical email from a Ph.D. student:**

I am limited with equimolar concentrations, as I would otherwise need a lot of material for the experiments, which are really expensive because of the dyes and proteins.

Courtesy of Senthil Perumal, Penn State University (Steven Benkovic lab)

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### Example: Experimental setup

ALL COMPONENTS PRESENT AT EQUAL CONCENTRATIONS

- pre-mix sliding clamp (C) + clamp loader (L) to form C.L complex;
- add DNA solution;
- observe the formation of C.L.DNA ternary complex over time

final concentrations:

**100 nM clamp** ... gp45 labeled with Cy5 acceptor dye  
**100 nM loader** ... gp44/62  
**100 nM DNA** ... primer labeled with Cy3 donor dye

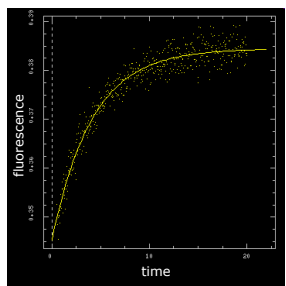
C.L complex has estimated  $K_d = 1 \text{ nM}$ , so C.L  $\rightarrow$  C+L dissociation upon adding DNA should be negligible

Courtesy of Senthil Perumal, Penn State University (Steven Benkovic lab)

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### Example: Raw data

JUST BECAUSE THE DATA FIT TO A MODEL DOES NOT MEAN THAT THE MODEL IS CORRECT!



raw fluorescence  $F$  fit to  $F = A_0 + A_1 \exp(-k t)$

exponential model fits the data well but it is theoretically **invalid!**

Courtesy of Senthil Perumal, Penn State University (Steven Benkovic lab)

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### Example: Anatomy of DynaFit scripts

DYNAFIT SOFTWARE IS DRIVEN BY TEXT "SCRIPTS" - MINIATURE "COMPUTER PROGRAMS"

```

DynaFit: fit-001.txt
File Edit View Help
Input Output

[[task]
task = fit
data = progress

[mechanism]
DNA + Clamp.Loader <=> Complex : kon koff

[constants]
kon = 1 ?
koff = 1 ?

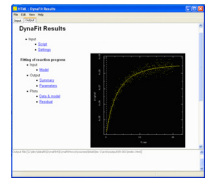
[concentrations]
DNA = 0.1
Clamp.Loader = 0.1
    
```

### Example: DynaFit tutorial

YOUR FIRST DYNAFIT DATA-ANALYSIS SESSION

TUTORIAL

1. Start DynaFit
2. Select menu "File ... Open" or press **Ctrl+O**
3. Navigate to file  
./courses/bkeb/lec-1/a+b/fit-001.txt
4. Select menu "File ... Try" or press **Ctrl+T**  
This is the initial estimate
5. Select menu "File ... Run" or press **Ctrl+U**  
Wait several seconds to finish the analysis
6. Select menu "View ... Results in External Browser"  
Navigate in the output files



### Example: Detailed explanation

A BIT OF THEORY

1. Reaction order
2. Units and dimensions (scaling)
3. The DynaFit model for biomolecular kinetics
4. Initial estimates of model parameters

### Molecularity and reaction order

IN PRACTICE WE ENCOUNTER ONLY ZERO-, FIRST-, AND SECOND-ORDER REACTIONS

| ORDER   | PHYSICAL MEANING                                    | NOTATION  | DYNAFIT NOTATION  |
|---|---|---|---|
| <b>zero-</b>  | constant-rate<br>influx or efflux                   |   | $X \rightarrow \quad : v$                               |
| <b>first-</b><br><i>uni-molecular<br/>monomolecular</i> | isomerization or<br>dissociation of<br>one molecule | $A \xrightarrow{k_1} B$<br>$A \xrightarrow{k'_1} B + C$ | $A \rightarrow B : k_1$<br>$C \rightarrow A + B : k'_1$ |
| <b>second-</b><br><i>bimolecular</i>                    | binding<br>(association) of<br>two molecules        | $A + B \xrightarrow{k_2} C$                             | $A + B \rightarrow C : k_2$                             |

### Reversible reactions and reaction mechanisms

DYNAFIT CAN HANDLE AN ARBITRARY NUMBER OF ELEMENTARY REACTIONS IN A MECHANISM

| REVERSIBLE REACTION  | DYNAFIT NOTATION   |
|--|--|
| $A + B \xrightleftharpoons[k_2]{k_1} C$  | $A + B <=> C : k_1 \quad k_2$  |
| MULTI-STEP MECHANISM   | DYNAFIT NOTATION   |
| $A \xrightleftharpoons[k_2]{+B \quad k_1} AB$<br>$A \xrightleftharpoons[k_5]{+C \quad k_3} AC$<br>$AC \xrightarrow{k_4} X$ | $A + B <=> AB : k_1 \quad k_2$<br>$A + C <=> AC : k_3 \quad k_4$<br>$AC \rightarrow X : k_5$ |

### Dimensions of rate constants

CAREFUL ABOUT DIMENSIONS OF RATE CONSTANTS! DIMENSIONAL ANALYSIS

forward and reverse reaction rates:

$$v_{\rightarrow} = k_1 [A] [B]$$

$$v_{\leftarrow} = k_2 [AB]$$

| quantity   | dimension            |
|------------|----------------------|
| $v$        | concentration / time |
| $[X]$      | concentration        |
| $k_1, k_2$ | ?                    |

EXAMPLE  
dimensional analysis of  $k_1$  (**bimolecular association** rate constant)

$$v_{\rightarrow} = k_1 [A] [B] \Rightarrow k_1 = \frac{v_{\rightarrow}}{[A] [B]} = \frac{\text{concentration} / \text{time}}{\text{concentration} \times \text{concentration}} = \frac{1}{\text{concentration} \times \text{time}}$$

### Dimensions of rate and equilibrium constants

**SUMMARY**

$A + B \xrightleftharpoons[k_2]{k_1} AB$

dissociation rate constant  $k_2$  **1/sec**  
 association rate constant  $k_1$  **1/(M sec)**  
 dissociation equilibrium constant  $K_d = k_2/k_1$  **M**  
 association equilibrium constant  $K_a = k_1/k_2$  **1/M**

---

$A \xrightleftharpoons[k_4]{k_3} A'$

→ rate constant  $k_3$  **1/sec**  
 ← rate constant  $k_4$  **1/sec**  
 → equilibrium constant  $K_{eq} = k_3/k_4$  **--**  
 ← equilibrium constant  $K_{eq} = k_4/k_3$  **--**

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### Example: Units (scaling) of rate constants

ALL UNITS ARE ARBITRARY BUT MUST BE IDENTICAL THROUGHOUT THE ENTIRE SCRIPT!

```
[mechanism]
DNA + Clamp.Loader <=> Complex : kon koff

[constants]
kon = 1 ?
koff = 1 ?

[concentrations]
DNA = 0.1
Clamp.Loader = 0.1
Complex = 1 ?

[responses]
Complex = 1 ?

[data]
file ./.../dl-edit.txt
offset auto ?
```

$kon = 1 \mu M^{-1} s^{-1} = 10^6 M^{-1} s^{-1}$   
 concentration units of rate constants:  $\mu M$   
 time units of rate constants: **sec**  
 100 nM

| time, s | signal, eV |
|---------|------------|
| 0.52    | 0.3526     |
| 0.56    | 0.3484     |
| 0.60    | 0.3485     |
| 0.64    | 0.3454     |
| 0.68    | 0.3499     |
| 0.72    | 0.3502     |
| ...     | ...        |

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### Example: The "response" coefficient

MOLAR "RESPONSE" = PROPORTIONALITY FACTOR LINKING CONCENTRATIONS TO SIGNAL

```
[mechanism]
DNA + Clamp.Loader <=> Complex : kon koff

[constants]
kon = 1 ?
koff = 1 ?

[concentrations]
DNA = 0.1
Clamp.Loader = 0.1
Complex = 1.00 ?

[responses]
Complex = 1.00 ?

[data]
file ./.../dl-edit.txt
offset auto ?
```

one concentration unit (in this case 1  $\mu M$ ) of Complex will produce an increase in the signal equal to 1.00 instrument units

| time, s | signal, eV |
|---------|------------|
| 0.52    | 0.3526     |
| 0.56    | 0.3484     |
| 0.60    | 0.3485     |
| 0.64    | 0.3454     |
| 0.68    | 0.3499     |
| 0.72    | 0.3502     |
| ...     | ...        |

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### Example: Initial estimates

NONLINEAR REGRESSION ANALYSIS ALWAYS REQUIRES INITIAL ESTIMATES OF THE SOLUTION

```
[mechanism]
DNA + Clamp.Loader <=> Complex : kon koff

[constants]
kon = 1 ?
koff = 1 ?

[concentrations]
DNA = 0.1
Clamp.Loader = 0.1
Complex = 1 ?

[responses]
Complex = 1 ?

[data]
file ./.../dl-edit.txt
offset auto ?
```

the initial estimate of rate constants

optimized model parameters

A VERY DIFFICULT PROBLEM:  
HOW TO GUESS "GOOD ENOUGH" INITIAL ESTIMATES OF RATE CONSTANTS?

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### Example: "Good" initial estimate

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### Example: "Good" initial estimate - results

Regression Summary

| Levenberg-Marquardt Algorithm |      | No. | Par#Set | Initial    | Final     | Std. Error | CV (%) | Note |
|-------------------------------|------|-----|---------|------------|-----------|------------|--------|------|
| #1                            | kon  | 1   |         | 2.202992   | 0.4650206 |            | 21.30  |      |
| #2                            | koff | 1   |         | 0.0298798  | 0.015363  |            | 51.48  |      |
| sum of squares                |      |     |         | 0.00230769 |           |            |        |      |

Optimized Parameters

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### Example: "Bad" initial estimate

100 times smaller / larger

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### Example: "Bad" initial estimate – results

| Regression Summary            |                  | Optimized Parameters |         |          |          |            |         |      |
|-------------------------------|------------------|----------------------|---------|----------|----------|------------|---------|------|
| Levenberg-Marquardt Algorithm |                  | No.                  | Par#Set | Initial  | Final    | Std. Error | CV (%)  | Note |
| #1                            | k <sub>on</sub>  | 100                  |         | 0.164621 | 3.3584e3 |            | 2055.89 |      |
| #2                            | k <sub>off</sub> | 0.01                 |         | 0.211276 | 0.609154 | 288.32     |         |      |
| sum of squares                |                  | 0.00235435           |         |          |          |            |         |      |

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### Example: "Good" vs. "Bad" results - comparison

|        | initial estimate                              | sum of squares | relative sum of sq. | "best-fit" constants   | K <sub>d</sub> , nM = k <sub>2</sub> /k <sub>1</sub> |
|--------|---|----------------|---------------------|--|--|
| "good" | k <sub>1</sub> = 1<br>k <sub>2</sub> = 1      | 0.002308       | 1.00                | k <sub>1</sub> = 2.2 ± 0.5<br>k <sub>2</sub> = 0.030 ± 0.015 | 13 nM  |
| "bad"  | k <sub>1</sub> = 100<br>k <sub>2</sub> = 0.01 | 0.002354       | 1.02                | k <sub>1</sub> = 0.2 ± 3.4<br>k <sub>2</sub> = 0.2 ± 0.6     | 1000 nM  |

DynaFit warnings from running the "bad" estimate:

|               |  |
|---------------|--|
| Error status  | 3  |
| Error message | WARNING: No convergence in Levenberg-Marquardt algorithm. Consider increasing the number of iterations. WARNING: Hessian matrix inversion failed. This problem is ill-conditioned. Parameter errors and model inference bands are suspect. To fix this problem, try removing at least one severely redundant parameter from the fitting model. |

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### Example: "Good" vs. "Bad" results - comparison

From "good" initial estimate      From "bad" initial estimate

not very encouraging!

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### Example: "Good" vs. "Bad" results - summary

- Initial estimates "off" by a factor of 100 can produce misleading results.
- The data/model overlay may "look good", but the results may be invalid.
- The same applies to the residual sum of squares (only 2% difference).
- The only indication that something went wrong might be:
  - huge standard errors of model parameters; and
  - various warnings from the least-squares fitter
- The simplest possible safeguard: Use several different initial estimates?

Disadvantage: how do we know which multiple estimates?

**LOOKING AHEAD**

DynaFit offers more reliable and convenient safeguards

- Global minimization
- Systematic combinatorial scan
- Confidence interval search

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### Summary and conclusions

**NUMERICAL MODELS IN BIOCHEMISTRY AND BIOPHYSICS: BETTER THAN ALGEBRAIC EQUATIONS**

- Numerical models are applicable to all experimental conditions. No more "large excess of this over that".
- Numerical models apply uniformly to all types of experiments:
  - reaction progress (kinetics);
  - equilibrium composition (binding);
  - enzyme catalysis.
- Numerical models can be automatically derived by computer. No more looking up algebraic equations – if they exist at all.
- Main disadvantage: requirement for specialized software. But DynaFit is free to academic users.
- Not a "silver bullet"! Example: the initial estimate problem. But this is not specific to numerical models (applies to algebraic models, too).

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