

# Numerical Enzymology

## Generalized Treatment of Kinetics & Equilibria

Petr Kuzmič, Ph.D.  
BioKin, Ltd.

### DYNAFIT SOFTWARE PACKAGE

#### 1. Overview of recent applications

#### 2. Selected examples

- ATPase cycle of Hsp90 Analog Trap1 (Leskovar *et al.*, 2008)
- Nucleotide binding to ClpB (Werbeck *et al.*, 2009)
- Clathrin uncoating (Rothnie *et al.*, 2011)

#### 3. Recent enhancements

##### - Optimal Experimental Design



### DYNAFIT software

#### NUMERICAL ENZYME KINETICS AND LIGAND BINDING

Kuzmic (1996) *Anal. Biochem.* **237**, 260-273.

ANALYTICAL BIOCHEMISTRY 237, 260 - 273 (1996)  
ARTICLE NO. 0238

#### Program DYNAFIT for the Analysis of Enzyme Kinetic Data: Application to HIV Proteinase

Petr Kuzmič

School of Pharmacy and Department of Chemistry, University of Wisconsin, 1101 University Avenue, Madison, Wisconsin 53706; and BioKin, Ltd., 1601 Adams Street, Madison, Wisconsin 53711

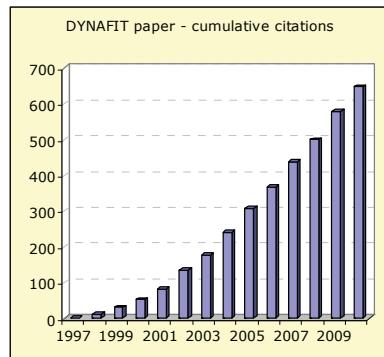
Received January 26, 1996

<http://www.biokin.com/dynafit>



## DynaFit: Citation analysis

JULY 2011: 683 BIBLIOGRAPHIC REFERENCES ("WEB OF SCIENCE")



Biochemistry (USA)  
*J. Biol. Chem.*

~65%  
~20%



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## A "Kinetic Compiler"

HOW DYNAFIT PROCESSES YOUR BIOCHEMICAL EQUATIONS



Input (plain text file):

```
E + S ---> ES : k1
ES ---> E + S : k2
ES ---> E + P : k3
```

Rate terms:

$k_1 \times [E] \times [S]$

$k_2 \times [ES]$

$k_3 \times [ES]$

Rate equations:

$$\frac{d[E]}{dt} = -k_1 \times [E] \times [S] + k_2 \times [ES] + k_3 \times [ES]$$

$$\frac{d[ES]}{dt} = +k_1 \times [E] \times [S] - k_2 \times [ES] - k_3 \times [ES]$$

Similarly for other species...



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## System of Simple, Simultaneous Equations

HOW DYNAFIT PROCESSES YOUR BIOCHEMICAL EQUATIONS

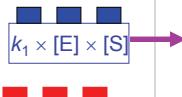


"The  method"  
of deriving rate equations

Input (plain text file):

```
E + S ---> ES : k1
ES ---> E + S : k2
ES ---> E + P : k3
```

Rate terms:

  
 $k_1 \times [E] \times [S]$   
 $k_2 \times [ES]$   
 $k_3 \times [ES]$

Rate equations:



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## DynaFit can analyze many types of experiments

MASS ACTION LAW AND MASS CONSERVATION LAW IS APPLIED TO DERIVE DIFFERENT MODELS

### EXPERIMENT

### DYNAFIT DERIVES A SYSTEM OF ...

**Reaction progress**

First-order ordinary differential equations

**Initial rates**

Nonlinear algebraic equations

**Equilibrium binding**

Nonlinear algebraic equations

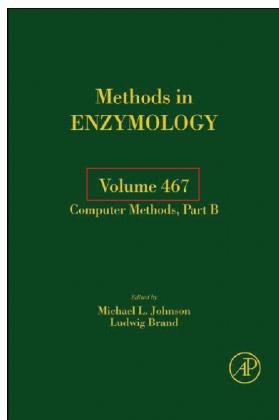


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## DynaFit: Recent enhancements

REVIEW (2009)



CHAPTER TEN

### DYNAFIT—A SOFTWARE PACKAGE FOR ENZYMOLOGY

Petr Kuzmič

Kuzmic, P. (2009) *Meth. Enzymol.* **467**, 248-280



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## DynaFit Example 1: Trap1 ATPase cycle

EXCELLENT EXAMPLE OF COMBINING "TRADITIONAL" (ALGEBRAIC) AND NUMERICAL (DYNAFIT) MODELS

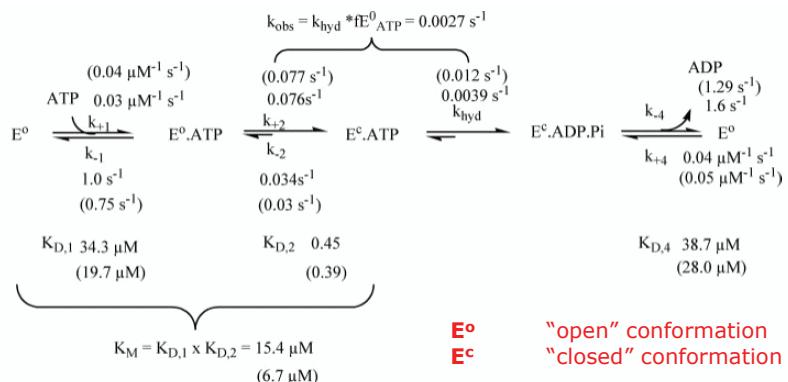


FIGURE 7. Kinetic model of Trap1 ATPase cycle at 25 °C

Leskovar et al. (2008)

"The ATPase cycle of the mitochondrial Hsp90 analog Trap1"  
*J. Biol. Chem.* **283**, 11677-688

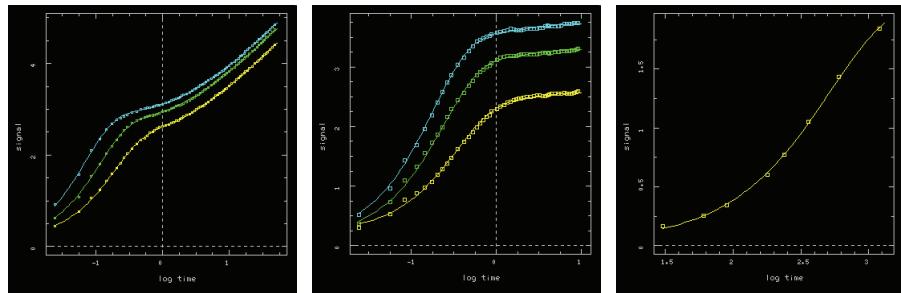


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## DynaFit Example 1: Trap1 ATPase cycle - experiments

THREE DIFFERENT TYPES OF EXPERIMENTS COMBINED



- varied [ATP analog]
- stopped flow fluorescence

- varied [ADP analog]
- stopped flow fluorescence

- single-turnover ATPase assay

Leskovar et al. (2008) *J. Biol. Chem.* **283**, 11677-688



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## DynaFit Example 1: Trap1 ATPase cycle - script

MECHANISM INCLUDES PHOTO-BLEACHING (ARTIFACT)

```
[task]
data = progress
task = fit

[mechanism]
Eo + ATP <====> Eo.ATP : kat kdt
Eo.ATP <=====> Ec.ATP : koc kco
Ec.ATP -----> Ec.ADp : khy
Ec.ADp <=====> Eo + ADP : kdd kad

PbT -----> PbT* : kbt
PbD -----> PbD* : kbd

[data]
directory      ./users/EDU/DE/MPIMF/Leskovar_A/...
extension      txt
plot           logarithmic
monitor        Eo, Eo.ATP, Ec.ATP, Ec.ADp, PbT*, PbD*
show concentrations of these species over time
```

**photo-bleaching  
is a first-order process**

Leskovar et al. (2008) *J. Biol. Chem.* **283**, 11677-688

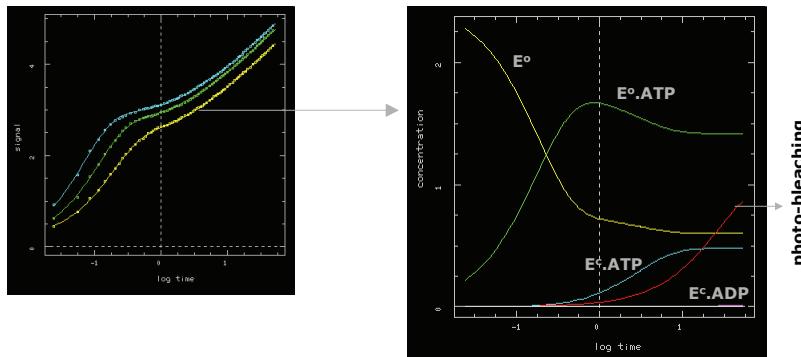


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## DynaFit Example 1: Trap1 – species concentrations

USEFUL WAY TO GAIN INSIGHT INTO THE MECHANISM



Leskovar et al. (2008) *J. Biol. Chem.* **283**, 11677-688

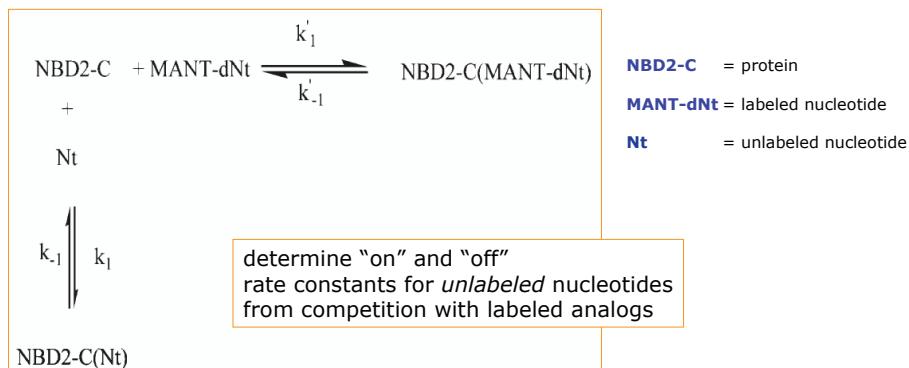


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## DynaFit Example 2: Nucleotide binding to ClpB

FROM THE SAME LAB (MAX-PLANCK INSTITUTE FOR MEDICAL RESEARCH, HEIDELBERG)



Werbeck et al. (2009)  
 "Nucleotide binding and allosteric modulation of the second AAA+ domain of ClpB probed by transient kinetic studies"  
*Biochemistry* **48**, 7240-7250

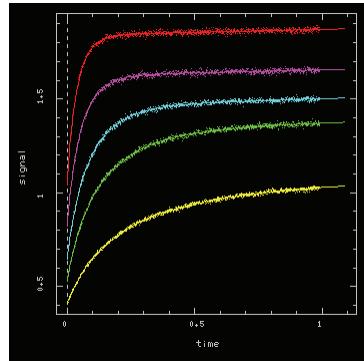


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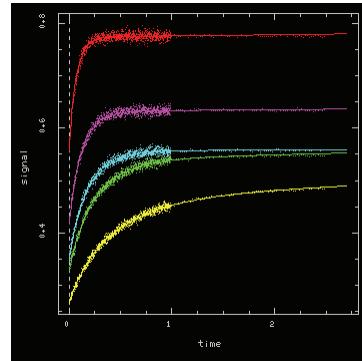
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## DynaFit Example 2: Nucleotide binding to ClpB - data

AGAIN COMBINE TWO DIFFERENT EXPERIMENTS (ONLY "LABELED" NUCLEOTIDE HERE)



- variable [ADP\*]
- constant [ClpB]



- constant [ADP\*]
- variable [ClpB]

Werbeck et al. (2009) *Biochemistry* **48**, 7240-7250



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## DynaFit Example 2: Nucleotide binding to ClpB script

THE DEVIL IS ALWAYS IN THE DETAIL

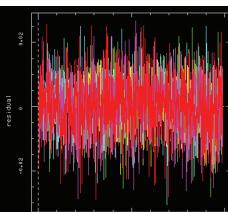
```
[task]
  data = progress
  task = fit
  model = simplest

[mechanism]
  P + mADP <=> P.mADP : k1  k-1
  ---> drift : v

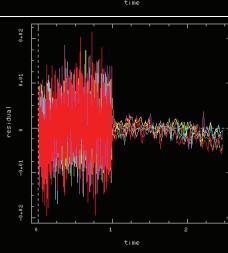
[constants]
  k1 = 5 ?
  k-1 = 0.1 ?
  v = 0.1 ?
```

"drift in the machine"

### Residuals



- variable [ADP\*]
- constant [ClpB]



- constant [ADP\*]
- variable [ClpB]

Werbeck et al. (2009) *Biochemistry* **48**, 7240-7250

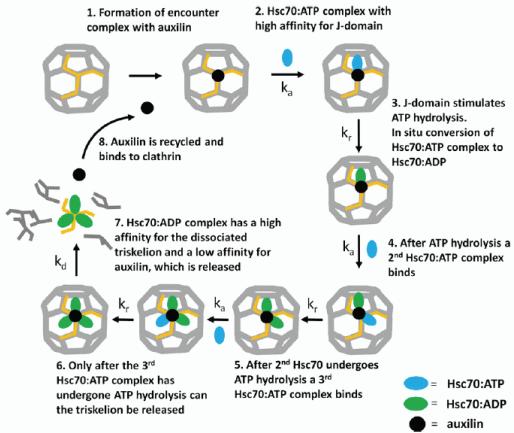


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## DynaFit Example 3: Clathrin uncoating kinetics

IN COLLABORATION WITH GUS CAMERON (BRISTOL)



**Rothnie et al. (2011)**

"A sequential mechanism for clathrin cage disassembly by 70-kDa heat-shock cognate protein (Hsc70) and auxilin"  
*Proc. Natl. Acad. Sci USA* **108**, 6927–6932



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## DynaFit Example 3: Clathrin uncoating - script

### MODEL DISCRIMINATION ANALYSIS

```
[task]
task = fit
data = progress
model = AAAH ?
[mechanism]
CA + T --> CAT : ka
CAT + T --> CATT : ka
CATT + T --> CATTT : ka
CATTT --> CADD : kr
CADD --> Prods : kd
...
[task]
task = fit
data = progress
model = AHAHAH ?
[mechanism]
CA + T --> CAT : ka
CAT --> CAD + Pi : kr
CAD + T --> CADT : ka
CADT --> CADD + Pi : kr
CADD + T --> CADDT : ka
CADDT --> CADD + Pi : kr
CADD --> Prods : kd
...
```

- Arbitrary number of models to compare
- Model selection based on two criteria:
  - **Akaike Information Criterion (AIC)**
  - F-test for nested models
- **Extreme caution** is required for interpretation
  - Both AIC and F-test are far from perfect
  - Both are based on many assumptions
  - One must use common sense
  - Look at the results only for *guidance*

**Rothnie et al. (2011)** *Proc. Natl. Acad. Sci USA* **108**, 6927–6932

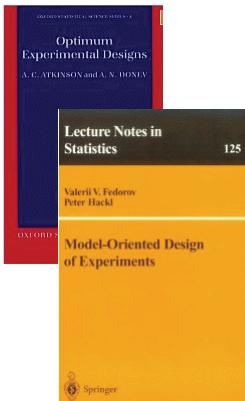


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## Optimal Experimental Design: Books

DOZENS OF BOOKS



- *Fedorov, V.V. (1972)*  
"Theory of Optimal Experiments"
- *Fedorov, V.V. & Hackl, P. (1997)*  
"Model-Oriented **Design** of Experiments"
- *Atkinson, A.C & Donev, A.N. (1992)*  
"Optimum Experimental **Designs**"
- *Endrenyi, L., Ed. (1981)*  
"**Design** and Analysis of **Enzyme** and Pharmacokinetics Experiments"



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## Optimal Experimental Design: Articles

HUNDREDS OF ARTICLES, INCLUDING IN ENZYMOLOGY

*J. theor. Biol.* (1981) **90**, 241–263

### Optimal Design of Experiments for the Estimation of Precise Hyperbolic Kinetic and Binding Parameters

LASZLO ENDRENYI AND FUNG-YEE CHAN

ANALYTICAL BIOCHEMISTRY **184**, 172–183 (1990)

### DESIGN: Computerized Optimization of Experimental Design for Estimating $K_d$ and $B_{\max}$ in Ligand Binding Experiments

G. Enrico Rovati,<sup>1</sup> David Rodbard, and Peter J. Munson<sup>2</sup>



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## Some theory: Fisher information matrix

"D-OPTIMAL" DESIGN: MAXIMIZE DETERMINANT OF THE FISHER INFORMATION MATRIX

$$\text{Fisher information matrix: } (\mathcal{I}(\theta))_{i,j} = -E \left[ \frac{\partial^2}{\partial \theta_i \partial \theta_j} \ln f(X; \theta) \middle| \theta \right]$$

**EXAMPLE:** Michaelis-Menten kinetics

**Model:**

$$v = V \frac{[S]}{[S] + K}$$

two parameters ( $M=2$ )

**Derivatives:** ("sensitivities")

$$s_V \equiv \frac{\partial v}{\partial V} = \frac{[S]}{[S] + K}$$

**Design:** four concentrations ( $N=4$ )

$$[S]_1, [S]_2, [S]_3, [S]_4$$

$$s_K \equiv \frac{\partial v}{\partial K} = -V \frac{[S]}{([S] + K)^2}$$



## Some theory: Fisher information matrix (contd.)

"D-OPTIMAL" DESIGN: MAXIMIZE DETERMINANT OF THE FISHER INFORMATION MATRIX

**Approximate** Fisher information matrix ( $M \times M$ ):

$$F_{i,j} = \sum_{k=1}^N s_i([S]_k) s_j([S]_k)$$

**EXAMPLE:** Michaelis-Menten kinetics

$$\mathbf{F} = \begin{pmatrix} \sum_{k=1}^N \left( \frac{[S]_k}{[S]_k + K} \right)^2 & \sum_{k=1}^N \left( -V \frac{[S]_k}{([S]_k + K)^2} \right) \left( \frac{[S]_k}{[S]_k + K} \right) \\ \sum_{k=1}^N \left( -V \frac{[S]_k}{([S]_k + K)^2} \right) \left( \frac{[S]_k}{[S]_k + K} \right) & \sum_{k=1}^N \left( -V \frac{[S]_k}{([S]_k + K)^2} \right)^2 \end{pmatrix}$$

$$\det \mathbf{F} = F_{11}F_{22} - F_{12}F_{21} \quad \text{determinant}$$

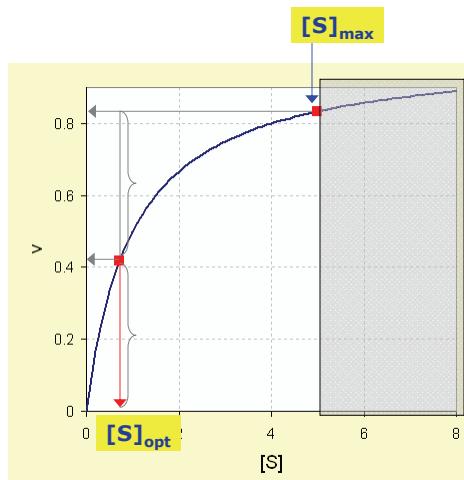
**"D-Optimal" Design:**

Maximize determinant of  $\mathbf{F}$  over design points  $[S]_1, \dots, [S]_4$ .



## Optimal Design for Michaelis-Menten kinetics

DUGGLEBY, R. (1979) J. THEOR. BIOL. 81, 671-684



**Model:**

$$v = V \frac{[S]}{[S] + K}$$

$$V = 1$$

$$K = 1$$

$$[S]_{\text{opt}} = \frac{[S]_{\text{max}} K}{[S]_{\text{max}} + 2(K)}$$

**K is assumed  
to be known !**

## Optimal Design: Basic assumptions

OPTIMAL DESIGN FOR ESTIMATING PARAMETERS IN THE GIVEN MODEL

TWO FAIRLY STRONG ASSUMPTIONS:

1. Assumed mathematical **model is correct** for the experiment
2. A fairly **good estimate** already exists for the model **parameters**



"Designed" experiments are most suitable for **follow-up** (verification) experiments.

## Optimal Experimental Design: Initial conditions

IN MANY **KINETIC** EXPERIMENTS THE OBSERVATION **TIME CANNOT BE CHOSEN**

### CONVENTIONAL EXPERIMENTAL DESIGN:

- Make an optimal choice of the **independent variable**:
  - Equilibrium experiments: **concentrations** of varied species
  - Kinetic experiments: **observation time**

### DYNAFIT MODIFICATION:

- Make an optimal choice of the **initial conditions**:
  - Kinetic experiments: **initial concentrations** of reactants

Assume that the **time points are given** by instrument setup.



## Optimal Experimental Design: DynaFit input file

**EXAMPLE: CLATHRIN UNCOATING KINETICS**

```
[task]
  task = design
  data = progress

[mechanism]
  CA + T -> CAT : ka
  CAT -> CAD + Pi : kr
  CAD + T -> CADT : ka
  CADT -> CADD + Pi : kr
  CADD + T -> CADDT : ka
  CADDT -> CADD + Pi : kr
  CADD -> Prods : kd

[constants]
  ka = 0.69 ?
  kr = 6.51 ?
  kd = 0.38 ?

[data]
  file run01 | concentration CA = 0.1, T = 1 ?? (0.001 .. 100)
  file run02 | concentration CA = 0.1, T = 1 ?? (0.001 .. 100)
  file run03 | concentration CA = 0.1, T = 1 ?? (0.001 .. 100)
  file run04 | concentration CA = 0.1, T = 1 ?? (0.001 .. 100)
  file run05 | concentration CA = 0.1, T = 1 ?? (0.001 .. 100)
  file run06 | concentration CA = 0.1, T = 1 ?? (0.001 .. 100)
  file run07 | concentration CA = 0.1, T = 1 ?? (0.001 .. 100)
  file run08 | concentration CA = 0.1, T = 1 ?? (0.001 .. 100)
```

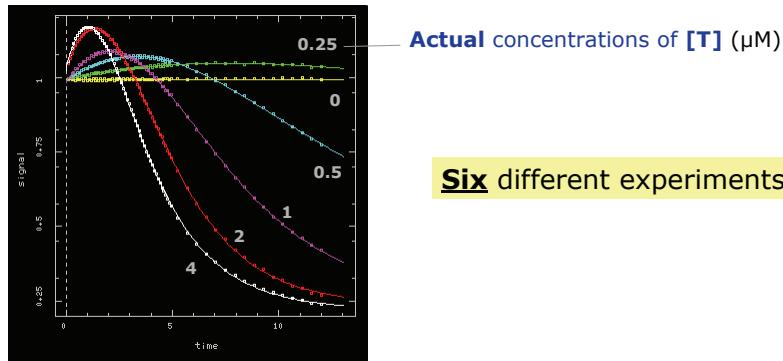
"Choose eight initial concentration of **T**  
such that the rate constants  $k_a$ ,  $k_r$ ,  $k_d$   
are determined most precisely."



## Optimal Experimental Design: Preliminary experiment

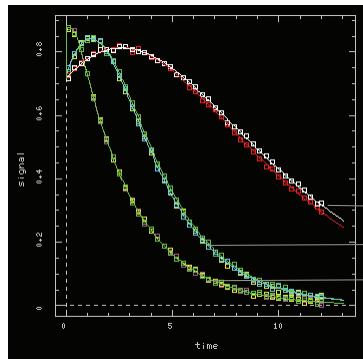
EXAMPLE: CLATHRIN UNCOATING KINETICS – ACTUAL DATA

Rothenie et al. (2011) Proc. Natl. Acad. Sci USA **108**, 6927–6932



## Optimal Experimental Design: DynaFit results

EXAMPLE: CLATHRIN UNCOATING KINETICS



Just **three** experiments would be sufficient for follow-up

## Optimal Experimental Design in DynaFit: Summary

NOT A SILVER BULLET !

- Useful for **follow-up (verification)** experiments only
  - Mechanistic model must be known already
  - Parameter estimates must also be known
- Takes a **very long time** to compute
  - Constrained global optimization: "Differential Evolution" algorithm
  - Clathrin design took 30-90 minutes
  - Many design problems take multiple hours of computation
- Critically depends on assumptions about **variance**
  - Usually we assume **constant variance** ("noise") of the signal
  - Must verify this by plotting **residuals against signal** (not the usual way)