


MASARYKOVA UNIVERZITA
 INNOVATION LECTURES
 (INNOIEC)

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Binding and Kinetics for Experimental Biologists
 Lecture 2
Evolutionary Computing: Initial Estimate Problem

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 WATERTOWN, MASSACHUSETTS, U.S.A.

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

 **EVROPSKÁ UNIE**  **esf**

 **MINISTERSTVO ŠKOLSTVÍ,
MLÁDEŽE A TĚLOVÝCHOVY**

 **OP Vzdělávání
pro konkurenční
schopnost**

INVESTICE DO ROZVOJE VZDĚLÁVÁNÍ



Lecture outline

- **The problem:**

Fitting nonlinear data usually requires an [initial estimate](#) of model parameters. This initial estimate must be [close enough](#) to the “true” values.

- **The solution:**

Use a data-fitting method that [does not depend](#) on initial estimates.

- **An implementation:**

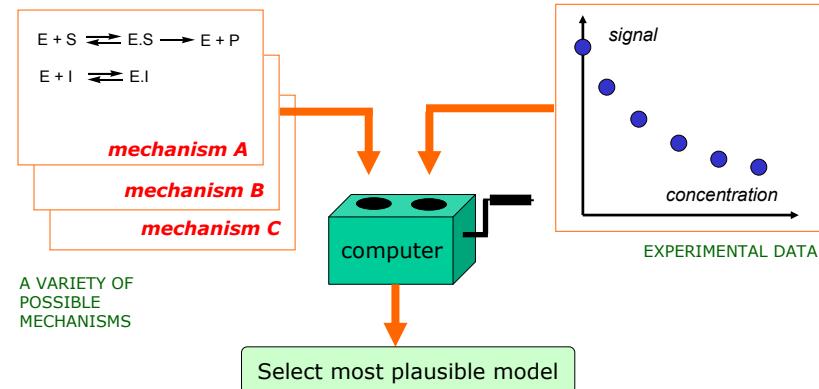
The [Differential Evolution](#) algorithm (Price *et al.*, 2005).

- **An example:**

Kinetics of [forked DNA](#) binding to the protein-protein complex formed by DNA-polymerase [sliding clamp \(gp45\)](#) and [clamp loader \(gp44/62\)](#).

The ultimate goal of analyzing kinetic / binding data

SELECT AMONG POSSIBLE MOLECULAR MECHANISMS

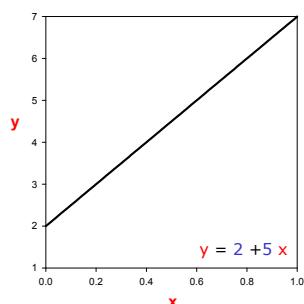


Most models in natural sciences are nonlinear

LINEAR VS. NONLINEAR MODELS

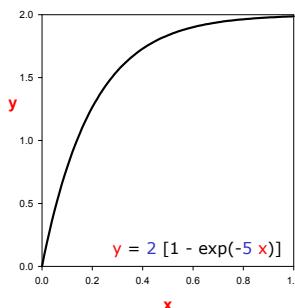
Linear

$$y = A + kx$$



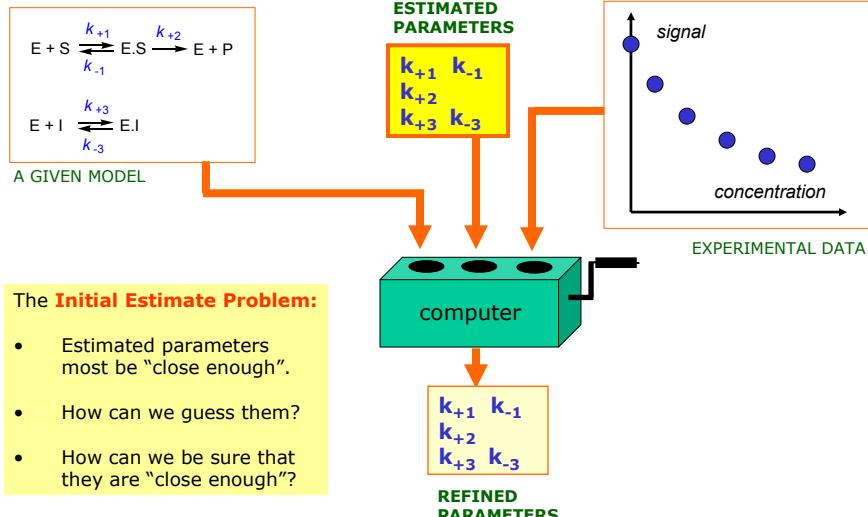
Nonlinear

$$y = A [1 - \exp(-kx)]$$

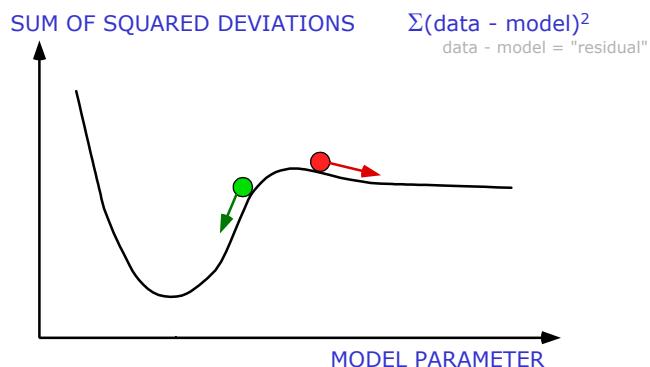


We need initial estimates of model parameters

NONLINEAR MODELS REQUIRE INITIAL ESTIMATES OF PARAMETERS



The crux of the problem: Finding *global* minima



- Least-squares fitting **only** goes "downhill"
- **How do we know where to start?**

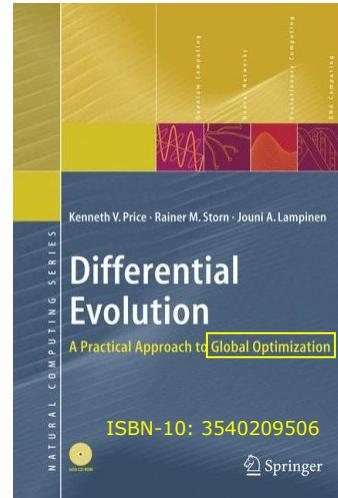


Charles Darwin to the rescue

BIOLOGICAL EVOLUTION IMITATED IN "DE"



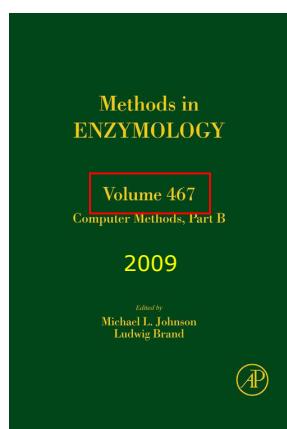
Charles Darwin (1809-1882)



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Specialized numerical software: *DynaFit*



CHAPTER TEN

DYNAFIT—A SOFTWARE PACKAGE FOR ENZYMOLOGY

Petr Kuzmič

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

DynaFit implements the
Differential Evolution algorithm
for global sum-of-squares minimization.

Kuzmic (2009) *Meth. Enzymol.*, **467**, 247-280



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Biological metaphor: "Gene, allele"

BIOLOGY

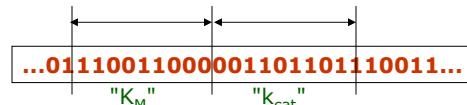
gene



four-letter alphabet
variable length

COMPUTER

- sequence of bits representing a number



- two letter alphabet
- fixed length (16 or 32 bits)

"Chromosome, genotype, phenotype"

BIOLOGY

genotype

...A[AGTCGGTT]CG[GAAGTCGGTT]TA...

↓
keratin ↓
oncoprotein

phenotype

COMPUTER

- particular combination of all model parameters

011010110110|011110011010|00111101101

V_{max}=1.23

K_M=4.56

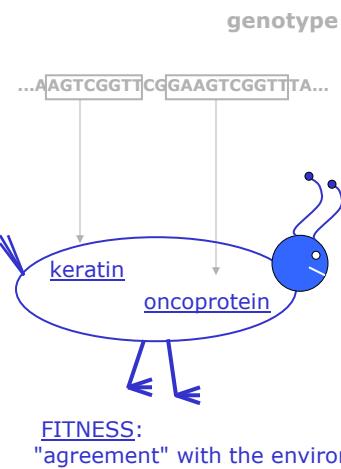
K_{is}=78.9

full set of parameters

$$v = V_{\max} \frac{[S]/K_M}{1 + [S]/K_M + [S]^2 / K_M K_{is}}$$

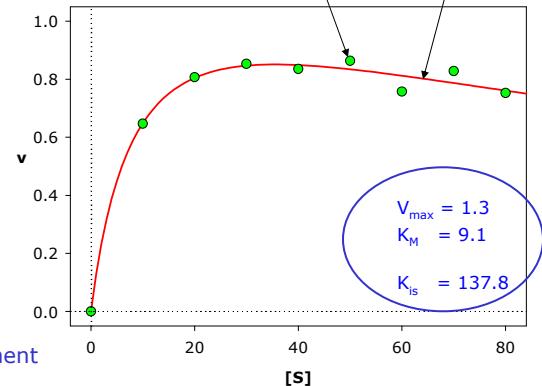
"Organism, fitness"

BIOLOGY



COMPUTER

- **FITNESS:** agreement between the data and the model



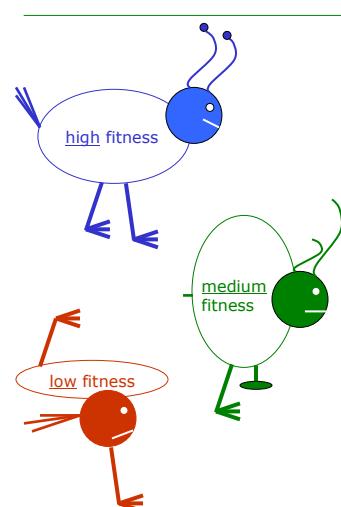
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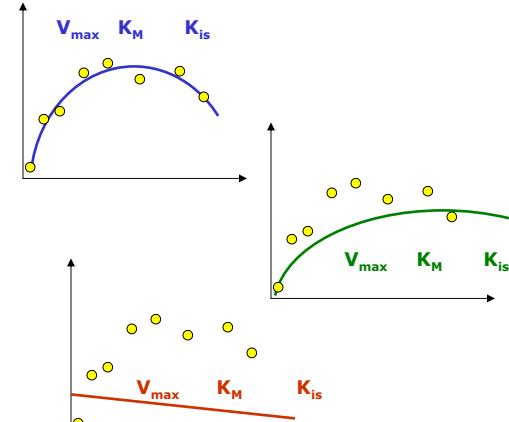
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"Population"

BIOLOGY



COMPUTER

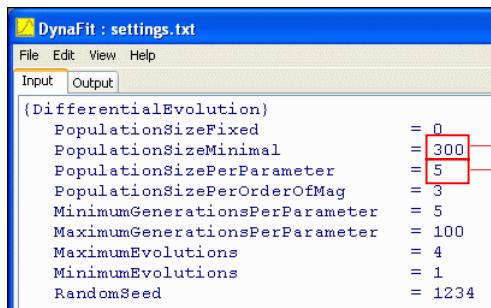


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DE Population size in DynaFit



DynaFit : settings.txt

```

File Edit View Help
Input Output
(DifferentialEvolution)
PopulationSizeFixed = 0
PopulationSizeMinimal = 300
PopulationSizePerParameter = 5
PopulationSizePerOrderOfMag = 3
MinimumGenerationsPerParameter = 5
MaximumGenerationsPerParameter = 100
MaximumEvolutions = 4
MinimumEvolutions = 1
RandomSeed = 1234

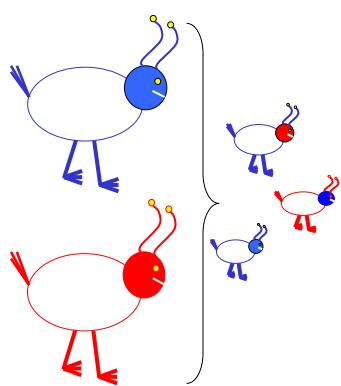
```

number of population members **per optimized model parameter**

number of population members **per order of magnitude**

"Sexual reproduction, crossover"

BIOLOGY



COMPUTER

random crossover point

mother

01101011011001111001101 00011111011

father

01101011011001111001101 11100011011

"sexual mating"
probability p_{cross}

child

011010110110 | 01111001101 | 11100011011

V_{max}

K_M

K_{ls}

"Mutation, genetic diversity"

BIOLOGY	COMPUTER												
	<p>father</p> <table border="1"> <tr> <td>01101011011</td> <td>001111001101</td> <td>11100011011</td> </tr> <tr> <td>V_{max}</td> <td>K_M</td> <td>K_{is}</td> </tr> </table> <p>mutation</p> <p>↓</p> <p>mutant father</p> <table border="1"> <tr> <td>11100111011</td> <td>001011010101</td> <td>11001011001</td> </tr> <tr> <td>$V_{max}^{(*)}$</td> <td>$K_M^{(*)}$</td> <td>$K_{is}^{(*)}$</td> </tr> </table>	01101011011	001111001101	11100011011	V_{max}	K_M	K_{is}	11100111011	001011010101	11001011001	$V_{max}^{(*)}$	$K_M^{(*)}$	$K_{is}^{(*)}$
01101011011	001111001101	11100011011											
V_{max}	K_M	K_{is}											
11100111011	001011010101	11001011001											
$V_{max}^{(*)}$	$K_M^{(*)}$	$K_{is}^{(*)}$											

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"Mutation, genetic diversity"

THE "DIFFERENTIAL" IN DIFFERENTIAL EVOLUTION ALGORITHM - **STEP 1**

Compute difference between two randomly chosen "auntie" phenotypes

aunt #1

01101011011	001111001101	11100011011
$V_{max}^{(1)}$	$K_M^{(1)}$	$K_{is}^{(1)}$

subtract

aunt #2

11100111011	001011010101	11001011001
$V_{max}^{(2)}$	$K_M^{(2)}$	$K_{is}^{(2)}$

↓

aunt #2 minus aunt #1

11100111011	001011010101	11001011001
$V_{max}^{(2)} - V_{max}^{(1)}$	$K_M^{(2)} - K_M^{(1)}$	$K_{is}^{(2)} - K_{is}^{(1)}$

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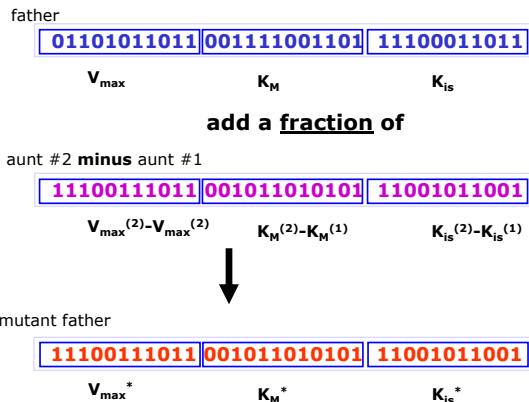
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"Mutation, genetic diversity"

THE "DIFFERENTIAL" IN DIFFERENTIAL EVOLUTION ALGORITHM - STEP 2

Add **weighted** difference between two "uncle" phenotypes to "father"



"Mutation, genetic diversity"

THE "DIFFERENTIAL" IN DIFFERENTIAL EVOLUTION ALGORITHM

EXAMPLE: Michaelis-Menten equation $v = V_{\max} \frac{[S]}{[S] + K_M}$

$$\text{"father"} \quad \text{"aunt 1"} \quad \text{"aunt 2"} \\ \downarrow \quad \downarrow \quad \downarrow \\ \text{"mutant father"} \longrightarrow K_M^* = K_M + F \times (K_M^{(1)} - K_M^{(2)}) \\ \uparrow \\ \text{weight (fraction)} \\ \text{mutation rate}$$

DE “undocumented” settings in DynaFit

DynaFit : settings-HIDDEN.txt

```

File Edit View Help
Input Output
{DifferentialEvolution}
CombineGenerations = n
ReplaceStragglersPercent = 0
Constrained = y
Strategy = 3
Weight = 0.8
Crossover = 1
Jitter = 0.01
Distribution = uniform
AddUserEstimate = n
Scaling = logarithmic
NormalDeviation = 0.5
ExponentialLambda = 2
ReportFrequency = 1
TestParameterRange = y
TestParameterRangeAll = y
TestParameterRangeFull = n
StopParameterRange = 0.01
TestCostFunctionRange = y
StopCostFunctionRange = 0.000001
TestCostFunctionChange = y
StopCostFunctionChange = 0.000001
TestCostFunctionChangeCount = 10

```

six different mutation strategies
fractional difference used in mutations
 $K_M^* = K_M + F \times (K_M^{(1)} - K_M^{(2)})$

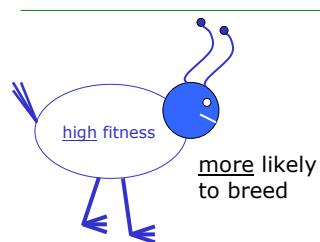
probability that “child” inherits “father’s” genes, not “mother’s” genes

These DE tuning constants are “undocumented” in the DynaFit distribution.

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“Selection”

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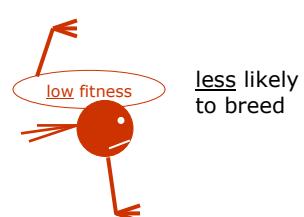


COMPUTER

low sum of squares

0110101101	100111100110	00011111011
V_{max}	K_M	K_{ls}

more likely to be carried to the next generation



high sum of squares

00000000001	111111111111	000000000000
V_{max}	K_M	K_{ls}

less likely to be carried to the next generation

Basic Differential Evolution Algorithm - Summary

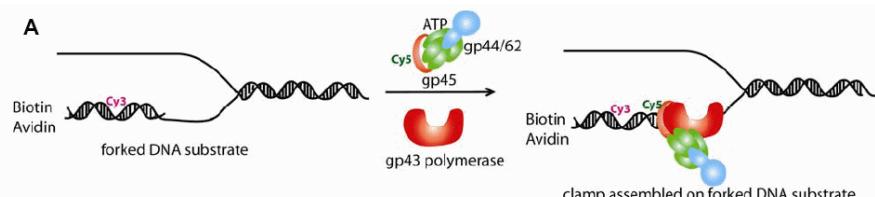
1 Randomly create the initial population (size N)

Repeat until almost all population members have very high fitness:

- 2 Evaluate fitness: sum of squares for all population members
- ↓ 3 Mutation: random gene modification (mutate *father*, weight F)
- ↓ 4 Sexual reproduction: random crossover with probability P_{cross}
- ↓ 5 Natural selection: keep *child* in gene pool if more fit than *mother*

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.

see Lecture 1 for details

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

Example: DynaFit script for Differential Evolution

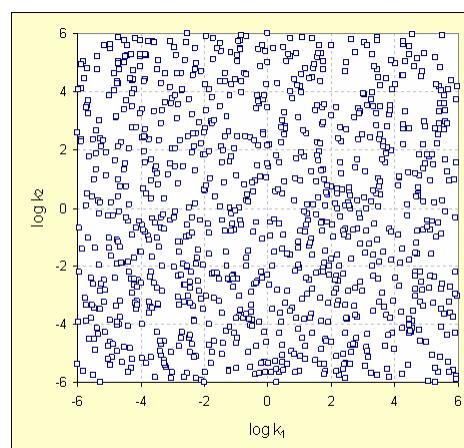
INSERT A SINGLE LINE IN THE [TASK] SECTION

```
DynaFit : fit-004.txt
File Edit View Help
Input Output
[task]
task = fit
data = progress
algorithm = differential-evolution
[mechanism]
DNA + Clamp.Loader <==> Complex : kon koff
[constants]
kon = 1 ?
koff = 1 ?
[responses]
Complex = 1 ? (0.01 .. 100)
[data]
file ./courses/bkeh/lec-1/a+b/data/d1-edit.txt
offset 0.3 ? (0.2 .. 0.4)
```



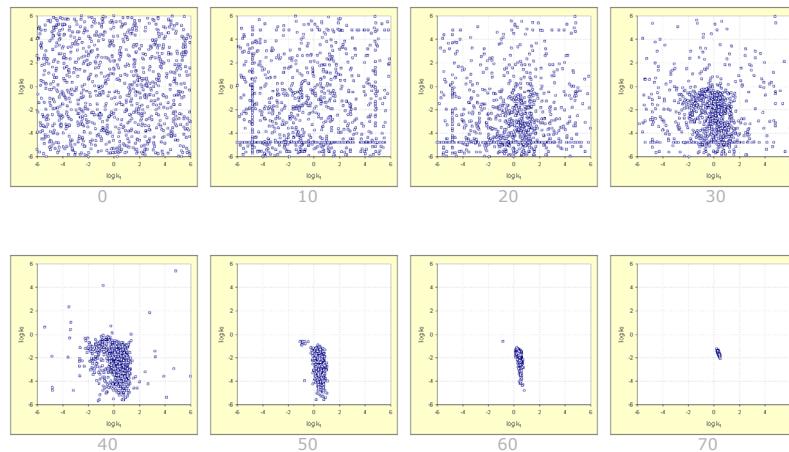
Example: Initial population

BOTH RATE CONSTANTS SPAN TWELVE ORDERS OF MAGNITUDE



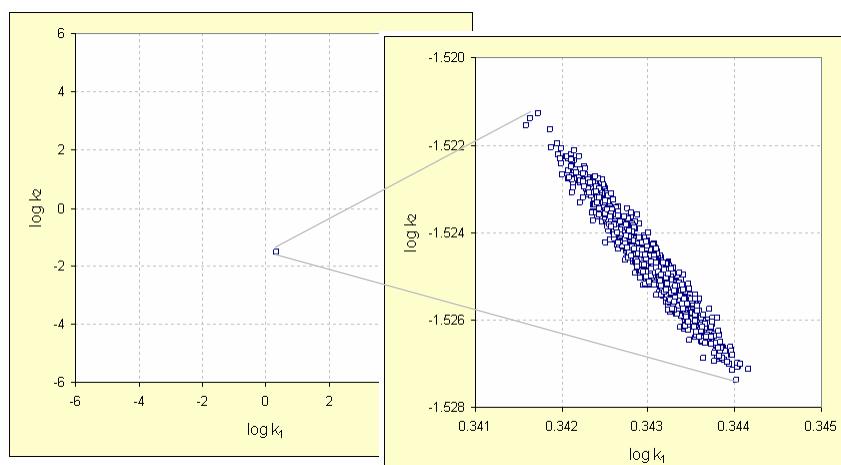
Example: The evolutionary process

SNAPSHOTS OF k_1 / k_2 CORRELATION DIAGRAM - SPACED BY 10 "GENERATIONS"



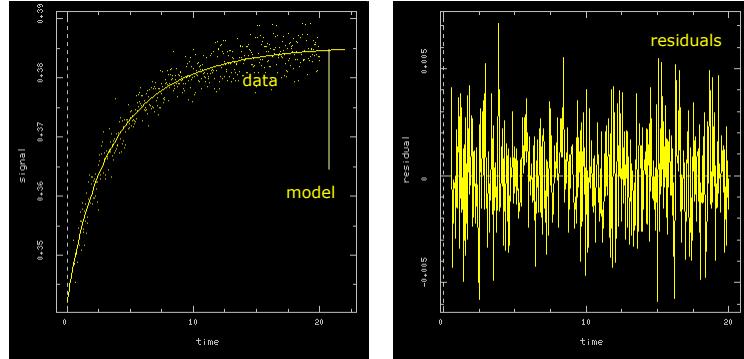
Example: Final population

BOTH RATE CONSTANTS SPAN AT MOST $\pm 30\%$ RANGE RELATIVE TO NOMINAL VALUE



Example: The “fittest” member of final population

THIS IS (PRESUMABLY) THE **GLOBAL MINIMUM** OF SUM-OF-SQUARES



Regression Summary

Differential Evolution (EvoDEPSL)

sum of squares **0.00230769**

Regression Summary

Levenberg-Marquardt Algorithm

sum of squares **0.00230769**

compare with
“good” estimate
from Lecture 1



Example: Comparison of DE and regular data fitting

DIFFERENTIAL EVOLUTION (DE) FOUND THE SAME FIT AS THE “GOOD” ESTIMATE

	initial estimate	sum of squares	relative sum of sq.	“best-fit” constants
lecture 1	“good” $k_1 = 1$ $k_2 = 1$	0.002308	1.00	$k_1 = 2.2 \pm 0.5$ $k_2 = 0.030 \pm 0.015$
	“bad” $k_1 = 100$ $k_2 = 0.01$	0.002354	1.02	$k_1 = 0.2 \pm 3.4$ $k_2 = 0.2 \pm 0.6$
1000 random estimates	$k_1 = 10^{-6} - 10^{+6}$ $k_2 = 10^{-6} - 10^{+6}$	0.002308	1.00	$k_1 = 2.2 \pm 0.5$ $k_2 = 0.030 \pm 0.015$



Significant disadvantage of DE: very slow

DYNAFIT CAN TAKE MULTIPLE DAYS TO RUN A COMPLEX PROBLEM

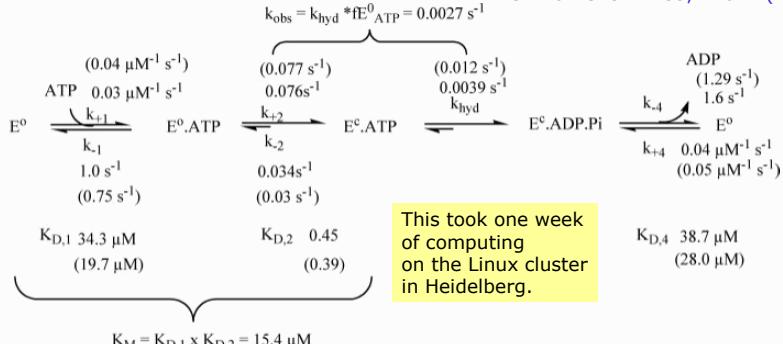
DynaFit 4.065 on DNA / clamp / clamp loader example:

algorithm	computation time	relative time
Levenberg-Marquardt with two restarts	0.88 sec	1
Differential Evolution with four restarts (population size: 1000)	12 min 31 sec	853



Example of Differential Evolution in DynaFit

J. Biol. Chem. 283, 11677 (2007)



THE JOURNAL OF BIOLOGICAL CHEMISTRY VOL. 283, NO. 17, pp. 11677–11688, April 25, 2008
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The ATPase Cycle of the Mitochondrial Hsp90 Analog Trap1*§

Received for publication, November 20, 2007, and in revised form, February 5, 2008. Published, JBC Papers in Press, February 20, 2008, DOI 10.1074/jbc.M709516200

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Example: Systematic scan of many initial estimates

CAREFUL! THIS IS FASTER THAN DIFFERENTIAL EVOLUTION BUT DOES NOT ALWAYS WORK

```
DynaFit : fit-007.txt
File Edit View Help
Input Output
[task]
task = estimate
data = progress

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

[constants]
kon = {0.001, 0.01, 0.1, 1, 10, 100, 1000} ?
koff = {0.001, 0.01, 0.1, 1, 10, 100, 1000} ?

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

ALGORITHM

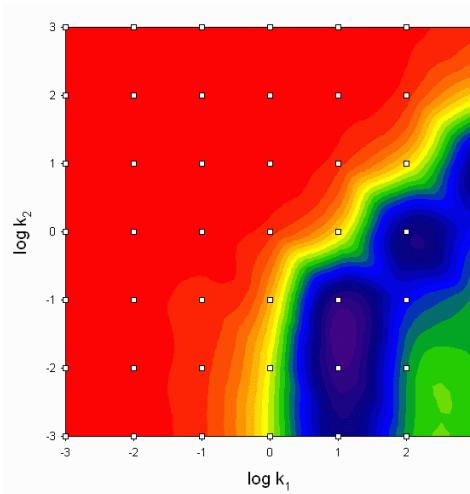
1. generate all possible combinations of rate constants
2. compute initial sum of squares for each combination
3. rank combinations by initial sum of squares
4. select the best **N** combinations
5. perform a full fit for those **N**
6. rank results again

$$7 \times 7 = 49 \text{ combinations of } k_{\text{on}} \text{ and } k_{\text{off}}$$



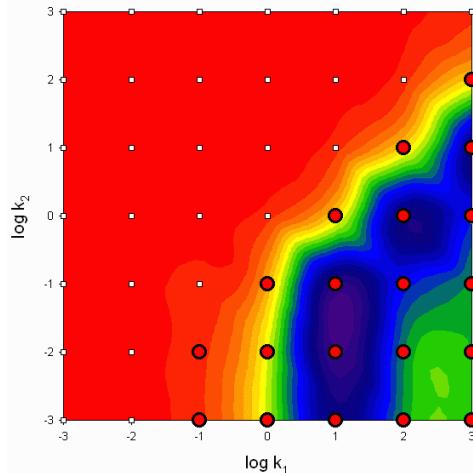
Example: Systematic scan – Phase 1

AFTER EVALUATING THE INITIAL SUM OF SQUARES FOR ALL 49 COMBINATIONS OF k_1 and k_2



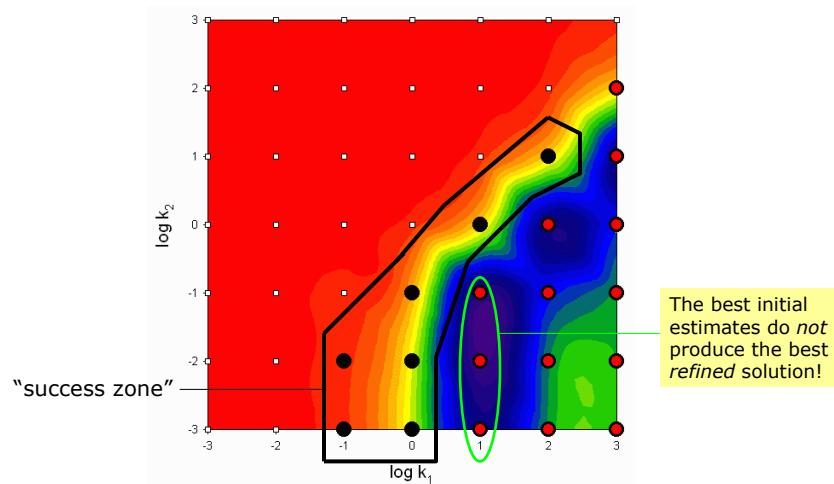
Example: Systematic scan – Phase 2

AFTER RANKING THE INITIAL ESTIMATES AND SELECTING 20 BEST ONES BY SUM OF SQUARES



Example: Systematic scan – Phase 3

AFTER PERFORMING FULL REFINEMENT FOR 20 BEST ESTIMATES OUT OF 49 TRIED



Summary and conclusions

1. Finding good-enough initial estimates is a very difficult problem.
2. One should use system-specific information as much as possible.
This includes using the literature and/or general principles for "intelligent" guesses.
3. Always use the "Try" method in DynaFit to display the initial fit.
Make sure that the initial estimate is at least approximately correct.
4. The Differential Evolution algorithm almost always helps.
However, it can be excruciatingly slow (running typically for multiple hours).
5. The systematic scan (`task = estimate`) sometimes helps.
However, the "best" initial estimates almost never produce the desired solution!
6. DynaFit is not a "silver bullet": You must still use your brain a lot.