

Universiteit Utrecht



# Evolutionary Systems Biology: multilevel evolution

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## Biology is changing fast....

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*One of the most fundamental patterns of scientific discovery is the revolution in thought that accompanies a new body of data*

Nigel Goldenfeld and Carl Woese  
Biology's next revolution  
Nature 445 (Jan 2007)

*Biology faces a quantum leap into the incomprehensible: the complexity of biology information processing networks will bring us in a counterintuitive world*

Paul Nurse  
Four great ideas in Biology  
*gene; evolution; cell; selforganization*  
webvideo Guardian (nov 2010)

**NEXT GENERATION BIOLOGY**

# Evolutionary Systems Biology

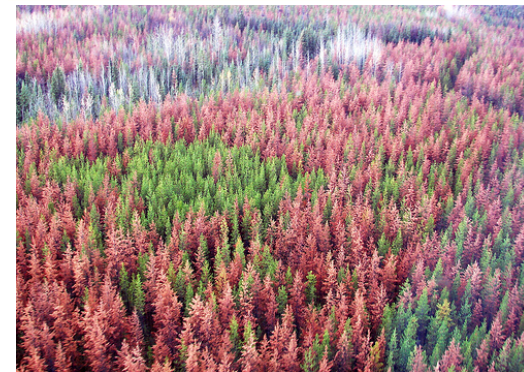
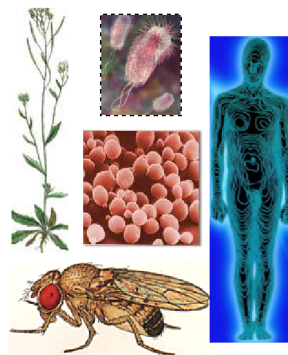
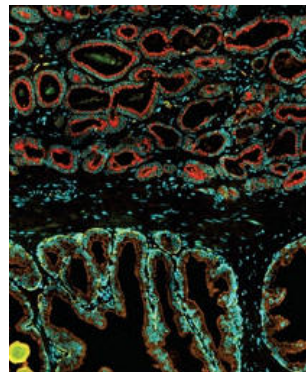
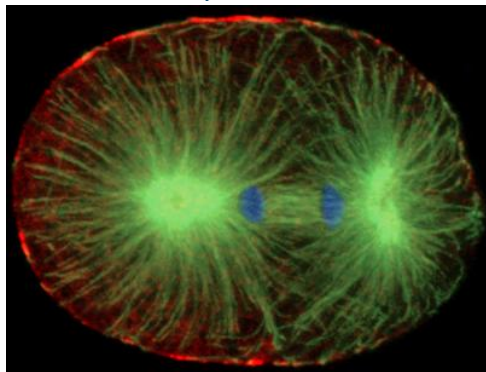
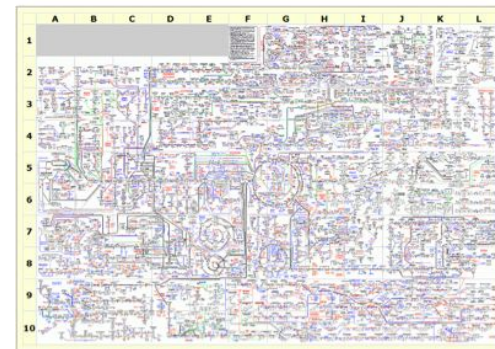
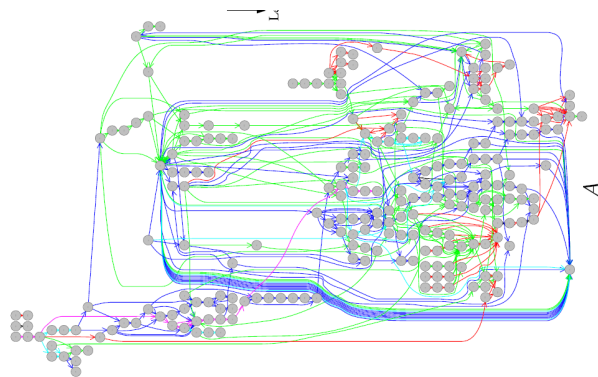
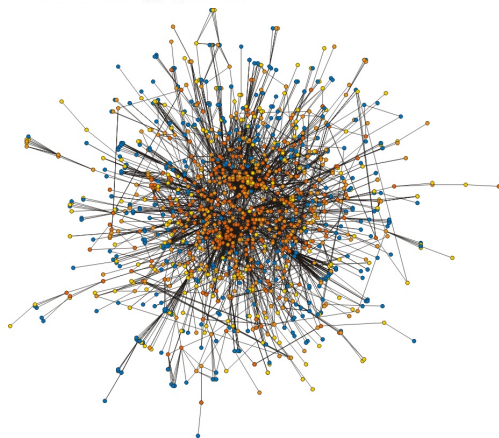
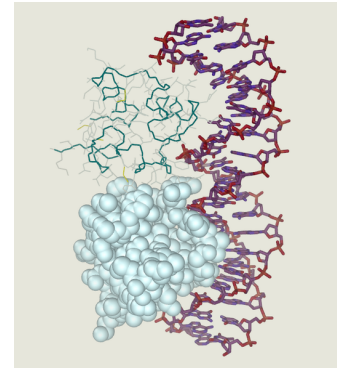
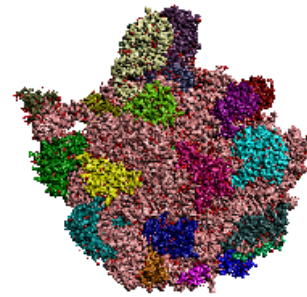
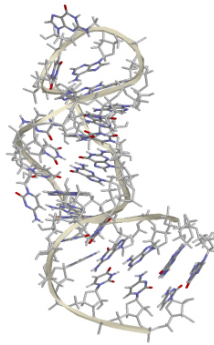
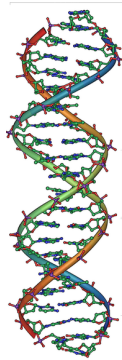
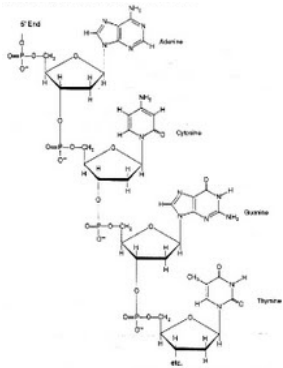
## multilevel evolution

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Using data 'tsunami' to reconstruct  
what DID happen in evolution  
*bioinformatic data analysis*

Using modeling to discover what DOES happen -  
through mutation/selection process  
*very often very counterintuitive*  
in multilevel setting

Experimental evolution + bioinformatic analysis of the data  
+ modeling



# Today

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*eco-evolutionary dynamics:*

emergence of new levels of selection  
through spatial pattern formation  
- evolution of cooperation/altruism

*Genome evolution*

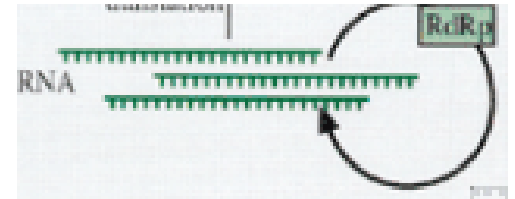
complex genotype-phenotype mapping help or hinder?  
- observed long term trends in evolution *generic* property of  
(multilevel) darwinian evolution?  
- evolution of evolvability.

# “Life is a self-sustained chemical system capable of undergoing Darwinian evolution” G.F. Joyce, 1994

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Simplest form: RNA-world

RNA both template and enzym



Joyce (and others) (back)evolve RNA world

e.g. evolve (engineer)

RNA which is RNA dependent RNA polymerase

(Wochner et al 2011: 95 nucleotides: not selfreplicating yet)

Here minimal model of minimal RNA world

study its dynamics *independent* from (bio)chemical properties

## minimal model of RNA world: RP system

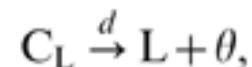
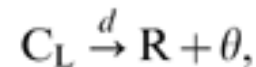
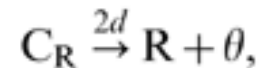
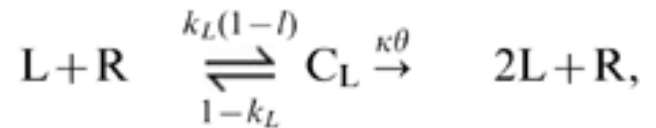
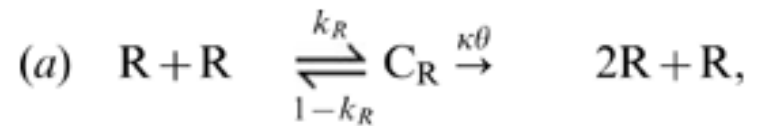
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R replicase

L other RNA (“parasitic”)  
 replicated when unfolded  
 ‘functional’ when folded  
 fraction  $l$  in folded state

*Evolve  $l$  and  $k_L$*

*(i.e. multiple (infinite) L species  
 one R species)*

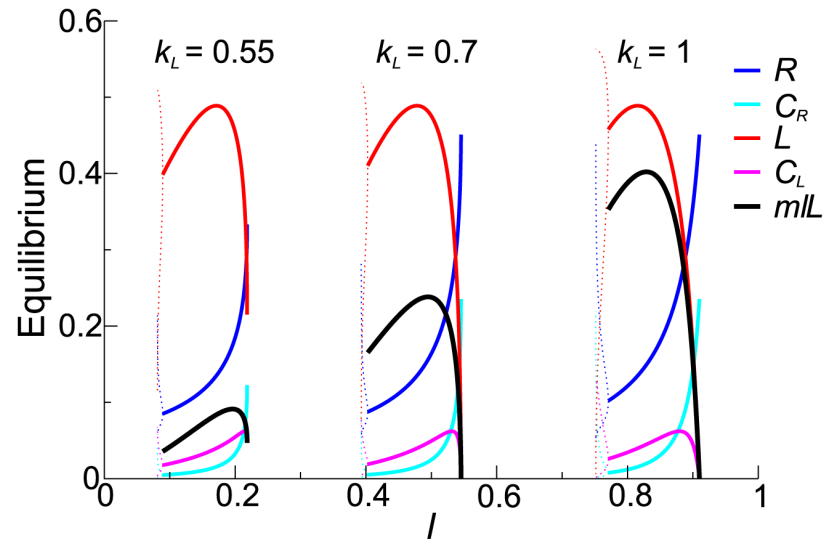


## Classical problem ODE model of RP system

evolutionary extinction because mutants of L which increase  $k_L$   
and/or decrease of I will outcompete L, and eventually  
outcompete R)

$$\begin{aligned}
 \dot{R} &= -2k_R R^2 + [2(1-k_R) + 3\kappa\theta + 2d]C_R - k_L RL \\
 &\quad + [(1-k_L) + \kappa\theta + d]C_L - dR, \\
 \dot{L} &= -k_L(1-I)RL + [(1-k_L) + 2\kappa\theta + d]C_L - dL, \quad (2) \\
 \dot{C}_R &= k_R R^2 - [(1-k_R) + \kappa\theta]C_R - 2dC_R, \\
 \dot{C}_L &= k_L(1-I)RL - [(1-k_L) + \kappa\theta]C_L - 2dC_L,
 \end{aligned}$$

$$k_R = .6$$



*intrinsic advantage of parasite (L)*



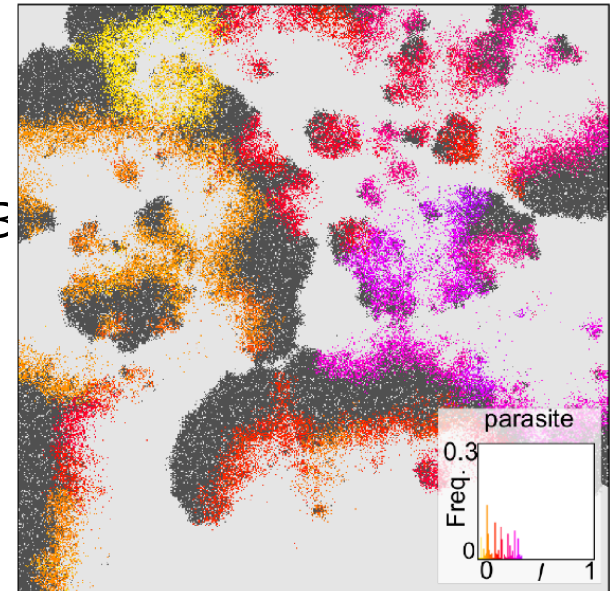
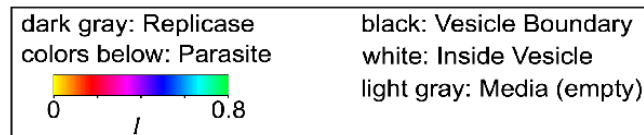
# better way to model RP system individual (particle) based, spatial model

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*better way to model evolving systems  
because very many types possible  
(less particles present than possible particles)*

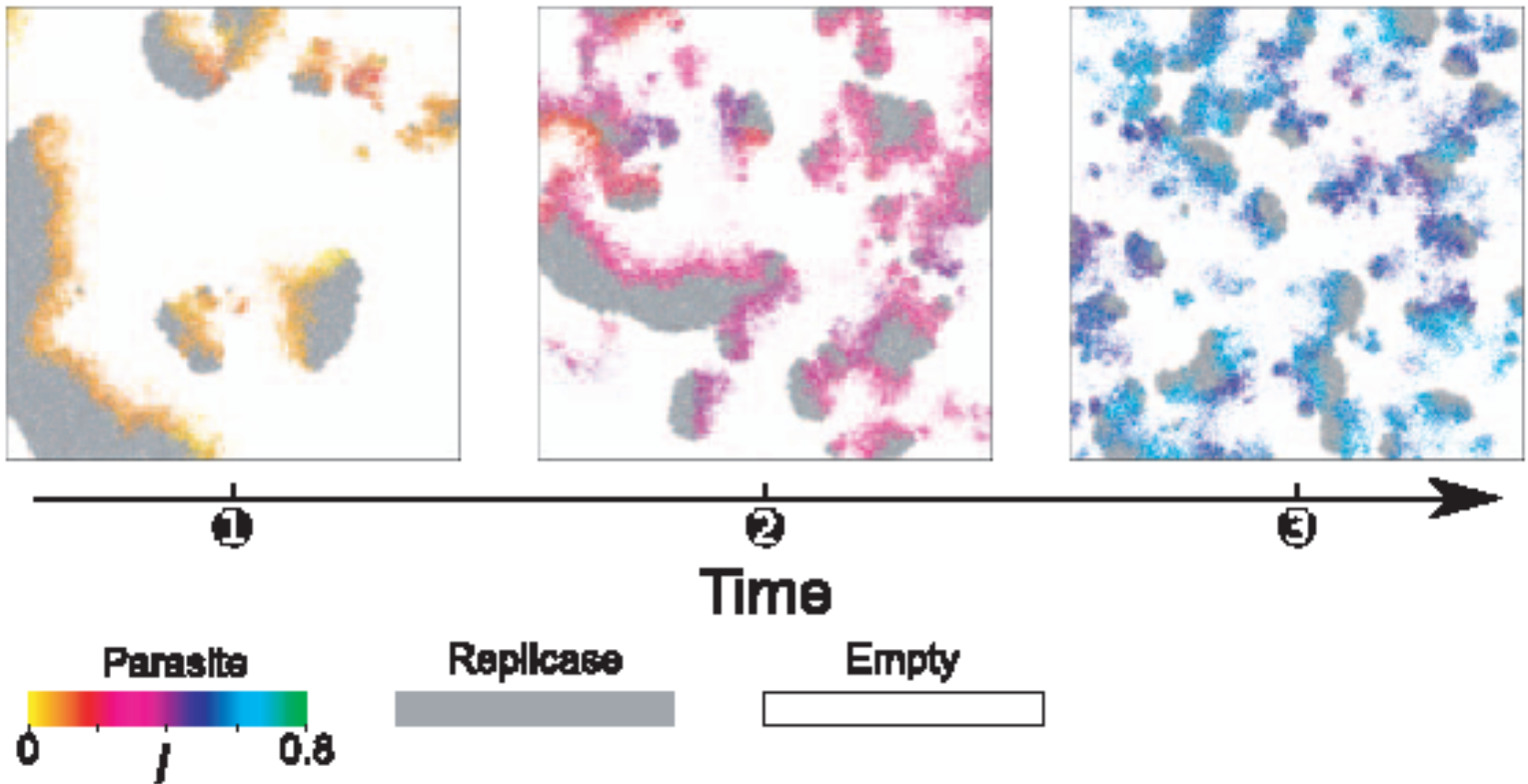
*better way because spatial setting more 'natural'*

*grid based stochastic CA model  
Monte Carlo step: N times  
choose random patch and random NE  
perform reaction or diffusion  
with prob. according to  
individual (evolving) parameters*



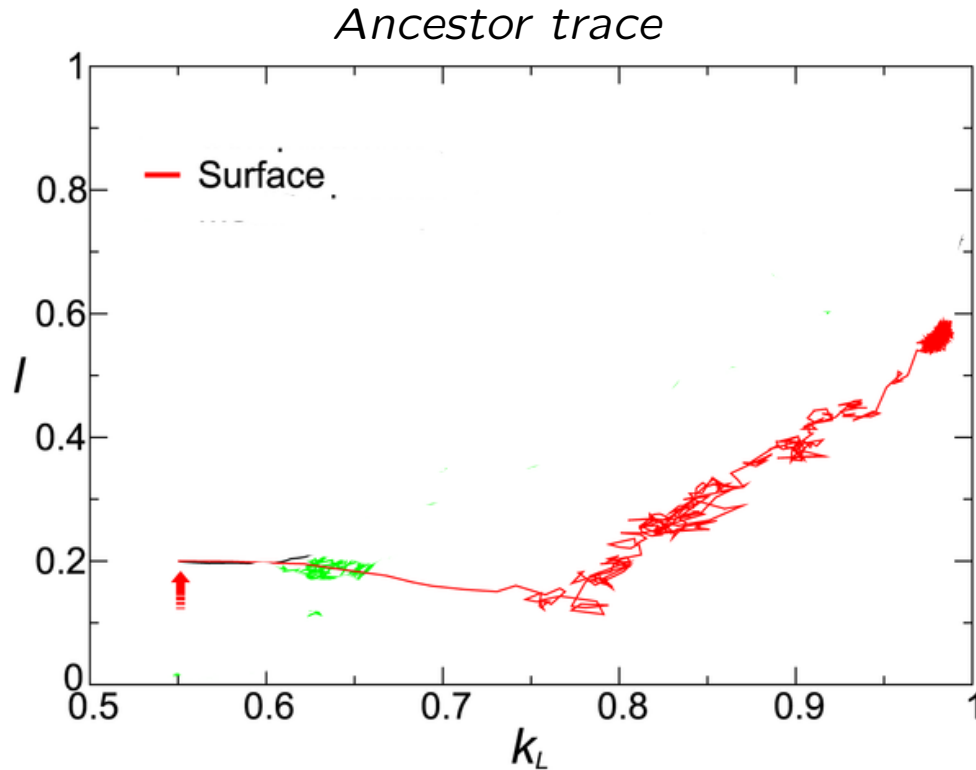
long term evolution: towards  
smaller waves more folded L

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Long term evolution (parameters)  
emergent 'trade-off'  $k_L$  and  $I$   
Maximizing  $I$  : potential 'new' function

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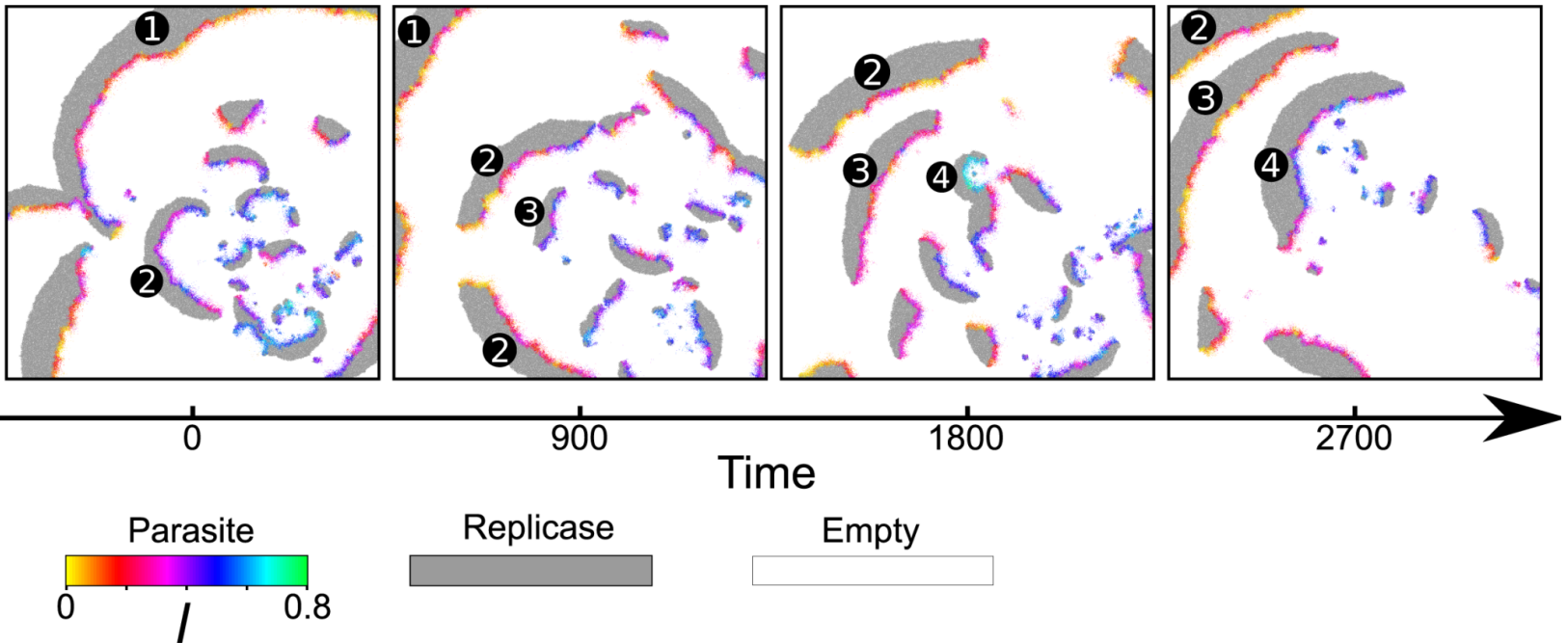
**WHY?**

*evolution of higher level entities*

# The waves of replicase and parasites are higher level “Darwinian” entities

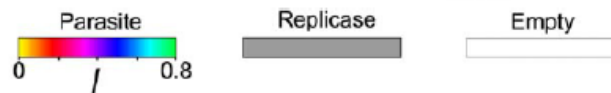
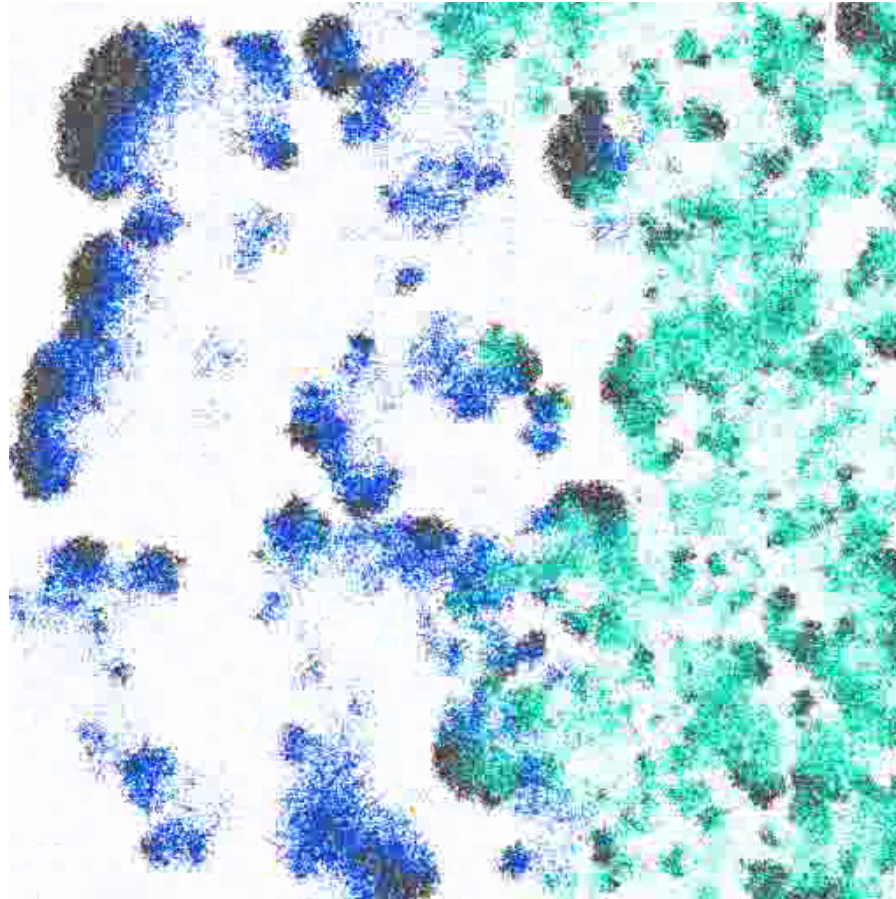
*Birth*  
*Maturation*  
*Death*  
*Mutation*  
*Selection*  
*Competing*

*maximizing birthrate*



# evolutionary attractor at “edge of chaos” (“border of order”)

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## 2 levels of Darwinian selection

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### Wave level selection

- Waves: long lived -  
( death not by parasites but by collision)
- Maximize Birthrate + growth rate of newborns
- Birthrate higher for high I ('escape')
- However higher birthrate – > more (smaller) waves
- – > increase collision! (= deathrate of waves))

### Individual level selection

- Within waves: parasites evolve towards 'nastiness' (low I)
- However viability maintained – – >  
“prudent” parasites
- because of higher level selection; which also
- 'frees' parasites to do other things (be folded)

*through parasites  
evolution of novel functionality*

**Not only “far away and long ago”  
Similar for Evolution and cooperation  
a classical problem in (too simplistic) evolutionary  
theory**

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*why not cheat?*

In simple ODE models cheaters destroy the cooperation

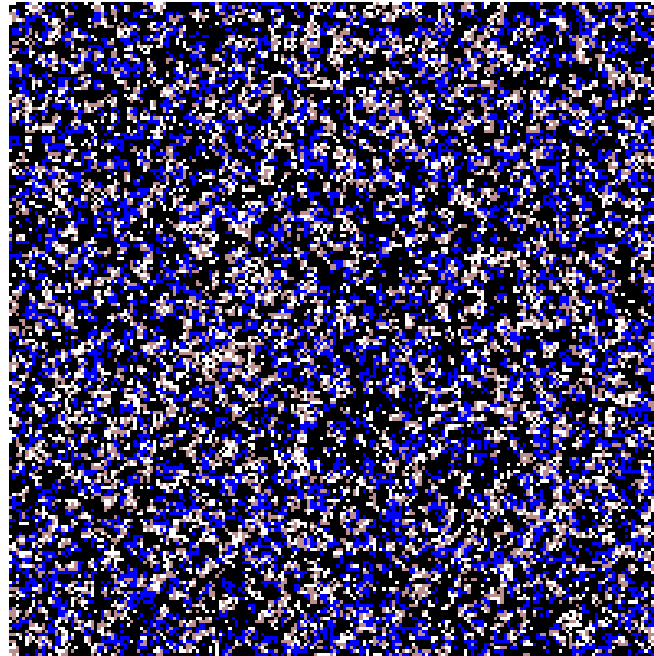
Nevertheless cooperation widespread  
e.g. figs/figwasps, dictyostelium, social insects ....

In spatial (CA) model cooperation does persist!

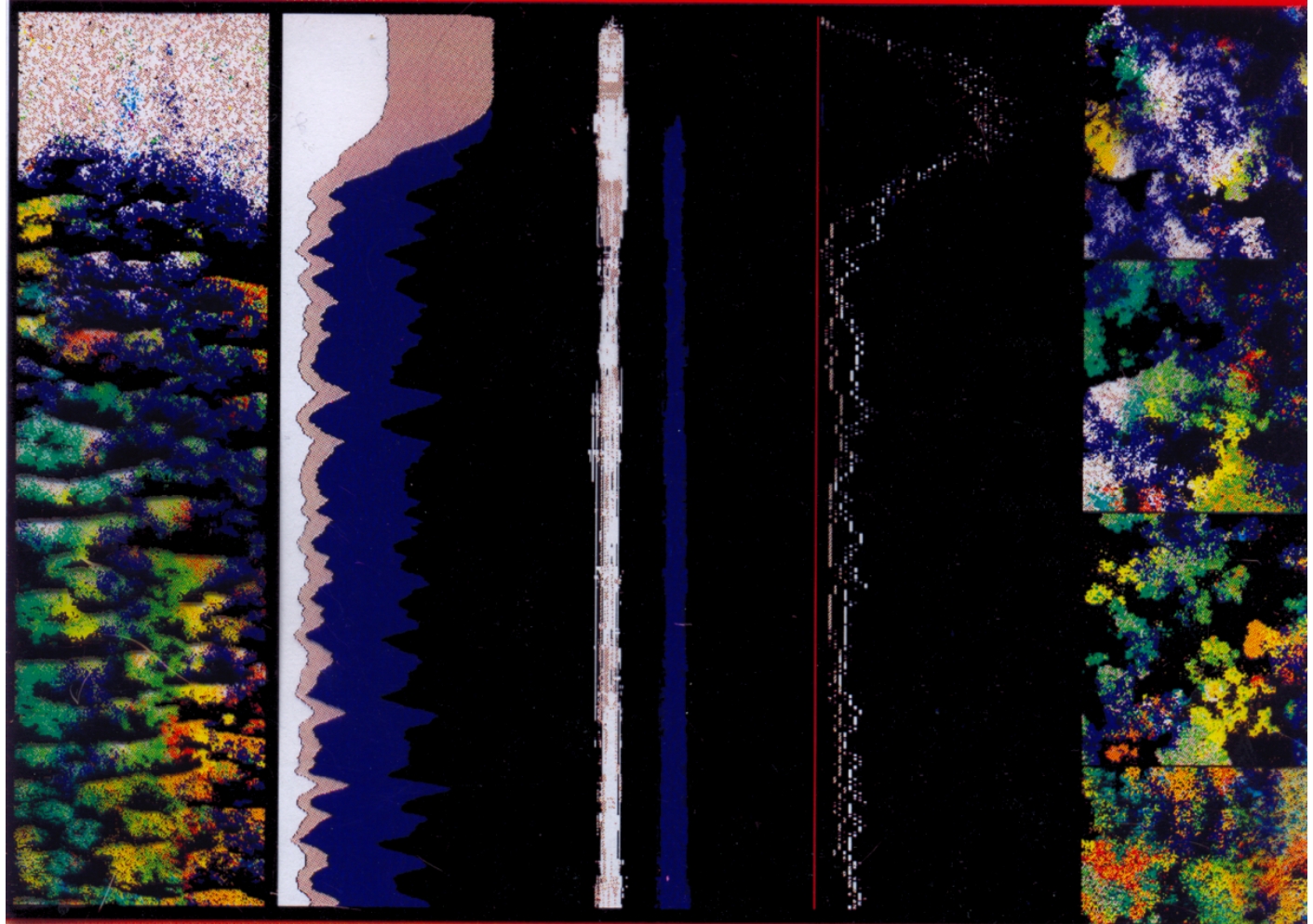
# Persistence of cooperation

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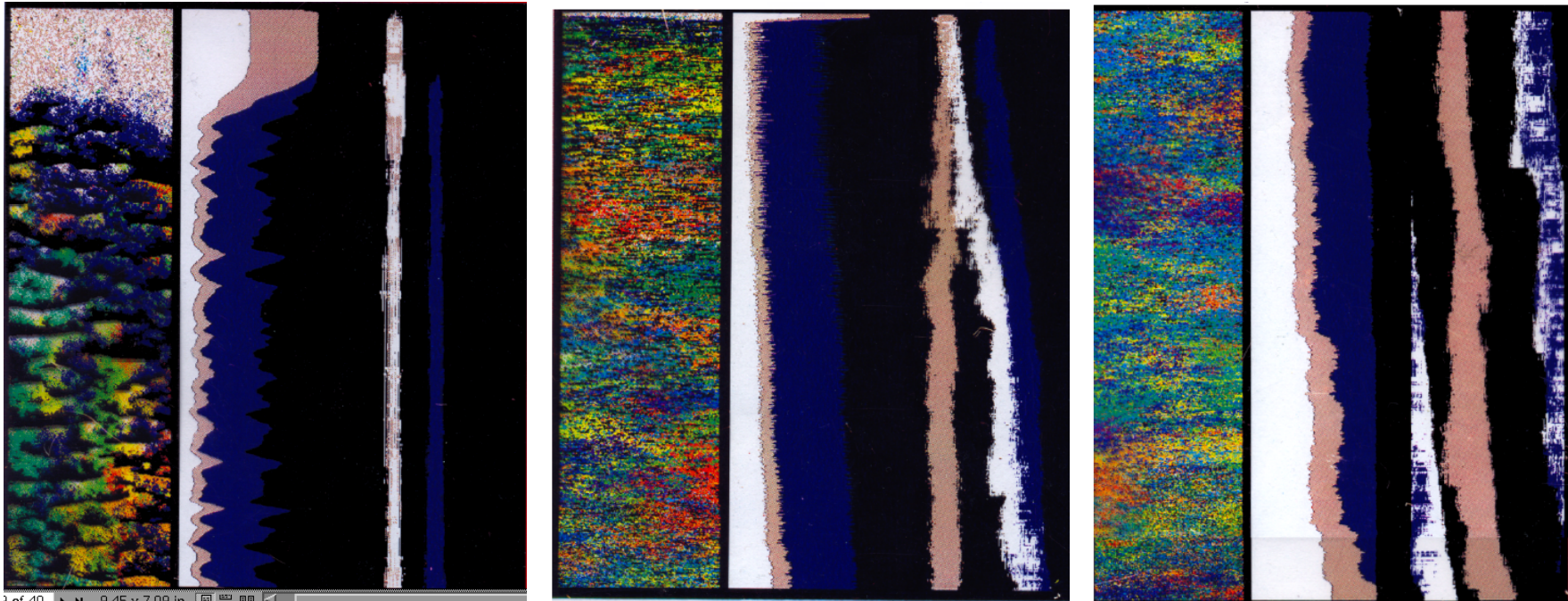
$A \rightleftharpoons B \rightarrow P$  Persistence in Ecological Time







long term evolution: extinction of cheater  
selection on spatial patterns,  
brown species B NO selection for getting more help



$A \rightleftharpoons B \rightarrow P$  Persistence in Ecological Time

# Phylogenetic reconstruction shows: Gene loss plays major role in evolution

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(reconstructed) Ancestral Genomes relatively large

Genes often present before their known present day function is realized.

Example HOX genes before differentiated bodyplan

Example Cell differentiation genes before multicellularity (cf Volvox)

Are these counterintuitive observations inherent to evolutionary processes?

**Study by modeling basic evolutionary processes**

# phylogenetic reconstruction of metabolic enzymes

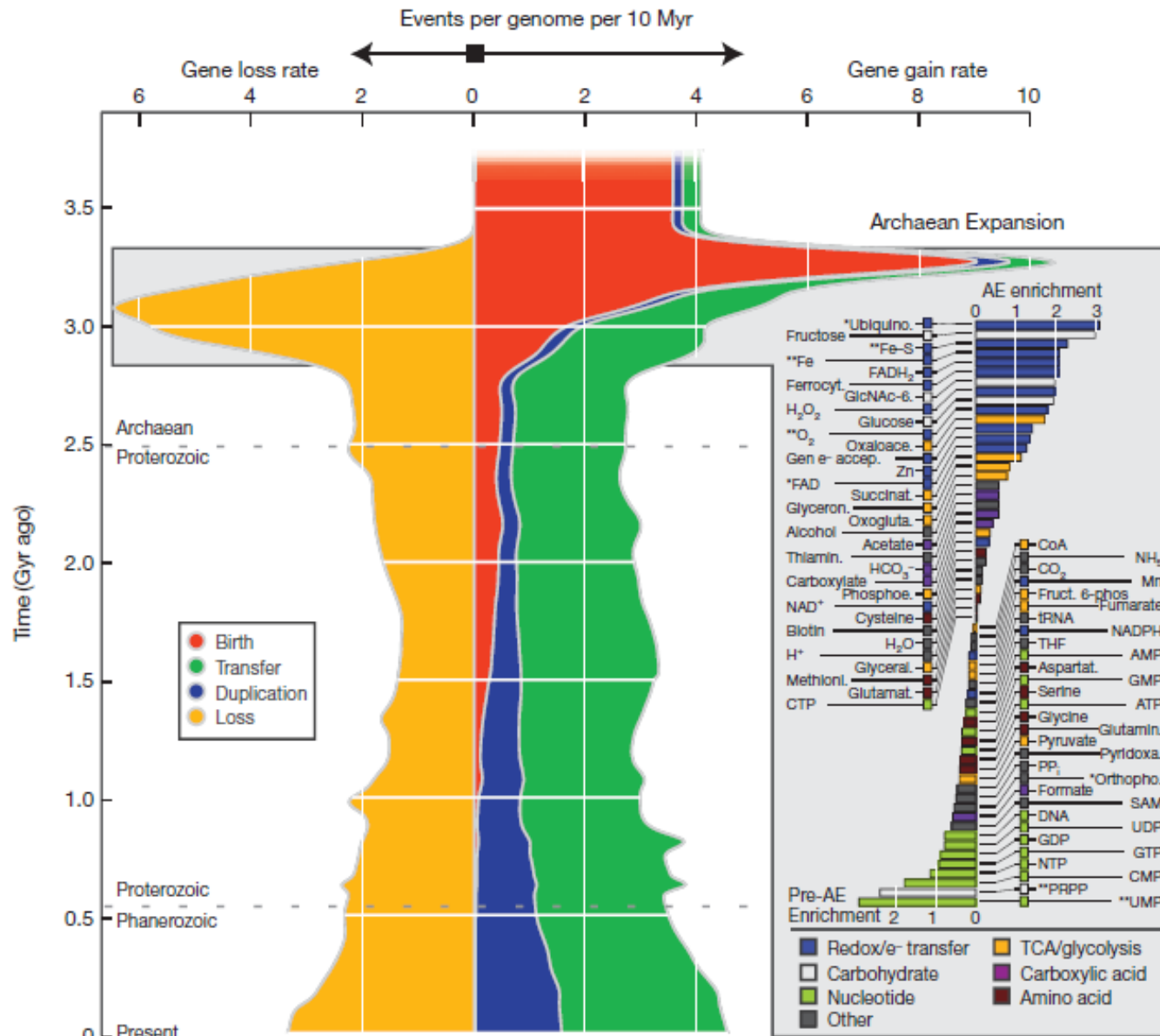
David and Alm, Nature 2010

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- make all gene trees (3983)
- reconcile gene trees on species tree  
minimizing number of 'events': innovation, loss, HGT, duplication and changes in genome sizes along the tree
- calibrate timing on fossil record

How did tot biospere metabolism change over the history of life?

**“big bang” in metabolic expansion and radiation**



# Gene loss as major evolutionary process

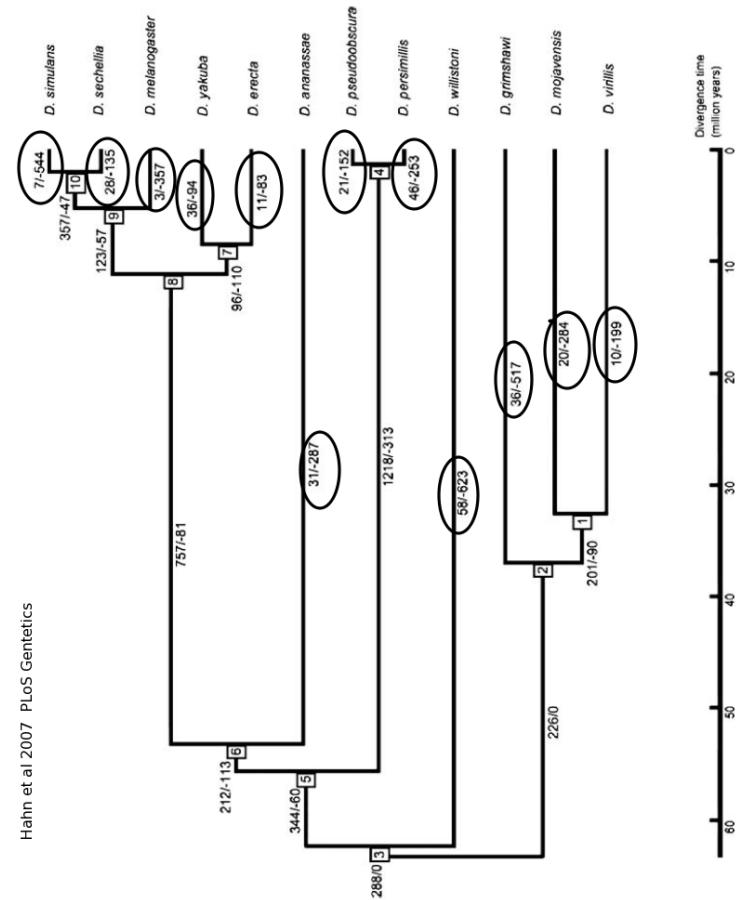
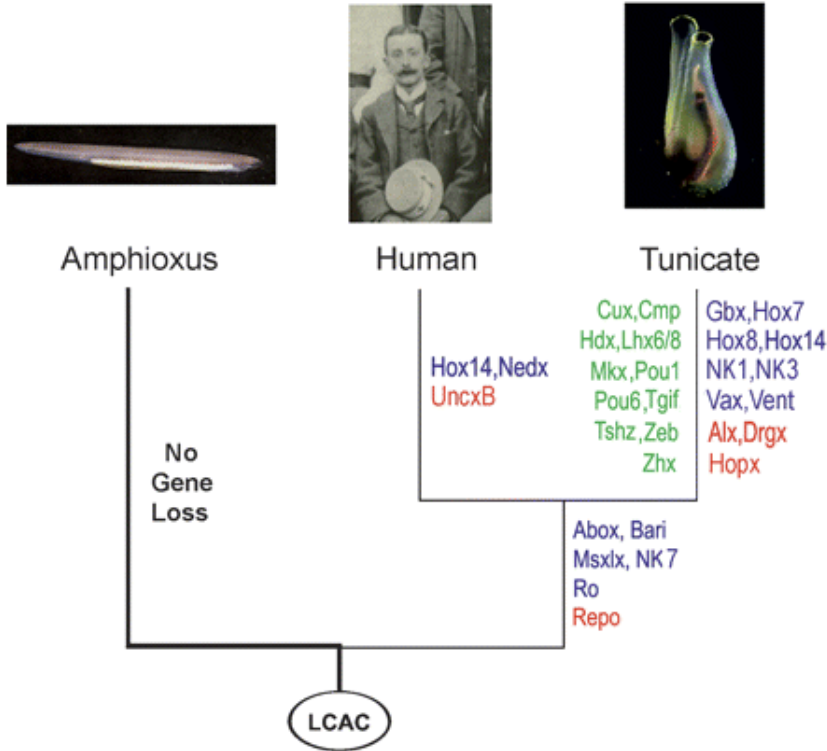


Figure 3. Lineage-Specific and Extinct Gene Families

Metazoa  
Loss of homeobox genes

*Drosophila* species  
gain/loss of genes

## Modeling genome evolution

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NOT like in ecological/immunological models in the course  
populations of identical individuals.

But (through mutations) all individuals may be unique.

Not ODE, but individual oriented models (like above) but  
moreover

Individuals: genotype - phenotype - fitness mapping  
can be dynamical system ODE (gene regulation, metabolism)

birth/death dependent on fitness

mutational operators: INDELS, substitutions (and/or parameter changes)

# Evolution of genome size in virtual cells

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based on “plausible” *minimal* multilevel ‘cell’  
mutations segmental duplications/ deletions, pointmutations  
fitness: *homestasis* (evolves regulatory adaptation)  
*evolving in varying environment*

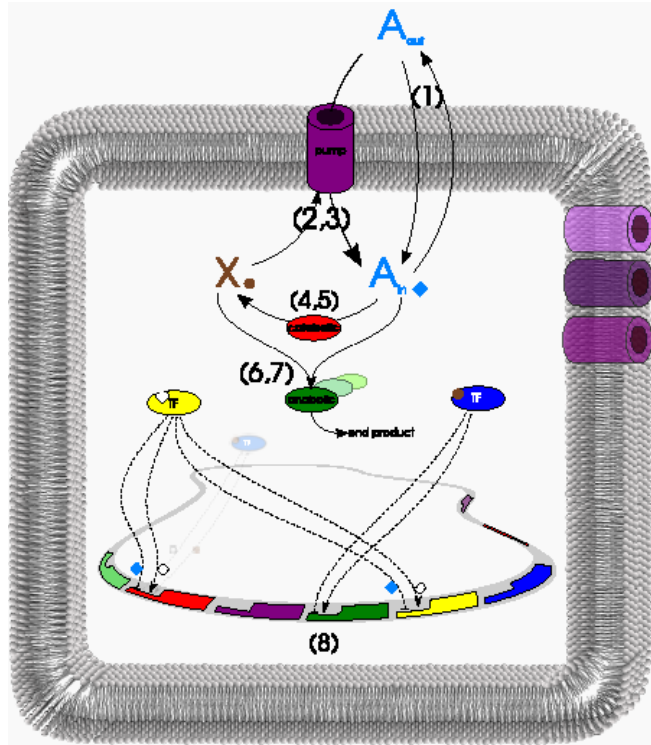
## Questions

Are some of the features seen in phylogenetic analysis observable in evolution of such cells?

Early complexity, dominance of gene loss



## virtual cell model (adapted from Neyfakh et al 2009 Biol Direct)



$$\frac{d[A]}{dt} = ([A]_{out} - [A])P_{perm} \quad (1)$$

$$\left( \frac{d[A]}{dt} = \frac{[A]_{out}[X]V_{max_p}[Prot]_p}{([A]_{out} + KA_p)([X] + KX_p)} \right) \quad (2)$$

$$\left( \frac{d[X]}{dt} = \frac{-d[A]}{dt} \right) \quad (3)$$

$$\left( \frac{d[A]}{dt} = \frac{-[Prot]_c[A]V_{max_c}}{[A] + KA_c} \right) \quad (4)$$

$$\left( \frac{d[X]}{dt} = \frac{-d[A]}{dt} E_{conv} \right) \quad (5)$$

$$\left( \frac{d[A]}{dt} = \frac{-[Prot]_a[A][X]V_{max_a}}{([A] + KA_a)([X] + KX_a)} \right) \quad (6)$$

$$\left( \frac{d[X]}{dt} = \frac{d[A]}{dt} \right) \quad (7)$$

$$\frac{d[Prot]}{dt} = Pr \cdot Reg - Degr[Prot] \quad (8)$$

### Processes modelled in the cell:

- diffusion (1) : **A** follows the gradient over the cell membrane
- pumping (2,3) : **pump enzymes** consume **X** to import **A**
- catabolism (4,5) : **catabolic enzymes** convert resource (**A**) into energy (**X**)
- anabolism (6,7) : **anabolic enzymes** consume **A** and **X** to produce building blocks
- protein production and degradation (8) : **TFs** regulate the rate of transcription of proteins; degradation takes place at a constant rate

## evolution of virtual cells

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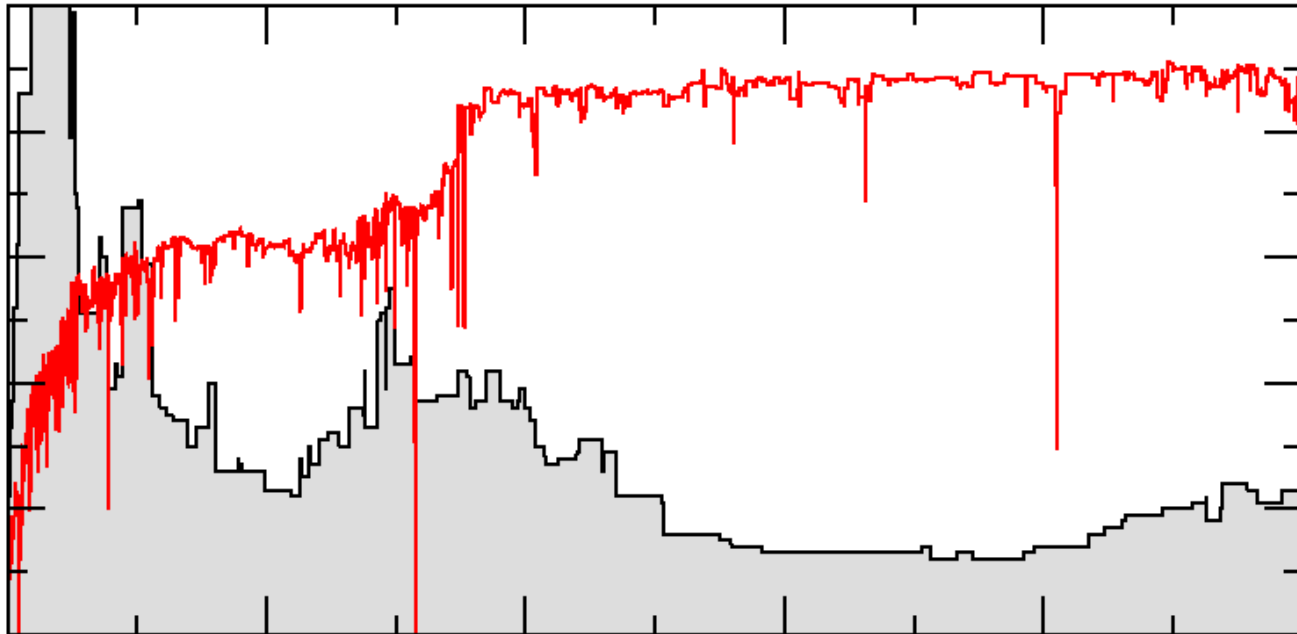
- Population of 1000 cells, 10000 generations
- external concentration of resource **A** fluctuates between .003 and 30
- *homeostasis*: Internal concentration should be kept at 1.
- Initial genome size ca 10 genes
- Mutational operators: duplication / deletion / rearrangement / point mutations
- ('sees' (only) 1-3 environments in lifetime - adapts to 'all')

**Typical evolutionary dynamics:  
Genome inflation(s) - followed by fitness increase -  
followed by stream lining - followed by genome size  
fluctuations**

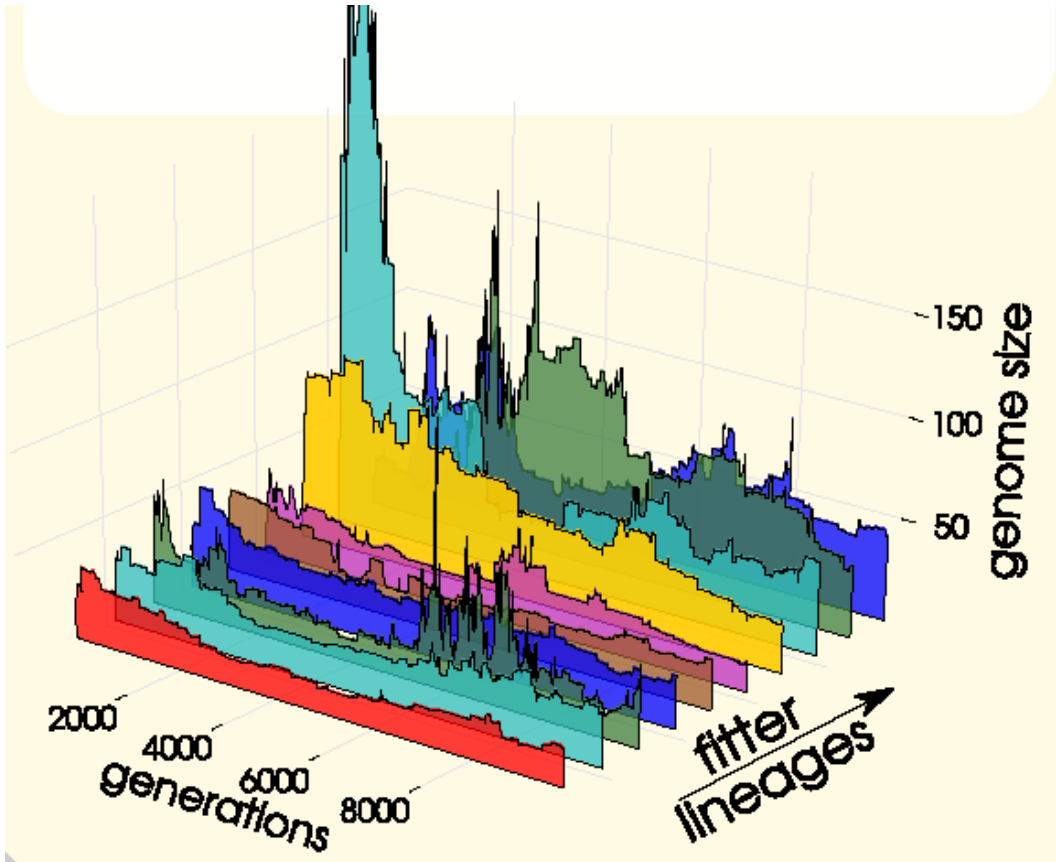
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## Genome size and fitness

Ancestor trace



## early genome inflation “generic” pattern

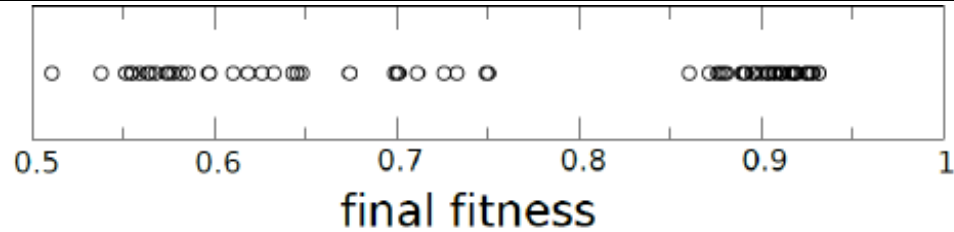


mutation type \ trend	genome inflation	streamlining	final fitness
point mutation	-		-
single gene dup/del	-	+	-
deletion bias	-	+	-

occurs in “better” runs  
in one param. setting

occurs in parameter settings  
which lead to “better” results

# Local landscapes, genome expansion and future fitness



Duplications			Deletions		
t=1-100	t=101-200	$\Delta F$	t=1-100	t=101-200	$\Delta F$
+	(+)	> 1.05	=	=	> 1.05
(+)	+	.95 – 1.05	=	+	.95 – 1.05
-	-	< .95	=	-	< .95
Genome Size			Fitness		
t=1-100	t=101-200		t=1-100	t=101-200	
+	+		=	=	

## Conclusions evolution of virtual cells

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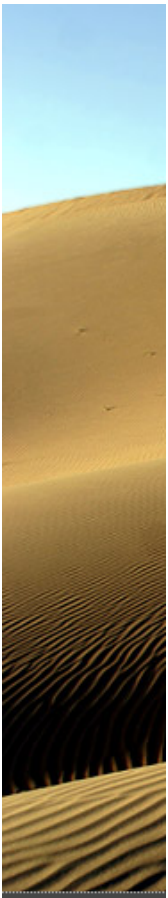
- early genome inflations,  
increases degrees of freedom and therewith adaptability
- followed by streamlining: fitness gain through gene loss
- Intricate interplay of neutral and adaptive processes:  
adaptation — > neutrality; neutrality — adaptation
- also other observables, eg effect of mutations, e.g.  
Evolved genotype phenotype mapping maximizes  
neutrality AND selection

*interesting (unexpected) but generic behaviour  
of mutation/selection*

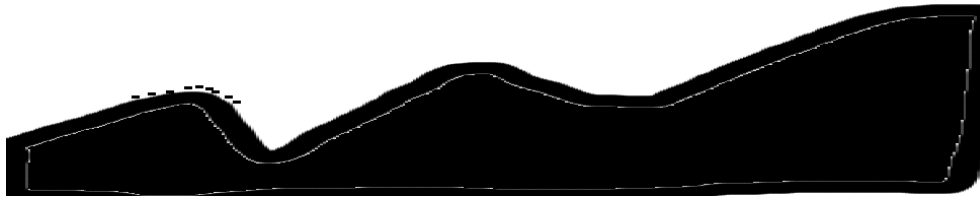


Rugged fitness landscape

Evolution “stuck on local optima??”



NO.....



DETOURS!



**Evolution not "far away and long ago"**  
**New insights through experimental evolution, high  
throughput data, bioinformatic analysis and  
evolutionary modeling**

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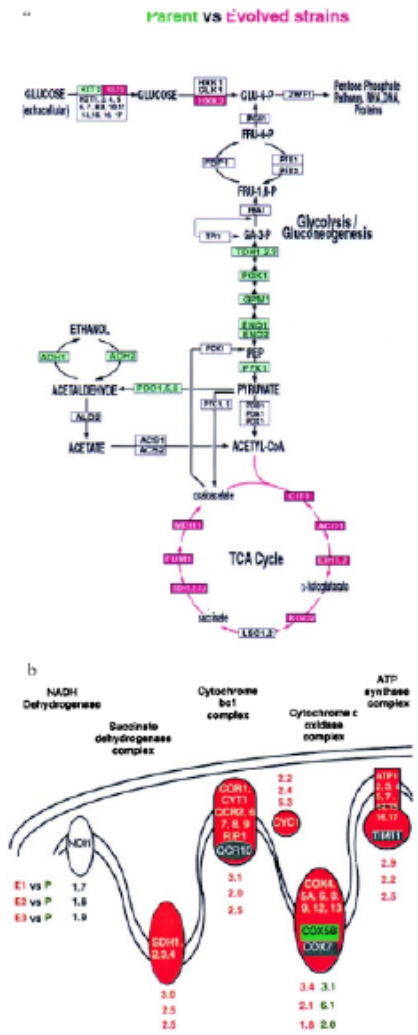
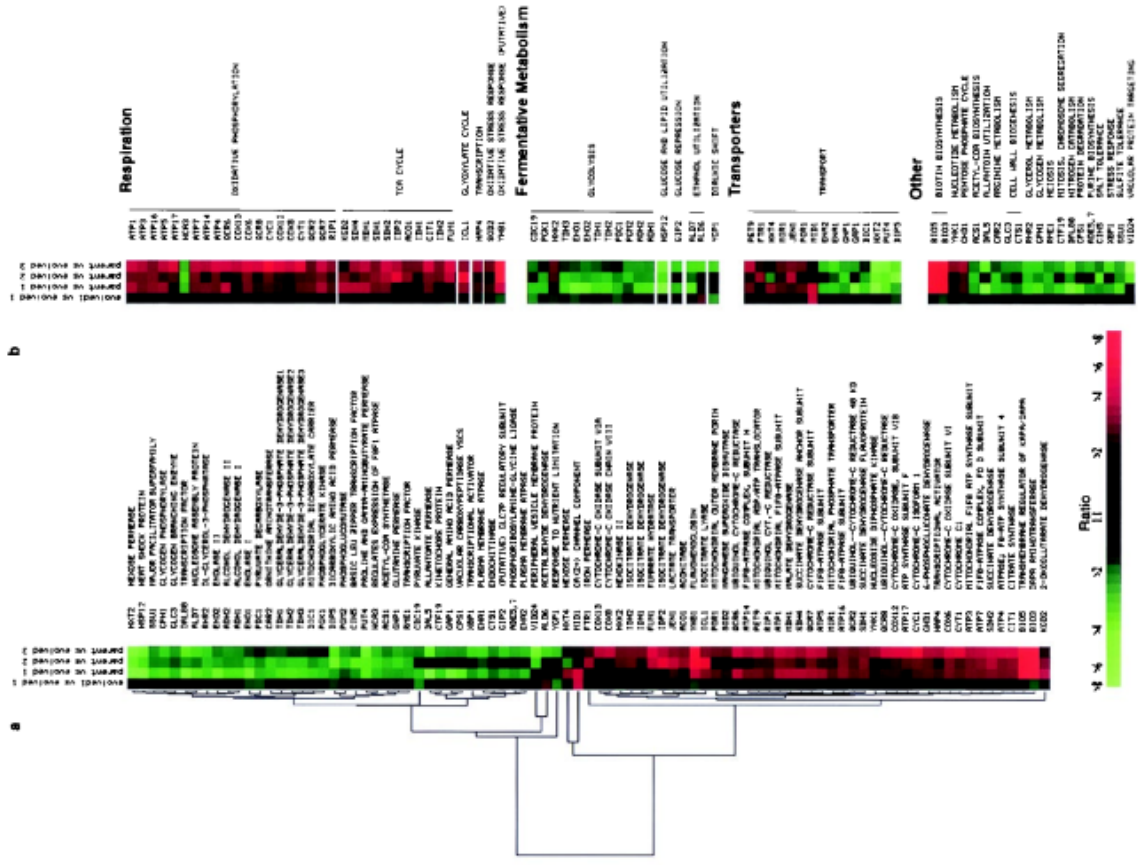
**Minor (?) transitions in evolution**  
**Yeast regulatory network evolution**  
**Some “surprising” observations from short term**  
**evolution experiments**  
**( Ferrea et al 1999, Dunham et al 2002)**

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- very efficient adaptation in short period
- major changes in gene expression in short evolutionary time: ca 600 genes differentially expressed in period that no more than 7 mutations expected
- changes in gene expression make “sense” with respect to adaptation
- resemble regulatory adaptation
- many gross chromosomal rearrangement (GCR)
- similar GCR in duplicate evol experiment

**evolved evolvability?**

# regulatory and/vs evolutionary 'adaptation' gene expression changes in strain evolved on low glucose med



**Are these properties of short term evolution a generic property of mutation/selection in evolving systems with explicit genome-network mapping?**

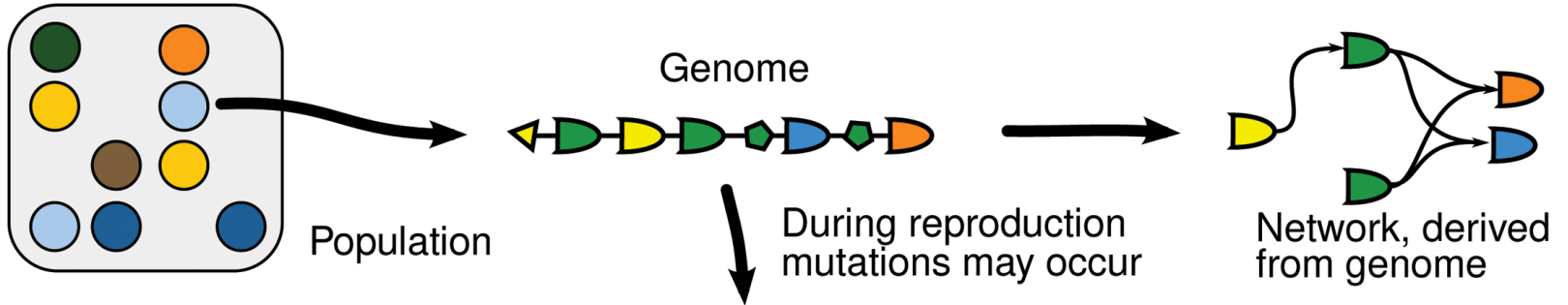
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By evolution of genome/transcriptome structure?

# Evolution of Regulation based mutational priming

Crombach and Hogeweg PLOS Comp Biol 2008

## A Overview of the model

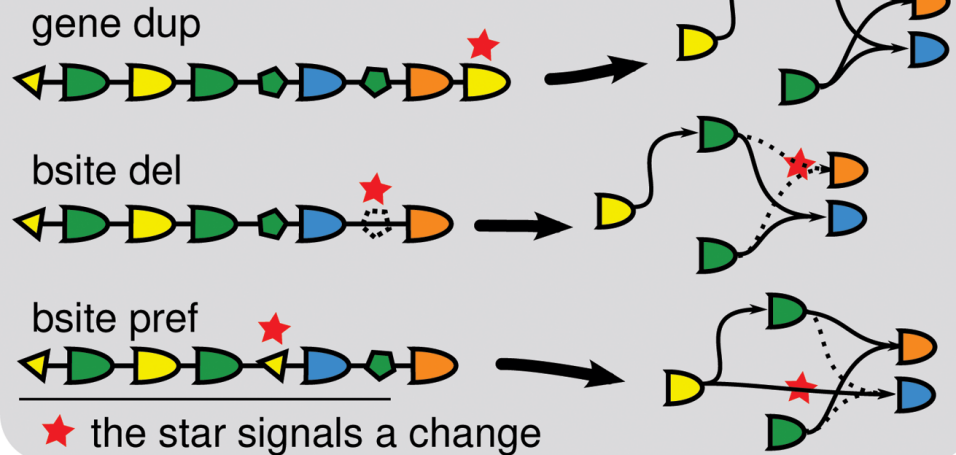


## C Evolutionary targets

		A	
		genes on	genes off
B	genes on		
	genes off		



## B Effect of mutations



## network dynamics and fitness

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Network update:

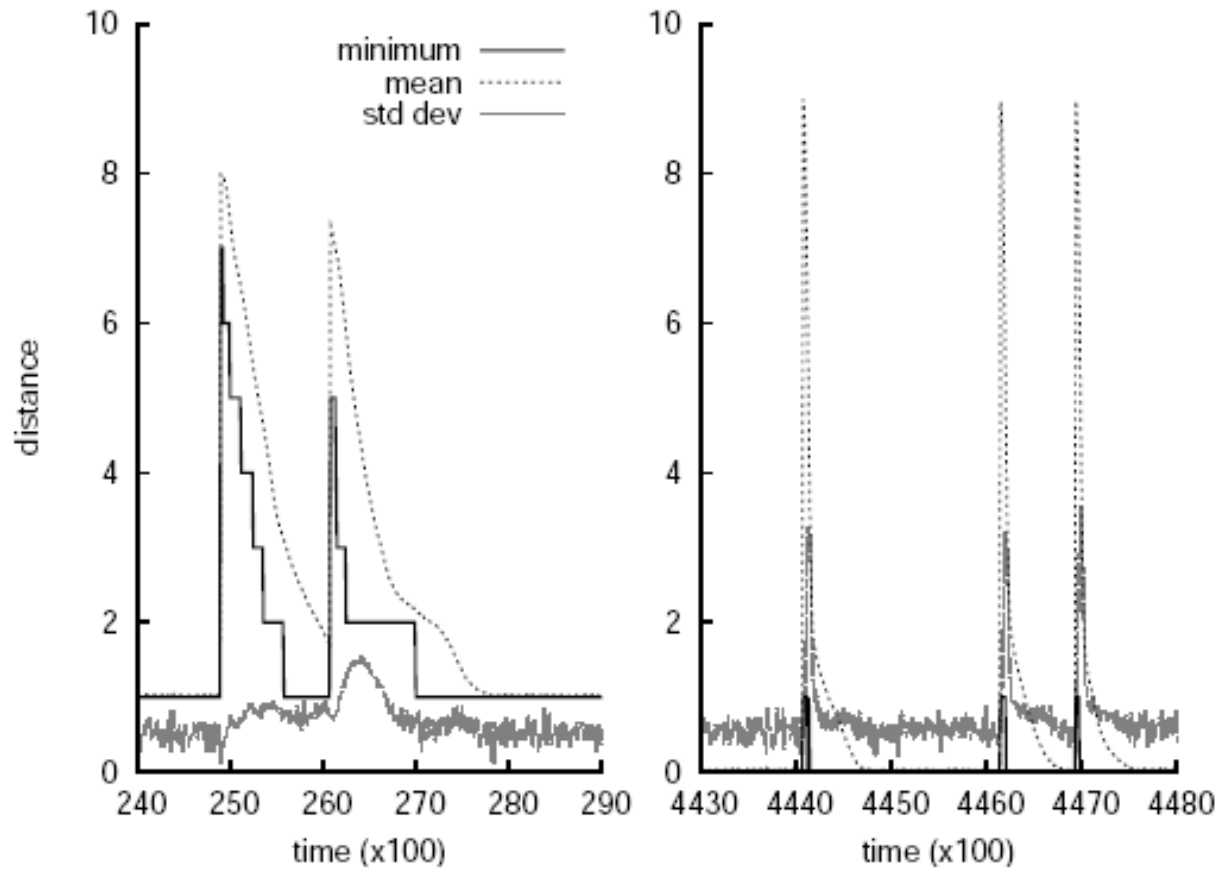
$$s_i^{t+1} = \begin{cases} 0 & \text{if } \sum_j w_{ij}s_j^t < \theta_i \\ s_i^t & \text{if } \sum_j w_{ij}s_j^t = \theta_i \\ 1 & \text{if } \sum_j w_{ij}s_j^t > \theta_i \end{cases}$$

fitness: distance to target

$$f = \left(1 - \frac{D}{D_{max}}\right)^p$$

# improved evolvability observed

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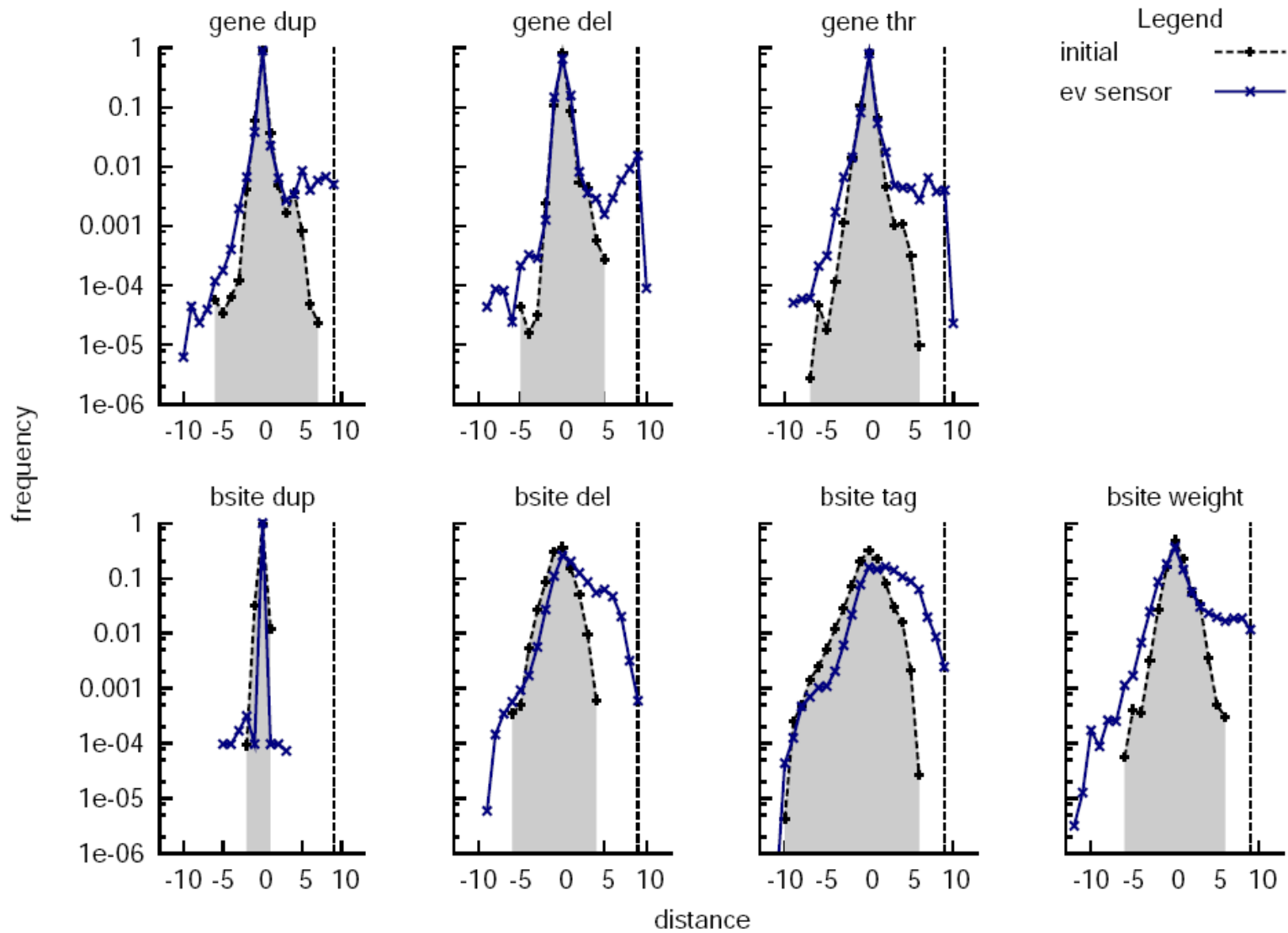


## **Hamming distance improvement to opposite target**

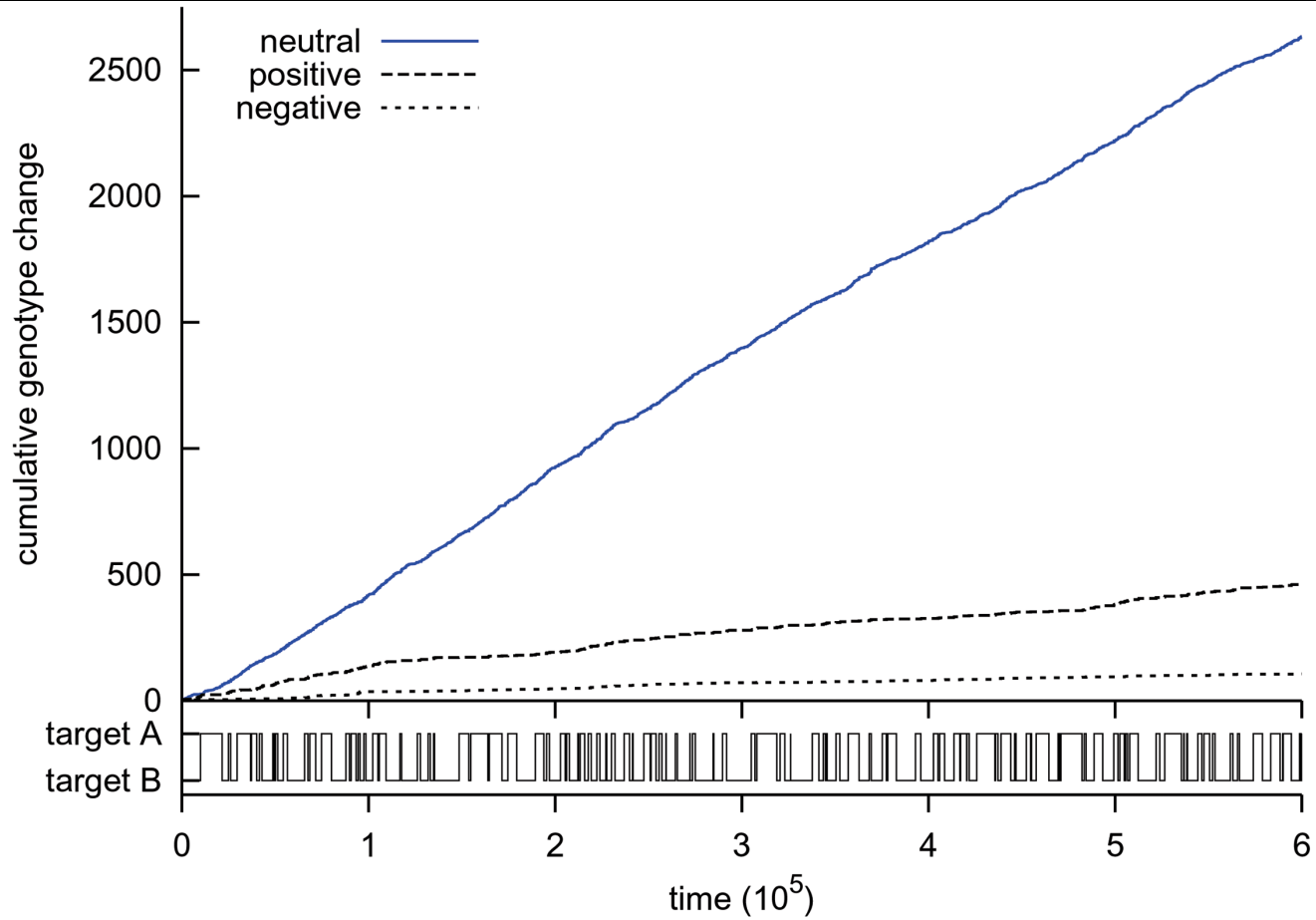
### **Regulatory Mutational Priming:**

Many different mutations lead to “beneficial” adaptation



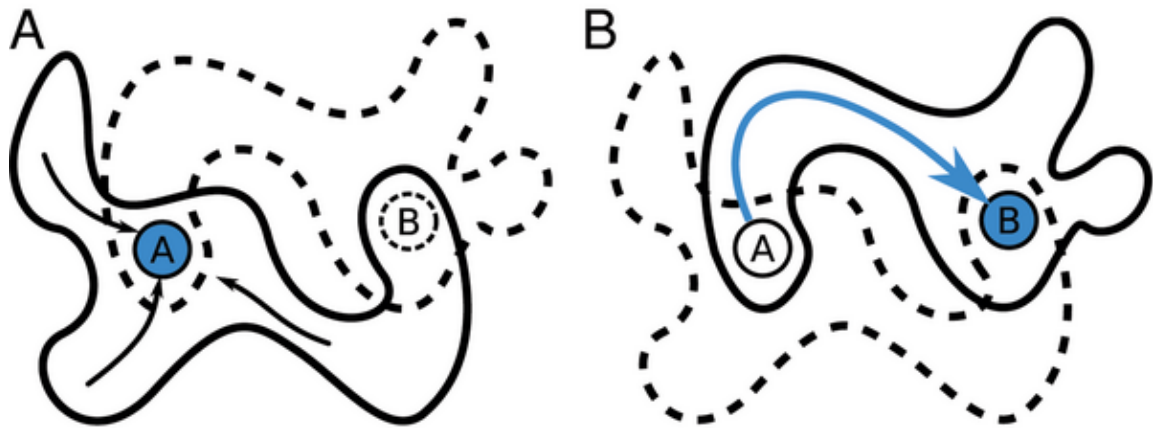


# Neutral drift far greater than adaptive change!



# evolution of evolvability and bases of attraction

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## conclusions

# Evolution of genomes and gene regulatory networks evolution of evolvability

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*Random mutations are not “random”  
in EVOLVED genomes*

- Transposon dynamics structures genomes creating hotspots for mutations and genome ordering. Long term evolution leads to genome structures such that short term evolution is facilitated
- Genotype to phenotype mapping through gene regulatory networks evolves such that (advantageous) attractor switching occurs (blow up of single mutations to large scale effects)

*Both these mechanisms appear to occur in Yeast*

# Overall summary/conclusions

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*Nurse: Biology faces a quantum leap into the incomprehensible: the complexity of biology information processing networks will bring us in a counterintuitive world*

Indeed but: **+ bioinformatic modeling**  
renders the counterintuitive comprehensible

- — — > profound new insight is major transitions in evolution
- — — > current functioning of organisms

**Next part of the course**

*Woese: "One of the most fundamental patterns of scientific discovery is the revolution in thought that accompanies a new body of data"*

Indeed High throughput data  
**+ bioinformatic data analysis**

- — — > profound new insight is major transitions in evolution
- — — > current functioning of organisms

