

Genome evolution: coding structures and evolvability

Evolution of coding structures (continued)

LAST TIME

Multilevel evolution in RNA world

- GP map based on Min. energy folding - fixed genome length
- spatial pattern formation

Evolution of multiple lineages (lower mutation rates)

Evolution of functional mutational neighborhood (at high mutation rates)

Evolution of multiple coding

ecosystem based solution to "do more" than individuals??

NOW

From prebiotic evolution to biotic evolution

Surprising observation explained from "first principles"

Evolution of genomes and transcription regulatory networks

Mutational operators Not only point mutations

evolution of evolvability (2)

Random mutation - non-random effects

Individual based vs ecosystem based “problem solving” predator-prey-scavenger coevolution

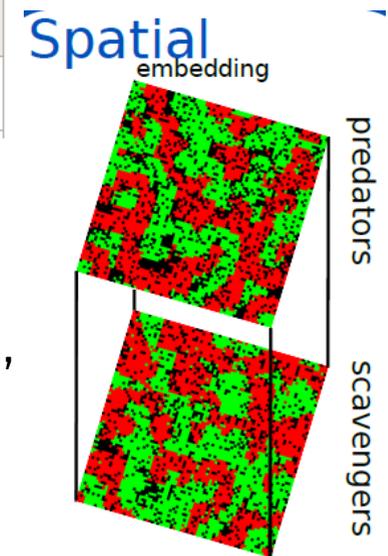
FK de Boer, P Hogeweg 2010

Problem: solve “function” - fully digest all possible prey
prey 2 continuous properties: $0 < X, Y < K$
Fully eaten when predator calculates $f(X, Y)$ correctly

e.g.

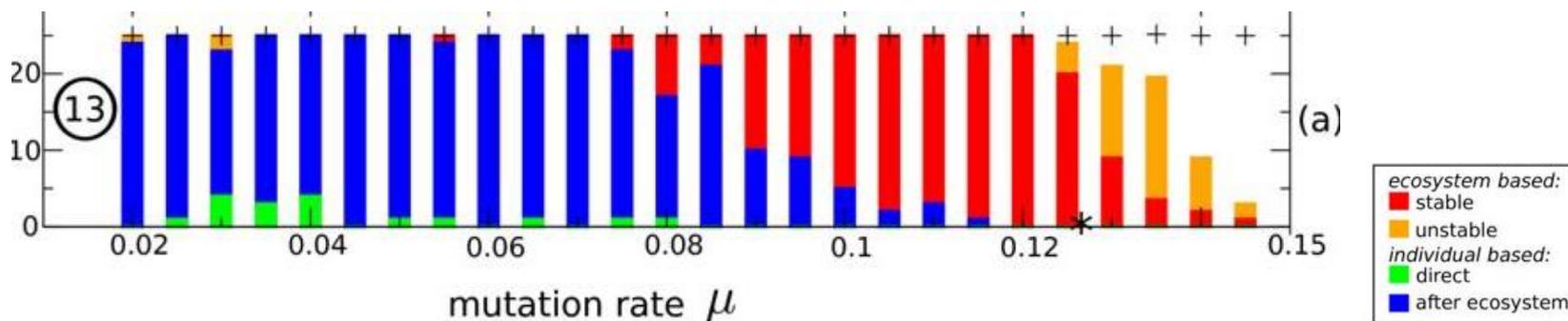
Evolutionary Target	Minimal Coding Example
$f(x, y) = x^3 + y^3 + 5x^2$	<code>(+ (* (* (+ x 5) x) x) (* (* y y) y))</code>

Fitness predator: how well it solves “its” prey
Fitness prey: how badly predator solves it
Fitness scavenger: How it solves “what is left”

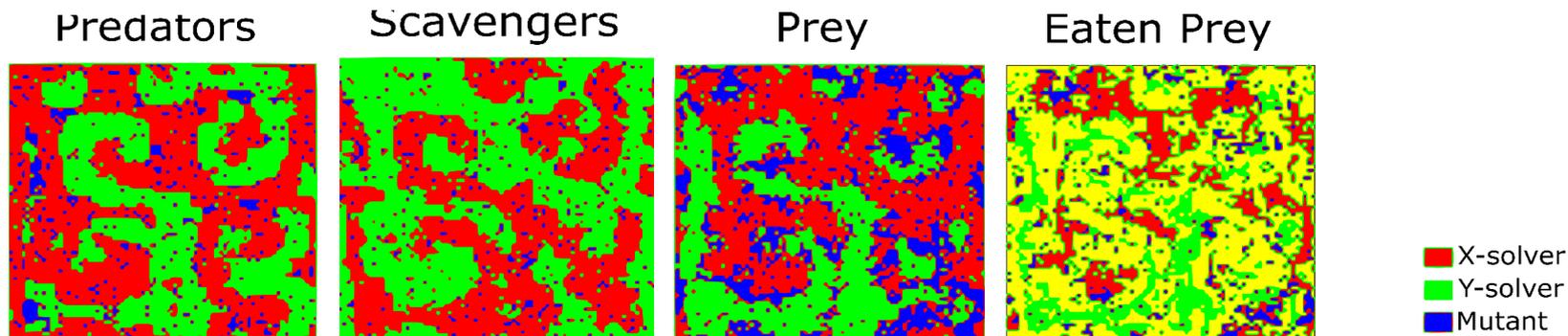


do individual predators, or does the ecosystems solve it

Ecosystem based solution 'easier' to evolve precedes individual based solution



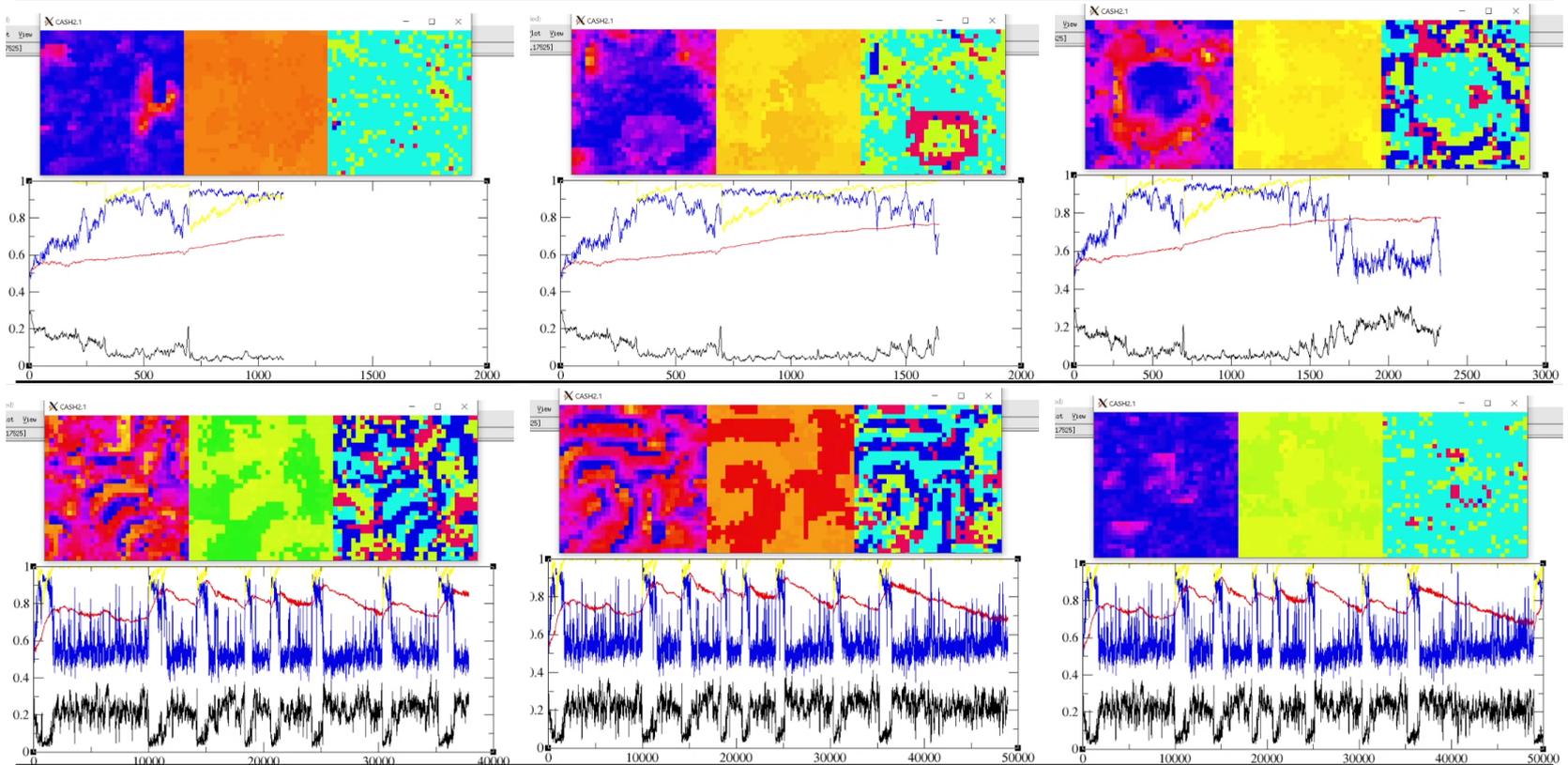
Two predator populations specializing on X or Y
 Two scavenger populations specializing on X or Y
 Two prey populations with high X or high Y values
 Self organize in spiral waves,
 X predator and Y scavenger pairs together digest prey fully
 (i.e. encode the target function correctly)



**Slightly beneficial 'genes' (bits) and
individual vs ecosystem based solutions (gene
conservation)**

Host-Parasite coevolution, above the information threshold

Fitness target: bitstring(256), differential weighted fitness/bit
Wouter Ubbink 2021: the helpful parasite



“the absolute of the host increases while its actual fitness decreases, and that of the parasite increases and the ecosystems loses bits”

*black: host actual fitness; blue parasite actual fitness;
red: host absolute fitness; yellow: collective fitness (bits in population)*

conclusion

“Cooperation” (getting something done together) through spatial selforganization

Division of labor among predators

Coerces prey into certain types

See less – > can do more:

Cooperative solution of “all” problems, by “seeing” only a subset of problems No direct or indirect fitness benefit for predators to give scavengers an eatable bite.

ecosystem based solution precedes individual based solution

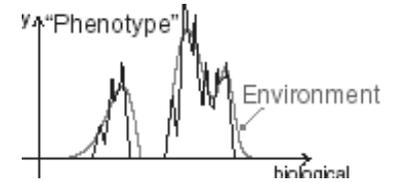
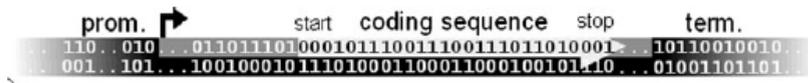
ecosystem based solution stable at high mutation rates

ecosystem based solution preserves slightly beneficial genes

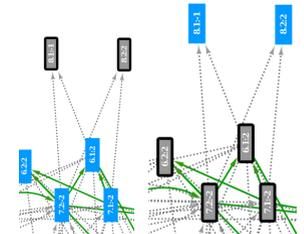
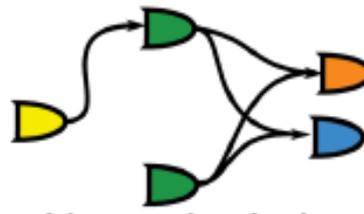
GENOME EVOLUTION: 3 (4) modeling frameworks

genome structure and genotype to phenotype/fitnessmap

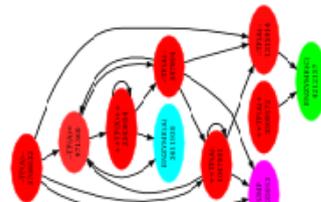
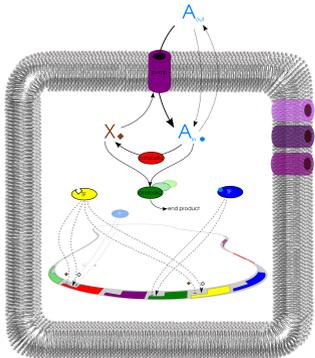
AEVOL bit (nucleotide) level coding of genome allows evolution of new genes!



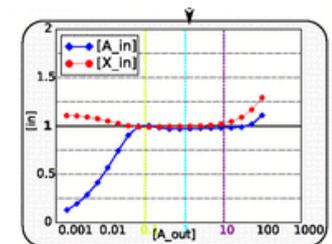
PoaS course grained genome allows multilevel (evolving) GPf mapping



PoaS + metabolism (Virtual cell) allows *indirect* sensing environment



$$\begin{aligned} \frac{d[A]}{dt} &= \frac{[A]_{out}[X]Vmax_p[Proc]_p}{([A]_{out} + K_{A_p})([X] + K_{X_p})} & (2) \\ \frac{d[X]}{dt} &= -d[A] & (3) \\ \frac{d[A]}{dt} &= \frac{-[Proc]_c[A]Vmax_c}{[A] + K_{A_c}} & (4) \\ \frac{d[X]}{dt} &= -d[A] \cdot K_{leak} & (5) \end{aligned}$$



Genome evolution

evolution transcription regulation networks

Observations

Structural features of transcription regulation networks

FFL

Very fast evolutionary adaptation in the lab (e.g. YEAST)

Involves massive changes at transcriptome level (ca 10%)

Involves GCR - and the same ones in multiple experiments

Are these generic features, to be expected from random mutations (with or without selection)?

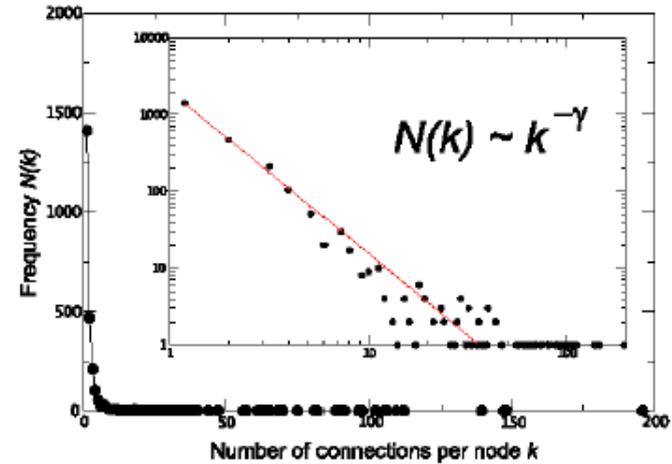
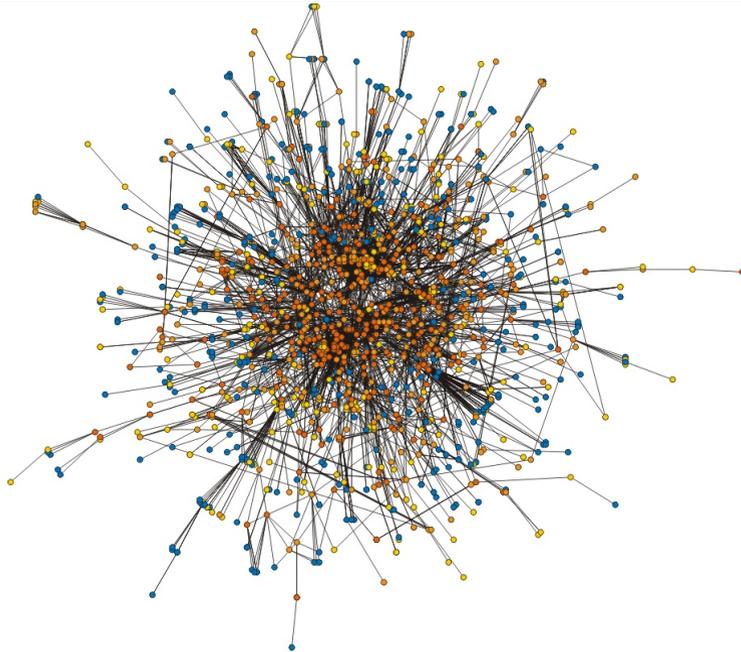
dynamics OF evolving and evolved Gene Regulatory Networks (GRN)

Structural features of transcription regulation networks
powerlaw and FFL

- Characterizing topology of GRN
- *What do we mean with “over-representation of...”*
- *Random mutations = / = randomization*

Importance of coding structure

Observed properties of the GRN of Yeast



Network	Nodes	Edges	N_{real}	$N_{\text{rand}} \pm \text{SD}$	Z score	N_{real}	$N_{\text{rand}} \pm \text{SD}$	Z score
Gene regulation (transcription)								
					Feed-forward loop		Bi-fan	
<i>E. coli</i>	424	519	40	7 ± 3	10	203	47 ± 12	13
<i>S. cerevisiae</i> *	685	1,052	70	11 ± 4	14	1812	300 ± 40	41

(also many other networks: neural networks, computer networks, (but not Eco-networks))

Genomic encoding of GRN

Modeling Mutational Dynamics

bag of genes with binding sites (BS)

BS deletion 8×10^{-3}

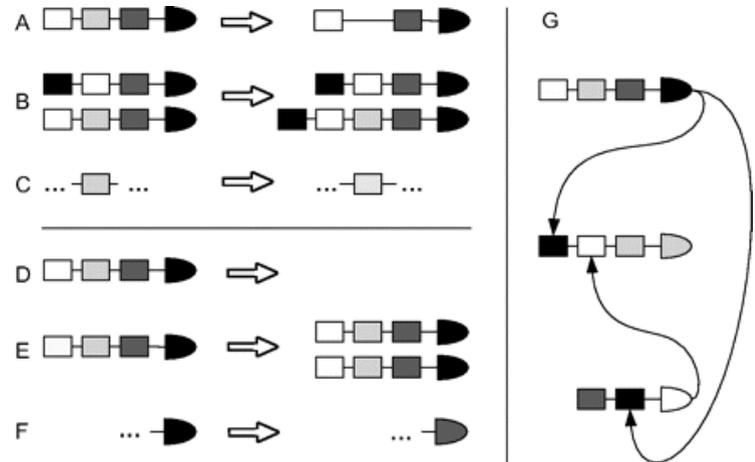
BS duplication 8×10^{-3}

Bs mutation 8×10^{-4}

Gene deletion 1×10^{-3}

Whole gene duplication 1×10^{-3}

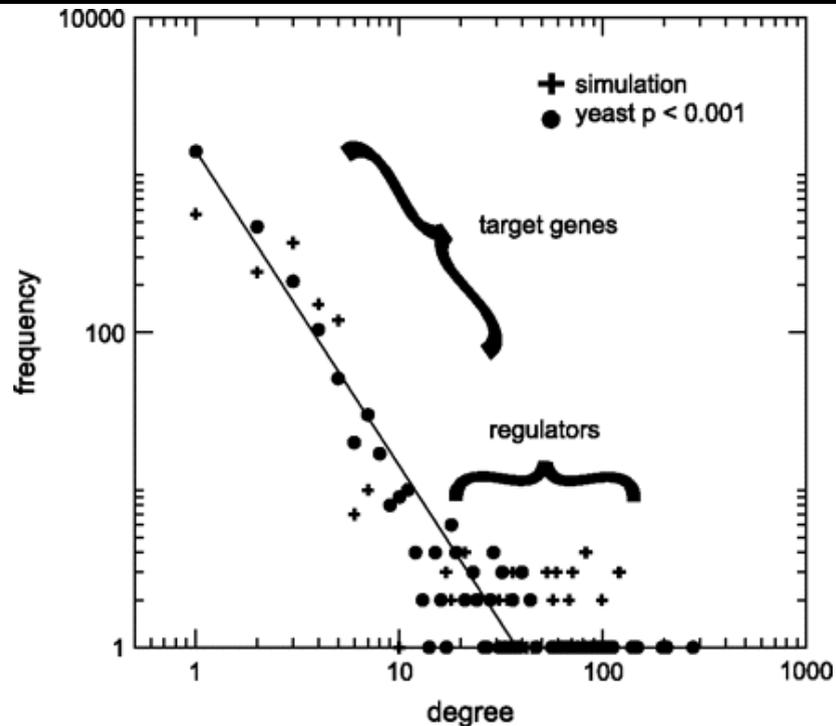
Protein divergence/innovation 5×10^{-3}



Parameters loosely chosen from literature, NOT FITTED.

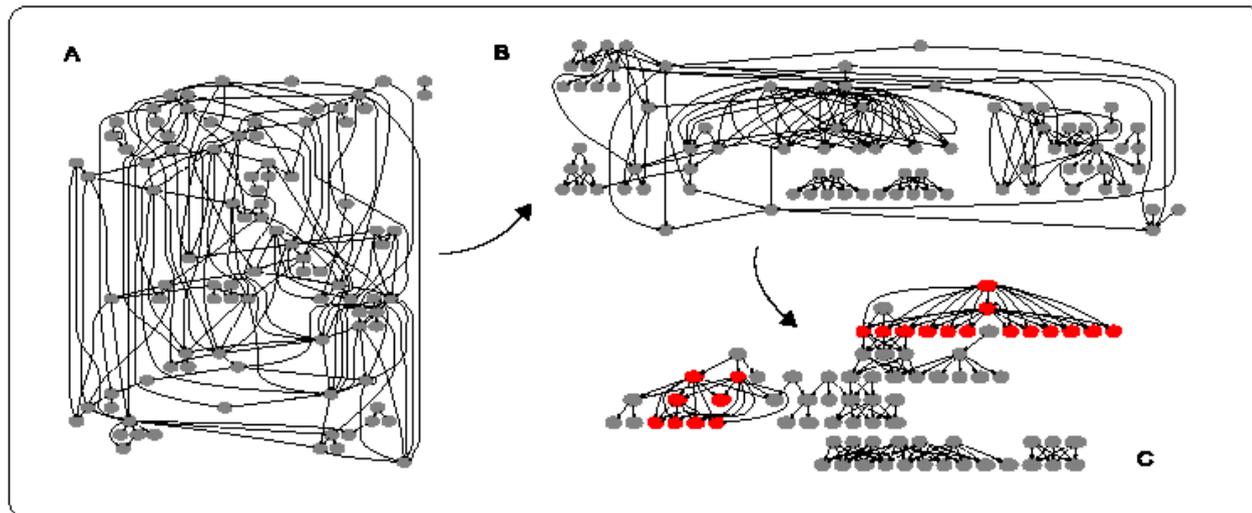
Feed-Forward Loop Circuits as a Side Effect of Genome Evolution Otto X. Cordero,
Paulien Hogeweg MBE 2006

mutational dynamics WITHOUT selection leads to
Powerlaw distribution of connections
with similar γ as Yeast GRN
for similar number of genes and TF (2000, 100)



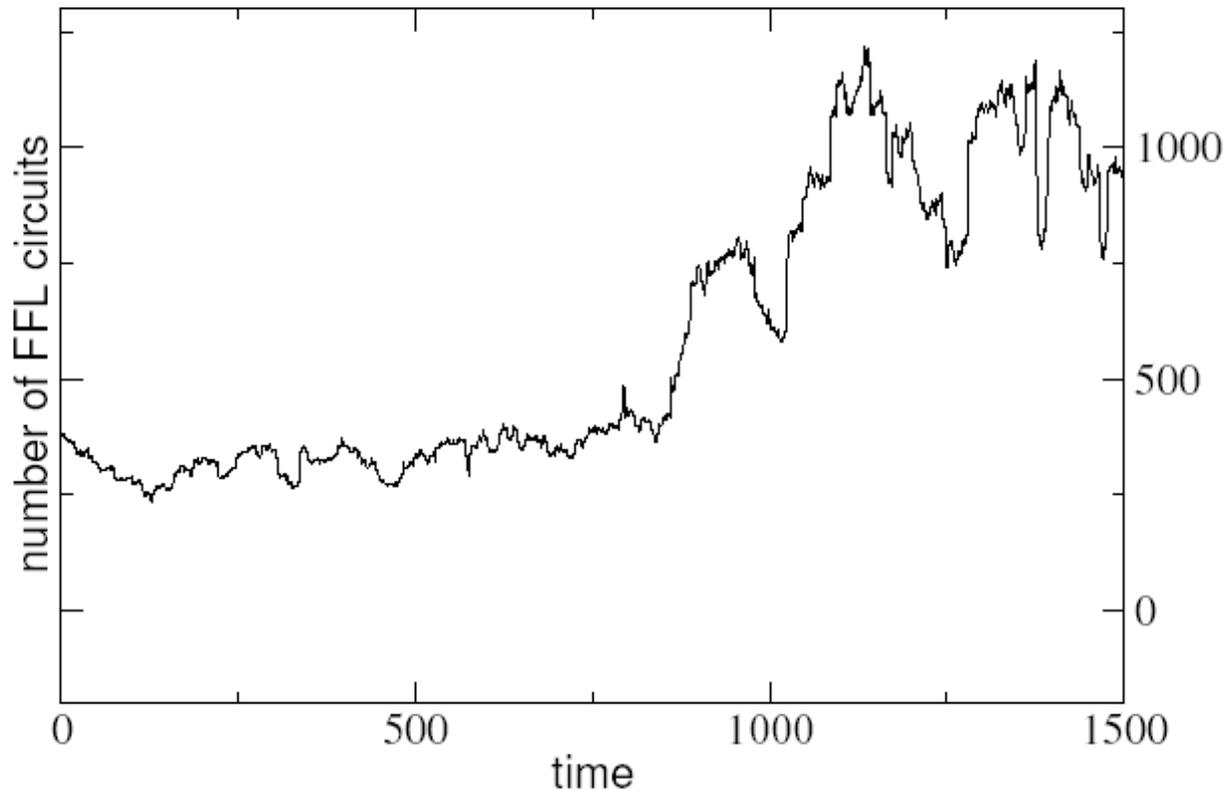
Toy model: Visualization of network restructuring during neutral evolution (hierarchical) structured network for free!

- ▶ Visualization of the network evolution (toy example):



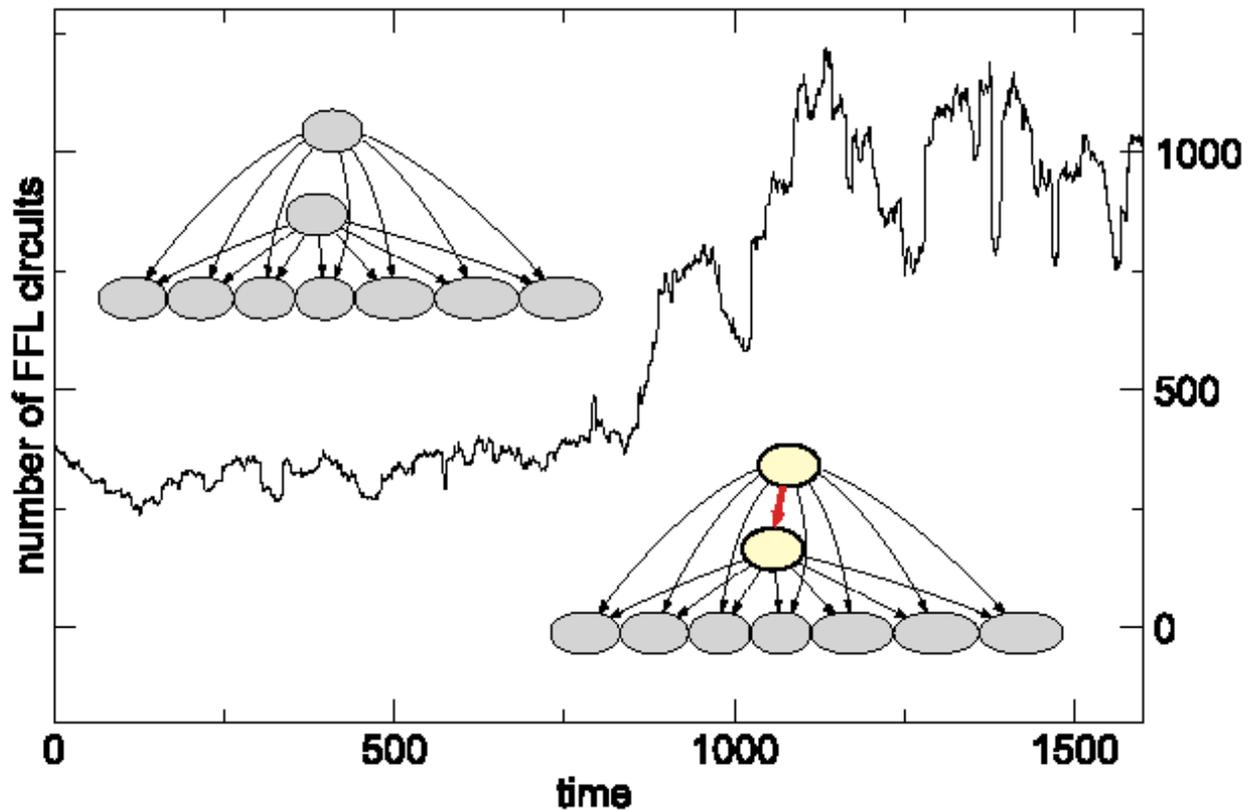
During evolution Sudden increase of FFL motifs: FFL as mutational signature

► Results of mutational dynamics at the microlevel:



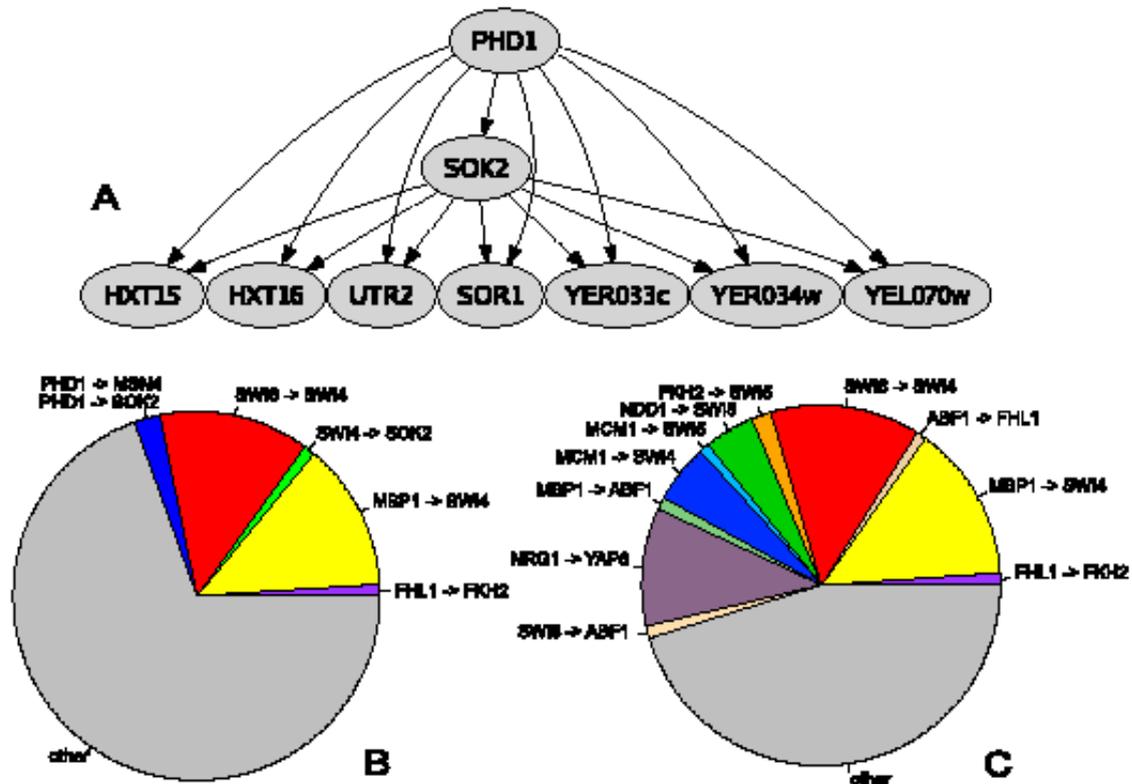
large increase of FFL motifs:
Originate through duplication of hub
+ new connection

► Mechanics of massive FFL formation:



Over-representation of FFL motifs in Yeast: Duplication + connection of Hub genes

- Evidence in the yeast network:



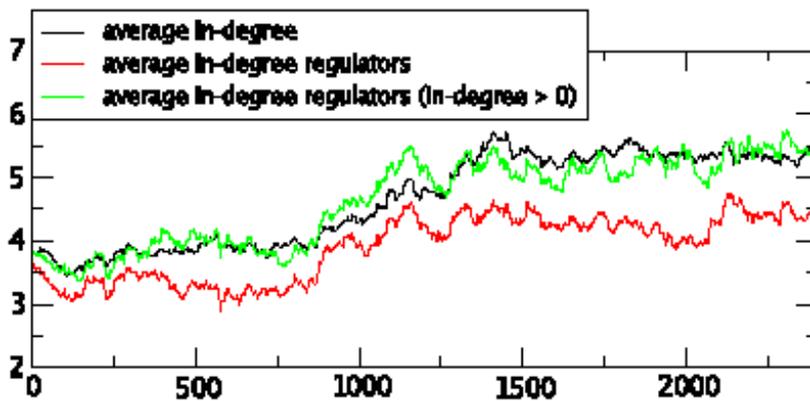
random mutations vs randomization

Randomization tests: keep everything the same EXCEPT feature to be tested

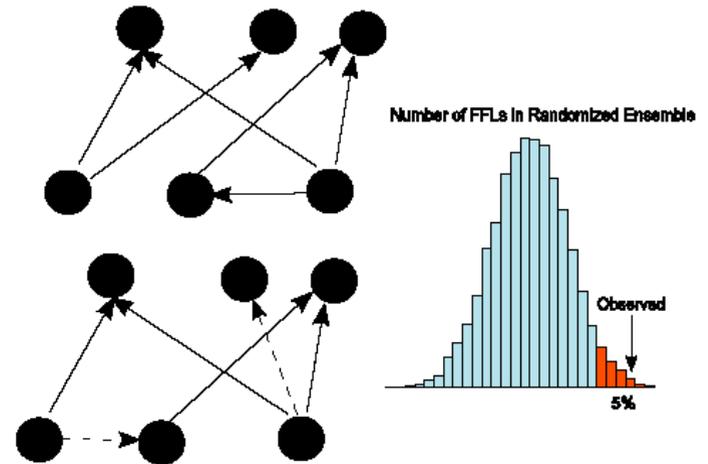
here: keep degree distribution
— test for FFL

BUT random mutations

do not conserve degree distribution



► Randomization test: swapping connections.



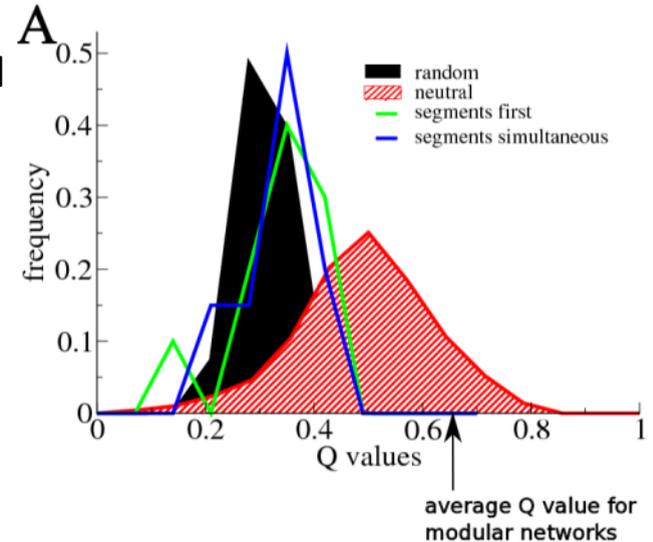
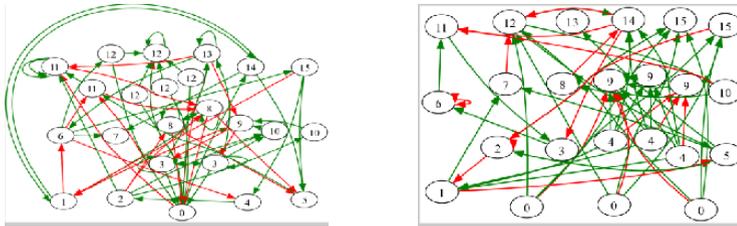
discussion/conclusions

Assessing relevance of observed evolved network structure in models or “real” life

Random mutations lead to non-random structure
With or without selection of “something”

Random mutations as stochastic dynamical system
goes to attractor

Example: Modularity in evolved GRN model
Drosophila segmentation/differentiation



ten Tusscher & Hogeweg 2011: Evolution of Networks for Body Plan Patterning;
Interplay of Modularity, Robustness and Evolvability

random networks as null model?

Yeast regulatory network evolution
Some “surprising” observations from short term
evolution experiments
(Ferea et al 1999, Dunham et al 2002)

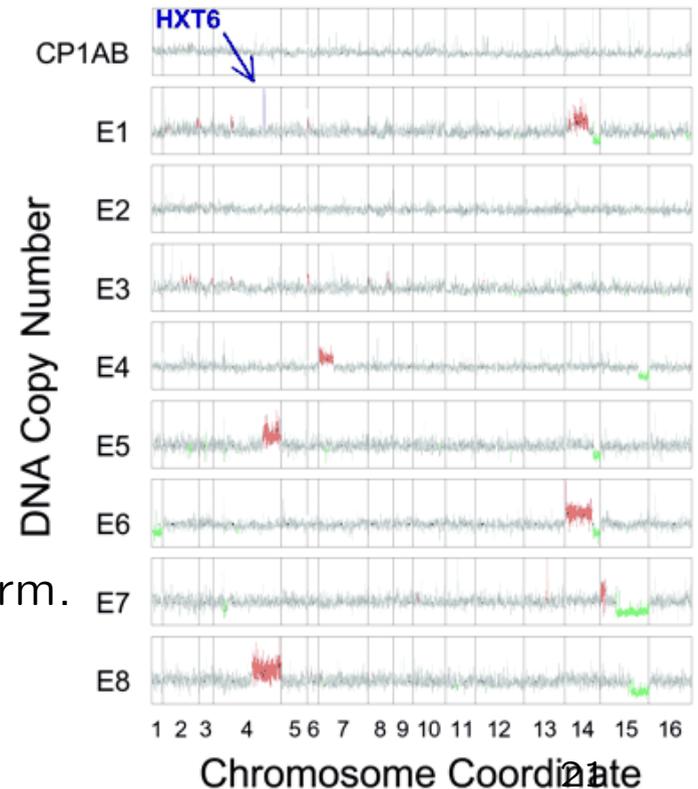
- very efficient adaptation in short period
- major changes in gene expression in short evolutionary time: ca 600 genes differentially expressed in period that no more than 7 mutations expected
- changes in gene expression make “sense” with respect to adaptation
- resemble regulatory adaptation
- many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

evolved evolvability?

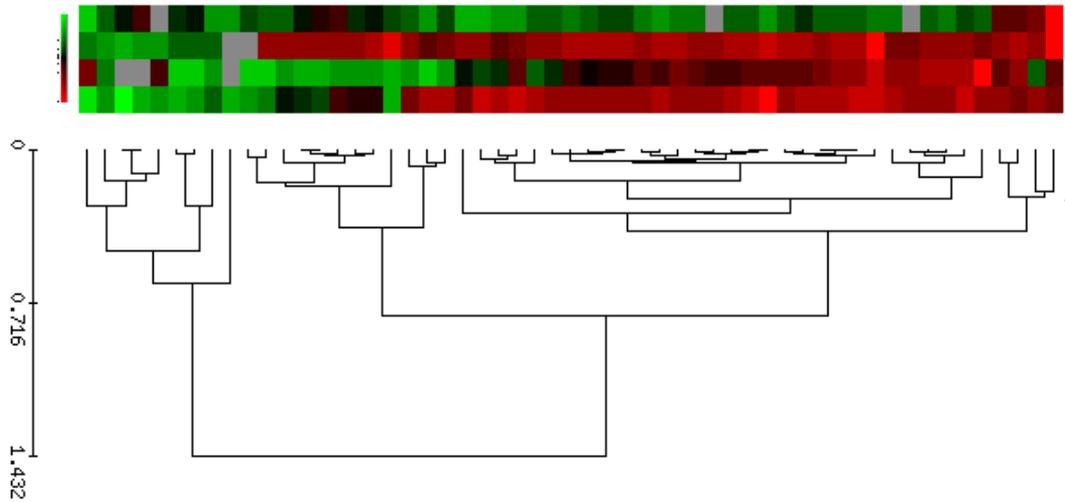
“Mutational priming” seen in yeast evolution

“Characteristic genome rearrangements in experimental evolution of *Saccharomyces cerevisiae*”
(Dunham et al PNAS 2002)

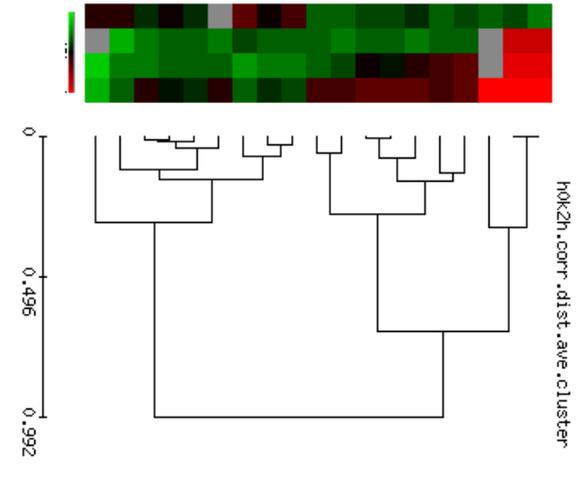
repeated duplication and loss
at the same breakpoints
3* in C14 near CIT1 (citrate synthetase)
3* in C4 amplific. high-affinity hexose perm.
transposon-related sequences at
the breakpoints.



overexpression of deleted genes, underexpression of duplicated genes



duplicated genes



deleted genes

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

By evolution of genome structure?

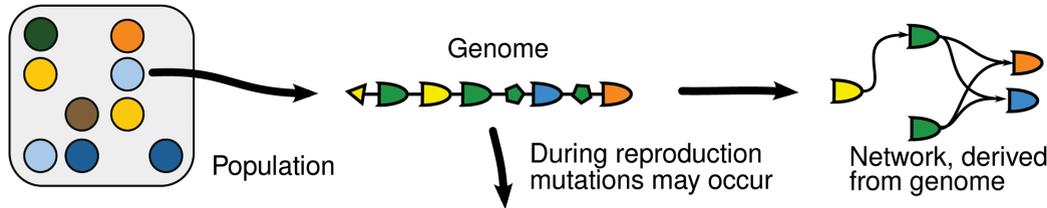
By evolution of transcriptome structure?

Crombach & H. 2007 MBE, 2008 PLOS-CompBio, Hajji Msc thesis

basic course grained representation of a genome chain of genes, TF binding sites, transposon,

“Pearls on a string model”

A Overview of the model

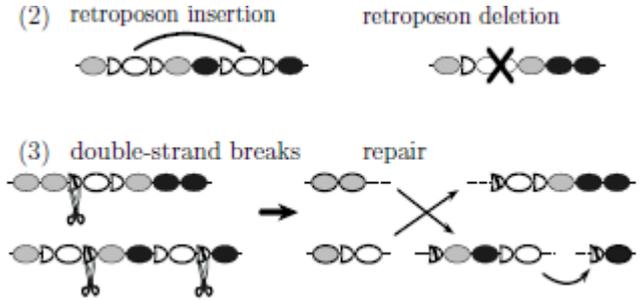
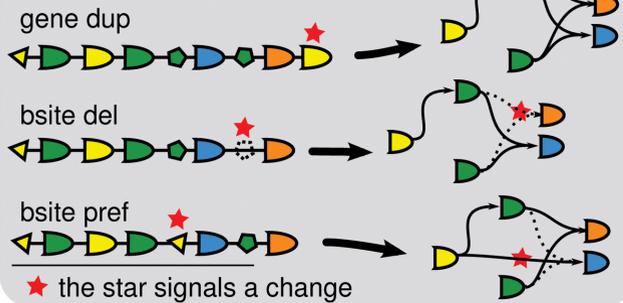


C Evolutionary targets

	A	genes on	genes off
B	genes on		
genes off			



B Effect of mutations



Genetic operators: beyond point mutations:

(DupDel: single genes, TBS; LCR: random, targeted)

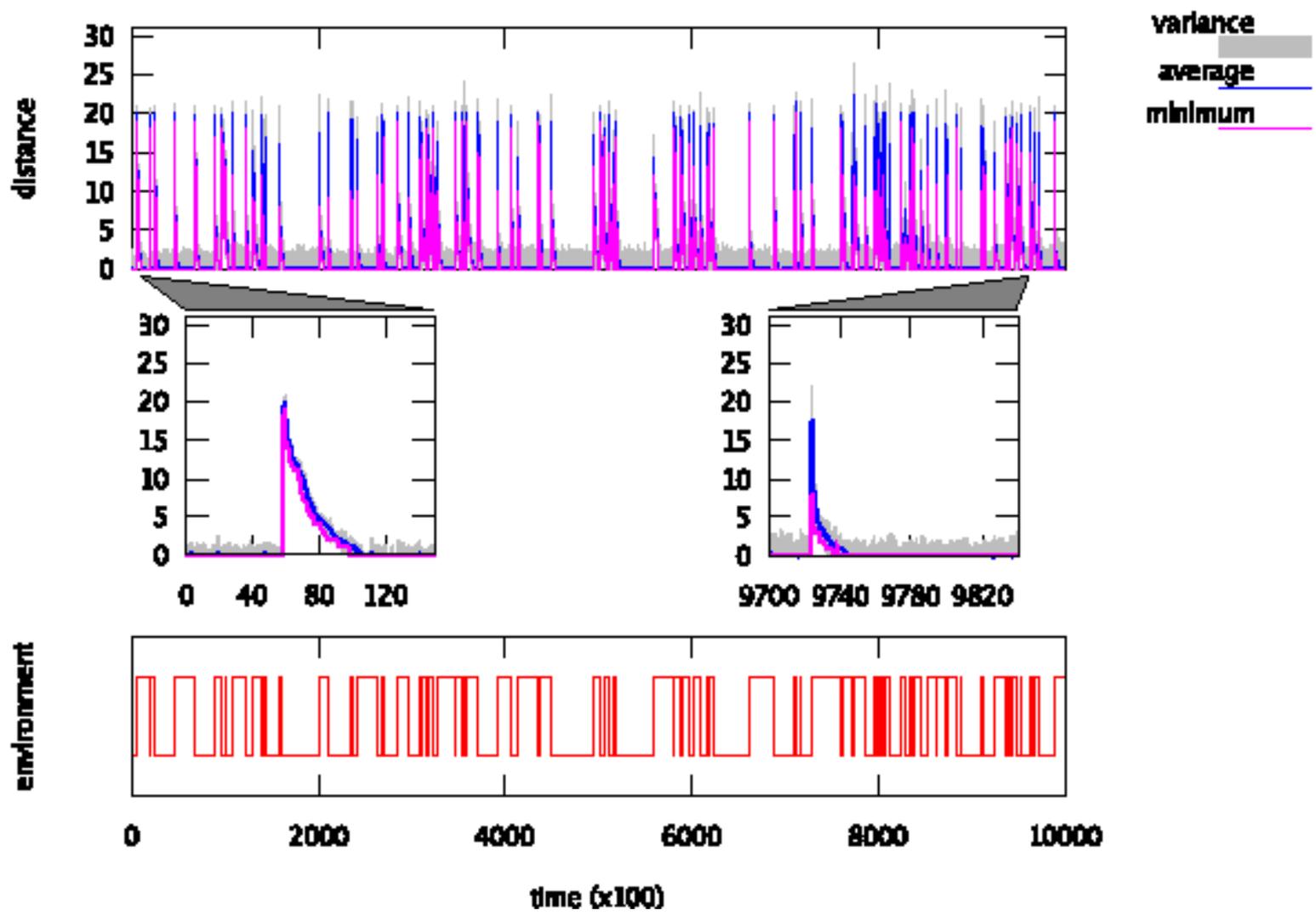
selforganization of genomes by transposon mutational dynamics evolution of evolvability

mutational dynamics

- gene duplication; gene deletion.
- transposon duplication;
- transposon deletion; leaves breakpoints
- double stranded breaks and repair
→ gross chromosomal rearrangement

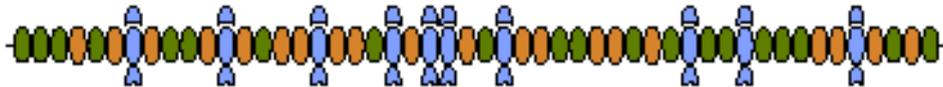
selection

- fluctuating environment
- need 2 copies of part of the genes in one environment



self organization of the genomes clustering of genes which need to be duplicated

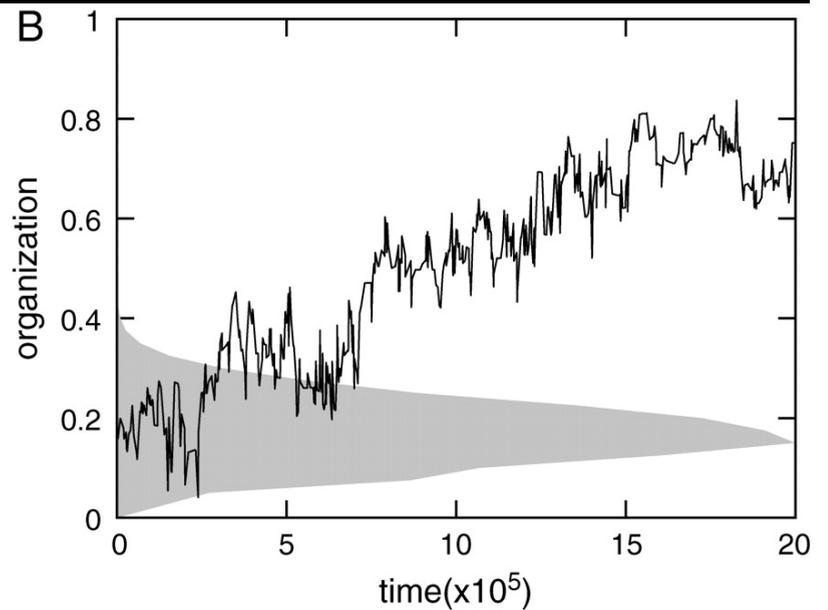
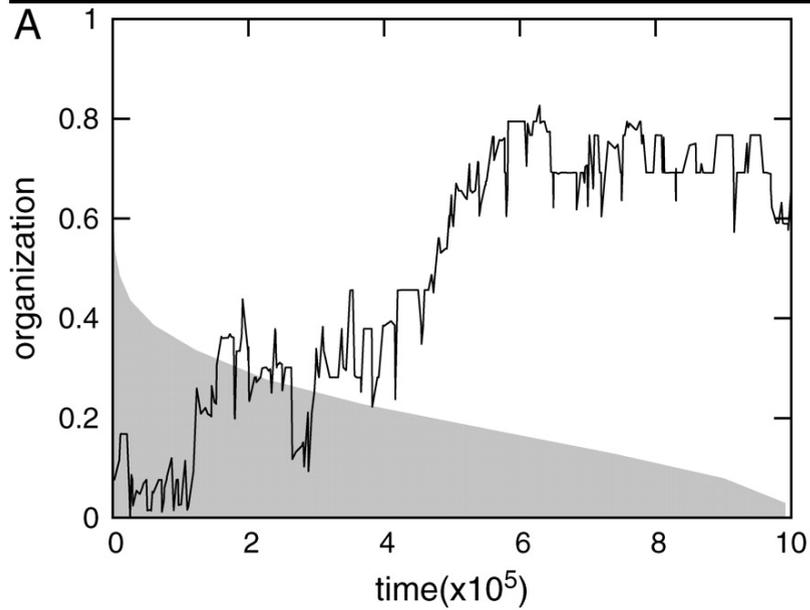
- Randomly generated genome (at start simulation):



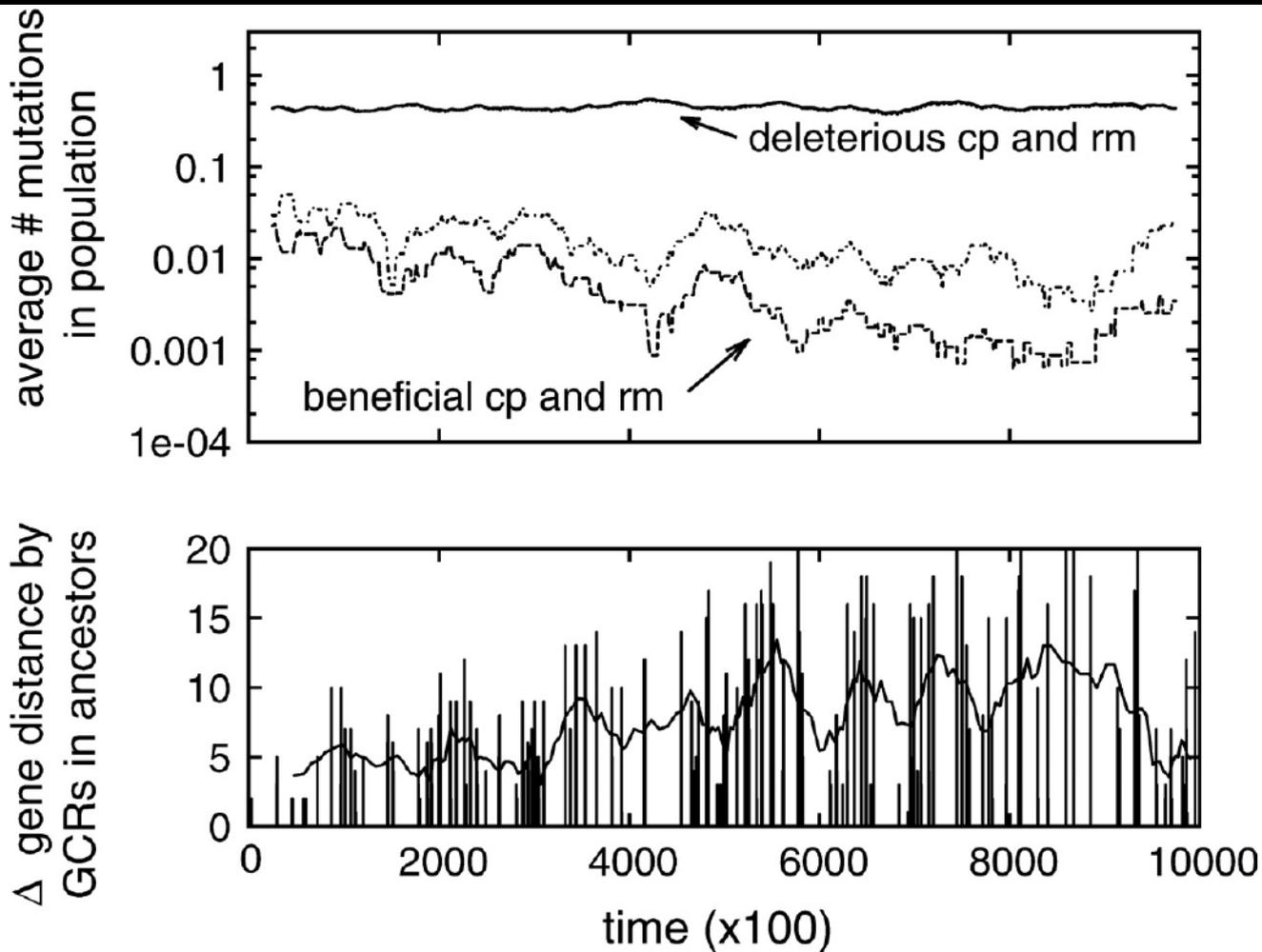
- Evolved genome, grouping of genes as a side effect:



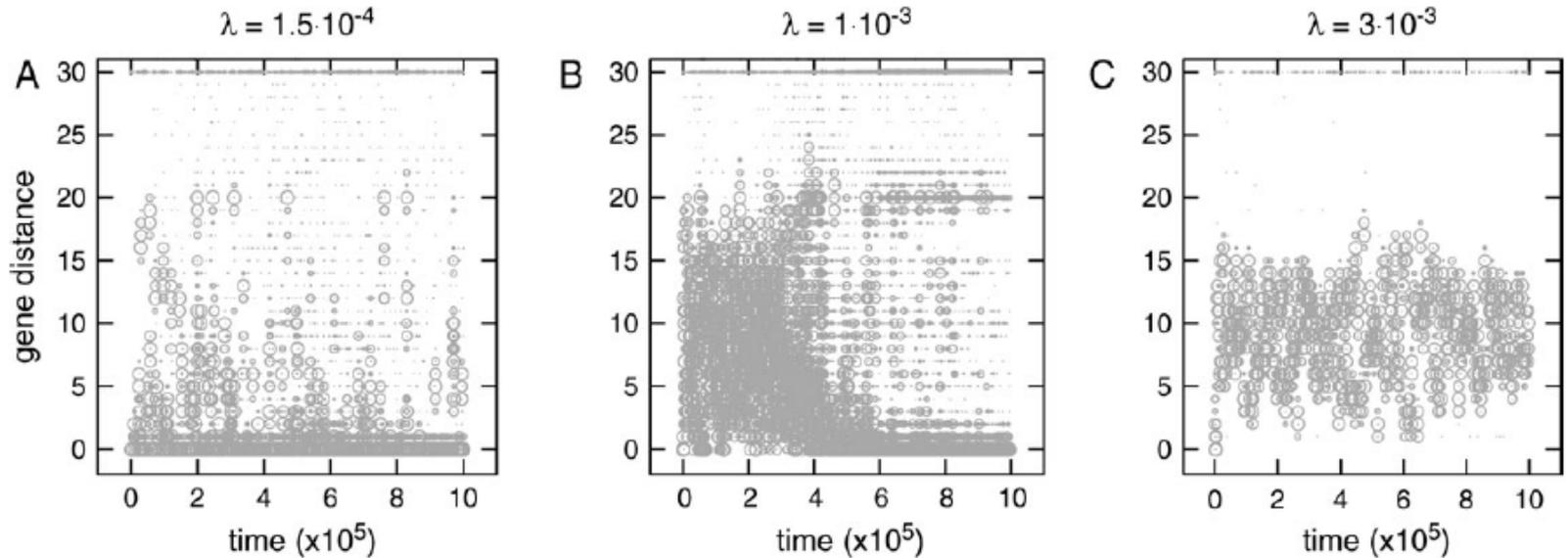
genome organization over time



mutations over time



Evolution of evolvability: timescales



genome organization evolves when no adaptation is possible
- and so enables adaptation

conclusions

Very simple demonstration of mutational priming through genome structuring

Yeast example also transposon remnants on breakpoints

Much pattern analysis research:

observation:

older transposons often in “important” (e.g. regulatory) regions

Evolution of Regulation based mutational priming

v

network dynamics and fitness

Network update:

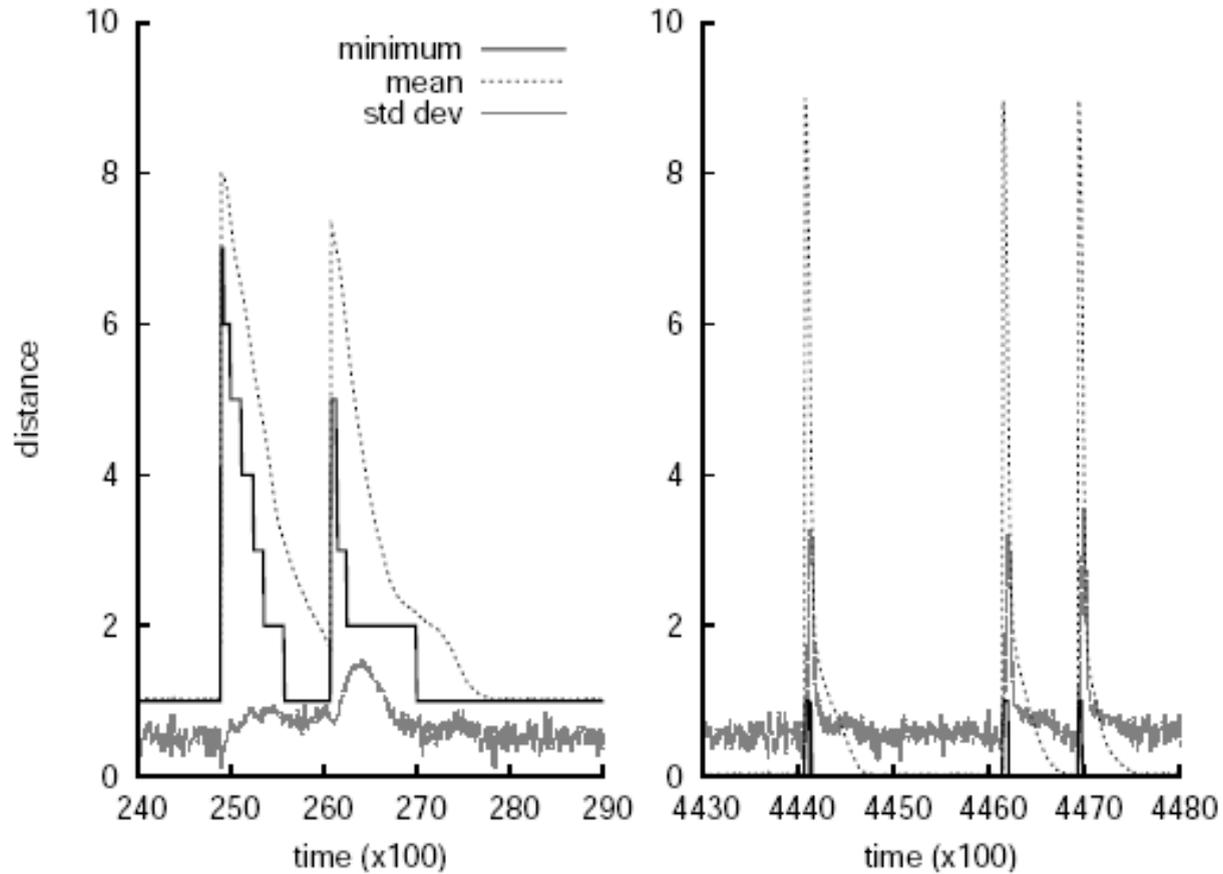
fitness:

$$s_i^{t+1} = \begin{cases} 0 & \text{if } \sum_j w_{ij}s_j^t < \theta_i \\ s_i^t & \text{if } \sum_j w_{ij}s_j^t = \theta_i \\ 1 & \text{if } \sum_j w_{ij}s_j^t > \theta_i \end{cases} \quad f = \left(1 - \frac{D}{D_{max}}\right)^p$$

Switching environment

Gene expression in attractor of GRN defines fitness
each environment defines in terms of target gene expression

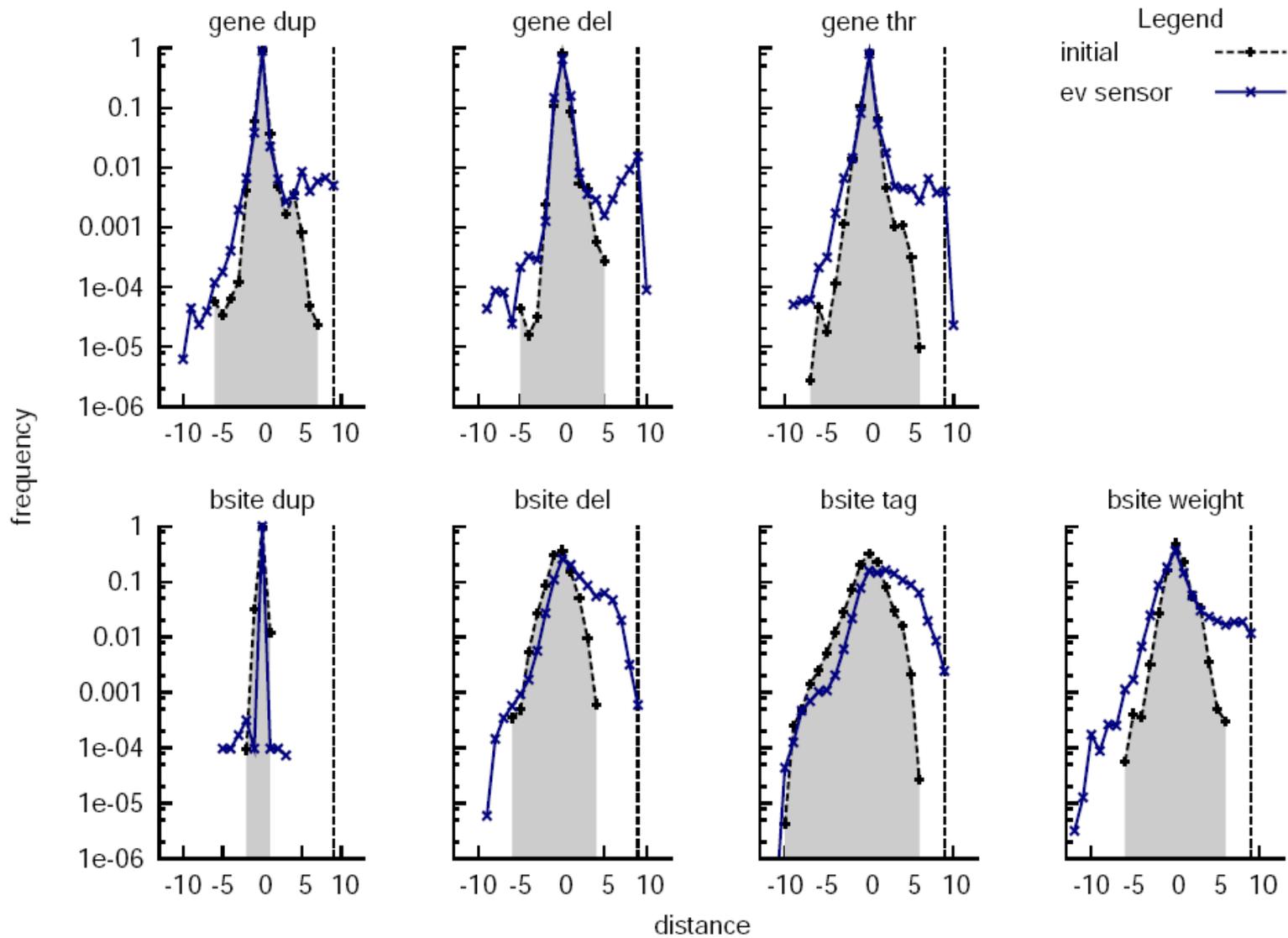
improved (faster) evolvability observed



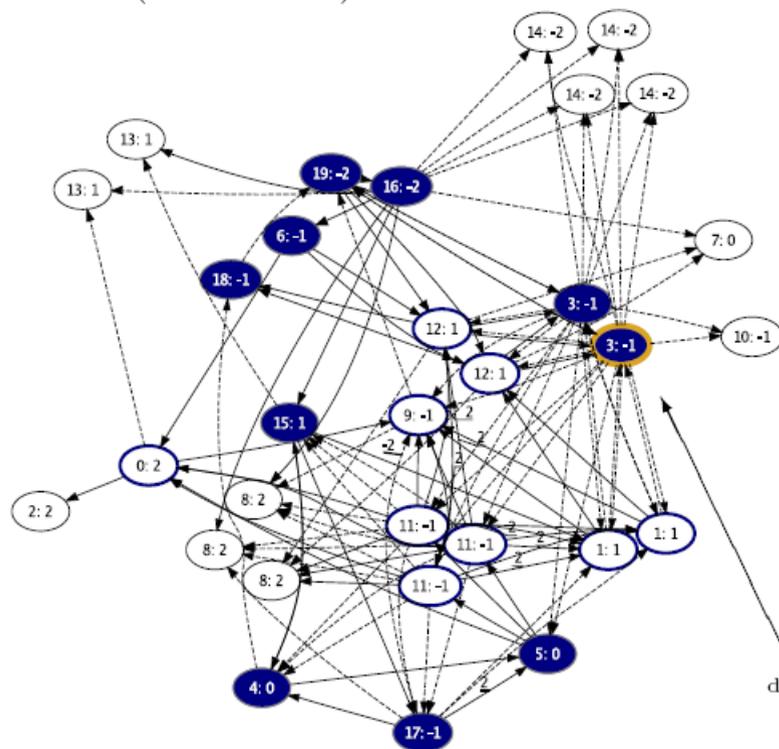
Hamming distance improvement to opposite target

Regulatory Mutational Priming:

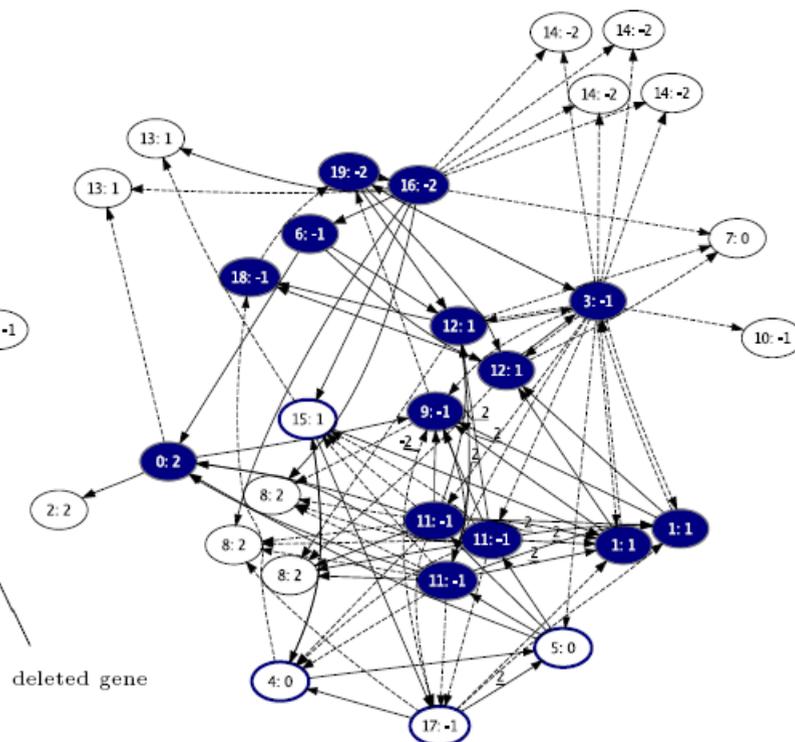
Many different mutations lead to “beneficial” adaptation

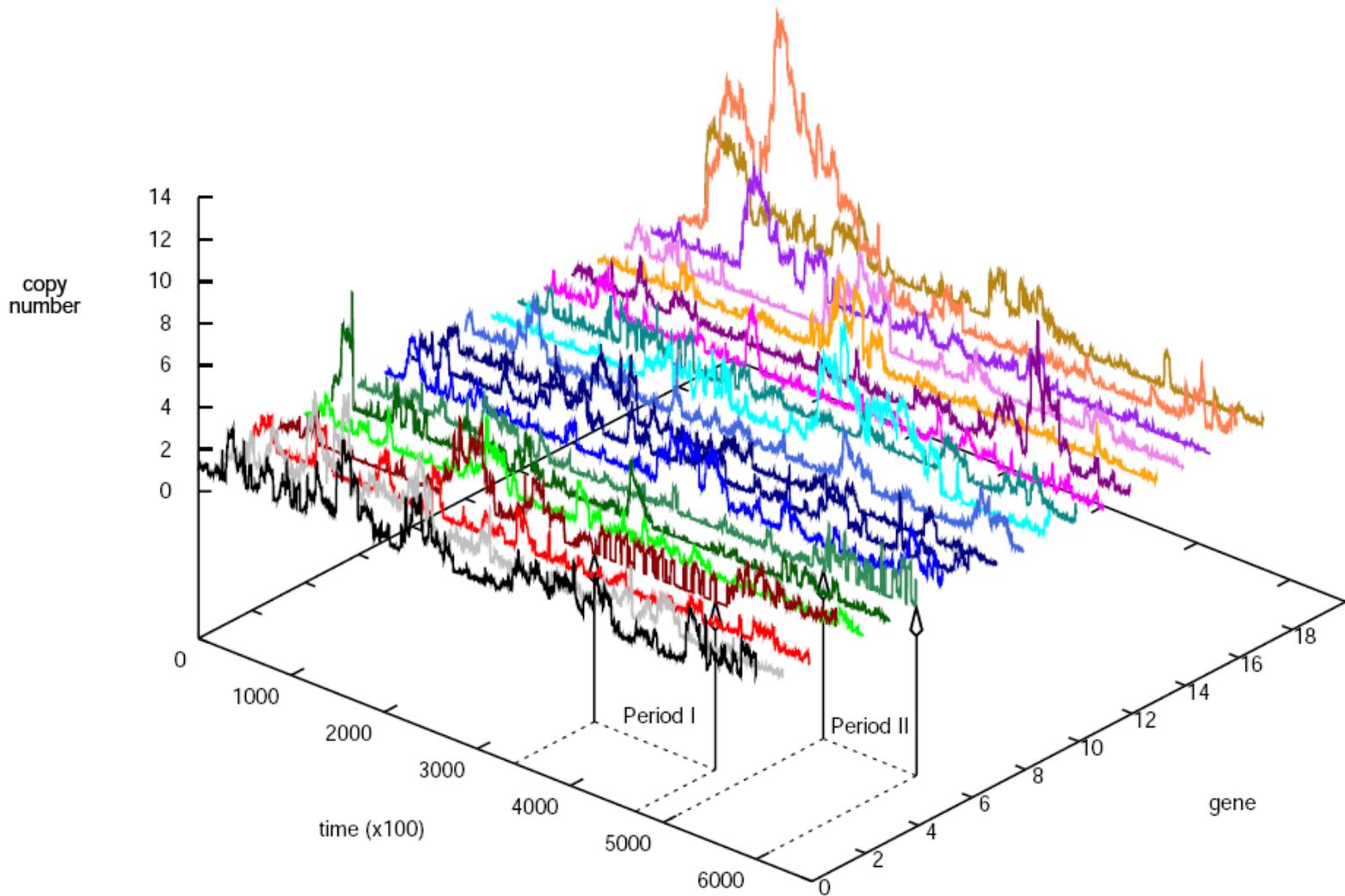


Parent (t = 457 755):



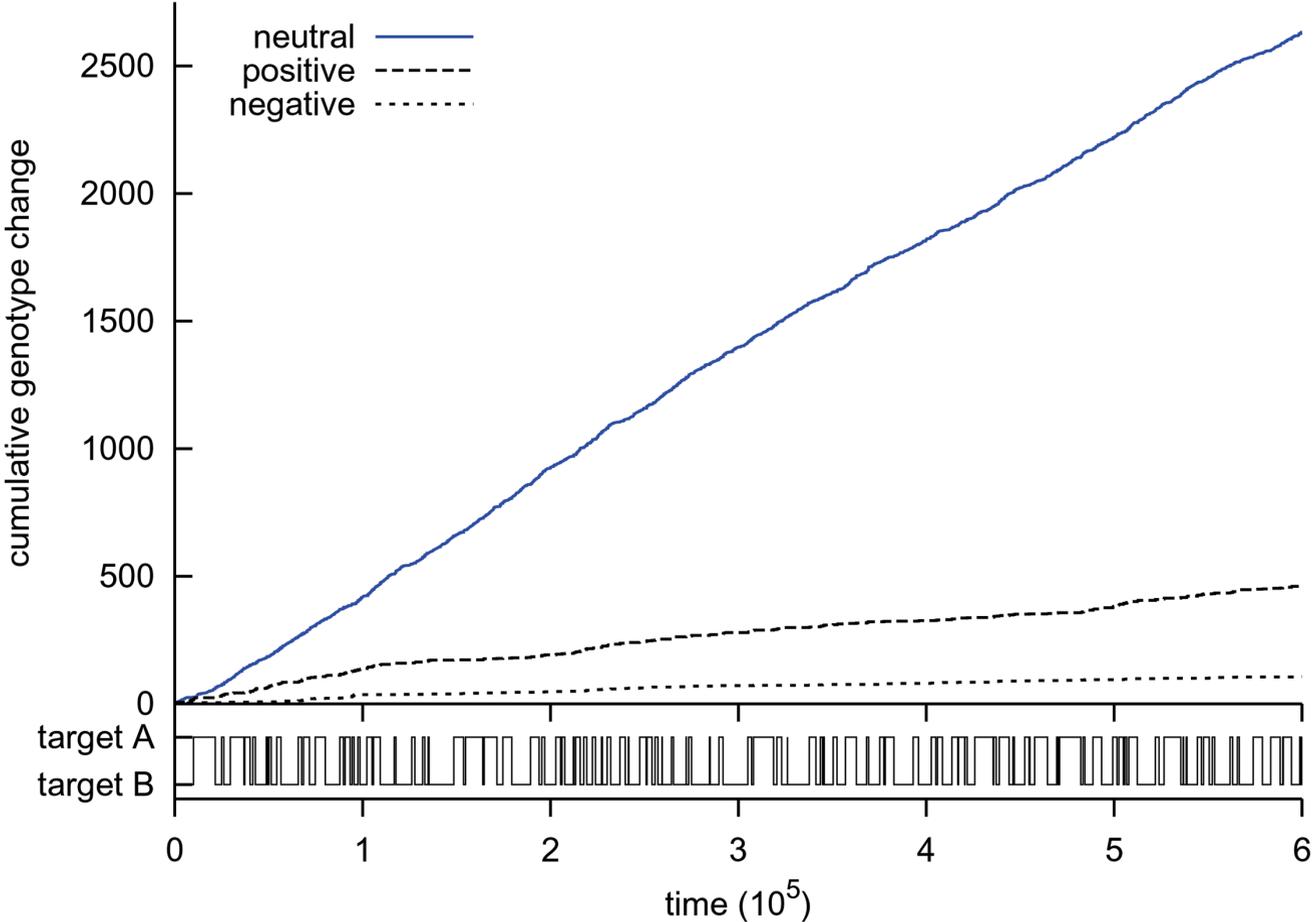
Child (t = 457 758):



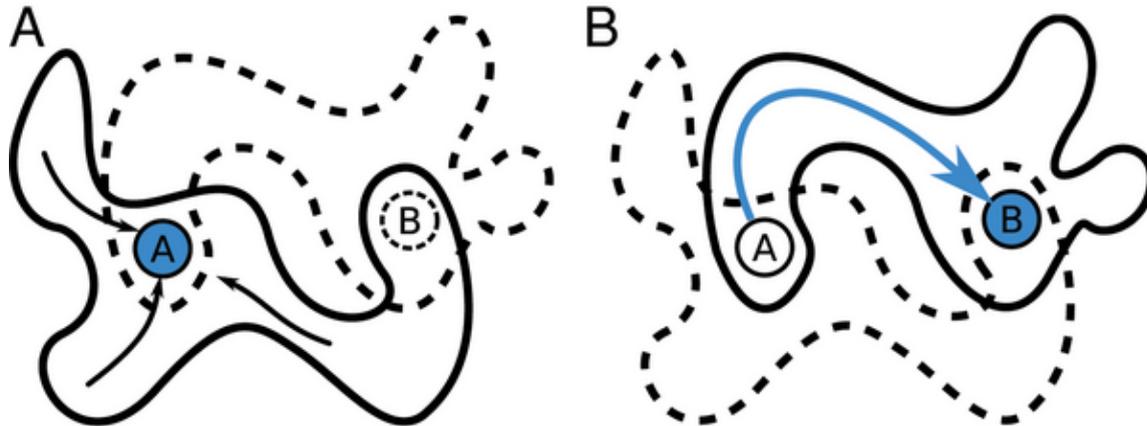


“evolutionary sensor”

Neutral drift far greater than adaptive change!



evolution of evolvability and bases of attraction



single/few Mutations destabilize attractor of env 1 and becomes state in domain of attraction of of env 2

conclusions

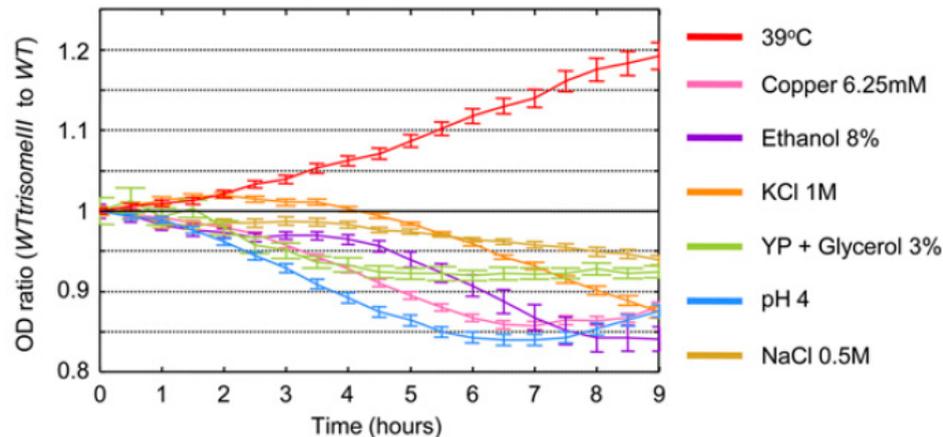
Evolution of genomes and gene regulatory networks evolution of evolvability

*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)
- State space may remain very similar despite attractor switch, but can also change drastically
- *“individual vs population based” evolution of evolvability*

Mutational Priming in Yeast: Genome organization and regulation adaptation to high temperature: short term vs long term effects

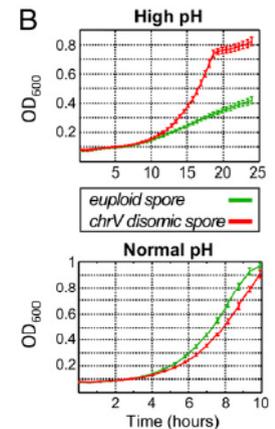
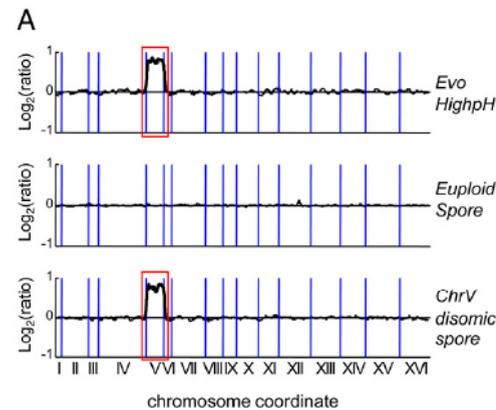
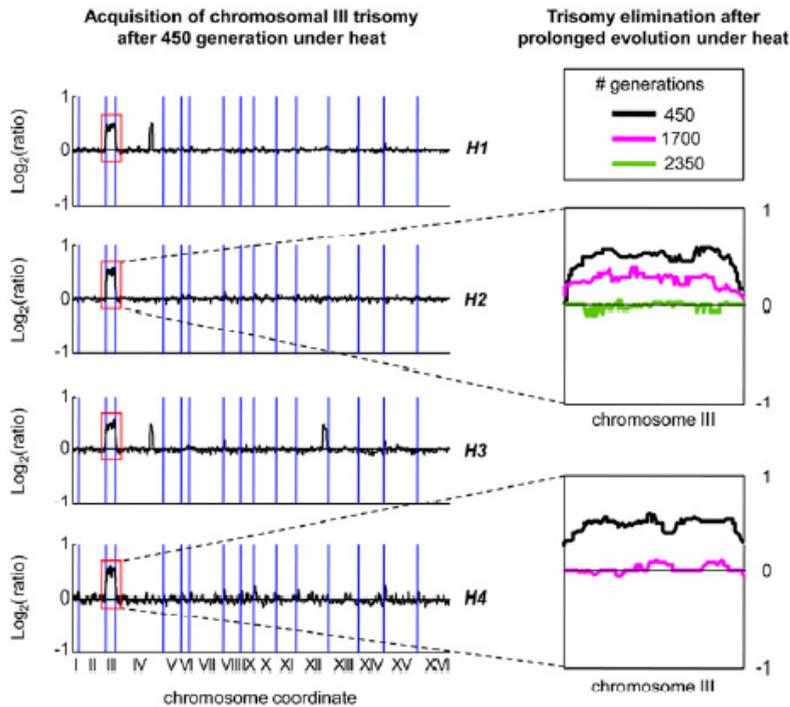
Diploid yeast adapted to 'normal' temperature of 30 C placed in 39 C. After 450 generations:
Increased growth rate (specific for temperature)



Next continued at 39 C

Yona et al, PNAS 2012: Chromosomal duplication is a transient evolutionary solution to stress

Yeast adaptation to high temperature by duplication of resp chromosome 3 (and at high ph chromosome 5)

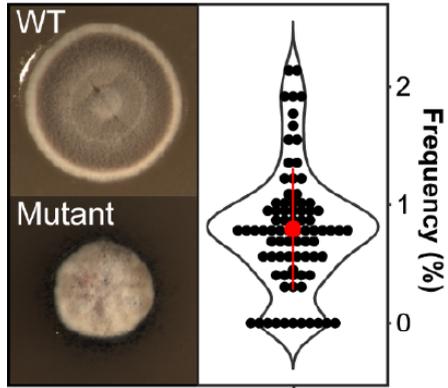


trisomy not retained

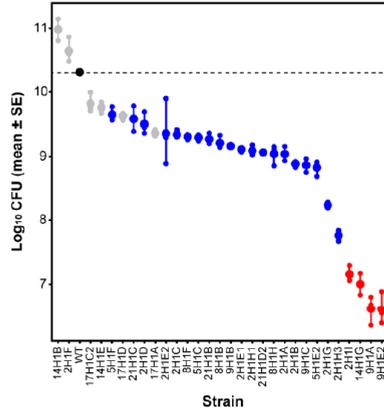
Yona et al, PNAS 2012: Chromosomal duplication is a transient evolutionary solution to stress

Mutational priming in *Streptomyces*: quasispecies

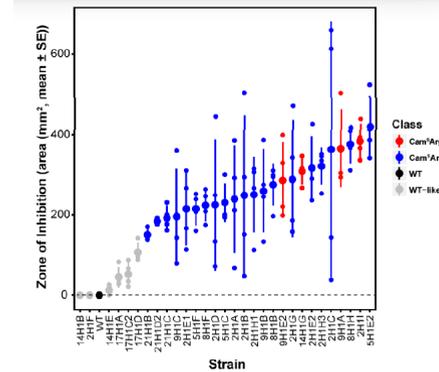
Antibiotic production in Streptomyces is organized by a division of labor through terminal genomic differentiation. Zhang ...Rozen 2020



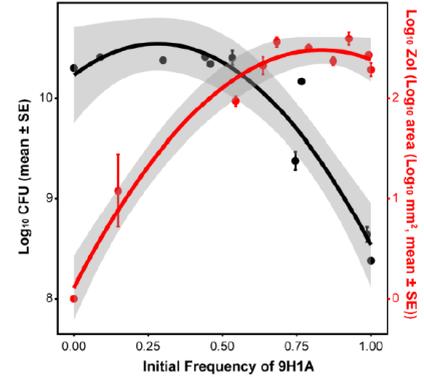
High Mut. Rate



Mut. fitness

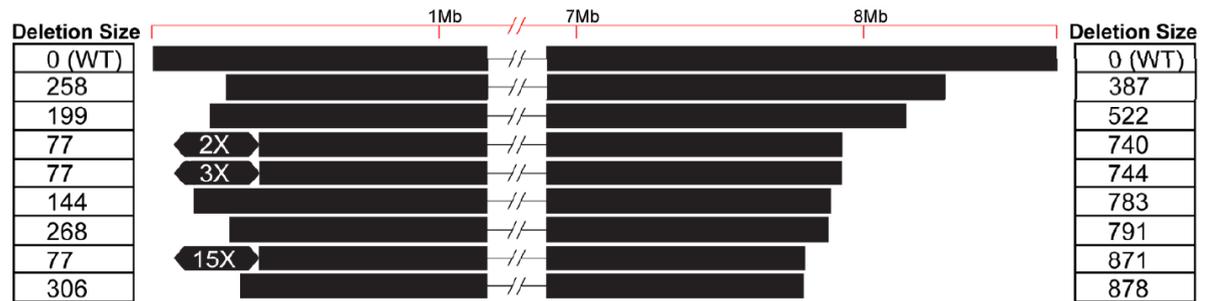
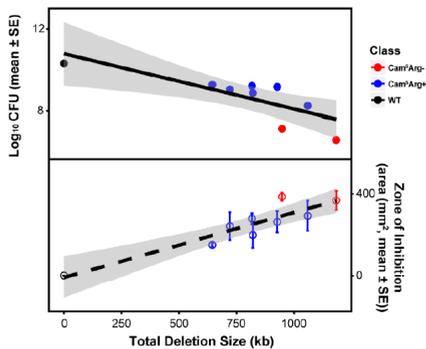


antibiotics production



QS fitness

only > 50% of mutants lowers colony fitness



Genome structure and targeted mutations