

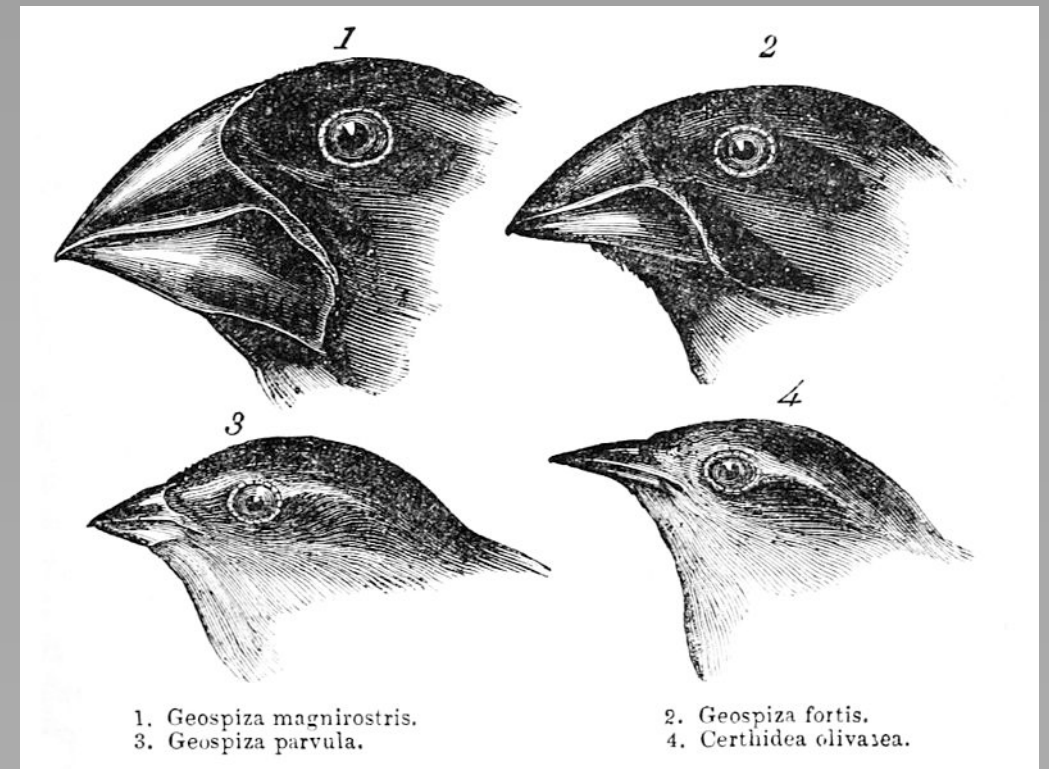
Non-equilibrium statistical physics, population genetics and evolution

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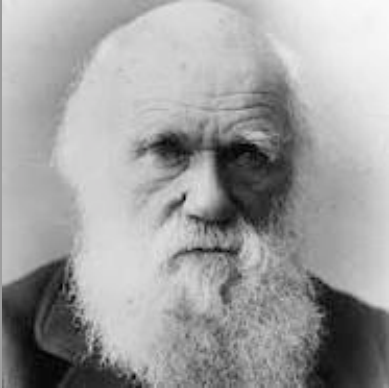
outline

- traditional view of population genetics:
mutations, recombination and selection
- questions of interest
- why all hasn't been solved yet?
- relations to statistical physics
 - spin glass (clonal interference)
 - polymers, path integrals, localization phenomena (phenotype switching)
- unusual kind of non-equilibrium statistical physics:
 - new processes like recombination
 - effects of discreteness appearing even in the “thermodynamical limit”

Darwin's finches



early ideas on evolution and the standard picture



Charles Darwin



Alfred Russel
Wallace



Gregor Mendel



Jean-Baptiste
Lamarck

$$g = \{s_1, \dots, s_L\}$$

genotype = spin configuration

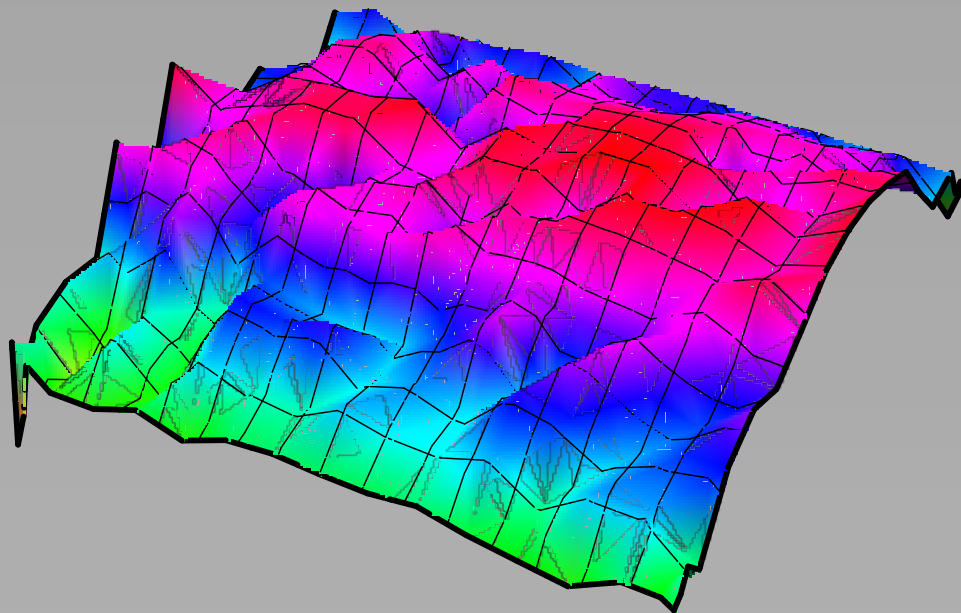
$$s_i \in \{+1, -1\}$$

allele = spin state

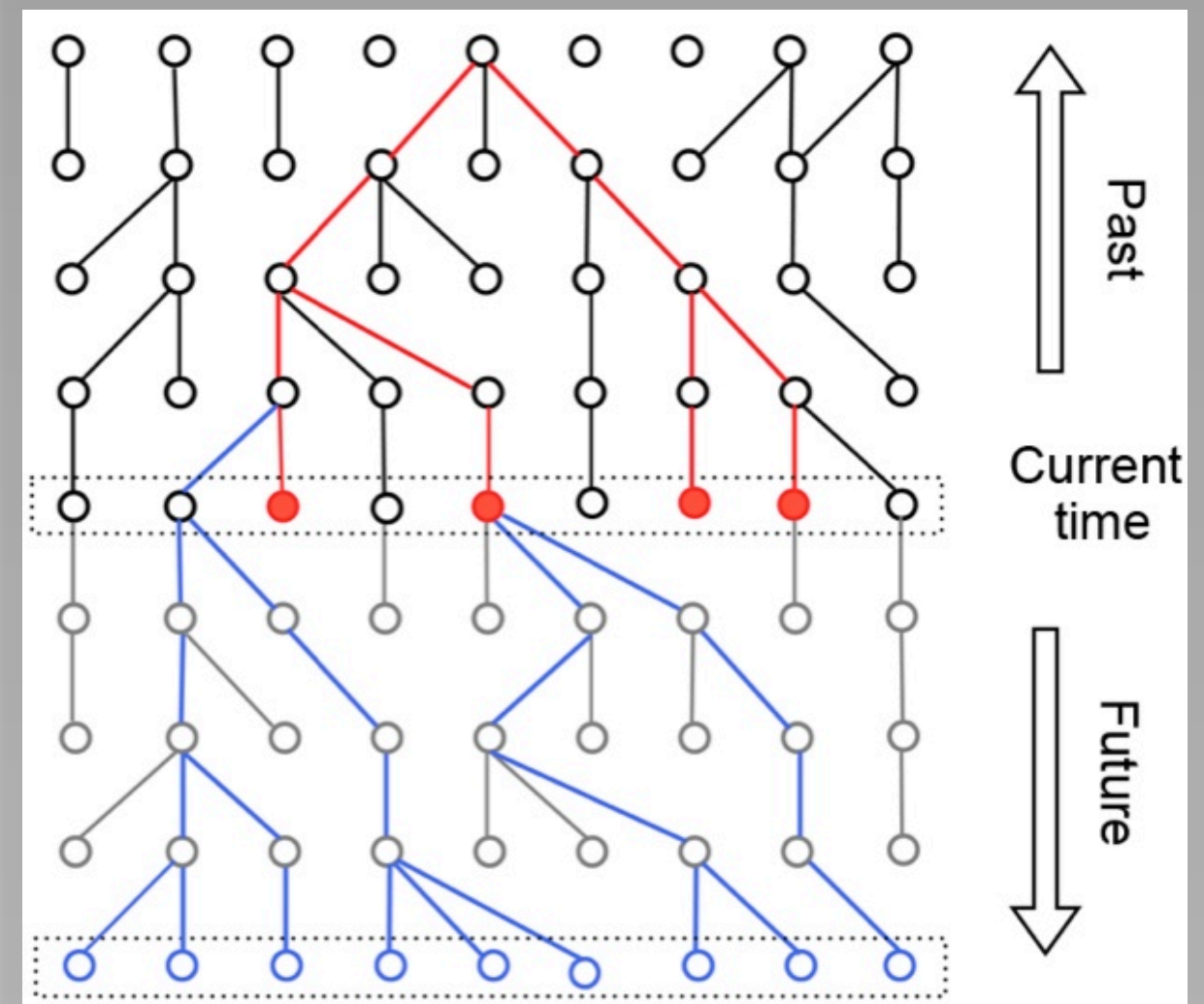
example: eye color

fitness landscape

$$F(g, \text{environment}) \quad \text{fitness} = \text{energy}$$



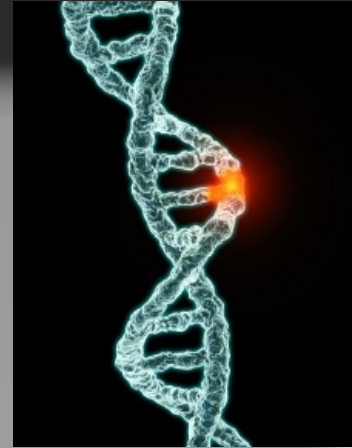
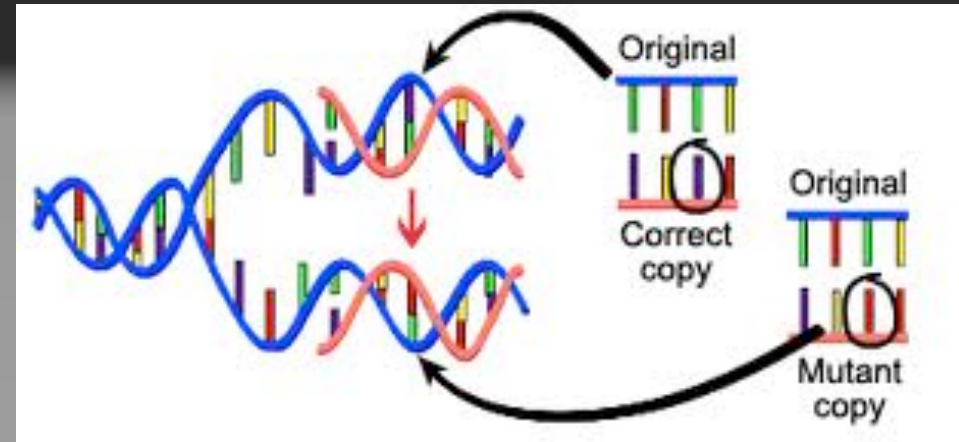
mutation creates variation: only favorable survive



mutation - spin flip

$$M : g \rightarrow Mg = \{s_1, \dots, Ms_i, \dots, s_L\}$$

clones = organisms' with the same genotype



successional mutations

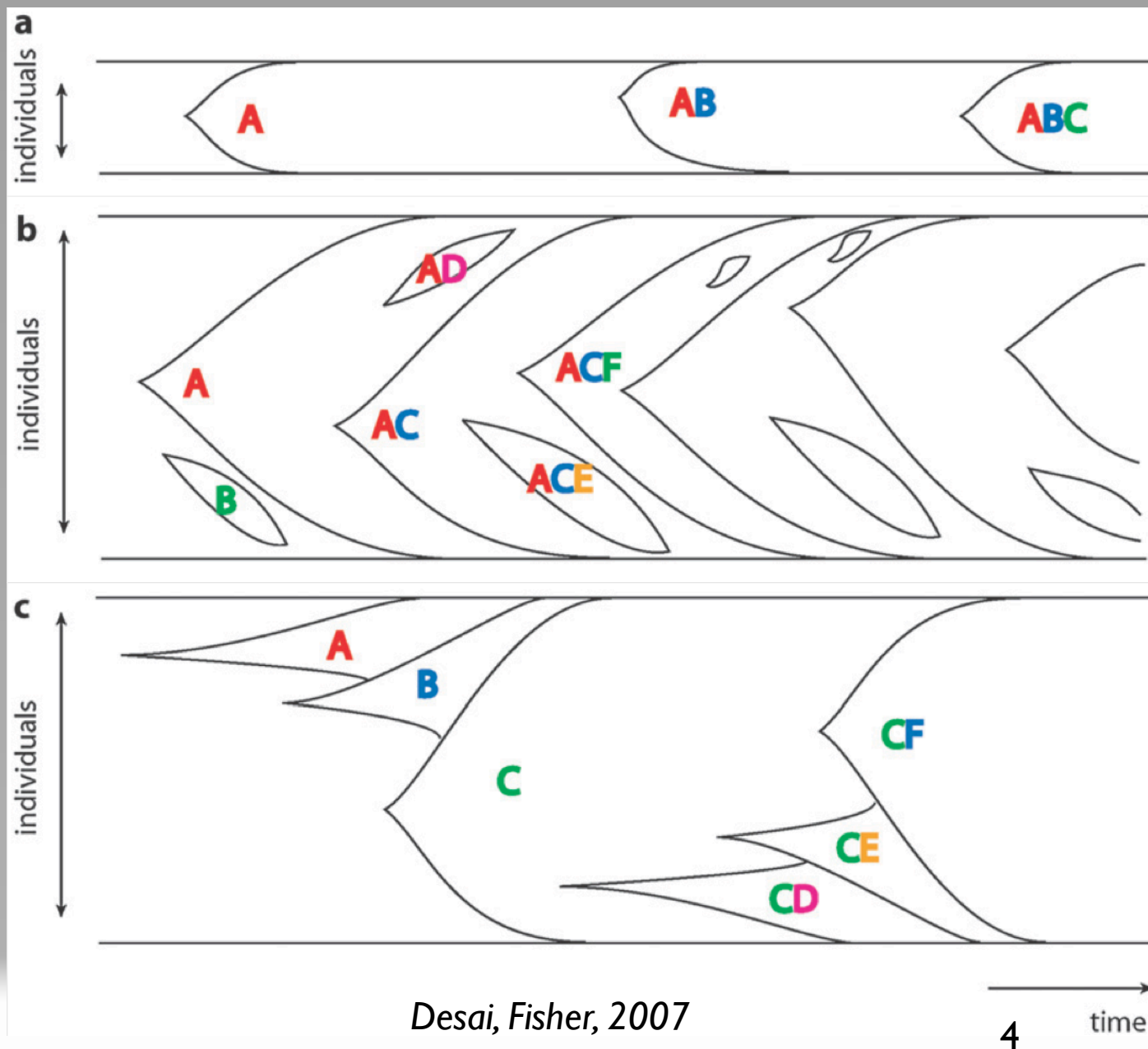
strong selection & weak mutations
present in small populations

concurrent mutations

strong selection & strong mutations
present in large populations
we see: clonal competition &
multiple mutations

clonal competition

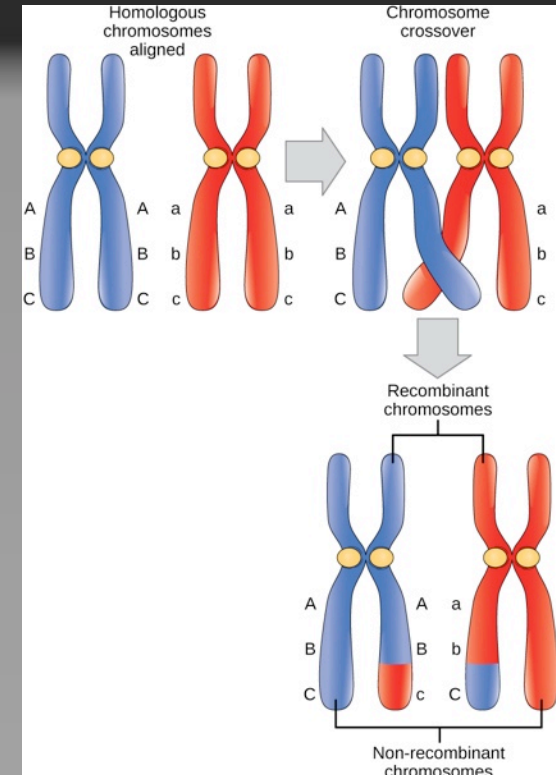
weak beneficial mutation
present in large populations
disregarding multiple mutations



recombination - no physics analog

humans and other organisms
with two sets of chromosomes

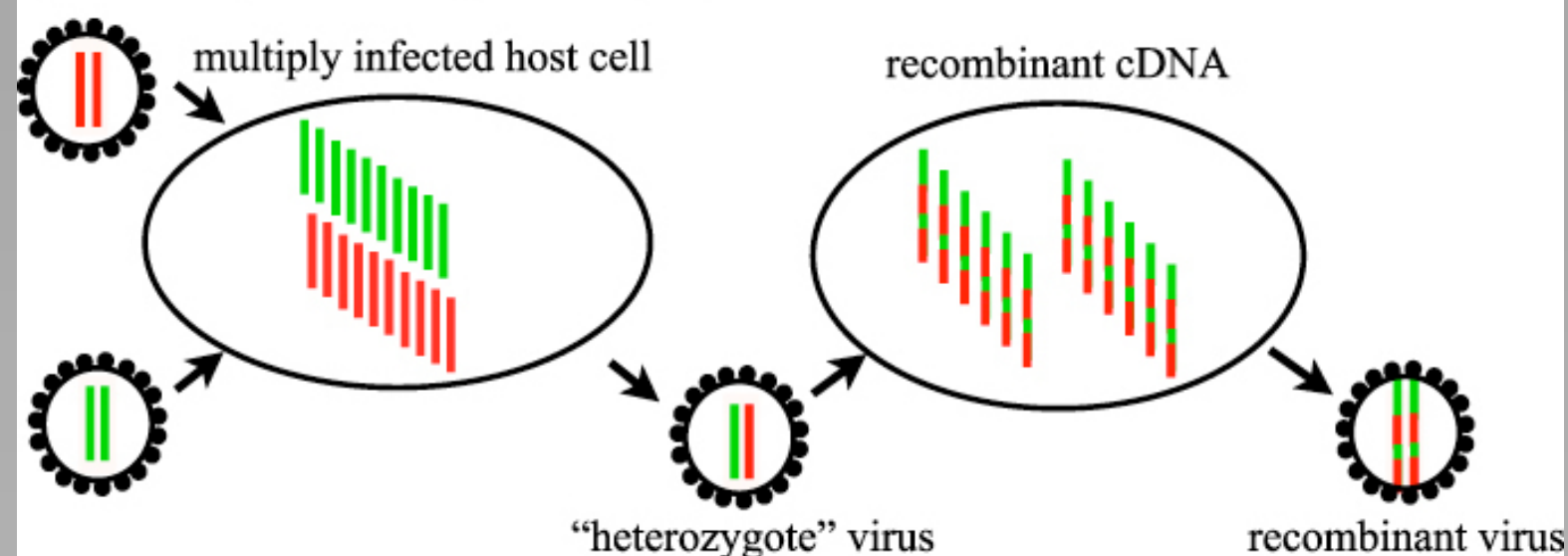
$$R : g^{(m)}, g^{(f)} \rightarrow g = \left\{ s_i | \xi^i s_i^{(m)} + (1 - \xi^i) s_i^{(f)} \right\}$$



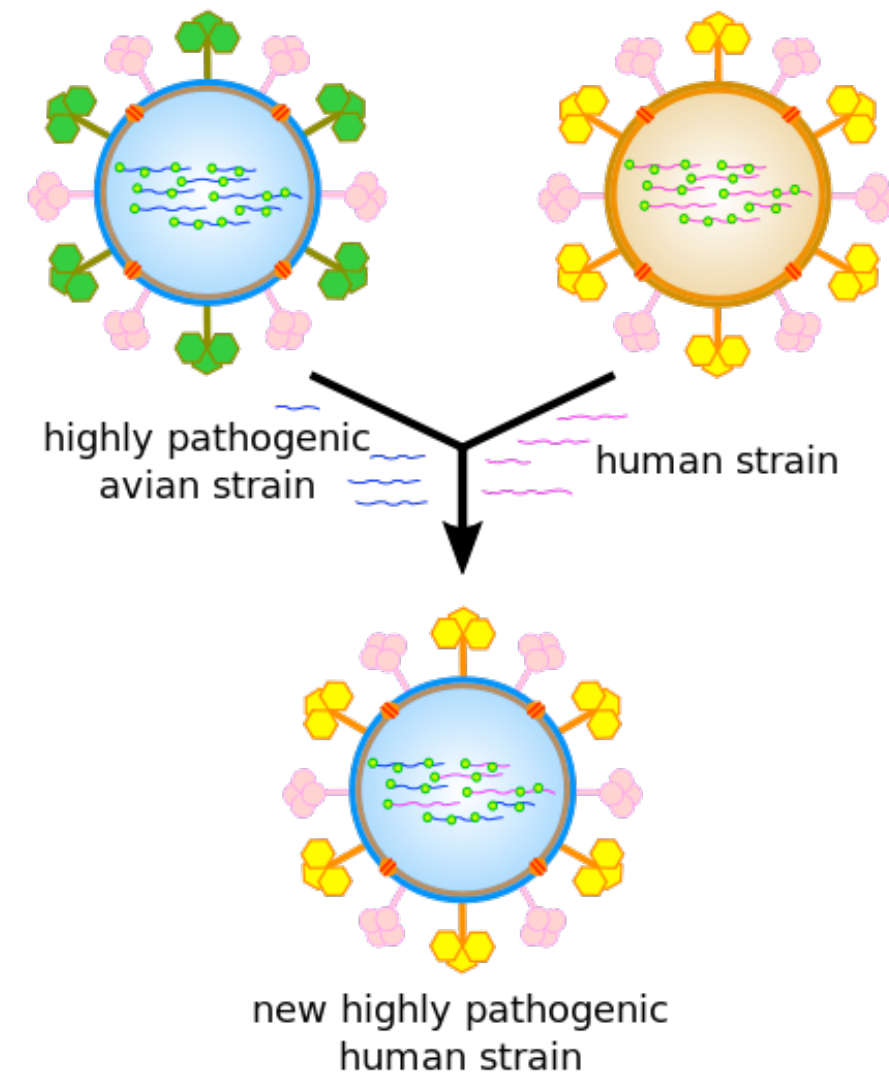
viruses
recombine in the hosts' cell

HIV: 2 RNA strains

Figure 5 from Vitaly V Ganusov et al J. Stat. Mech. (2013) P01010



influenza: 7-8 RNA strains

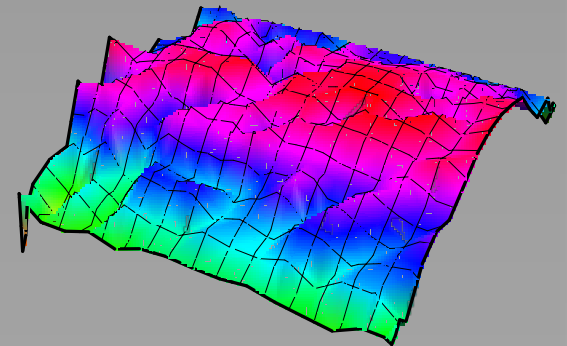


selection

$F(g, \text{environment})$ proxy for fitness = the expected reproductive success

spin interactions

$$F = \bar{F} + \sum_i f_i s_i + \sum_{i < j} f_{ij} s_i s_j + \sum_{i < j < k} f_{ijk} s_i s_j s_k + \dots$$



n_g number of individuals with genotype g

$$\dot{n}_g(t) = (F_g - \bar{F})n_g(t) + \text{noise}$$

$$\bar{F} = \frac{1}{N} \sum_g n_g F_g \quad \text{chemical potential}$$

$$\frac{n_g(t)}{N} = \frac{e^{F_g t - \int \bar{F} dt}}{N} = \frac{e^{F_g t}}{Z}$$

Boltzmann statistics
time = inverse temperature

$1 \ll N \ll 2^L$ population = random sample from genotype space

$\rho(F,t)$ fitness distribution

$$\rho(F, t) = \sum_g \delta(F - F_g)$$
$$\langle F \rangle \equiv \int dF p(F, t) F$$



R.A. Fisher

Fundamental theorem of natural selection

$$\frac{d}{dt} \langle F \rangle = \langle (F - \langle F \rangle)^2 \rangle$$

variance = selection strength, since only n_g with $F_g > \bar{F}$ grow

$$n_g(t) = e^{F_g t - \int \bar{F} dt}$$

perturbative:

weak selection & weak interactions compared to mutation & recombination that act to decorrelate spins (bio: alleles)

$$F = \underbrace{\bar{F} + \sum_i f_i s_i}_{F_0 \text{ non-interacting (bio: additive)}} + \underbrace{\sum_{i < j} f_{ij} s_i s_j + \sum_{i < j < k} f_{ijk} s_i s_j s_k + \dots}_{F_{int} \text{ interactions (bio: epistasis)}}$$

fitness distribution

$$\rho(F, t) = p(F_0, t) \omega(F_{int}, t)$$

separable solution:

distributions for non-interacting and interacting part

description with first two moments:

$$\begin{aligned} \frac{d}{dt} \langle s_i \rangle &\approx \text{function of } \langle s_i \rangle, \langle s_i s_j \rangle \\ \langle s_i s_j \rangle &\approx \text{function of } \langle s_i \rangle, \langle s_j \rangle \end{aligned}$$

Quasi-linkage equilibrium is allele competition

Neher, Shraiman 2009

no mutations

recombination rate r

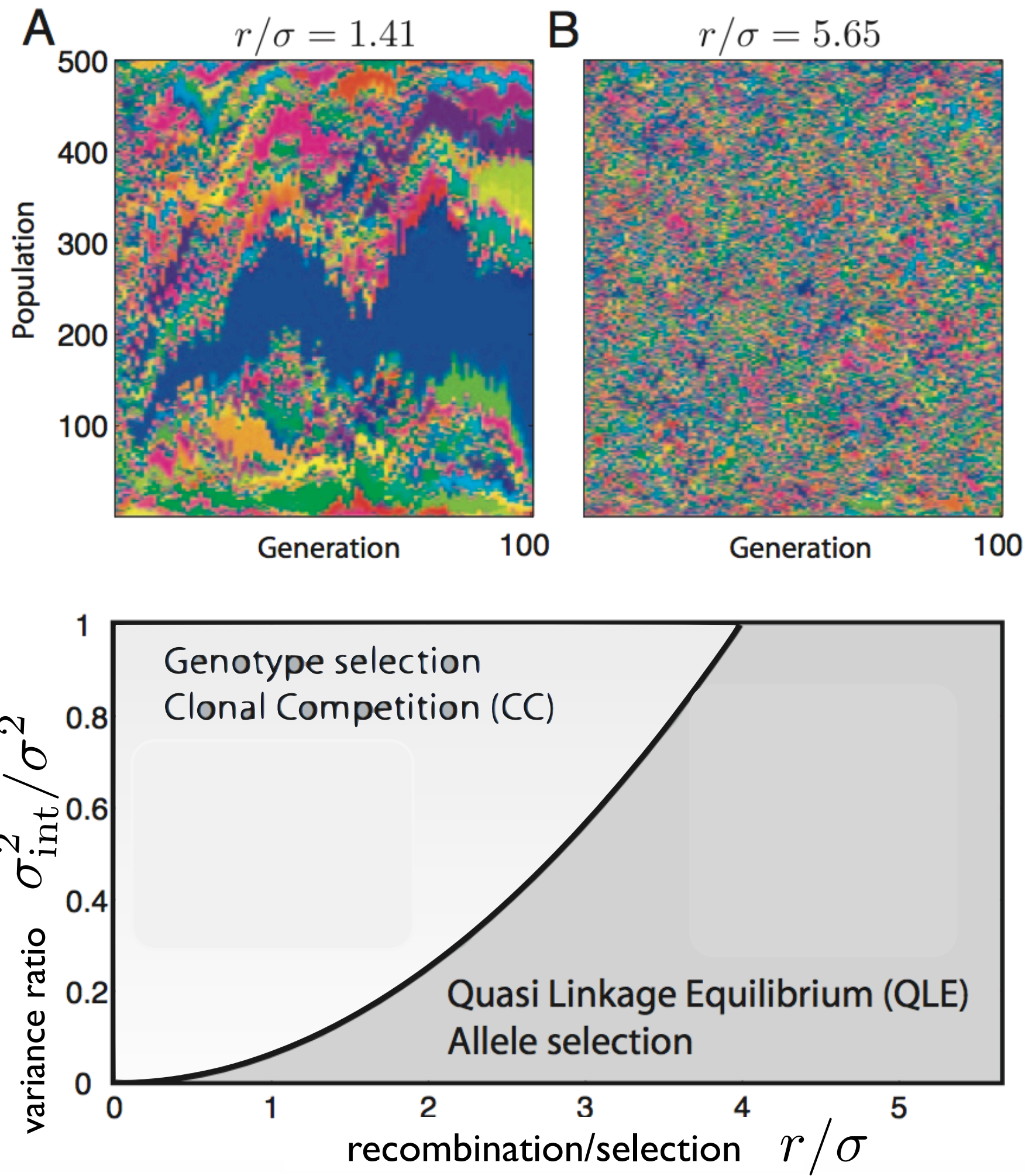
selection strength = variance of F
 $= \sigma^2$

perturbation: $r/\sigma > 1$

color = different allele
same colors grouped together

$$F = F_0 + F_{\text{int}}$$

$$\sigma_{\text{int}}^2 = \text{variance } F_{\text{int}}$$

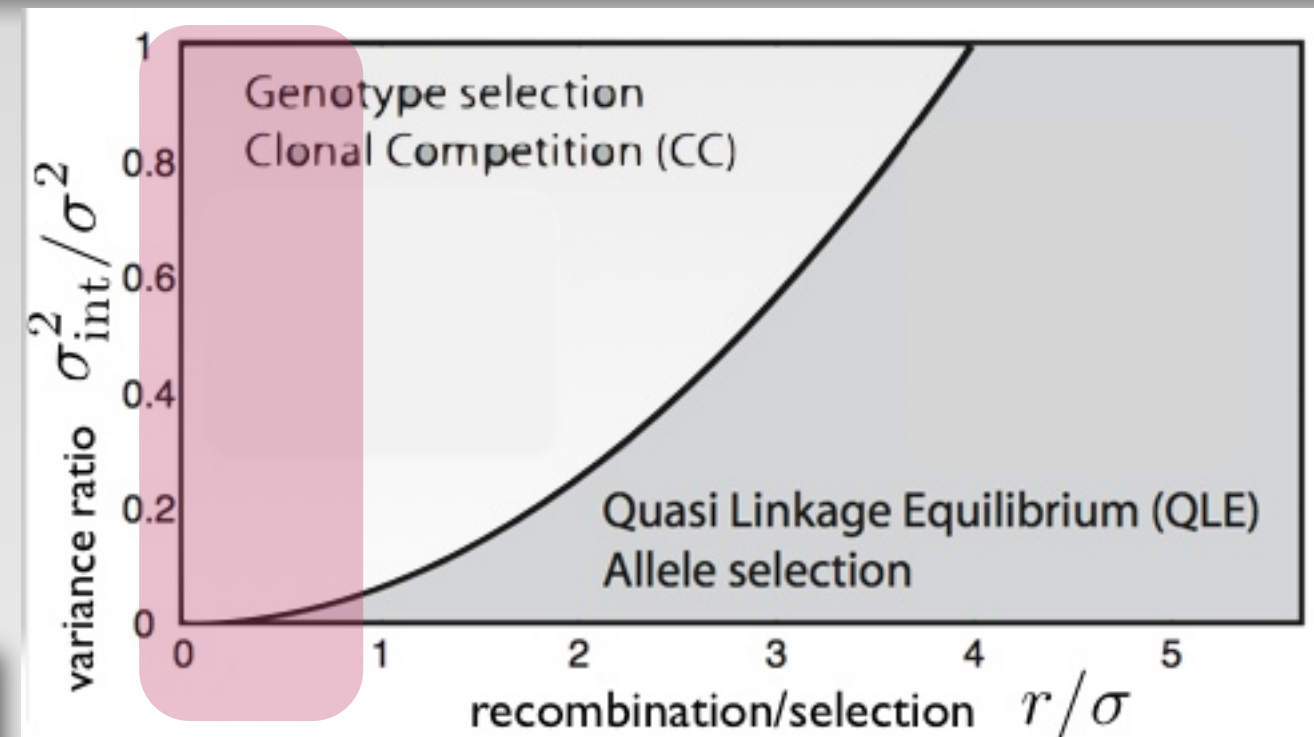


Clonal Competition

Neher, MV, Mezard, B. Shraiman 2009

Quasi-Linkage Equilibrium	Clonal Condensation
weakly nonlinear perturbative	strongly interacting non-perturbative

How do large clones form and persist?

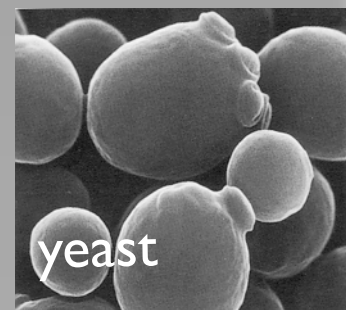


Relevant for:

facultatively recombining populations like:

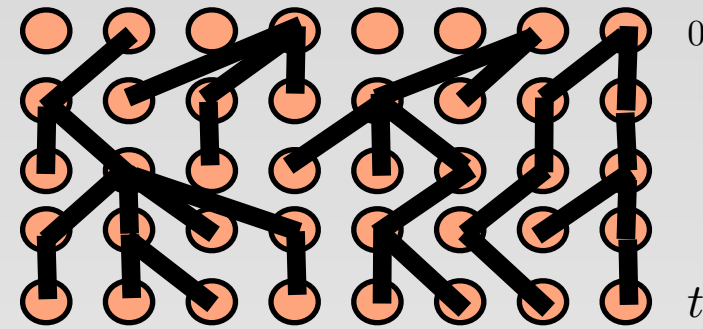
- fungi
- microorganisms
- plants
- contiguous parts of chromosome
- pathogens such as HIV and influenza
- nematodes

rapidly adapting populations,
hybridization zone, population bottlenecks, HIV, influenza

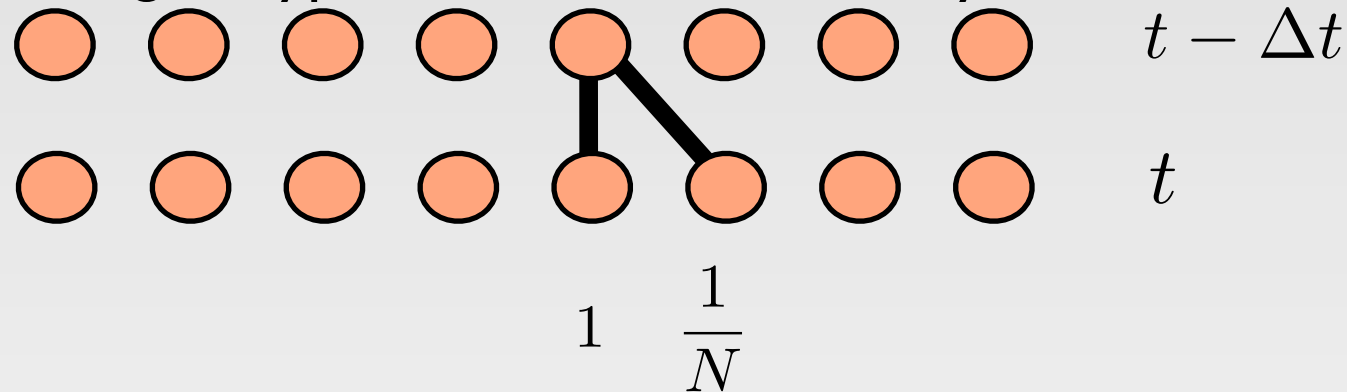


Coalescence rate = good measure of distance in genotype space

$Y(t)$ = probability that in a population of N individuals 2 are from the same clone (have a common ancestor)



if two genotypes are identical then they had a common ancestor in the past



$$\mathcal{O}(N^{-1})$$

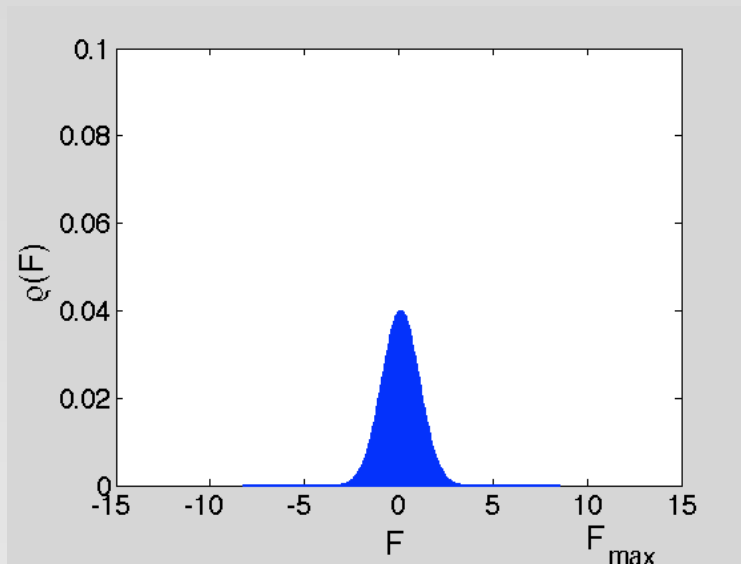
probability of two individuals in the present generation, having the same ancestor one generation in the past

actually:

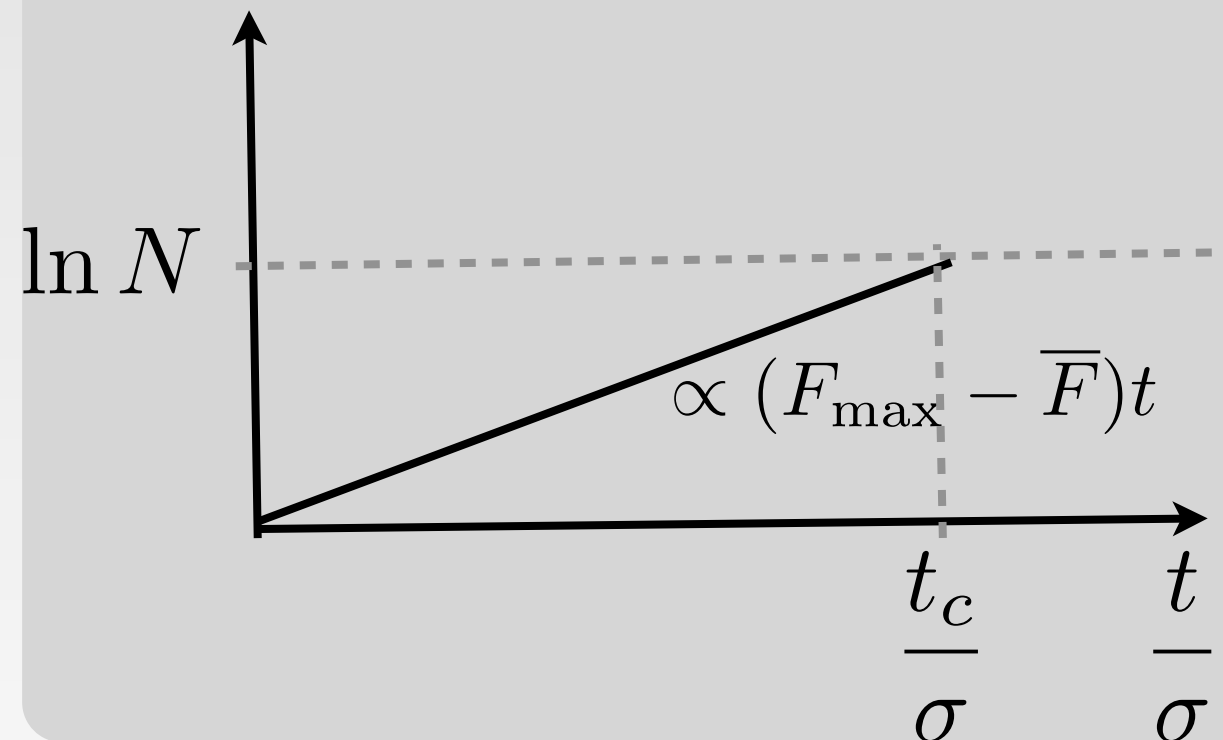
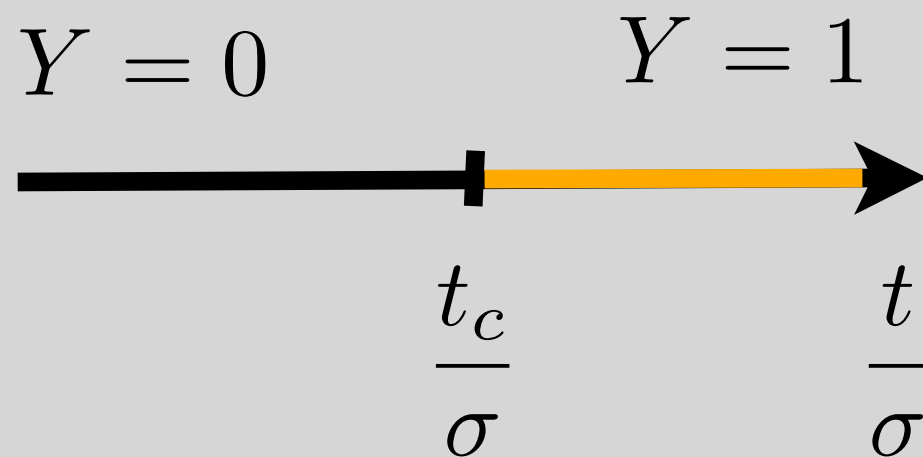
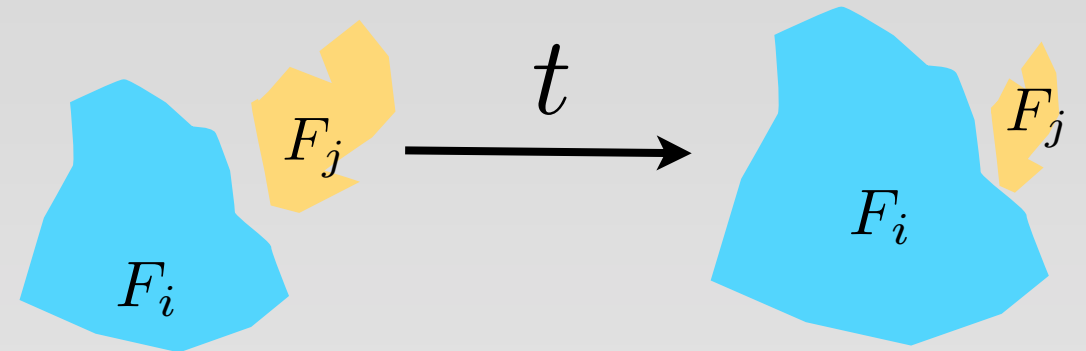
$$Y(t, \dots) = \begin{cases} \mathcal{O}(N^{-1}) \\ \mathcal{O}(1) \end{cases}$$

clonal condensation few clones grow to form a significant fraction of the population.

Selection



the most fit
genotype
overtakes the
whole population



“condensation time”

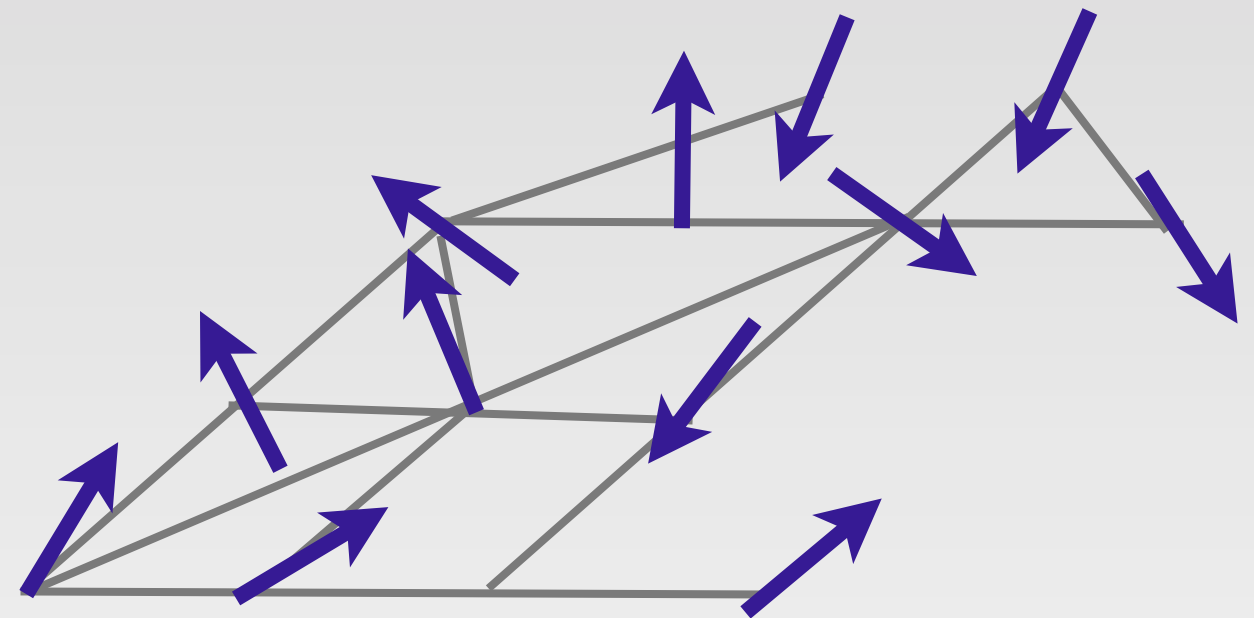
time for the fittest clone to over takes the whole population

the Random Energy Model and spin glasses

$$\rho(F) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{F^2}{2\sigma^2}}$$

many spins contribute independently; Central Limit Theorem

statistics of clones is identical to that of the Random Energy Model

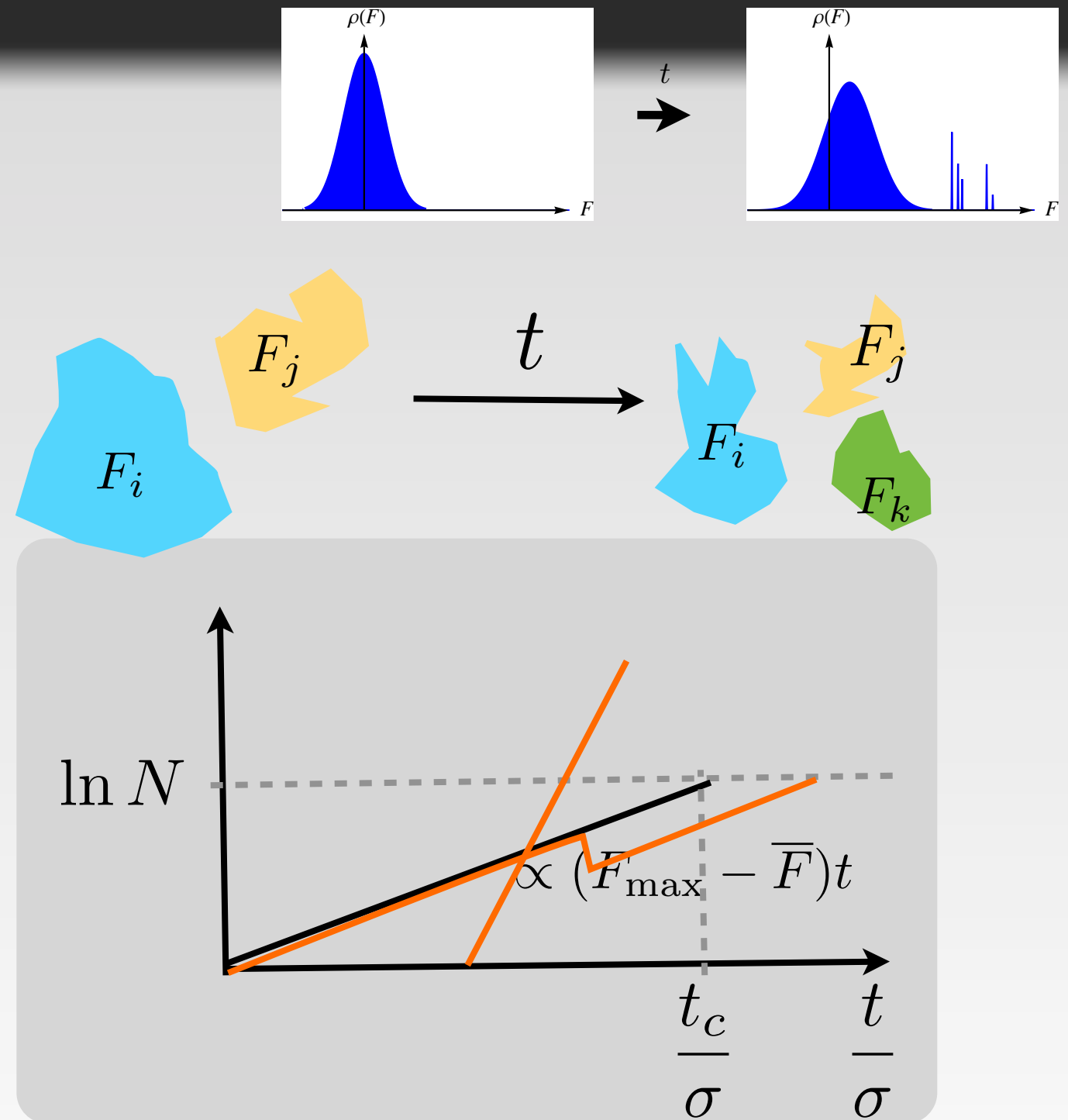
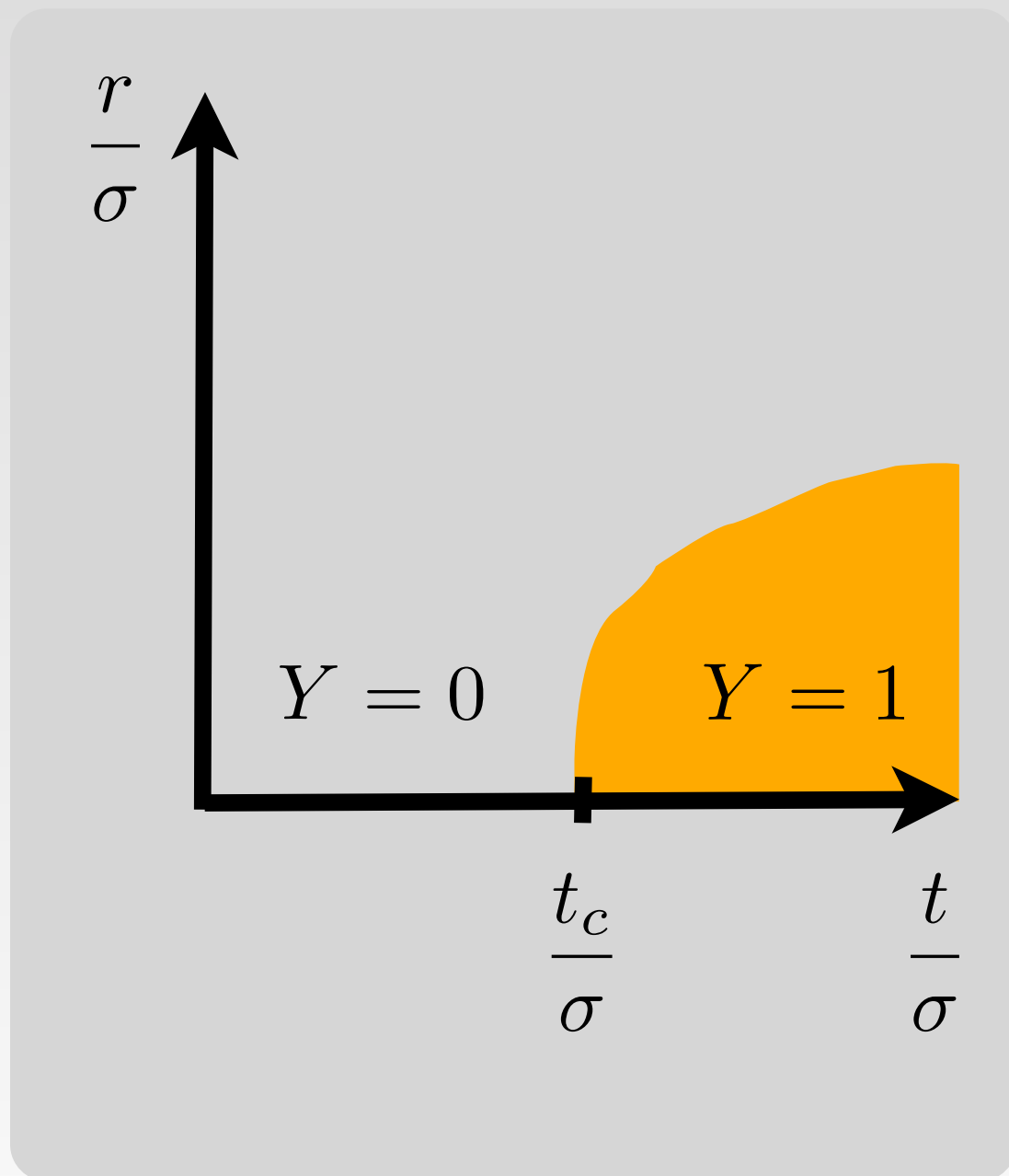


spin glass: disordered magnet with frustrated interactions and stochastic positions of spins

Random Energy Model = a set of configurations and an energy functional over those configurations.

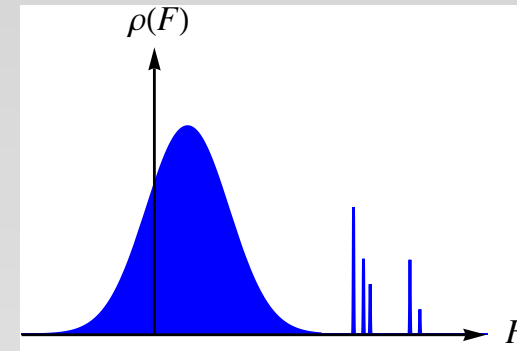
$\{E_i\}$ are i.i.d. random variables drawn from $\rho(E)$
“sample” = particular realization of such a process

Recombination - reshuffles genetic variations and produces novel combinations of existing alleles



$$\dot{n}(\mathbf{g}) = (F(\{\mathbf{g}\}) - \bar{F}(t)) n(\mathbf{g}) + r \left[N^{-1} \sum_{\mathbf{g}', \mathbf{g}''} K(\mathbf{g}|\mathbf{g}', \mathbf{g}'') n(\mathbf{g}') n(\mathbf{g}'') - n(\mathbf{g}) \right]$$

Condensation $r = 0$



At short t the averages are dominated by vicinity of the peak of $\rho(F)$

At $t > t_c$ the dominant contribution shifts to the leading edge of the distribution

With time the population shifts to fitter and fitter genotypes and eventually condenses.

$$\langle Y_t \rangle = \left\langle \sum_i \left(\frac{n_i(t)}{\sum_j n_j(t)} \right)^2 \right\rangle$$

probability of two individuals
being identical
spin-glass order parameter

$$\langle Y_t \rangle = \left\langle \sum_{i=1}^N \int_0^\infty dz z e^{2(F_i - \bar{F})t - z \sum_{i'=1}^N e^{(F_{i'} - r - \bar{F})t}} \right\rangle$$

$$\langle Y_t \rangle = N \int_0^\infty dz z \int dF_i \rho(F_i) n_i^2(t) e^{-z n_i(t)} \left[\int dF_j \rho(F_j) e^{-z n_j(t)} \right]^{N-1}$$

Effective model

$$\sum_{i=1}^{N-M} e^{(F_i - r)t - \int_0^t dt' \bar{F}(t')} + \sum_{j=1}^M e^{(F_j - r)(t - t_j) - \int_{t_j}^t dt' \bar{F}(t')} = N$$

forefathers

recombinants

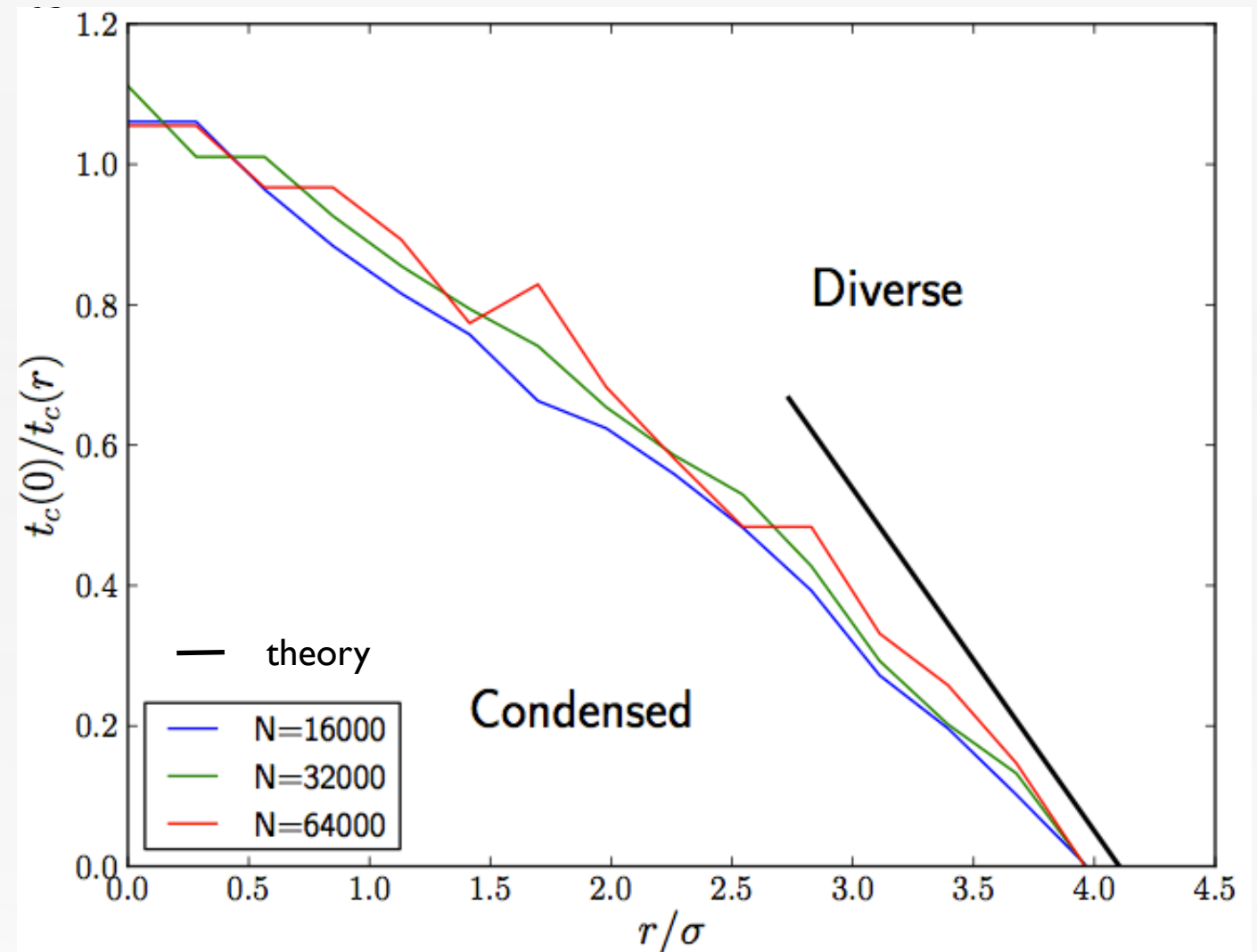
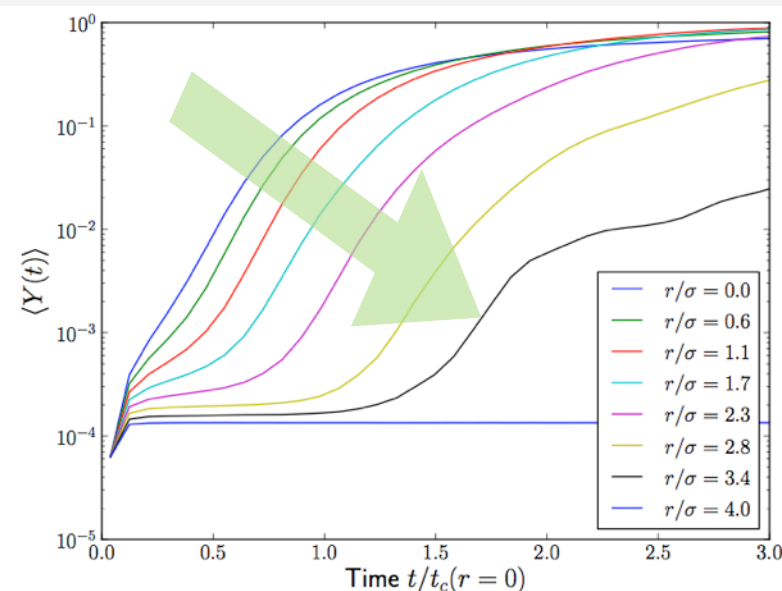
$$\langle Y_t \rangle = \left\langle \sum_i \left(\frac{n_i(t)}{\sum_j n_j(t)} \right)^2 \right\rangle$$

averaging:

- statistics of fitness
- Poisson process - arrival of M recombinants at times t_j

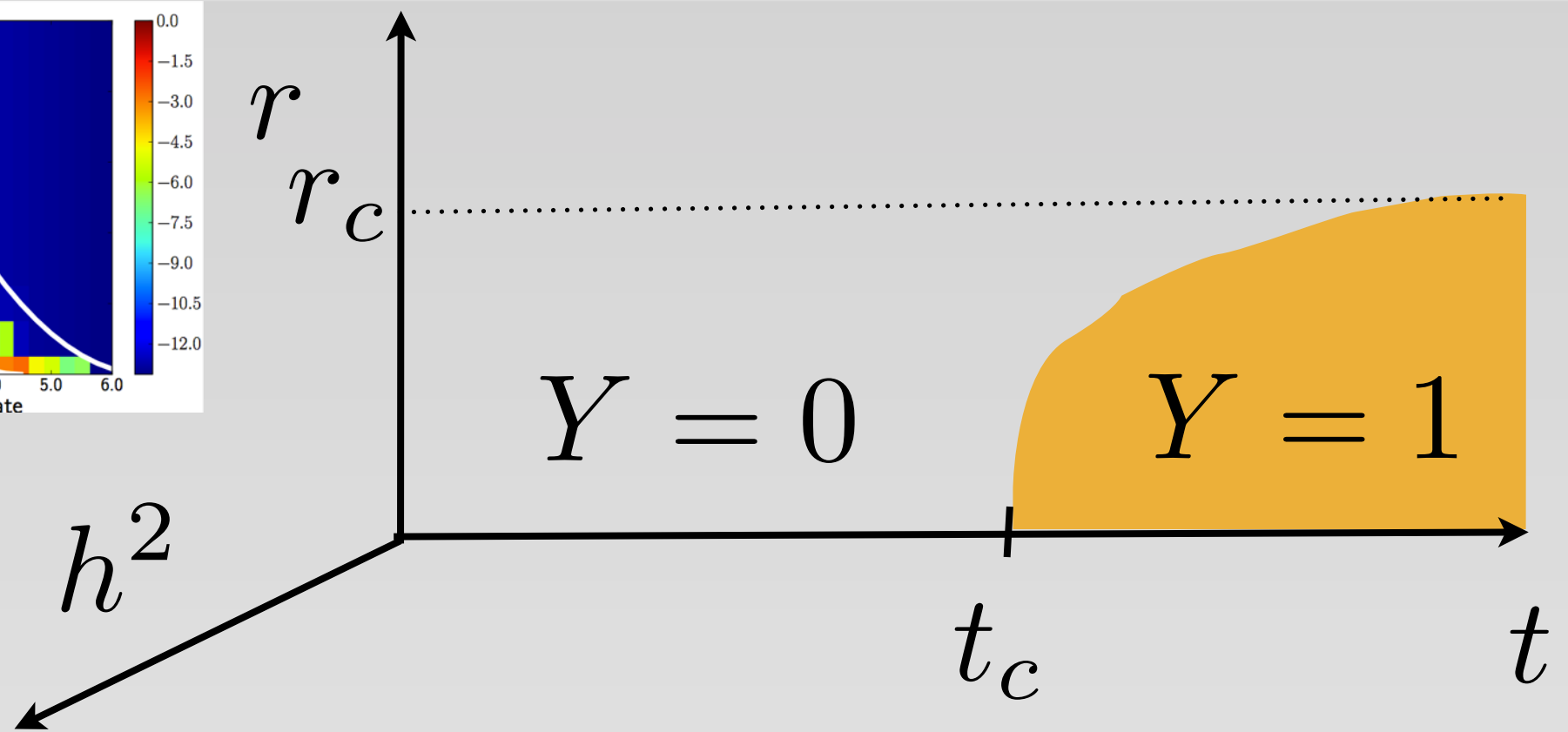
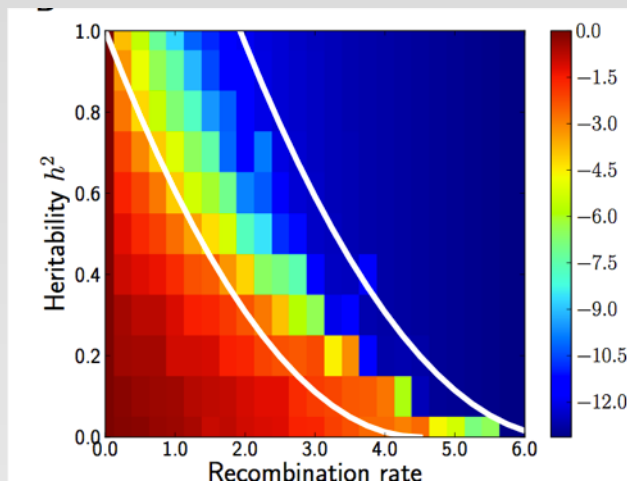
$$\frac{t_c(r)}{t_c} = \frac{1}{1 - r/r_c}$$

$$r_c \equiv \sigma \sqrt{2 \ln(N)}$$



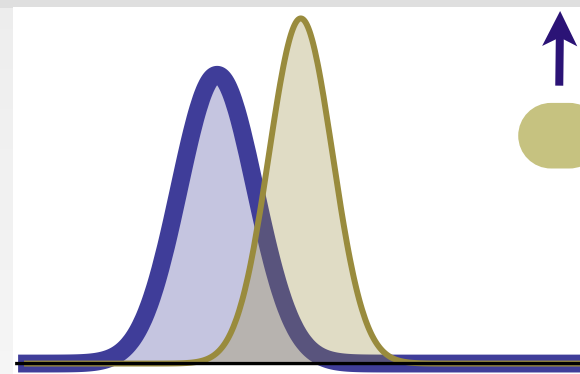
Similar behavior observed in Barton 1983, Franklin and Levontin, 1970

Heritability - how does the fitness of recombinants relate to that of the parents.

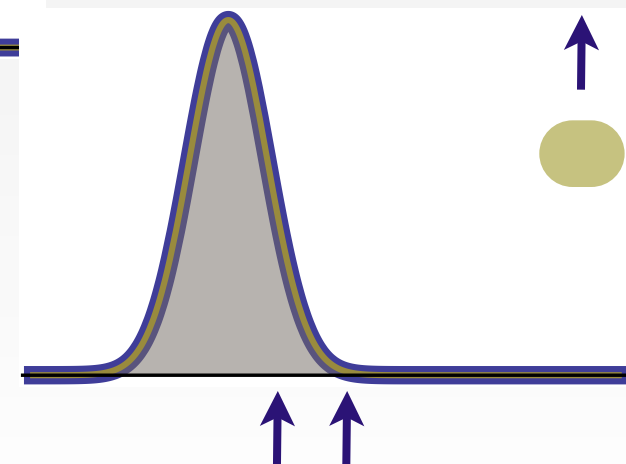


highly heritable trait

trait of low heritability



parents
offspring



parents
offspring

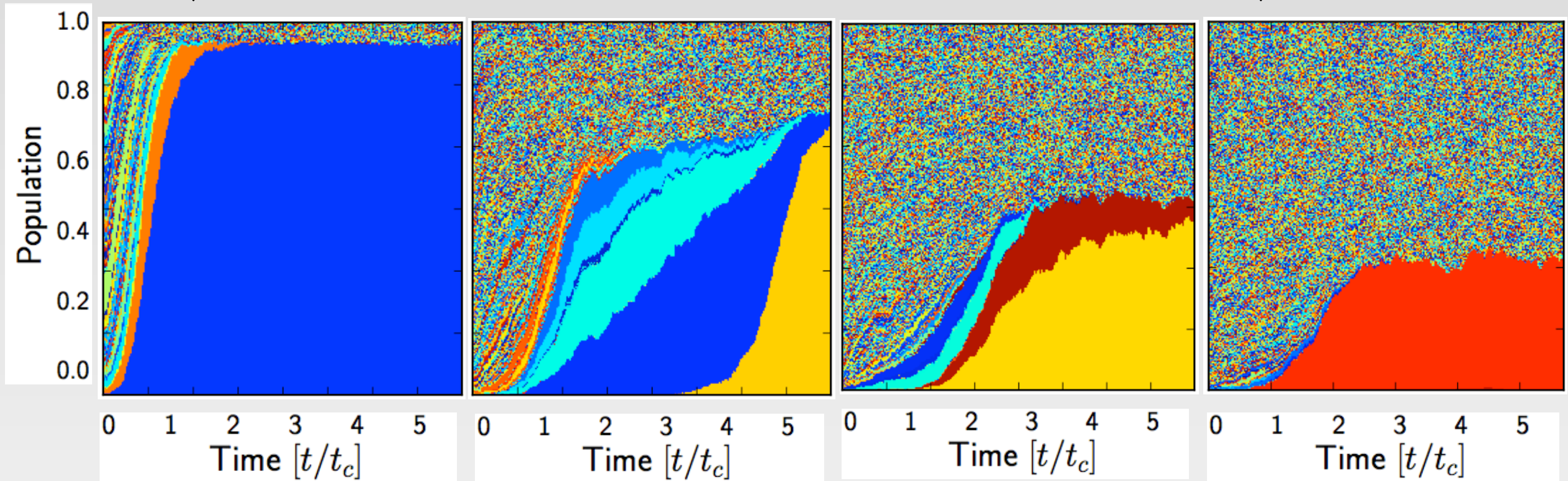
without heritability

$$r/\sigma = 0.2$$

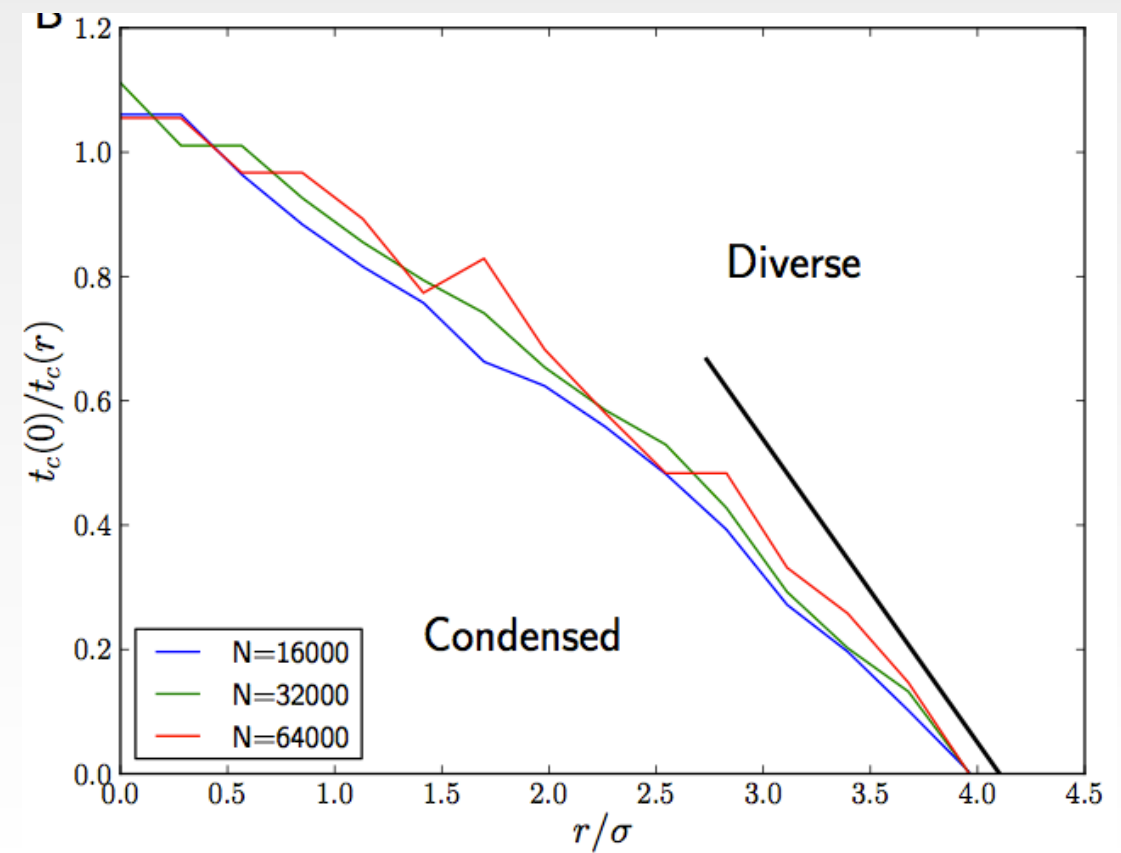
$$r/\sigma = 1.0$$

$$r/\sigma = 1.8$$

$$r/\sigma = 2.4$$



clone = color
same colors grouped
increasing recombination rate delays
the transition to condensed state



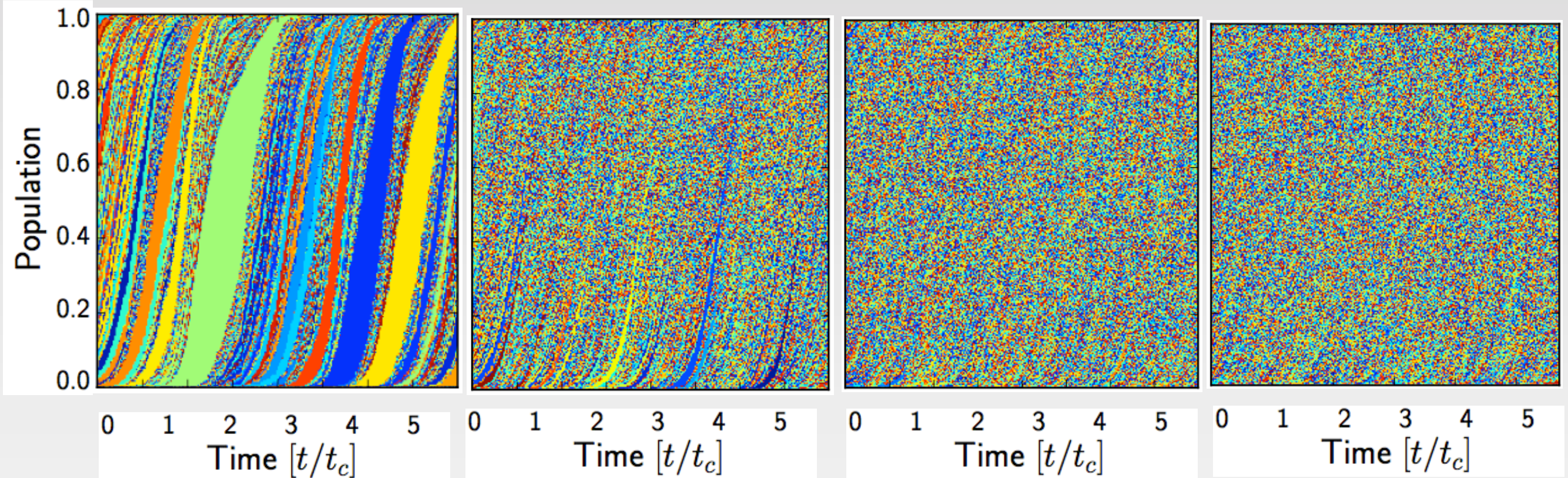
high heritability

$$r/\sigma = 0.2$$

$$r/\sigma = 1.0$$

$$r/\sigma = 1.8$$

$$r/\sigma = 2.4$$



Large clones cease to exist.
Most population is made out of short lived genotypes.

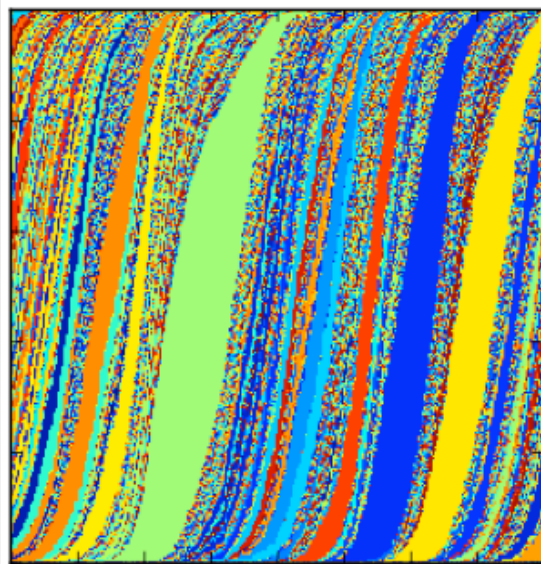
$$t_c = \sigma^{-1} \sqrt{2 \ln(N)}$$

Traveling solutions for additive fitness $h^2 = 1$

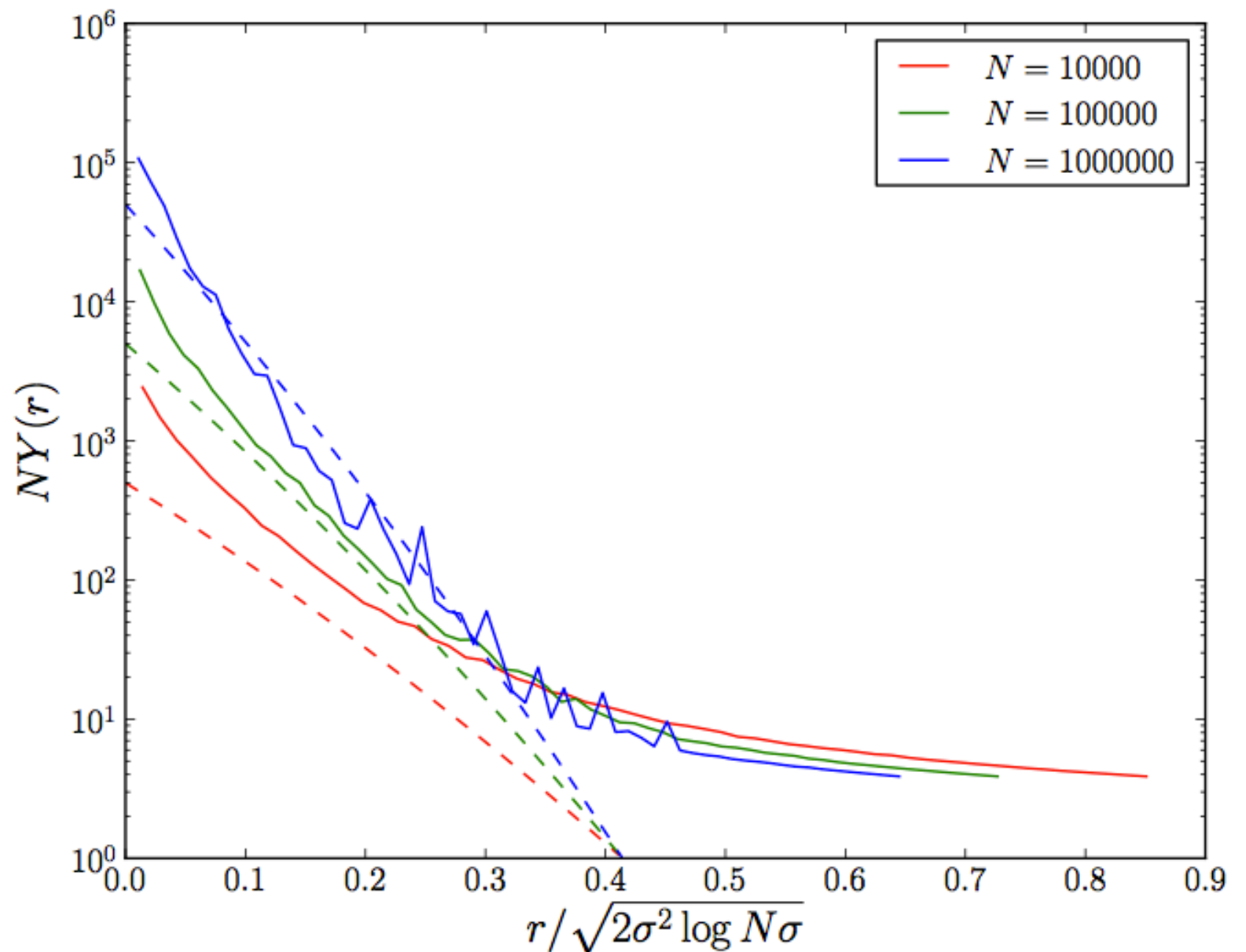
No aging = old genotypes are replaced; dominant genotypes have a finite characteristic age τ_j

$$n_j = e^{(A_j - r)\tau_j - \frac{\tau_j^2 \sigma^2}{2}}$$

$$\langle Y(r) \rangle \approx 2N^{-1} + re^{-r\sigma^{-1}\sqrt{2\ln N} - \frac{r^2}{2\sigma^2}}$$



(Cohen et al., 2005a;
Desai and Fisher, 2007;
Hallatschek, 2011; Neher et al., 2010;
Rouzine et al., 2003;
Tsimring et al., 1996)



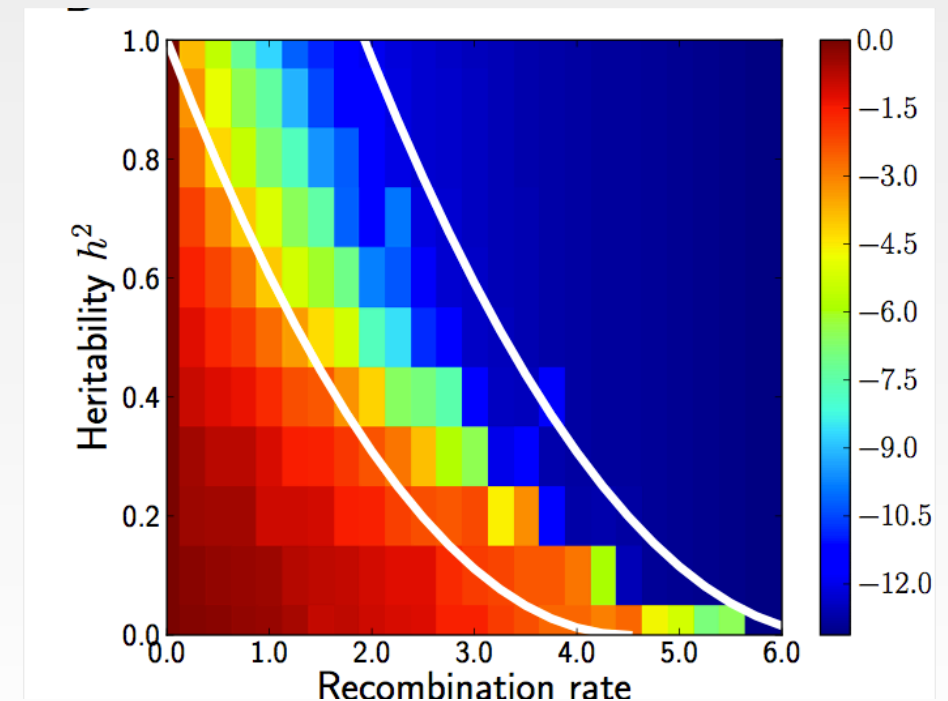
Future directions: clonal condensation

Connection to real world populations: heritability (between 0 and 1), large number of loci.

- more complicated models would reveal more structure than this simple “dust/clone” dichotomy

- better understanding of a “mixed” phase $0 < h^2 < 1$
- adding mutations
- REM \longrightarrow Sherrington - Kirkpatrick
- dynamics population getting homogeneous and diverse

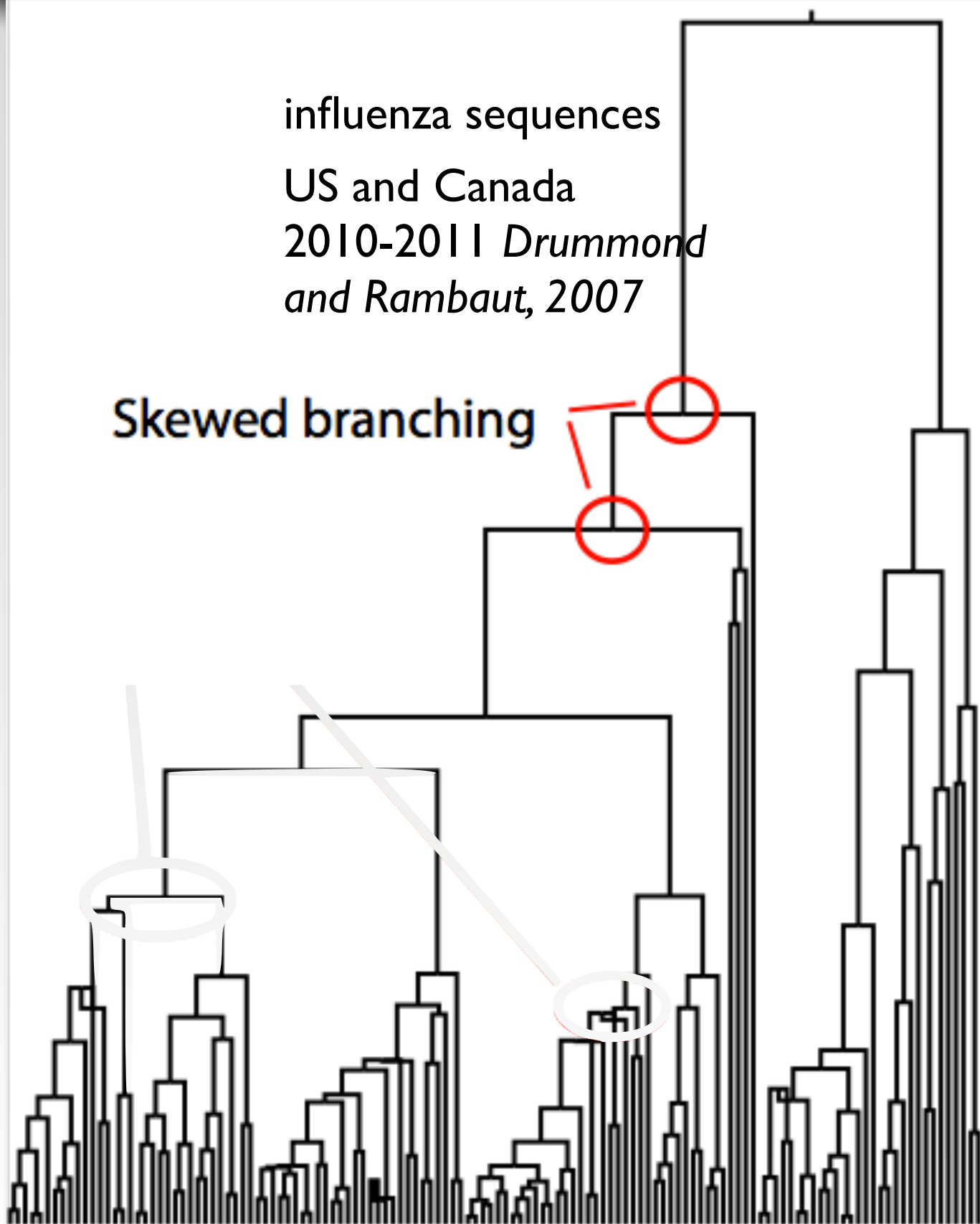
$$\log \langle Y_{\infty}(r, h) \rangle$$



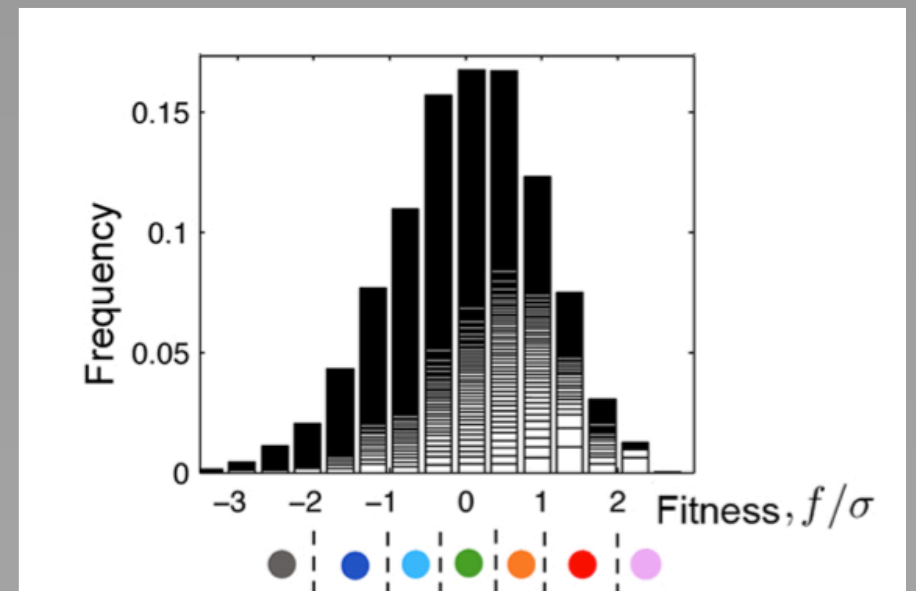
Inference of fitness of the leaves from genealogical trees

influenza sequences
US and Canada
2010-2011 *Drummond
and Rambaut, 2007*

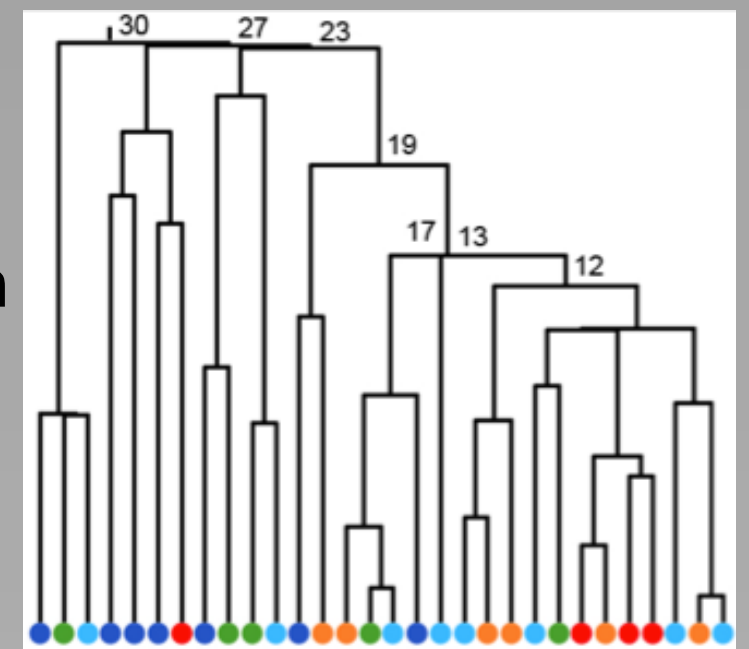
Skewed branching



Dayarian, Shraiman 2012



heuristic
approach



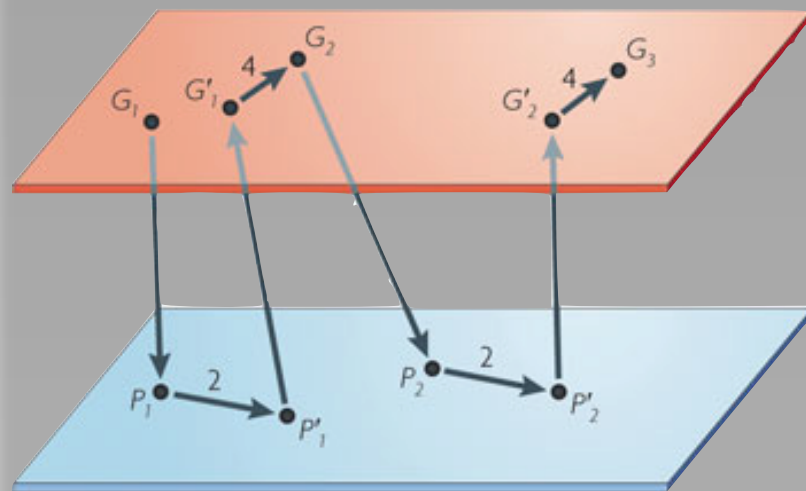
Phenotype internal set of states of individual organisms

set of organisms' traits

example: stripes, color, biochemical or physiological properties, behavior...



phenotype-genotype map



+ environment

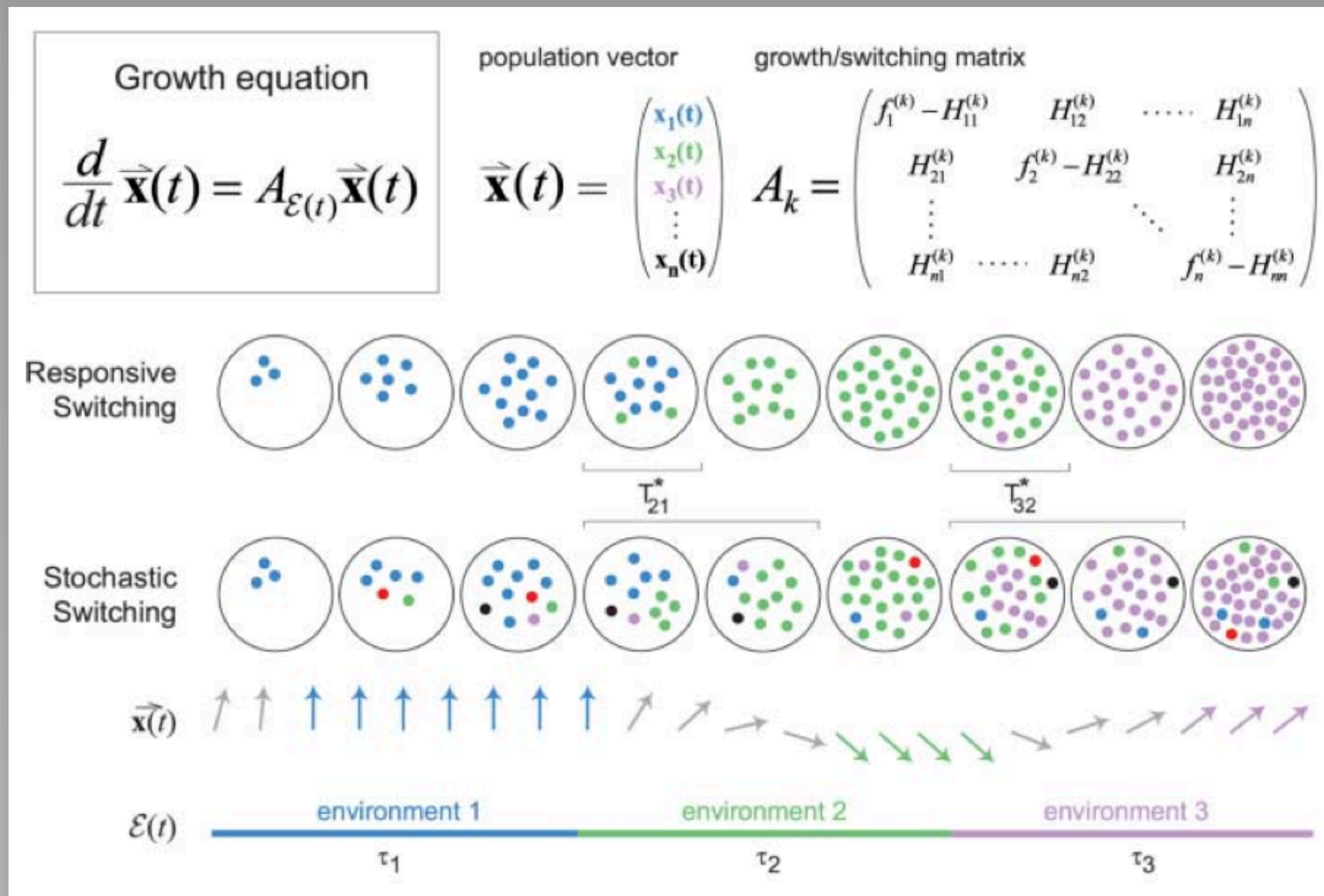
phenotypic switching

environmental fluctuations = external forcing

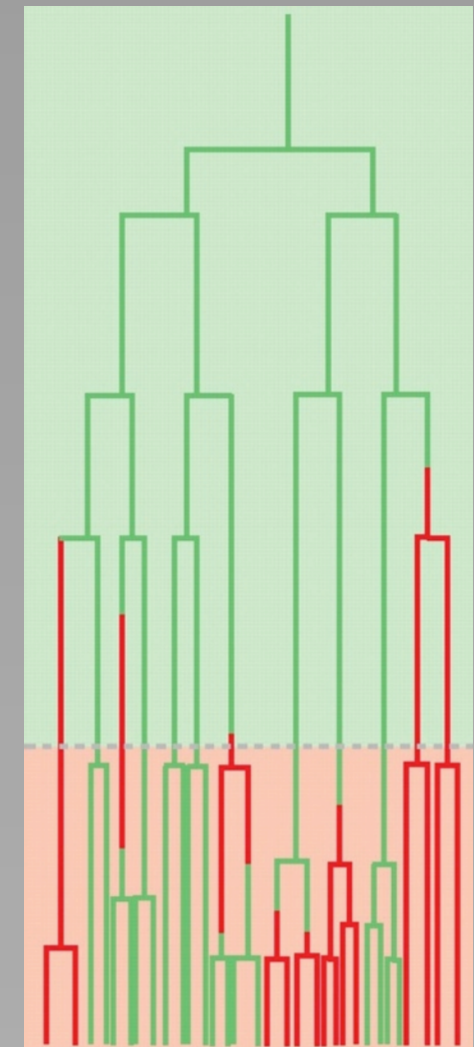
temporal fluctuations drive the system away from equilibrium

equilibrium perspective - individual histories (trajectories in the phenotypic space observed in the population)

Phenotypic switching



Responsive and
stochastic switching
Kussell, Leibler 2005

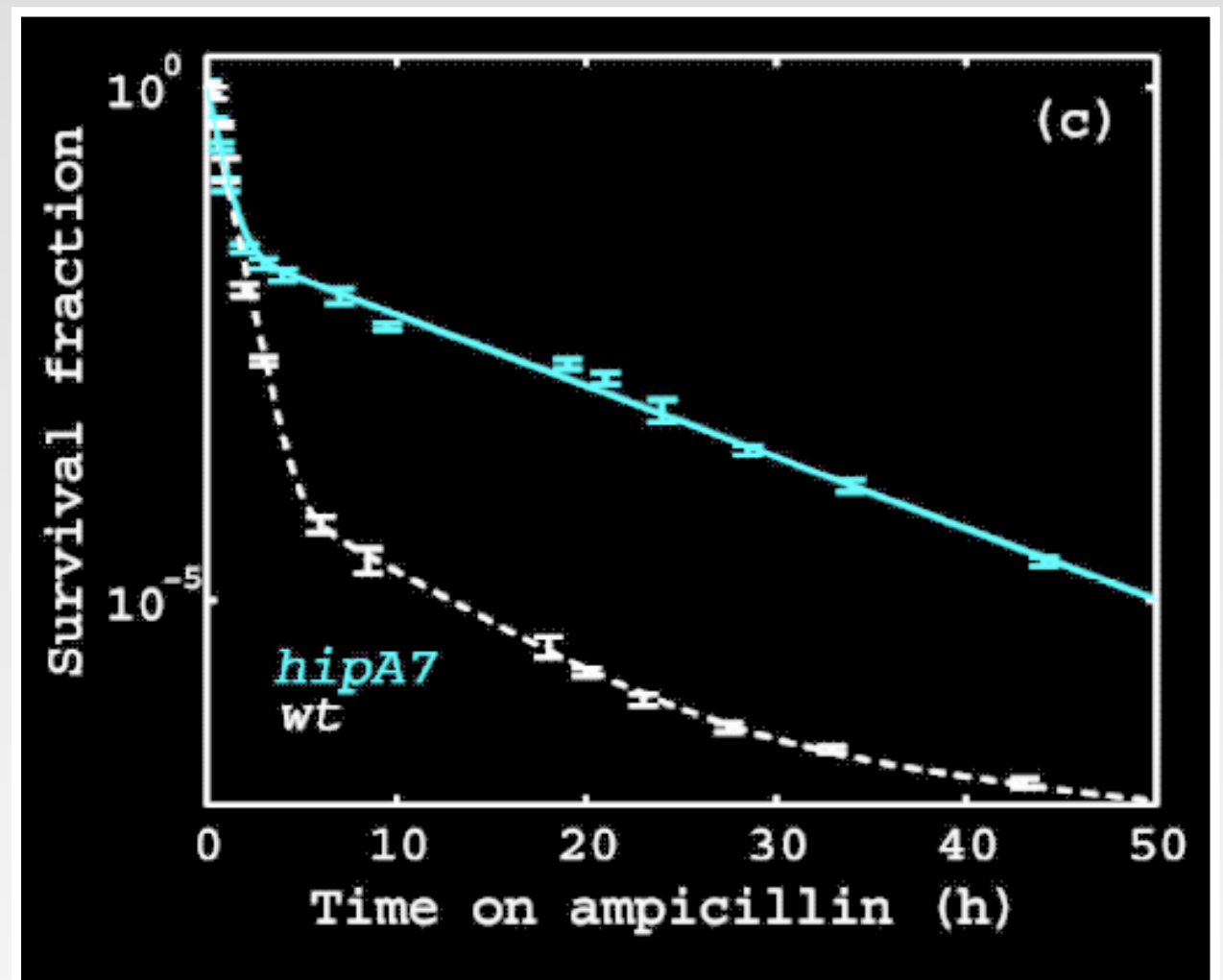
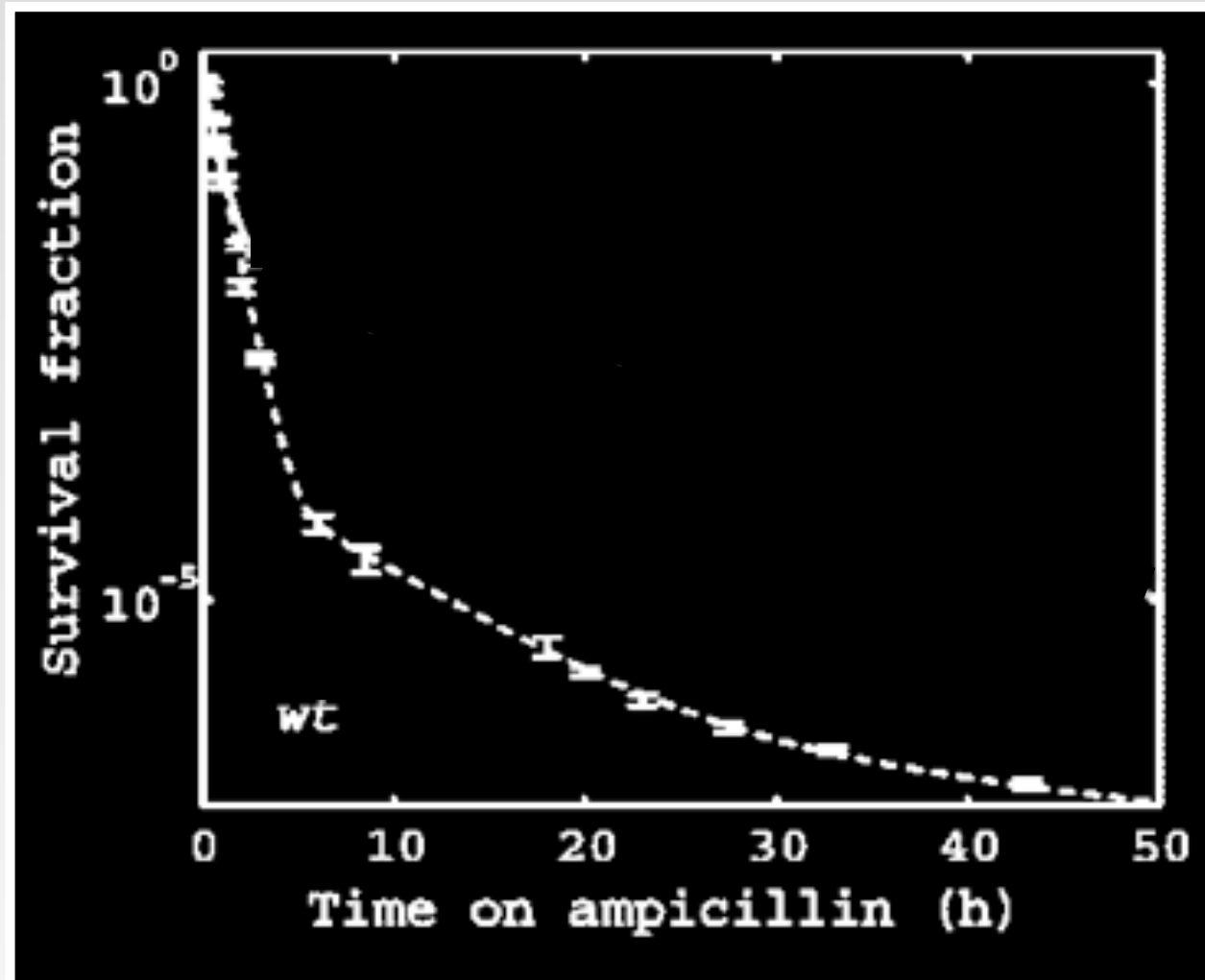


Kussell, Leibler, 2010

Persistors

different behavior, dividing more slowly

antibiotic persistence



Balaban et al, 2004

hipA7 - high persistence mutants

Mutators individuals with a much higher mutation rate in a population (up to 10^5 higher)

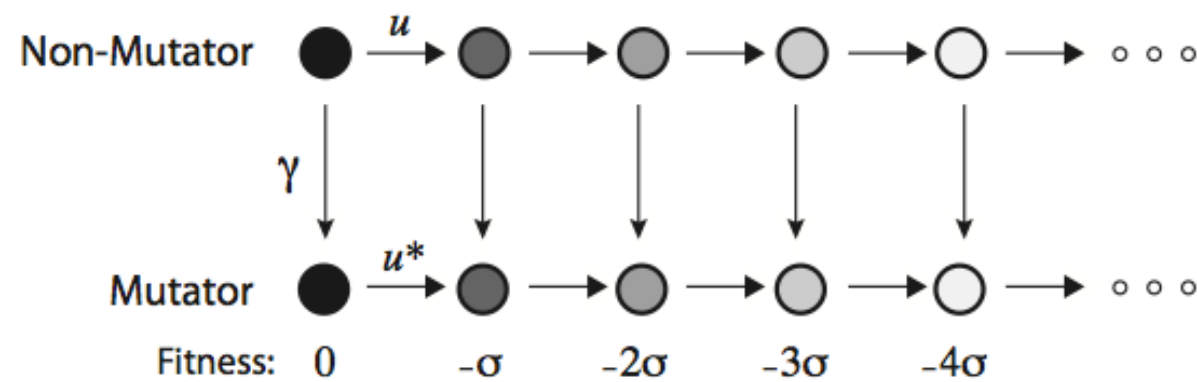
Desai, Fisher 2010

Travis, Travis 2002

Lenski experiments - evolving *E. coli* since 1988

constant environment

Mutator Dynamics Model

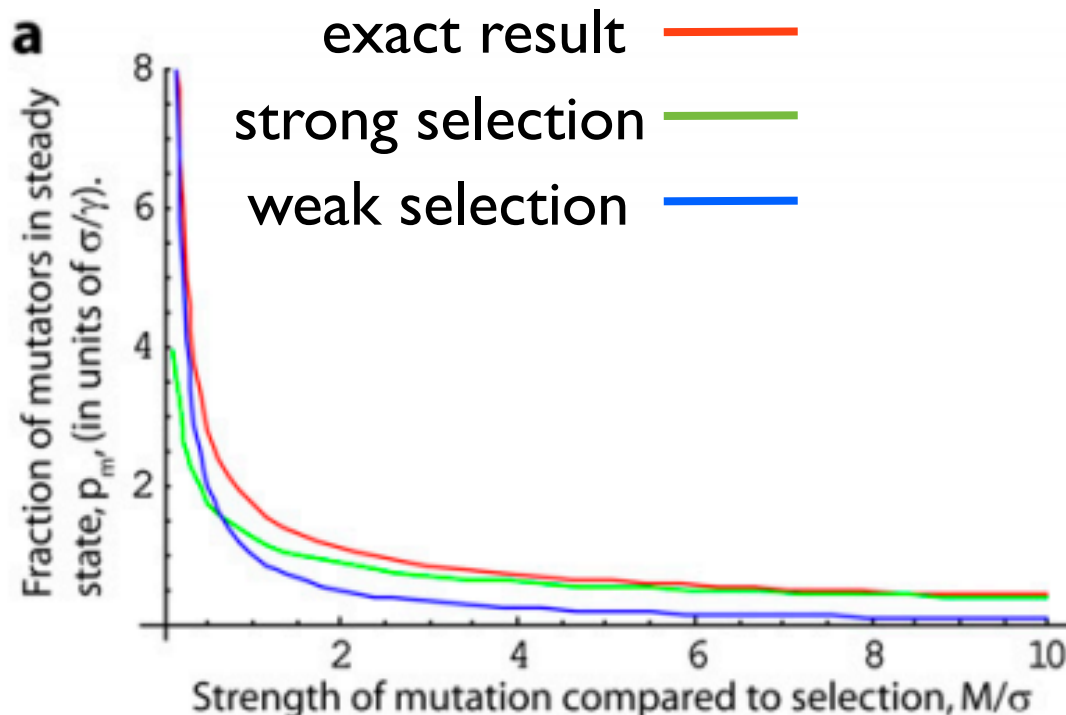


Desai, Fisher 2010: Even in situations where selection on average acts against mutators, so they cannot stably invade, the mutators can still occasionally generate beneficial mutations and hence be important to the evolution of the population.



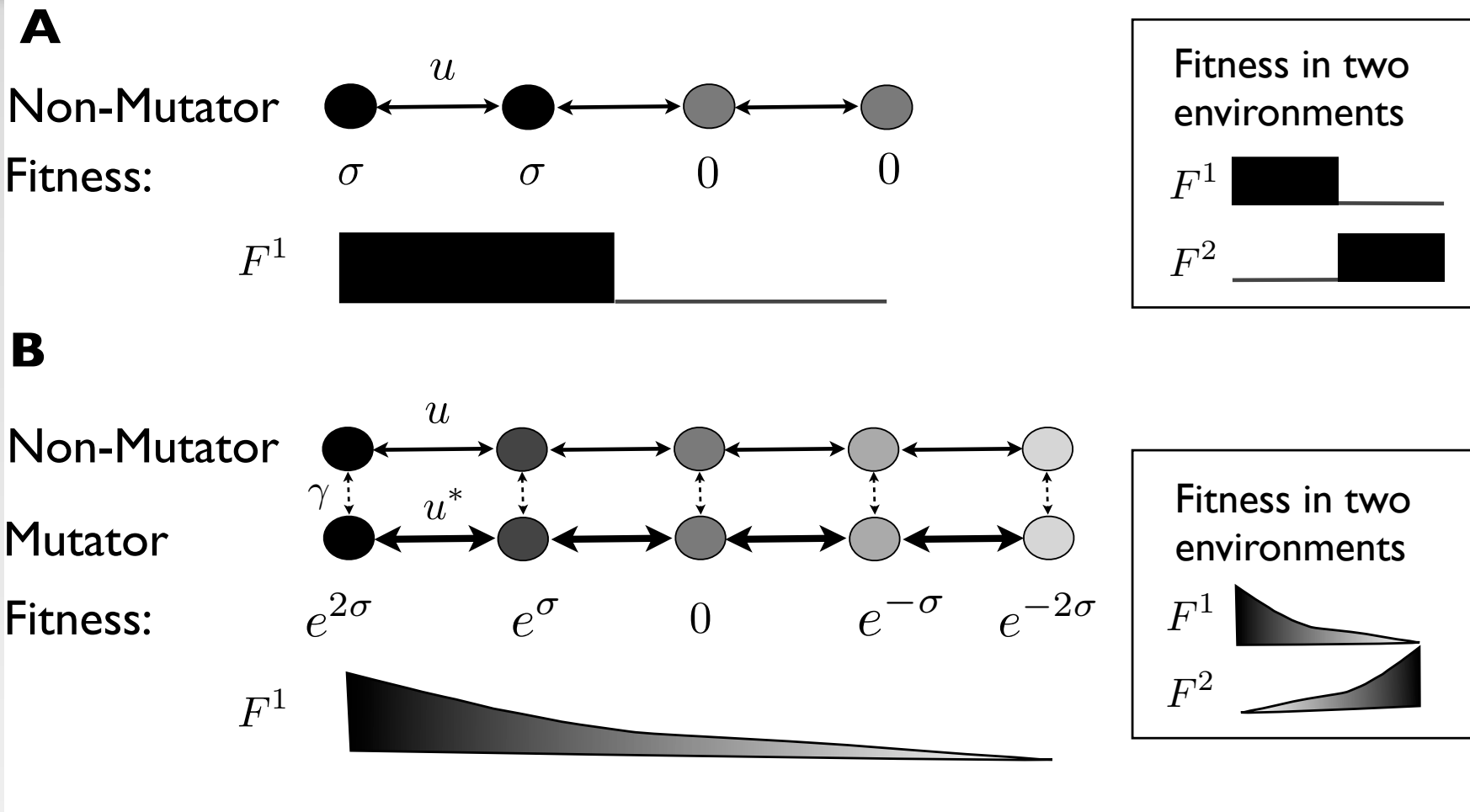
Desai, Fisher 2010: findings confirmed by R. Lenski experiments:

$$p_m \propto \gamma/M$$



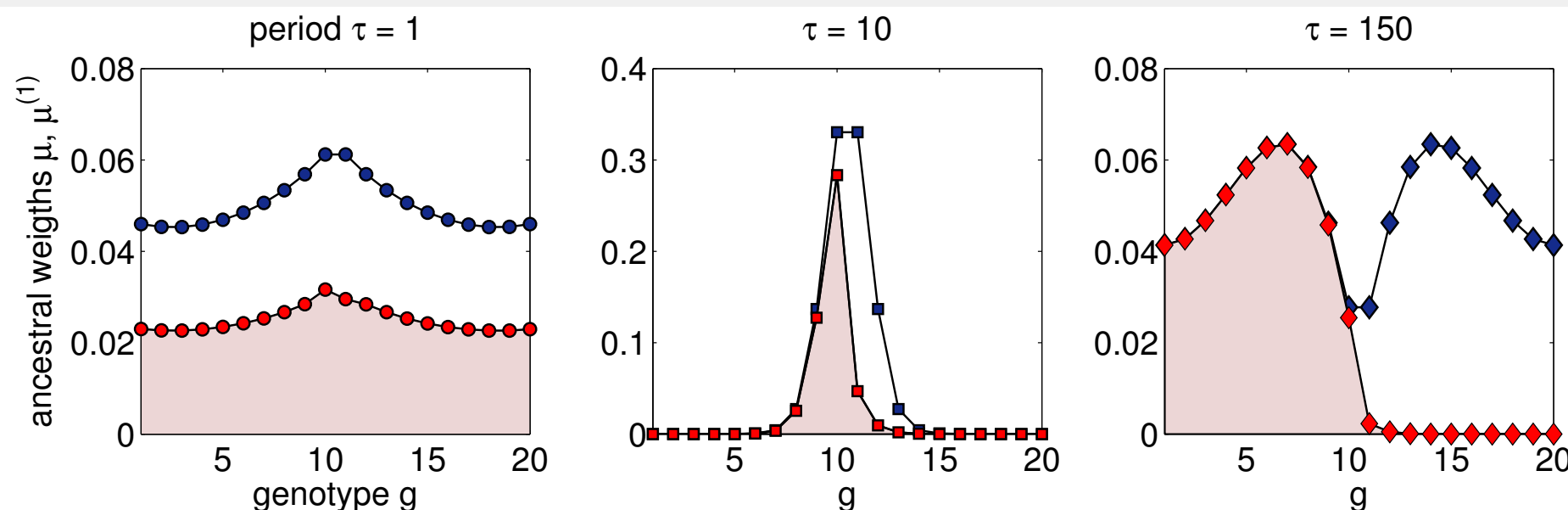
Mutators in fluctuating environments

Kussell, Leibler 2005; Kussell, Leibler, Grosberg, 2006; Travis Travis 2002



Ancestral distributions of non-mutators in the periodic case of two environments

Analogous problem: heteropolymer localization on the interface



localization for:

$$\frac{1}{f} < \tau < \frac{L^2}{4u}$$

- **missing:** better theory and quantitative aspects of population genetics; relevant timescales?
- **present:** abundance of data.
- **difficulties:** numerous timescales, quenched disorder, incomplete statistics (we only see a single realization of the outcome of the evolution)
- **relevance:** drug resistance, disease evolution, origin of life
- **Population genetics - on statistical mechanics language** - relations to polymers, path integrals, localization phenomena naturally emerge, non-trivial “thermodynamic limit”

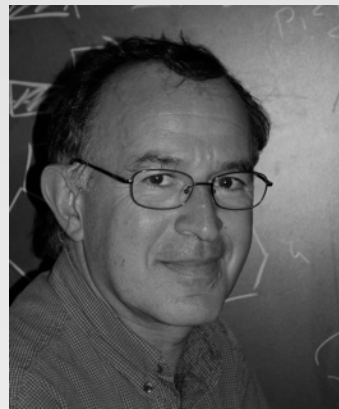
Collaborators



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