Statistically characterizing antibody diversity

work with Thierry Mora, William Bialek, Curt Callan

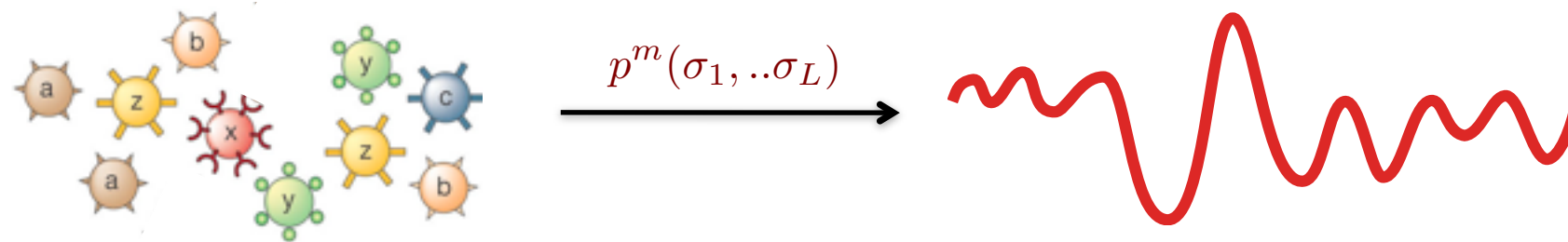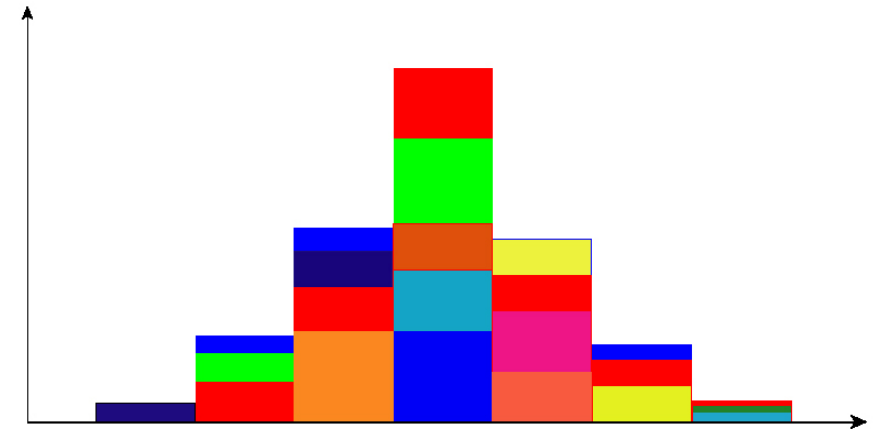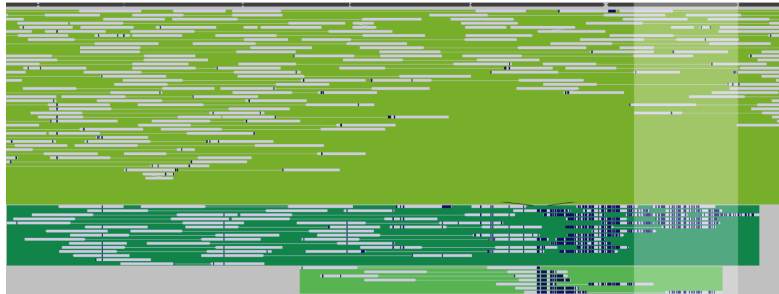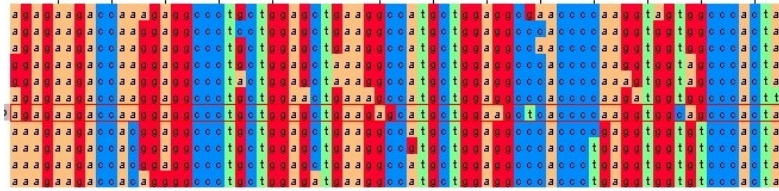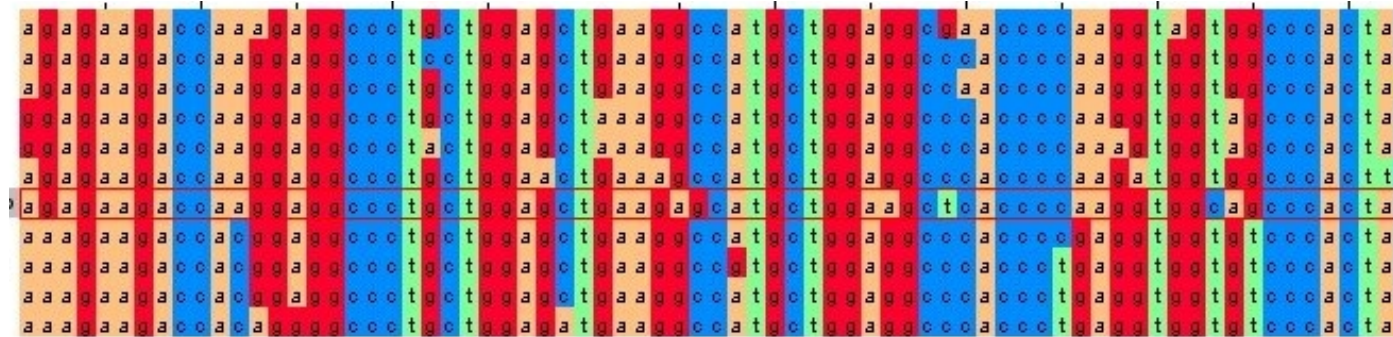# The effects of negative selection on the evolution of linked sites

Aleksandra M Walczak
Laboratoire de Physique Théorique - ENS, CNRS

Michael Desai, Harvard
Joshua Plotkin, University of Pennsylvania
Lauren Nicolaisen, Harvard

Inferring evolutionary processes

Understanding evolutionary processes:

- test consistency of data with null models
- currently: easy to use neutral or weak selection models
- disagreement: selection, demography, geography ...

Goal: develop null models with selection

- test consistency of data with null models with selection
- rule out models also when neutrality does not apply
- infer selective parameters from data

Evolutionary scenarios

Genetic Drift
Well understood
But what do deviations mean?
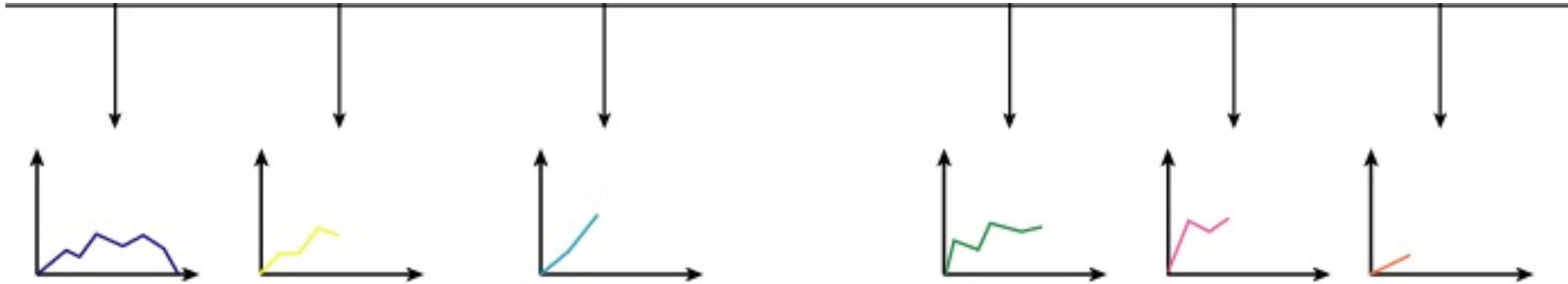
Natural Selection
Reduces diversity

Demography
Bottlenecks, expansions reduce diversity

Geography
Environmental structure increases diversity

What should we look at? What do we expect?

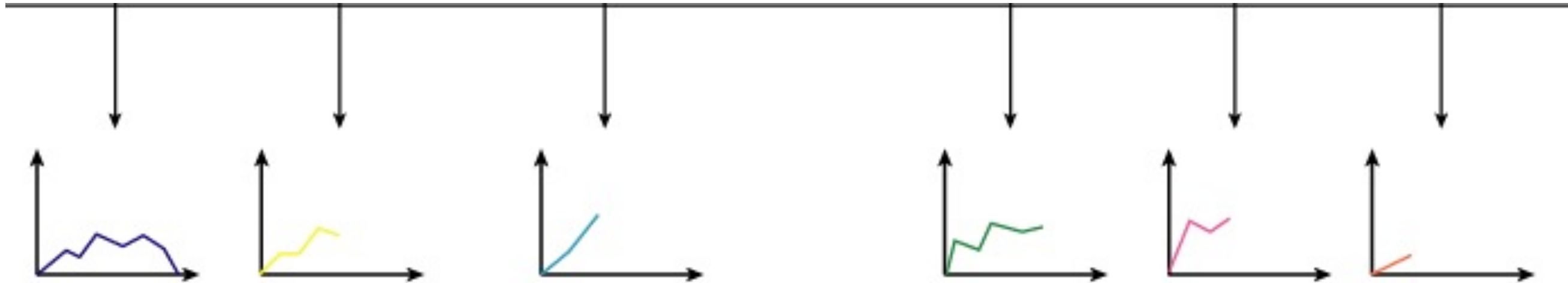# Model the fate of each site in the genome

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Calculate the fate of each mutant forward in time.

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Calculate the fate of each mutant forward in time.

Fate of each mutation is not in steady state

But there is a steady state distribution of the distribution of mutant frequencies

Nearby mutations are not independent

```
                Base Position

                11111112222222333344
                14567023345911124491346l2
                59479030537323661878008OO
Haplotypes      74421520693615O6795967889

Chimpanzee      CCGGTTATGCCGAGAATACGGCGCC
    A           --ACCC--TGT--AC-CC-----T-
    B           --ACCC--TGT--AC-C------T-
    B1          --ACCC--TGT--AC-C---A--T-
    C           ---CCC--TGT--AC-C------T-
    D           -A-----C--*-T-----T--T---
    E           TA-----C----------T--T---
    F           -A----CC----------TA-----
    G           -A-----C-------G--T---C-T
    H           -A----CC--*----G--T---C--
    I           -A-----C--*A------T-A-C--
    J           -A-----C--*-------T------
```

[Harris and Hey 1999]

Strong correlations between mutations.

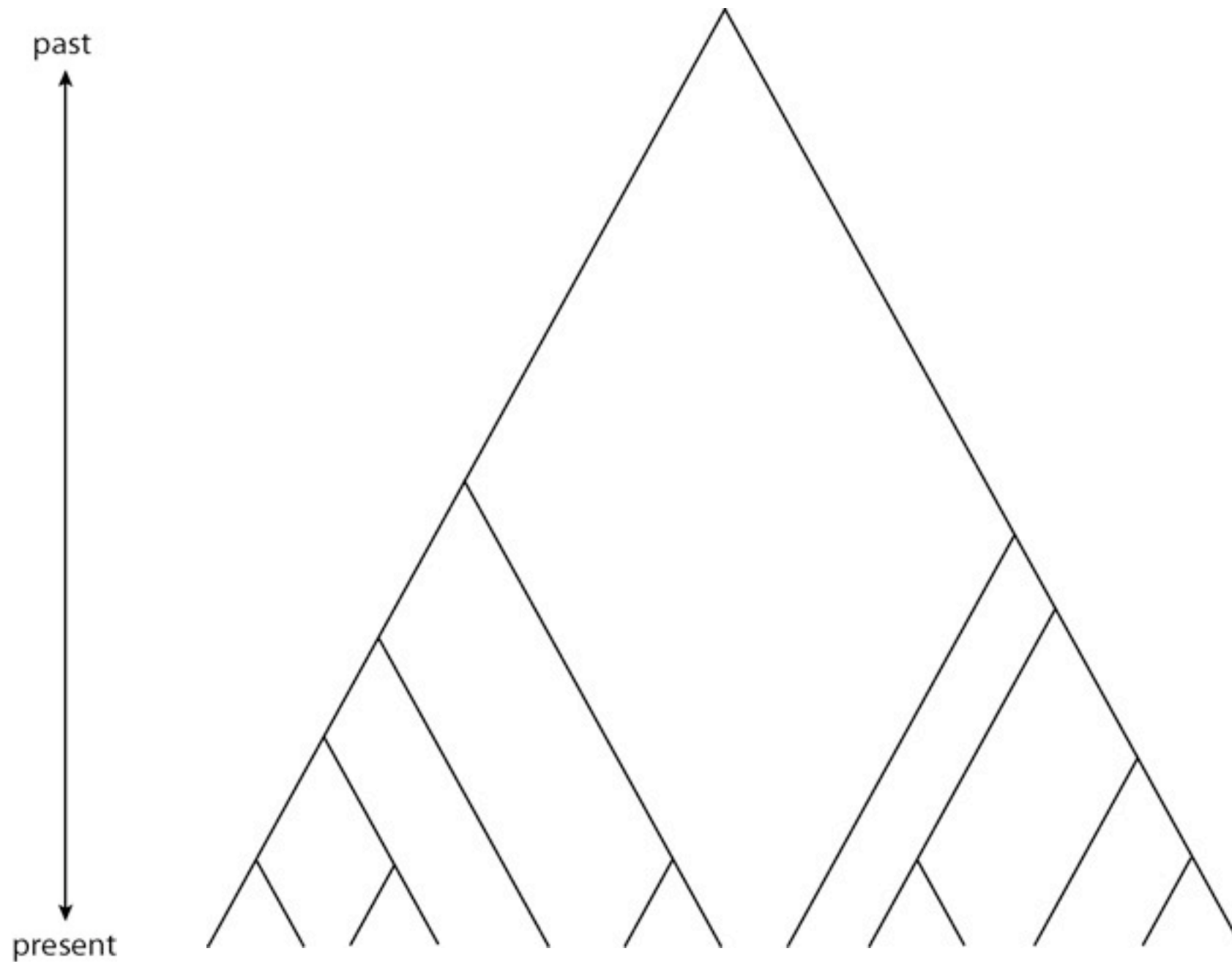Mutations are physically *linked*.

Recombination breaks linkage.

No recombination - fully linked sites
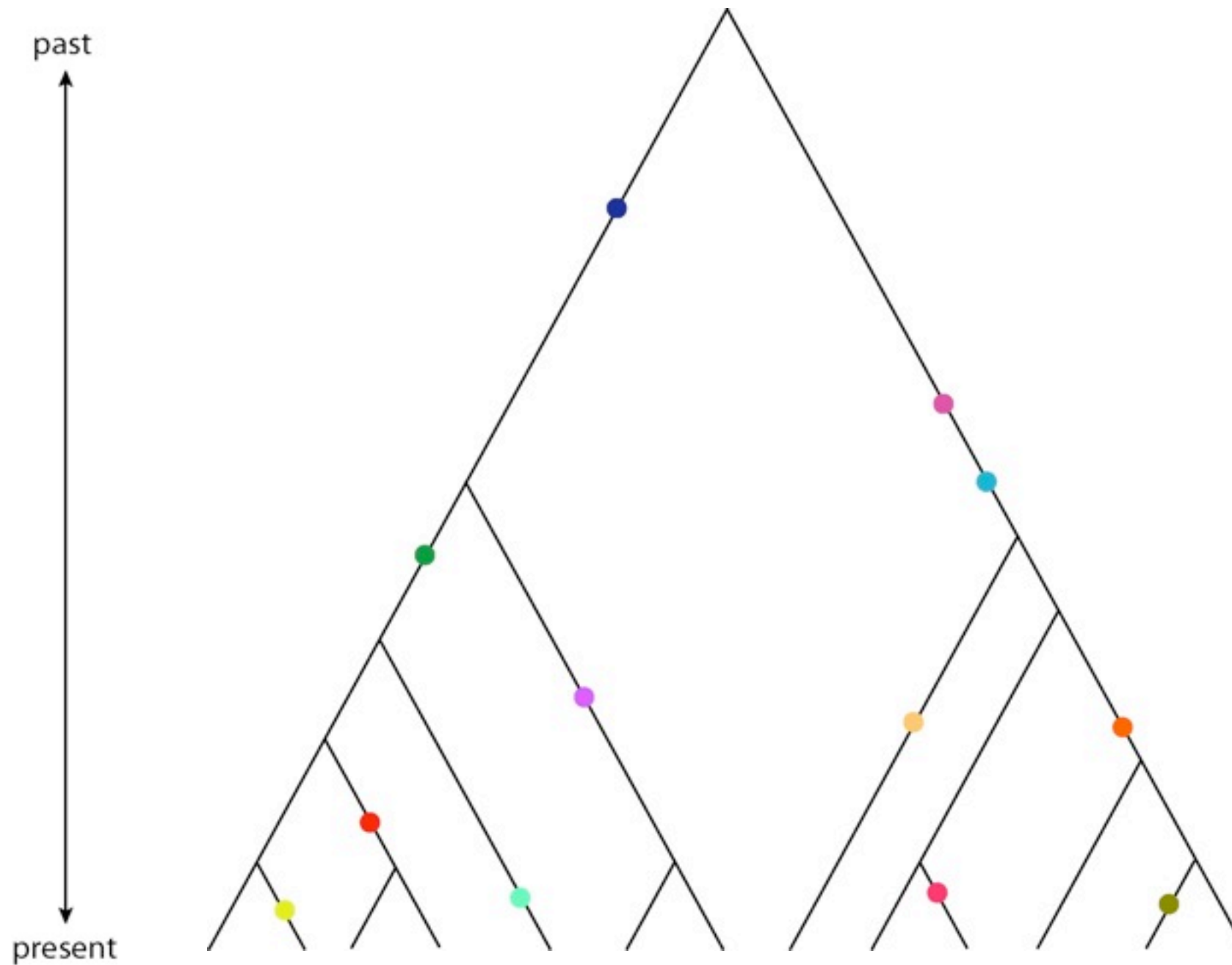
No selection: *Coalescent Theory*

The whole sequence shares a common genealogy.

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The whole sequence shares a common genealogy.

Cannot easily handle selection, despite 20 years of effort.

Comparison to the neutral null model

Is this data consistent with neutral well-mixed random-mating population?

What can we infer about the evolutionary history of this population?

Comparison to the neutral null model
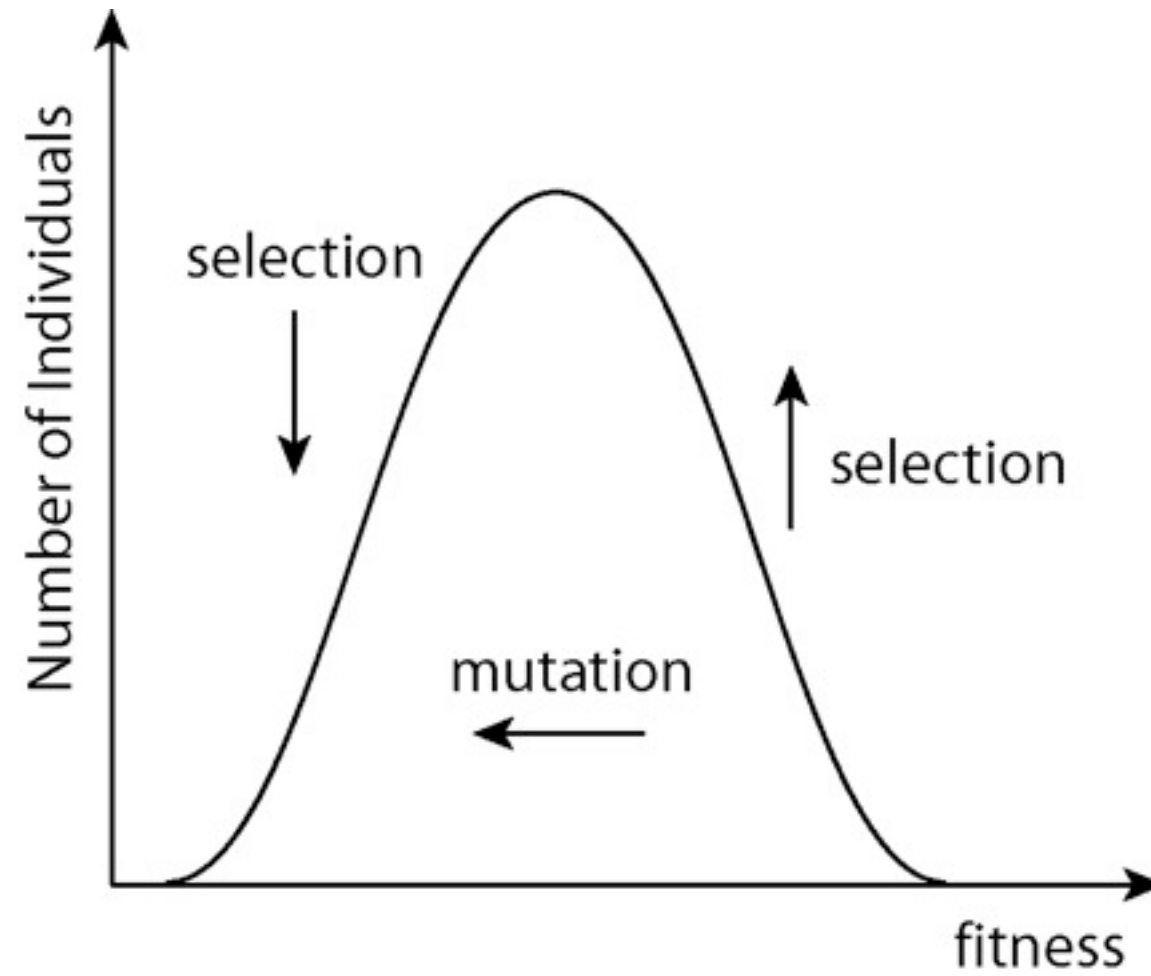

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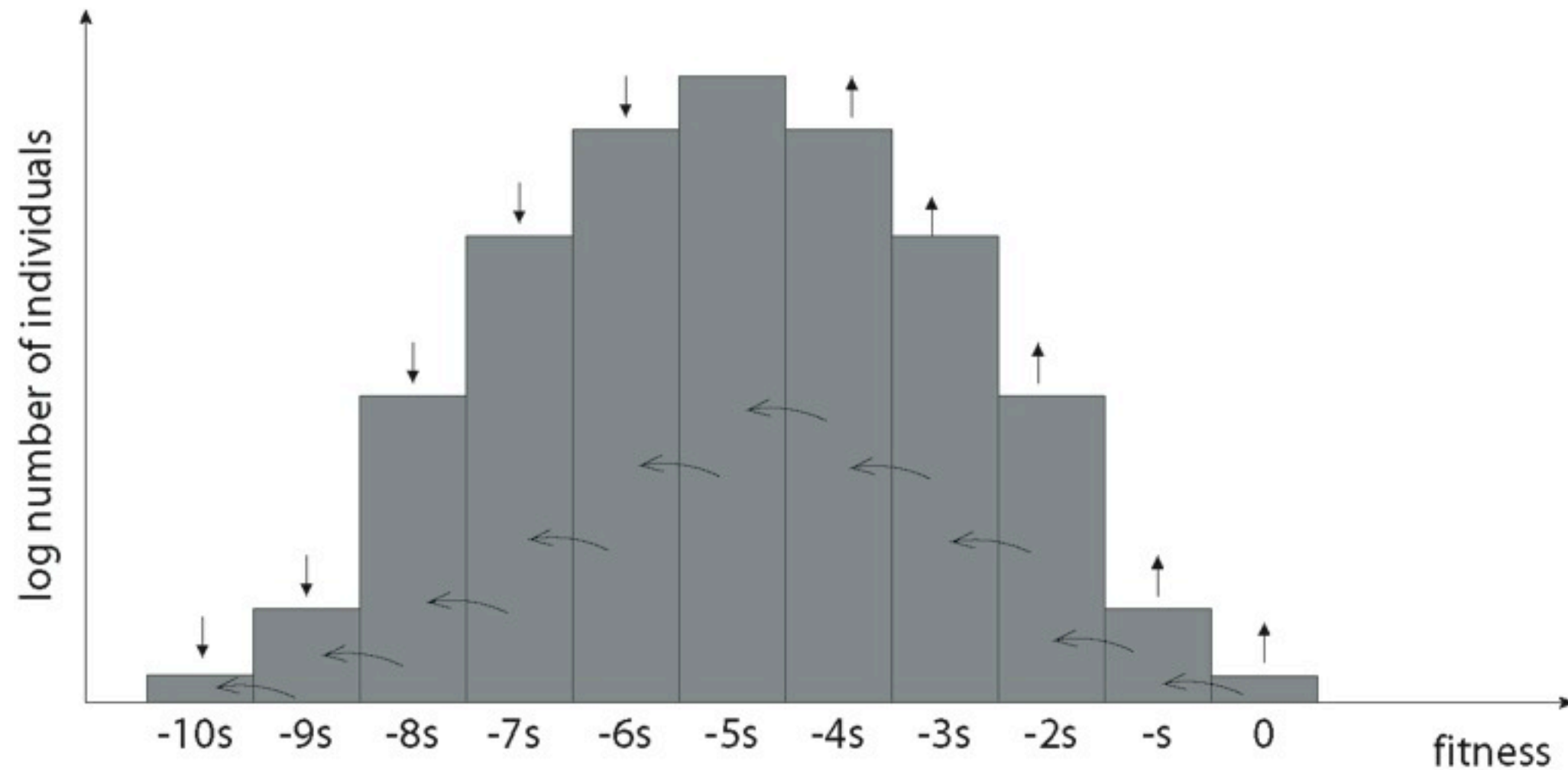

Are the coalescent trees that lead to some aspect of the observed diversity likely?

Evolution of the fitness distribution

Balance between mutations and selection in each class:
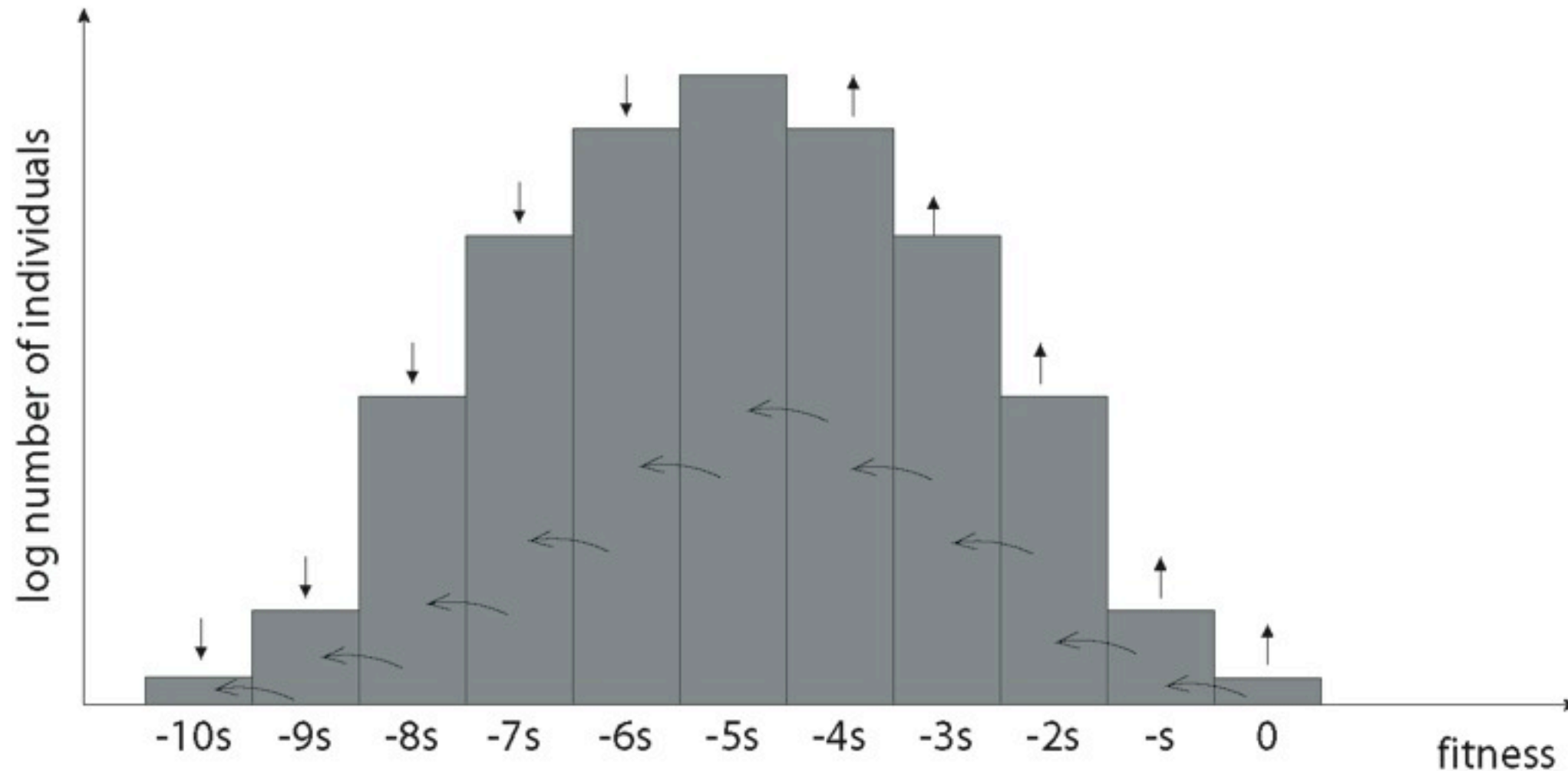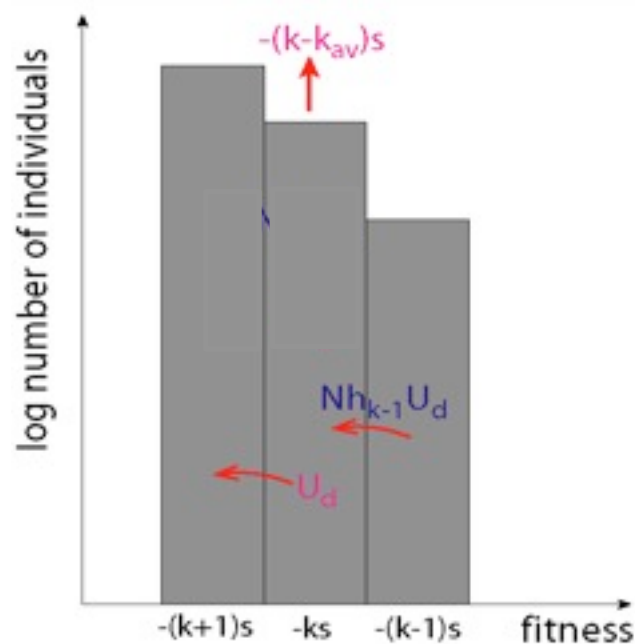Deterministic steady state fitness distribution.

Evolution of the fitness distribution

Balance between mutations and selection in each class:
Deterministic steady state fitness distribution.

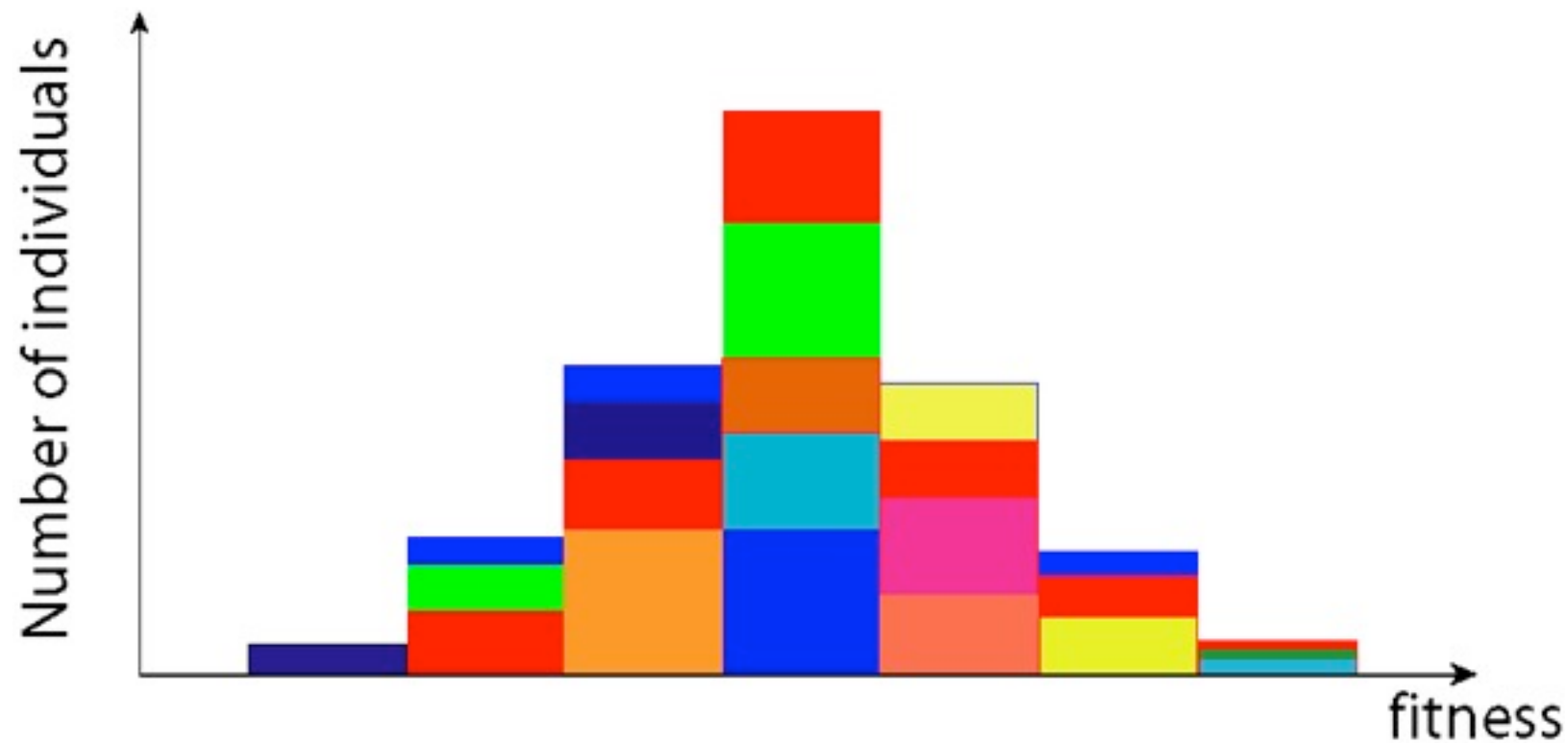$$\frac{dh_k(t)}{dt} = U_d h_{k-1} - U_d h_k - s(k - k_{av})h_k$$

In steady state:

$$\hat{h}_k = e^{-U_d/s}\frac{U_d^k}{k!s}$$

Many fluctuating lineages maintain the balance

- each fitness class is not genetically homogenous
- each class composed of many lineages
- different alleles with the same total fitness

Each class is maintained by flux in of new mutant alleles as old alleles drift and go extinct.

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- mutation decoupled from selection
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New alleles created at (mutation) rate:

$$\frac{\theta_k}{2} = N h_{k-1} U_d + N h_k U_n$$

per genome
per generation

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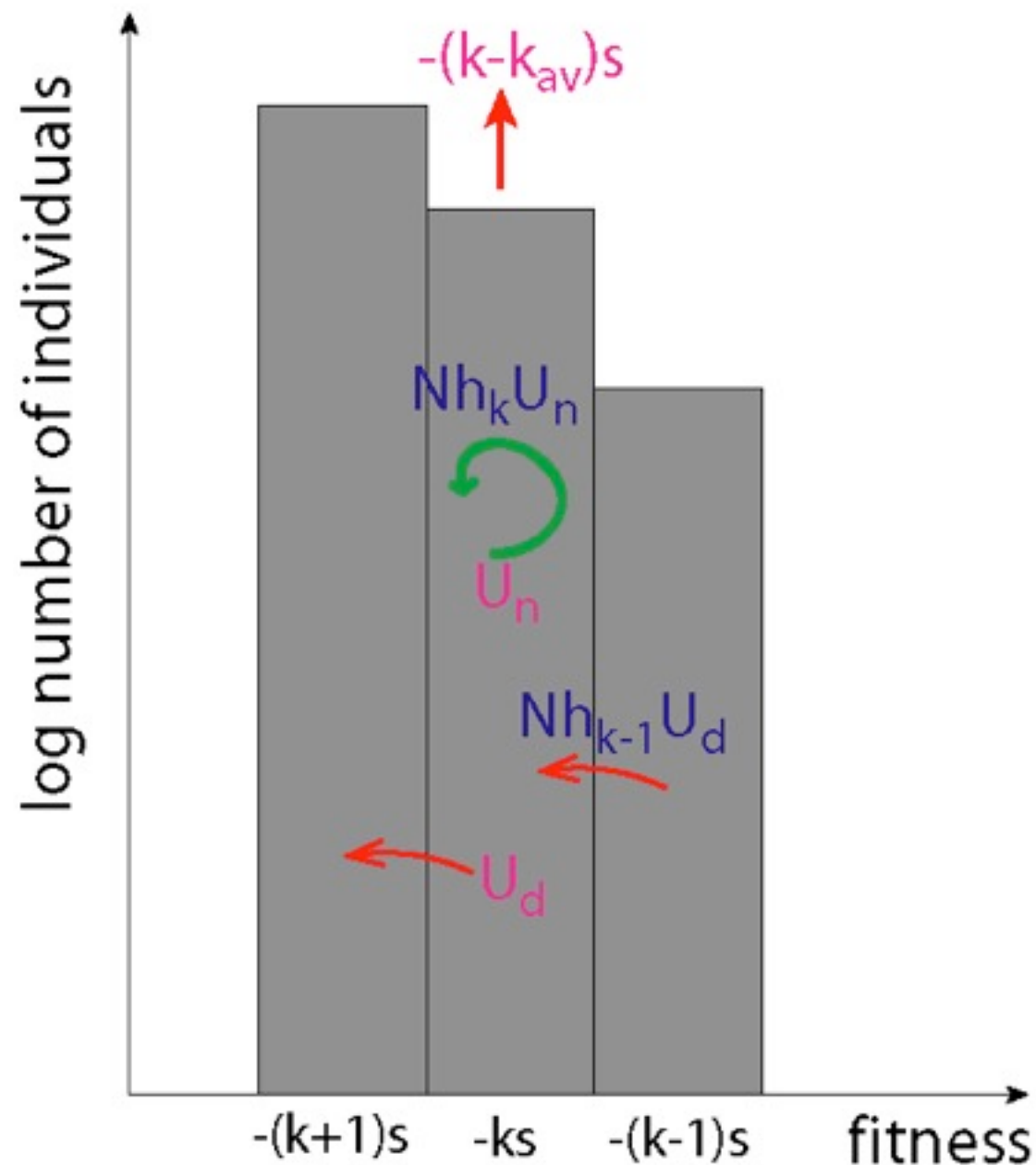Each class is maintained by flux in of new mutant alleles as old alleles drift and go extinct.

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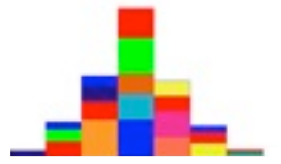

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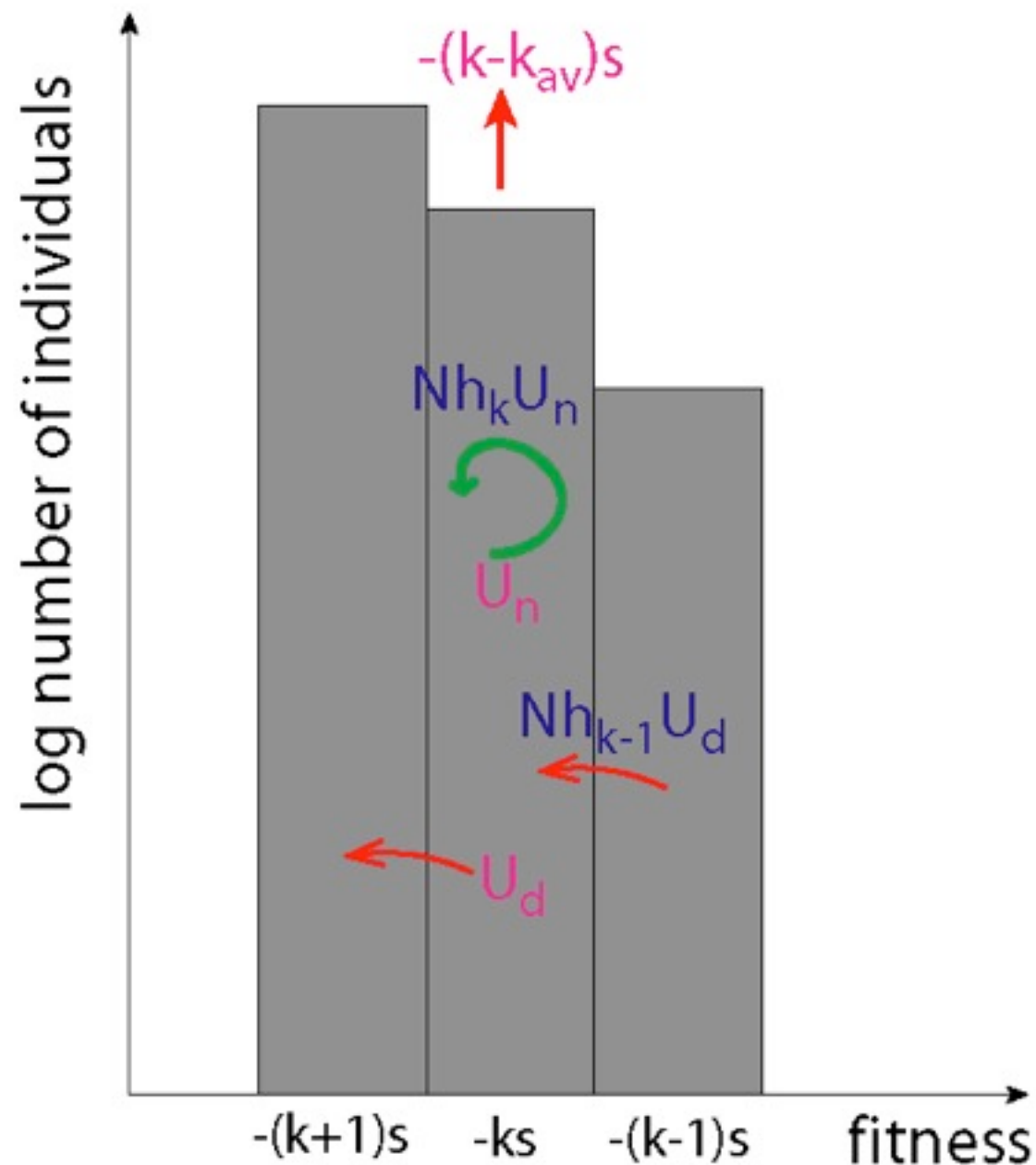New alleles created at (mutation) rate:

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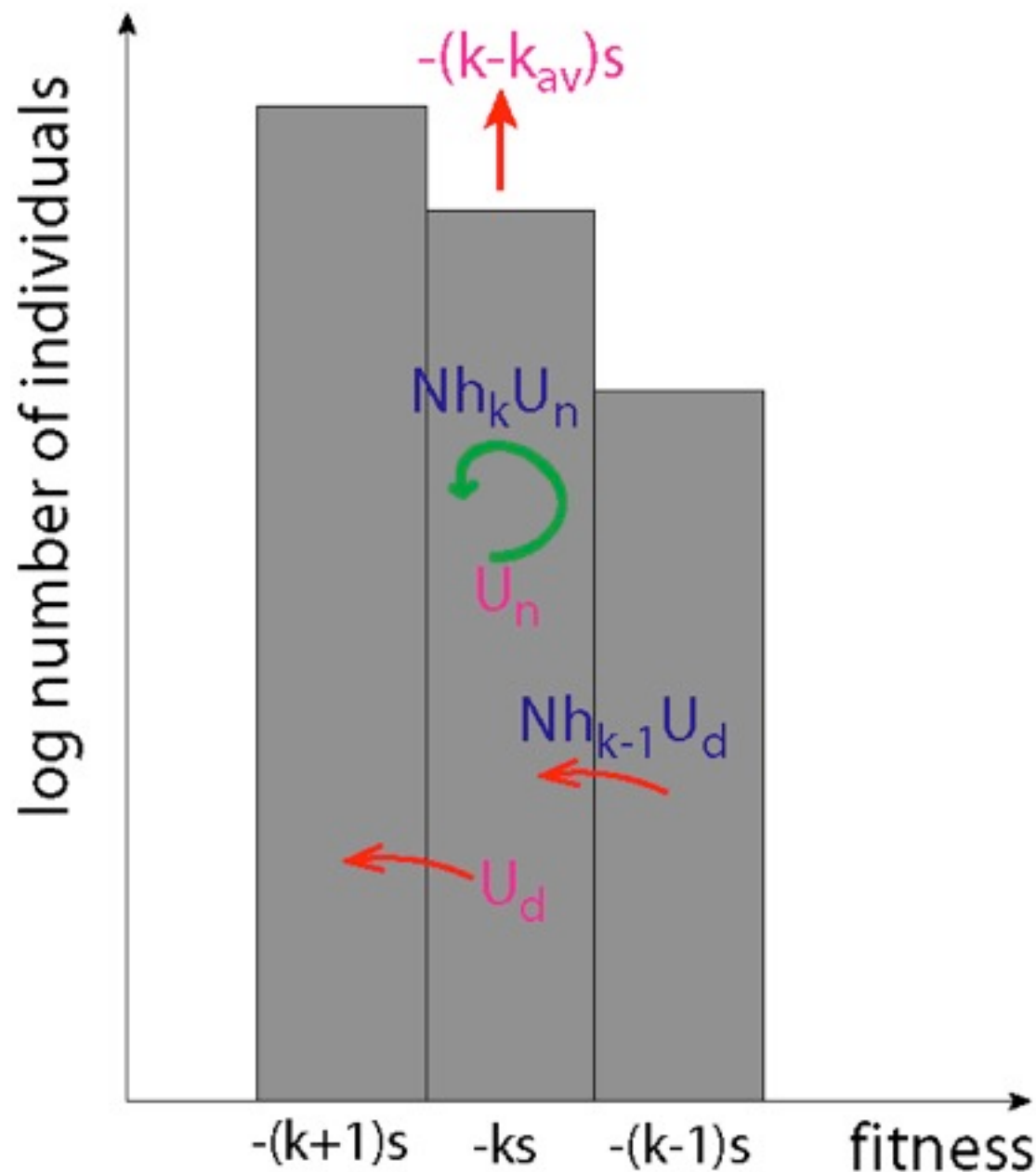
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-(k-k_{av})s

Nh_kU_n

U_n

$Nh_{k-1}U_d$

U_d

log number of individuals

-(k+1)s   -ks   -(k-1)s   fitness

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- older individuals must die out to conserve steady state fitness distribution
- k=0 class drifts neutrally - fitness advantage balanced by loss of individuals to less fit classes

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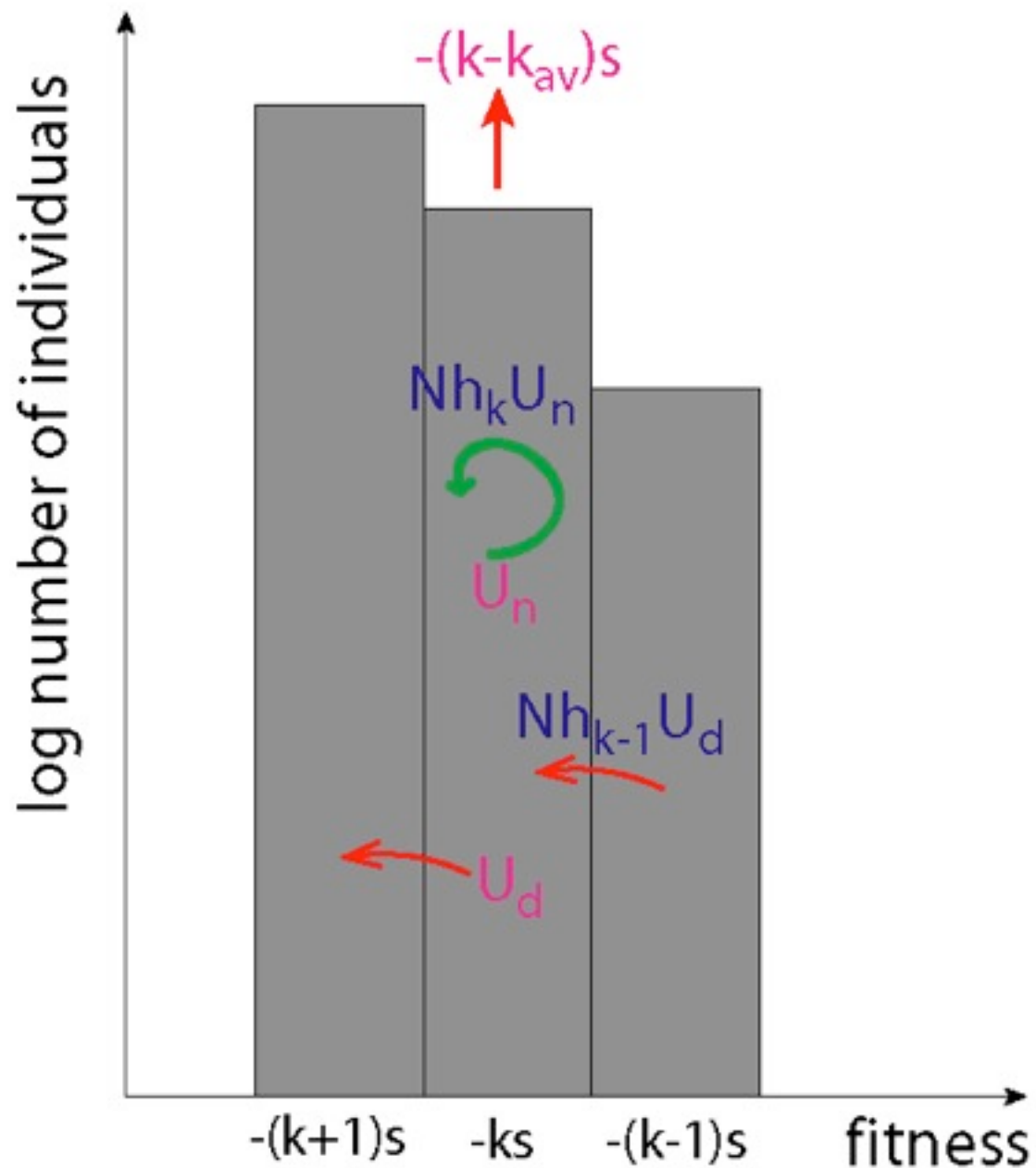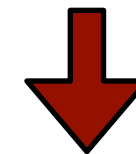Experience effective selective pressure:

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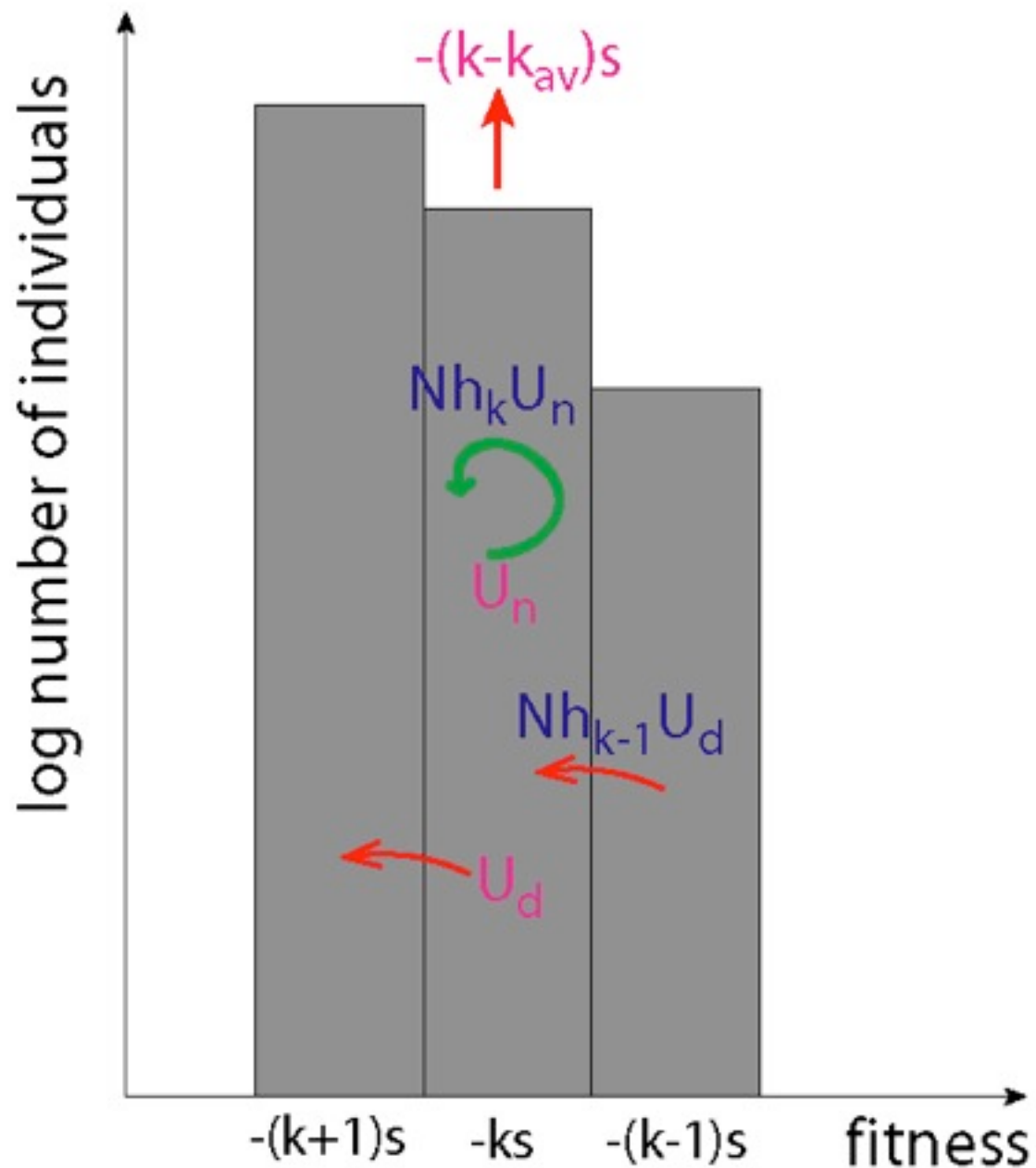
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with neutral mutations:

- $s_k < 0$, effective selection even more negative
- even $s_0 < 0$, all classes effectively selected against!

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Balance between creation and destruction of alleles

Balance between creation and destruction of alleles

Fluctuations of particular mutations are *not* independent. Fluctuations of alleles *are*. [26]

Balance between creation and destruction of alleles

→ Distribution of probability of seeing an allele frequency x:

$$f_k(x)dx = \theta_k \frac{1 - e^{-2Ns_k(1-x)}}{(1 - e^{-2Ns_k})x(1-x)}dx$$

Poisson Random Field (PRF) gives distribution of lineages in given fitness class

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self-consistency condition - fluctuations of alleles affect the mean fitness and the rate of mutations to less-fit alleles

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PRF - qualitatively determines the intensity of selection on a particular gene

The model: $p(x; x_0, t)$ probability distribution of derived allele frequency x at time t, given x_0 at time t_0

D x s

O $1 - x$

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$q(x_0; x, t)$ backward equation: $\partial_t q(x_0; x, t) = v(x_0) \dfrac{\partial q(x_0; x, t)}{\partial x_0} + \dfrac{D(x_0)}{2} \dfrac{\partial^2 q(x_0; x, t)}{\partial x_0^2}$

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absorbing boundary conditions
at $x = 1$ or $x = 0$

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$$T(x_0) = \int_0^1 \tilde{f}(x; x_0)dx$$ - mean time until absorption (MFPT)

mean time derived allele frequency spends in the interval $(x, x + dx)$:

$$\tilde{f}(x) = \frac{1 - e^{2Ns(1-x)}}{1 - e^{2Ns}}\frac{2}{x(1 - x)}$$

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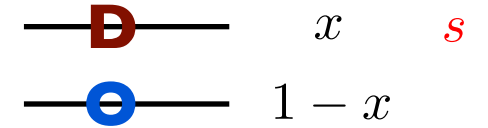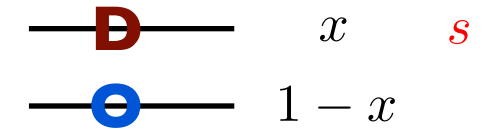
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- mutations arise at Poisson times
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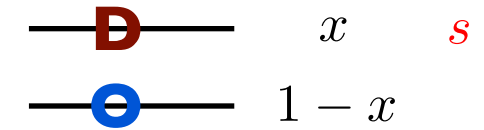→ mean time derived allele frequency spends in the interval $(x, x + dx)$:

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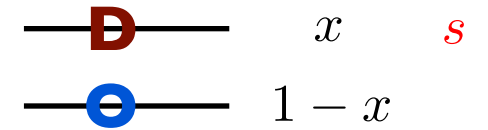
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$$\int_{x_1}^{x_2} \theta \tilde{f}(x)dx = \int_{x_1}^{x_2} f(x)dx$$ - expected number of sites with derived allele/lineage frequency in a given range:

$$f(x)dx = \theta\frac{1 - e^{2Ns(1-x)}}{(1 - e^{2Ns})x(1 - x)}dx$$

θ **-** per site mutation rate

PRF - qualitatively determines the intensity of selection on a particular gene

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$q(x_0; x, t)$

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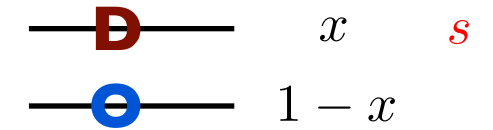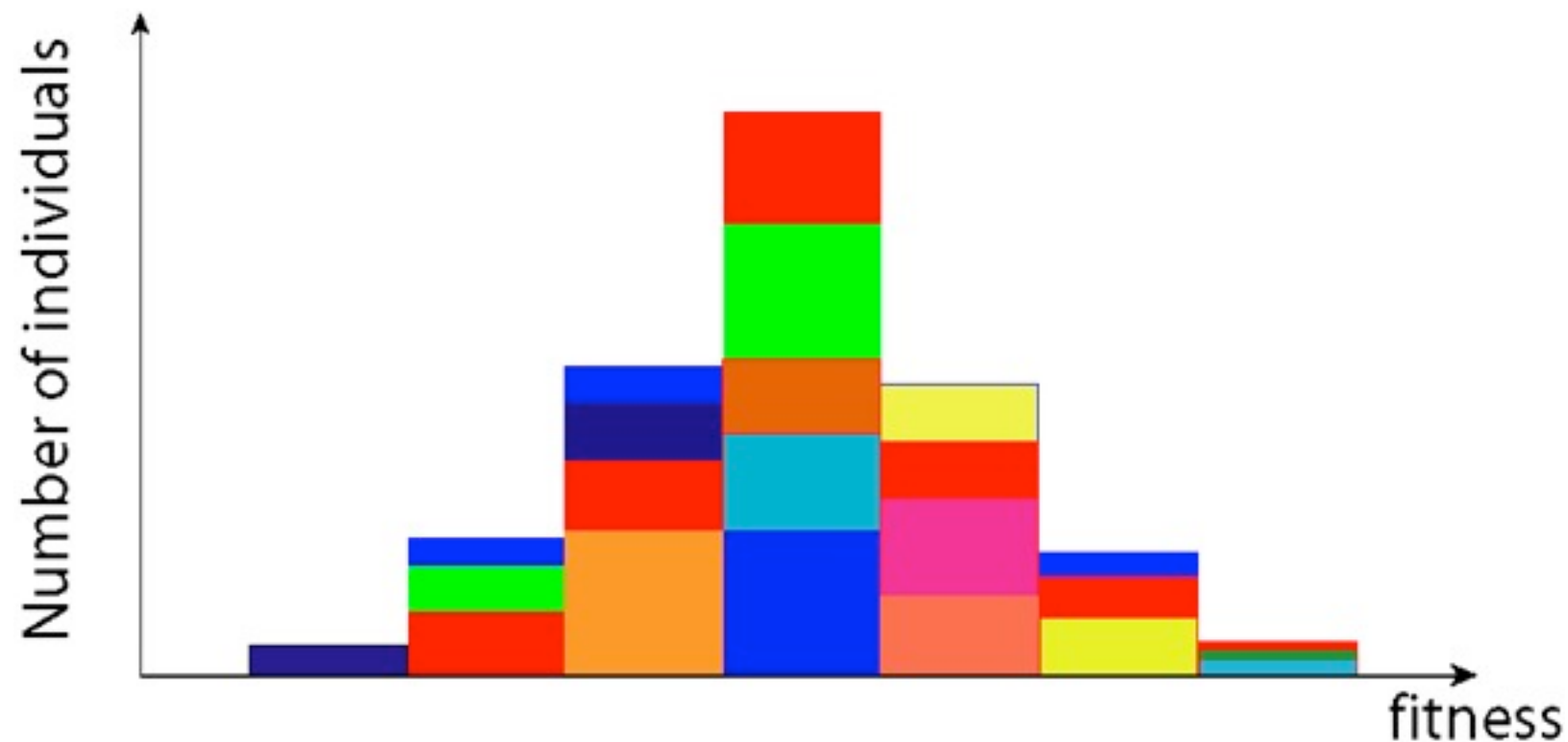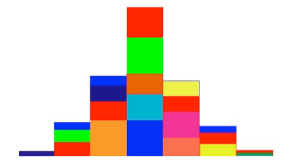
$$f(x) dx = \theta \frac{1 - e^{2Ns(1-x)}}{(1 - e^{2Ns}) x(1-x)} dx$$

θ - per site mutation rate

The number of sites that have i copies of the derived allele are Poisson distributed with mean:

probability that the site has i copies in the sample $\longrightarrow$ $\int \binom{n}{i} x^i (1-x)^{n-i} f(x) dx$

[hence Poisson Random Field]

33

Balance between creation and destruction of alleles
→ Distribution of probability of seeing an allele frequency x:

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Poisson Random Field (PRF) gives
distribution of lineages in given fitness class
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self-consistency condition - fluctuations of alleles affect the
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Fluctuations of particular mutations are *not* independent. Fluctuations of alleles *are*. **34**

Allelic diversity within each class

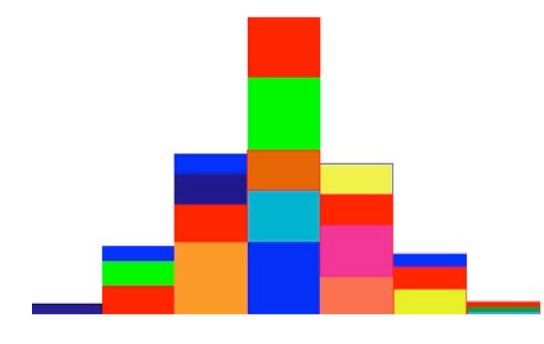

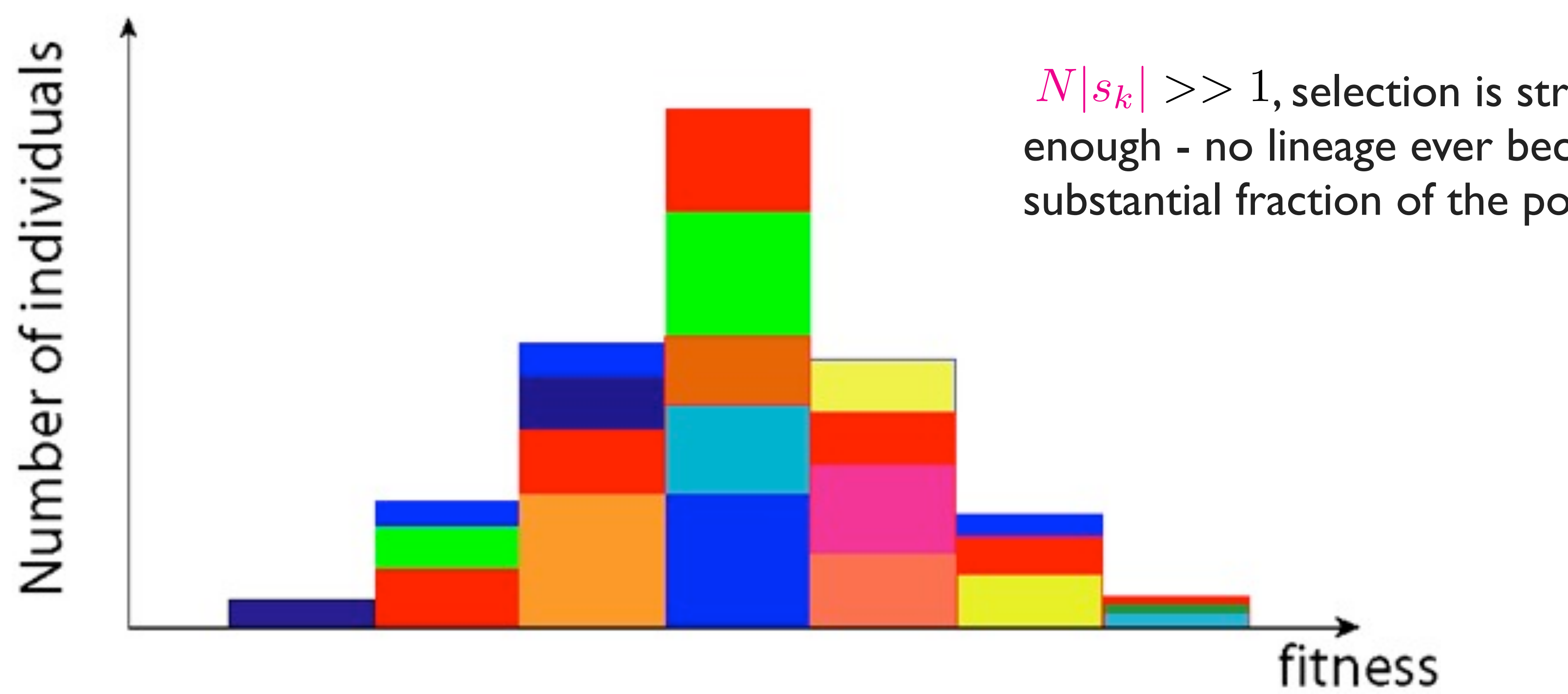

$N|s_k| >> 1$, selection is strong enough - no lineage ever becomes a substantial fraction of the population

Balance between creation and destruction of alleles

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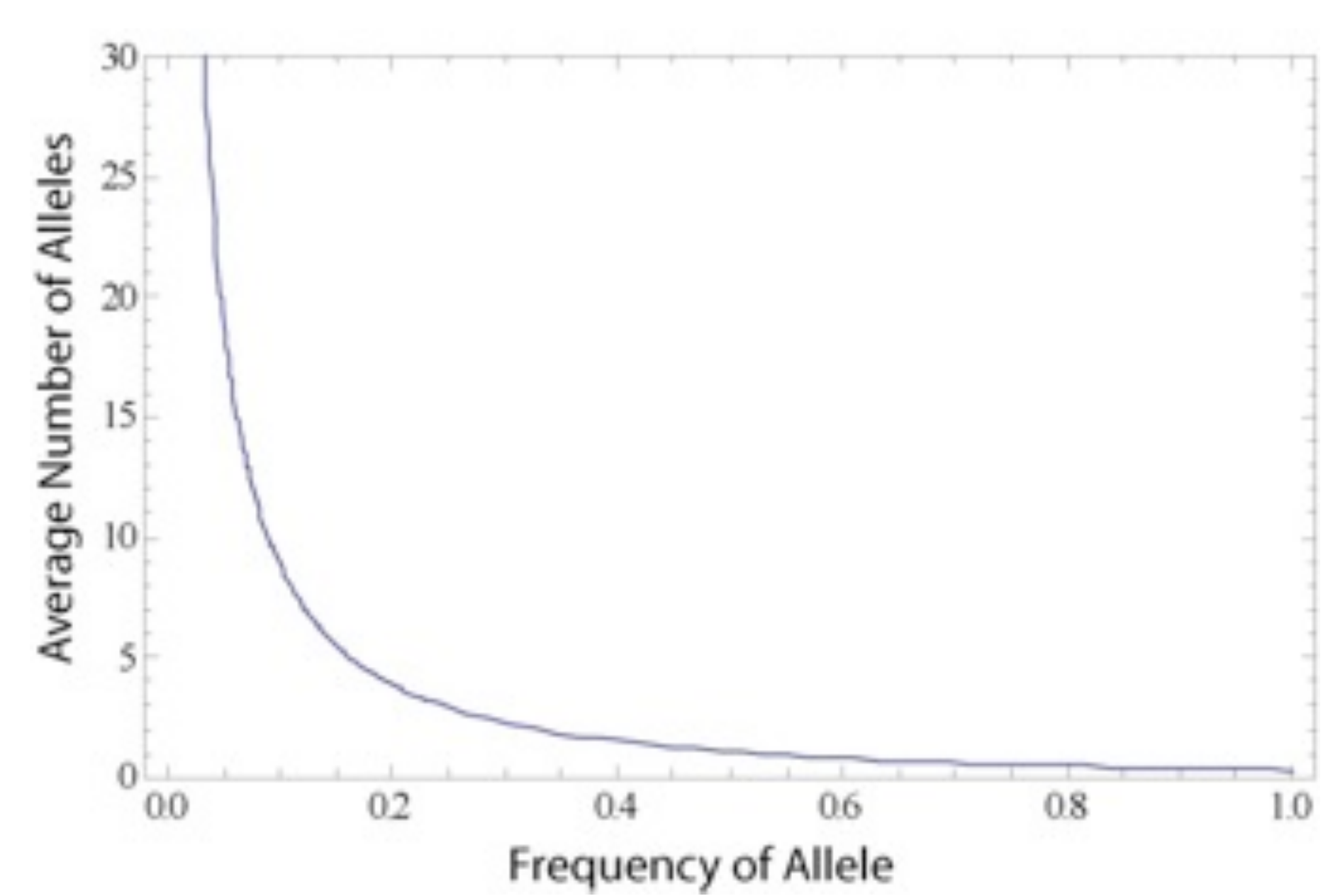

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steady state distribution of fitness classes

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$$\int_0^1 \frac{1 - e^{-2\gamma_k x}}{x} dx = \frac{1 - e^{-2\gamma_k}}{2|\gamma_k|}$$

$$N|s_k| >> 1 \qquad\qquad s_k = -U_d - U_n - (k - k_{av})s$$

Self-consistency condition

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close to k_{av}:

$N(U_d + U_n) >> 1$

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$NU_d \gg 1$ **or** $NU_n \gg 1$ $\longleftarrow$ self-consistency holds

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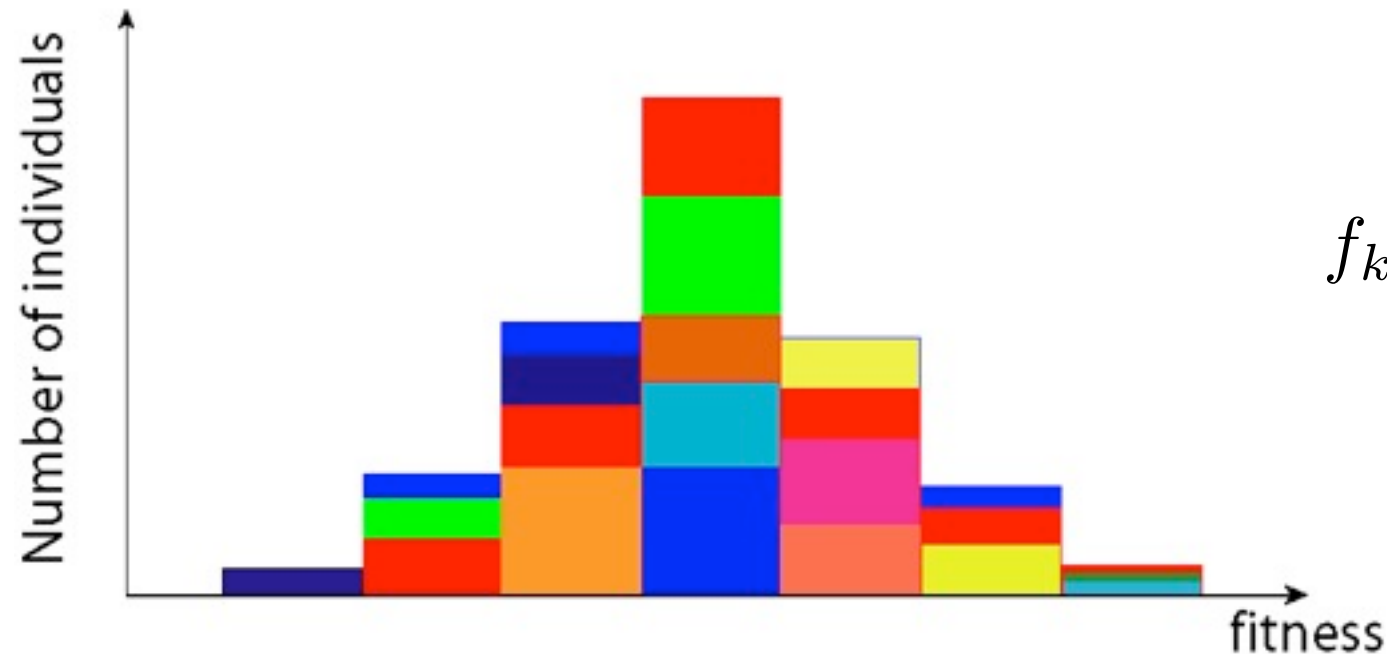close to k_{av}:
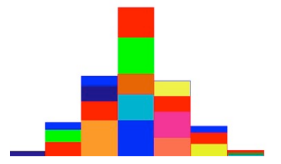
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$$NU_d >> 1 \quad \text{or} \quad NU_n >> 1 \quad \longleftarrow \quad \text{self-consistency holds}$$

For $N(U_d + U_n) < 1$ PRF breaks down :

- the growth of some mutants is limited by size of population
- lineages are no longer independent

Number of individuals

$$f_k(x)dx = \theta_k \frac{1 - e^{-2Ns_k(1-x)}}{(1 - e^{-2Ns_k})x(1-x)}dx$$

fitness

Sample n individuals.
What is the probability of a particular allelic configuration?

(n_1 individuals with allele 1, n_2 individuals with allele 2,....)

Homozygosity:
$$Q_2 = \sum_k \int x^2 f_k(x)dx = \sum_{k=0}^{\infty} \frac{h_k}{2Ns_k}$$

Sample n=2 individuals. What is the probability that they have the same genotype?

"Bizygosity":
$$Q_{2,1} = \sum_k \int 3x^2(1-x)f_k(x)dx$$
$$= 3\sum_{k=0}^{\infty} \frac{h_k}{2Ns_k}(1 - \frac{1}{Ns_k})$$

Sample n=3 individuals. What is the probability that two have the same alleles and one is different?

Sample n individuals.
What is the probability of a particular allelic configuration?

(n_1 individuals with allele 1, n_2 individuals with allele 2,....)

➡ generalization of Ewens Sampling Formula (ESF)

$$P(n_1, ..., n_2) = \frac{n!}{\theta(\theta + 1)...(\theta + n - 1)} \prod_{j=1}^{n} \frac{\theta^{n_j}}{j^{n_j} n_j!}$$

- neutral model
- steady state with respect to mutation and drift
- infinite alleles
- sample size n<<N - population size

Sample n individuals.
What is the probability of a particular allelic configuration?

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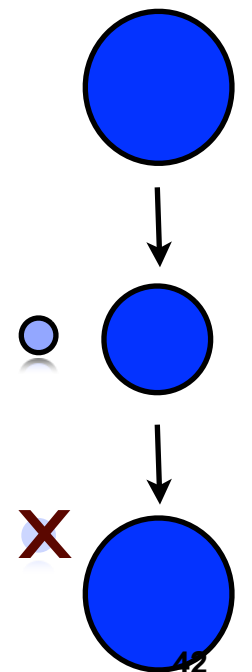➡ generalization of Ewens Sampling Formula (ESF)

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- neutral model
- steady state with respect to mutation and drift
- infinite alleles
- sample size n<<N - population size

Effective Population Size Approximation (EPS):

- deleterious mutations are purged quickly from the population
- all individuals are recently descended from neutral individuals
- only the zero-class matters
- results in neutral population with an effective reduced population size
- makes predictions about diversity at individual sites
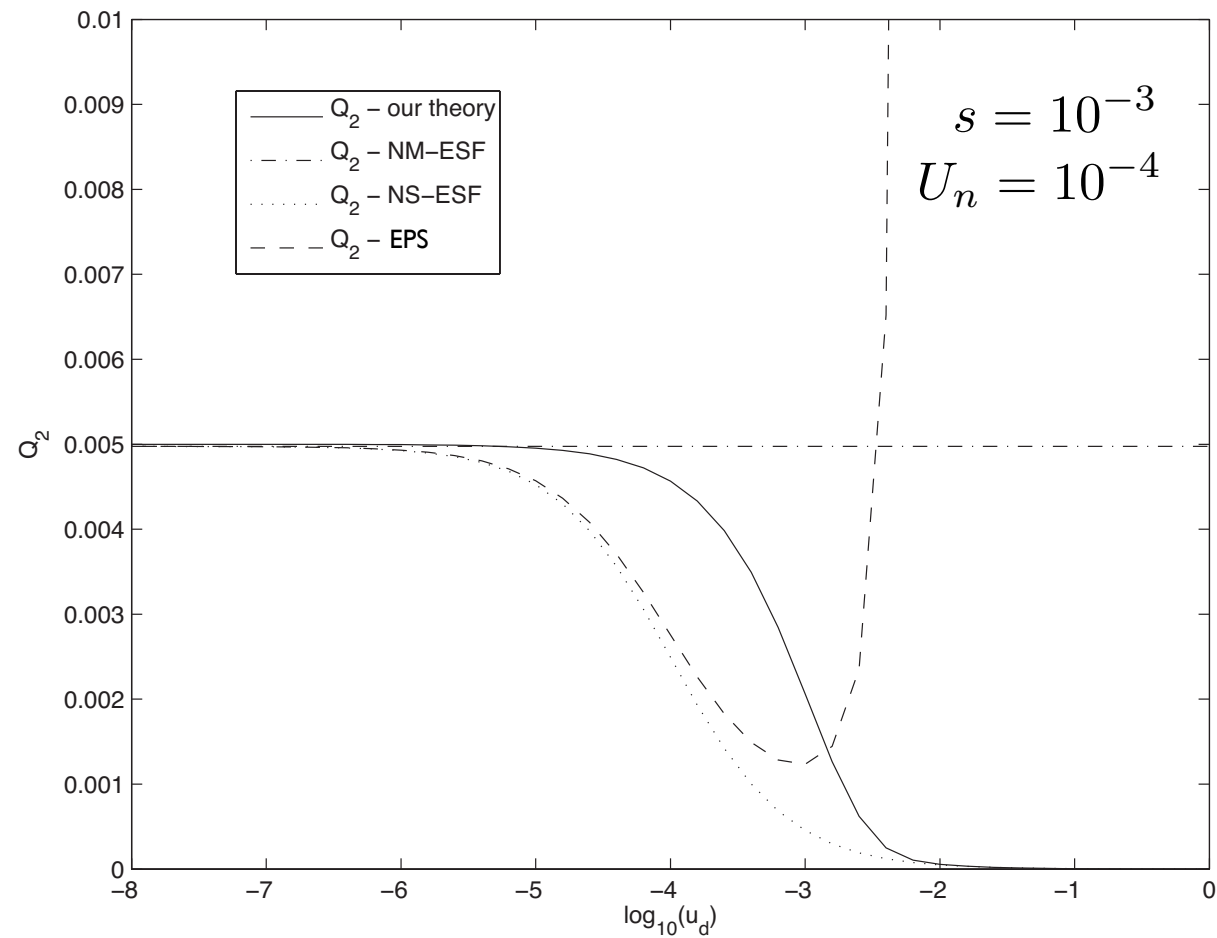- only makes predictions for neutral sites

$$N_e = N h_0 = N e^{-U_d/s}$$

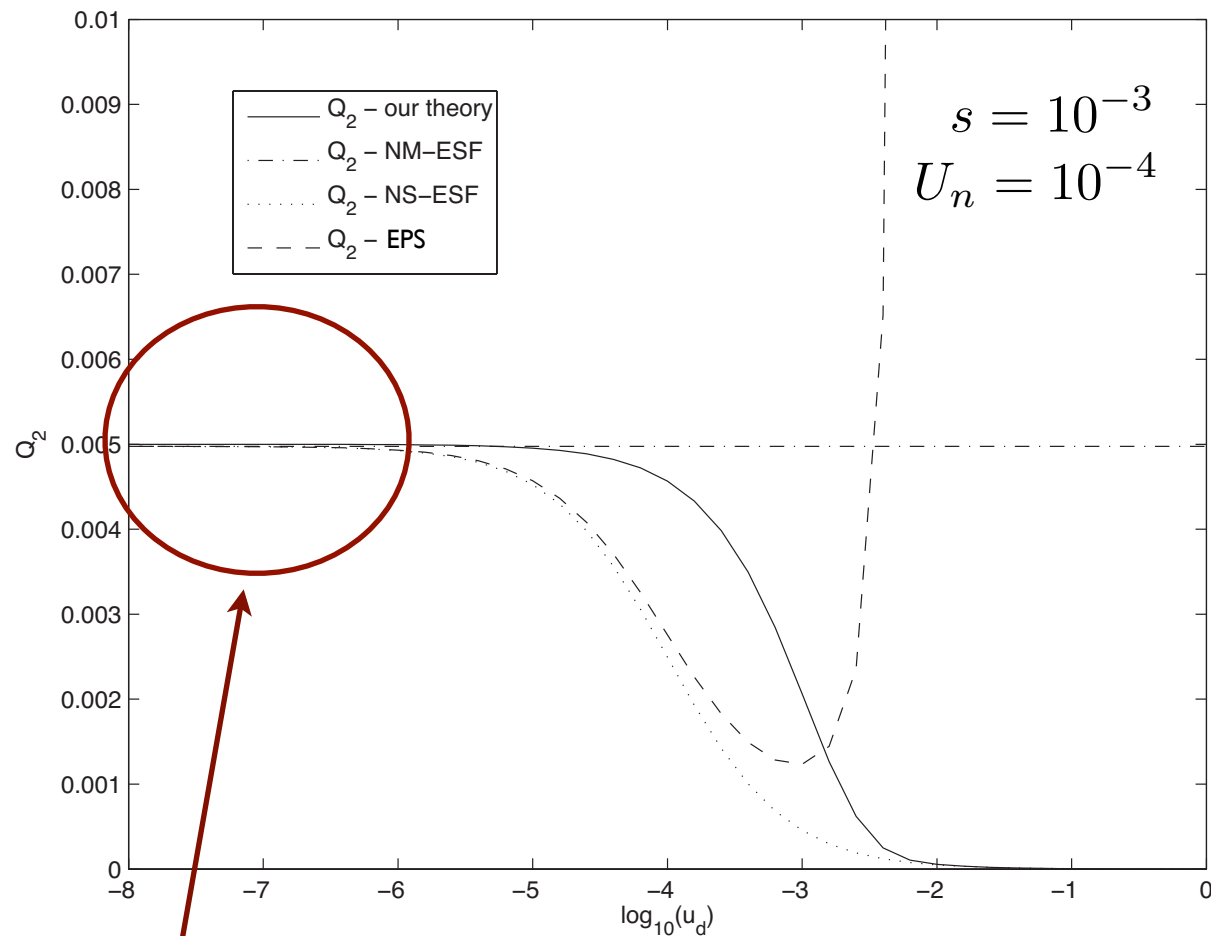Homozygosity: $Q_2 = \sum_k \int x^2 f_k(x)dx = \sum_{k=0}^{\infty} \frac{h_k}{2Ns_k}$

$s = 10^{-3}$
$U_n = 10^{-4}$

Homozygosity: $\quad Q_2 = \sum_k \int x^2 f_k(x) dx = \sum_{k=0}^{\infty} \frac{h_k}{2Ns_k}$

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Legend:
- Q_2 – our theory
- Q_2 – NM–ESF
- Q_2 – NS–ESF
- Q_2 – EPS

(vertical axis: Q_2; horizontal axis: $\log_{10}(u_d)$)

neutral case, $U_d = 0$: $\qquad \theta = 2NU_n >> 1$

$$Q_2 = \frac{1}{\theta}$$
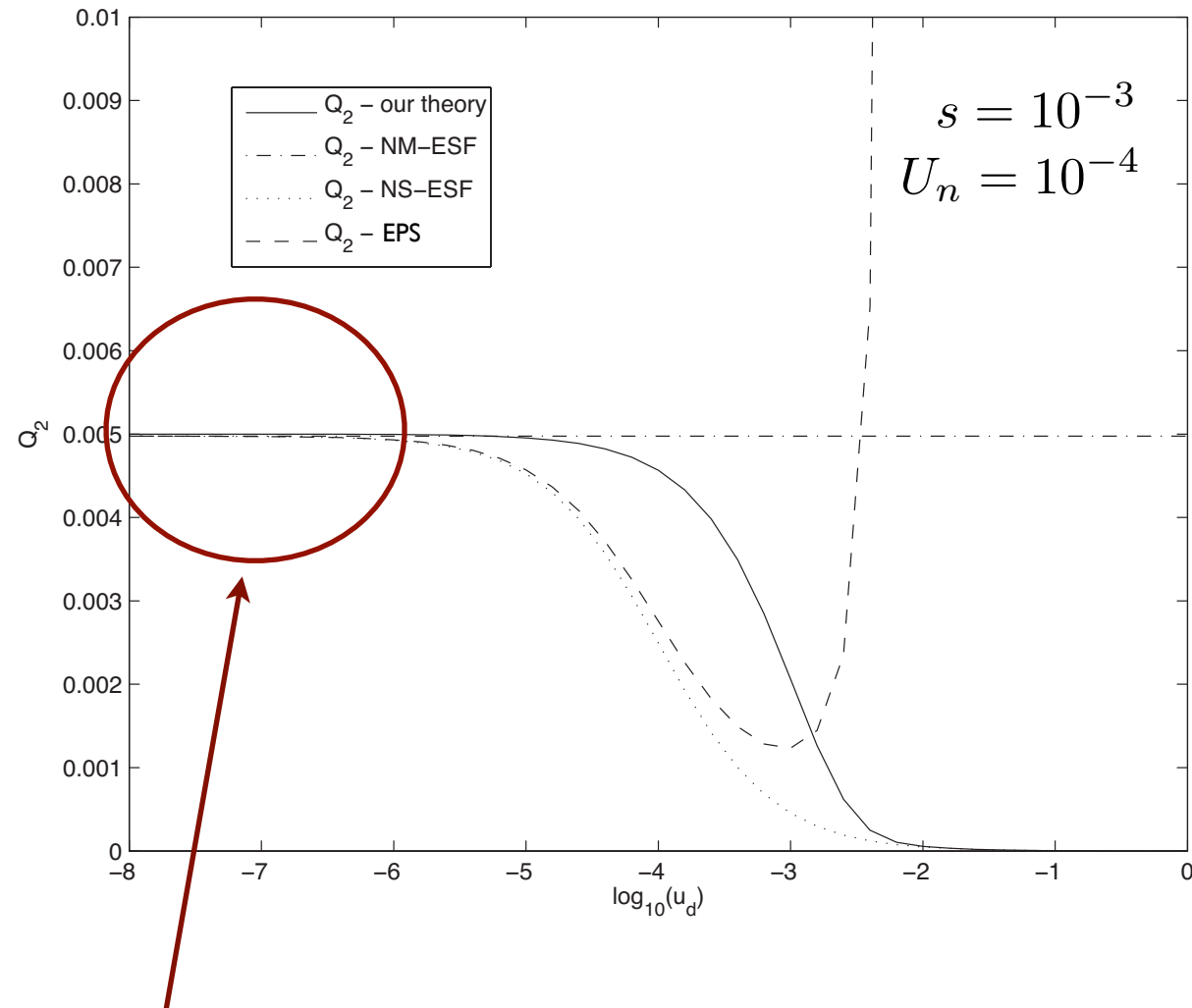
$$Q_2^{ESF} = \frac{1}{1 + \theta}$$

- all neutral models agree: ESF, BGS

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EPS - change in reduced effective population size of "neutral" population:

$$\theta = 2N(U_n + U_d)e^{-U_d/|s|}$$

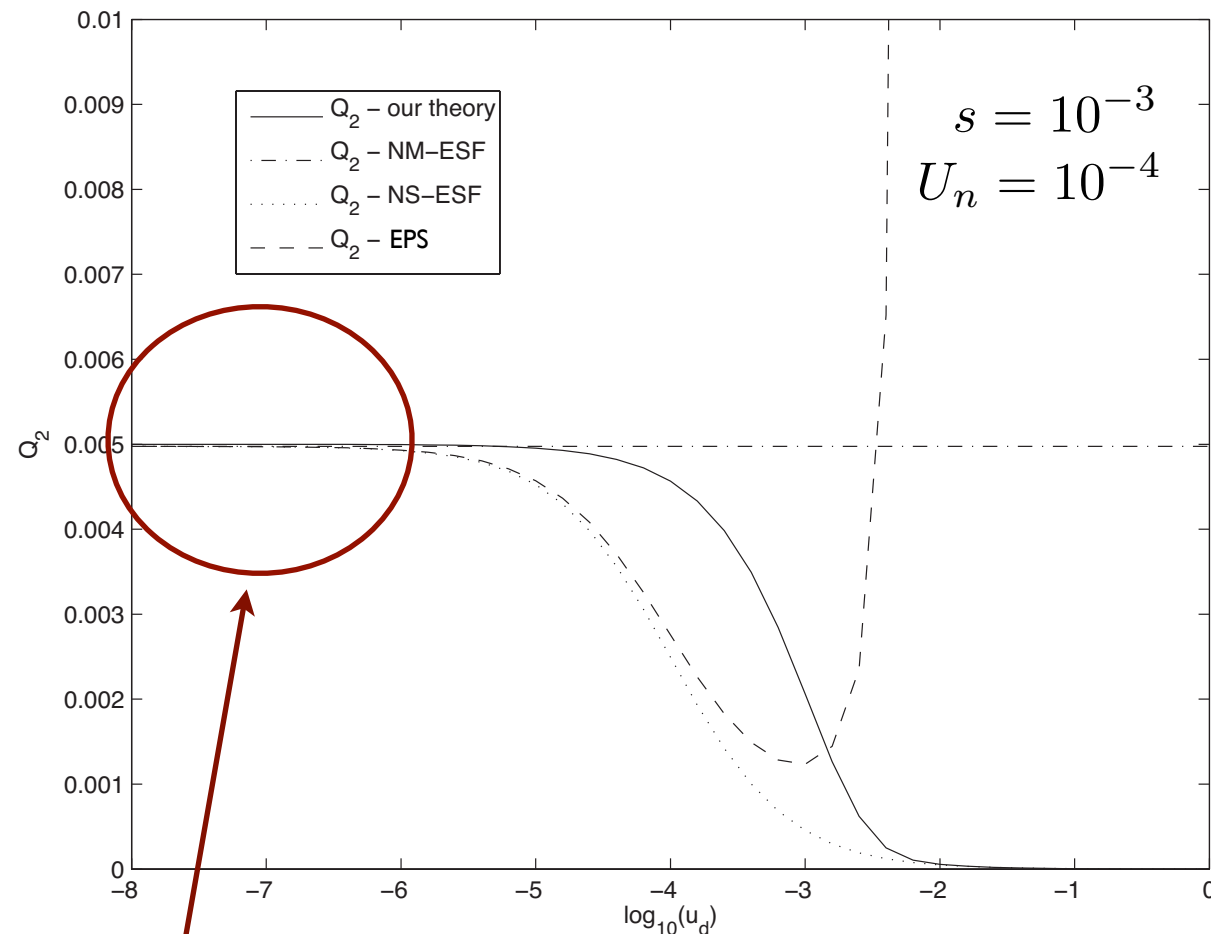neutral case, $U_d = 0$: $\qquad \theta = 2NU_n >> 1$

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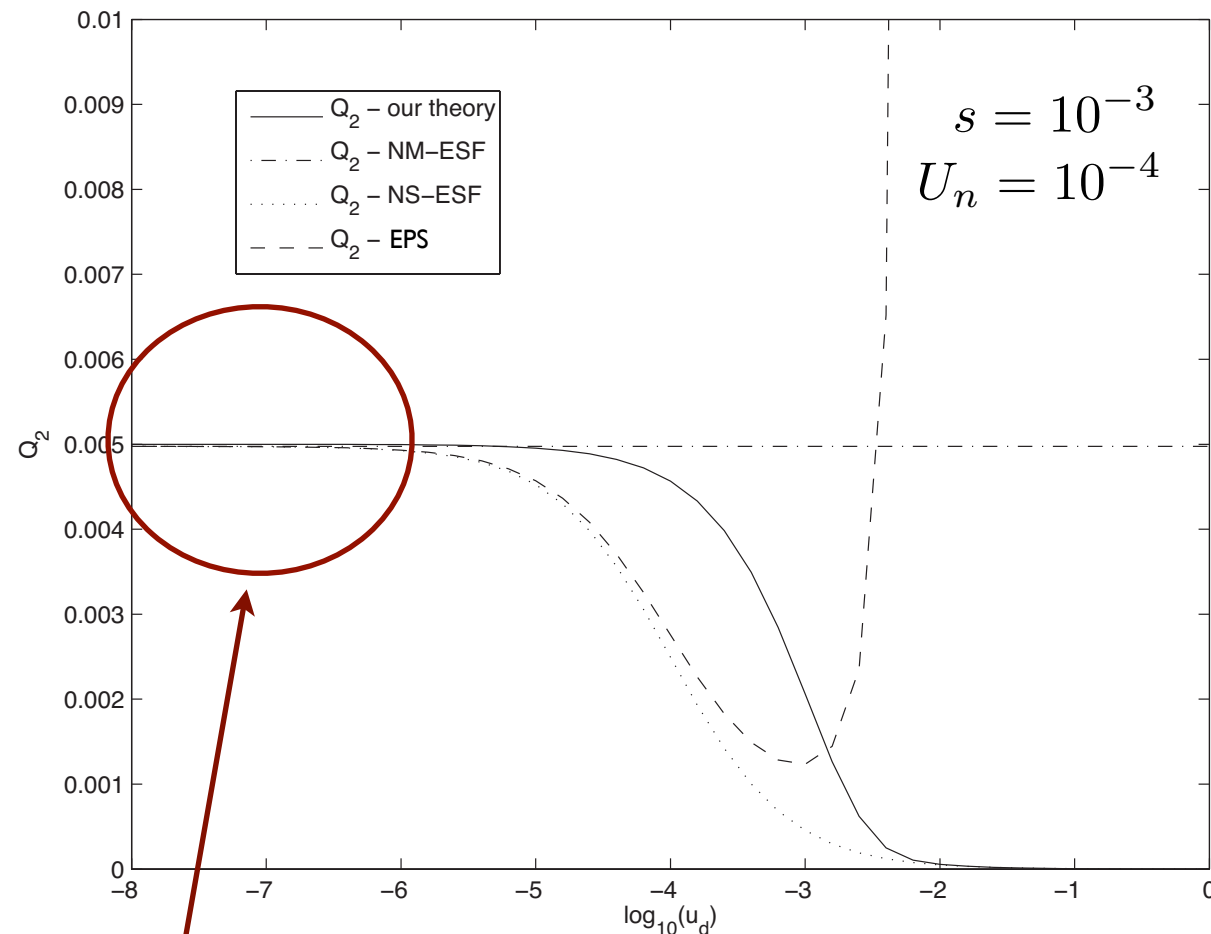neutral case, $U_d = 0$: $\qquad \theta = 2NU_n >> 1$

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Expected genetic variation

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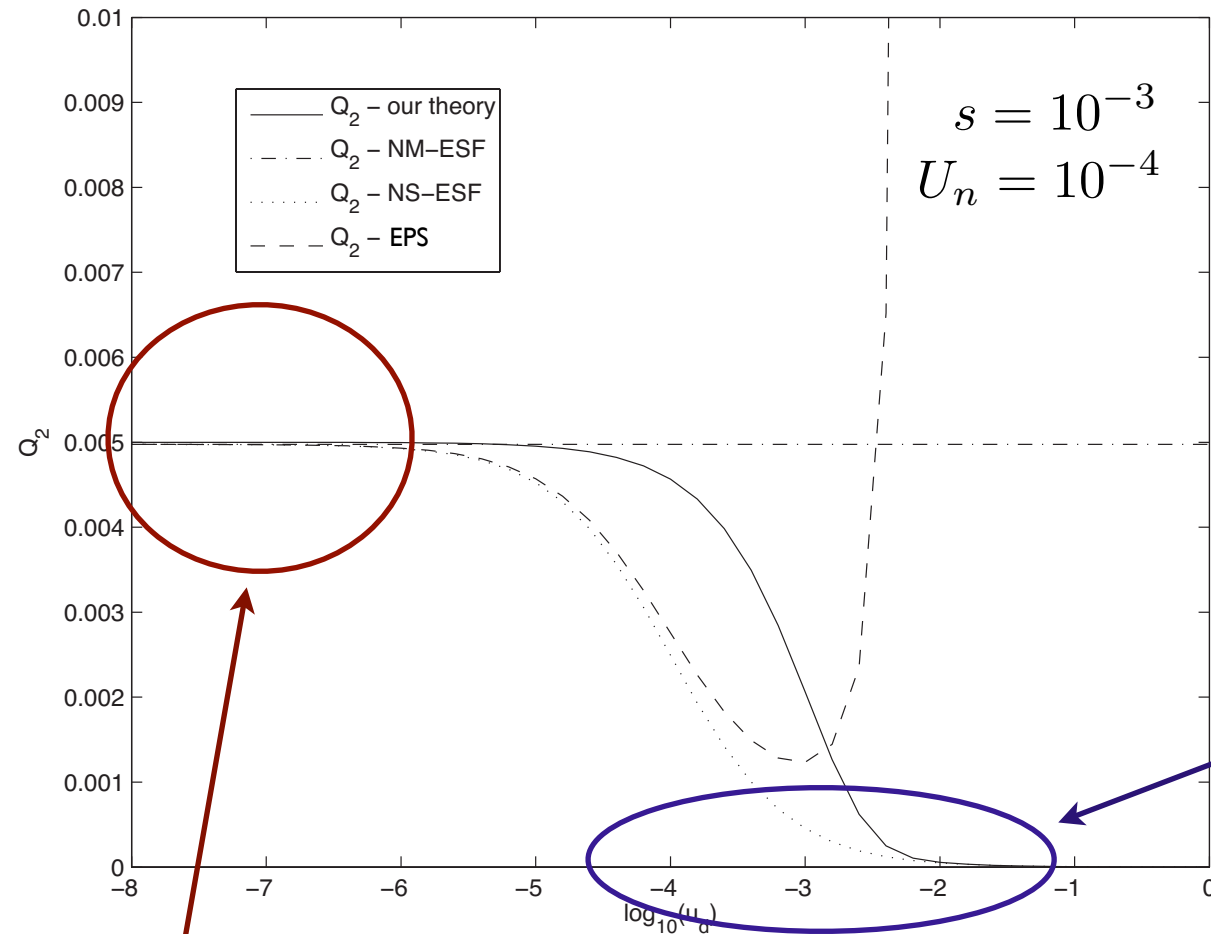NM-ESF - neglect deleterious mutations: $\theta = 2NU_n$

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• deleterious mutations decrease the homozygozity, $U_d \approx s$
• deleterious mutations decrease homozygosity less than neutral ones (they must eventually die)
• deleterious mutations are not rare for $U_d > s$, NM-ESF breaks down
• for $U_d > s$ still significant difference between NS-ESF and our results
• important parameter: are mutations purged slowly enough to matter $U_d \approx s$
• contrary to intuition from EPS, more deleterious mutations cannot decrease diversity

47

Homozygosity:
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neutral case, $U_d = 0$: $\qquad \theta = 2NU_n \gg 1$
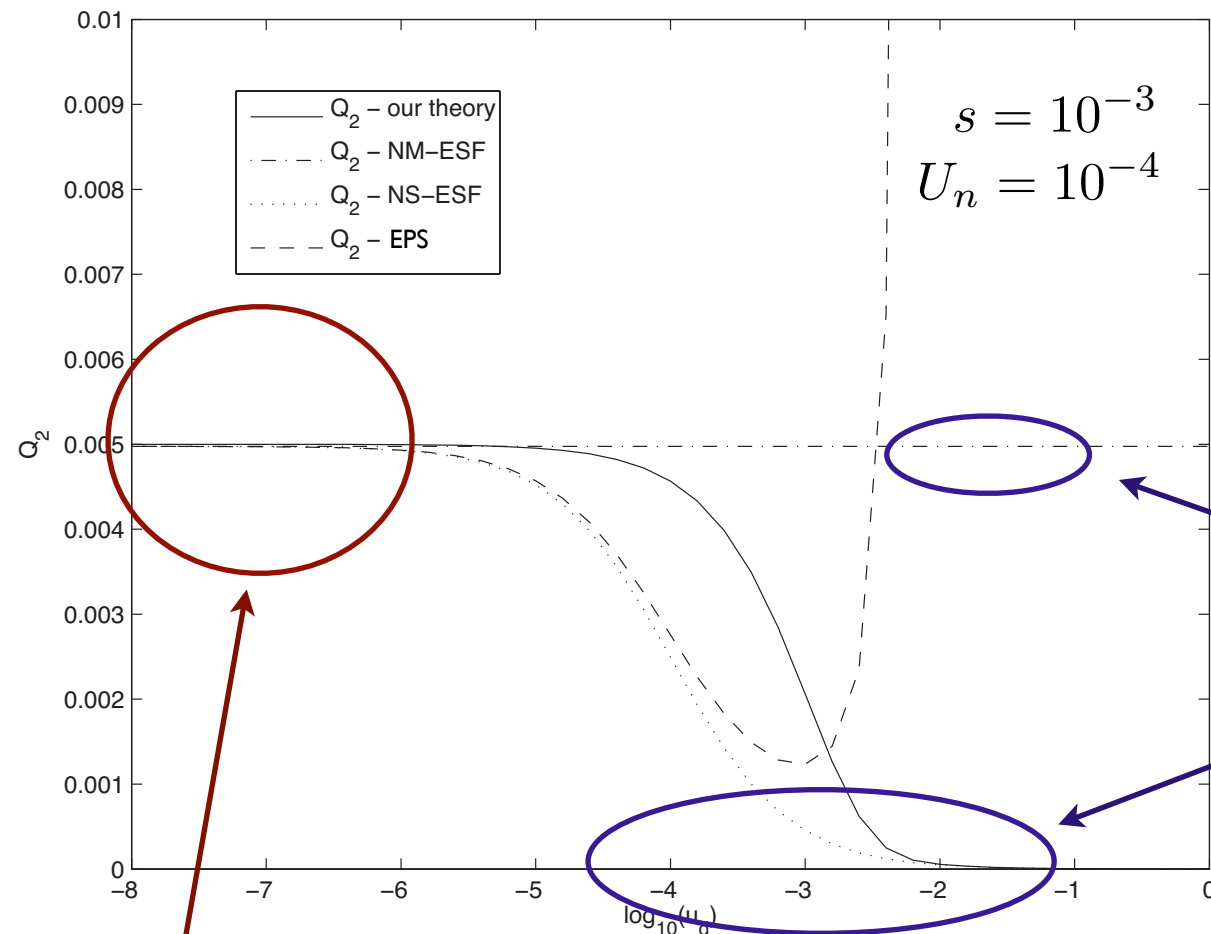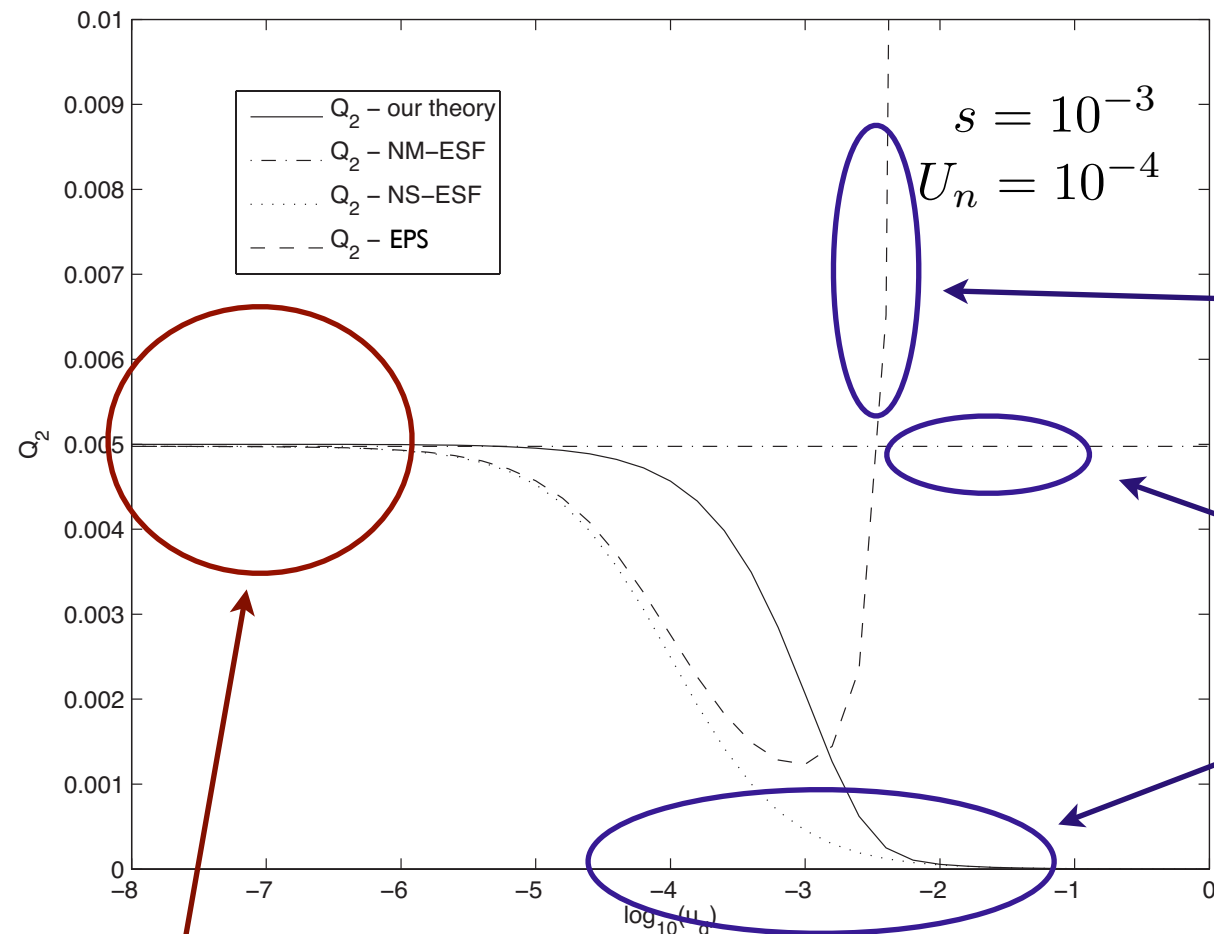
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Homozygosity:
$$Q_2 = \sum_k \int x^2 f_k(x)dx = \sum_{k=0}^{\infty} \frac{h_k}{2Ns_k}$$

$$U_d = 10^{-4.5}$$
$$U_n = 10^{-4}$$

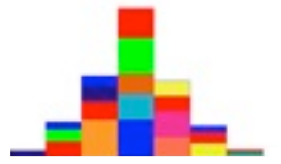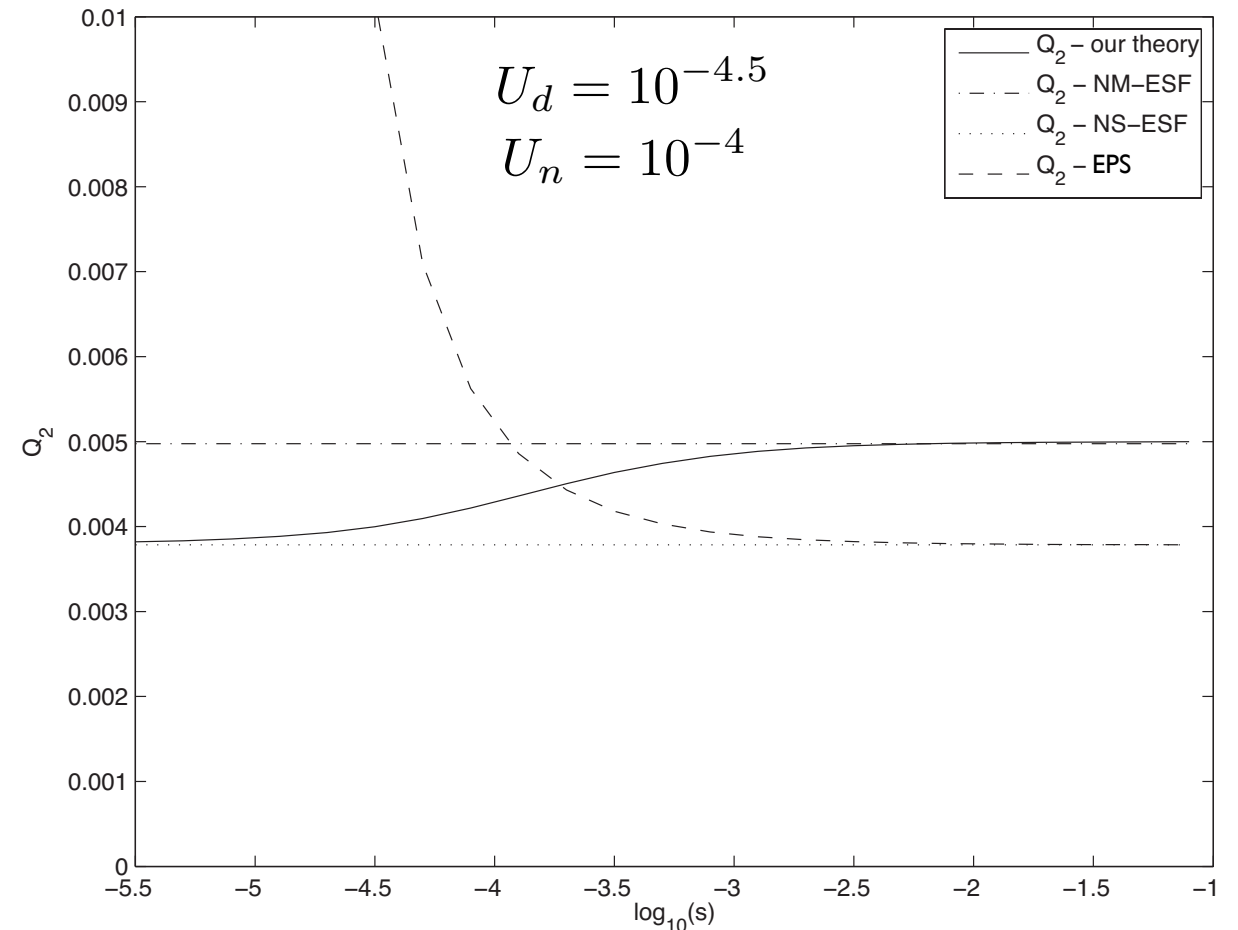NM-ESF - neglect deleterious mutations: $\theta = 2NU_n$

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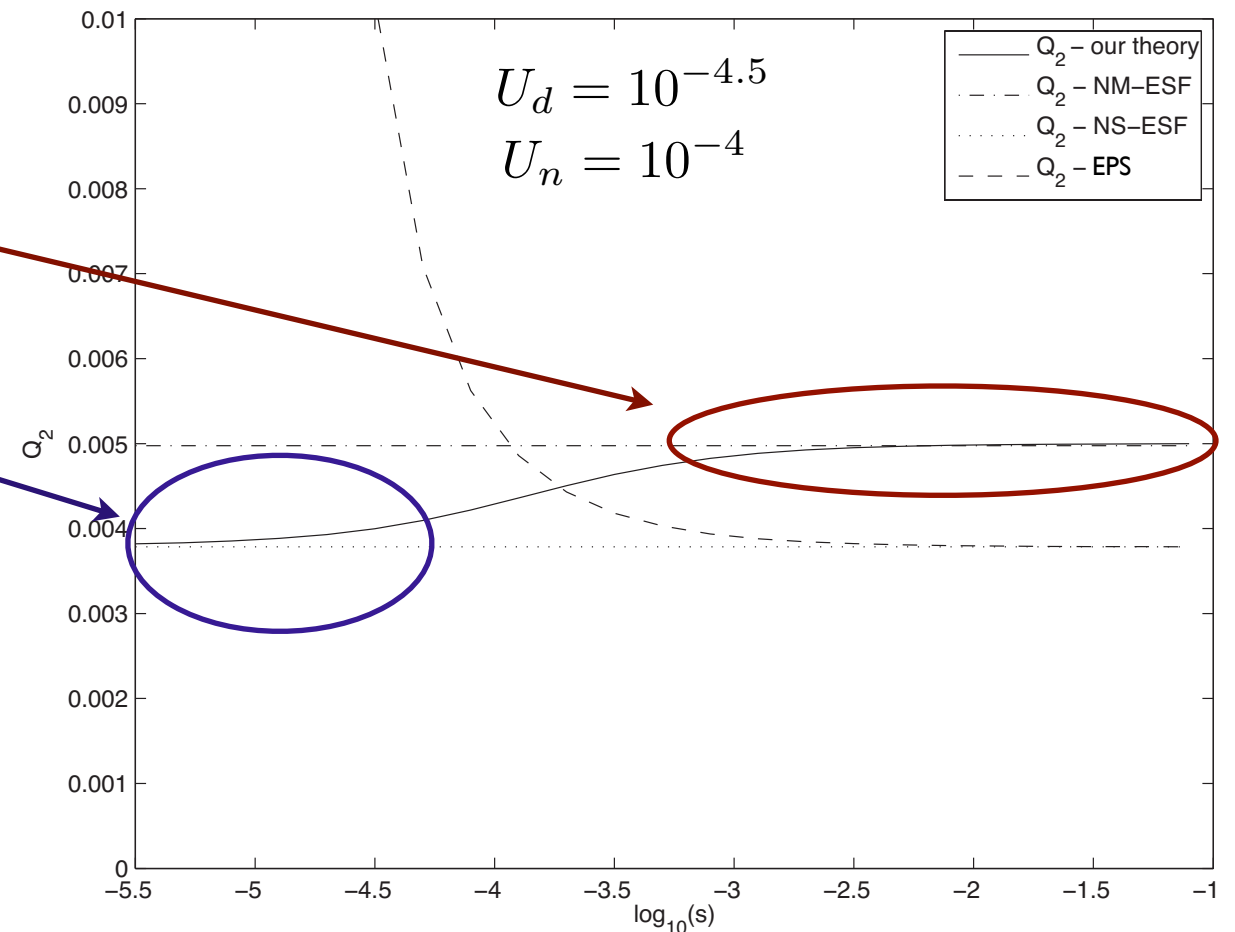
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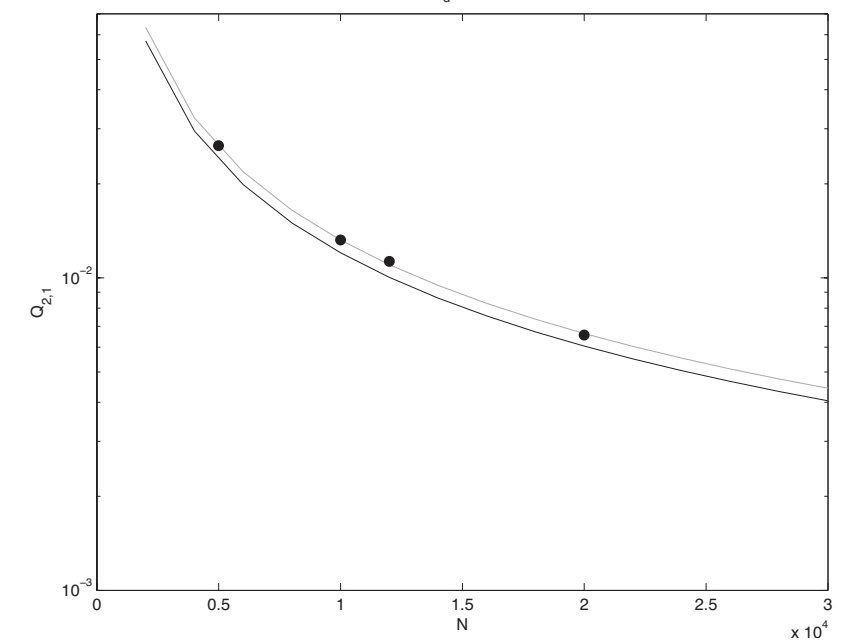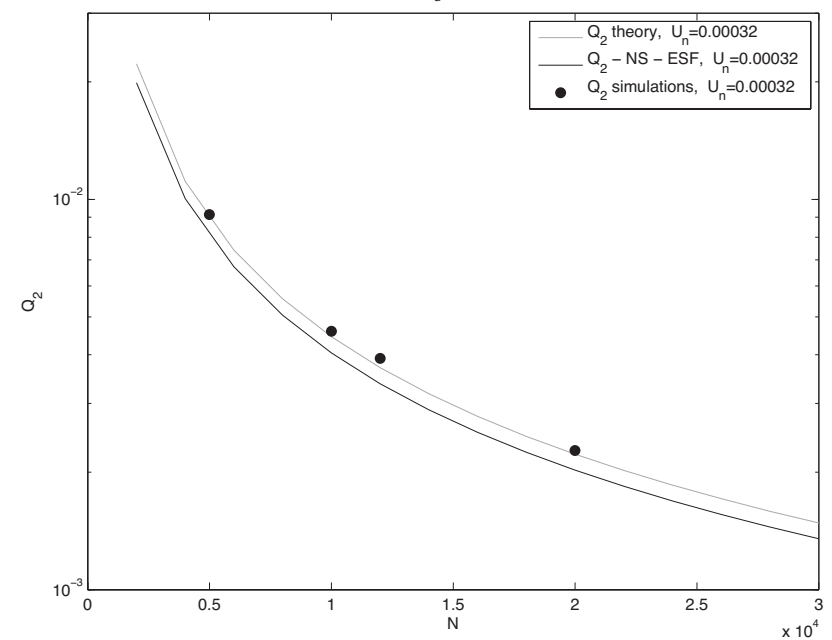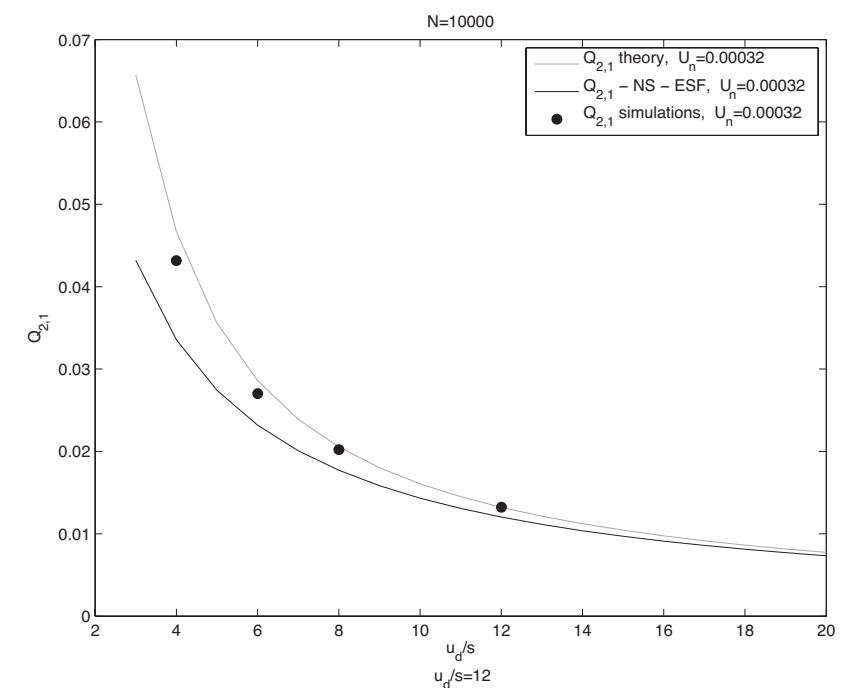$$U_d = 10^{-4.5}$$
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- for strong selection mutations eliminated quickly - neutral mutations dominate - NM-ESF holds
- for very weak selection, deleterious mutations are like neutral - NS-ESF holds
- regions where no neutral theory holds
- EPS underestimates size of most fit for weak selection

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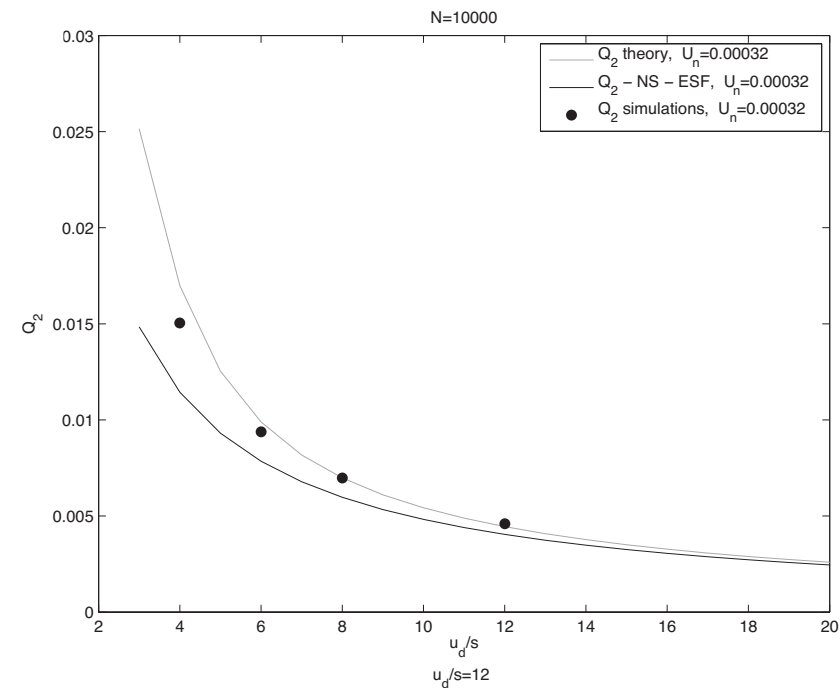
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- MC of Wright-Fisher population
- constant size N
- N individuals sampled with replacement in each generation
- sampling according to relative fitness in the population
- Poisson number of deleterious and neutral mutations introduced in each generation
- mutations introduced randomly and independently among individuals
- keep track of frequencies of all genotypes
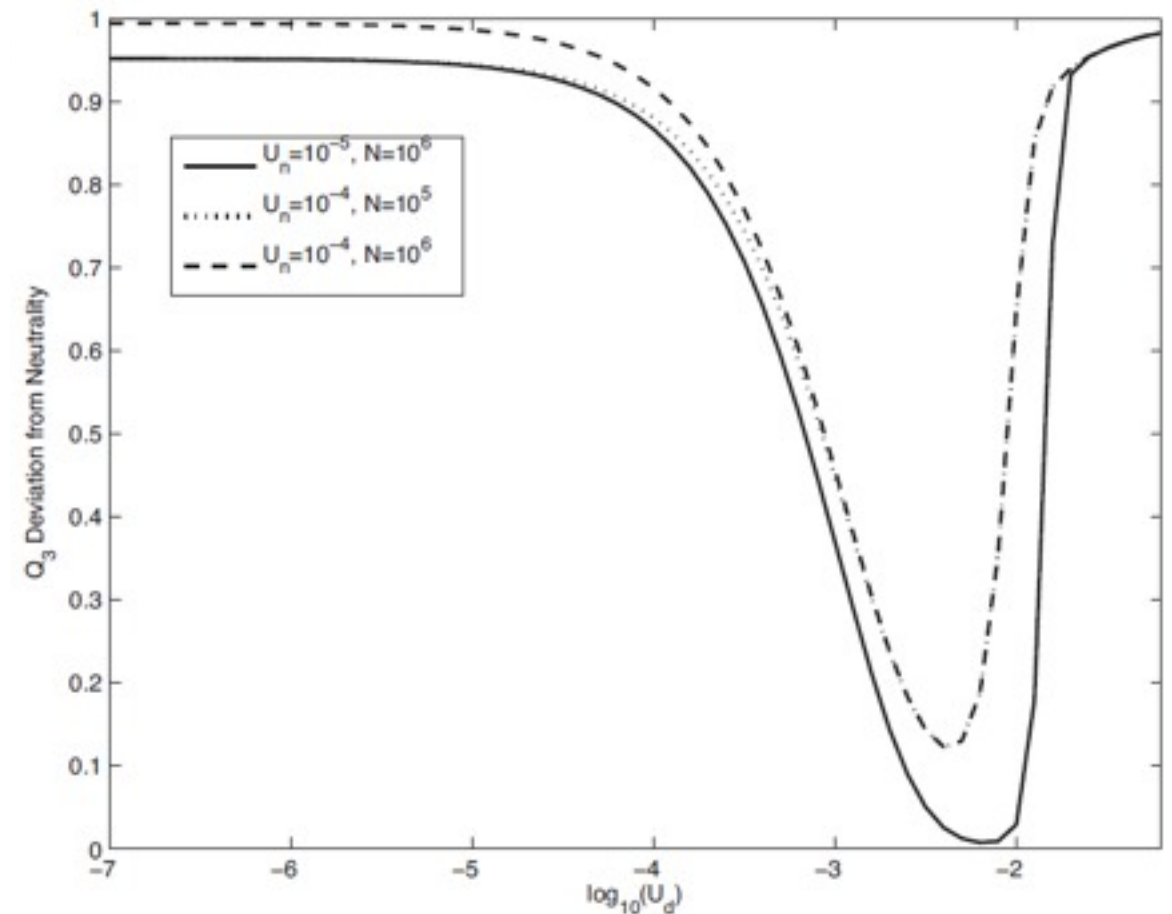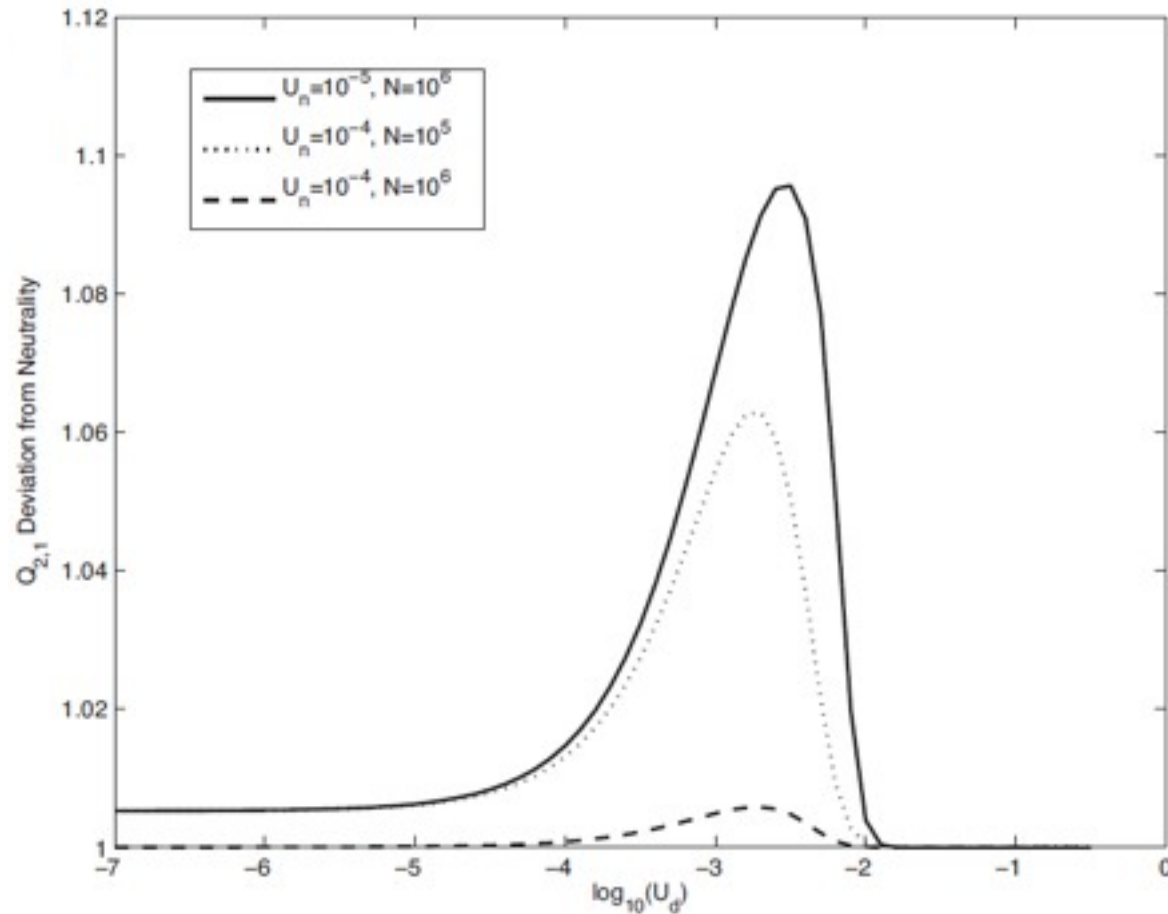- genotype - set of mutation sites

neutral ESF result:

$$Q_2^{ESF} = \frac{1}{1 + \theta_e}$$

compute effective
mutation rate:

$$\theta_e$$

calculate other
statistics:

$$Q_{2,1}^e, Q_3^e,$$

Compute expected Q$_{2,1}$ or Q$_3$ Given Q$_2$

Expected deviation from neutral ratio $\quad Q_{2,1}^e/Q_{2,1}$

$$Q_3^e/Q_3$$

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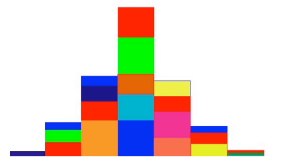There is no effective population size that reproduces the statistics consistently

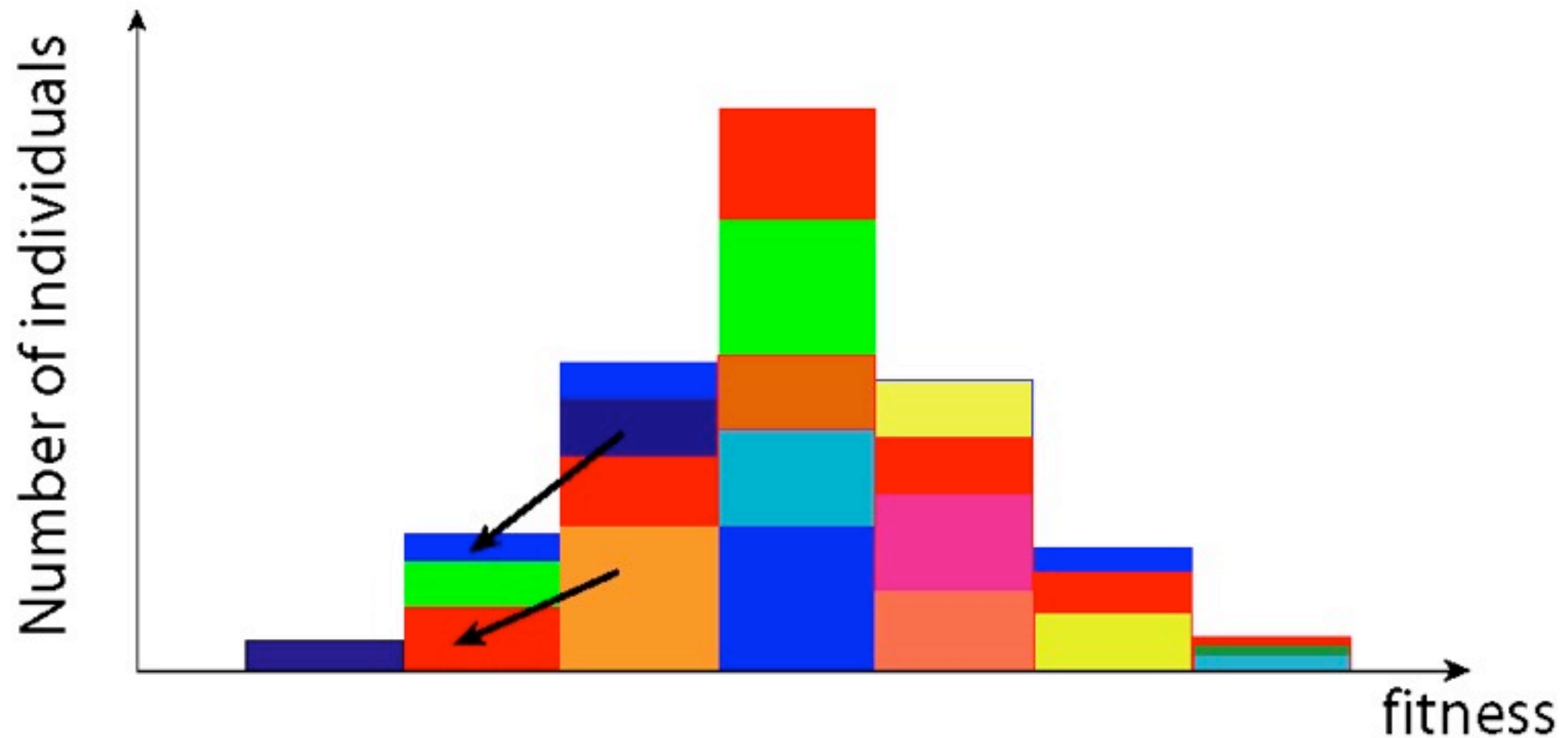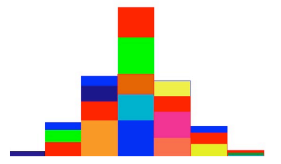We now know the probability of different allelic configurations

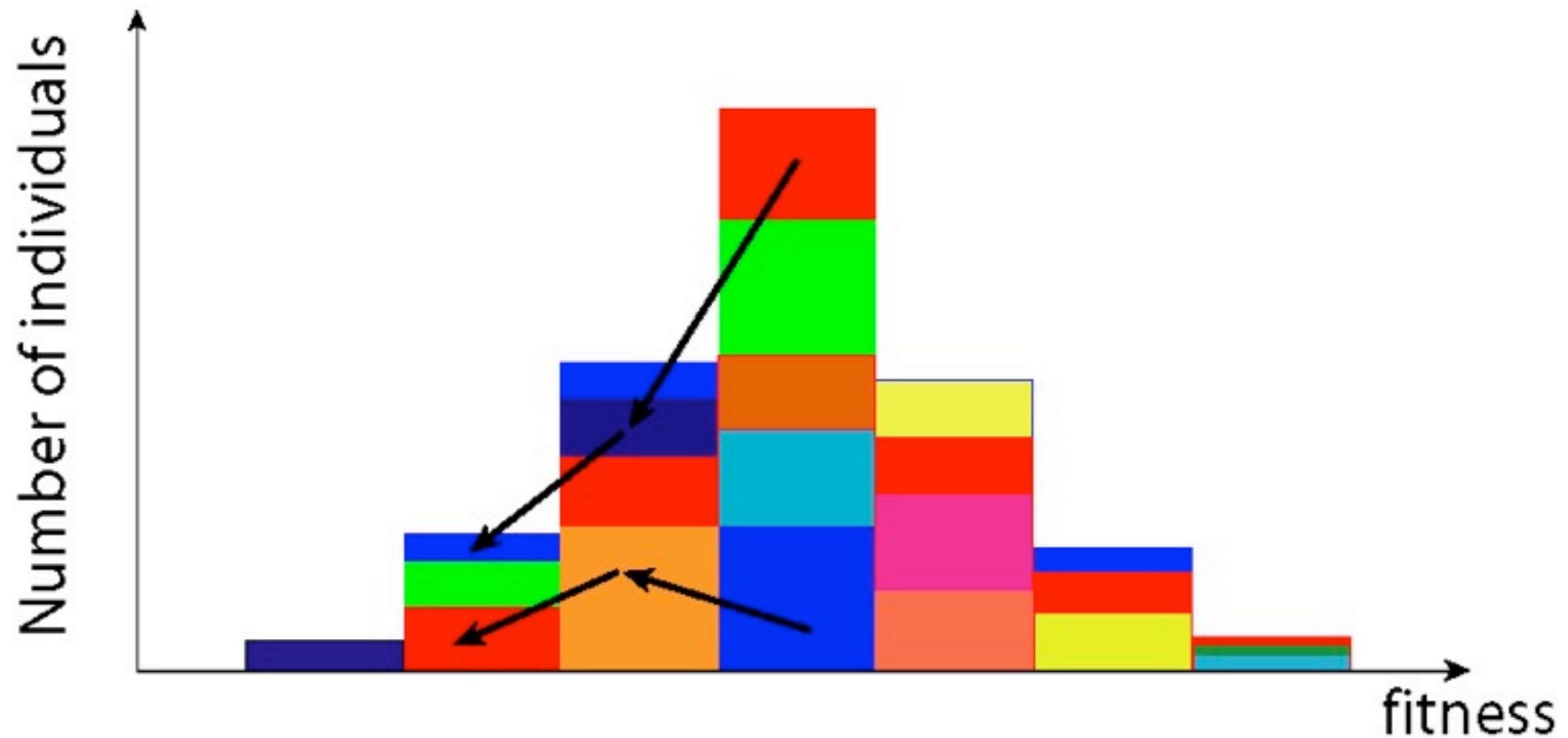What is the relationship among alleles?
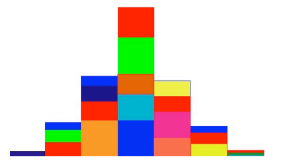
An effective coalescent approach

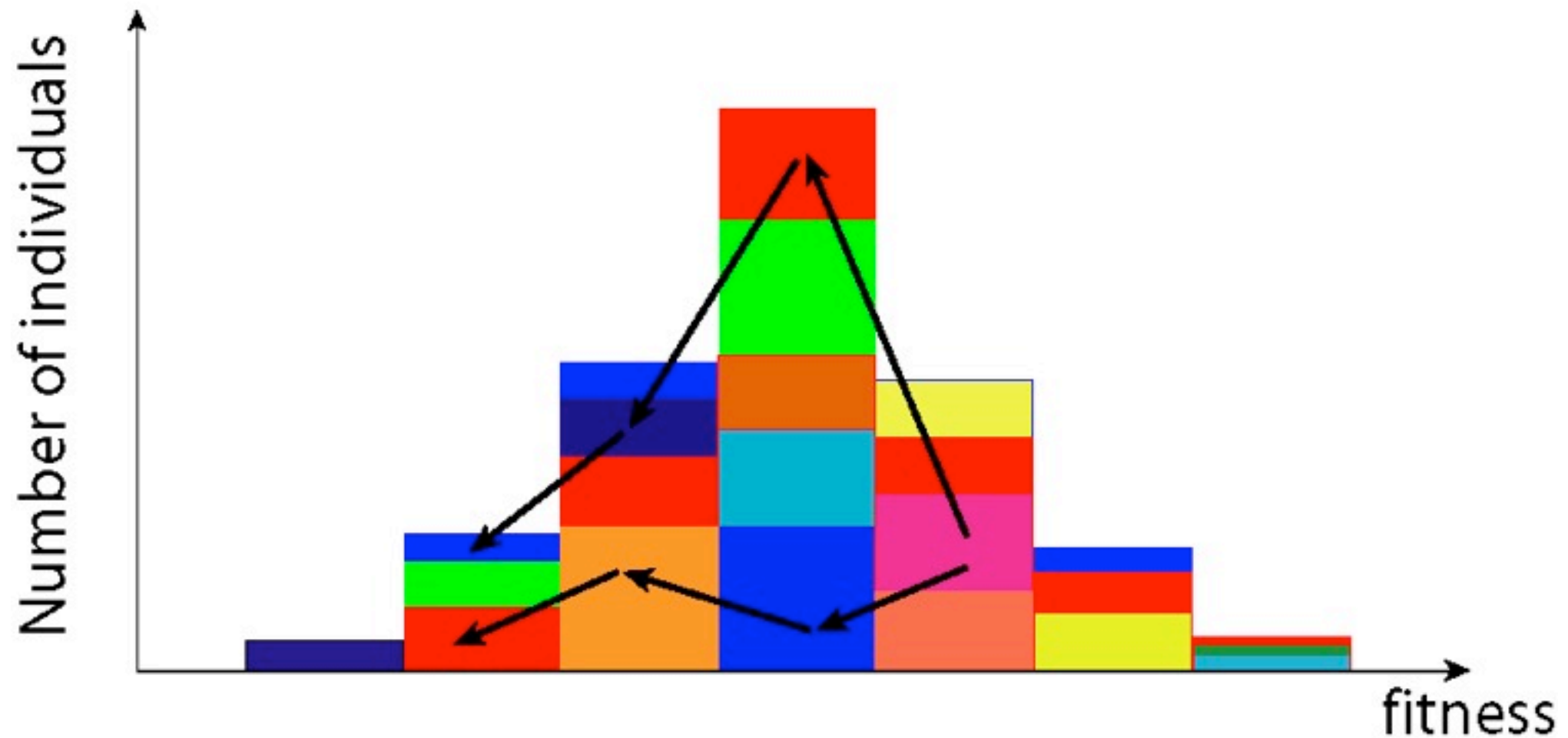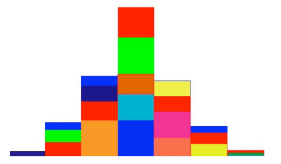Trace the ancestry of each individual through the fitness distribution

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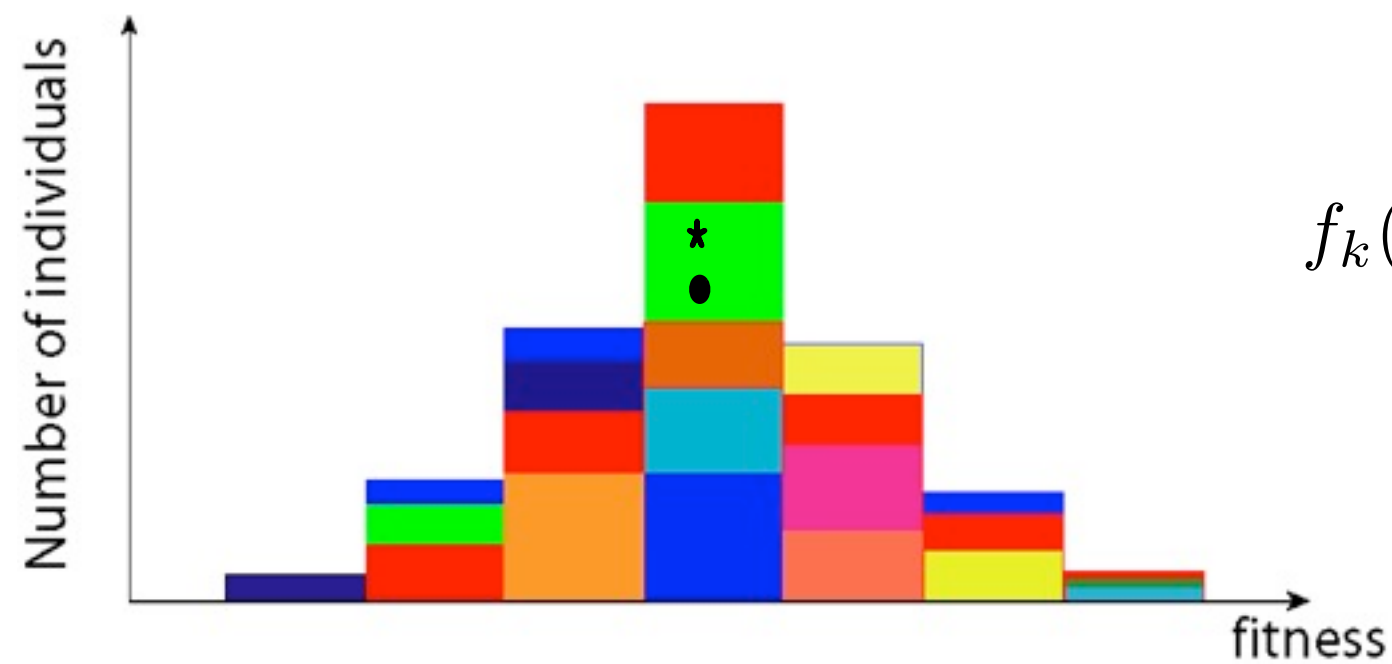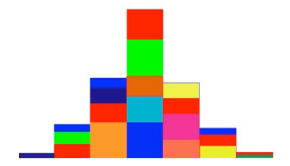
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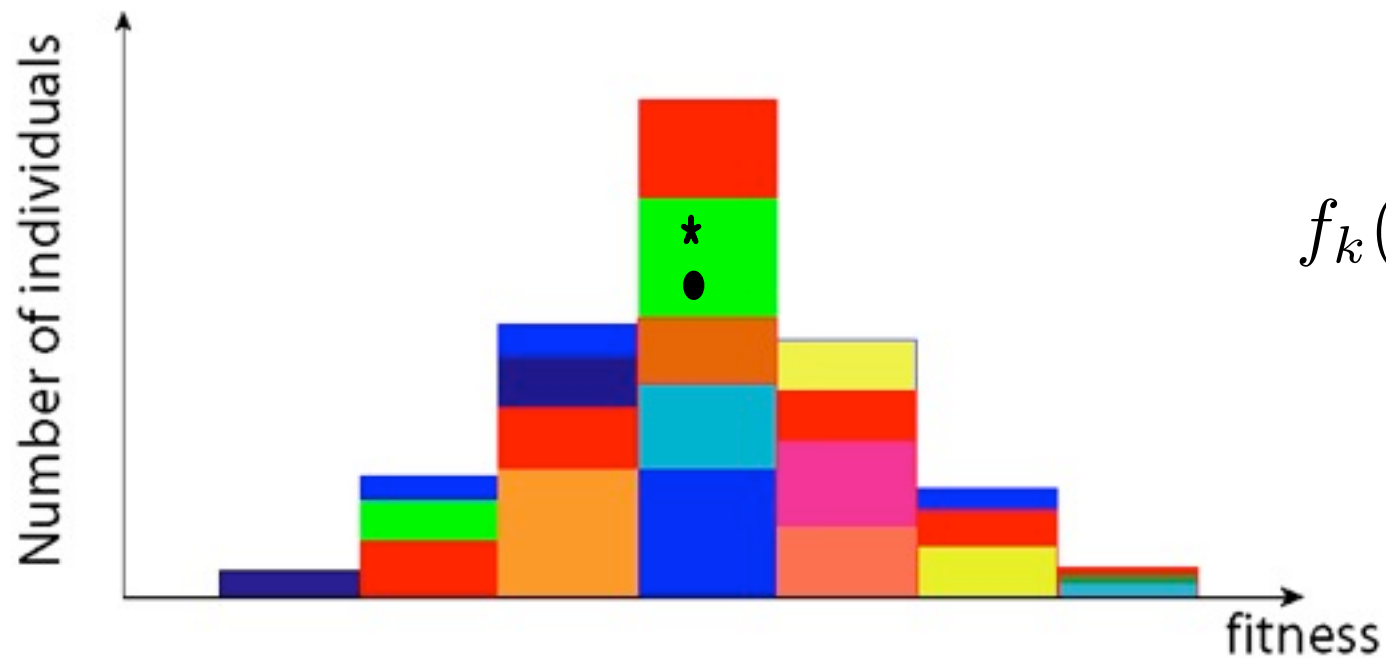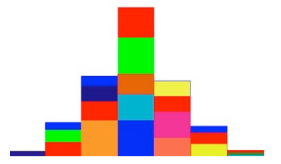
$$f_k(x)dx = \theta_k \frac{1 - e^{-2Ns_k(1-x)}}{(1 - e^{-2Ns_k})x(1-x)}dx$$

Sample 2 individuals from class k

$$f_k(x)dx = \theta_k \frac{1 - e^{-2Ns_k(1-x)}}{(1 - e^{-2Ns_k})x(1-x)}dx$$

Sample 2 individuals from class k

Coalescent probability in class k:

$$P_c^{k,k \to k} = \int \frac{x^2}{h_k^2} f_k(x)dx$$

$$f_k(x)dx = \theta_k \frac{1 - e^{-2Ns_k(1-x)}}{(1 - e^{-2Ns_k})x(1-x)}dx$$

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Coalescent probability in class k:

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Coalescent probability in class k-1:

$$P_c^{k,k \to k-1} = \int \frac{x f_{k-1}}{h_{k-1}} \frac{y G_{k-1}(y \to x, |t_2 - t_1|)}{h_{k-1}} Q_{k,k}^{k-1}(t_1, t_2) dx dy dt_1 dt_2$$

probability an individual comes from class k and lineage with frequency x

probability that a lineage in class k-1 changes in frequency from x to y in time |t_2-t_1|

joint distribution of times t_1 and t_2 - times when lineages in class k where founded by mutations
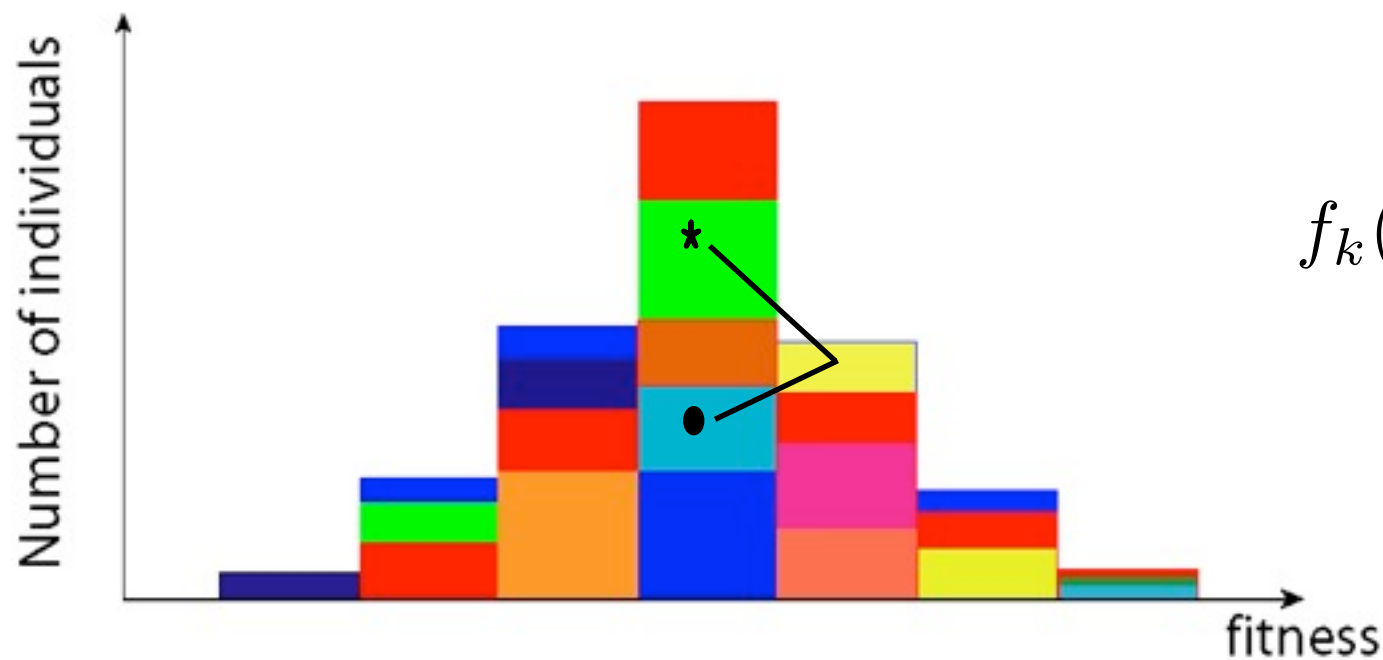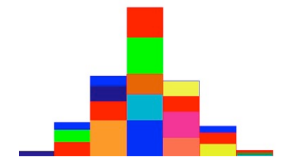
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Sample 2 individuals from class k

Coalescent probability in class k: $\qquad P_c^{k,k \to k} = \int \frac{x^2}{h_k^2} f_k(x)dx$
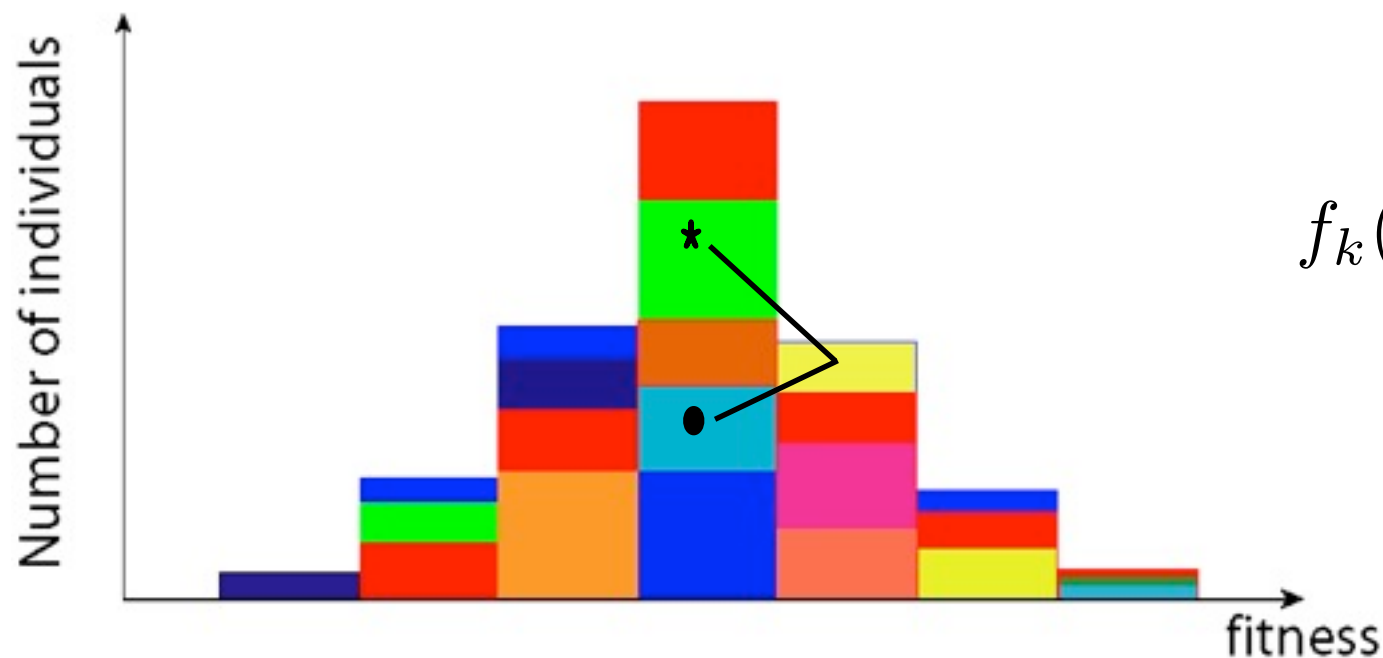
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$$P_c^{k,k \to k-2} = ...$$

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Sample 2 individuals from class k

Coalescent probability in class k:

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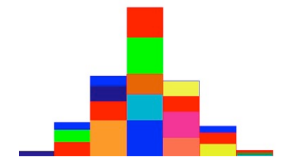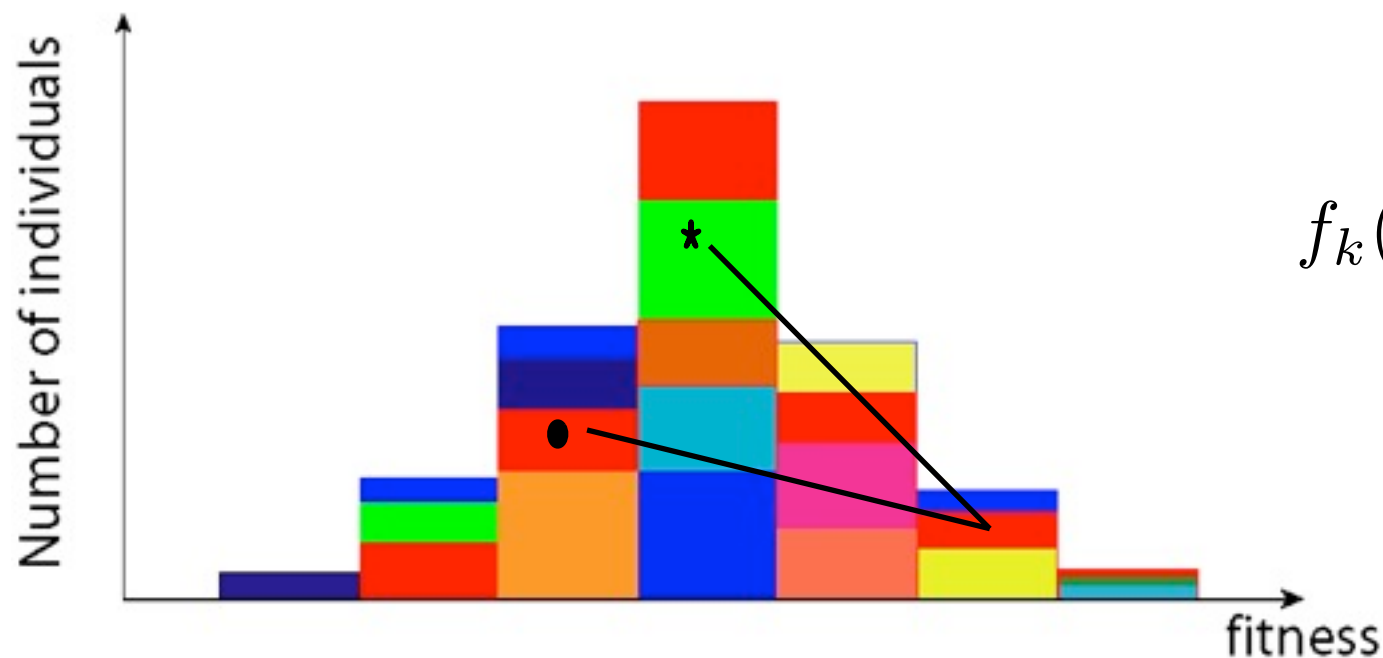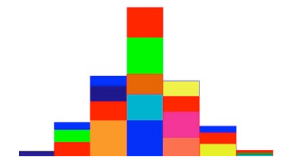
General coalescent probability in class k-ℓ:

$$P_c^{k,k+m \to k-\ell} = \int \frac{x f_{k-\ell}}{h_{k-l}} \frac{y G_{k-\ell}(y \to x, |t_2 - t_1|)}{h_{k-l}} Q_{k,k+m}^{k-\ell}(t_1, t_2) dx \, dy \, dt_1 \, dt_2$$

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joint distribution of times t_1 and t_2 - times when lineages in class k where founded by mutations

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Non-conditional approximation

$$P_c^{k,k+m\to k-\ell} = \int \frac{xf_{k-\ell}}{h_{k-l}} \frac{yG_{k-\ell}(y\to x, |t_2 - t_1|)}{h_{k-l}} Q_{k,k+m}^{k-\ell}(t_1, t_2) dx\,dy\,dt_1\,dt_2$$

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⬇ y integral is just mean y - deterministic result for the change in the frequency of the lineage

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Assume:

- **non-conditional approximation:** the times at which the two individuals moved from one fitness class to another is independent

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[generally not true because moving between fitness classes assumes no coalescence - but small correction]

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+

distribution of mutant timings:

$$Q_k^{k-\ell}(t) = Q_k^{k-1}(t) * Q_k^{k-2}(t) * \ldots Q_{k-\ell+1}^{k-\ell}(t) \quad \text{and} \quad Q_{k-\ell+1}^{k-\ell}(t) = s(k - \ell + 1) e^{-s(k-\ell+1)t}$$

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evaluate many integrals
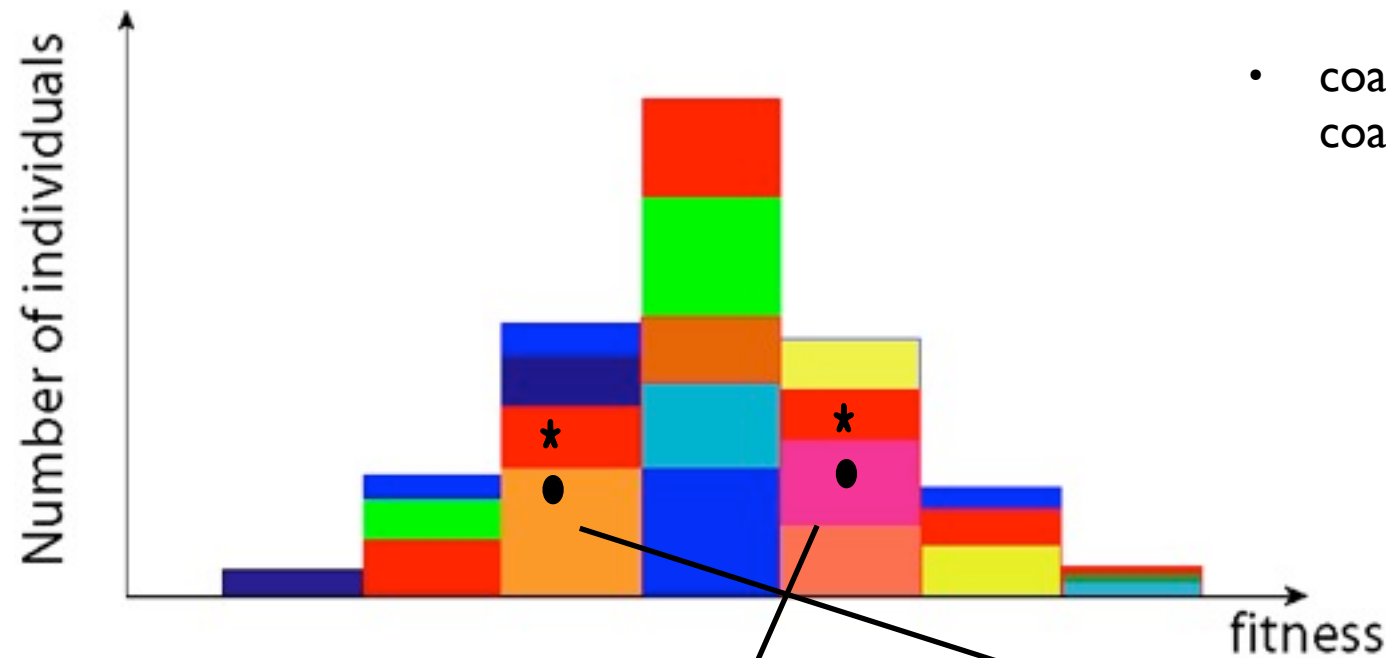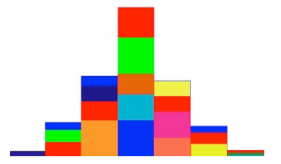
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In non-conditional approximation: $\quad P_c^{k,k+m \to k-\ell} = \dfrac{1}{N h_{k-\ell} s(k-\ell)} A_\ell^{k,m}$

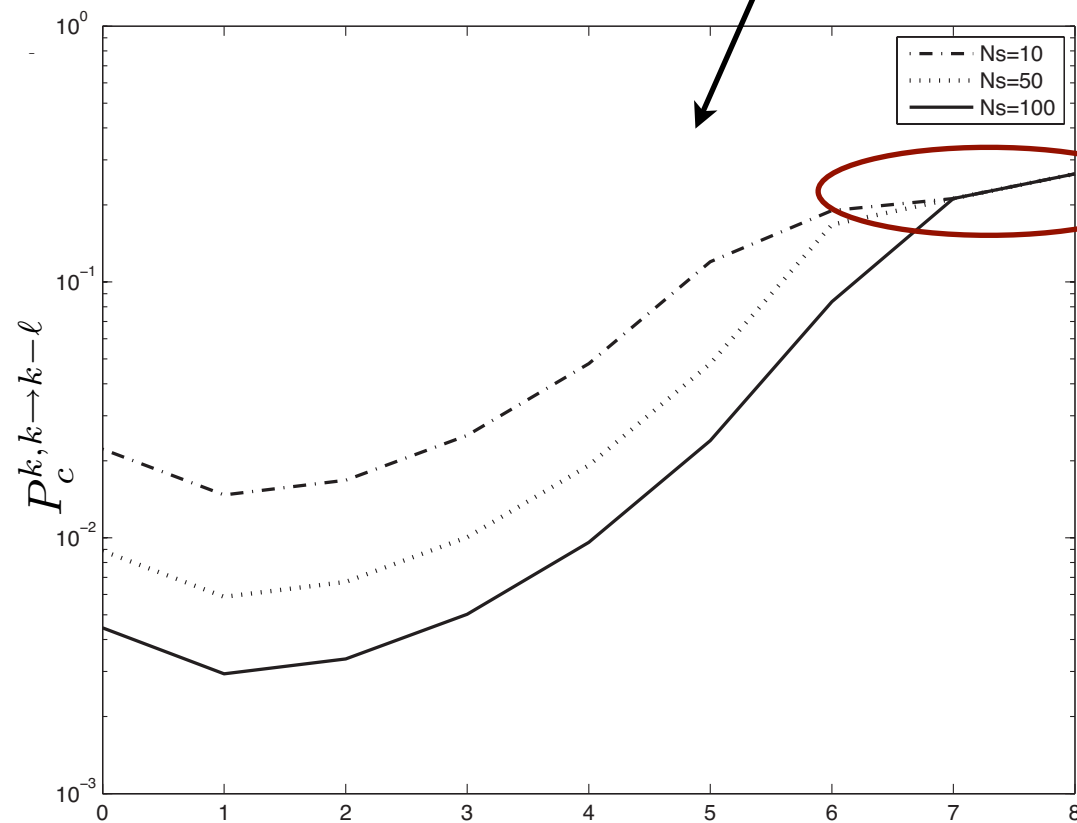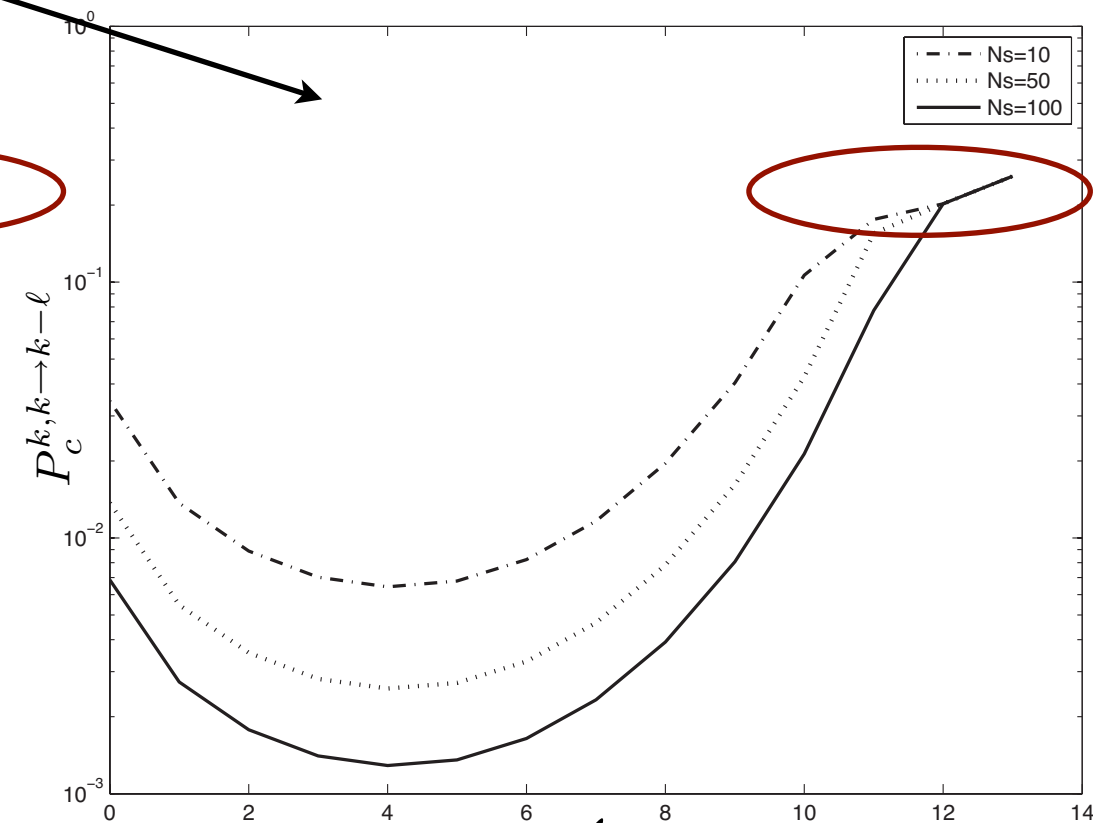Easy formula for coefficient: $\quad A_\ell^{k,m} = \dfrac{\binom{k+m}{k-\ell}\binom{k}{k-\ell}}{\binom{2k+m}{2\ell+m}}$

- coalescence probability increases for longer steptimes - coalescence in more fit classes is more likely

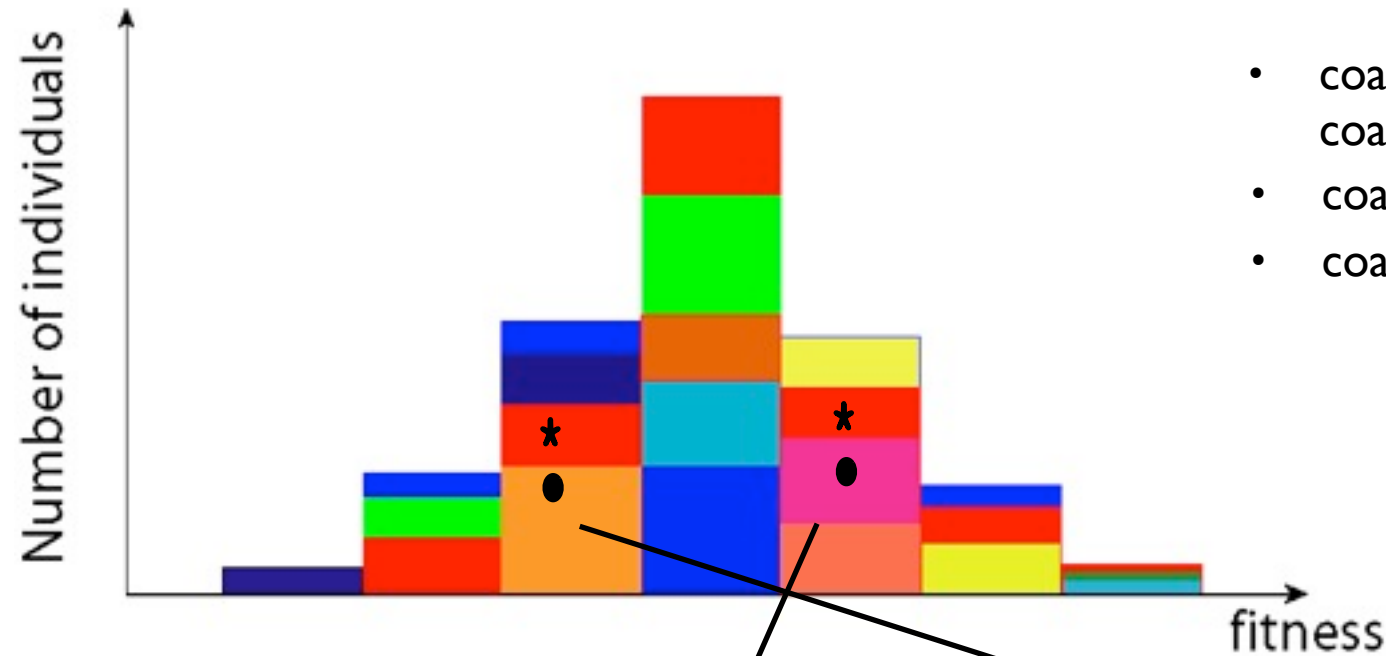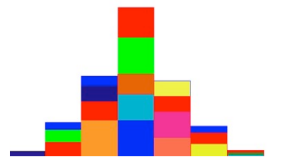$$P_c^{k,k+m \to k-\ell} = \frac{1}{Nh_{k-\ell}s(k-\ell)} A_\ell^{k,m}$$
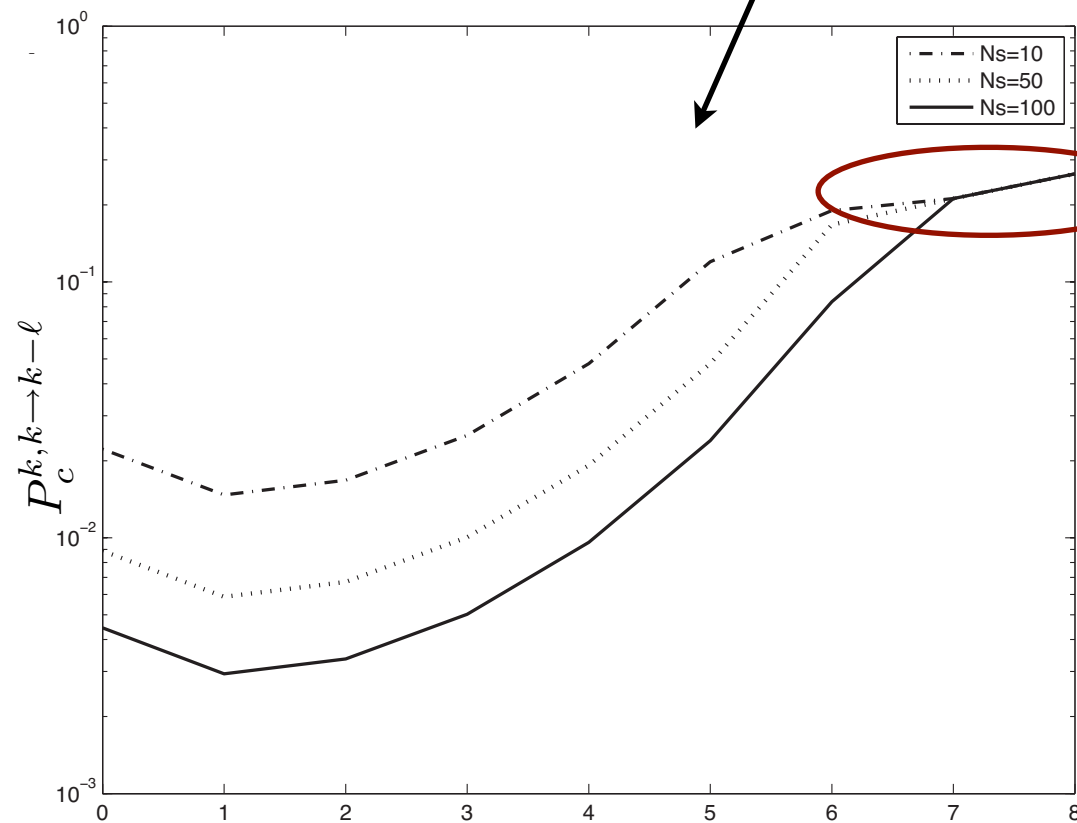
Sampled k just right of mean
(more fit than mean)
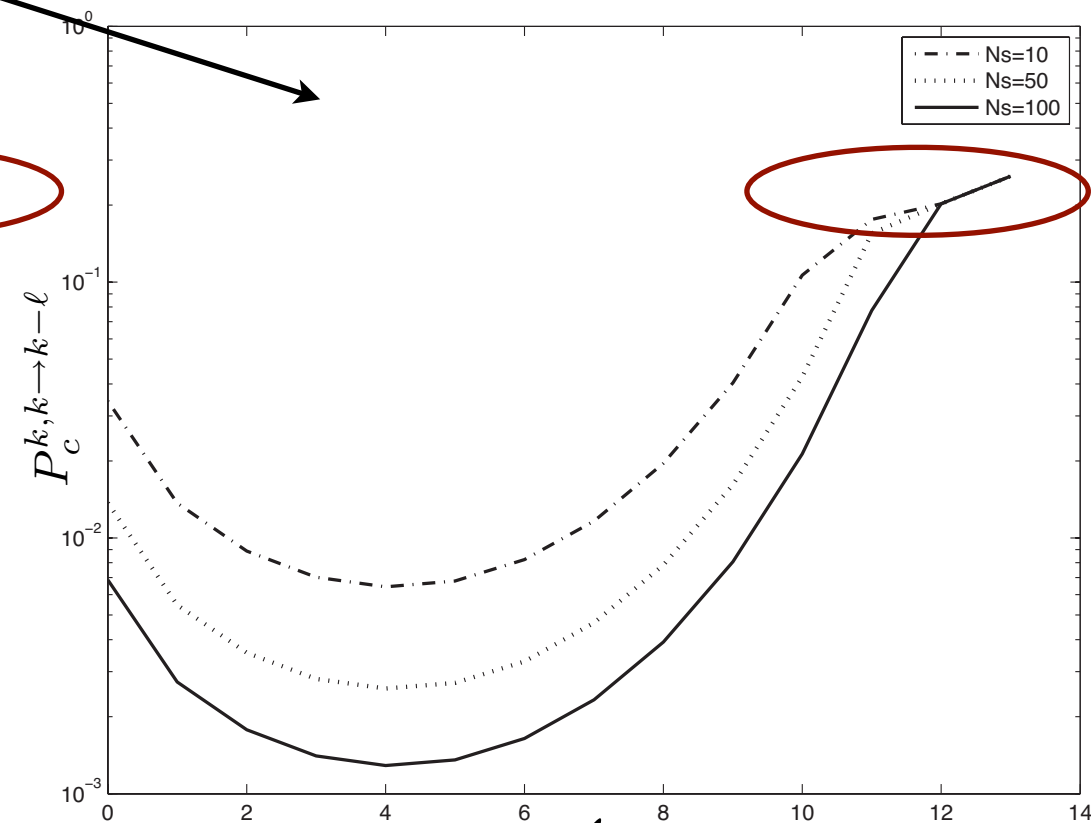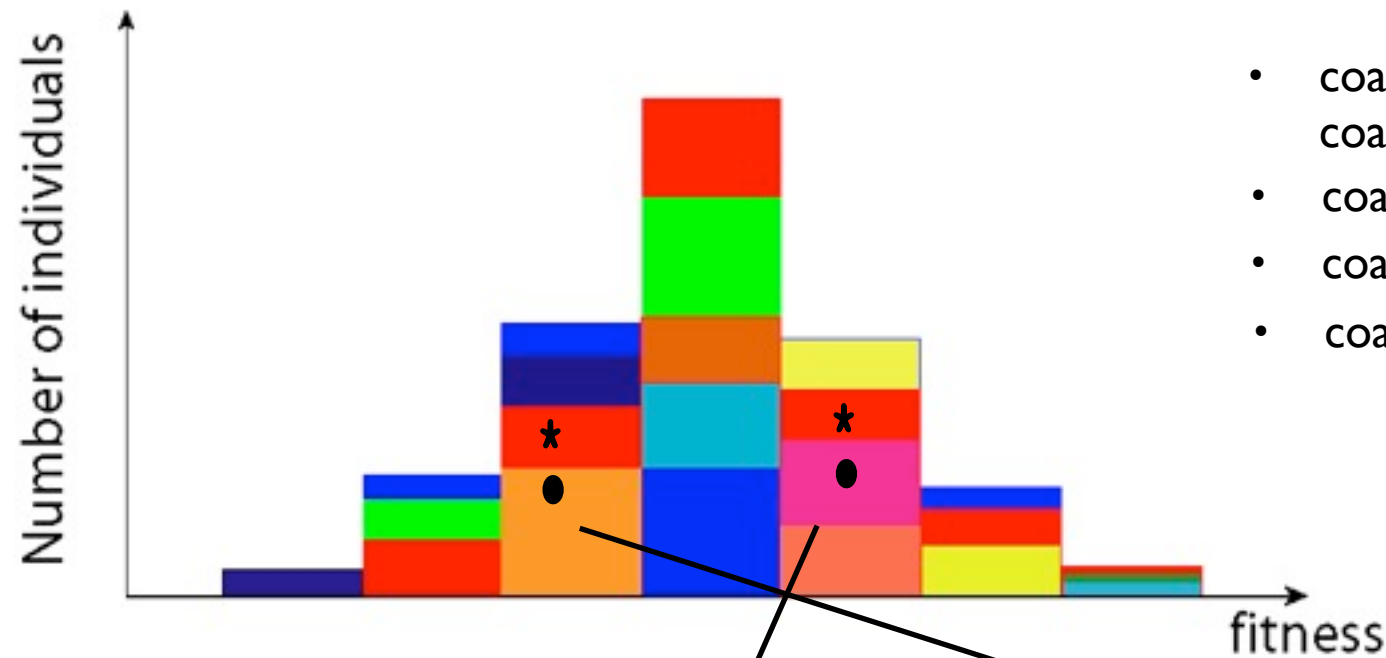
Sampled k left of mean
(less fit than mean)

- coalescence probability increases for longer steptimes - coalescence in more fit classes is more likely
- coalescence probability decreases with selection
- coalescence probability decreases with population size

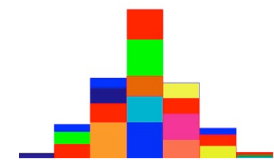$$P_c^{k,k+m \to k-\ell} = \frac{1}{Nh_{k-\ell}s(k-\ell)} A_\ell^{k,m}$$

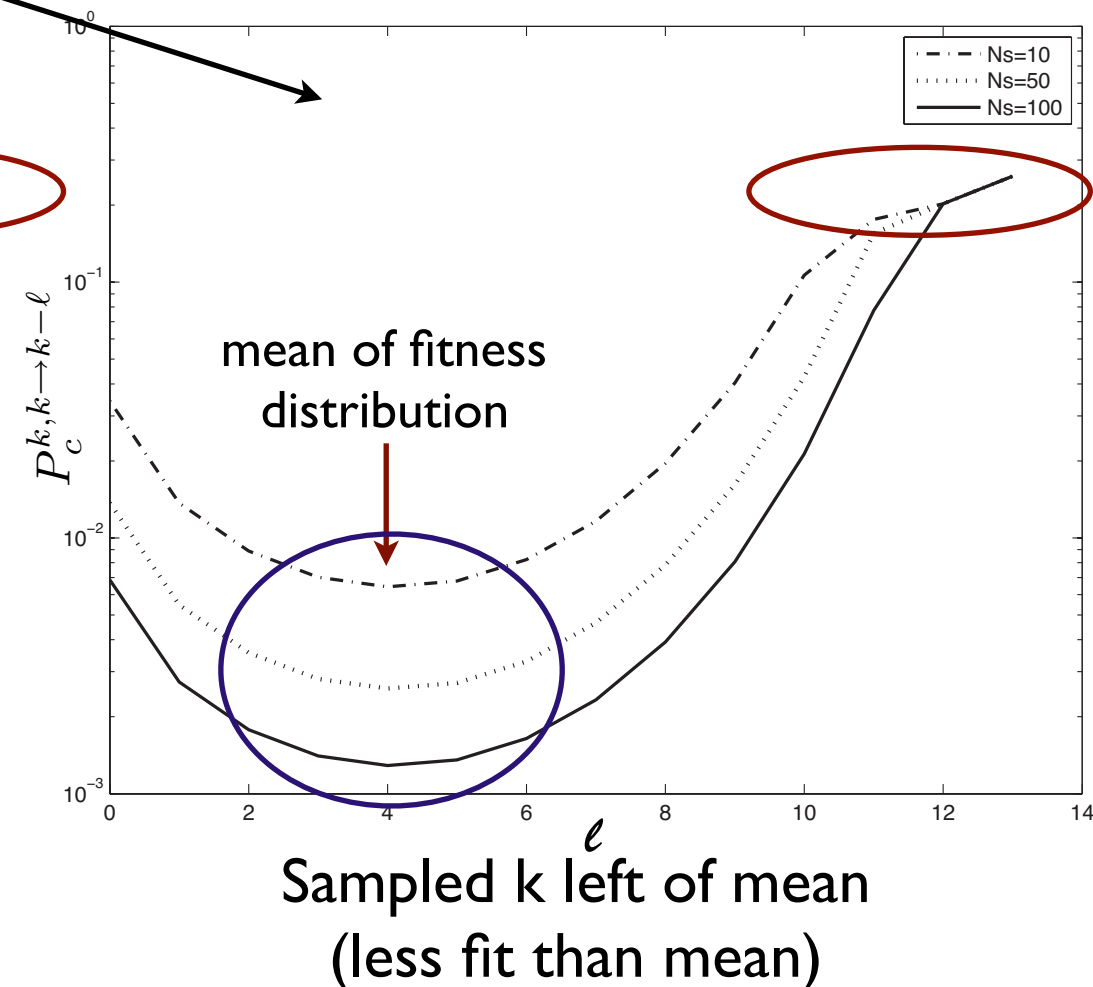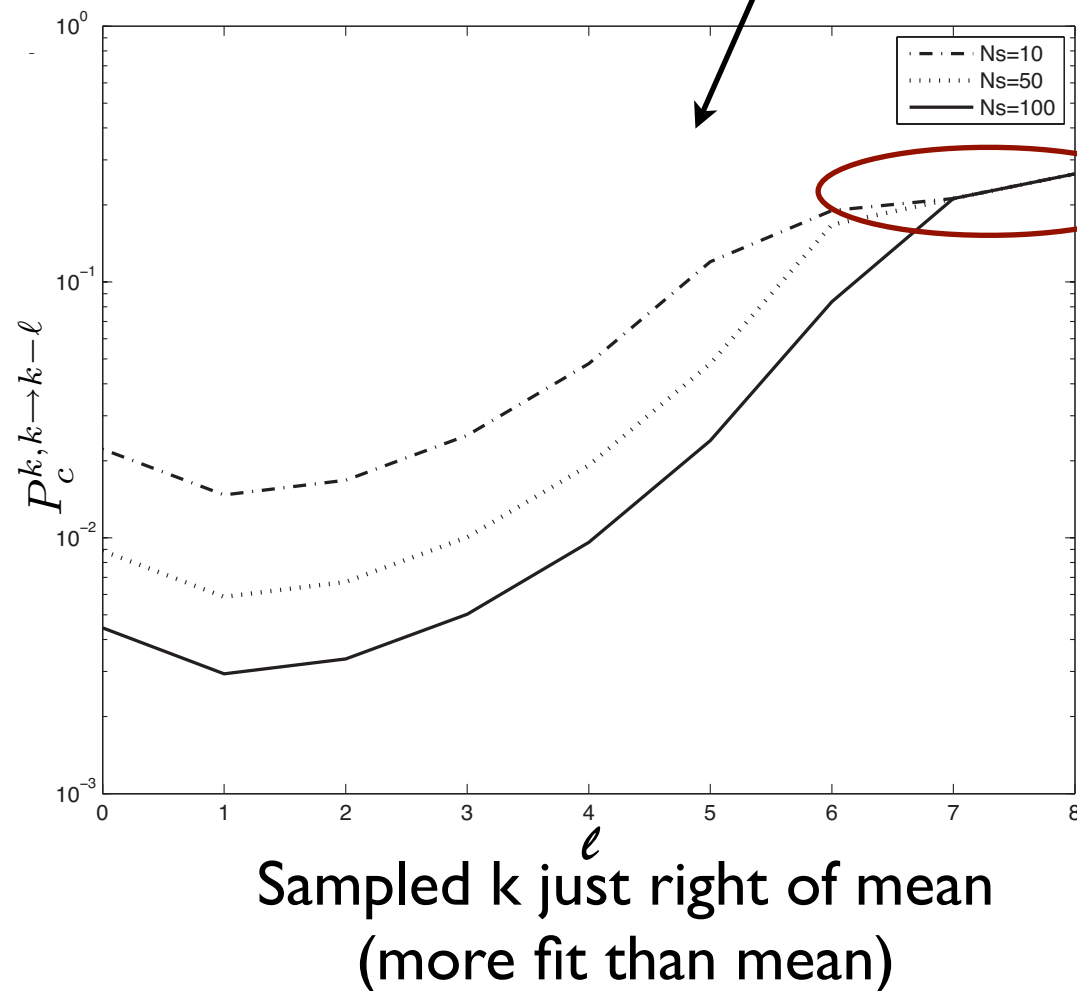**Sampled k just right of mean (more fit than mean)**

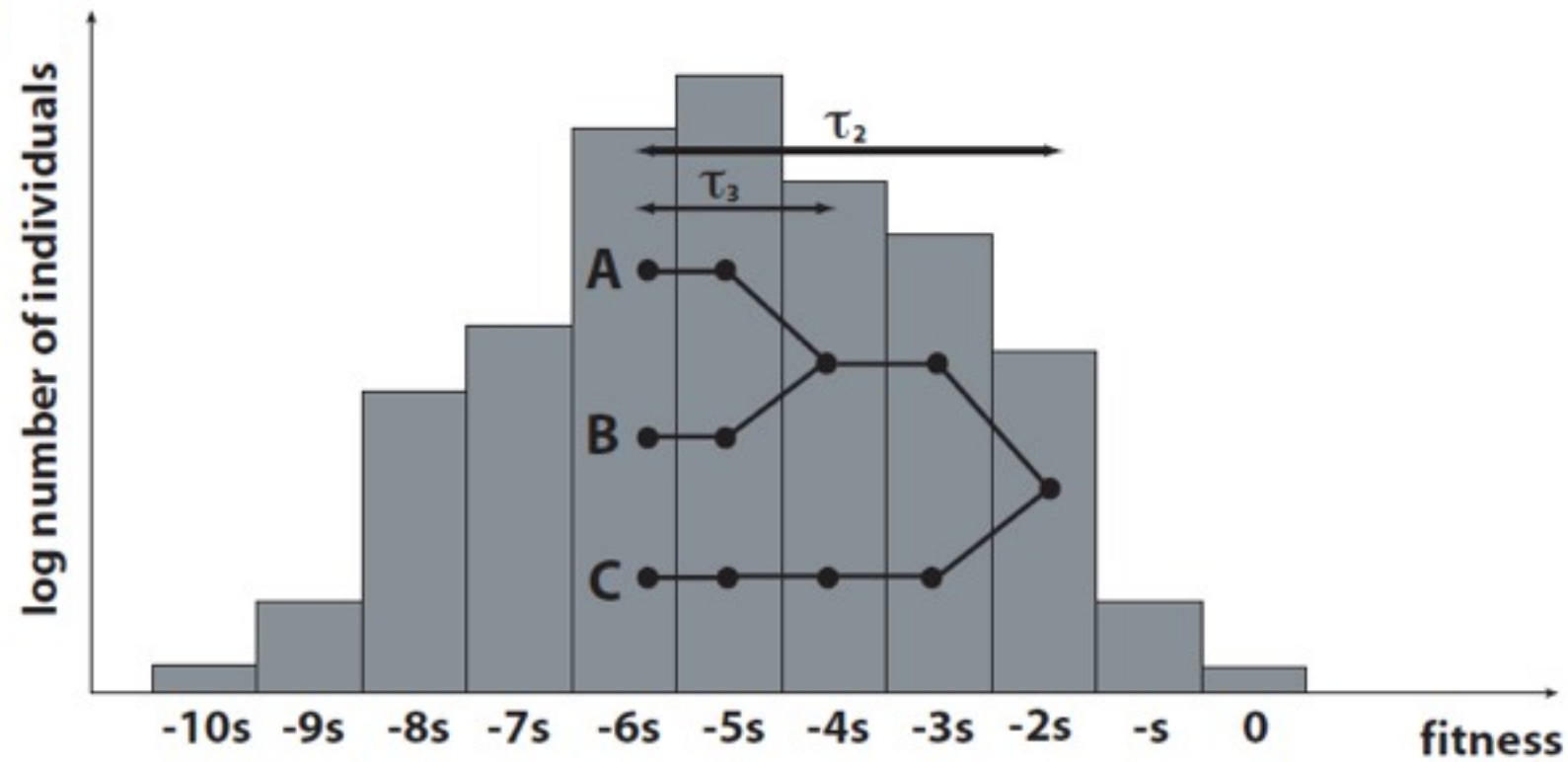**Sampled k left of mean (less fit than mean)**

- coalescence probability increases for longer steptimes - coalescence in more fit classes is more likely
- coalescence probability decreases with selection
- coalescence probability decreases with population size
- coalescence probability is less likely in most probable class

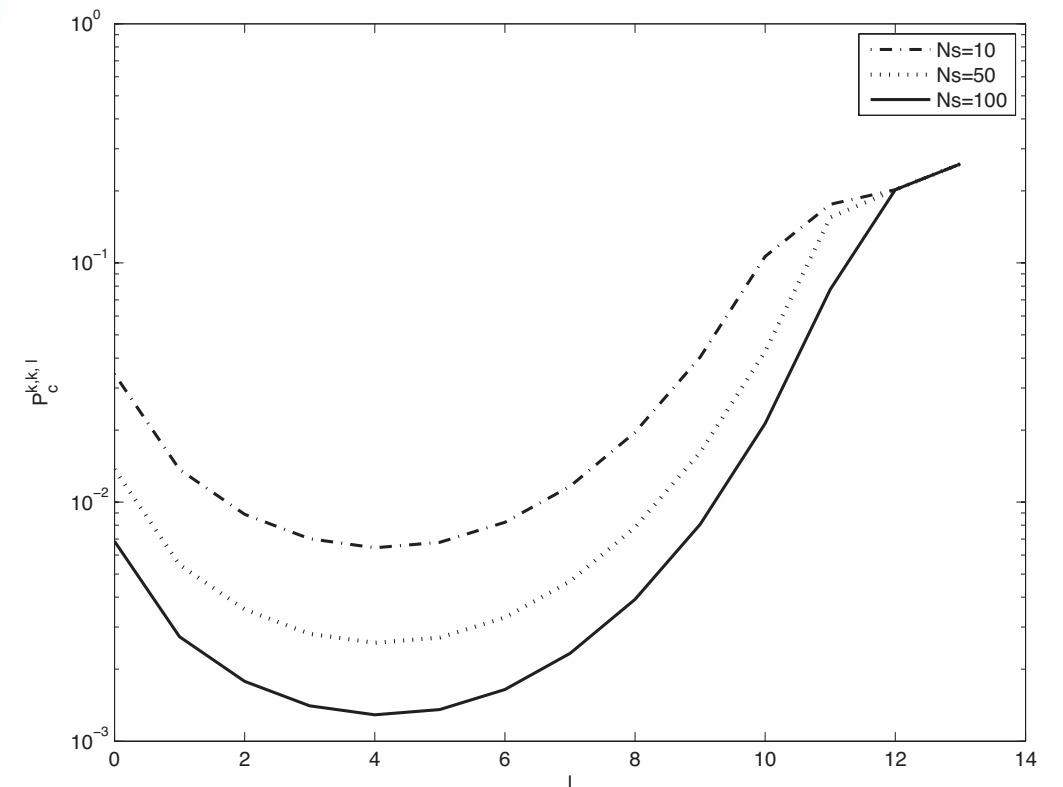$$P_c^{k,k+m \to k-\ell} = \frac{1}{Nh_{k-\ell}s(k-\ell)} A_\ell^{k,m}$$

Sampled k just right of mean
(more fit than mean)

Sampled k left of mean
(less fit than mean)

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Comparison to variable population size

$$P_c^{k,k+m\to k-\ell} = \frac{1}{Nh_{k-\ell}s(k-\ell)} A_\ell^{k,m}$$

$$P_c^{k,k+m\to k-\ell} = \frac{1}{n_{k-\ell}s_{k-\ell}} A_\ell^{k,m}$$

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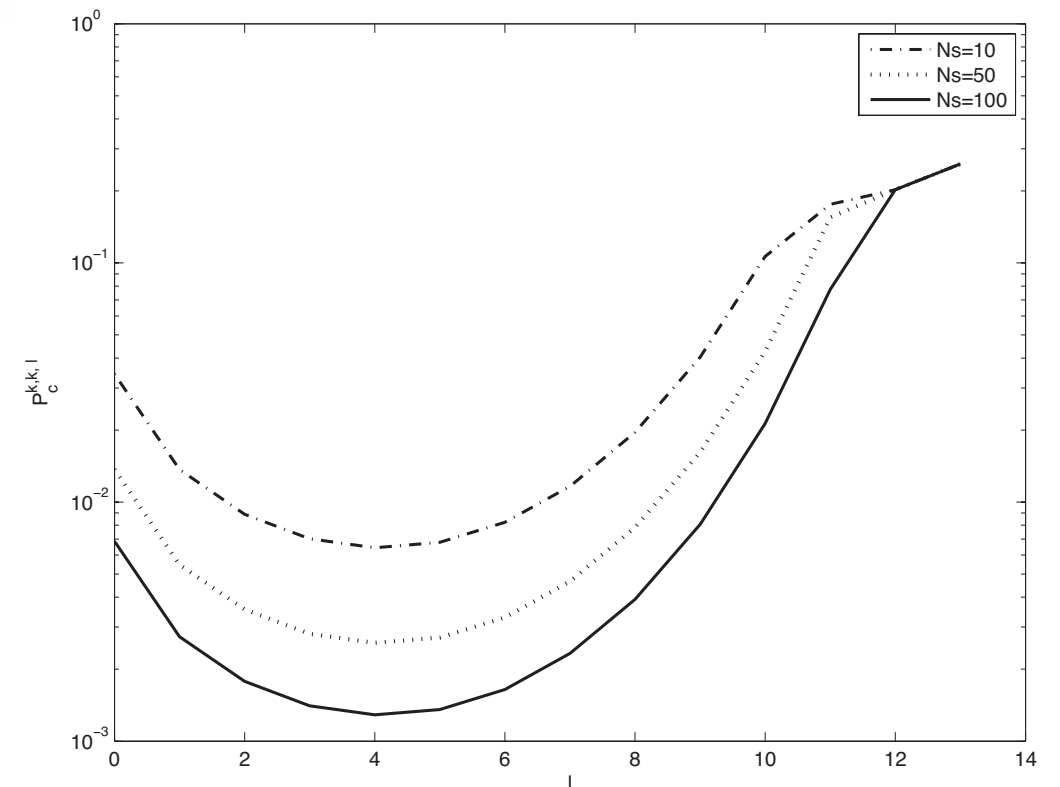# Comparison to variable population size

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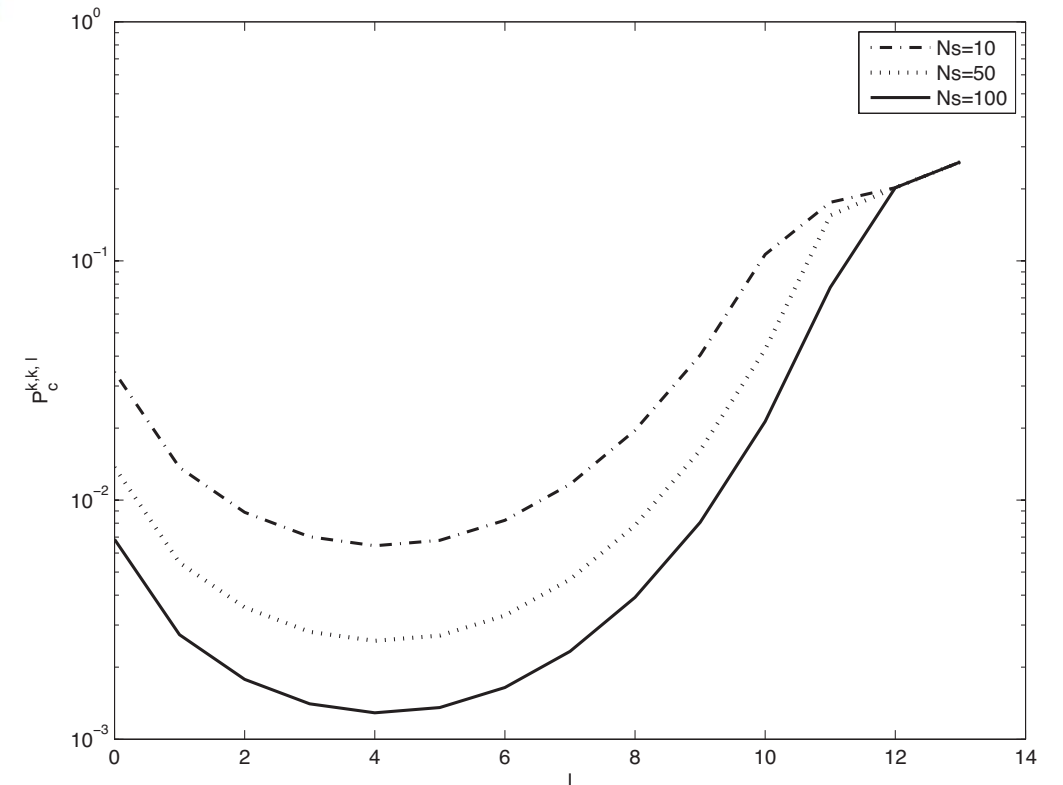
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lineage spends ~$1/s_k$ generations in each class

→ per generation coalescence probability in class k is $1/n_k$

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historically varying population size - different effective population sizes depending on initial position in fitness distribution

Comparison to variable population size

log number of individuals

τ_2

τ_3

A

B

C

-10s -9s -8s -7s -6s -5s -4s -3s -2s -s 0 fitness

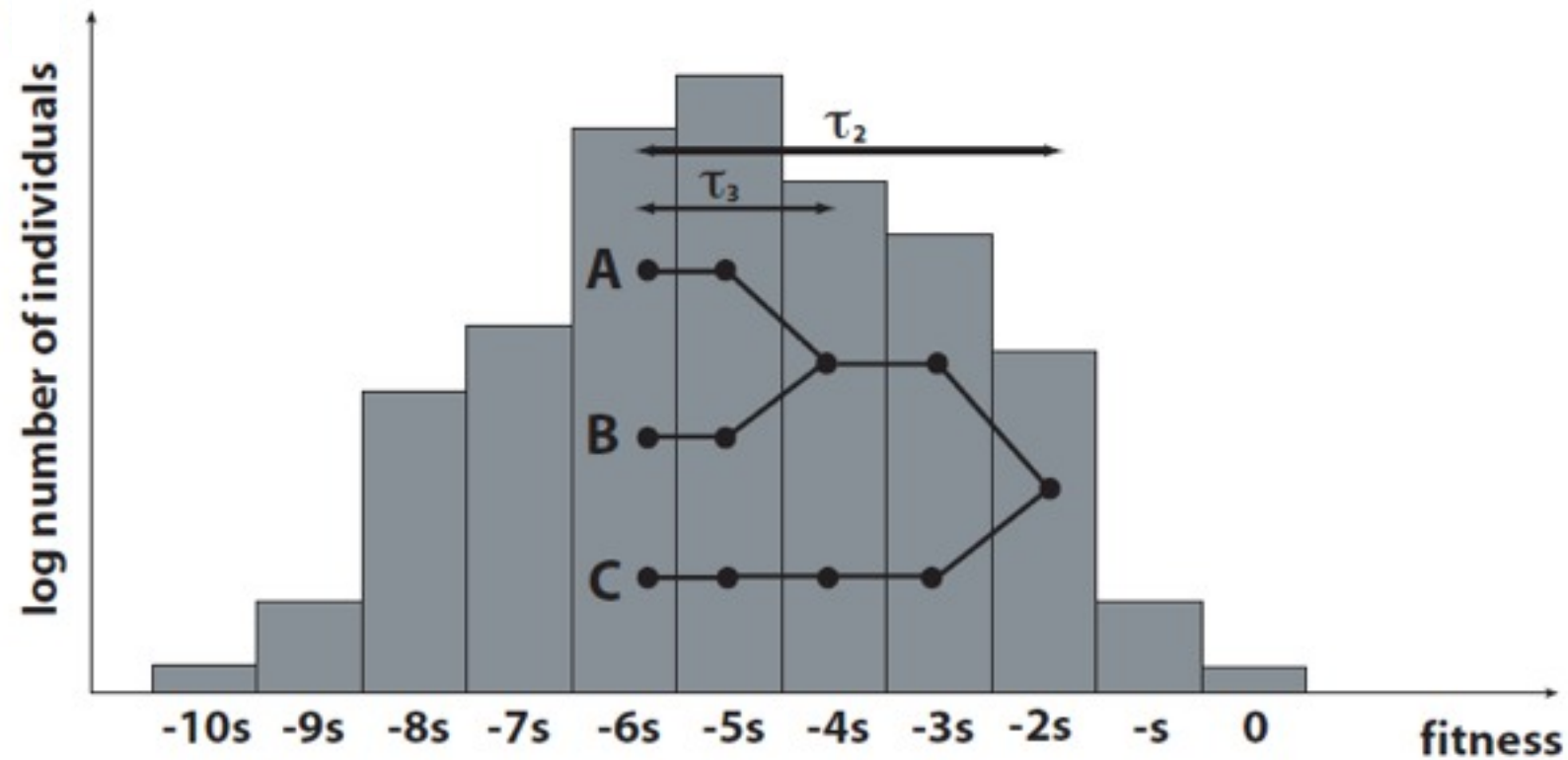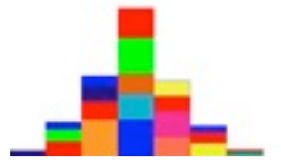$$P_c^{k,k+m \to k-\ell} = \frac{1}{Nh_{k-\ell}s(k-\ell)} A_\ell^{k,m}$$

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historically varying population size - different
effective population sizes depending on initial
position in fitness distribution

n_k

Ns=10
Ns=50
Ns=100

less fit than mean:
"weird" varying
population size

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Comparison to variable population size

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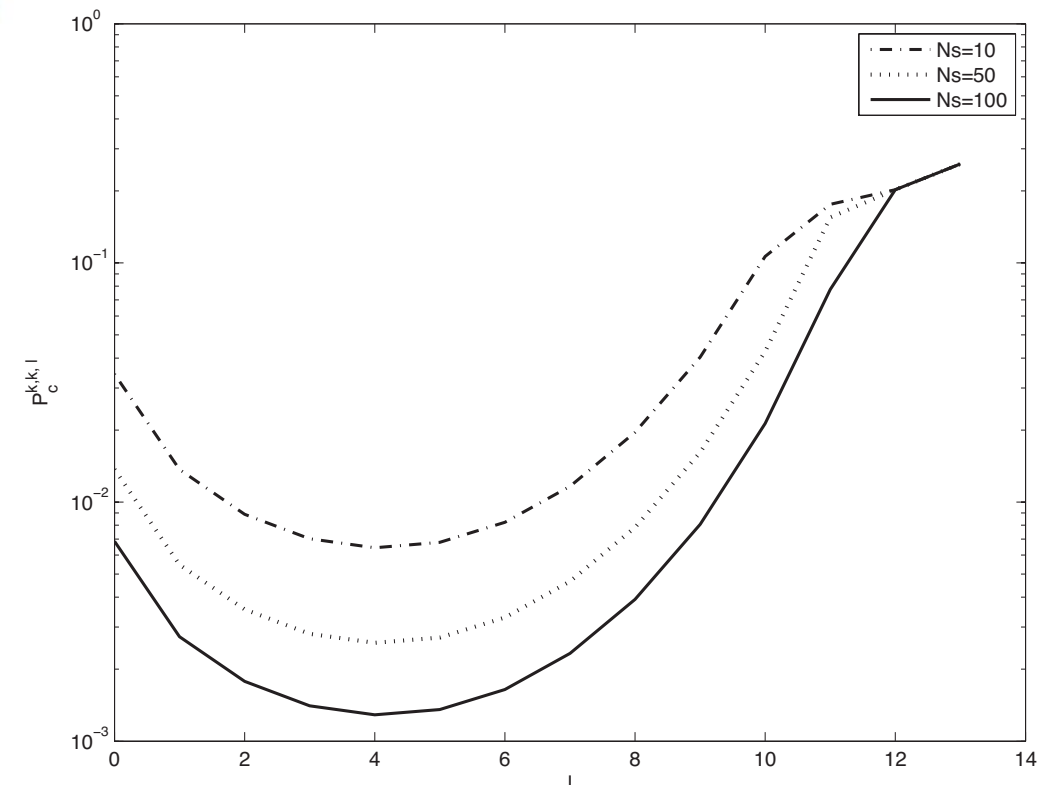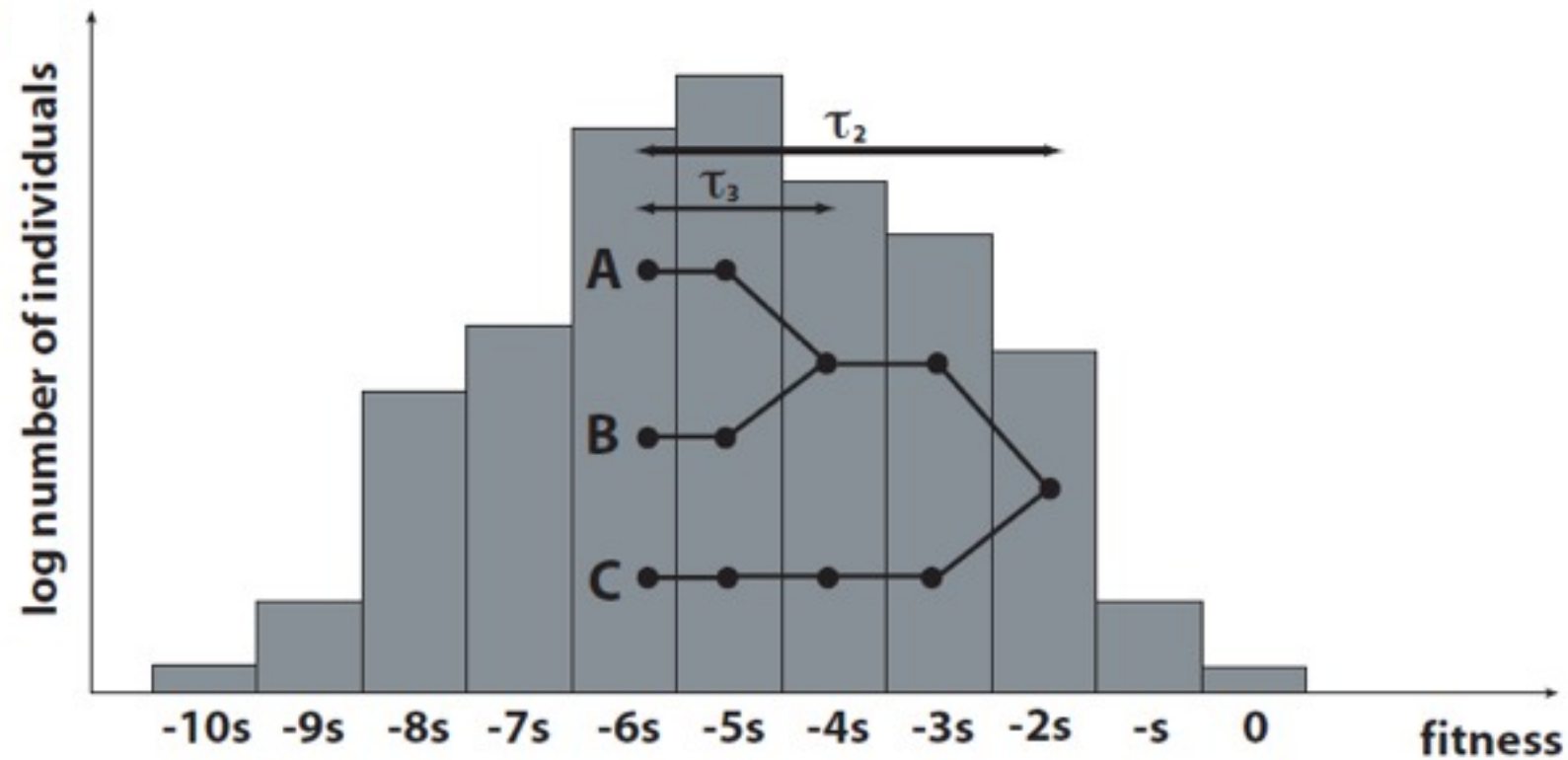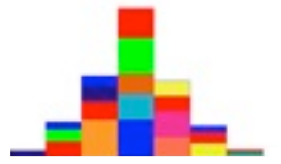
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more fit than mean: "simple" varying population size
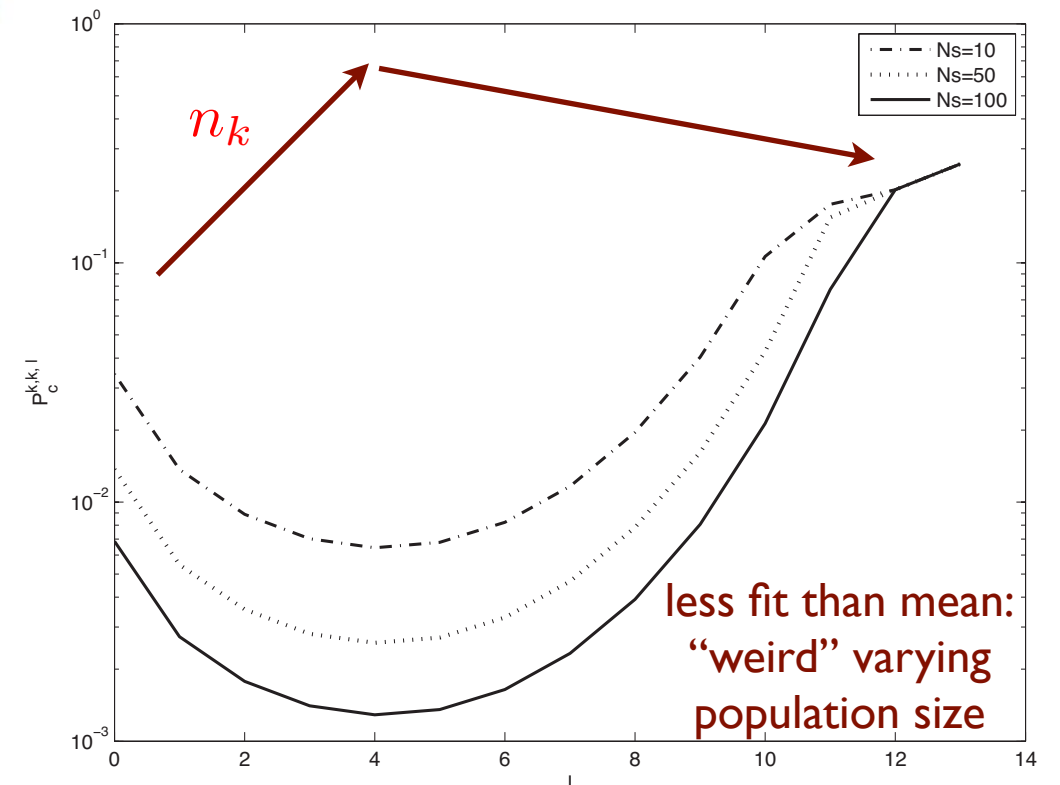
less fit than mean: "weird" varying population size

$$P_c^{k,k+m\to k-\ell} = \frac{1}{N h_{k-\ell} s(k-\ell)} A_\ell^{k,m}$$

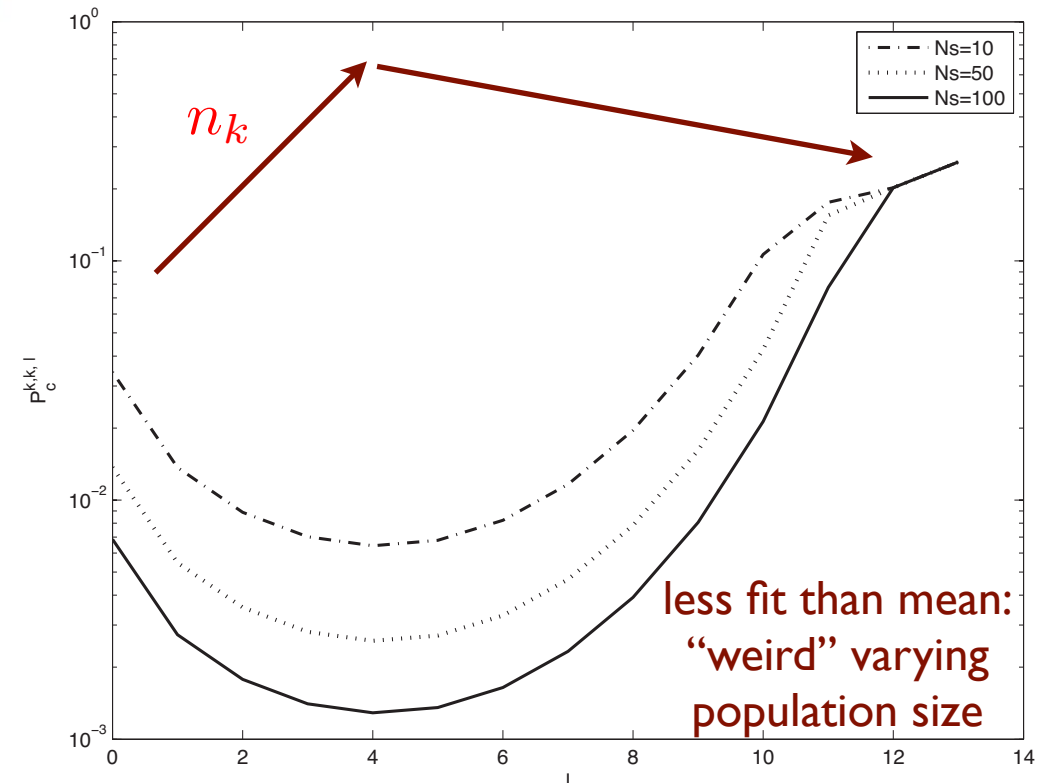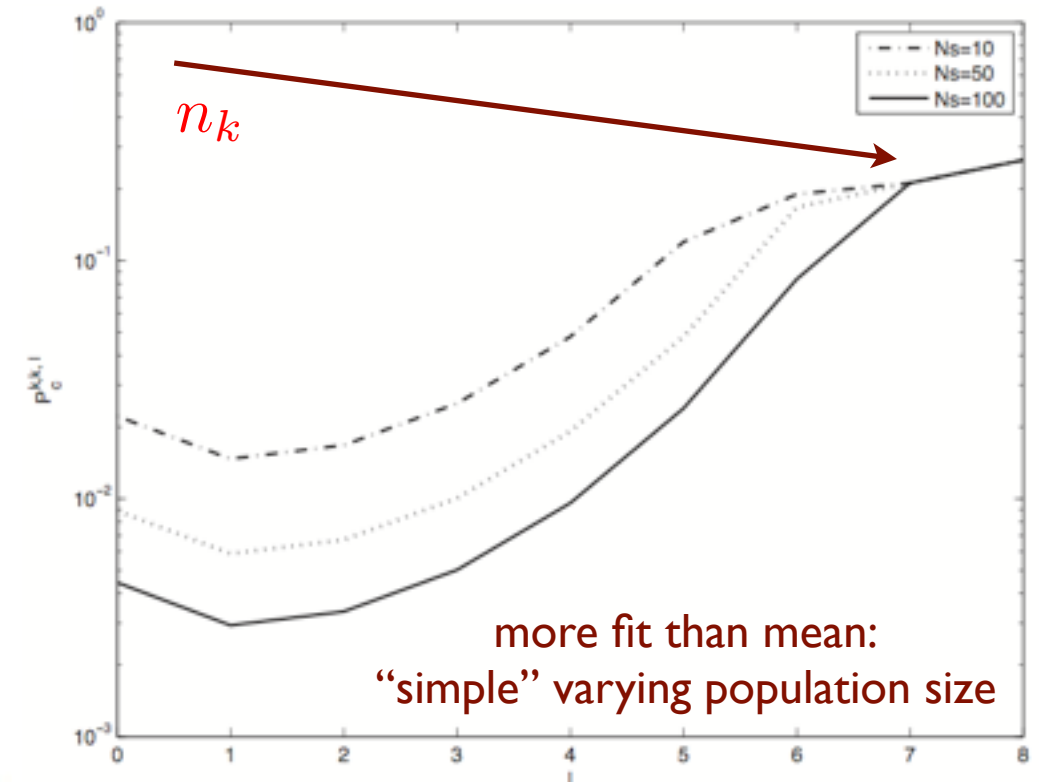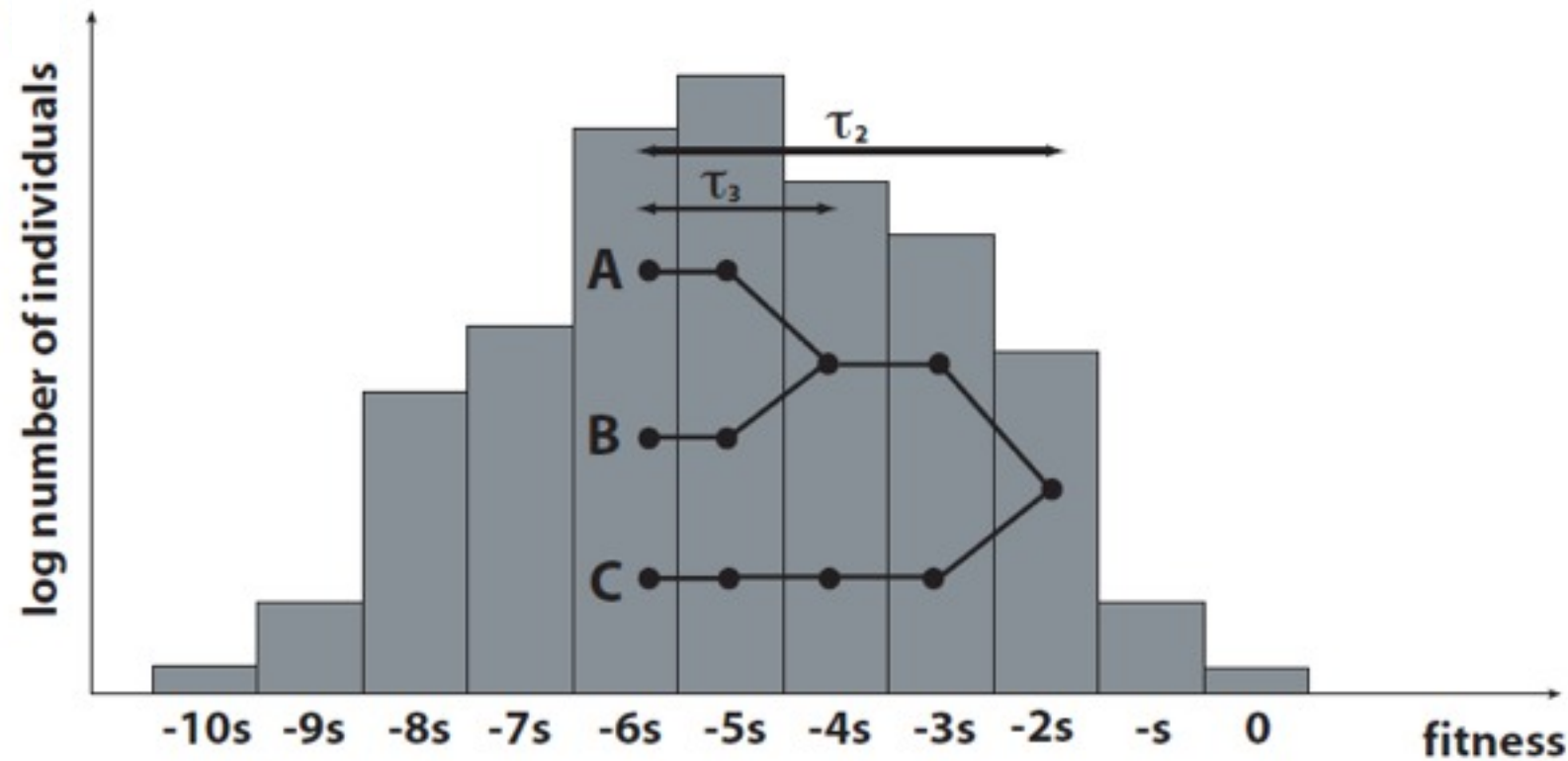$$P_c^{k,k+m\to k-\ell} = \frac{1}{n_{k-\ell} s_{k-\ell}} A_\ell^{k,m}$$

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historically varying population size - different effective population sizes depending on initial position in fitness distribution

n_k

more fit than mean: "simple" varying population size

n_k

less fit than mean: "weird" varying population size

→ really strange variation in population size for two individuals from different classes

π_d – per site heterozygosity at deleterious sites – distance in number of mutations between individuals

$$P(\pi_{AB} = 4) = P(\tau_3 = 2) = P_c^{k,k\to k-2}(1 - P_c^{k,k\to k-1})(1 - P_c^{k,k\to k})$$

coalesced k-2 classes ago

did not coalesce k-1 class ago

did not coalesce in k class

π_d – per site heterozygosity at deleterious sites – distance in number of mutations between individuals

$$P(\pi_{AB} = 4) = P(\tau_3 = 2) = P_c^{k,k \to k-2}(1 - P_c^{k,k \to k-1})(1 - P_c^{k,k \to k})$$

coalesced k-2 classes ago

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did not coalesce in k class

Analogous expressions apply for k, k', k"

$$P(\tau = \ell) = P(\pi_d = 2\ell + m) = P_c^{k,k+m \to k-\ell} \prod_{j=0}^{\ell-1}(1 - P_c^{k,k+m \to k-j})$$

Average over distribution of k, k', k":

$$\rho(\pi_d) = \sum_{\ell=0}^{\pi_d/2} \sum_{k=0}^{\infty} H(k, k+m = k + \pi_d - 2\ell) P_k^{k+m=k+\pi_d-2\ell}(\tau = \ell)$$

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Left figure axes: $<\pi_d>$ vs U_d/s, legend: Ns=5, Ns=10, Ns=100, Ns=500

Right figure: MCP: $< \pi_d > = 2U_d/s$, axes $<\pi_d>$ vs Ns, legend: $U_d/s=2$, $U_d/s=6$, $U_d/s=10$, $U_d/s=14$

- large selection - weak N dependence
- mean coalescence path approximation for large N and large U_d/s (weaker selection) :
    - large number of lineages in each fitness class - coalescence events unlikely
    - all coalescence happens in zeroth class (like in EPS)
    - coalescence time is dominated by time it takes to get to zeroth class (unlike EPS)
- for small N - larger probability to coalesce in bulk - smaller $<\pi_d>$

$k = 0$

$k = U_d/s$

$$\rho(\pi_d) = \sum_{\ell=0}^{\pi_d/2} \sum_{k=0}^{\infty} H(k, k+m = k + \pi_d - 2\ell) P_k^{k+m=k+\pi_d-2\ell}(\tau = \ell)$$

$u_d/s = 6$

— Theory

----- Simulation (neglect least-loaded)

········ Simulation (include least-loaded)

π_d − distance in number of mutations between individuals

MCP:$< \pi_d > = 2U_d/s$

$u_d/s = 8$

— Theory

----- Simulation (neglect least-loaded)

········ Simulation (include least-loaded)

$\pi_{\blacklozenge\blacklozenge} = 2$

$\pi_{\blacklozenge\blacklozenge} = 4$

FGS: $\rho(\pi_d = r) = \sum_{k=r-k-m} H(k, k+m)$

$\qquad\qquad = e^{-2U_d/s} \dfrac{1}{r!} \left(\dfrac{2U_d}{s}\right)^r$

• need to translate step-times into real times to get the distribution of actual coalescence time between two randomly chosen individuals $\Psi(t)$

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$$\Psi(t) = \sum_{k=0}^{\infty} \sum_{m=0}^{\infty} \sum_{\ell=0}^{k} \Psi(t|, k+m, \ell) \phi_k^{k+m}(\tau = \ell) H(k, k+m)$$

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distribution of actual coalescence
time conditional on them coalescing

probability to coalesce
ℓ steps ago

average over class
frequencies

Effective time to real times and neutral diversity

- need to translate step-times into real times to get the distribution of **actual coalescence time** between two randomly chosen individuals $\Psi(t)$

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$\left(\begin{array}{c} \text{longer of the actual} \\ \text{mutation times+time for} \\ \text{coalescence in class k-}\ell \end{array} \right) \sim$ distribution of actual coalescence time conditional on them coalescing   probability to coalesce ℓ steps ago   average over class frequencies

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$\begin{pmatrix} \text{longer of the actual} \\ \text{mutation times+time for} \\ \text{coalescence in class k-}\ell \end{pmatrix} \sim$ distribution of actual coalescence   probability to coalesce   average over class
time conditional on them coalescing ℓ steps ago   frequencies

• as in the traditional coalescent - neutral mutations distributed according to a Poisson process where time is drawn from distribution of coalescence times (branch lengths)

$$\rho(\pi_n) = \int \frac{[2U_n t]^{\pi_n}}{\pi_n!} e^{-2U_n t} \Psi(t) dt$$

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distribution of actual coalescence time conditional on them coalescing

probability to coalesce ℓ steps ago

average over class frequencies

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$$\frac{U_d}{s} = 6$$

$$\frac{U_d}{s} = 8$$

Effective time to real times and neutral diversity

- need to translate step-times into real times to get the distribution of actual coalescence time between two randomly chosen individuals $\Psi(t)$

$$\Psi(t) = \sum_{k=0}^{\infty} \sum_{m=0}^{\infty} \sum_{\ell=0}^{k} \Psi(t|,k+m,\ell) \phi_k^{k+m}(\tau = \ell) H(k,k+m)$$

$\left(\begin{array}{c} \text{longer of the actual} \\ \text{mutation times+time for} \\ \text{coalescence in class k-}\ell \end{array} \right) \sim$ distribution of actual coalescence time conditional on them coalescing   probability to coalesce ℓ steps ago   average over class frequencies
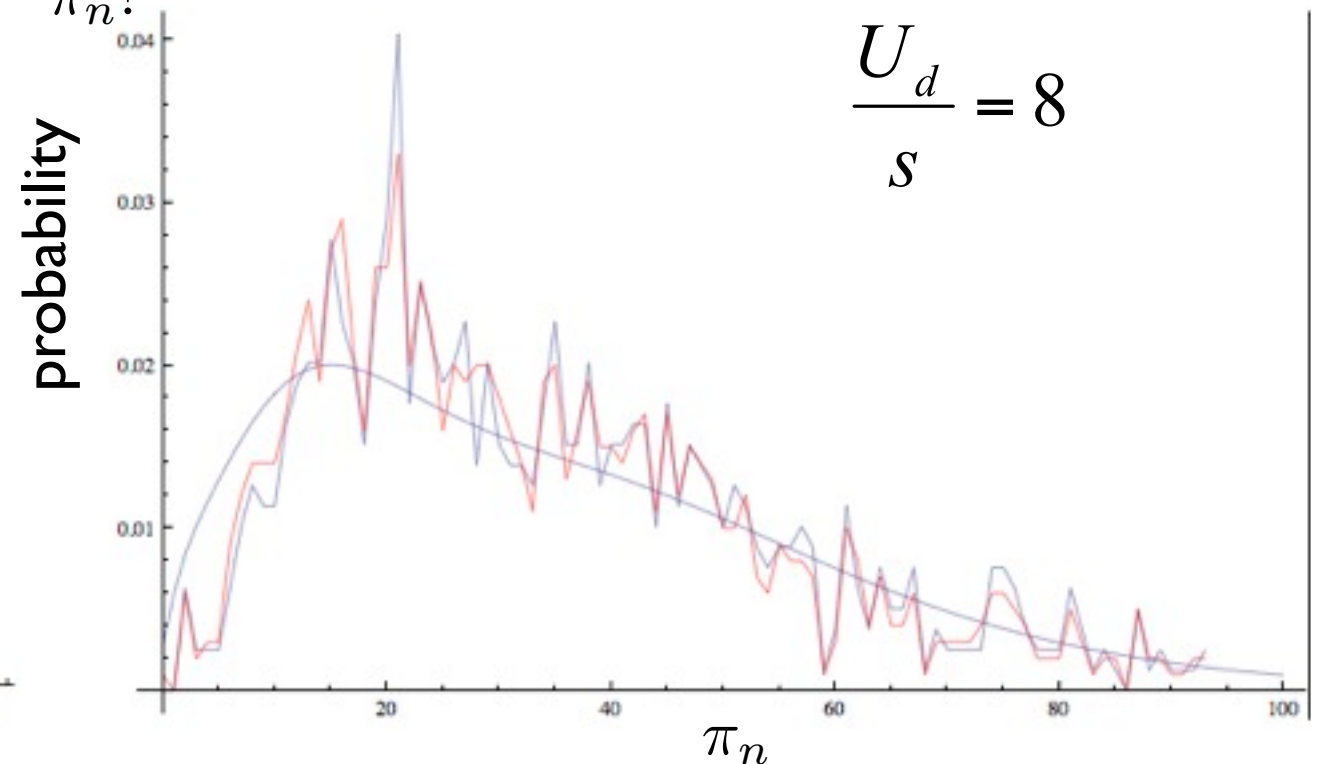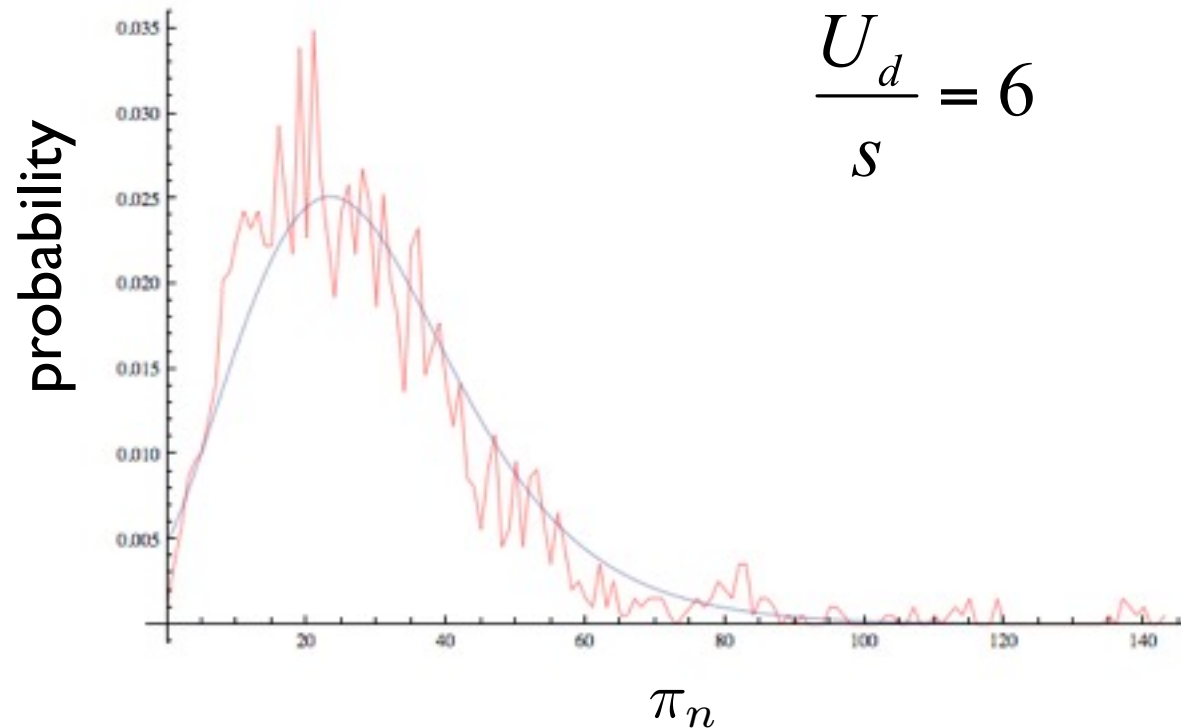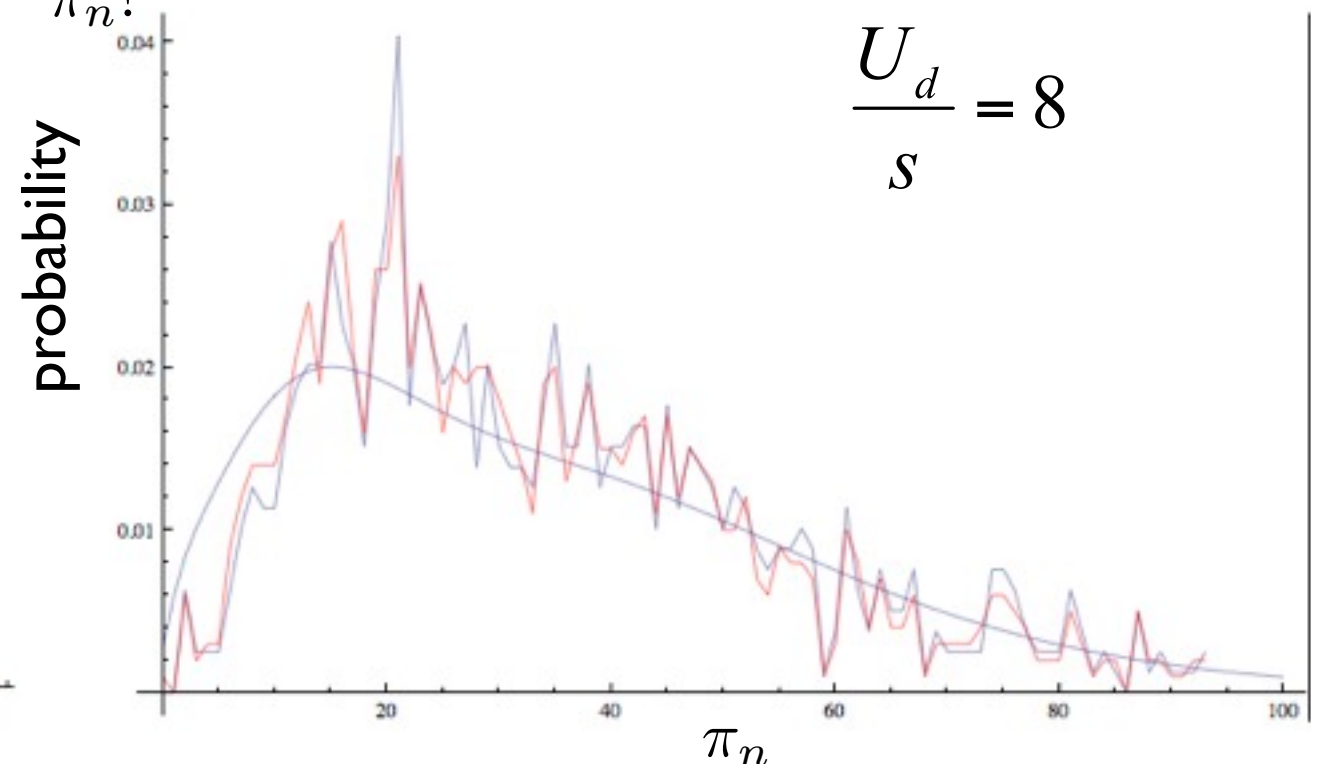
- as in the traditional coalescent - neutral mutations distributed according to a Poisson process where time is drawn from distribution of coalescence times (branch lengths)

$$\rho(\pi_n) = \int \frac{[2U_n t]^{\pi_n}}{\pi_n!} e^{-2U_n t} \Psi(t) dt$$

$$\frac{U_d}{s} = 6$$

$$\frac{U_d}{s} = 8$$

- non-zero peak in distribution - unlikely for two individuals to be extremely closely related - from peak in fitness distribution
- non-exponential distribution - difference from neutral case

We can now calculate the expected distribution of any statistic describing variation when negative selection is operating.

We know a bit more about what we're looking for.

Summary

- **expansion of coalescence framework to negative selection**
  - idea: effectively see how individuals **move** through fitness distribution
  - do not **follow** individual ancestry
  - count time is **steptimes**
- **the genetic variability cannot be mimicked by effective population size**
- **approach works for weak and strong selection**
  - **strong** selection: reproduce results of background selection
  - **weak** selection: deviations from neutrality, background selection predictions
  - **weak** selection: heterozygosity signatures clearly distinct from neutral models
- **coalescent probabilities depend on time varying ancestry dependent effective population size**
- **mean coalescence path approximation** - weak selection, large N
  - coalescence in zeroth class determined by time to get there
  - no N dependence
- **beneficial mutations**
- **positive and negative selection**

PRF

N_e

U_d/s

n_k

$$< \pi_d > = 2U_d/s$$

MM Desai, AM Walczak, JB Plotkin, arXiv:1010.2478v1

MM Desai, AM Walczak, LE Nicolaisen, JB Plotkin, arXiv:1010.2479v1