

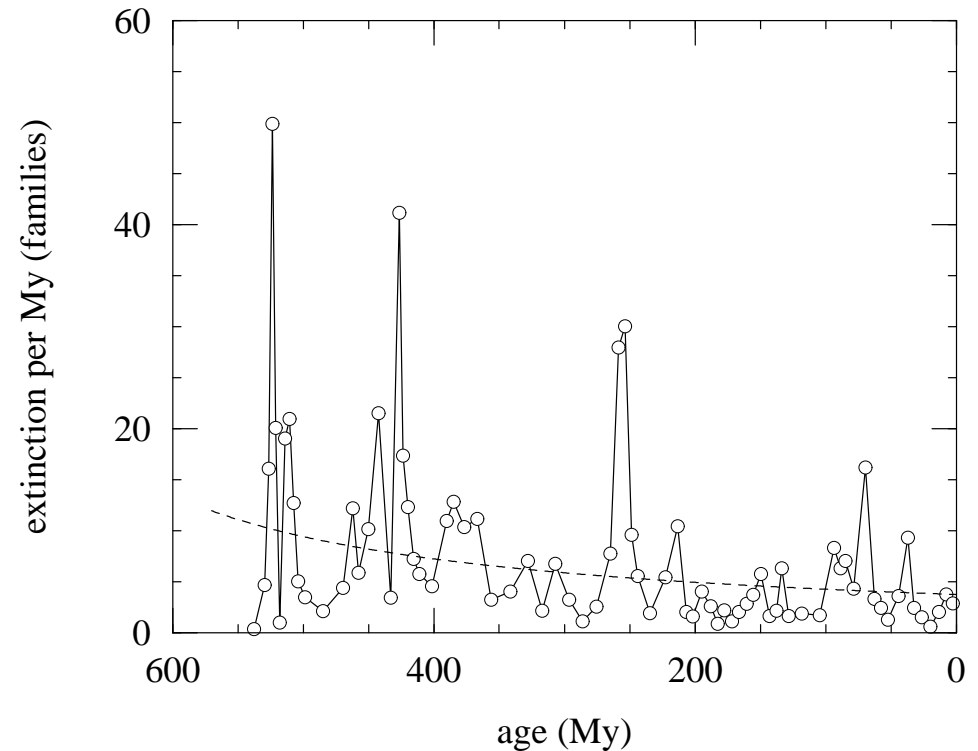
Regimes of asexual evolution in rugged fitness landscapes

- Macroevolution and microevolution
- The adaptive landscape
- Sequence space, fitness and Fisher-Wright sampling
- Regimes of asexual population dynamics
- Evolutionary trajectories in the quasispecies model

Joint work with Kavita Jain JSTAT (2005) P04008, [q-bio.PE/0606025](#), [q-bio.PE/0508008](#)
Supported by DFG within SFB/TR 12 “Symmetries and Universality in Mesoscopic Systems”
& SFB 680 “Molecular Basis of Evolutionary Innovations”

Macroevolution: Episodic patterns in the fossil record

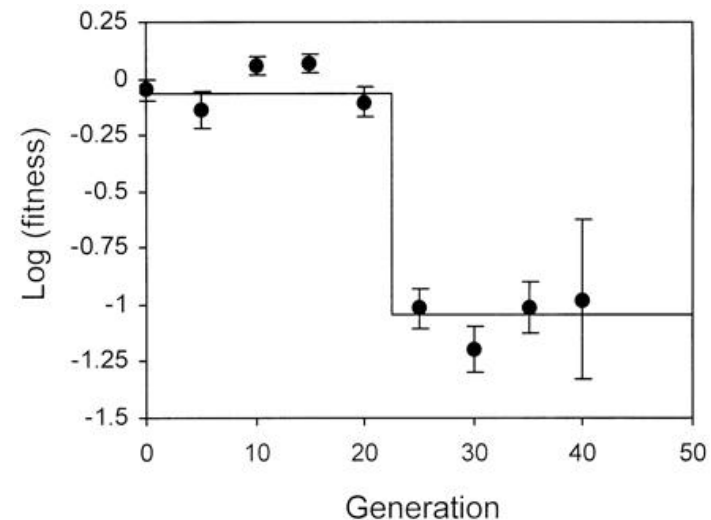
- Number of extinct families of known marine organisms (Sepkoski, 1992)
- Extinction rate is intermittent with power law event size distribution
- Decline of the average extinction rate \Rightarrow **non-stationary** dynamics



Newman & Eble (1999)

Microevolution: Adaptation of viral populations

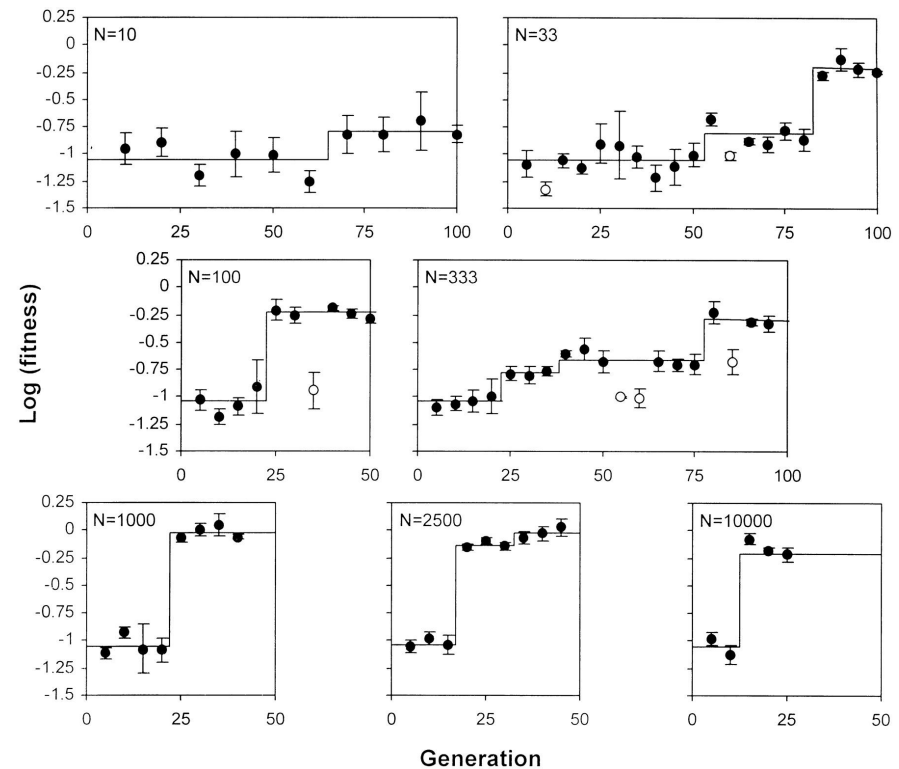
- Fitness **decline** induced in populations of bacteriophage $\phi 6$ by successive population bottlenecks
- Fitness recovery through increasing population size
- Step-like fitness changes reflect single deleterious or beneficial (compensatory) mutations
- Evidence for ruggedness of the fitness landscape



C.L. Burch, L. Chao,
Genetics **151**, 921 (1999)

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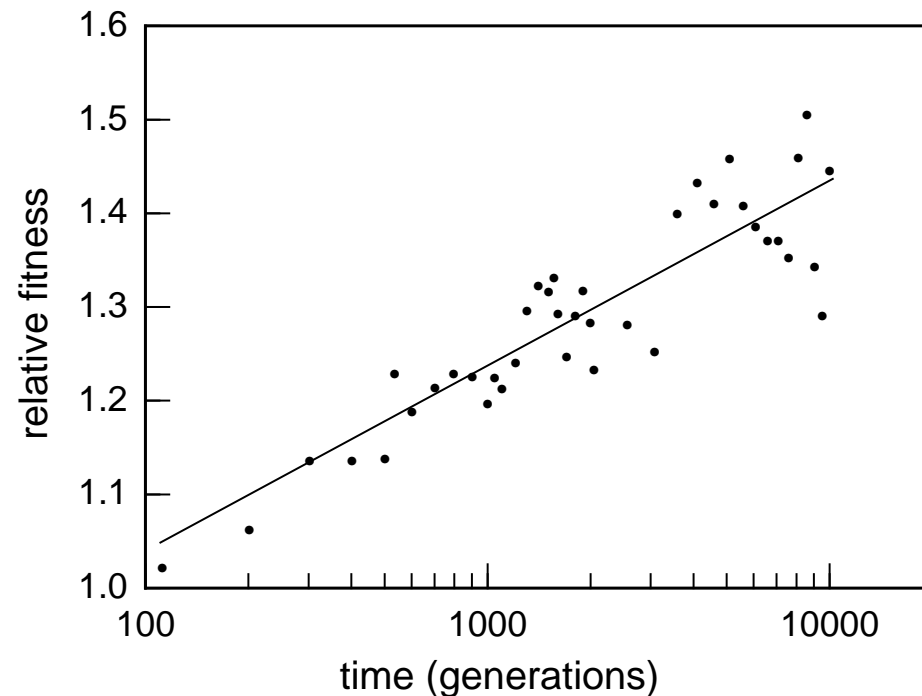


C.L. Burch, L. Chao,
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Microevolution: Adaptation of bacterial populations

S.F. Elena, R.E. Lenski, Nature Reviews Genetics 4, 457 (2003)

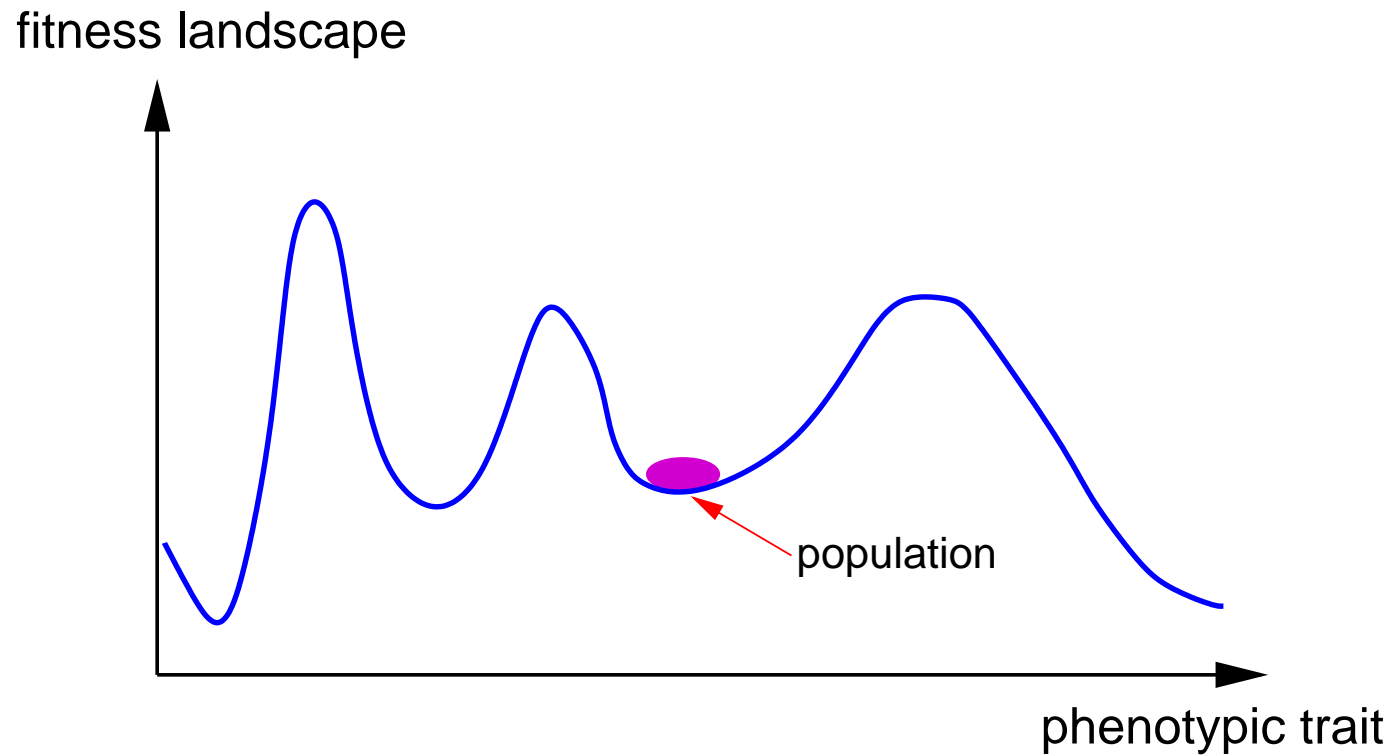
- 12 populations of *E. coli* propagated in identical environments
- Step-wise increase of cell size and fitness reflects selection of rare beneficial mutations; 4 - 6 steps in 10000 generations
- Fitness increase slows down, but rate of genetic change does not
- Parallel and divergent changes



fit: Sibani, Brandt & Alstrøm (1998)

Adaptation as a hill-climbing process

S. Wright, 1932

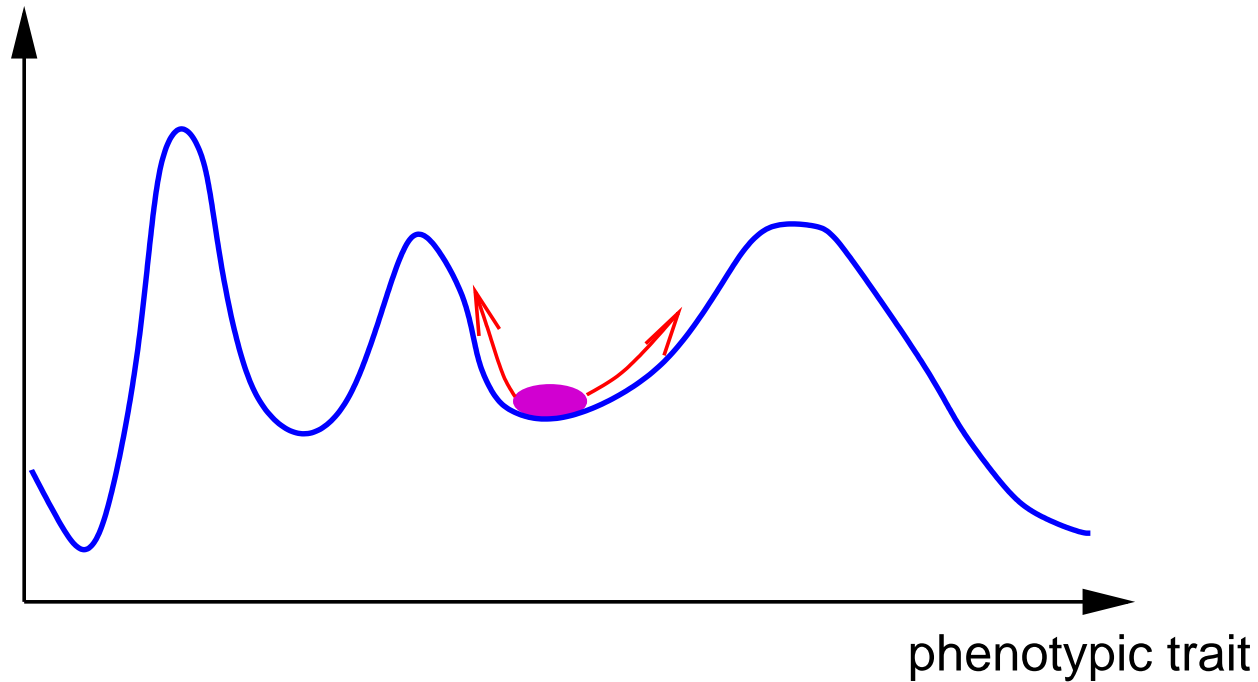


- Initial population placed in a new environment

Adaptation as a hill-climbing process

S. Wright, 1932

fitness landscape

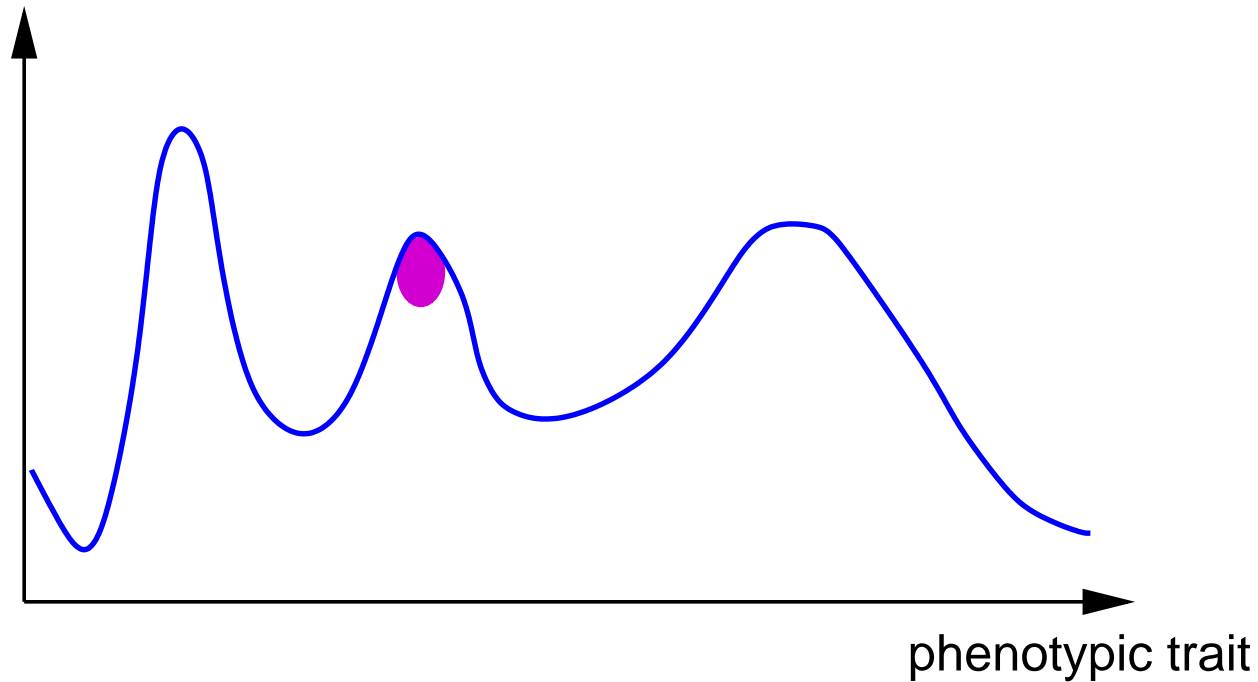


- Rapid adaptation to a nearby fitness maximum

Adaptation as a hill-climbing process

S. Wright, 1932

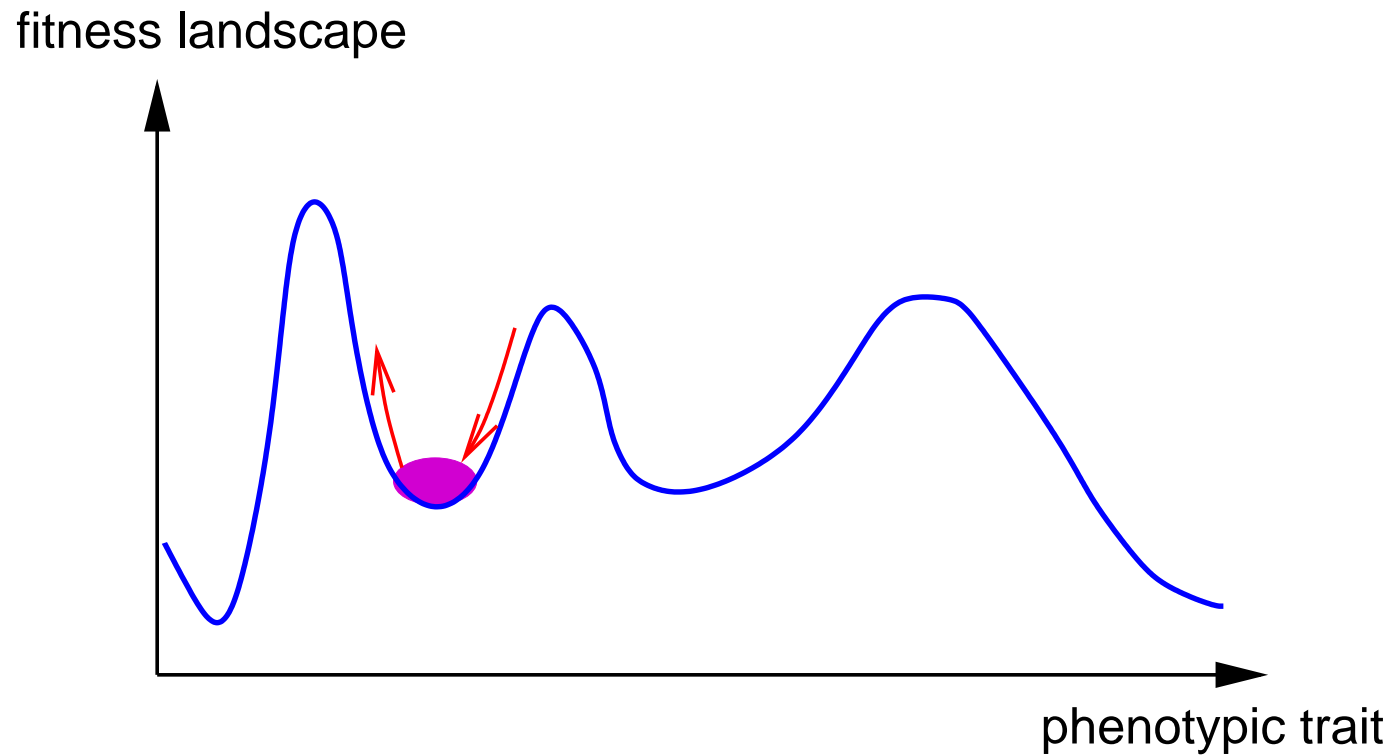
fitness landscape



- Stabilizing selection at the fitness peak; mutation-selection balance

Adaptation as a hill-climbing process

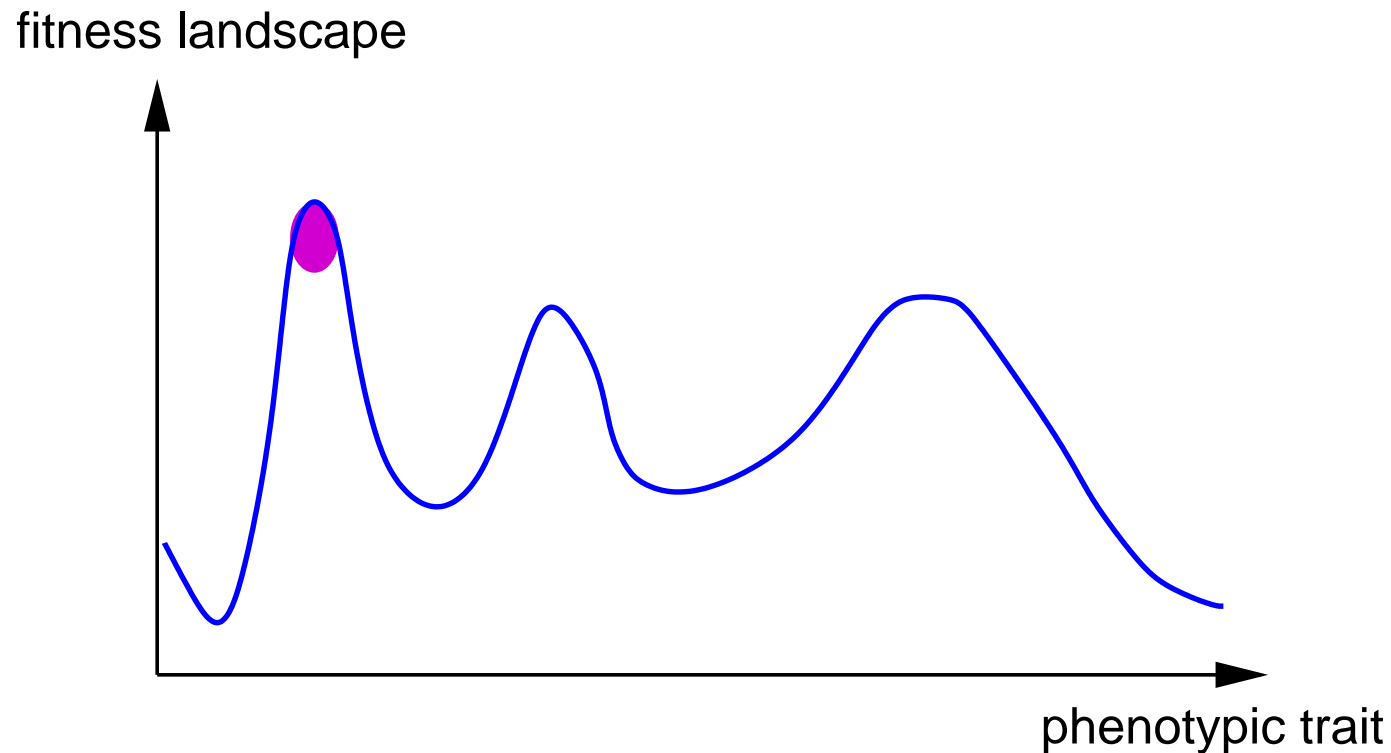
S. Wright, 1932



- Rare, rapid transition through non-adaptive zone to a higher fitness peak:

Adaptation as a hill-climbing process

S. Wright, 1932

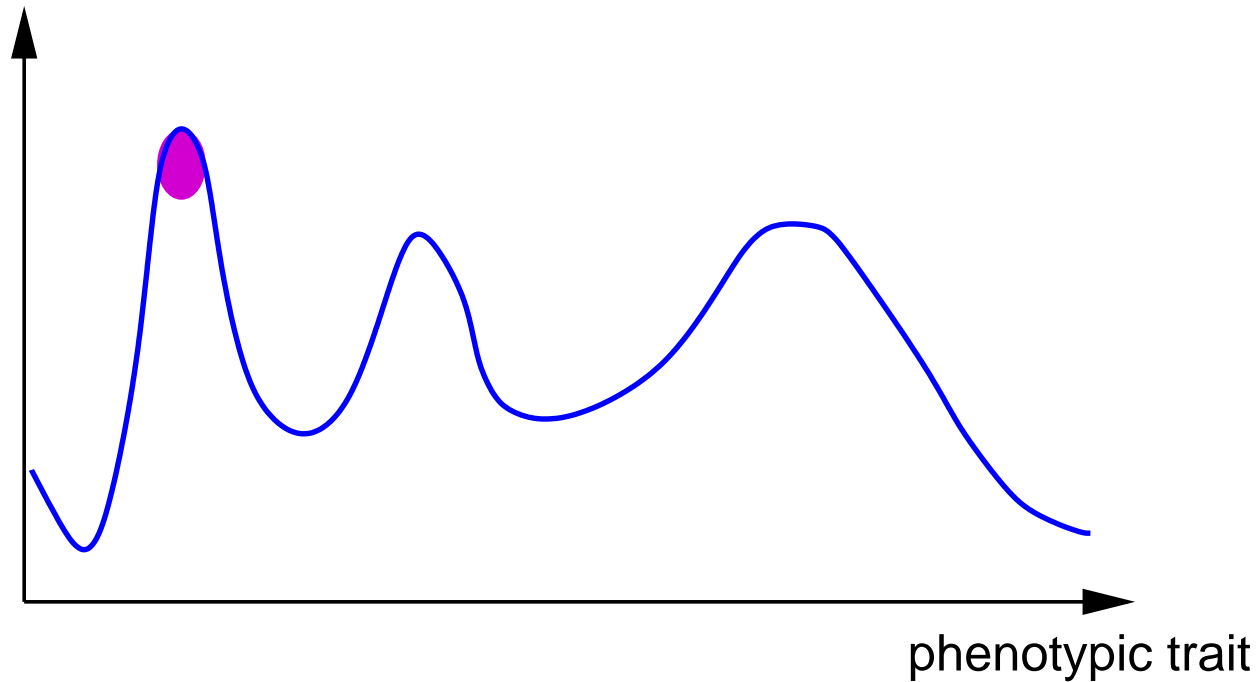


- Rare, rapid transition through non-adaptive zone to a higher fitness peak:
Quantum evolution (G.G. Simpson, 1944)

Adaptation as a hill-climbing process

S. Wright, 1932

fitness landscape



- **Goal:** To turn this mental imagery into a theory of adaptation

Sequence space

- Each individual carries a genetic sequence of length L

$\sigma = (\sigma_1, \sigma_2, \dots, \sigma_L)$ with $\sigma_i = A, C, G, T$ **genotype**

Simplification: Binary sequences $\sigma_i = 0, 1$

- **Point mutations** change individual letters in the sequence:

$$\sigma_i \rightarrow 1 - \sigma_i$$

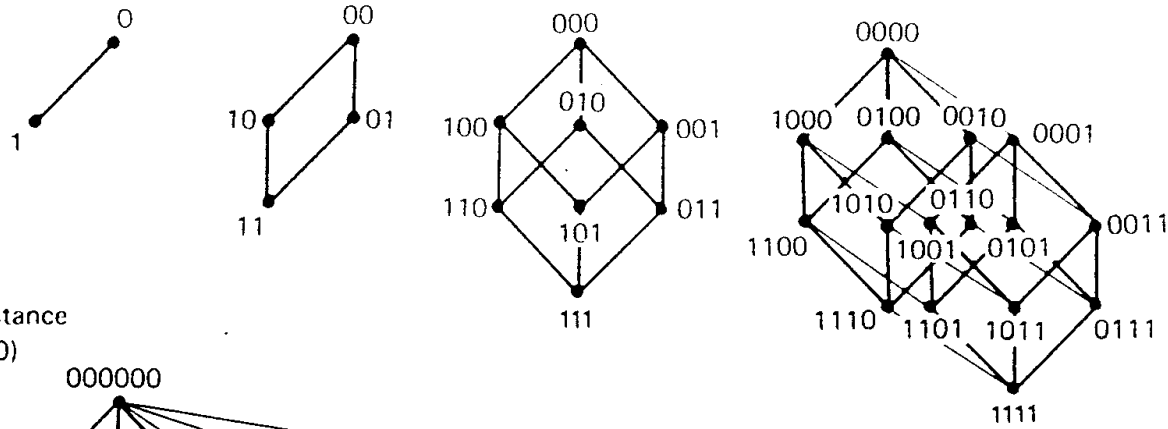
- The **Hamming distance** between two sequences σ, σ' is the number of letters in which they differ:

$$d(\sigma, \sigma') = \sum_{i=1}^L (\sigma_i - \sigma'_i)^2$$

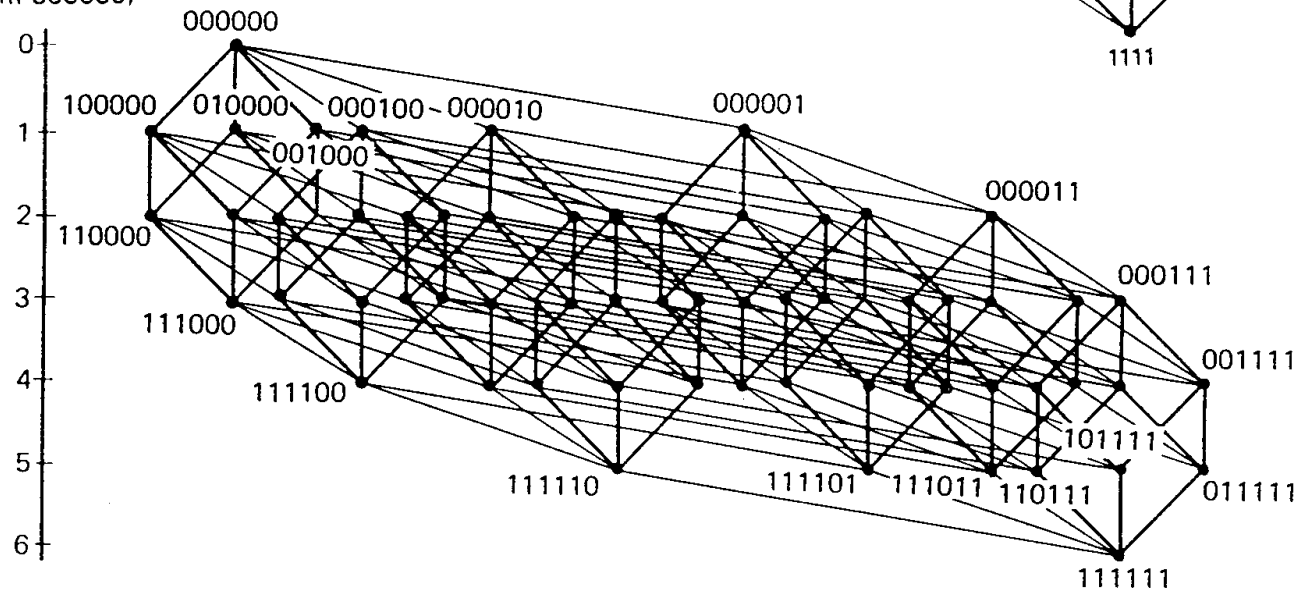
- Total number of sequences: $S = 2^L$ **volume**
Maximal distance between two sequences: $L = \log_2(S)$ **diameter**

\Rightarrow infinite dimensionality!

L-dimensional hypercubes/Hamming graphs

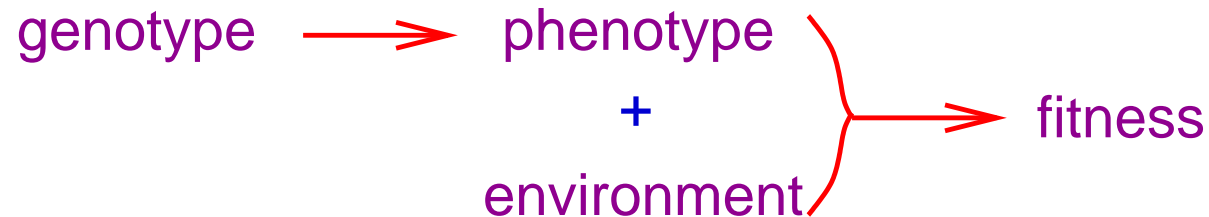


Hamming Distance
(from 000000)



Fitness

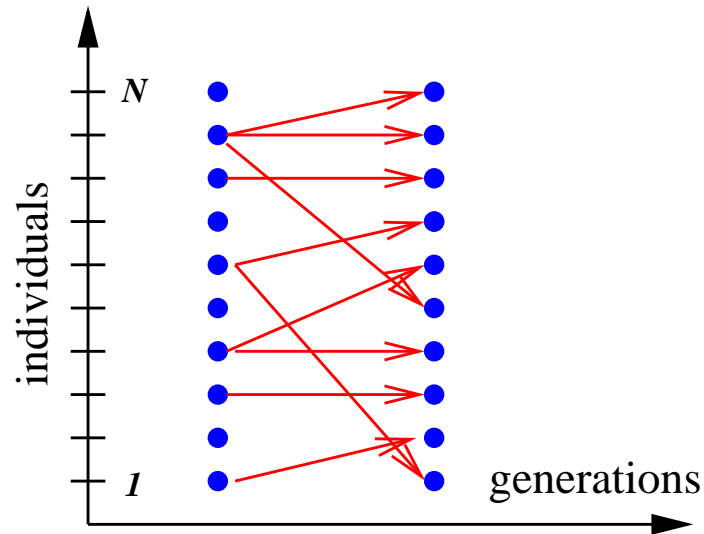
- The fitness $W(\sigma)$ of genotype σ is the **expected number of offspring** of an individual carrying σ
- The mapping $\sigma \rightarrow W(\sigma)$ is very complicated:



- Single peak landscapes: $W(\sigma) = w(d(\sigma, \sigma_0))$ σ_0 : master sequence
- Maximally rugged fitness landscape: Fitnesses $W(\sigma)$ are **uncorrelated quenched random variables** drawn from a common distribution
 - \Rightarrow random energy model (REM) of spin glasses (Derrida, 1981)
 - \Rightarrow house of cards model of population genetics (Kingman, 1977)

Evolution of asexual populations

Basic model: Stochastic Fisher-Wright sampling of a finite population



- Each individual chooses an ancestor from the preceding generation
- Genotype σ is chosen with probability $\sim W(\sigma)$
- Point mutations occur with probability μ per site and generation
- Interplay of disorder with two distinct sources of fluctuations ($\sim 1/N, \mu$)

Fixation

- In the absence of mutations ($\mu = 0$) the population becomes genetically homogeneous (**monomorphic**) for $t \rightarrow \infty$
- When a single mutant σ' is introduced into a monomorphic population with genotype σ , the outcome for $t \rightarrow \infty$ is either fixation (**all σ'**) or loss of the mutation (**all σ**)
- Fixation of **deleterious** mutations ($W(\sigma') < W(\sigma)$) is exponentially unlikely for large N
- **Beneficial** mutations ($W(\sigma') > W(\sigma)$) are fixed with probability

$$\Pi(\sigma', \sigma) = 1 - \frac{W(\sigma)}{W(\sigma')}$$

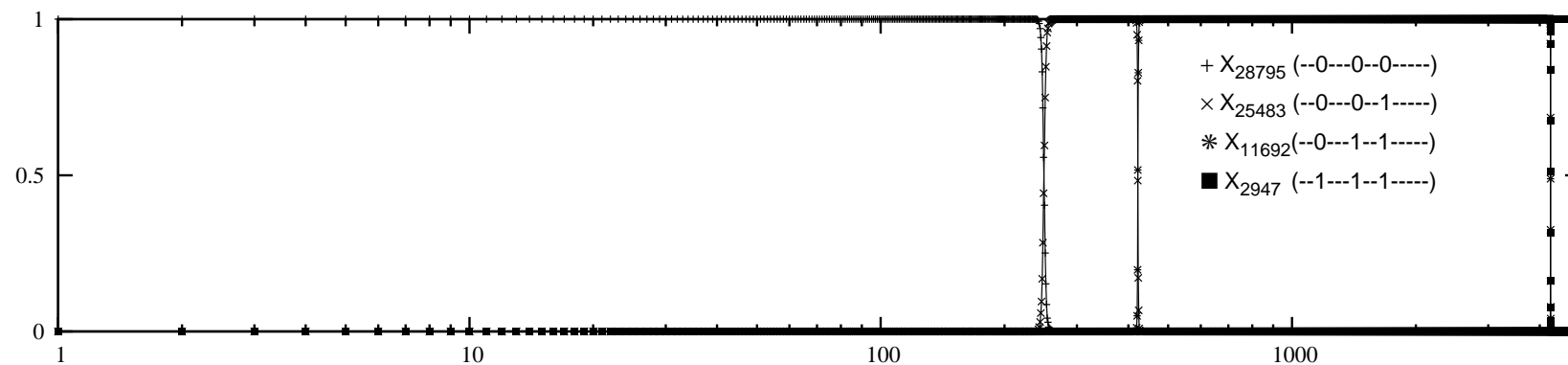
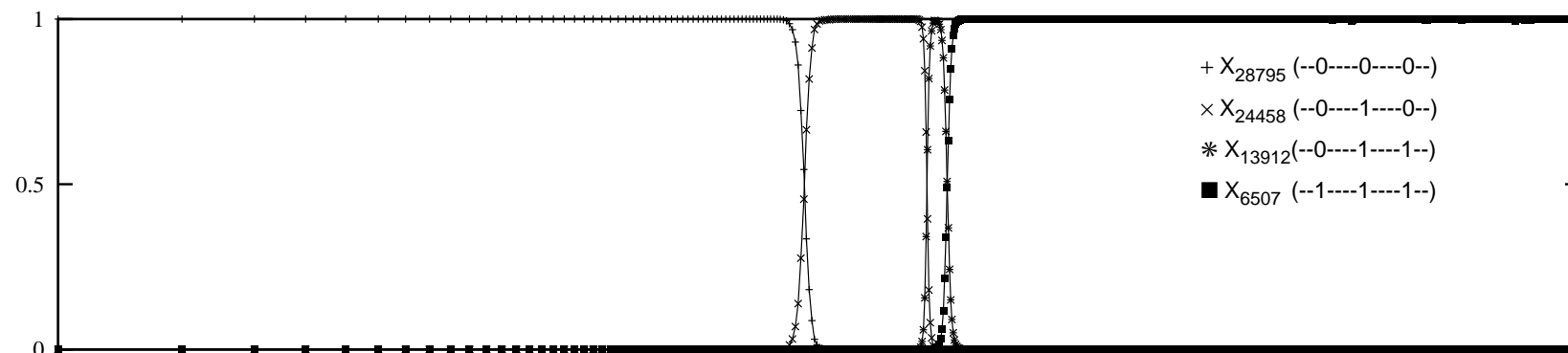
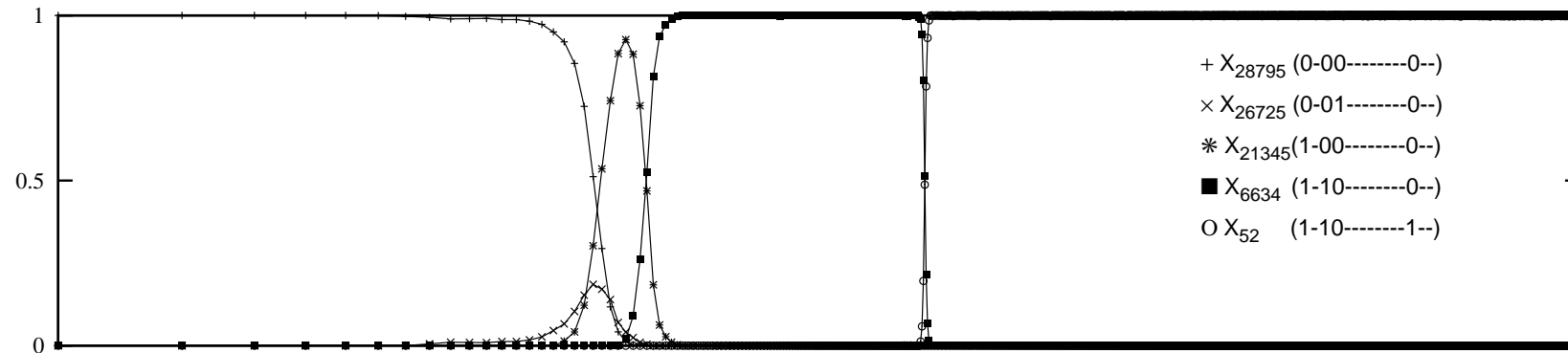
Neutral evolution and adaptive walks

- In a flat fitness landscape [$W = \text{const.}$] the population diffuses randomly in sequence space (**genetic drift**)
- Average genetic distance between two individuals: (Derrida & Peliti 1991)

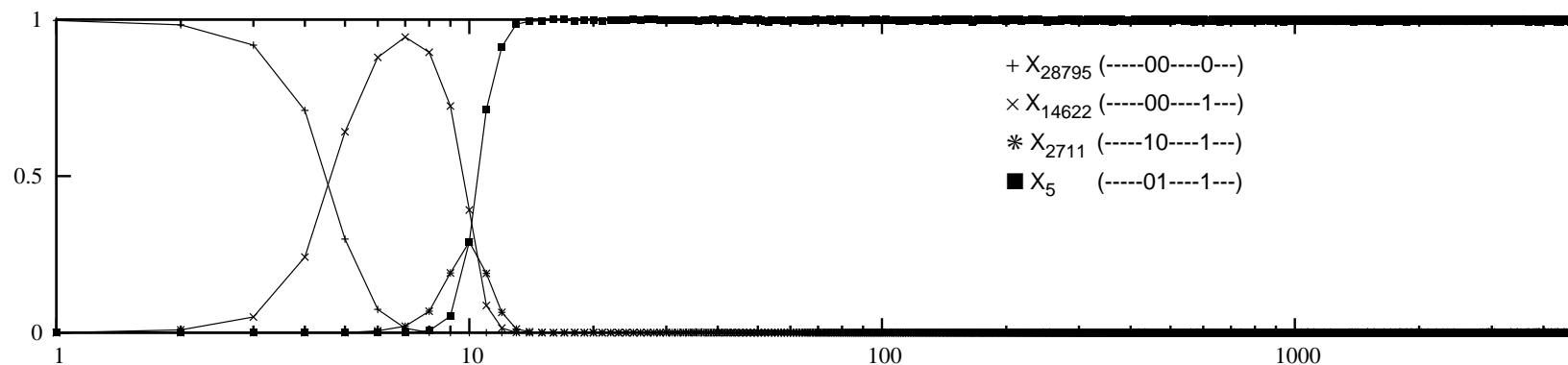
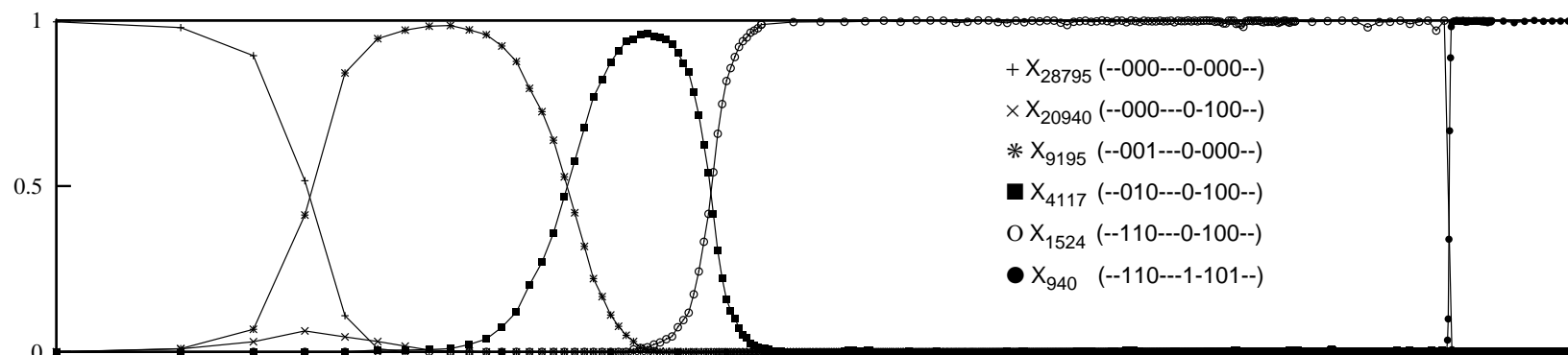
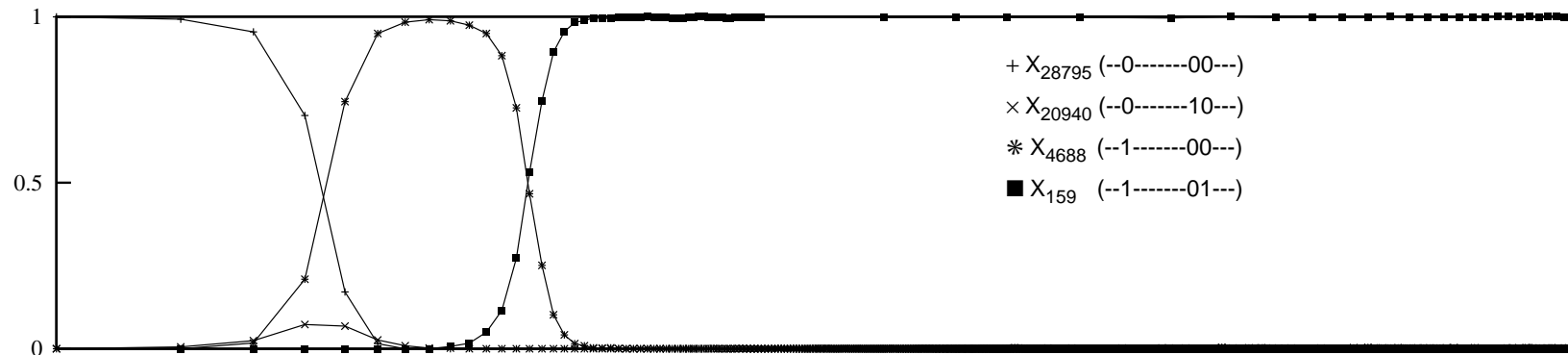
$$\langle d(\sigma, \sigma') \rangle = \frac{L}{2} \left(\frac{4\mu N}{1 + 4\mu N} \right)$$

- $L\mu N \ll 1 \Rightarrow \langle d \rangle \ll 1 \Rightarrow$ population is monomorphic most of the time and evolves by fixation of rare **neutral** mutations
- In the presence of selection [$W \neq \text{const.}$] the population performs an uphill adaptive walk by fixation of rare **beneficial** mutations (Gillespie, 1984)
- Adaptive walks terminate at local fitness maxima

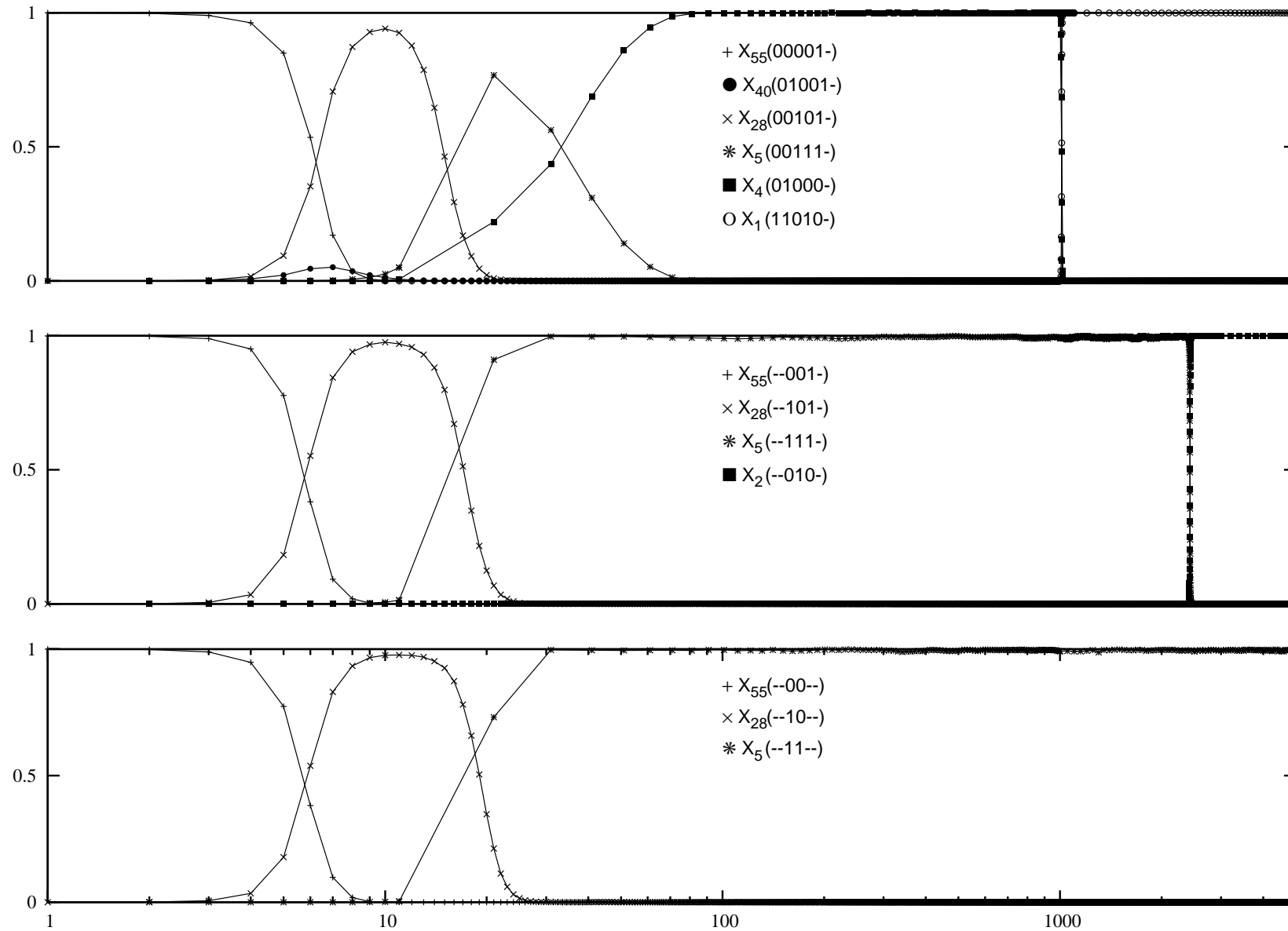
Adaptive walk: $L = 15, N = 1024, \mu = 10^{-5}, 10^{-6}, 10^{-7}$



“Clonal interference”: $L = 15, N = 1024, \mu = 10^{-4}$



Locally deterministic evolution: $L = 6, N = 16384, \mu = 10^{-4}$



Classification of evolutionary regimes

K. Jain, JK, q-bio.PE/0606025

Key parameters: Population size N , mutation probability μ per site & generation, sequence length L

- $LN\mu$: Number of mutants produced per generation
- $LN\mu \ll 1$: Adaptive walk of a monomorphic population
- $1 \ll LN\mu \ll L$: Stochastic regime with clonal interference
- $LN\mu > L$: Deterministic evolution within a shell of size

$$d_{\text{eff}} = \frac{\ln N}{|\ln \mu|}$$

around the dominating genotype

- $d_{\text{eff}} \rightarrow L$: Fully deterministic evolution \Rightarrow **quasispecies dynamics**

The quasispecies model

M. Eigen, *Naturwissenschaften* **58**, 465 (1972); K. Jain, JK, q-bio.PE/0508008

- Unnormalized population fraction $Z(\sigma, t)$ satisfies linear time evolution

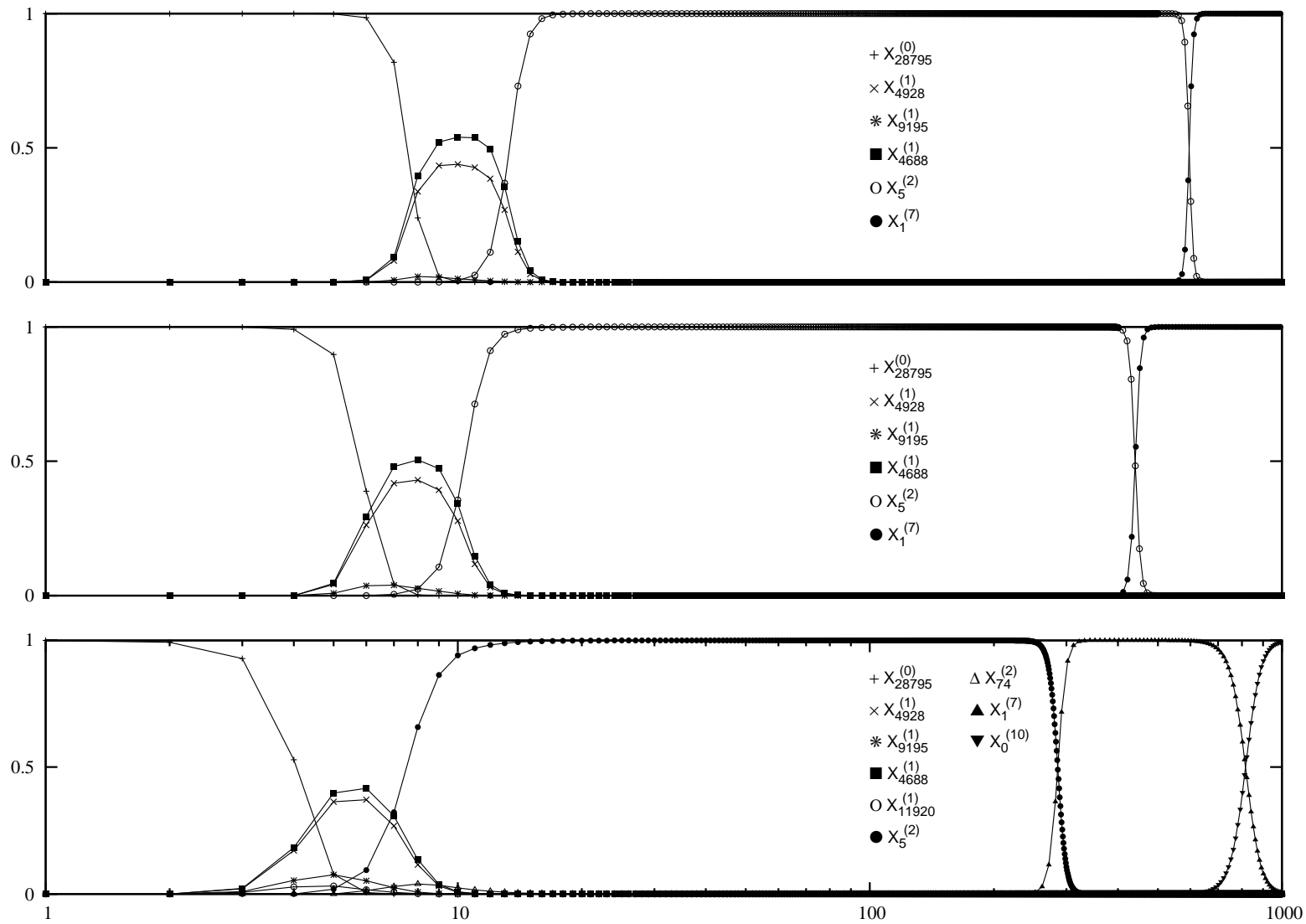
$$Z(\sigma, t+1) = \sum_{\sigma'} W(\sigma') M^{(\mu)}(\sigma', \sigma) Z(\sigma', t)$$

$$\vec{Z}(t+1) = \hat{E}^{(\mu)} \vec{Z}(t), \quad \hat{E}^{(\mu)} = \hat{W} \hat{M}^{(\mu)}, \quad \hat{W} : \text{diagonal, random}$$

- Mutation matrix: $M^{(\mu)}(\sigma', \sigma) = \mu^{d(\sigma, \sigma')} (1 - \mu)^{L-d(\sigma, \sigma')} \approx \mu^{d(\sigma, \sigma')}$
- Initial condition $Z(\sigma, 0) = \delta_{\sigma, \sigma_0} \Rightarrow$ after one generation $Z(\sigma, 1) \sim \mu^{d(\sigma, \sigma_0)}$
 \Rightarrow population fraction drops to $1/N$ at distance

$$d \approx d_{\text{eff}} = \frac{\ln N}{|\ln \mu|}$$

- Evolutionary trajectories are deterministic and independent of μ :



$$L = 15, \mu = 10^{-8}, 10^{-6}, 10^{-4}$$

The strong selection limit

JK, C. Karl, Physica A 318, 137 (2003)

- For $\mu \rightarrow 0$ mutations are irrelevant after the first generation:

$$\vec{Z}(t) = \{\hat{E}^{(\mu)}\}^t \vec{Z}(0) \rightarrow \{\hat{E}^{(0)}\}^{t-1} \vec{Z}(1)$$

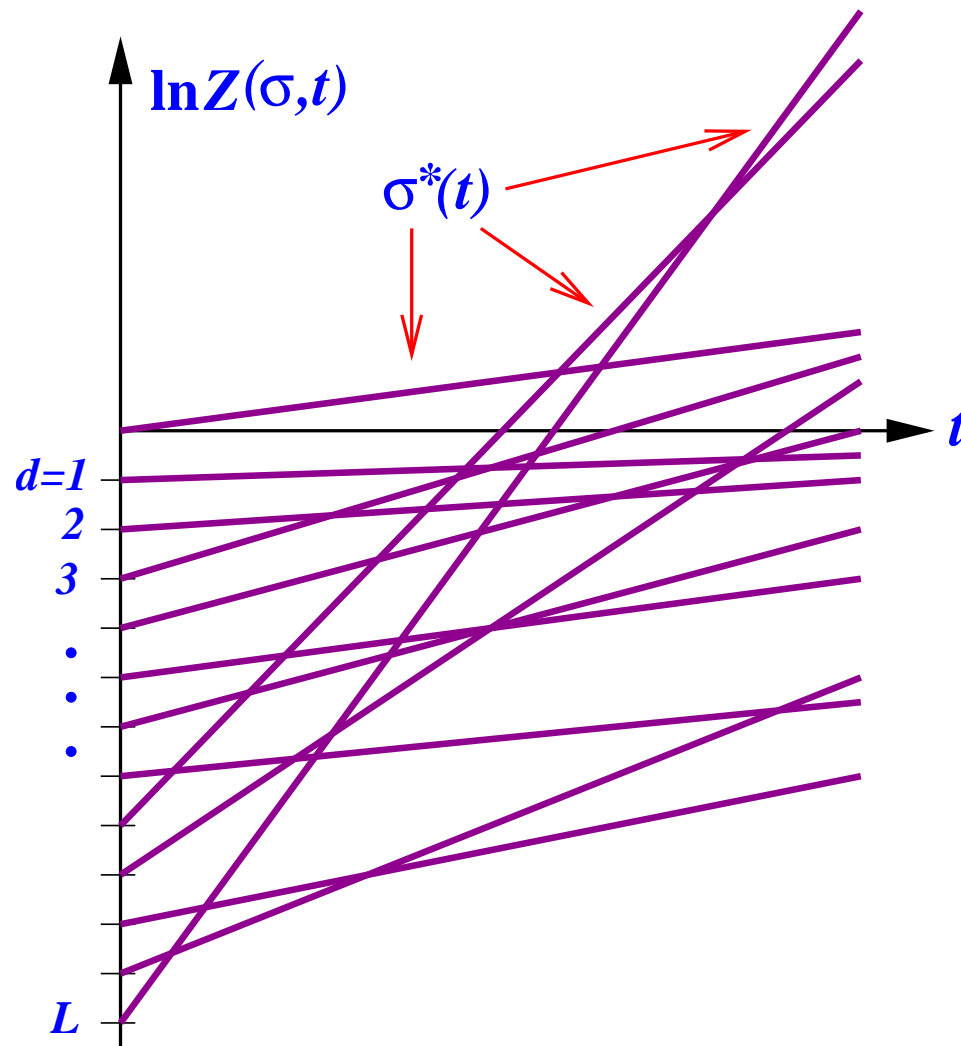
$$\Rightarrow Z(\sigma, t) \approx W(\sigma)^{t-1} W(\sigma_0) \mu^{d(\sigma, \sigma_0)}$$

- Logarithmic population variables evolve linearly in time:

$$\ln Z(\sigma, t) \approx \ln W(\sigma)(t-1) - |\ln \mu| d(\sigma, \sigma_0)$$

- Change in μ is compensated by rescaling of time with $|\ln \mu|$
- Most populated genotype $\sigma^*(t)$ is determined by maximizing $\ln Z(\sigma, t)$ with respect to σ

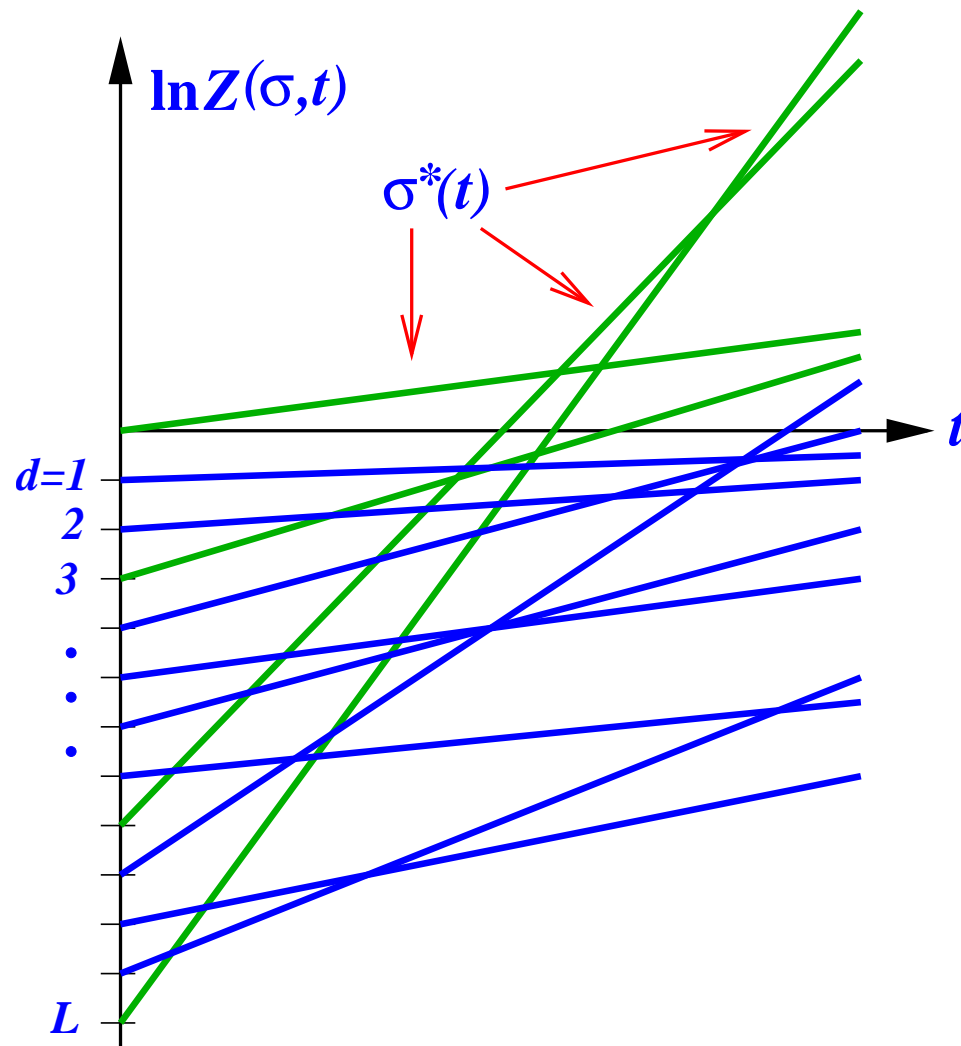
Geometry of the evolutionary race



$$\ln Z(\sigma, t) = \ln W(\sigma)(t - 1) - d(\sigma_0, \sigma)$$

- Only the most fit mutant in each **shell** of constant Hamming distance $d(\sigma, \sigma_0)$ needs to be considered \Rightarrow system size $2^L \rightarrow L$

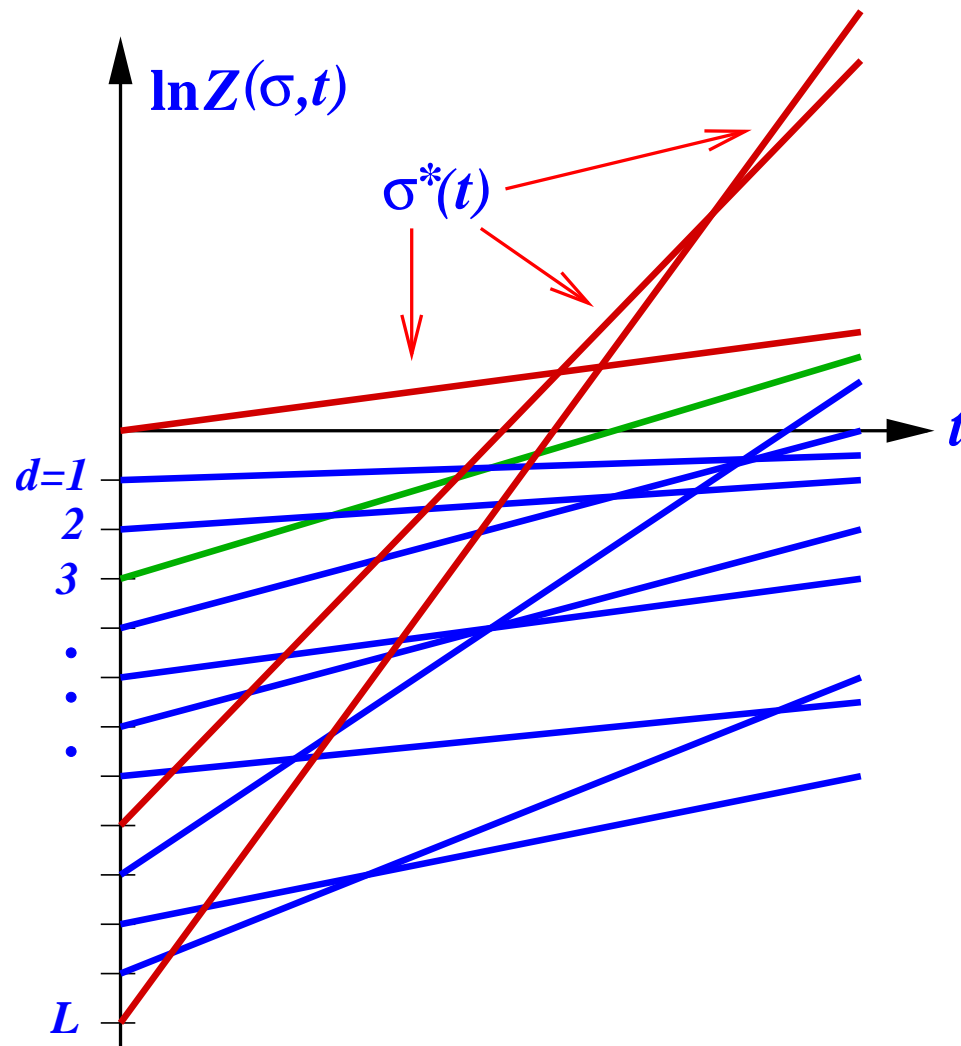
Geometry of the evolutionary race



$$\ln Z(\sigma, t) = \ln W(\sigma)(t-1) - d(\sigma_0, \sigma)$$

- **Contenders** and **spectators**: Contenders are fitness records σ_k with $W(\sigma_{k+1}) > W(\sigma_k)$

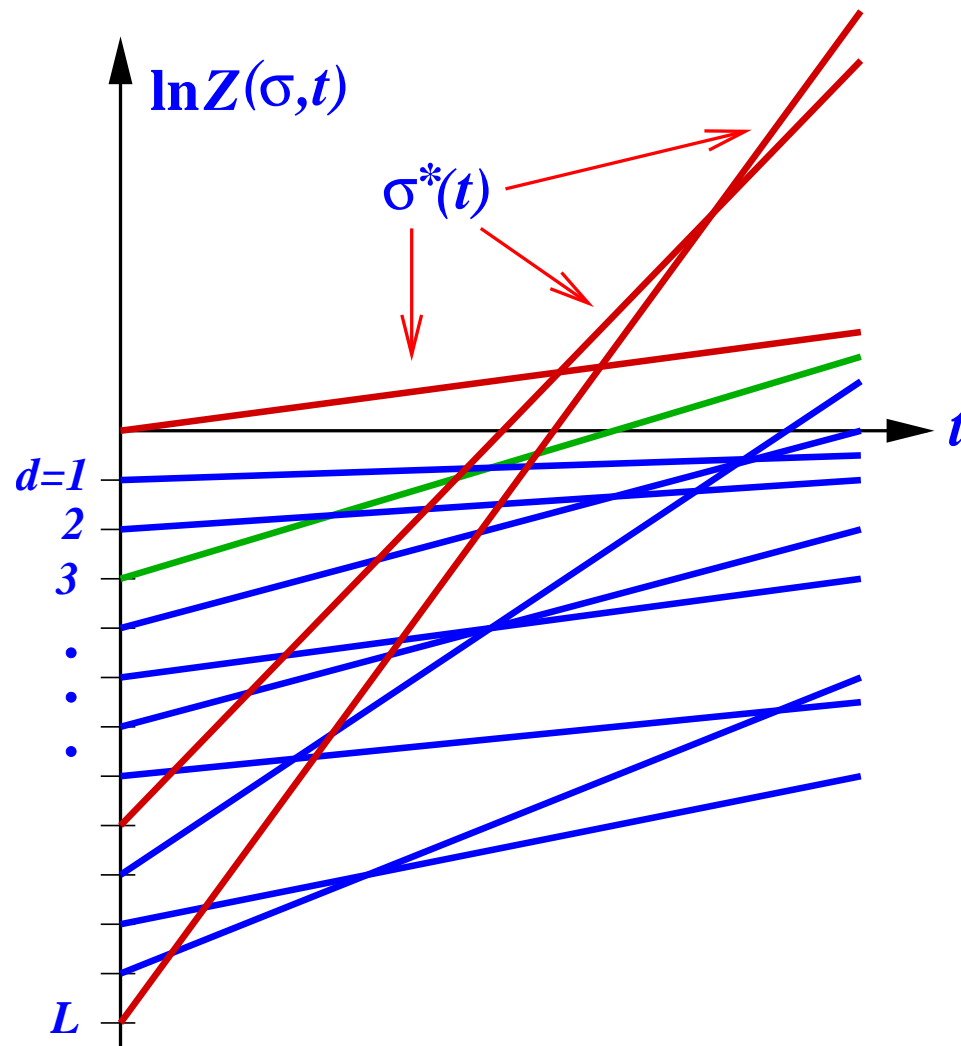
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- **Contenders** and **spectators**: Contenders are fitness records σ_k with $W(\sigma_{k+1}) > W(\sigma_k)$
- Not all contenders are **winners**; some are bypassed

Geometry of the evolutionary race



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- **Contenders** and **spectators**: Contenders are fitness records σ_k with $W(\sigma_{k+1}) > W(\sigma_k)$
- Not all contenders are **winners**; some are bypassed
- The evolutionary trajectory is composed of **non-bypassed fitness records**; how many?

Bypassing for i.i.d. shell fitnesses

- Simplification: Replace logarithmic shell fitnesses

$$F_k = \max_{d(\sigma, \sigma_0)=k} \ln W(\sigma)$$

by i. i. d. random variables from a distribution $P(F)$

- The number of records is $\ln L$ with Poisson distribution
- Numerical conjecture: Bypassing probability $1 - \beta$ with $\beta < 1$ depending on the tail of $P(F)$ JK, C. Karl (2003)

(i) exponential-like: $\beta = 1/2$

(ii) power law: $\beta = (\delta - 1)/(2\delta - 1)$ $P(F) \sim F^{-1-\delta}$

(iii) bounded: $\beta = (2 + \nu)/(3 + 2\nu)$ $P(F) \sim (F_{\max} - F)^\nu$

- Fluctuations in the number of non-bypassed records are sub-Poissonian
- Analytic proof by first-passage techniques

C. Sire, S.N. Majumdar, D.S. Dean, J. Stat. Mech. (2006) L07001

Bypassing in sequence space

K. Jain, JK, JSTAT (2005) P04008

- Number of sequences at distance d from σ_0 is $\binom{L}{d}$
- Probability to find a new fitness record at d : (M.C.K. Yang, 1975)

$$P(d) = \frac{\binom{L}{d}}{\sum_{k=1}^d \binom{L}{k}} \approx \frac{1 - 2d/L}{1 - d/L} \text{ for } L, d \rightarrow \infty, d/L < 1/2$$

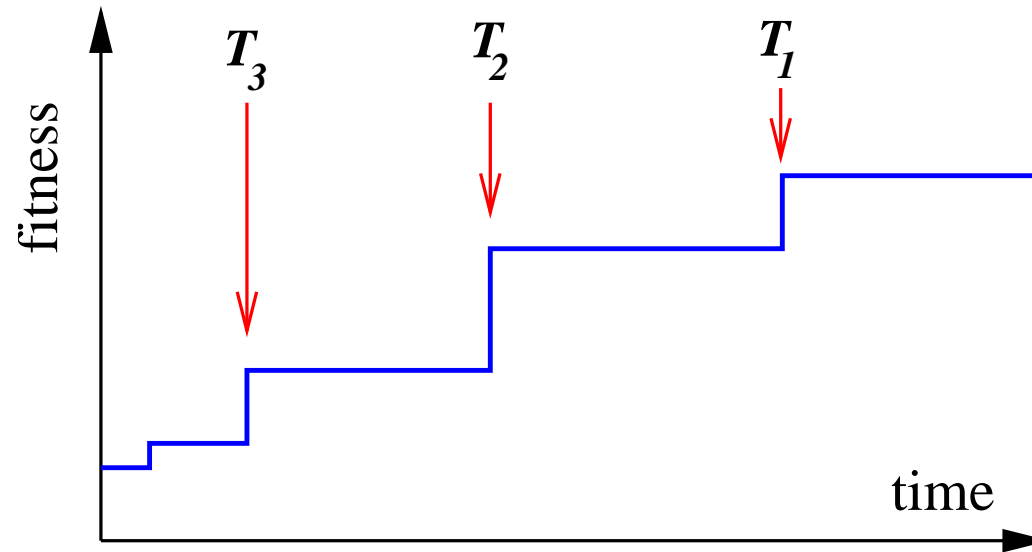
independent of the fitness distribution

- Total number of records \mathcal{R} has mean and variance

$$\langle \mathcal{R} \rangle = (1 - \ln 2)N \approx 0.307...N, \quad \langle \mathcal{R}^2 \rangle - \langle \mathcal{R} \rangle^2 = (3 \ln 2 - 2)N \approx 0.0794...N$$

- Bypassing reduces this to $\mathcal{O}(\sqrt{L})$ for exponential-like fitness distributions, and to $\mathcal{O}(1)$ for power-law fitness distributions

Distribution of evolution times



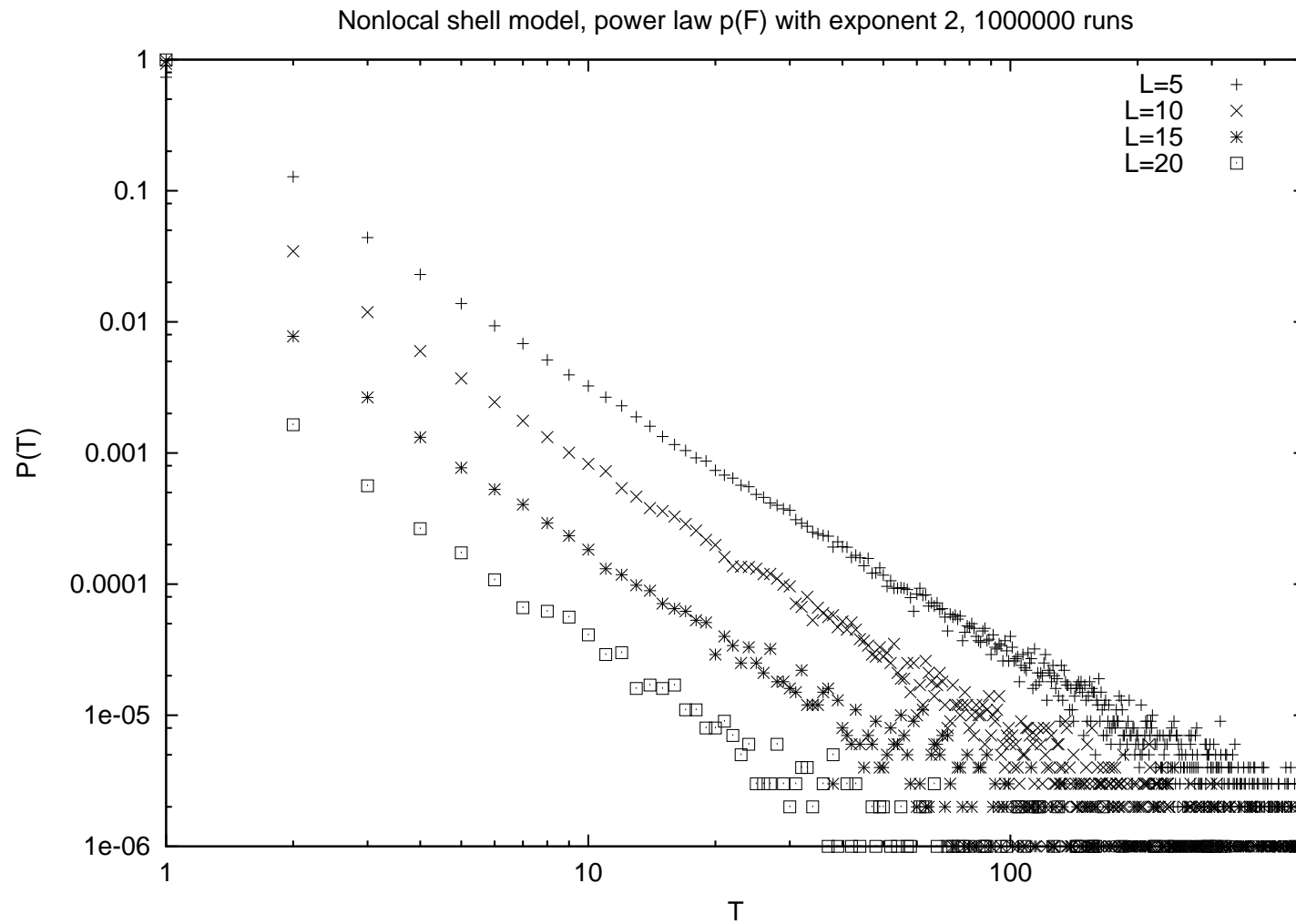
- Global fitness maximum is reached at time T_1
- **Universal** power law tail of distribution of time T_k of the k 'th last jump:

$$P_k(T_k) \sim T_k^{-(1+k)}$$

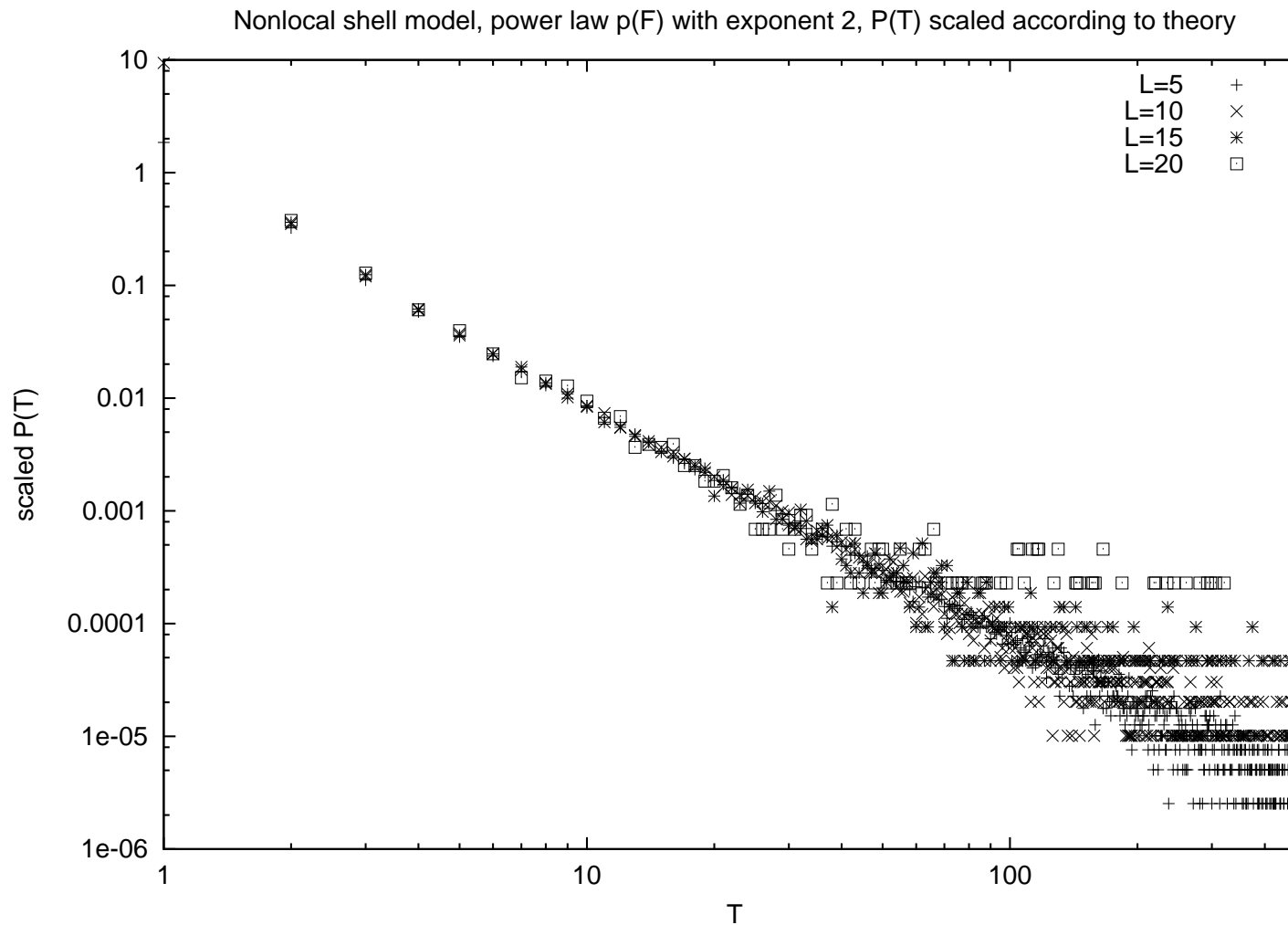
with prefactor depending on fitness distribution and sequence length.

- Expected evolution time is **infinite**: $\langle T_1 \rangle = \infty$

- Exponential fitness distribution: $P_1(T) \sim \sqrt{L}/T^2$
- Power law fitness distribution: $P(F) \sim F^{-(\delta+1)} \Rightarrow P_1(T) \sim \sqrt{L} 2^{-L/\delta} / T^2$



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\Rightarrow All intermediate maxima are bypassed in the power law case!

Summary

- Evolutionary biology & population genetics a rich source of stochastic many-body problems with disorder
- State of the art largely restricted to neutral evolution, small number of sites ($L = 1, 2$), non-interacting (=multiplicative) fitness landscapes
- Distinct fluctuations from mutations & sampling noise
- Different kinds of “thermodynamic limits”:
 - $N \rightarrow \infty \Rightarrow$ quasispecies model
 - $L \rightarrow \infty \Rightarrow$ infinite sites model
- Future directions (in SFB 680)
 - (i) Speed of adaptation in smooth landscapes
 - (ii) Timing of adaptive events in the infinite sites model
 - (iii) Modes of reproduction in diploids (sexual, parthenogenetic, selfing)