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Historical prologue

The work on a molecular theory of evolution started more than 40 years ago

DIE NATURWISSENSCHAFTEN

58. Jahrgang, 1971

Heft 10 Oktober

Selforganization of Matter and the Evolution of Biological Macromolecules

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I. Introduction

I.1. „Cause and Effect“

The question about the origin of life often appears as a question about "cause and effect". Physical theories of macroscopic processes usually involve answers to such questions, even if a statistical interpretation is given to the relation between "cause" and "effect". It is mainly due to the nature of this question that many scientists believe that our present physics does not offer any obvious explanation for the existence of life.

* Partly presented as the "Robbins Lectures" at Pomona College, California, in spring 1970.

which even in its simplest forms always appears to be associated with complex macroscopic (i.e. multimolecular) systems, such as the living cell.

As a consequence of the exciting discoveries of "molecular biology", a common version of the above question is: *Which came first, the protein or the nucleic acid?*—a modern variant of the old "chicken-and-egg" problem. The term "first" is usually meant to define a causal rather than a temporal relationship, and the words "protein" and "nucleic acid" may be substituted by "function" and "information". The question in this form, when applied to the interplay of nucleic acids and proteins as presently encountered in the living cell, leads ad absurdum, because "function"

1971

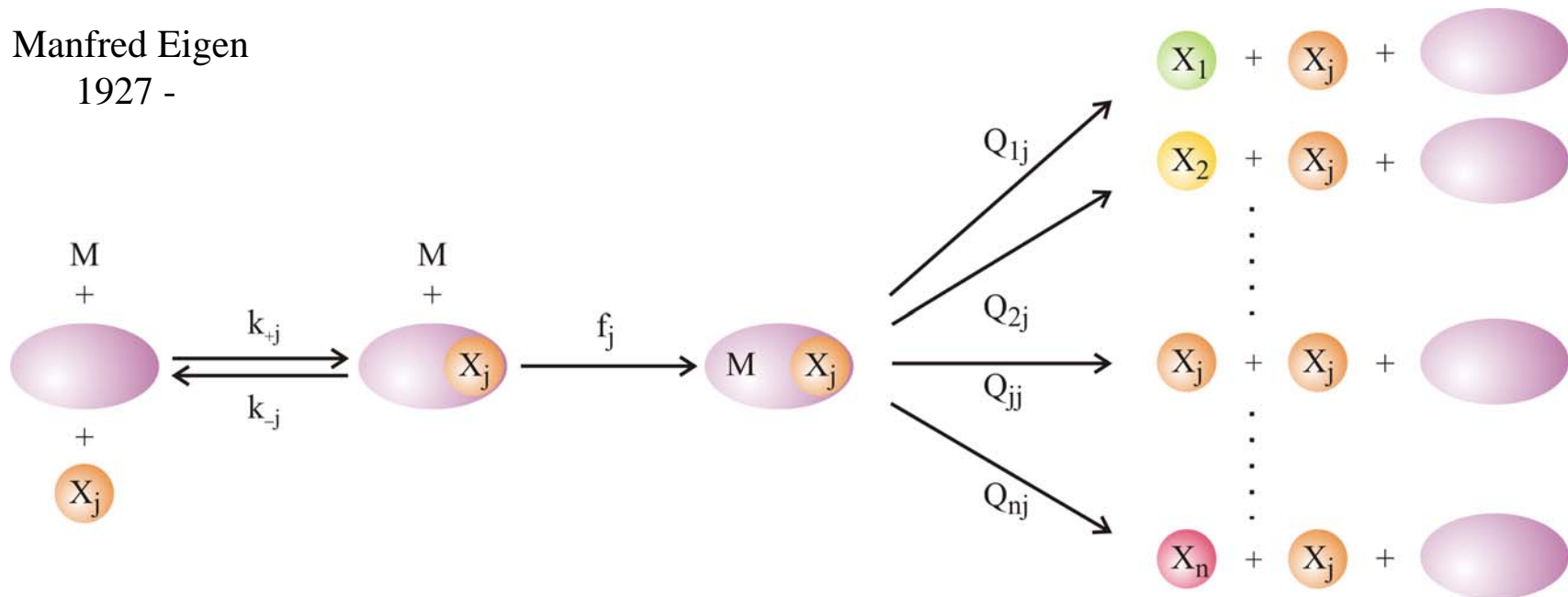
Chemical kinetics of molecular evolution



Manfred Eigen
1927 -

$$\frac{dx_j}{dt} = \sum_{i=1}^n W_{ji} x_i - x_j \Phi ; j = 1, 2, \dots, n$$

$$\Phi = \sum_{i=1}^n f_i x_i / \sum_{i=1}^n x_i$$



Mutation and (correct) replication as parallel chemical reactions

M. Eigen. 1971. *Naturwissenschaften* 58:465,

M. Eigen & P. Schuster. 1977. *Naturwissenschaften* 64:541, 65:7 und 65:341

Molecular Evolution

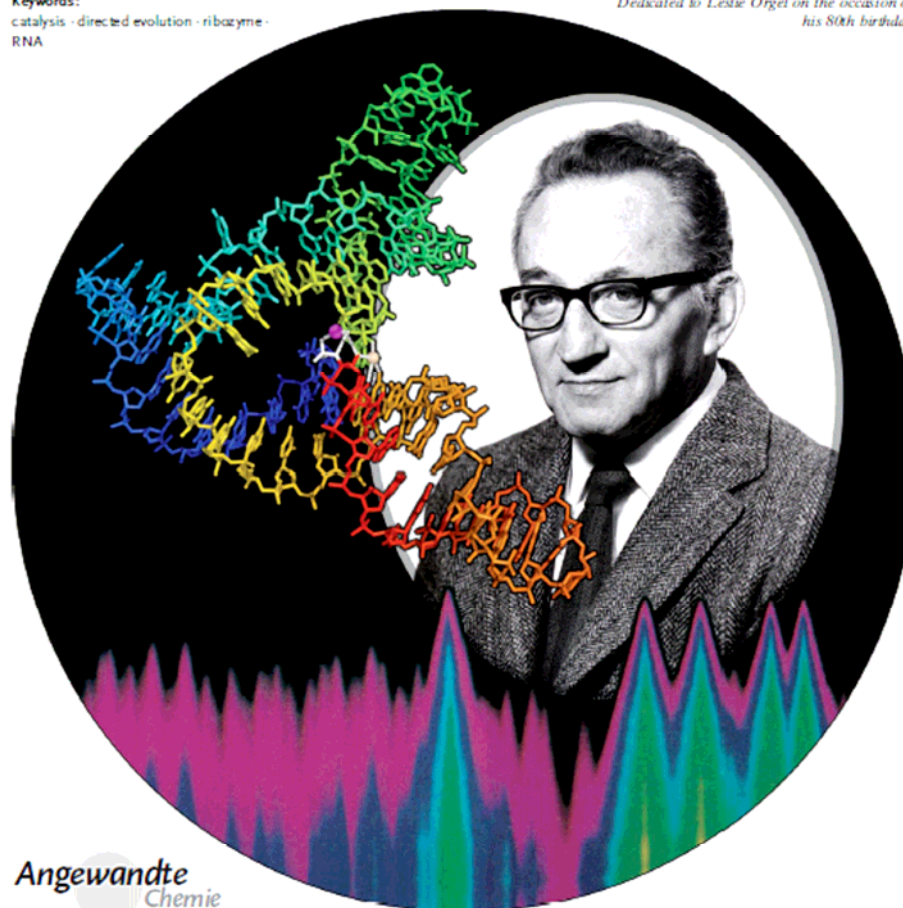
DOI: 10.1002/anie.200701369

Forty Years of In Vitro Evolution**

Gerald F. Joyce*

Keywords:
catalysis · directed evolution · ribozyme ·
RNA

Dedicated to Leslie Orgel on the occasion of
his 80th birthday



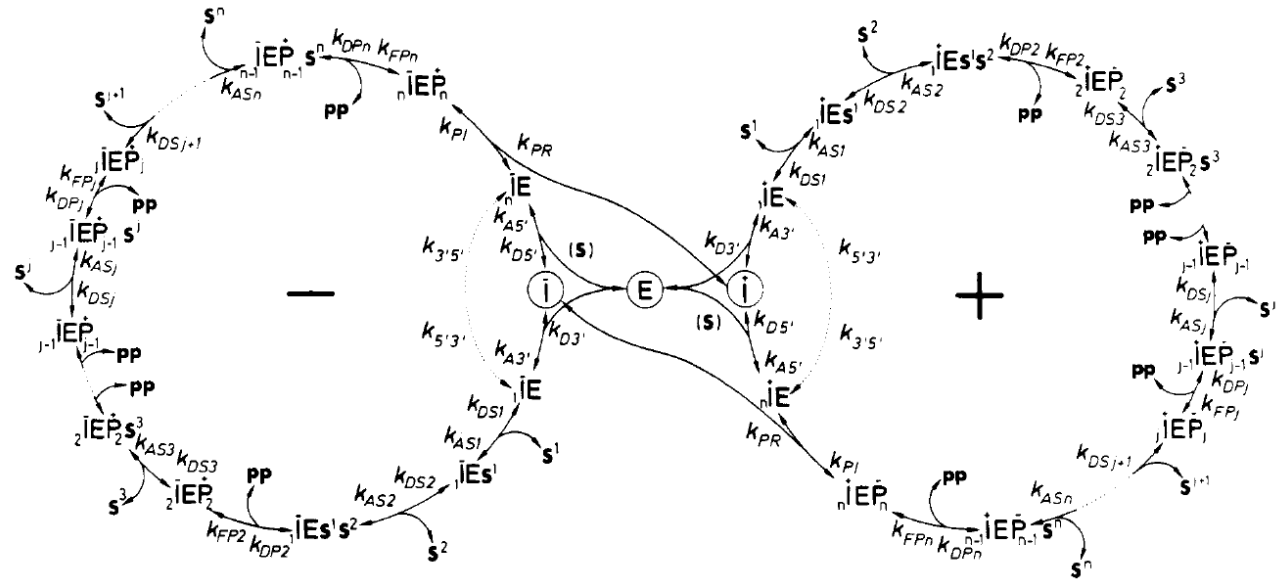
Sol Spiegelman,
1914 - 1983

Evolution in the test tube:

G.F. Joyce, *Angew.Chem.Int.Ed.*
46 (2007), 6420-6436

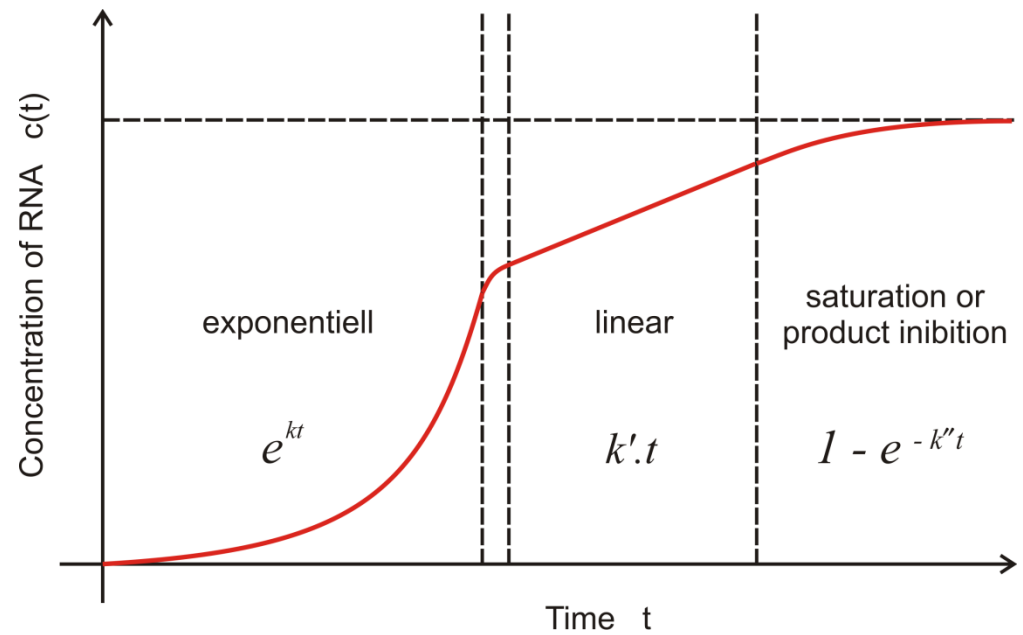


Christof K. Biebricher,
1941-2009



Kinetics of RNA replication

C.K. Biebricher, M. Eigen, W.C. Gardiner, Jr.
Biochemistry **22**:2544-2559, 1983



```

      C           U           AGC           UU  GG
pppGGG ACCCCCC UCGGGGGGUCACCUCGCGU UAGCUACGCGAGGG AAA
HOA CCC  UGGGGGG AAA AGCCCCCCCAGUGGAGCGCA AUCGAUGC GCUCCC UUU CC
      UU
    
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plus SV-11 (115 b)

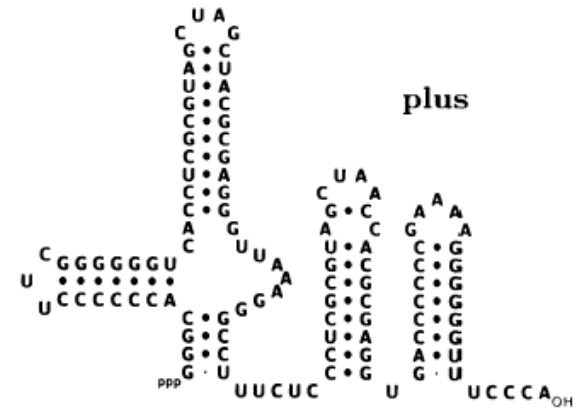
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      AA           UUU           A           AG
pppGGG ACCCCCC UCGGGGGGUCACCUCGCGUGGUUAGCUACGCGAGGG GAA G
HOA CCC  UGGGGGG A AGCCCCCCCAGUGGAGCGCAUCGAUCGAUGC GCUCCC AA UUU C
      G
    
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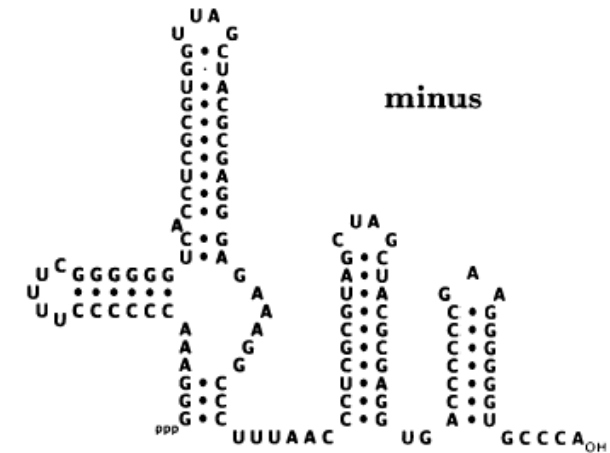
minus

stable

does not replicate!



SV-11 (115 b)

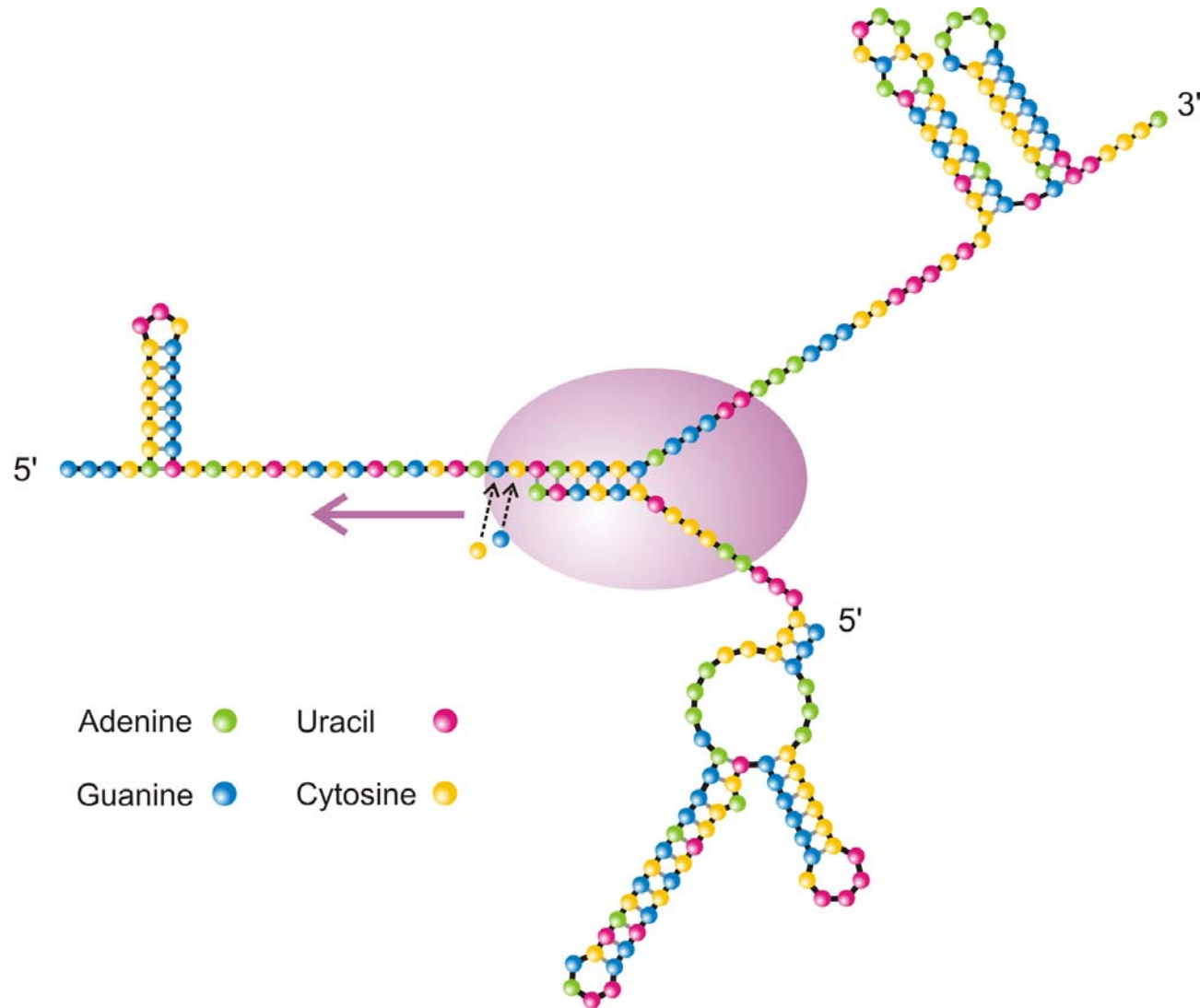


minus

metastable

replicates!

C.K. Biebricher, R. Luce. 1992. *In vitro* recombination and terminal recombination of RNA by Qβ replicase. *The EMBO Journal* 11:5129-5135.



Charles Weissmann
1931-

RNA replication by Q β -replicase

C. Weissmann, *The making of a phage.*

FEBS Letters **40** (1974), S10-S18

The Hypercycle

A Principle of Natural Self-Organization

Part A: Emergence of the Hypercycle

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This paper is the first part of a trilogy, which comprises a detailed study of a special type of functional organization and demonstrates its relevance with respect to the origin and evolution of life. Self-replicative macromolecules, such as RNA or DNA in a suitable environment exhibit a behavior, which we may call Darwinian and which can be formally represented by the concept of the quasi-species. A quasi-species is defined as a given distribution of macromolecular species with closely interrelated sequences, dominated by one or several (degenerate) master copies. External constraints enforce the selection of the best adapted distribution, commonly referred to as the wild-type. Most important for Darwinian behavior are the criteria for lateral stability of the quasi-species. If these criteria are violated, the information stored in the nucleotide sequence of the master copy will disintegrate irreversibly leading to an error catastrophe. As a consequence, selection and evolution of RNA or DNA molecules is limited with respect to the amount of information that can be stored in a single replicative unit. An analysis of experimental data regarding RNA and DNA replication at various levels of organization reveals, that a sufficient amount of information for the build up of a translation machinery can be gained only via integration of several different replicative units (or reproductive cycles) through functional linkages. A stable functional integration then will raise the system to a new level of organization and thereby enlarge its information capacity considerably. The hypercycle appears to be such a form of organization.

Preview on Part B: The Abstract Hypercycle

The mathematical analysis of dynamical systems using methods of differential topology, yields the result that there is only one type of mechanism which fulfills the following requirements: The information stored in each single replicative unit (or reproductive cycle) must be maintained, i.e., the respective master copies must compete favorably with their error distributions. Despite their competitive behavior these units must establish a cooperation which includes all functionally integrated species. On the other hand, the cycle as a whole must continue to compete strongly with any other single entity or linked ensemble which does not contribute to its integrated function. These requirements are crucial for a selection of the best adapted functionally linked ensemble and its evolutive optimization. Only

hypercyclic organizations are able to fulfil these requirements. Non-cycle linkages among the autonomous reproduction cycles, such as chains or branched, tree-like networks are devoid of such properties. The mathematical methods used for proving these assertions are fixed-point, Lyapunov- and trajectory analysis in higher-dimensional phase spaces, spanned by the concentration coordinates of the cooperating partners. The self-organizing properties of hypercycles are elucidated, using analytical as well as numerical techniques.

Preview on Part C: The Realistic Hypercycle

A realistic model of a hypercycle relevant with respect to the origin of the genetic code and the translation machinery is presented. It includes the following features referring to natural systems: 1) The hypercycle has a self-enriching circle structure to admit an organization with finite probability under prebiotic conditions. 2) It permits a continuous emergence from closely interrelated (t-RNA-like) precursors, originally being members of a stable RNA quasi-species and having been amplified to a level of higher abundance. 3) The organizational structure and the properties of single functional units of this hypercycle are still reflected in the present genetic code in the translation apparatus of the prokaryotic cell as well as in certain bacterial viruses.

I. The Paradigm of Unity and Diversity in Evolution

Why do millions of species, plants and animals, exist, while there is only one basic molecular machinery of the cell: one universal genetic code and unique chiralities of the macromolecules?

The geneticists of our day would not hesitate to give an immediate answer to the first part of this question. Diversity of species is the outcome of the tremendous branching process of evolution with its myriads of single steps of reproduction and mutation. It in-

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Molecular Quasi-Species¹

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The molecular quasi-species model describes the physicochemical organization of monomers into an ensemble of heteropolymers with combinatorial complexity by ongoing template polymerization. Polynucleotides belong to the simplest class of such molecules. The quasi-species itself represents the stationary distribution of macromolecular sequences maintained by chemical reactions effecting error-prone replication and by transport processes. It is obtained deterministically, by mass-action kinetics, as the dominant eigenvalue of a culture matrix, W , which is derived directly from chemical rate coefficients, but it also exhibits stochastic features, being composed to a significant fraction of unique individual macromolecular sequences. The quasi-species model demonstrates how macromolecular information originates through specific nonequilibrium autocatalytic reactions and thus forms a bridge between reaction kinetics and molecular evolution. Selection and evolutionary optimization appear as new features in physical chemistry. Concentration bias in the production of mutants is a new concept in population genetics, relevant to frequently mutating populations, which is shown to greatly enhance the optimization properties. The present theory relates to asexually replicating ensembles, but this restriction is not essential. A sharp transition is exhibited between a drifting population of essentially random macromolecular sequences and a localized population of close relatives. This transition at a threshold error rate was found to depend on sequence lengths, distributions of selective values, and population sizes. It has been determined generically for complex landscapes and for special cases, and, it was shown to persist generally in the presence of nearly neutral mutants. Replication dynamics has much in common with the equilibrium statistics of complex spin systems: the error threshold is equivalent to a magnetic order-disorder transition. A rational function of the replication accuracy plays the role of temperature. Experimental data obtained from test-tube evolution of polynucleotides and from studies of natural virus populations support the quasi-species model. The error threshold seems to set a limit to the genome lengths of several classes of RNA viruses. In addition, the results are relevant even in eucaryotes where they contribute to the exon-intron debate.

1. Molecular Selection

Our knowledge of physical and chemical systems is, in a final analysis, based on models derived from repeatable experiments. While none of the classic and rather besieged list of properties rounded up to support the intuition of a distinction between the living and nonliving—metabolism, self-reproduction, irritability, and adaptability, for example—intrinsically limit the application of the scientific method, a determining role by unique or individual entities comes into conflict with the requirement of repeatability. Combinatorial variety, such as that in heteropolymers based on even very small numbers of different bases, even just two, readily provides numbers of different entities so enormous that neither consecutive nor parallel physical realization is possible. The physical chemistry of finite systems of such macromolecules must deal with both known regularities and the advent of unique copolymeric sequences. Normally this would present no difficulty in a statistical mechanical analysis of typical behavior, where rare events play no significant role, but with autocatalytic polymerization processes even unique single molecules may be amplified to determine the fate of the entire system. Potentially creative, self-organizing around unique events, the dynamics of this simplest living chemical system is invested with regularities that both allow and limit efficient adaptation. The quasi-species model is a study of these regularities.

The fundamental regularity in living organisms that has invited explanation is adaptation. Why are organisms so well fitted to their environments? At a more chemical level, why are enzymes

optimal catalysts? Darwin's theory of natural selection has provided biologists with a framework for the answer to this question. The present model is constructed along Darwinian lines but in terms of specific macromolecules, chemical reactions, and physical processes that make the notion of survival of the fittest precise. Not only does the model give an understanding of the physical limitations of adaptation, but also it provides new insight into the role of chance in the process. For an understanding of the structure of this minimal chemical model it is first necessary to recall the conceptual basis of Darwin's theory.

Darwin recognized that new inheritable adaptive properties were not induced by the environment but arose independently in the production of offspring. Lasting adaptive changes in a population could only come about by natural selection of the heritable traits or genotype based on the full characteristics or phenotype relevant for producing offspring. A process of chance, i.e., uncorrelated with the developed phenotype, controls changes in the genotype from one generation to the next and generates the diversity necessary for selection. Three factors have probably prevented chemists from gaining a clear insight into these phenomena in the past, despite the discovery of the polymeric nature of the genotype (DNA): the complexity of a minimum replication phenotype, the problem of dealing with a huge number of variants, and the nonequilibrium nature of these ongoing processes.

The formulation of a tractable chemical model based on Darwin's principle may be understood in several steps:

1. The major constituents of the system have to be inherently self-reproductive. Only two classes of molecules are presently

¹This is an abridged account of the quasi-species theory that has been submitted in comprehensive form to *Advances in Chemical Physics*.

(1) Eigen, M.; McCaskill, J. S.; Schuster, P. *Adv. Chem. Phys.*, in press.

1977

1988

Chemical kinetics of molecular evolution (continued)



Preface

Antiviral strategy on the horizon

Error catastrophe had its conceptual origins in the middle of the XXth century, when the consequences of mutations on enzymes involved in protein synthesis, as a theory of aging. In those times biological processes were generally perceived differently from today. Infectious diseases were regarded as a fleeting nuisance which would be eliminated through the use of antibiotics and antiviral agents. Microbial variation, although known in some cases, was not thought to be a significant problem for disease control. Variation in differentiated organisms was seen as resulting essentially from exchanges of genetic material associated with sexual reproduction. The problem was to unveil the mechanisms of inheritance, expression of genetic information and metabolism. Few saw that genetic change is occurring at present in all organisms, and still fewer recognized Darwinian principles as essential to the biology of pathogenic viruses and cells. Population geneticists rarely used bacteria or viruses as experimental systems to define concepts in biological evolution. The extent of genetic polymorphism among individuals of the same biological species came as a surprise when the first results on comparison of electrophoretic mobility of enzymes were obtained. With the advent of *in vitro* DNA recombination, and rapid nucleic acid sequencing techniques, molecular analyses of genomes reinforced the conclusion of extreme inter-individual genetic variation within the same species. Now, due largely to spectacular progress in comparative genomics, we see cellular DNAs, both prokaryotic and eukaryotic, as highly dynamic. Most cellular processes, including such essential information-bearing and transferring events as genome replication, transcription and translation, are increasingly perceived as inherently inaccurate. Viruses, and in particular RNA viruses, are among the most extreme examples of exploitation of replication inaccuracy for survival.

Error catastrophe, or the loss of meaningful genetic information through excess genetic variation, was formulated in quantitative terms as a consequence of quasispecies theory, which was first developed to explain self-organization and adaptability of primitive replicons in early stages of life. Recently, a conceptual extension of error catastrophe that could be defined as “induced genetic deterioration” has emerged as

a possible antiviral strategy. This is the topic of the current special issue of *Virus Research*.

Few would nowadays doubt that one of the major obstacles for the control of viral disease is short-term adaptability of viral pathogens. Adaptability of viruses follows the same Darwinian principles that have shaped biological evolution over eons, that is, repeated rounds of reproduction with genetic variation, competition and selection, often perturbed by random events such as statistical fluctuations in population size. However, with viruses the consequences of the operation of these very same Darwinian principles are felt within very short times. Short-term evolution (within hours and days) can be also observed with some cellular pathogens, with subsets of normal cells, and cancer cells. The nature of RNA viral pathogens begs for alternative antiviral strategies, and forcing the virus to cross the critical error threshold for maintenance of genetic information is one of them.

The contributions to this volume have been chosen to reflect different lines of evidence (both theoretical and experimental) on which antiviral designs based on genetic deterioration inflicted upon viruses are being constructed. Theoretical studies have explored the copying fidelity conditions that must be fulfilled by any information-bearing replication system for the essential genetic information to be transmitted to progeny. Closely related to the theoretical developments have been numerous experimental studies on quasispecies dynamics and their multiple biological manifestations. The latter can be summarized by saying that RNA viruses, by virtue of existing as mutant spectra rather than defined genetic entities, remarkably expand their potential to overcome selective pressures intended to limit their replication. Indeed, the use of antiviral inhibitors in clinical practice and the design of vaccines for a number of major RNA virus-associated diseases, are currently presided by a sense of uncertainty. Another line of growing research is the enzymology of copying fidelity by viral replicases, aimed at understanding the molecular basis of mutagenic activities. Error catastrophe as a potential new antiviral strategy received an important impulse by the observation that ribavirin (a licensed antiviral nucleoside analogue) may be exerting, in some systems, its antiviral activity through enhanced mutagenesis.

This has encouraged investigations on new mutagenic base analogues, some of them used in anticancer chemotherapy. Some chapters summarize these important biochemical studies on cell entry pathways and metabolism of mutagenic agents, that may find new applications as antiviral agents.

This volume intends to be basically a progress report, an introduction to a new avenue of research, and a realistic appraisal of the many issues that remain to be investigated. In this respect, I can envisage (not without many uncertainties) at least three lines of needed research: (i) One on further understanding of quasispecies dynamics in infected individuals to learn more on how to apply combinations of virus-specific mutagens and inhibitors in an effective way, finding synergistic combinations and avoiding antagonistic ones as well as severe clinical side effects. (ii) Another on a deeper understanding of the metabolism of mutagenic agents, in particular base and nucleoside analogues. This includes identification of the transporters that carry them into cells, an understanding of their metabolic processing, intracellular stability and alterations of nucleotide pools, among other issues. (iii) Still another line of needed research is the development of new mutagenic agents specific for viruses, showing no (or limited) toxicity for cells. Some advances may come from links with anticancer research, but others should result from the designs of new molecules, based on the structures of viral polymerases. I really hope that the reader finds this issue not only to be an interesting and useful review of the current situation in the field, but also a stimulating exposure to the major problems to be faced.

The idea to prepare this special issue came as a kind invitation of Ulrich Desselberger, former Editor of *Virus Research*, and then taken enthusiastically by Luis Enjuanes, recently appointed as Editor of *Virus Research*. I take this opportunity to thank Ulrich, Luis and the Editor-in-Chief of *Virus Research*, Brian Mahy, for their continued interest and support to the research on virus evolution over the years.

My thanks go also to the 19 authors who despite their busy schedules have taken time to prepare excellent manuscripts, to Elsevier staff for their prompt responses to my requests, and, last but not least, to Ms. Lucía Horrillo from Centro de Biología Molecular “Severo Ochoa” for her patient dealing with the correspondence with authors and the final organization of the issue.

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Esteban Domingo
1943 -

Review

Quasispecies Made Simple

Vol.1(6), e61, 2005,
pp.450 – 460.

J. J. Bull, Lauren Ancel Meyers, Michael Lachmann*

JOURNAL OF VIROLOGY, Mar. 2007, p. 2930–2939
0022-538X/07/\$08.00+0 doi:10.1128/JVI.01624-06
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Vol. 81, No. 6

Theory of Lethal Mutagenesis for Viruses^{∇†}

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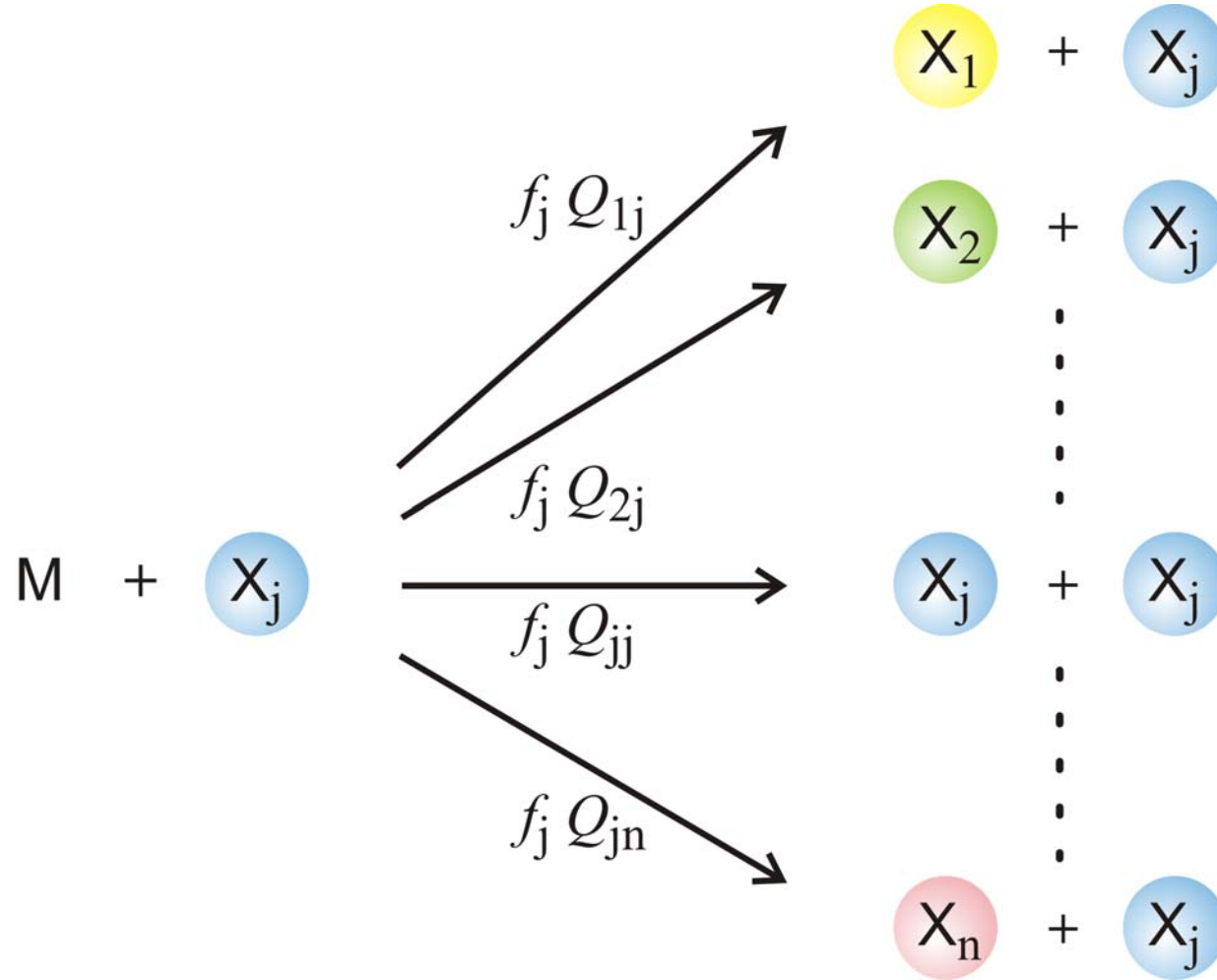
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Mutation is the basis of adaptation. Yet, most mutations are detrimental, and elevating mutation rates will impair a population's fitness in the short term. The latter realization has led to the concept of lethal mutagenesis for curing viral infections, and work with drugs such as ribavirin has supported this perspective. As yet, there is no formal theory of lethal mutagenesis, although reference is commonly made to Eigen's error catastrophe theory. Here, we propose a theory of lethal mutagenesis. With an obvious parallel to the epidemiological threshold for eradication of a disease, a sufficient condition for lethal mutagenesis is that each viral genotype produces, on average, less than one progeny virus that goes on to infect a new cell. The extinction threshold involves an evolutionary component based on the mutation rate, but it also includes an ecological component, so the threshold cannot be calculated from the mutation rate alone. The genetic evolution of a large population undergoing mutagenesis is independent of whether the population is declining or stable, so there is no runaway accumulation of mutations or genetic signature for lethal mutagenesis that distinguishes it from a level of mutagenesis under which the population is maintained. To detect lethal mutagenesis, accurate measurements of the genome-wide mutation rate and the number of progeny per infected cell that go on to infect new cells are needed. We discuss three methods for estimating the former. Estimating the latter is more challenging, but broad limits to this estimate may be feasible.

Error threshold versus lethal mutagenesis

1. Complexity in molecular evolution
2. The error threshold
3. Simple landscapes and error thresholds
4. ‚Realistic‘ fitness landscapes
5. Quasispecies on realistic landscapes
6. Neutrality and consensus sequences

1. **Complexity in molecular evolution**
2. The error threshold
3. Simple landscapes and error thresholds
4. ‚Realistic‘ fitness landscapes
5. Quasispecies on realistic landscapes
6. Neutrality and consensus sequences



Chemical kinetics of replication and mutation as parallel reactions

$$\frac{dx_j}{dt} = \sum_{i=1}^n W_{ji} x_i - x_j \Phi = \sum_{i=1}^n Q_{ji} f_i x_i - x_j \Phi; \quad j=1,2,\dots,n$$

$$\Phi = \sum_{i=1}^n f_i x_i / \sum_{i=1}^n x_i$$

Decomposition of matrix W

$$W = \begin{pmatrix} w_{11} & w_{12} & \dots & w_{1n} \\ w_{21} & w_{22} & \dots & w_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ w_{n1} & w_{n2} & \dots & w_{nn} \end{pmatrix} = \mathbf{Q} \cdot \mathbf{F} \text{ with}$$

$$\mathbf{Q} = \begin{pmatrix} Q_{11} & Q_{12} & \dots & Q_{1n} \\ Q_{21} & Q_{22} & \dots & Q_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ Q_{n1} & Q_{n2} & \dots & Q_{nn} \end{pmatrix} \text{ and } \mathbf{F} = \begin{pmatrix} f_1 & 0 & \dots & 0 \\ 0 & f_2 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & f_n \end{pmatrix}$$

Factorization of the value matrix W separates **mutation** and **fitness** effects.

Mutation-selection equation: $[I_i] = x_i \geq 0, f_i \geq 0, Q_{ij} \geq 0$

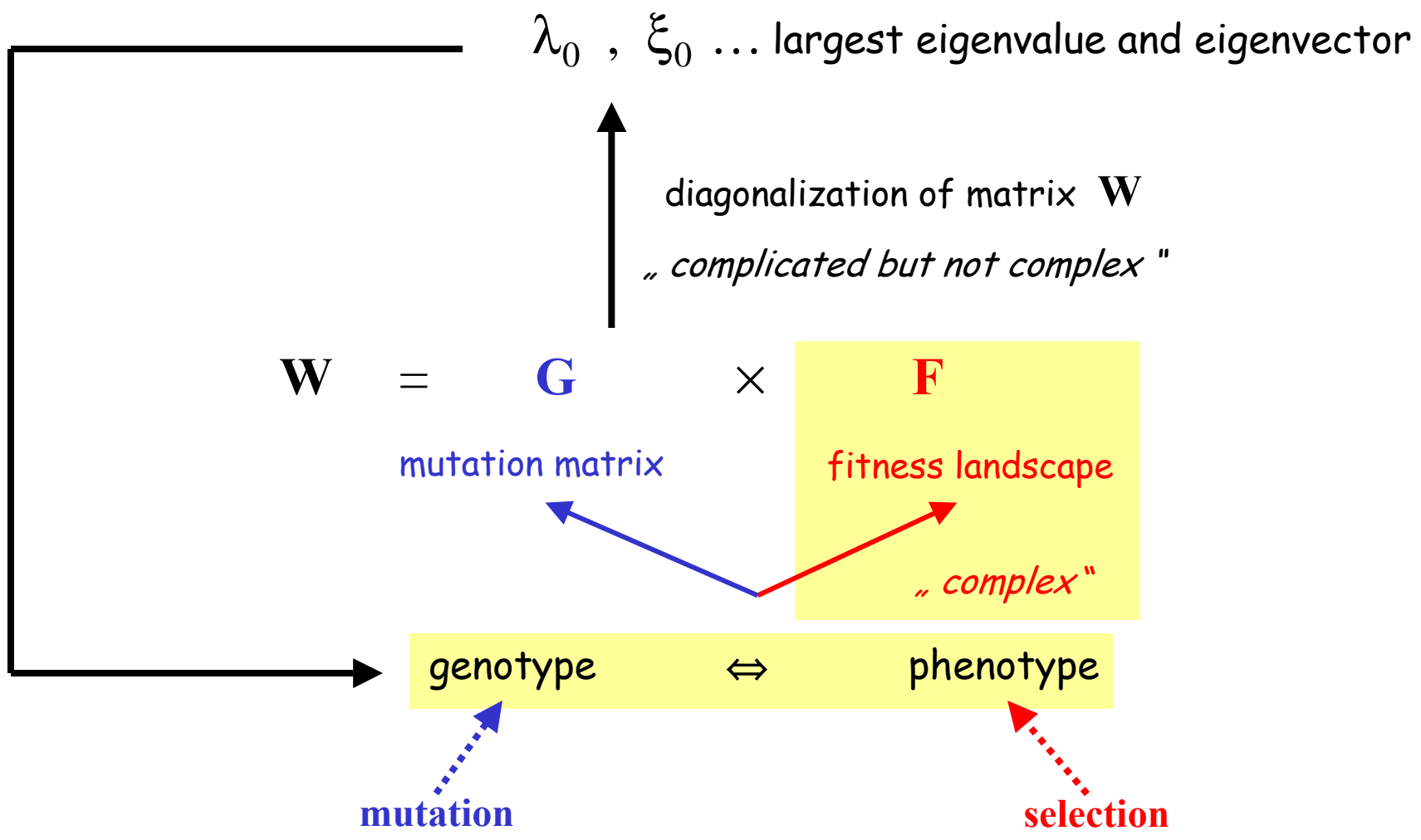
$$\frac{dx_i}{dt} = \sum_{j=1}^n Q_{ij} f_j x_j - x_i \phi, \quad i=1,2,\dots,n; \quad \sum_{i=1}^n x_i = 1; \quad \phi = \sum_{j=1}^n f_j x_j = \bar{f}$$

solutions are obtained after integrating factor transformation by means of an eigenvalue problem

$$x_i(t) = \frac{\sum_{k=0}^{n-1} \ell_{ik} \cdot c_k(0) \cdot \exp(\lambda_k t)}{\sum_{j=1}^n \sum_{k=0}^{n-1} \ell_{jk} \cdot c_k(0) \cdot \exp(\lambda_k t)}; \quad i=1,2,\dots,n; \quad c_k(0) = \sum_{i=1}^n h_{ki} x_i(0)$$

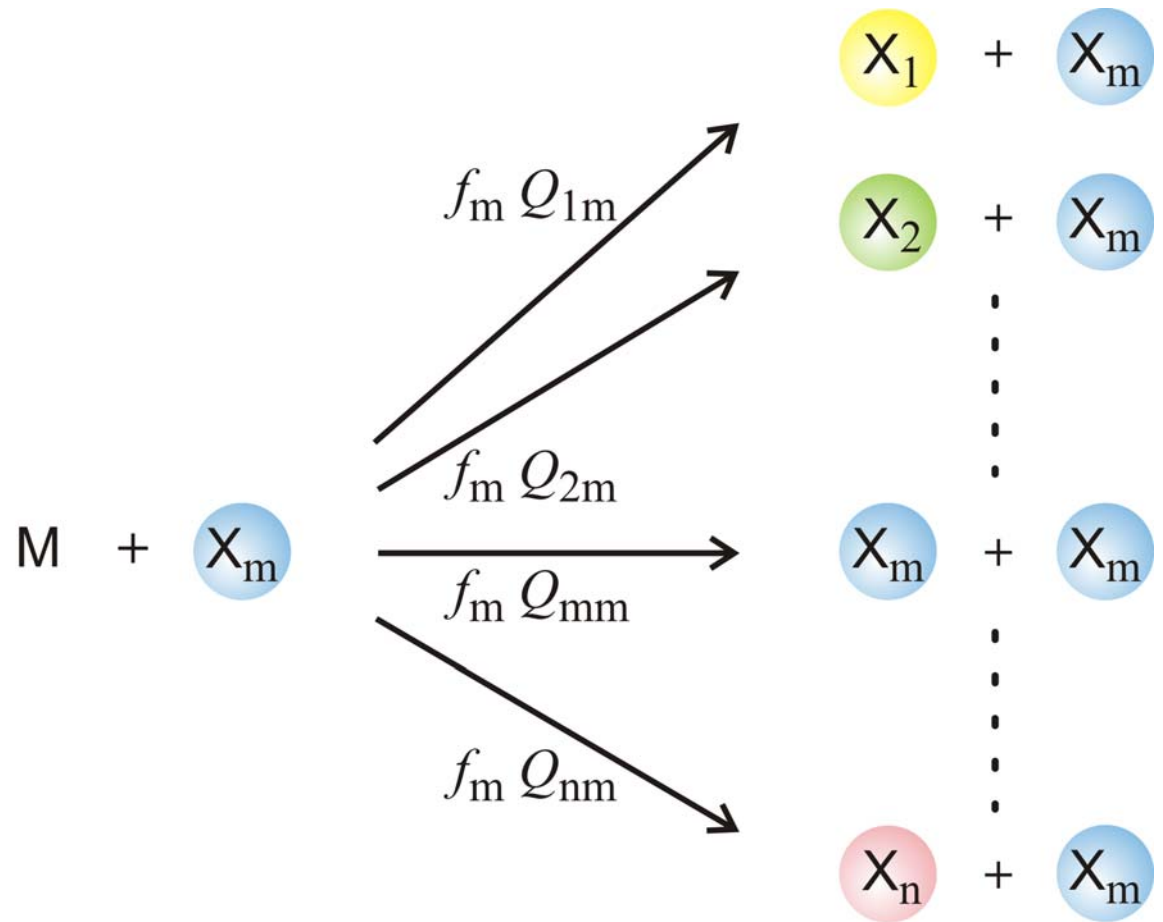
$$W \div \{f_i Q_{ij}; i, j=1,2,\dots,n\}; \quad L = \{\ell_{ij}; i, j=1,2,\dots,n\}; \quad L^{-1} = H = \{h_{ij}; i, j=1,2,\dots,n\}$$

$$L^{-1} \cdot W \cdot L = \Lambda = \{\lambda_k; k=0,1,\dots,n-1\}$$

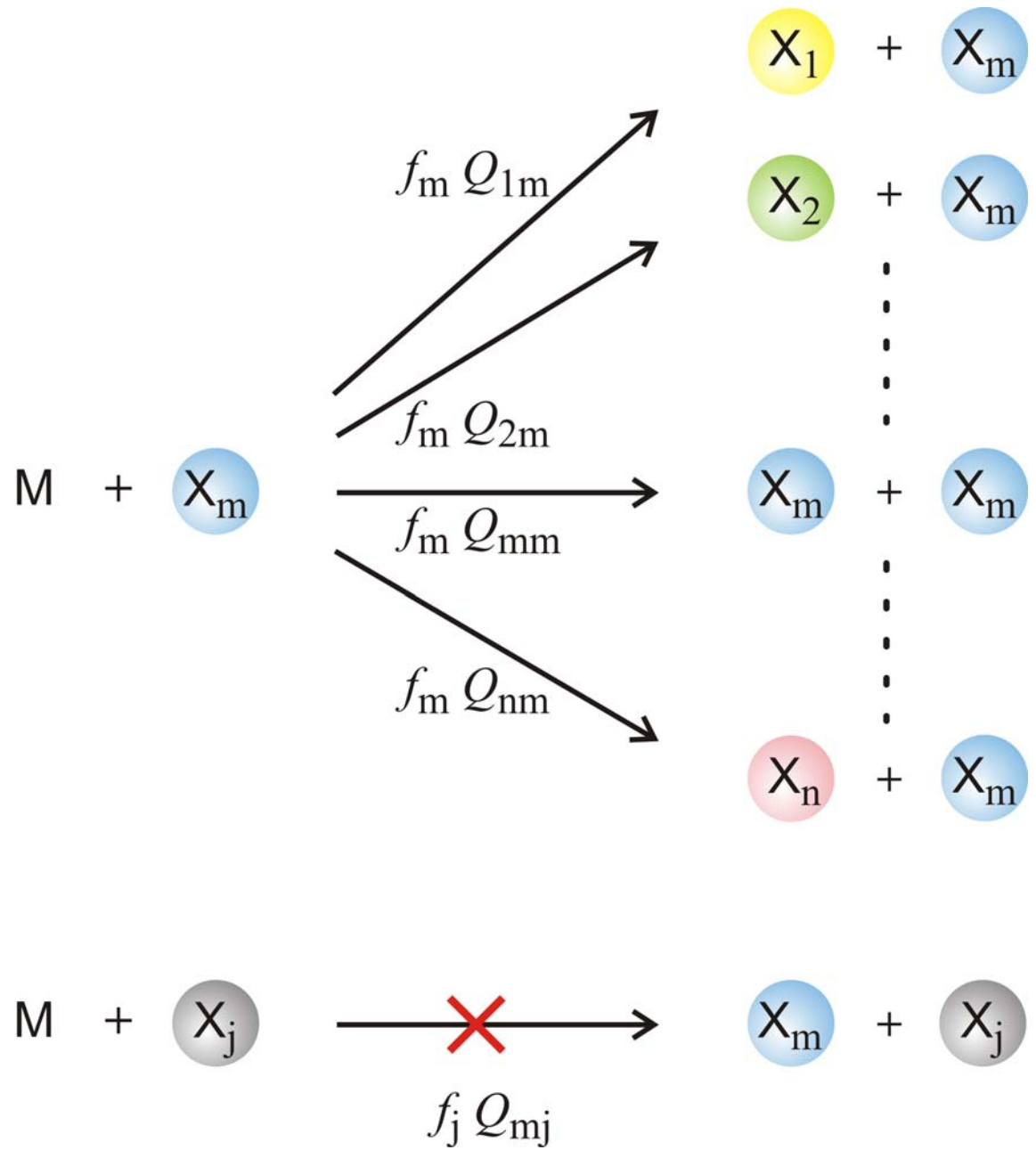


Complexity in molecular evolution

1. Complexity in molecular evolution
- 2. The error threshold**
3. Simple landscapes and error thresholds
4. ‚Realistic‘ fitness landscapes
5. Quasispecies on realistic landscapes
6. Neutrality and consensus sequences



The no-mutational backflow or zeroth order approximation



The no-mutational backflow or zeroth order approximation

$$\frac{dx_m^{(0)}}{dt} = x_m^{(0)} (Q_{mm} f_m - \phi(t)) = 0 \quad \text{and} \quad \phi(t) = Q_{mm} f_m$$

$$\bar{x}_m^{(0)} = \frac{Q_{mm} - \sigma_m^{-1}}{1 - \sigma_m^{-1}} = \frac{1}{\sigma_m - 1} (\sigma_m (1-p)^n - 1)$$

$$\bar{x}_m^{(0)} = 0 \quad \Rightarrow \quad (1-p)^n = \sigma_m^{-1} \quad \text{and} \quad p_{cr} \approx 1 - (\sigma_m)^{-1/n}$$

$$\sigma_m = \frac{f_m}{\bar{f}_{-m}} \quad \text{and} \quad \bar{f}_{-m} = \frac{1}{(1-x_m)} \sum_{i=1, i \neq m}^N x_i f_i$$

The ,no-mutational-backflow‘ or zeroth order approximation

Chain length and error threshold

$$Q_{mm} \cdot \sigma_m = (1-p)^n \cdot \sigma_m \geq 1 \Rightarrow n \cdot \ln(1-p) \geq -\ln \sigma_m$$

$$p \dots \text{constant} : n_{\max} \approx \frac{\ln \sigma_m}{p}$$

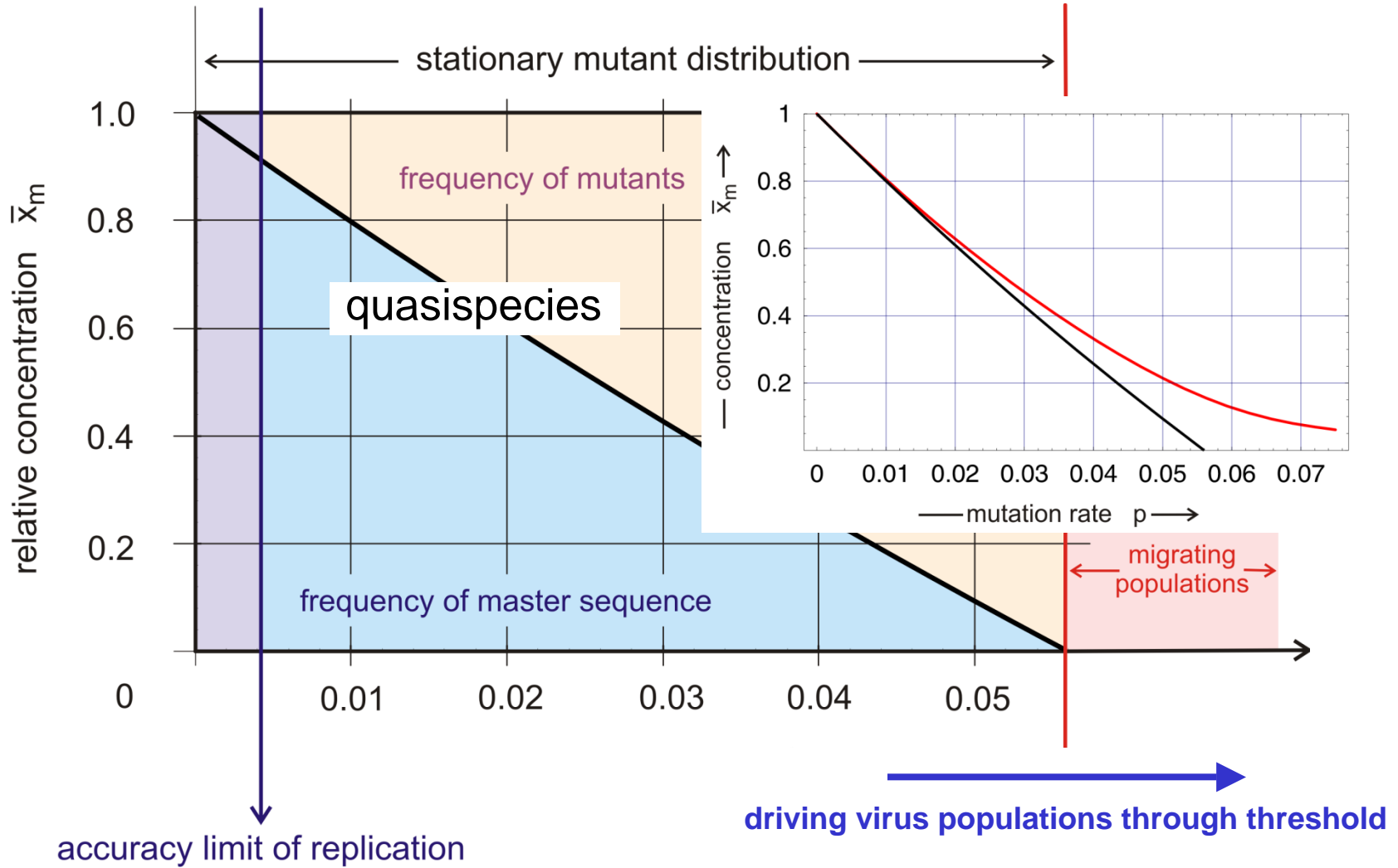
$$n \dots \text{constant} : p_{\max} \approx \frac{\ln \sigma_m}{n}$$

$Q_{mm} = (1-p)^n$... replication accuracy

p ... error rate

n ... chain length

$$\sigma_m = \frac{f_m}{\sum_{j \neq m} x_j f_j / (1 - x_m)} \dots \text{superiority of master sequence}$$



The error threshold in replication and mutation

1. Complexity in molecular evolution
2. The error threshold
- 3. Simple landscapes and error thresholds**
4. ‚Realistic‘ fitness landscapes
5. Quasispecies on realistic landscapes
6. Neutrality and consensus sequences

NOTES AND COMMENTS

SURFACES OF SELECTIVE VALUE REVISITED

Provine, in his generally favorable discussion of my shifting-balance theory of evolution, severely criticized the concept of "surfaces of selective value" (1986, p. 307). I think that he was looking for something more mathematical than was intended. Professor E. M. East, as organizer of the program of the Sixth International Congress of Genetics (held in 1932 in Ithaca, New York), had asked me to present a brief, nonmathematical account of the views on evolution that I had presented in a long (63-page) paper in 1931. I agreed to do this.

Most early geneticists thought of the phenotype as if it were a mosaic of unit characters, each determined by a single locus, with effects as conspicuous as those that they used in their experiments. They thought of alleles as having constant relative selective values. The consequences of this assumption were worked out most exhaustively by Haldane in a series of papers beginning in 1924 and summarized in 1932. In addition, he worked out less fully some of the consequences of various other assumptions, also summarized in this book.

Sewall Wright. 1931. Evolution in Mendelian populations.
Genetics 16:97-159.

-- --. 1932. The roles of mutation, inbreeding, crossbreeding,
and selection in evolution. In: D.F.Jones, ed. *Proceedings of
the Sixth International Congress on Genetics, Vol.I.* Brooklyn
Botanical Garden. Ithaca, NY, pp. 356-366.

-- --. 1988. Surfaces of selective value revisited.
The American Naturalist 131:115-131.

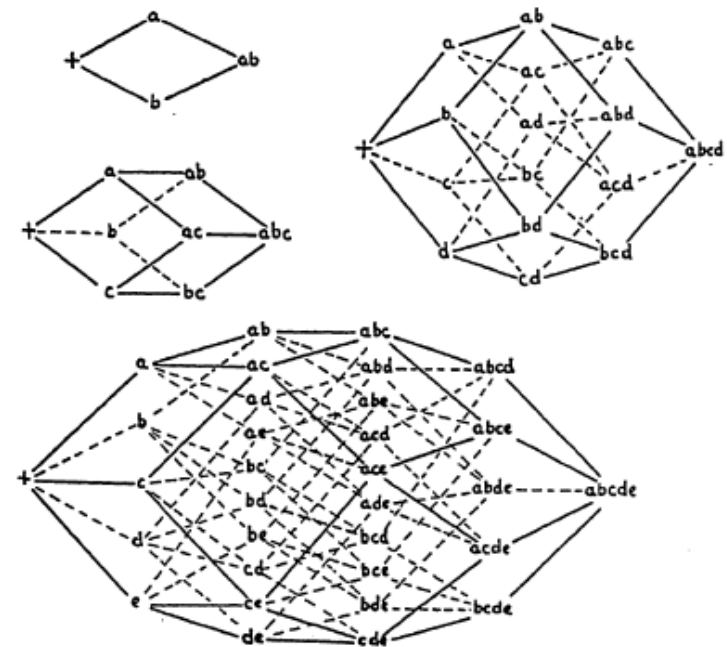


FIG. 1.—The combinations of from 2 to 5 paired allelomorphs.

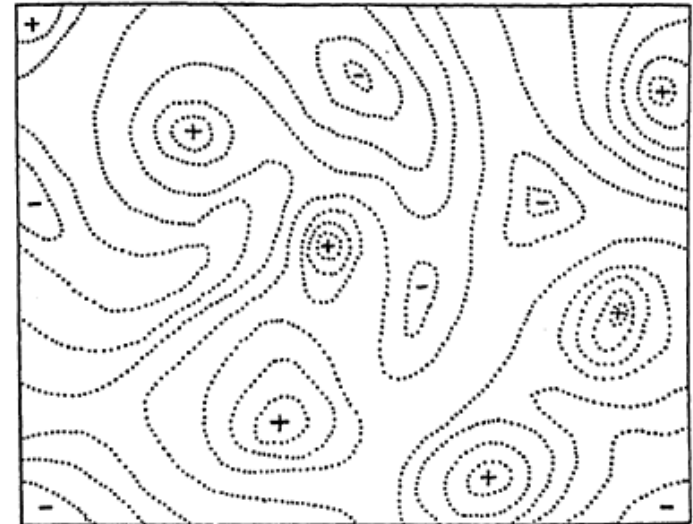
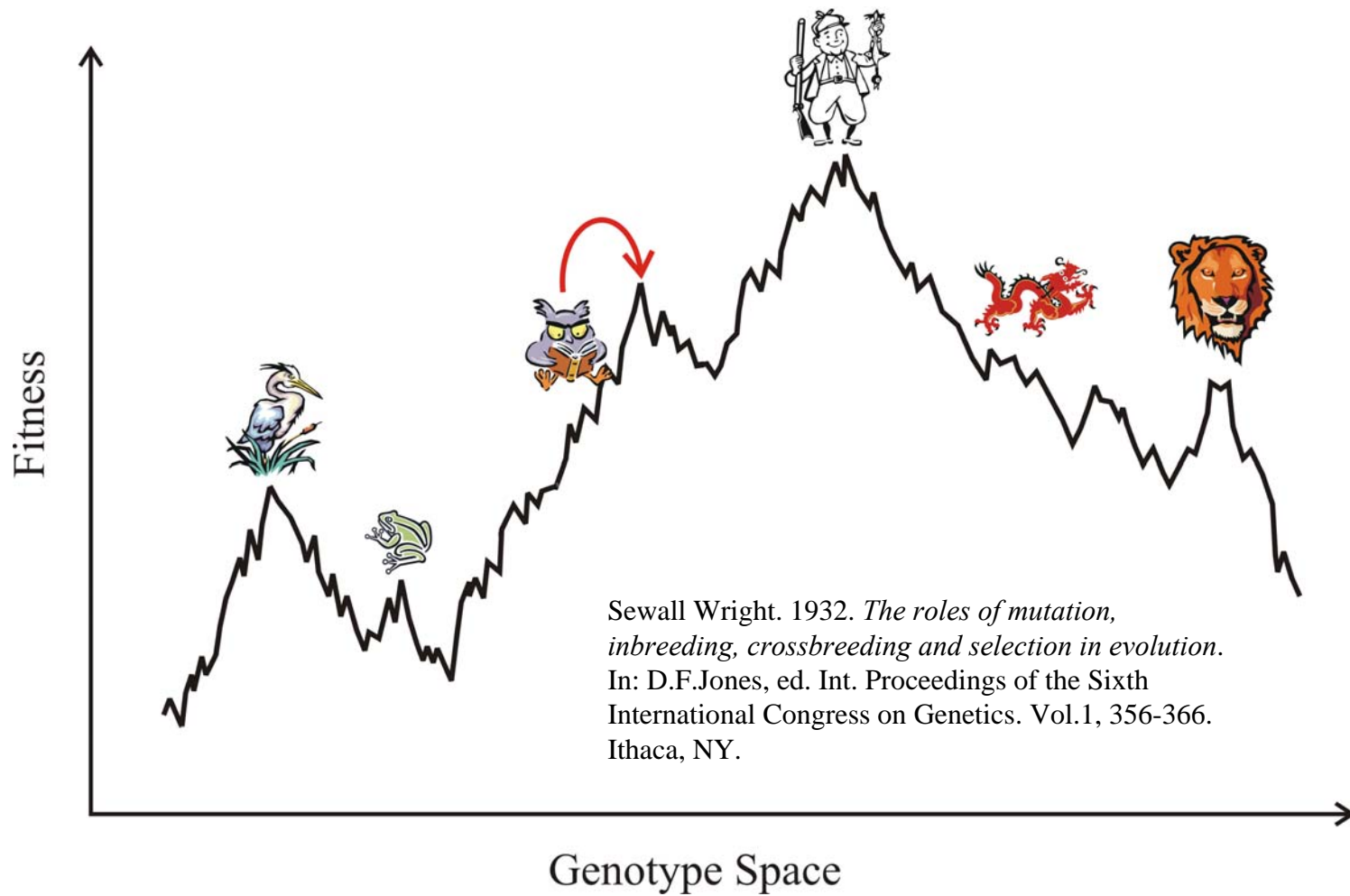
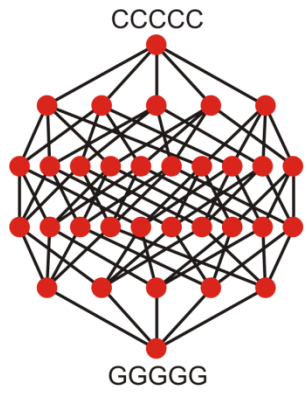


FIG. 2.—Diagrammatic representation of the field of gene combinations in two dimensions instead of many thousands. Dotted lines represent contours with respect to adaptiveness.



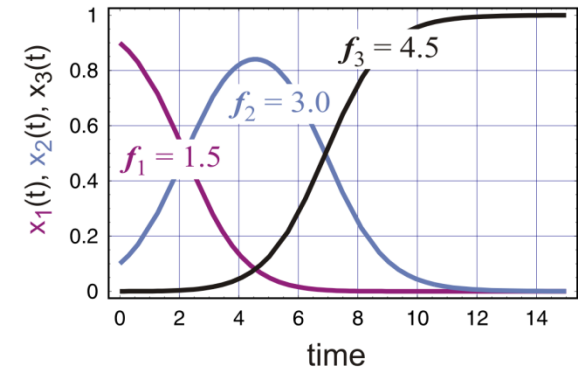
Sewall Wright's fitness landscape as metaphor for Darwinian evolution



sequence space

S

sequence

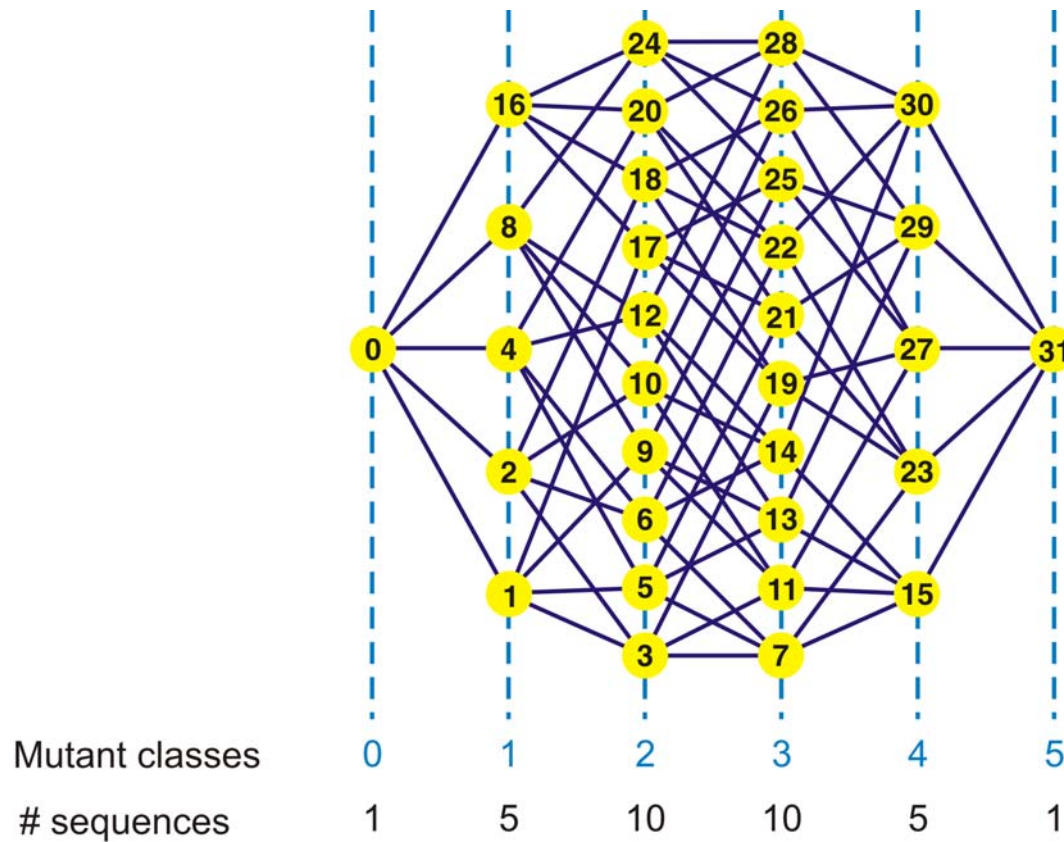


parameter space

$f = \Psi(Y)$

function

The landscape model



Binary sequences are encoded by their decimal equivalents:

C = 0 and **G** = 1, for example,

"0" ≡ 00000 = **CCCCC**,

"14" ≡ 01110 = **CGGGC**,

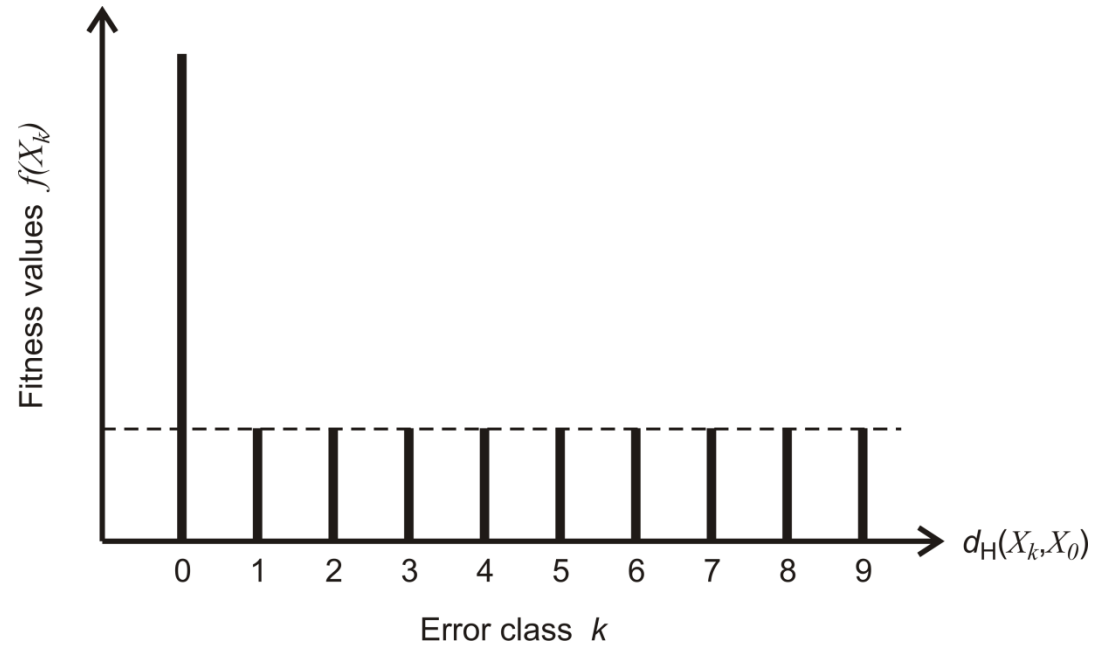
"29" ≡ 11101 = **GGGCG**, etc.

Concentrations of entire error classes: $[\Gamma_k] = y_k(p)$, $k = 0, 1, \dots, n$

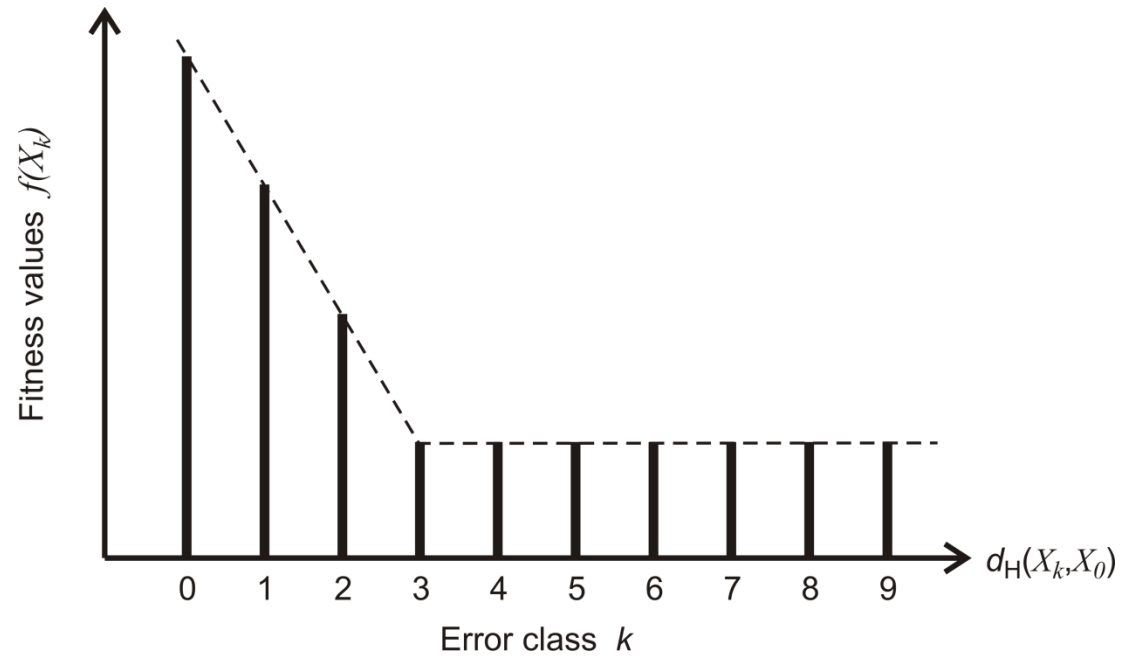
$$y_k(p) = \sum_{i=1, d_H(X_i, X_k)=k}^N x_i(p), \quad |\Gamma_k| = \binom{n}{k}$$

The simple landscape model

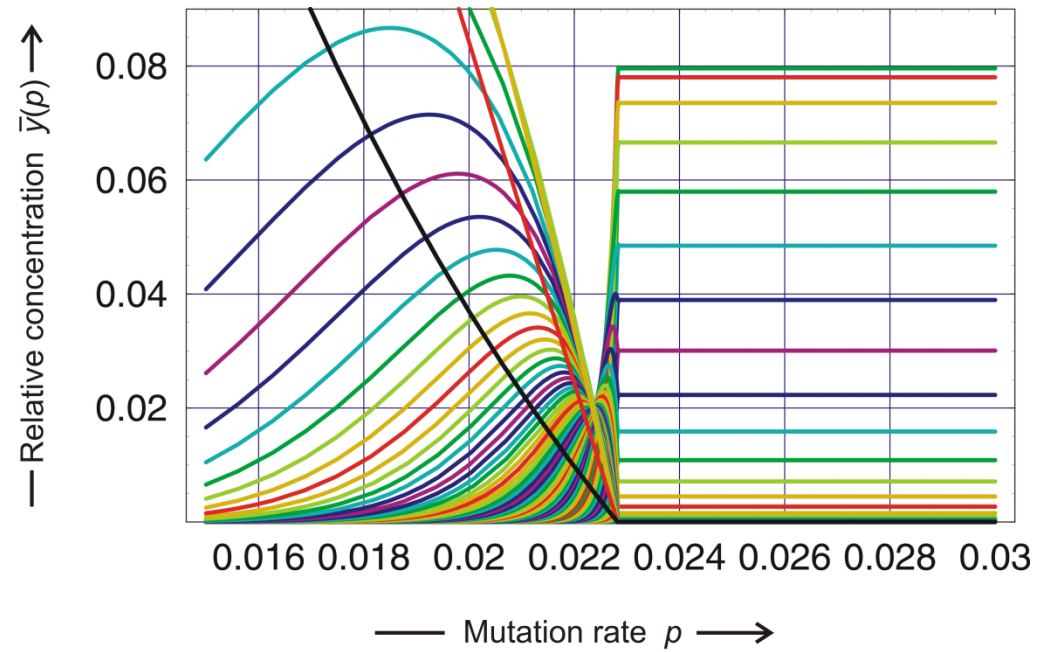
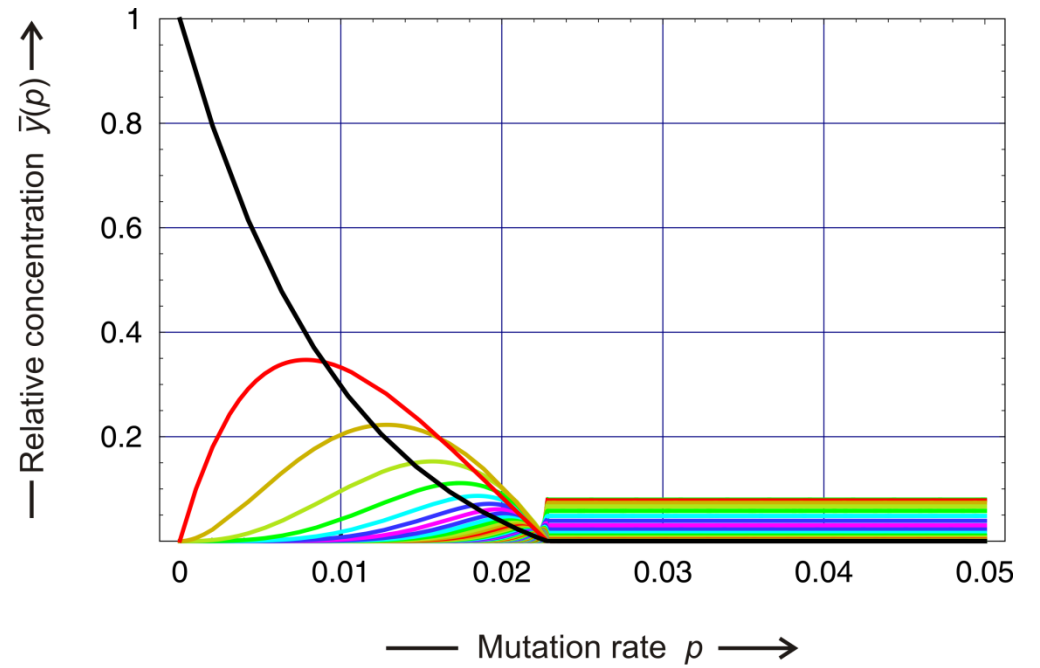
single peak landscape



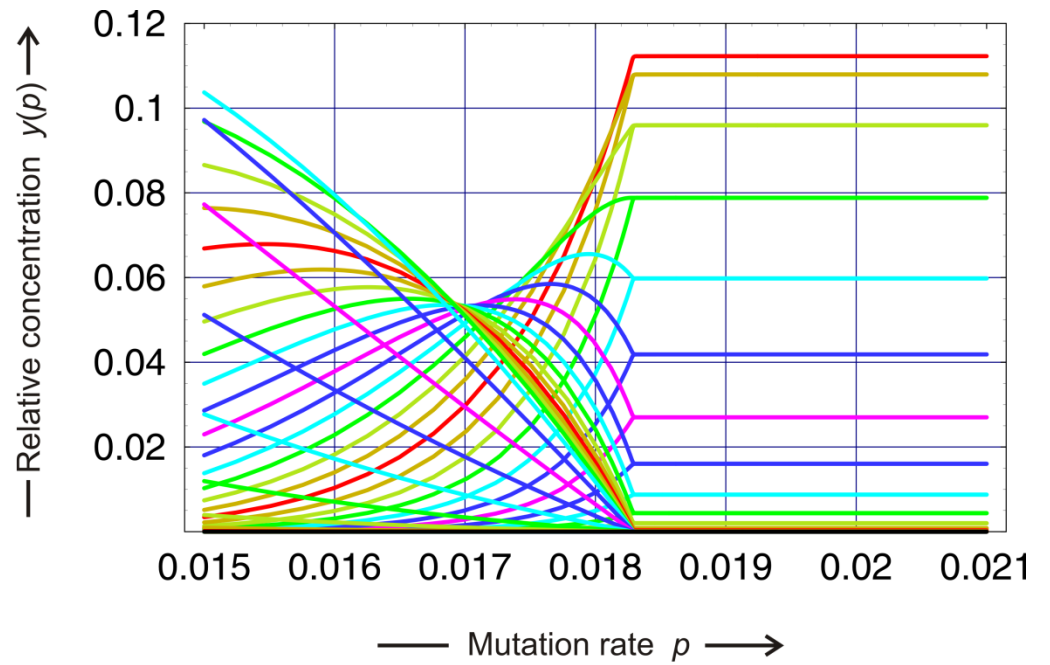
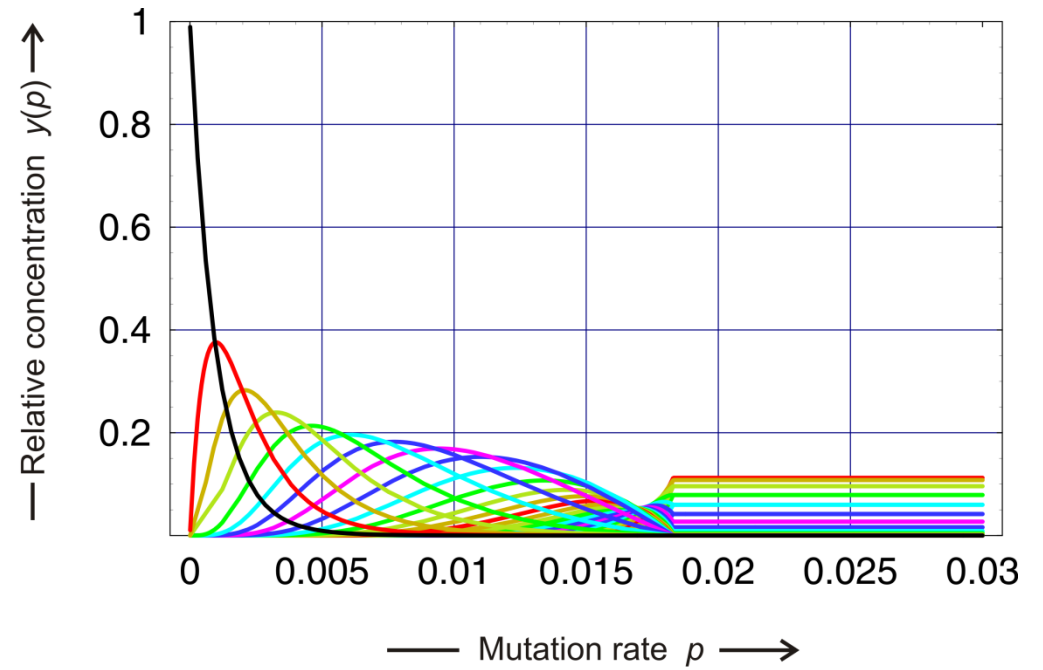
step linear landscape



Model fitness landscapes I

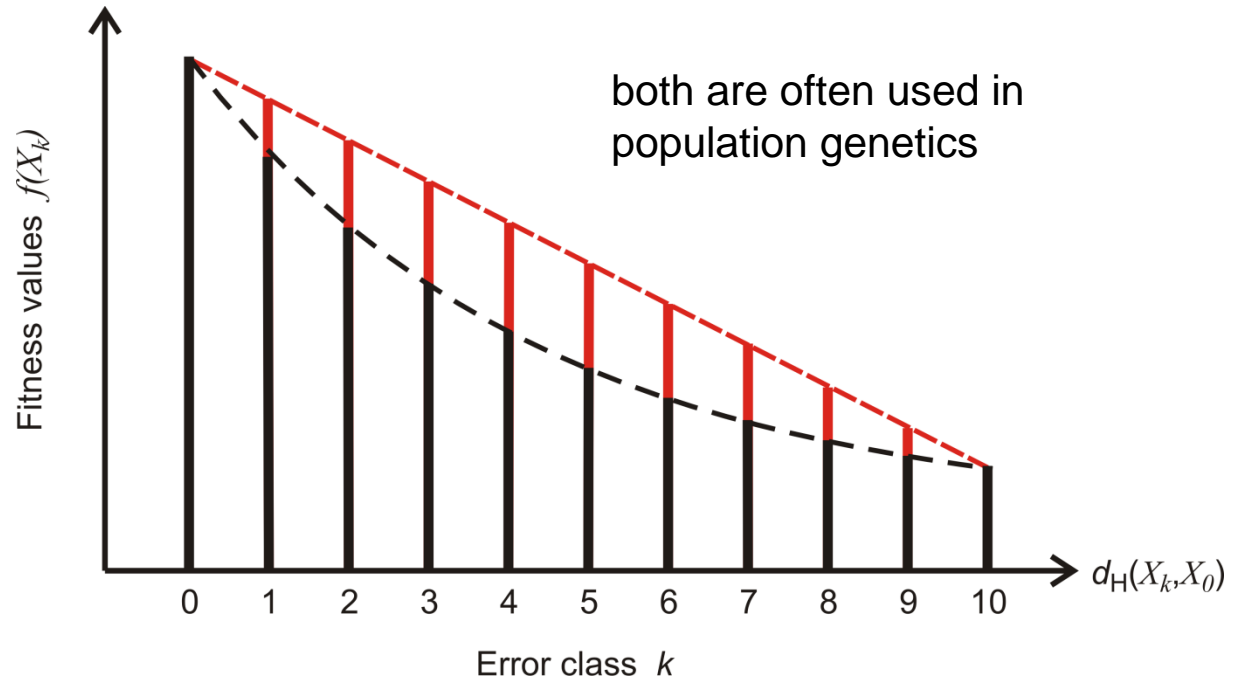


Error threshold on the single peak landscape

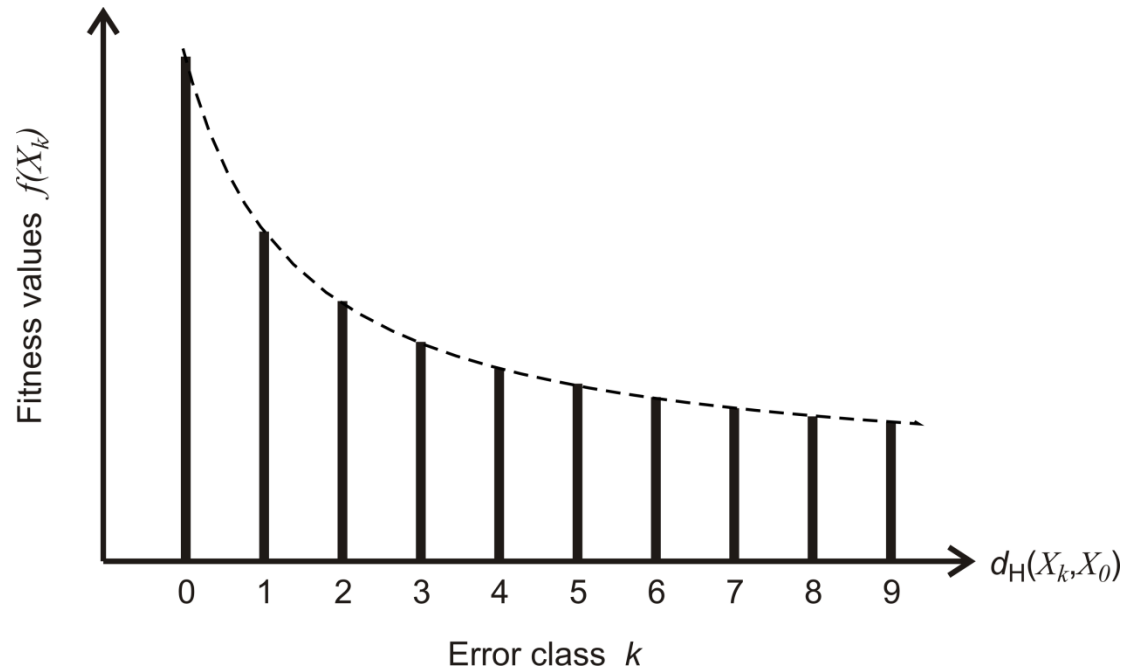


Error threshold on the step linear landscape

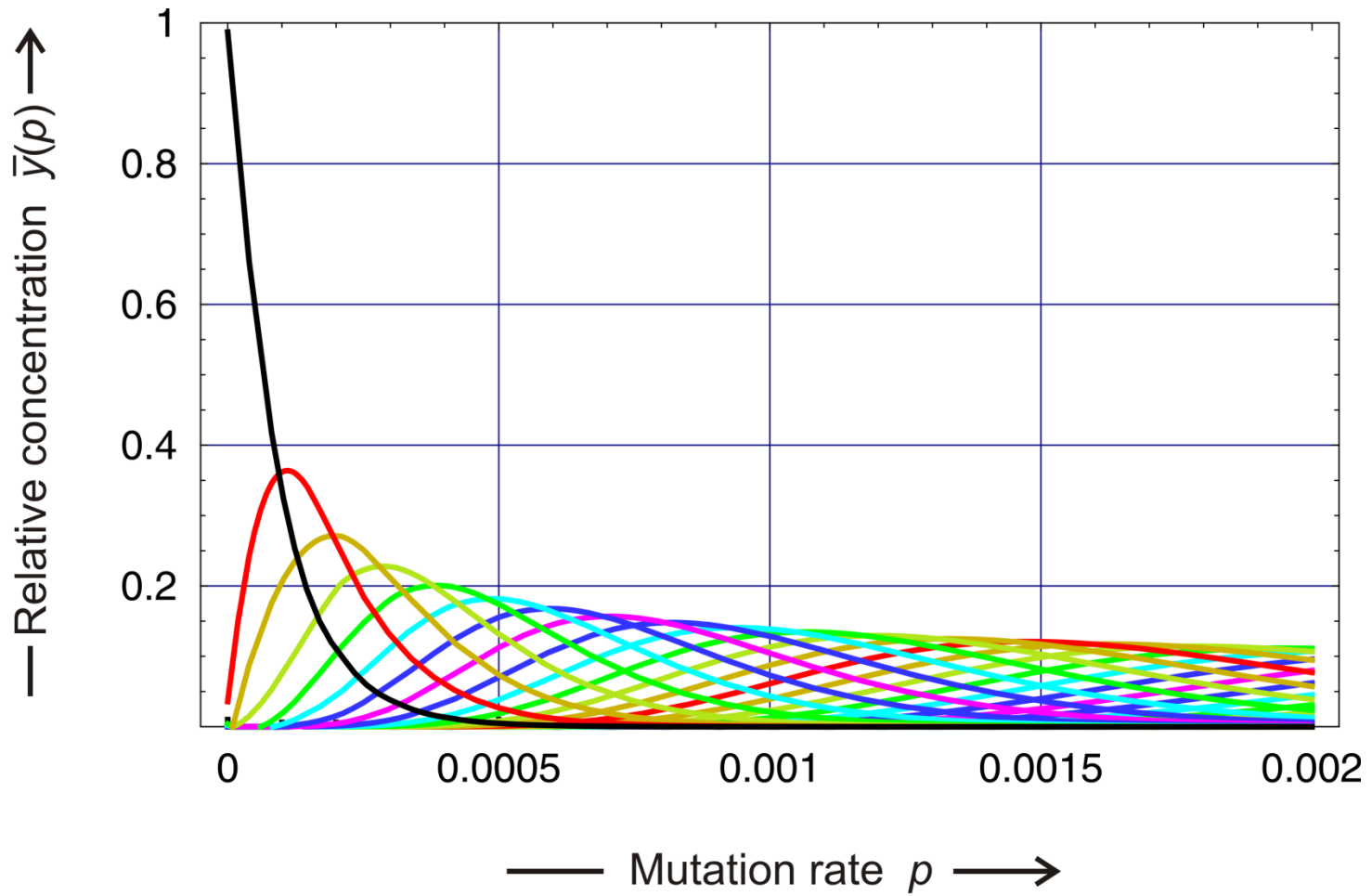
linear and
multiplicative



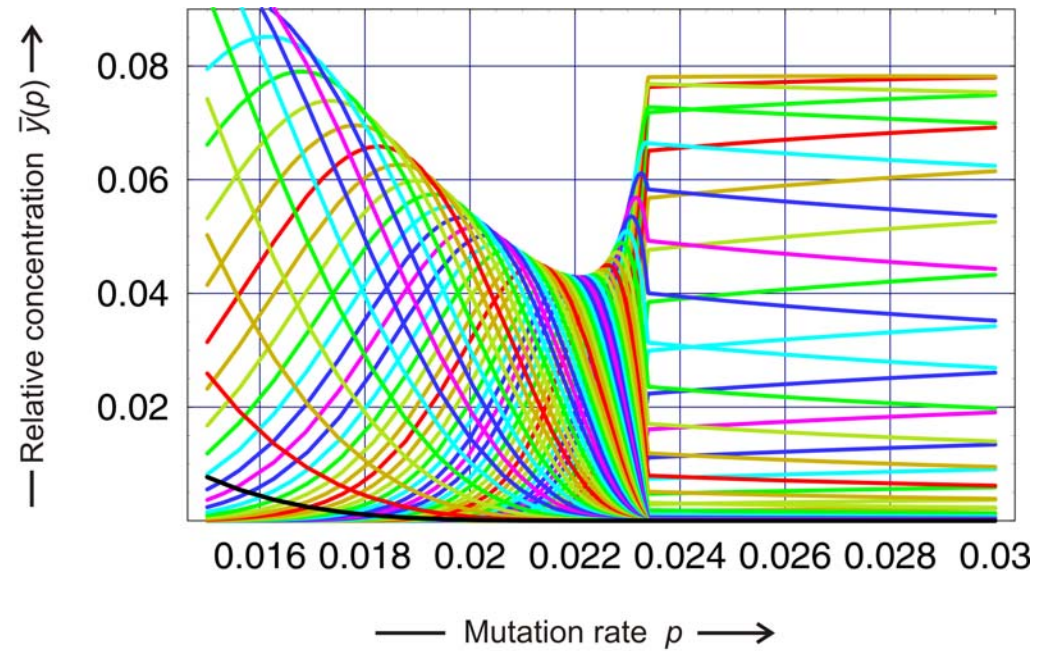
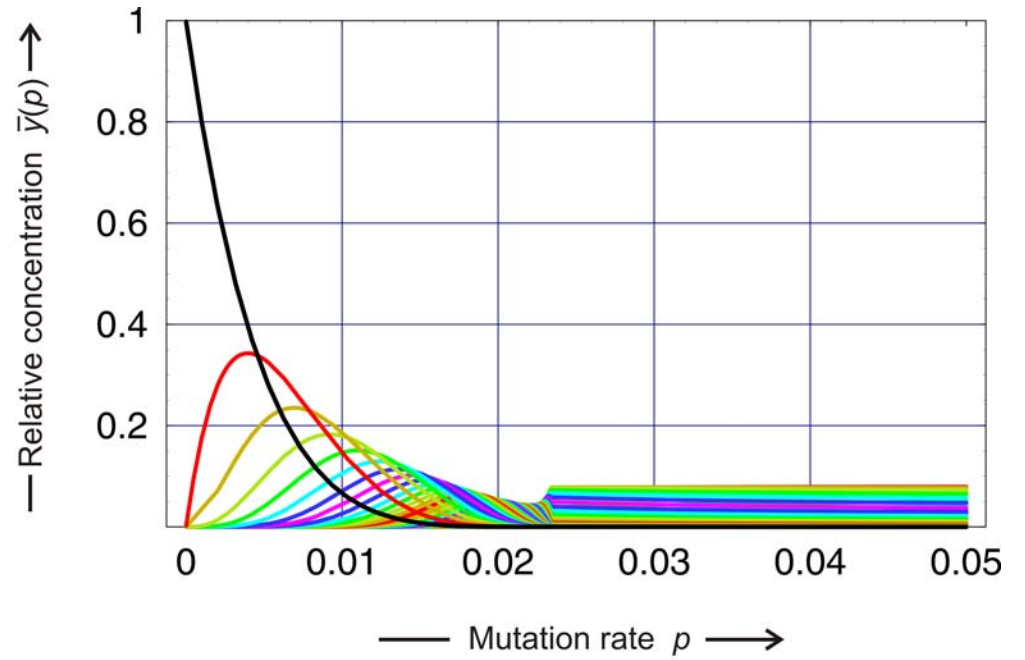
hyperbolic



Model fitness landscapes II



The linear fitness landscape shows no error threshold



Error threshold on the hyperbolic landscape

The error threshold can be separated into three phenomena:

1. **Steep decrease** in the concentration **of the master sequence** to very small values.
2. **Sharp change** in the stationary concentration of the **quasispecies** distribution.
3. **Transition to the uniform distribution** at small mutation rates.

The error threshold can be separated into three phenomena:

1. **Steep decrease** in the concentration **of the master sequence** to very small values.
2. **Sharp change** in the stationary concentration of the **quasispecies** distribution.
3. **Transition to the uniform distribution** at small mutation rates.

All three phenomena coincide for the quasispecies on the single peak fitness landscape.

Make things as simple as possible,
but not simpler !

Albert Einstein

Albert Einstein's razor, precise reference is unknown.

1. Complexity in molecular evolution
2. The error threshold
3. Simple landscapes and error thresholds
4. **„Realistic“ fitness landscapes**
5. Quasispecies on realistic landscapes
6. Neutrality and consensus sequences

Realistic fitness landscapes

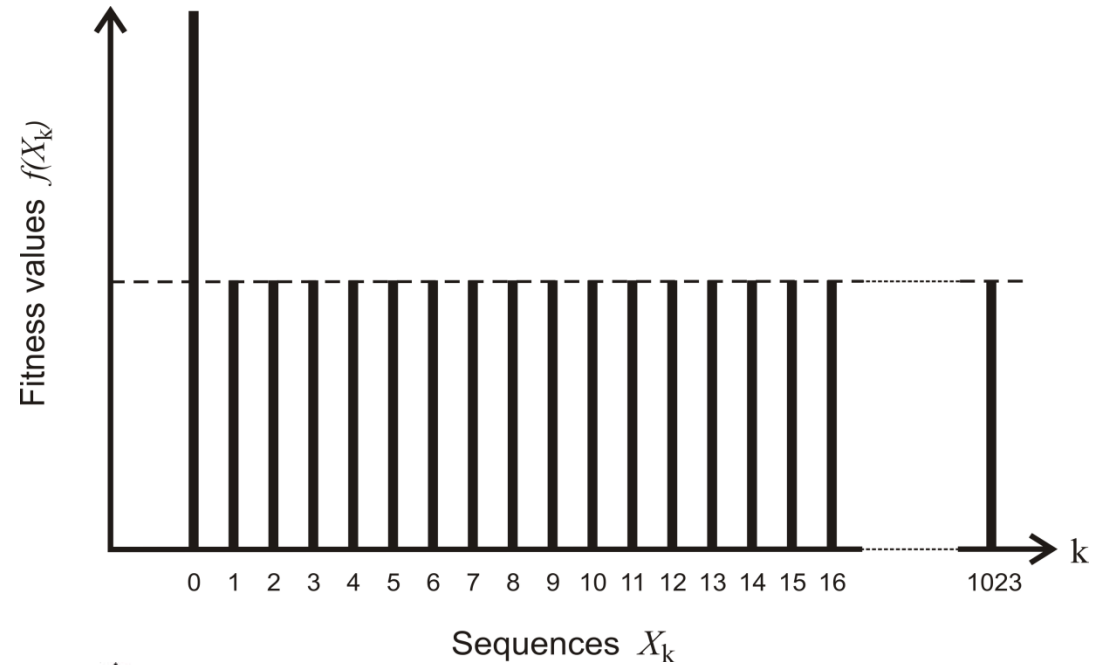
1. Ruggedness: nearby lying genotypes may develop into very different phenotypes

2. Neutrality: many different genotypes give rise to phenotypes with identical selection behavior

3. Combinatorial explosion: the number of possible genomes is prohibitive for systematic searches

Facit: Any successful and applicable theory of molecular evolution must be able to predict evolutionary dynamics from a small or at least in practice measurable number of fitness values.

single peak landscape



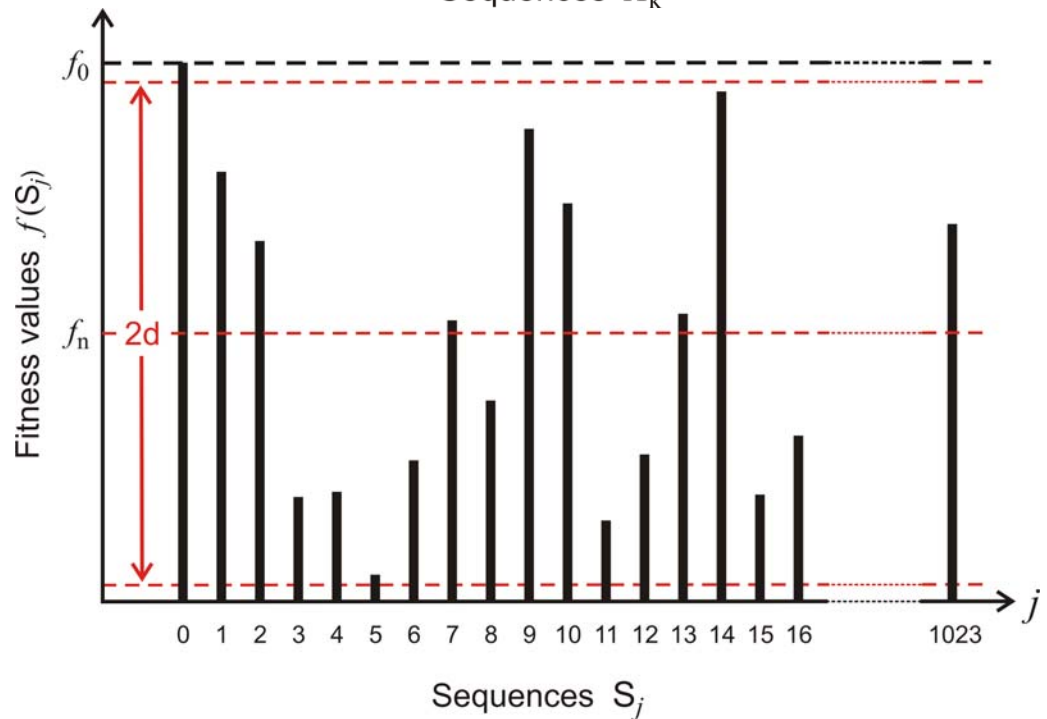
$$f(S_j) = f_n + 2d(f_0 - f_n) \left(\eta_j^{(s)} - 0.5 \right)$$

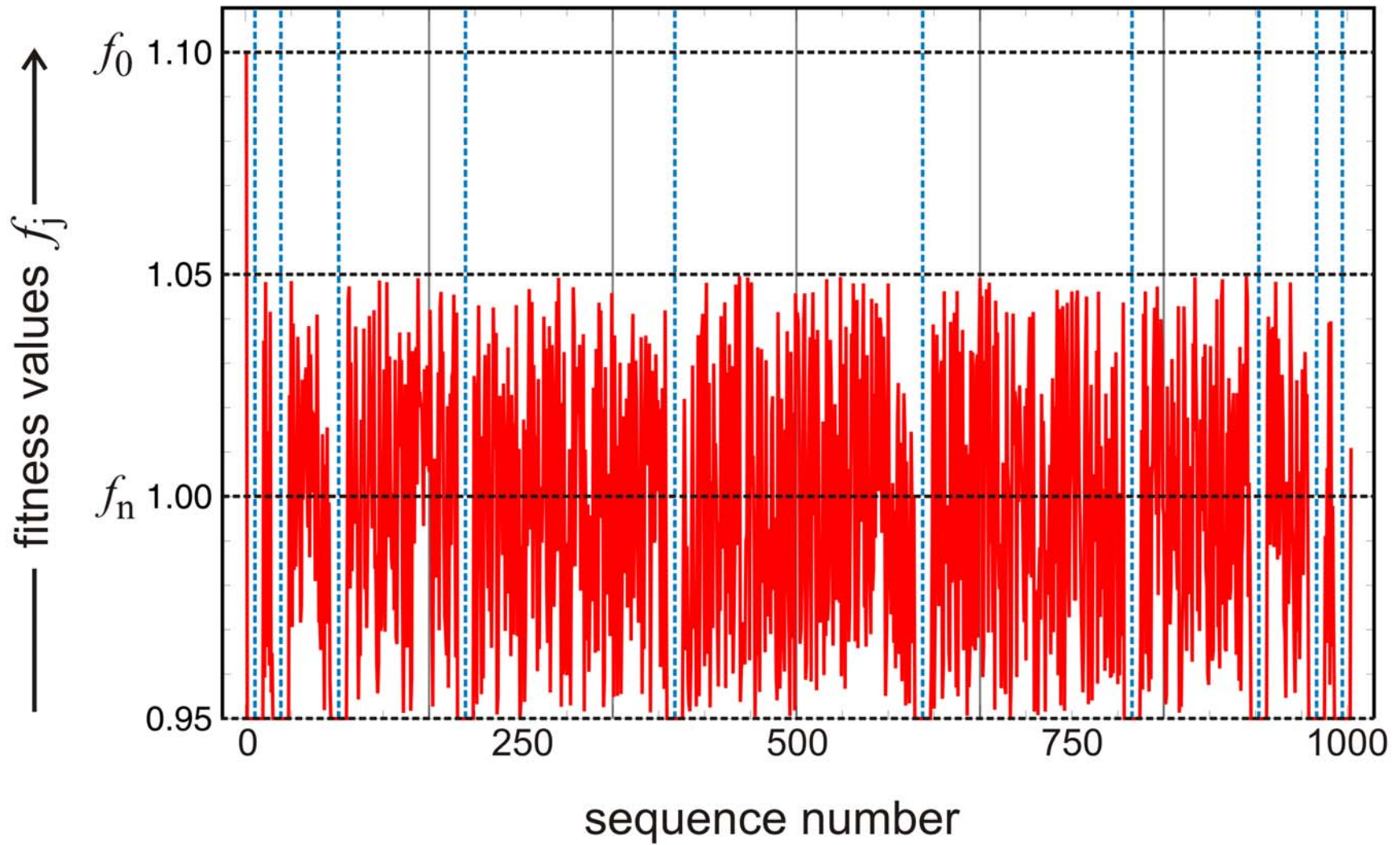
$$j = 1, 2, \dots, N; j \neq m,$$

η ... random number; s ... seeds

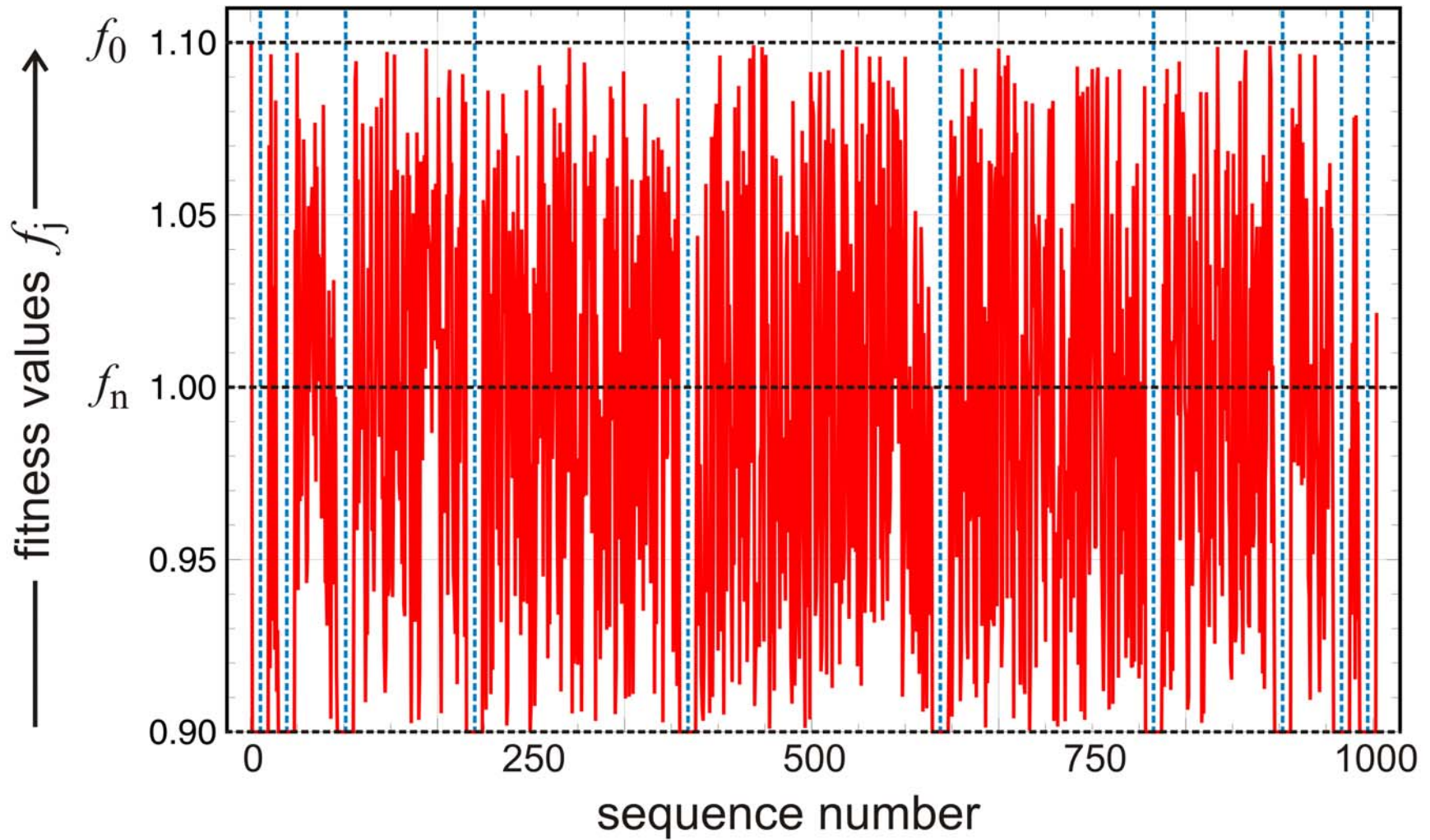
„realistic“ landscape

Rugged fitness landscapes
over individual binary sequences
with $n = 10$

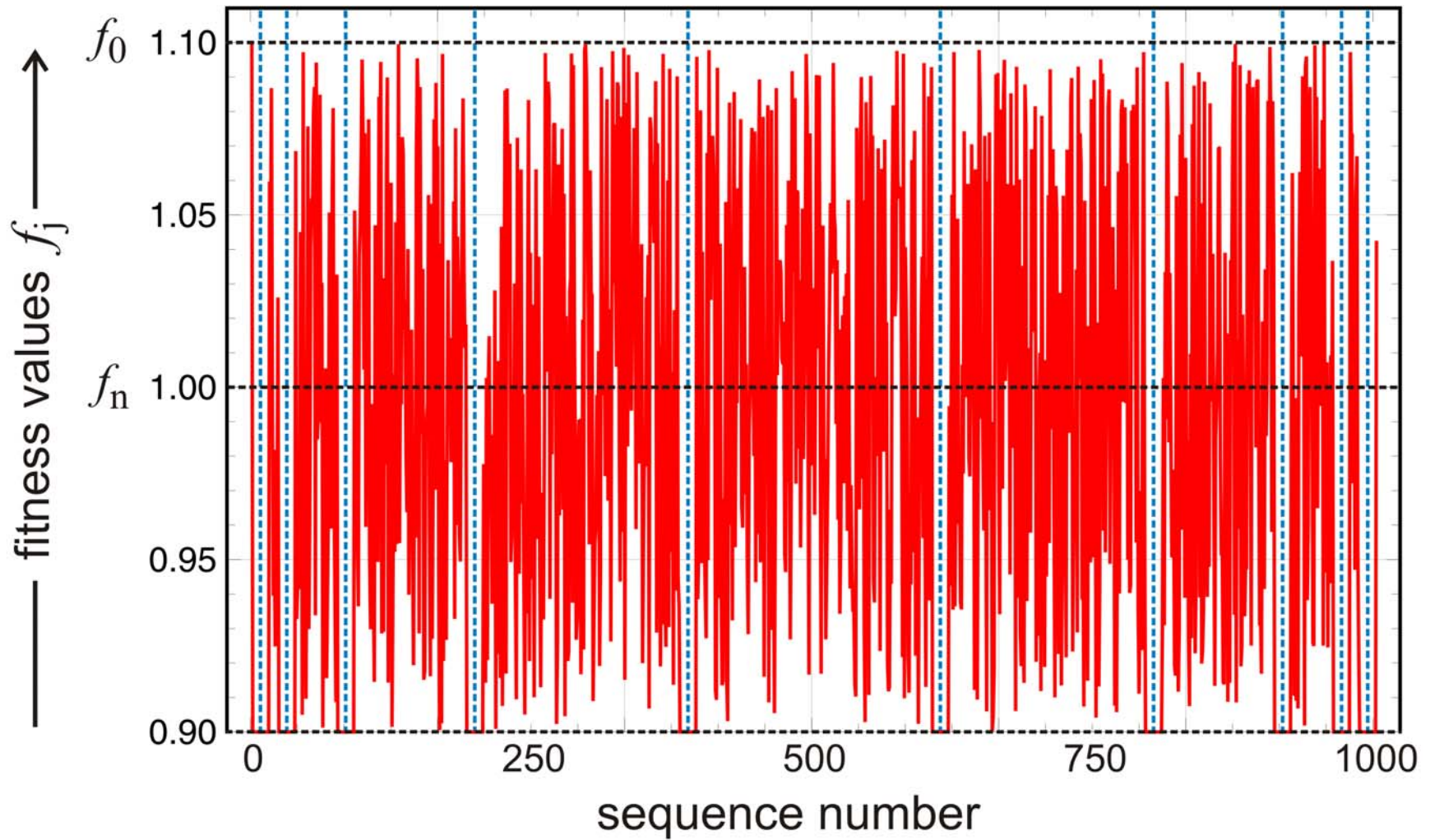




Random distribution of fitness values: $d = 0.5$ and $s = 919$

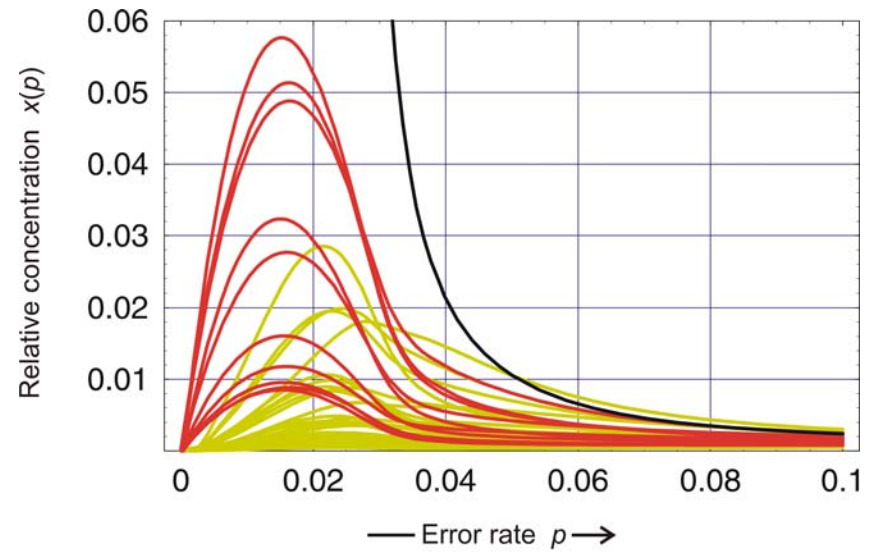
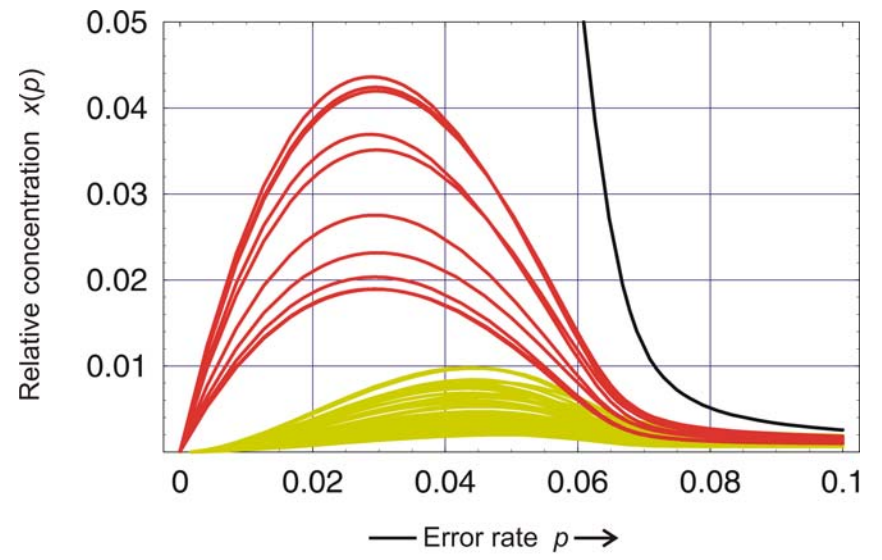
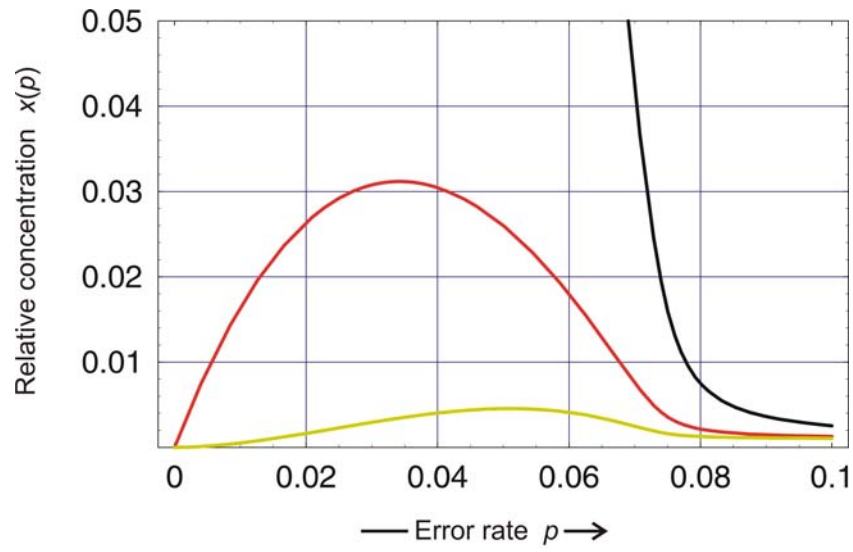


Random distribution of fitness values: $d = 1.0$ and $s = 919$



Random distribution of fitness values: $d = 1.0$ and $s = 637$

1. Complexity in molecular evolution
2. The error threshold
3. Simple landscapes and error thresholds
4. ‚Realistic‘ fitness landscapes
- 5. Quasispecies on realistic landscapes**
6. Neutrality and consensus sequences

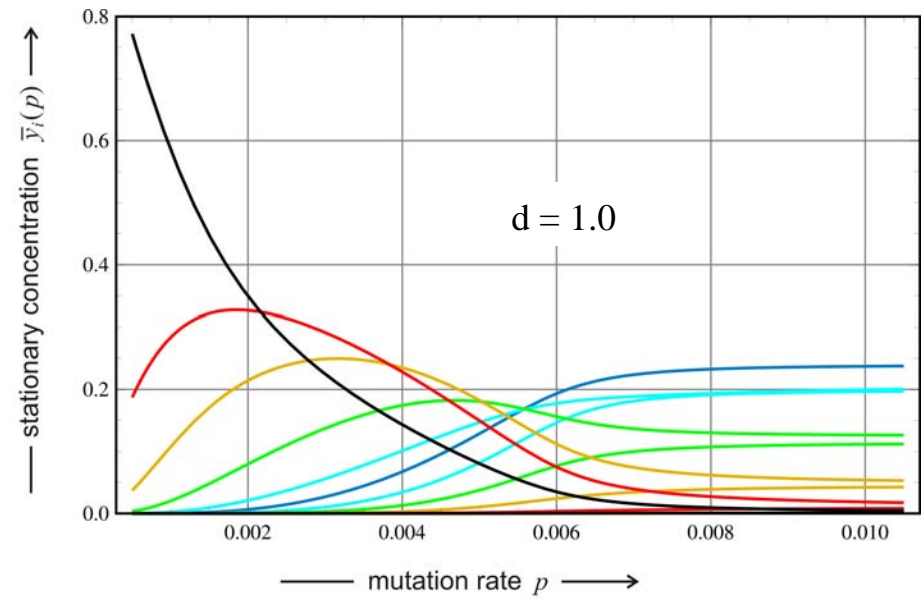
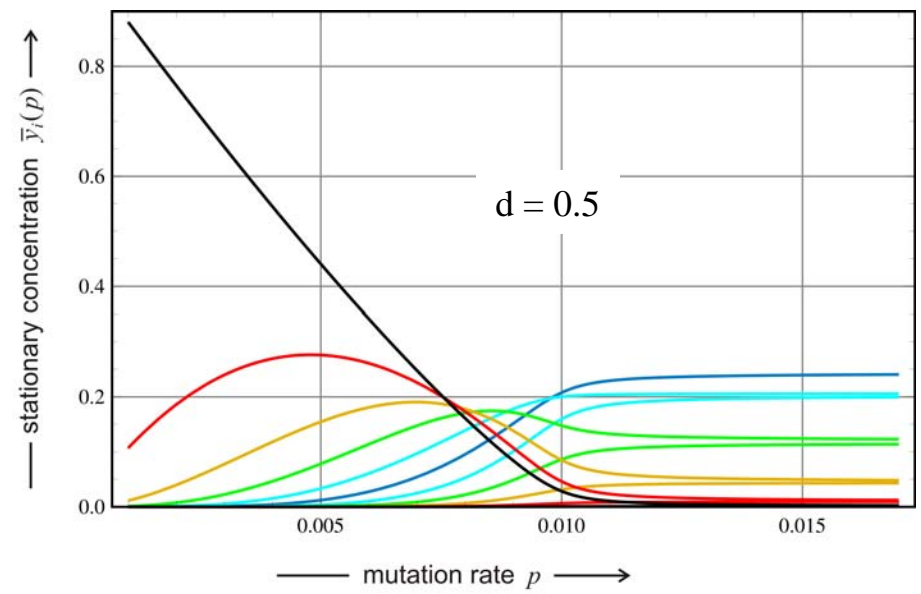
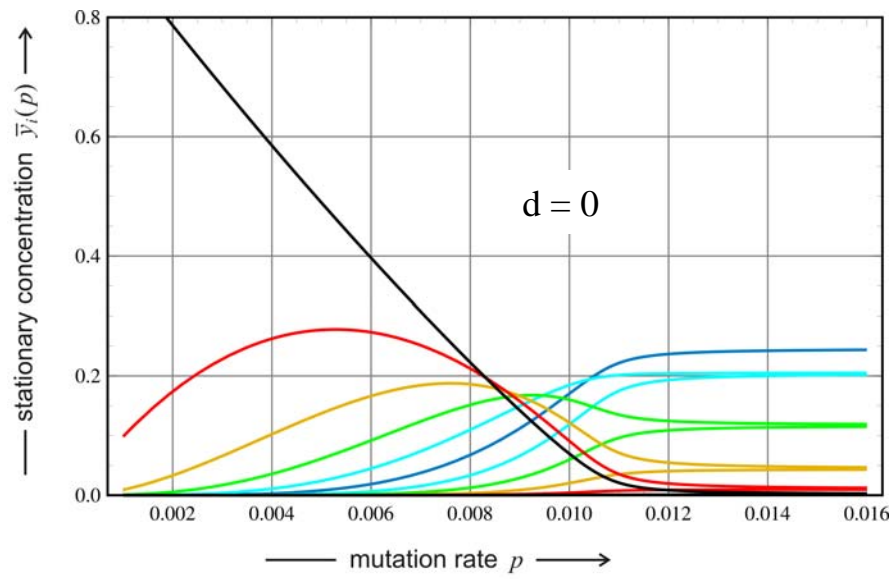


Error threshold: Individual sequences

$n = 10$, $\sigma = 2$, $s = 491$ and $d = 0, 0.5, 0.9375$

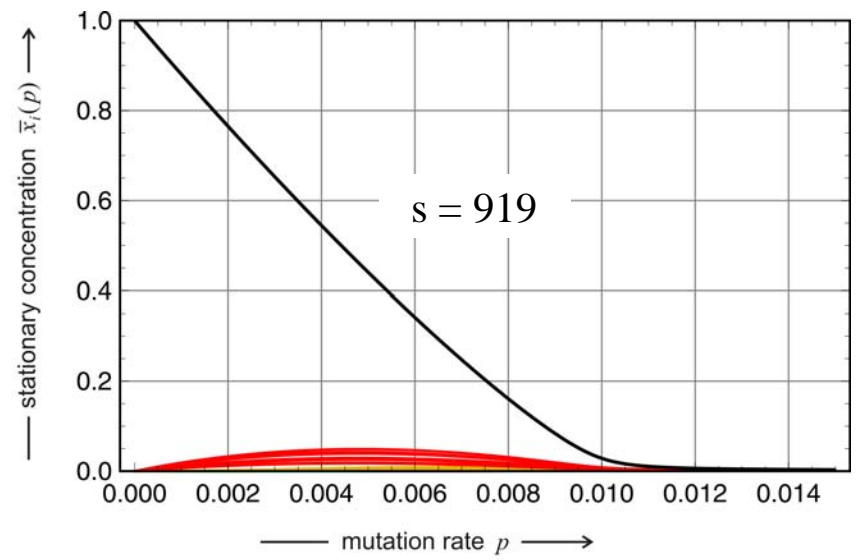
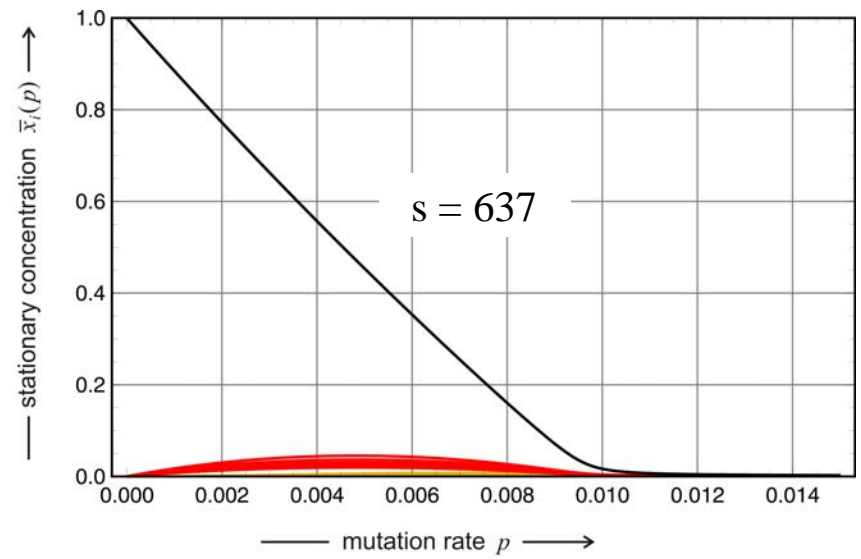
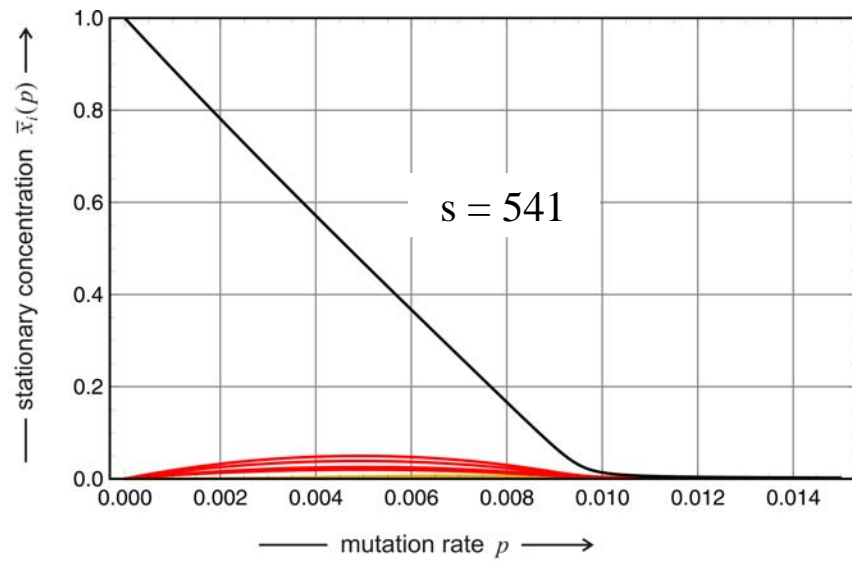
Do ,realistic' landscapes sustain error thresholds?

Three criteria: 1. **steep decrease** of master concentration,
2. **phase transition** like behavior, and
3. transition to the **uniform distribution**.



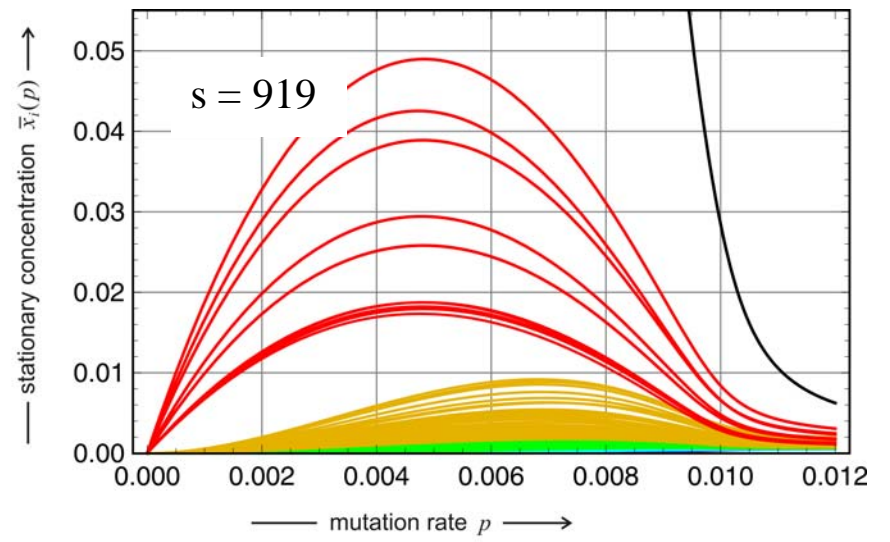
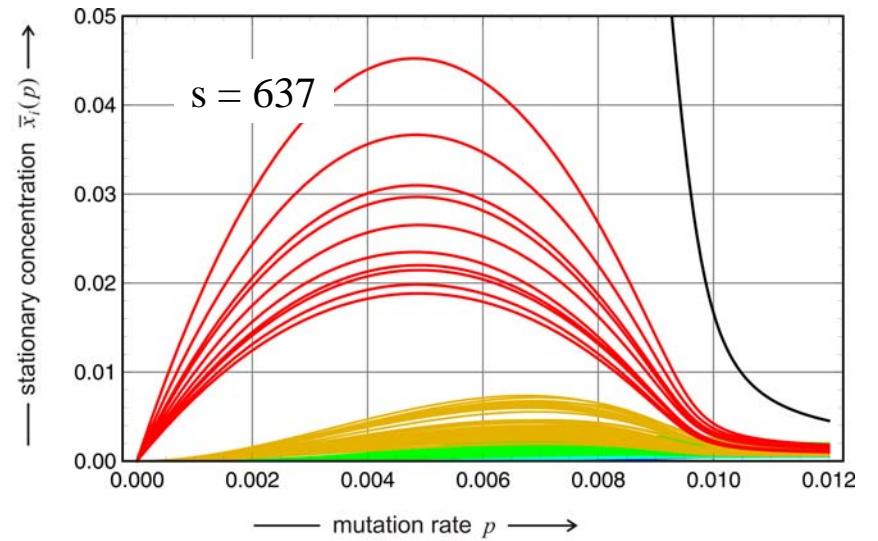
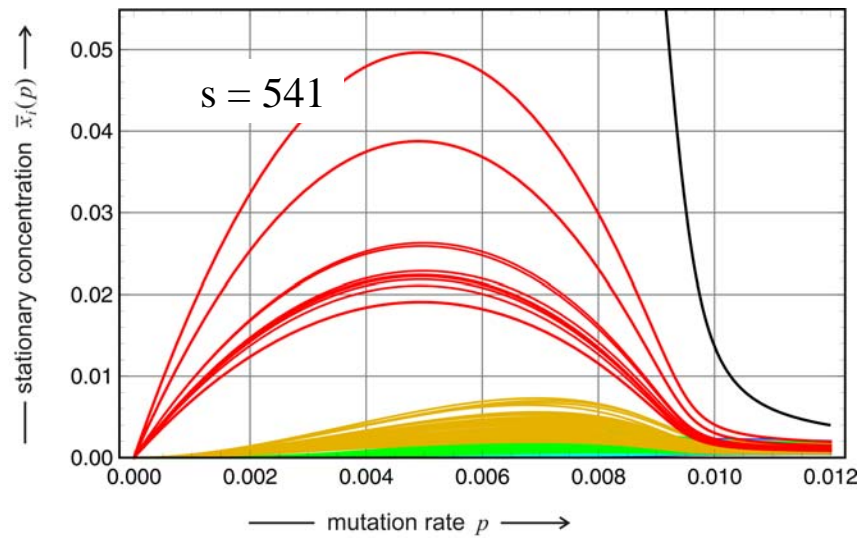
Error threshold on a 'realistic' landscape

$$n = 10, f_0 = 1.1, f_n = 1.0, s = 919$$



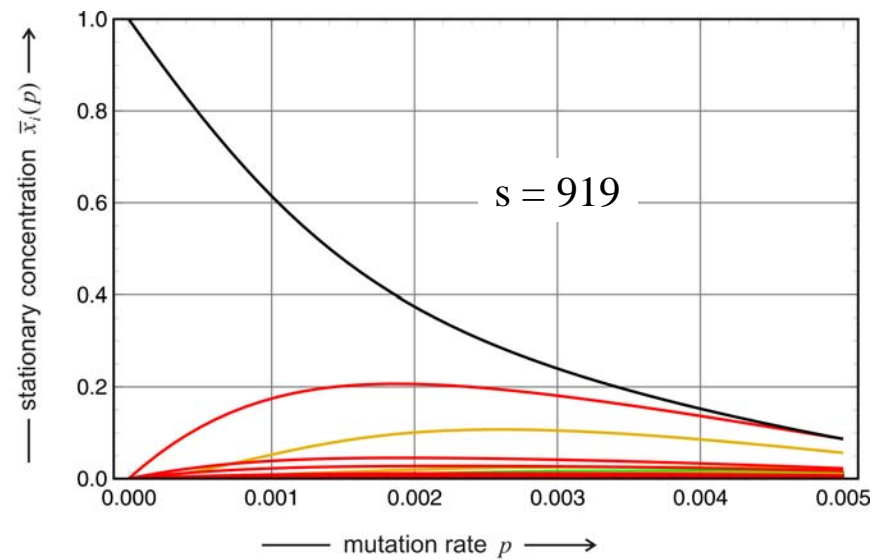
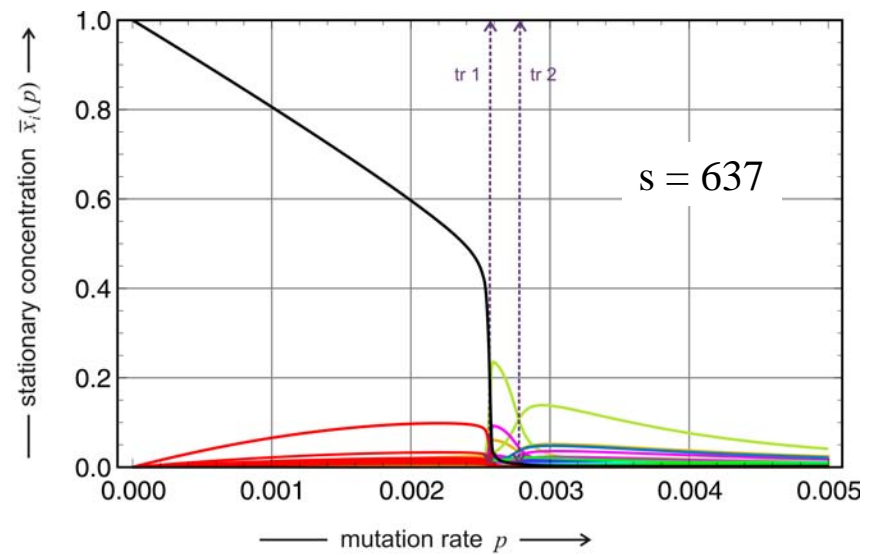
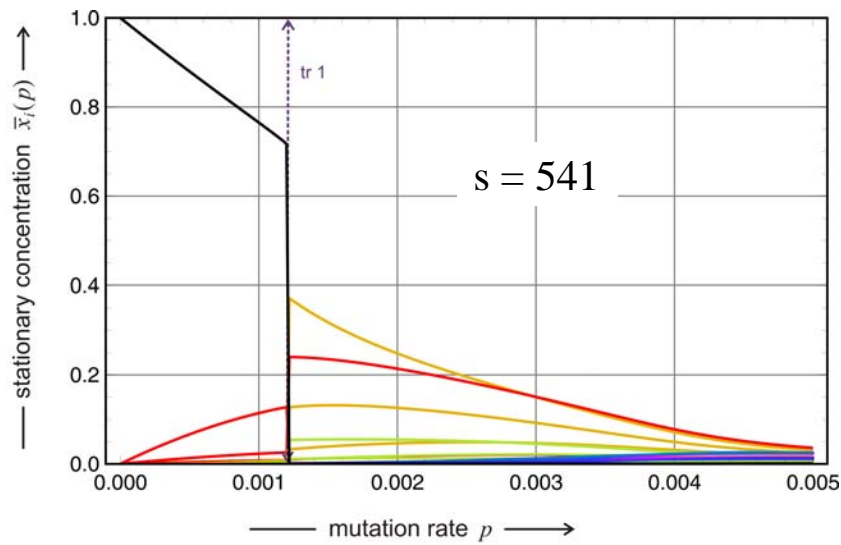
Error threshold on ,realistic‘ landscapes

$$n = 10, f_0 = 1.1, f_n = 1.0, d = 0.5$$



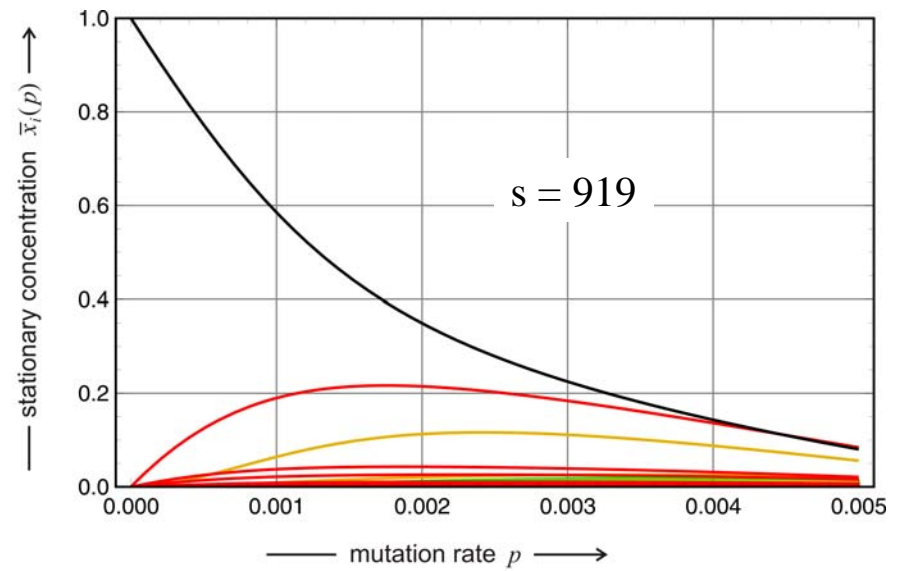
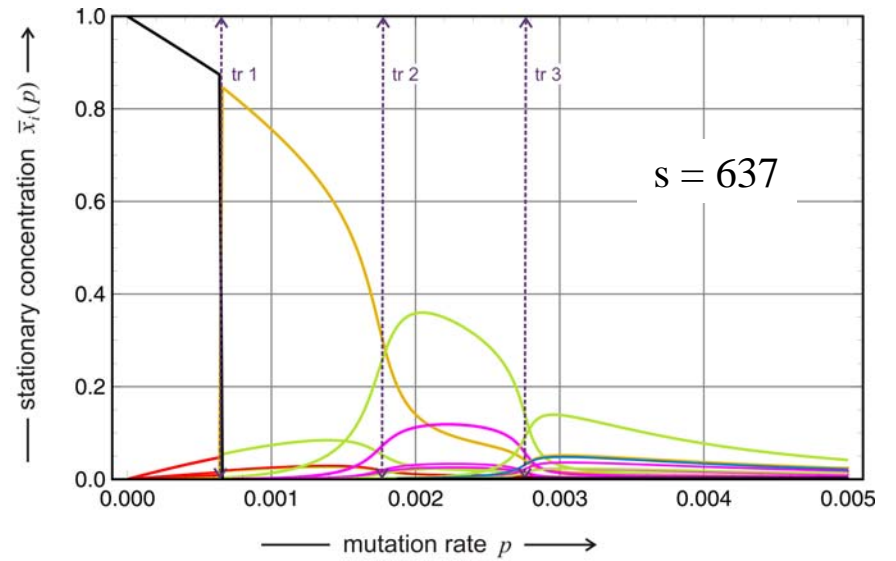
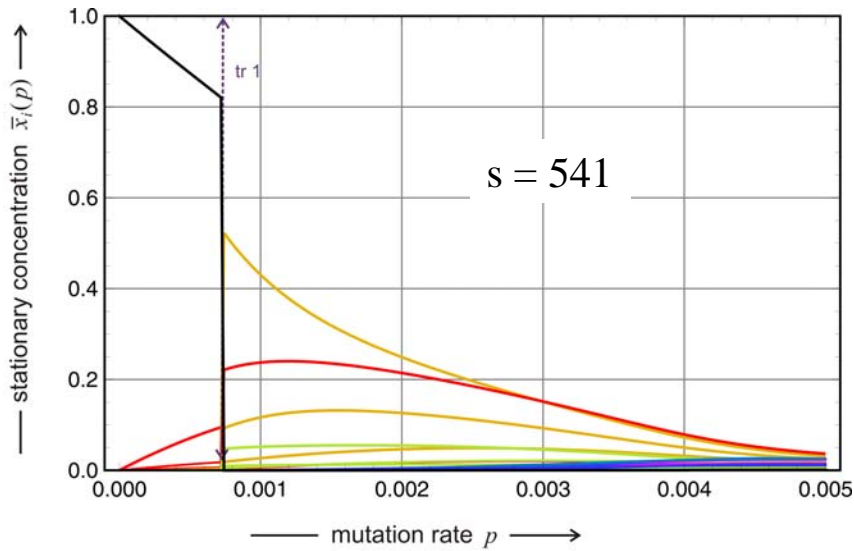
Error threshold on ,realistic‘ landscapes

$$n = 10, f_0 = 1.1, f_n = 1.0, d = 0.5$$



Error threshold on ,realistic‘ landscapes

$$n = 10, f_0 = 1.1, f_n = 1.0, d = 0.995$$



Error threshold on ,realistic‘ landscapes

$$n = 10, f_0 = 1.1, f_n = 1.0, d = 1.0$$

Two questions:

1. Can we predict mutational behavior of quasispecies from fitness landscapes?
2. What is the evolutionary consequence of the occurrence of mutationally stable and unstable quasispecies?

Landscape analysis 10-1.1-1.0-1-637

Class	Mean fitness	Max.fitness	Seq.	f0-fn	Neigh.mean	Neigh.max	Seq.	Class
0	1.10000	1.10000	0	0.00000	0.99786	1.08691	4	1

Landscape analysis through the evaluation of single point mutation neighborhoods

Landscape analysis 10-1.1-1.0-1-637

Class	Mean fitness	Max.fitness	Seq.	f0-fn	Neigh.mean	Neigh.max	Seq.	Class
0	1.10000	1.10000	0	0.00000	0.99786	1.08691	4	1
1	0.99786	1.08691	4	0.01309	0.99575	1.08499	68	2

Landscape analysis through the evaluation of single point mutation neighborhoods

Landscape analysis 10-1.1-1.0-1-637

Class	Mean fitness	Max.fitness	Seq.	f0-fn	Neigh.mean	Neigh.max	Seq.	Class
0	1.10000	1.10000	0	0.00000	0.99786	1.08691	4	1
1	0.99786	1.08691	4	0.01309	0.99575	1.08499	68	2
2	0.99417	1.09731	768	0.00269	0.99184	1.08998	769	3

Landscape analysis through the evaluation of single point mutation neighborhoods

Landscape analysis 10-1.1-1.0-1-637

Class	Mean fitness	Max.fitness	Seq.	f0-fn	Neigh.mean	Neigh.max	Seq.	Class
0	1.10000	1.10000	0	0.00000	0.99786	1.08691	4	1
1	0.99786	1.08691	4	0.01309	0.99575	1.08499	68	2
2	0.99417	1.09731	768	0.00269	0.99184	1.08998	769	3
3	1.00138	1.09966	19	0.00034	1.00521	1.08891	275	4
4	0.99981	1.09953	960	0.00047	0.99409	1.07539	968	5
5	1.00346	1.09794	391	0.00206	1.01215	1.07379	903	6
6	1.00218	1.09799	462	0.00201	1.01235	1.08449	334	5
7	0.99668	1.09971	923	0.00029	0.99776	1.09311	667	6
8	1.00614	1.09999	1003	0.00001	0.99310	1.08863	995	7
9	1.02155	1.09735	511	0.00265	0.97927	1.06224	447	8
10	1.04209	1.04209	1023	0.05791	1.02155	1.09735	511	9

Landscape analysis through the evaluation of single point mutation neighborhoods

Landscape analysis 10-1.1-1.0-1-919

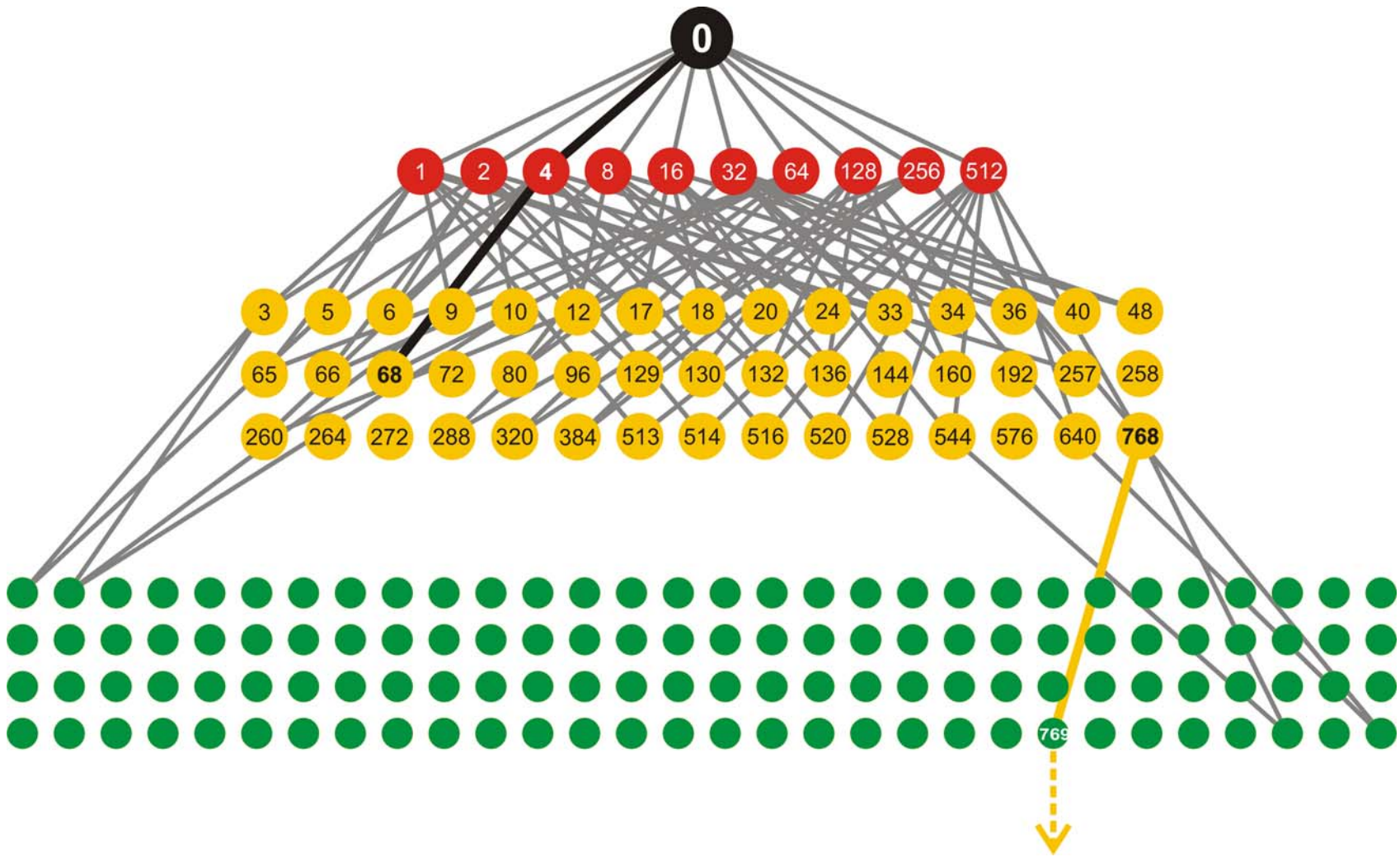
Class	Mean fitness	Max.fitness	Seq.	f0-fn	Neigh.mean	Neigh.max	Seq.	Class
0	1.10000	1.10000	0	0.00000	0.98992	1.09659	4	1
1	0.98992	1.09659	4	0.00341	1.03912	1.09703	516	2
2	1.00480	1.09703	516	0.00297	0.99848	1.09659	4	1

Landscape analysis through the evaluation of single point mutation neighborhoods

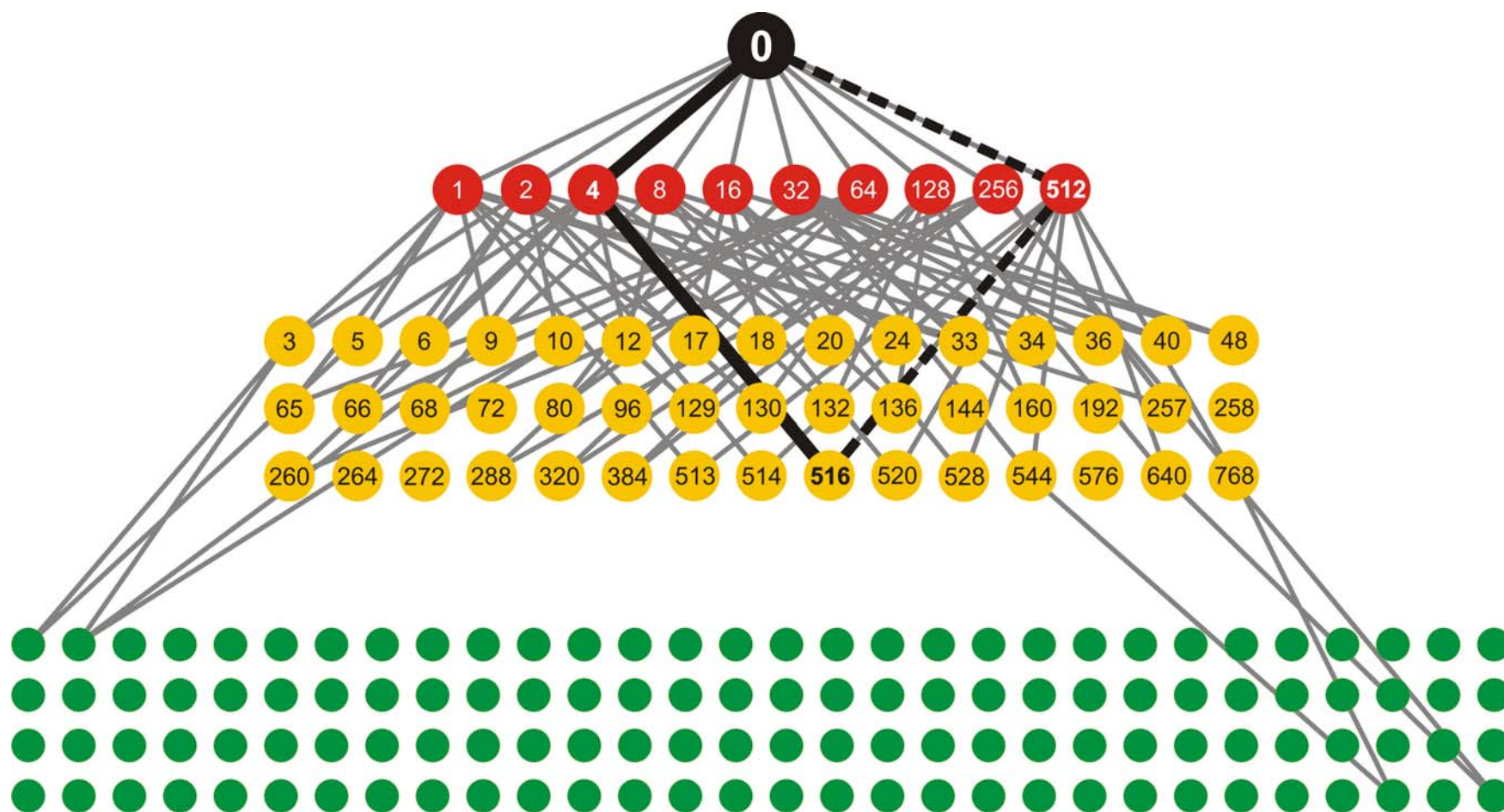
Landscape analysis 10-1.1-1.0-1-919

Class	Mean fitness	Max.fitness	Seq.	f0-fn	Neigh.mean	Neigh.max	Seq.	Class
0	1.10000	1.10000	0	0.00000	0.98992	1.09659	4	1
1	0.98992	1.09659	4	0.00341	1.03912	1.09703	516	2
2	1.00480	1.09703	516	0.00297	0.99848	1.09659	4	1
3	1.00575	1.09827	112	0.00173	0.99391	1.09340	624	4
4	0.99763	1.09850	801	0.00150	0.99919	1.09729	769	3
5	0.99339	1.09924	570	0.00076	0.98717	1.06809	634	6
6	0.99719	1.09829	573	0.00171	0.97527	1.09874	829	7
7	0.99683	1.09912	247	0.00088	1.00176	1.07528	503	8
8	1.00649	1.09670	703	0.00330	0.99227	1.06191	671	7
9	0.98467	1.07890	1015	0.02110	1.01749	1.09640	951	8
10	1.02104	1.02104	1023	0.07896	0.98467	1.07890	1015	9

Landscape analysis through the evaluation of single point mutation neighborhoods



Determination of the dominant mutation flow: $d = 1$, $s = 637$



Determination of the dominant mutation flow: $d = 1$, $s = 919$

1. Complexity in molecular evolution
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Motoo Kimuras population genetics of neutral evolution.

Evolutionary rate at the molecular level.
Nature **217**: 624-626, 1955.

The Neutral Theory of Molecular Evolution.
Cambridge University Press. Cambridge, UK, 1983.

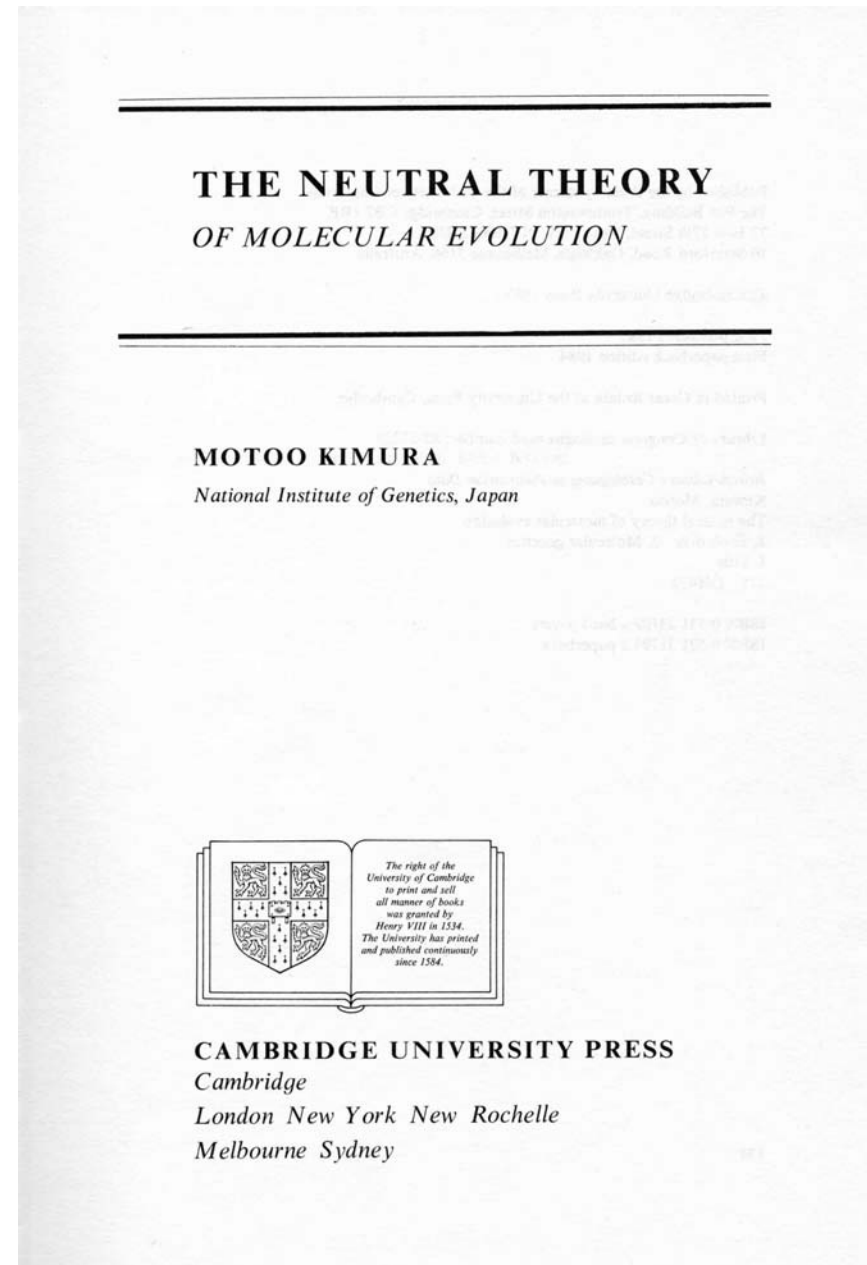
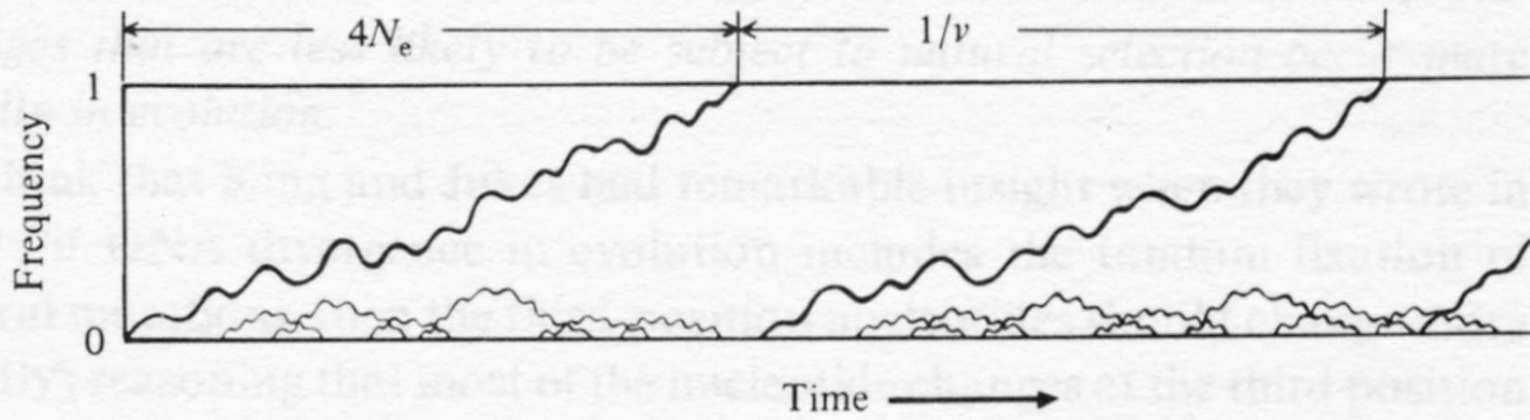


Fig. 3.1. Behavior of mutant genes following their appearance in a finite population. Courses of change in the frequencies of mutants destined to fixation are depicted by thick paths. N_e stands for the effective population size and v is the mutation rate.



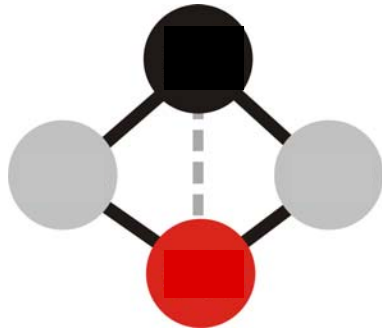
Motoo Kimura

Is the Kimura scenario correct for frequent mutations?



$$d_H = 1$$

$$\lim_{p \rightarrow 0} x_1(p) = x_2(p) = 0.5$$



$$d_H = 2$$

$$\lim_{p \rightarrow 0} x_1(p) = \alpha / (1 + \alpha)$$

$$\lim_{p \rightarrow 0} x_2(p) = 1 / (1 + \alpha)$$

$$d_H \geq 3$$

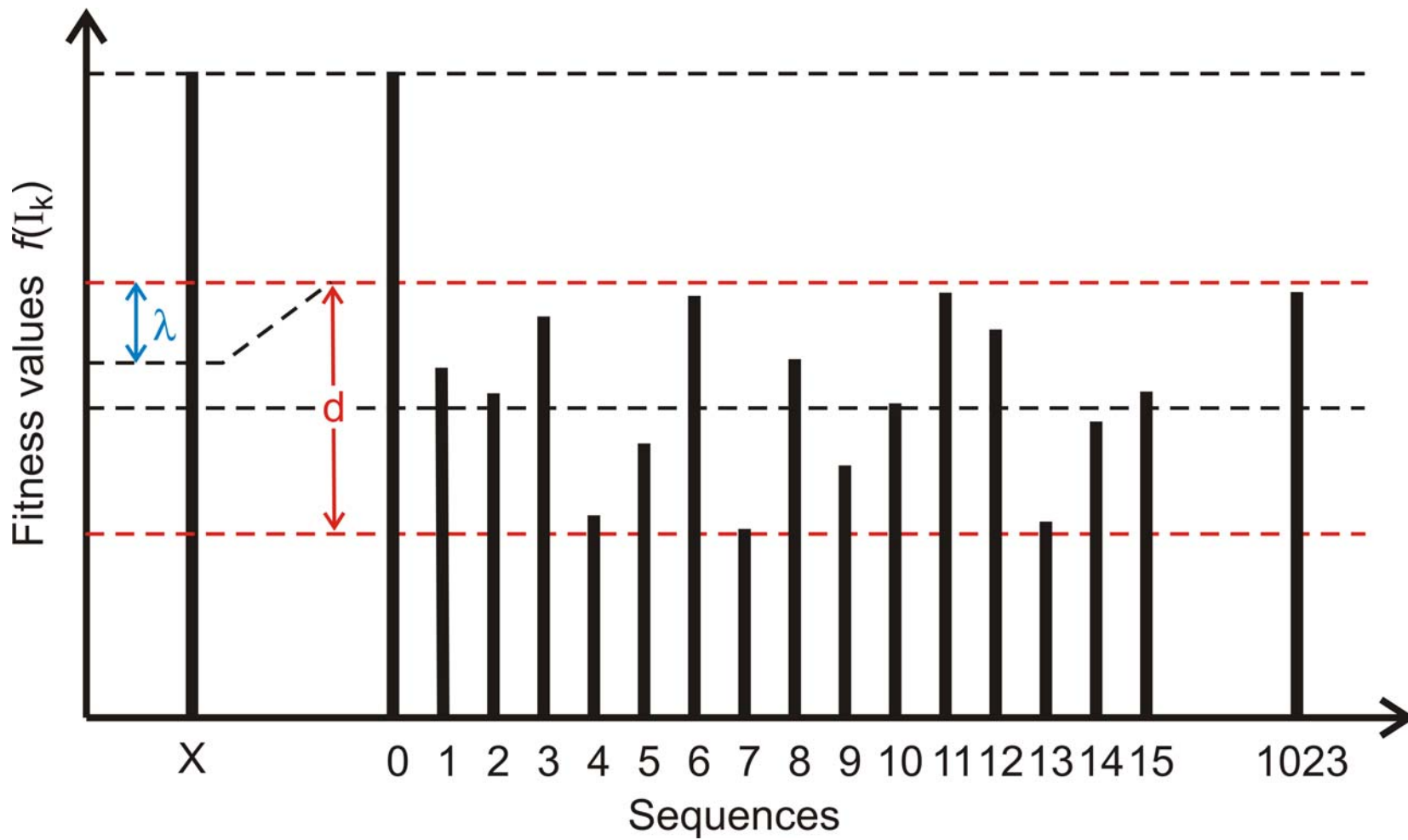
$$\lim_{p \rightarrow 0} x_1(p) = 1, \lim_{p \rightarrow 0} x_2(p) = 0 \text{ or}$$

$$\lim_{p \rightarrow 0} x_1(p) = 0, \lim_{p \rightarrow 0} x_2(p) = 1$$

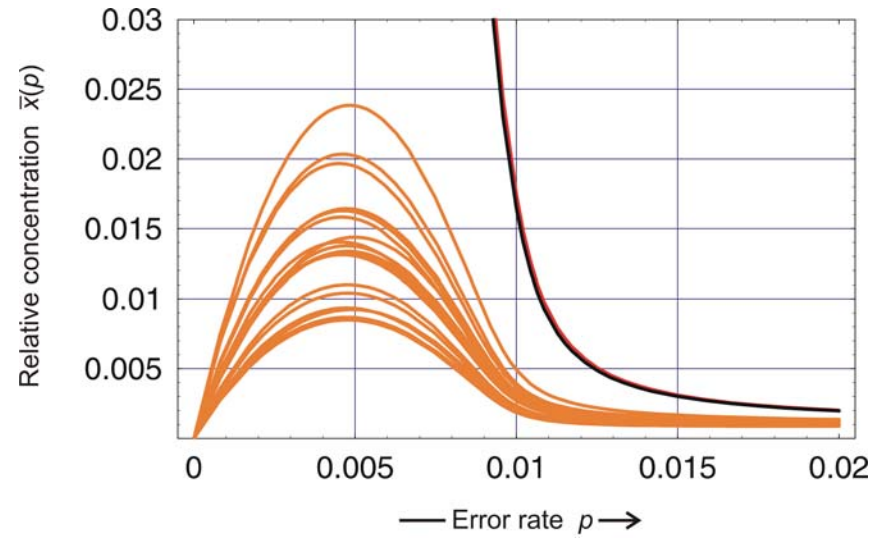
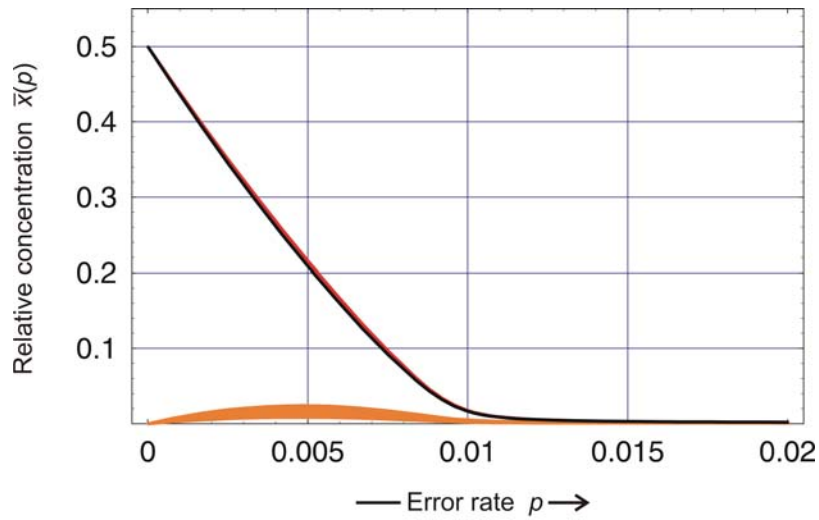
Pairs of neutral sequences in replication networks

Random fixation in the sense of Motoo Kimura

P. Schuster, J. Swetina. 1988. Bull. Math. Biol. 50:635-650



A fitness landscape including neutrality

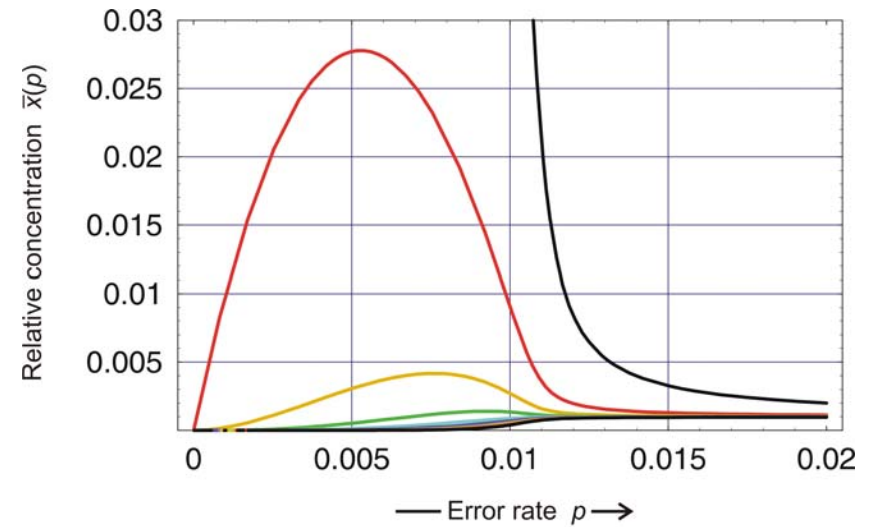


Neutral network

$\lambda = 0.01, s = 367$

Neutral network: Individual sequences

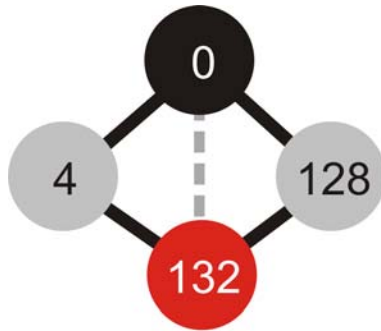
$n = 10, \sigma = 1.1, d = 1.0$



..... ACAUGCGAA
 AUAUACGAA
 ACAUGCGCA
 GCAUACGAA
 ACAUGC UAA
 ACAUGC GAG
 ACACGCGAA
 ACGUACGAA
 ACAUAGGAA
 ACAUACGAA

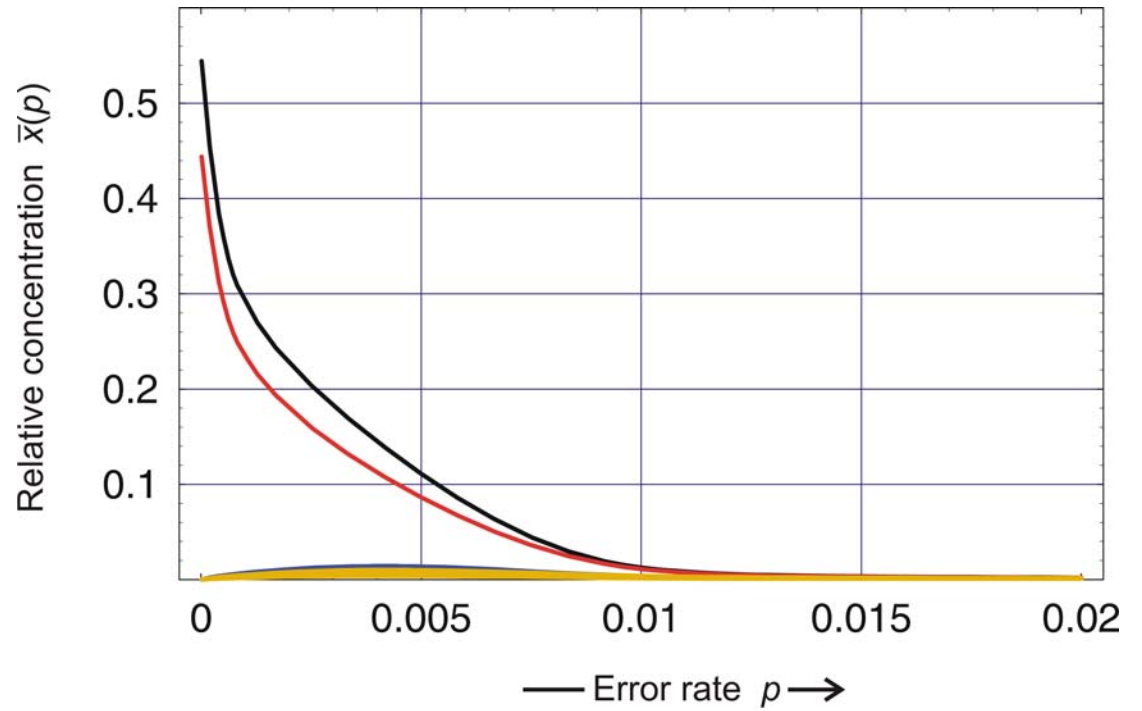
.....ACAU $\begin{matrix} G \\ A \end{matrix}$ CGAA.....

Consensus sequence of a quasispecies of two strongly coupled sequences of Hamming distance $d_H(X_i, X_j) = 1$.



Neutral network

$\lambda = 0.01, s = 877$



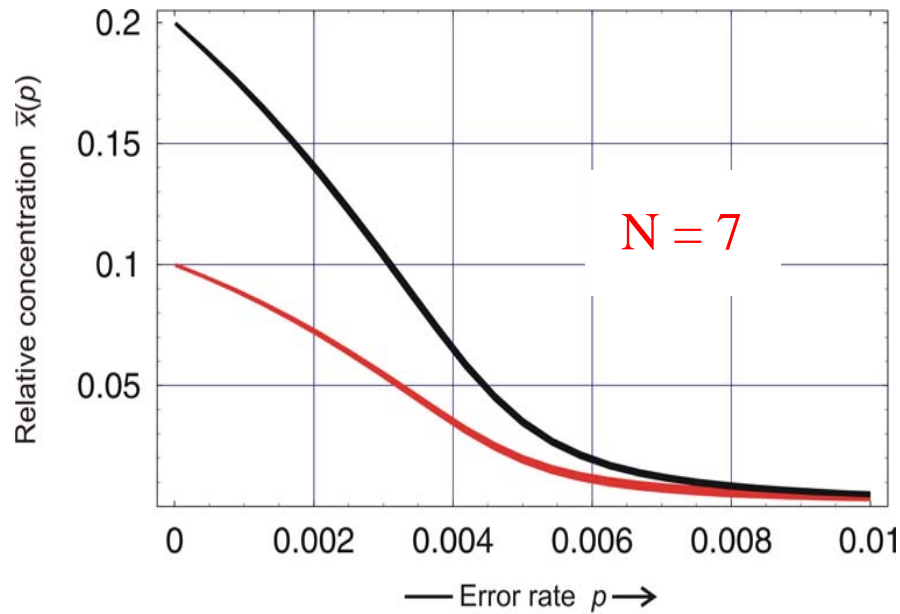
Neutral network: Individual sequences

$n = 10, \sigma = 1.1, d = 1.0$

..... ACAUGAUUCCCGAA
 AUAAUACCU CGAA
 ACAUAAUUC CCGCA
 GCAUAAUUUCU CGAA
 ACAUGAUUCCCUAA
 ACAUAAGUCCCGAG
 ACACGAUUC CCGAA
 ACGUAAUUCU CGAA
 ACAUGC UUCUAGAA
 ACAUAAUUC CCGAA
 AUAAUUCUCGGAA
 ACAAAUGCCCGUA

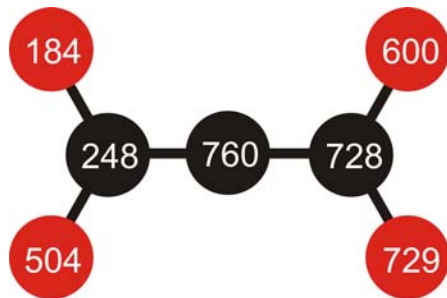
..... ACAU^A_GAUUC C^C_UCGAA

Consensus sequence of a quasispecies of two strongly coupled sequences of
 Hamming distance $d_H(X_i, X_j) = 2$.



Perturbation matrix W

$$W = \begin{pmatrix} f & 0 & \varepsilon & 0 & 0 & 0 & 0 \\ 0 & f & \varepsilon & 0 & 0 & 0 & 0 \\ \varepsilon & \varepsilon & f & \varepsilon & 0 & 0 & 0 \\ 0 & 0 & \varepsilon & f & \varepsilon & 0 & 0 \\ 0 & 0 & 0 & \varepsilon & f & \varepsilon & \varepsilon \\ 0 & 0 & 0 & 0 & \varepsilon & f & 0 \\ 0 & 0 & 0 & 0 & \varepsilon & 0 & f \end{pmatrix}$$



Neutral network

$$\lambda = 0.10, s = 229$$

Adjacency matrix

Largest eigenvector of W

$$\xi_0 = (0.1, 0.1, 0.2, 0.2, 0.2, 0.1, 0.1) .$$

Neutral networks with increasing λ : $\lambda = 0.10, s = 229$

Theory cannot remove complexity, but it shows what kind of „regular“ behavior can be expected and what experiments have to be done to get a grasp on the irregularities.

Manfred Eigen,

Preface to E. Domingo,
C.R. Parrish, J.J.Holland, eds.
Origin and Evolution of Viruses.
Academic Press 2008

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Thank you for your attention!

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