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PROBABILISTIC STRUCTURES IN EVOLUTION

DFG SPP 1590

COLLABORATIVE RESEARCH CENTER | SFB 680 Molecular Basis of Evolutionary Innovations

## Random paths in evolutionary biology

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- Biological contexts for the KPZ equation
- Paths on the hypercube
- Accessibility percolation
"Nonequilibrium dynamics of stochastic and quantum integrable systems" KITP Santa Barbara, February 18, 2016


## Eden growth



Eden 1961

Genetic segregation in growing bacterial colonies
Hallatschek et al., PNAS 2007


## Sector boundaries display superdiffusive KPZ fluctuations

Hallatschek et al., PNAS 2007




## Fitness of a population in a linear habitat

J. Otwinowski, S. Boettcher, PRE 84:011925 (2011)


## Universal fitness distributions

J. Otwinowski, JK, Phys. Biol. 11:056003 (2014)


- Three universal distributions for flat, droplet and stationary initial conditions


## SHE and PAM

- Under the Cole-Hopf transformation $\psi(x, t)=\exp \left[\frac{\lambda}{2 v} h(x, t)\right]$ the KPZ equation tranforms into the stochastic heat equation (SHE)

$$
\frac{\partial \psi}{\partial t}=v \nabla^{2} \psi+\frac{\lambda}{2 v} \eta(x, t) \psi
$$

- Via the Feynman-Kac formula this establishes the relation to directed polymers in random media (DPRM) and first passage percolation (FPP)
- When the noise is independent of time ("columnar DPRM") the problem is known as the parabolic Anderson model (PAM) with a natural biological intepretation:

$$
x \rightarrow \text { phenotype, } \quad \psi(x, t) \rightarrow \text { population density, } \quad \eta(x) \rightarrow \text { fitness }
$$

- However in that context the dynamics should properly be defined on the space of genetic sequences rather than on $\mathbb{R}^{d}$ or $\mathbb{Z}^{d}$


## Sequence spaces

- Genetic information is encoded in DNA-sequences consisting of four different nucleotide bases


## ..ACTATCCATCTACTACTCCCAGGAATCTCGATCCTACCTAC...

- The sequence space consists of all $4^{L}$ sequences of length $L$
- Typical genome lengths:
$L \sim 10^{3}$ (viruses), $L \sim 10^{6}$ (bacteria), $L \sim 10^{9}$ (higher organisms)
- Proteins are sequences of 20 amino acids with $L \sim 10^{2}$
- Coarse-grained representation of classical genetics: $L$ genes that are present as different alleles; often it is sufficient to distinguish between wild type ( 0 ) and mutant ( 1 ) $\Rightarrow$ binary sequences
- Hamming distance: Two sequences are nearest neighbors if they differ in a single letter (mutation)


## Hamming spaces/hypercubes for $L=1-6$



## Fitness landscapes

- A fitness landscape assigns a fitness value $f(\sigma)$ to each genotype sequence $\sigma=\left(\sigma_{1} \sigma_{2} . . \sigma_{L}\right)$ with $\sigma_{i} \in\{0,1\}$
- Evolution is a hill-climbing process in the fitness landscape
- Example: $L=2$



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- Example: $L=3$

- 3 ! = 6 directed $(0 \rightarrow 1)$ and 18 undirected mutational pathways from $\sigma^{(0)}=(000)$ to $\sigma^{(1)}=(111)$


## Random fitness landscapes

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


Simple choice: Assign fitnesses at random to genotypes

- Fitnesses as i.i.d. random variables $\Rightarrow$ Kingman's house-of-cards model

Kingman 1978, Kauffman \& Levin 1987

- Equivalent to Derrida's Random Energy Model of spin glasses Derrida 1981
- Correlated landscapes can be generated along similar lines (e.g., the spin-glass-like NK-models)


## PAM on the random hypercube: Adaptive flights

K. Jain, JK, JSTAT 2005; K. Jain, PRE 2007

- Under PAM dynamics the population concentrates on sites with exceptionally high fitness
- An adaptive trajectory consists of a sequence of long-ranged "tunneling" events between such sites that terminates at the global maximum
- The number of jumps is $\mathscr{O}(\sqrt{L})$ for Gumbel-class fitness distributions and $\mathscr{O}(1)$ for power-law distributions
- The distribution of the time $T_{k}$ of the $k^{\prime}$ th last jump has a universal power law tail

$$
\operatorname{Prob}\left[T_{k}>t\right] \sim t^{-k}, \quad k=1,2,3, \ldots
$$

which implies that the expected time to the maximum is infinite

- This scenario is however biologically meaningless because it relies on exponentially small population densities


## Evolutionary accessibility

J. Franke et al., PLoS Comp. Biol. 7 (2011) e1002134

- In moderately large populations, adaptive trajectories are constrained to move uphill in single mutational steps $\Rightarrow$ a pathway connecting two genotypes is accessible if fitness increases monotonically in each step
- Example: Mutational pathways from (1111) to (0000) in two 4-locus subgraphs of an 8 -dimensional empirical fitness landscape for the filamentous fungus Aspergillus niger

no directed pathway accessible


6 out of 24 pathways accessible

## Pathways to antibiotic resistance

D.M. Weinreich et al., Science 312, 111 (2006)


- 5 mutations increase resistance to a new drug by $\sim 10^{5}$
- 18 out of 5 ! = 120 directed mutational pathways are accessible, and only few of them have appreciable weight


## Pathways to antibiotic resistance

De Pristo et al., Mol. Biol. Evol. 24:1608 (2007)


- 27 out of 18651552840 undirected pathways are accessible


## Pathways to drug resistance in malaria

E.R. Lozovsky et al., Proc. Natl. Acad. Sci. USA 106, 12025 (2009)


- 4 ! $=24$ pathways, 10 (red) are monotonic in resistance
- Dominating pathways are realized in natural populations


## Accessibility percolation

- Directed or undirected graph $G$ with nodes $x \in G$ and distance $d(\cdot, \cdot)$
- Assign a nondegenerate real random variable $f(x)$ to each node
- A path is a string of nodes $x_{0} \rightarrow x_{1} \rightarrow x_{2} \rightarrow \ldots \rightarrow x_{N}$ such that $d\left(x_{i}, x_{i+1}\right)=1$ for all $i$
- A path is called accessible if $f$ increases monotonically along the path, i.e. $f\left(x_{0}\right)<f\left(x_{1}\right)<\ldots<f\left(x_{N}\right)$
- Accessibility percolation is concerned with the existence of global paths that connect the global maximum $x_{\max }$ of $f(x)$ to the node at maximal distance $D \equiv \max _{x \in G} d\left(x_{\max }, x\right)$
- In the standard setting $G$ is the hypercube and the $f(x)$ are i.i.d. random variables


## Directed random hypercube

- Assign maximal fitness $f=1$ to $\sigma^{(1)} \equiv(1,1, \ldots, 1)$ and i.i.d. $U(0,1)$ RV's to all other sites
- What is the expected number of directed accessible pathways from a site at distance $d$ to $\sigma^{(1)}$ ?
- The total number of paths is $d$ !, and a given path consists of $d$ i.i.d. fitness values $f_{0}, \ldots ., f_{d-1}$; it is accessible iff $f_{0}<f_{1} \ldots .<f_{d-1}$
- Since all $d$ ! permutations of the $d$ random variables are equally likely, the probability for this event is $1 / d$ !

$$
\Rightarrow \mathbb{E}\left(n_{\mathrm{acc}}\right)=\frac{1}{d!} \times d!=1
$$

- This applies in particular for $d=L$


## Distribution of the number of accessible paths

J. Franke et al., PLoS Comp. Biol. 7 (2011) e1002134


- "Condensation of probability" at $n_{\text {acc }}=0$
$\Rightarrow$ mean is not representative of the typical behavior
- Constraining initial fitness to $f_{0}=0$ massively increases the accessibility


## Transition as a function of initial fitness

- Conditioned on initial fitness $f_{0} \in[0,1)$ the expected number of accessible paths is

$$
\mathbb{E}\left(n_{\text {acc }}\right)=\frac{\left(1-f_{0}\right)^{L-1}}{(L-1)!} \times L!=L\left(1-f_{0}\right)^{L-1}
$$

which diverges/vanishes asymptotically for large $L$ when $f_{0}<\frac{\ln L}{L} / f_{0}>\frac{\ln L}{L}$

- This implies that the existence of accessible paths becomes likely at $f_{0} \sim \frac{\ln L}{L}$, in the sense that Hegarty \& Martinsson, Ann. Appl. Prob. 2014

$$
\lim _{L \rightarrow \infty} \operatorname{Prob}\left[n_{\text {acc }}>0\right]=\left\{\begin{array}{l}
0 \text { for } f_{0}>\frac{\ln L}{L} \\
1 \text { for } f_{0}<\frac{\ln L}{L} .
\end{array}\right.
$$

- Proof uses estimate of second moment of $n_{\text {acc }}$ and the bounds

$$
\mathbb{E}\left(n_{\mathrm{acc}}\right) \geq \operatorname{Prob}\left[n_{\mathrm{acc}}>0\right] \geq \frac{\mathbb{E}\left(n_{\mathrm{acc}}\right)^{2}}{\mathbb{E}\left(n_{\mathrm{acc}}^{2}\right)}
$$

## Accessibility percolation on trees

S. Nowak, JK, EPL 101, 66004 (2013)

- Consider a regular tree with branching number $b$ and height $h$ equipped with i.i.d. RV's on the nodes
- Let $n_{\text {acc }}$ denote the number of accessible paths from the root to the leaves
- First and second moments are given by

$$
\mathbb{E}\left(n_{\mathrm{acc}}\right)=\frac{b^{h}}{h!}, \quad \mathbb{E}\left(n_{\mathrm{acc}}^{2}\right)=\mathbb{E}\left(n_{\mathrm{acc}}\right)+\frac{b-1}{b} \sum_{k=1}^{h}\binom{2 k}{k} \frac{b^{h+k}}{(h+k)!}
$$

- Scaling $b, h \rightarrow \infty$ at fixed $\alpha=b / h$ it follows that accessibility percolation occurs at some $\alpha_{c} \in[1 / e, 1]$,
- Refined analysis shows that $\alpha_{c}=1 / e$ which corresponds exactly to the hypercube geometry


## Effect of downhill steps

- Two scenarios for allowing downhill steps along the path:
- unconditional: ...f $f_{i-2}<f_{i-1}>f_{i}<f_{i+1}<\ldots$ for some $i$
- conditional: $f_{i-1}>f_{i}<f_{i+1}$ but $f_{i+1}>f_{i-1}$
- Expected number of accessible paths in the two cases are

$$
\mathbb{E}^{\mathrm{uc}}\left(n_{\mathrm{acc}}\right)=2^{L}-L, \quad \mathbb{E}^{\mathrm{c}}\left(n_{\mathrm{acc}}\right)=1+\frac{1}{2} L(L-1)
$$

- In the unconditional case accessible paths exist almost surely for any initial fitness when $L \rightarrow \infty$, whereas in the conditional case the accessibility threshold is

$$
f_{0} \sim \frac{(2 p+1) \ln L}{L}
$$

when $p$ downhill steps are allowed for

## Accessibility percolation on the undirected hypercube

J. Berestycki, É. Brunet, Z. Shi, arXiv:1401.6894

- A general undirected path from $\sigma^{(0)}$ to $\sigma^{(1)}$ consists of $L+2 p$ steps where $p \geq 0$ is the number of backsteps (mutational reversions)
- The expected number of accessible paths conditioned on starting fitness $f_{0}$ is

$$
\mathbb{E}\left(n_{\mathrm{acc}}\right)=\sum_{p \geq 0} a_{L, p} \frac{\left(1-f_{0}\right)^{L+2 p-1}}{(L+2 p-1)!}
$$

where $a_{L, p}$ is the number of paths with $p$ backsteps.

- Analyzing the asymptotics of the $a_{L, p}$ it is shown that

$$
\lim _{L \rightarrow \infty}\left[\mathbb{E}\left(n_{\mathrm{acc}}\right)\right]^{1 / L}=\sinh \left(1-f_{0}\right)
$$

which suggests a finite accessibility threshold $f_{0}^{*}=1-\sinh ^{-1}(1) \approx$ 0.11863...

## A link to first passage percolation

A. Martinsson, arXiv:1501.02206

- Graph $G$ with i.i.d. $\mathrm{U}(0,1)$ random waiting times $\tau(x)$ assigned to nodes $x$
- The first passage time from a distinguished node $x^{(0)}$ to $x$ is

$$
T(x)=\min _{\pi}\left[\sum_{y \in \pi \backslash\left\{x^{(0)}, x\right\}} \tau(y)\right]
$$

where $\pi$ is a path from $x^{(0)}$ to $x$

- Then the fitnesses $f(\sigma)$ defined as the fractional part of $f_{0}+T(x)$ are i.i.d. $U(0,1)$ RV's, and as a consequence

$$
\operatorname{Prob}\left[n_{\mathrm{acc}}\left(x^{(0)} \rightarrow x\right)>0\right]=\operatorname{Prob}\left[T(x)<1-f_{0}\right]
$$

- It follows that the first passage time on the oriented (unoriented) hypercube converges to $1\left(1-f_{0}^{*}=\sinh ^{-1}(1) \approx 0.88137 \ldots\right)$ for large $L$.


## The role of backsteps in empirical data

M. Josupeit, JK, in preparation


- Comparison of subgraph analysis of an empirical data set with the rough Mt. Fuji model defined by $f(\sigma)=c d\left(\sigma, \sigma^{(0)}\right)+\eta_{\sigma}$ with $U(0,1)$ RV's $\eta_{\sigma}$
- Accessibility is dominated by direct paths for small $L$


## Summary

- A new type of random path problem motivated by evolutionary biology
- "Critical" role of hypercube geometry
- Provides a tool to interpret empirical fitness landscapes J.A.G.M. de Visser, JK, Nat. Rev. Gen. 15:480 (2014)
- Focus so far on the existence of paths rather than on the distribution of path weights


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