Darwin's daemon and a mechanistic approach to evolution

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with R. Hanel and P. Klimek

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What is evolution?

• Universal mechanism how diversity of a system changes endogenously

biological, ecological, chemical, social, industrial, economical, financial, historical



What is an evolutionary process ?

Evolution is a three step process

- Step 1: new thing comes into being into a given environment
- Step 2: new thing has the chance to interact with environment. Result of this interaction: get selected or destroyed
- Step 3: if new thing gets selected (survives) in this environment it becomes part of this environment it becomes part of the definition of the environment for all future new and arriving elements

Evolution is not physics



Evolution is not physics

How does physics work?

Since Newton physics follows a recipe

(1) write down differential equations: $\frac{d^2x}{dt^2} = mg$ (2) fix boundary conditions: $x(0) = x_0$, $v(0) = v_0$ (3) solve equations: $x(t) = \frac{1}{2}mgt^2 + v_0t + x_0$ (4) test the solution in nature: drop stone, measure x(t), compare

Two problems

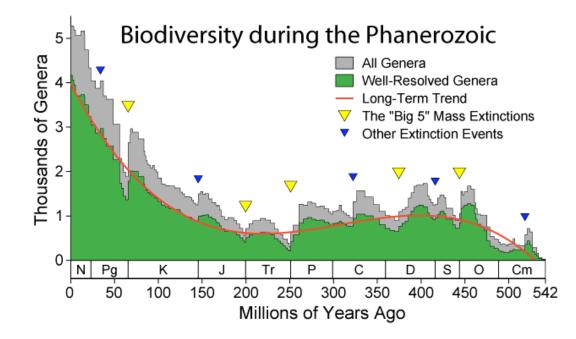
(1) can not fix boundary conditions

(2) can not prestate potential outcomes

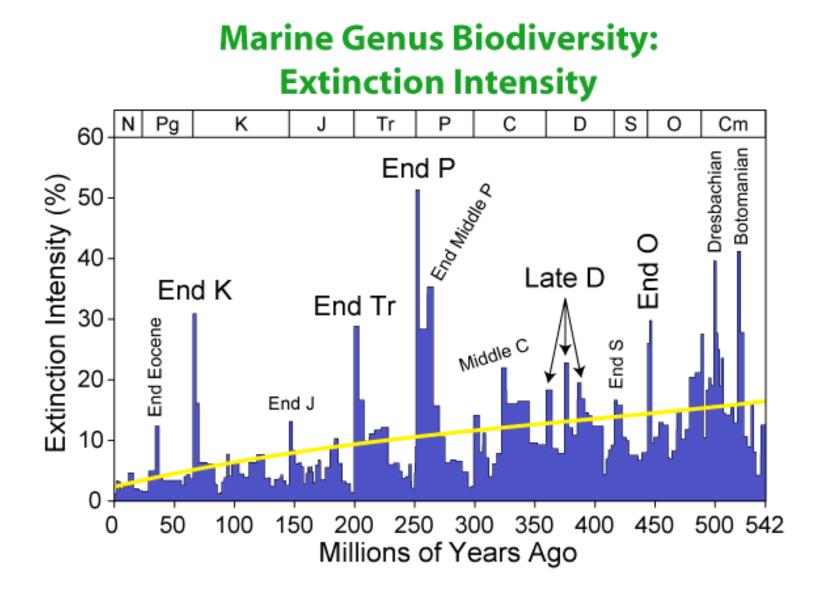


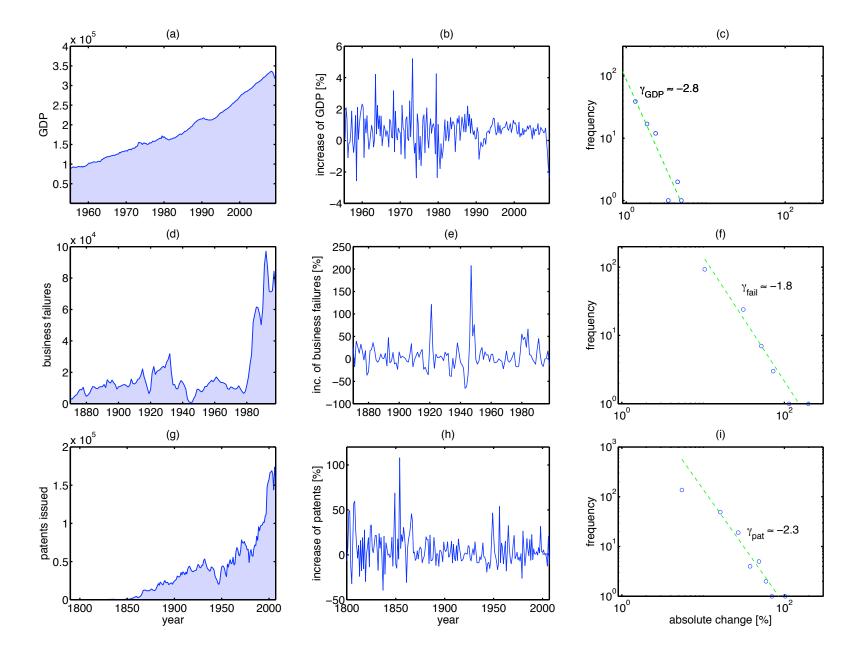
Facts on evolution dynamics?

- Evolution involves fast and dramatic changes of diversity
- Bursts of massive creation and extinction happen with little / no precursor
- Transitional species are practically missing











Traditional view on evolutionary dynamics

Species i produces species k under the influence of species j – can be expressed by a network equation of the type

$$\dot{x}_k = \sum_{i,j} \alpha_{kij} x_i x_j - x_k \Phi \quad , \quad \Phi = \sum_{i,j,k} \alpha_{kij} x_i x_j$$

• entities: x_i • rules: α_{kij} (density of entries r)

replication	x_{i}	\rightarrow	$2x_i$	f_i
competition	$x_i + x_j$	\rightarrow	x_{j}	p_{ij}
mutation	x_{i}	\rightarrow	x_{j}	q_{ij}
recombination	$x_j + x_l$	\rightarrow	x_i	$lpha_{ijl}$
development	$x_j + x_l$	\rightarrow	p_i	d_{ijl}



Problems

What can you measure?

- *x*: YES
- α: NO

Even if one could \rightarrow fundamental problem of predictive value of traditional of evolutionary dynamics



Darwin's daemon – limitations of the fitness concept

Fitness-based population dynamics can not make quantitative, falsifiable predictions on long-term behavior of evolutionary systems

Thought experiment: *Darwin's Demon*

Suppose a demon has exact knowledge on the abundance and fitness of each species in the universe. (fitness \equiv functional dependence of its proliferation rate on all other species). Daemon can measure fitness landscape exactly

What can the demon *predict*? Surprisingly little

New species is created \rightarrow part of environment \rightarrow environment changes

Demon has information on environment existing *before* arrival of new species

Thus the demon might have an exact description of the *current* biosphere, but with the advent of each new species this description loses accuracy.

Fitness always encodes $a \ posteriori$ info \rightarrow useless for falsifiable predictions



Needed: co-evolution of fitness and species

Not fruitful to predict future fitness from present fitness

Instead one should understand how species and their fitness landscapes co-construct each other – *co-evolve*

We propose a variational principle in a spin-model-like setup

We derive a functional which is minimized under the most general evolutionary formulation of a dynamical system, i.e. evolutionary trajectories causally emerge as a minimization of a functional



Three components of a model

System is characterized by

- \bullet States: open system \rightarrow infinite state vector. For simplicity binary
- Interactions: mutual influences of things \rightarrow production rate tensor. binary
- Stochastic component (β)



Ingredients: Species

Species exist or not

 $\sigma_i(t) = 1$ species *i* exists at *t*

 $\sigma_k(t) = 0$ species k does not exist at t: because not produced yet, or got eliminated from the system

N-dimensional vector $\vec{\sigma}(t)$ contains all thinkable species

Species has to be produced by its components at every timestep

Production of species i can happen only if all necessary components (parts) are simultaneously available

Diversity of system: $D(t) = \frac{1}{N} \sum_{i=1}^{N} \sigma_i(t)$



How do new species (goods, ideas, ...) appear?

New things emerge through combination and substitution of existing things

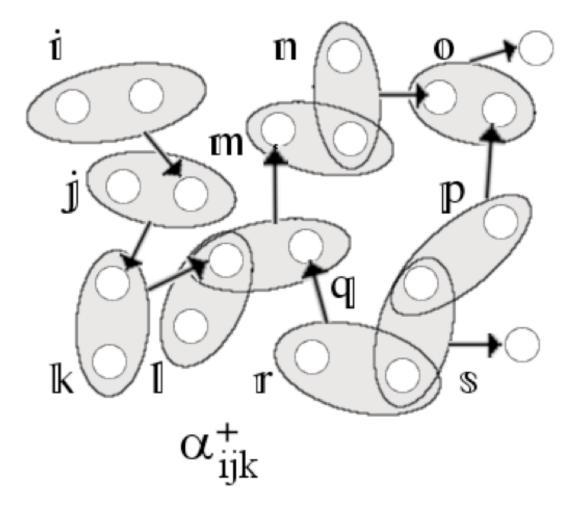
Examples

- \exists rule: a blacksmith can make a knife from a block of steel + a hammer
- \exists rule: hydrogen with oxygen can form water
- \exists rule: a cock and a hen can be the parents of a chicken
- $\neg \exists$ rule: a fish and a dog can be the parents of a chicken
- $\neg \exists$ rule: two blocks of U 235 can be welded into one big block of uranium

Whether a thing i can be produced from components j and k is encoded in the production map, $\alpha^+_{ijk}=0$ or 1



Ingredients: Production map





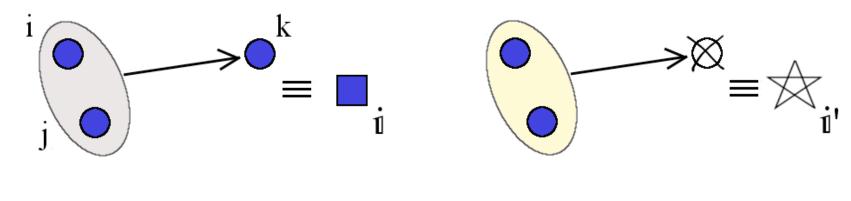
Ingredients: Production

If possible to produce thing *i* from *j* and *k*, call it a production ($\alpha_{ijk}^+ = 1$)

If production actually produces i, call it active production ($\sigma_i(t) = \sigma_j(t) = \sigma_k(t) = \alpha^+_{ijk} = 1$)

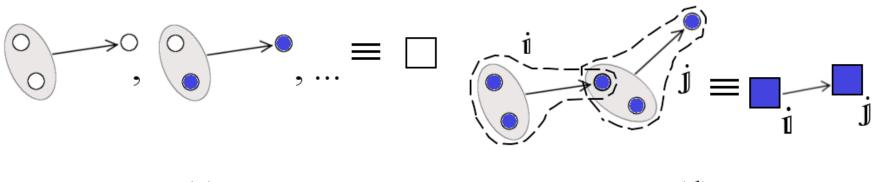
Production process: $\sigma_i(t+1) = \alpha_{kij}^+ \sigma_j(t) \sigma_k(t)$











(c)





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Ingredients: Selection/Competition/Destruction

New species, good,... can cause competition with existing one: think of good which can be produced cheaper or that is more robust etc., will drive the other from the market, biosphere,...

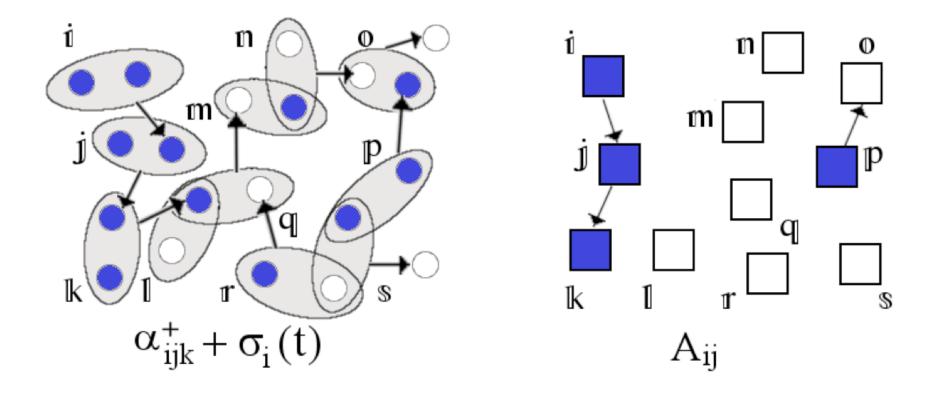
Mechanism: combination of things i and j produces a thing l which then drives product k out.

active destruction: if $\sigma_i(t) = \sigma_j(t) = \sigma_k(t) = \alpha_{ijk}^- = 1$

Destruction process (if $\sigma_i(t) = 1$): $\sigma_i(t+1) = 1 - \alpha_{jki}^- \sigma_j(t) \sigma_k(t)$



The active production network



Distinguish production rules (α^+) and the active production networks A(t)



Spontaneous inventions/mutations

- From time to time spontaneous ideas or inventions happen
- \bullet Introduce probability p with which a non-existing good gets spontaneously invented or an existing good is spontaneously annihilated



Summary of the model

- Phase space: $\Gamma \equiv \{0,1\}^N = \{ \boldsymbol{\sigma} \mid \sigma_i \in \{0,1\}, 1 \le i \le N \}$, $N \gg 1$
- Define quadratic forms: $\Delta_i(\boldsymbol{\sigma}) \equiv \sum_{j,k}^N \alpha_{ijk} \sigma_j \sigma_k$ with $\alpha_{ijk} \in \{-1, 0, 1\}$
- Select a positive value $p \leq 1$
- Dynamical update is the map $F: \Gamma \to \Gamma$ via the difference equation

$$\boldsymbol{\sigma}(t+1) = F(\boldsymbol{\sigma}(t)) \qquad , \qquad F = \boldsymbol{\Psi} \circ \boldsymbol{\Phi}$$

• Deterministic part $\Phi(\sigma) \equiv \mathbf{x}$ where $x_i = 1(0)$ when $\Delta_i(\sigma) > 0 (< 0)$, and $x_i = \sigma_i$, when $\Delta_i(\sigma) = 0$.

• Stochastic part $\Psi \equiv \mathbf{x}$, where $x_i = 1 - \sigma_i$ with probability p, and $x_i = \sigma_i$ else



An evolutionary algorithm

Update from time t to t+1

- pick a thing *i* at random (random sequential update)
- count all active productive and destructive influences on i

For example: If $N_i^{\text{prod}} > N_i^{\text{destr}} \longrightarrow \text{produce } i$

- with probability p switch the state of $\sigma_i(t+1)$
- continue with next thing

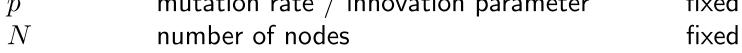
Initial condition (t = 0): fraction of randomly chosen initial things which exist

For simplicity: α_{ijk} binary random matrices, characterized by r^{\pm}



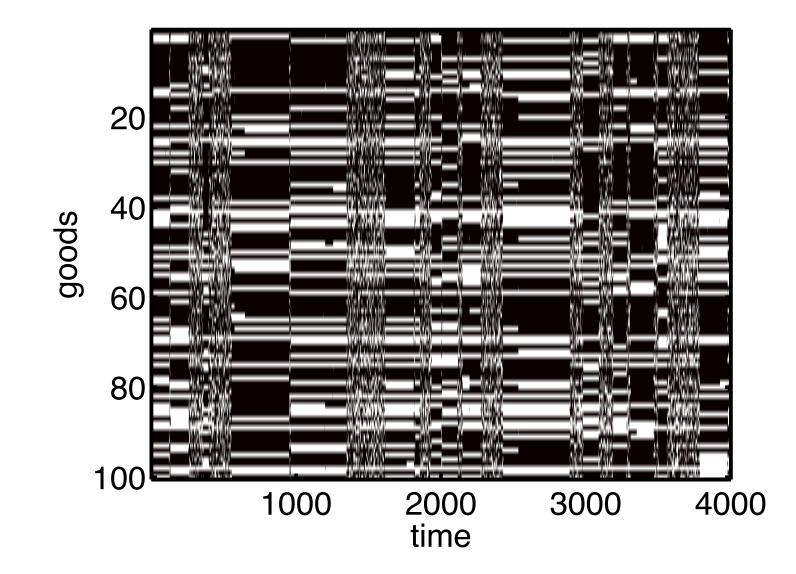
Variables and parameters

Variable	Description	
$\sigma_i(t)$	state of species i . exists / does not exist	dynamic
D(t)	diversity at time t	dynamic
A(t)	active production network	dynamic
~ /		-
Parameter		
α^{\pm}	productive/destructive interaction topology	fixed
r^{\pm}	rule densities	fixed
p	mutation rate / innovation parameter	fixed
		_



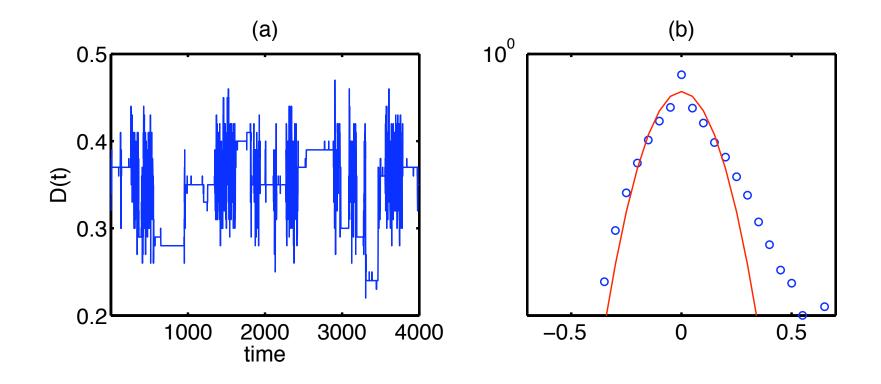


Dynamics



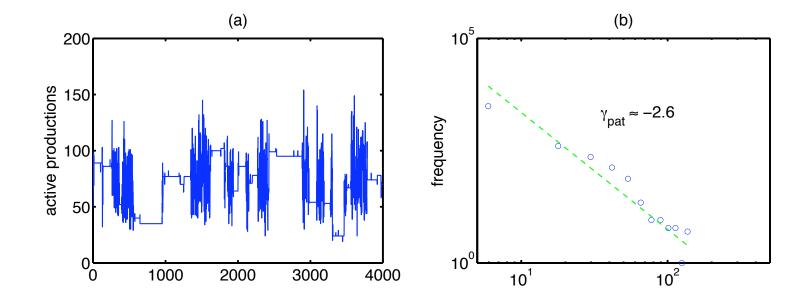


Diversity dynamics



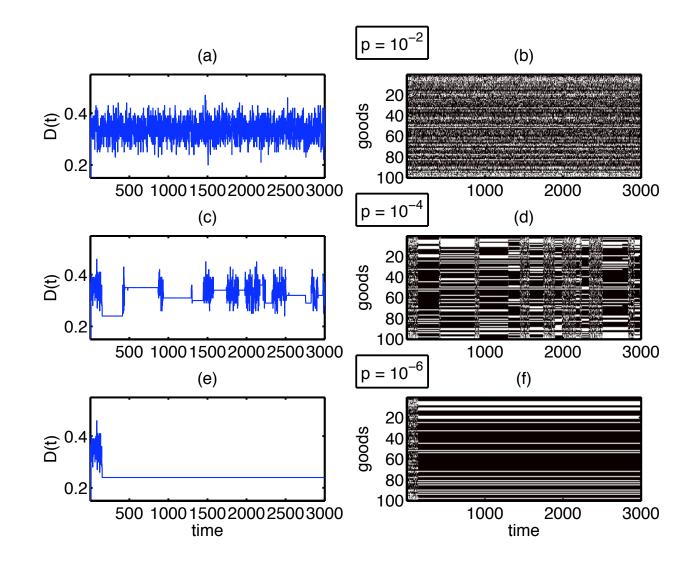


Production dynamics – 'GDP'





Dependence on invention rates





Model variants

• More realistic competition

hierarchical suppression as a more realistic mechanism of competition

• Topology of production and destruction networks

scale-free production/destruction networks Presence of hubs stabilizes system: increases lifetime of plateaus

• Asymmetry in production and destruction

easier to destroy than to 'create': one destructive influence stops production

• Modular structure of production/destruction networks

use modules m (up to ten) with different random topologies r_m^{\pm} linked by a variable number of connecting links



Model variants II

• Finite lifetime of things

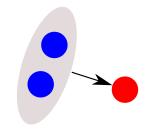
things decay with probability λ (can serve as a stochastic driving force, for p=0)

• Bounded rationality

if a product *can* get produced it does not mean that it actually *will* get produced. To incorporate this possibility we say that if a good can get produced, it will actually get produced with a probability q

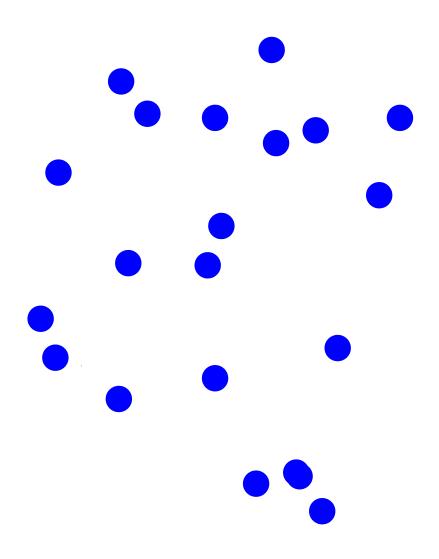
• Variations in the update: robust for parallel update



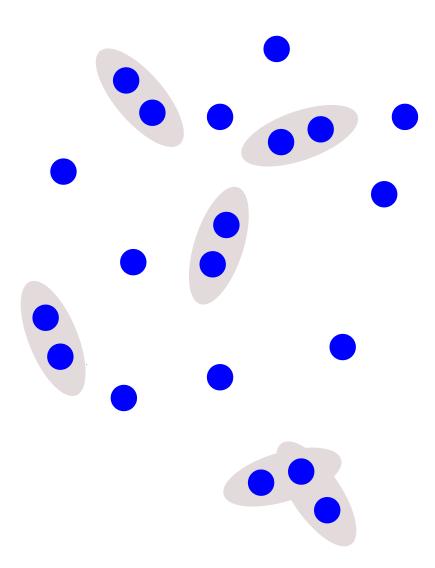




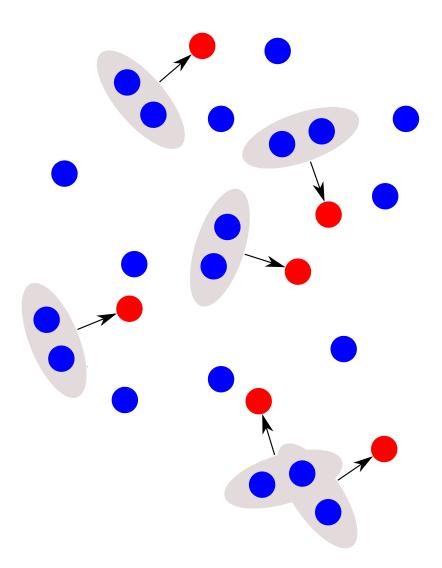
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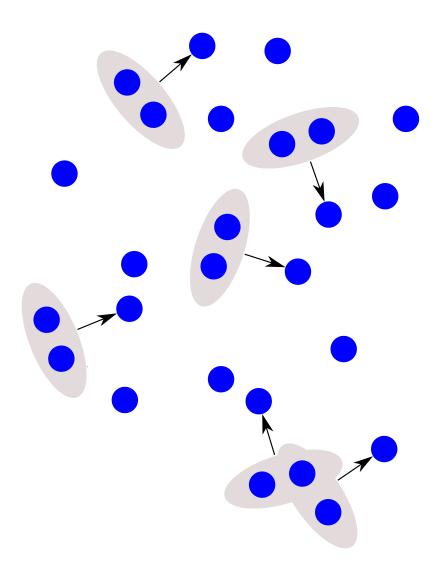




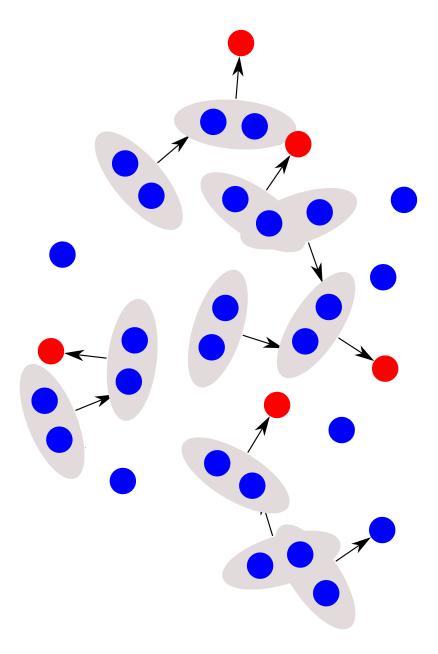














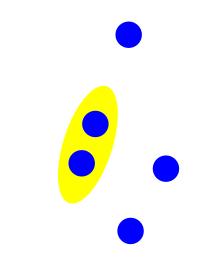
Competition ?

• How to get competition without introducing new ingredients

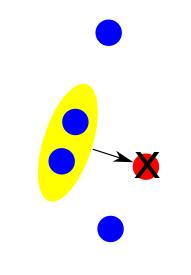
 $\bullet \rightarrow destructive \ pairs$

Taboo: utility, fitness, niches, selection pressure, ...





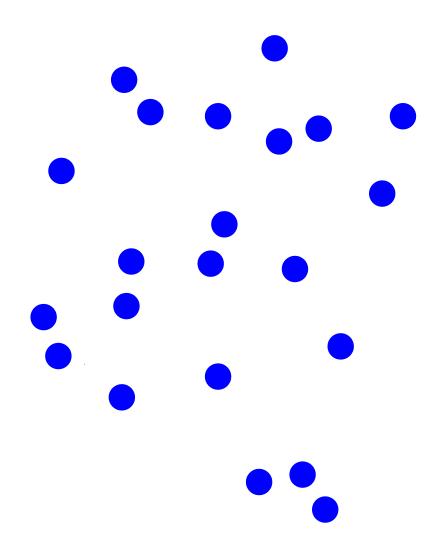




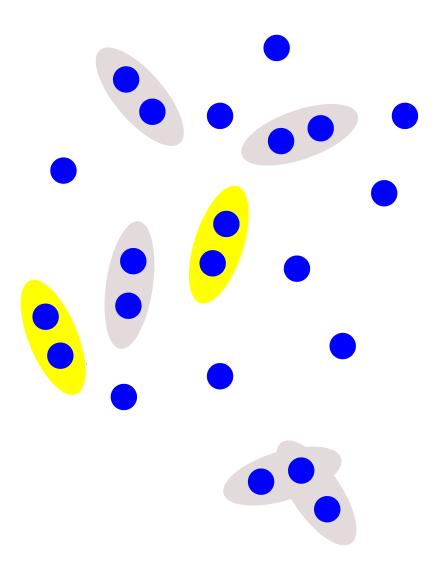


Combine α_{ijk}^+ (constructive) and α_{ijk}^- (destructive) pairs for implicit competition mechanism

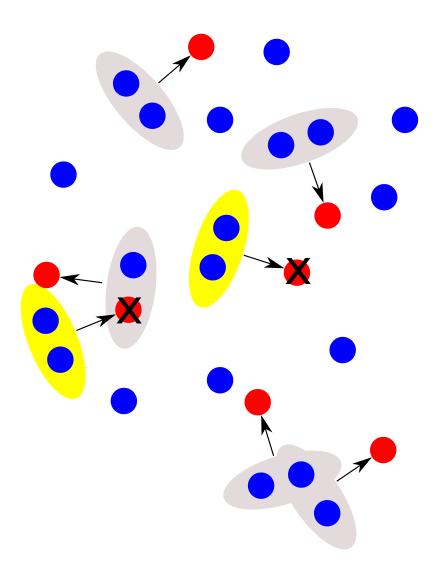




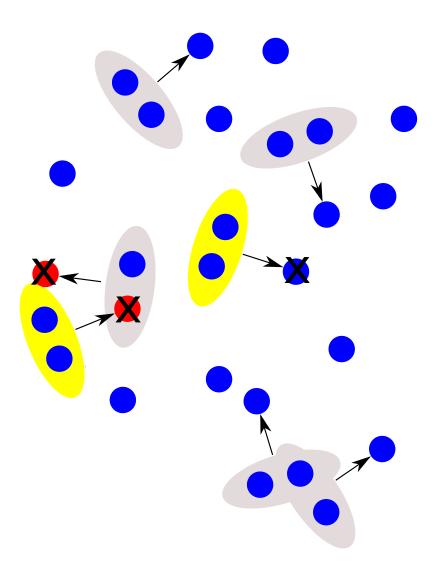




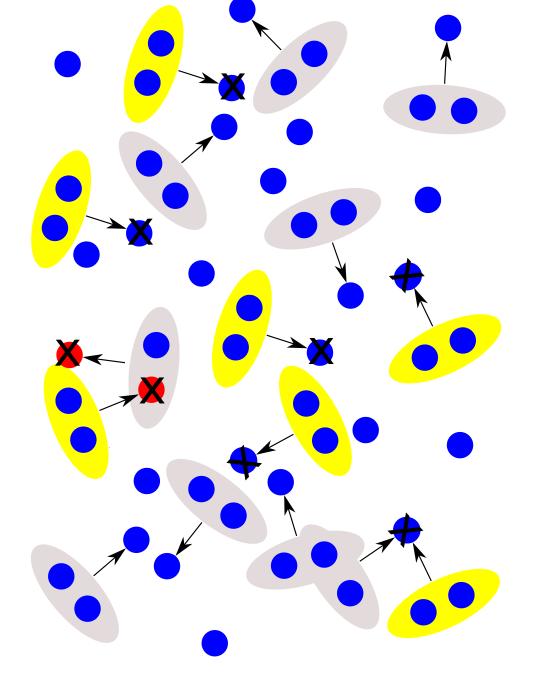




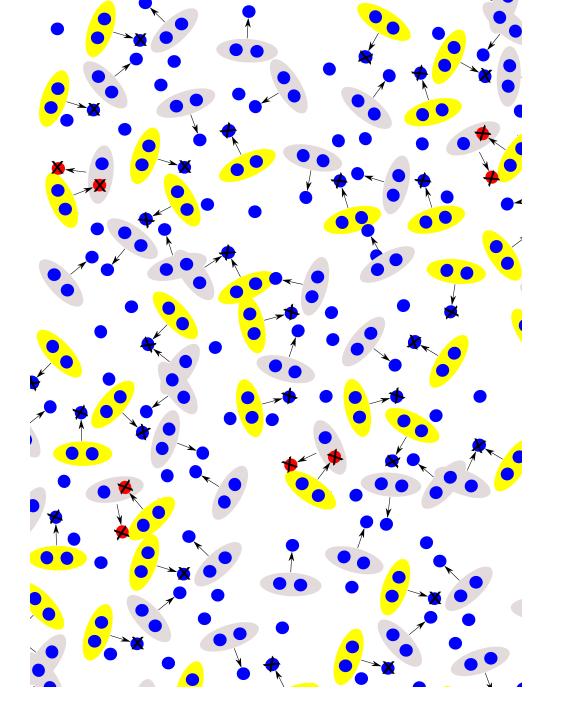




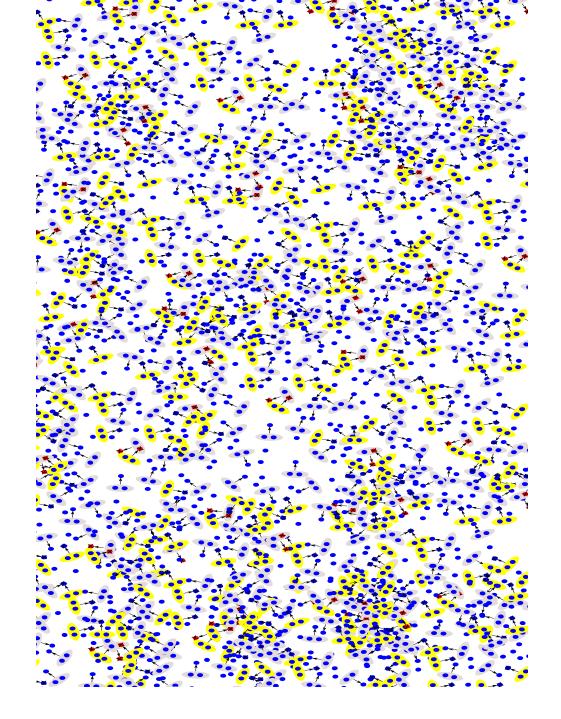




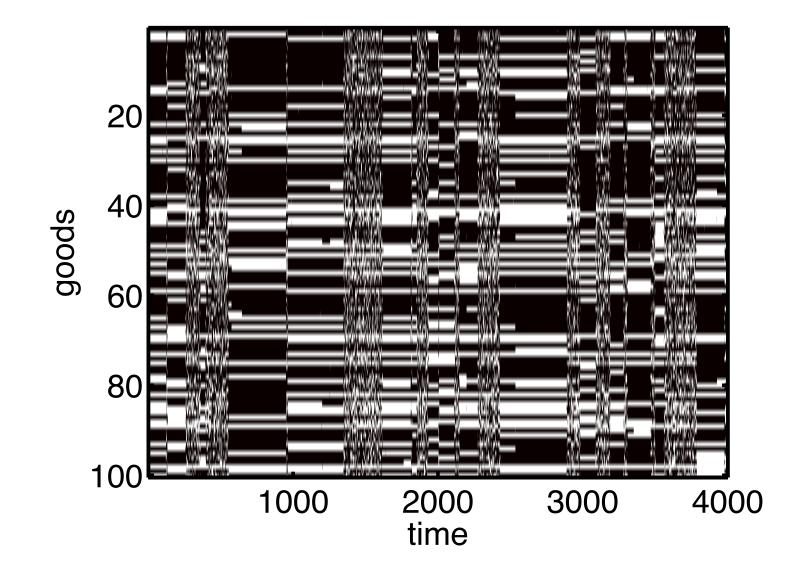






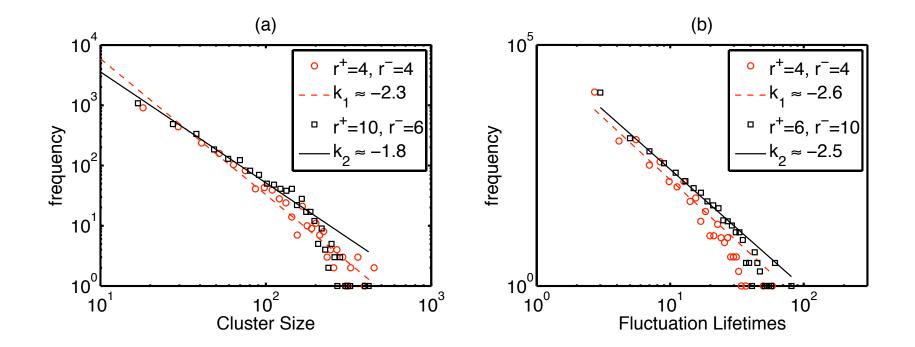








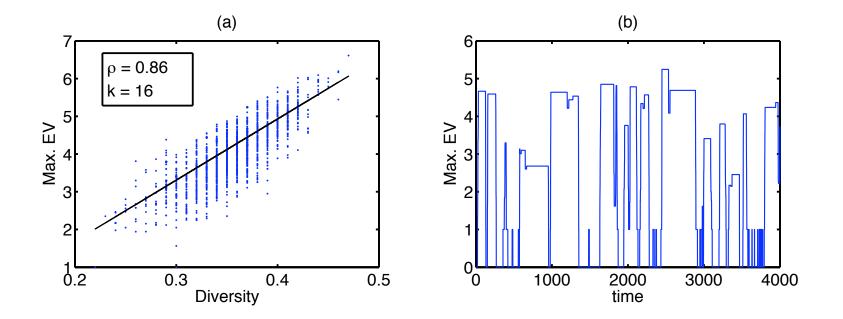
Understanding the dynamics I: it's a sandpile



Set p = 0 – let system relax – flip one state – measure avalanche of activity



Understanding the dynamics II: Eigenvalues & Keystone species

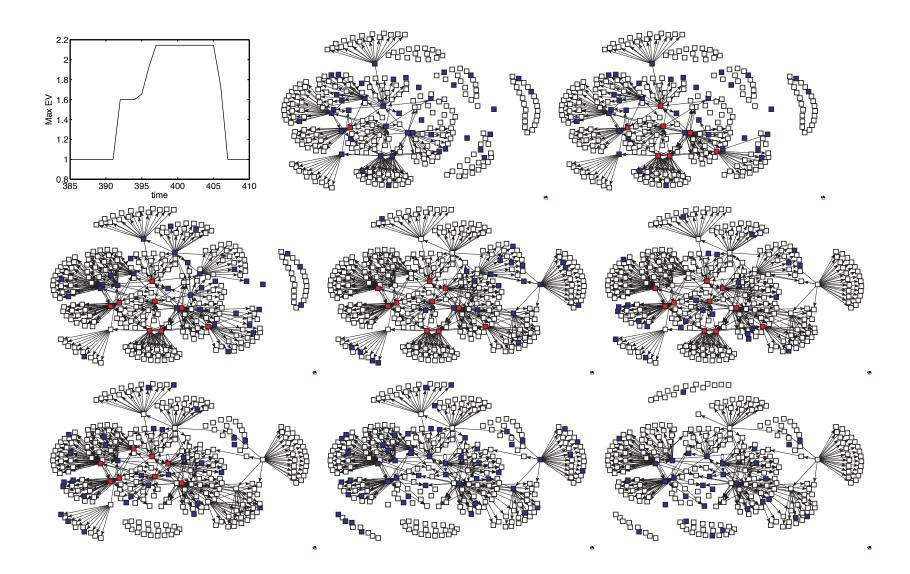


Relate topology of active production network with diversity

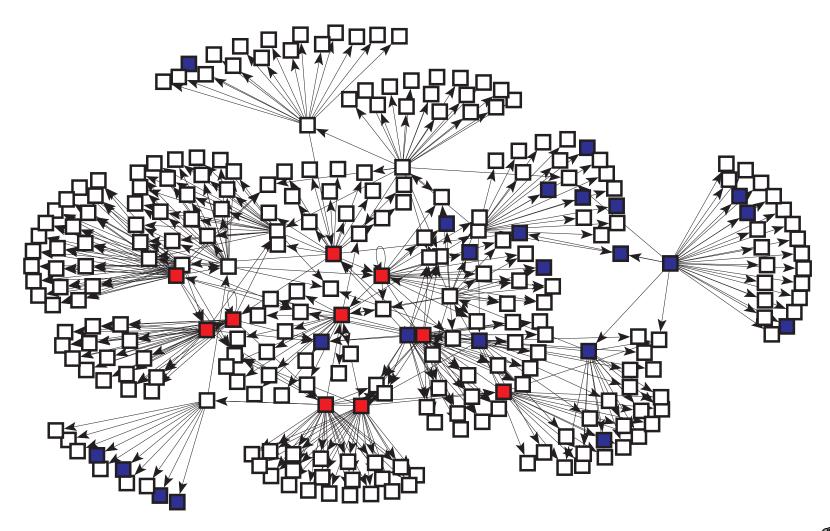
Similar to what was found in Jain&Krishna model of biological evolution Kill keystone species (in autocatalytic core) \rightarrow diversity collapses



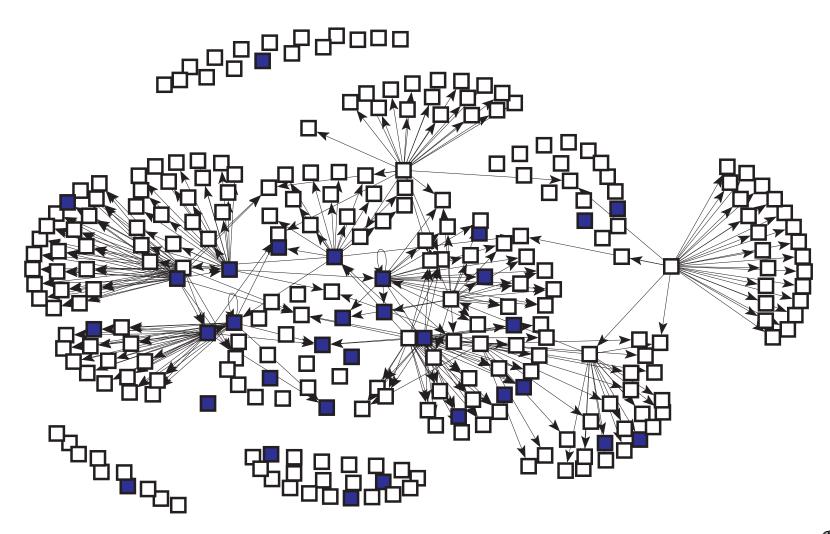
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Pajek



Toward a unified mathematical framework

At each time each element i experiences one of three cases:

(i) creation
$$\sigma_i(t) = 0 \rightarrow \sigma_i(t+1) = 1$$

(ii) annihilation $\sigma_i(t) = 1 \rightarrow \sigma_i(t+1) = 0$
(iii) nothing $\sigma_i(t) = \sigma_i(t+1)$

Imagine a function, $f_i(\vec{\sigma}(t)): \{0,1\}^N \to \mathbb{R}$ to indicate

(i)
$$f_i(\vec{\sigma}(t)) > 0 \Rightarrow \sigma_i(t+1) = 1$$

(ii) $f_i(\vec{\sigma}(t)) < 0 \Rightarrow \sigma_i(t+1) = 0$
(iii) $f_i(\vec{\sigma}(t)) = 0 \Rightarrow \sigma_i(t+1) = \sigma_i(t)$

Use ramp function $R(x) \equiv \max(0, x)$

$$\sigma_i(t+1) = \sigma_i(t) + \Delta \sigma_i(t) ,$$

$$\Delta \sigma_i(t) = \operatorname{sgn} \left[(1 - \sigma_i(t)) R(f_i(\vec{\sigma}(t))) - \sigma_i(t) R(-f_i(\vec{\sigma}(t))) \right]$$



A variational principle for diversity dynamics

Consider displacement: $\sigma'_i(t) = \sigma_i(t) + \delta \sigma_i(t)$

- Quadratic distance: $K_i(\sigma'_i(t), \sigma_i(t)) \equiv \frac{\mu}{2} \left[\sigma'_i(t) \sigma_i(t)\right]^2$
- Potential: $V_i(\sigma'_i(t), \vec{\sigma}(t)) \equiv |(1 \sigma'_i(t))R(f_i(\vec{\sigma}(t))) \sigma'_i(t)R(-f_i(\vec{\sigma}(t)))|$
- Balance function: $B_i \equiv K_i + V_i$ contains all full dynamical information
- Variational principle (Theorem):

Given $\vec{\sigma}(t)$ the solution $\sigma_i(t+1)$ is identical to the value of $\sigma'_i(t)$ for which B_i assumes its minimum, i.e.

$$\sigma_i(t+1) = \operatorname*{argmin}_{\sigma'_i(t)} \left[B_i \left(\sigma'_i(t), \vec{\sigma}(t) \right) \right]$$

with $\mathop{\rm argmin}_x \left[f(x)\right]$ the x for which f(x) takes its minimum



Stochastic formulation

Probability to find a configuration $\sigma_i(t)$

$$p(\sigma_i(t)) \propto e^{-\beta B_i(\vec{\sigma}(t))} \qquad \beta \equiv 1/T$$

Demand detailed balance

$$\frac{p(\sigma_i(t) \to \hat{\sigma}_i(t))}{p(\hat{\sigma}_i(t) \to \sigma_i(t))} = \frac{p(\hat{\sigma}_i(t))}{p(\sigma_i(t))} = e^{-\beta(\hat{B}_i - B_i)}$$

with $\hat{B}_i \equiv B_i(\hat{\sigma}_i(t), \sigma(t)_{j \neq i})$

Now use e.g. Metropolis transition probabilities



Mean-field approximation

Expectation value $q_i(t) = \langle \sigma_i(t) \rangle$

Assume probability factorizes: $p(\vec{\sigma}(t)) = \prod_i p_i(\sigma_i(t))$

Boltzmann-Gibbs entropy s for element i is

$$s(\sigma_i(t)) = -\langle \ln p_i(\sigma_i(t)) \rangle \equiv s(q_i(t))$$

Free energy

$$\phi(q_i(t)) = \langle B_i \rangle_{p(\vec{\sigma}(t))} - \frac{s(q_i(t))}{\beta}$$

Asymptotic state of species $i,~q_i(t\rightarrow\infty)\equiv q_i$ at minimum in ϕ

Necessary condition
$$\partial \phi(q_i) / \partial q_i = 0$$
, is $\frac{\partial \langle B_i \rangle}{\partial q_i} + \frac{1}{\beta} \ln \left(\frac{q_i}{1 - q_i} \right) = 0$

$$q_i = \frac{1}{2} \left\{ \tanh\left[-\frac{\beta}{2} \frac{\partial \langle B_i \rangle}{\partial q_i}\right] + 1 \right\}$$

The self-consistent solution yields the asymptotic configuration



Mean-field approximation II

Calculate $\langle B_i \rangle$ for random interaction topologies rule densities: r^{\pm}

constructive / desructive set size: n^\pm

$$\langle K_i \rangle_{p(\sigma)} = \frac{1}{2} \left((1 - q_i)p^+ + q_i p^- \right)^2$$

and

$$\langle V_i \rangle_{p(\sigma)} = |(1 - q_i) R(f_i) - q_i R(-f_i)|$$

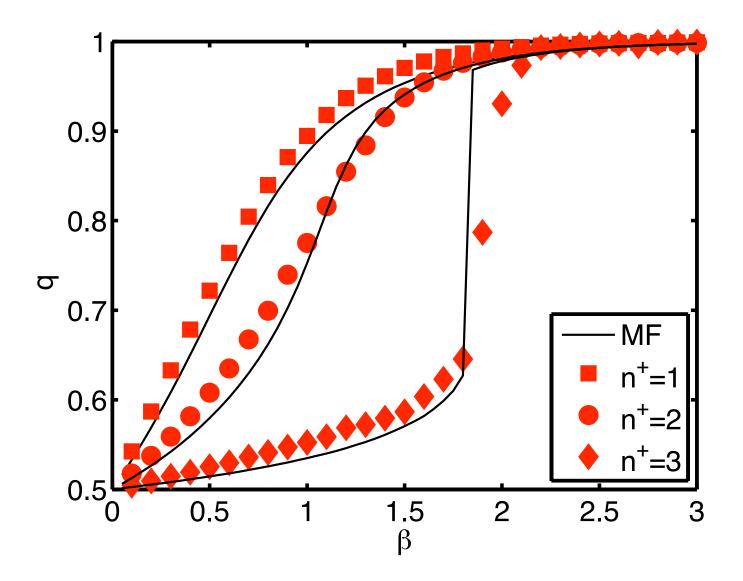
Take derivative w.r.t. q_i . MF result is the self-consistent solution for q

$$q = \frac{1}{2} \left\{ \tanh\left[\frac{\beta}{2} \left(r^{+} q^{n^{+}} - r^{-} q^{n^{-}} + \left[(1-q)p^{+} + qp^{-}\right](p^{+} - p^{-})\right)\right] + 1 \right\}$$



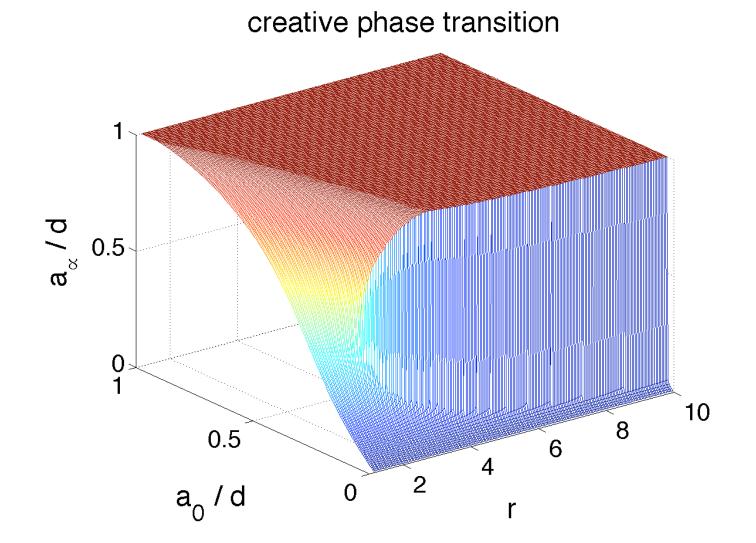
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Mean-field approximation III

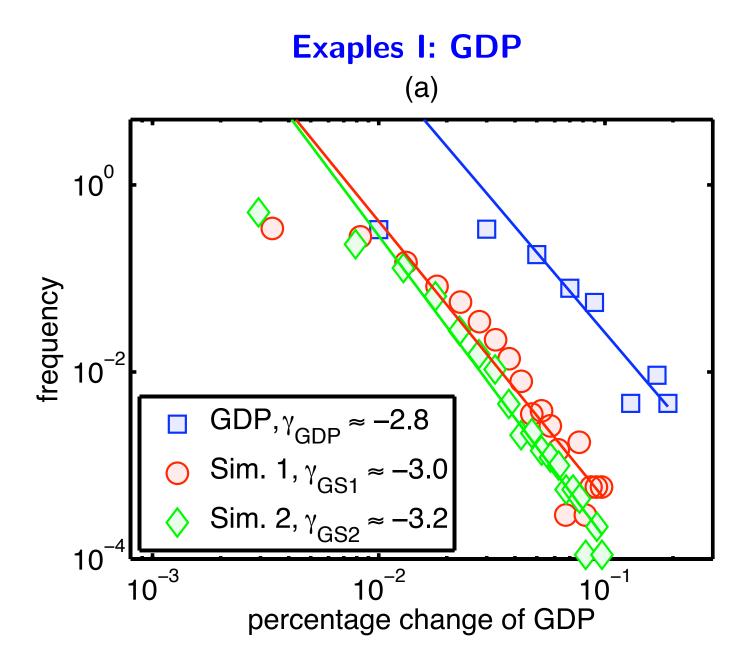




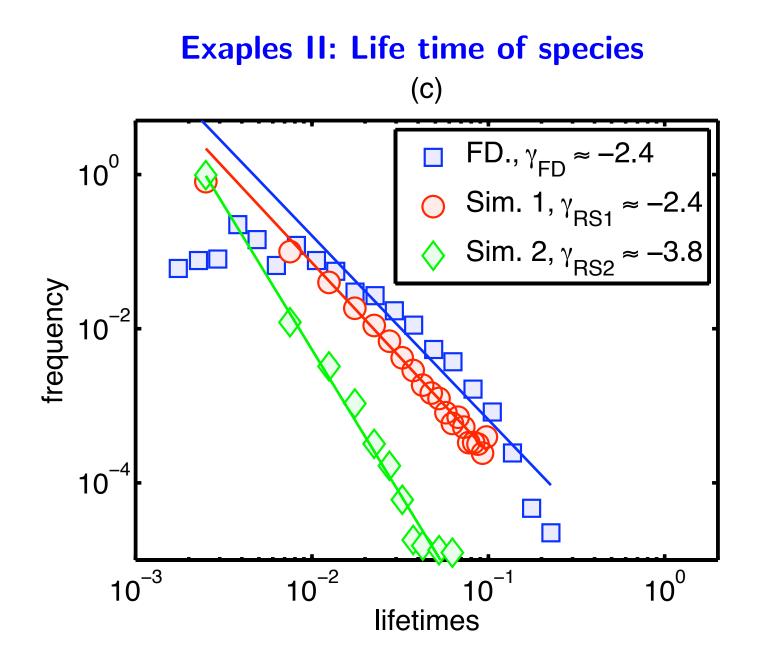
Phase diagrams of evolution systems



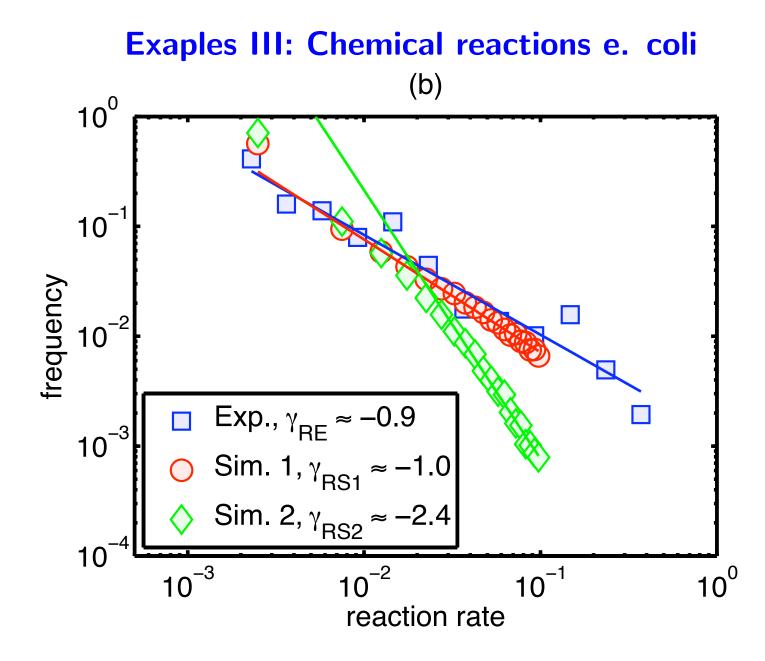














Conclusions

- identified problem with traditional evolutionary thinking
- proposed model: overlay production map with states
- formulate as a spin system
- compute its phase diagrams and compare to simulations
- show that actual timeseries are within predictions of the model

