



Are Copying and Innovation Enough?

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Outline

- Part 1: Simple Copying Model
 - Why Copying?
 - Introduce a simple network model with only copying and innovation
 - Show how this model can be solved exactly for all times
 - Properties of the network: Homogeneity
- Part 2: Generalisations
 - Additional vertex graphs
 - Update Dynamics
 - Additional Vertex types (time permitting!)

Why Copying?

Local process yet produces macroscopic features

- Copying can lead to preferential attachment i.e. with probability proportional to degree k [e.g. T.S. Evans & J.P. Sarämäki (2005)]→ scale free behaviour

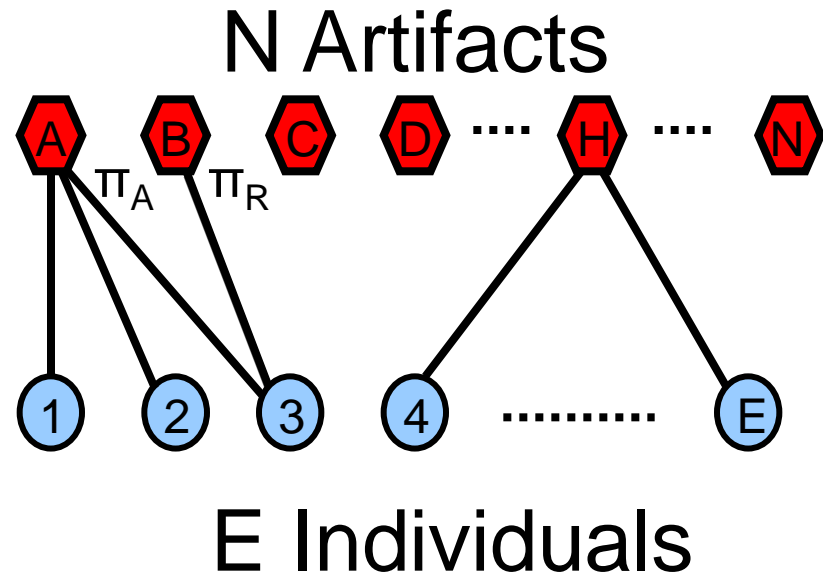
Copying and Innovation natural mechanisms in many models

- Gene Frequencies: Inheritance and Mutation [M. Kimura & J.F. Crow (1964)]
- Urn models
- Voter Models: e.g. Language Extinction [Stauffer et al. (2007)]
- Cultural Transmission: Baby names, dog breeds, pop music... [e.g. Bentley et al. 2007]

Interesting from a mathematical point of view – simple models with just copying and innovation can be exactly solved.

Model - Simple non-growing bi-partite network

- Edges fixed to Individuals but Artifact side free to rewire
- Choose an Individual to remove edge, with probability π_R
- Choose Artifact to attach edge to, with probability π_A
- Only then perform rewiring




Model - Mean Degree Distribution Master Equation

- Interested in degree distribution $n(k,t)$ (and $p(k,t)=n(k,t)/N$) of Artifact Vertices.


Mean Field Equation:

$$\begin{aligned}
 n(k, t + 1) - n(k, t) = & + n(k + 1, t) \Pi_R(k + 1) [1 - \Pi_A(k + 1)] \\
 & - n(k, t) \Pi_A(k) [1 - \Pi_A(k)] \\
 & - n(k, t) \Pi_R(k) [1 - \Pi_R(k)] \\
 & + n(k - 1, t) \Pi_A(k - 1) [1 - \Pi_R(k - 1)]
 \end{aligned}$$

Number of edges
attaching to vertex
of degree (k-1)



Probability of
NOT reattaching
to same vertex



Model - Mean Degree Distribution Master Equation

- Good approximation when vertex correlations small
- For Probabilities $\propto k^\beta/z_\beta$, exact when normalisations constant, i.e.

$$\sum_k n_i(k) k^\beta = \left\langle \sum_k n_i(k) k^\beta \right\rangle$$

- ONLY when $\beta=0$ or $\beta=1$, thus most general π_R and π_A

$$\Pi_R = \frac{k}{E}, \quad \Pi_A = p_r \frac{1}{N} + p_p \frac{k}{E}$$

Only Random or Preferential Attachment ($p_r + p_p=1$)

Model - Solutions

- Mean-field equation is linear
→ Encode degree distribution in Generating function

$$G(z, t) = \sum_{k=0}^E z^k n(k, t) = \sum_{m=0}^E c_m (\lambda_m)^t G^{(m)}(z)$$

- Re-write mean-field equation as differential equation:

$$z(1-z)G^{(m)''}(z) + [c - (a+b+1)z]G^{(m)'}(z) - \left[ab - \frac{(\lambda_m - 1)}{1-z} b(c-a-1) \right] G^{(m)}(z) = 0$$

Coefficients: $a = \frac{p_r}{p_p} \langle k \rangle$, $b = -E$, $c = 1 + a + b - \frac{p_r}{p_p} E$

Model - Solutions

- Expand $G^{(m)}(z)$ as a polynomial in $(1-z)$. Leads to Hypergeometric differential equations with solutions

$$G^{(m)}(z) = (1-z)^m F(a+m, b+m; c; z)$$

and Eigenvalues given by

$$\lambda_m = 1 - m(m-1) \frac{p_p}{E^2} - m \frac{p_r}{E}$$

$$\lambda_m \geq \lambda_{m+1}$$

Model - Equilibrium Distribution

- Can recover degree distribution from generating function.

$$n(k) = \frac{1}{k!} \left. \frac{d^k G(z)}{dz^k} \right|_{z=0} = A \frac{\Gamma(k + \bar{K})}{\Gamma(k + 1)} \frac{\Gamma(E - \bar{E} - \bar{K} - k)}{\Gamma(E + 1 - k)}$$
$$\bar{K} = \frac{p_r}{p_p} \langle k \rangle \quad \bar{E} = \frac{p_r}{p_p} E$$

A – Normalisation. Ratio of Gamma functions.

- Exact for all parameter values of E, N and p_r .

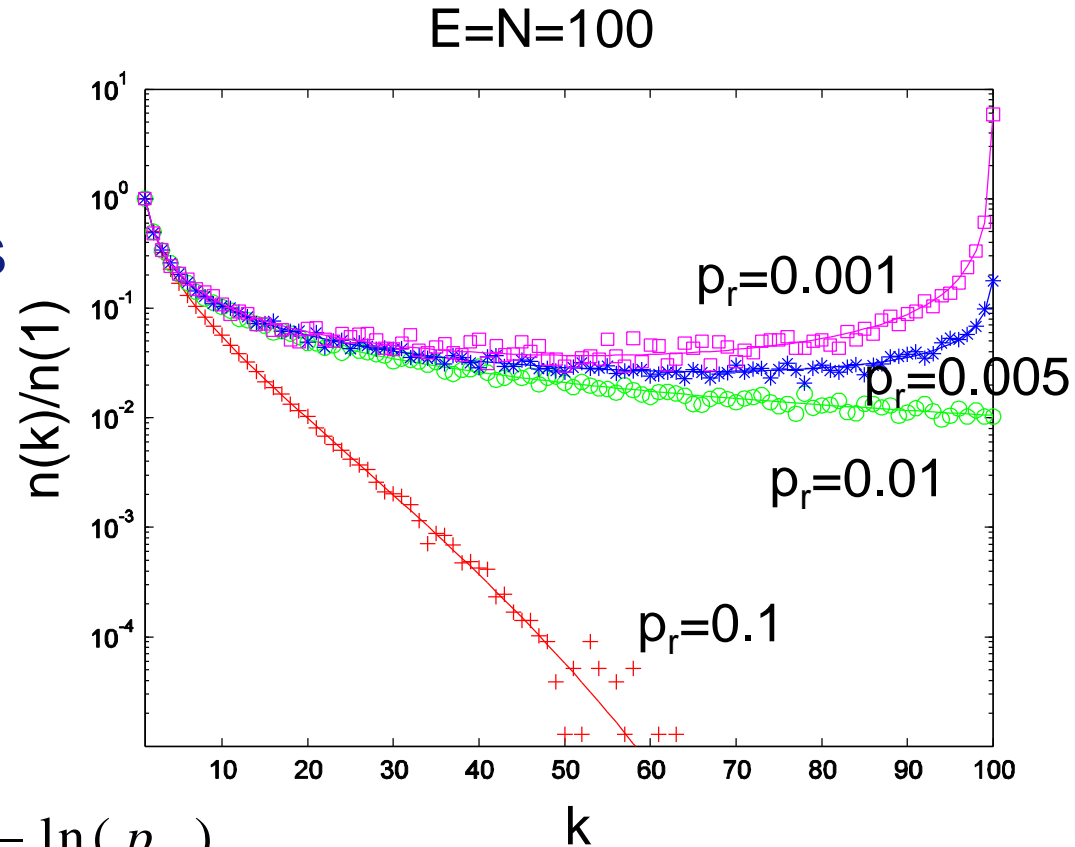
Model - Equilibrium Distribution

- Numerical: Averaged over 10^5 realisations
Lines: Analytic Results
- $p_r \geq 1/E$: power law

$$\lim_{k \rightarrow \infty} n(k) = k^{-\gamma} \exp(-\zeta k)$$

$$\text{with power } \gamma = 1 - \frac{p_r}{p_p} \langle k \rangle$$

$$\text{and exponential cutoff } \zeta = -\ln(p_p)$$



Homogeneity Measures

- Call the probability that two *distinct* edges chosen at random connected to same Artifact, $F_2(t)$

$$F_2(t) = \sum_{k=0}^E \frac{k(k-1)}{E(E-1)} n(k,t) = \frac{1}{E(E-1)} \left. \frac{d^2 G(x,t)}{dx^2} \right|_{x=1}$$

- E fixes $c_1=0$

$\longrightarrow F_2(t) = F_2(\infty) + (\lambda_2)^t (F_2(0) - F_2(\infty))$
 $F_2(0)$ fixed by initial conditions

$$F_2(\infty) = \frac{1 + p_r (\langle k \rangle - 1)}{1 + p_r (E - 1)}$$

- Generalise to n distinct edges $F_n(t) = \frac{\Gamma(E+1-n)}{\Gamma(E+1)} \left. \frac{d^n G(x,t)}{dx^n} \right|_{x=1}$

- $g_n^{(m)}$: n^{th} derivatives of $G^{(m)}(z)$ at $z=1$. Only non zero for $m \leq n$ so contributions only from first $n+1$ eigenfunctions:

Homogeneity Measures

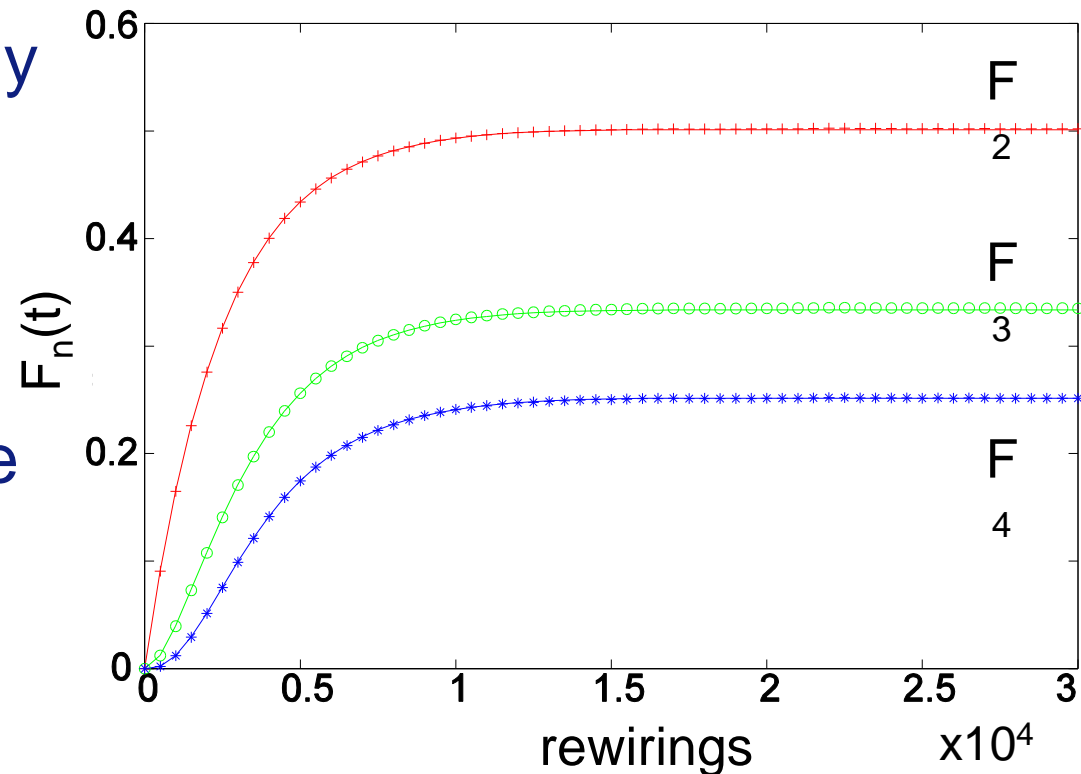
- F_n measures can fix coefficients c_m iteratively

$$\frac{\Gamma(E + 1 - n)}{\Gamma(E + 1)} \sum_{m=0}^n c_m g_n^{(m)} = F_n(0)$$

$$g_n^{(m)} = 0, \quad m > n$$

- Network becomes more homogenous with time

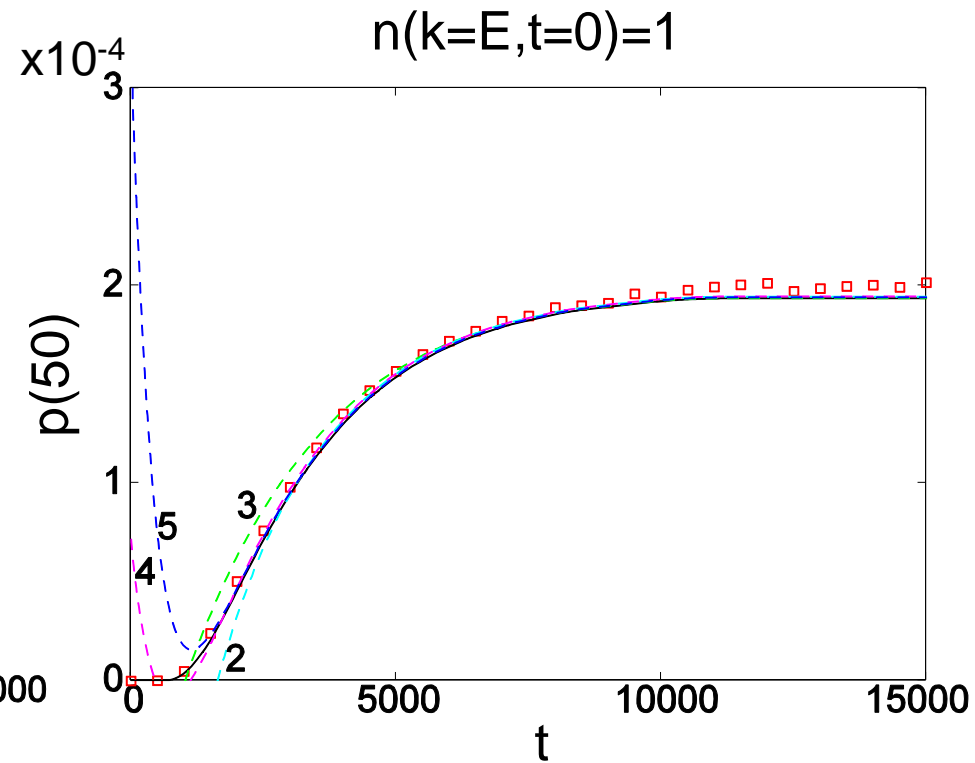
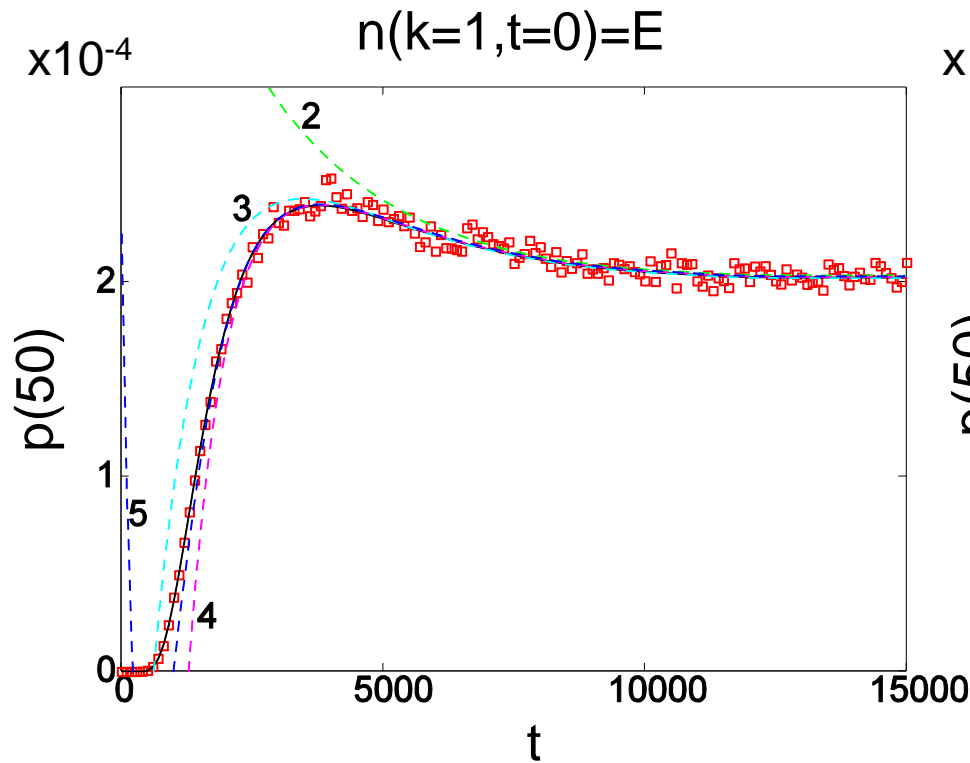
$E=N=100$, $p_r=1/E$. $n(k)=\delta_{k,1}$ averaged over 10^5 runs



Time Dependence

- Low Eigenvalue contribution to the degree distribution evolution

$N=E=100$, $p_r=1/E$



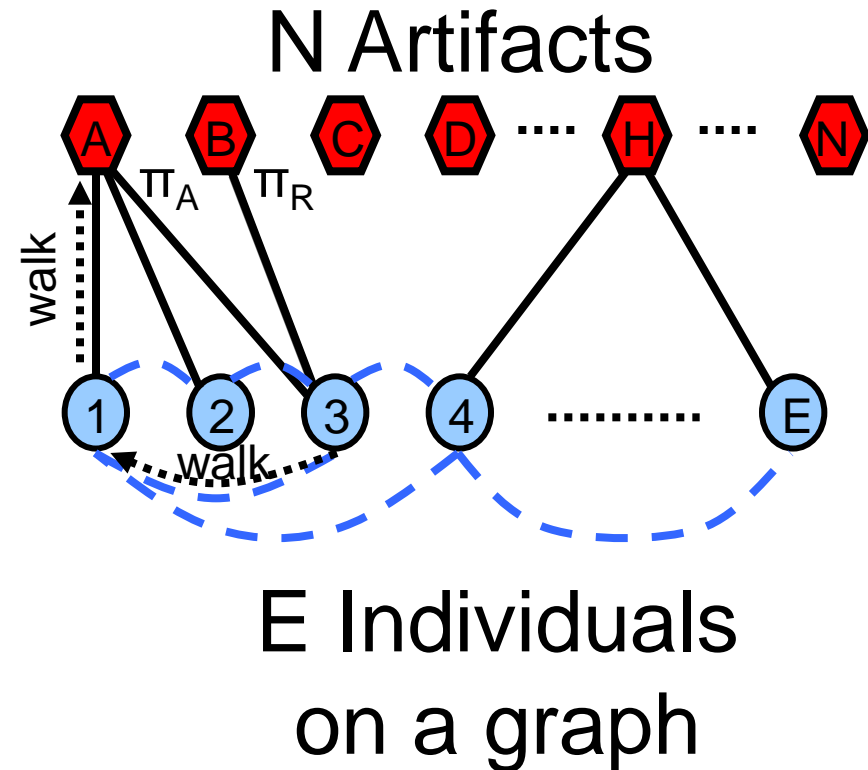
Part II:

Generalisations

- Add a graph to the Individuals Vertices
- Different update dynamics
 - Simple example: 2 step update
 - Multi edge rewiring
- Different Types of Individuals

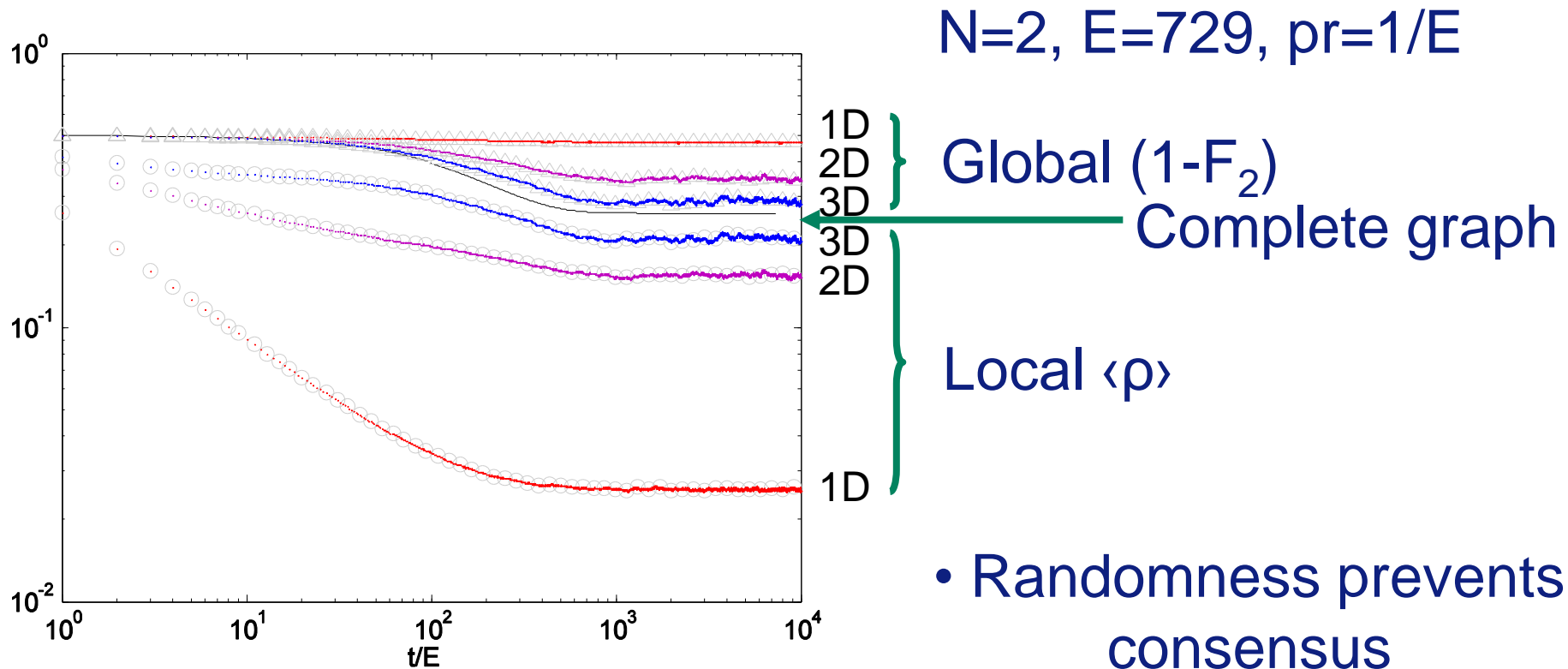
Network of Individuals

- Individuals vertices now connected by a graph
- Choose removal edge with $\pi_R = k/E$
- Preferential Attachment is replaced by a random walk on Individuals network
- Copy vertex choice
- Rewire



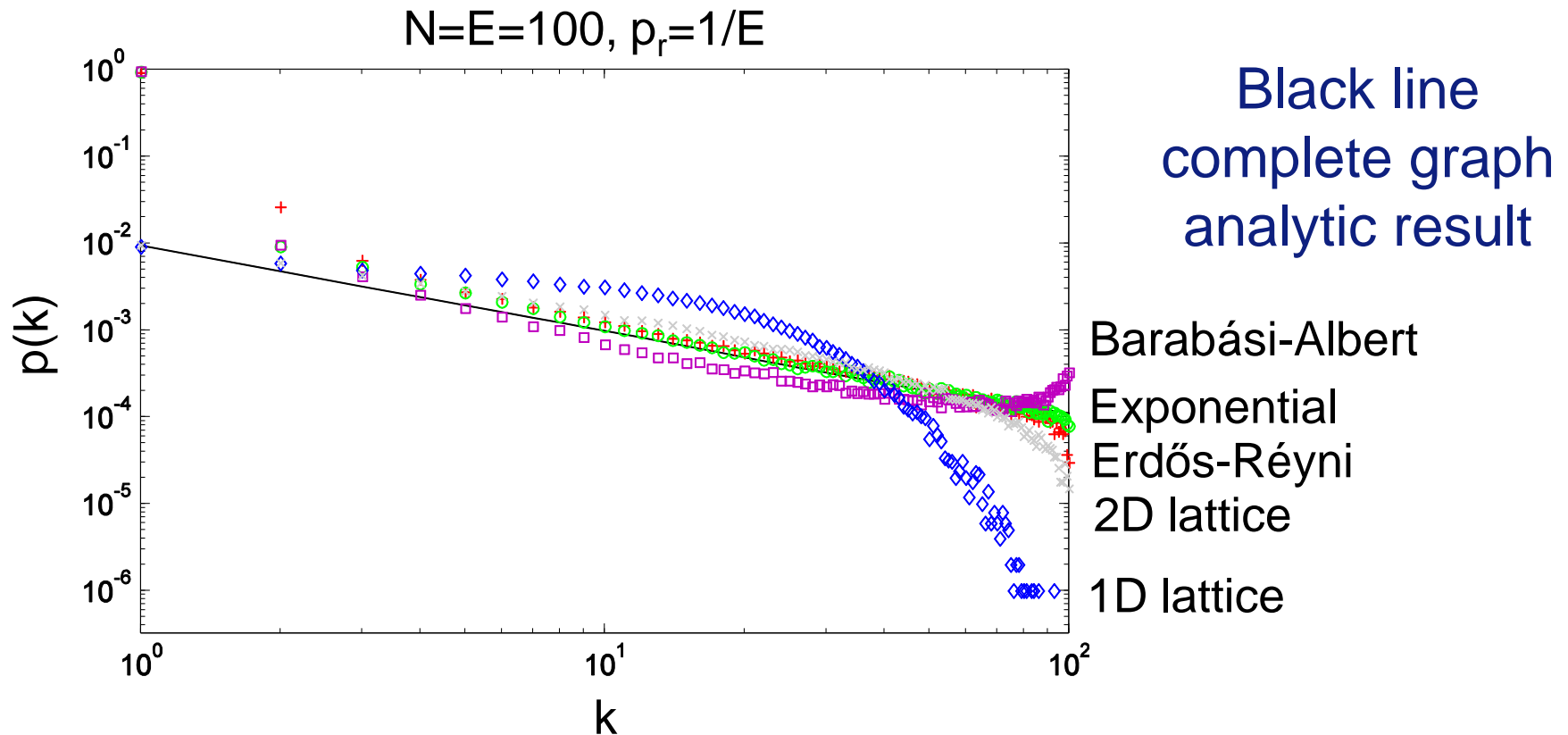
Network of Individuals – Homogeneity

- $N=2$: traditional Voter Model with randomness
- Average interface density $\langle \rho \rangle_t$
- $(1-F_2(t))$ vs. $\langle \rho \rangle_t \leftrightarrow$ Global vs. local



Network of Individuals – Degree Distribution

- Degree Distribution similar except for 1D lattice



Alternate Update dynamics – two step update

Simple Example. Step 1: Remove edge

Step 2: Attach new edge

- New attachment probability: $\tilde{\Pi}_A = p_r \frac{1}{N} + p_p \frac{k}{E-1}$
- Modified master equation.

However...

Solution still similar to before

$$G^{(m)}(z) = (1-z)^m F(\tilde{a} + m, b + m; \tilde{c}; z)$$

Minor change in parameters $a \rightarrow \tilde{a} = \frac{p_r}{p_p} \frac{E-1}{N}$, $b = -E$,
 $c \rightarrow \tilde{c} = 1 + \tilde{a} + b - \frac{p_r}{p_p} E$

Alternate Update dynamics – Multi Edge Rewiring

In more realistic models up to date information may not be available for each entity

- Parameter X controls number of distinct edges to be rewired at each timestep
- Now choose edges at Random or Sequentially
- Extremes: $X=1 \rightarrow$ Simple Copy Model already discussed
 $X=E \rightarrow$ Bentley *et al.* generational rewiring,
Fisher-Wright Model

Alternate Update dynamics – Multi Edge Rewiring

Sequential and Random updates.

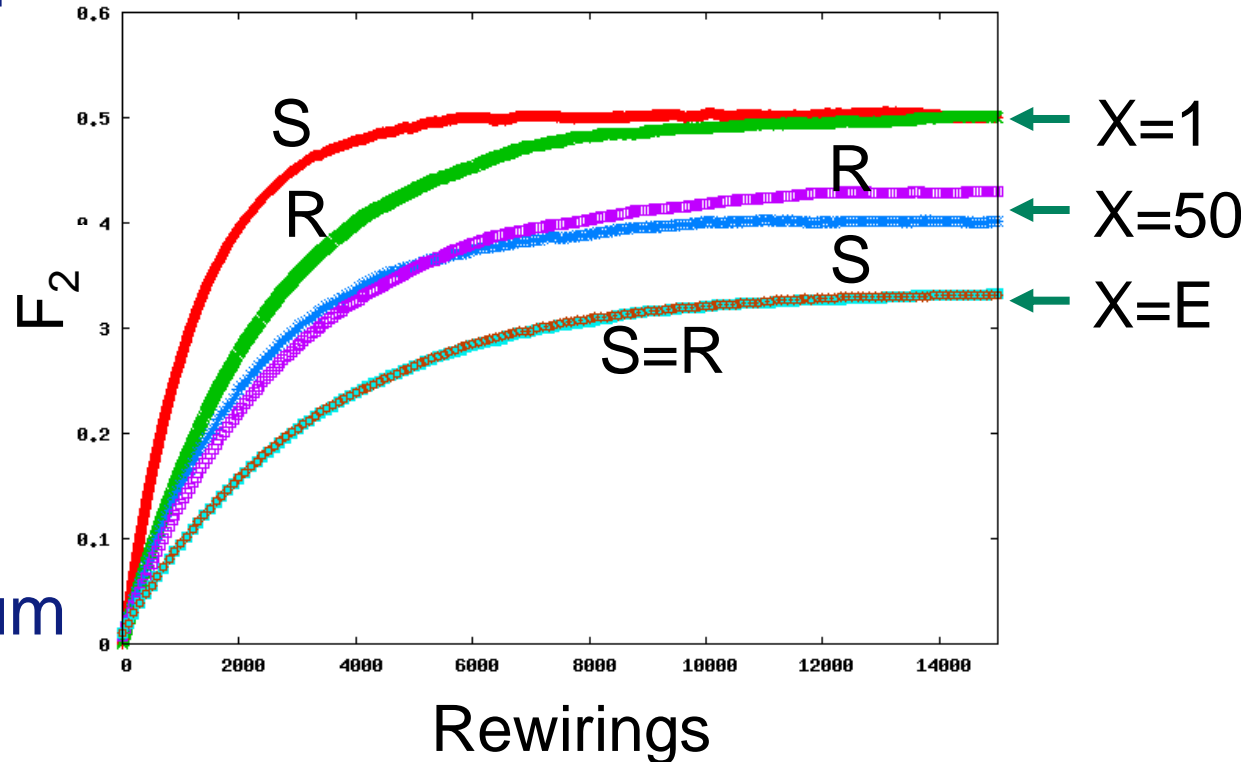
$E=N=100, p_r=1/E$

- Numerical results:

- $X=1$: Equilibrium identical, but sequential update faster

- $X=E/2$: Similar timescales, but different Equilibrium F_2

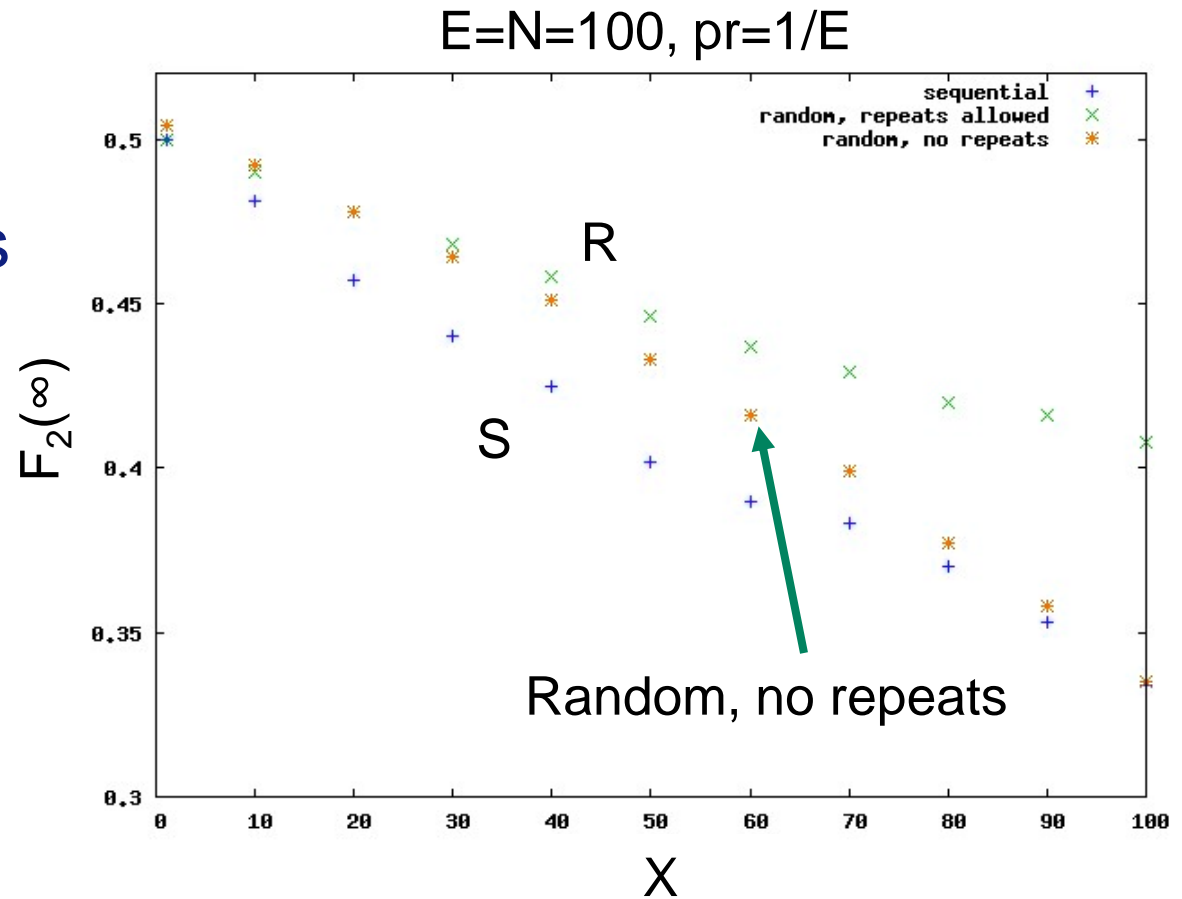
- $X=E$: Sequential = Random



$$F_2(\infty) = \frac{p_p^2 + (1 - p_p^2) \langle k \rangle}{p_p^2 + (1 - p_p^2) E}, \quad \lambda_2 = \frac{p_p^2 (E - 1)}{E}$$

Alternate Update dynamics – Multi Edge Rewiring

- Homogeneity decreases with number of edges rewired



Two Individual Types

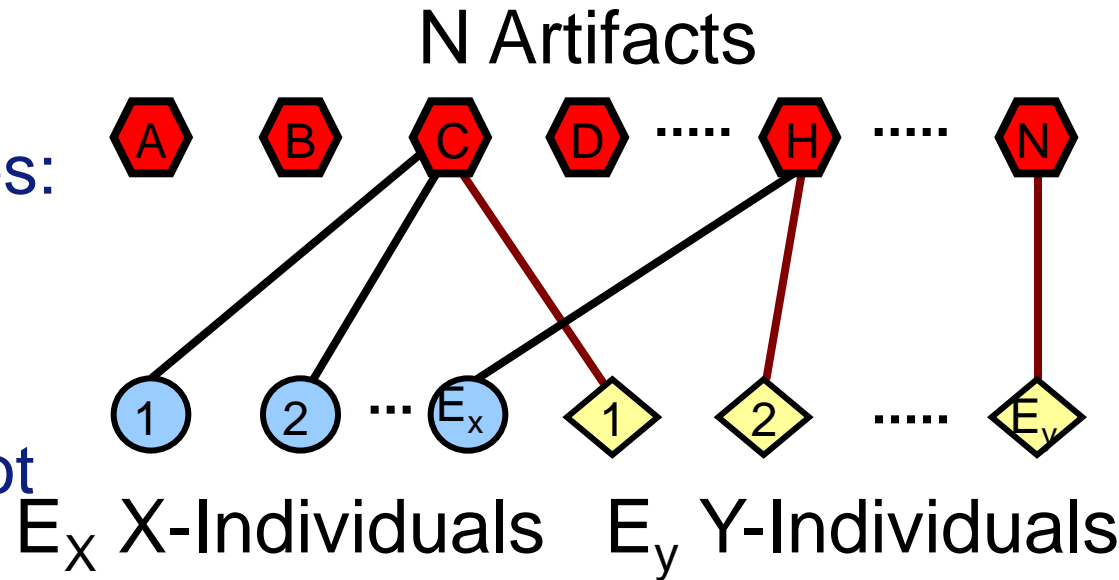
Artifacts have k_x and k_y degree and now four independent probabilities:

$$p_{rx} + p_{p_{xx}} + p_{p_{xy}} = 1,$$

$$p_{ry} + p_{p_{yy}} + p_{p_{yx}} = 1.$$

- Complete solutions not available

- Some progress can be made on F_{mn} measures, but lead to lengthy algebraic solutions



Summary

- Simple models can be solved Exactly
- Even with just copying and innovation, a large amount of variation in modelling is possible
- Rewiring dynamics is important
- Many other models can be mapped to a simple network model, and some properties studied analytically, e.g. Voter Model

References

- T.S.Evans & ADKP
“Exact Solution for the Time Evolution of Network Rewiring Models”
Phys. Rev. E **75** (2007) 056101
[**cond-mat/0612214**]
- T.S.Evans & ADKP
“Network Rewiring Models”
Networks and Heterogeneous Media 3, 2 (2008) 221.
arXiv:0707.3783
- T.S.Evans & ADKP
“Exact Solutions for Models of Cultural Transmission and Network Rewiring” (for ECCS06)
[**physics/0608052**]
- T.S.Evans, ADKP, & T.You (in prep)

Solutions – Time Dependence

- Now $n(k, t) = \frac{1}{k!} \left. \frac{d^k G(z, t)}{dz^k} \right|_{z=0}$

- Must include contribution from all eigenvalues

$$G^{(m)}(z) = \sum_{k=0}^E z^k \omega^{(m)}(k)$$

- Cumbersome, but can in principle give full time dependence exactly

$$\omega^{(m)}(k) = (-1)^m \frac{\Gamma(k+1)}{\Gamma(k+1-m)} \frac{\Gamma(c+k)}{\Gamma(c+k-m)} \frac{\Gamma(a)}{\Gamma(a+m)} \frac{\Gamma(b)}{\Gamma(b+m)}$$

$$\times {}_3F_2(-m, a+k, b+k; k+1-m, k+c-m; 1) \omega^{(0)}(k)$$