

#### Are Copying and Innovation Enough?

Main authors on different aspects of work discussed here:-

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#### Notation

- I will focus on Simple Graphs with multiple edges allowed (no values or directions on edges, no values for vertices)
- **N** = number of vertices in graph
- **E** = number of edges in graph
- **k** = degree of a vertex
- <k> = average degree = 2E/N
- Degree Distribution

   n(k) = number of vertices with degree k
   p(k) = n(k)/N = normalised distribution

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**Degree** *k*=2

# Growing Scale-Free Networks

- Random Walks and Copying
  - The Origin of Scale-Free Networks?
- Copying Model
  - General Features
  - Equilibrium Solutions
  - Time Evolution
- Summary



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Long Tails = Hubs





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### Growth with Preferential Attachment

(Yule 1925, 1944; Simon 1955; Price 1965,1976; Barabasi,Albert 1999)

- Add new vertex attached to one end of <sup>1</sup>/<sub>2</sub><k> new edges
- 2. Attach other ends to existing vertices chosen with by picking random end of an existing edge chosen randomly, so probability is

Π(k) = k / (2E) Preferential Attachment "Rich get Richer"

Π**(k) 2**/(2E) 5/(2E) **4/(2E) 2**/(2E)

> **Result: Scale-Free** *n(k)* ~ *k*<sup>-γ</sup> γ=3

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#### Scale-Free Growing Model comments

- Network not essential *k*=frequency of previous choices
- Generalised attachment probability

$$\Pi(k) = (1 - p_r) \frac{k}{2E} + p_r \frac{1}{N},$$

Preferential Attachment

Random Attachment

with new vertex added fraction  $\epsilon$  of the time gives power laws with powers from 2 to  $\infty$ 

 BUT if lim<sub>k</sub> Π(k) ∝ k<sup>α</sup> for any α≠1 then a power law degree distribution is not produced!

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### Walking to a Scale-Free Network

[Saramäki, Kaski 2004; TSE, Saramäki 2004]

- Add a new vertex with <sup>1</sup>/<sub>2</sub><k>
  new edges
- 2. Attach to existing vertices, found by executing a random walk on the network of *L* steps



è Probability of arriving at a vertex
 ∞ number of ways of arriving at vertex

# = k, the degree

 $\Rightarrow Preferential Attachment \gamma = 3$ 

(Can also mix in random attachment with probability  $p_r$ )

Naturalness of the Random Walk algorithm

Automatically gives preferential attachment for any shape network and hence tends to a scale-free network

- Uses only LOCAL information at each vertex
  - Simon/Barabasi-Albert models use global information in their normalisation
- Uses structure of Network to produce the networks
  - a self-organising mechanism
     e.g. informal requests for work on the film actor's social network
     e.g. finding links to other web pages when writing a new one

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• Walks of length ONE are usually sufficient to generate reasonable scale-free networks

 $\Rightarrow$  Degree Correlation Length  $\xi < 1 < d$  (any distance scale)

Is the Walk Algorithm Robust? YES

- Different starting points
- Vary length of walks per edge keep L=<L> fixed
  Vary edges added per vertex keep <k> fixed
- Allow multiple edges

Good Power Laws but power varies by 10% or 20%





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### Copying Model – General Features

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# Copying

Copying is an intrinsically local process, no global information used yet produces distinctive macroscopic features.

o Preferential Attachment ⇔ Copying

 e.g via random walk (TSE+Saramäki 2005)

 o Rewiring of Networks of fixed size (*N*, *E*)

 vs. Growing Networks

o Example of how to get exact solutions for finite sized graphs at any time

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A Simple Model of Cultural Transmission

- Fixed population of *E* individuals
- Each person chooses one of **N** artifacts
  - Artifacts have no intrinsic benefit
  - e.g. pedigree dog, shoe style, name for baby
- At each time step, one random person updates their choice using one of two methods:-

(a) COPYING someone else's choice

(b) INNOVATING, picking an artifact at random it will be one no one else has chosen if *N* large

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The Model as Network Rewiring

- Removal: Choose an individual at random
   = choosing departure artifact with probability Π<sub>R</sub>=(k/E)
   = preferential removal from artifacts
- Attachment: Choose an arrival artifact with probability  $\Pi_{A}(k) = (1-p_{r})\Pi_{copy} + p_{r}\Pi_{innovate}$ copying probability innovation probability
- Rewire: Only after these choices are made.



### Evidence for this model

- Registrations of pedigree dogs
- Baby name registrations
- Changes in top 100 of popular music charts
- Applied to archaeological pot shards



## each breed of pedigree dog

See Neiman (1995); Bentley, Maschner (2000,2001); Bentley, Hahn, Shennan (2004); Bentley, Shennan (2003,2005); Hahn, Bentley (2003); Herzog, Bentley, Hahn (2004); Bentley, Lipo, Herzog, Hahn (2007).

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### Relationship to Other Systems

- Unipartite Graph Rewiring [Watts & Strogatz 1998]
- Gene Frequencies [Kimura & Crow, 1964] Inheritence and Mutation (genes not memes)
- Speciation ['Tangled Nature' Christensen et al 2002]
- Family Names [Zanette & Manrubia, 2001]
  - Inheritence and New Immigrants
- Language Extinction [Stauffer et al. 2006]
- Minority Game variant

[Anghel et al, 2004]

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**Relationship to Statistical Physics Models** 

Some parameter values of other models are equivalent to our model:

- Urn Models [Bernoulli 1713, ..., Ohkubo et al. 2005]
- Zero Range Processes (Misanthrope version) [review M.R.Evans & Hanney 2005; Pulkkinen & Merikoski 2005]
- Voter Models [Liggett 1999, ..., Sood & Redner 2005]
- Backgammon/Balls-in-Boxes

applied to glasses [Ritort 1995], wealth distributions, simplicical gravity

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#### Urn models as networks



Mean Degree Distribution Master Equation

Usually one also uses a mean field *approximation* very accurate for many models (low vertex correlations)

$$n(k,t+1) - n(k,t) =+ n(k+1,t)\Pi_{R}(k+1)[1 - \Pi_{A}(k+1)]$$

$$(1-\Pi) \text{ terms}_{lnvariably}_{ignored} - n(k,t)\Pi_{A}(k)[1 - \Pi_{A}(k)] - n(k,t)\Pi_{R}(k)[1 - \Pi_{R}(k)] + n(k-1,t)\Pi_{A}(k-1)[1 - \Pi_{R}(k-1)]$$
Number of edges attaching to a vertex of degree (k-1)
$$Probability of NOT reattaching to same vertex$$

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Can the Mean Degree Distribution equation be

$$\left(\frac{n_i(k)k^{\beta}}{\sum_k n_i(k)k^{\beta}}\right) \neq \left\langle n_i(k)k^{\beta} \right\rangle \left(\frac{1}{\sum_k n_i(k)k^{\beta}}\right)$$
exact?

Normalisation of probabilities not usually same for different instances *i* 

EXACT  
only if 
$$\sum_{k} n_i(k)k^{\beta} = \left\langle \sum_{k} n_i(k)k^{\beta} \right\rangle \begin{cases} \beta = 0 \\ \text{or} \\ \beta = 1 \end{cases}$$

### Random or preferential attachment only

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**Exact Solution** 

Exploit linearity and break into eigenfunctions:-

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

 $\Rightarrow \text{ Find Hypergeometric equations and solutions:-} \\ \textbf{Eigenfunctions } G^{(m)}(z) = (1) T^{n} \mathcal{F}(a - \Phi b, b - An; c; 0) 2 3 \\ \textbf{Hypergeometric function} \end{aligned}$ 

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

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

**c**<sub>m</sub> are constants fixed by initial conditions

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### **Exact Equilibrium Solution**

$$n(k) = A \frac{\Gamma(k+\overline{K})}{\Gamma(k+1)} \frac{\Gamma(E-\overline{E}-\overline{K}-k)}{\Gamma(E+1-k)} \qquad \overline{K} = \frac{p_r}{p_p} \langle k \rangle$$
$$\overline{E} = \frac{p_r}{p_p} E$$

A is ratio of four  $\Gamma$  functions

- Simple ratios of  $\Gamma$  functions
- Similar to those found for growing networks but second fraction is only found for network rewiring with correct master equation
- Only approximate solutions known previously

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Large Degree Equilibrium Behaviour – Large p<sub>r</sub> Case

For  $p_r > p_* \sim 1/E$ 

(on average at least one edge attached to a randomly chosen artifact per generation)

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

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

Power below one but in data indistinguishable from one if <k> << 10

$$\zeta = -\ln(1-p_r)$$

**Exponential Cutoff** 

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Large Degree Equilibrium Behaviour – Small p<sub>r</sub> Case

For 
$$p_r < p_* \sim 1/E$$

(on average if all edges have been rewired once no edge is attached to a randomly chosen artifact per generation)

- Degree distribution rises near **k=E**
- $\Rightarrow$ In extreme case  $p_r=0$  all the edges are attached to ONE artifact

- a CONDENSATION or FIXATION

$$n(k) = A \left( \frac{\Gamma(k + \overline{K})}{\Gamma(k + 1)} \right) \left[ \frac{\Gamma(E - \overline{E} - \overline{K} - k)}{\Gamma(E + 1 - k)} \right]$$
  
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Blows

up

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#### Log-Log plot of typical **p(k)**



#### **Tangled Nature**

- Tangled Nature model of speciation with a particular focus on extinctions and intermittency in extinction patterns [Christensen et al, 2002; Laird & Jensen 2007]
- Each species (= artifact) identified by genes. Species die and reproduce (with inheritance/copying) and mutation (random attachment). Reproduction probability depends on interaction with other species.

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Network Version of Tangled Nature

 Simplified network model of Tangled Nature deletes whole nodes (species) and creates new ones with some inheritance = copying of old links



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#### Network Version of Tangled Nature (2)





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Minority Game - Leaders and Followers

- At each step each individual chooses one or zero

   the *minority* choice wins
- Choices are made based on one of a large but finite number of strategies using finite history

   each strategy is a different artifact
- Individuals may follow their own prediction or they may follow the prediction from the most successful nearest neighbour in an ER random graph of individuals
  - i.e. they *copy* the strategy of a neighbour
     [Anghel et al. PRL 92 (2004) 058701]

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Minority Game – Leaders and Followers

**Question:** 

How many individuals - *followers* – are using the same strategy, one which belongs to a particular individual – a *leader*?

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**Minority Game - Leaders and Followers** 

Plot *n(k)* the average of the number of strategies (of some leader) used by *k* individuals (followers). Various system sizes and various ER random graphs.



Minority Game Example - Leaders and Followers

### This Minority Game variant again shows how **copying** can arise naturally

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#### Equilibrium with a Network of Individuals

#### Qualitative behaviour largely unchanged except for 1d Lattice



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Exact Time Evolution Known

Exact solution for generating function known at all times and any finite parameter in terms of standard functions

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

constants c<sub>m</sub> fixed by initial conditions

Eigenvalues

Eigenfunctions are ratios of Gamma functions

$$\lambda_{m} = 1 - m(m-1) \frac{p_{p}}{E^{2}} - m \frac{p_{r}}{E}$$

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Homogeneity Measures  $F_n$ 

*n*-th derivatives of generating function gives measures of homogeneity

$$\mathbf{F}_{n}(t) \coloneqq \frac{\Gamma(E+1-n)}{\Gamma(E+1)} \frac{d^{n} \mathbf{G}(z,t)}{dz^{n}} \bigg|_{z=1}$$

- These are simple known ratios of  $\Gamma$  functions
- Related to *m*-th moments  $\mu_n$  via Stirling numbers

$$F_n(t) = N \frac{\Gamma(E+1+n)}{\Gamma(E+1)} \sum_{m=0}^n S_m^n \mu_m$$

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### **F**<sub>n</sub> Homogeneity Measures

$$F_n(t) = \sum_{k=0}^{E} \frac{k}{E} \frac{(k-1)}{(E-1)} \lfloor \frac{(k-n+1)}{(E-n+1)} n(k)$$

- These equal the probability of choosing n different individuals connected to the same artifact
- $\Rightarrow F_n = 0 \text{ if no artifact chosen more than once} \\ F_n = 1 \text{ if all individuals attached to same artifact}$

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Time Dependence and  $F_n$  Homogeneity Measures

 $F_n$  = probability that n different individuals have chosen the same artifact







E=N=100,  $p_r=0.01 \cong p_*$ , Points: average of 10<sup>5</sup> simulations Lines: exact mean field prediction Start: n(k)= $\delta_{k,1}$ 

*F<sub>n</sub>* increase as homogeneity increases with time

Time dependence of averages predicted very accurately, deviations less than 1% Phase transitions in real time

- Bipartite graph can be projected onto a unipartite graph of the artifact vertices
- Artifact degree distribution *p(k)* is the degree distribution for a random graph



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Graph Transition in Real Time

Infinite Random Graphs have a phase transition (e.g. appearance of GCC - Giant Connected Component) at Z(t)=1 where

$$z(t) = \frac{\langle k^2 \rangle}{\langle k \rangle} - 1 = (E - 1)F_2(t)$$

 $F_{2}(t) = F_{2}(\infty) + (\lambda_{2})^{t} (F_{2}) (\Phi F_{2}) (\Phi$ 

 $F_2$  is the probability that two randomly chosen stubs are attached to the same (artifact) vertex

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#### Phase Transition in Molloy-Reed projection



Phase Transition in Molloy-Reed projection

- For  $N=E=10^5$ ,  $p_r=0$ , initial  $F_2(0)=0$
- *z*(*t*)=1 at *t*=0.50000 (2) as predicted
- Transition at t/E = 0.535 (5)
- At transition *z*(*t*)=1.06 (1) not *z*(*t*)=1
- Average distance and diameter of GCC maximum at this point and second derivative of number of vertices in GCC zero at this point (within errors)
- ⇒ Finite size effects clearly present
- ⇒ Can follow a system through a phase transition in time exactly

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#### **Tangled Nature**

Tangled nature models can show periodic extinctions'

#### Simple copying model seems to show show less stability



### **Generalisations of Model**

- Different ways to update the model
  - Exact analytic results still possible
- Add a graph to the individual vertices
  - choose who to copy using individual's network
- Add a graph to the artifact vertices
  - mutations/innovations limited by metric in an artifact space
- Different types of individual
  - update their choice and copy/innovate at different rates

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#### Summary

- Preferential Attachment
  - = Making Random Walk on Network
  - = Copying Choice made by neighbour
- Applied to network rewiring can get exact solutions for any finite sized graph at any time
  - Related to many other situations where reached size of system is constant (at least over short time averages) and where there are *two* processes
    1) copying/inheritance/preferential attachment
    2) innovation/mutation/random attachment

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