Milgram's experiment in various networks

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1 Introduction

- sociology (the strength of weak ties, terrorism)
- computer science (Internet, WWW)

1.1 Milgram's experiment

- A group of individuals was asked to send a letter
- to a target person in Boston via an acquaintance
- who was supposed to be closer to the target,
- than the sender.
- The mean length of the letter chain was less than seven.

1.2 Small world

- $d \equiv$ the mean length of the shortest path between two vertices
- The growth of *d* slower than any positive power of *N* is called a small-world effect.
- By definition, a networks is a small world if it shows the small-world effect.

1.3 Dictionary

- **a graph** \equiv a set of nodes and a set of edges
- a simple graph \equiv a graph without loops and without multiple edges
- a forest \equiv a graph without cycles
- a tree \equiv a compact forest
- a classical random graph (CRG) ≡ Erdös–Rényi and/or Gilbert model

By **growing** we mean adding subsequent nodes to an already existing graph with *m* links:

- $m = 1 \rightarrow \text{trees}$
- $m = 2 \rightarrow \text{simple graphs}$

P(q) = probability that a new node will be attached to existing node q

- for **exponential** networks P(q) is uniform
- for scale-free (preferential, Albert–Barabási)
 networks P(q) is proportional to the node
 degree (i.e. its number of edges)

- a network containing N nodes is fully characterized by its connectivity matrix C: c_N(i, j) = 1 if the nodes i, j are linked together, and c_N(i, j) = 0 elsewhere
- in a distance matrix S, the matrix element $s_N(i, j)$ is the number of links along the shortest path from i^{th} to j^{th}

1.4 Search strategy

- In this kind of contact experiments, to find an appropriate next person in the path is a nontrivial task, and several strategies are possible.
- One of the most obvious is to find a person most connected, i.e. a neighbouring node with the highest degree (MCNS).

- This strategy has been shown to be effective in networks with power-law degree distribution, but not in random graphs.
- Here we ask, how the MCNS is effective in the exponential networks.

2 Numerical approach

For growing networks the starting point is a matrix **S** for the tree of linked together two nodes:

$$\mathbf{S} = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix}$$

Selecting a node to link a new node is equivalent to select a number q of column/row of the matrix.



 $\forall 1 \leq i \leq N : s_{N+1}(N+1,i) =$ (1a) $s_{N+1}(i, N+1) = s_N(q, i) + 1$

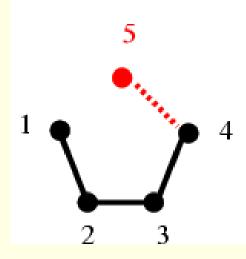
and

$$s_{N+1}(N+1,N+1) = 0$$
 (1b)

Distance matrix evolution for trees:

| 0 | 1 | 2 | 3 |
|---|---|---|---|
| 1 | 0 | 1 | 2 |
| 2 | 1 | 0 | 1 |
| 3 | 2 | 1 | 0 |

| 0 | 1 | 2 | 3 | 4 |
|---|---|---|---|---|
| 1 | 0 | 1 | 2 | 3 |
| 2 | 1 | 0 | 1 | 2 |
| 3 | 2 | 1 | 0 | 1 |
| 4 | 3 | 2 | 1 | 0 |



2.2 Simple graphs

- In simple graphs cyclic paths are possible and the distance matrix **S** is to be rebuilt when adding each node. The $(N+1)^{\text{th}}$ node is added to existing nodes *p* and $q \neq p$:
 - $\forall 1 \leq i, j \leq N : s_{N+1}(i,j)$
 - $=\min\left(s_N(i,j),s_N(i,p)+2+s_N(q,j)\right),$

(2a)

 $s_N(i,q) + 2 + s_N(p,j)$.

For new,
$$(N+1)^{\text{th}}$$
, column/row
 $\forall 1 \le i \le N : s_{N+1}(N+1,i) = s_{N+1}(i,N+1)$
 $= \min(s_N(p,i),s_N(q,i)) + 1$
(2b)

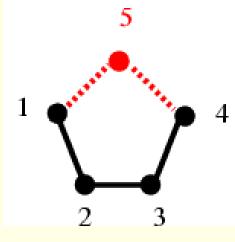
and again for the diagonal element

$$s_{N+1}(N+1,N+1) = 0$$
 (2c)

Distance matrix evolution for simple graphs:

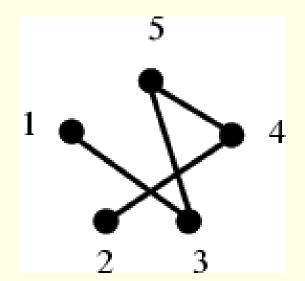
| 0 | 1 | 2 | 3 |
|---|---|---|---|
| 1 | 0 | 1 | 2 |
| 2 | 1 | 0 | 1 |
| 3 | 2 | 1 | 0 |

| 0 | 1 | 2 | 2 | 1 |
|---|---|---|---|---|
| 1 | 0 | 1 | 2 | 2 |
| 2 | 1 | 0 | 1 | 2 |
| 2 | 2 | 1 | 0 | 1 |
| 1 | 2 | 2 | 1 | 0 |



2.3 CRG

$$N = 5, L = 4, p = \frac{2L}{N(N-1)} = 0.4$$



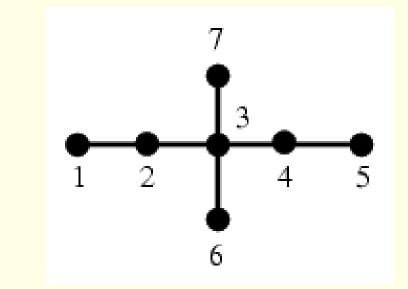
- We start the simulation with an $N \times N$ matrix with all non-diagonal elements equal to N.
- We go through all non-diagonal elements of
 S and set s(i, j < i) equal to one with the
 probability p.
- The matrix **S** is kept symmetric.
- Each time, when a new edge is added, we have to rebuild the whole matrix S due to link

between nodes i^{th} and j^{th} : $\forall 1 \le m, n \le N : s(m, n) = \min(s(m, n),$

$$s(m,i) + 1 + s(j,n), s(m,j) + 1 + s(i,n)).$$
(3)

 After such a procedure the matrix S_{N×N} contains elements equal to N only if graph is not connected.

2.4 The Kertész list



| 1 2 2 | 3 3 | 3 3 4 | 4 5 | 6 7 |
|-------|-----|-------|-----|-----|
|-------|-----|-------|-----|-----|

2.5 Network characteristics

- In a connected graph, nodes can be characterised locally (with their degree k) or globally (e.g. with their average length path ξ to other nodes).
- Here we investigate how ξ depends on k.

• k_i is # of '1' in i^{th} row/column,

•
$$\xi_i \equiv [N^{-1} \sum_{j=1}^N s_N(i,j)],$$

•
$$d = [N^{-2} \sum_{i=1}^{N} \sum_{j=1}^{N} s_N(i,j)] = [N^{-1} \sum_{i=1}^{N} \xi_i],$$

• $\xi(k) \equiv$ an average of ξ_i for nodes with given degree $k = k_i$,

where $[\cdots]$ is an average over N_{run} different matrices.

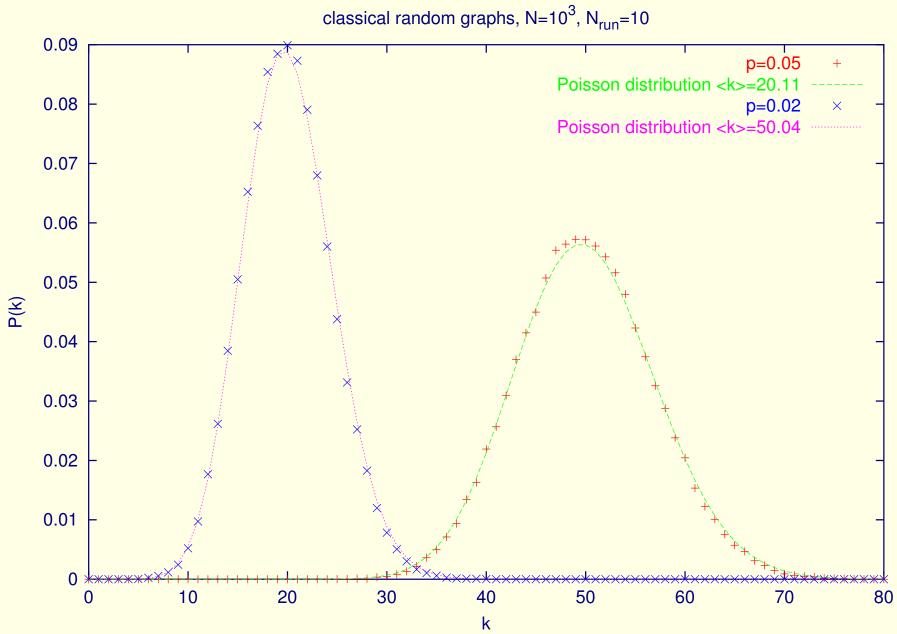
3 Results

- Each network contains at least a thousand of nodes ($N = 10^3$).
- The results are typically averaged over $N_{\rm run} = 10^7$, 10^3 and 100 different networks for trees, simple graphs and CRG, respectively.

3.1 Degree distribution

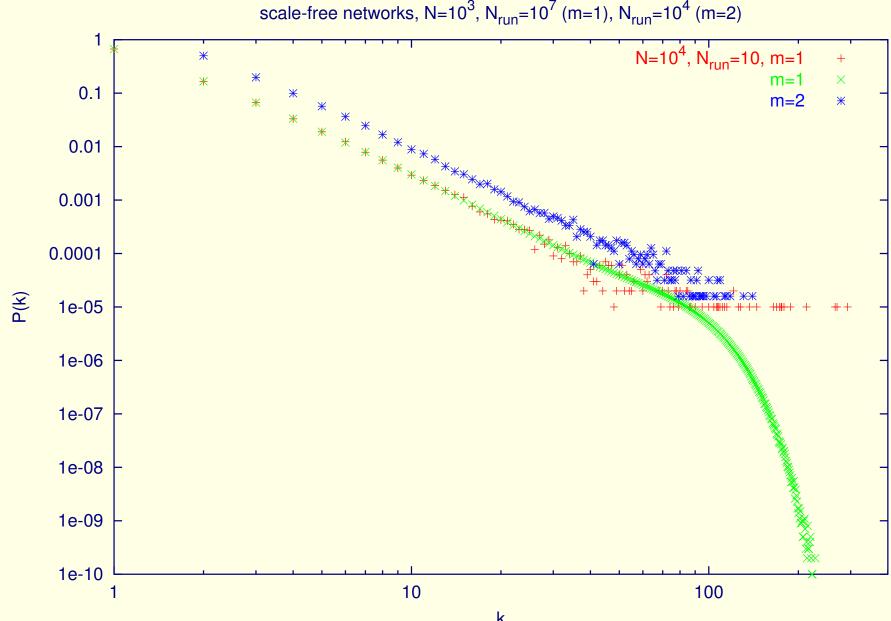
3.1.1 CRG

The degree distribution for CRG follows the Poisson distribution $P(k) = \exp(-\langle k \rangle) \cdot \langle k \rangle^k / k!$, with $\langle k \rangle \approx 20$ and $\langle k \rangle \approx 50$ for p = 0.02 and p = 0.05, respectively.



3.1.2 Scale-free natworks

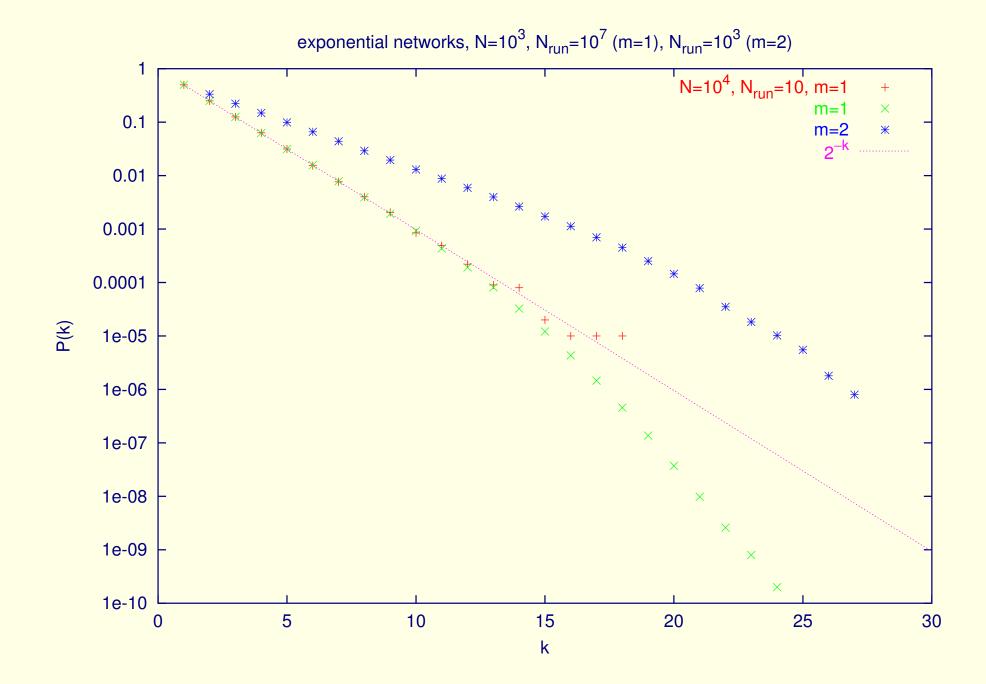
For the scale-free networks we reproduce $P(k) \propto k^{-\gamma}$ with $\gamma \approx 2.7$, while the theoretical value is 3.0. The numerical reduction of γ is known to be caused by the finite-size effect.



scale-free networks, N=10³, N_{run}=10⁷ (m=1), N_{run}=10⁴ (m=2)

3.1.3 Exponential networks

For the exponential trees the node degree distribution is verified to be $P(k) = 2^{-k}$.



3.2 Small-world effect

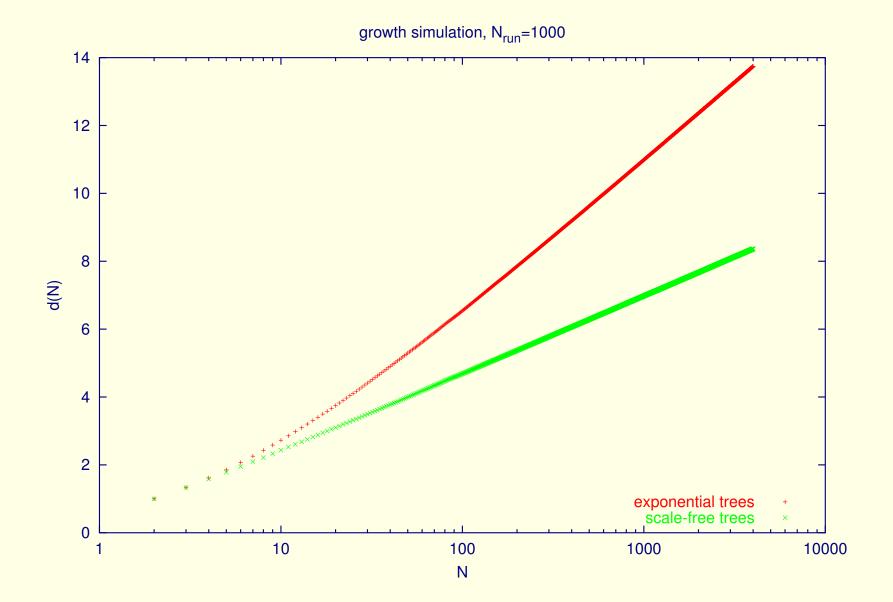


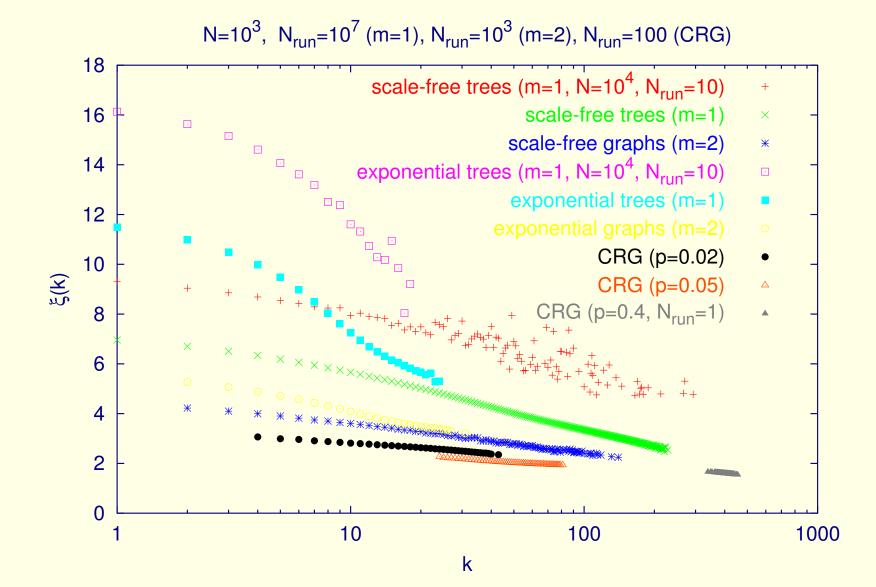
Table 1: The mean distance $d(N) = a \ln N + b$ for different evolving exponential networks.

| т | 1 | 2 |
|---|-------|------|
| a | 2.00 | 0.71 |
| b | -2.84 | 0.16 |

Table 2: The mean distance $d(N) = a \ln N + b$ for different evolving scale-free networks.

| т | 1 | 2 |
|---|-------|------|
| a | 1.00 | 0.48 |
| b | -0.08 | 0.83 |

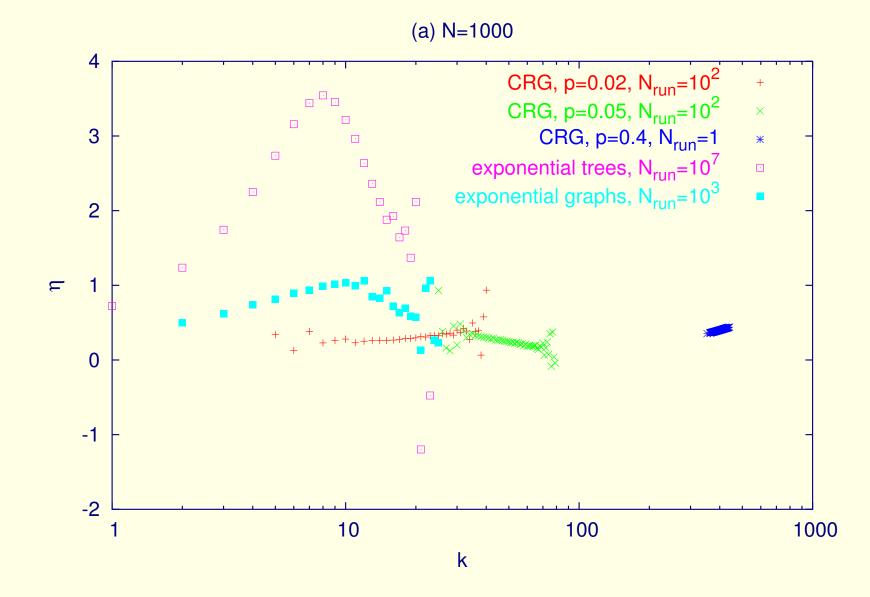
3.3 Effectiveness of MCNS

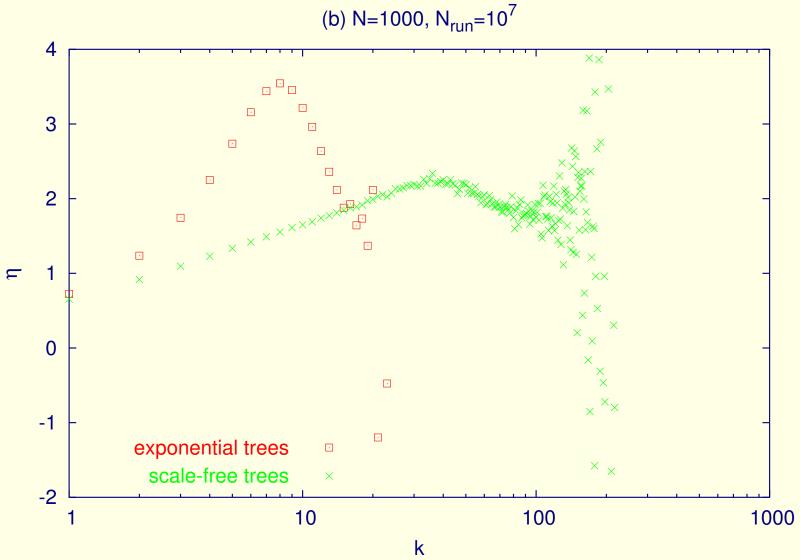


The slope of the curve $\xi(k)$ brings an information, how this strategy is effective for a given network. The effectiveness of MCNS for nodes of given *k* can be evaluated by an index

$$\eta = -\frac{\partial \xi}{\partial \ln k}.$$

4





4 Conclusioins

- For the random graphs the mean distance ξ practically does not depend on k, and the index η is close to zero.
- MCNS works better for scale-free networks than for CRG.
- The new result is that MCNS applied in an exponential tree is even more effective, than in the scale-free tree.

- In the scale-free networks, local fluctuations of the degree are enhanced by subsequent linkings.
- The multiple centres of high degree can be created, and the growing concentrates on these centres.
- Then, MCNS can be misleading, as it leads always to a local centre; however, sometimes the target is somewhere else.

- This enhancement is absent in the exponential networks, and that is why MCNS works better there. We note that this argumentation works well for trees.
- For other systems, there is more than one path between each pair of nodes, and any educated but general strategy cannot replace the knowledge where the target is.

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