

### Complex Network Systems

Small worlds

Ilche Georgievski

2019/2020 Winter

#### **SMALL-WORLD PHENOMENON**

# Milgram's experiment



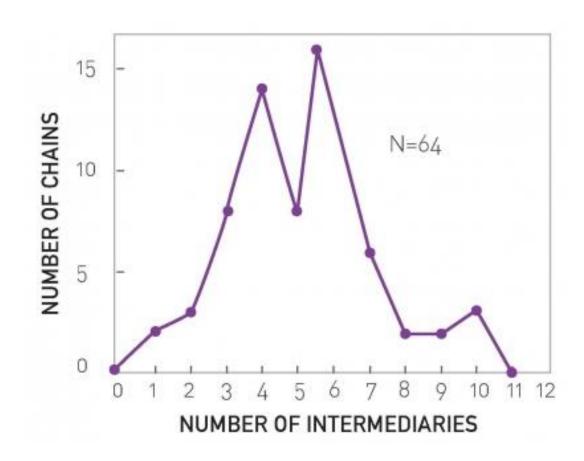
Stanley, Milgram, The Small-World Problem, Psychology Today, May 1967, 1(1): 61-67.

#### **Instructions:**

MA

Given a target individual (stockbroker in Boston), pass the message to a person you correspond with who is "closest" to the target.

### Milgram's experiment

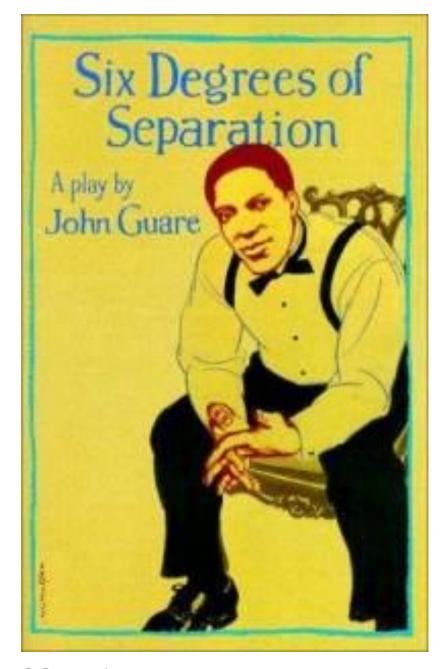


#### **Outcome:**

20% of initiated chains reached target

average chain length = 6.5 mean chain length = 5.2

Stanley, Milgram, The Small-World Problem, Psychology Today, May 1967, 1(1): 61-67.



"Everybody on this planet is separated by only six other people. Six degrees of separation. Between us and everybody else on this planet. The president of the United States. A gondolier in Venice.... It's not just the big names. It's anyone. A native in a rain forest. A Tierra del Fuegan. An Eskimo. I am bound to everyone on this planet by a trail of six people. It's a profound thought. How every person is a new door, opening up into other worlds."

### Interpreting Milgram's experiment

Is 6 is a *surprising* number? In the 1960s? Today? Why?





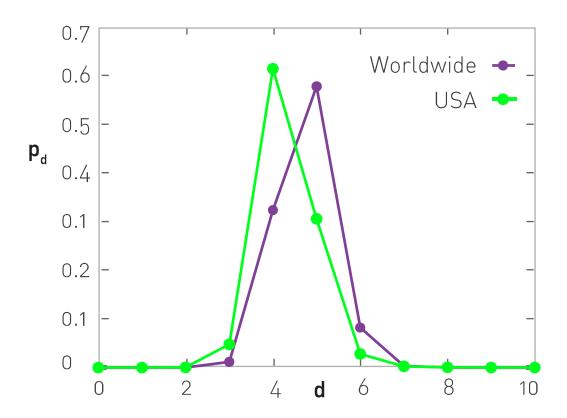
18 targets13 different countries

60 000+ participants24 163 message chains384 reached their targets

average path length = 4.0

Dodds, Peter Sheridan, Muhamad, Roby, Watts, Duncan J., An Experimental Study of Search in Global Social Networks, *Science*, 08 Aug 2003 : 827-829.

## Facebook experiment



721 million active users
68 billion symmetric friendship links

average path length = 4.74

L. Backstrom, P. Boldi, M. Rosa, J. Ugander, and S. Vigna. Four degrees of separation. In ACM Web Science 2012: Conference Proceedings, pages 45–54. ACM Press, 2012.

# Interpreting Milgram's experiment

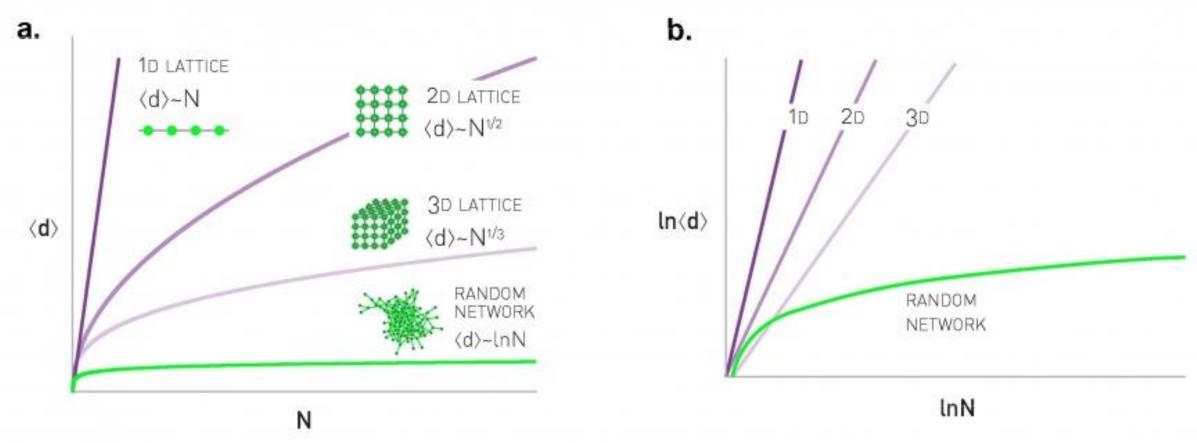
- If social networks were random...?
  - ~1000 acquaintances/person
  - ~1000 choices 1st link
  - $\sim 1000^2 = 1,000,000$  potential 2nd degree neighbors
  - $\sim 1000^3 = 1,000,000,000$  potential 3rd degree neighbors

de Sola Pool, Ithiel and Kochen, Malfred, Contacts and Influence, *Social Networks*, I (1978/79): 5-51.

### Question

- If the network were completely cliquish, that is all your friends of friends were also directly your friends, what would be true:
  - a) None of your friendship edges would be part of a triangle (closed triad)
  - Vb) It would be impossible to reach any node outside the clique by following directed edges
    - c) Your shortest path to your friends' friends would be 2

# <d> is diameter Why are small worlds surprising?



In regular lattices path lengths are significantly longer than in a random network

Average path length or diameter depends logarithmically on the system size

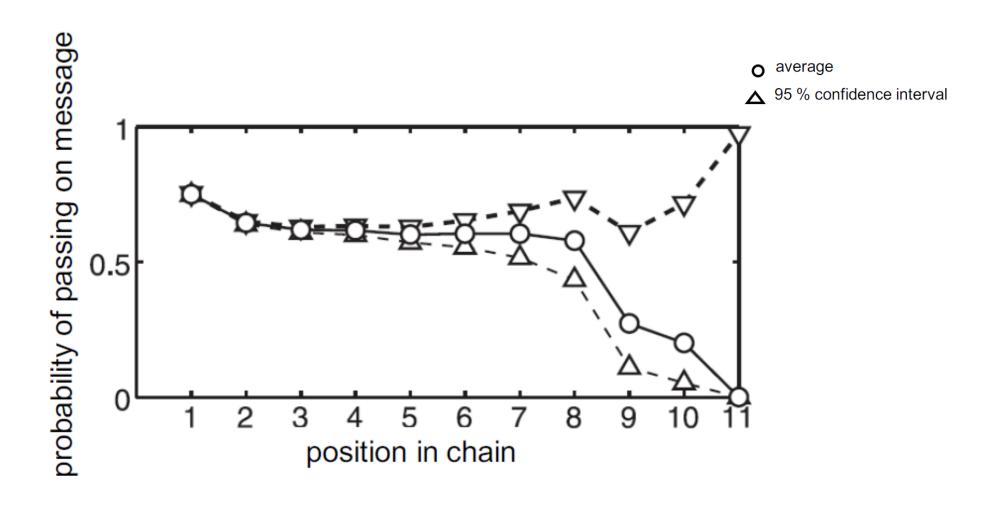
### Accuracy of distances

• Is 6 an accurate number?

no, there are survival bias

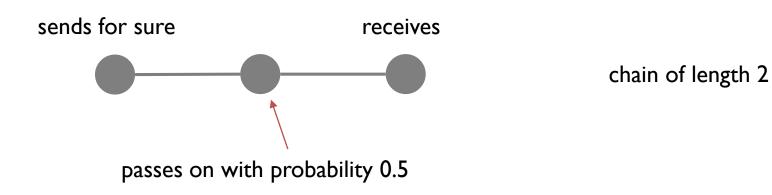
- What bias is introduced by uncompleted chains?
  - are longer or shorter chains more likely to be completed?

### Attrition



### Question

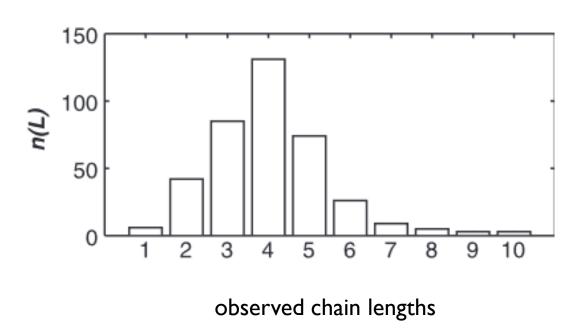
- If each person in the chain has 0.5 probability of passing the message on, what is the likelihood of a chain of length 2 being completed?

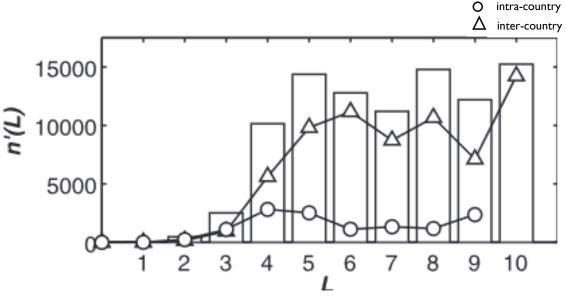


### Question

- If each person in the chain has 0.5 probability of passing the message on, what is the likelihood of a chain of length 5 being completed?
  - a) 1/2
  - b) 1/4
  - c) 1/8
  - d) I/I6

### Estimating the true distance





'recovered' histogram of path lengths

### Accuracy of distances

• Is 6 an accurate number?

- Do people find the shortest paths?
  - accuracy of small-world chains in social networks by Killworth et al.: less than optimal choice for the next link in chain is made  $\frac{1}{2}$  of the time

Watts-Strogatz model

#### **SMALL-WORLD MODEL**

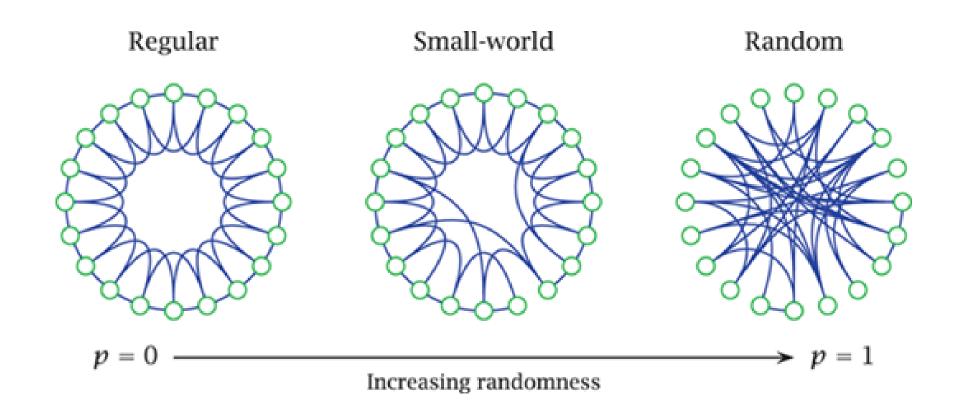
#### **Models**

- Watts and Strogatz
  - Clustering and short paths

- Kleinberg
  - Geographical structure

- Watts, Dodds and Newman
  - Hierarchical structure

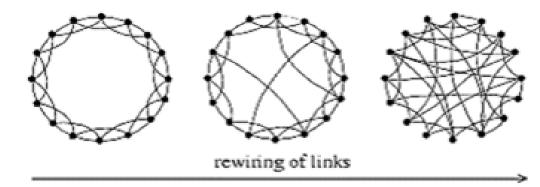
### Watts-Strogatz model



Olofsen, Erik and Dahan, Albert, Big Brain, Small World?, Anesthesiology, 2015, 122(1):8-11.

### Generating a small-world graph

- The lattice has a high, but fixed, clustering coefficient
- The lattice has a high average path length
- The model begins with a regular lattice and starts adding random edges (through rewiring)
  - **Rewiring**: take an edge, change one of its end-points randomly



### Constructing small-world networks

#### Algorithm 4.1 Small-World Generation Algorithm

**Require:** Number of nodes |V|, mean degree k, parameter p

- 1: **return** A small-world graph G(V, E)
- 2: G = A regular ring lattice with |V| nodes and degree k
- 3: **for** node  $v_i$  (starting from  $v_1$ ), and all edges  $e(v_i, v_j)$ , i < j **do**
- 4:  $v_k$  = Select a node from V uniformly at random.
- 5: **if** rewiring  $e(v_i, v_j)$  to  $e(v_i, v_k)$  does not create loops in the graph or multiple edges between  $v_i$  and  $v_k$  **then**
- 6: rewire  $e(v_i, v_j)$  with probability  $p: E = E \{e(v_i, v_j)\}, E = E \cup \{e(v_i, v_k)\};$
- 7: **end if**
- 8: end for
- 9: Return G(V, E)

As in many network generation algorithms:

- Disallow loops
- Disallow multiple edges

#### **SMALL-WORLD MODEL PROPERTIES**

### Degree distribution

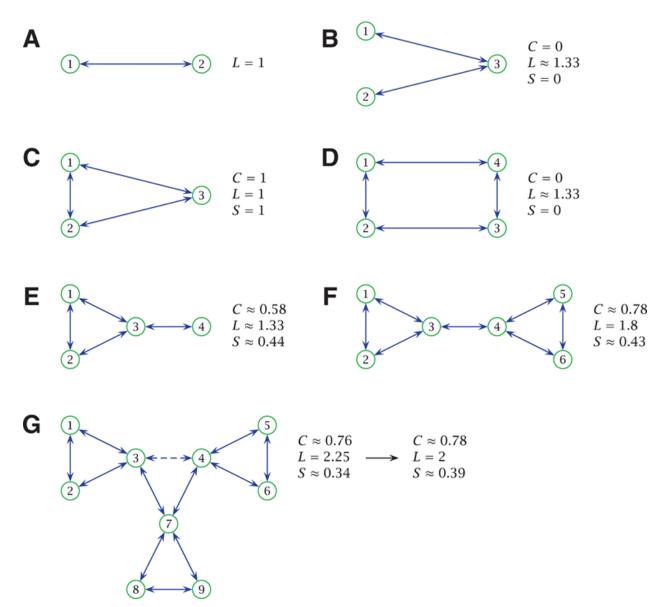
 What can we say about distribution of degrees in small-world graphs?

• In the graph generated by the small-world model, most nodes have similar degrees due to the underlying lattice

### Regular lattice vs. random graph

#### Regular lattice

- Clustering coefficient: high
- Average path length: high
- Random graph
  - Clustering coefficient: low
  - Average path length: ok



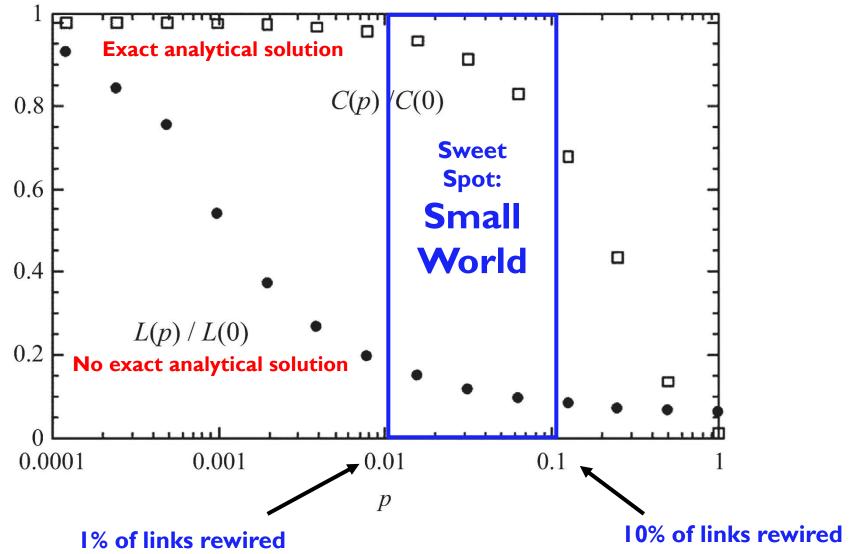
Olofsen, Erik and Dahan, Albert, Big Brain, Small World?, Anesthesiology, 2015, 122(1):8-11.

### What happens in between?

- Small average path length means small clustering?
- Large average path length means large clustering?

- Numerical simulation
  - -As we increase p from 0 to 1, we observe
    - Fast decrease of average distance
    - Slow decrease in clustering coefficient

### Change in clustering coefficient/Avg. path length

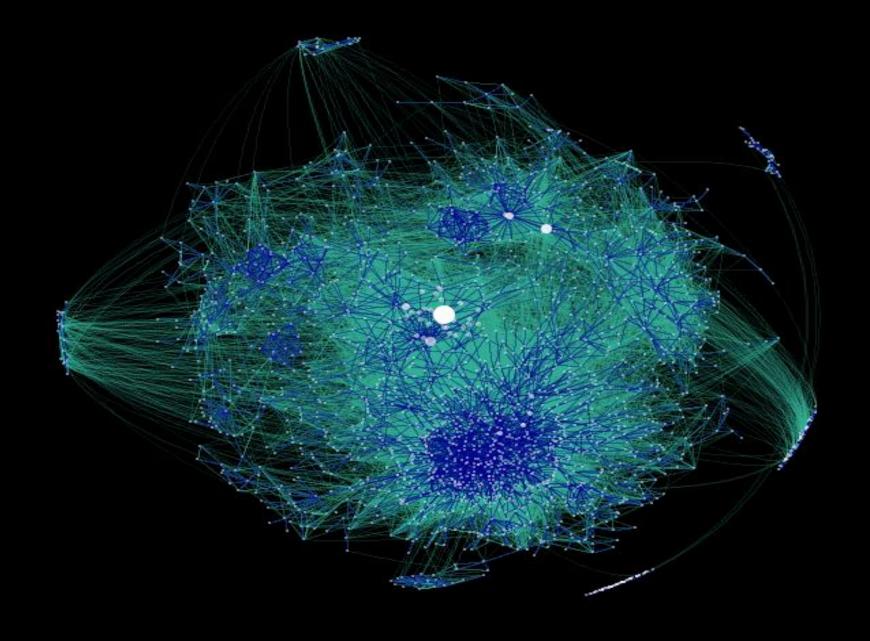


# Modelling with the small-world model

- Given a real-world network in which average degree k and clustering coefficient  $\mathcal C$  are given
  - we set C(p) = C
  - we calculate p using

$$C(p) \approx (1-p)^3 C(0)$$

• Given p, k, and n = |V|, we can simulate the small-world model



Network	N	L	<b>(k)</b>	⟨d⟩	d <sub>max</sub>	lnN/ln∢k>
Internet	192,244	609,066	6.34	6.98	26	6.58
WWW	325,729	1,497,134	4.60	11.27	93	8.31
Power Grid	4,941	6,594	2.67	18.99	46	8.66
Mobile-Phone Calls	36,595	91,826	2.51	11.72	39	11.42
Email	57,194	103,731	1.81	5.88	18	18.4
Science Collaboration	23,133	93,437	8.08	5.35	15	4.81
Actor Network	702,388	29,397,908	83.71	3.91	14	3.04
Citation Network	449,673	4,707,958	10.43	11.21	42	5.55
E. Coli Metabolism	1,039	5,802	5.58	2.98	8	4.04
Protein Interactions	2,018	2,930	2.90	5.61	14	7.14

#### Sources

- Adamic, L. Social and Information Network Analysis, CS224W, Stanford University (2015), http://snap.stanford.edu/class/cs224w-2015/
- Zafarani, R., Abbasi, M.A. and Liu, H. Social Media Mining: An Introduction, Cambridge University Press, 2014.
- Barabási, A. Network Science, http://networksciencebook.com