

Small World

Networks, Dynamics and the Small World Phenomenon Duncan J. Watts

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for S2_4 Mining the Social Web Mining for web2.0 PhD Course at Aalborg
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Small World Phenomenon – Informal Definition

- A and B (complete strangers) for the first time.
- After some conversation, they find out that A and B have a mutual acquaintance.
- Famous: six degrees of separation.

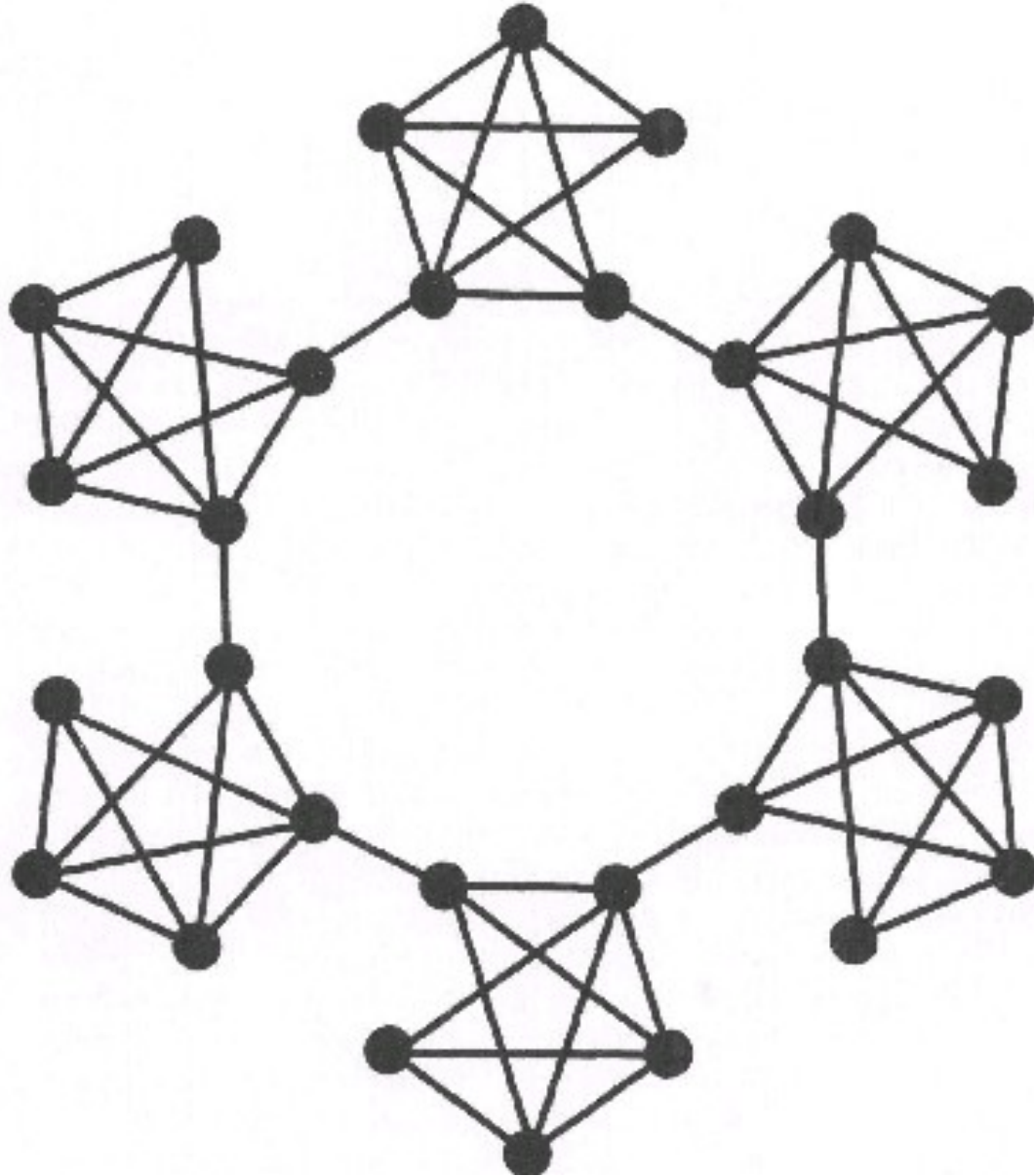
Properties

- The network is:
 - Numerically large, $n \gg 1$. (real world: n billions)
 - Sparse, each person knows k people (hundred thousand times smaller than n)
 - Decentralised, no dominant central vertex.
 $k_{\max} \ll n$
 - Highly clustered

Formal Definition

- Population n is fixed.
- Average degree k of vertices are
 - sparse: $k_{\max} \ll n$,
 - sufficiently dense: $k_{\max} \gg 1$.
- Connected: Any vertex can be reached from any vertex.

Most Clustered Sparse Graph - Caveman



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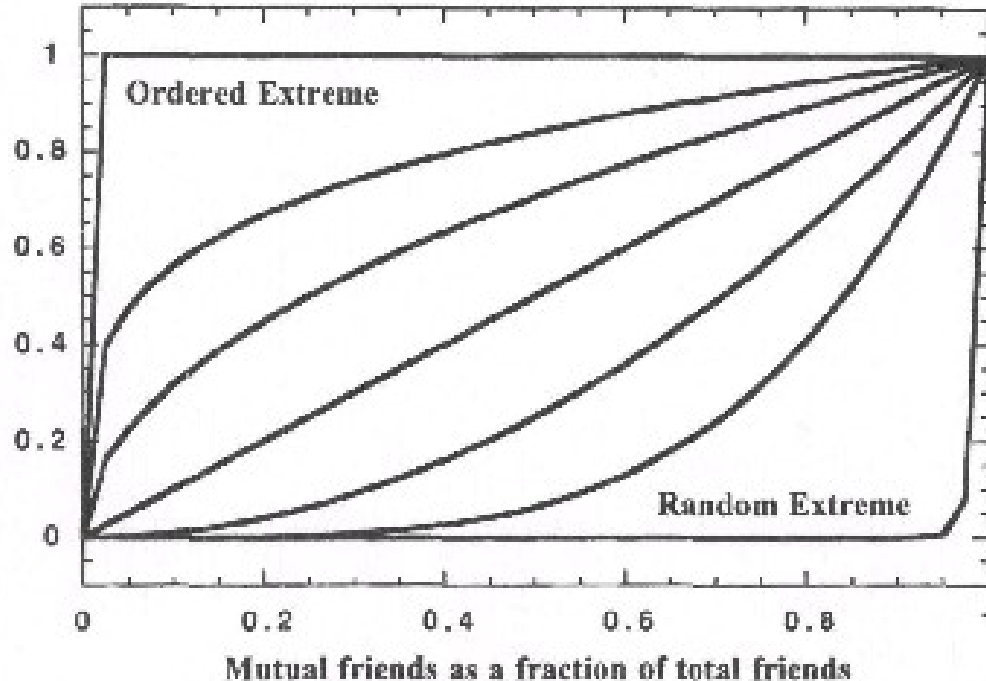
- People become friends only by acquaintance
- Clustering Coefficient $C = \frac{1-6}{(k^2-1)}$
 - Goes towards 1 as k increases.
- Average Path length $L = \frac{n}{2(k+1)}$
 - When $n \gg k$, L is large.
- Highly cluster / locally ordered

Random Graph

- People become friends by complete random
- Clustering Coefficient $C = k/n$
- Average Path Length $L = \ln(n)/\ln(k)$
- $L_{\text{random}} \ll L_{\text{caveman}}$
- Note:
 - Sparse Condition implies $k \ll n$.
 - Graphs with long characteristics, short path length
- clustering vanishes for large n

Real World

- In between Complete Ordered (Caveman) and Random
- Potential Graphs – alpha model



alpha model

- Abstract model
- Each alpha, enourmos finite number of graphs.
 - Small alpha => isolated graphs, L = infinite
- To deal with this
 - Minimal structure
 - No special vertices (no star, chain etc.)
 - Minimally connected
 - No more edges than necessary

Definitions – Small world

- large- n , sparse connected, decentralised ($n \gg k_{\max} \gg 1$). $L = L_{\text{random}}$. $C \gg C_{\text{random}}$
- Range, $r =$ second shortest path.
- Shortcut, edge $r > 2$
- $\Phi =$ Fraction of all edges that are shortcuts.
- Contraction = Pair of vertices that share one and only one common neighbor.
- **Adding contractions e.g. caveman makes a graph closer to small world.**

Evidence of small worlds

- Milgrams experiments suggest it
 - The letter experiments explained yesterday
 - We cannot always find the shortest chain
- Mapping the exact networks is impossible
 - There might not be an exact network
 - Global information not available locally
- The graph does not have to be a social network

Evidence of small worlds

- Movie actor collaboration
 - Vertex: Movie actor, Edge: Two actors appearing in same movie
 - $n = 226000$, $k=61$
- Power Grid
 - Well connected grid -> efficiency and robustness
 - $n = 4941$, $k=2.94$
- Neural network of *C. elegans*
 - Roundworm :-)
 - $n = 282$, $k = 14$

Results from studies

CHARACTERISTIC PATH LENGTH (L) AND CLUSTERING COEFFICIENT (C) FOR THREE REAL NETWORKS

	L_{Actual}	L_{Random}	C_{Actual}	C_{Random}
Movie actors	3.65	2.99	.79	.00027
Power grid	18.7	12.4	.080	.005
<i>C. elegans</i>	2.65	2.25	.28	.05

- Length close to average for random
- Clustering significantly higher
- The three networks exhibits the same characteristics
 - though order of magnitude difference in size
- Small worlds are very real

Dynamical systems on small worlds

- Small local changes in the graph can have dramatic impact on global properties
- Can this happen to dynamical systems connected in a small-world manner?
 - Or: Does modelling dynamical systems as small-worlds reveal anything new
- Dynamical systems are a bit different

Disease spreading 1/2

- Modelling disease spreading as a graph
- Previous work has modelled population as a uniform mix and with simple structured patterns
 - Structure heavily impacts disease and parasites
 - If the population structure is small-world like, it could be used for better predictions
- Set up a model for this
 - p = infectious degree
 - $p < 0.11$: disease die quickly

Disease spreading 2/2

- Sat up a model for this (cont.)
 - $0.11 \leq p \leq 0.5$, depends on structure
 - $p > 0.5$, disease take over population
- Number of infected depends on topology as well as infections degree
 - Not exactly news
 - New: The required changes to the structure are subtle
- Small world spread disease faster
 - almost as fast as random, significantly faster than caveman graph

Small worlds

- Class of graphs which:
 - $n \gg 1$
 - sparsely connected
 - high clustering
 - low pair length
 - structure is somewhere between random and ordered
- model came from social network, but found several other places
 - especially in biology and technology