

CMSC5741 Big Data Tech. & Apps.

Lecture 10: Mining Social- Assignment Project Exam Help Network Graphs

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The Chinese University of Hong Kong

Prediction of 2016 US Election

POLITICS

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**2016 ELECTION PREDICTIONS:
HILLARY CLINTON TOPS
DONLAD TRUMP IN
ELECTORAL COLLEGE,
FORECASTERS AGREE**

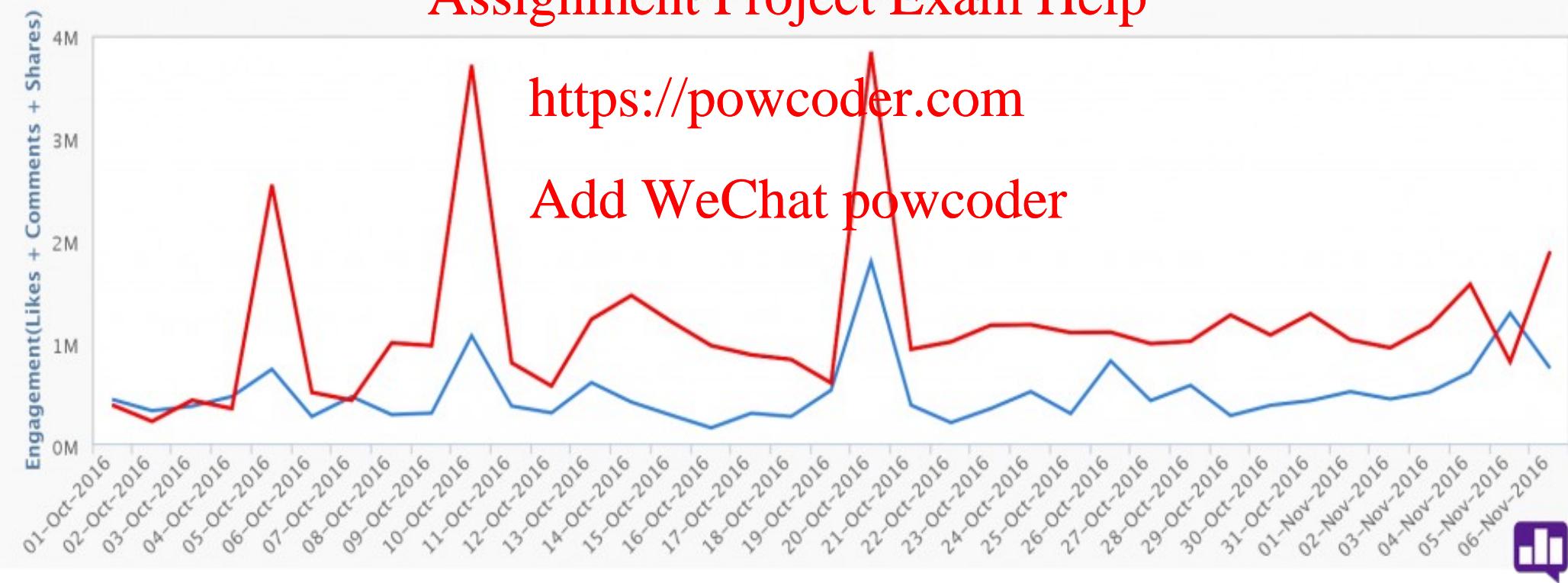
Social Network Prediction of 2016 US Election

Engagement Over Time
A comparison of Post Engagement over time for all accounts

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Outline

- Introduction
- Social Networks as Graph
 - Basics of Graph Theory
 - Important properties of social networks as graphs
 - Analysis of a real-world social network
- Community detection algorithms
 - Girvan-Newman algorithm
 - Spectral clustering
- Conclusions

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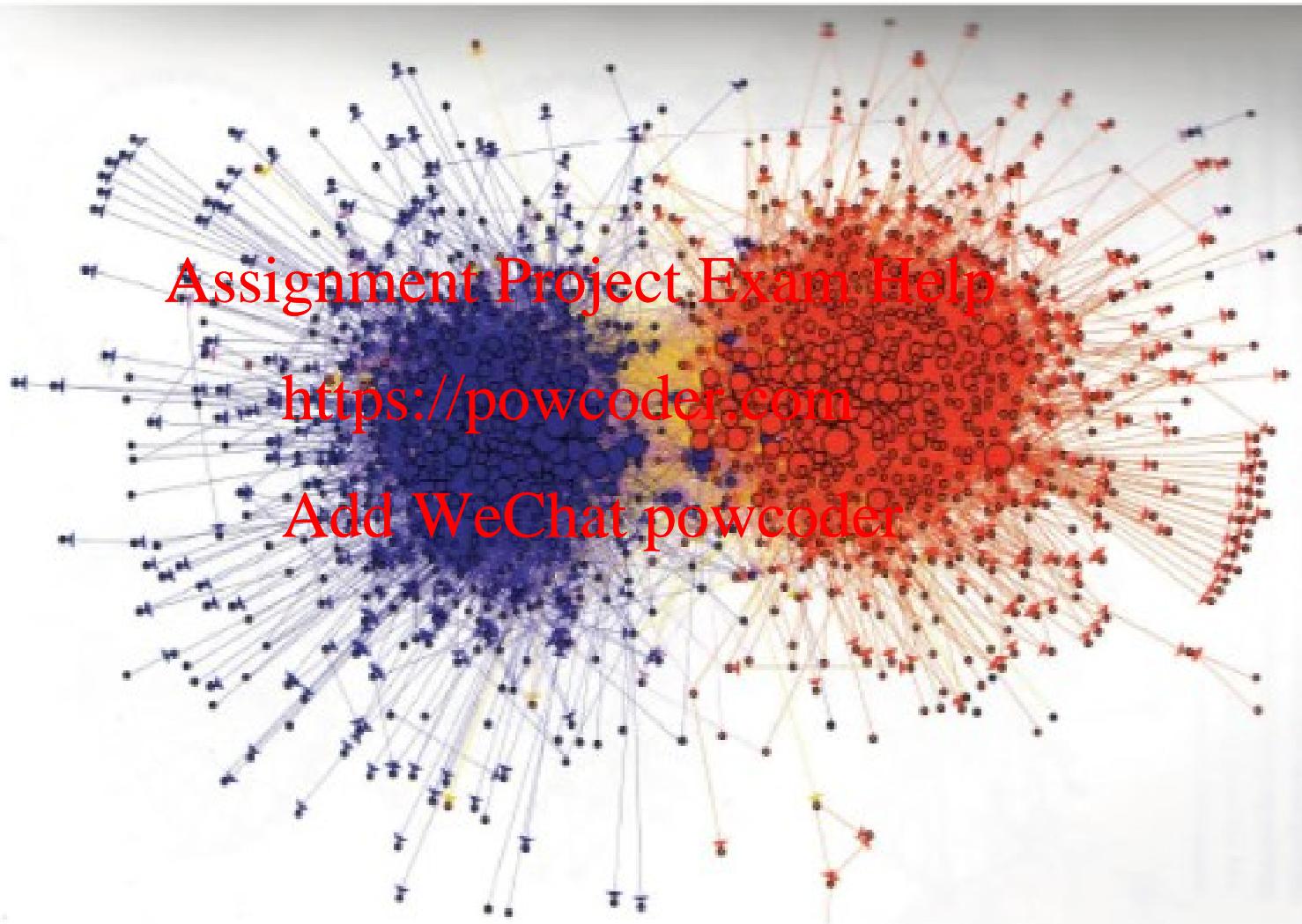
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Networks: Communication



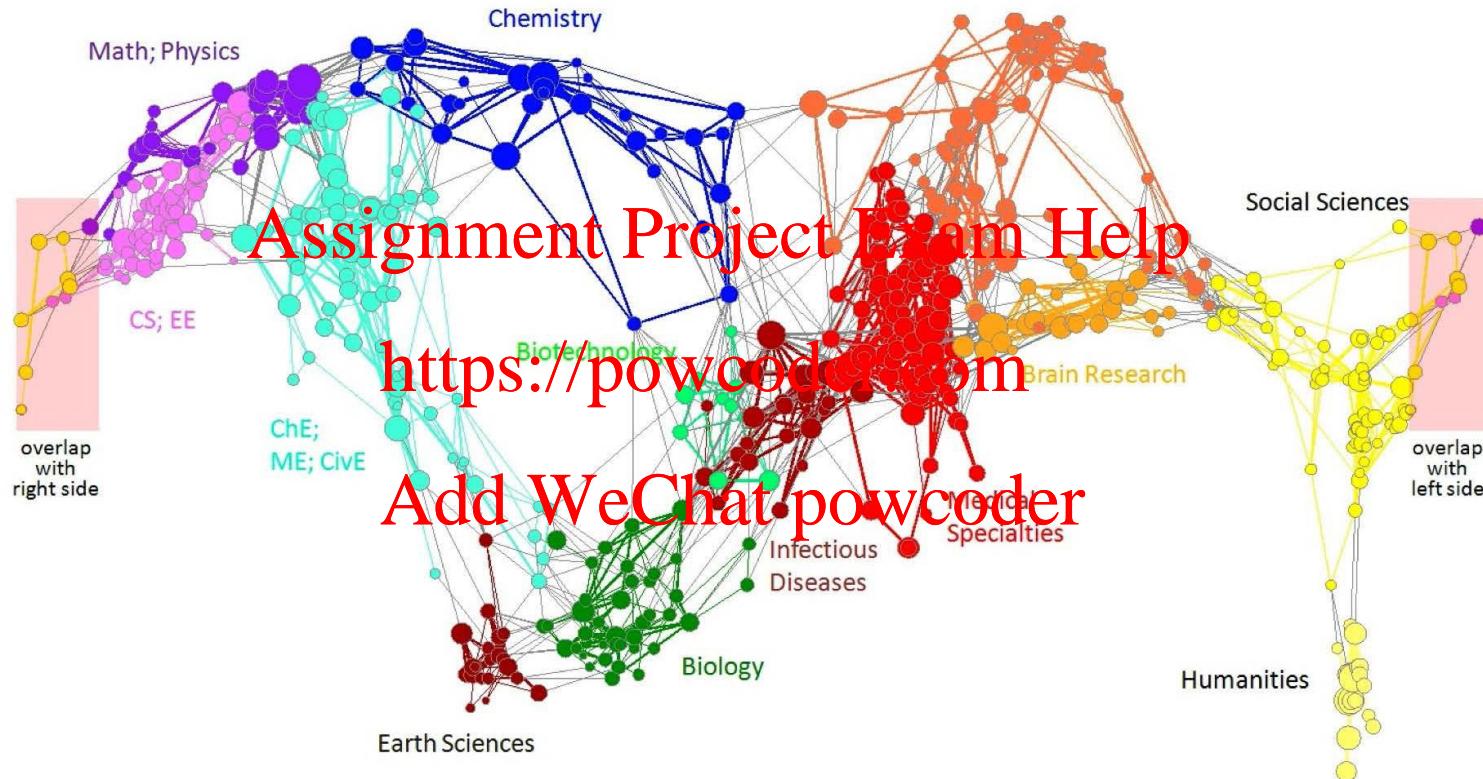
Graph of the Internet (Autonomous Systems)
Power-law degrees [Faloutsos-Faloutsos-Faloutsos, 1999]
Robustness [Doyle-Willinger, 2005]

Networks: Media



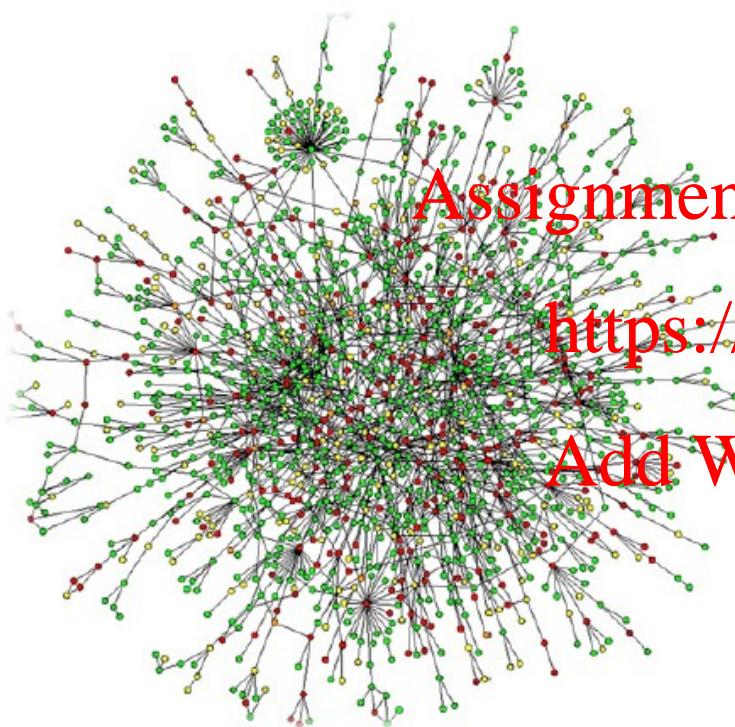
Connections between political blogs
Polarization of the network [Adamic-Glance, 2005]

Network: Science



Citation networks and Maps of science
[Börner et al., 2012]

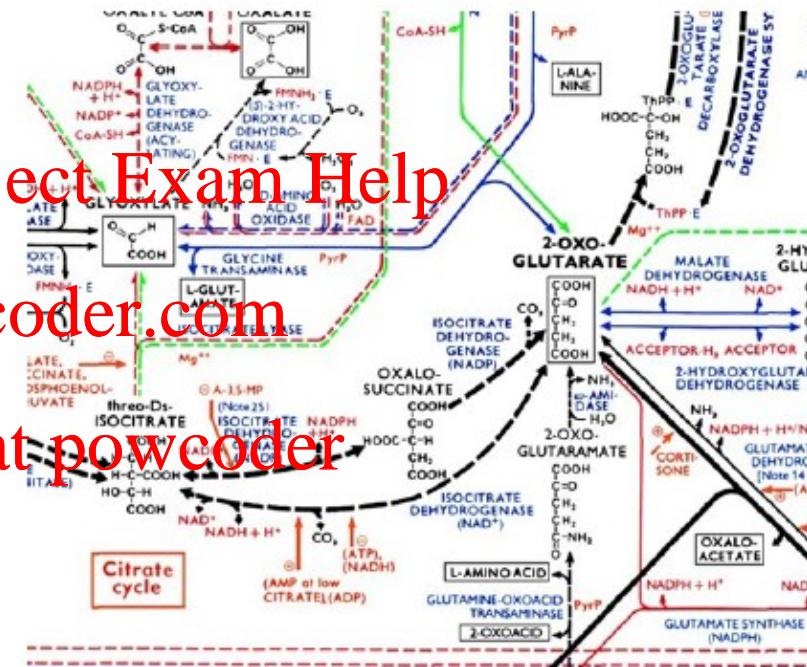
Networks: Biology



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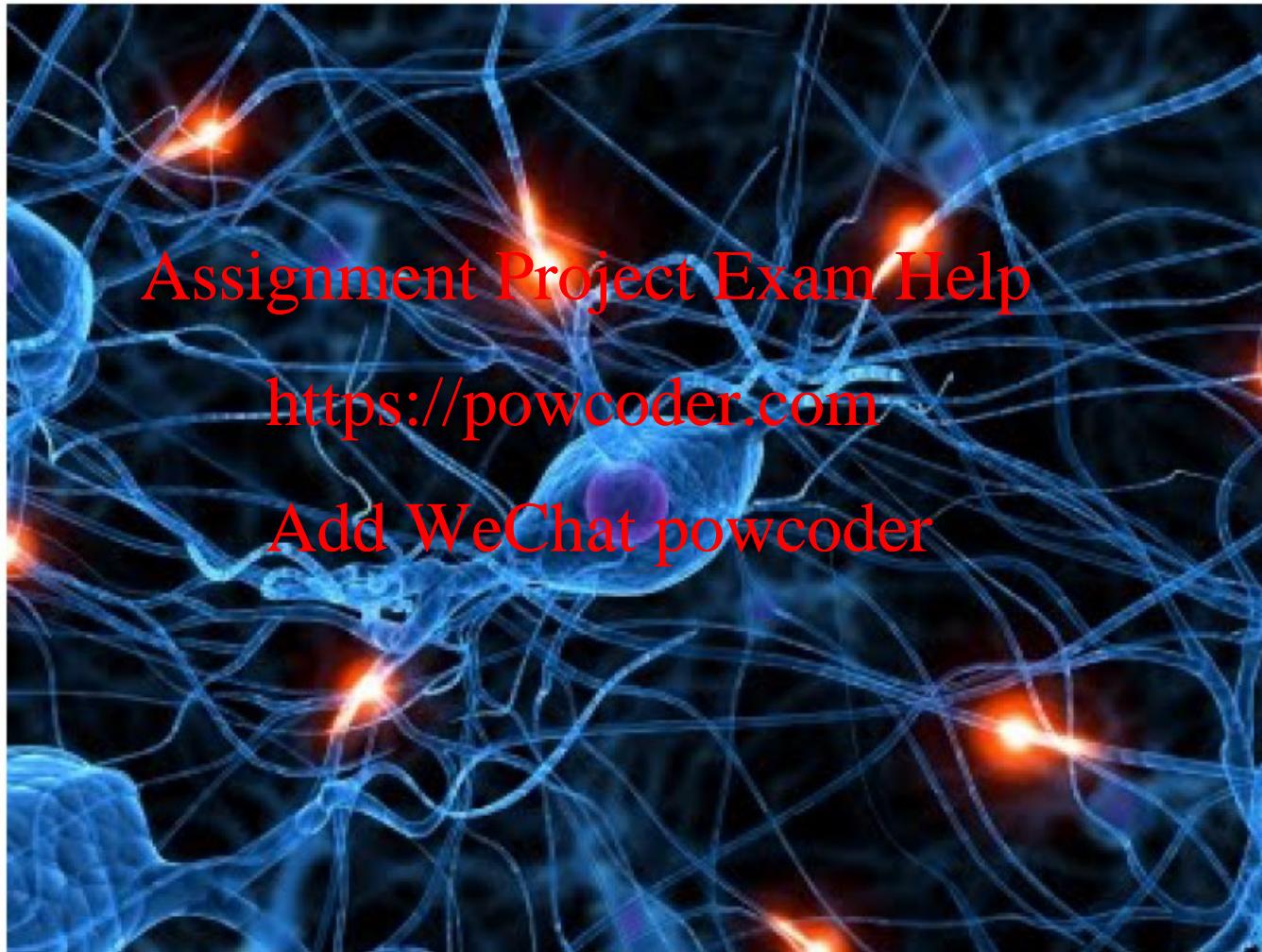


Protein-Protein Interaction Networks:

Nodes: Proteins
Edges: 'physical' interactions

Metabolic networks:
Nodes: Metabolites and enzymes Edges:
Chemical reactions

Networks: Brain



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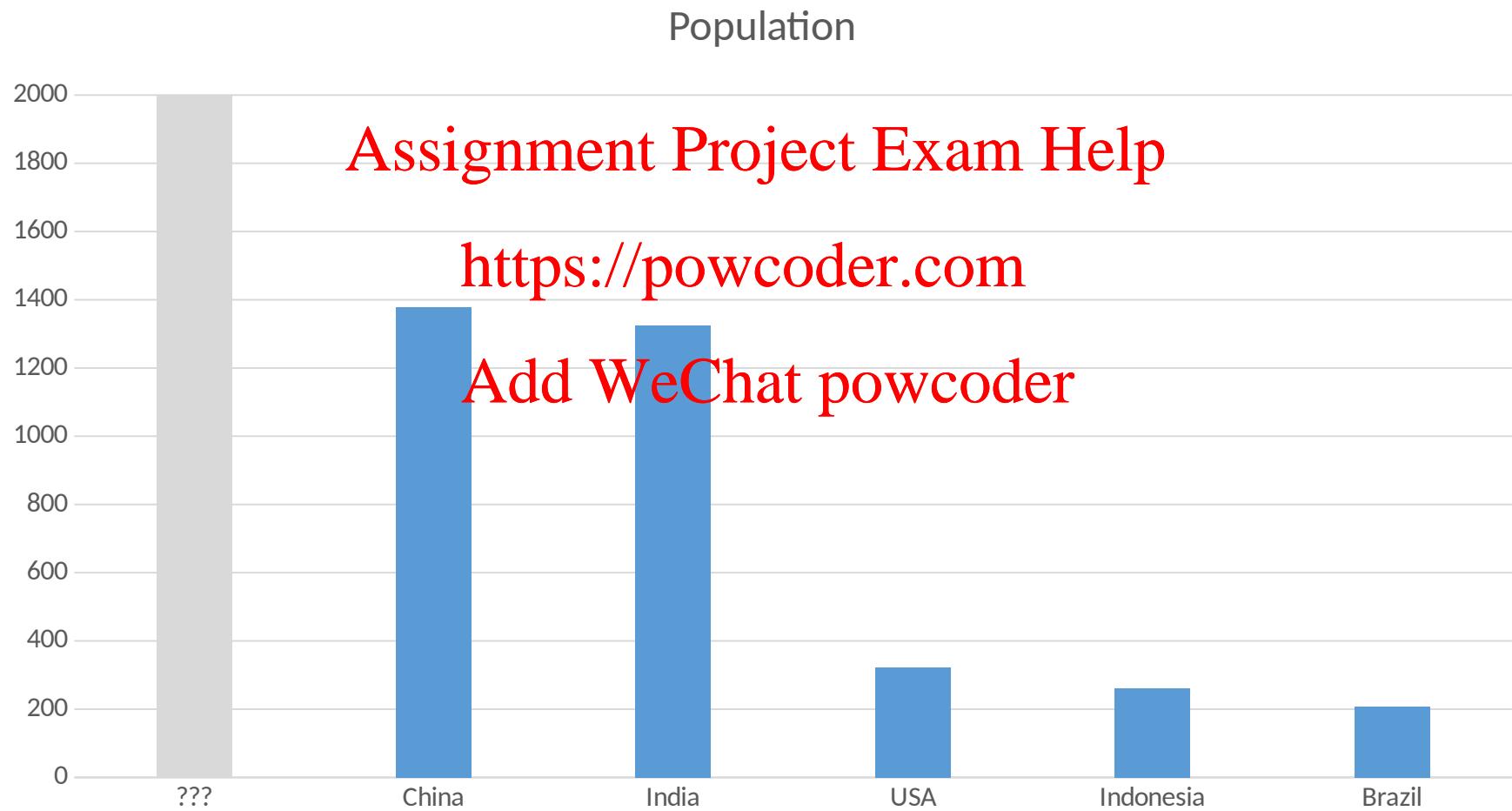
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**Human brain has between 10-100
billion neurons [Sporns, 2011]**

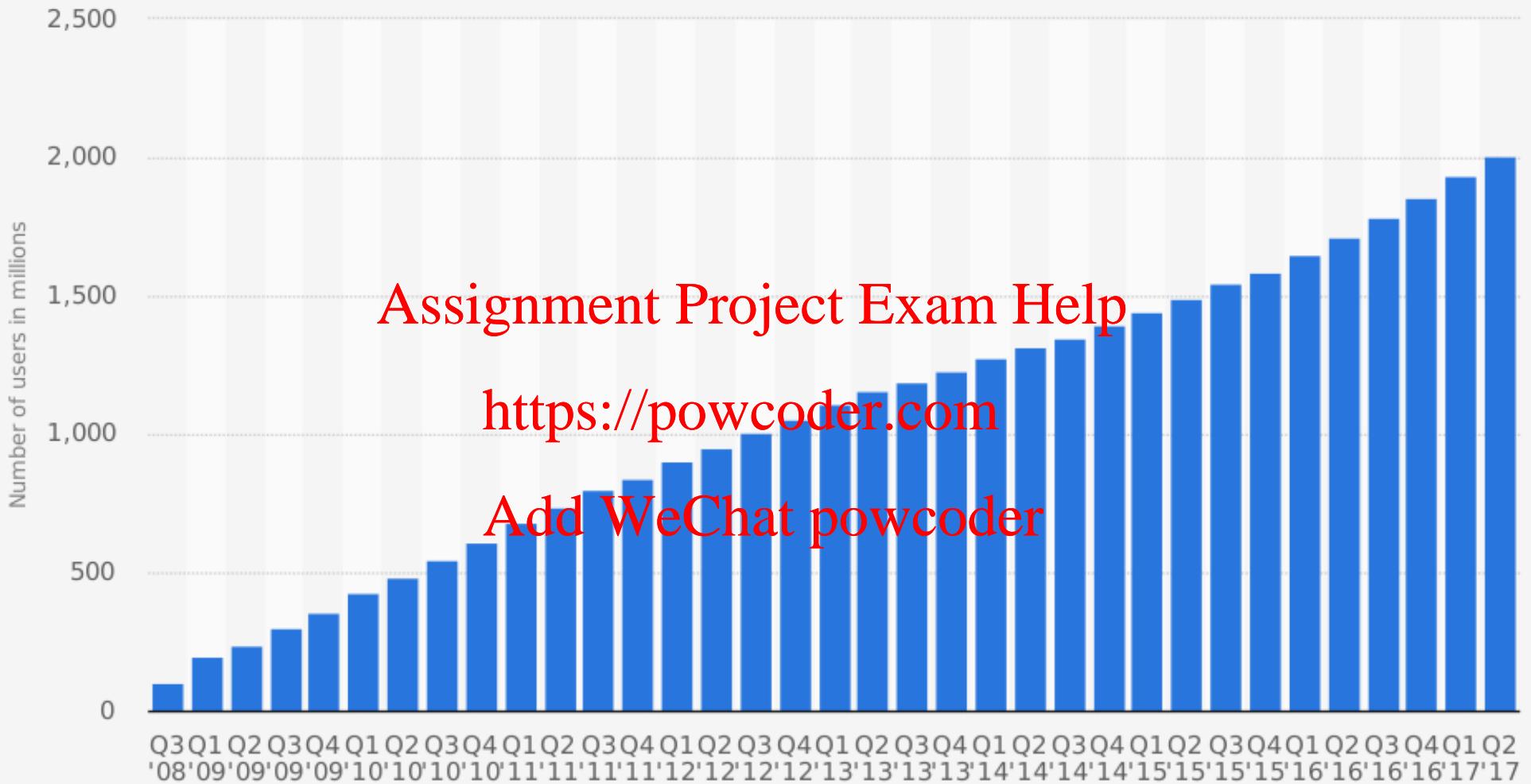
Networks: Social



Country with Largest World Population?



Number of monthly active Facebook users worldwide as of 2nd quarter 2017 (in millions)



Source
Facebook
© Statista 2017

Additional Information:
Worldwide; Facebook; 3rd quarter 2008 to 2nd quarter 2017

If Facebook were a Country..

The Republic of Facebook

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If Facebook were a country....

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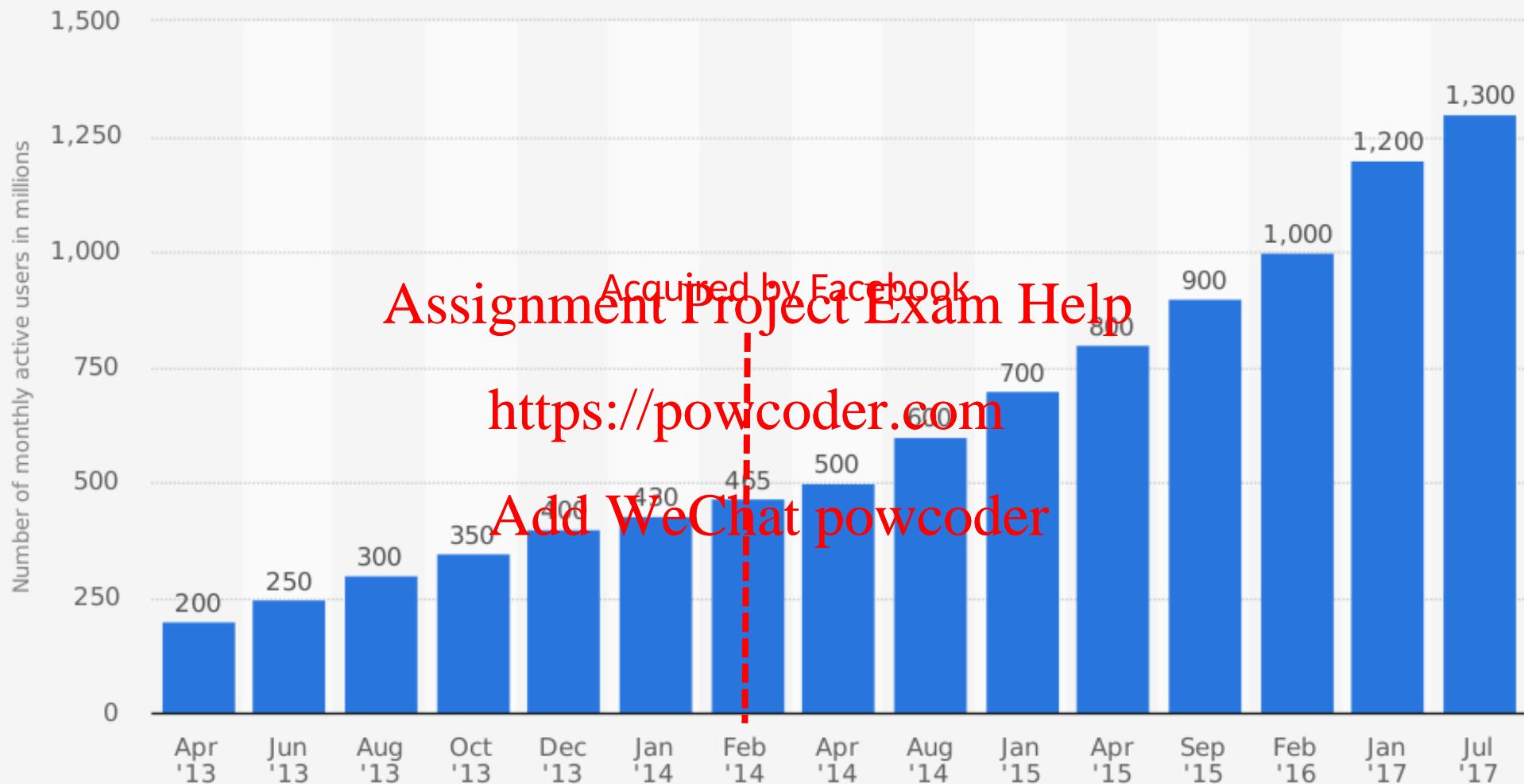
It would be home to 1 in 7 of the world's entire population

4

Sources
www.newsroom.facebook.com/Key-Facts
www.en.wikipedia.org/wiki/World_Population

www.blogsession.co.uk

Number of monthly active WhatsApp users worldwide from April 2013 to July 2017 (in millions)



Source

WhatsApp; Facebook; TechCrunch
© Statista 2017

Additional Information:

Worldwide; WhatsApp; April 2013 to July 2017

Networks!

- Behind each such system there is an intricate wiring diagram, **a network**, that defines the **interactions** between the components

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- **We will never understand these systems unless we understand the networks behind them!**

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Social Web - The Lab of Knowledge

ar



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The web is our “laboratory” for
knowledge and understanding
humanity

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Networks: Problems

- **Social networks:**

- Link prediction, friend recommendation
- Social circle detection, community detection
- Social recommendations
- Identifying influential nodes, viral marketing

- **Communication networks:**

- Intrusion detection, fraud
- Churn prediction

- **Information networks:**

- Navigational aids

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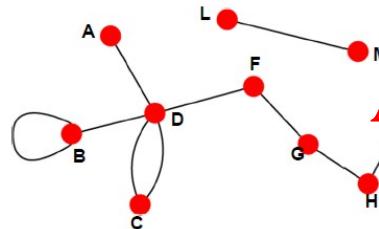
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Social Network as Graph

- **Undirected graphs**

- Links: undirected
(symmetrical, reciprocal
relations)

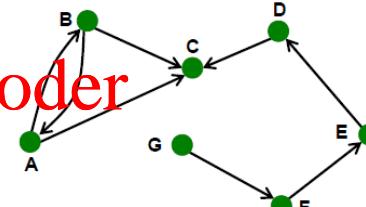


- **Directed graphs**

- Links: directed
(asymmetrical relations)

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- Undirected links:

- Collaborations
- Friendship on Facebook

- Directed links:

- Phone calls
- Following on Twitter

Adjacency Matrix

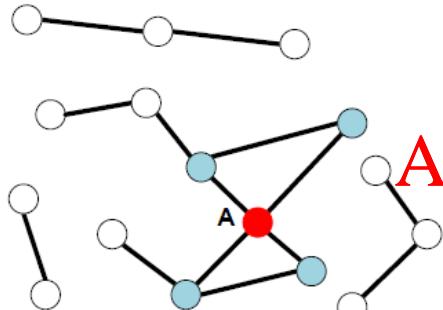


$A_{ij} = 1$ if there is a link from node i to node j

$A_{ij} = 0$ otherwise

Node Degrees

Undirected



- Node degree : the number of edges adjacent to node

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- Avg. degree: $k = \langle k \rangle = \frac{1}{N} \sum_{i=1}^N k_i = \frac{2E}{N}$

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- In directed networks we define an **in? degree** and **out?degree**.
 - The (total) degree of a node is the sum of in? and out?degrees.

$$k_C^{in} = 2 \quad k_C^{out} = 1 \quad k_C = 3$$

$$\bar{k} = \frac{E}{N} \quad \bar{k}^{in} = \bar{k}^{out}$$

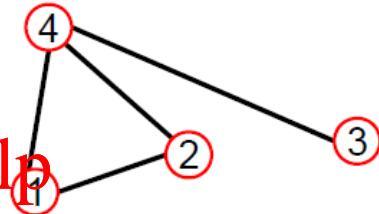
Degree Matrix

- This graph has entries only on the diagonal.

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- The entry for row and column i is the degree of the i th node

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$$D = \begin{pmatrix} 2 & 0 & 0 & 0 \\ 0 & 2 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 3 \end{pmatrix}$$

Laplacian Matrix

- Suppose a graph as adjacency matrix and degree matrix
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- Laplacian matrix is defined as .
- the Laplacian matrix
 - Has the same entries as on the diagonal Add WeChat powcoder
 - Off the diagonal, at row and column ,
 - has -1 if there is an edge between nodes and
 - 0 if not.

In-class Practice 1

- Go to [practice](#)

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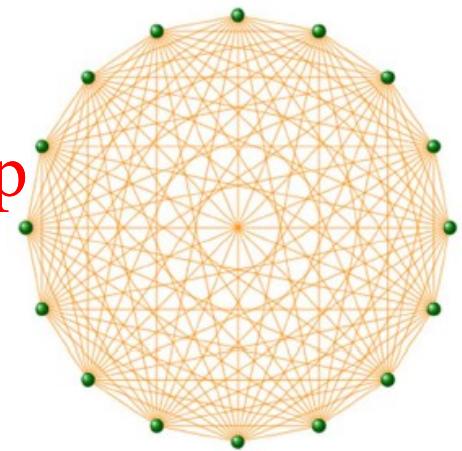
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Complete Graph

- The **maximum number of edges** in an undirected graph on **N** nodes is

$$E_{\max} = \binom{N}{2} = \frac{N(N-1)}{2}$$

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- A graph with the number of edges is a **complete graph**

- and its average degree is
- N-1**

Networks are Sparse Graphs

- Most real-world networks are sparse

- $E \ll E_{\max}$ (or $k \ll N-1$)

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WWW (Stanford-Berkeley):	$N=319,717$	$\langle k \rangle=9.65$
Social networks (LinkedIn):	$N=6,946,668$	$\langle k \rangle=8.87$
Communication (MSN IM):	$N=242,720,596$	$\langle k \rangle=11.1$
Coauthorships (DBLP):	$N=317,080$	$\langle k \rangle=6.62$
Internet (AS-Skitter):	$N=1,719,031$	$\langle k \rangle=14.91$
Roads (California):	$N=1,957,027$	$\langle k \rangle=2.82$
Proteins (<i>S. Cerevisiae</i>):	$N=1,870$	$\langle k \rangle=2.39$

(Source: Leskovec et al., Internet Mathematics, 2009)

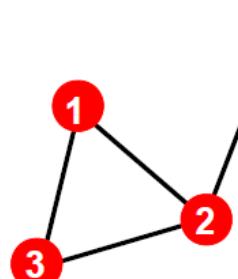
- Consequence: Adjacency matrix is filled with zeros!

- Density of the matrix (E/N^2):

WWW= $1.51 \cdot 10^{-5}$, MSN IM Instant Messenger = $2.27 \cdot 10^{-8}$

More Types of Graphs

Unweighted



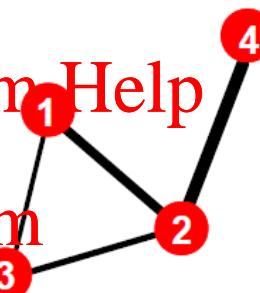
$$A_{ij} = \begin{pmatrix} 0 & 1 & 1 & 0 \\ 1 & 0 & 1 & 1 \\ 1 & 1 & 0 & 0 \\ 0 & 1 & 0 & 0 \end{pmatrix}$$

$$A_{ii} = 0 \quad A_{ij} = A_{ji}$$

$$E = \frac{1}{2} \sum_{i,j=1}^N A_{ij} \quad \bar{k} = \frac{2E}{N}$$

Examples: Friendship, Hyperlink

Weighted



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$$A_{ij} = \begin{pmatrix} 0 & 2 & 0.5 & 0 \\ 2 & 0 & 1 & 4 \\ 0.5 & 1 & 0 & 0 \\ 0 & 4 & 0 & 0 \end{pmatrix}$$

$$A_{ii} = 0$$

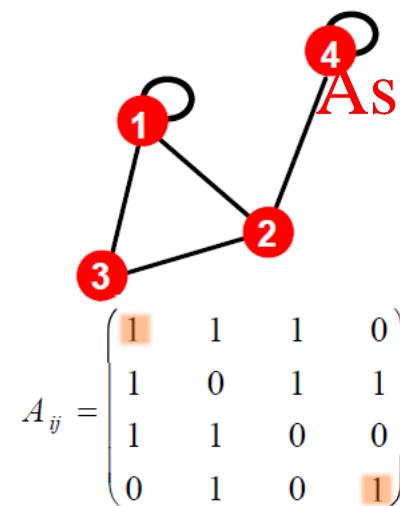
$$A_{ij} = A_{ji}$$

$$E = \frac{1}{2} \sum_{i,j=1}^N \text{nonzero}(A_{ij}) \quad \bar{k} = \frac{2E}{N}$$

Examples: Collaboration, Internet, Roads

More Types of Graphs

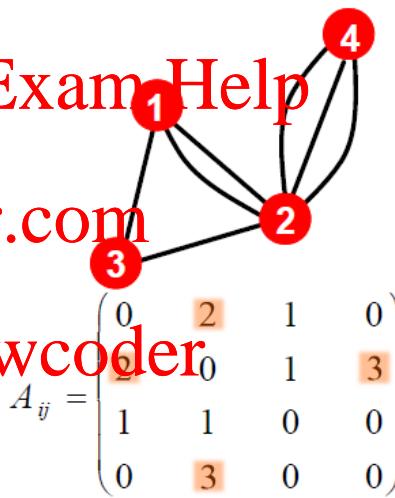
Self-edges (Self-loops)



$$E = \frac{1}{2} \sum_{i,j=1, i \neq j}^N A_{ij} + \sum_{i=1}^N A_{ii} \quad ?$$

Examples: Proteins, Hyperlink

Multigraph



$$E = \frac{1}{2} \sum_{i,j=1}^N \text{nonzero}(A_{ij}) \quad \bar{k} = \frac{2E}{N}$$

Examples: Communication, Collaboration

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Network Representations

WWW >> directed multigraph with self-edges

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Facebook friendships >> undirected, unweighted

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Citation networks >> unweighted, directed, acyclic

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Collaboration networks >> undirected multigraph or weighted graph

Mobile phone calls >> directed, (weighted?) multigraph

Protein Interactions >> undirected, unweighted with self-interactions

Bipartite Graph

- **Bipartite graph** is a graph whose nodes can be divided into two disjoint sets U and V such that every link connects a node in U to one in V ; that is, U and V are **independent sets**.

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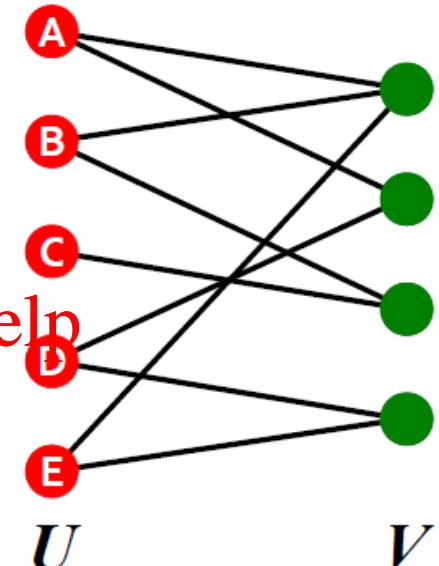
- **Examples:**

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- Authors \rightarrow papers (they authored)
- Actors \rightarrow Movies (they appeared in)
- Users \rightarrow Movies (they rated)

- **“Folded” networks:**

- Author collaboration networks
- Movie co-rating networks



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<https://powcoder.com> as graphs

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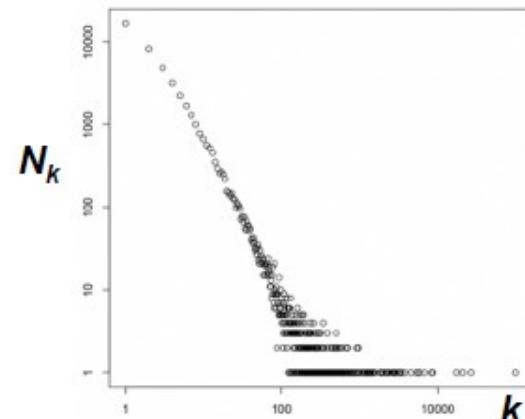
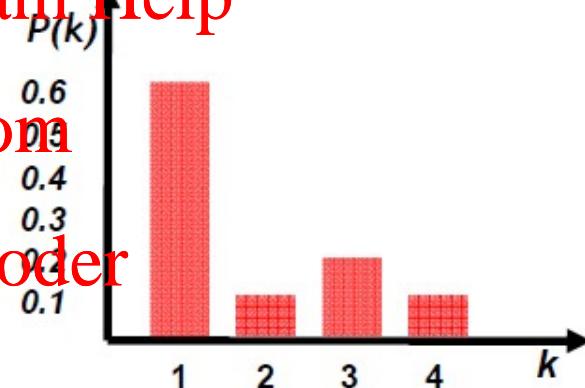
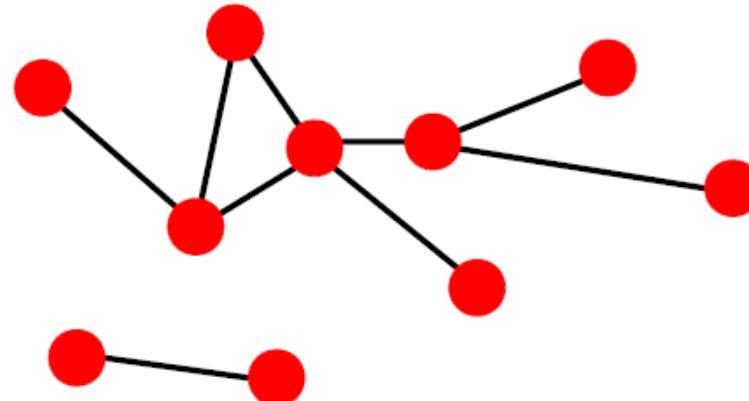
Degree Distribution

- Degree distribution : Probability that a randomly chosen node has degree k
 - $N_k = \# \text{ nodes with degree } k$
- Normalized:
 - $P(k) = N_k / N \rightarrow \text{plot}$

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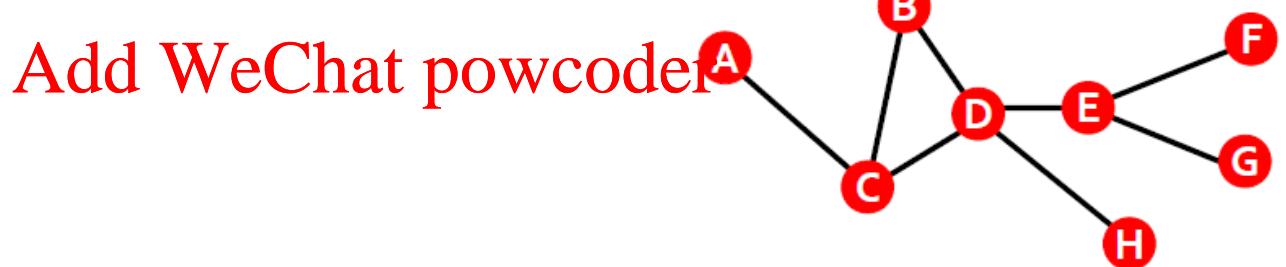
Paths in a Graph

- A **path** is a sequence of nodes in which each node is linked to the next one

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$$P_n = \{i_0, i_1, i_2, \dots, i_n\} \quad P_n = \{(i_0, i_1), (i_1, i_2), (i_2, i_3), \dots, (i_{n-1}, i_n)\}$$

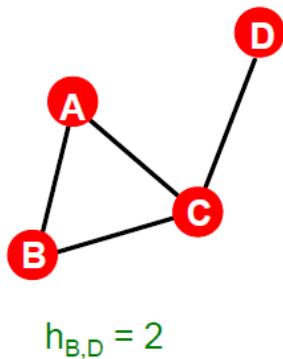
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- Path can intersect itself and pass through the same edge multiple times

- E.g.: ACBDCDEG
- In a directed graph a path can only follow the direction of the “arrow”

Distance in a Graph



- **Distance (shortest path, geodesic)** between a pair of nodes is defined as the number of edges along the shortest path connecting the nodes
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 - *If the two nodes are disconnected, the distance is usually defined as infinite
- In **directed graphs** paths need to follow the direction of the arrows
 - Consequence: Distance is **not symmetric**:
 $h_{A,C} \neq h_{C,A}$

Network Diameter

- **Diameter**: the maximum (shortest path) distance between any pair of nodes in a graph
- **Average path length** for a connected graph (component) or a strongly connected component of a directed graph:

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$$\bar{h} = \frac{1}{2E_{\max}} \sum_{i,j \neq i} h_{ij}$$
$$E_{\max} = \binom{N}{2} = \frac{N(N-1)}{2}$$

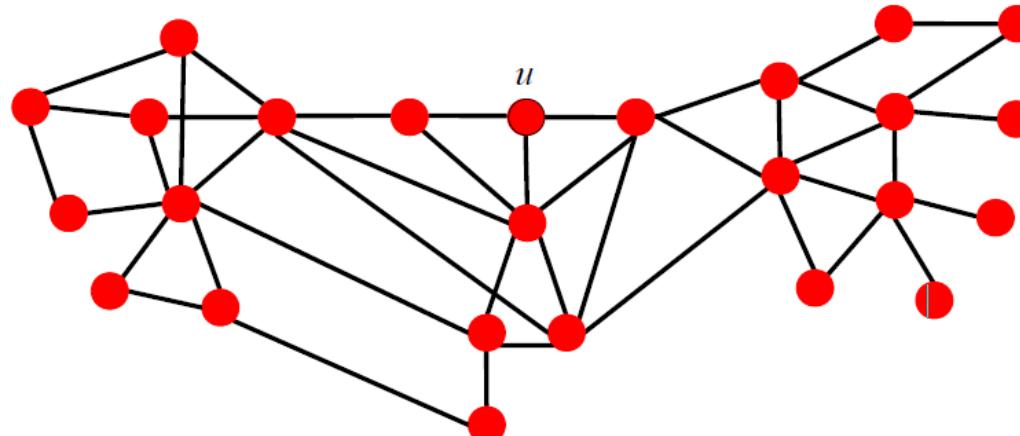
where h_{ij} is the distance from node i to node j

- Many times we compute the average only over the connected pairs of nodes (we ignore “infinite” length paths)

Finding the Shortest Path

- **BreadthFirst Search:**

- Start with node u , mark it to be at distance $h_u(u)=0$, add u to the queue
- While the queue not empty:
 - Take node v off the queue, put its unmarked neighbors w into the queue and mark $h_u(w)=h_u(v)+1$



Clustering Coefficient

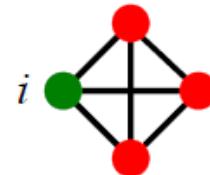
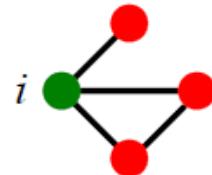
- **Clustering coefficient:**

- What portion of i 's neighbors are connected?
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- Node i with degree k_i

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where e_i is the number of edges

between the neighbors of node i .



Clustering Coefficient

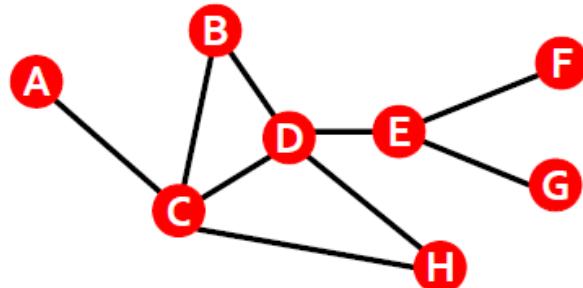
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where e_i is the number of edges

between the neighbors of node i .
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$$k_B=2, \quad e_B=1,$$

$$k_D=4, \quad e_D=2,$$

Key Network Properties

- Degree distribution:
- Path length:
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- Clustering coefficient:
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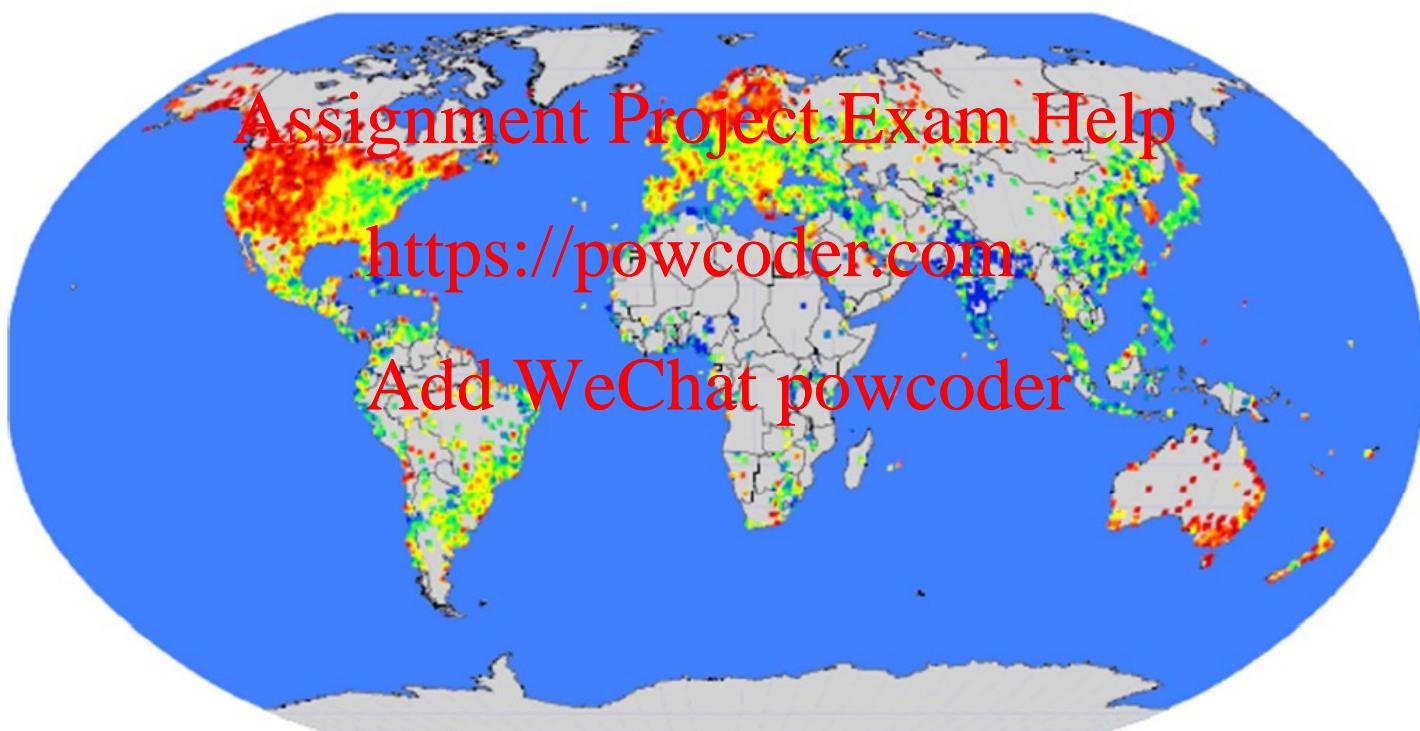
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The MSN Messenger

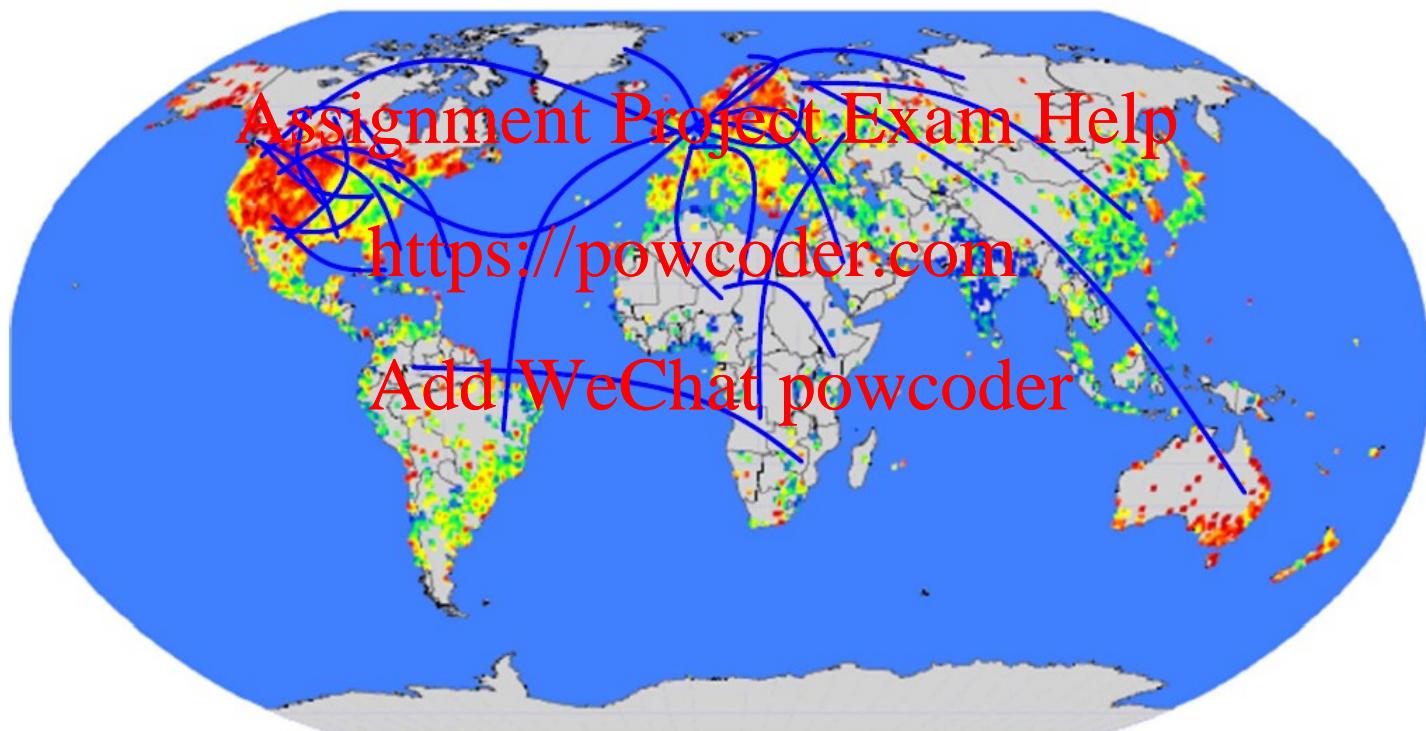
- **MSN Messenger activity in June 2006:**
 - 150Gb/day (compressed)
 - 4.5Tb / month
 - 245 million users logged in
 - 180 million users engaged in conversations
 - More than 30 billion conversations
 - More than 255 billion exchanged messages



Communication: Geography

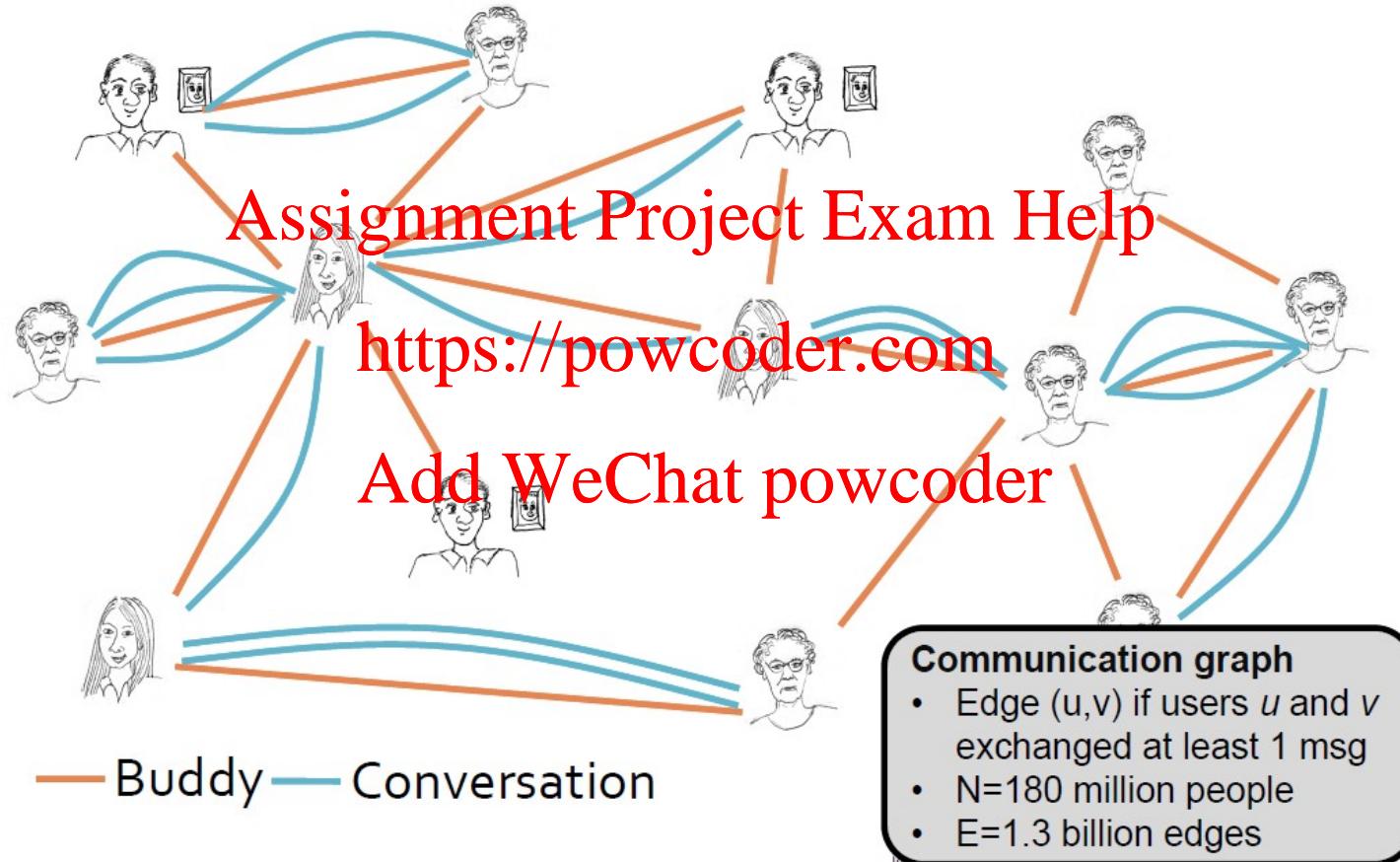


Communication Network

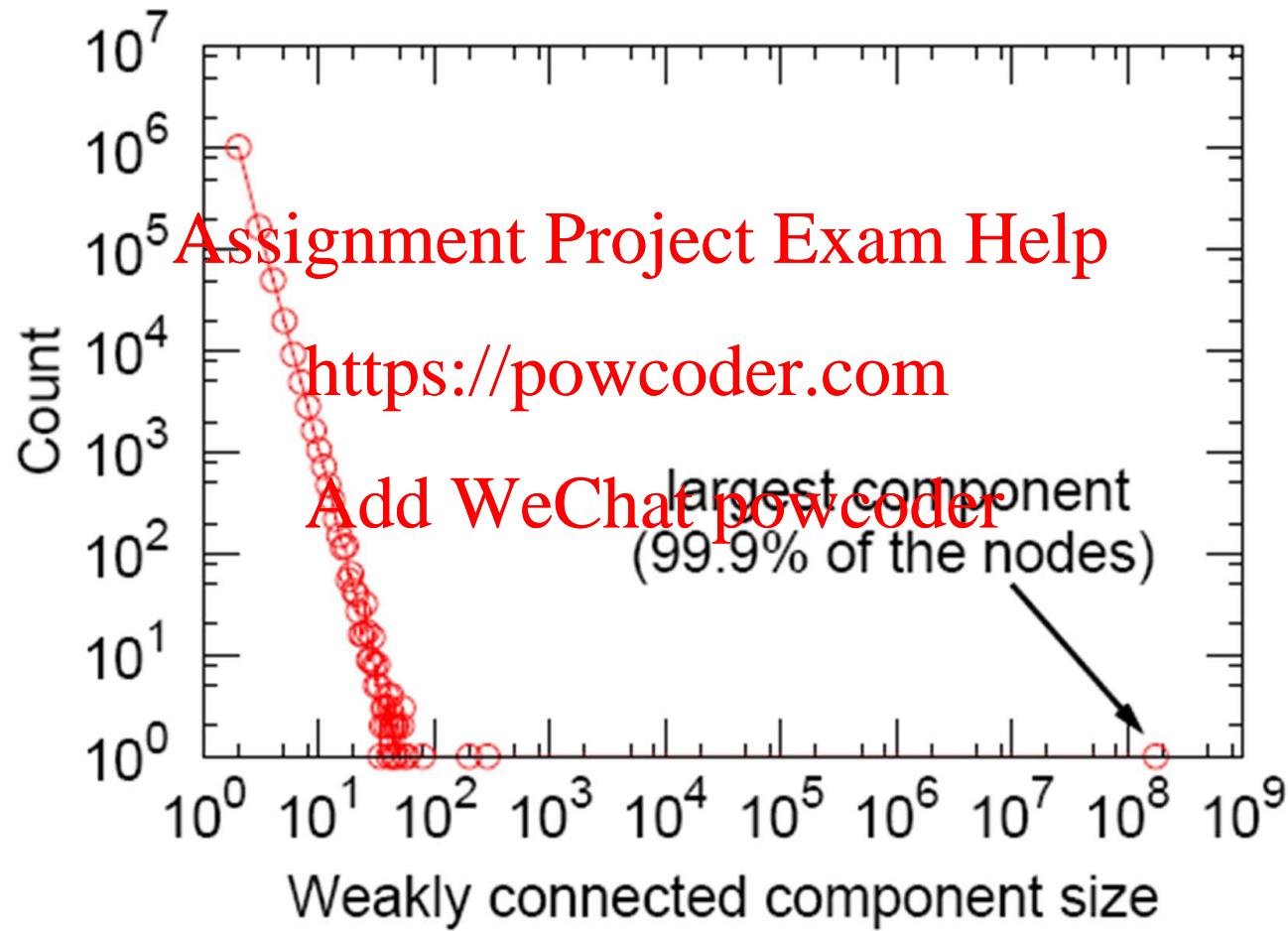


Network: 180M people, 1.3B edges

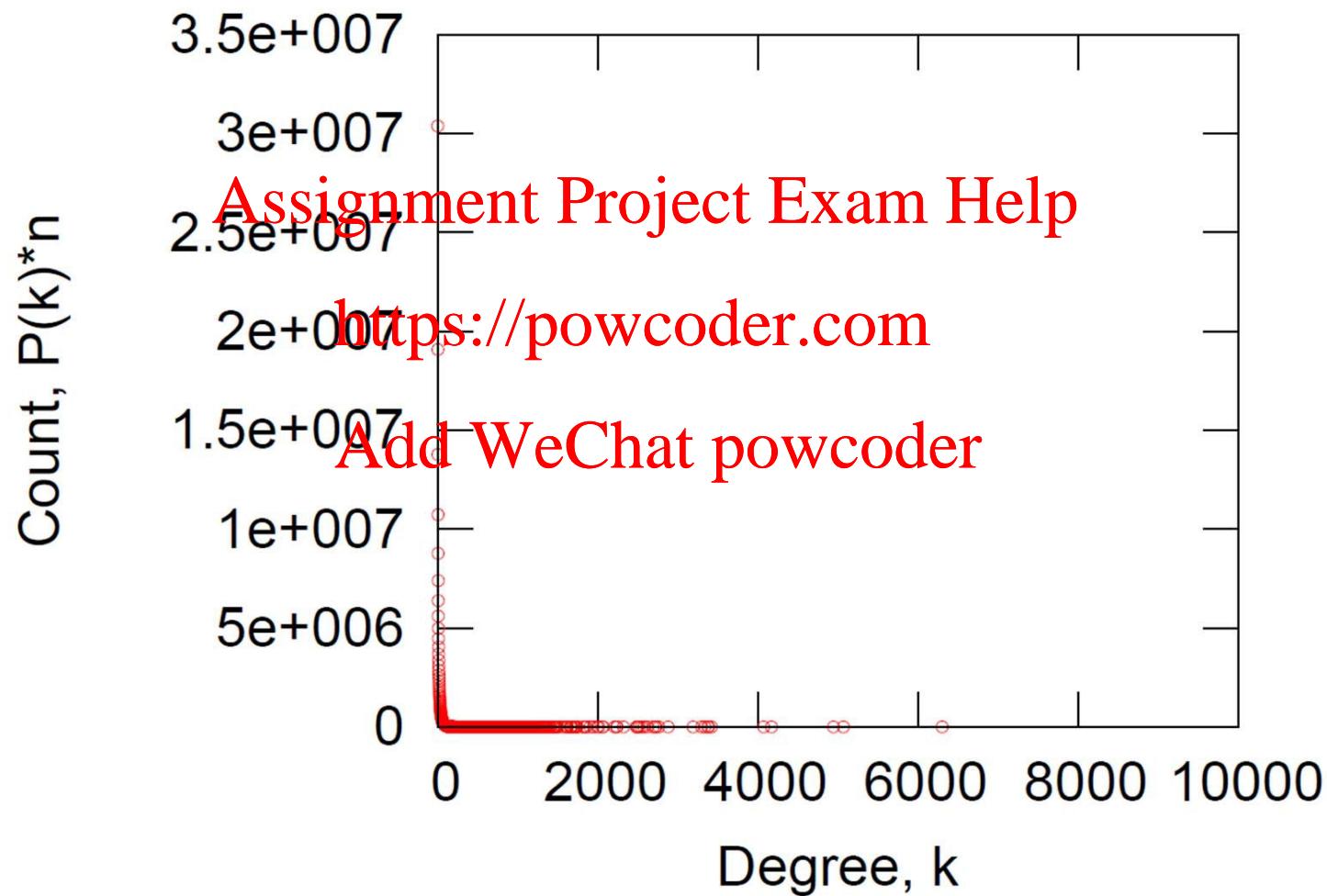
Messaging as a Network



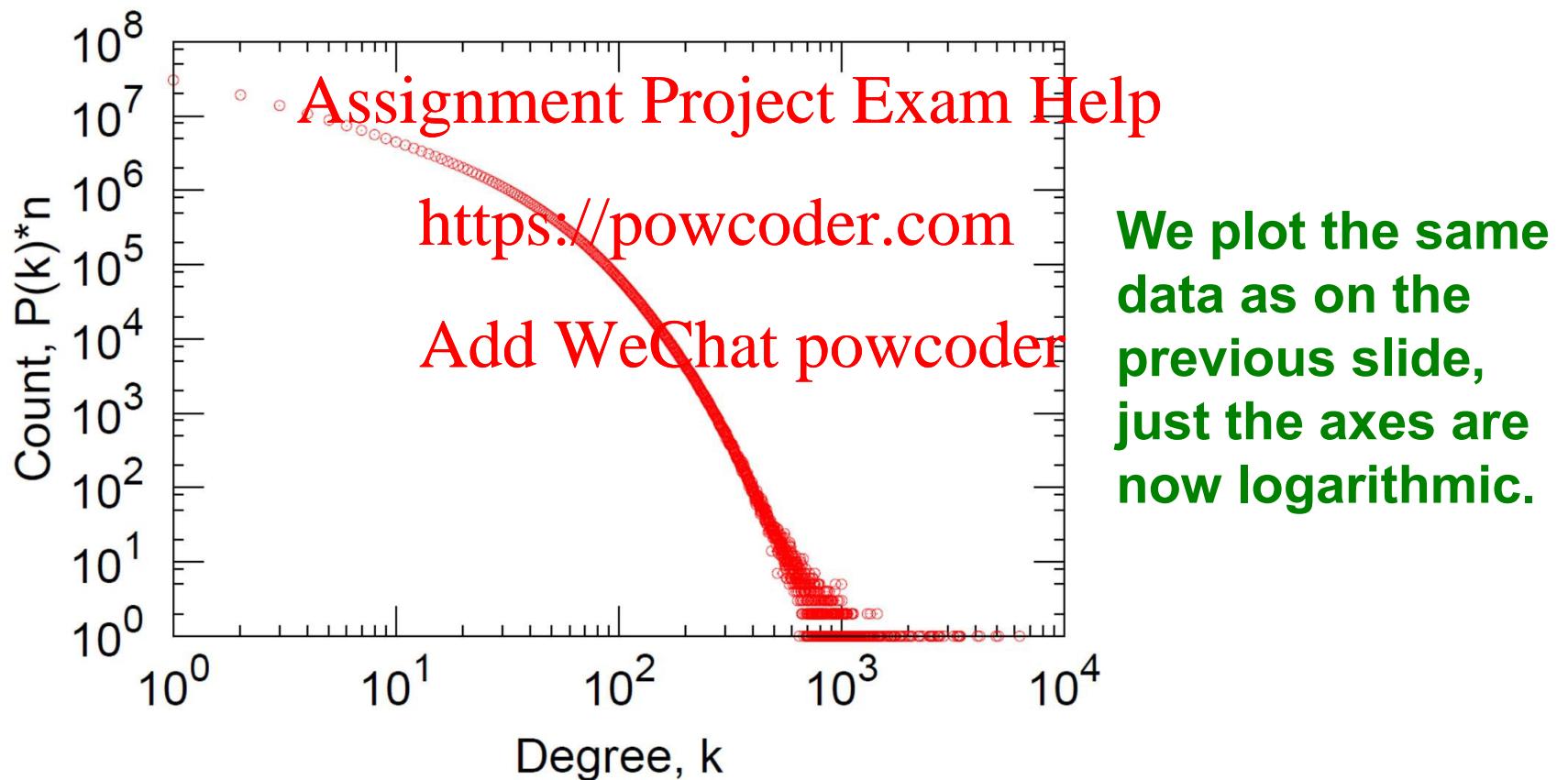
MSN Network: Connectivity



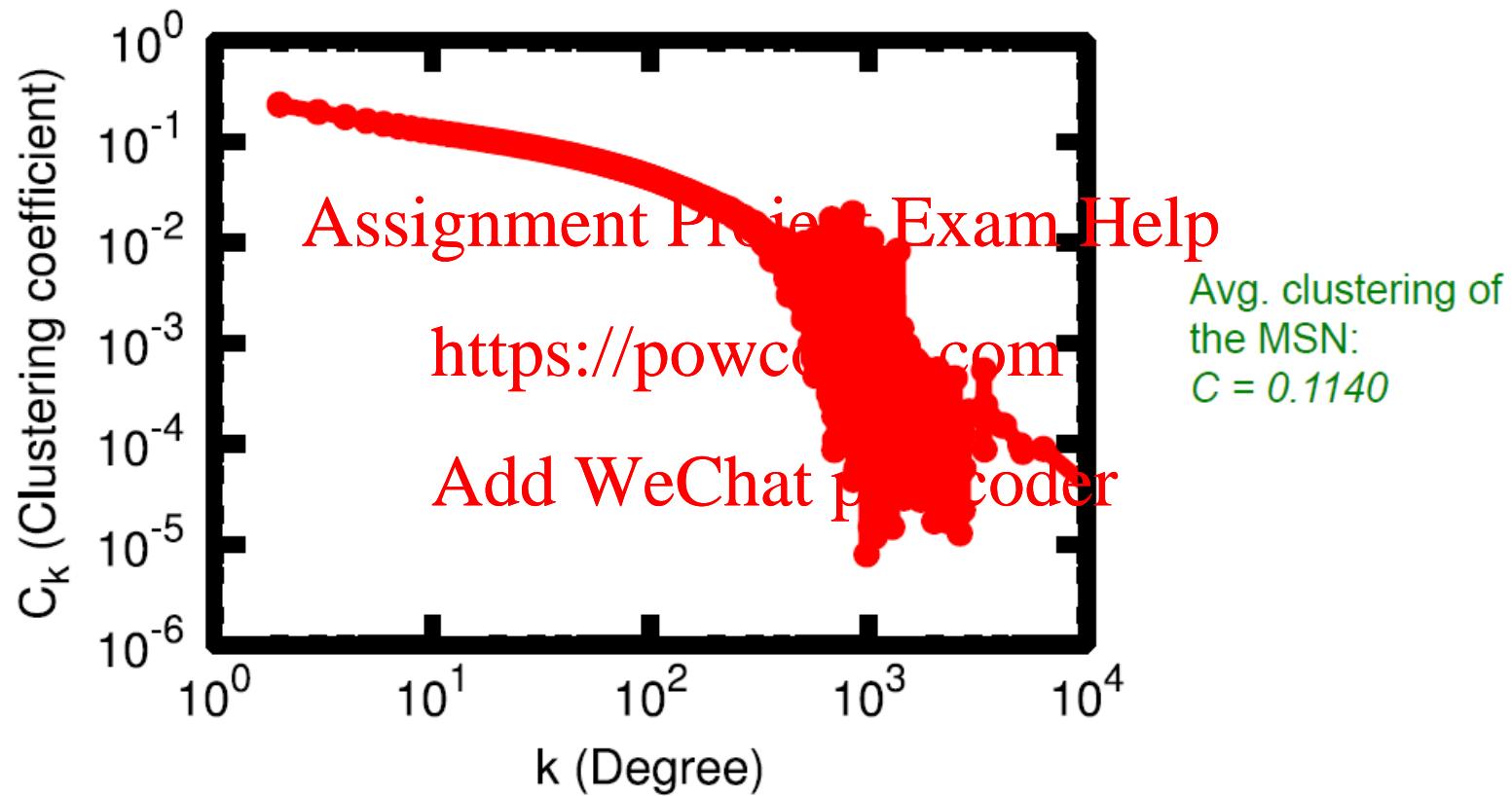
MSN: Degree Distribution



MSN: Log-Log Degree Distribution

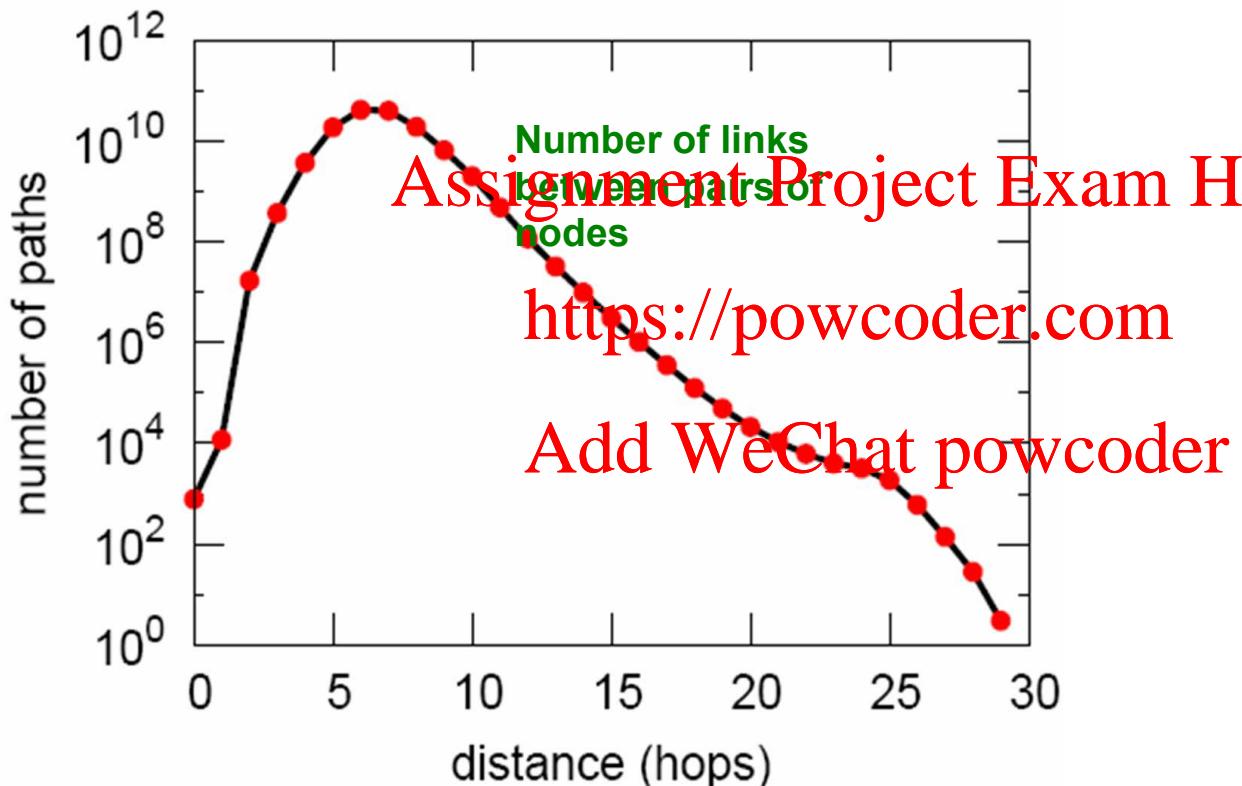


MSN: Clustering



$$C_k: \text{average } C_i \text{ of nodes } i \text{ of degree } k: C_k = \frac{1}{N_k} \sum_{i:k_i=k} C_i$$

MSN: Diameter



Steps	#Nodes
0	1
1	10
2	78
3	3,96
4	8,648
5	3,299,252
6	28,395,849
7	79,059,497
8	52,995,778
9	10,321,008
10	1,955,007
11	518,410
12	149,945
13	44,616
14	13,740
15	4,476
16	1,542
17	536
18	167
19	71
20	29
21	16
22	10
23	3
24	2
25	3

MSN: Key Network Properties

- Degree distribution: heavily skewed. avg degree = 14.4
- Path length: 6.6
- Clustering coefficient: 0.11

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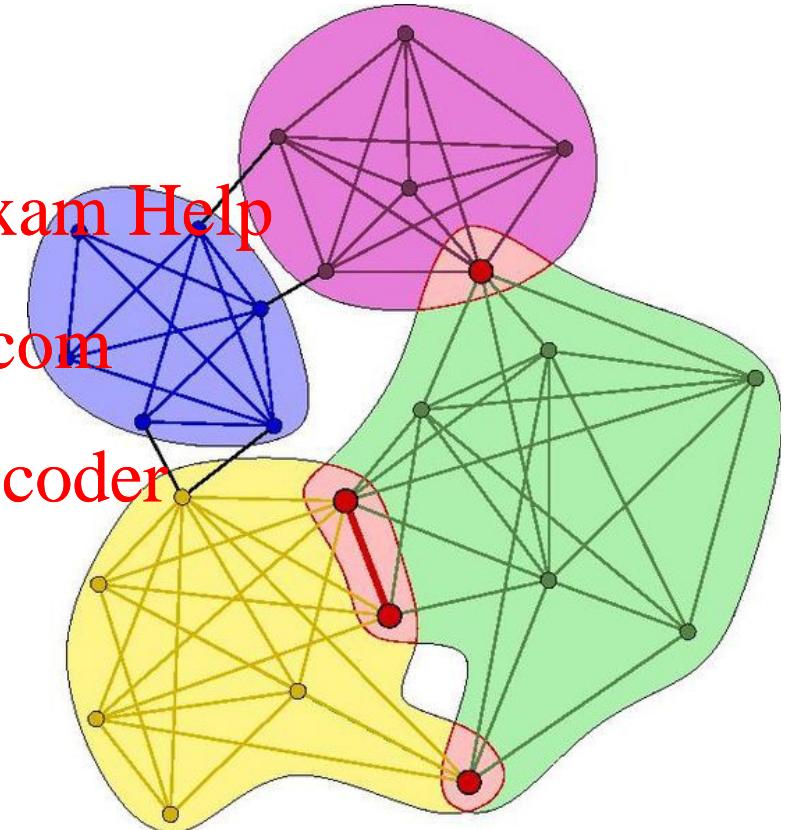
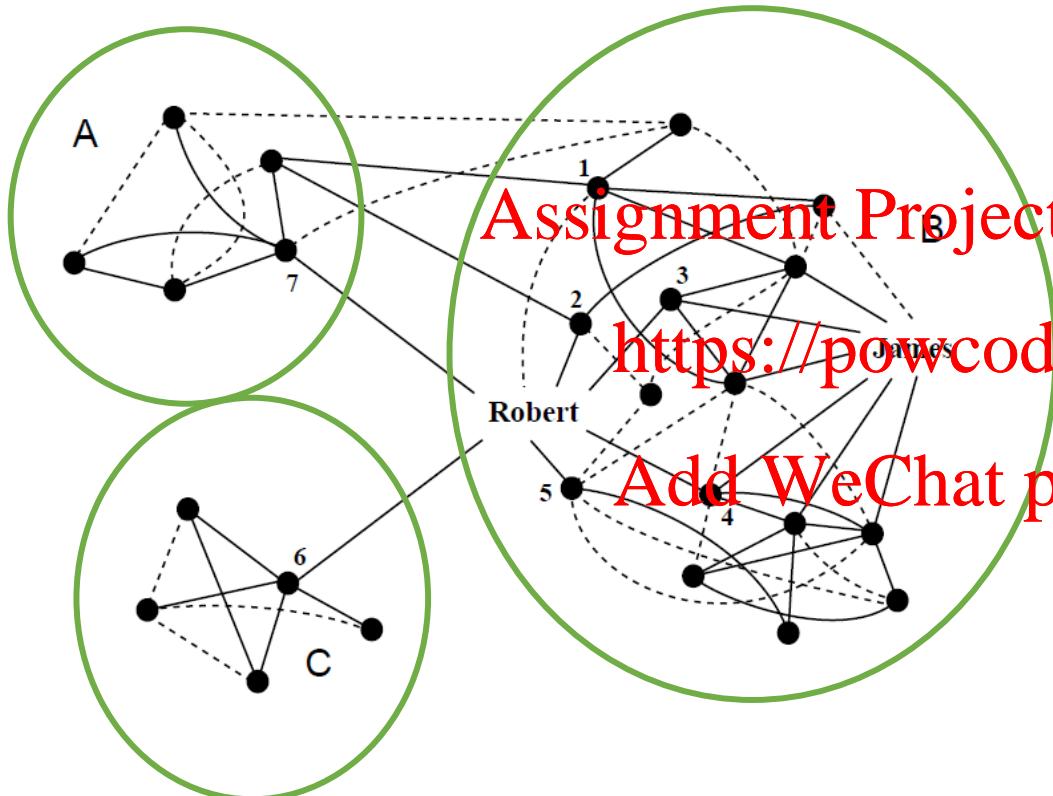
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Community Detection

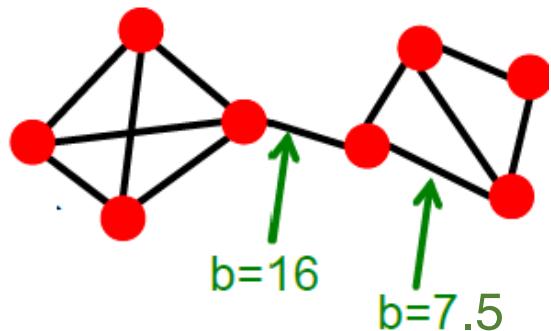
How to find communities?



We will work with **undirected** (unweighted) networks

Betweenness: Strength of Ties

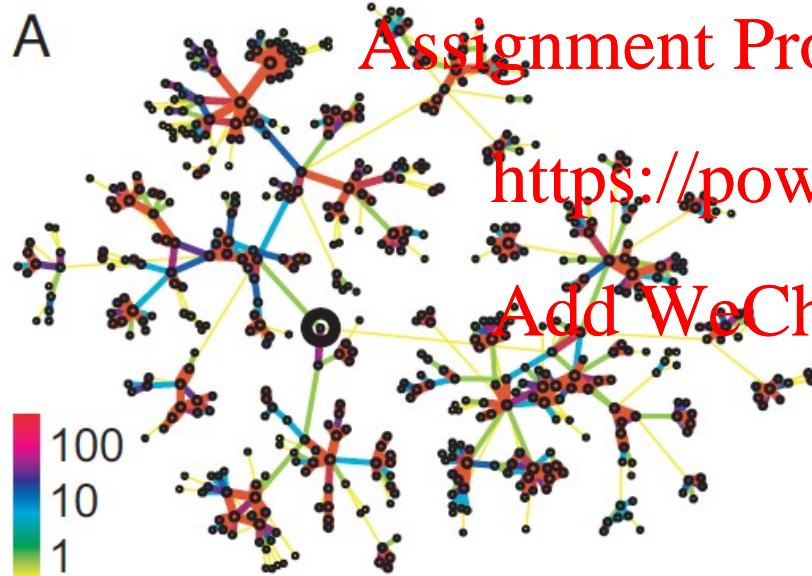
- **Edge betweenness**: number of shortest paths (among all pair of vertices) passing over the edge.
- If there is more than one shortest path between a pair of nodes, each path is assigned equal weight such that the total weight of all the paths is equal to 1.
<https://powcoder.com>



Add WeChat [powcoder](#) for simple algorithm for calculating betweenness.

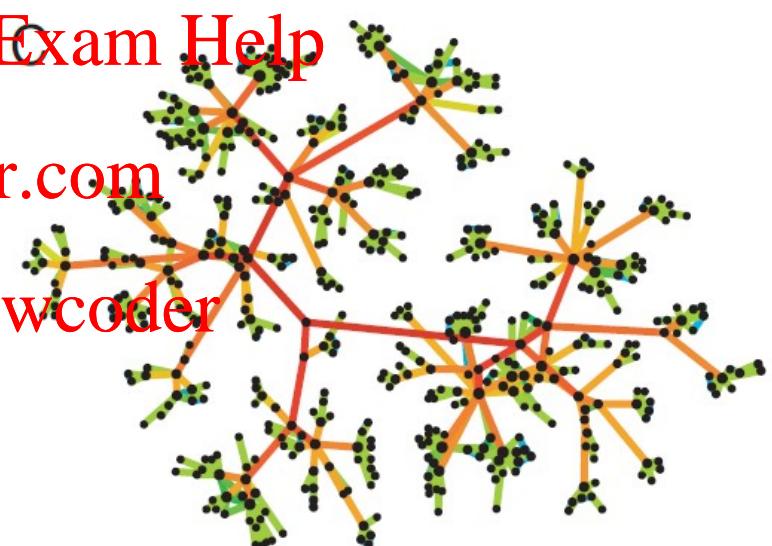
1. Repeat for each vertex
 - a) compute its shortest path to all other vertices.
 - b) for each edge, count the number of the shortest paths passing over the edge
2. Sum up the results from each vertex.

Betweenness: Strength of Ties



Calculation for 1 vertex

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Calculation for all vertices

In-class Practice 2

- Go to [practice](#)

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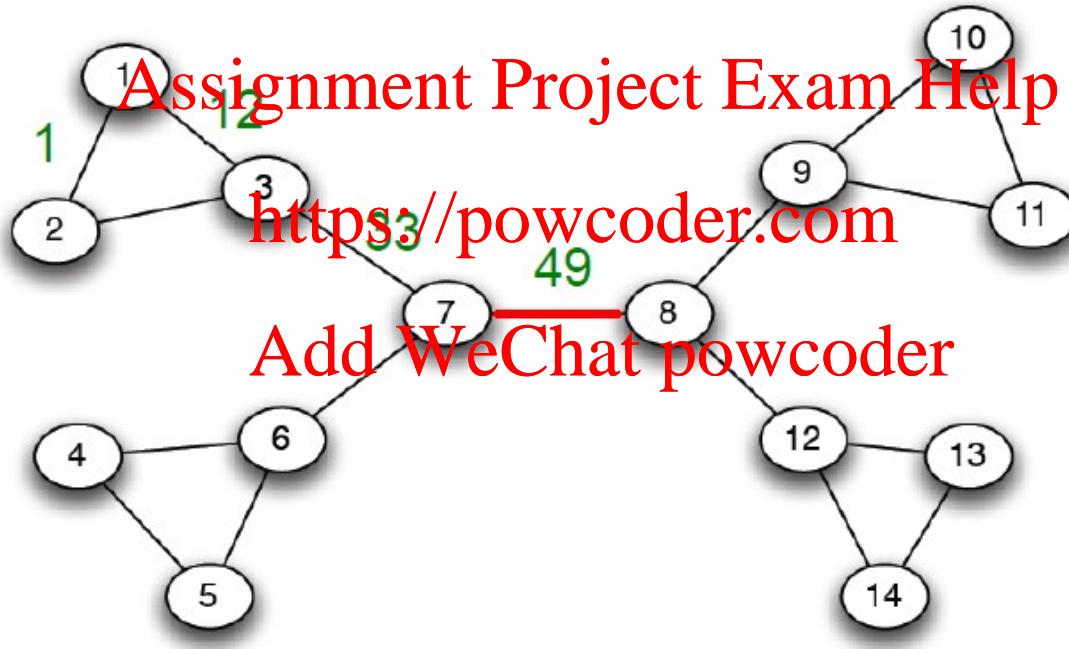
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Method for Community Detection: Girvan-Newman

- Divisive hierarchical clustering based on the notion of edge **betweenness**:
 - Number of shortest paths passing through the edge
- Girvan-Newman <https://powcoder.com>
 - Undirected unweighted networks
 - Repeat until no edges are left:
 - Calculate betweenness of edges
 - Remove edges with highest betweenness
 - Connected components are communities
 - Gives a hierarchical decomposition of the network

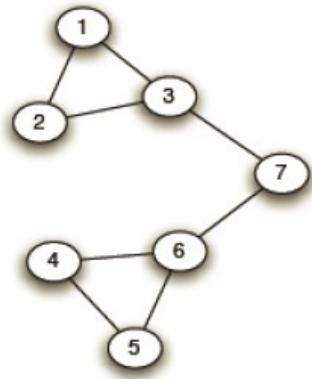
Girvan-Newman: Example



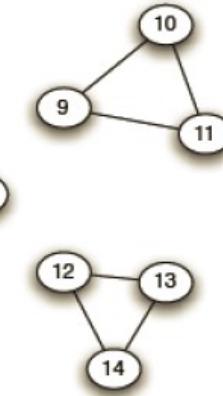
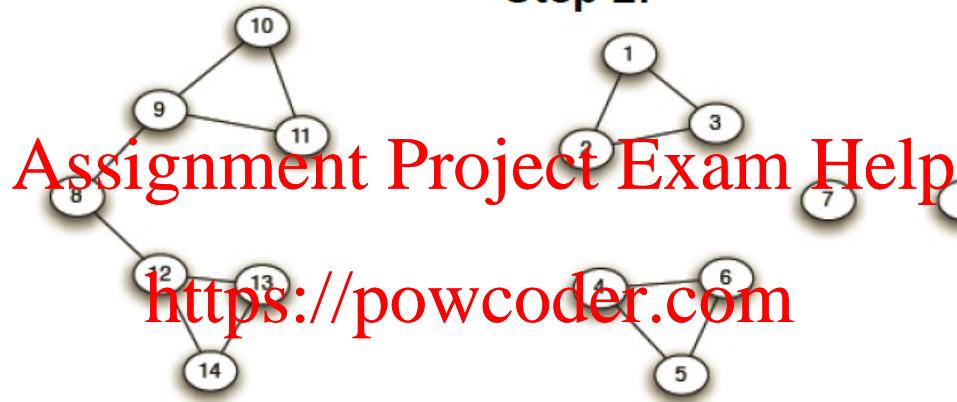
Need to re-compute betweenness at every step

Girvan-Newman: Example

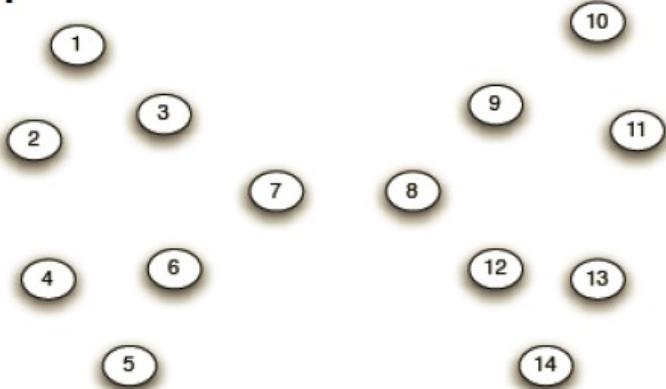
Step 1:



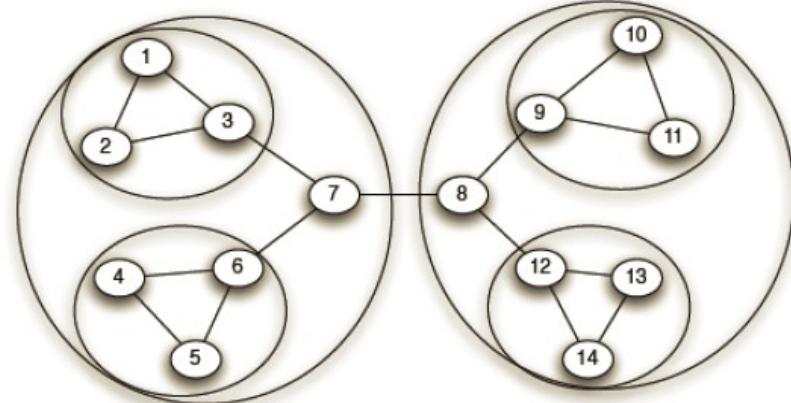
Step 2:



Step 3:

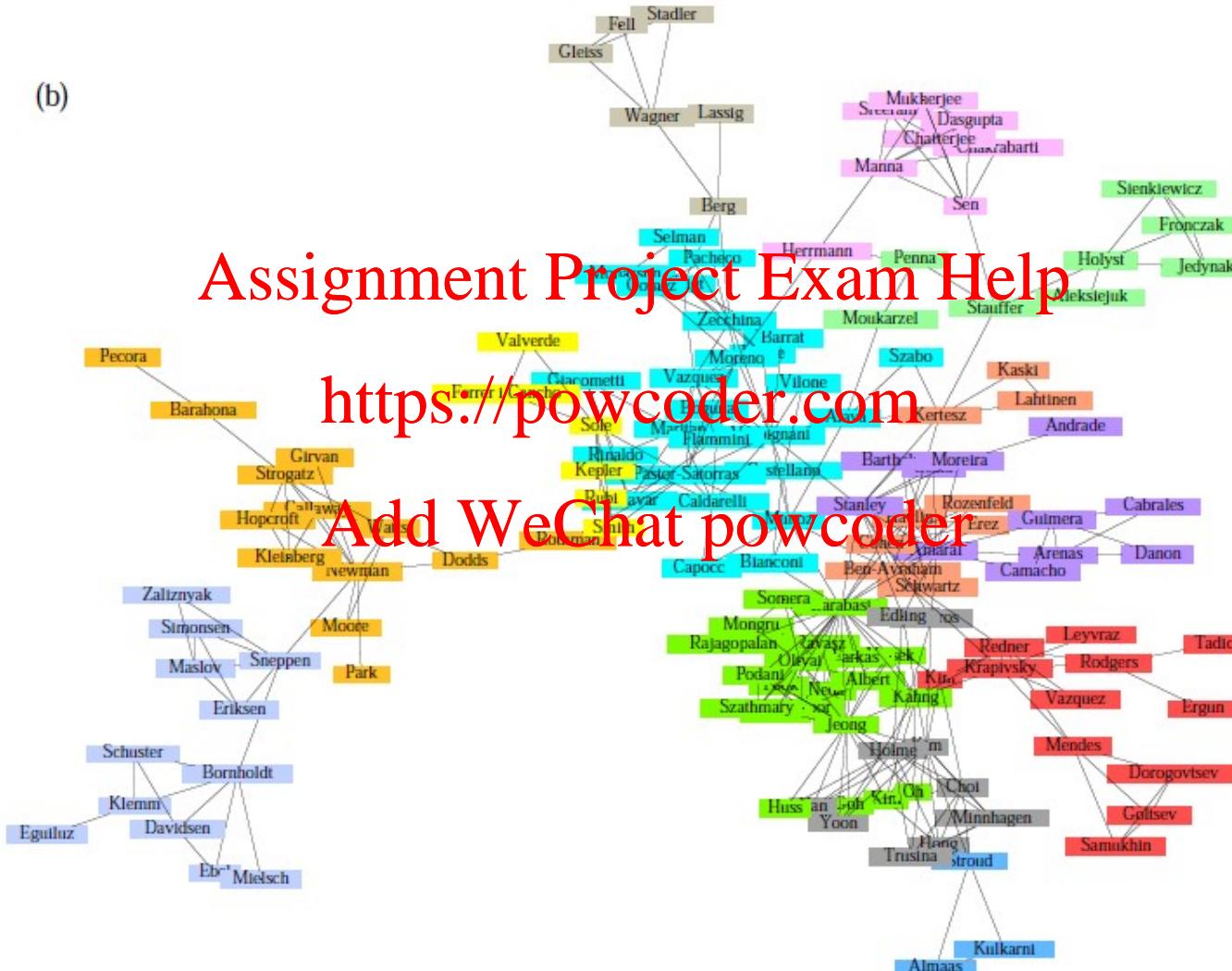


Hierarchical network decomposition:



Girvan-Newman

(b)



Problems of Girvan-Newman

- Computing betweenness is slow.
- No theoretical guarantees.
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- Not widely applied.
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Outline

- Introduction
- Social Networks as Graph
 - Basics of Graph Theory
 - Important properties of social networks as graphs
 - Analysis of a real-world social network
- Community detection algorithms
 - Girvan-Newman algorithm
 - Spectral clustering
- Conclusions

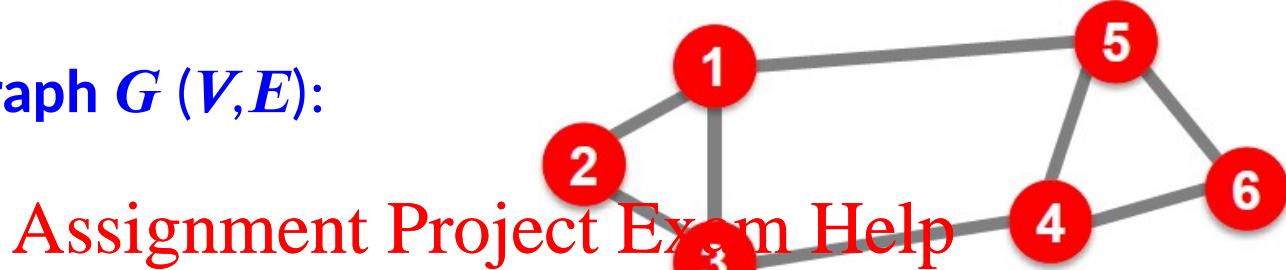
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Graph Partitioning

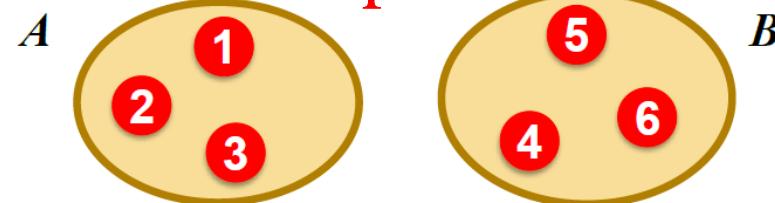
- Undirected graph $G(V,E)$:



- Bi-partitioning task: <https://powcoder.com>

- Divide vertices into two disjoint groups A, B

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- Questions:

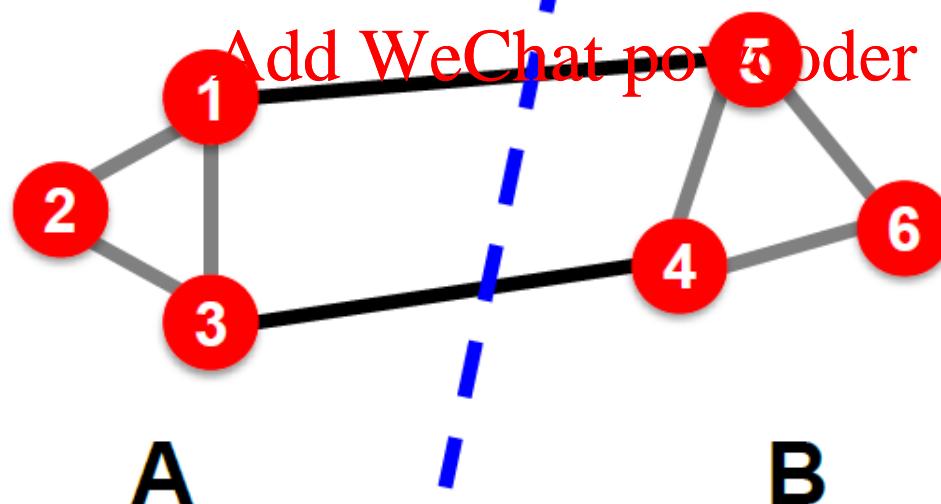
- How can we define a “good” partition of G ?
 - How can we efficiently identify such a partition?

Graph Partitioning

- **What makes a good partition?**

- Maximize the number of within-group connections
- Minimize the number of between-group connections

<https://powcoder.com>



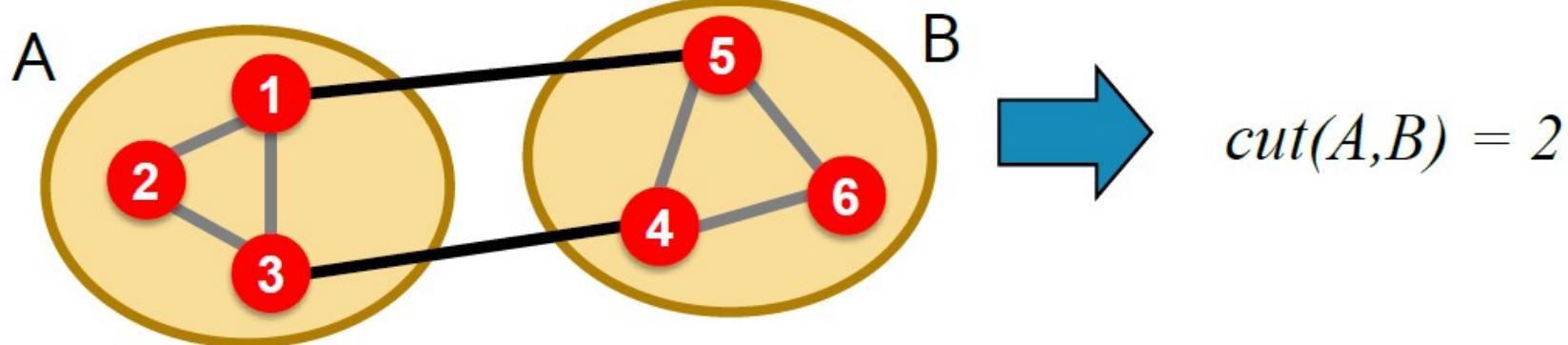
Graph Cuts

- Express partitioning objectives as a function of the “edge cut” of the partition
- Cut: Set of edges with only one vertex in a group:

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$$\text{cut}(A, B) = \sum_{i \in A, j \in B} w_{ij}$$

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$$\text{cut}(A, B) = 2$$

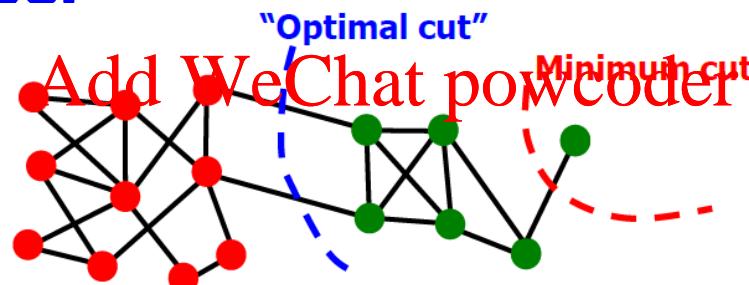
Graph Cut Criterion

- **Criterion: Minimum-cut**

- Minimize weight of connections between groups

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- **Degenerate case:** <https://powcoder.com>



- **Problem:**

- Only considers external cluster connections
- Does not consider internal cluster connectivity

Graph Cut Criterion

- **Criterion: Normalized-cut** [Shi-Malik, '97]
 - Connectivity between groups relative to the density of each group
<https://powcoder.com>
 - $\text{vol}(A)$: total weight of the edges with at least one endpoint in A
- **Why use this criterion?**
 - Produces more balanced partitions
- **How do we efficiently find a good partition?**
 - **Problem:** Computing optimal cut is NP-hard

Spectral Graph Partitioning

- A : adjacency matrix of undirected G
 - $A_{ij} = 1$ if (i, j) is an edge, else 0
- x is a vector in \mathbb{R}^n with components (x_1, \dots, x_n) . Think of it as a label/value of each node of G
- What is the meaning of Ax ?

[Assignment Project Exam Help](https://powcoder.com)
<https://powcoder.com>

$$\begin{bmatrix} a_{11} & \dots & a_{1n} \\ \vdots & & \vdots \\ a_{n1} & \dots & a_{nn} \end{bmatrix} \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix} = \begin{bmatrix} y_1 \\ \vdots \\ y_n \end{bmatrix} \quad y_i = \sum_{j=1}^n A_{ij} x_j = \sum_{(i,j) \in E} x_j$$

- Entry y_i is a sum of labels x_j of neighbors of i

What's the Meaning of Ax ?

$$\begin{bmatrix} a_{11} & \dots & a_{1n} \\ \vdots & & \vdots \\ a_{n1} & \dots & a_{nn} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} = \lambda \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix}$$

- ***j*th coordinate of Ax :**

- Sum of the x -values of neighbors of j
- Make this a new value at node j

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$$A \cdot x = \lambda \cdot x$$

- **Spectral Graph Theory** <https://powcoder.com>

- Analyze the “spectrum” of matrix representing G
- **Spectrum:** Eigenvectors x_i of a graph, ordered by the magnitude (strength) of their corresponding eigenvalues λ_i :

$$\Lambda = \{\lambda_1, \lambda_2, \dots, \lambda_n\}$$

$$\lambda_1 \leq \lambda_2 \leq \dots \leq \lambda_n$$

Example: d-Regular Graph

- Suppose all nodes in G have degree d and G is connected
 - What are some eigenvalues/vectors of G ? <https://powcoder.com>
 - $A \cdot x = \lambda \cdot x$ What is λ ? What x ?
 - Let's try: $x = (1, 1, \dots, 1)$
 - Then: $A \cdot x = d \cdot d \cdot \dots \cdot d = \lambda \cdot x$. So: $\lambda = d$
 - We found eigenpair of G : $x = (1, 1, \dots, 1)$, $\lambda = d$

Remember the meaning of $y = A \cdot x$:

$$y_i = \sum_{j=1}^n A_{ij} x_j = \sum_{(i,j) \in E} x_j$$

d is the Largest Eigenvalue of A

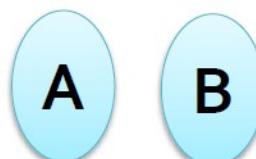
- G is d -regular connected, A is its adjacency matrix
- **Claim:**
 - d is largest eigenvalue of A
 - d has multiplicity of 1 (there is only 1 eigenvector associated with eigenvalue d)
- **Proof: Why no eigenvalue $d' > d$?**
 - To obtain d we needed $x_i = x_j$ for every i, j
 - This means $x = c \cdot (1, 1, \dots, 1)$ for some const. c
 - **Define:** S = nodes i with maximum possible value of x_i
 - Then consider some vector y which is not a multiple of vector $(1, \dots, 1)$. So not all nodes i (with labels y_i) are in S
 - Consider some node $j \in S$ and a neighbor $i \notin S$ then node j gets a value strictly less than d
 - So y is not eigenvector! And so d is the largest eigenvalue!

Example: Graph on 2 Components

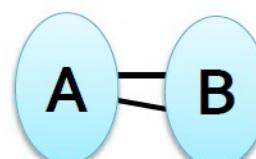
- What if G is not connected?
 - G has 2 components, each d-regular
- **What are some eigenvectors?**
 - x = Put all 1s on A and 0s on B or vice versa

$$x' = (\underbrace{1, \dots, 1}_{|A|}, \underbrace{0, \dots, 0}_{|B|}) \text{ then } A \cdot x' = (d, \dots, d, 0, \dots, 0)$$
$$x'' = (0, \dots, 0, \underbrace{1, \dots, 1}_{|A|}) \text{ then } A \cdot x'' = (0, \dots, 0, d, \dots, d)$$

- And so in both cases the corresponding
- A bit of intuition



$$\lambda_n = \lambda_{n-1}$$



$$\lambda_n - \lambda_{n-1} \approx 0$$

2nd largest eigenval.
 λ_{n-1} now has
value very close
to λ_n

More Intuition

- More Intuitions

A B
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$$\lambda_n = \lambda_{n-1}$$

$$\lambda_n - \lambda_{n-1} \approx 0$$

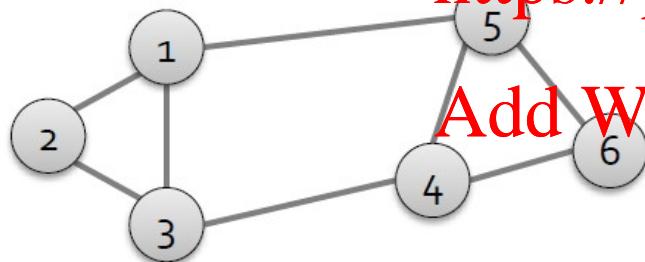
2nd largest eigenval.
 λ_{n-1} now has
value very close
to λ_n

- If the graph is connected (right example) then we already know that $x_n = (1, \dots, 1)$ is an eigenvector
 - Since eigenvectors are orthogonal then the components of x_{n-1} sum to 0.
 - So we can look at the eigenvector of the 2nd largest eigenvalue and declare nodes with positive label in A and negative label in B. **But there is still lots to sort out.**

Matrix Representation

- **Adjacency matrix (A):**

- $n \times n$ matrix
- $A = [a_{ij}]$, $a_{ij} = 1$ if edge between node i and j



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	1	2	3	4	5	6
1	0	1	1	0	1	0
2	1	0	1	0	0	0
3	1	1	0	1	0	0
4	0	0	1	0	1	1
5	1	0	0	1	0	1
6	0	0	0	1	1	0

- **Important properties:**

- Symmetric matrix
- Eigenvectors are real and orthogonal

Matrix Representation

- **Degree matrix (D):**

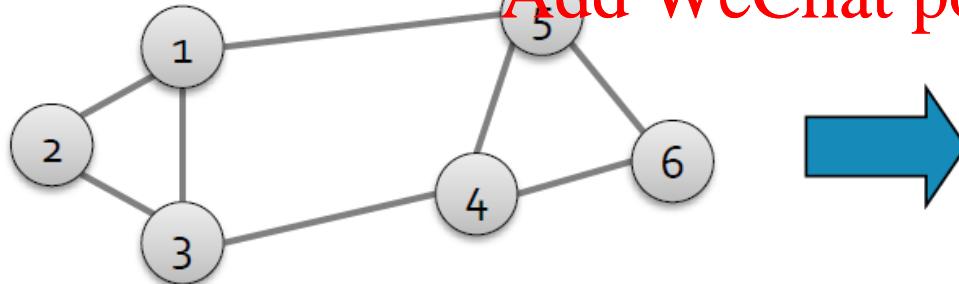
- $n \times n$ diagonal matrix

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- $D = [d_{ii}]$, d_{ii} = degree of node i

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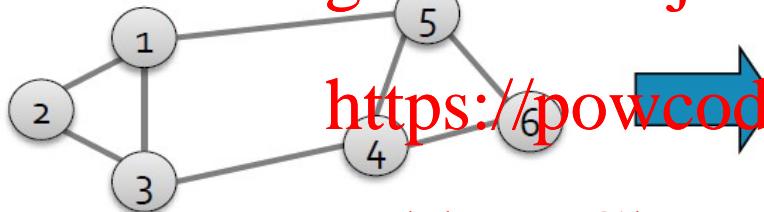


	1	2	3	4	5	6
1	0	0	0	0	0	0
2	0	2	0	0	0	0
3	0	0	3	0	0	0
4	0	0	0	3	0	0
5	0	0	0	0	3	0
6	0	0	0	0	0	2

Matrix Representation

- **Laplacian matrix (L):**

- $n \times n$ symmetric matrix



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	1	2	3	4	5	6
1	3	-1	-1	0	-1	0
2	-1	2	-1	0	0	0
3	-1	-1	3	-1	0	0
4	0	0	-1	3	-1	-1
5	-1	0	0	-1	3	-1
6	0	0	0	-1	-1	2

- **What is trivial Eigenpair?**

$$L = D - A$$

- $x=(1, \dots, 1)$ then $L \cdot x = 0$ and so $\lambda = \lambda_1 = 0$

- **Important properties:**

- **Eigenvalues** are non-negative real numbers
- **Eigenvectors** are real and orthogonal

Facts about Laplacian L

- (a) All eigenvalues are ≥ 0
- (b) for every x Assignment Project Exam Help
- (c) <https://powcoder.com>
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- That is, L is **positive semi-definite**

as Optimization Problem

- Fact: For symmetric matrix L :

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- What is the meaning of $\min x^T L x$ on G ?

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- $x^T L x = \sum_{i,j=1}^n L_{ij} x_i x_j = \sum_{i,j=1}^n (D_{ij} - A_{ij}) x_i x_j$
- $= \sum_i D_{ii} x_i^2 - \sum_{(i,j) \in E} 2x_i x_j$
- $= \sum_{(i,j) \in E} (x_i^2 + x_j^2 - 2x_i x_j)$

as Optimization Problem

- **What else do we know about x ?**

- x is unit vector:

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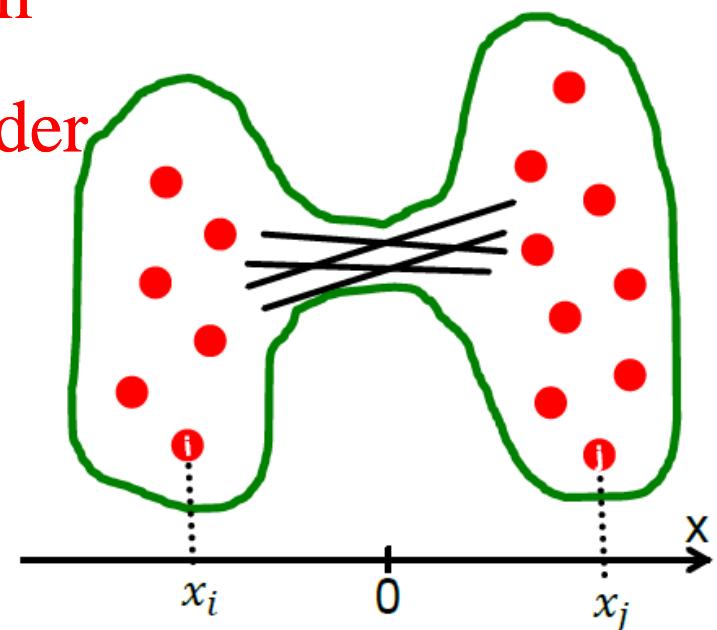
- x is orthogonal to 1st eigenvector $(1, \dots, 1)$ thus:

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- Remember

We want to assign values x_i to nodes i such that few edges cross 0. (we want x_i and x_j to subtract each other)



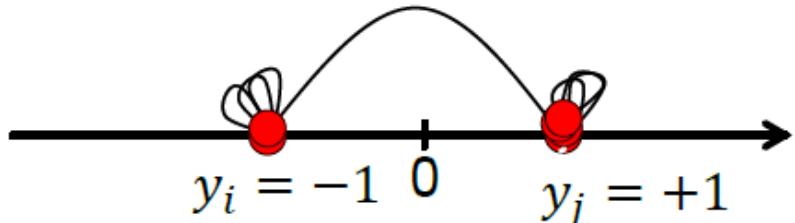
Finding Optimal Cut [Fielder'73]

- Back to finding the optimal cut
- Express partition (A, B) as a vector

$$y_i = \begin{cases} +1 & \text{if } i \in A \\ -1 & \text{if } i \in B \end{cases}$$

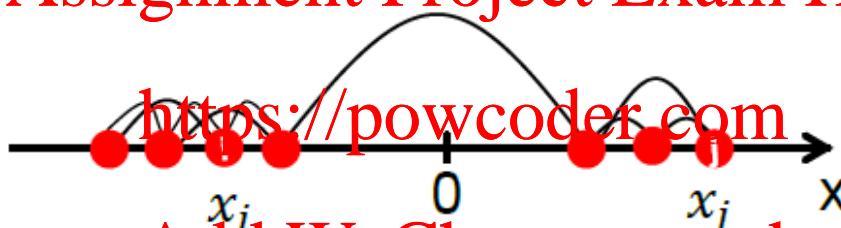
- We can minimize the cut of the partition by finding a non-trivial vector x that **minimizes**

Can't solve exactly. Let's relax y and allow it to take any real value.



Rayleigh Theorem

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- The minimum value of $f(y)$ is given by the 2nd smallest eigenvalue λ_2 of the Laplacian matrix L
- The optimal solution for $f(y)$ is given by the corresponding eigenvector x , referred as the **Fiedler vector**

Approx. Guarantee of Spectral (details)

- Suppose there is a partition of \mathbf{G} into \mathbf{A} and \mathbf{B} where $A \leq |B|$, s.t. then .
 - This is the approximation guarantee of the spectral clustering.
 - It says the cut spectral is at most 2 away from the optimal one of score .

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So Far..

- **How to define a “good” partition of a graph?**
 - Minimize a given graph cut criterion
- **How to efficiently identify such a partition?** <https://powcoder.com> Add WeChat powcoder
 - Approximate using information provided by the eigenvalues and eigenvectors of a graph
- **Spectral Clustering**

Spectral Clustering Algorithms

- Three basic stages:

1) Pre-processing

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- Construct a matrix representation of the graph

2) Decomposition

- <https://powcoder.com>
- Compute eigenvalues and eigenvectors of the matrix
 - Map each point to a lower-dimensional representation based on one or more eigenvectors

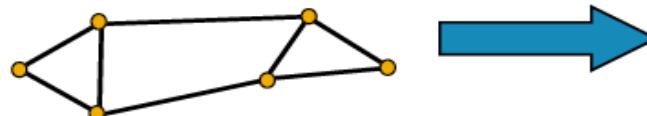
3) Grouping

- Assign points to two or more clusters, based on the new representation

Spectral Partition Algorithms

1) Pre-processing:

- Build Laplacian matrix L of the graph



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	1	2	3	4	5	6
1	3	-1	-1	0	-1	0
2	-1	2	-1	0	0	0
3	-1	-1	3	-1	0	0
4	0	0	-1	3	-1	-1
5	-1	0	0	-1	3	-1
6	0	0	0	-1	-1	2

2)

Decomposition:

- Find eigenvalues λ and eigenvectors x of the matrix L

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0.0
1.0
3.0
3.0
4.0
5.0

0.4	0.3	-0.5	-0.2	-0.4	-0.5
0.4	0.6	0.4	-0.4	0.4	0.0
0.4	-0.3	0.1	0.6	-0.4	0.5
0.4	-0.3	0.1	0.6	0.4	-0.5
0.4	-0.3	-0.5	-0.2	0.4	0.5
0.4	-0.6	0.4	-0.4	-0.4	0.0

- Map vertices to corresponding components of λ_2

1	0.3
2	0.6
3	0.3
4	-0.3
5	-0.3
6	-0.6



How do we now find the clusters?

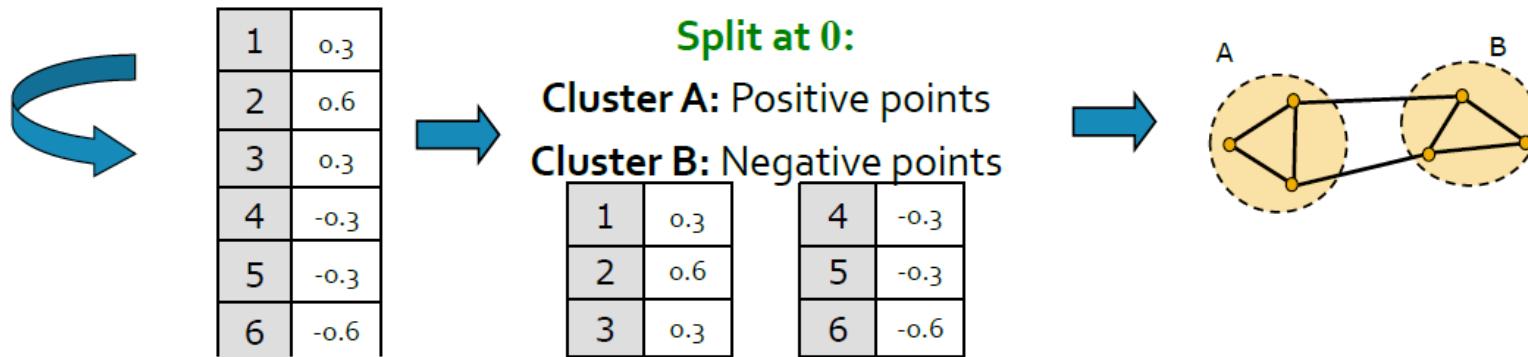
Spectral Partition Algorithms

- **3) Grouping**

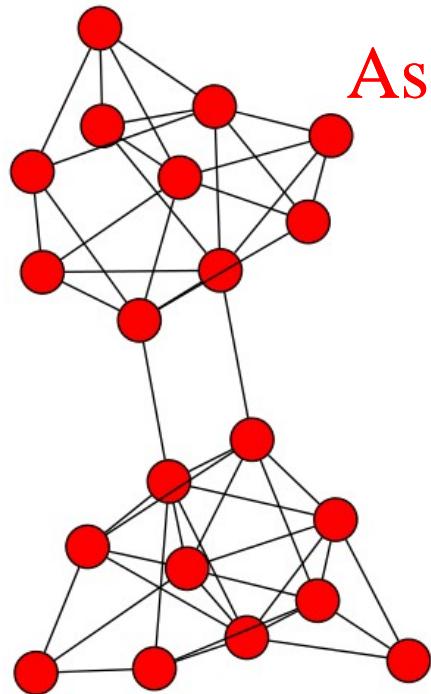
- Sort components of reduced 1-dimensional vector
- Identify clusters by splitting the sorted vector in two

- **How to choose assignment Project Exam Help**

- Naïve approaches:
 - Split at 0 or median value
- More expensive approaches:
 - Attempt to minimize normalized cut in 1-dimension (sweep over ordering of nodes induced by the eigenvector)



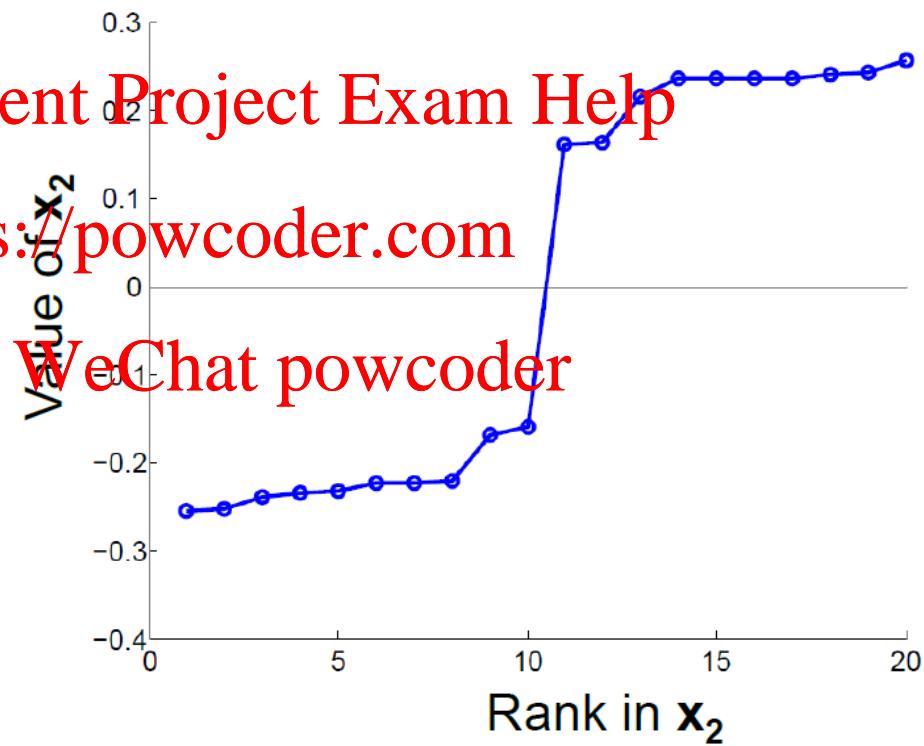
Example: Spectral Partitioning



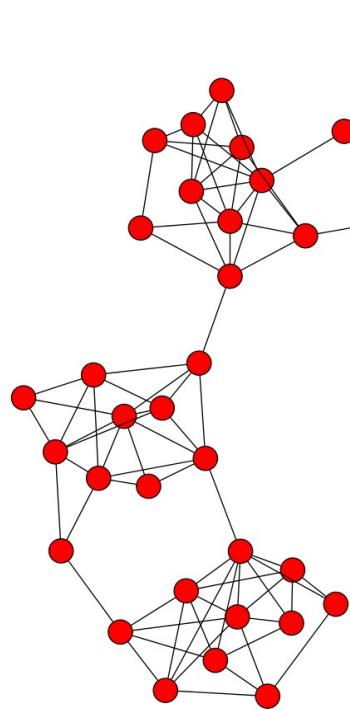
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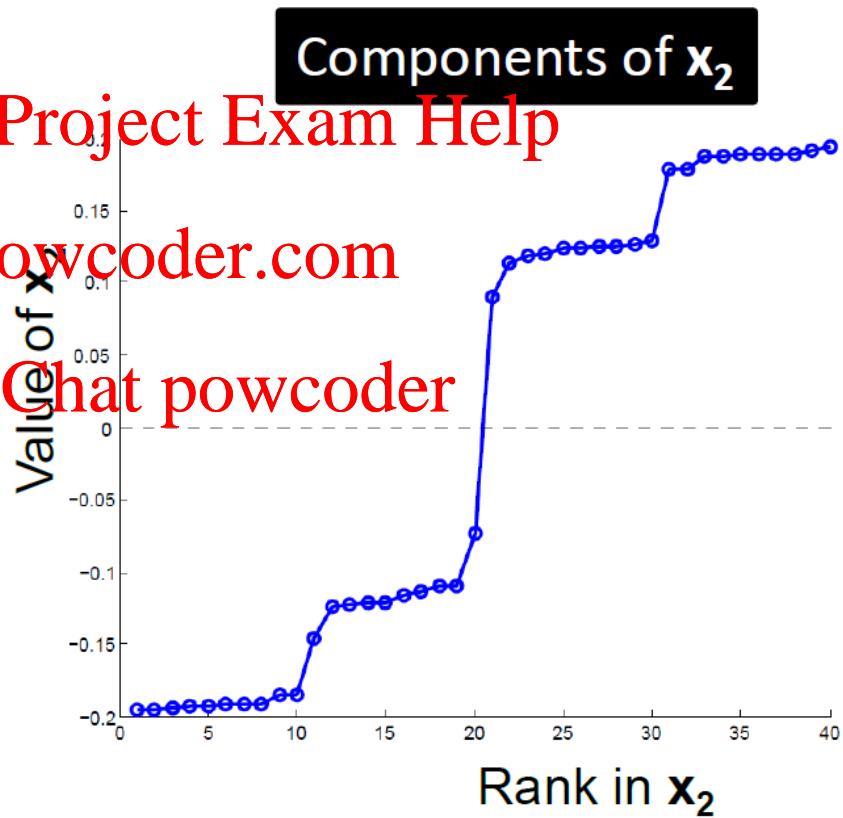
Example: Spectral Partitioning



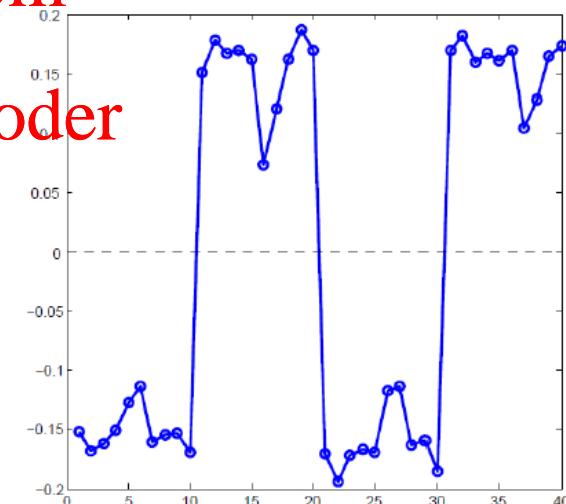
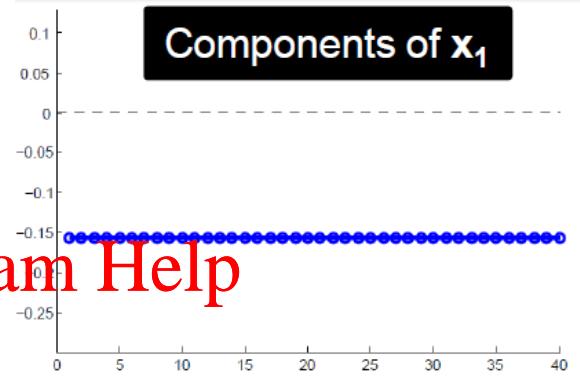
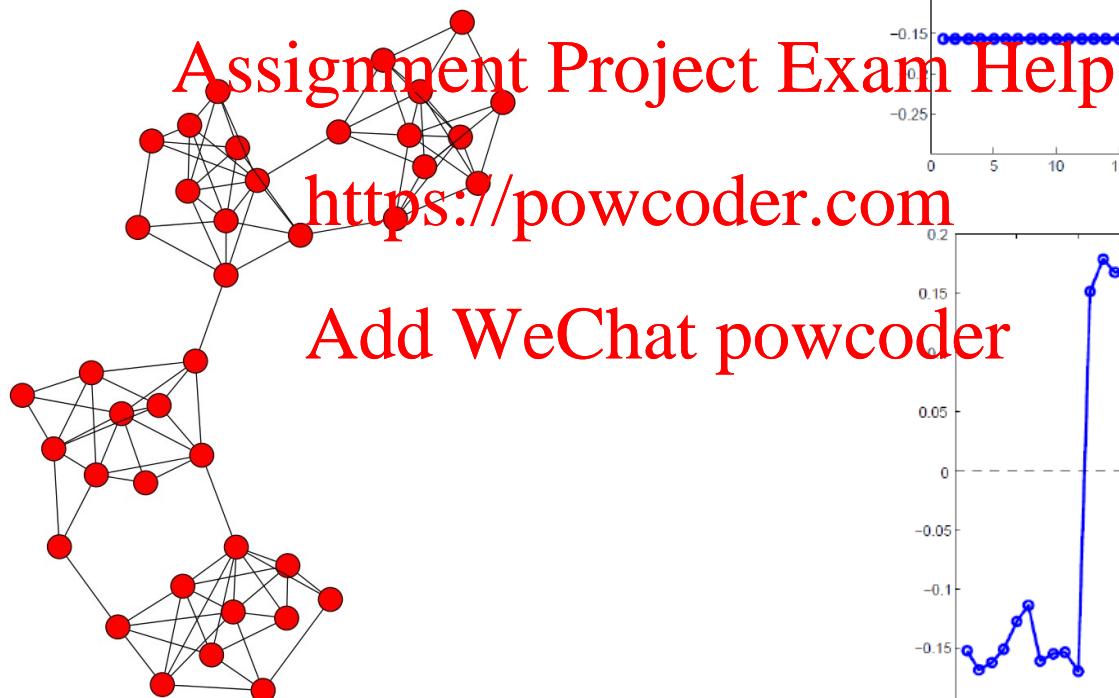
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Example: Spectral Partitioning



Components of x_3

Outline

- Introduction
- Social Networks as Graph
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Conclusions

- How do we reason about networks?
 - Empirical: Study network data to find organizational principles. How do we measure and quantify networks?
- Mathematical models:
 - Graph theory, statistical models
 - allow us to understand behaviors and distinguish surprising from expected phenomena
- Algorithms
 - for analyzing graphs
 - Hard computational challenges

Conclusions: This is Just a Beginning

Observations

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Small diameter,
Edge clustering

Scale-free

Strength of weak ties,
Core-periphery

Densification power law,
Shrinking diameters

Patterns of signed edge
creation

Viral Marketing, Blogosphere,
Memetracking

Models

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Small-world model,
Erdős-Renyi model

Preferential attachment,
Copying model

Kronecker Graphs

Microscopic model of
evolving networks

Structural balance,
Theory of status

Independent cascade model,
Game theoretic model

Algorithms

Decentralized search

PageRank, Hubs and
authorities

Community detection:
Girvan-Newman, Modularity

Link prediction,
Supervised random walks

Models for predicting
edge signs

Influence maximization,
Outbreak detection, LIM

One-Slide Takeaway

- Key concepts of social graphs
 - Undirected/Directed Graph/Adjacency Matrix
 - Shortest path/Diameter
 - Degree distribution, Path length, Clustering Coefficient
 - Real world graph
 - Degree distribution: high-skewed
 - Most path lengths are small
- Community detection algorithms
 - Purpose of community detection
 - Spectral clustering

Reference

- Adamic, Lada A., and Eytan Adar. "Friends and neighbors on the web." *Social networks* 25.3 (2003): 211-230.
- Kossinets, Gueorgi, and Duncan J. Watts. "Empirical analysis of an evolving social network." *Science* 311.5757 (2006): 88-90. APA
- Liben-Nowell, David, et al. "Geographic routing in social networks." *Proceedings of the National Academy of Sciences of the United States of America* 102.33 (2005): 11623-11628.
- Leskovec, Jure, and Eric Horvitz. "Planetary-scale views on a large instant-messaging network." *Proceedings of the 17th international conference on World Wide Web*. ACM, 2008.
- J. Shi and J. Malik, “Normalized cuts and image segmentation,” *IEEE Trans. on Pattern Analysis and Machine Intelligence*, 22:8 (2000), pp. 888–905.
- S. Fortunato, “Community detection in graphs,” *Physics Reports* 486:3–5(2010), pp. 75–174.
- M. Girvan and M.E.J. Newman, “Community structure in social and biological networks,” *Proc. Natl. Acad. Sci.* 99 (2002), pp. 7821–7826.

Reference

- U. von Luxburg, “A tutorial on spectral clustering,” Statistics and Computing bf17:4 (2007), 2007, pp. 395–416.
- J. Leskovec, K.J. Lang, A. Dasgupta, and M.W. Mahoney, “Community structure in large networks: natural cluster sizes and the absence of large well-defined clusters,” <http://arxiv.org/abs/0810.1355>.
- L. Backstrom and J. Leskovec, “Supervised random walks: predicting and recommending links in social networks,” Proc. Fourth ACM Intl. Conf. on Web Search and Data Mining (2011), pp. 635–644.
- Predicting Positive and Negative Links in Online Social Networks by J. Leskovec, D. Huttenlocher, J. Kleinberg. ACM WWW International Conference on World Wide Web (WWW), 2010.
- Learning to Discover Social Circles in Ego Networks by J. McAuley, J. Leskovec. Neural Information Processing Systems (NIPS), 2012.
- Defining and Evaluating Network Communities based on Ground-truth by J. Yang, J. Leskovec. IEEE International Conference On Data Mining (ICDM), 2012.
- The Life and Death of Online Groups: Predicting Group Growth and Longevity by S. Kairam, D. Wang, J. Leskovec. ACM International Conference on Web Search and Data Mining (WSDM), 2012

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In-class Practice 1

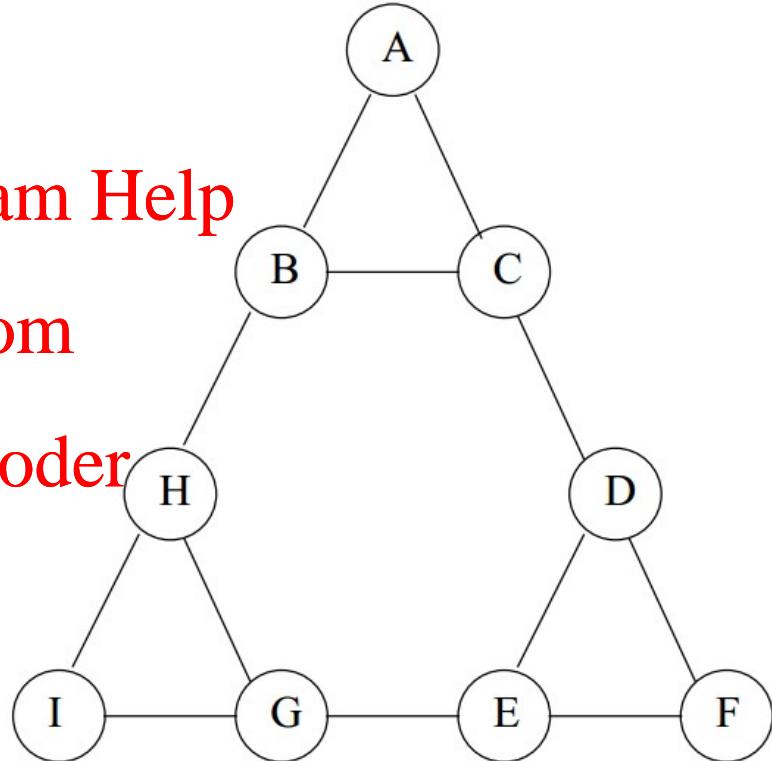
- For the graph on the right, compute:

- The adjacency matrix
- The degree matrix
- The Laplacian matrix

Assignment Project Exam Help

<https://powcoder.com>

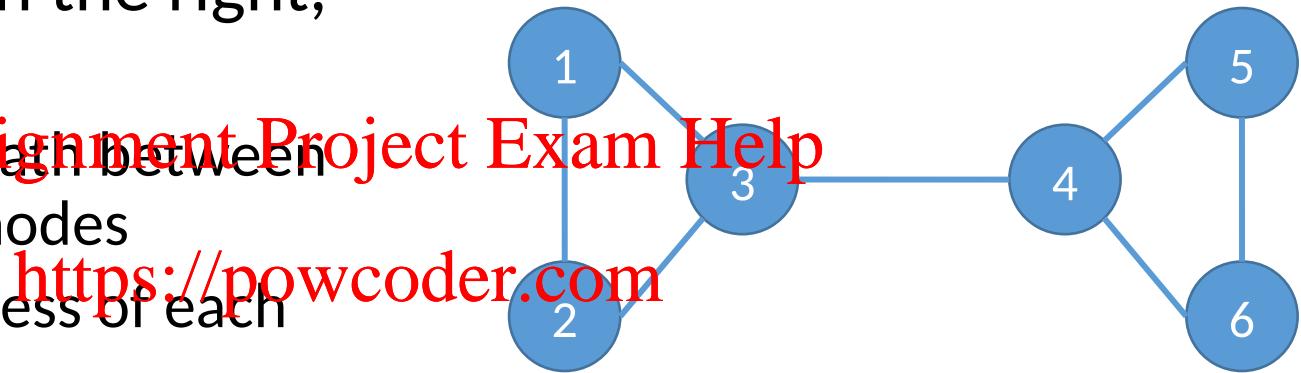
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In-class Practice 2

- For the graph on the right, compute:

- The shortest path between each pairs of nodes
- The betweenness of each edge



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