Social Network Analysis: Introduction to the Course

Even Semester of Academic Year 2022-23
Instructors: 1) Sankita Patel 2) Suhani Chauhan

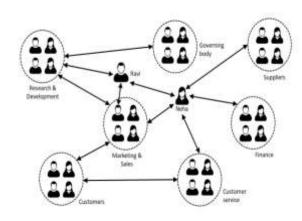
Slide Credits: Teaching Material on Social Network Analysis by Tanmoy Chakraborty, Wiley, 2021

Motivation, Curriculum, Teaching Scheme and Course Outcome

Books to refer

- Social Network Analysis by Tanmoy Chakraborty, Wiley (Slide credits to this book)
- Network Science by Barabasi, Cambridge University Press
- Slide Credits: Teaching Material on Social Network Analysis by Tanmoy Chakraborty, Wiley, 2021

What is Social Network Analysis?



Network:

An abstract representation of relations among entities

Social Network:

A simplified representation of the social structure characterized by actors and ties

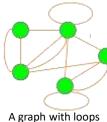
Social Network Analysis:

The application of networks and graph theory to analyze the relations present in a society

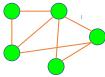
Network: Definition

- A network, also referred to as a *graph*, is defined as an ordered pair G(V, E), where V is a set of nodes (also referred to as vertices or entities), and E is a set of edges (also referred to as links or relations) joining the nodes.
- ☐ Depending on the nature of application, the above definition may be revised or augmented, as follows:
 - the nature of edges may vary undirected (also called symmetric, or reversible) edges, directed (also called asymmetric, or irreversible) edges, or hyperedges, etc.
 - both the nodes and/or the links are associated with one or more attributes/features like weights, timestamps, textual features, etc.
- An edge in a graph may have same node as end nodes. Such edges of a graph are called self loops (or, simply, loops).
- A graph may have more than one edge joining a pair of nodes. Such edges are called parallel edges.

Network: Definition (contd...)

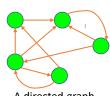


A graph with loops and parallel edges

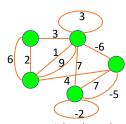


A simple graph

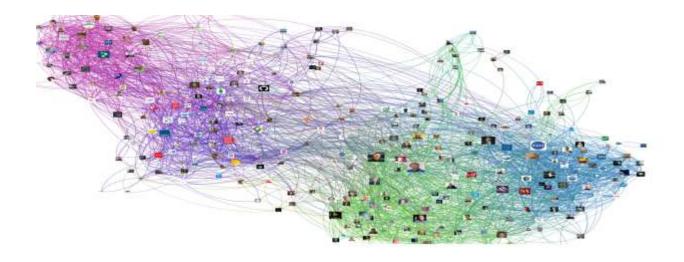
- A graph having neither self loops nor parallel edges are called a simple graph.
- ☐ A graph having directed edges (i.e. links having a direction) is called a directed graph. Directed edges are also referred to as arcs.
- A graph having no direction in its edges is called an undirected graph.
- ☐ A graph having weights associated with its edges are called a weighted graph. A weighted graph can be directed as well as undirected.
- A graph having its nodes and/or edges attributed with feature values is called an attributed graph.



A directed graph



A weighted graph



A sample of Twitter follower-followee network

(image source: http://allthingsgraphed.com/2014/11/02/twitter-friends-network/)

Social Network Analysis: Key Features

- Required Knowledge Domains
 - $lue{}$ Sociology $lue{}$ Psychology $lue{}$ Mathematics $lue{}$ Statistics $lue{}$ Computer Science
- Study Benefits
- ☐ To know the way social interactions influence a network
- ☐ To learn how the information flows inside a network
- ☐ To characterize roles of the individuals in a network
- ☐ To characterize communities inside a network
- ☐ To characterize the evolution of a network

Key Application Areas

■ Healthcare

- Combating Epidemics
- Mass Vaccination

□ Social Media & E-Commerce

- ❖ Friend & Follow Recommendation
- **❖** Know Your Customers
- ❖ Recommendation & viral marketing

■Web & Cyberspace

- **❖**Search engine optimization
- ❖ Malware detection
- **❖**Spam detection

☐ Police & Military

- ❖ Fighting cyber crimes
- ❖ Fighting terrorism
- ❖ Network-centric warfare

☐ Scientific Research & Academic Collaboration

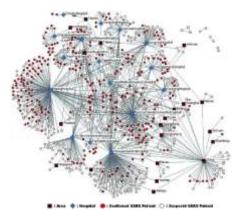
- Ranking scientific publications
- Ranking scientific authors
- Ranking publication venues

■ Miscellaneous

- **❖**Computer-supported collaborative learning
- Complex project management

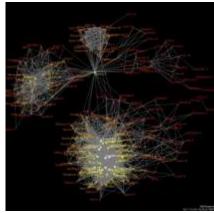
SNA Applications: Healthcare

□ Modeling the spread of infectious diseases
 □ Contact tracing during epidemic outbreak to identify possible patients
 □ Identify and isolate super-spreaders
 □ Planning lockdown schedule
 □ Identify vulnerable population during vaccination
 □ Planning vaccination schedule, etc.



2003 SARS contact Network in Taiwan https://doi.org/10.1007/978-1-4419-6892-0 15

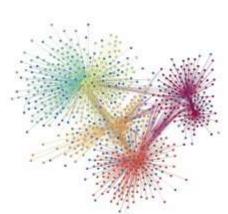
SNA Applications: Social Media



- ☐ Identifying friendship structures in online social media
- ☐ Recommending friends, and groups, or pages to follow
- ☐ Identifying information propagation patterns in social networks

An Example Map of a Facebook Friendship Network https://mathconceptions.wordpress.com/2012/01/16/application-snippet-friendship-and-influence-in-social-networks/

SNA Applications: E-Commerce



- ☐ Customer profiling to Know the customers
- ☐ Product/Service recommendation based on customer profile
- ☐ Instigating viral marketing by pinpointing influential players
- "People like you buy", "Frequently bought with this", or "Frequently browsed", "Trending" are common buzzwords



Influential Communities in Social Network

https://towardsdatascience.com/influential-communities-in-social-network-simplified-fe5050dbe5a4

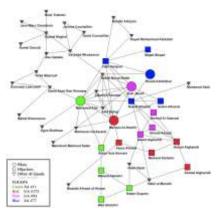
SNA Applications: Malware & SPAM Detection



- ☐ Modeling malwares using Graph representations
- ☐ System call graphs, malware similarity network, etc. are typical examples
- ☐ These graphs are large due to volume of networks
- ☐ Malware detection through network analysis
- □SPAM detection

https://searchsecurity.techtarget.com/definition/malware

SNA Applications: Cybercrimes & Terrorism



9/11 Terrorist Network http://www.orgnet.com/hijackers.html

- Online fraud, fake news propagation, cyber bullying/trolling, sharing pornographic materials, etc. rising with growth of social media
- ☐ Terrorists often use social media to communicate as well as to brainwash innocent people
- ☐ These people often span across countries and use untraceable communication devices
- ☐ Tracking cyber criminals in conventional methods are difficult due to user anonymity, fake accounts, lack of cyber laws, etc.
- ☐ Social network analysis techniques help nabbing these criminals

SNA Applications: Network Centric Warfare



Network Centric Warfare

- ☐ Rising popularity of social network analysis influence the military doctrines
- Various military think-tanks proposed for networkcentric warfare principle using social network analysis methods
- ☐ It is claimed that Saddam Hussein has been captured from his hideout exploiting network analysis techniques
- ☐ It is also claimed US Navy Seal Team Six assassinated Osama Bin Laden by tracking his secret hiding location in a similar manner

http://www.indiandefencereview.com/news/network-centric-warfare/

SNA Applications: Scientific Research & Academic Collaboration



https://library.bu.edu/citedreferences

- Scientific authors cite (refer) the works of other authors in the publications to authenticate their claims
- ☐ Finding the dynamics of these citations attracted social scientists
- Various networks of scholarly articles may be formed exploiting this relationship
 - ☐ Paper-paper citation network
 - ☐ Paper-paper co-citation network, etc.
- Various popular metrics are outcome of analysing these networks.
 - ☐ Publication related: H-index, i-10 index, g-index, etc.
 - ☐ Venue related: impact factor, CORE rank, etc.

SNA Applications: Scientific Research & Academic Collaboration



https://scholarlykitchen.sspnet.org/2017/04/07/updated-figures-scale-nature-researchers-use-scholarly-collaboration-networks/

- ☐ Scientific authors collaborate with one another to improve research quality
- Various scholar networks may be formed using these relationships
 - ☐ Author collaboration network
 - Author citation network
 - ☐ Author co-citation network, etc.
- ☐ Information retrieved from these networks may be used to measure authors' research quality

SNA Applications: Computer-supported Collaborative Learning

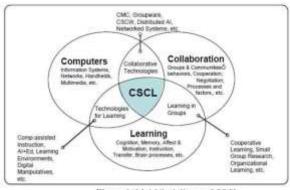


Figure 1: Multidisciplinary of CSCL

- Pedagogical process of observation where students learn progressively through active group interaction using ICT
- ☐ SNA techniques used to extract relationship between various actors (human and non-human) of CSCL
- Study insights are used to improve the students' learning outcome and user experience

https://www.semanticscholar.org/paper/The-Characteristics-of-the-Computer-Supported-a-on-Hashim-Ismail/42176e6bf76dd15a2c9874e6fa8696e153a3f554

SNA Applications: Organizational Network Analysis

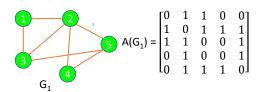


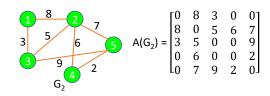
https://www.i4cp.com/productivity-blog/what-organizationalnetwork-analysis-is-and-how-it-benefits-companies

- ☐ Classic (linear) organization charts do not effectively describe the real social network of an organization
- ☐ The informal networks, historically considered an annoying problem and an indicator of undisciplined attitude of workers, carry huge potential if applied suitably
- ONA provides information on how to improve performance in the organization
- ONA represents the complete set of real relationships between the players
 - who is in touch with whom
 - specific features of each player
 - ☐ Type and intensity of relationship, etc.

Network Representation: Adjacency Matrix

 \square An adjacency matrix $A=\left(a_{ij}\right)$ for a graph G(V,E) is a square matrix of dimension $|V|\times |V|$ such that each element a_{ij} of A indicates the existence of an edge between the node v_i and node v_j (also the weight of the corresponding edge in case of a weighted graph) in G.



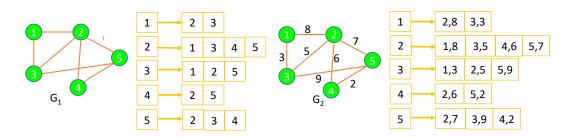


Network Representation: Adjacency Matrix (Cont...)

advantages □ Easy to implement and follow □ Addition, and removal of an edge require O(1) time □ Query to an edge require O(1) time disadvantages □ Consumes $O(|V|^2)$ storage space, even if the graph is sparse □ Addition or removal of a node require $O(|V|^2)$ time

Network Representation: Adjacency List

 \square An adjacency list for a graph G(V,E) is a collection of unordered lists such that each node correspond to a list from the collection that indicates the set of neighbours of the node. Every entry in an adjacency list A_i for node v_i in the graph is a node adjacent to node v_i



Network Representation: Adjacency List (Cont...)

advantages

- \square Require O(|V|+|E|) space
- □Insertion of vertex and edge require O(1) time
- \square Removal of vertex require O(|V|+|E|) time

disadvantages

- \square Removal of edge require O(|E|) time
- \square Query to an edge require O(|V|) time

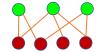
Network Types: Link-centric View Unipartite Network



- □ Consists of a vertex set **V** and an edge set **E**. There is no restriction on the formation of edges between nodes of the network
- □ Example: An organizational LAN, where nodes are the devices, and edges are the local area links.
- ☐Used to model the situation when links can join any pair of nodes of the network

Network Types: Link-centric View Bipartite Network

□Consists of a vertex set *V* that is divided into two sets *V1* and *V2* that are disjoint and independent. Each edge of the network connects a vertex in *V1* to another vertex in *V2*



- ☐ Example: An e-commerce user-product network. One part consists of the users, the other part consists of the products, the links are based on the basis of who bought what.
- ☐Generalization of Bipartite network is n-partite networks, where the vertex set is partitioned into n number of part, an edges join a node from one part with a node from another part.

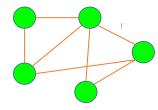
Network Types: Link-centric View Signed Networks

□Consists of a set of nodes V, a set of edges E, and a function $f: E \rightarrow \{+, -\}$ that assigns each edge a positive or a negative sign



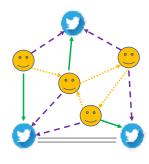
- □ Example: Consider a social media website that allows users to tag other users as friends or foes. The positive edges are friendship links and negative links are between foes.
- □studied specifically in the context of balance and status theory which determines the stability or existence of certain types of structural patterns in a network.

Network Types: Node and Link-centric View: Homogeneous Network



- □Consists of a set of nodes V, all of which are of same type, and a set of edges *E*, all of which are of same type
- □ Example: Follower-Followee network of any kind.
 - ❖ Nodes are the users of the platform (all nodes are of same type),
 - Links are the follower-followee link between these nodes (all links are of same type)

Network Types: Node and Link-centric View: Heterogeneous Network

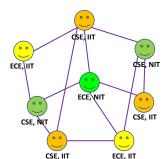


 \square Consists of a set of nodes V and a set of edges E, and two associated mapping functions, f_v and f_e , for nodes and edges, respectively. f_v maps a node to a node type and f_e maps an edge to an edge type

- $oldsymbol{\square}$ Example: We consider a specific instance of Twitter network like the figure.
 - ☐ Two types of nodes:
 - ☐ Twitter Users
 - ☐ Tweet Posts
 - ☐ Four types of edges representing four types of relations between these nodes:
 - ☐ .Posted-by (User Post: Directed links)
 - ☐ Followed-by (User User: Directed links)
 - ☐ Similar (Post Post: Undirected links)
 - ☐ Retweet (User Post: Directed links)

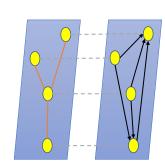
Network Types: Node and Link-centric View: Attributed Network

- \square Consists of a set of nodes V and a set of edges E, and two associated mapping functions, f_v and f_e , for nodes and edges, respectively. f_v maps a node to a node attribute vector and and f_e maps an edge to an edge attribute vector
- ☐ Example: We consider a specific instance of Facebook network like the figure
 - □ Nodes are some Facebook users
 - ☐ Edges are given by Facebook friendship relationship between these users
 - □Node attributes are the users' academic affiliations
 - ☐ There is no edge attribute in this network
- ☐ The example is a node-attributed network



Network Types: Node and Link-centric View: Multidimensional Network

- ☐ A special type of multilayer network where each layer represents a particular type of relationship among nodes
- ☐ Example 1: A special instance of Twitter network:
 - ☐ Nodes are Twitter users in both layer
 - □ Layer 1 edges: user user similarity links (based on mutual interests) Undirected links
 - □Layer 2 edges: user user follower-followee links Directed links
- □ Note: In this example, each layer is node homogeneous.



Network Types: Node and Link-centric View: Multidimensional Network (Cont...)

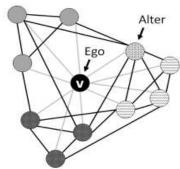
- □ Example 2: Customer product relationships as a multidimensional complex network system
 - ☐ Layer 1: One type of nodes, one type of edges:
 - Nodes: Customers
 - ☐ Edges: Customer Social Interaction Undirected links
 - ☐ Layer 2: Two types of nodes, two types of nodes
 - ☐ Nodes:
 - ☐ High performance Cars,
 - ☐ Fuel efficient cars
 - ☐ Edges:
 - ☐ Car feature association Undirected links
 - ☐ Car preference association Directed links
 - ☐ Inter-layer edges: two types of edges
 - ☐ Customer Car: Purchase decision Undirected links
 - ☐ Customer Car: Consideration decision Undirected links



https://www.semanticscholar.org/paper/Modeling-customer-preferences-using-network-in-Wang-Chen/539c7f0632041903521b8cbc42eabd27b8844673

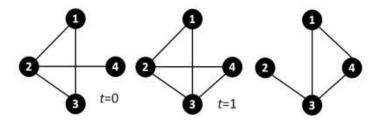
Network Types: Local View: Ego-centric Network

- \square A network of the form G(V, E, u), that corresponds to a node $u \in V$, usually known as the 'ego', and consists of the node u as the central node, the nodes that are connected directly to the node u, usually known as the 'alters', and the induced subgraph for the same.
 - ☐ Example: A subgraph of a Facebook Friendship Network:
 - ☐ Ego node corresponds to a user,
 - ☐ Alter nodes are his Facebook friends from different capacities and affiliations.



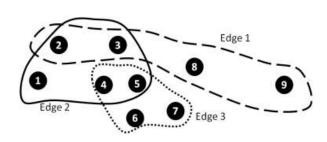
Network Types: Temporal View: Time-varying Network

- \square Consists of a set of nodes V and a set of edges E where each edge $e_{i,j} \in E$ is represented by a three- tuple $e_{i,j} = \{v_i, v_j, t_{ij}\}$. Here, v_i and v_j are two end-points, and t_{ij} indicates the persistence duration of the edge e_{ij}
- ☐ Example: Person-to-person communication network over a span of time. The visible components are snapshots of the network at different time instances.



Network Types: Generalized View: Hypergraph

- \square Defined by a set of nodes V and a set of edge or hyperedges E, where each hyperedge e connects multiple nodes of the hypergraph
- ☐ Example: A special representation of Coauthoship Network:
 - ❖ Nodes are authors
 - Papers are hyperedges connecting the coauthors of the paper



Popular Real-world Networks

■ Social Network

- ❖ Telephone call network
- ❖ Email message network
- ❖ Film actor collaboration network
- ❖ Academic co-authorship network

☐ Biological Network

- Protein-protein interaction networks
- ❖ Genetic regulatory networks
- ❖ Neural networks
- ❖ Metabolic networks
- ❖ Food Web
- Cell signalling networks

☐ Information Network

- ❖ World Wide Web (WWW)
- Citation network

☐ Technological Network

- ❖ Electric power grids
- ❖ Networks of airline routes
- ❖ Network of Railway Routes
- ❖ Electronic circuits
- Delivery networks of postoffice/Courier
- ❖ The Internet

☐ Language Network

Network formed by using the persons speaking a particular language

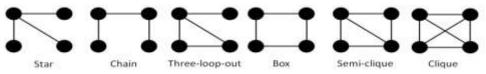
Levels of Social Network Analysis: Microscopic Level

■We begin by analyzing how a pair of nodes interacts and gradually trace the interactions at the group level or subgraph level.	
□ Dyadic level → Interaction patterns among two nodes	
☐ Examined properties: homophily, reciprocity, social equality, mutuality, etc.	
☐ Derived global statistics: assortativity, mixing coefficient, etc.	
□ Triadic level → Interaction patterns among three nodes	
□Examined properties: triadic closure	
☐ Derived network properties: Clustering Coefficient, local bridges, etc.	

□ Ego-centric circles → Interaction pattern between ego node with its alters

Levels of Social Network Analysis: Mesoscopic Level

- ☐ Mesoscopic analysis is an intermediary between microscopic and macroscopic analyses, which mostly deals with a subset of the entire population.
- □ Communities → Formed due to frequent interactions among homogeneous nodes in a network
 - ❖ Within a community, the nodes exhibit a particular kind of dynamicity
 - Across communities, the dynamic behaviour differs
- □ Network Motifs → Subgraphs that repeat themselves frequently within or across a network
 - ☐ Highly effective in capturing functional properties in a network



Undirected motifs with size 4 and their names

Levels of Social Network Analysis: Macroscopic Level

- ☐ At macroscopic level, we deal with the entire network as a whole and try to understand the micro-level dynamics by exploring the overall graph property.
- ☐ Features of Interest:
 - ■Connectedness,
 - ☐ Diameter or Average path length,
 - ☐ Degree Distribution,
 - ☐ Edge Density, etc.
- ■Example:
 - □We find that the diameter of a network is too small ⇒ network may look like a star, or a clique
 - □We further find that overall edge density is too high ⇒ network looks like a clique

Graph Visualization Tools

■ Web-based tools

- Pollinode: https://www.polinode.com (Non-open source application)
- NodeGoat: https://nodegoat.net (Non-open source application)
- Linkage: https://linkage.fr (Open source application)
- EchoDemo: https://osome.iuni.iu.edu/demos/echo (Nonopen source application)
- Palladio: https://hdlab.stanford.edu/palladio (Open source application)

■ Standalone tools

- NDlib-Viz: https://ndlib.readthedocs.io/en/latest/viz/ndlibviz.html (Open source application)
- CytoScape: https://cytoscape.org (Open source application)
- Gephi: https://gephi.org (Open source application)
- Vizster: http://vis.stanford.edu/jheer/projects/vizster source application)
- SparklingGraph: https://sparkling-graph.github.io (Open source application)