

Social Computing: Introduction

Mainack Mondal

CS 60017
Autumn 2021



- Course logistics
- Social networks in the offline and online worlds
- Research challenges on social networks / social media

Instructors

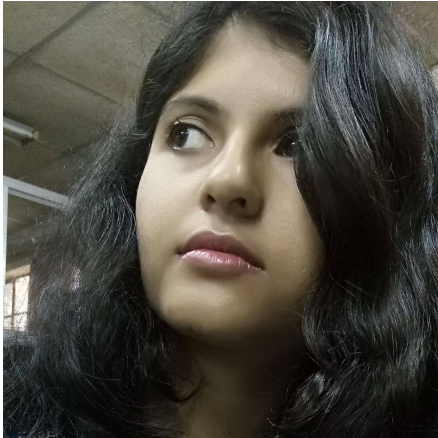


- **Mainack Mondal:** social systems, usable security and privacy, system security and privacy
 - Office: CSE 316
 - Also teaching Usable Security and Privacy

TAs (1)



Paramita Das



Rima Hazra

TAs (2)



Sayantan Adak



Jagdish Singh Samant

TAs (3)



Ankan Mullick

Website

- <https://cse.iitkgp.ac.in/~mainack/courses/2021-autumn/socialcomp/>
 - Read the instructions carefully
 - Check the announcements
 - Regularly visit

Course timings

- Credit : 3 – 0 – 0
- Wednesday 11:00 am - 11:55 am
- Thursday 12:00 noon - 12:55 pm
- Friday 8:00 am – 9 am

Mode of teaching

- Live lectures via zoom
 - Zoom link will be announced in MS classroom
 - Might not be recorded, please attend classes
- Pre-recorded lectures
 - We will upload the recorded lectures via MS teams

CSE Moodle

- All submissions will be via CSE moodle unless otherwise stated
 - CSE Moodle
 - <https://moodlecse.iitkgp.ac.in/moodle/login/index.php>
 - Course name: Social Computing (Autumn 2021-22)
 - Enrollment key: STU60017 (checking with the admins)

Course evaluation: Test / viva

- Three tests / Viva (40%)
 - Syllabus : Everything until that point
 - Dates are in the webpage

Course evaluation: Assignment

- 40% of the evaluation
 - 3/4 in number
 - Programming based
 - Will test how well can you apply the knowledge acquired in the class

Course evaluation: Research papers

- 20% of the evaluation
- Requirements:
 - Present research papers (in class)
 - Present in groups
 - The covered papers will be part of the test
 - More details will be announced shortly

Course logistics

- Questions?

- Course logistics
- Social networks in the offline and online worlds
- Research challenges on social networks / social media

Social networks in off-line world

- Social networks studied for several decades
 - Friendship networks among students of a school, members of a club, ...
 - Collaboration networks among scientists, movie actors, ...
 - Citation networks: scientists / papers referring to other scientists / papers

What are possible rules? Ideas

- Proximity – how close the person is sitting or living
- Common interest
- Organization
- Common friends – mutual friends
- Common experiences

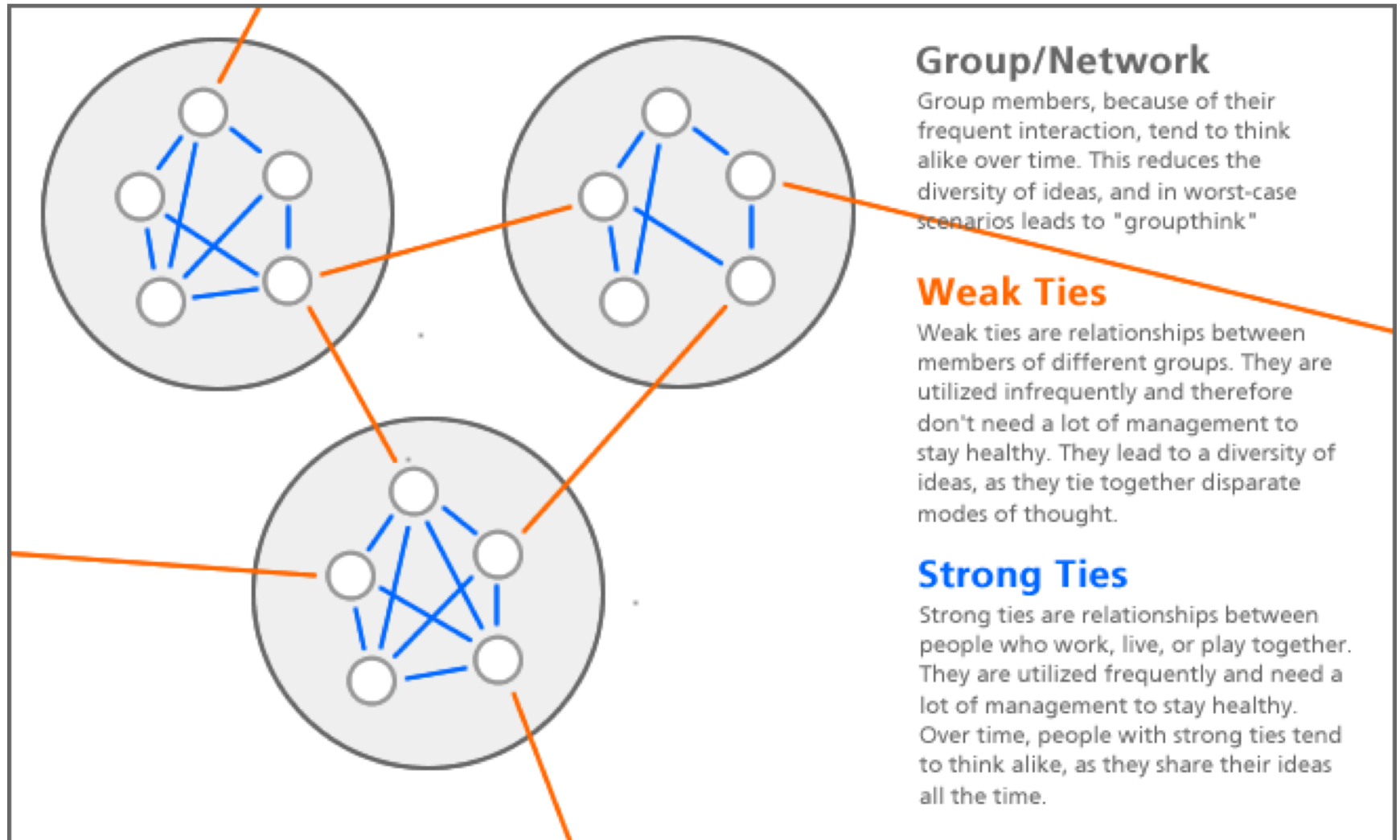
Sociological theories

- Several sociological theories developed
 - Homophily – birds of a feather flock together
 - Six degrees of separation - Milgram's experiments (1967)
 - Strength of weak ties (1973)
 - Spread of epidemics / conventions / news / rumors

Milgram's experiment in 1967

- Sent packets to people in Omaha, Nebraska and Wichita, Kansas
 - You need to get the packets to a specific person in Boston
 - If you know the recipient, send the packet directly to him
 - If not, think of a friend you know, who is likely to be closer to the recipient in Boston; sign your name to a roster, and send the packet to your friend
- Boston recipient examined the roster and saw how many steps it took for the letter to arrive
- 64 letters reached recipient, **average number of links: between 5 and 6**

Strength of ties

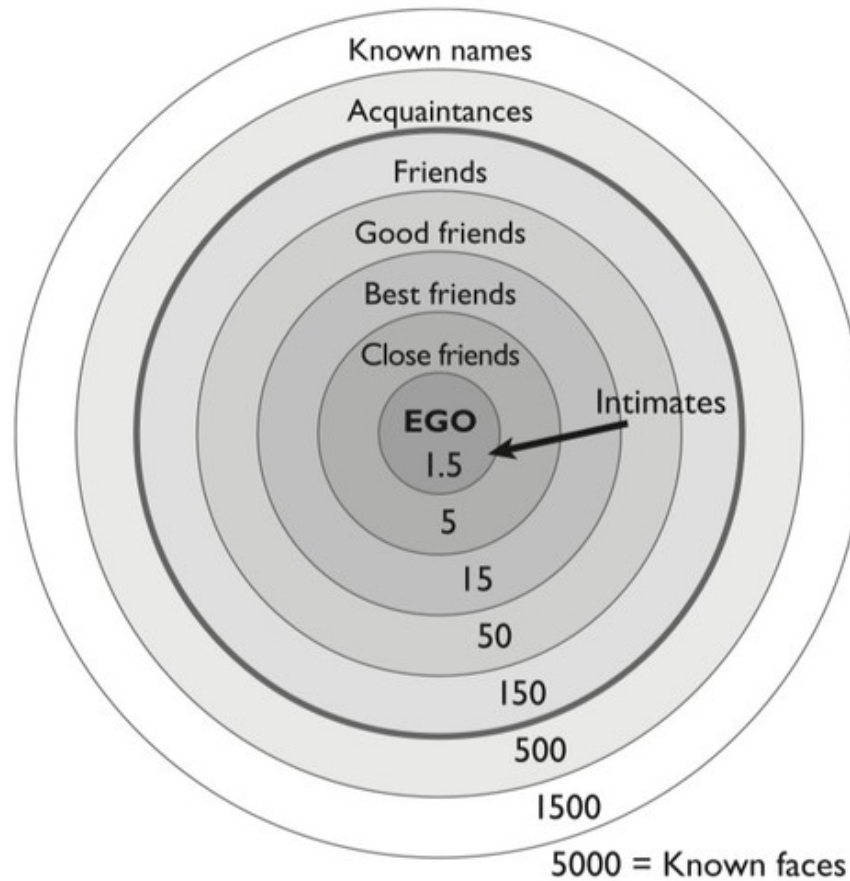


Dunbar number and circles

Dunbar number and circles

- Robin Dunbar checked the frequency of sending Christmas cards over the years for a number of people
 - The more frequently sent, the closer they are
 - Then via statistical analysis he found that these frequencies are discretized
 - There are “levels” of friendship

Dunbar number and circles



Today's online social networks



Online Social Networks (also called social media and OSN)

- Among the most popular sites on today's Web
- Billions of users world-wide
 - Celebrities, media houses, politicians, commoners, ...
 - Spammers, cyber-bullies, hatemongers, ...
- Huge impact
 - Advertisers reach large population at minimal cost

OSN in research

- Huge data readily available
 - Volume – networks of billions of users, petabytes of user generated content every day
 - Variety – text, image, speech, video, ...
 - Velocity – thousands of posts / minute during major events
- Automated data collection rather than surveys

Multi-disciplinary research on OSNs

- Computer networks & distributed systems
- Sociology, social psychology, linguistics, ...
- Network science, complex network theory
- Data mining, machine learning, information retrieval, natural language processing, privacy...

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THIS
COURSE

Two important aspects in a social network / social media

- The network structure
- The content
- This course
 - First part – network structure – we will consider only simple networks
 - Second part – content – we will focus on only textual content

Social networks vs. Social media

- “What is Twitter, a Social Network or a News Media?”, Kwak et al., WWW 2010
- What is the difference between social network and social media?
 - Social network – interpersonal links, interactions in focus
 - Social media – information exchange in focus
- Sometimes these two influence each other
 - Case study: Why do FB sometimes show things that you were texting about on whatsapp (which is encrypted)?

- Course logistics
- Social networks in the offline and online worlds
- Research challenges on online social networks (OSN) / social media

1. Modelling OSN

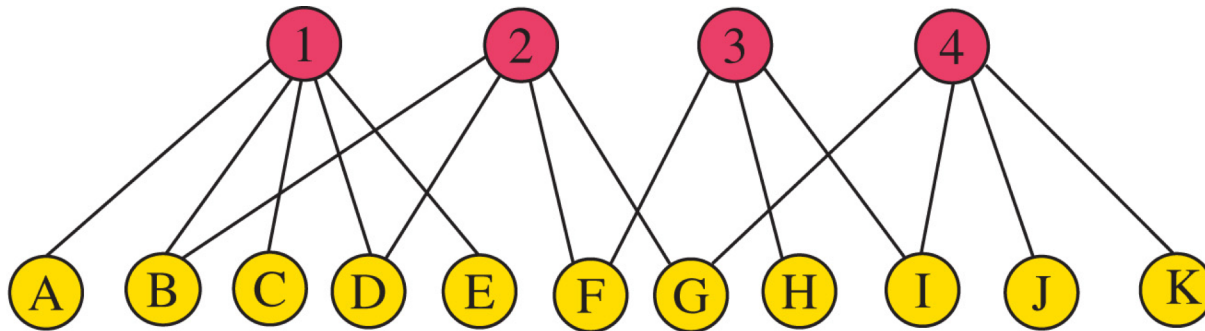
How to model / represent OSNs?

- Most common representation: a graph
 - Nodes: users, edges: social links
- Undirected networks: Facebook
- Directed networks: Twitter
- Weighted networks
 - Edge-weights usually measure “strength” of social link, e.g., number of interactions

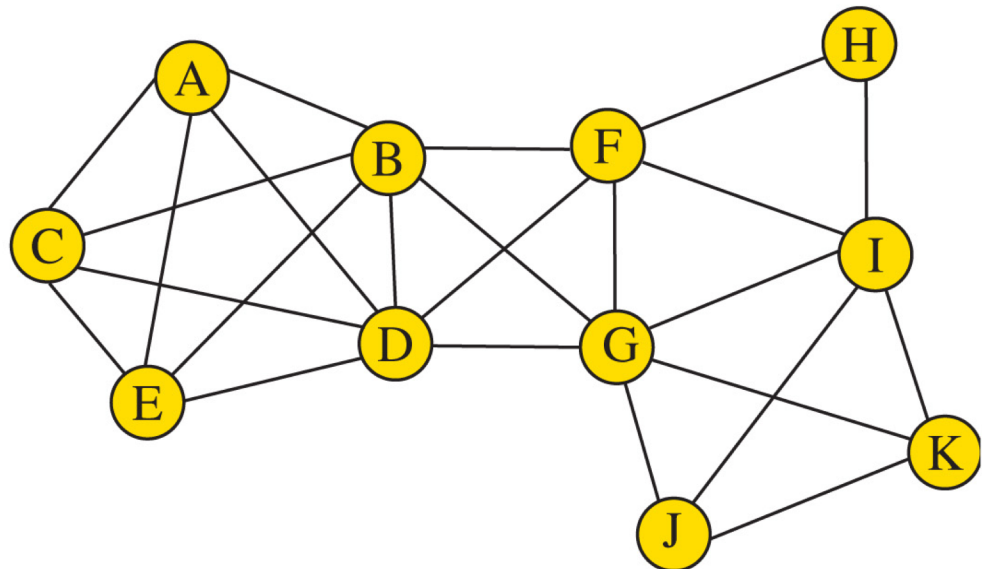
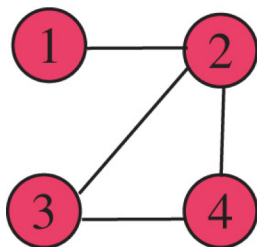
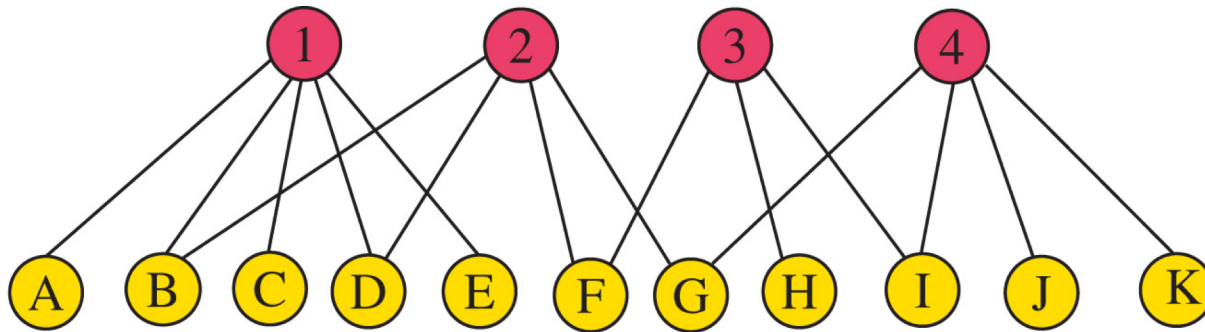
Graph models of OSNs

- Other varieties of networks
 - Networks among blogs, videos, ...
 - **Bipartite networks**, e.g., viewer-video model of Youtube
 - **Folksonomy**: Users annotate resources with tags, modeled as tri-partite hypergraphs [Cattuto, AI Communications 2007]

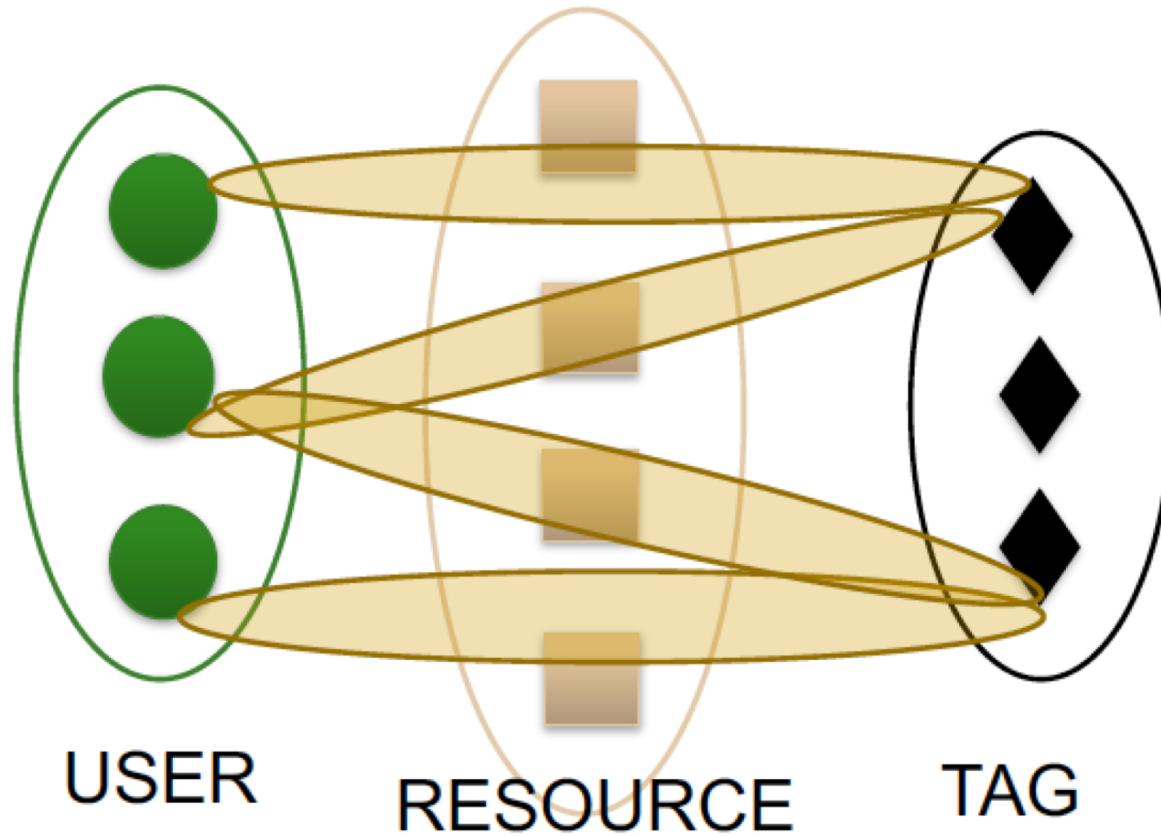
Bipartite networks and projections



Bipartite networks and projections



Tri-partite model for folksonomies



Sociological differences between offline and online graph

- Sociological theories investigated on OSNs
 - Homophily, strength of weak ties [Grabowicz, Plos ONE, 2012]
 - Emergence and spread of conventions [Kooti, ICWSM 2012]
- OSNs different from offline SNs in some aspects
 - Offline social networks – an individual can maintain only a certain number of meaningful social links – Dunbar number (~150); OSNs – almost zero cost of maintaining
 - social links - many more links can be maintained
 - Important users readily connect to many ordinary ones
 - Geographical distance does not matter

Locality of friendship in Facebook



- <http://www.techprone.com/facebook-displays-visually-its-impact-and-spread>

2. Properties of online social graphs

Network properties of OSNs [Mislove, IMC 2007]

- Most users have few links, few have many links
 - Degree distributions: power-law, exponential, ...
- Presence of numerous triangles (transitivity)
- Small-world, e.g., 6 degrees of separation
- Assortativity, homophily

Explaining the network properties

- What nature of link-creation dynamics explain the empirically observed properties of OSNs?
- Several evolution models proposed
 - Global rules, e.g., Preferential Attachment [Barabasi, Science 1999]
 - Local rules, e.g., triangle closure [Kleinberg, ICWSM 2010], random walk starting from a node [Vasquez, PRE 2003]
 - Biased Preferential attachment, based on different types of users: inactive, linkers, inviters [Kumar, KDD06]
 - Co-evolution of social and content networks [Singer, Making Sense of Microposts, 2012]

Dynamic network properties

- Dynamic nature: how do properties of OSNs change with time?
 - Network density varies non-monotonically [Kumar, KDD06]
 - Assortativity varies non-monotonically [Hu, Physics Letters A, 2009]
- Models to explain temporal variation of properties

Link analysis

- Classification of social links
 - Strong and weak links (e.g. based on level of interaction) [Gilbert, CHI'09][Wilson, EuroSys09][Valafar, WOSN09] [Xiang, WWW10]
 - Some OSNs allow positive and negative links (friends and enemies)
- Variation of strength of links with time [Viswanath, WOSN09]

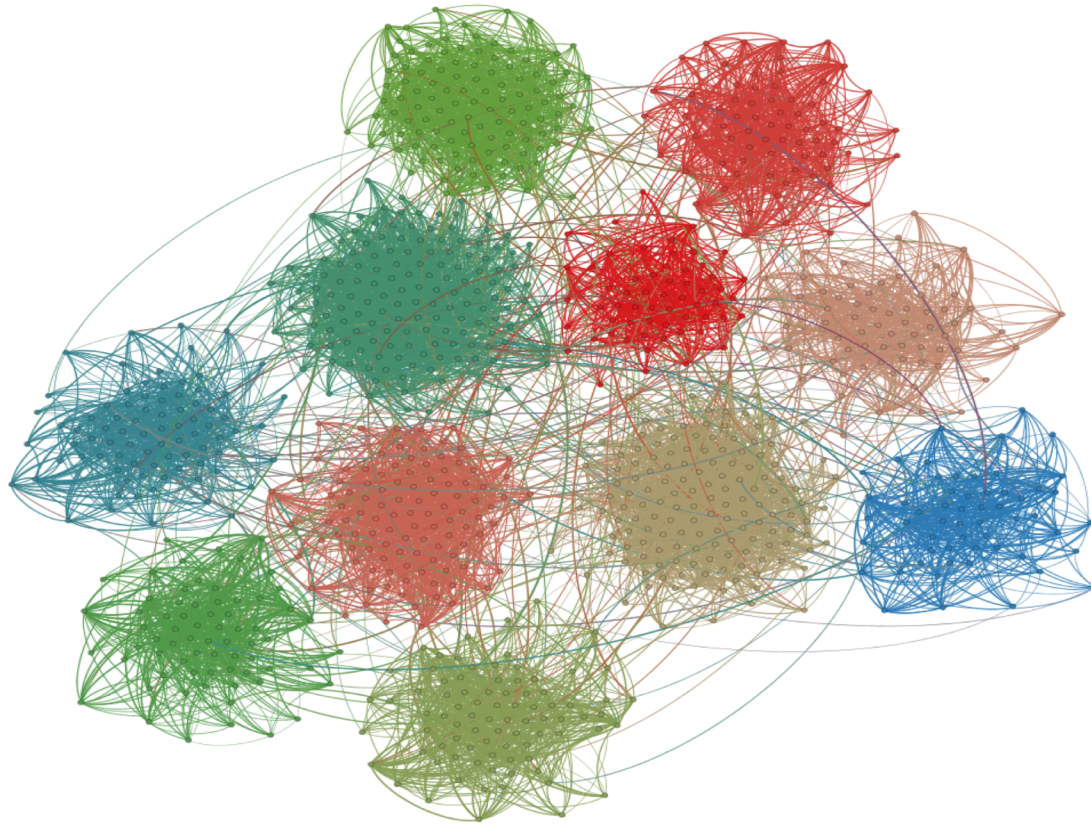
Centrality (importance) of nodes

- How important is a node in a network?
 - How influential is a person in a social network?
 - How important is a website on the Web?
- Many proposed centrality metrics
 - Degree centrality
 - Closeness centrality
 - Betweenness centrality
 - Eigenvector centrality, PageRank

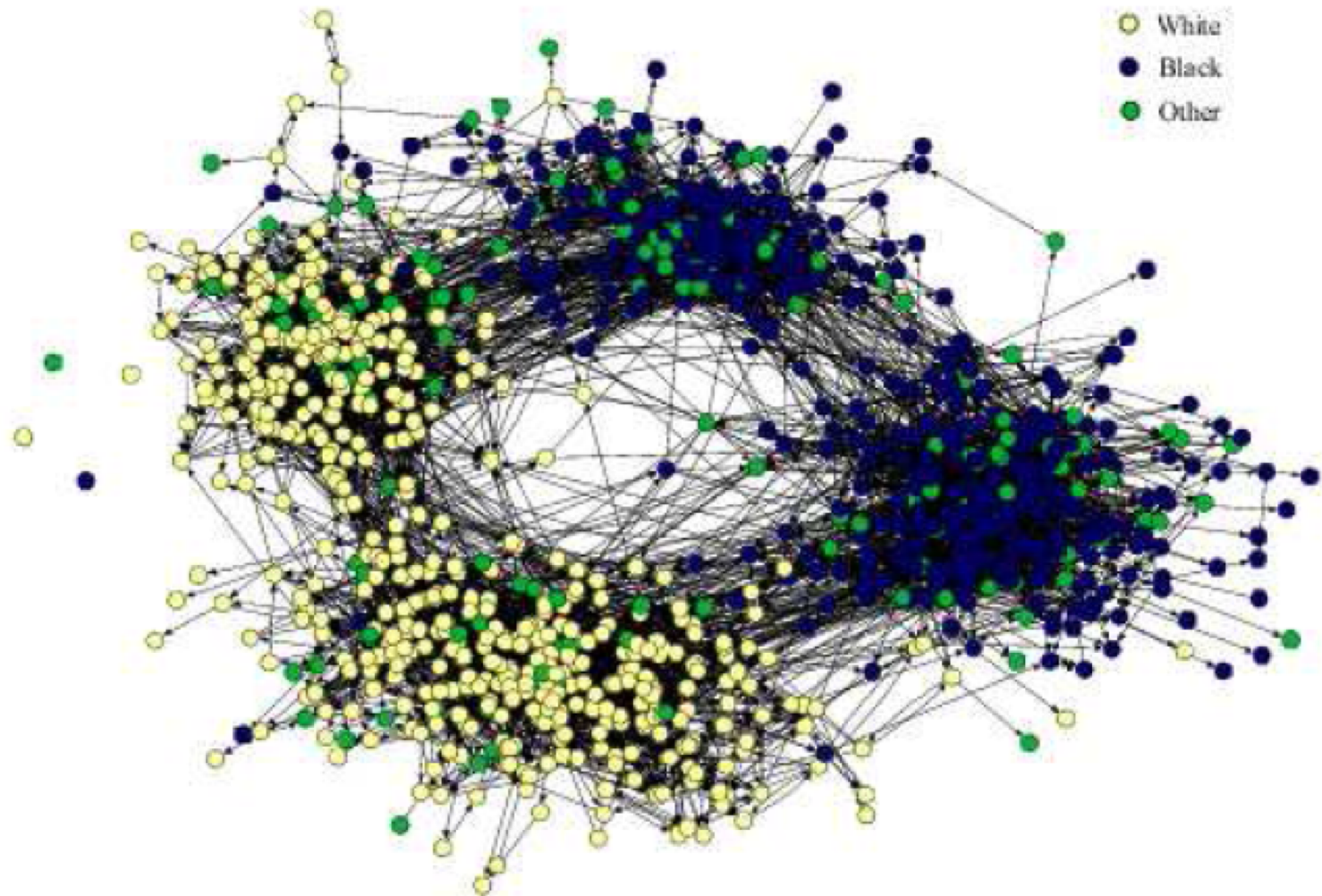
Community detection / clustering

- Identifying communities of ‘similar’ users
 - Traditionally, only rely on network structure: several algorithms
[Fortunato, Physics Reports 2010] [Leskovec, WWW10]
 - Content can also be leveraged in case of OSNs
- Dynamic communities: how do communities change with time? [Mitra, Computer Networks, 2012]

Community example



Friendship network among students in a US school

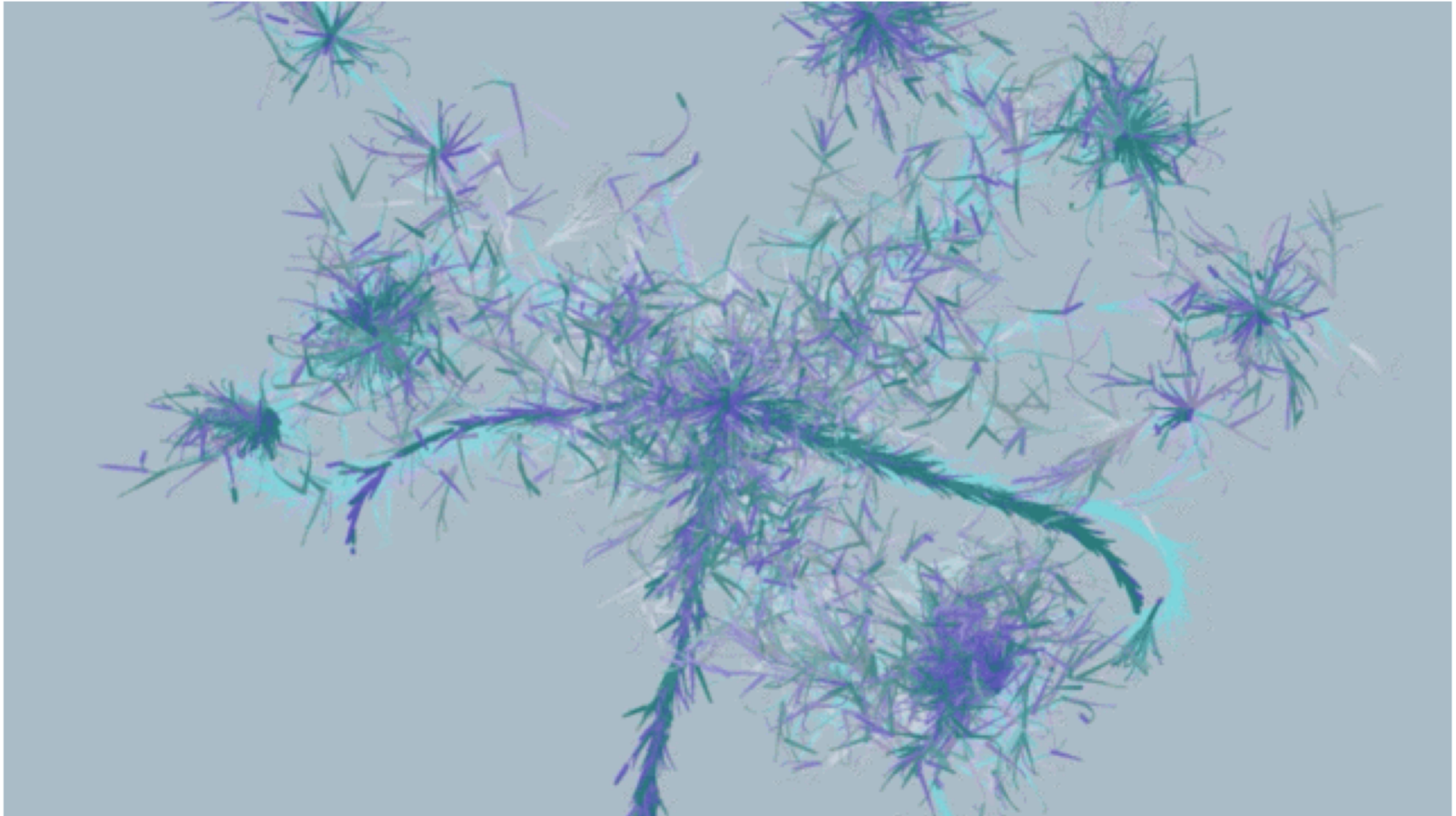


3. Modeling spread of information in OSN

Information spread / diffusion

- Understanding information spread / diffusion in OSNs
[Cha, WWW 2009] [Lerman, ICWSM 2010] [Bakshy, WWW 2012]
 - To what extent does information (news) spread?
 - How fast? Along which links?
 - Who are the most influential in spreading information?
 - How does a topic / video become viral?

Spread of viral images in Facebook



<http://www.gizmodo.com.au/2012/10/how-viral-images-spread-on-facebook-visualised/>

4. Utilizing content uploaded in OSN

Utilizing information content in OSNs

- Recommendation and search
- Identifying influential users / experts
- Misinformation detection
- Authority identification
- Enabling private communication

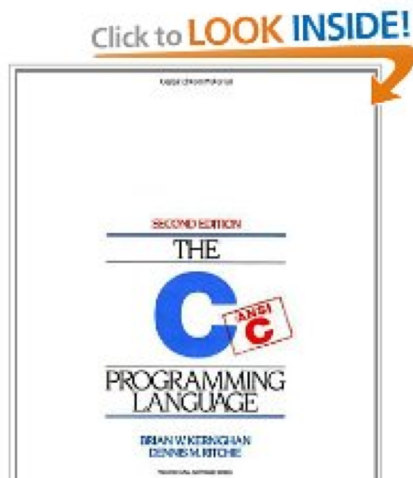
Recommendation and search

- Help users discover interesting content, friends, groups
- **Motivation:** The amount of information has become so large that it is impossible for an individual user to find out on her own interesting content / friends / groups
- Recommend friends, groups to join [Chen, WWW09], resources [Konstas, SIGIR09], tags [Sen, WWW09][Song, SIGIR08]
- Personalized answers to queries [Xu, SIGIR08] [Bao, WWW07] [Mislove, HotNets06]

Recommendation

- Two broad ways
 - Content-based, e.g., based on your profile information (e.g., you study in IITKGP), or some keywords given by you while creating account
 - Collaborative filtering – identify “similar” users or items
 - how to find “similar” users or items?

Recommendation of books in Amazon



C Programming Language (2nd Edition)

[Brian W. Kernighan](#) (Author), [Dennis M. Ritchie](#) (Author)

★★★★★ (367 customer reviews)

Buy New

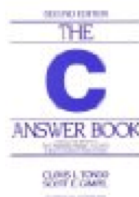
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Social recommendations

Similar friends might have similar tastes



Identify influential users / experts

- Several metric of influence: #followers or #friends, PageRank, number of times retweeted [Cha, ICWSM10]
- Identifying topical experts [Weng, WSDM10] [Pal, WSDM11] [Ghosh, SIGIR12]
- How to measure topic-specific expertise / interests of users?

Emotion / opinion mining

- Identify user's emotion / opinion from posts
 - Identify opinion on movies / political issues [Fang, WSDM12]
 - Comparison among different methods[Goncalves, COSN13]
 - Summarization of opinions [Ganesan, WWW12]
 - Twitter used to predict election results [Tumasjan, ICWSM10]

Various types misinformation on social media

- Spam, phishing, ...
- Hate speech (against a particular social / religious /ethnic group), cyberbullying, ...
- Fake news, rumors, ...

Spam detection

- Identify spam / users with malicious intentions [Heymann, IEEE Internet Computing 2007]
 - Identify spam in Facebook [Gao, IMC10] , Twitter [Lee, SIGIR10], Youtube [Benevenuto, SIGIR09], blogs [Shin, Infocom11], ...
 - Sybil detection [Yu, SIGCOMM 2006][Viswanath, SIGCOMM10]
- Identifying trustworthy entities, e.g., reviews, ratings [Chandra, Trustcom 2012]

Fake news – some ideas

- First step – identify claims / factual statements that need to be verified – NLP features used
- Second step – verify the claims
- How to verify?
 - From trustworthy information sources, e.g., claims about COVID19 can be verified from medical research papers
 - Crowdsourcing from trusted people, e.g., relief workers present in the region of a disaster

Privacy of social content

- What information to share / what not to share?
- What privacy settings to choose while uploading?
 - Online social media does not have the geographical separation of online world – context collapse
- Privacy of old content
 - Privacy preferences get changed over time – users almost never go back and reevaluate privacy settings

Thank you

<http://cse.iitkgp.ac.in/~mainack>