



Mining Social Graphs

Text, Web and Social Media Analytics Lab

Prof. Dr. Diana Hristova

Exercise 7: Social Media



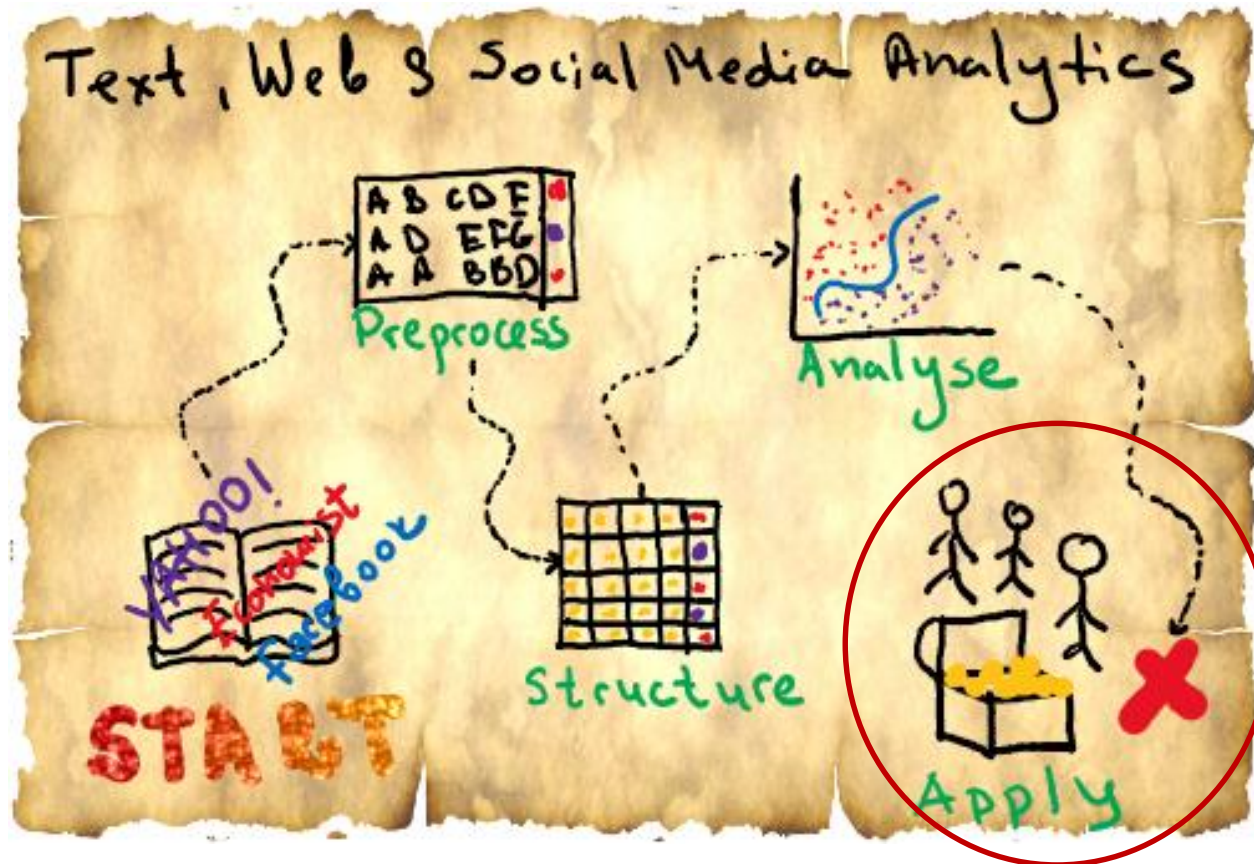
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Can one group present please?



What did we learn last week?

Social Media Analytics: Treasury map



Course structure



Date	Lecture	Exercise
12.04.2021	Introduction	Technical Installation
19.04.2021	Text Preprocessing	Projects kick-off
26.04.2021	Text Representation	Preprocessing Newsgroups
03.05.2021	Text Representation (2)	Text Representation Newsgroups
10.05.2021	Text Classification	Text Representation Newsgroups (2)
17.05.2021	Text Clustering/Capgemni	Newsgroups Topic Classification
31.05.2021	Text Mining in Social Media	Newsgroups Topic Clustering
07.06.2021	Mining Social Graphs	Sentiment Analysis and Time Series in Twitter
14.06.2021	Projects Status Update	Projects Status Update
21.06.2021	Web Analytics	Mining Social Graphs in Twitter
28.06.2021	Mock Exam	Web Analytics in E-commerce
05.07.2021	Final Presentation	Final Presentation
19.07.2021	Submit Code & Written report	
t.b.a.	Exam	

Projects Status Update



- ☐ Each group should discuss the following points:
 1. What is the task?
 2. Why does it generate added value?
 3. How does the data look like (high-level)?
 4. What is the planned approach?
 5. First results
 6. Issues
- ☐ Teams with projects with companies should send their slides for approval to the corresponding company representative in advance.
- ☐ Please plan with 20 minutes for your presentation including Q&A.

Projects Status Update: Preliminary Schedule



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- Group 1 (Signavio): 08:05-08:25
- Group 2 (Signavio): 08:30-08:50
- Group 3 (Strayz): 08:55-09:15

Break: 10 Minutes

- Group 4 (Aam): 09:25-09:45
- Group 5 (BayLev): 09:50-10:10
- Group 6 (Zalando): 10:15-10:35



What will we learn today?

At the end of this lecture, you will:

1. Know the motivation behind analysing social media graphs.
2. Be able to represent social media graphs as networks and calculate the following network metrics:
 - Degree centrality
 - Betweenness centrality

Motivation: Why Analyse Social Media Graphs?

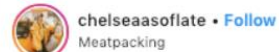


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Which politician is likely to influence more people on Twitter?



Motivation: Why Analyse Social Media Graphs? (2)



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Meatpacking

chelseasoflate #ad Welcome to @StarbucksReserveRoastery! After years in NYC, Starbucks takes over Meatpacking District with a new concept that will blow your minds! ☕ From Main Bar, where freshly roasted Starbucks Reserve beans are brewed into the perfect cup of coffee, to the Arriviamo Bar - where you can enjoy nice coffee-inspired cocktails - the Roastery is legit the most magical for coffee lovers. Head over to the to learn more! 🍷❤️ #StarbucksRoastery #NewYorkRoastery

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audreymadstowe 🌟 Trying to get ahead of my gifts for family this year... ❤️ and recently grabbed the @gillette Limited Edition Gift Packs to gift my brother and Josh. 🎁 I picked these up in 5 mins from Walgreens and they're the perfect, easy gift to give the men in your life. Bonus, the razor is sleek and comes with a stand to ensure your bathroom is staying tidy. Walgreens is doing a BOGO 50% off until 11/24 and \$2 off these limited edition packs from 12/16 - 12/22 if you need something last minute. 🎁🎁🎁 #sponsored #gillettepartner

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audreymadstowe @ari_speed2 pacsun!! 🎁
faithfullyfarmy The best razors!!
monicsutter Perfect gift for the

508 likes

NOVEMBER 21, 2018

AT?!? This is so
need to check this

Cmon Starbucks

What's

3,615,5



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zachking When your wingmen are on point to help you impress the love of your life. #DateNight #MeetAnOwner like me and other real @Subaru_USA owners on MeetAnOwner.com #ad

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ygorgedroso @ch.mlopes vou te dar um desse
jack_tequilis @carlos.gabriel94 @kah_barbossa assim é bom ter um carro hahahaha

carlos.gabriel94 @jack_tequilis topperson hahahaha

carsquin What?

agiletp ntaap wkwk @_deninn.31

@ramadhani_fw @d.ferdimas

petiquay11 How to lost your friends for ever 🤔

8,257,567 views

MARCH 18, 2017

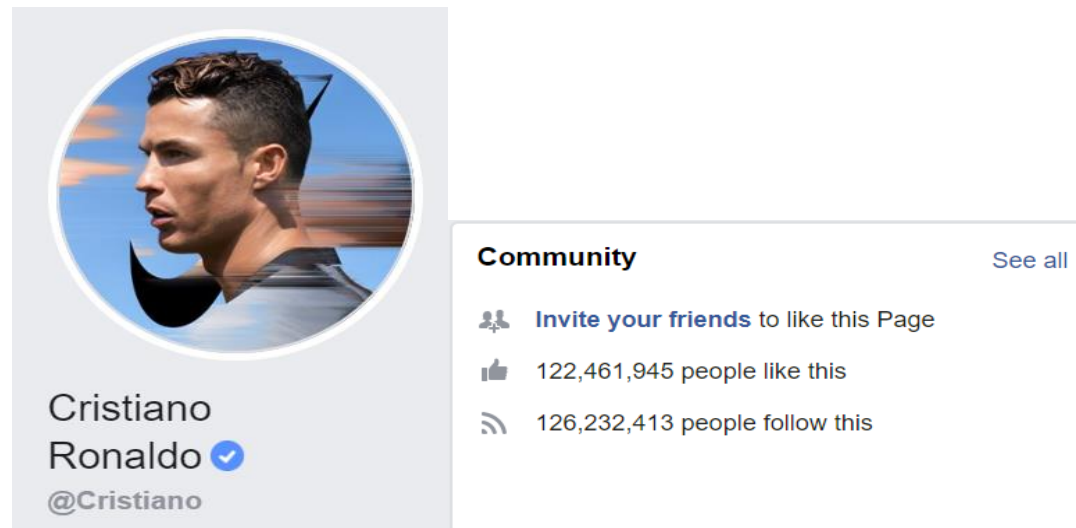
Log in to like or comment.

?

How to choose your social media partners?

Motivation: Why Analyse Social Media Graphs? (3)

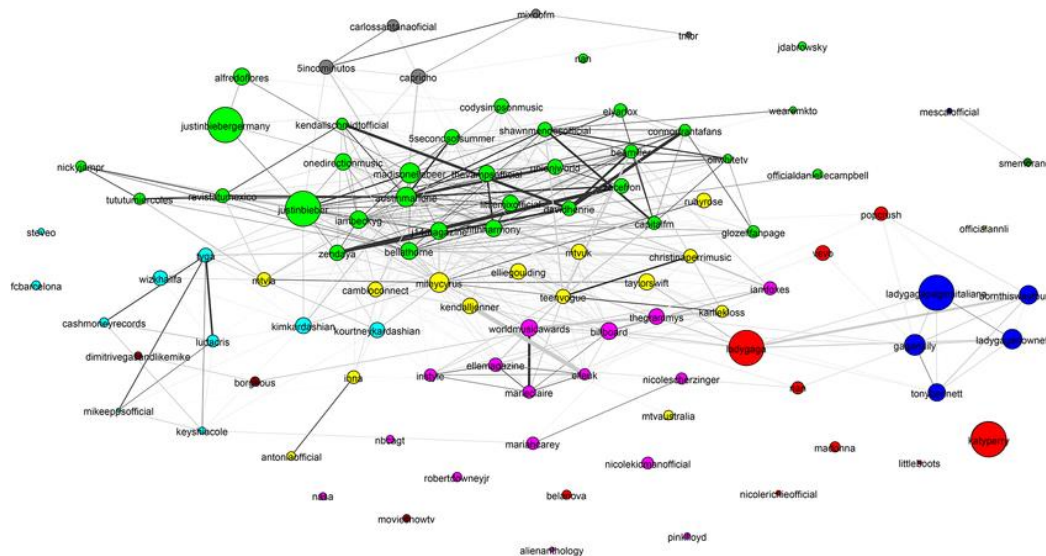
- Analysing social media graphs can help you identify important users in the network
- Two types of important users in social media graphs we consider here:
 1. Users with many direct connections i.e., followers, retweets, likes, comments → Use to reach many customers directly e.g. by advertising with them



Motivation: Why Analyse Social Media Graphs? (4)

Two types of important users in social media graphs:

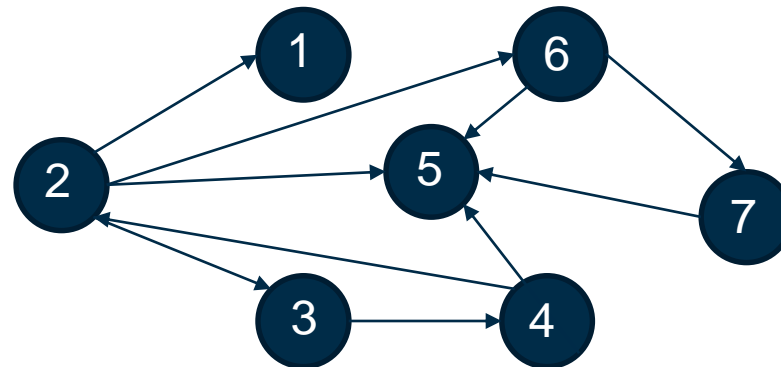
2. Users who are central nodes for information spreading i.e. they transfer ideas between multiple groups (brokers) → use them to reach new customer groups with different interests (e.g. not football fans)



Chamberlain BP, Levy-Kramer J, Humby C, Deisenroth MP (2018) Real-time community detection in full social networks on a laptop. PLoS ONE 13(1): e0188702. <https://doi.org/10.1371/journal.pone.0188702>

Analysing Social Media Graphs: Network Analysis

- In order to determine the most important users, **network analysis** is applied.
- A network consists of a set of **nodes** (e.g. users, tweets) and **edges** that connect them.
- An edge shows a relationship between two nodes (e.g. comment, retweet, following, friends).



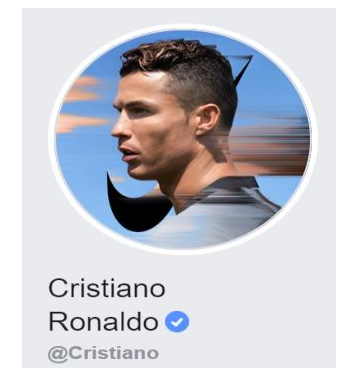
- An edge can be directed (i.e. relationship **from** a source node **to** a sink node) or undirected (i.e. mutual relationship **between** nodes).
- **Focus here:** directed edges and therefore directed networks



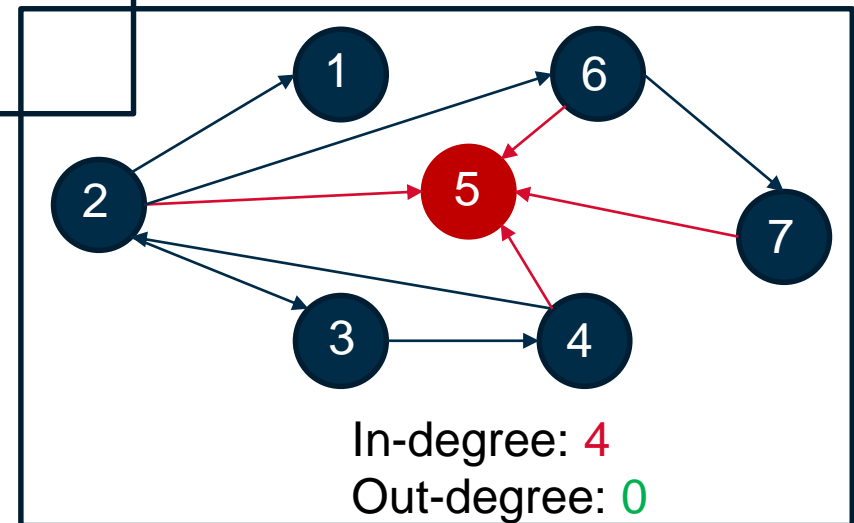
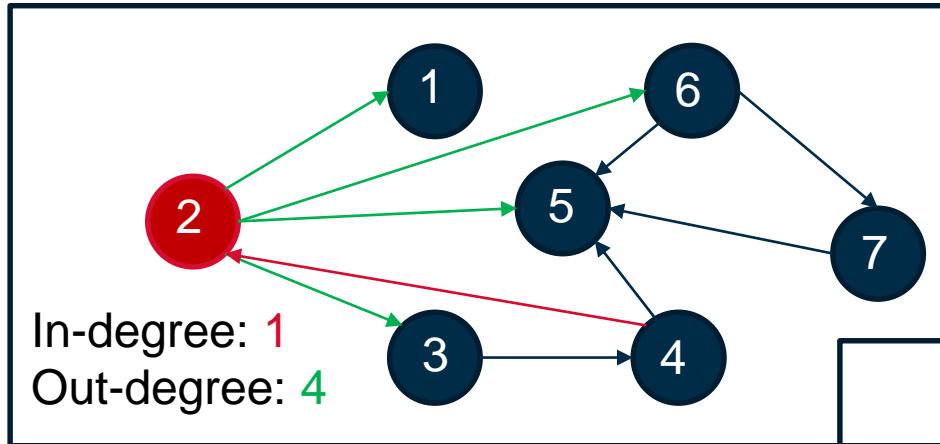
Analysing Social Media Graphs: Network Metrics

- There are different metrics to measure the role of a node (user/tweet) in a network.
- **Centrality:** measures a node's "importance" in the network
- Two types of centrality in this lecture: **degree centrality** and **betweenness centrality**
- **Degree centrality:** number of edges connected to a node i.e. "...the ability of a node to communicate directly with others." (Estrada Knight 2015, p. 143)
 - In-degree: how many edges go **into** a node (e.g. how often someone retweets)
 - Out-degree: how many edges go **out of** a node (e.g. how often is someone retweeted)

Ernesto Estrada and Philip A. Knight (2015): A First Course in Network Theory, Oxford University Press, 272 pp., £29.99, ISBN 9780198726463. In: *Statistical Papers* 58 (4), S. 1283–1284. DOI: 10.1007/s00362-017-0961-1.



Network Metrics: Degree Centrality



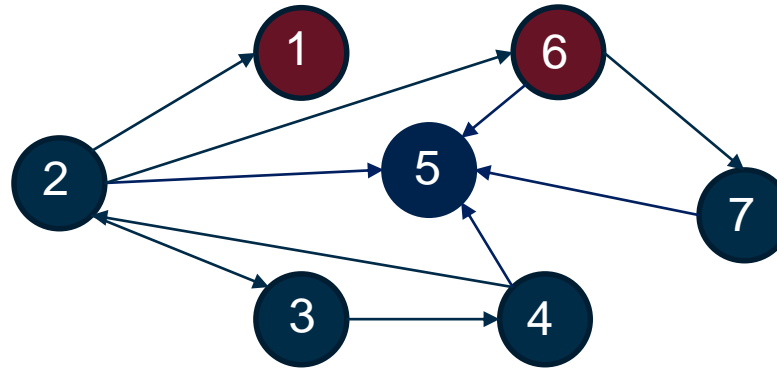
- ➔ User 2 is being often retweeted.
- ➔ User 5 retweets very often.

Network Metrics: Degree centrality

Exercise



What is the in- and out-degree of the below nodes?



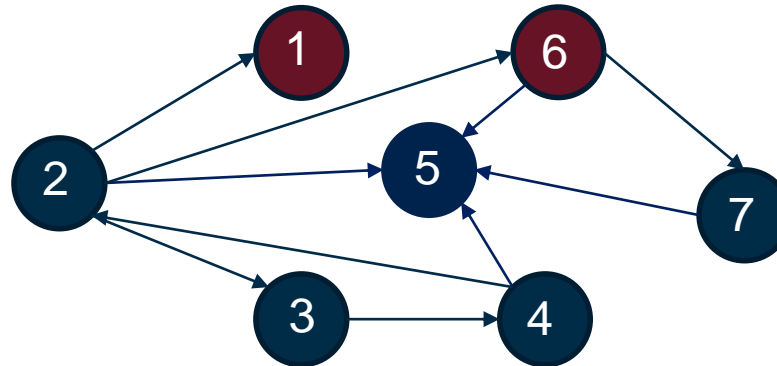
Node	In-degree	Out-degree
1		
6		

Network Metrics: Degree centrality

Exercise



What is the in- and out-degree of the below nodes?



Node	In-degree	Out-degree
1	1	0
6	1	2

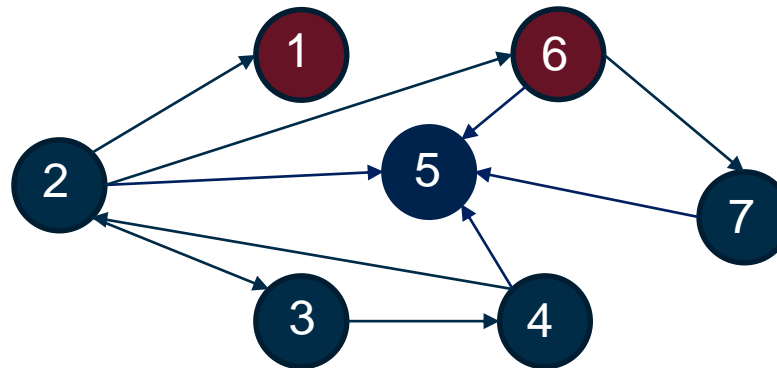
! **Note:** degree centrality is often normalised by dividing over $(n-1)$ with n =number of nodes

Network Metrics: Degree centrality

Exercise



What is the in- and out-degree of the below nodes?



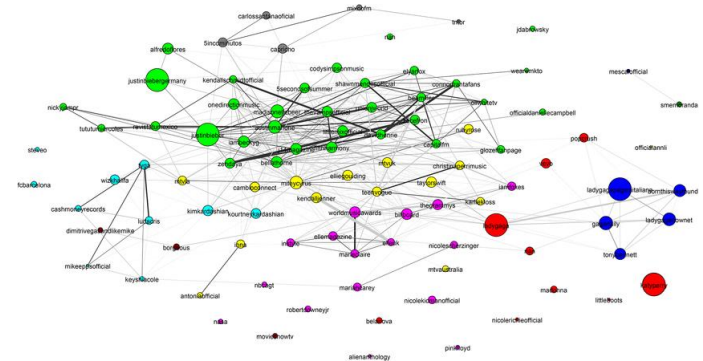
Node	In-degree	Out-degree
1	1	0
6	1	2



What is a disadvantage of degree centrality?

Network Metrics: Betweenness Centrality

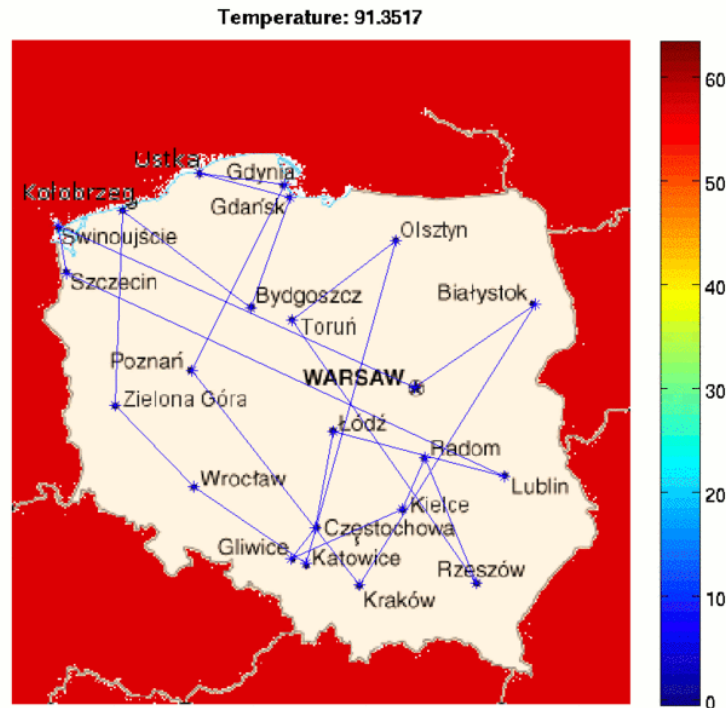
- Degree centrality** focuses on the user's importance, but does not consider the **flow of information** to other users who are not directly related to her. → Betweenness centrality
- Betweenness centrality:** „... a measure of the influence of a vertex over the **flow of information** between every pair of vertices under the assumption that information primarily flows over the **shortest paths** between them.” (p. 1)¹
- Shortest path:** what is the quickest connection (e.g. the least number of users) between two nodes? → we assume information is spread efficiently



¹Unnithan, Raghavan, et al. "Betweenness centrality in some classes of graphs." *International Journal of Combinatorics* 2014 (2014).

Network Metrics: Betweenness Centrality (2)

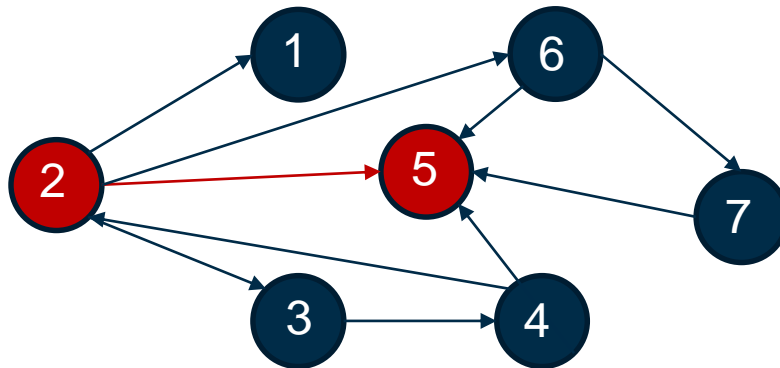
- Betweenness centrality is based on **transportation theory**.



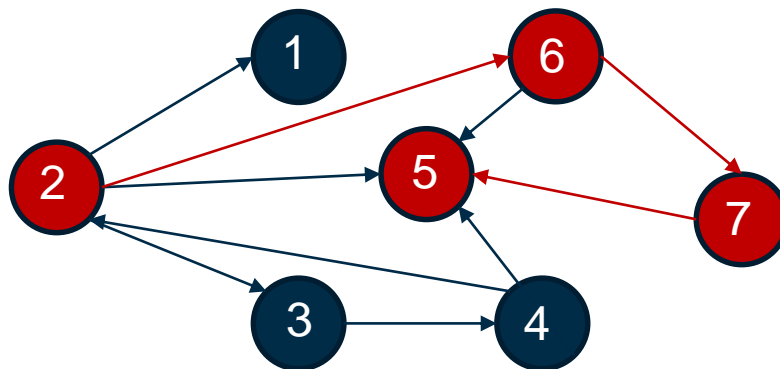
https://commons.wikimedia.org/wiki/File:Sa_poland_tsp.gif



Network Metrics: Betweenness Centrality (3)



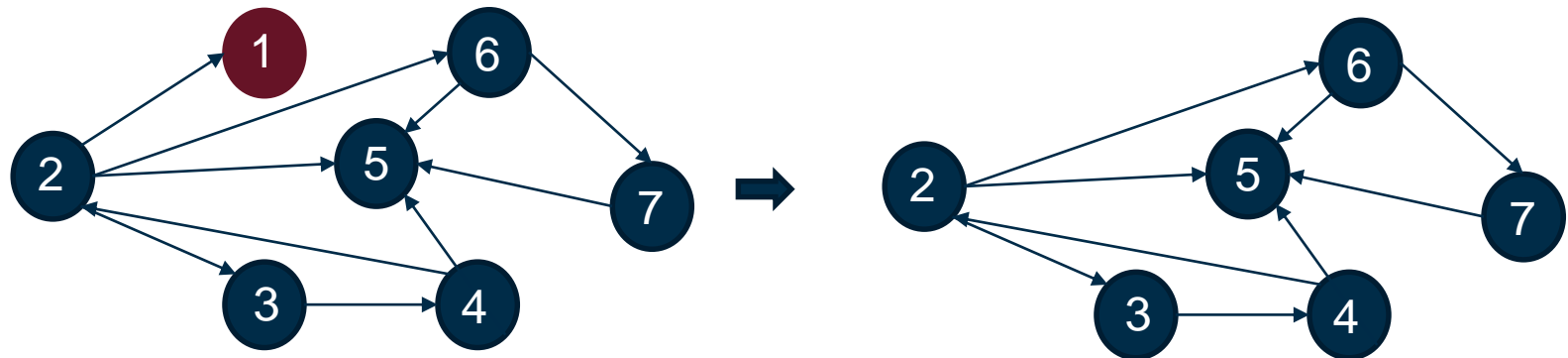
Shortest path from 2 to 5



Not the shortest path from 2 to 5

Network Metrics: Betweenness Centrality Calculation

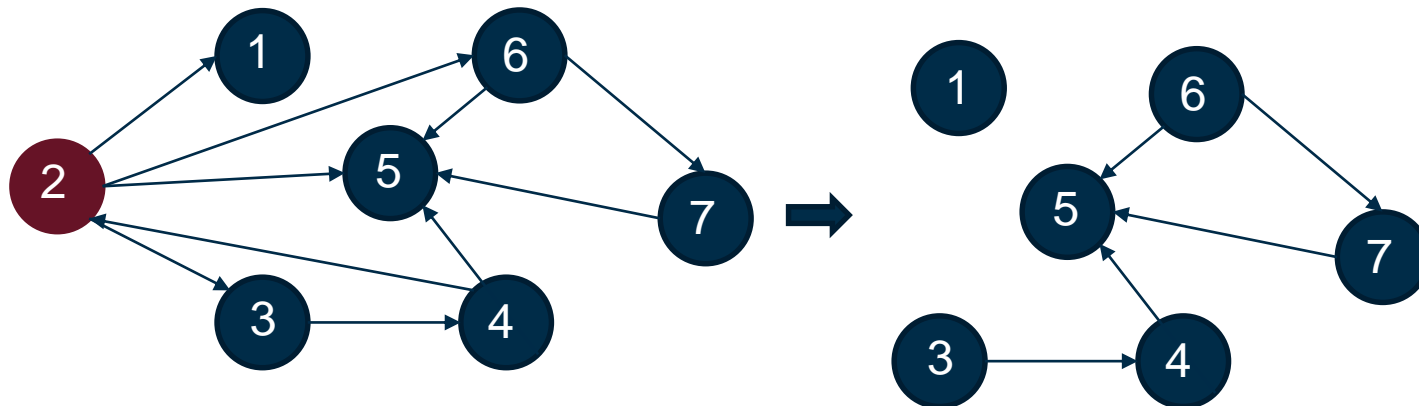
- **Betweenness centrality:** „... a measure of the influence of a vertex over the **flow of information** between every pair of vertices under the assumption that information primarily flows over the **shortest paths** between them.” (p. 1)¹
- **Betweenness centrality calculation:** for a given node, determine how many times it will interrupt the shortest-path flow of information in the network, if removed.
- **Example:** remove Node 1



- Node 1 interrupts the shortest path zero times
- ➔ removing it is not so crucial for the information flow in the network

Network Metrics: Betweenness Centrality Calculation (2)

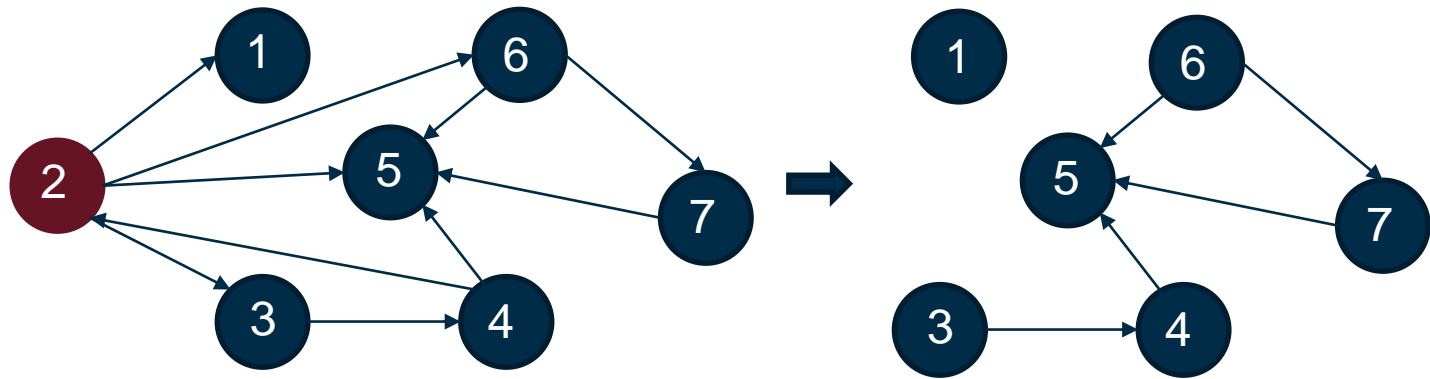
- **Example:** remove Node 2



- Node 2 interrupts the shortest path from:
 - Node 3 to Node 1, 6 and 7
 - Node 4 to Node 1, 3, 6, and 7
- ➔ 7 times
- ➔ Removing it is very important for the flow of information.

Network Metrics: Betweenness Centrality Calculation (3)

- In order to be able to compare the betweenness centrality of different networks, it is normalised by $(n-1)(n-2)$, where n is the number of nodes in the network (here 7).

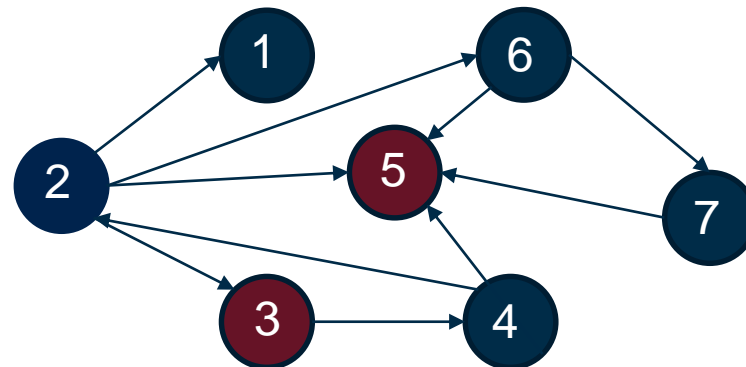


- Node 2 interrupts the shortest path 7 times → normalised betweenness centrality = $7/30$

Network Metrics: Betweenness Centrality Exercise

- What is the normalised and the not normalised betweenness of the below nodes?

Node	Betweenness (not normalised)	Betweenness (normalised)
3		
5		

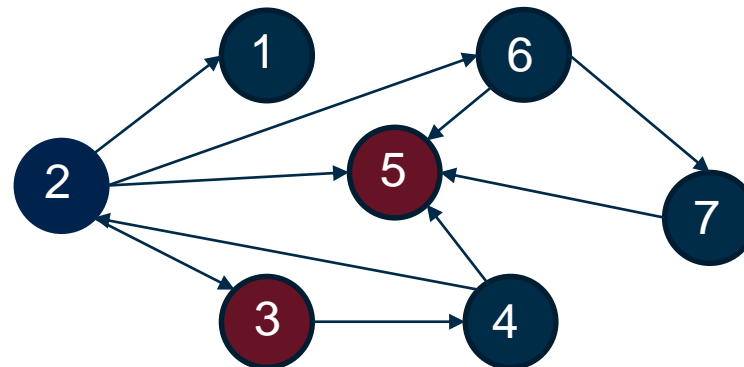




Network Metrics: Betweenness Centrality Exercise

- What is the normalised and the not normalised betweenness of the below nodes?

Node	Betweenness (not normalised)	Betweenness (normalised)
3	1	0.03
5	0	0



Summary:

- Mining social graphs is important because it helps identify important users and their influence on the flow of information.
- Social graphs can be modelled with network analysis tools and analysed with network metrics.
- Degree centrality is determined the number of direct connections of a user.
- Betweenness centrality determines the importance of the user for the flow of information in the network. It can be normalized for comparison between networks.

Outlook: Social graphs' analytics can be combined with web analytics to determine optimal marketing strategies.



Questions?

Exercise 8



In a minute, six break-out rooms will be created. Choose the room that corresponds to your group in Moodle e.g. Room 1= Group 1. In your project group discuss and document the solution for Exercise 8 (in Moodle).