

# Assignment of EE 4389

February 22, 2018

Consider the high school Face-to-Face contact data<sup>1</sup>, which is given in the following format: each row "a b t" denotes a contact (a temporal link) between node a and b at time step t. This contact network is sampled/measured once every 20 seconds. Thus, each time step has a duration of 20s, which is though not relevant for our analysis in this assignment. We denote this temporal network as  $G_{data}$ .

A. Explore the topological features of the network  $G$  that is aggregated over all the  $T = 7375$  steps. Specifically, the aggregated network  $G$  is composed of all the nodes that have ever appeared in the dataset and any two nodes are connected by a link (unweighted) if they have at least a contact over the whole period.

Compute the following network properties for  $G$ .

- 1) What is the number of nodes  $N$ , the number of links  $L$ , the link density  $p$ , the average degree  $E[D]$  and the degree variance  $Var[D]$ ?
- 2) Plot the degree distribution. Which network model, Erdős-Rényi (ER) random graphs or scale-free networks, could better model this network? Why?
- 3) What is the degree correlation (assortativity)  $\rho_D$ ? What is its physical meaning?
- 4) What is the clustering coefficient  $C$ ?
- 5) What is the average hopcount  $E[H]$  of the shortest paths between all node pairs? What is the diameter  $H_{max}$ ?
- 6) Has this network the small-world property? Justify your conclusion quantitatively.
- 7) What is the largest eigenvalue (spectral radius)  $\lambda_1$  of the adjacency matrix?
- 8) What is the second smallest eigenvalue  $\mu_{N-1}$  of the Laplacian matrix (algebraic connectivity)?

Hint: All metrics computed for the network  $G$  are recommended to put into a table.

B. Information spreading on a temporal network

We consider the following information spreading process, which is actually a simplified Susceptible-Infected model but on a temporal network. Initially, at time  $t = 0$ , a single node  $s$  is infected meaning that this node possesses the information whereas all the other nodes are Susceptible, thus have not yet perceived the information. Node  $s$  is also called the seed of the information. Whenever an infected node  $i$  is in contact with a susceptible node  $j$  at any time step  $t$ , the susceptible node becomes infected during the same time step and could possibly infect other nodes only since the next time step via its contacts with susceptible nodes. Once a node becomes infected, it stays infected forever. For example, assume that the seed node has its first contact, e.g. with a node  $m$  at time  $t = 5$ . Although node  $s$  gets infected since  $t = 0$ , it infects a second node, i.e. node  $m$  only at  $t = 5$  when it contacts  $m$ . Infection happens only when an infected node and a susceptible node are in contact. The number of infected nodes is non-decreasing over time.

Simulate the information spreading process on the given temporal network  $G_{data}$  for  $N$  iterations. Each iteration starts with a different seed node infected at  $t = 0$  and ends at  $t = T = 7375$  the last time step that the network is measured. Record the number of infected nodes  $I(t)$  over time  $t$  for each iteration.

- 9) Taking all the  $N$  iterations into count, plot the average number of infected nodes  $E[I(t)]$  together with its error bar (standard deviation  $\sqrt{Var[I(t)]}$ ) as a function of the time step  $t$ .

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<sup>1</sup>R. Mastrandrea, J. Fournet, A. Barrat, Contact patterns in a high school: a comparison between data collected using wearable sensors, contact diaries and friendship surveys. PLoS ONE 10(9): e0136497 (2015)

10) How influential a node is as a seed node could be partially reflected by, e.g. the time it takes to reach/infect 80% of the total nodes when this node is selected as the seed node. The shorter the time is, the more influential the seed node is. Using this standard to rank the influence of all the nodes and record the ranking in a vector  $R = [R_{(1)}, R_{(2)}, \dots, R_{(N)}]$  where  $R_{(i)}$  is the node index of the  $i$ -th most influential seed node and  $R_{(1)}$  is the most influential node that infects 80% nodes in the shortest time. Note that you don't need to provide this vector in your report.

11) We are going to explore which nodal level network feature could well suggest the nodal influence discussed in 10). Compute the degree and clustering coefficient of each node in the aggregated network  $G$  and rank the importance of the nodes according to these two centrality metrics respectively. You obtain the ordered vector  $D = [D_{(1)}, D_{(2)}, \dots, D_{(N)}]$  and  $C = [C_{(1)}, C_{(2)}, \dots, C_{(N)}]$ , where  $D_{(i)}$  is the node having the  $i$ -th highest degree and  $C_{(i)}$  is the node with the  $i$ -th highest clustering coefficient. How precise a centrality metric e.g. the degree could predict seed nodes' influence could be quantified by the top  $f$  recognition rate  $r_{RD}(f) = \frac{|R_f \cap D_f|}{|R_f|}$  where  $R_f$  and  $D_f$  are the sets of nodes ranking in the top  $f$  fraction according to their influence and degree respectively and  $|R_f| = fN$  is the number of nodes in  $R_f$ . Plot  $r_{RD}(f)$  and  $r_{RC}(f)$  as a function of  $f$  where  $f = 0.05, 0.1, 0.15, \dots, 0.5$ . Which metric, the degree or the clustering coefficient could better predict the influence of the nodes? Why?

12) Propose another two nodal/centrality features that could possibly well predict nodes' influence. The previous two features are all based on the aggregated network. At least one feature that you are going to propose should take nodal temporal features into account. Compare the two features you proposed and the two features proposed in question 11): which feature better/badly reflects how influential a node is and why? Hint: Similar to 11) plot  $r_{RM}(f)$  as a function of  $f$  for each metric  $M$  you proposed.

13) Reflection on B: What can be improved in the aforementioned analysis method to discover which nodal feature could better reflect how influential a node is as a seed node?

C. Influence of temporal network features on information spreading.

14) Construct the following three temporal networks.  $G_2$  is exactly the same as  $G_{data}$  except that the time stamps describing when each temporal link (contact) appears in  $G_{data}$  are randomized in  $G_2$ . In other words,  $G_2$  is constructed by copying all the temporal links from  $G_{data}$  but their time stamps are randomly re-shuffled or equivalently, randomly reassigned to the temporal links. The number of contacts between each node pair is the same between  $G_{data}$  and  $G_2$ . [A time stamp vector  $v$ , whose length equals the number of contacts, can be randomly reshuffled to a vector  $v_2$  by assigning each element in  $v$  to a randomly selected position in vector  $v_2$  while avoiding more than one elements from  $v$  assigned to the same position in  $v_2$ ].

$G_3$  is constructed by the following steps:  $G_3^*$  has the same topology as  $G$ , which is an unweighted network. Second, assign the time stamps in  $G_{data}$  to the linked node pairs (links) in  $G_3^*$ , randomly. A link in  $G_3^*$  may receive more than one time stamps, meaning that the two nodes contact more than once. A link  $G_3^*$  receives no time stamp means that there is no contact between the corresponding two nodes.  $G_3$  is composed of all these contacts.

15) Simulate exactly the same information spreading process on  $G_2$  and  $G_3$  as described in B. On each temporal network,  $N$  iterations of the spreading processes are simulated and each iteration starts at a different seed node. Plot the average number of infected nodes  $E[I(t)]$  and the standard deviation  $\sqrt{Var[I(t)]}$  as a function of the time step  $t$  for  $G_{data}$ ,  $G_2$  and  $G_3$  respectively. Compare and rank the information spreading performance (e.g. prevalence or speed of the spread) on these three temporal networks. Interpret/explain your observation. For example, which temporal network features could possibly explain the different spreading performance on these temporal networks?