

Practical Assignment 5: Gephi and Social Network Basic Analysis

Behavior Dynamics in Social Networks 2017

PS.: Do you use Github? You can get this assignment from the repository.

This assignment is divided in 3 parts. The idea is to teach how to work with Gephi, a Social Analysis software, and also to do some data mining in Twitter.

Part A: Introduction to Gephi

A.1. Set up and introduction to Gephi

This part is dedicated to make yourself acquainted to the Gephi and its environment. You are going to install Gephi and take a look on the tutorial before doing the activities. Then you are going to follow the steps below.

The entire data set is found together with the assignment.

Gephi Download

- Gephi 0.9.2 - <https://gephi.org/users/download/>
- **Required:** Java Runtime Environment - <http://www.java.com/nl/download/index.jsp>
- **Remark:** update your Gephi when requested.

Some students were having problems to make Gephi work. If your system is saying that Java 1.8 is not installed, and you've installed it, just find the file where it is installed (usually inside the Program Files folder) and go to the `gephi/etc/gephi.conf` file on your computer and set the path for the `jdkhome` to the folde where your Java is installed.

Gephi Tutorials

- Quick Start Tutorial - <https://gephi.org/users/quick-start/>
- Optional - <https://gephi.org/users/tutorial-layouts/>
- Optional - <https://gephi.org/users/tutorial-visualization/>

When you open Gephi it should be possible to see this data sets:

- Les Miserables.gexf
- Java.gexf
- Power Grid.gml

Before we begin with the practical assignment below, it is necessary become acquainted with the Gephi environment. This leads us to step 1.

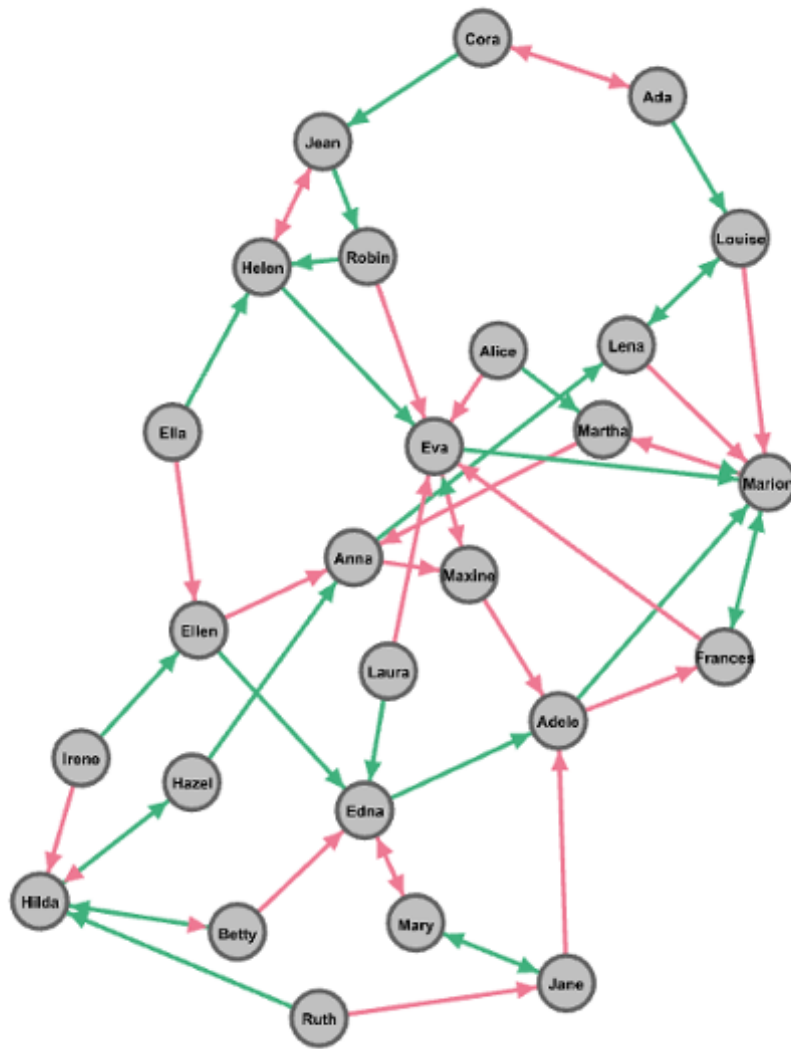


Fig 1: Girls' school dormitory dining-table partners, 1st (pink) and 2nd (green) choices (Moreno, The sociometry reader, 1960).

Step 1: Go through the quick start tutorial: <https://gephi.org/users/quick-start/>

The network for this tutorial '*LesMiserables.gefx*' can also be downloaded from Canvas.

Ps.: It seems that the tutorial provided by Gephi is not updated to the newest version. Feel free to skip steps that are not related due to changes in the interface

of the software.

A.2. The dining-table partners Network

Now we are going to look at the network we will use for this practical assignment.

The network in figure 1 shows an example of sociometric research. It depicts the choices of twenty-six girls living in one *cottage* (dormitory) at a New York state training school. The girls were asked to choose the ones they liked the best as their dining-table partners. First and second choices are selected only.

In this sociogram, each girl in the dormitory is represented by a circle. For the sake of identification, the girls' names are written next to the circles. Each arc (arrow) represents a choice. The girl who chooses a peer as a dining-table companion sends an arc toward her. Irene (in the bottom right of the figure), for instance, chose Hilda (pink arrow) as her favorite dining-table partner and Ellen (green arrow) as her second choice, as indicated by the colors of each arrow (de Nooy, Mrvar, Batagelj, & Granovetter, 2012).

Now that you know the concepts nodes (or vertices, sites, actors), degrees and lines (or edges, arcs, links, bonds, ties, relations), we will look at some basic analysis and representation options for the network.

Step 2: Download and open the dataset file 'DiningTablePartners.gephi' in Gephi. Open the file as a directed network when asked.

As you can see the graphical representation of the network needs some work. First, we want to change the layout of the nodes, and then we want to add labels to the nodes.

Step 3: Choose the 'Force Atlas' layout algorithm under the layout tab and run it until the network has become stable in its new form.

Step 4: Click on the 'Show node labels' option at the bottom of the screen and adjust the size with the 'Font size scale' so that it looks in proportion.

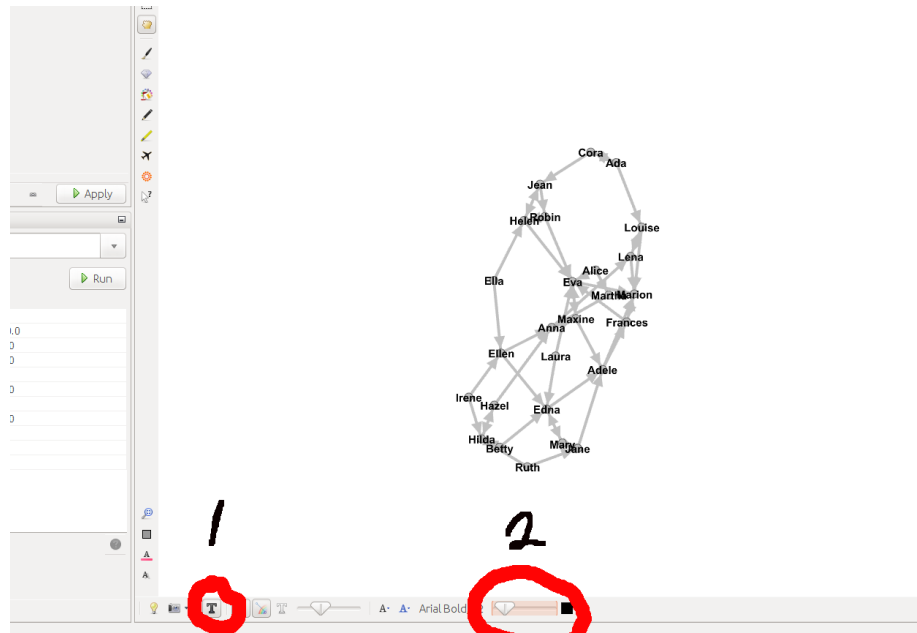


Fig 2: Tools to show labels and change their size.

Because this is a directed graph, we can calculate the Degree, In-Degree and Out-Degree values.

Step 5: Run the ‘Average degree’ option under the ‘Statistics’ tab.

Remark: After each initiated calculation, these results can also be found and edited in the ‘Data Laboratory’.

Question 1:

- Give the average out-degree value and comment why this is a logical value?
- Given the in-degree distribution, how many girls have the highest value, and what is this value?
- What does the in-degree value mean in the context of this network?
- Who is the most isolated person in the group? How did you define isolation using the degrees?
- If we want to help the girls to change their routine and be more active, we need to know who are the most important nodes in the network, and try to make sure that these people are being affected by some sort of health lifestyle program. Who would you pick from this group? Explain the reasons and argue in favor of your choice(s).

Now we want to give the nodes different sizes corresponding to their in-degree values.

Step 6: Under the ‘Ranking’ tab, under ‘Nodes’, select In-Degree as input. Next to the ‘Nodes’ tab you can select four ways to display the input in the network, which are by color, by size/weight, by label color, and by label size. Select the size/weight option. For minimum size we select 2 and for maximum size we choose 10. Then click ‘Apply’.

Question 2:

- a. Does this representation helps you to find who are the most isolated or popular nodes?
- b. Which other metrics would you use to define the position of a person in a network?

Now, we want to know how many, and which nodes, clusters of nodes, or components, in the directed network are **strongly connected**. A directed network is strongly connected if there is a path from each node in the graph to every other node, specifically a path in each direction, a path from a to b and a path from b to a . Identifying strongly connected components gives an idea about how information might flow. E.g. if someone had some piece of gossip, how would this information spread? For calculating the connected components, Gephi uses a depth-first search algorithm written by Robert Tarjan (1972).

The **weakly connected components** are the set of nodes that have a path from each node to the others in a “relaxed” set up of the network. The relaxed condition is basically found ignoring the direction of the arrows, and treating all connections as bidirectional. Then the same is calculated as before: if there is a path from each node in the graph to every other node, specifically a path in each direction, a path from a to b and a path from b to a then we have a weakly connected component.

Step 7: Run the ‘Connected Components’ option under the ‘Statistics’ tab.

The connected components can be graphically represented in the network, which will be our next step.

Step 8: Under the ‘Partition’ tab, choose the ‘Nodes’ option. Select the refresh option so the connected component analysis becomes available in the selection menu next to it. Select the ‘Strongly-Connected ID’ from the selection menu. Now click ‘apply’.

Remark: By right-clicking near the colors of the different components, it is possible to randomize colors, if you want to change them. Also you can click on each color to select specific colors for each component.

Question 3:

- a. How many strongly connected components (or subgraphs) are found?
- b. There is only one weakly connected component. Can you elaborate on the reason why there is only one?
- c. Looking at the colors of the nodes, is that possible to identify isolated nodes in the network? Are they different from what you got in the previous question?
- d. Is that possible to identify important nodes that can play a very influential role in the network using the colors of the components? Elaborate on how to do that.

Now we would like to introduce the notion of centrality of nodes within a network. There are various types of measures of the centrality of a node that determines the relative importance of a node in the network. In the current network the centrality of the node would represent the popularity of the girl within the group. Various methods of calculating the centrality include:

- **Degree centrality:** the number of edges a node has determines the centrality value;
- **Closeness centrality:** the sum of the shortest paths of node s to all other nodes determines the closeness centrality value;
- **Betweenness centrality:** vertices that have a high probability to occur on a randomly chosen shortest path between two randomly chosen nodes have a high betweenness centrality; and
- **Eigenvector centrality:** relative scores are assigned to all nodes in the network based on the concept that connections to high-scoring nodes contribute more to the score of the specific node than equal connections to low-scoring nodes.

For this assignment we will use the **betweenness centrality measure** for determining the sizes of the nodes. This leads us to the next step.

Step 9: Run the 'Avg. Path Length' option under the 'Statistics' tab. When asked choose the 'Directed' option and leave the 'Normalize Centralities in [0,1]' option unchecked. Within this calculation the Betweenness Centrality, Closeness Centrality and the Eccentricity are calculated.

Step 10: Under the 'Ranking' tab, choose 'Nodes' and choose the size/weight option. Choose 'Betweenness Centrality' from the selection menu. After setting the minimum size to 2 and the maximum size to 10, click 'Apply'.

Remark: Earlier we used the In-Degree values to determine the size of the nodes, and now we let the Betweenness Centrality value to determine the nodes. Mind the changes of the size of some of the nodes, that became bigger or smaller after step 9. If you did not see the changes, you can reselect In-Degree and click 'Apply' again, to see them. Do not forget to select Betweenness Centrality as rank afterwards.

Question 4:

- a. Who has the highest betweenness centrality value, and what is this value?
- b. Can you explain the changes in size of the nodes between the In-Degree measure and the Betweenness Centrality measure?
- c. If you aim to find influential people in the network, who would you consider choosing based on the centrality?
- d. If you aim to find isolated people in the network, who would you consider choosing based on the centrality?
- e. Elaborate about your choices using the different metrics for isolated nodes and for influential people in the network and propose a good method to identify them based on the metrics studied here. Which metric is the most important in your opinion? Is there a way to combine different metrics to find this nodes? How would you do that?

Finally, we want to output the network in PNG format. If you want to learn more about layout and visualization options, you can look at the Tutorial about Layouts and the Tutorial about visualization.

Step 11: Click on the ‘Preview’ tab at the top of the screen. Choose the ‘Show Labels option’ under ‘Node Labels’. Under ‘Node Labels’ you can also adjust the font, and font-size. Unclick the proportional size, to unrelate the size of the nodes with size of the labels. Click ‘export: SVG/PDF/PNG’, give the output file a name and choose PNG as output format. Copy the PNG file in your document where you wrote down the answers to the questions. Probably your output will look similar to figure 2.

Question 1:

- a. How many nodes are in this network? How many relations are in this network?
 - b. What is the average degree? Which person has the most connections / degrees?
 - c. Who is the most influential person in the network according to Betweenness centrality calculation?
 - d. How would you rank people if you were trying to target a part of the network for advertisement of a new device for health lifestyle? Who would be the best candidates? Elaborate about your strategy.
 - e. Imagine we have the information about the exchange of messages between all the nodes in the networks. How can we use this information to assess the strength of the connections? How this information would change your marketing strategy in question 1d?
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Step 2: Turn on the labels for each node and adjust the size of the labels to your liking.

Step 3: Apply a node ranking calculation that calculates the size for each node related to the Betweenness centrality for each node.

Now we introduce a method which extracts the community structure of large networks. It is a heuristic method based on modularity optimization. For more information about this method, see Blondel, Guillaume, Lambiotte & Lefebvre (2008).

Step 4: Click on the calculation option 'Modularity' under 'Statistics', with the options 'Randomize' on and with resolution '1.0'.

Step 5: Apply a node partition calculation that attributes the color for each community to each node.

Question 2:

- a. How many communities are in this network?
- b. How big is the largest community (how many nodes)?
- c. Consider you are doing a study about spread of opinions about some TV show competition on Facebook. You have 4 candidates to win the show, and is willing to understand how the opinion about the candidates and their support evolves in a web media based data set. You can collect what people are sharing and the reactions to others' opinions. How could you analyse this data set considering that there are communities not connected but interacting at the same time? How would you evaluate the spread of opinions using a data set like this one?

Step 6: Play with the ‘Force atlas’ layout algorithm until you find a suitable graphical layout for the network. Also play with the ‘attraction distribution’ and ‘adjust by sizes’ options.

Step 7: Output the network to PNG format when you are pleased with the graphical representation of the network. Copy the PNG file in your document where you wrote down the answers to the questions.

Step 8: Save the document with all your answers and graphical output and submit it in Canvas.

Possibly your output will look similar to figure 3.

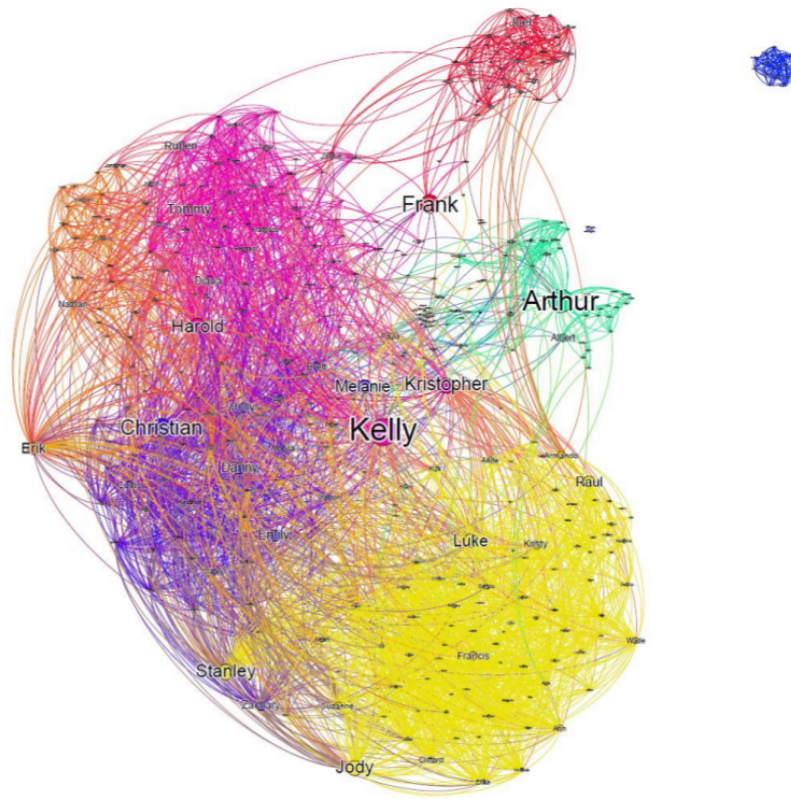


Fig 3: Facebook network

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