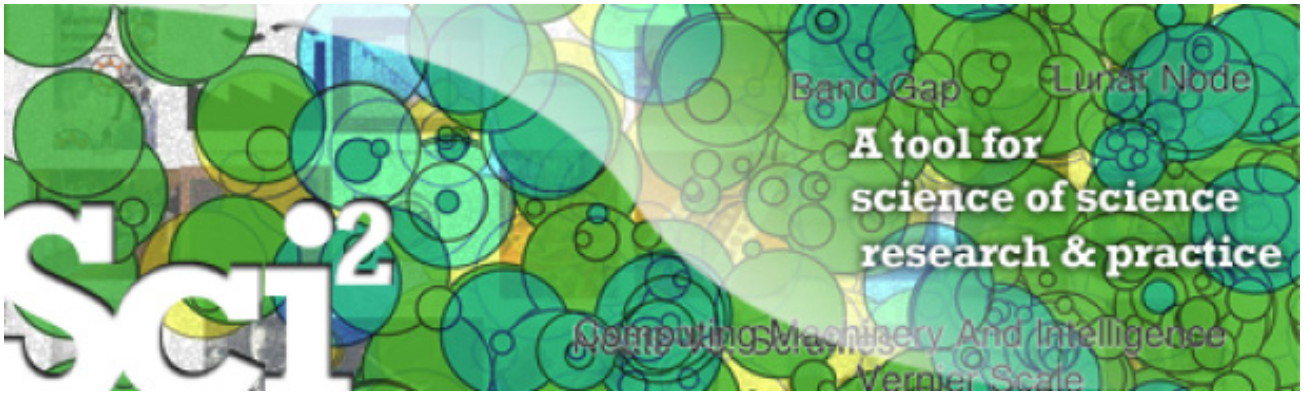
Visualizing Data with the Science of Science (Sci2) Tool



Ted Polley
Research & Editorial Assistant
Cyberinfrastructure for Network Science Center
School of Informatics and Computing
Indiana University Bloomington
http://cns.iu.edu

Presentation Overview

Why should we visualize?
How should we visualize?
Introduction to Sci2
- Introduction
- Macroscopes
- OSGi & Cyberinfrastructure Shell
- Types and levels of analysis
- File formats supported by Sci2
- User Interface
- Supported tools
- Visualizations
- Sci2 Adoption

Brainstorming Session
Break for lunch

Hands-on with Sci2
- Installing Sci2
- Needs Driven Workflow Design
- Introduction to Networks
- Visualizing the Florentine Dataset
- Evolving Co-Authorship Networks
- Visualize Geographic Distribution of Clients
- Visualize Client Brand Networks

Discussion/Questions

Why should we Visualize?

Humans are surprisingly good at:

- Pattern matching – especially when it comes to identifying trends, gaps, and outliers. We can use this to predict the future.

- Determining placement, orientation, shape, size, color etc.

Ultimately, visualizations provide access to insight much quicker than simply examining the raw data, or even data that has been statistically analyzed. This insight can lead to informed decision making.

Adapted from Noah Iliinsky's Keynote address at the OCLC Symposium, *Four Pillars of Data Visualization*, at ACRL 2013 in Indianapolis, Indiana.

Why should we Visualize?

How should we Visualize?

Step 1. **Purpose**: identify a clear purpose and focus for the visualization

Step 2. **Content**: obtain the best possible data that suits the purpose

Step 3. **Structure**: identify the best structure for the visualization that suits the purpose

Step 4. **Format**: identity the best way to represent the data so that it is easily digested

Adapted from Noah Iliinsky's Keynote address at the OCLC Symposium, *Four Pillars of Data Visualization*, at ACRL 2013 in Indianapolis, Indiana.

Introduction to Sci2

The Science of Science (Sci2) Tool is an open-source modular toolset originally designed for the study of science. However it has many uses that support temporal, geospatial, topical, and network analysis and visualization of scholarly datasets.

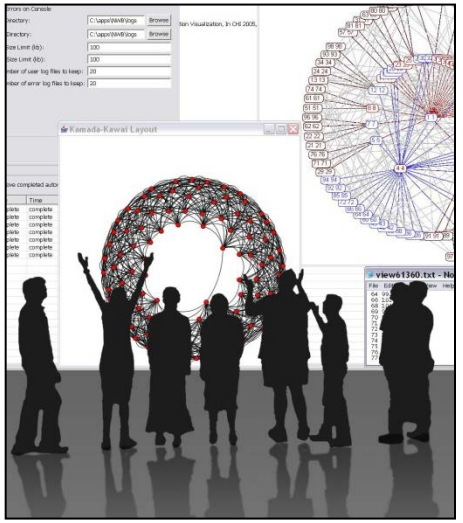
Macroscopes

Decision making in science, industry, and politics, as well as in daily life, requires that we make sense of the massive amounts of data that result from complex systems.

Rather than making things larger or smaller, **macroscopes let us observe what is too great, slow, or complex for us to comprehend or sometimes even notice.**

Microscopes

Telescopes



Macroscopes

Plug-and-Play Macroscopes

While microscopes and telescopes are physical instruments, macroscopes are **continuously changing bundles of software plugins**

Macroscopes make it easy to

- Simply drop plugins into the tool and they appear in the menu, ready to use
- Sharing algorithm components, tools, or novel interfaces becomes as easy as sharing images on Flickr or videos on YouTube

OSGi & Cyberinfrastructure Shell (CIShell)

- CIShell (http://cishell.org) is an open source software specification for the integration and utilization of datasets, algorithms, and tools
- It extends the Open Services Gateway Initiative (OSGi) (http://osgi.org), a standardized, modularized service platform
- CIShell provides “sockets” into which algorithms, tools, and datasets can be plugged using a wizard-driven process

OSGi & Cyberinfrastructure Shell (CIShell)

Developers

Alg

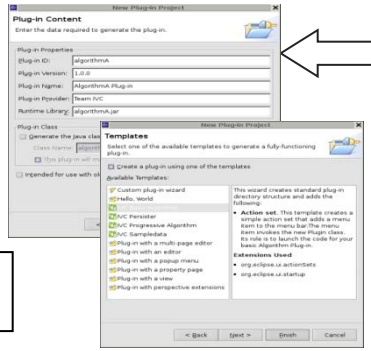
Alg

Alg

Tool

Tool

CIShell Wizards

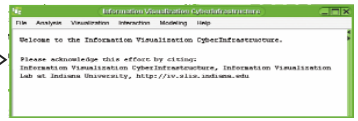


CIShell

Users

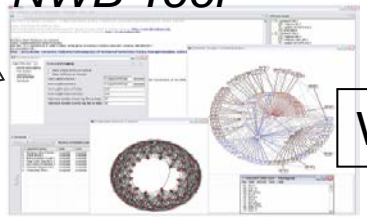
Workflow

Sci2Tool



Workflow

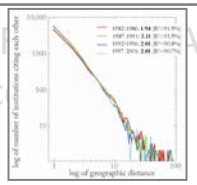
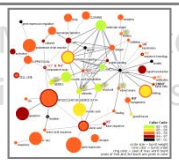
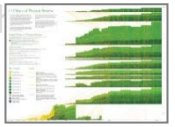
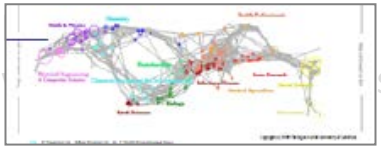
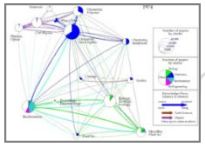
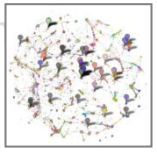
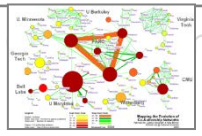
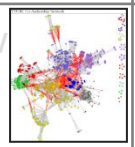
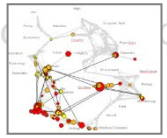
NWB Tool



Workflow

Workflow

Type of Analysis vs. Level of Analysis

	Micro/Individual (1-100 records)	***Meso/Local (101–10,000 records)***	***Macro/Global (10,000 < records)***
Statistical Analysis/Profiling	Individual person and their expertise profiles	Larger labs, centers, universities, research domains, or states	All of NSF, NIH, NEA, all of science 
Temporal Analysis (When)	Funding portfolio of one individual	 opic bursts s of PNAS	113 Years of PNAS Research 
Geospatial Analysis (Where)	Career trajectory of one individual	Mapping a sta intellectual la pe 	PNAS 
Topical Analysis (What)	 s.	 flows in research	VxOrd/Topic m NIH funding 
Network Analysis (With Whom?)	NSF C rk of one in 	 netw 	NIH's core c cy 

Sci2 Tool – Supported Data Formats

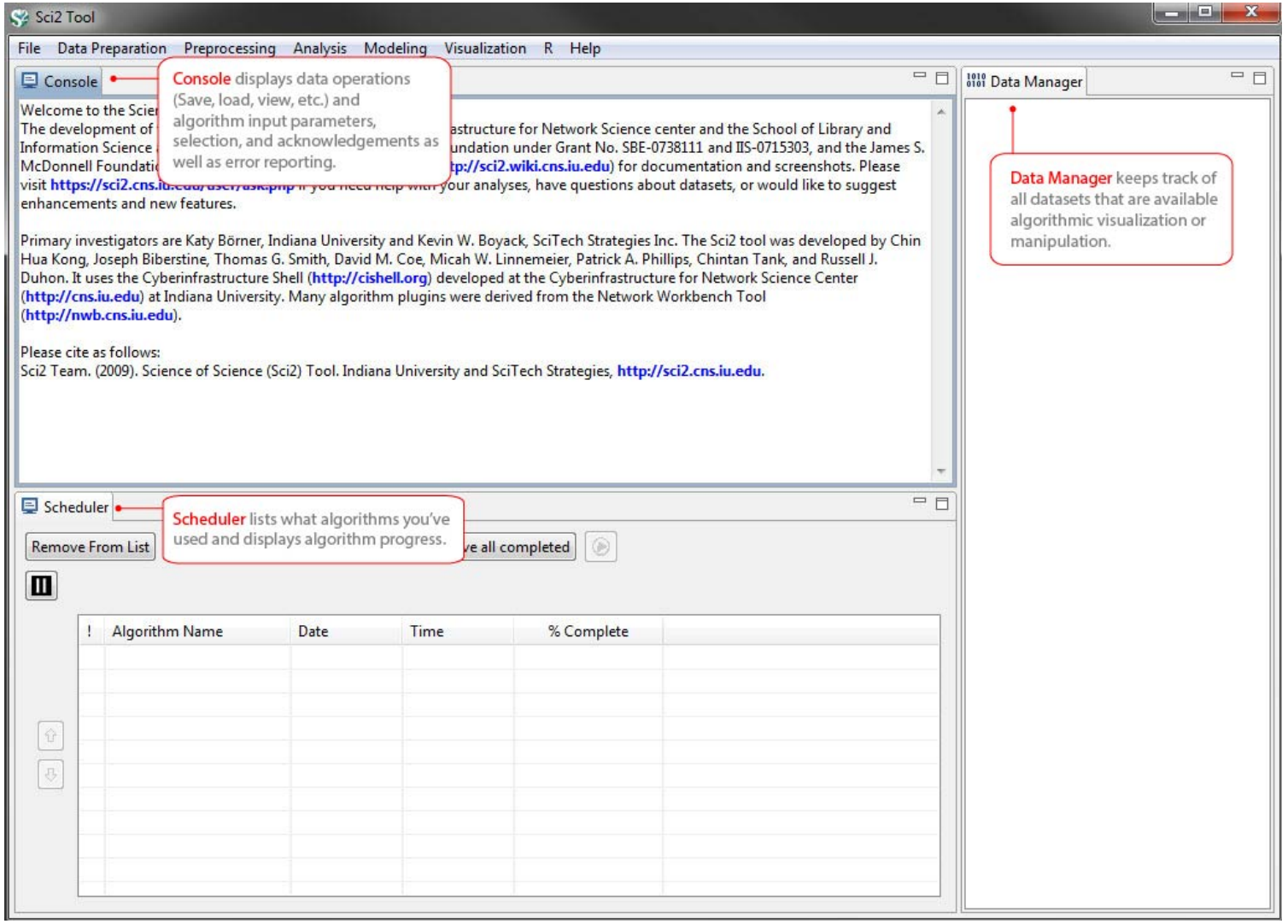
Input:

- Network Formats
- GraphML (*.xml or *.graphml)
- XGMML (*.xml)
- Pajek .NET (*.net)
- NWB (*.nwb)
- Scientometric Formats
- ISI (*.isi)
- Bibtex (*.bib)
- Endnote Export Format (*.enw)
- Scopus csv (*.scopus)
- NSF csv (*.nsf)
- Other Formats
- Pajek Matrix (*.mat)
- TreeML (*.xml)
- Edgelist (*.edge)
- CSV (*.csv)

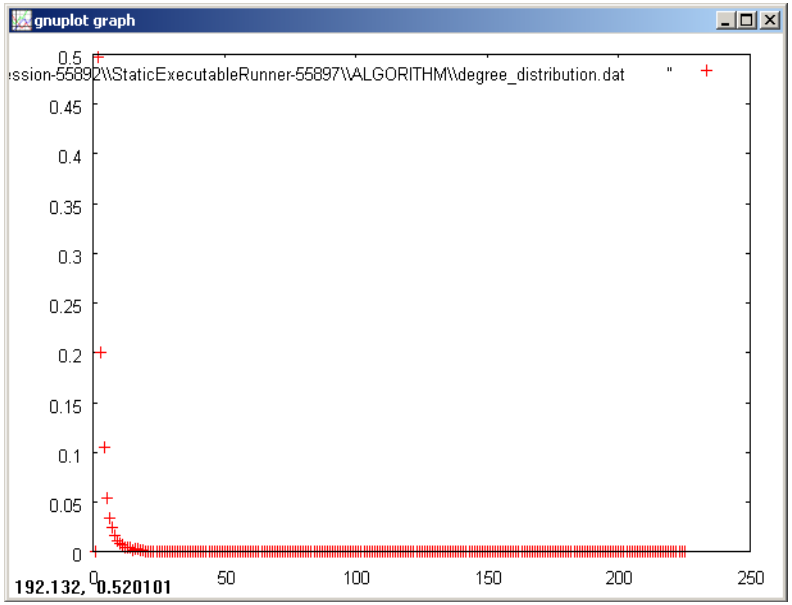
Output:

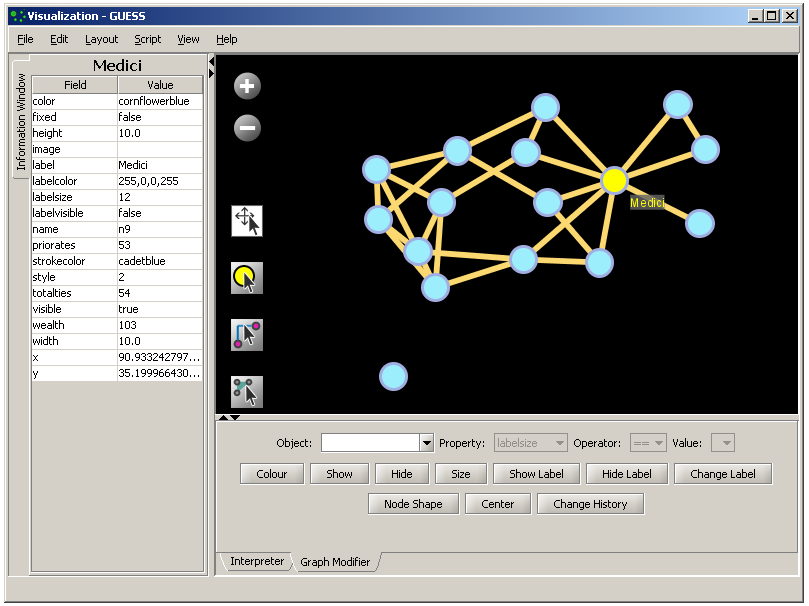
- Network File Formats
- GraphML (*.xml or *.graphml)
- Pajek .MAT (*.mat)
- Pajek .NET (*.net)
- NWB (*.nwb)
- XGMML (*.xml)
- CSV (*.csv)
- JPEG (*.jpg)
- PDF (*.pdf)
- PostScript (*.ps)

User Interface



Supported Tools





Gnuplot

portable command-line driven interactive data and function plotting utility http://www.gnuplot.info/.

GUESS

exploratory data analysis and visualization tool for graphs and networks.

https://nwb.slis.indiana.edu/community/?n=VisualizeData.GUESS.

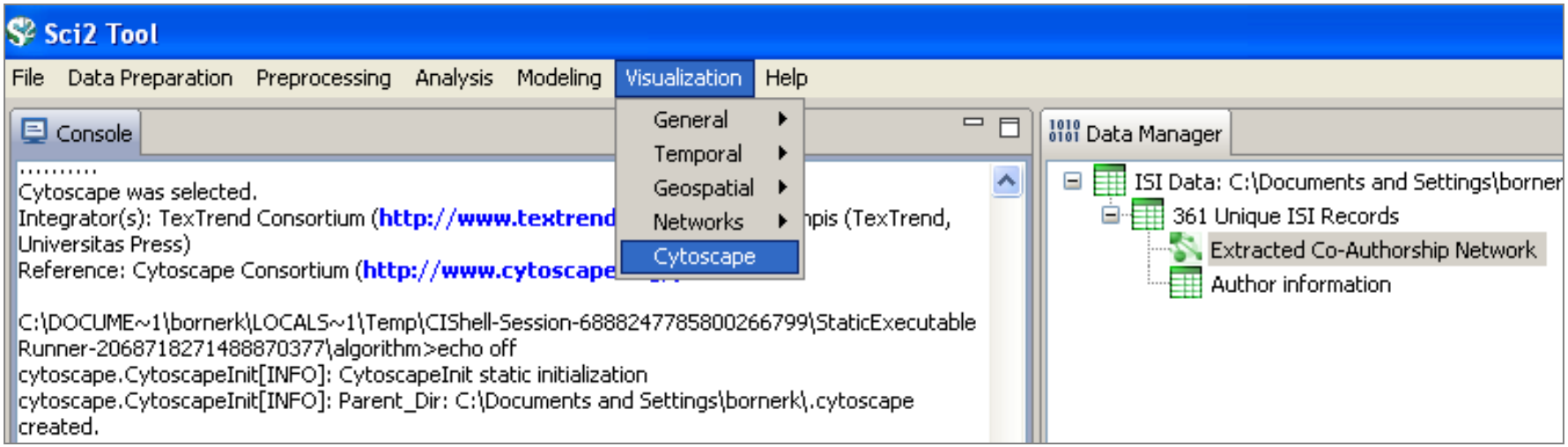
Supported Tools

Adding more layout algorithms and network visualization interactivity via Cytoscape http://www.cytoscape.org.

Simply add *org.textrend.visualization.cytoscape_0.0.3.jar* into your /plugin directory.

Restart Sci² Tool

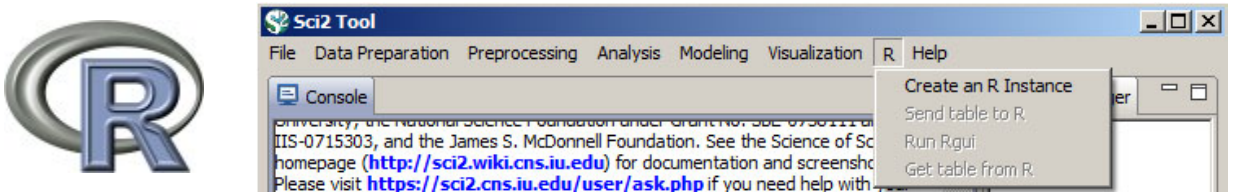
Cytoscape now shows in the Visualization Menu



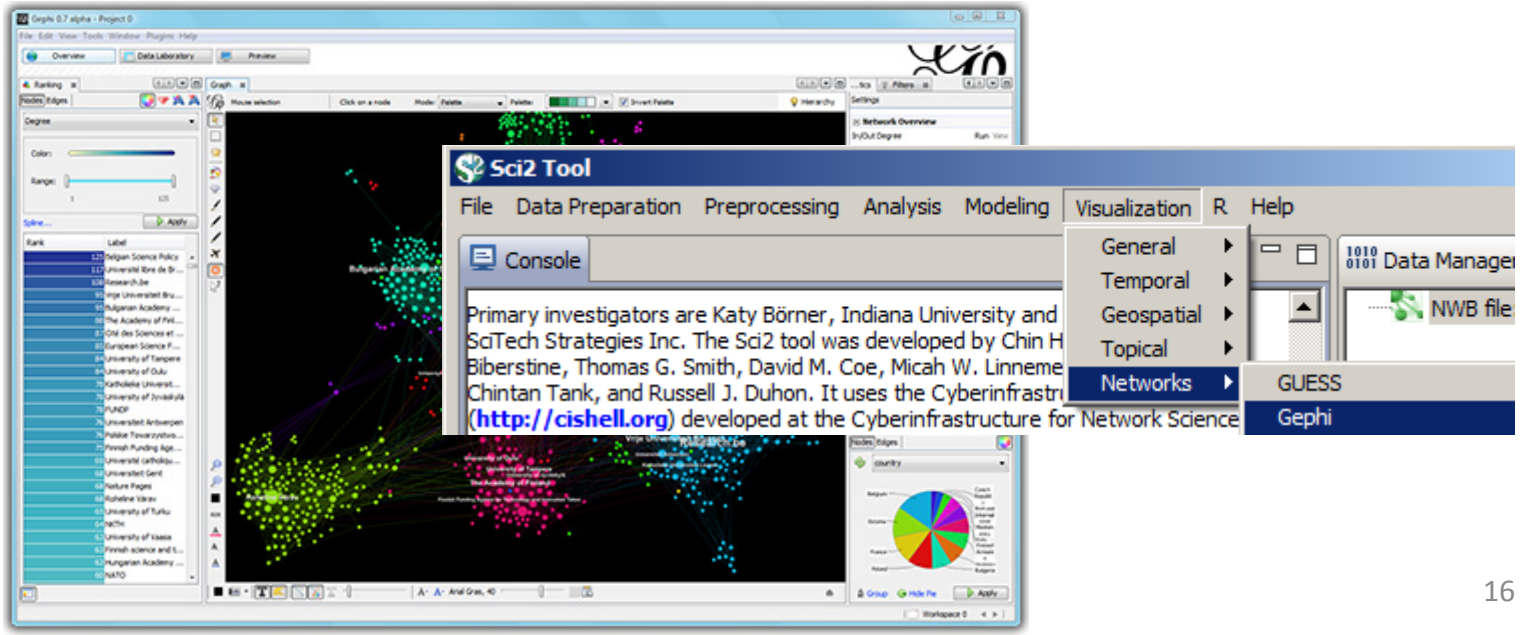
Select a network in Data Manager, run Cytoscape and the tool will start with this network loaded.

Bridged Tools

R statistical tool

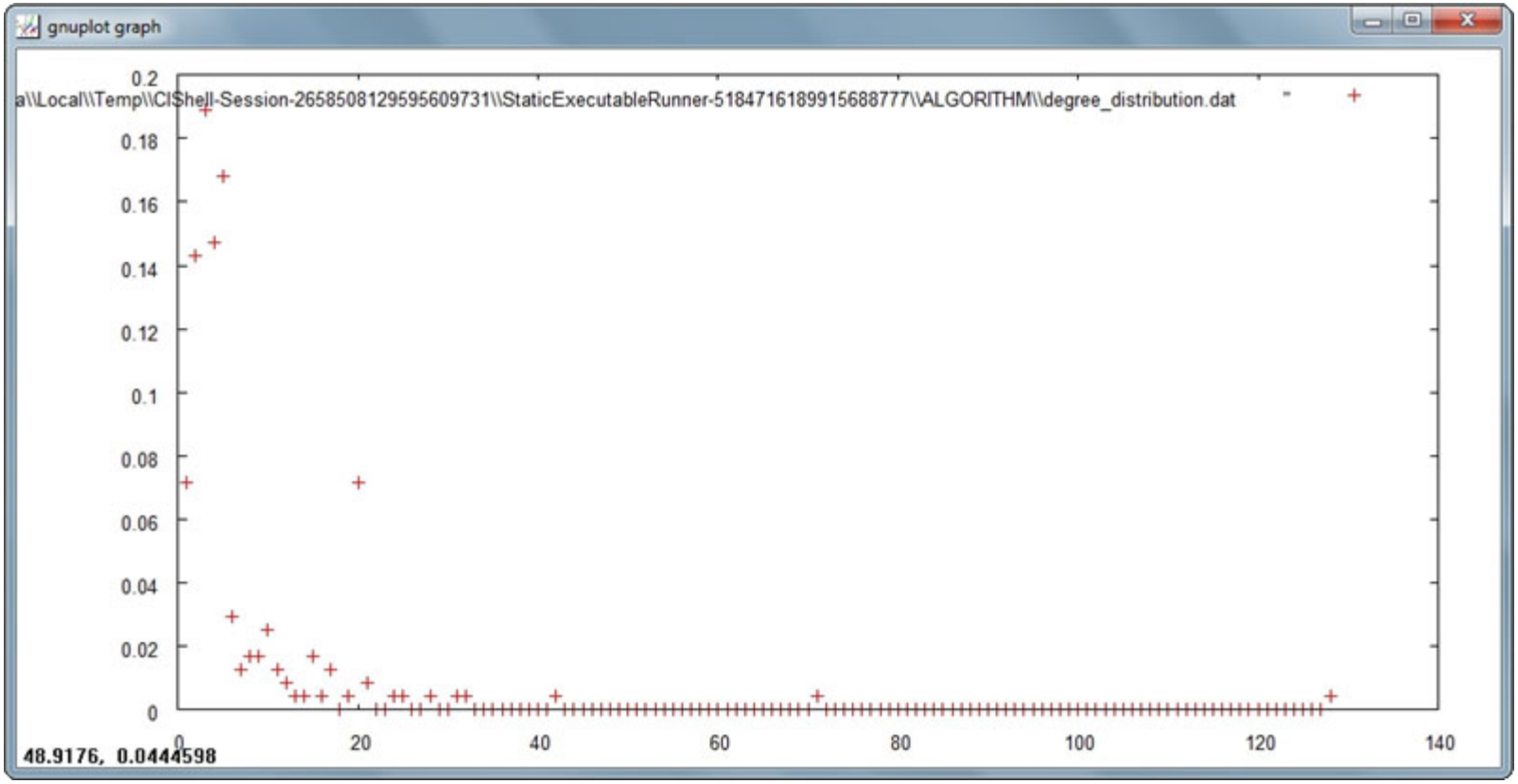


Gephi visualization tool



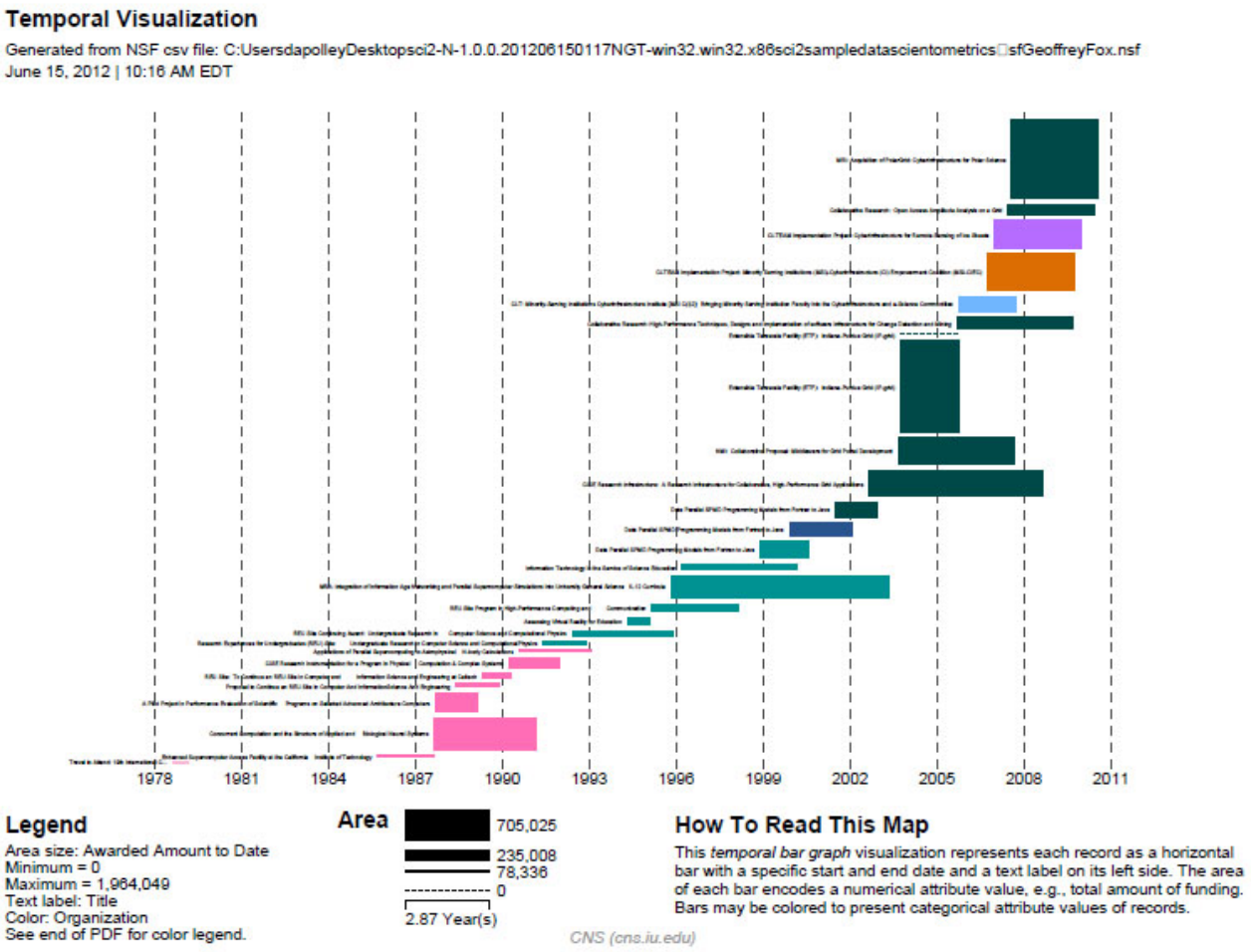
Sci2 Visualizations: *General*

Use GnuPlot to visualize the degree distribution of a co-authorship network extracted from ISI data…



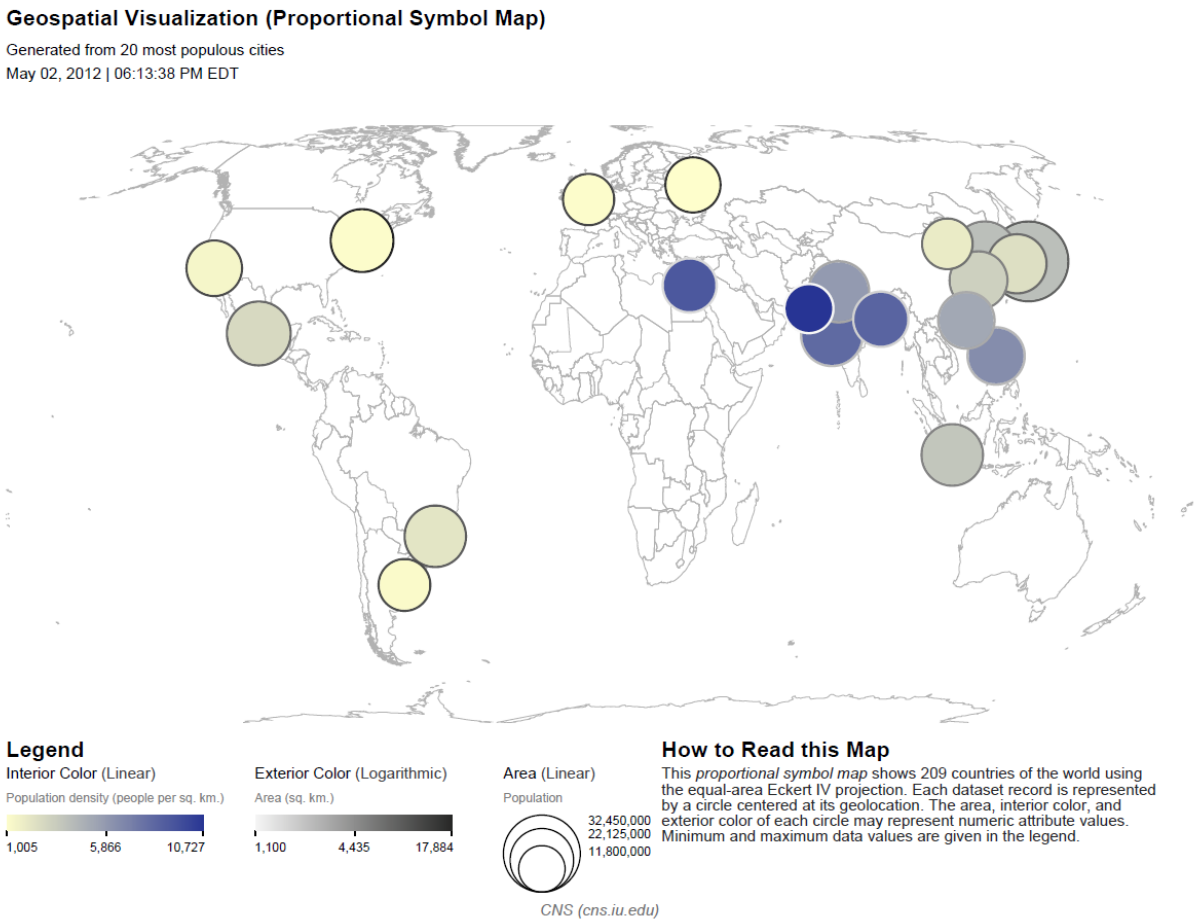
Sci2 Tool Visualizations: *Temporal*

Use Temporal Bar Graph to visualize NSF funding profiles over time…



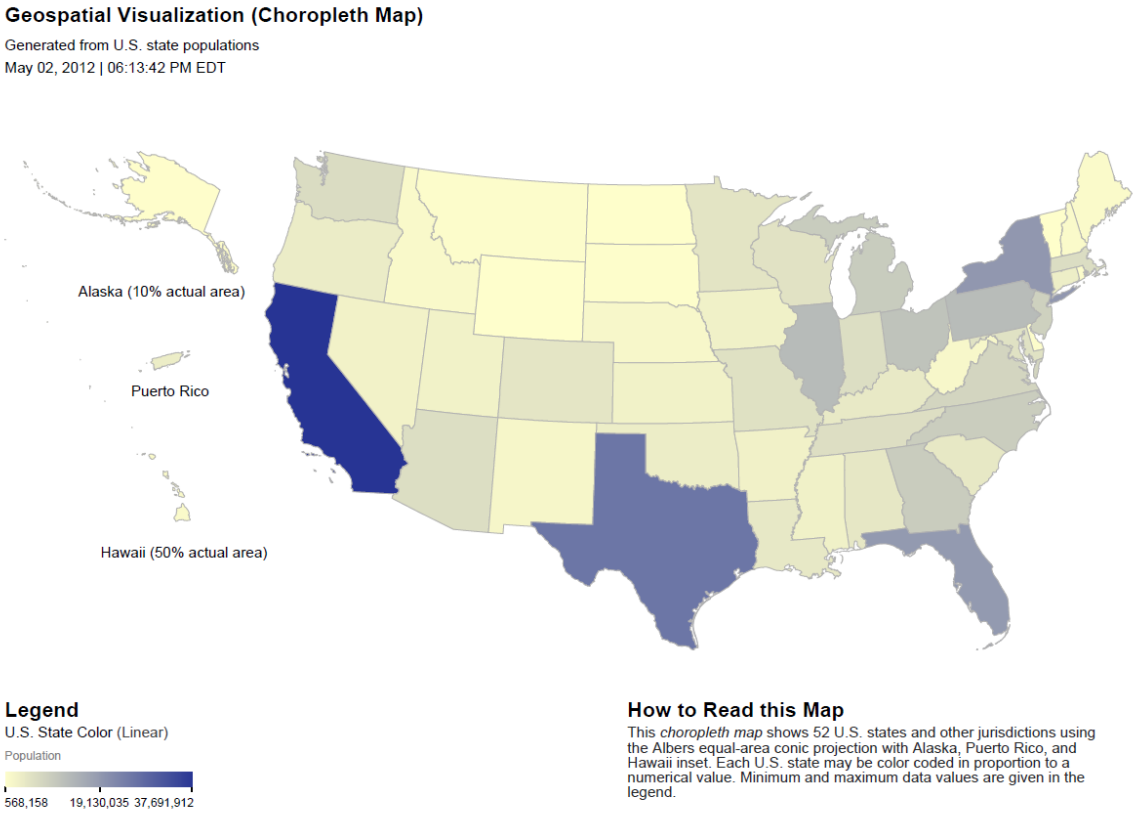
Sci2 Tool Visualizations: *Geospatial*

Use the Proportional Symbol Map to size and color symbols proportionally to numeric data, in this case the 20 most populated cities around the world…



Sci2 Tool Visualizations: *Geospatial*

Use the Choropleth Map to color regions proportionally to numeric data, in this case the US by state population…



Sci2 Tool Visualizations: *Geospatial*

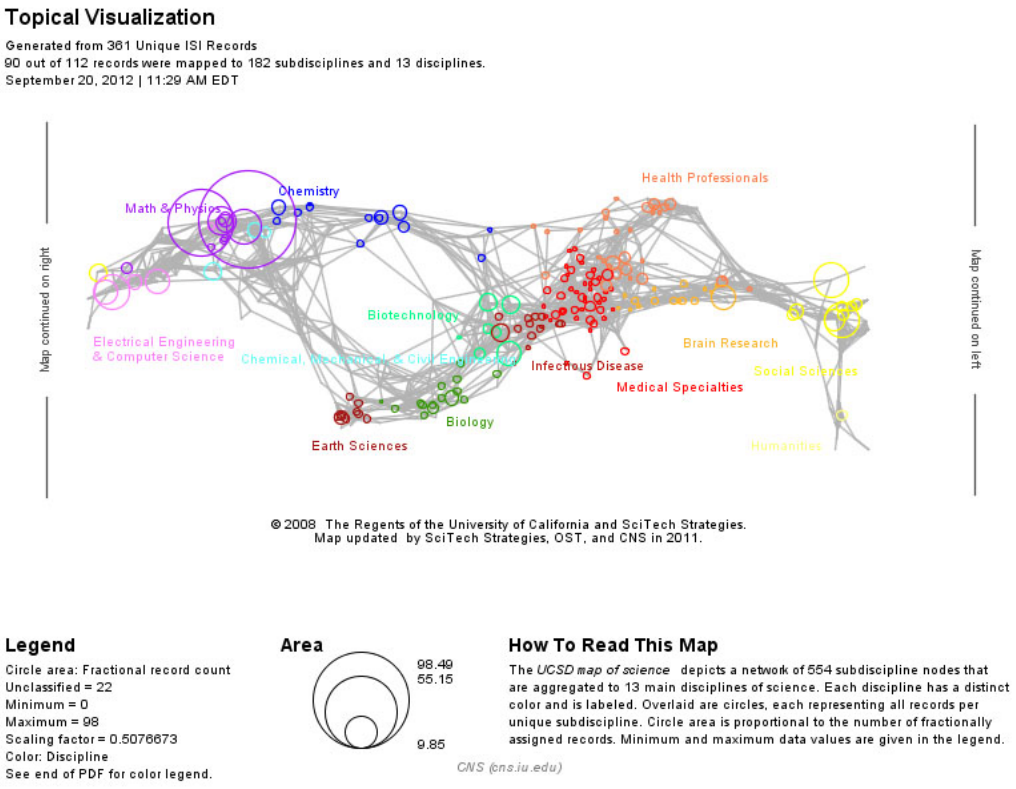
Overlay a geospatial network on a base map, in this case Albert-László Barabási and his collaborators…

Geo Map ()
Eckert IV Projection
Apr 19, 2012 | 11:14:48 AM

Created with Sci² Tool | Cyberinfrastructure for Network Science Center (http://cns.iu.edu)

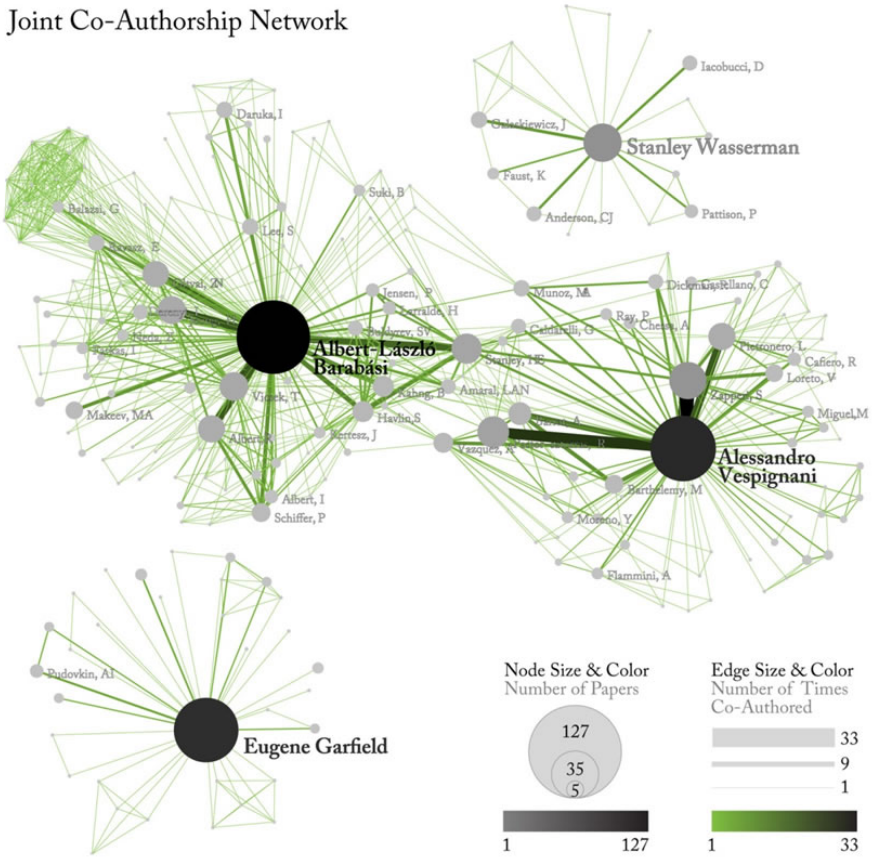
Sci2 Tool Visualizations: *Topical*

Use the Map of Science via Journals visualization a network drawn the result of mapping a dataset's journals to the underlying sub-discipline(s) those journals contain…



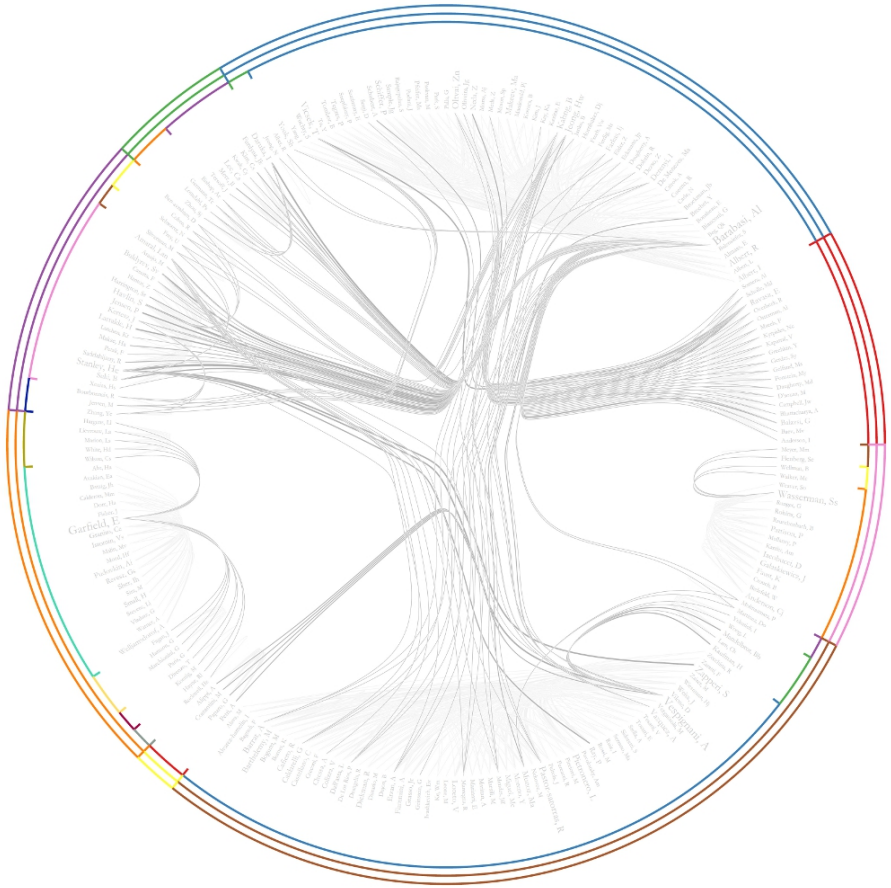
Sci2 Tool Visualizations: *Networks*

Use GUESS to visualize networks, such as this co-authorship network extracted from ISI data…



Sci2 Tool Visualizations: *Networks*

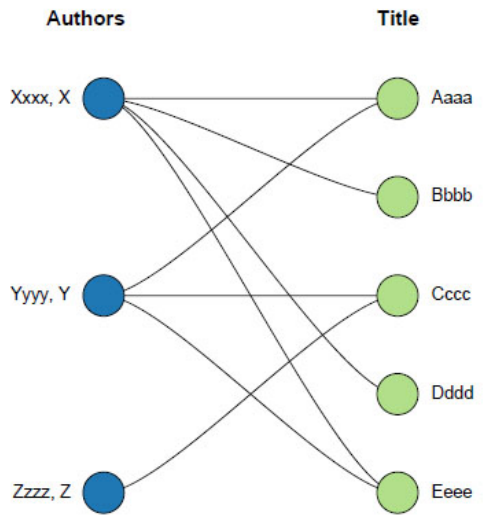
Use Circular Hierarchy to visualize networks with community attributes appended…



Sci2 Tool Visualizations: *Networks*

Use the Bipartite Network visualization to create a network of authors and publication titles…

Network Visualization
Generated from Bipartite network from Authors and Title.2
September 20, 2012 | 11:04 AM EDT



Legend
Sorted by
Left side:
Alphabetical
Right side:
Alphabetical

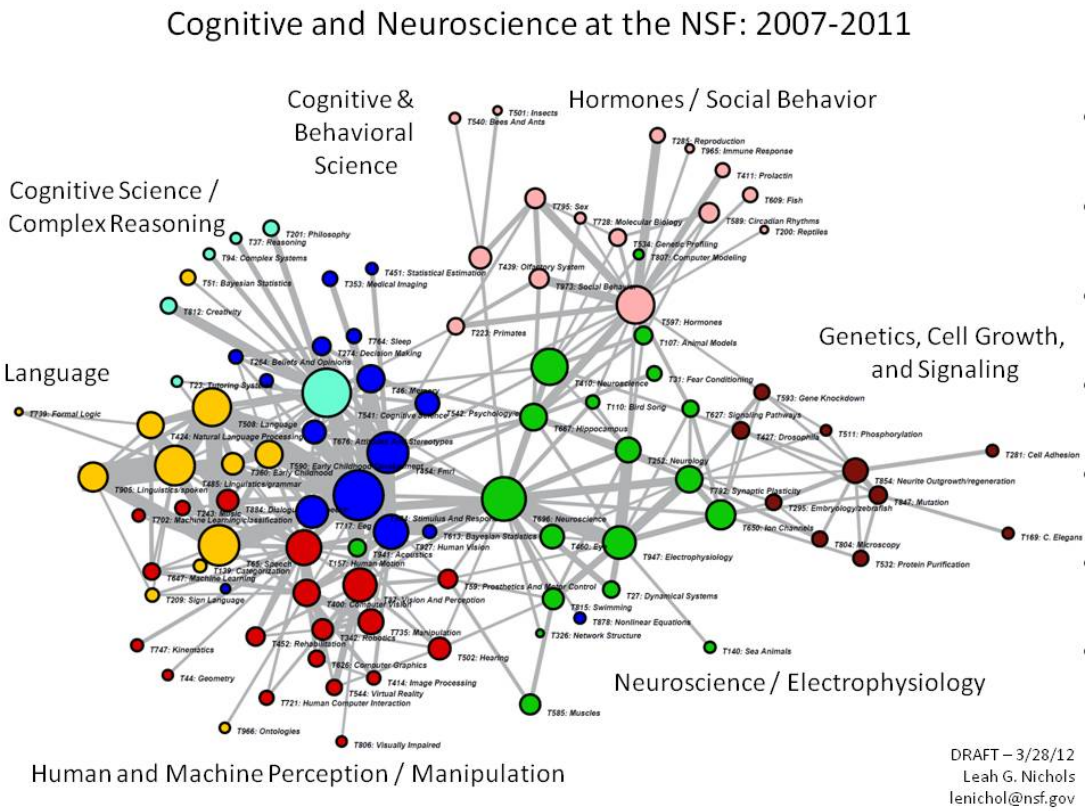
How To Read This Map
This *bipartite network* shows two record types and their interconnections. Each record is represented by a labeled circle that is size coded by a numerical attribute value. Records of each type are vertically aligned and sorted, e.g., by node size or alphabetically. Links between records of different type may be weighted as represented by line thickness.

CNS (cns.iu.edu)

Sci2 Adoption

Topic co-occurrence network of the 2885 cognitive and neuroscience NSF projects funded between 2007 and 2011.

The nodes are labeled based on how the awards were tagged. The nodes are scaled by number of awards (max = 355) with a particular tag and the edges are scaled on number of co-occurrences (max =91) of those tags. The node colors differentiate the different communities of awards, which allows you to identify topic areas.



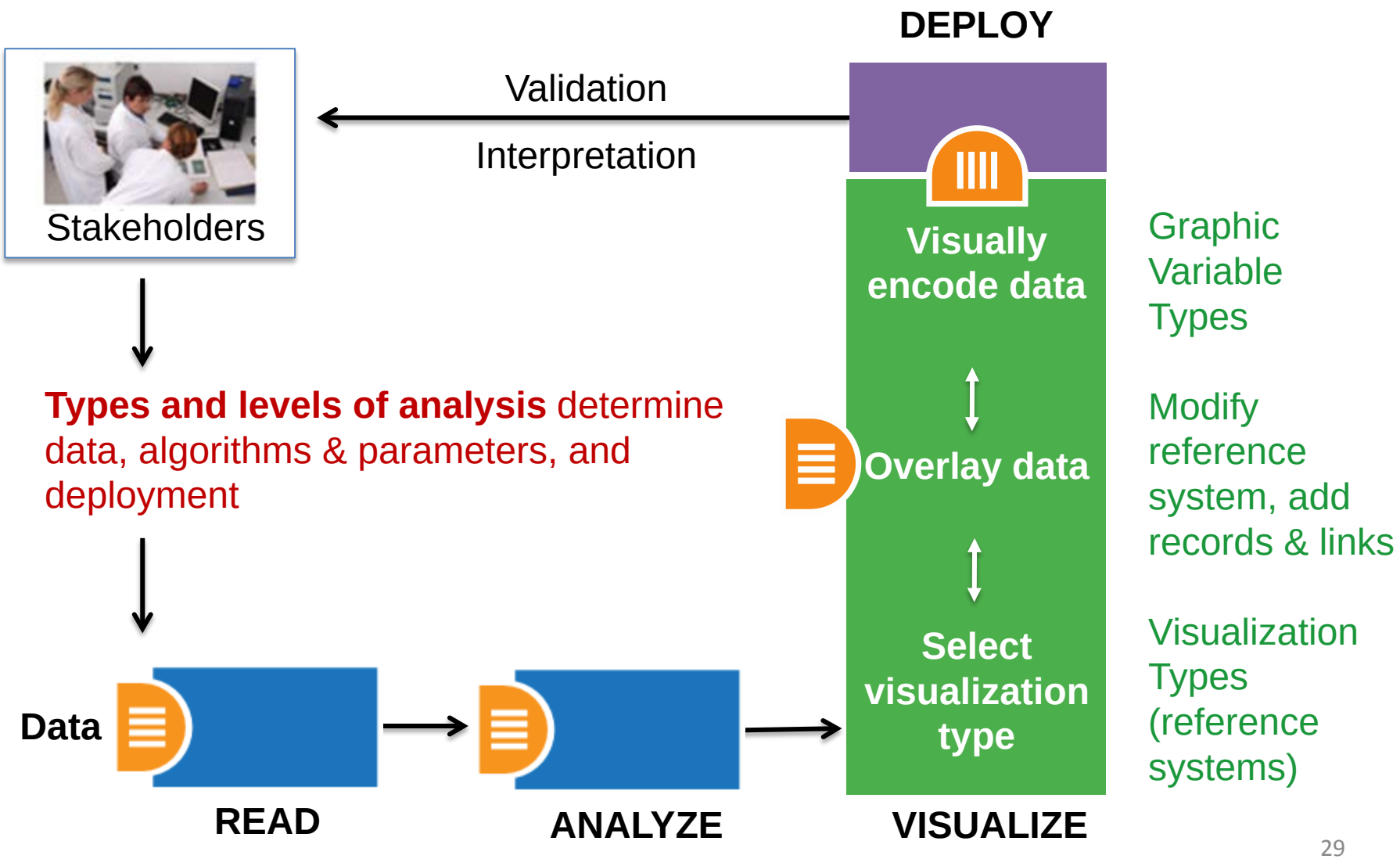
Sci2 Adoption

This is … an **entirely new way of characterizing and understanding the NSF portfolio**. This is in part because this enables **analysis of the content of the awards/proposals independent of the institutional structure.** One can quickly identify ALL of the Cog/Neuro awards throughout the entire NSF portfolio – so it captures research in all of the unexpected institutional places. This method also allows one to **easily identify areas of parallel or potentially collaborative research being funded by different institutional structures** and … to identify potential areas for advancing science by facilitating collaborations.

Leah G. Nichols, NSF

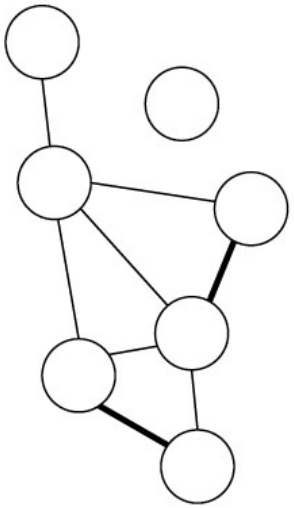
Questions?

Needs-Driven Workflow Design



Introduction to Networks

Undirected Networks



Nodes:

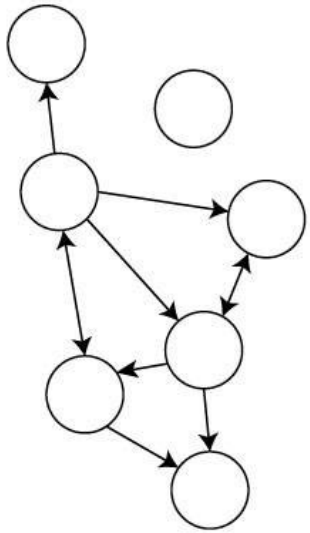
Edges:

Node Degree:
Number of edges connected to nodes

Isolates:
Nodes that are not connected to the rest of the network

Edge Weight:
Demonstrates relative importance of relationships

Directed Networks



Edge Direction:
Directional relationship is represented by arrows

In-Degree:
Number of incoming edges

Out-Degree:
Number of outgoing edges

Visualizing the Florentine Dataset

This example will demonstrate how to visualize data using Sci2. In this workflow we will be working with Padgett's Florentine families dataset which includes 16 different Italian families from the early 15th century. Each family is represented by a node in the network and families are connected by edges that represent either a marriage or business/lending ties. Each node (family) has several attributes: wealth (in thousands of lira), number of priorates (seats on the civic council between 1282-1344), and total ties (total number of business ties and marriages in the dataset).

“Substantively, the data include families who were locked in a struggle for political control of the city of Florence around 1430. Two factions were dominant in this struggle: one revolved around the infamous Medici family, the other around the powerful Strozzis.”

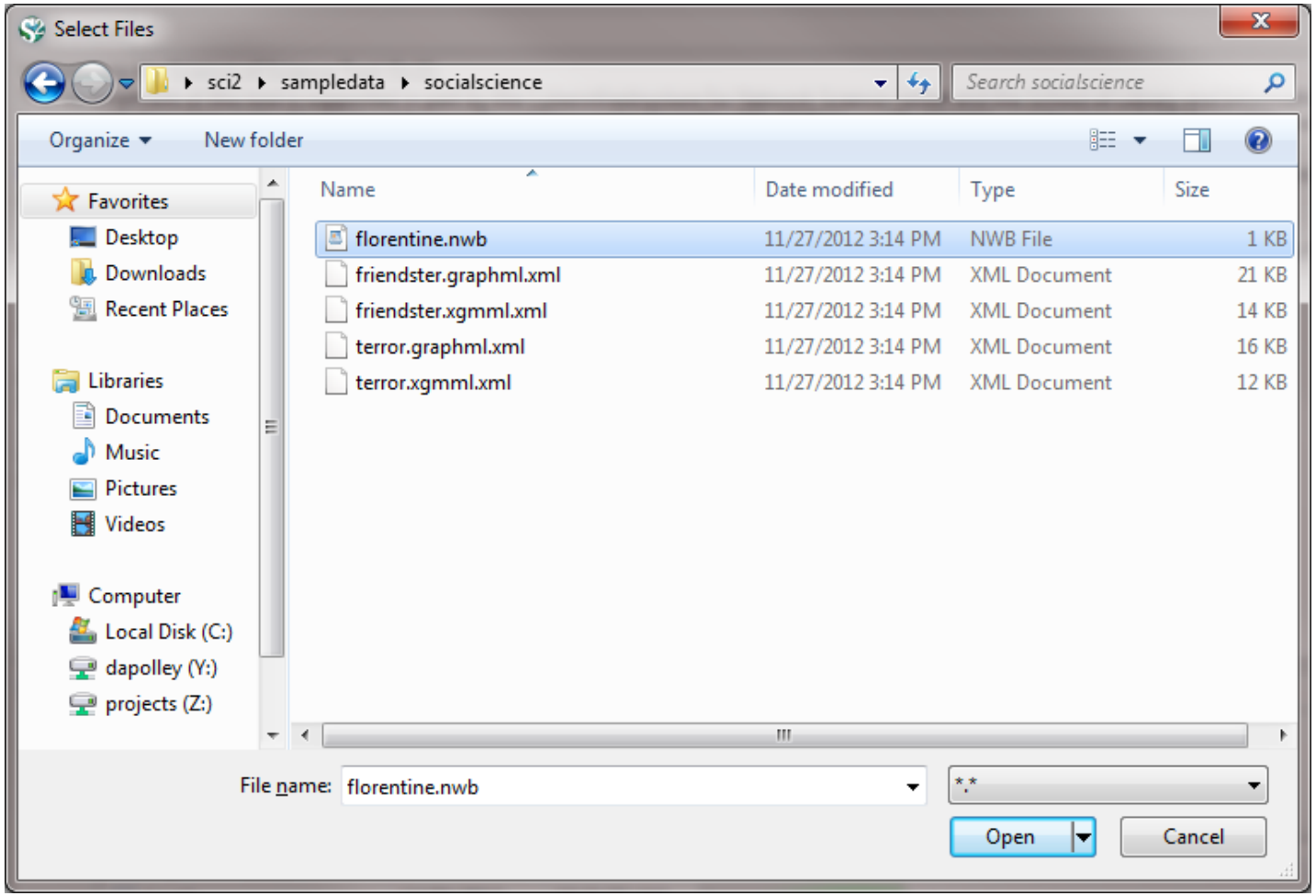
More info at http://svitsrv25.epfl.ch/R-doc/library/ergm/html/florentine.html

Visualizing the Florentine Dataset

```
*Nodes
id*int label*string wealth*int totalities*int priorates*int
1 "Acciaiuoli" 10 2 53
2 "Albizzi" 36 3 65
3 "Barbadori" 55 14 0
4 "Bischeri" 44 9 12
5 "Castellani" 20 18 22
6 "Ginori" 32 9 0
7 "Guadagni" 8 14 21
8 "Lamberteschi" 42 14 0
9 "Medici" 103 54 53
10 "Pazzi" 48 7 0
11 "Peruzzi" 49 32 42
12 "Pucci" 3 1 0
13 "Ridolfi" 27 4 38
14 "Salviati" 10 5 35
15 "Strozzi" 146 29 74
16 "Tornabuoni" 48 7 0
*UndirectedEdges
source*int target*int marriage*string business*string
9 1 "T" "F"
6 2 "T" "F"
7 2 "T" "F"
9 2 "T" "F"
5 3 "T" "T"
```

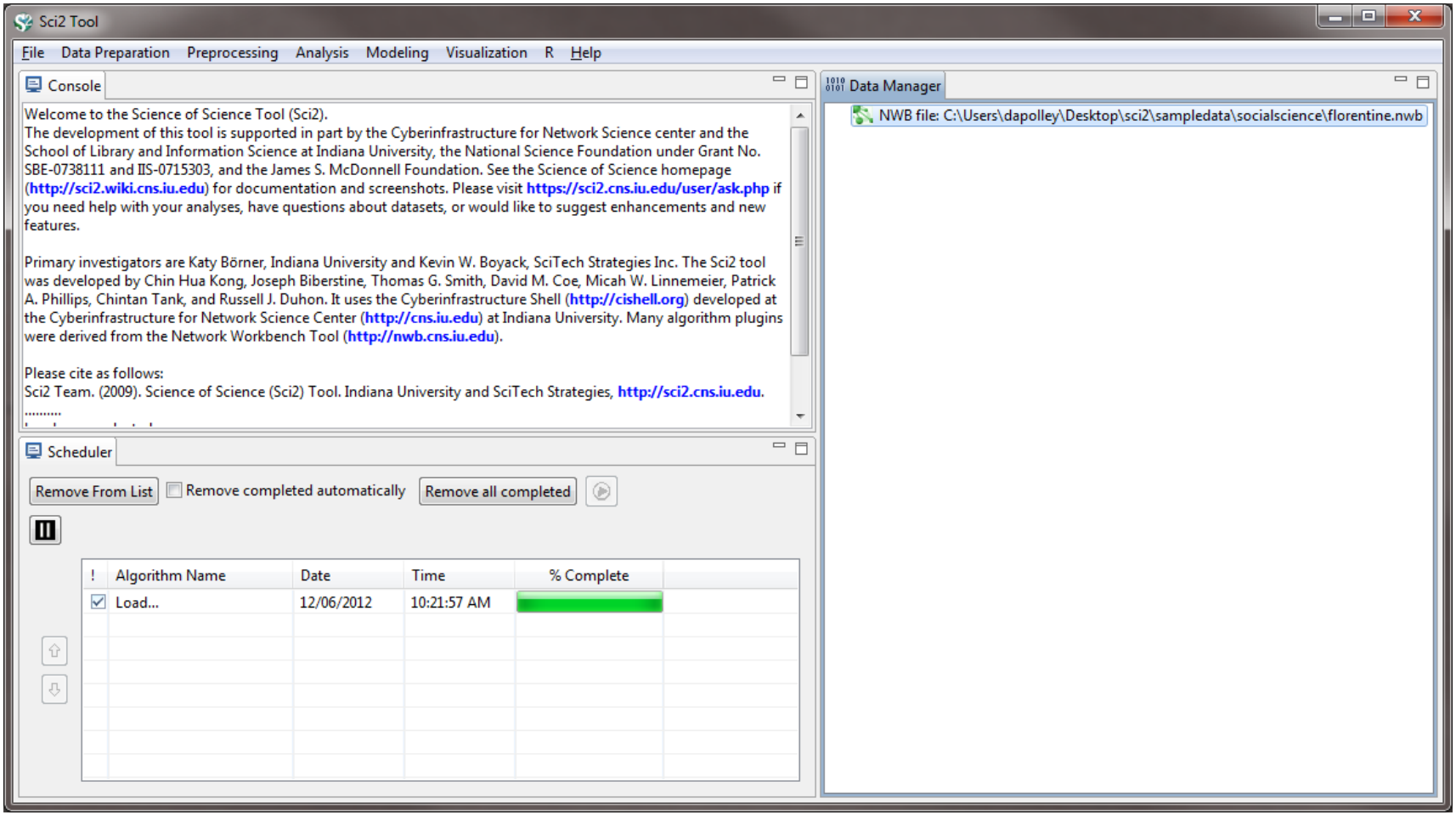
Visualizing the Florentine Dataset

First, load the florentine.nwb by following *File* > *Load* > ***yoursci2directory****/sampledata/scientometrics/endnote/florentine.nwb.*



Visualizing the Florentine Dataset

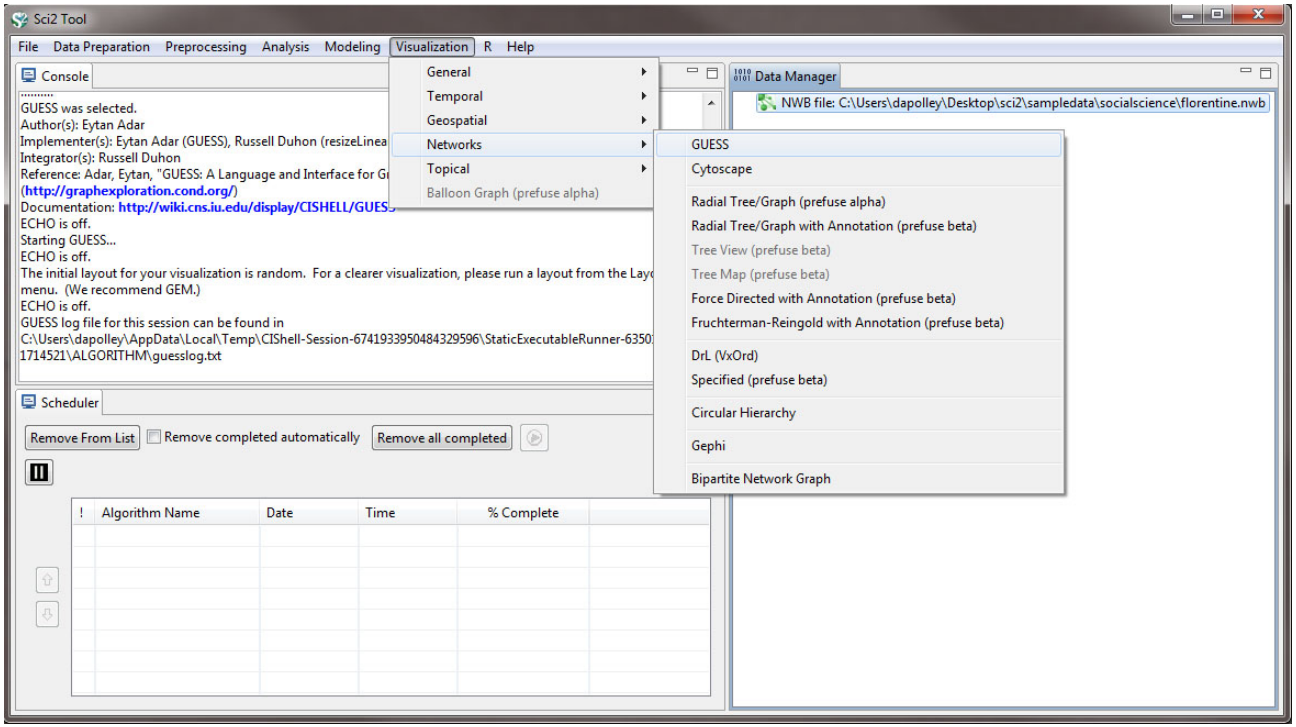
Once you have loaded the data in Sci2, it will appear in the Data Manager.



Visualizing the Florentine Dataset

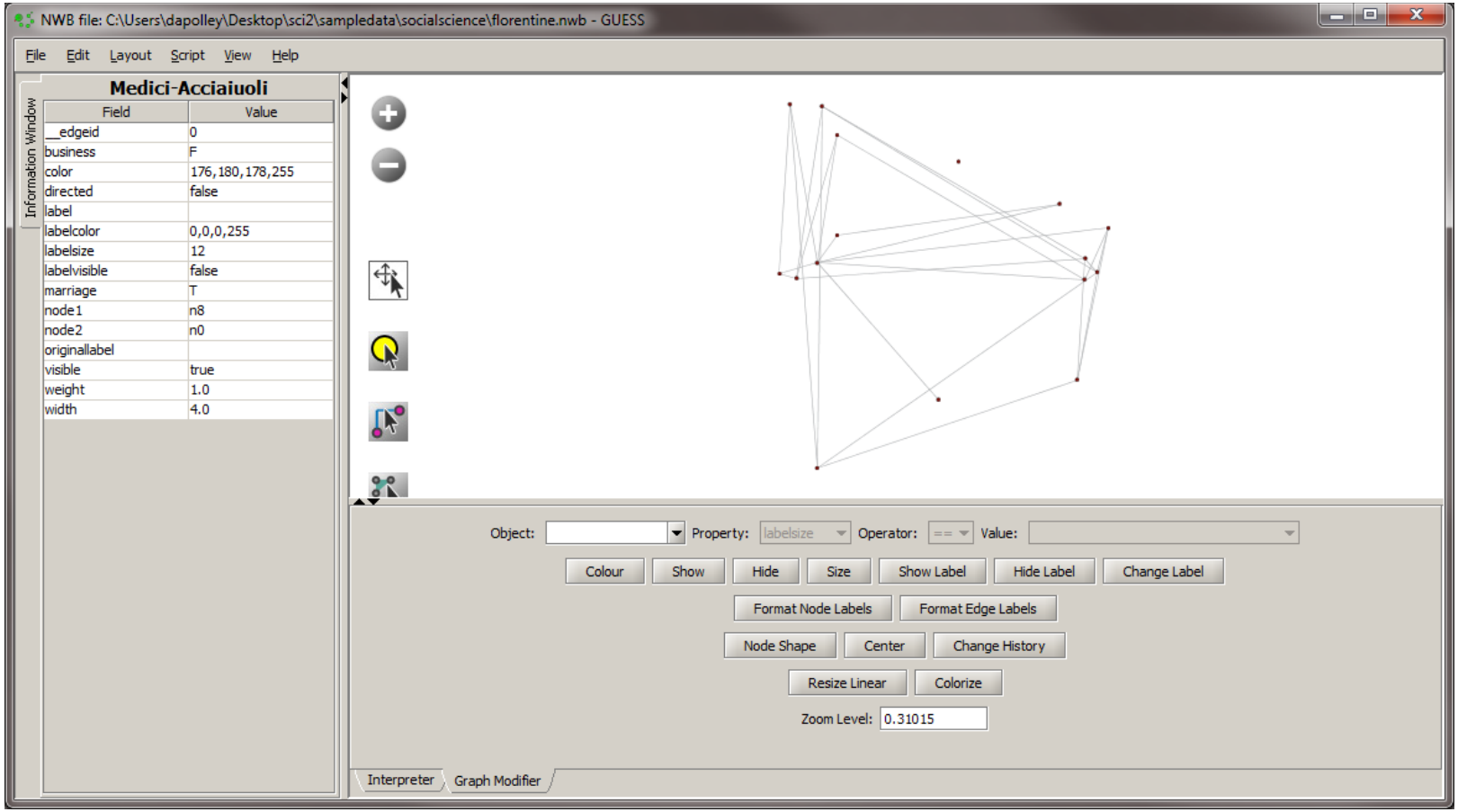
For this workflow we will skip straight to the visualization step, since the network file that we loaded already has the attributes we are interested in visualizing (wealth, priorates, and totalities). For other datasets, you will likely need to extract networks and run some type of analysis to answer the questions in which you are interested.

To visualize this network select the file from the Data Manager and run *Visualization > Networks > GUESS.*



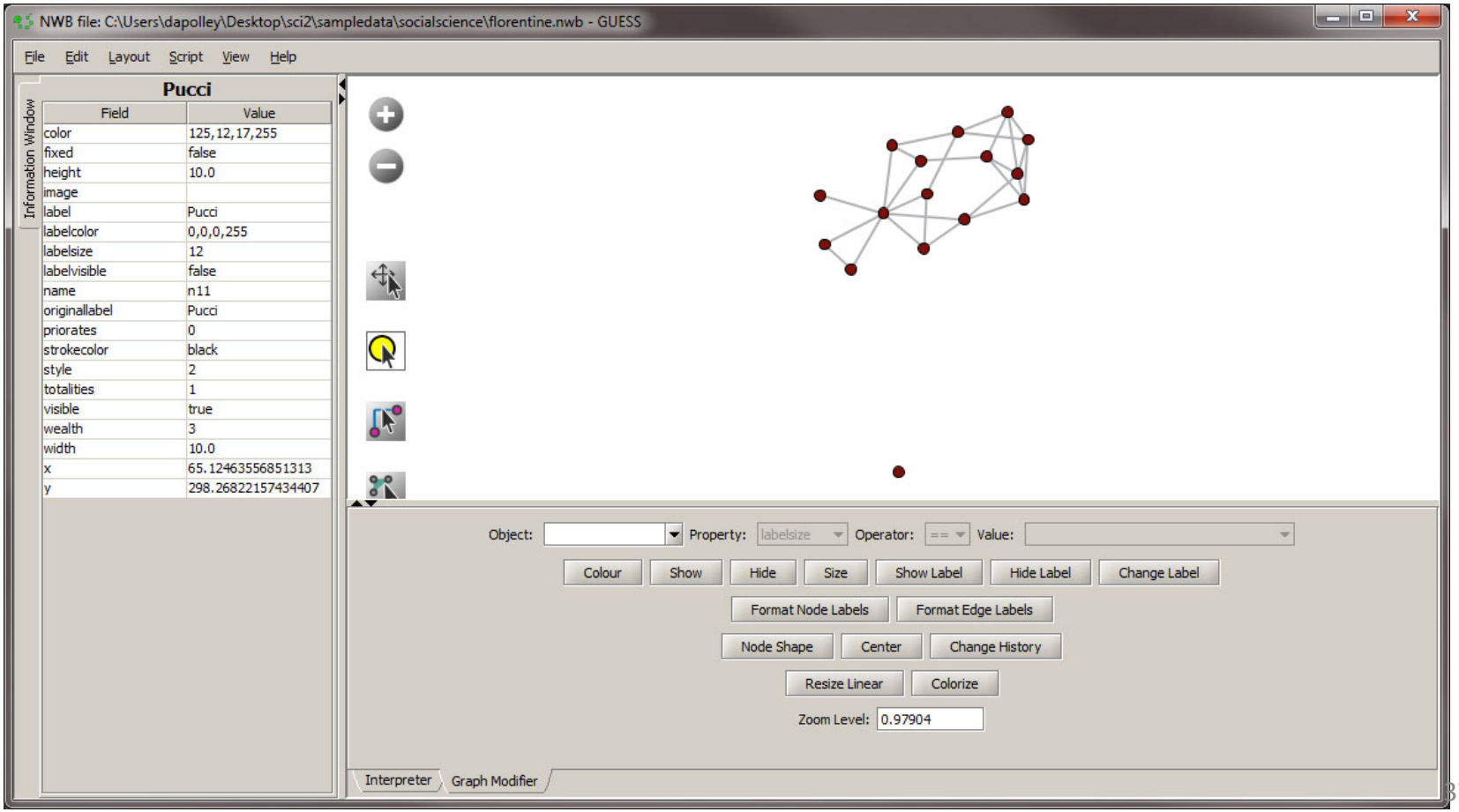
Visualizing the Florentine Dataset

When the network is loaded in GUESS it will be laid out randomly.



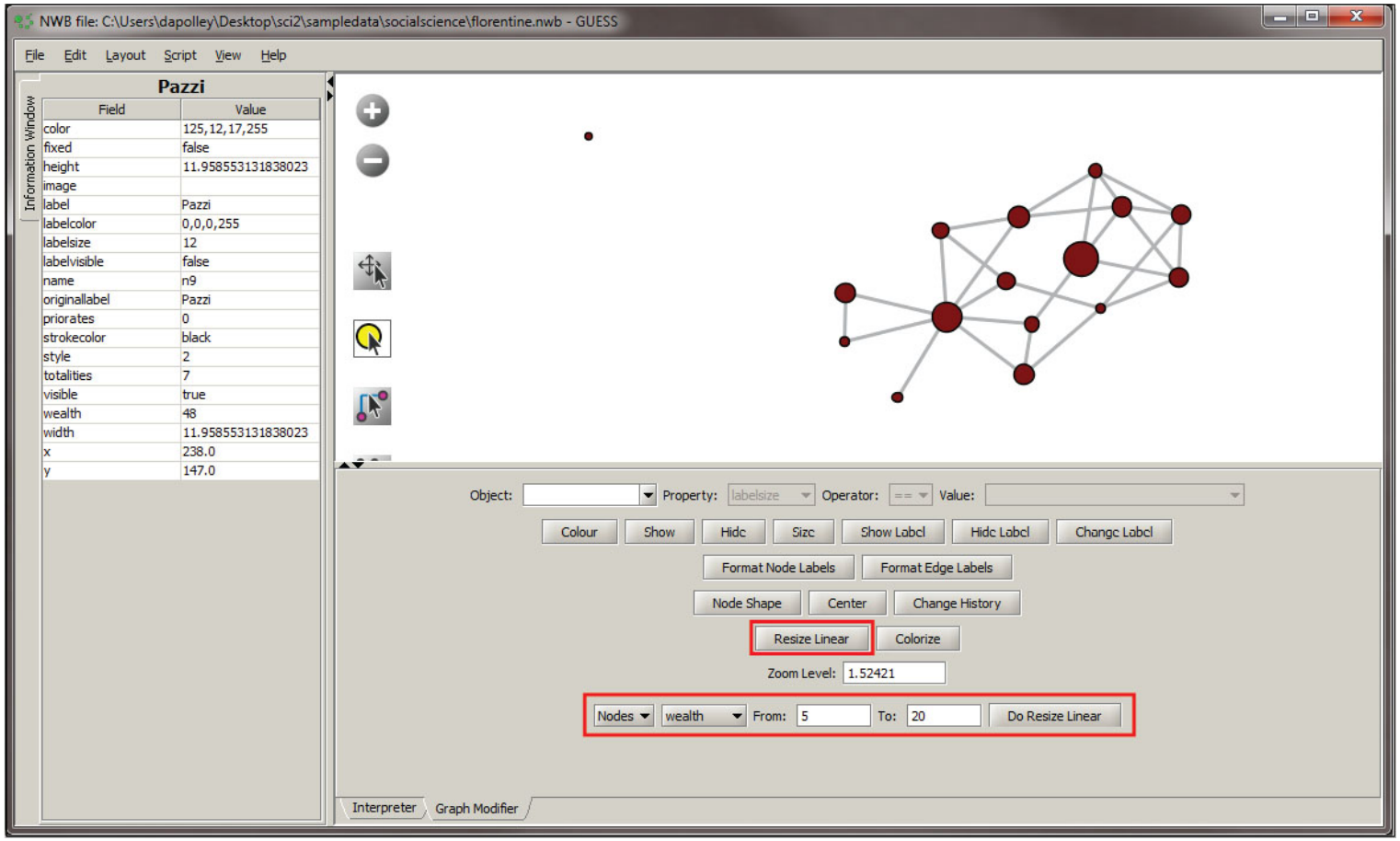
Visualizing the Florentine Dataset

The first step in enhancing this network visualization is to apply a different layout. For this visualization we will use the GEM layout *Layout > GEM.* You will notice that the GEM layout is random, you can run it multiple times and the network will appear slightly different each time.



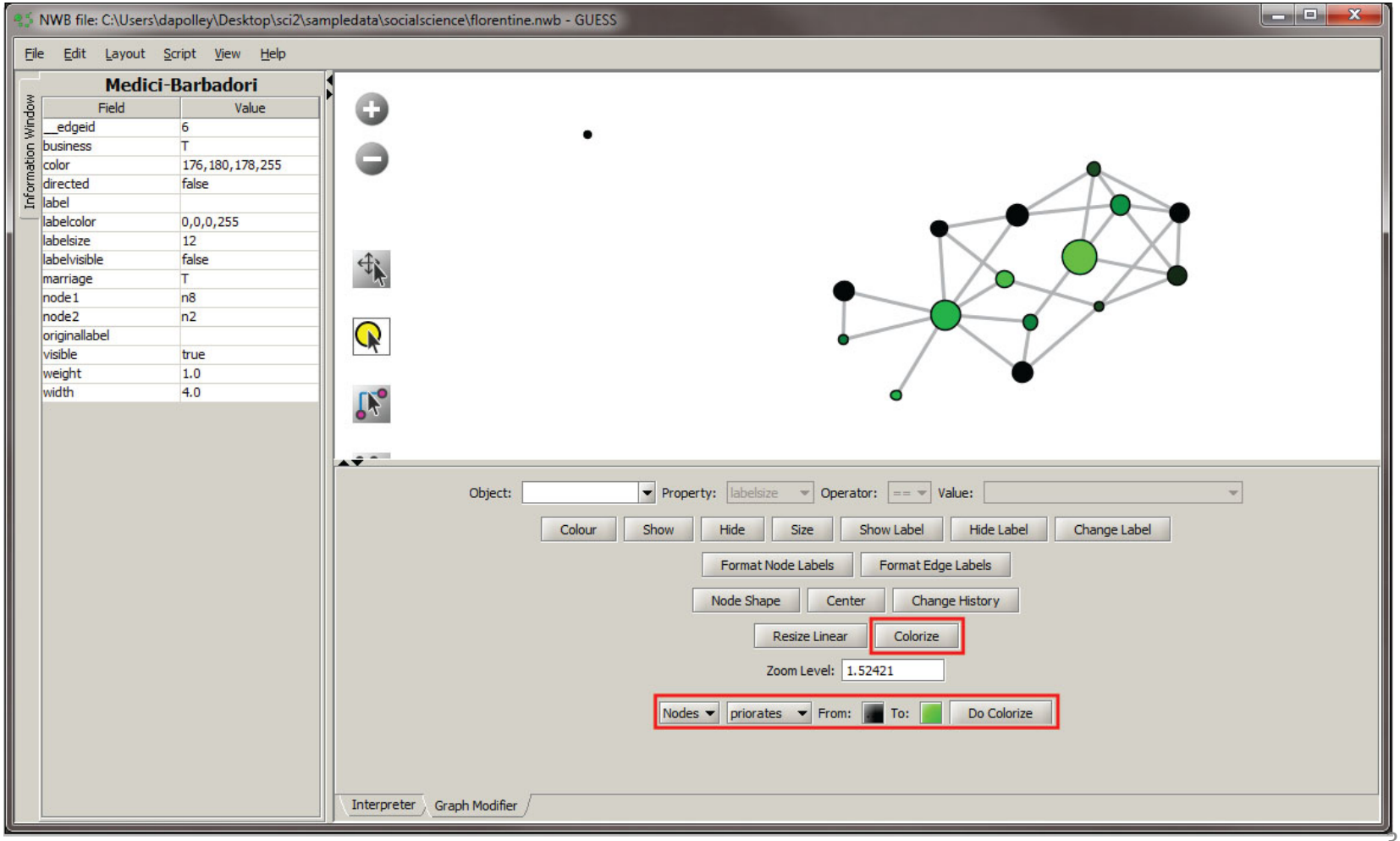
Visualizing the Florentine Dataset

The next step will be to resize the nodes based on the wealth attribute. To do this resize select the *Resize Linear* button and set the parameters to those shown below.



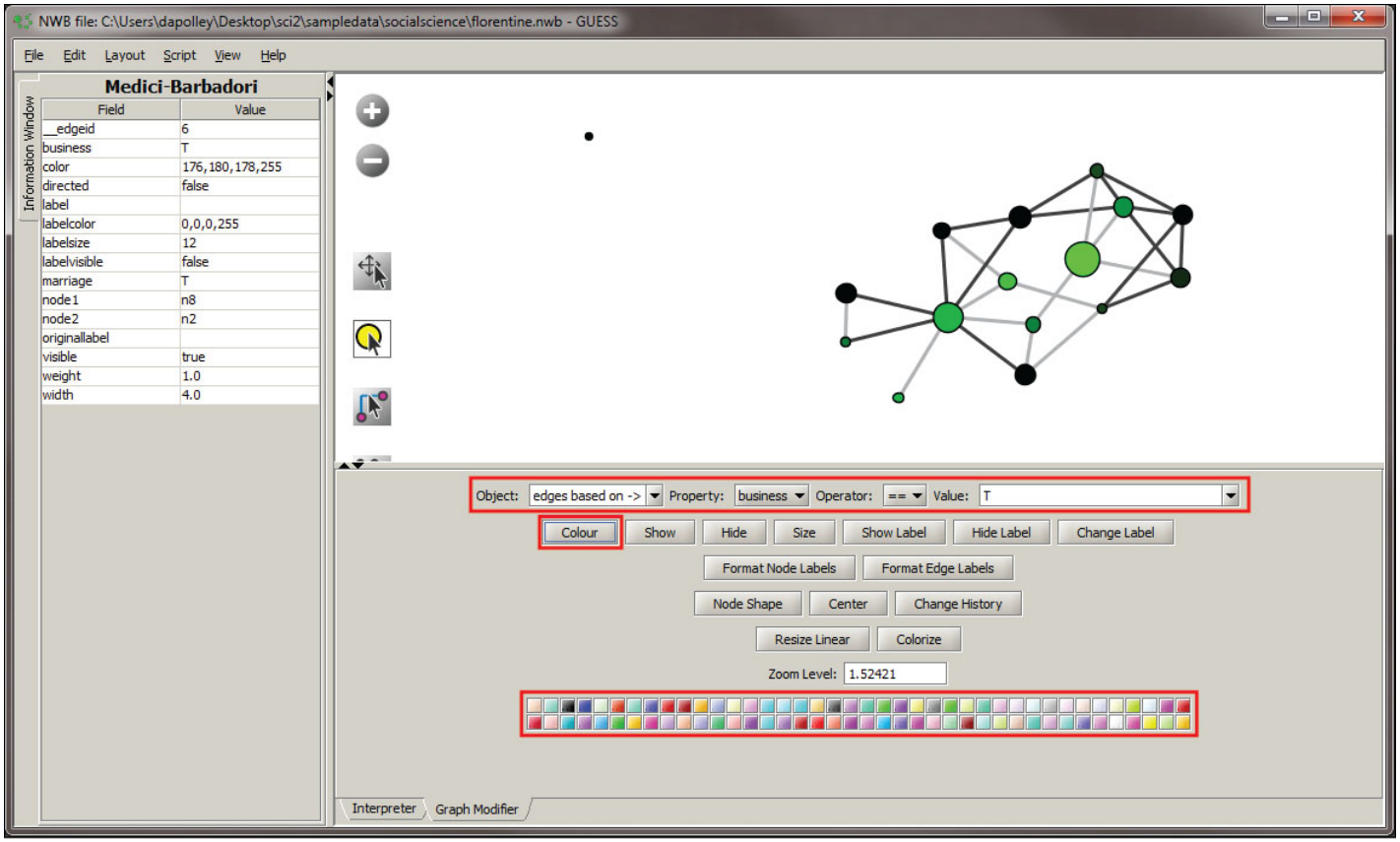
Visualizing the Florentine Dataset

Next we will colorize the nodes based on priorates to add an additional dimension to this visualization.



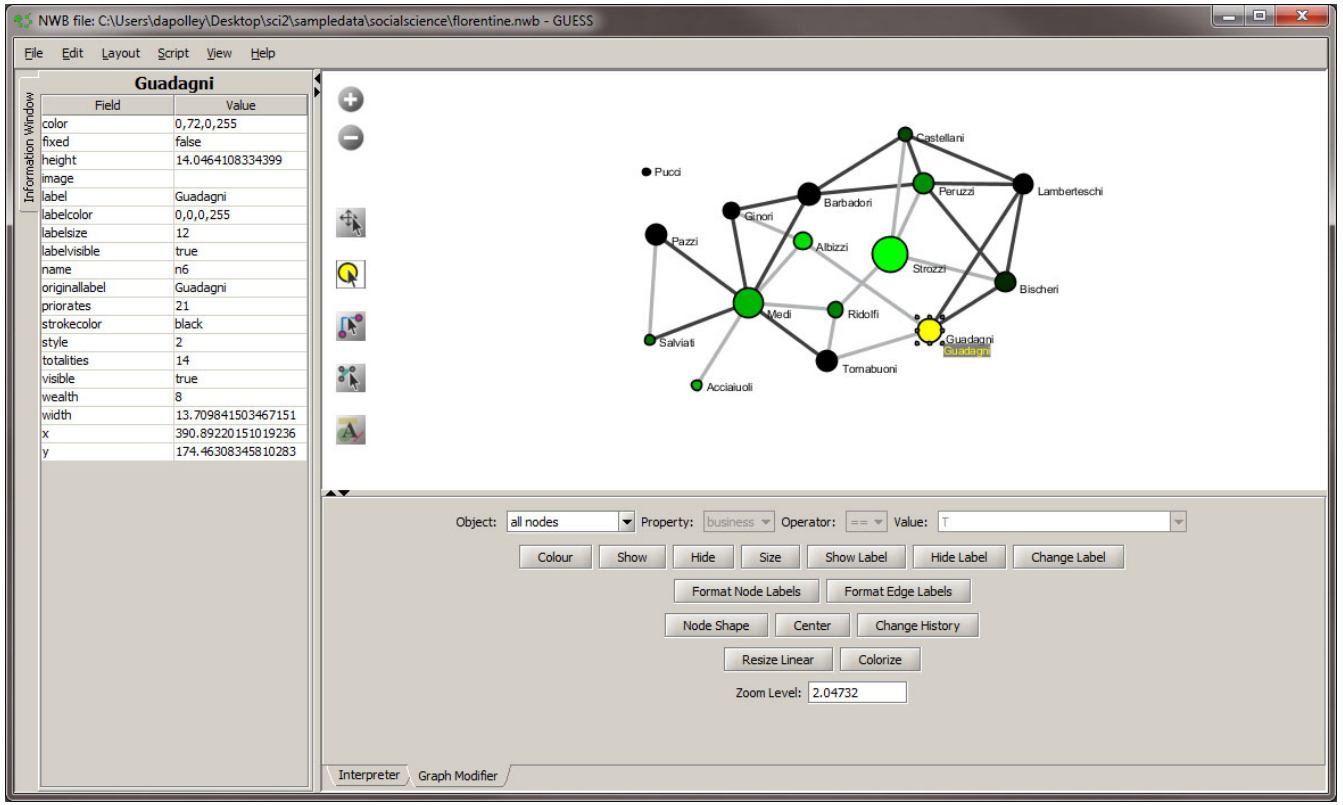
Visualizing the Florentine Dataset

Next we will color the edges to show the type of relationship between the families. To do this, you will need to select the *Object* edges *based on* ->, set the property to *marriage*, the operator to ==, and the value to *T*. Next, click the *Color* button and you can select the color of your choice from the pallet that will appear at the bottom of the Graph Modifier pane.



Visualizing the Florentine Dataset

You can repeat this process for the *business* property if you want to, or you can leave the edges that represent business ties the default color. In this workflow we will leave them the default color, light gray. The final step is to show all the labels. To do this, you will need to select the "Object" all nodes and the click the *Show Label* button and the labels will appear in the visualization.



Visualizing the Florentine Dataset

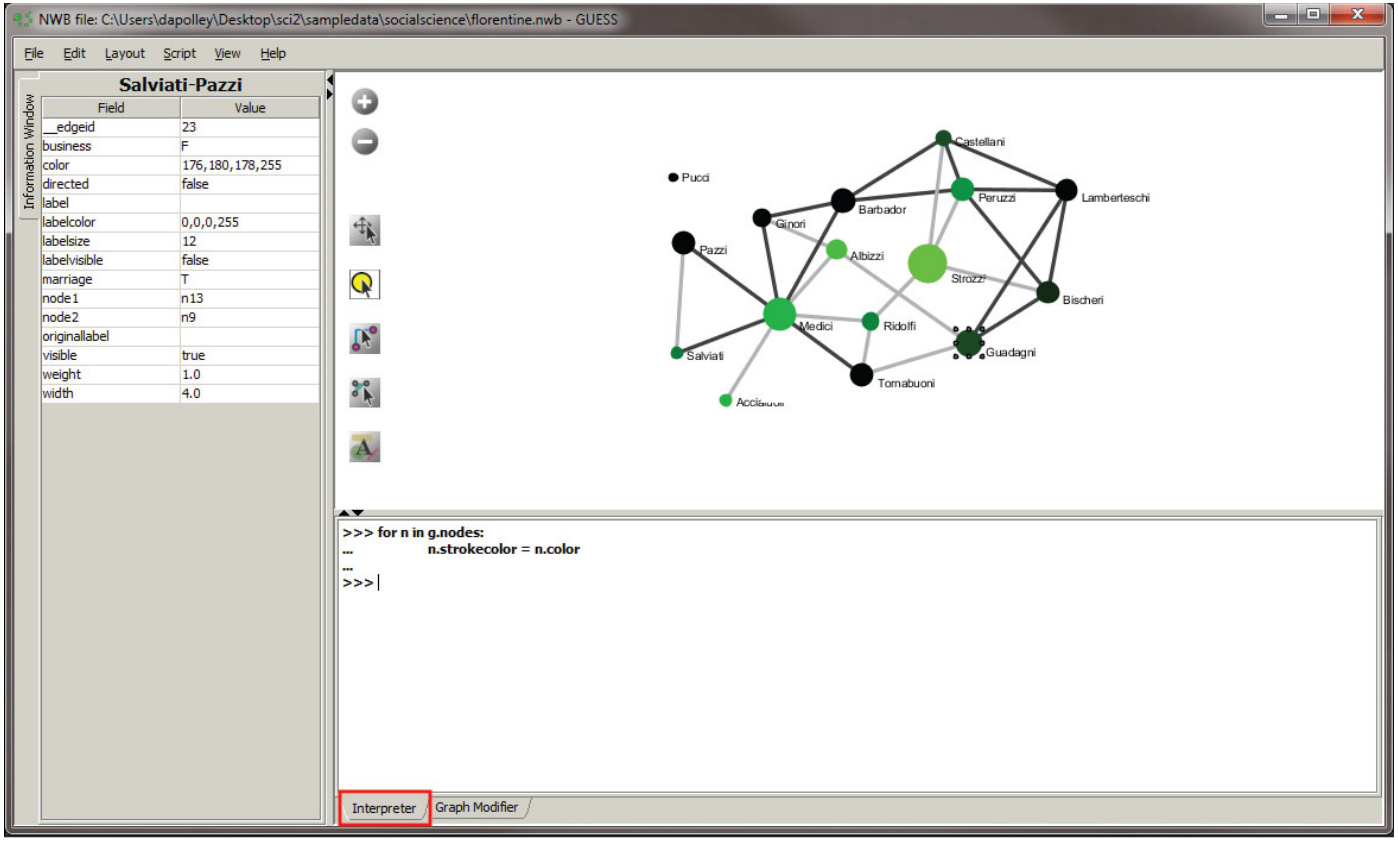
Since the GEM layout is random and all the nodes are spaced more or less evenly apart, you do not have to worry about disrupting the layout. However, other layout algorithms may space the nodes according to specific attributes of the network. Manually moving around nodes in this case would disrupt the layout of the network and distort the meaning of the visualization.

The last thing we want to do to our network is color the border of the nodes the same as the nodes themselves. This is not as crucial for networks with only a few nodes, but as the size of your network increases it can become difficult to read with the thick black lines around every node. To color those the same as the node go to the *Interpreter* tab at the bottom of the GUESS window and type in the following commands:

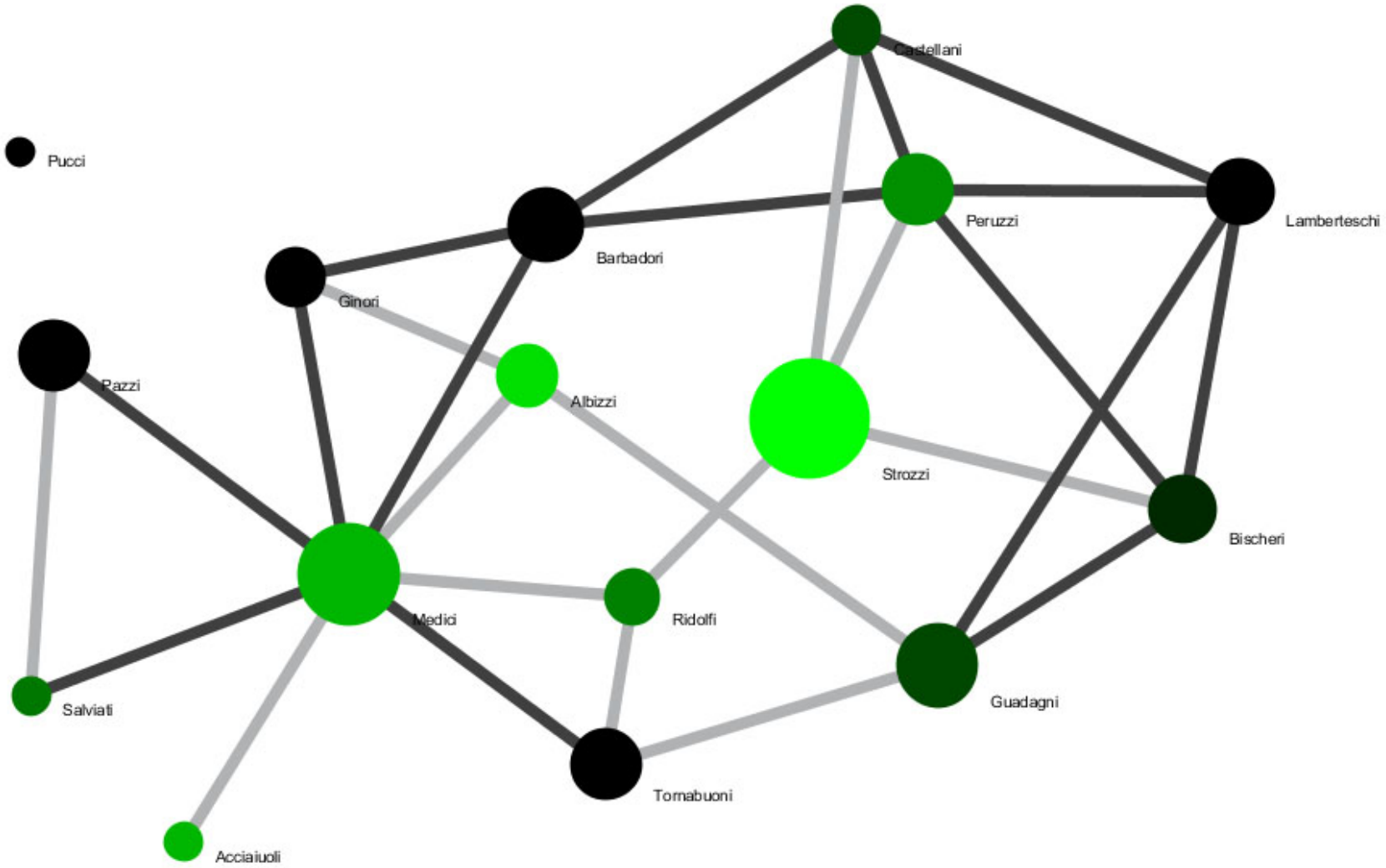
```
for n in g.nodes:
        n.strokecolor = n.color
```

Visualizing the Florentine Dataset

This code basically tells GUESS that for every node (*n*) in this graph of nodes (*g.nodes)* make the border color of the nodes (*n.strokecolor*) equal to the node color (*n.color)*. After you type the first line you will need to hit the "Tab" key before you start typing the second line of code.



Visualizing the Florentine Dataset



Questions?

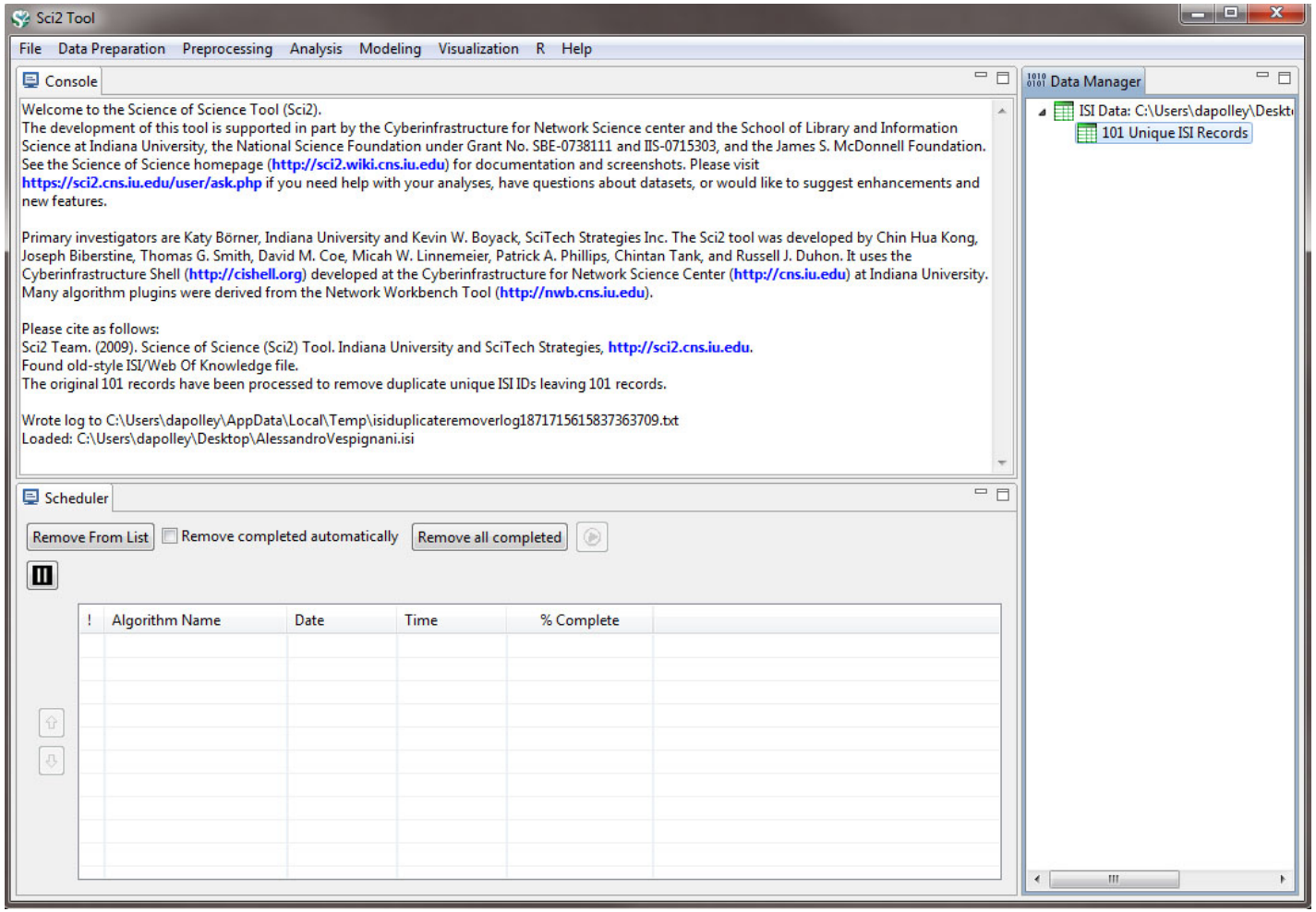
Temporal Analysis: *Evolving Co-Authorship Network*

For this analysis we will be studying the evolution of Alessandro Vespignani's co-authorship network over time. We will see his network of collaborators grow from 1990 to 2006, giving us a sense of how his scholarly output has grown. Each node in the network will represent an author in the data set and the edges that connect them will be weighted based on how many times they have collaborated.

Evolving Co-Authorship Network

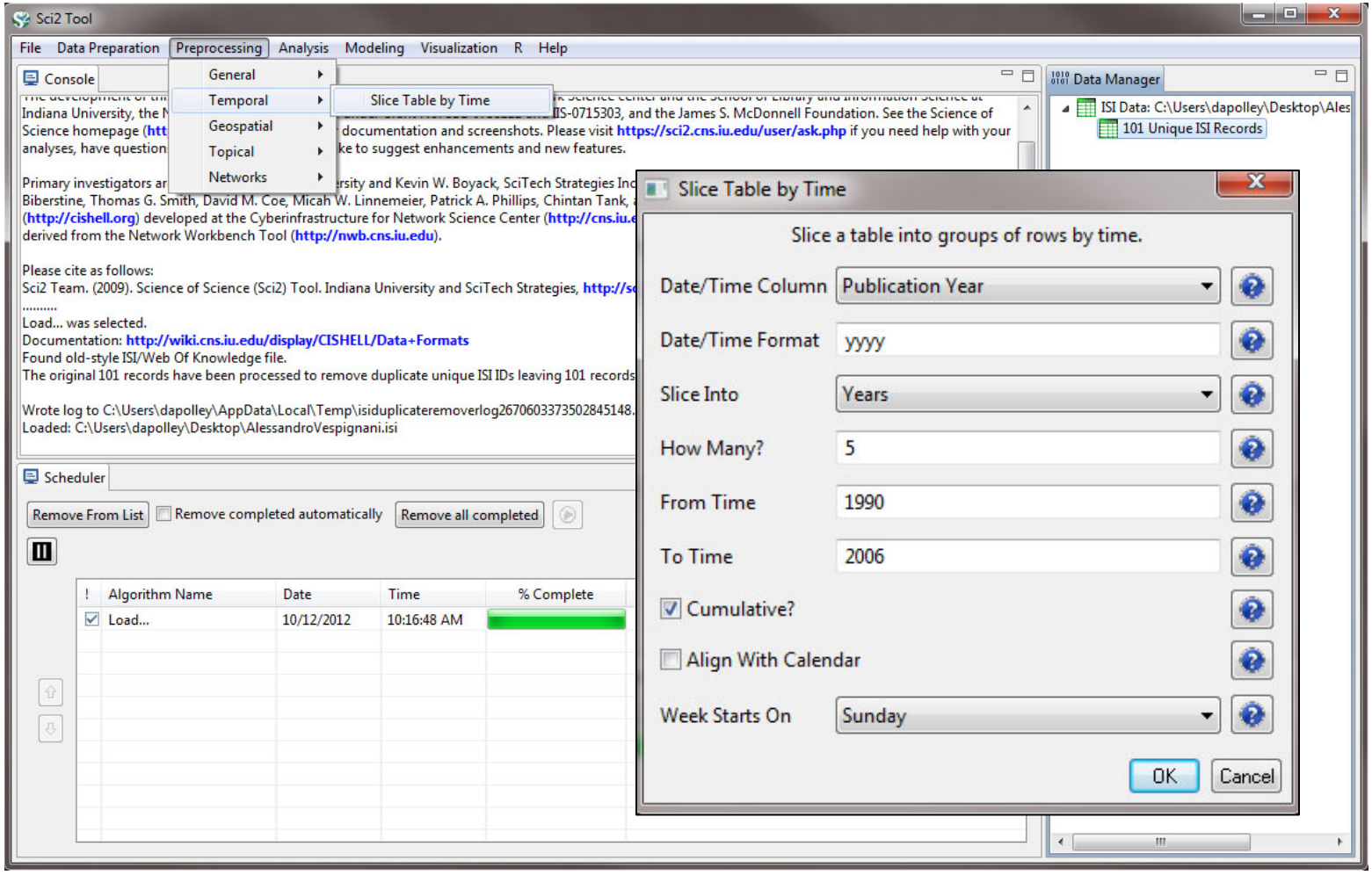
File > Load > AlessandroVespignani.isi

Load this file from the sample data folder you copied from the flash drive.



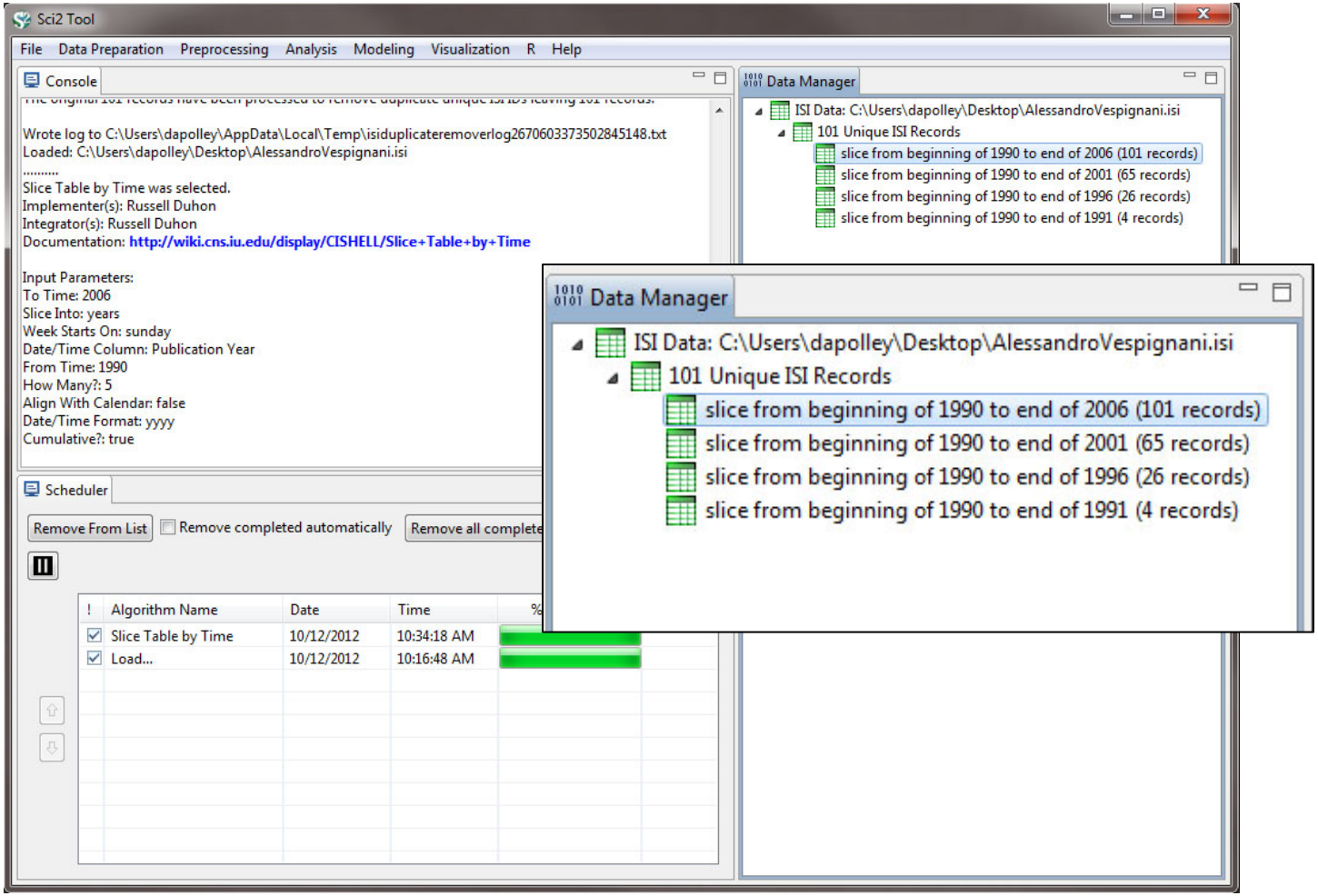
Evolving Co-Authorship Network

Select *Preprocessing* > *Temporal* > *Slice Table by Time* and choose the parameters shown at the right.



Evolving Co-Authorship Network

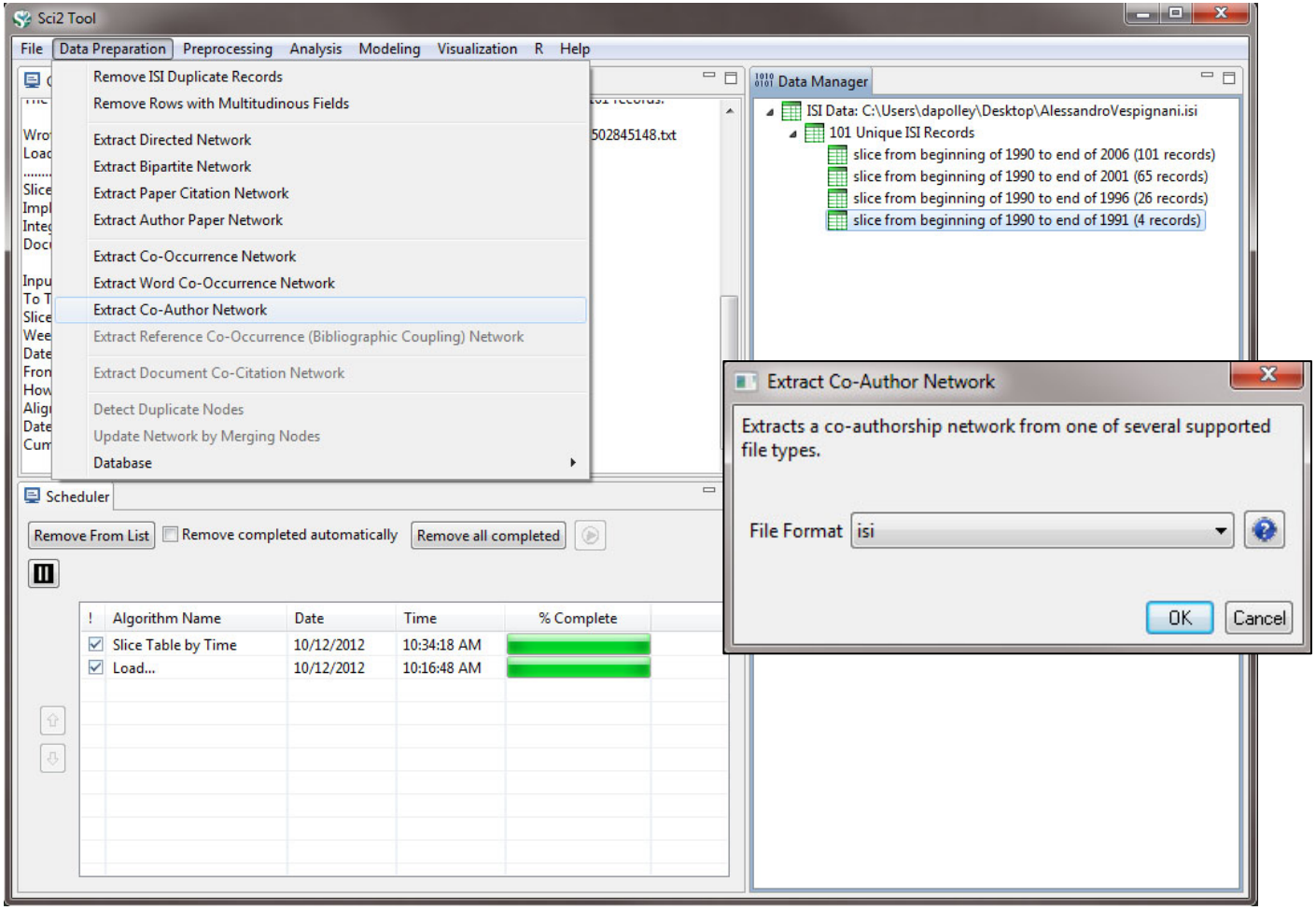
Now that the algorithm has been run, you will notice the original dataset has been divided into four tables that cumulatively display the evolution of this data.



Evolving Co-Authorship Network

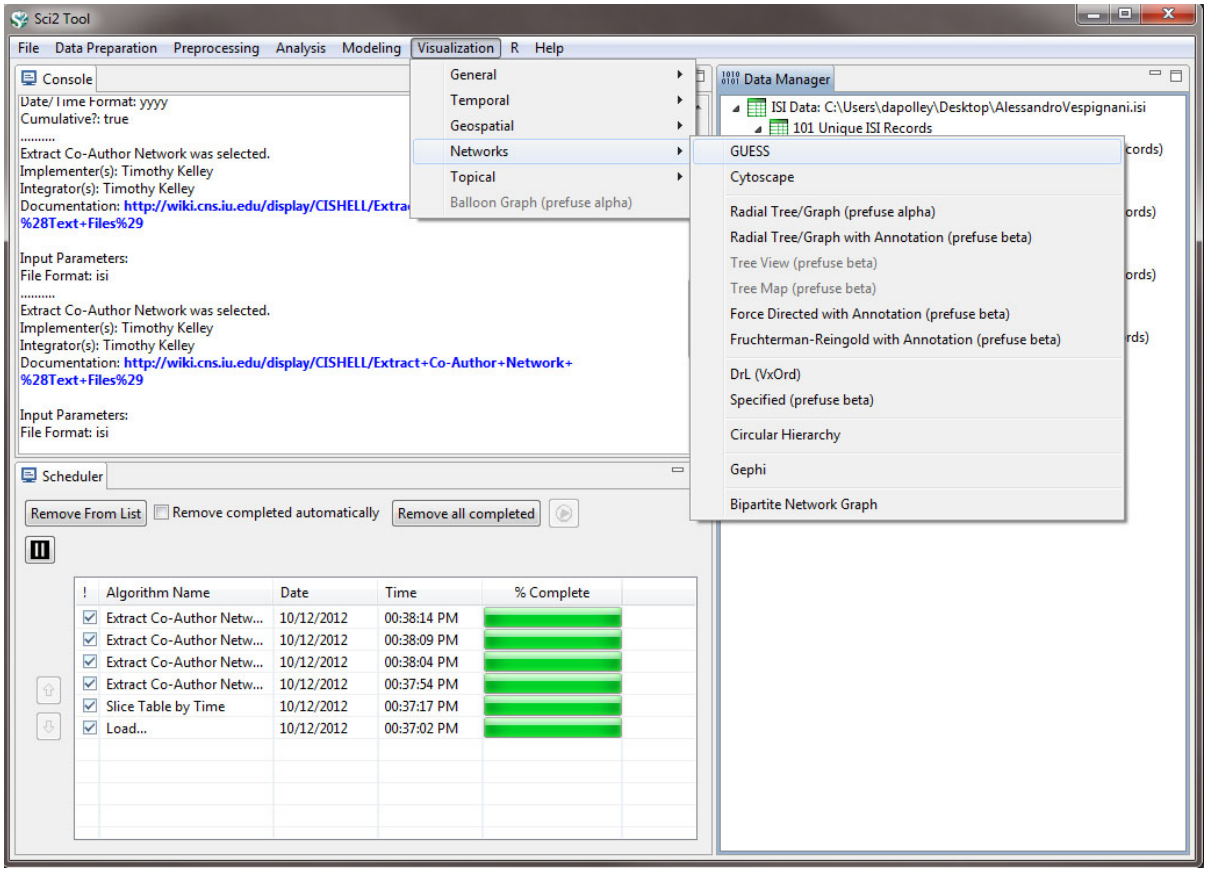
Select the first table and run *Data Preparation > Extract Co-Author Network*

Repeat this step for each of the tables in the Data Manager



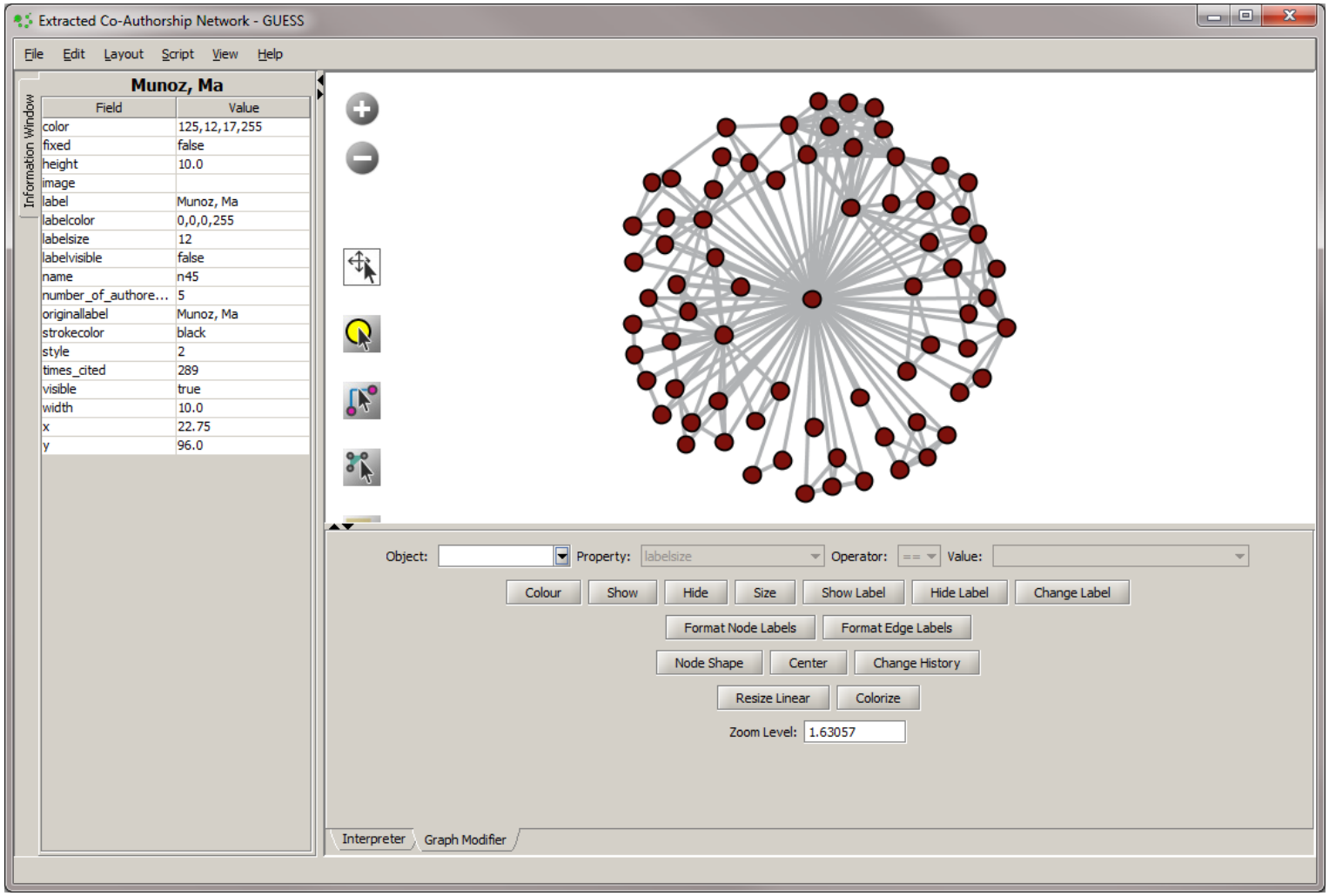
Evolving Co-Authorship Network

Select the first extracted co-author network and run *Visualization* > *Networks* > *GUESS* starting with the network that spans 1990 to 2006 because we will export these node positions and use them to layout the other networks.



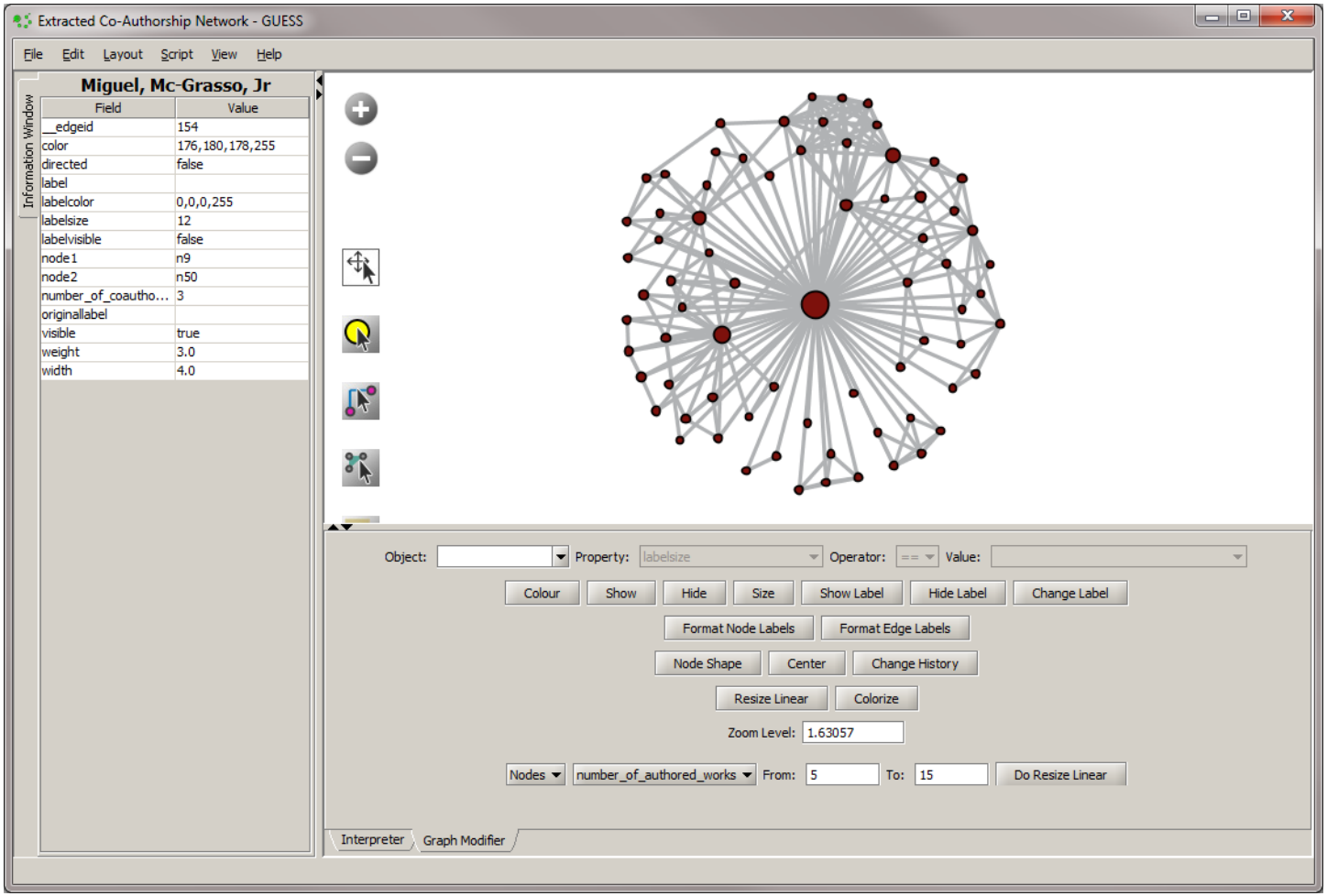
Evolving Co-Authorship Network

The network will be loaded in with a random layout in GUESS
To change the layout select *Layout > Gem*



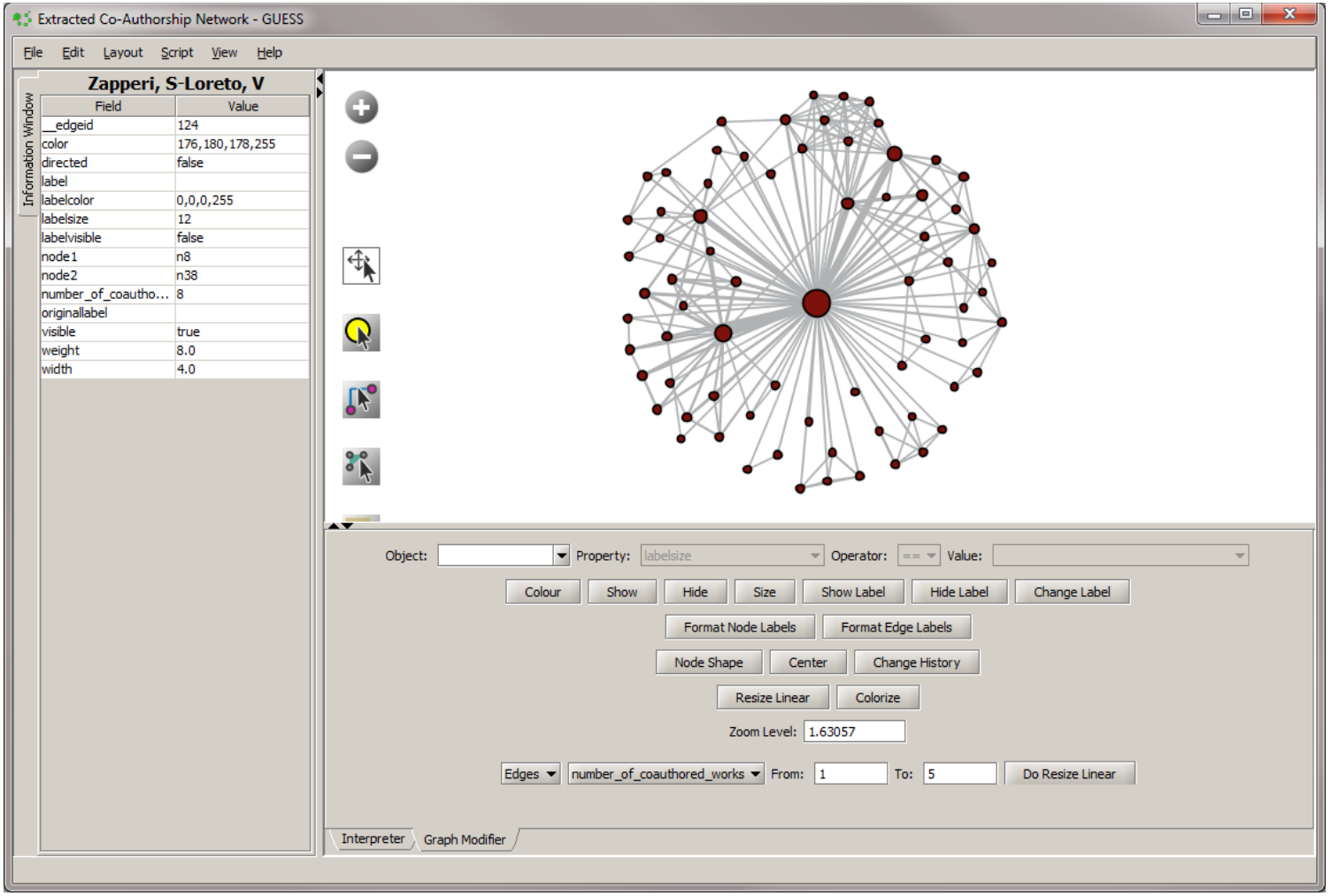
Evolving Co-Authorship Network

Resize the nodes based on *number_of_authored_works*
Set the parameters from 5 to 15 and click *Do Resize Linear*



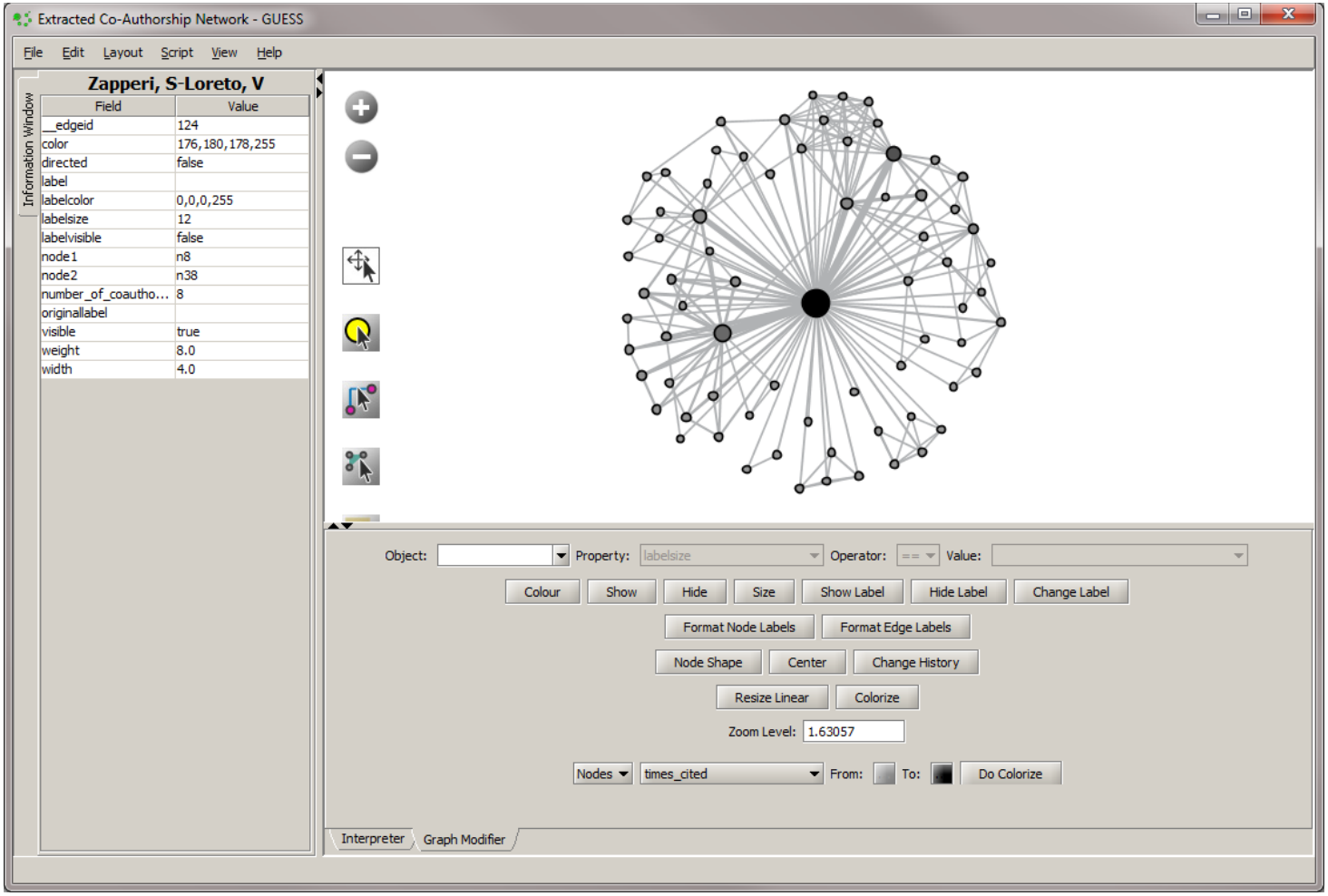
Evolving Co-Authorship Network

Resize the edges based on *number_of_coauthored_works*
Set the parameters from 1 to 5 and click *Do Resize Linear*



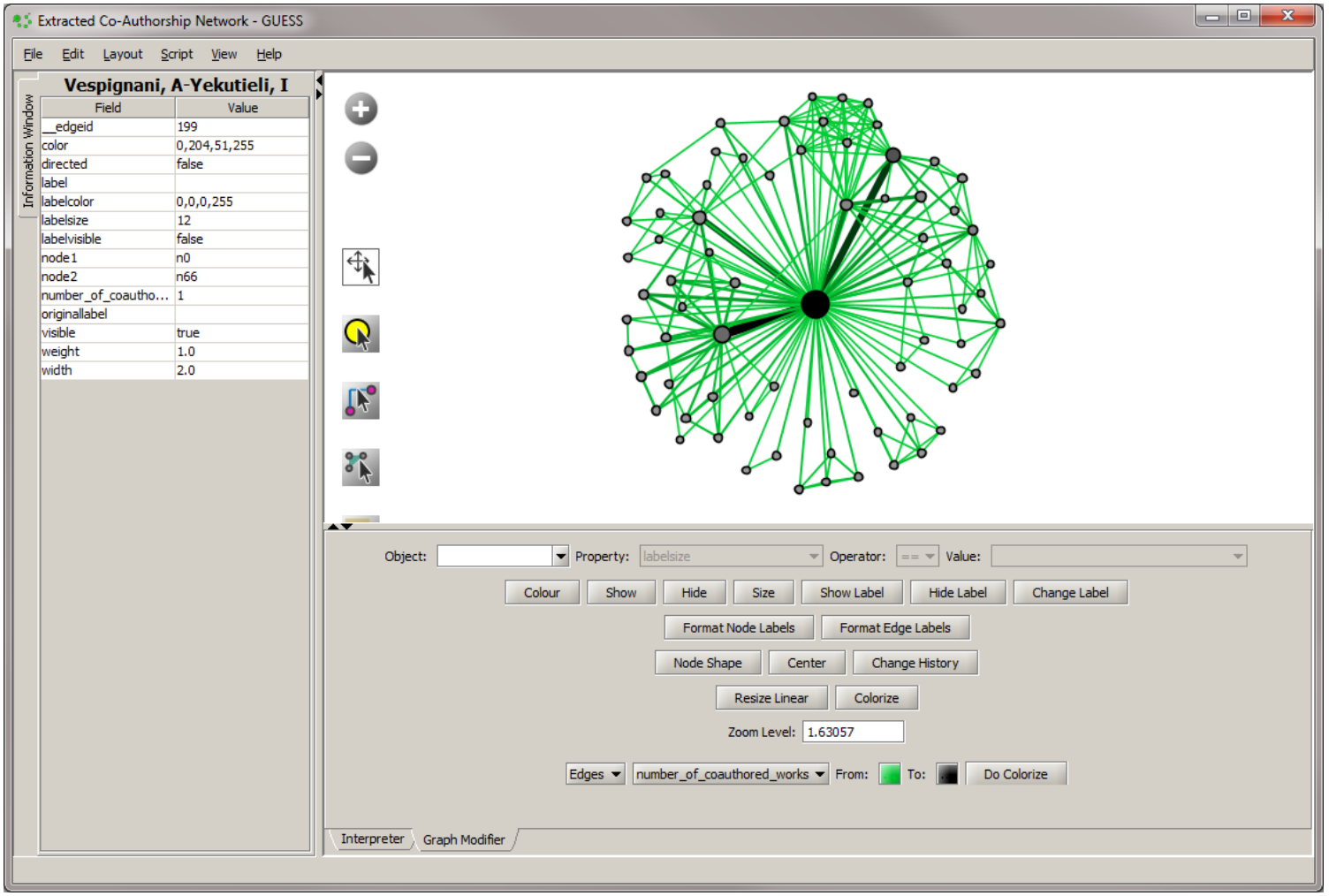
Evolving Co-Authorship Network

Colorize the nodes based on *times_cited*
Set the parameters from *Gray* to *Black* and click *Do Colorize*



Evolving Co-Authorship Network

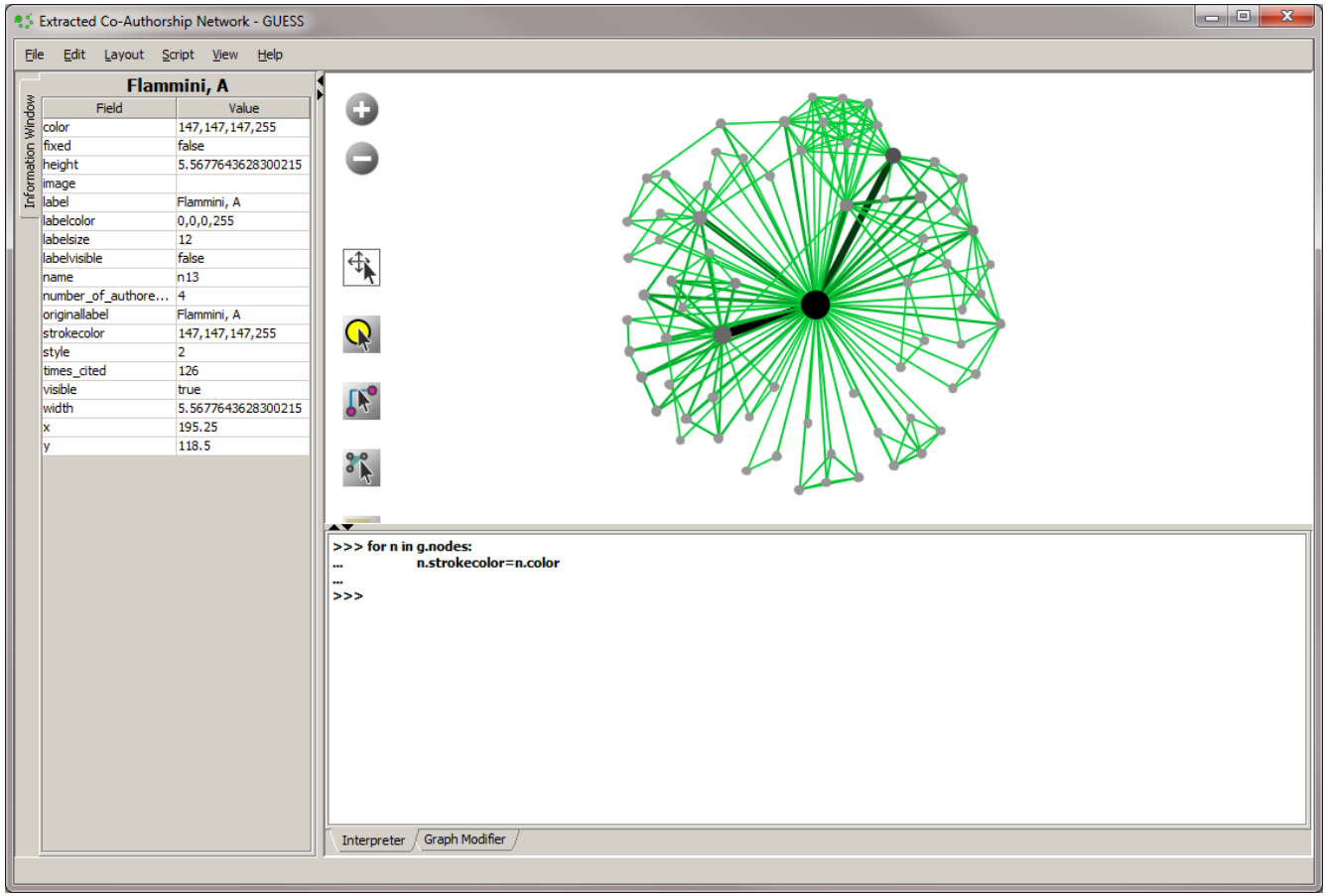
Colorize the edges based on *number_of_coauthored_works*
Set the parameters from *Green* to *Black* and click *Do Colorize*



Evolving Co-Authorship Network

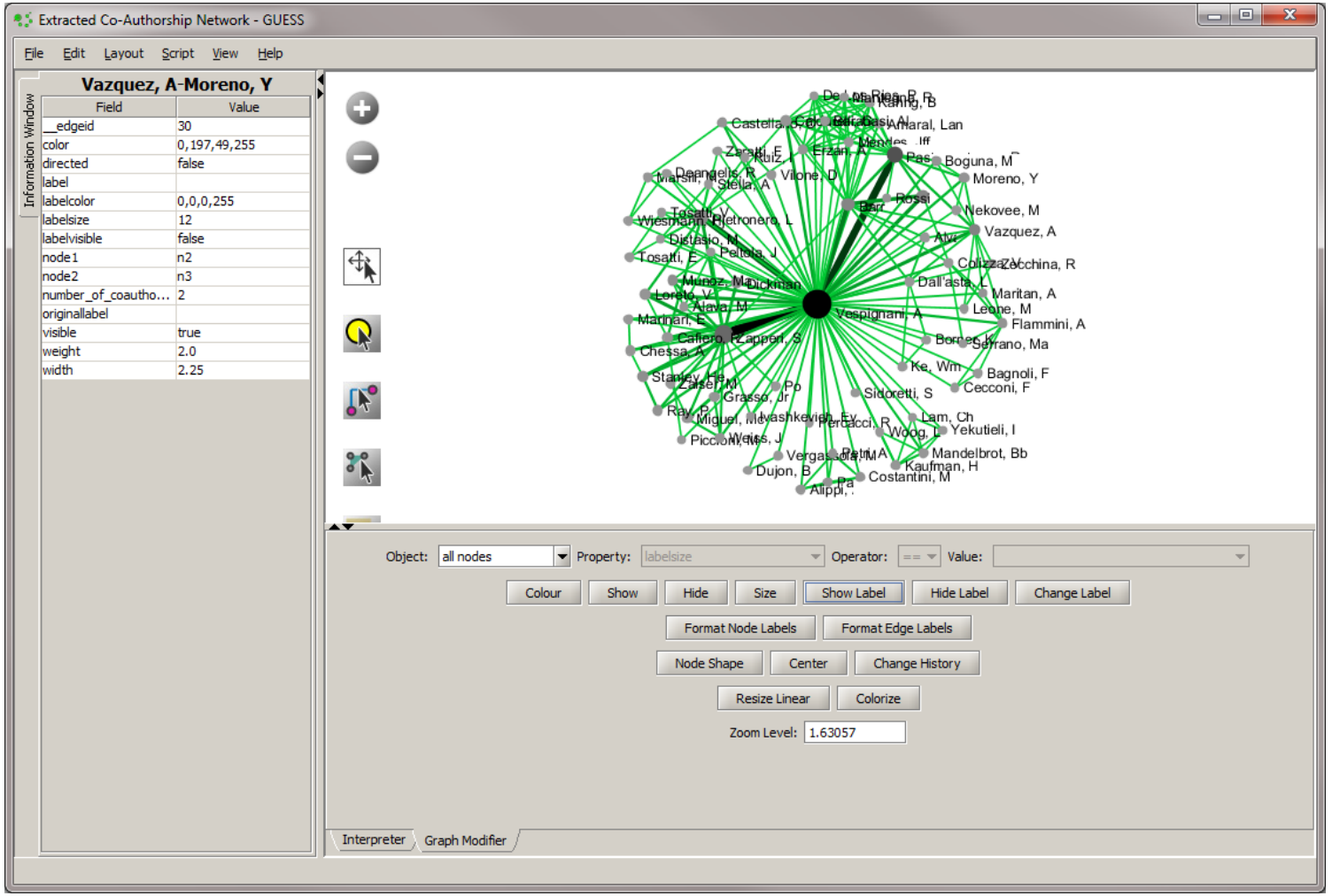
If you want to remove the borders from the nodes, type the following commands in into the interpreter:

```
for n in g.nodes:
        n.strokecolor=n.color
```



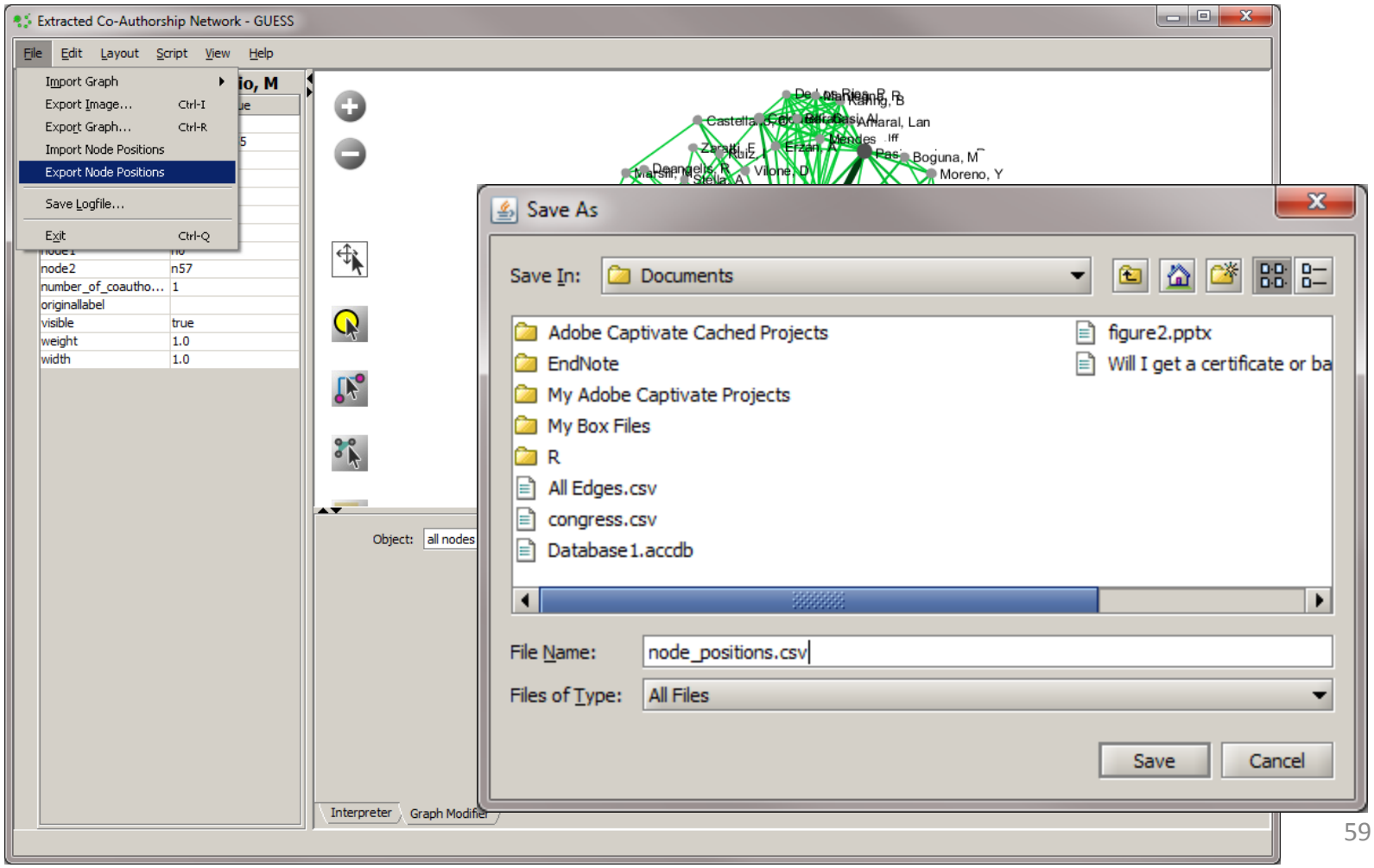
Evolving Co-Authorship Network

Finally add the labels to the nodes by selecting object: *all nodes* and then click Show Label



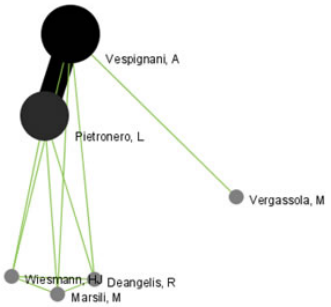
Evolving Co-Authorship Network

To save the node positions of the current layout so that the layout is consistent across all time slices select *File* > *Export Node Positions* and save the file as a CSV file.

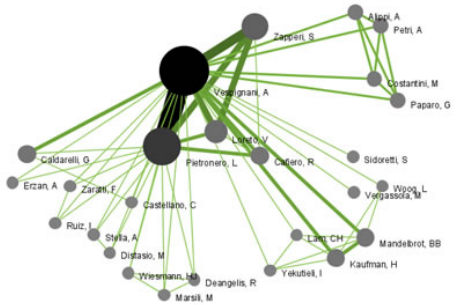


Evolving Co-Authorship Network

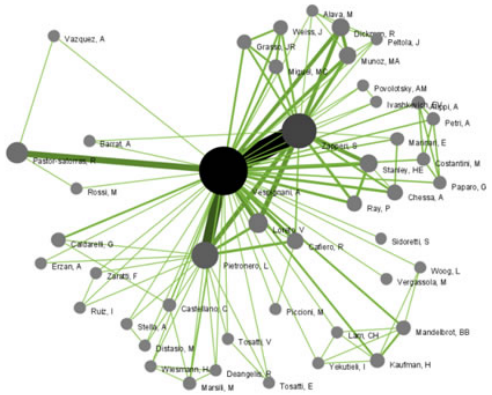
Now when you go to visualize the other three networks you will want to import the node the node positions using *File > Import Node Positions* and the network will be laid out accordingly. When the networks are displayed side-by-side you can see an evolution.



1990-1991



1990-1996



1990-2001

1990-2006

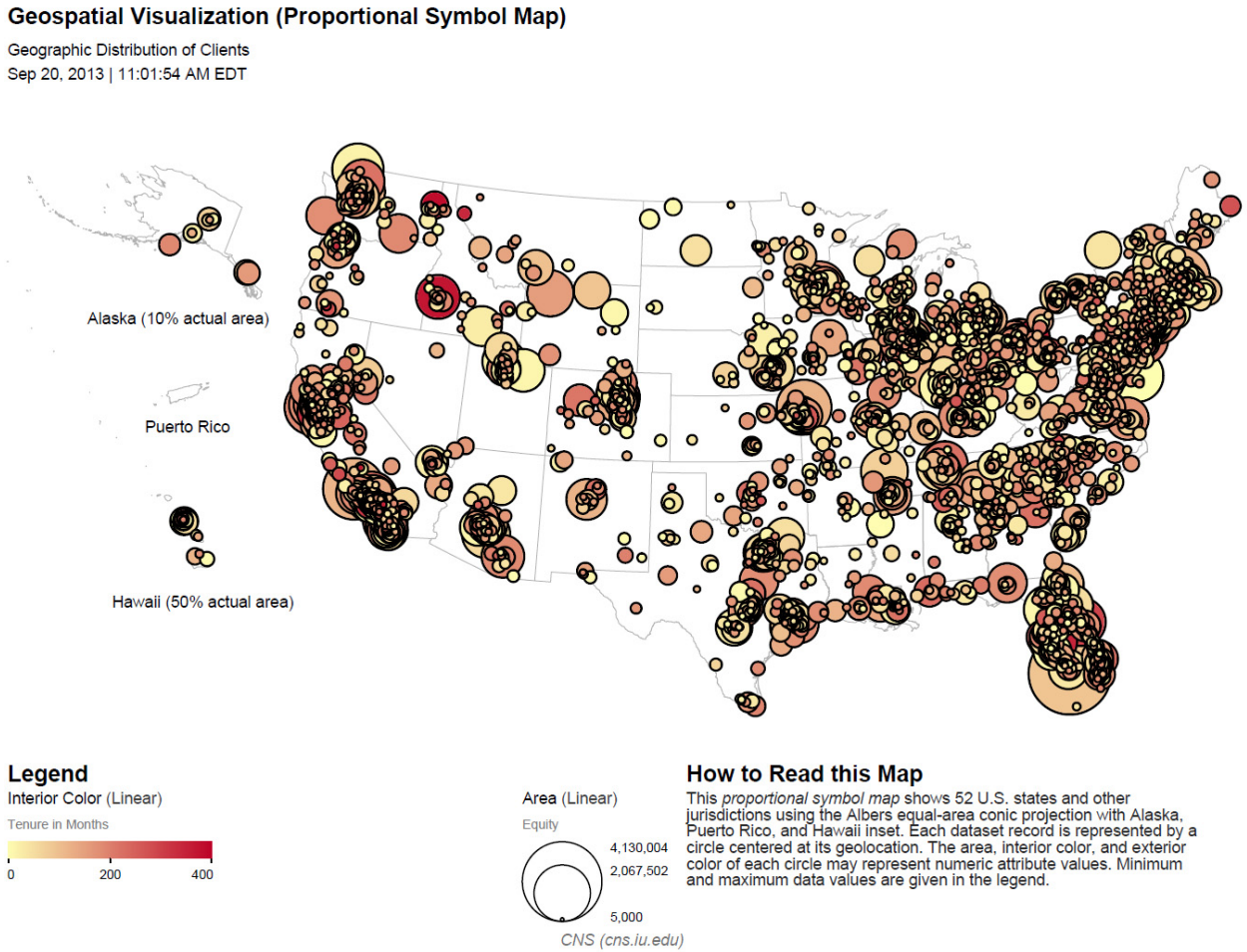
Questions?

Geospatial Analysis: *Proportional Symbol Map of Clients*

For this analysis we will be exploring the geographic distribution of clients and encoding various attributes of the data, such as equity and tenure as a TD Ameritrade client. The sample data has been provided to you prior to the workshop.

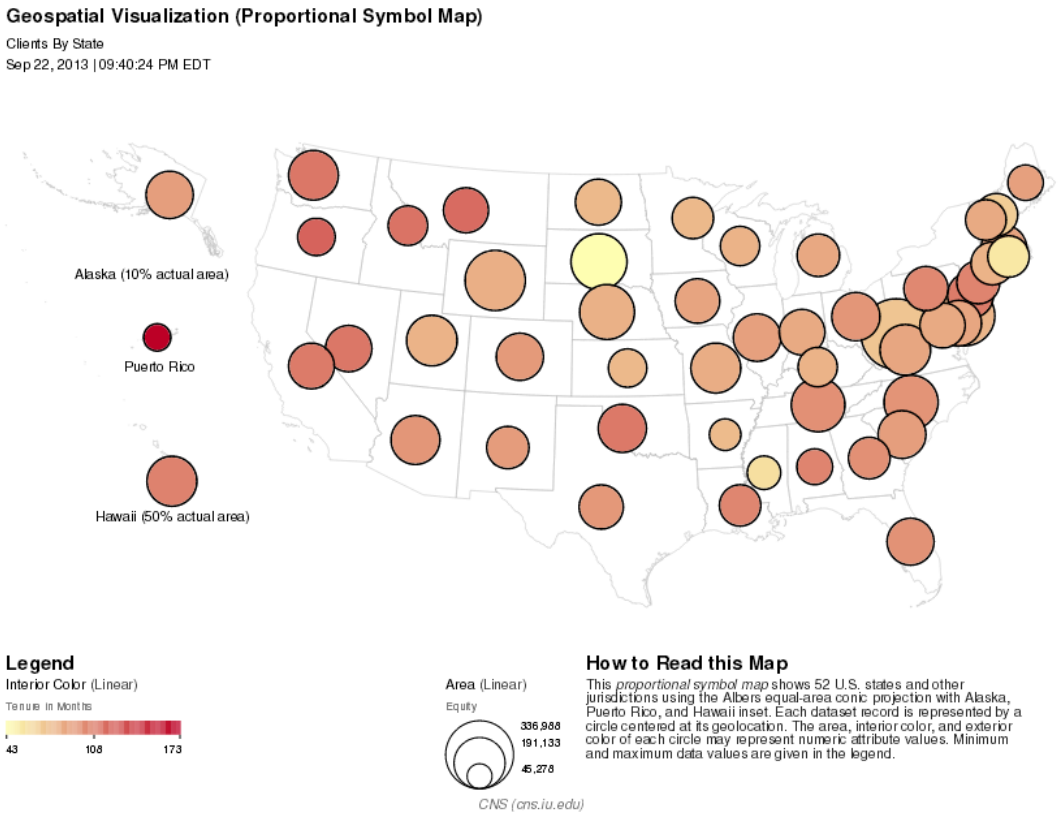
Geographic Distribution of Clients

How much is too much?



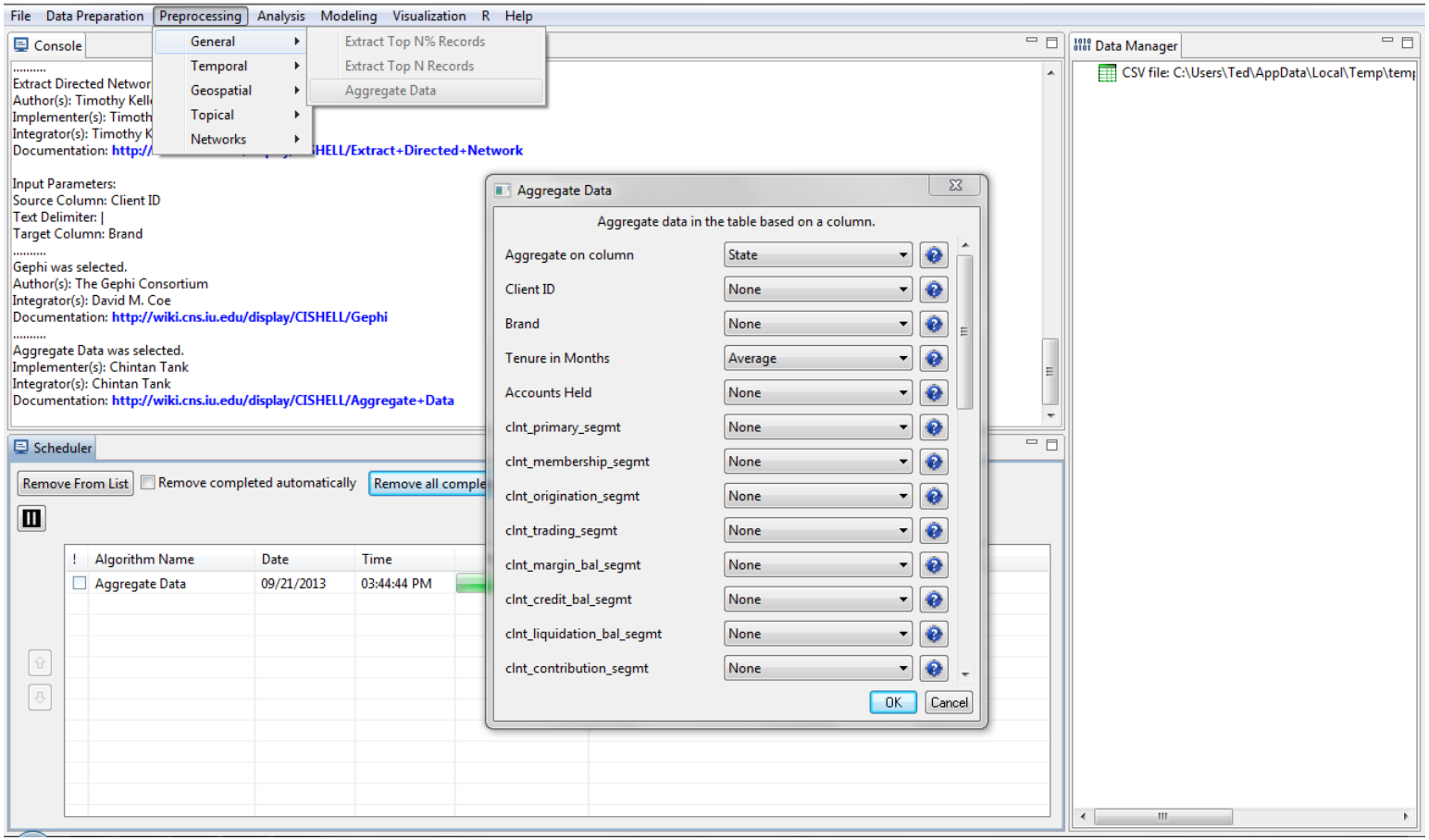
Geographic Distribution of Clients

Aggregate clients by state



Geographic Distribution of Clients

Preprocessing > *General* > *Aggregate Data* and set the Aggregate on Column to State, find the Average Tenure in Months and Equity for the clients in these states.



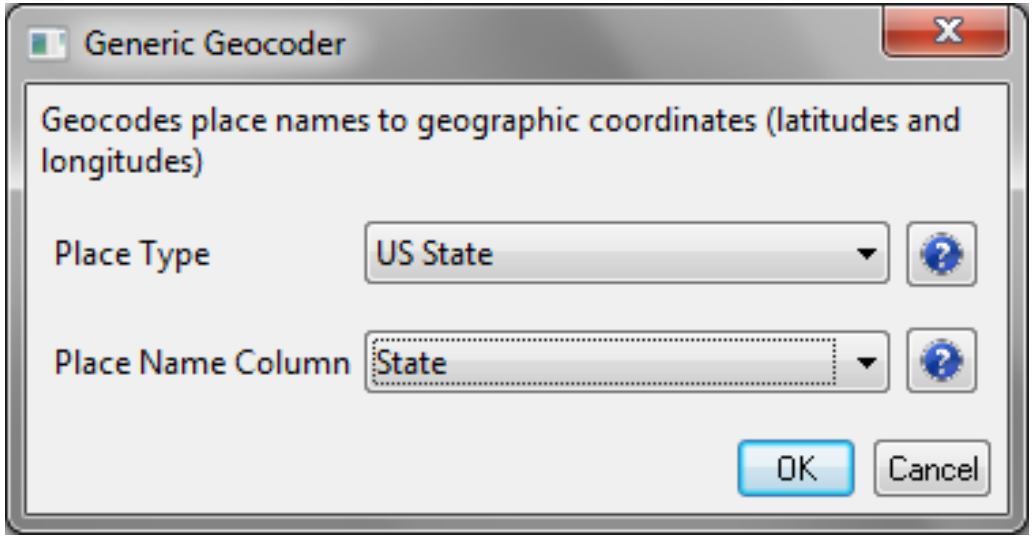
Geographic Distribution of Clients

The result is a simplified dataset with the each state present. The *Count* column gives the total number of clients in each state for this dataset. The *Tenure in Months* column gives the average tenure in months for those clients, and the *Equity* column gives the average equity for clients in those states.

	A	B	C	D
1	Tenure in Months	Equity	State	Count
2	106	165412	HI	27
3	98	145186	FL	308
4	95	128056	TX	322
5	104	111266	LA	41
6	105	81466	AL	55
7	59	70413	MS	9
8	99	112742	GA	114
9	96	158025	AZ	117
10	92	150764	SC	56
11	110	137150	CA	891
12	93	116514	NM	24
13	97	194911	NC	116
14	111	153475	OK	35
15	77	62068	AR	19
16	99	191937	TN	54
17	112	141320	NV	32

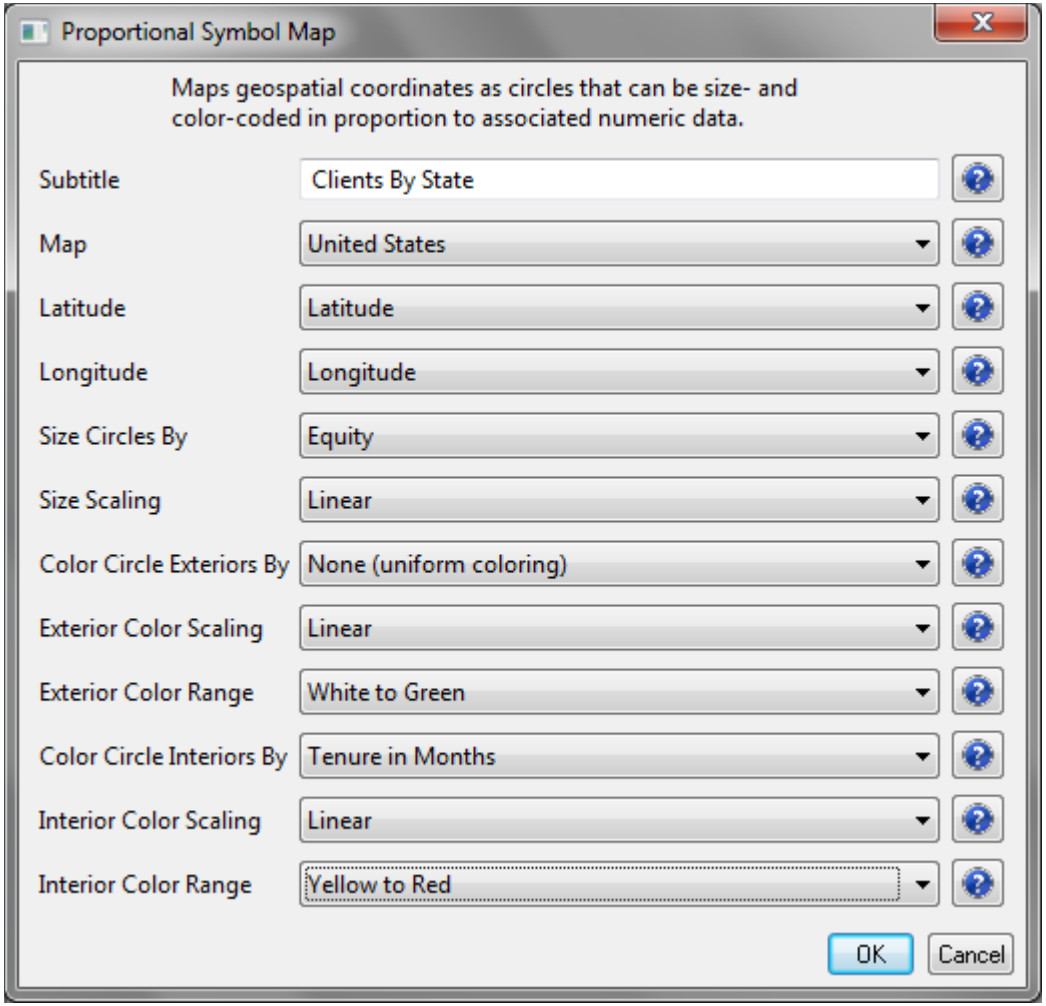
Geographic Distribution of Clients

Next, select *Analysis* > *Geospatial* > *Generic Geocoder*



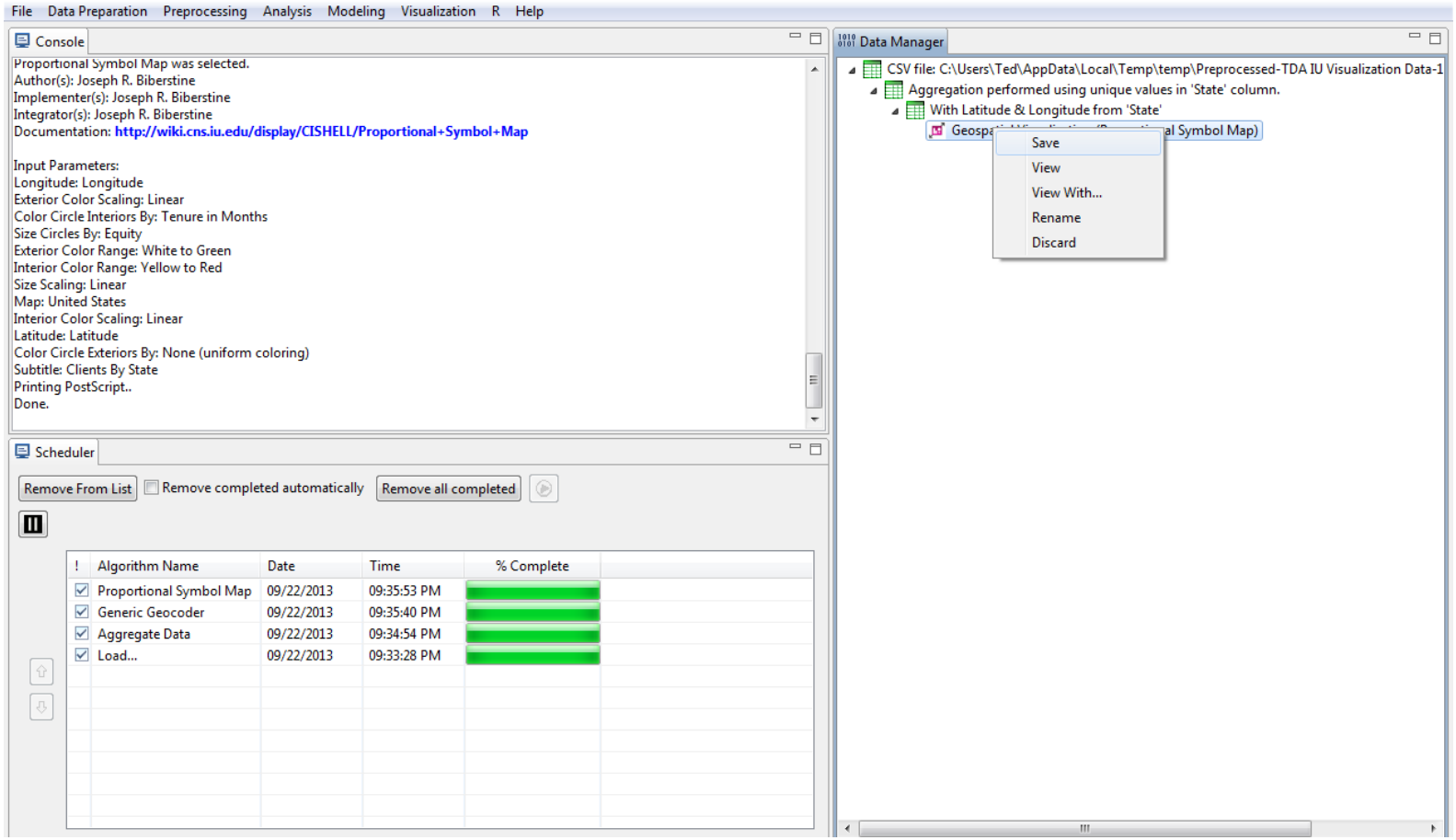
Geographic Distribution of Clients

Next, select *Visualization* > *Geospatial* > *Proportional Symbol Map*



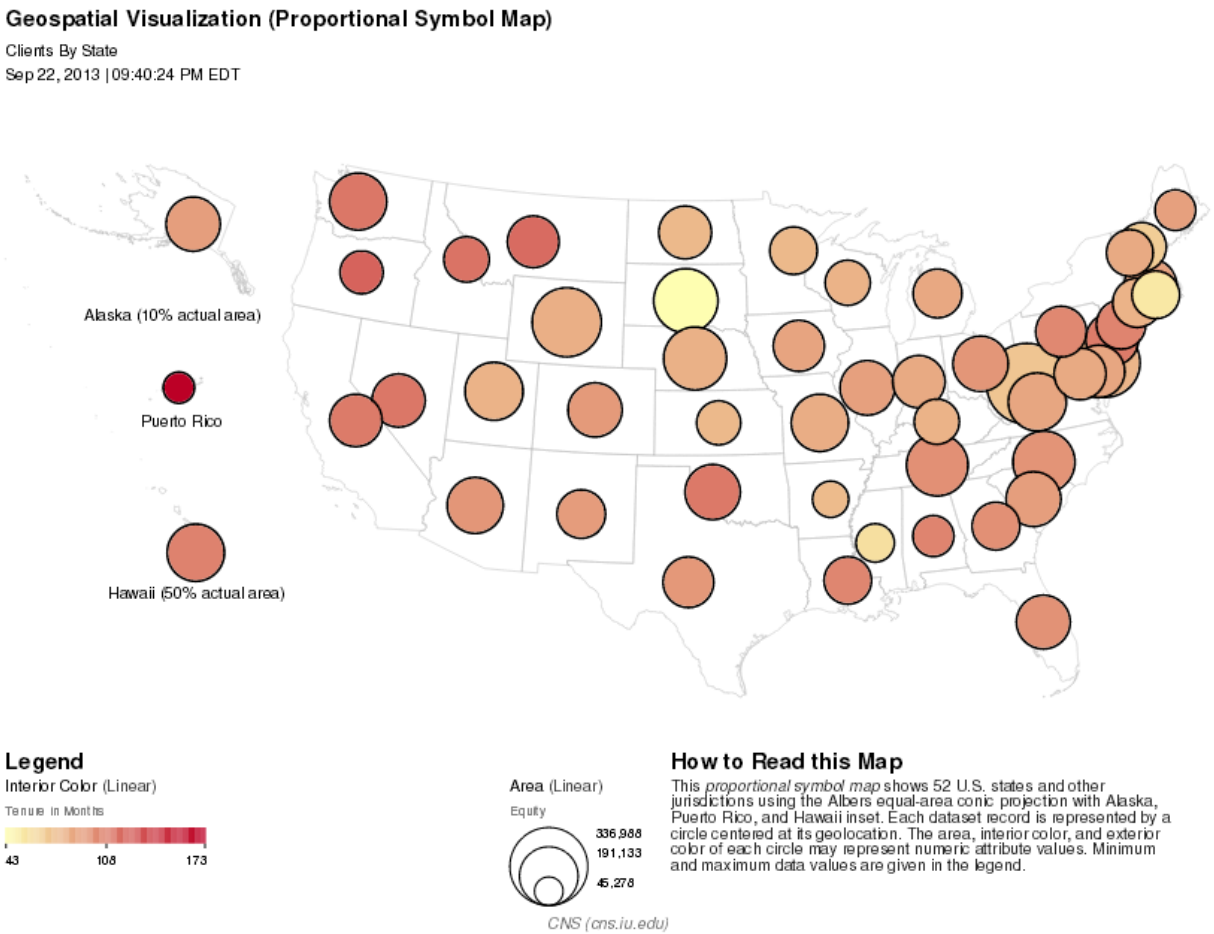
Geographic Distribution of Clients

The result is a PostScript file in the Data Manager. Right-click on the file and select *Save*.



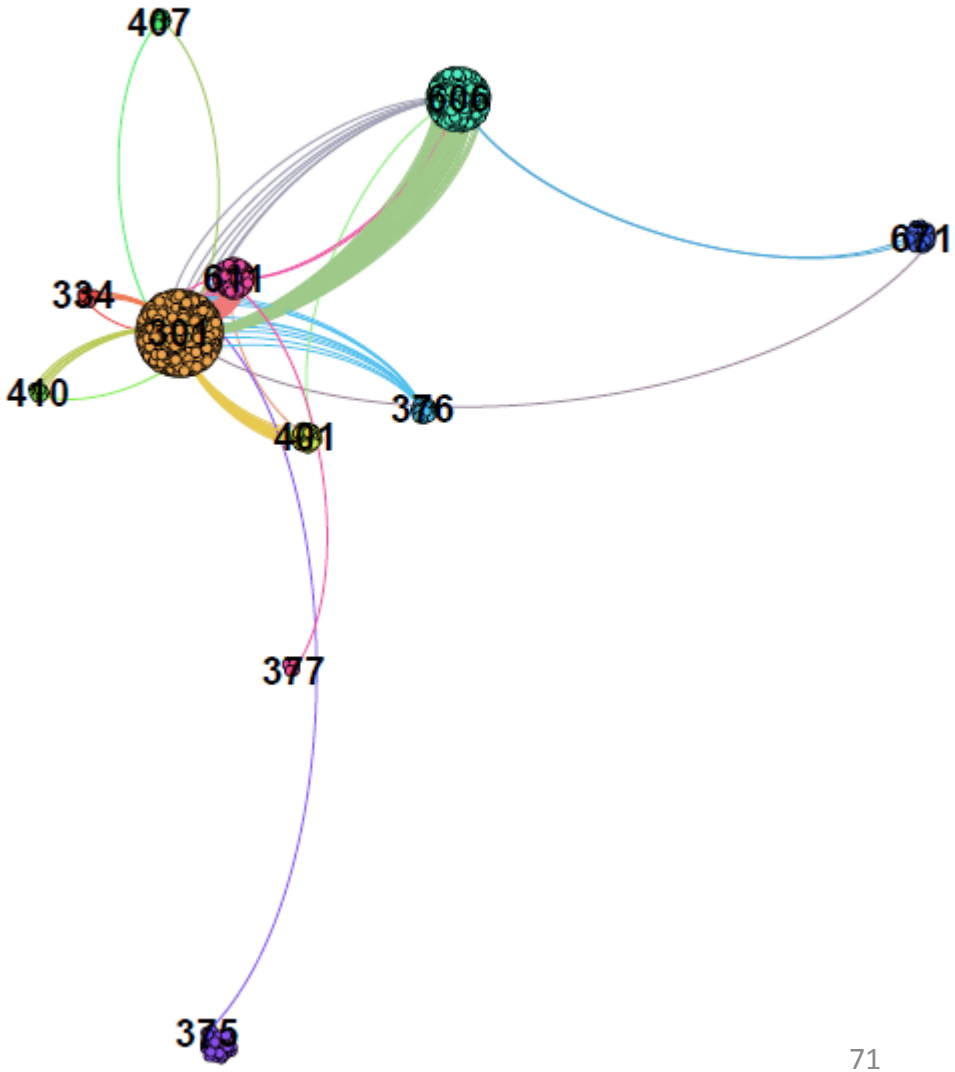
Geographic Distribution of Clients

You will have to convert the PostScript file to a PDF to view it. You can use Adobe Distiller to convert, or an online service, such as http://ps2pdf.com.



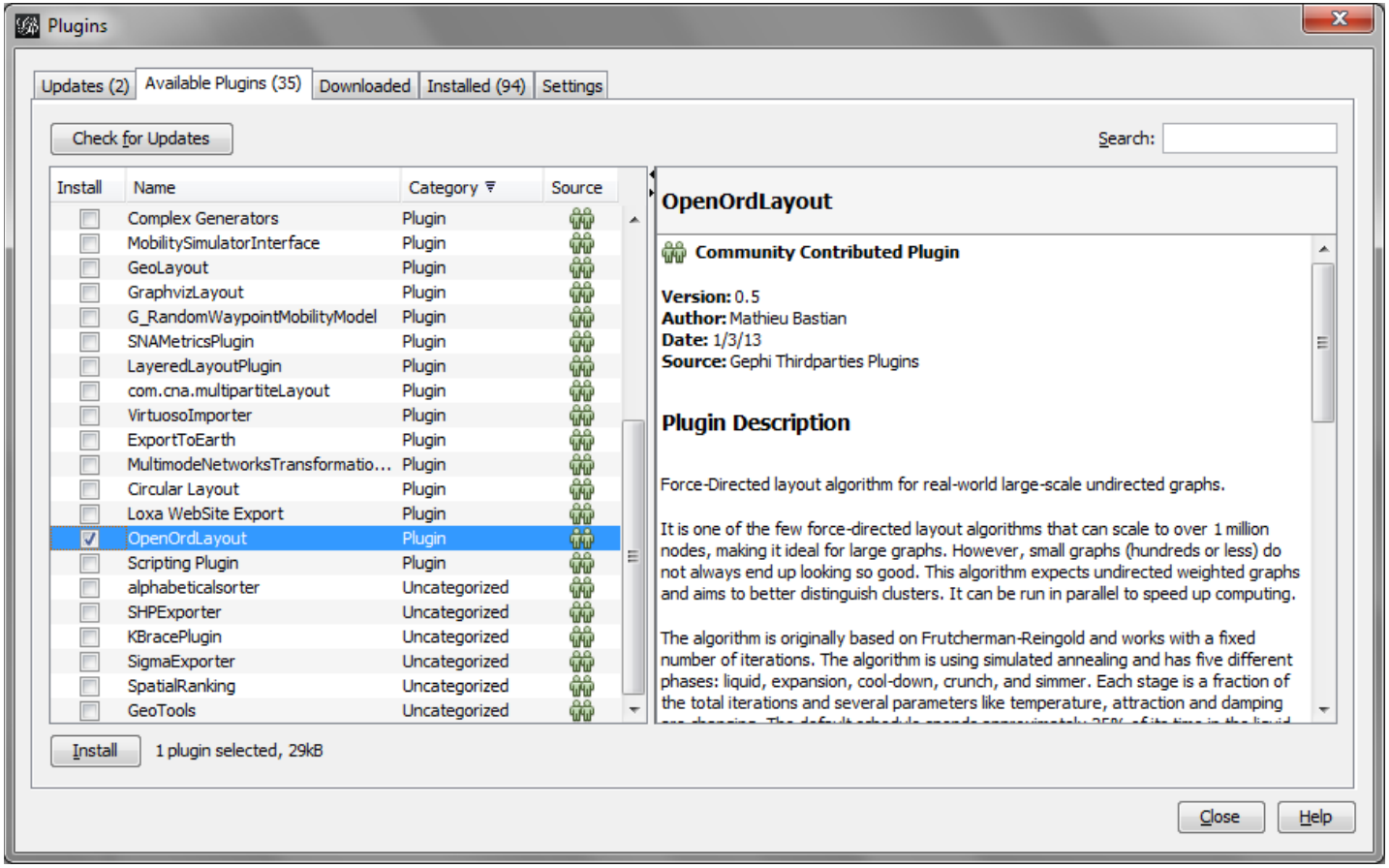
Client Brand Network

Extract a directed network showing the connections between clients and brands.



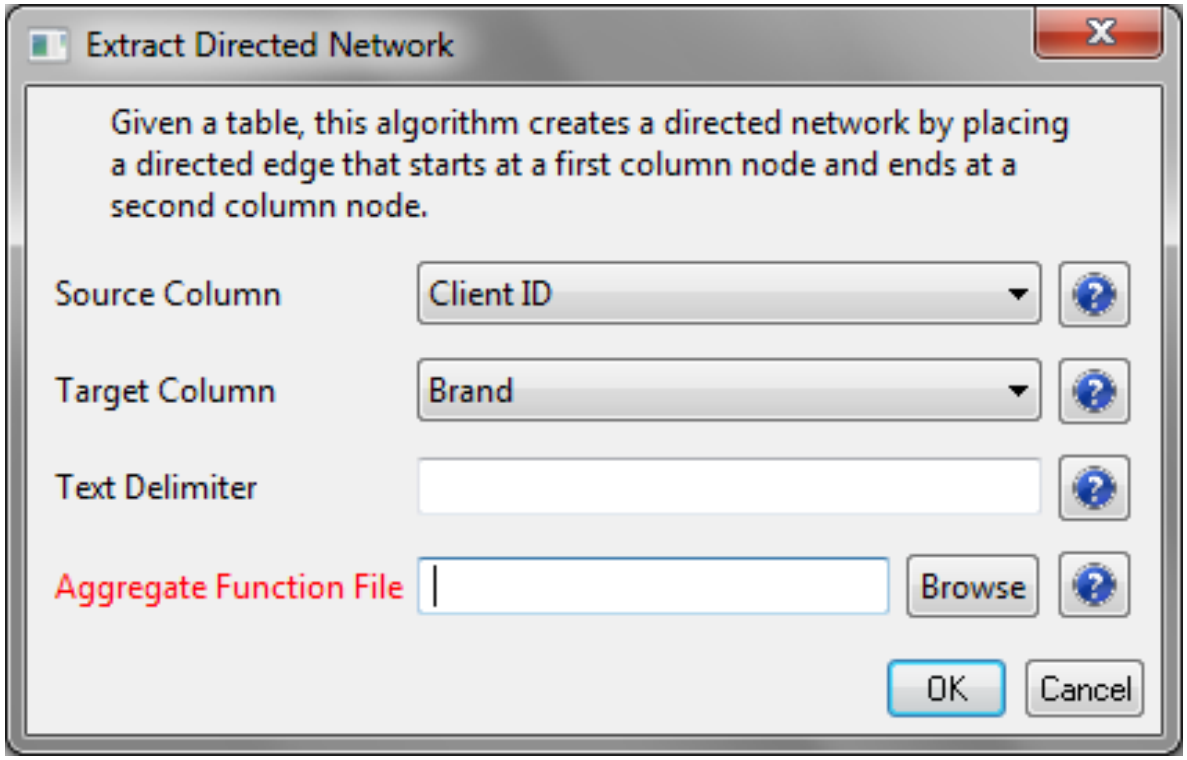
Client Brand Network

This network visualization requires the OpenOrd Network Layout in Gephi. It is available as a plugin. Simply open Gephi and select, *Tools > Plugins.* Then select the Available Plugins tab and find the OpenOrdLayout plugin. Install this plugin and restart Gephi.



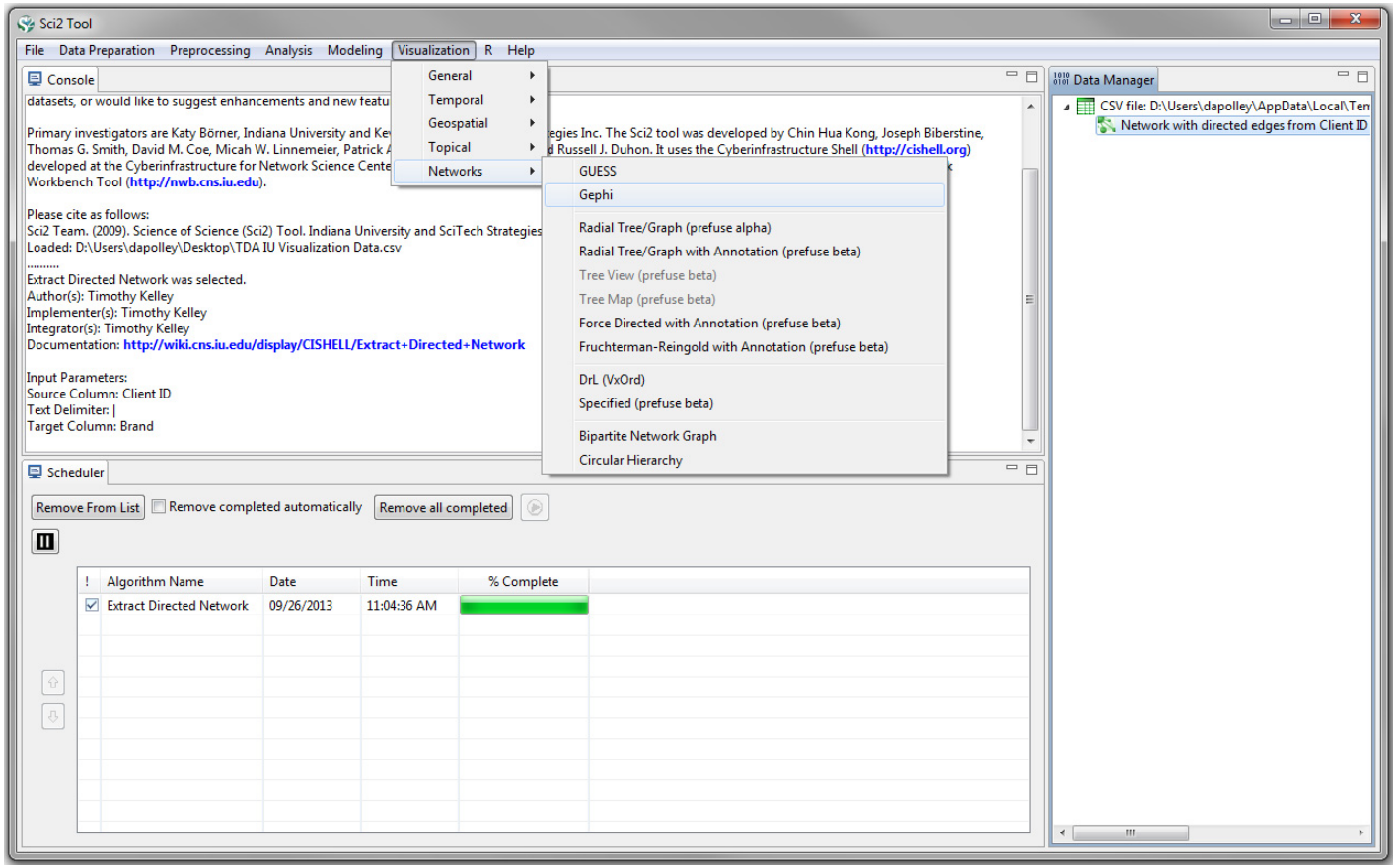
Client Brand Network

Select, *Data Preparation > Extract Directed Network*



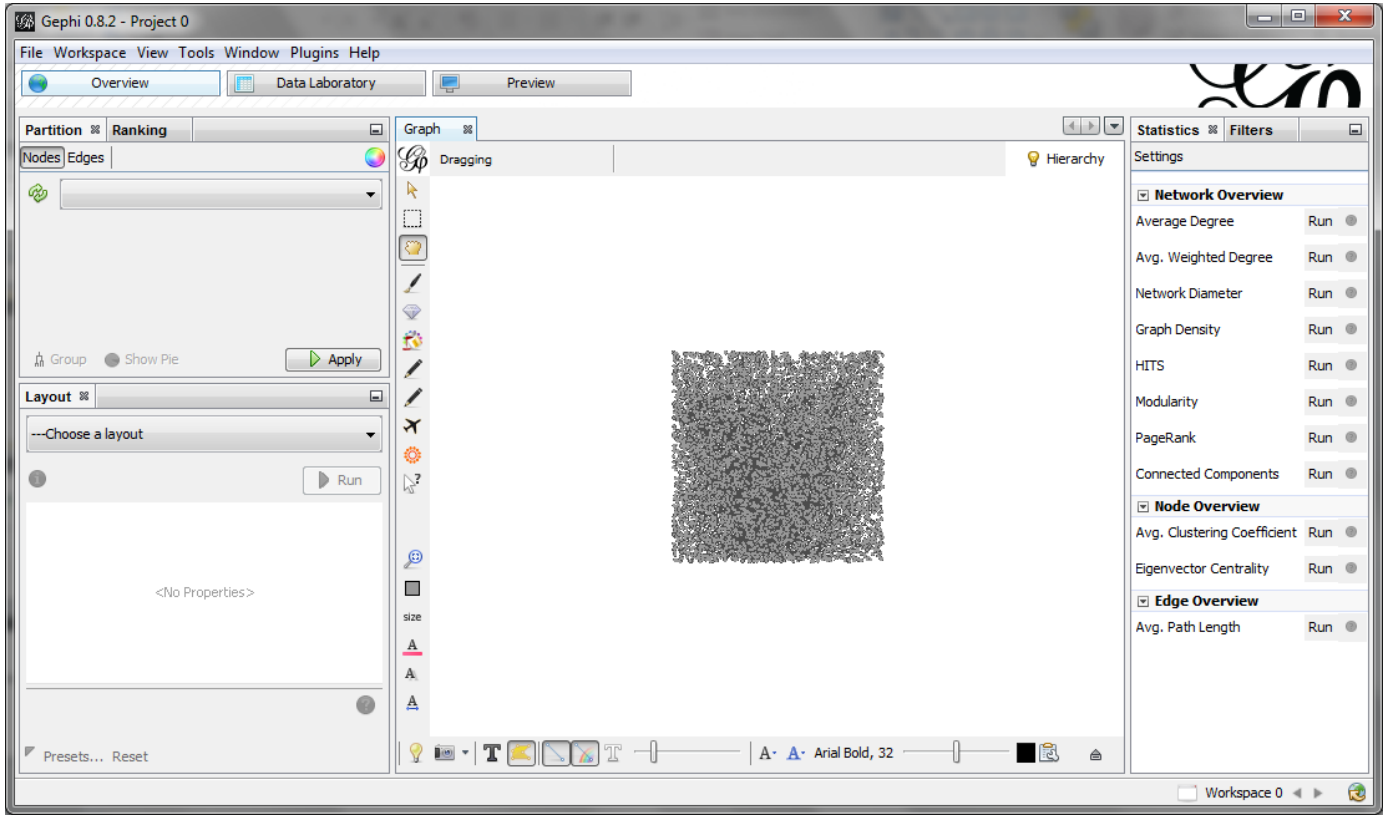
Client Brand Network

Visualize the network with Gephi, *Visualization> Networks > Gephi*



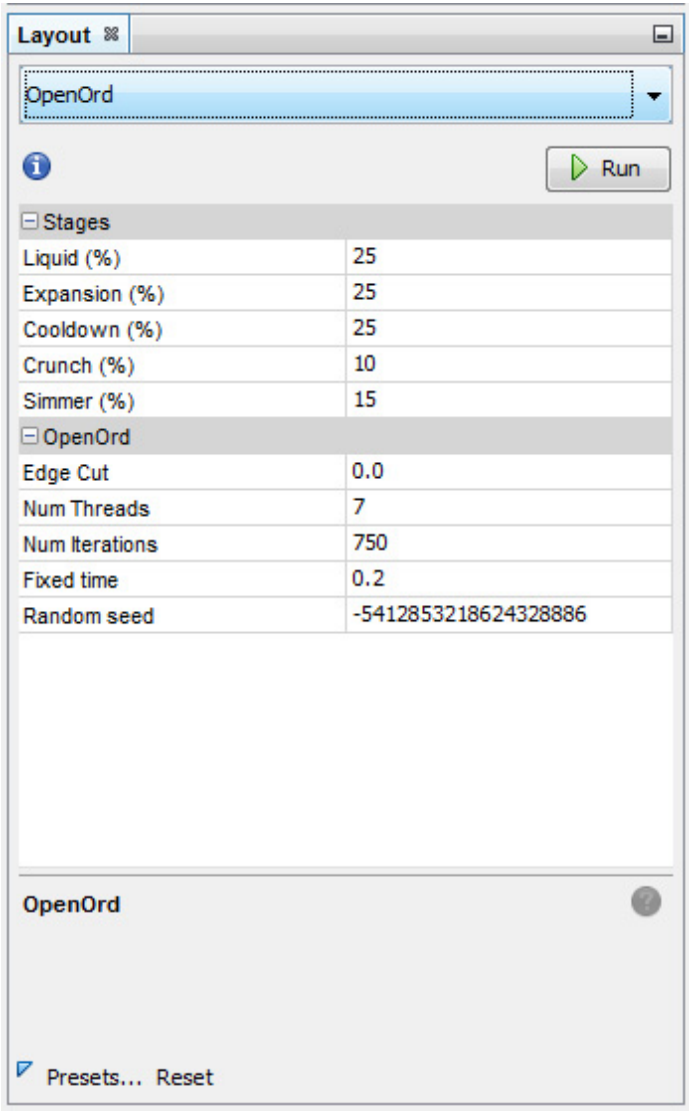
Client Brand Network

Gephi will layout the nodes randomly



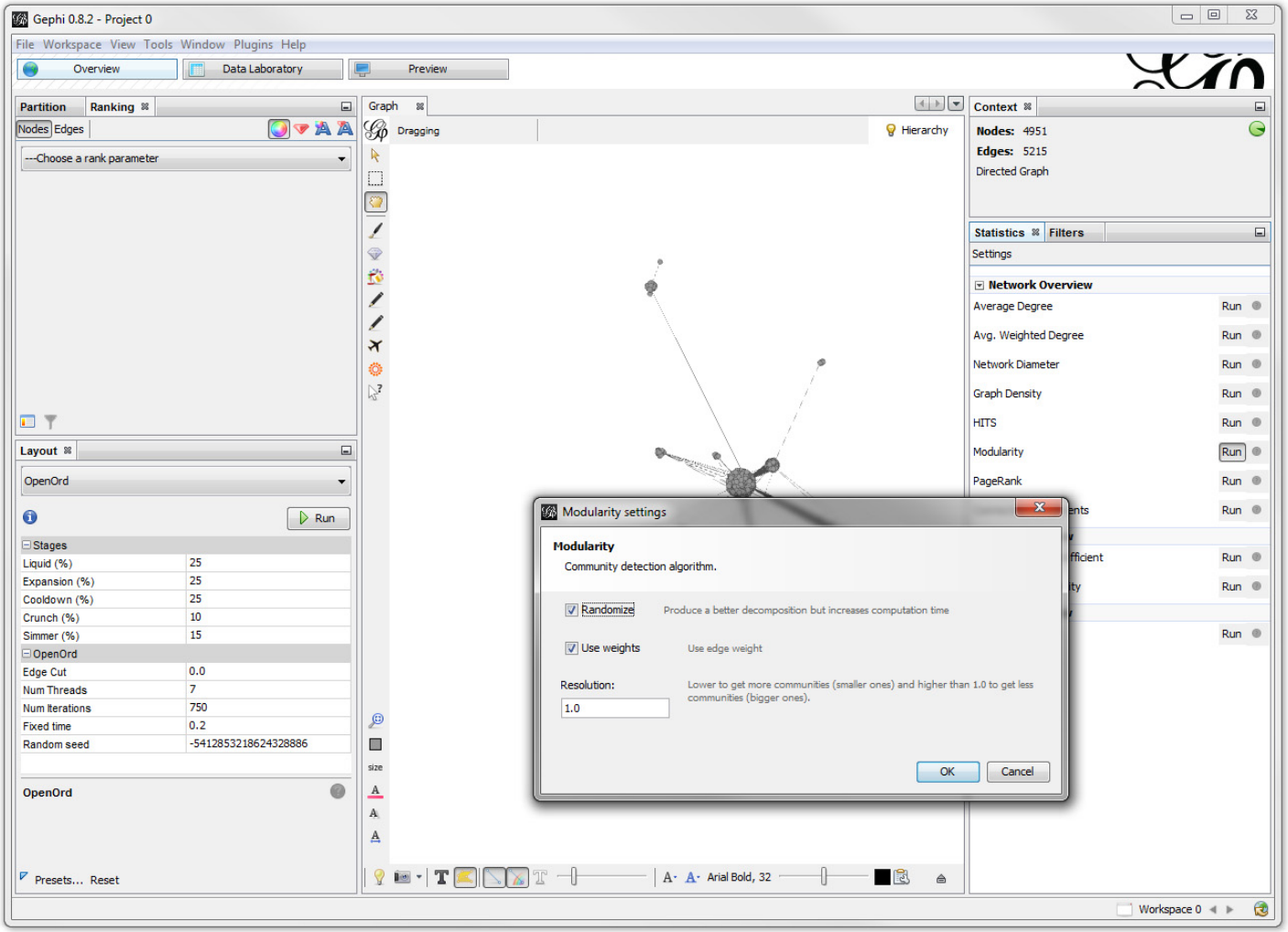
Client Brand Network

Select the OpenOrd layout. This layout is extremely efficient for large networks. Set the Edge Cut parameter to 0.0. This will result in a slightly less clustered result.



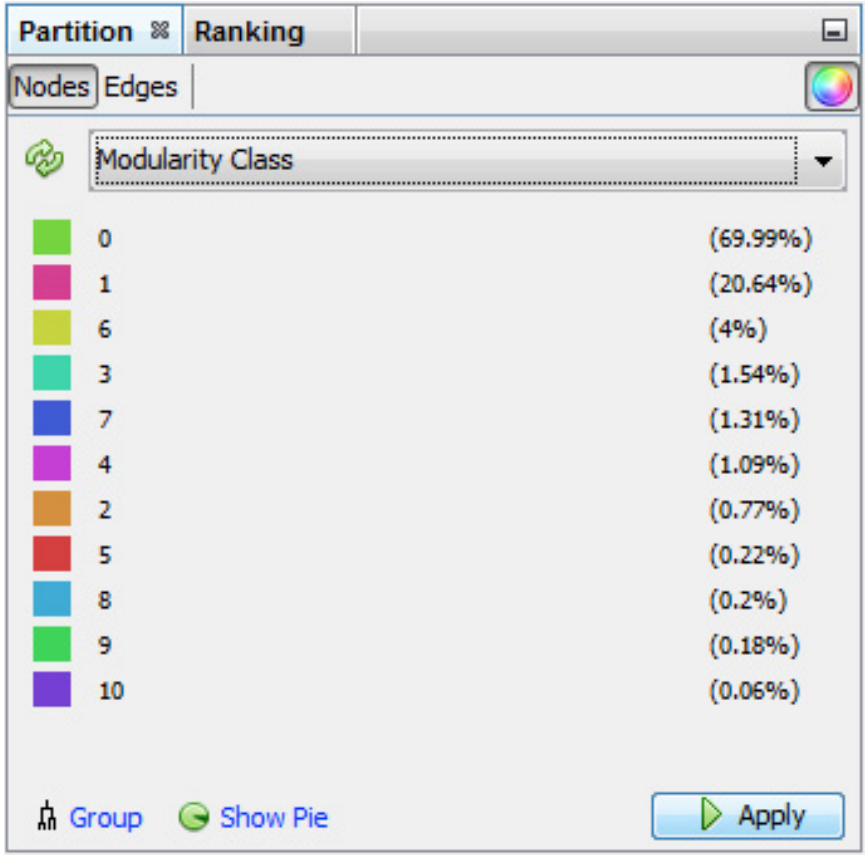
Client Brand Network

In the statistics tab, run Modularity. This identifies communities of clients based on the brands with which they are associated.



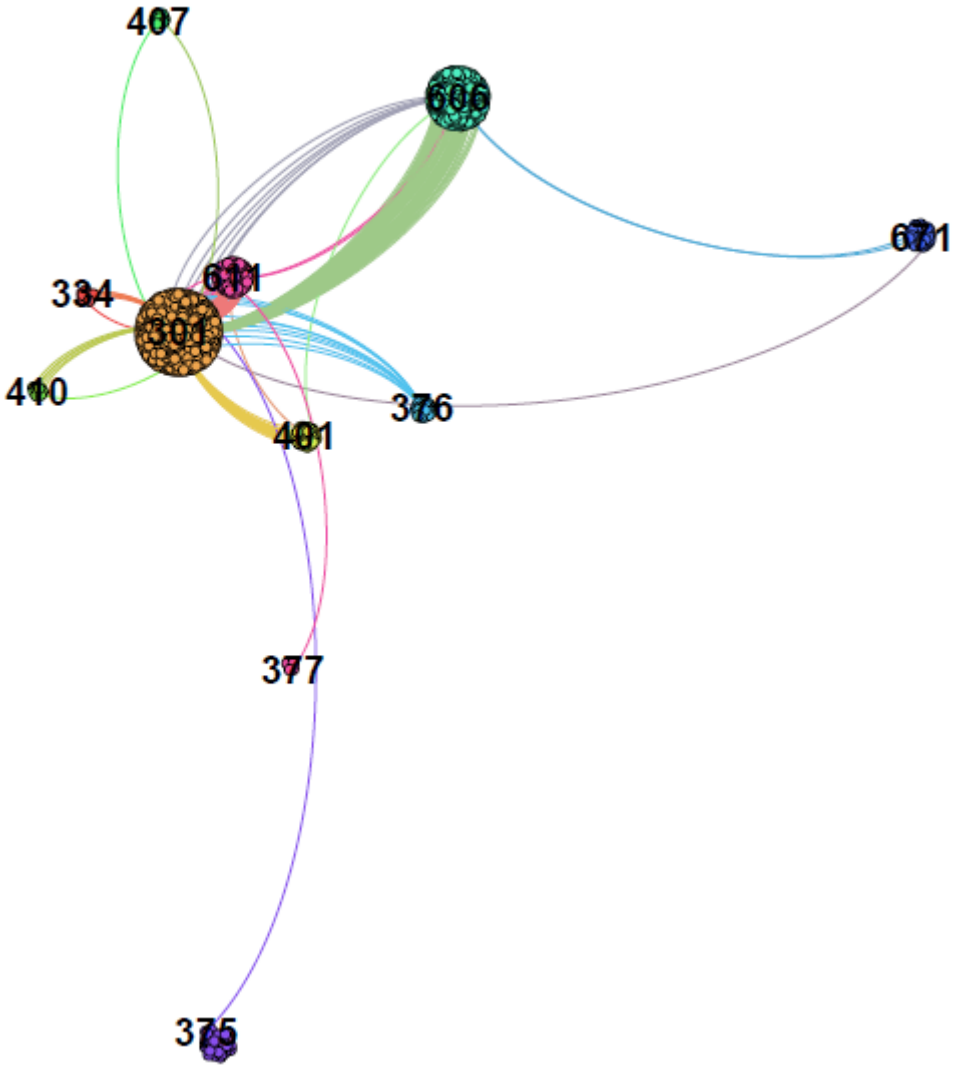
Client Brand Network

Use the partition tab in the upper right-hand corner. Select Modularity Class. You may have to click refresh. Select Apply and the colors shown will be applied to each modularity class.



Client Brand Network

This presents a network where the nodes are distinct clusters based on brand. You can determine brand popularity based on cluster size. Each brand cluster will have a distinct color. There are also a lot of connections between brand clusters.



Brainstorming/Questions?