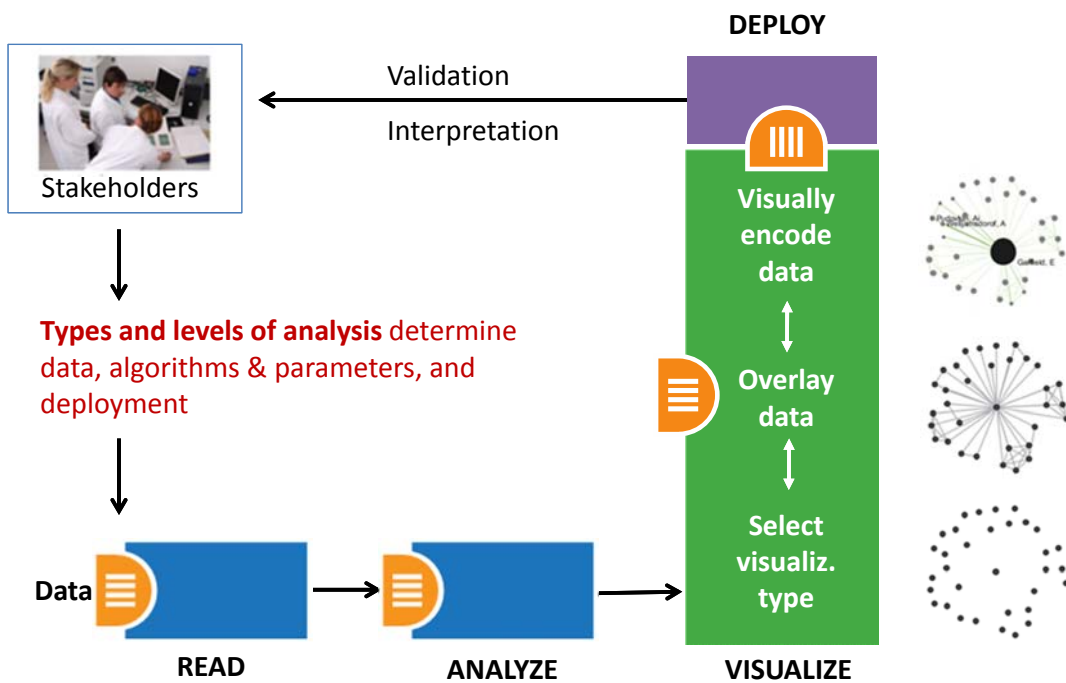


Needs-Driven Workflow Design



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Information Visualization MOOC

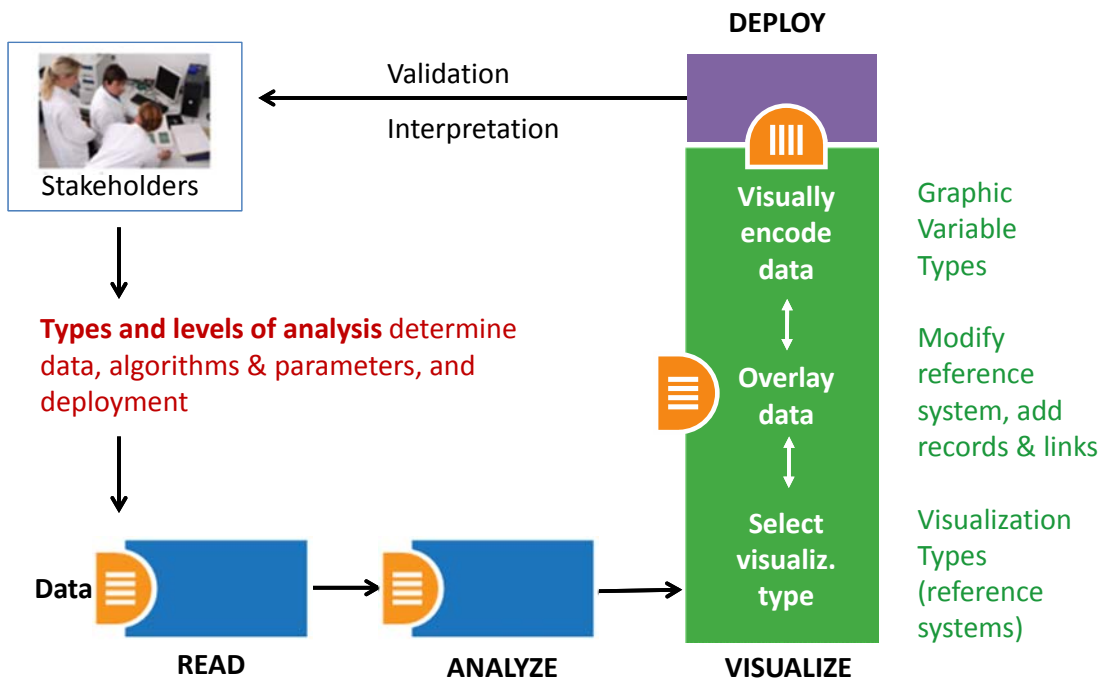
Unit 6 – “With Whom:” Networks

Workflow Design

Reference:

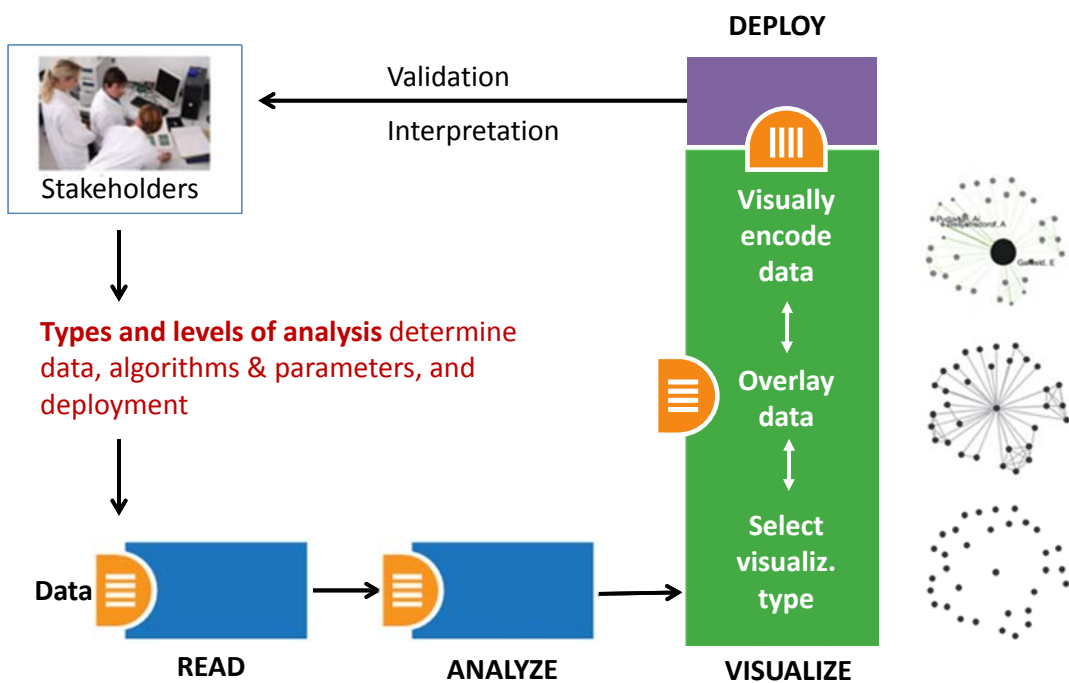
Börner, Katy, Soma Sanyal, and Alessandro Vespignani. 2007. "[Network Science](#)." Chap. 12 in *Annual Review of Information Science & Technology*, edited by Blaise Cronin, 537-607. Medford, NJ: Information Today, Inc./American Society for Information Science and Technology.

Needs-Driven Workflow Design



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Needs-Driven Workflow Design



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Read Data

Sample Data:

- UCINET datasets, <https://sites.google.com/site/ucinetsoftware/datasets>
- Pajek Datasets, <http://pajek.imfm.si/doku.php?id=data:index>
- Gephi datasets, <http://wiki.gephi.org/index.php/Datasets>
- CASOS Datasets, http://www.casos.cs.cmu.edu/computational_tools/datasets
- Stanford Large Network Dataset Collection, <http://snap.stanford.edu/data/>
- Tore Opsahl's Datasets, <http://toreopsahl.com/datasets/>
- Sci2 Datasets, <http://sci2.wiki.cns.iu.edu/display/SCI2TUTORIAL/2.5+Sample+Datasets> and general data sources, <http://sci2.wiki.cns.iu.edu/display/SCI2TUTORIAL/8.1+Datasets>

Major Data Formats:

See next slide

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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)

Image Formats

- JPEG (*.jpg)
- PDF (*.pdf)
- PostScript (*.ps)

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Preprocess Data

- Unification, see Unit 4
- Network extraction
- Delete isolate nodes
- Remove self-loops
- Threshold—e.g., extract nodes/edges above or below value
- Merge two networks
- Pathfinder network scaling, see [Backbone Identification](#)

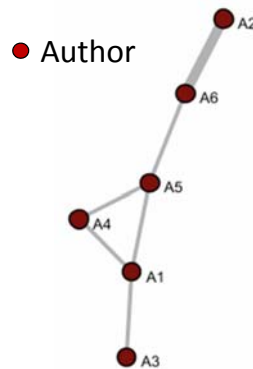
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Preprocessing—Network Extraction

Weighted, undirected co-occurrence network.

↓

	A	B	C	D
1	Paper	Authors	References	Year
2	P1	A1		1970
3	P2	A2;A6	P1	1980
4	P3	A1;A3	P1;P2	1990
5	P4	A1;A4;A5	P2	1995
6	P5	A5;A6	P1;P2;P3;P4	1995
7	P6	A2;A6	P5	2000



*Vertices 6

1 A1

2 A6

3 A2

4 A3

5 A5

6 A4

*Edges 6

2 3 2

1 4 1

1 5 1

5 6 1

1 6 1

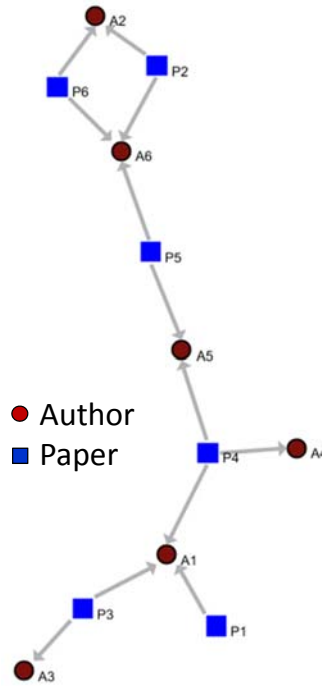
2 5 1

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Preprocessing—Network Extraction

Unweighted, directed bimodal network.

	A	B	C	D
1	Paper	Authors	References	Year
2	P1	A1		1970
3	P2	A2;A6	P1	1980
4	P3	A1;A3	P1;P2	1990
5	P4	A1;A4;A5	P2	1995
6	P5	A5;A6	P1;P2;P3;P4	1995
7	P6	A2;A6	P5	2000

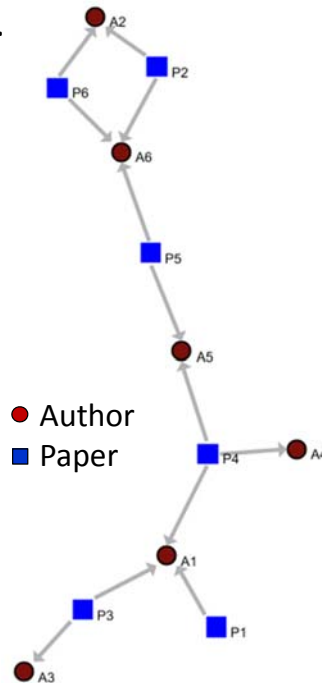


- *Vertices 12
- 1 P1 bipartitetype "Paper"
- 2 A1 bipartitetype "Authors"
- 3 P2 bipartitetype "Paper"
- 4 A2 bipartitetype "Authors"
- 5 A6 bipartitetype "Authors"
- 6 P3 bipartitetype "Paper"
- 7 A3 bipartitetype "Authors"
- 8 P4 bipartitetype "Paper"
- 9 A4 bipartitetype "Authors"
- 10 A5 bipartitetype "Authors"
- 11 P5 bipartitetype "Paper"
- 12 P6 bipartitetype "Paper"
- *Arcs
- 1 2
- 3 4
- 3 5
- 6 2
- 6 7
- 8 2
- 8 10
- 8 9
- 11 5
- 11 10
- 12 4
- 12 5

Preprocessing—Network Extraction

Unweighted, directed network of two types.
Calculate node degrees.

	A	B	C	D
1	Paper	Authors	References	Year
2	P1	A1		1970
3	P2	A2;A6	P1	1980
4	P3	A1;A3	P1;P2	1990
5	P4	A1;A4;A5	P2	1995
6	P5	A5;A6	P1;P2;P3;P4	1995
7	P6	A2;A6	P5	2000

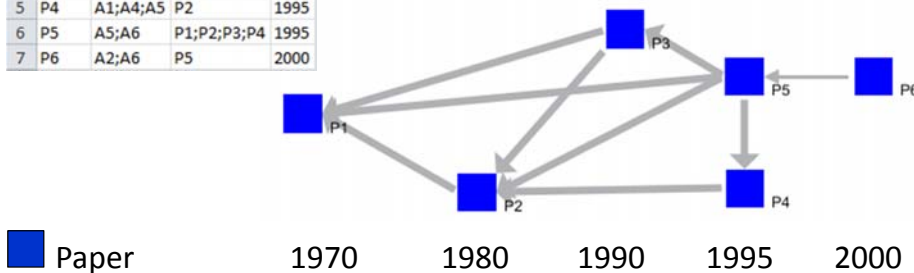


- *Vertices 12
- 1 P1 indegree 0
- 2 A1 indegree 3
- 3 P2 indegree 0
- 4 A2 indegree 2
- 5 A6 indegree 3
- 6 P3 indegree 0
- 7 A3 indegree 1
- 8 P4 indegree 0
- 9 A4 indegree 1
- 10 A5 indegree 2
- 11 P5 indegree 0
- 12 P6 indegree 0
- *Arcs
- 1 2
- 3 4
- 3 5
- 6 2
- 6 7
- 8 10
- 8 2
- 8 9
- 11 10
- 11 5
- 12 4
- 12 5

Preprocessing—Network Extraction

Unweighted, directed paper-citation network.
Arcs go from papers to references.

	A	B	C	D
1	Paper	Authors	References	Year
2	P1	A1		1970
3	P2	A2;A6	P1	1980
4	P3	A1;A3	P1;P2	1990
5	P4	A1;A4;A5	P2	1995
6	P5	A5;A6	P1;P2;P3;P4	1995
7	P6	A2;A6	P5	2000



*Vertices 6

- 1 P1
- 2 P2
- 3 P3
- 4 P4
- 5 P5
- 6 P6

*Arcs

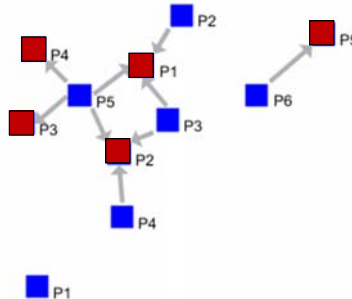
- 2 1
- 3 1
- 3 2
- 4 2
- 5 4
- 5 3
- 5 1
- 5 2
- 6 5

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Preprocessing—Network Extraction

Unweighted, directed bipartite network.

	A	B	C	D
1	Paper	Authors	References	Year
2	P1	A1		1970
3	P2	A2;A6	P1	1980
4	P3	A1;A3	P1;P2	1990
5	P4	A1;A4;A5	P2	1995
6	P5	A5;A6	P1;P2;P3;P4	1995
7	P6	A2;A6	P5	2000



*Vertices 11

- 1 P1 bipartitetype "Paper"
- 2 P2 bipartitetype "Paper"
- 3 P1 bipartitetype "References"
- 4 P3 bipartitetype "Paper"
- 5 P2 bipartitetype "References"
- 6 P4 bipartitetype "Paper"
- 7 P5 bipartitetype "Paper"
- 8 P4 bipartitetype "References"
- 9 P3 bipartitetype "References"
- 10 P6 bipartitetype "Paper"
- 11 P5 bipartitetype "References"

*Arcs

- 2 3
- 4 3
- 4 5
- 6 5
- 7 3
- 7 9
- 7 5
- 7 8
- 10 11

WRONG!!!

■ ■ Paper

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Network Analysis

Calculate

- Node and link properties
- Network properties
- Statistical properties
- Network types

- Extract relevant subtrees

- Calculate error and attack tolerance
- Compute clusters and backbones, see later parts in this unit

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Visualization Goals

Representing hierarchical data

- Structural information
- Content information

Objectives

- Efficient space utilization
- Comprehension
- Interactivity
- Aesthetics

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Example: Collaboration Network

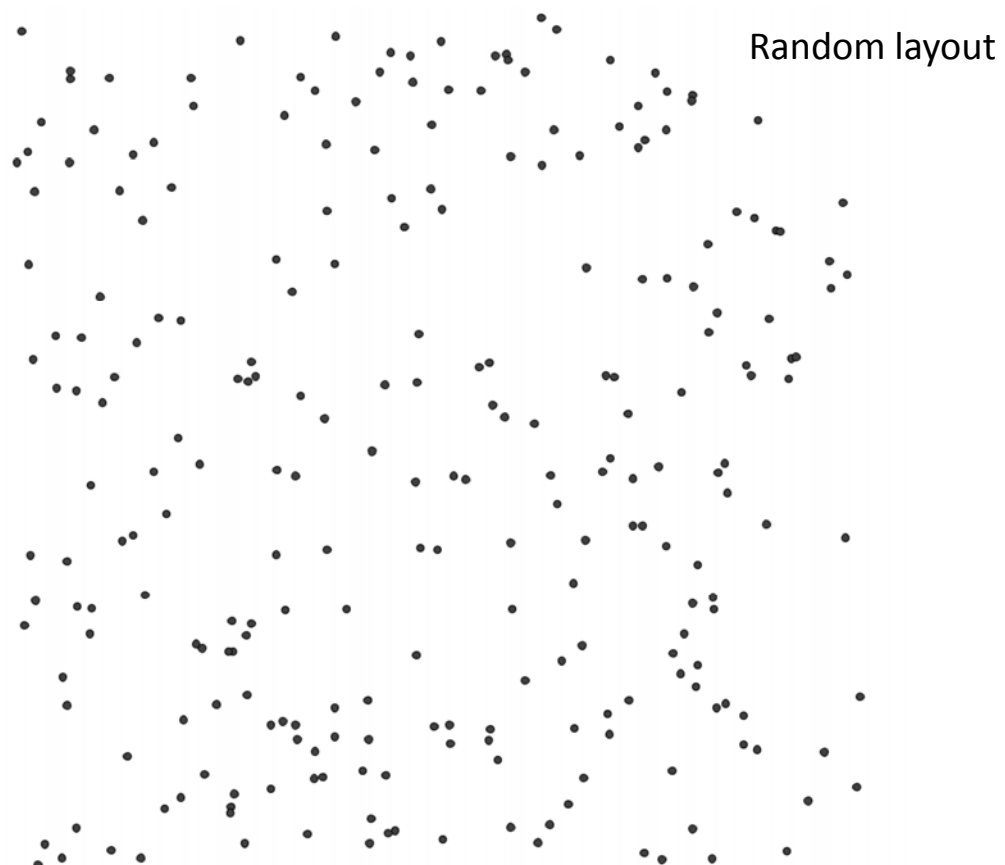
- Random layout
- Circular layout
- Generalized Expectation-Maximization (GEM) layout

First, determine node layout

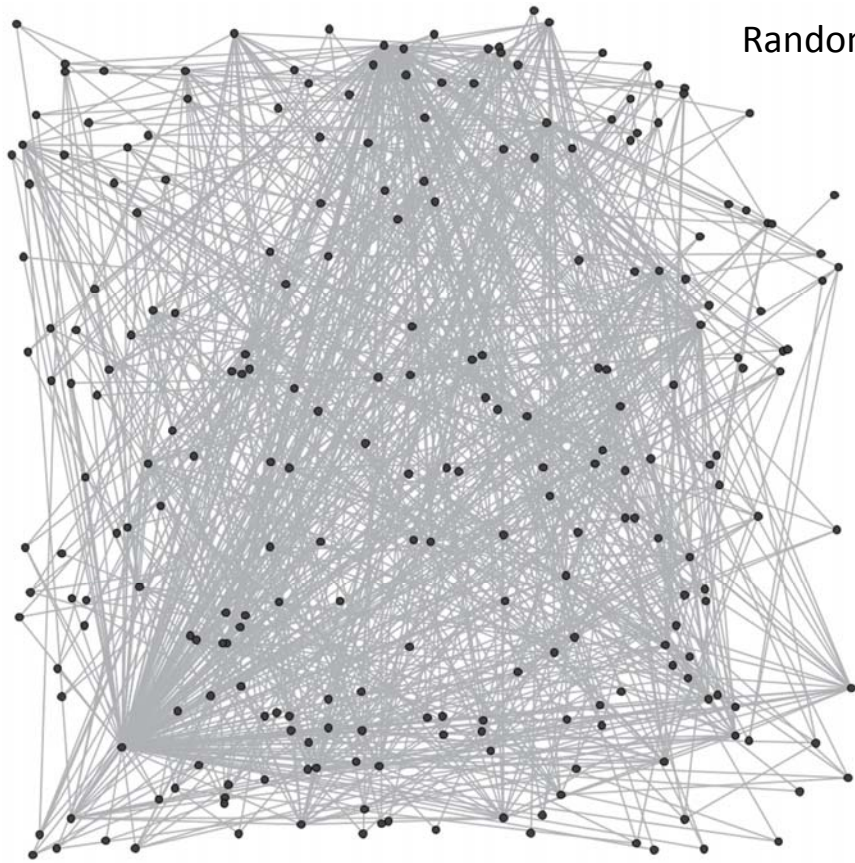
Then add nodes and linkages

For GEM layout, color- and size-code nodes and links and add a legend.

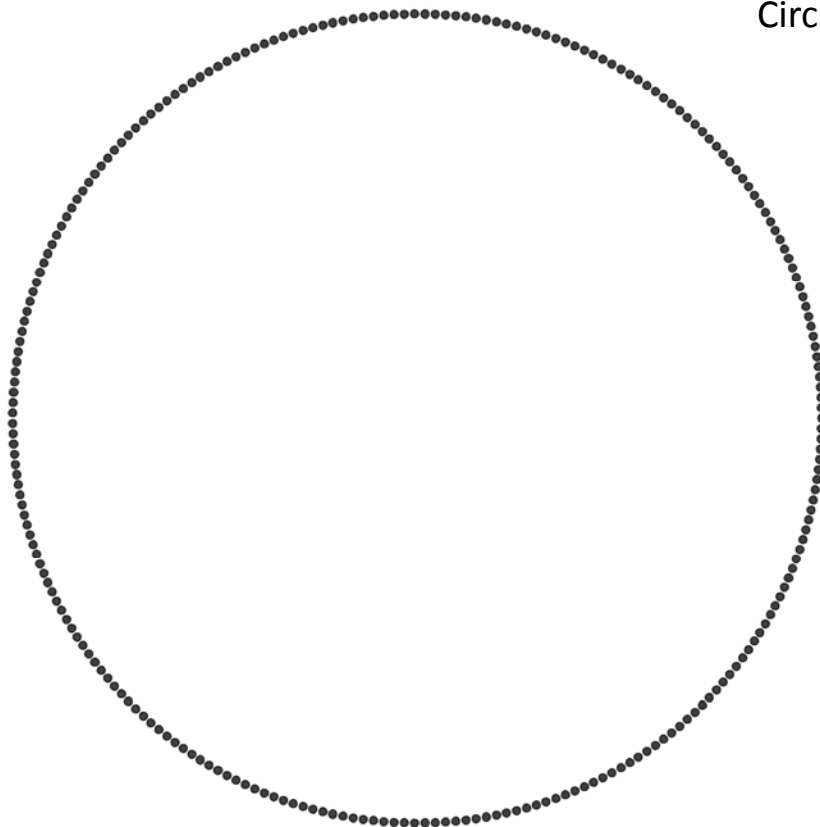
41



42

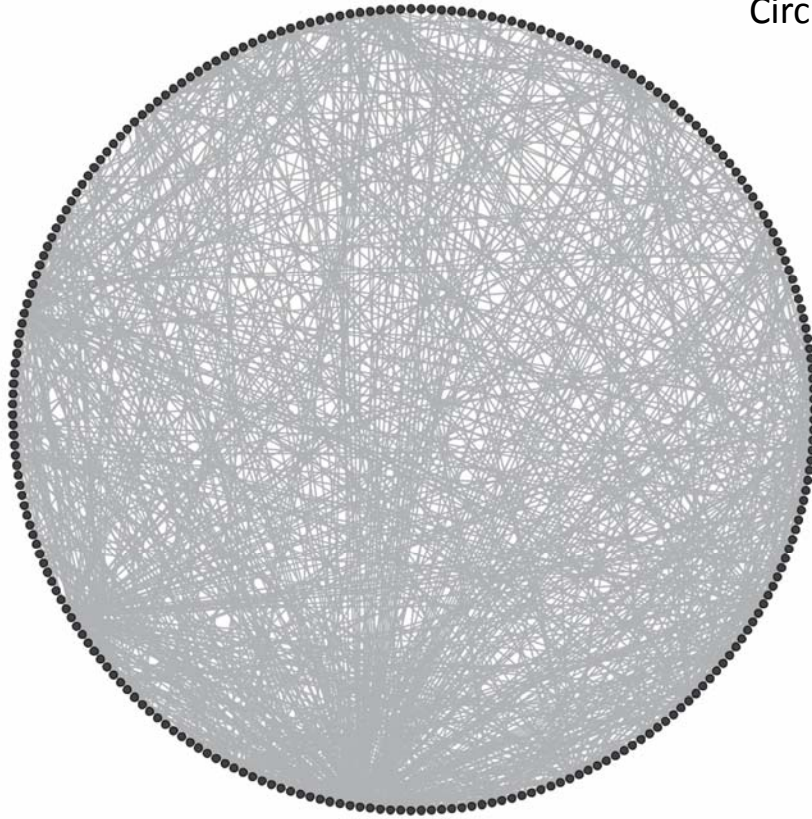


Random layout



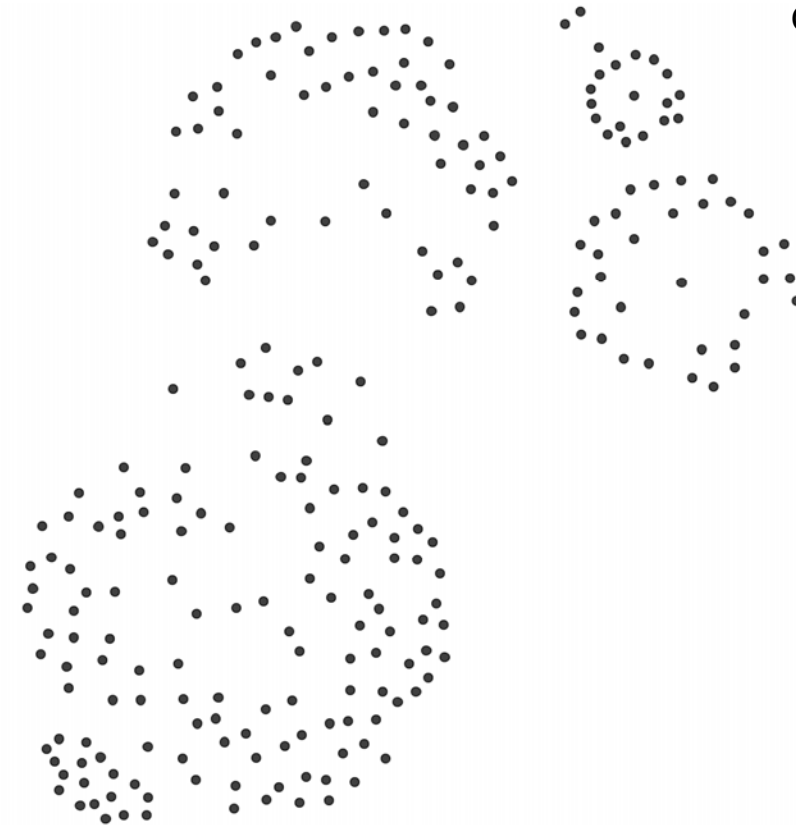
Circular layout

Circular layout

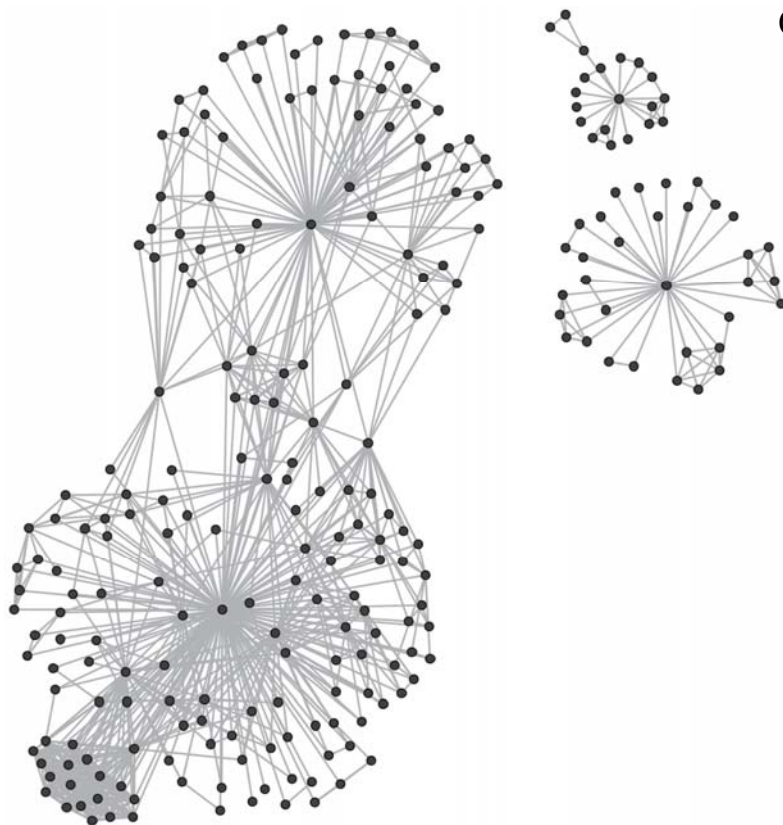


45

GEM layout



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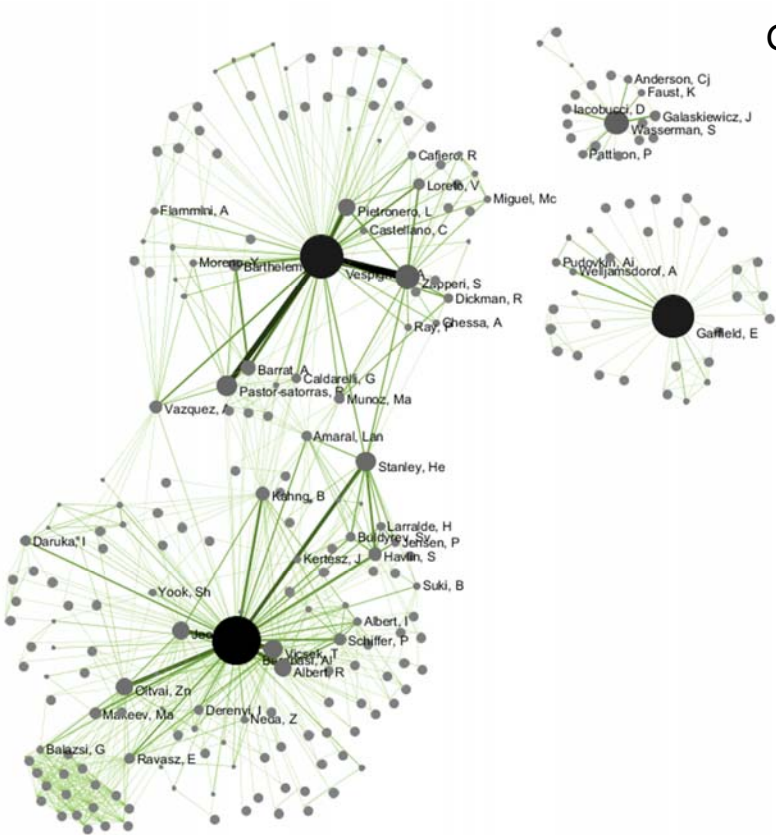


GEM layout

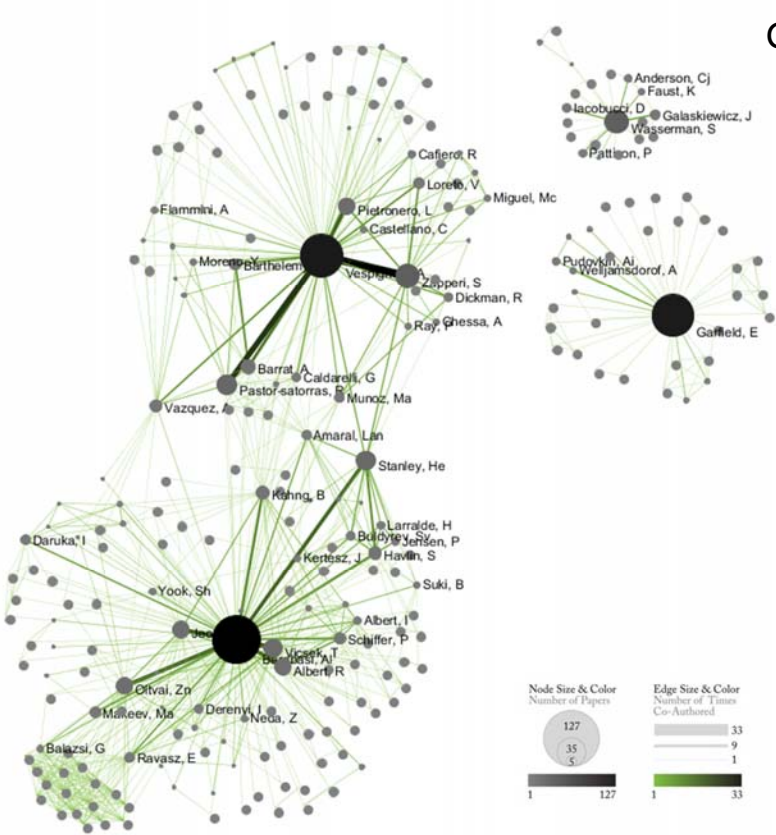


GEM layout

GEM layout

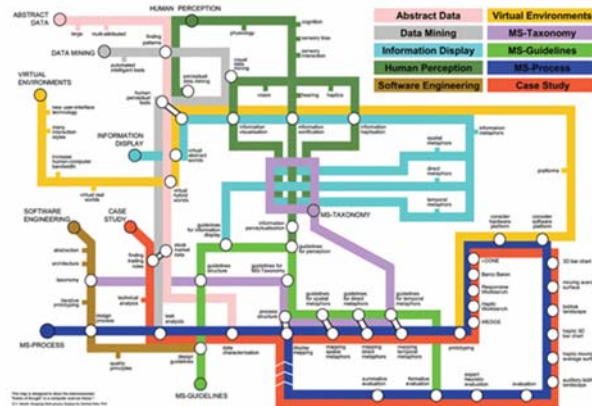


GEM layout



Aesthetic Criteria for Graph Drawing

- Maximize symmetry
- Evenly distributed nodes
- Uniform edge lengths
- Minimized edge crossings
- Orthogonal drawings
- Minimize area, bends, slopes, angles
- Maximize consistent flow direction (in directed networks)



Optimization criteria may be relaxed to speed up layout process.

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Visualizing Large Networks

Discover landmark nodes based on

- Existing node attributes—e.g., frequency of access
- Connectivity (hubs & authorities)
- Depth in a hierarchy

Identify strong (and weak) links

Identify backbone, see [Backbone Identification](#)

Show clusters, see [Clustering](#)

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Interacting With Networks

Modify focusing parameters while continuously providing visual feedback and updating display (fast computer response).

- Conditioning: filter, set background variables and display foreground parameters
- Identification: highlight, color, shape code
- Parameter control: line thickness, length, color legend, time slider, and animation control
- Navigation: bird's-eye view, zoom, and pan
- Information requests: mouse over or click on a node to retrieve more details or collapse/expand a subnetwork