

Final project

Design new visualization method (e.g. software)

- Pose problem, Implement creative solution
- Design studies/evaluations less common but also possible (talk to us)

Deliverables

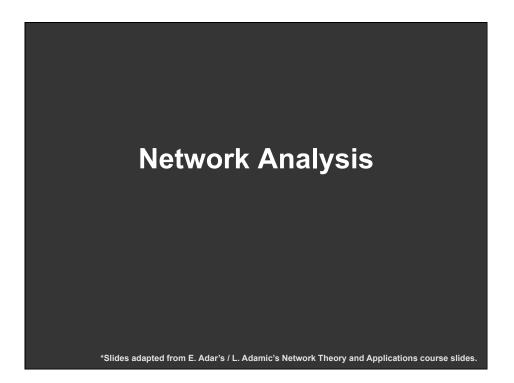
- Implementation of solution
- **6**-8 page paper in format of conference paper submission
- Project progress presentations

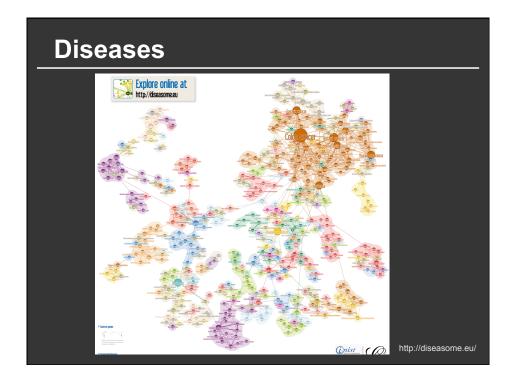
Schedule

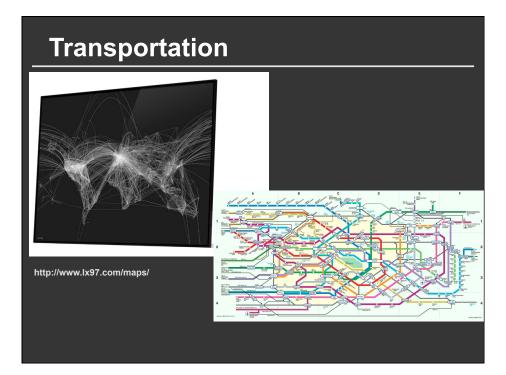
- Project proposal: 5/11
- Project progress presentation: 5/23 in class (3-4 min) slide presentation
- Final poster presentation: 6/3 12:15-3:15pm Location: Lathrop 282
- Final paper: 6/5 11:59pm

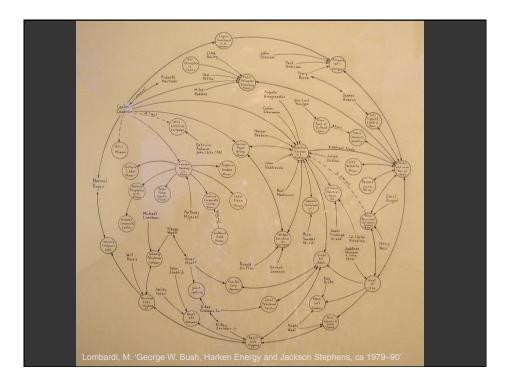
Grading

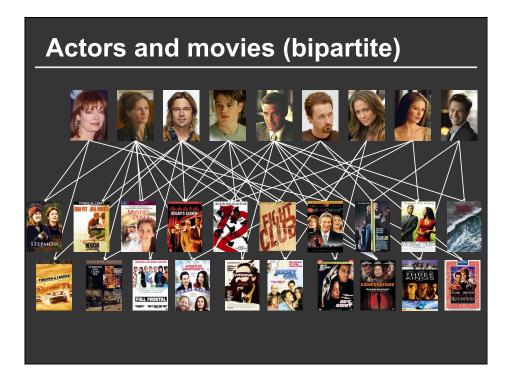
- Groups of up to 3 people, graded individually
- Clearly report responsibilities of each member

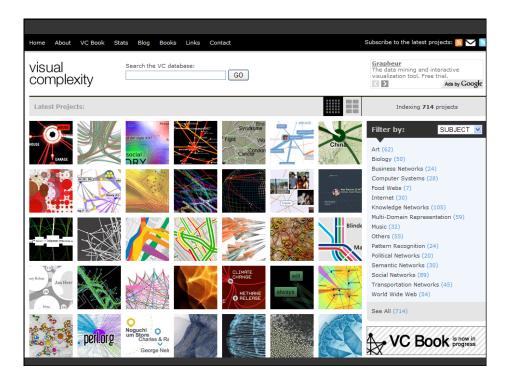


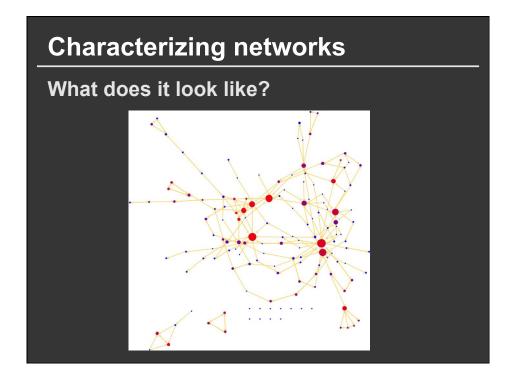


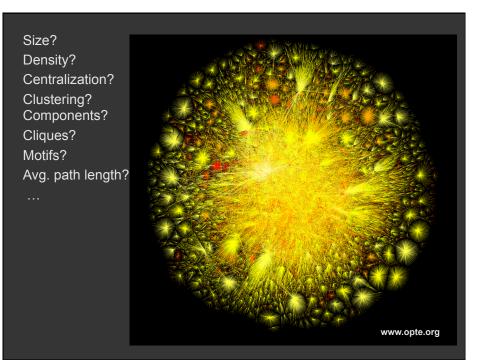










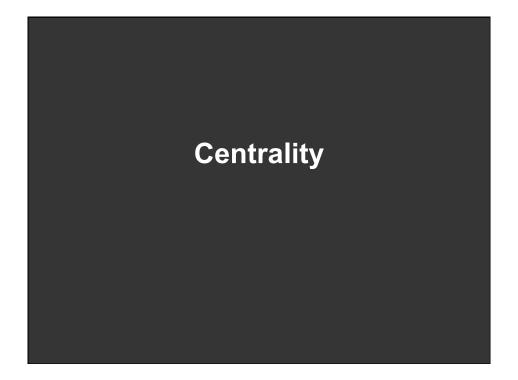


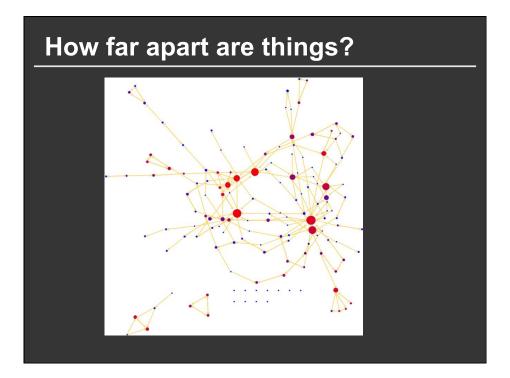
Topics

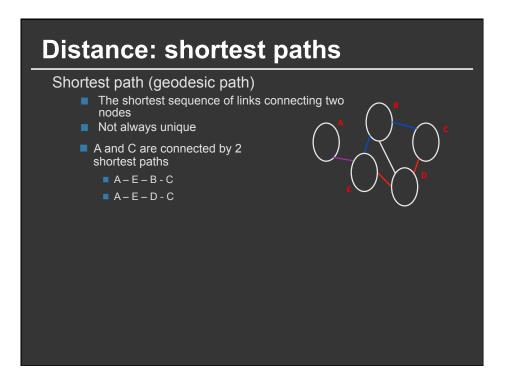
Network Analysis

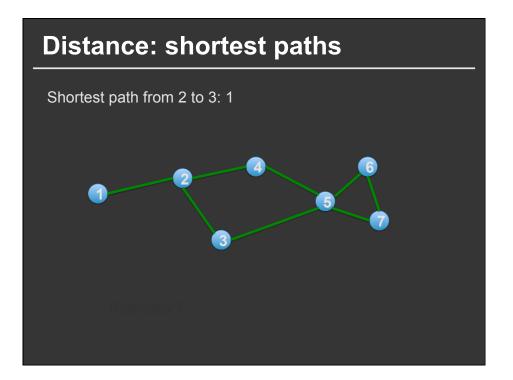
- Centrality / centralization
- Community structure
- Pattern identification
- Models

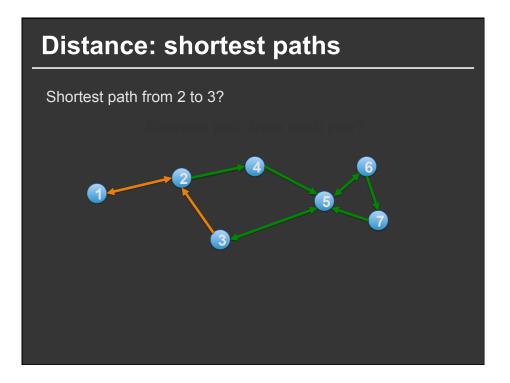
Tools for Network EDA

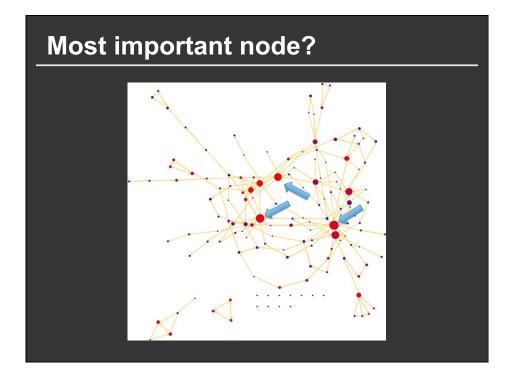


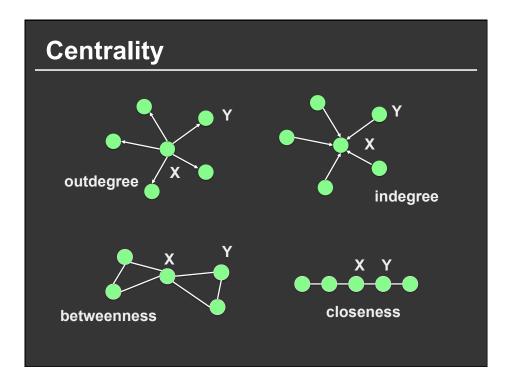


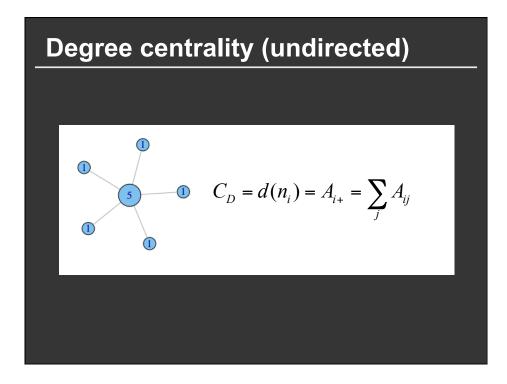


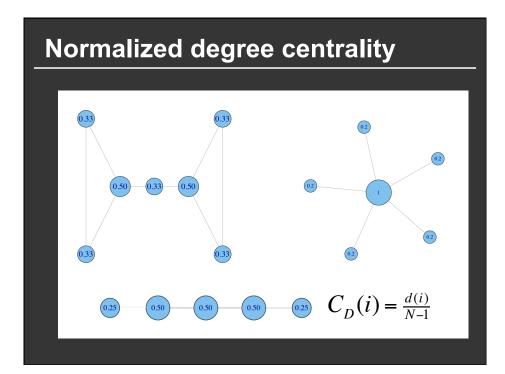








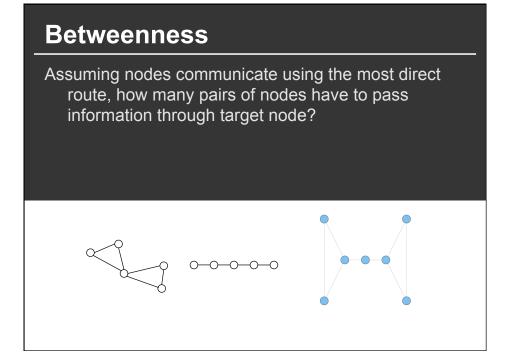


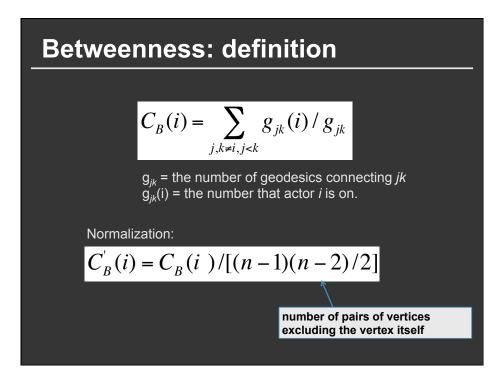


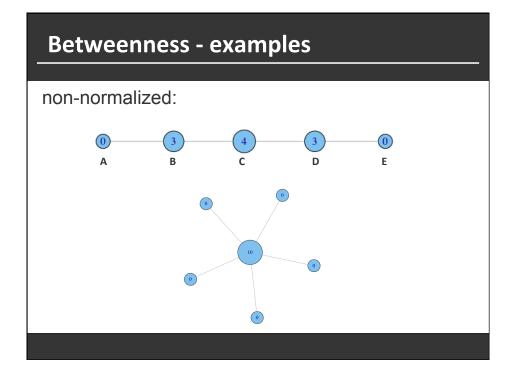
When is degree not sufficient?

Ability to broker between groups

Likelihood that information originating anywhere in the network reaches you







When are C_d, C_b not sufficient?

Likelihood that information originating anywhere in the network reaches you

Closeness: definition

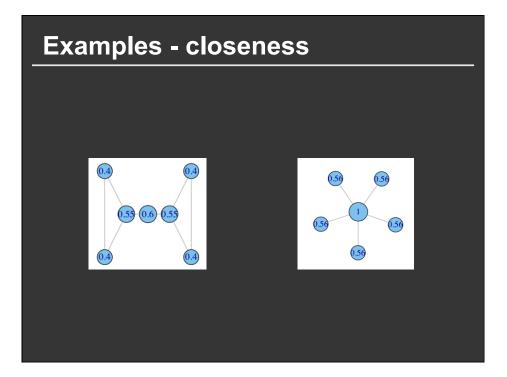
Being close to the center of the graph

Closeness Centrality:

$$C_{c}(i) = \left[\sum_{j=1, j \neq i}^{N} d(i, j)\right]^{-1}$$

Normalized Closeness Centrality

$$C_{C}(i) = (C_{C}(i)) / (N-1) = \frac{N-1}{\sum_{j=1, j \neq i}^{N} d(i, j)}$$



Centrality in directed networks

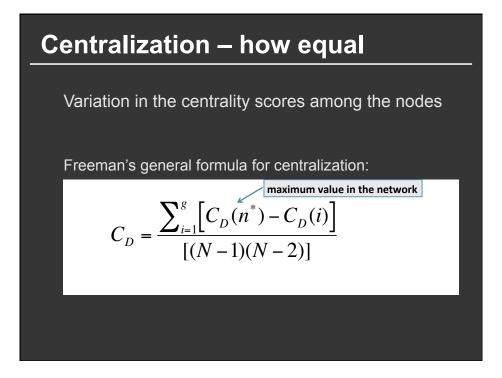
Prestige ~ indegree centrality

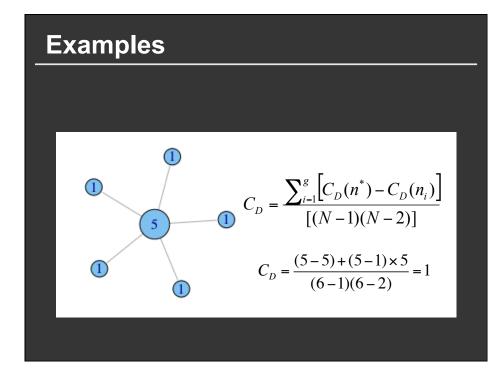
Closeness ~ consider nodes from which target node can be reached Influence range ~ nodes reachable from target node Betweenness ~ consider directed shortest paths

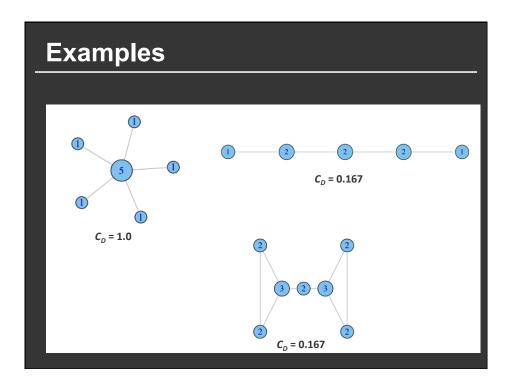
Straight-forward modifications to equations for non-directed graphs

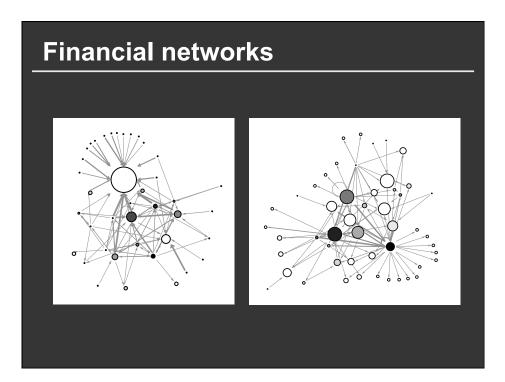
Characterizing nodes

	Low Degree	Low Closeness	Low Betweenness
High Degree		Node embedded in cluster that is far from the rest of the network	Node's connections are redundant - communication bypasses him/her
High Closeness	Node links to a small number of important/active other nodes.		Many paths likely to be in network; node is near many people, but so are many others
High Betweenness	Node's few ties are crucial for network flow	Rare. Node monopolizes the ties from a small number of people to many others.	

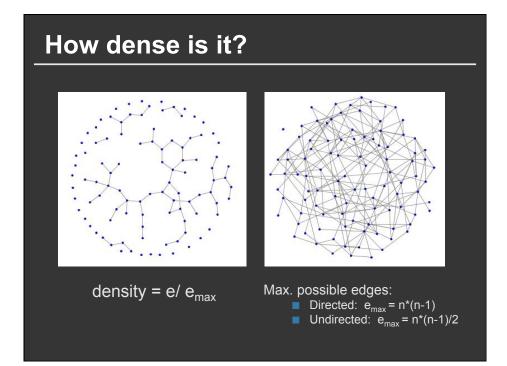


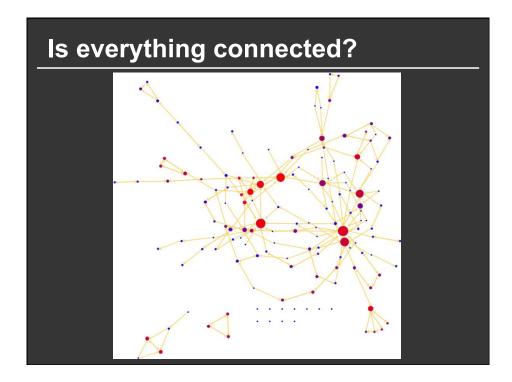


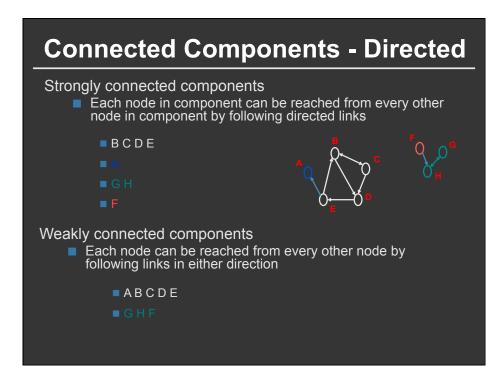


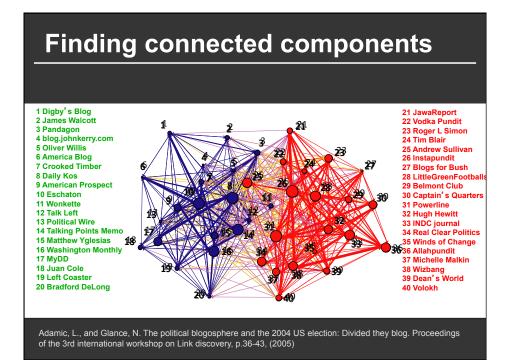


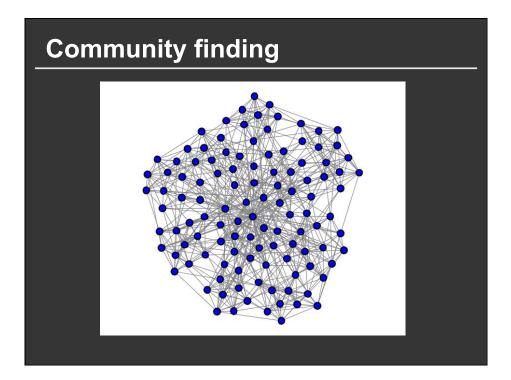








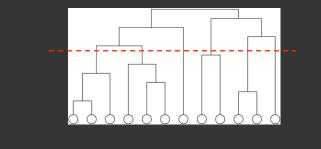


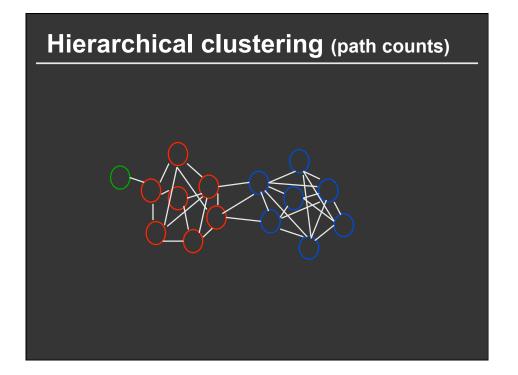


Hierarchical clustering

Process:

- Calculate affinity weights *W* for all pairs of vertices
- Start: *N* disconnected vertices
- Adding edges (one by one) between pairs of clusters in order of decreasing weight (use closest distance to compare clusters)
- Result: nested components

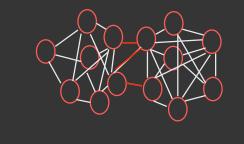


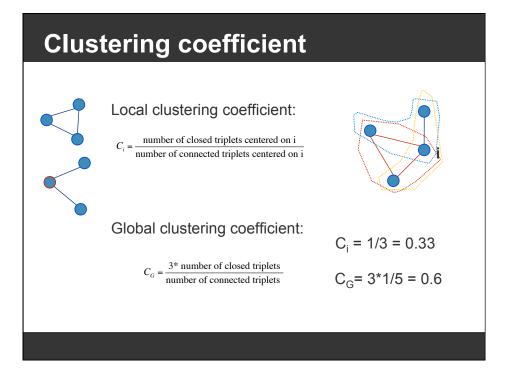


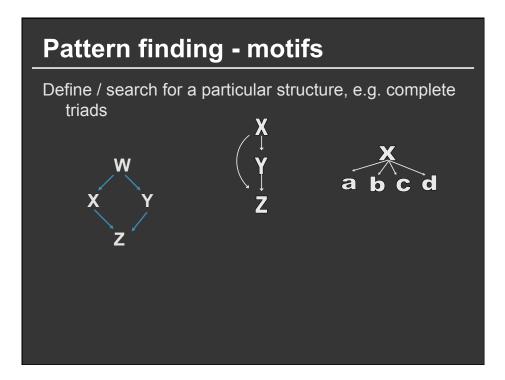
Betweenness clustering

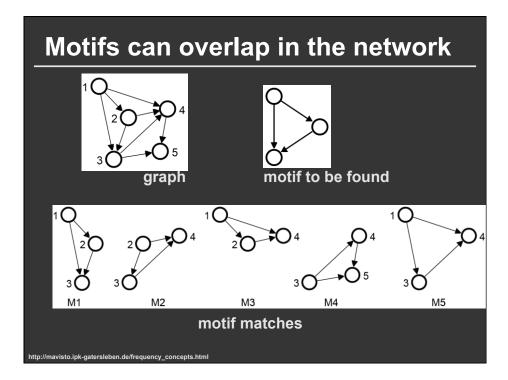
Girvan and Newman 2002 iterative algorithm:

- Compute *C_b* of all *edges*
- Remove edge *i* where $C_b(i) == max(C_b)$
- Recalculate betweenness

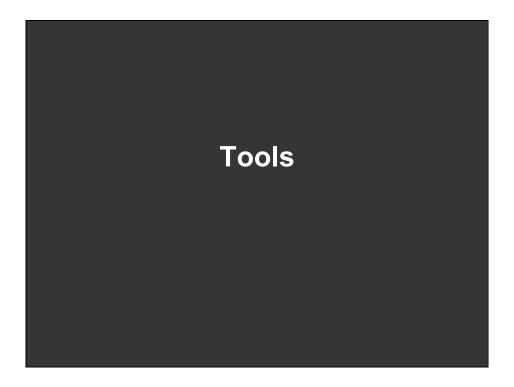








4 node subgraphs Ň \mathbb{Z} **** K, 000 A



Network EDA

Structure

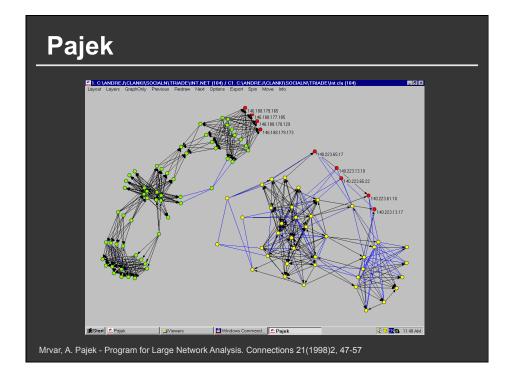
- Centralization
- Density
- Clustering, components
- Motifs
- Comparison to models

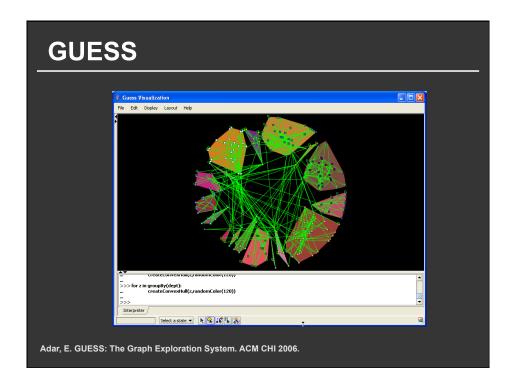
Attributes

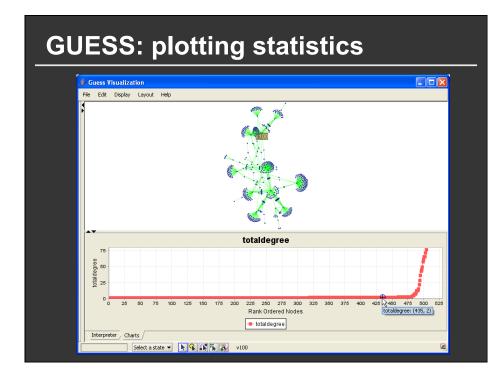
• Nodes / links / communities

Useful features:

- Associate attributes
- Node/graph level centrality
- Filter on statistics
- Examine distributions
- Identify components, clusters
- Define and search for patterns
- Create random graphs, calculate statistics
- Map statistics to visual features (color, size, weight)
- Track nodes and groups of interest
- Zoom and pan in large graphs







SocialAction

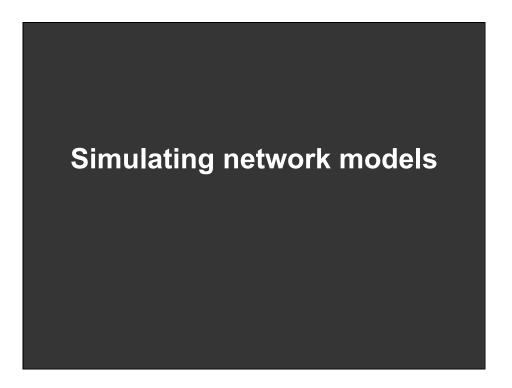
Challenge:

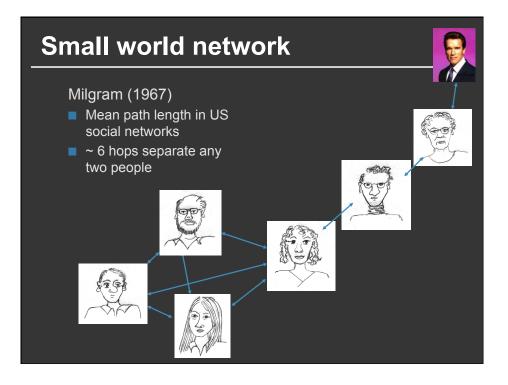
User directedness + number of statistical features leads to opportunistic analysis in most tools

Solution:

- Provide overview
- Use attribute ranking and coordinated views
- Aggregate networks, identify communities
- View bi-, tripartite (etc.) networks separately
- Access to matrix overview
- Keep nodes in place

Perer, A. and Shneiderman, B. Balancing systematic and flexible exploration of social networks. InfoVis 2006.

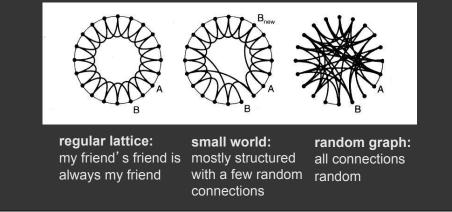




Small world networks

Watts and Strogatz 1998

a few random links in an otherwise structured graph make the network a small world



Defining small world phenomenon

 $C_{\text{network}} >> C_{\text{random graph}}$

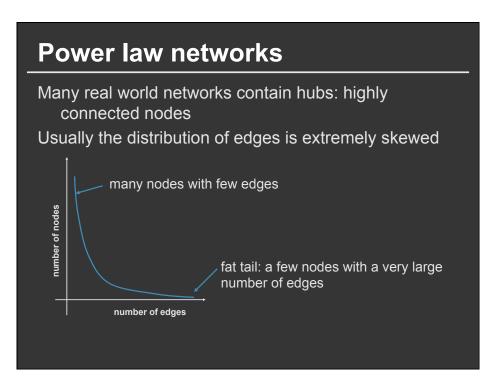
 $l_{\text{network}} \approx \ln(N)$

Pattern:

- high clustering
- Iow mean shortest path

Examples

- neural network of C. elegans,
- semantic networks of languages,
- actor collaboration graph
- food webs



Summary

Structural analysis

- Centrality
- Community structure
- Pattern finding
- → Widely applicable across domains

Tools for network EDA

- Calculate, filter on statistics
- View graph plus matrix, histograms, etc.
- Overview plus details on demand
- Highlight user-defined nodes of interest, consistent positions