Network data Visualization Analysis & Design Network Data (Ch 9)

Tamara Munzner

Department of Computer Science University of British Columbia @tamaramunzner

- edge crossings, node overlaps
- distances between topological neighbor nodes

- -angular distance between different edges
- emphasize symmetry
- similar graph structures should look similar in layout

networks Dataset Types -model relationships → Tables between things · aka graphs -two kinds of items, both can have attributes nodes links • tree -special case - no cycles • one parent per node



-compare centrality/importance measures - identify clusters / communities attribute based tasks (similar to table data) -find distributions, .. combination tasks, incorporating both - example: find friends-of-friends who like cats

• topology: find all adjacent nodes of given node

• attributes: check if has-pet (node attribute) == cat

topology based tasks

-find (topological) neighbors

-find paths

Node-link diagrams nodes: point marks

- links: line marks
- straight lines or arcs
- -connections between nodes
- intuitive & familiar
- -most common
- many, many variants



Force-directed placement





Criteria for good node-link layouts

- -total drawing area
- edge bends
- maximize
- -aspect ratio disparities

• most criteria NP-hard individually

Criteria conflict

- many criteria directly conflict with each other



Uniform edge

Optimization-based layouts • formulate layout problem as optimization problem

- convert criteria into weighted cost function
- -F(layout) = a*[crossing counts] + b*[drawing space used]+...

Network tasks: topology-based and attribute-based

- use known optimization techniques to find layout at minimal cost
- energy-based physics models
- -force-directed placement -spring embedders

• physics model -links = springs pull together -nodes = magnets repulse apart algorithm -place vertices in random locations -while not equilibrium · calculate force on vertex - sum of » pairwise repulsion of all nodes » attraction between connected nodes • move vertex by c * vertex_force http://mbostock.github.com/d3/ex/force.html

Force-directed placement properties strengths

- reasonable layout for small, sparse graphs - clusters typically visible
- edge length uniformity
- weaknesses
- nondeterministic
- -computationally expensive: O(n^3) for n nodes • each step is n^2, takes ~n cycles to reach equilibrium
- naive FD doesn't scale well beyond IK nodes - iterative progress: engaging but distracting

· visual encoding -link connection marks, node point marks considerations d3-force testing ground

-proximity semantics?

-spatial position: no meaning directly encoded left free to minimize crossings

Idiom: force-directed placement

- · sometimes meaningful · sometimes arbitrary, artifact of layout algorithm
- · tension with length -long edges more visually salient than short
- scalability

- explore topology; locate paths, clusters

-node/edge density E < 4N

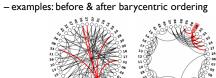
Node order is crucial: Reordering



• restricted node-link layouts: lay out nodes around circle or along line • data

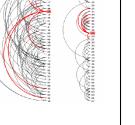
Idiom: circular layouts / arc diagrams (node-link)

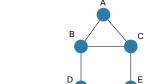
- original: network - derived: node ordering attribute (global computation)
- · considerations: node ordering crucial to avoid excessive clutter from edge crossings



http://brofs.etsmtl.ca/mmcguffin/research/2012-mcguffin-simpleNetVis/mcguffin-2012-simpleNetVis.pdf



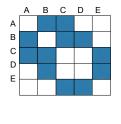




Adjacency matrix representations

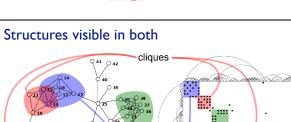
· derive adjacency matrix from network

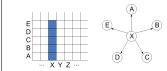










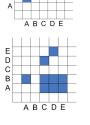


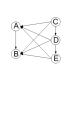
ABCDE

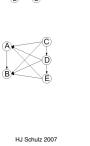
Adjacency matrix examples

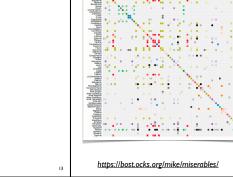


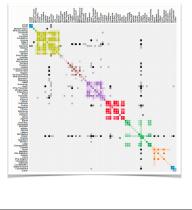






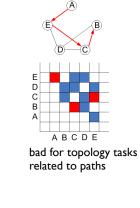


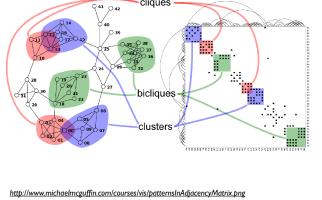


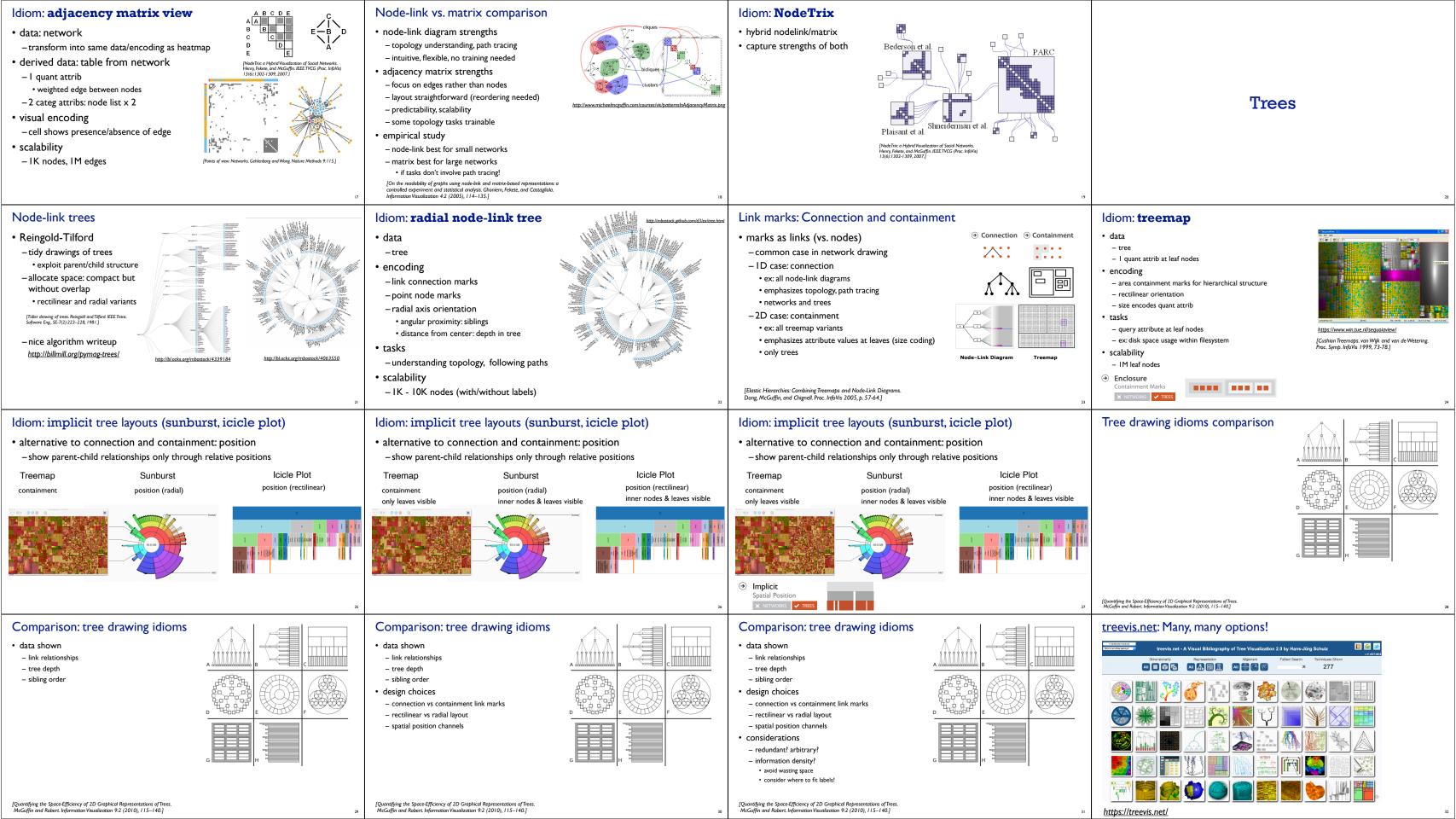


A B C D E F G H good for topology tasks related to neighborhoods (node 1-hop neighbors)

Adjacency matrix







Arrange networks and trees







→ Adjacency Matrix



considerations

technique

scalability

• data: compound graph - original: network

- derived: cluster hierarchy atop it

• same: fundamental use of space

not shown explicitly

-nodes, edges: IK-10K -hairball problem eventually hits

-better algorithm for same encoding

• hierarchy used for algorithm speed/quality but



Idiom: **sfdp** (multi-level force-directed placement)

Visualization Analysis & Design

Network Data (Ch 9) II

Tamara Munzner

Department of Computer Science University of British Columbia

[Schulz 2004]

Hierarchical edge bundling

Multilevel networks

real vertexvirtual vertex

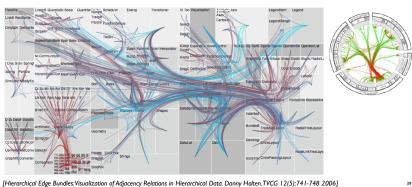
____ internal spring

- virtual spring

• works for any layout: treemap vs radial

derive cluster hierarchy of metanodes

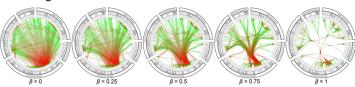
on top of original graph nodes



@tamaramunzner

Idiom: hierarchical edge bundling

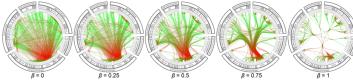
- data
- any layout of compound network
- cluster hierarchy: class package structure
- derived: bundles of edges with same source/destination (multi-level)
- idiom: curve edge routes according to bundles



[Efficient and high quality force-directed graph drawing Hu.The Mathematica Journal 10:37–71, 2005.]

- - network: software classes (nodes), import/export between classes (links)

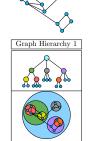
- task: edge clutter reduction



[Hierarchical Edge Bundles: Visualization of Adjacency Relations in Hierarchical Data. Danny Holten.TVCG 12(5):741-748 2006]

Idiom: GrouseFlocks

- · data: compound network
- network
- cluster hierarchy atop it
- derived or interactively chosen
- visual encoding
- -connection marks for network links
- -containment marks for hierarchy
- -point marks for nodes
- dynamic interaction
 - -select individual metanodes in hierarchy to expand/ contract



[GrouseFlocks: Steerable Exploration of Graph Hierarchy Space. Archambault, Munzner, and Auber. IEEE TVCG 14(4):900-913, 2008.]