## Chap 9: Arrange Networks Paper: Topological Fisheye Networks

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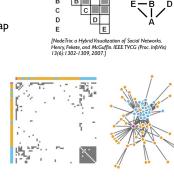
nformation Visualization, CPSC 547 Oct 8 2014

http://www.cs.ubc.ca/~tmm/courses/547-14/#chap9

#### Idiom: adjacency matrix view

data: network

- -transform into same data/encoding as heatmap
- derived data: table from network
- I quant attrib
- · weighted edge between nodes -2 categ attribs: node list x 2
- visual encoding
- cell shows presence/absence of edge
- scalability
- IK nodes, IM edges



### Connection vs. adjacency comparison

adjacency matrix strengths

Arrange networks and trees

Node-link Diagrams

Adjacency Matrix

Enclosure

- predictability, scalability, supports reordering - some topology tasks trainable
- node-link diagram strengths -topology understanding, path tracing
- -intuitive, no training needed
- empirical study
- node-link best for small networks
- matrix best for large networks
- if tasks don't involve topological structure! [On the readability of graphs using node-link and matrix-based representations: a controlled experiment and statistical analysis. Ghoniem, Fekete, and Castagliola. Information Visualization 4:2

Tree drawing idioms comparison

# - node/edge density E < 4N

#### Idiom: radial node-link tree

Idiom: force-directed placement

- link connection marks, node point marks

· left free to minimize crossings

- proximity semantics?

· sometimes meaningfu

· tension with length

- spatial position: no meaning directly encoded

• sometimes arbitrary, artifact of layout algorithm

- long edges more visually salient than short

- explore topology; locate paths, clusters

- data
- -tree encoding

scalability

visual encoding

considerations

- -link connection marks
- -point node marks
- radial axis orientation
- angular proximity: siblings · distance from center: depth in tree
- tasks
- -understanding topology, following paths
- · scalability
- IK IOK nodes

http://mbostock.github.com/d3/ex/tree.htm

http://mbostock.github.com/d3/ex/force.html

#### Idiom: treemap

- nodes, edges: IK-10K

- original: network

considerations

- derived: cluster hierarchy atop it

· same: fundamental use of space

- hairball problem eventually hits

-better algorithm for same encoding technique

· hierarchy used for algorithm speed/quality but

• (more on algorithm vs encoding in afternoon)

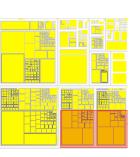
data

scalability

- I quant attrib at leaf nodes

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

- encoding -area containment marks for hierarchical structure
- rectilinear orientation
- size encodes quant attrib • tasks
- -query attribute at leaf nodes
- scalability
- IM leaf nodes



[Efficient and high quality force-directed graph drawing

Hu.The Mathematica Journal 10:37-71, 2005.]

### Link marks: Connection and Containment

- marks as links (vs. nodes) -common case in network drawing
- ID case: connection
- ex: all node-link diagrams · emphasizes topology, path tracing
- networks and trees

Topological Fisheye Views

-user changed selected focus point

- -2D case: containment • ex: all treemap variants
- emphasizes attribute values at leaves (size coding)

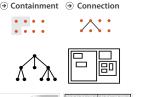
-input: laid-out network (spatial positions for nodes)

- output: multilevel hierarchy from graph coarsening

-hybrid view made from cut through several hierarchy levels

only trees

visual encoding



[Elastic Hierarchies: Combining Treemaps and Node-Link

[Fig 4.7.Tobological Fisheve Views for

and North, IEEE TVCG 11(4), p 457-468,

Diagrams. Dong, McGuffin, and Chignell. Proc. InfoVis

### data shown

- link relationships - tree depth
- sibling order

(2005), 114-135,1

- design choices - connection vs containment link marks
- rectilinear vs radial layout
- spatial position channels
- considerations - redundant? arbitrary?
- information density?
- avoid wasting space
- [Quantifying the Space-Efficiency of 2D Graphical Representations of Trees, McGuffin and Robert, Information Visualization 9:2 (2010), 115-140.]

#### Idiom: GrouseFlocks data: compound graphs

- network
- -cluster hierarchy atop it
- · derived or interactively chosen
- visual encoding -connection marks for network links
- -containment marks for hierarchy
- point marks for nodes

Coarsening requirements

uniform cluster/metanode size

- dynamic interaction
- -select individual metanodes in hierarchy to expand/ contract

match coarse and fine layout geometries





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

## Further reading

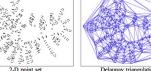
- Chap 9: Arrange Networks and Trees
- Simple Algorithms for Network Visualization: A Tutorial. McGuffin. Tsinghua Science and
- Technology (Special Issue on Visualization and Computer Graphics) 17:4 (2012), 383-
- LNCS Tutorial, 2025, edited by M. Kaufmann and D. Wagner, LNCS Tutorial, 2025, pp. 71– 86. Springer-Verlag, 2001
- Perceptual Guidelines for Creating Rectangular Treemaps. Kong, Heer, and Agrawala.

- Visualization Analysis and Design. Munzner. AK Peters / CRC Press, Oct 2014. • Visual Analysis of Large Graphs: State-of-the-Art and Future Research Challenges. von
- Landesberger et al. Computer Graphics Forum 30:6 (2011), 1719–1749.
- Drawing on Physical Analogies. Brandes. In Drawing Graphs: Methods and Models,
- Treevis.net: A Tree Visualization Reference. Schulz. IEEE Computer Graphics and Applications 31:6 (2011), 11–15. http://www.treevis.net
- IEEE Trans. Visualization and Computer Graphics (Proc. InfoVis) 16:6 (2010), 990–998.

### Coarsening strategy

- must preserve graph-theoretic properties
- use both topology and geometry -topological distance (hops away)
- -geometric distance but not just proximity alone! • just contracting nodes/edges could create new cycles
- · derived data: proximity graph









## Topological Fisheye Views

- -input: laid-out network (spatial positions for nodes) - output: multilevel hierarchy from graph coarsening
- -user changed selected focus point visual encoding
- hybrid view made from cut through several hierarchy levels







[Fig 4,8.Topological Fisheye Views for and North IEEE TVCG 11(4) b 457-468

[Fig 3.Topological Fisheye Views for Visualizing Large Graphs. Gansner, Koren and North, IEEE TVCG 11(4), p 457-468, 2005]

scalable

what **not** to do!

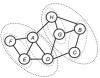
#### Candidate pairs: neighbors in original and proximity graph

- proximity graph: compromise between larger DT and smaller RNG
- better than original graph neighbors alone
- slow for cases like star graph
- maximize weighted sum of
- geometric proximity
- goal: preserve geometry
- cluster size
- goal: keep uniform cluster size
- gourneep annerm claster size
- normalized connection strengthgoal: preserve topology
- neighborhood similarity
- goal: preserve topology
- degree
- goal: penalize high-degree nodes to avoid salient artifacts and computational problems

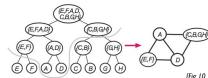
#### Hybrid graph creation

cut through coarsening hierarchy to get active nodes

 animated transitions between states



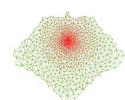


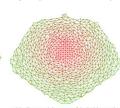


[Fig 10, 12.Topological Fisheye Views for Visualizing Large Graphs. Gansner, Koren and North, IEEE TVCG 11(4), p 457-468, 2005]

#### Final distortion

- geometric distortion for uniform density
- (colorcoded by hierarchy depth just to illustrate algorithm)
- compare to original
- -compare to simple topologically unaware fisheye distortion







(b) default layout of hybrid graph (c) distorted layout of hybrid graph

[Fig 2,15. Topological Fisheye Views for Visualizing Large Graphs. Gansner, Koren and North, IEEE TVCG 11 (4), p. 457-468, 2005]

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