

# Chap 9:Arrange Networks

## Paper: Topological Fisheye Networks

**Tamara Munzner**

Department of Computer Science  
University of British Columbia

*Information Visualization, CPSC 547*  
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<http://www.cs.ubc.ca/~tmm/courses/547-14/#chap9>

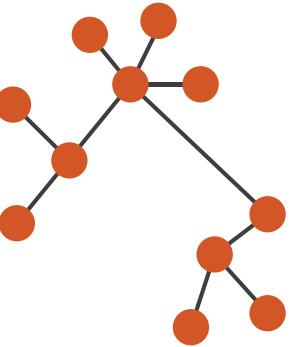
# Arrange networks and trees

## → Node-link Diagrams

Connections and Marks

NETWORKS

TREES



## → Adjacency Matrix

Derived Table

NETWORKS

TREES

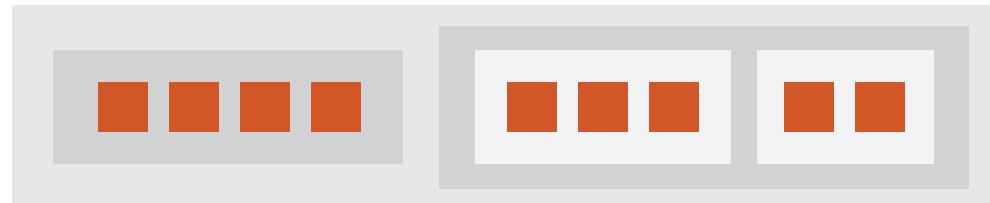
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## → Enclosure

Containment Marks

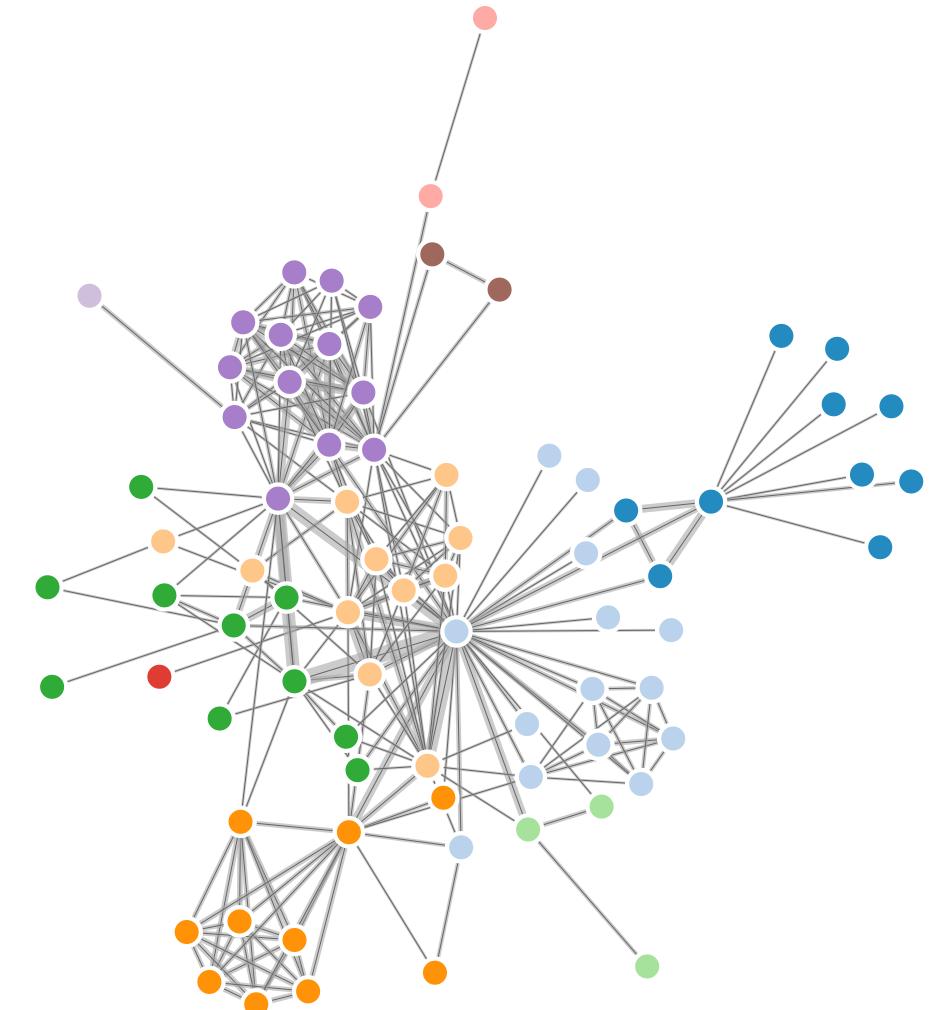
NETWORKS

TREES



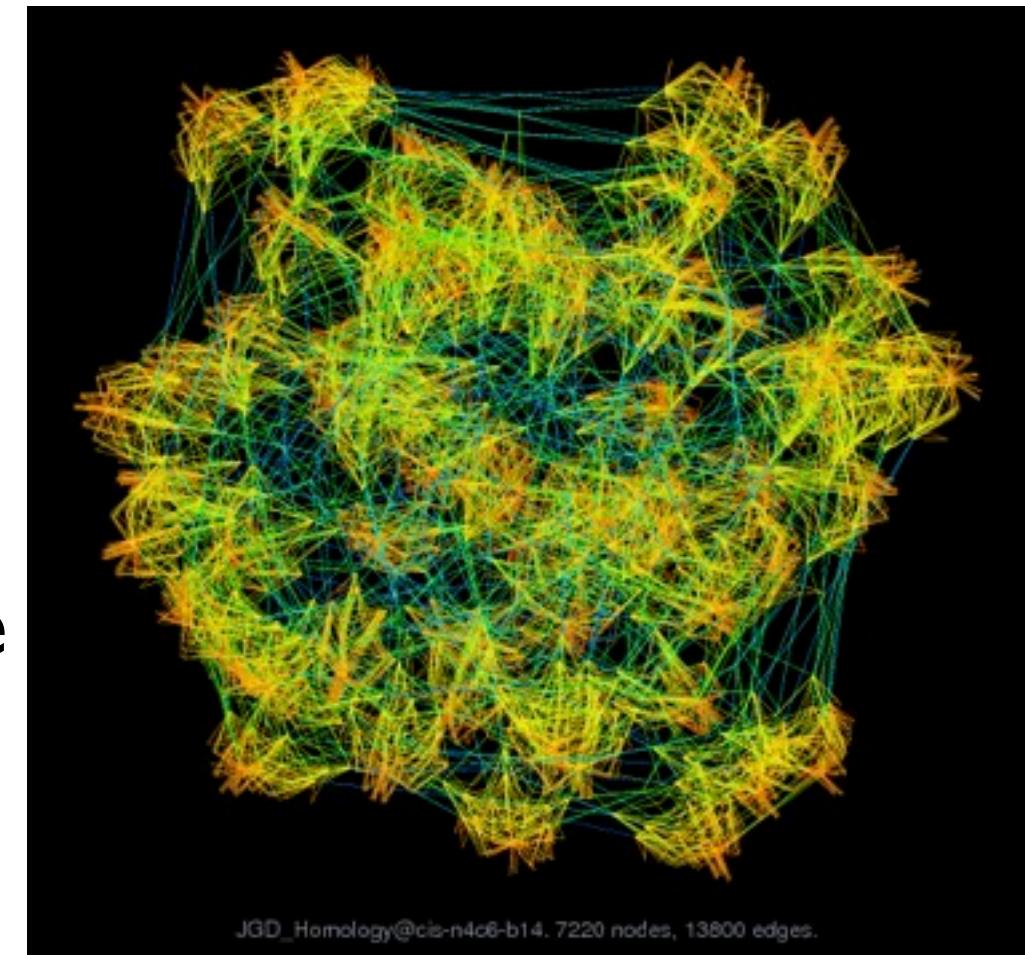
# Idiom: force-directed placement

- visual encoding
  - link connection marks, node point marks
- considerations
  - spatial position: no meaning directly encoded
    - left free to minimize crossings
  - proximity semantics?
    - sometimes meaningful
    - sometimes arbitrary, artifact of layout algorithm
    - tension with length
      - long edges more visually salient than short
- tasks
  - explore topology; locate paths, clusters
- scalability
  - node/edge density  $E < 4N$

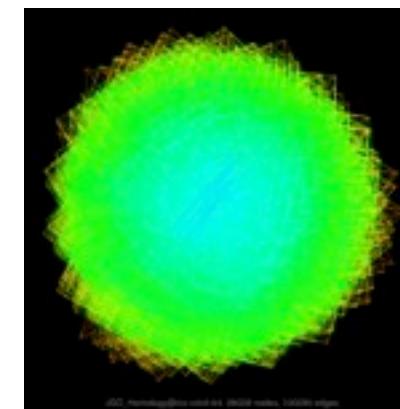


# Idiom: **sfdp** (multi-level force-directed placement)

- data
  - original: network
  - derived: cluster hierarchy atop it
- considerations
  - better algorithm for same encoding technique
    - same: fundamental use of space
    - hierarchy used for algorithm speed/quality but not shown explicitly
    - (more on algorithm vs encoding in afternoon)
- scalability
  - nodes, edges: 1K-10K
  - hairball problem eventually hits



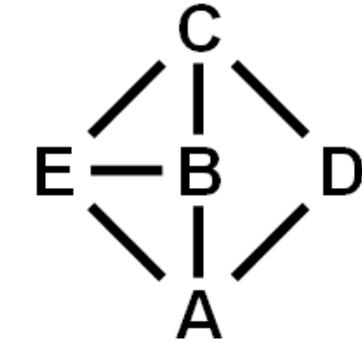
[Efficient and high quality force-directed graph drawing.  
Hu. *The Mathematica Journal* 10:37–71, 2005.]



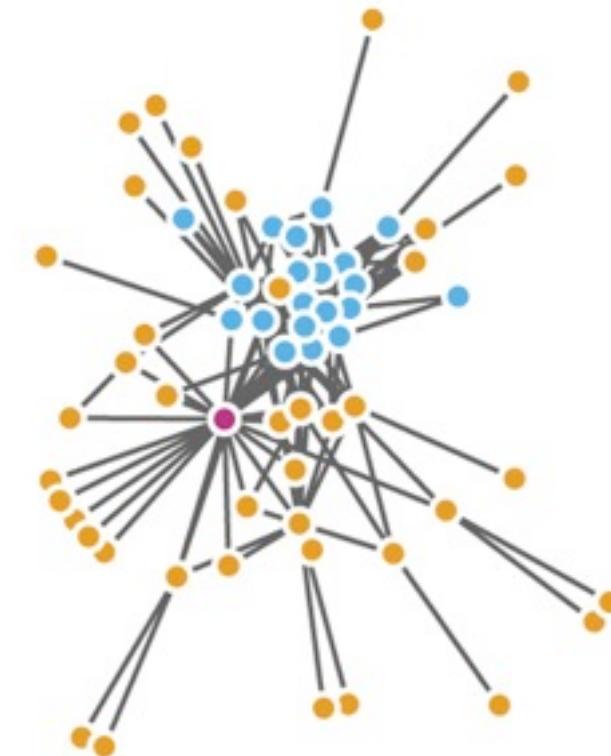
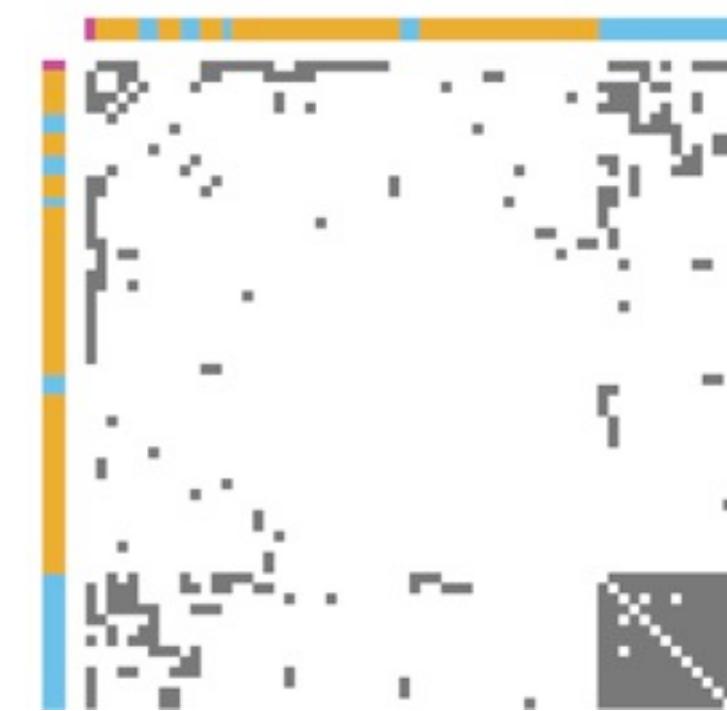
# Idiom: adjacency matrix view

- data: network
  - transform into same data/encoding as heatmap
- derived data: table from network
  - 1 quant attrib
    - weighted edge between nodes
  - 2 categ attribs: node list x 2
- visual encoding
  - cell shows presence/absence of edge
- scalability
  - 1K nodes, 1M edges

	A	B	C	D	E
A	A				
B		B			
C			C		
D				D	
E					E



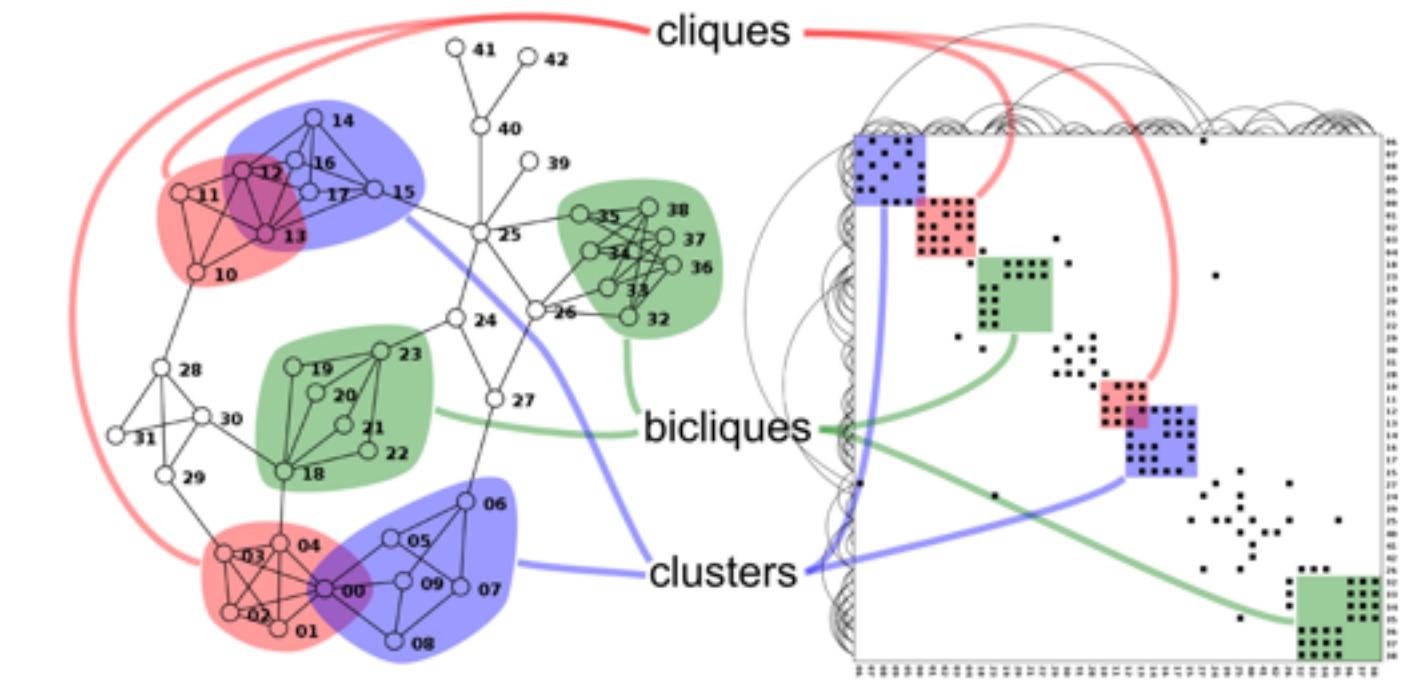
[NodeTrix: a Hybrid Visualization of Social Networks.  
Henry, Fekete, and McGuffin. IEEE TVCG (Proc. InfoVis)  
13(6):1302-1309, 2007.]



[Points of view: Networks. Gehlenborg and Wong. Nature Methods 9:115.]

# Connection vs. adjacency comparison

- adjacency matrix strengths
  - 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!

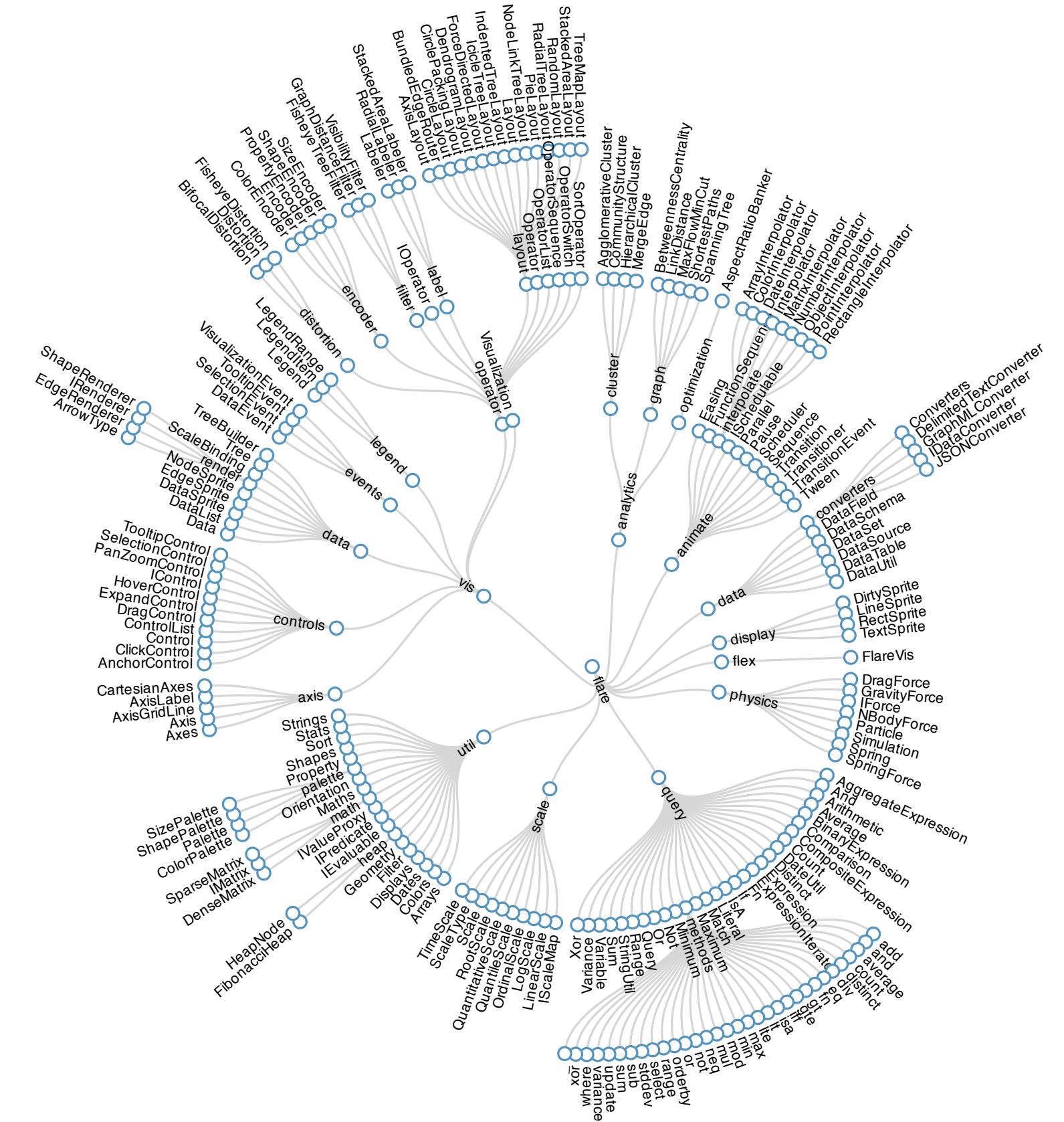


<http://www.michaelmcguffin.com/courses/vis/patternsInAdjacencyMatrix.png>

[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 (2005), 114–135.]

# Idiom: radial node-link tree

- data
  - tree
- encoding
  - link connection marks
  - point node marks
  - radial axis orientation
    - angular proximity: siblings
    - distance from center: depth in tree
- tasks
  - understanding topology, following paths
- scalability
  - 1K - 10K nodes



# Idiom: treemap

- data
  - tree
  - 1 quant attrib at leaf nodes
- encoding
  - area containment marks for hierarchical structure
  - rectilinear orientation
  - size encodes quant attrib
- tasks
  - query attribute at leaf nodes
- scalability
  - 1M leaf nodes

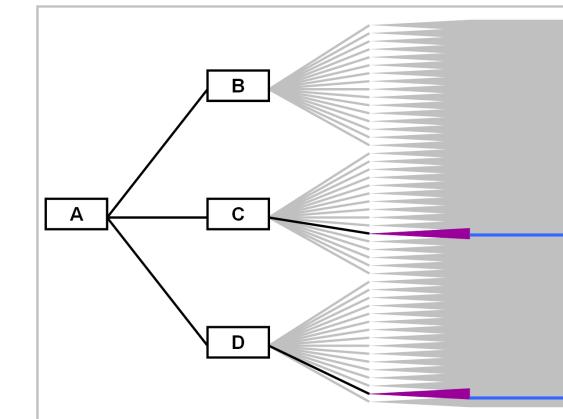
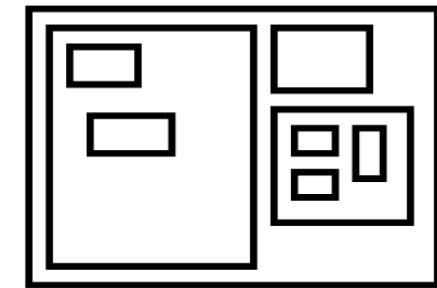
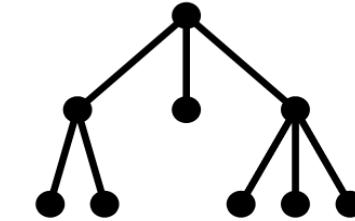
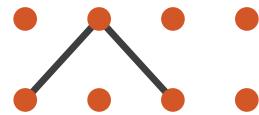
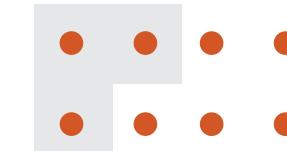


[http://tulip.labri.fr/Documentation/3\\_7/userHandbook/html/ch06.html](http://tulip.labri.fr/Documentation/3_7/userHandbook/html/ch06.html)

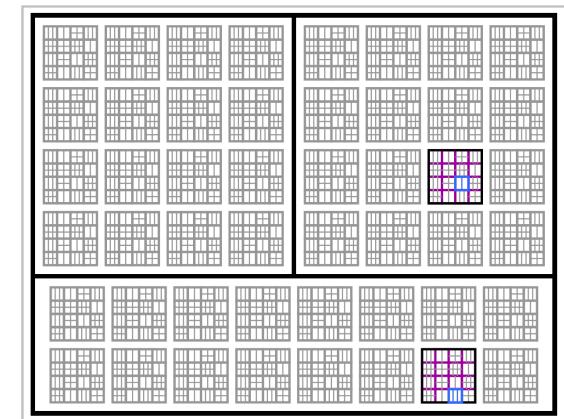
# Link marks: Connection and Containment

- marks as links (vs. nodes)
  - common case in network drawing
  - 1D case: connection
    - ex: all node-link diagrams
    - emphasizes topology, path tracing
    - networks and trees
  - 2D case: containment
    - ex: all treemap variants
    - emphasizes attribute values at leaves (size coding)
    - only trees

→ Containment → Connection



Node-Link Diagram

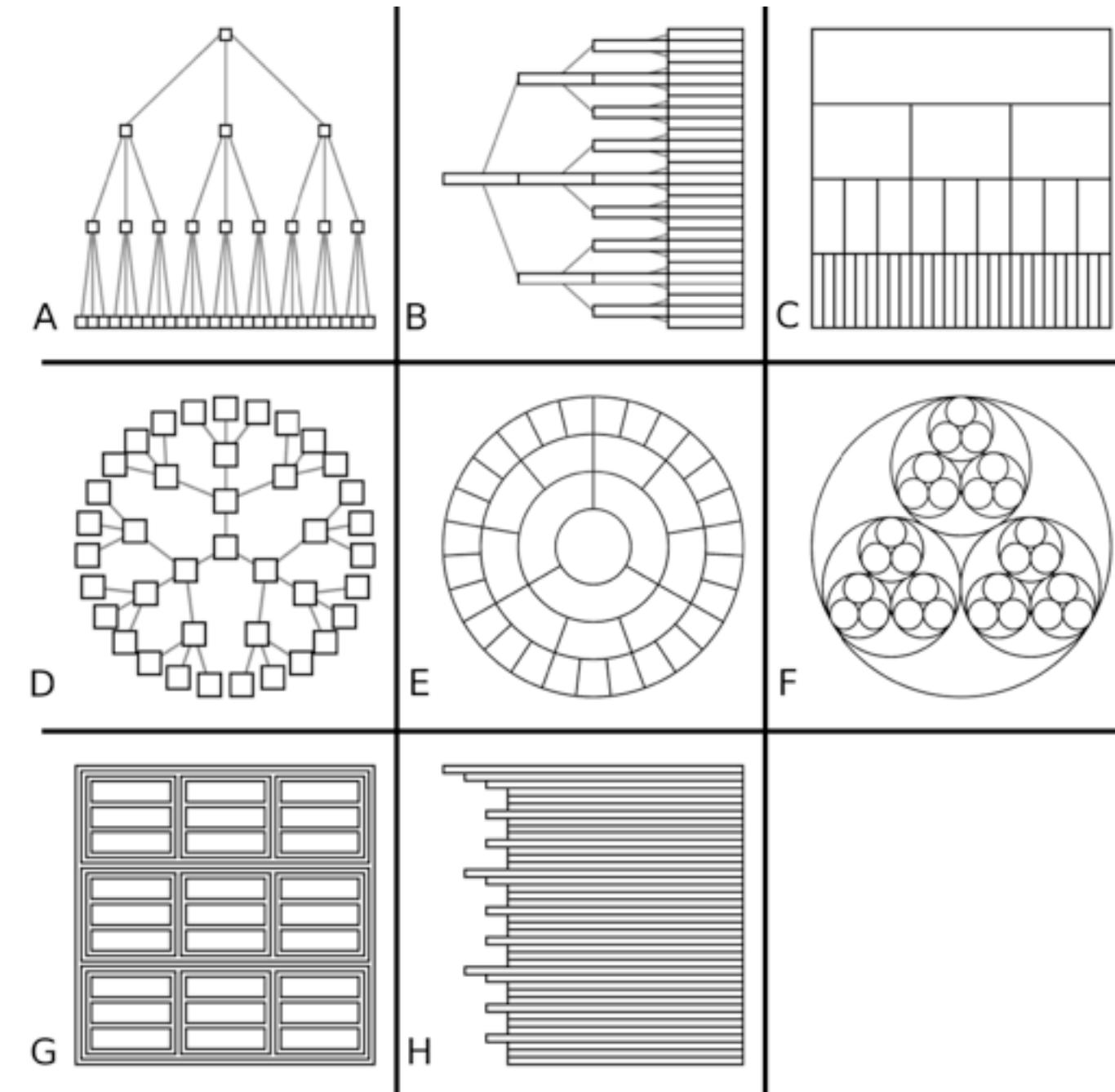


Treemap

[*Elastic Hierarchies: Combining Treemaps and Node-Link Diagrams*. Dong, McGuffin, and Chignell. Proc. InfoVis 2005, p. 57-64.]

# Tree drawing idioms comparison

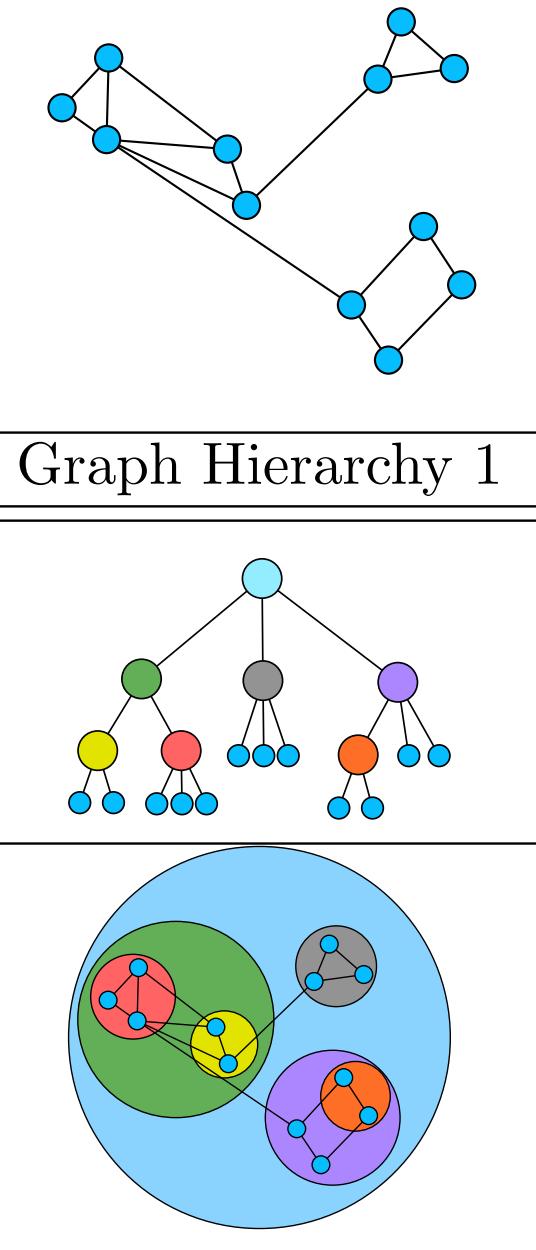
- data shown
  - link relationships
  - tree depth
  - sibling order
- design choices
  - connection vs containment link marks
  - rectilinear vs radial layout
  - spatial position channels
- considerations
  - redundant? arbitrary?
  - information density?
    - avoid wasting space



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



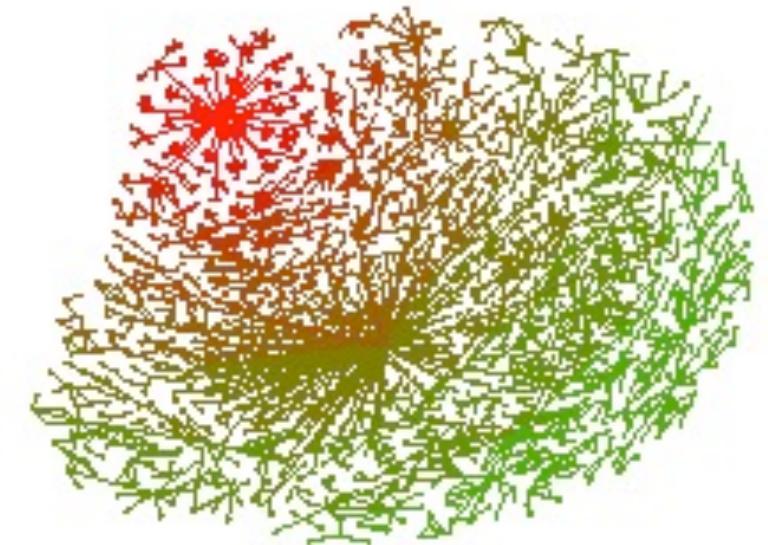
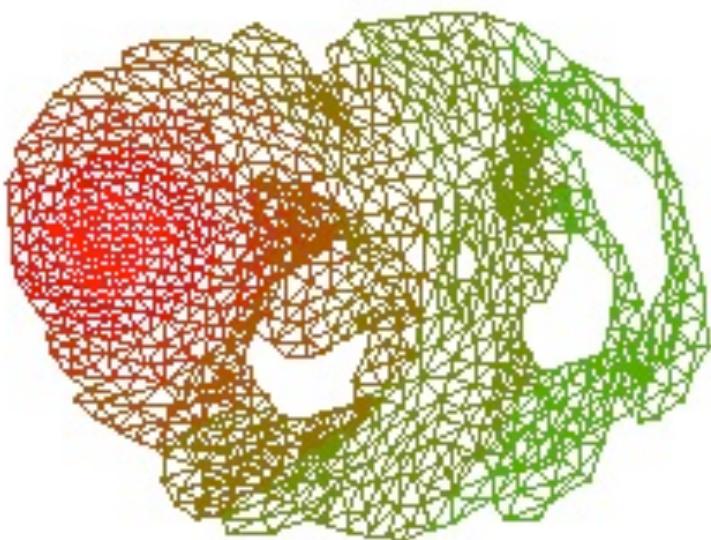
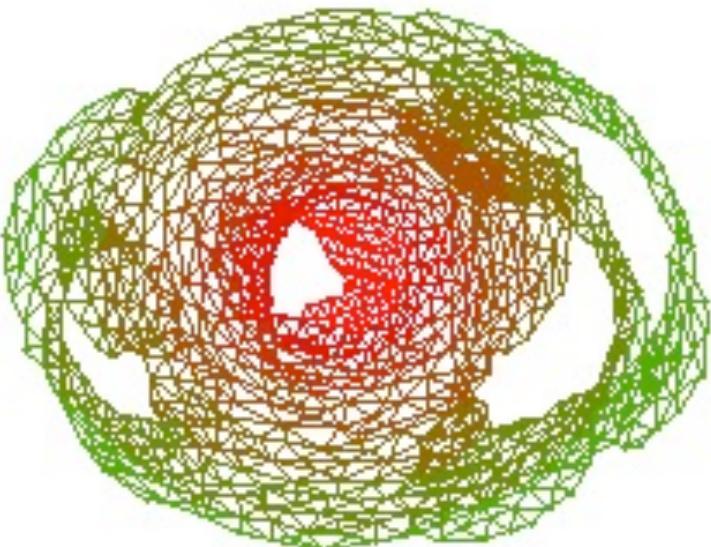
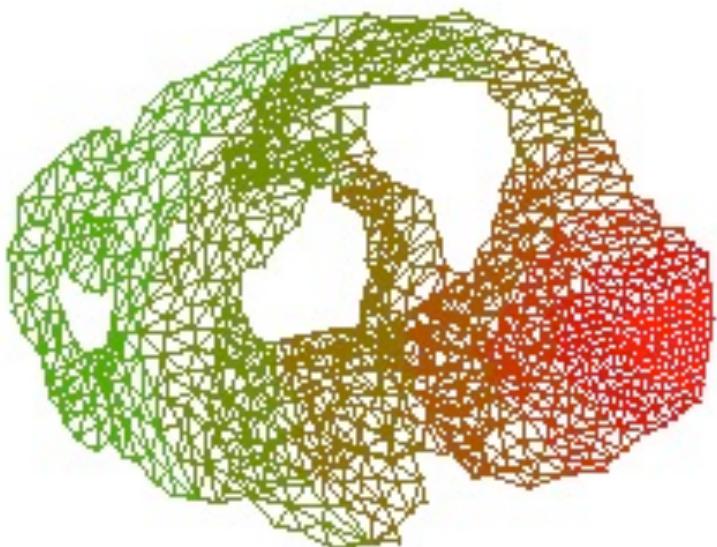
[*GrouseFlocks: Steerable Exploration of Graph Hierarchy Space*. Archambault, Munzner, and Auber. *IEEE TVCG* 14(4): 900-913, 2008.]

# Further reading

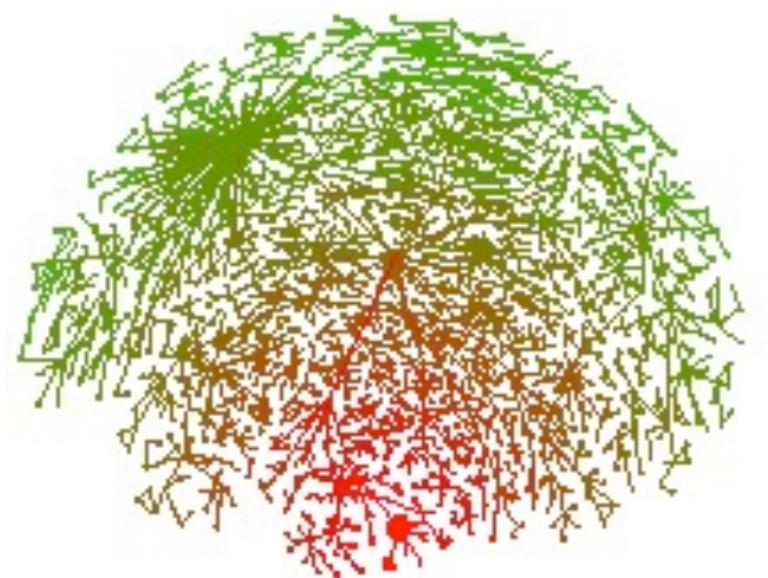
- Visualization Analysis and Design. Munzner. AK Peters / CRC Press, Oct 2014.
  - *Chap 9:Arrange Networks and Trees*
- Visual Analysis of Large Graphs: State-of-the-Art and Future Research Challenges. von Landesberger et al. Computer Graphics Forum 30:6 (2011), 1719–1749.
- Simple Algorithms for Network Visualization: A Tutorial. McGuffin. Tsinghua Science and Technology (Special Issue on Visualization and Computer Graphics) 17:4 (2012), 383–398.
- Drawing on Physical Analogies. Brandes. In Drawing Graphs: Methods and Models, LNCS Tutorial, 2025, edited by M. Kaufmann and D. Wagner, LNCS Tutorial, 2025, pp. 71–86. Springer-Verlag, 2001.
- Treevis.net: A Tree Visualization Reference. Schulz. IEEE Computer Graphics and Applications 31:6 (2011), 11–15. <http://www.treevis.net>
- Perceptual Guidelines for Creating Rectangular Treemaps. Kong, Heer, and Agrawala. IEEE Trans. Visualization and Computer Graphics (Proc. InfoVis) 16:6 (2010), 990–998.

# Topological Fisheye Views

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



focus on a top-left portion

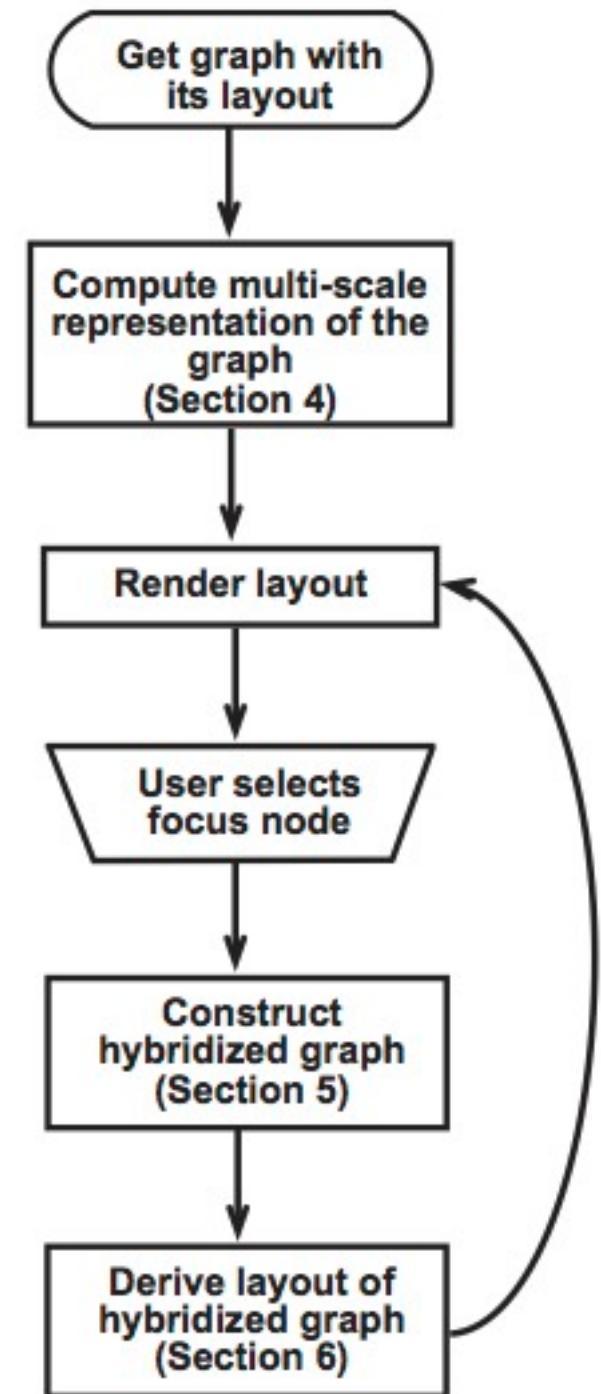
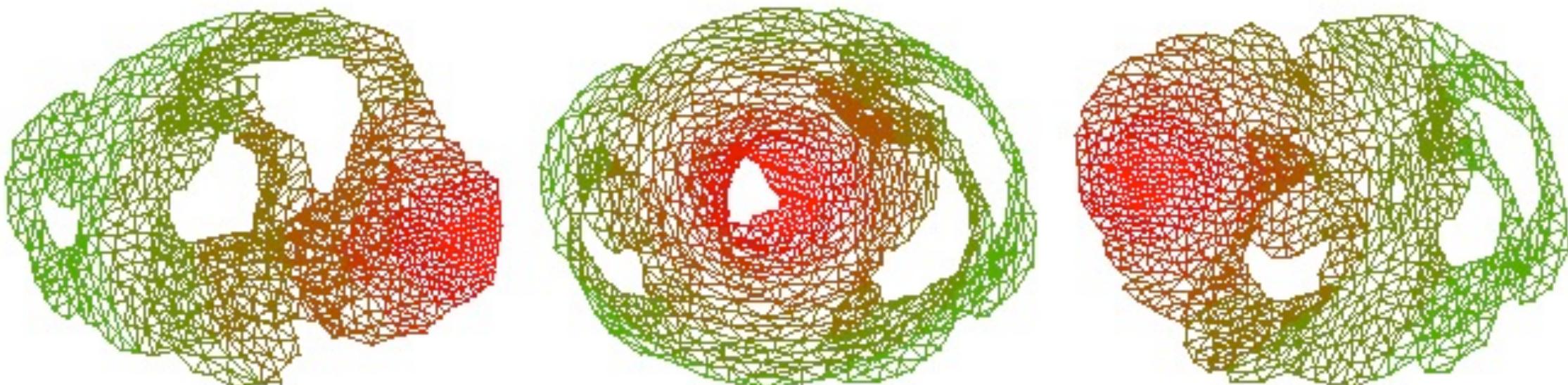


focus on a bottom portion

[Fig 4.7. Topological Fisheye Views for Visualizing Large Graphs. Gansner, Koren and North, IEEE TVCG 11(4), p 457-468, 2005]

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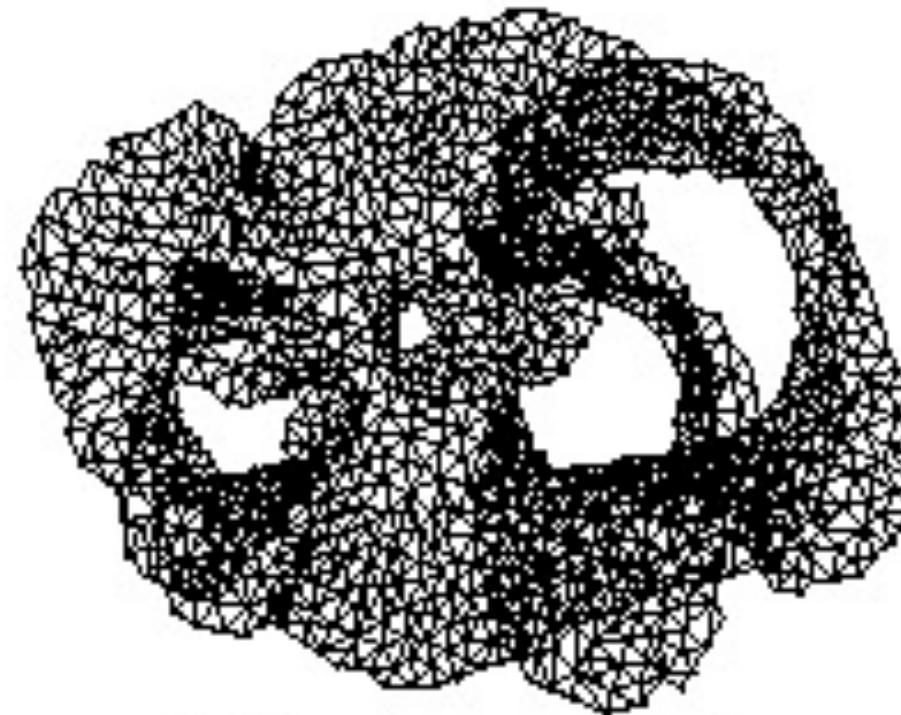
[Fig 4.8. Topological Fisheye Views for Visualizing Large Graphs. Gansner, Koren and North, IEEE TVCG 11(4), p 457-468, 2005]

# Coarsening requirements

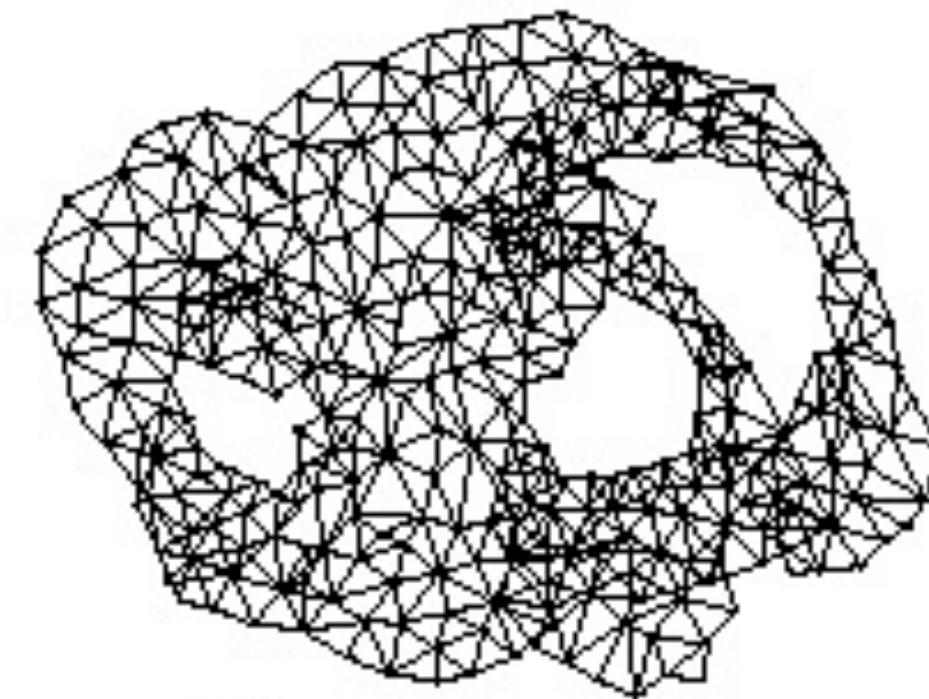
- uniform cluster/metanode size
- match coarse and fine layout geometries
- scalable



4394-node approximation



1223-node approximation

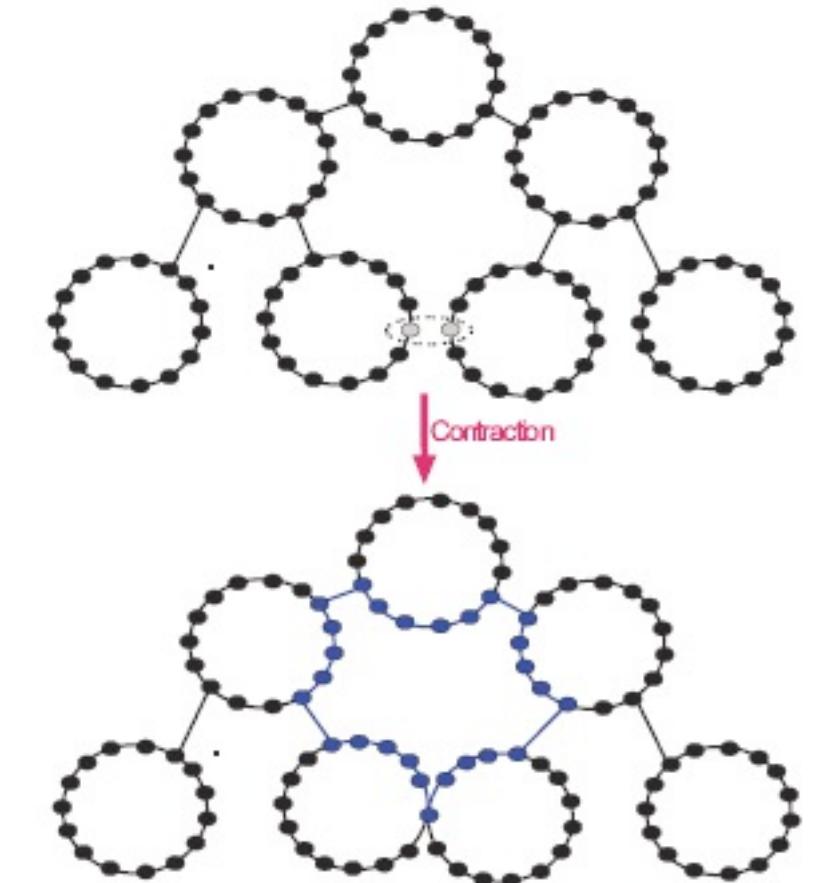
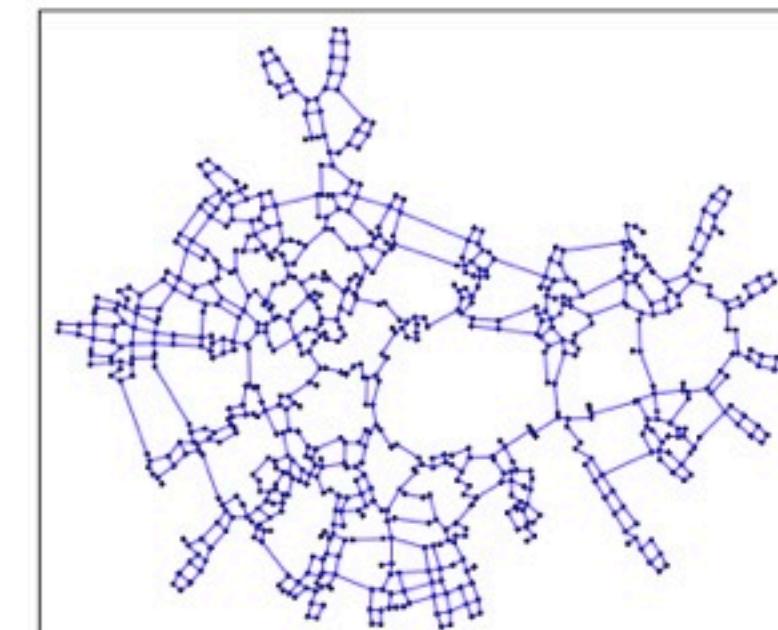
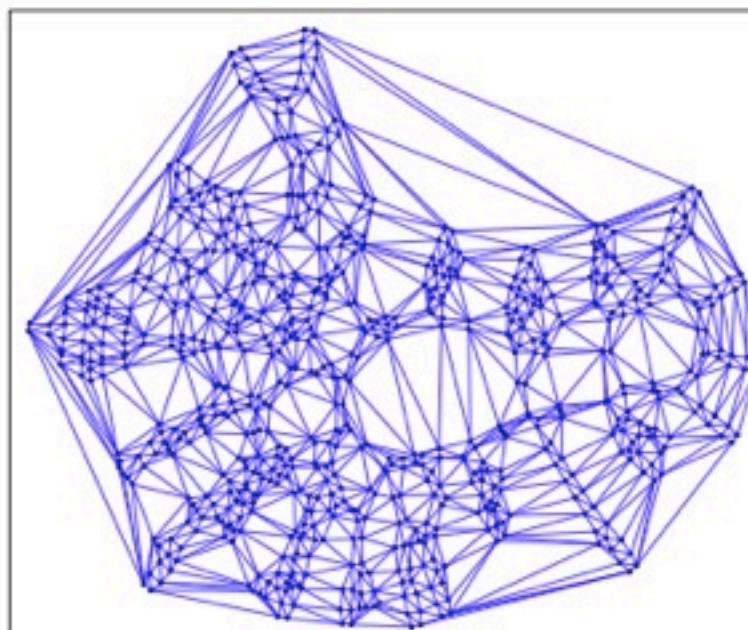


341-node approximation

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

# 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



what **not** to do!

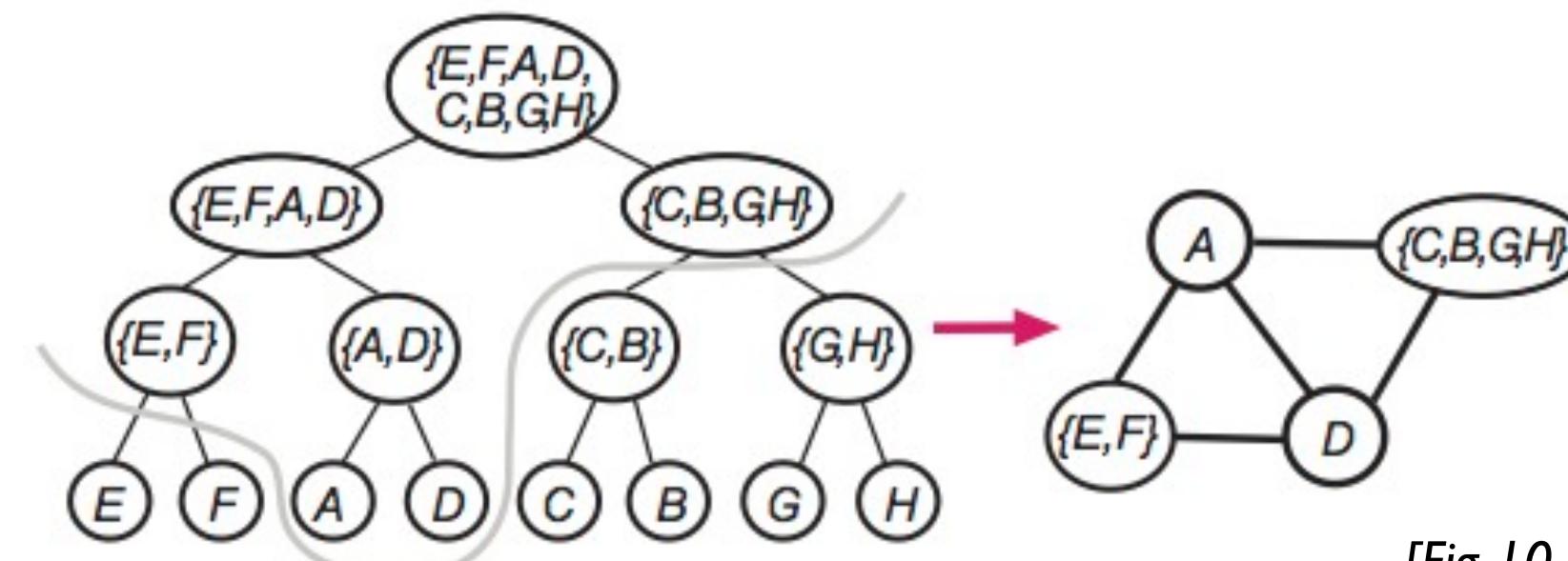
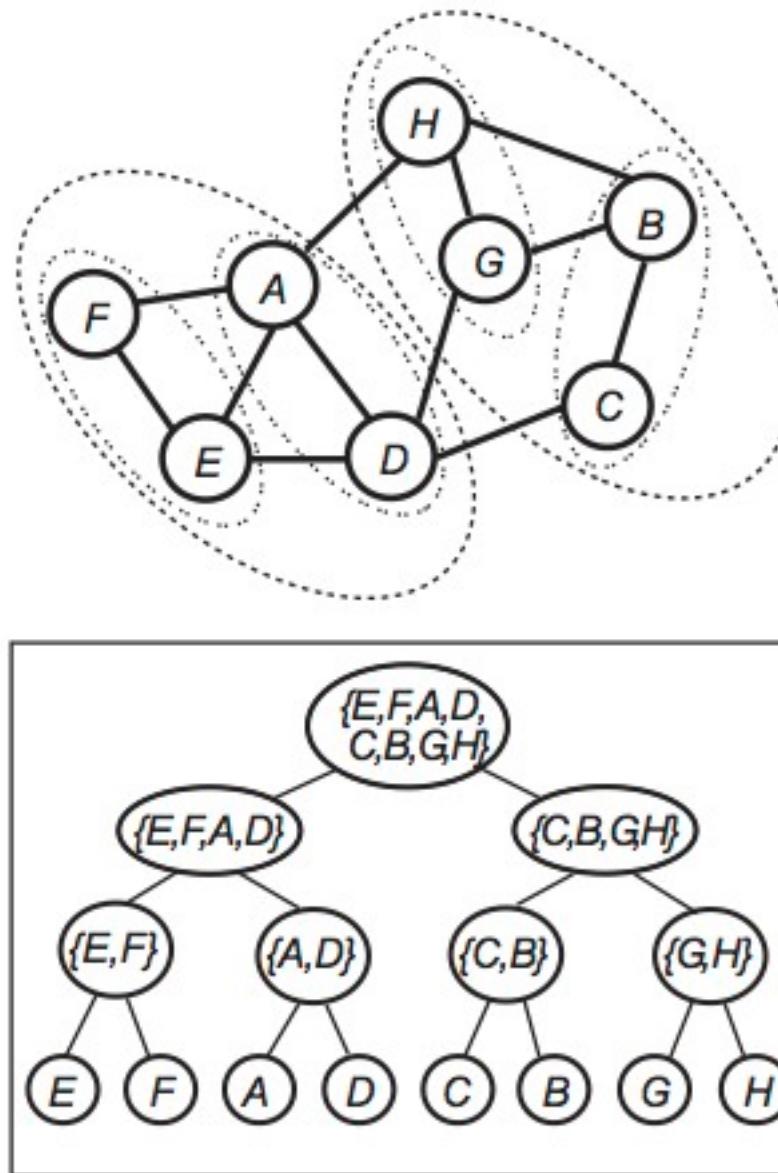
[Fig 10, 12. Topological Fisheye Views for Visualizing Large Graphs. Gansner, Koren and North, IEEE TVCG 11(4), p 457-468, 2005]

# 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
  - normalized connection strength
    - goal: preserve topology
  - neighborhood similarity
    - goal: preserve topology
  - degree
    - goal: penalize high-degree nodes to avoid salient artifacts and computational problems

# Hybrid graph creation

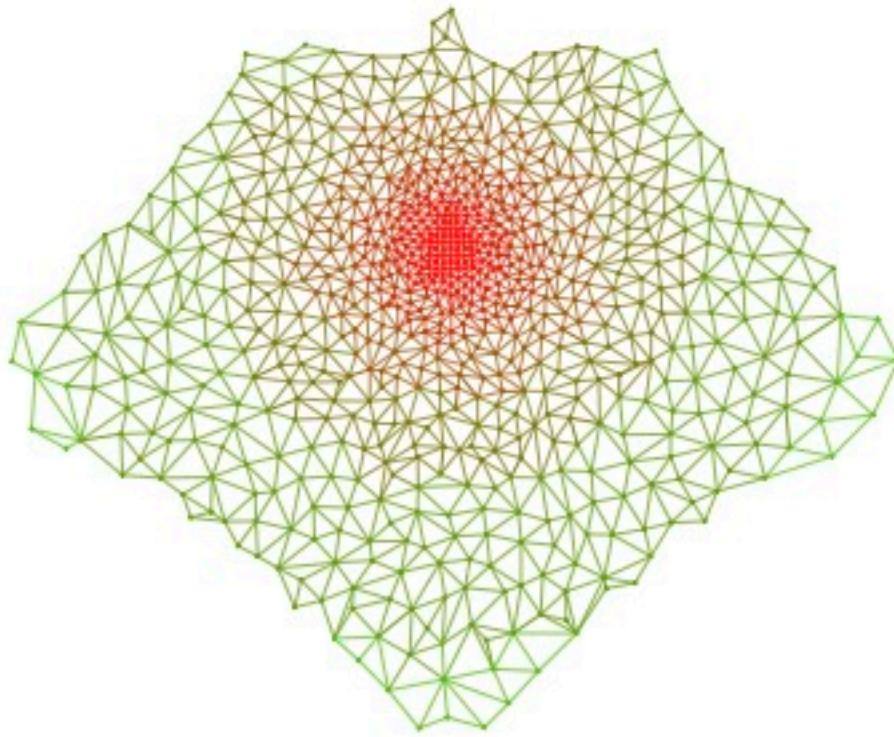
- cut through coarsening hierarchy to get active nodes
  - animated transitions between states



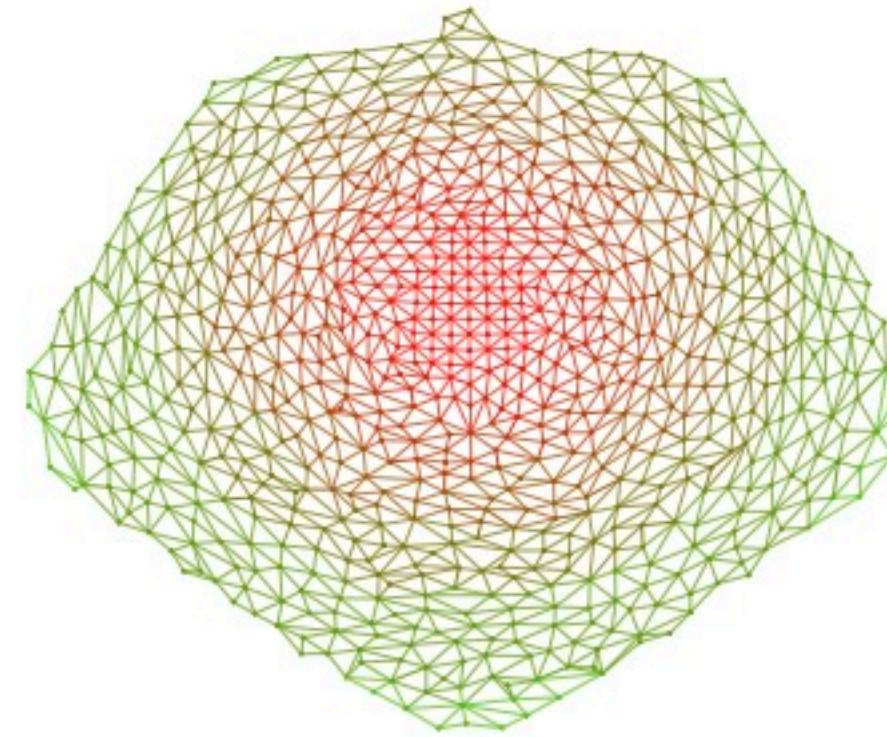
[Fig 10, 12. Topological Fisheye Views for Visualizing Large Graphs. Gansner, Koren and North, IEEE TVCG 11(4), p 457-468, 2005]

# Final distortion

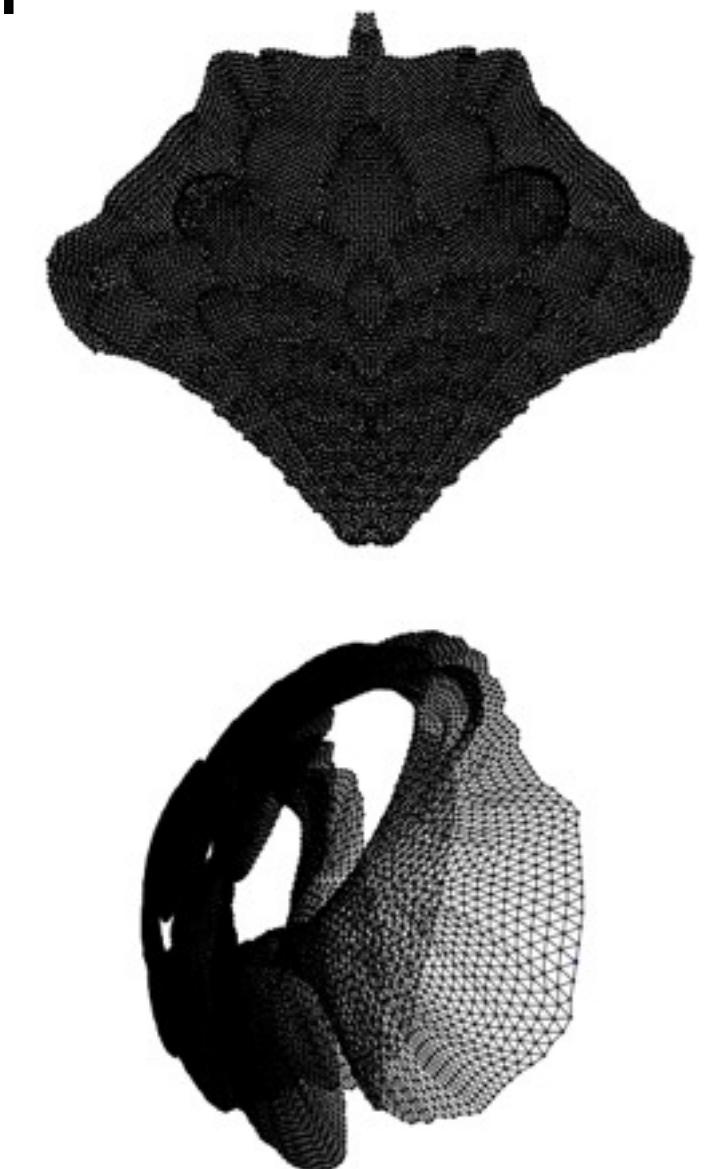
- geometric distortion for uniform density
- (colorcoded by hierarchy depth just to illustrate algorithm)
  - compare to original
  - compare to simple topologically unaware fisheye distortion



**(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]