

Information Visualization Networks & Trees 1/2

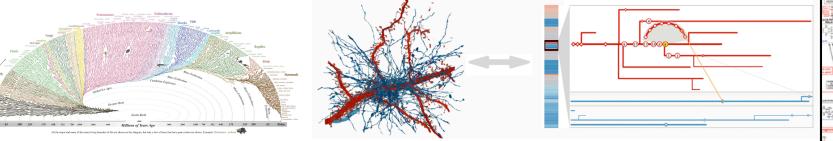
Tamara Munzner
Department of Computer Science
University of British Columbia

Lect 14/15, Feb 27 & Mar 3 2020

<https://www.cs.ubc.ca/~tmm/courses/436V-20>

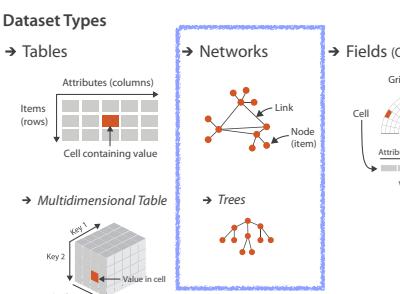
Applications of networks: biological networks

- interactions between genes, proteins, and chemical products
- the brain: connections between neurons
- your ancestry: the relations between you and your family
- phylogeny: the evolutionary relationships of life



Network data

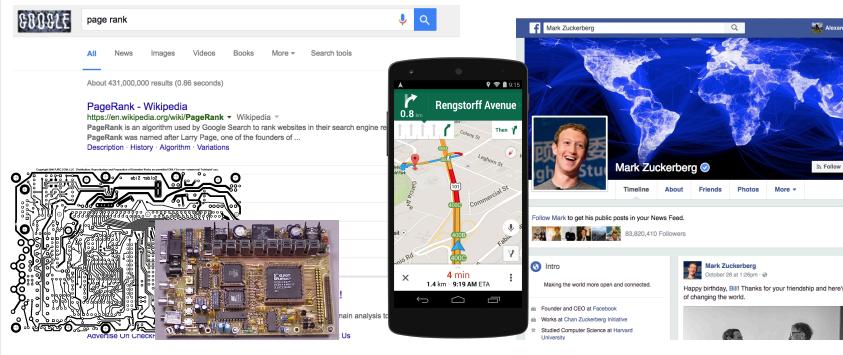
- networks
 - model relationships between things
 - aka graphs
 - two kinds of items, both can have attributes
 - nodes
 - links
- tree
 - special case
 - no cycles
 - one parent per node



Networks

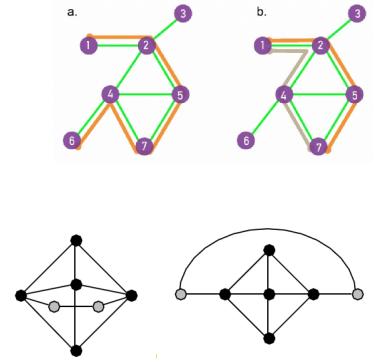
Applications of networks

- without networks, couldn't have any of these:



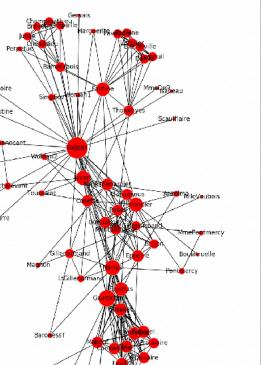
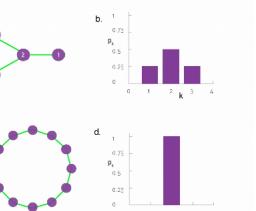
Network tasks: topology-based

- topological structure of network
 - path following
 - path is route along links
 - hops from one node to another
 - path length is number of links along route
 - shortest path connects nodes i & j with smallest # of hops
- topology vs geometry
 - topological hops different from geometric distance given specific layout
 - topology does not depend on layout
 - geometry does



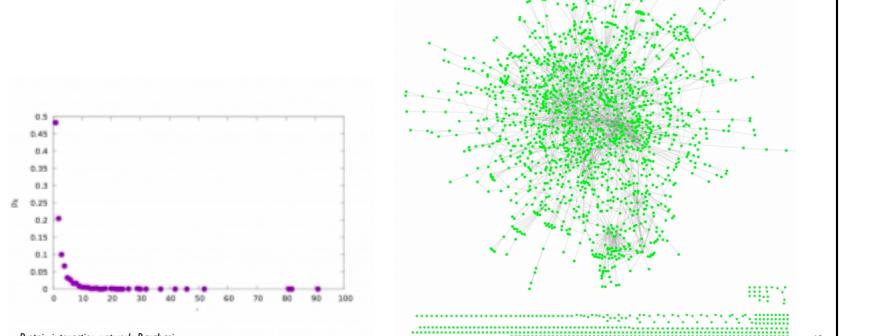
Network tasks: topology-based

- topological structure of network
 - node importance metrics
 - node degree: attribute on nodes
 - number of links connected to a node
 - local measure of importance
 - average degree, degree distribution



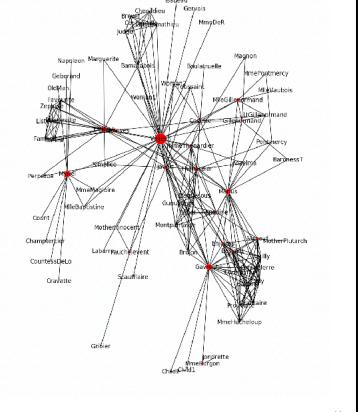
Degree distribution

- real network
 - power law distributions are common

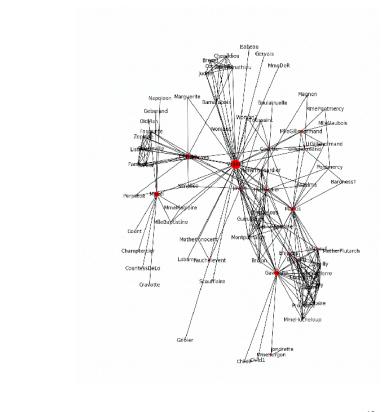
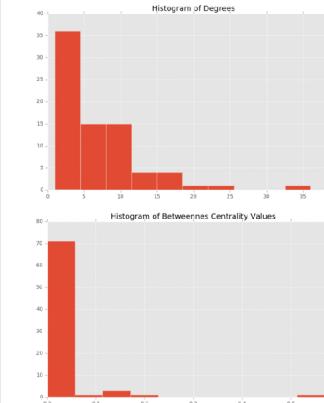


Network tasks: topology-based

- topological structure of network
 - node importance metrics
 - betweenness centrality: attribute on nodes
 - how many shortest paths pass through a node
 - global measure of importance
 - good measure for overall relevance of node in network

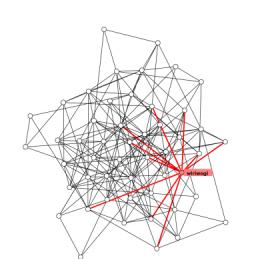


Centrality measures: Degree vs betweenness centrality

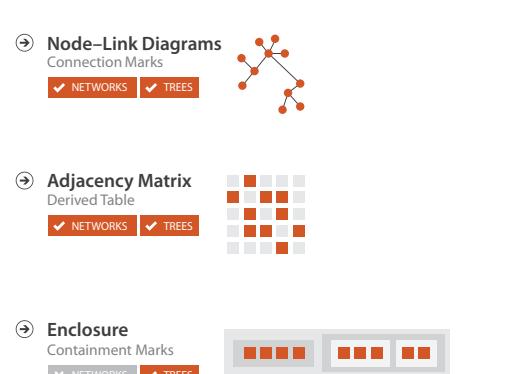


Network tasks: attribute-based vs topology-based

- topology based tasks
 - find paths
 - find (topological) neighbors
 - compare centrality/importance measures
 - identify clusters / communities
- attribute based tasks (similar to table data)
 - find extreme values, ...
- combination tasks - incorporating both
- example: locate - find single or multiple nodes/links with a given property
 - topology: find all adjacent nodes of given node
 - attributes: find edges with maximum edge weight

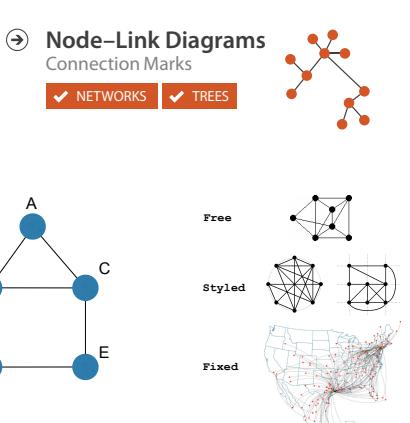


Three kinds of network visual encodings



Node-link diagrams

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

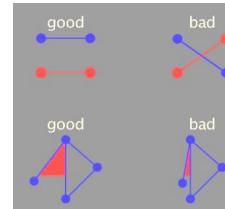


Exercise

- sketch an aesthetically pleasing node-link diagram of this network
 - there are five nodes: A,B,C,D,E
 - each row in the table describes an edge
- A B
C D
C B
A D
A C
B D
D E
A E
- Socrative quiz: pick true when done [~5 min]

Criteria for good node-link layouts

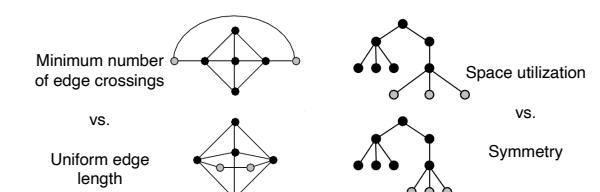
- minimize
 - edge crossings
 - distances between topological neighbor nodes
 - total drawing area
 - edge bends
 - edge length disparities (sometimes)
- maximize
 - angular distance between different edges
 - aspect ratio disparities
- emphasize symmetry
 - similar graph structures should look similar in layout



17

Criteria conflict

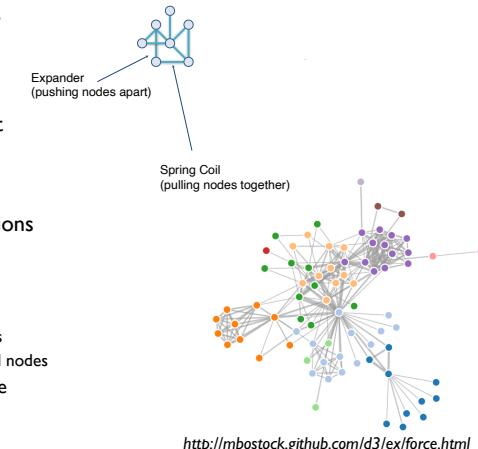
- most criteria NP-hard individually
- many criteria directly conflict with each other



18

Optimization-based layouts

- formulate layout problem as optimization problem
- convert criteria into weighted cost function
 - $F(\text{layout}) = a * [\text{crossing counts}] + b * [\text{drawing space used}] + \dots$
- use known optimization techniques to find layout at minimal cost
 - energy-based physics models
 - force-directed placement
 - spring embedders



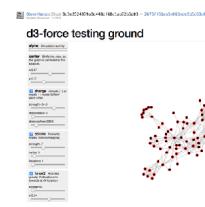
Force-directed placement

- 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 $c * \text{vertex_force}$

19

Force-directed placement properties

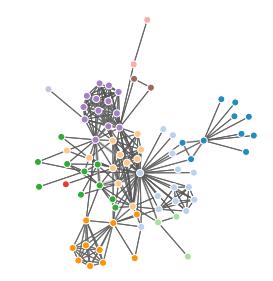
- strengths
 - reasonable layout for small, sparse graphs
 - clusters typically visible
 - edge length uniformity
- weaknesses
 - nondeterministic
 - computationally expensive: $O(n^3)$ for n nodes
 - each step is n^2 , takes $\sim n$ cycles to reach equilibrium
 - naive FD doesn't scale well beyond 1K nodes
 - iterative progress: engaging but distracting

<https://blocks.org/steveharoz/8c3e2524079a8c440df60c1db72b5d03>

21

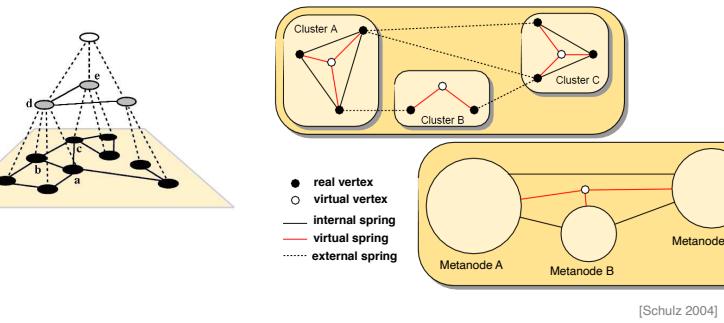
Idiom: force-directed placement

- 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 $E < 4N$

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

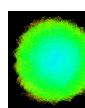
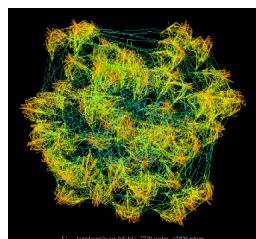
Multilevel approaches

- derive cluster hierarchy of metanodes on top of original graph nodes



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

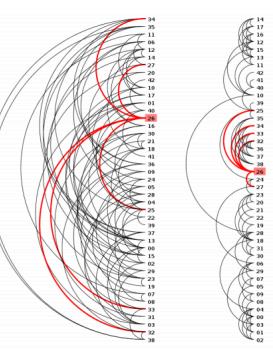
- data
 - original: network
 - derived: cluster hierarchy atop it
- considerations
 - better algorithm for same encoding technique
 - same: fundamental use of space
 - hierarchy used for algorithm speed/quality but not shown explicitly
- scalability
 - nodes, edges: 1K-10K
 - hairball problem eventually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



24

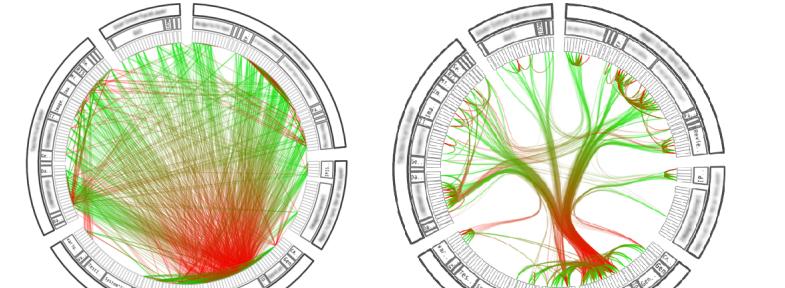
Restricted layouts: Circular, arc

- lay out nodes around circle or along line
 - circular layouts
 - arc diagrams
- node ordering crucial to avoid excessive clutter from edge crossings
 - barycentric ordering before & after
 - derived attribute: global computation

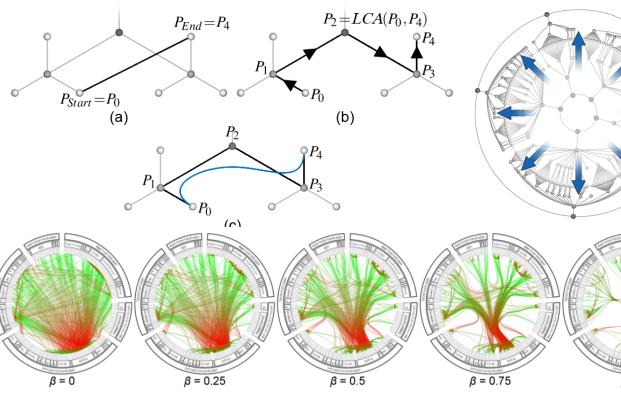
<http://profs.etsmtl.ca/mmcguffin/research/2012-mcguffin-simpleNetVis/mcguffin-2012-simpleNetVis.pdf>

25

Edge clutter reduction: hierarchical edge bundling

[http://Hierarchical.Edge.Bundles:Visualization.of.Adjacency.Relations.in.Hierarchical.Data.Danny.Holten.TVCG.12\(5\):741-748.2006](http://Hierarchical.Edge.Bundles:Visualization.of.Adjacency.Relations.in.Hierarchical.Data.Danny.Holten.TVCG.12(5):741-748.2006)

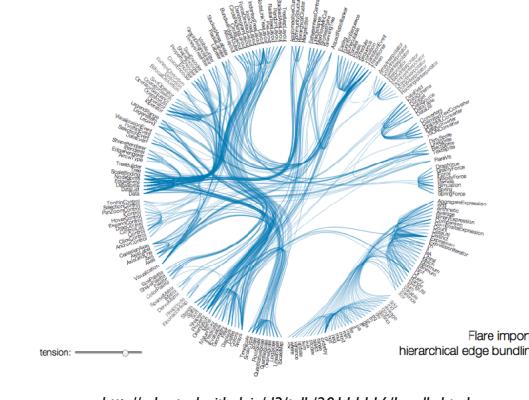
Hierarchical edge bundling



26

27

Bundle strength



28

Fixed layouts: Geographic

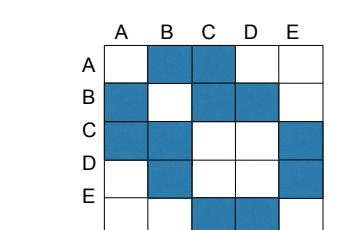
- lay out network nodes using given/fixed spatial data
 - route edges accordingly
 - edge bundling also applicable

<https://www.facebook.com/notes/facebook-engineering/visualizing-friendships/469716398919>

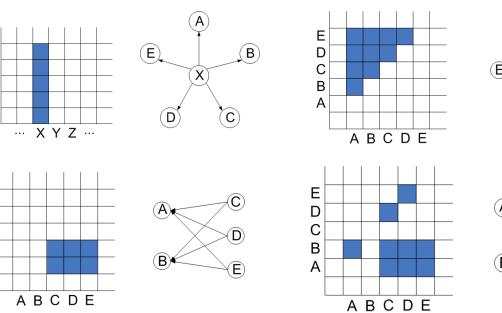
29

Adjacency matrix representations

- derive adjacency matrix from network



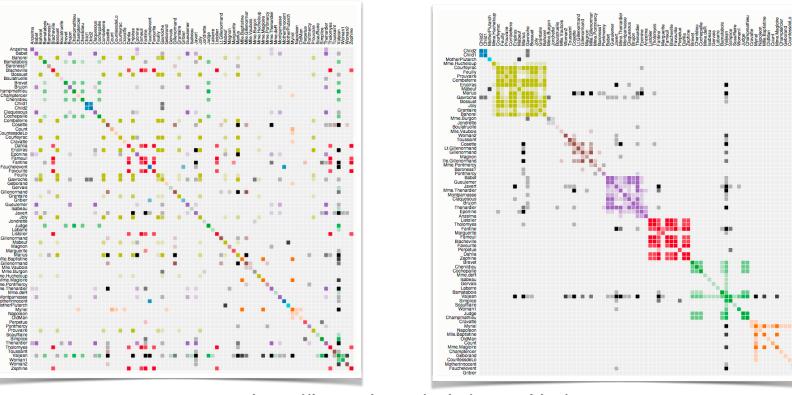
Adjacency matrix examples



30

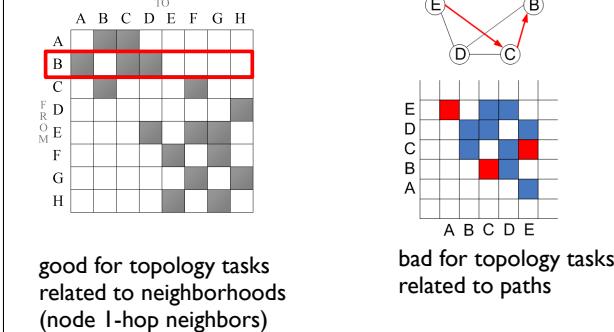
31

Node order is crucial: Reordering

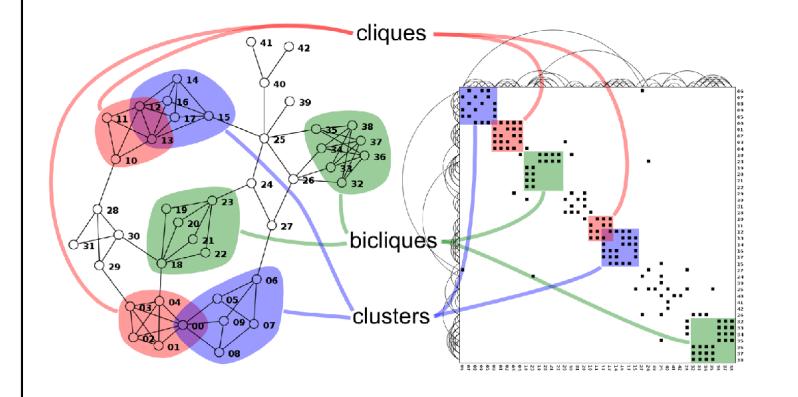


32

Adjacency matrix

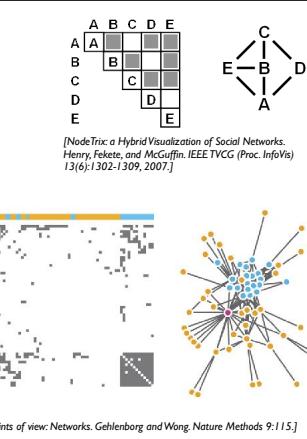


Structures visible in both



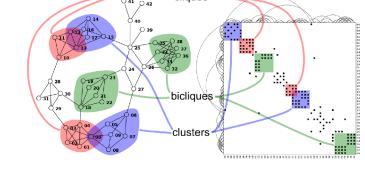
Idiom: adjacency matrix view

- 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



Node-link vs. matrix comparison

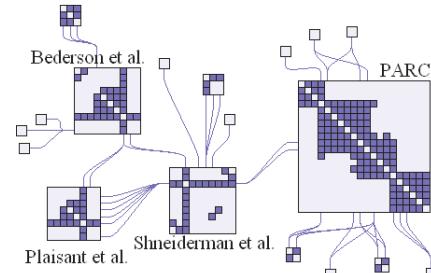
- node-link diagram strengths
 - topology understanding, path tracing
 - intuitive, flexible, no training needed
- adjacency matrix strengths
 - focus on edges rather than nodes
 - layout straightforward (reordering needed)
 - predictability, scalability
 - some topology tasks trainable
- empirical study
 - node-link best for small networks
 - matrix best for large networks
 - if tasks don't involve path tracing!



[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: NodeTrix

- hybrid nodelink/matrix
- capture strengths of both

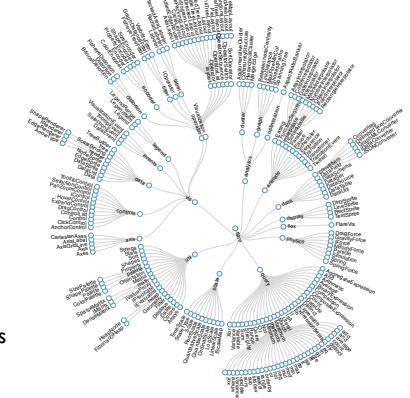


NodeTrix
[Henry et al. 2007]

Trees

Idiom: radial node-link tree

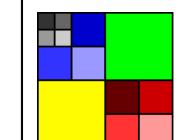
- data
- tree
- encoding
 - link connection marks
 - point node marks
 - radial axis orientation
 - angular proximity: siblings
 - distance from center: depth in tree
- tasks
 - understanding topology, following paths
- scalability
 - 1K - 10K nodes



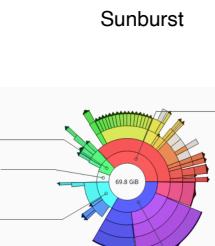
<http://mbostock.github.io/d3/ex/tree.html>

Containment (implicit) layouts

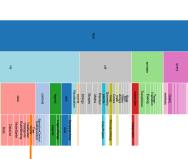
Treemap



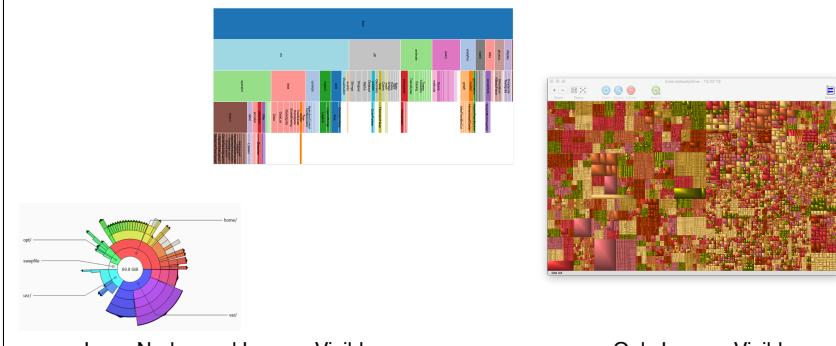
Sunburst



Icicle Plot

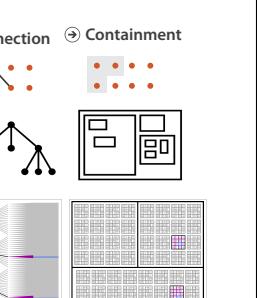


Containment: Approaches compared



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



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

Tree exercise

Here is part of a directory structure used for the material for this class and the relative file size.

datavis-17/
lectures/
 Intro.key (110 MB)
 perception/
 Perception.key (113 MB)
 Blindness.mov (15MB)
 Data.key (12 MB)
 Graphs.key (180 MB)
exams/
 Exam1-solution.doc (5MB)
 Exam1.doc (1MB)
exercise/
 Graph.doc (3MB)
 Graph-video.doc (210MB)

Sketch two different visualizations that show both, the directory structure and the size of the directories and the contained files.

- socrative: true when done with both

Idiom: treemap

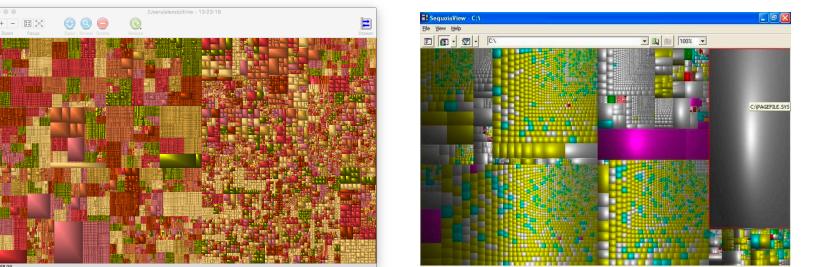
- data
 - tree
 - 1 quant attrib at leaf nodes
- encoding
 - area containment marks for hierarchical structure
 - rectilinear orientation
 - size encodes quant attrib
- tasks
 - query attribute at leaf nodes
- scalability
 - 1M leaf nodes



http://tulip.labri.fr/Documentation/3_7/userHandbook/html/ch06.html

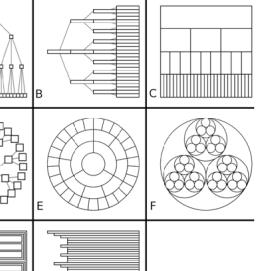
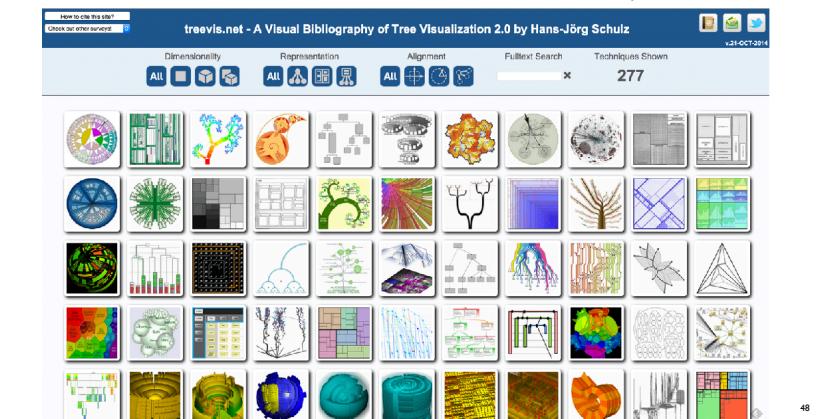
Treemap software: disk space

Mac: GrandPerspective Windows: Sequoia View



treevis.net: many, many options

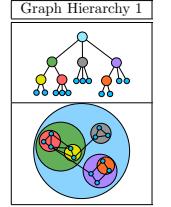
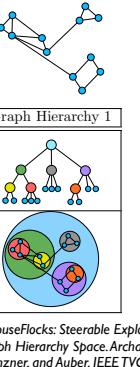
<https://treevis.net/>



[Quantifying the Space-Efficiency of 2D Graphical Representations of Trees. McGuffin and Robert. Information Visualization 9:2 (2010), 115-140.]

Idiom: GrouseFlocks

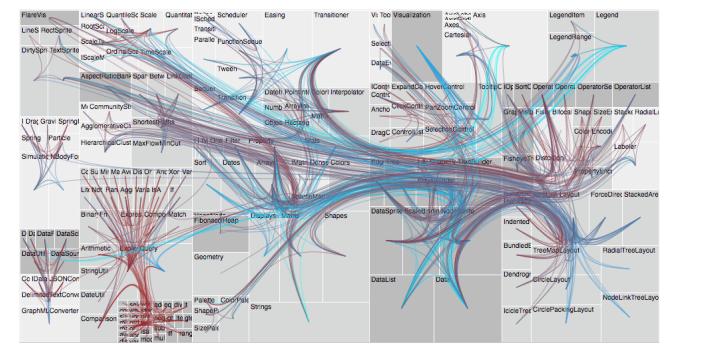
- data: compound graphs
 - network
 - cluster hierarchy atop it
 - derived or interactively chosen
- visual encoding
 - connection marks for network links
 - containment marks for hierarchy
 - point marks for nodes
- dynamic interaction
 - select individual metanodes in hierarchy to expand/contract



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

49

Hierarchical edge bundling: treemap vs radial



Credits

- Visualization Analysis and Design (Ch 9)
- Alex Lex & Miriah Meyer, <http://dataviscourse.net/>

50

51