



# CS-GY 6313 B: Information Visualization

10/10/2024

# Logistics

- Assignment 2 due next week



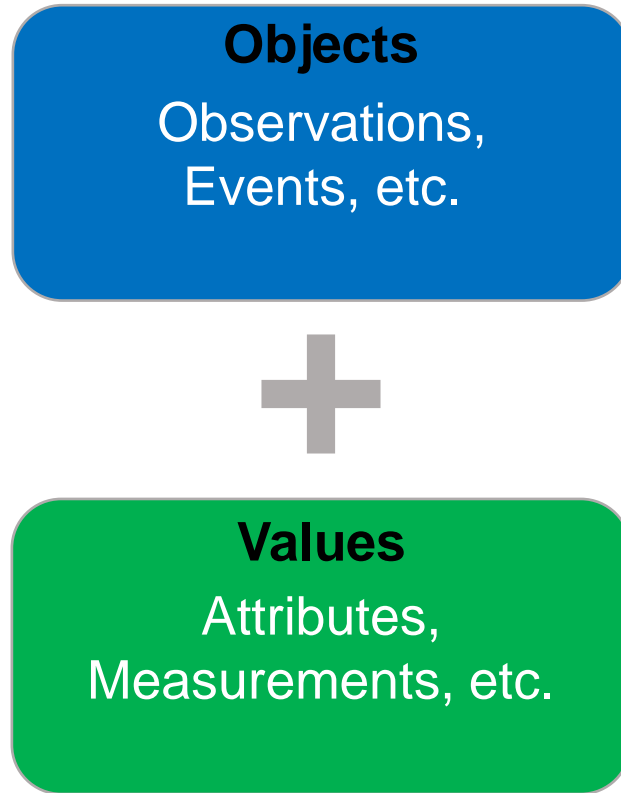


# 2D Viz: Networks and Trees

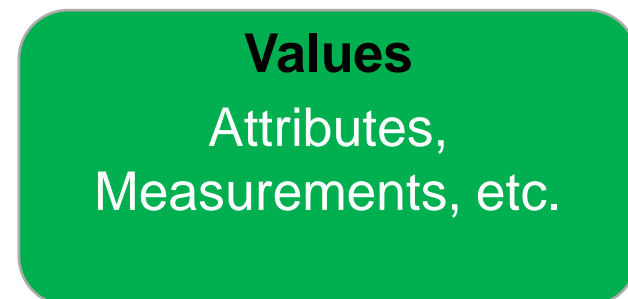
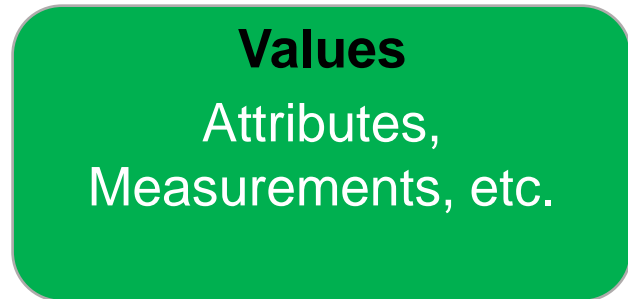
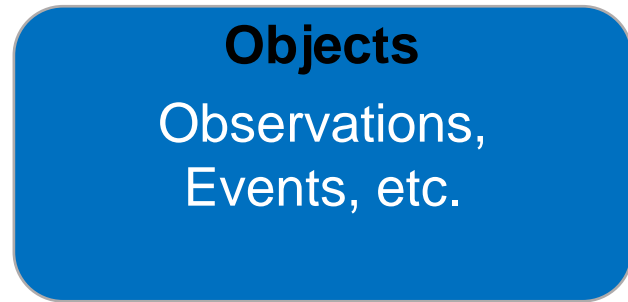
# Network data

- A way to model *relationships* between entities
- Also known as:
  - Graph
  - Node-link graph

# Tabular data



# Network data



# Network data

Nodes

Label	Value
A	5
B	10
C	15

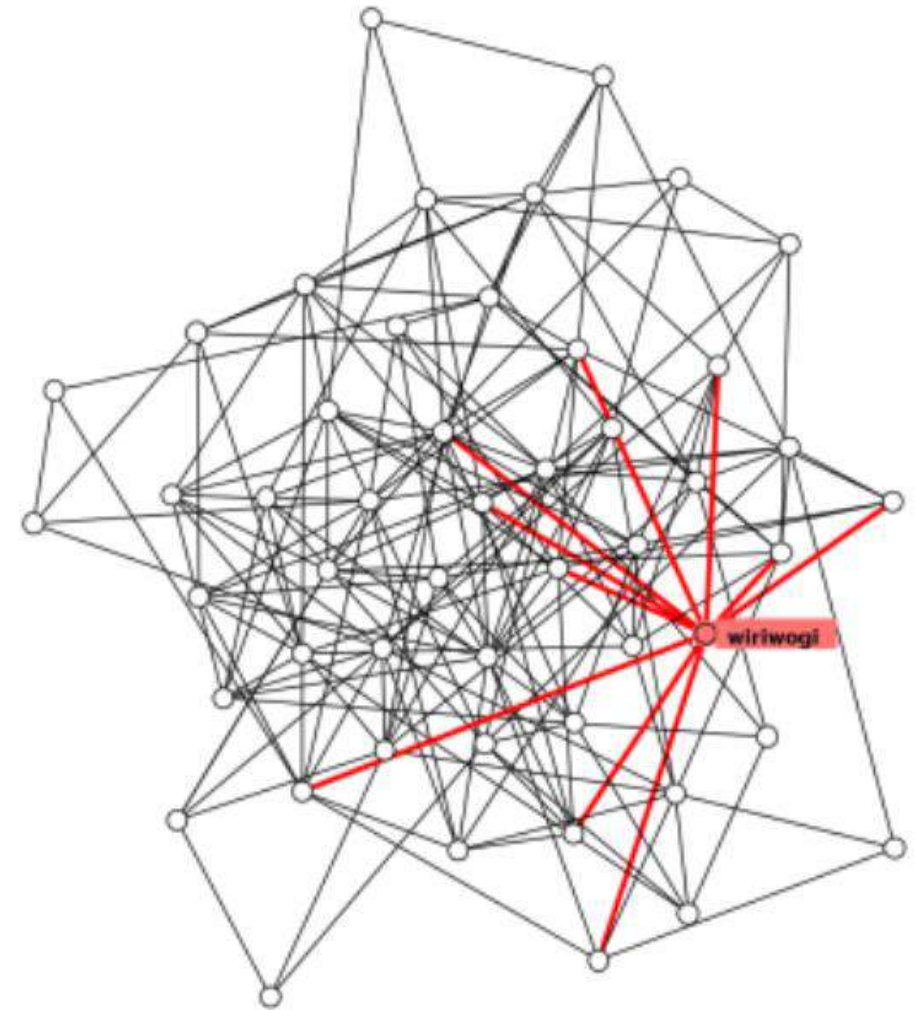


Links

N1	N2	Value
A	B	9
A	C	27
B	C	18

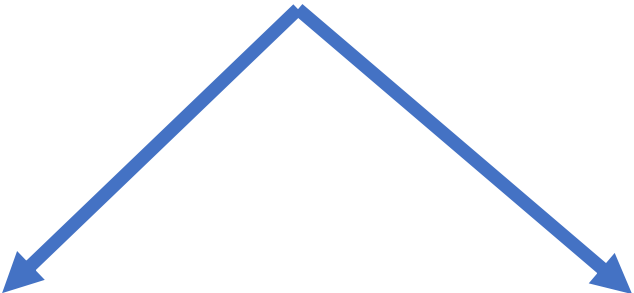
# Network tasks: topology-based and attribute-based

- Topology-based tasks:
  - Find paths
  - Find (topological) neighbors
  - Compare centrality
  - Identify clusters/communities
- Attribute-based tasks:
  - Find distributions, extreme values, etc.
- Combination tasks:
  - Use both attribute and connection data!
  - Example?
    - Find all friends-of-friends who like cats
      - Topology: friends of friends (path from you to the other node)
      - Attribute: does the person (node) like cats?

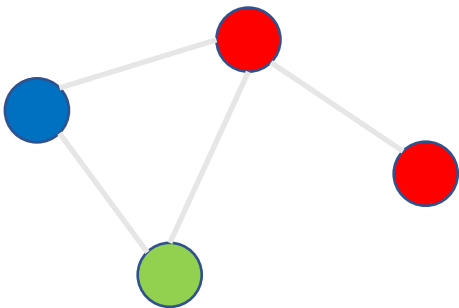




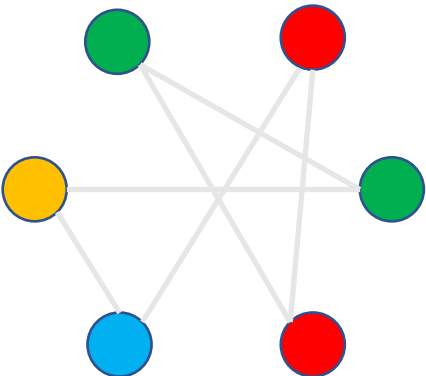
# Node-Link Diagram



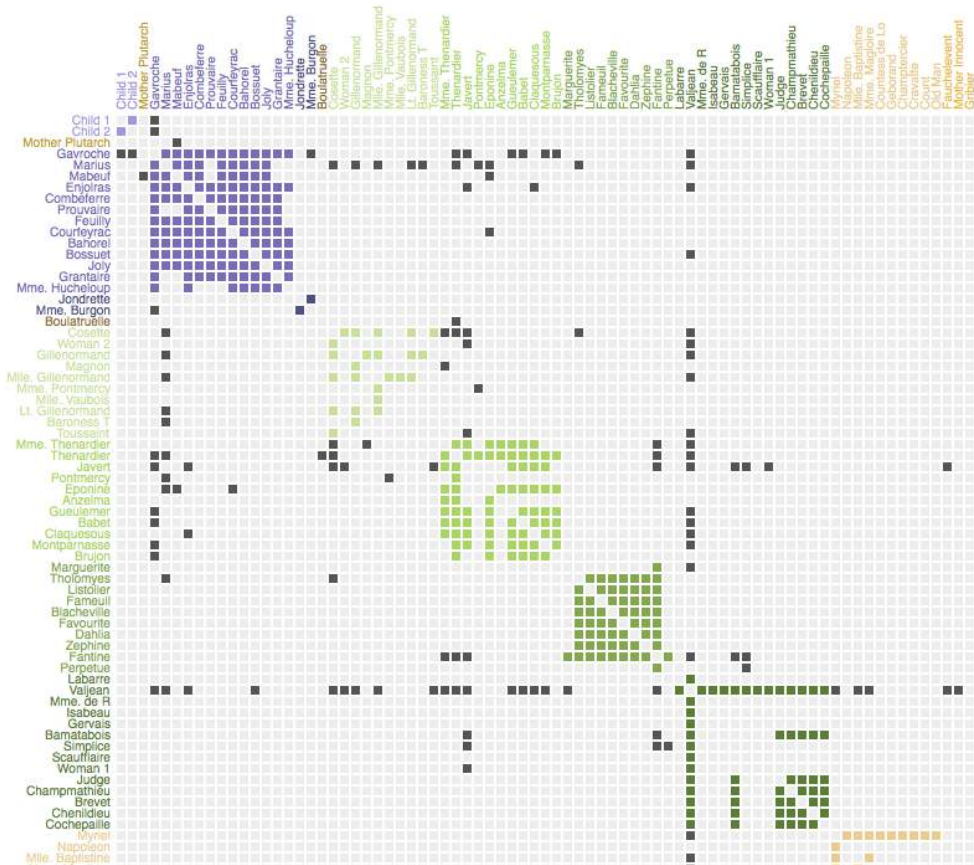
Force-Directed  
Layouts



Fixed Layouts



# Matrices

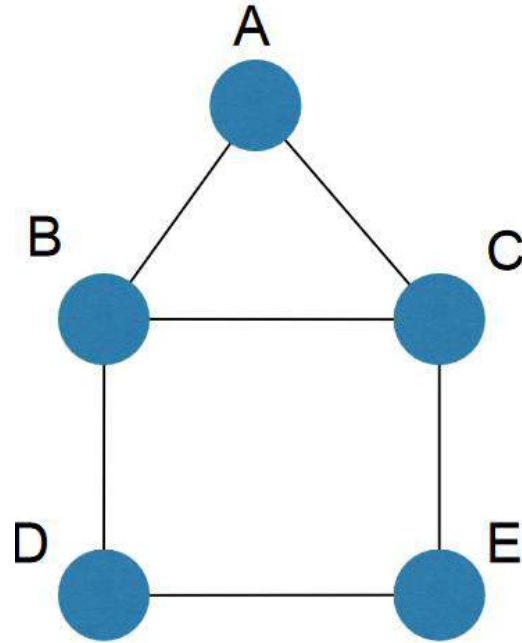




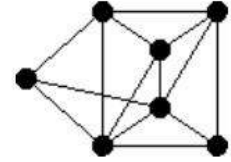
# Node-link diagrams

# Node-link diagrams

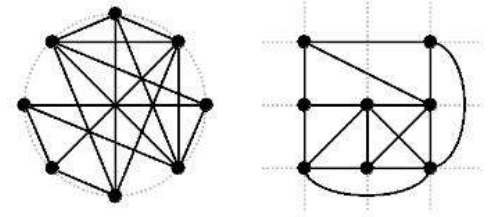
- Nodes: point marks
- Links: line marks
  - Straight lines or arcs
  - Connections between nodes
- Intuitive and familiar
  - Most common
  - Many, many variants



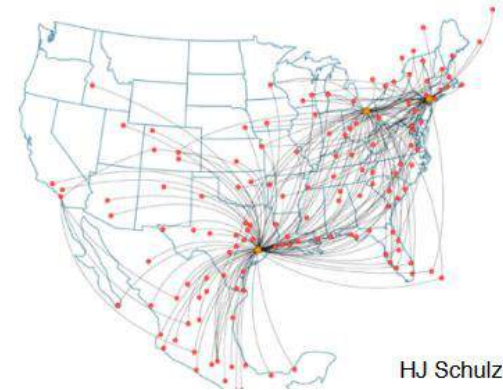
**Free**



**Styled**

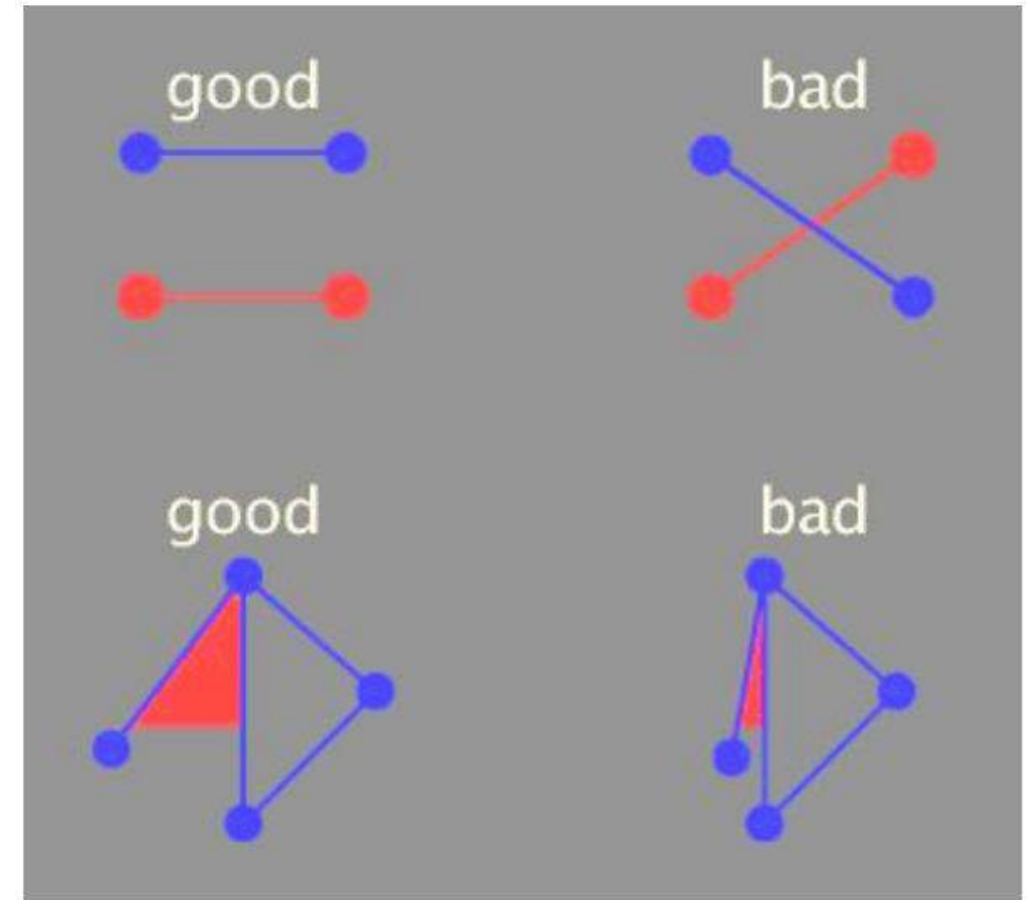


**Fixed**



# What makes for a good node-link layout?

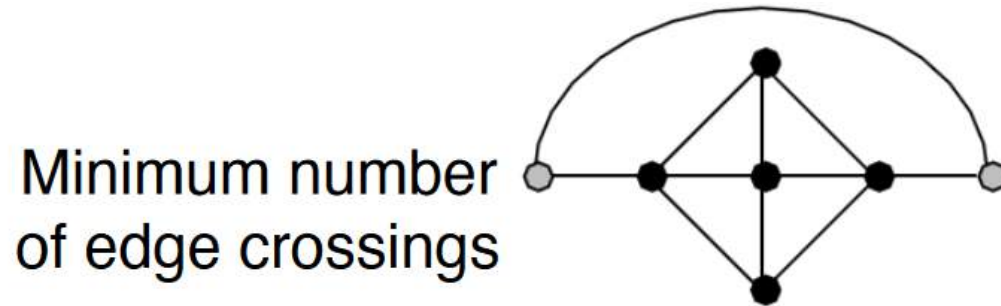
- Minimize:
  - Edge crossings
  - Node overlaps
  - Distances between topological neighbor nodes
  - Total drawing area (white space)
  - Edge bends
- Maximize:
  - Angular distance between edges
  - Aspect ratio disparities
- Emphasize symmetry
  - Similar graph structures should look similar in layout





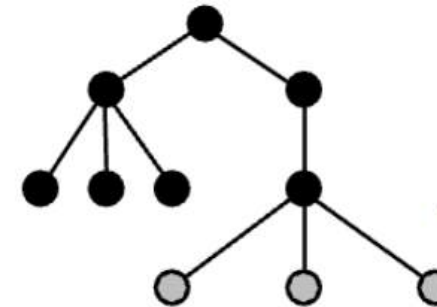
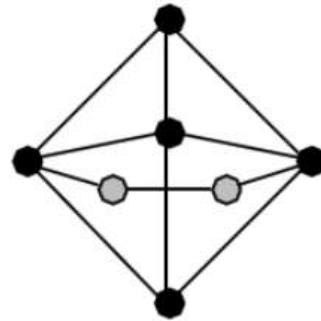
# Criteria conflict

- Meeting all of those criteria is hard
- Most criteria are NP-hard just on their own
- Many criteria directly conflict with each other
- Solution: use heuristics



vs.

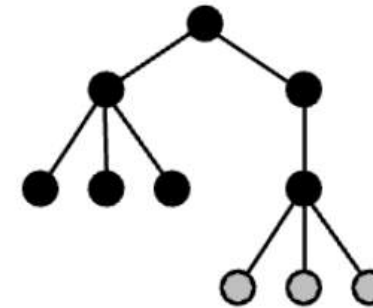
Uniform edge length



Space utilization

vs.

Symmetry



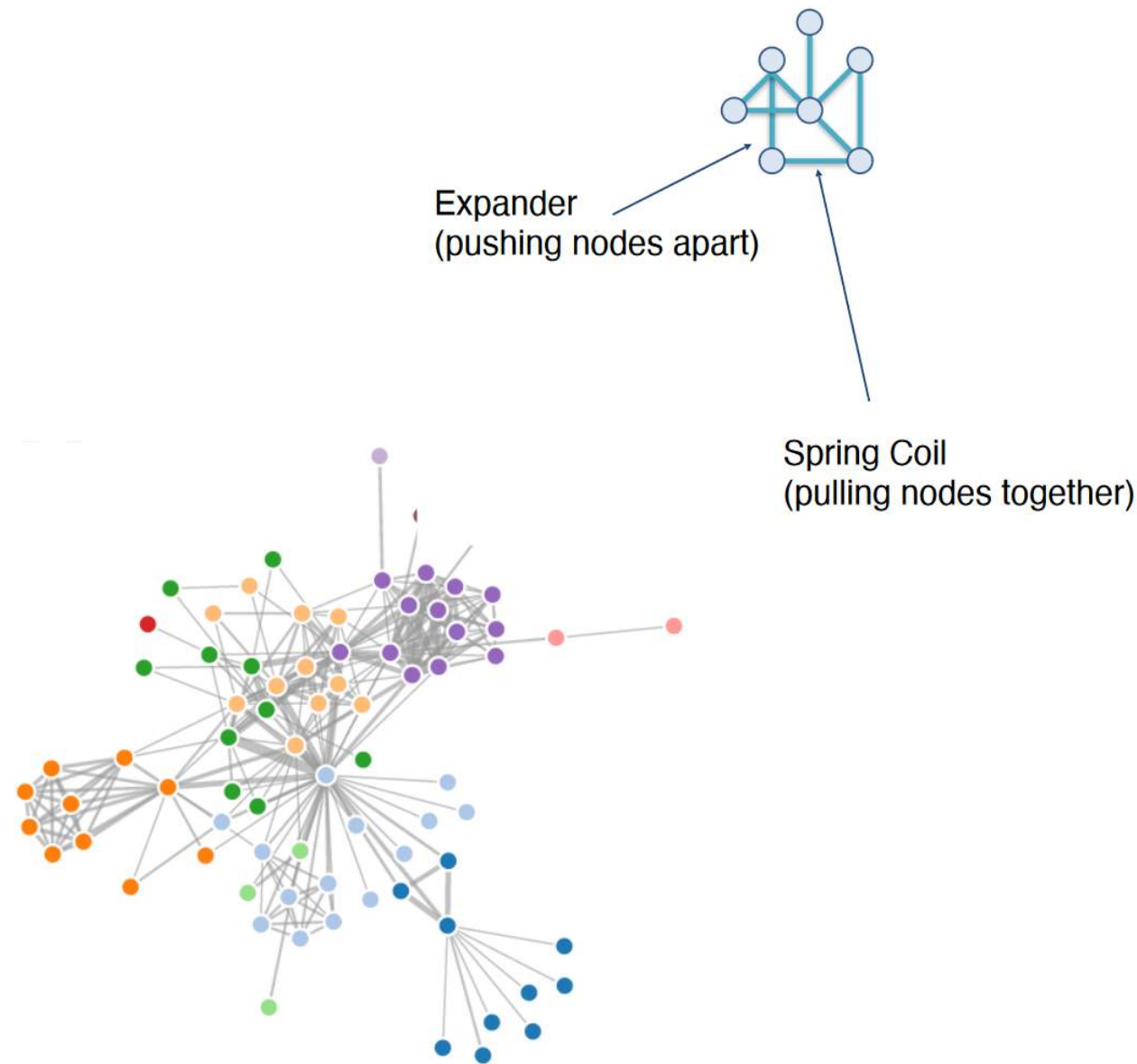
# Optimization-based layouts

- 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 away
- Algorithm
  - Place vertices in random locations
  - While not in equilibrium:
    - For each vertex:
      - Calculate force on vertex
        - Sum of:
          - Pairwise repulsion of all nodes
          - Attraction between connected nodes
      - Move vertex by  $c * \text{vertex\_force}$

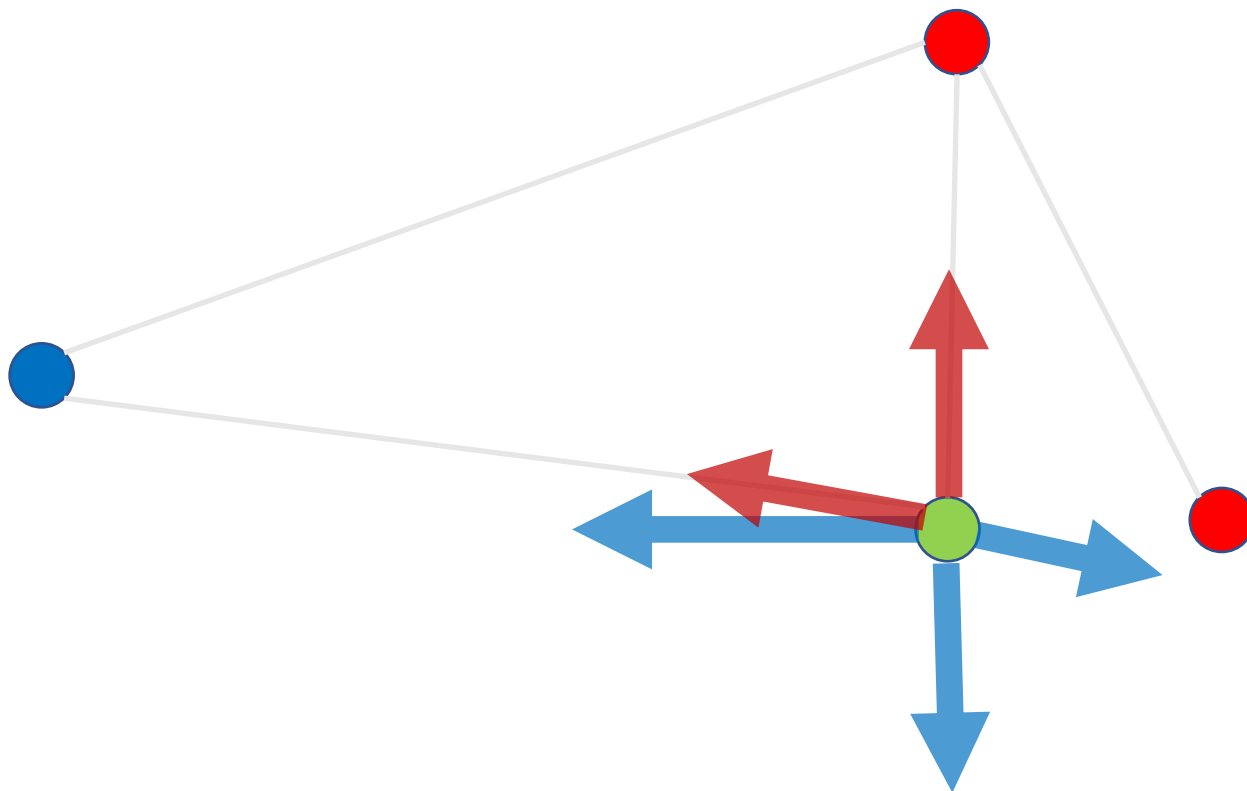


<https://observablehq.com/@d3/force-directed-graph-component>

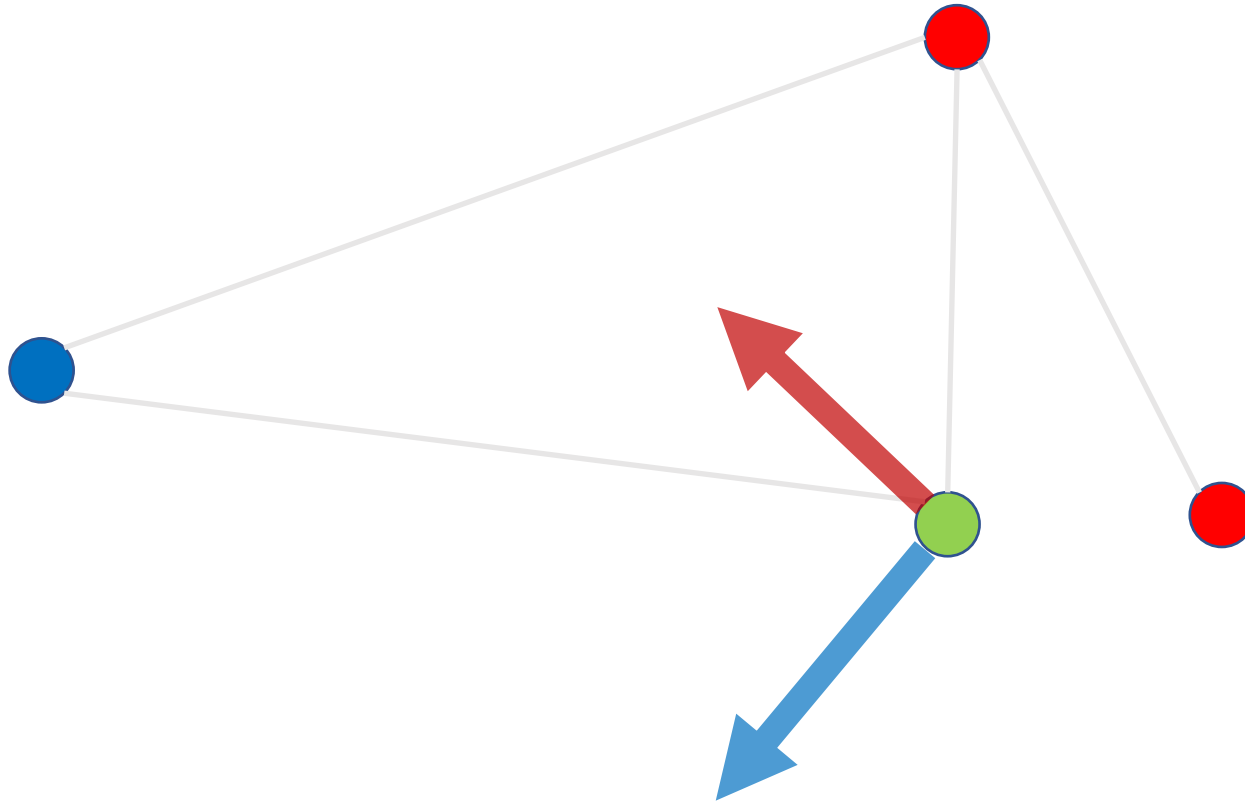
**Initialize with random layout**



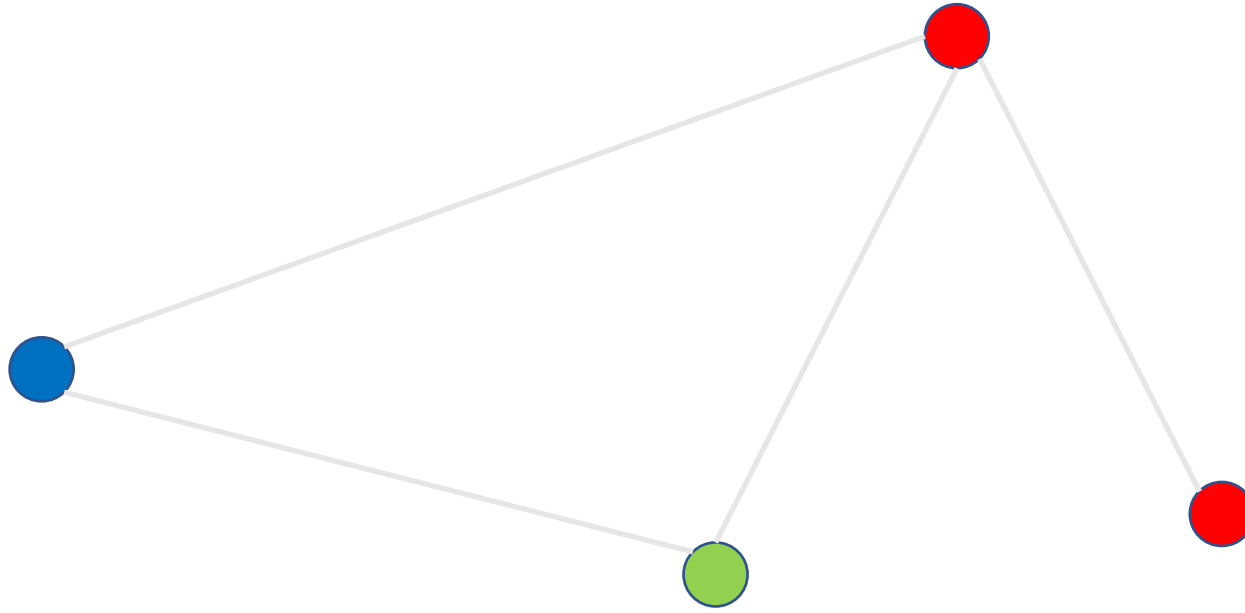
**Compute attraction and repulsion forces per node**



**Compute the net force per node**



**Move the node according to the net force – Hooke's Law**



**And repeat...**

# Force-directed placement properties

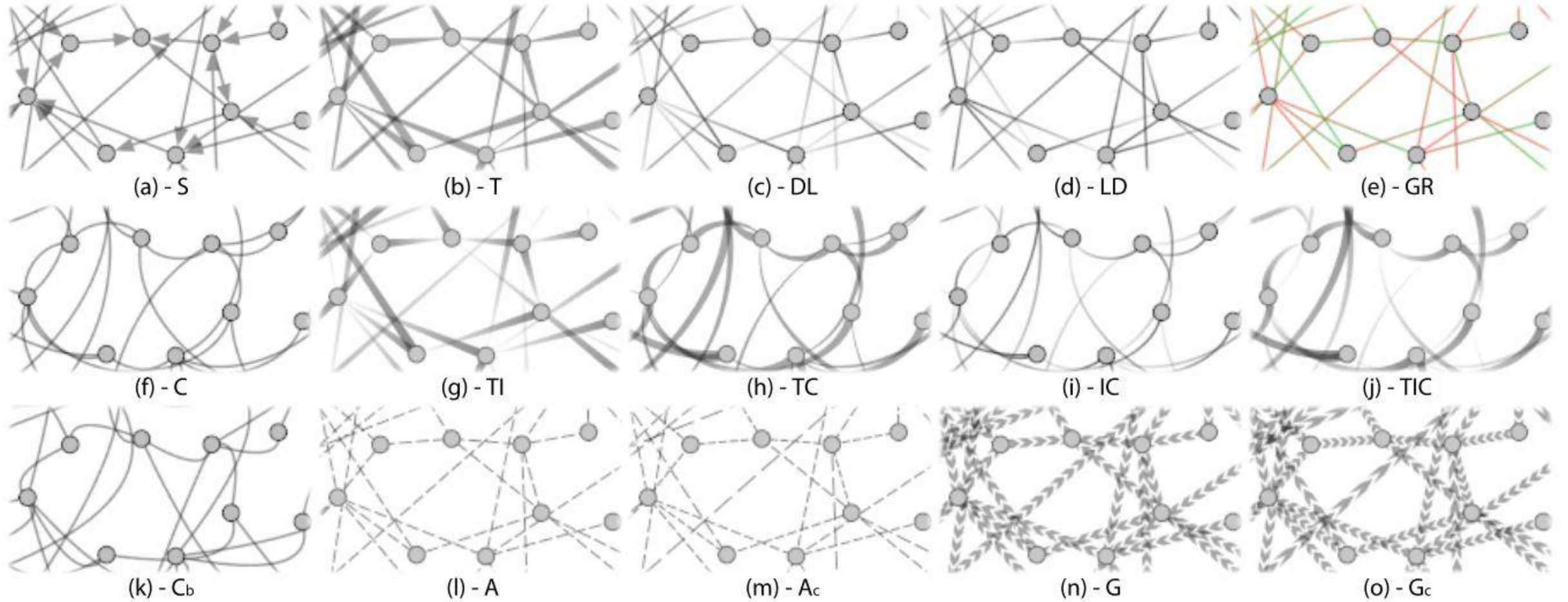
- Strengths:
  - Reasonable layout for small, sparse graphs
  - Clusters typically visible
  - Edge length uniformity
- Weaknesses:
  - Nondeterministic
  - Computationally expensive:  $O(n^3)$  for  $n$  nodes
    - Each step is  $n^2$ , takes  $\sim n$  cycles to reach equilibrium
  - Naïve FD doesn't scale well beyond  $\sim 1000$  nodes
  - Iterative progress of viz: cool but distracting
  - Visual complexity:  $E < 4N$ 
    - Max # of edges in a graph is  $N^N$ , so  $4N$  is quite sparse!



# Force-directed placement

- Considerations:
  - Spatial position: no meaning directly encoded!
    - Left free to minimize crossings
  - Semantics of proximity?
    - Sometimes meaningful
    - Sometimes arbitrary (artifact of layout algorithm)

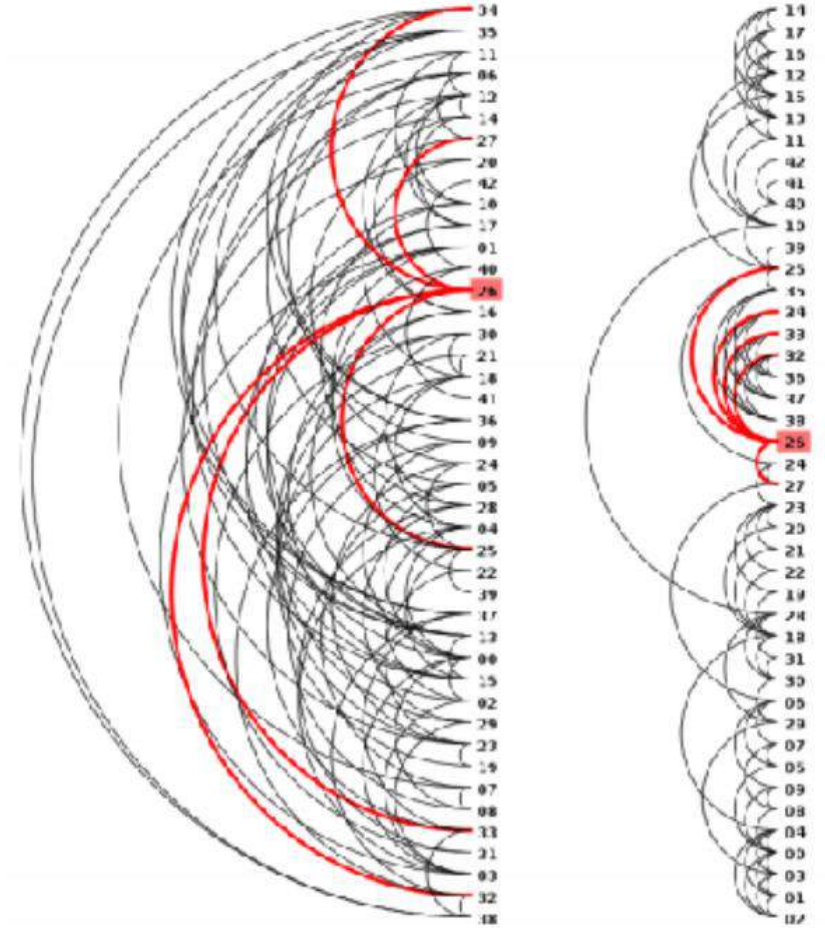
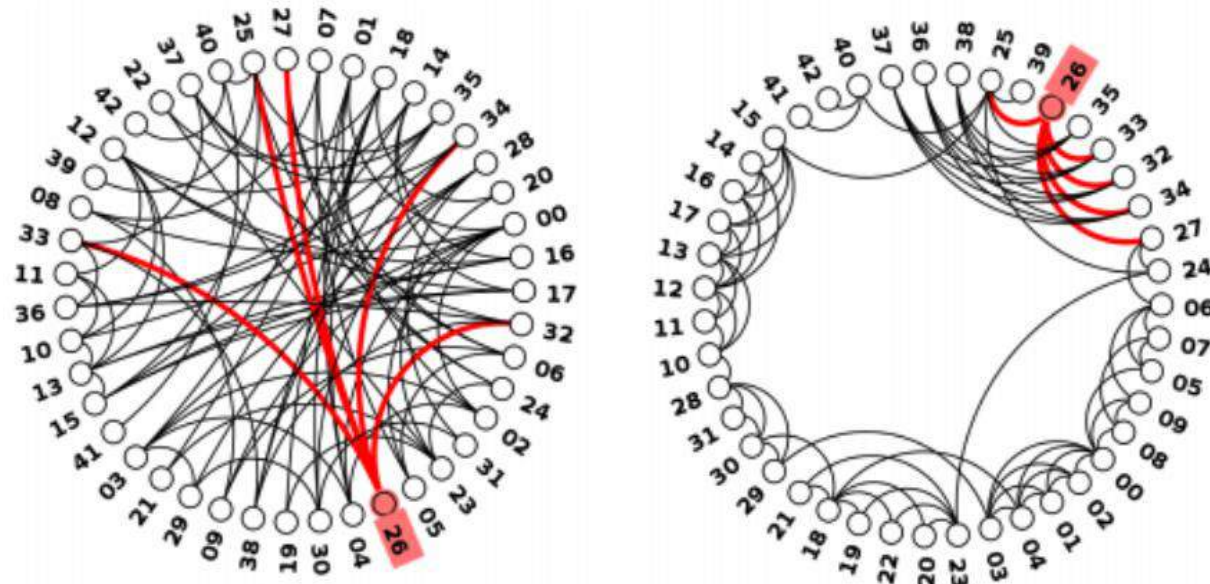
# Directed graphs



Holten, Danny, et al. "An extended evaluation of the readability of tapered, animated, and textured directed-edge representations in node-link graphs."  
*IEEE Pacific Visualization Symposium (PacificVis)*, 2011.

# Fixed layout: circular layout/arc diagram (node-link)

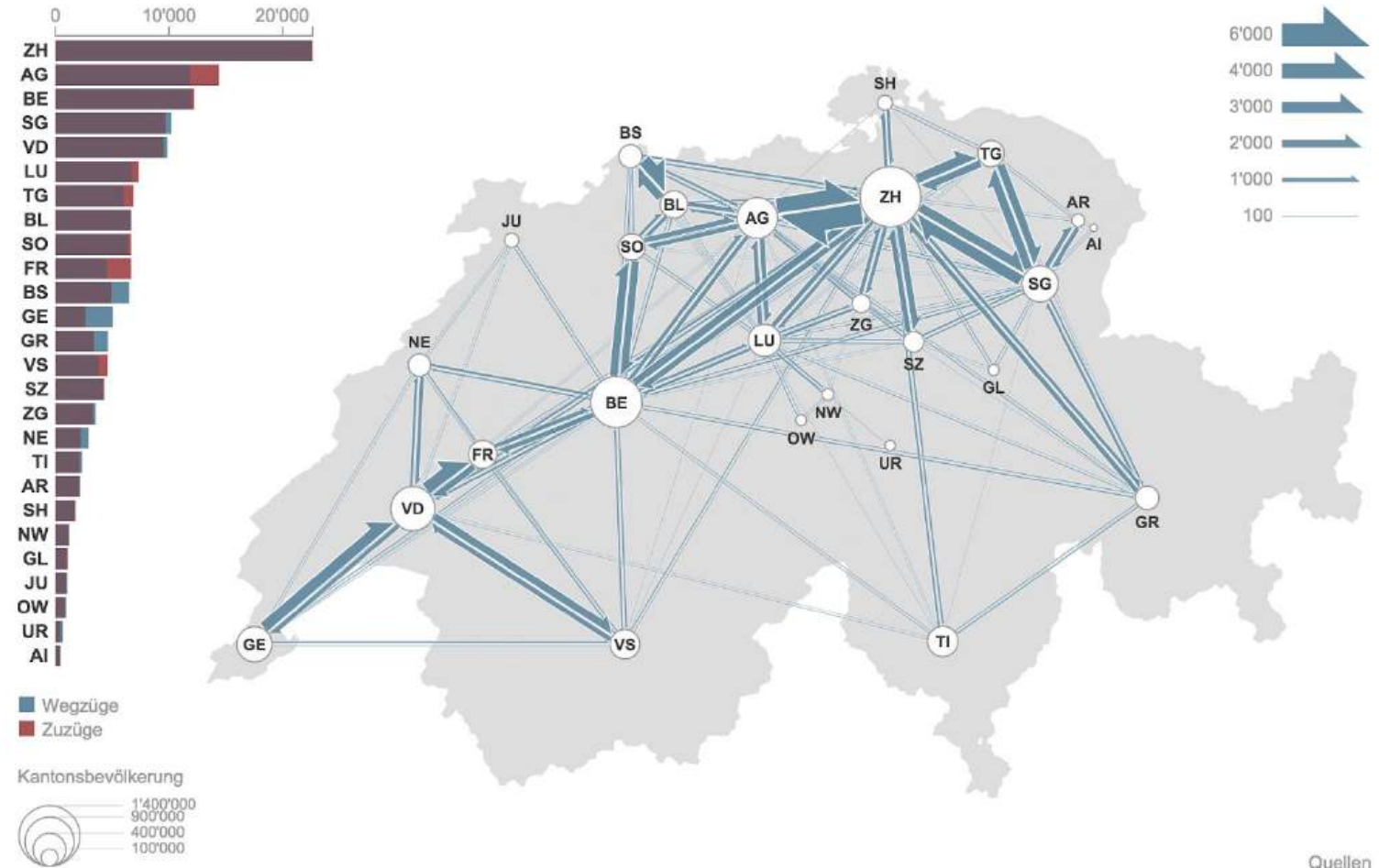
- Alternative node-link layouts: lay out nodes around circle or along line
- Data:
  - Original: network
  - Derived: node ordering attribute (global computation)
- Considerations: node ordering is crucial to avoid excessive clutter from edge crossing
  - Example: Barycentric ordering



# Fixed layout: spatial

- Nodes are placed based on geographical coordinates
- Gains all the benefits of spatial visualizations:
  - Familiarity
  - Semantic meaning of position

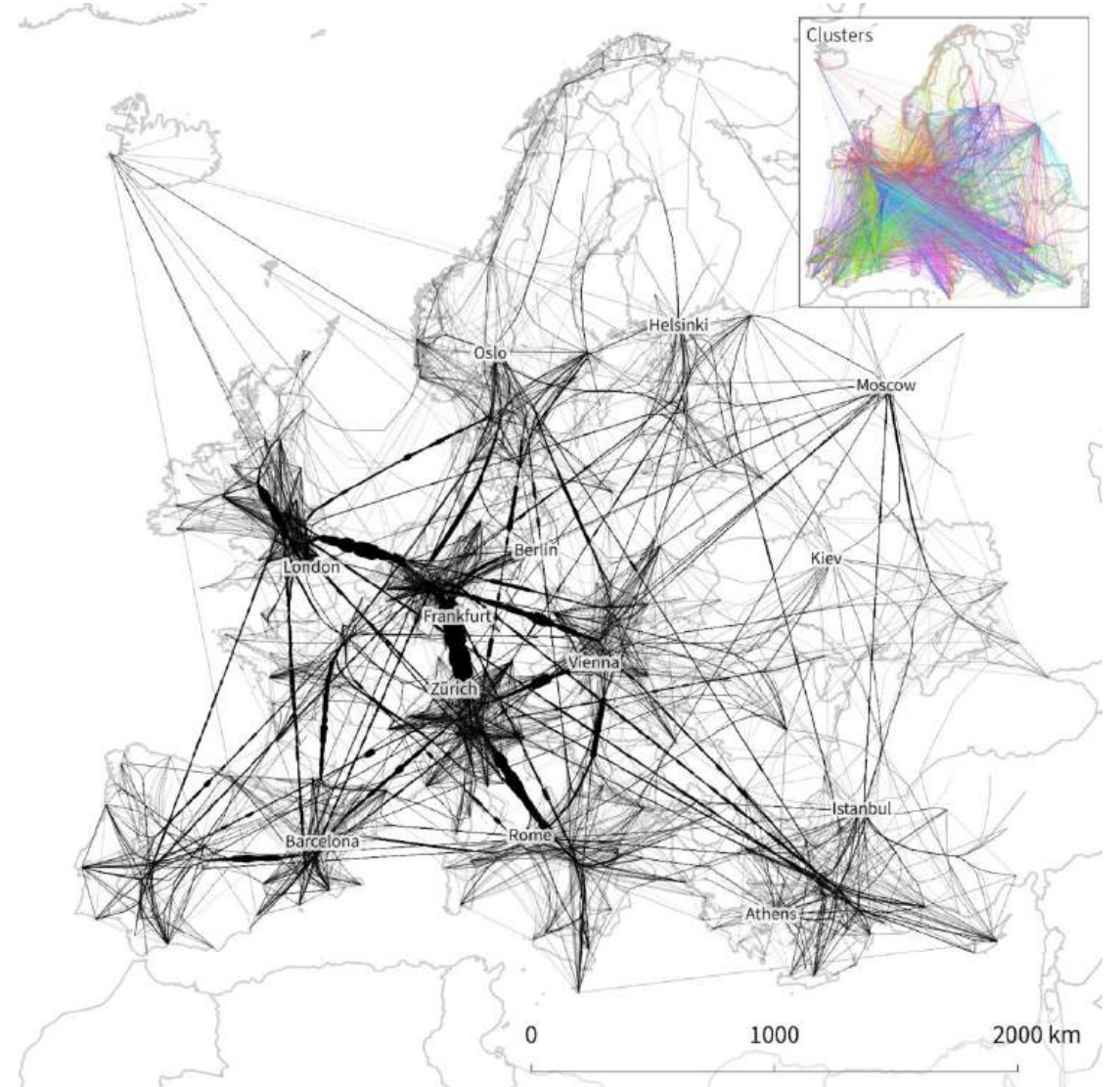
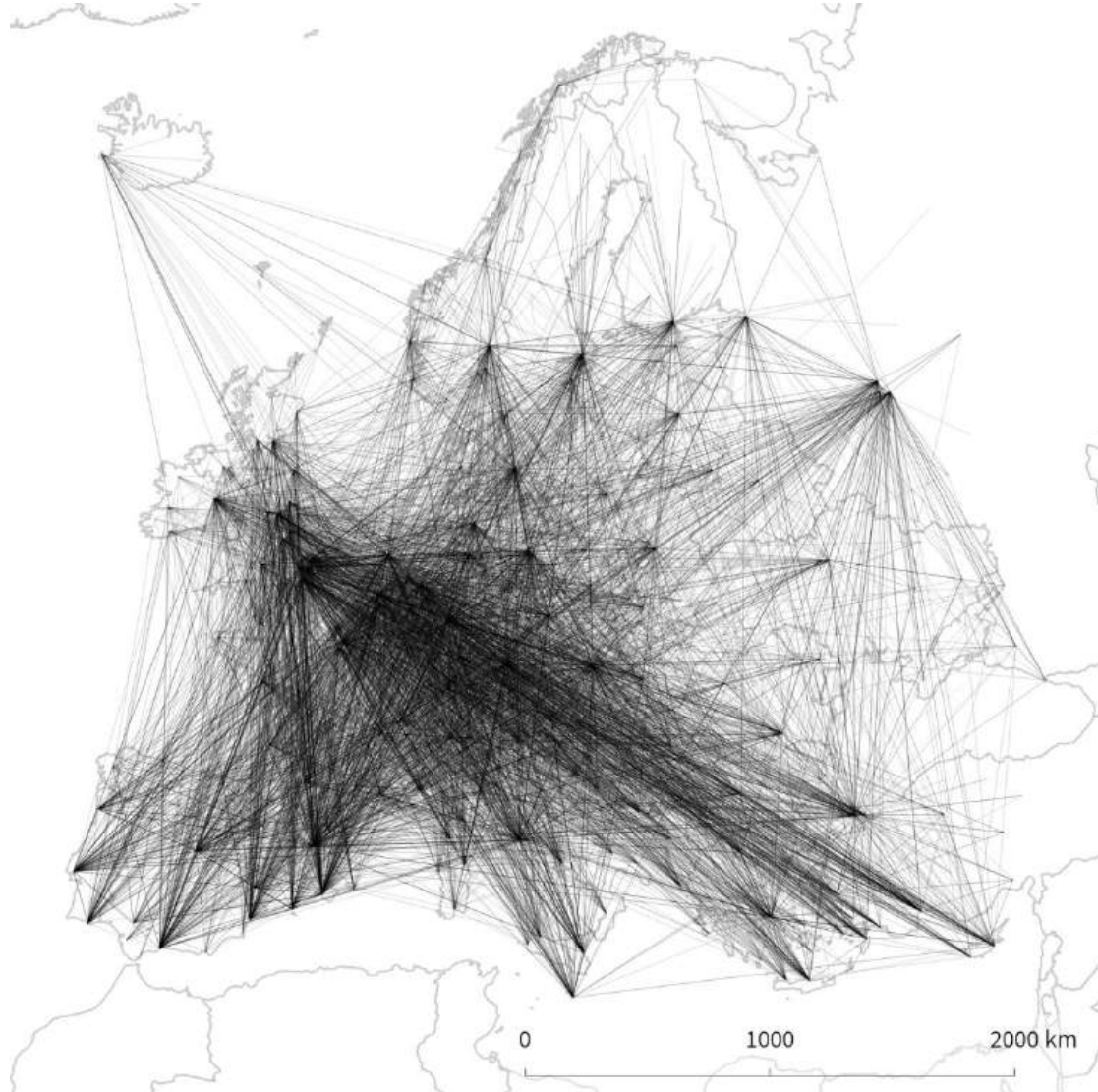
Die Wanderungsbewegungen zwischen den Kantonen im Jahr 2011





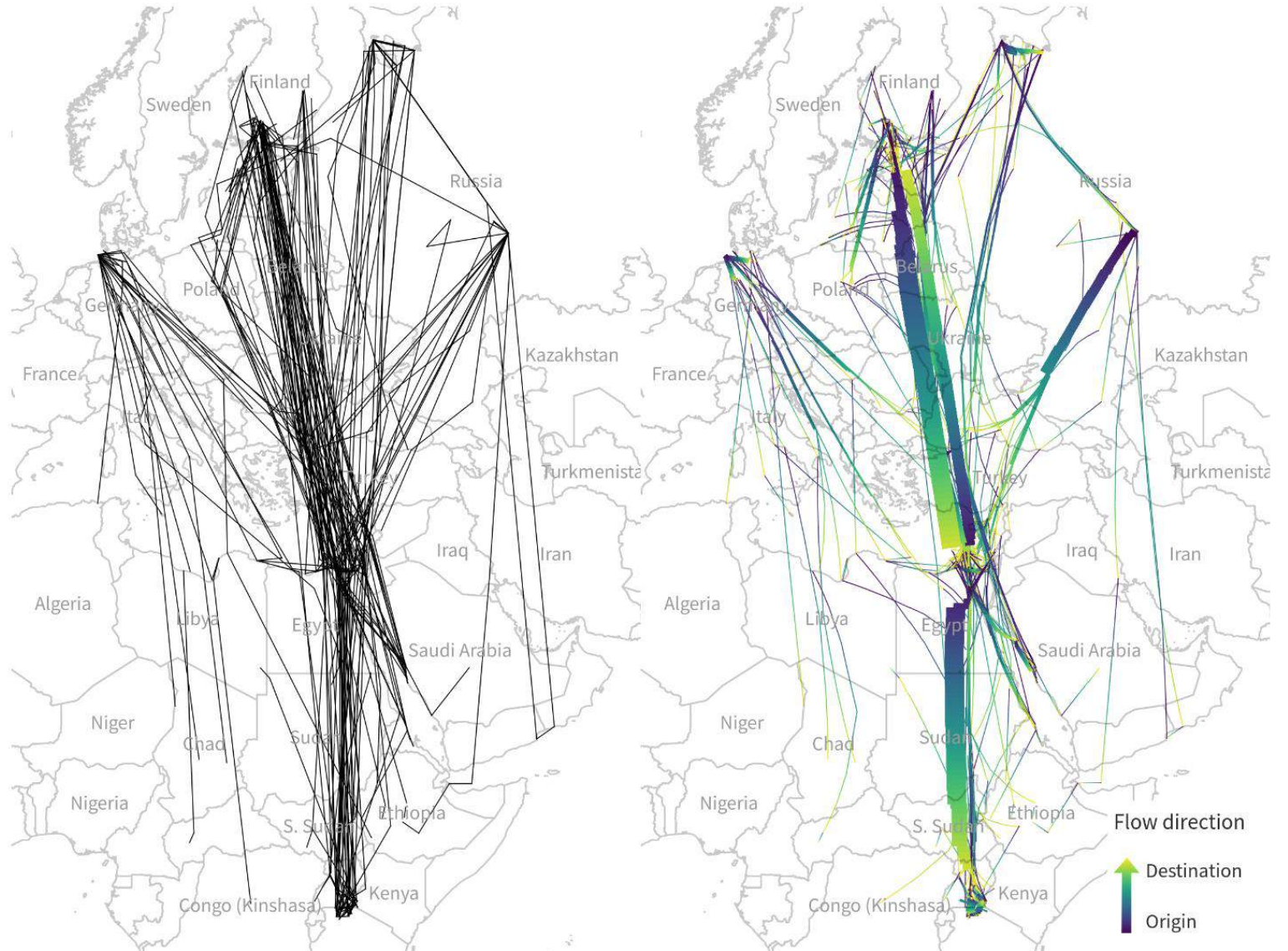
# Fixed layout: spatial

- Watch out for clutter!



# Reducing clutter

- Edge bundling





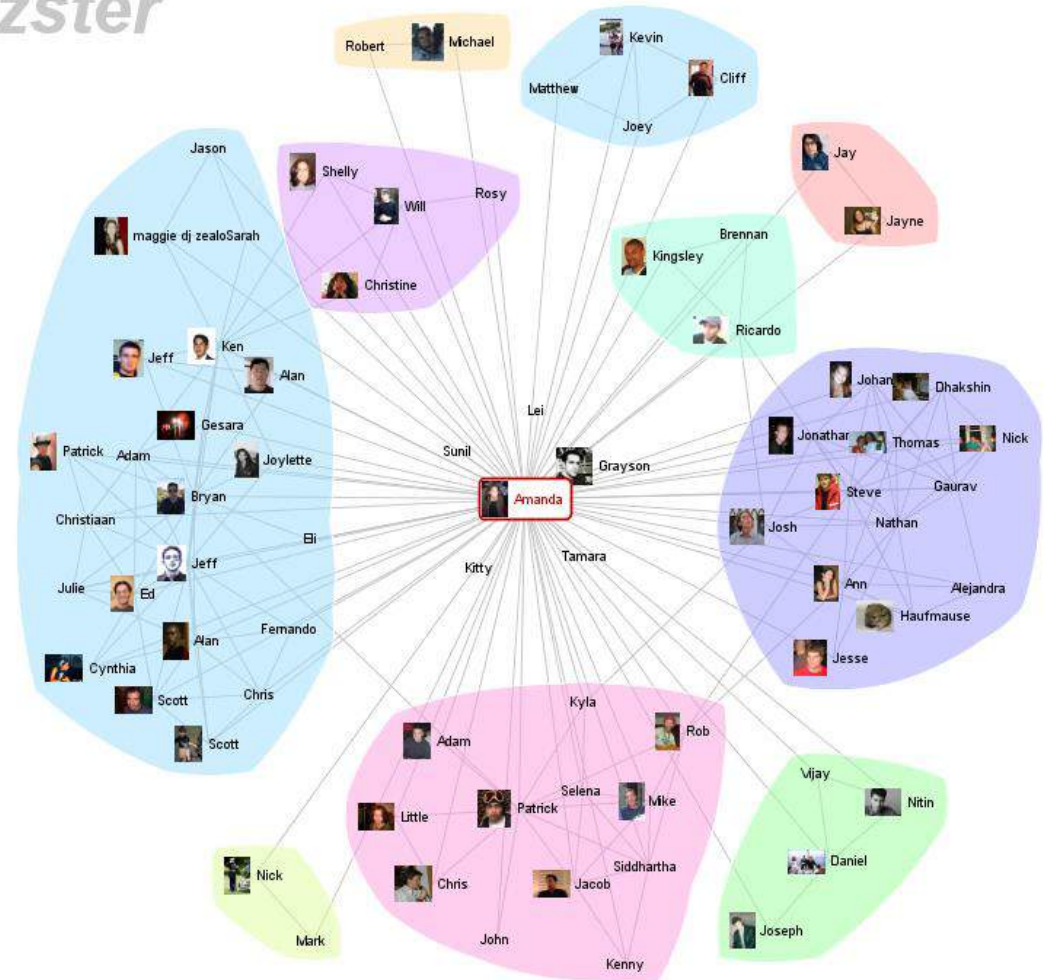
# Reducing clutter

- Edge bundling
- Clustering

*vizster*

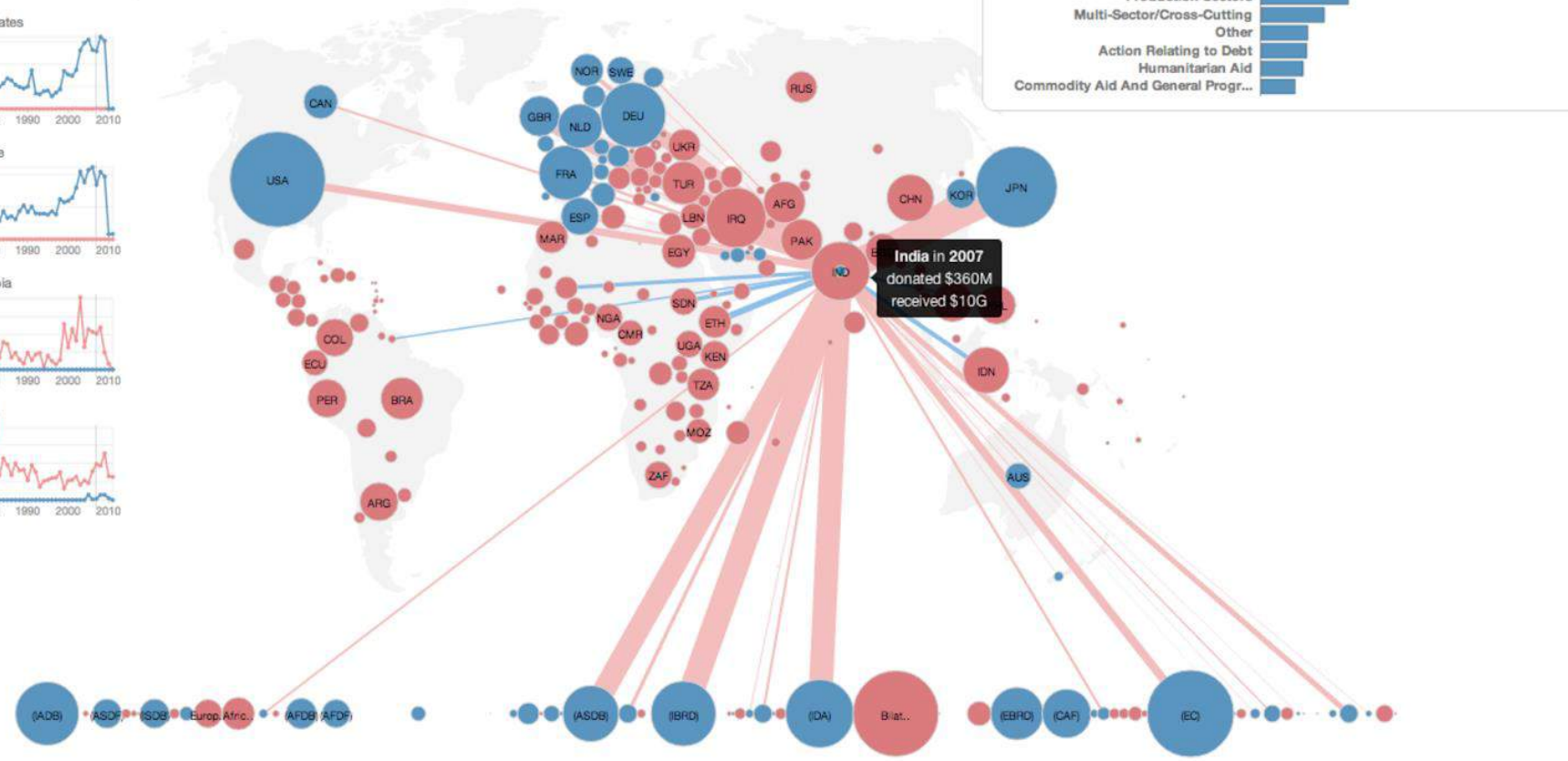
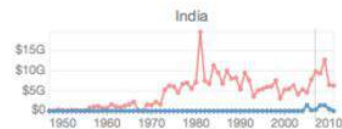
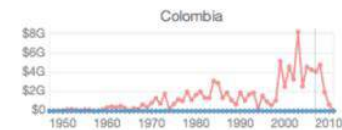
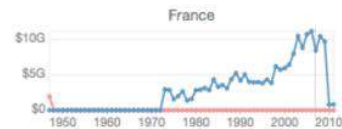
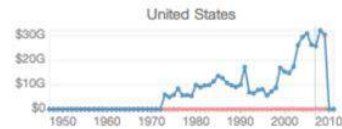


*vizster*



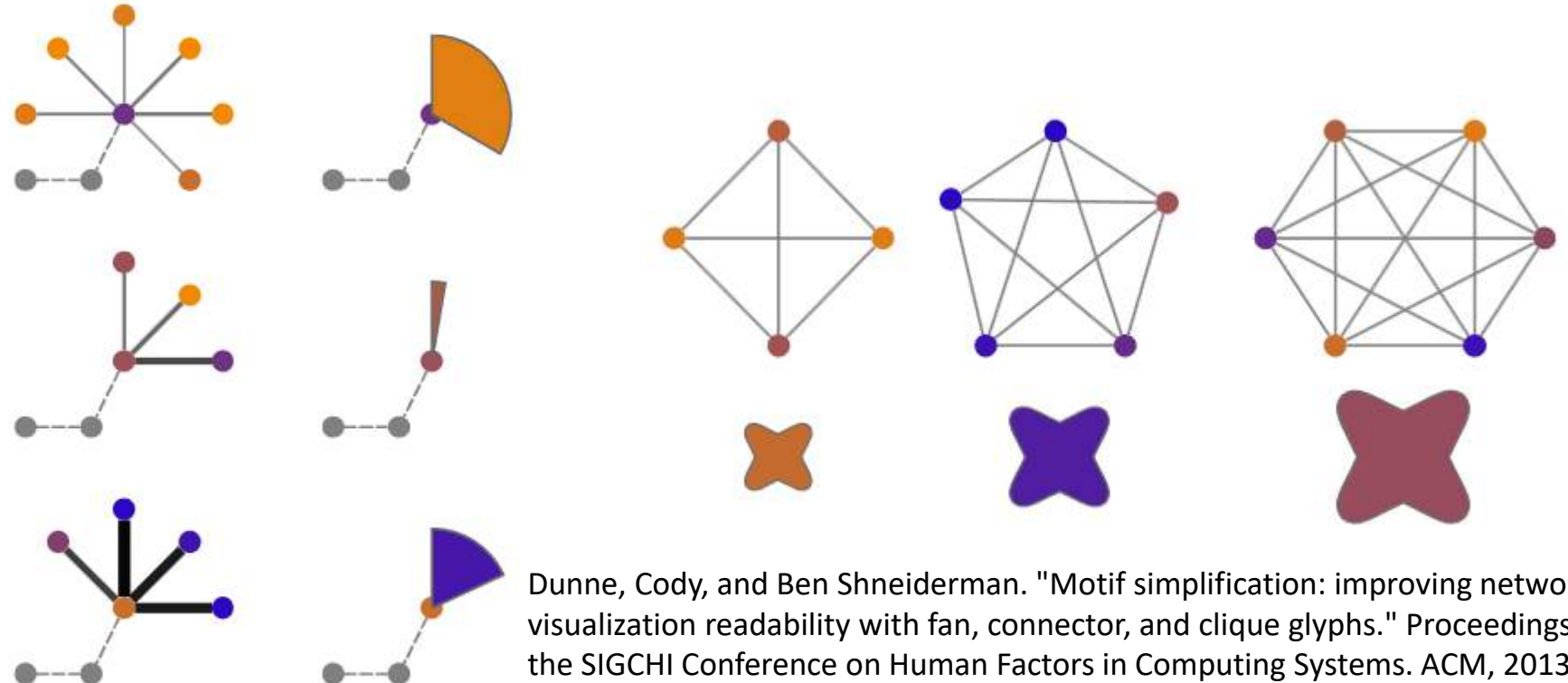
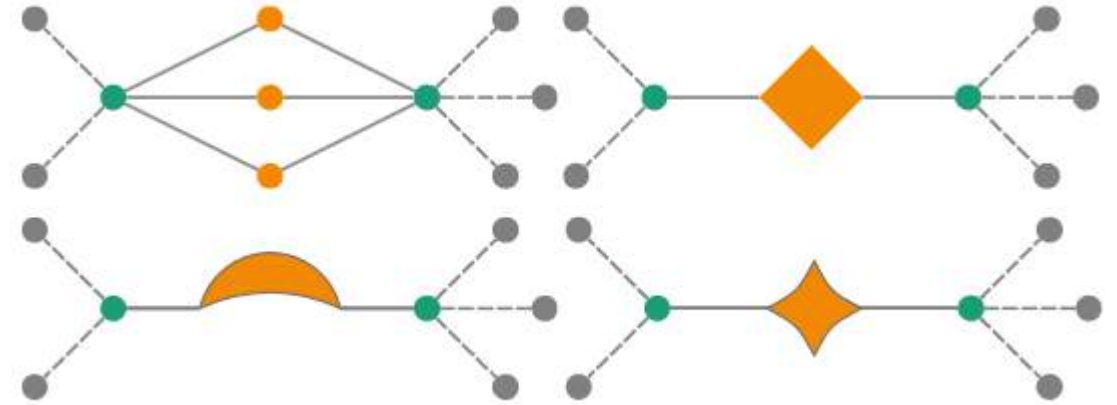
# Reducing clutter

- Edge bundling
- Clustering
- Edges on demand



# Reducing clutter

- Edge bundling
- Clustering
- Removing Edges
- Aggregation/Simplification

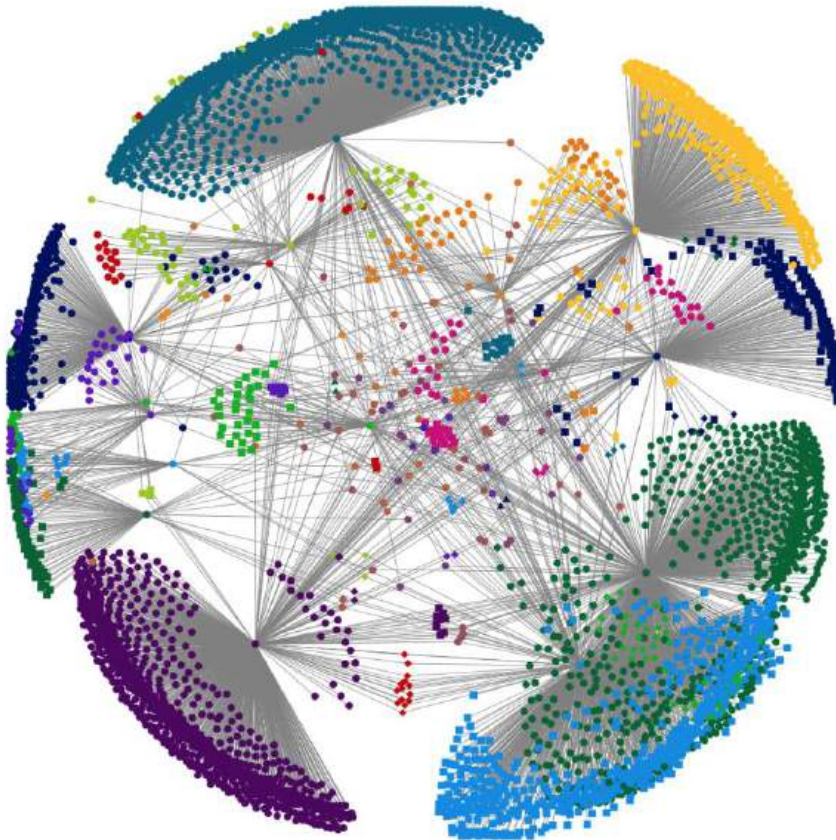


Dunne, Cody, and Ben Shneiderman. "Motif simplification: improving network visualization readability with fan, connector, and clique glyphs." Proceedings of the SIGCHI Conference on Human Factors in Computing Systems. ACM, 2013.

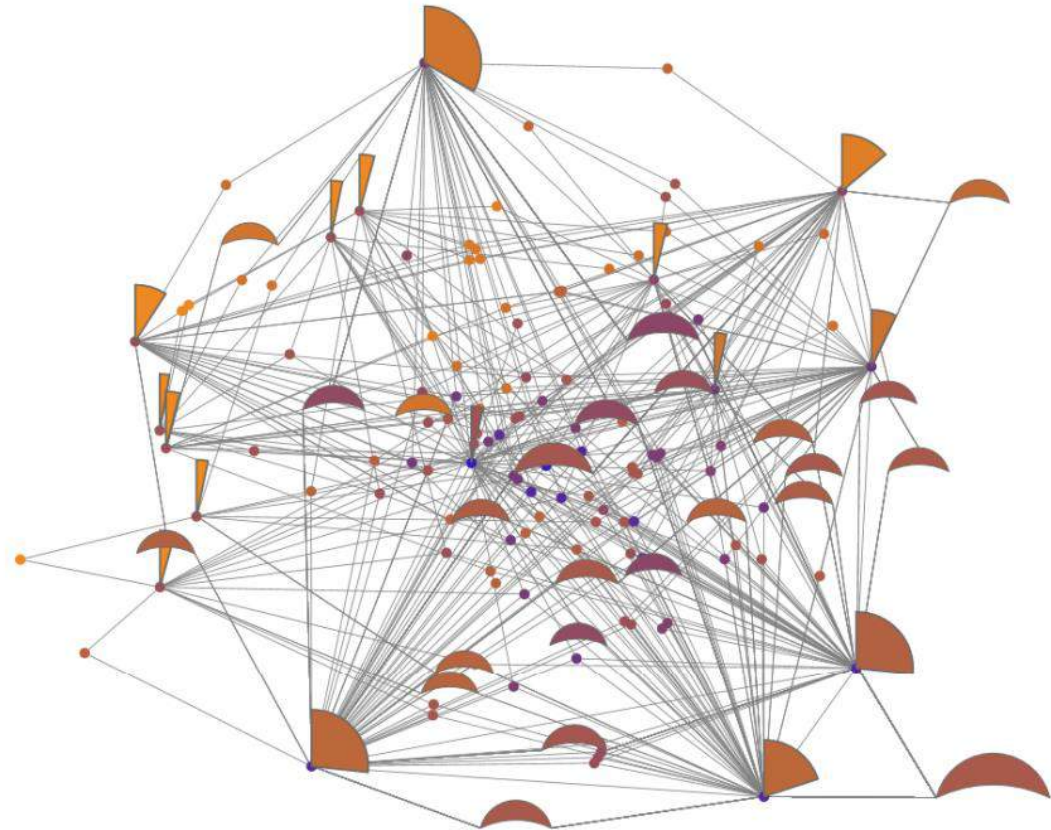


# Reducing clutter

- Edge bundling
- Clustering
- Removing Edges
- Aggregation/Simplification

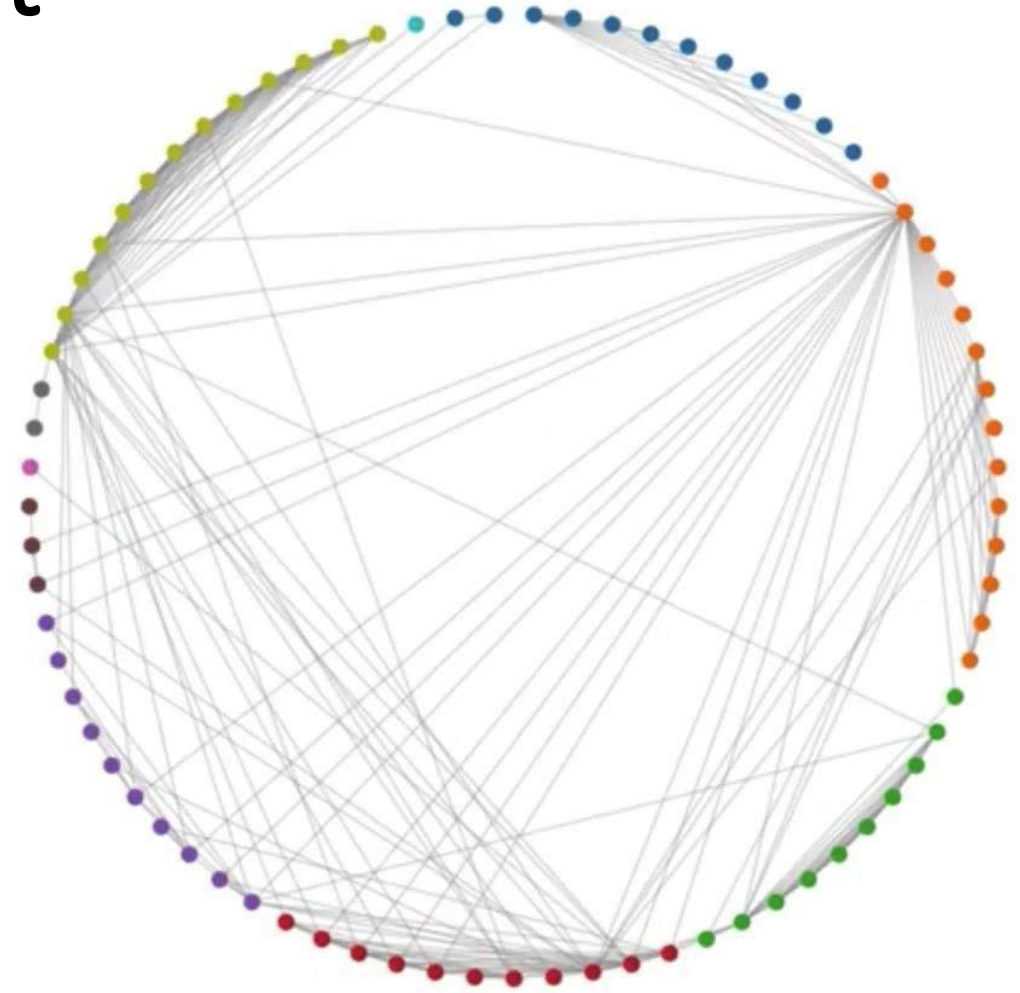
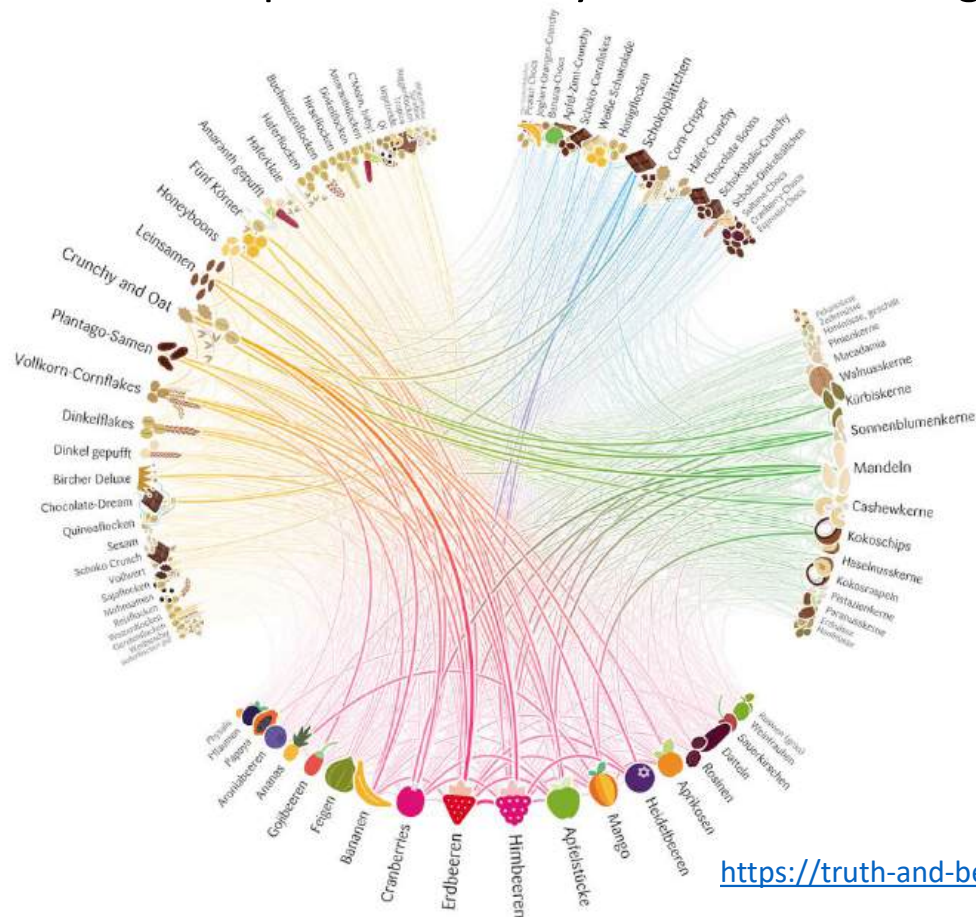


Dunne, Cody, and Ben Shneiderman. "Motif simplification: improving network visualization readability with fan, connector, and clique glyphs." Proceedings of the SIGCHI Conference on Human Factors in Computing Systems. ACM, 2013.



# Force-directed vs fixed layout

- Force-directed:
  - Reveals the **structure** of the network
- Fixed layout:
  - Improves visibility of nodes and edges





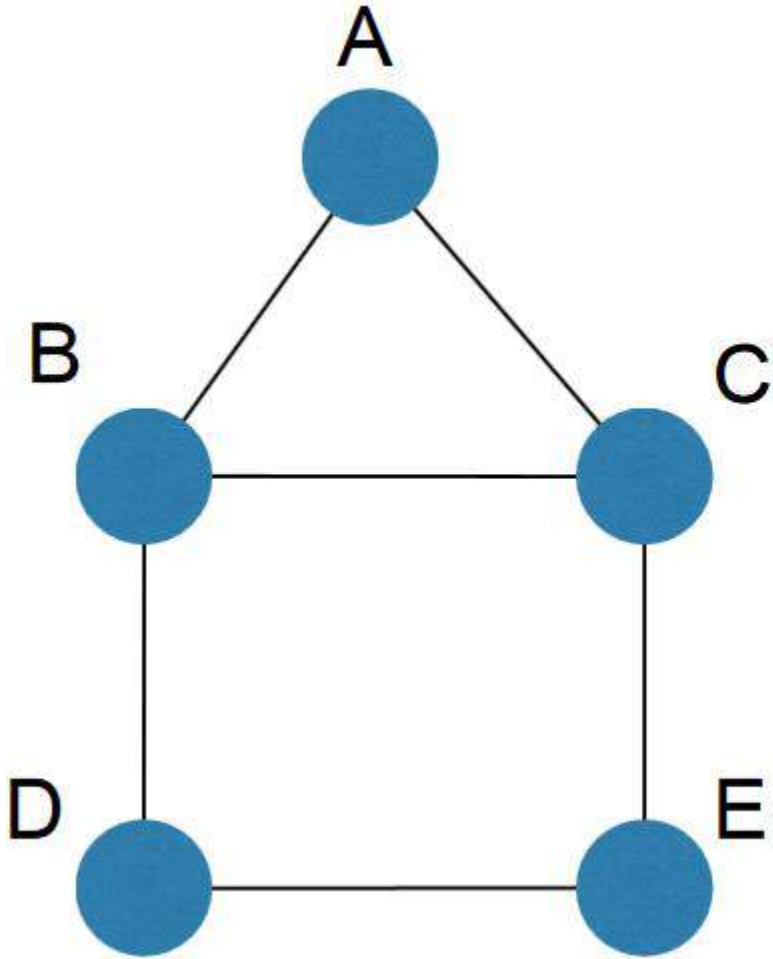


# Matrices



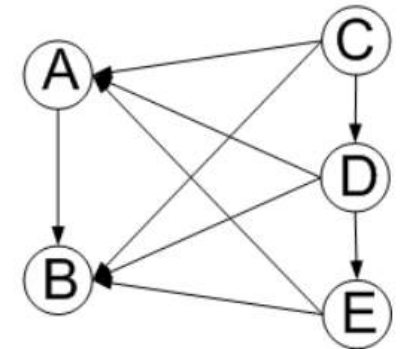
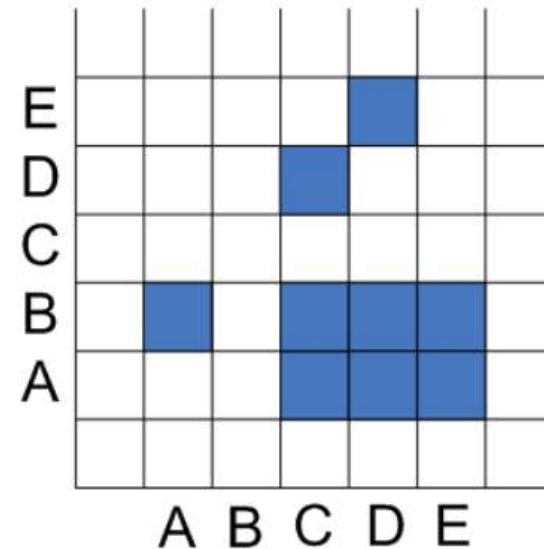
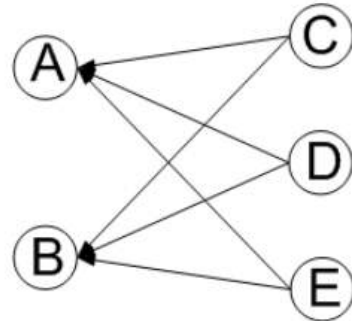
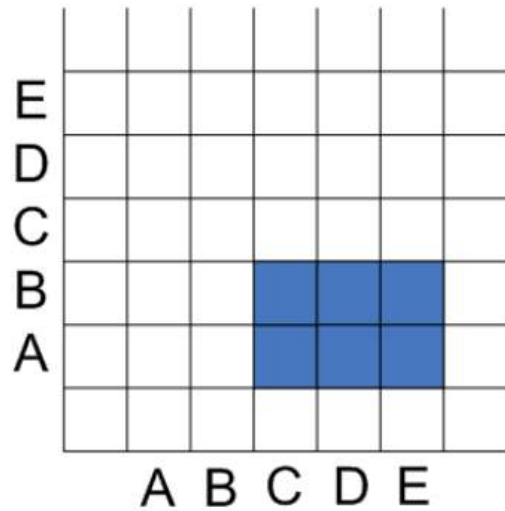
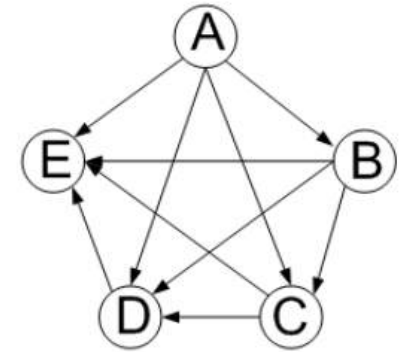
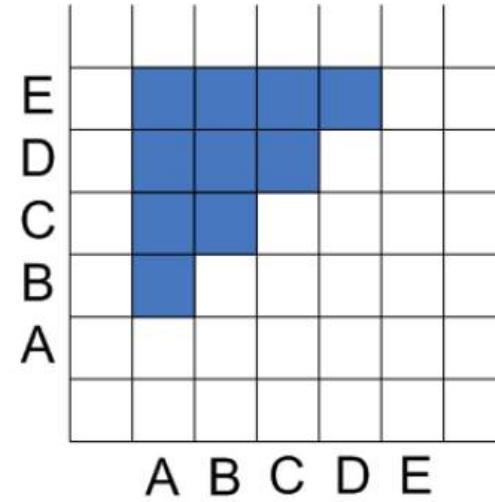
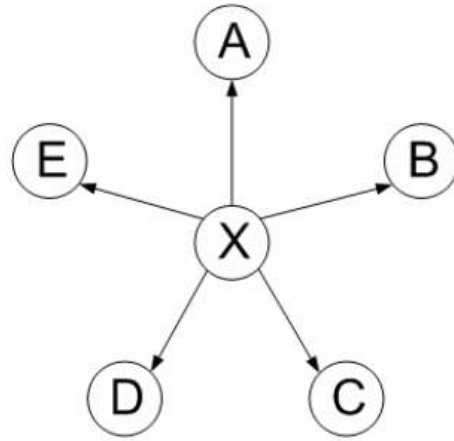
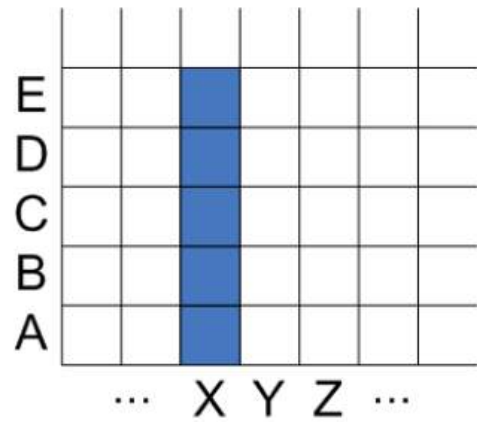
# Adjacency matrix representations

- Derive adjacency matrix from network connections
  - Basically turning it into a table

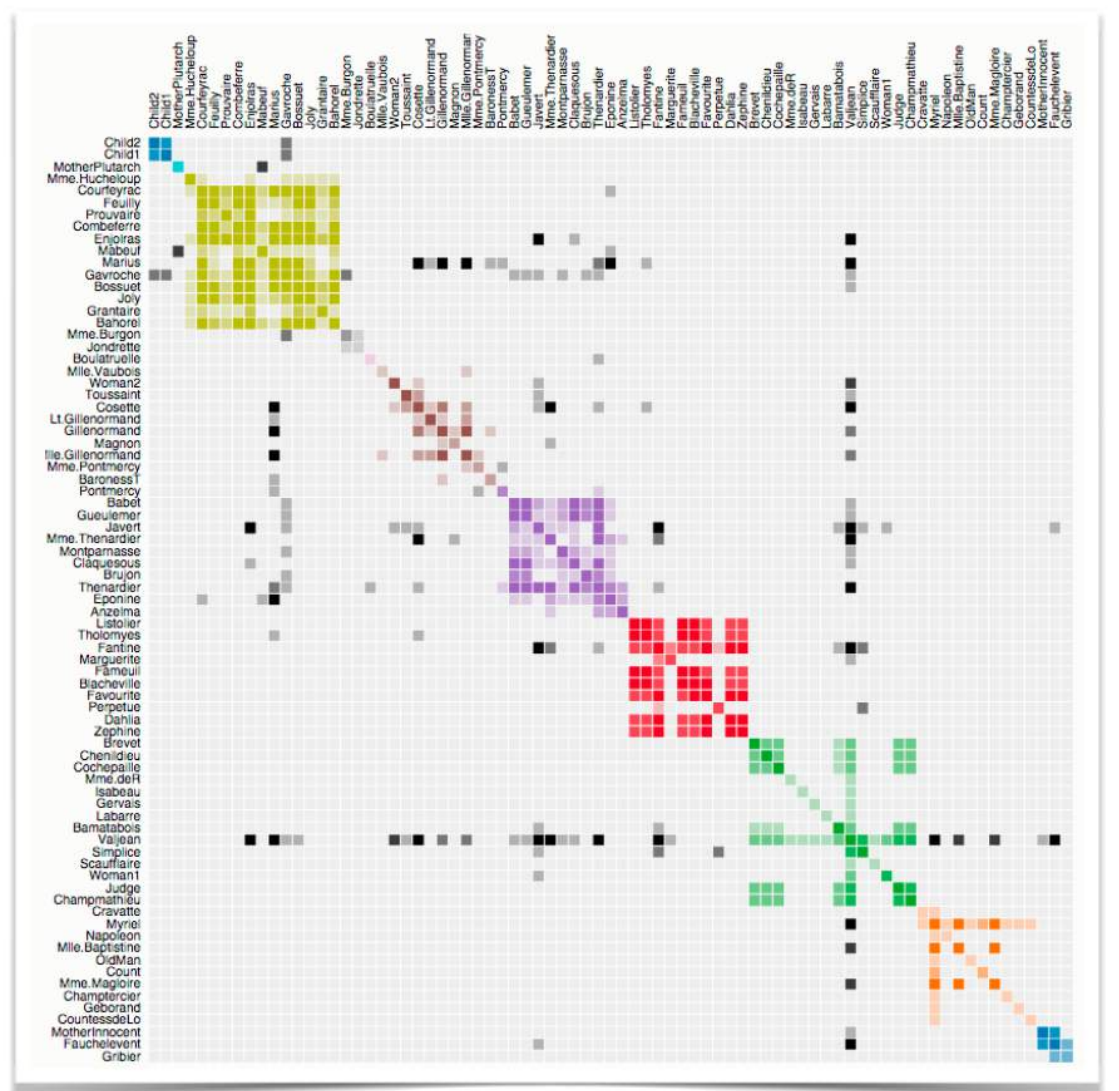


	A	B	C	D	E
A					
B					
C					
D					
E					

# Adjacency matrix examples



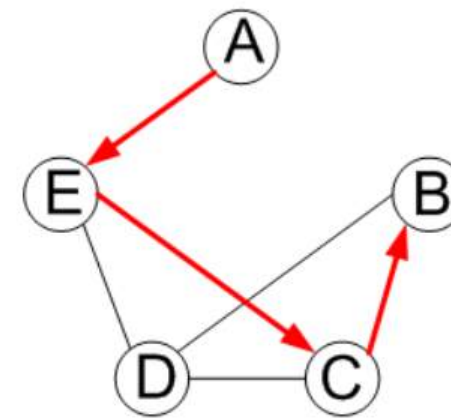
- Just like circular/arc node-link diagrams



# Adjacency matrix pros and cons

		TO							
		A	B	C	D	E	F	G	H
FROM	A								
	B								
	C								
	D								
	E								
	F								
	G								
	H								

Good for neighborhood tasks  
(node 1-hop neighbors)



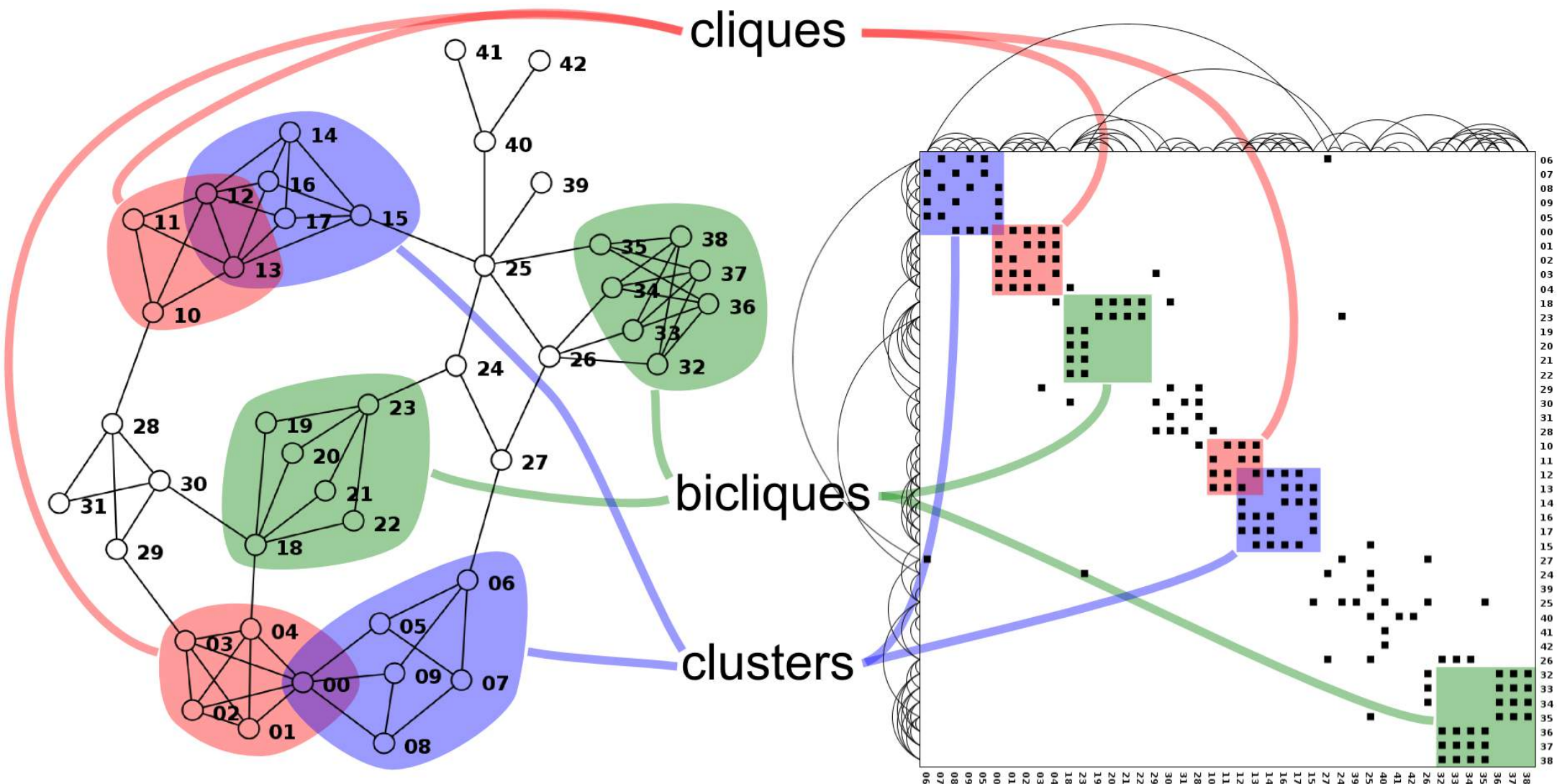
	A	B	C	D	E
E					
D					
C					
B					
A					

Bad for tasks related to paths



# Structure in both

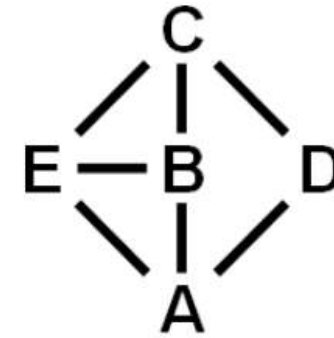
- Typically need to be taught to recognize



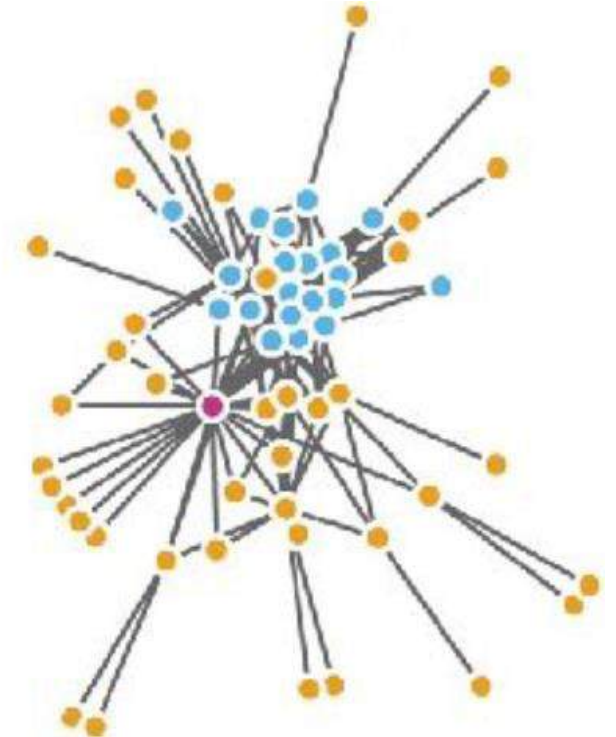
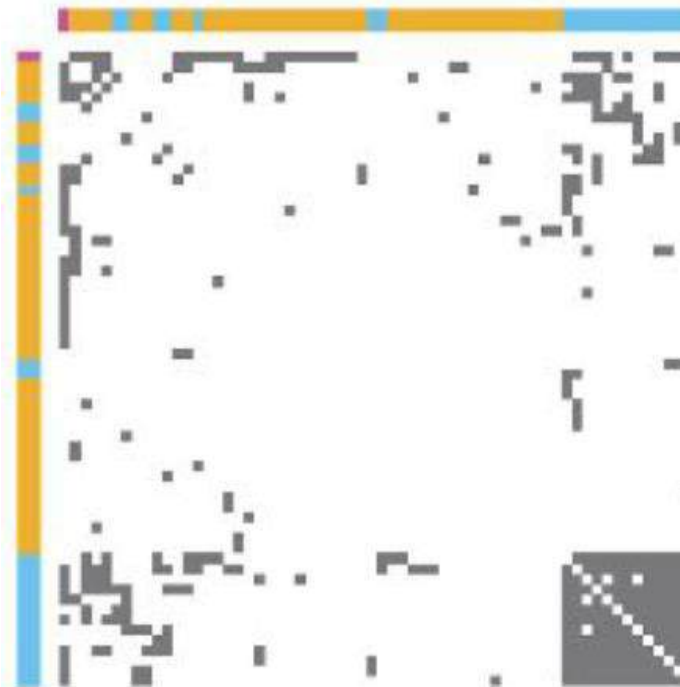
# Adjacency matrix

- Data: network
- Derived data: table from network
  - 1 quant. attribute
    - Weighted edge between nodes
  - 2 categorical attributes: node list 2x
- Visual encoding:
  - Cell shows presence/absence of edge
- Scalability
  - 1000 nodes, 1M edges (no clutter!)

	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

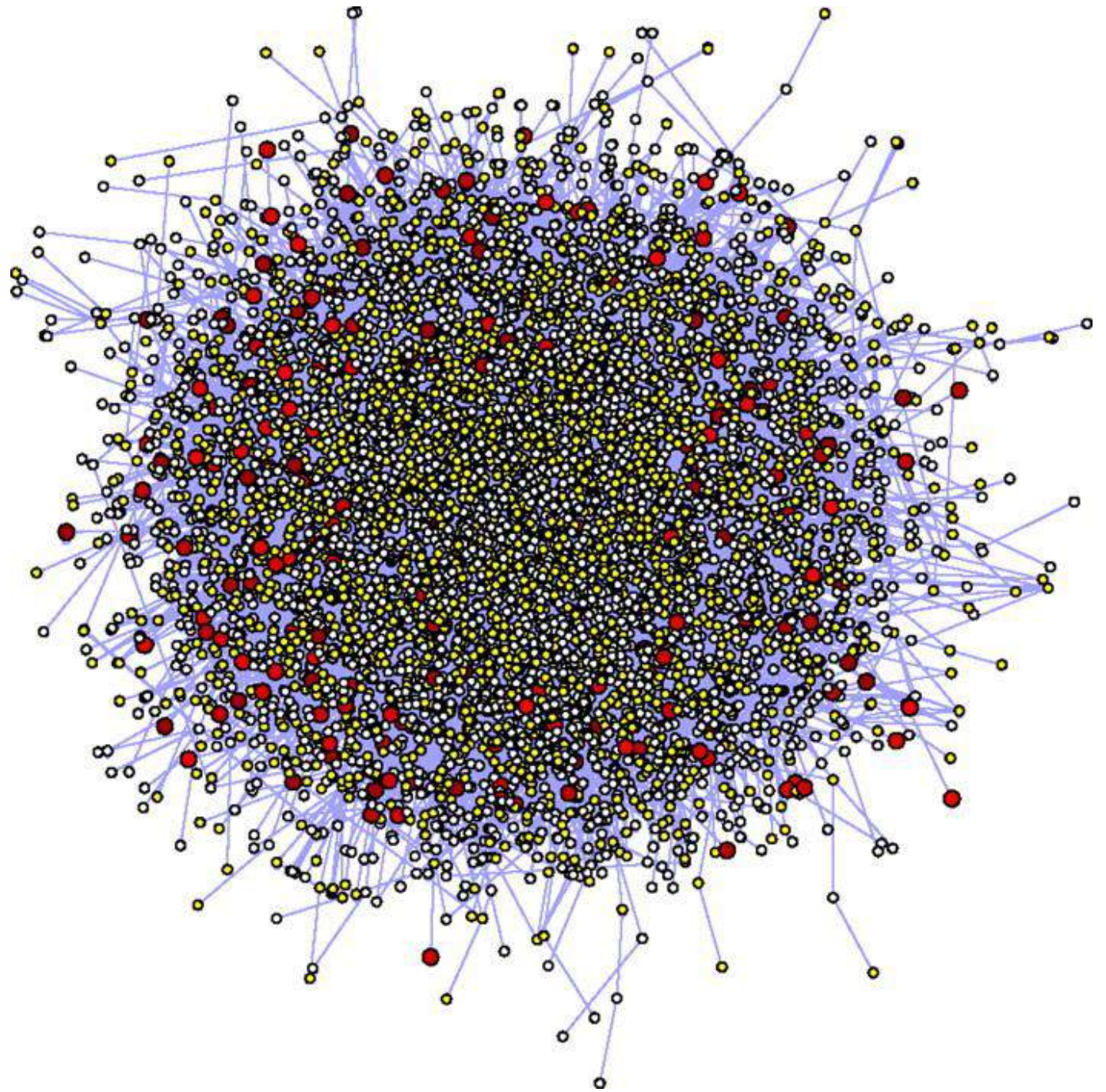


Points of view: Networks. Gehlenborg and Wong. Nature Methods 9:115



# Adjacency matrix

- Data: network
- Derived data: table from network
  - 1 quant. attribute
    - Weighted edge between nodes
  - 2 categorical attributes: node list 2x
- Visual encoding:
  - Cell shows presence/absence of edge
- Scalability
  - 1000 nodes, 1M edges (no clutter!)
- Avoid hairball effect!

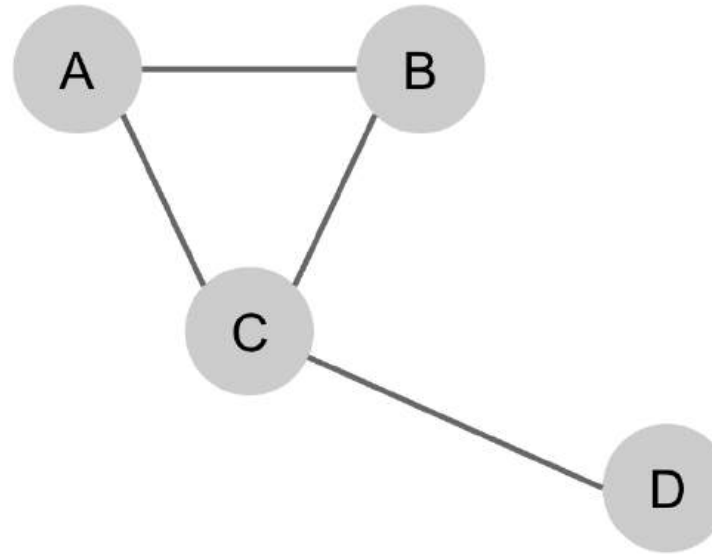




# Adjacency matrix

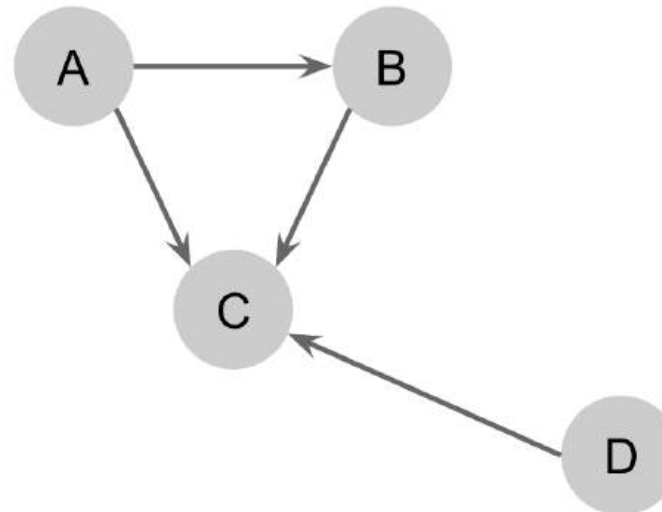
- Can also show directed graphs

Undirected  
Network



	A	B	C	D
A				
B				
C				
D				

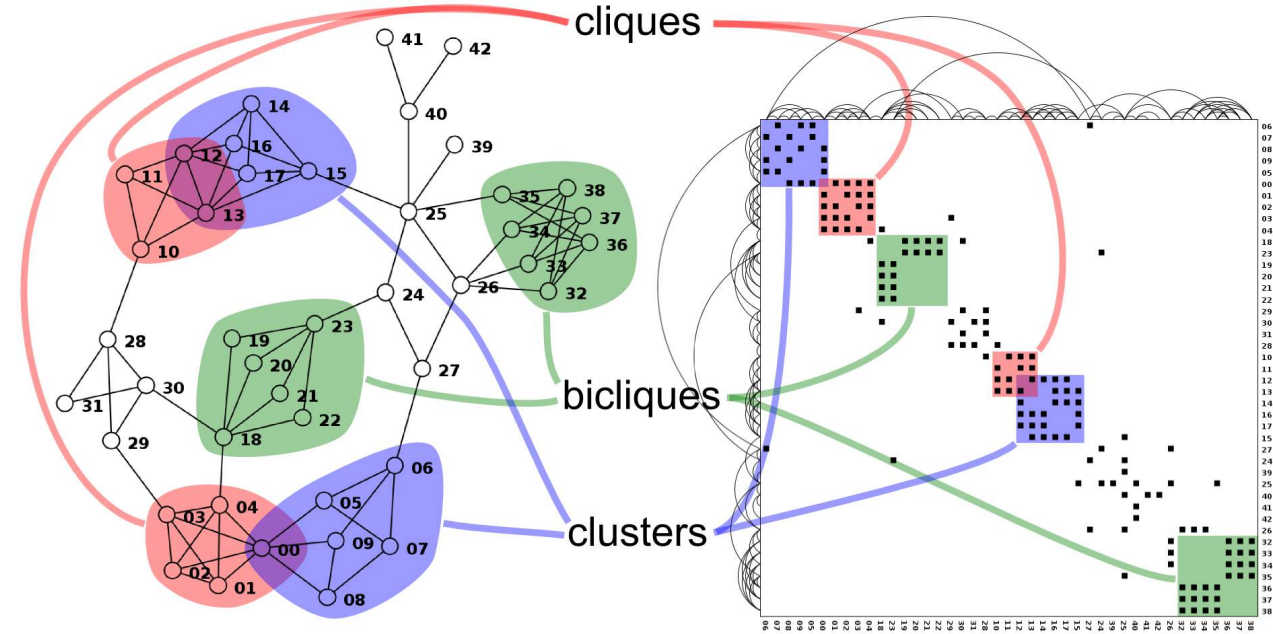
Directed  
Network



	A	B	C	D
A				
B				
C				
D				

# Node-link vs matrix comparison

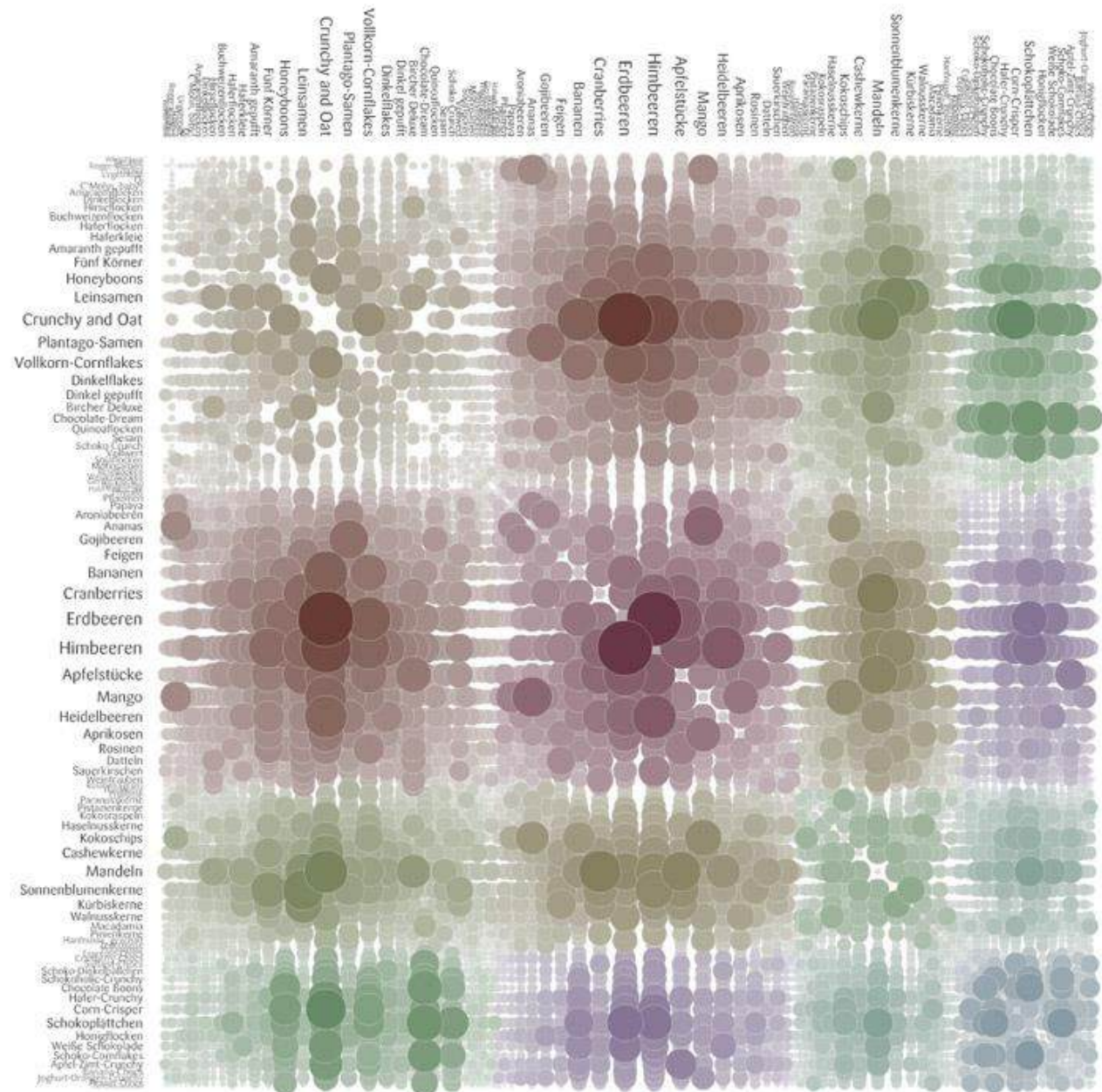
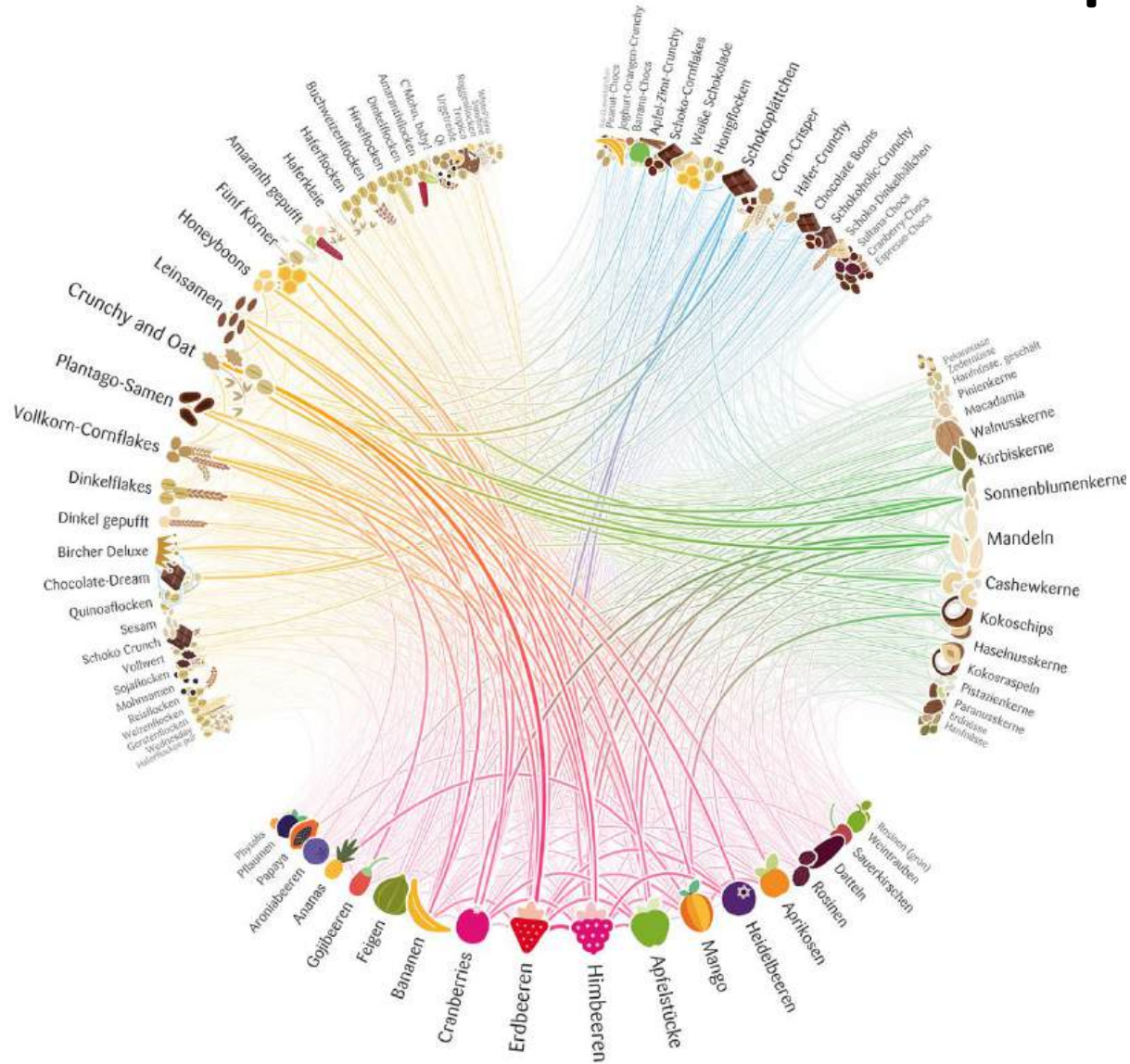
- Node-link diagram strengths
  - Topology understanding, path tracing
  - Intuitive, flexible, no training needed
- Adjacency matrix strengths
  - Focus on edges rather than nodes
  - Layout is straightforward (reordering needed)
  - Predictable space requirements, scalability
  - Some topology tasks are easy with training
- Empirical study:
  - Node-link best for small networks
  - Matrix best for large networks
    - As long as no 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



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

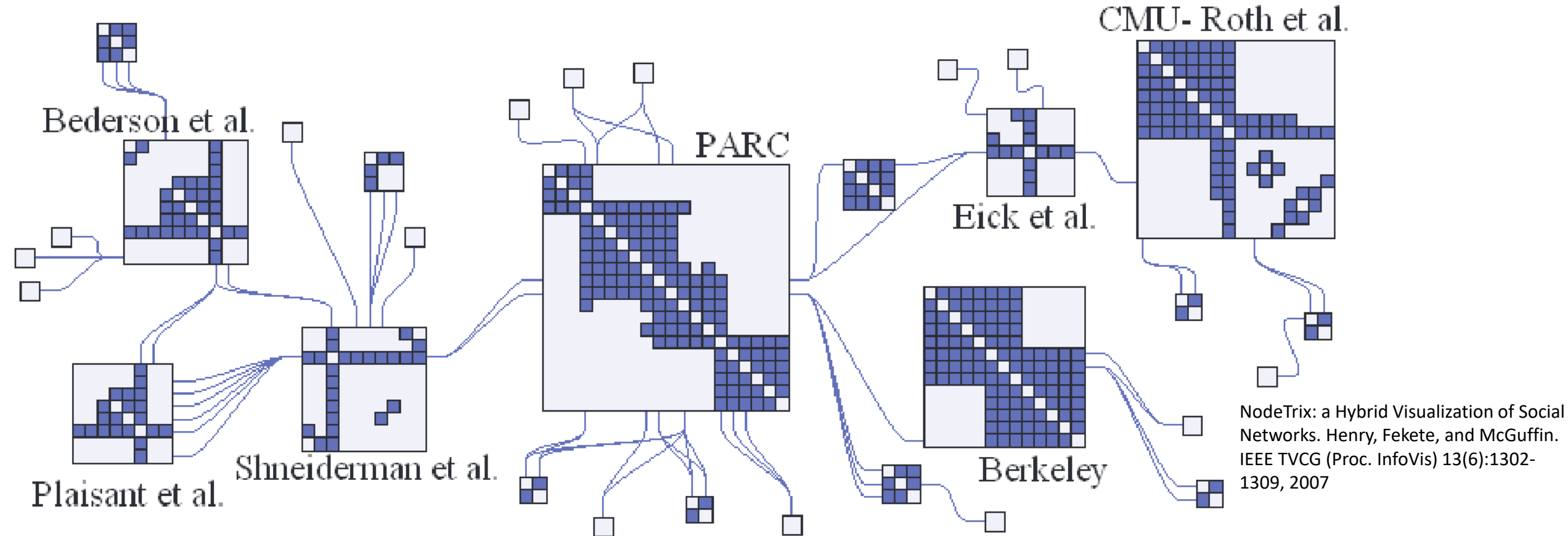


# Node-link vs matrix comparison



# NodeTrix

- What if I need path tracing for a large network?
- NodeTrix: hybrid node-link/matrix
- Captures strengths of both
- <https://www.youtube.com/watch?v=7G3MxyOcHKQ>



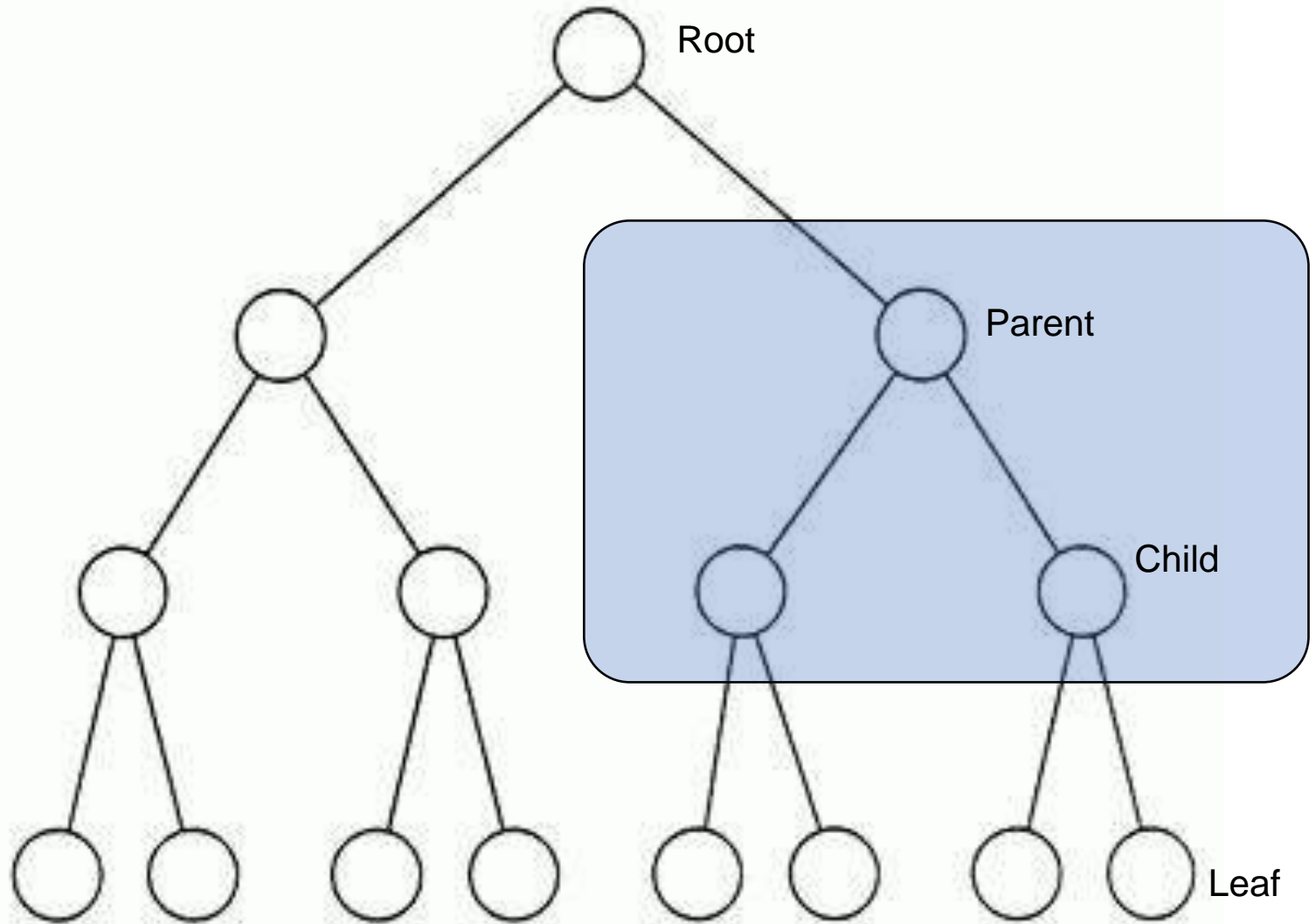




# Trees

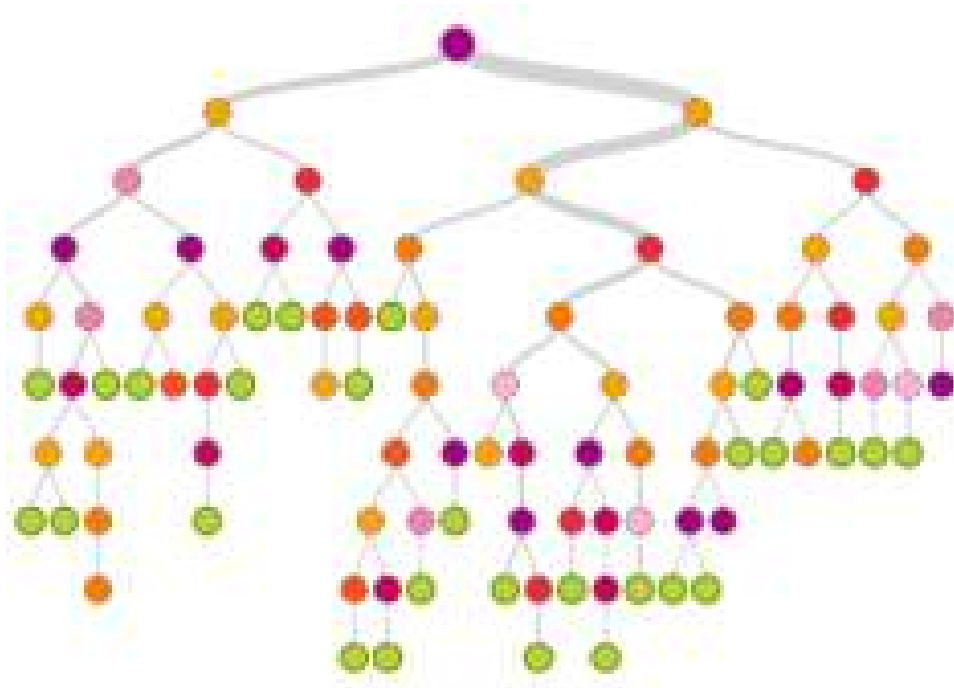
# Node-link trees

- Applications of trees:
  - File structure
  - Evolutionary tree
  - Language structure
  - More?



# Visualizing trees

## Node-link

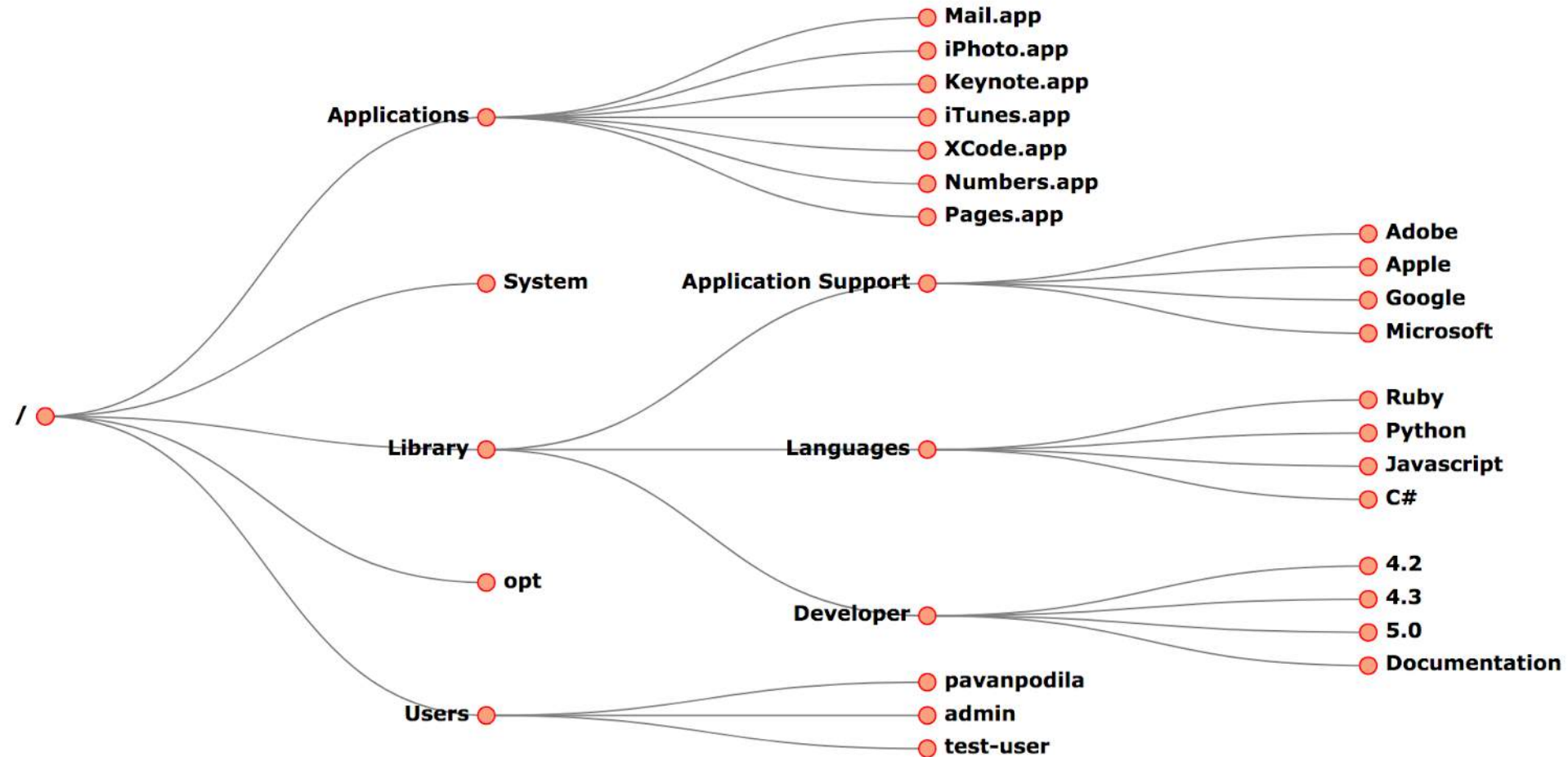


## Containment

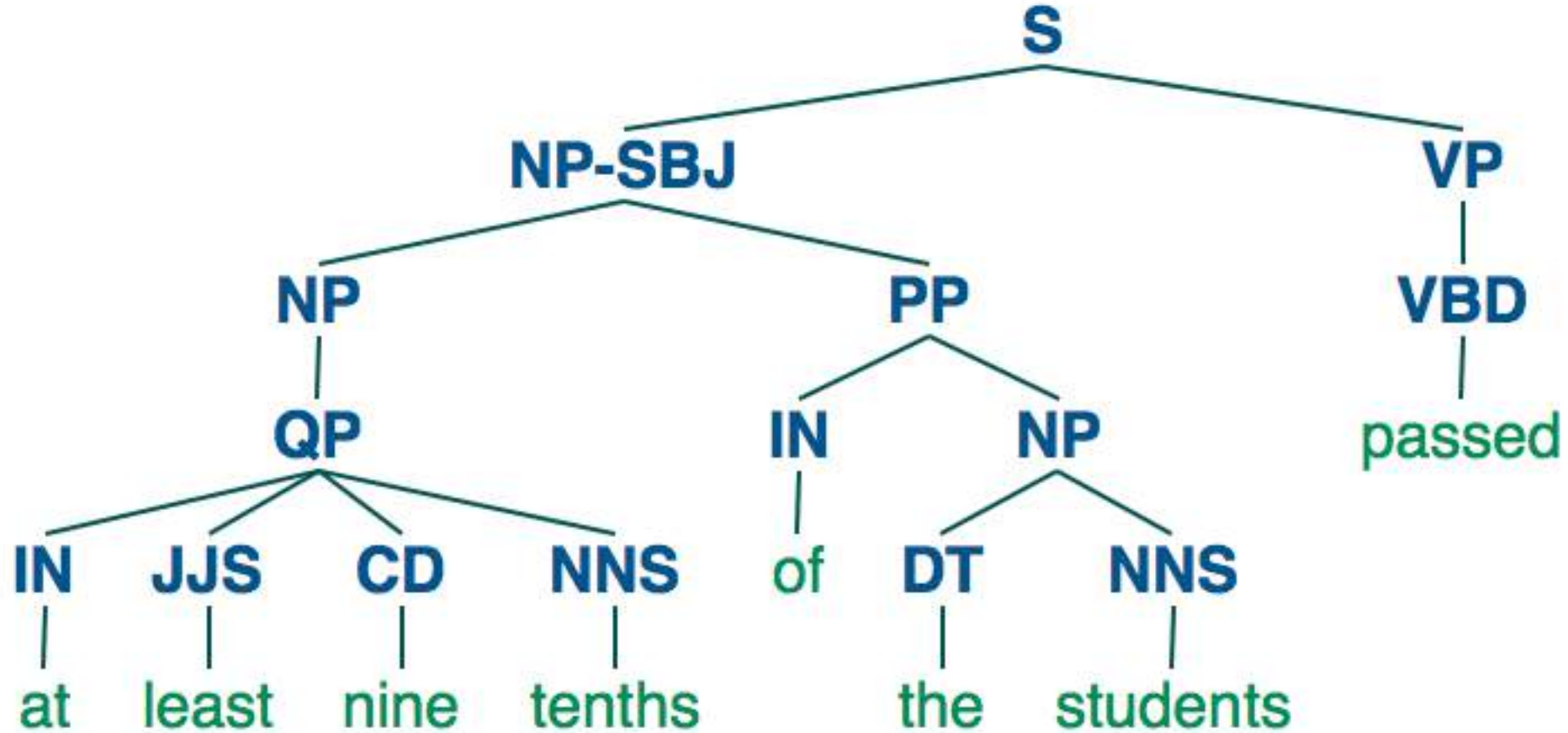




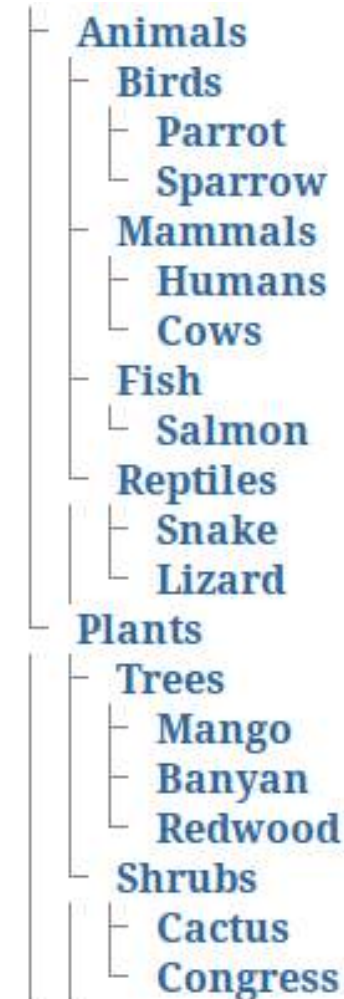
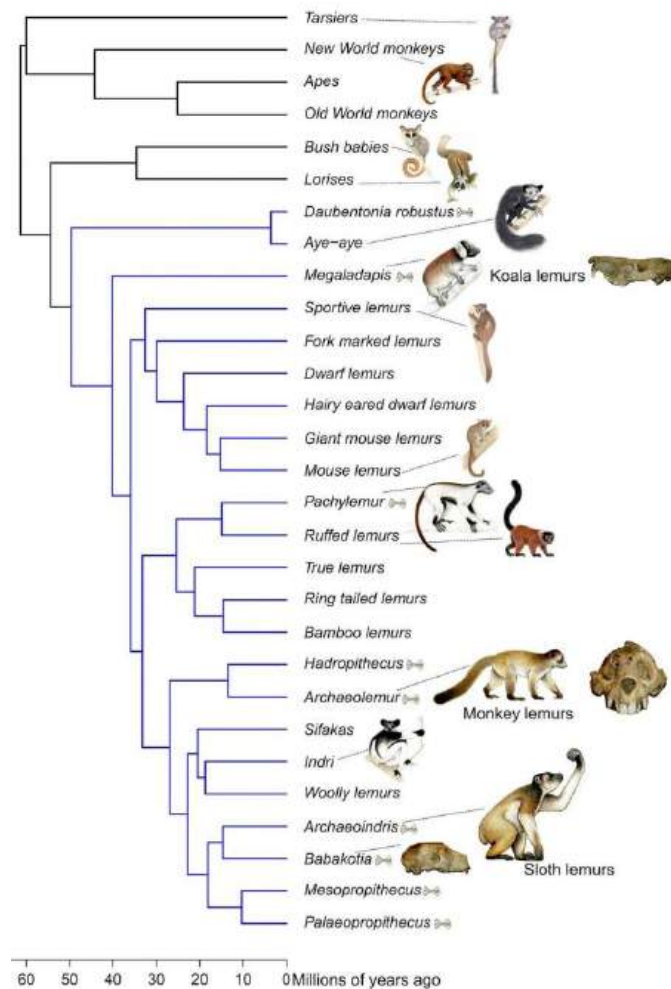
# Node-link trees: file system



# Node-link trees: language parser

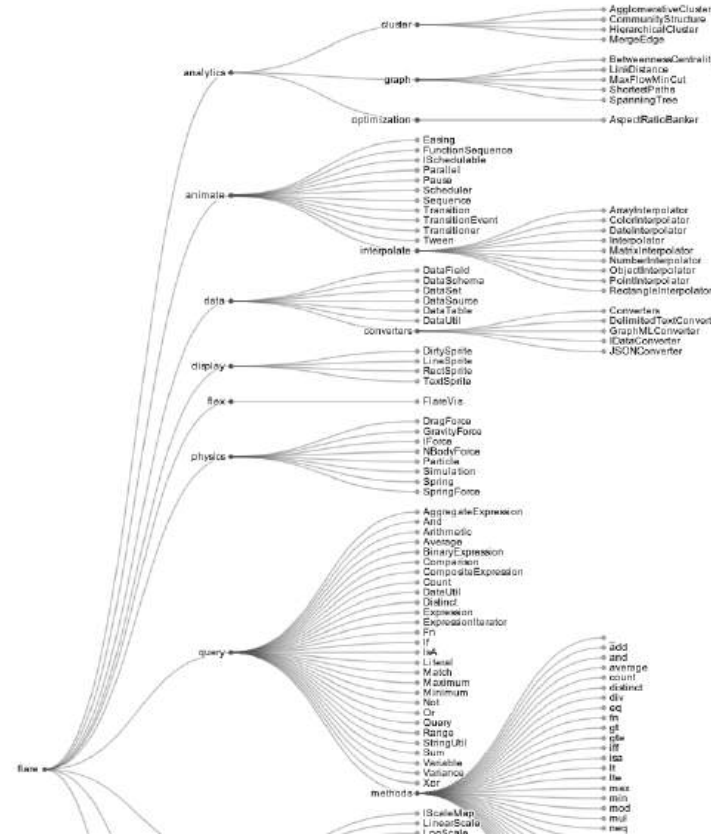


# Node-link trees: species ancestry

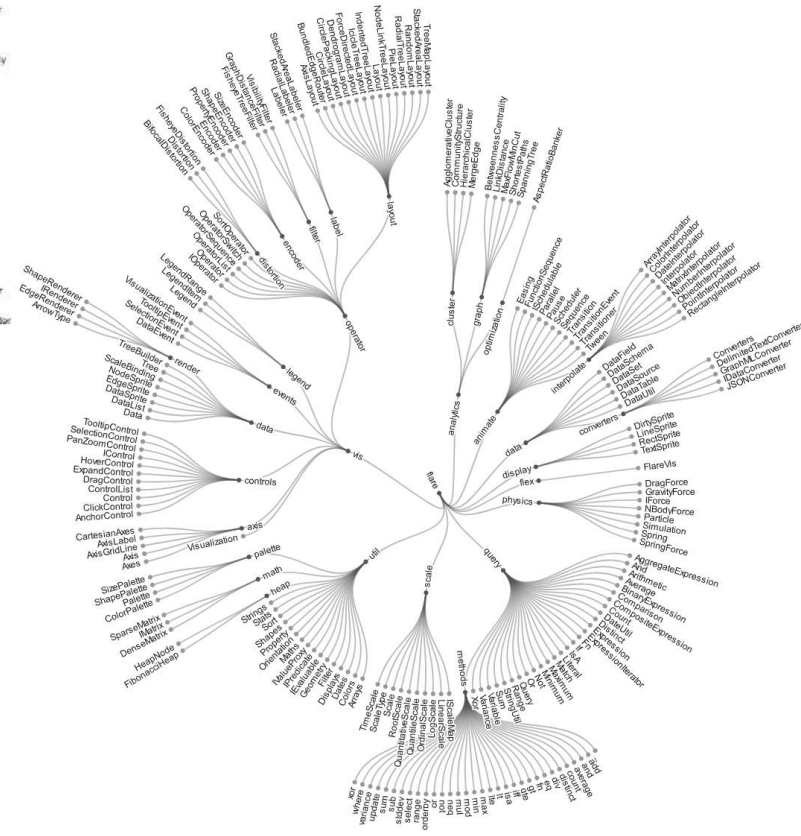


# Node-link trees

- Special type of network
- Reingold-Tilford algorithm
  - Tidy drawings of trees
    - Exploit parent-child structure
  - Allocate space: compact but without overlap
    - Rectilinear and radial variants
- Algorithm writeup:  
<https://llimlib.github.io/pymag-trees/>



<https://observablehq.com/@d3/tree/2>

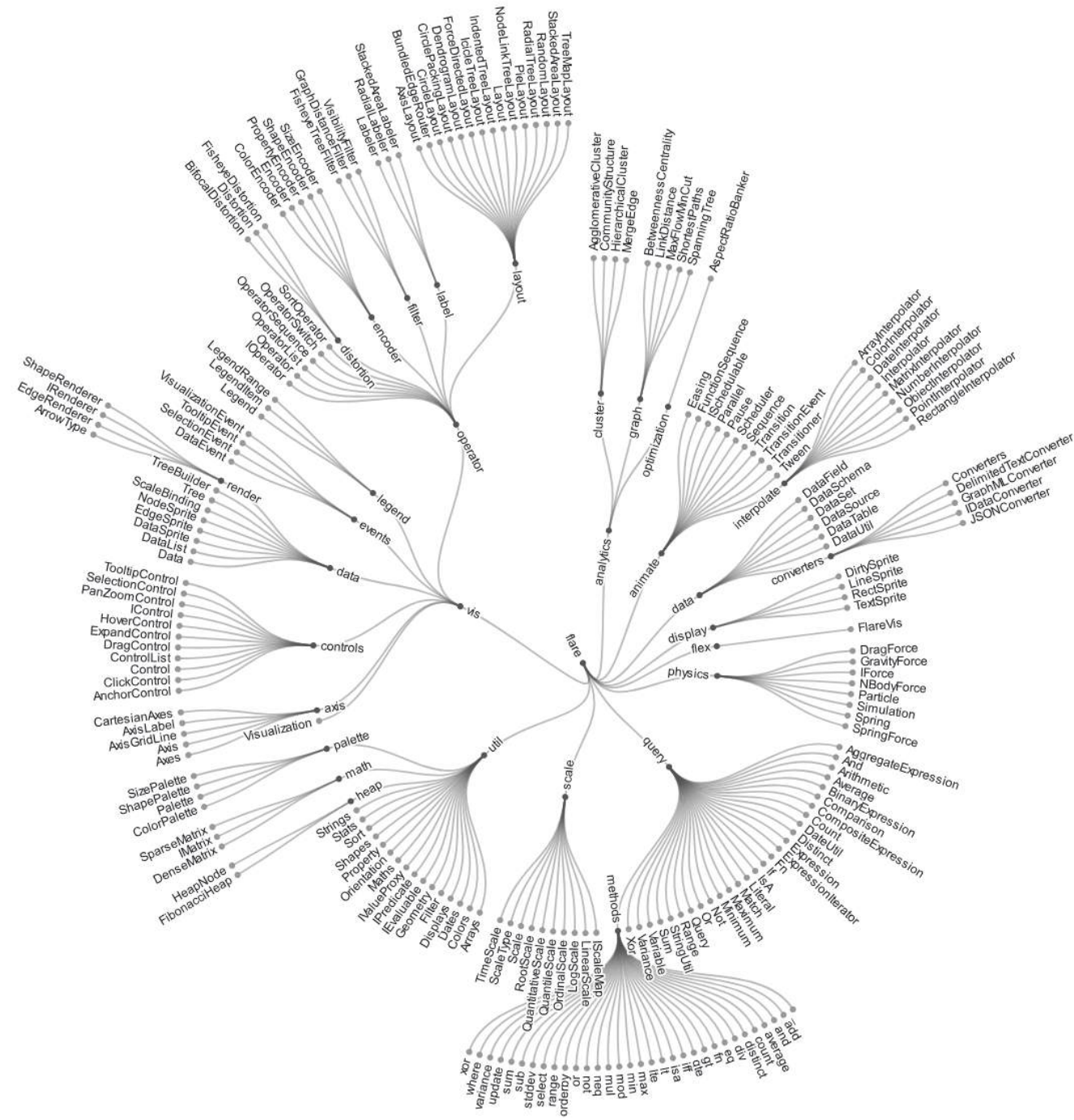


<https://observablehq.com/@d3/radial-tree/2>



# Radial node-link trees

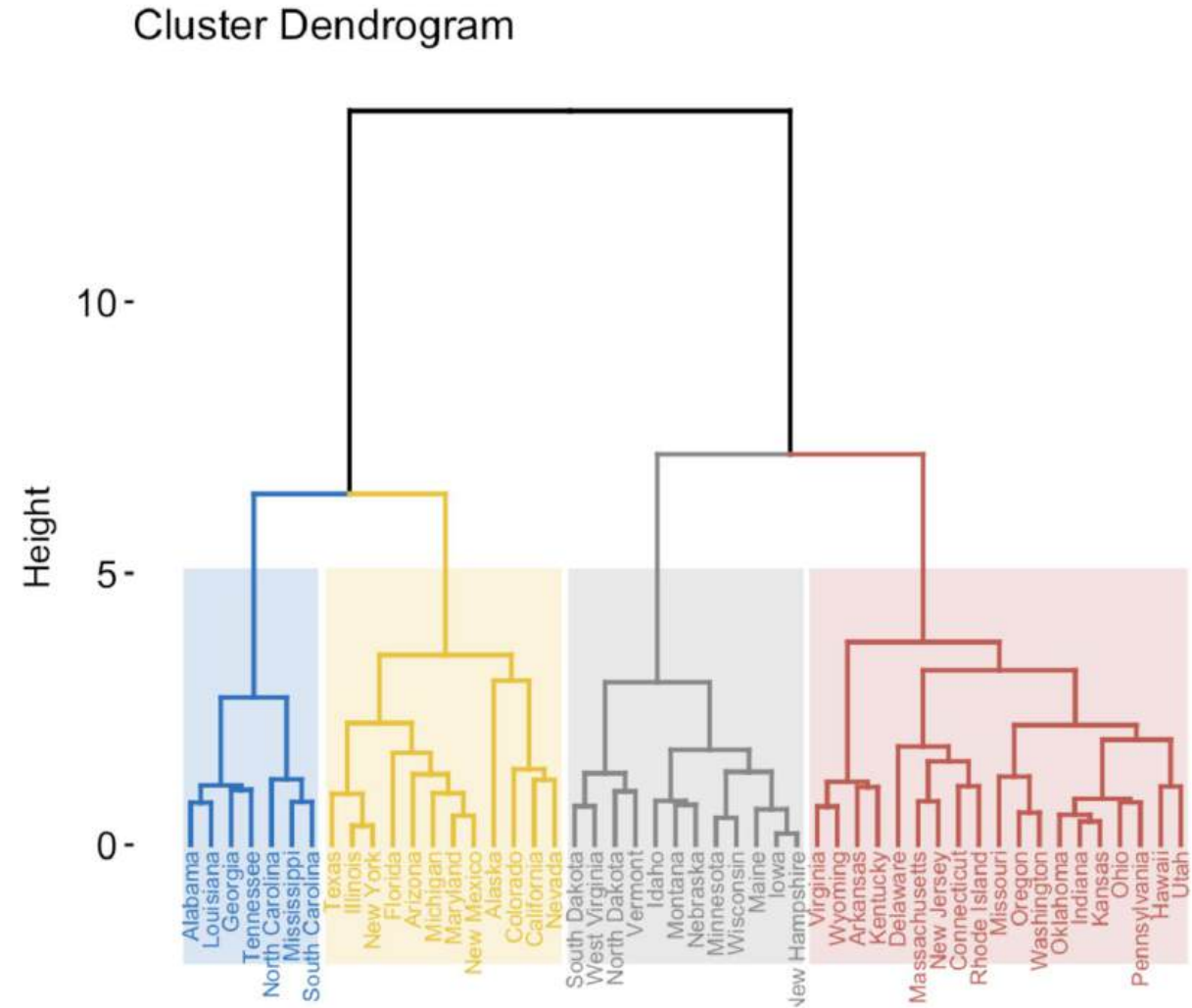
- Data:
  - Tree
- Encoding:
  - Point node marks
  - Link connection marks
  - Radial axis orientation
    - Angular proximity → siblings
    - Distance from center → depth in tree
- Tasks:
  - Understand topology, follow paths
- Scalability:
  - 1k – 10k nodes (with/without labels)



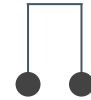
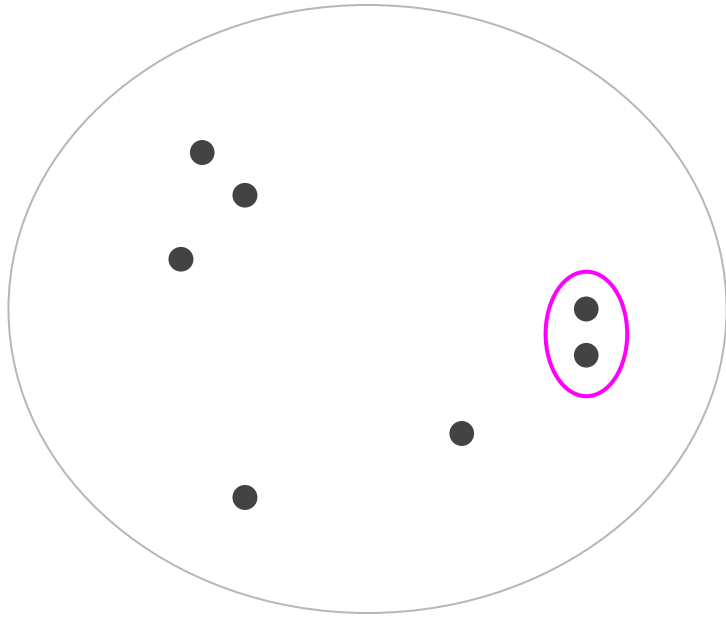


# Dendrogram

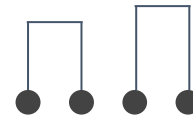
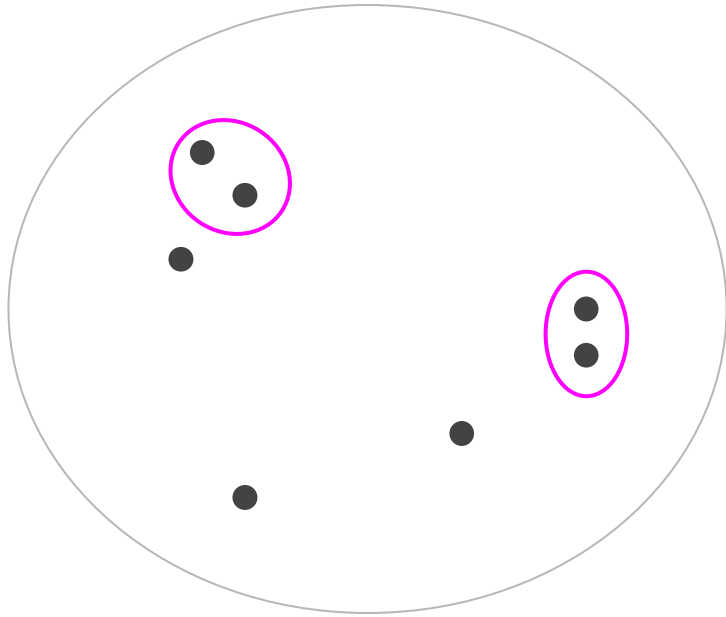
- Represents **hierarchical clusters**
  - Data processing algorithm to organize objects into a hierarchy according to a similarity metric defined between the objects.
- Binary tree!



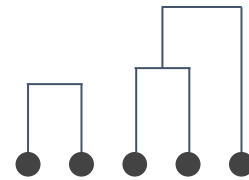
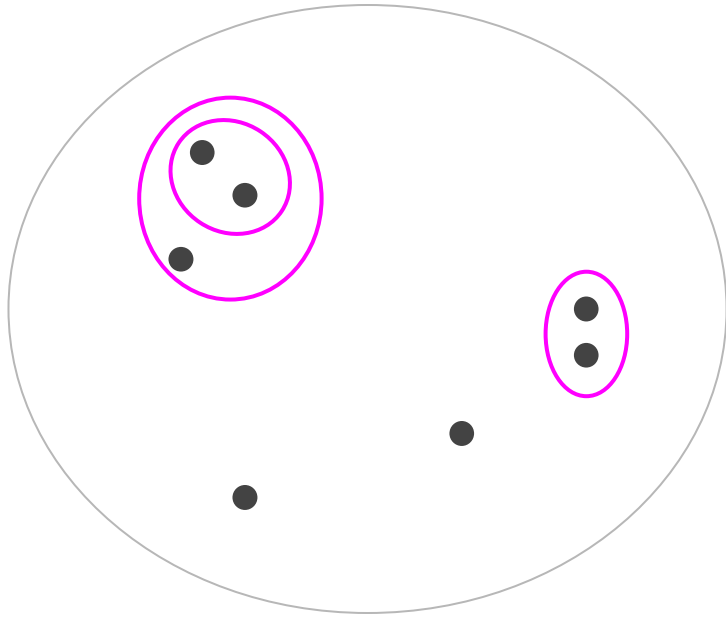
# Hierarchical Clustering



# Hierarchical Clustering

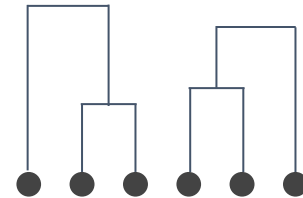
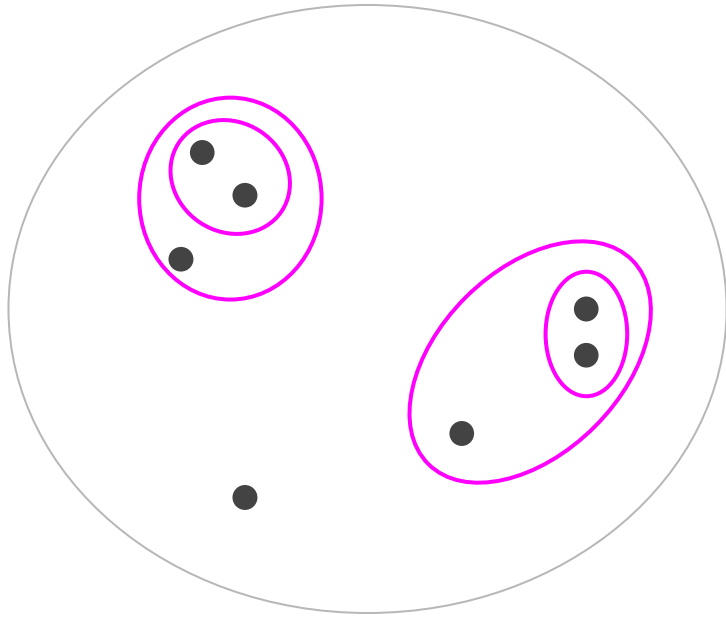


# Hierarchical Clustering

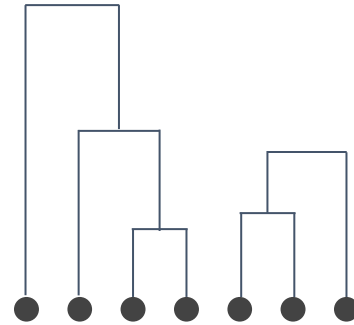
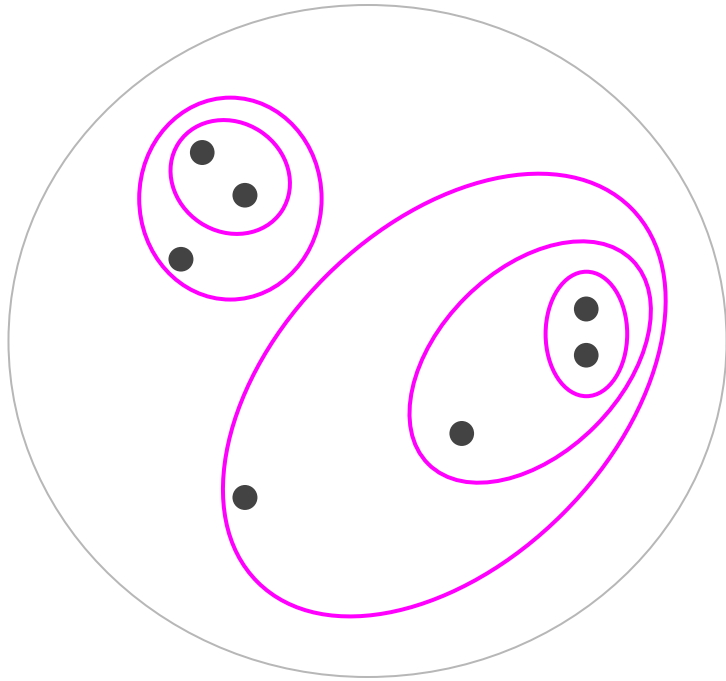




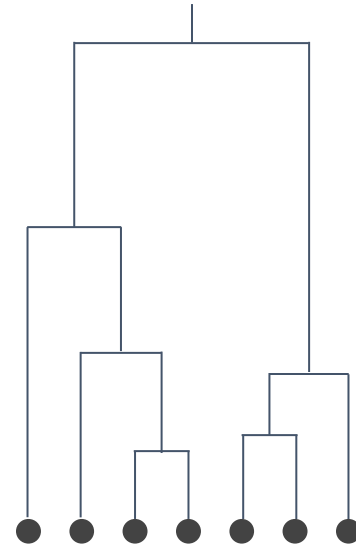
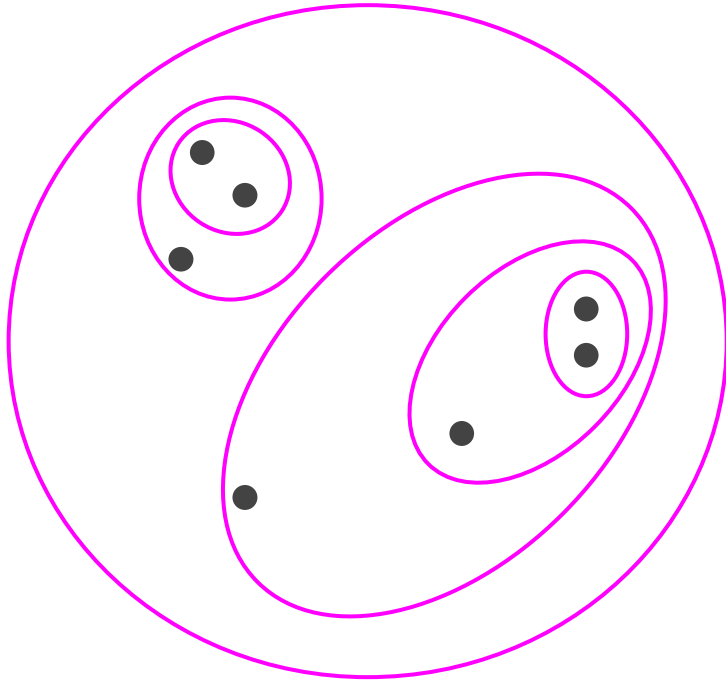
# Hierarchical Clustering



# Hierarchical Clustering

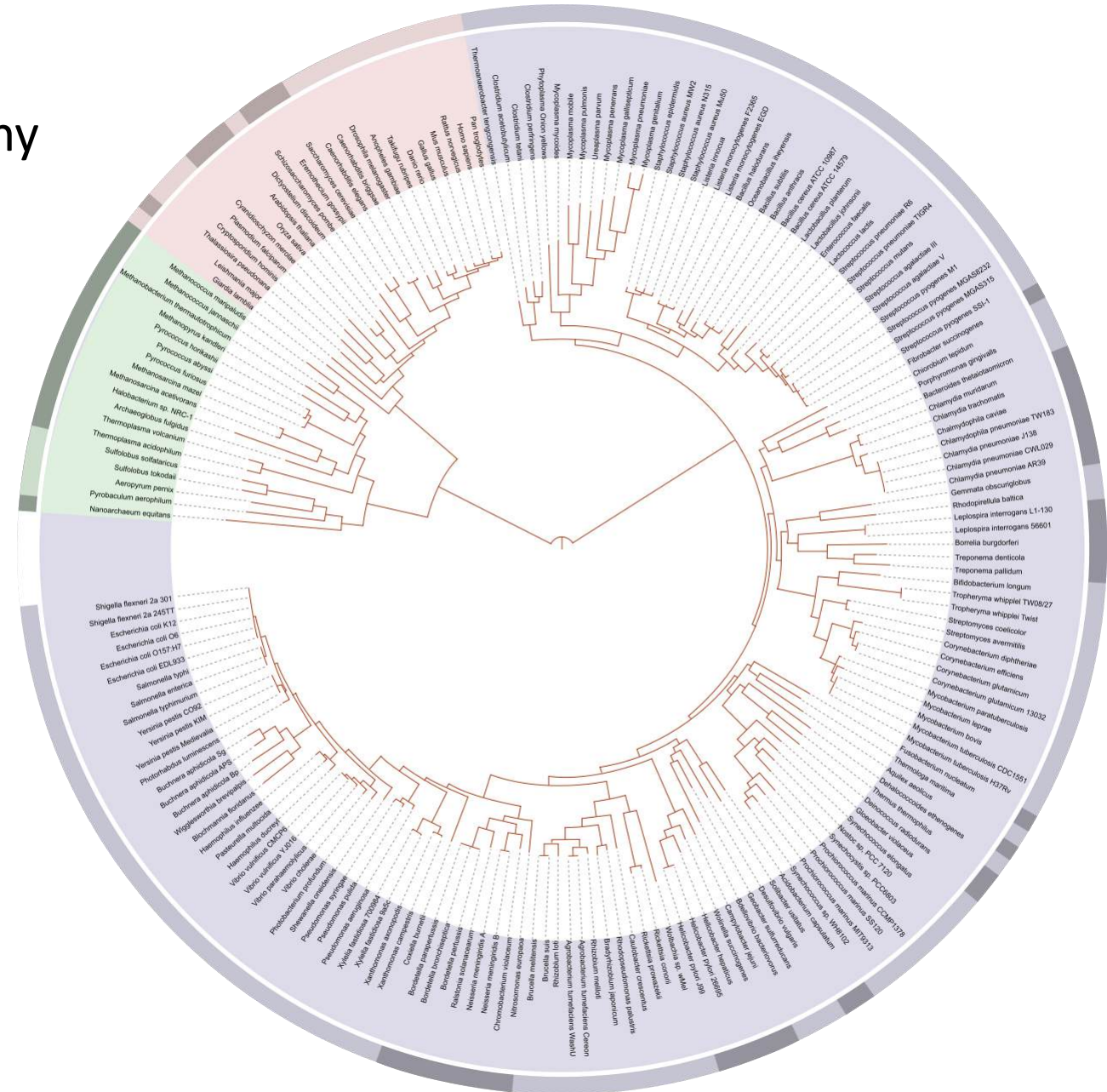


# Hierarchical Clustering



# Hierarchical Clustering

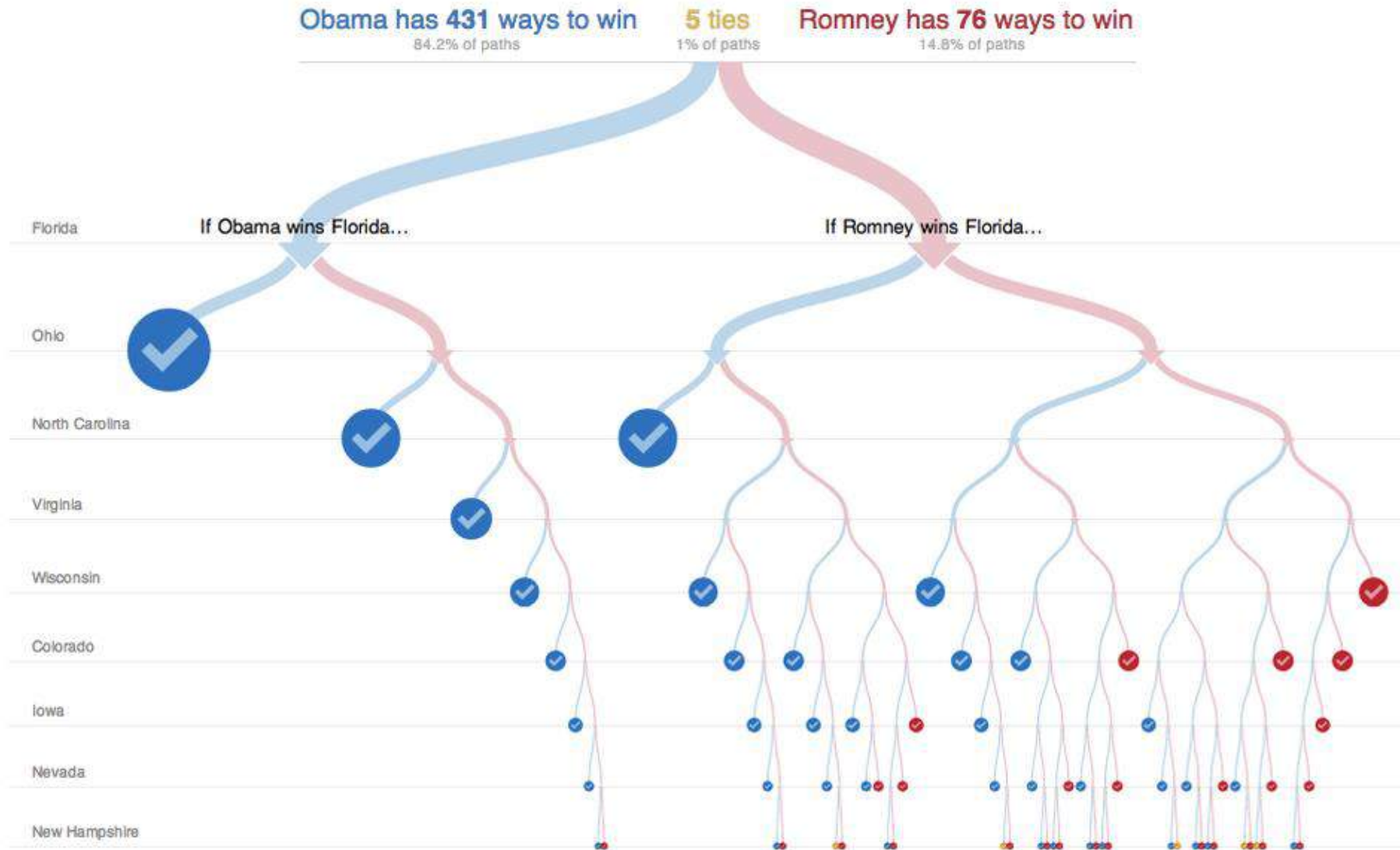
- Useful to group together and visualize any set of complex objects on top of which a distance function can be defined
  - Eg: Phylogenetic Trees Computed from “Genetic Distance” from multiple sequence alignments.





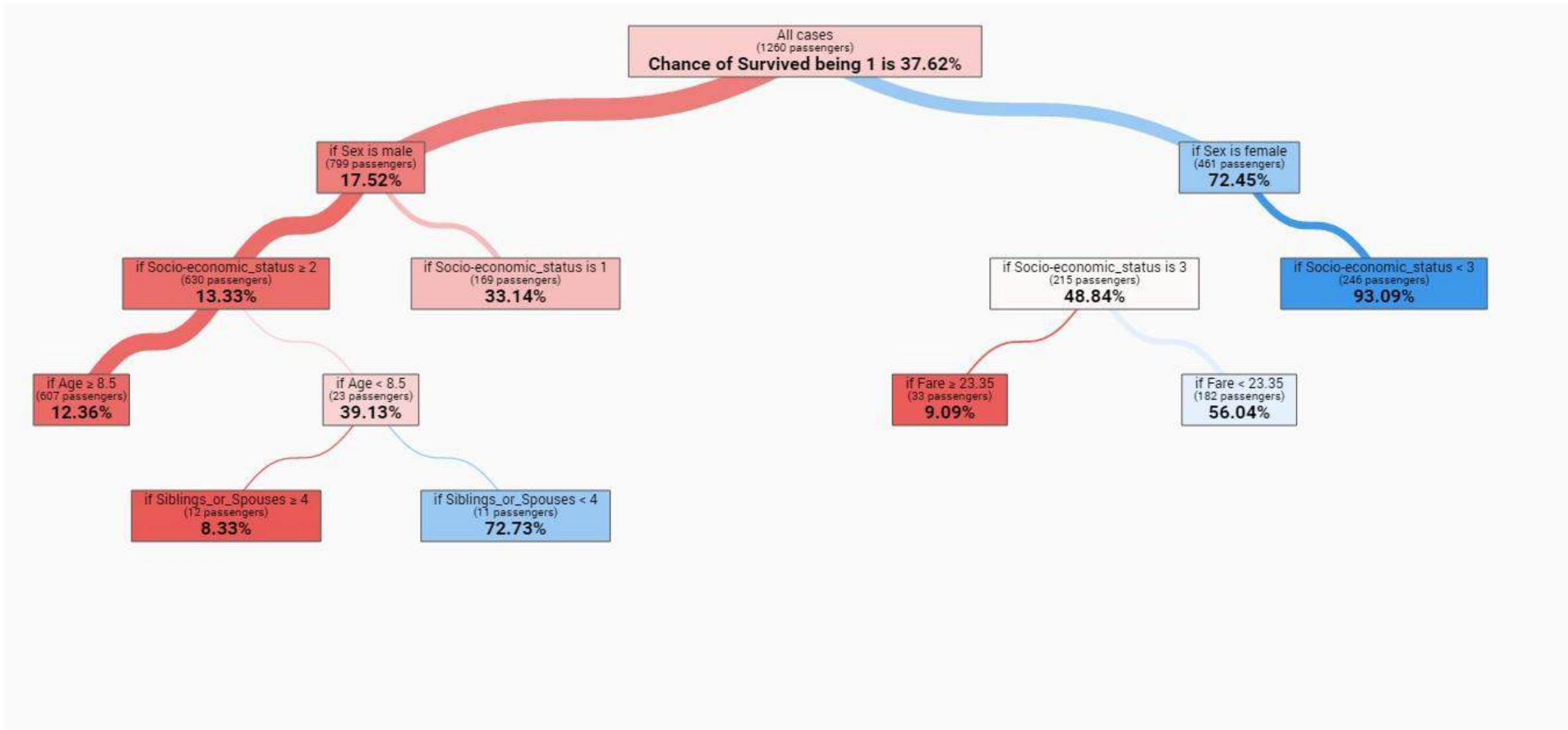
# Decision trees

- Visualize different combinations of decisions/events/outcomes



# Decision trees

- Visualize different combinations of decisions/events/outcomes
- Can be built automatically from data

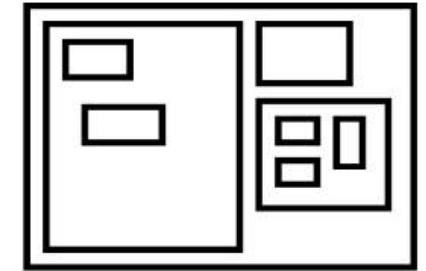
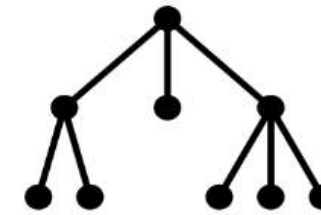
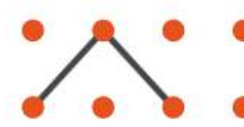


# Alternative: link marks w. connection & containment

- What we've seen so far:

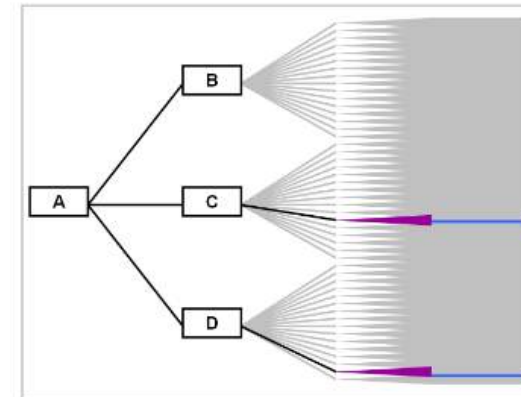
- 1D case: connection (lines)
  - Eg: all node-link diagrams
  - Emphasizes topology, path tracing
  - Networks and trees

→ Connection → Containment



- Alternative: marks as links (rather than nodes)

- Common case in network drawings
- 2D case: containment (areas)
  - Eg: treemaps
  - Emphasizes attribute values at leaves (size encoding)
  - Only works for trees



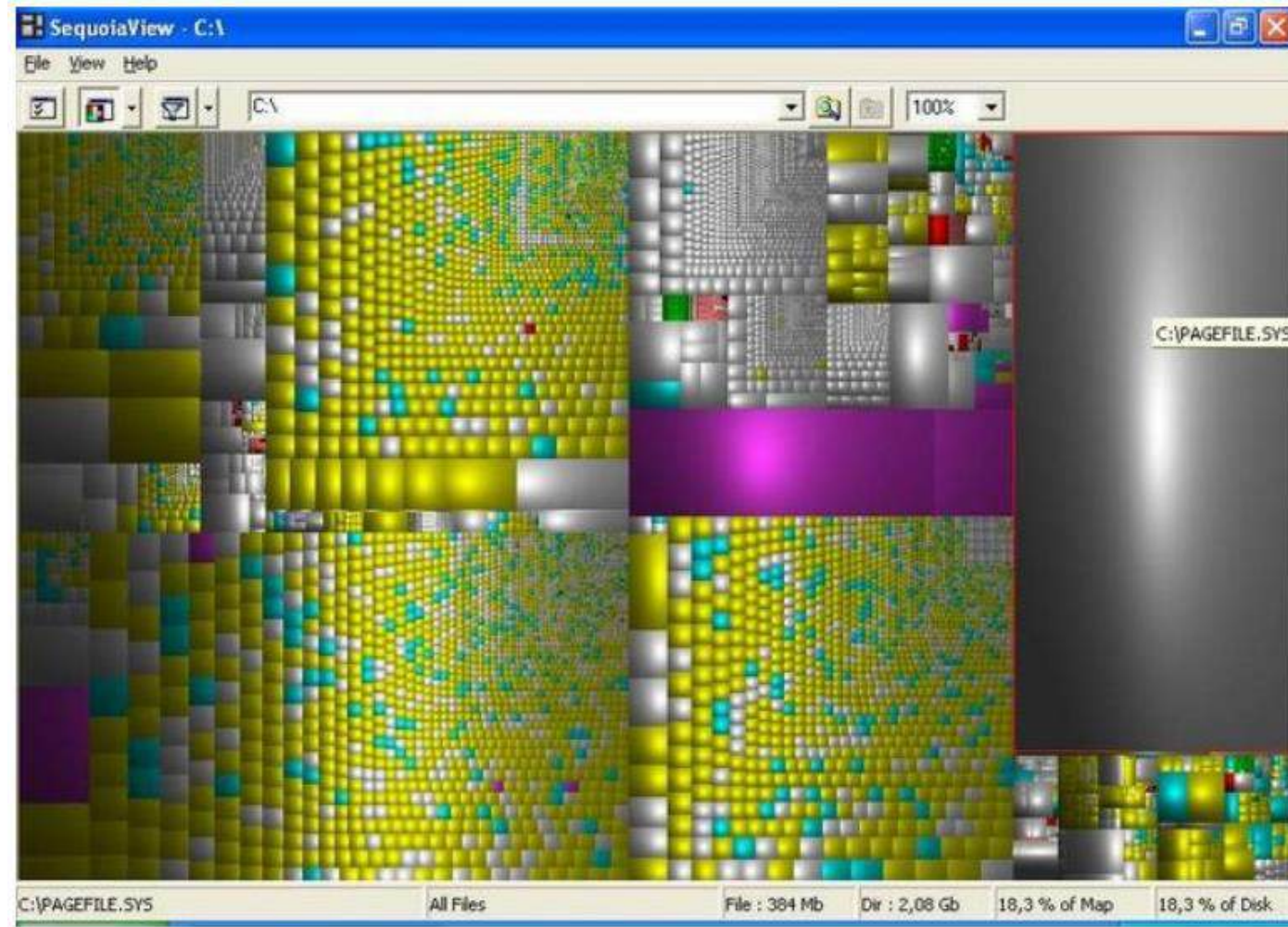
**Node-Link Diagram**



**Treemap**

# Treemap

- Data:
  - Tree
  - 1 quant. attribute at leaf nodes
- Encoding
  - Area containment marks for hierarchical structure
  - Rectilinear orientation
  - Size encodes quant. Attributes
- Tasks:
  - Query attribute at leaf nodes
  - Eg: disk space usage
- Scalability
  - 1M leaf nodes



Cushion Treemaps. van Wijk and van de Wetering. Proc. Symp. InfoVis 1999, 73-78.  
<https://sequoiaview.win.tue.nl/>

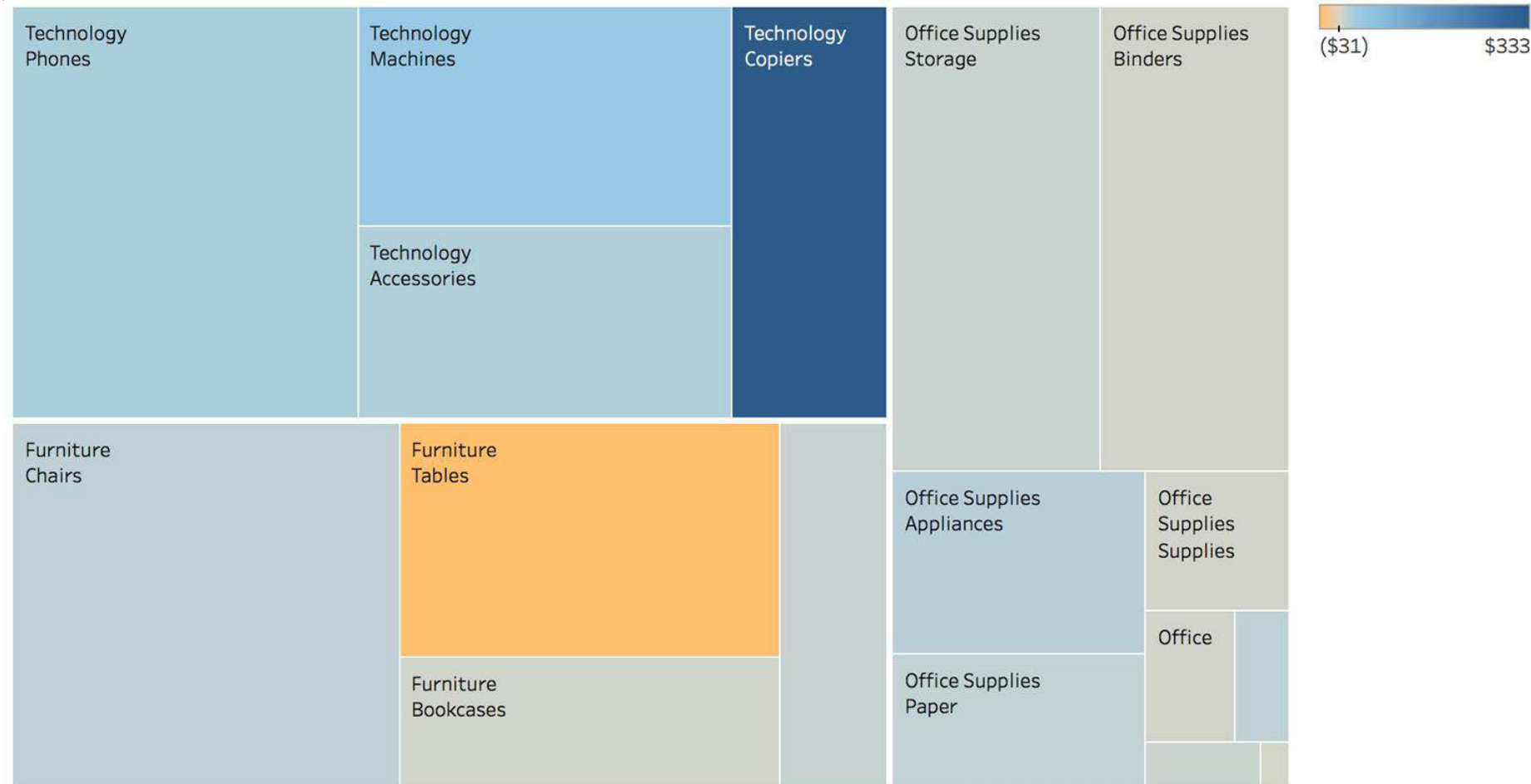


- What information can be visualized with a tree map?
  - Area: Quantity
  - Color: Quantity/Category
  - Hierarchy: Nesting

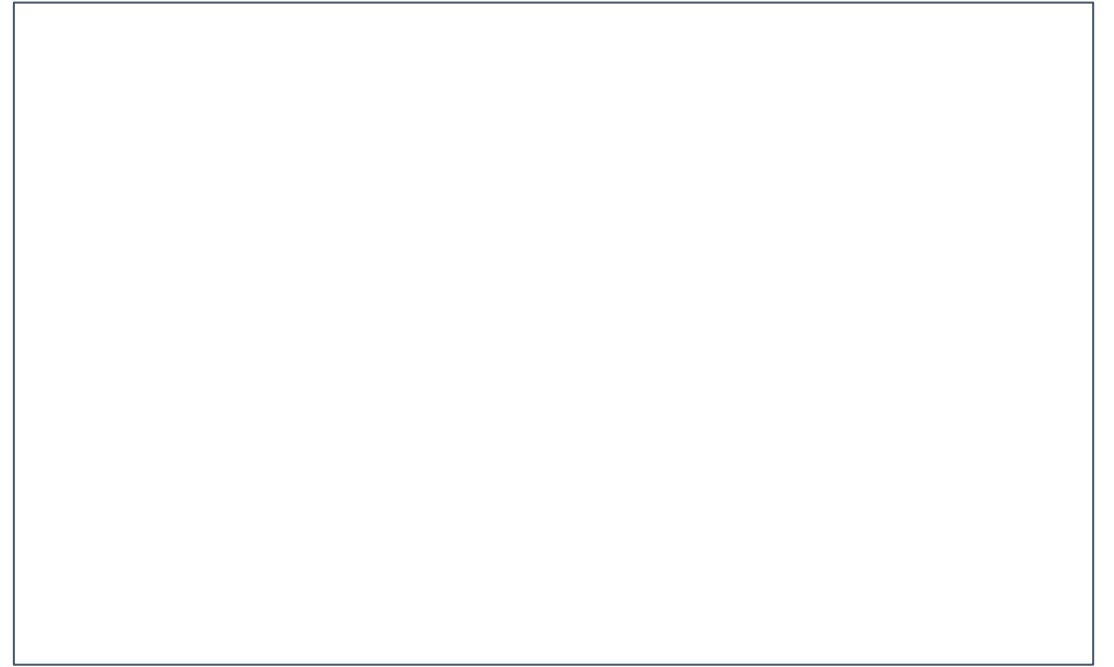
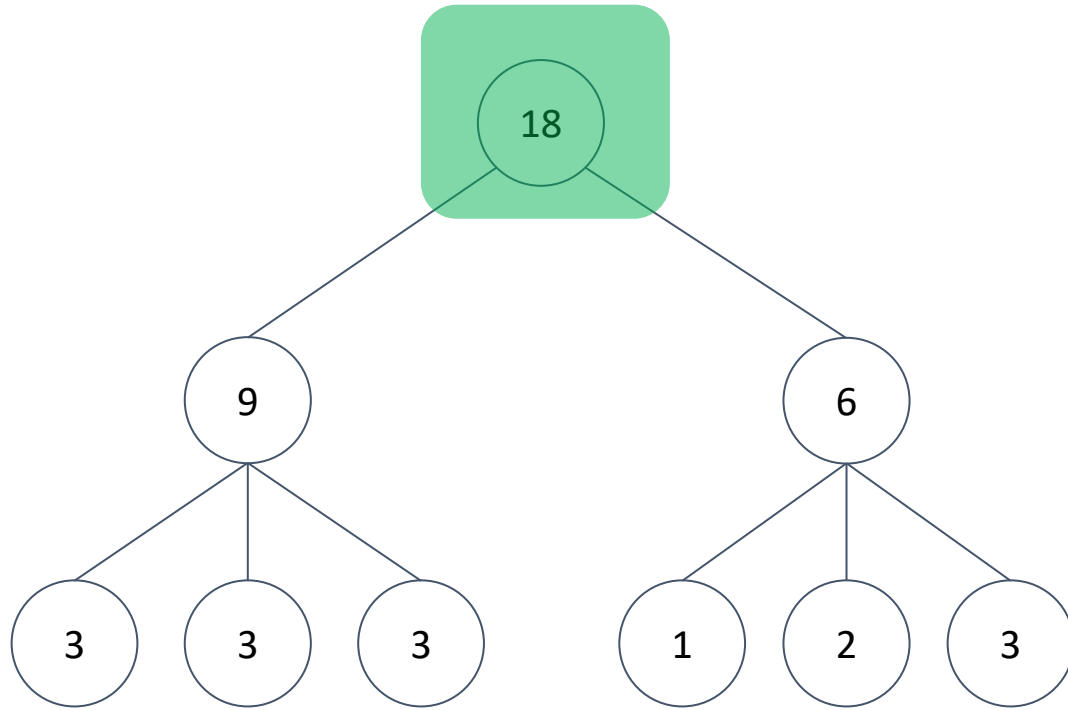


# Treemap

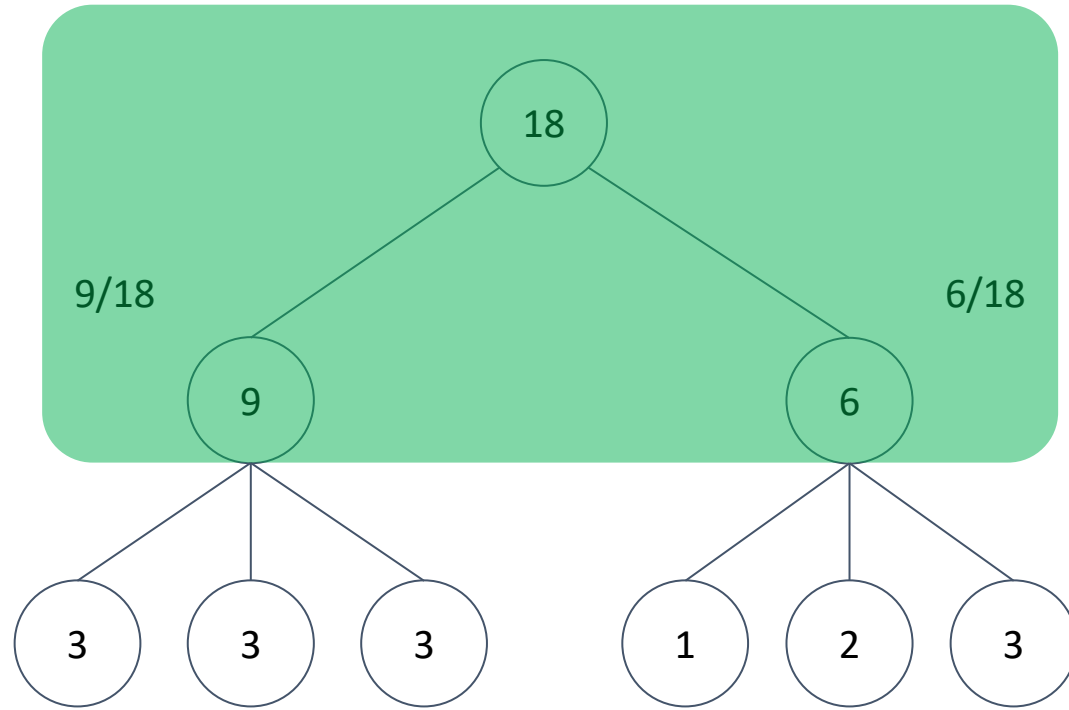
- What information can be visualized with a tree map?
  - Area: Quantity
  - Color: Quantity/Category
  - Hierarchy: Nesting



# Treemap construction

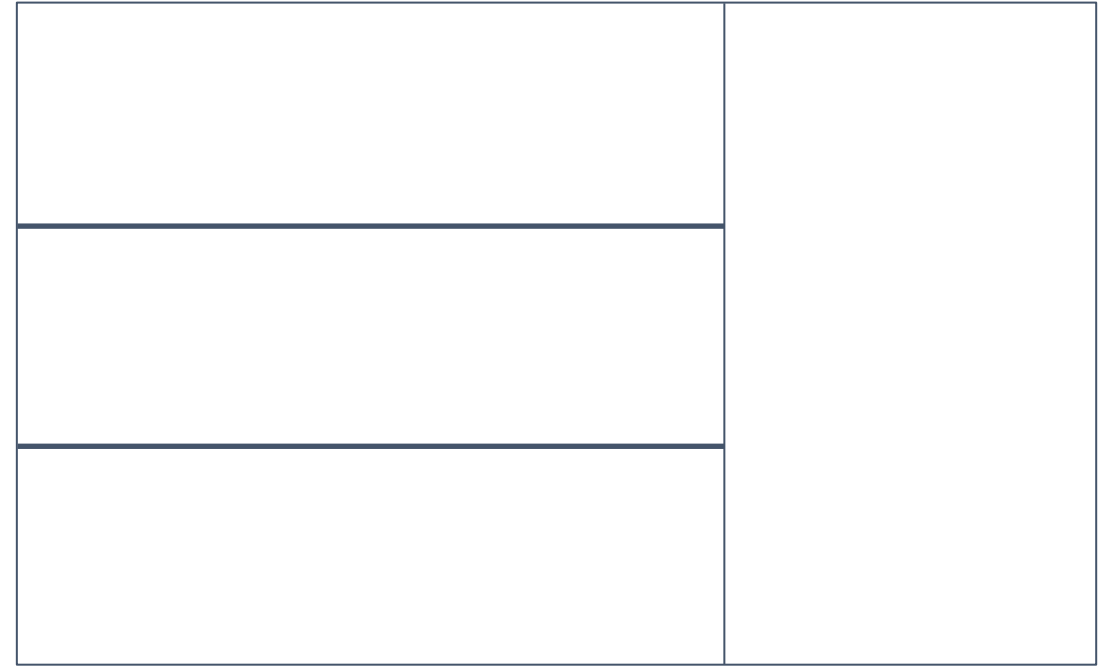
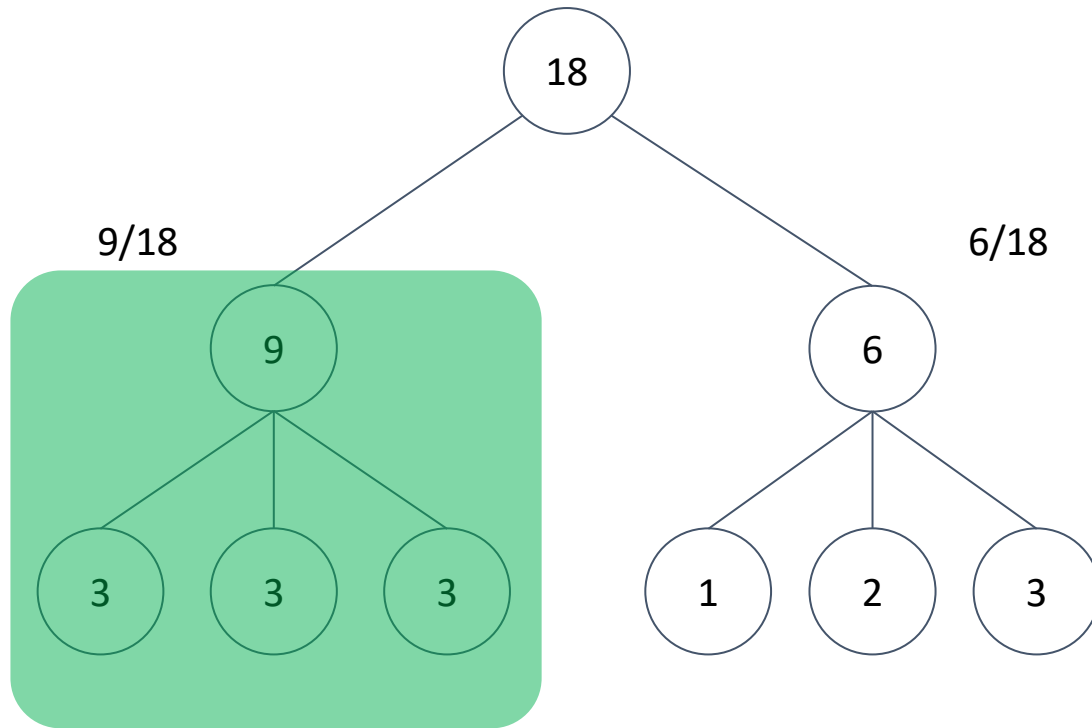


# Treemap construction

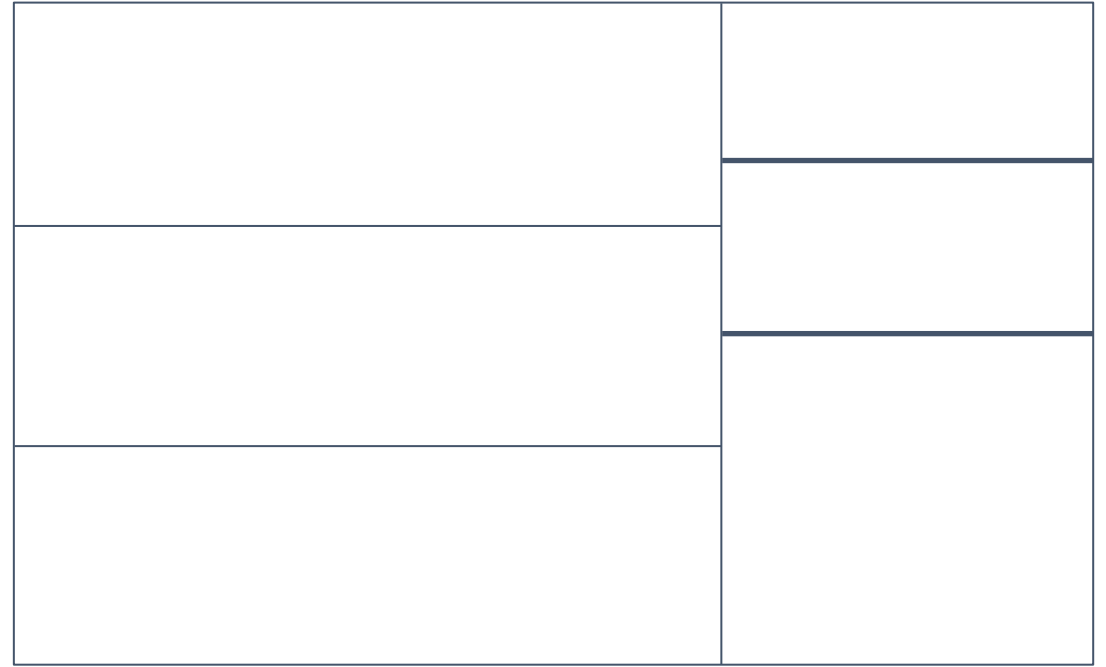
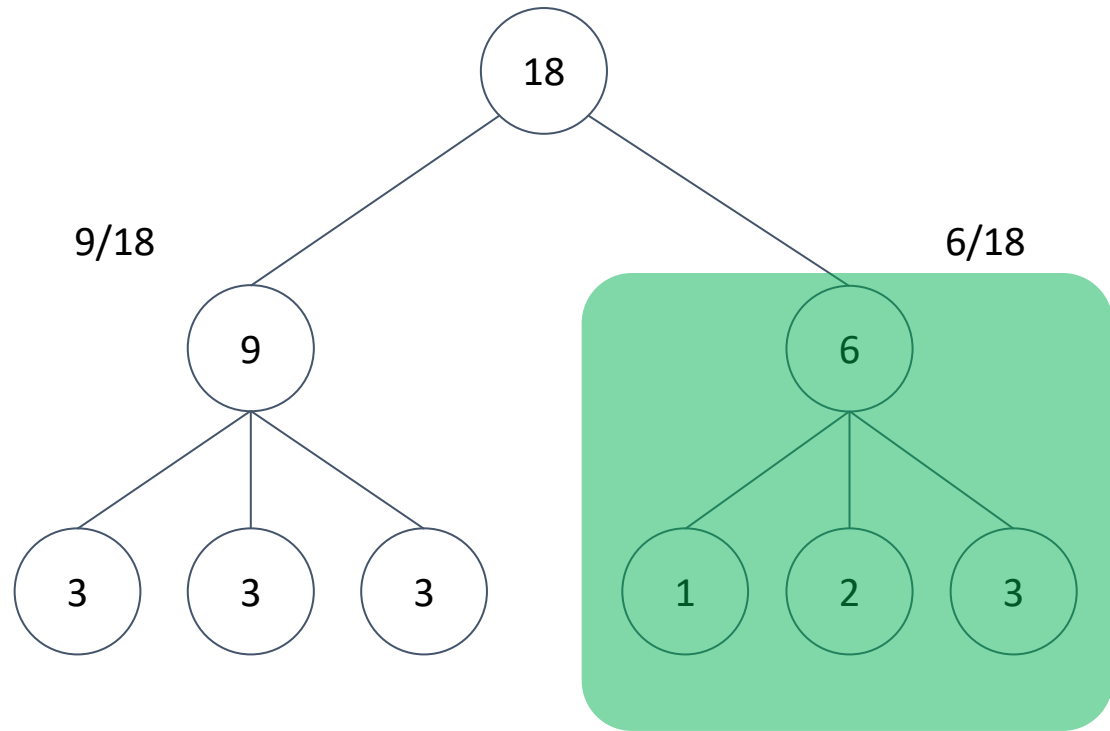




# Treemap construction

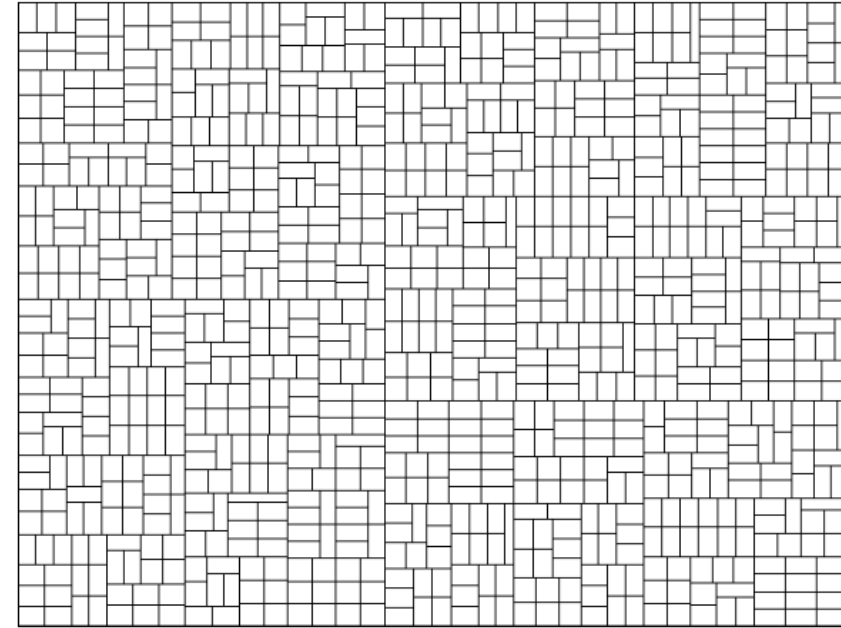
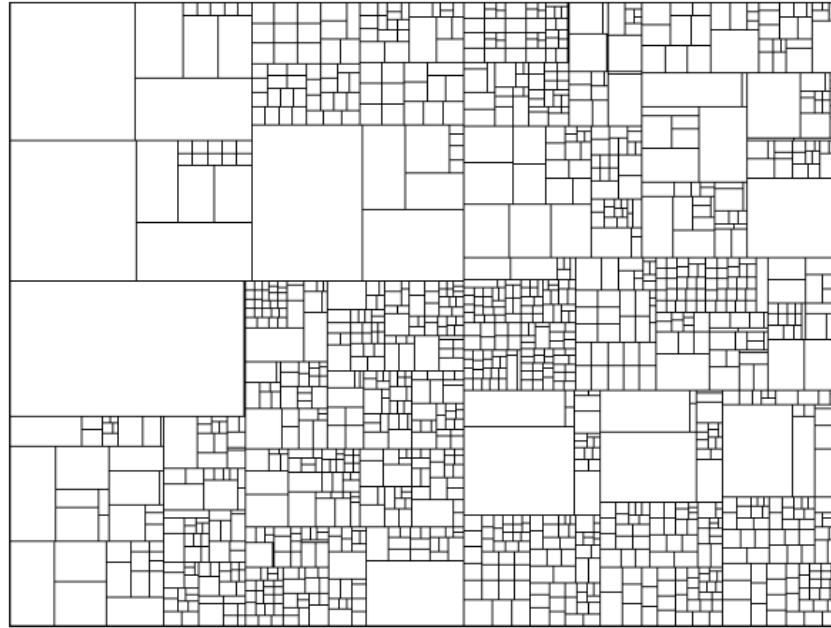


# Treemap construction



# Treemap drawback

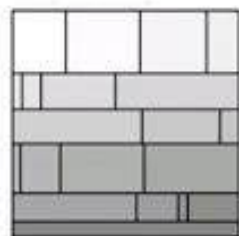
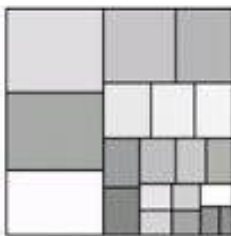
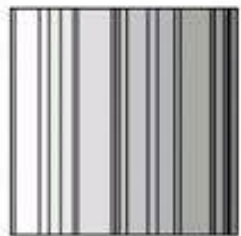
- When the rectangles have very different aspect ratios (proportion of height vs. width), comparing areas gets harder (especially with thin elongated rectangles).
- Solution: Squarified Tree Maps.



SliceAndDice

Squarified

Strip

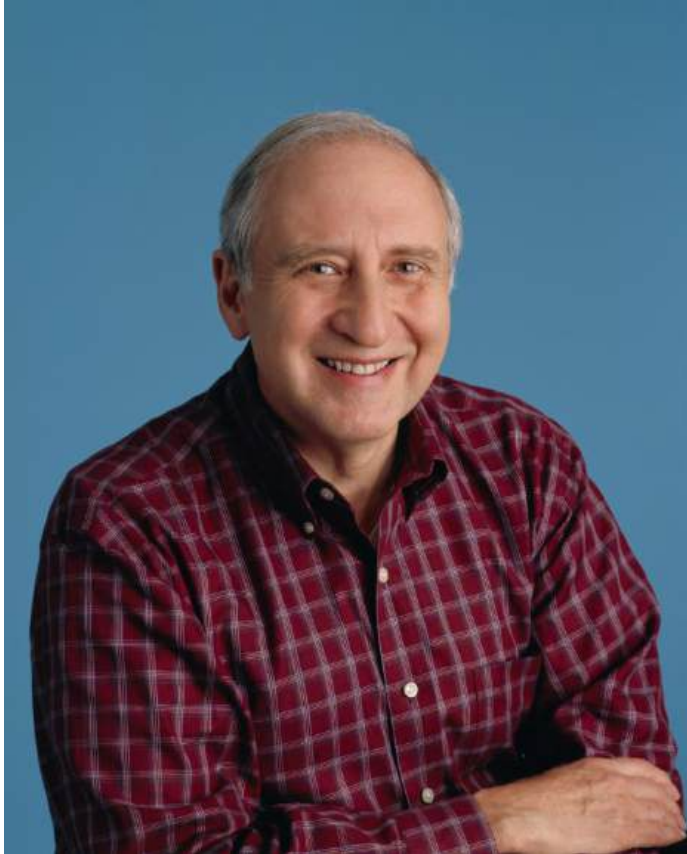


# Treemap summary

- Advantages:
  - Node visibility
  - No overlapping marks
  - Supports size and color channels
- Disadvantages:
  - Size is not the most accurate channel
  - Structures can be hard to discern



# Treemap lore



*“During 1990, in response to the common problem of a filled hard disk, I became obsessed with the idea of producing a compact visualization of directory tree structures.”* - Ben Shneiderman

<http://www.cs.umd.edu/hcil/treemap-history/>

## Alternative: implicit tree layout

- Alternative to node-link trees (connection) and treemaps (containment)
  - Show parent-child relationship only through relative positions

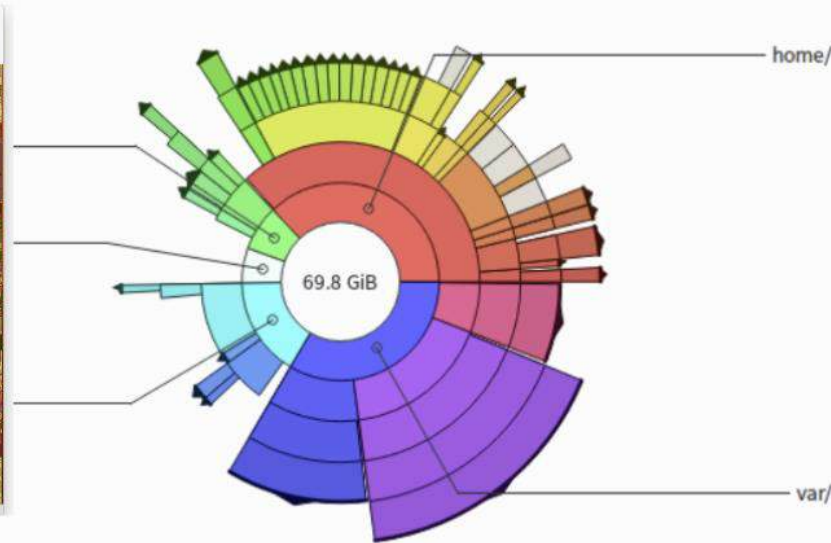
# Treemap

- Containment
- Only leaves visible



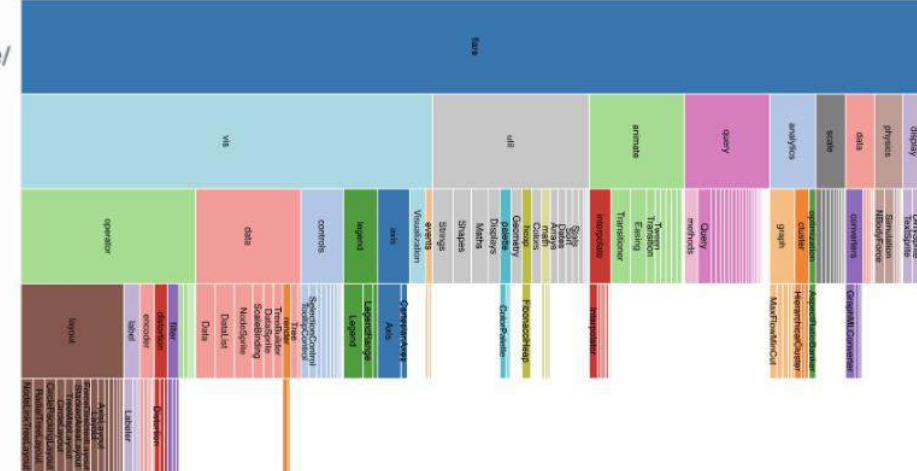
# Sunburst

- Position (radial)
- Inner nodes & leaves visible



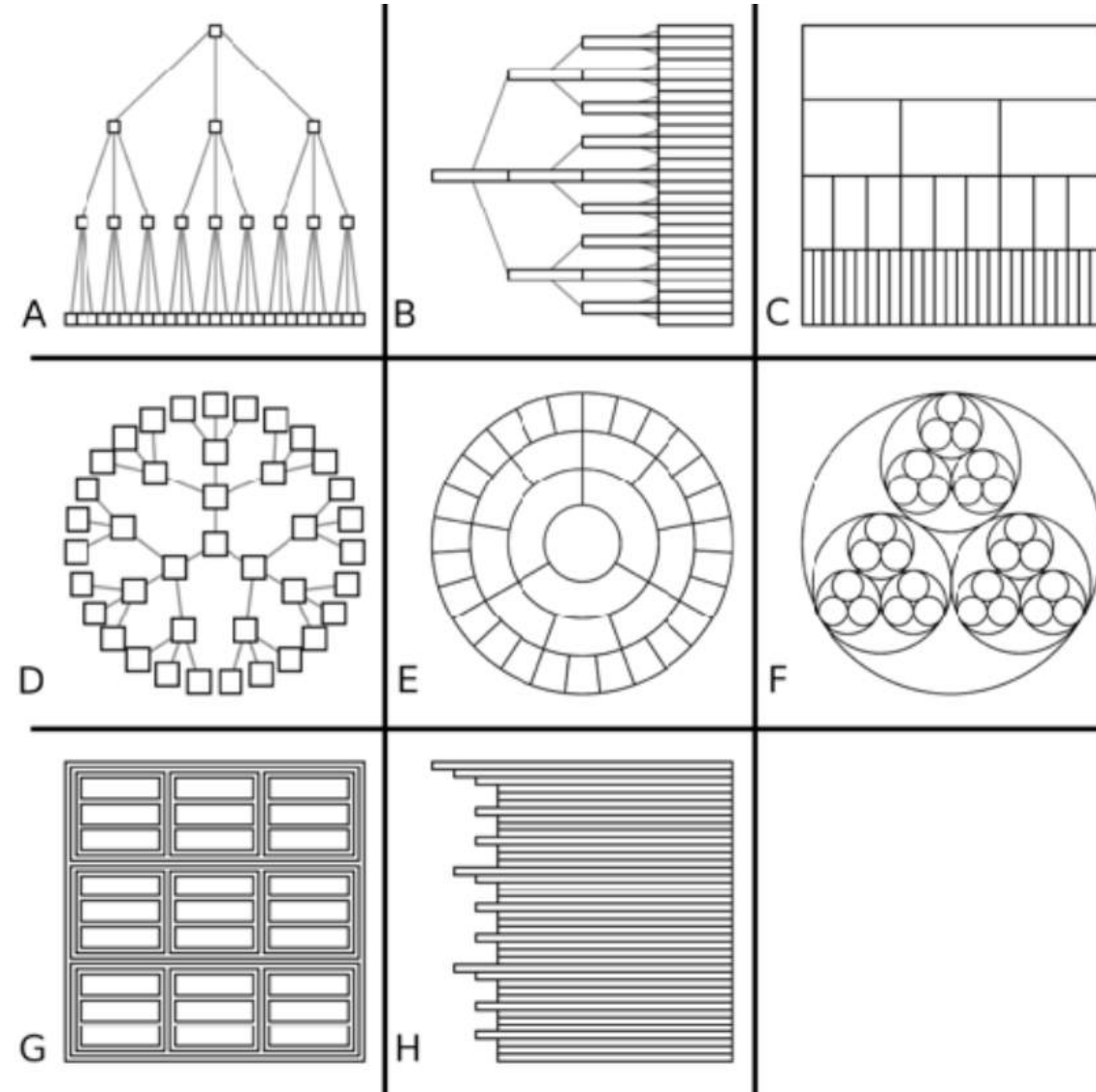
## Icicle plot

- Position (rectilinear)
- Inner nodes & leaves visible



# Tree viz considerations & comparisons

- Data shown
  - Link relationships
  - Tree depth
  - Sibling order
- Design choices
  - Connection vs containment link marks
  - Rectilinear vs radial layout
  - Spatial position channels
- Considerations
  - Redundant encoding? Arbitrary encoding?
  - Information density?
    - Avoid wasting space, but don't overlap
    - Consider where to fit labels!



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



# treevis.net: Many, many options!

<https://treevis.net/>

How to cite this site?

Check out other surveys!

**treevis.net - A Visual Bibliography of Tree Visualization 2.0 by Hans-Jörg Schulz**

v.21-OCT-2014

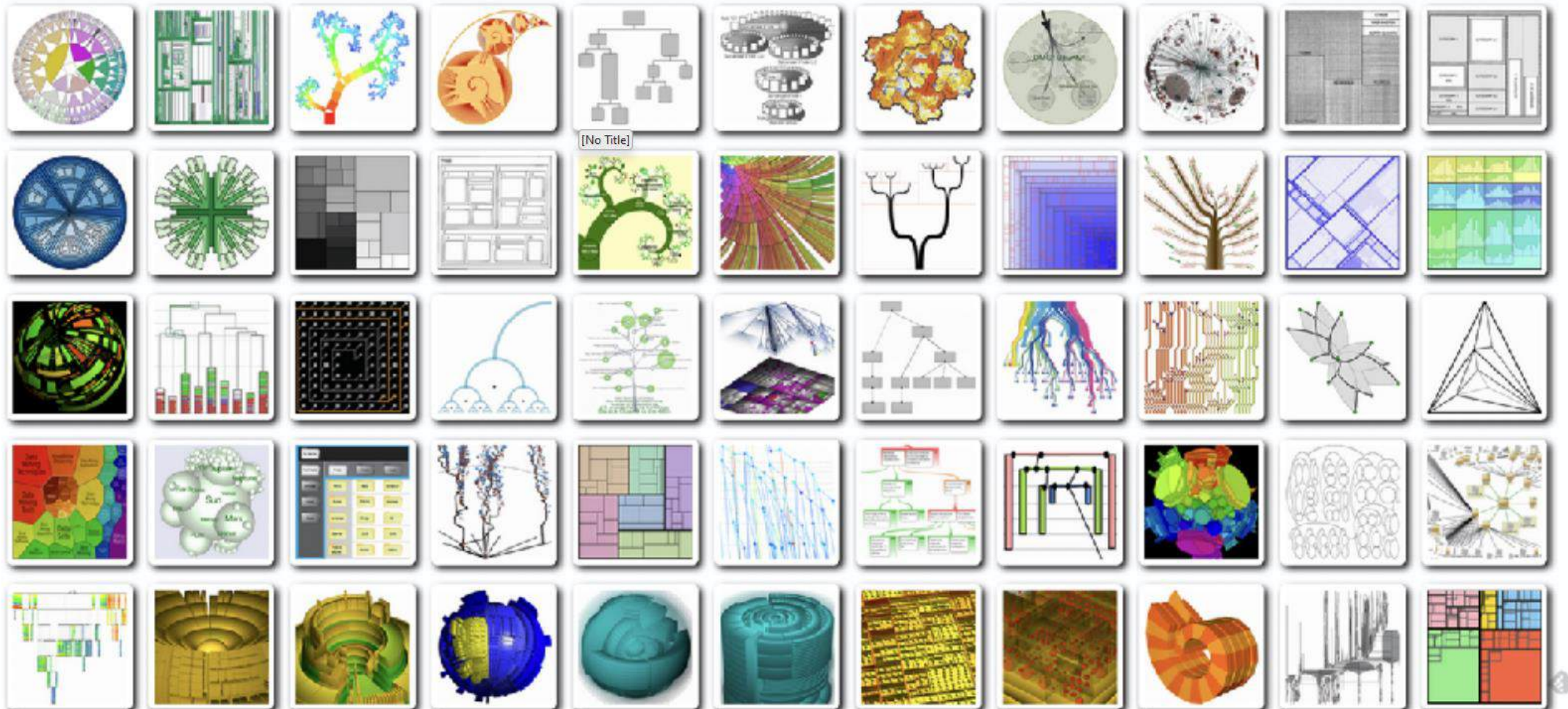
Dimensionality: All

Representation: All

Alignment: All

Fulltext Search:

Techniques Shown: **277**

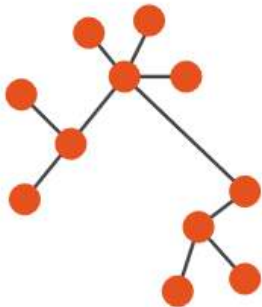




# Summary

➔ **Node–Link Diagrams**  
Connection Marks

✓ NETWORKS    ✓ TREES



➔ **Adjacency Matrix**  
Derived Table

✓ NETWORKS    ✓ TREES



➔ **Enclosure**  
Containment Marks

✗ NETWORKS    ✓ TREES



➔ **Implicit**  
Spatial Position

✗ NETWORKS    ✓ TREES





Break