

# NETWORK VISUALIZATION

CS 448B | Fall 2024

MANEESH AGRAWALA

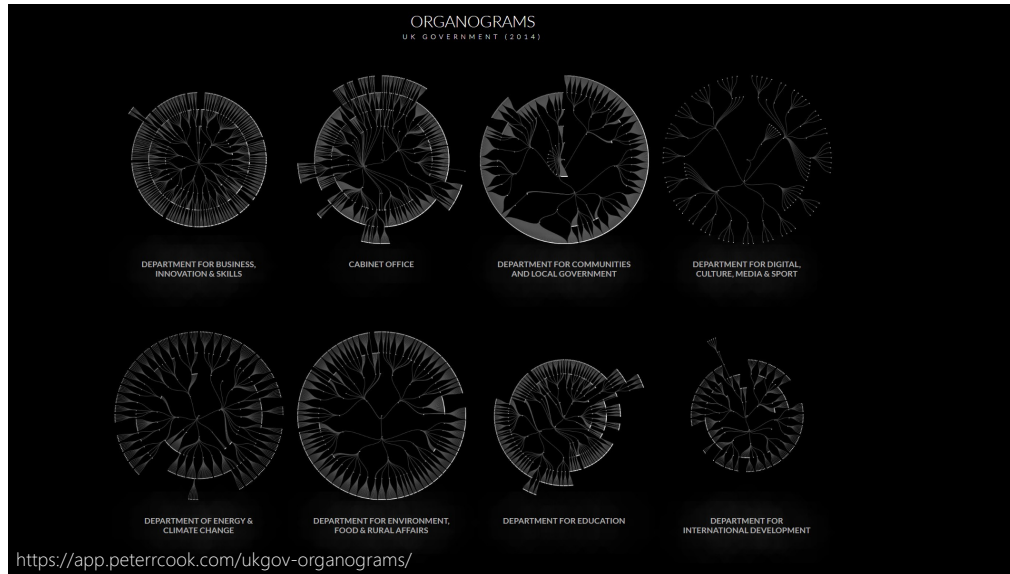
1

## READING RESPONSE: QUESTIONS/THOUGHTS

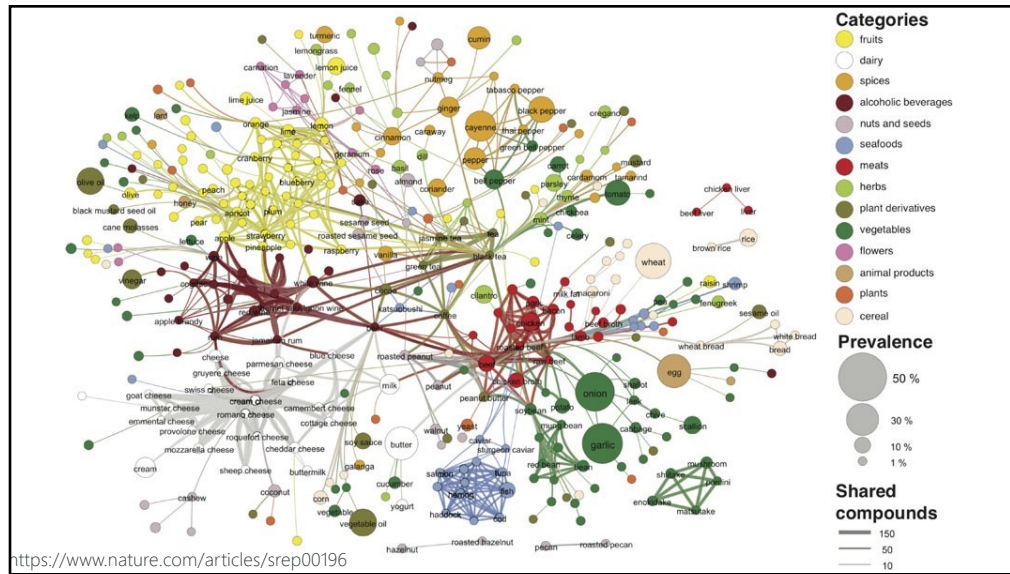
While Segel and Heer's martini glass structure actively encourages users to construct personal narratives from the data, Tufte's focus remains on the clear communication of data, with personal interpretation emerging incidentally. This contrast between the two sets of authors raises an interesting question: ***Could actively encouraging personal narratives in data visualization introduce unintended biases, especially when exploring complex or sensitive topics?***

***I'm curious how we should think picking color for communicating across cultures.*** The article by Heer and Stone mentions that categorical color perception is affected by language. Additionally, we saw in lecture results from the world color survey that show south pacific respondents distinguished blue and green far more than respondents from Mexico. Clearly, to what degree we view colors as distinct is subjective and contingent upon our culture and our language. I wonder, how much might these linguistic and cultural differences affect how viewers perceive our visualizations? ***How should we choose colors for our visualizations if we're trying to reach a multicultural audience?***

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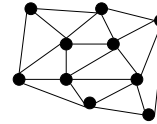


4

## GRAPHS AND TREES

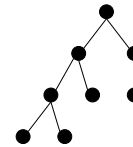
### Graphs

Model relations among data  
*Nodes and edges*



### Trees

Graphs with hierarchical structure  
 Connected graph with  $N-1$  edges  
 Nodes as *parents* and *children*



5

## SPATIAL LAYOUT

**Primary concern – positioning of nodes and edges**

**Often (but not always) goal is to depict structure**

- Connectivity, path-following
- Topological distance
- Clustering/grouping
- Ordering (e.g., hierarchy level)

6

## NETWORK ANALYSIS TASKS [Pretorius 2013]

**Structure-based:** relationships and connectivity

**Attribute-based:** properties associated with node or link

**Browsing:** follow paths in the data

**Estimation:** summarization and temporal changes

8

## NETWORK ANALYSIS TASKS [Pretorius 2013]

**Structure-based:** relationships and connectivity

Find all the friends of friends of Kermit

Find all the people who are friends of Animal and Gonzo

Find shortest path between two people: Six degrees of separation

**Attribute-based:** properties associated with node or link

**Browsing:** follow paths in the data

**Estimation:** summarization and temporal changes

9

## NETWORK ANALYSIS TASKS [Pretorius 2013]

### Structure-based: relationships and connectivity

Find all the friends of friends of Kermit

Find all the people who are friends of Animal and Gonzo

Find shortest path between two people: Six degrees of separation

### Attribute-based: properties associated with node or link

Find all friends of Fozzie that are students at Stanford (node property)

Find all friends of Fozzie that are their family (link property)

### Browsing: follow paths in the data

### Estimation: summarization and temporal changes

10

## NETWORK ANALYSIS TASKS [Pretorius 2013]

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### Browsing: follow paths in the data

Find Kermit's friend with first name Beaker and then find Beaker's mentor Bunsen

### Estimation: summarization and temporal changes

11

## NETWORK ANALYSIS TASKS [Pretorius 2013]

### Structure-based: relationships and connectivity

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### Browsing: follow paths in the data

Find Kermit's friend with first name Beaker and then find Beaker's mentor Bunsen

### Estimation: summarization and temporal changes

How does Miss Piggy's friend group change over the course of a year

12

## TODAY

### Learning Objectives

1. Techniques for visualizing trees
2. Techniques of laying out graphs
3. Alternative techniques for visualizing node-link data

13

# TREE VISUALIZATION

14

## COMMON TYPES OF TREE VISUALIZATION

### Indentation

Linear list, indentation encodes depth



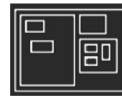
### Node-Link diagrams

Nodes connected by lines/curves



### Enclosure diagrams

Represent hierarchy by enclosure



### Layering

Layering and alignment



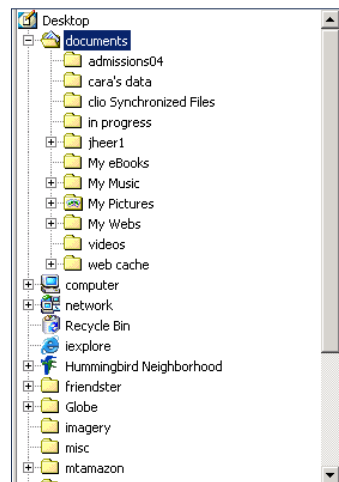
Tree layout is fast:  $O(n)$  or  $O(n \log n)$ , enabling real-time layout for interaction

15

# INDENTATION

16

## INDENTATION

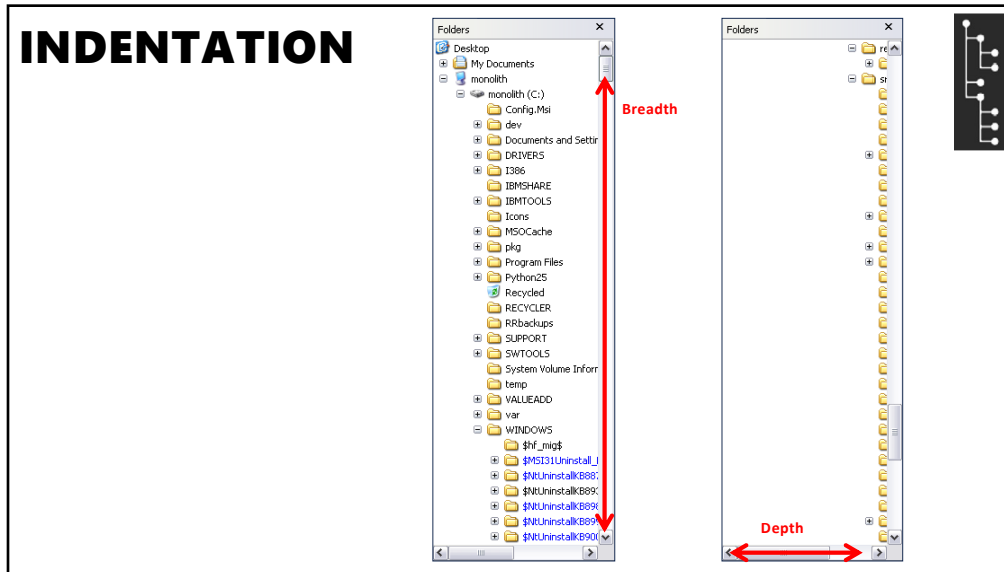


- Items along vertically spaced rows
- Indentation shows parent/child relationships
- Often used in interfaces
- Breadth/depth contend for space
- Often requires scrolling

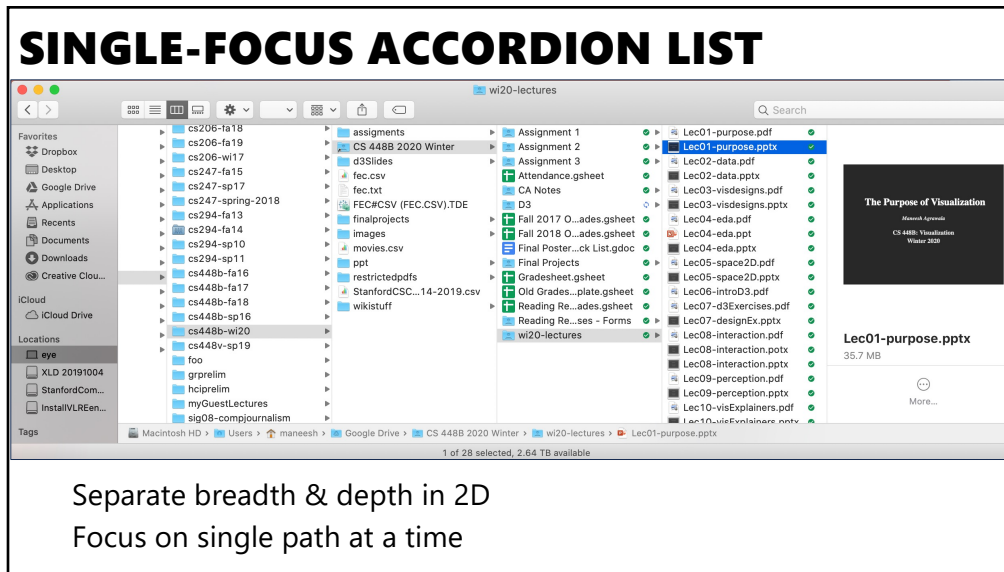


17



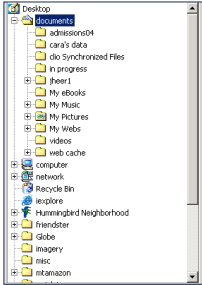
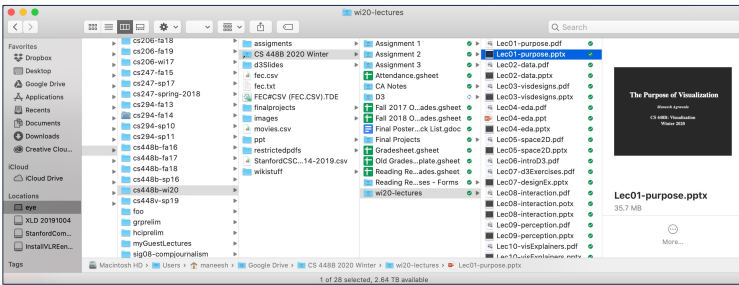


18



19

## WHAT TASKS ARE THESE GOOD FOR?

**Benefits**  
 Navigation + Browsing, Parent-child relationships

**Disadvantages**  
 Network overview, Estimation, Comparison

20

# NODE-LINK DIAGRAMS

21

## NODE-LINK DIAGRAMS



Nodes distributed in space, connected by straight or curved lines

Use 2D space to break apart breadth and depth

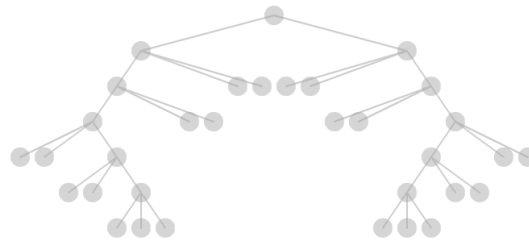
Space used to communicate hierarchical orientation

(e.g., towards *authority* or *generality*)

22

## BASIC RECURSIVE LAYOUT

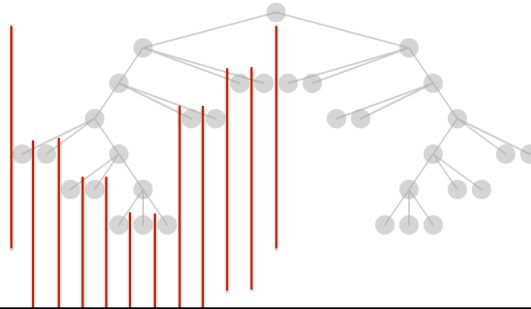
Repeatedly divide space for subtrees by leaf count



23

## BASIC RECURSIVE LAYOUT

Repeatedly divide space for subtrees by leaf count



24

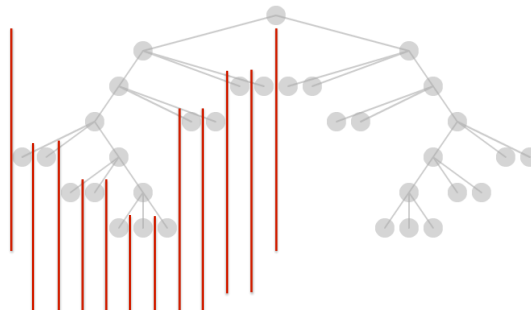
## BASIC RECURSIVE LAYOUT

Repeatedly divide space for subtrees by leaf count

Breadth of tree along one dimension

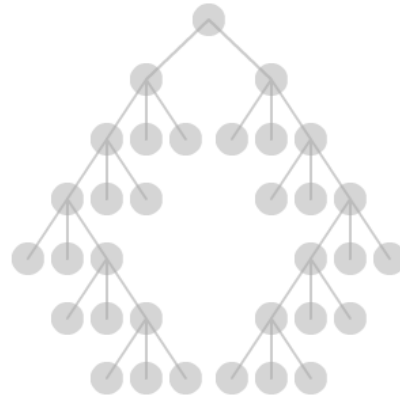
Depth along the other dimension

**Problem:** Exponential growth of breadth



25

## REINGOLD & TILFORD'S "TIDY" LAYOUT



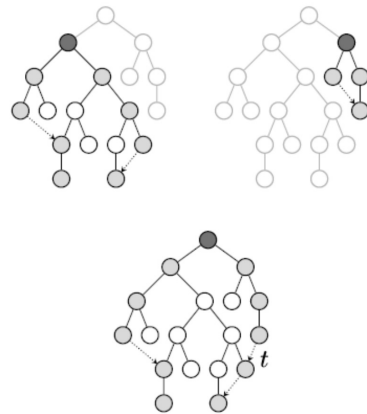
**Goal:** maximize density and symmetry

Originally for binary trees, extended by Walker to cover general case

Corrected by Buchheim et al. to achieve a linear time algorithm

26

## REINGOLD & TILFORD LAYOUT



### Design Considerations

Clearly encode depth

No edge crossings

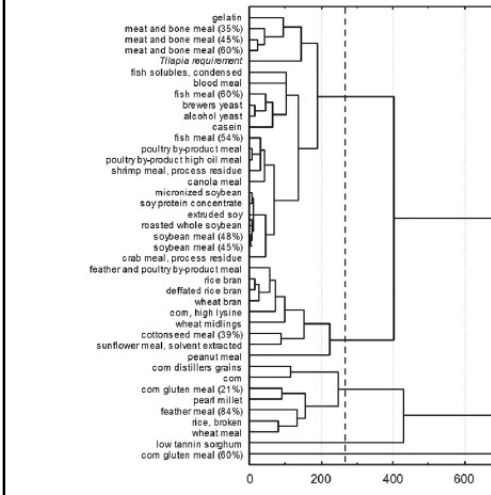
Draw isomorphic subtrees identically (same shape)

Preserve layout ordering and symmetry

**Compact space saving layout (don't waste space)**

27

## CLUSTER DENDROGRAMS



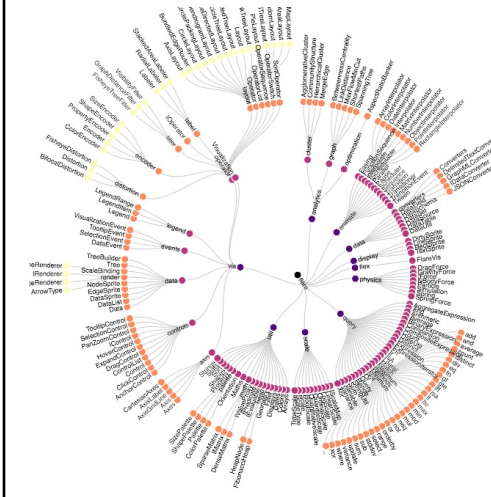
Depicts cluster trees produced by hierarchical clustering algorithms

Leaf nodes arranged in line, internal node depth indicates order/value at which clusters merge

Apply basic recursive layout with orthogonal two-segment edges

59

## RADIAL LAYOUT



Node-link diagram in polar coordinates

Radius encodes depth, root at center

Angular sectors assigned to subtrees (basic recursive approach)

Reingold-Tilford approach can also be applied here

60

# ANALYSIS TASKS: FOCUS + CONTEXT

61

## VISUALIZING LARGE HIERARCHIES

The image compares two methods for visualizing large hierarchies. On the left, an 'Indented Layout' is shown as a screenshot of a Windows File Explorer window. It displays a tree structure of folders and files with varying levels of indentation. On the right, a 'Reingold-Tilford Layout' is shown as a complex, branching tree diagram. This layout uses a consistent horizontal distance between nodes to represent hierarchical levels, which helps in identifying patterns and relationships across a large dataset.

**Indented Layout**

**Reingold-Tilford Layout**

62

## MORE NODES, MORE PROBLEMS...

### Scale

Tree breadth often grows exponentially  
Even with tidier layout, quickly run out of space

### Possible solutions

Filtering  
Scrolling or Panning  
Zooming  
Aggregation  
Focus+Context

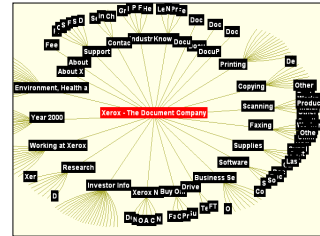
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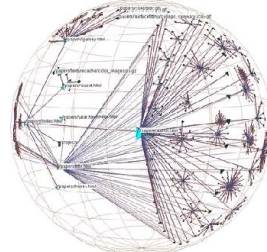
## HYPERBOLIC LAYOUT



Perform tree layout in hyperbolic space, then project the result on to the Euclidean plane

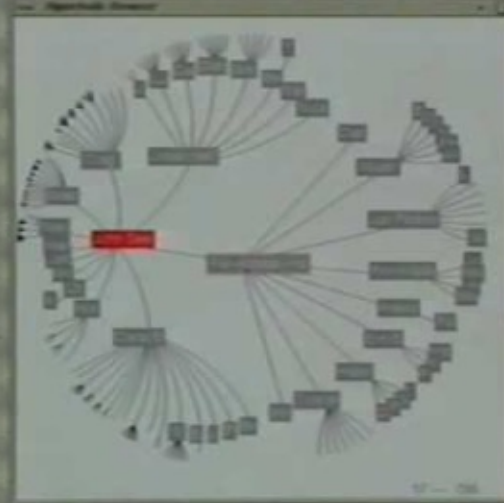
Why? Like tree breadth, the hyperbolic plane expands exponentially!

Also computable in 3D, projected into a sphere

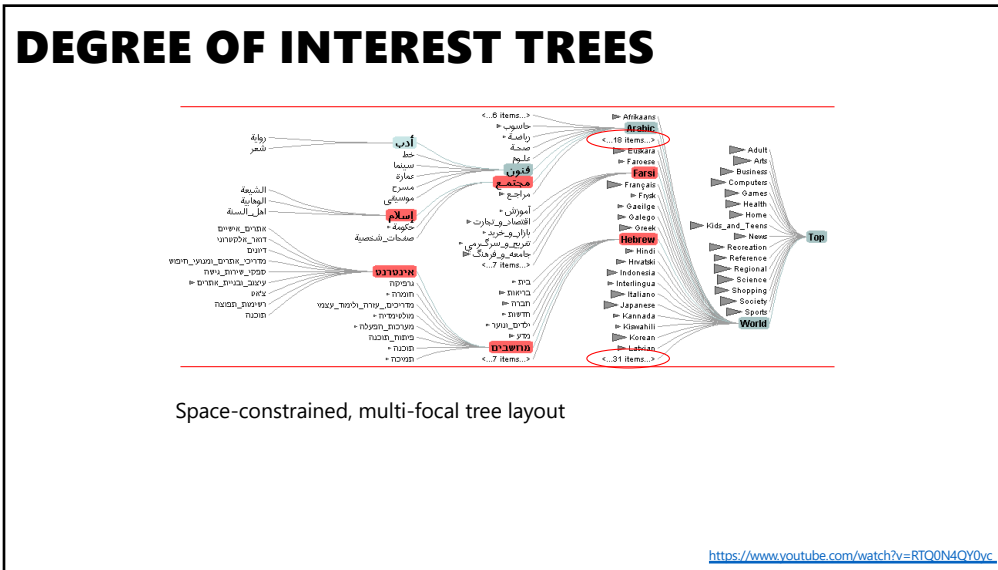


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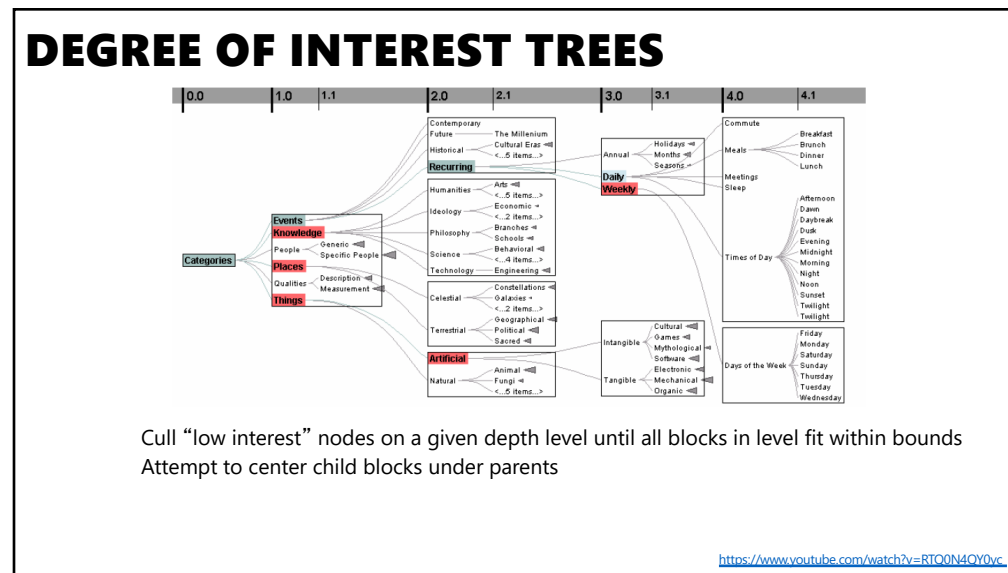
## HYPERBOLIC LAYOUT [Rao 1995]



66



67



68

## INDENTATION & NODE-LINK DIAGRAMS

Encode structure in **2D space** (breadth/depth)

### Benefits

Clearly depicts node relationships / structure  
Structure-based or browsing tasks

### Problems

Even with tidy layout, quickly run out of space

### Missing

Attribute-based encodings

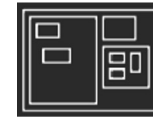
69



**ENCLOSURE**

70

## ENCLOSURE DIAGRAMS



Encode structure using **spatial enclosure**  
 Popularly known as **treemaps**

### Benefits

Provides a single view of an entire tree  
 Easier to spot large/small nodes

### Problems

Difficult to accurately read structure/depth

71

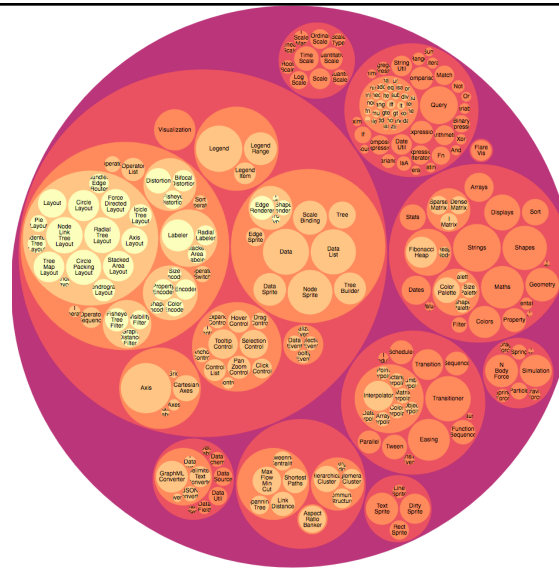
## CIRCLE PACKING LAYOUT

Nodes represented as sized circles

Nesting to show parent-child relationships

### Problems?

Inefficient use of space  
 Parent size misleading?



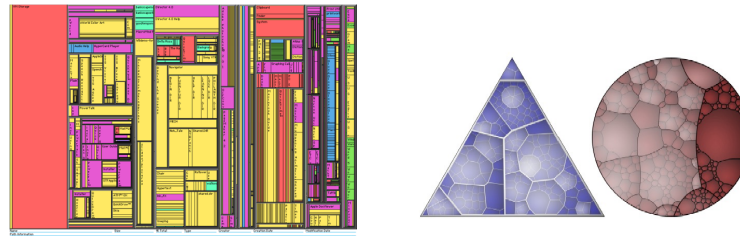
72

# TREEMAPS

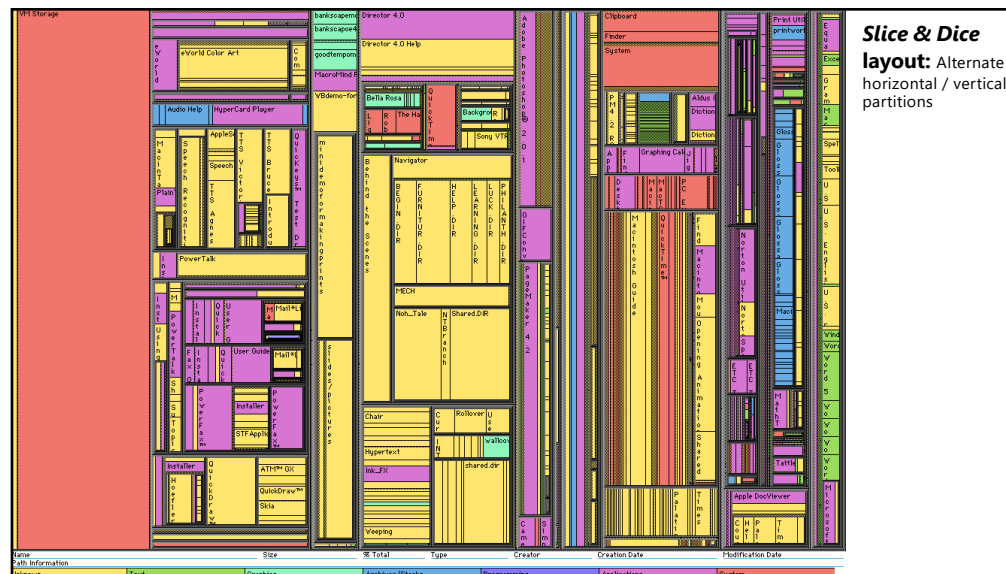
Hierarchy visualization that emphasizes values of nodes via area encoding

Partition 2D space such that leaf nodes have sizes proportional to data values

First algorithms designed to show file sizes on a hard drive [Shneiderman 1990]



74



75



76

## SQUARIFIED TREEMAPS [Bruls 2000]

Slice & Dice layout suffers from extreme aspect ratios. How might we do better?

Squarified layout: greedy optimization with objective of square rectangles. Slice/dice within siblings; alternate whenever ratio worsens

77

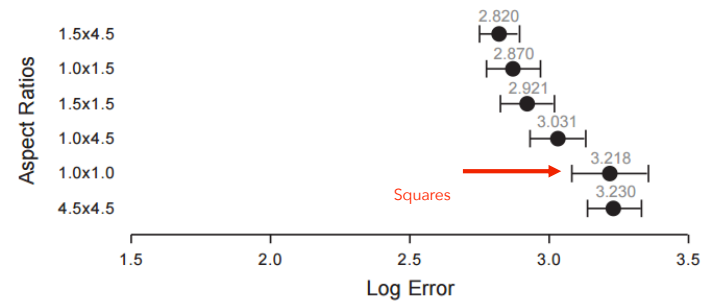
## WHY SQUARES?

### Posited Benefits of 1:1 Aspect Ratios

1. Minimize perimeter, reducing border ink  
*Mathematically true!*
2. Easier to select with a mouse cursor.  
*Validated by empirical research & Fitt's Law!*
3. Similar aspect ratios are easier to compare.  
*Seems intuitive, but is this true?*

78

## COMPARISON: ERROR VS. ASPECT RATIO



### Study by Kong et al. 2010

Comparison of squares has higher error! Comparison of extreme ratios even worse  
Perhaps squaring works well because it fails to meet its objective?

79

## WHY SQUARES?

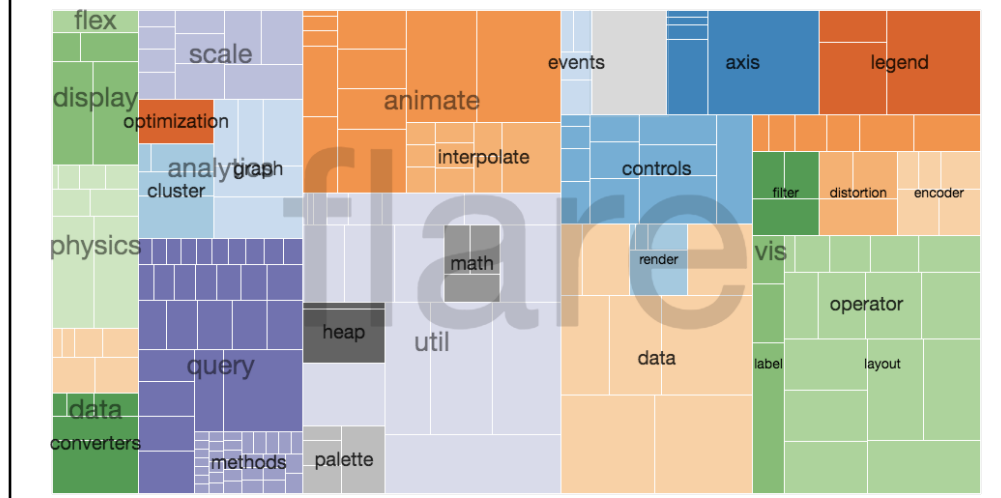
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1. Minimize perimeter, reducing border ink  
*Mathematically true!*
2. Easier to select with a mouse cursor.  
*Validated by empirical research & Fitt's Law!*
3. Similar aspect ratios are easier to compare.  
*Seems intuitive, but is this true?*  
*Extreme ratios & squares-only, are more inaccurate.*  
*Balanced ratios better? Target golden ratio?*

80

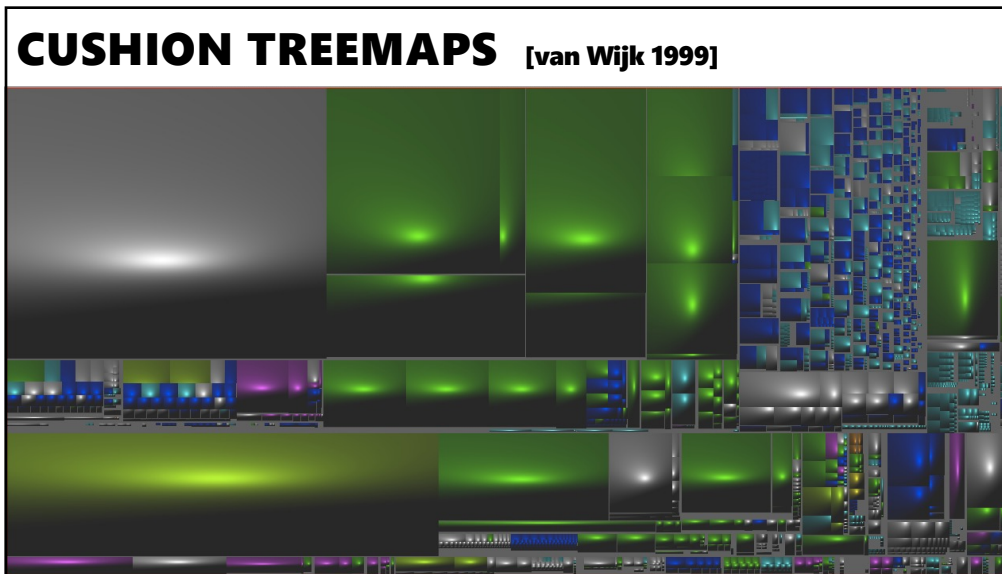
## INTERACTIVE EXAMPLE

<https://vega.github.io/vega/examples/treemap/>

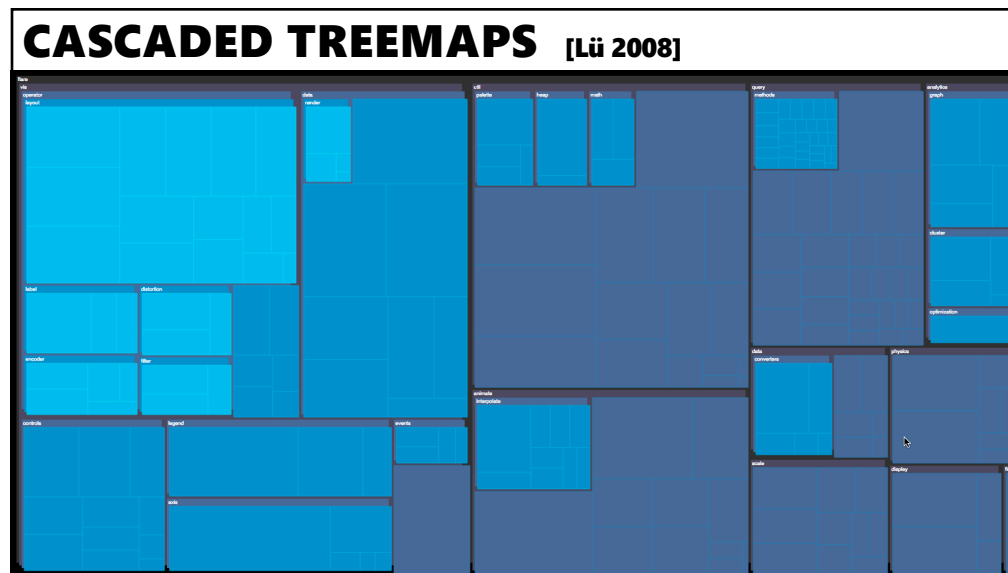


81





84

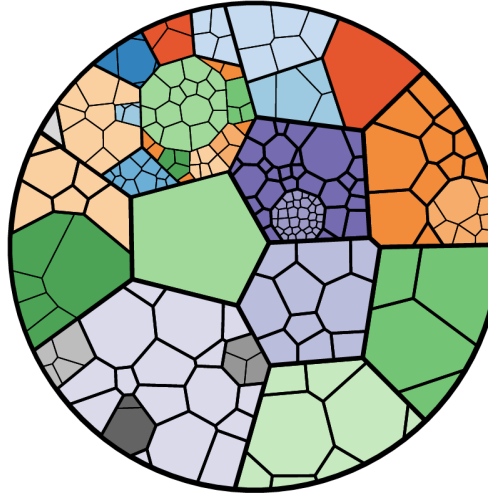


85

## VORONOI TREEMAPS [Balzer 2005]

Treemaps with arbitrary polygonal shapes and boundary

Uses iterative, weighted Voronoi tessellations to generate cells with value-proportional areas



86

## LAYERING

88

## LAYERED DIAGRAMS



Signify tree structure using

- Layering
- Adjacency
- Alignment

Involves recursive sub-division of space

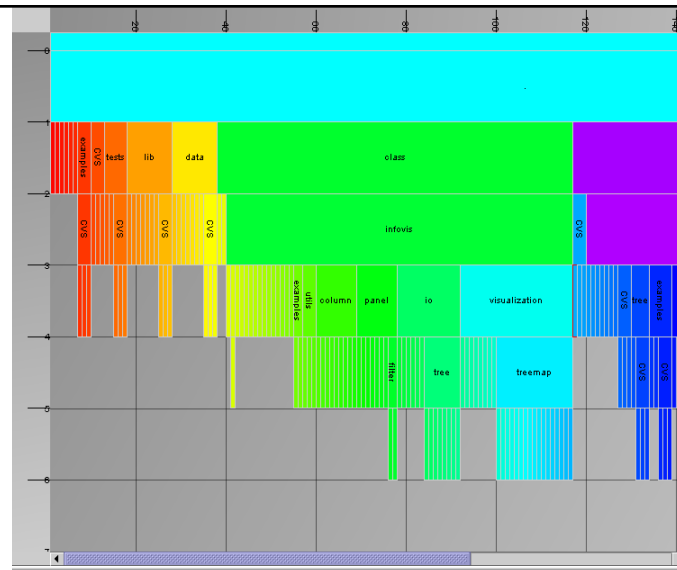
Leaf nodes may be sized by value, parent size visualizes sum of descendant leaf values

89

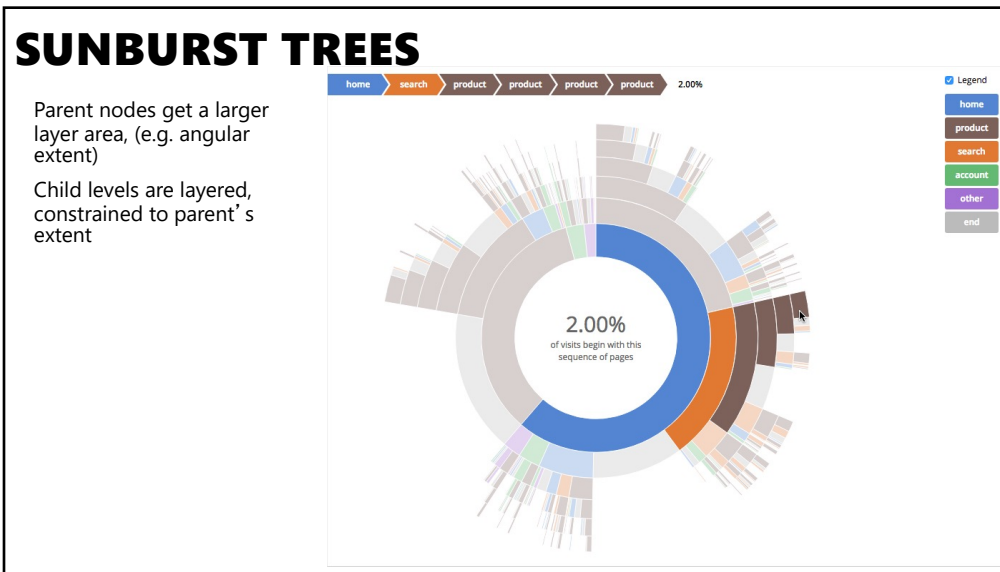
## ICICLE TREES

Parent nodes get a larger layer area, (e.g. horizontal extent)

Child levels are layered, constrained to parent's extent



90



91

## ANOTHER USE OF LAYERED TREES

		Coffee			Espresso			
		Amaretto	Columbian	Decaf Irish Cr..	Caffe Latte	Caffe Mocha	Decaf Espresso	Regular Espre..
Central	Colorado	█	█	█		█	█	
	Illinois		█			█		
	Iowa		█			█		
	Missouri		█			█		
	Ohio		█			█		
Wisconsin		█			█			
East	Connecticut		█			█		
	Florida		█	█		█		
	Massachusetts		█	█		█		█
	New Hamps..	█	█			█		█
	New York		█			█		█
South	Louisiana		█			█		
	New Mexico		█			█		
	Oklahoma		█			█		
	Texas		█		█	█		
West	California	█	█	█	█	█	█	
	Nevada		█			█		
	Oregon		█			█		
	Utah		█			█		
	Washington		█			█		
		-20K 0K 20K	-20K 0K 20K	-20K 0K 20K	-20K 0K 20K	-20K 0K 20K	-20K 0K 20K	-20K 0K 20K
		SUM(Profit)	SUM(Profit)	SUM(Profit)	SUM(Profit)	SUM(Profit)	SUM(Profit)	

92

# ANNOUNCEMENTS

94

## FINAL PROJECT

### Design Reviews Dec 2 and Dec 4

#### Data analysis/explainer

Analyze dataset in depth & make a visual explainer

#### Deliverables

An article with multiple different interactive visualizations

Short video (2 min) demoing and explaining the project

#### Schedule

Project proposal: Today!

Design Review and Feedback: 10<sup>th</sup> week of quarter, 12/2 and 12/4

Final code and video: Sun 12/8 8pm

#### Grading

Groups of up to 3 people, graded individually

Clearly report responsibilities of each member

95

## FINAL PROJECT GUIDELINES

### Consider the audience

Your visual explainer should be of interest to a group of people beyond your immediate circle (an explainer about your own Spotify data unlikely be of interest to others you don't know)

### Pick relatively less explored topics/datasets

Do some research on what has already been done for the topic/dataset(s)

Certain data like songs (e.g. Spotify) or movies (e.g. IMDB) are already well analyzed and should be avoided, unless you want to try to take a very different angle or use innovative analysis methods

### Develop a narrative

In the early stages of the analysis process, try to uncover patterns to help you form and shape a narrative through-line for the explainer

96

## FINAL PROJECT GUIDELINES

### Design visualization interactions

Choose base visualizations that can support a high level of interactivity  
Bubble charts, tree maps, and word clouds typically aren't the most effective choices

Design interactive features that would enable viewers to interact with the data in a way that strengthens your narrative

Tooltip is typically not enough interaction

Draw inspiration from sites like the New York Times and the Pudding

97

## NODE-LINK GRAPH LAYOUT

98

## NODE-LINK GRAPH VISUALIZATION

Nodes connected by lines/curves

**Sugiyama-Style Layout** - arranged by depth

**Force-Directed Layout** - physical simulation

**Attribute-Driven Layout** - arranged by value

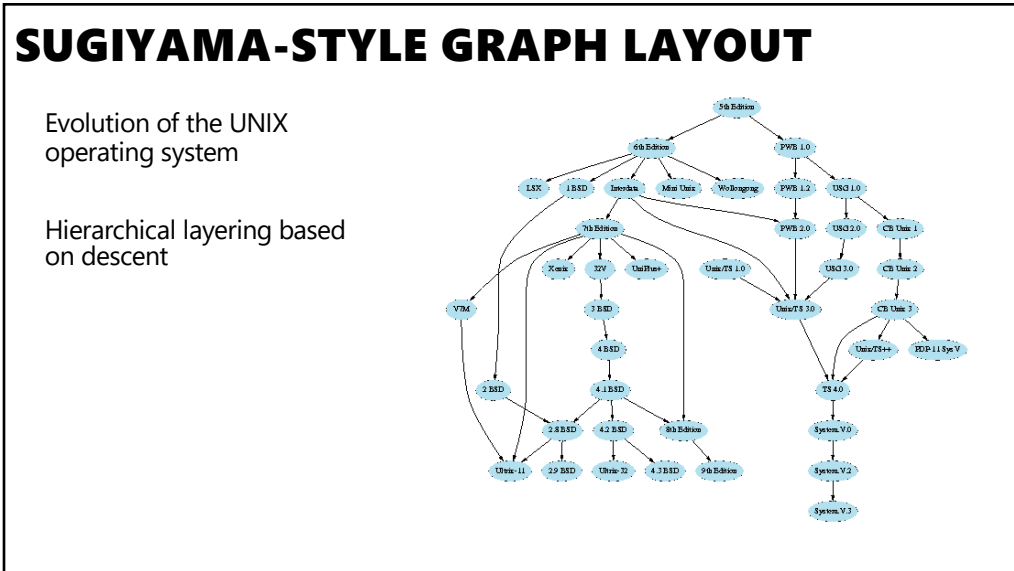
**Constraint-Based Layout** – optimization

**Arc Diagrams** - aligned layout

99

# SUGIYAMA-STYLE LAYOUT

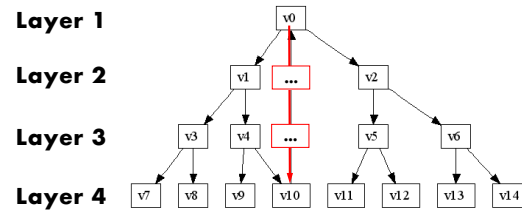
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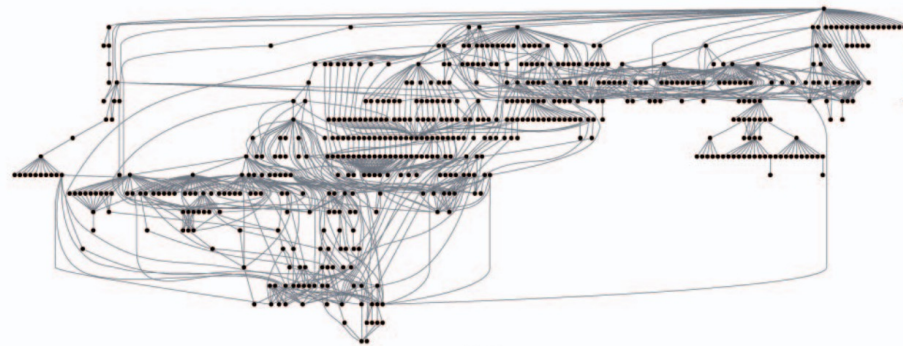
## SUGIYAMA-STYLE GRAPH LAYOUT



- Reverse some edges to remove cycles (if not already a DAG)
- Assign nodes to hierarchy layers → Longest path layering
  - Create dummy nodes to “fill in” missing layers
- Arrange nodes within layer, minimize edge crossings
- Route edges – layout splines if needed

102

## PRODUCES HIERARCHICAL LAYOUT



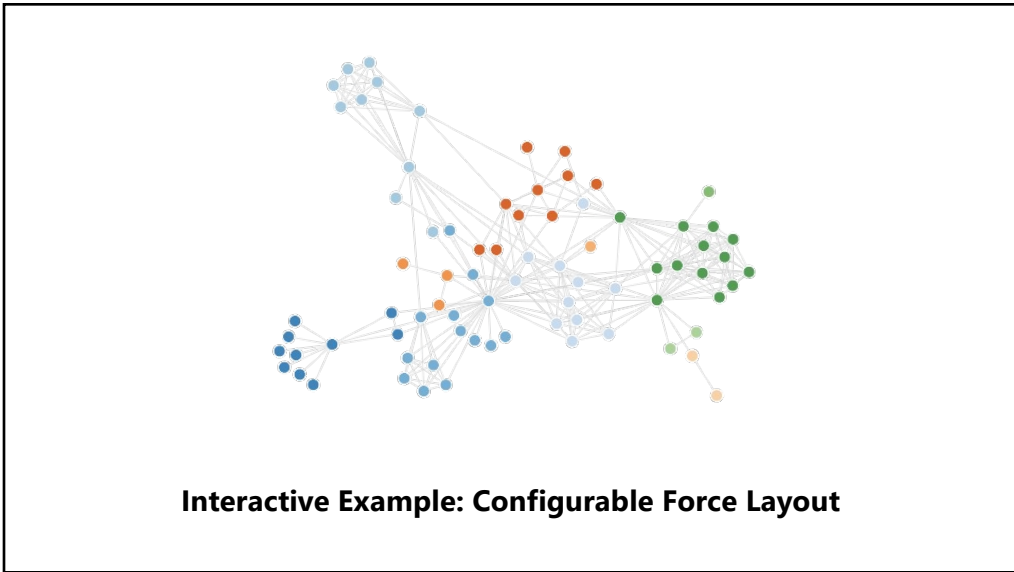
### Sugiyama-style layout emphasizes hierarchy

- However, cycles in the graph may not be as apparent, and hierarchy may mislead
- Long edges can impede perception of proximity

103

# FORCE-DIRECTED LAYOUT

104

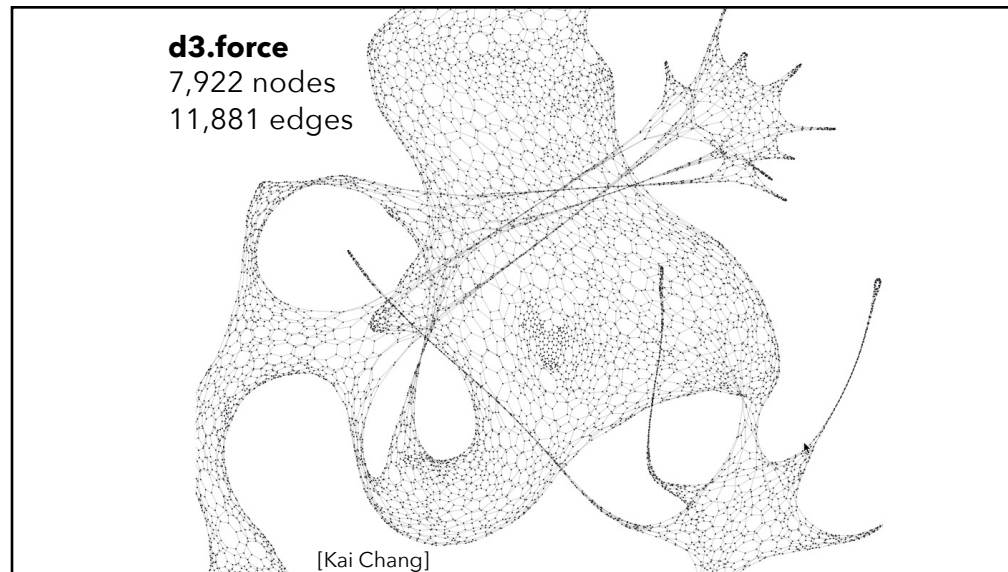


105

# Use the Force!

<http://mbostock.github.io/d3/talk/20110921/>

107



108

## LAYOUT BY PHYSICS SIMULATION

Nodes = charged particles  
with air resistance  
Edges = springs

$$F = q_i * q_j / d_{ij}^2$$

$$F = -b * v_i$$

$$F = k * (L - d_{ij})$$

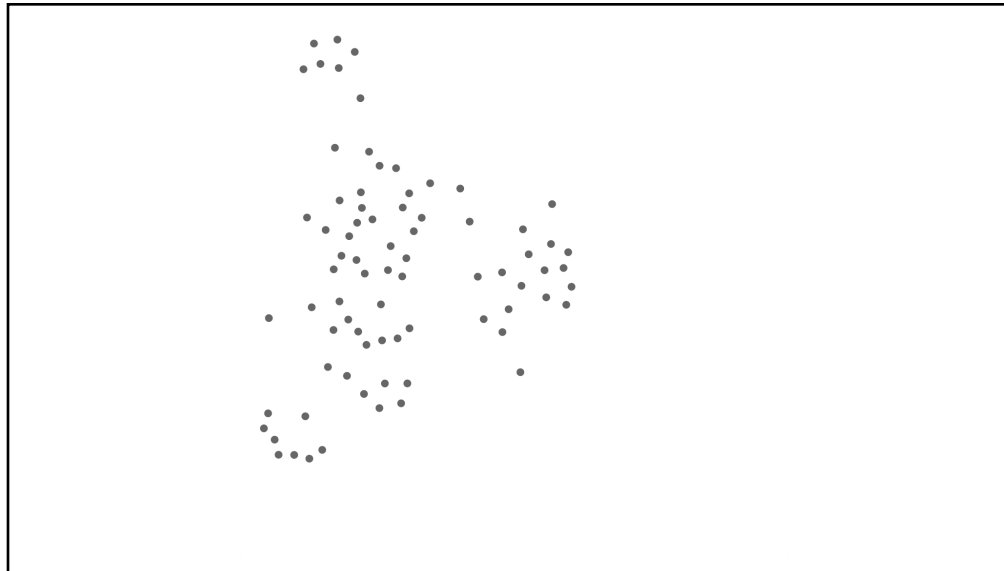
At each timestep, calculate forces acting on nodes.  
Integrate for updated velocities and positions.

D3's force layout uses **velocity Verlet** integration

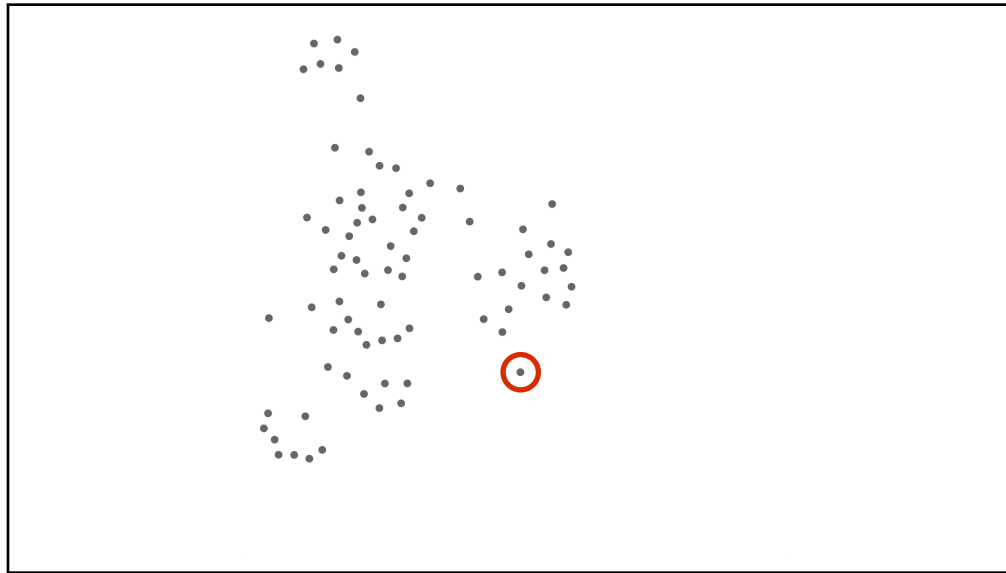
Assume uniform mass  $m$  and timestep  $\Delta t$ :  
 $F = ma \rightarrow F = a \rightarrow F = \Delta v / \Delta t \rightarrow F = \Delta v$

*Forces simplify to velocity offsets!*

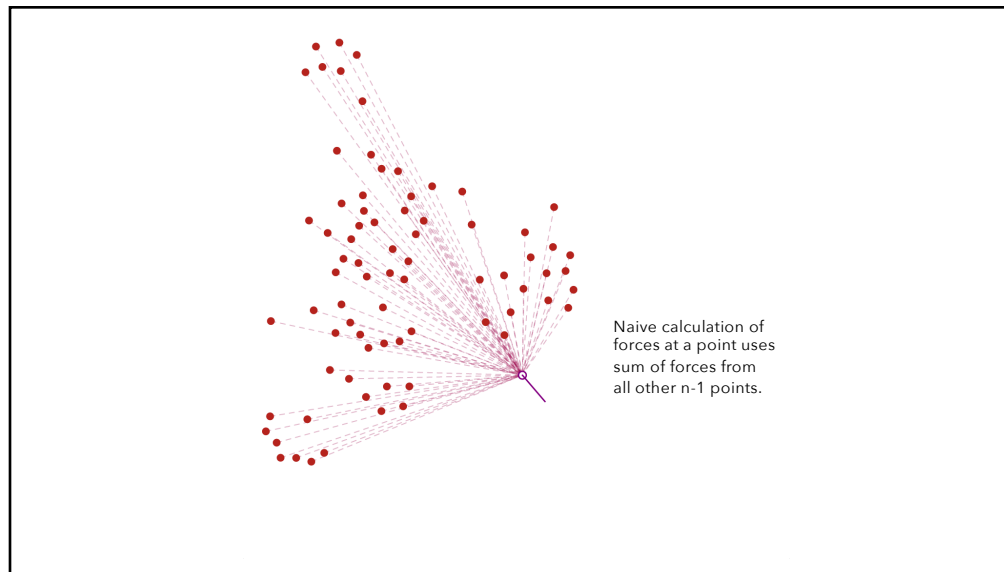
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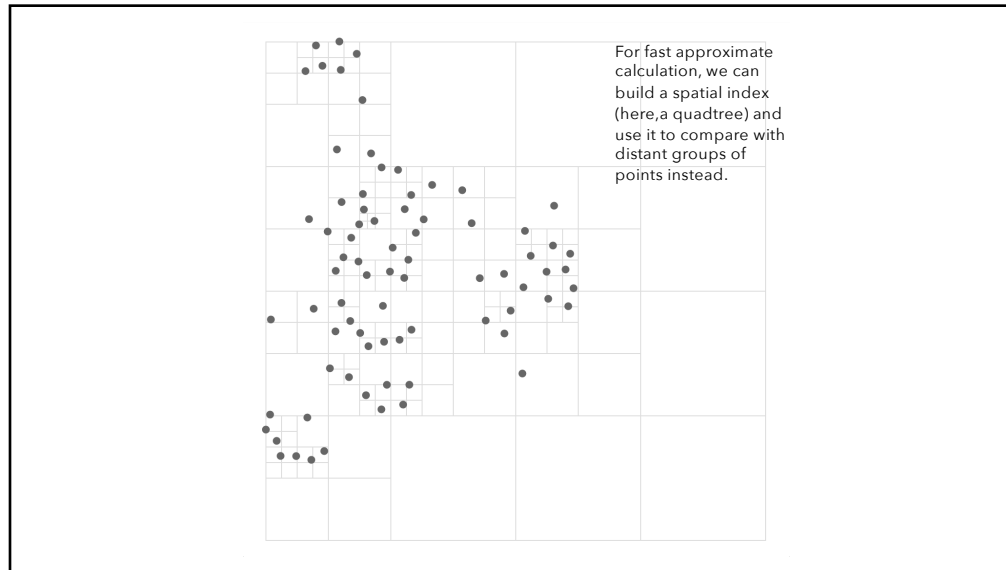
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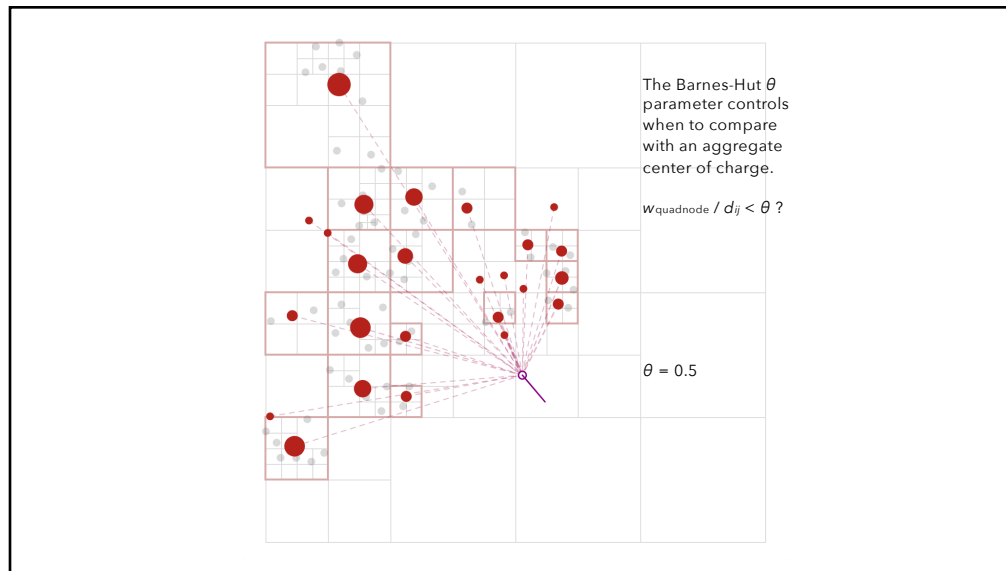
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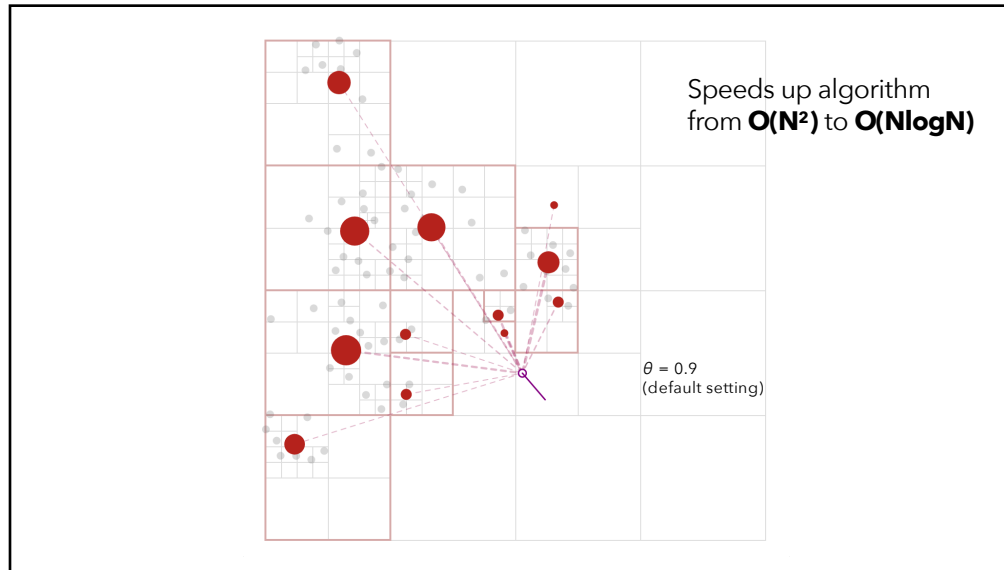
112



113



114



115

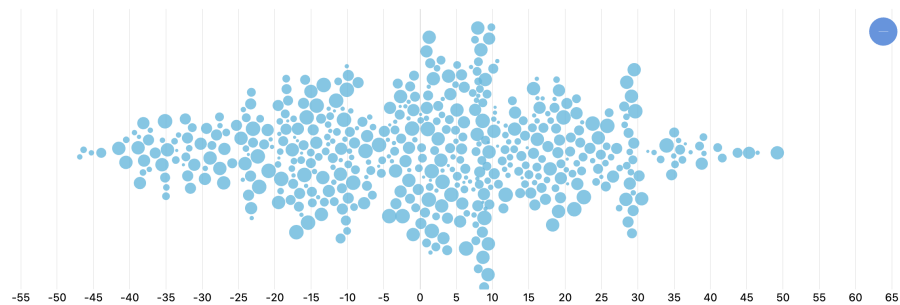
## CUSTOMIZED FORCE LAYOUTS <https://www.amcharts.com/demos/beeswarm/>

Different forces can be composed to create variety of custom layouts

A **beeswarm plot** can be made by combining:

Attractive **X** and **Y** forces to draw nodes of a certain category to a desired point

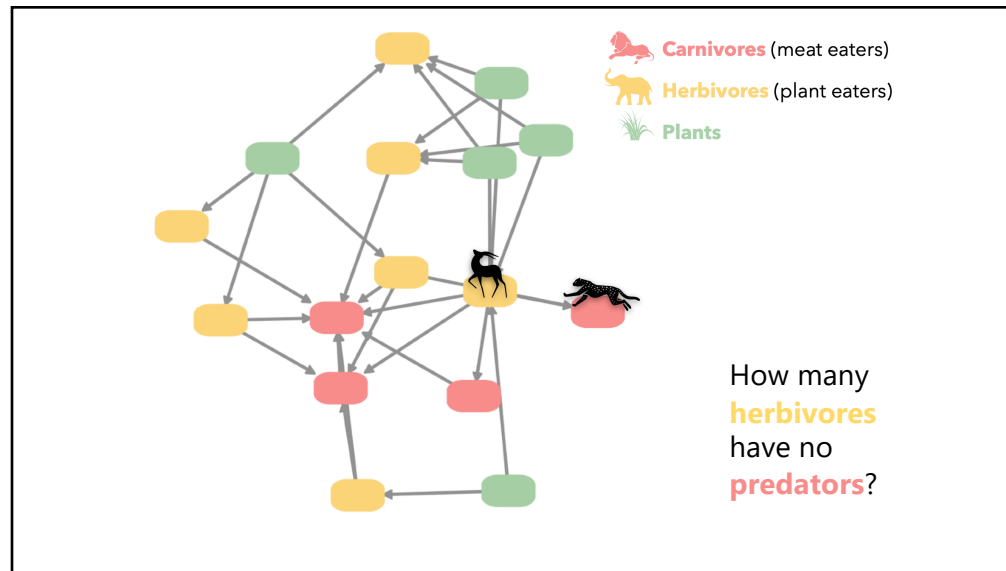
**Collide** force to detect collision & remove overlap



119

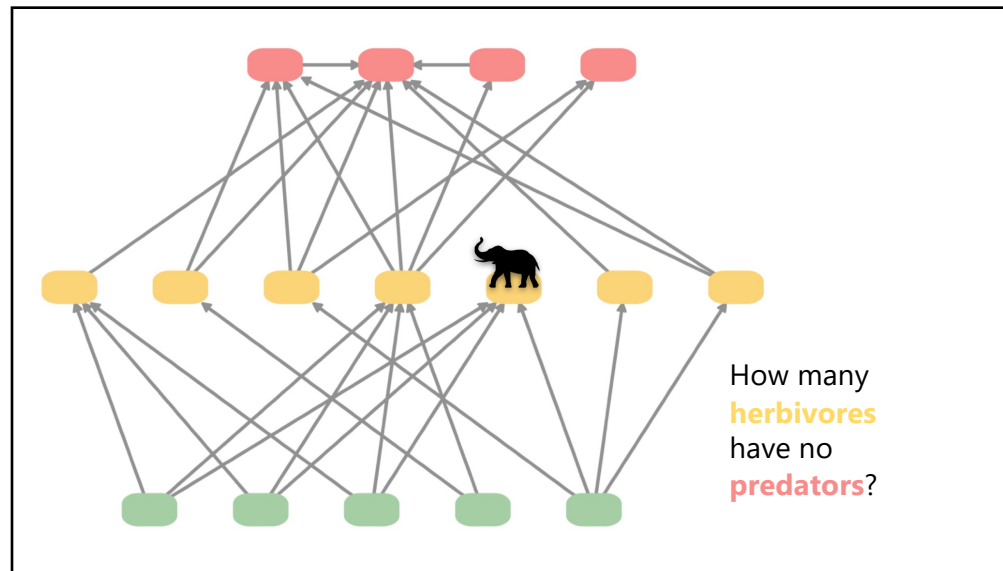
## ATTRIBUTE-DRIVEN LAYOUT

120



121





122

## ATTRIBUTE-DRIVEN LAYOUT

Large node-link diagrams **get messy!**

Can we exploit additional structure?

*Idea:* Use **data fields/attributes** associated with nodes or edges to perform layout (e.g., scatter plot based on node values)

Attributes may also be statistical properties of the graph

Can apply dynamic queries & brushing on attributes/fields to explore...

123

## ATTRIBUTE-DRIVEN LAYOUT

### The “Skitter” Layout

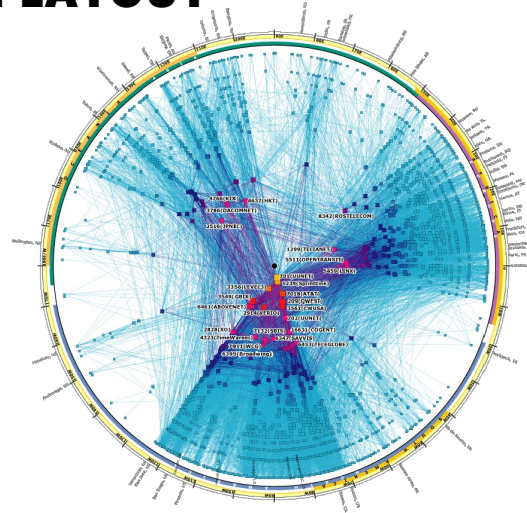
Internet Connectivity  
Radial Scatterplot

**Angle = Longitude**

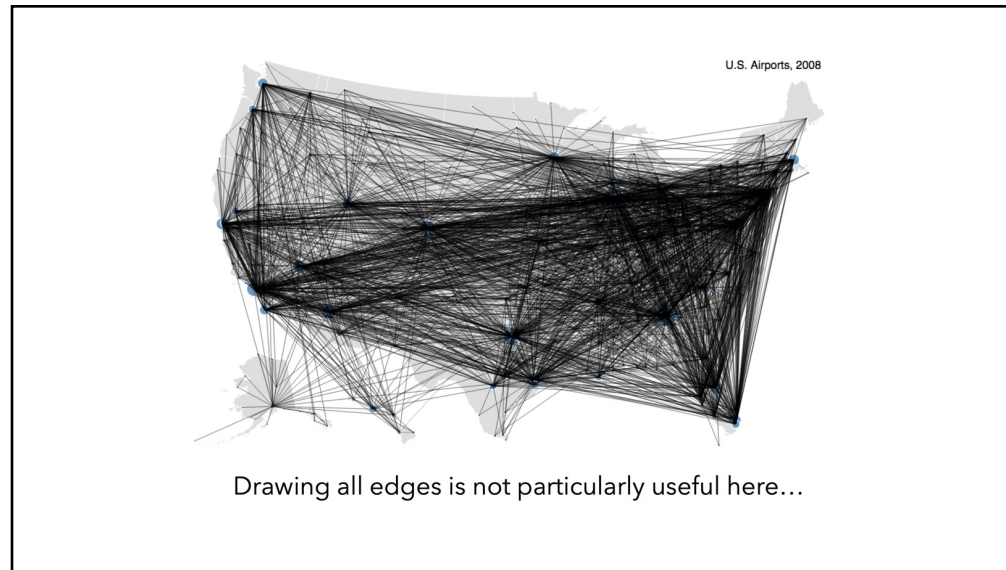
Geography

**Radius = Degree**

# of connections  
(a statistic of the nodes)

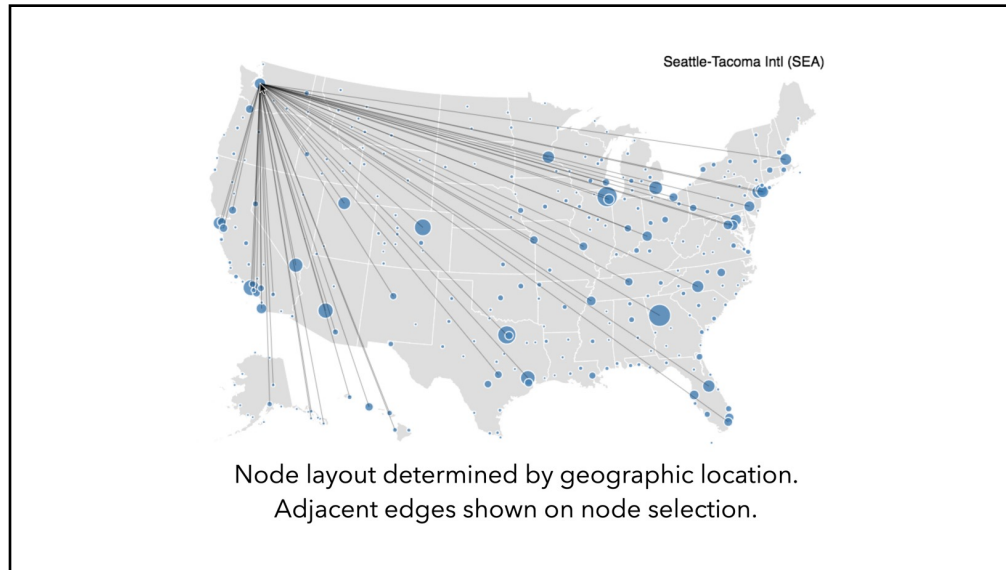


124

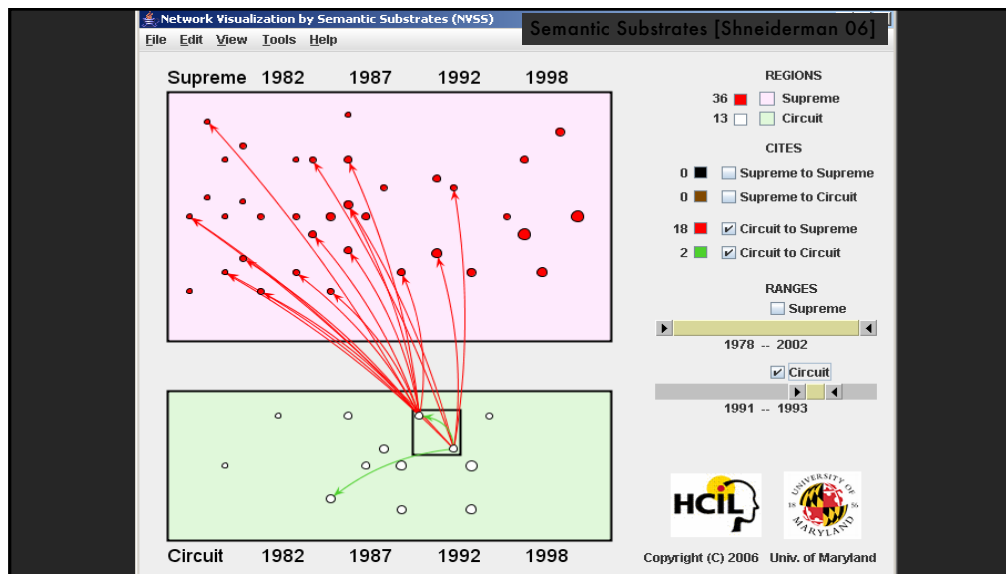


Drawing all edges is not particularly useful here...

125



126



127

## CONSTRAINT-BASED LAYOUT

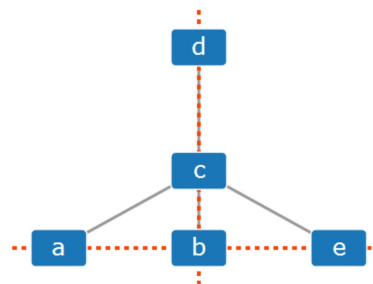
134

## CONSTRAINT-BASED LAYOUT

### Treat layout as an optimization problem

Define layout using an *energy model* along with *constraint equations* the layout should obey

Use optimization algorithms to solve:



### Position Constraints

a must be to the **left** of b

d, c, and b must have the same **x position**

a, b, and e must have the same **y position**

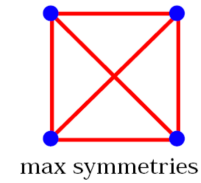
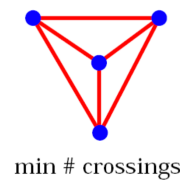
135

## OPTIMIZING AESTHETICS

- Minimize edge crossings
- Minimize area
- Minimize line bends
- Minimize line slopes
- Maximize smallest angle between edges
- Maximize symmetry

### but, can't do it all

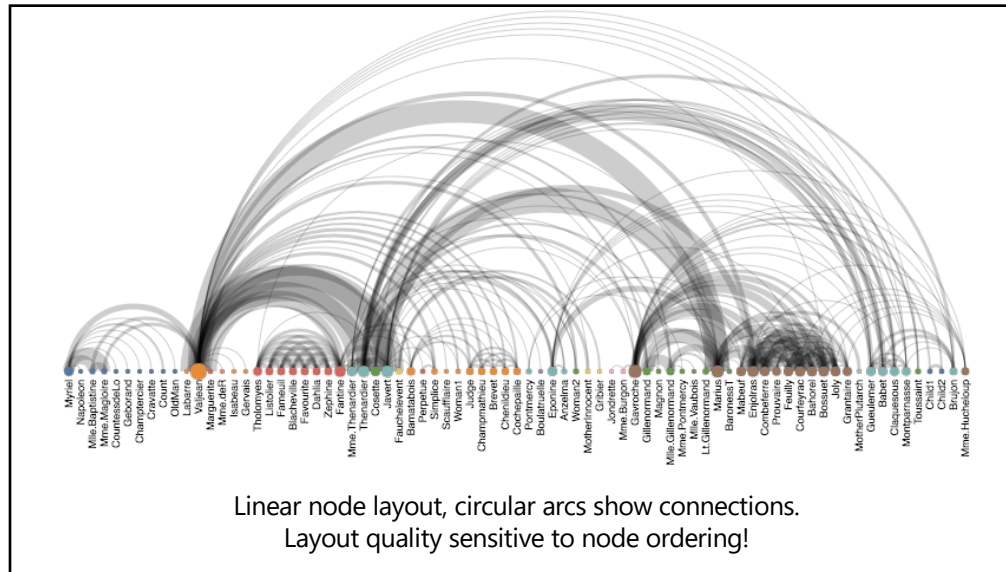
Optimizing these criteria is often NP-Hard, and requires approximations



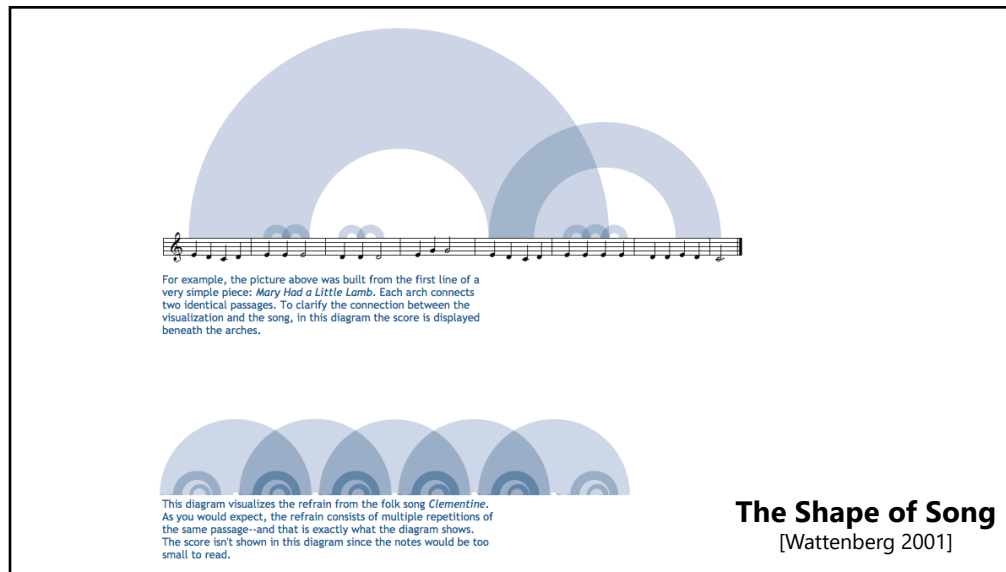
136

## ARC DIAGRAMS

138



139



140

## NODE-LINK GRAPH VISUALIZATION

**Sugiyama-Style Layout** - arranged by depth

**Force-Directed Layout** - physical simulation

**Attribute-Driven Layout** - arranged by value

**Constraint-Based Layout** – optimization

**Arc Diagrams** - aligned layout

141

## NODE-LINK GRAPH VISUALIZATION

**Sugiyama-Style Layout** - arranged by depth

**Good:** Structure-based analysis of hierarchical relationships

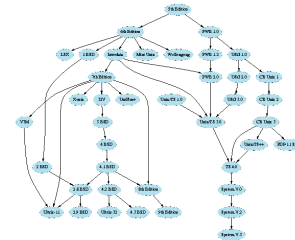
**Bad:** Browsing and path following due to long edges

**Force-Directed Layout** - physical simulation

**Attribute-Driven Layout** - arranged by value

**Constraint-Based Layout** – optimization

**Arc Diagrams** - aligned layout



142

## NODE-LINK GRAPH VISUALIZATION

**Sugiyama-Style Layout** - arranged by depth

**Good:** Structure-based analysis of hierarchical relationships

**Bad:** Browsing and path following due to long edges

**Force-Directed Layout** - physical simulation

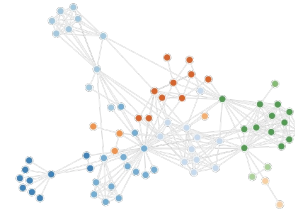
**Good:** Structure-based analysis of closely related elements

**Bad:** Browsing and summarization of dense networks

**Attribute-Driven Layout** - arranged by value

**Constraint-Based Layout** - optimization

**Arc Diagrams** - aligned layout



143

## NODE-LINK GRAPH VISUALIZATION

**Sugiyama-Style Layout** - arranged by depth

**Good:** Structure-based analysis of hierarchical relationships

**Bad:** Browsing and path following due to long edges

**Force-Directed Layout** - physical simulation

**Good:** Structure-based analysis of closely related elements

**Bad:** Browsing and summarization of dense networks

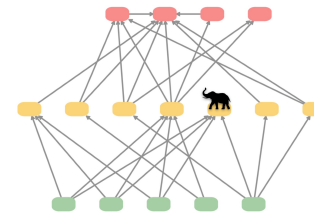
**Attribute-Driven Layout** - arranged by value

**Good:** Enables attribute-based analysis tasks

**Bad:** Difficult to design layouts appropriate to revealing attributes and network structure

**Constraint-Based Layout** - optimization

**Arc Diagrams** - aligned layout



144



## NODE-LINK GRAPH VISUALIZATION

### Sugiyama-Style Layout - arranged by depth

**Good:** Structure-based analysis of hierarchical relationships

**Bad:** Browsing and path following due to long edges

### Force-Directed Layout - physical simulation

**Good:** Structure-based analysis of closely related elements

**Bad:** Browsing and summarization of dense networks

### Attribute-Driven Layout - arranged by value

**Good:** Enables attribute-based analysis tasks

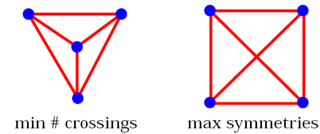
**Bad:** Difficult to design layouts appropriate to revealing attributes and network structure

### Constraint-Based Layout – optimization

**Good:** Graph layout based on structural/aesthetic properties

**Bad:** Difficult to select appropriate constraints

### Arc Diagrams - aligned layout



145

## NODE-LINK GRAPH VISUALIZATION

### Sugiyama-Style Layout - arranged by depth

**Good:** Structure-based analysis of hierarchical relationships

**Bad:** Browsing and path following due to long edges

### Force-Directed Layout - physical simulation

**Good:** Structure-based analysis of closely related elements

**Bad:** Browsing and summarization of dense networks

### Attribute-Driven Layout - arranged by value

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**Bad:** Difficult to design layouts appropriate to revealing attributes and network structure

### Constraint-Based Layout – optimization

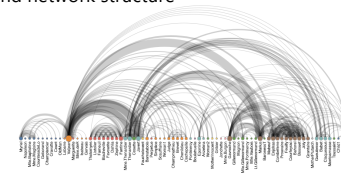
**Good:** Graph layout based on structural/aesthetic properties

**Bad:** Difficult to select appropriate constraints

### Arc Diagrams - aligned layout

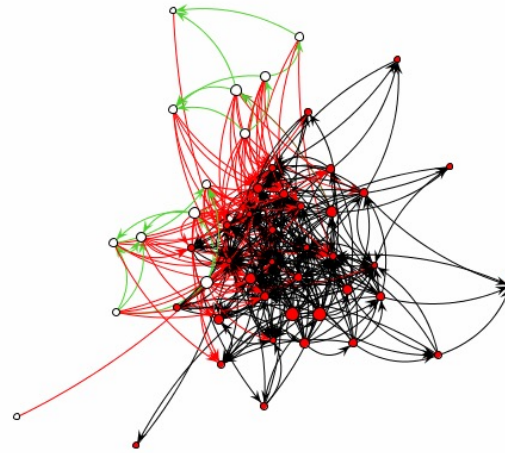
**Good:** Summarization and comparison of overall structure

**Bad:** Order matters for node layout; Structure-based and path following



146

## LIMITATIONS OF NODE-LINK LAYOUTS



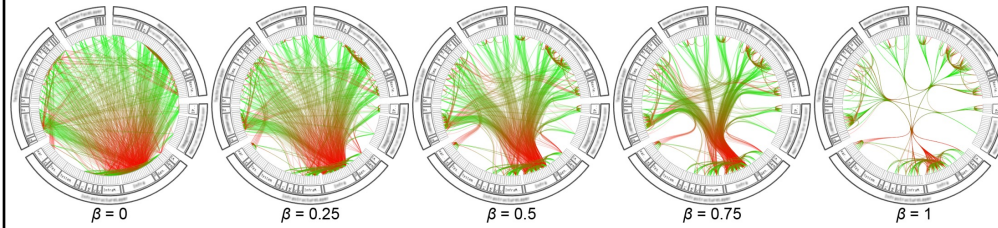
Edge crossings and occlusions!  
Poor scalability...

147

## HIERARCHICAL EDGE BUNDLING

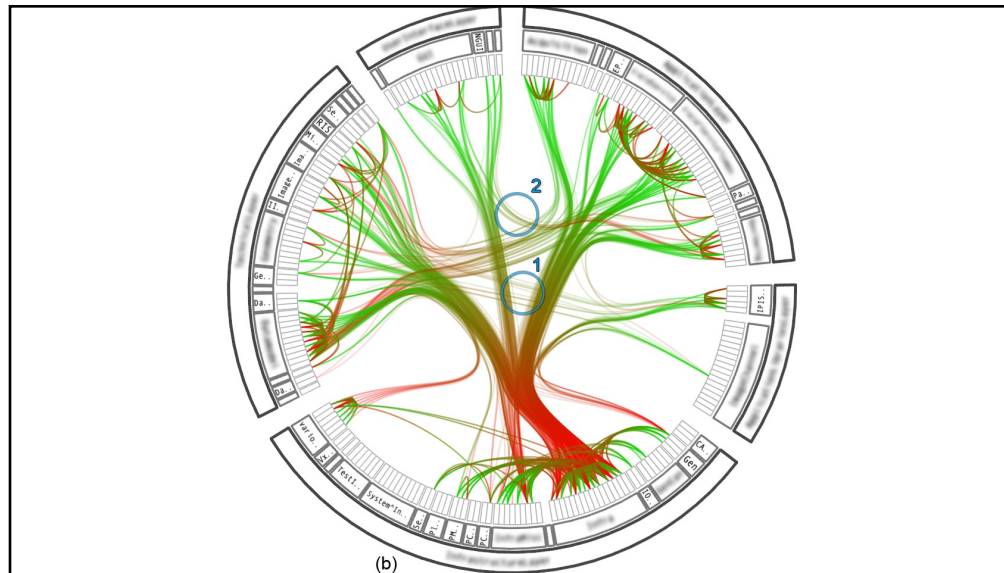
148

## HIERARCHICAL EDGE BUNDLING



Given a tree with additional *adjacency* edges (usually between leaves)  
 Bundle edges with varying amounts of tension – helping to reveal common connections between subtrees

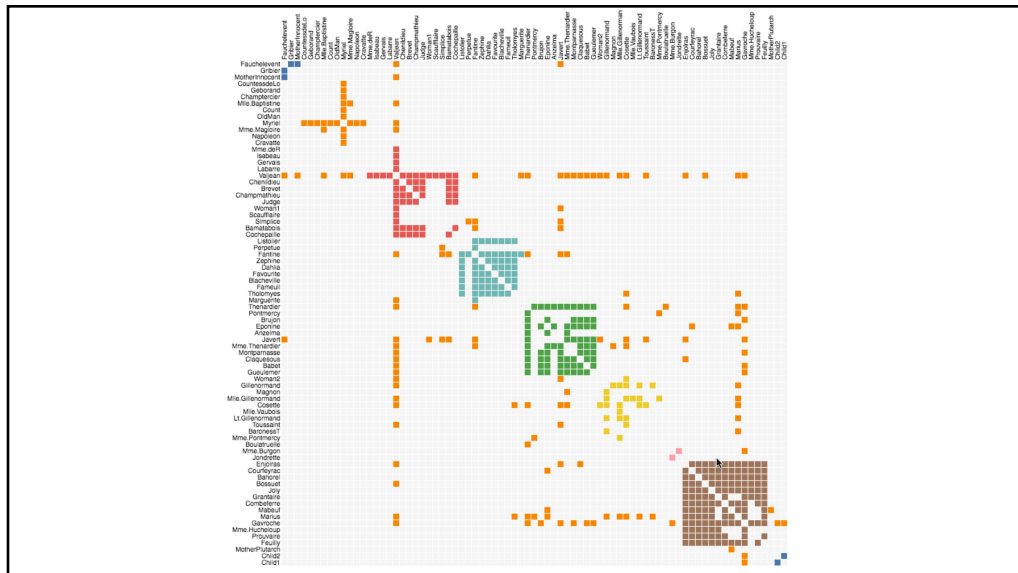
150



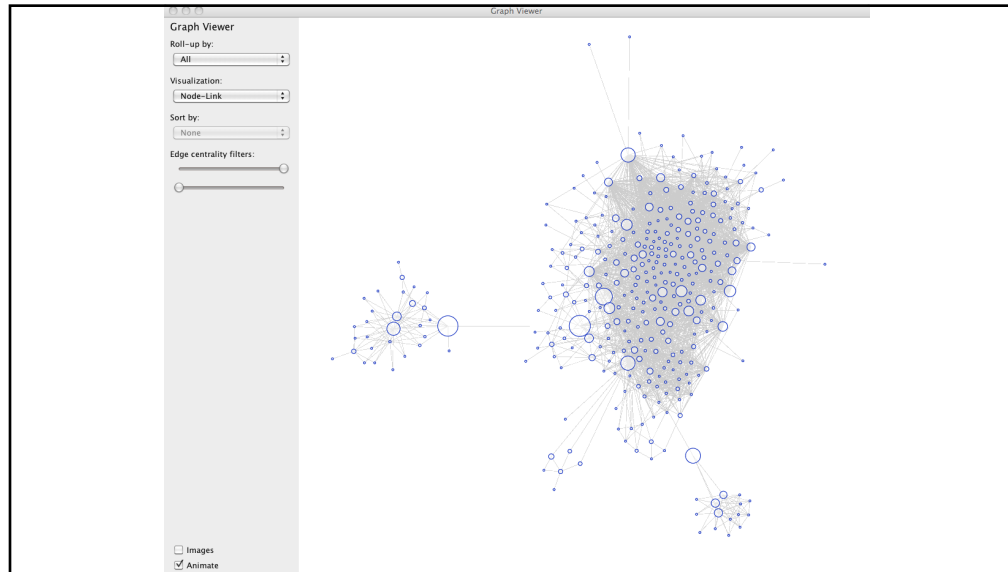
151

# MATRIX DIAGRAMS

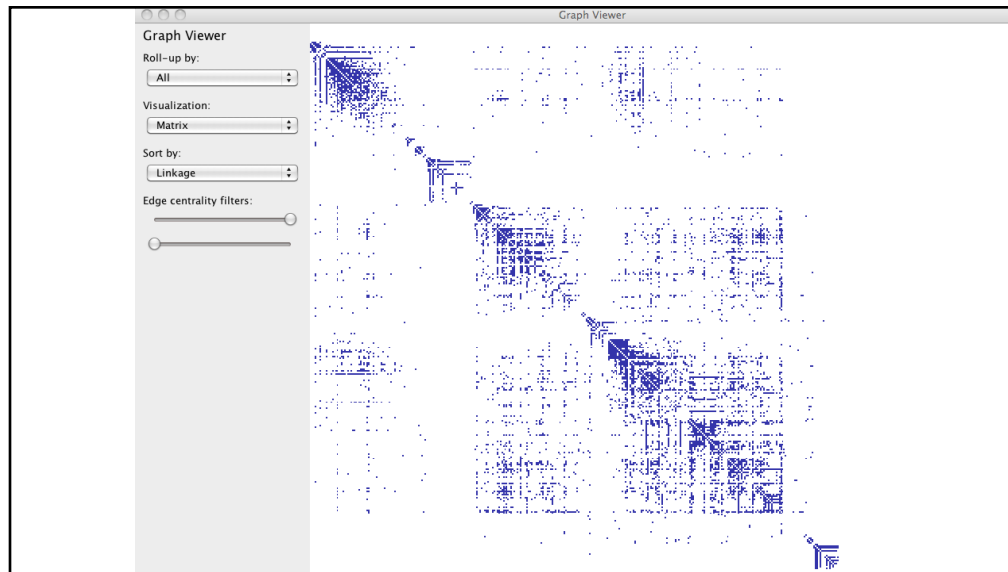
156



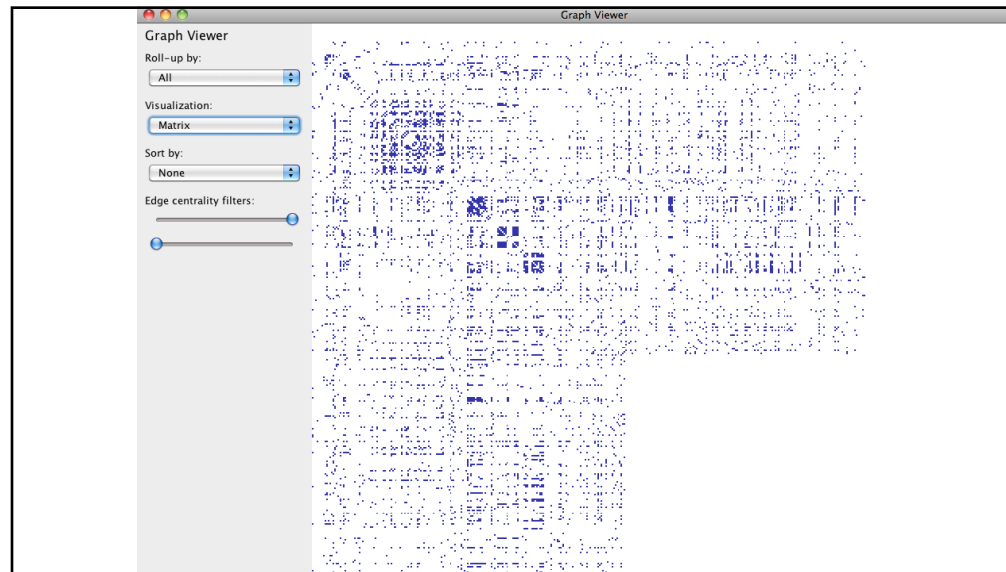
157



158



159



160

## SUMMARY: TREES AND NETWORKS

### Tree Layout

Indented / Node-Link / Enclosure / Layers

Focus+Context techniques for scale

### Graph Layout

Sugiyama Layout

Force-Directed Layout

Attribute-Driven Layout

Constraint Layout

Arc Diagrams

Matrix Diagrams

161