15 Views of a Node-Link Graph: An InfoVis Portfolio

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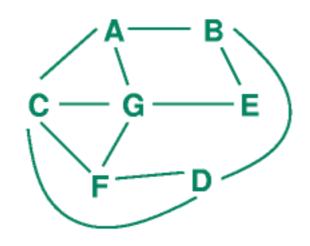
University of British Columbia Department of Computer Science

Information Esthetics Lecture Series One 14 July 2005

15 Views of a Node-Link Graph: An InfoVis Portfolio

node-link graph

- common abstraction: nodes connected by edges
- trees are special case: hierarchy with no cycles





infovis: information visualization

- visual representation of abstract data computer-based: interactivity possible
- · help human perform some task more effectively

Information Esthetics

manifesto

 information content can enhance esthetic experience and esthetic consideration can enhance information content

successful infovis as example

 design guidelines for visual encoding as interplay of perception, cognition, esthetics, and data/task characterization

Visual Channels

visual attribute of geometric mark

· position, color, size, shape, orientation, ...

separable vs. integral



color position

color motion

color

shape orientation y-size yellow-blue

size x-size red-green

Outline

Introduction

15 Views

- Traditional Graphs
- Nontraditional Representations
- Focus+Context Trees

Wrapup

1: Edge List

data: semantic network from Hofstadter book Godel, Escher, Bach

nodes: topics

· links: discussion of ideas together in book

Turing – Halting problem Halting problem - Unpredictably long Halting problem - Infinity searches Infinity - Paradoxes BlooP and FlooP – Unpredictably long Paradoxes - Lewis Carroll searches Infinity - Lewis Carroll BlooP and FlooP - Recursion Tarski - Truth vs. provability Infinity - Unpredictably long searches Tarski - Epimenides Infinity – Recursion Tarski - Undecidability Infinity – Zeno Paradoxes - Self-ref Infinity – Paradoxes Lewis Carroll - Zeno Epimenides - Tarski Epimenides – Paradoxes Lewis Carroll – Wordplay Epimenides – Self-ref Halting problem - Decision procedures BlooP and FlooP - AI [...]

strengths

easy to create

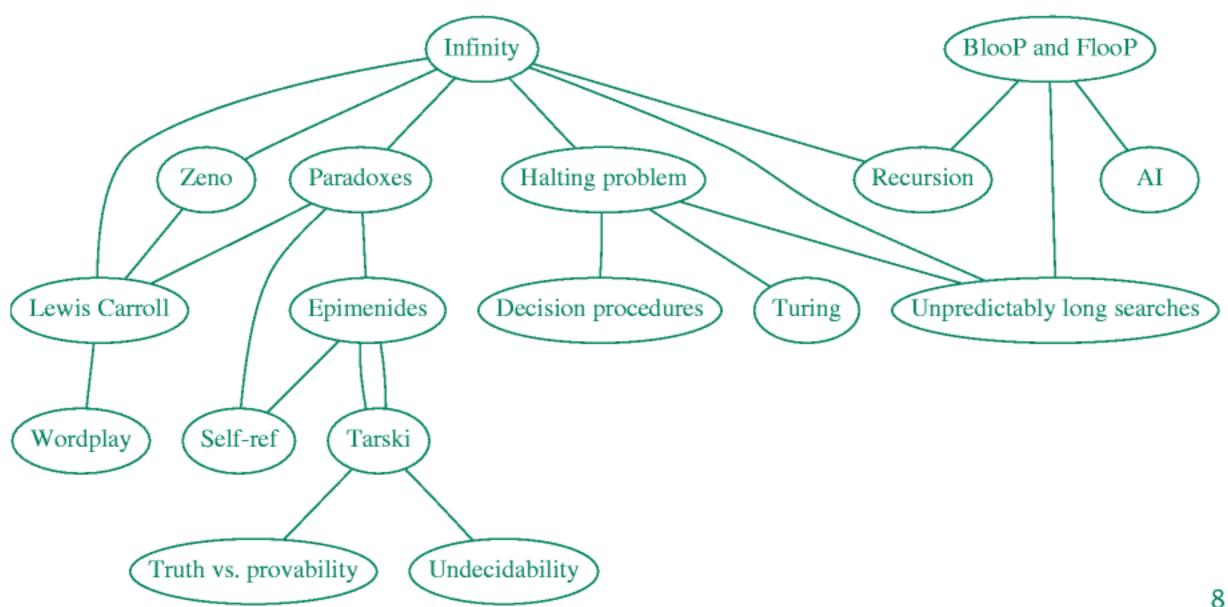
weaknesses

- · requires too much memory and cognition
- · does not exploit human perceptual system

Visual External Representation

read off answers from node-link graph drawing

- connections drawn between nodes
- offload cognition to visual system



2: Hand-Drawn

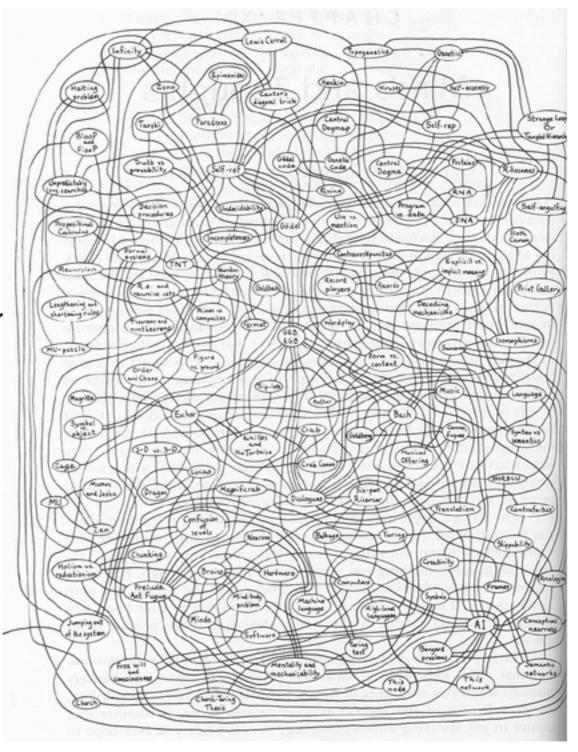
data: GEB semantic network

strengths

- high information density ratio of marks to whitespace foreground vs. background layer
- subtleties of spatial layout

weaknesses

hours or days to create



3: Dot

data: semantic network

automatically compute positions for nodes, edges

strengths

- fast: one second to create
- careful routing of curved edges

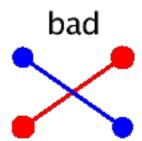
weaknesses

low information density can't read labels



[Gansner, Koutsofois, North and Vo. A Technique for Drawing Directed Graphs. IEEE Trans. Software Engineering, 19(3):21₁₀229]

good



minimize

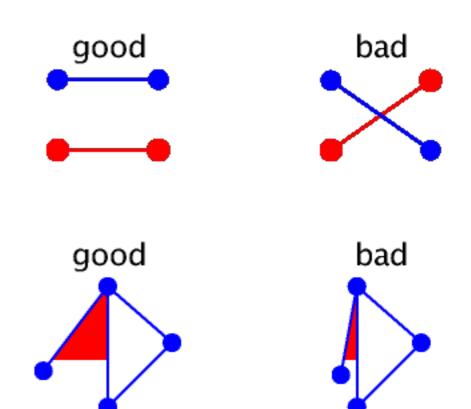
· crossings, area, bends/curves

minimize

· crossings, area, bends/curves

maximize

· angular resolution, symmetry

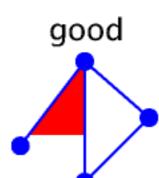


crossings, area, bends/curves

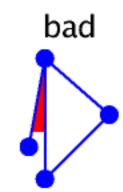
minimize

maximize

· angular resolution, symmetry



good



bad

most criteria individually NP-hard

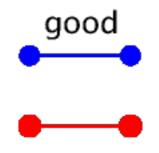
- cannot just compute optimal answer
- · heuristics: try to find something reasonable

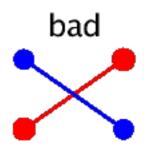
minimize

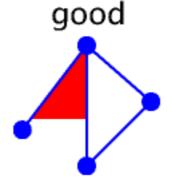
crossings, area, bends/curves

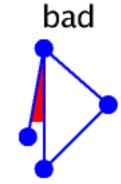


· angular resolution, symmetry





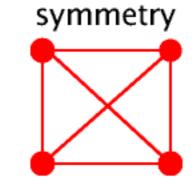


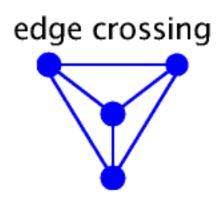


most criteria individually NP-hard

- · cannot just compute optimal answer
- heuristics: try to find something reasonable

criteria mutually incompatible





[Ware, Purchase, Colpys, and McGill. Cognitive Measures of Graph Aesthetics. Information Visualization 1(2):103-110, Palgrave 2002] [Brandenburg. Nice Drawings of Graphs are Computationally Hard. Visualization in Human-Computer Interaction, Springer Verlag 1988]

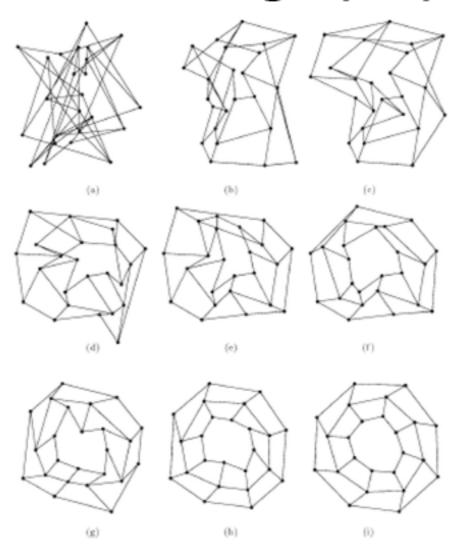
4: Force-Directed Placement

nodes: repel like magnets

edges: attract like springs

· start from random positions, run to convergence

encoding: geometric for graph proximity



strengths

- · intuitive model
- many mathematical approaches

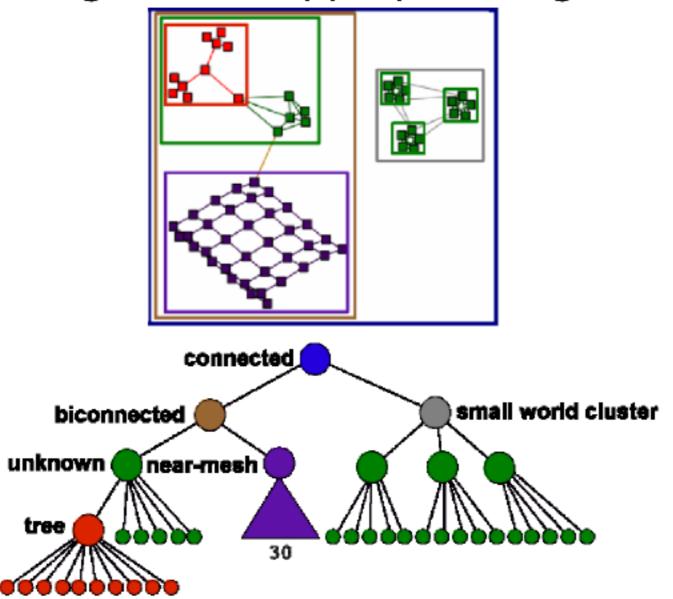
weaknesses

does not scale to large datasets

5: TopoLayout

multilevel decomposition and layout

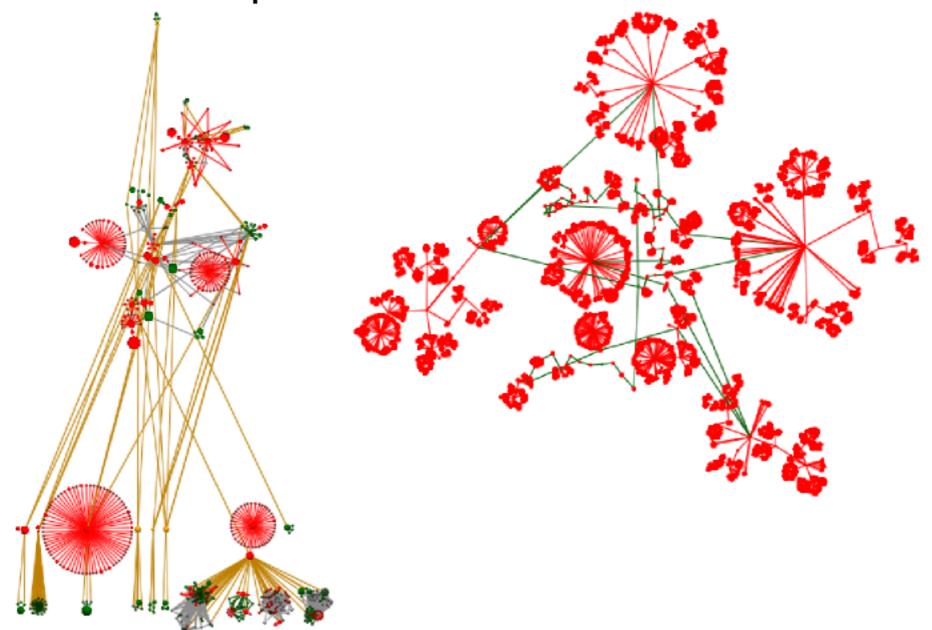
- automatic detection of topological features chop into hierarchy of manageable pieces
- · lay out using feature-appropriate algorithms



Multilevel Hierarchies

data: web sites, network backbones

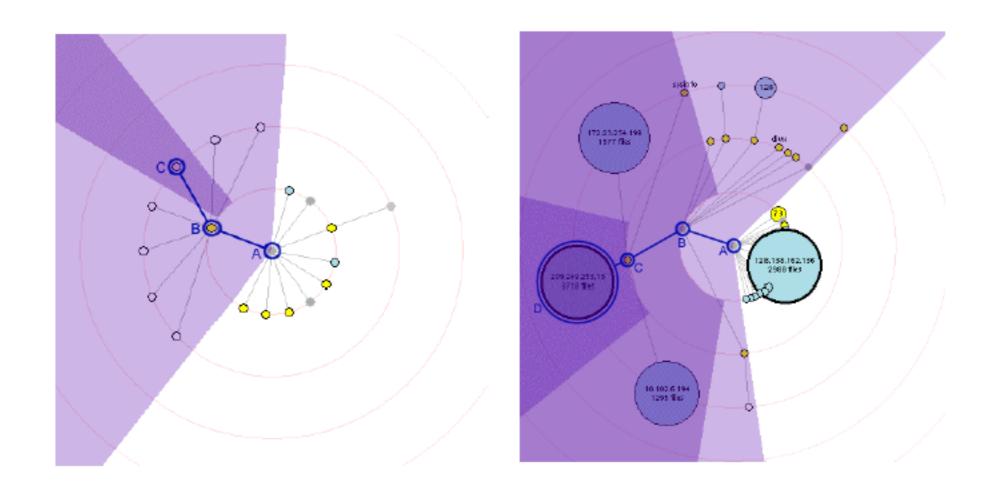
- · strengths: handles large class of graphs
- · weaknesses: poor if no detectable features



6: Animated Radial Layouts

dynamic graphs that change over time

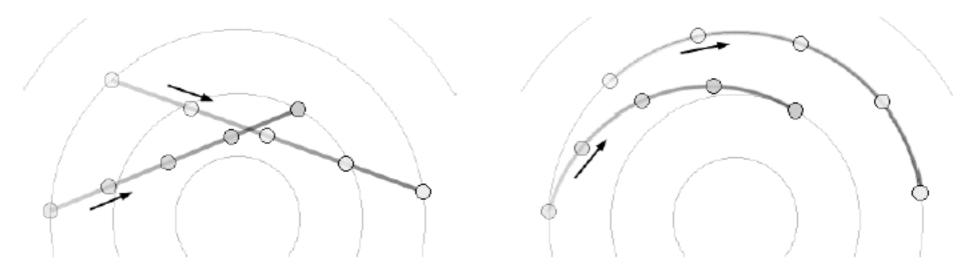
- · minimize visual changes
- stay true to current dataset structure [video: www.sims.berkeley.edu/~ping/gv]



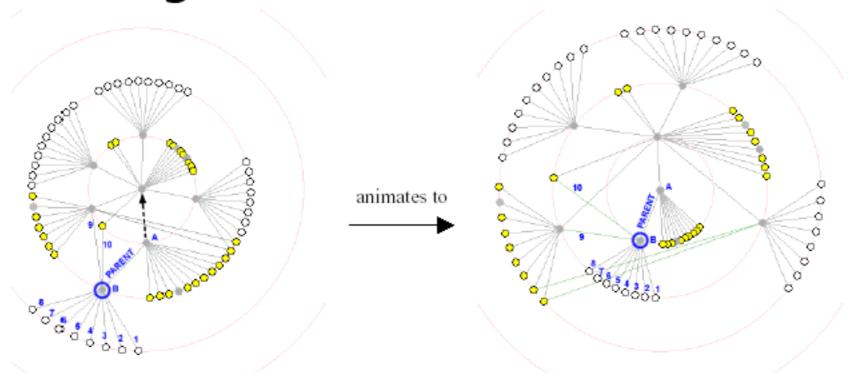
[Yee, Fisher, Dhamija, and Hearst. Animated Exploration of Graphs with Radial Layods. Proc. InfoVis 2001. bailando.sims.berkeley.edu/papers/infovis01.htm]

Animation

polar interpolation



maintain neighbor order



[Yee, Fisher, Dhamija, and Hearst. Animated Exploration of Graphs with Radial Layo 20. Proc. InfoVis 2001. bailando.sims.berkeley.edu/papers/infovis01.htm]

strengths

· smoother transtions

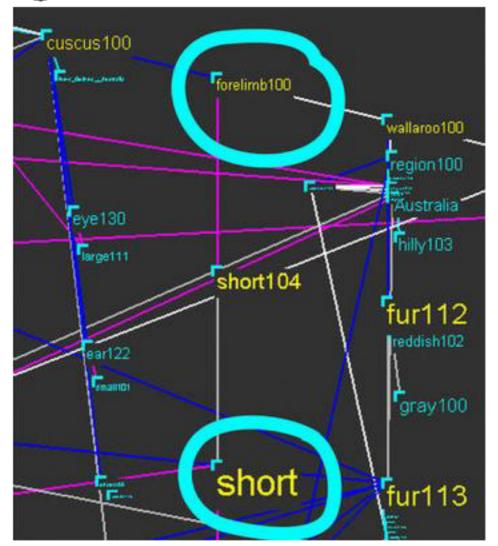
weaknesses

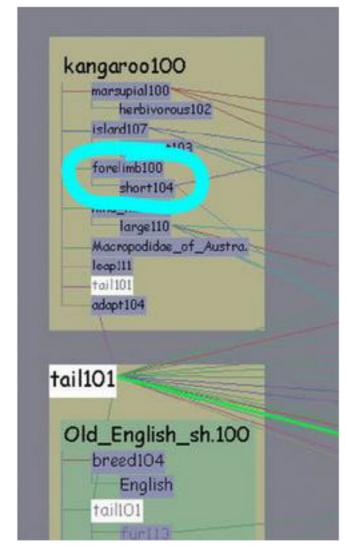
· not scalable to large datasets

7: Constellation

data: semantic network from dictionary entry

- nodes: English words, links: used together in entry information density
 - · design tradeoff with visual salience





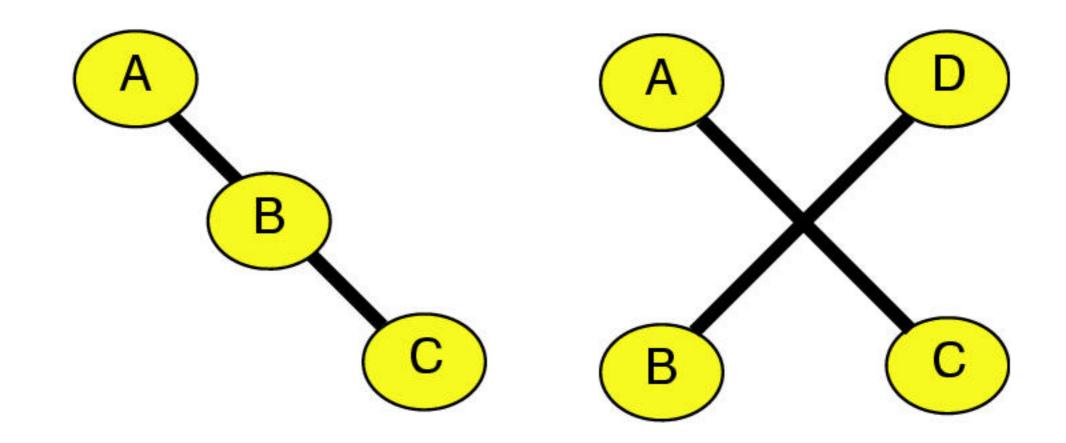
[Munzner, Guimbretiere and Robertson. Constellation: A Visualization Tool For Linguizzic Queries from MindNet. Proc. InfoVis 1999. graphics.stanford.edu/papers/const]

Traditional Layout

avoid crossings

· considered "aesthetic criterion"

reason: avoid false attachments



ambiguity

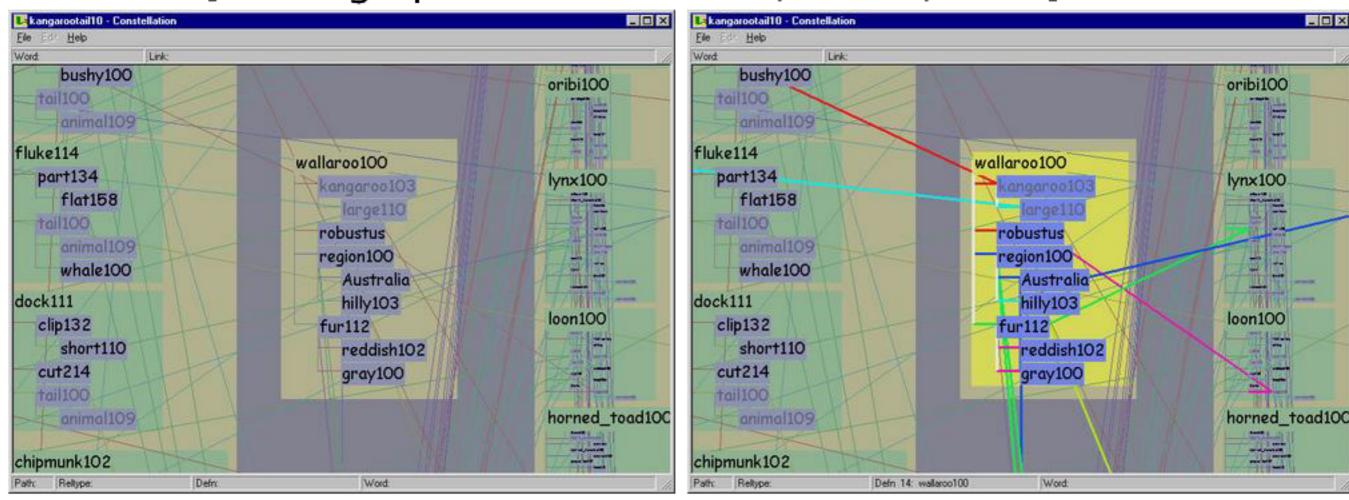
artifact salience

[graphics.stanford.edu/papers/munzner_thesis/html/node10.html#tradlayoutfig²³

Selective Emphasis

highlight sets of boxes and edges

- additional perceptual channels based on interaction avoid perception of false attachments
 - avoid hidden state
 - [video: graphics.stanford.edu/videos/const]



[graphics.stanford.edu/papers/munzner_thesis/html/node10.html#selemphfig]24

strengths

- highly specialized
- good information density in final version
- perceptual layering successful

weaknesses

- highly specialized
- custom system design is expensive

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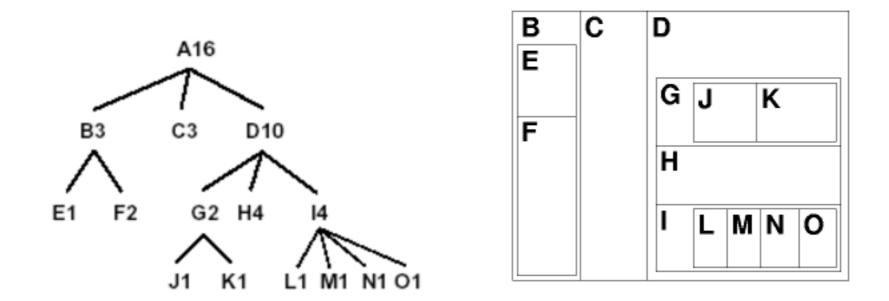
Wrapup

8: Treemaps

data: filesystems, stock performance

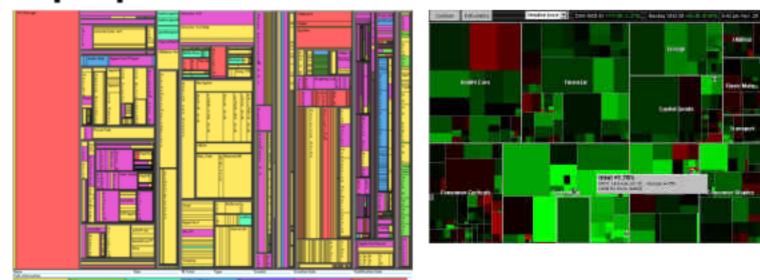
show structure with containment not connection

size according to node attribute



[Johnson and Shneiderman. Treemaps: A Space-Filling Approach to the Visualization of Hierarchical Information Structures. Proc. IEEE Visualization 1991.]

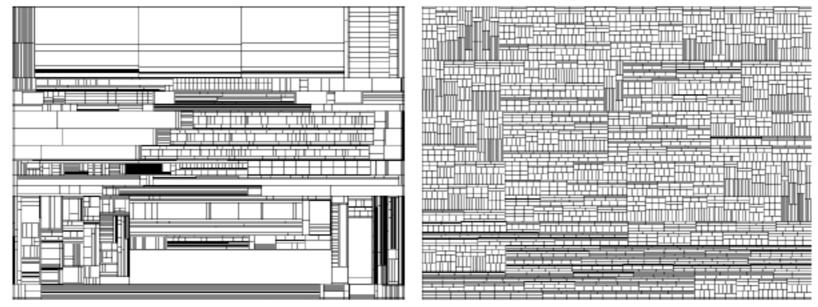
strength: popout for extreme attributes



[Johnson and Shneiderman. Treemaps: A Space-Filling Approach to the Visualization of Hierarchical Information Structures. Proc. IEEE Visualization 1991.]

[www.smartmoney.com/marketmap]

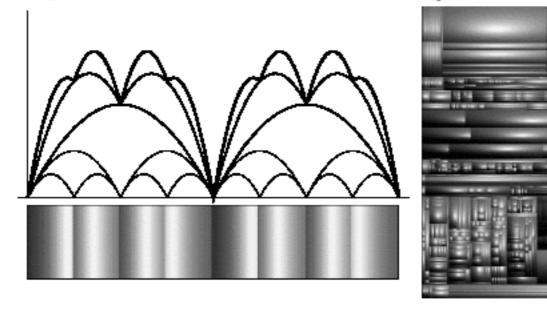
weaknesses: difficulties seeing structure

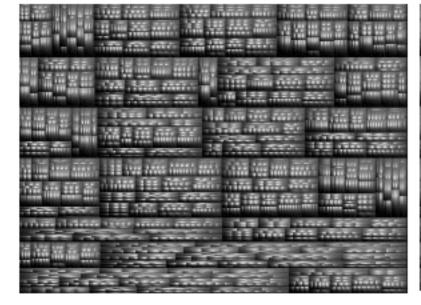


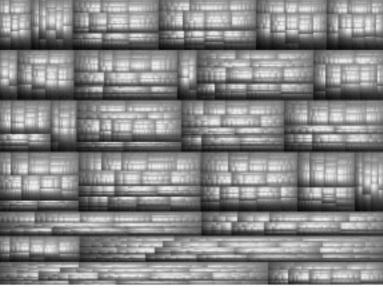
9: Cushion Treemaps

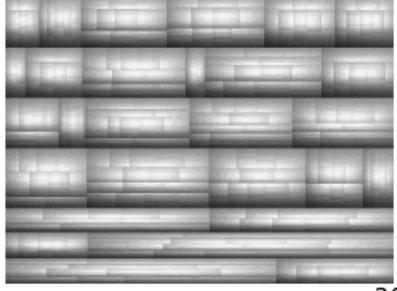
data: filesystems show structure with shading

· scale parameter controls global vs. local









29

strengths

- shows more topological structure than plain treemaps
- keeps power to show attribute outliers
- · allows color to be used to encode other info

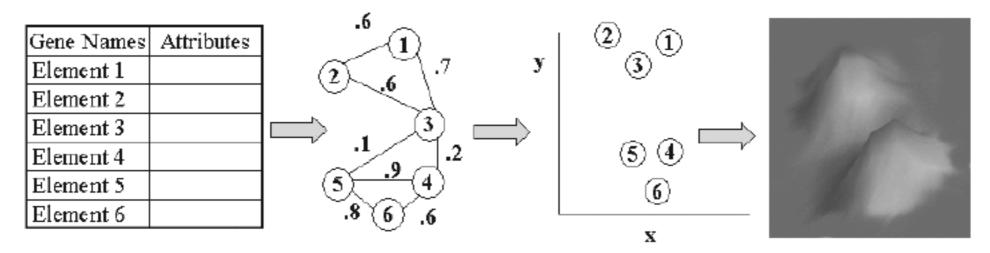
weaknesses

 still considerably worse than node-link representation for showing topological structure

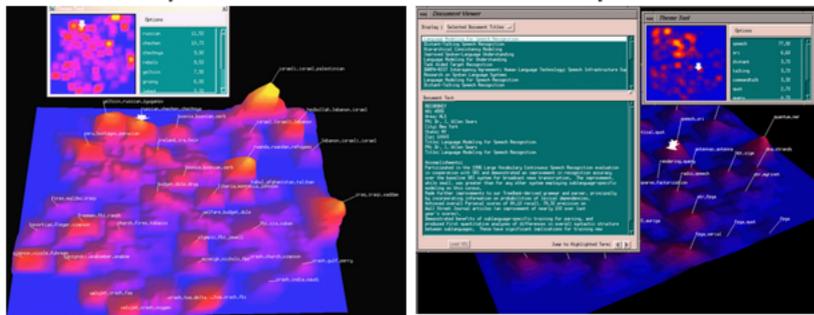
10: Themescapes

data: news stories, gene expression

from graph to terrain



Davidson et al. Cluster Stability and the Use of Noise in Interpretation of Clustering. InfoVis 01



[Wise et al. Visualizing the non-visual: spatial analysis and interaction with information from text documents. Proc. InfoVis 1995. www.pnl.gov/infoviz/graphics.html]

strengths

- terrain model intuitive for people
- good for overview

weaknesses

 possibly misleading implication of continuous data typically made from discrete samples

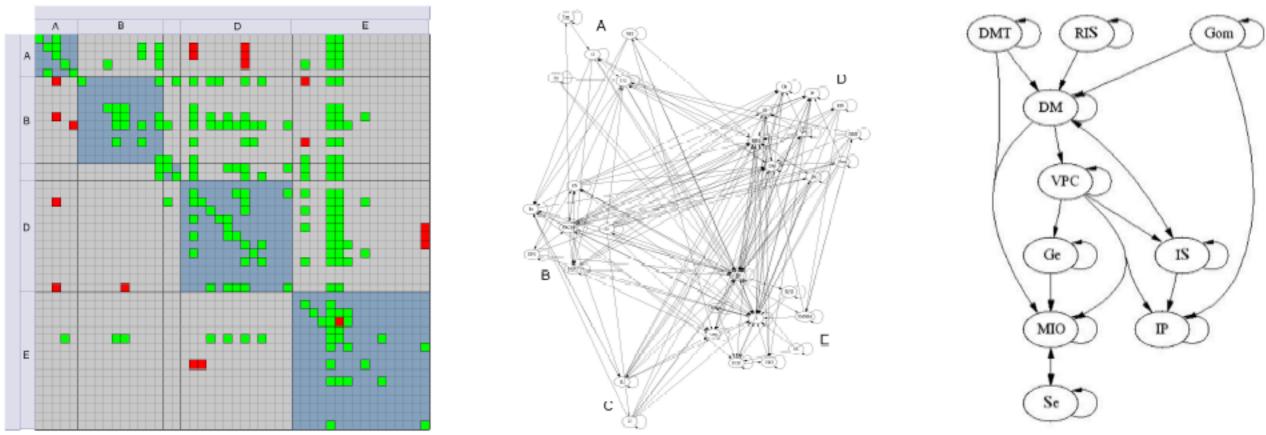
11: Multilevel Call Matrices

data: large software project

link matrix vs. node-link network

force-directed matrix

layered subset (dot)

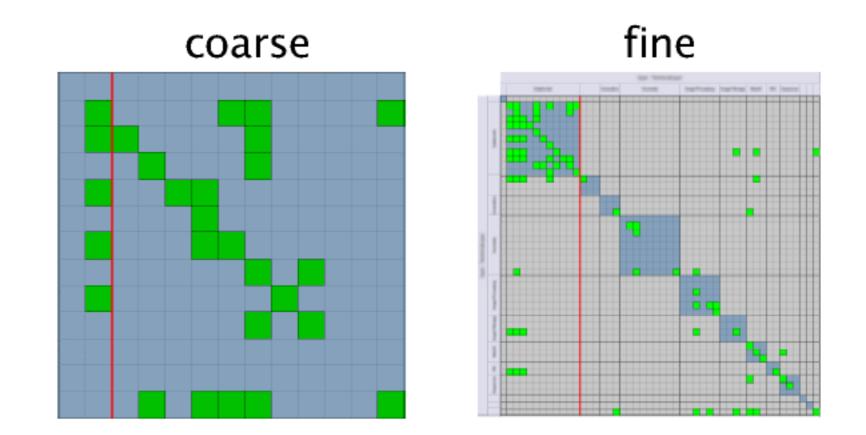


- position: box shows link between nodes in row/column
- color: calls not in specification in red

[van Ham. Using Multilevel Call Matrices in Large Software Projects. Proc. InfoVis 20@3]

Abstraction Levels

matrices: uniform, recursive, stable



[van Ham. Using Multilevel Call Matrices in Large Software Projects. Proc. InfoVis 2003]

strengths: tasks successfully supported

- spotting unwanted calls in implementation but not specification
- previous summary shown to be incomplete

weaknesses

matrix views poor for some tasks

[Ghoniem, Fekete, and Castagliola. A Comparison of the Readability of Graphs Using Node-Link and Matrix-Based Representations. Proc. InfoVis 2004]

Outline

Visual Encoding

15 Views

- Traditional Graphs
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- Focus+Context Trees

Wrapup

Focus+Context

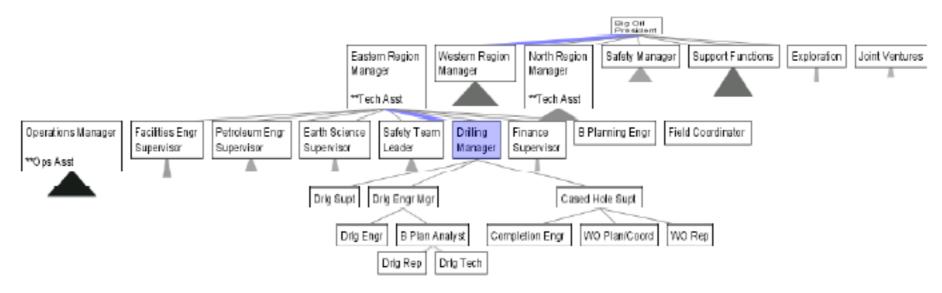
combine overview, details into integrated view

- · vs. single detail view
- · vs. multiple linked windows

12: SpaceTree

data: org charts, species relationships interaction: expand/contract

· [demo: www.cs.umd.edu/hcil/spacetree]



strengths

animated transitions easy to follow weakness

cannot have multiple areas of focus

13: 2D Hyperbolic Trees

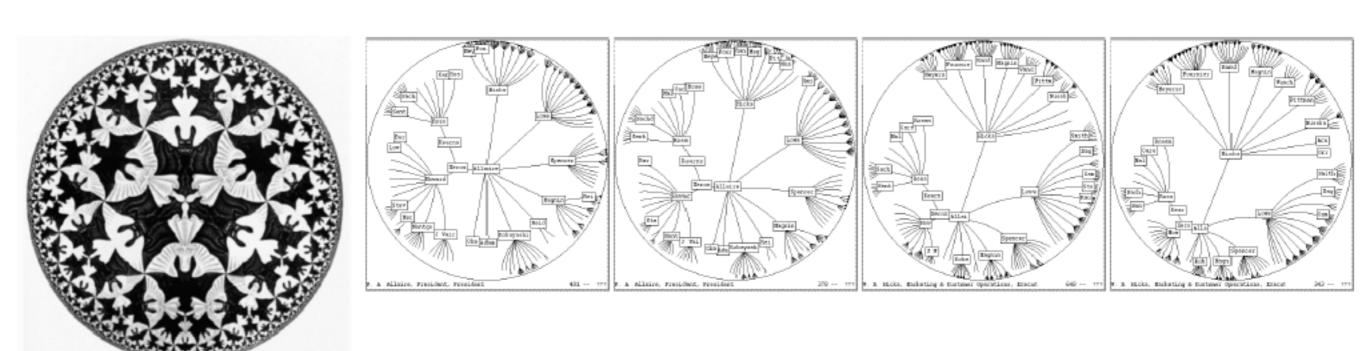
data: org charts, web sites

node: document

link: hyperlink between pages

carefully chosen distortion

- · fisheye effect: single focus from hyperbolic geometry
- · [demo: ucjeps.berkeley.edu/map2.html]



[The Hyperbolic Browser: A Focus + Context Technique for Visualizing Large Hierarchies. Lamping and Rao, Proc SIGCHI '95. http://citeseer.nj.nec.com/lamping95focuscontext.h351]

Critique

strengths

· scales to over 10,000 nodes

weaknesses

- distortion poor for distance judgement tasks
- · still possible to get lost in large graphs

14: H3

data: web sites, species relationships 3D fisheye from hyperbolic geometry [demo: graphics.stanford.edu/~munzner/h3]



[Munzner. H3: Laying Out Large Directed Graphs in 3D Hyperbolic Space. Proc. InfoVis 1997. graphics.stanford.edu/papers/h3/]

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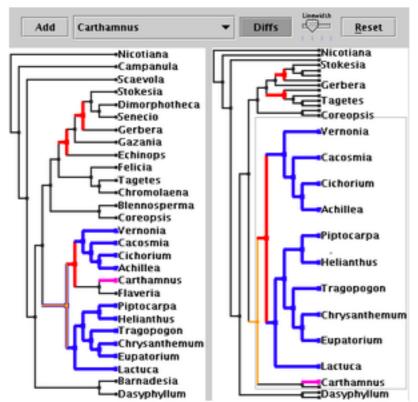
15: TreeJuxtaposer

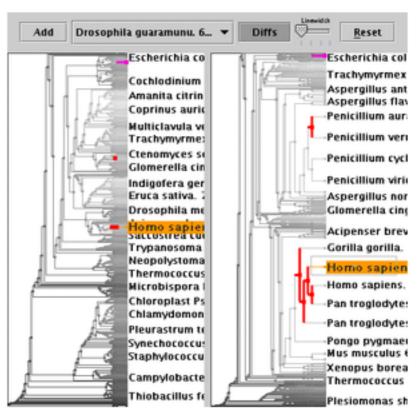
data: species evolutionary relationships

task: side by side comparison

accordion drawing

- guaranteed visibility of landmarks
- stretch and squish navigation
- [demo: olduvai.sf.net/tj]





[Munzner et al. TreeJuxtaposer: Scalable Tree Comparison using Focus+Context with Guaranteed Visibility. SIGGRAPH 2003. www.cs.ubc.ca/~tmm/papers/tj] 43

Guaranteed Visibility

drawing colored marks

- easy with small datasets
- hard with big datasets

reasons a mark could be invisible

- outside the window
- underneath other marks
- smaller than a pixel

benefits of GV

- minimizes amount of navigation required
- guides necessary navigation choices
- provides visible landmarks

Critique

strengths

- scalability to millions of nodes guaranteed frame rate
- guaranteed visibility
- supports multiple focus areas

weaknesses

- stretch and squish navigation inappropriate for tasks requiring distance estimation
- · computationally intensive

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Hard Problems

designing within huge space of possibilities

scalability

- · size of dataset
- number of pixels
- · kinds of data

dynamic data

characterizing Focus+Context

· how and when does it help

Grand Challenge

"visual Google for nontextual data"

not search for images

web search made available text data useable

· for general and surprising uses beyond original intent

infovis browsing

· could make available nontext data useful/visible

More Information

this talk

· www.cs.ubc.ca/~tmm/talks.html#ie05

my grad course

www.cs.ubc.ca/~tmm/courses/infovis

conferences

- InfoVis symposia: www.infovis.org/symposia.php
 IEEE Sympoxium on Information Visualization
- Graph Drawing conferences: www.gd2005.org