

University of British Columbia CPSC 314 Computer Graphics Jan-Apr 2013

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Visualization

http://www.ugrad.cs.ubc.ca/~cs314/Vjan2013

Nonspatial/Information Visualization

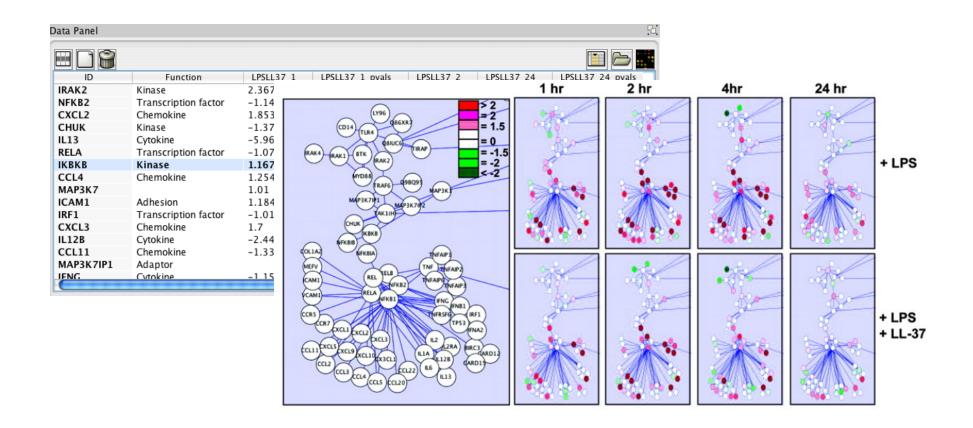
Reading

- FCG Chap 27
 - N/A 2nd edition, available online at

http://www.cs.ubc.ca/labs/imager/tr/2009/VisChapter

Why Do Visualization?

- pictures help us think
 - substitute perception for cognition
 - external memory: free up limited cognitive/memory resources for higher-level problems



Information Visualization

- interactive visual representation of abstract data
 - help human perform some task more effectively
- bridging many fields
 - computer graphics: interact in realtime
 - cognitive psychology: find appropriate representation
 - HCI: use task to guide design and evaluation
- external representation
 - reduces load on working memory
 - offload cognition
 - familiar example: multiplication/division
 - infovis example: topic graphs

External Representation: Topic Graphs

hard to find topics two hops away from target

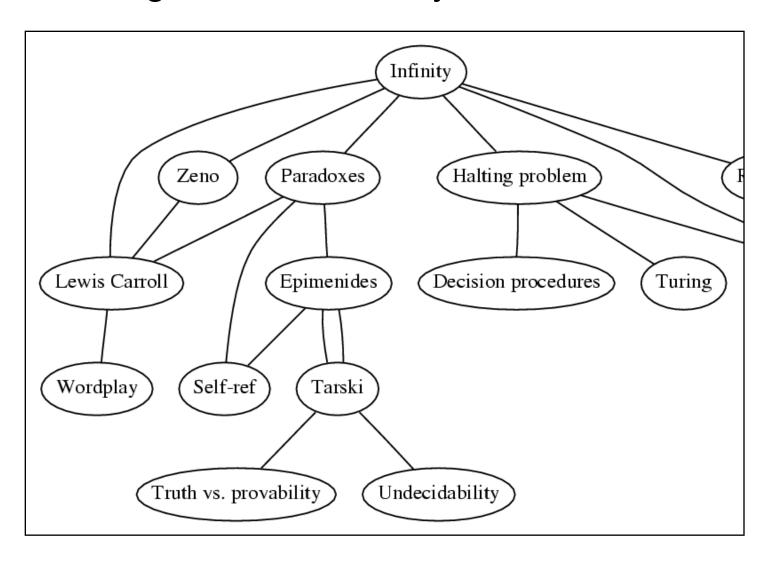
[Godel, Escher, Bach: The Eternal Golden Braid. Hofstadter 1979]

- Paradoxes Lewis Carroll
- Turing Halting problem
- Halting problem Infinity
- Paradoxes Infinity
- Infinity Lewis Carroll
- Infinity Unpredictably long searches
- Infinity Recursion
- Infinity Zeno
- Infinity Paradoxes
- Lewis Carroll Zeno
- Lewis Carroll Wordplay

- Halting problem Decision procedures
- BlooP and FlooP Al
- Halting problem Unpredictably long searches
- BlooP and FlooP Unpredictably long searches
- BlooP and FlooP Recursion
- Tarski Truth vs. provability
- Tarski Epimenides
- Tarski Undecidability
- Paradoxes Self-ref
- [...]

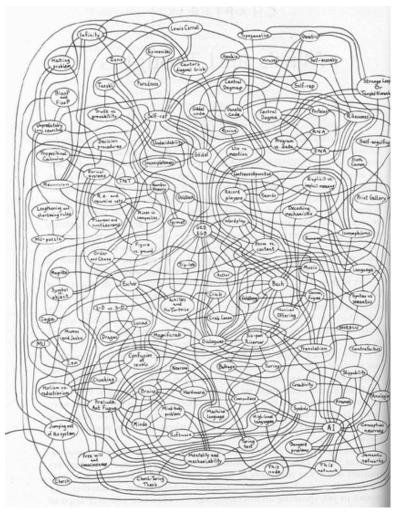
External Representation: Topic Graphs

offload cognition to visual system



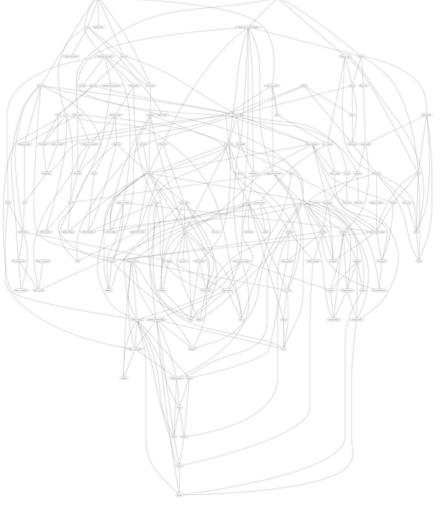
Automatic Node-Link Graph Layout

manual: hours, days



[Godel, Escher, Bach. Hofstadter 1979]

automatic: seconds



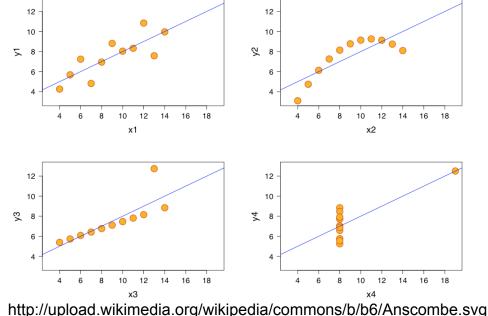
[dot, Gansner et al, 1973.]

When To Do Vis?

- need a human in the loop
 - augment, not replace, human cognition
 - for problems that cannot be (completely) automated
- simple summary not adequate
 - statistics may not adequately characterize complexity of dataset distribution

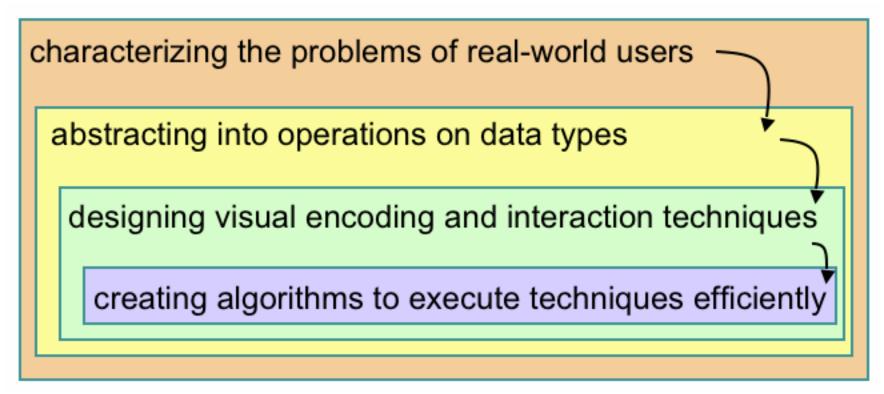
Anscombe's quartet: same

- mean
- variance
- correlation coefficient
- linear regression line



Visualization Design Layers

depends on both data and task



Visual Encoding

marks: geometric primitives points lines areas

position

size

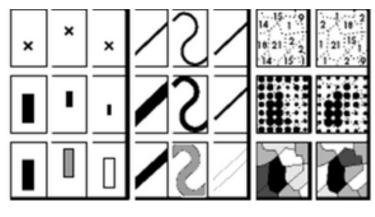
grey level

texture

color

orientation

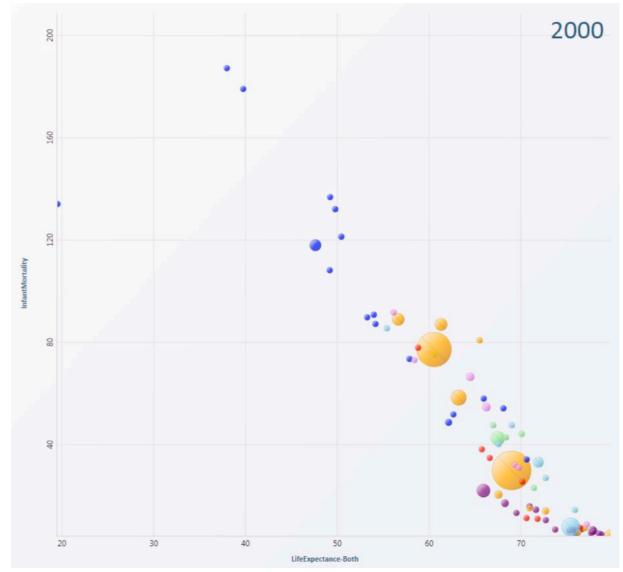
shape



- attributes
 - parameters control mark appearance
 - separable channels flowing from retina to brain

Visual Encoding Example: Scatterplot

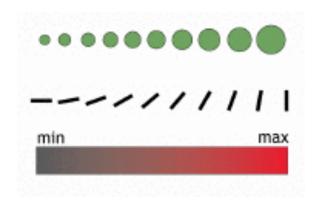
- x position
- y position
- hue
- size



Robertson et al. Effectiveness of Animation in Trend Visualization. IEEE TVCG (Proc. InfoVis08) 14:6 (2008), 1325–1332.

Data Types

- quantitative
 - lengths: 10 inches, 17 inches, 23 inches



- ordered
 - sizes: small, medium, large
 - days: Mon, Tue, Wed, ...

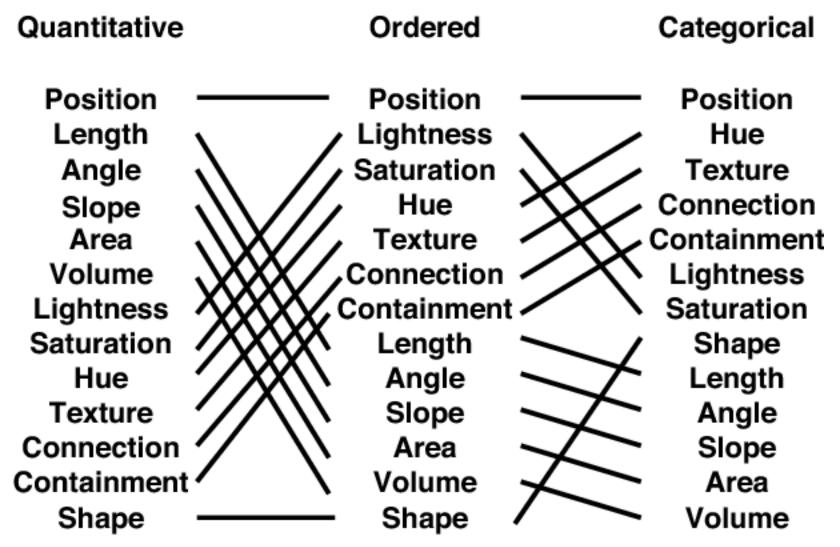


- categorical
 - fruit: apples, oranges, bananas



[Stolte and Hanrahan. Polaris: A System for Query, Analysis and Visualization of Multi-dimensional Relational Databases. Proc InfoVis 2000. graphics.stanford.edu/projects/polaris/]

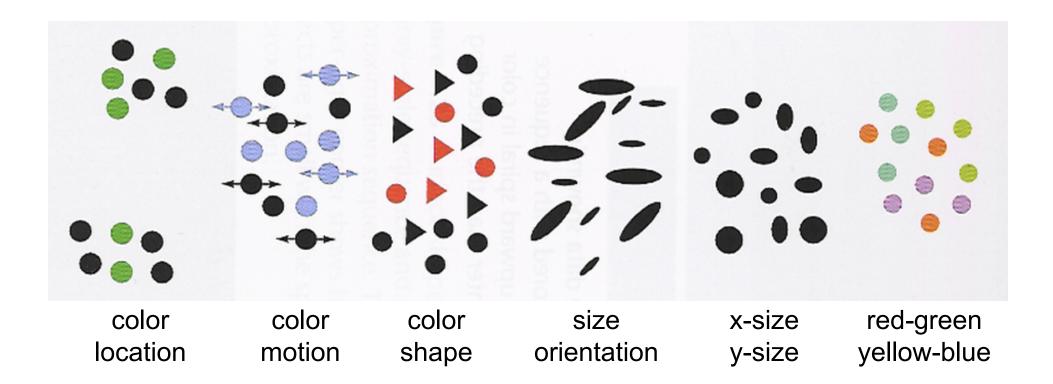
Channel Ranking Varies By Data Type



[Mackinlay, Automating the Design of Graphical Presentations of Relational Information, ACM TOG 5:2, 1986]

Integral vs. Separable Dimensions

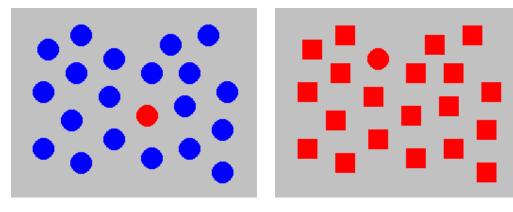
not all dimensions separable



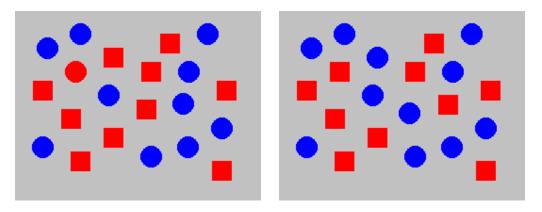
[Colin Ware, Information Visualization: Perception for Design. Morgan Kaufmann 1999.]

Preattentive Visual Channels

color alone, shape alone: preattentive

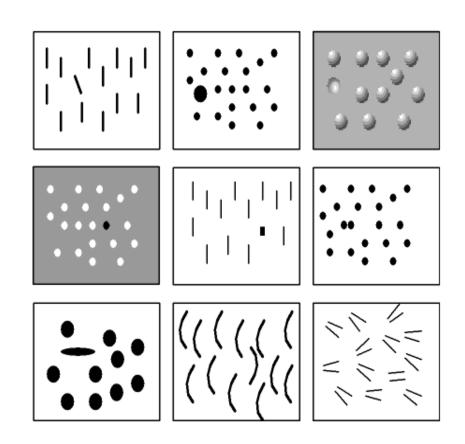


- combined color and shape: requires attention
 - search speed linear with distractor count



Preattentive Visual Channels

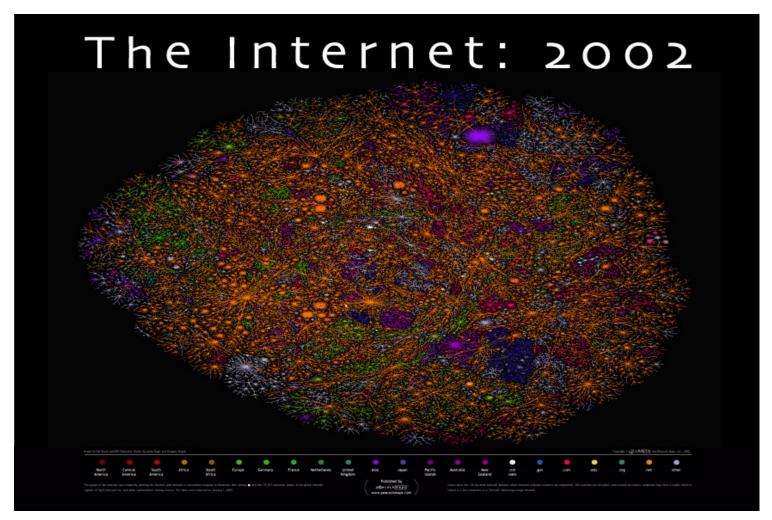
- preattentive channels include
 - hue
 - shape
 - texture
 - length
 - width
 - size
 - orientation
 - curvature
 - intersection
 - intensity
 - flicker
 - direction of motion
 - stereoscopic depth
 - lighting direction
 - many more...



[Healey, [www.csc.ncsu.edu/faculty/healey/PP/PP.html]

Coloring Categorical Data

• 22 colors, but only ~8 distinguishable



Coloring Categorical Data

- discrete small patches separated in space
- limited distinguishability: around 8-14
 - channel dynamic range low
 - best to choose bins explicitly
- maximal saturation for small areas

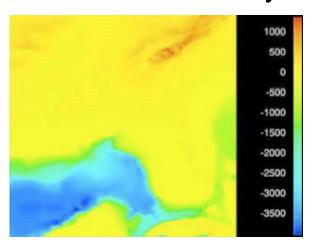


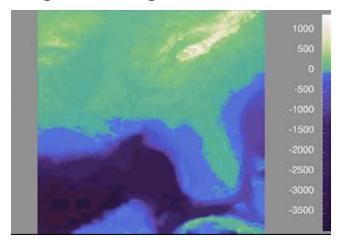


[Colin Ware, Information Visualization: Perception for Design. Morgan Kaufmann 1999.]

Quantitative Colormaps

- dangers of rainbows
 - perceptually nonlinear
 - arbitrary not innate ordering
- other approaches
 - explicitly segmented colormaps
 - monotonically increasing/(decreasing) luminance, plus hue to semantically distinguish regions

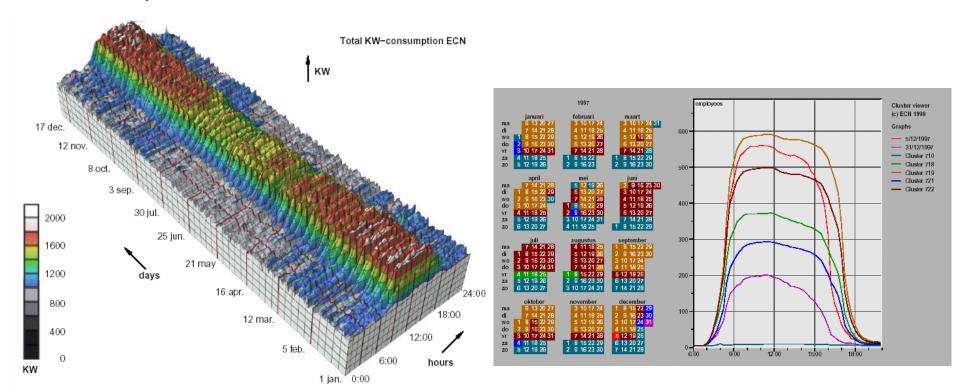




Rogowitz and Treinish. Data Visualization: The End of the Rainbow. IEEE Spectrum 35(12):52-59, Dec 1998.

3D vs 2D Representations

- curve comparison difficult: perspective distortion, occlusion
 - dataset is abstract, not inherently spatial
 - after data transformation to clusters, linked 2D views of representative curves show more



[van Wijk and van Selow, Cluster and Calendar based Visualization of Time Series Data, InfoVis99

Space vs Time: Showing Change

- animation: show time using temporal change
 - good: show process
 - good: flip between two things
 - bad: flip between between many things
 - interference between intermediate frames



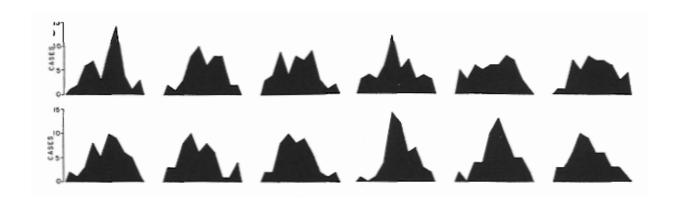






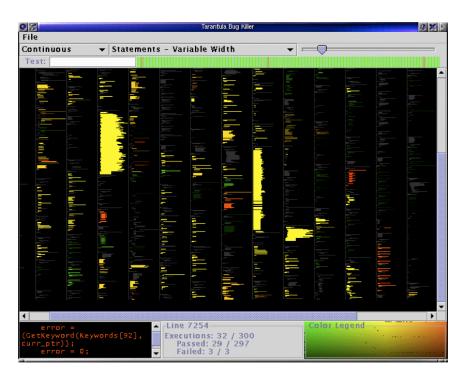
Space vs Time: Showing Change

- small multiples: show time using space
 - overview: show each time step in array
 - compare: side by side easier than temporal
 - external cognition vs internal memory
 - general technique, not just for temporal changes



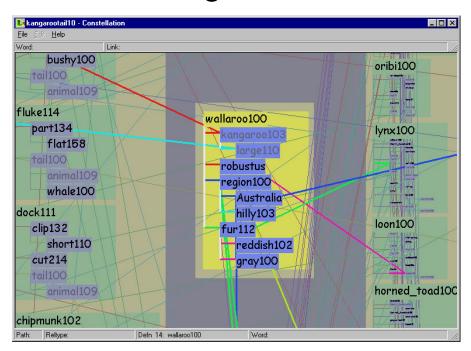
Composite Views

- pixel-oriented views
 - overviews with high information density



[Jones, Harrold, and Stasko. Visualization of Test Information to Assist Fault Localization. Proc. ICSE 2002, p 467-477.]

- superimposing/layering
 - shared coordinate frame
 - redundant visual encoding

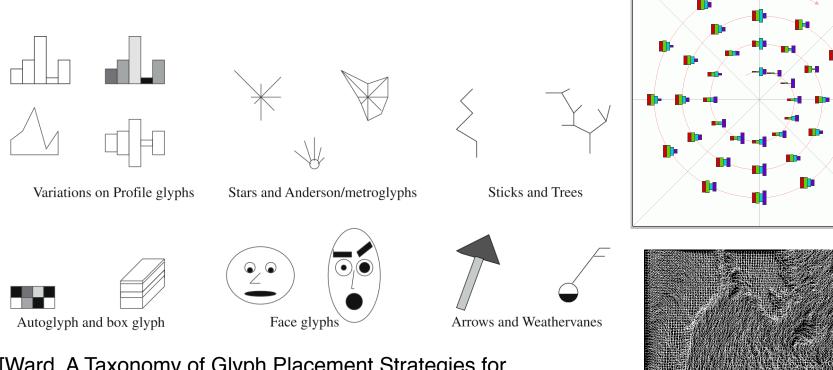


[Munzner. Interactive Visualization of Large 24 Graphs and Networks. Stanford CS, 2000]

Composite Views: Glyphs

internal structure where subregions have different

visual channel encodings

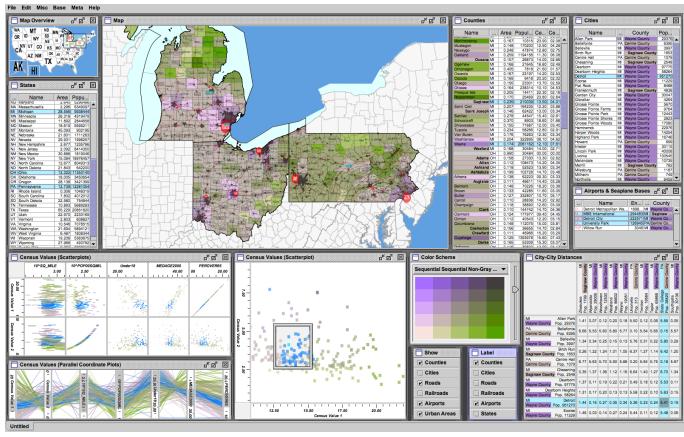


[Ward. A Taxonomy of Glyph Placement Strategies for Multidimensional Data Visualization. Information Visualization Journal 1:3-4 (2002), 194--210.]

[Smith, Grinstein, and Bergeron. Interactive data exploration with a supercomputer. Proc. IEEE Visualization, p 248-254, 1991.]

Adjacent: Multiple Views

- different visual encodings show different aspects of the data
- linked highlighting to show where contiguous in one view distributed within another

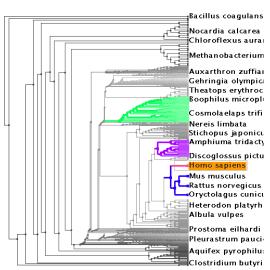


Adjacent Views

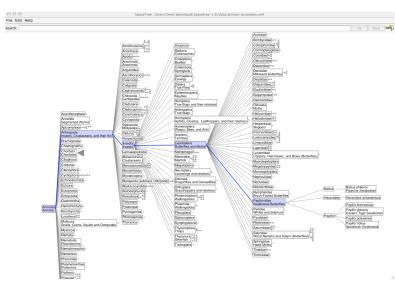
- overview and detail
 - same visual encoding, different resolutions
- small multiples
 - same visual encoding, different data

Data Reduction

- overviews as aggregation
- focus+context
 - show details embedded within context
 - distortion: TreeJuxtaposer video
 - filtering: SpaceTree demo



[Munzner et al. TreeJuxtaposer: Scalable Tree Comparison using Focus+Context with Guaranteed Visibility. Proc SIGGRAPH 2003, p 453-462]



[Plaisant, Grosjean, and Bederson. SpaceTree: Supporting Exploration in Large Node Link Tree, Design Evolution and Empirical Evaluation. Proc. InfoVis 2002

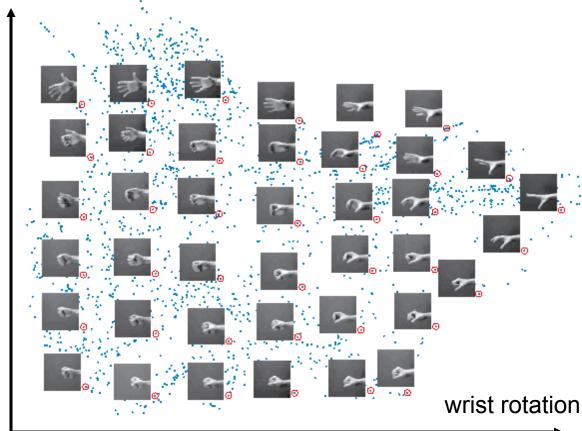
Dimensionality Reduction

- mapping from high-dimensional space into space of fewer dimensions
 - generate new synthetic dimensions
- why is lower-dimensional approximation useful?
 - assume true/intrinsic dimensionality of dataset is (much) lower than measured dimensionality!
 - only indirect measurement possible?
 - fisheries: want spawn rates.
 have water color, air temp, catch rates...
 - sparse data in verbose space?
 - documents: word occurrence vectors.
 10K+ dimensions, want dozens of topic clusters

DR Example: Image Database

- 4096 D (pixels) to 2D (hand gesture)
 - no semantics of new synthetic dimensions from alg.
 - assigned by humans after inspecting results

finger extension



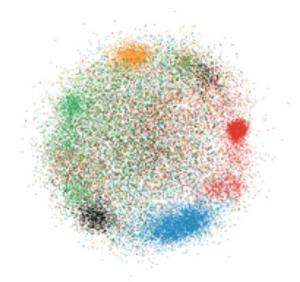
DR Technique: MDS

- multidimensional scaling
 - minimize differences between interpoint distances in high and low dimensions
- minimize objective function: stress

$$stress(D,\Delta) = \sqrt{rac{\sum_{ij} (d_{ij} - \delta_{ij})^2}{\sum_{ij} \delta_{ij}^2}}$$

D: matrix of lowD distances d_{ij}

Δ: matrix of hiD distances

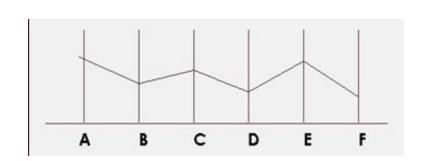


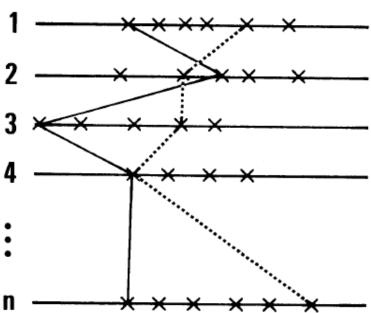
Glimmer: MDS on the GPU

[Ingram, Munzner, Olano. Glimmer: Multiscale MDS on the GPU. IEEE TVCG 15(2):249-261, 2009.

Parallel Coordinates

- only two orthogonal axes in the plane
- instead, use parallel axes!

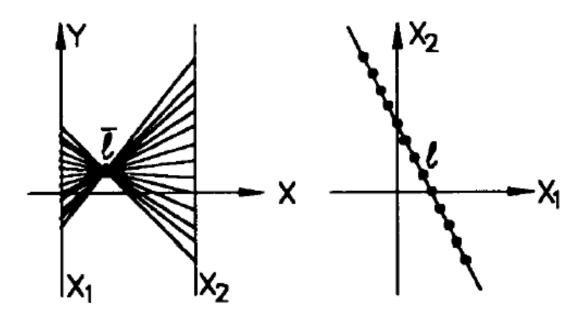




[Hyperdimensional Data Analysis Using Parallel Coordinates. Edward J. Wegman. Journal of the American Statistical Association, Vol. 85, No. 411. (Sep., 1990), pp. 664-675.]

Parallel Coordinates

- point in Cartesian coords is line in par coords
- point in par coords is line in Cartesian n-space



[Inselberg and Dimdale. Parallel Coordinates: A Tool for Visualizing Multi-Dimensional Geometry. IEEE Visualization '90.]

Par Coords: Correllation

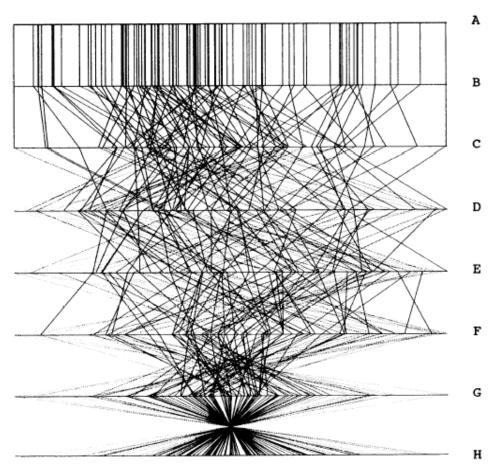
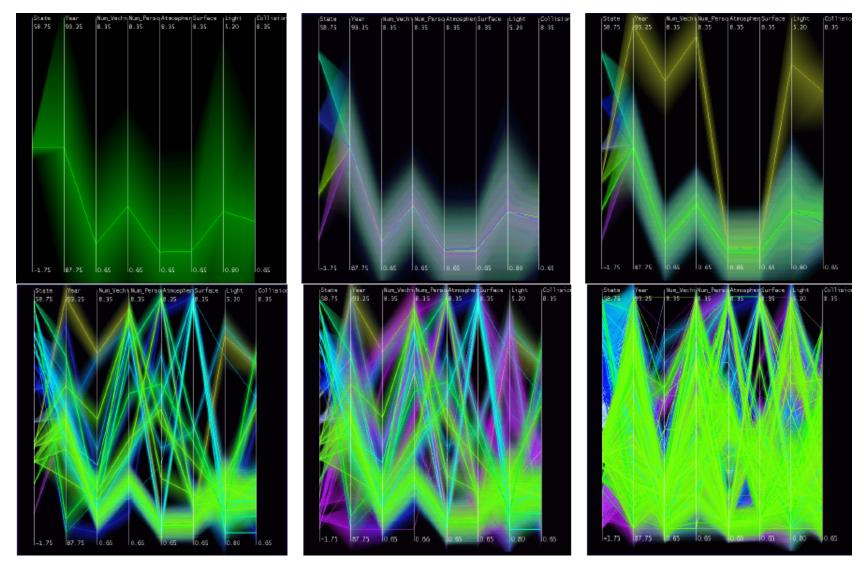


Figure 3. Parallel Coordinate Plot of Six-Dimensional Data Illustrating Correlations of $\rho=1,.8,.2,0,-.2,-.8,$ and -1.

[Hyperdimensional Data Analysis Using Parallel Coordinates. Wegman. Journal of the American Statistical Association, Vol. 85, No. 411. (Sep., 1990), pp. 664-675.]

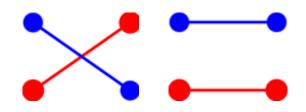
Hierarchical Parallel Coords: LOD

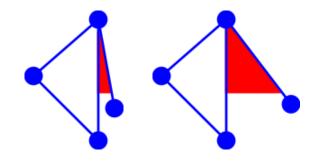


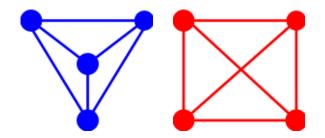
[Hierarchical Parallel Coordinates for Visualizing Large Multivariate Data Sets. Fua, Ward, and Rundensteiner. IEEE Visualization '99.]

Node-Link Graph Layout

- minimize
 - crossings, area, bends/curves
- maximize
 - angular resolution, symmetry
- most criteria individually NP-hard
 - cannot just compute optimal answer
 - heuristics: try to find something reasonable
- criteria mutually incompatible

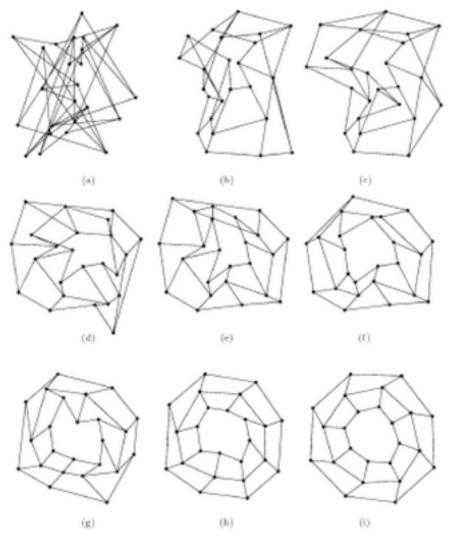






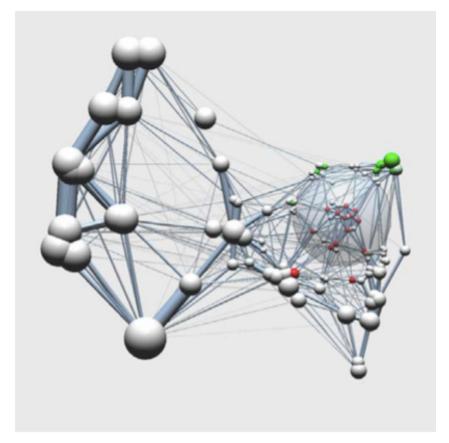
Force-Directed Placement

- nodes: repel like magnets
- edges: attract like springs
 - start from random positions, run to convergence
- very well studied area!
 - many people reinvent the wheel



Interactive Graph Exploration

geometric and semantic fisheye



van Ham and van Wijk. Interactive Visualization of Small World Graphs.

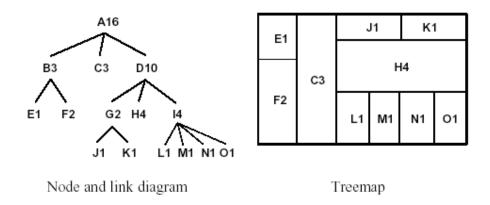
Proc. InfoVis 2005

Treemaps

containment rather than connection

emphasize node attributes, not topological

structure

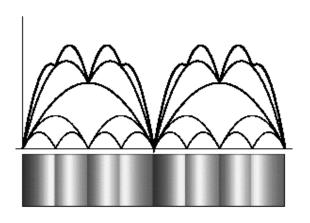


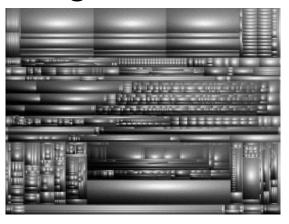
[van Wijk and van de Wetering. Cushion Treemaps. Proc InfoVis 1999]

[Fekete and Plaisant. Interactive Information Visualization of a Million Items. Proc InfoVis 2002.

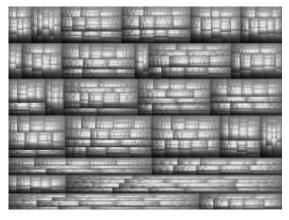
Cushion Treemaps

- show structure with shading
 - single parameter controls global vs local view











Now What?

Beyond 314: Other Graphics Courses

- 424: Geometric Modelling
 - was offered this year
- 426: Computer Animation
 - will be offered next year
- 514: Image-Based Rendering Heidrich
- 526: Algorithmic Animation van de Panne
- 533A: Digital Geometry Sheffer
- 533B: Animation Physics Bridson
- 547: Information Visualization Munzner

Beyond UBC CS

- SIGGRAPH conference back in Vancouver August 2014!
 - 15K-20K people: incredible combination of research, entertainment, art
 - Electronic Theater, Exhibit, ETech, ...
 - pricey: but student rate, student volunteer program
- local SIGGRAPH chapter
 - talk series, SPARK FX festival, ...
 - http://siggraph.ca