

Visualization Principles

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Twitter

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<http://www.cs.ubc.ca/~tmm/talks.html#twitter> | 2

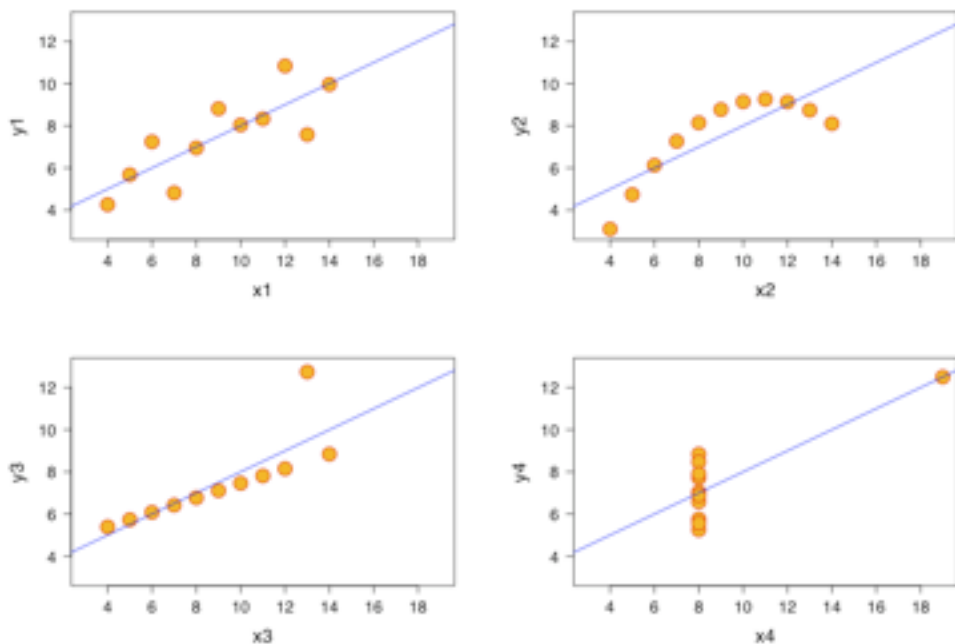
Defining visualization

computer-based visualization systems provide visual representations of datasets intended to help people carry out some task more effectively

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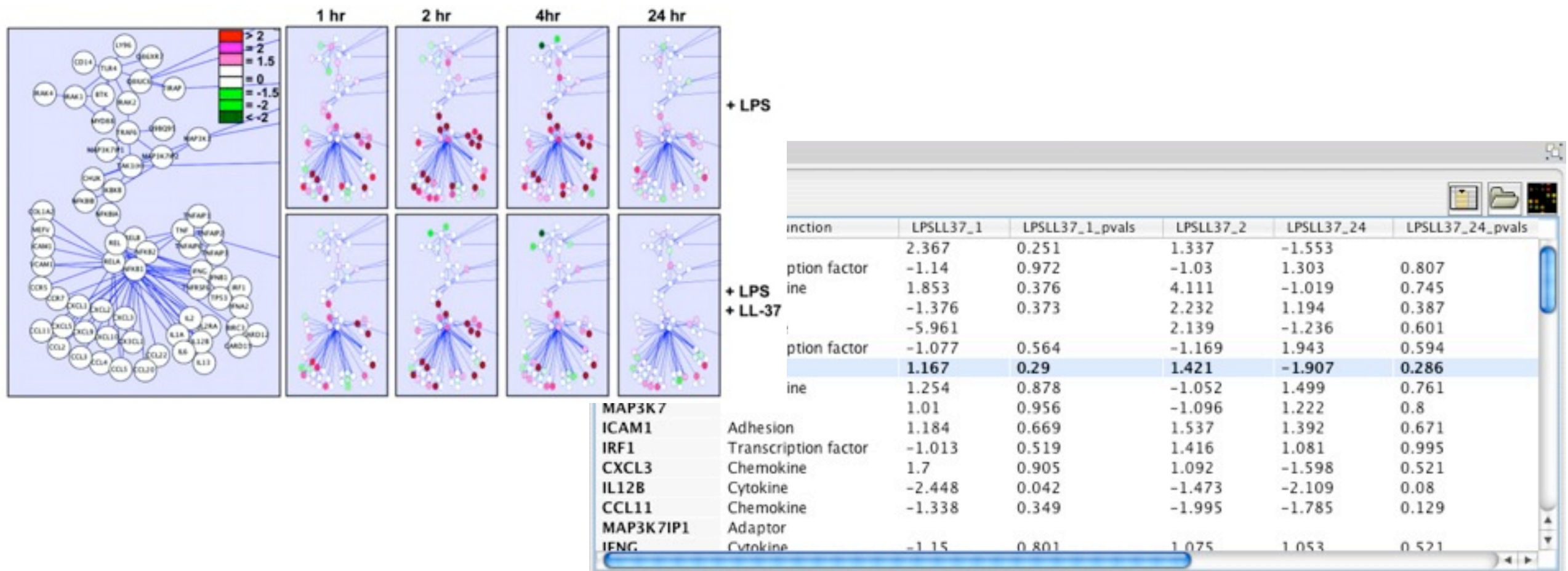
Identical statistics

x mean	9.0
x variance	10.0
y mean	7.50
y variance	3.75
x/y correlation	0.816

Defining visualization

computer-based visualization systems provide visual representations of datasets intended to help people carry out some task more effectively

- human in the loop needs the details
- external representation: perception vs cognition



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- human in the loop needs the details
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- intended task

Defining visualization

computer-based visualization systems provide visual representations of datasets intended to help people carry out some task more effectively

- human in the loop needs the details
- external representation: perception vs cognition
- intended task
- measurable definitions of effectiveness

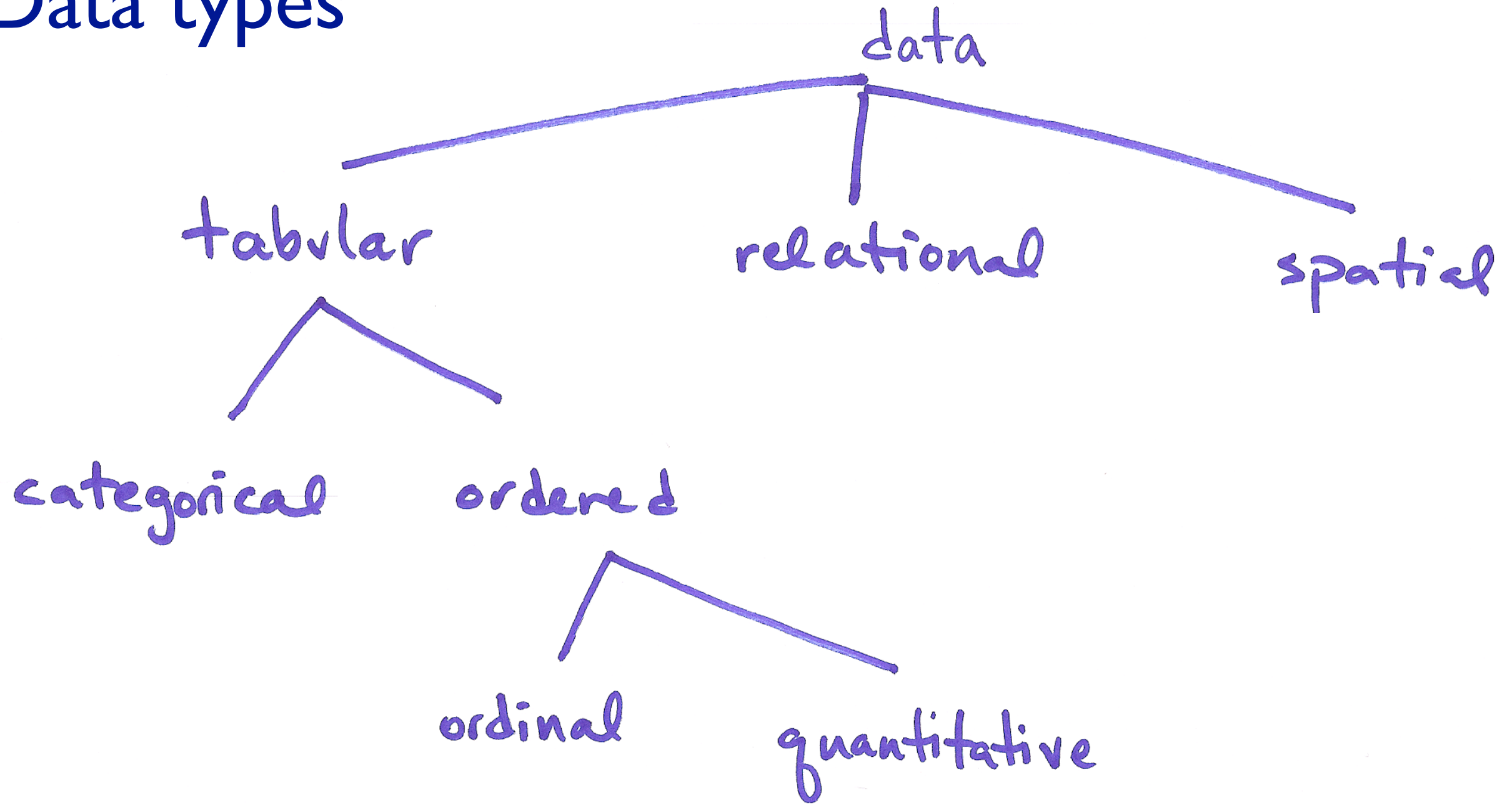
Visualization design space

- huge space of design alternatives
 - tradeoffs abound
- many possibilities now known to be ineffective
 - avoid random walk through parameter space
 - avoid some of our past mistakes
 - extensive experimentation has already been done
- guidelines continue to evolve
 - we reflect on lessons learned in design studies
 - iterative refinement usually wise

Principles

- know your visual channel types and ranks
- categorical color constraints
- power of the plane
- danger of depth
- resolution beats immersion
- eyes beat memory
- validate against the right threat

Data types



Visual encoding

- analyze
showing abstract data dimensions

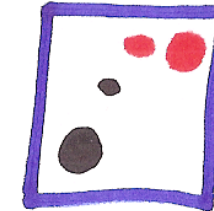
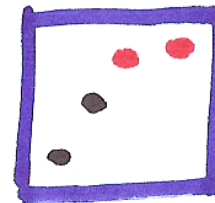
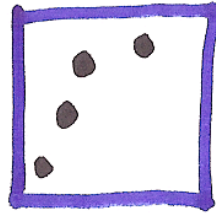
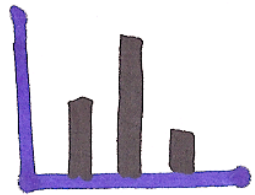


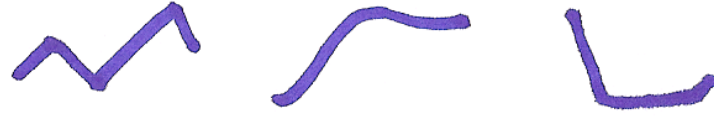
Image theory

- marks : geometric primitives

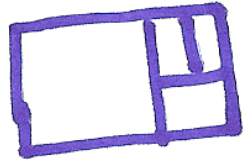
- points



- lines



- areas



- visual channels: control appearance of marks

- position

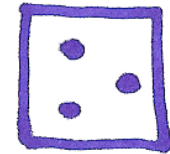
horizontal



vertical



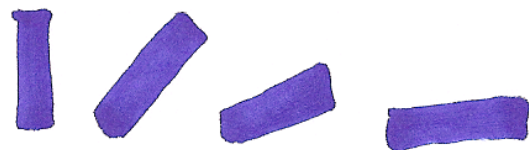
both



- color



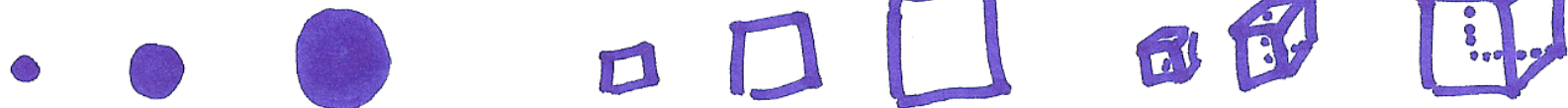
- tilt



- shape

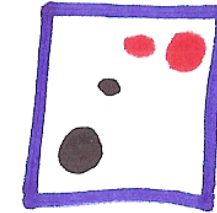
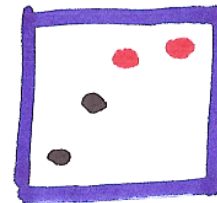
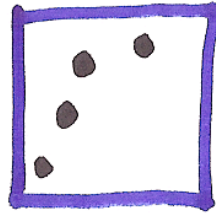
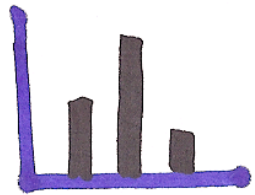


- size



Visual encoding

- analyze as combination of marks and channels showing abstract data dimensions



Visual channel types and rankings

Categorical
What/where

Ordered: Ordinal/Quantitative
How much

Relational, Same Category
Grouping

Power of the plane: only position works for all!

Categorical
What/where

planar position 

color hue 

shape 

stipple pattern 

Relationship, Same Category
Grouping

containment (2D) 

connection (1D) 

Similarity (other channels) 


Proximity (position) 

Ordered: Ordinal/Quantitative
How much

position on common scale 

position on unaligned scale 

length (1D size) 

tilt, angle 

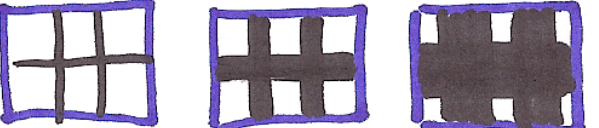
area (2D size) 

curvature 

volume (3D size) 

lightness black/white 

color saturation 

stipple density 

Ranking differs for all other channels

Categorical
What/where

planar position 

color hue 

shape 


stipple pattern 

Ordered: Ordinal/Quantitative
How much

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tilt, angle 

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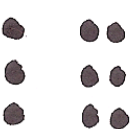
stipple density 

Relation, Same Category
Grouping

Containment (2D) 

Connection (1D) 

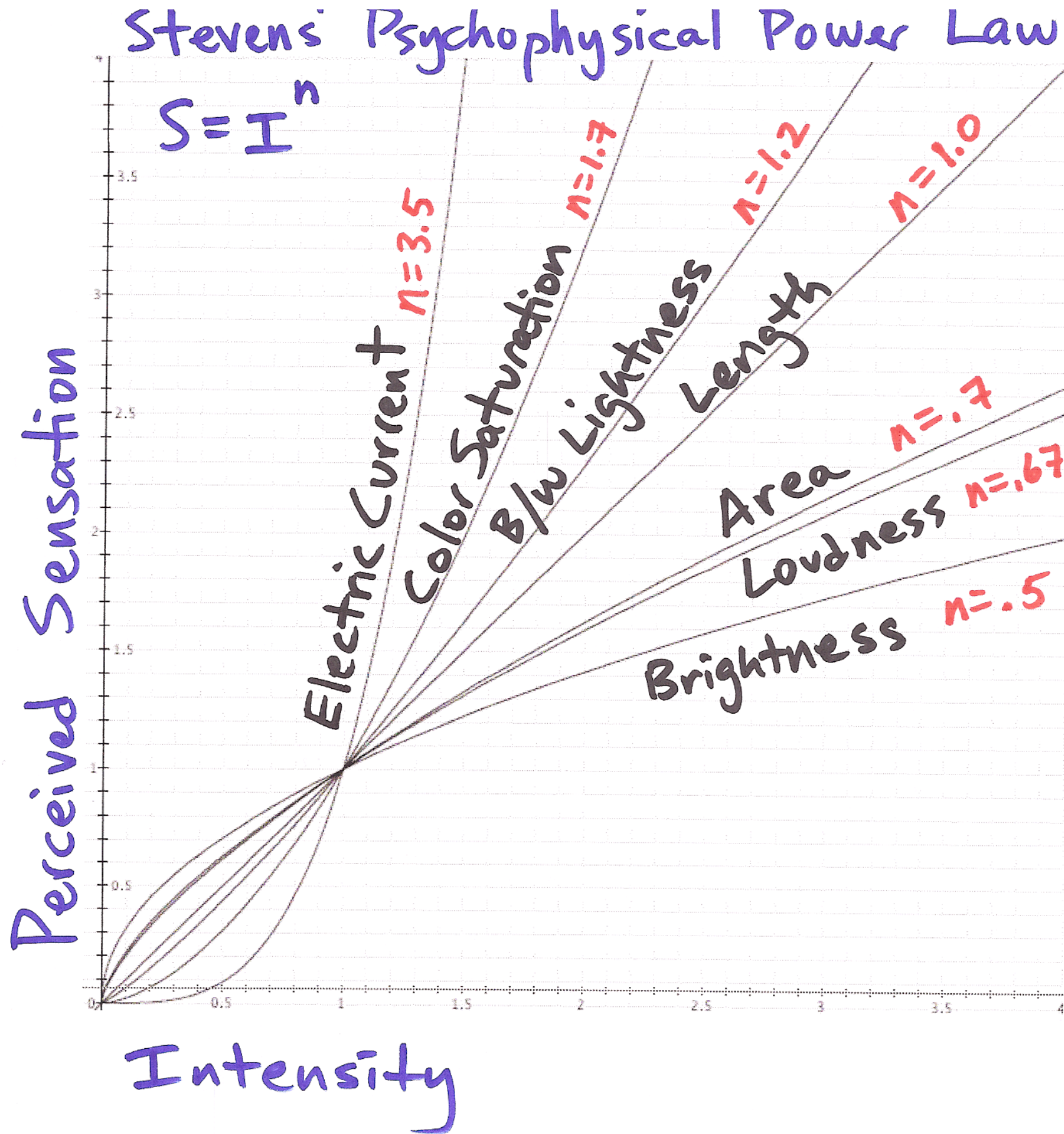
Similarity (other channels) 

Proximity (position) 

Channel rankings

- effectiveness principle: encode most important attributes with highest ranked channels [Mackinlay 86]
- where do rankings come from?
 - accuracy, discriminability, separability, popout

Accuracy

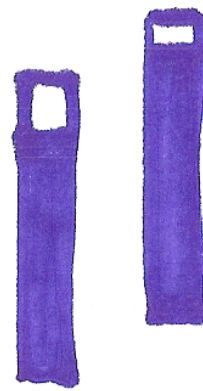


Accuracy

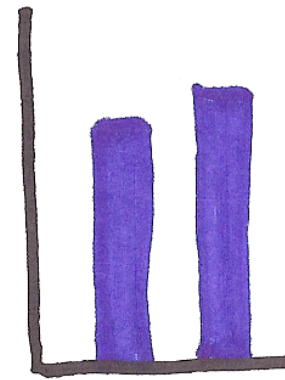
- position along common scale



no scale



framed

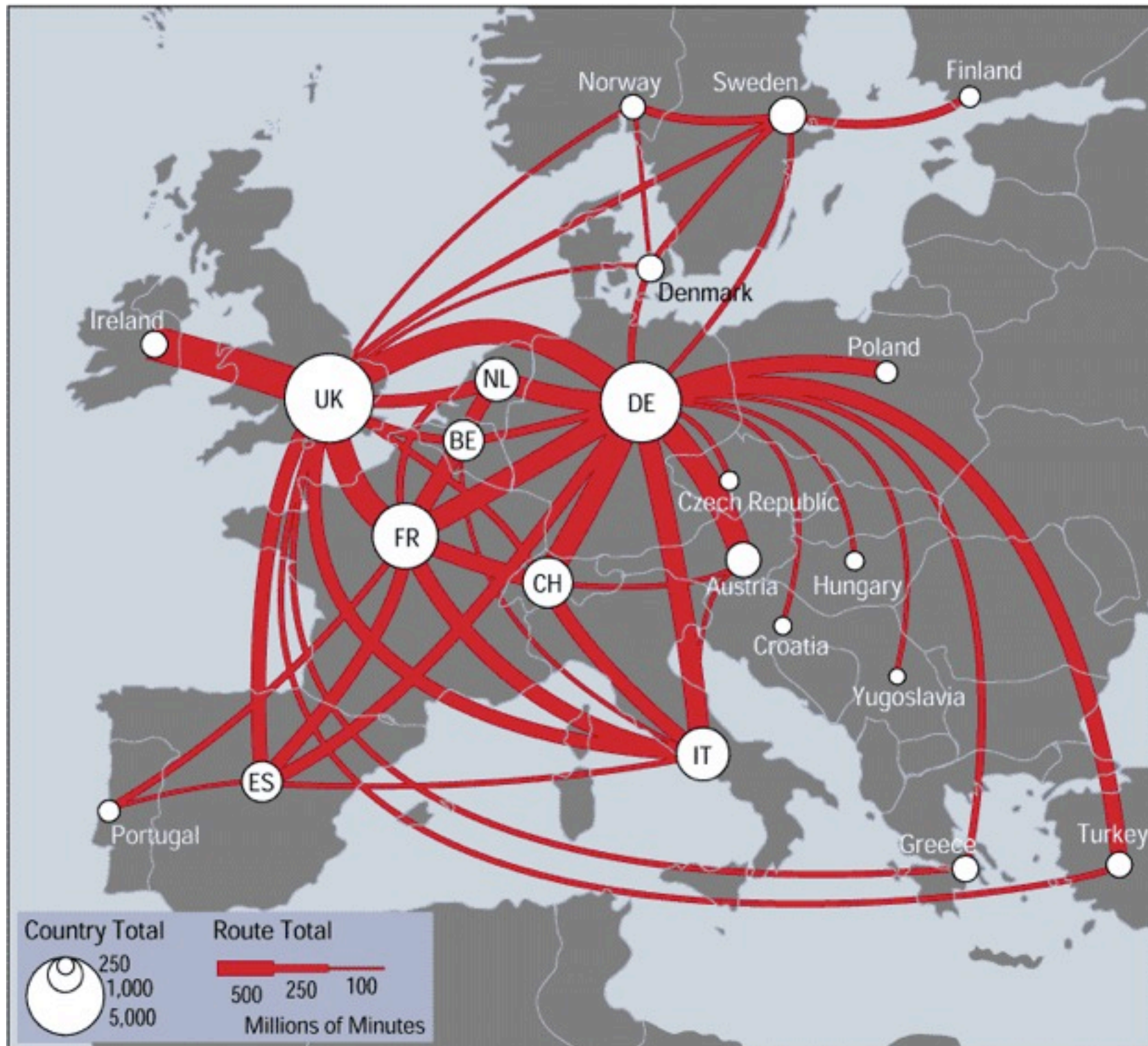


aligned

- frame increases accuracy [Cleveland 84]
- Weber's Law: relative judgements
 - filled rectangles differ by 1:9
 - white rectangles differ by 1:2

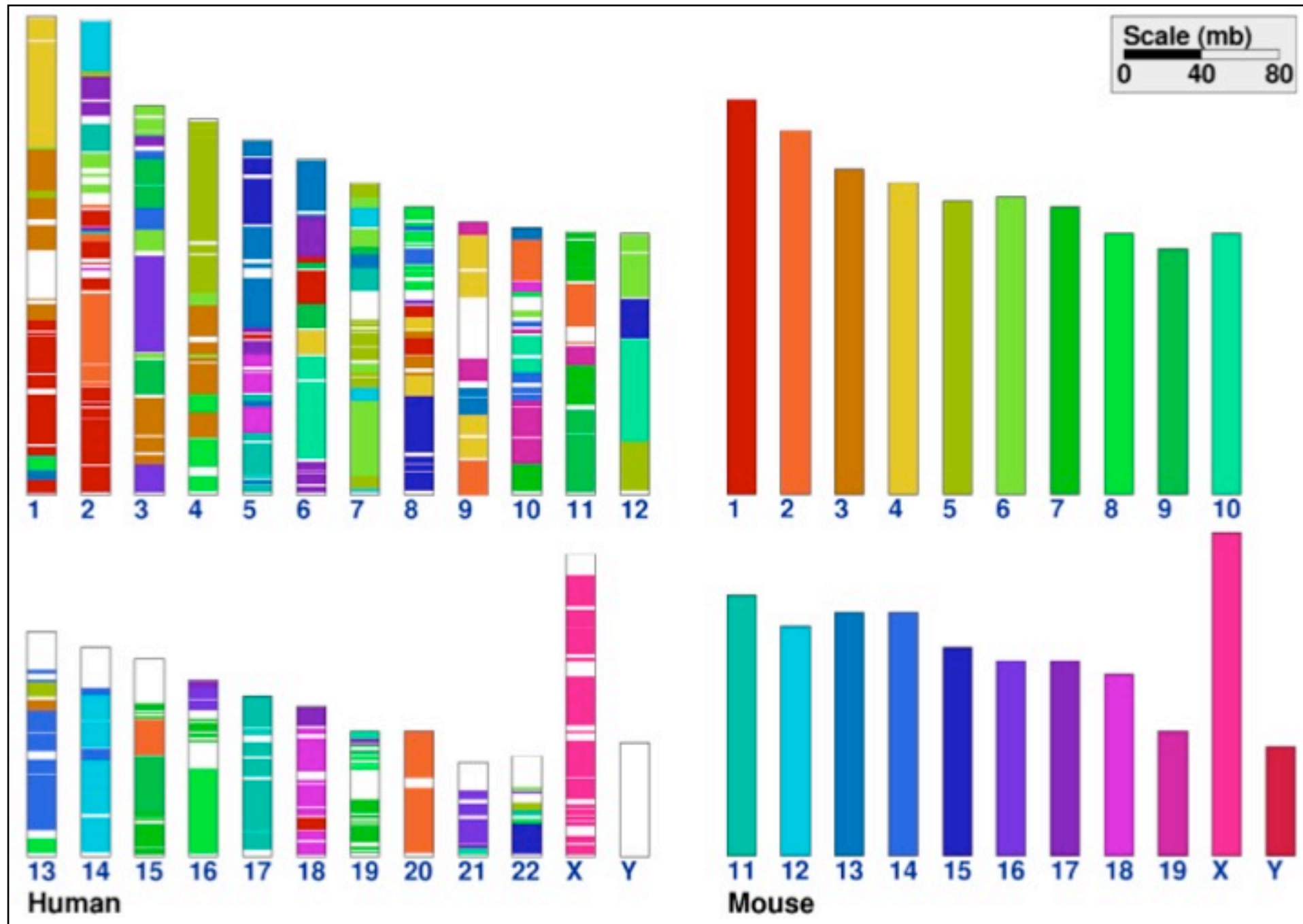
Discriminability: How many usable steps?

- linewidth: only a few



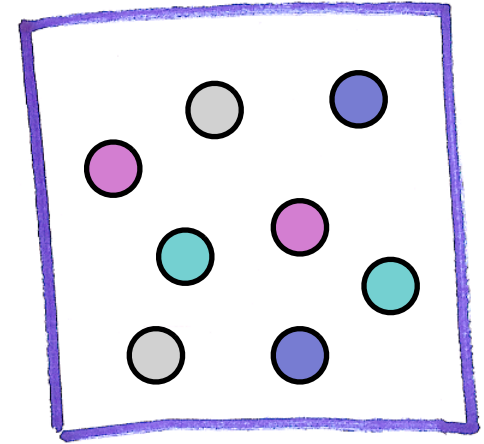
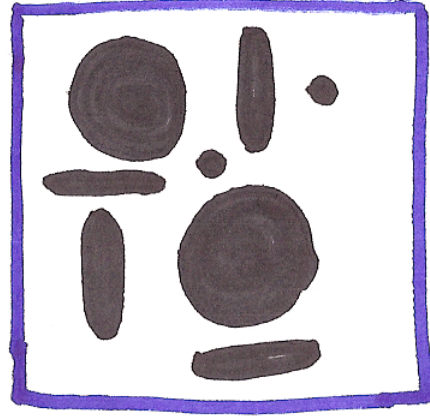
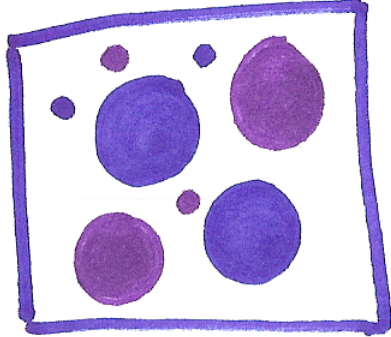
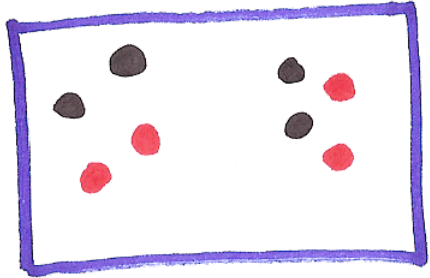
Discriminability: Categorical color constraints

- noncontiguous small regions of color: only 6-12 bins



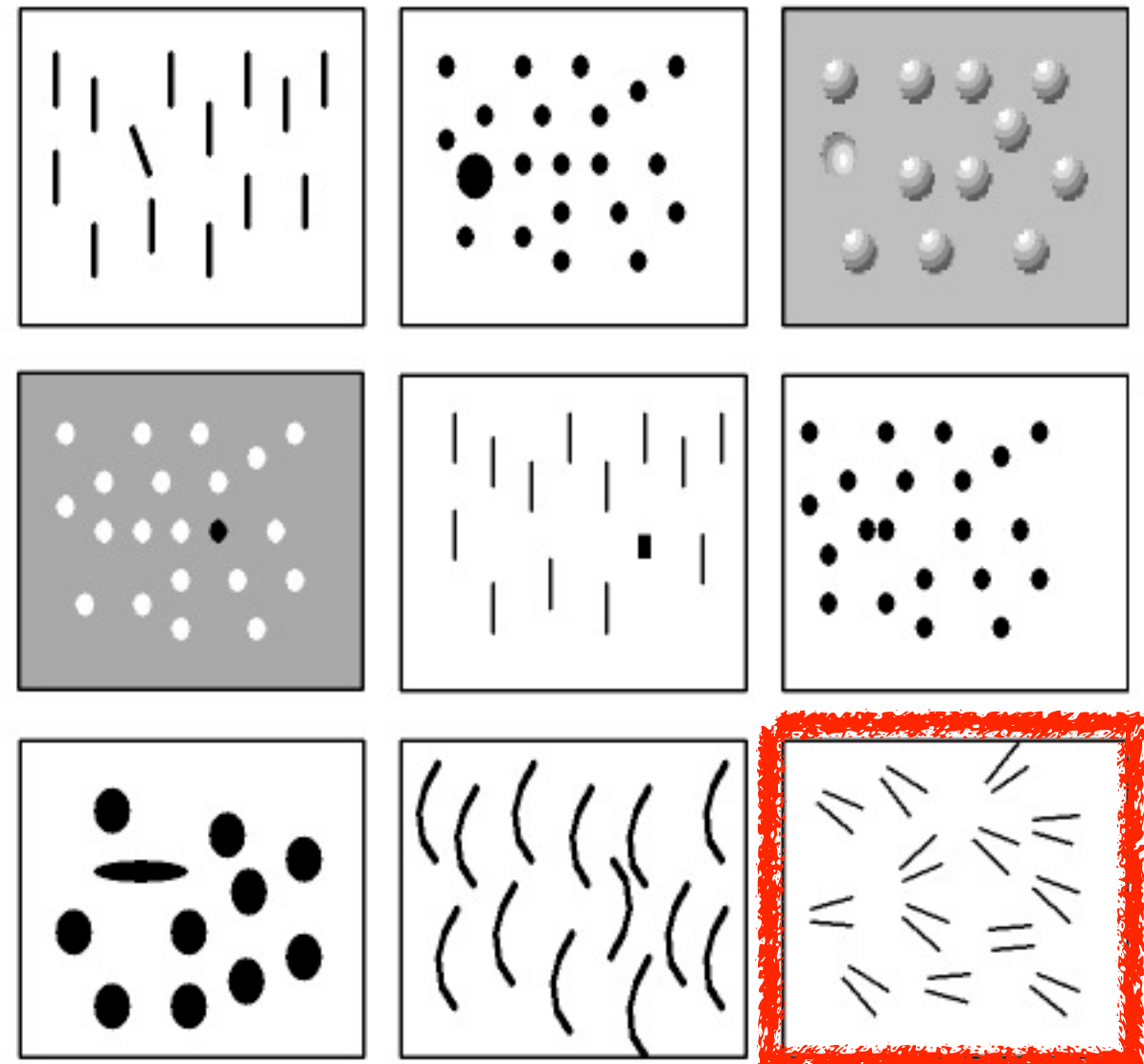
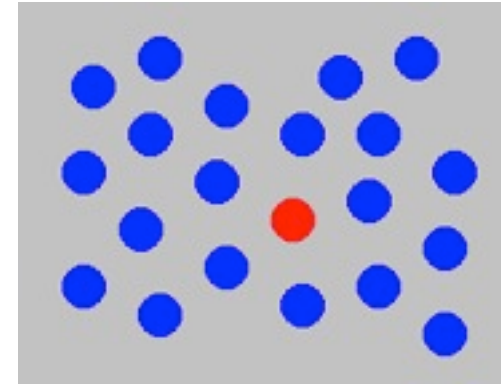
Cinteny: flexible analysis and visualization of synteny and genome rearrangements in multiple organisms. Sinha and Meller. Bioinformatics 2007

Separability vs. integrality



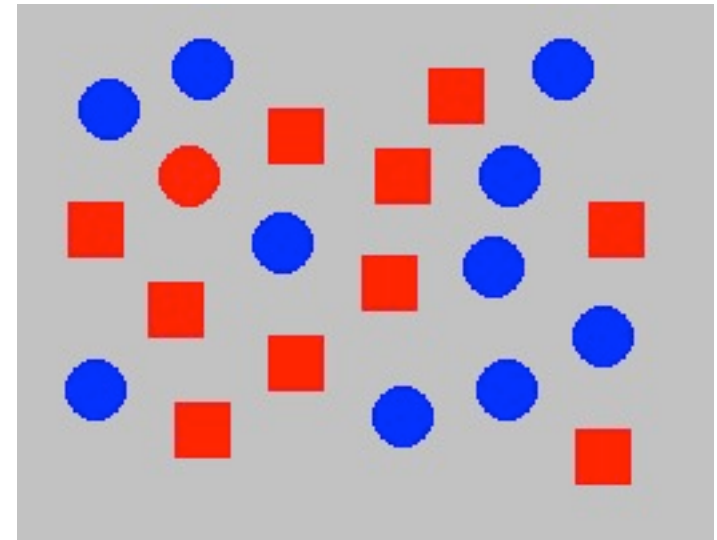
Popout: Most channels

- parallel processing on most channels
 - sufficiently different item noticed immediately, independent of distractor count
- some channels have no popout: serial search required

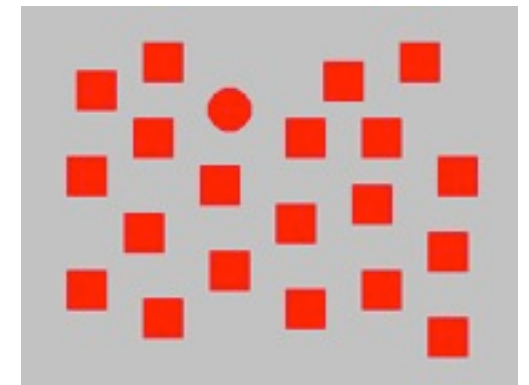


Popout limits

- only one channel at a time
 - combination searches are serial
 - most channel pairs
 - all channel triplets, etc



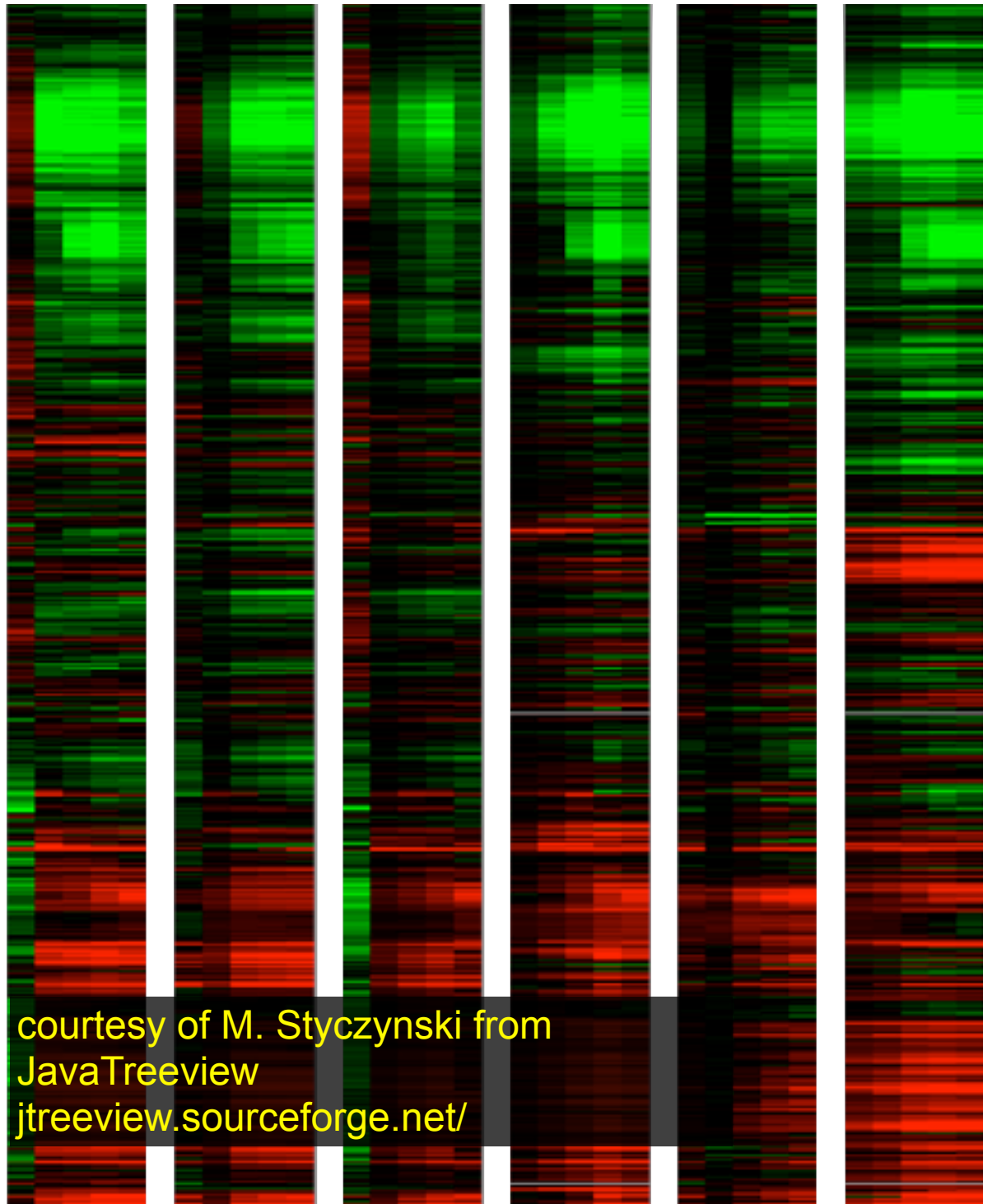
- within channel, speed depends on which channel and how different item is from surroundings
 - ‘sufficiently different’: context dependent



Encoding example: Heatmaps vs. curvemaps

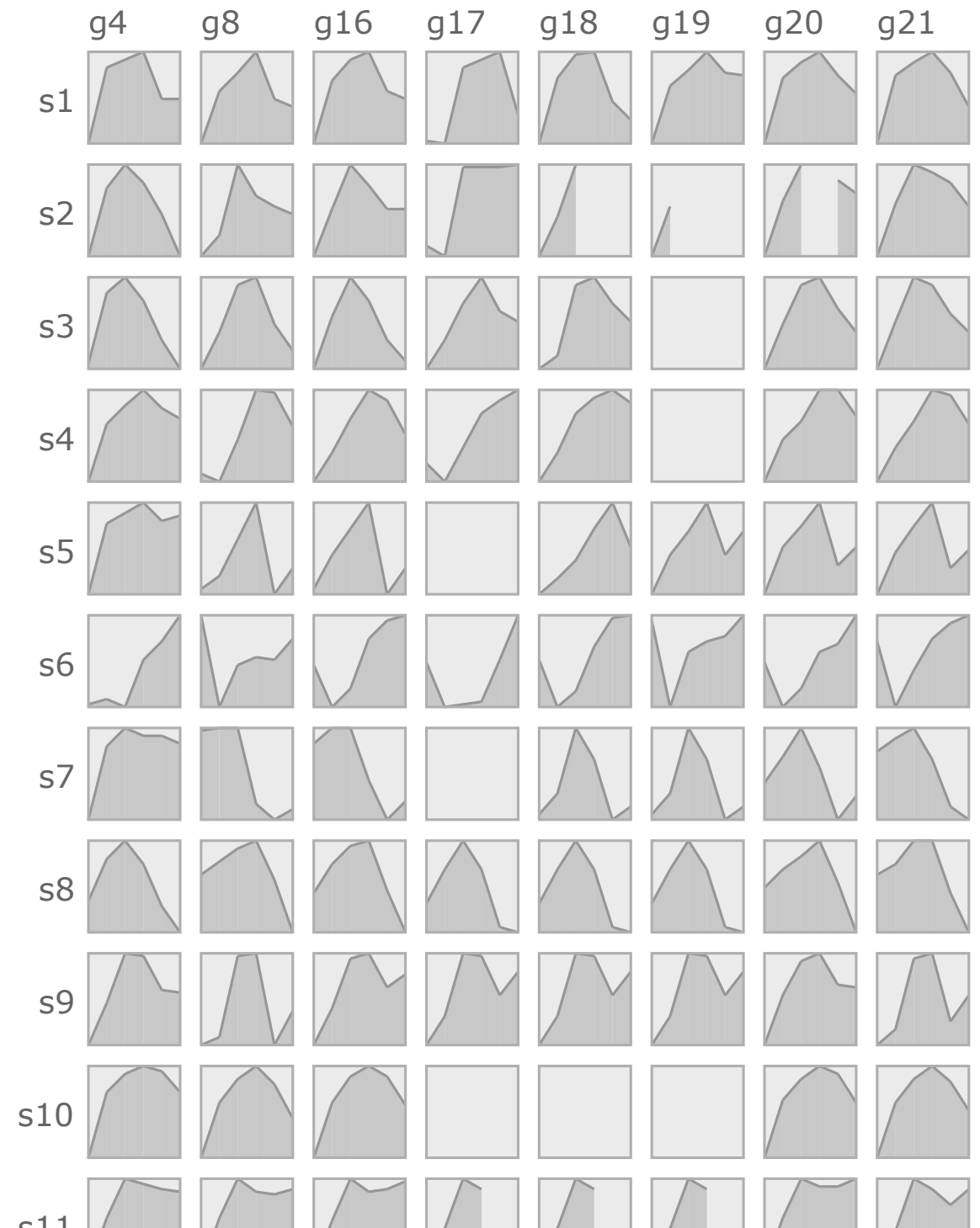
- color traditional, but spatial position outranks it

heatmap



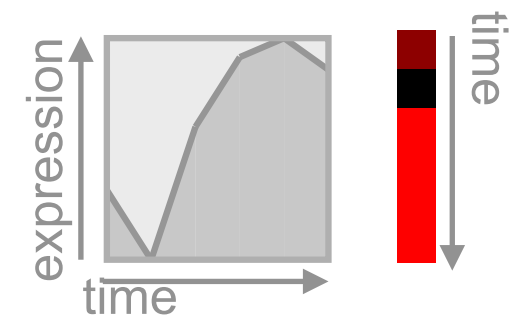
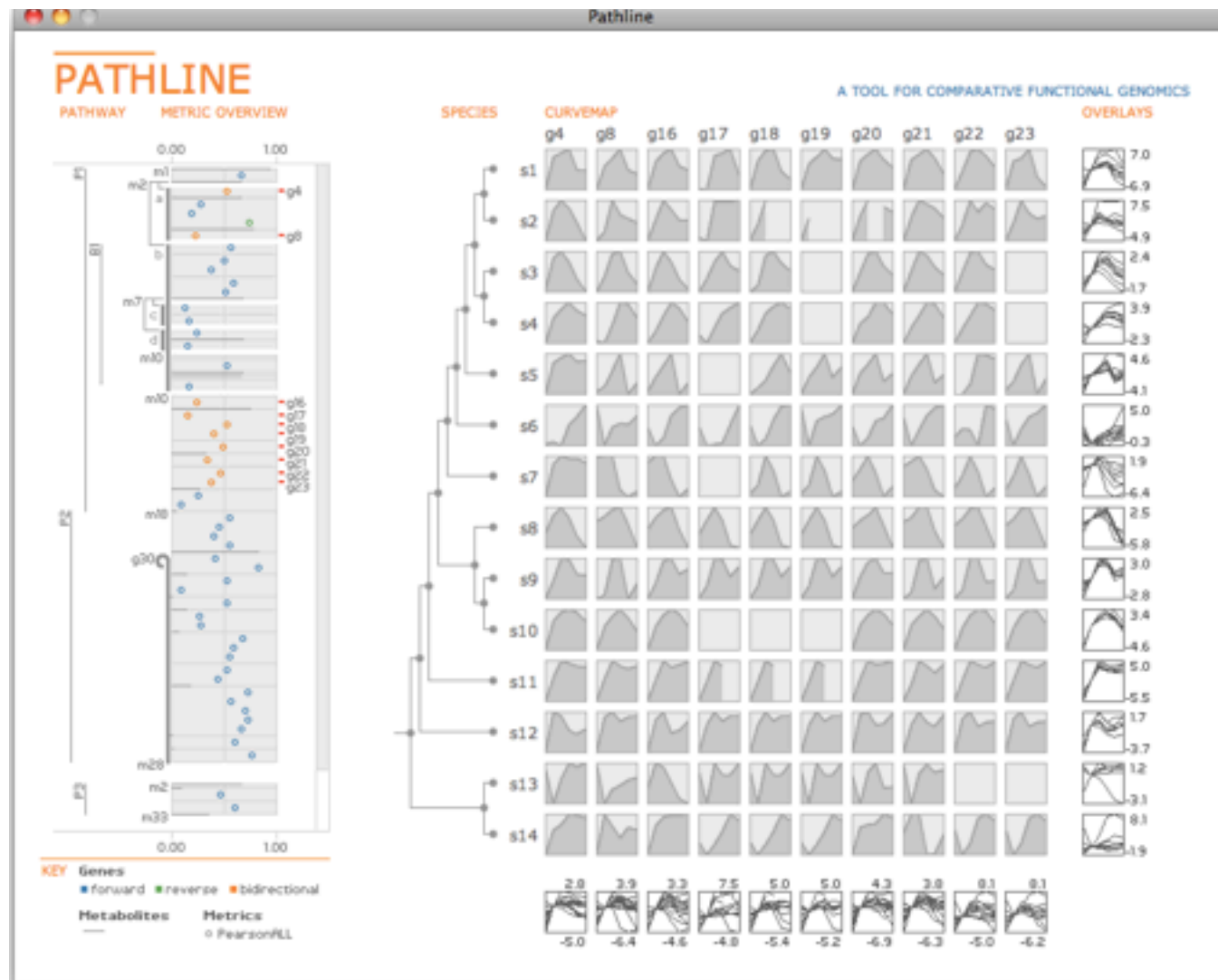
courtesy of M. Styczynski from
JavaTreeview
jtreeview.sourceforge.net/

curvemaps



Curvemap

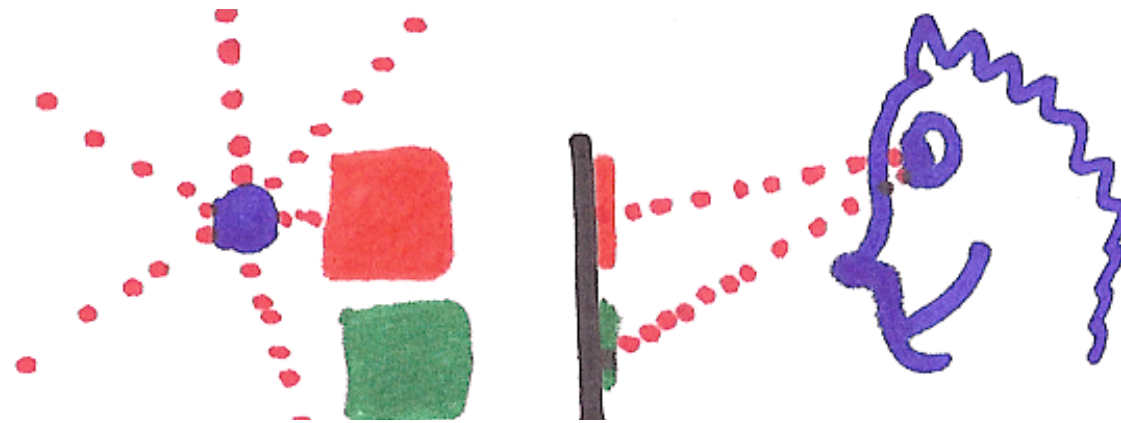
- shape perception easier for filled framed line charts than colored boxes



Pathline: A Tool for Comparative Functional Genomics.
Meyer, Wong, Styczynski, Munzner, Pfister. EuroVis 2010.

Dangers of depth

- rankings for **planar** spatial position, not depth!
- we don't really live in 3D: we **see** in 2.05D
 - up/down and sideways: image plane
 - acquire more info quickly from eye movements
 - away: depth into scene
 - only acquire more info from head/body motion

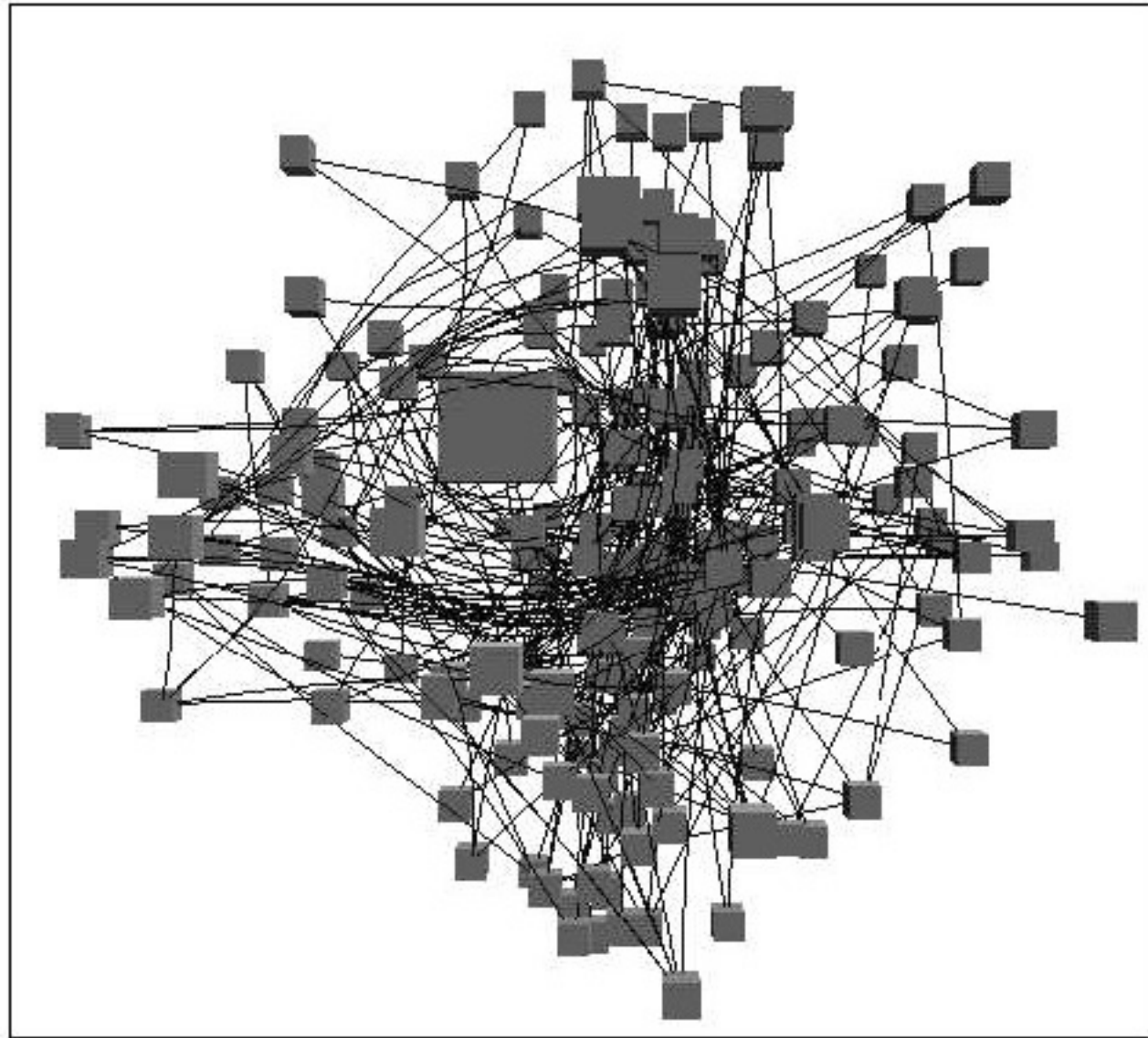


- further reading

Visual Thinking for Design (Chap 5). Colin Ware. 2008

Dangers of depth: difficulties of 3D

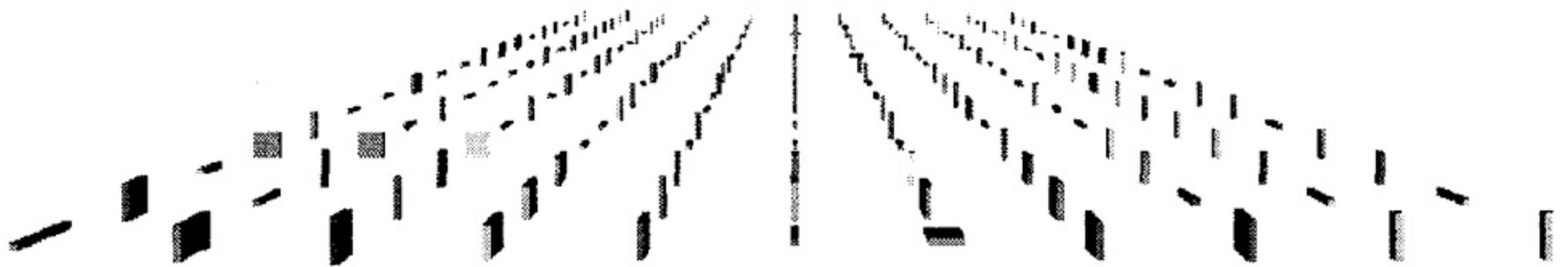
- occlusion
- interaction complexity



Distortion Viewing Techniques for 3D Data. Carpendale et al. InfoVis I 1996.

Dangers of depth: difficulties of 3D

- perspective distortion
 - interferes with all size channel encodings
 - power of the plane is lost!



*Visualizing the Results of Multimedia Web Search Engines.
Mukherjea, Hirata, and Hara. InfoVis 96*

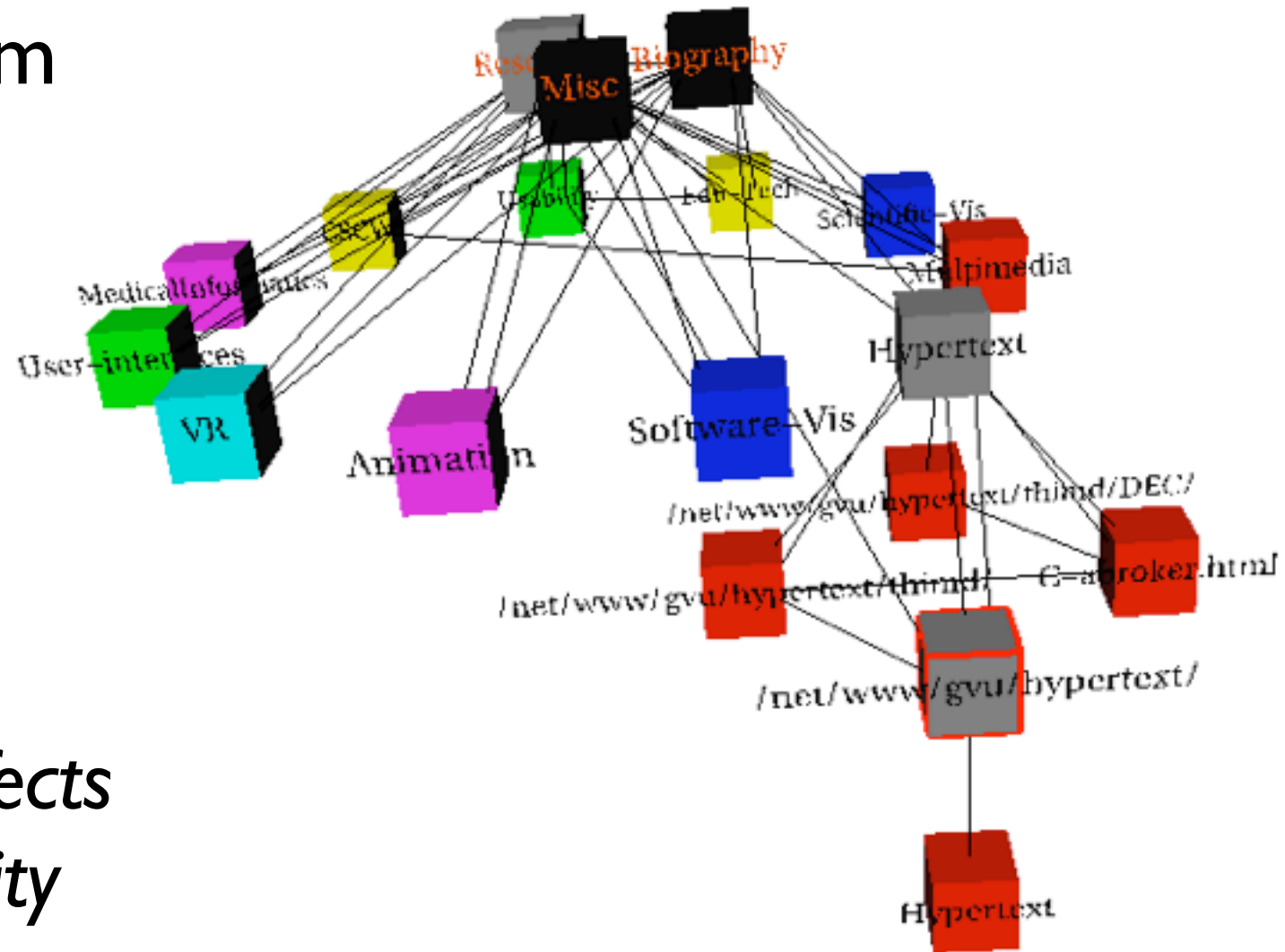
Dangers of depth: difficulties of 3D

- text legibility
 - far worse when tilted from image plane

- further reading

Exploring and Reducing the Effects of Orientation on Text Readability in Volumetric Displays.

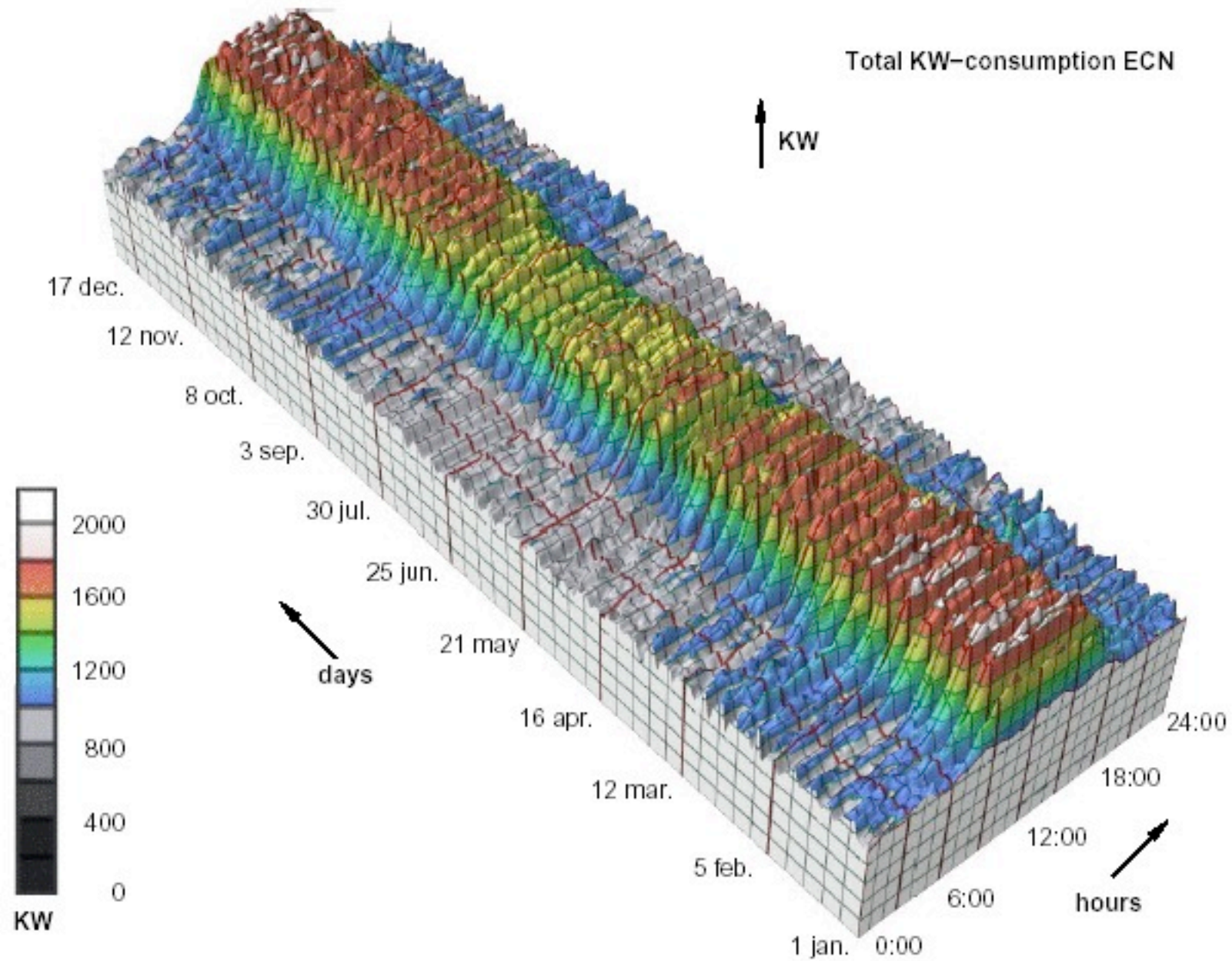
Grossman et al. CHI 2007



Visualizing the World-Wide Web with the Navigational View Builder.
Mukherjea and Foley. Computer Networks and ISDN Systems, 1995.

Dangers of depth example

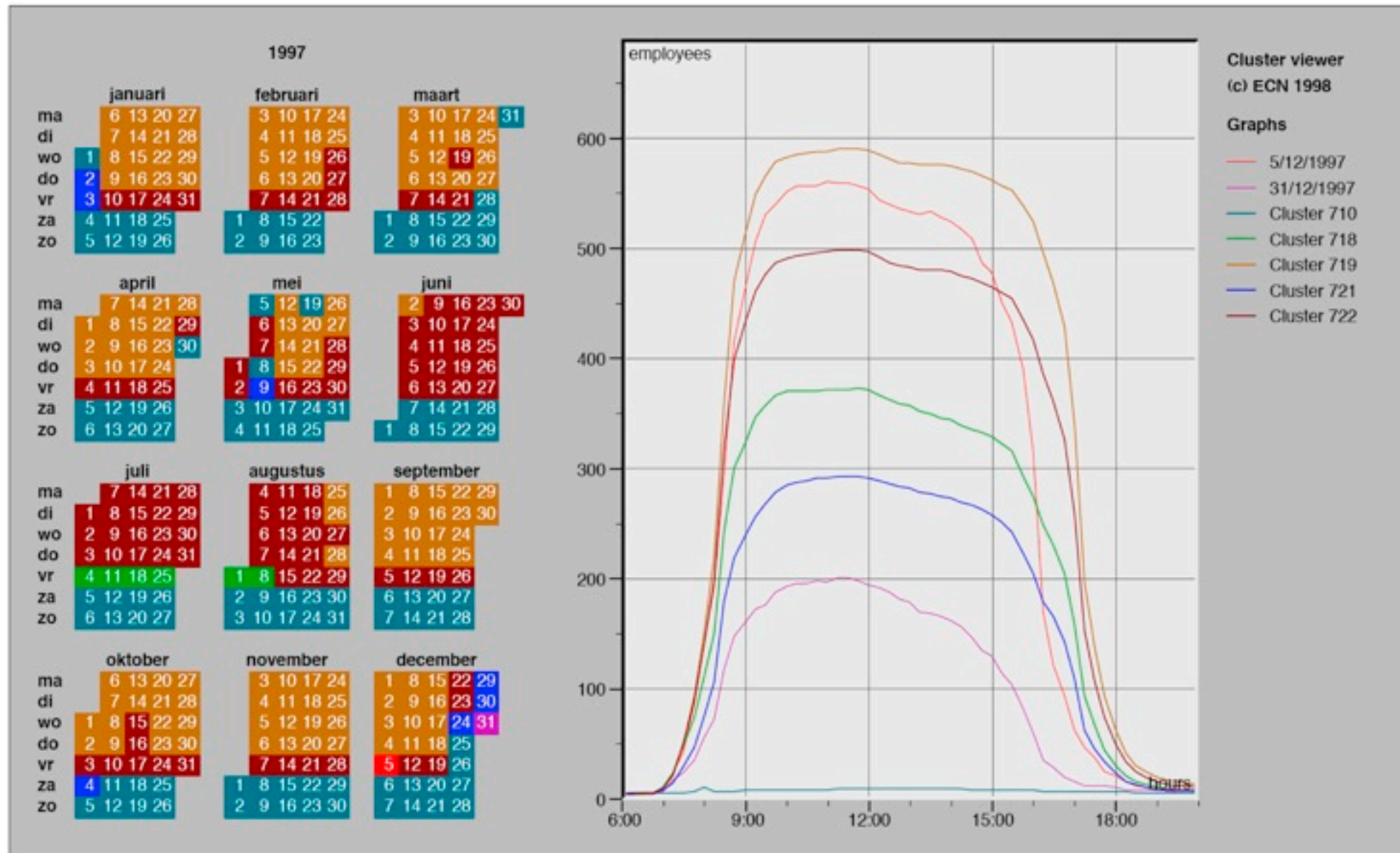
- extruded curves: detailed comparisons impossible



*Cluster and Calendar based Visualization of Time Series Data.
van Wijk and van Selow, Proc InfoVis 99.*

Transformation to suitable abstraction

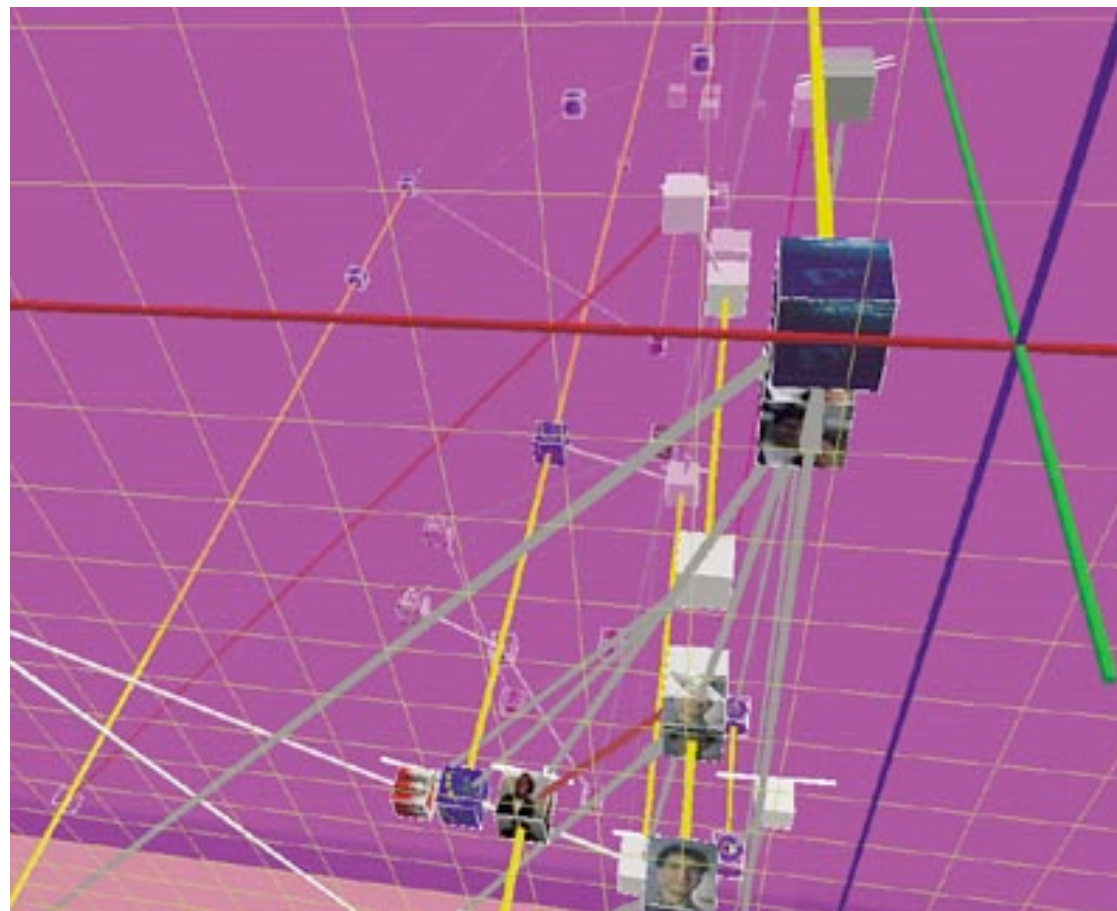
- derived data: clusters
- multiple views: calendar, superimposed 2D curves



*Cluster and Calendar based Visualization of Time Series Data.
van Wijk and van Selow, Proc InfoVis 99.*

Dangers of depth: must justify

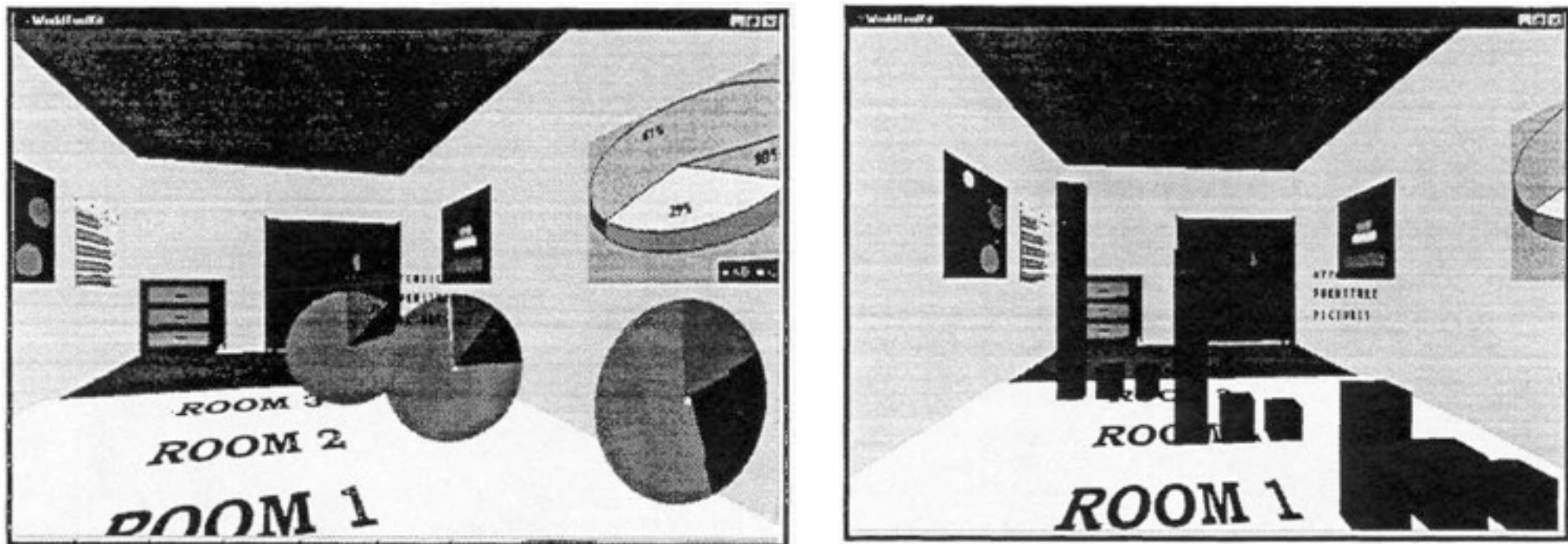
- 3D legitimate for true 3D spatial data
- 3D needs very careful justification **for abstract data**
 - enthusiasm in 1990s, but now skepticism
 - be especially careful with 3D for point clouds or networks



WEBPATH-a three dimensional Web history. Frecon and Smith. InfoVis 1999

Resolution beats immersion

- immersion typically not helpful **for abstract data**
 - do not need sense of presence or stereoscopic 3D
- resolution much more important
 - pixels are the scarcest resource
 - desktop also better for workflow integration
- virtual reality for abstract data very difficult to justify



*Development of an information visualization tool using virtual reality.
Kirner and Martins. Symp Applied Computing 2000*

Eyes beat memory

- principle: external cognition vs. internal memory
 - easy to compare by moving eyes between side-by-side views
 - harder to compare visible item to memory of what you saw
- implications for animation
 - great for choreographed storytelling
 - great for transitions between two states
 - poor for many states with changes everywhere
 - consider small multiples instead

literal

abstract

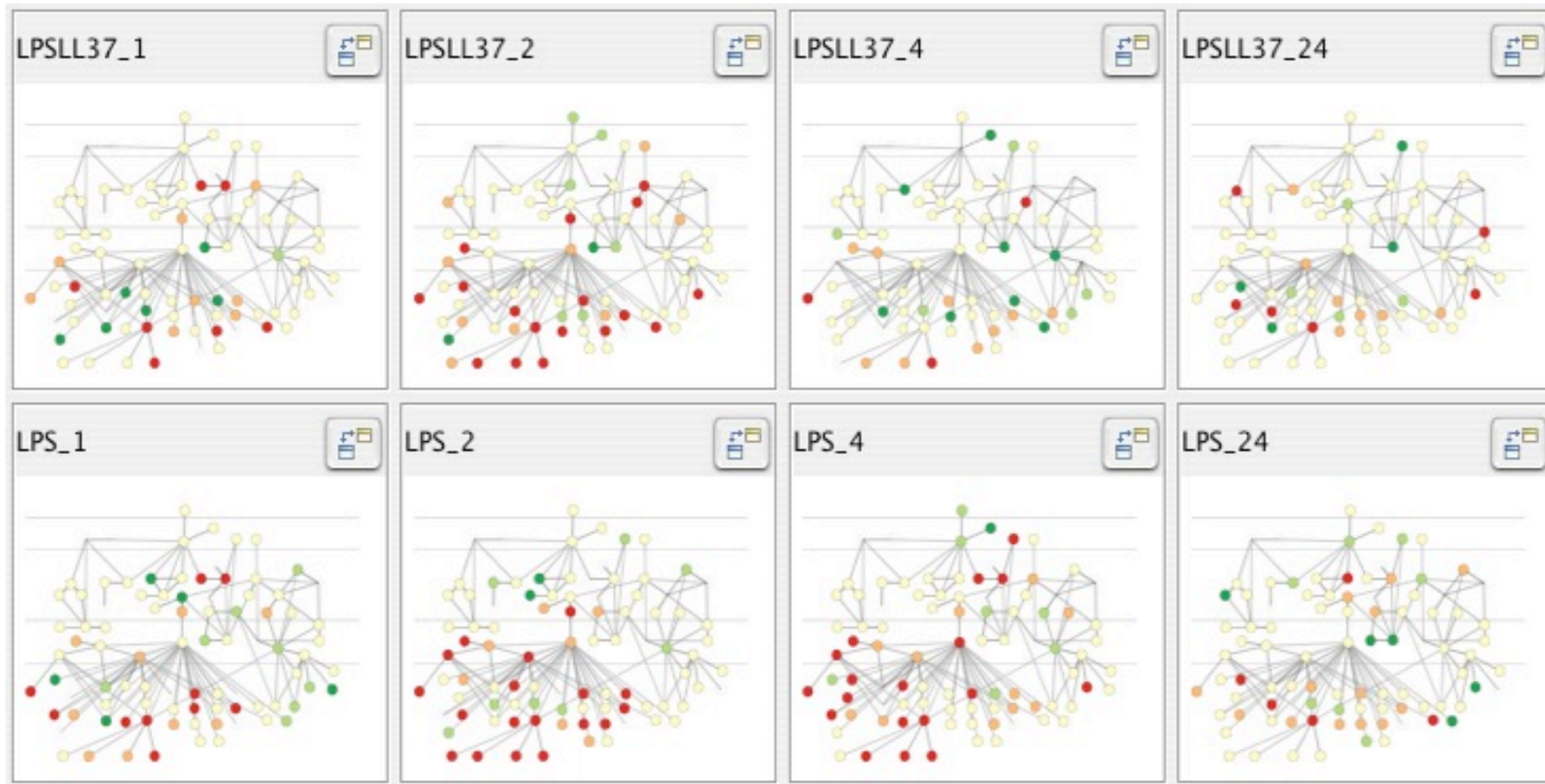
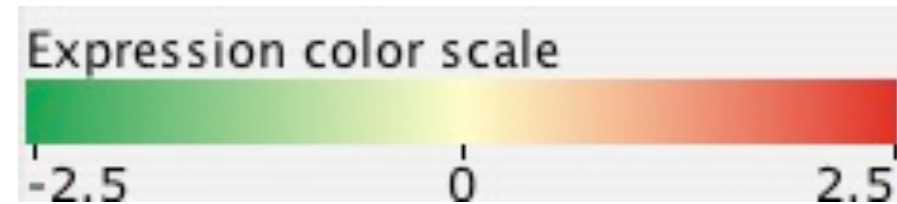
animation

small multiples



Small multiples example: Cerebral

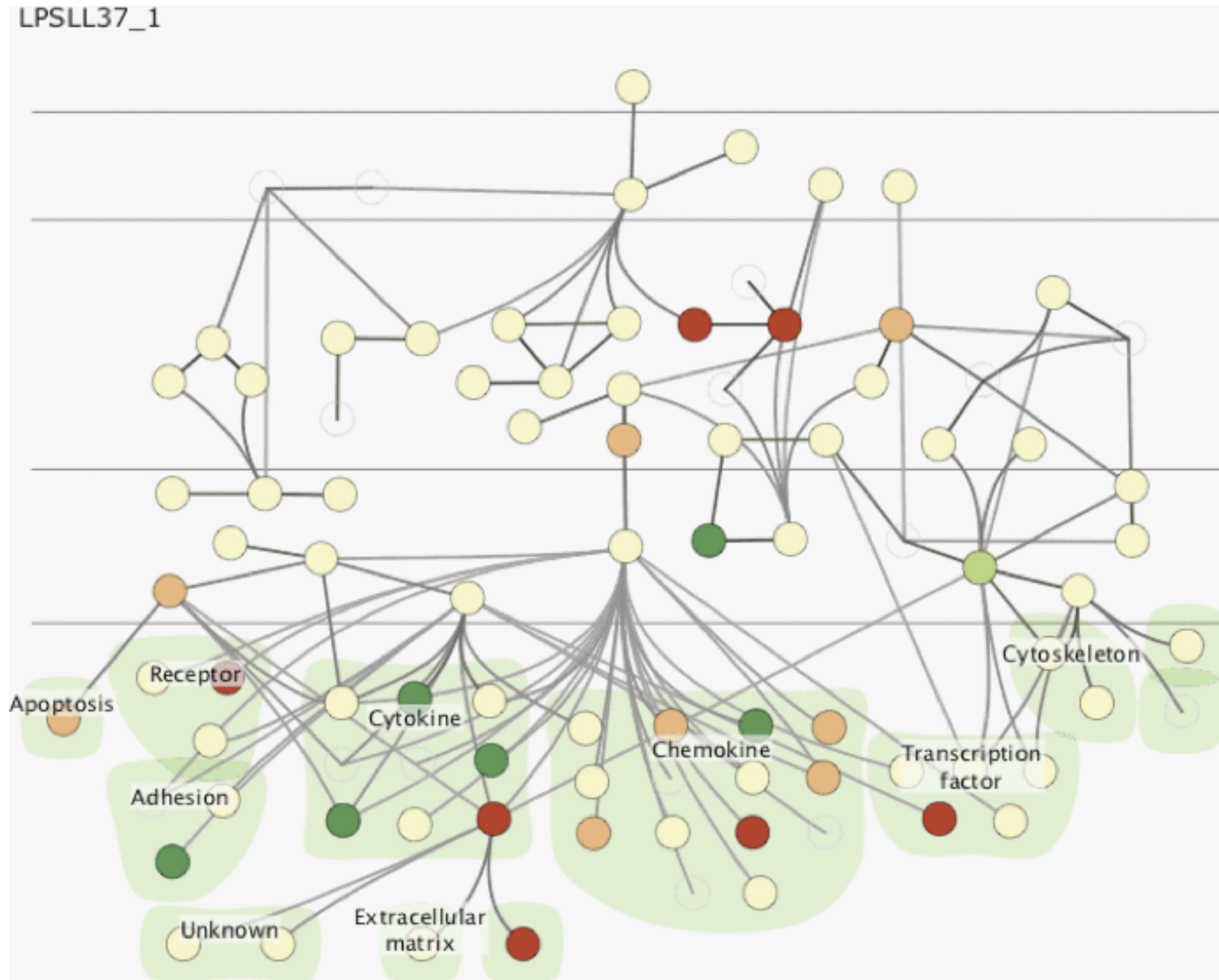
- small multiples: one graph instance per experimental condition
 - same spatial layout
 - color differently, by condition



Cerebral: Visualizing Multiple Experimental Conditions on a Graph with Biological Context. Barsky, Munzner, Gardy, Kincaid. IEEE InfoVis 2008.

Why not animation?

- global comparison difficult



Why not animation?

- further reading

*Animation: can it facilitate? Tversky et al.
Intl Journ Human-Computer Studies, 57(4):247-262, 2002.*

Beyond encoding and interaction

- three more levels of design questions
 - different threats to validity at each level
- validate against the right threat

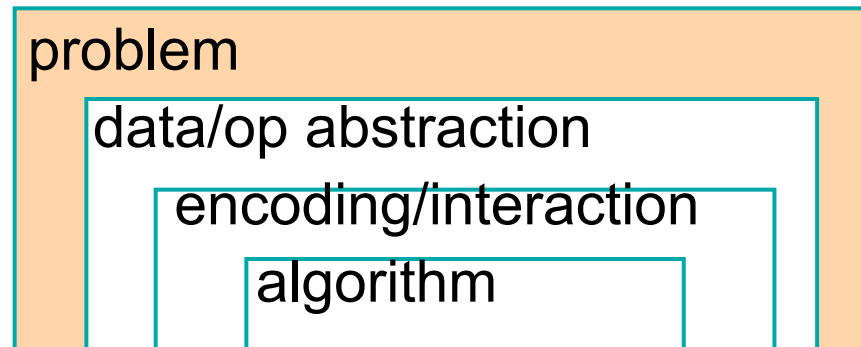
problem: you misunderstood their needs

abstraction: you're showing them the wrong thing

encoding: the way you show it doesn't work

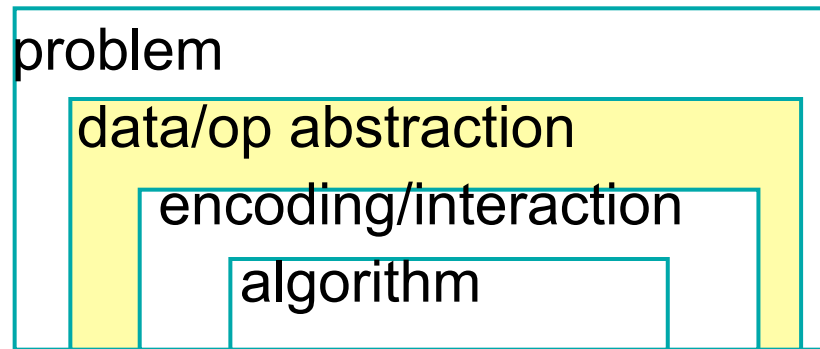
algorithm: your code is too slow

Characterizing problems of real-world users



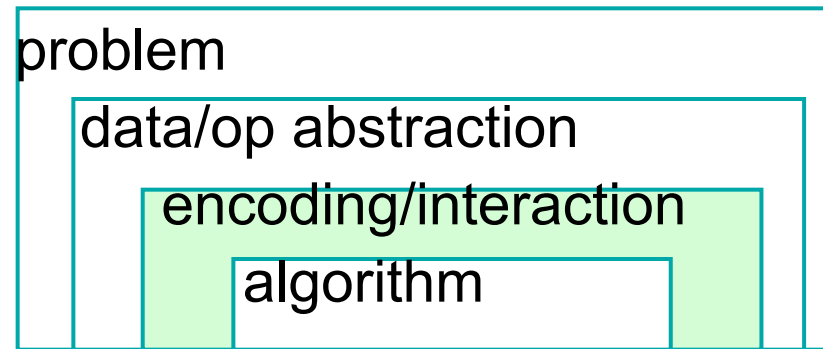
- **identify a problem amenable to vis**
 - provide novel capabilities
 - speed up existing workflow
- **validation**
 - immediate: interview and observe target users
 - downstream: notice adoption rates

Abstracting into operations on data types



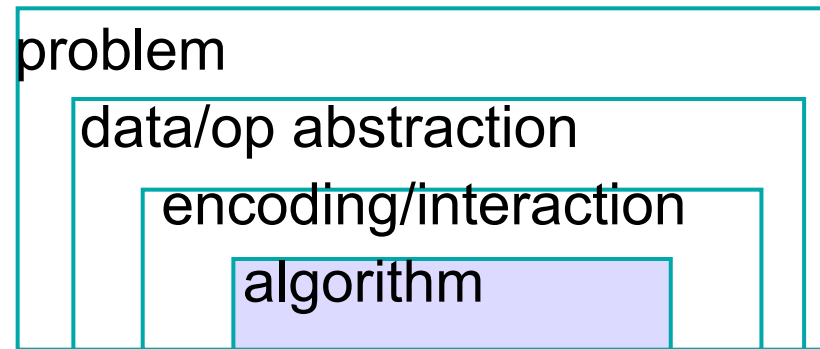
- abstract from domain-specific to generic
- operations
 - sorting, filtering, browsing, comparing, finding trend/outlier, characterizing distributions, finding correlation...
- data types
 - tables of numbers, relational networks, spatial
 - transform into useful configuration: derived data
- validation
 - deploy in the field and observe usage

Designing visual encoding, interaction techniques



- **visual encoding: drawings they are shown**
- **interaction: how they manipulate drawings**
- **validation**
 - immediate: careful justification wrt known principles
 - downstream: qualitative or quantitative analysis of results
 - downstream: lab study measuring time/error on given task
- **focus of this talk**

Creating algorithms to execute techniques



- automatically carry out specification
- validation
 - immediate: complexity analysis
 - downstream: benchmarks for system time, memory

Danger of validation mismatch

- cannot show encoding good with system timings
- cannot show abstraction good with lab study

problem validate: observe target users

encoding validate: justify design wrt alternatives

algorithm validate: measure system time

encoding validate: lab study, qualitative analysis

abstraction validate: observe real usage in field

Principles recap

- know your visual channel types and ranks
- categorical color constraints
- power of the plane
- danger of depth
- resolution beats immersion
- eyes beat memory

- validate against the right threat

More information

- vis intro book chapter
 - principles in more depth
 - also, techniques!

<http://www.cs.ubc.ca/~tmm/papers.html#akpchapter>

- papers, videos, software, talks, courses

<http://www.cs.ubc.ca/~tmm>

- this talk

<http://www.cs.ubc.ca/~tmm/talks.html#twitter12>