

Visualization Analysis & Design

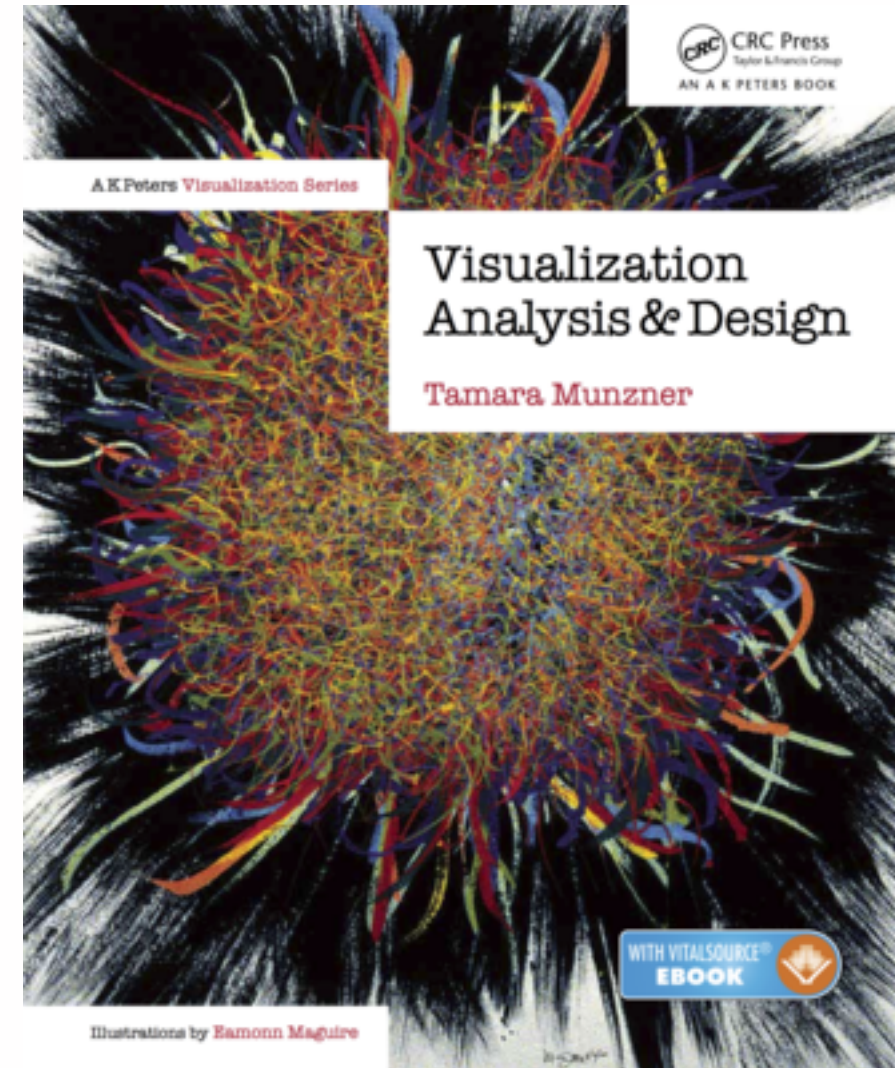
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*UBC STAT 545A Guest Lecture
October 22 2015, Vancouver BC*

<http://www.cs.ubc.ca/~tmm/talks.html#vad15bryan>

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Defining visualization (vis)

Computer-based visualization systems provide visual representations of datasets designed to help people carry out tasks more effectively.

Why?...

Why have a human in the loop?

Computer-based visualization systems provide visual representations of datasets designed to help people carry out tasks more effectively.

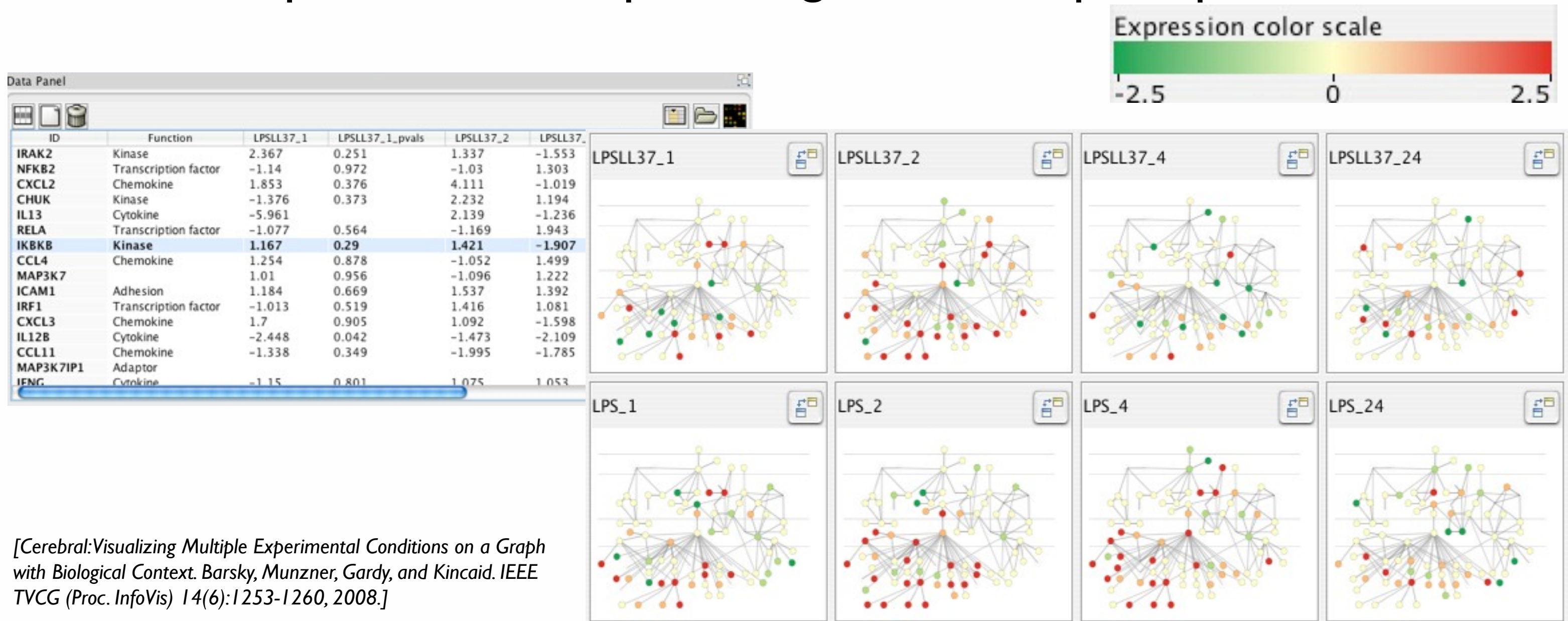
Visualization is suitable when there is a need to augment human capabilities rather than replace people with computational decision-making methods.

- don't need vis when fully automatic solution exists and is trusted
- many analysis problems ill-specified
 - don't know exactly what questions to ask in advance
- possibilities
 - long-term use for end users (e.g. exploratory analysis of scientific data)
 - presentation of known results
 - stepping stone to better understanding of requirements before developing models
 - help developers of automatic solution refine/debug, determine parameters
 - help end users of automatic solutions verify, build trust

Why use an external representation?

Computer-based visualization systems provide **visual representations** of datasets designed to help people carry out tasks more effectively.

- external representation: replace cognition with perception



[Cerebral: Visualizing Multiple Experimental Conditions on a Graph with Biological Context. Barsky, Munzner, Gardy, and Kincaid. IEEE TVCG (Proc. InfoVis) 14(6):1253-1260, 2008.]

Why represent all the data?

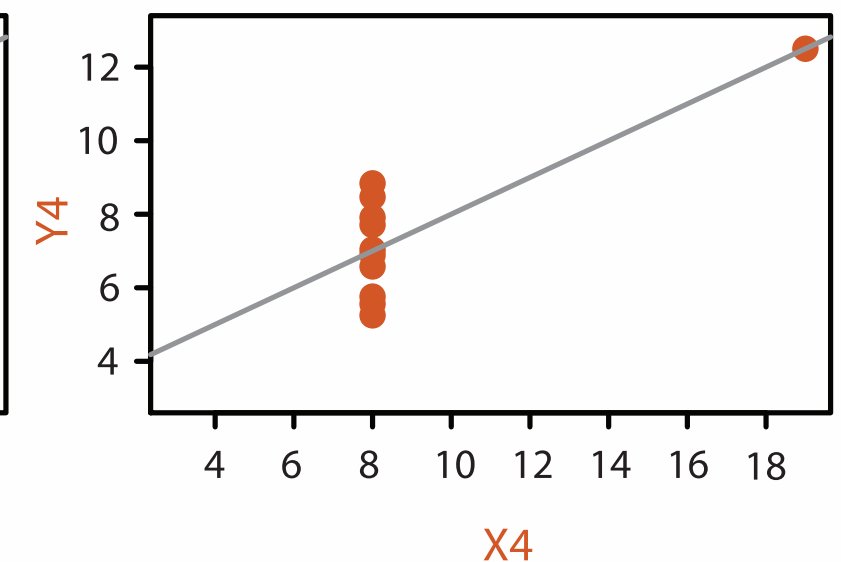
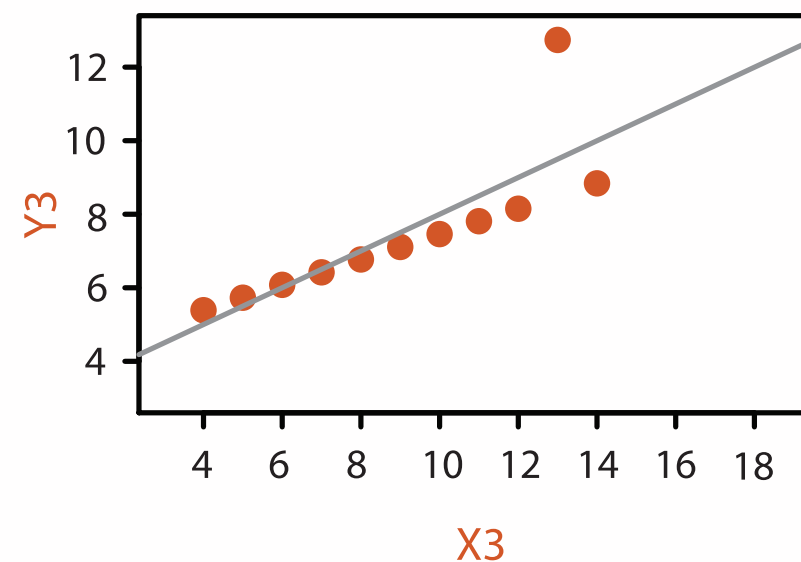
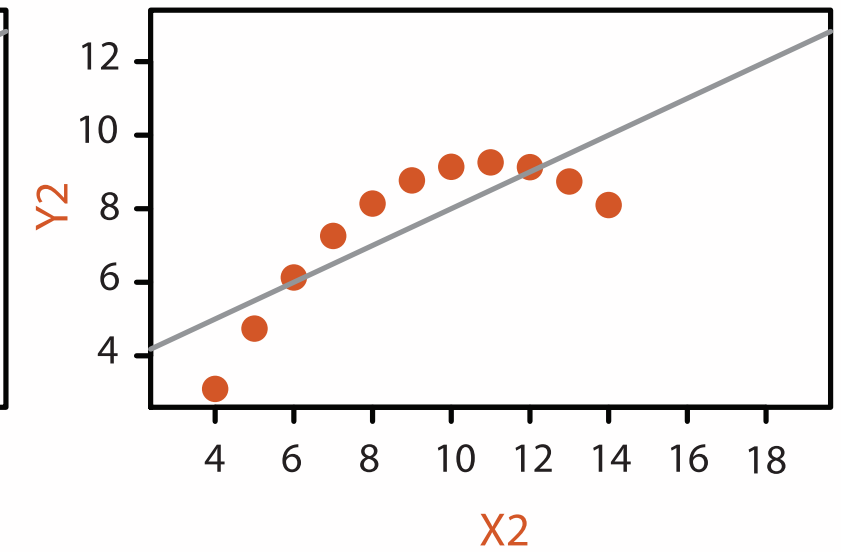
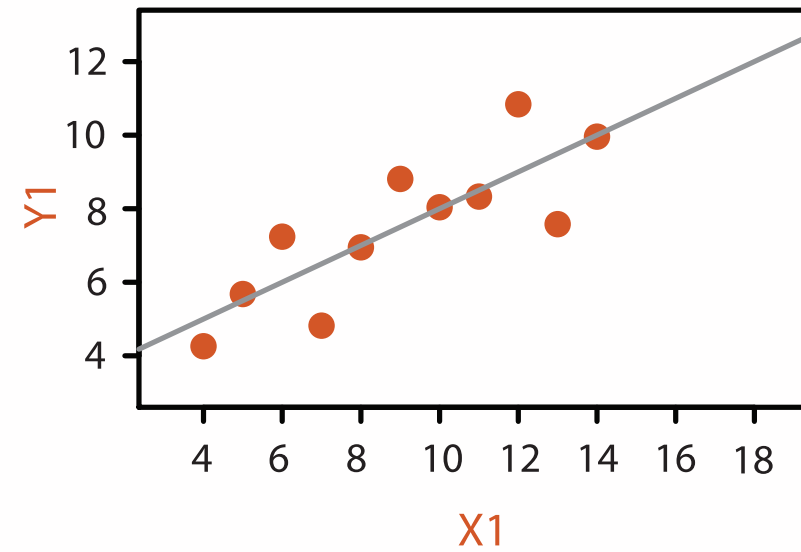
Computer-based visualization systems provide visual representations of datasets designed to help people carry out tasks more effectively.

- summaries lose information, details matter
 - confirm expected and find unexpected patterns
 - assess validity of statistical model

Anscombe's Quartet

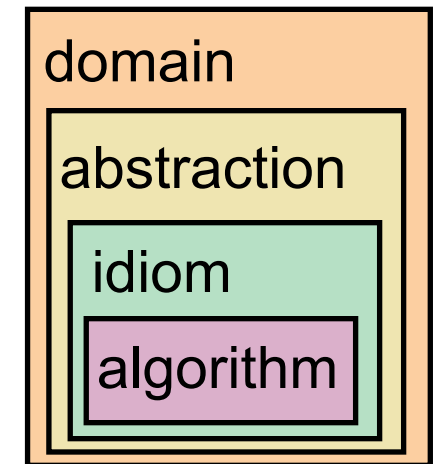
Identical statistics

x mean	9
x variance	10
y mean	8
y variance	4
x/y correlation	1

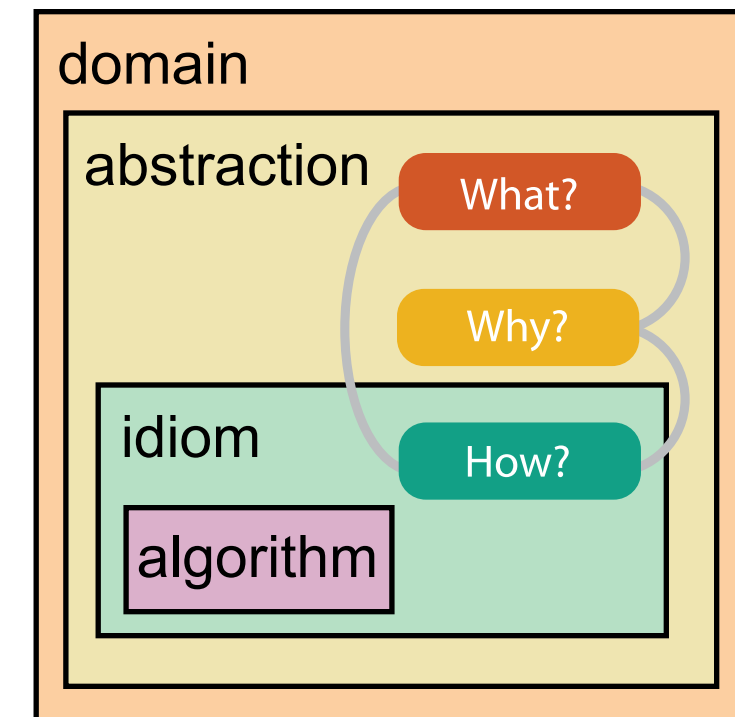


Analysis framework: Four levels, three questions

- *domain* situation
 - who are the target users?
- *abstraction*
 - translate from specifics of domain to vocabulary of vis
- **what** is shown? **data abstraction**
 - often don't just draw what you're given: transform to new form
- **why** is the user looking at it? **task abstraction**
- *idiom*
 - **how** is it shown?
 - **visual encoding idiom**: how to draw
 - **interaction idiom**: how to manipulate
- *algorithm*
 - efficient computation



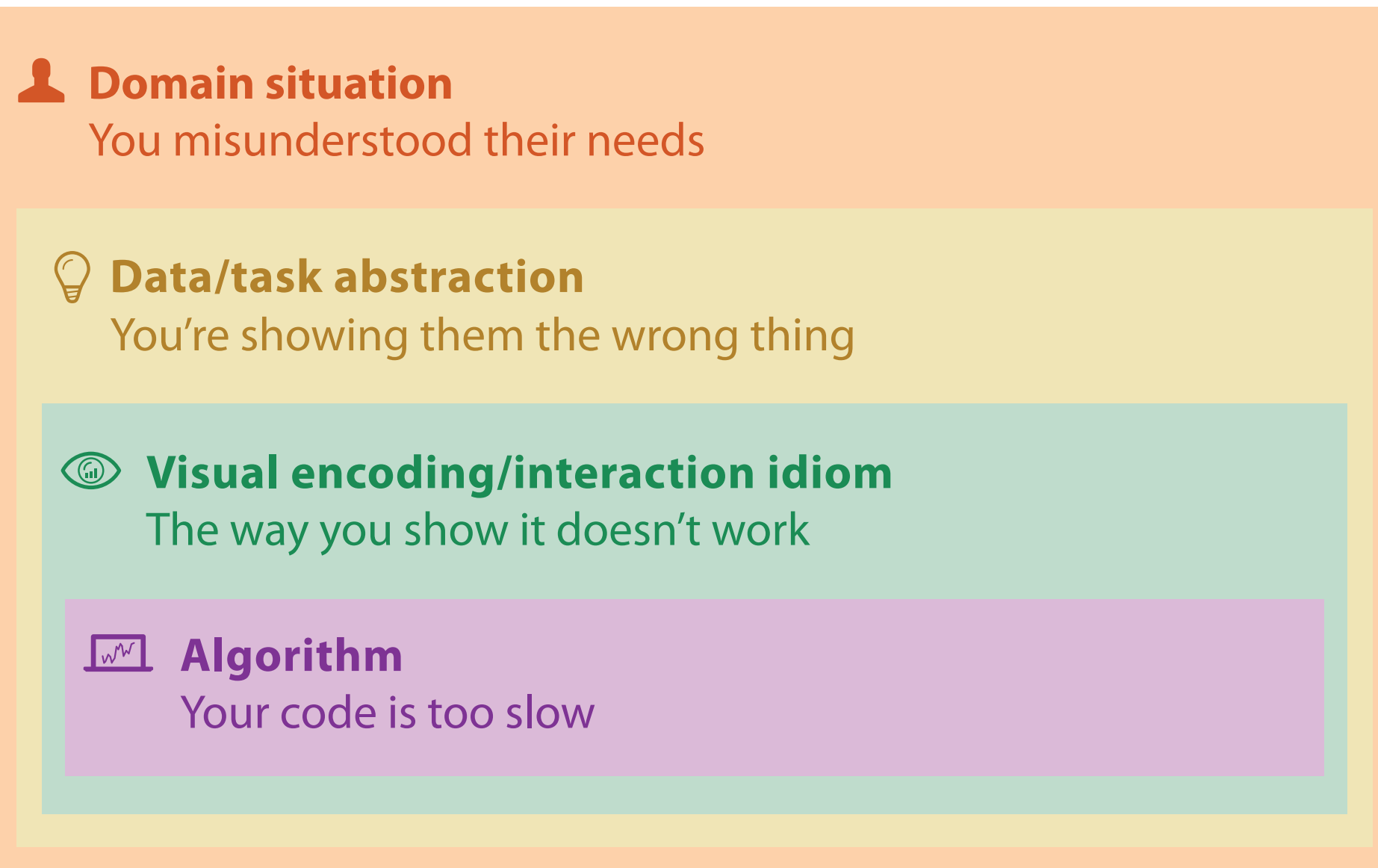
[A Nested Model of Visualization Design and Validation.
Munzner. *IEEE TVCG* 15(6):921-928, 2009 (Proc. InfoVis 2009).]



[A Multi-Level Typology of Abstract Visualization Tasks
Brehmer and Munzner. *IEEE TVCG* 19(12):2376-2385, 2013 (Proc. InfoVis 2013).]

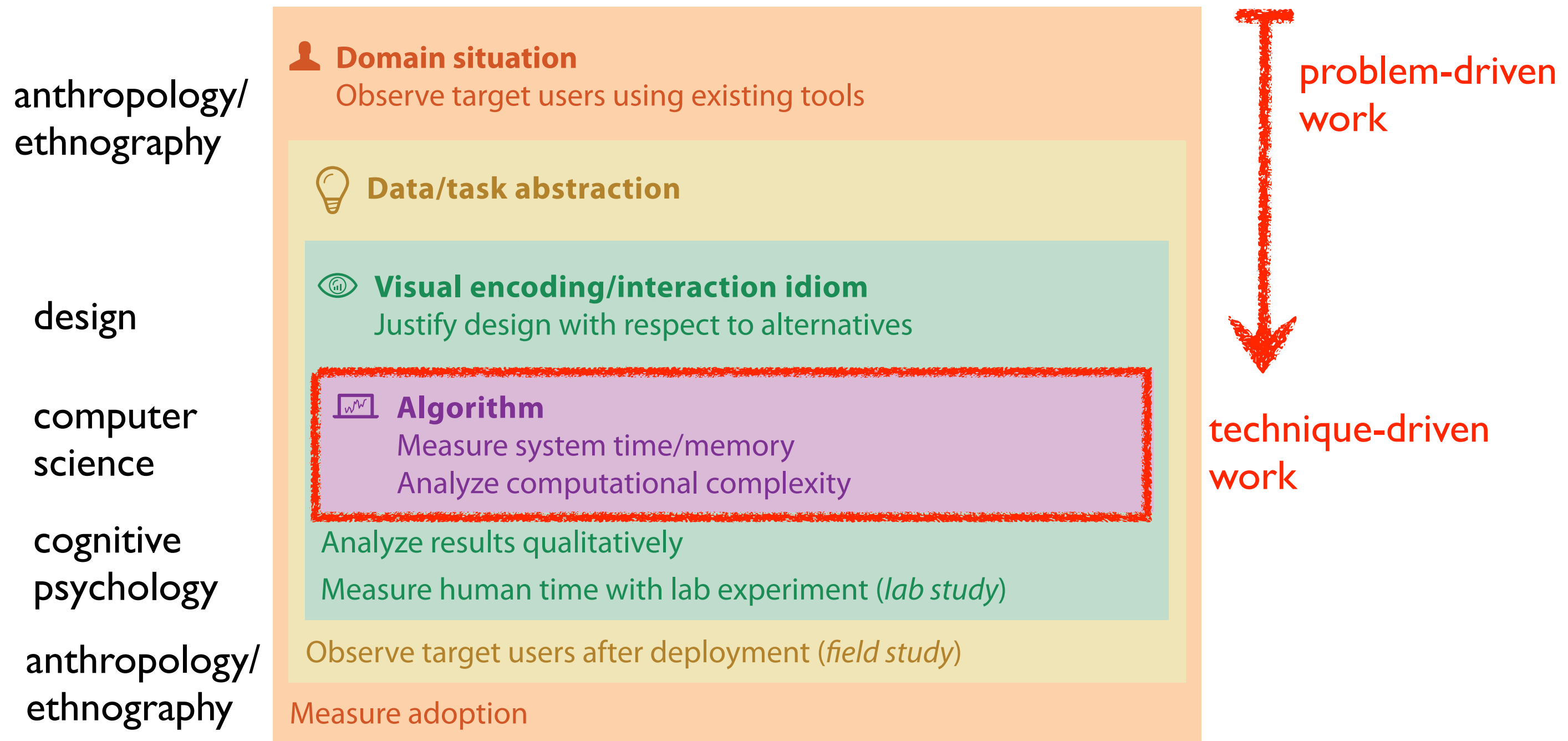
Why is validation difficult?

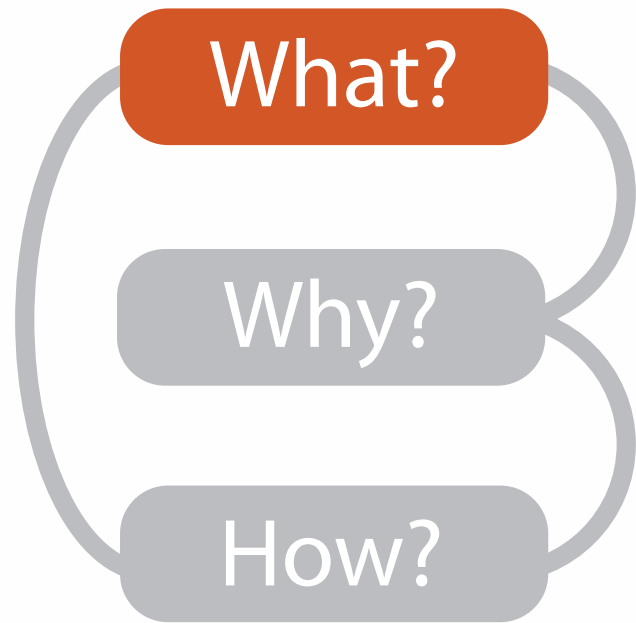
- different ways to get it wrong at each level



Why is validation difficult?

- solution: use methods from different fields at each level





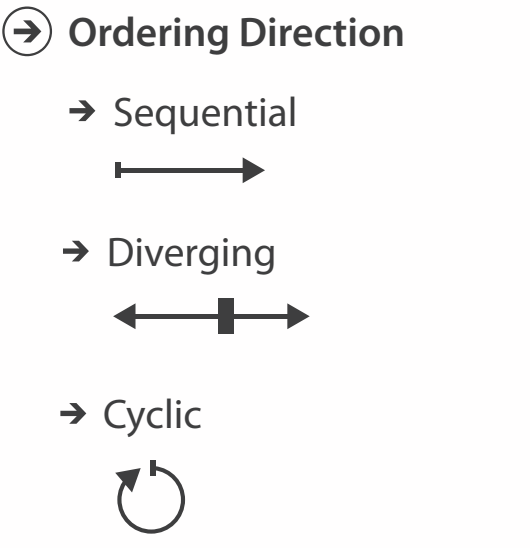
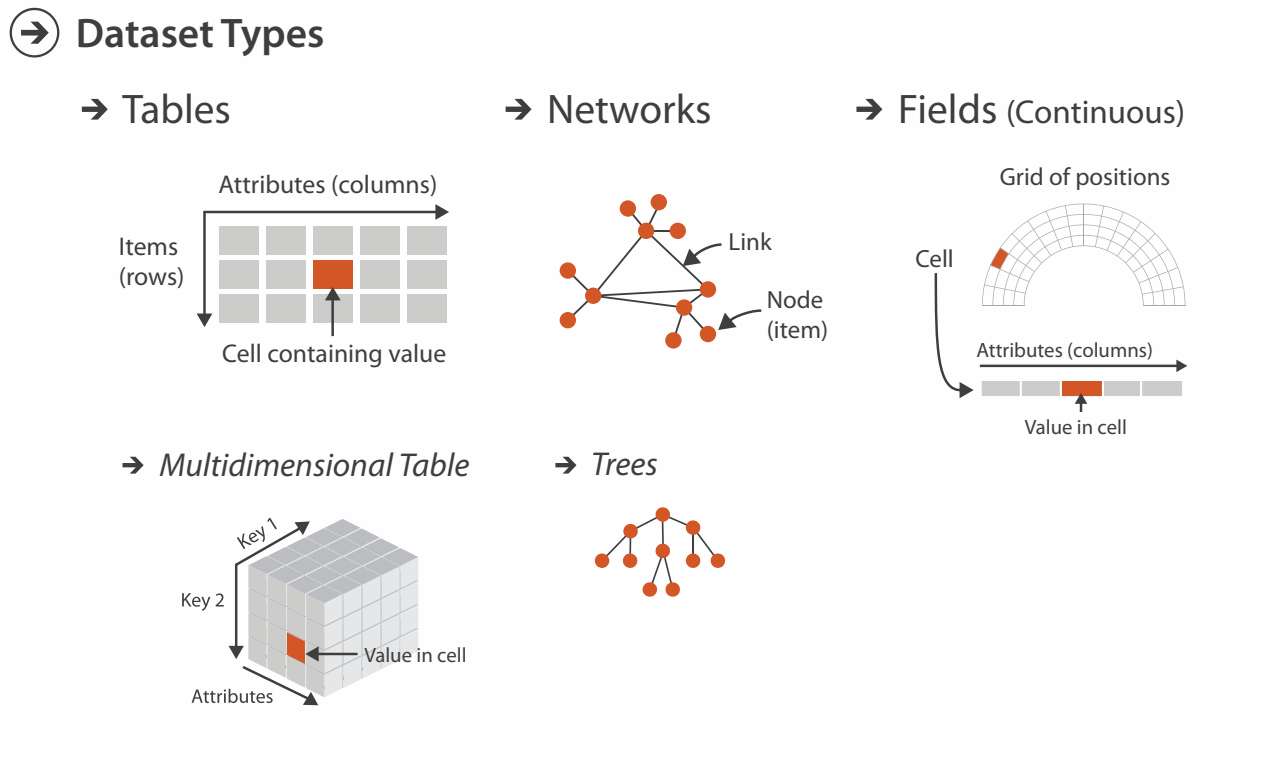
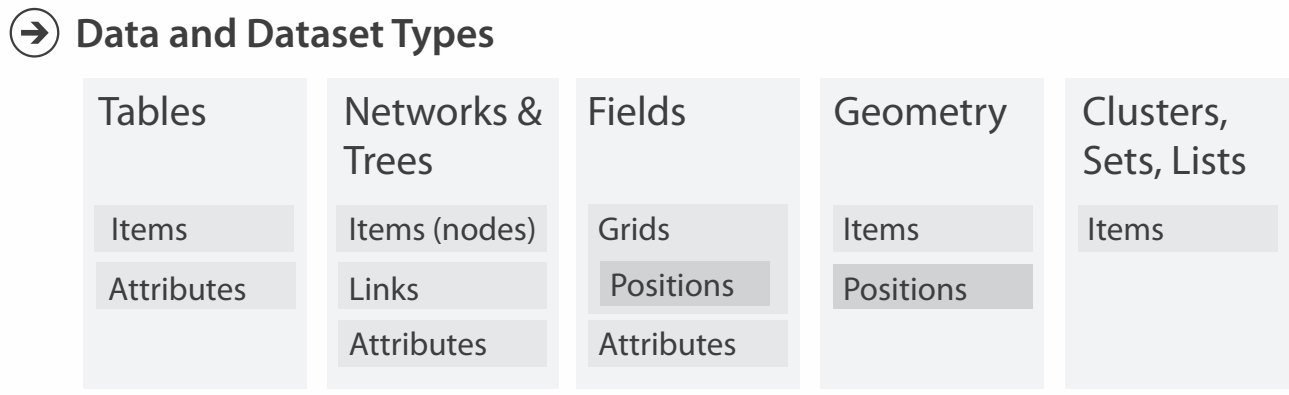
What?

Datasets

Attributes

- ➔ Data Types
 - ➔ Items
 - ➔ Attributes
 - ➔ Links
 - ➔ Positions
 - ➔ Grids

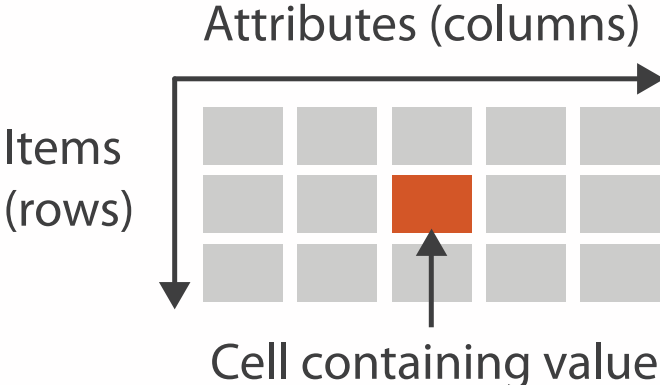
- ➔ Attribute Types
 - ➔ Categorical
 - + ● ■ ▲
 - ➔ Ordered
 - ➔ Ordinal
 - 👕 👕 👕
 - ➔ Quantitative
 - ┆ ┆ ┆



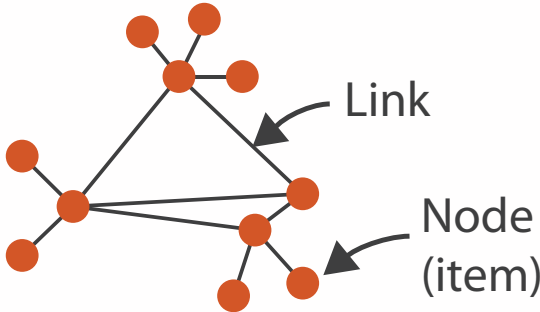
Types: Datasets and data

→ Dataset Types

→ Tables

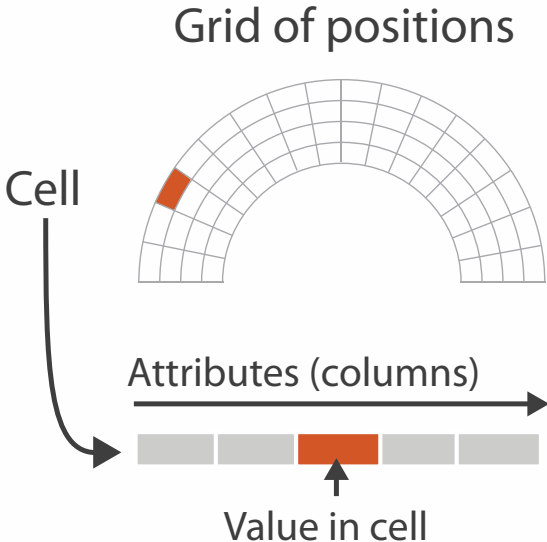


→ Networks



→ Spatial

→ Fields (Continuous)



→ Geometry (Spatial)



→ Attribute Types

→ Categorical

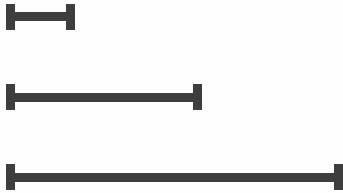


→ Ordered

→ Ordinal



→ Quantitative



Why?

👉 Actions

🎯 Targets

➔ **Analyze**

- ➔ Consume
 - ➔ Discover
 - ➔ Present
 - ➔ Enjoy
- ➔ Produce
 - ➔ Annotate
 - ➔ Record
 - ➔ Derive

➔ **All Data**

- ➔ Trends
- ➔ Outliers
- ➔ Features

➔ **Attributes**

- ➔ One
 - ➔ Distribution
 - ➔ Extremes
- ➔ Many
 - ➔ Dependency
 - ➔ Correlation
 - ➔ Similarity

➔ **Search**

	Target known	Target unknown
Location known	<i>Lookup</i>	<i>Browse</i>
Location unknown	<i>Locate</i>	<i>Explore</i>

➔ **Network Data**

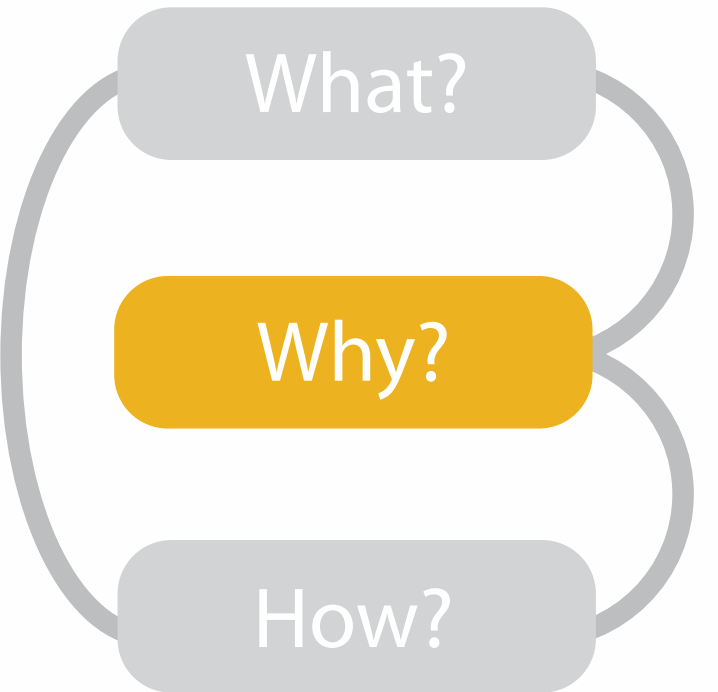
- ➔ Topology
- ➔ Paths

➔ **Query**

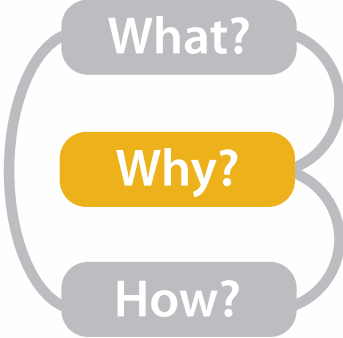
- ➔ Identify
- ➔ Compare
- ➔ Summarize

➔ **Spatial Data**

- ➔ Shape



- {action, target} pairs
 - discover distribution
 - compare trends
 - locate outliers
 - browse topology



Actions: Analyze

- consume
 - discover vs present
 - classic split
 - aka explore vs explain
 - enjoy
- produce
 - newcomer
 - aka casual, social
- produce
 - annotate, record
 - derive
 - crucial design choice

→ Analyze

→ Consume

→ Discover



→ Present

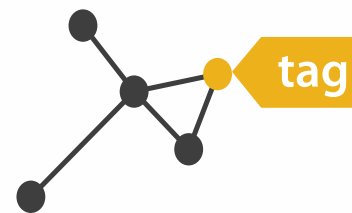


→ Enjoy



→ Produce

→ Annotate



→ Record

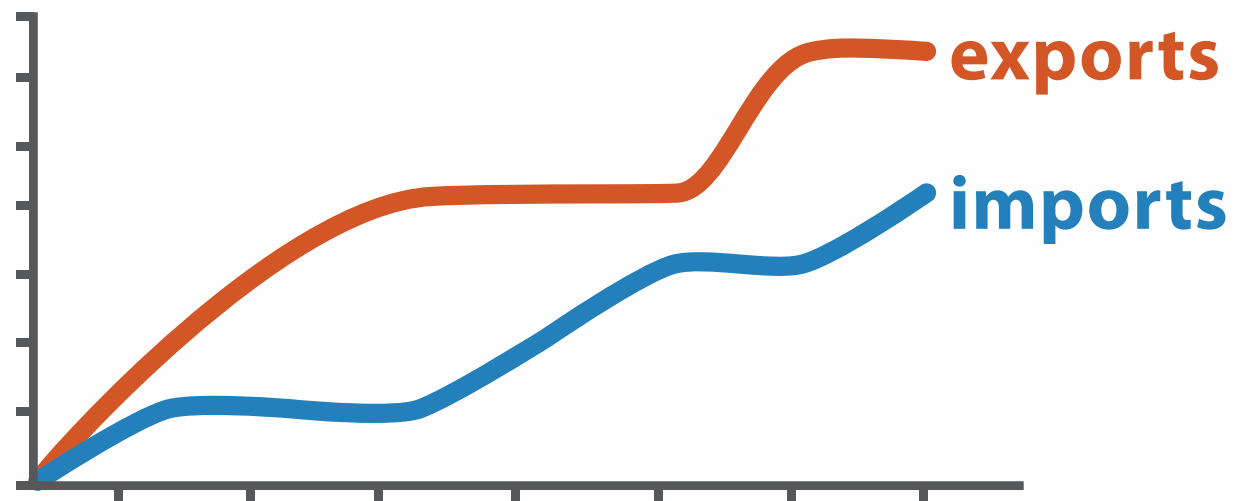


→ Derive

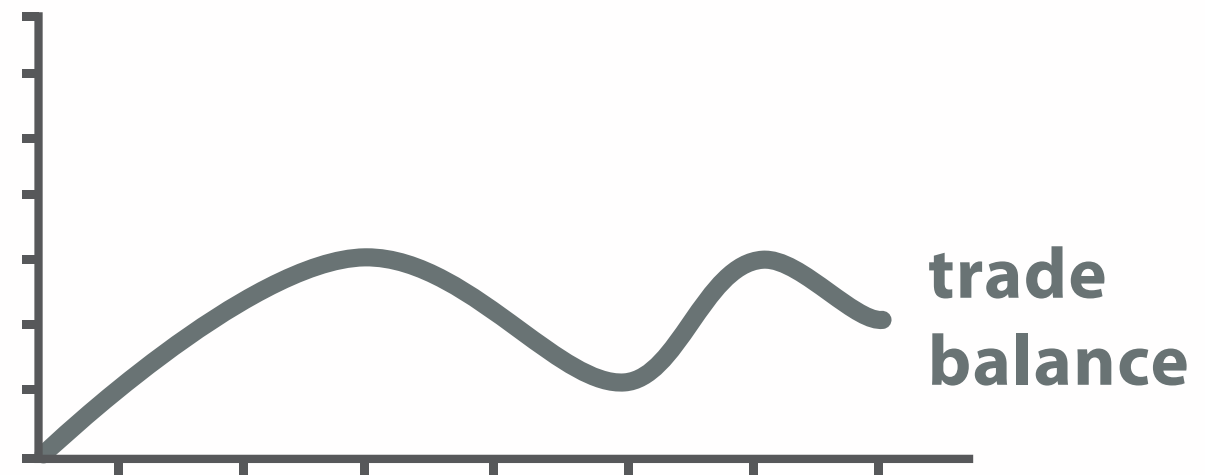


Derive

- don't just draw what you're given!
 - decide what the right thing to show is
 - create it with a series of transformations from the original dataset
 - draw that
- one of the four major strategies for handling complexity



Original Data



$$\text{trade balance} = \text{exports} - \text{imports}$$

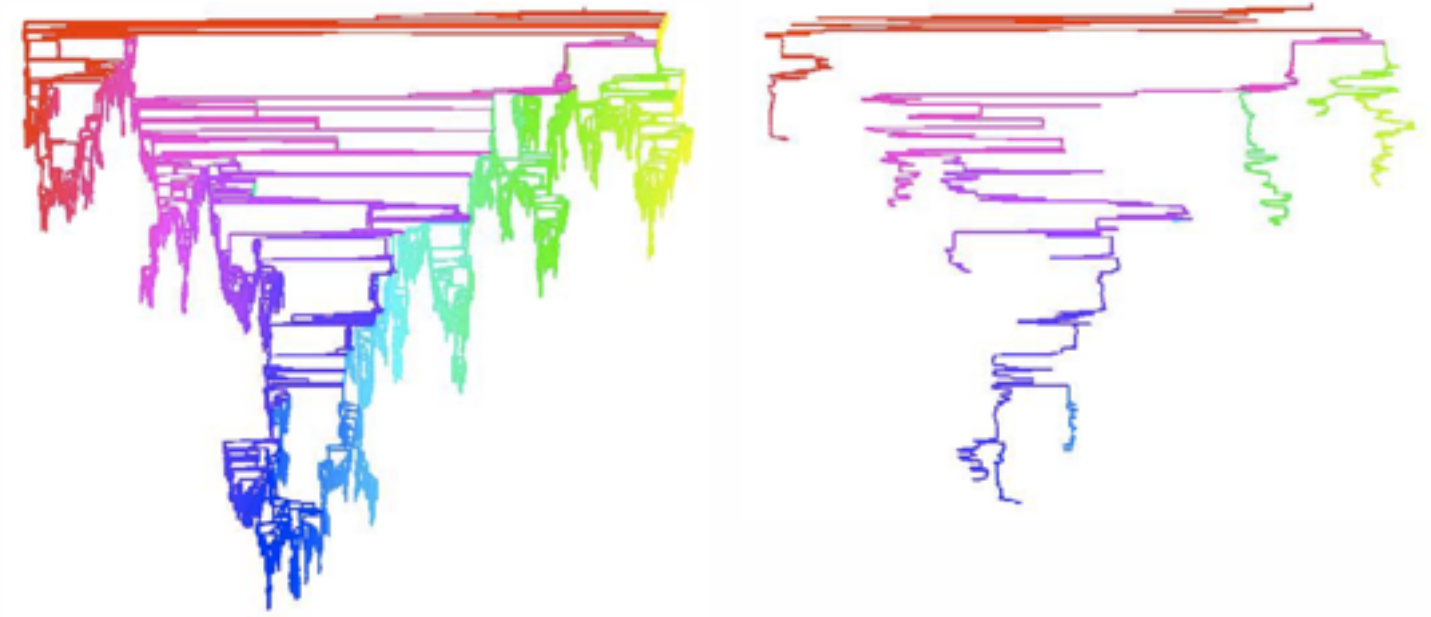
Derived Data

Analysis example: Derive one attribute

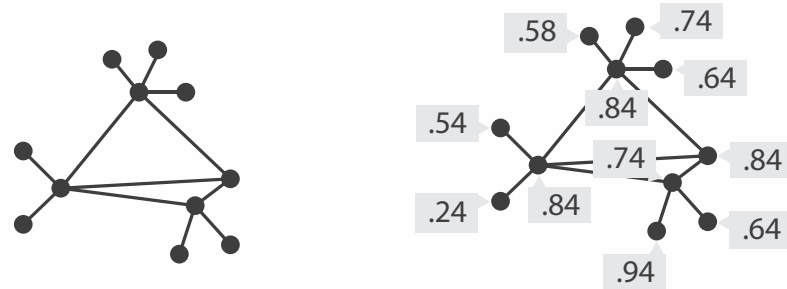
- Strahler number

- centrality metric for trees/networks
- derived quantitative attribute
- draw top 5K of 500K for good skeleton

[Using Strahler numbers for real time visual exploration of huge graphs. Auber. Proc. Intl. Conf. Computer Vision and Graphics, pp. 56–69, 2002.]



Task 1



In
Tree

➔

Out
Quantitative
attribute on nodes

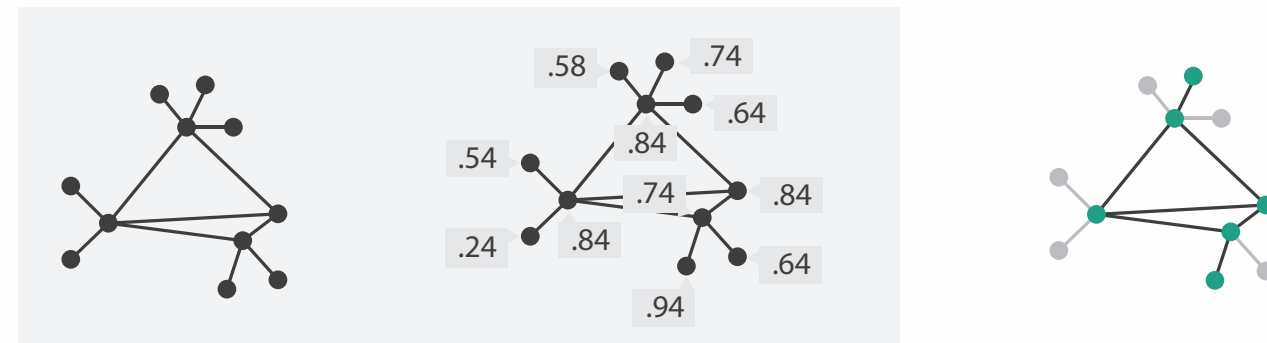
What?

- ➔ In Tree
- ➔ Out Quantitative attribute on nodes

Why?

- ➔ Derive

Task 2



In
Tree

+

In
Quantitative
attribute on nodes

➔

Out
Filtered Tree
Removed
unimportant parts

What?

- ➔ In Tree
- ➔ In Quantitative attribute on nodes
- ➔ Out Filtered Tree

Why?

- ➔ Summarize
- ➔ Topology





How?

- ➔ Reduce
- ➔ Filter

Actions: Search, query

- what does user know?
 - target, location
- how much of the data matters?
 - one, some, all

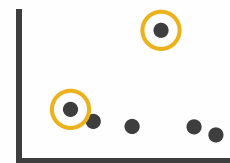
➔ Search

	Target known	Target unknown
Location known	 <i>Lookup</i>	 <i>Browse</i>
Location unknown	 <i>Locate</i>	 <i>Explore</i>

- independent choices for each of these three levels
 - analyze, search, query
 - mix and match

➔ Query

➔ Identify



➔ Compare



↓ ↑



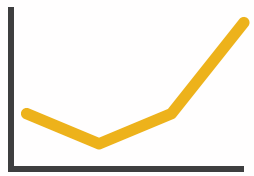
➔ Summarize



Targets

→ All Data

→ Trends



→ Outliers



→ Features



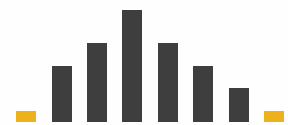
→ Attributes

→ One

→ *Distribution*

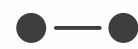


→ *Extremes*

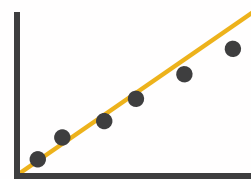


→ Many

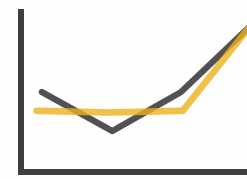
→ *Dependency*



→ *Correlation*

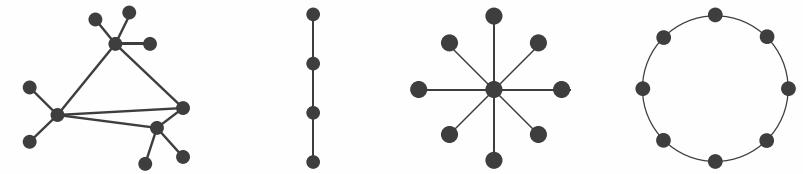


→ *Similarity*

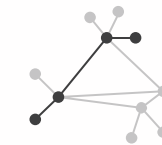


→ Network Data

→ Topology

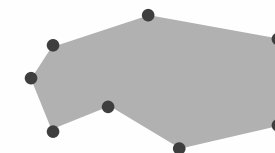


→ *Paths*



→ Spatial Data

→ Shape



How?

Encode

→ Arrange

→ Express



→ Separate



→ Order



→ Align



→ Use



→ Map

from **categorical** and **ordered** attributes

→ Color

→ Hue



→ Saturation



→ Luminance



→ Size, Angle, Curvature, ...



→ Shape



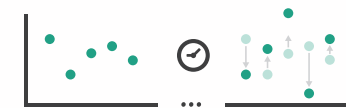
→ Motion

Direction, Rate, Frequency, ...



Manipulate

→ Change



→ Select



→ Navigate

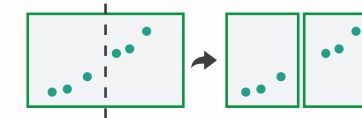


Facet

→ Juxtapose



→ Partition



→ Superimpose



Reduce

→ Filter



→ Aggregate



→ Embed



What?

Why?

How?

How to encode: Arrange space, map channels

Encode

① Arrange

→ Express



→ Order



→ Use



→ Separate



→ Align



② Map

from **categorical** and **ordered** attributes

→ Color

→ Hue



→ Saturation



→ Luminance



→ Size, Angle, Curvature, ...



→ Shape



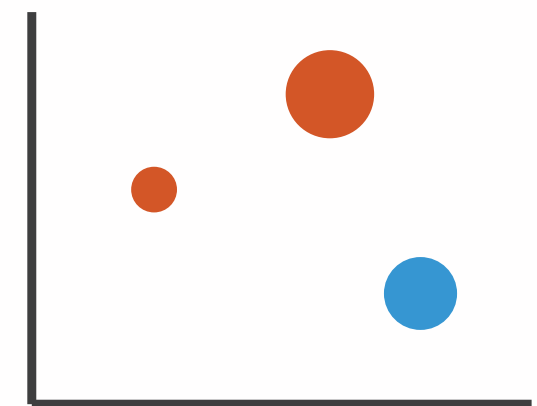
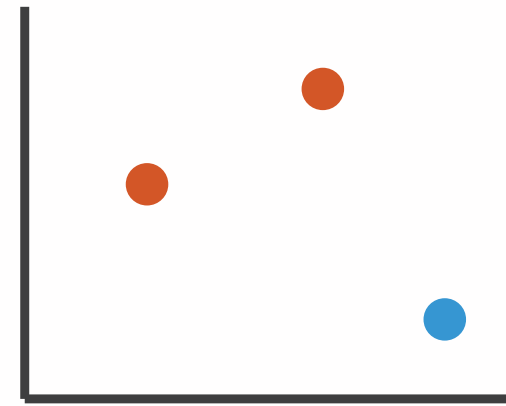
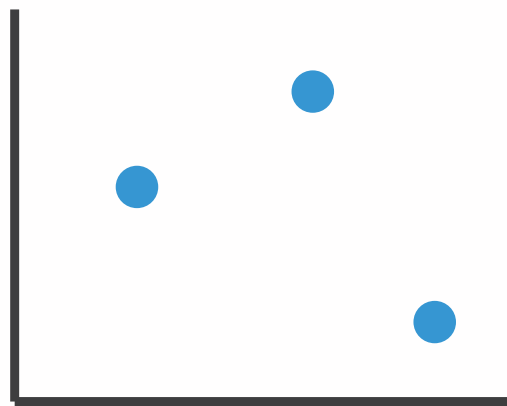
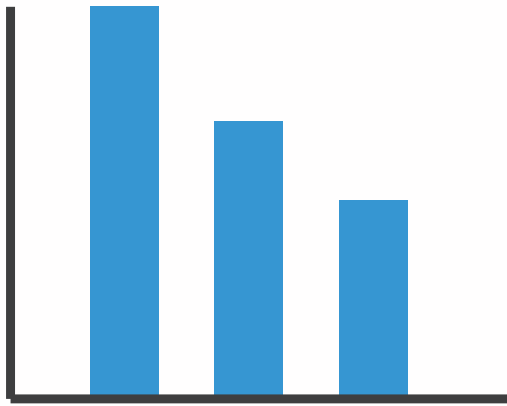
→ Motion

Direction, Rate, Frequency, ...



Encoding visually

- analyze idiom structure



Definitions: Marks and channels

- marks

– geometric primitives

→ Points



→ Lines



→ Areas



- channels

– control appearance of marks

→ Position

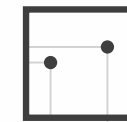
→ Horizontal



→ Vertical



→ Both



→ Color



→ Shape



→ Tilt



→ Size

→ Length



→ Area

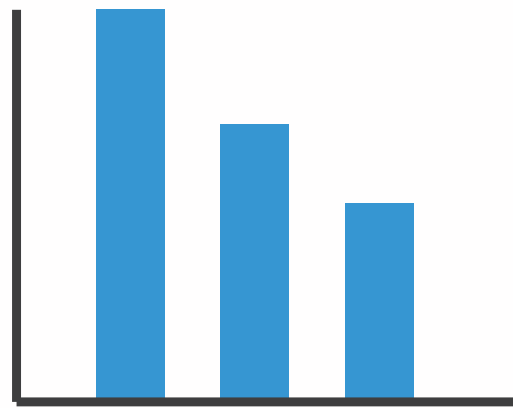


→ Volume



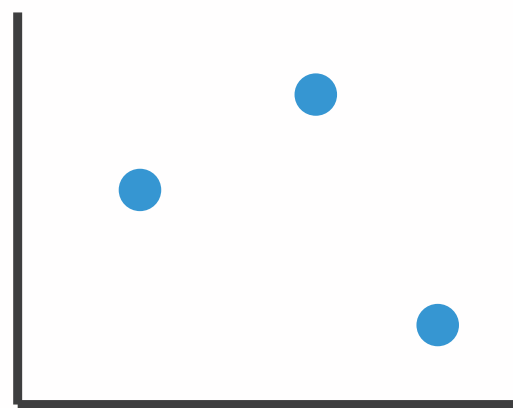
Encoding visually with marks and channels

- analyze idiom structure
 - as combination of marks and channels



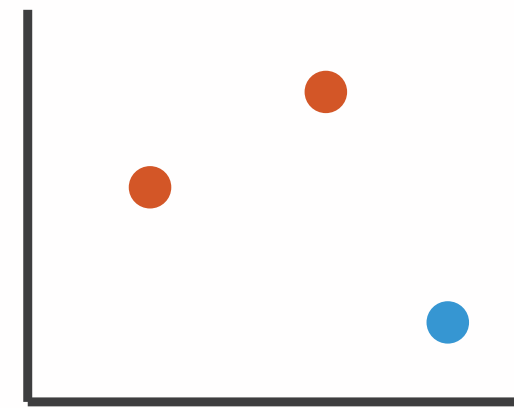
1:
vertical position

mark: line



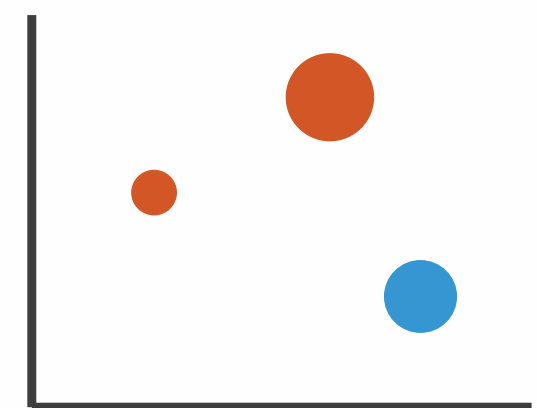
2:
vertical position
horizontal position

mark: point



3:
vertical position
horizontal position
color hue

mark: point



4:
vertical position
horizontal position
color hue
size (area)

mark: point

Channels

Position on common scale



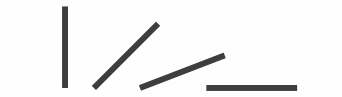
Position on unaligned scale



Length (1D size)



Tilt/angle



Area (2D size)



Depth (3D position)



Color luminance



Color saturation



Curvature



Volume (3D size)



Same

Spatial region



Color hue



Motion



Shape



Channels: Matching Types

➔ Magnitude Channels: Ordered Attributes

Position on common scale 

Position on unaligned scale 

Length (1D size) 

Tilt/angle 

Area (2D size) 

Depth (3D position) 

Color luminance 

Color saturation 

Curvature 

Volume (3D size) 

Same
Same

➔ Identity Channels: Categorical Attributes

Spatial region 

Color hue 

Motion 

Shape 

- **expressiveness principle**
 - match channel and data characteristics

Channels: Rankings

➔ Magnitude Channels: Ordered Attributes



➔ Identity Channels: Categorical Attributes



Best

Effectiveness

Least

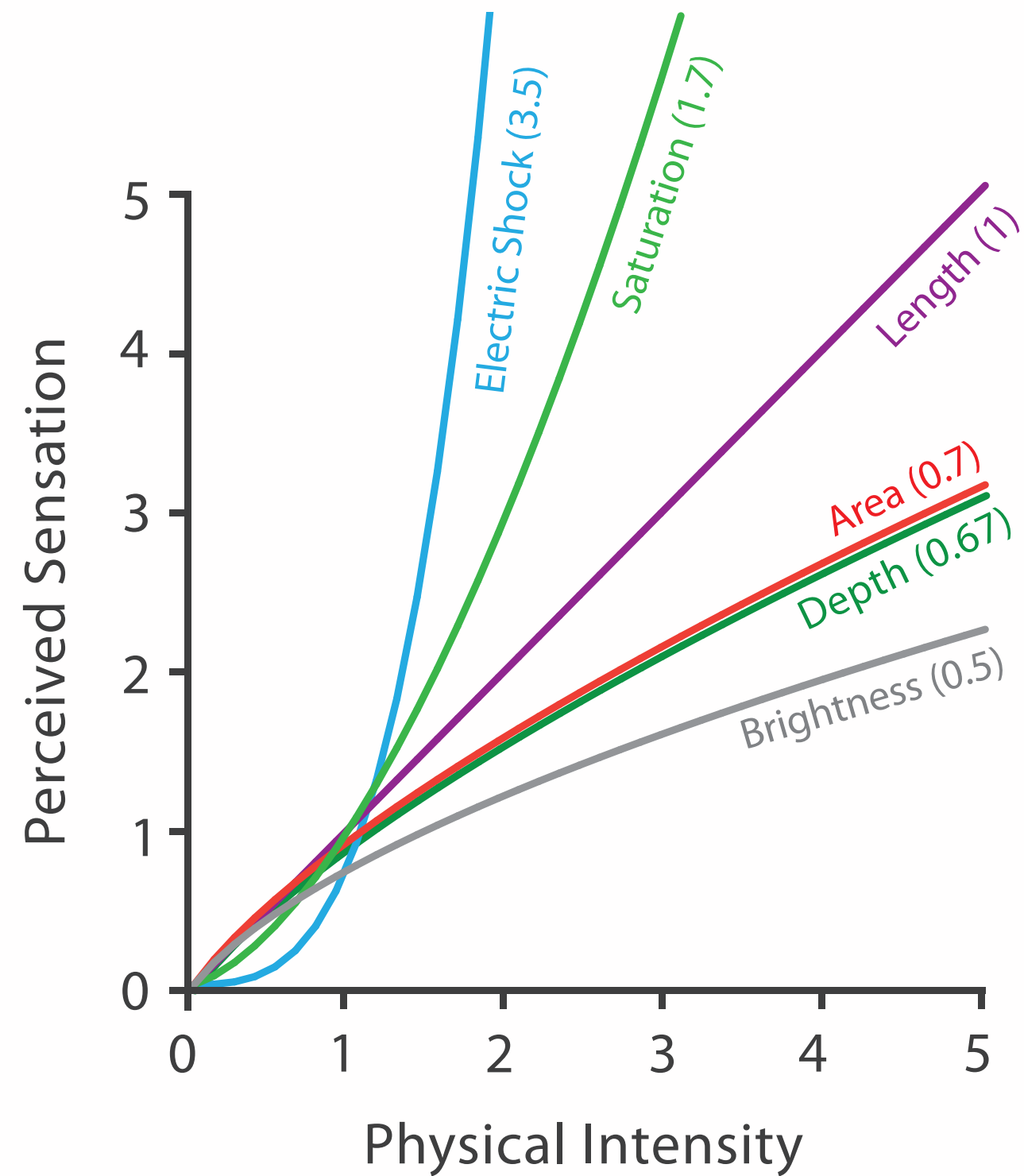
Same

Same

- **expressiveness principle**
 - match channel and data characteristics
- **effectiveness principle**
 - encode most important attributes with highest ranked channels

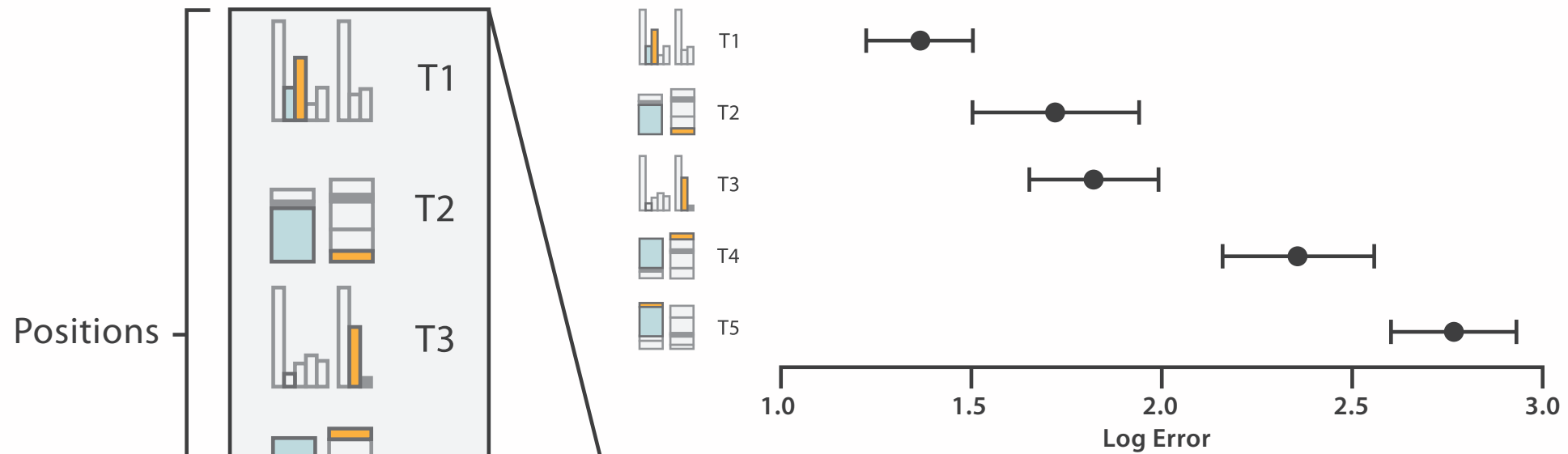
Accuracy: Fundamental Theory

Steven's Psychophysical Power Law: $S = I^N$

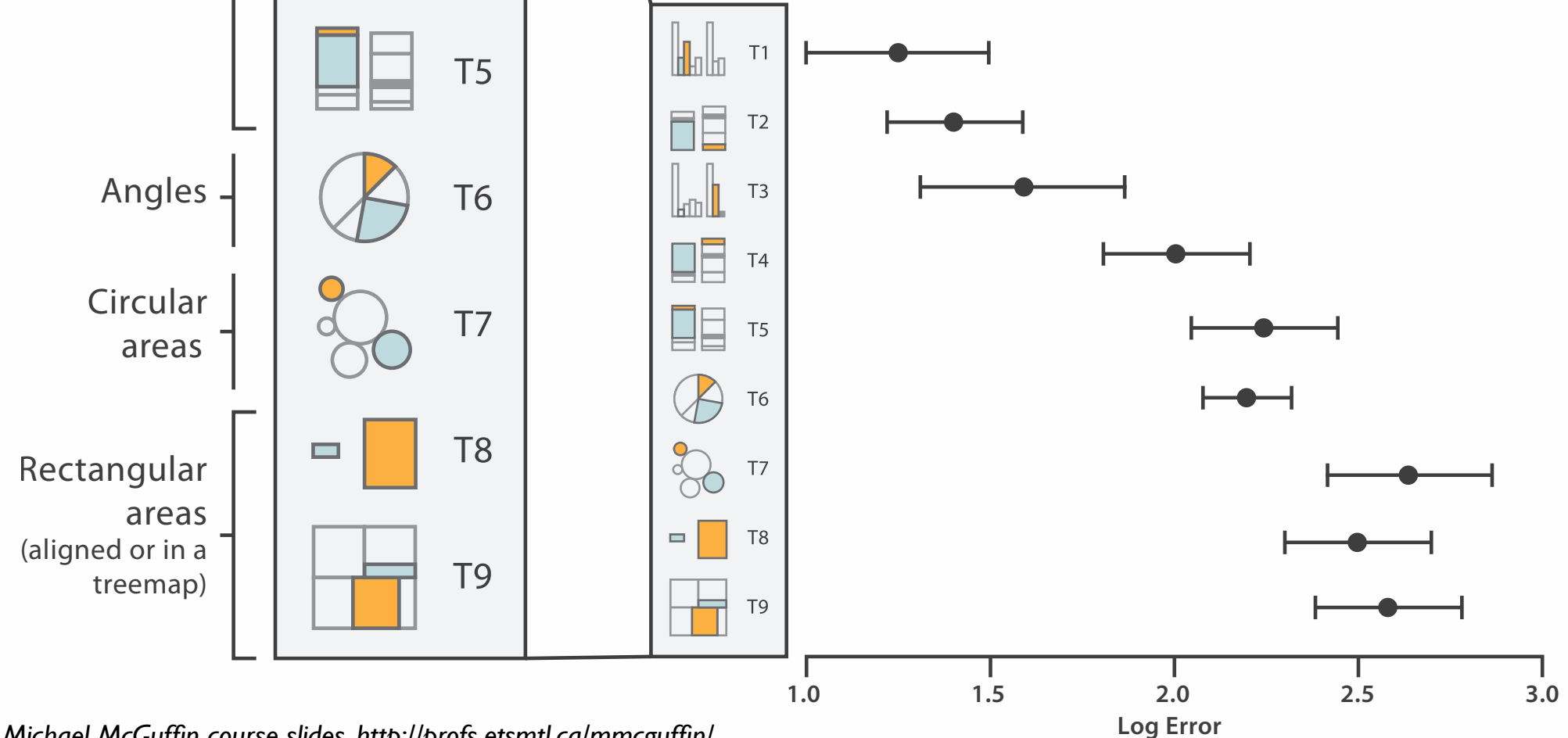


Accuracy: Vis experiments

Cleveland & McGill's Results



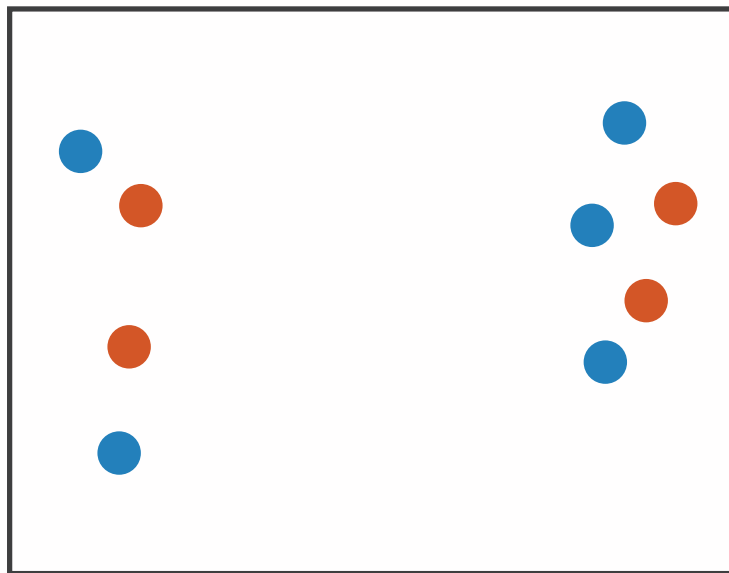
Crowdsourced Results



[Crowdsourcing Graphical Perception: Using Mechanical Turk to Assess Visualization Design. Heer and Bostock. Proc ACM Conf. Human Factors in Computing Systems (CHI) 2010, p. 203–212.]

Separability vs. Integrality

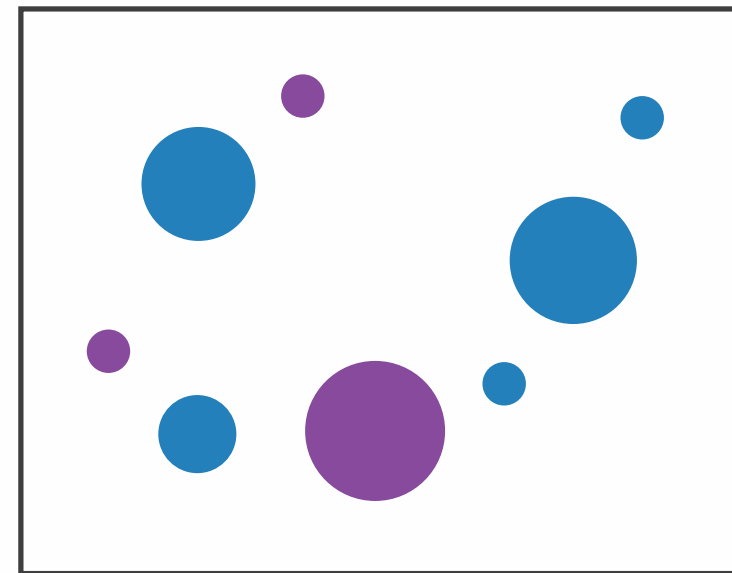
Position
+ Hue (Color)



Fully separable

2 groups each

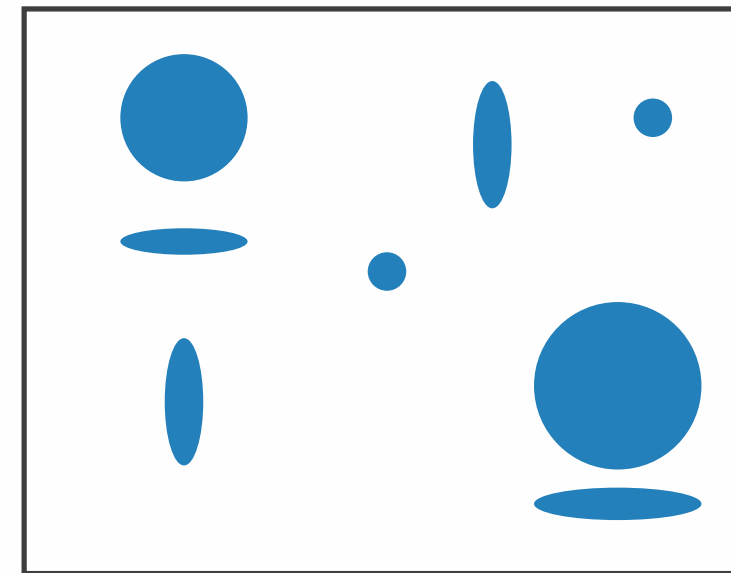
Size
+ Hue (Color)



Some interference

2 groups each

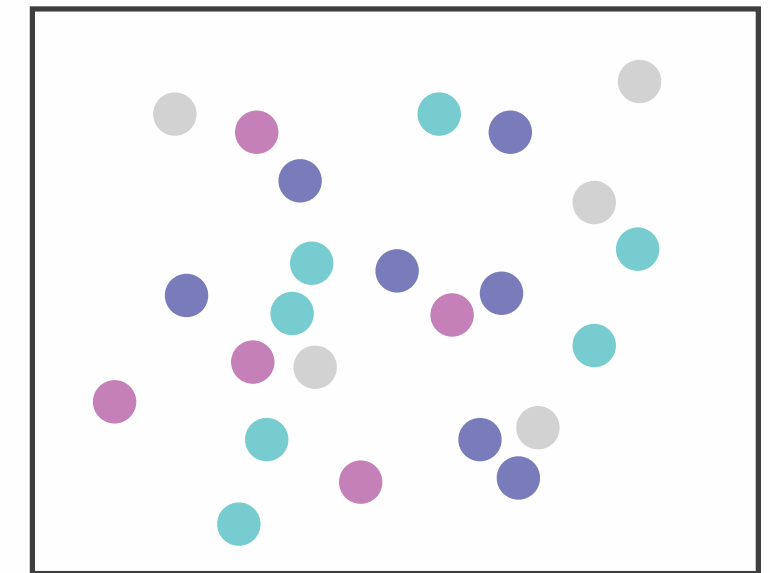
Width
+ Height



Some/significant
interference

3 groups total:
integral area

Red
+ Green



Major interference

4 groups total:
integral hue

Grouping

- containment
- connection

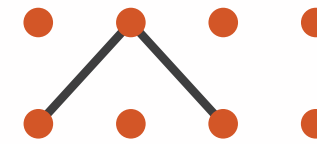
- proximity
 - same spatial region
- similarity
 - same values as other categorical channels

Marks as Links

➔ Containment



➔ Connection



➔ Identity Channels: Categorical Attributes

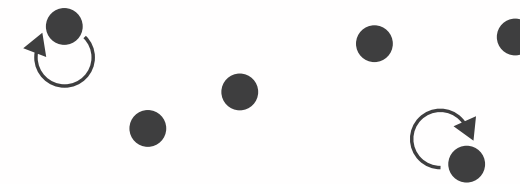
Spatial region



Color hue



Motion



Shape



How to encode: Arrange position and region

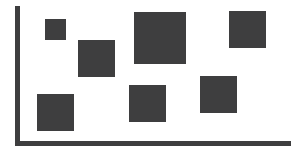
Encode

→ Arrange

→ Express



→ Separate



→ Order



→ Align



→ Use



→ Map

from **categorical** and **ordered** attributes

→ Color

→ Hue



→ Saturation



→ Luminance



→ Size, Angle, Curvature, ...

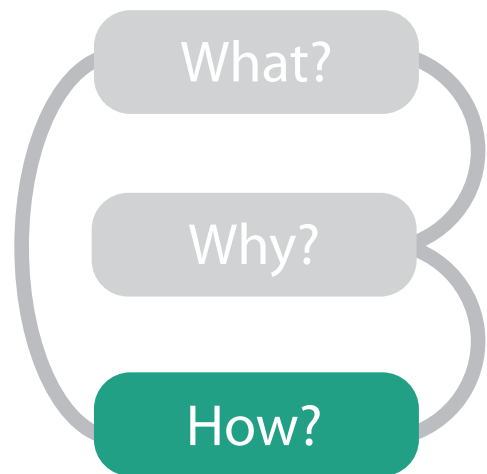


→ Shape



→ Motion

Direction, Rate, Frequency, ...



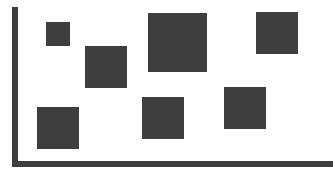
Arrange tables

② Express Values



② Separate, Order, Align Regions

→ Separate



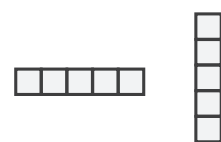
→ Order



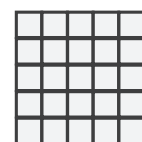
→ Align



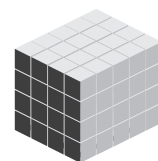
→ 1 Key
List



→ 2 Keys
Matrix



→ 3 Keys
Volume

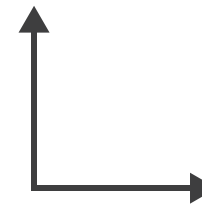


→ Many Keys
Recursive Subdivision

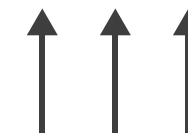


② Axis Orientation

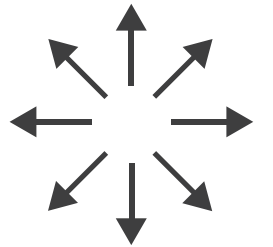
→ Rectilinear



→ Parallel

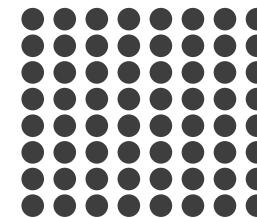


→ Radial



② Layout Density

→ Dense



→ Space-Filling



Idioms: dot chart, line chart

- one key, one value

- data

- 2 quant attribs

- mark: points

- dot plot: + line connection marks between them

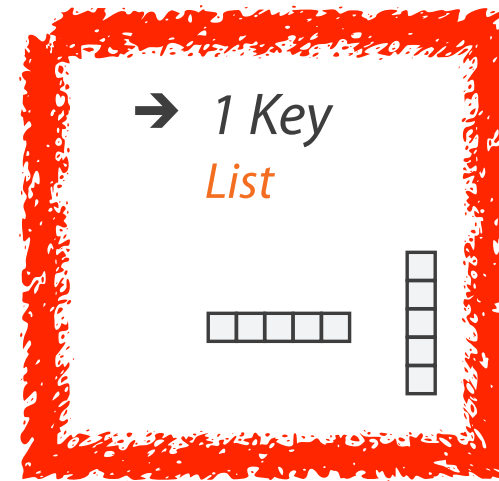
- channels

- aligned lengths to express quant value
- separated and ordered by key attrib into horizontal regions

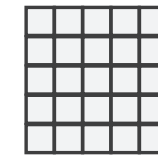
- task

- find trend

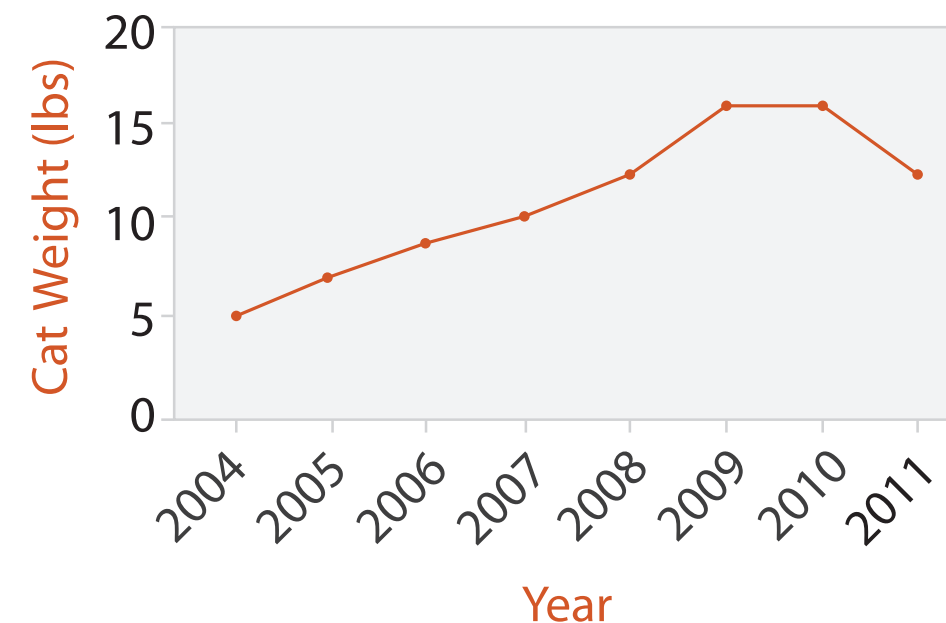
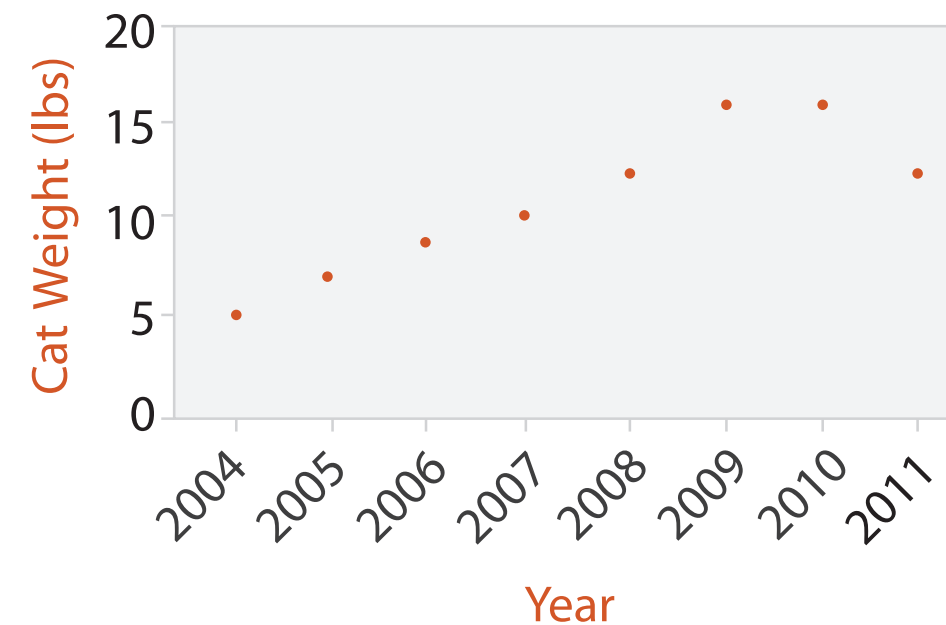
- connection marks emphasize ordering of items along key axis by explicitly showing relationship between one item and the next



→ 2 Keys
Matrix

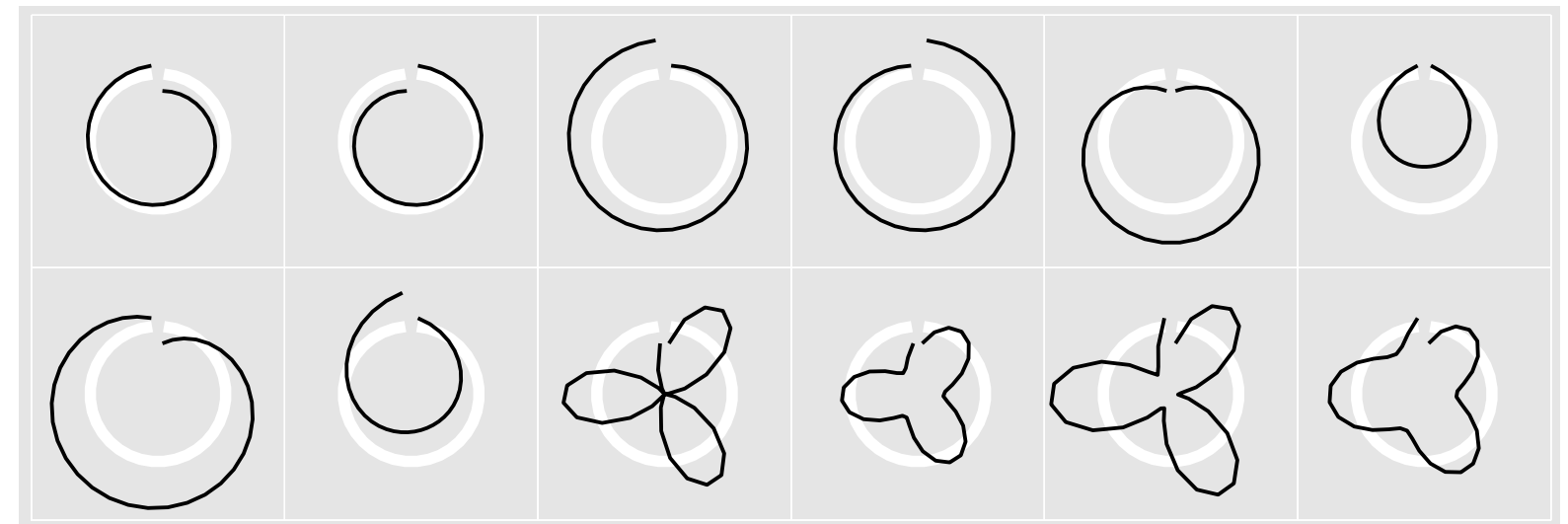
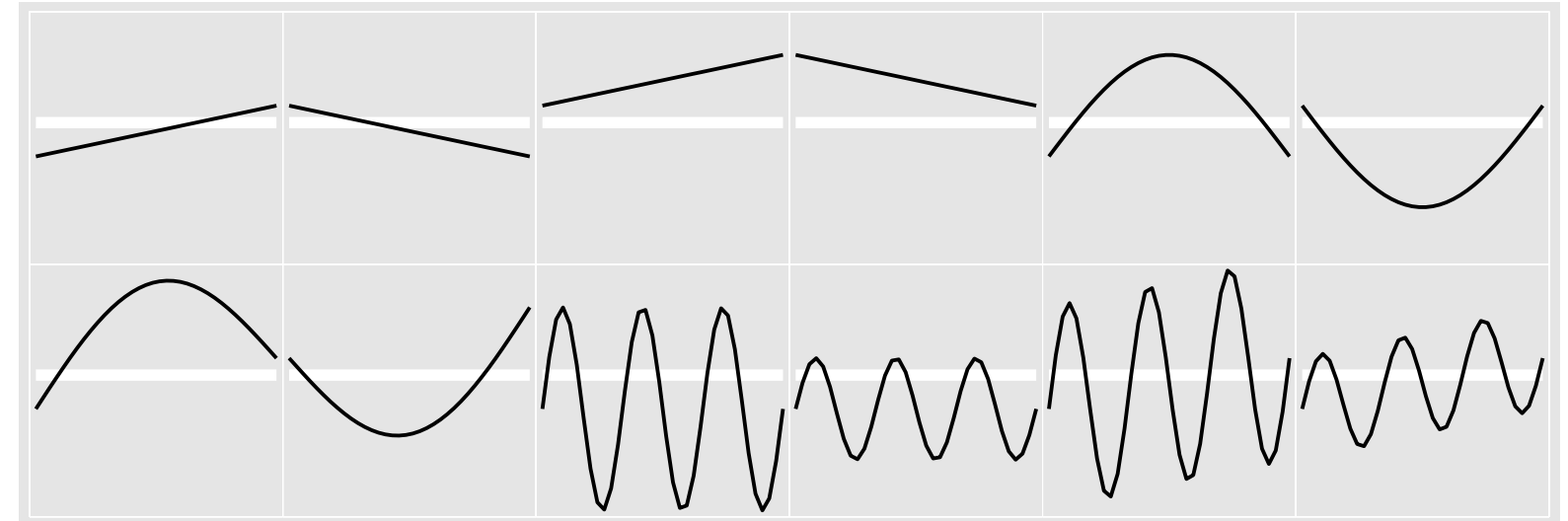


→ Many Keys
Recursive Subdivision

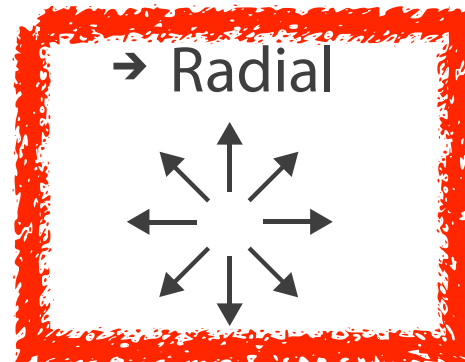
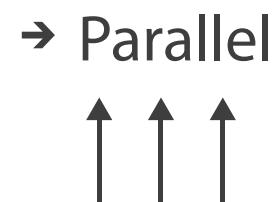
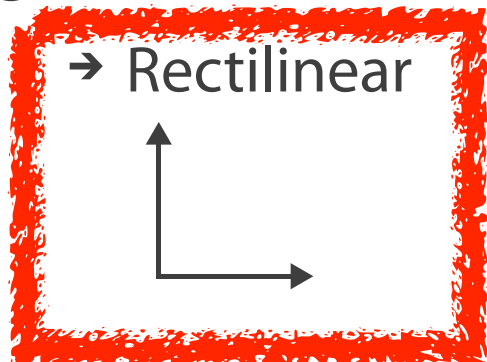


Idiom: glyphmaps

- rectilinear good for linear vs nonlinear trends
- radial good for cyclic patterns



➔ Axis Orientation



[Glyph-maps for Visually Exploring Temporal Patterns in Climate Data and Models. Wickham, Hofmann, Wickham, and Cook. *Environmetrics* 23:5 (2012), 382–393.]

Idiom: heatmap

- two keys, one value

- data

- 2 categ attribs (gene, experimental condition)
- 1 quant attrib (expression levels)

- marks: area

- separate and align in 2D matrix
 - indexed by 2 categorical attributes

- channels

- color by quant attrib
 - (ordered diverging colormap)

- task

- find clusters, outliers

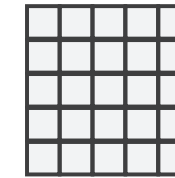
- scalability

- 1M items, 100s of categ levels, ~10 quant attrib levels

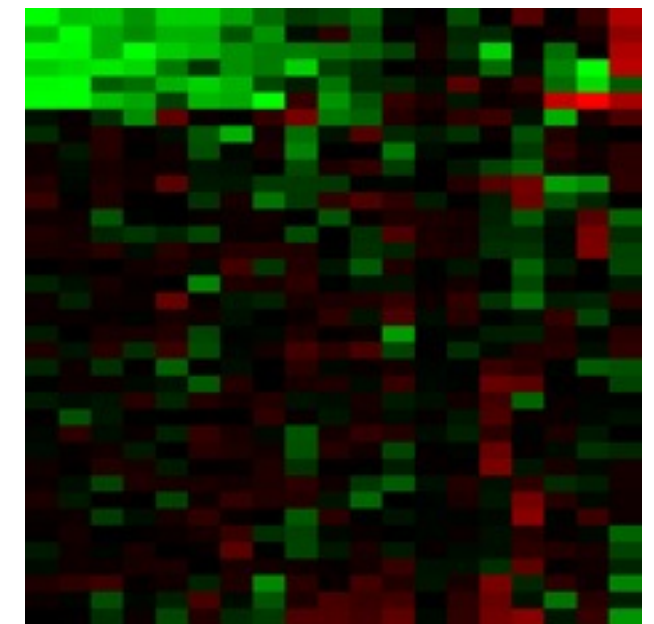
→ 1 Key
List



→ 2 Keys
Matrix

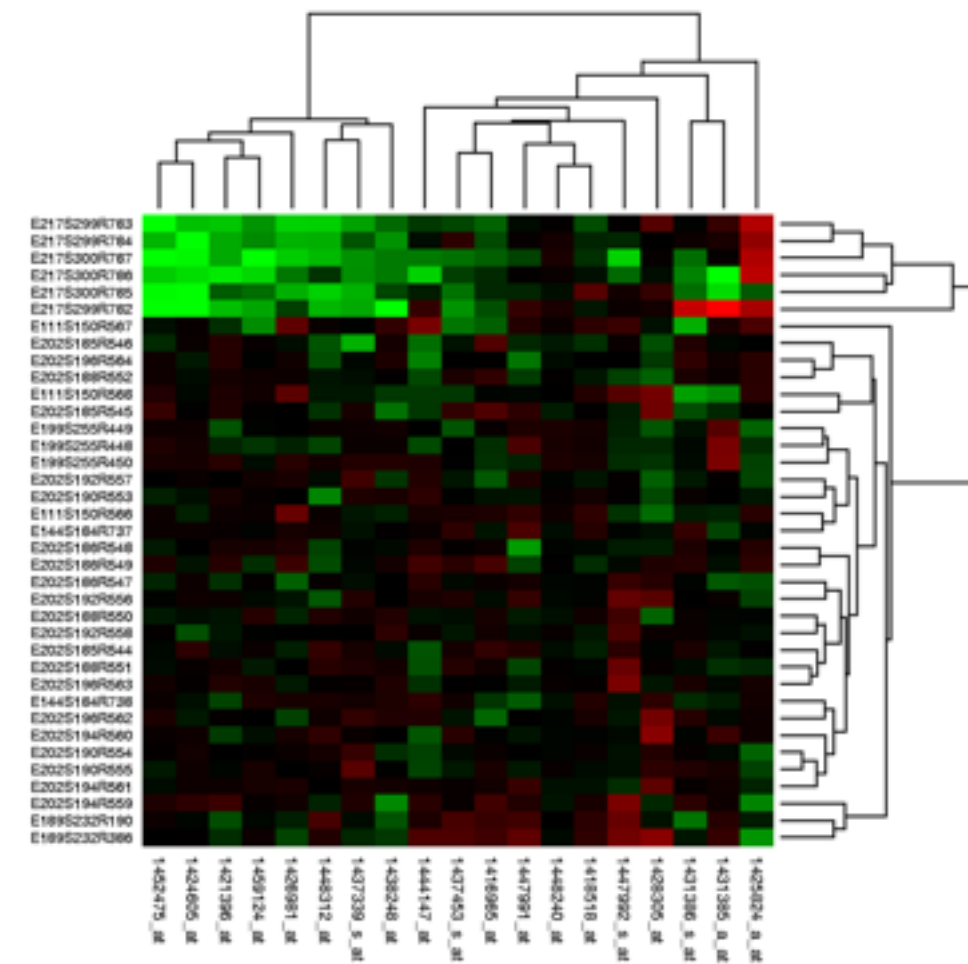


→ Many Keys
Recursive Subdivision



Idiom: cluster heatmap

- in addition
 - derived data
 - 2 cluster hierarchies
 - dendrogram
 - parent-child relationships in tree with connection line marks
 - leaves aligned so interior branch heights easy to compare
 - heatmap
 - marks (re-)ordered by cluster hierarchy traversal



Arrange spatial data

→ Use Given

→ Geometry

→ *Geographic*

→ *Other Derived*

→ Spatial Fields

→ *Scalar Fields (one value per cell)*

→ *Isocontours*

→ *Direct Volume Rendering*

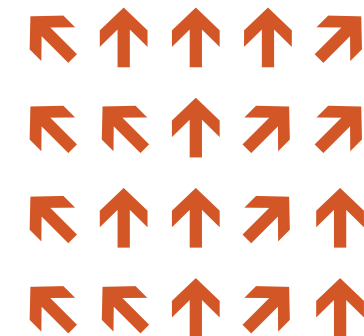
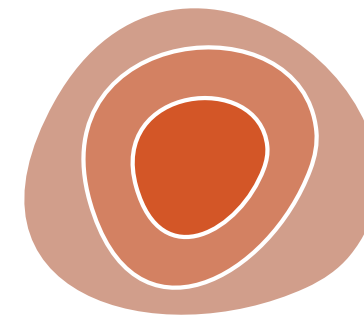
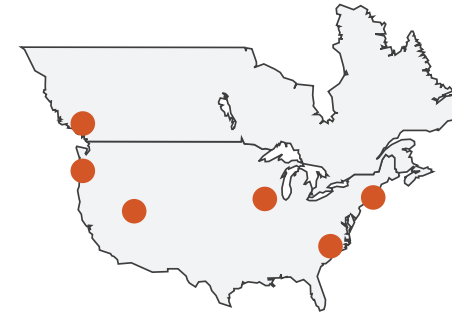
→ *Vector and Tensor Fields (many values per cell)*

→ *Flow Glyphs (local)*

→ *Geometric (sparse seeds)*

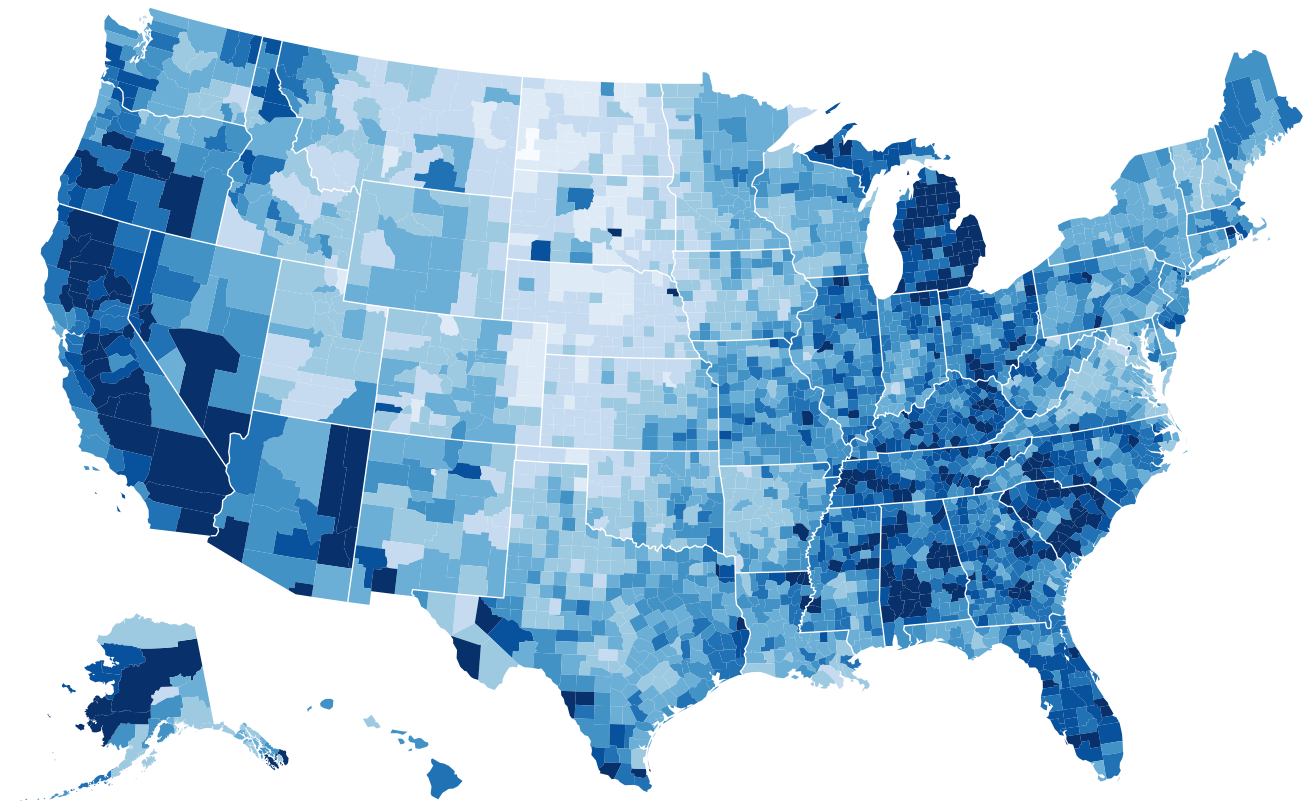
→ *Textures (dense seeds)*

→ *Features (globally derived)*



Idiom: **choropleth map**

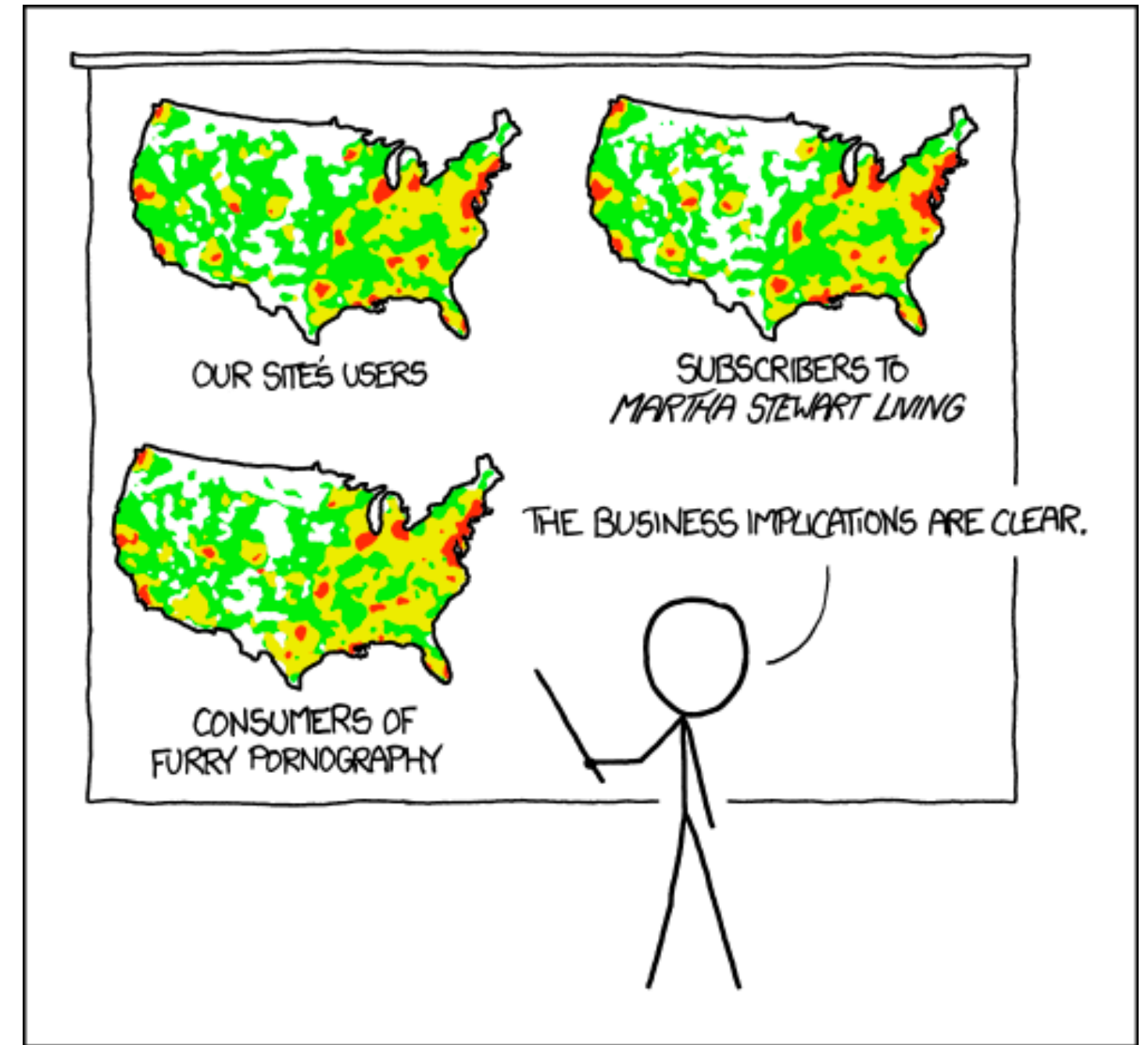
- **use** given spatial data
 - when central task is understanding spatial relationships
- **data**
 - geographic geometry
 - table with 1 quant attribute per region
- **encoding**
 - use given geometry for area mark boundaries
 - sequential segmented colormap



<http://bl.ocks.org/mbostock/4060606>

Population maps trickiness

- beware!

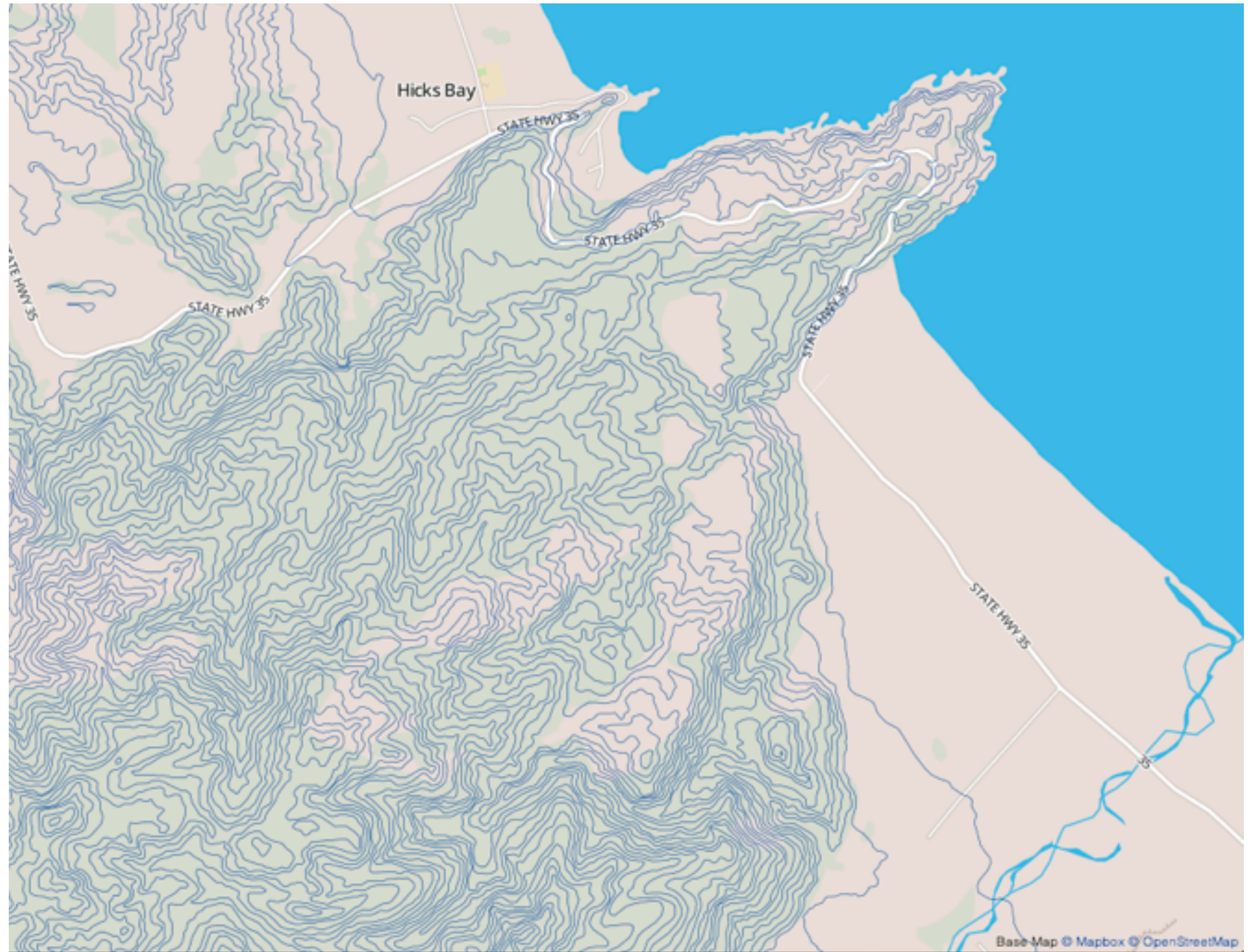


PET PEEVE #208:
GEOGRAPHIC PROFILE MAPS WHICH ARE
BASICALLY JUST POPULATION MAPS

[<https://xkcd.com/1138>]

Idiom: **topographic map**

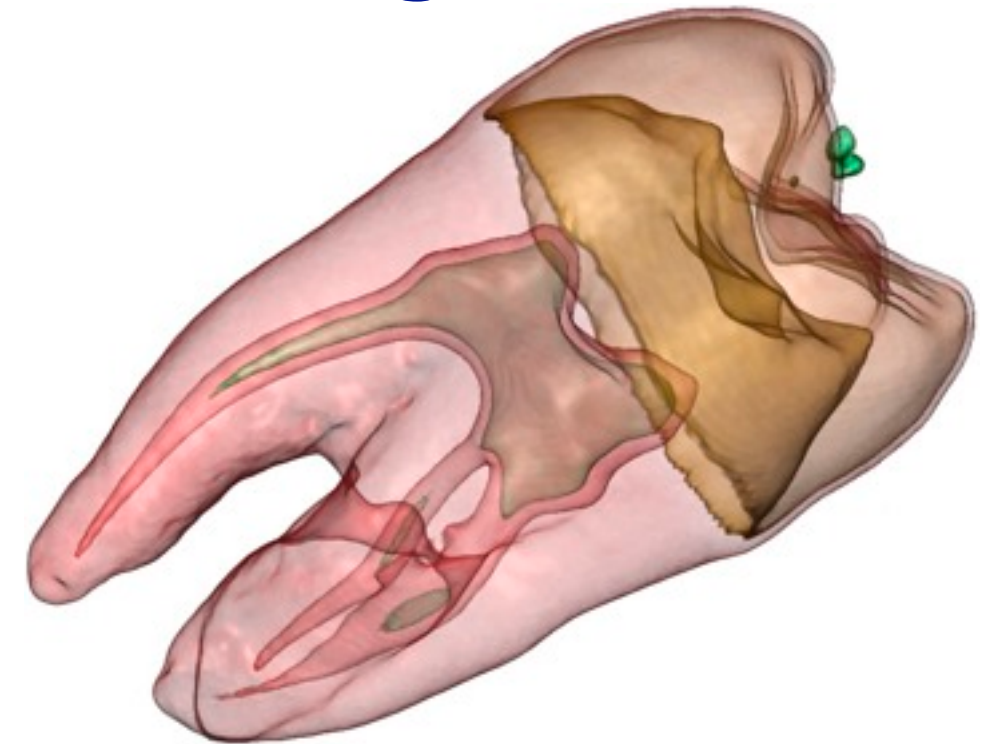
- data
 - geographic geometry
 - scalar spatial field
 - 1 quant attribute per grid cell
- derived data
 - isoline geometry
 - isocontours computed for specific levels of scalar values



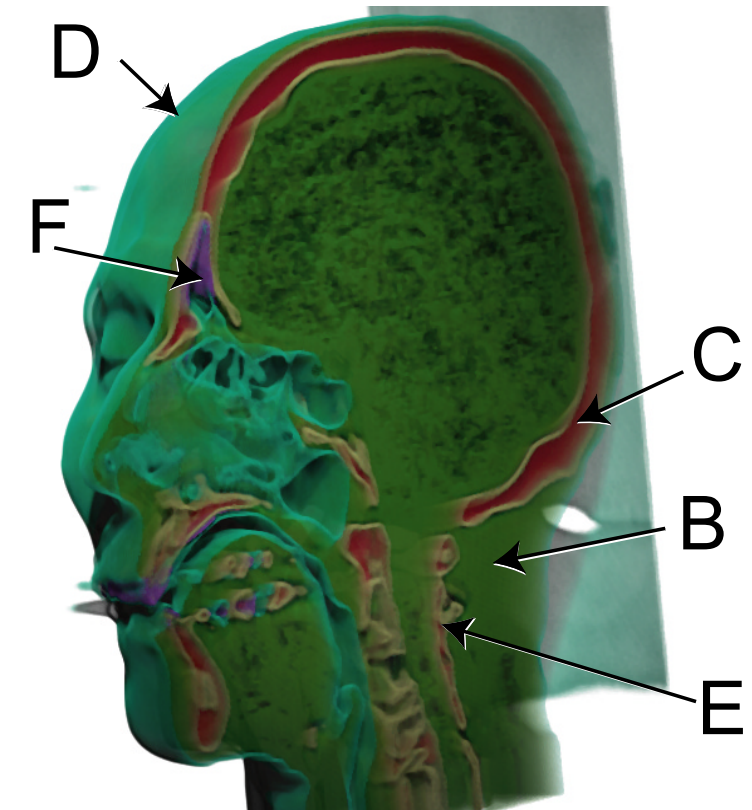
Land Information New Zealand Data Service

Idioms: **isosurfaces**, **direct volume rendering**

- **data**
 - scalar spatial field
 - 1 quant attribute per grid cell
- **task**
 - shape understanding, spatial relationships
- **isosurface**
 - derived data: isocontours computed for specific levels of scalar values
- **direct volume rendering**
 - transfer function maps scalar values to color, opacity
 - no derived geometry



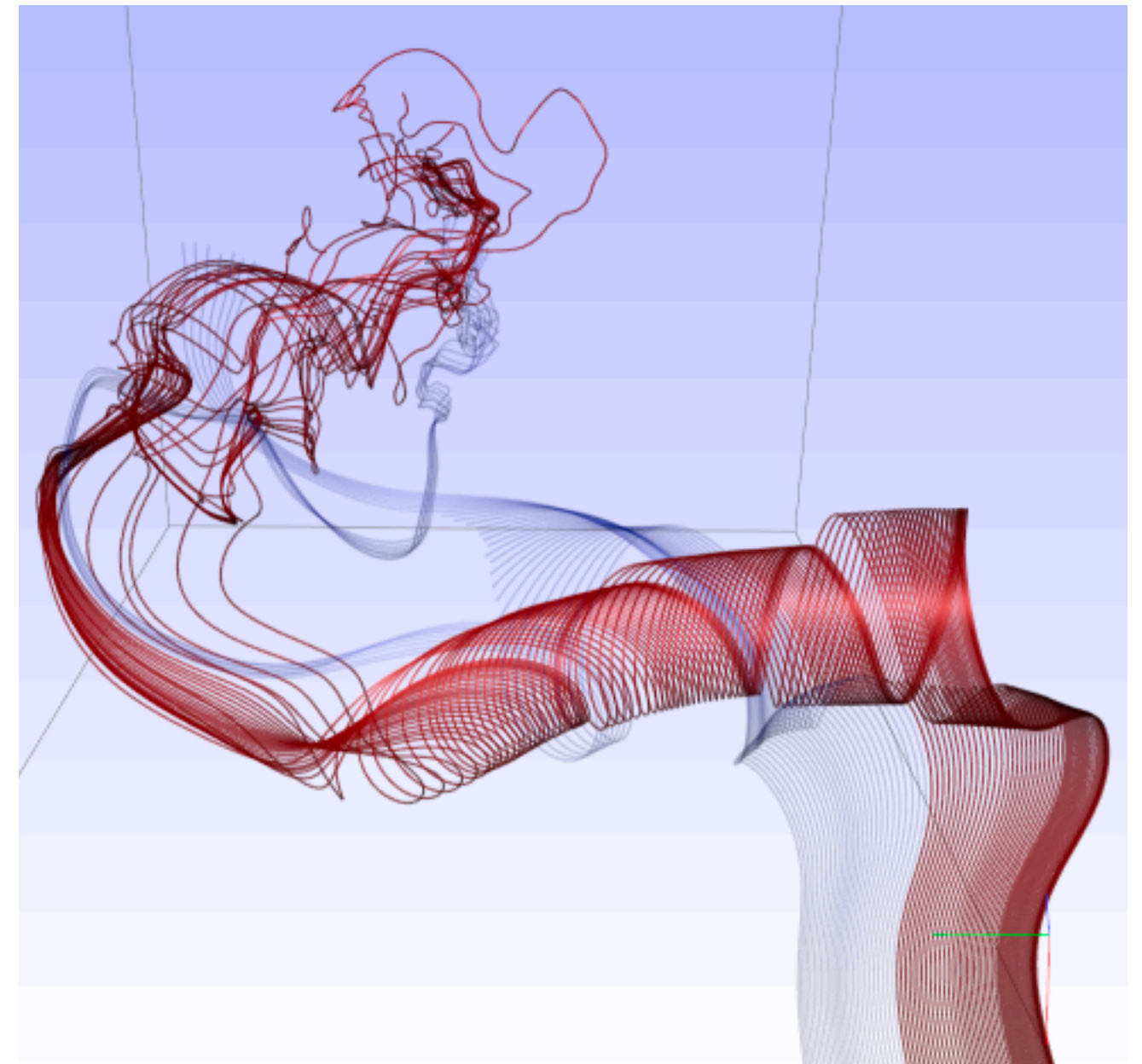
[Interactive Volume Rendering Techniques. Kniss. Master's thesis, University of Utah Computer Science, 2002.]



[Multidimensional Transfer Functions for Volume Rendering. Kniss, Kindlmann, and Hansen. In The Visualization Handbook, edited by Charles Hansen and Christopher Johnson, pp. 189–210. Elsevier, 2005.]

Idiom: **similarity-clustered streamlines**

- data
 - 3D vector field
- derived data (from field)
 - streamlines: trajectory particle will follow
- derived data (per streamline)
 - curvature, torsion, tortuosity
 - signature: complex weighted combination
 - compute cluster hierarchy across all signatures
 - encode: color and opacity by cluster
- tasks
 - find features, query shape
- scalability
 - millions of samples, hundreds of streamlines



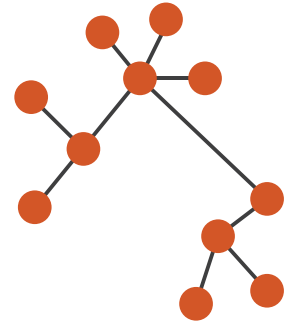
[Similarity Measures for Enhancing Interactive Streamline Seeding. McLoughlin, Jones, Laramee, Malki, Masters, and Hansen. IEEE Trans. Visualization and Computer Graphics 19:8 (2013), 1342–1353.]

Arrange networks and trees

→ Node–Link Diagrams Connection Marks

✓ NETWORKS

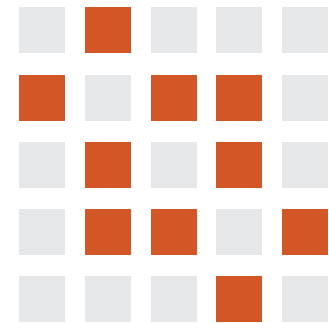
✓ TREES



→ Adjacency Matrix Derived Table

✓ NETWORKS

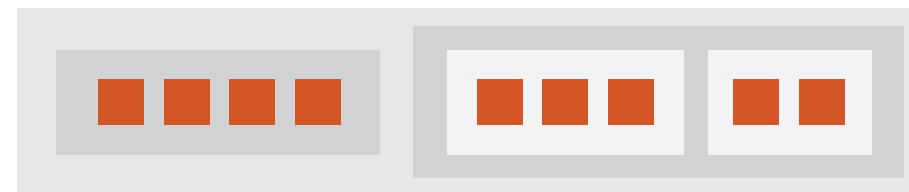
✓ TREES



→ Enclosure Containment Marks

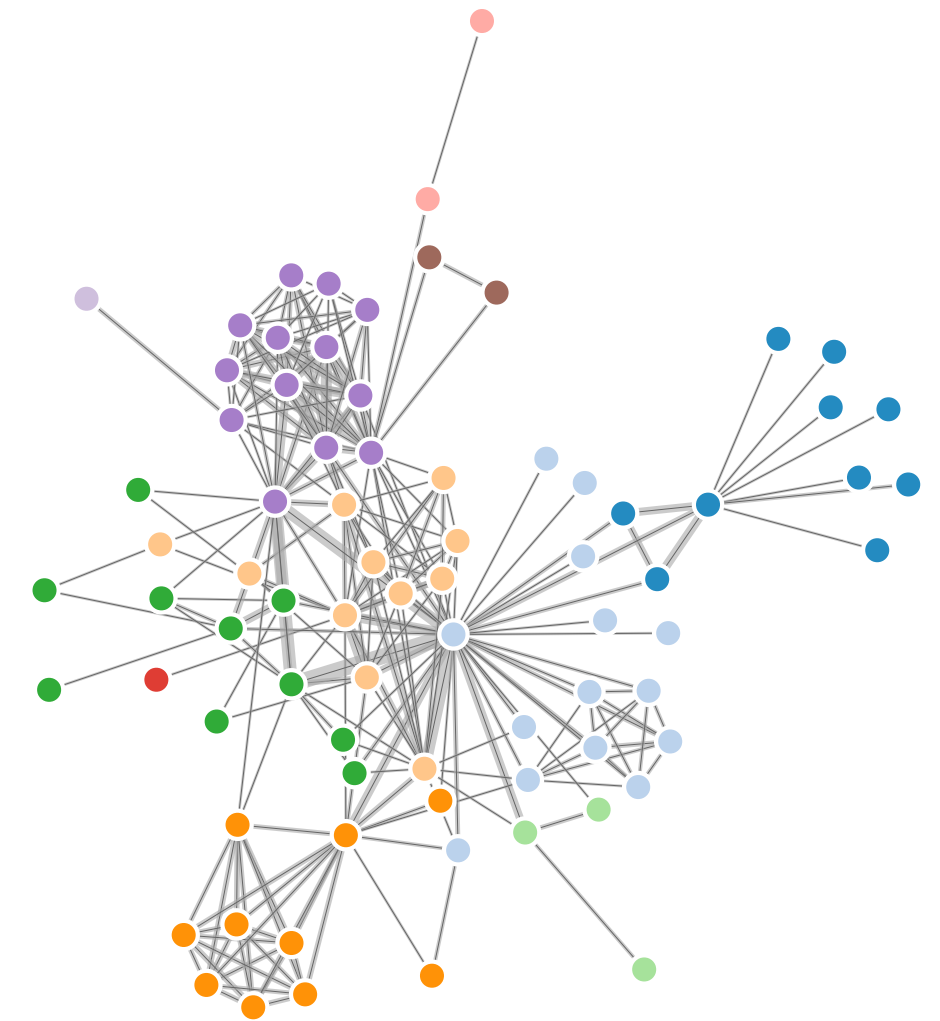
✗ NETWORKS

✓ TREES



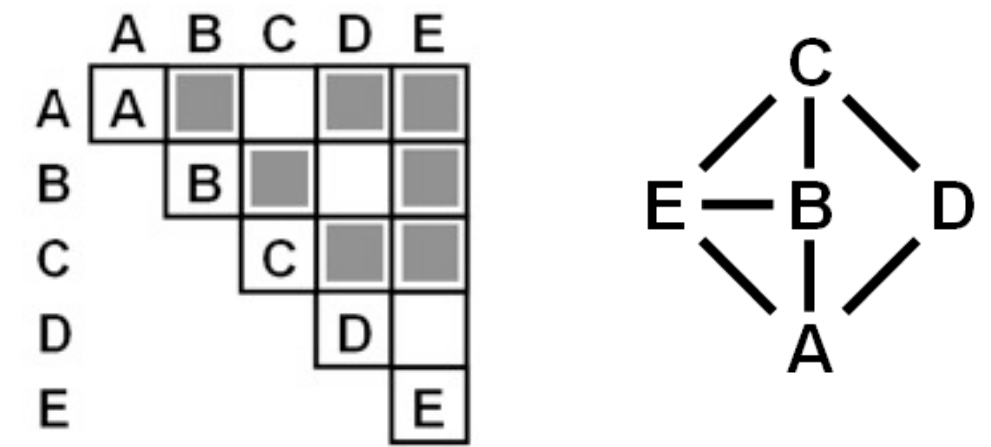
Idiom: **force-directed placement**

- visual encoding
 - link connection marks, node point marks
- considerations
 - spatial position: no meaning directly encoded
 - left free to minimize crossings
 - proximity semantics?
 - sometimes meaningful
 - sometimes arbitrary, artifact of layout algorithm
 - tension with length
 - long edges more visually salient than short
- tasks
 - explore topology; locate paths, clusters
- scalability
 - node/edge density $E < 4N$



Idiom: adjacency matrix view

- data: network
 - transform into same data/encoding as heatmap
- derived data: table from network
 - 1 quant attrib
 - weighted edge between nodes
 - 2 categ attribs: node list x 2
- visual encoding
 - cell shows presence/absence of edge
- scalability
 - 1K nodes, 1M edges



[NodeTrix: a Hybrid Visualization of Social Networks. Henry, Fekete, and McGuffin. IEEE TVCG (Proc. InfoVis) 13(6):1302-1309, 2007.]

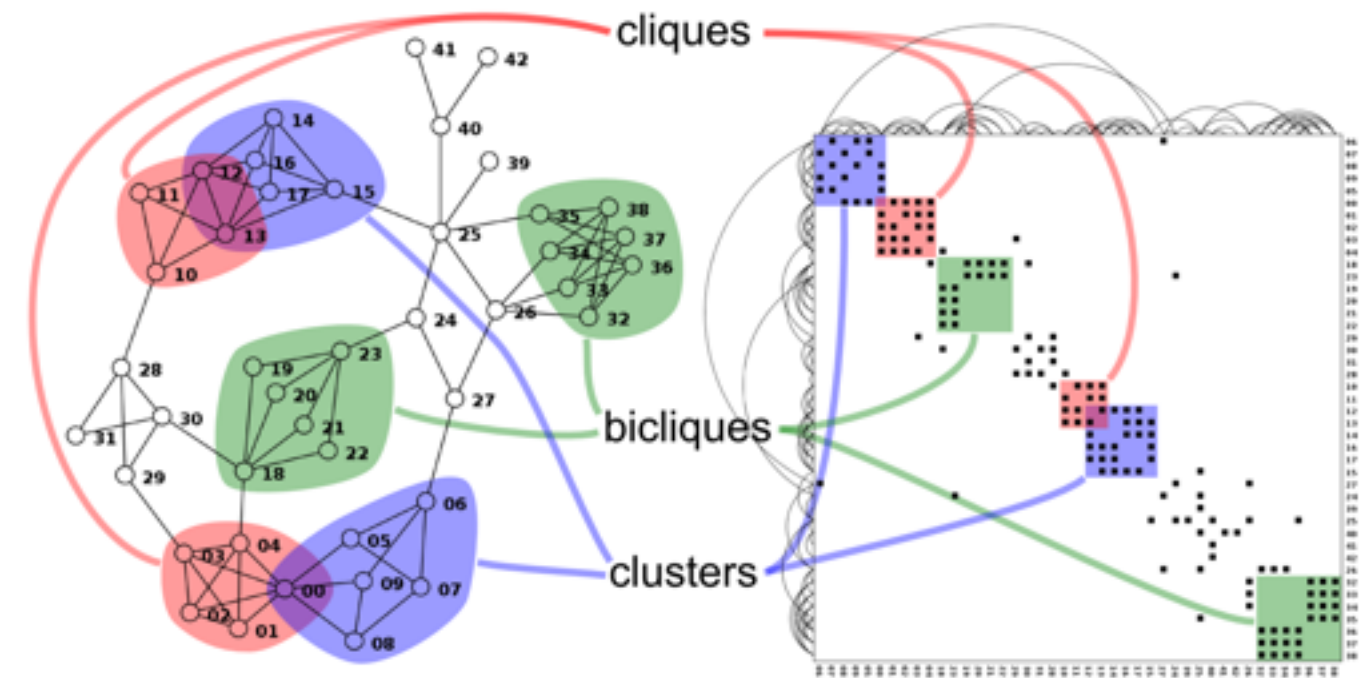


[Points of view: Networks. Gehlenborg and Wong. Nature Methods 9:115.]

Connection vs. adjacency comparison

- adjacency matrix strengths
 - predictability, scalability, supports reordering
 - some topology tasks trainable
- node-link diagram strengths
 - topology understanding, path tracing
 - intuitive, no training needed
- empirical study
 - node-link best for small networks
 - matrix best for large networks
 - if tasks don't involve topological structure!

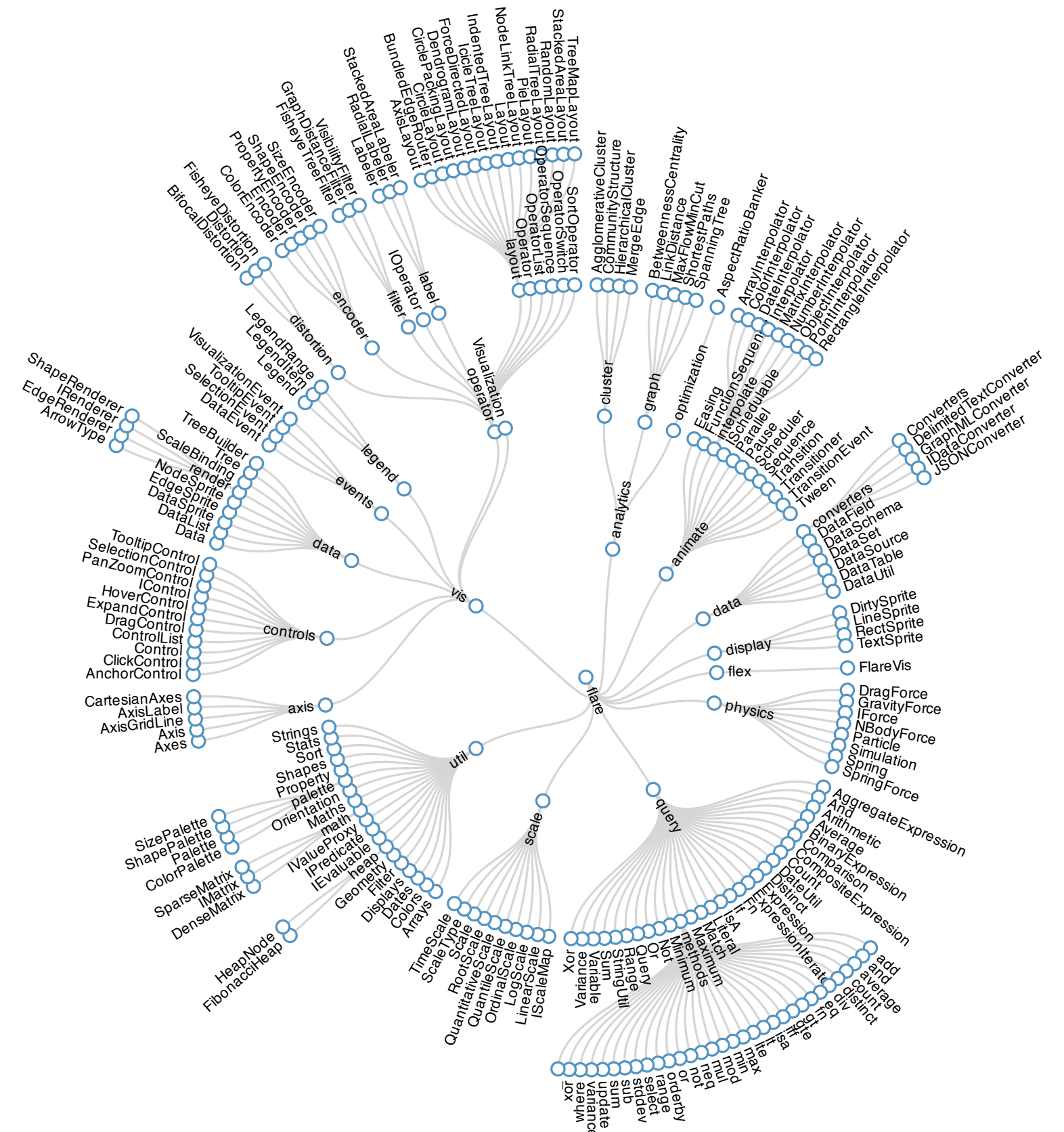
[On the readability of graphs using node-link and matrix-based representations: a controlled experiment and statistical analysis. Ghoniem, Fekete, and Castagliola. Information Visualization 4:2 (2005), 114–135.]



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

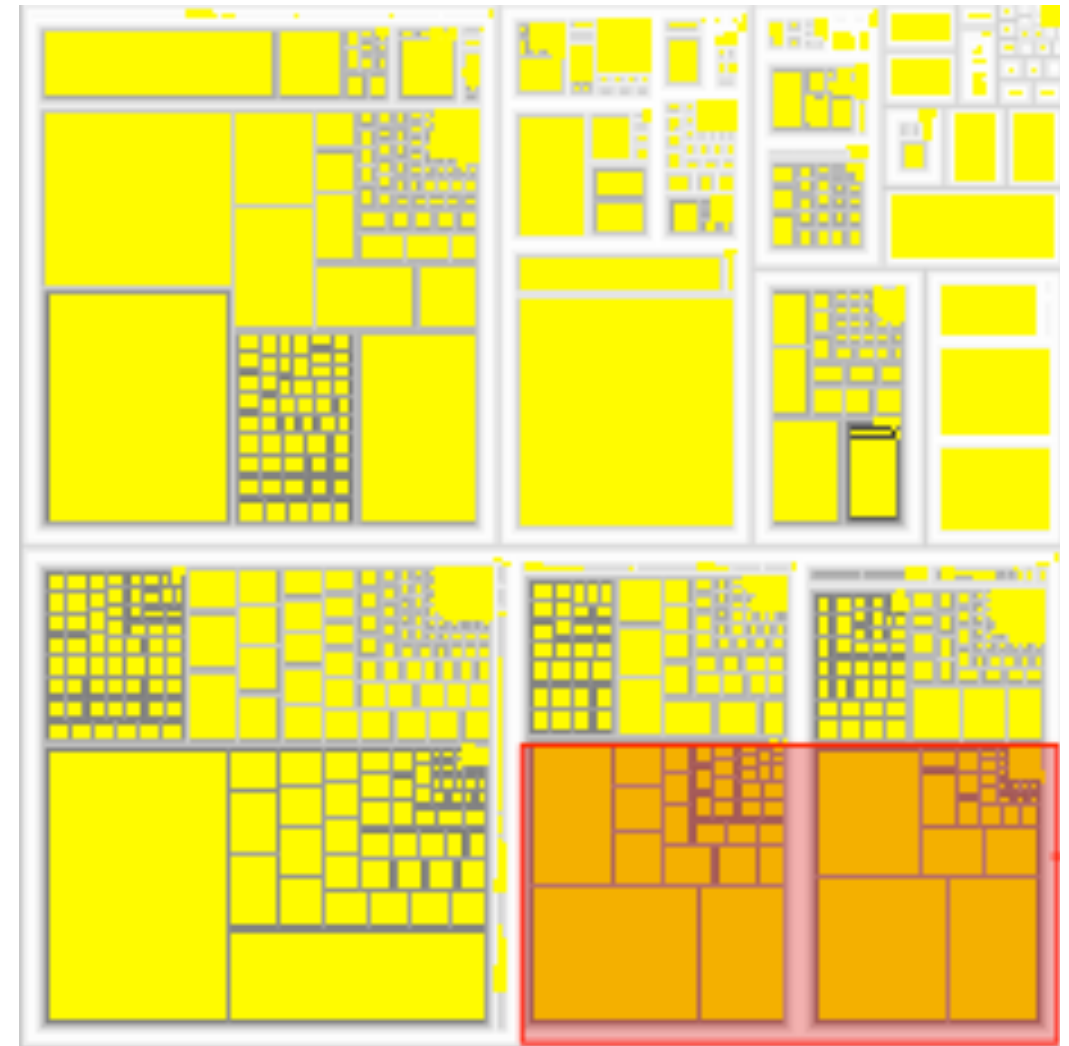
Idiom: radial node-link tree

- data
 - tree
- encoding
 - link connection marks
 - point node marks
 - radial axis orientation
 - angular proximity: siblings
 - distance from center: depth in tree
- tasks
 - understanding topology, following paths
- scalability
 - 1K - 10K nodes



Idiom: **treemap**

- **data**
 - tree
 - 1 quant attrib at leaf nodes
- **encoding**
 - area containment marks for hierarchical structure
 - rectilinear orientation
 - size encodes quant attrib
- **tasks**
 - query attribute at leaf nodes
- **scalability**
 - 1M leaf nodes

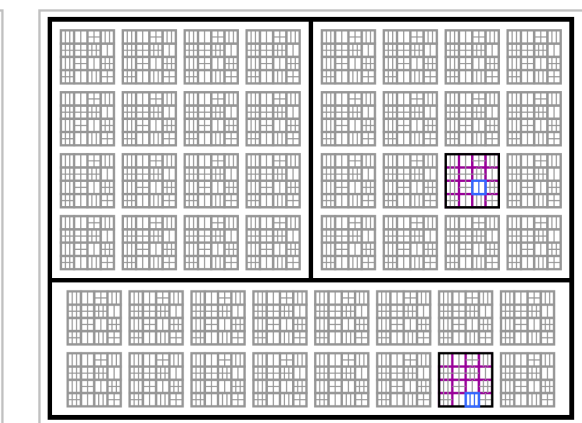
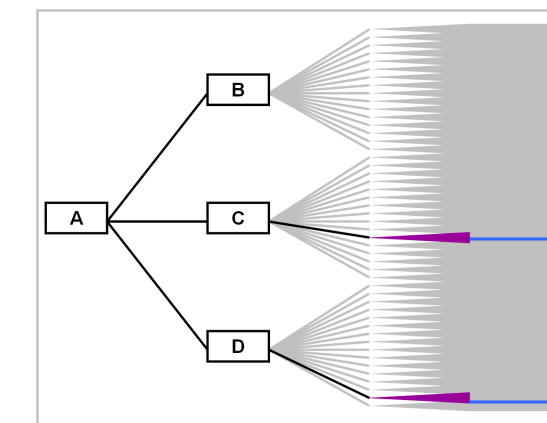
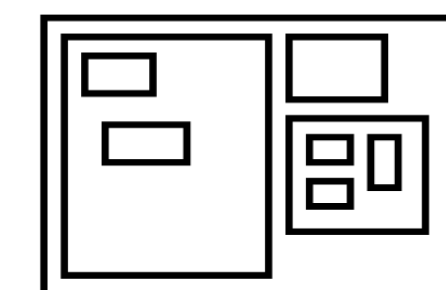
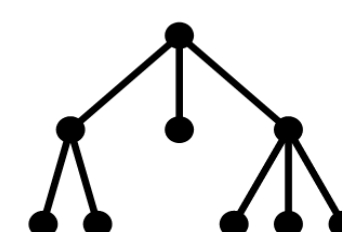
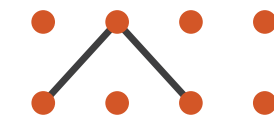
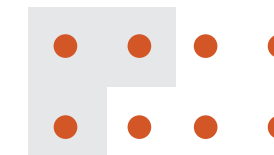


http://tulip.labri.fr/Documentation/3_7/userHandbook/html/ch06.html

Connection vs. containment comparison

- marks as links (vs. nodes)
 - common case in network drawing
 - 1D case: connection
 - ex: all node-link diagrams
 - emphasizes topology, path tracing
 - networks and trees
 - 2D case: containment
 - ex: all treemap variants
 - emphasizes attribute values at leaves (size coding)
 - only trees

➔ Containment ➔ Connection



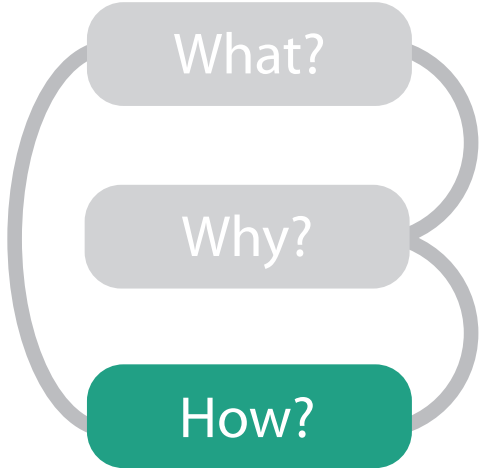
Node-Link Diagram

Treemap

[Elastic Hierarchies: Combining Treemaps and Node-Link Diagrams. Dong, McGuffin, and Chignell. Proc. InfoVis 2005, p. 57-64.]

How to encode: Mapping color

Encode



→ Arrange

→ Express



→ Separate



→ Order



→ Align



→ Use



→ Map

from **categorical** and **ordered** attributes

→ Color

→ Hue



→ Saturation



→ Luminance



→ Size, Angle, Curvature, ...



→ Shape



→ Motion

Direction, Rate, Frequency, ...

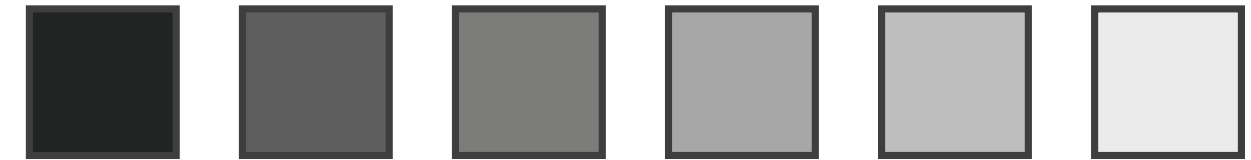


Color: Luminance, saturation, hue

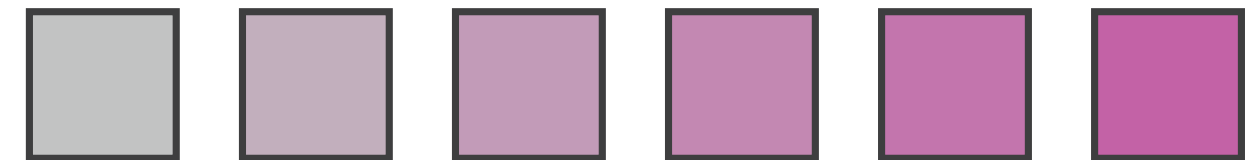
- 3 channels
 - identity for categorical
 - hue
 - magnitude for ordered
 - luminance
 - saturation
- RGB: poor for encoding
- HSL: better, but beware
 - lightness \neq luminance



Luminance



Saturation



Hue

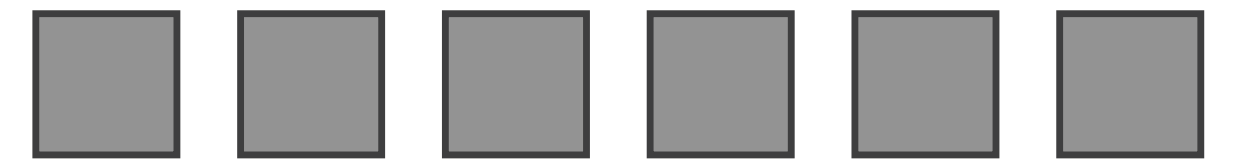


Corners of the RGB color cube

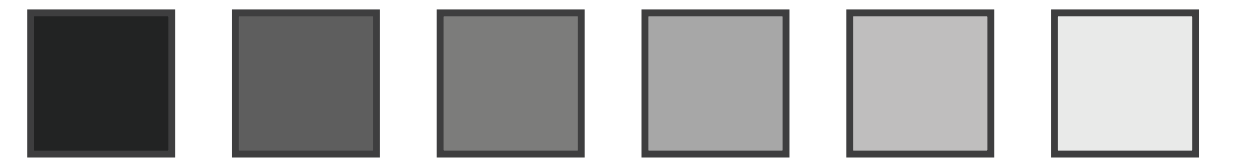


L from HLS

All the same

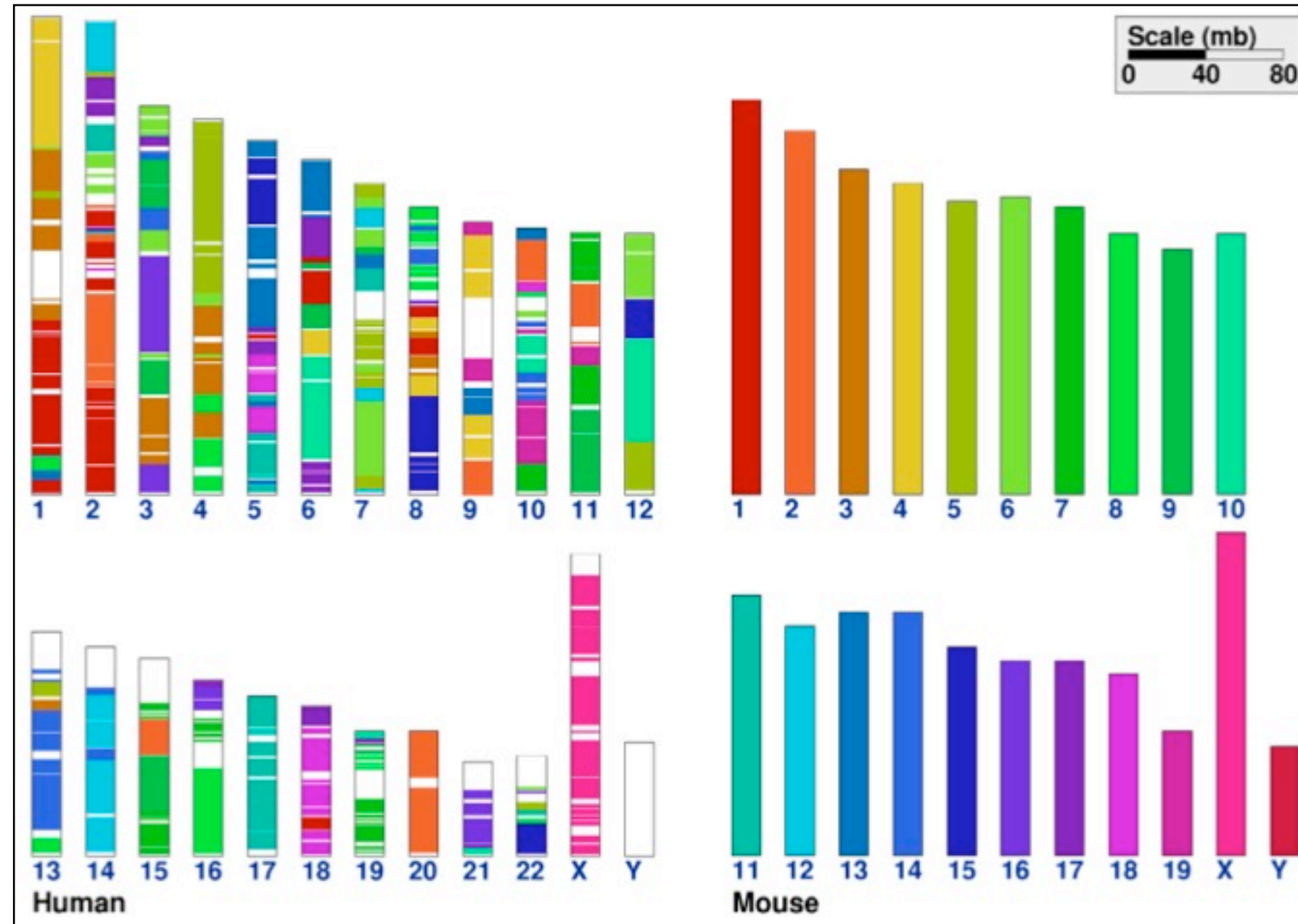


Luminance values



Categorical color: Discriminability constraints

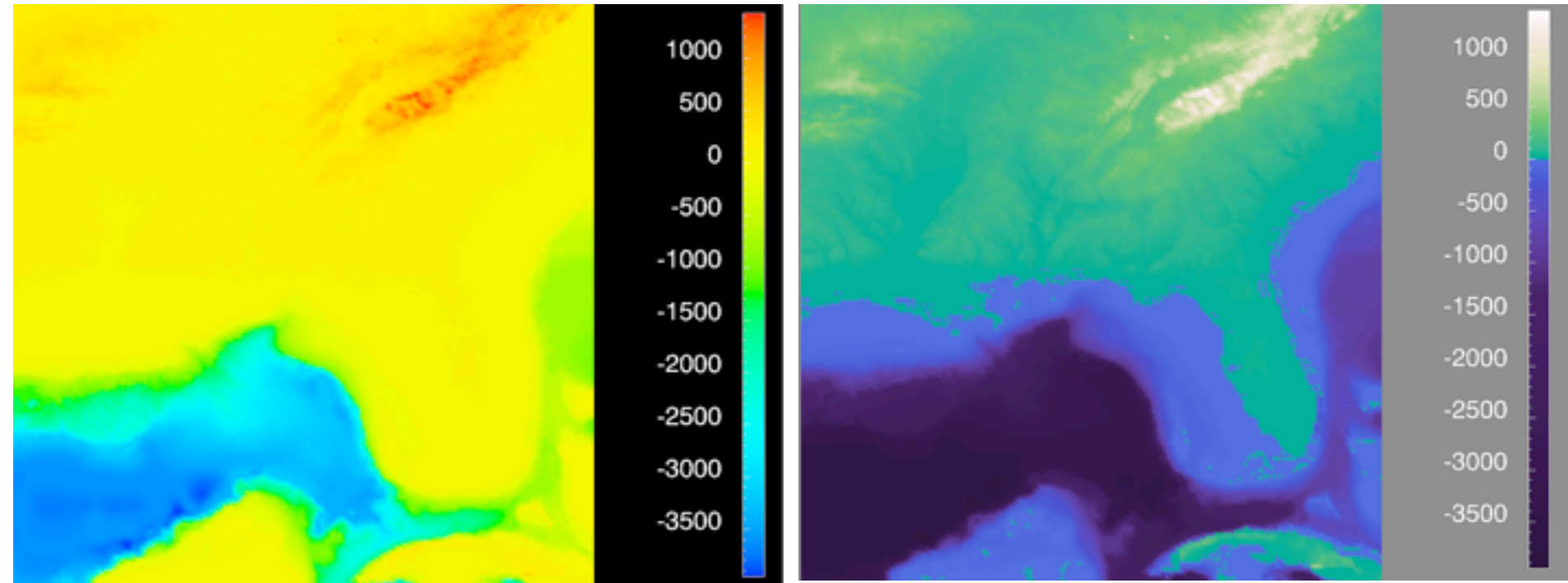
- noncontiguous small regions of color: only 6-12 bins



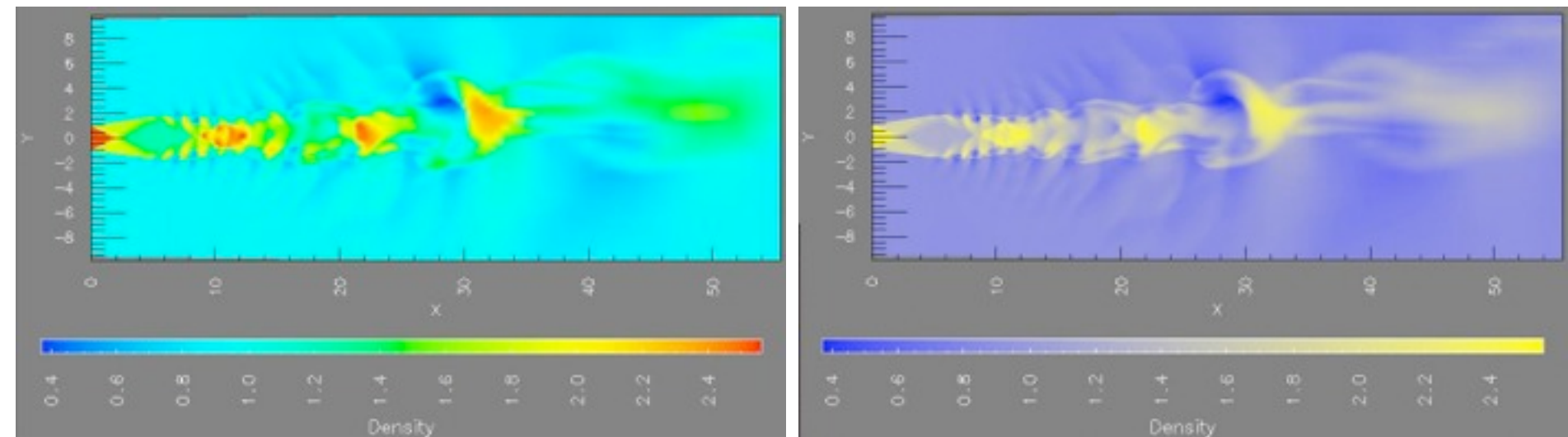
[Cinteny: flexible analysis and visualization of synteny and genome rearrangements in multiple organisms. Sinha and Meller. BMC Bioinformatics, 8:82, 2007.]

Ordered color: Rainbow is poor default

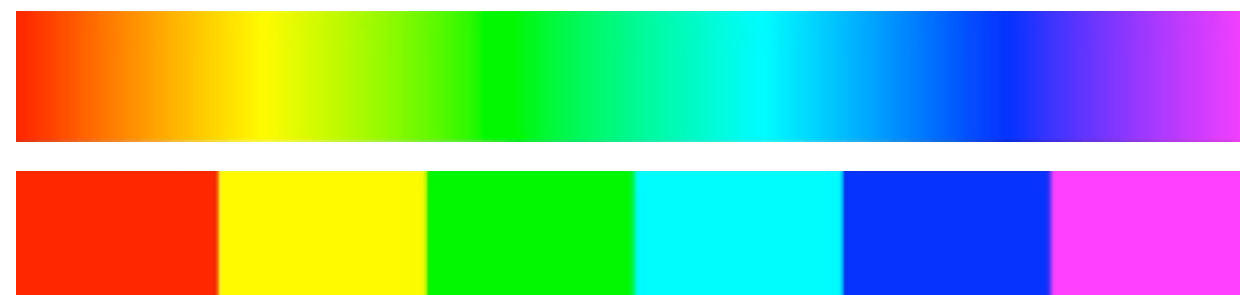
- problems
 - perceptually unordered
 - perceptually nonlinear
- benefits
 - fine-grained structure visible and nameable
- alternatives
 - fewer hues for large-scale structure
 - multiple hues with monotonically increasing luminance for fine-grained
 - segmented rainbows good for categorical, ok for binned



[Why Should Engineers Be Worried About Color? Treinish and Rogowitz 1998. <http://www.research.ibm.com/people/lloyd/color/color.HTM>]



[A Rule-based Tool for Assisting Colormap Selection. Bergman, Rogowitz, and Treinish. Proc. IEEE Visualization (Vis), pp. 118–125, 1995.]



[Transfer Functions in Direct Volume Rendering: Design, Interface, Interaction. Kindlmann. SIGGRAPH 2002 Course Notes]

How?

Encode

→ Arrange

→ Express



→ Order



→ Use



→ Separate



→ Align



→ Map

from **categorical** and **ordered** attributes

→ Color

→ Hue



→ Saturation



→ Luminance



→ Size, Angle, Curvature, ...



→ Shape



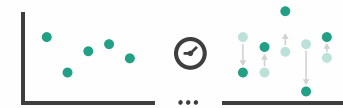
→ Motion

Direction, Rate, Frequency, ...



Manipulate

→ Change



→ Select

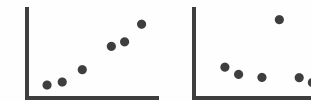


→ Navigate

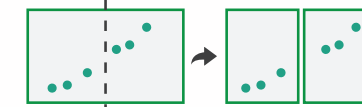


Facet

→ Juxtapose



→ Partition



→ Superimpose



Reduce

→ Filter



→ Aggregate



→ Embed



What?

Why?

How?

How to handle complexity: 3 more strategies + 1 previous

Manipulate

➔ Change



➔ Select

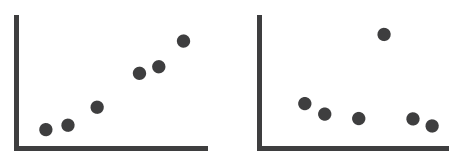


➔ Navigate

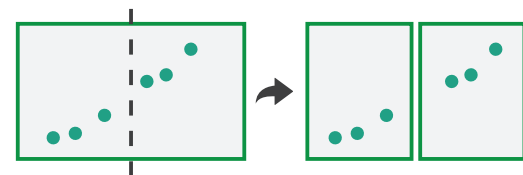


Facet

➔ Juxtapose



➔ Partition



➔ Superimpose

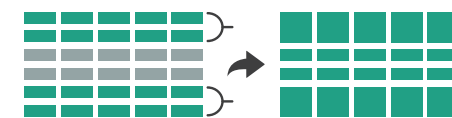


Reduce

➔ Filter



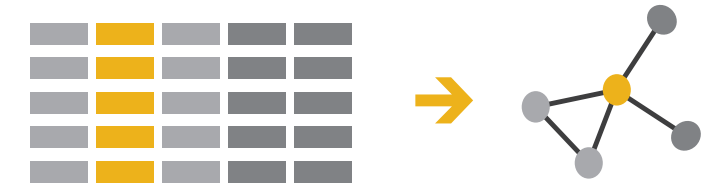
➔ Aggregate



➔ Embed



➔ *Derive*




- change view over time
- facet across multiple views
- reduce items/attributes within single view
- derive new data to show within view

How to handle complexity: 3 more strategies

+ 1 previous

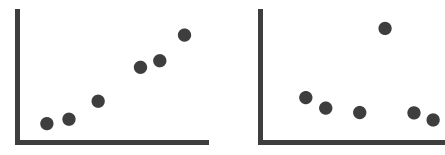
Manipulate

→ **Change**



Facet

→ **Juxtapose**

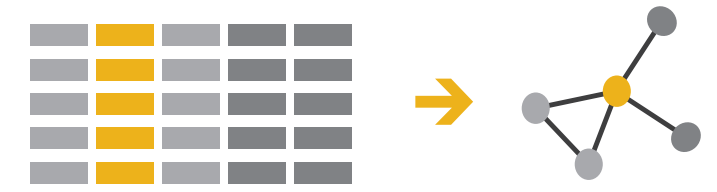


Reduce

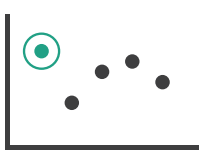
→ **Filter**



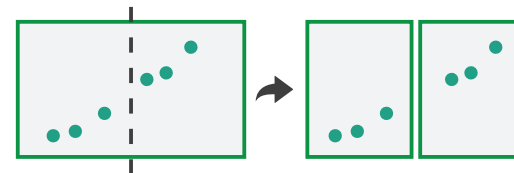
→ *Derive*



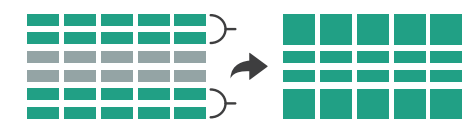
→ **Select**



→ **Partition**



→ **Aggregate**

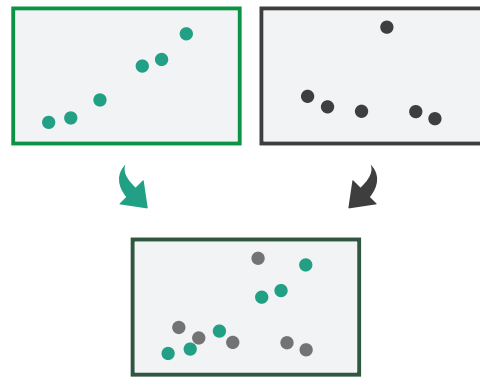


- **change over time**
- most obvious & flexible of the 4 strategies


→ **Navigate**



→ **Superimpose**

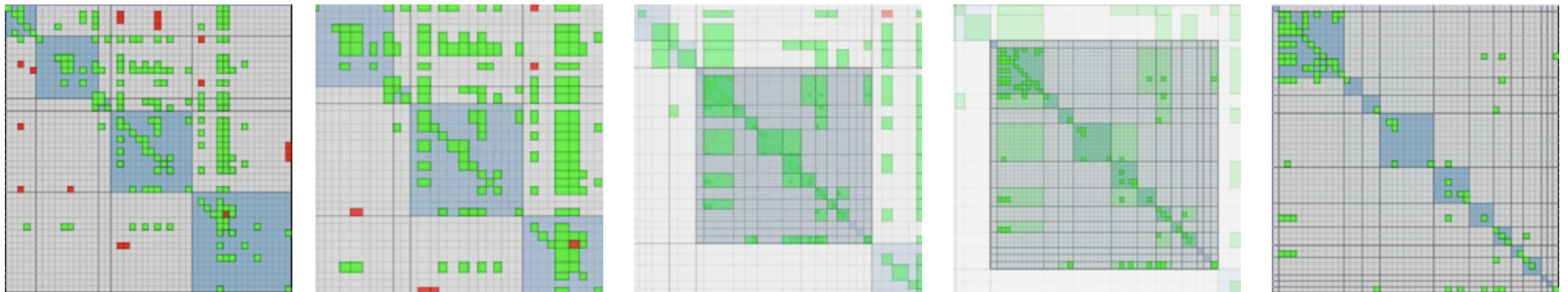


→ **Embed**



Idiom: **Animated transitions**

- smooth transition from one state to another
 - alternative to jump cuts
 - support for item tracking when amount of change is limited
- example: multilevel matrix views
 - scope of what is shown narrows down
 - middle block stretches to fill space, additional structure appears within
 - other blocks squish down to increasingly aggregated representations



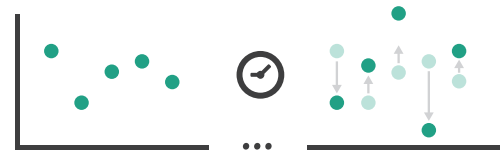
[Using Multilevel Call Matrices in Large Software Projects. van Ham. Proc. IEEE Symp. Information Visualization (InfoVis), pp. 227–232, 2003.]

How to handle complexity: 3 more strategies

+ 1 previous

Manipulate

➔ Change



➔ Select

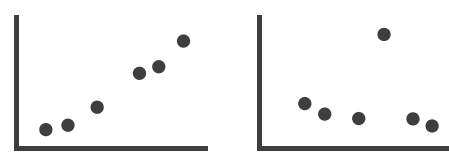


➔ Navigate

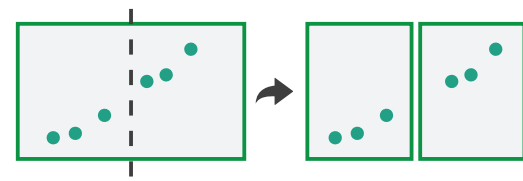


Facet

➔ Juxtapose



➔ Partition



➔ Superimpose



Reduce

➔ Filter



➔ Aggregate



➔ Embed



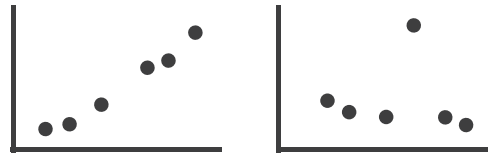
➔ *Derive*



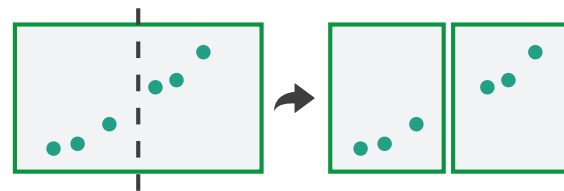
- facet data across multiple views

Facet

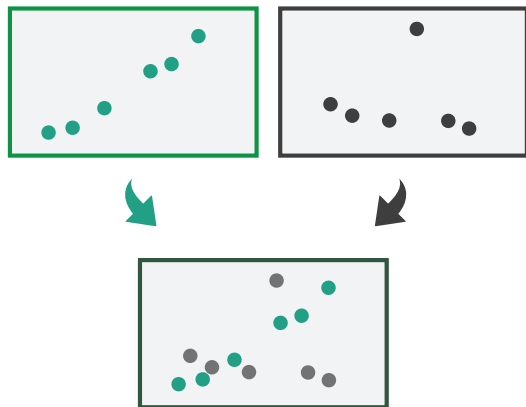
→ Juxtapose



→ Partition



→ Superimpose



→ Coordinate Multiple Side By Side Views

→ Share Encoding: Same/Different

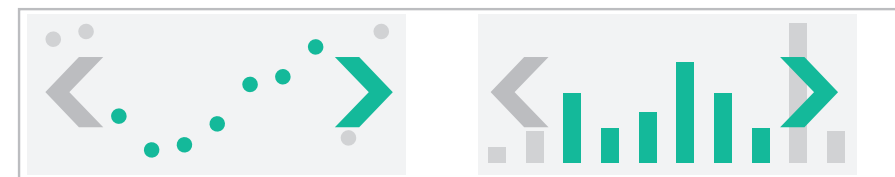
→ *Linked Highlighting*



→ Share Data: All/Subset/None



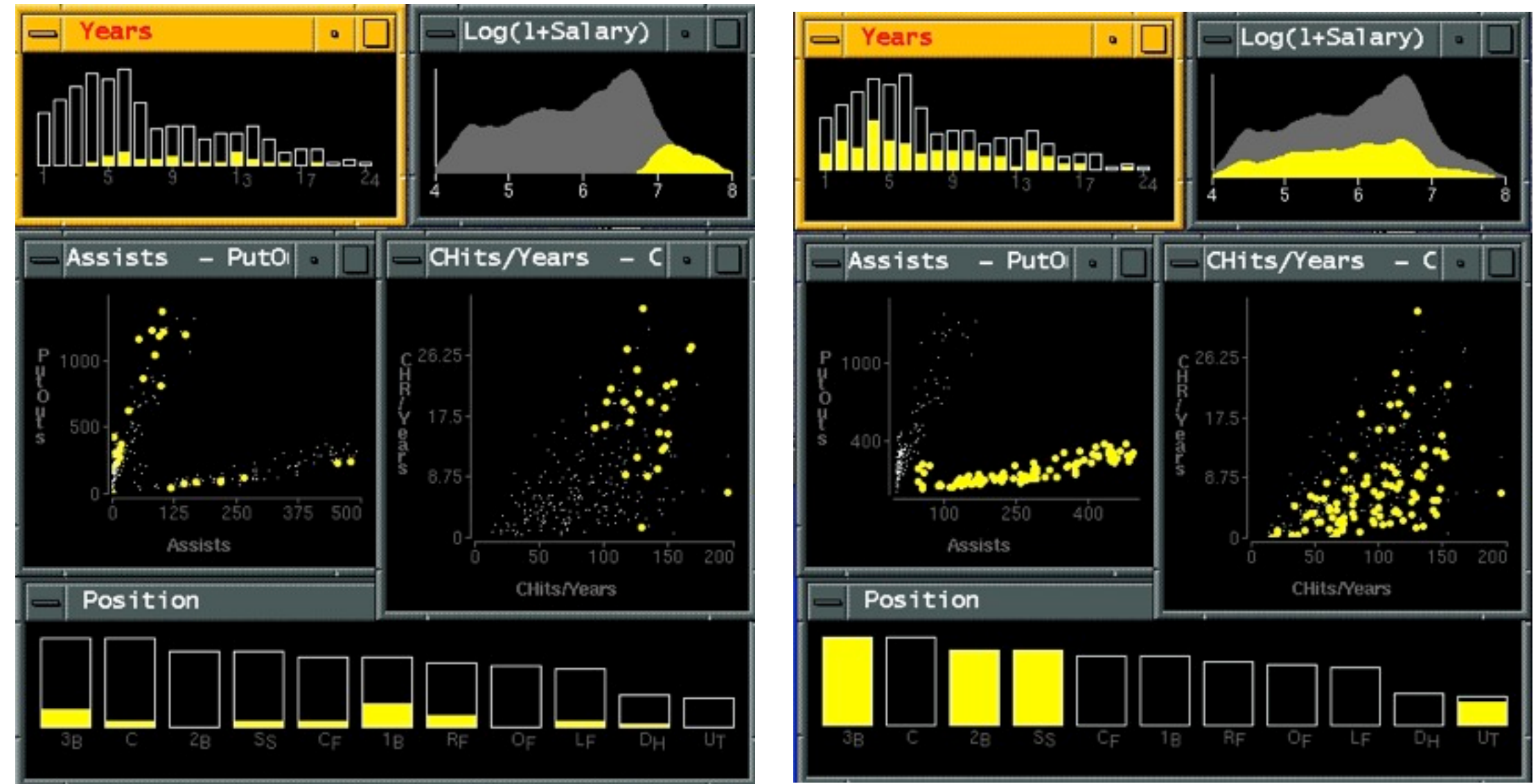
→ Share Navigation



Idiom: **Linked highlighting**

System: **EDV**

- see how regions contiguous in one view are distributed within another
 - powerful and pervasive interaction idiom
- encoding: different
 - **multiform**
- data: all shared



[Visual Exploration of Large Structured Datasets. Wills. Proc. New Techniques and Trends in Statistics (NTTS), pp. 237–246. IOS Press, 1995.]

Idiom: **bird's-eye maps**

System: **Google Maps**

- encoding: same
- data: subset shared
- navigation: shared
 - bidirectional linking

- differences
 - viewpoint
 - (size)

- **overview-detail**

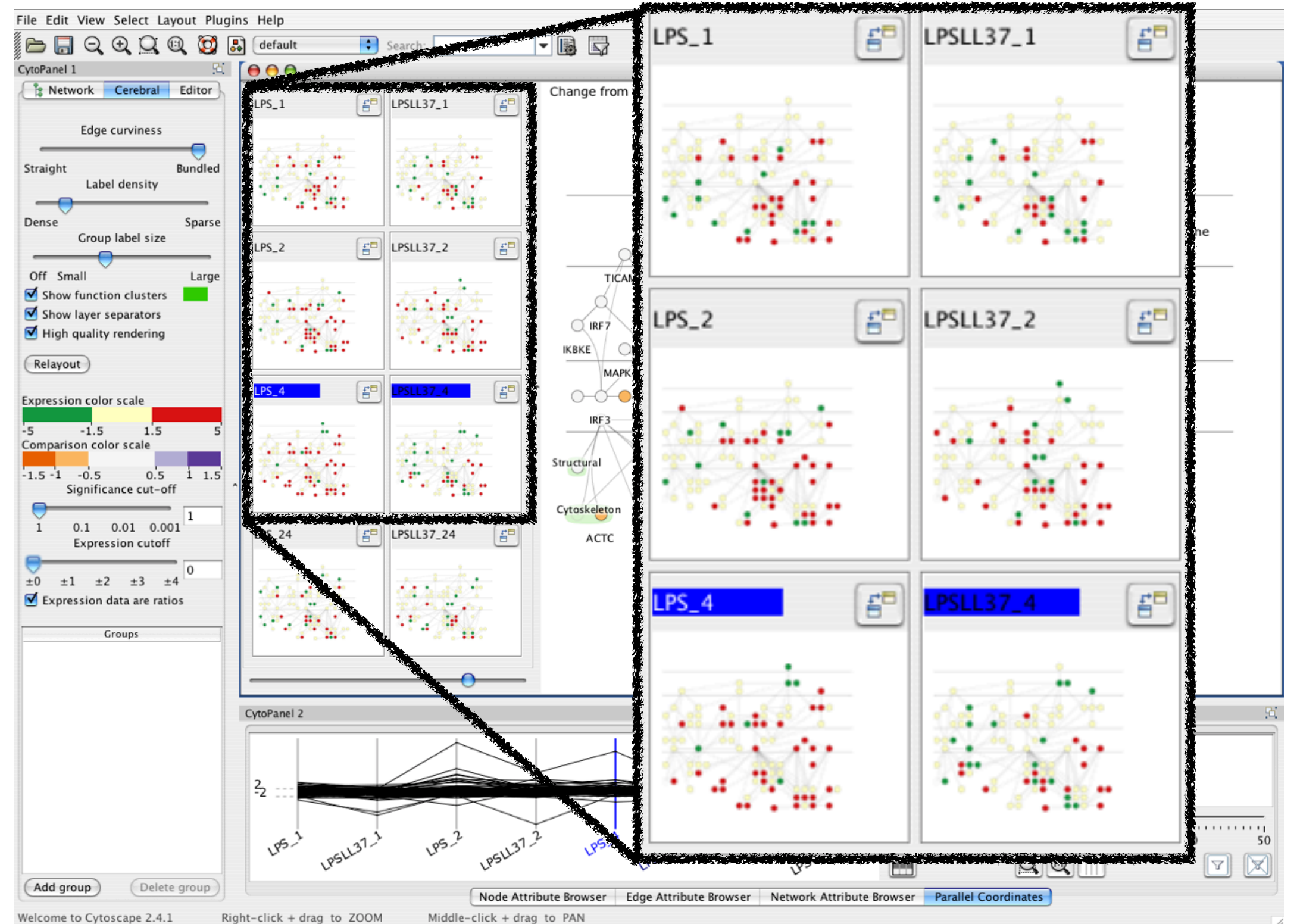


[A Review of Overview+Detail, Zooming, and Focus+Context Interfaces. Cockburn, Karlson, and Bederson. *ACM Computing Surveys* 41:1 (2008), 1–31.]

Idiom: **Small multiples**



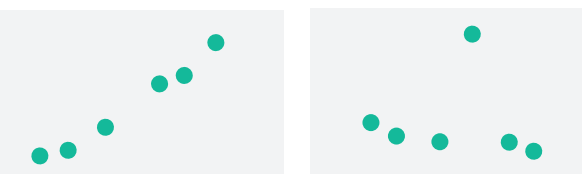



System: **Cerebral**

- encoding: same
- data: none shared
 - different attributes for node colors
 - (same network layout)
- navigation: shared



[Cerebral: Visualizing Multiple Experimental Conditions on a Graph with Biological Context. Barsky, Munzner, Gardy, and Kincaid. *IEEE Trans. Visualization and Computer Graphics (Proc. InfoVis 2008)* 14:6 (2008), 1253–1260.]

Coordinate views: Design choice interaction

		Data		
		All	Subset	None
Encoding	Same	 <p>Redundant</p>	 <p>Overview/ Detail</p>	 <p>Small Multiples</p>
	Different	 <p>Multiform</p>	 <p>Multiform, Overview/ Detail</p>	 <p>No Linkage</p>

- why juxtapose views?

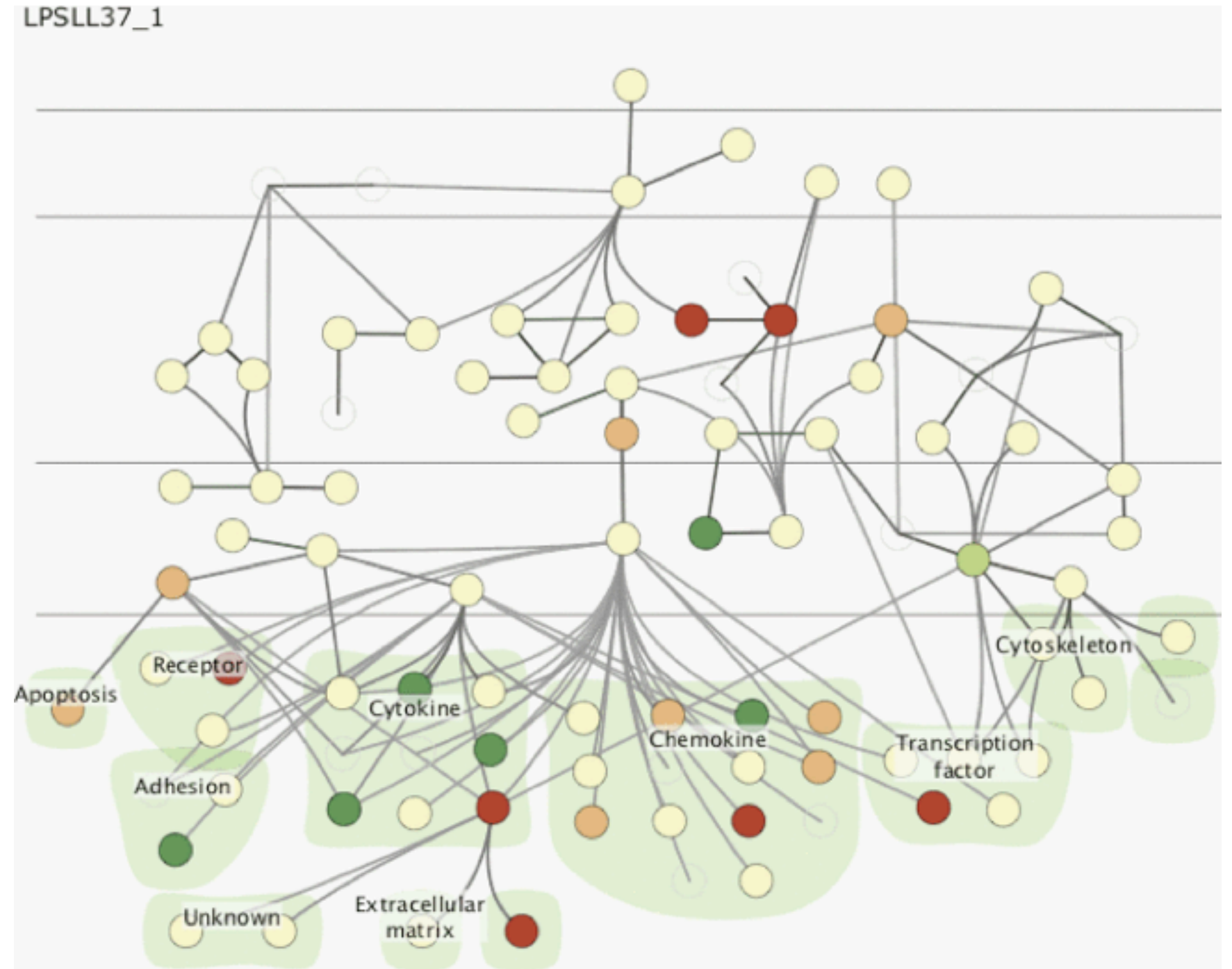
- benefits: eyes vs memory

- lower cognitive load to move eyes between 2 views than remembering previous state with single changing view

- costs: display area, 2 views side by side each have only half the area of one view

Why not animation?

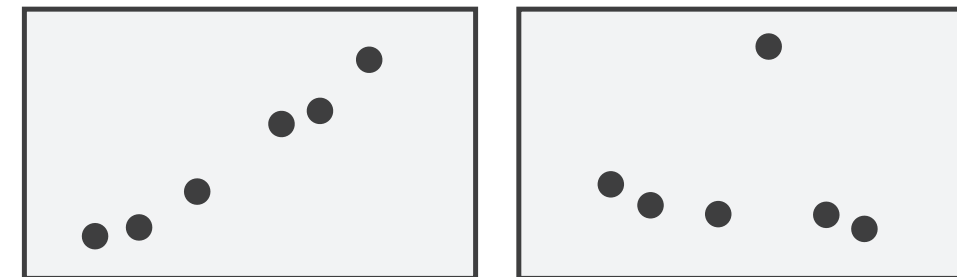
- disparate frames and regions: comparison difficult
 - vs contiguous frames
 - vs small region
 - vs coherent motion of group
- safe special case
 - animated transitions



Partition into views

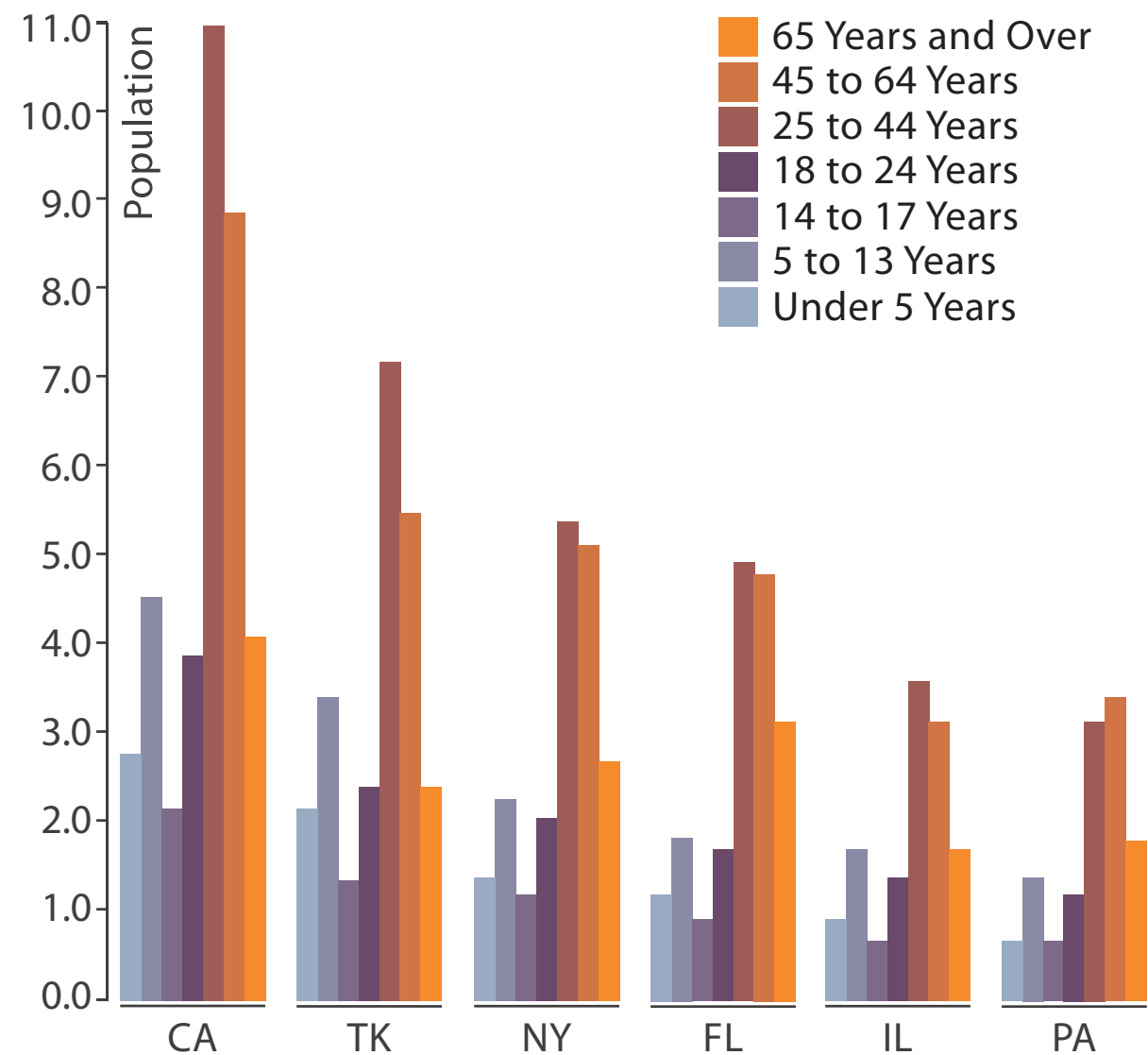
- how to divide data between views
 - encodes association between items using spatial proximity
 - major implications for what patterns are visible
 - split according to attributes
- design choices
 - how many splits
 - all the way down: one mark per region?
 - stop earlier, for more complex structure within region?
 - order in which attribs used to split
 - how many views

➔ Partition into Side-by-Side Views

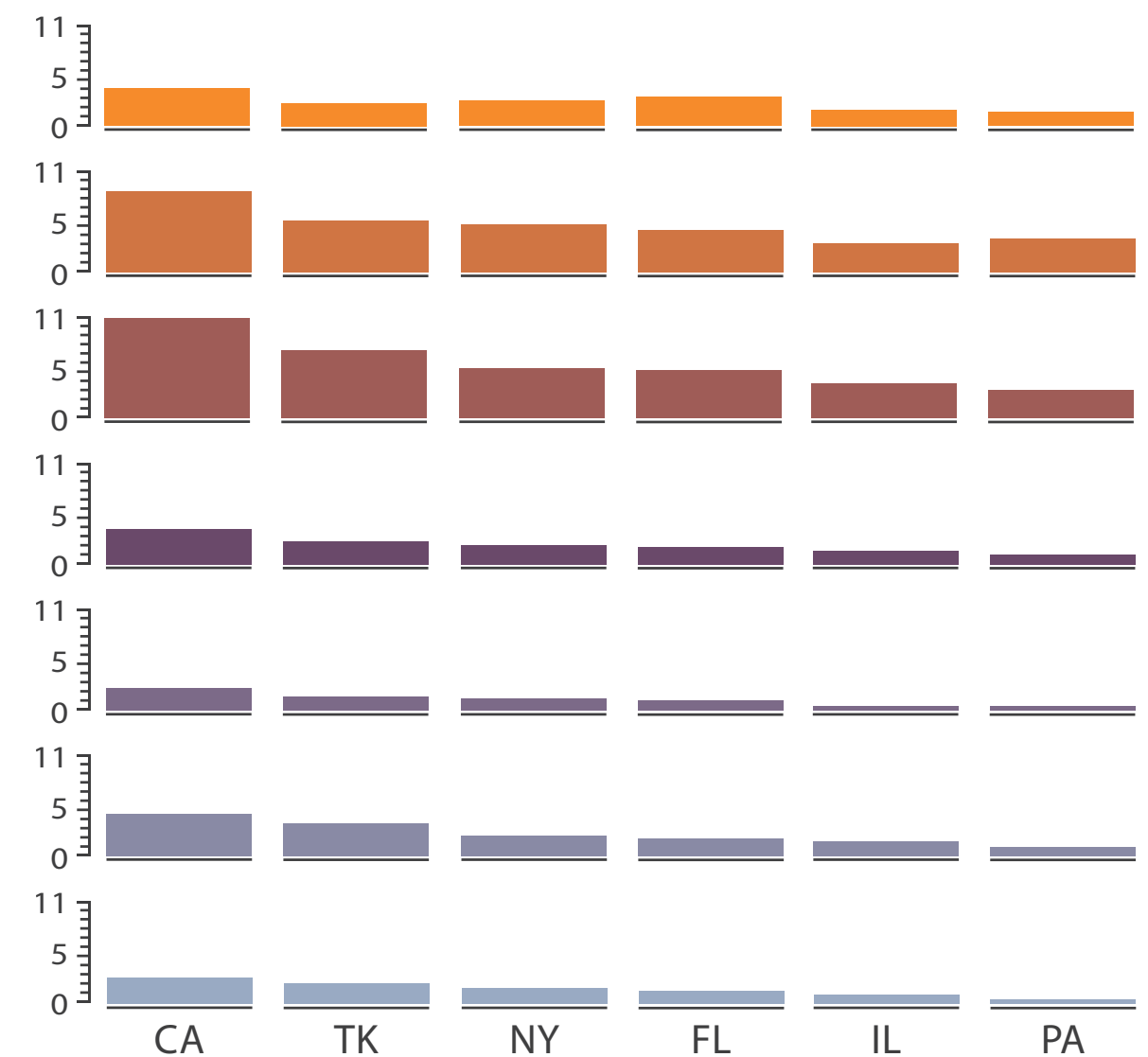


Partitioning: List alignment

- single bar chart with grouped bars
 - split by state into regions
 - complex glyph within each region showing all ages
 - compare: easy within state, hard across ages



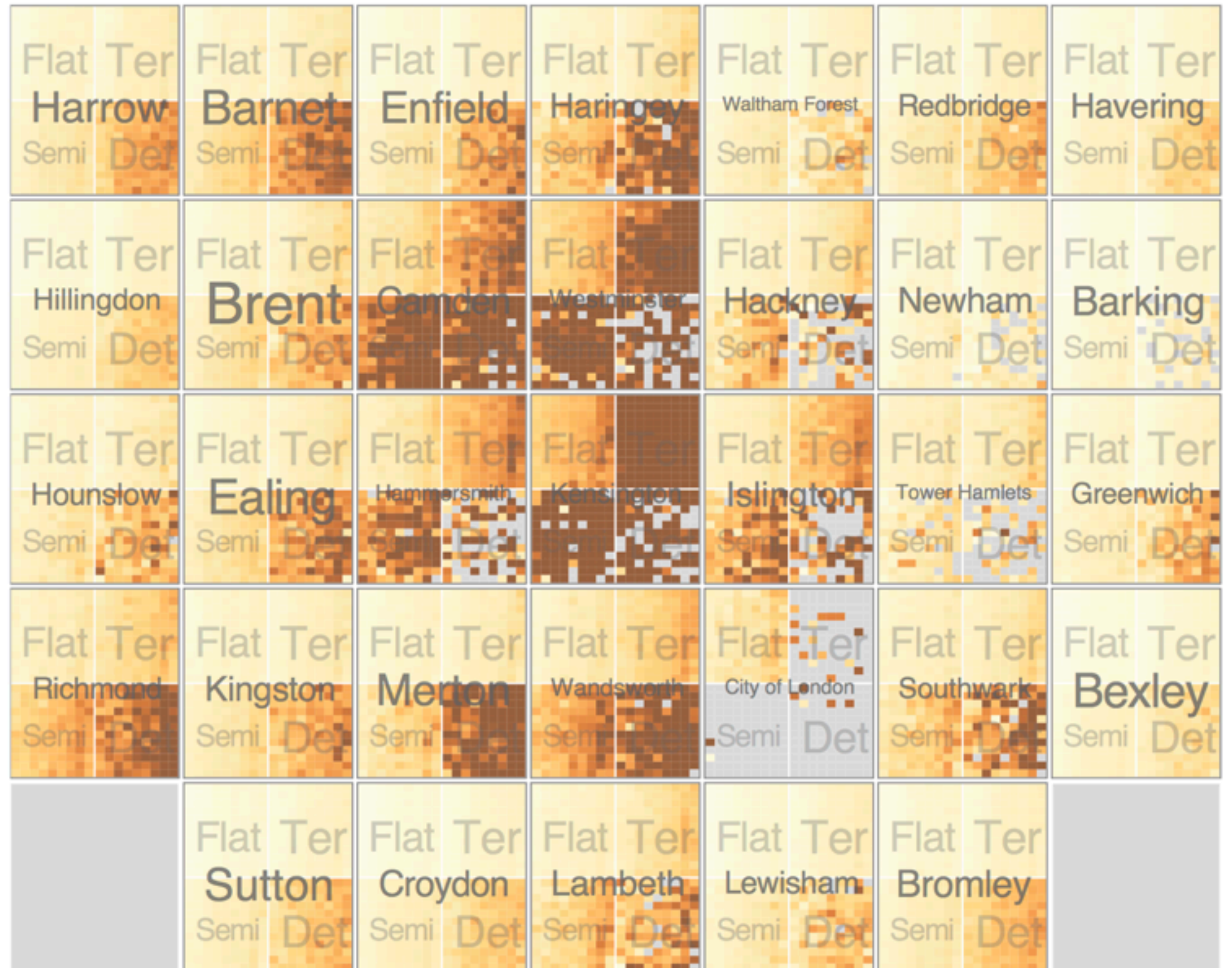
- small-multiple bar charts
 - split by age into regions
 - one chart per region
 - compare: easy within age, harder across states



Partitioning: Recursive subdivision

System: **HIVE**

- split by neighborhood
- then by type
- then time
 - years as rows
 - months as columns
- color by price
- neighborhood patterns
 - where it's expensive
 - where you pay much more for detached type



Partitioning: Recursive subdivision

System: **HIVE**

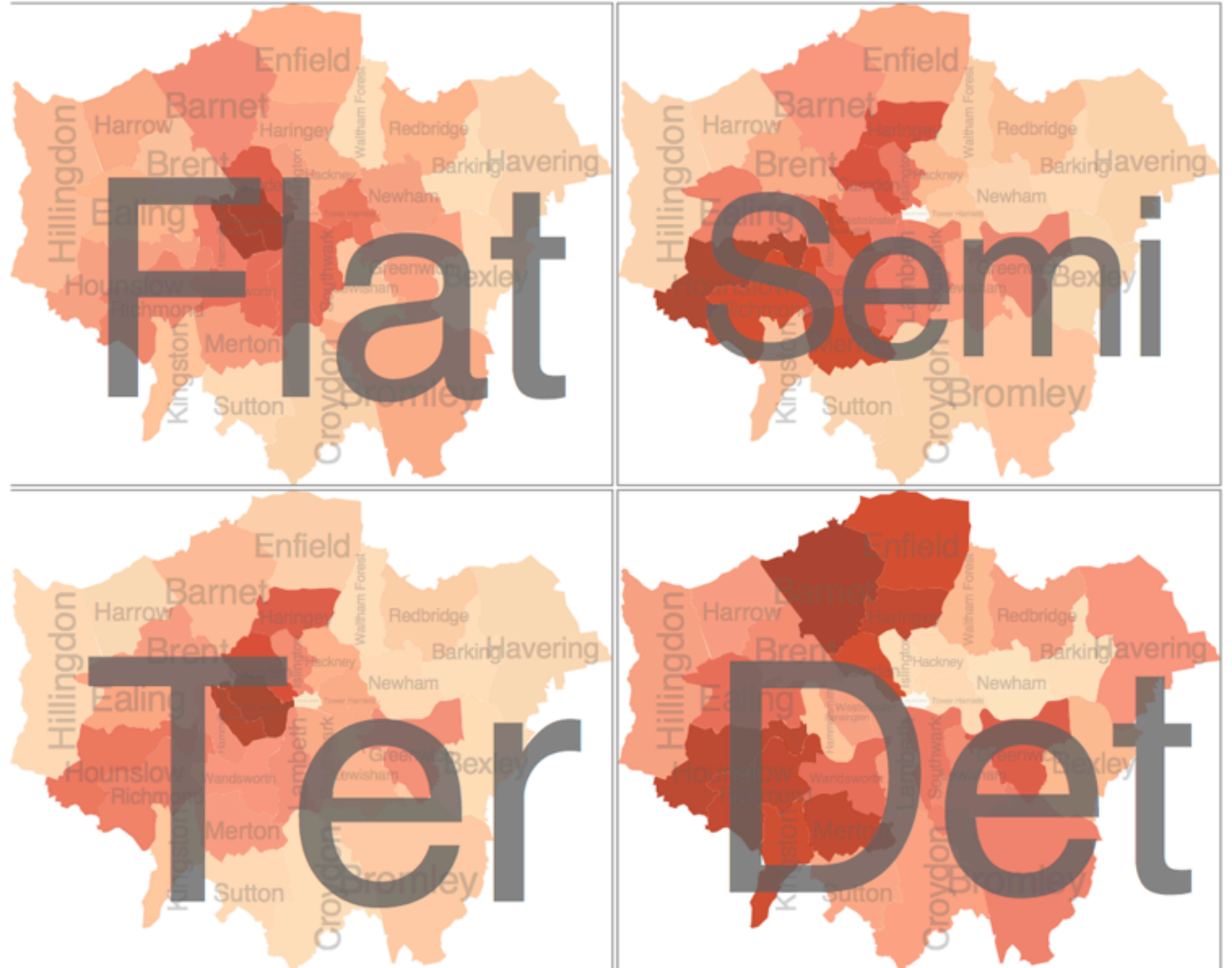
- switch order of splits
 - type then neighborhood
- switch color
 - by price variation
- type patterns
 - within specific type, which neighborhoods inconsistent



Partitioning: Recursive subdivision

System: **HIVE**

- different encoding for second-level regions
 - choropleth maps

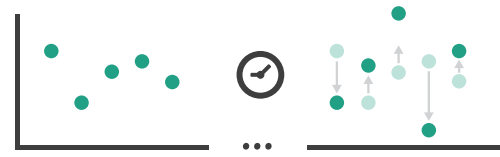


How to handle complexity: 3 more strategies

+ 1 previous

Manipulate

➔ Change



➔ Select

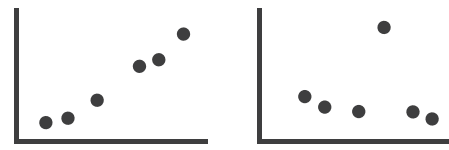


➔ Navigate

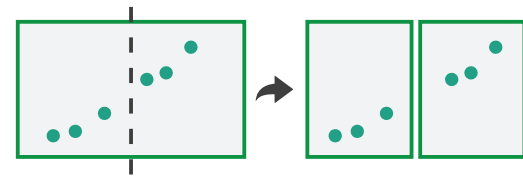


Facet

➔ Juxtapose



➔ Partition



➔ Superimpose



Reduce

➔ Filter



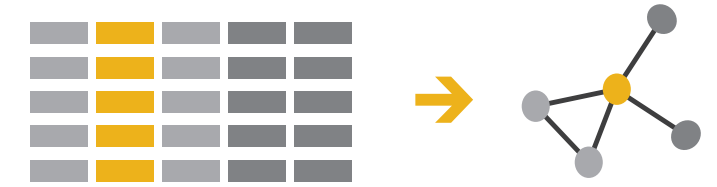
➔ Aggregate



➔ Embed



➔ *Derive*



- reduce what is shown within single view

Reduce items and attributes

- reduce/increase: inverses
- filter
 - pro: straightforward and intuitive
 - to understand and compute
 - con: out of sight, out of mind
- aggregation
 - pro: inform about whole set
 - con: difficult to avoid losing signal
- not mutually exclusive
 - combine filter, aggregate
 - combine reduce, facet, change, derive

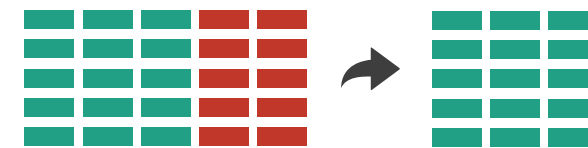
Reducing Items and Attributes

→ Filter

→ Items

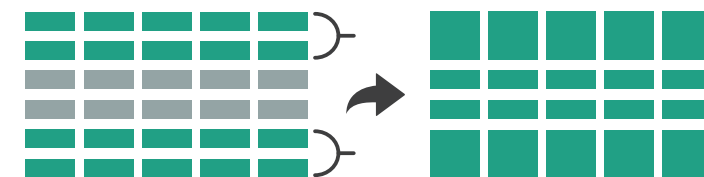


→ Attributes

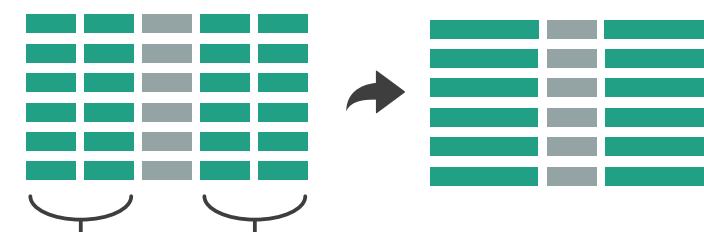


→ Aggregate

→ Items



→ Attributes

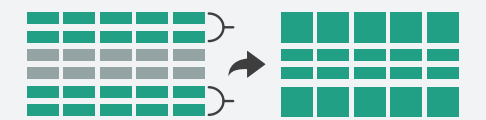


Reduce

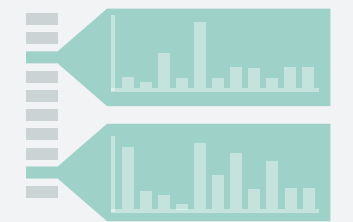
→ Filter



→ Aggregate

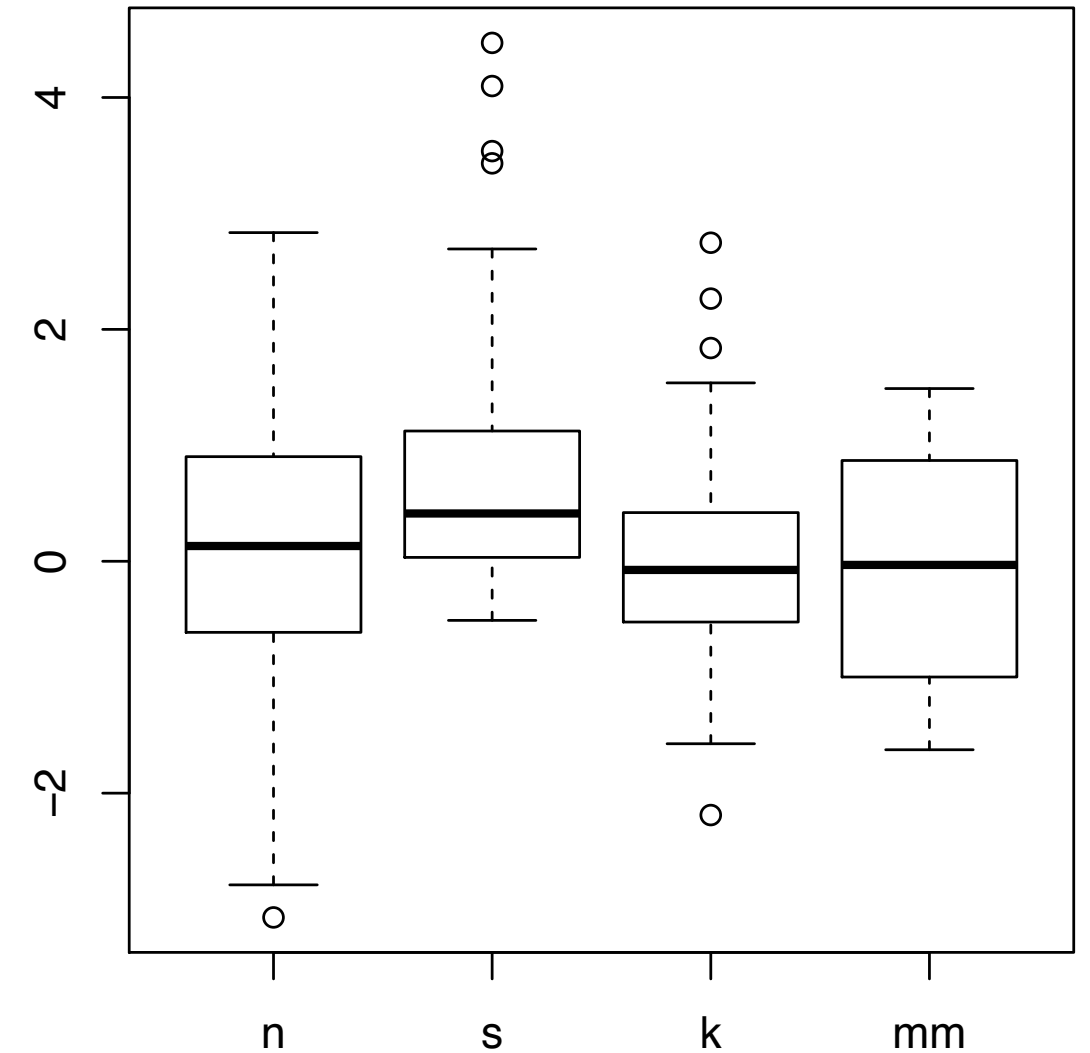


→ Embed



Idiom: **boxplot**

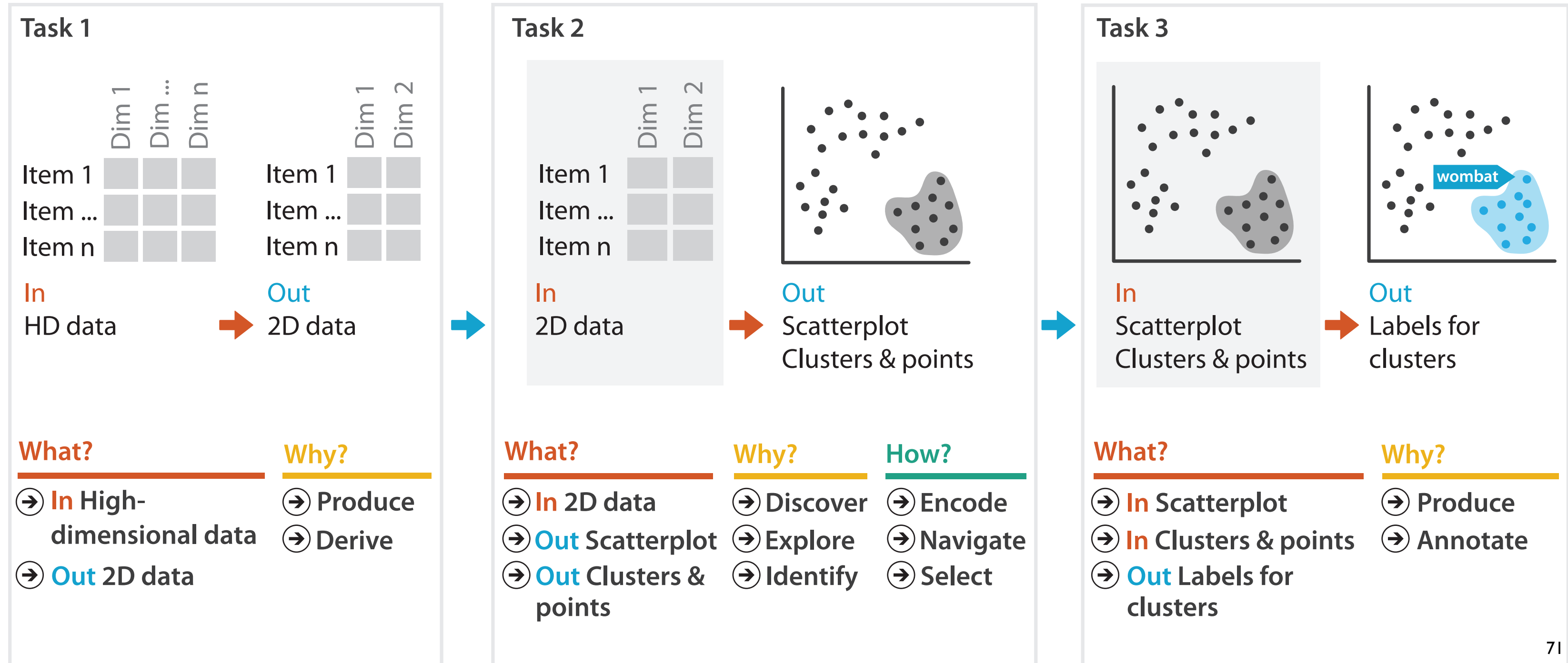
- static item aggregation
- task: find distribution
- data: table
- derived data
 - 5 quant attribs
 - median: central line
 - lower and upper quartile: boxes
 - lower upper fences: whiskers
 - values beyond which items are outliers
 - outliers beyond fence cutoffs explicitly shown



[40 years of boxplots. Wickham and Stryjewski. 2012. had.co.nz]

Idiom: Dimensionality reduction for documents

- attribute aggregation
 - derive low-dimensional target space from high-dimensional measured space



What?

Datasets

Attributes

domain

abstraction

What?

Why?

idiom

How?

algorithm

Why?

Actions

Targets

→ Data Types

→ Items

→ Data and D

Tables

Items

Attributes

→ Analyze

→ Consume

→ Discover



→ Present



→ Enjoy



→ All Data

→ Trends



→ Outliers

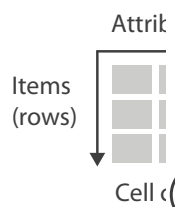


→ Features



→ Dataset Typ

→ Tables



→ Produce

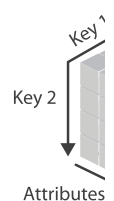
→ Annotate



→ Search

	Tar
Location known	••
Location unknown	<••

→ Multidir



→ Geometr



→ Query

→ Identify



How?

Encode

Manipulate

Facet

Reduce

→ Arrange

→ Express



→ Separate



→ Order



→ Align



→ Use



→ Map

from **categorical** and **ordered** attributes

→ Color

→ Hue



→ Saturation



→ Luminance



→ Size, Angle, Curvature, ...



→ Shape

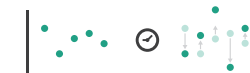


→ Motion

Direction, Rate, Frequency, ...



→ Change



→ Select



→ Navigate



→ Juxtapose



→ Partition



→ Superimpose



→ Filter



→ Aggregate



→ Embed



What?

Why?

More Information

[@tamaramunzner](https://twitter.com/tamaramunzner)

- this talk

<http://www.cs.ubc.ca/~tmm/talks.html#vad15bryan>

- book page (including tutorial lecture slides)

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

– 20% promo code for book+ebook combo:
HVN17

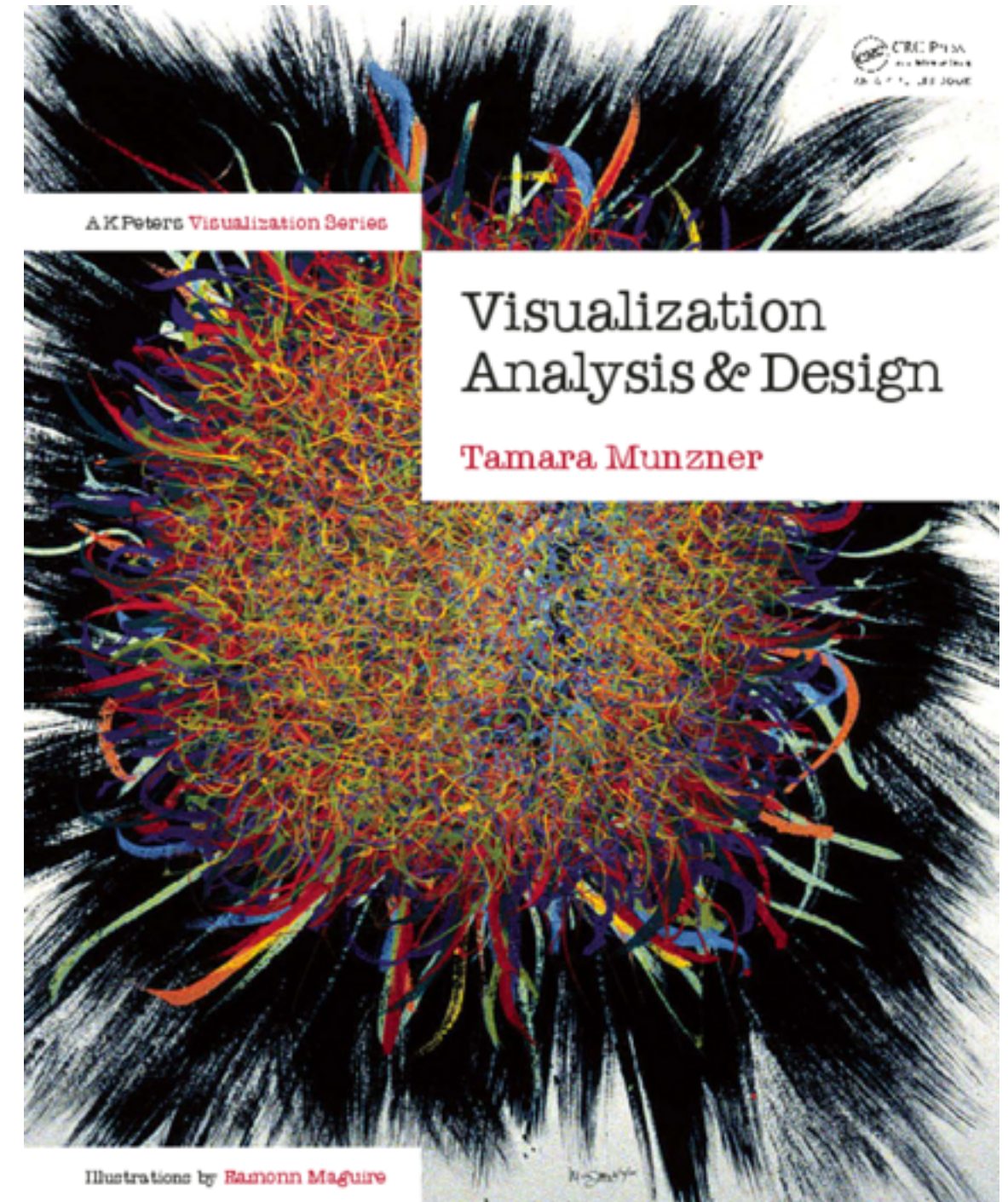
– <http://www.crcpress.com/product/isbn/9781466508910>

– illustrations: Eamonn Maguire

- papers, videos, software, talks, full courses

<http://www.cs.ubc.ca/group/infovis>

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



Visualization Analysis and Design.
Munzner. A K Peters Visualization Series, CRC Press, Visualization Series, 2014.