

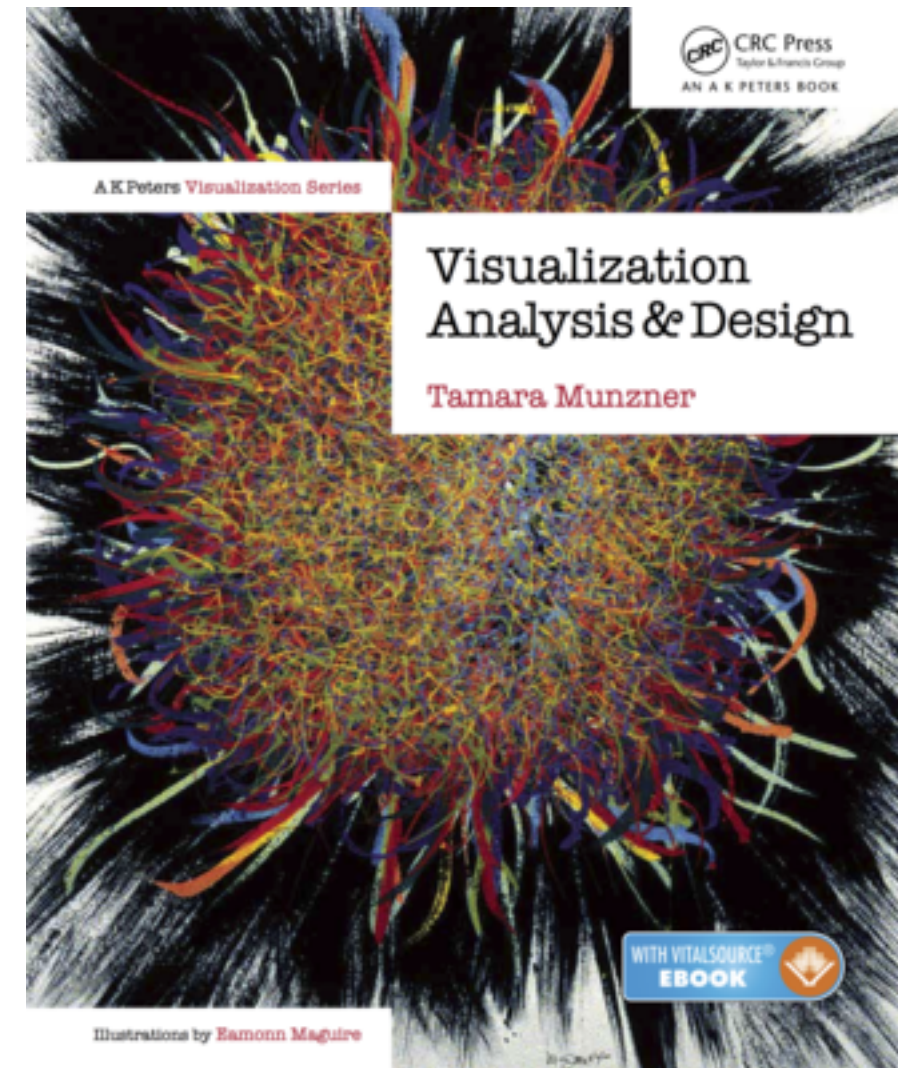
Visualization Analysis & Design

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*Guest Lecture, STAT 545
Oct 19 2017, Vancouver BC*

www.cs.ubc.ca/~tmm/talks.html#vad17stat545



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

Visualization (vis) defined & motivated

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

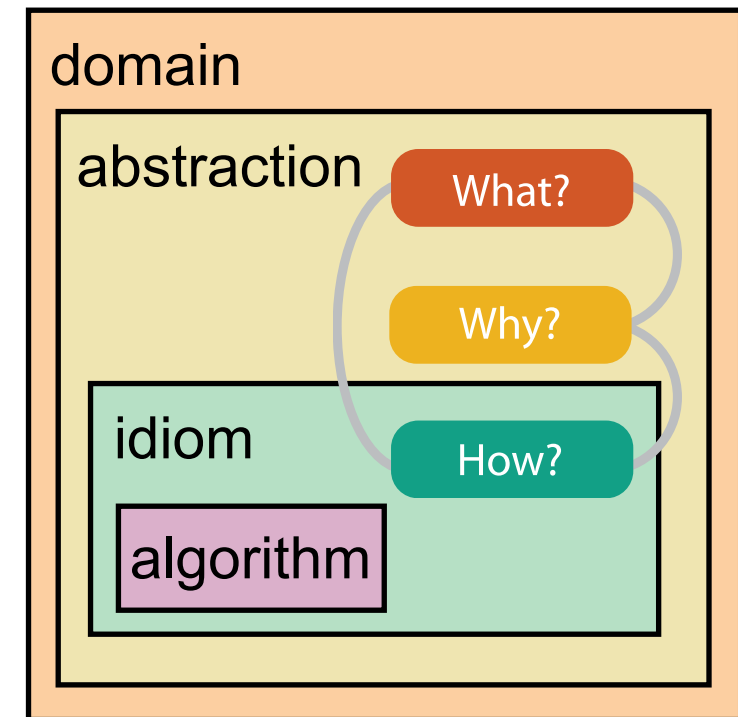
Visualization is suitable when there is a need to augment human capabilities rather than replace people with computational decision-making methods.

- human in the loop needs the details
 - doesn't know exactly what questions to ask in advance
 - longterm exploratory analysis
 - *speed up* through human-in-the-loop visual data analysis
 - presentation of known results
 - stepping stone towards automation: refining, trustbuilding
- intended task, measurable definitions of effectiveness

Nested model: Four levels of vis design

- *domain situation*
 - who are the target users?
- *abstraction*
 - translate from specifics of domain to vocabulary of vis
 - **what** is shown? **data abstraction**
 - **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



Domain situation

You misunderstood their needs



Data/task abstraction

You're showing them the wrong thing



Visual encoding/interaction idiom

The way you show it doesn't work

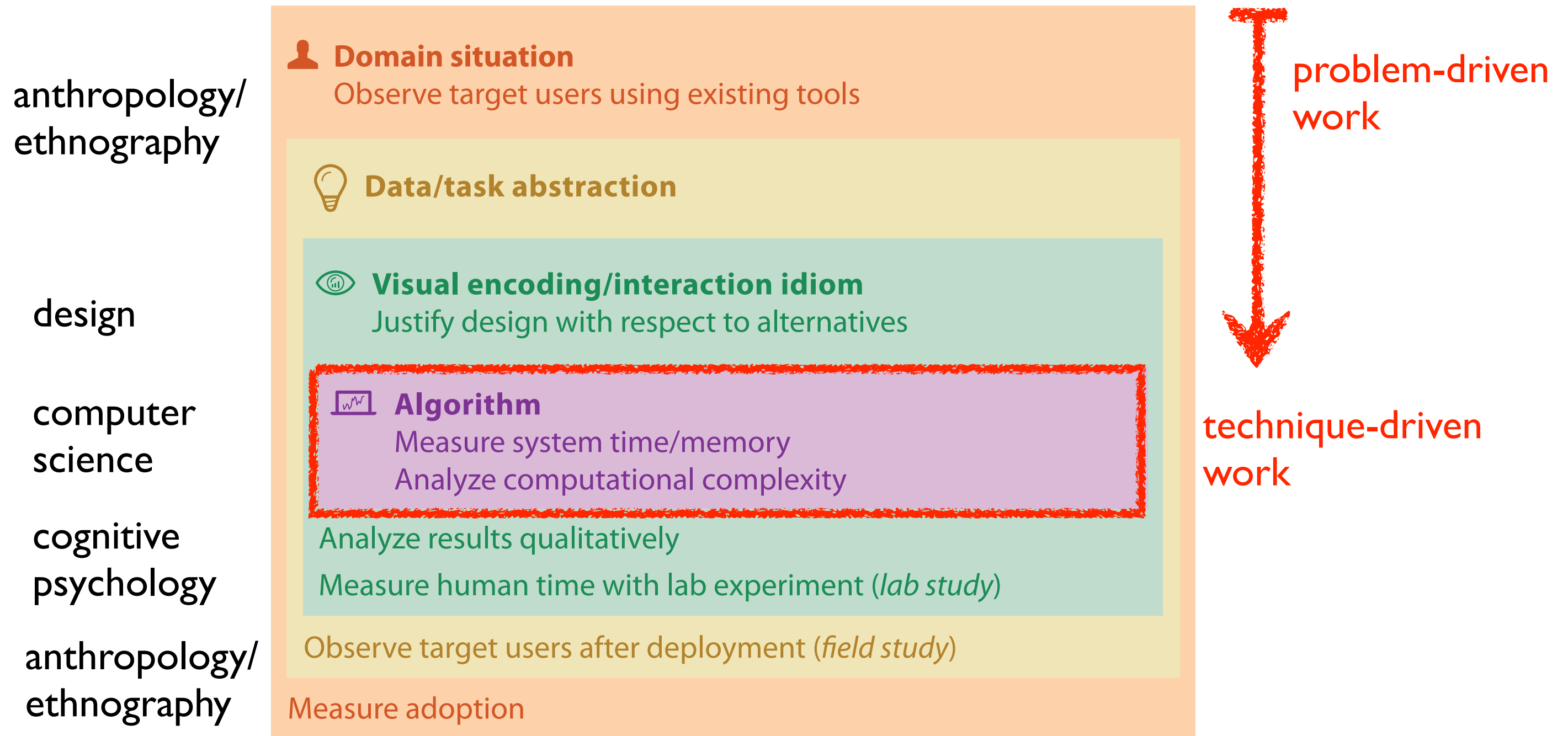


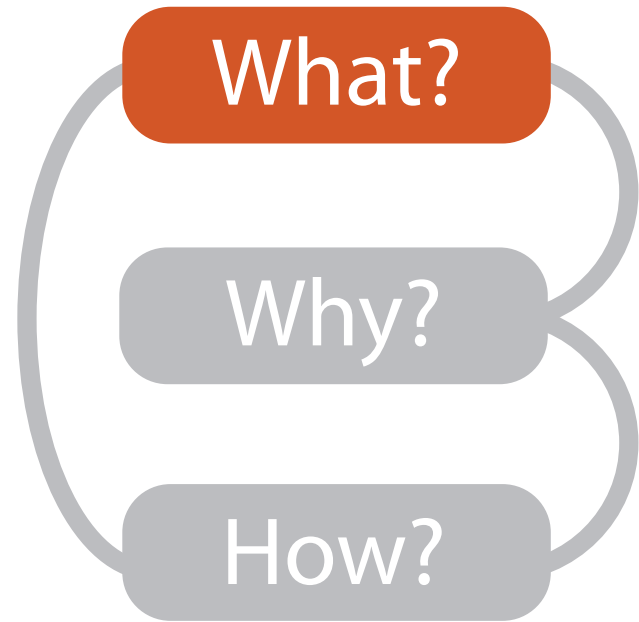
Algorithm

Your code is too slow

Why is validation difficult?

- solution: use methods from different fields at each level





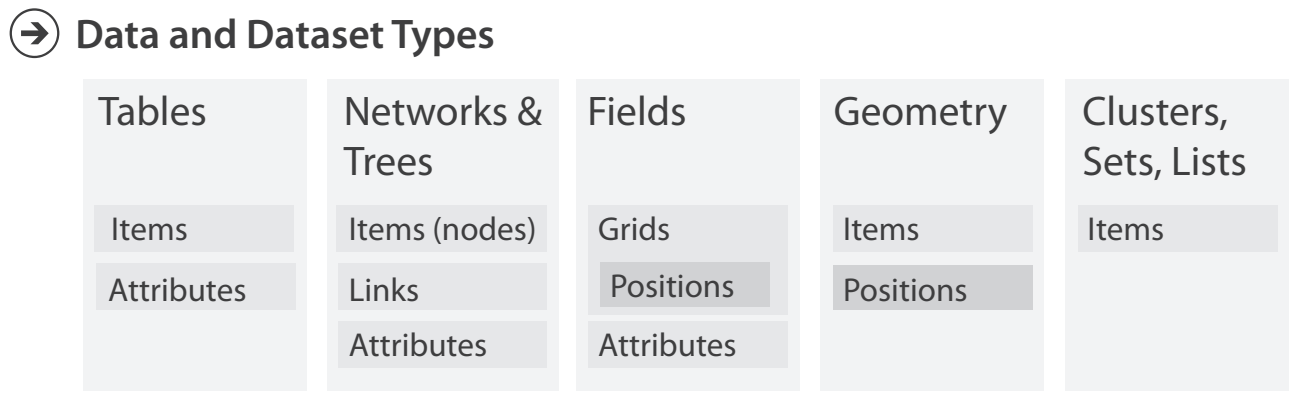
What?

Datasets

Attributes

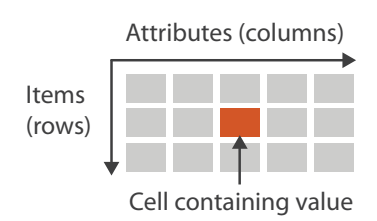
- ➔ Data Types
 - ➔ Items
 - ➔ Attributes
 - ➔ Links
 - ➔ Positions
 - ➔ Grids

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

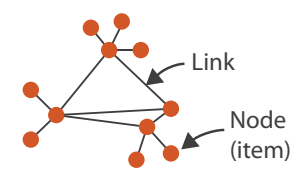


➔ Dataset Types

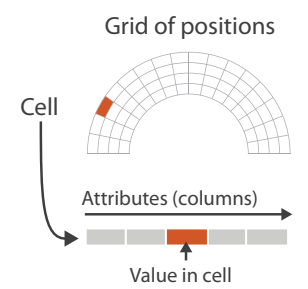
➔ Tables



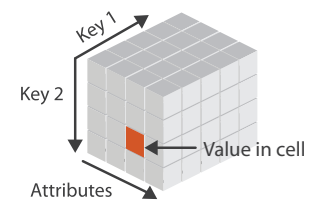
➔ Networks



➔ Fields (Continuous)



➔ Multidimensional Table



➔ Trees



➔ Geometry (Spatial)



➔ Ordering Direction

- ➔ Sequential
 -
- ➔ Diverging
 - ←→
- ➔ Cyclic
 - ↻

➔ Dataset Availability

➔ Static



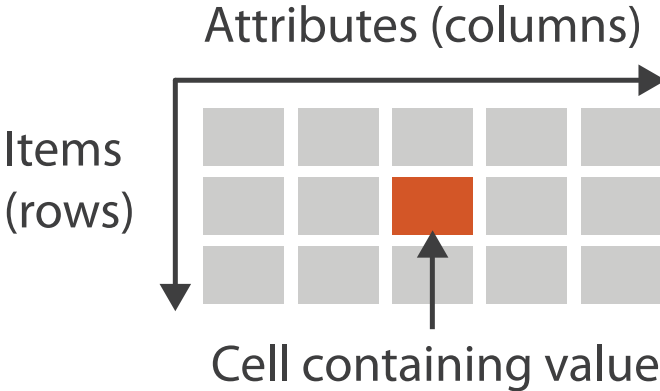
➔ Dynamic



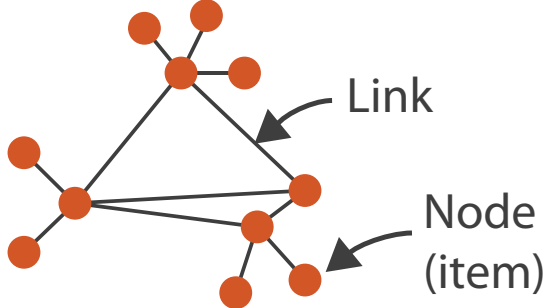
Types: Datasets and data

→ Dataset Types

→ Tables



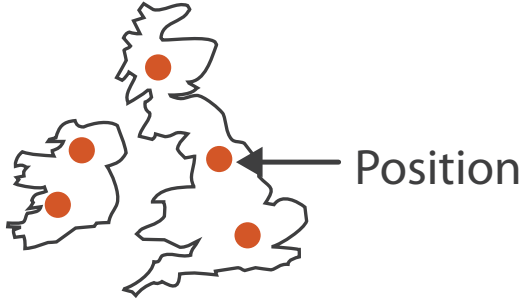
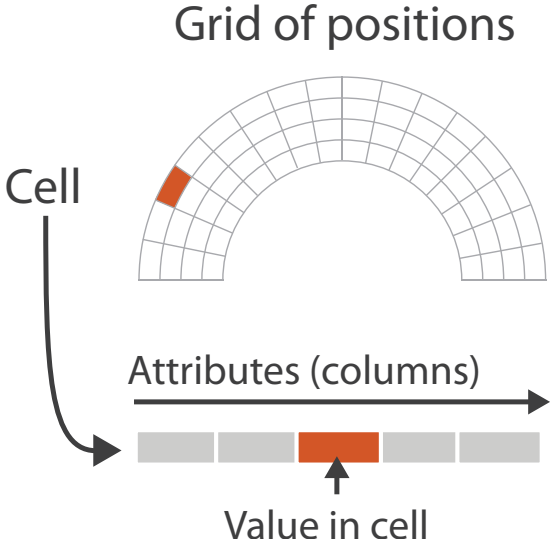
→ Networks



→ Spatial

→ Fields (Continuous)

→ Geometry (Spatial)



→ Attribute Types

→ Categorical

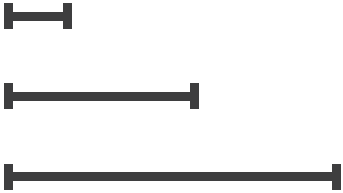


→ Ordered

→ Ordinal



→ Quantitative



→ Ordering Direction

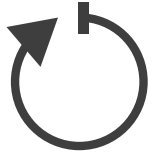
→ Sequential



→ Diverging









→ Cyclic






👉 Actions

🎯 Targets




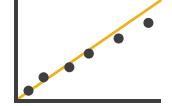
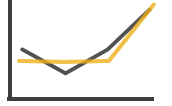
➔ **Analyze**

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





➔ **All Data**

- ➔ Trends 
- ➔ Outliers 
- ➔ Features 


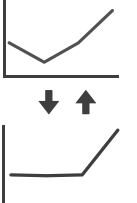

➔ **Attributes**

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

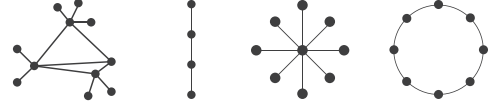

➔ **Search**

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


➔ **Query**

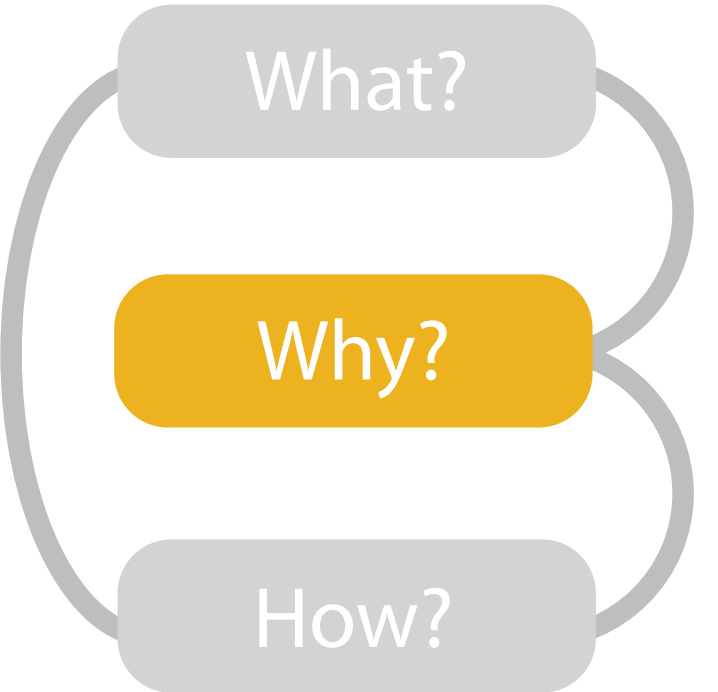
- ➔ Identify 
- ➔ Compare 
- ➔ Summarize 

➔ **Network Data**

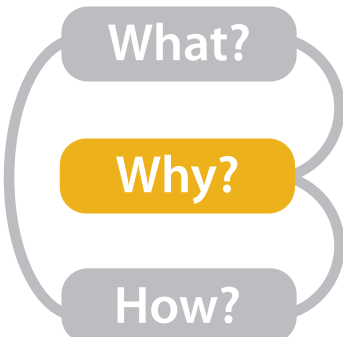
- ➔ Topology 
- ➔ Paths 

➔ **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

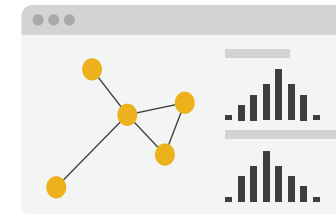
→ Analyze

→ Consume

→ *Discover*



→ *Present*

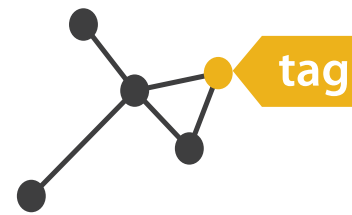


→ *Enjoy*



→ Produce

→ *Annotate*



→ *Record*

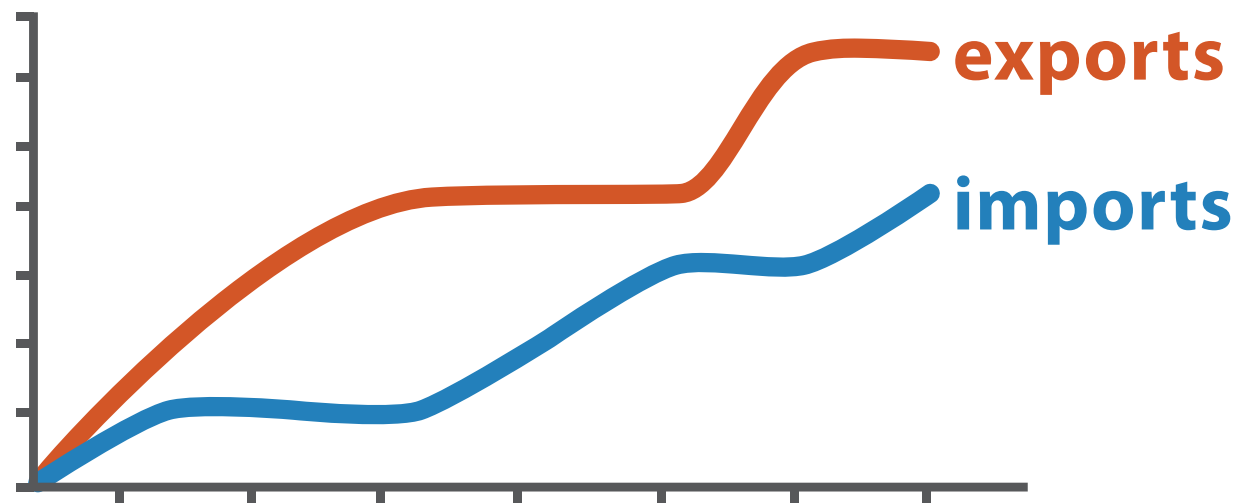


→ *Derive*

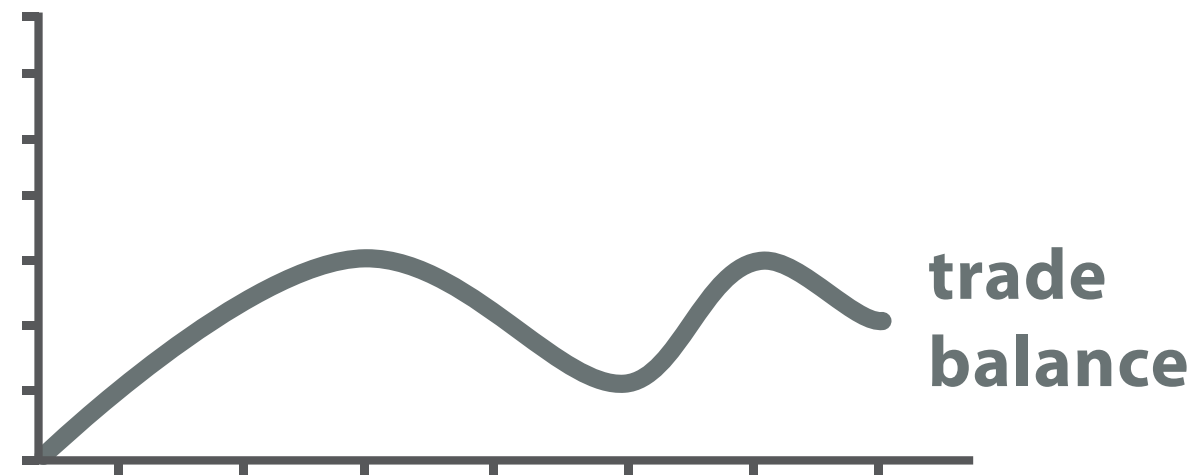


Derive

- 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



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

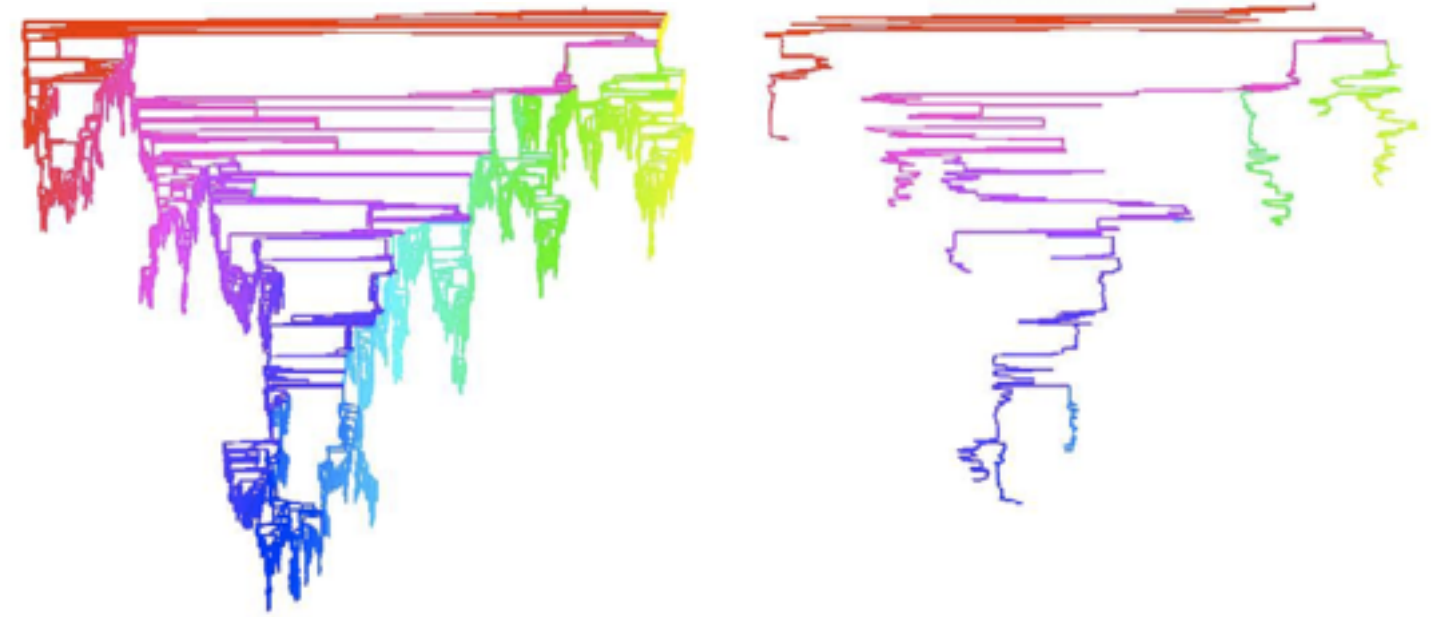
Derived Data

Analysis example: Derive one attribute

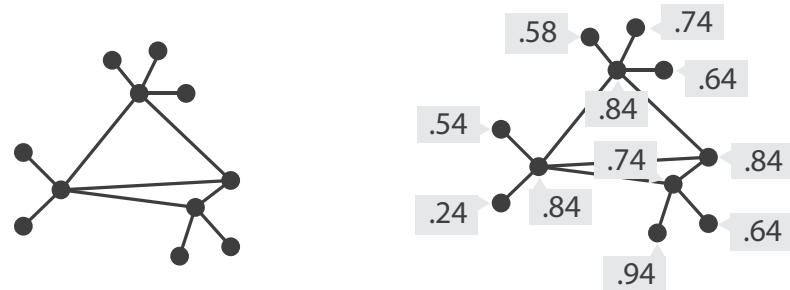
- Strahler number

- 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

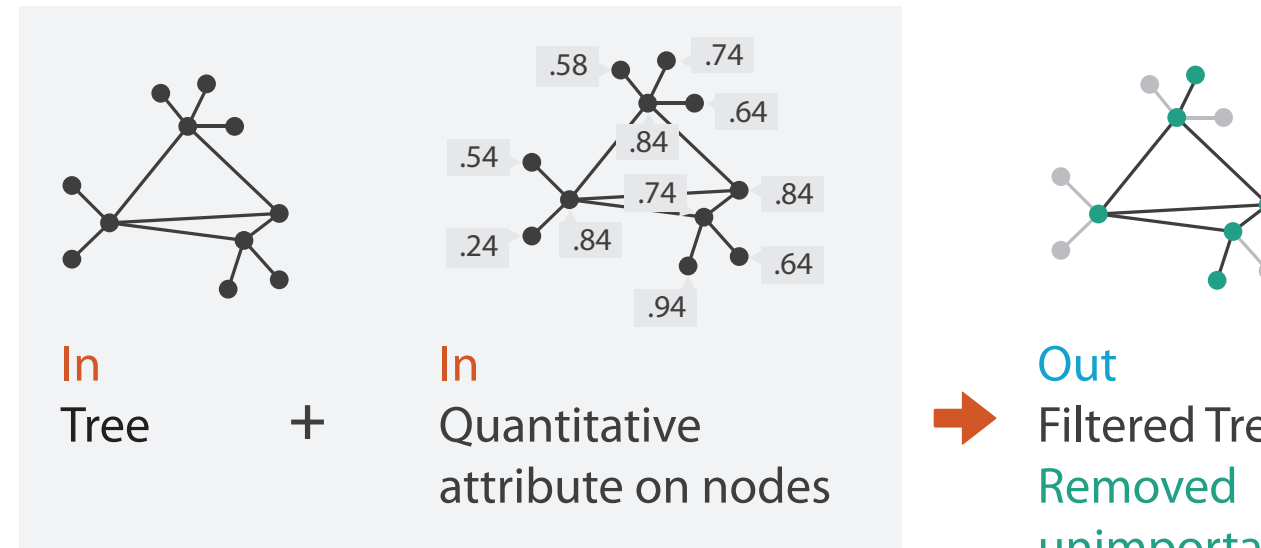
What?

- ➔ 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: Analyze, Query

- analyze

 - consume

 - discover vs present
 - aka explore vs explain

 - enjoy

 - aka casual, social

 - produce

 - annotate, record, derive

- query

 - how much data matters?

 - one, some, all

- independent choices

 - analyze, query, (search)

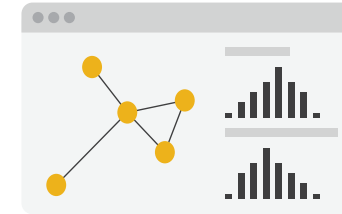
→ Analyze

→ Consume

→ Discover



→ Present

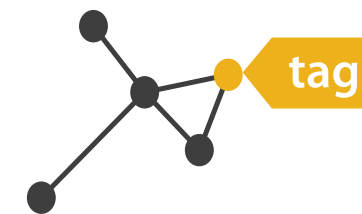


→ Enjoy



→ Produce

→ Annotate



→ Record

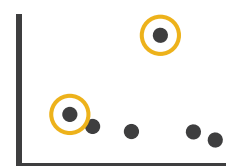


→ Derive

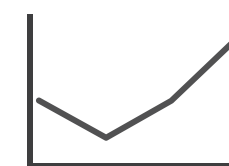


→ Query

→ Identify



→ Compare



→ Summarize



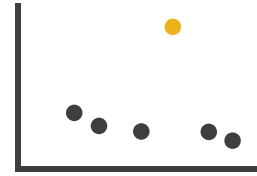
Why: Targets

→ All Data

→ Trends



→ Outliers



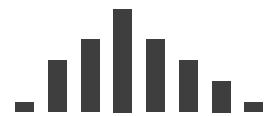
→ Features



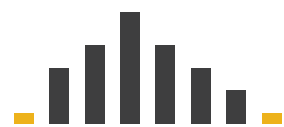
→ Attributes

→ One

→ *Distribution*



→ *Extremes*

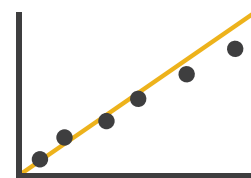


→ Many

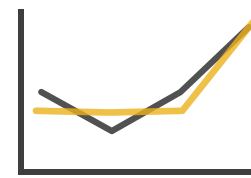
→ *Dependency*



→ *Correlation*

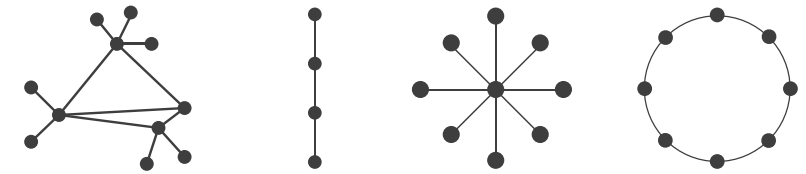


→ *Similarity*



→ Network Data

→ Topology

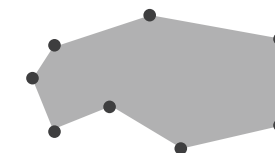


→ *Paths*



→ 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



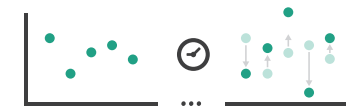
→ Motion

Direction, Rate, Frequency, ...

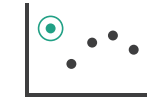


Manipulate

→ Change



→ Select



→ Navigate

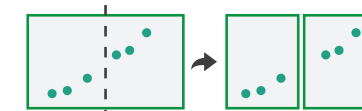


Facet

→ Juxtapose



→ Partition



→ Superimpose



Reduce

→ Filter



→ Aggregate



→ Embed



What?

Why?

How?

How to encode: Arrange space, map channels

Encode

➔ Arrange

➔ Express



➔ Separate



➔ Order



➔ Align



➔ Use



➔ Map

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

➔ Color

➔ Hue



➔ Saturation



➔ Luminance



➔ Size, Angle, Curvature, ...



➔ Shape



➔ Motion

Direction, Rate, Frequency, ...



Definitions: Marks and channels

- marks

- geometric primitives

→ Points



→ Lines



→ Areas



- channels

- control appearance of marks

- can redundantly code with multiple channels

→ Position

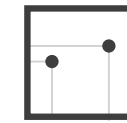
→ Horizontal



→ Vertical



→ Both



→ Color



→ Shape



→ Tilt



→ Size

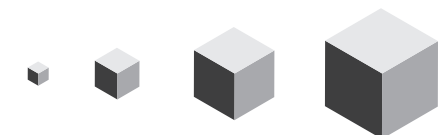
→ Length



→ Area

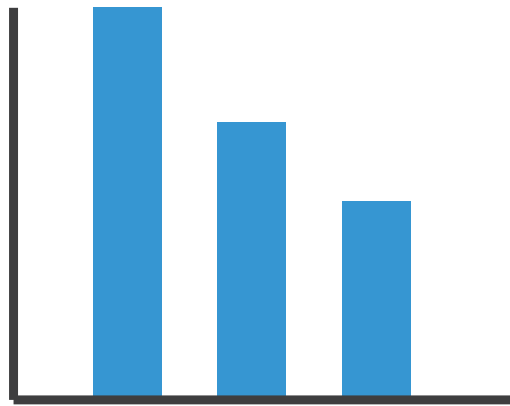


→ Volume



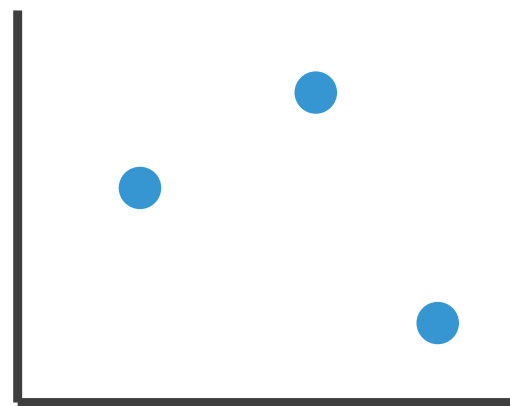
Visual encoding

- analyze idiom structure
 - as combination of marks and channels



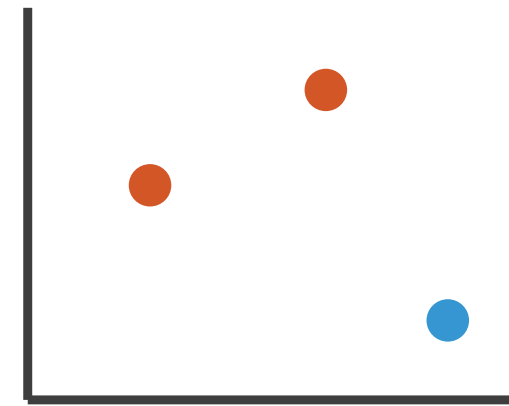
1:
vertical position

mark: line



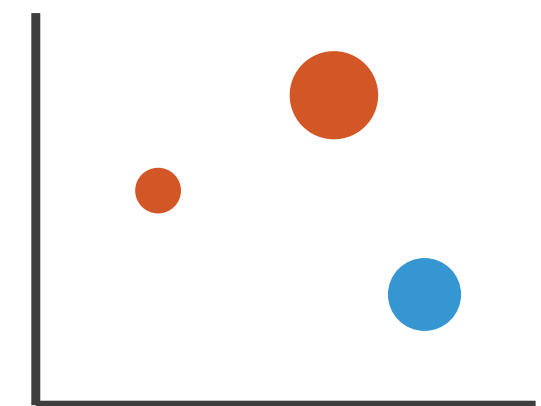
2:
vertical position
horizontal position

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



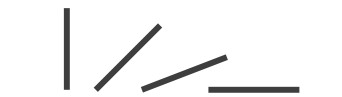
Position on unaligned scale



Length (1D size)



Tilt/angle



Area (2D size)



Depth (3D position)



Color luminance



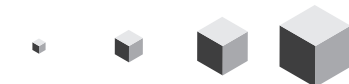
Color saturation



Curvature



Volume (3D size)



Same

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 

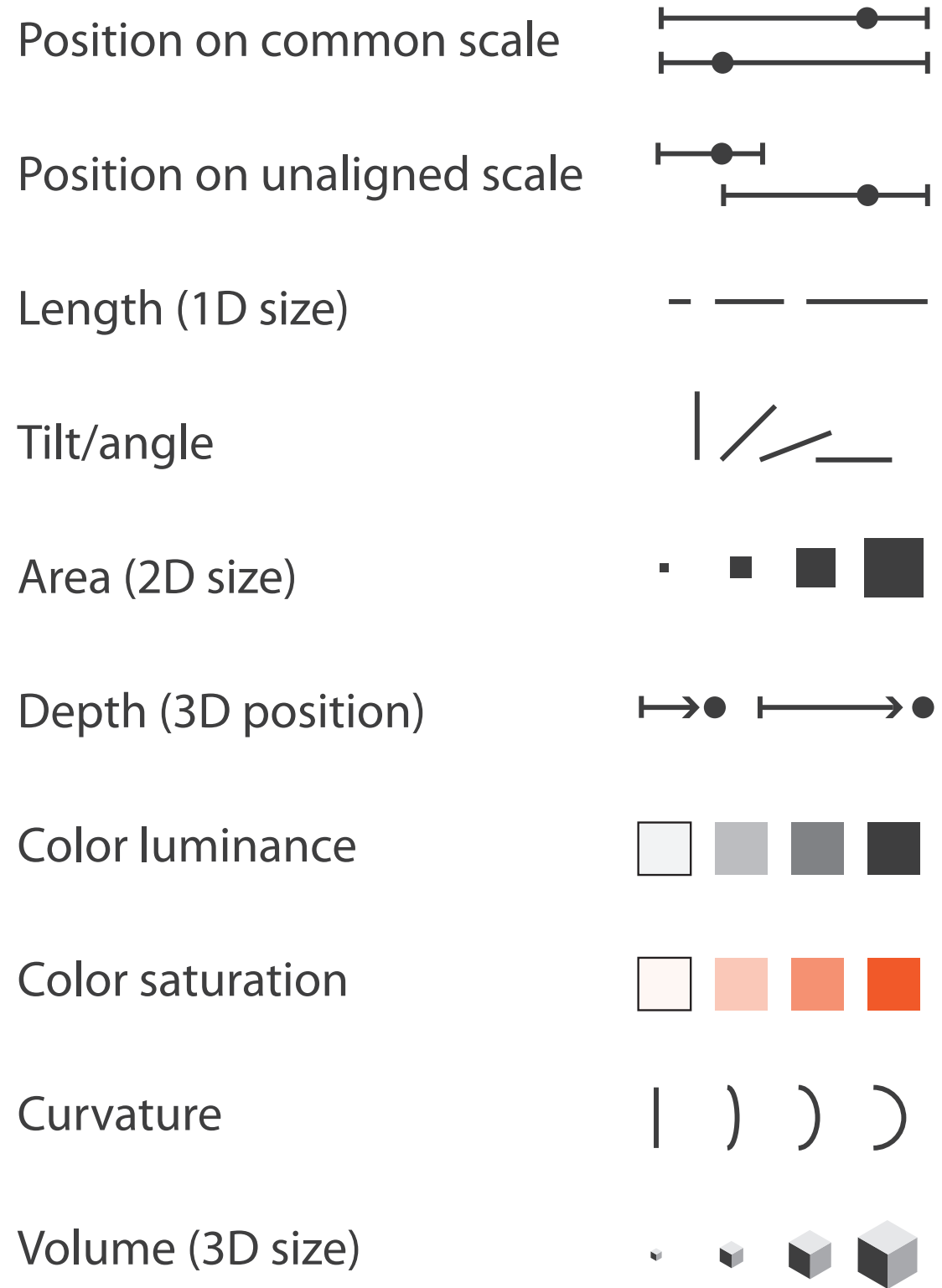
Motion 

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

Channels: Expressiveness types and effectiveness rankings

➔ Magnitude Channels: Ordered Attributes

Position on common scale



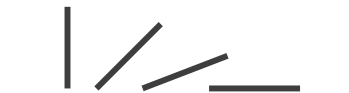
Position on unaligned scale



Length (1D size)



Tilt/angle



Area (2D size)



Depth (3D position)



Color luminance



Color saturation



Curvature



Volume (3D size)



Same

➔ Identity Channels: Categorical Attributes

Spatial region



Color hue



Motion



Shape



Best

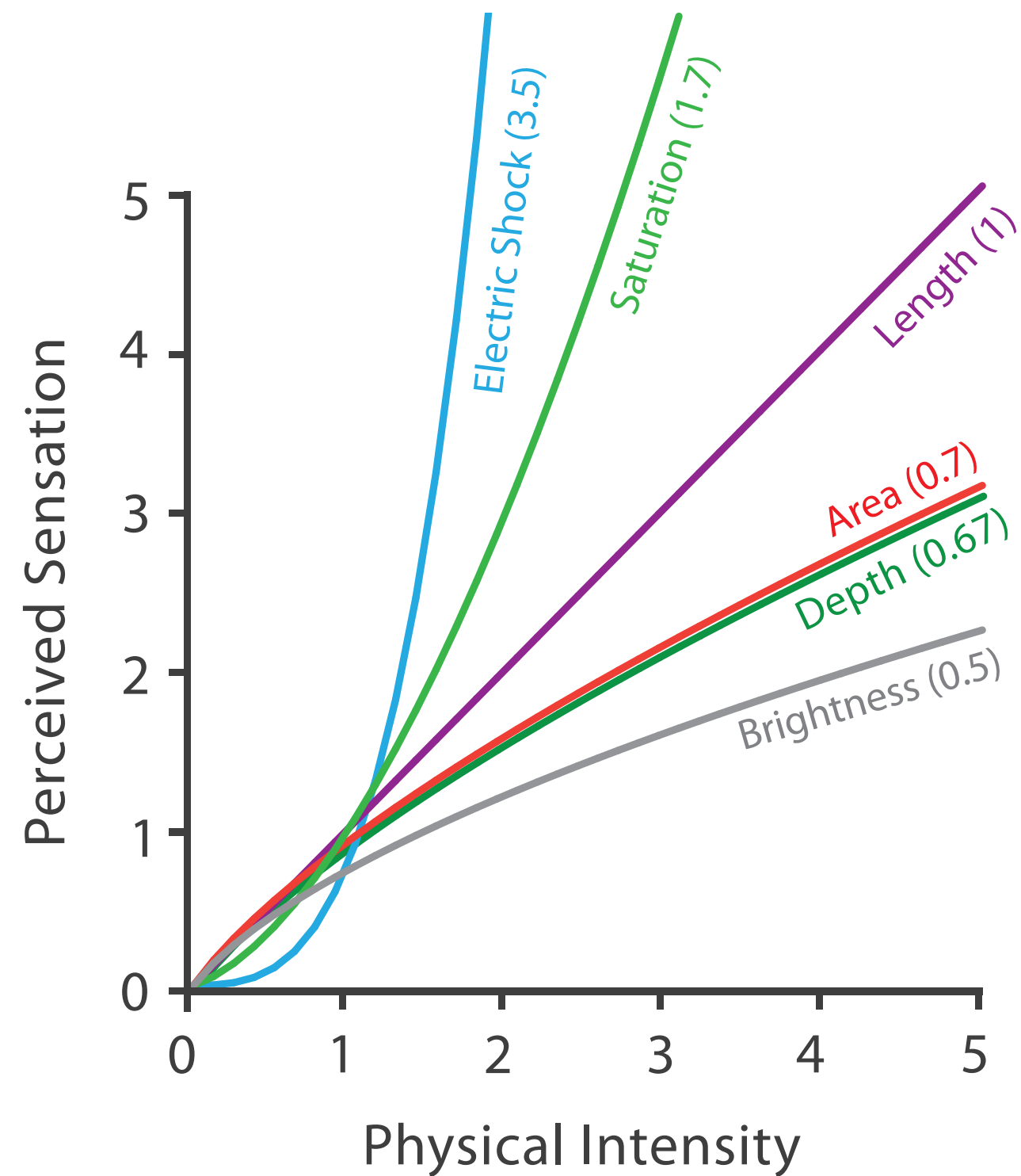
Effectiveness

Least

- **expressiveness principle**
 - match channel and data characteristics
- **effectiveness principle**
 - encode most important attributes with highest ranked channels
 - spatial position ranks high for both

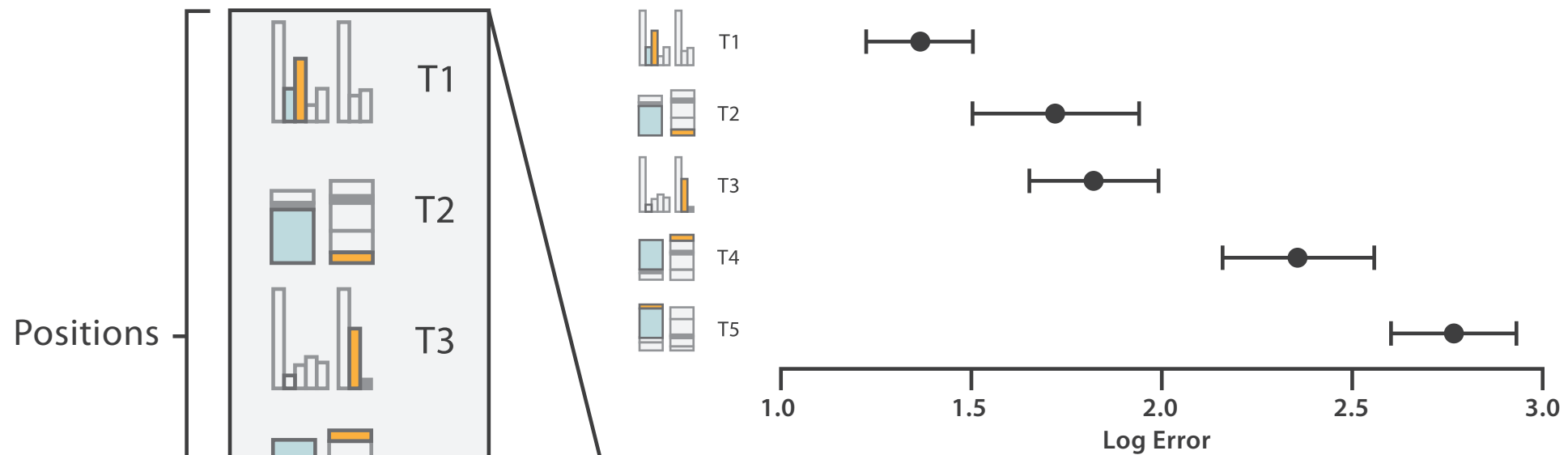
Accuracy: Fundamental Theory

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

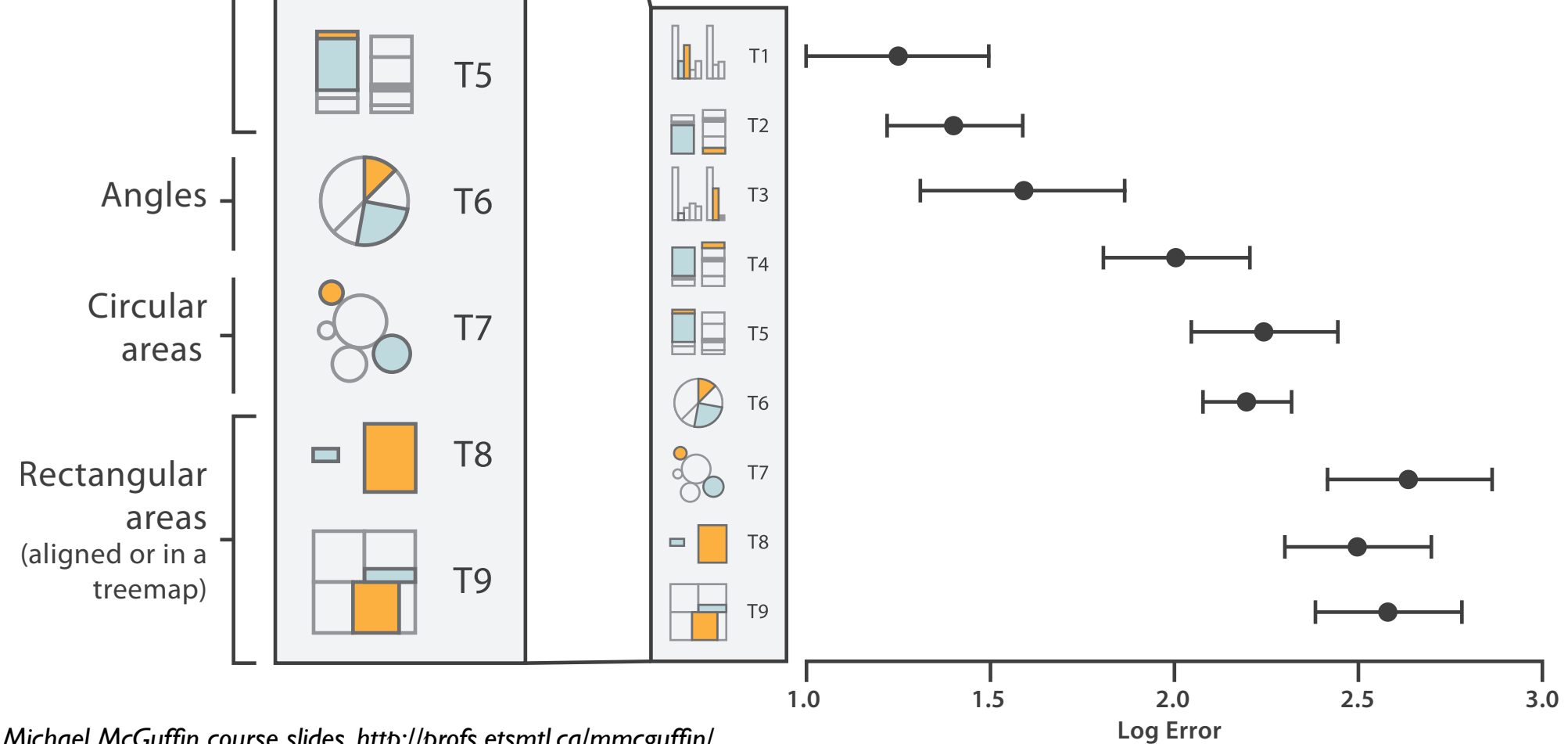


Accuracy: Vis experiments

Cleveland & McGill's Results



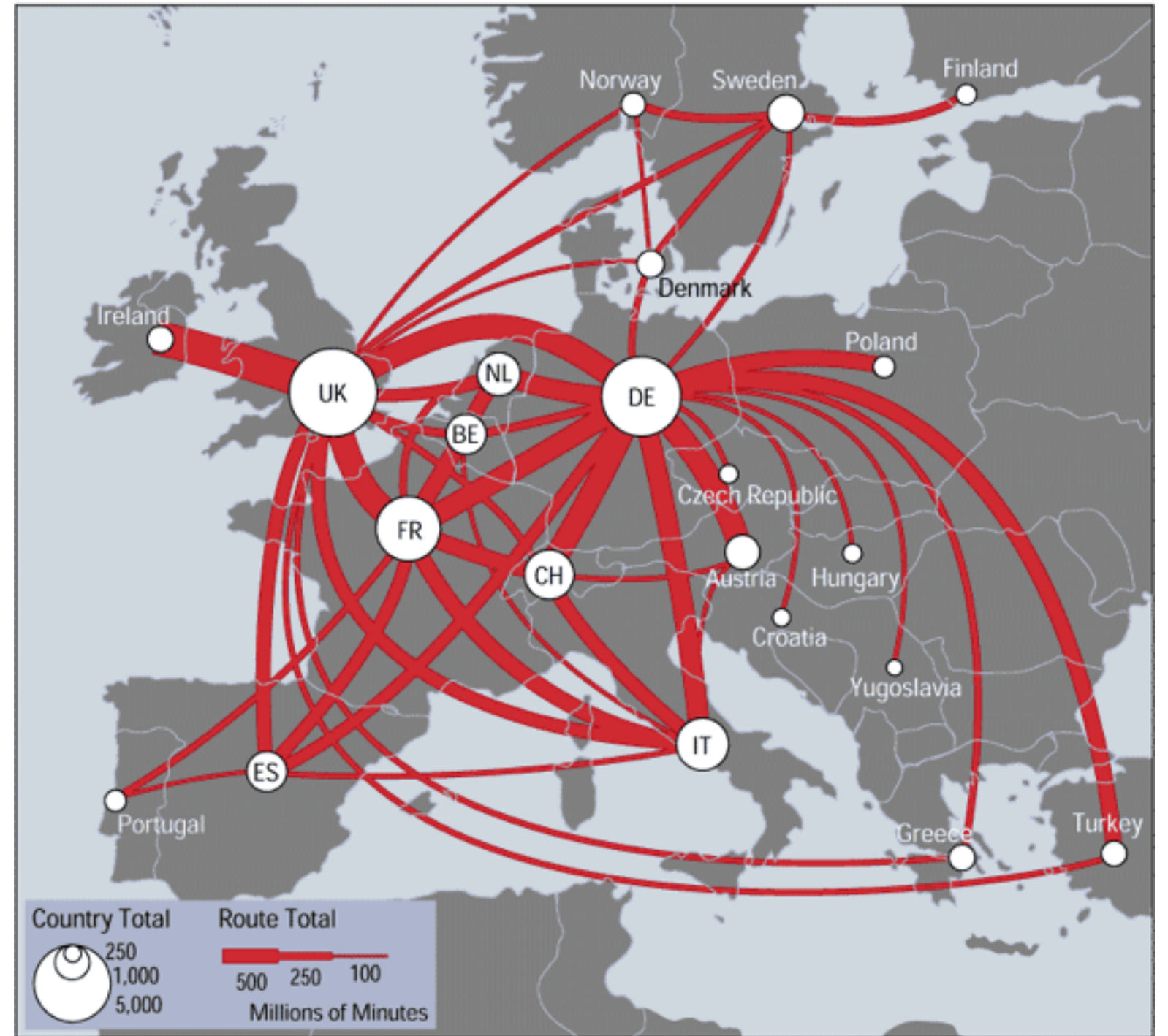
Crowdsourced Results



[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.]

Discriminability: How many usable steps?

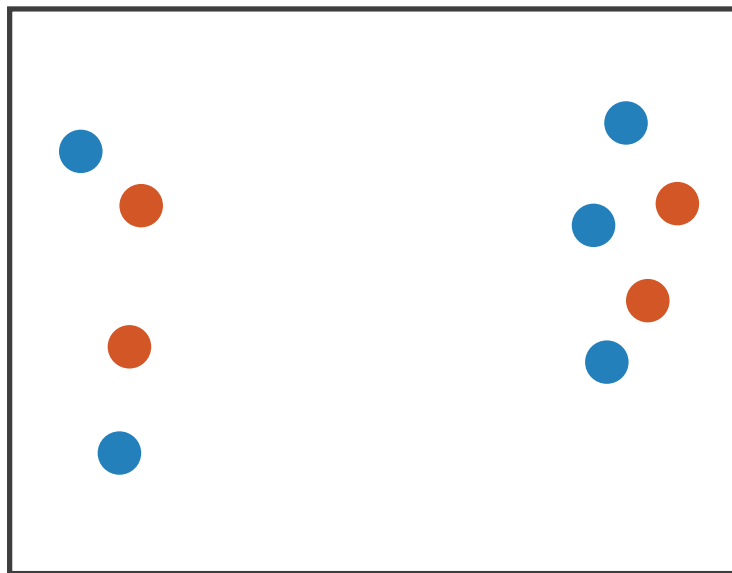
- must be sufficient for number of attribute levels to show
 - linewidth: few bins but salient



[mappa.mundi.net/maps/maps_014/telegeography.html]

Separability vs. Integrality

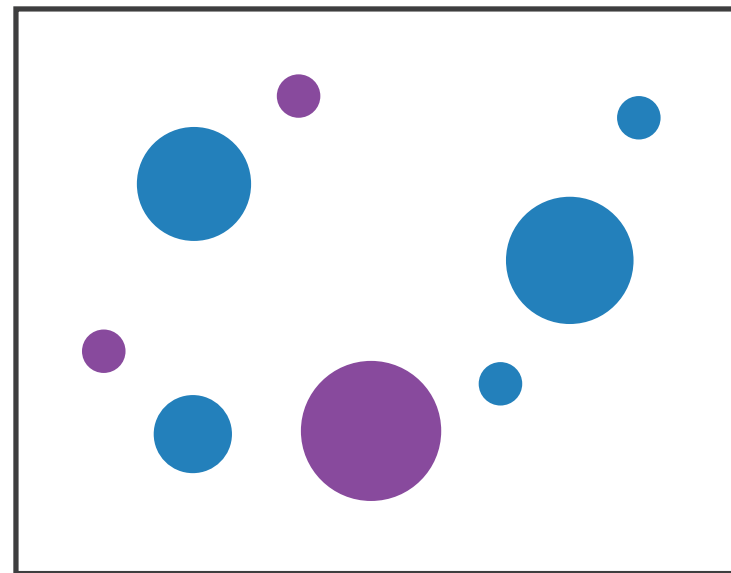
Position
+ Hue (Color)



Fully separable

2 groups each

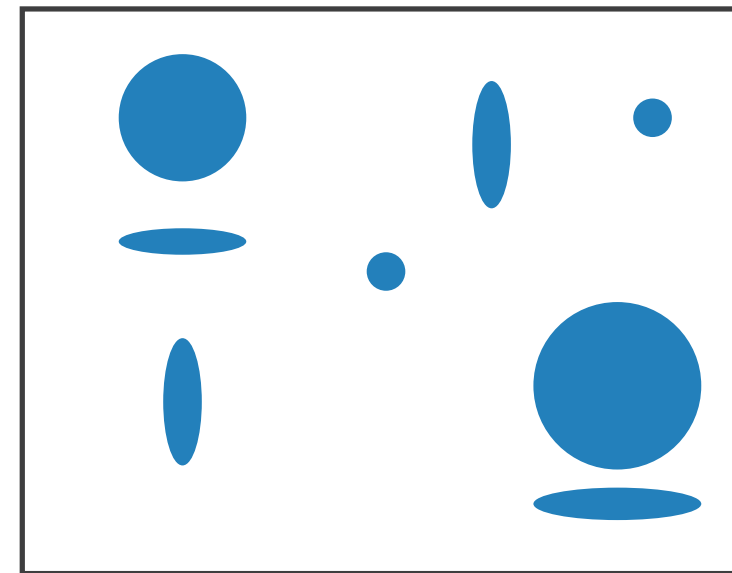
Size
+ Hue (Color)



Some interference

2 groups each

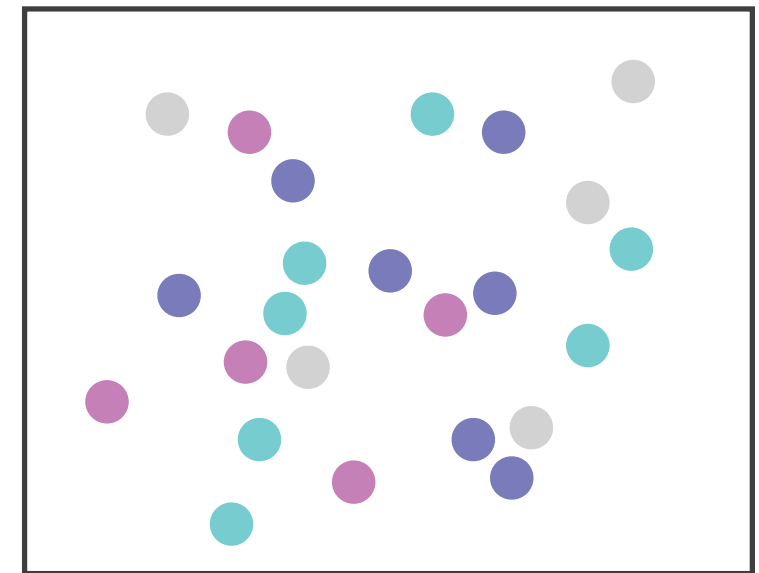
Width
+ Height



Some/significant
interference

3 groups total:
integral area

Red
+ Green



Major interference

4 groups total:
integral hue

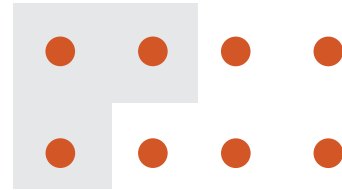
Grouping

- containment
- connection

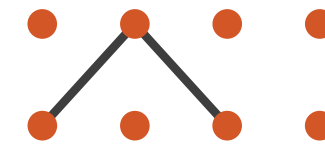
- proximity
 - same spatial region
- similarity
 - same values as other categorical channels

Marks as Links

➔ Containment



➔ Connection



➔ Identity Channels: Categorical Attributes

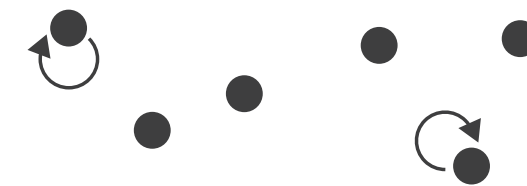
Spatial region



Color hue



Motion

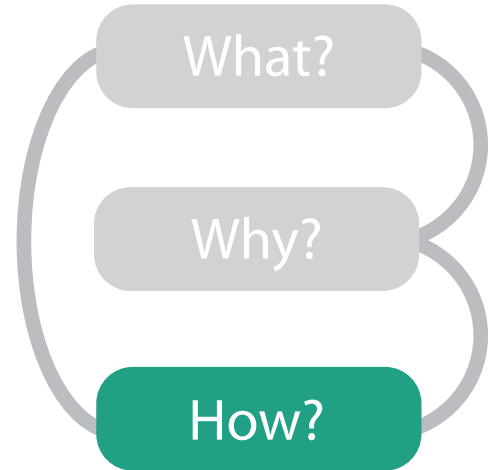


Shape



Idiom design choices: Encode

Encode



→ Arrange

→ Express



→ Separate



→ Order



→ Align



→ Use



→ Map

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

→ Color

→ Hue



→ Saturation



→ Luminance



→ Size, Angle, Curvature, ...



→ Shape

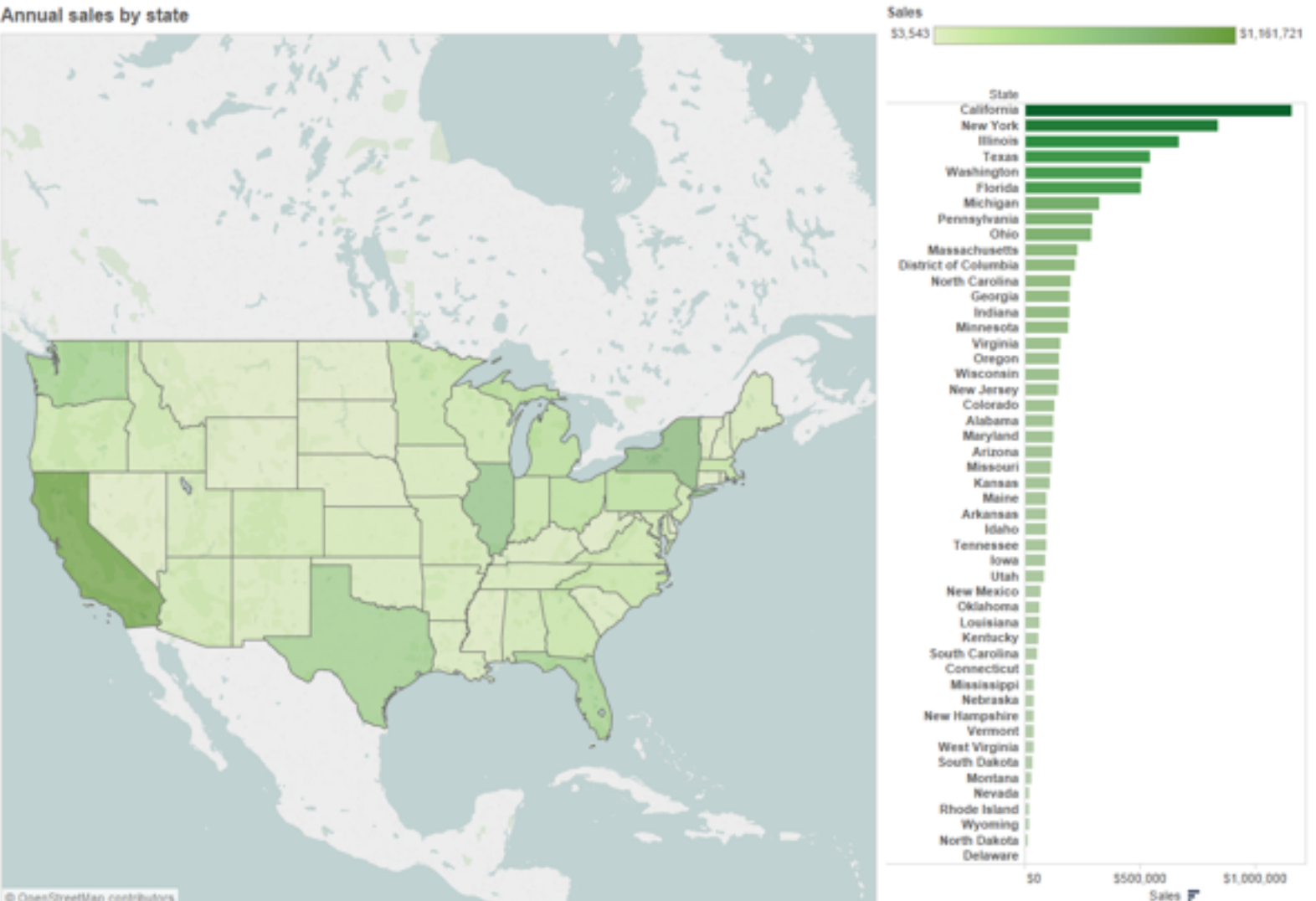
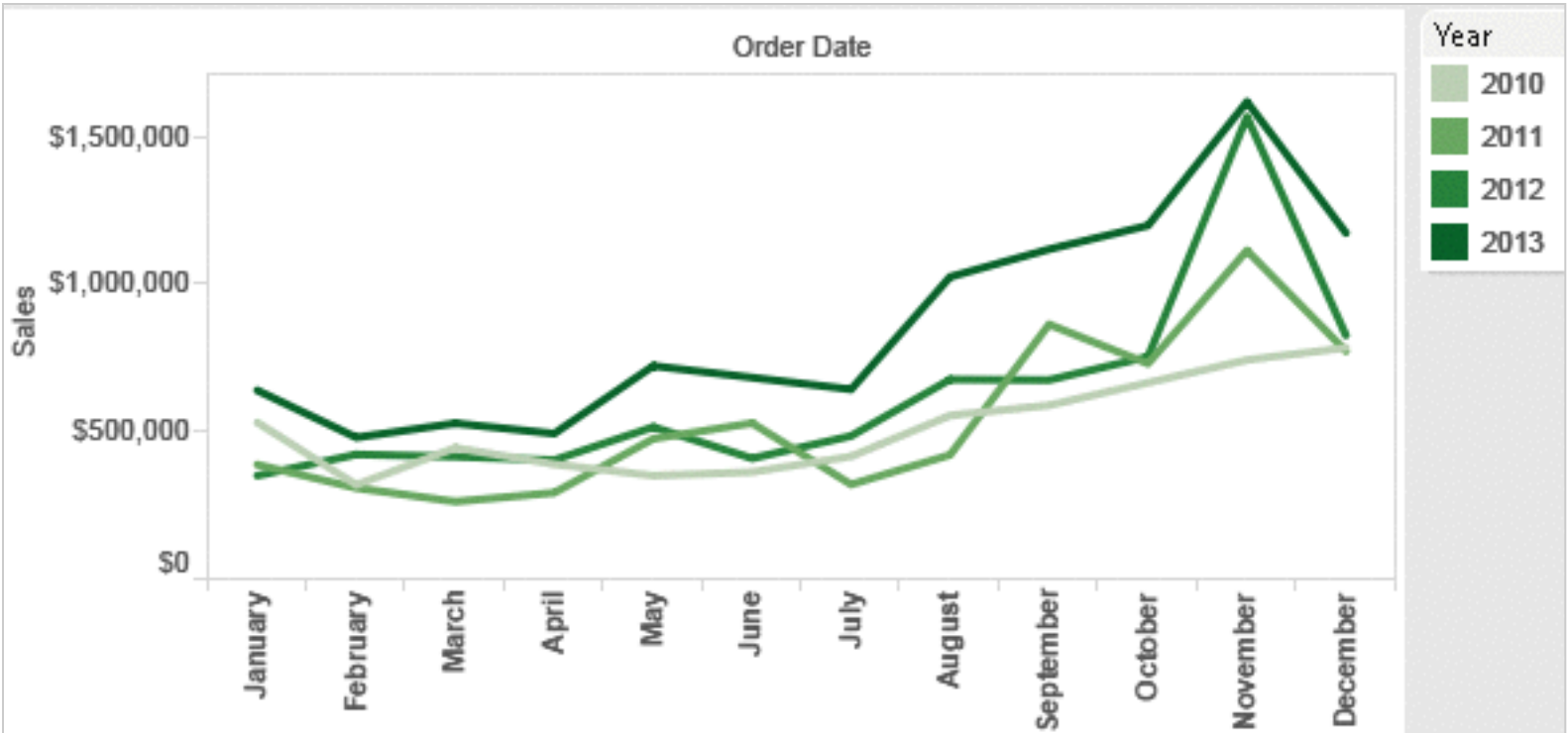
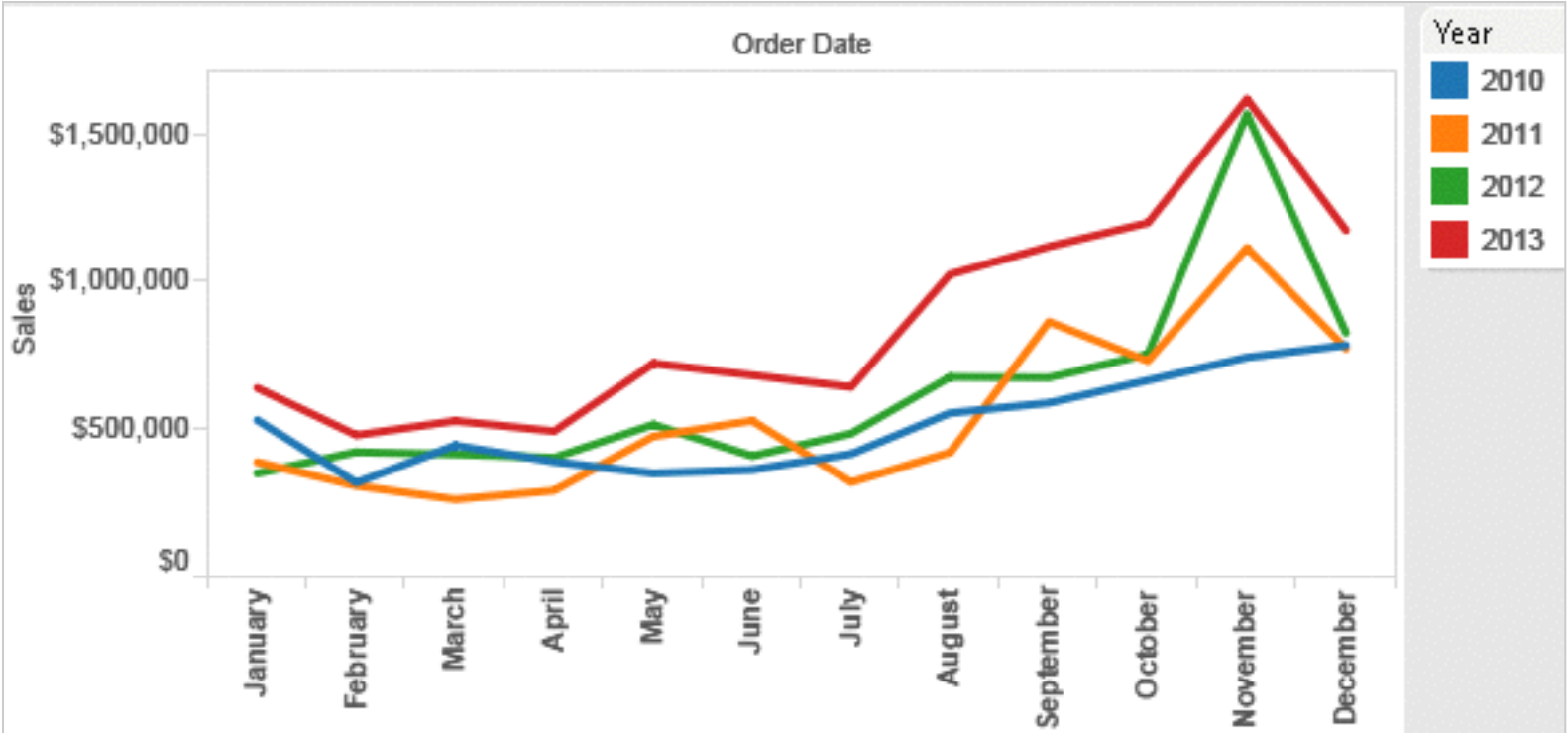


→ Motion

Direction, Rate, Frequency, ...



Categorical vs ordered color



[Seriously Colorful: Advanced Color Principles & Practices. Stone.Tableau Customer Conference 2014.]

Decomposing color

- first rule of color: do not talk about color!
 - color is confusing if treated as monolithic

- decompose into three channels

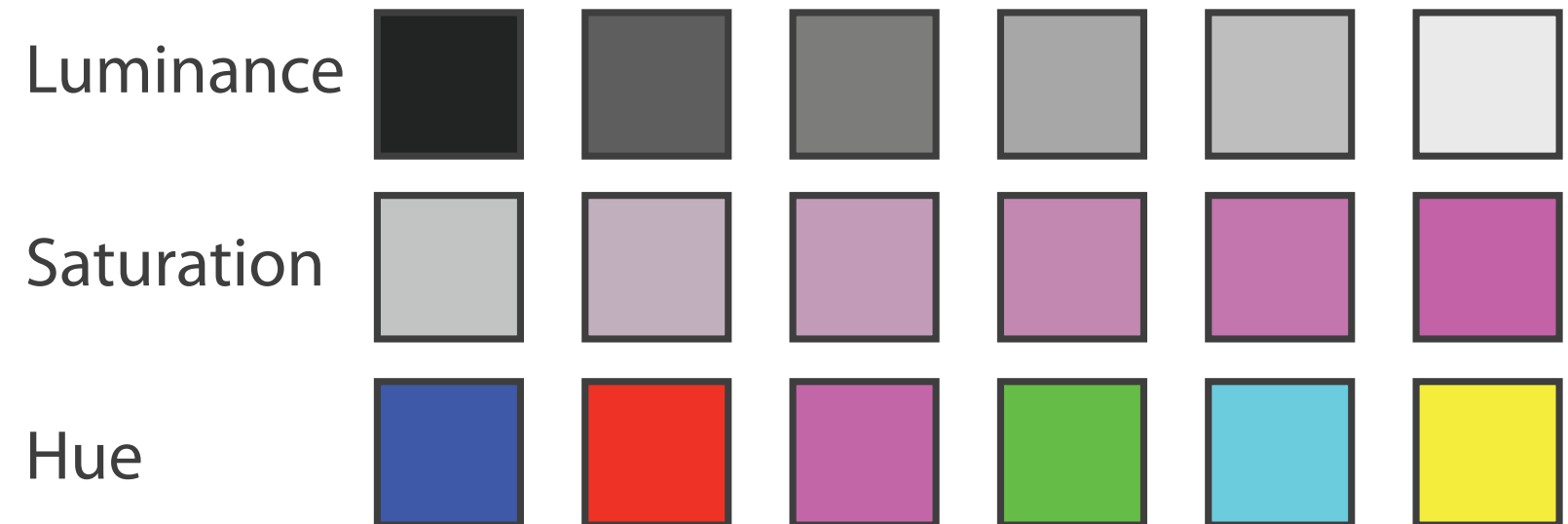
- ordered can show magnitude

- luminance: how bright

- saturation: how colorful

- categorical can show identity

- hue: what color



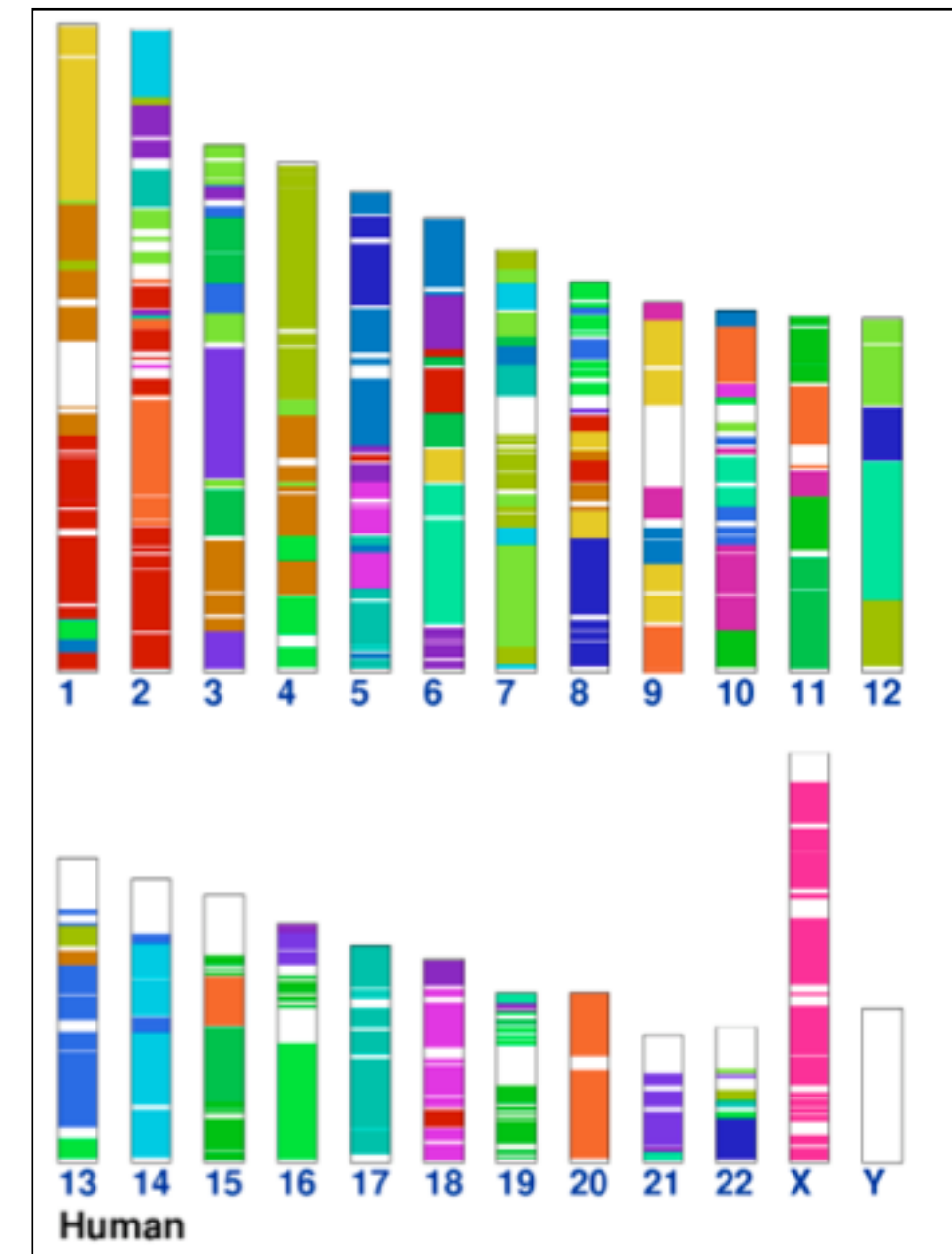
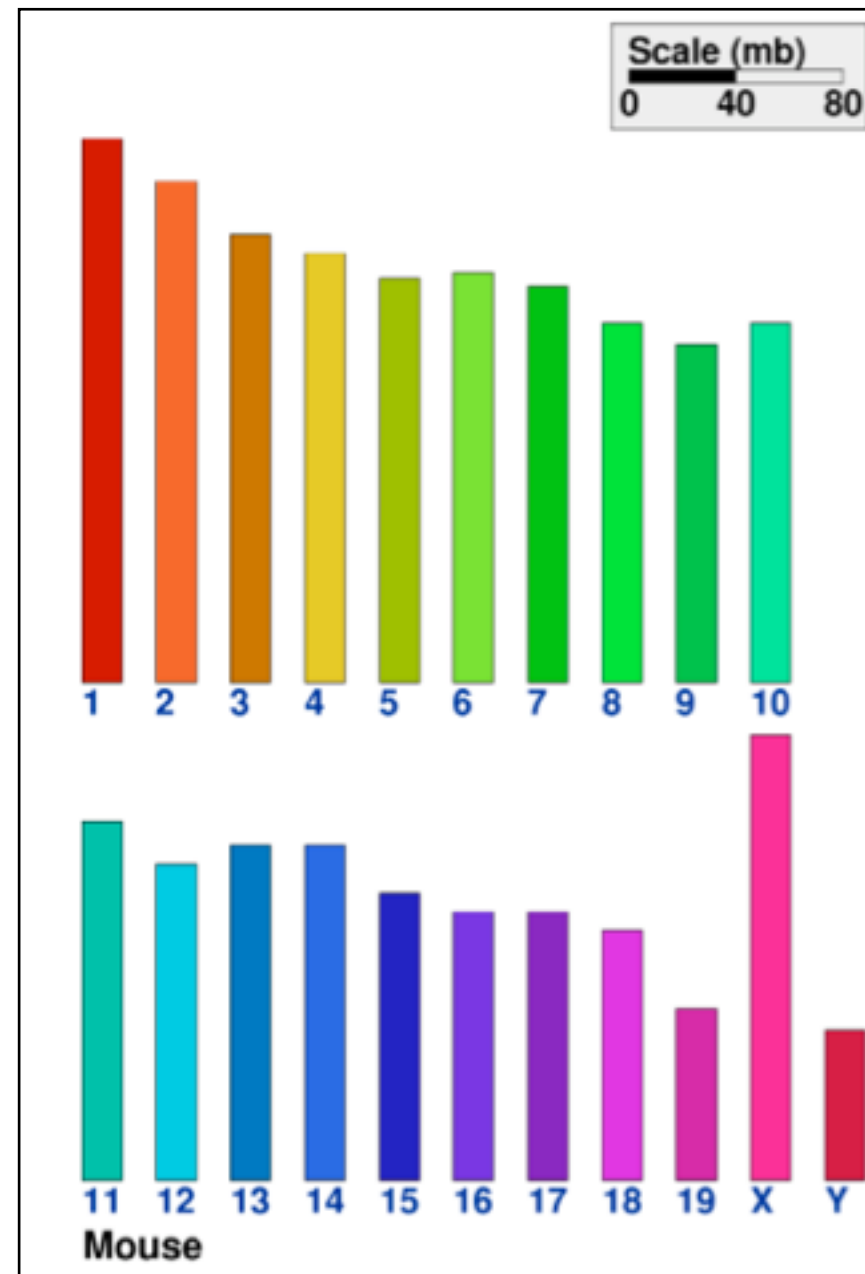
- channels have different properties

- what they convey directly to perceptual system

- how much they can convey: how many discriminable bins can we use?

Categorical color: limited number of discriminable bins

- human perception built on relative comparisons
 - great if color contiguous
 - surprisingly bad for absolute comparisons
- noncontiguous small regions of color
 - fewer bins than you want
 - rule of thumb: 6-12 bins, including background and highlights

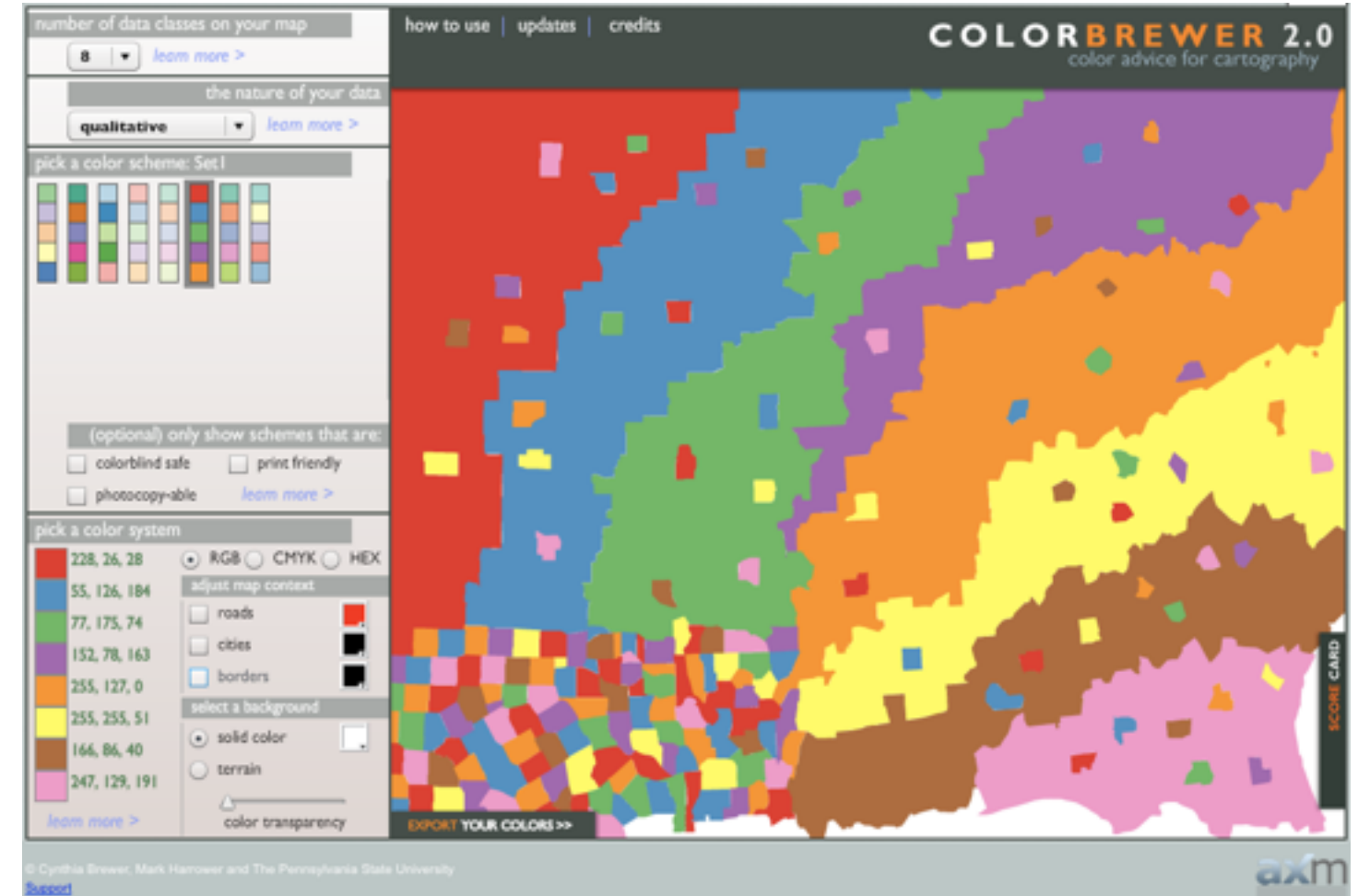
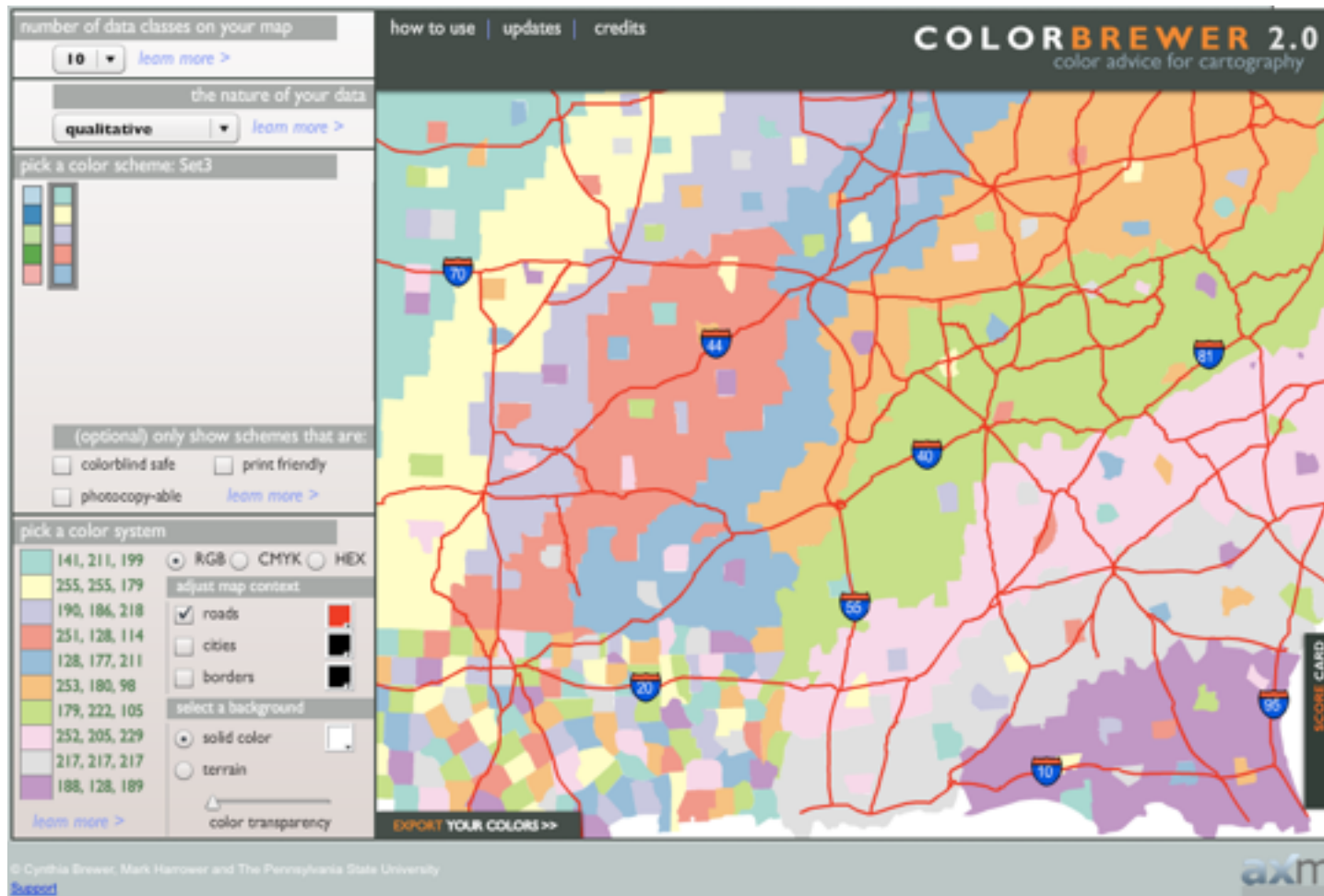


–alternatives? this afternoon!

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

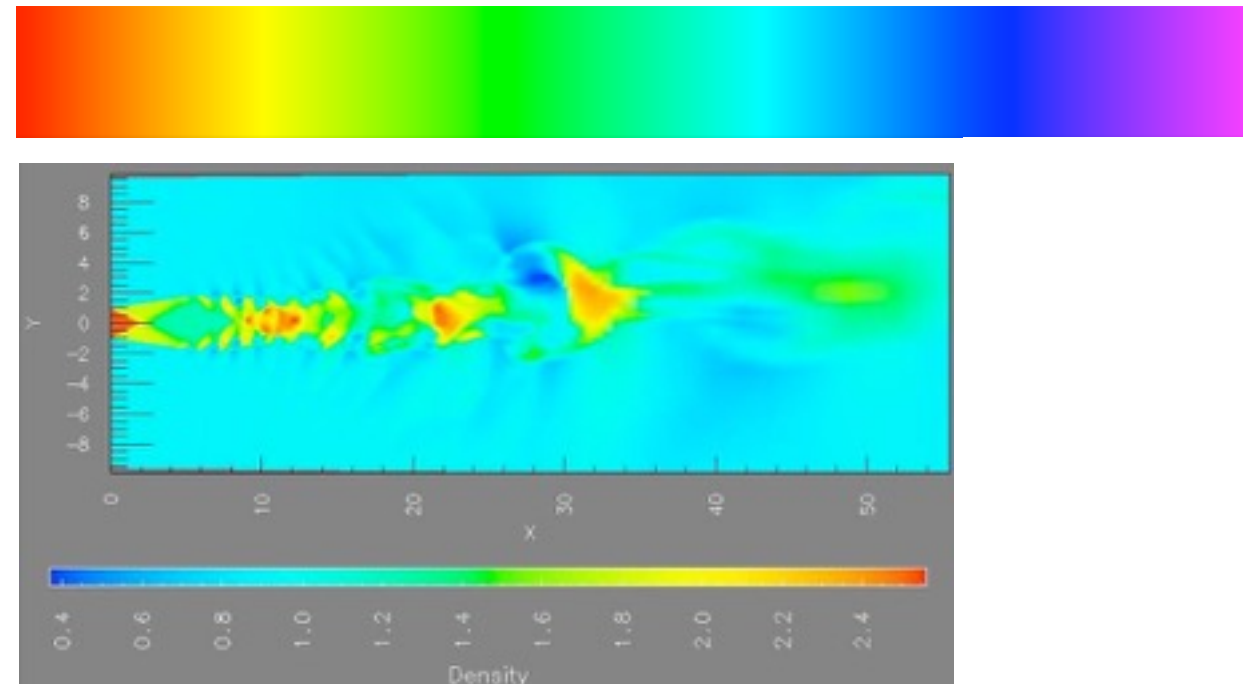
ColorBrewer

- <http://www.colorbrewer2.org>
- saturation and area example: size affects salience!

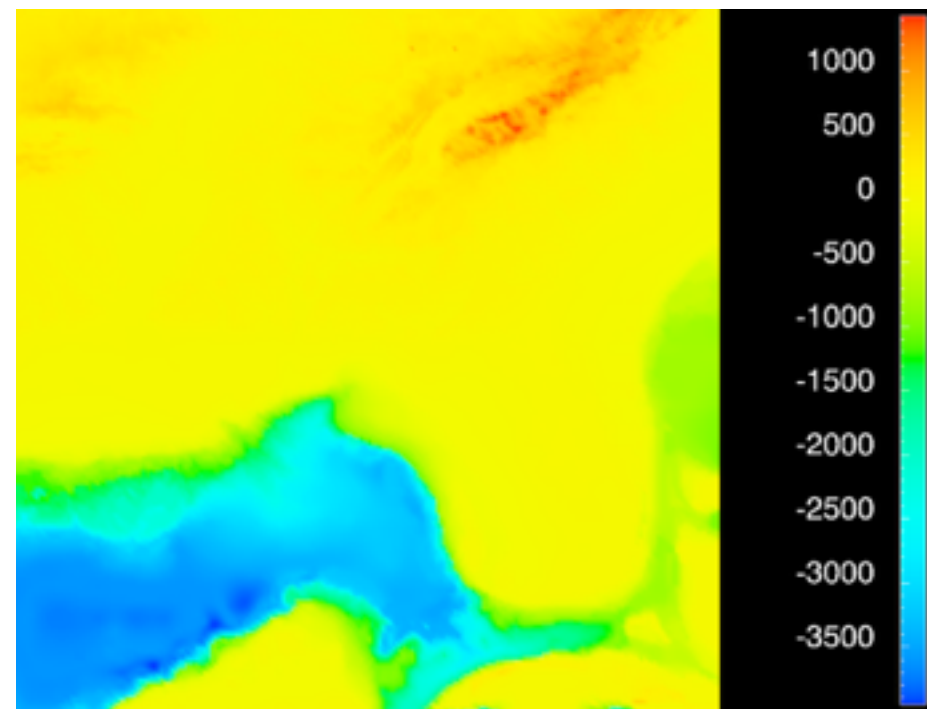


Ordered color: Rainbow is poor default

- problems
 - perceptually unordered
 - perceptually nonlinear
- benefits
 - fine-grained structure visible and nameable



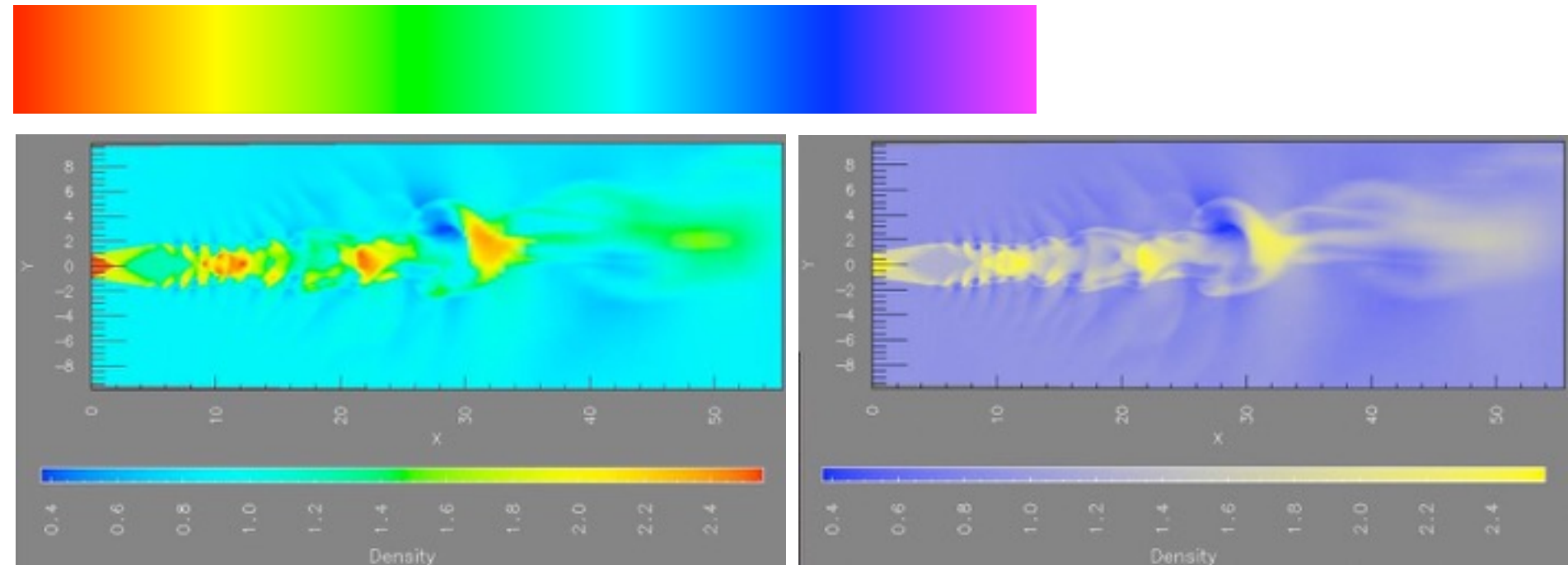
[A Rule-based Tool for Assisting Colormap Selection. Bergman, Rogowitz, and Treinish. Proc. IEEE Visualization (Vis), pp. 118–125, 1995.]



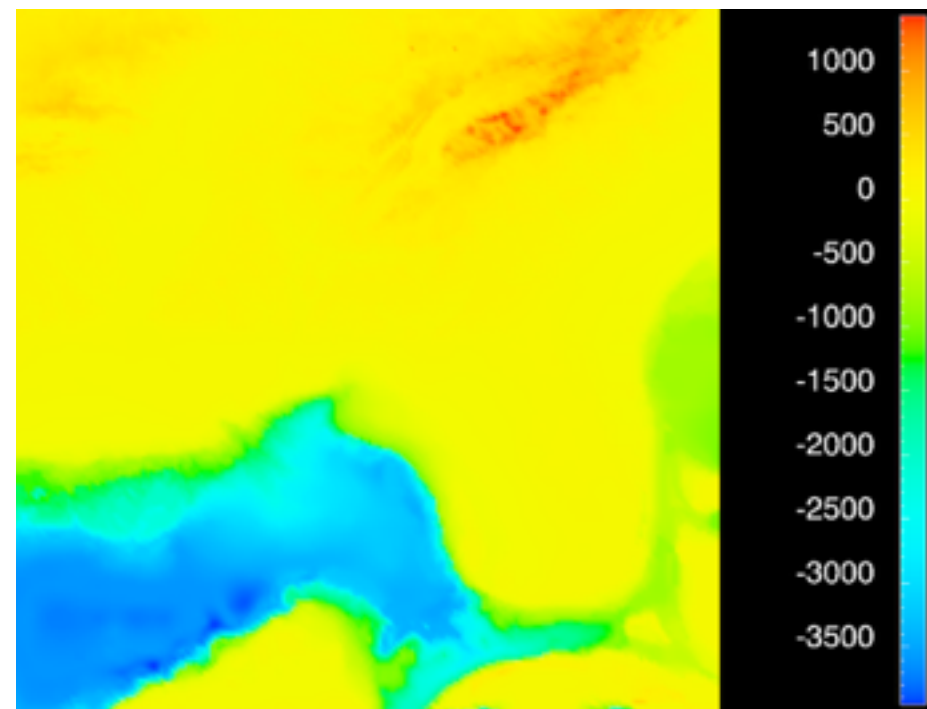
[Why Should Engineers Be Worried About Color? Treinish and Rogowitz 1998. <http://www.research.ibm.com/people/lloyd/color/color.HTM>]

Ordered color: Rainbow is poor default

- problems
 - perceptually unordered
 - perceptually nonlinear
- benefits
 - fine-grained structure visible and nameable
- alternatives
 - large-scale structure: fewer hues



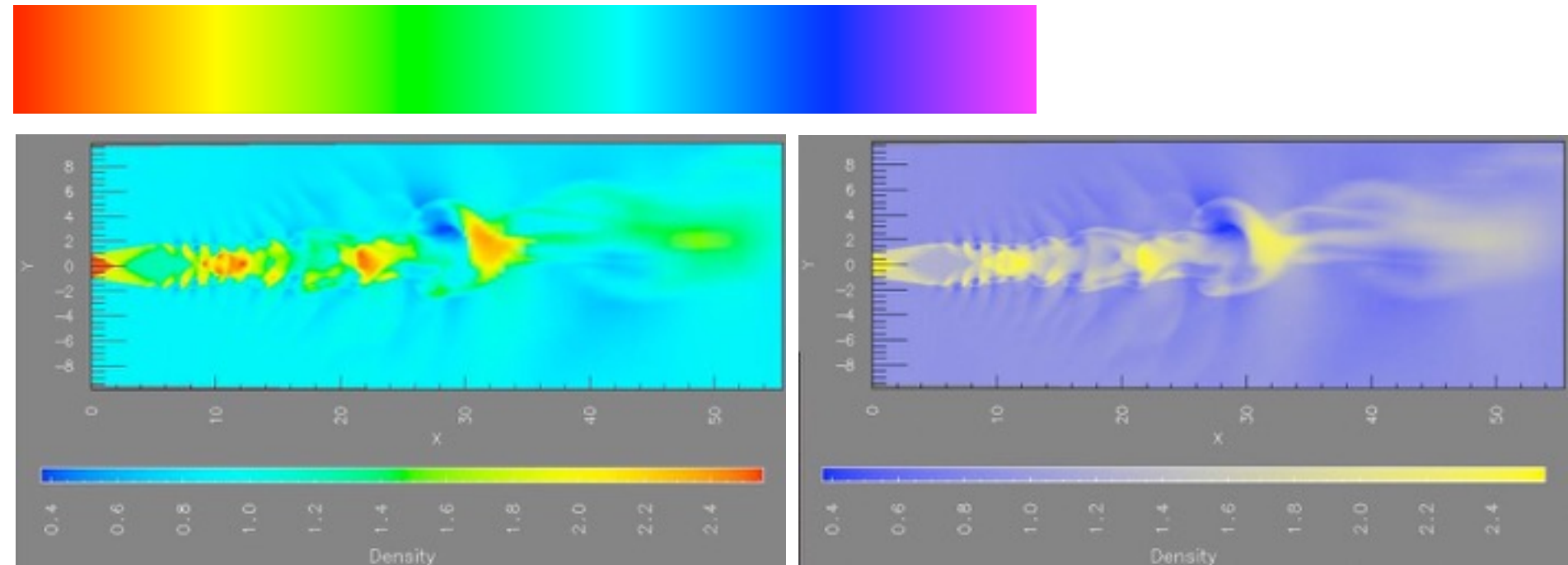
[A Rule-based Tool for Assisting Colormap Selection. Bergman, Rogowitz, and Treinish. Proc. IEEE Visualization (Vis), pp. 118–125, 1995.]



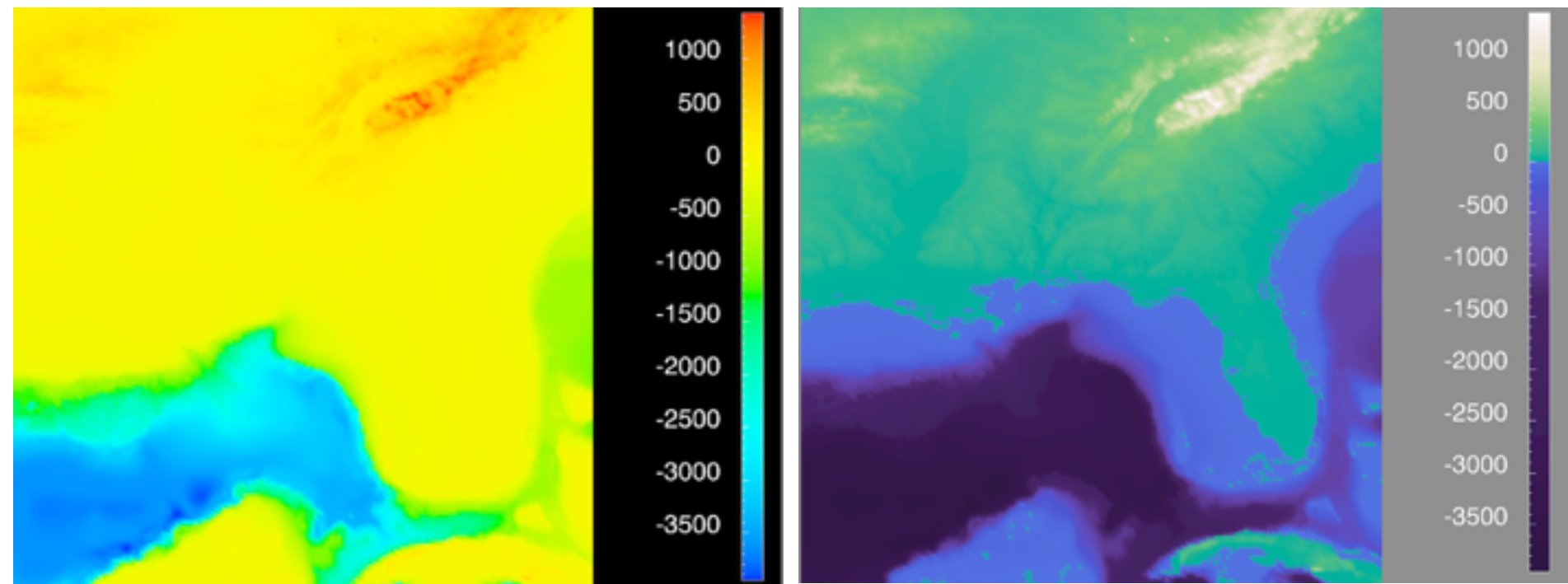
[Why Should Engineers Be Worried About Color? Treinish and Rogowitz 1998. <http://www.research.ibm.com/people/lloyd/color/color.HTM>]

Ordered color: Rainbow is poor default

- problems
 - perceptually unordered
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- benefits
 - fine-grained structure visible and nameable
- alternatives
 - large-scale structure: fewer hues
 - fine structure: multiple hues with monotonically increasing luminance [eg viridis R/python]



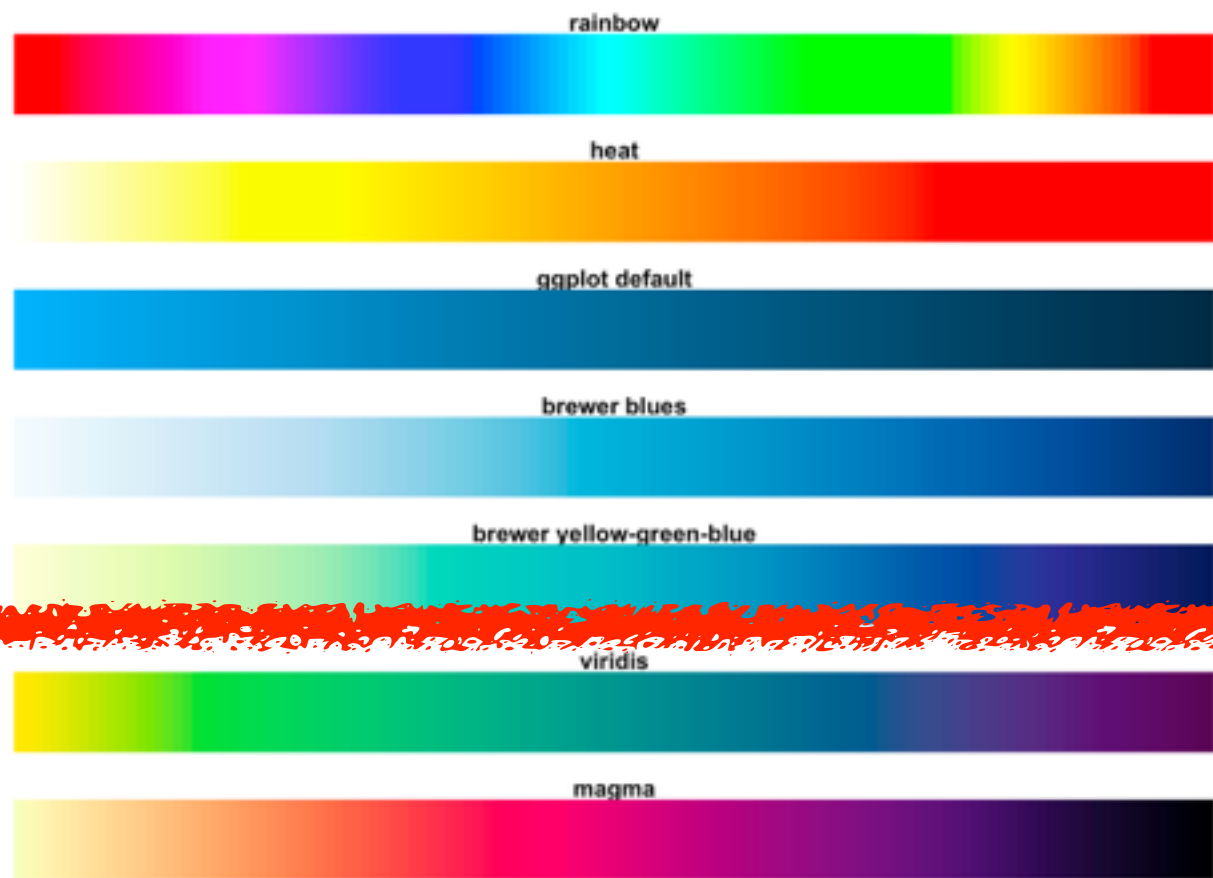
[A Rule-based Tool for Assisting Colormap Selection. Bergman, Rogowitz, and Treinish. Proc. IEEE Visualization (Vis), pp. 118–125, 1995.]



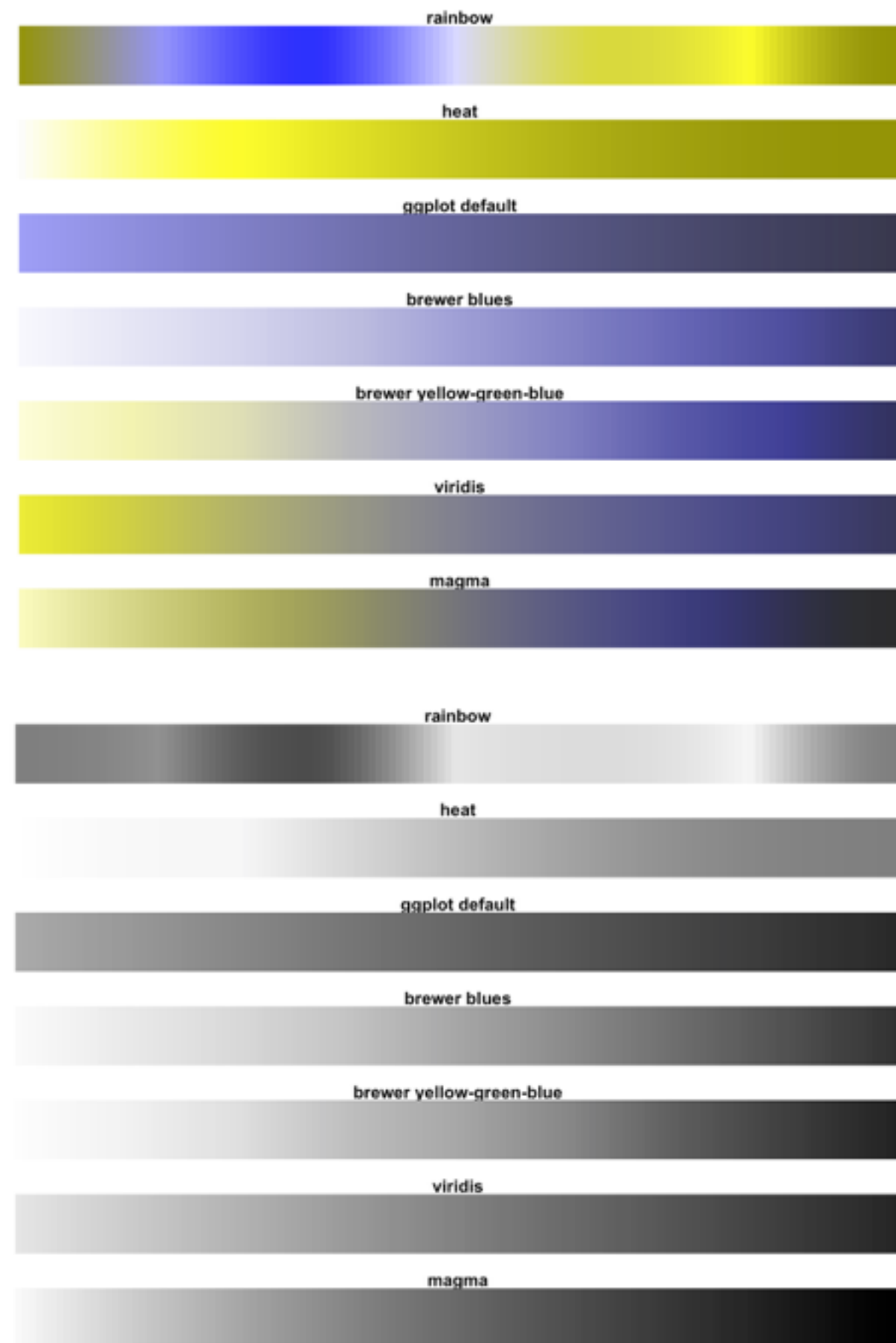
[Why Should Engineers Be Worried About Color? Treinish and Rogowitz 1998. <http://www.research.ibm.com/people/lloyd/color/color.HTM>]

Viridis

- colorful, perceptually uniform, colorblind-safe, monotonically increasing luminance

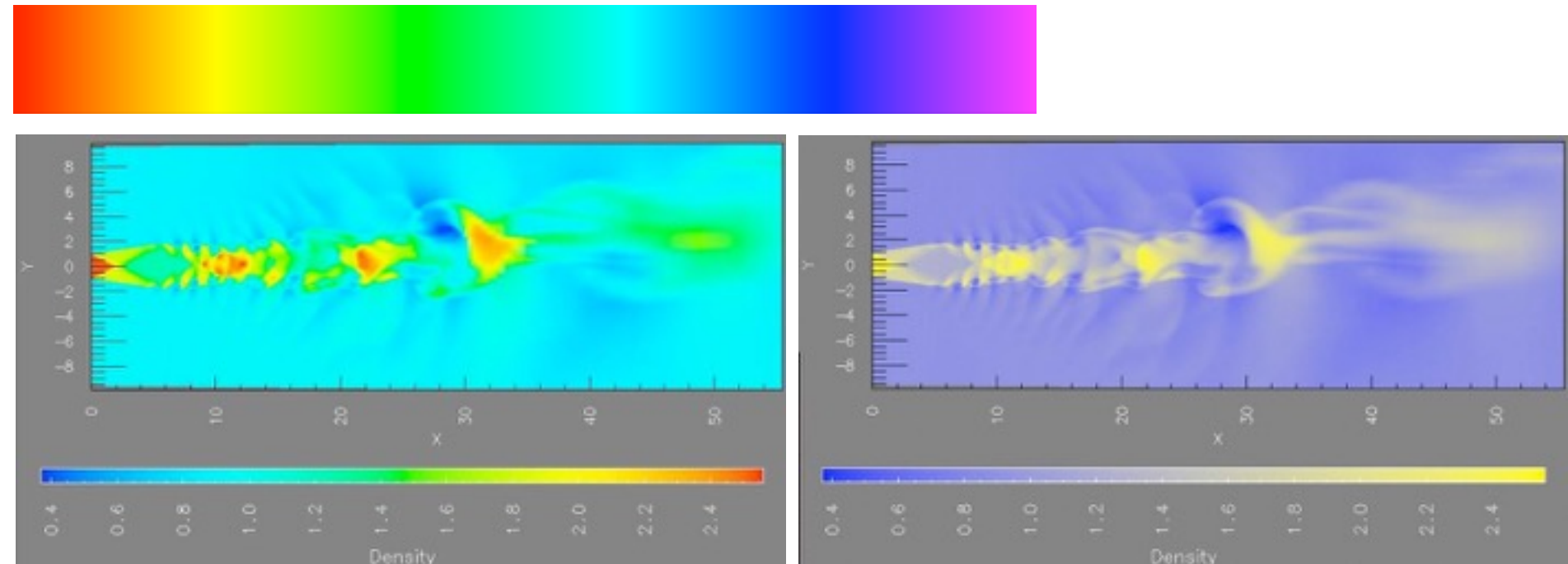


<https://cran.r-project.org/web/packages/viridis/vignettes/intro-to-viridis.html>

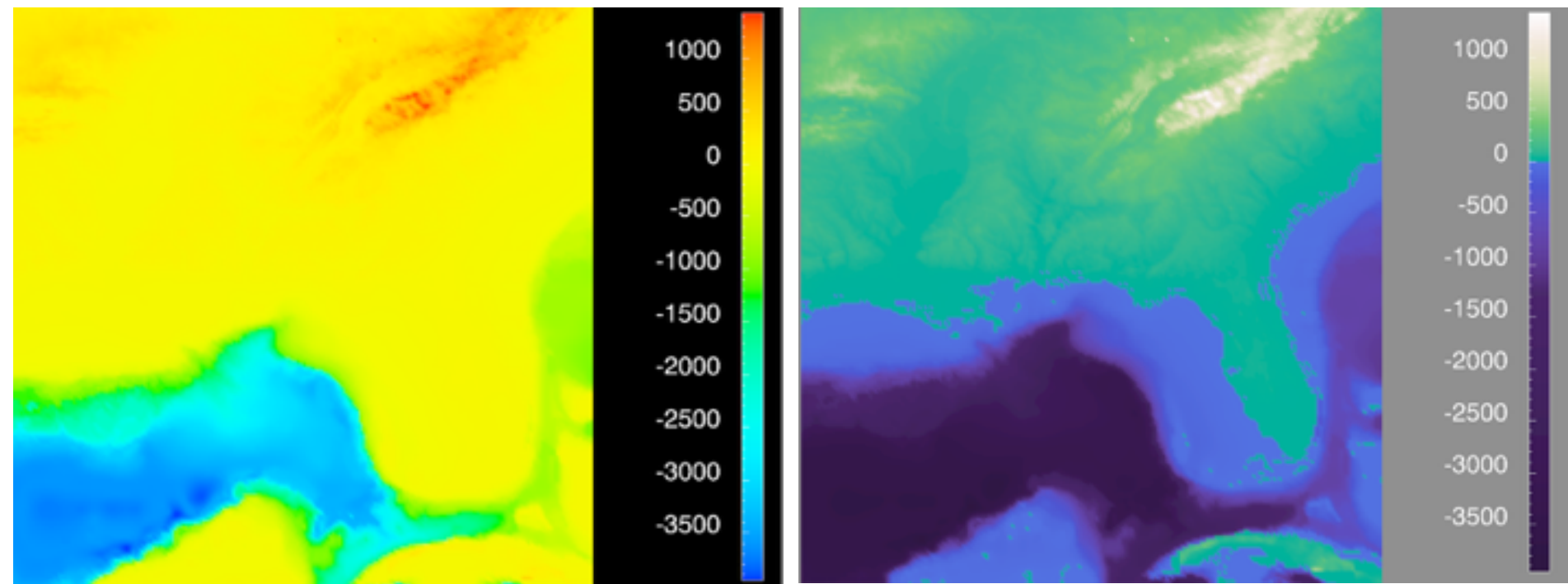


Ordered color: Rainbow is poor default

- problems
 - perceptually unordered
 - perceptually nonlinear
- benefits
 - fine-grained structure visible and nameable
- alternatives
 - large-scale structure: fewer hues
 - fine structure: multiple hues with monotonically increasing luminance [eg viridis R/python]
 - segmented rainbows for binned or categorical



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



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



How?

Encode

→ Arrange

→ Express



→ Separate



→ Order



→ Align



→ Use



→ Map

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

→ Color

→ Hue



→ Saturation



→ Luminance



→ Size, Angle, Curvature, ...



→ Shape



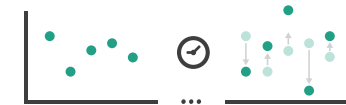
→ Motion

Direction, Rate, Frequency, ...

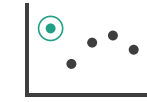


Manipulate

→ Change



→ Select



→ Navigate

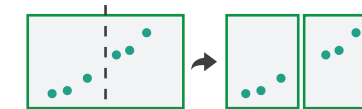


Facet

→ Juxtapose



→ Partition



→ Superimpose



Reduce

→ Filter



→ Aggregate



→ Embed



What?

Why?

How?

Encode tables: Arrange space

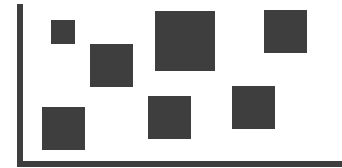
Encode

➔ Arrange

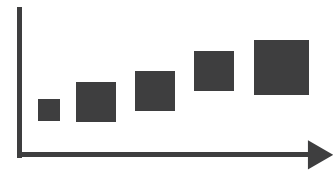
➔ Express



➔ Separate



➔ Order



➔ Align



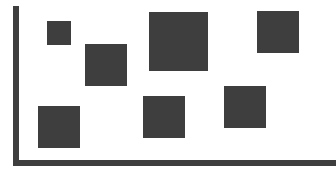
Arrange tables

→ Express Values

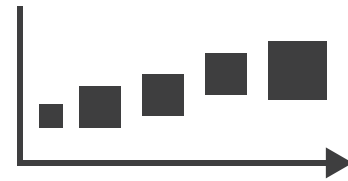


→ Separate, Order, Align Regions

→ Separate



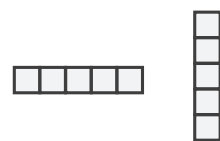
→ Order



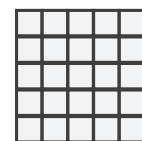
→ Align



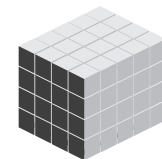
→ 1 Key *List*



→ 2 Keys *Matrix*



→ 3 Keys *Volume*

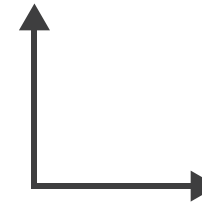


→ Many Keys *Recursive Subdivision*

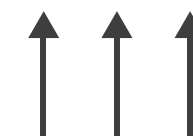


→ Axis Orientation

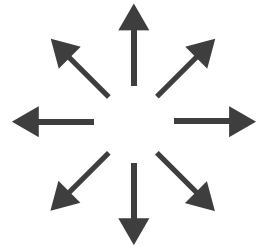
→ Rectilinear



→ Parallel

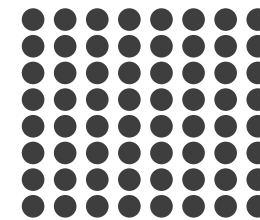


→ Radial



→ Layout Density

→ Dense



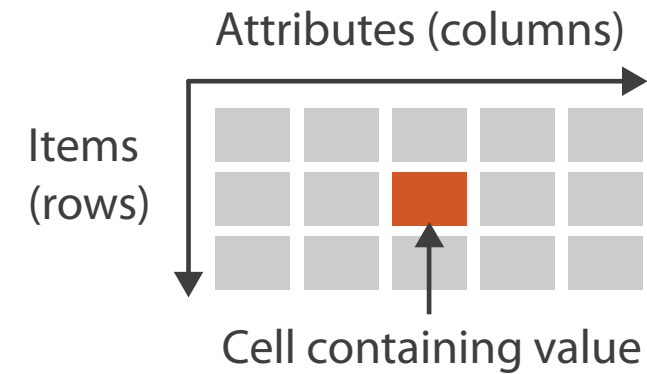
→ Space-Filling



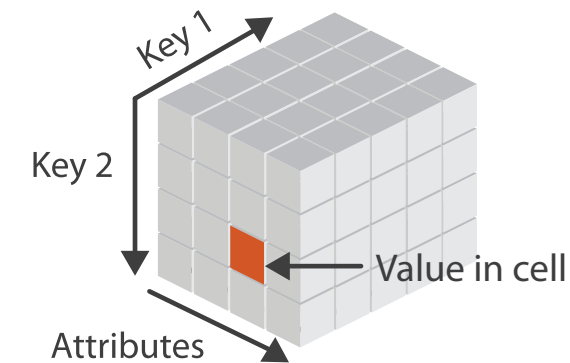
Keys and values

- key
 - independent attribute
 - used as unique index to look up items
 - simple tables: 1 key
 - multidimensional tables: multiple keys
- value
 - dependent attribute, value of cell
- classify arrangements by key count
 - 0, 1, 2, many...

→ Tables



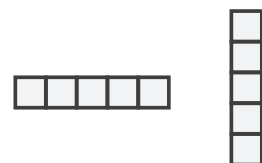
→ Multidimensional Table



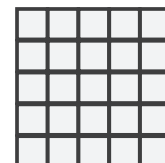
⊕ Express Values



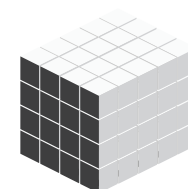
→ 1 Key
List



→ 2 Keys
Matrix



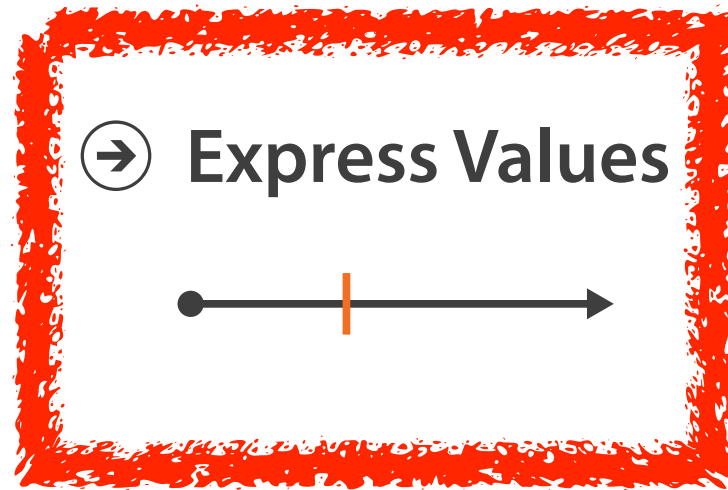
→ 3 Keys
Volume



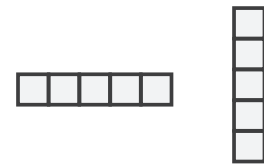
→ Many Keys
Recursive Subdivision



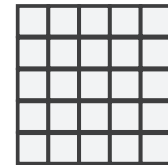
0 Keys



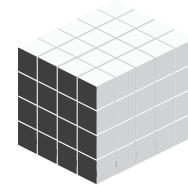
→ 1 Key
List



→ 2 Keys
Matrix



→ 3 Keys
Volume



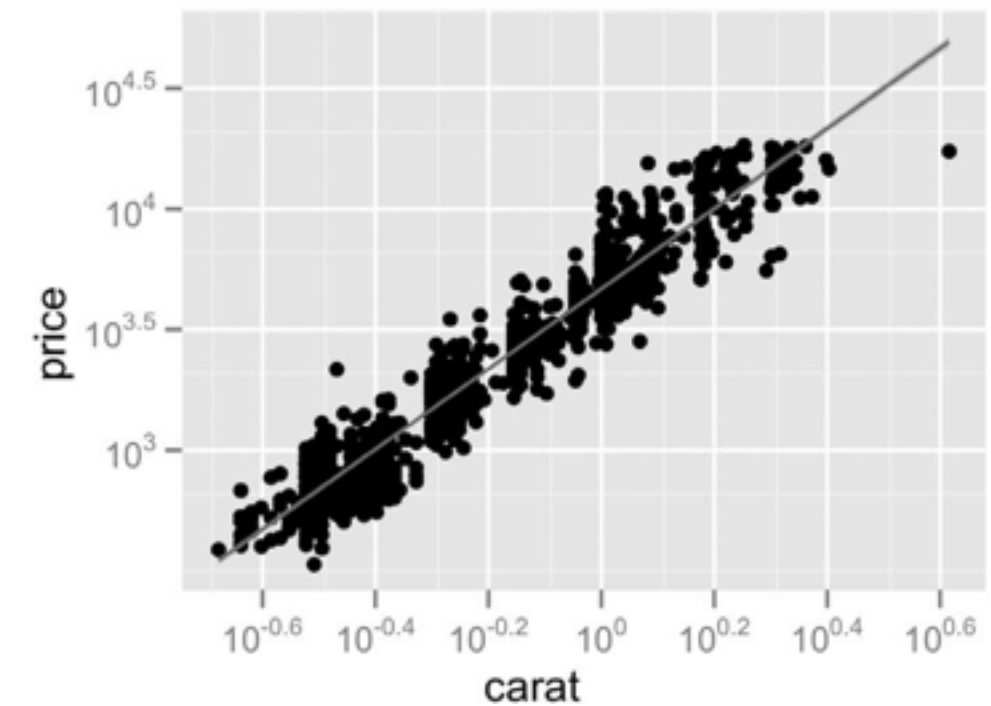
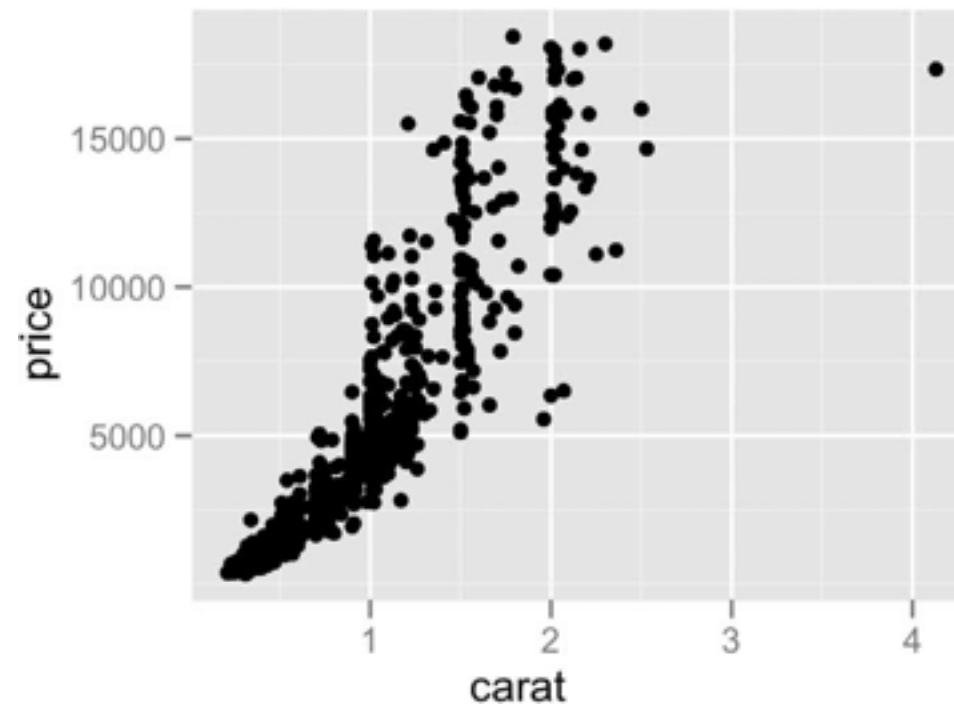
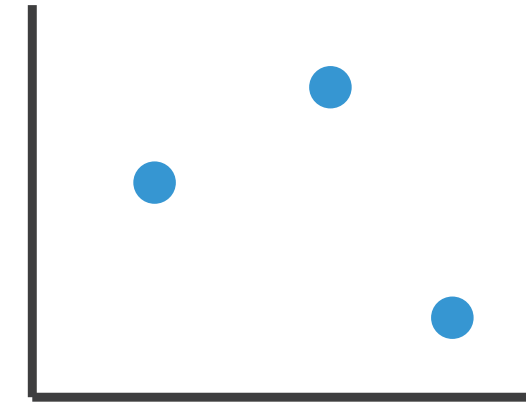
→ Many Keys
Recursive Subdivision



Idiom: scatterplot

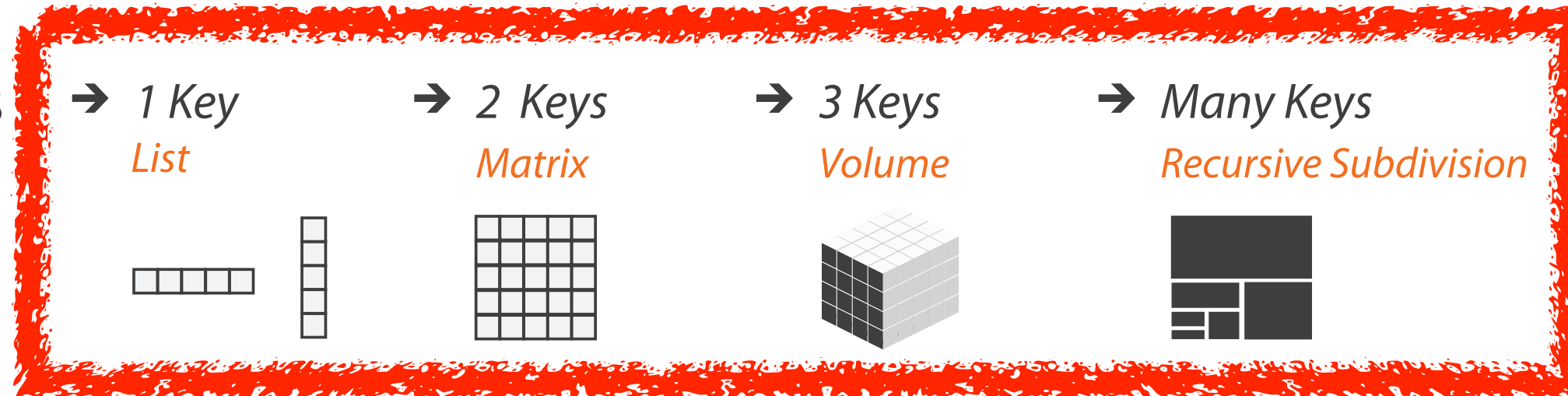
- *express* values
 - quantitative attributes
- no keys, only values
 - data
 - 2 quant attribs
 - mark: points
 - channels
 - horiz + vert position
 - tasks
 - find trends, outliers, distribution, correlation, clusters
 - scalability
 - hundreds of items

⊞ Express Values



Some keys

→ Express Values

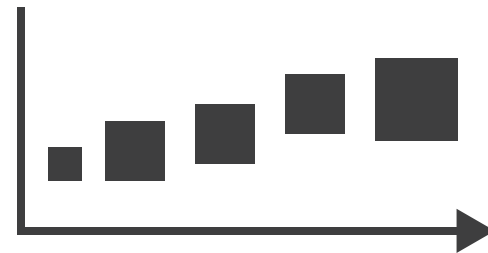


Some keys: Categorical regions

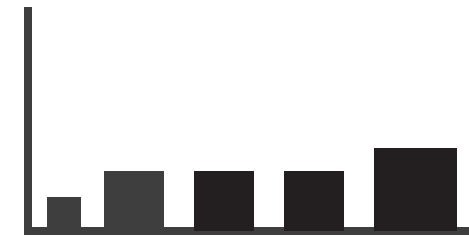
→ Separate



→ Order

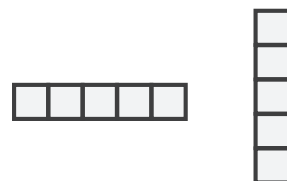


→ Align

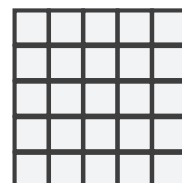


- **regions:** contiguous bounded areas distinct from each other
 - using space to *separate* (proximity)
 - following expressiveness principle for categorical attributes
- use ordered attribute to *order* and *align* regions

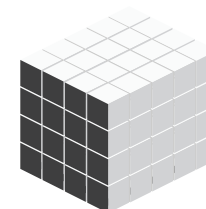
→ 1 Key
List



→ 2 Keys
Matrix



→ 3 Keys
Volume



→ Many Keys
Recursive Subdivision



Idiom: bar chart

- one key, one value

- data

- 1 categ attrib, 1 quant attrib

- mark: lines

- channels

- length to express quant value
- spatial regions: one per mark
 - separated horizontally, aligned vertically
 - ordered by quant attrib

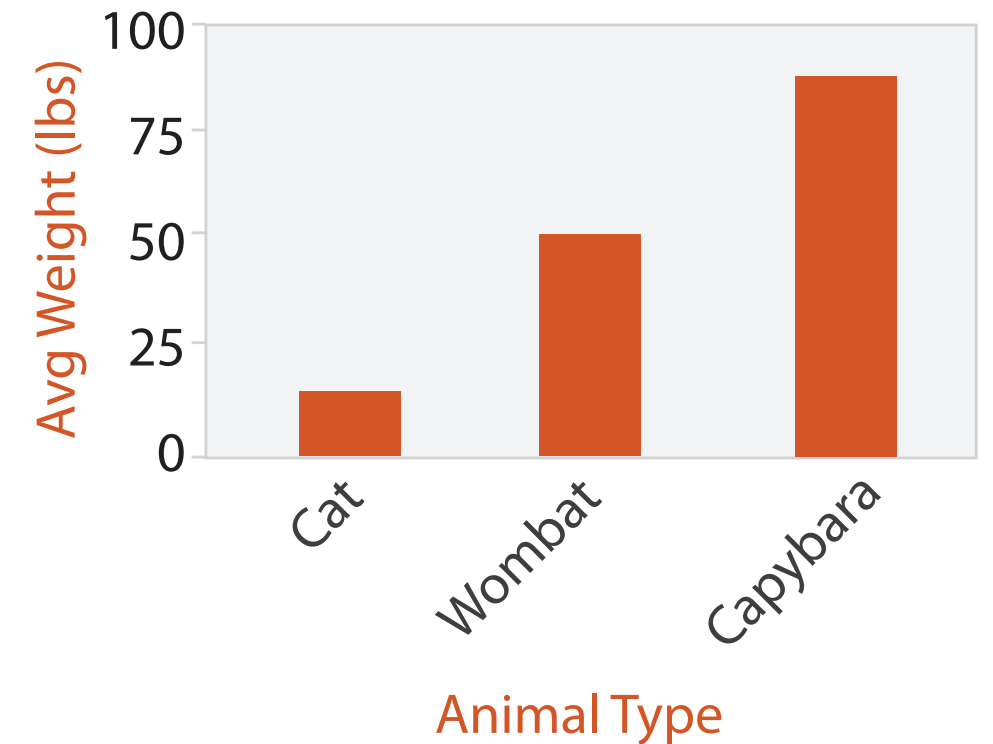
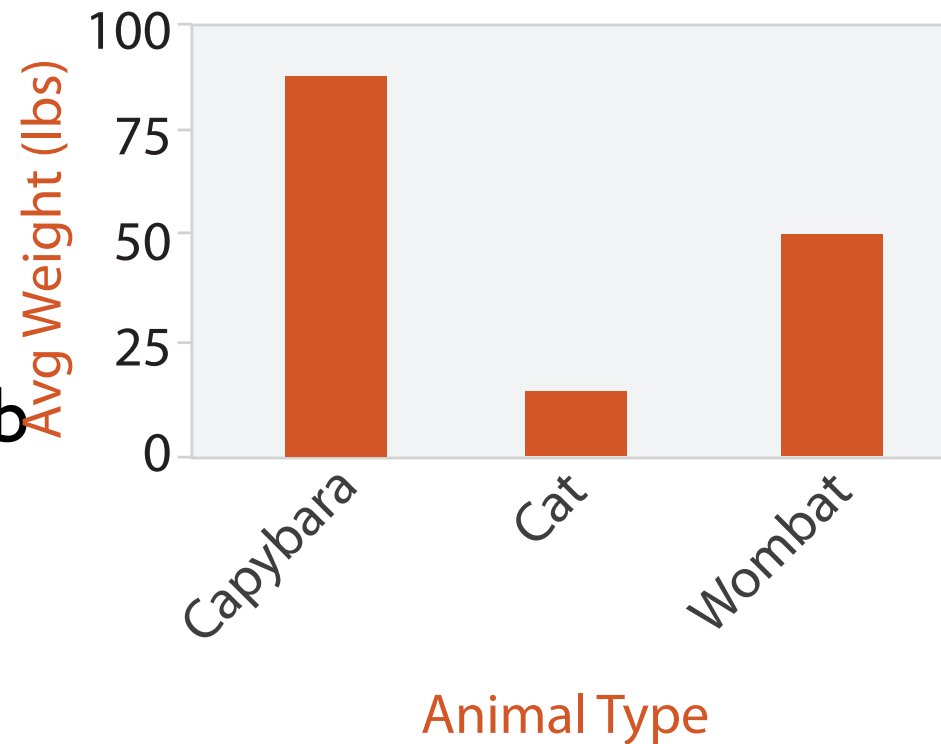
» by label (alphabetical), by length attrib (data-driven)

- task

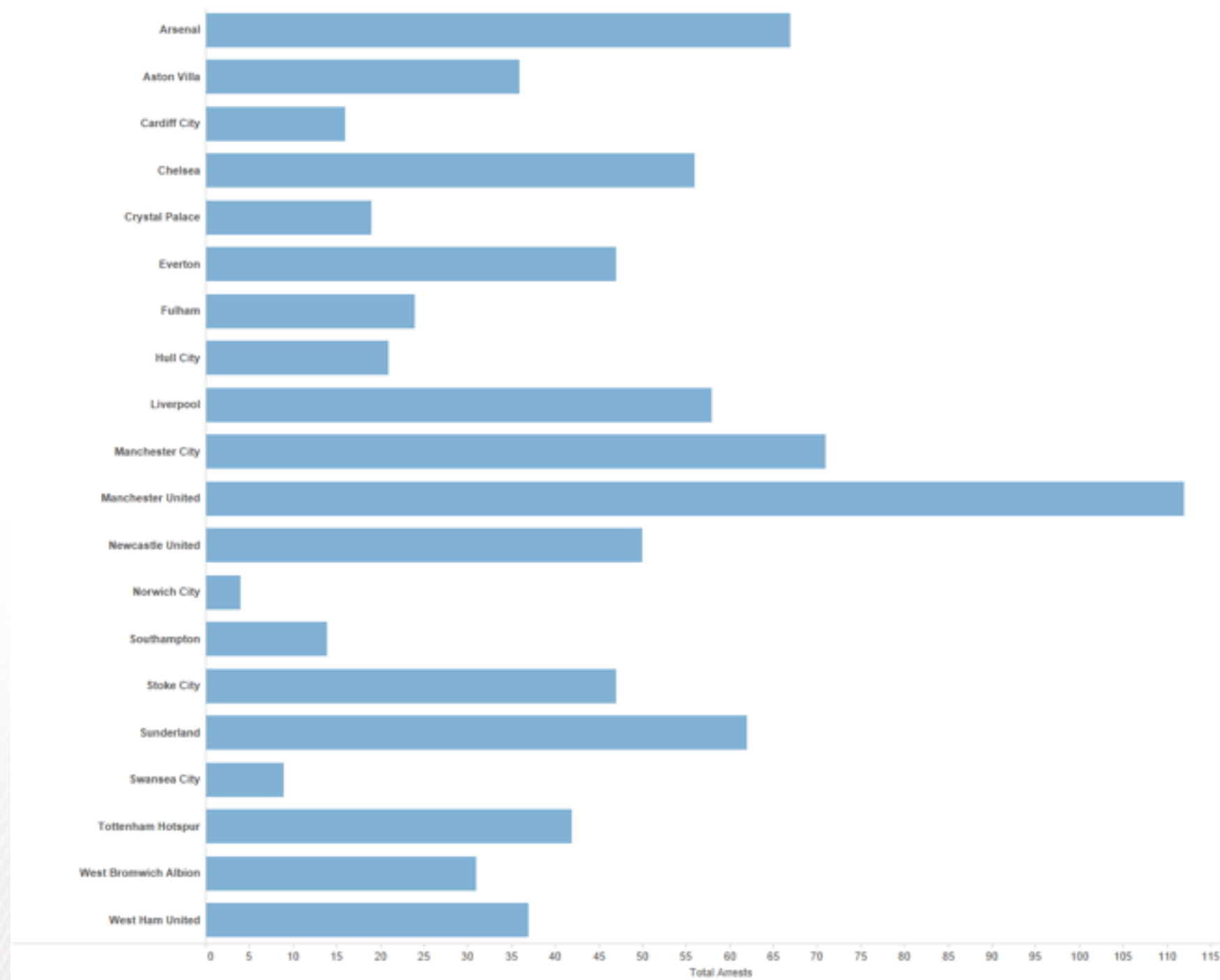
- compare, lookup values

- scalability

- dozens to hundreds of levels for key attrib



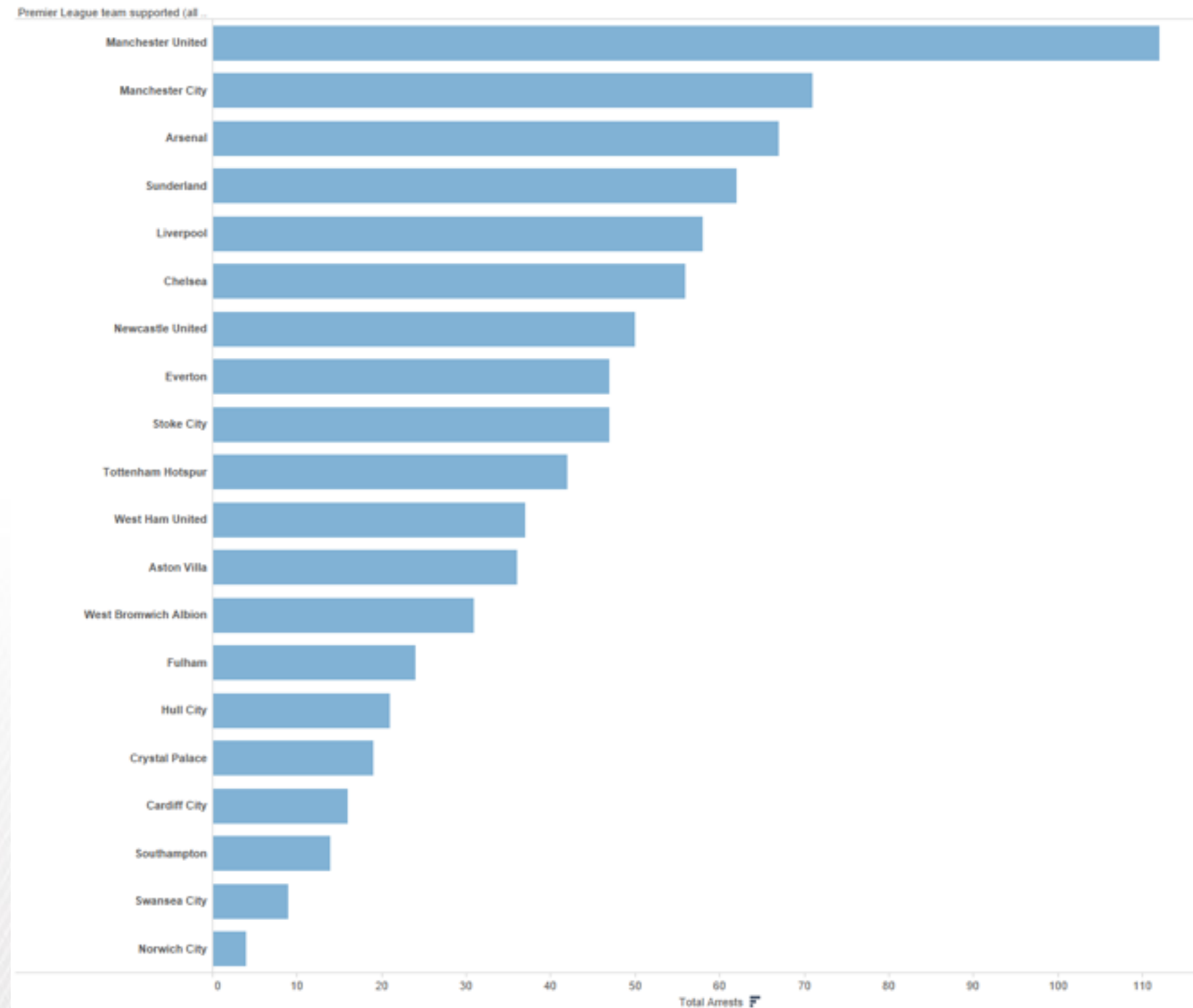
Separated and Aligned but not Ordered



LIMITATION: Hard to know rank. What's the 4th most? The 7th?

[Slide courtesy of Ben Jones]

Separated, Aligned and Ordered



[Slide courtesy of Ben Jones]

Separated but not Ordered or Aligned



LIMITATION: Hard to make comparisons

[Slide courtesy of Ben Jones]

Idiom: line chart / dot plot

- one key, one value

- data

- 2 quant attribs

- mark: points

- line connection marks between them

- channels

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

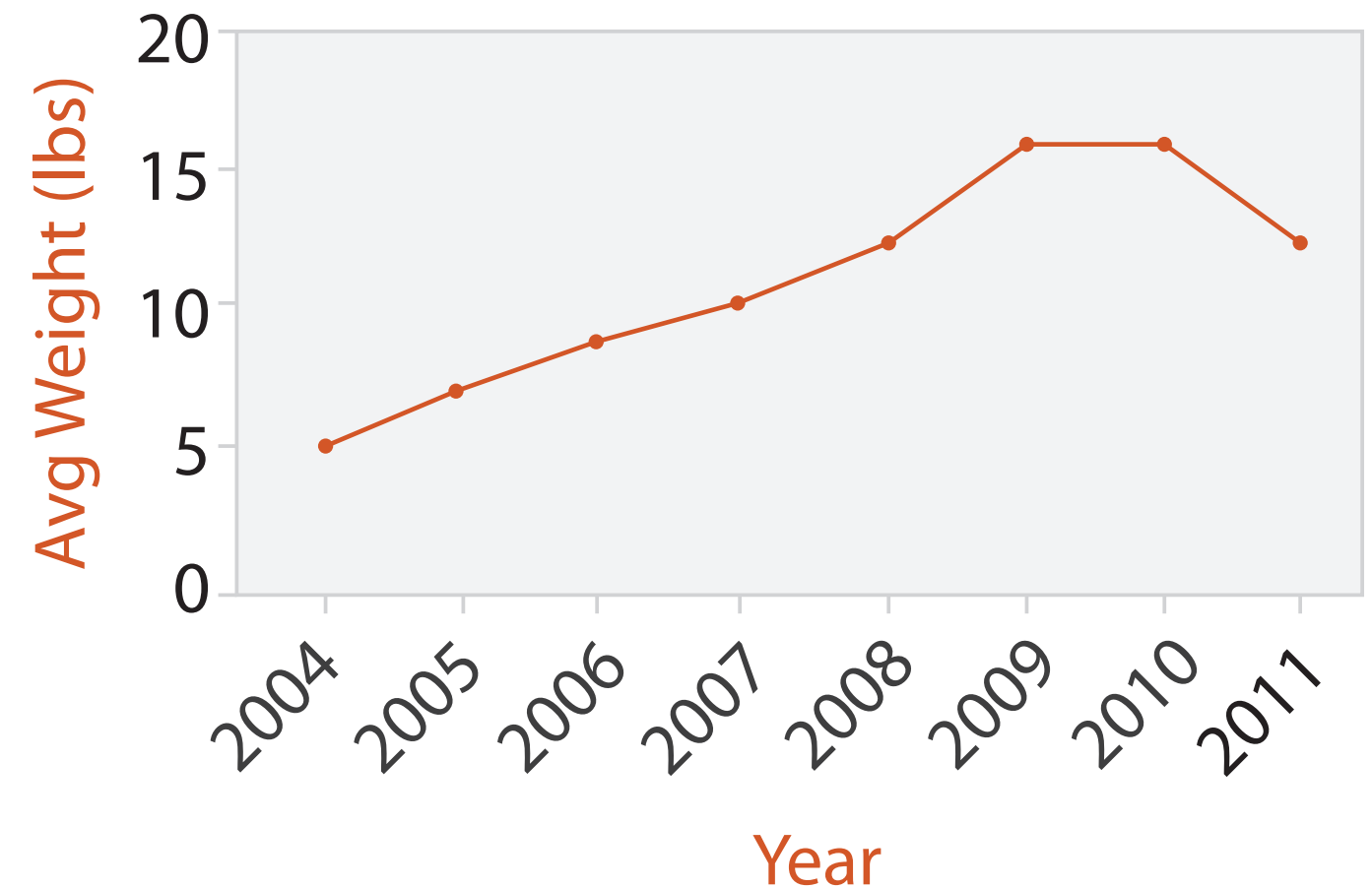
- task

- find trend

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

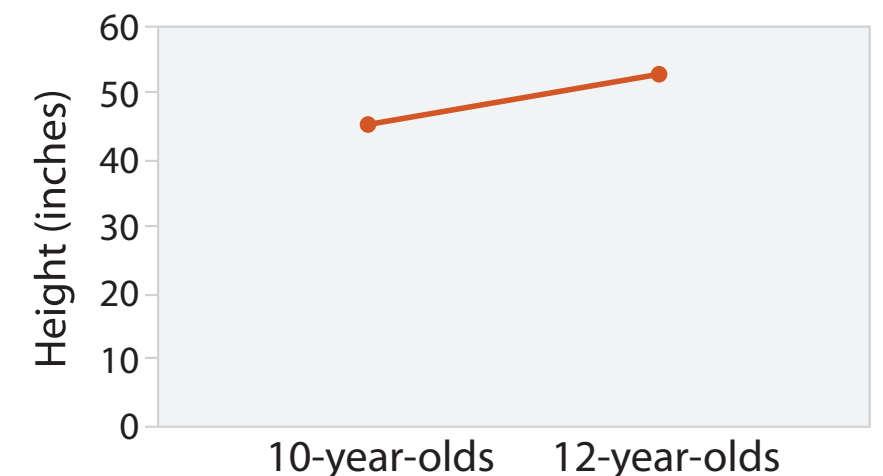
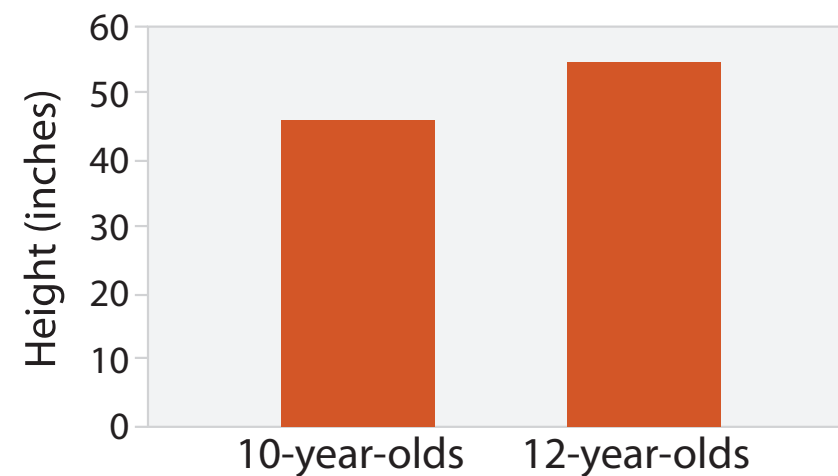
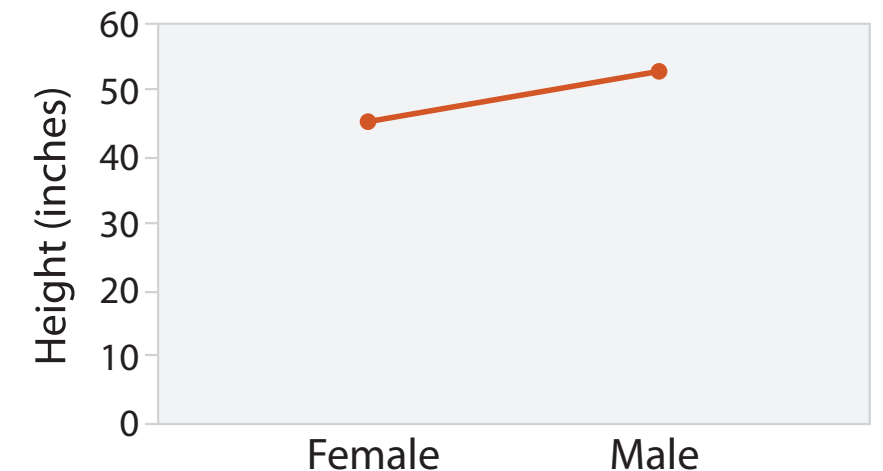
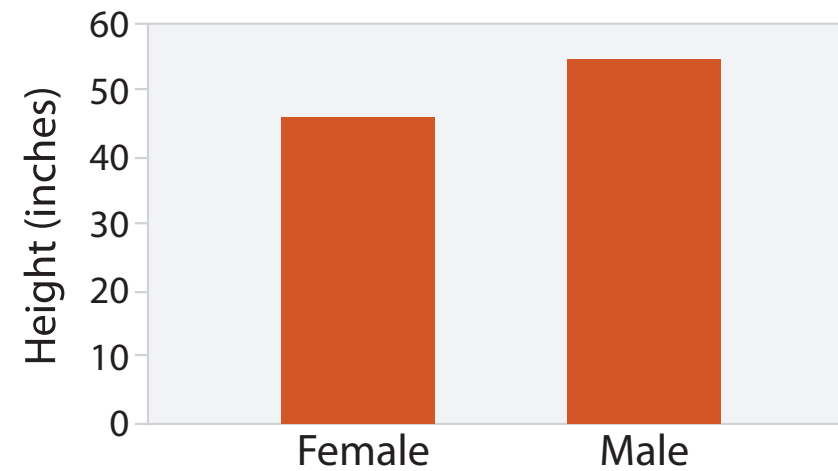
- scalability

- hundreds of key levels, hundreds of value levels



Choosing bar vs line charts

- depends on type of key attrib
 - bar charts if categorical
 - line charts if ordered
- do not use line charts for categorical key attribs
 - violates expressiveness principle
 - implication of trend so strong that it overrides semantics!
 - “The more male a person is, the taller he/she is”



after [Bars and Lines: A Study of Graphic Communication. Zacks and Tversky. *Memory and Cognition* 27:6 (1999), 1073–1079.]

2 Keys

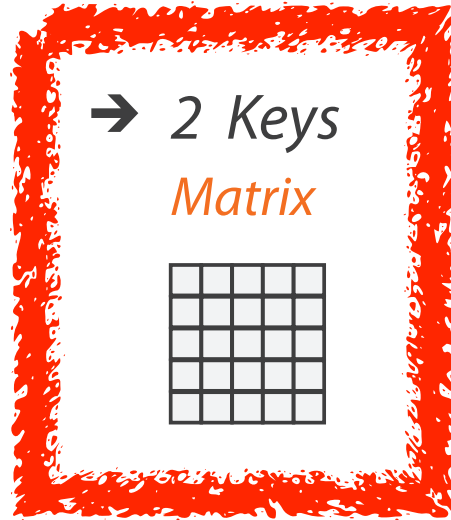
➔ Express Values



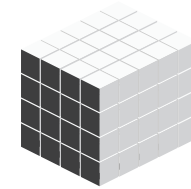
➔ 1 Key
List



➔ 2 Keys
Matrix



➔ 3 Keys
Volume



➔ Many Keys
Recursive Subdivision



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

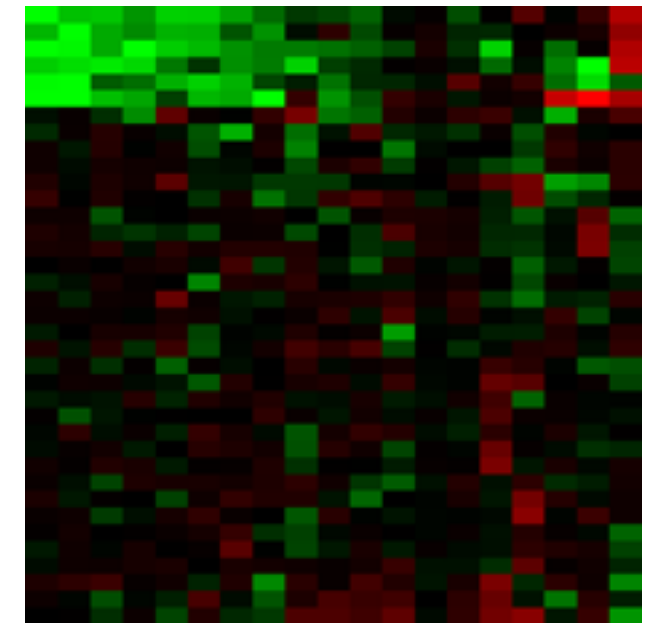
- color by quant attrib
 - (ordered diverging colormap)

- task

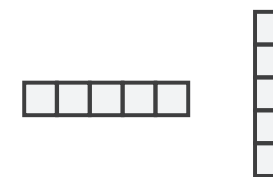
- find clusters, outliers

- scalability

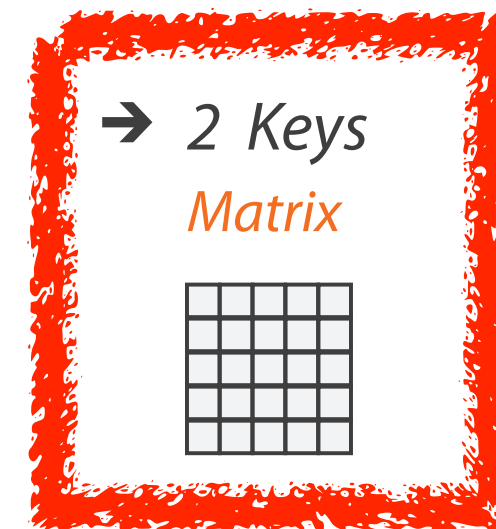
- 1M items, 100s of categ levels, ~10 quant attrib levels



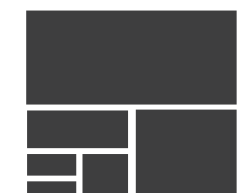
→ 1 Key
List



→ 2 Keys
Matrix

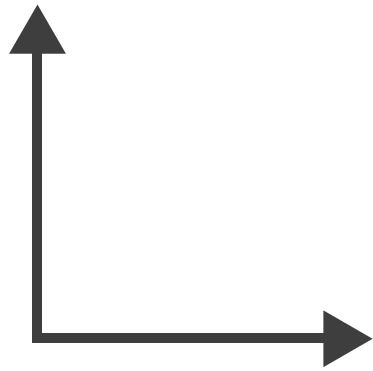


→ Many Keys
Recursive Subdivision

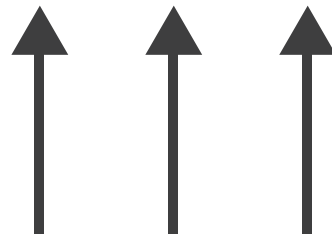


➔ Axis Orientation

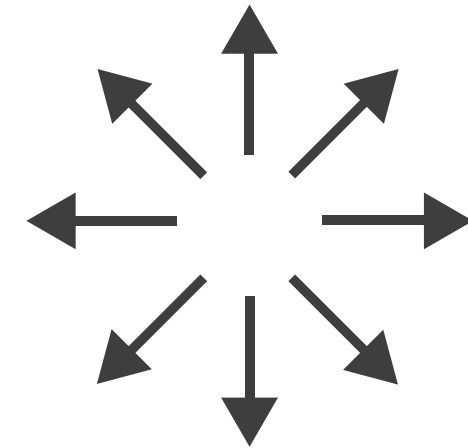
➔ Rectilinear



➔ Parallel

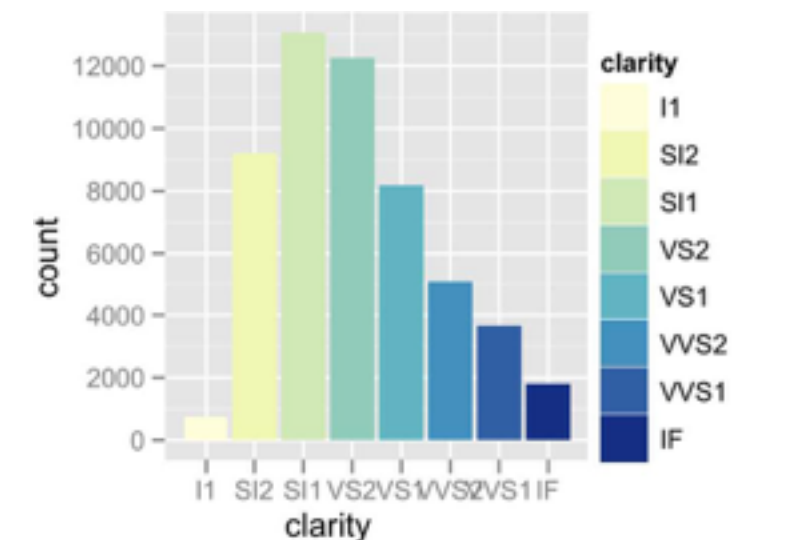
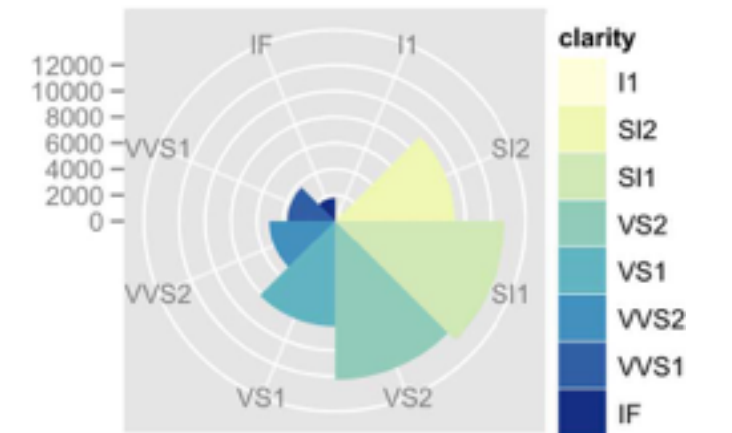
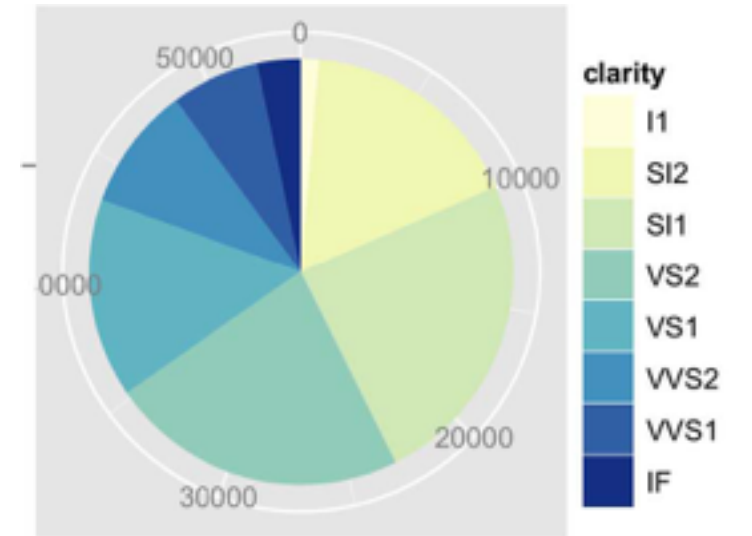


➔ Radial



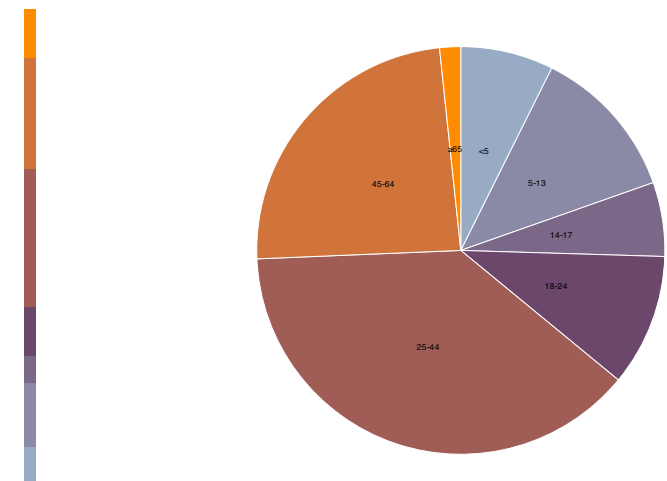
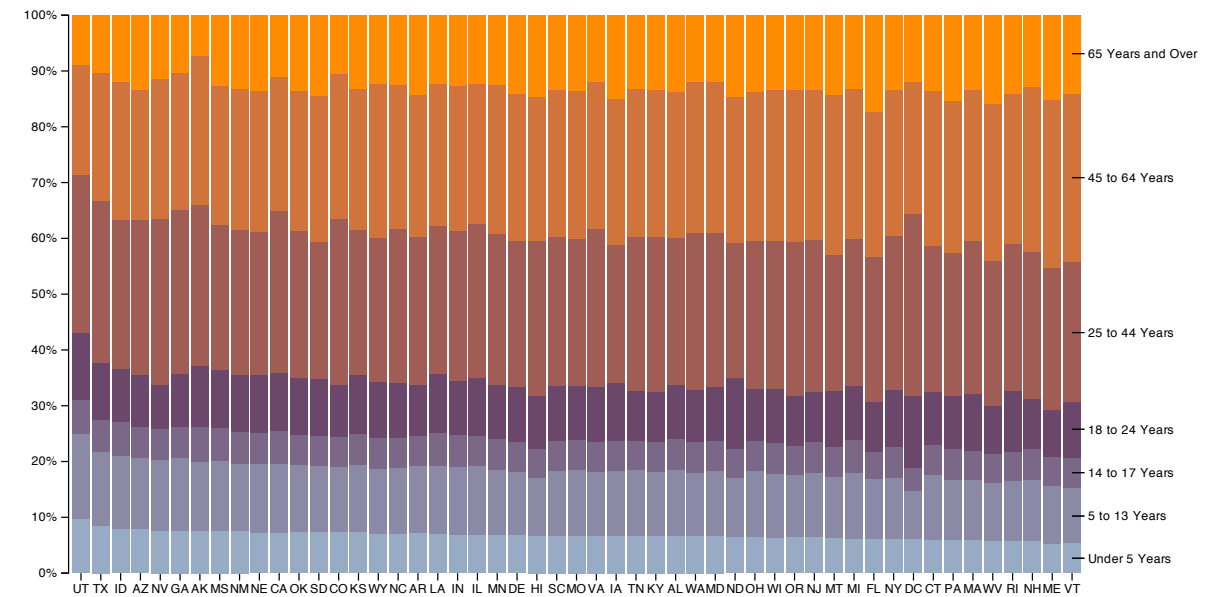
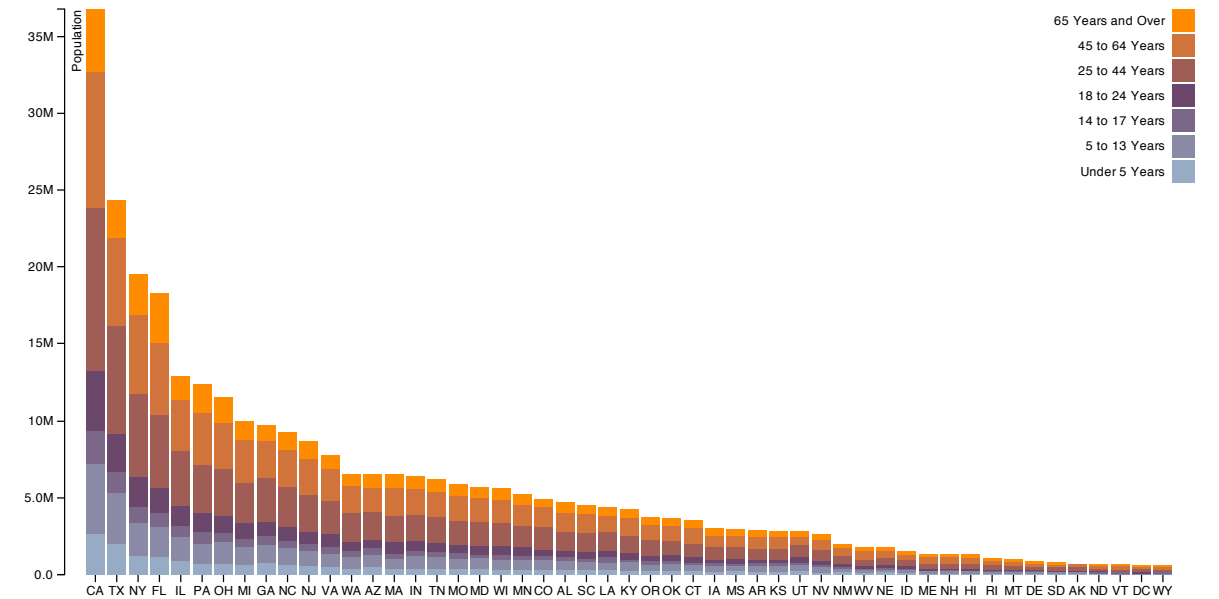
Idioms: pie chart, polar area chart

- pie chart
 - area marks with angle channel
 - accuracy: angle/area less accurate than line length
 - arclength also less accurate than line length
- polar area chart
 - area marks with length channel
 - more direct analog to bar charts
- data
 - I categ key attrib, I quant value attrib
- task
 - part-to-whole judgements



Idioms: **normalized stacked bar chart**

- task
 - part-to-whole judgements
- **normalized stacked bar chart**
 - stacked bar chart, normalized to full vert height
 - single stacked bar equivalent to full pie
 - high information density: requires narrow rectangle
- **pie chart**
 - information density: requires large circle



<http://bl.ocks.org/mbostock/3886208>,

<http://bl.ocks.org/mbostock/3887235>,

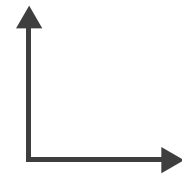
<http://bl.ocks.org/mbostock/3886394>.

Idiom: glyphmaps

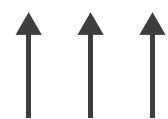
- rectilinear good for linear vs nonlinear trends
- radial good for cyclic patterns

➔ Axis Orientation

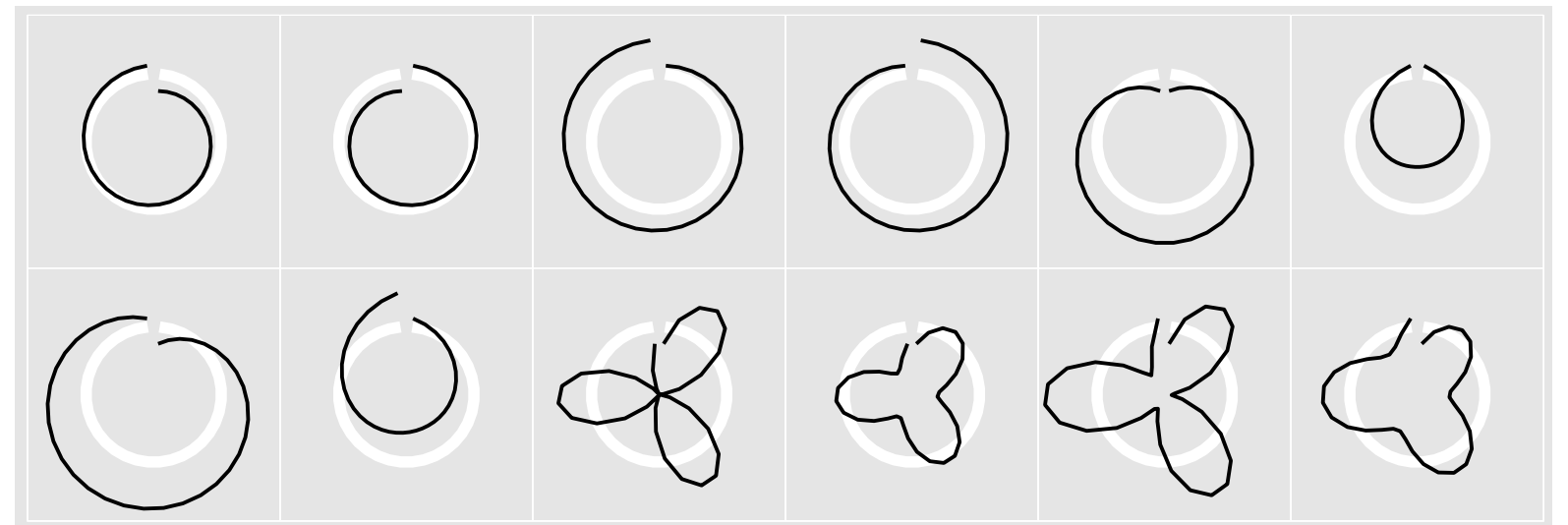
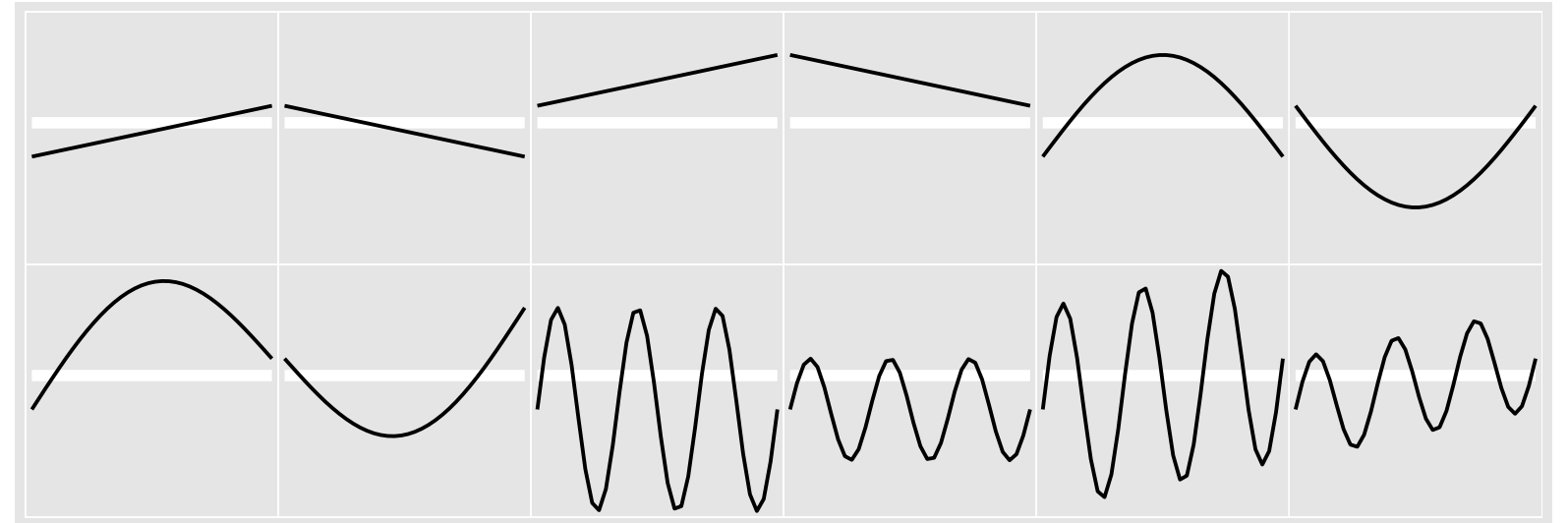
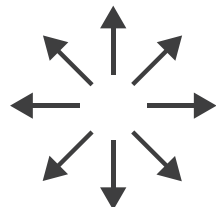
➔ Rectilinear



➔ Parallel



➔ Radial



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

Arrange spatial data

→ Use Given

→ Geometry

→ *Geographic*

→ *Other Derived*

→ Spatial Fields

→ *Scalar Fields (one value per cell)*

→ *Isocontours*

→ *Direct Volume Rendering*

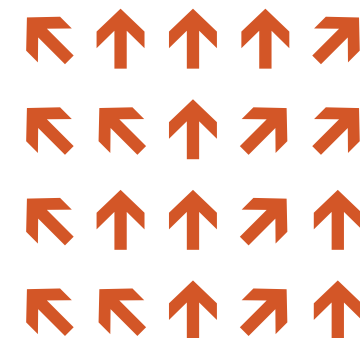
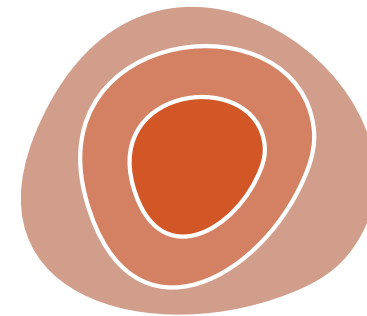
→ *Vector and Tensor Fields (many values per cell)*

→ *Flow Glyphs (local)*

→ *Geometric (sparse seeds)*

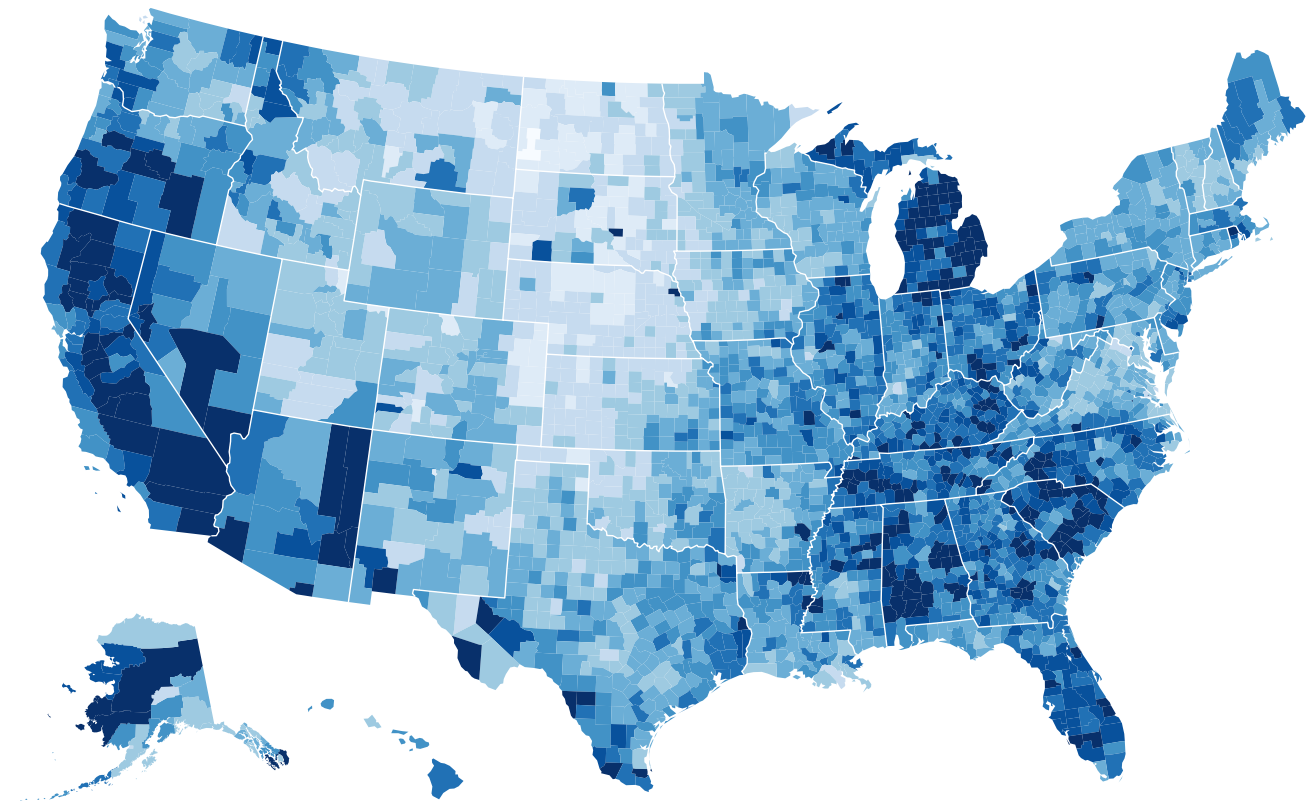
→ *Textures (dense seeds)*

→ *Features (globally derived)*



Idiom: **choropleth map**

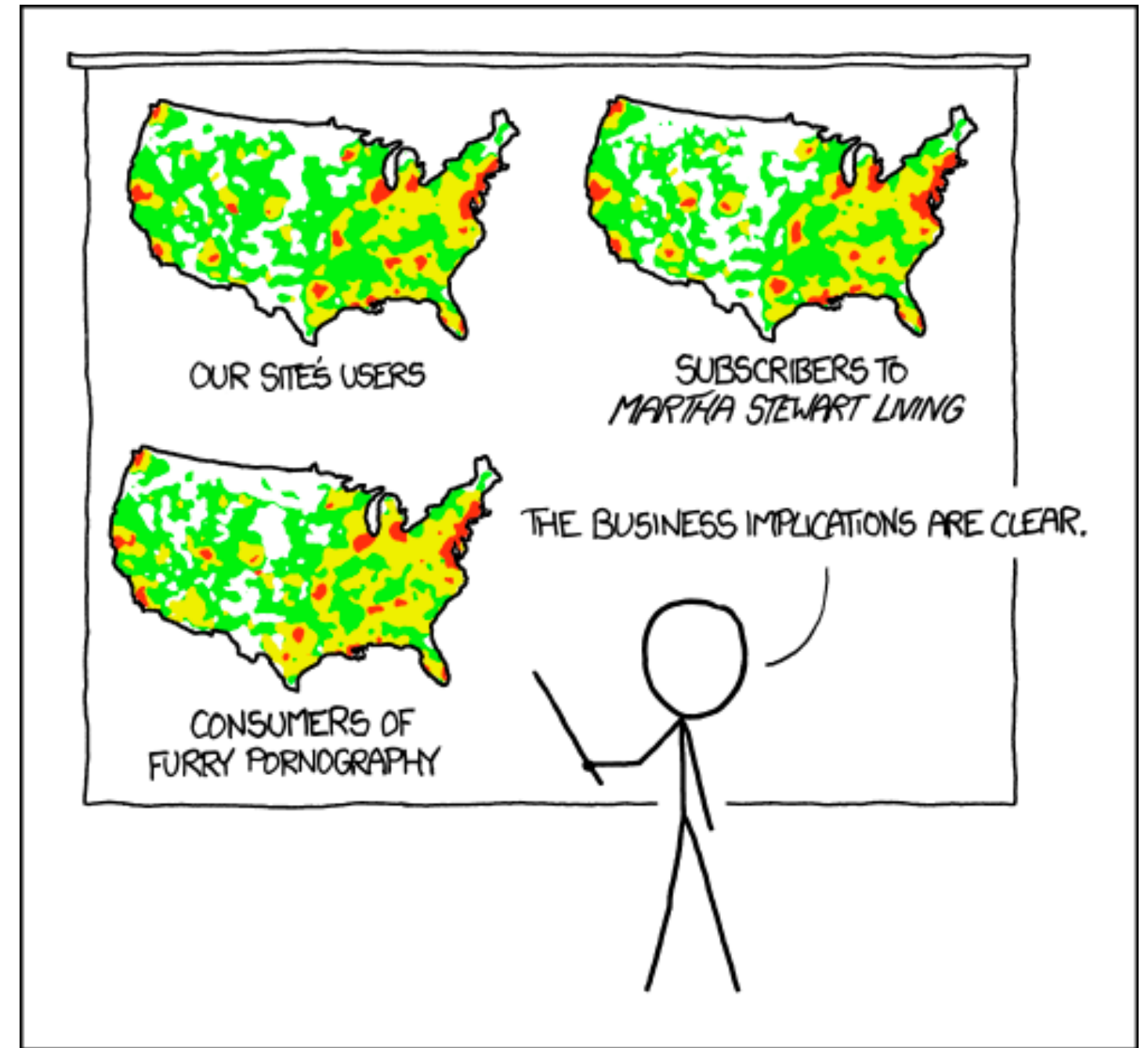
- *use* given spatial data
 - when central task is understanding spatial relationships
- data
 - geographic geometry
 - table with 1 quant attribute per region
- encoding
 - use given geometry for area mark boundaries
 - sequential segmented colormap *[more later]*
 - (geographic heat map)



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

Population maps trickiness

- beware!
 - absolute vs relative again
 - population density vs per capita
 - investigate with Ben Jones Tableau Public demo
 - <http://public.tableau.com/profile/ben.jones#!/vizhome/PopVsFin/PopVsFin>
- Are Maps of Financial Variables just Population Maps?*
- yes, unless you look at per capita (relative) numbers

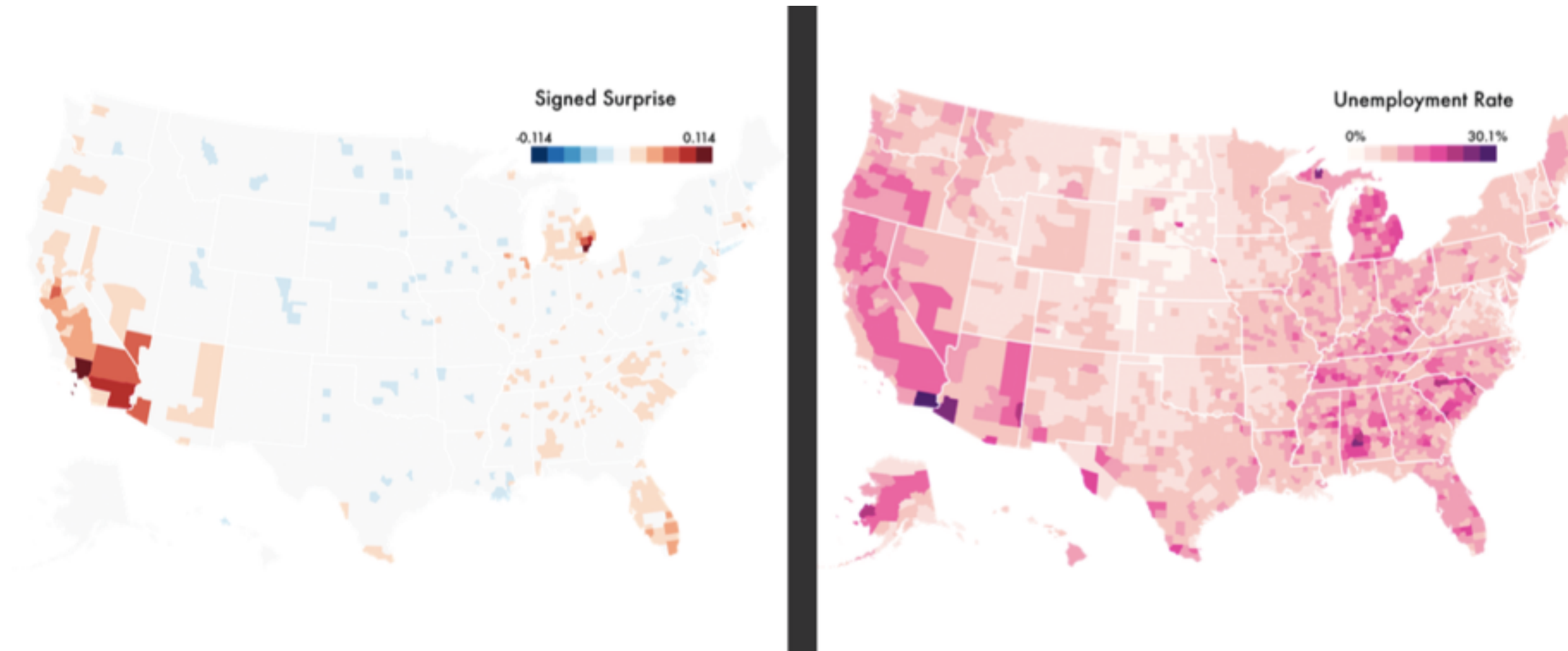


PET PEEVE #208:
GEOGRAPHIC PROFILE MAPS WHICH ARE
BASICALLY JUST POPULATION MAPS

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

Idiom: Bayesian surprise maps

- use models of expectations to highlight surprising values
- confounds (population) and variance (sparsity)



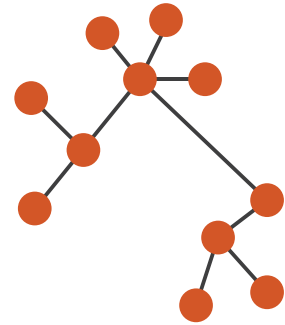
[Surprise! Bayesian Weighting for De-Biasing Thematic Maps. Correll and Heer. Proc InfoVis 2016]

Arrange networks and trees

→ Node–Link Diagrams Connection Marks

✓ NETWORKS

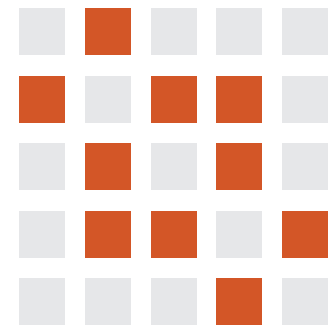
✓ TREES



→ Adjacency Matrix Derived Table

✓ NETWORKS

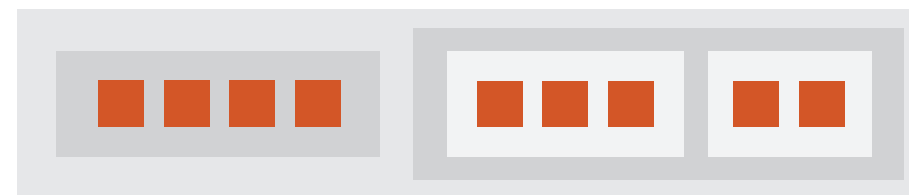
✓ TREES



→ Enclosure Containment Marks

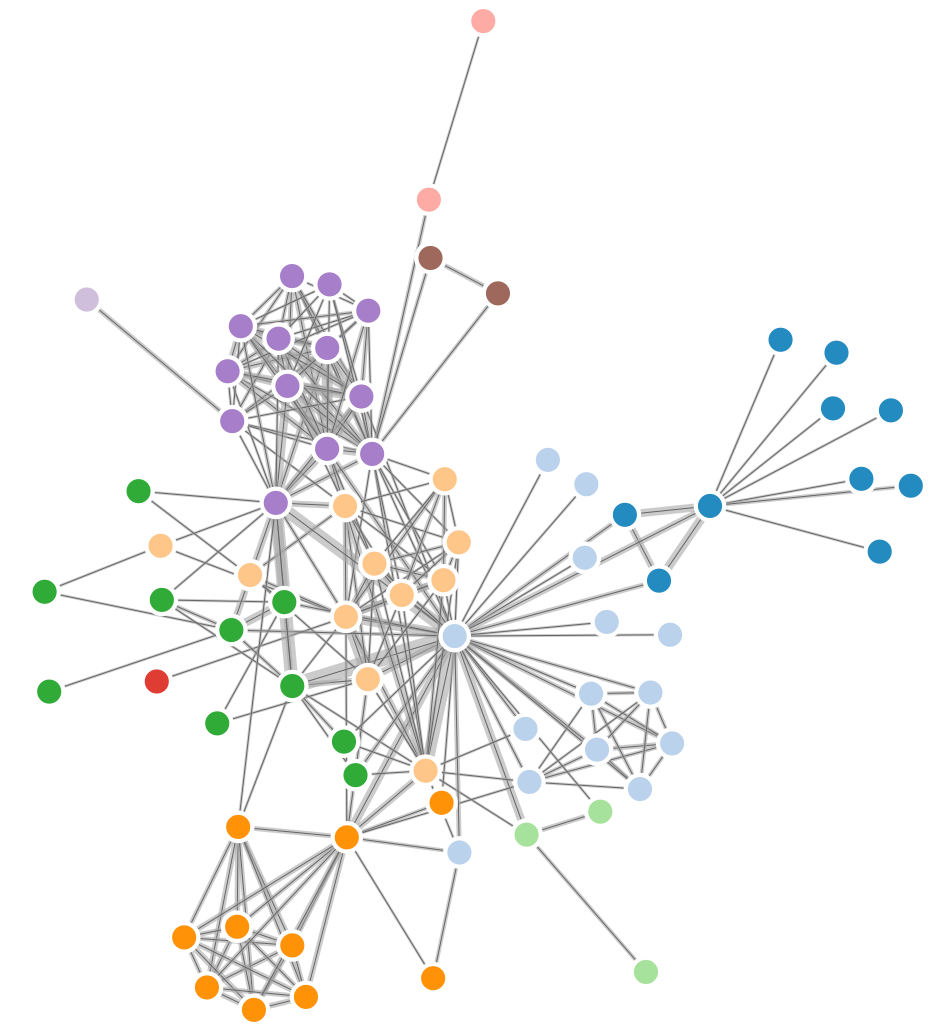
✗ NETWORKS

✓ TREES



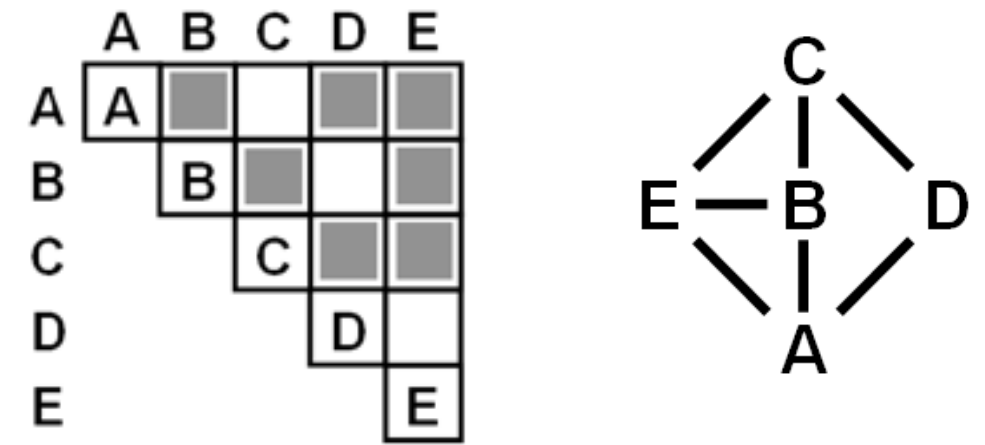
Idiom: **force-directed placement**

- 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



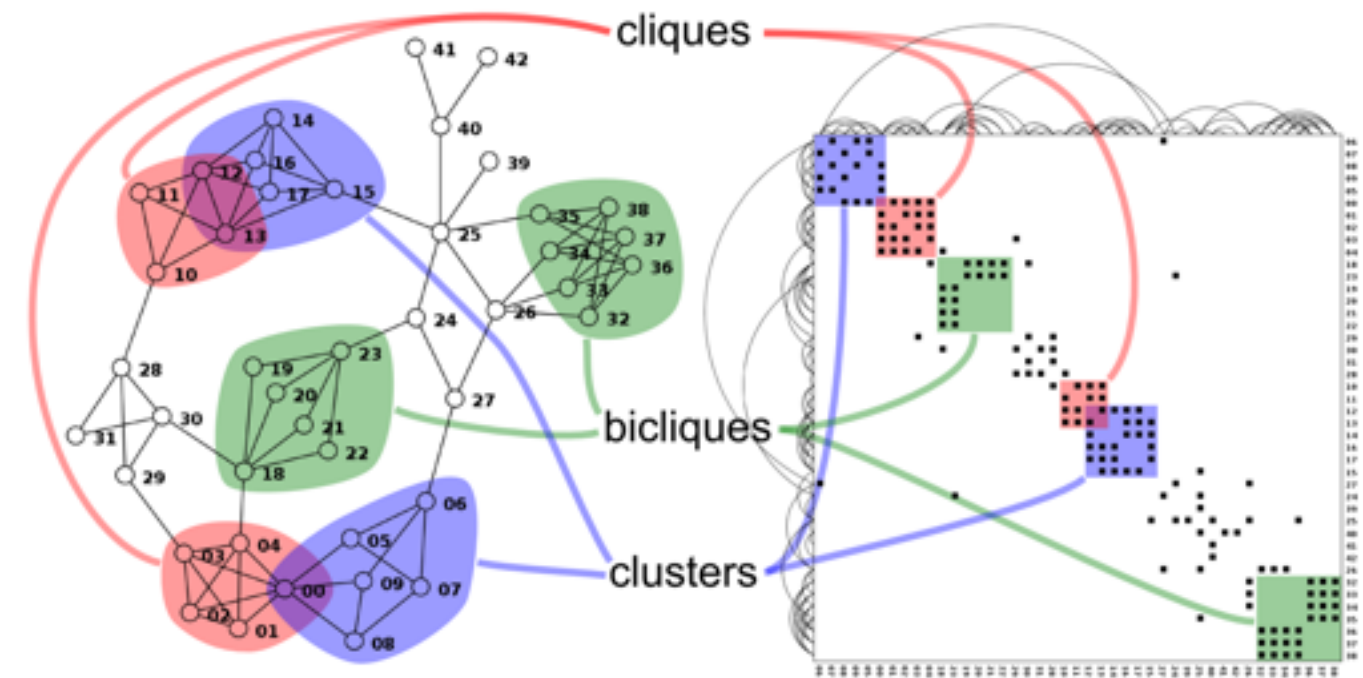
[NodeTrix: a Hybrid Visualization of Social Networks. Henry, Fekete, and McGuffin. IEEE TVCG (Proc. InfoVis) 13(6):1302-1309, 2007.]



[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!



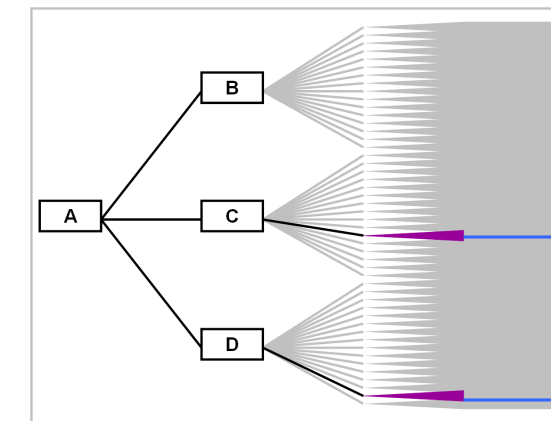
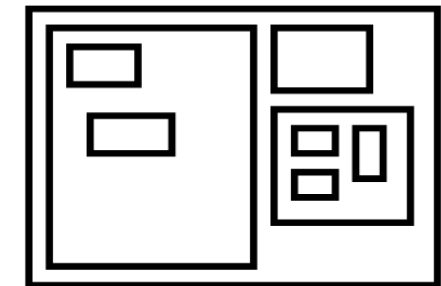
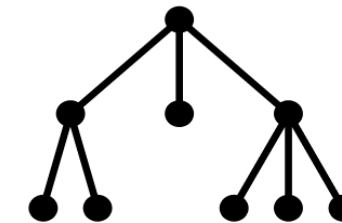
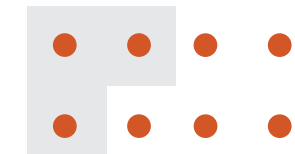
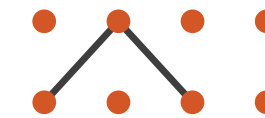
<http://www.michaelmcguffin.com/courses/vis/patternsInAdjacencyMatrix.png>

[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.]

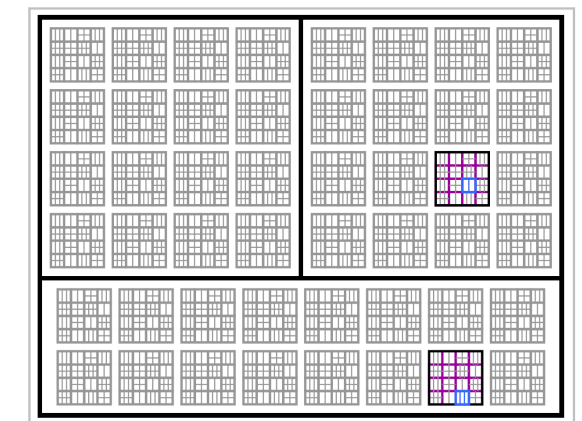
Link marks: Connection and containment

- 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

➔ Connection ➔ Containment



Node-Link Diagram



Treemap

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

How?

Encode

→ Arrange

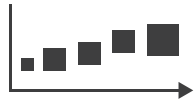
→ Express



→ Separate



→ Order



→ Align



→ Use



→ Map

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

→ Color

→ Hue



→ Saturation



→ Luminance



→ Size, Angle, Curvature, ...



→ Shape



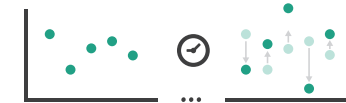
→ Motion

Direction, Rate, Frequency, ...

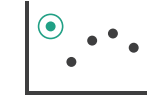


Manipulate

→ Change



→ Select



→ Navigate

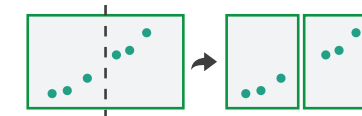


Facet

→ Juxtapose



→ Partition

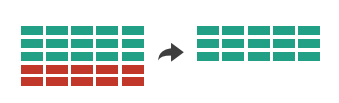


→ Superimpose

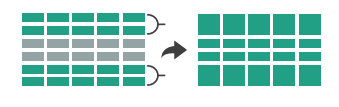


Reduce

→ Filter



→ Aggregate



→ Embed



What?

Why?

How?

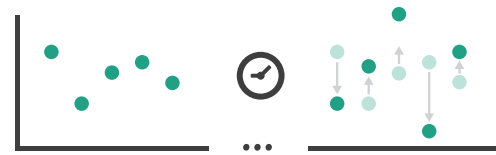
How to handle complexity: 1 previous strategy + 3 more

→ *Derive*

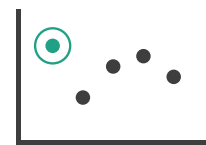


Manipulate

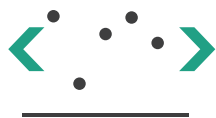
→ Change



→ Select

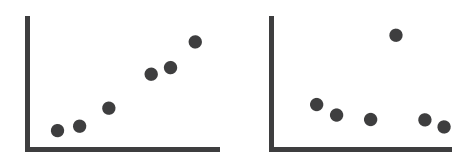


→ Navigate

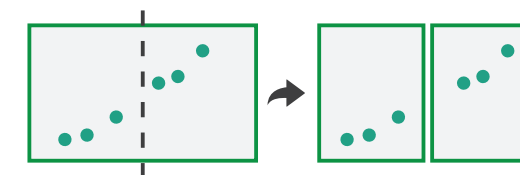


Facet

→ Juxtapose



→ Partition

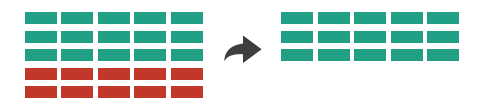


→ Superimpose



Reduce

→ Filter



→ Aggregate



→ Embed



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

More Information

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

- this talk

www.cs.ubc.ca/~tmm/talks.html#vad17stat545

- book page (including tutorial lecture slides)

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

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

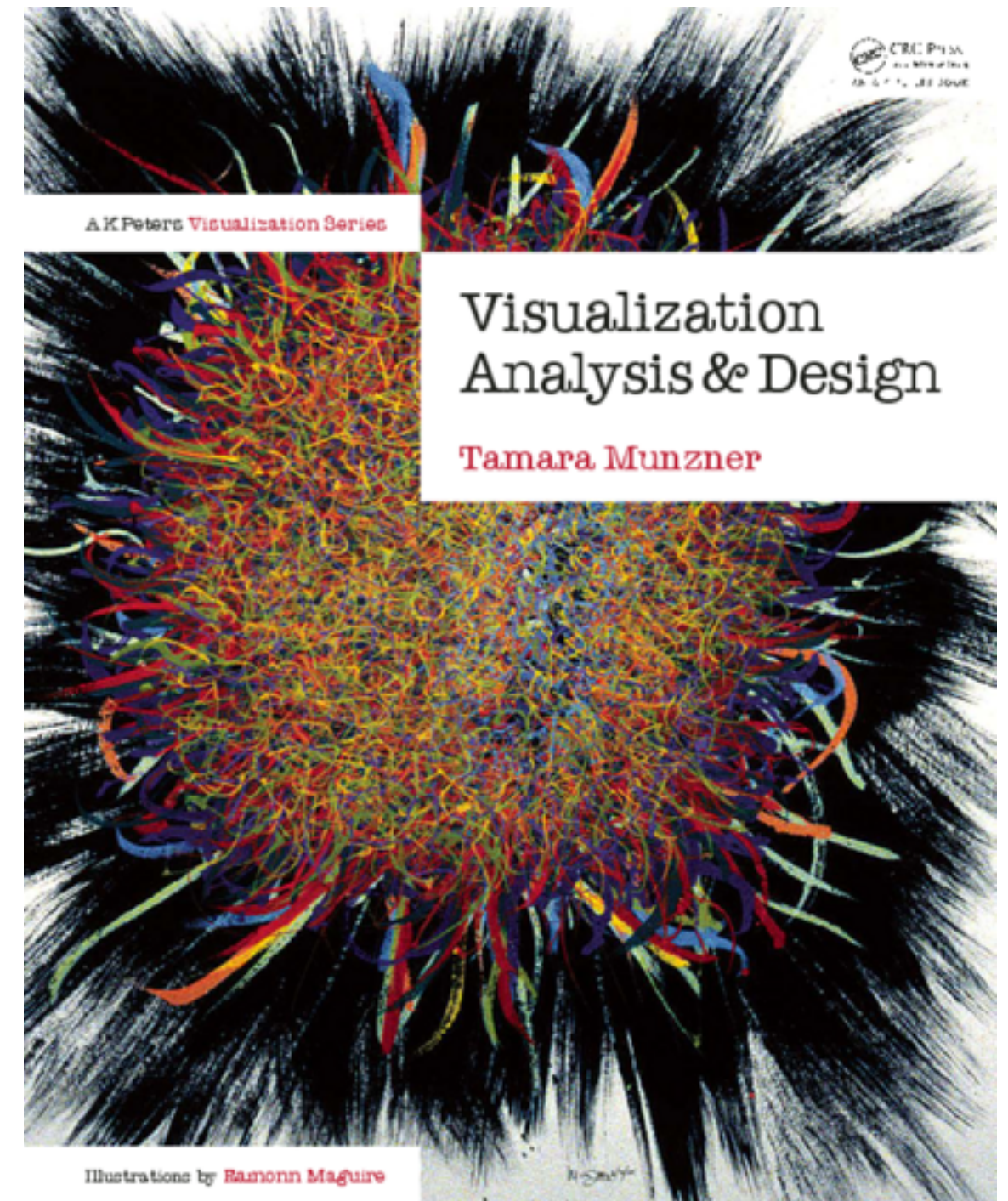
– <http://www.crcpress.com/product/isbn/9781466508910>

– illustrations: Eamonn Maguire

- papers, videos, software, talks, 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.