

Week 1: Intro, Tasks and Data, Marks and Channels

Tamara Munzner

Department of Computer Science
University of British Columbia

JRNL 520H, Special Topics in Contemporary Journalism: Data Visualization
Week 1: 13 September 2016

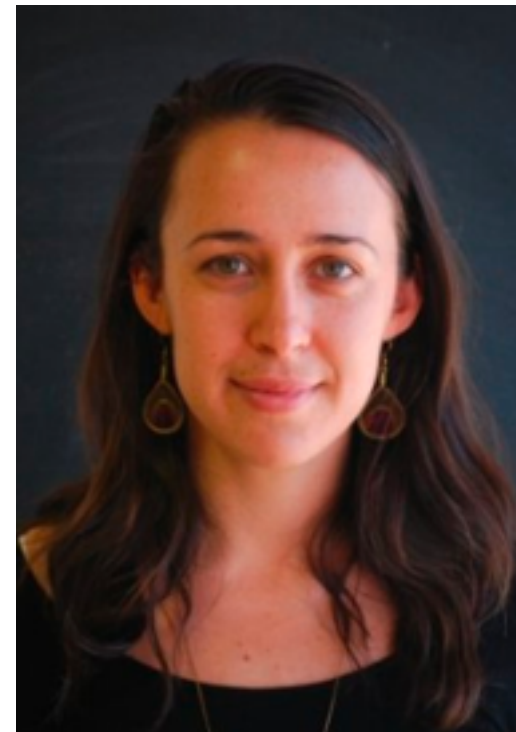
<http://www.cs.ubc.ca/~tmm/courses/journ16>

Who's who

- Instructor: Tamara Munzner
 - UBC Computer Science



- Instructor: Caitlin Havlak
 - Discourse Media



Class time

- 6 weeks, Sep 13 - Oct 18
 - once/week, 3 hr session 9:30am-12:30pm
- standard week
 - foundations lecture/discussion: 80 min
 - break: 15 min
 - demos: 45 min
 - lab: 30 min
- office hrs: 1-3pm most weeks

Structure

- participation, 10%
 - attend lectures and demos, discuss
 - tell us in advance if you'll miss class (and why)
 - tell when us recover if you were ill
- homework, 90%
 - gradual transition from structured to open-ended
 - 60%: 5 assignments
 - best 4 out of 5 marks used, so 15% each
 - start in lab time, finish over the subsequent week
 - due just before next class session (9am)
 - some solo, some in groups of 2
 - 30%: final assignment
 - find your own interesting data and design your own visualization for it

Further reading

- optional textbook for following up on visualization foundations lectures
 - Tamara Munzner. Visualization Analysis and Design. CRC Press, 2014.
 - <http://www.cs.ubc.ca/~tmm/vadbook/>
 - library has multiple ebook copies
 - to buy yourself, see course page
- optional textbook for more about Tableau software
 - Ben Jones, Communicating Data with Tableau. O'Reilly, 2014.
 - <http://dataremixed.com/books/cdwt/>
- optional papers/books
 - links and references posted on course page
 - if DL links, use library EZproxy from off campus

Finding us

- office hours in Sing Tao bldg
 - 1-3pm Tuesdays: Tamara and/or Caitlin
 - by appointment: Tamara in ICICS/CS bldg Room X661
- email other times
 - tmm@cs.ubc.ca, caitlin@discoursemedia.org
- course page is font of all information
 - don't forget to refresh, frequent updates
 - <http://www.cs.ubc.ca/~tmm/courses/journ16>

Topics

- Week 1
 - Intro
 - Tasks and Data
 - Marks and Channels
- Week 2
 - Arrange Data Tables
- Week 3
 - Color
 - Arrange Spatial Data
- Week 4
 - Manipulate, Facet, Reduce
- Week 5
 - Wrangle
 - Stories
 - Rules of Thumb
- Week 6
 - Networks
 - Regression Lines
 - Vis in Newsrooms

Introduction: 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.

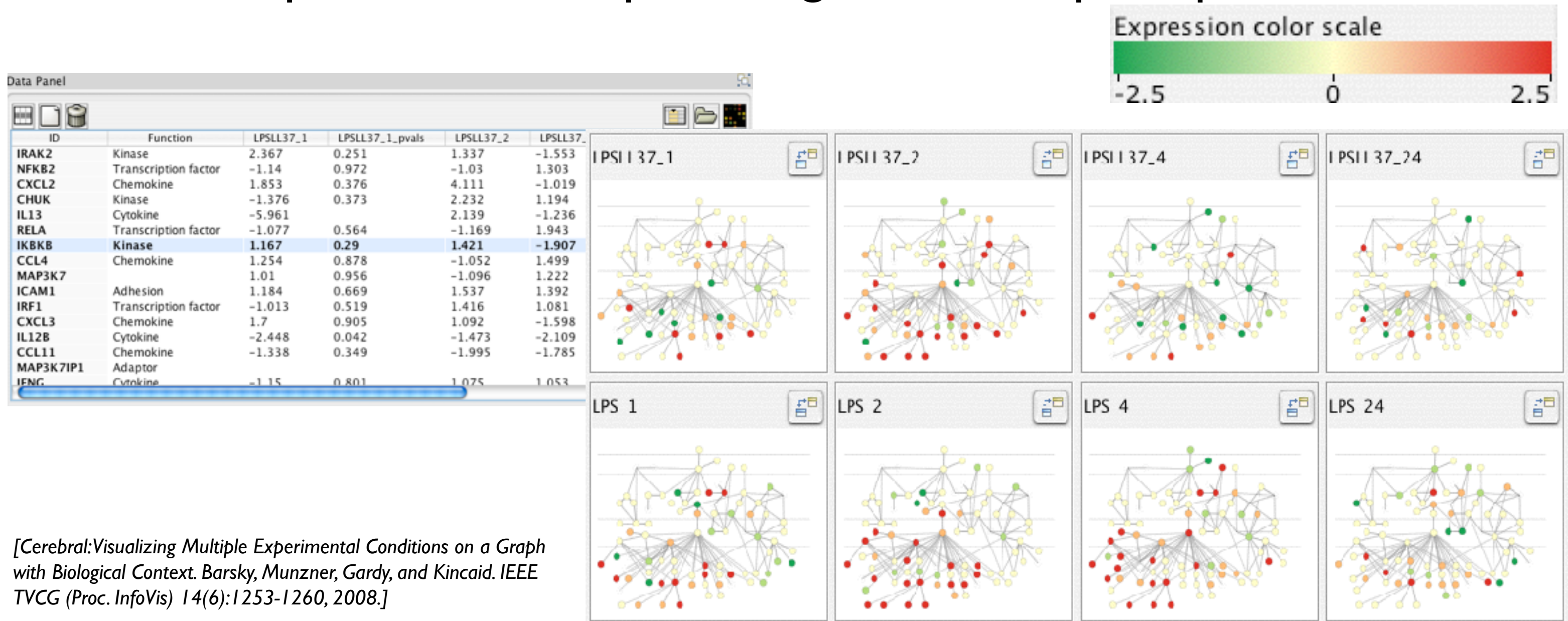
Visualization is suitable when there is a need to augment human capabilities rather than replace people with computational decision-making methods.

- 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 depend on vision?

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

- human visual system is high-bandwidth channel to brain
 - overview possible due to background processing
 - subjective experience of seeing everything simultaneously
 - significant processing occurs in parallel and pre-attentively
- sound: lower bandwidth and different semantics
 - overview not supported
 - subjective experience of sequential stream
- touch/haptics: impoverished record/replay capacity
 - only very low-bandwidth communication thus far
- taste, smell: no viable record/replay devices

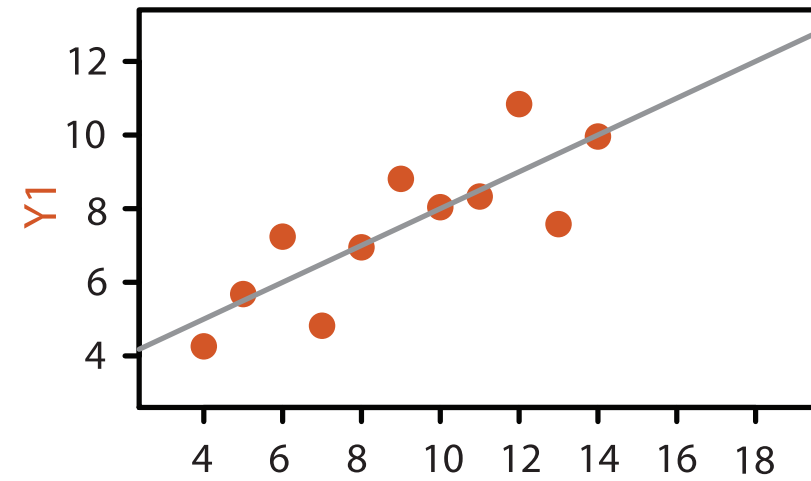
Why show the data in detail?

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

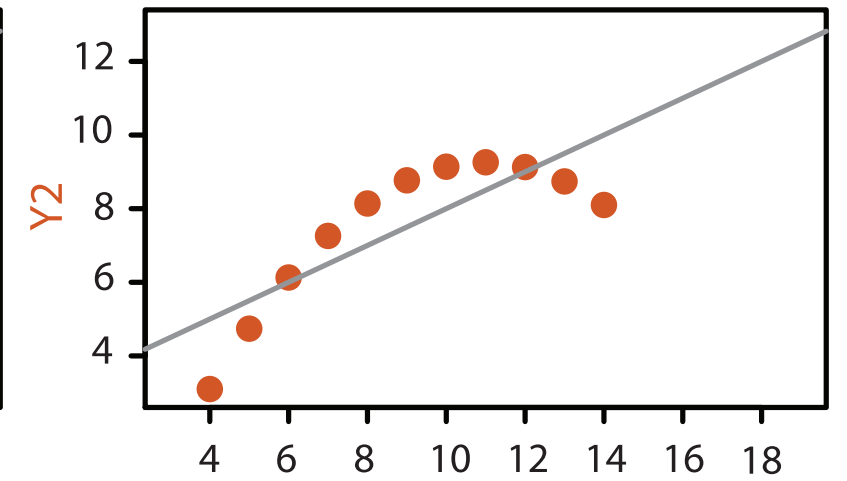
Anscombe's Quartet

Identical statistics

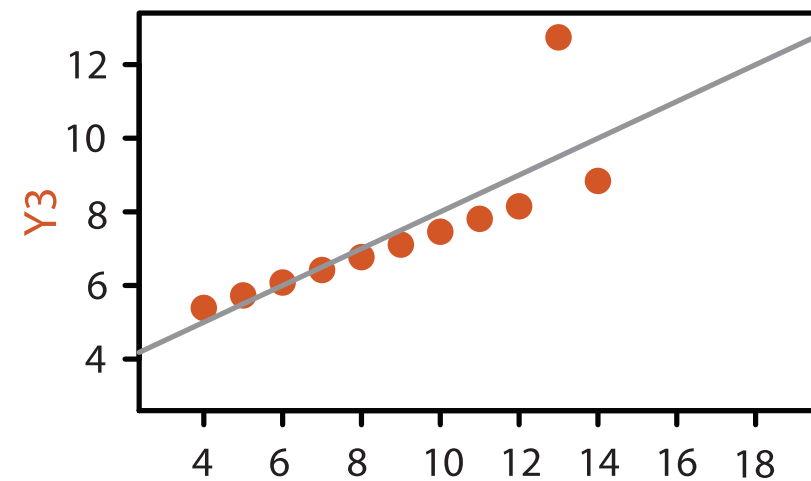
x mean	9
x variance	10
y mean	7.5
y variance	3.75
x/y correlation	0.816



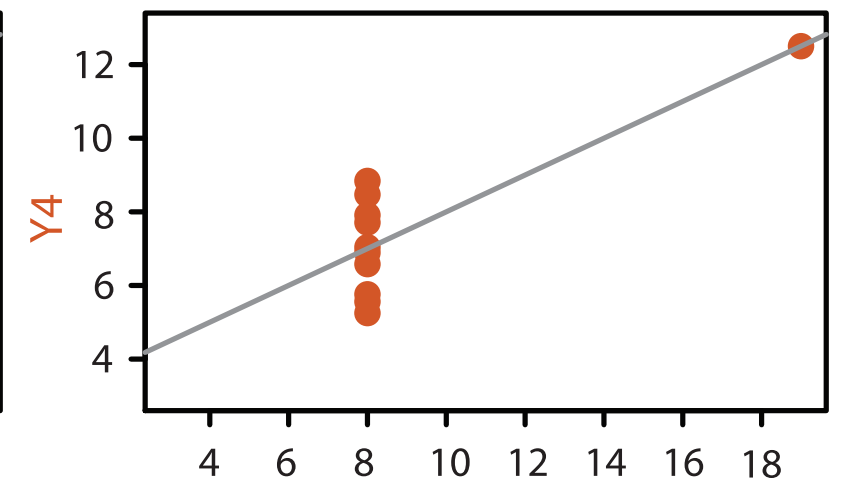
X1



X2



X3



X4

Why focus on tasks and effectiveness?

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

- tasks serve as constraint on design (as does data)
 - idioms do not serve all tasks equally!
 - challenge: recast tasks from domain-specific vocabulary to abstract forms
- most possibilities ineffective
 - validation is necessary, but tricky
 - increases chance of finding good solutions if you understand full space of possibilities
- what counts as effective?
 - novel: enable entirely new kinds of analysis
 - faster: speed up existing workflows

What resource limitations are we faced with?

Vis designers must take into account three very different kinds of resource limitations: those of computers, of humans, and of displays.

- computational limits
 - processing time
 - system memory
- human limits
 - human attention and memory
- display limits
 - pixels are precious resource, the most constrained resource
 - **information density**: ratio of space used to encode info vs unused whitespace
 - tradeoff between clutter and wasting space, find sweet spot between dense and sparse

Why analyze?

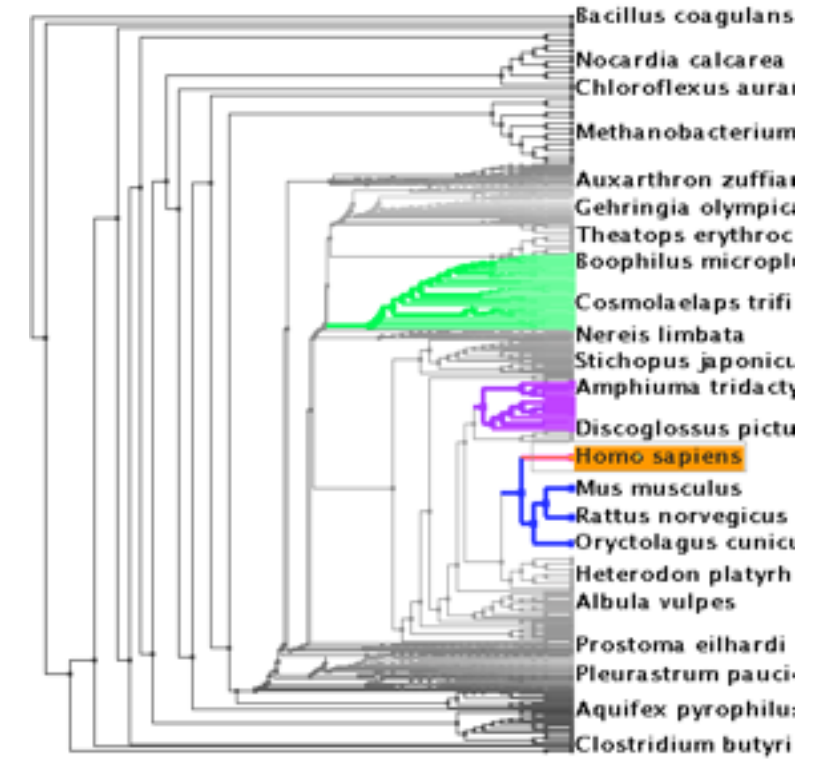
- imposes structure on huge design space
 - scaffold to help you think systematically about choices
 - analyzing existing as stepping stone to designing new
 - most possibilities ineffective for particular task/data combination

SpaceTree



[SpaceTree: Supporting Exploration in Large Node Link Tree, Design Evolution and Empirical Evaluation. Grosjean, Plaisant, and Bederson. Proc. InfoVis 2002, p 57–64.]

TreeJuxtaposer



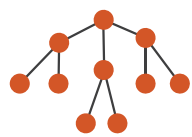
[TreeJuxtaposer: Scalable Tree Comparison Using Focus +Context With Guaranteed Visibility. ACM Trans. on Graphics (Proc. SIGGRAPH) 22:453– 462, 2003.]

What?

Why?

How?

→ Tree



→ Actions

→ Present → Locate → Identify



→ Targets

→ Path between two nodes



→ SpaceTree

→ Encode → Navigate → Select → Filter → Aggregate



→ TreeJuxtaposer

→ Encode → Navigate → Select → Arrange



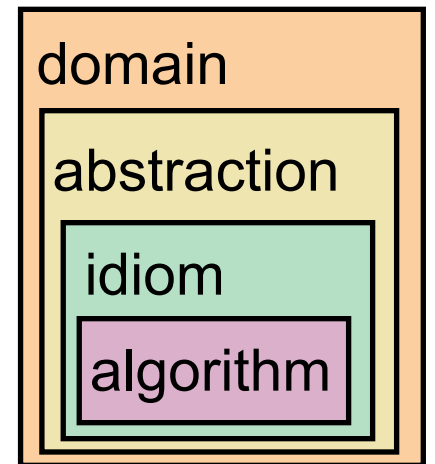
What?

Why?

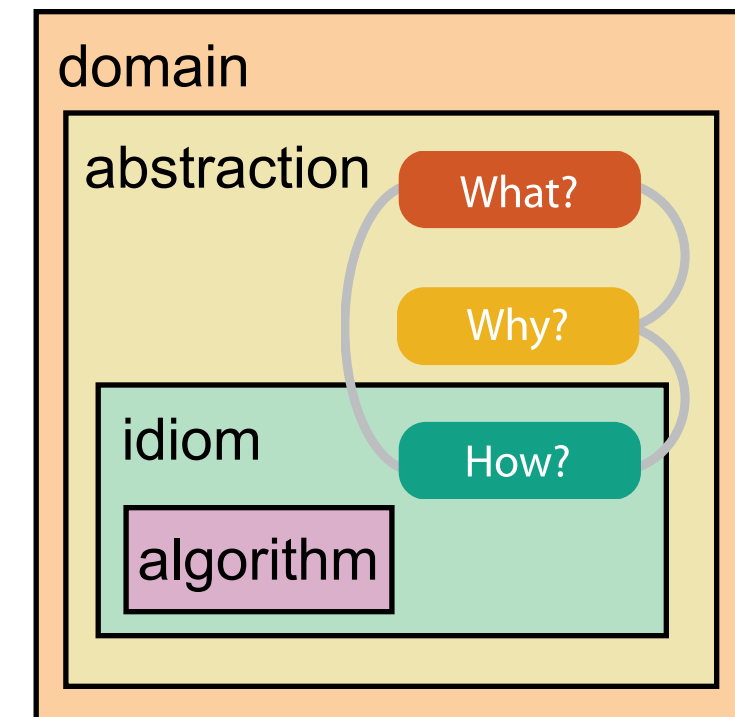
How?

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



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

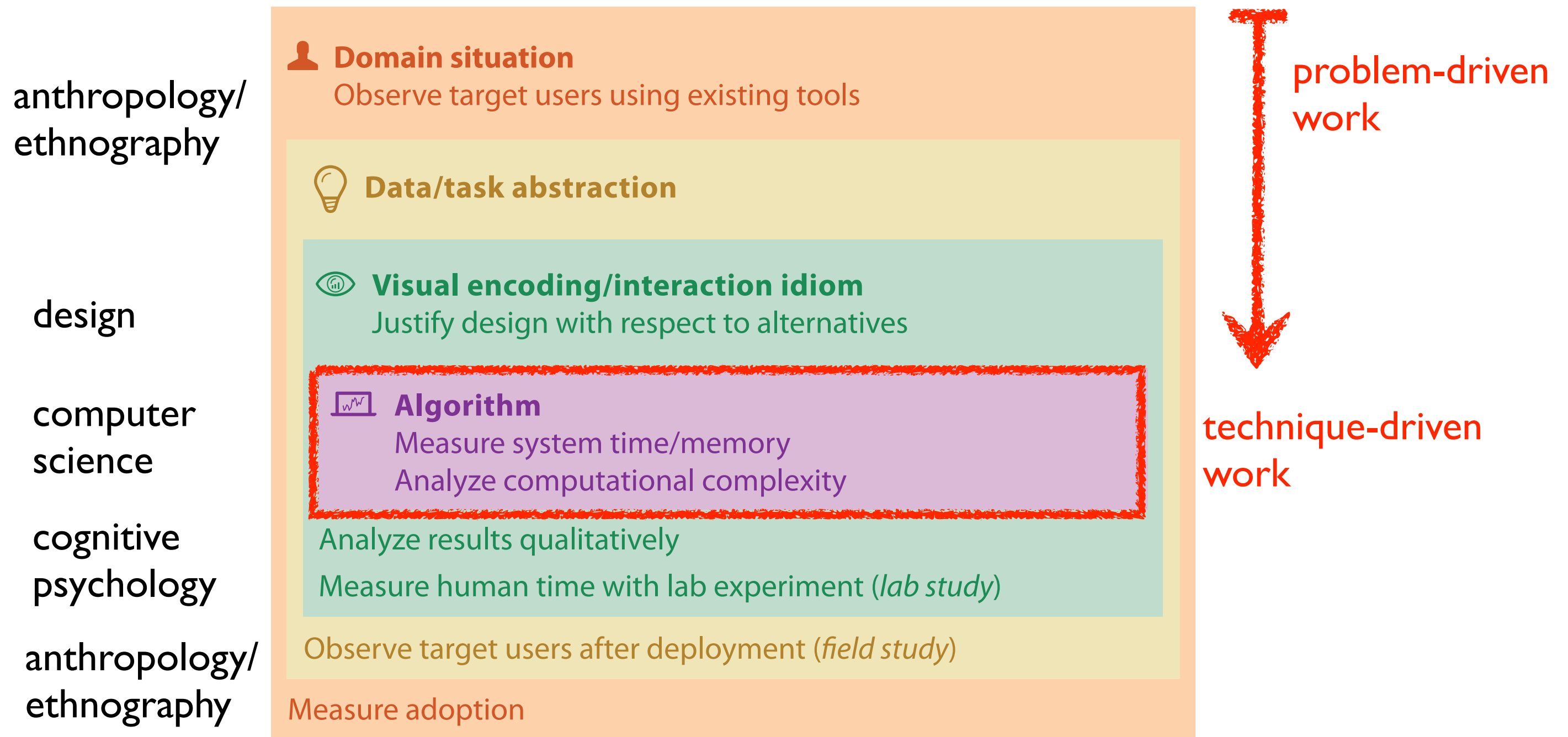


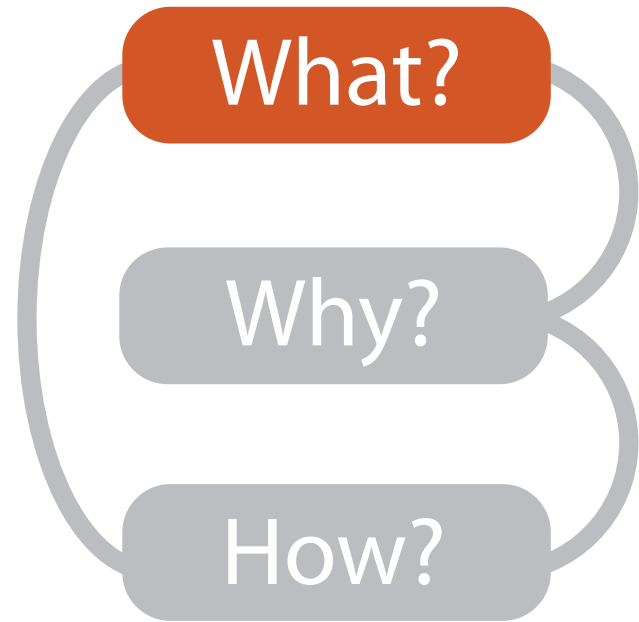
Algorithm

Your code is too slow

Why is validation difficult?

- solution: use methods from different fields at each level





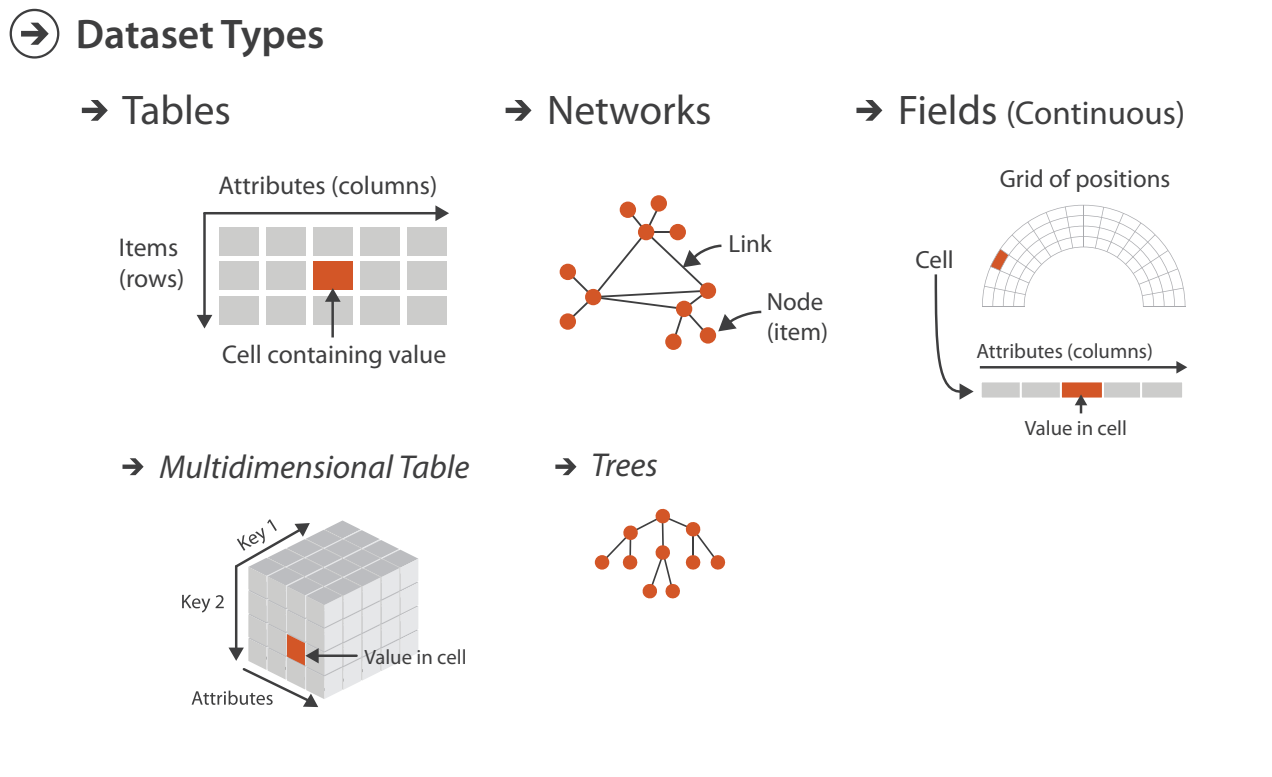
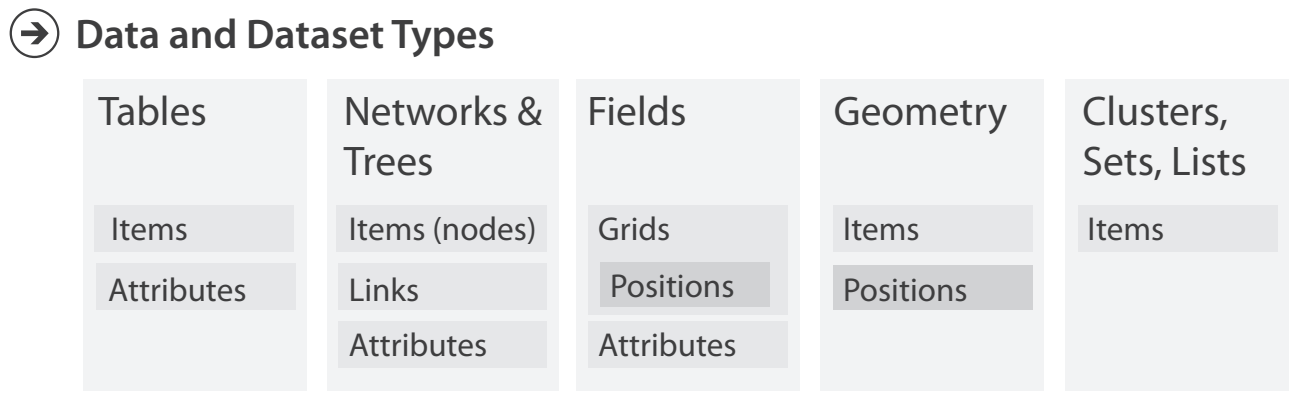
What?

Datasets

Attributes

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

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



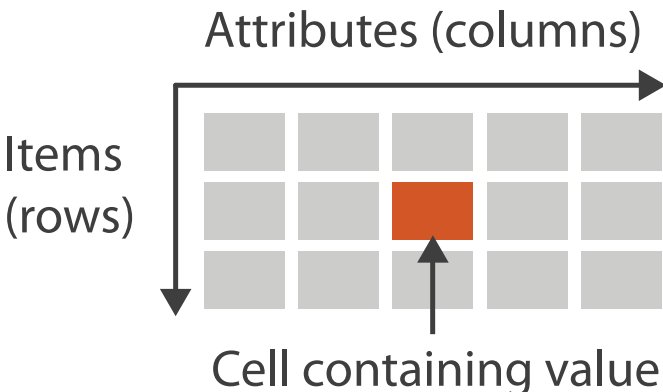
- Ordering Direction
 - Sequential
 -
 - Diverging
 - ←→
 - Cyclic
 - ↻



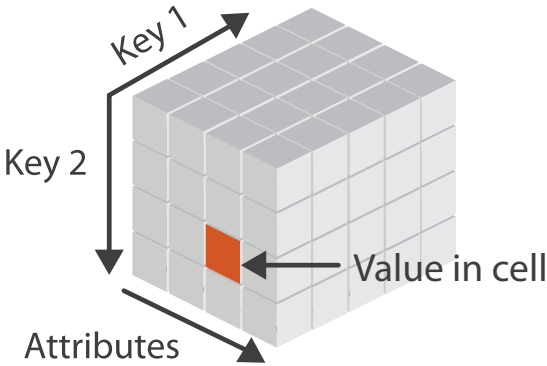
Three major datatypes

→ Dataset Types

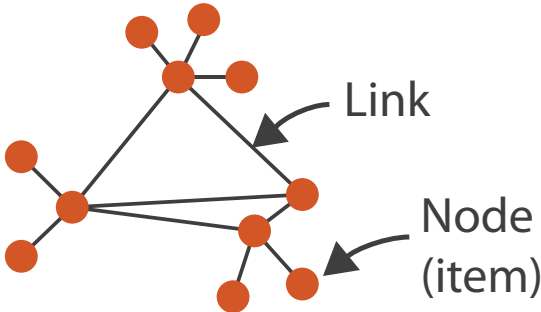
→ Tables



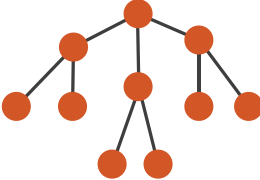
→ Multidimensional Table



→ Networks

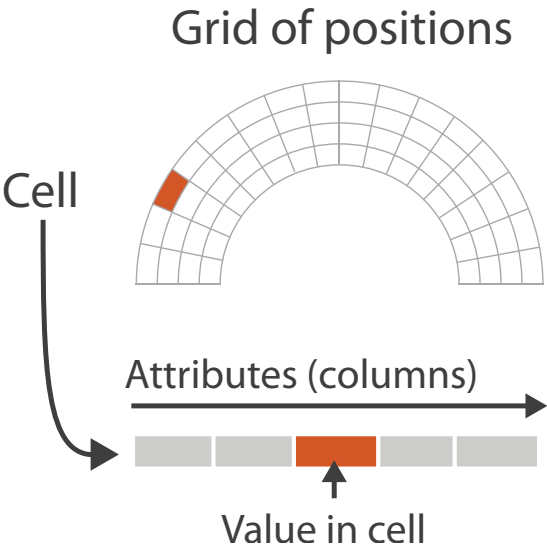


→ Trees

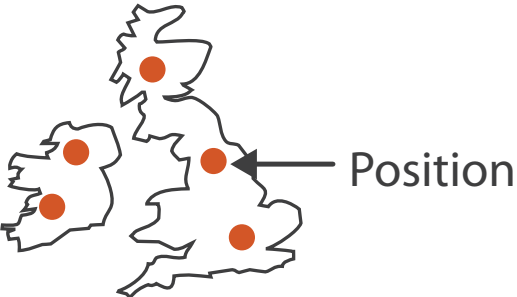


→ Spatial

→ Fields (Continuous)



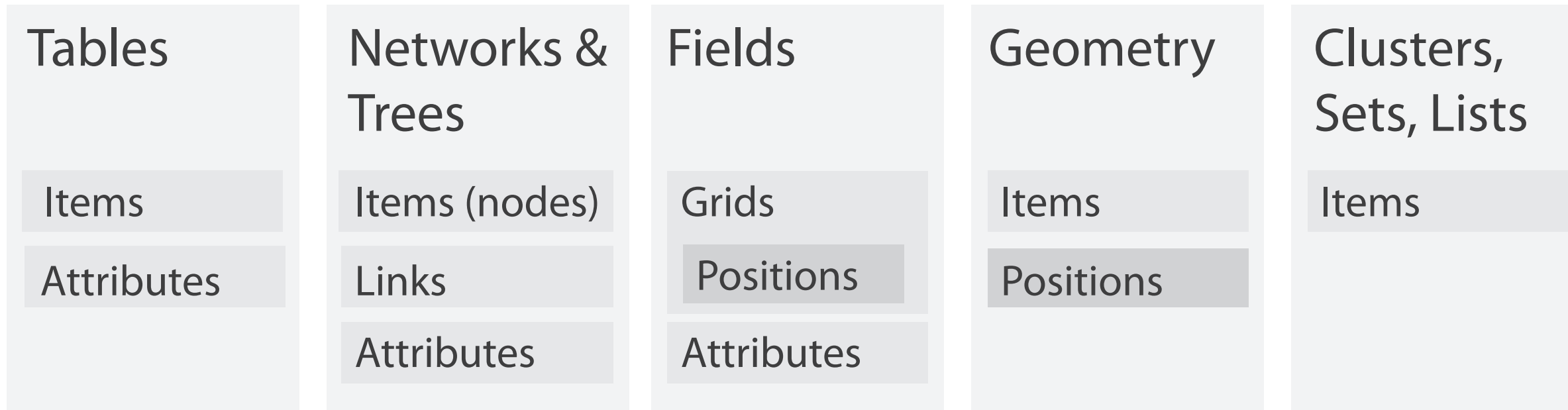
→ Geometry (Spatial)



- visualization vs computer graphics
 - geometry is design decision

Dataset and data types

→ Data and Dataset Types



→ Data Types

→ Items → Attributes → Links → Positions → Grids

→ Dataset Availability

→ Static



→ Dynamic



Attribute types

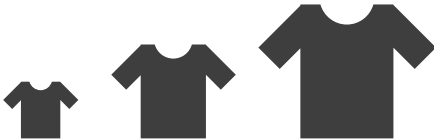
➔ Attribute Types

➔ Categorical

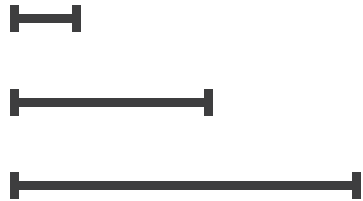


➔ Ordered

➔ *Ordinal*



➔ *Quantitative*



➔ Ordering Direction

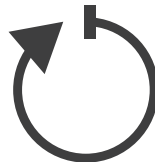
➔ Sequential

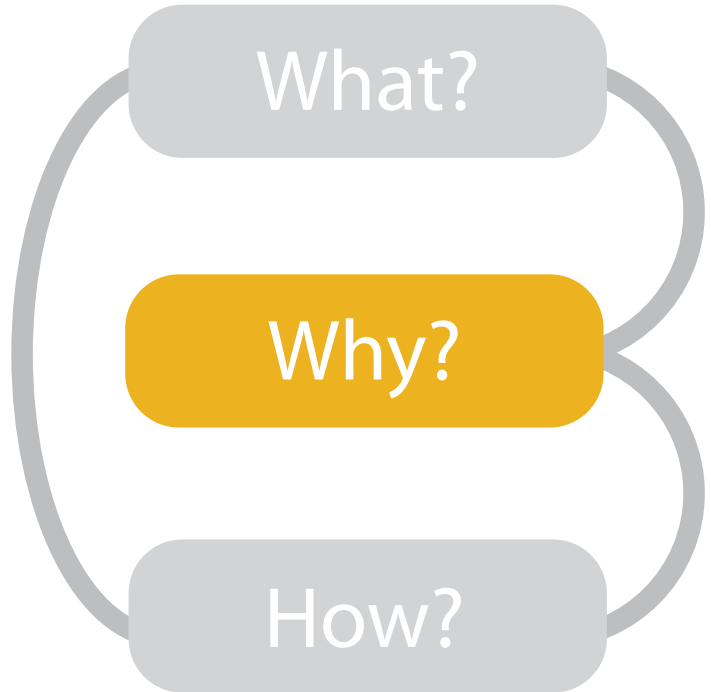


➔ Diverging



➔ 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	Lookup	Browse
Location unknown	Locate	Explore

➔ **Query**

- ➔ Identify
- ➔ Compare
- ➔ Summarize

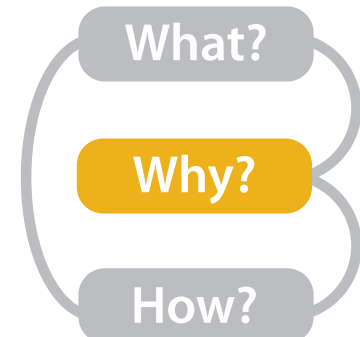
➔ **Network Data**

- ➔ 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

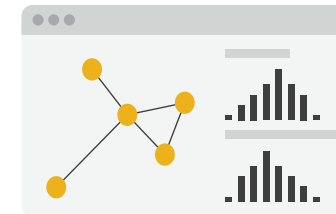
→ Analyze

→ Consume

→ Discover



→ Present

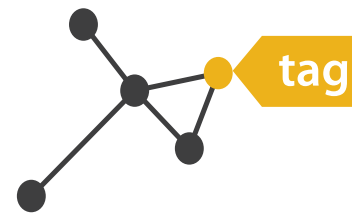


→ Enjoy



→ Produce

→ Annotate



→ Record

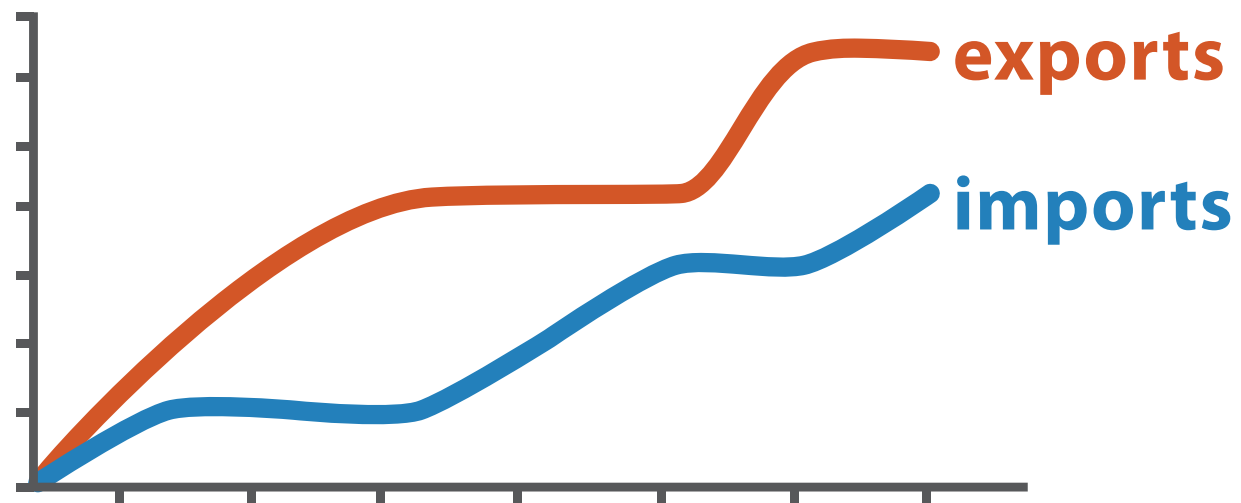


→ Derive

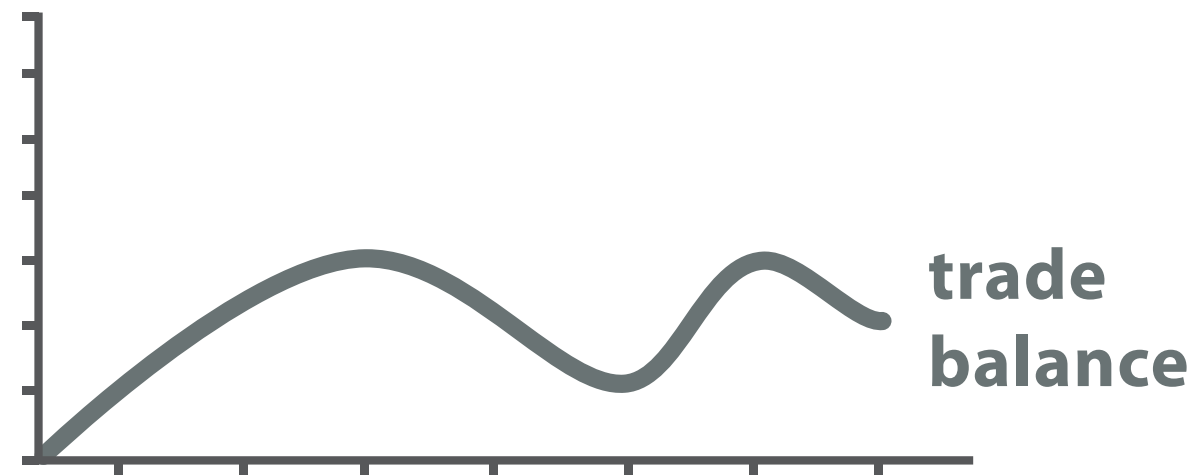


Derive

- 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





Actions: Search, query

- what does user know? → Search

- target, location

- how much of the data matters?

- one, some, all

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

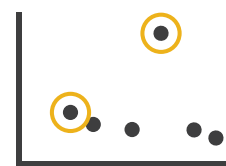
→ Query

- independent choices for each of these three levels

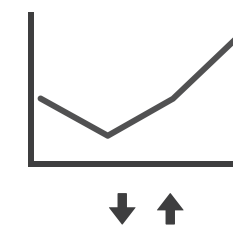
- analyze, search, query

- mix and match

→ Identify



→ Compare



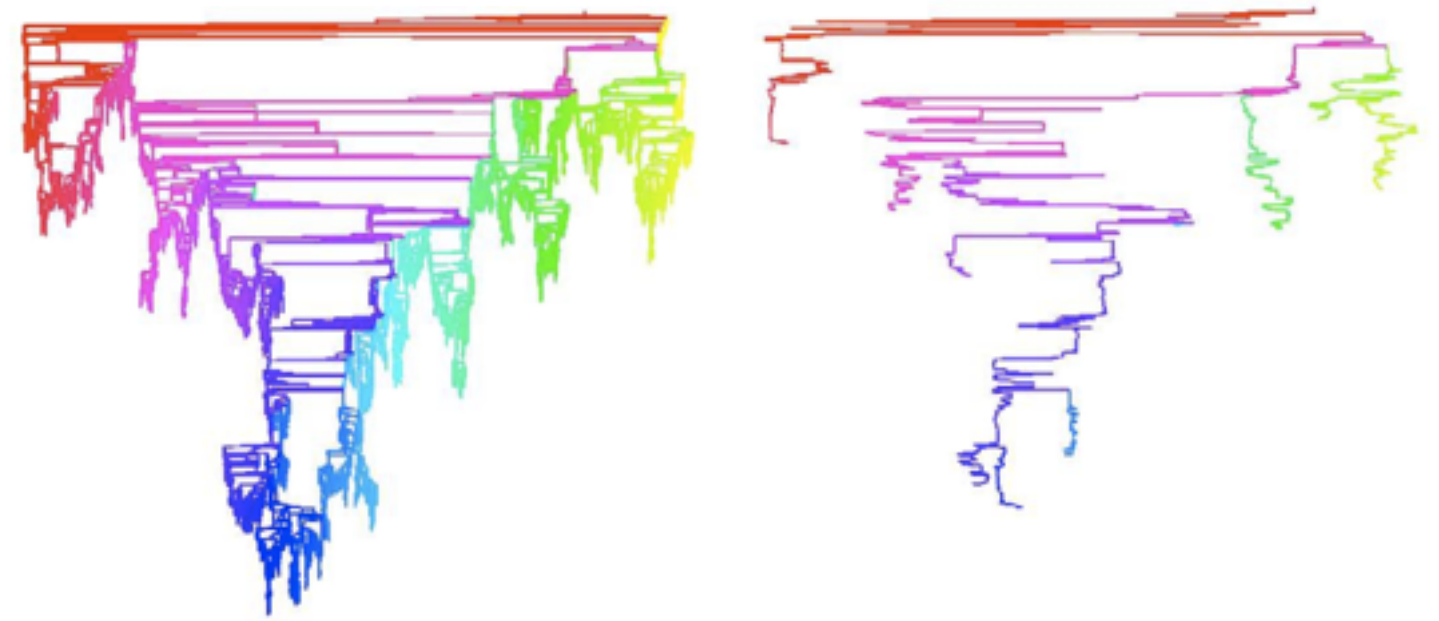
→ Summarize



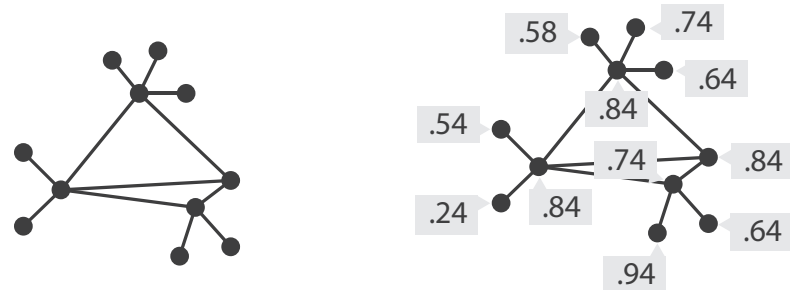
Analysis example: Derive one attribute

- 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

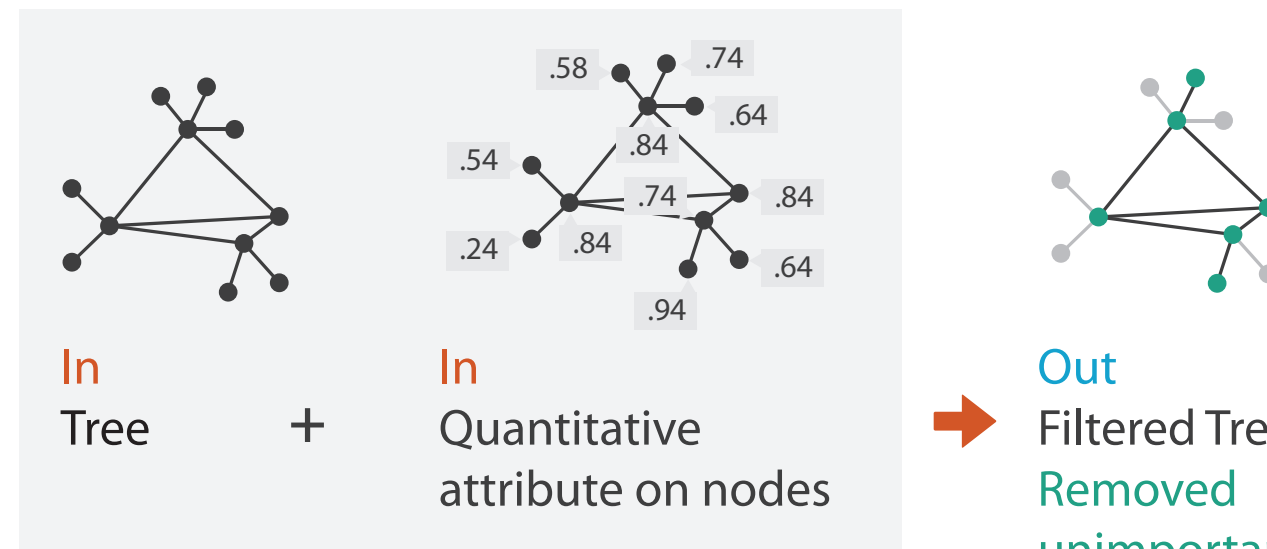
What?

- ➔ 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

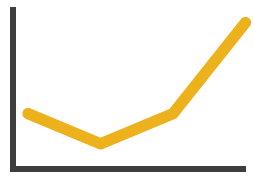
How?

- ➔ Reduce
- ➔ Filter

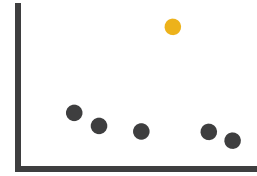
Why: Targets

→ All Data

→ Trends



→ Outliers



→ Features



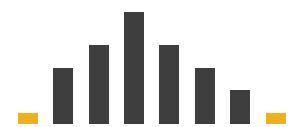
→ Attributes

→ One

→ *Distribution*



→ *Extremes*

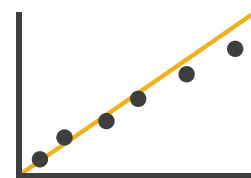


→ Many

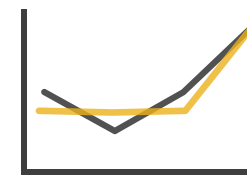
→ *Dependency*



→ *Correlation*

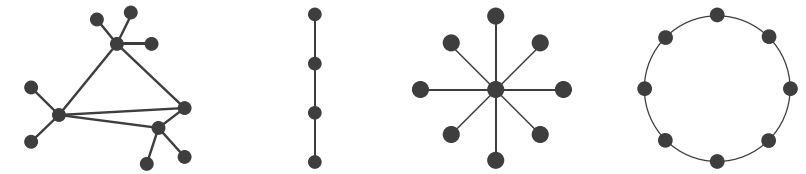


→ *Similarity*



→ Network Data

→ Topology

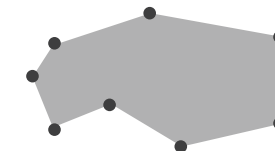


→ *Paths*



→ 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



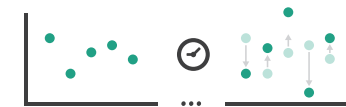
→ Motion

Direction, Rate, Frequency, ...

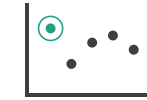


Manipulate

→ Change



→ Select



→ Navigate

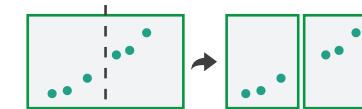


Facet

→ Juxtapose



→ Partition



→ Superimpose



Reduce

→ Filter



→ Aggregate



→ Embed



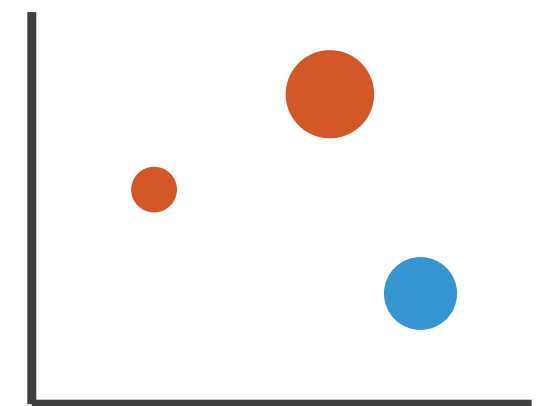
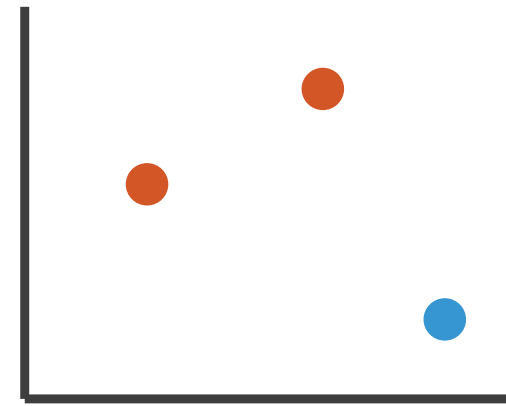
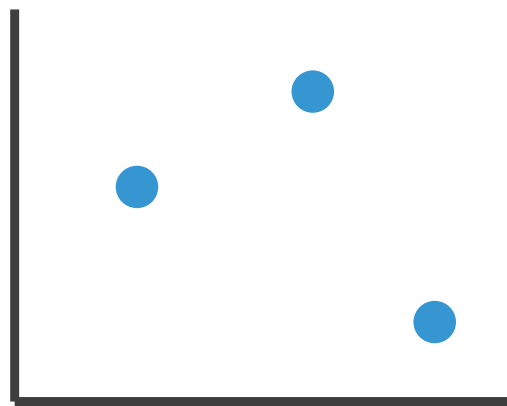
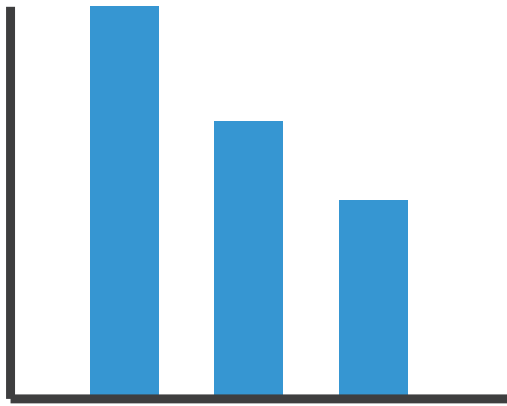
What?

Why?

How?

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

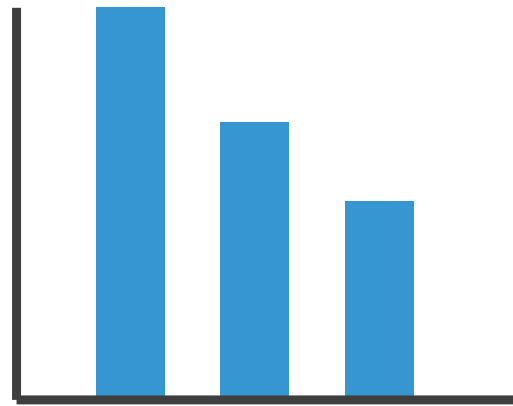


→ Volume



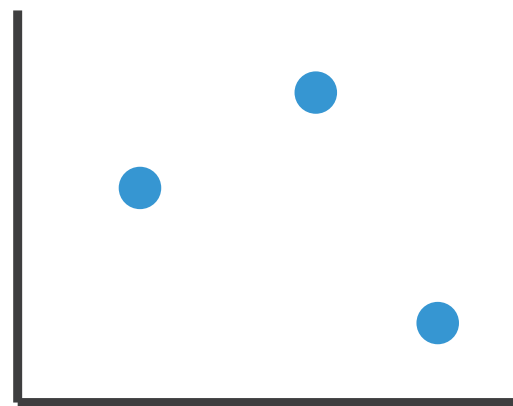
Encoding visually with marks and channels

- analyze idiom structure
 - as combination of marks and channels



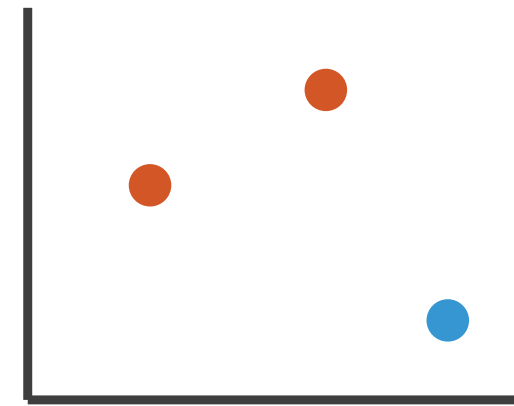
1:
vertical position

mark: line



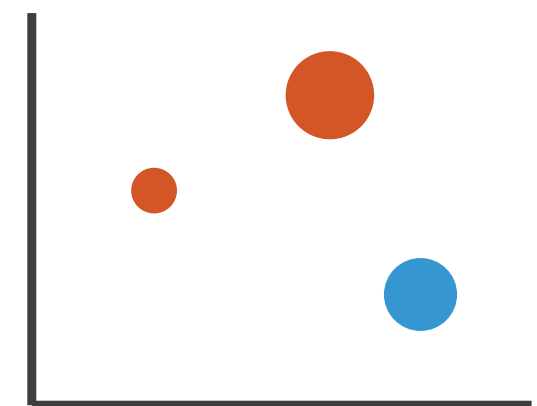
2:
vertical position
horizontal position

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

Same

Spatial region



Color hue



Motion



Shape



Channels: Rankings

➔ Magnitude Channels: Ordered Attributes



➔ Identity Channels: Categorical Attributes



Best
Effectiveness
Least

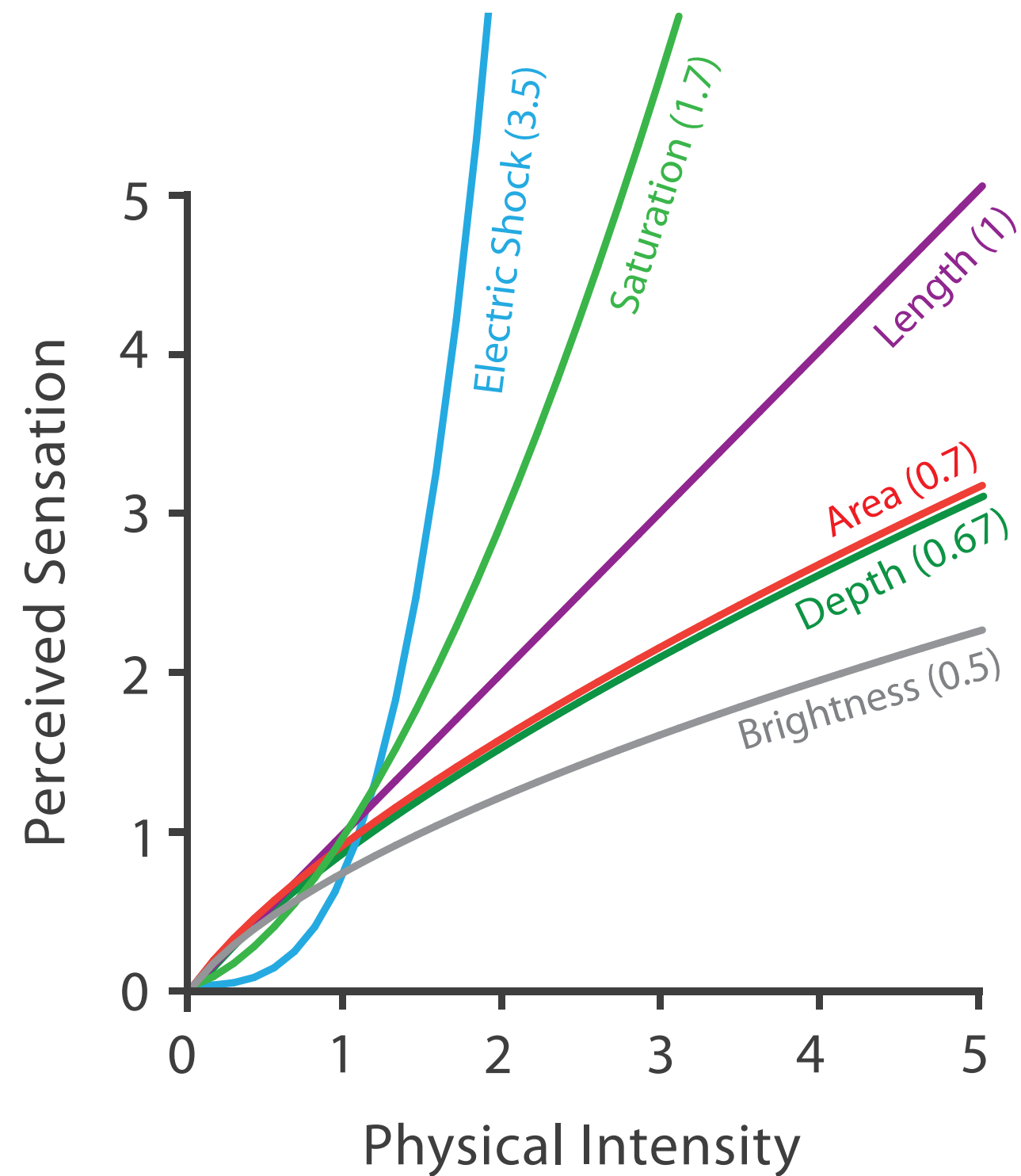
Same

Same

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

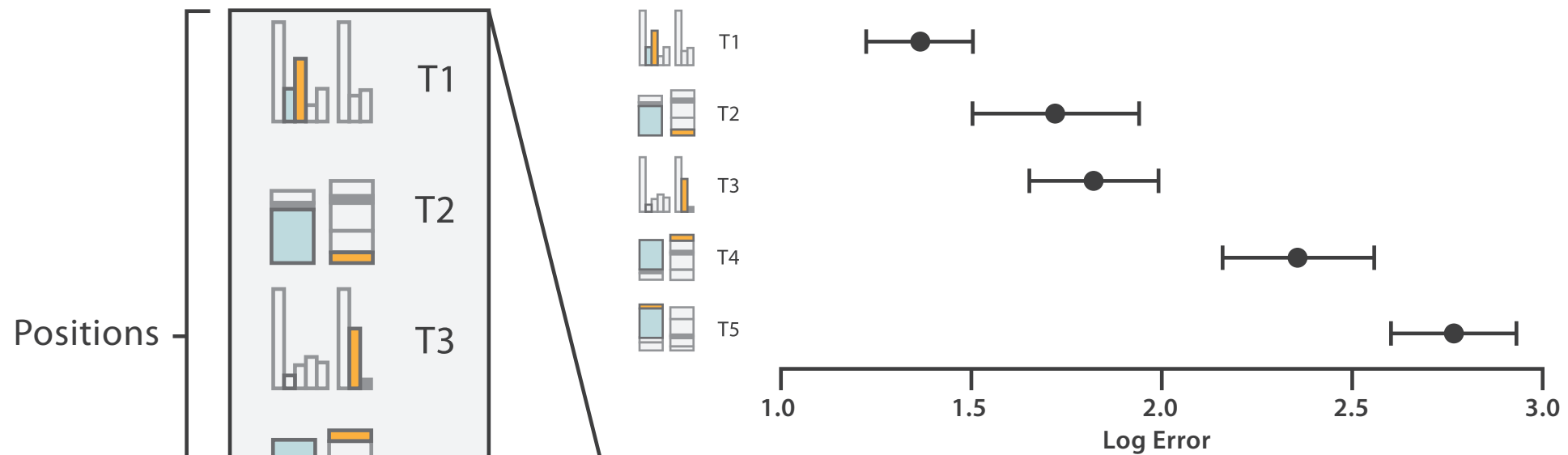
Accuracy: Fundamental Theory

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

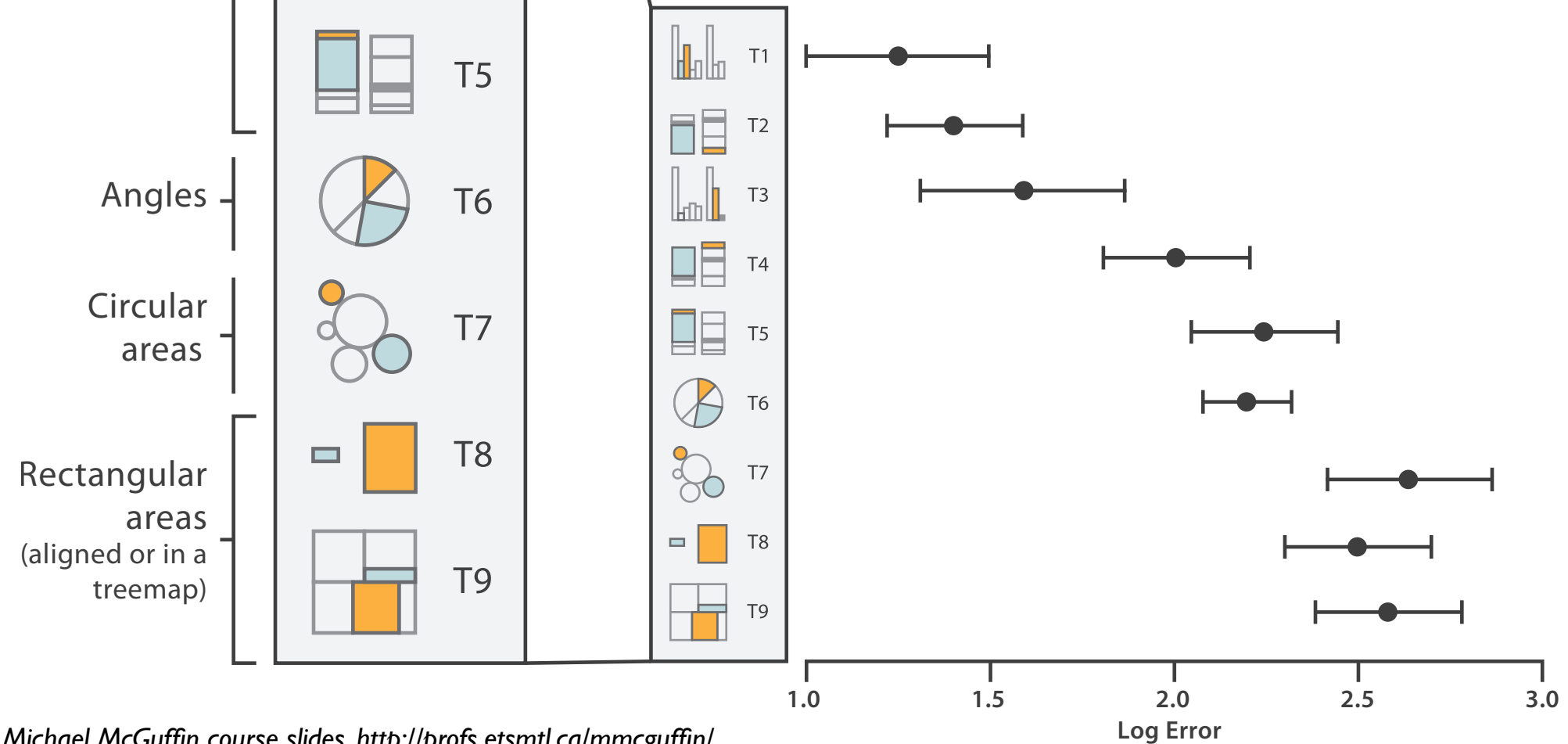


Accuracy: Vis experiments

Cleveland & McGill's Results



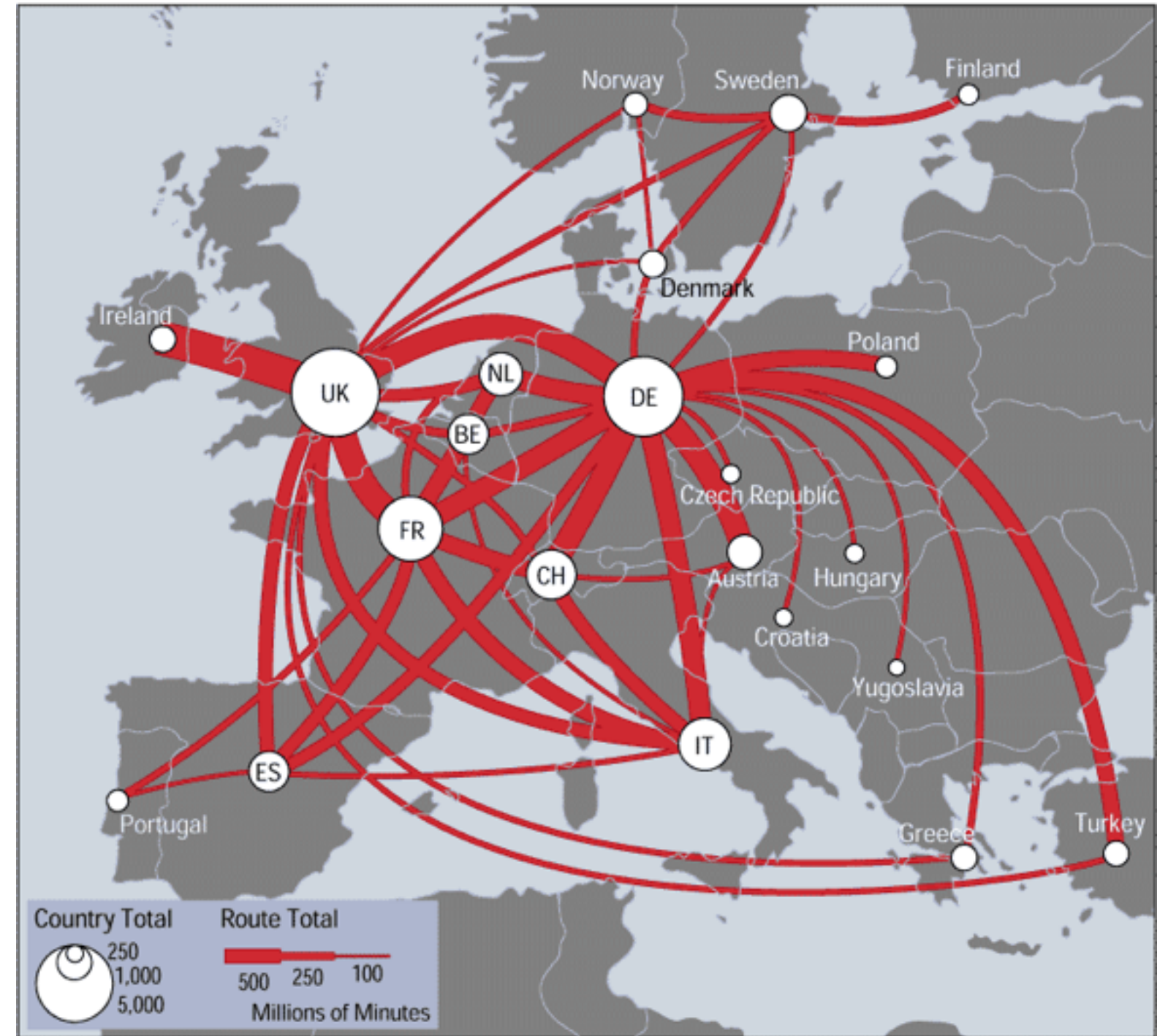
Crowdsourced Results



[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?

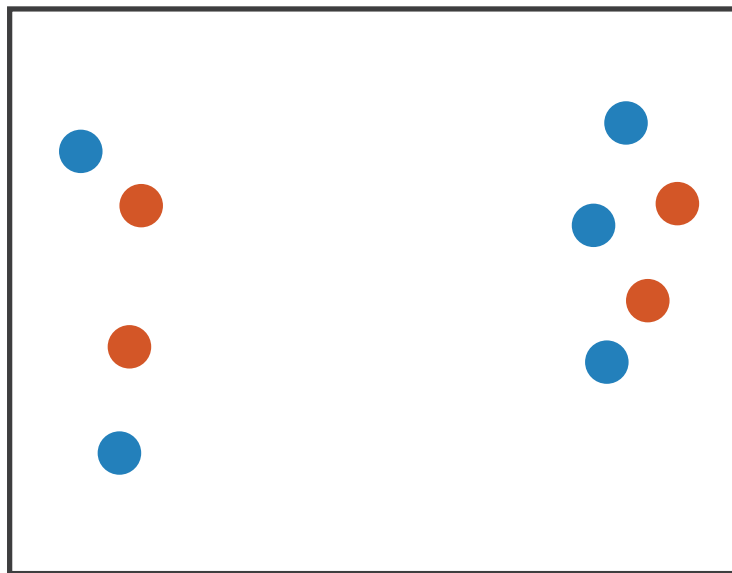
- must be sufficient for number of attribute levels to show
 - linewidth: few bins



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

Separability vs. Integrality

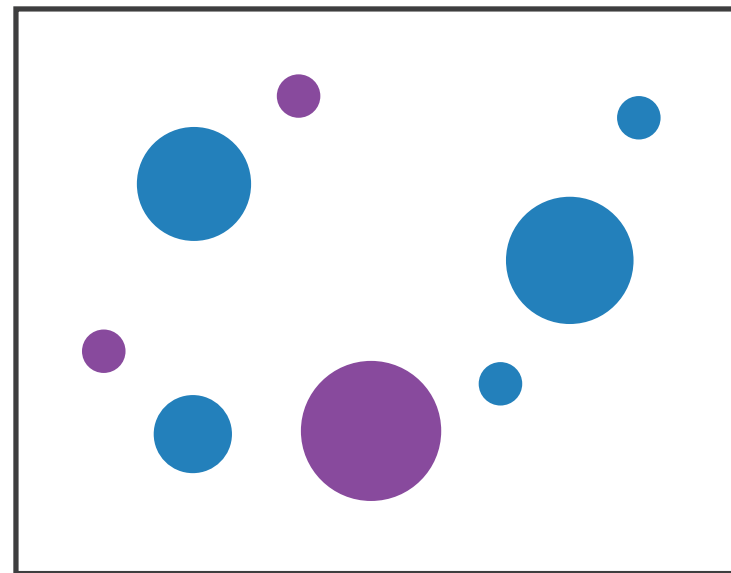
Position
+ Hue (Color)



Fully separable

2 groups each

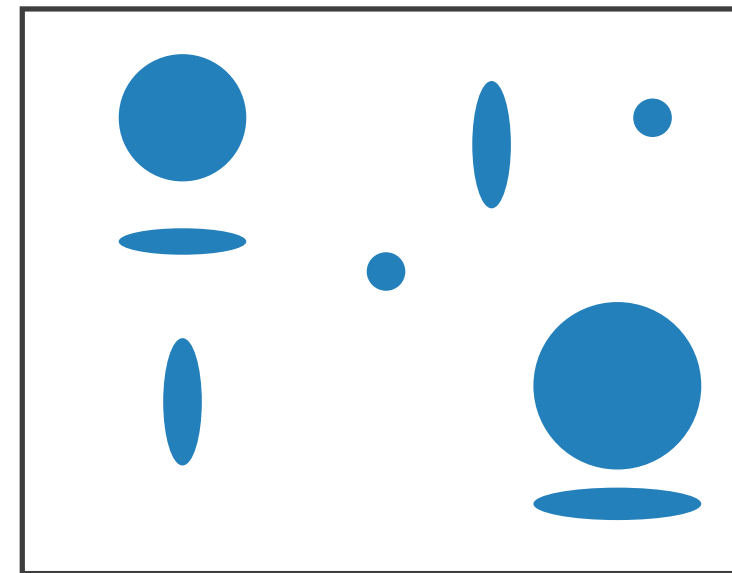
Size
+ Hue (Color)



Some interference

2 groups each

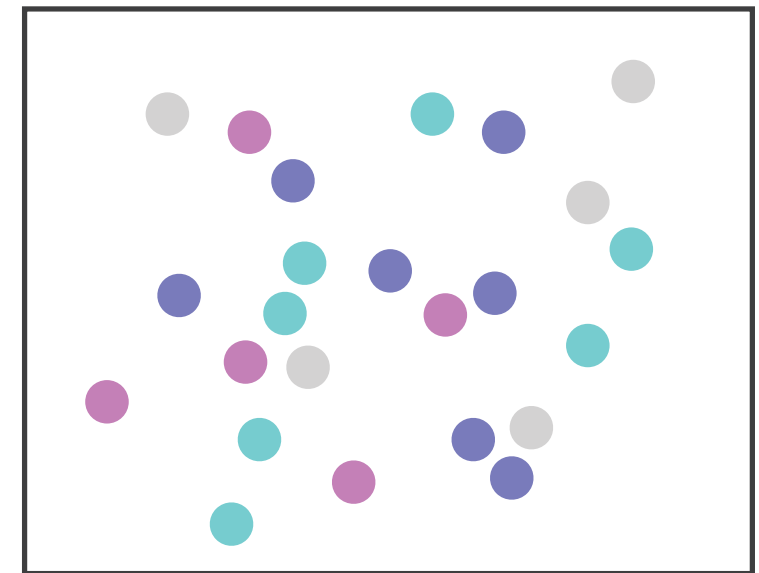
Width
+ Height



Some/significant
interference

3 groups total:
integral area

Red
+ Green

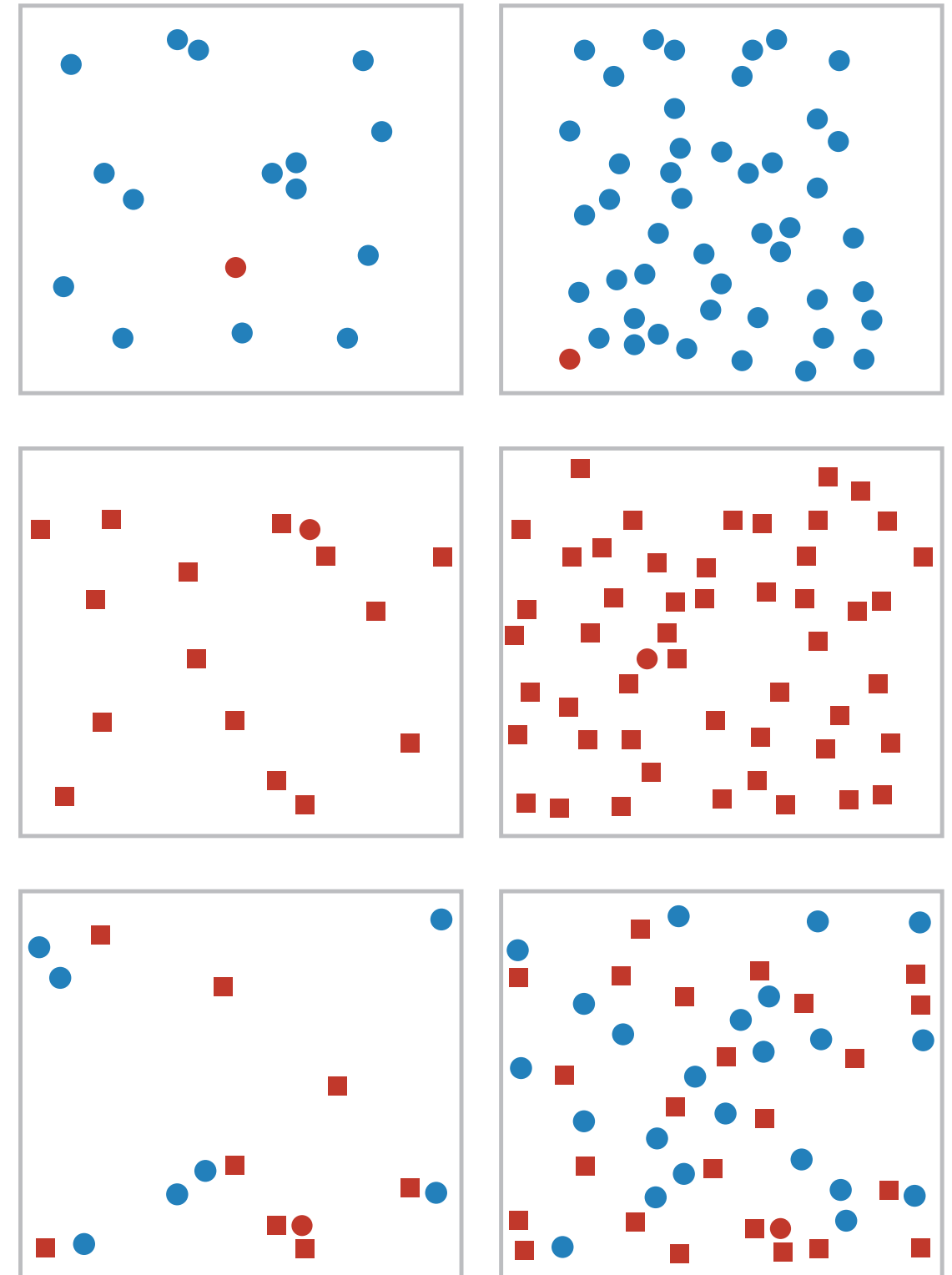


Major interference

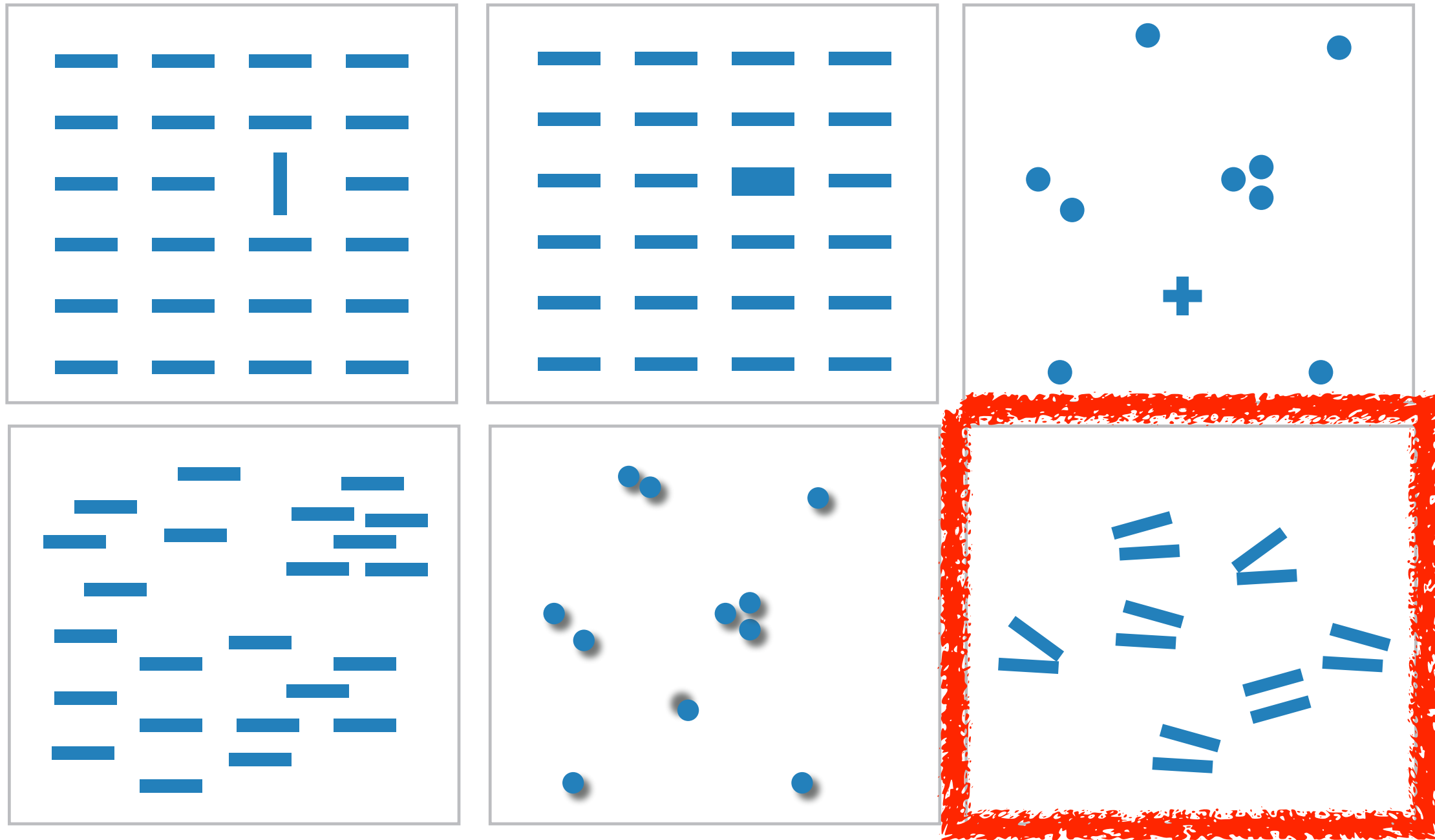
4 groups total:
integral hue

Popout

- find the red dot
 - how long does it take?
- parallel processing on many individual channels
 - speed independent of distractor count
 - speed depends on channel and amount of difference from distractors
- serial search for (almost all) combinations
 - speed depends on number of distractors



Popout



- many channels: tilt, size, shape, proximity, shadow direction, ...
- but not all! parallel line pairs do not pop out from tilted pairs

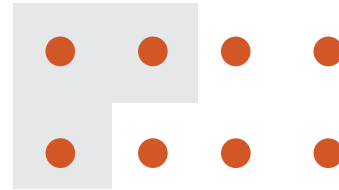
Grouping

- containment
- connection

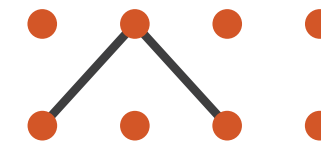
- proximity
 - same spatial region
- similarity
 - same values as other categorical channels

Marks as Links

➔ Containment



➔ Connection



➔ Identity Channels: Categorical Attributes

Spatial region



Color hue



Motion



Shape

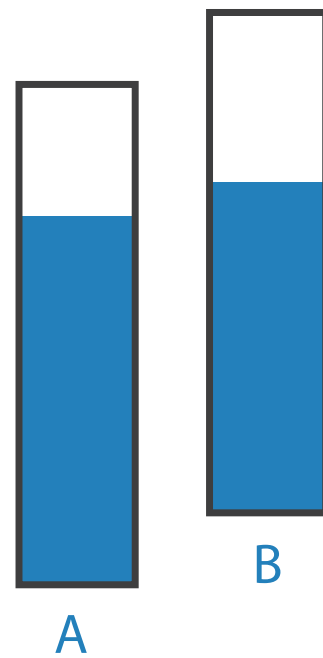


Relative vs. absolute judgements

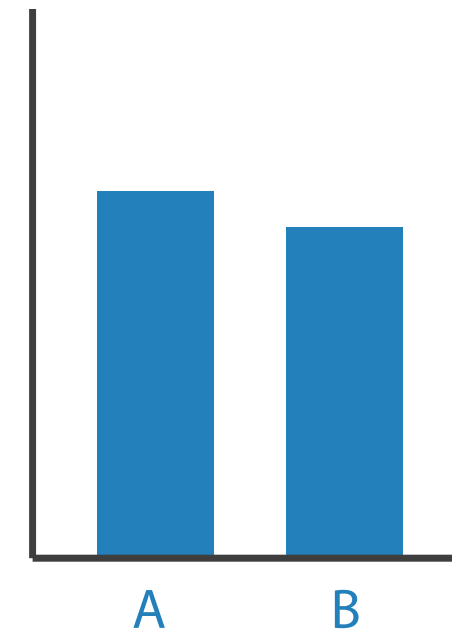
- perceptual system mostly operates with relative judgements, not absolute
 - that's why accuracy increases with common frame/scale and alignment
 - Weber's Law: ratio of increment to background is constant
 - filled rectangles differ in length by 1:9, difficult judgement
 - white rectangles differ in length by 1:2, easy judgement



length



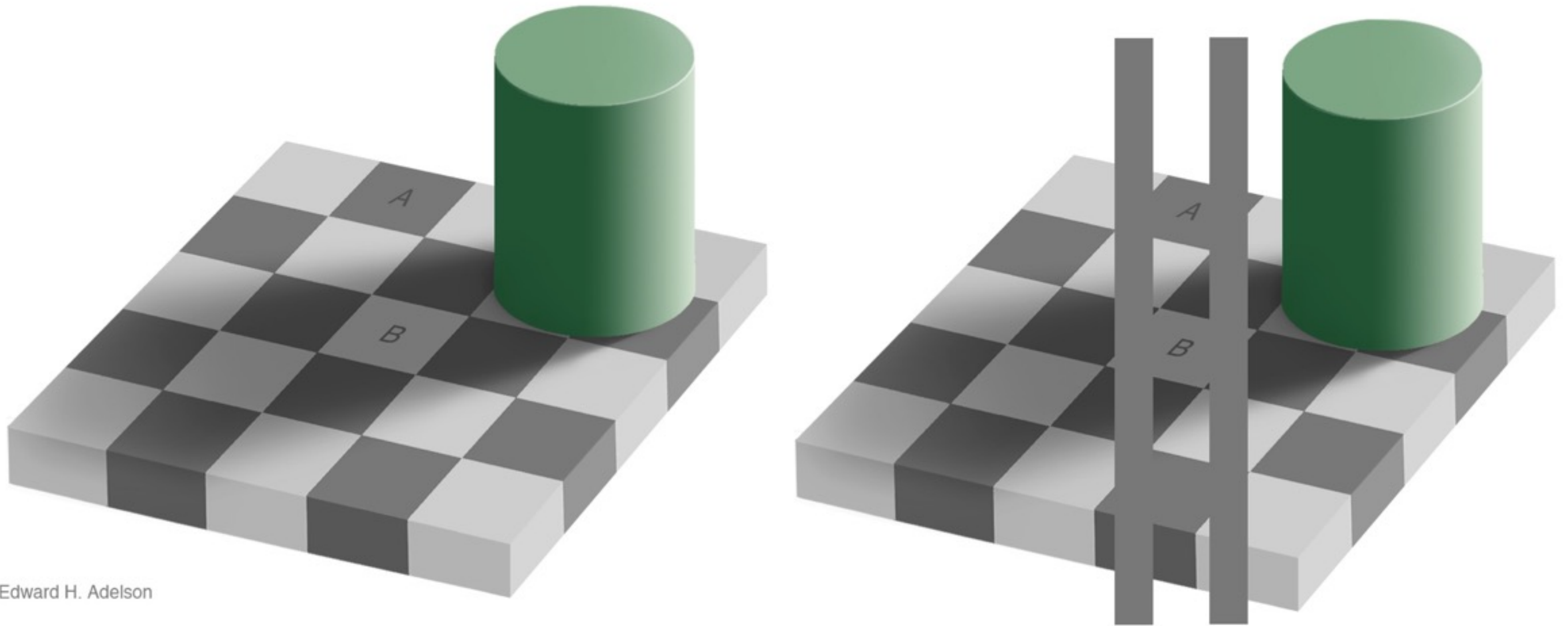
position along
unaligned
common scale



position along
aligned scale

Relative luminance judgements

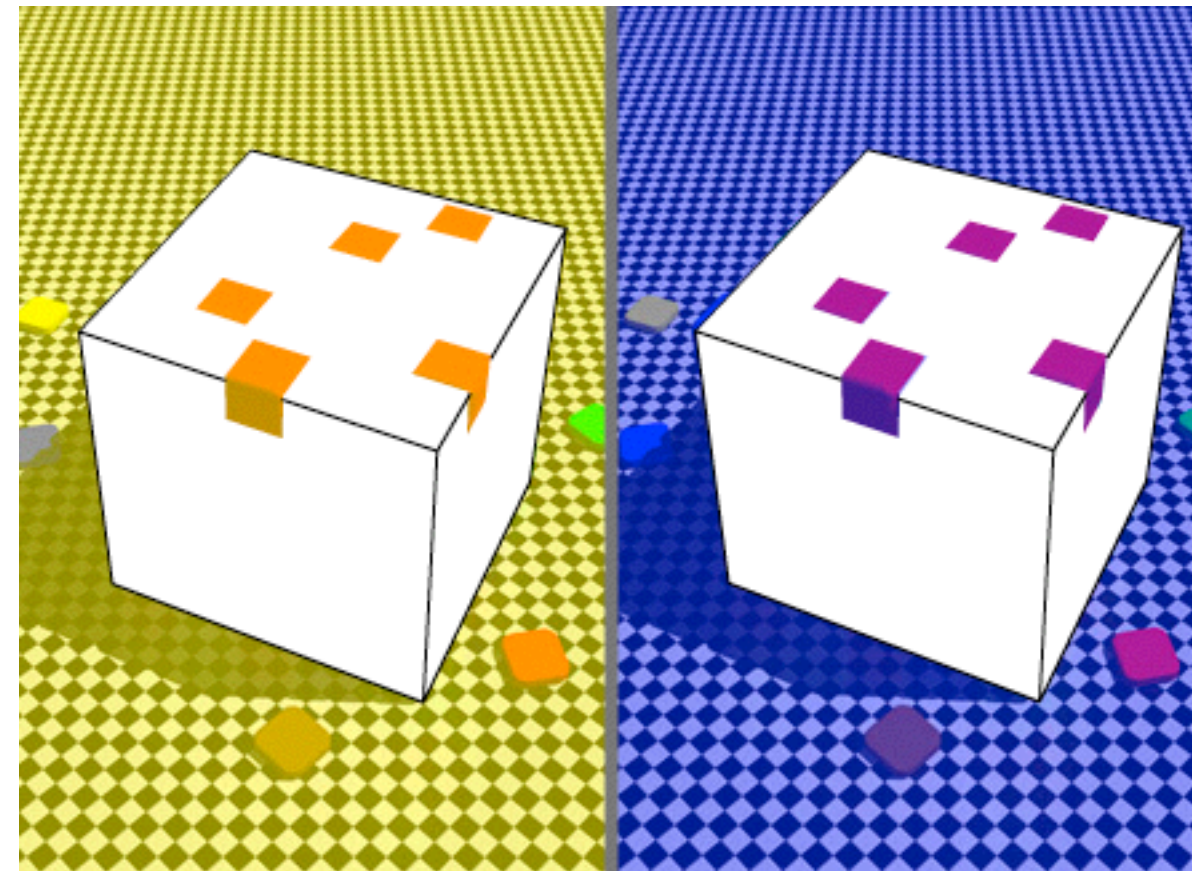
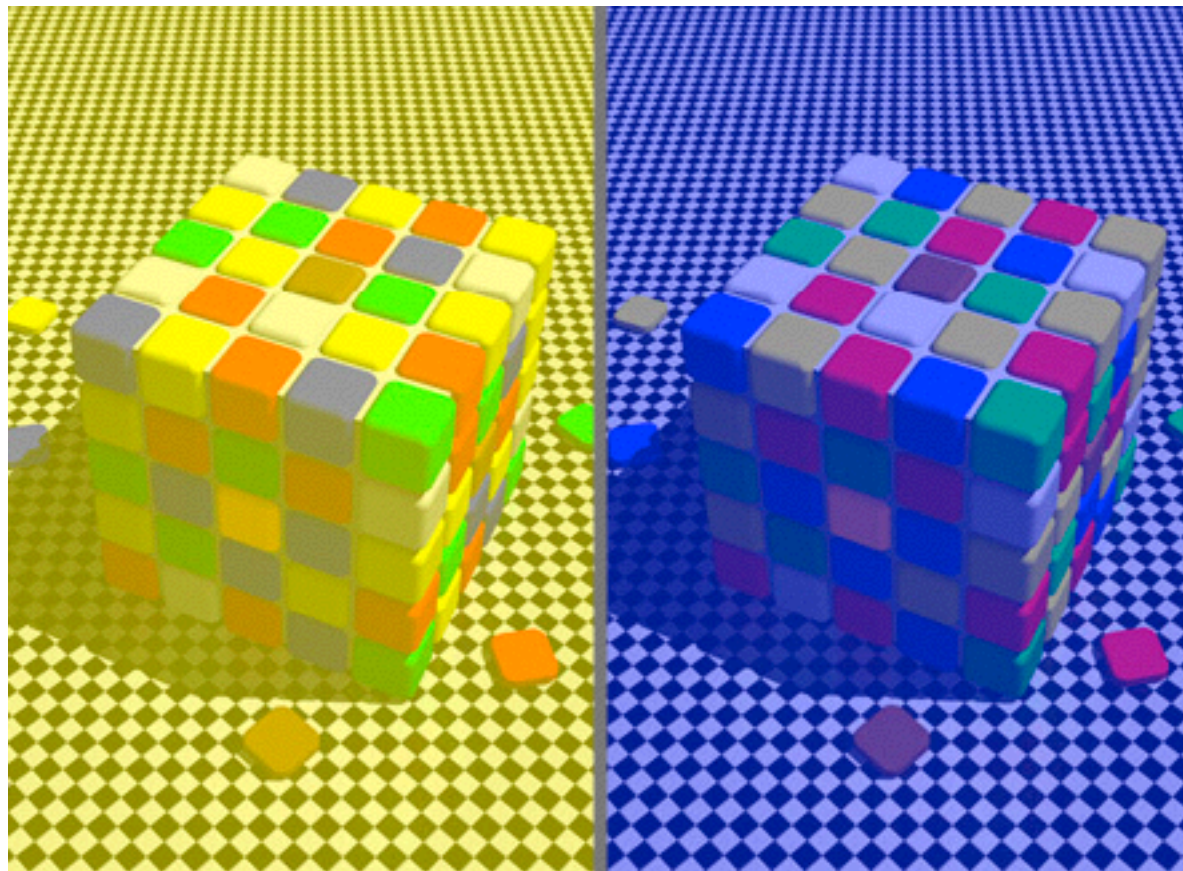
- perception of luminance is contextual based on contrast with surroundings



Edward H. Adelson

Relative color judgements

- color constancy across broad range of illumination conditions



Further reading

- Visualization Analysis and Design. Tamara Munzner. CRC Press, 2014.
 - *Chap 1, What's Vis, and Why Do It?*
 - *Chap 2, What: Data Abstraction*
 - *Chap 3, Why: Task Abstraction*
 - *Chap 4, Analysis: Four Levels for Validation*
 - *Chap 5, Marks and Channels*
- Crowdsourcing Graphical Perception: Using Mechanical Turk to Assess Visualization Design. Jeffrey Heer and Michael Bostock. Proc. CHI 2010
- Perception in Vision web page with demos, Christopher Healey.
- Visual Thinking for Design. Colin Ware. Morgan Kaufmann, 2008.

Next

- Break (15 min)
- Demos (45 min)
 - Caitlin will walk through Tableau demos
 - you follow along step by step on your own laptop
 - Tamara will rove the room to help out folks who get stuck
- Lab (30 min)
 - you'll get started on Tableau assignment

Demo 1: Basic Visual Encoding & Dashboarding

- Tableau Lessons

- Dimensions (**categorical**) and Measures (**quantitative**)
- drag and drop to create visual encodings
- combining multiple charts side by side into dashboards

- Big Ideas

- see different patterns with different visual encodings

Demo 2: Vancouver Election Results

- Tableau Lessons
 - sorting along axis
 - disaggregate into multiple charts

- Big Ideas
 - absolute numbers can sometimes mislead
 - check hunches with relative percentages!

Demo 3: Vancouver Crime

- Tableau Lessons
 - multiple pills on a shelf, pill ordering
 - show filters
 - undo
 - duplicate & rename tabs
- Big Ideas
 - underlying causes can be tricky to understand

Demo 4: Back to the Future

- Tableau Lessons
 - simple analytics: totals
 - more disaggregation practice
 - Show Me

- Big Ideas
 - beyond simple bars
 - challenges of missing data

Assignment

- Music Sales
 - work through workbook on your own
 - submit finished version (in workbook .twbx format)
- Vancouver Crime
 - analyze further on your own
 - write up brief news story (submit in PDF format)
 - < 500 words
 - up to 2 screenshots from Tableau
 - write up reflections (submit in PDF format)
 - discuss dead ends
 - include Tableau screenshots
- submit before next class (9am Tue Sep 20)
 - email tmm@cs.ubc.ca and caitlin@discoursemedia.org with subject JOURN Week 1