

# 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*

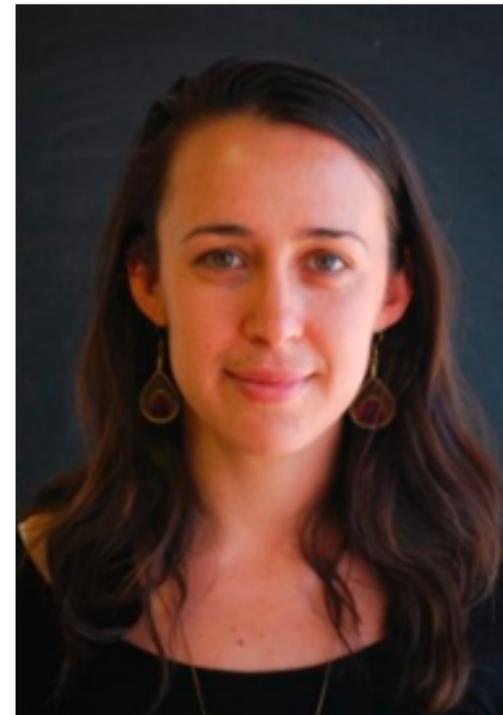
<http://www.cs.ubc.ca/~tmm/courses/journ16>

# 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](mailto:tmm@cs.ubc.ca), [caitlin@discoursemedia.org](mailto: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.**

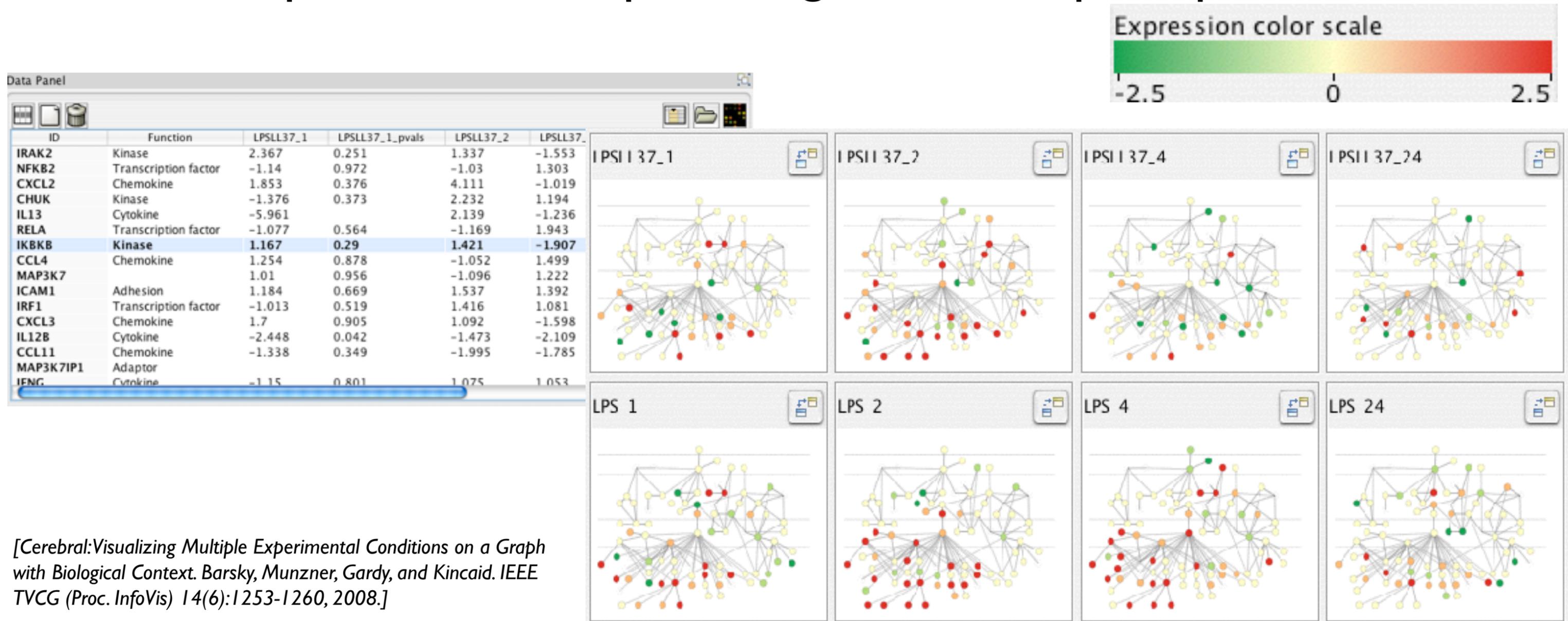
**Visualization is suitable when there is a need to augment human capabilities rather than replace people with computational decision-making methods.**

- 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

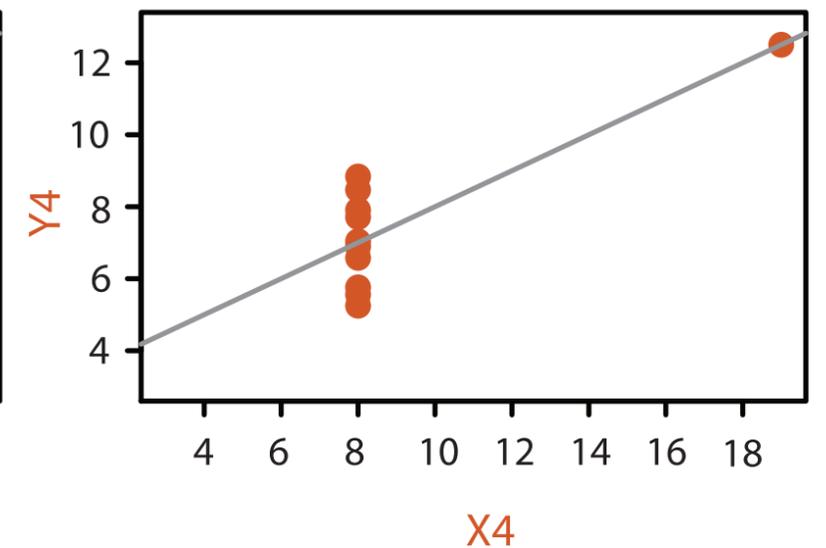
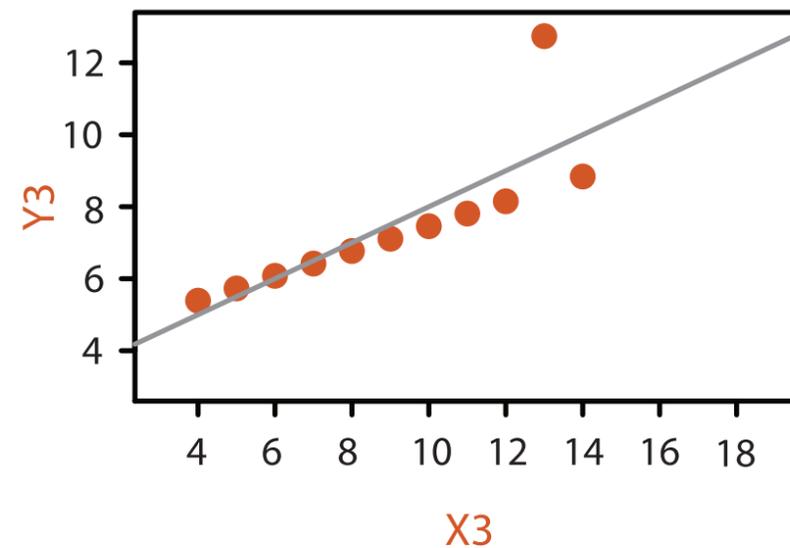
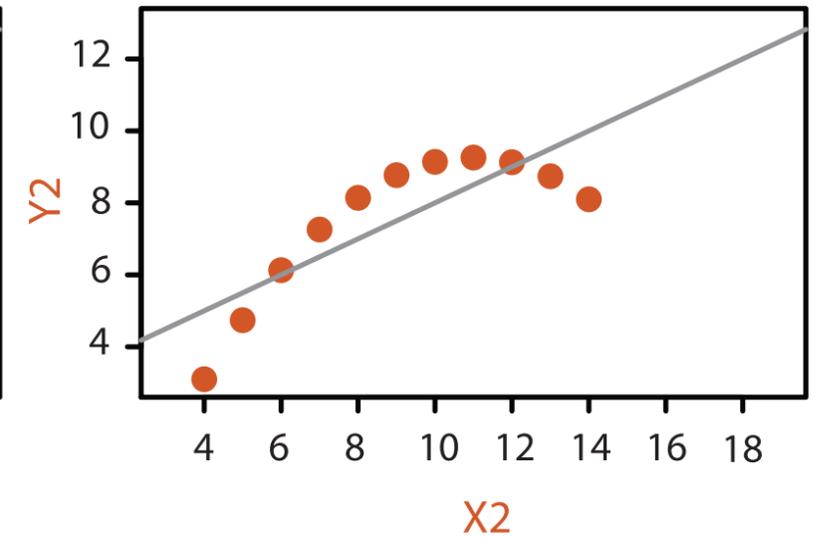
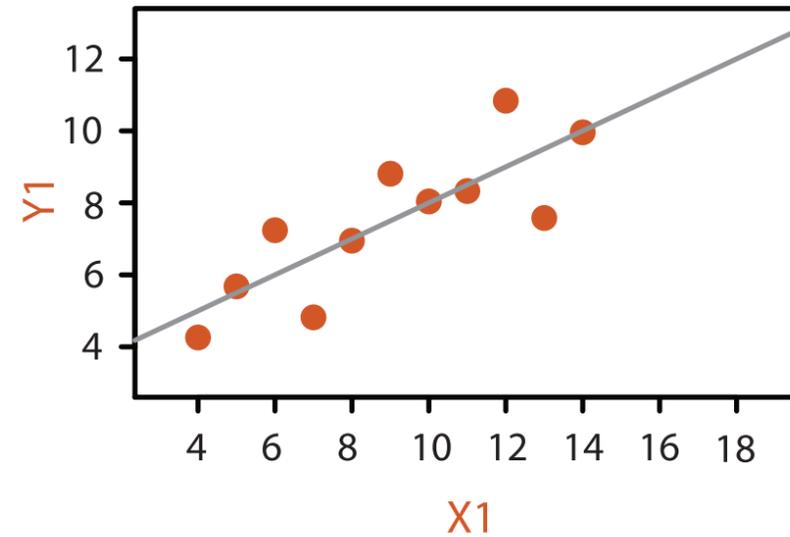
# Why show the data in detail?

- 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



# 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?

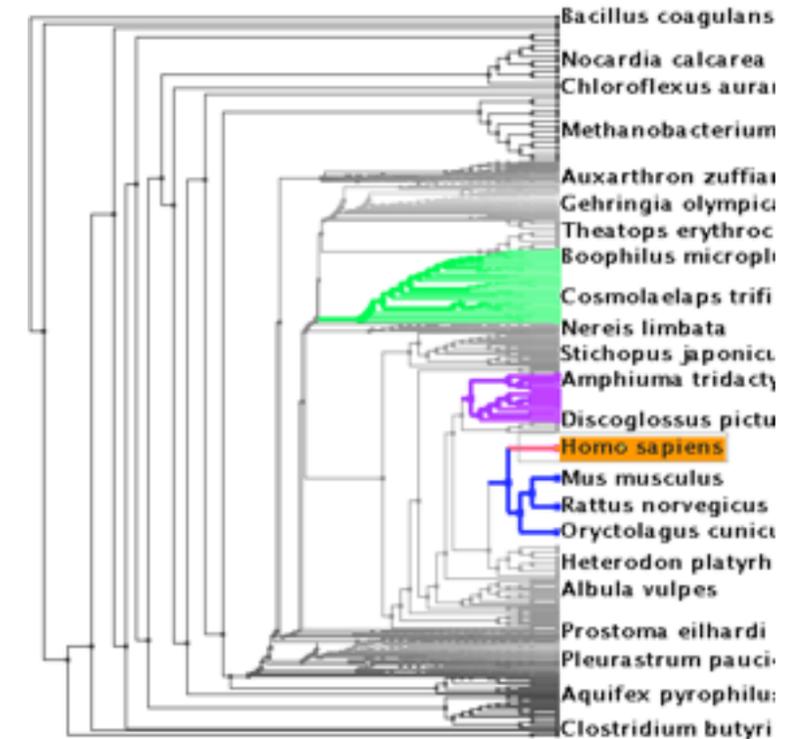
- imposes structure on huge design space
  - scaffold to help you think systematically about choices
  - analyzing existing as stepping stone to designing new
  - most possibilities ineffective for particular task/data combination

## 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



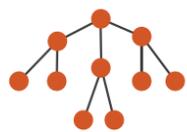
[TreeJuxtaposer: Scalable Tree Comparison Using Focus +Context With Guaranteed Visibility. ACM Trans. on Graphics (Proc. SIGGRAPH) 22:453– 462, 2003.]

### What?

### Why?

### How?

#### → Tree



#### → Actions

→ Present → Locate → Identify



#### → Targets

→ Path between two nodes



#### → SpaceTree

→ Encode → Navigate → Select → Filter → Aggregate



#### → TreeJuxtaposer

→ Encode → Navigate → Select → Arrange



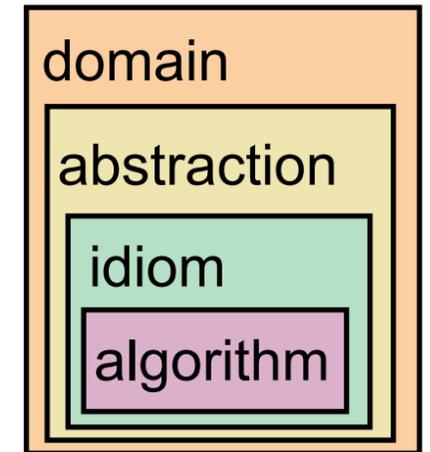
What?

Why?

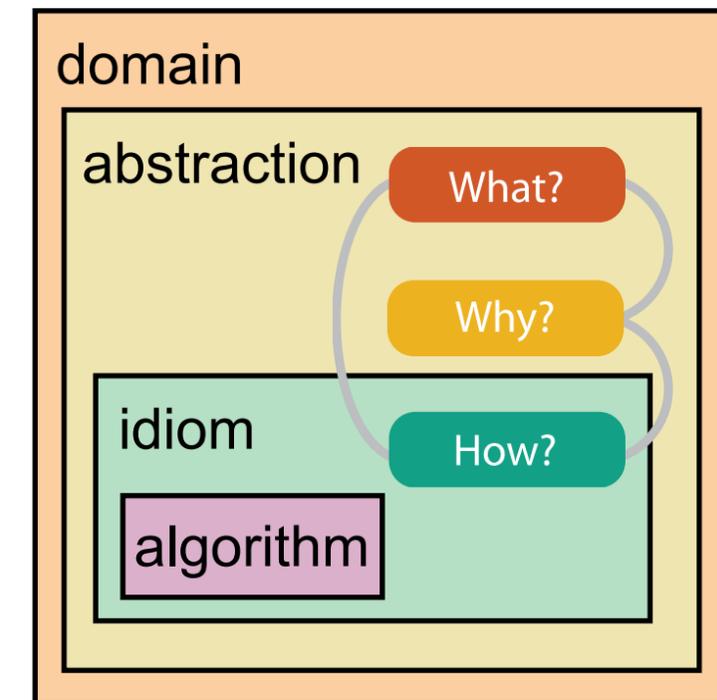
How?

# 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

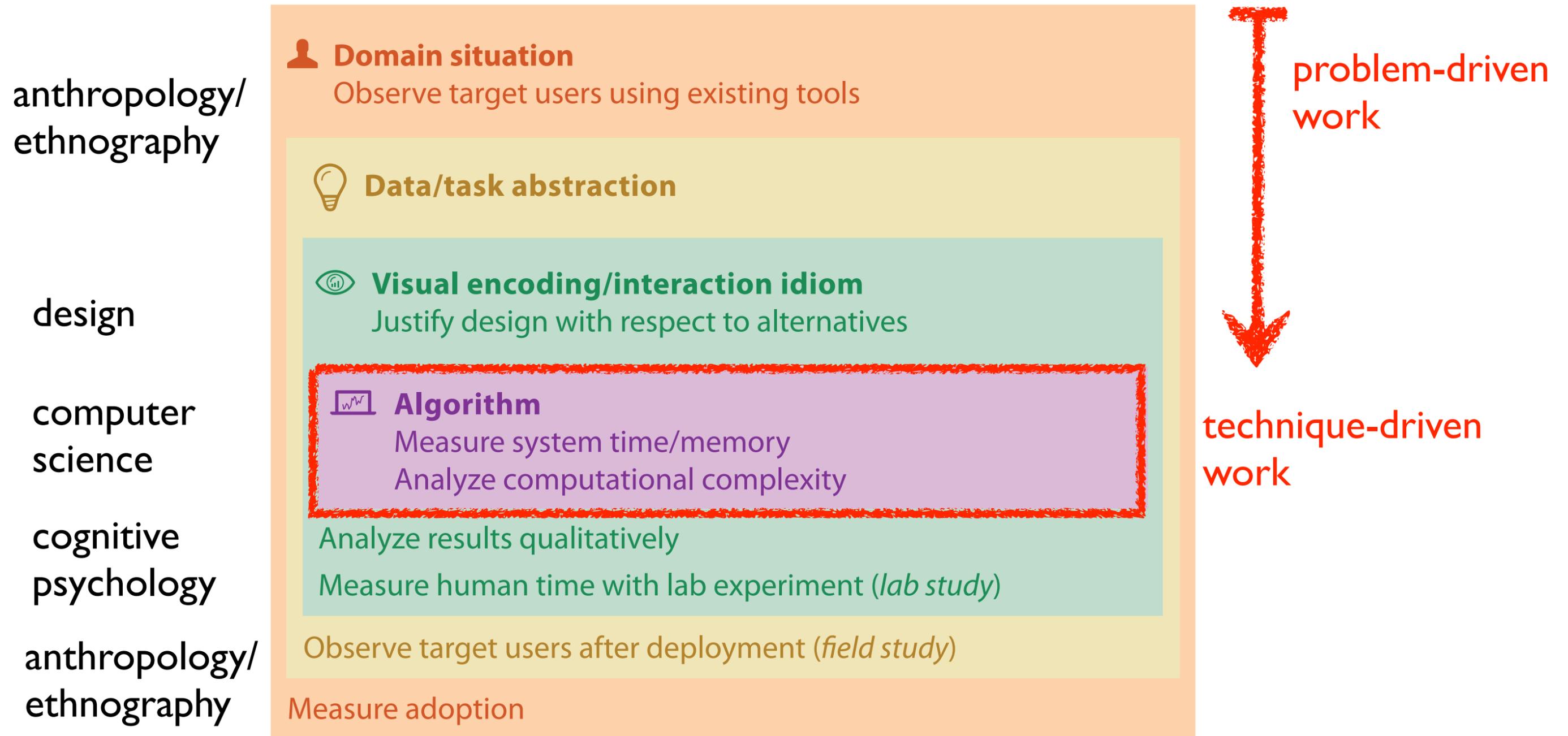


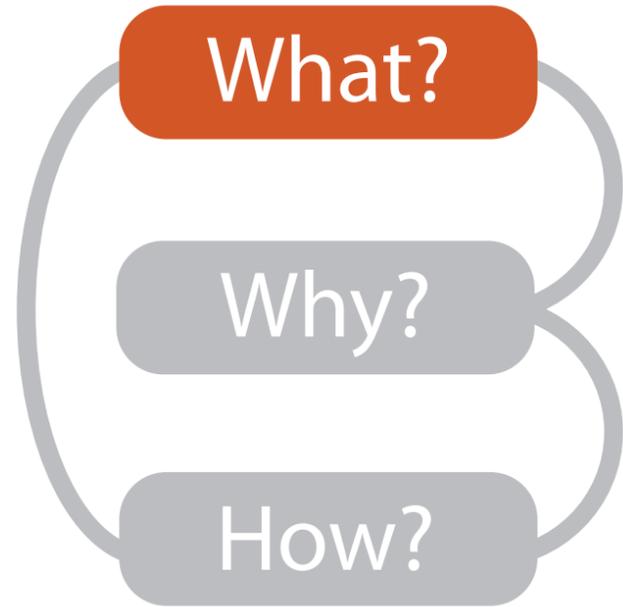
## Algorithm

Your code is too slow

# Why is validation difficult?

- solution: use methods from different fields at each level





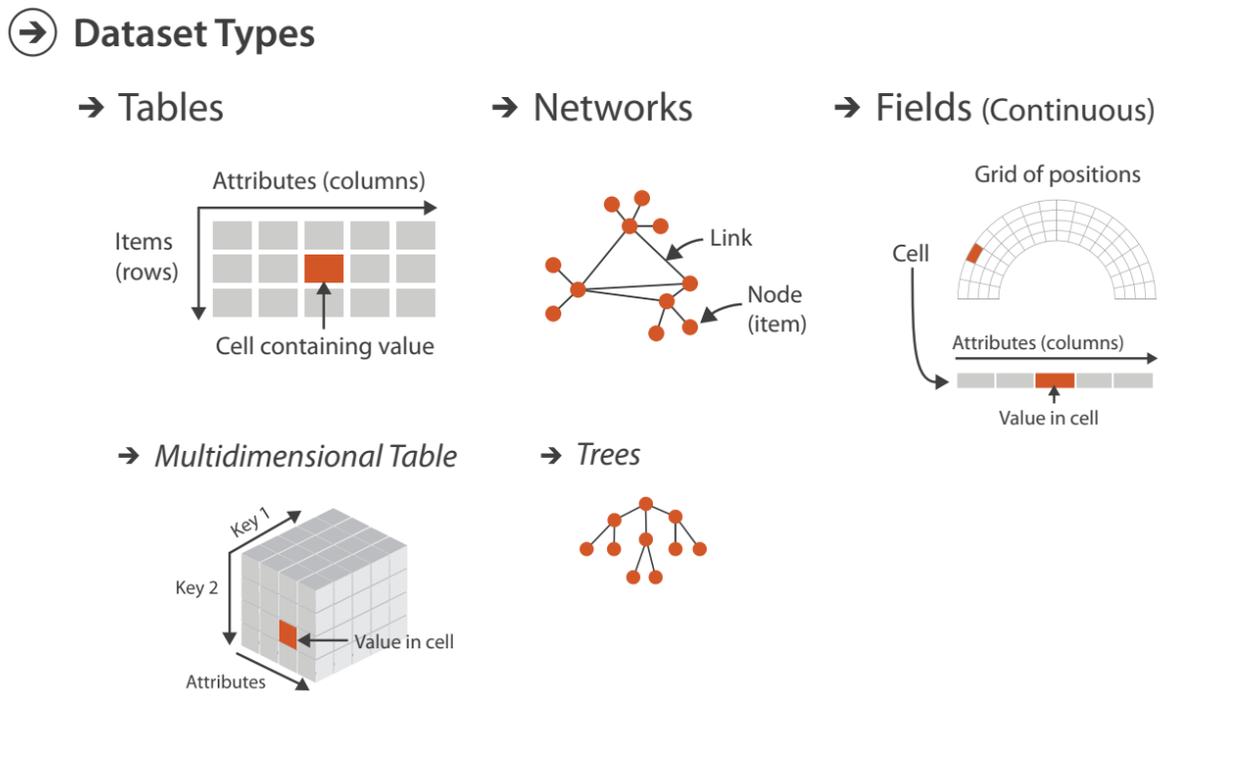
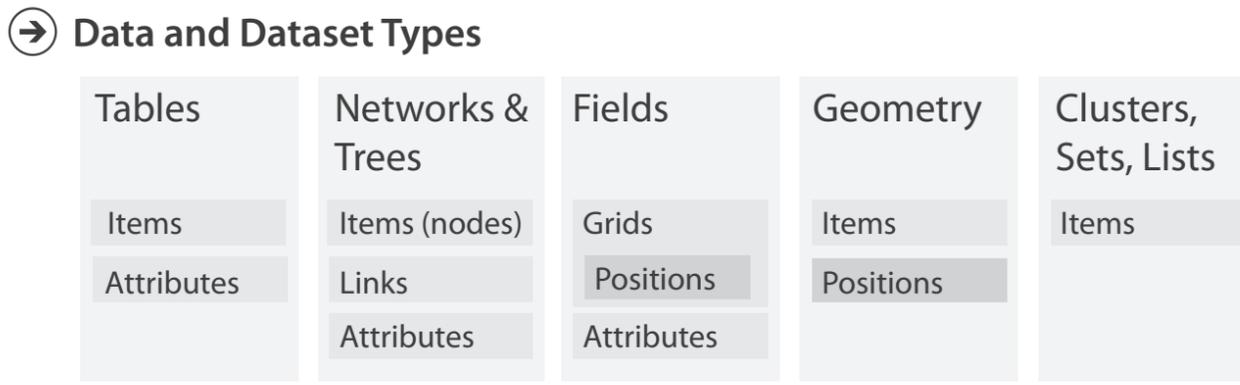
# What?

## Datasets

## Attributes

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

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



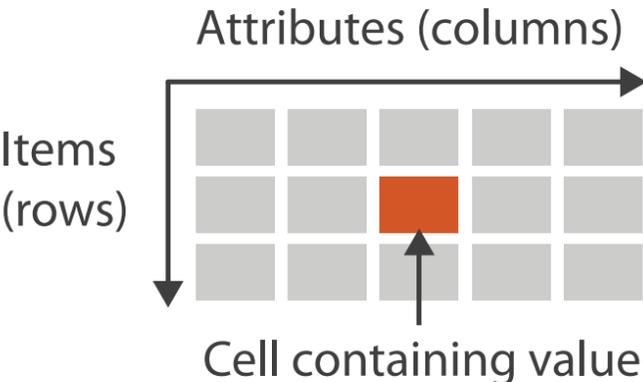
- ➔ Ordering Direction
  - ➔ Sequential
    -
  - ➔ Diverging
    - ←→
  - ➔ Cyclic
    - ↻



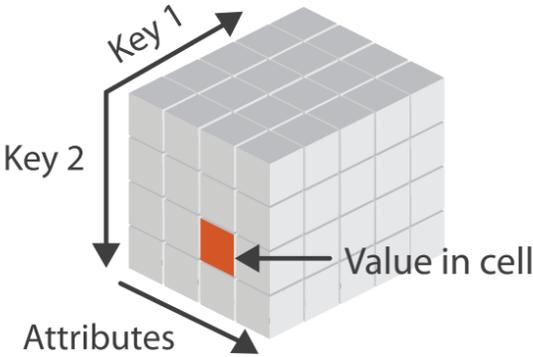
# Three major datatypes

## → Dataset Types

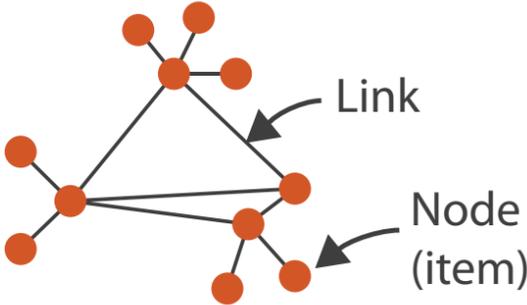
### → Tables



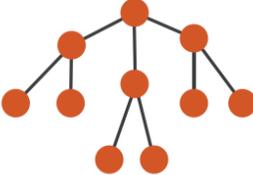
### → Multidimensional Table



### → Networks

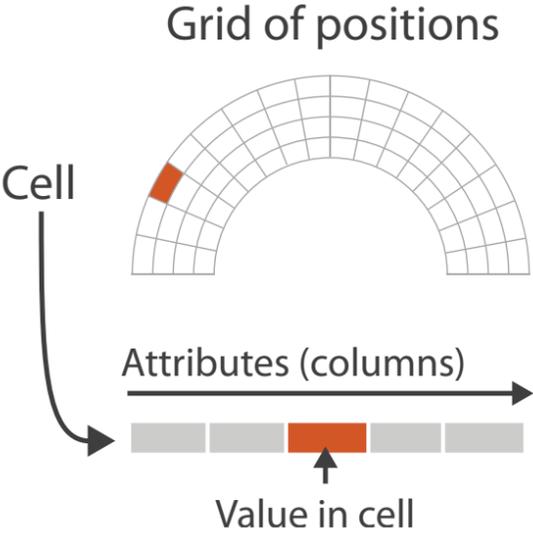


### → Trees



### → Spatial

#### → Fields (Continuous)



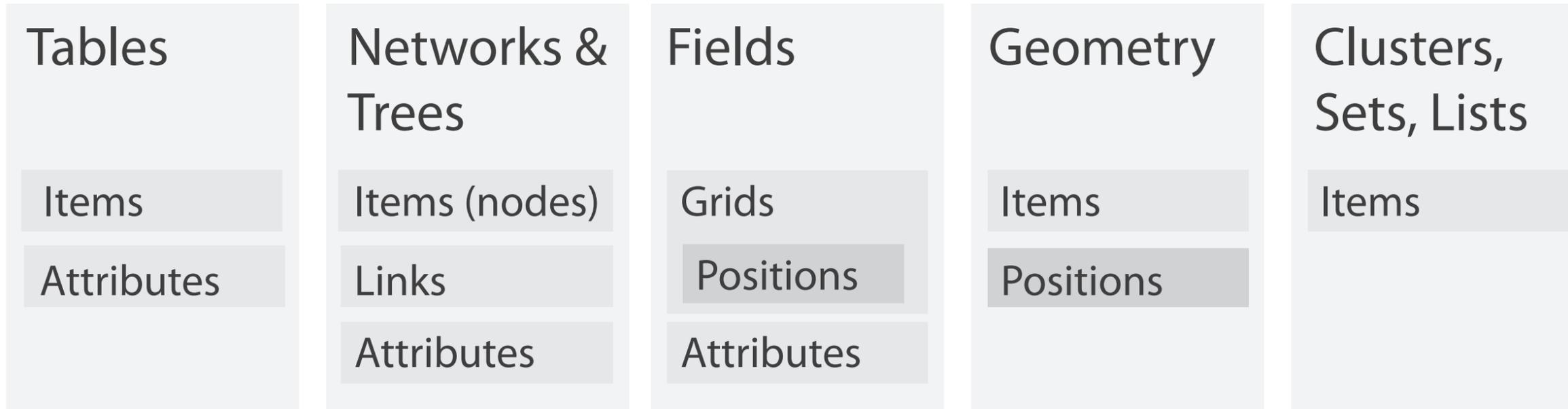
#### → Geometry (Spatial)



- 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

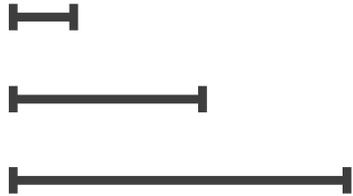


➔ Ordered

➔ *Ordinal*



➔ *Quantitative*



## ➔ Ordering Direction

➔ Sequential

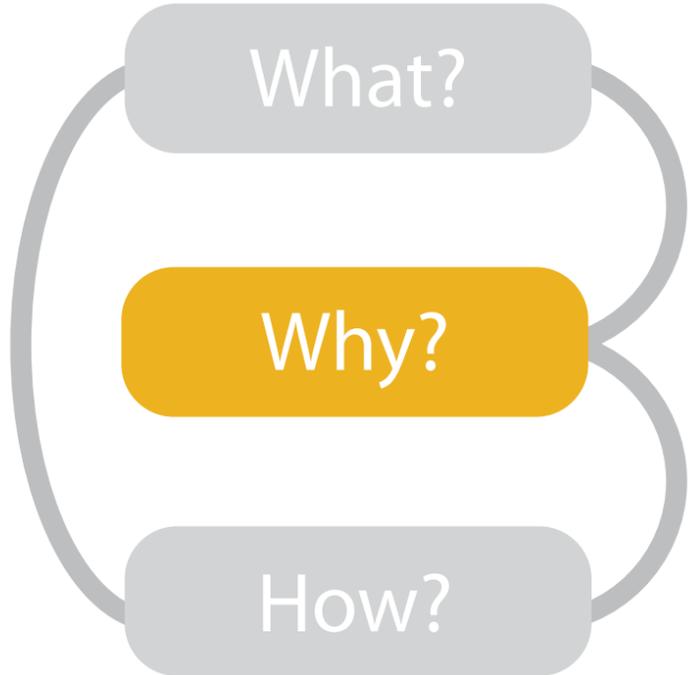


➔ Diverging



➔ Cyclic





## 👉 Actions

## 🎯 Targets

➔ **Analyze**

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

➔ **Search**

	Target known	Target unknown
Location known	Lookup	Browse
Location unknown	Locate	Explore

➔ **Query**

- ➔ Identify
- ➔ Compare
- ➔ Summarize

➔ **All Data**

- ➔ Trends
- ➔ Outliers
- ➔ Features

➔ **Attributes**

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

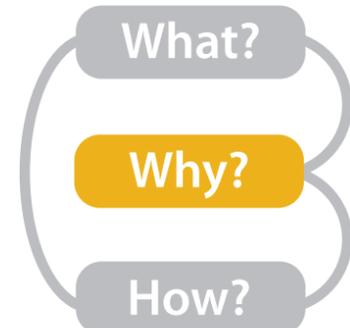
➔ **Network Data**

- ➔ 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*

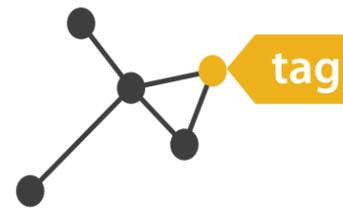


➔ *Enjoy*



### ➔ Produce

➔ *Annotate*



➔ *Record*

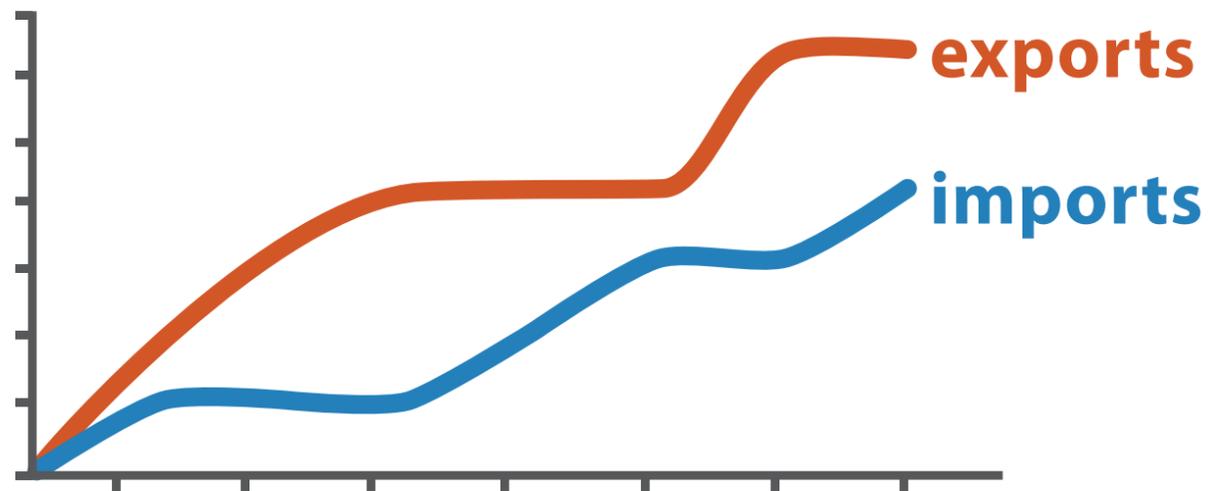


➔ *Derive*

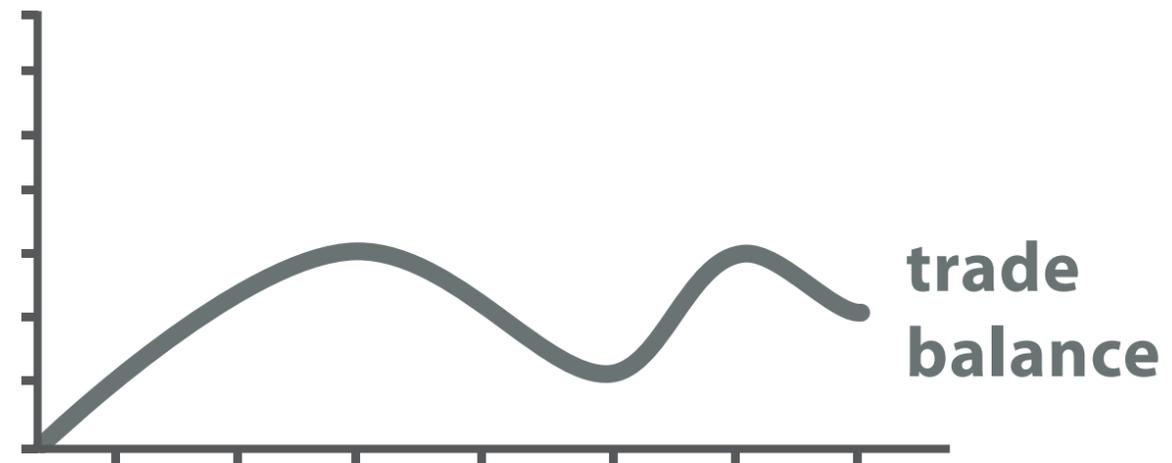


# Derive

- 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>

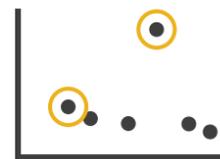
- independent choices for each of these three levels

–analyze, search, query

–mix and match

## → Query

→ Identify



→ Compare



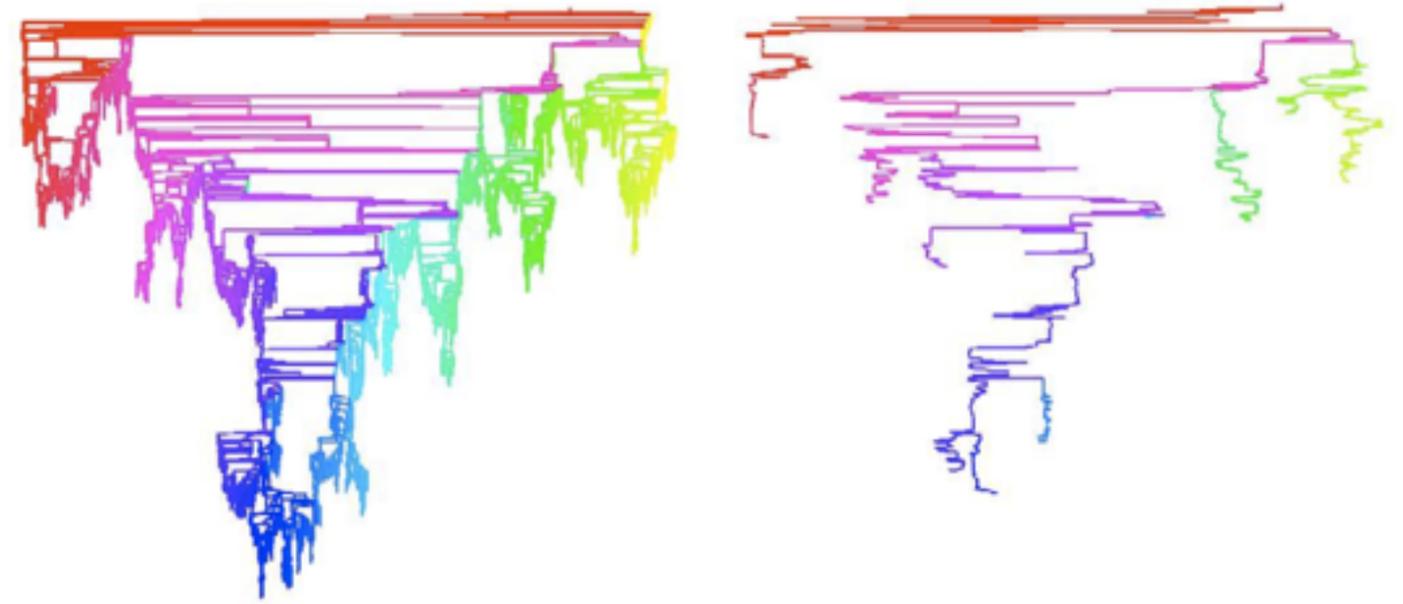
→ Summarize



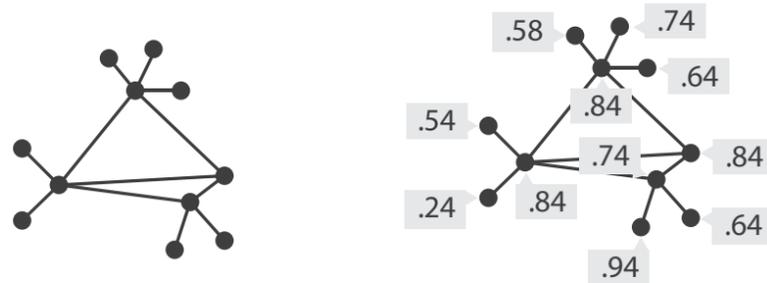
# Analysis example: Derive one attribute

- 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

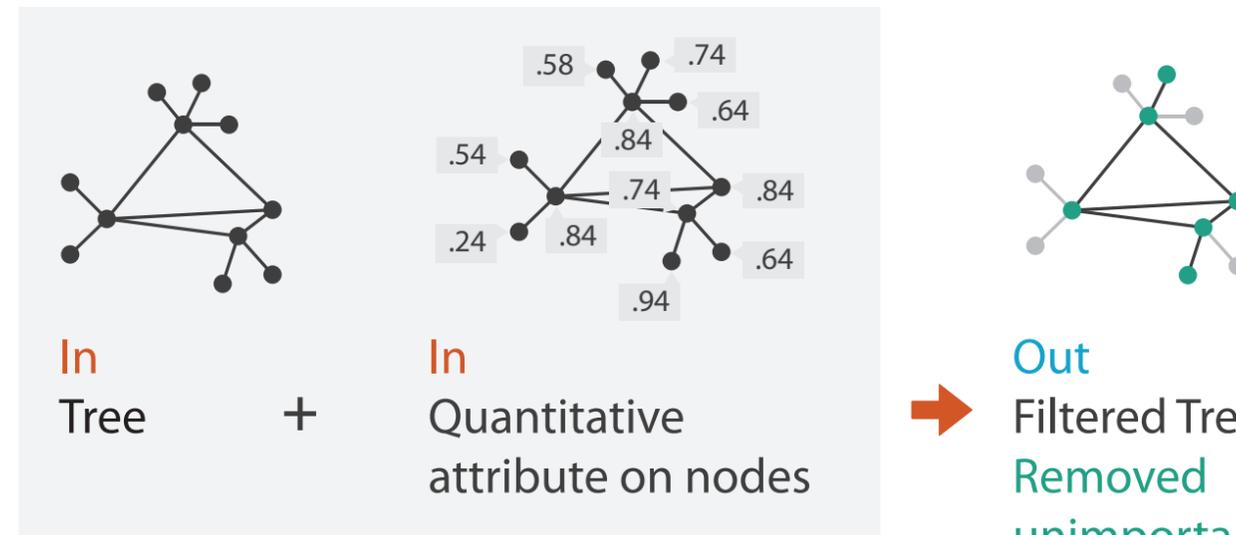
### What?

- ➔ 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



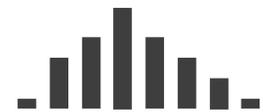
→ Features



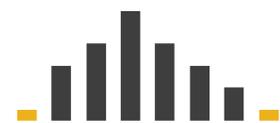
## → Attributes

→ One

→ *Distribution*



→ *Extremes*

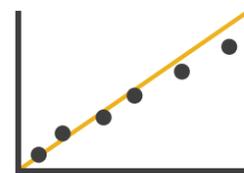


→ Many

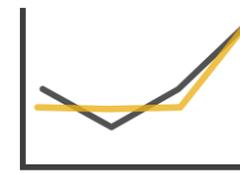
→ *Dependency*



→ *Correlation*

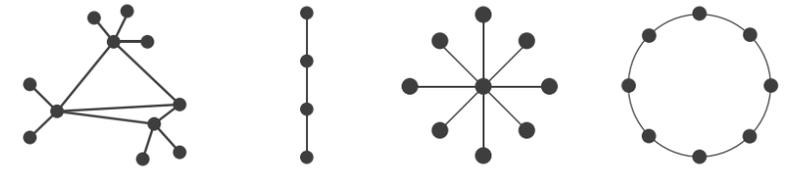


→ *Similarity*



## → Network Data

→ Topology

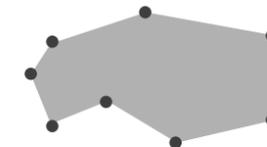


→ *Paths*



## → 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



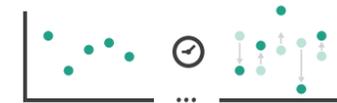
→ Motion

*Direction, Rate, Frequency, ...*



## Manipulate

### → Change



### → Select



### → Navigate



## Facet

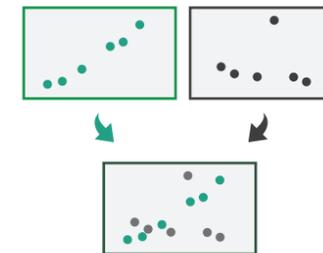
### → Juxtapose



### → Partition

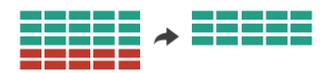


### → Superimpose

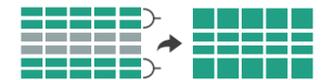


## Reduce

### → Filter



### → Aggregate



### → Embed



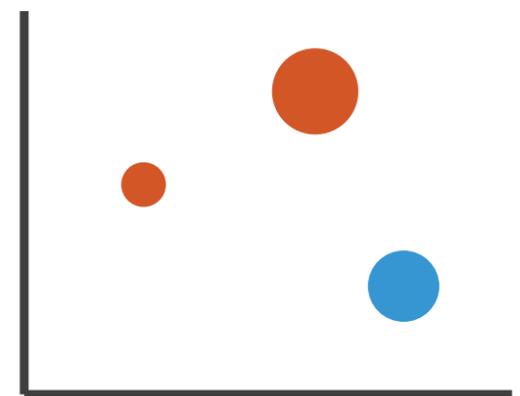
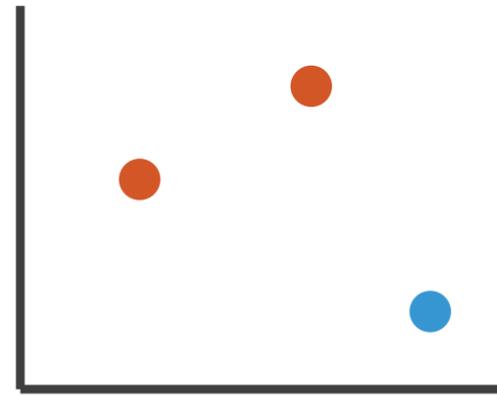
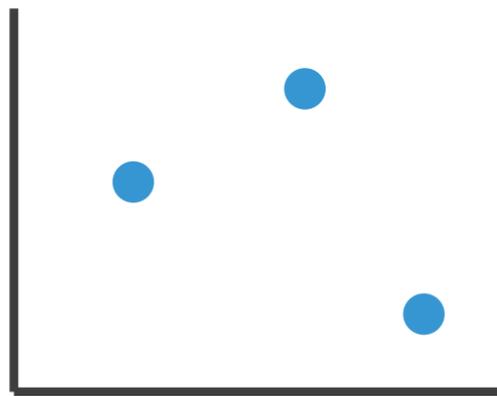
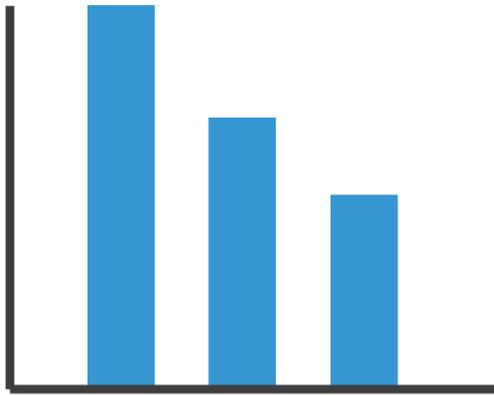
What?

Why?

How?

# Encoding visually

- analyze idiom structure



# Definitions: Marks and channels

- marks

  - geometric primitives

→ Points



→ Lines



→ Areas



- channels

  - control appearance of marks

→ Position

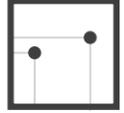
→ Horizontal



→ Vertical



→ Both



→ Color



→ Shape



→ Tilt



→ Size

→ Length



→ Area

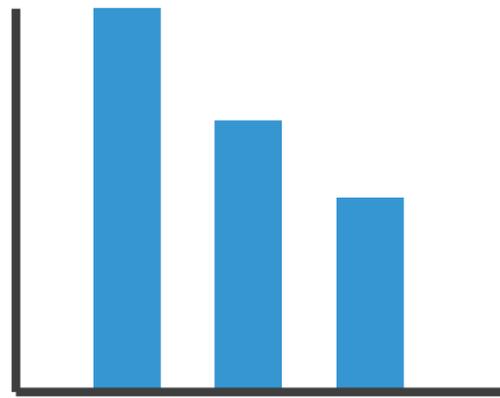


→ Volume



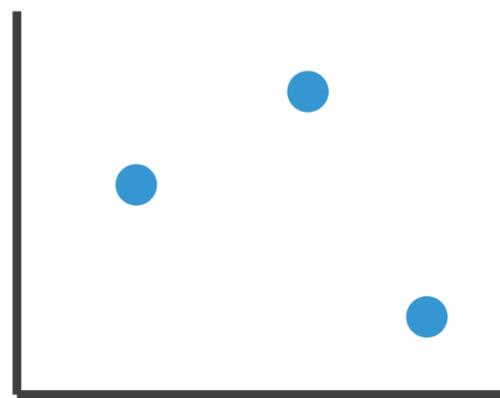
# Encoding visually with marks and channels

- analyze idiom structure
  - as combination of marks and channels



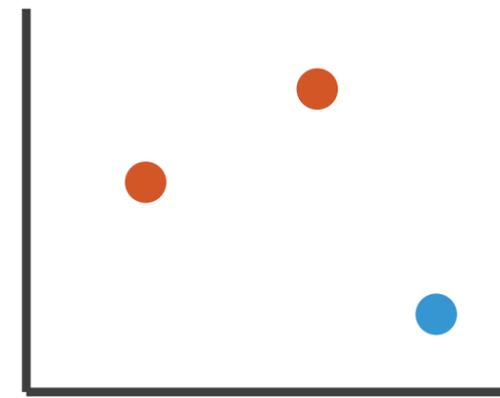
1:  
vertical position

mark: line



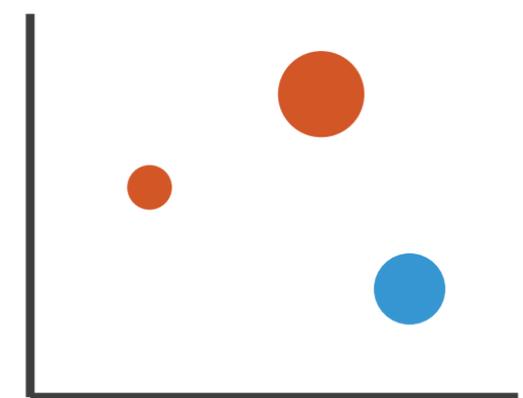
2:  
vertical position  
horizontal position

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



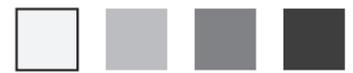
Area (2D size)



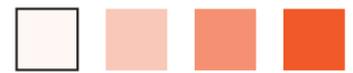
Depth (3D position)



Color luminance



Color saturation



Curvature



Volume (3D size)



Same

Same

Spatial region



Color hue



Motion

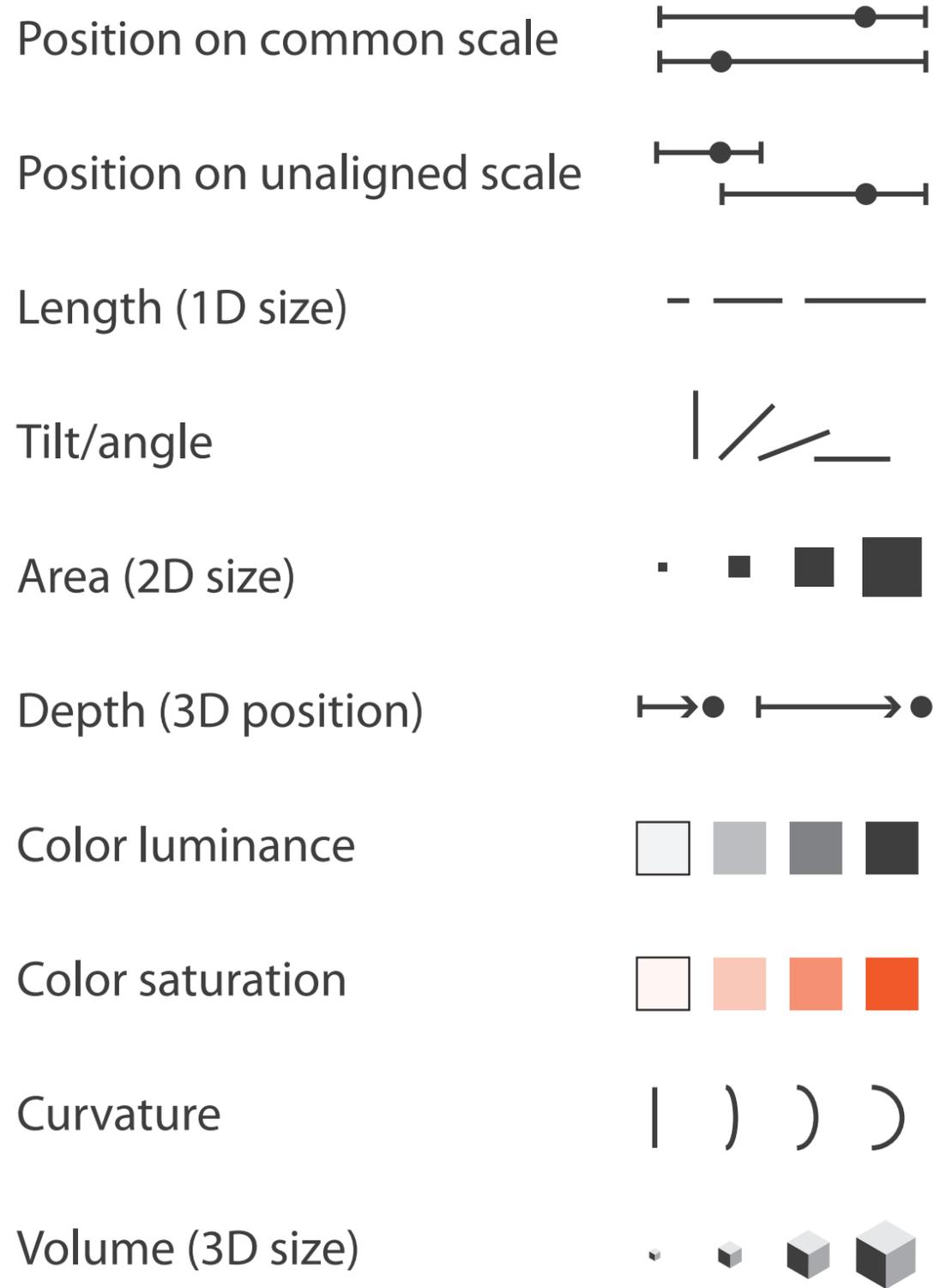


Shape



# Channels: Rankings

## ➔ Magnitude Channels: Ordered Attributes



## ➔ Identity Channels: Categorical Attributes

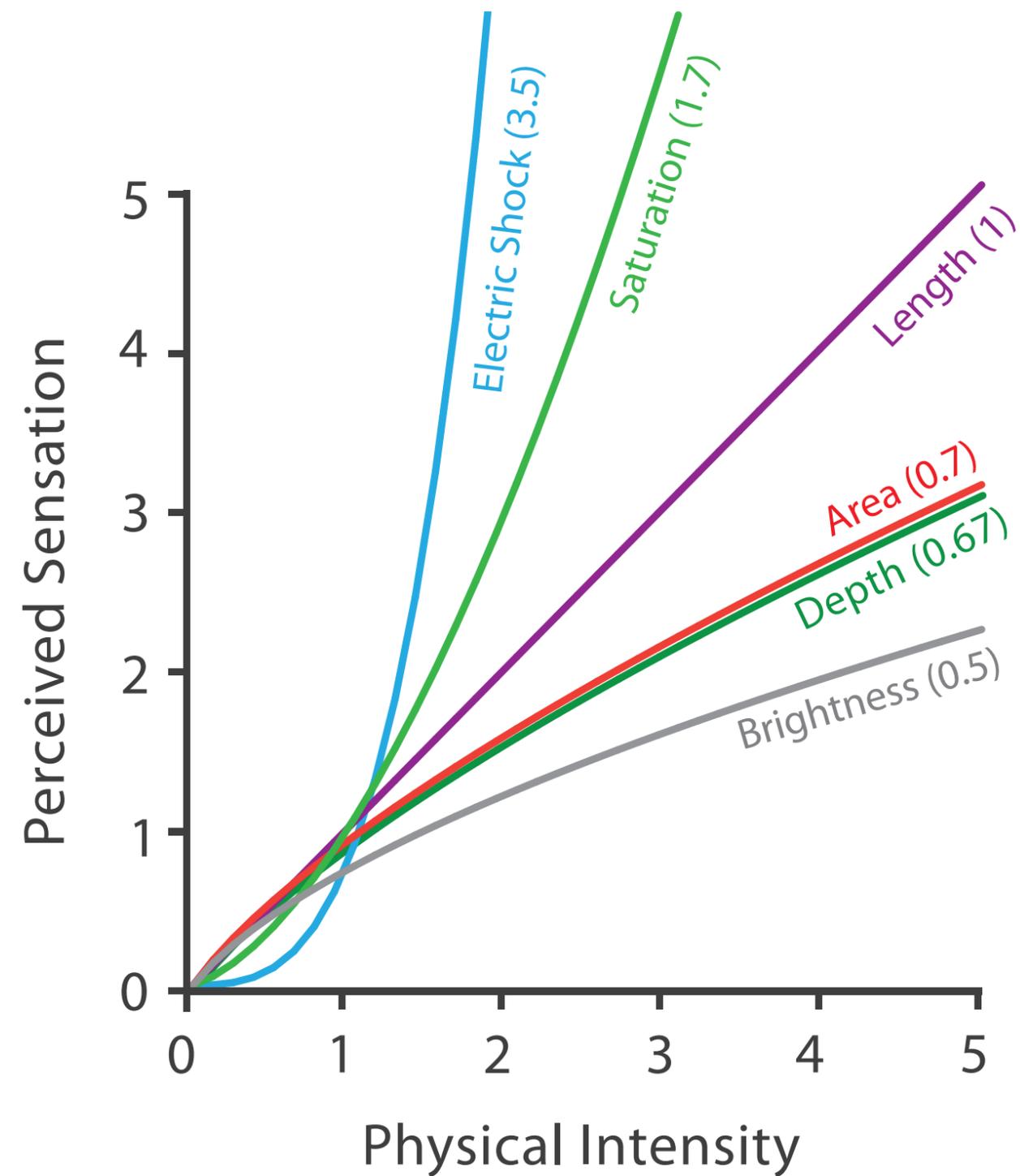


Best  
Effectiveness  
Least

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

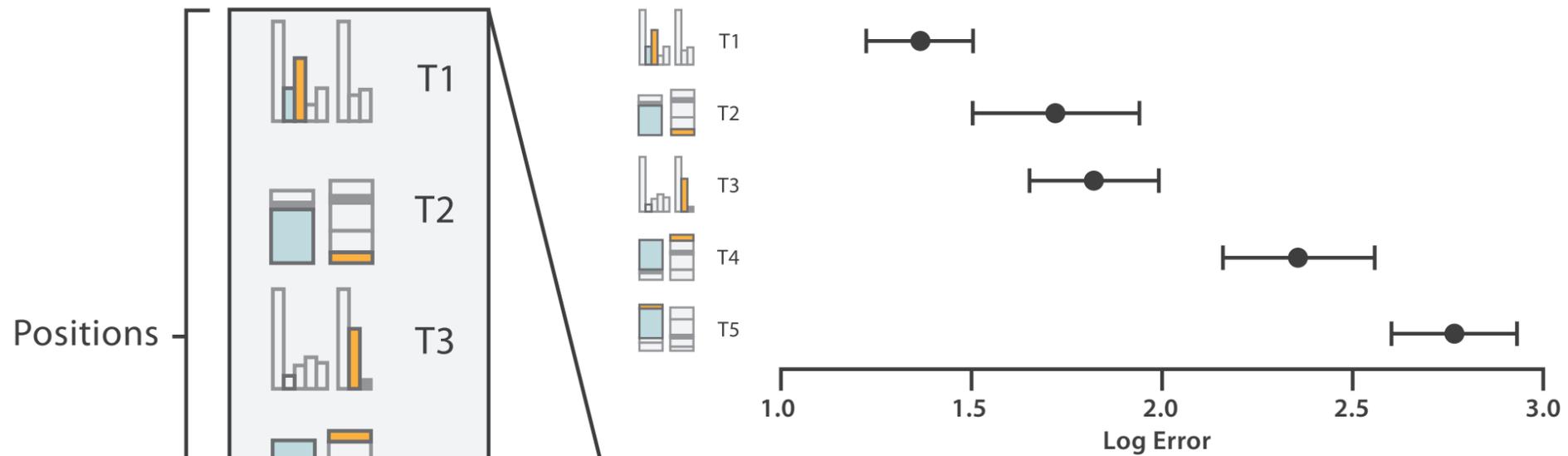
# Accuracy: Fundamental Theory

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

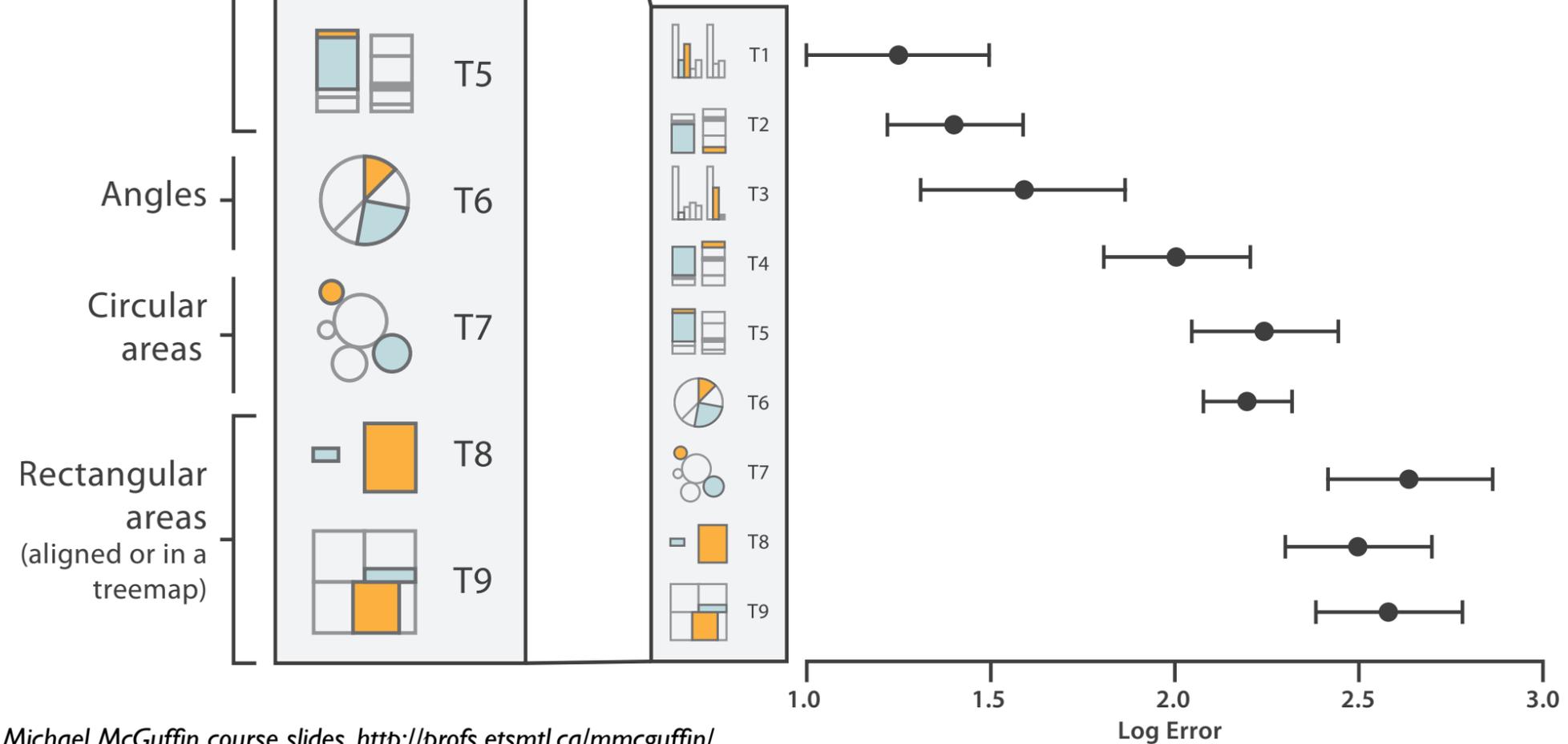


# Accuracy: Vis experiments

Cleveland & McGill's Results



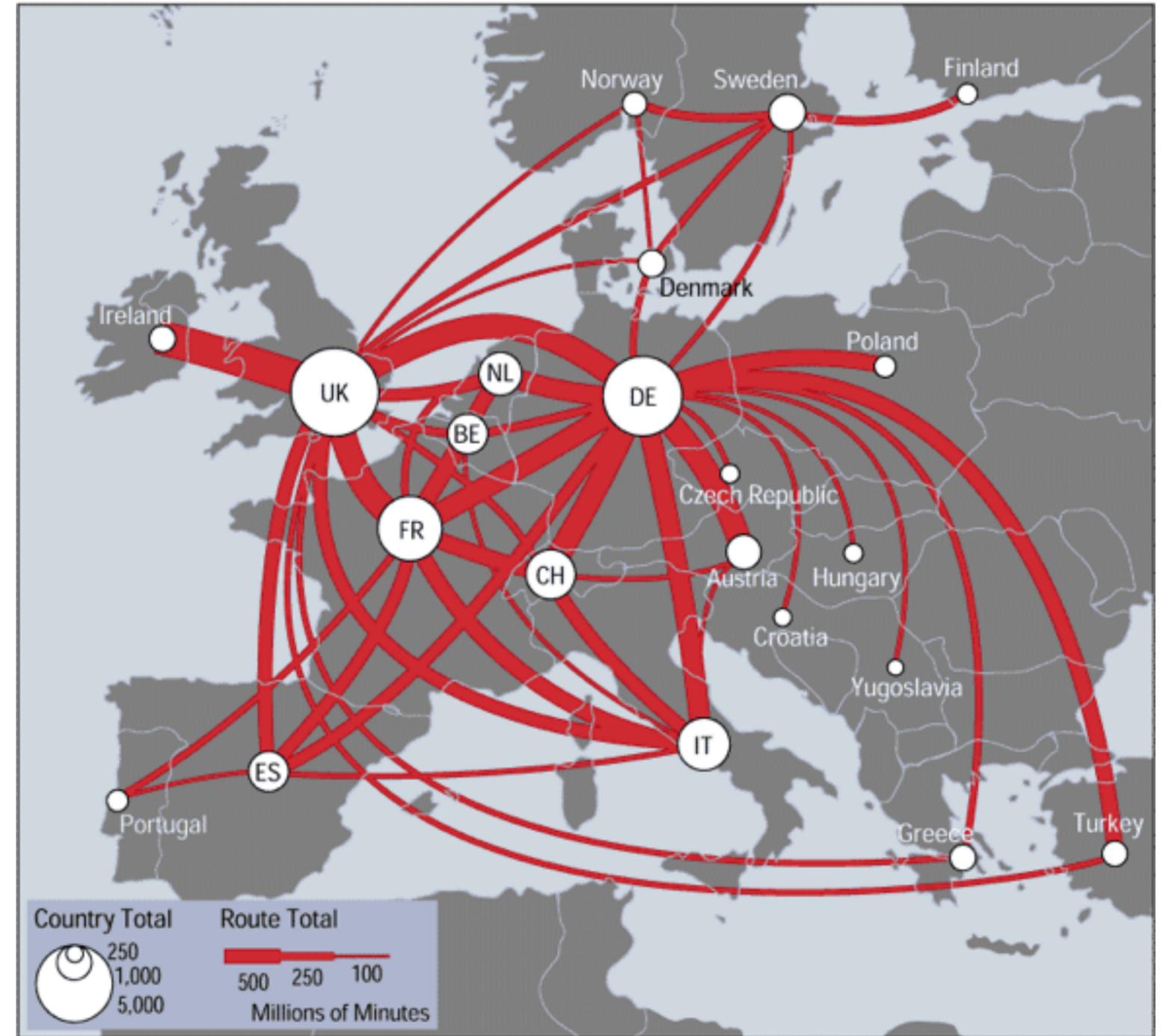
Crowdsourced Results



*[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?

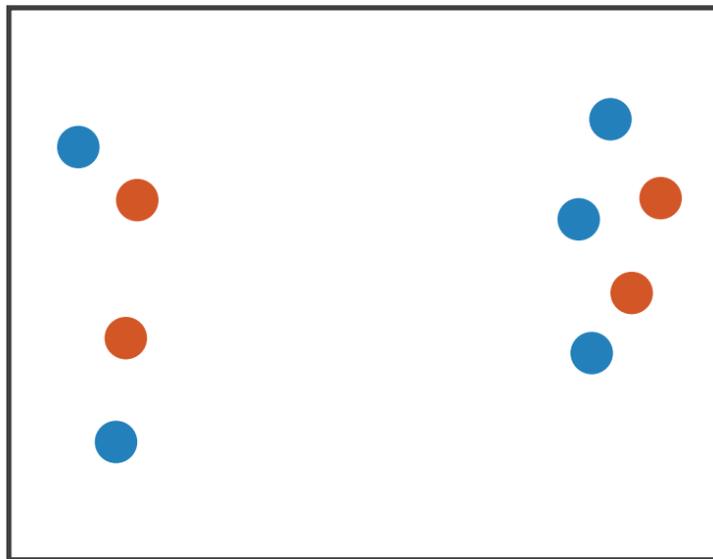
- must be sufficient for number of attribute levels to show
  - linewidth: few bins



[[mappa.mundi.net/maps/maps\\_014/telegeography.html](http://mappa.mundi.net/maps/maps_014/telegeography.html)]

# Separability vs. Integrality

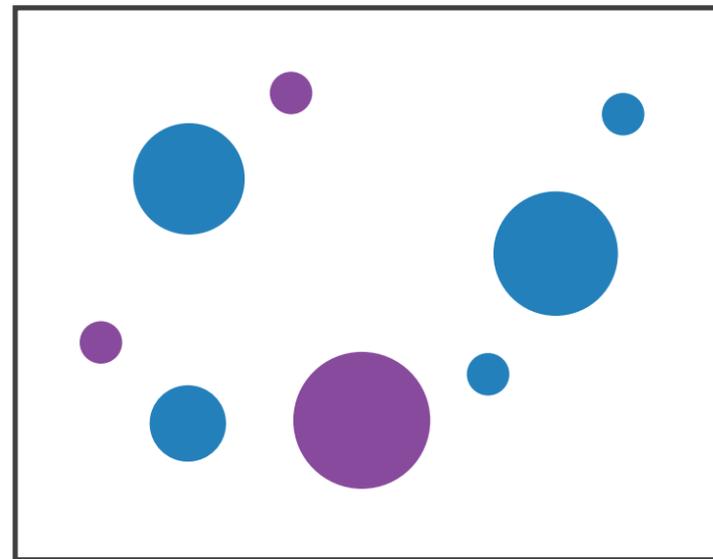
Position  
+ Hue (Color)



Fully separable

2 groups each

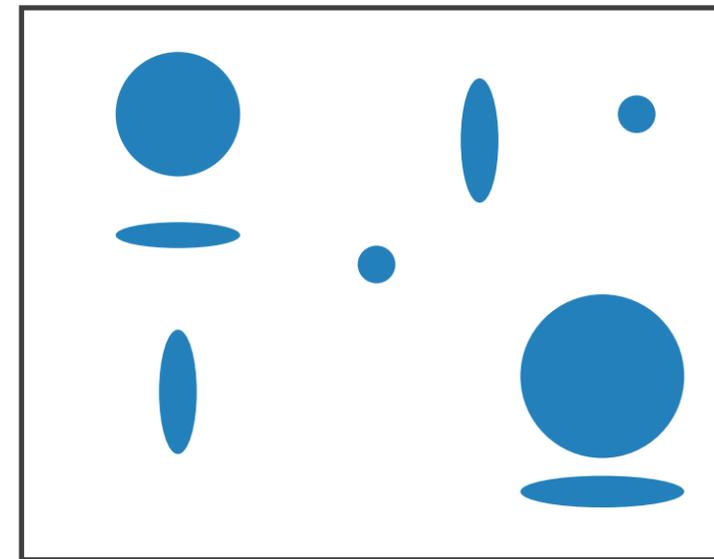
Size  
+ Hue (Color)



Some interference

2 groups each

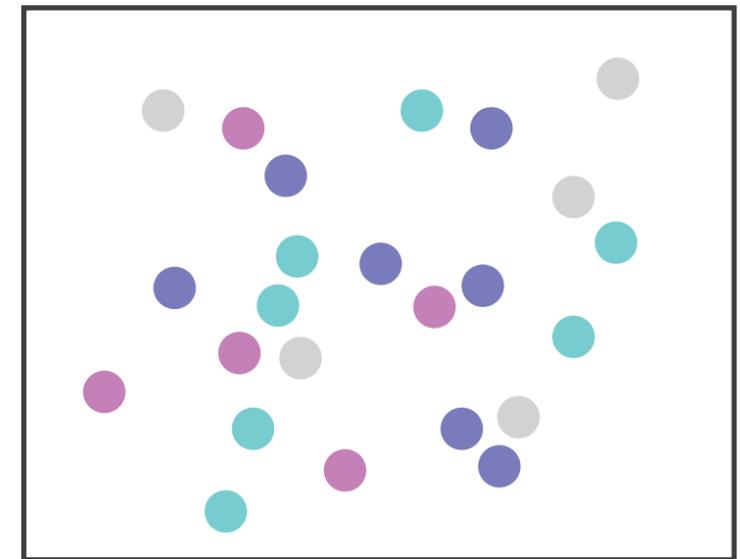
Width  
+ Height



Some/significant  
interference

3 groups total:  
integral area

Red  
+ Green

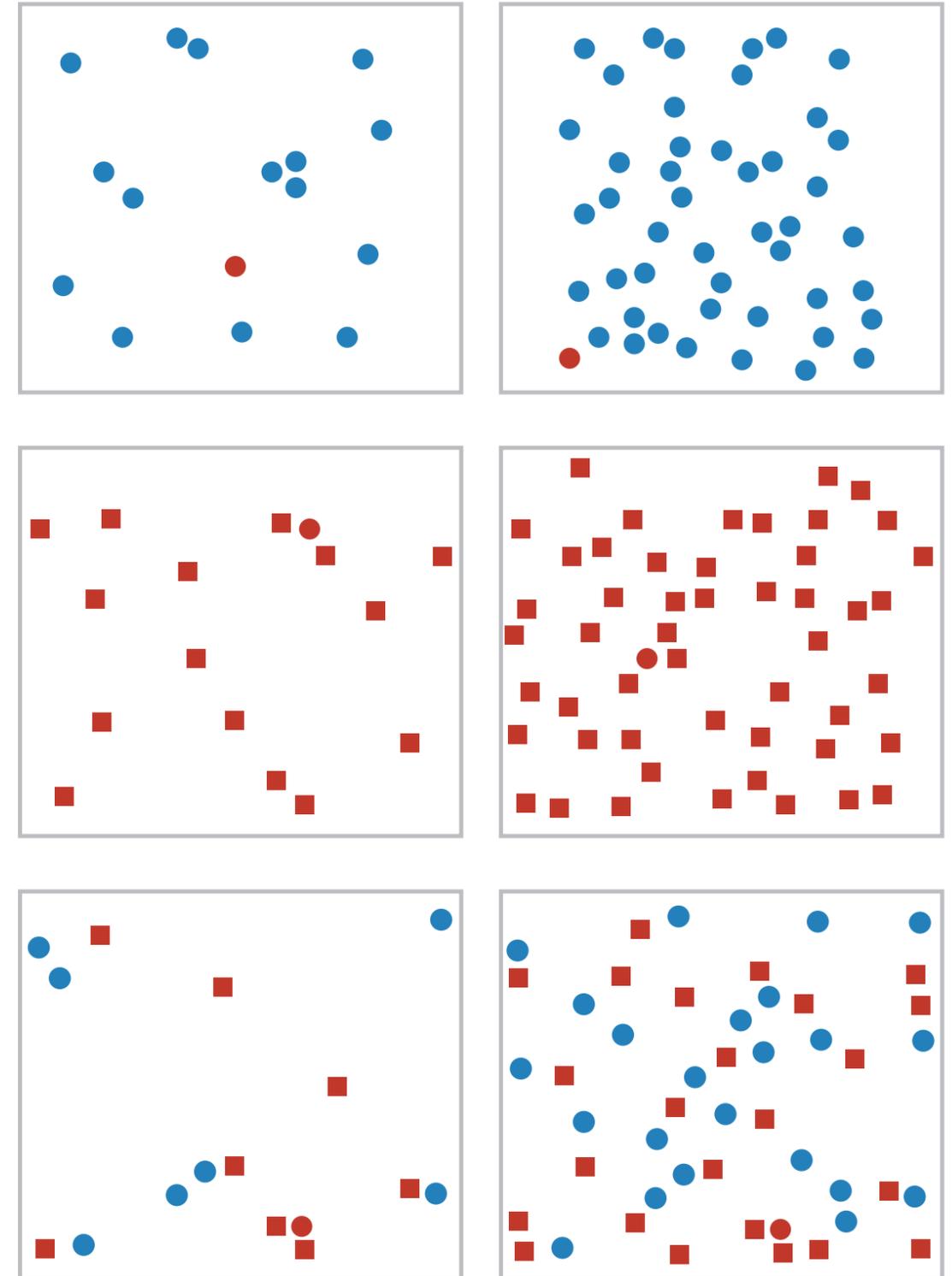


Major interference

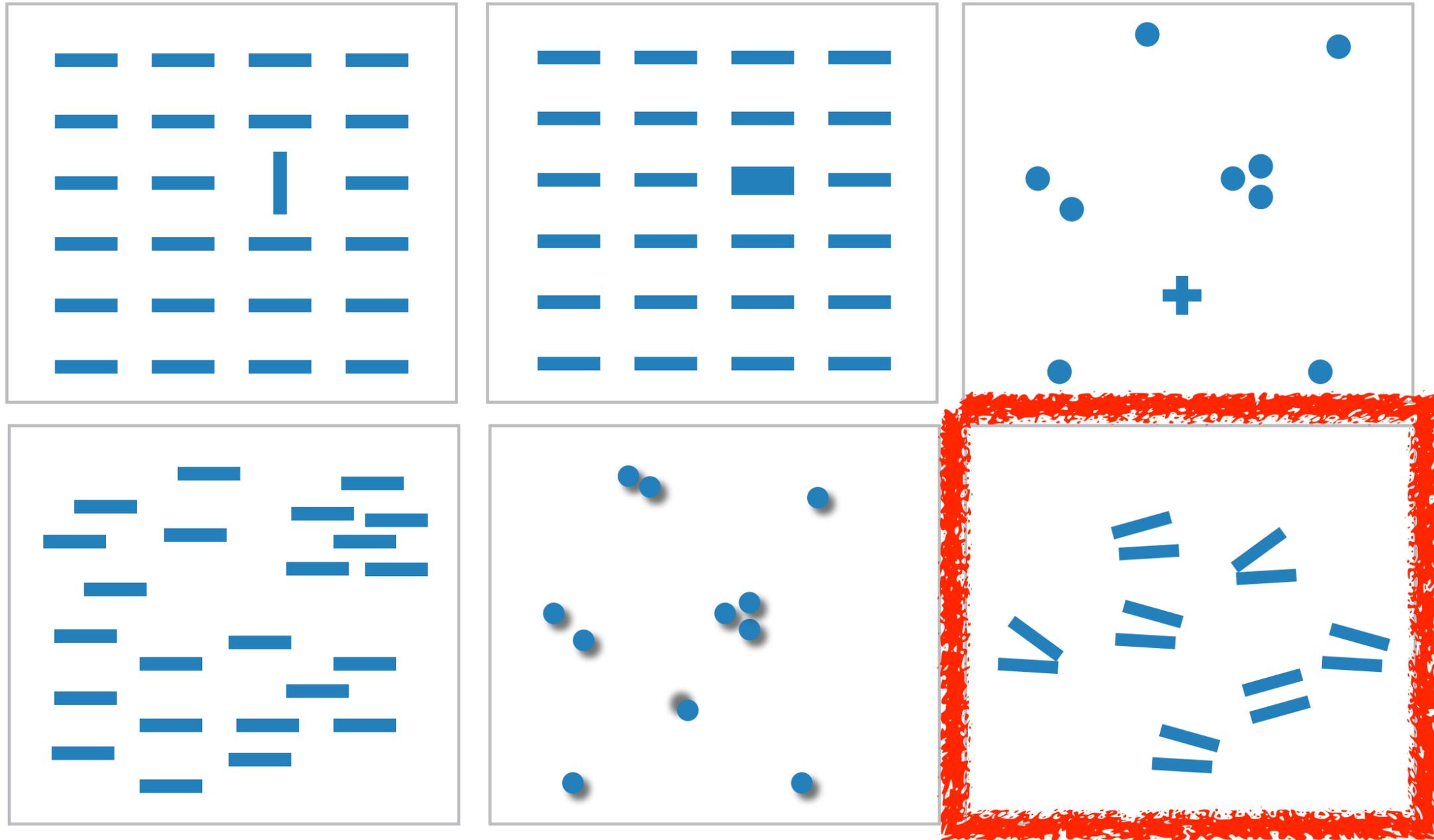
4 groups total:  
integral hue

# 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

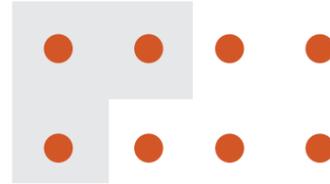
# Grouping

- containment
- connection

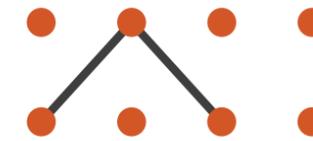
- proximity
  - same spatial region
- similarity
  - same values as other categorical channels

## Marks as Links

### ➔ Containment



### ➔ Connection



### ➔ Identity Channels: Categorical Attributes

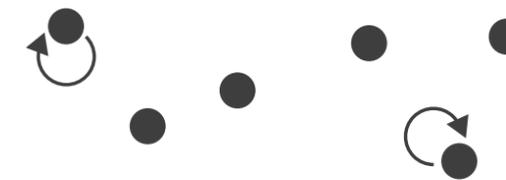
Spatial region



Color hue



Motion



Shape

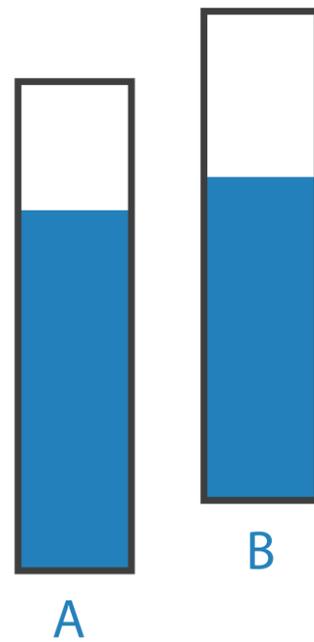


# Relative vs. absolute judgements

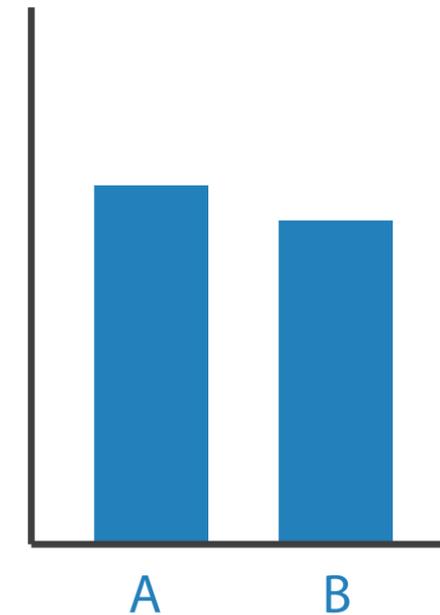
- perceptual system mostly operates with relative judgements, not absolute
  - that's why accuracy increases with common frame/scale and alignment
  - Weber's Law: ratio of increment to background is constant
    - filled rectangles differ in length by 1:9, difficult judgement
    - white rectangles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



length



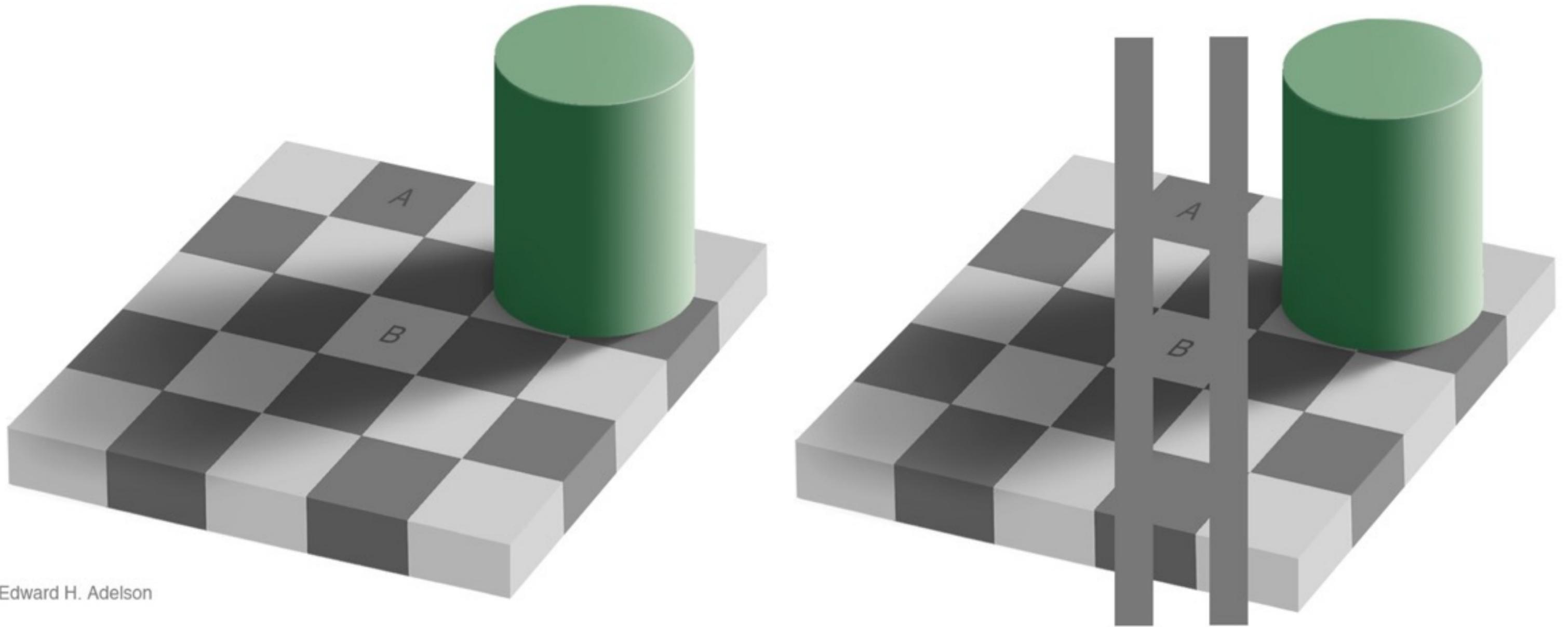
position along  
unaligned  
common scale



position along  
aligned scale

# Relative luminance judgements

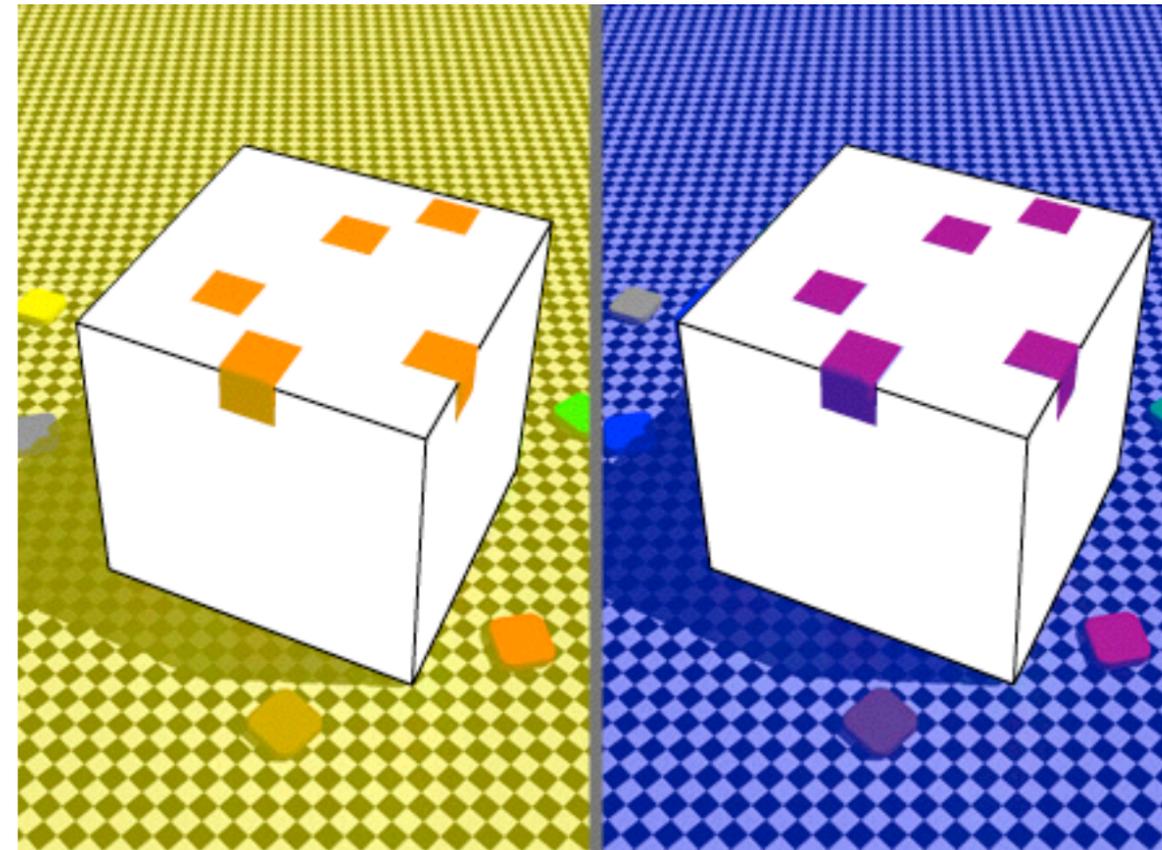
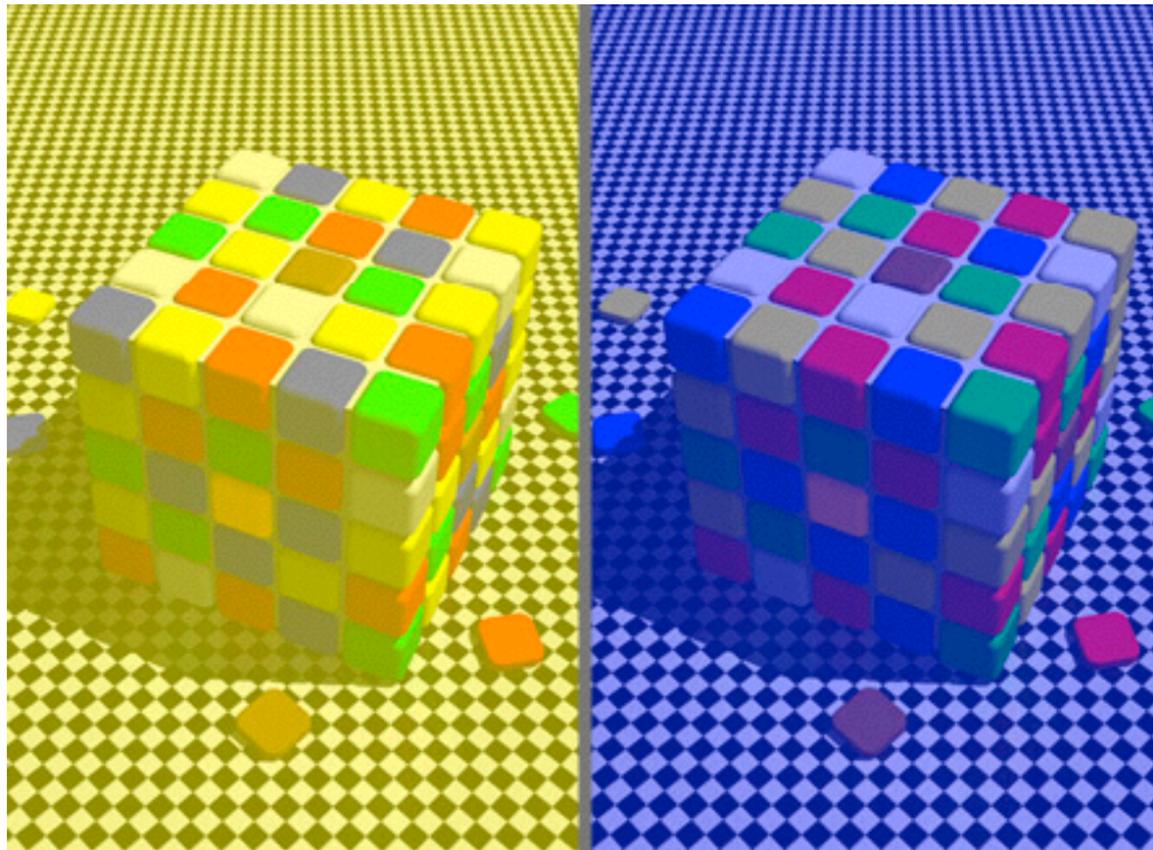
- perception of luminance is contextual based on contrast with surroundings



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](mailto:tmm@cs.ubc.ca) and [caitlin@discoursemedia.org](mailto:caitlin@discoursemedia.org) with subject JOURN Week 1