

Week 1: Intro, Marks and Channels

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University of British Columbia

JRNL 520M, Special Topics in Contemporary Journalism: Visualization for Journalists

Week 1: 15 September 2015

<http://www.cs.ubc.ca/~tmm/courses/journ15>

Who's who

- Instructor: Tamara Munzner
 - UBC Computer Science



- Journalistic kibitzer: Alfred Hermida
 - UBC Journalism



- Guest lecturer and significant labs help: Robert Kosara
 - Research Scientist, Tableau Software
 - previously UNC Charlotte Computer Science



Class time

- 6 weeks, Sep 15 - Oct 20
 - 1 3-hr session per week
- standard week
 - foundations lecture/discussion: 90 min
 - break: 15 min
 - demos: 30 min
 - lab: 45 min
- demo-intensive weeks
 - Week 1 & Week 4: longer demo from guest lecturer Robert Kosara
 - foundations 60 min, break 15 min, demos 60 min, lab 45 min

Structure

- participation
 - attendance and discussion in class, 16%
 - tell me in advance if you'll miss class (and why)
 - tell when you recover if you were ill
- homework, 84%
 - 6 assignments, 14% each
 - start in lab
 - finish over one week
 - due at start of next class session
 - some solo, some in groups of 2
 - gradual transition from structured to open-ended
 - final assignment: find your own interesting data and design your own visualization for it
- draft plan, may change as pilot continues!

Further reading

- optional textbook for following up on lecture topics
 - 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 papers/books
 - links and references posted on course page
 - if DL links, use library EZproxy from off campus

Finding me

- email is the best way to reach me: tmm@cs.ubc.ca
- office hours by appointment
 - X661 (X-Wing of ICICS/CS bldg)
- course page is font of all information
 - don't forget to refresh, frequent updates
 - <http://www.cs.ubc.ca/~tmm/courses/journ15>

Topics

- Week 1
 - Intro
 - Marks and Channels
 - Demo: Tableau I, Kosara
- Week 2
 - Task and Data Abstractions
 - Arrange Tables
 - Demo: TBD
- Week 3
 - Color
 - Arrange Spatial Data
 - Demo: Text Tools & Resources, Brehmer
- Week 4
 - Arrange Networks
 - Demo: Tableau II, Kosara
- Week 5
 - Facet Into Multiple Views
 - Reduce Items and Attributes
 - Demo: TBD
- Week 6
 - Rules of Thumb
 - Putting It All Together
 - Demo: TBD

VAD Ch 1: What's Vis and Why Do It?

- Why have a human in the decision-making loop?
- Why have a computer in the loop?
- Why use an external representation?
- Why depend on vision?
- Why show the data in detail?
- Why is the vis idiom design space so huge?
- Why focus on tasks and effectiveness?
- Why are there resource limitations?
- Why analyze vis?

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.

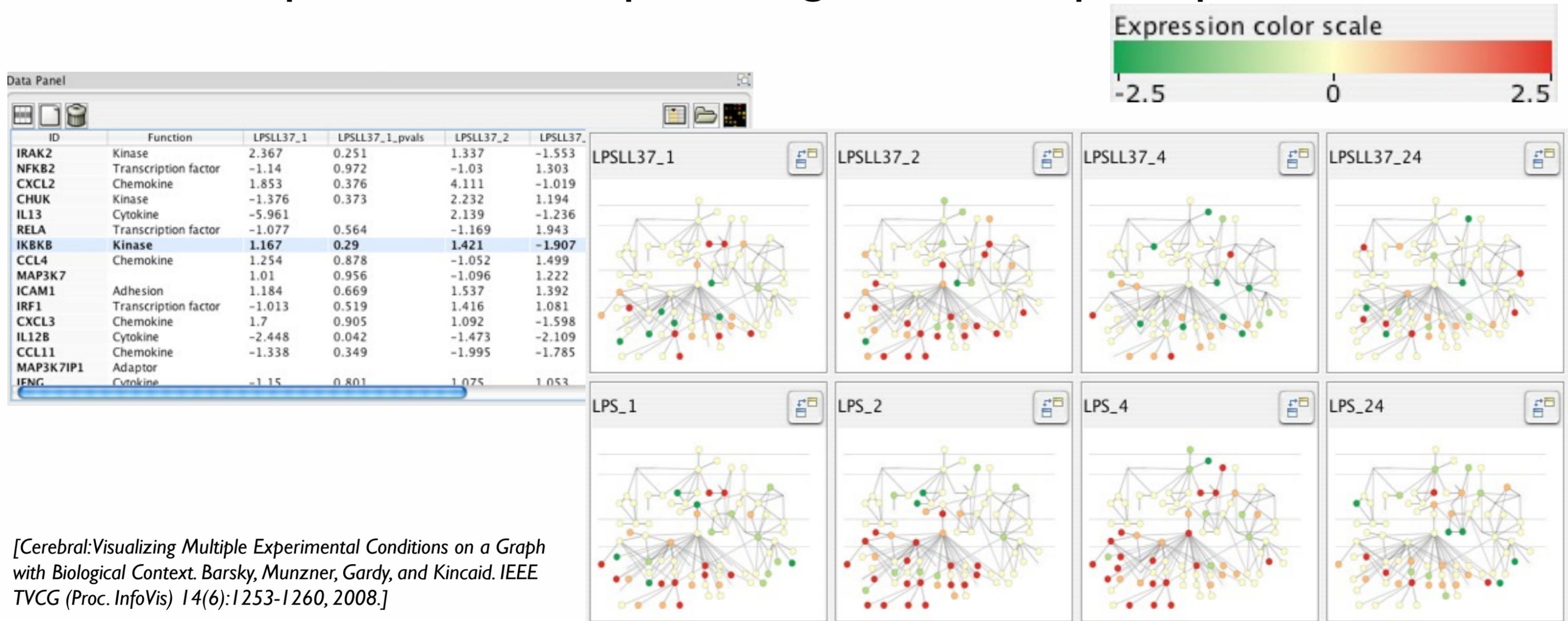
Visualization is suitable when there is a need to augment human capabilities rather than replace people with computational decision-making methods.

- 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

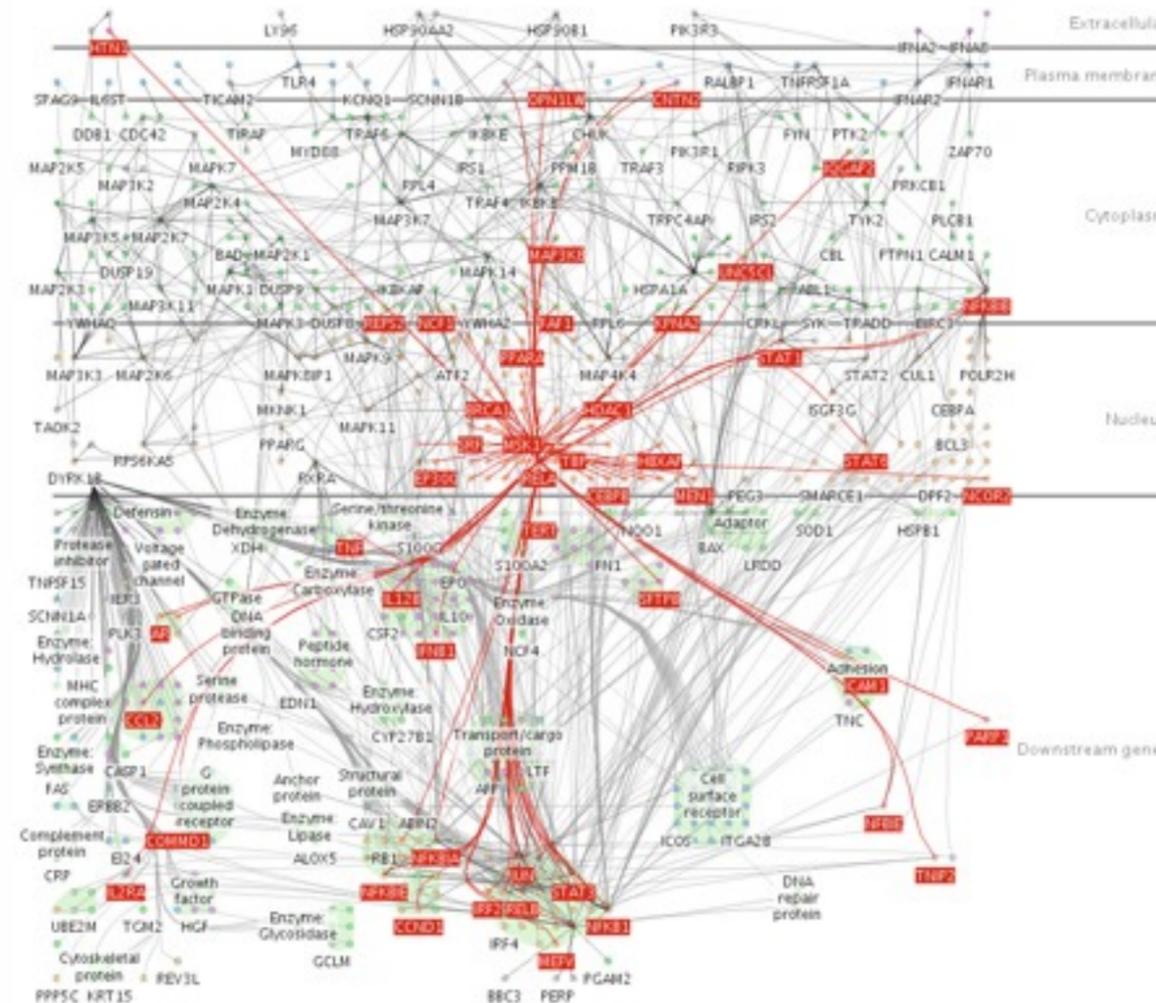
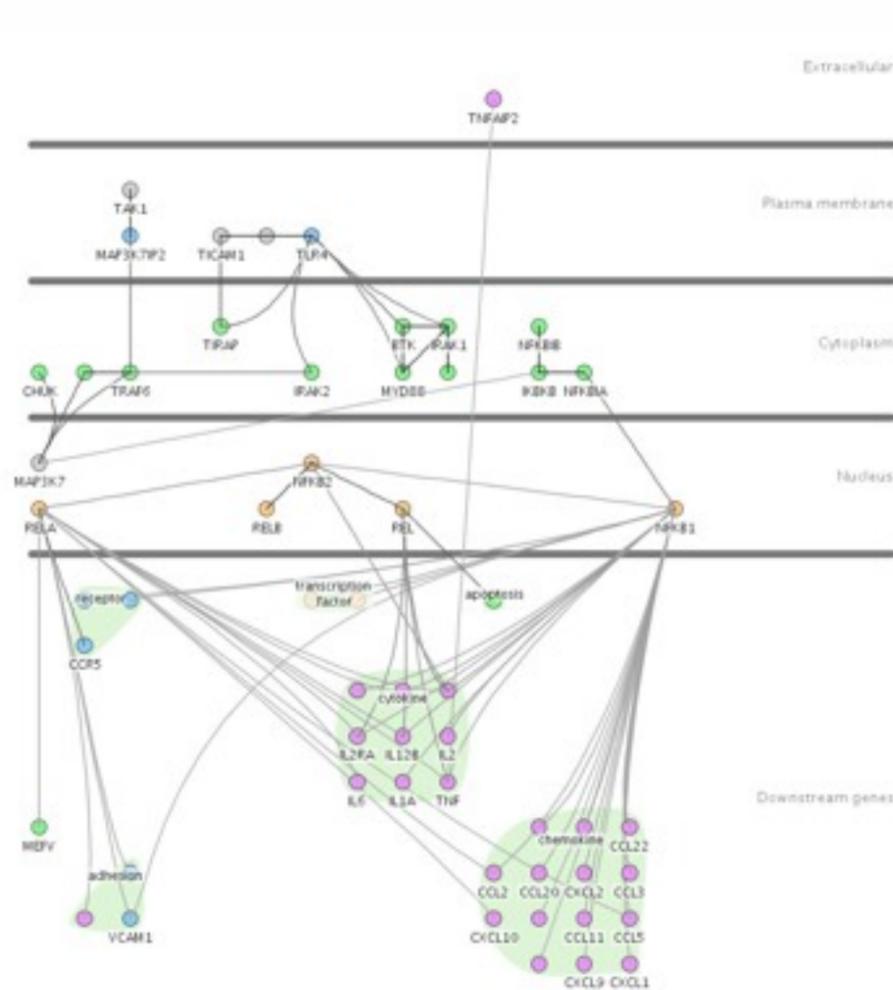


[Cerebral: Visualizing Multiple Experimental Conditions on a Graph with Biological Context. Barsky, Munzner, Gardy, and Kincaid. IEEE TVCG (Proc. InfoVis) 14(6):1253-1260, 2008.]

Why have a computer in the loop?

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

- beyond human patience: scale to large datasets, support interactivity
 - consider: what aspects of hand-drawn diagrams are important?



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

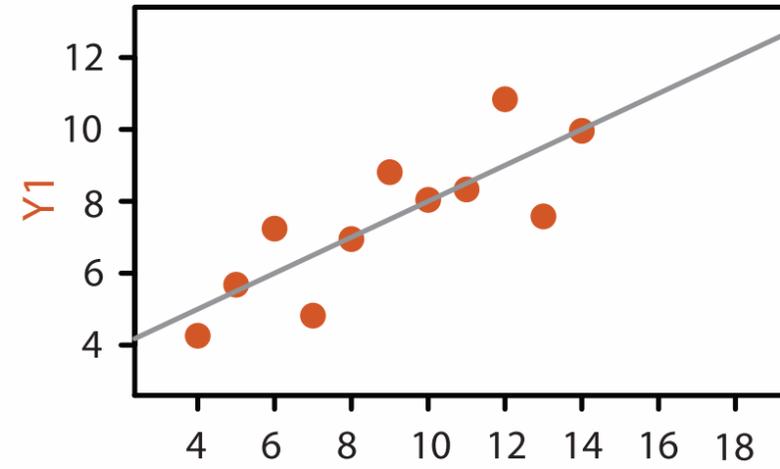
Why show the data in detail?

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

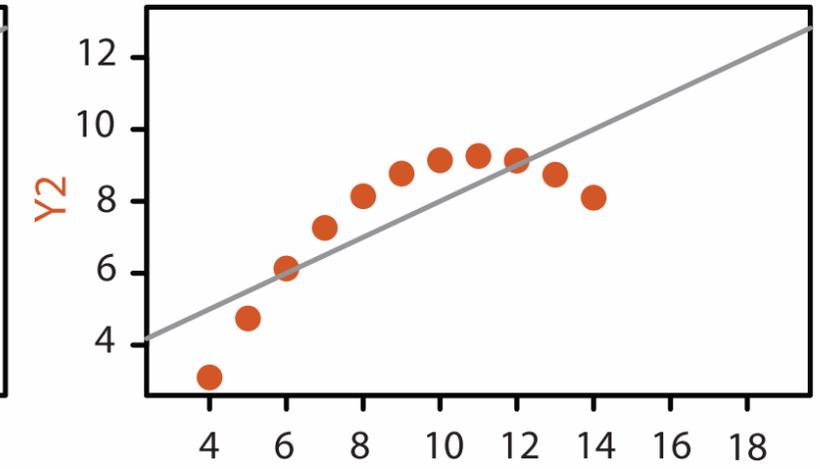
Anscombe's Quartet

Identical statistics

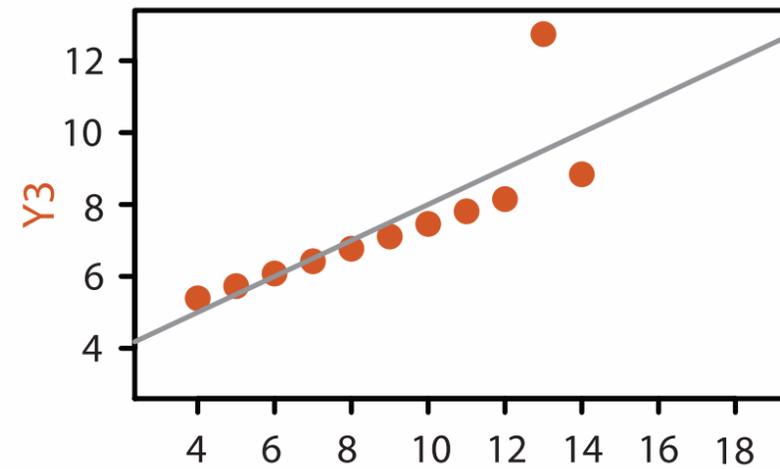
x mean	9
x variance	10
y mean	8
y variance	4
x/y correlation	1



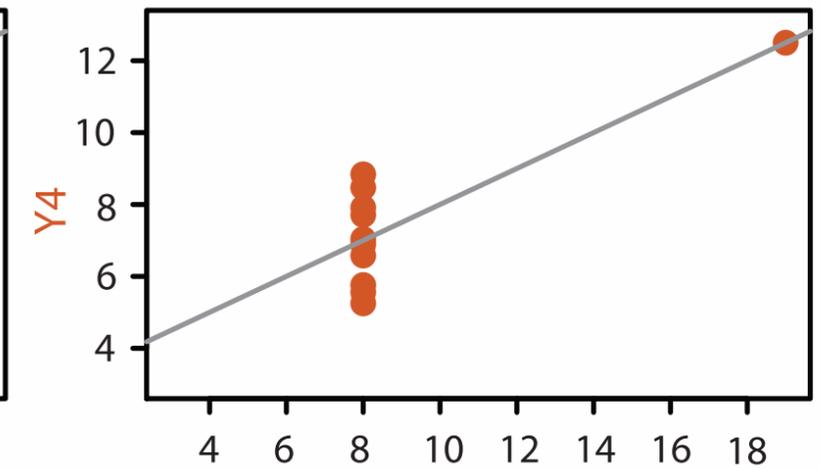
X1



X2



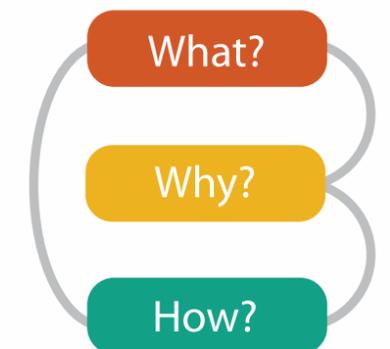
X3



X4

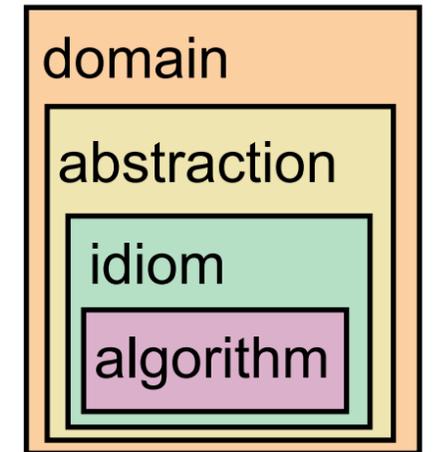
Why analyze?

- huge design space
 - visual encoding: combinatorial explosion of choices
 - add interaction: even bigger
 - add data abstraction transformation: truly enormous
- most possibilities ineffective for particular task/data combination
 - implication: avoid random walk, be guided by principles
- analysis framework: scaffold to think systematically about design space
 - ensure that consideration space encompasses full scope of possibilities
 - improve chances that selected solution is good not mediocre
 - next week's focus: abstractions and idioms, what-why-how

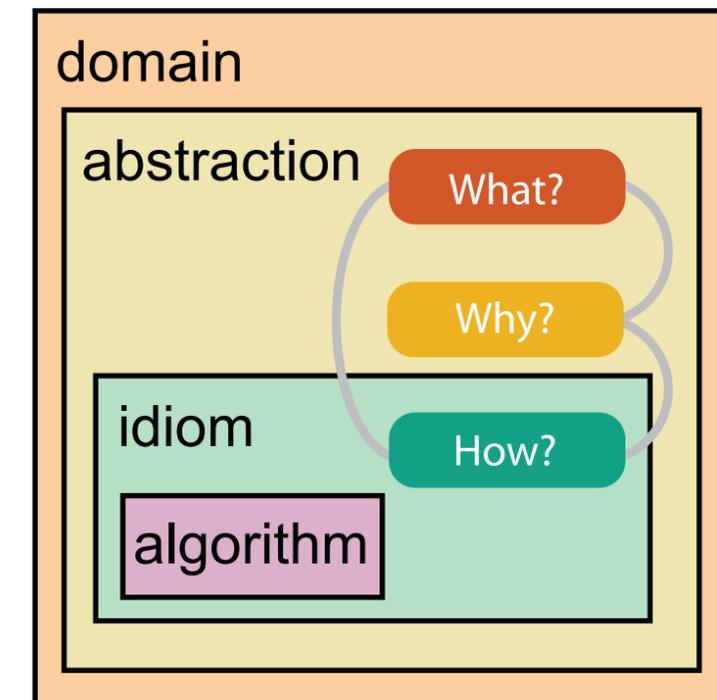


Analysis framework: Four levels, three questions

- *domain* situation
 - who are the target users?
- *abstraction*
 - translate from specifics of domain to vocabulary of vis
 - **what** is shown? **data abstraction**
 - **why** is the user looking at it? **task abstraction**
- *idiom*
 - **how** is it shown?
 - **visual encoding idiom**: how to draw
 - **interaction idiom**: how to manipulate
- *algorithm*
 - efficient computation



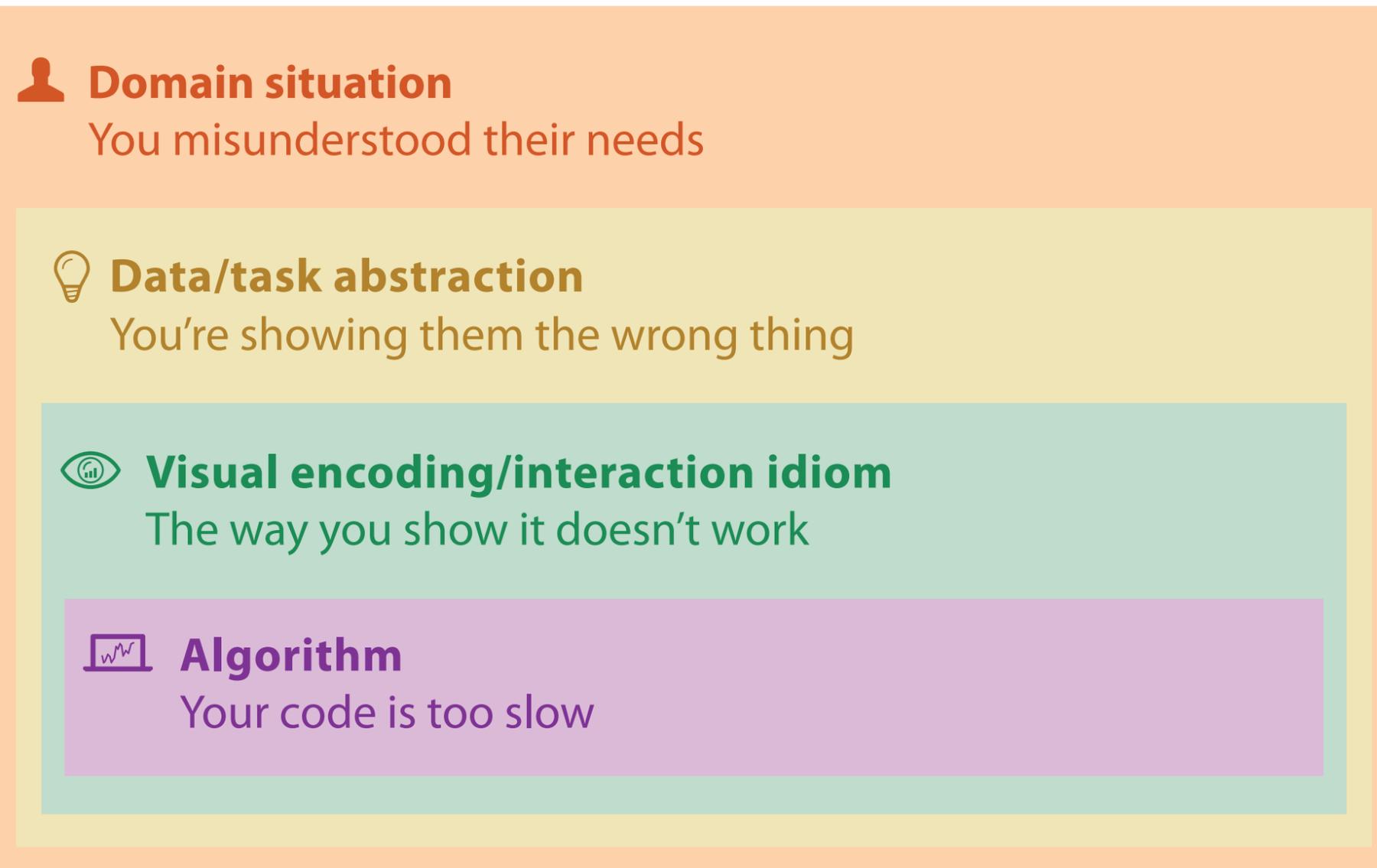
[A Nested Model of Visualization Design and Validation. Munzner. *IEEE TVCG* 15(6):921-928, 2009 (Proc. InfoVis 2009).]



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

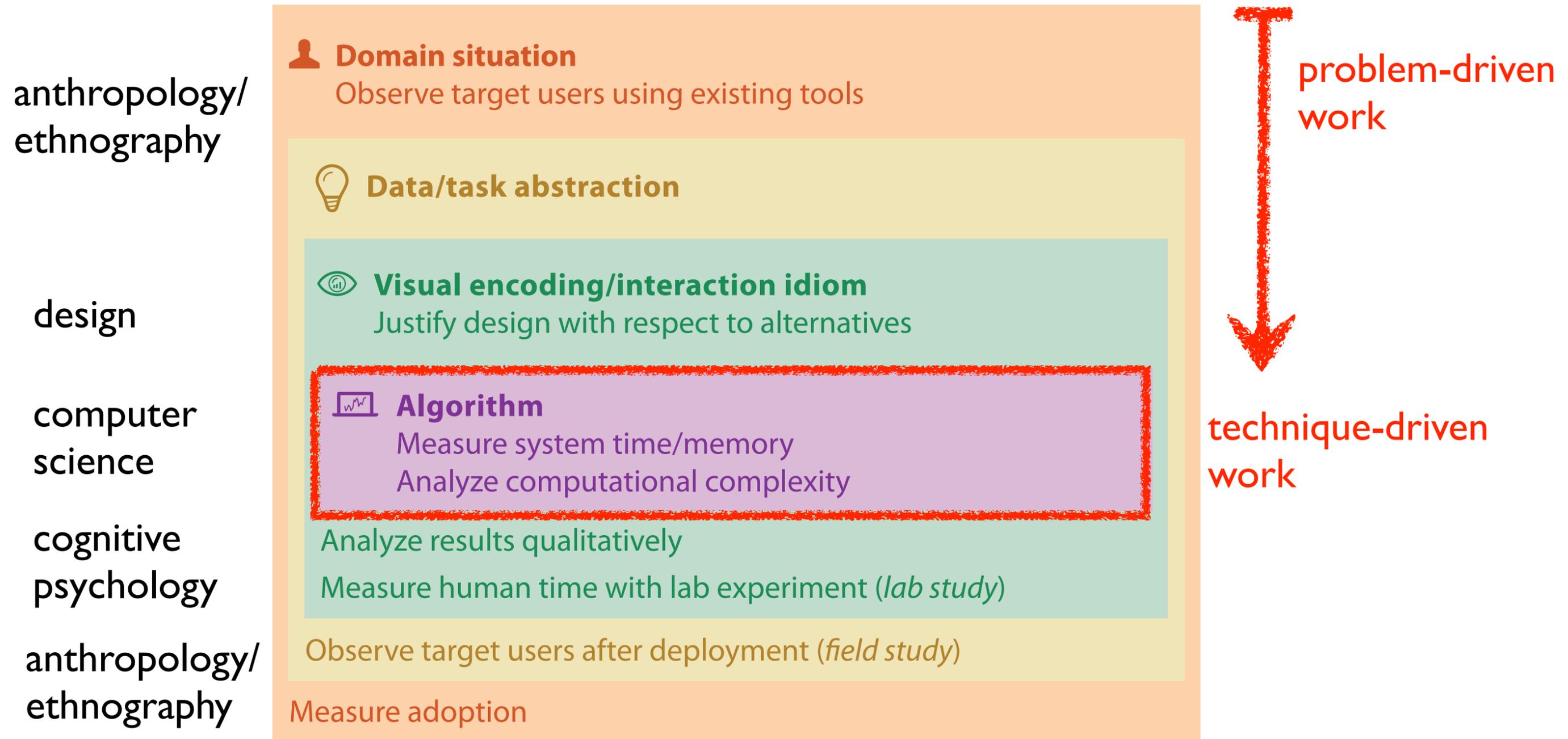
Why is validation difficult?

- different ways to get it wrong at each level



Why is validation difficult?

- solution: use methods from different fields at each level



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

Why are there resource limitations?

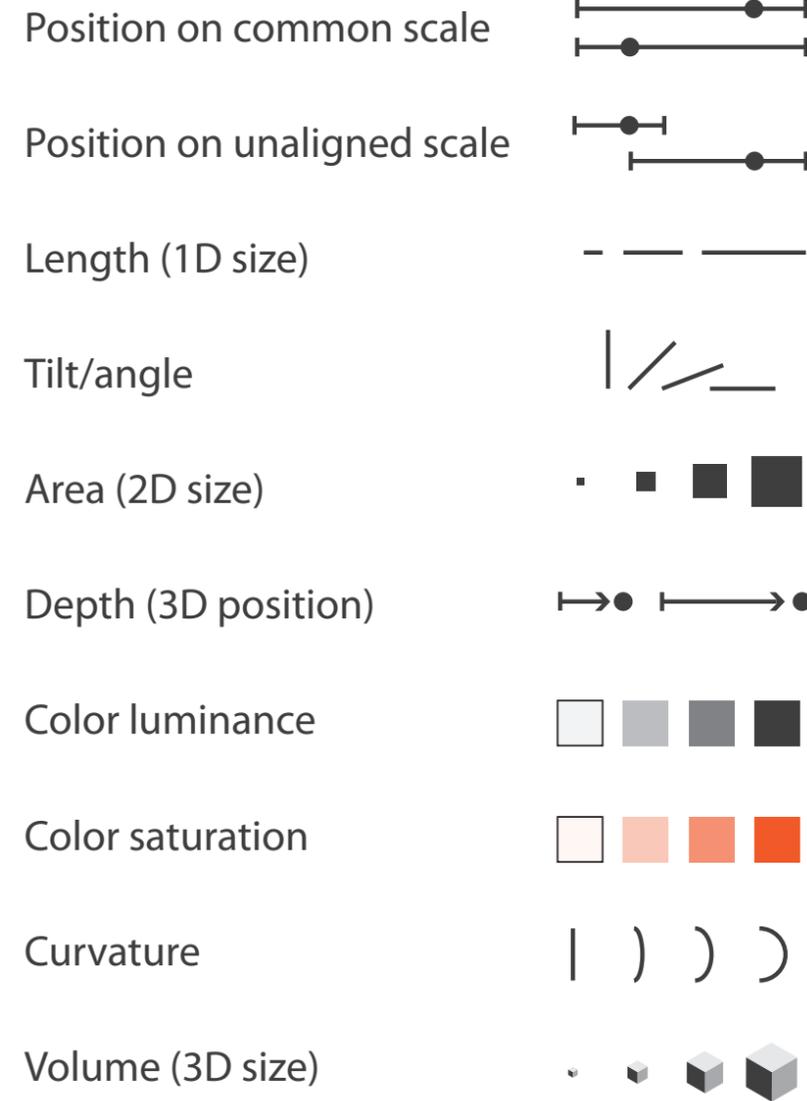
Vis designers must take into account three very different kinds of resource limitations: those of computers, of humans, and of displays.

- 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

VAD Ch 5: Marks and Channels

Channels: Expressiveness Types and Effectiveness Ranks

➔ **Magnitude Channels: Ordered Attributes**



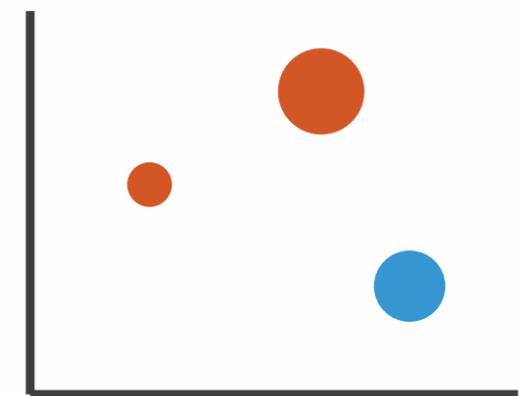
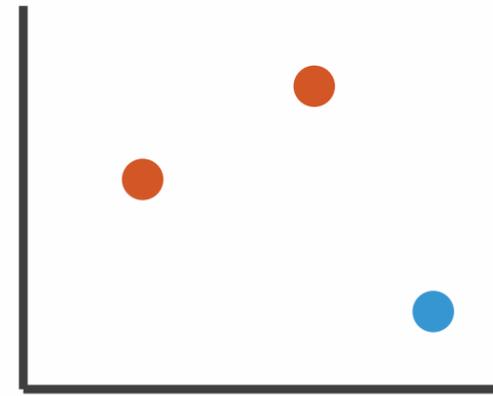
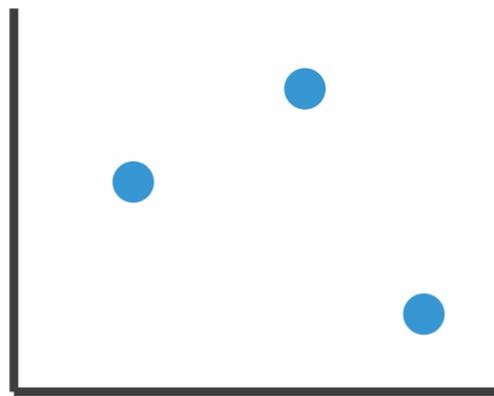
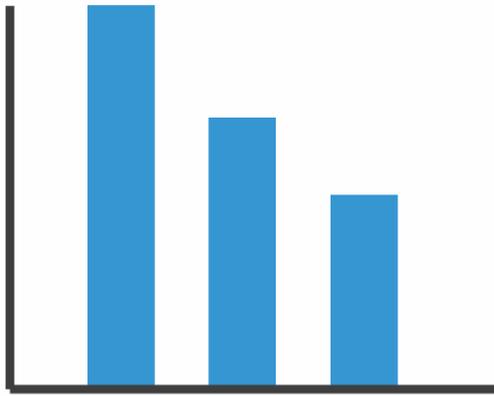
➔ **Identity Channels: Categorical Attributes**



[VAD Fig 5.1]

Encoding visually

- analyze idiom structure



Definitions: Marks and channels

- marks

 - geometric primitives

→ Points



→ Lines



→ Areas



- channels

 - control appearance of marks

→ Position

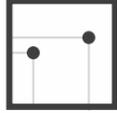
→ Horizontal



→ Vertical



→ Both



→ Color



→ Shape



→ Tilt



→ Size

→ Length



→ Area

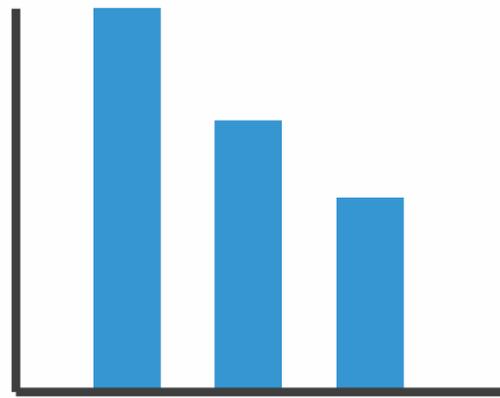


→ Volume



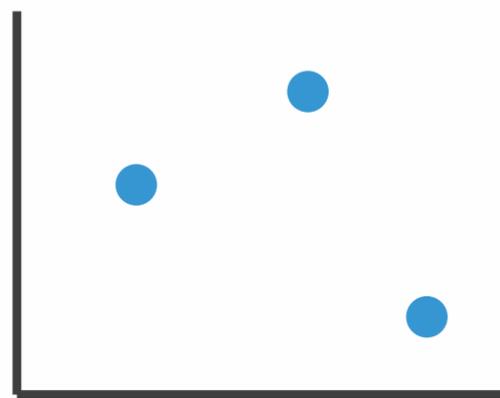
Encoding visually with marks and channels

- analyze idiom structure
 - as combination of marks and channels



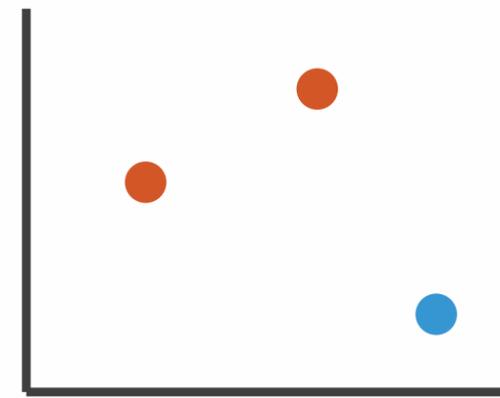
1:
vertical position

mark: line



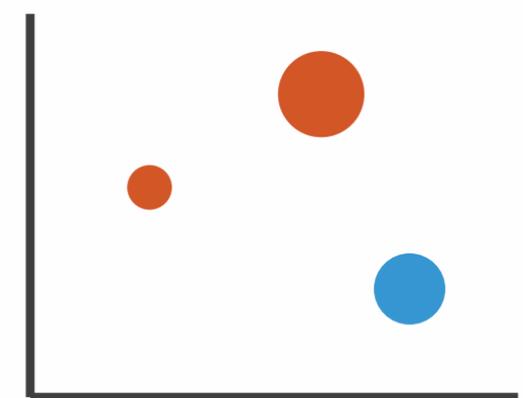
2:
vertical position
horizontal position

mark: point



3:
vertical position
horizontal position
color hue

mark: point



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

mark: point

Channels

Position on common scale



Position on unaligned scale



Length (1D size)



Tilt/angle



Area (2D size)



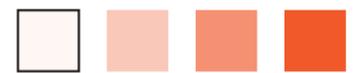
Depth (3D position)



Color luminance



Color saturation



Curvature



Volume (3D size)



Same

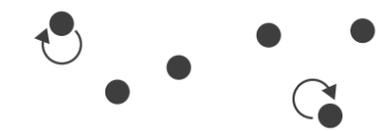
Spatial region



Color hue



Motion

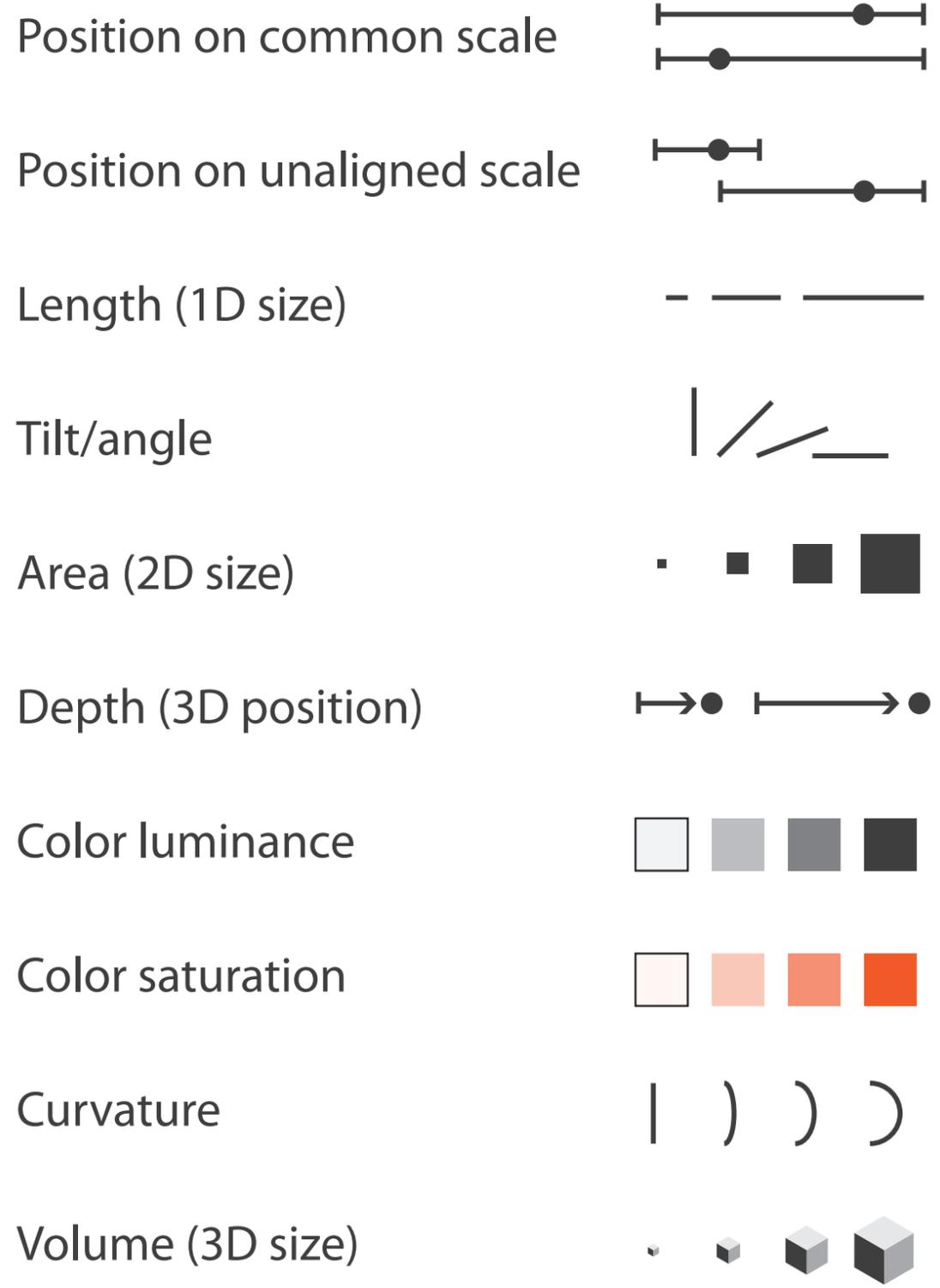


Shape



Channels: Rankings

➔ Magnitude Channels: Ordered Attributes



➔ Identity Channels: Categorical Attributes



Best

Effectiveness

Least

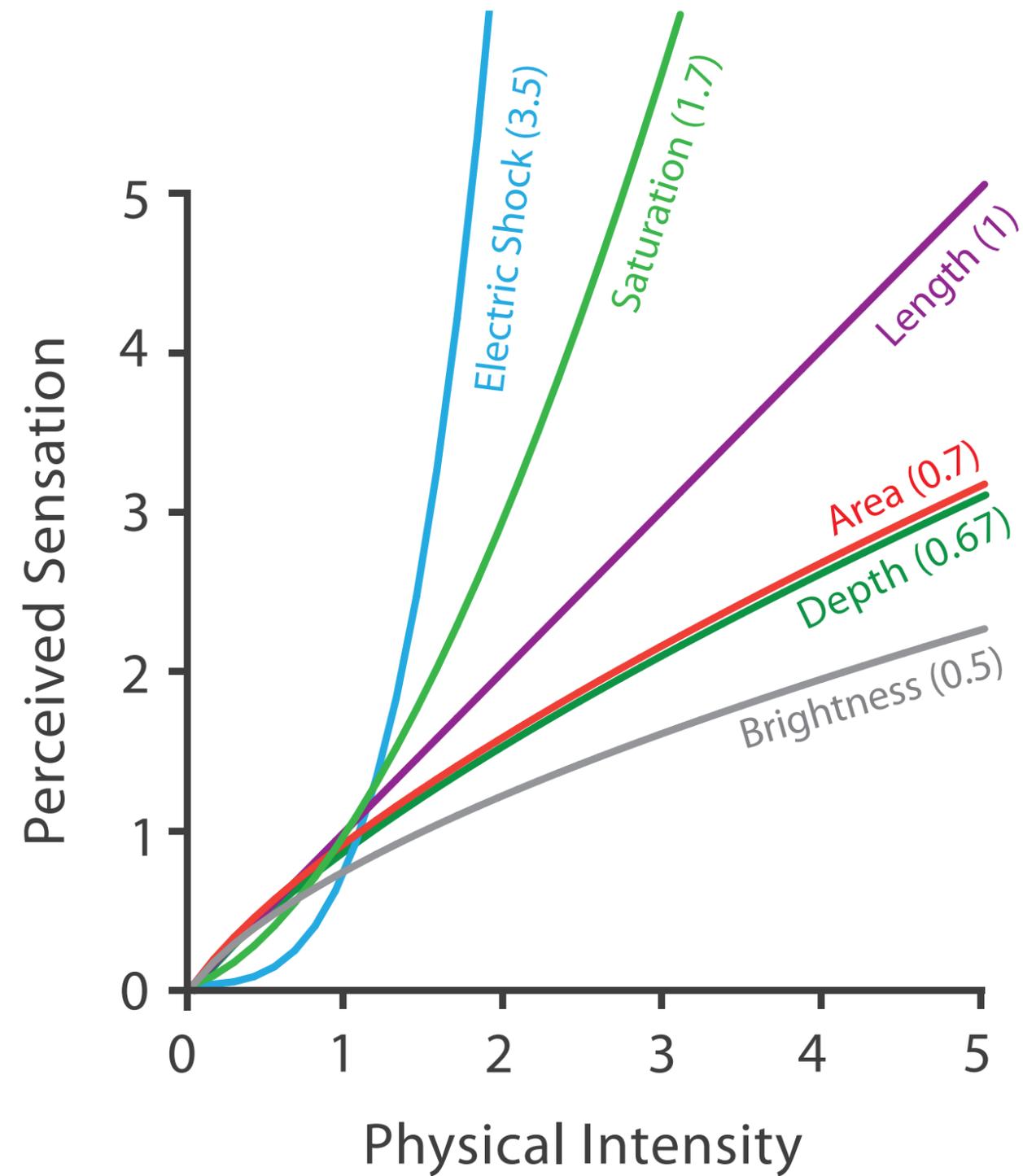
Same

Same

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

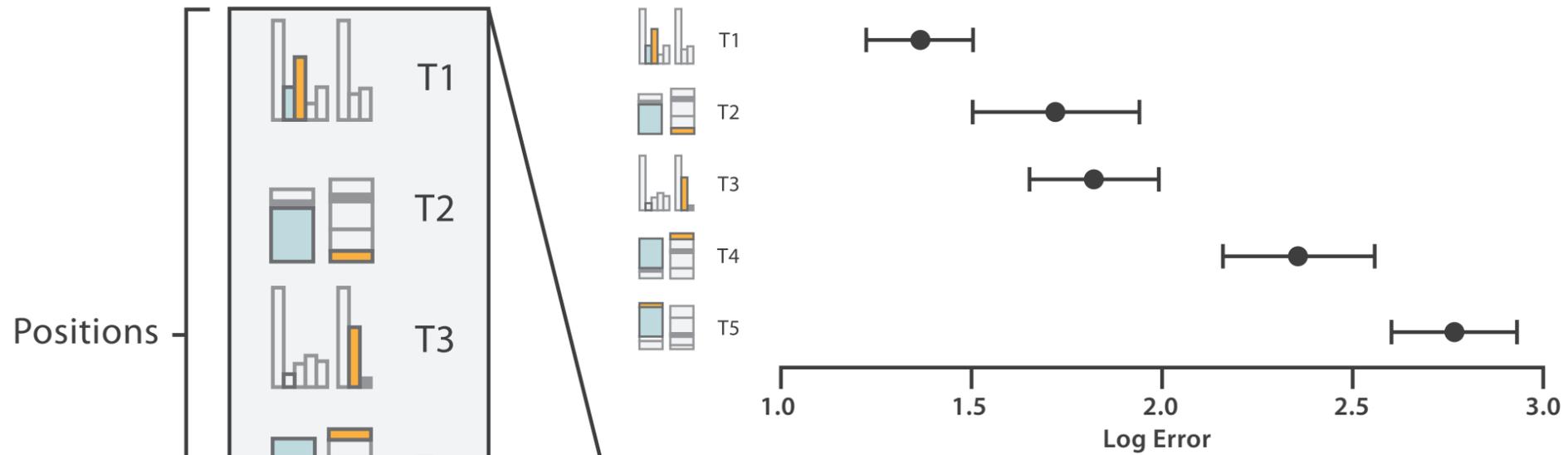
Accuracy: Fundamental Theory

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

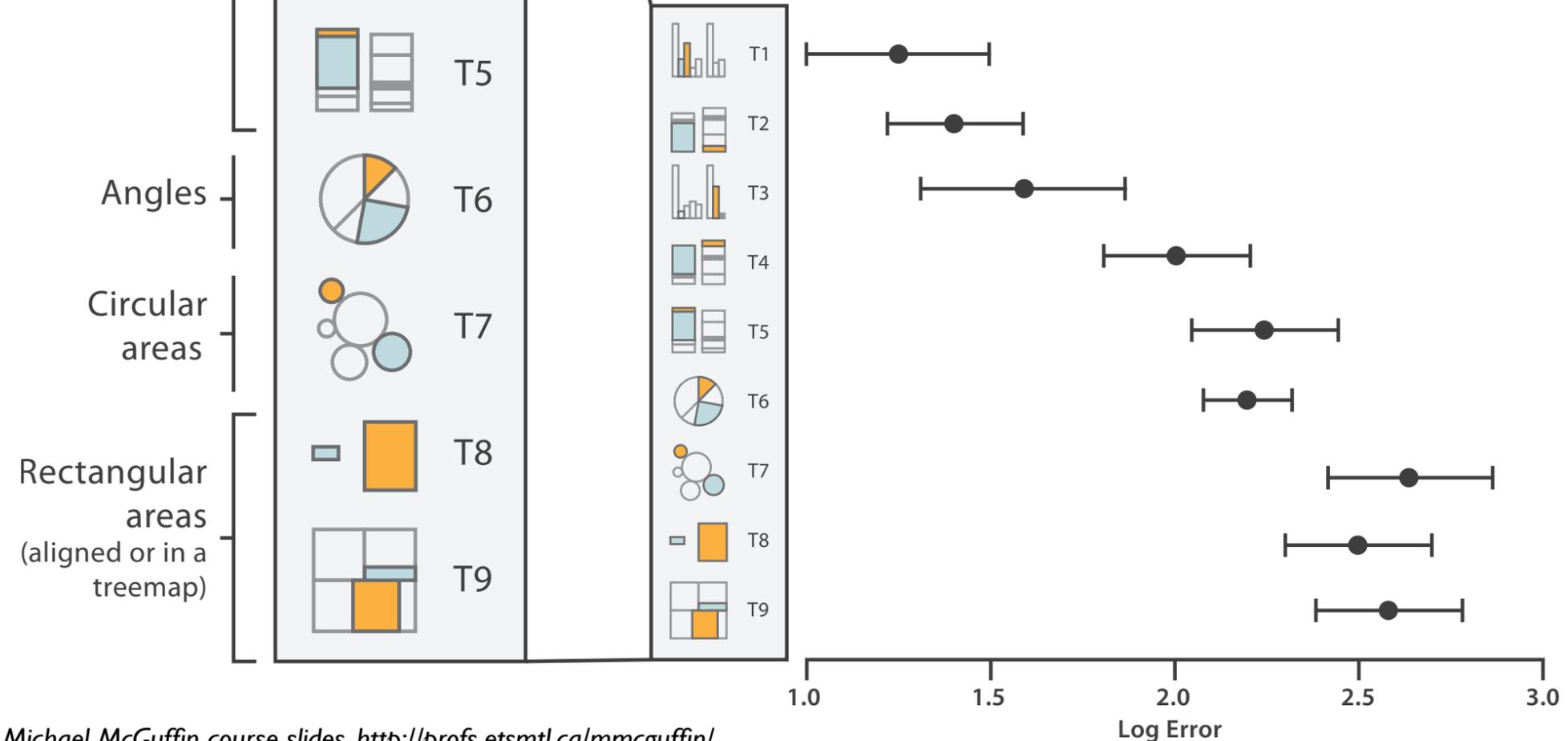


Accuracy: Vis experiments

Cleveland & McGill's Results



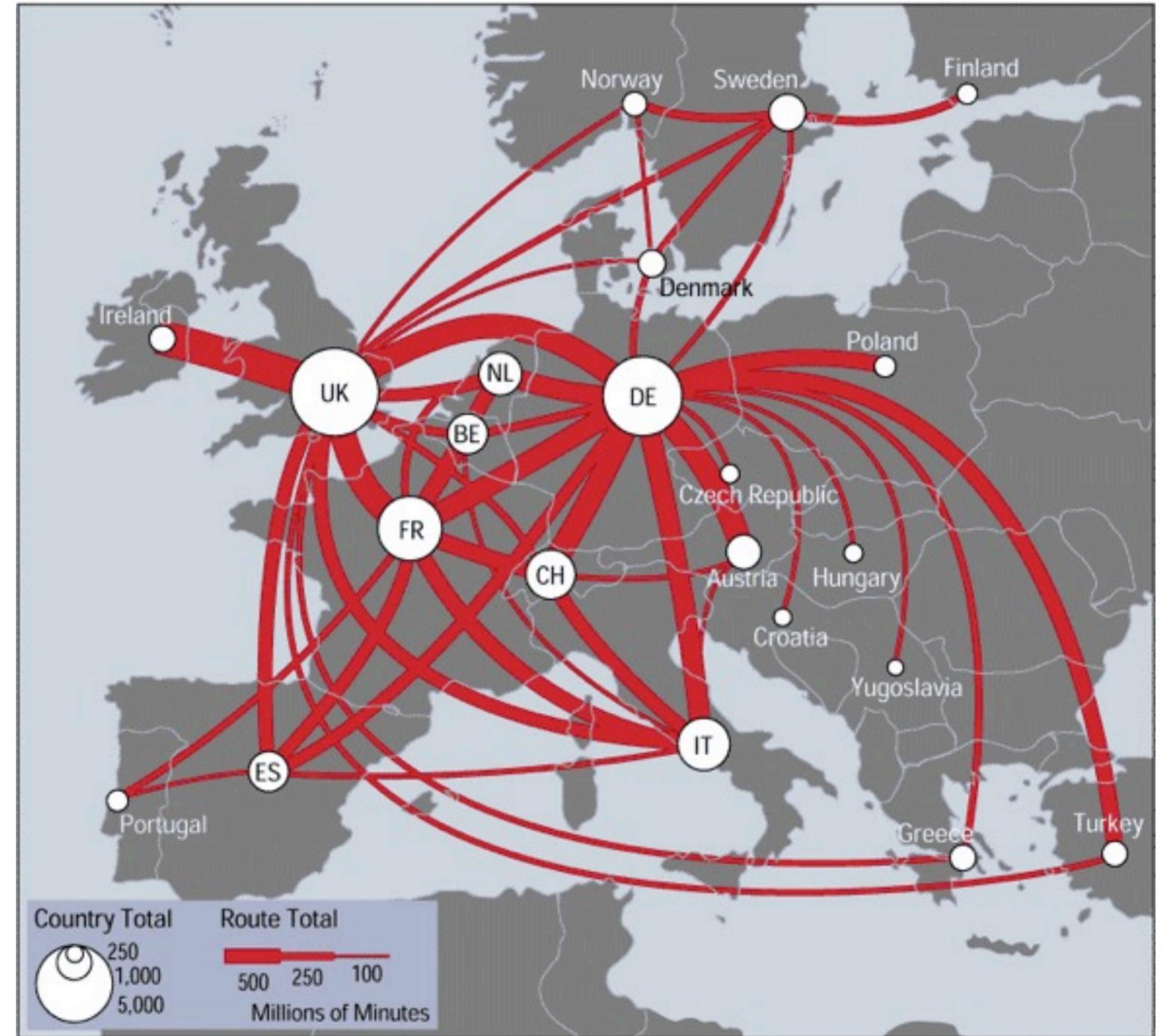
Crowdsourced Results



[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?

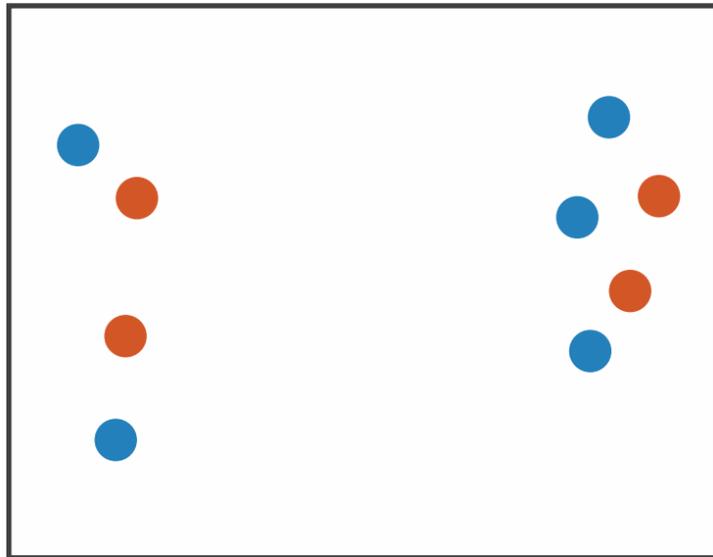
- must be sufficient for number of attribute levels to show
 - linewidth: few bins



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

Separability vs. Integrality

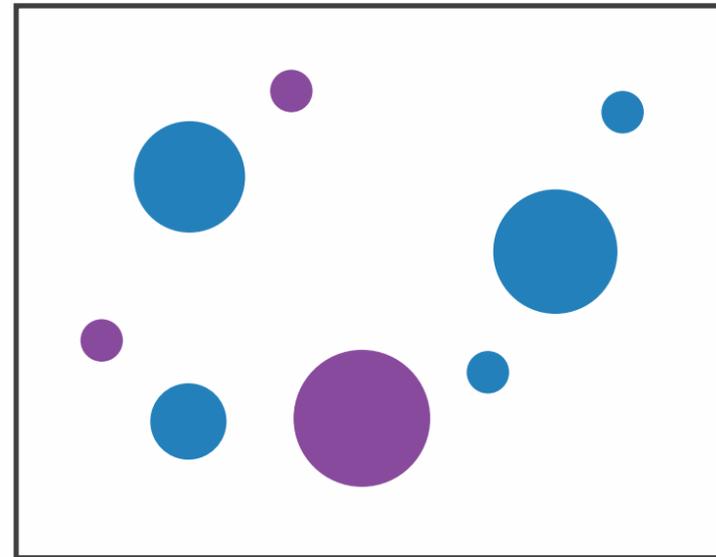
Position
+ Hue (Color)



Fully separable

2 groups each

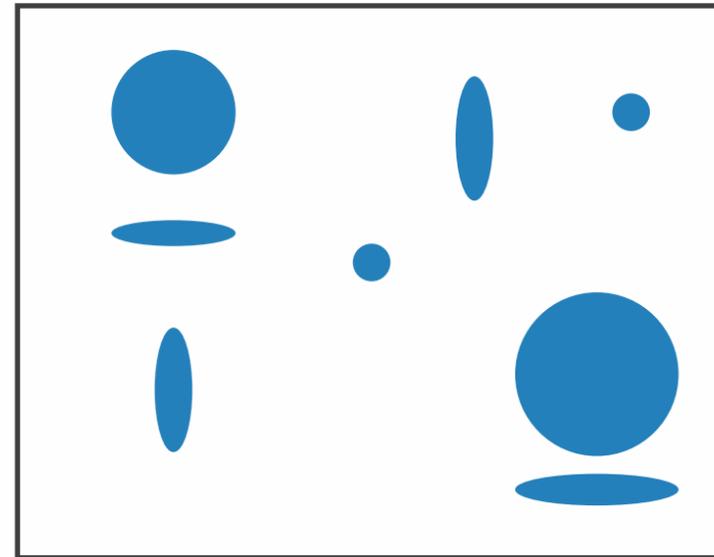
Size
+ Hue (Color)



Some interference

2 groups each

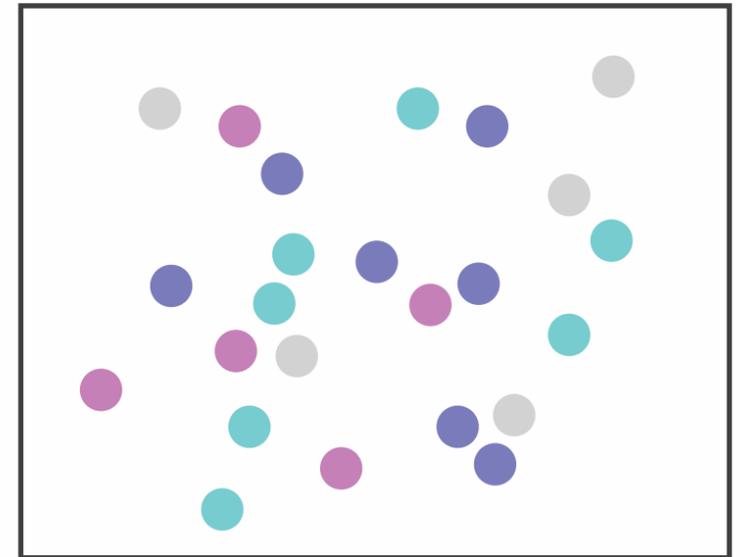
Width
+ Height



Some/significant
interference

3 groups total:
integral area

Red
+ Green

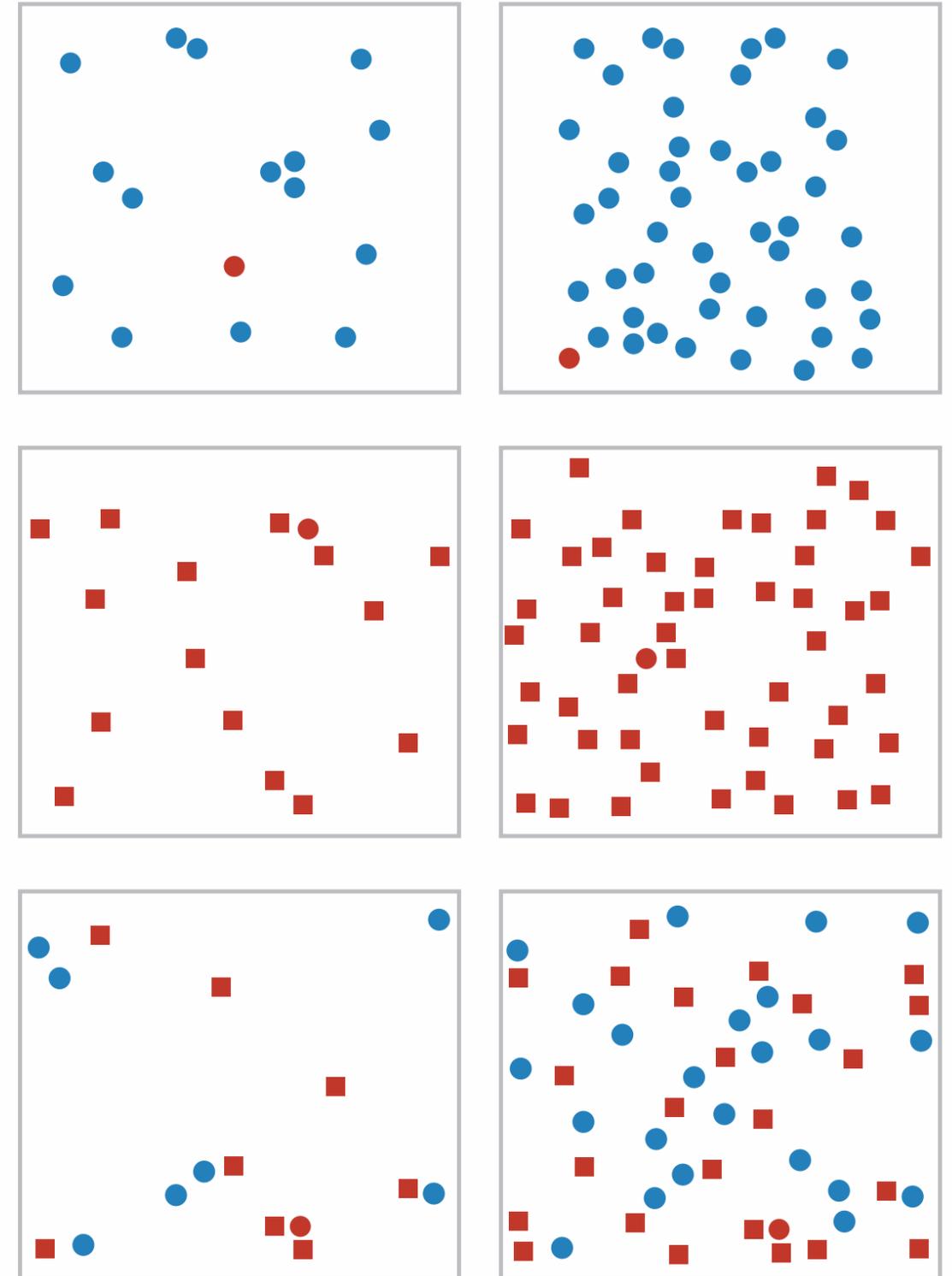


Major interference

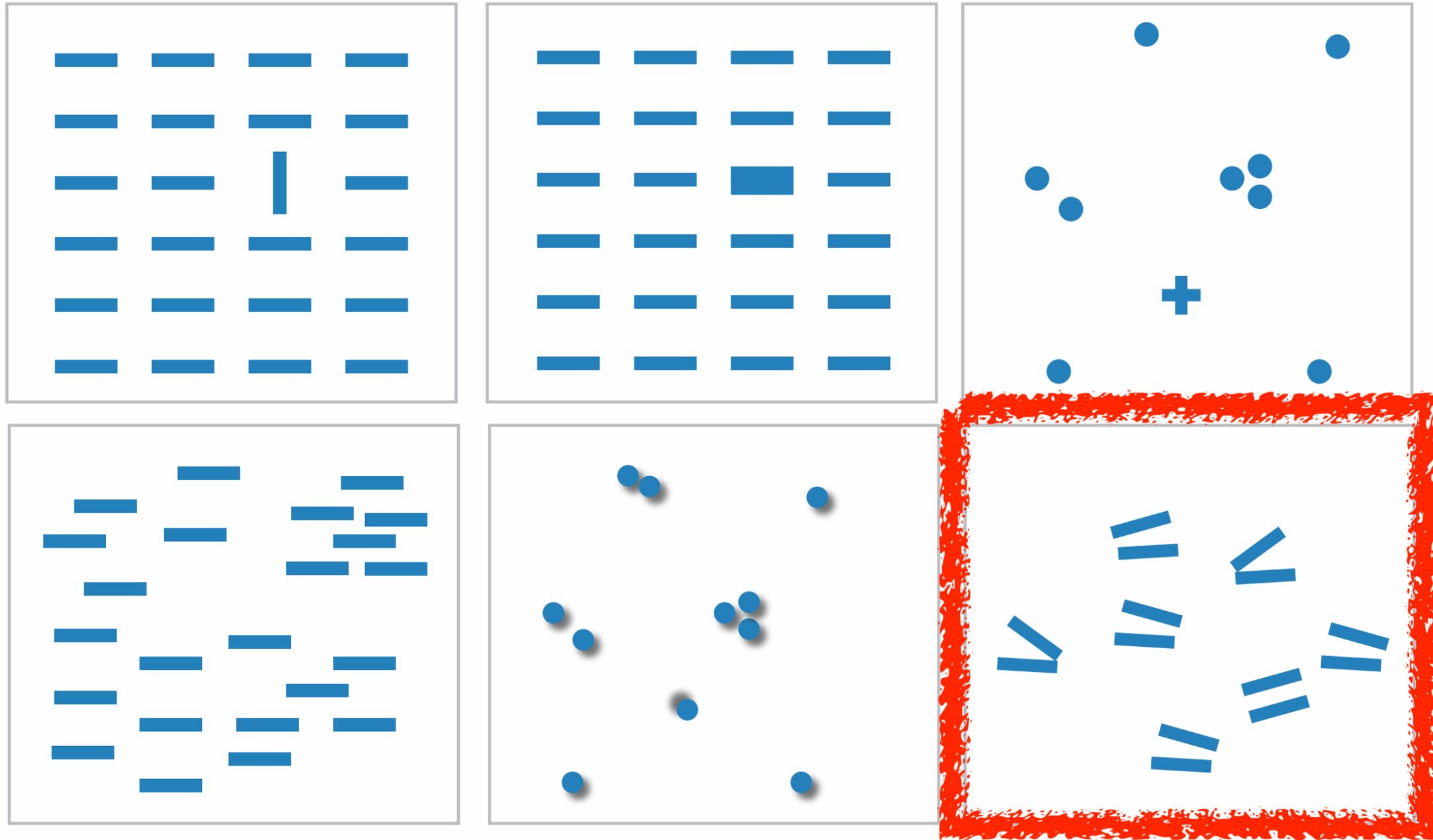
4 groups total:
integral hue

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

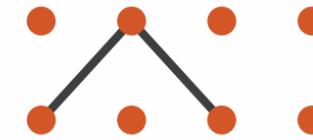
- proximity
 - same spatial region
- similarity
 - same values as other categorical channels

Marks as Links

➔ Containment



➔ Connection



➔ Identity Channels: Categorical Attributes

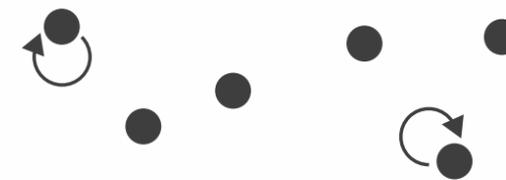
Spatial region



Color hue



Motion



Shape

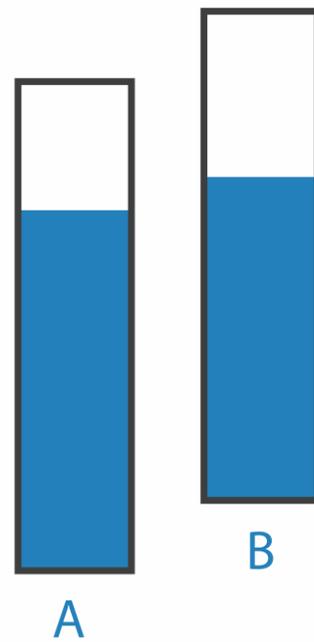


Relative vs. absolute judgements

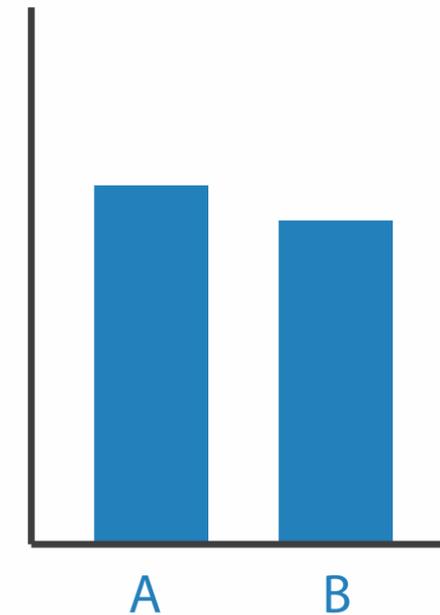
- perceptual system mostly operates with relative judgements, not absolute
 - that's why accuracy increases with common frame/scale and alignment
 - Weber's Law: ratio of increment to background is constant
 - filled rectangles differ in length by 1:9, difficult judgement
 - white rectangles differ in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



length



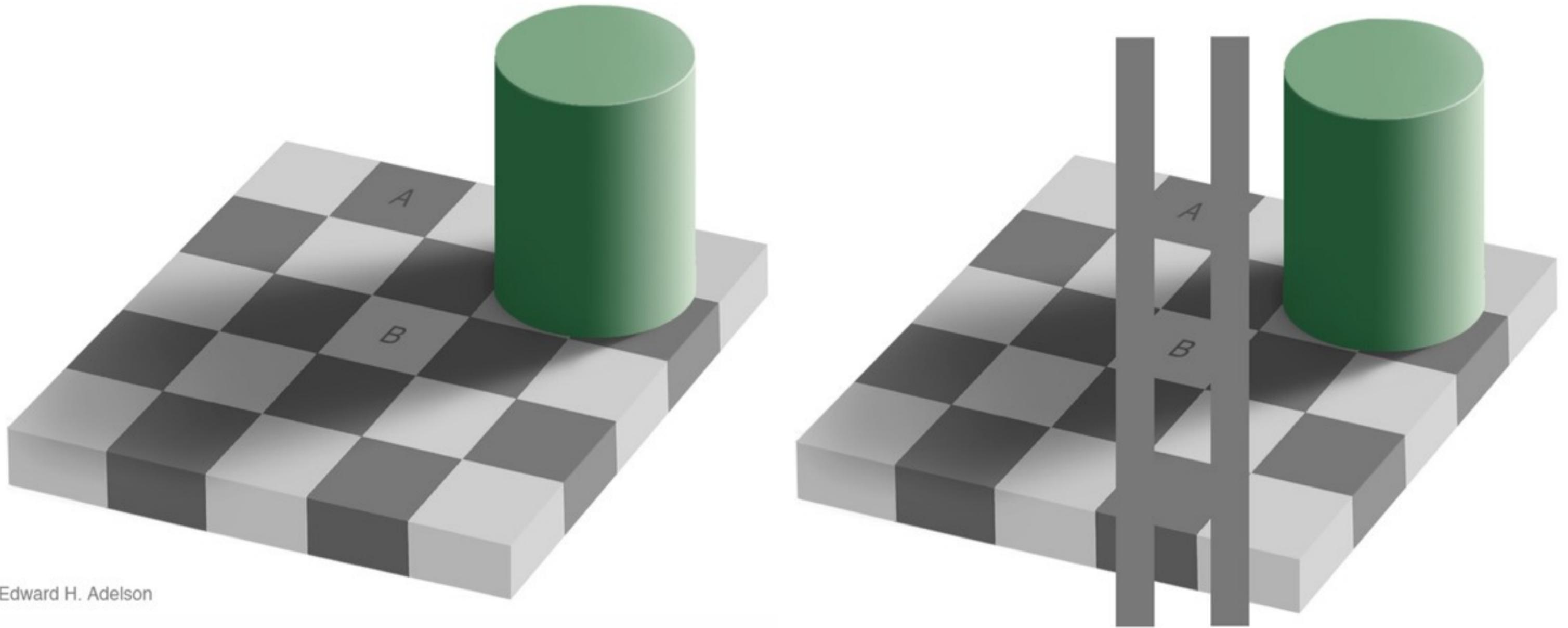
position along
unaligned
common scale



position along
aligned scale

Relative luminance judgements

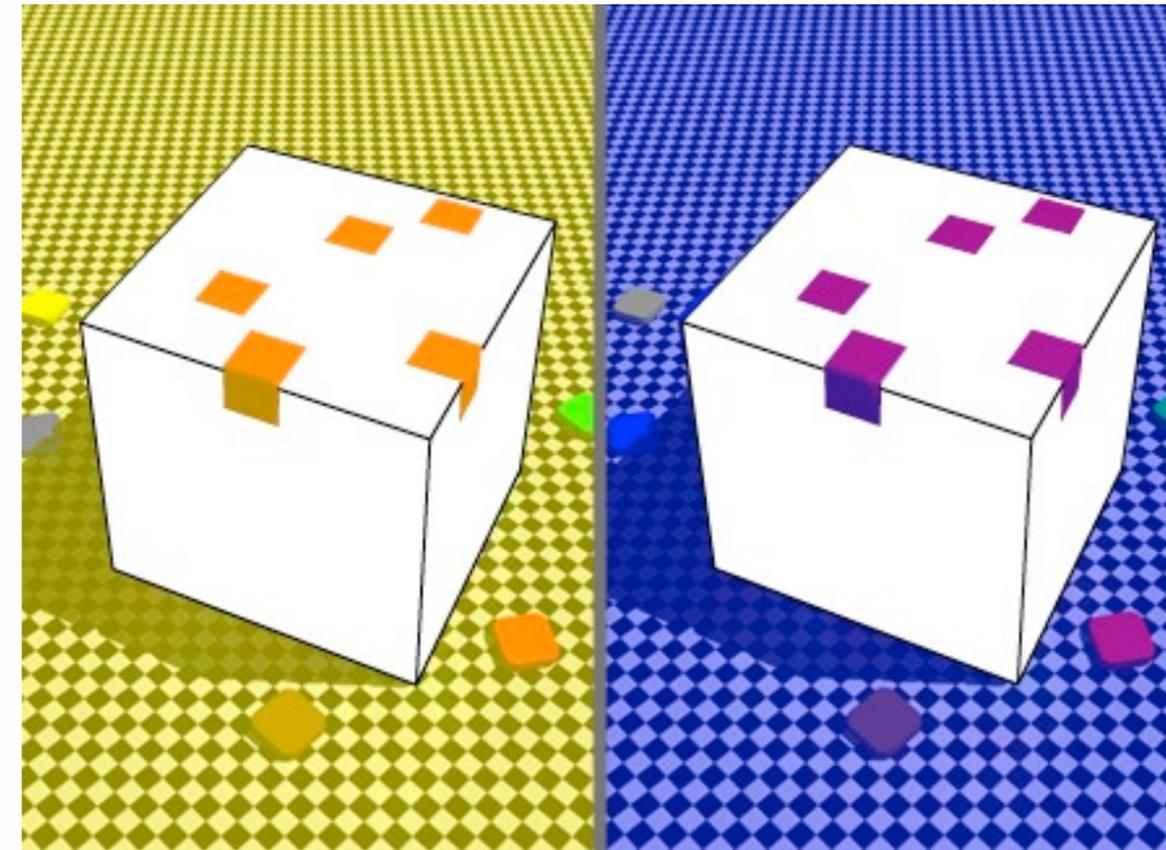
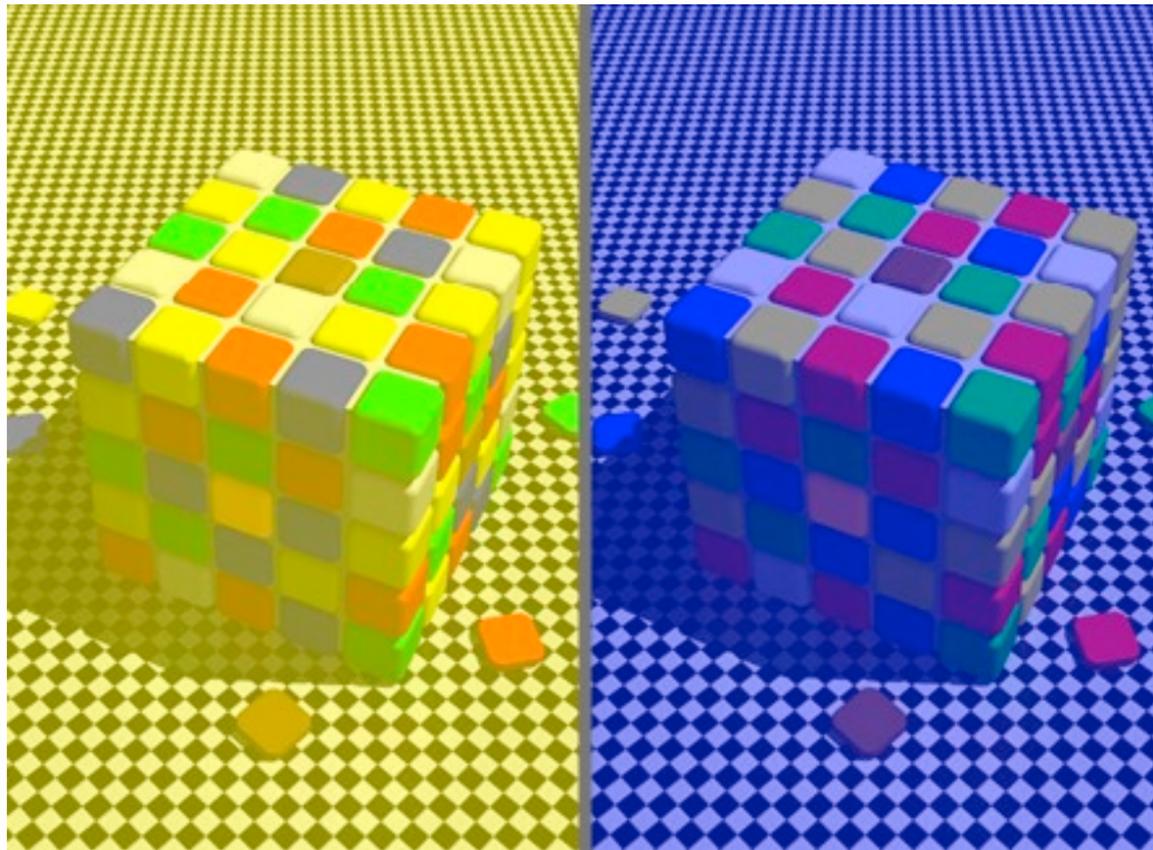
- perception of luminance is contextual based on contrast with surroundings



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 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.

Now

- Break (15 min)
- Demo: Guest lecture/demo from Robert Kosara on Tableau
- Lab: you'll try it!

Lab/Assignment (Updated after class)

- install Tableau on your own laptop
 - using course key from me or individual license key that you request personally
- work through Vienna tutorial (data: Chicago crime 2015, US forest fires)
- work through intro tutorial (data: music sales)
- download 1033 dataset from Tableau Public
 - play with it based on what you learned from Robert's demo
- pick three datasets from Tableau public
 - visualize them with Tableau with what you learned from demo and tutorials, also try at least two new features for each
- submit next week
 - by 9am Tue, email tmm@cs.ubc.ca with subject JOURN Week 1
 - reflections on what you've found in the 7 datasets
 - text illustrated by screenshots of what you've created, in PDF format
 - what did you find in the vis?
 - could you tell a story to others? could you get a sense of the story for yourself? did you find nothing useful?