

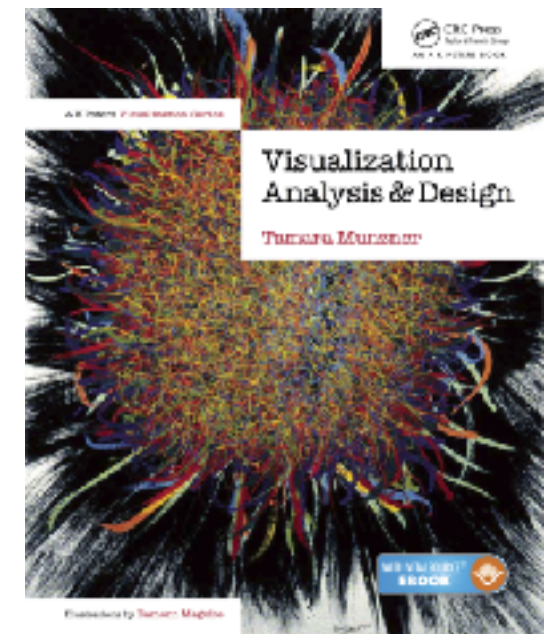
# Visualization Analysis & Design

## *Marks & Channels (Ch 5) I*

**Tamara Munzner**

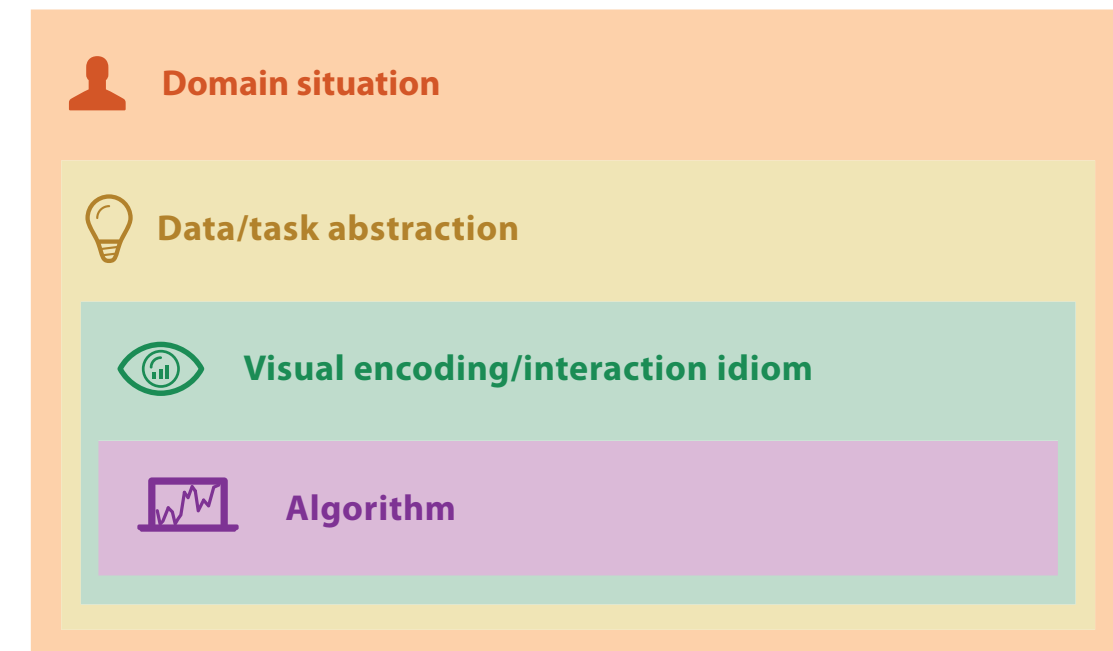
Department of Computer Science  
University of British Columbia

[@tamaramunzner](#)



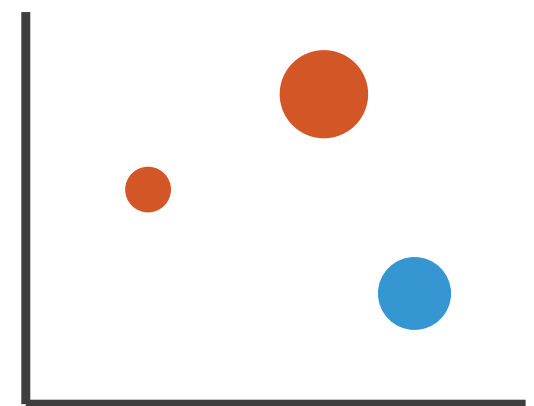
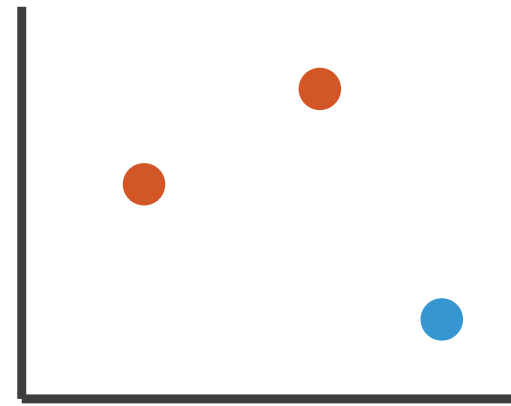
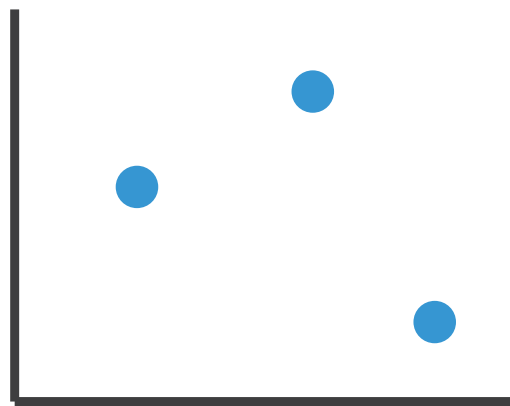
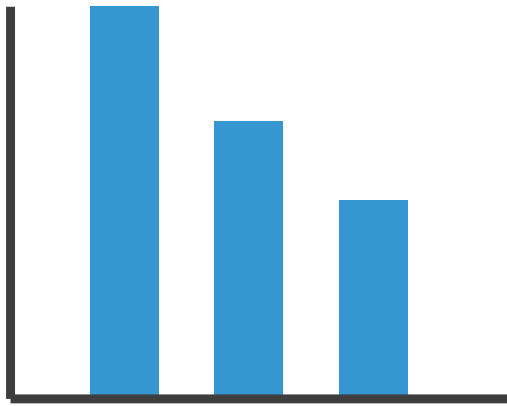
# Visual encoding

- how to systematically analyze idiom structure?



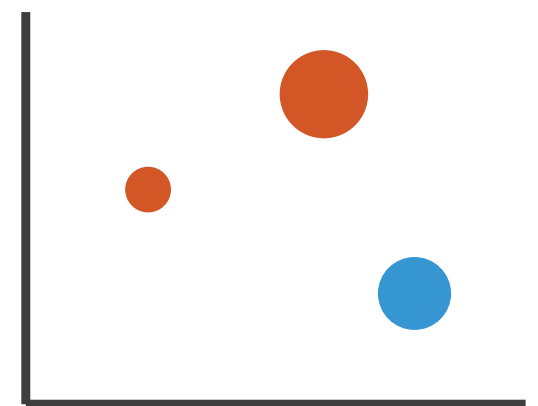
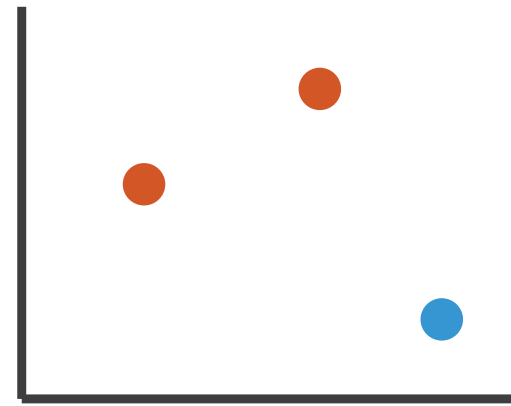
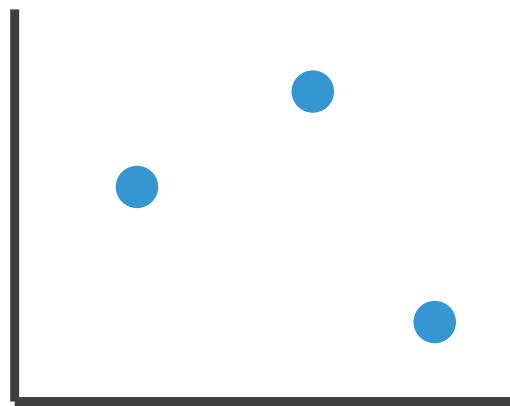
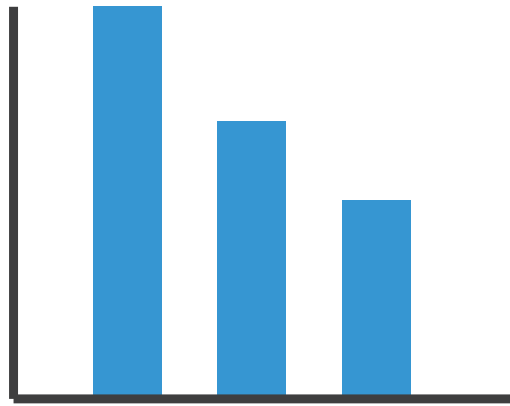
# Visual encoding

- how to systematically analyze idiom structure?



# Visual encoding

- how to systematically analyze idiom structure?



- marks & channels
  - marks: represent items or links
  - channels: change appearance of marks based on attributes

# Marks for items

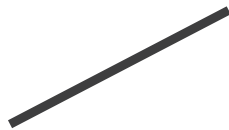
- basic geometric elements

➔ Points



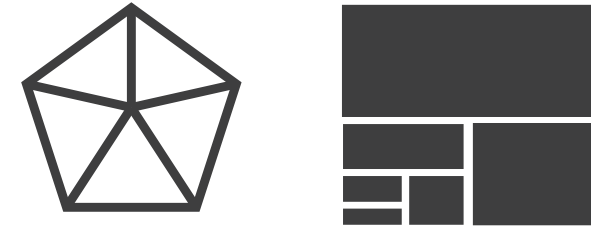
0D

➔ Lines



1D

➔ Interlocking Areas

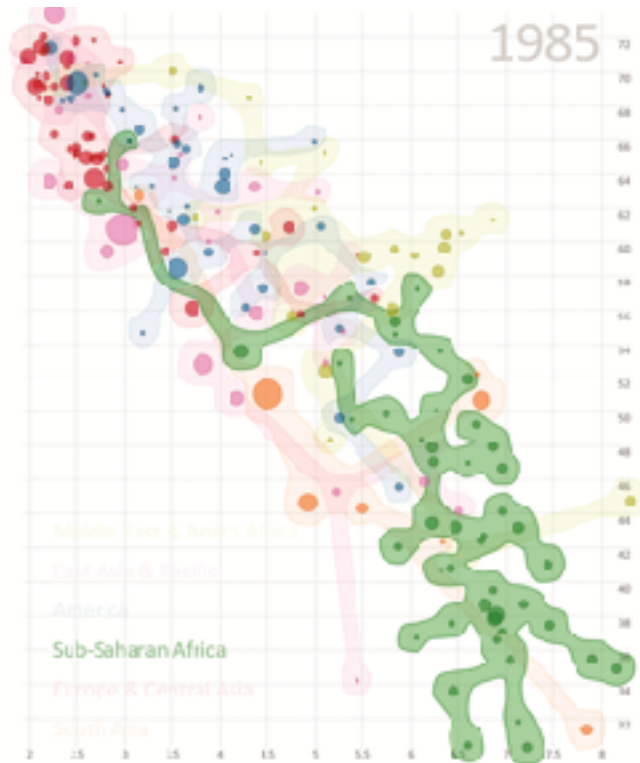
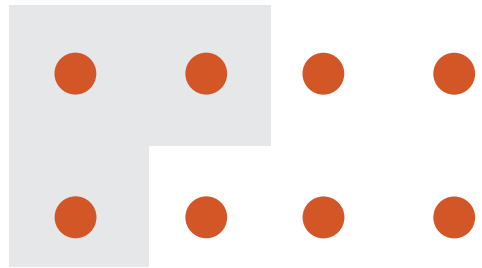


2D

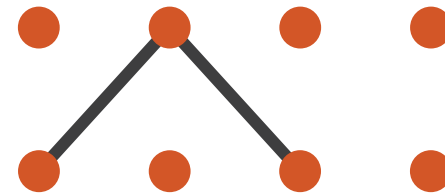
- 3D mark: volume, rarely used

# Marks for links

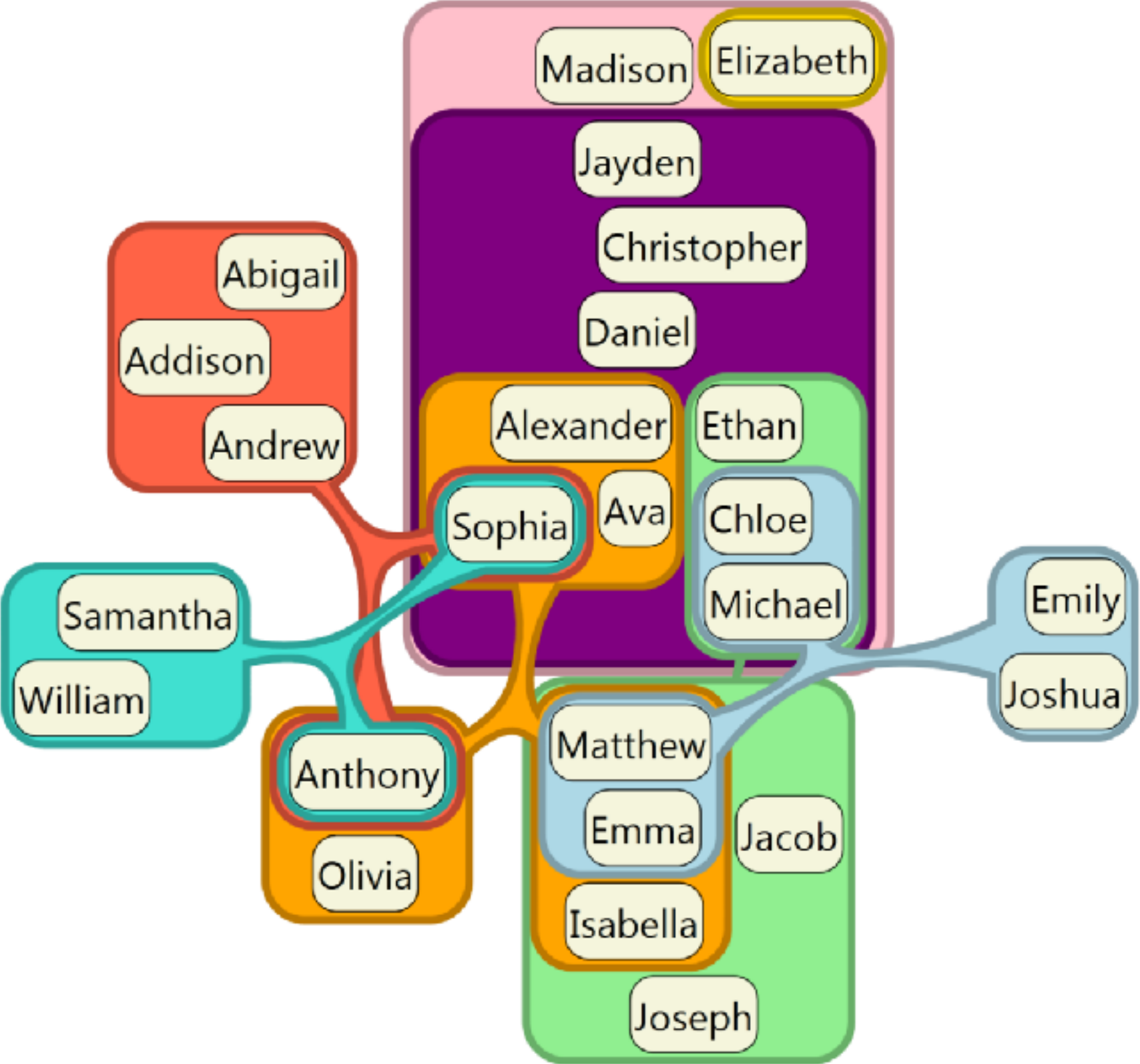
## ➔ Containment



## ➔ Connection



# Containment can be nested



[[Untangling Euler Diagrams, Riche and Dwyer, 2010](#)]

# Channels

- control appearance of marks
  - proportional to or based on attributes
- many names
  - **visual channels**
  - visual variables
  - retinal channels
  - visual dimensions
  - ...

## → Position

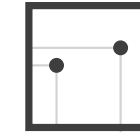
→ Horizontal



→ Vertical



→ Both



## → Shape



## → Size

→ Length



→ Area



→ Volume



## → Color



## → Tilt





# Definitions: Marks and channels

- marks
  - geometric primitives

→ Points



→ Lines



→ Areas



# Definitions: Marks and channels

- marks
  - geometric primitives
- channels
  - control appearance of marks

→ Points



→ Lines



→ Interlocking Areas



→ Position

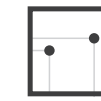
→ Horizontal



→ Vertical



→ Both



→ Color



→ Shape



→ Tilt



→ Size

→ Length



→ Area



→ Volume



# Definitions: Marks and channels

- marks
  - geometric primitives
- channels
  - control appearance of marks
- channel properties differ
  - type & amount of information that can be conveyed to human perceptual system

→ Points



→ Lines



→ Interlocking Areas



→ Position

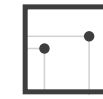
→ Horizontal



→ Vertical



→ Both



→ Color



→ Shape



→ Tilt



→ Size

→ Length



→ Area

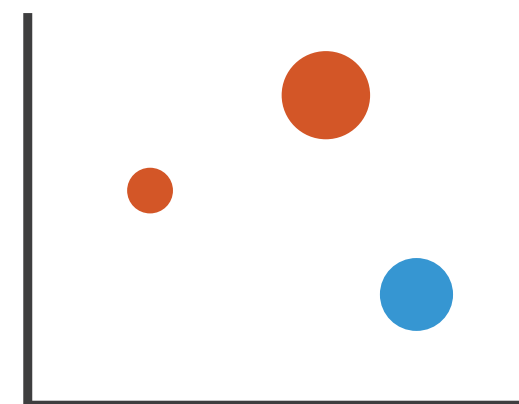
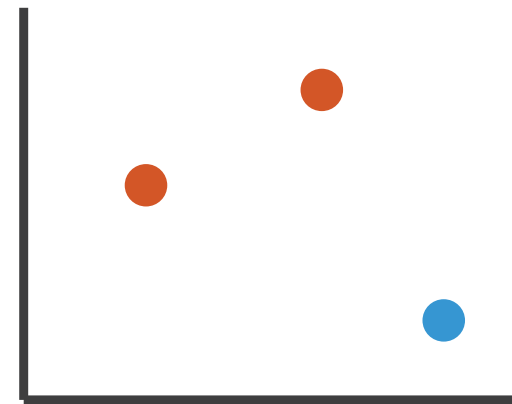
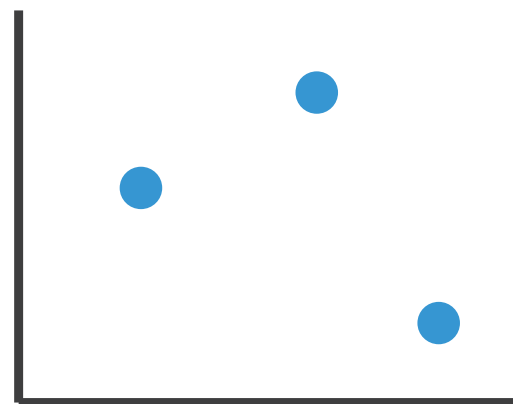
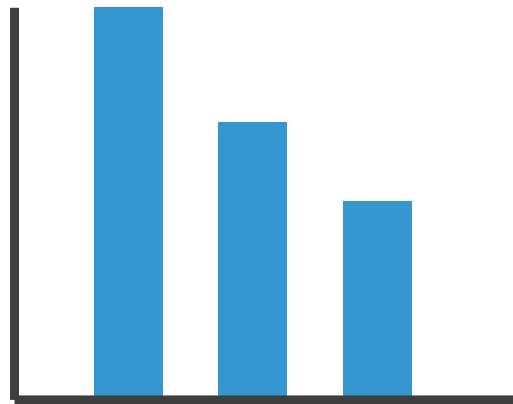


→ Volume



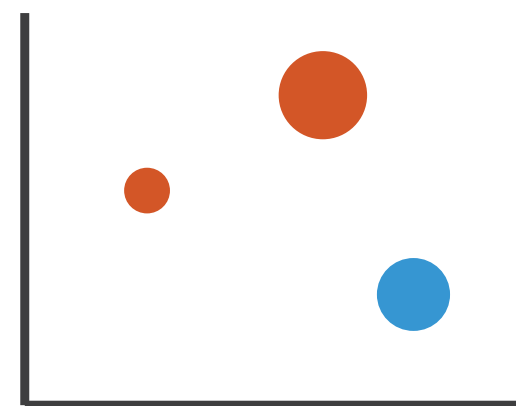
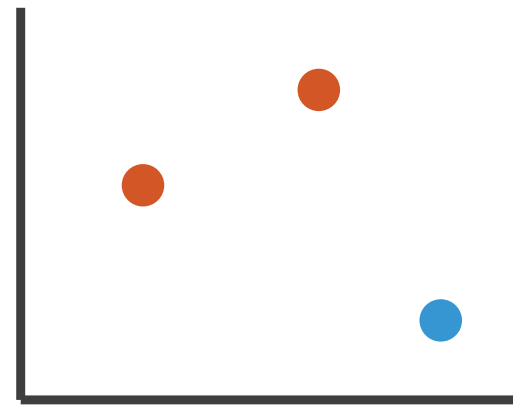
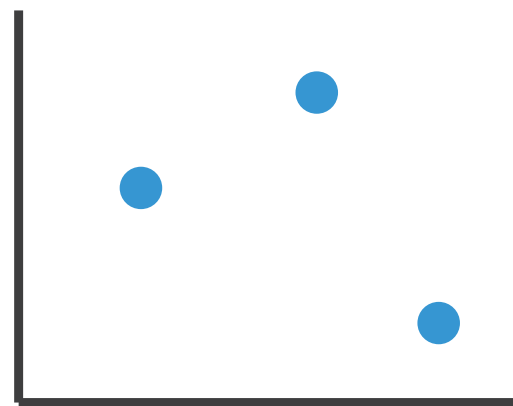
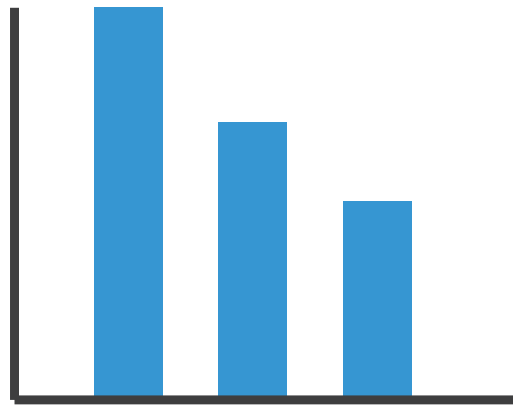
# Visual encoding

- analyze idiom structure as combination of marks and channels



# Visual encoding

- analyze idiom structure as combination of marks and channels

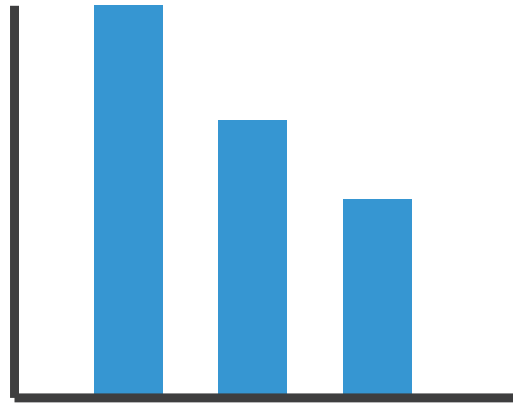


1:  
vertical position

mark: line

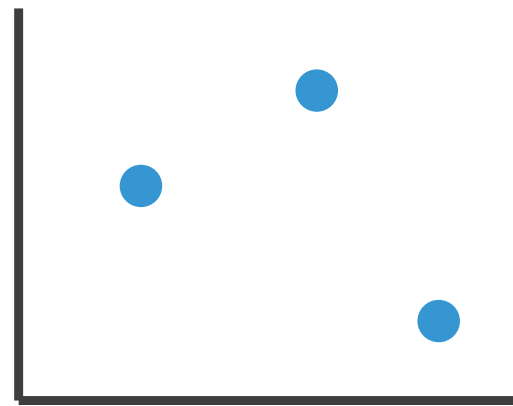
# Visual encoding

- analyze idiom structure as combination of marks and channels



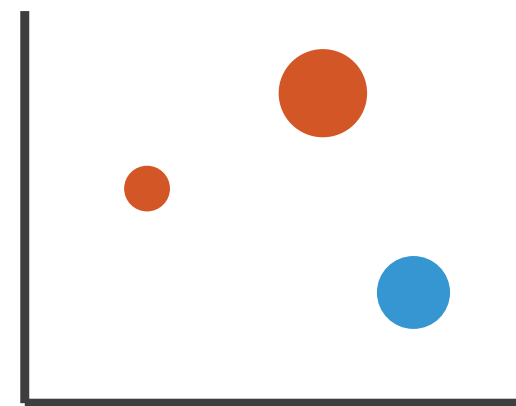
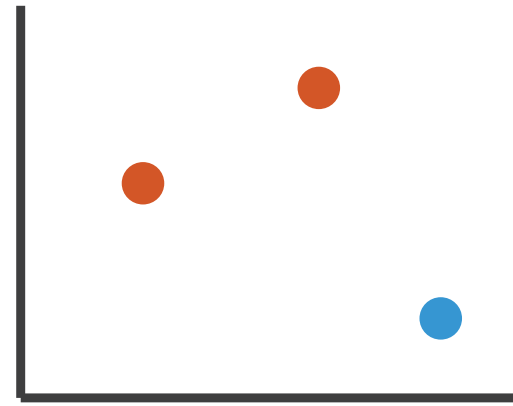
1:  
vertical position

mark: line



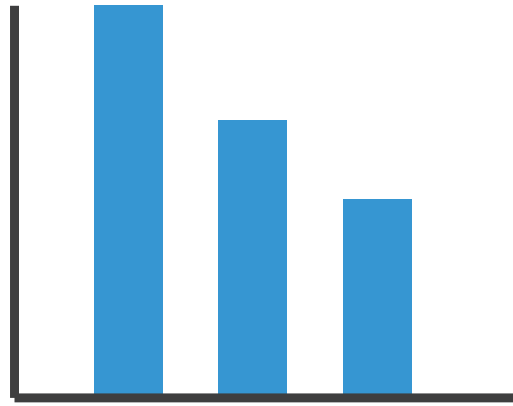
2:  
vertical position  
horizontal position

mark: point



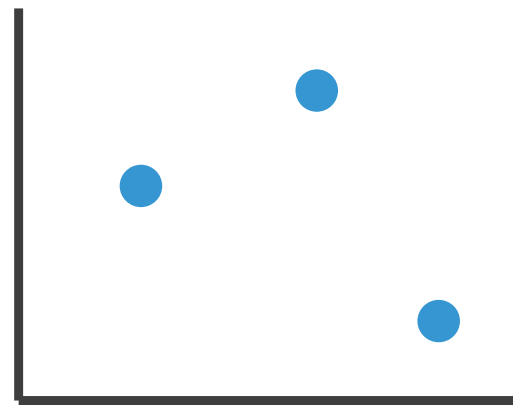
# Visual encoding

- analyze idiom structure as combination of marks and channels



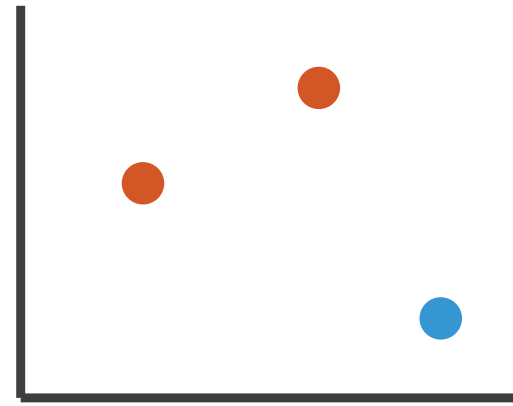
1:  
vertical position

mark: line



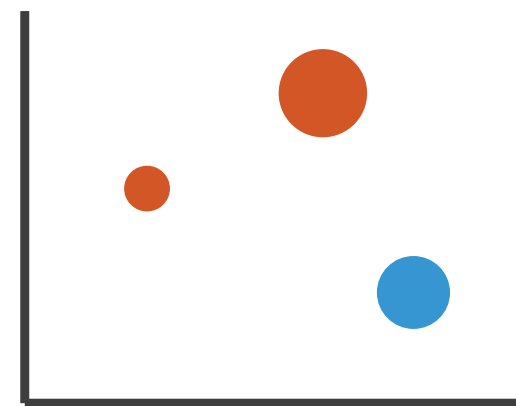
2:  
vertical position  
horizontal position

mark: point



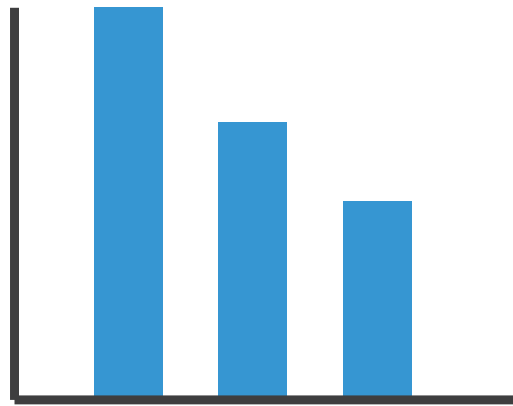
3:  
vertical position  
horizontal position  
color hue

mark: point



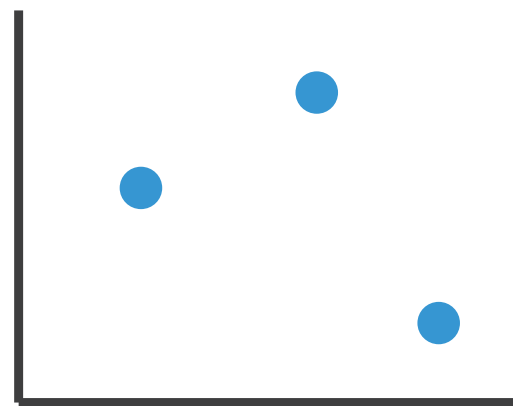
# Visual encoding

- analyze idiom structure as combination of marks and channels



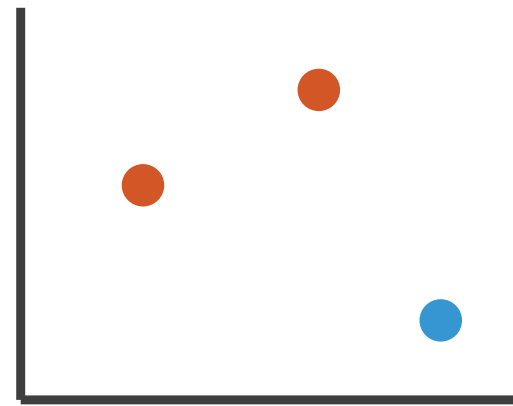
1:  
vertical position

mark: line



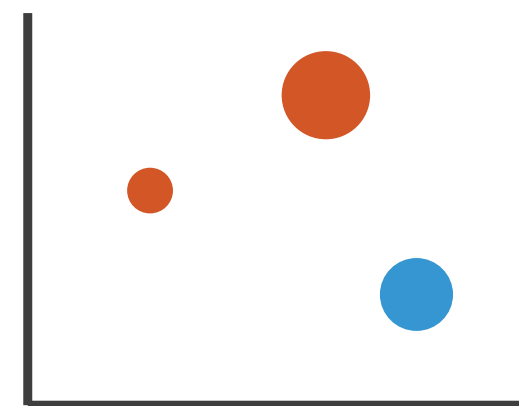
2:  
vertical position  
horizontal position

mark: point



3:  
vertical position  
horizontal position  
color hue

mark: point



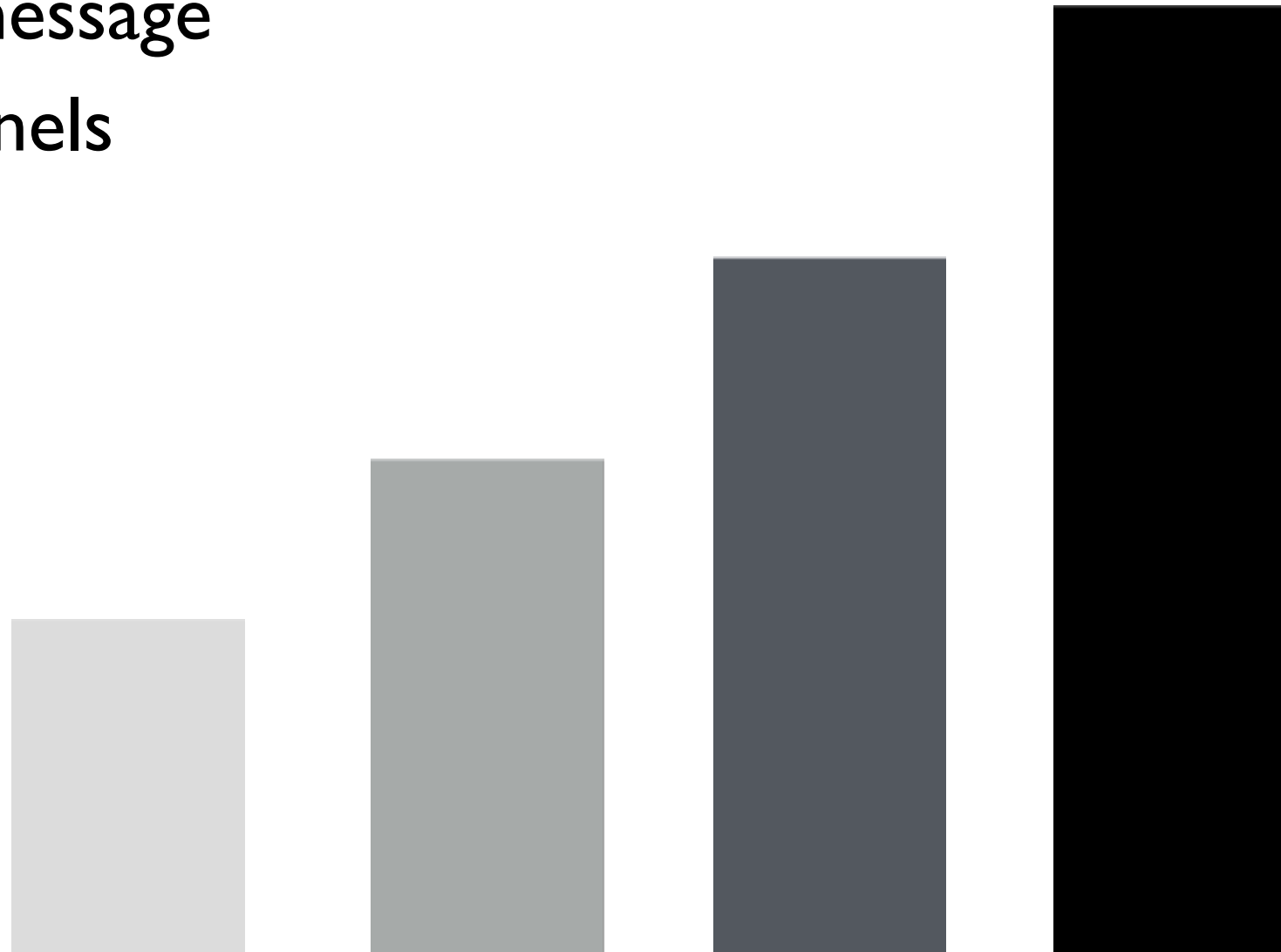
4:  
vertical position  
horizontal position  
color hue  
size (area)

mark: point



# Redundant encoding

- multiple channels
  - sends stronger message
  - but uses up channels



Length and Luminance

# Marks as constraints

- math view: geometric primitives have dimensions

→ Points

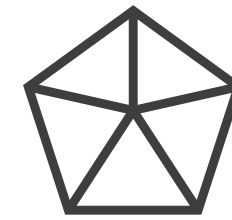
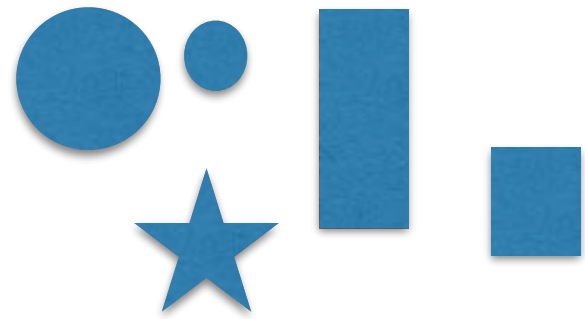
0D

→ Lines

1D

→ Interlocking Areas

2D



# Marks as constraints

- math view: geometric primitives have dimensions

→ Points

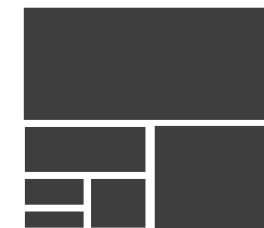
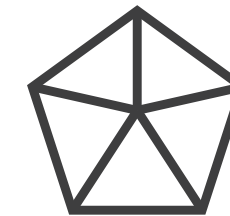
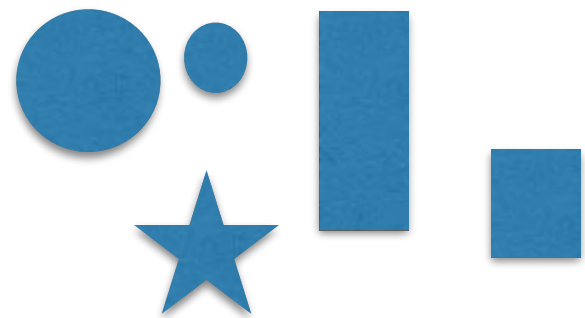
0D

→ Lines

1D

→ Interlocking Areas

2D



- constraint view: mark type constrains what else can be encoded
  - points: 0 constraints on size, can encode more attributes w/ size & shape
  - lines: 1 constraint on size (length), can still size code other way (width)
  - interlocking areas: 2 constraints on size (length/width), cannot size or shape code
    - interlocking: size, shape, position

# Marks as constraints

- math view: geometric primitives have dimensions

➔ Points

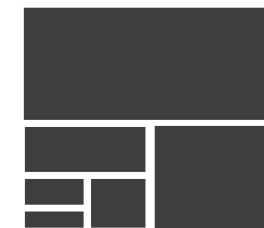
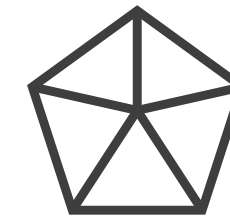
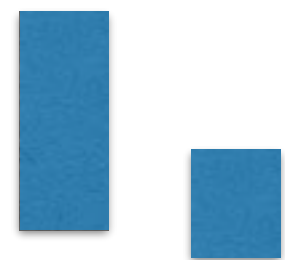
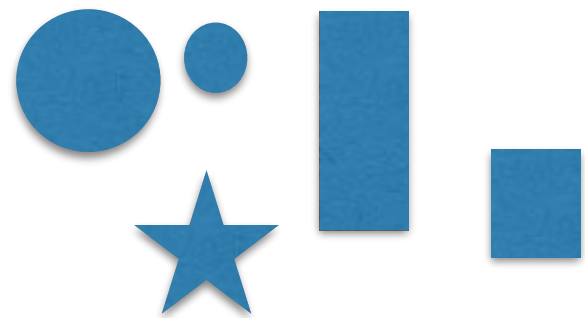
0D

➔ Lines

1D

➔ Interlocking Areas

2D



- constraint view: mark type constrains what else can be encoded
  - points: 0 constraints on size, can encode more attributes w/ size & shape
  - lines: 1 constraint on size (length), can still size code other way (width)
  - interlocking areas: 2 constraints on size (length/width), cannot size or shape code
    - interlocking: size, shape, position
- quick check: can you size-code another attribute
  - or is size/shape in use?

# Scope of analysis

- simplifying assumptions: one mark per item, single view
- later on
  - multiple views
  - multiple marks in a region (glyph)
  - some items not represented by marks (aggregation and filtering)

When to use which channel?

**expressiveness**

match channel type to data type

**effectiveness**

some channels are better than others

# Channels: Rankings

Position on common scale



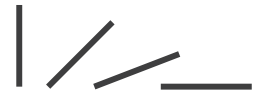
Position on unaligned scale



Length (1D size)



Tilt/angle



Area (2D size)



Depth (3D position)



Color luminance



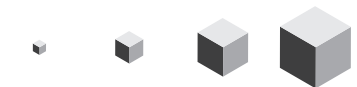
Color saturation



Curvature



Volume (3D size)



Same

Spatial region



Color hue



Motion



Shape



# Channels: Rankings

## ➔ Magnitude Channels: Ordered Attributes

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**
  - match channel and data characteristics




# Channels: 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  
Same

## ➔ Identity Channels: Categorical Attributes

Spatial region 

Color hue 

Motion 

Shape 

### ➔ Attribute Types

➔ Categorical



➔ Ordered

➔ Ordinal



➔ Quantitative



- **expressiveness**

- match channel and data characteristics

- magnitude for ordered

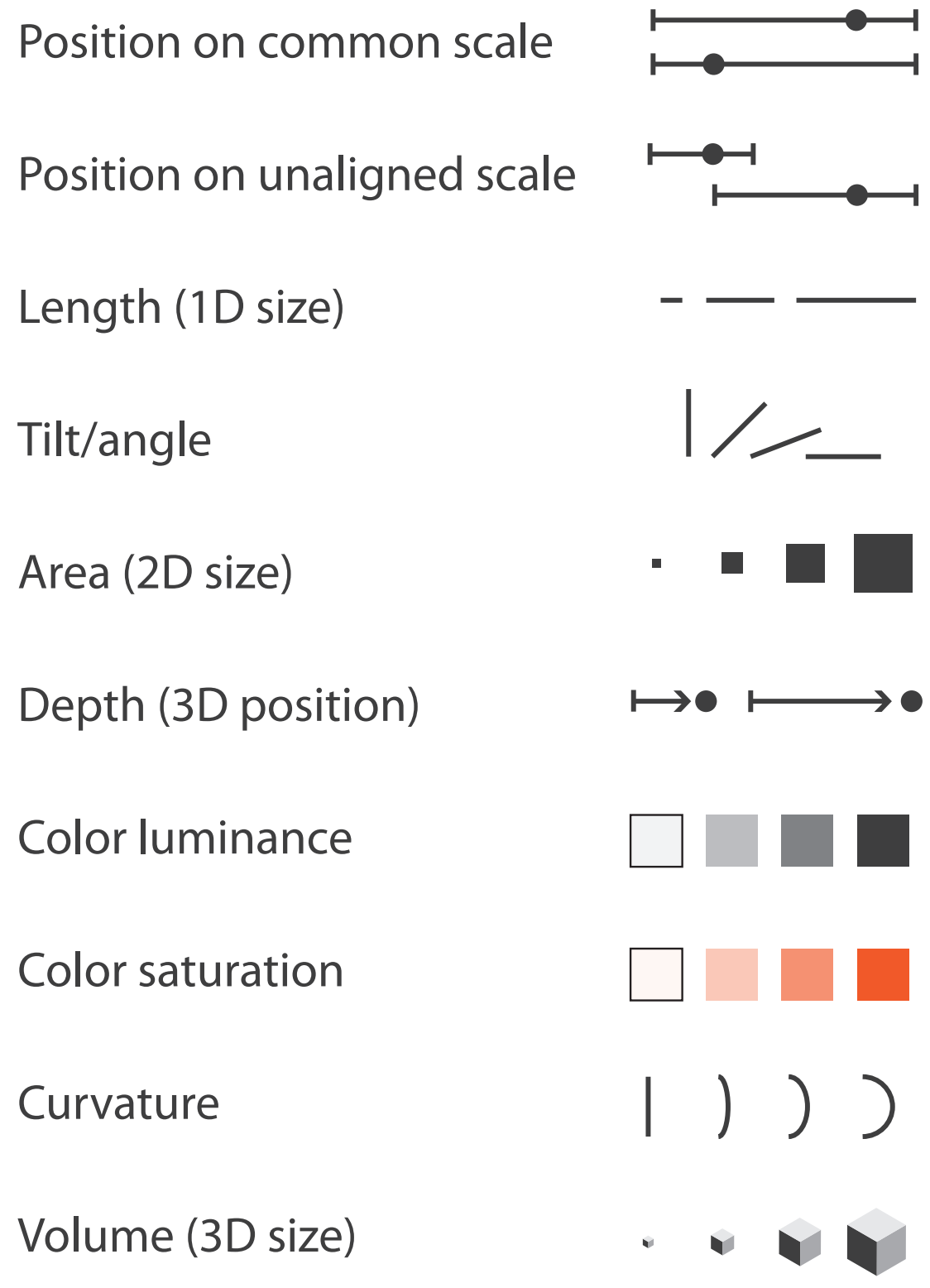
- how much? which rank?

- identity for categorical

- what?

# Channels: Rankings

## ➔ Magnitude Channels: Ordered Attributes



## ➔ Identity Channels: Categorical Attributes



Best

Effectiveness

Least

Same

Same

- **expressiveness**
  - match channel and data characteristics
- **effectiveness**
  - channels differ in accuracy of perception

# Channels: Rankings

## ➔ Magnitude Channels: Ordered Attributes

Position on common scale



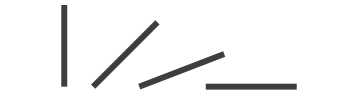
Position on unaligned scale



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**
  - match channel and data characteristics
- **effectiveness**
  - channels differ in accuracy of perception
  - spatial position ranks high for both

# Grouping

- containment
- connection

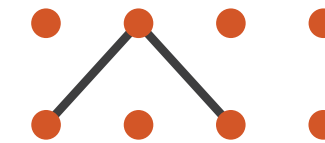
- proximity
  - same spatial region
- similarity
  - same values as other categorical channels

## Marks as Links

### ➔ Containment



### ➔ Connection



### ➔ Identity Channels: Categorical Attributes

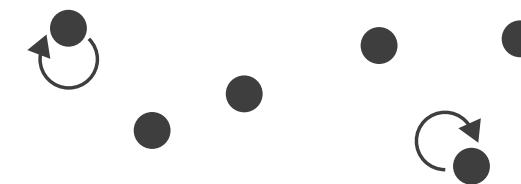
Spatial region



Color hue



Motion



Shape



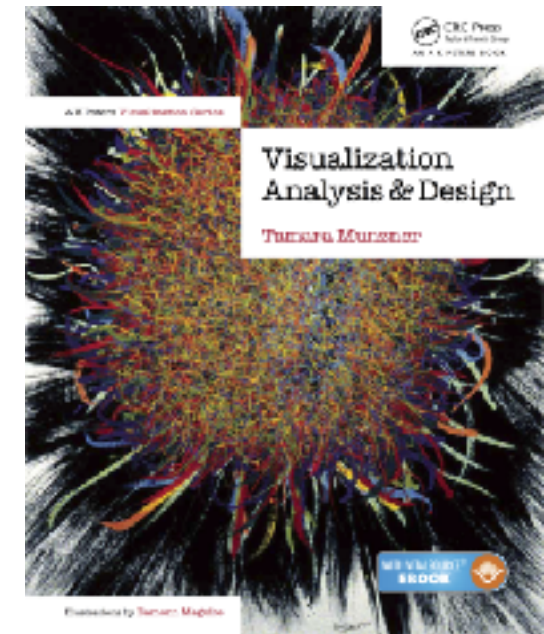
# Visualization Analysis & Design

## *Marks & Channels (Ch 5) II*

**Tamara Munzner**

Department of Computer Science  
University of British Columbia

[@tamaramunzner](#)



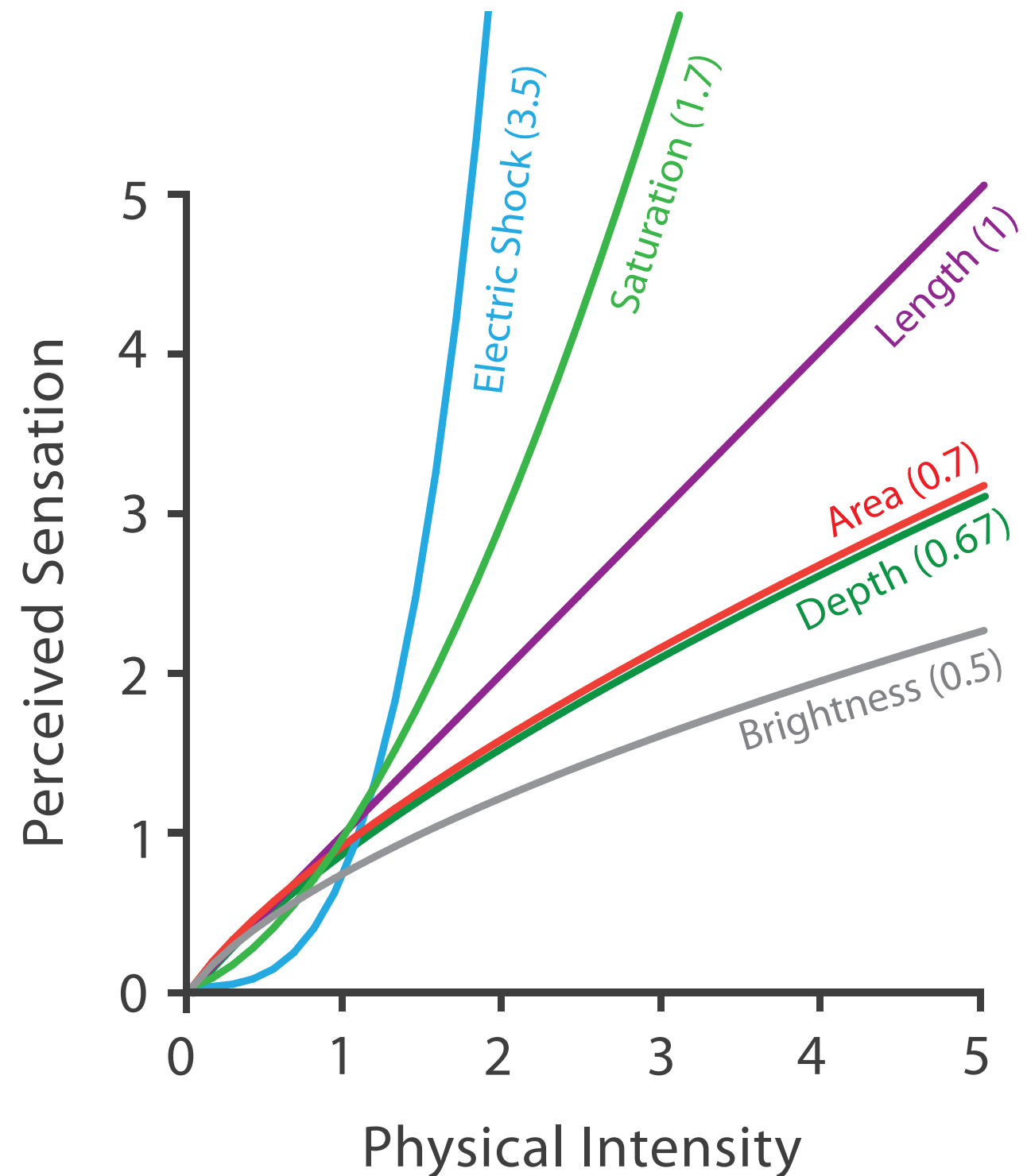
# Channel effectiveness

- accuracy: how precisely can we tell the difference between encoded items?
- discriminability: how many unique steps can we perceive?
- separability: is our ability to use this channel affected by another one?
- popout: can things jump out using this channel?

# Accuracy: Fundamental theory

- length is accurate: linear
- others magnified or compressed
  - exponent characterizes

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

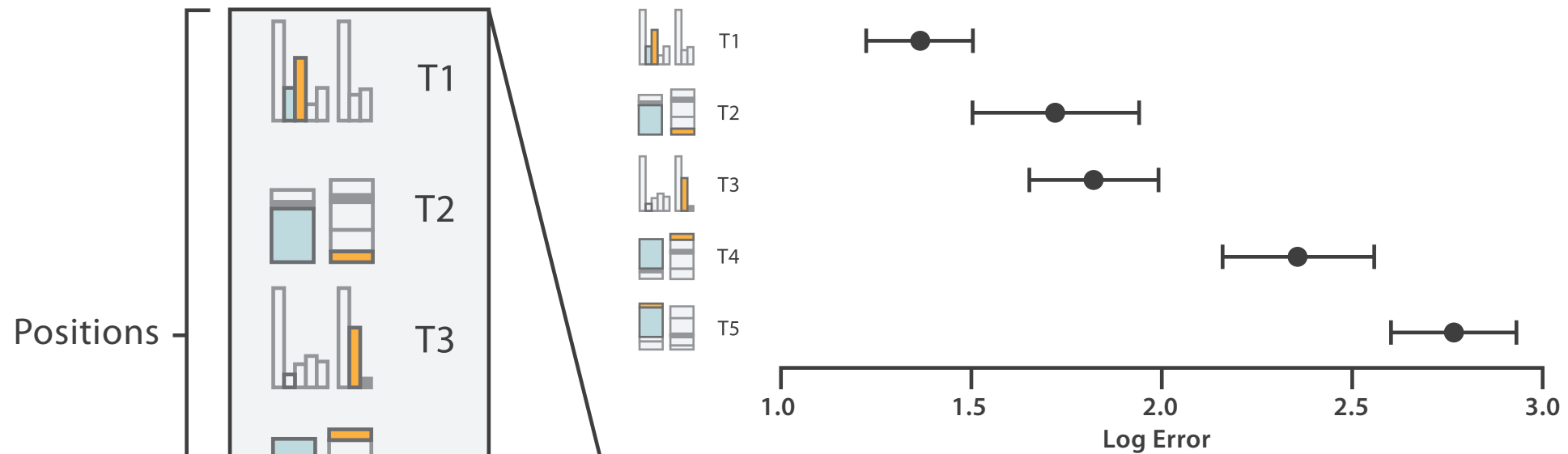


$S$  = sensation

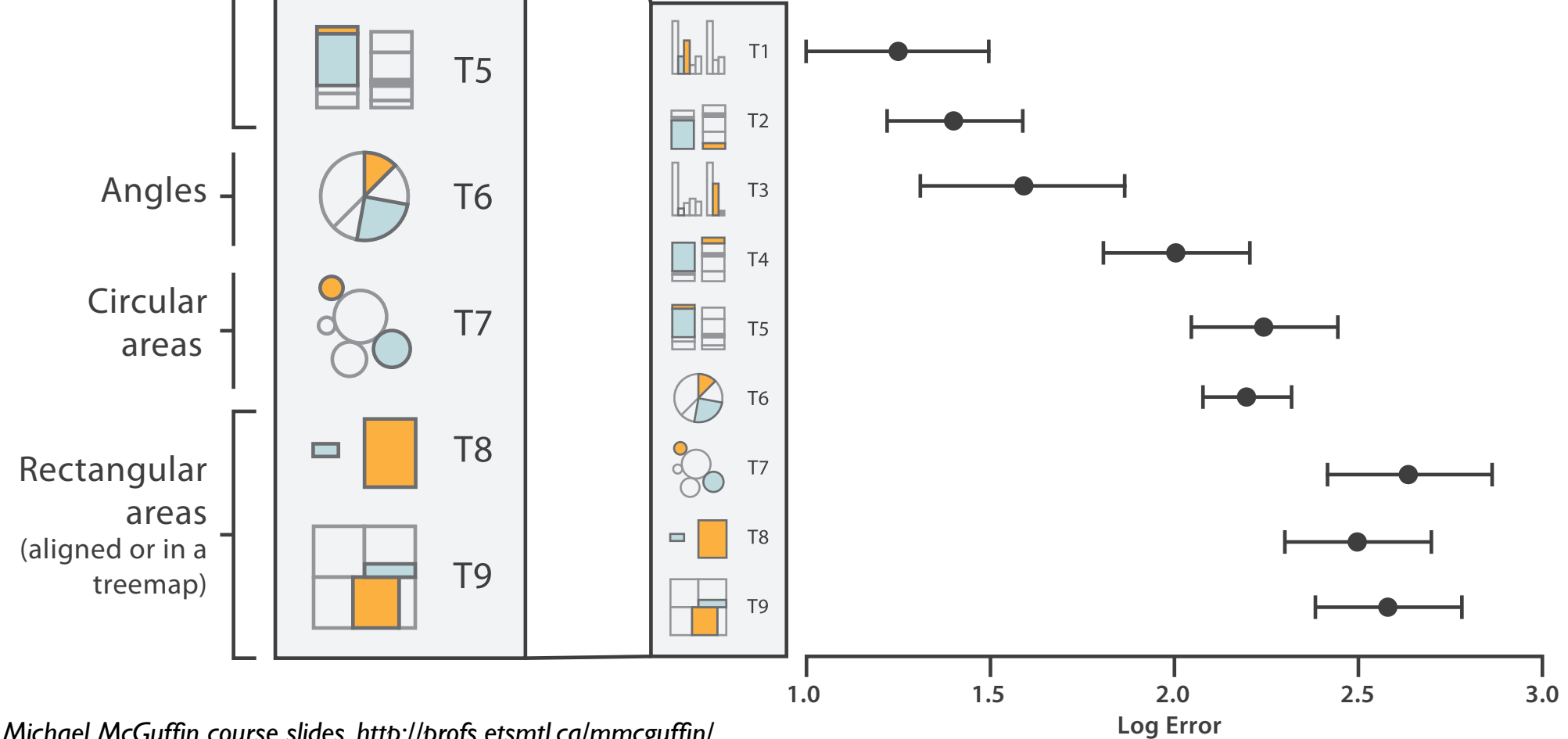
$I$  = intensity

# Accuracy: Vis experiments

Cleveland & McGill's Results



Crowdsourced Results

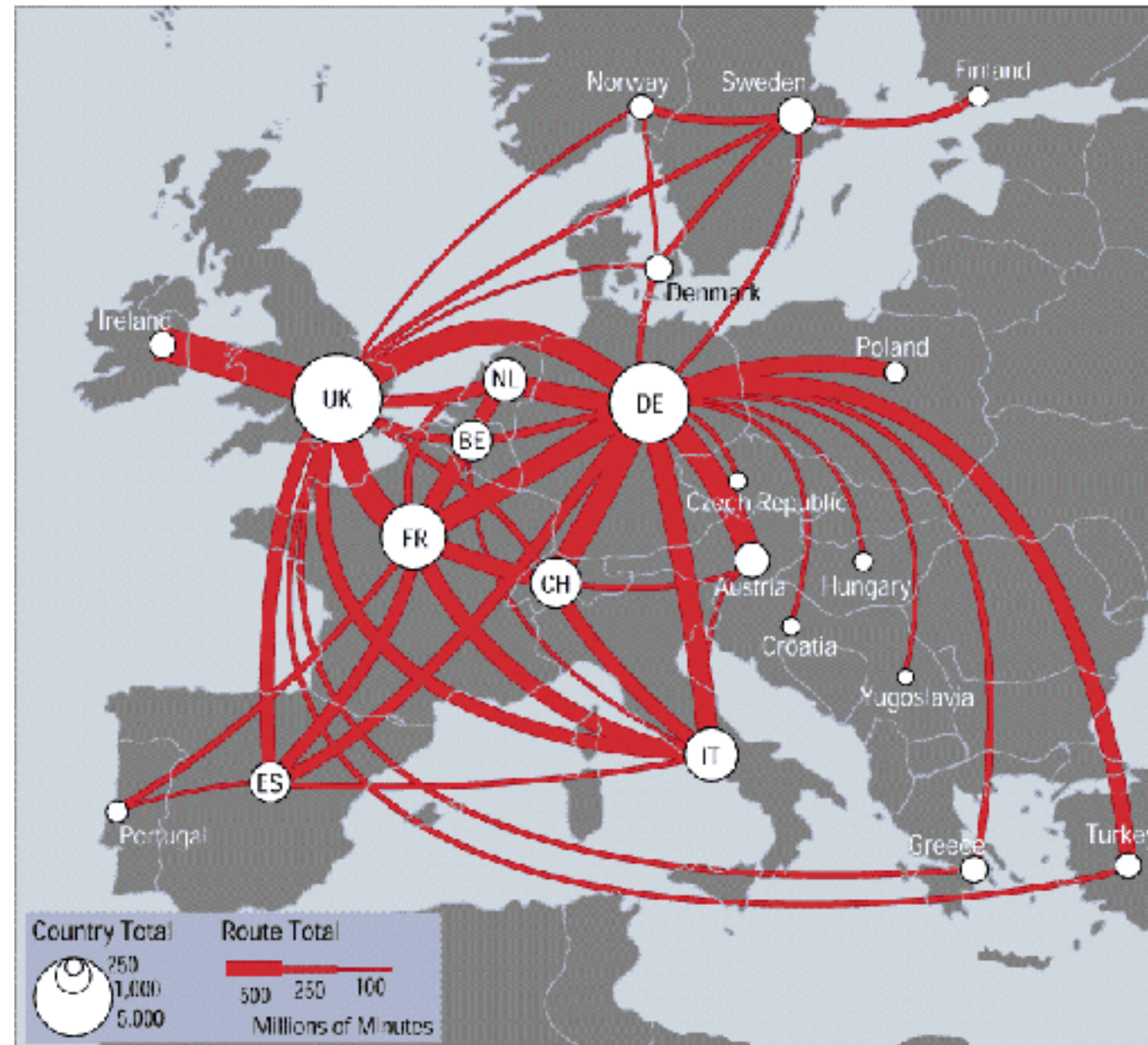


*[Crowdsourcing Graphical Perception: Using Mechanical Turk to Assess Visualization Design. Heer and Bostock. Proc ACM Conf. Human Factors in Computing Systems (CHI) 2010, p. 203–212.]*



# Discriminability: How many usable steps?

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

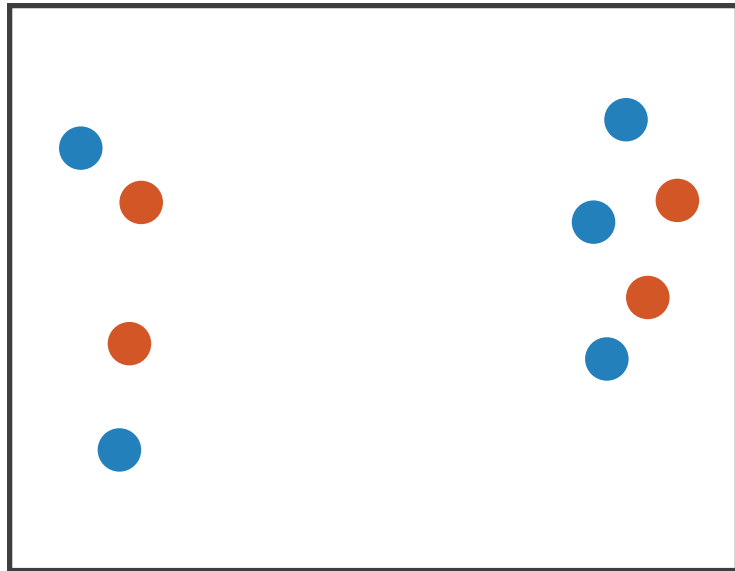


[[mappa.mundi.net/maps/maps\\_014/telegeography.html](http://mappa.mundi.net/maps/maps_014/telegeography.html)]

# Separability vs. Integrality

Position

+ Hue (Color)

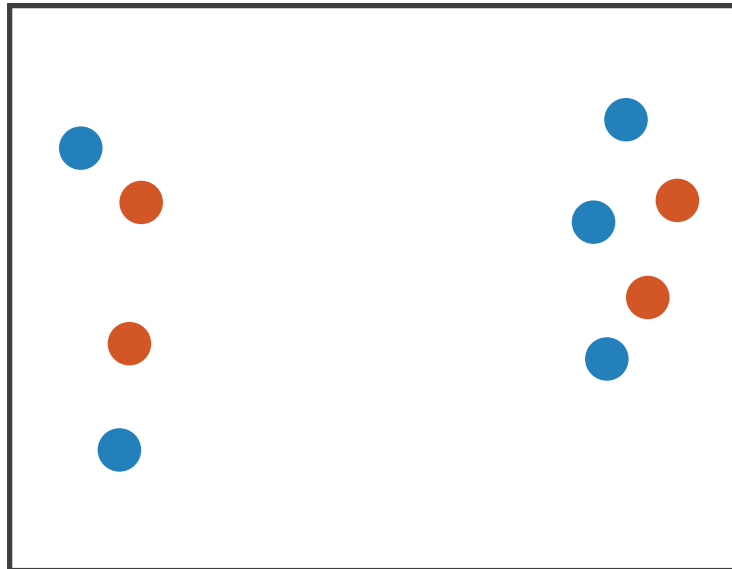


Fully separable

2 groups each

# Separability vs. Integrality

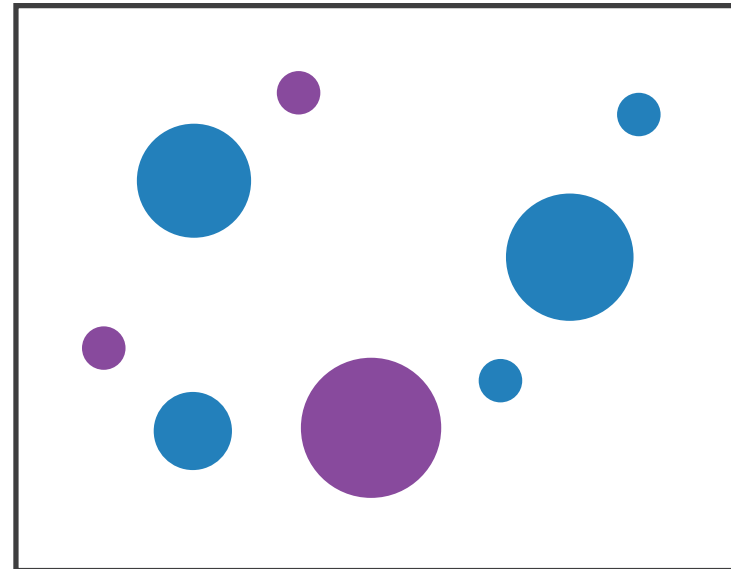
Position  
+ Hue (Color)



Fully separable

2 groups each

Size  
+ Hue (Color)

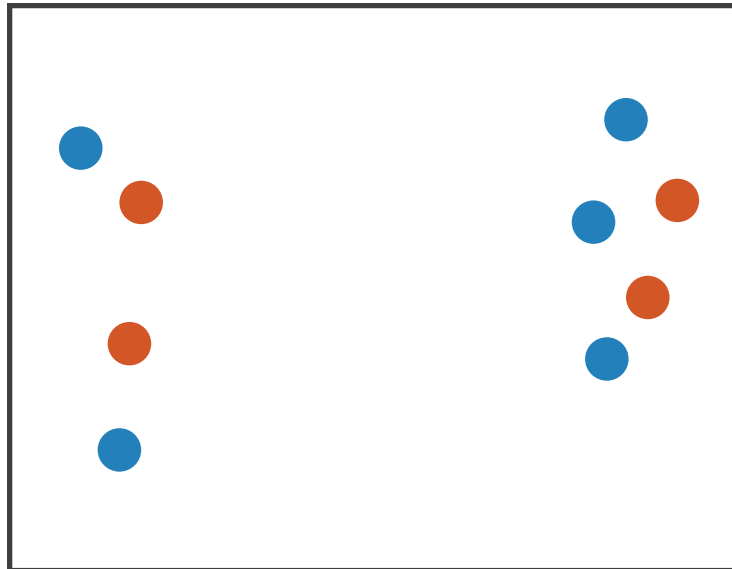


Some interference

2 groups each

# Separability vs. Integrality

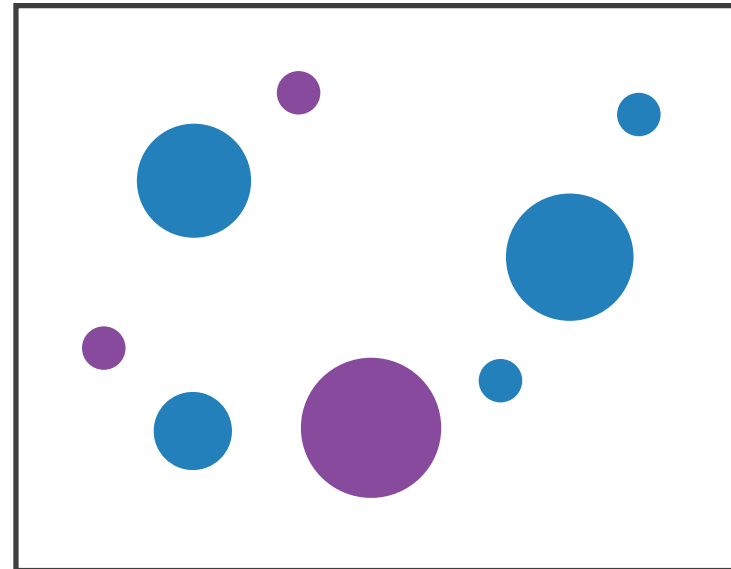
Position  
+ Hue (Color)



Fully separable

2 groups each

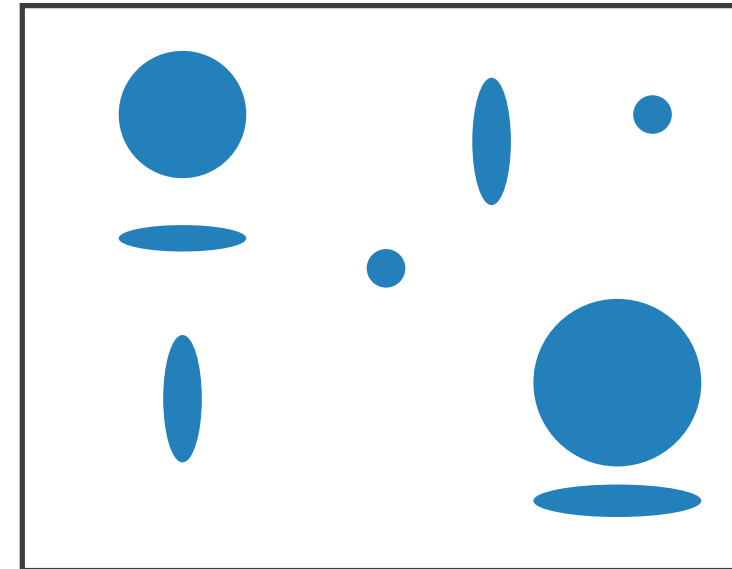
Size  
+ Hue (Color)



Some interference

2 groups each

Width  
+ Height

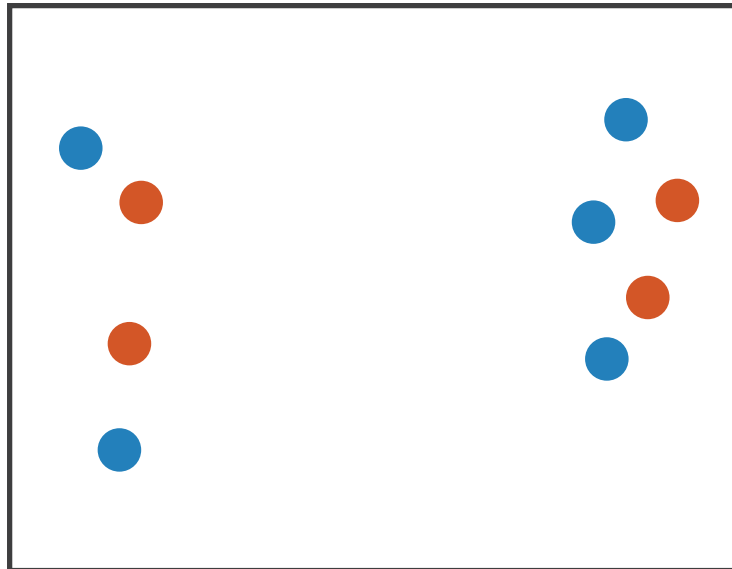


Some/significant  
interference

3 groups total:  
integral area

# Separability vs. Integrality

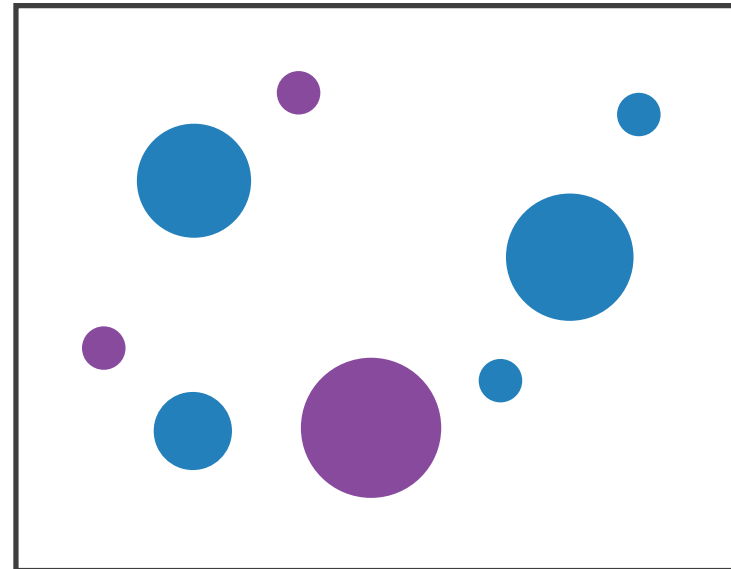
Position  
+ Hue (Color)



Fully separable

2 groups each

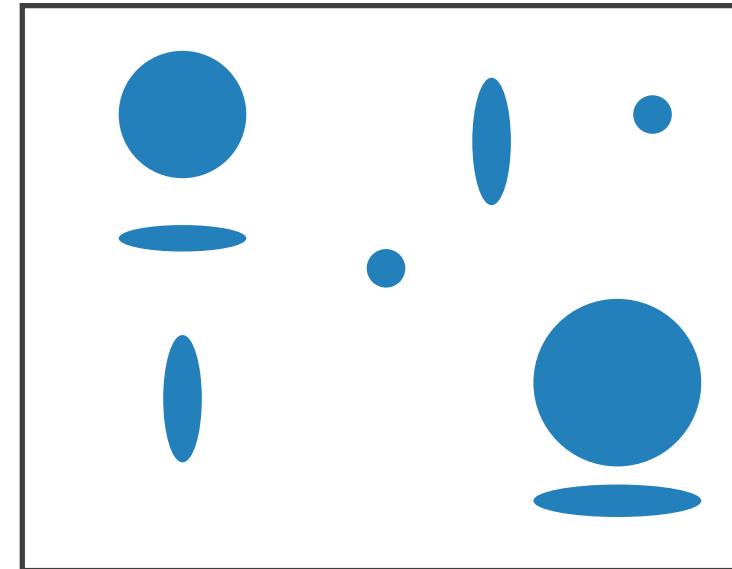
Size  
+ Hue (Color)



Some interference

2 groups each

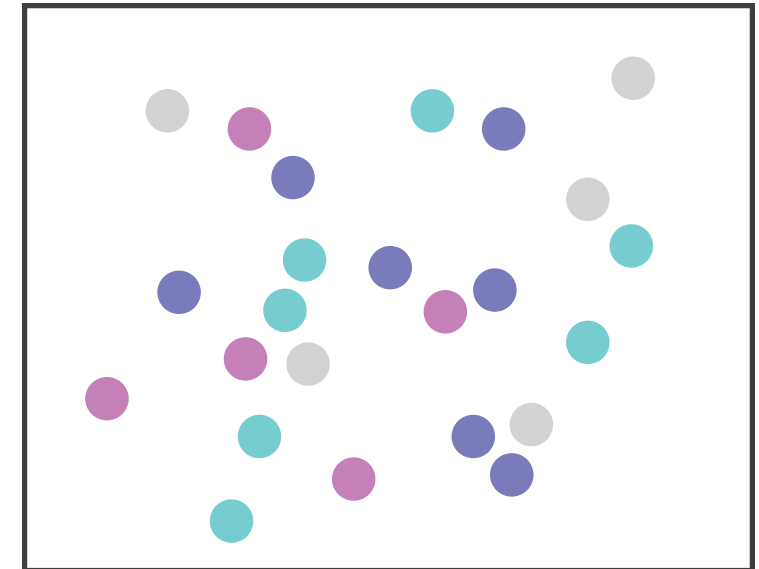
Width  
+ Height



Some/significant  
interference

3 groups total:  
integral area

Red  
+ Green



Major interference

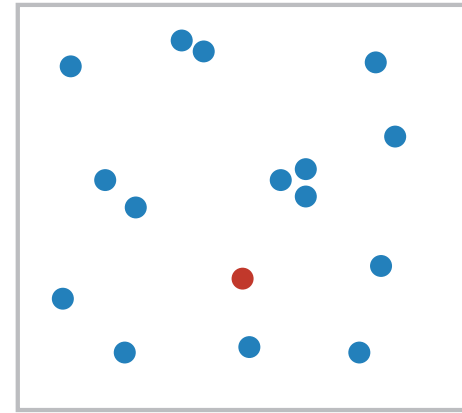
4 groups total:  
integral hue

# Popout

- find the red dot
  - how long does it take?

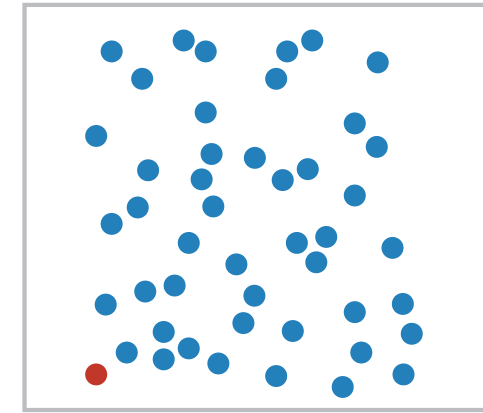
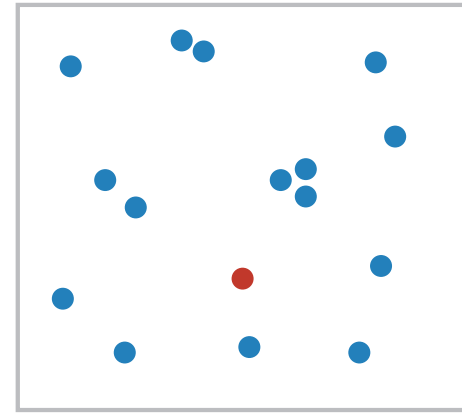
# Popout

- find the red dot
  - how long does it take?



# Popout

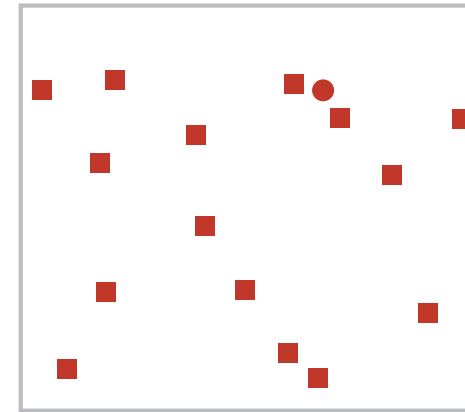
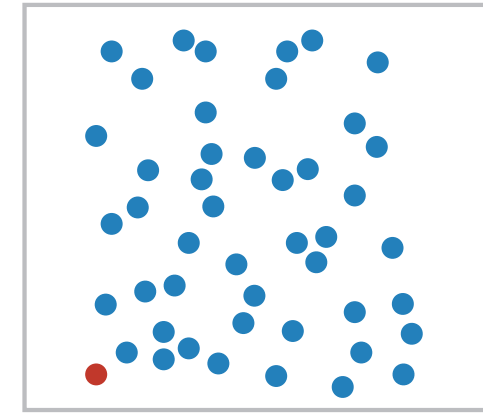
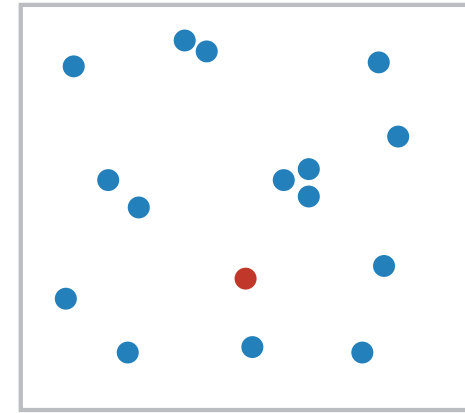
- find the red dot
  - how long does it take?





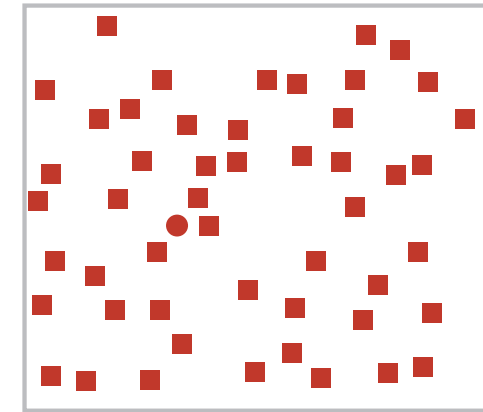
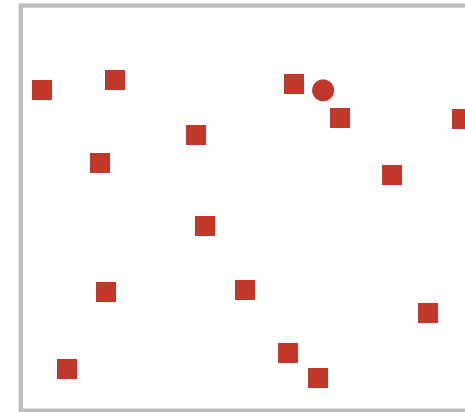
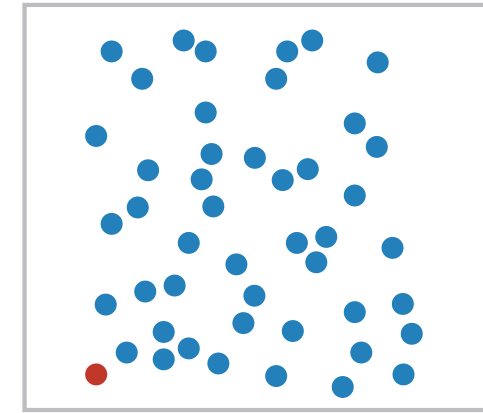
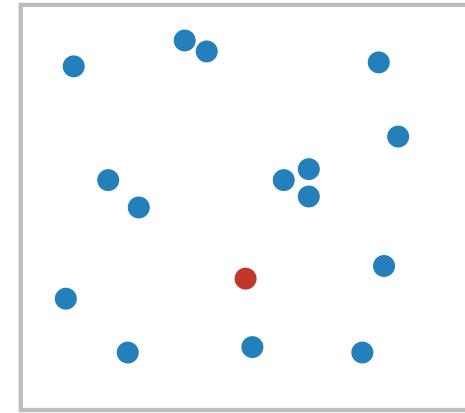
# Popout

- find the red dot
  - how long does it take?



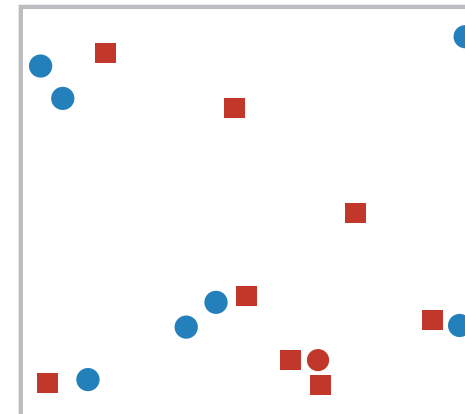
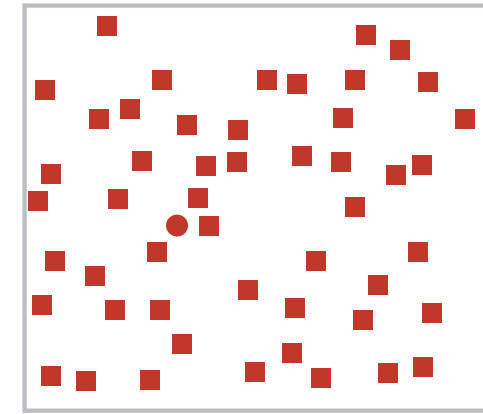
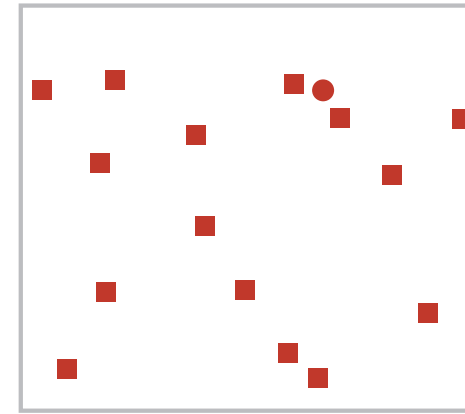
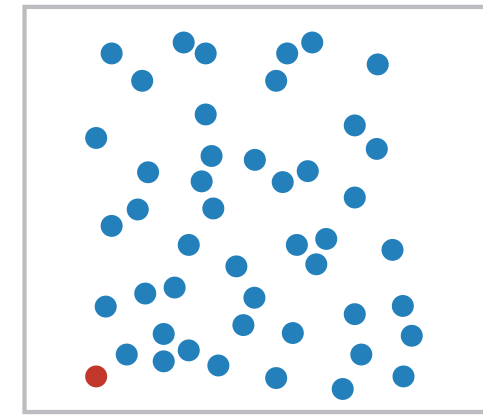
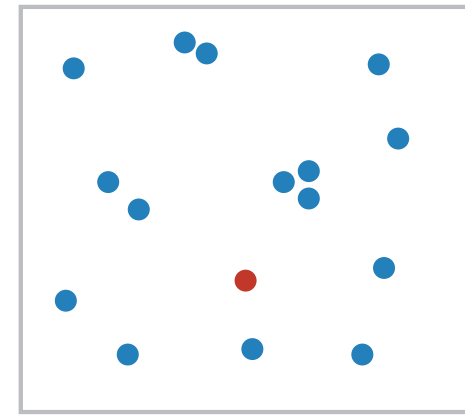
# Popout

- find the red dot
  - how long does it take?



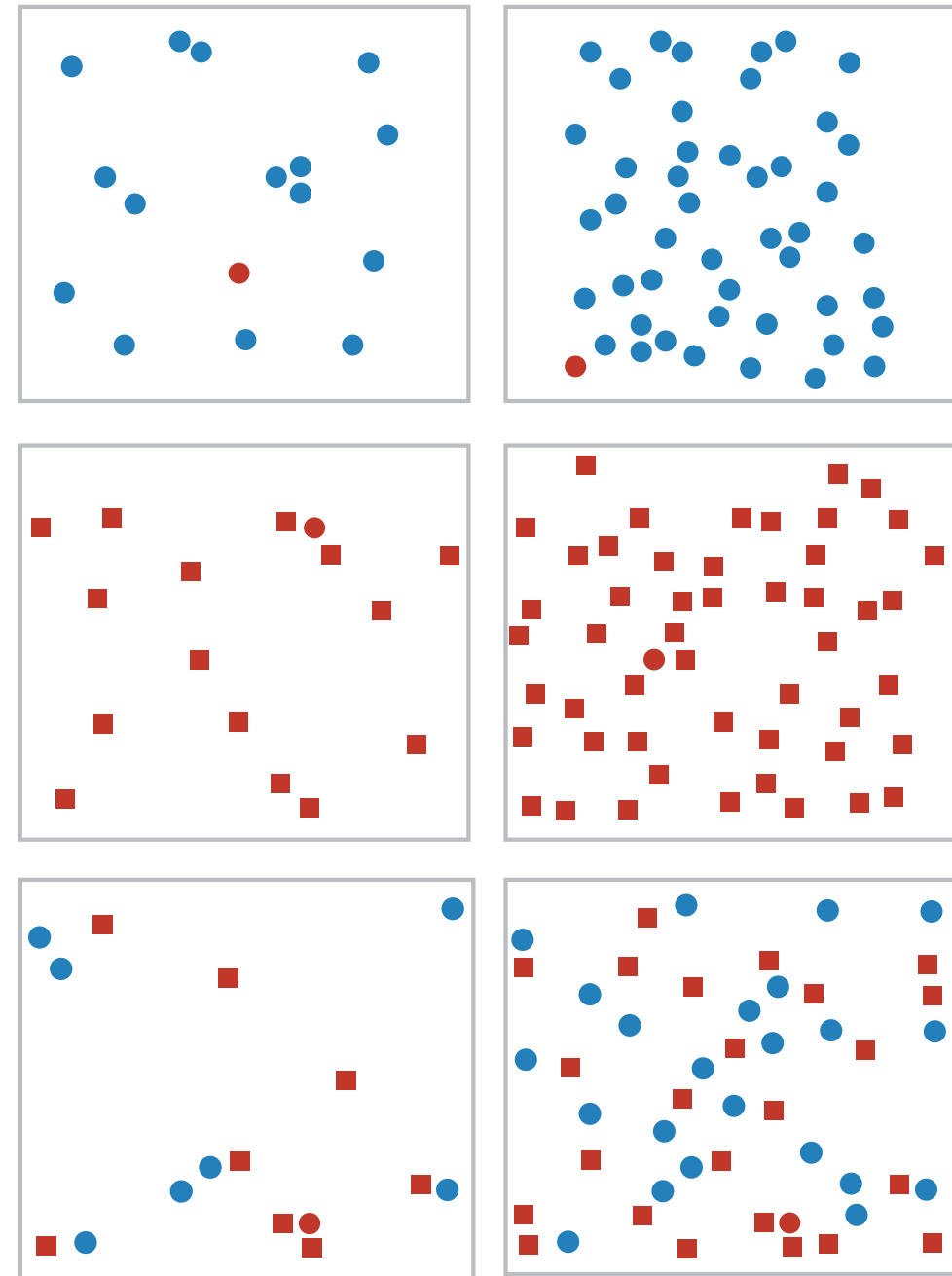
# Popout

- find the red dot
  - how long does it take?



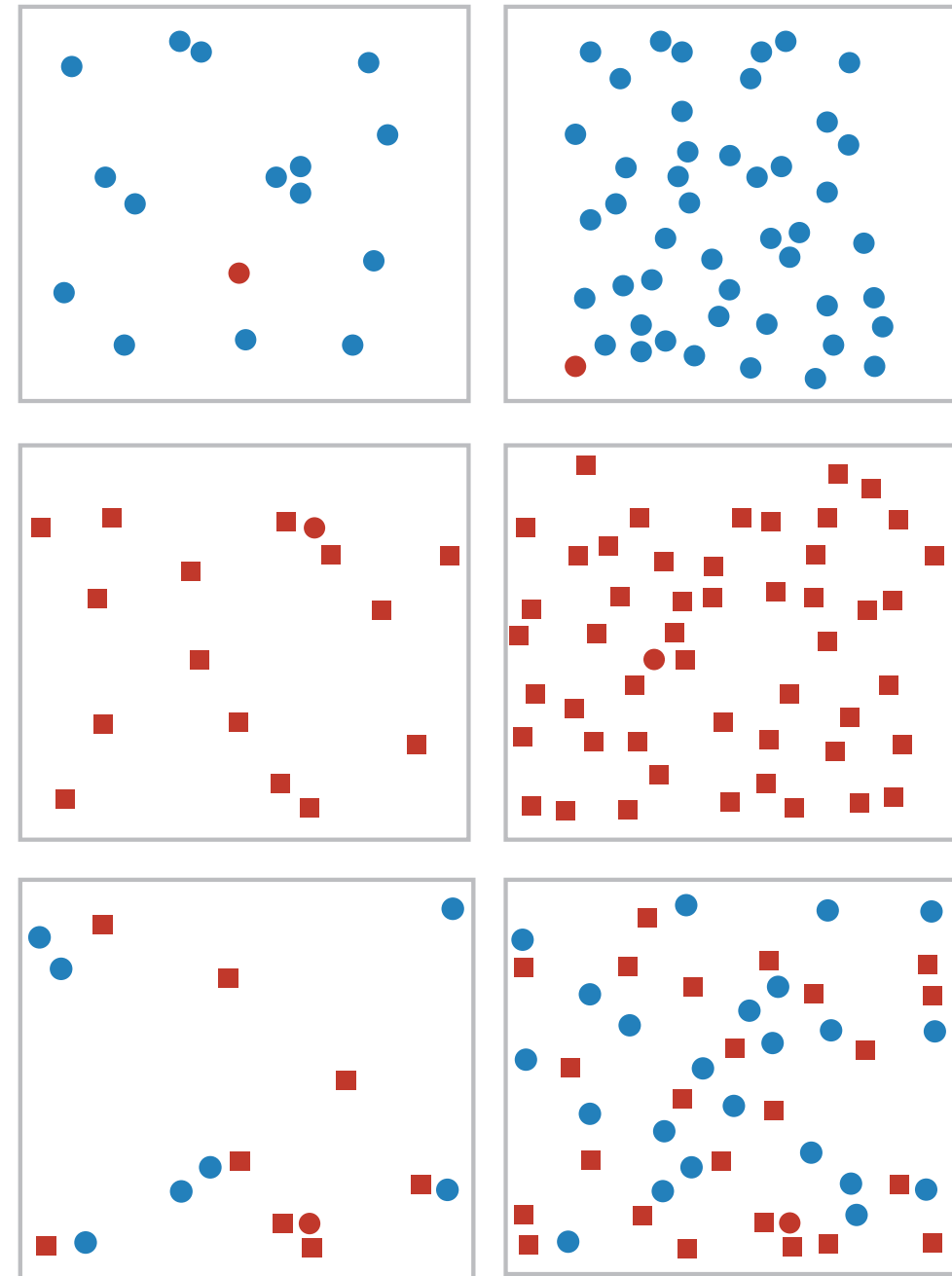
# Popout

- find the red dot
  - how long does it take?

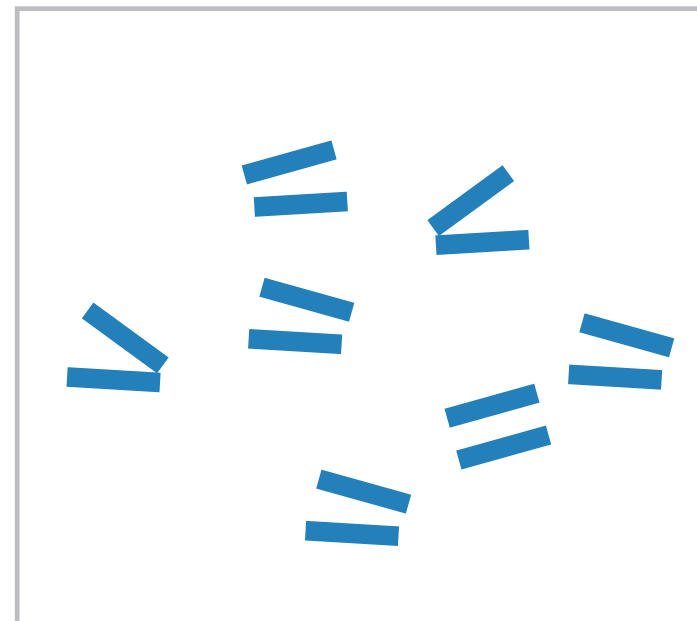
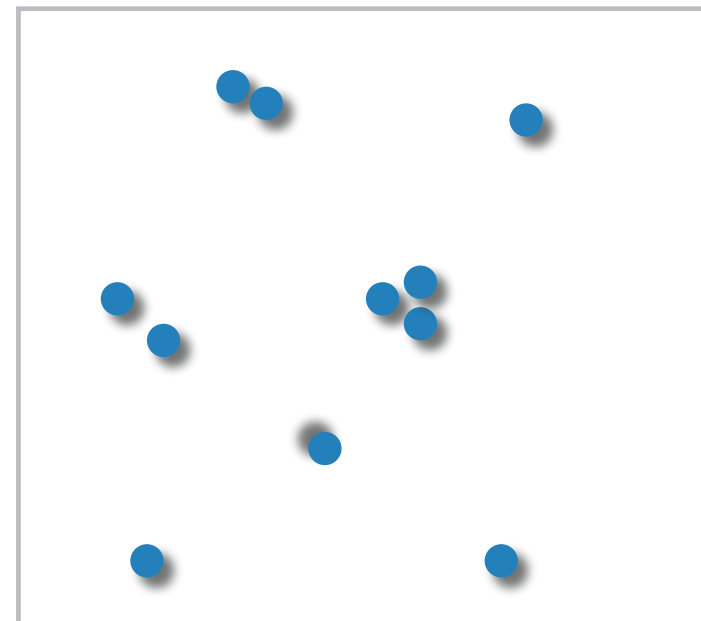
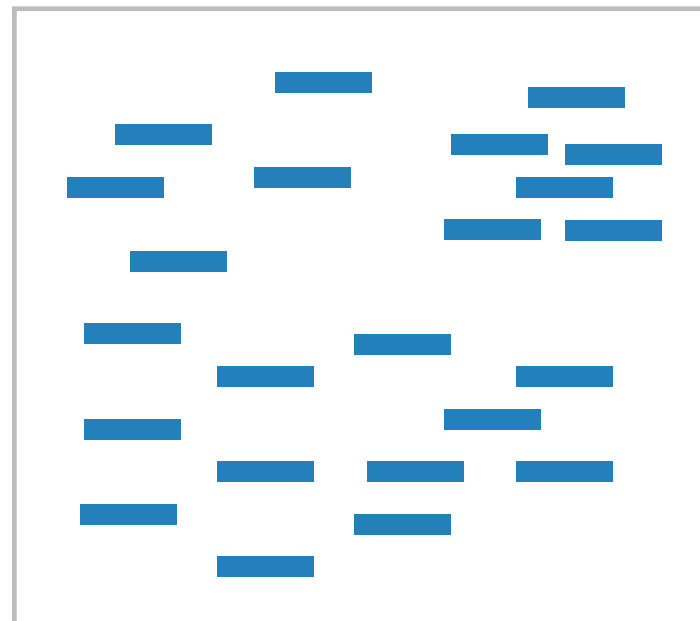
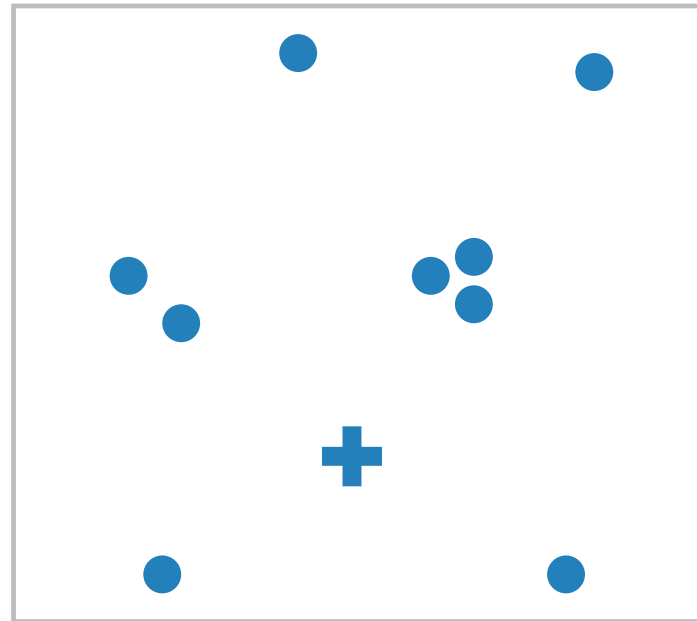
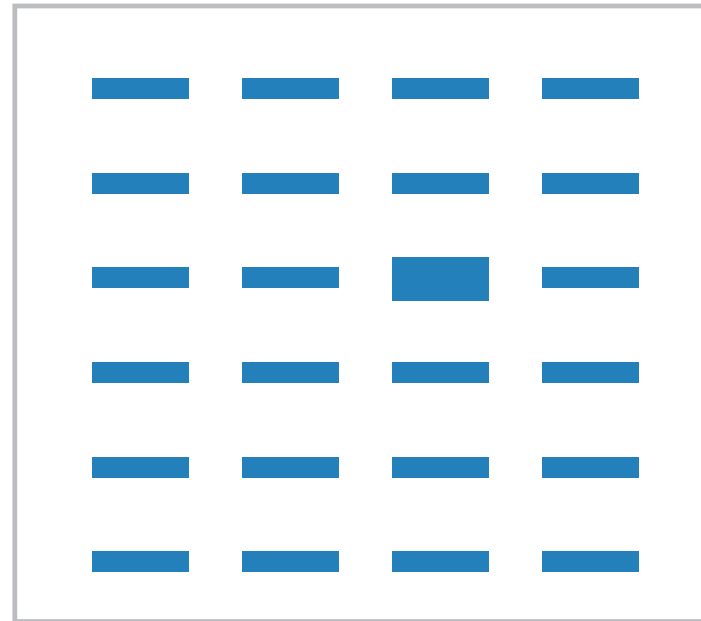
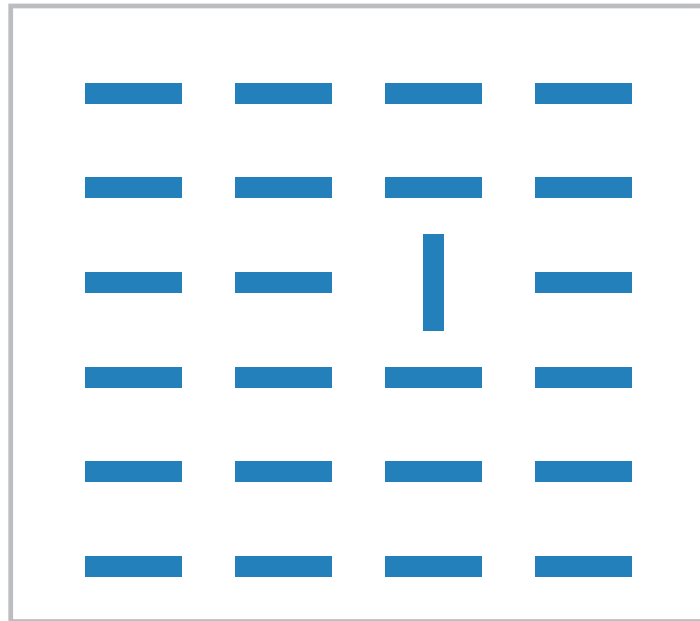


# Popout

- 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

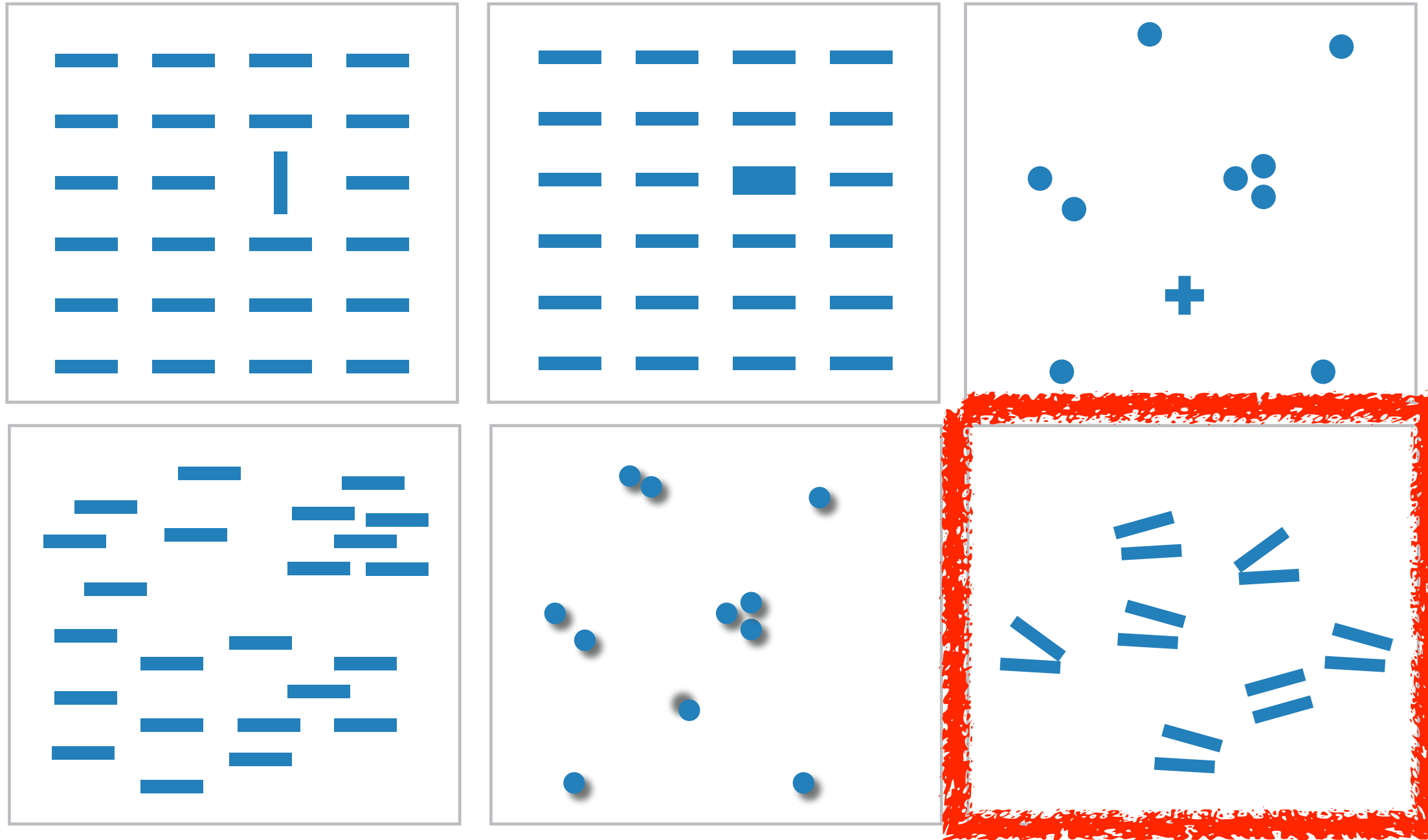


# Popout



- many channels
  - tilt, size, shape, proximity, shadow direction, ...

# Popout



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

# Factors affecting accuracy

- alignment
- distractors
- distance
- common scale / alignment





# Relative vs. absolute judgements

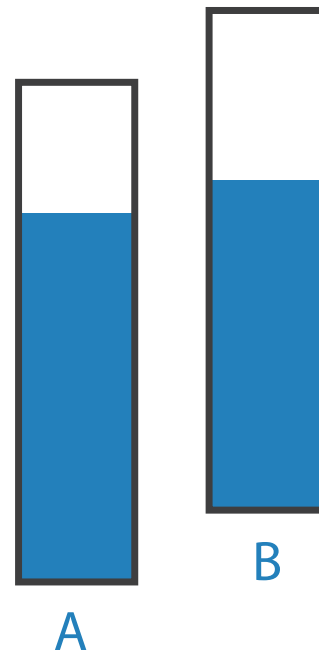
- perceptual system mostly operates with relative judgements, not absolute

# Relative vs. absolute judgements

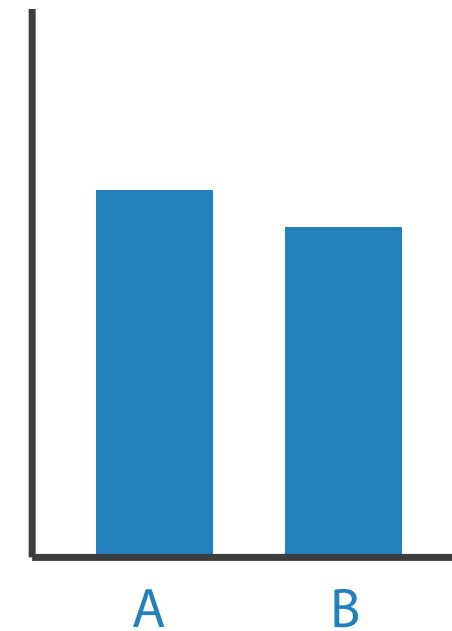
- perceptual system mostly operates with relative judgements, not absolute
  - that's why accuracy increases with common frame/scale and alignment



length



position along  
unaligned  
common scale



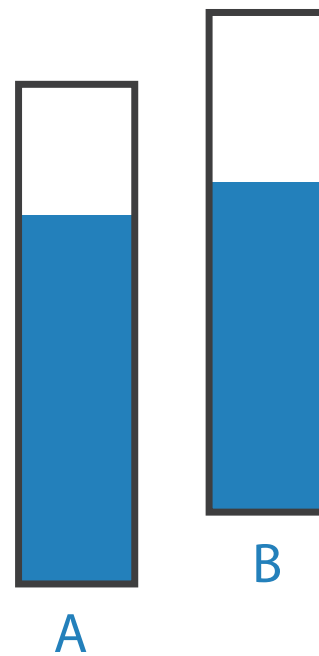
position along  
aligned scale

# Relative vs. absolute judgements

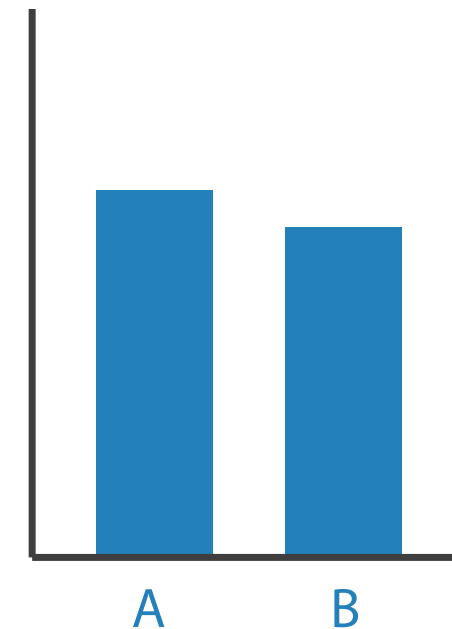
- perceptual system mostly operates with relative judgements, not absolute
  - that's why accuracy increases with common frame/scale and alignment
  - Weber's Law: ratio of increment to background is constant



length



position along  
unaligned  
common scale



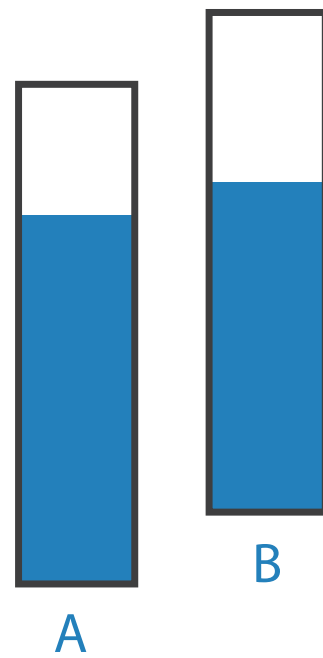
position along  
aligned scale

# Relative vs. absolute judgements

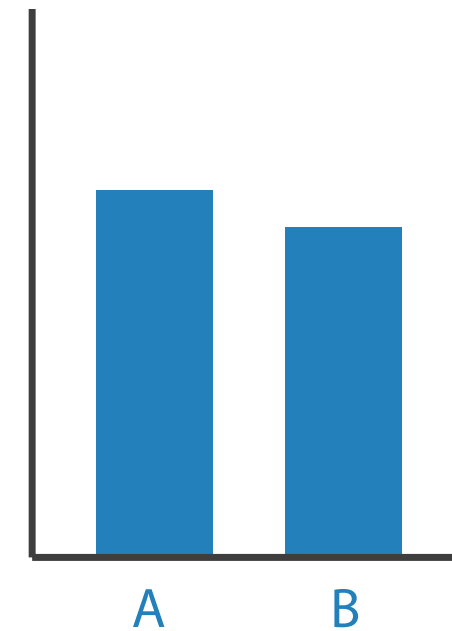
- perceptual system mostly operates with relative judgements, not absolute
  - that's why accuracy increases with common frame/scale and alignment
  - Weber's Law: ratio of increment to background is constant
    - filled rectangles differ in length by 1:9, difficult judgement
    - white rectangles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



length



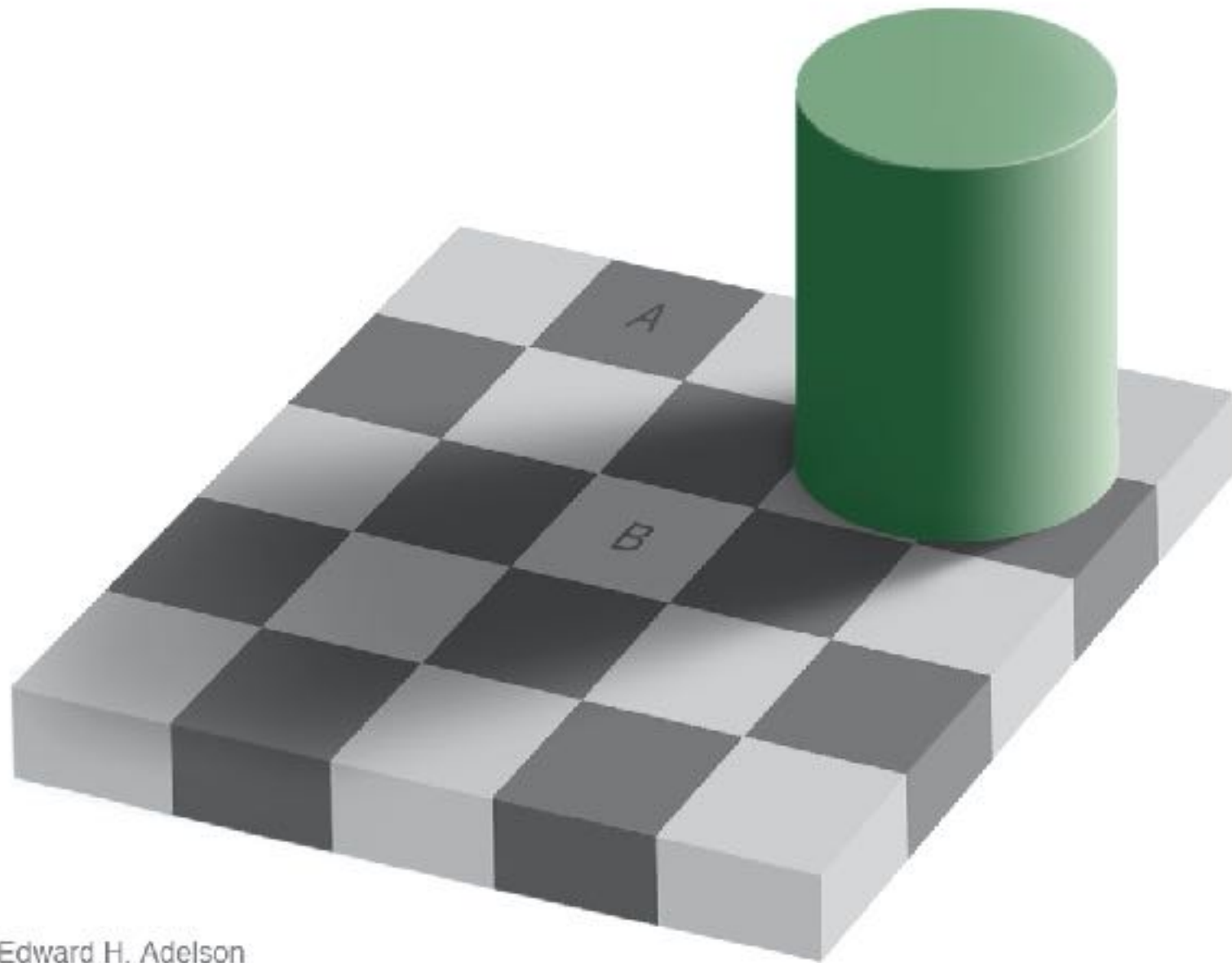
position along  
unaligned  
common scale



position along  
aligned scale

# Relative luminance judgements

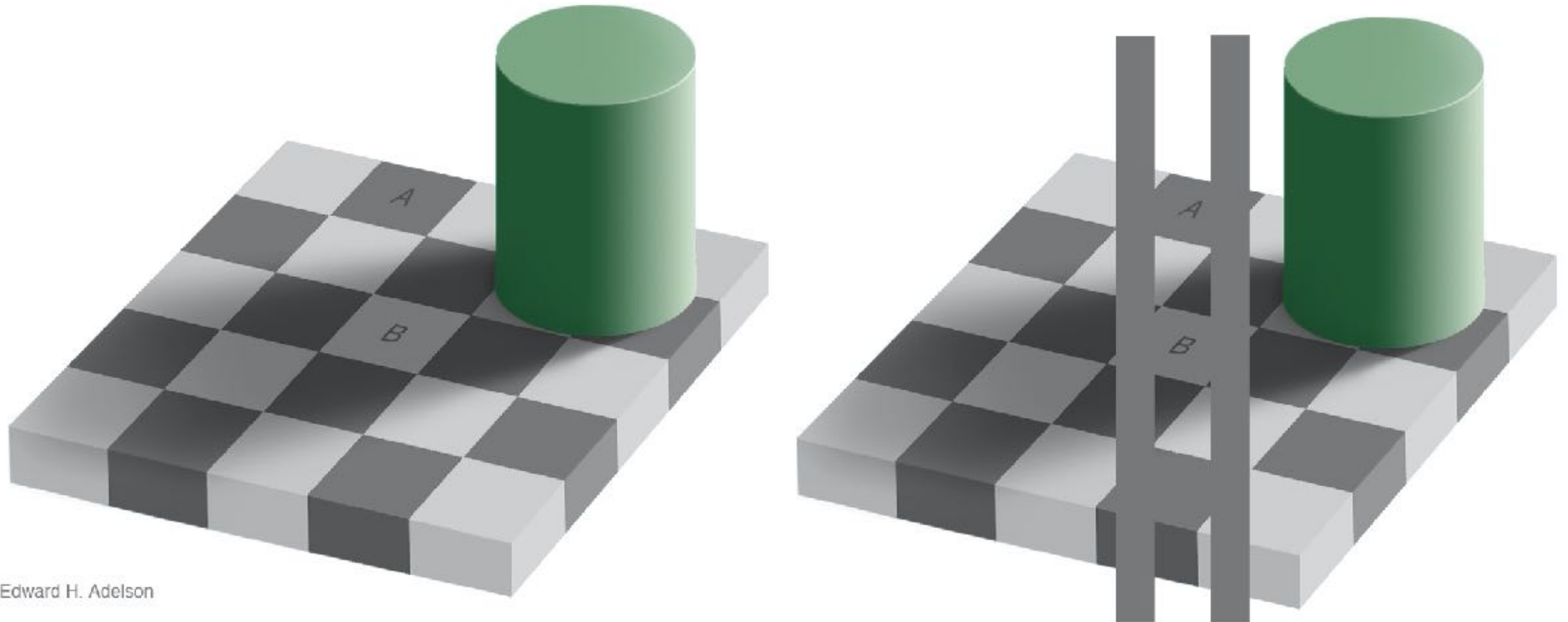
- perception of luminance is contextual based on contrast with surroundings



Edward H. Adelson

# Relative luminance judgements

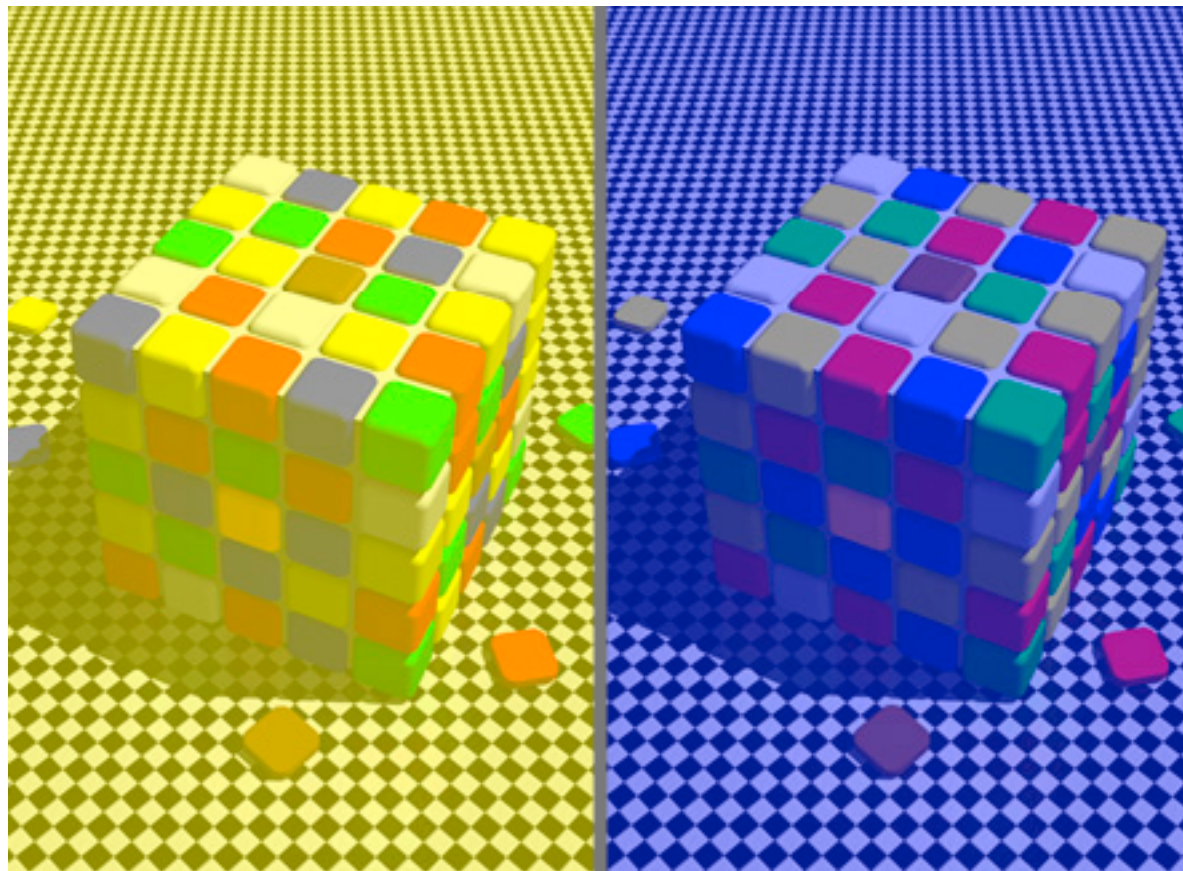
- perception of luminance is contextual based on contrast with surroundings



Edward H. Adelson

# Relative color judgements

- color constancy across broad range of illumination conditions





# Relative color judgements

- color constancy across broad range of illumination conditions

