

Computation and Visualization for Analytics

Spring 2021

Week 4.2

Visual Encoding

- Marks and Channels (*representation*)
- Channel Expressiveness and Effectiveness (*presentation*)
- Relative vs. Absolute Judgements (*presentation*)

Why Visualize?

- Explore
- Explain
 - > Memory limits
 - > Visual limits
 - > Choice limits

Explanatory Visualization

- Representation

Involves making decisions about how you are going to portray your data visually so that the subject understanding it offers can be made accessible to your audience. It is all about charts and the act of selecting the right chart to show the features of your data that you think are most relevant. (*Andy Kirk*)

- Presentation

Involves effective presentation to audience taking into consideration limits of visual perception

Representation (Marks)

➔ Points



➔ Lines



➔ Areas



Representation (Marks-points)

Age
10
20
30
40



Representation (Marks-**lines**)

State	
MA	_____
NH	_____
NY	_____
GA	_____

Representation (Marks-area, shapes)

State
MA
NH
NY
GA



Representation (Channels)

Position

Shape

Tilt

Color

Size

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)

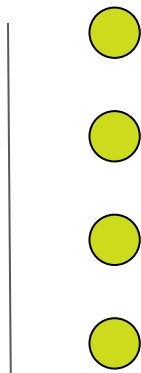


Same

Same

Representation (Mark-point, Channel-**position**)

Age
10
20
30
40



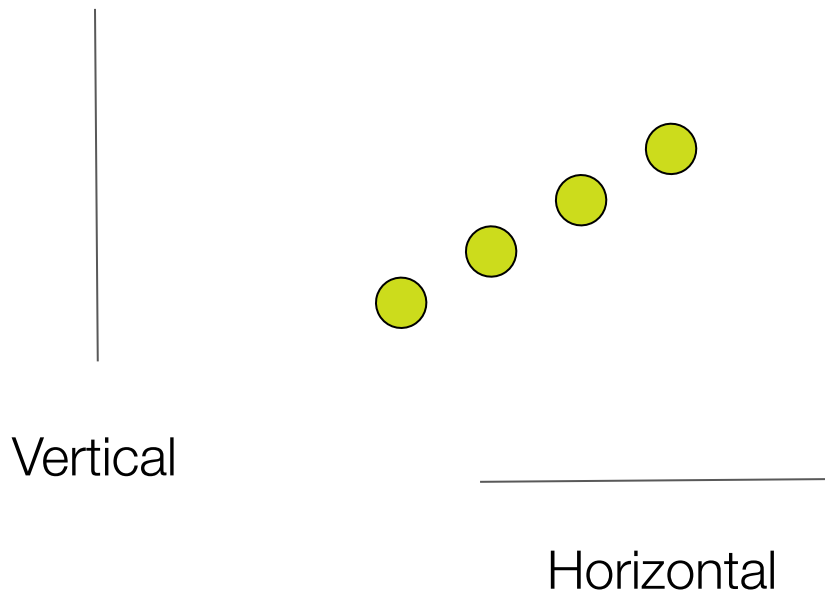
Vertical



Horizontal

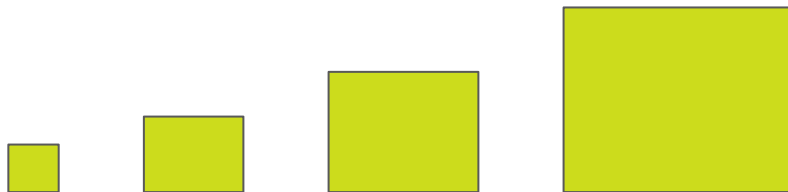
Representation (Mark-point, Channel-**position**)

Age	Height
10	20
20	40
30	60
40	80



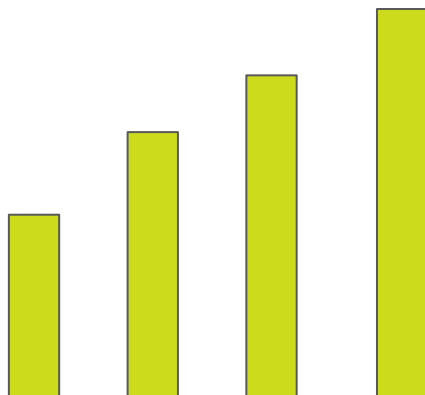
Representation (Mark-area, Channel-size)

Age
10
20
30
40



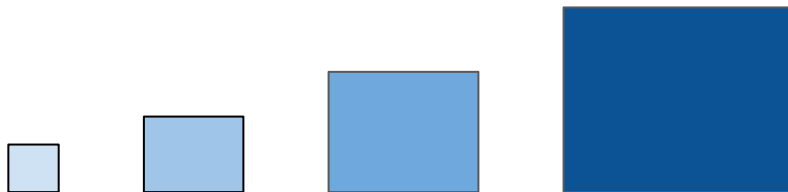
Representation (Mark-line, Channel-size)

Age
10
20
30
40



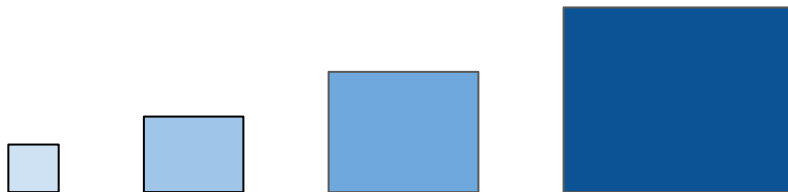
Representation (Mark-area, Channel-color saturation)

Age
10
20
30
40



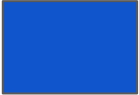



Representation (Mark-area, Channel-color saturation)

Income
10000
20000
30000
40000



Representation (Mark-area, Channel-color hue)

State	
MA	
NH	
NY	
GA	

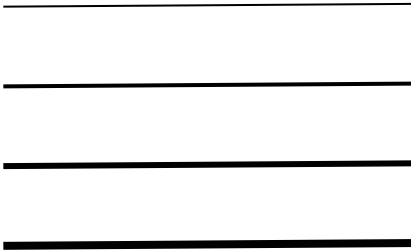
Representation (Mark-area, Channel-shape)

State
MA
NH
NY
GA



Representation (Mark-line, Channel-thickness)

Source	Destination	#Flights
BOS	LGA	1000
BOS	DTW	2000
BOS	ATL	3000
BOS	ORD	4000



Marks and Channels

Type of variable	Examples	Appropriate scale	Description
quantitative/numerical continuous	1.3, 5.7, 83, 1.5×10^{-2}	continuous	Arbitrary numerical values. These can be integers, rational numbers, or real numbers.
quantitative/numerical discrete	1, 2, 3, 4	discrete	Numbers in discrete units. These are most commonly but not necessarily integers. For example, the numbers 0.5, 1.0, 1.5 could also be treated as discrete if intermediate values cannot exist in the given dataset.
qualitative/categorical unordered	dog, cat, fish	discrete	Categories without order. These are discrete and unique categories that have no inherent order. These variables are also called <i>factors</i> .
qualitative/categorical ordered	good, fair, poor	discrete	Categories with order. These are discrete and unique categories with an order. For example, “fair” always lies between “good” and “poor”. These variables are also called <i>ordered factors</i> .
date or time	Jan. 5 2018, 8:03am	continuous or discrete	Specific days and/or times. Also generic dates, such as July 4 or Dec. 25 (without year).
text	The quick brown fox jumps over the lazy dog.	none, or discrete	Free-form text. Can be treated as categorical if needed.

Continuous Variables

#Flights

1000

2000

3000

4000

Channels: Expressiveness Types and Effectiveness Ranks

➔ 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)



➔ Identity Channels: Categorical Attributes

Spatial region



Color hue



Motion



Shape



Most

Effectiveness

Least

Same

Same

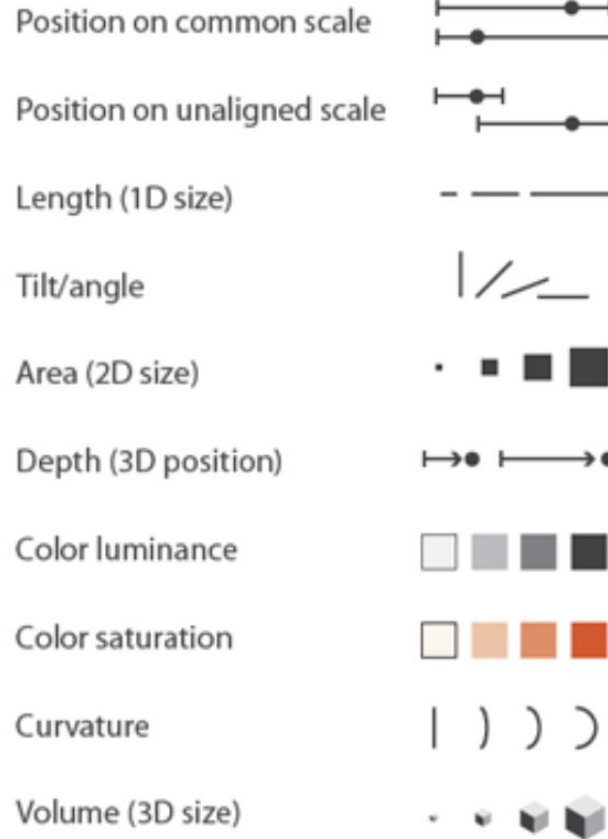
Munzner, T. (2014). *Visualization analysis and design*. CRC press.

Categorical Variables

State
MA
NH
NY
GA

Channels: Expressiveness Types and Effectiveness Ranks

➔ Magnitude Channels: Ordered Attributes



➔ Identity Channels: Categorical Attributes



Munzner, T. (2014). *Visualization analysis and design*. CRC press.

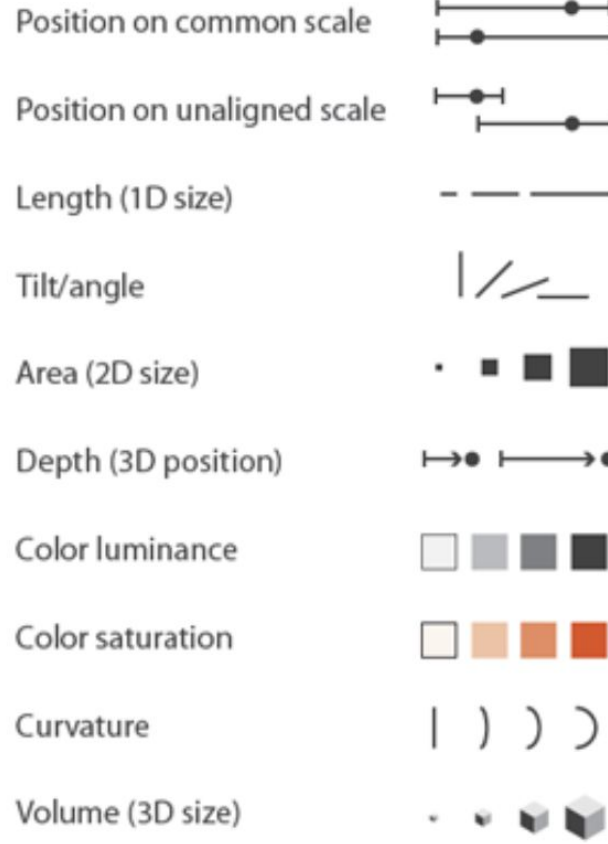
Categorical Variables

Income
Low
Med
High

Performance
Poor
Satisfactory
Good

Channels: Expressiveness Types and Effectiveness Ranks

➔ Magnitude Channels: Ordered Attributes



➔ Identity Channels: Categorical Attributes



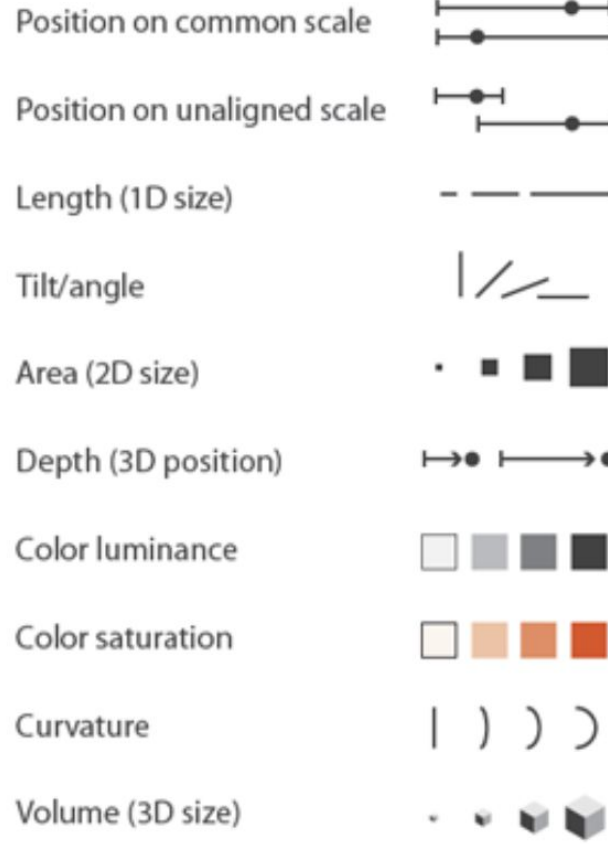
Munzner, T. (2014). *Visualization analysis and design*. CRC press.

Date Variables

Date
1/1/2021
1/2/2021
1/3/2021
1/4/2021

Channels: Expressiveness Types and Effectiveness Ranks

➔ Magnitude Channels: Ordered Attributes



➔ Identity Channels: Categorical Attributes



Munzner, T. (2014). *Visualization analysis and design*. CRC press.

Date Variables

Day
Sunday
Monday
Tuesday
Wednesday

Channels: Expressiveness Types and Effectiveness Ranks

➔ 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)



➔ Identity Channels: Categorical Attributes

Spatial region



Color hue



Motion



Shape



Most

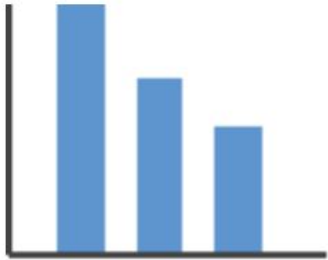
Effectiveness

Least

Same

Same

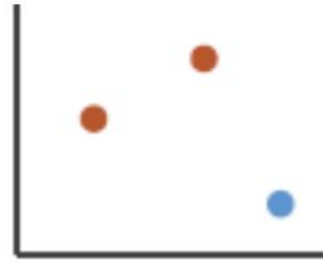
Multivariate Encoding



(a)



(b)



(c)



(d)

Channel Expressiveness and Effectiveness

- The **expressiveness** principle dictates that the visual encoding should express all of, and only, the information in the dataset attributes.*
- The **effectiveness** principle dictates that the importance of the attribute should match the salience of the channel; that is, its noticeability.*
- In other words, the most important attributes should be encoded with the most effective channels in order to be most noticeable, and then decreasingly important attributes can be matched with less effective channels.*

* Munzner, T. (2014). *Visualization analysis and design*. CRC press.

Channel Effectiveness (**Accuracy**)

- **Accuracy:** how close is human perceptual judgement to some objective measurement of the stimulus?^{*}
- We perceive different visual channels with different levels of accuracy; they are not all equally distinguishable.^{*}

^{*} Munzner, T. (2014). *Visualization analysis and design*. CRC press.

Channel Effectiveness (Accuracy)

- **Psychophysics**, the subfield of psychology devoted to the systematic measurement of general human perception
- “Our responses to the sensory experience of magnitude are characterizable by power laws, where the exponent depends on the exact sensory modality: most stimuli are magnified or compressed, with few remaining unchanged” - (*Tamara Munzner*)

Channel Effectiveness (Accuracy)

■ **Steven's Power law** $\psi(I) = kI^a$

where ***I*** is the intensity or strength of the stimulus in physical units (energy, weight, pressure, mixture proportions, etc.), **$\psi(I)$** is the magnitude of the sensation evoked by the stimulus, ***a*** is an exponent that depends on the type of stimulation or sensory modality, and ***k*** is a proportionality constant that depends on the units used

Channel Effectiveness (Accuracy)

- Steven's Power law

$$\psi(I) = kI^a$$

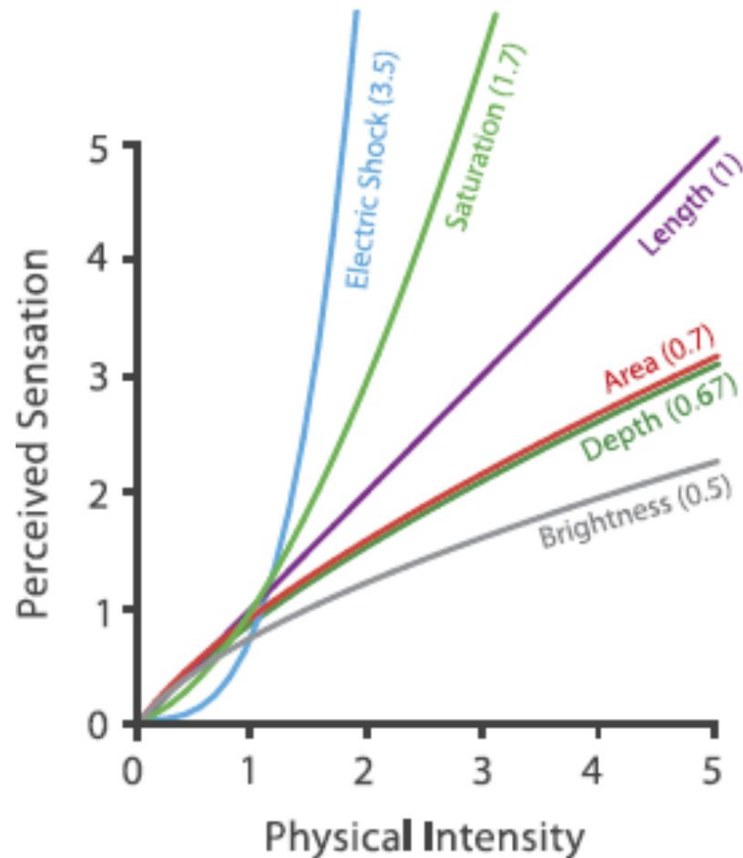


Image from Munzner, T. (2014). *Visualization analysis and design*. CRC press.

https://en.wikipedia.org/wiki/Stevens%27s_power_law

Channel Effectiveness (Accuracy)

- Steven's Power law

$$\psi(I) = kI^a$$

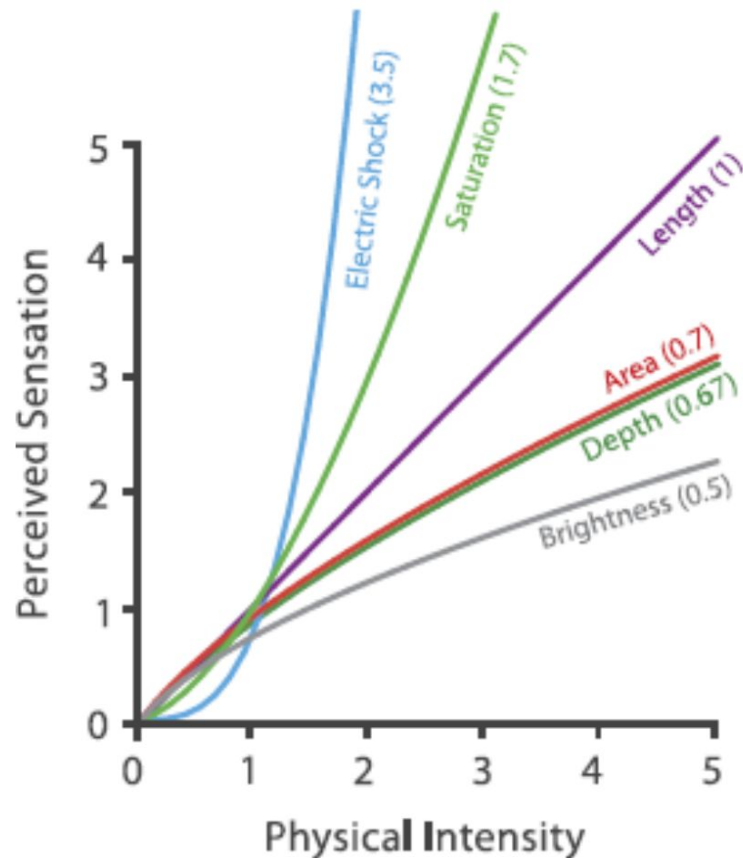
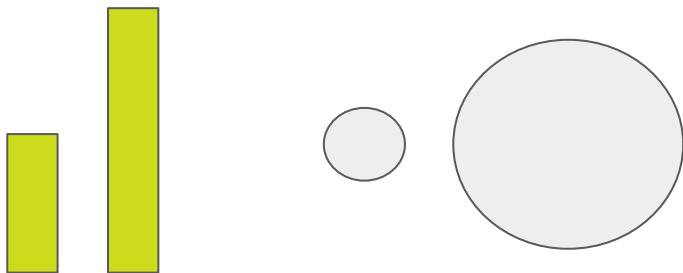
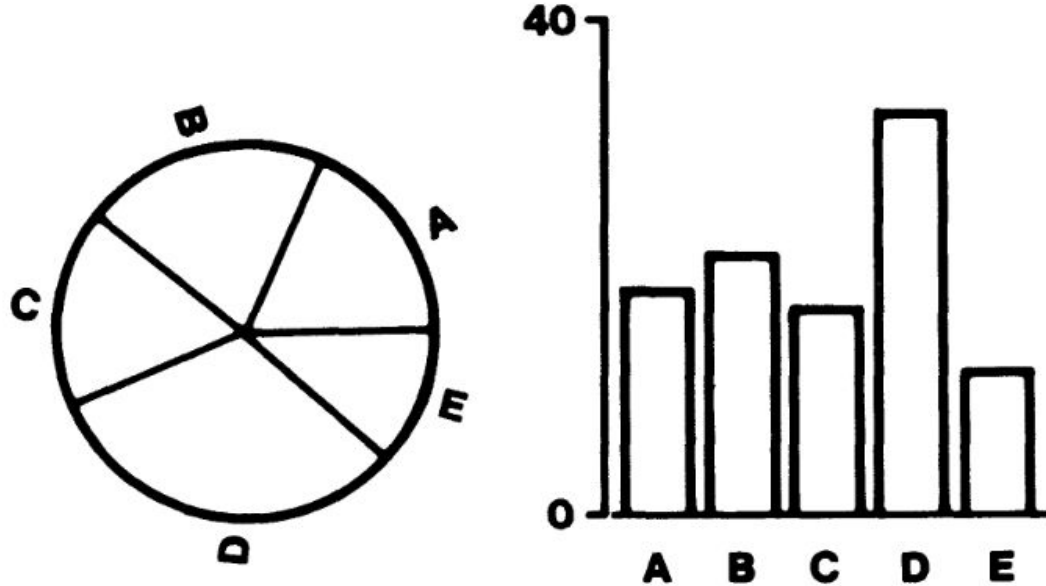


Image from Munzner, T. (2014). *Visualization analysis and design*. CRC press.

https://en.wikipedia.org/wiki/Stevens%27s_power_law

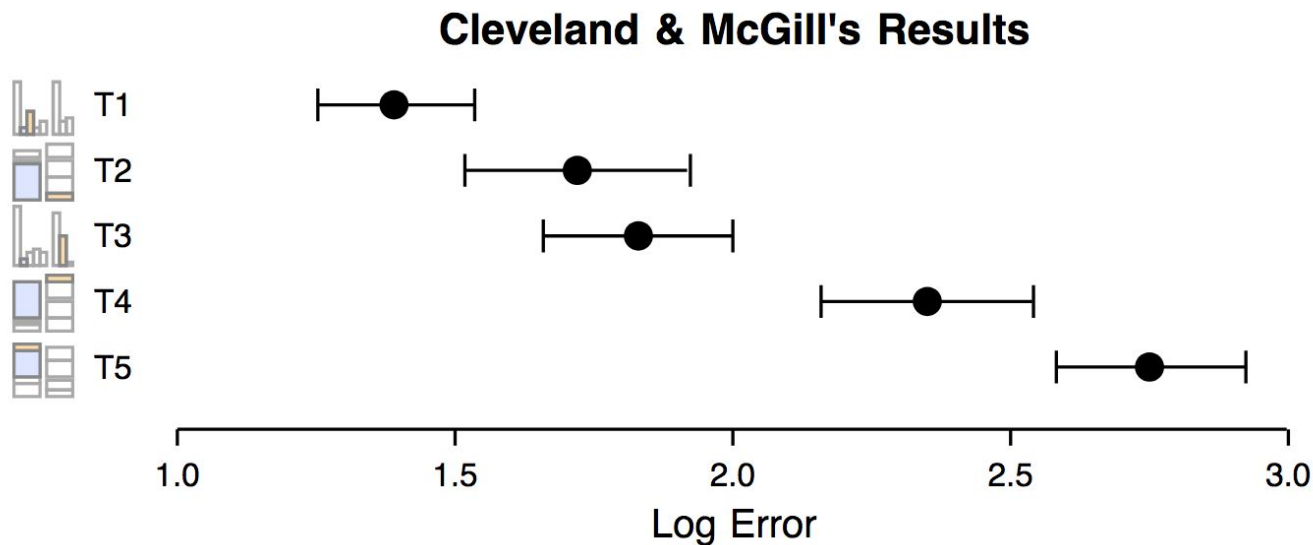
Channel Effectiveness (**Accuracy**)

Cleveland McGill: Ranking Accuracy, 1984



Channel Effectiveness (Accuracy)

Cleveland McGill: Ranking Accuracy, 1984



Channel Effectiveness (Accuracy)

Heer and Bostock 2010

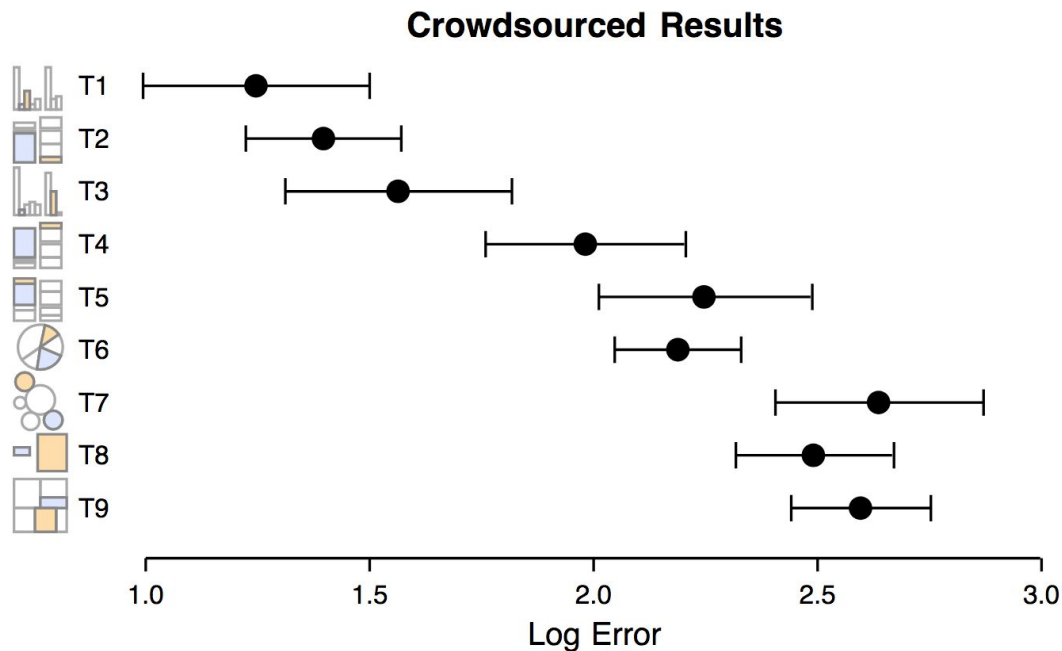


Image Source: <https://www.cs.ubc.ca/~tmm/courses/533-11/slides/eval.jessica.pdf>

Conference Article: <https://dl.acm.org/doi/10.1145/1753326.1753357>

Channel Effectiveness (**Discriminability**)

- If you encode data using a particular visual channel, are the differences between items perceptible to the human as intended?*
- The characterization of visual channel thus should quantify the number of bins that are available for use within a visual channel, where each bin is a distinguishable step or level from the other* - (Tamara Munzner)

* Munzner, T. (2014). *Visualization analysis and design*. CRC press.

Channel Effectiveness (Discriminability)

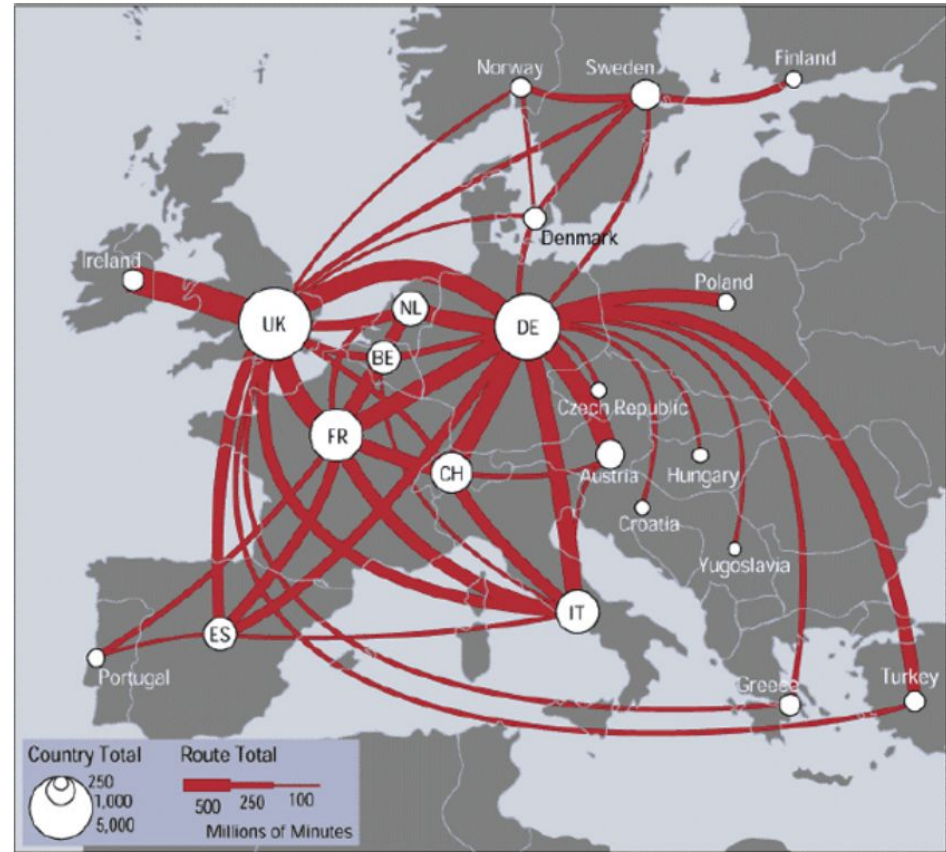


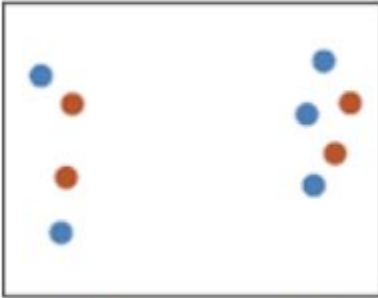
Image Source: Munzner, T. (2014). *Visualization analysis and design*. CRC press.

Channel Effectiveness (**Separability**)

- One cannot treat all visual channels as completely independent from each other, because some have dependencies and interactions with others. -(*Tamara Munzner*)
- One must consider a continuum of potential interactions between channels for each pair, ranging from the orthogonal and independent separable channels to the inextricably combined integral channels.-(*Tamara Munzner*)

Channel Effectiveness (Separability)

Position
+ Hue (Color)



Fully separable

Size
+ Hue (Color)



Some interference

Width
+ Height



Some/significant
interference

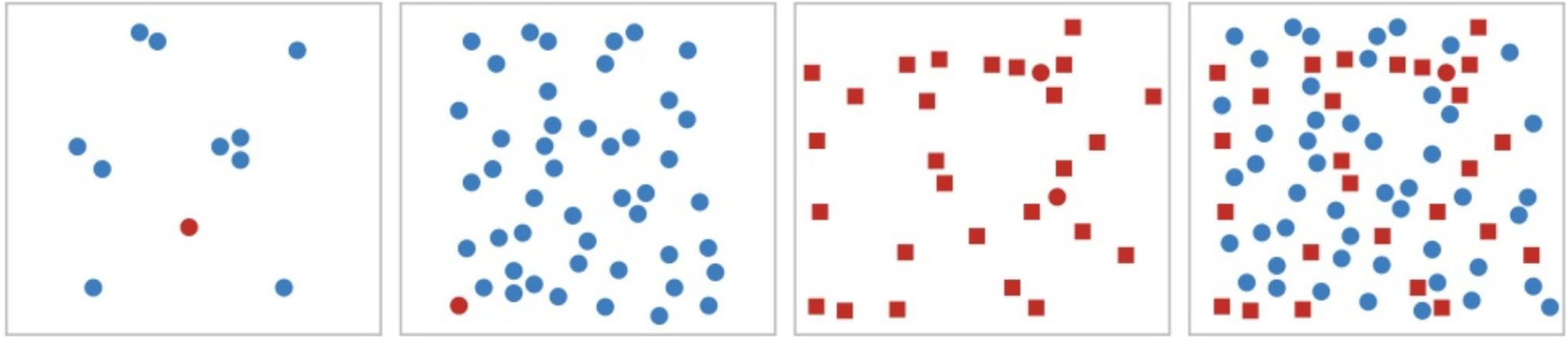
Red
+ Green



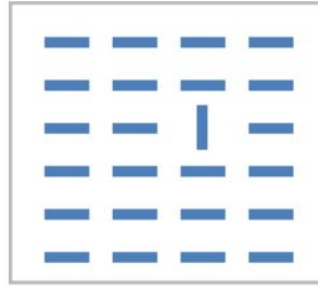
Major interference

Channel Effectiveness (Popout)

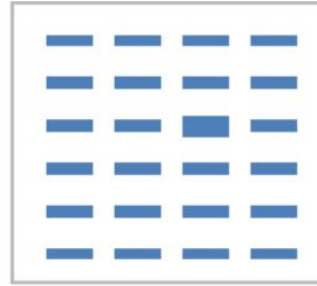
- Visual popout is often called preattentive processing or tunable detection - (*Tamara Munzer*)



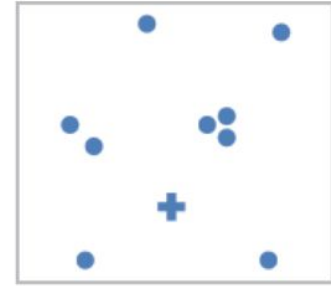
Channel Effectiveness (Popout)



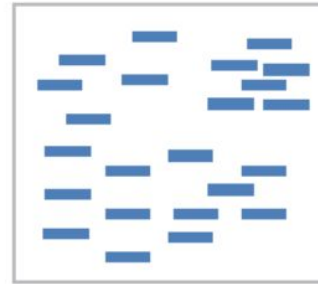
(a)



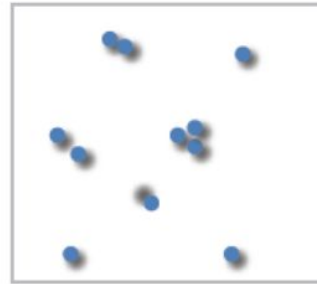
(b)



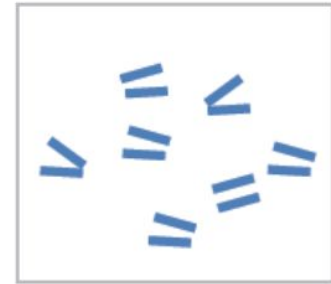
(c)



(d)



(e)

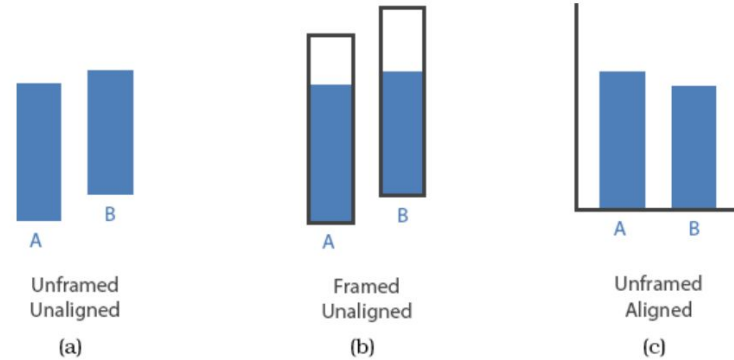


(f)

Channel Effectiveness

(Relative vs. Absolute)

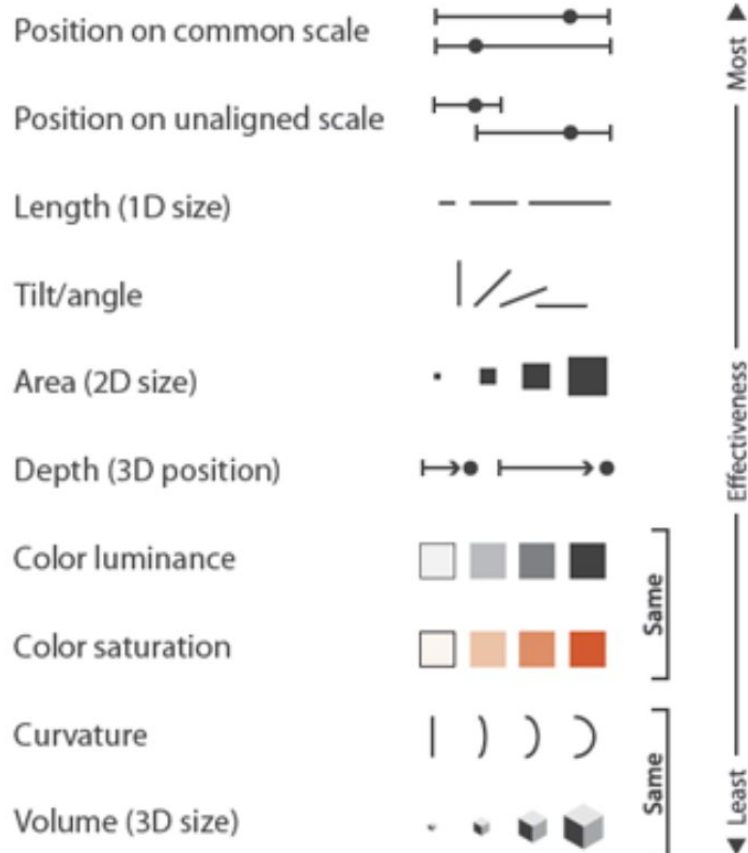
- The human perceptual system is fundamentally based on relative judgements, not absolute ones this principle is known as Weber's Law - *Tamara Munzner*
- For instance, the amount of length difference we can detect is a percentage of the object's length - *Tamara Munzner*



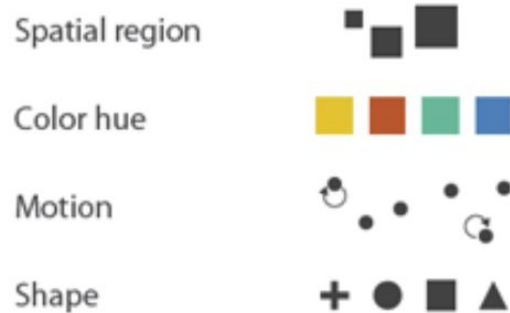
Weber's Law states that we judge based on relative, not absolute differences. (a) The lengths of unframed, unaligned rectangles of slightly different sizes are hard to compare. (b) Adding a frame allows us to compare the very different sizes of the unfilled rectangles between the bar and frame tops. (c) Aligning the bars also makes the judgement easy. Redrawn and extended after [Cleveland and McGill 84a, Figure 12].

Channels: Expressiveness Types and Effectiveness Ranks

➔ **Magnitude** Channels: **Ordered** Attributes



➔ **Identity** Channels: **Categorical** Attributes



Next Week (**Week 5.1**)

- Assignment A2 will be posted today (02/11)
- Quiz 1 (Introduction to data vis, visual encoding))
- Week 5.1 (Visualizing amounts)