

# Arranging space - I

Visualization  
Sebastian Ratzenböck

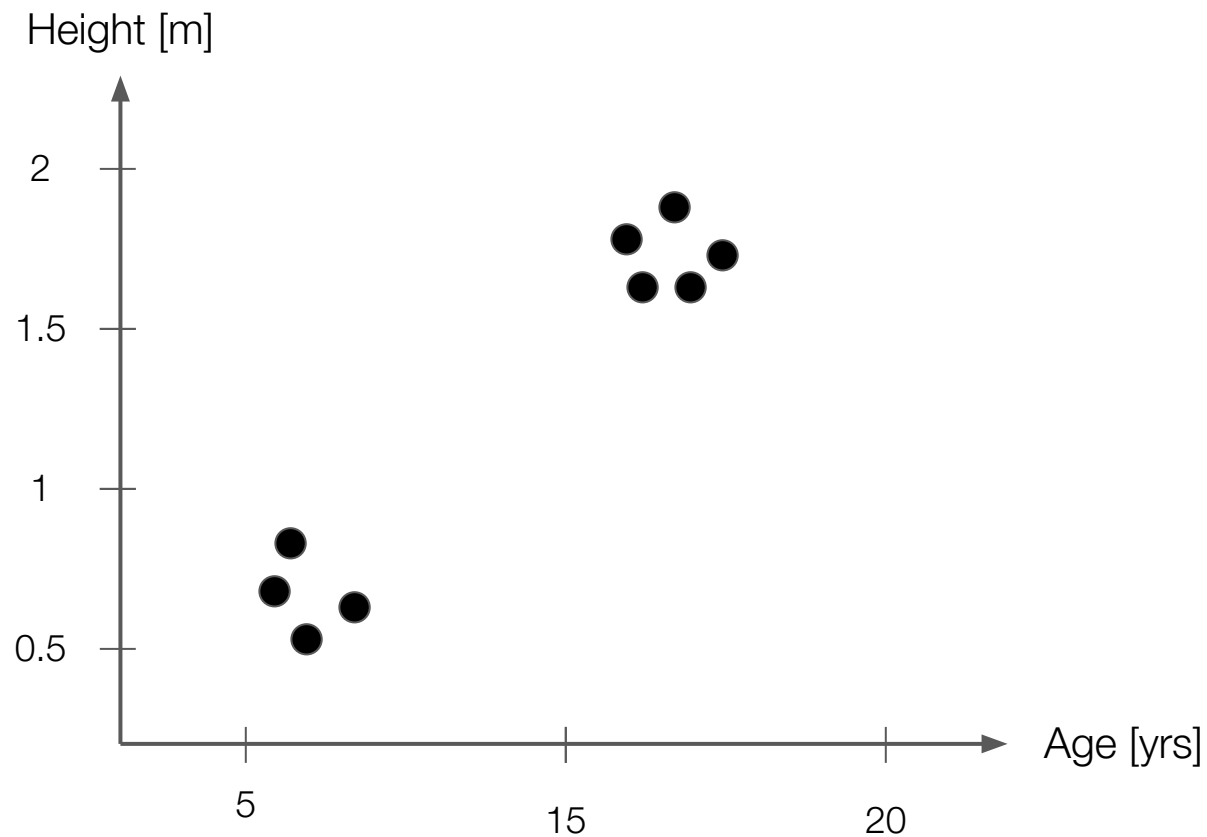
# What do we mean by arranging space?

- Spatial channels for visual encoding
- Most effective encoding choice ↔ tied to effectiveness principle

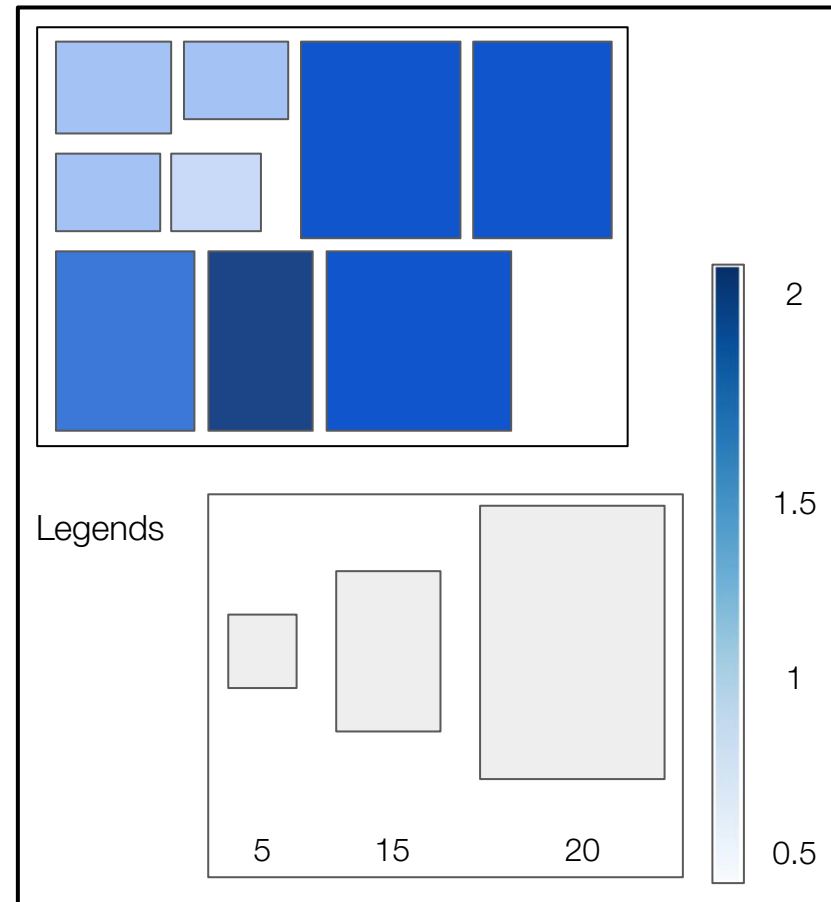
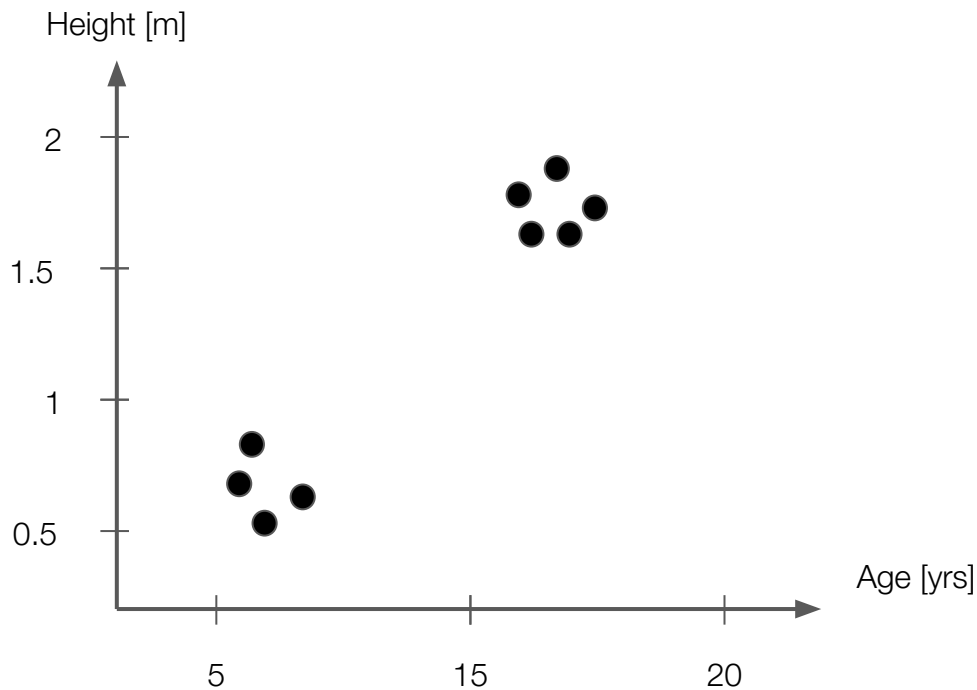
## **Effectiveness principle**

The most important attributes should be encoded with the most effective channels in order to be most noticeable.

# What do we mean by arranging space?



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# What do we mean by arranging space?

- Spatial channels for visual encoding
- Most effective encoding choice  $\leftrightarrow$  tied to **effectiveness principle**
- Spatial channel dominates the users mental model of the data set.

# Readings

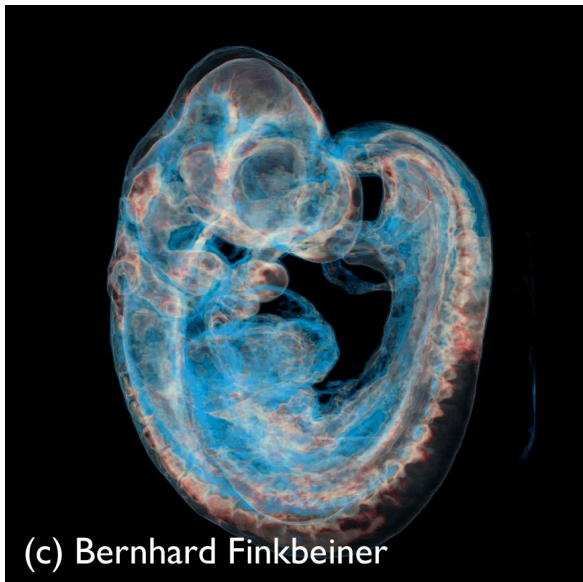
- Munzner, “Visualization Analysis and Design”:
  - Chapter 7 (Arrange Tables)
  - Chapter 8 (Arrange Spatial Data)

# Overview

- Spatial attributes / keys
- quantitative vs. categorical attributes
- Keys: the importance of ordering
  - list (1D) vs. matrix (2D) vs. partition / subdivide
  - (multiple D)
- Spatial layout
  - Rectilinear
  - Parallel
  - Radial
- Spacefilling

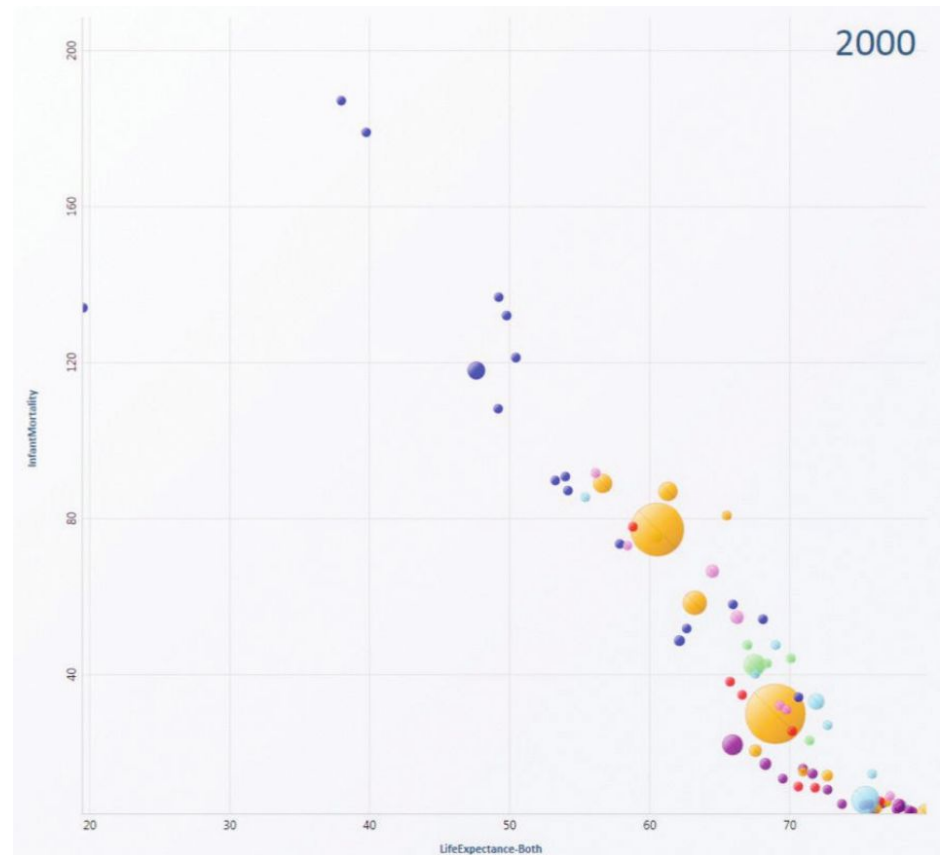
# Spatial attributes

- 1D - line graphs
- 2D - cartographic views
- 3D - volumes
- scalar vs. vector vs. tensor data



# Quantitative vs. categorical

- quantitative: express
  - e.g. dot plot, scatterplot

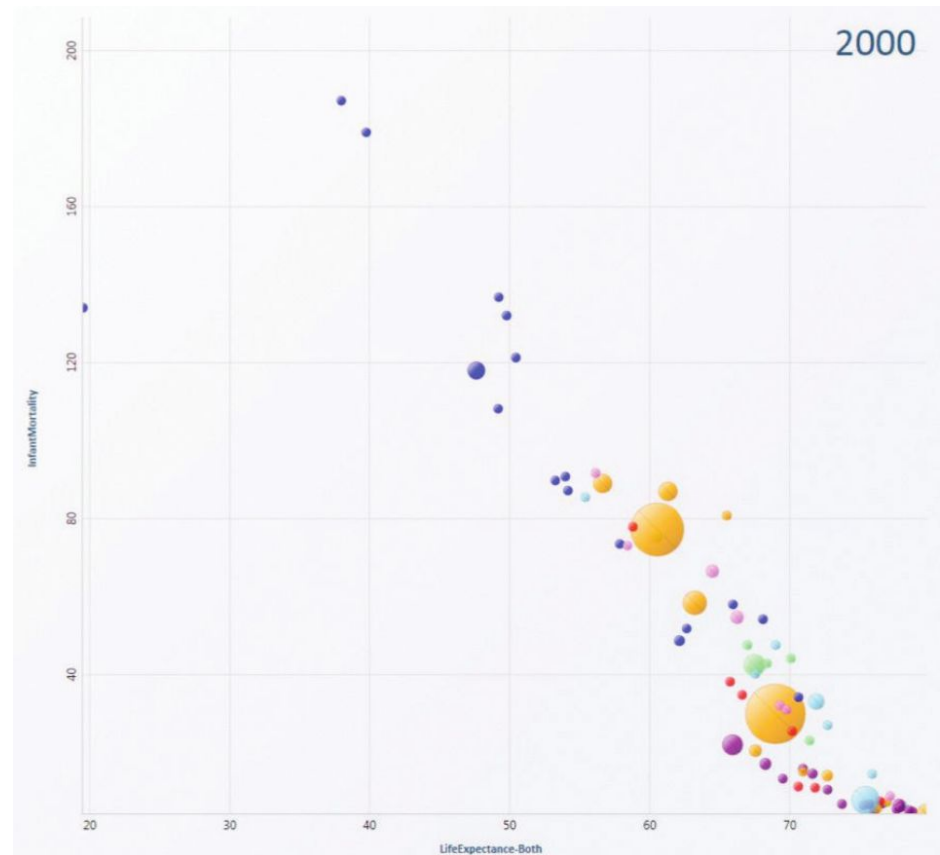


# Quantitative vs. categorical

- quantitative: express
  - e.g. dot plot, scatterplot



How does the **effectiveness principle** apply here?



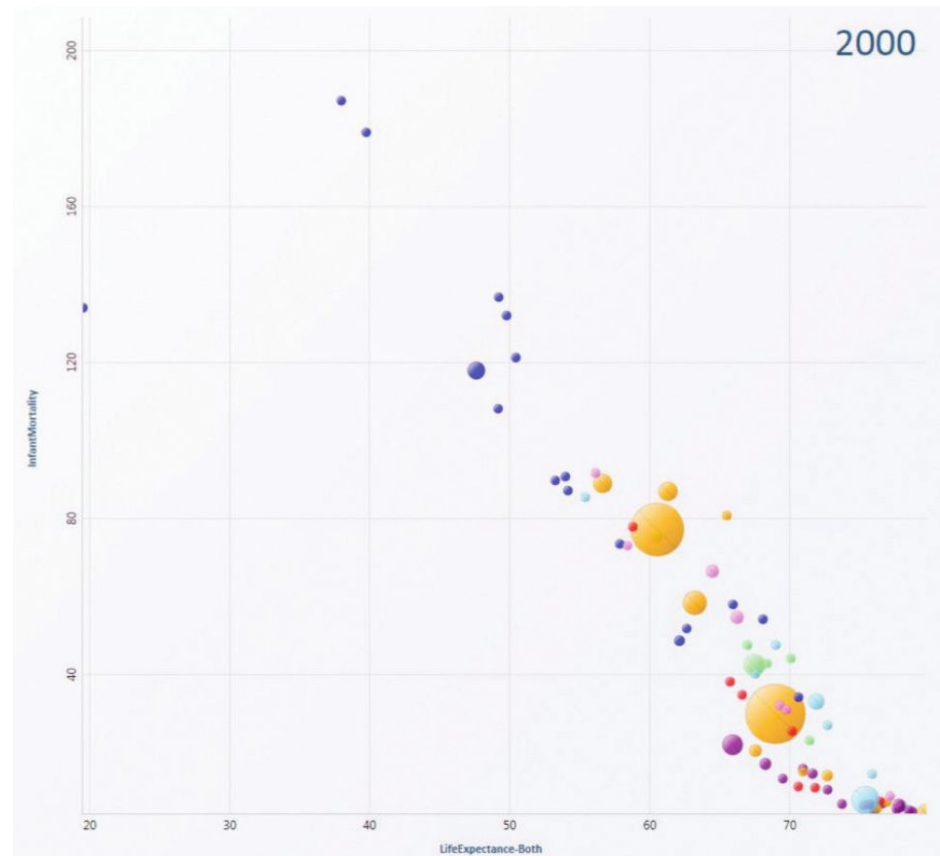
# Quantitative vs. categorical

- quantitative: express
  - e.g. dot plot, scatterplot



## Attributes

- Life expectancy
- Infant mortality
- Country size
- Continent




# Quantitative vs. categorical

- Encoding categorical attributes like quantitative ones forces order that does not exist!  
↳ violates principle of expressiveness

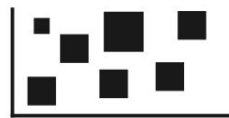
## **Principle of expressiveness**

The visual encoding should express all of, and only, the information in the dataset attributes.

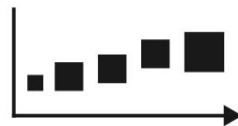
# Quantitative vs. categorical

- quantitative: express 
- e.g. dot plot, scatterplot
- Categorical
  - separate 2D space into regions
  - order along an axis
  - align along an axis

→ Separate



→ Order



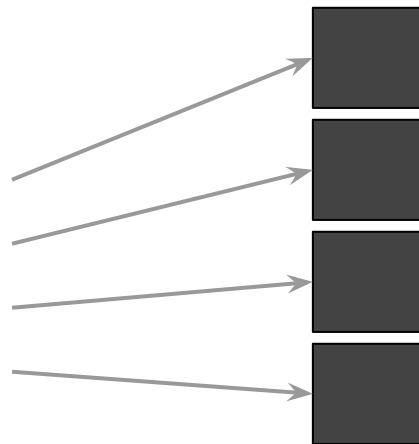
→ Align



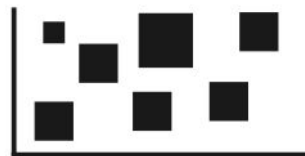
# Separate 2D space into regions

- Regions: bounded areas distinct from another
- Draw all items with same value into one region

	A	B
1	Name	Age
2	Kevin	15
3	Lucy	24
4	Marc	11
5	Maria	13



→ Separate



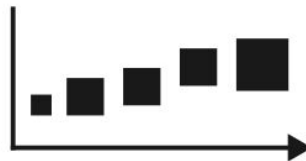
# Quantitative vs. categorical

- Categorical
  - separate 2D space into regions
  - order along an axis
  - align along an axis

→ Separate



→ Order



→ Align



# 1D keys

- 1 key - 1 value → one region per key-value pair

	A	B
1	Name	Age
2	Kevin	15
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4	Marc	11
5	Maria	13



→ Arrange region into 1D list alignment

# 1D keys

- 1 key - 1 value → one region per key-value pair

	A	B
1	Name	Age
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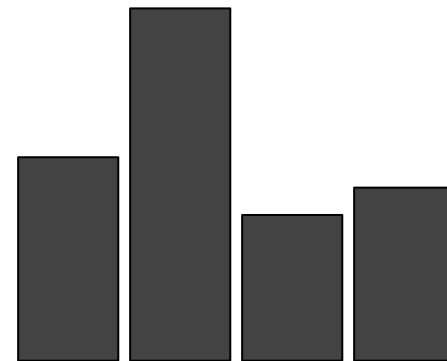


→ Arrange region into 1D list alignment

# 1D keys

- 1 key - 1 value → one region per key-value pair

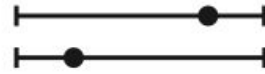
	A	B
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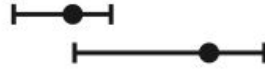
→ Arrange region into 1D list alignment

## ➔ Magnitude Channels: Ordered Attributes

Position on common scale



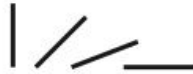
Position on unaligned scale



Length (1D size)



Tilt/angle



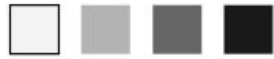
Area (2D size)



Depth (3D position)



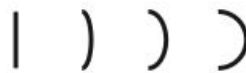
Color luminance



Color saturation



Curvature



Volume (3D size)



Same

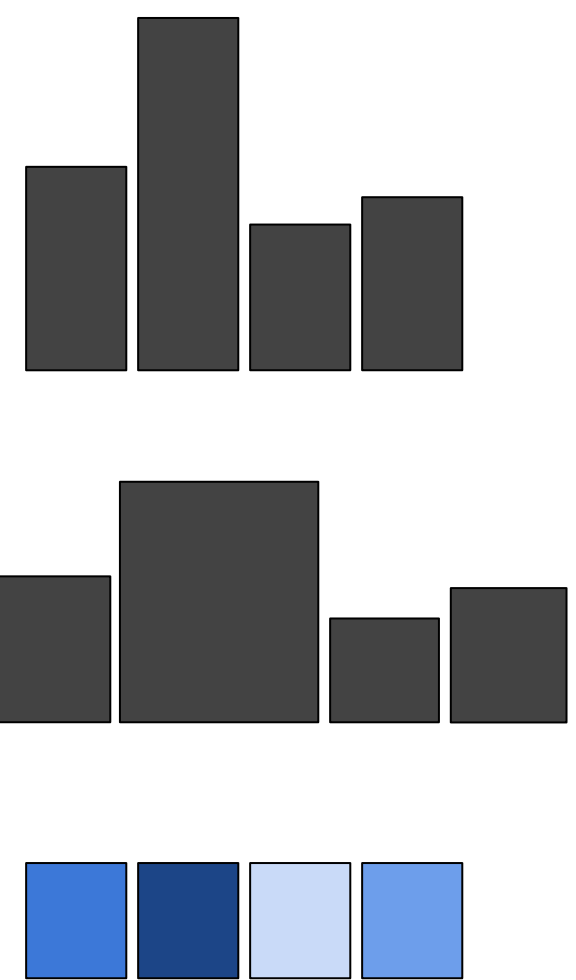
Same

Same

Most

Effectiveness

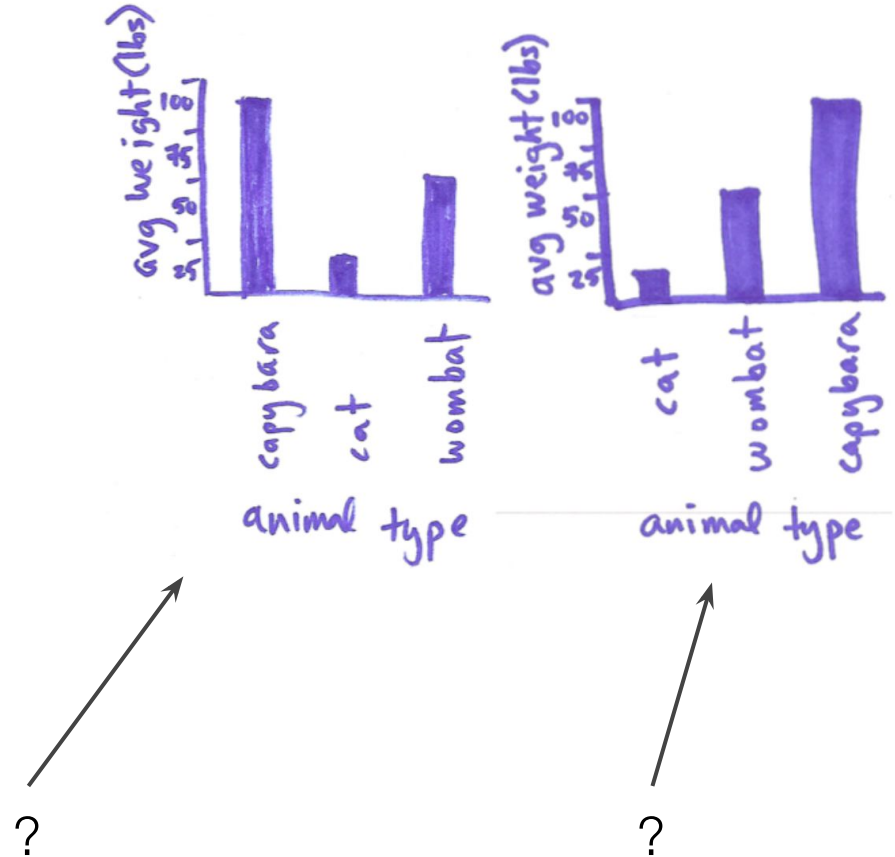
Least



# 1D keys

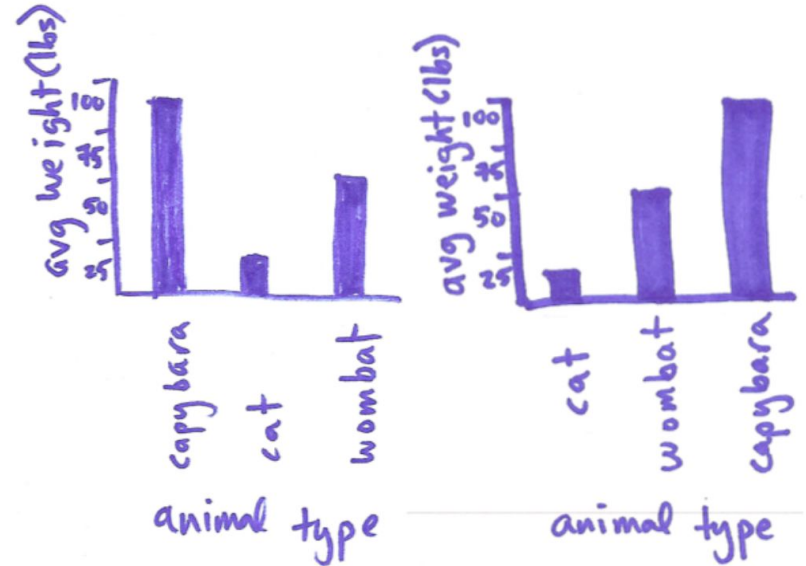
- categorical: bar charts
  - aligned / ordered

**Bar chart**  
Line marks  
Length channel



# 1D keys

- categorical: bar charts
  - aligned / ordered

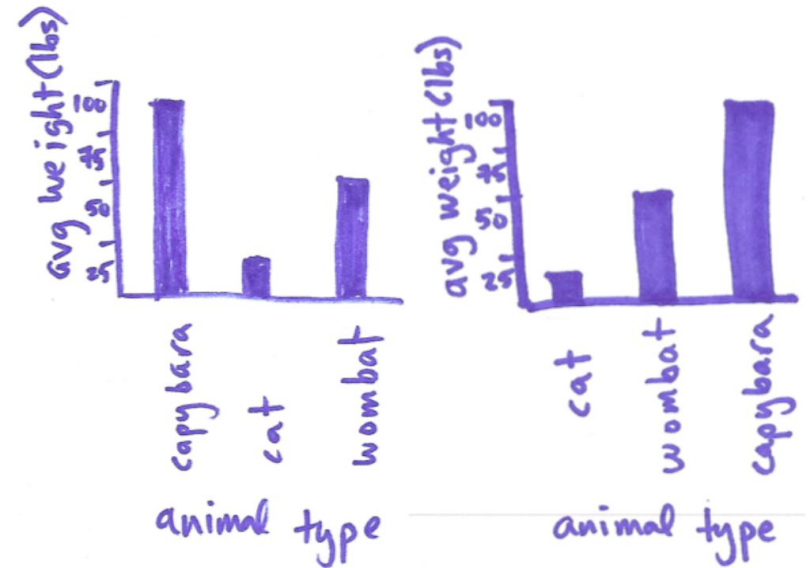


Lookup value

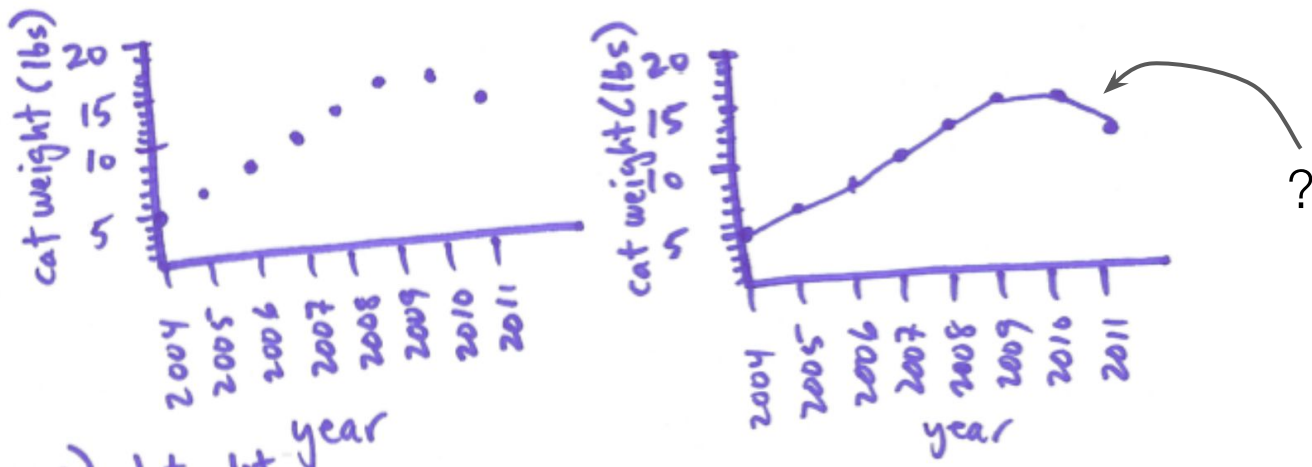
Spot trends

# 1D keys

- categorical: bar charts
  - aligned / ordered

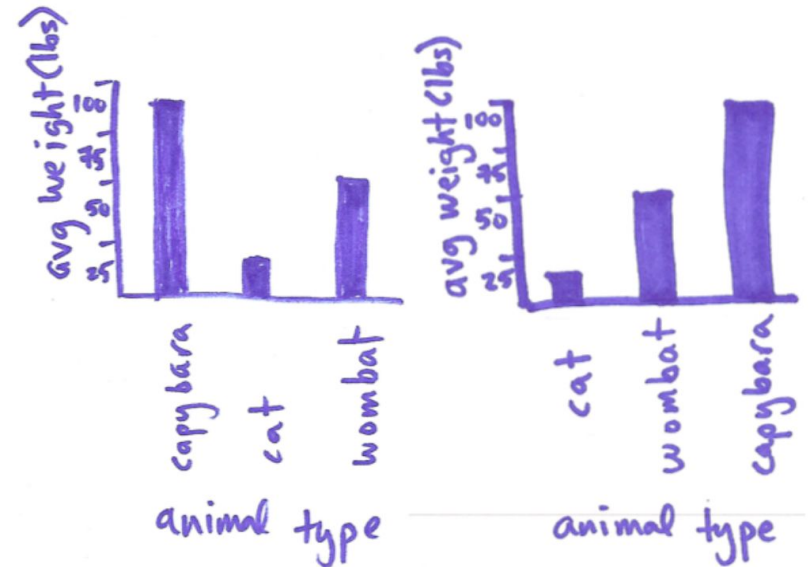


- quantitative/ ordered: dot plot / line chart

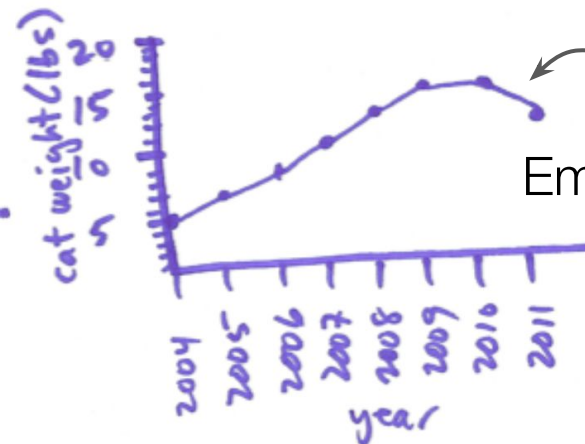
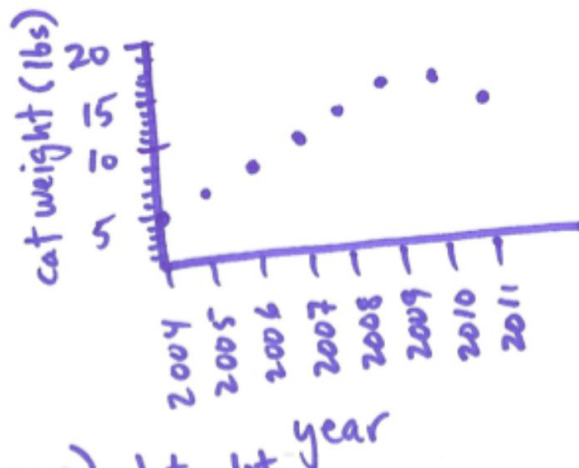


# 1D keys

- categorical: bar charts
  - aligned / ordered



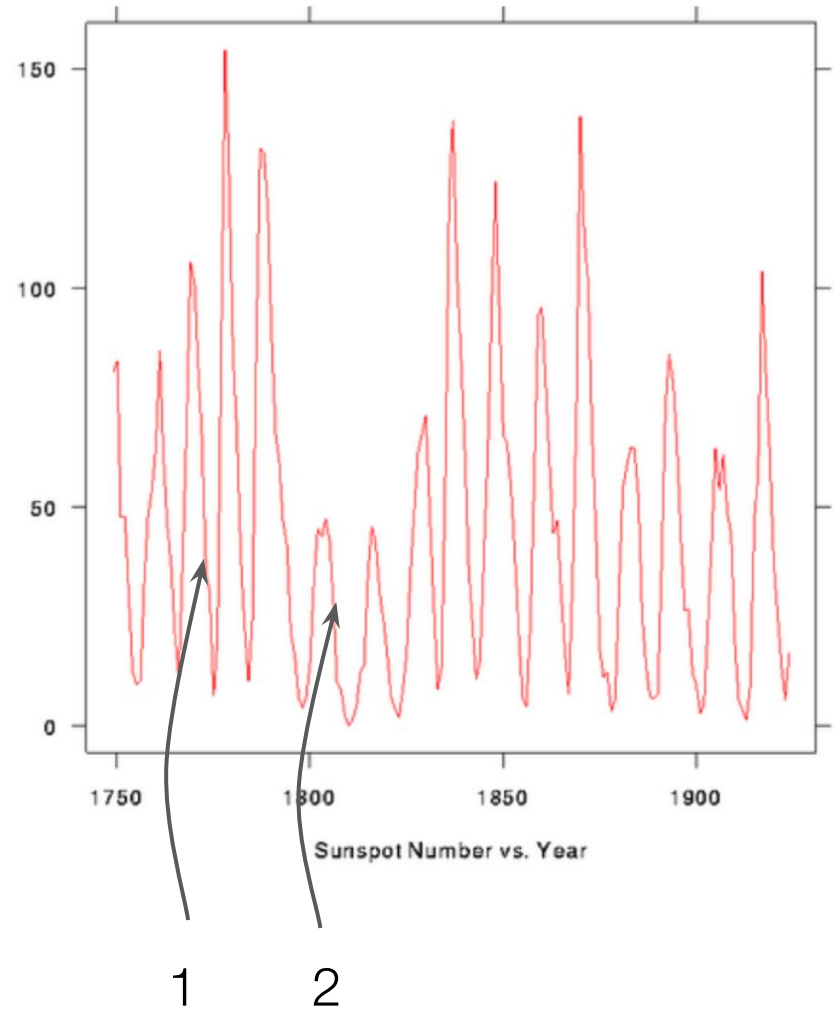
- quantitative/ ordered: dot plot / line chart



Emphasize trends

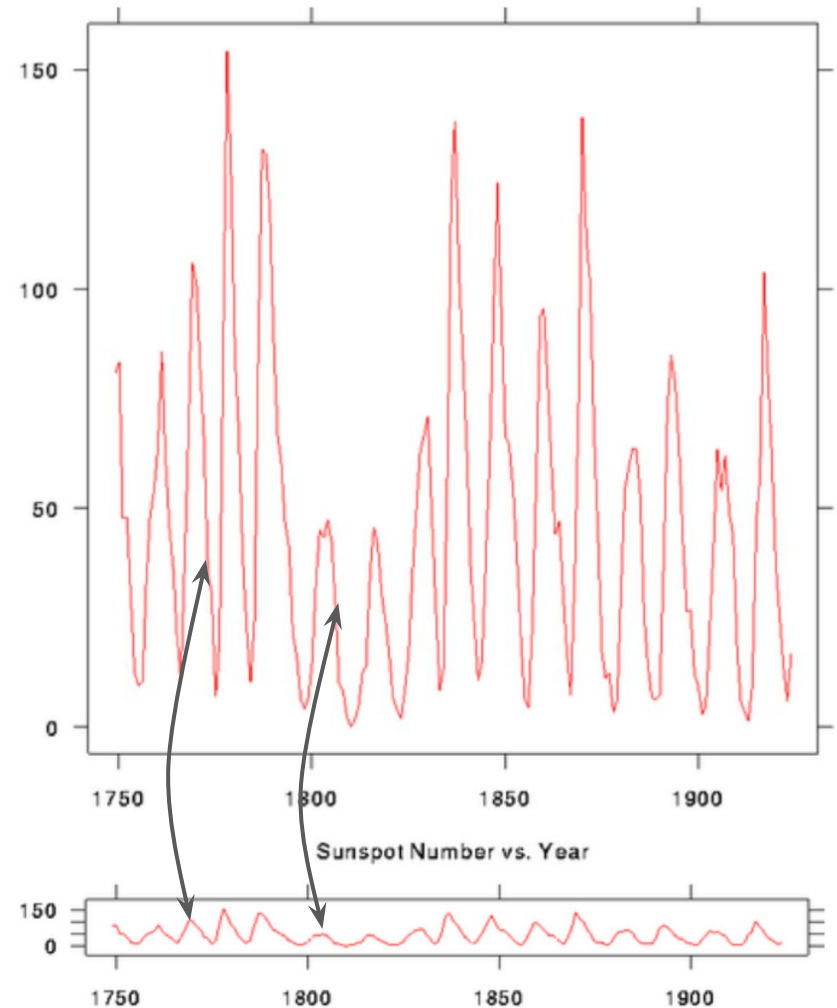
# Line plots: aspect ratio

- Aspect ratio is important
- Which of these two lines is steeper, 1 or 2?



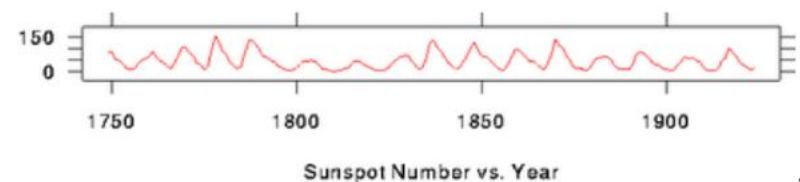
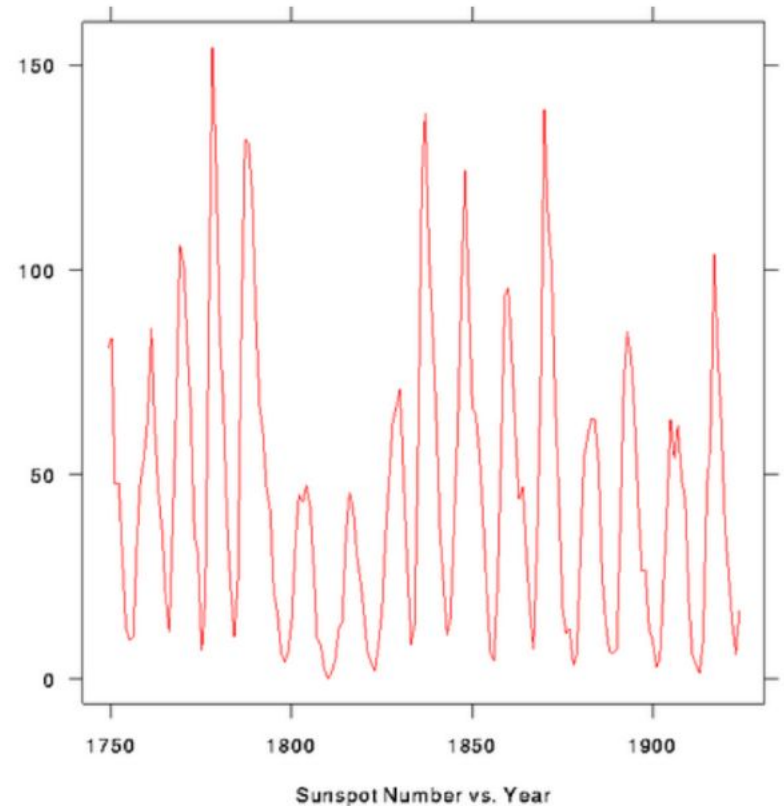
# Line plots: aspect ratio

- Aspect ratio is important
- Which of these two lines is steeper, 1 or 2?
- Did the aspect ratio influence the task?



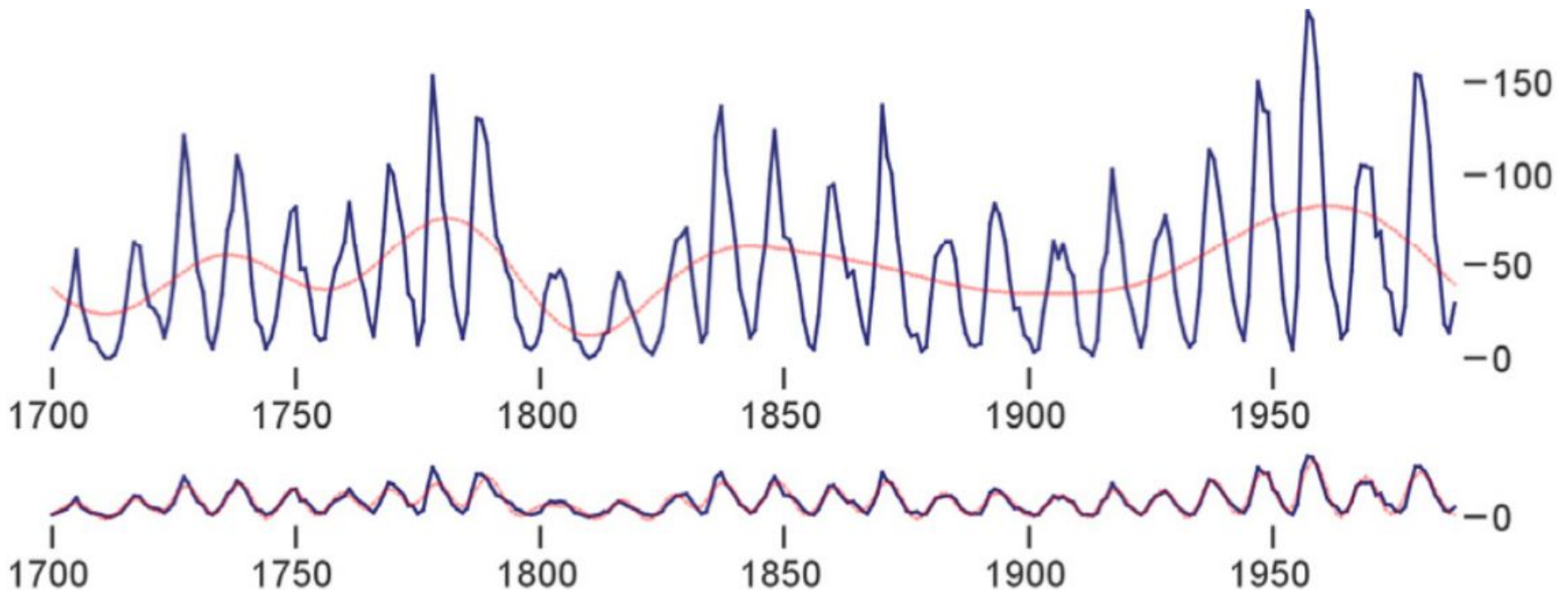
# Banking to 45 degrees

- Perceptual principle:
  - Most accurate angle judgement at  $45^\circ$
- Maximize the number of line segments that fall close to  $45^\circ$   
→ adjust aspect ratio accordingly



# Banking to 45 degrees

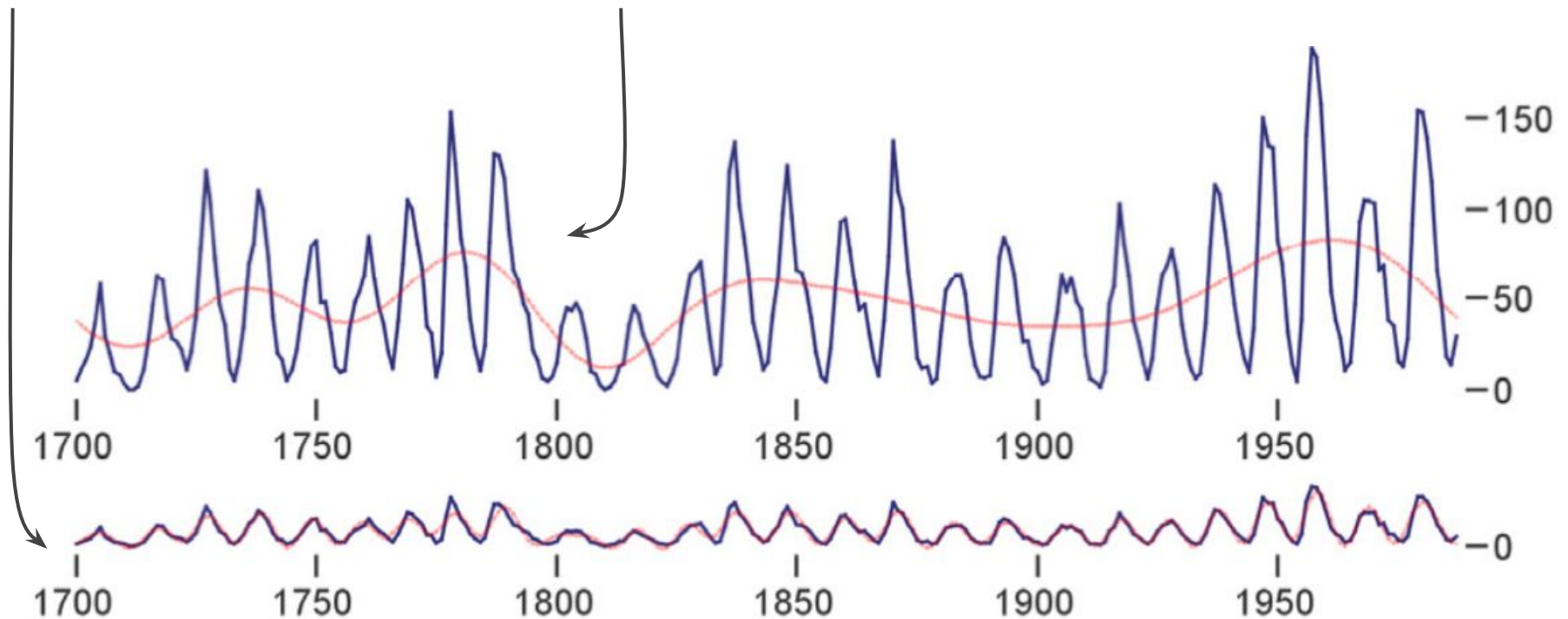
What is the difference between the two ARs?



# Banking to 45 degrees

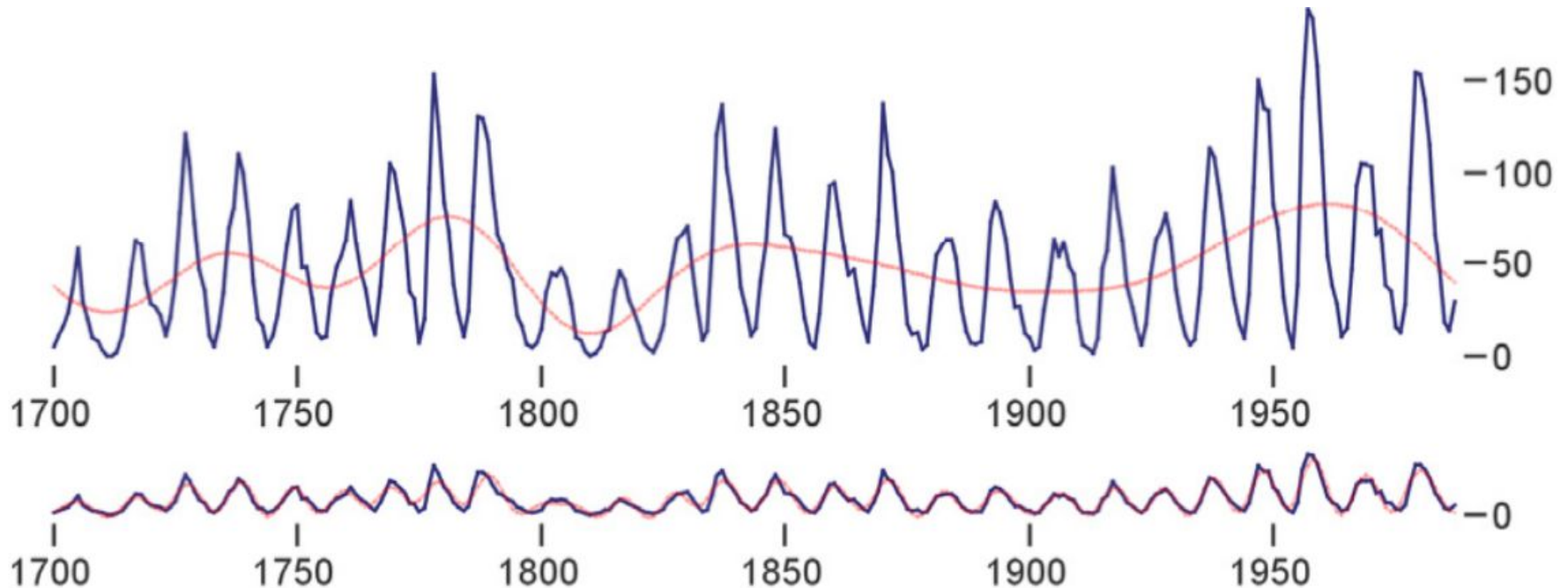
Fast  
changes

Slow changes



# Banking to 45 degrees

- frequency analysis → find dominant frequencies
- Each frequency corresponds to aspect ratio
- AR = 3.96 vs. AR = 22.35



# Quantitative vs. categorical

- Encoding categorical attributes like quantitative ones forces order that does not exist!\*
- ↳ violates principle of expressiveness

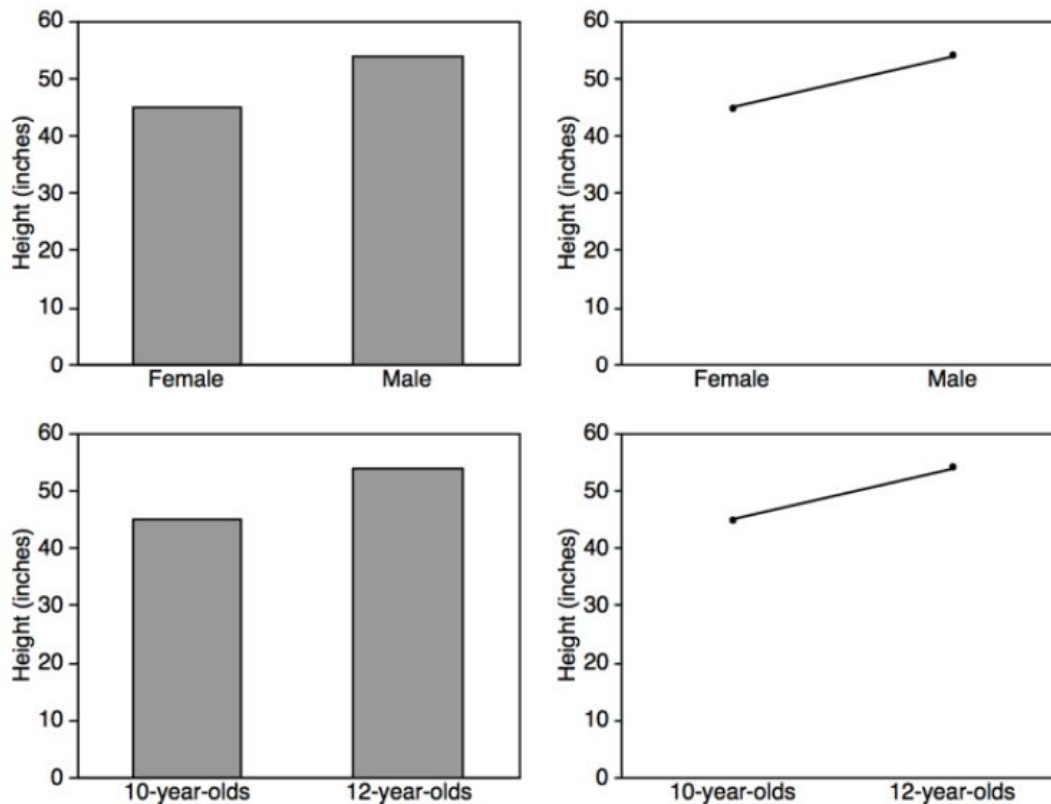
## **Principle of expressiveness**

The visual encoding should express all of, and only, the information in the dataset attributes.

\* This implication is in fact so strong that it can override common knowledge.

# Bar vs. line charts

- line implies trend, not appropriate for categorical data

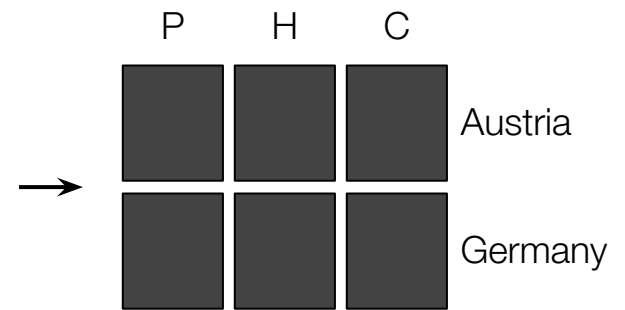


# 2D keys

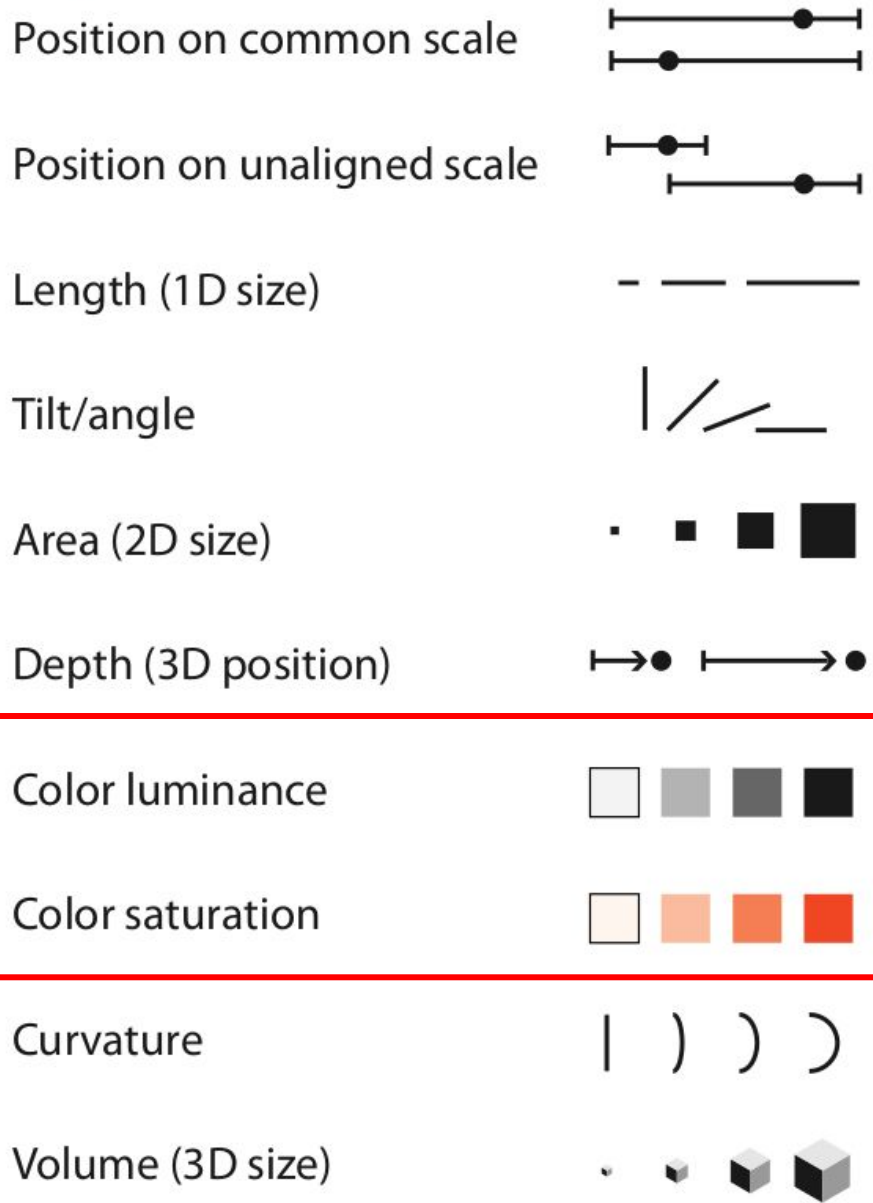
# 2D keys

- 2 keys - 1 value  $\rightarrow$  2D matrix alignment

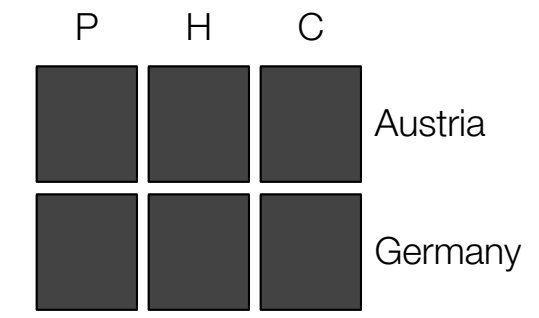
Country	Education level	Percentage
Austria	Primary school	99
Austria	High school	60
Austria	College	20
Germany	Primary school	98
Germany	High school	65
Germany	College	25



➔ **Magnitude Channels: Ordered Attributes**

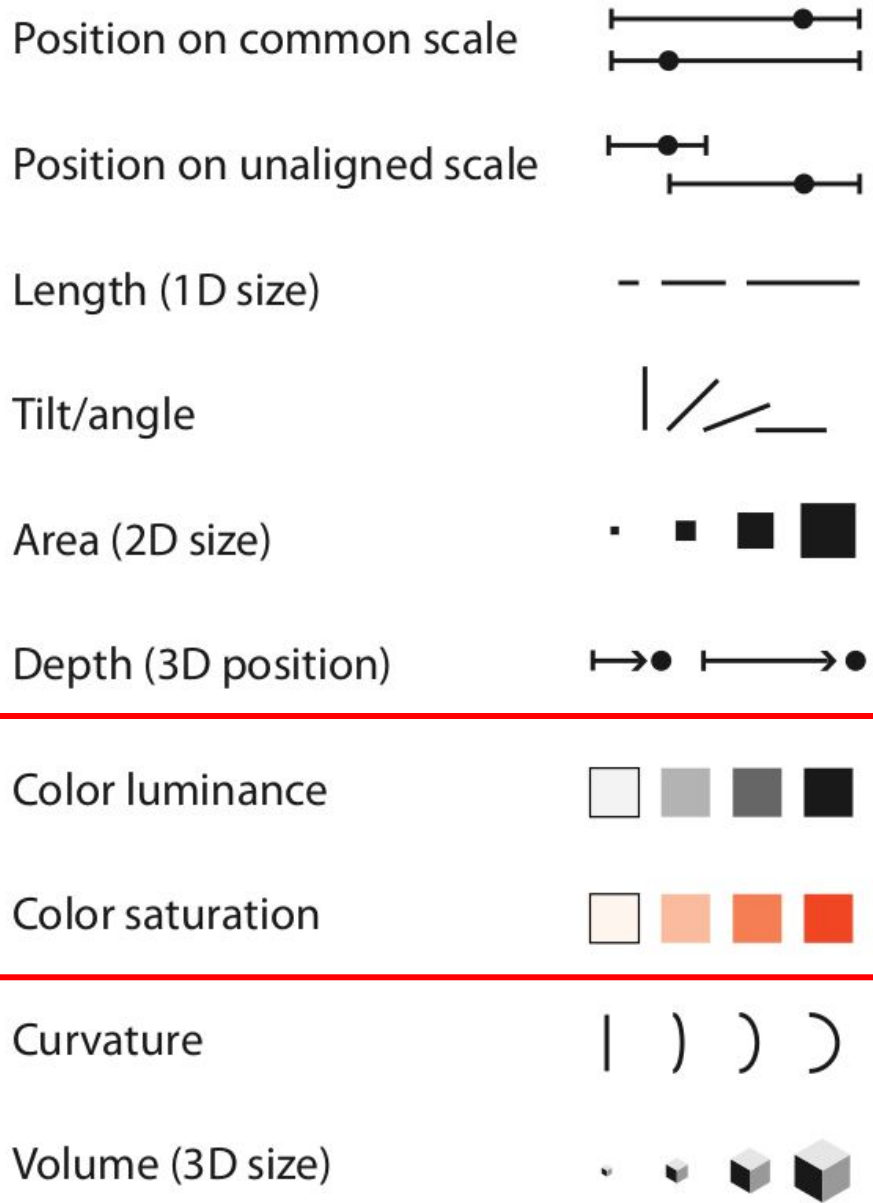


▲ Most  
Effectiveness  
Least ▼

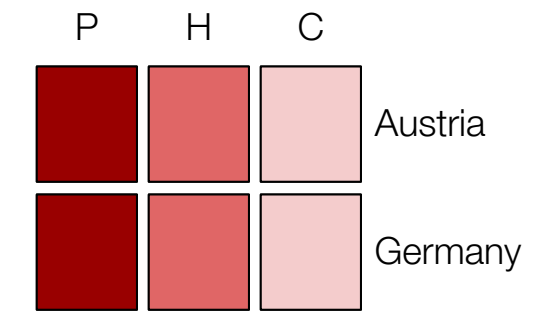


→ Heatmap

➔ **Magnitude Channels: Ordered Attributes**



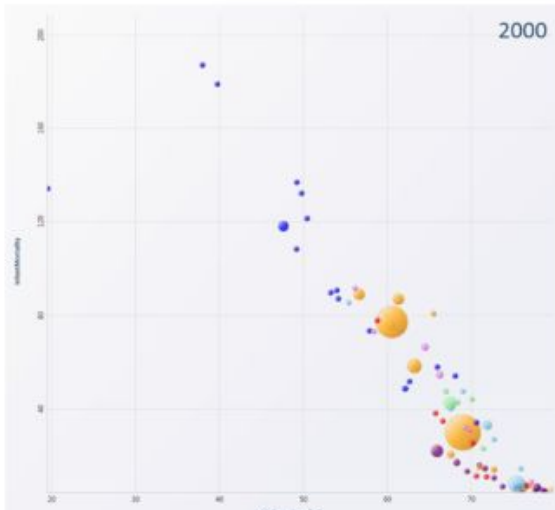
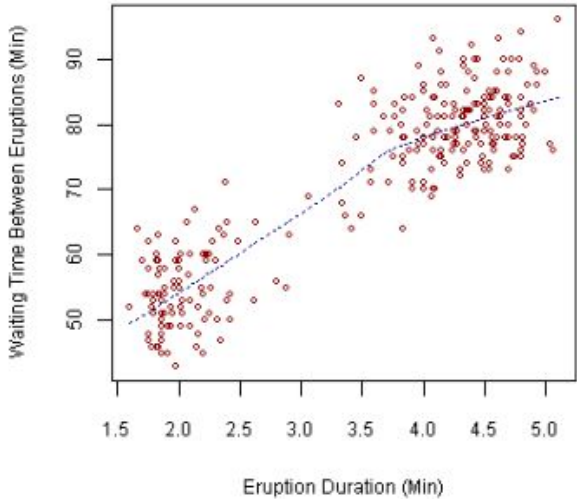
▲ Most  
Effectiveness  
Least ▼



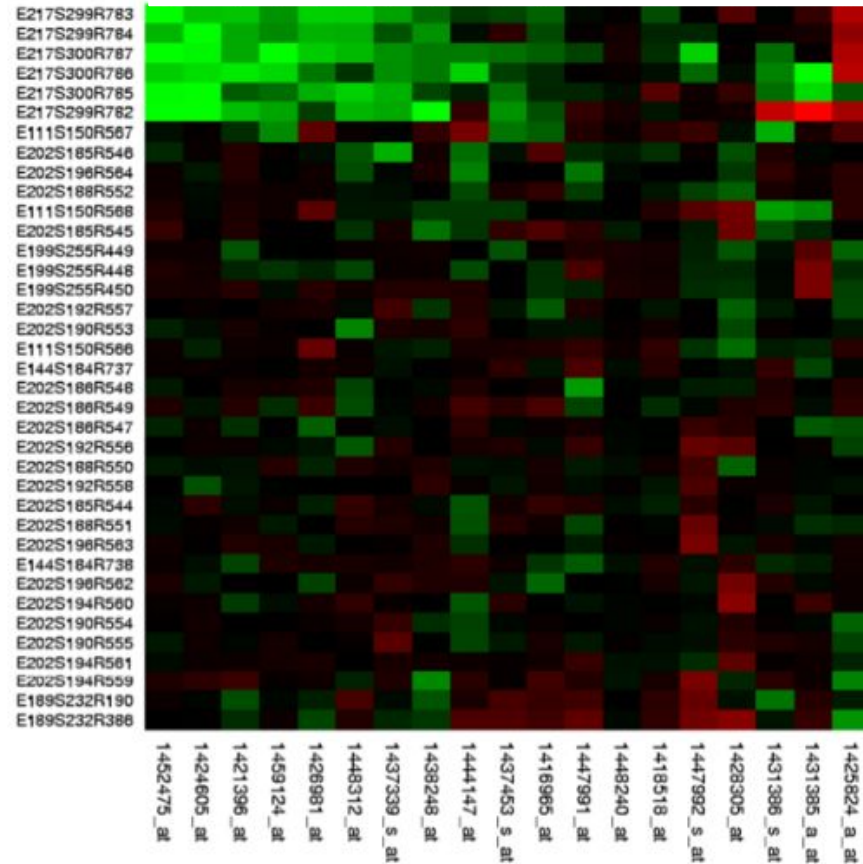
→ Heatmap

# 2D keys examples: tasks, size

Old Faithful Eruptions

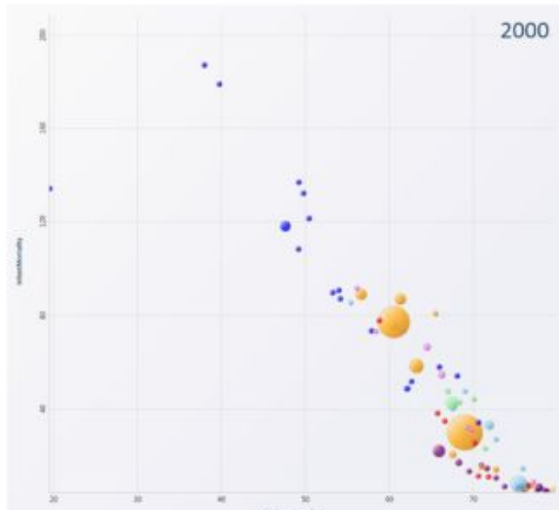
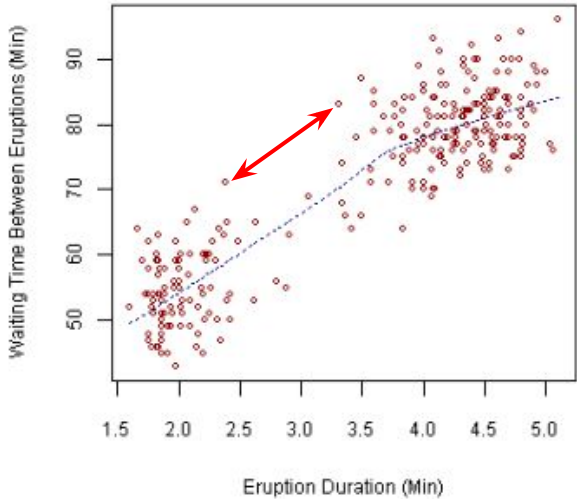


Heatmap

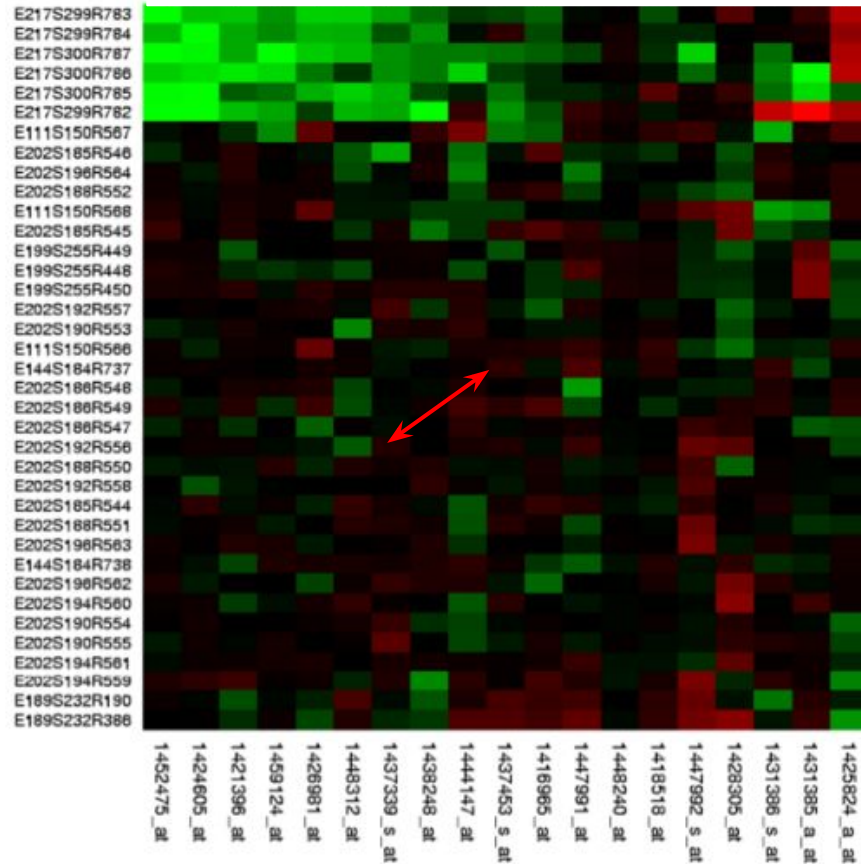


# Meaning of distances

Old Faithful Eruptions

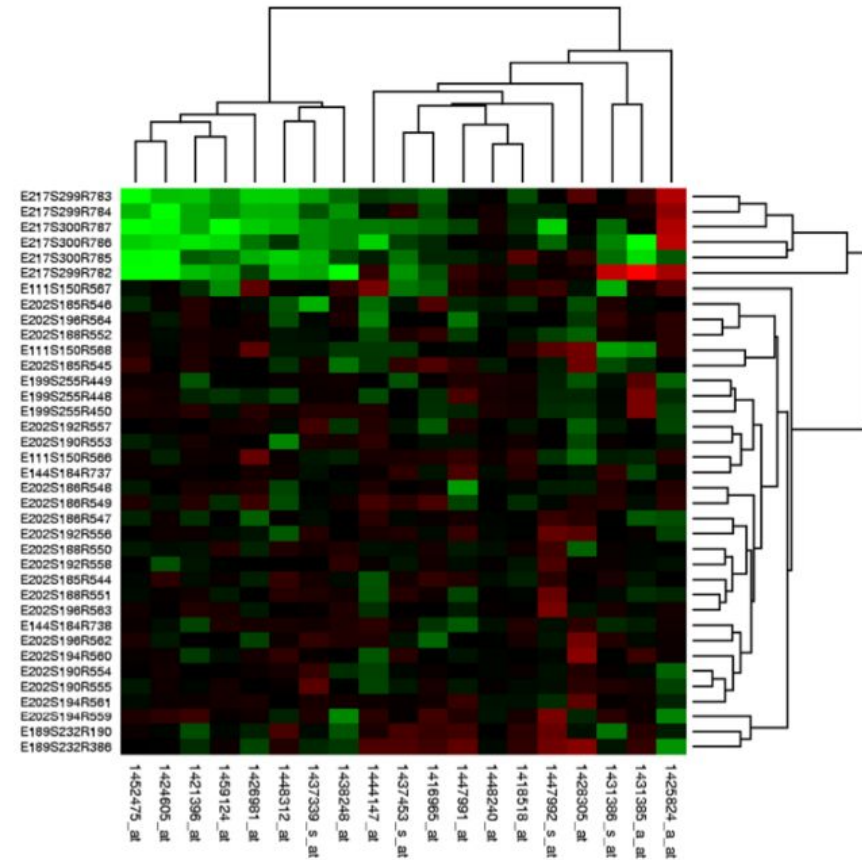


Heatmap



# Heatmaps

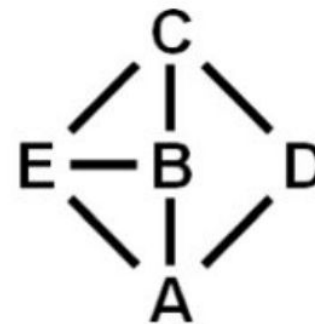
- Very compact
    - Limit: area mark=pixel
    - Good data overview
  - Find structure in data
    - Outliers, clusters, etc.
    - Needs sorted rows & columns
- cluster heatmap



# 2D keys - network matrix

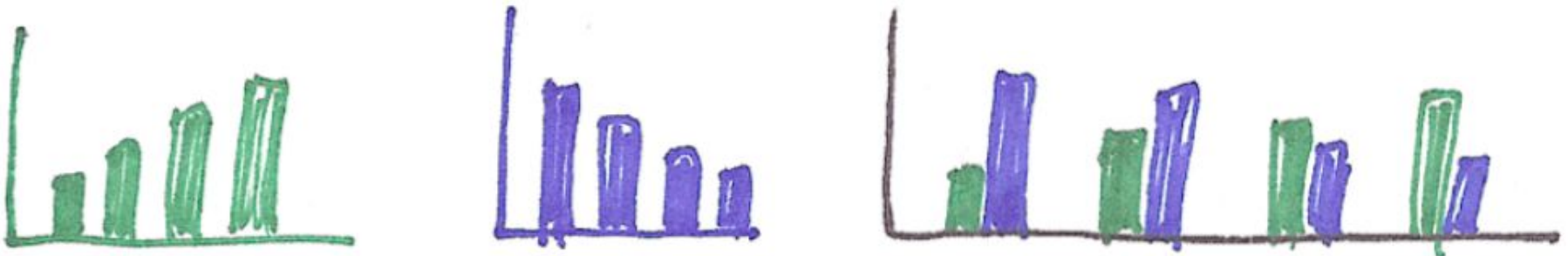
- Value attribute: link exists yes/no
- Weighted networks: e.g. color coding
- Undirected networks: half matrix
- Directed networks: full matrix
- Avoids hairball effect

	A	B	C	D	E
A	A	■		■	■
B		B	■		■
C			C	■	■
D				D	
E					E

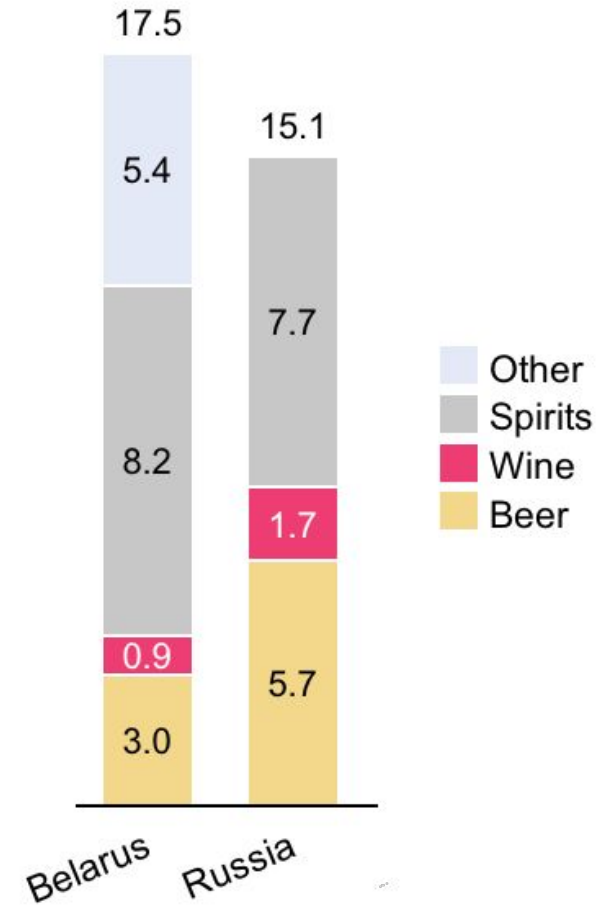
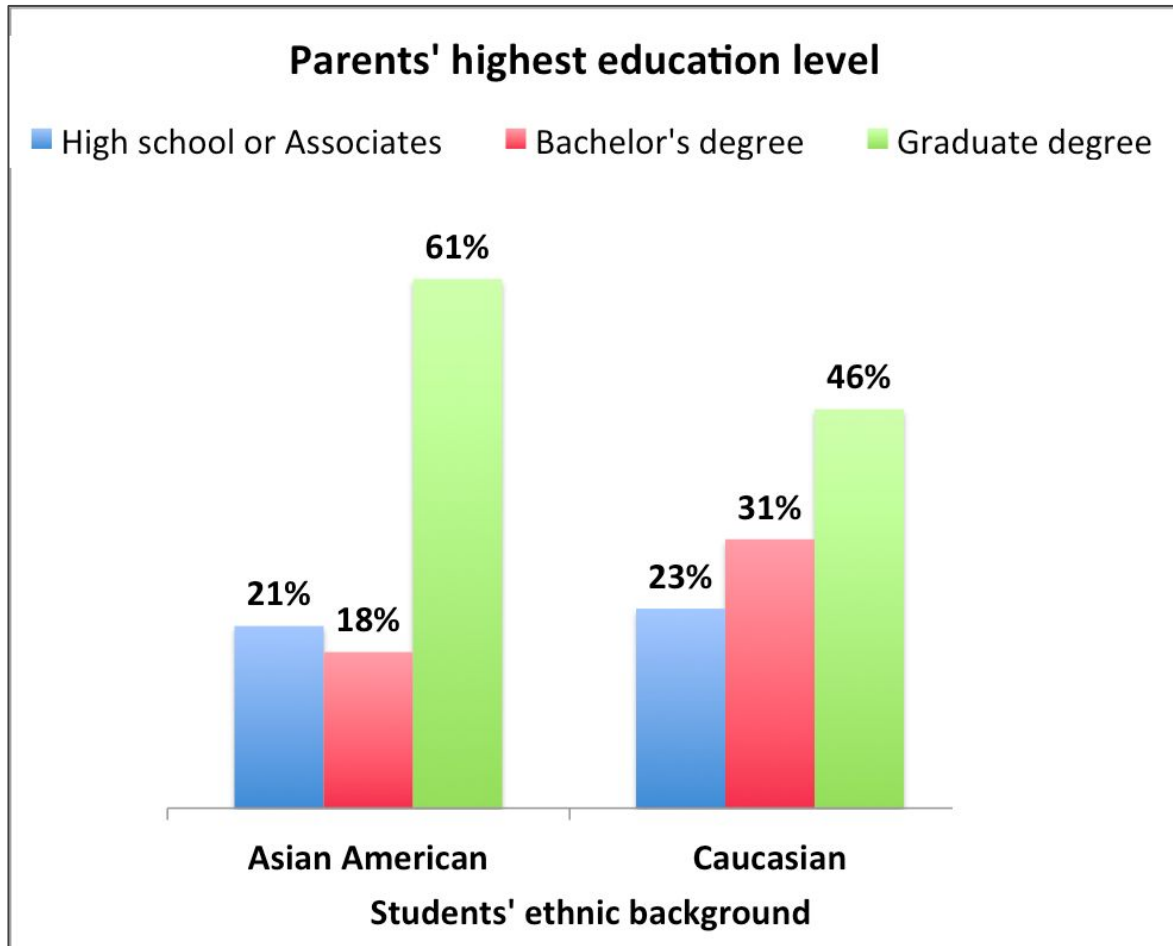


# multiple keys - partitioning / subdivide

- e.g. 2 keys
  - use two perpendicular axis OR
  - use alignment on one axis
    - separate by A first and then by B (left)
    - separate by B first and then by A (right)
- also known as **dimensional stacking**



# 2D keys - 1D list alignment



© <https://cra.org/crm/wp-content/uploads/sites/7/2015/06/Feb-2015.png>

Source: Wikipedia, 2010

# Graphical perception

- Human capacity to visually interpret information on graphs and charts.
  - ↳ Graph & encoding choice determines ability to obtain “correct” information from chart

# Experiment: graphical perception

- Goal: Investigate effect the different layouts on the ability to compare
- Study:
  - make a quick visual judgment
  - judge the percentage difference of two highlighted bars
  - 3 seconds for each task

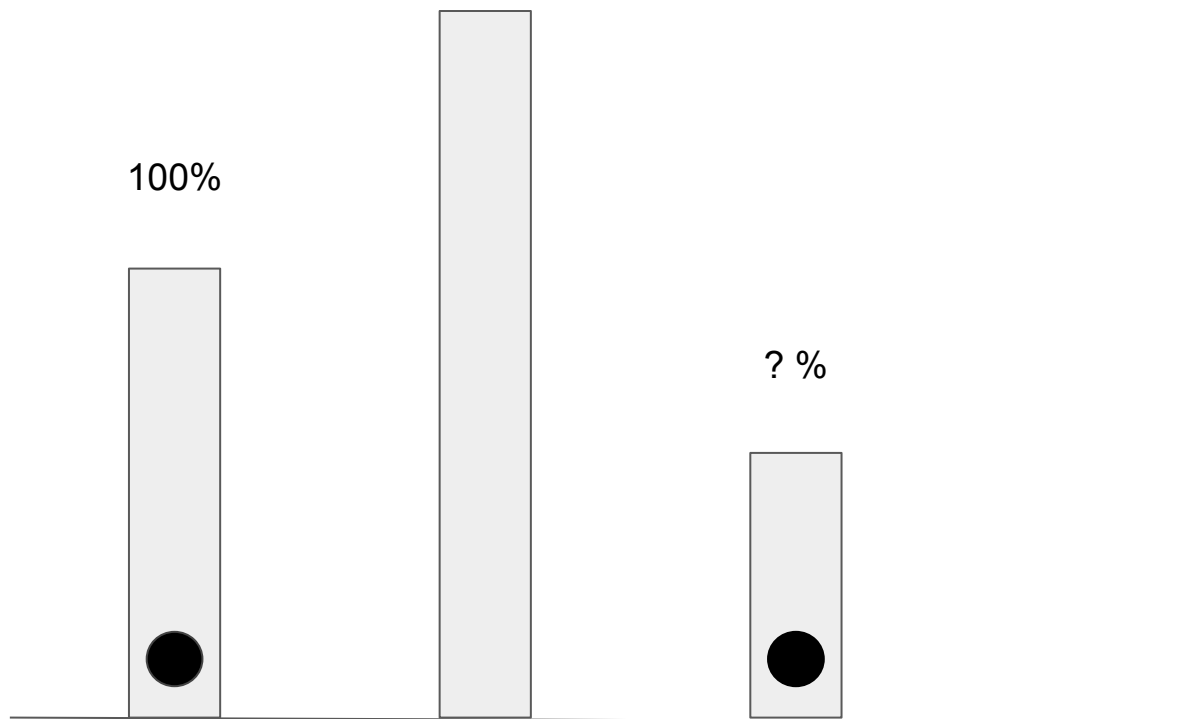
T1

- 5
- 10
- 15
- 20
- 25
- 30
- 35
- 40
- 45
- 50
- 55
- 60
- 65
- 70
- 75
- 80
- 85
- 90
- 95

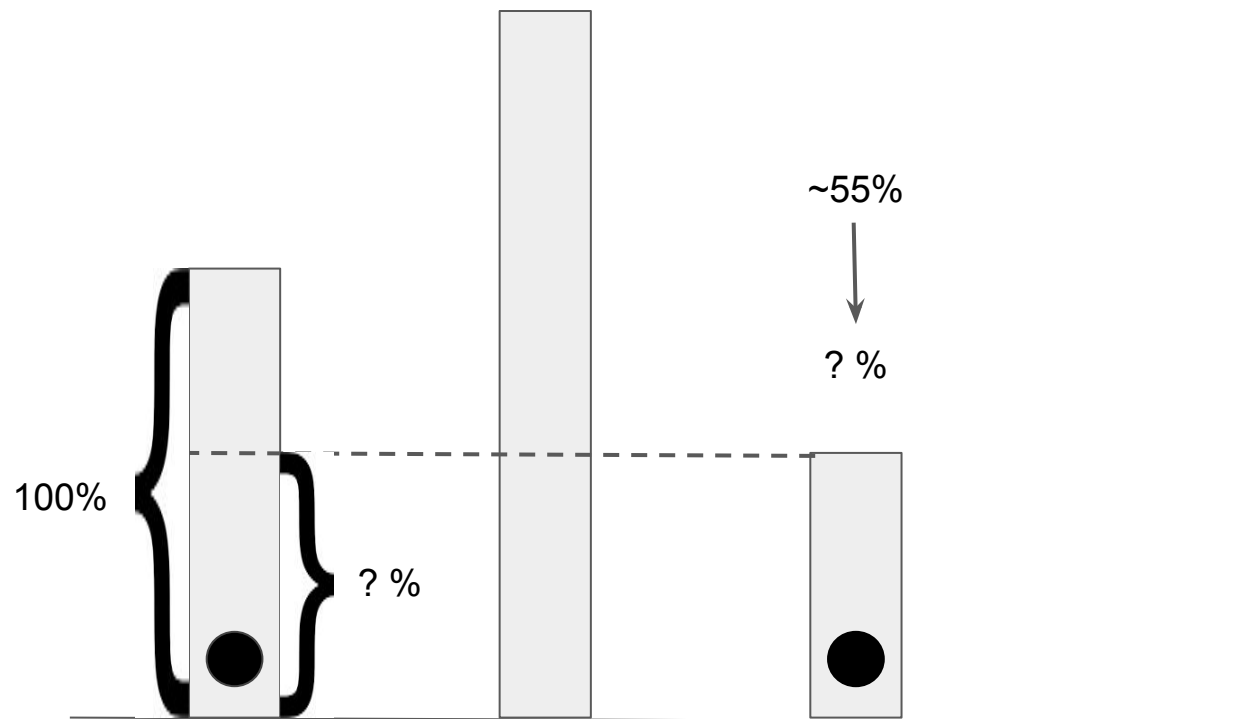
See results

vote

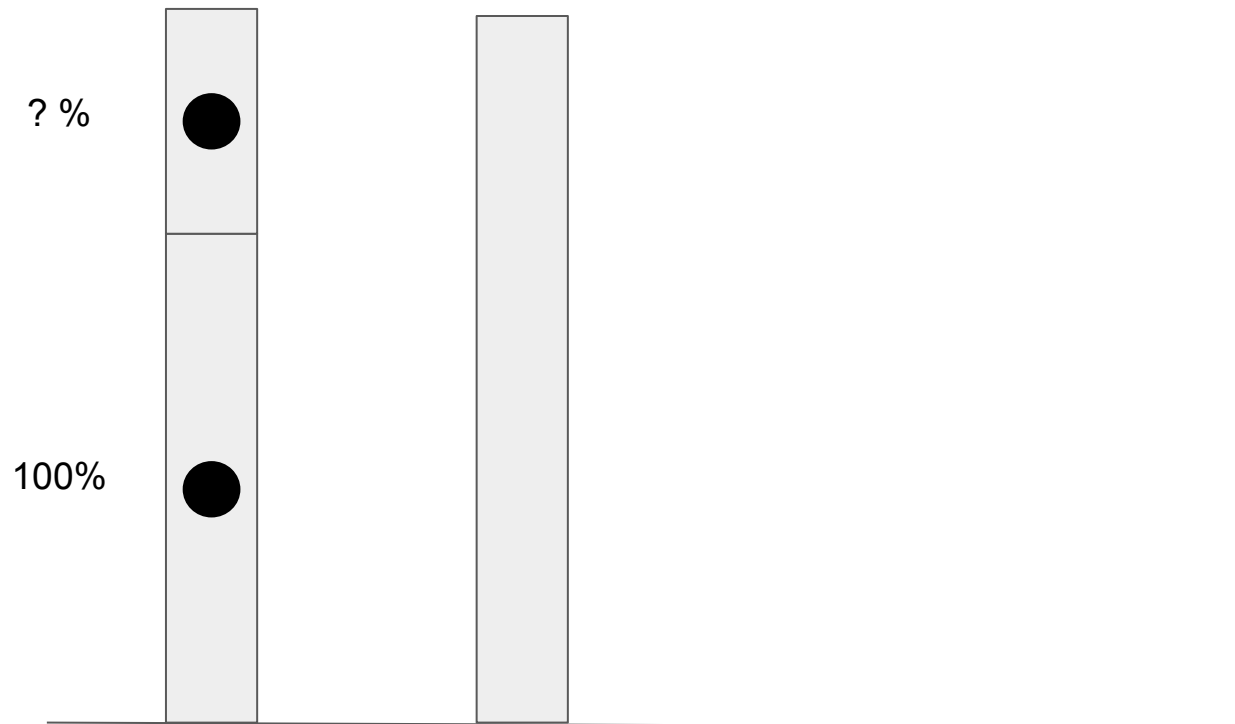
# Task layout



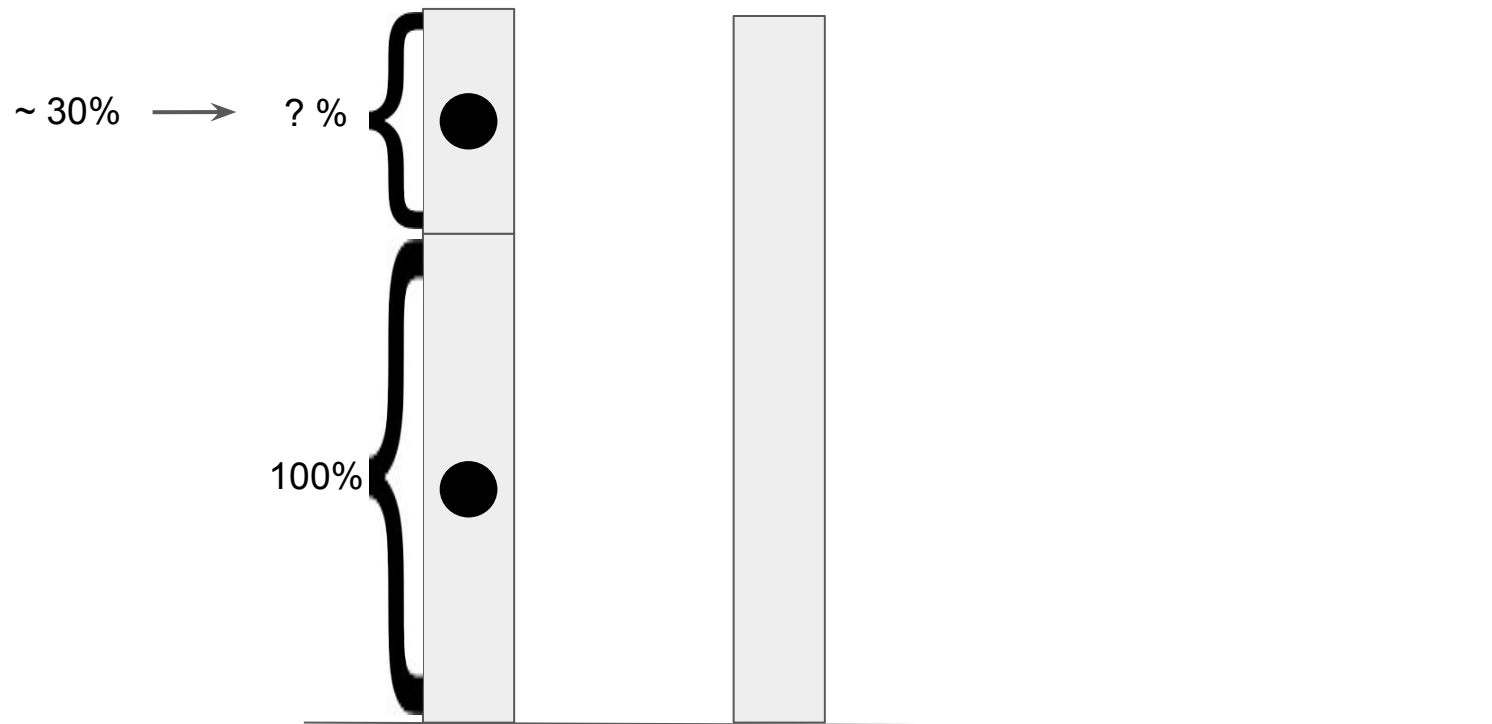
# Task layout



# Task layout



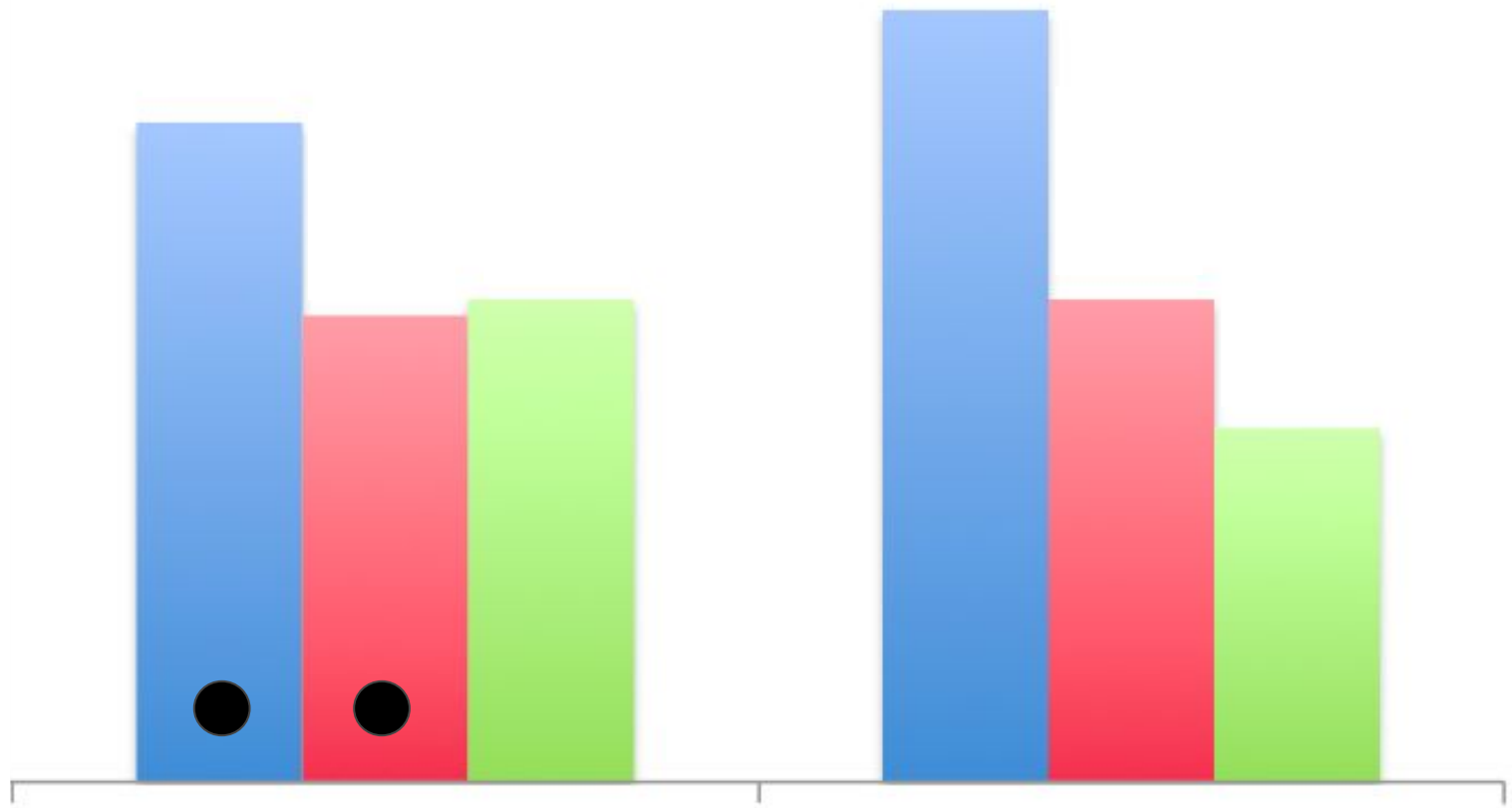
# Task layout







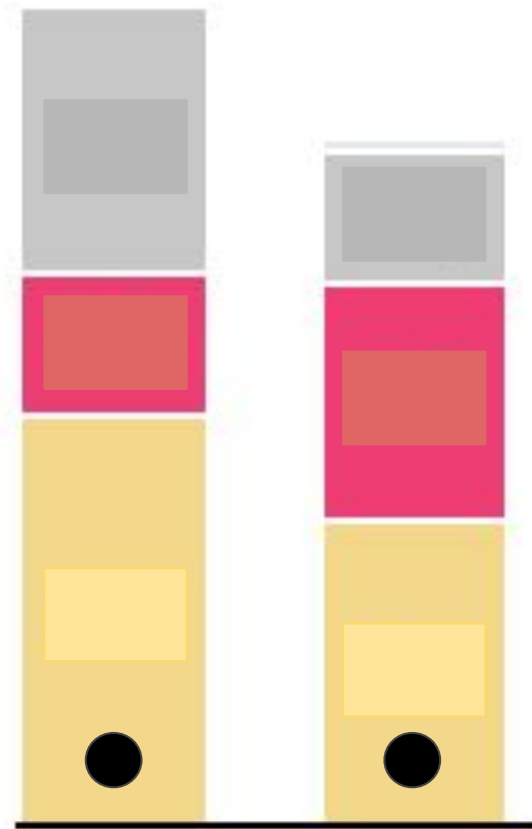
# T1







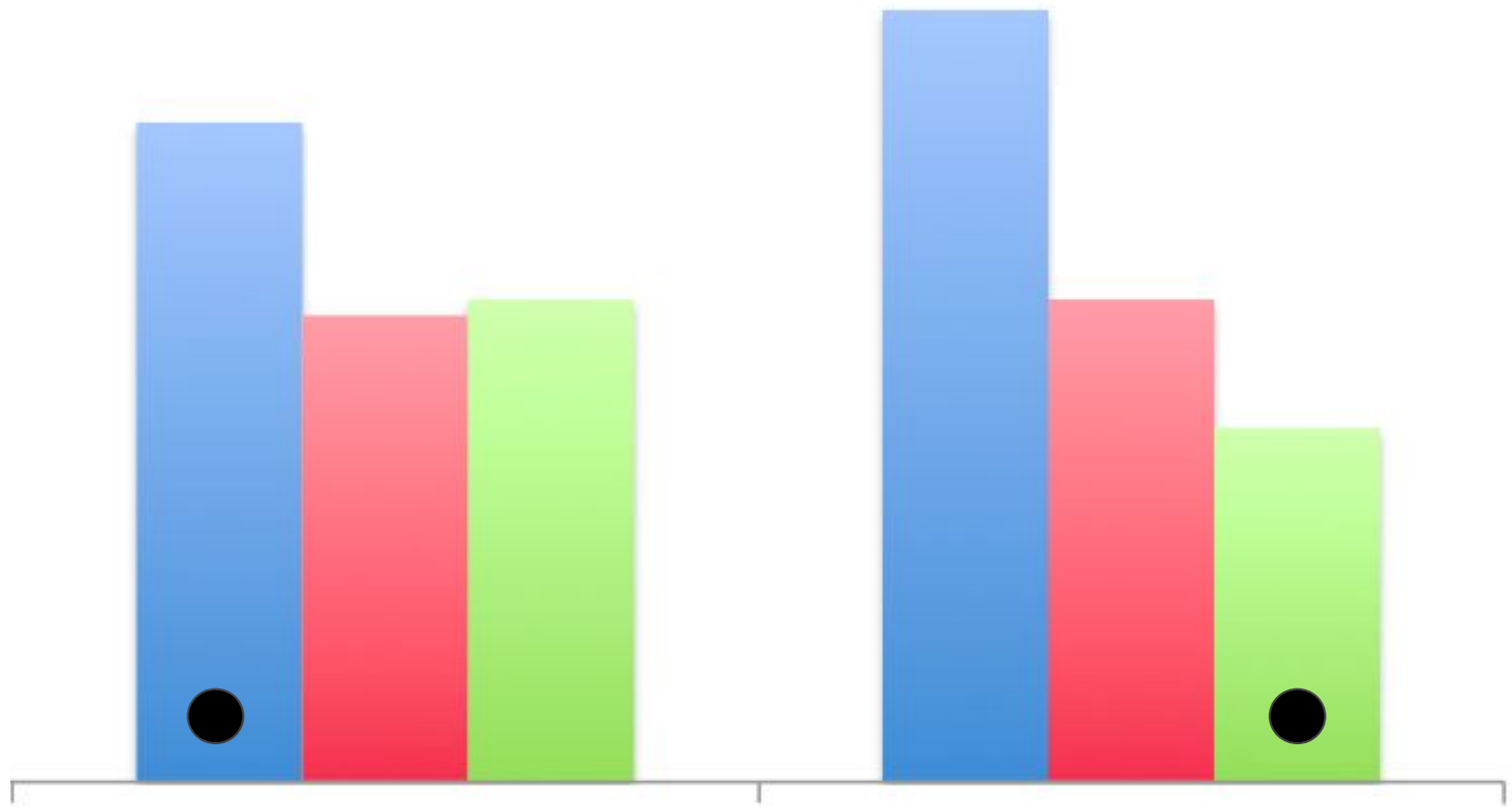
# T2







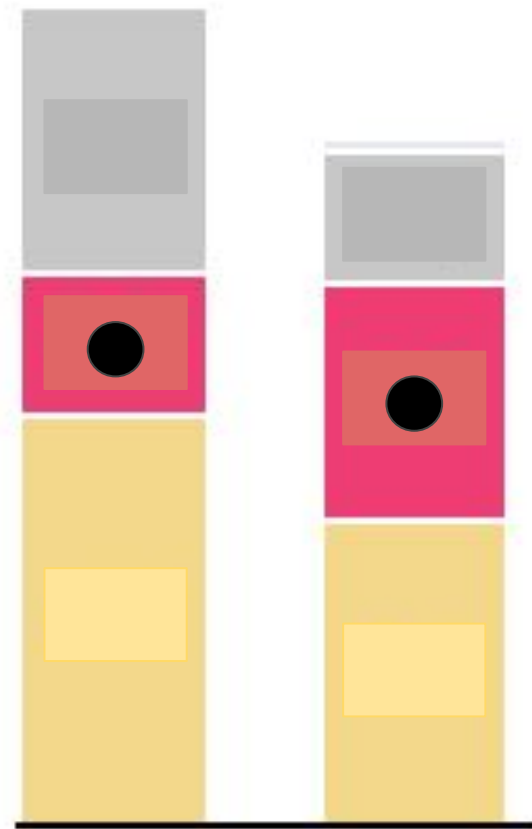
# T3







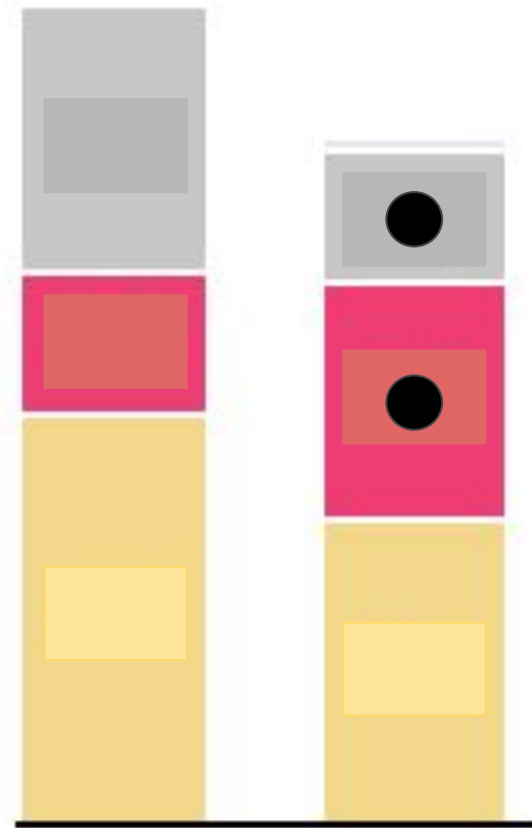
# T4







# T5



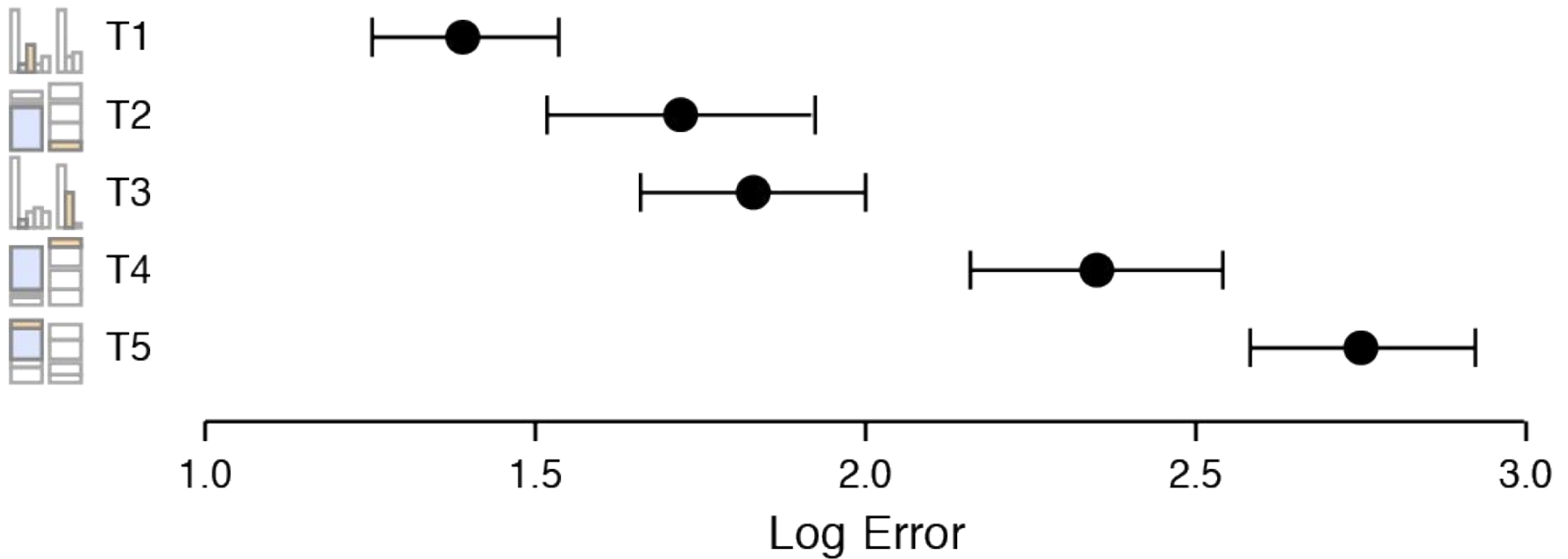


# Discussion

- Gather impressions
  - What made a task particularly easy or difficult?
  - What do you expect the results will be?

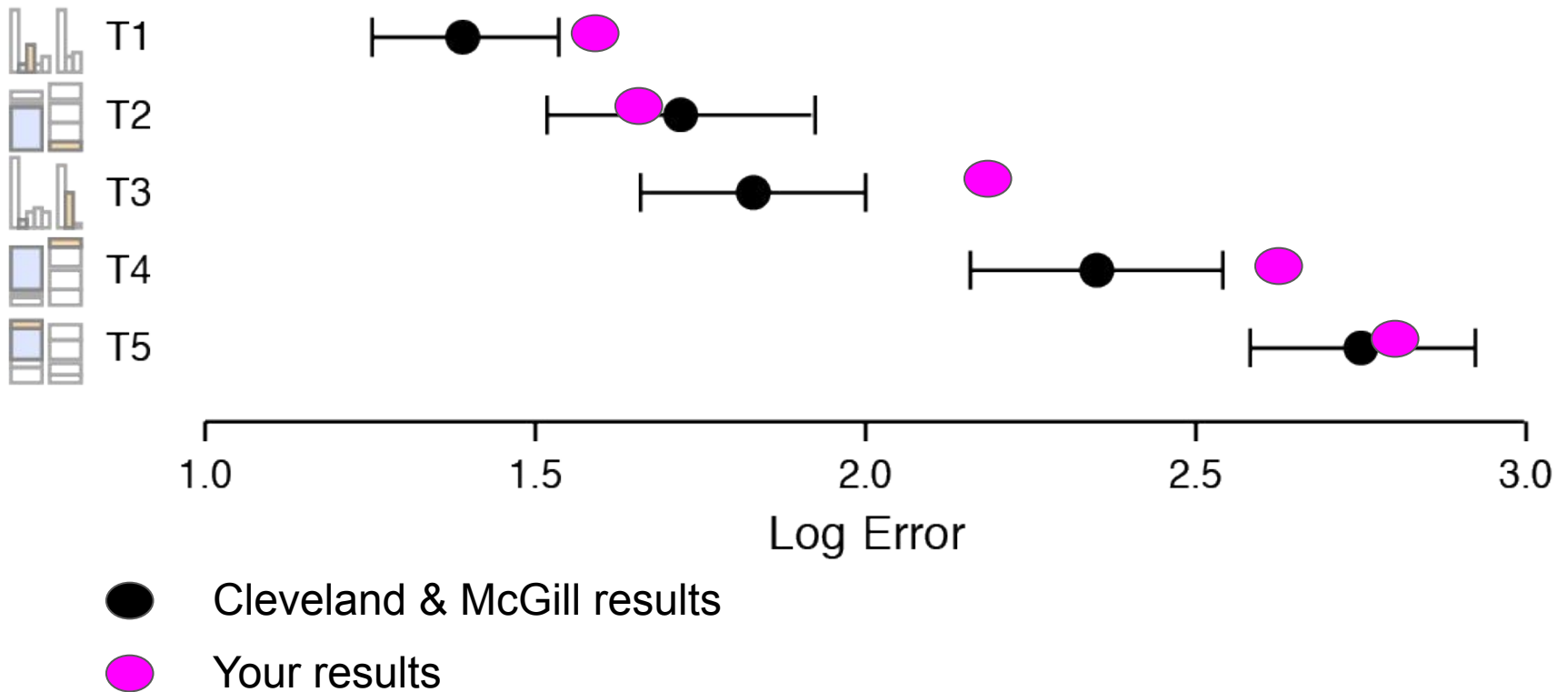
# Results

## Cleveland & McGill's Results



# Results

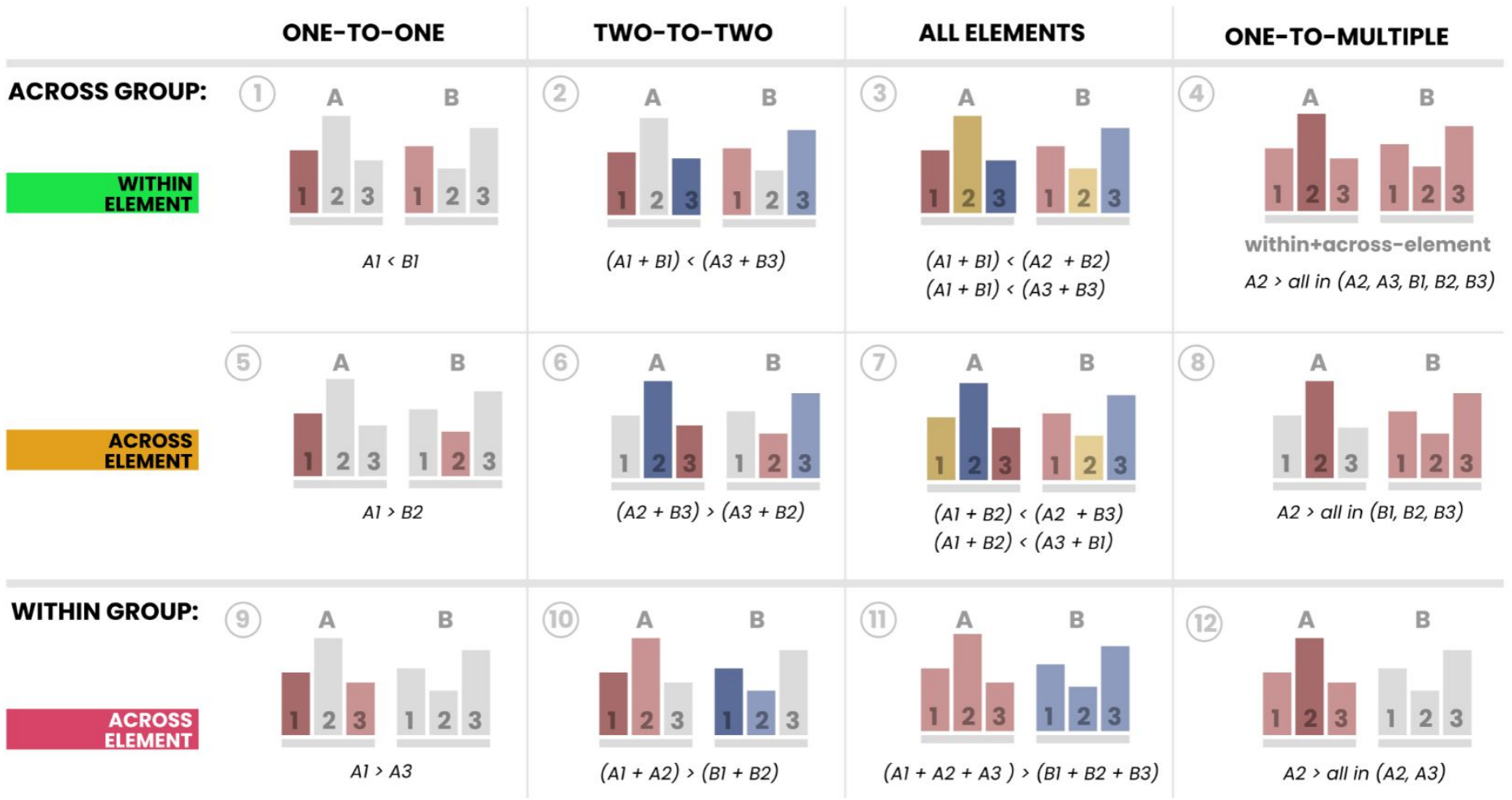
## Cleveland & McGill's Results




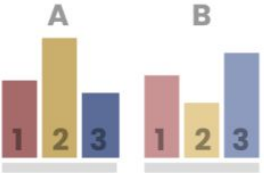


# Results summary

- Comparing by using **aligned position channel** against a common scale is the most accurately perceived visual channel.
- Comparing by using **unaligned position channel** adds mental overhead of due to different start and the end of each bar → reduced accuracy.

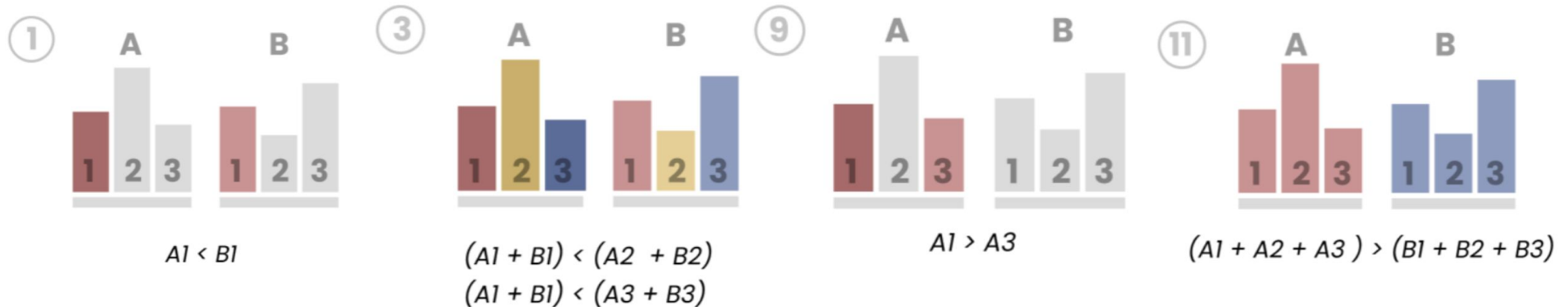
# Apply what you just learned



# Apply what you just learned

	ONE-TO-ONE	TWO-TO-TWO	ALL ELEMENTS	ONE-TO-MULTIPLE
<b>ACROSS GROUP:</b> <span style="background-color: #00FF00; padding: 2px;">WITHIN ELEMENT</span>	①  $A1 < B1$		③  $(A1 + B1) < (A2 + B2)$ $(A1 + B1) < (A3 + B3)$	
<span style="background-color: #FFA500; padding: 2px;">ACROSS ELEMENT</span>				
<b>WITHIN GROUP:</b> <span style="background-color: #DC143C; padding: 2px;">ACROSS ELEMENT</span>	⑨  $A1 > A3$		⑪  $(A1 + A2 + A3) > (B1 + B2 + B3)$	

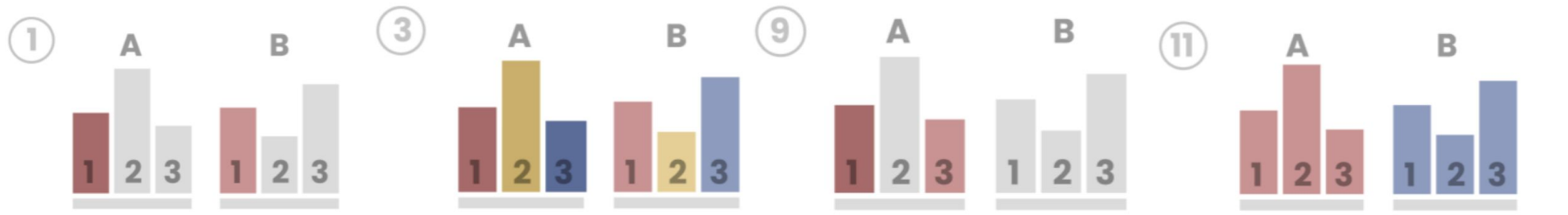
# Apply what you just learned



A & B Companies | 1,2,3 Continents | Bar height: revenue

1. Comparing revenue between companies A & B in continent 1.
2. Comparing total revenue of A & B in given continent.
3. Comparing revenue of company A among different continents.
4. Comparing the global revenue of company A to B

# Apply what you just learned

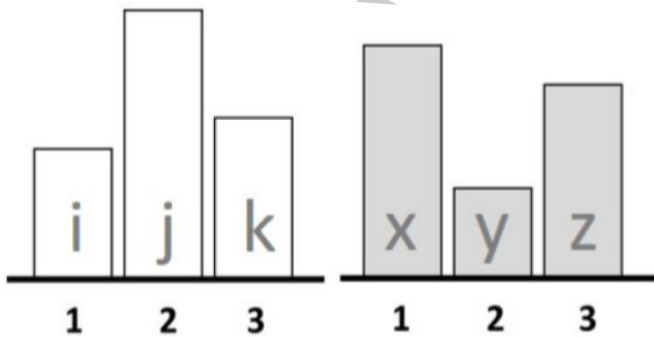


$A_1 < B_1$

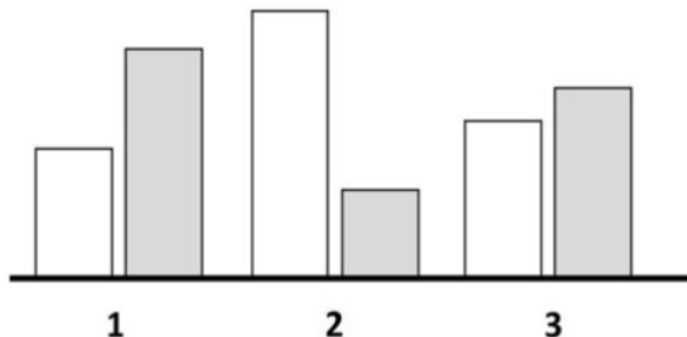
$(A_1 + B_1) < (A_2 + B_2)$   
 $(A_1 + B_1) < (A_3 + B_3)$

$A_1 > A_3$

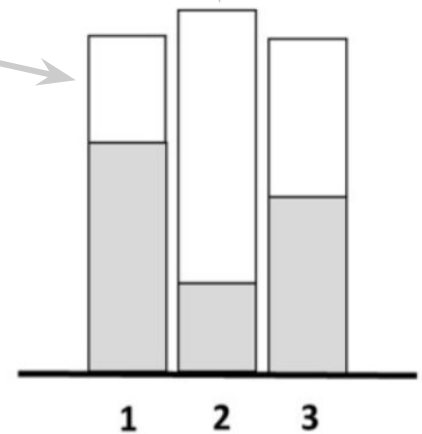
$(A_1 + A_2 + A_3) > (B_1 + B_2 + B_3)$



Adjacent



Dimensional stacking



Stacked