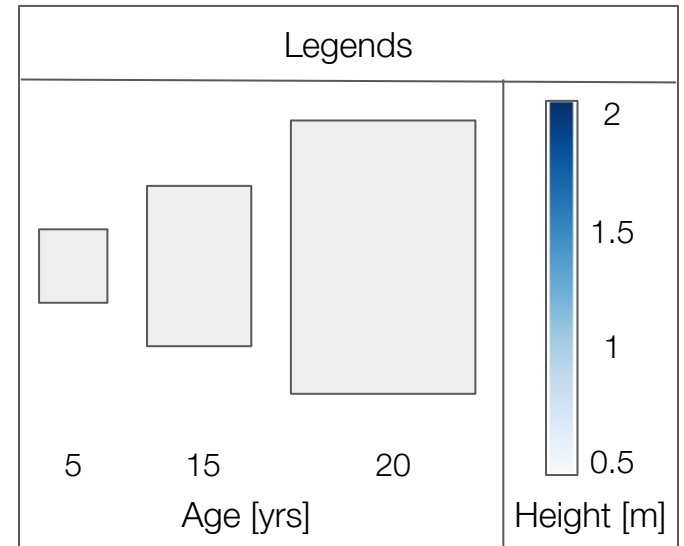
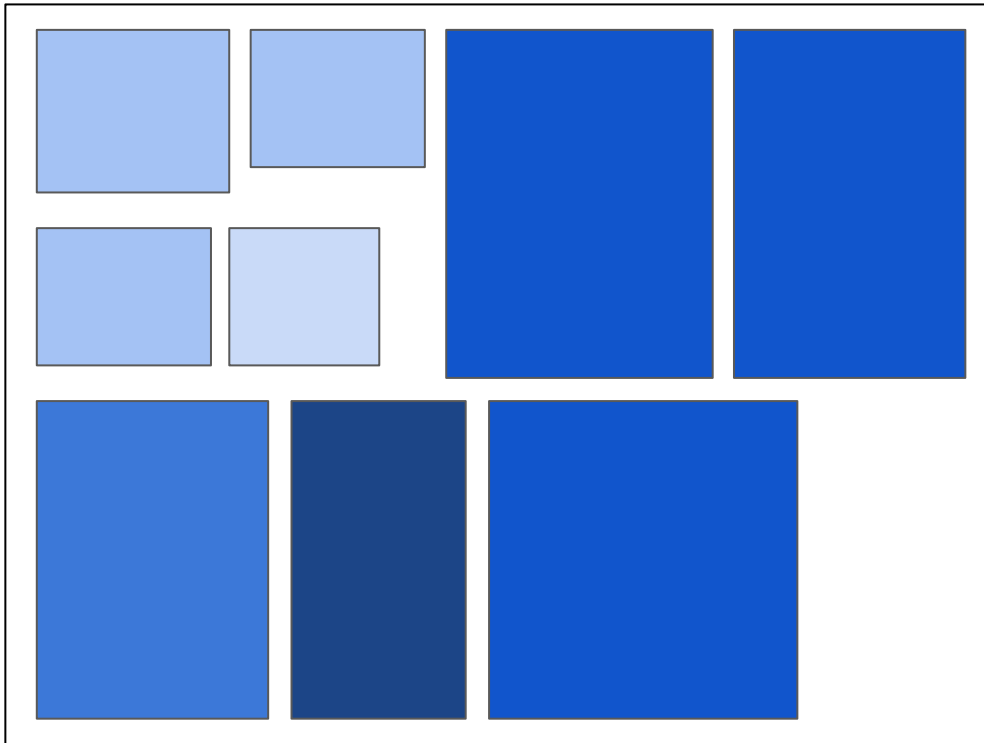


# Arranging space - I

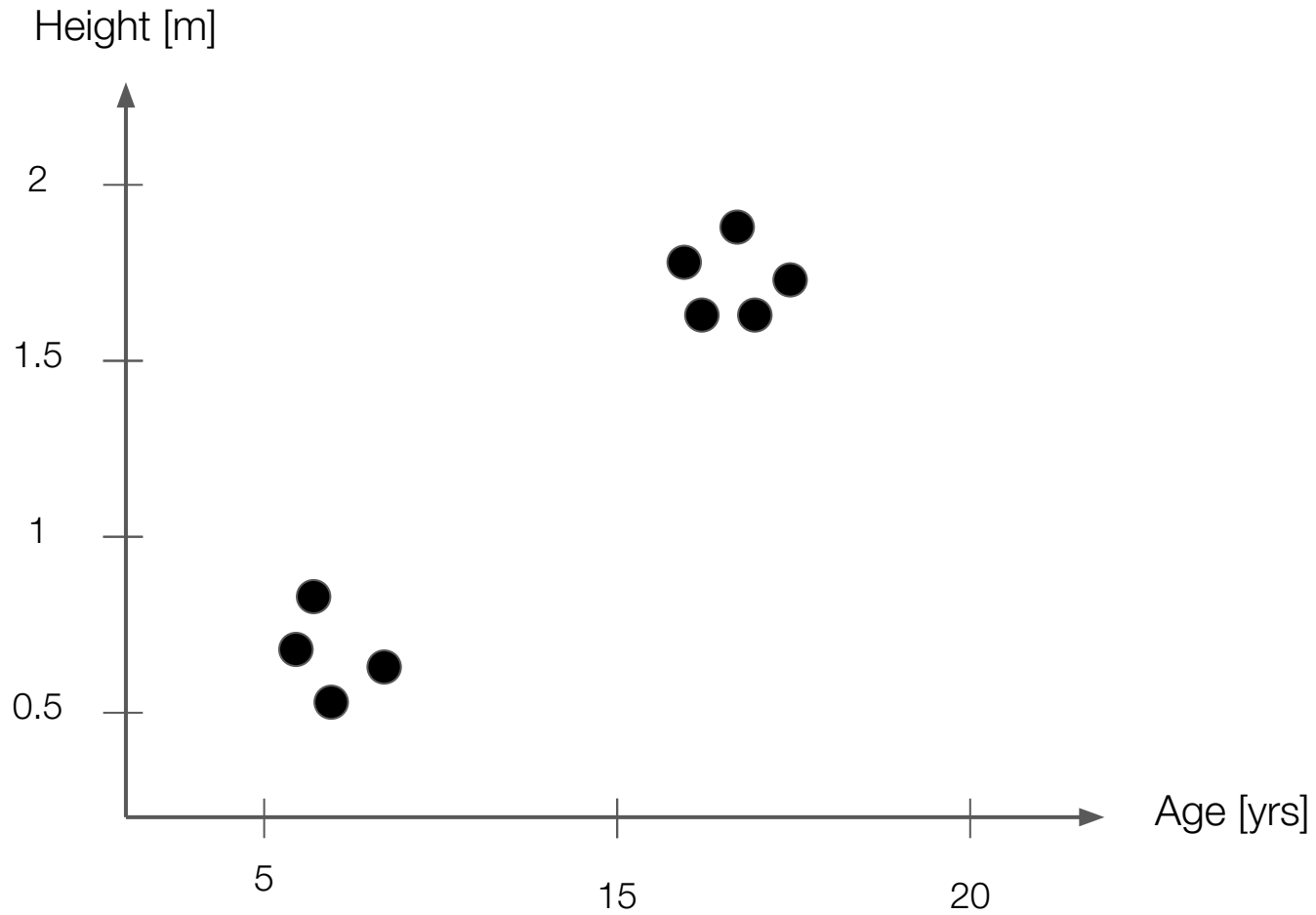
Visualization  
Sebastian Ratzenböck

What is the “best” visualization  
for the following data?

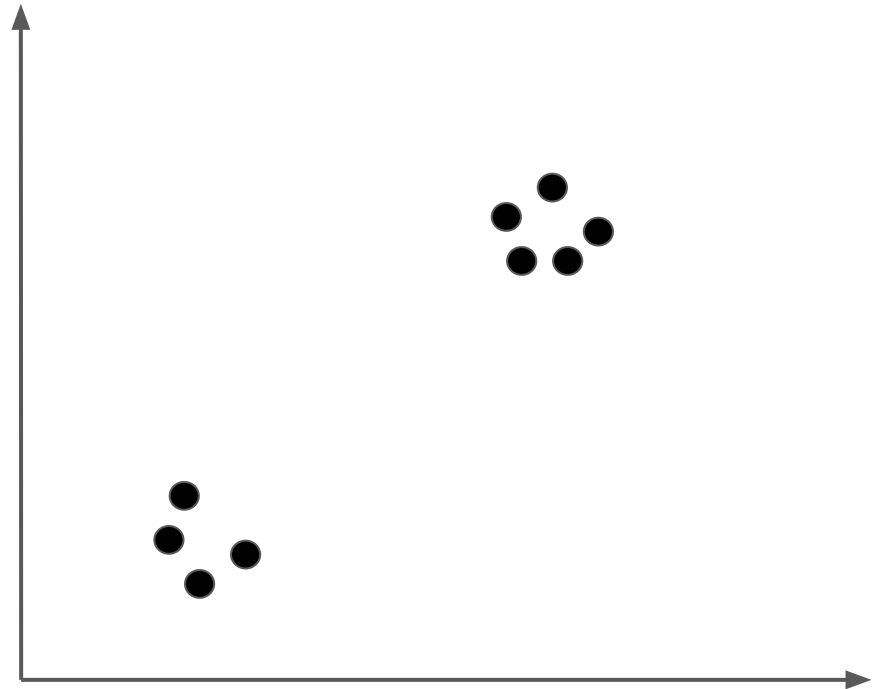
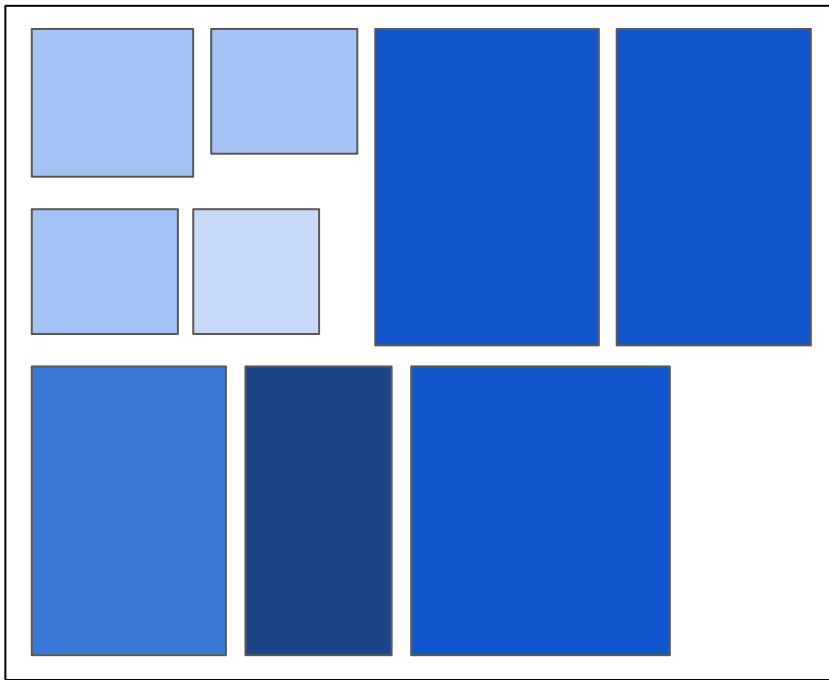
# VIS 1



# VIS 2



# VIS 1 or 2?

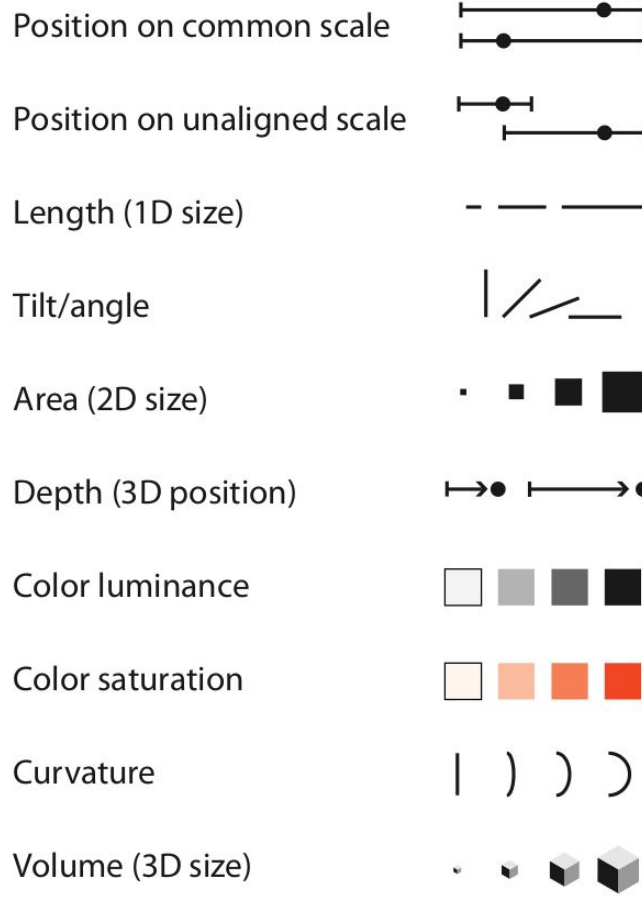


Why?

# Recap: effectiveness ranking

## Channels

### ➔ Magnitude Channels: Ordered Attributes



### ➔ Identity Channels: Categorical Attributes

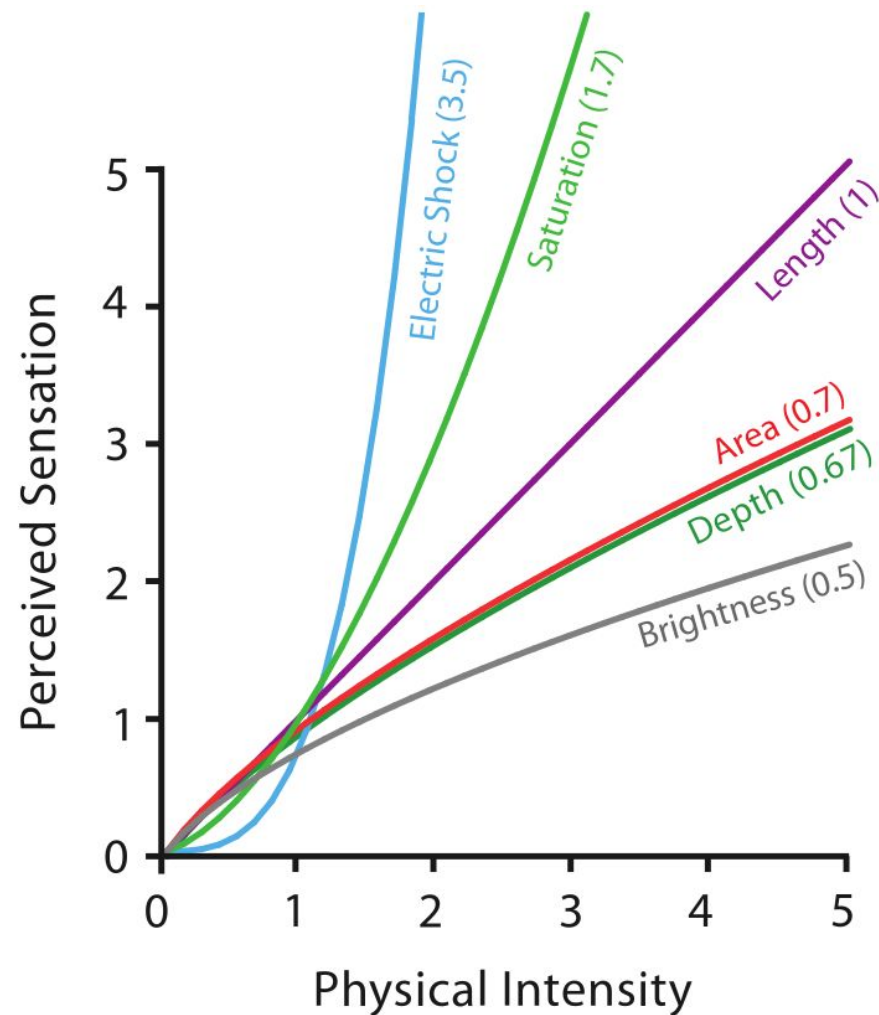


# Recap: effectiveness ranking

- How do we determine the ranking?
  - Accuracy
  - Discriminability
  - Separability
  - Popout



# Recap: Steven's psychophysical law



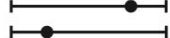
What do we mean  
by arranging space?

# Arranging space

- Spatial channels for visual encoding
- Most effective encoding choice

## ➔ Magnitude Channels: Ordered Attributes

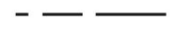
Position on common scale



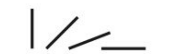
Position on unaligned scale



Length (1D size)



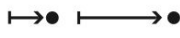
Tilt/angle



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

# Arranging space

- Spatial channels for visual encoding
- Most effective encoding choice  $\leftrightarrow$  tied to effectiveness principle

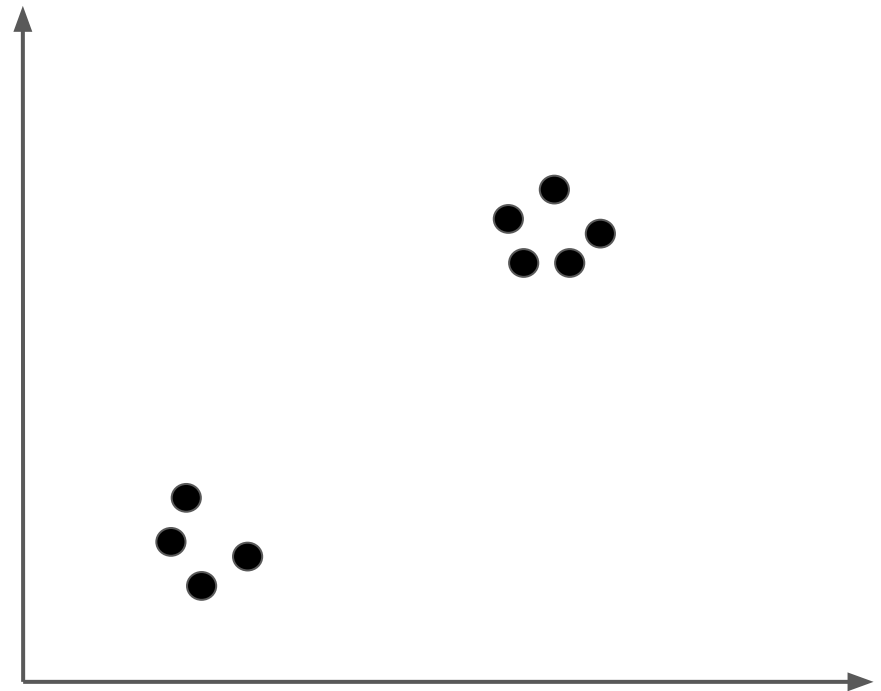
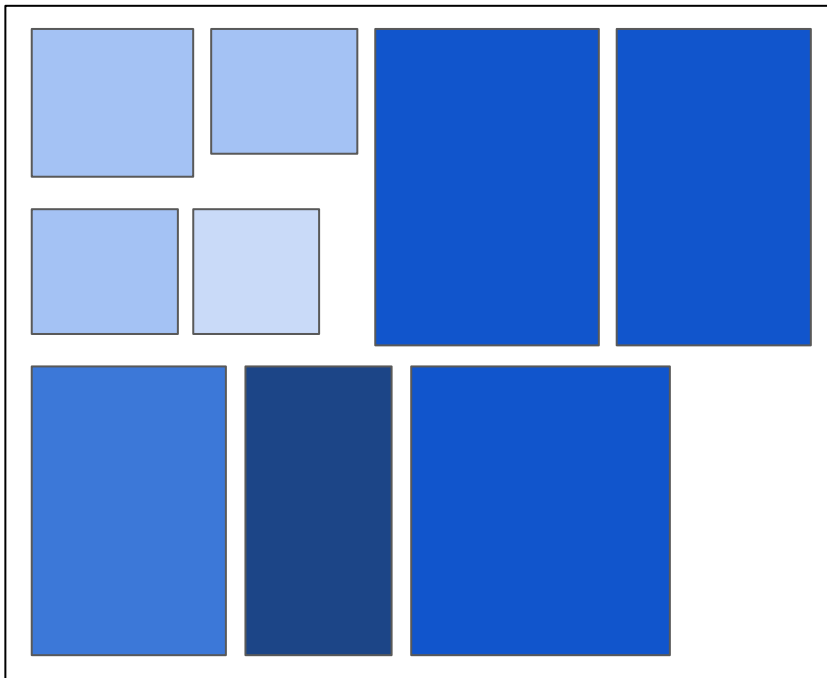
## **Effectiveness principle**

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

# 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.

# What's more “natural”?



# Readings

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

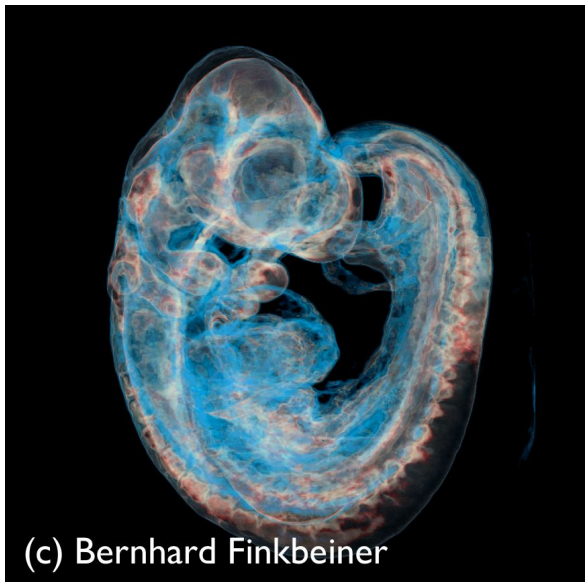
# Overview

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



# Spatial attributes

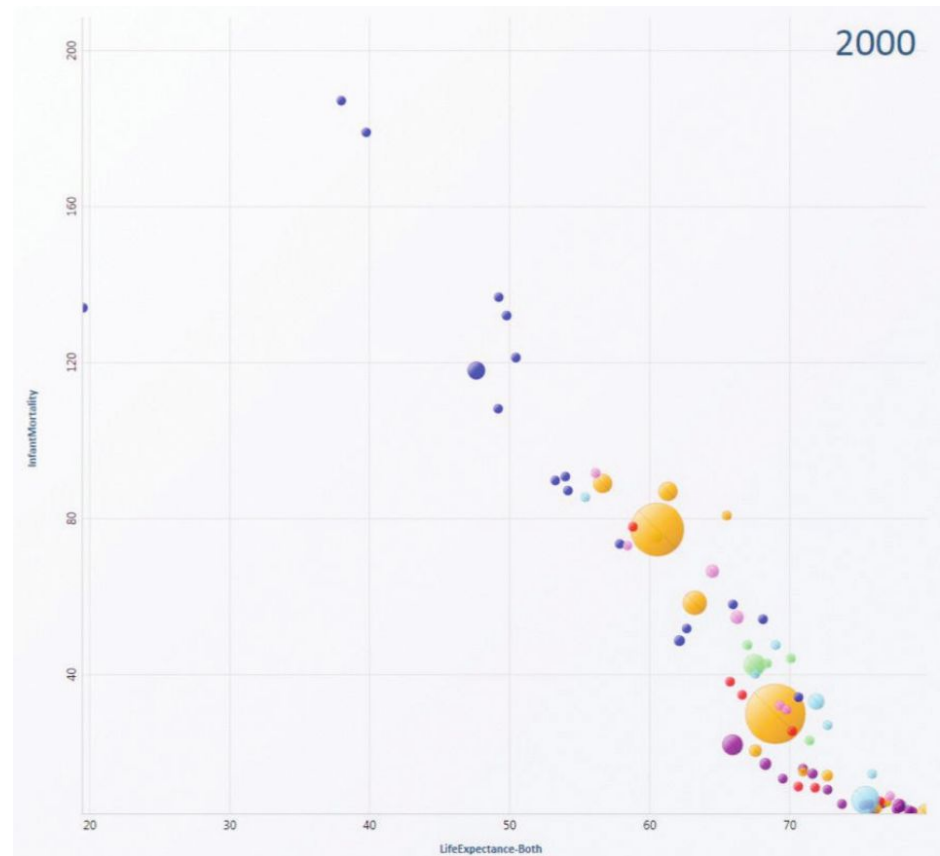
- 1D - line graphs
- 2D - cartographic views
- 3D - volumes
- scalar vs. vector vs. tensor data



# Quantitative vs. categorical data

# Quantitative vs. categorical

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

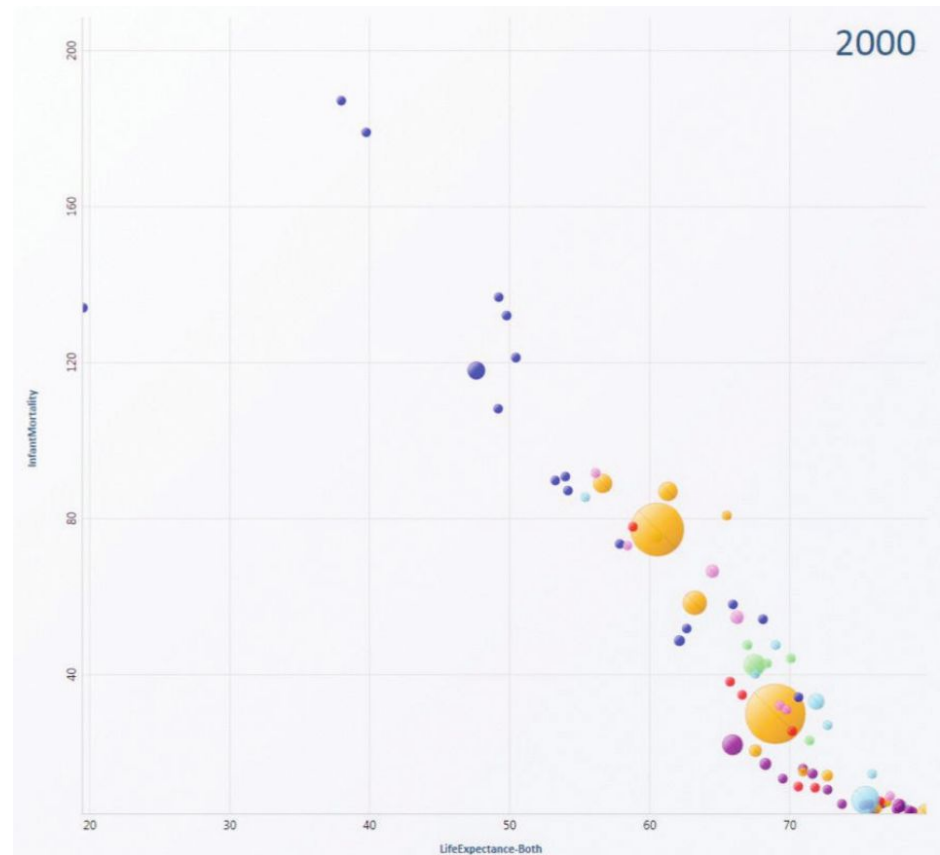


# Quantitative vs. categorical

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



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



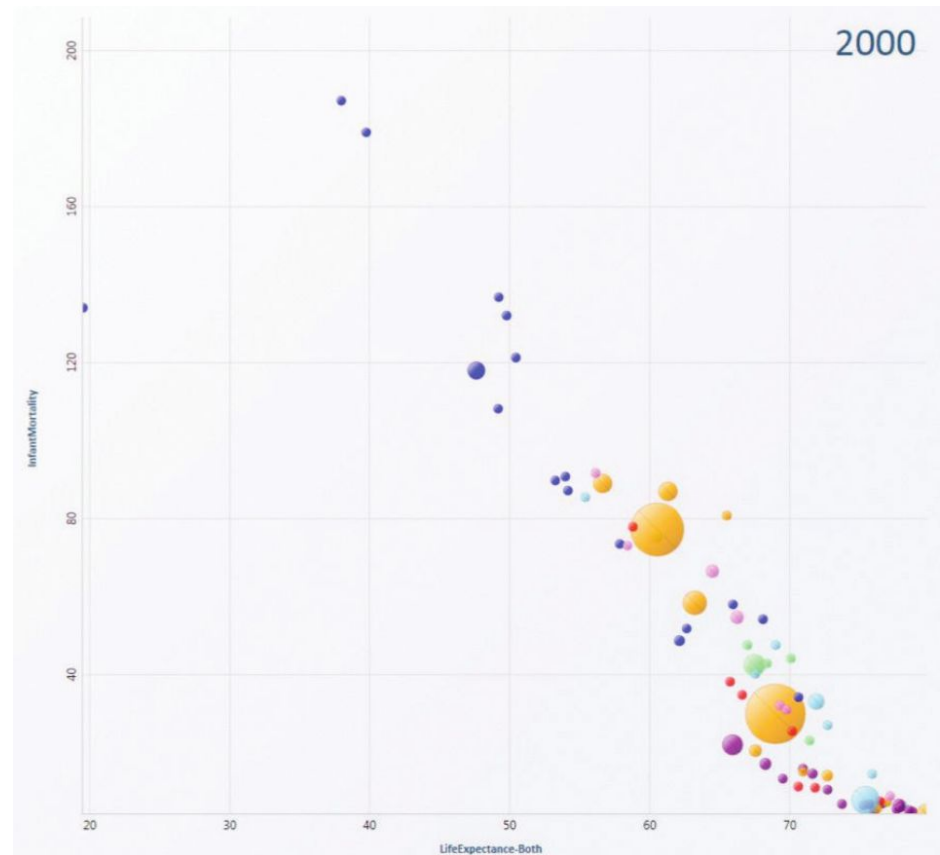
# Quantitative vs. categorical

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



## Attributes

- Life expectancy
- Infant mortality
- Country size
- Continent



How to use the spatial  
channel for **categorical**  
data?


# Quantitative vs. categorical

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

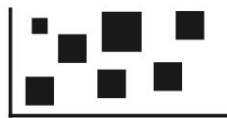
## **Principle of expressiveness**

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

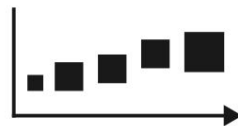
# Quantitative vs. categorical

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

→ Separate



→ Order



→ Align

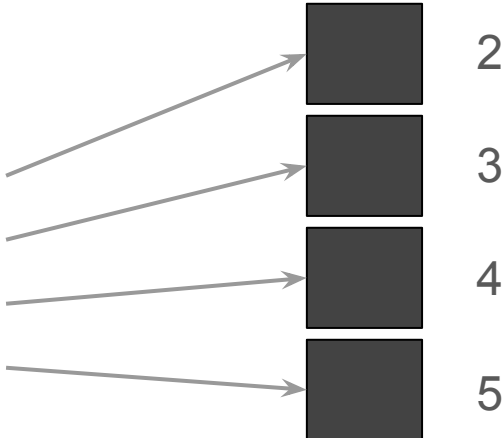




# Separate 2D space into regions

- 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



A diagram showing four arrows pointing from the rows of the table to a vertical stack of four dark gray rectangular regions. The regions are labeled 2, 3, 4, and 5 on the right side. The arrows originate from the rows for Kevin (15), Lucy (24), Marc (11), and Maria (13) respectively.

→ Separate



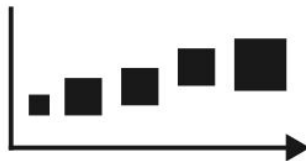
# Quantitative vs. categorical

- 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
3	Lucy	24
4	Marc	11
5	Maria	13



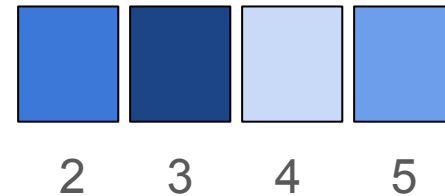
→ Arrange region into 1D list alignment

→ Now add channel

# 1D keys

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

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



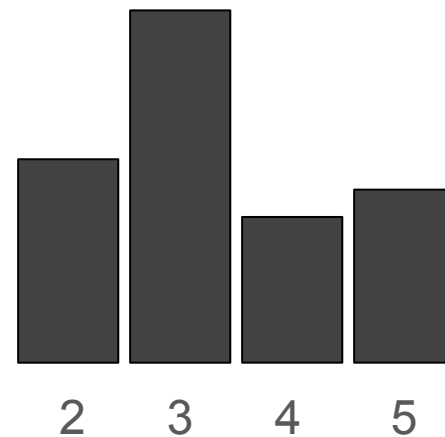
→ Arrange region into 1D list alignment

→ Now add channel: Color

# 1D keys

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

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

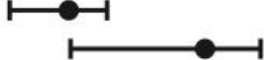


→ Arrange region into 1D list alignment

→ Now add channel: Length

## ➔ Magnitude Channels: Ordered Attributes

Position on common scale 

Position on unaligned scale 

Length (1D size) 

Tilt/angle 

Area (2D size) 

Depth (3D position) 

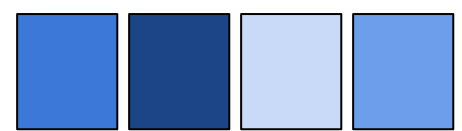
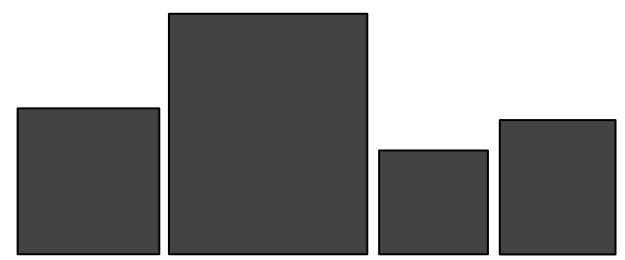
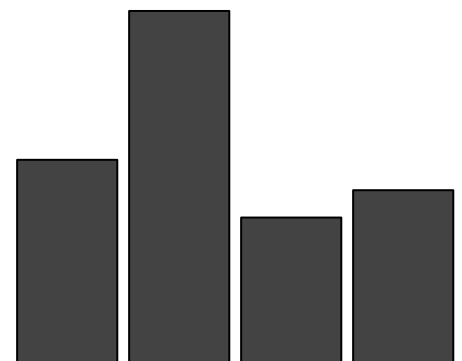
Color luminance 

Color saturation 

Curvature 

Volume (3D size) 

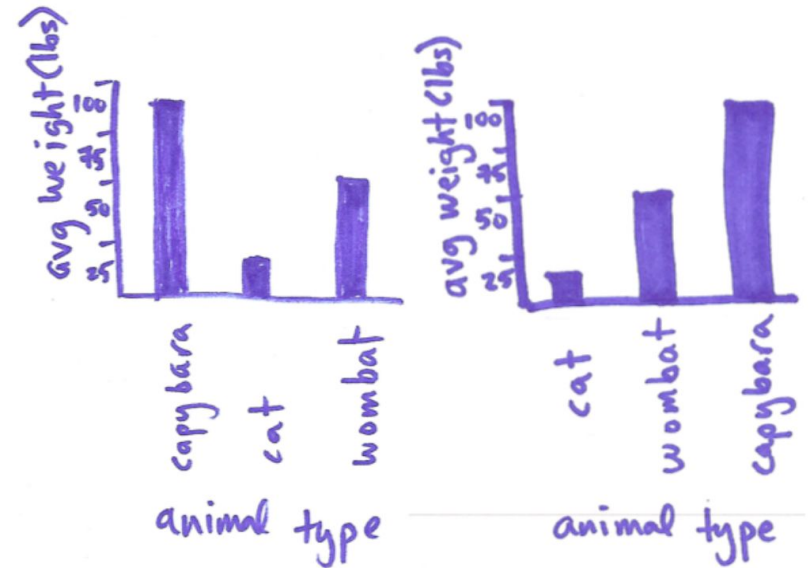
Most  
Effectiveness  
Least



# 1D keys

- categorical: bar charts
  - aligned / ordered

**Bar chart**  
Line marks  
Length channel

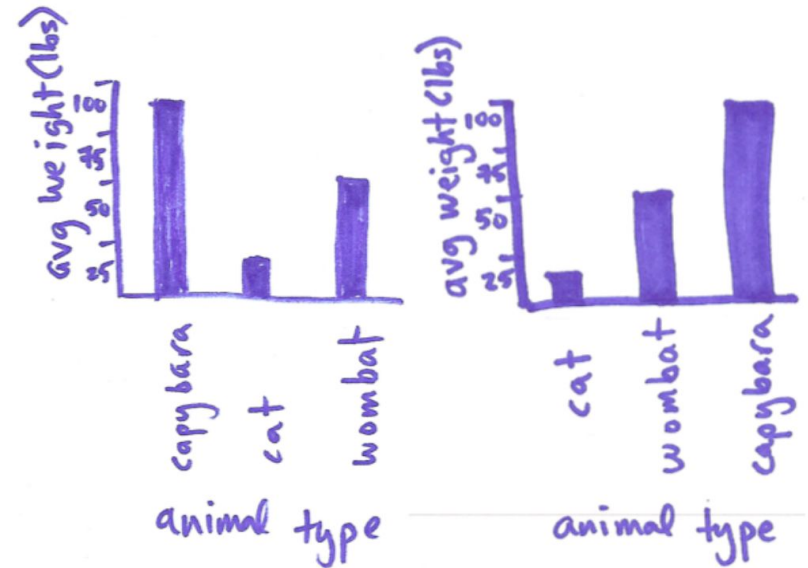


?

?

# 1D keys

- categorical: bar charts
  - aligned / ordered



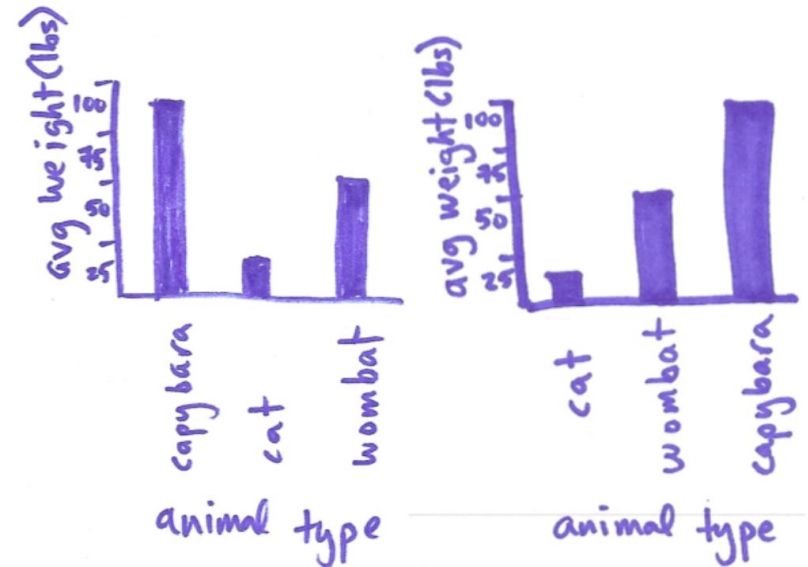
Lookup value

Spot trends

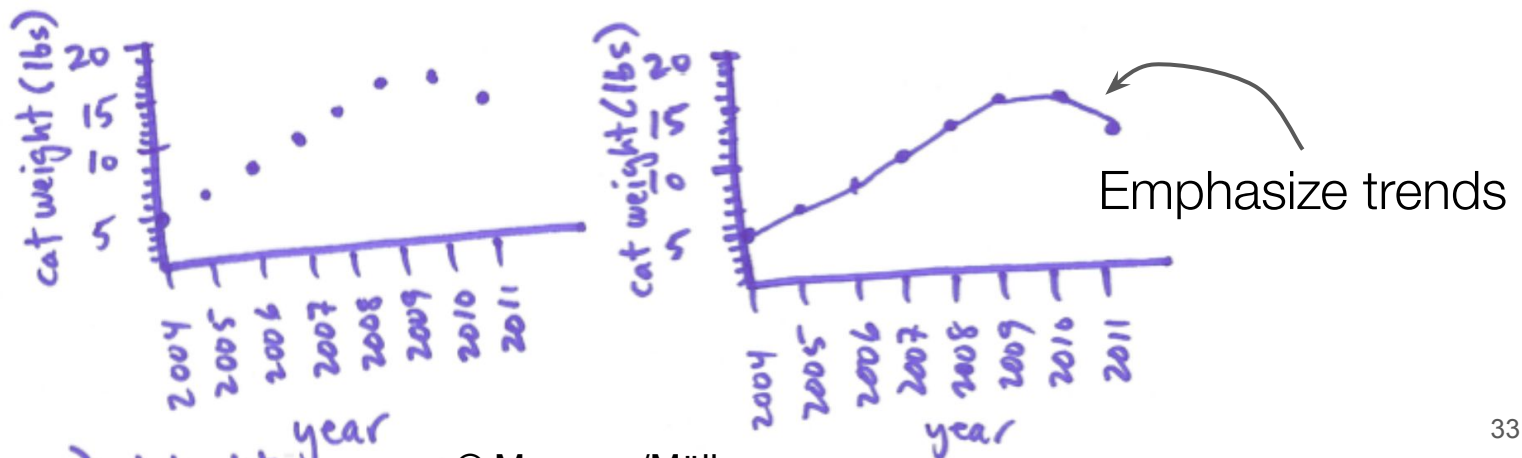


# 1D keys

- categorical: bar charts
  - aligned / ordered



- quantitative/ ordered: dot plot / line chart

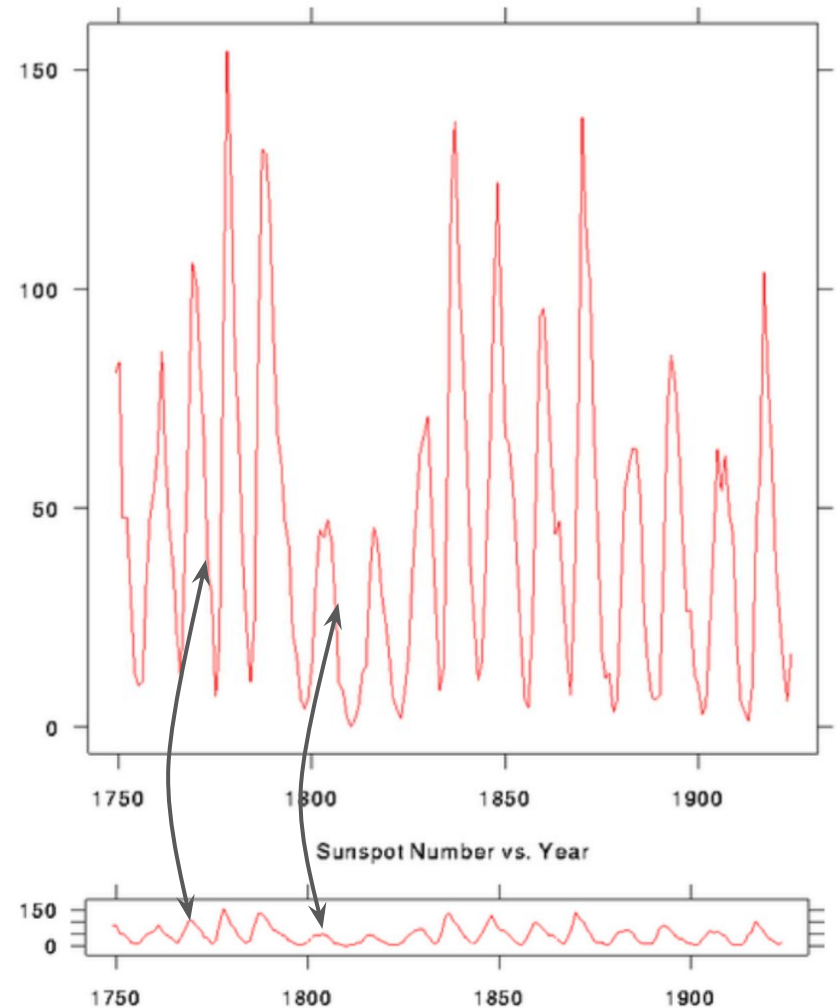


# Line graphs

1. Aspect ratio
2. Expressiveness principle

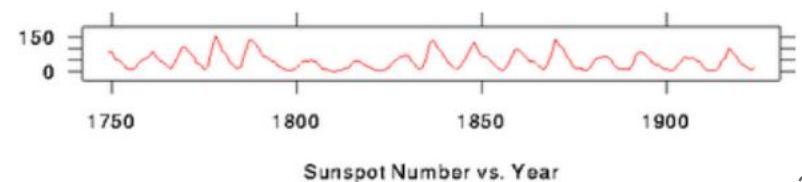
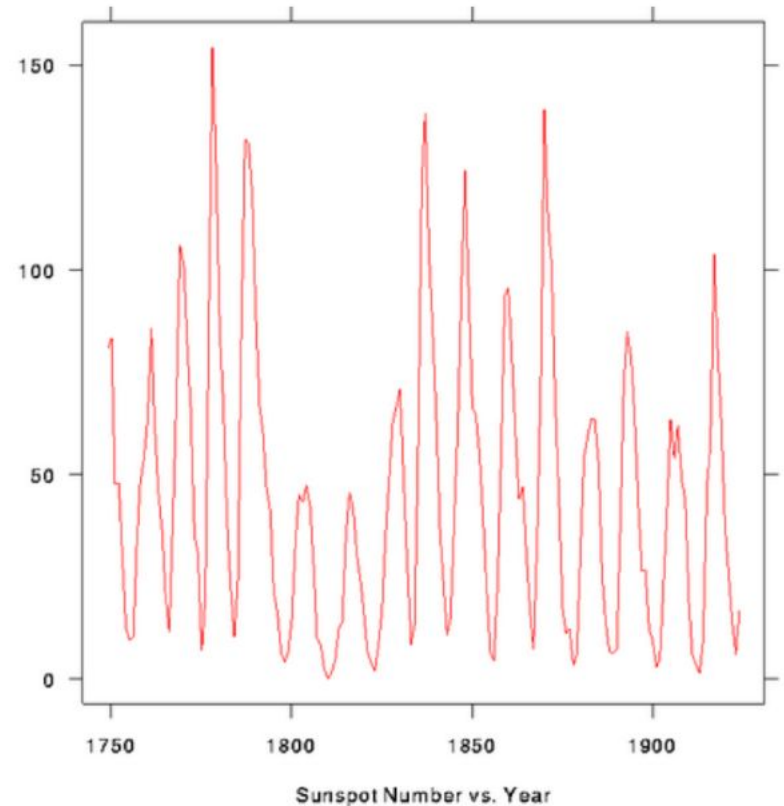
# Line plots: aspect ratio

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



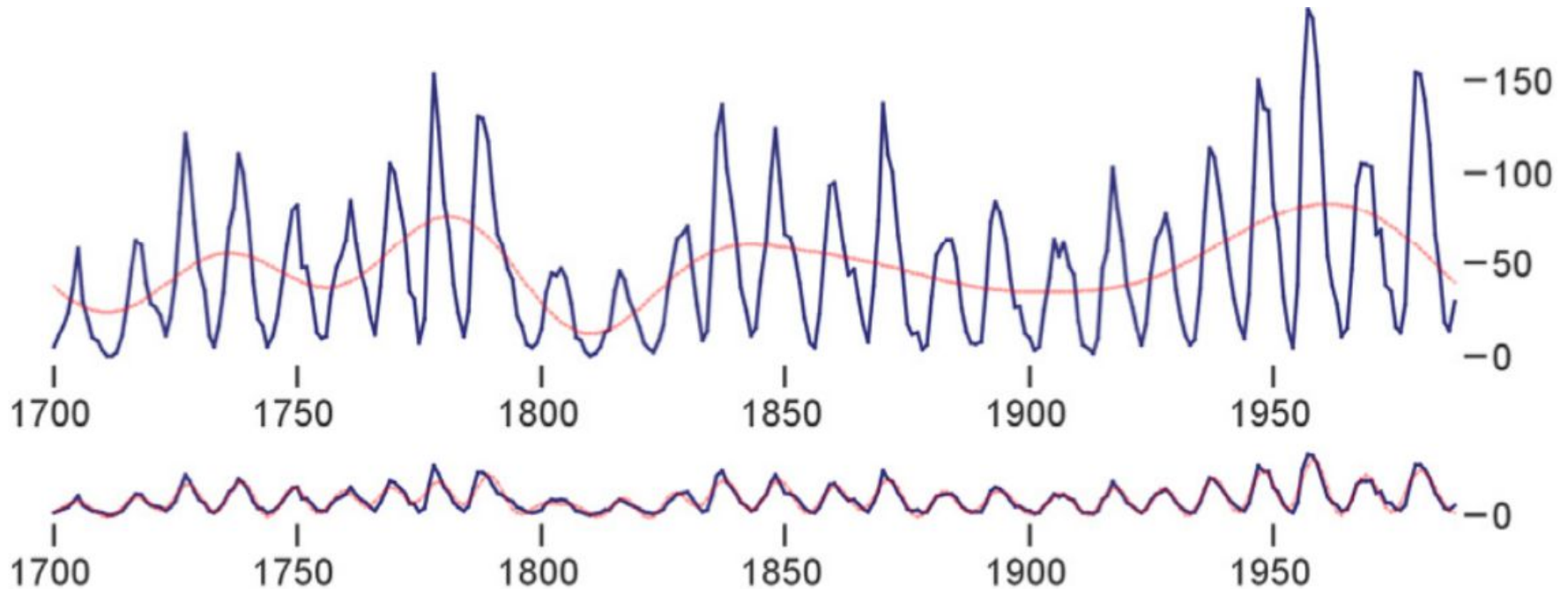
# Banking to 45 degrees

- 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

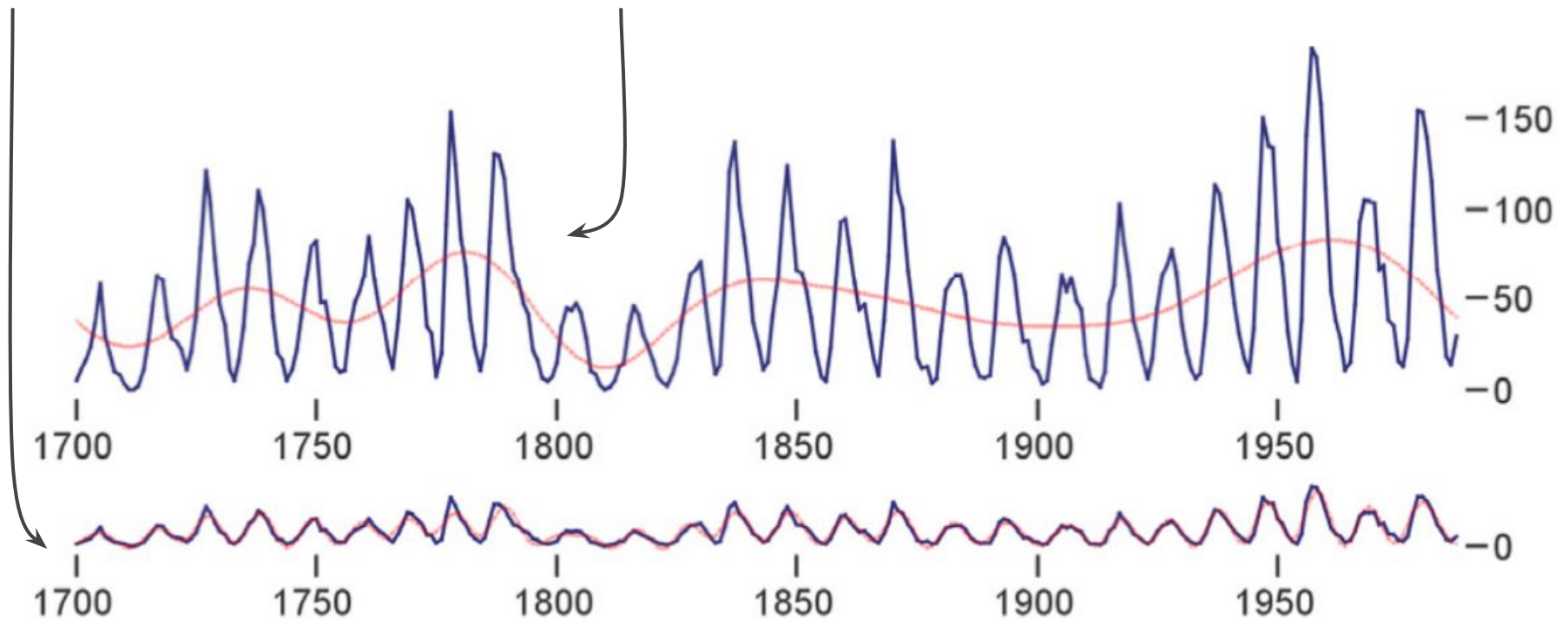
What is the difference between the two ARs?



# Banking to 45 degrees

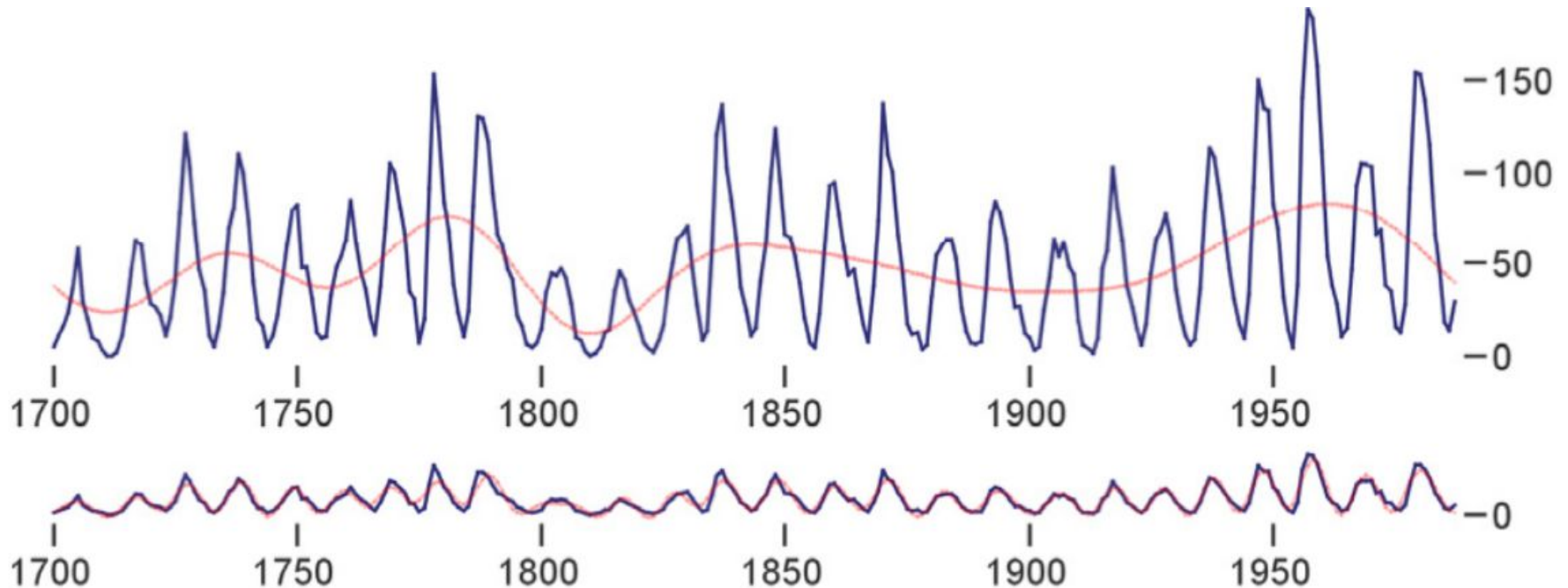
Fast  
changes

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.



# Quantitative vs. categorical

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

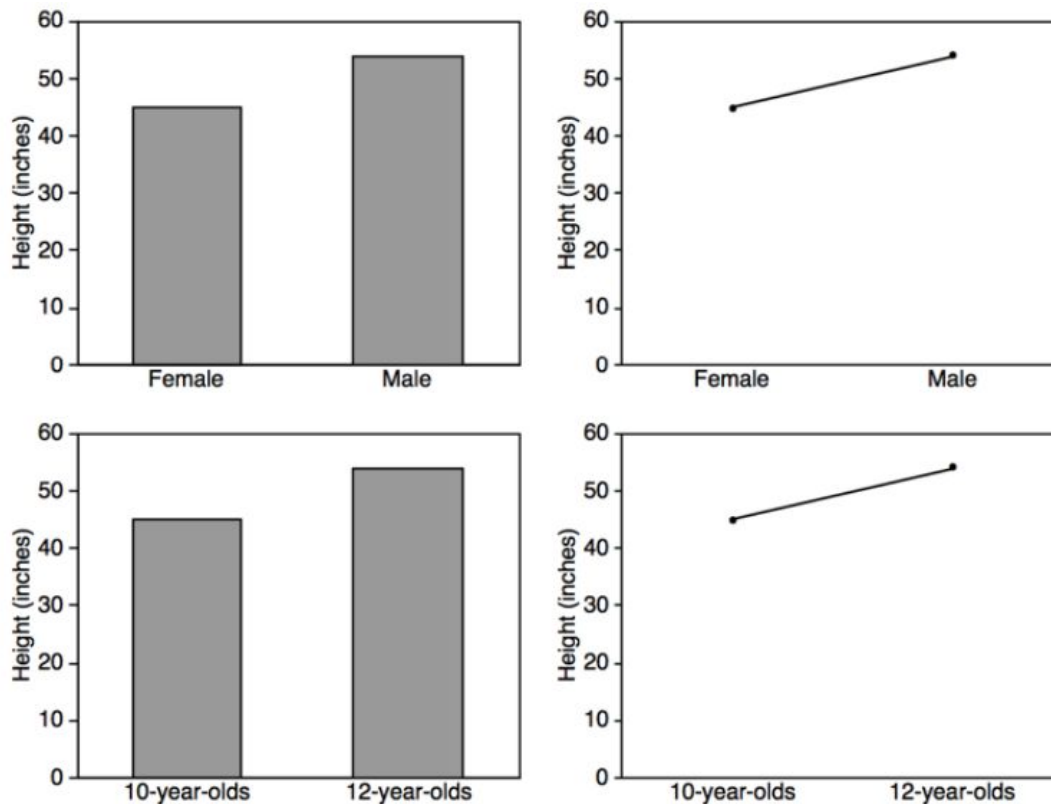
## **Principle of expressiveness**

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

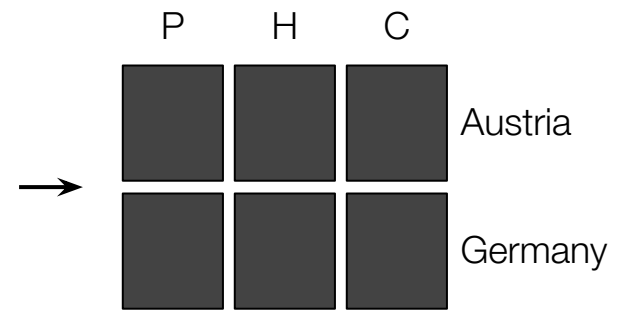


2D keys

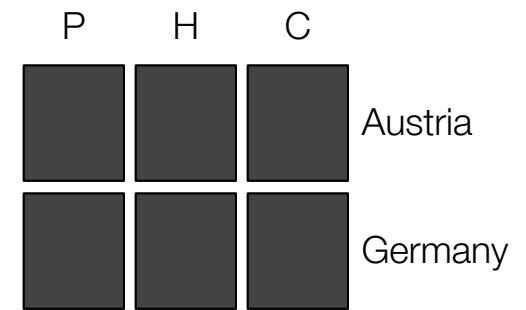
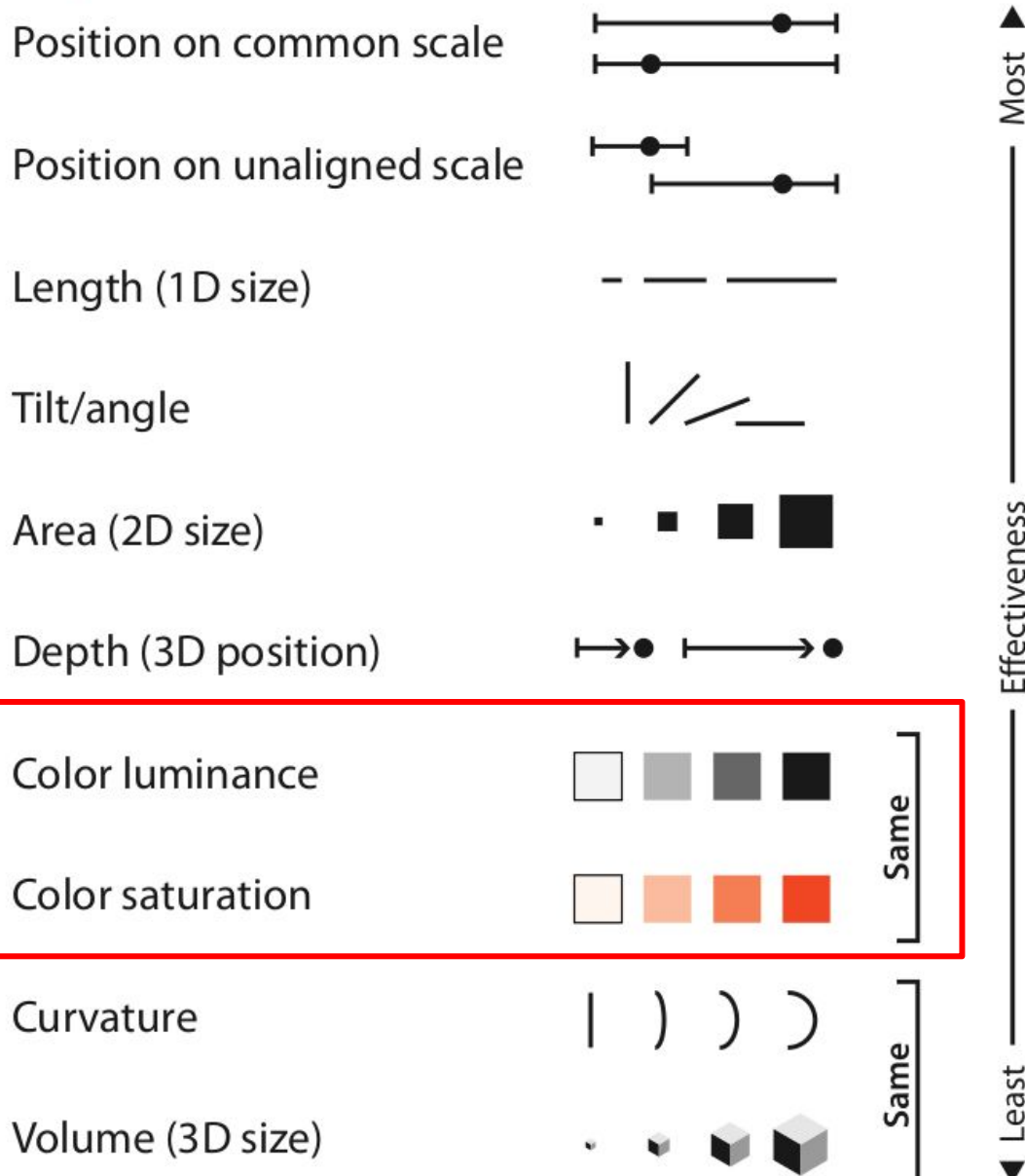
# 2D keys

- 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

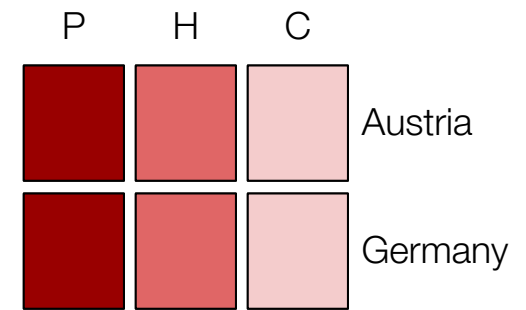
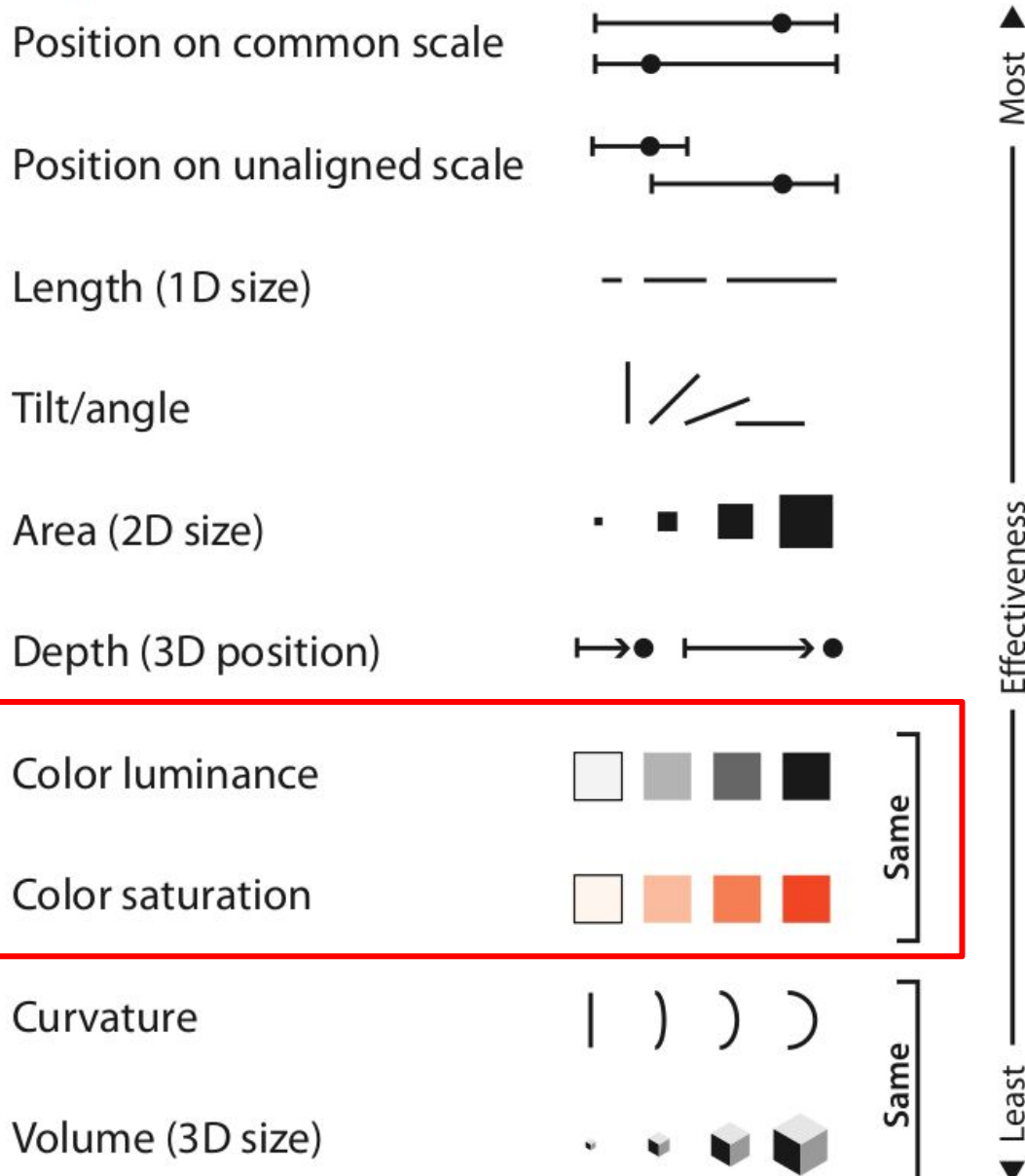


## ➔ Magnitude Channels: Ordered Attributes



→ Heatmap

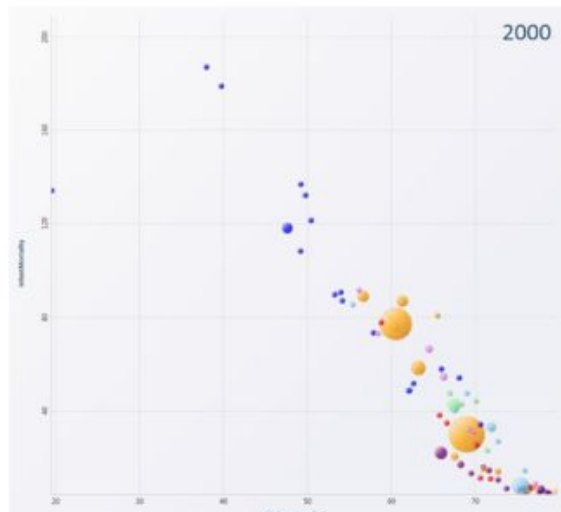
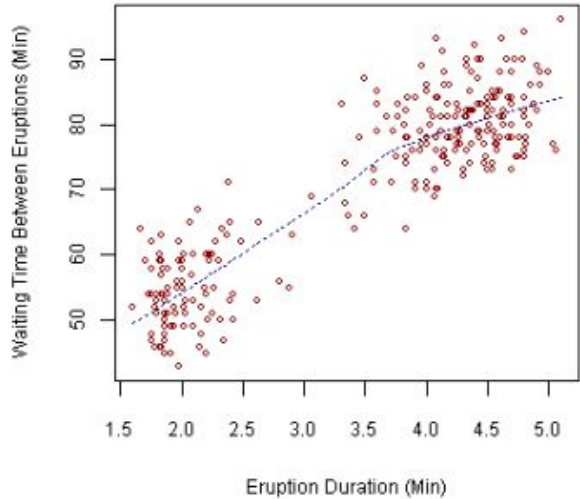
## ➔ Magnitude Channels: Ordered Attributes



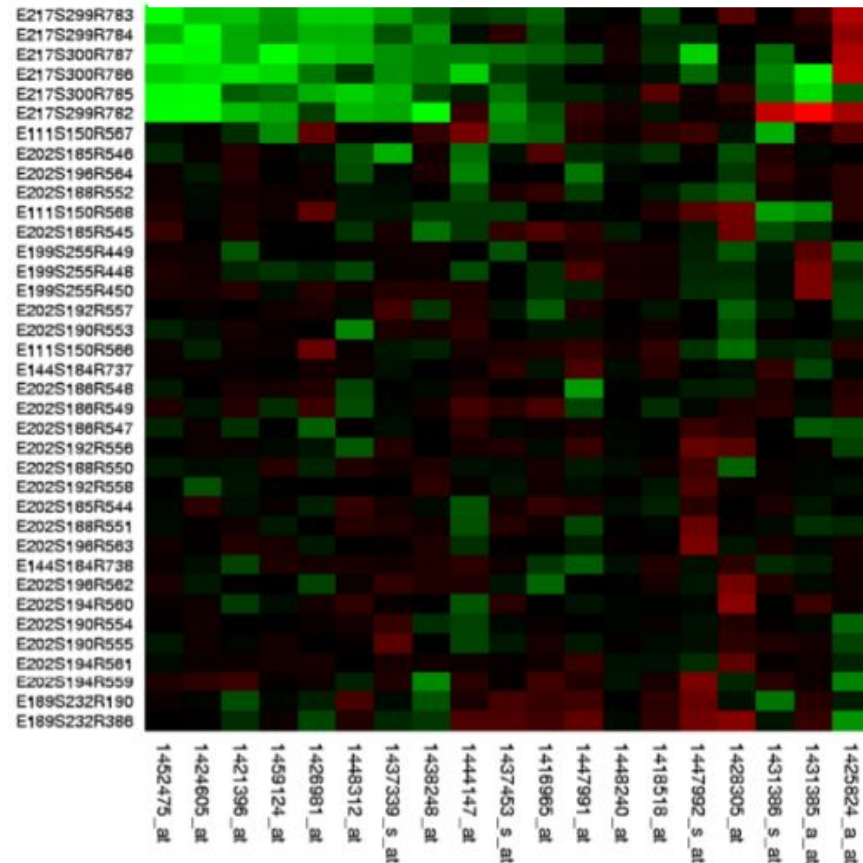
→ Heatmap

# 2D keys examples: tasks, size

Old Faithful Eruptions

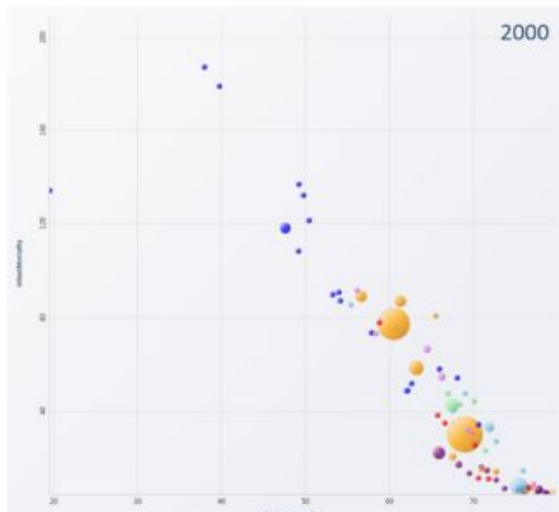
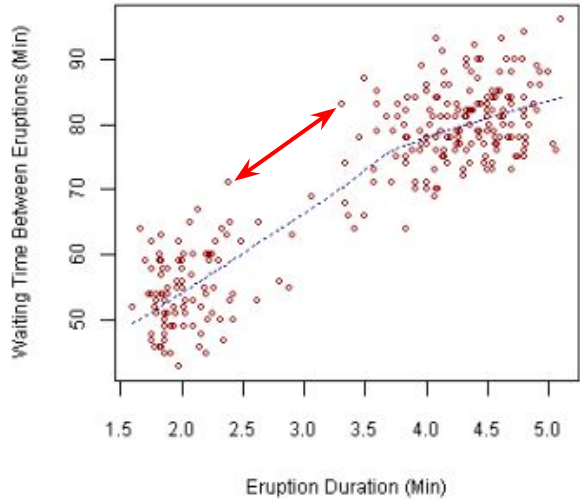


Heatmap

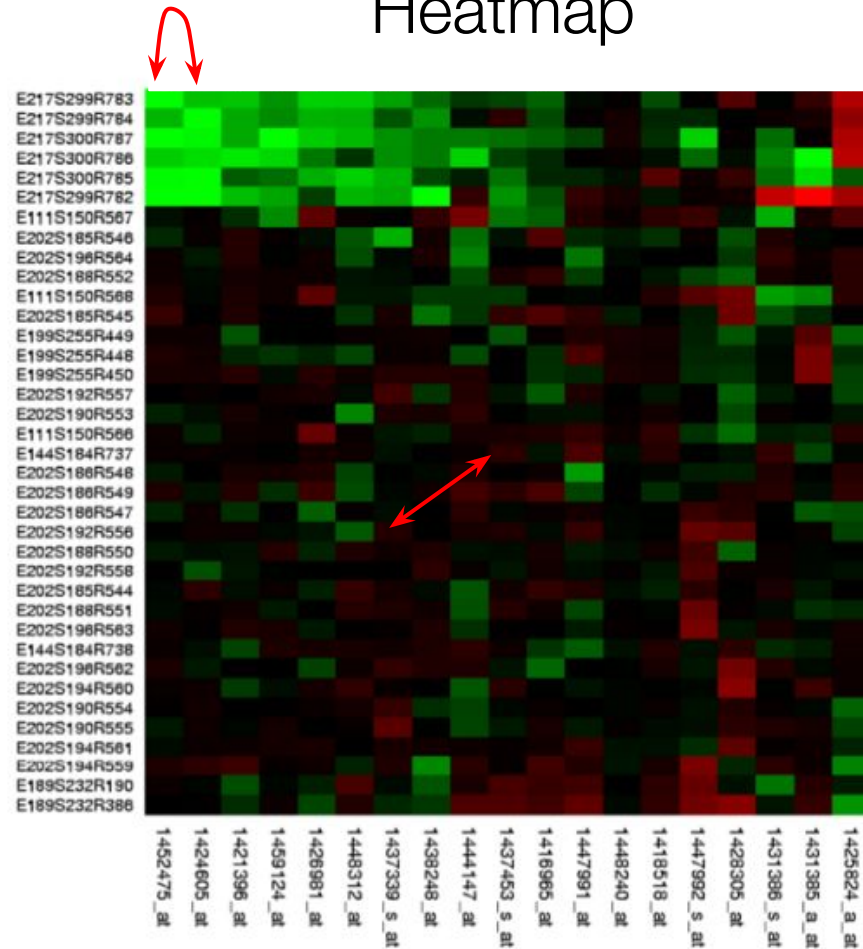


# Meaning of distances

Old Faithful Eruptions



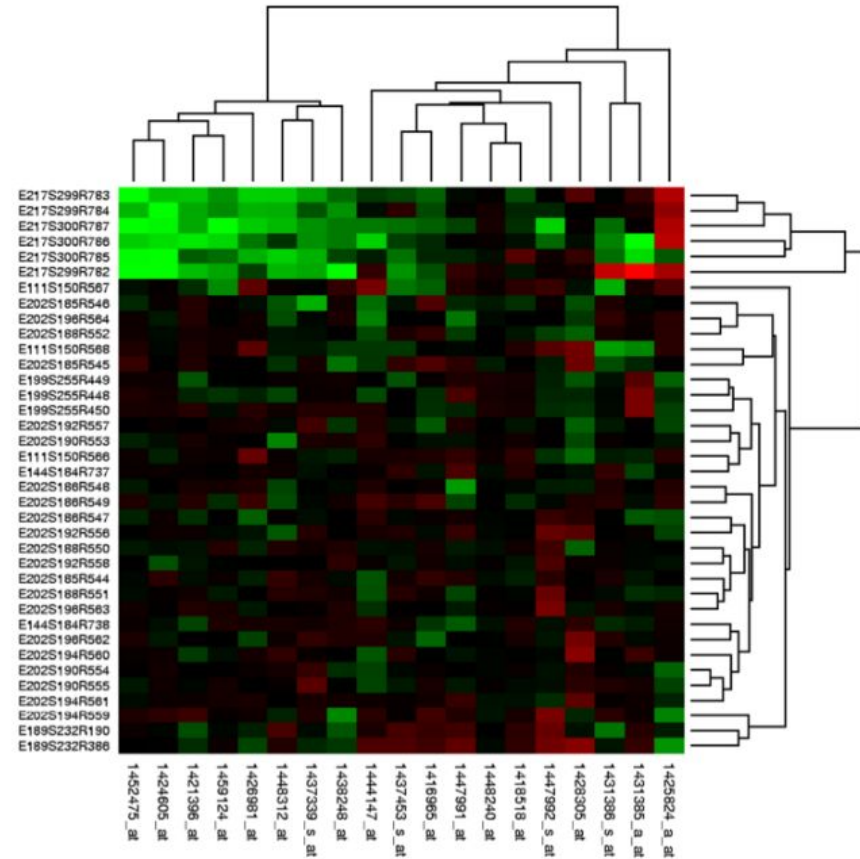
Heatmap





# Heatmaps

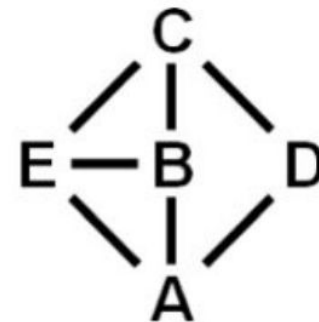
- Find structure in data
  - Outliers, clusters, etc.
  - Needs sorted rows & columns  
→ cluster heatmap
- Very compact
  - Limit: area mark=pixel
  - Good data overview



# 2D matrix arrangement - graph

- 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

# Multiple keys: partitioning / subdivide

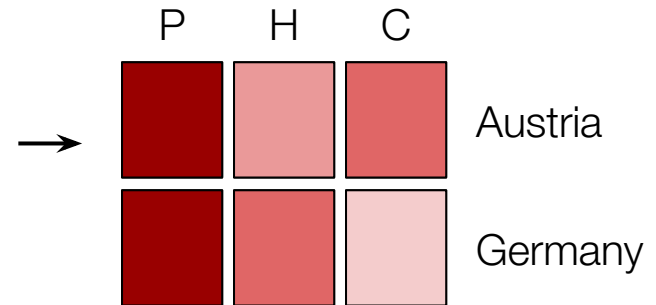
- e.g. 2 keys
  - use two perpendicular axes OR
  - use alignment on one axes
    - separate by key 1 first and then by key 2
- also known as **dimensional stacking**



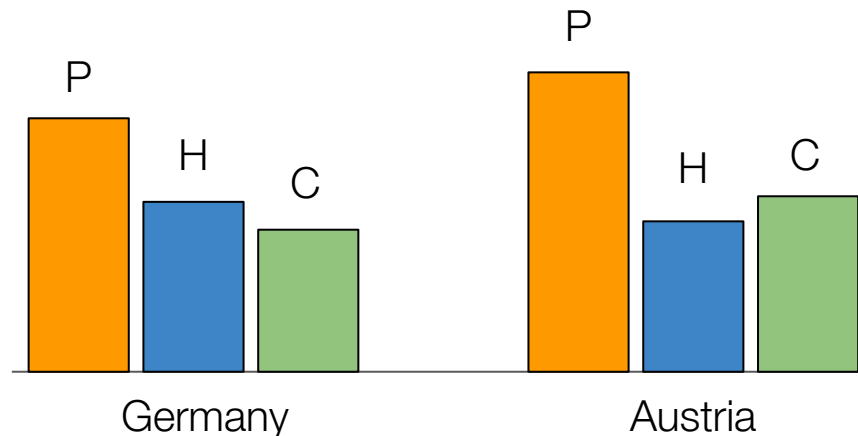
# Dimensional stacking: 2D

two perpendicular axes

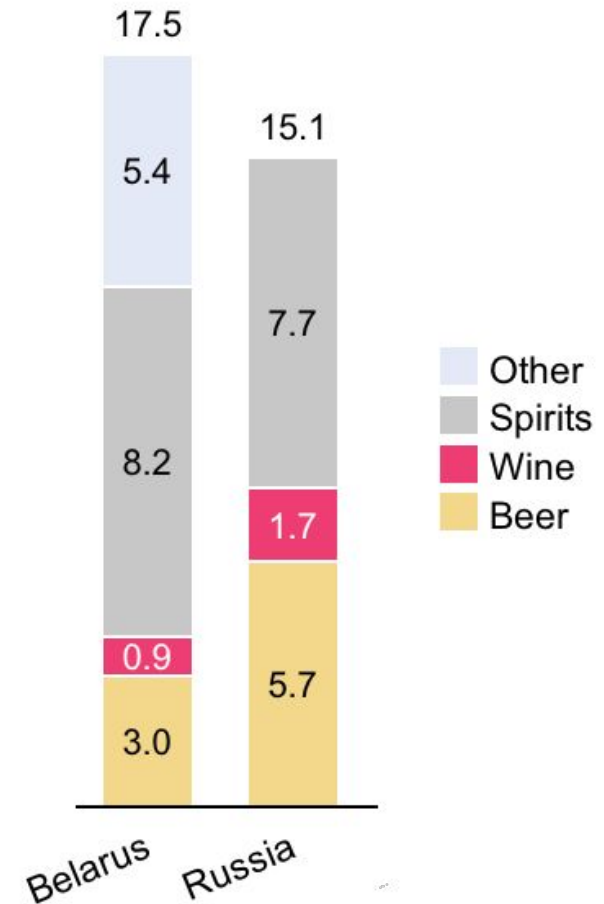
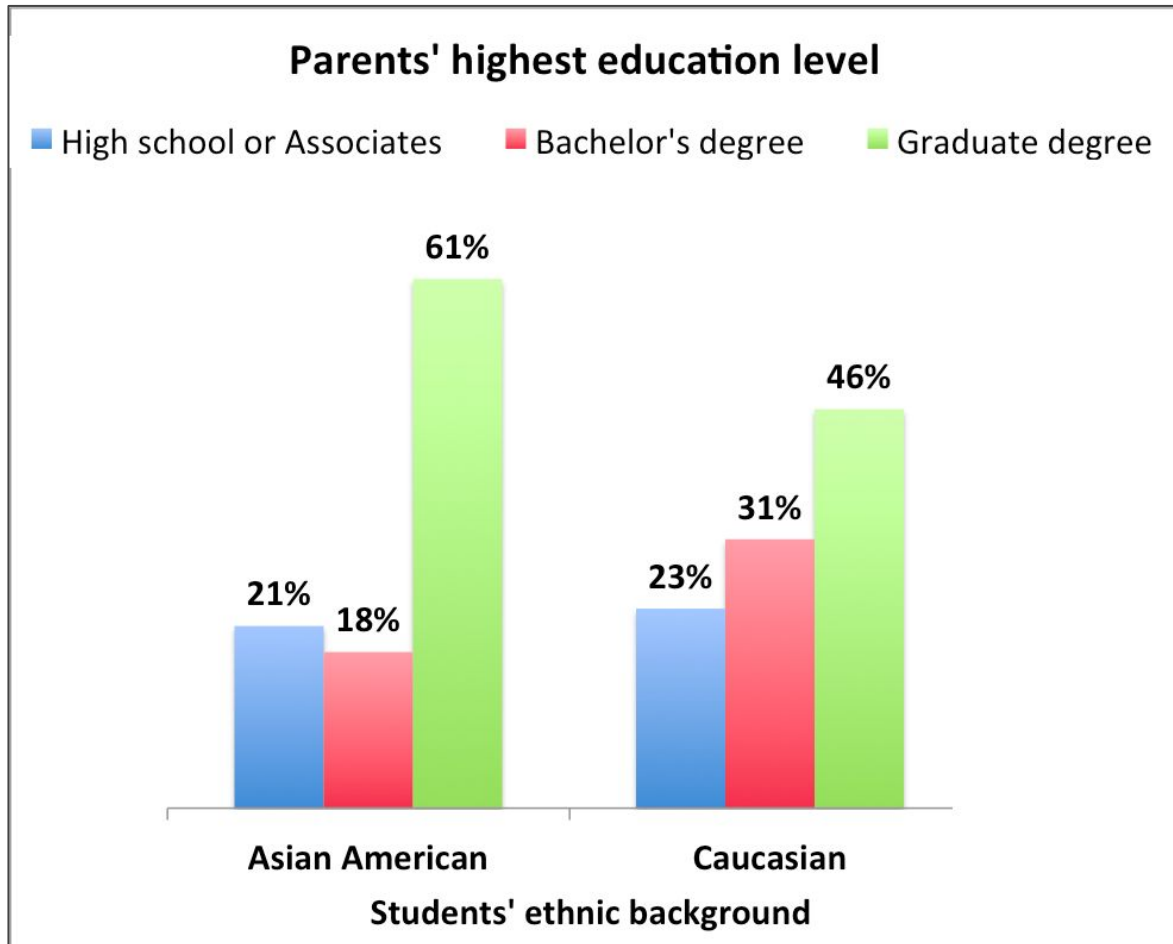
Country	Education level	Percentage
Austria	Primary school	99
Austria	High school	60
Austria	College	20
Germany	Primary school	98
Germany	High school	65
Germany	College	25



alignment on one axes



# 2D keys - 1D list alignment



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

Source: Wikipedia, 2010

What is the “correct/best” way  
to display your data?

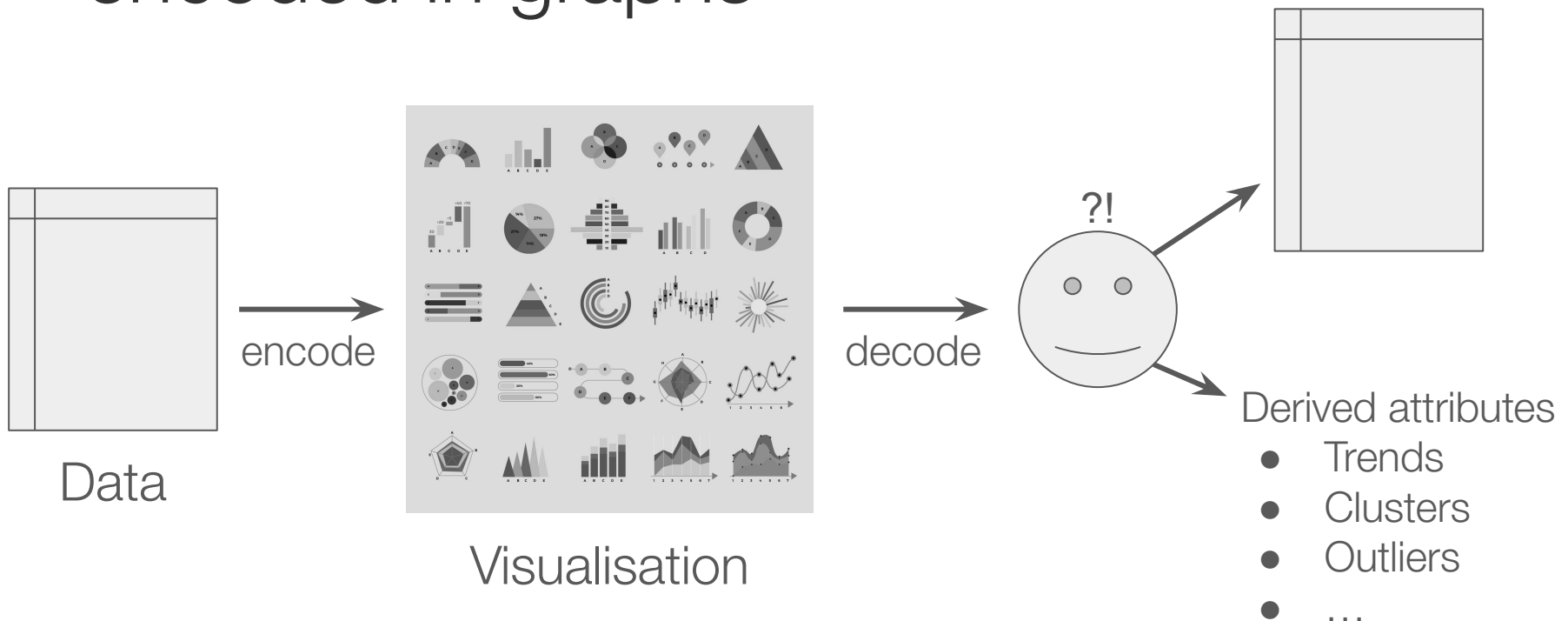
# Graphical perception

- The visual decoding of information encoded in graphs



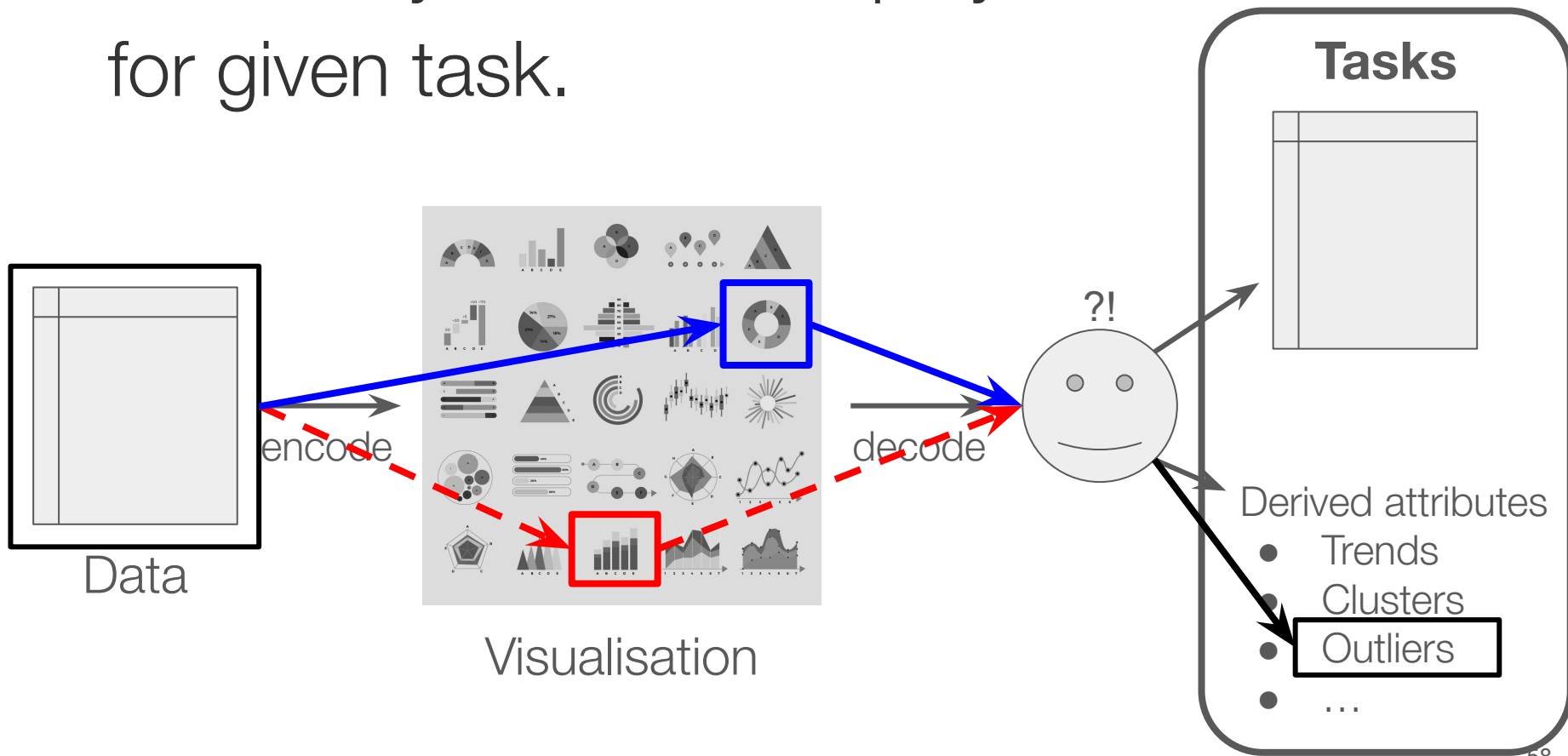
# Graphical perception

- The visual decoding of information encoded in graphs



# “Best” way to visualise data

- accurately decodes displayed information for given task.



# Graphical perception

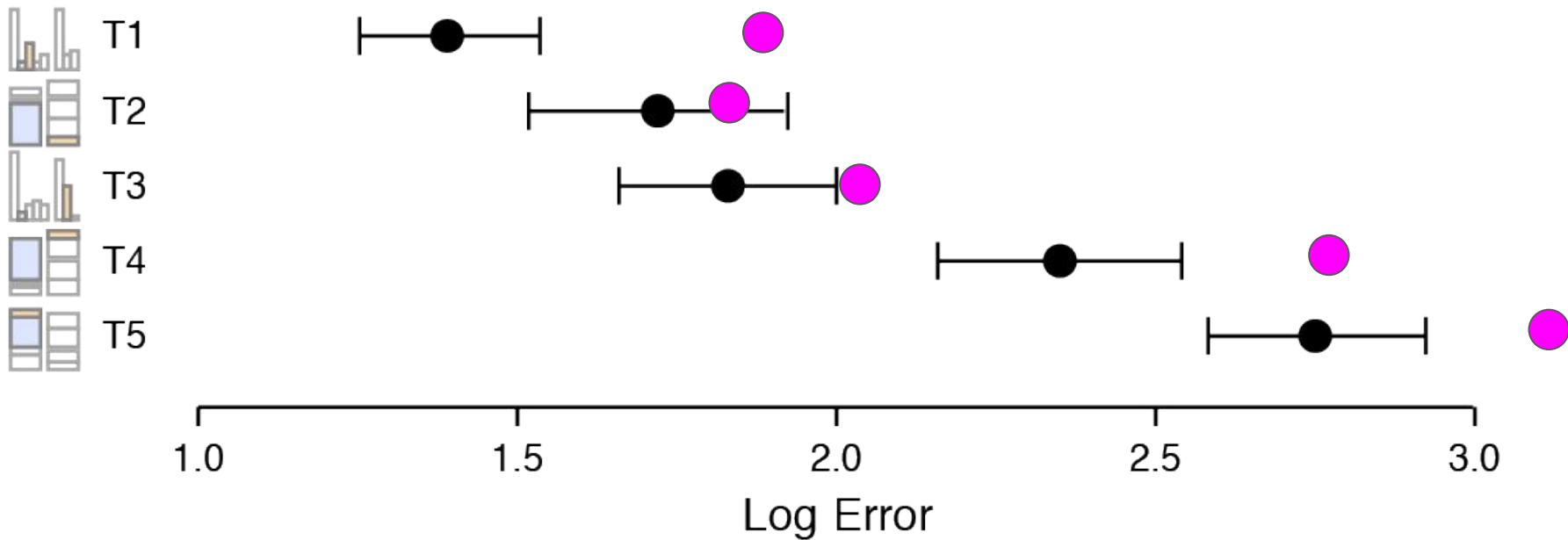
- Visual decoding of information encoded in graphs
- Human capacity to visually interpret information on graphs and charts → *aim to maximize it*
  - ↳ Graph & encoding choice determines ability to obtain “correct” information from chart

# Experiment: graphical perception

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

# Results

## Cleveland & McGill's Results



● Cleveland & McGill's results

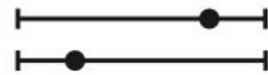
● Your results

# Results summary

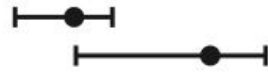
- Comparing by using **aligned position channel** against a common scale is the most accurately perceived visual channel.
- Comparing by using **unaligned position channel** adds mental overhead of due to different start and the end of each bar → reduced accuracy.

## ➔ Magnitude Channels: Ordered Attributes

Position on common scale



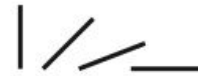
Position on unaligned scale



Length (1D size)



Tilt/angle



Area (2D size)



Depth (3D position)



Color luminance



Color saturation



Curvature



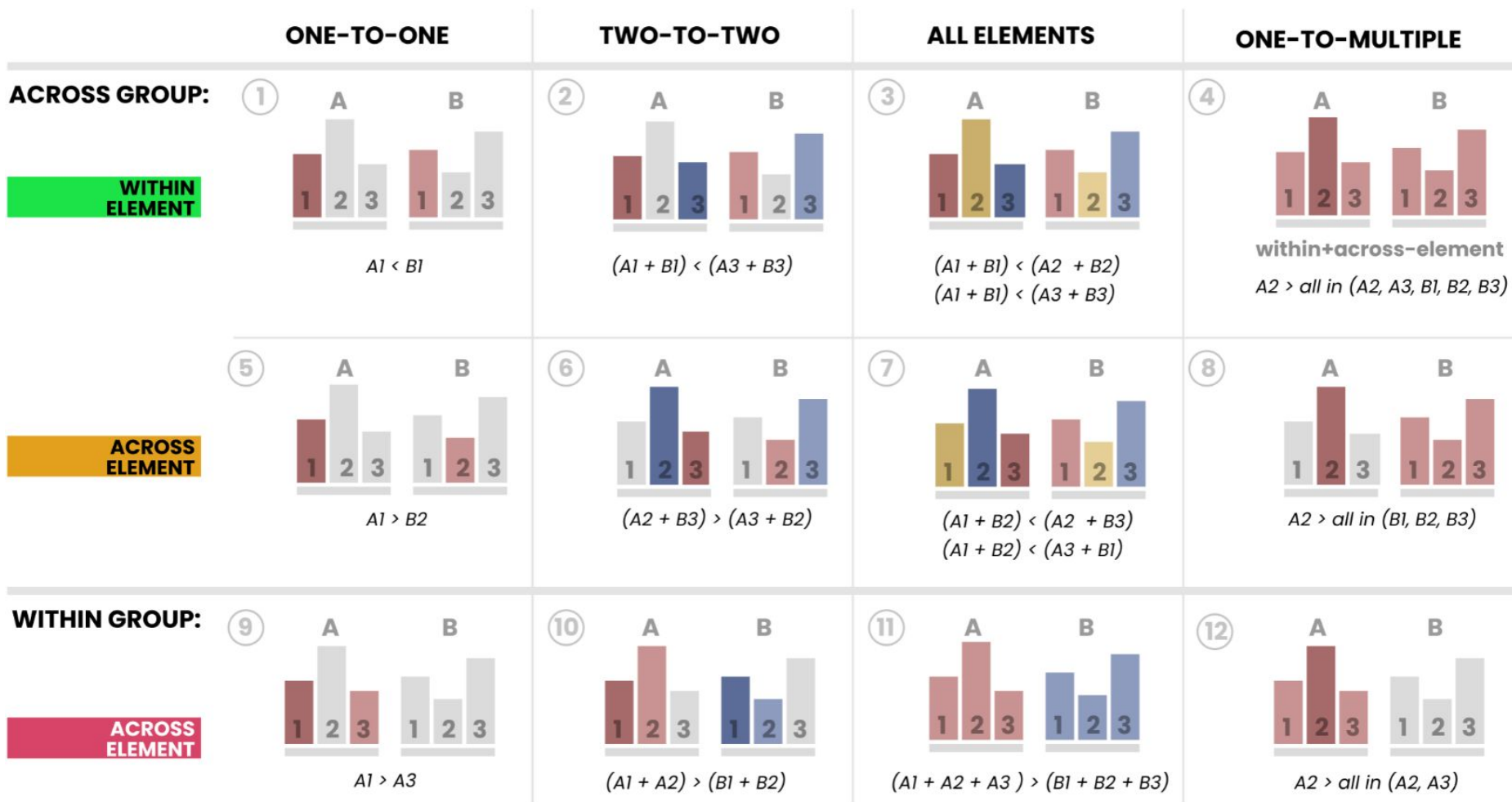
Volume (3D size)



Same


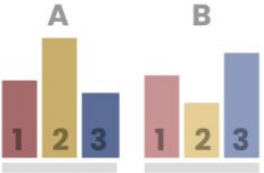


Effectiveness  
Most  
Least

# Apply what you just learned

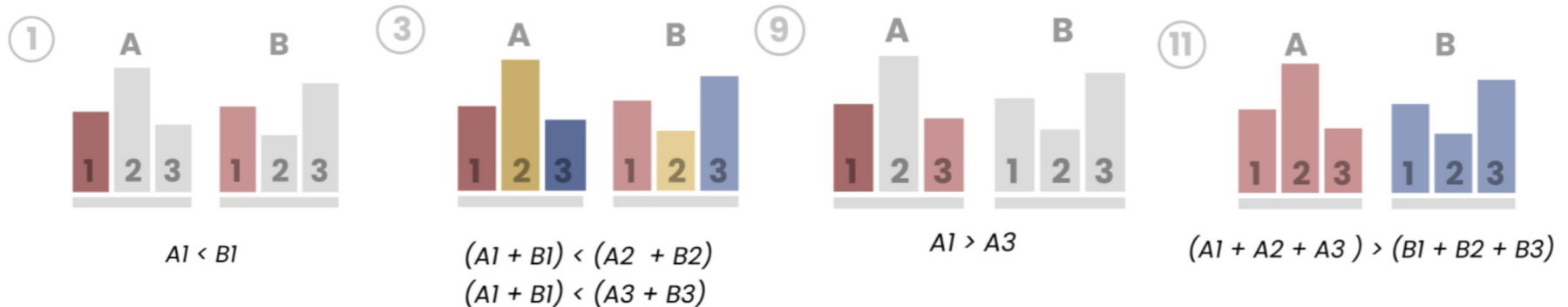




# Apply what you just learned

	ONE-TO-ONE	TWO-TO-TWO	ALL ELEMENTS	ONE-TO-MULTIPLE
<b>ACROSS GROUP:</b> ①  <b>WITHIN ELEMENT</b>	 <p><math>A1 &lt; B1</math></p>		 <p> <math>(A1 + B1) &lt; (A2 + B2)</math>  <math>(A1 + B1) &lt; (A3 + B3)</math> </p>	
<b>ACROSS ELEMENT</b>				
<b>WITHIN GROUP:</b> ⑨  <b>ACROSS ELEMENT</b>	 <p><math>A1 &gt; A3</math></p>		 <p><math>(A1 + A2 + A3) &gt; (B1 + B2 + B3)</math></p>	

# Apply what you just learned

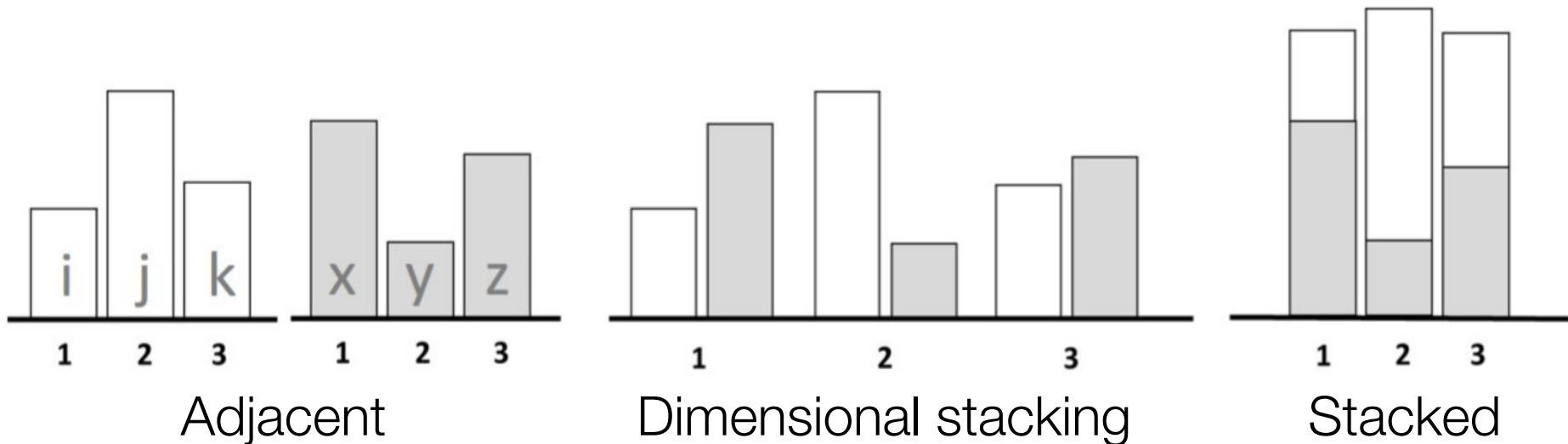
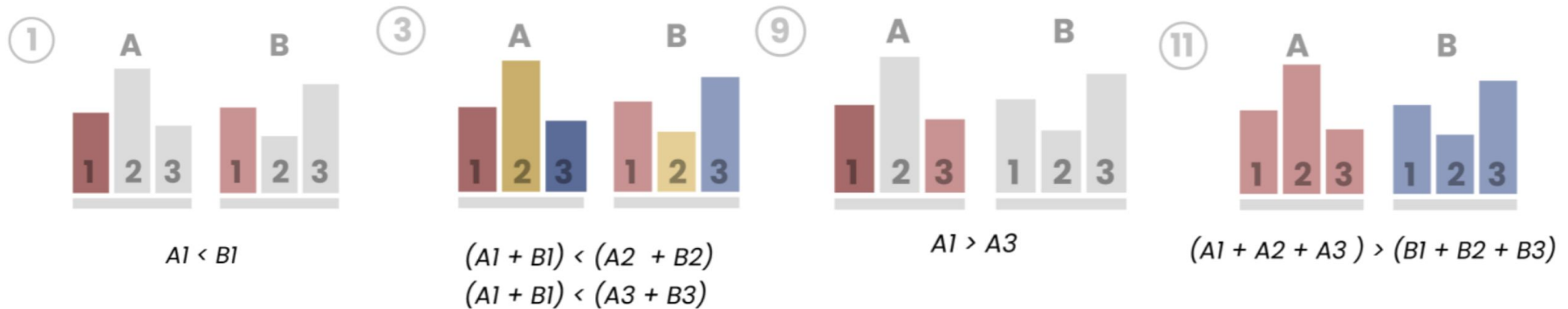


A & B Companies | 1,2,3 Continents | Bar height: revenue

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

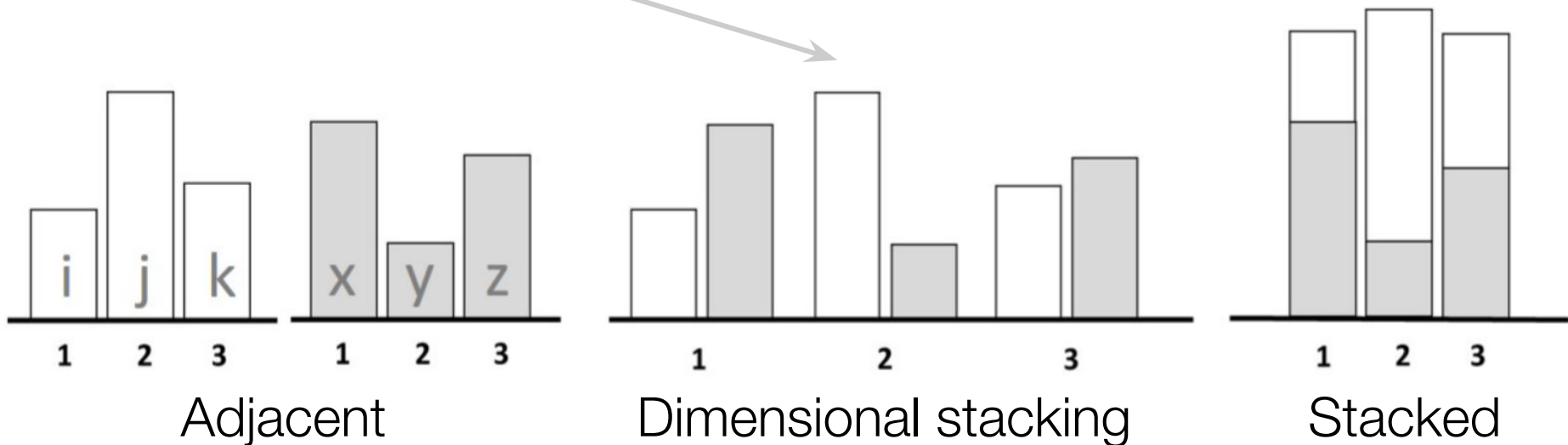
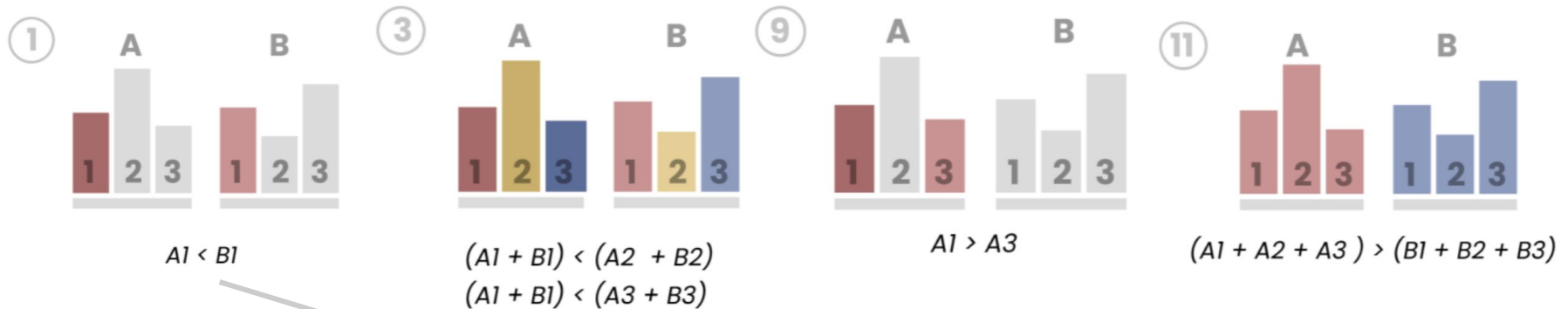
A & B Companies | 1,2,3 Continents | Bar height: revenue

## Comparing revenue between companies A & B in continent 1.



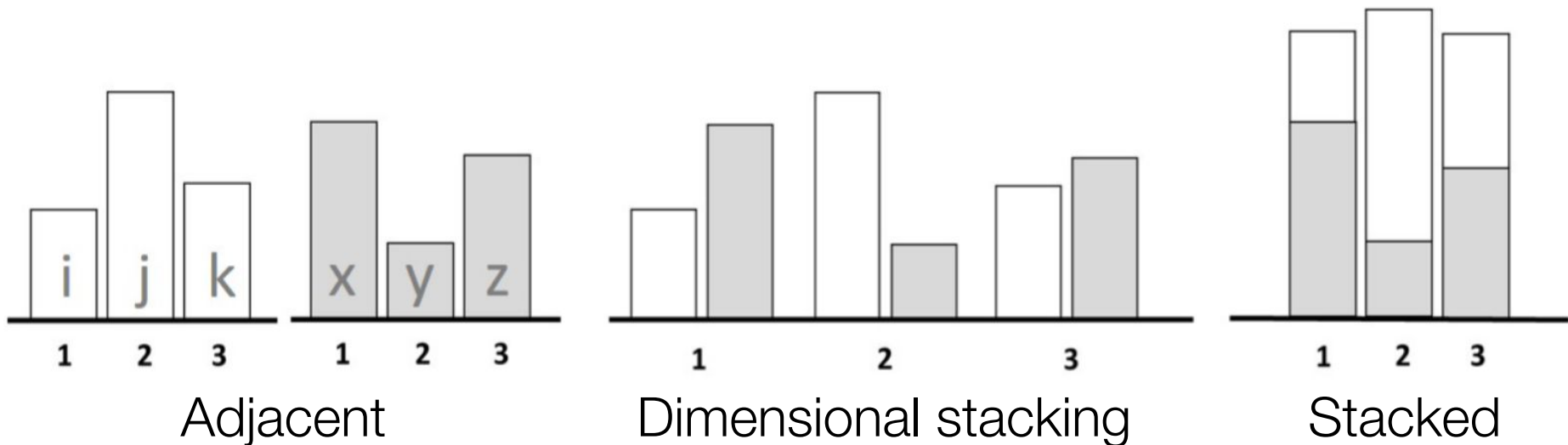
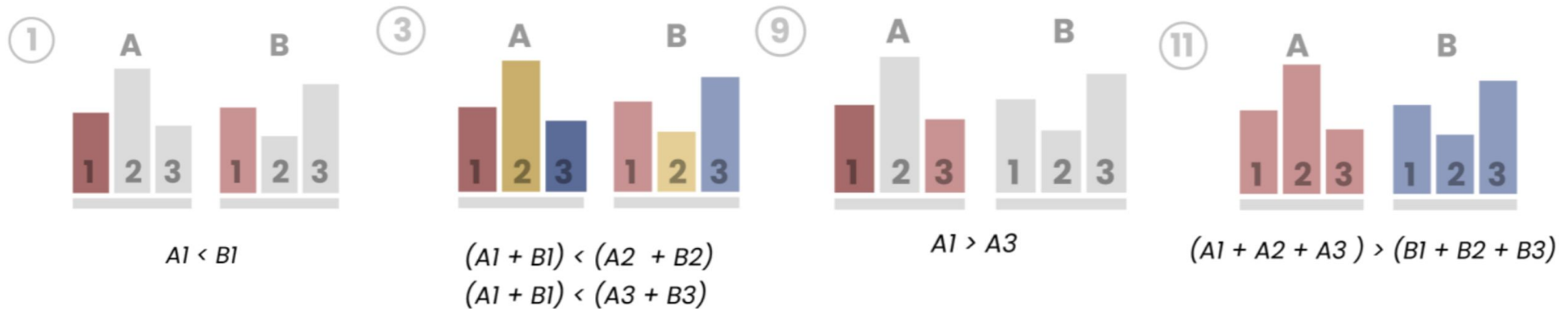
A & B Companies | 1,2,3 Continents | Bar height: revenue

## Comparing revenue between companies A & B in continent 1.



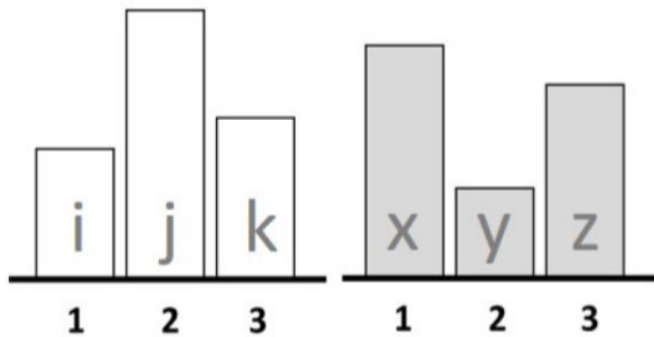
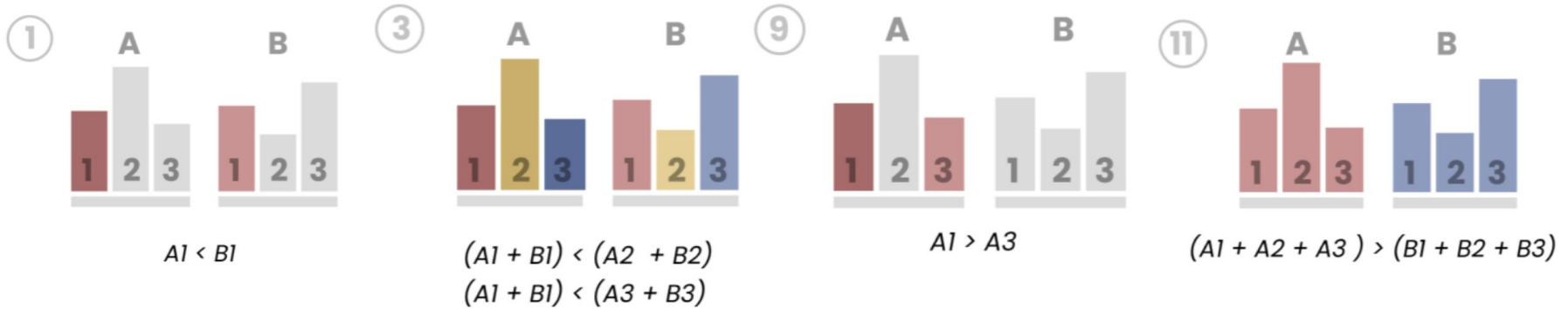
A & B Companies | 1,2,3 Continents | Bar height: revenue

**Comparing total revenue of A & B in given continent.**

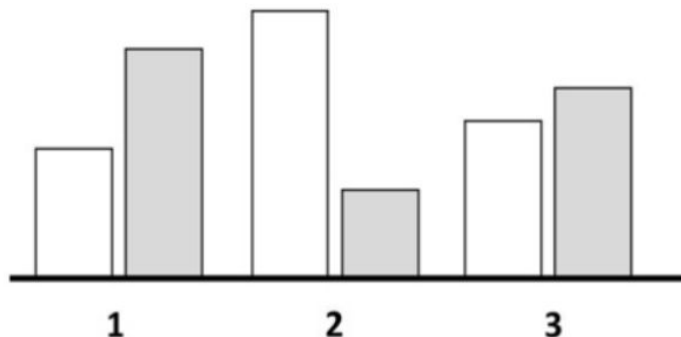


A & B Companies | 1,2,3 Continents | Bar height: revenue

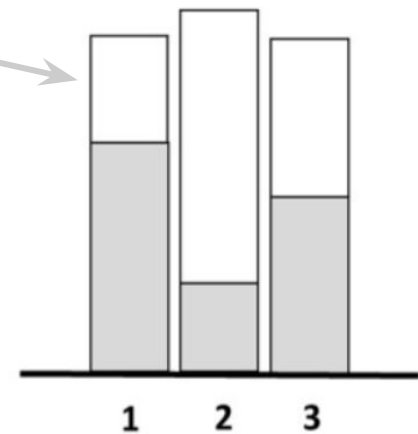
**Comparing total revenue of A & B in given continent.**



Adjacent



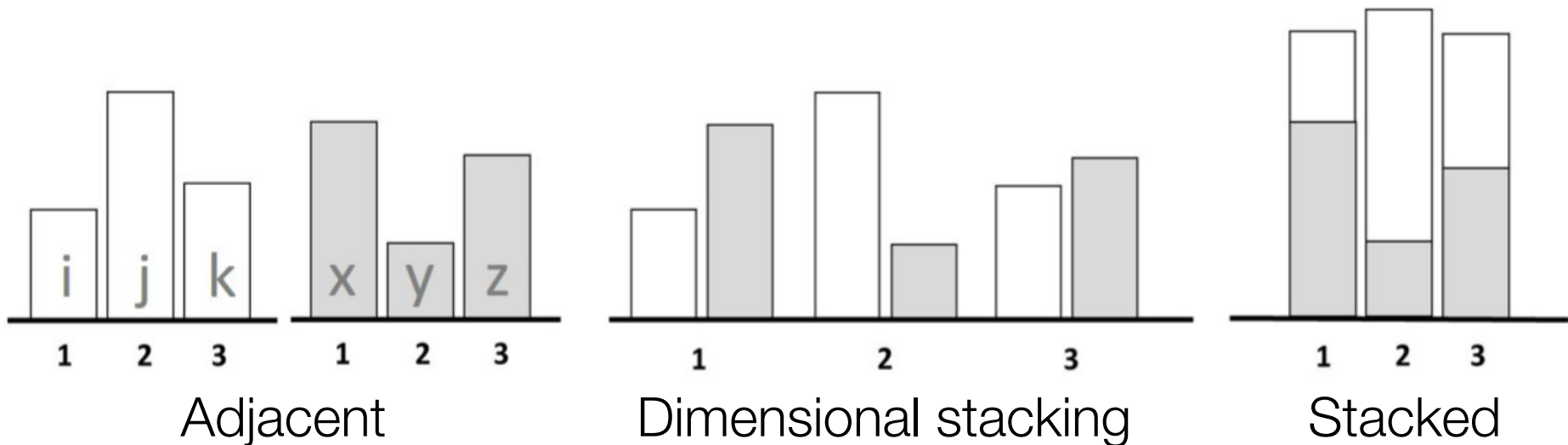
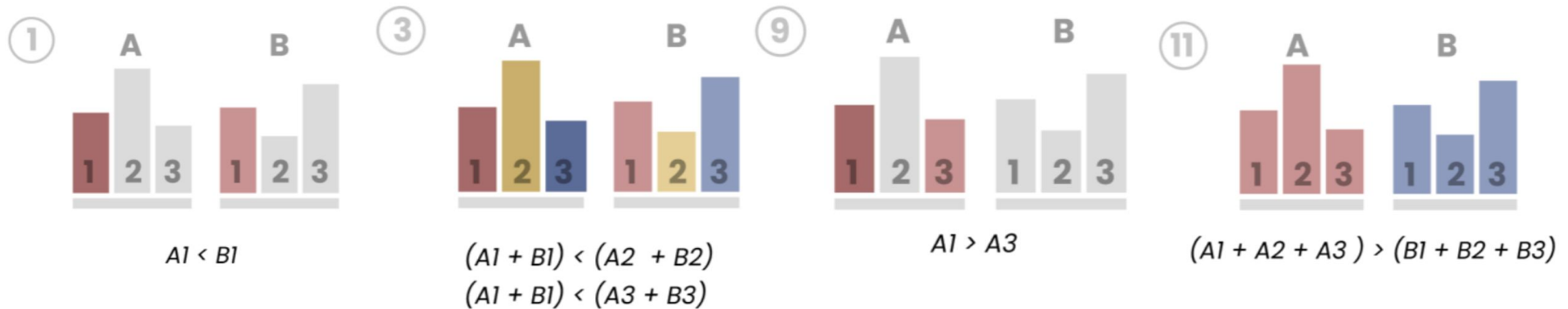
Dimensional stacking



Stacked

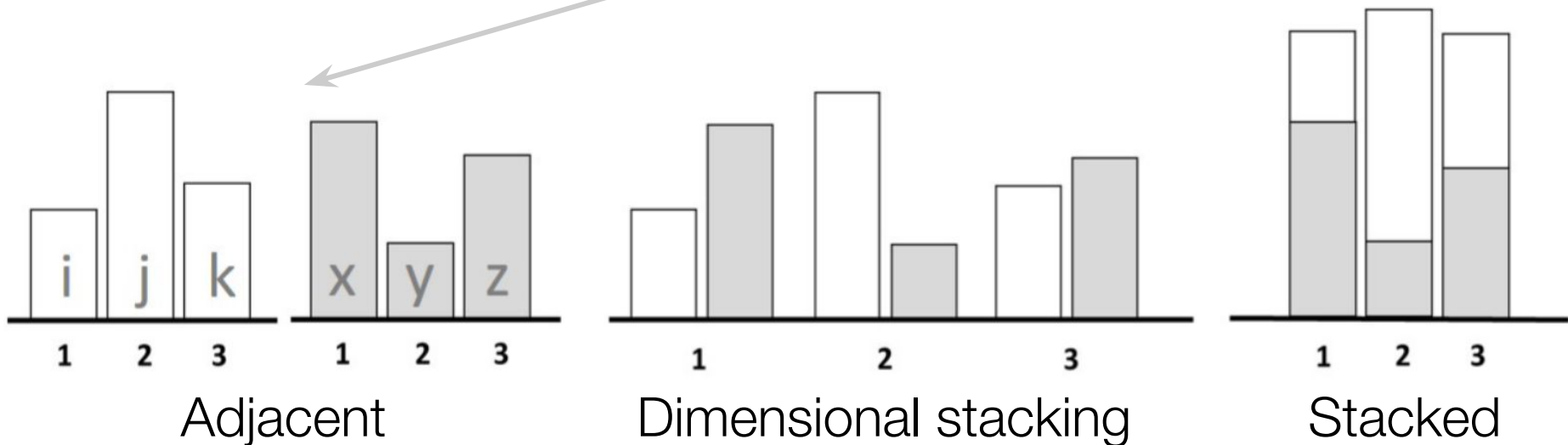
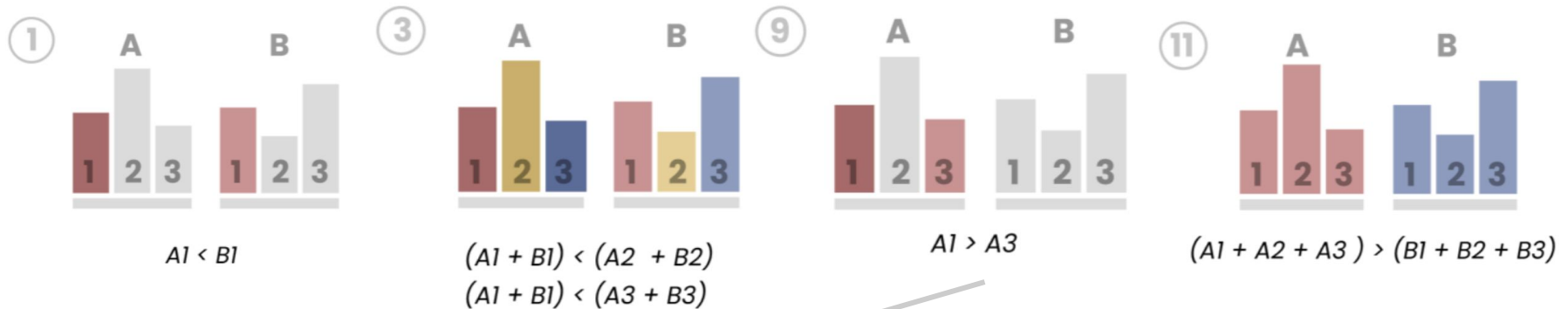
A & B Companies | 1,2,3 Continents | Bar height: revenue

## Comparing revenue of company A among different continents.



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

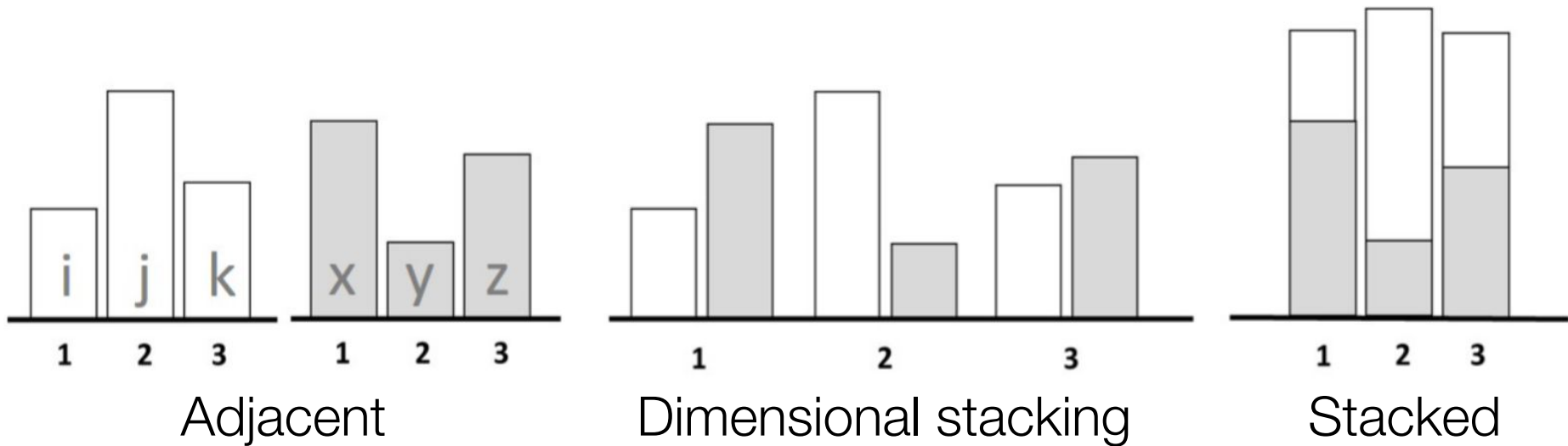
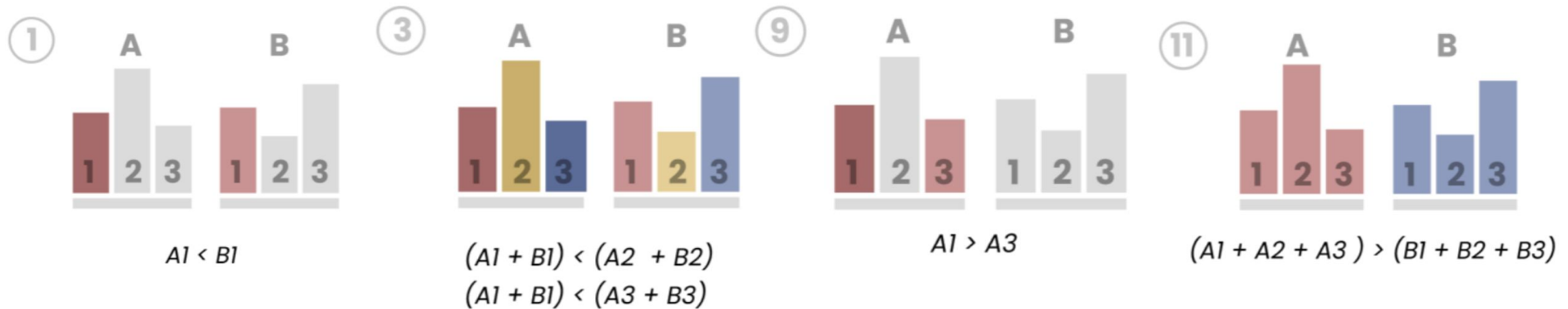
## Comparing revenue of company A among different continents.





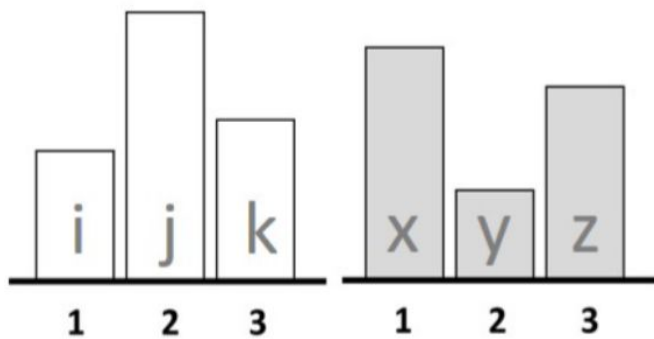
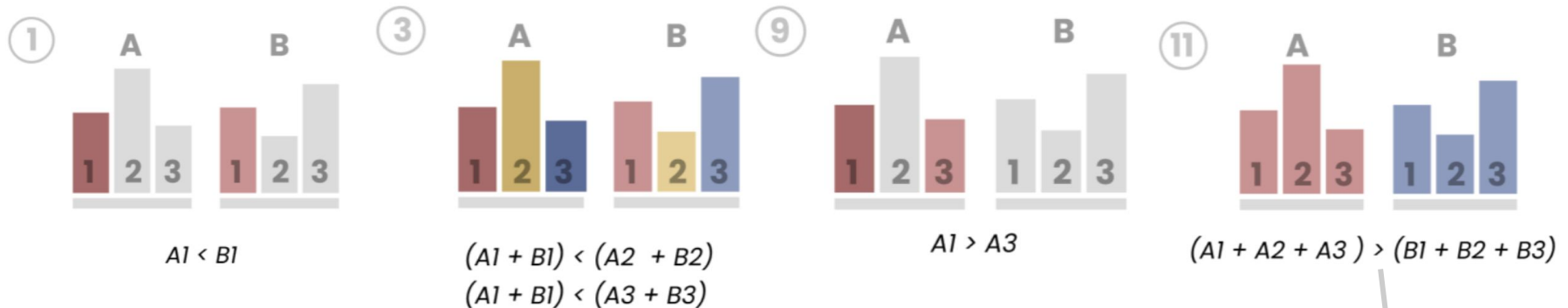
# A & B Companies | 1,2,3 Continents | Bar height: revenue

## Comparing the global revenue of company A to B.

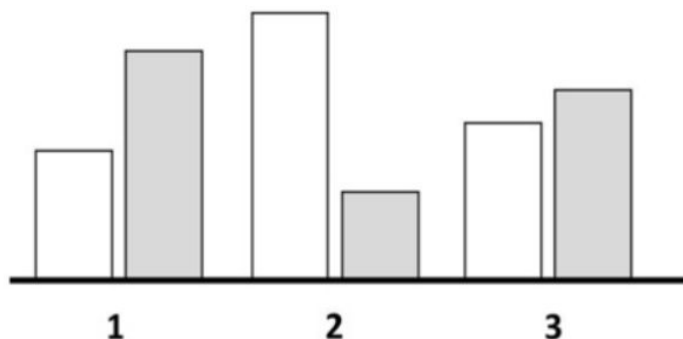


# A & B Companies | 1,2,3 Continents | Bar height: revenue

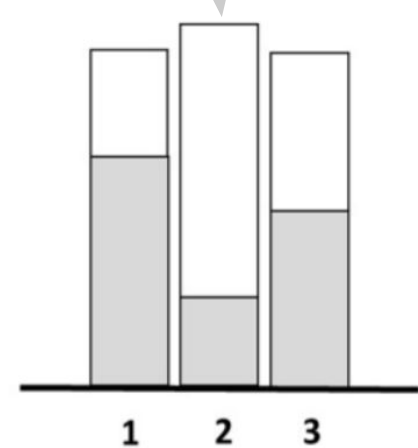
## Comparing the global revenue of company A to B.



Adjacent



Dimensional stacking



Stacked