

Perception

DSC 106: Data Visualization

Sam Lau
UC San Diego

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#5872 640



Announcements

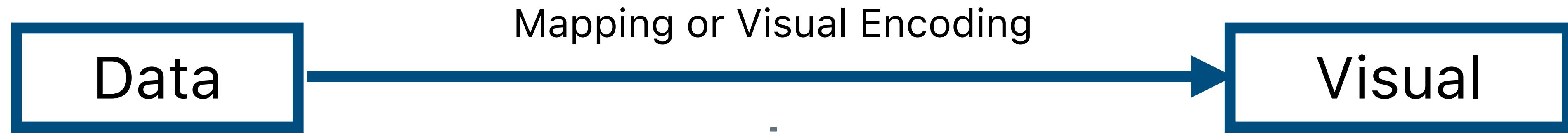
Lab 2 due tomorrow 1/19.

Project 1 also due tomorrow 1/19.



FAQs:

1. How does project grading work? You get 9/10 points if you follow all the project requirements. Can get more if your project goes above and beyond requirements (see project page for more details).
2. OH now have signup forms to distribute checkoffs, see Ed for more details.



Expressiveness

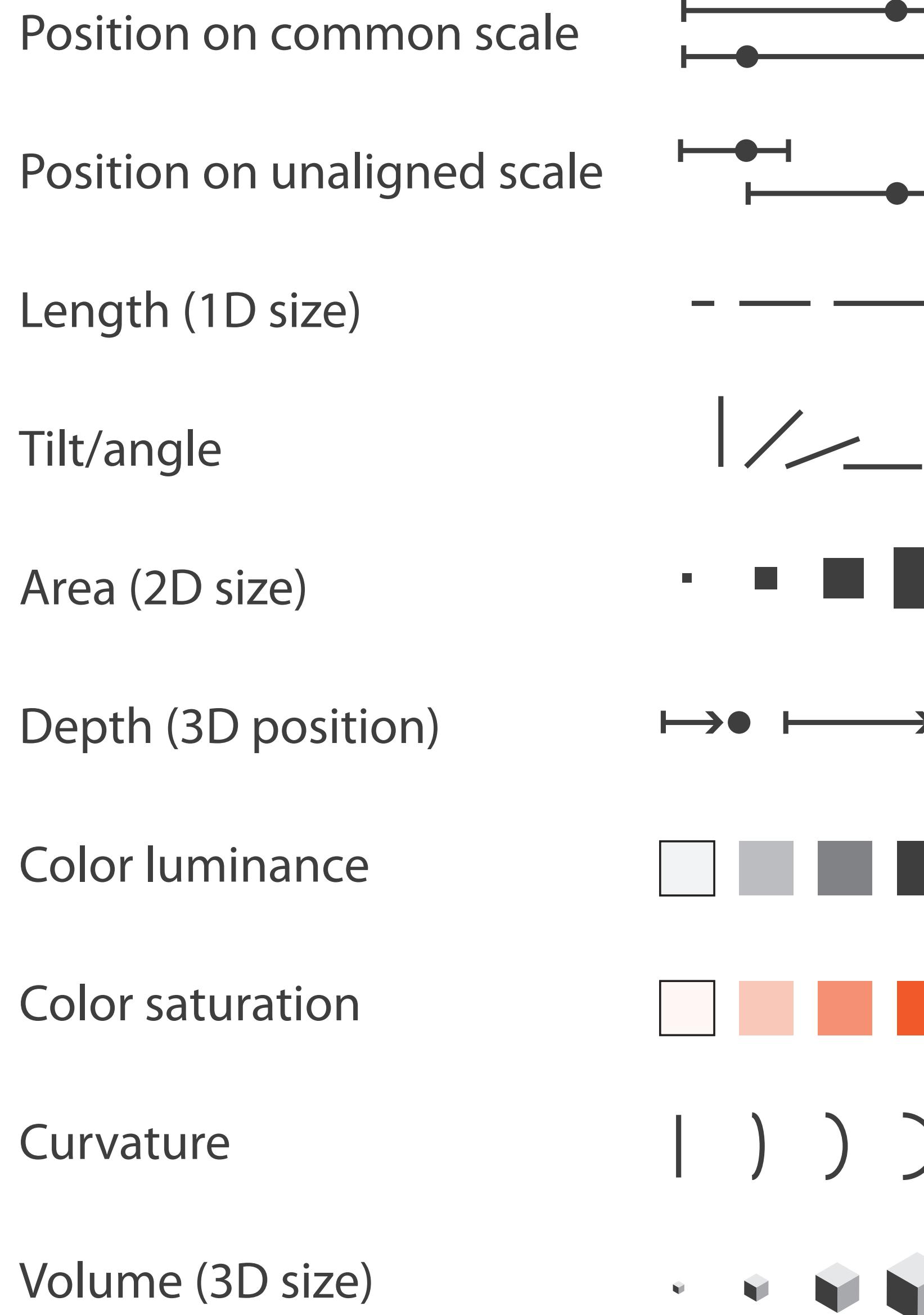
A set of facts is *expressible* in a visual language if the sentences (i.e. the visualizations) in the language express *all the facts in the set of data, and only the facts in the data.*

Effectiveness

A visualization is more *effective* than another if the information it conveys *is more readily perceived* than the information in the other visualization

Channels: Expressiveness Types and Effectiveness Ranks

→ Magnitude Channels: Ordered Attributes



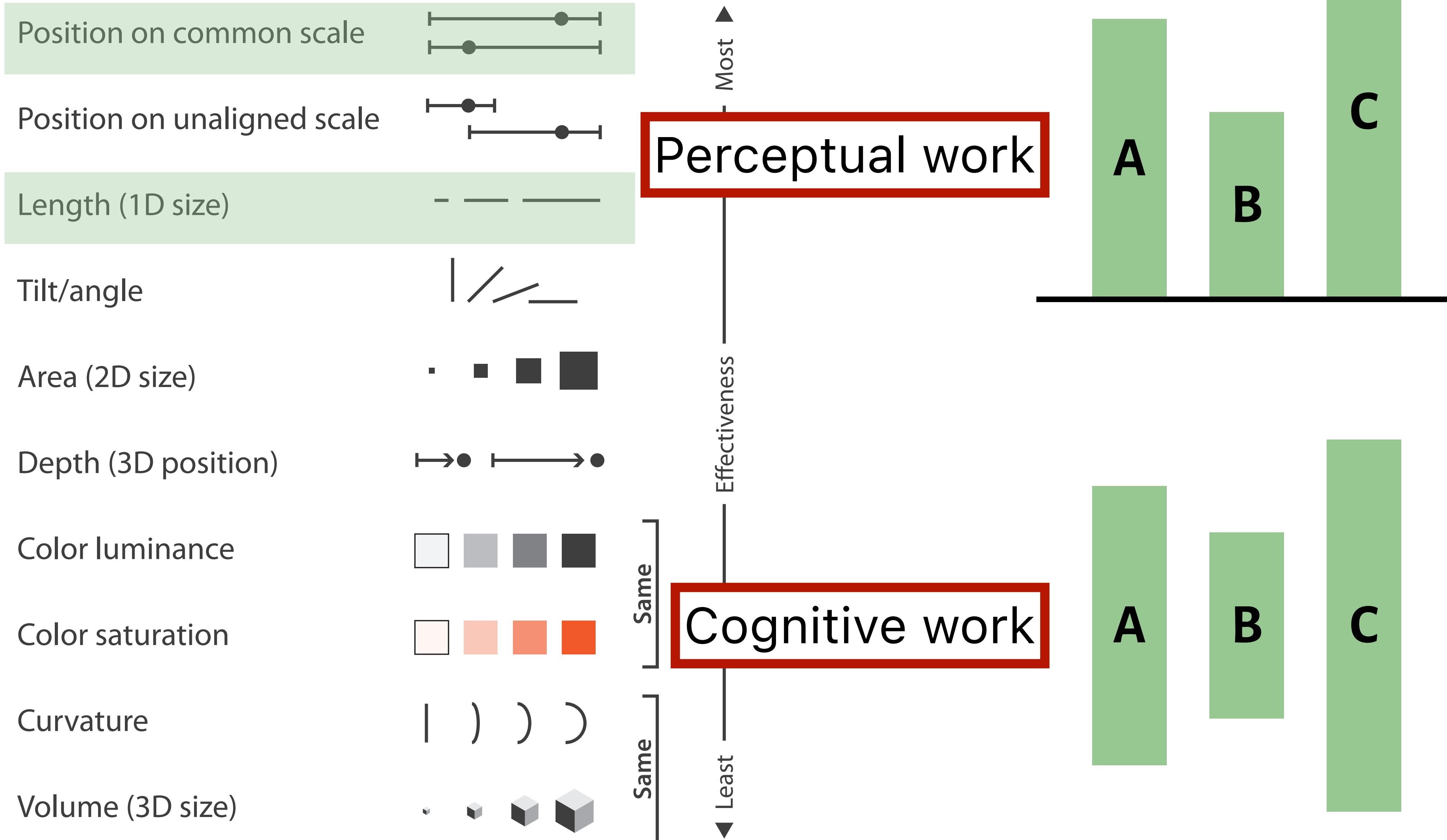
→ Identity Channels: Categorical Attributes



Tamara Munzner, *Visualization Analysis and Design* (2014).

Channels: Expressiveness Types and Effectiveness Ranks

→ **Magnitude Channels:** O or Q attributes



Graphical Perception

The ability of viewers to interpret visual (graphical) encodings of information and thereby decode information in graphs.

Signal Detection

Magnitude Estimation

Pre-Attentive Processing

Selective Attention

Gestalt Grouping

Signal Detection

Discriminability: how easy is it
to tell two things apart?

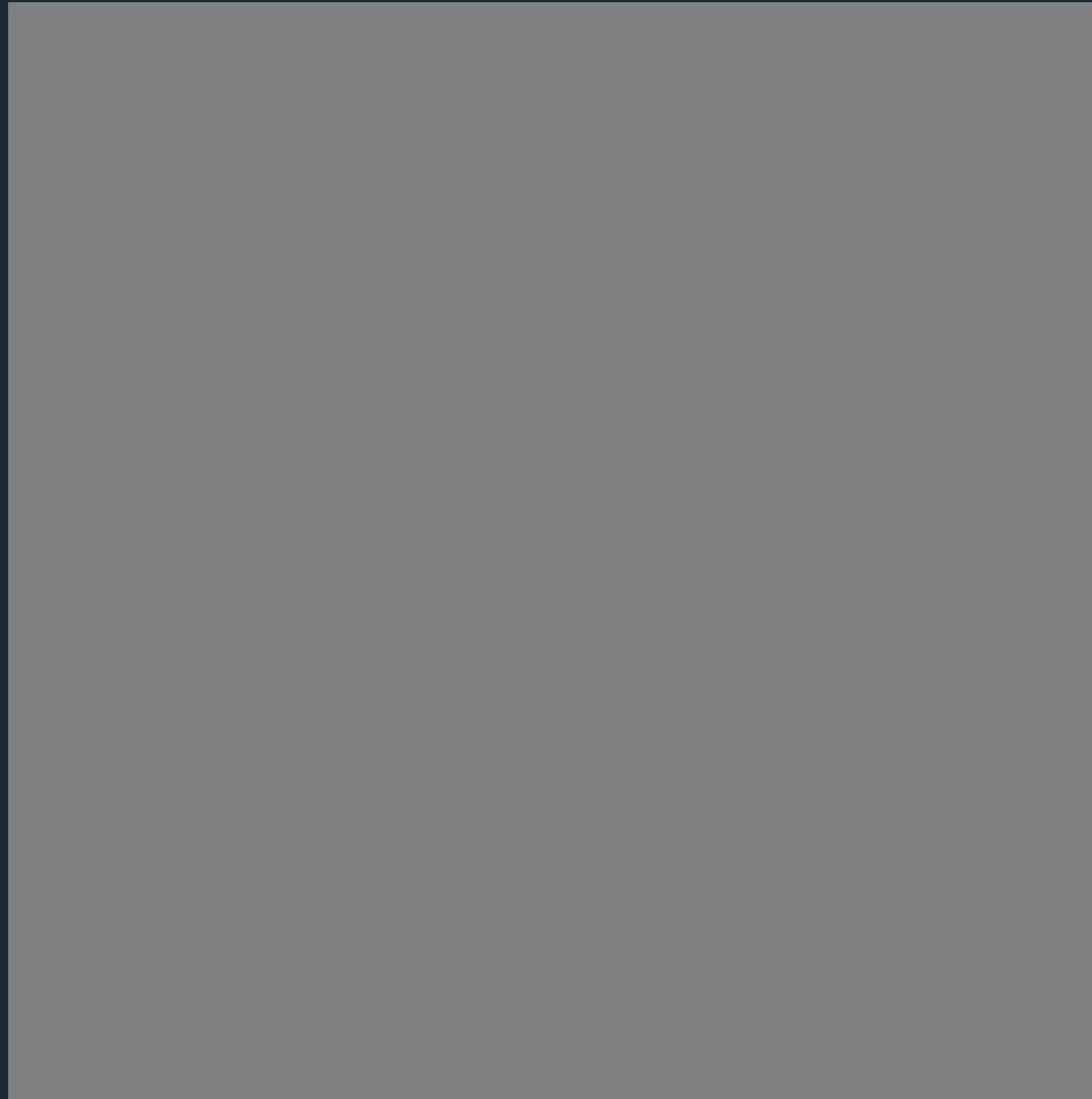
Magnitude Estimation

Pre-Attentive Processing

Selective Attention

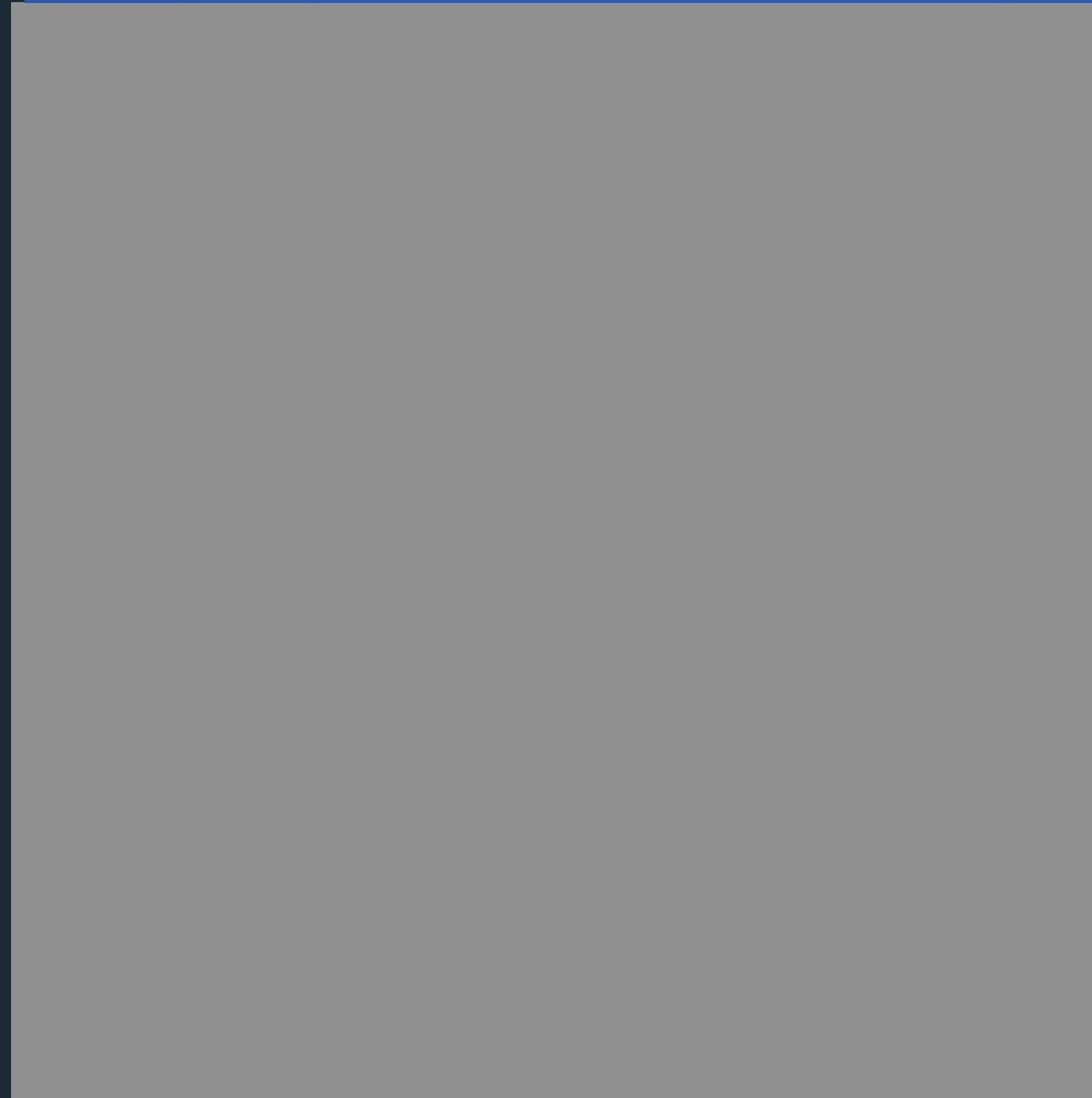
Gestalt Grouping

Which is brighter?



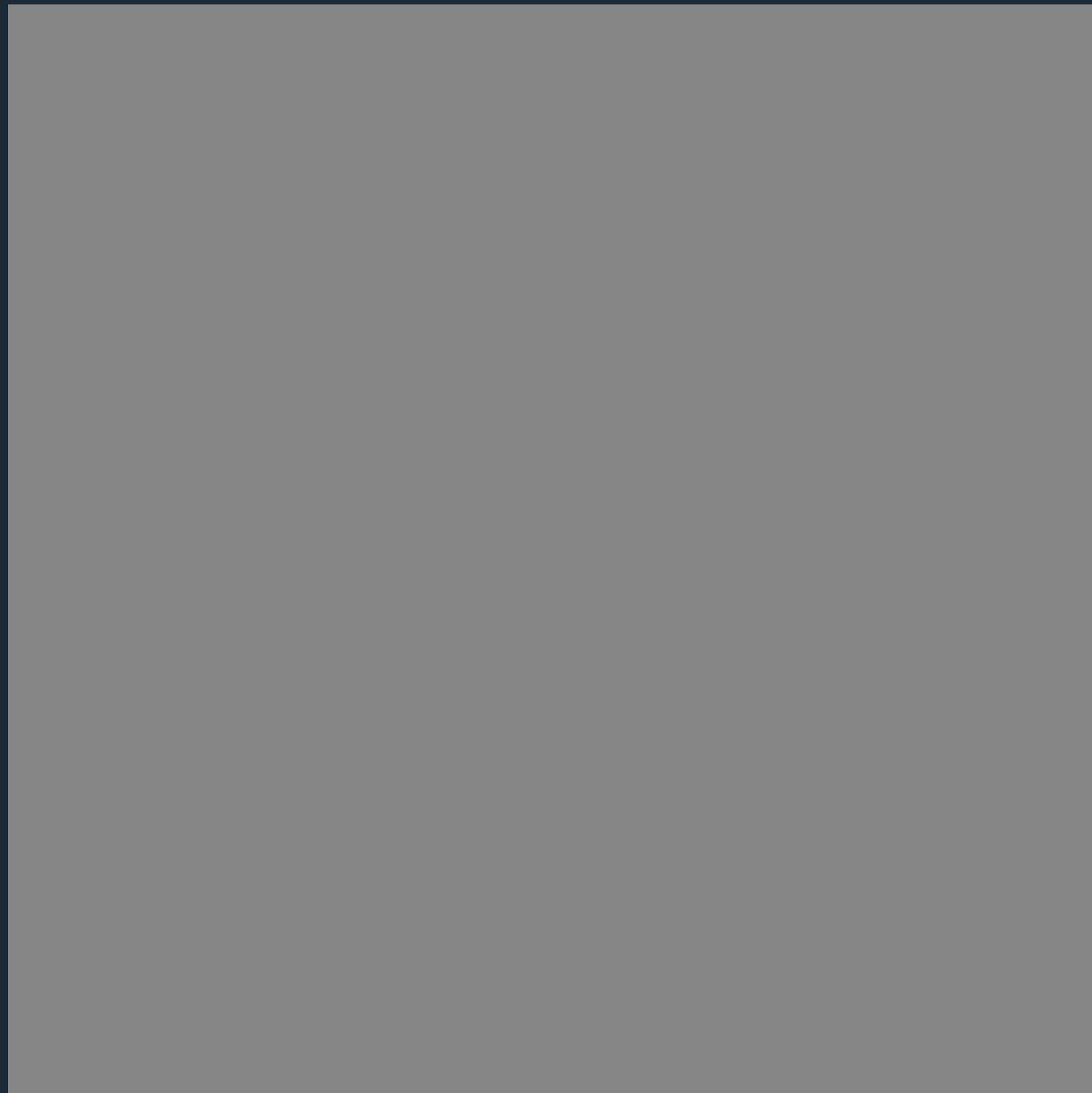
rgb(128, 128, 128)

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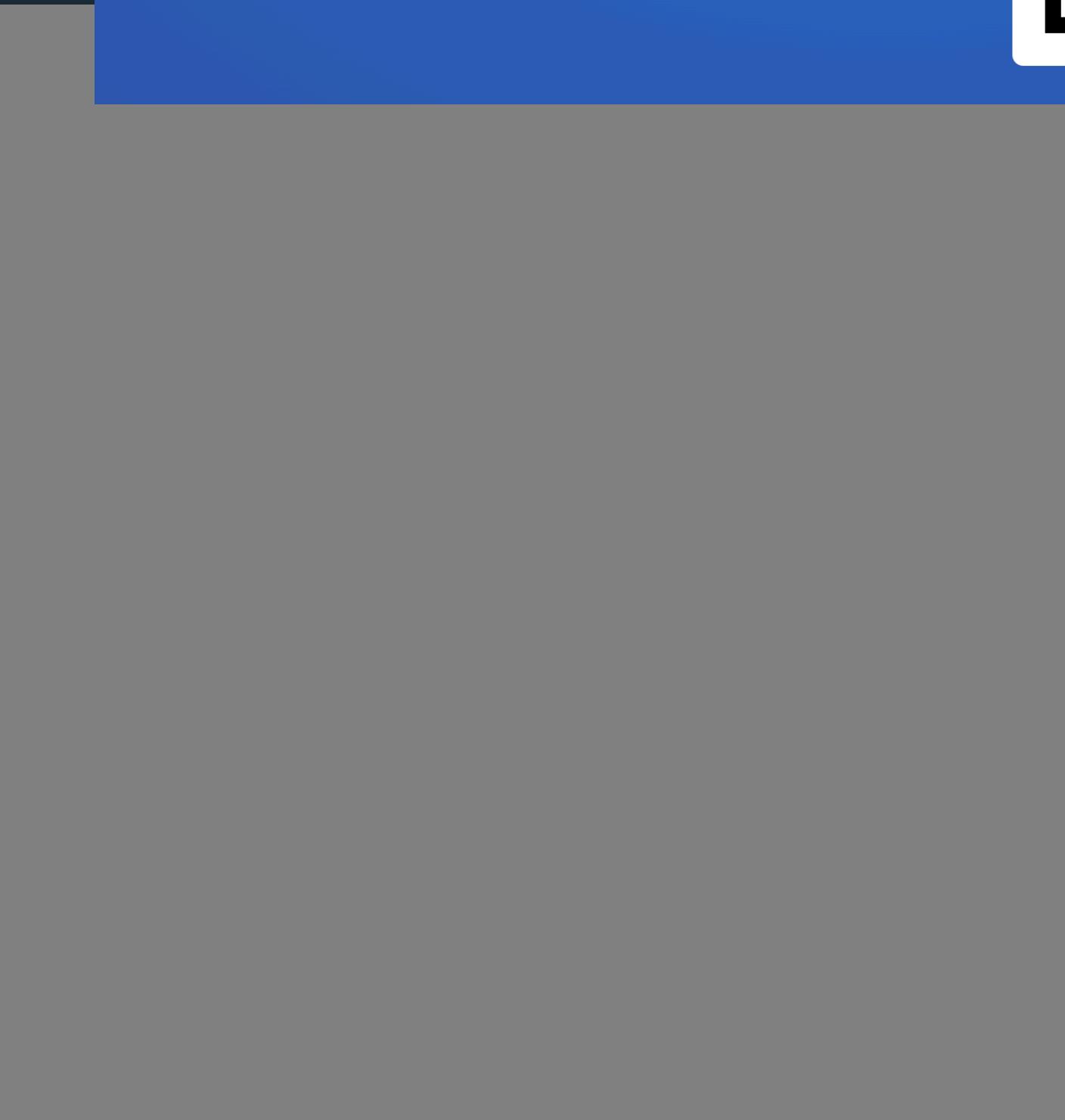


rgb(144, 144, 144)

Which is brighter?



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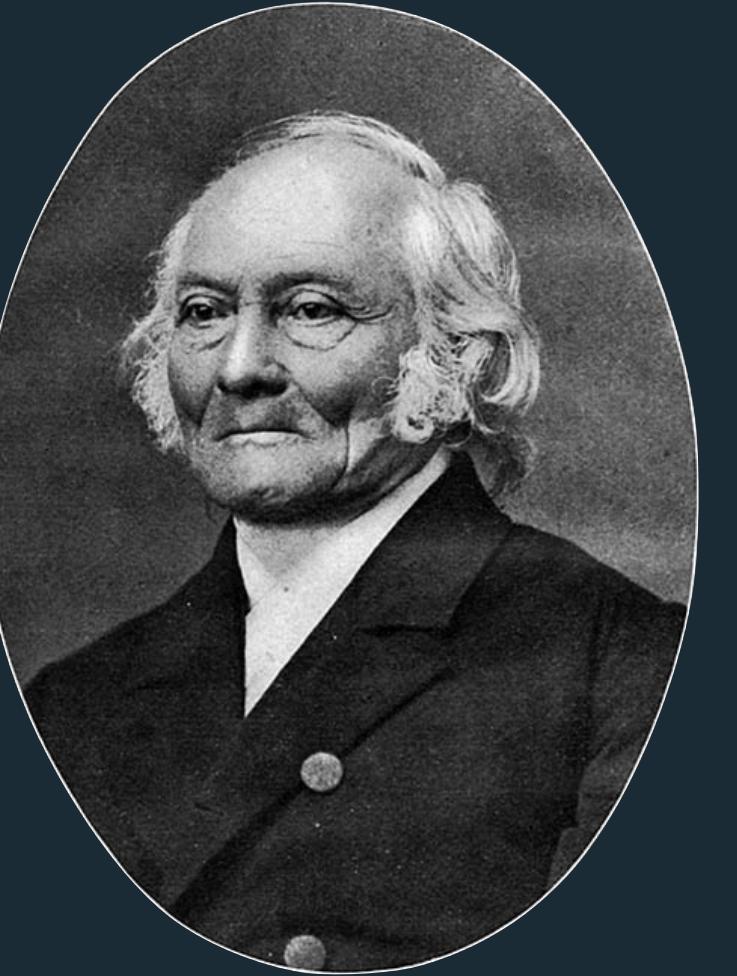


Just Noticeable Difference (jnd)

Scale Factor
(Determined Empirically)

$$\Delta S = k \frac{\Delta I}{I}$$

Perceived Change Change of Intensity Physical Intensity



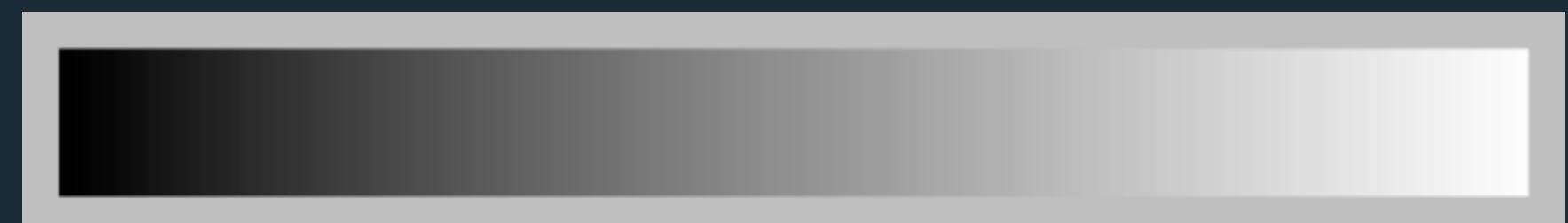
Ernst Weber

(1795 – 1878)

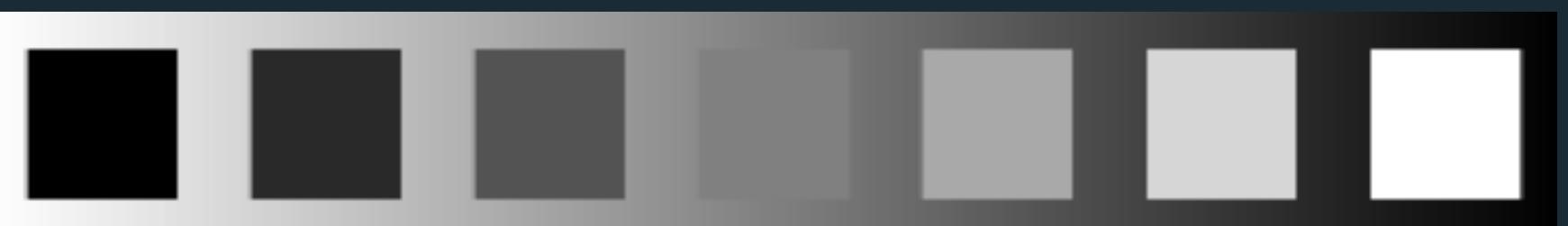
German physician
and a founder of
experimental
psychology.

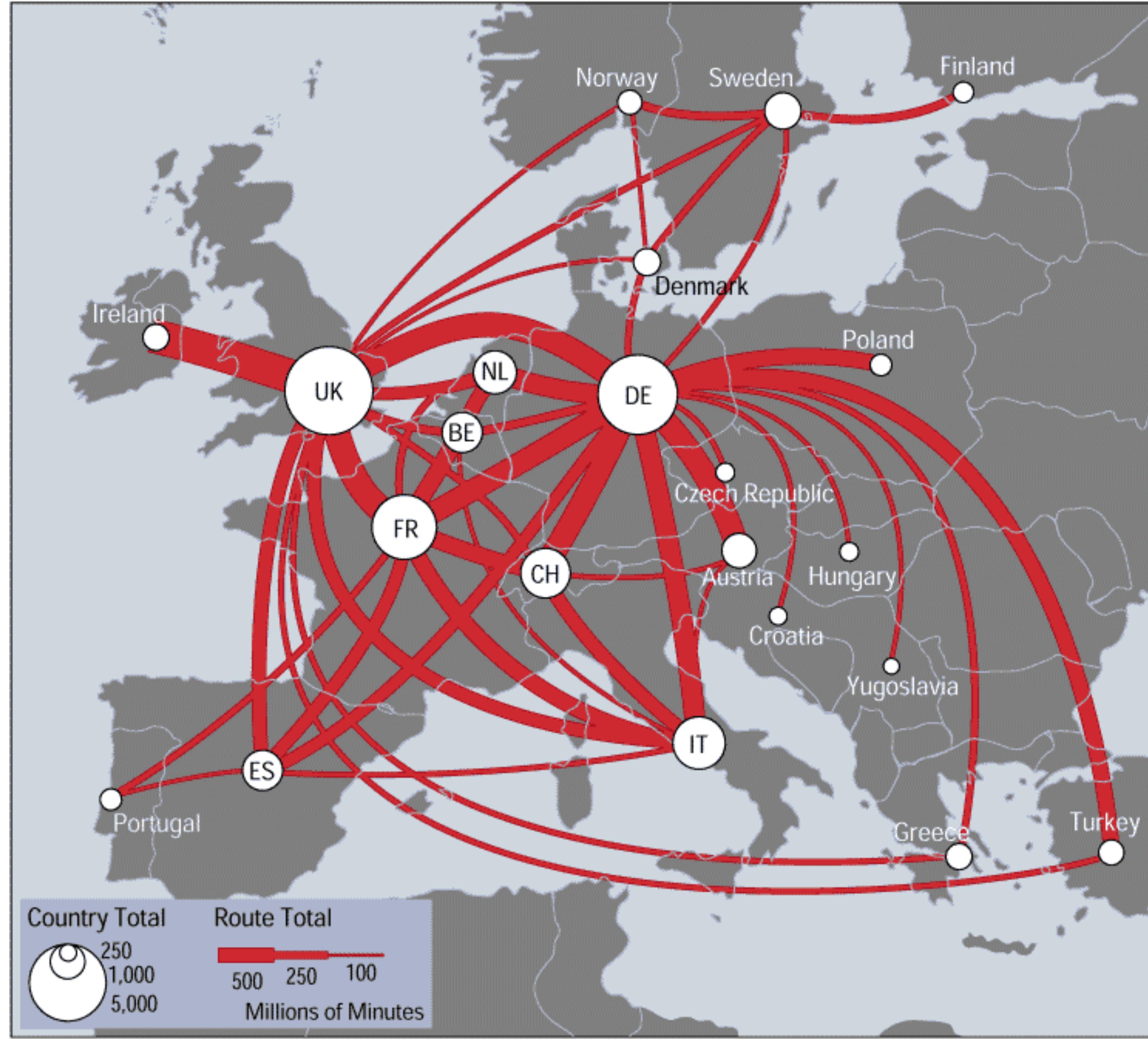
Ratios more important than magnitude.

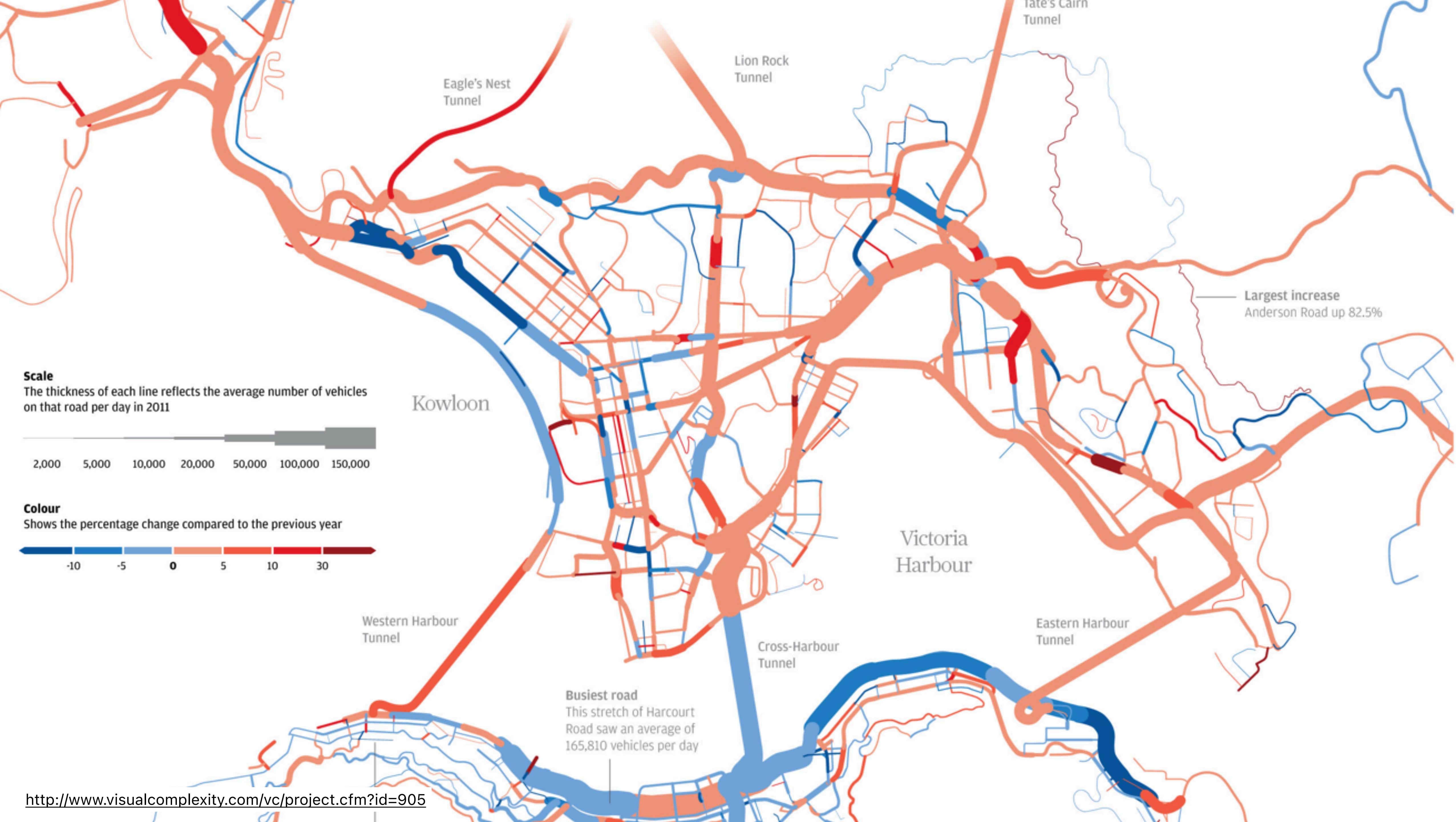
Most continuous variation in stimuli are perceived in discrete steps.



vs.







Signal Detection

Discriminability: how easy is it
to tell two things apart?

Magnitude Estimation

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Selective Attention

Gestalt Grouping

Signal Detection

Magnitude Estimation

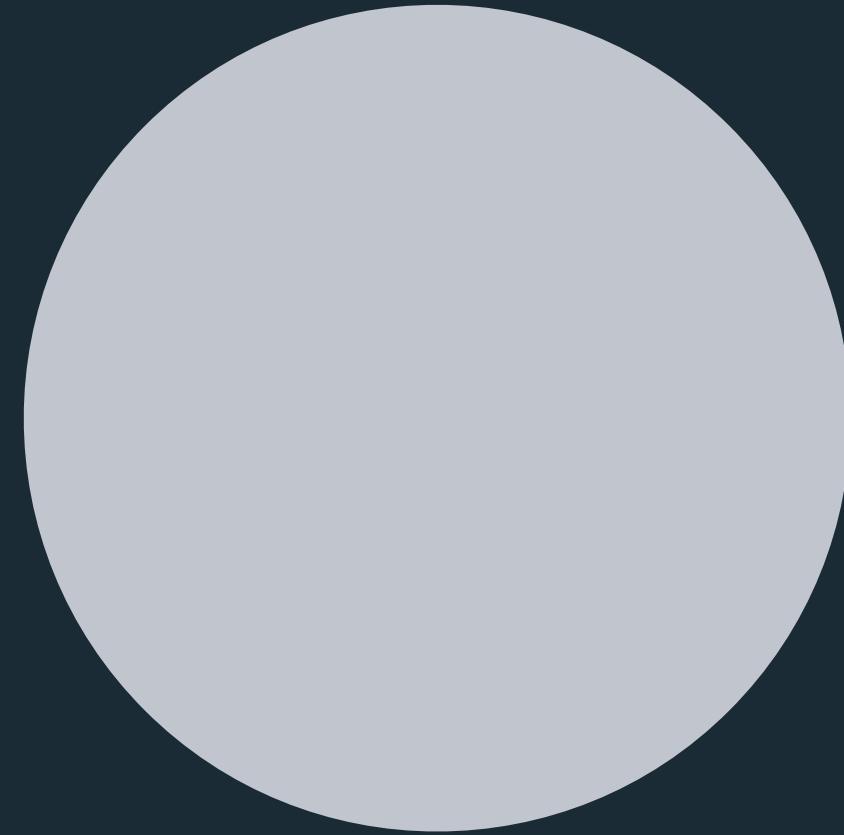
Accuracy: how correctly can we read off values?

Pre-Attentive Processing

Selective Attention

Gestalt Grouping

How much larger is the area of the big circle?



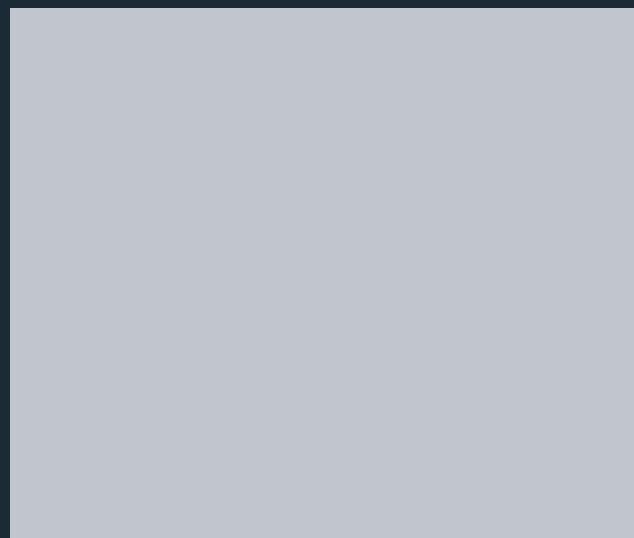
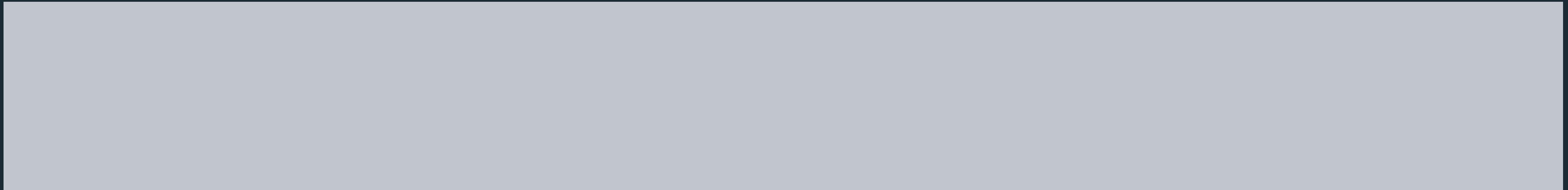
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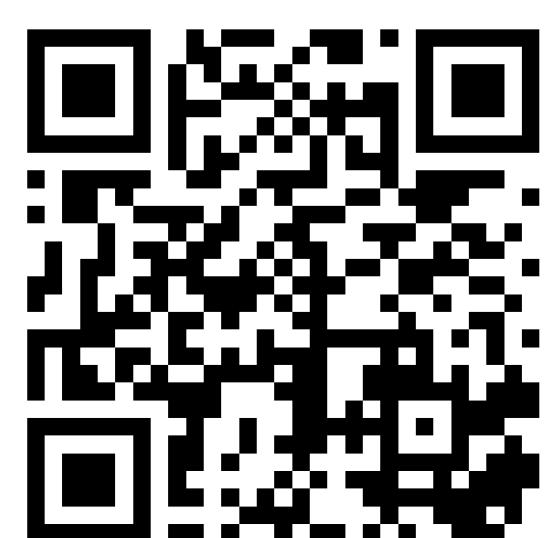
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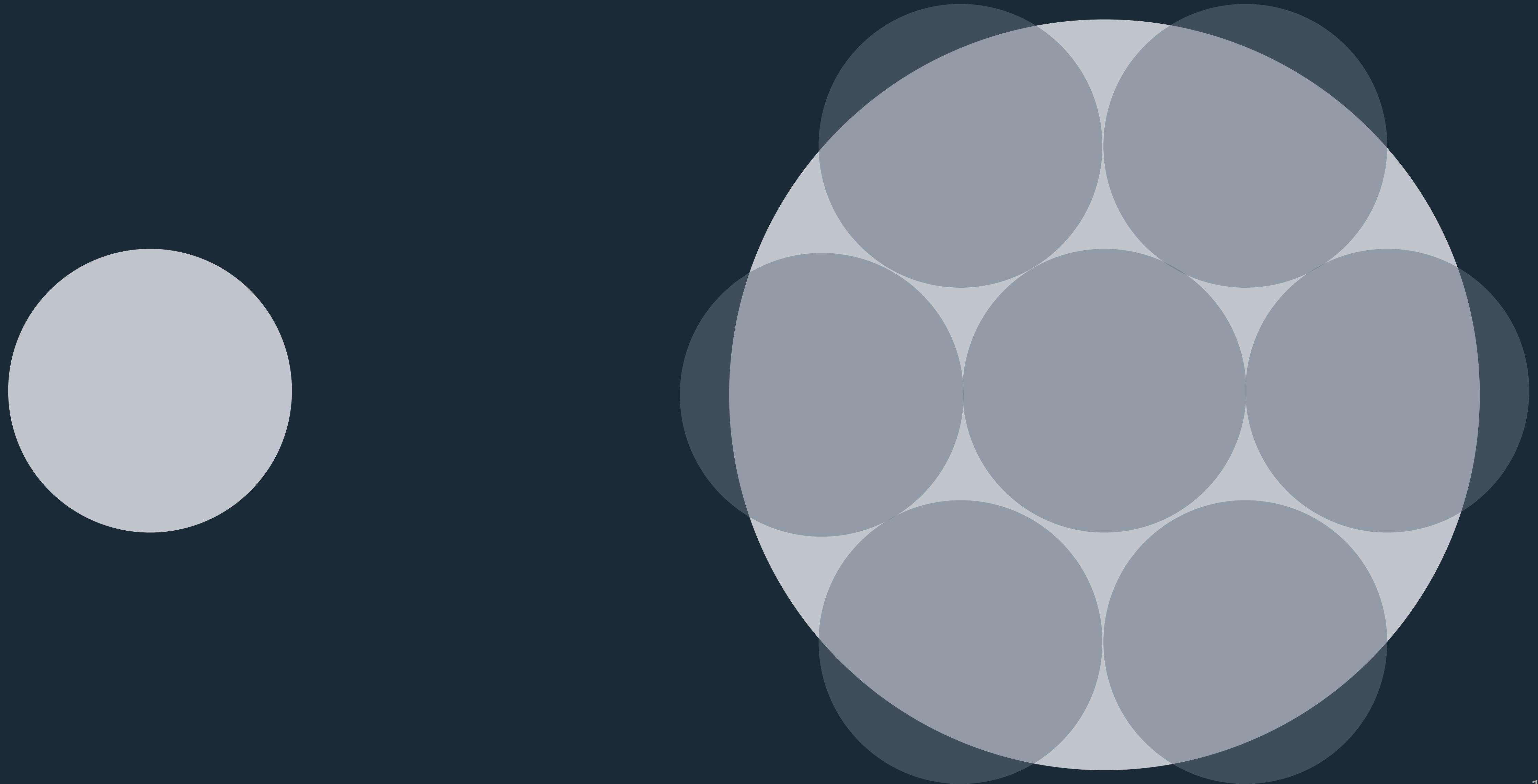
How much longer is the big bar?



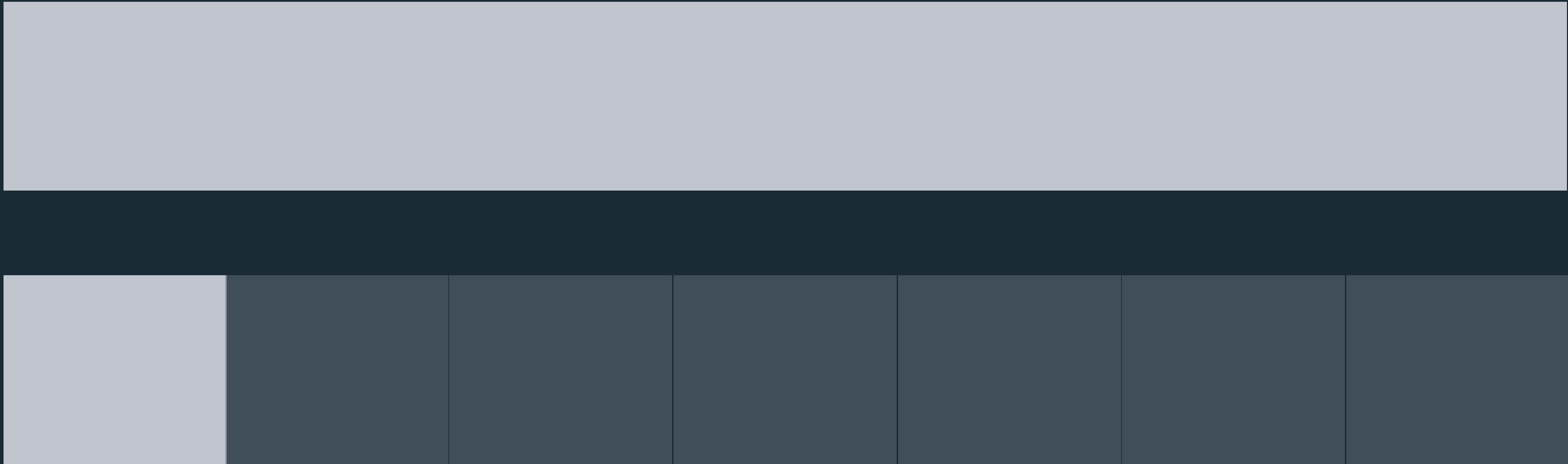
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How much larger is the area of the big circle?



How much longer is the big bar?



Stevens' Power Law



$$S = IP^p$$

Physical Intensity

Perceived Sensation

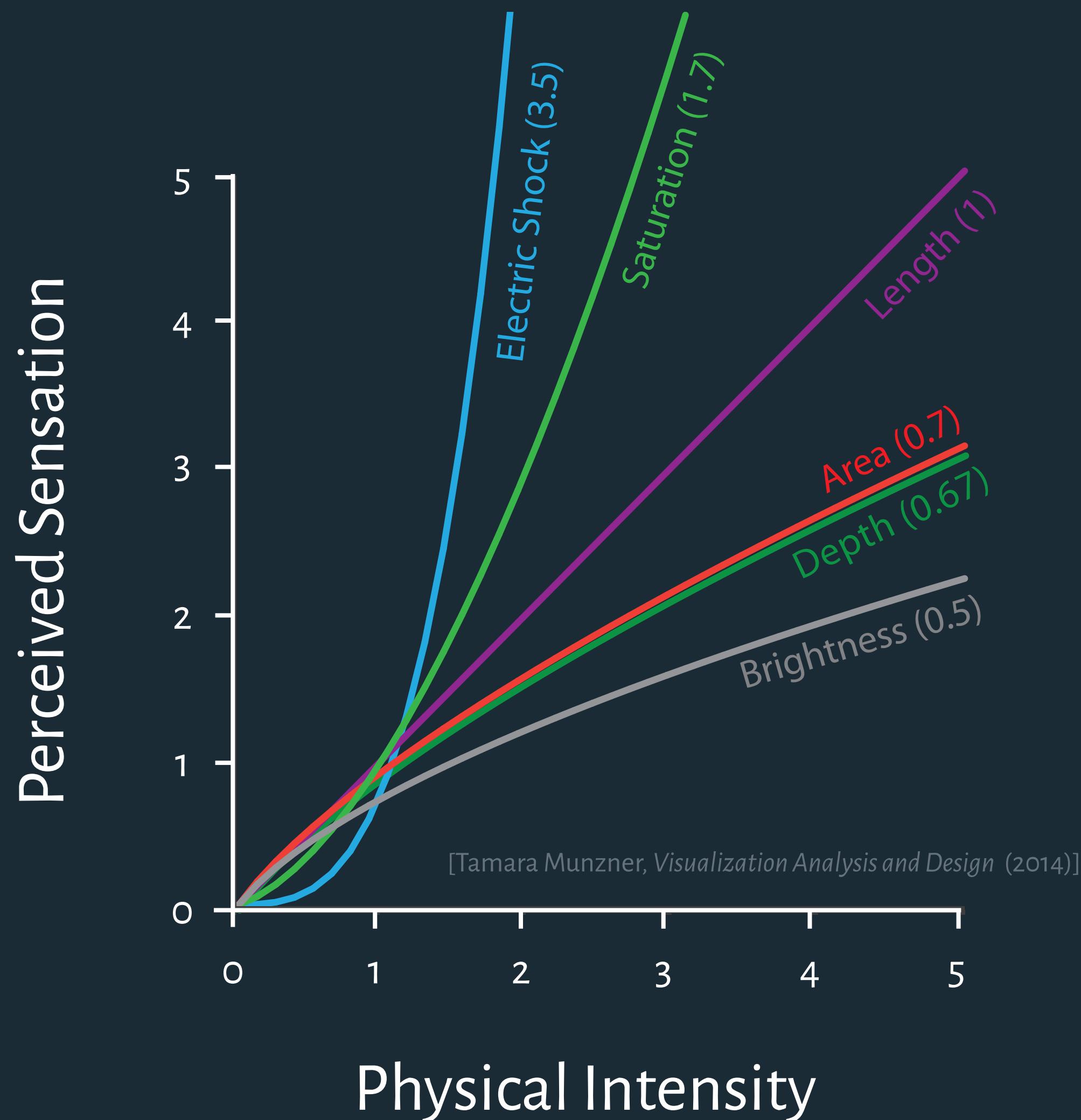
Exponent
(Determined Empirically)

$p < 1$ = underestimation
 $p > 1$ = overestimation

Predicts bias, not necessarily accuracy!

S. S. Stevens (1906 – 1972)

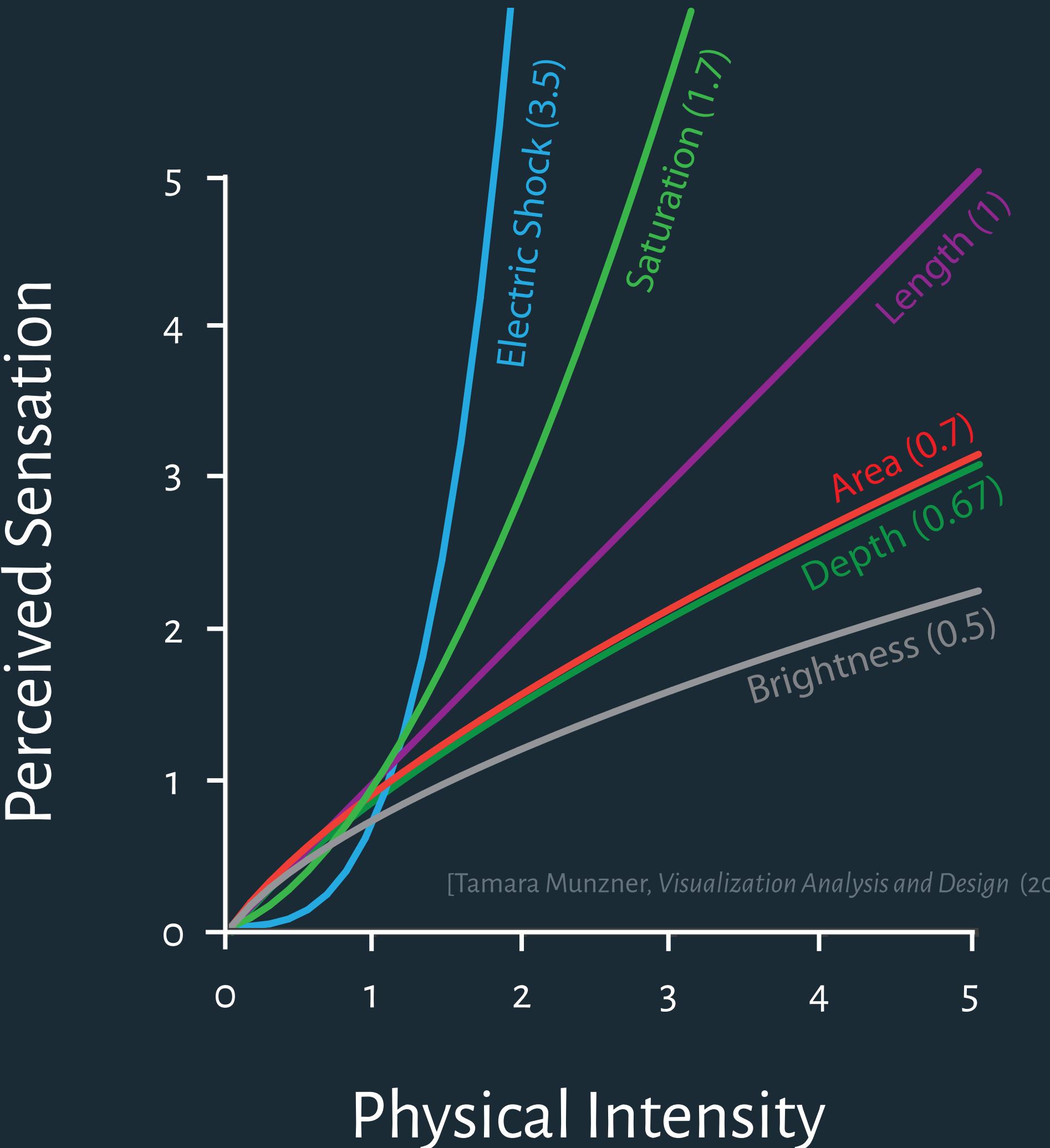
American psychologist, founded Harvard's Psychoacoustics Lab.



Stevens' Power Law

$$S = I^P$$

Sensation	Exponent
Loudness	0.6
Brightness	0.33
Smell	0.55 (Coffee) – 0.6 (Heptane)
Taste	0.6 (Saccharin) – 1.3 (Salt)
Temperature	1.0 (Cold) – 1.6 (Warm)
Vibration	0.6 (250 Hz) – 0.95 (60 Hz)
Duration	1.1
Pressure	1.1
Heaviness	1.45
Electric Shock	3.5



Graphical Perception Studies

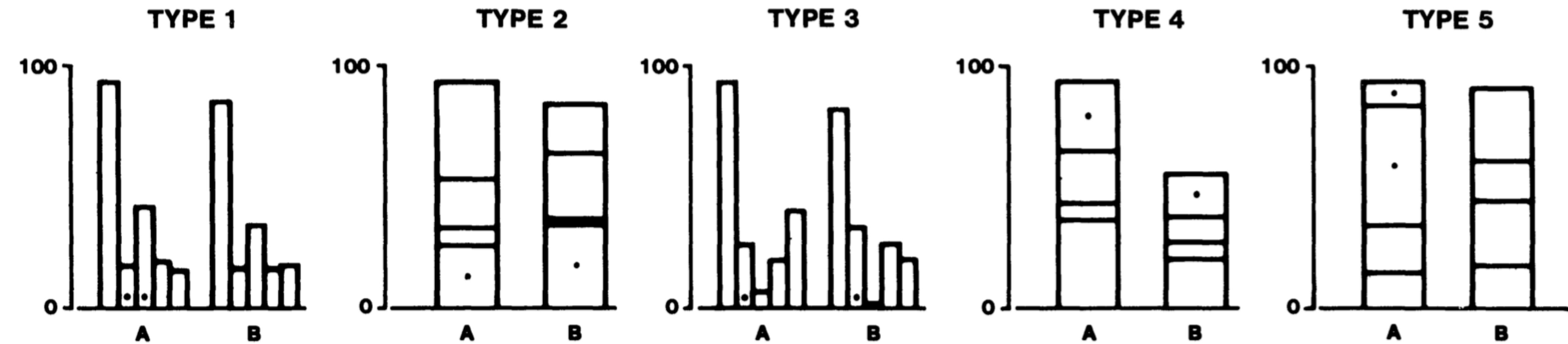


Figure 4. Graphs from position-length experiment.

What proportion is the smaller marked section of the larger?

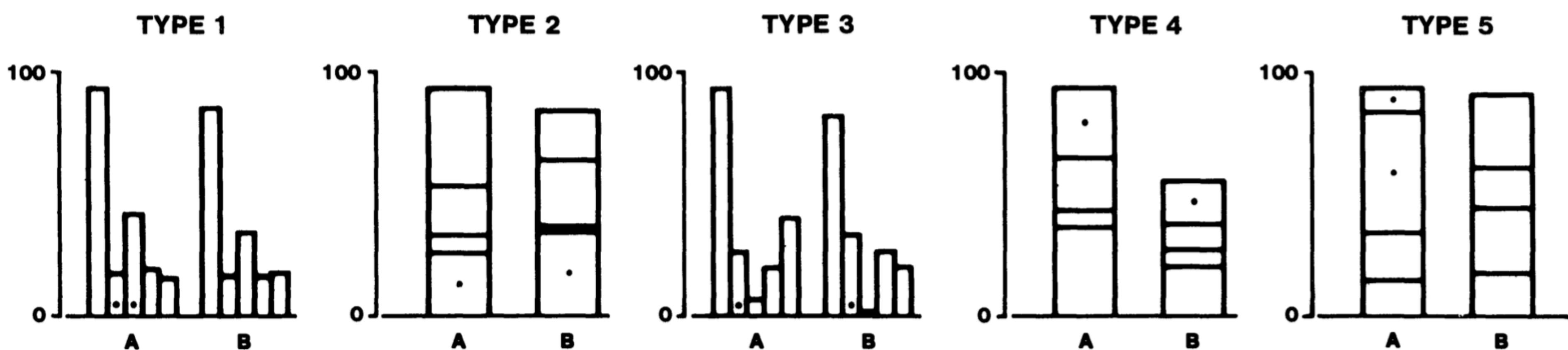
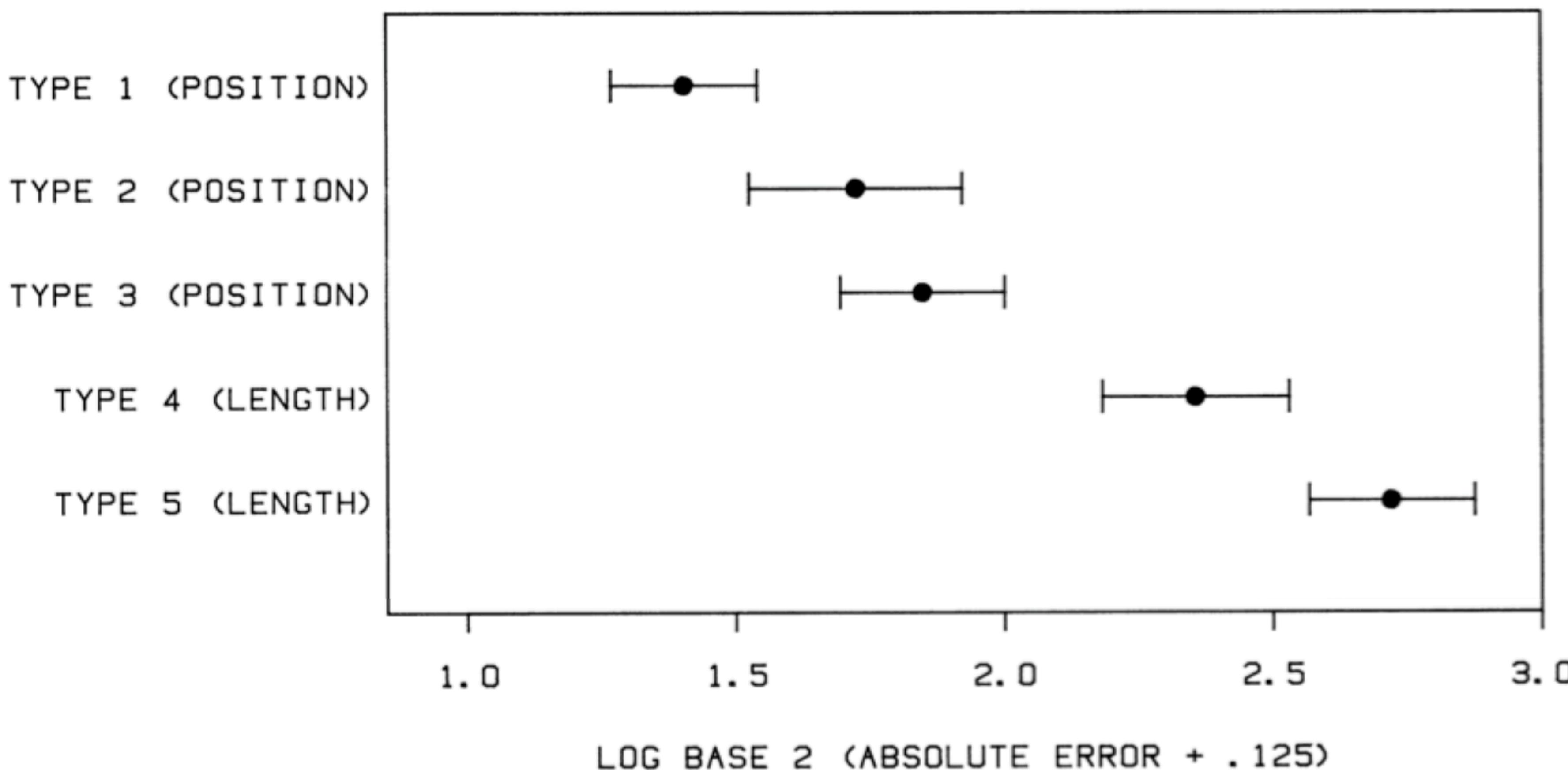
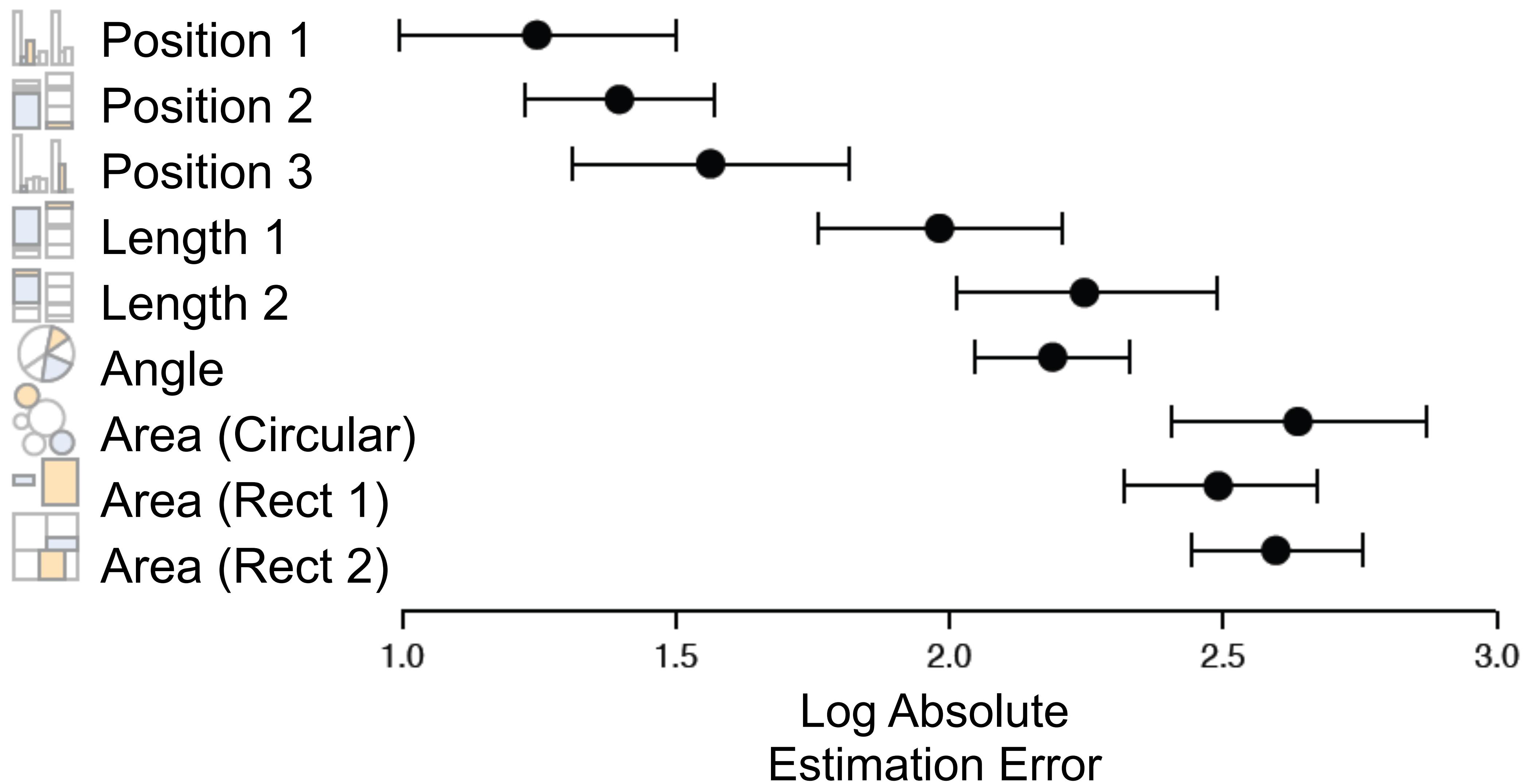


Figure 4. Graphs from position-length experiment.





Signal Detection

Magnitude Estimation

Pre-Attentive Processing

Selective Attention

Gestalt Grouping

Accuracy: how correctly can we read off values?

Signal Detection

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Pop Out: how easy is it to spot some values from the rest?

How many 3's?

1281768756138976546984506985604982826762
9809858458224509856458945098450980943585
90910302099059595772564675050678904567
8845789809821677654876364908560912949686

How many 3's?

1281768756138976546984506985604982826762
9809858458224509856458945098450980943585
90910302099059595772564675050678904567
8845789809821677654876364908560912949686

Pre-Attentive Processing

How immediately does our visual system perceive differences in a scene?

Pre-Attentive: immediately recognize variation with little or no conscious effort (<200–250 ms).

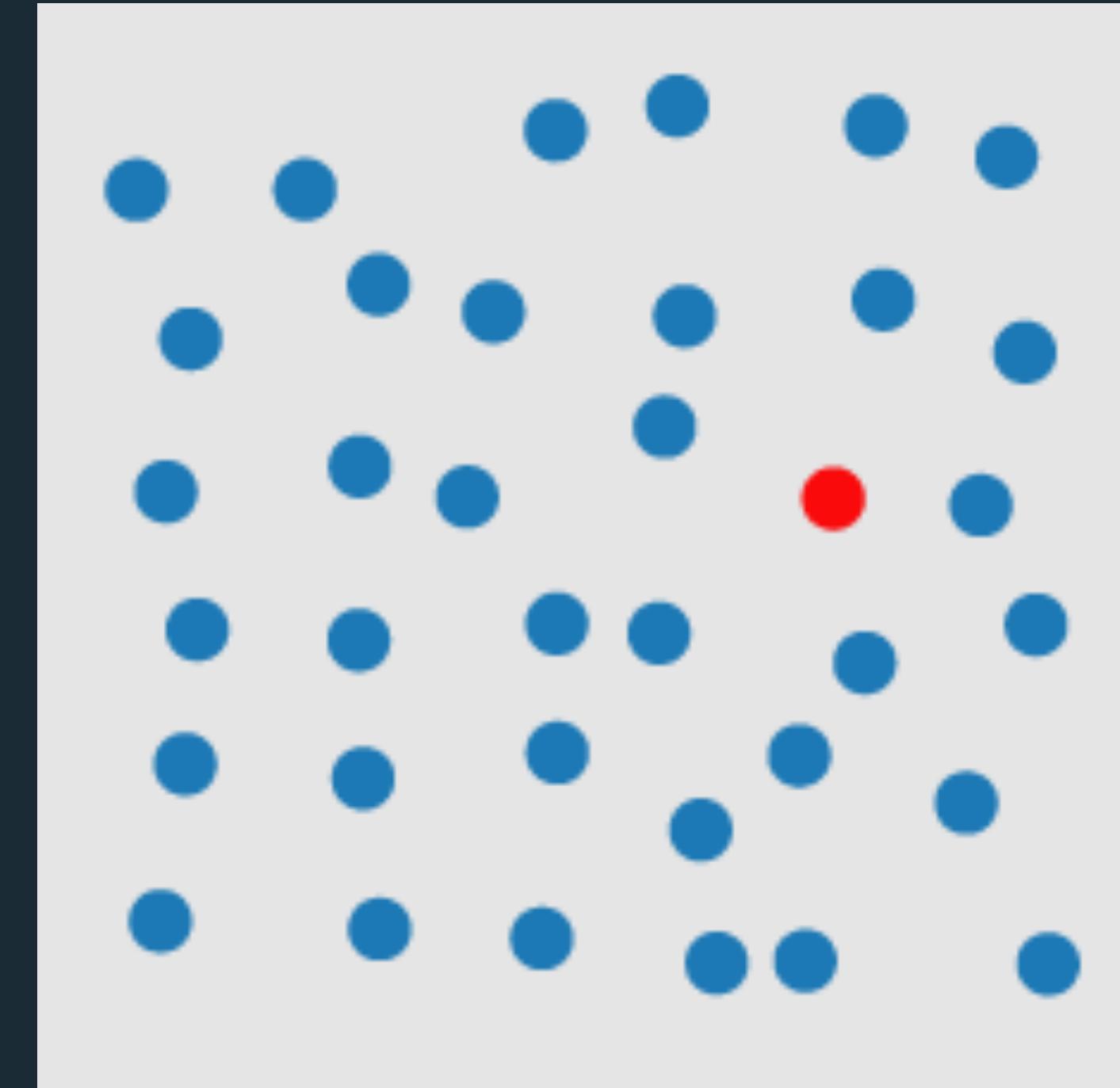
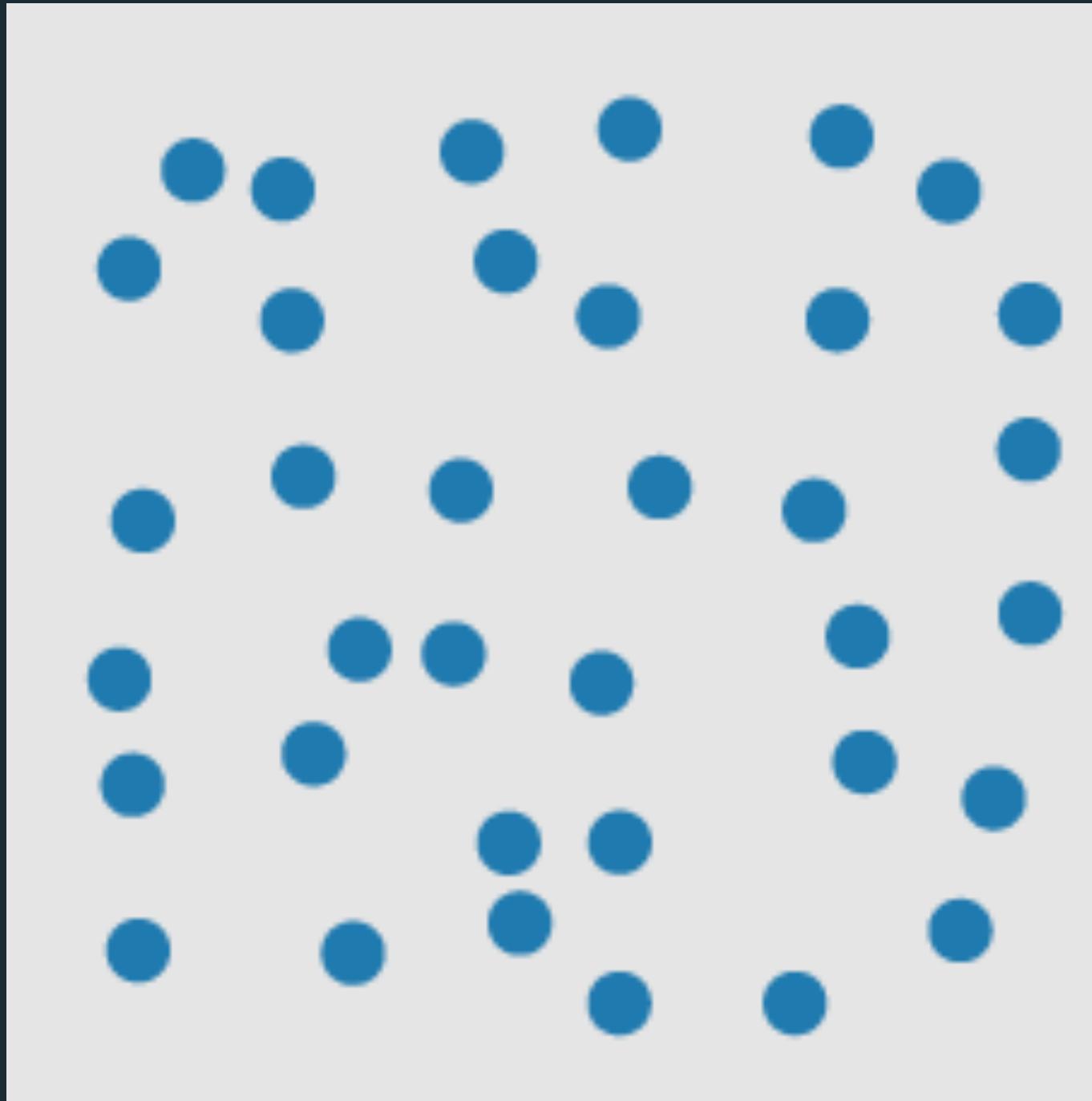
Attentive: Takes some deliberate effort to perceive differences.

Pre-Attentive Processing

Pre-Attentive: immediately recognize variation with little or no conscious effort (<200–250 ms).

Attentive: Takes some deliberate effort to perceive differences.

Visual Pop-Out: Color



[Healey & Enns 2012]

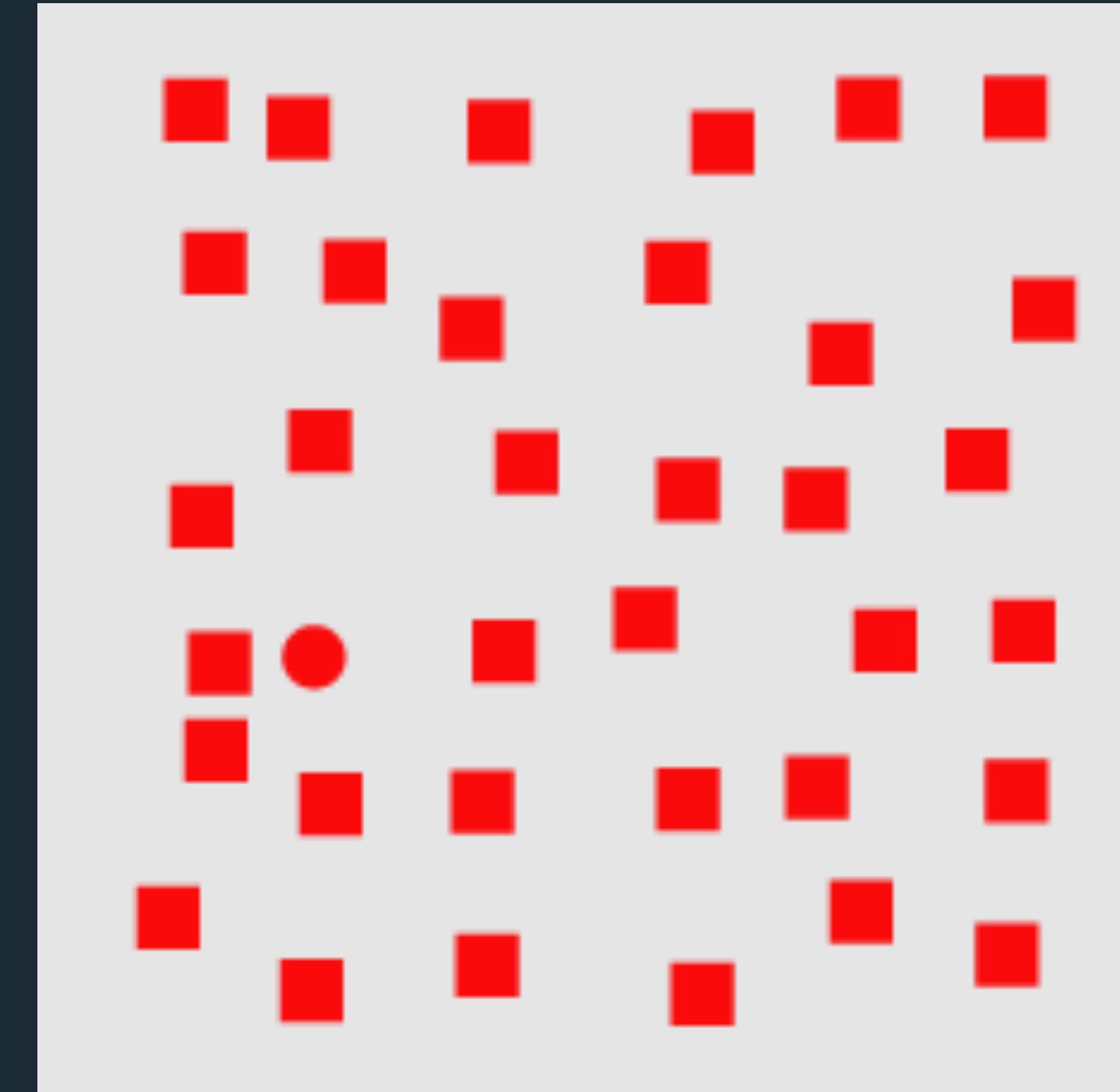
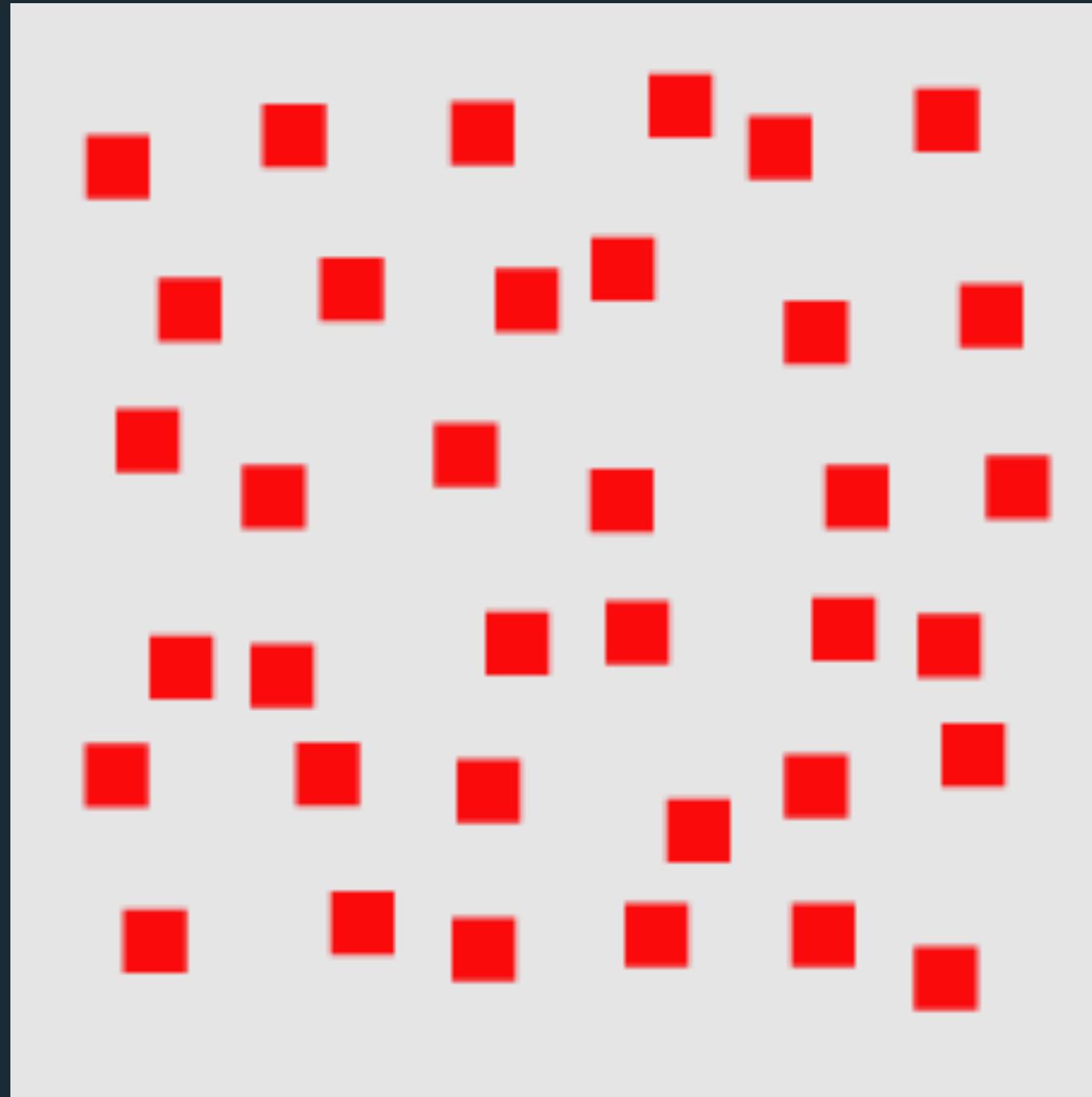
Pre-Attentive Processing

Visual Pop-Out: Color

Pre-Attentive: immediately recognize variation with little or no conscious effort (<200–250 ms).

Attentive: Takes some deliberate effort to perceive differences.

Visual Pop-Out: Shape



[Healey & Enns 2012]

Pre-Attentive Processing

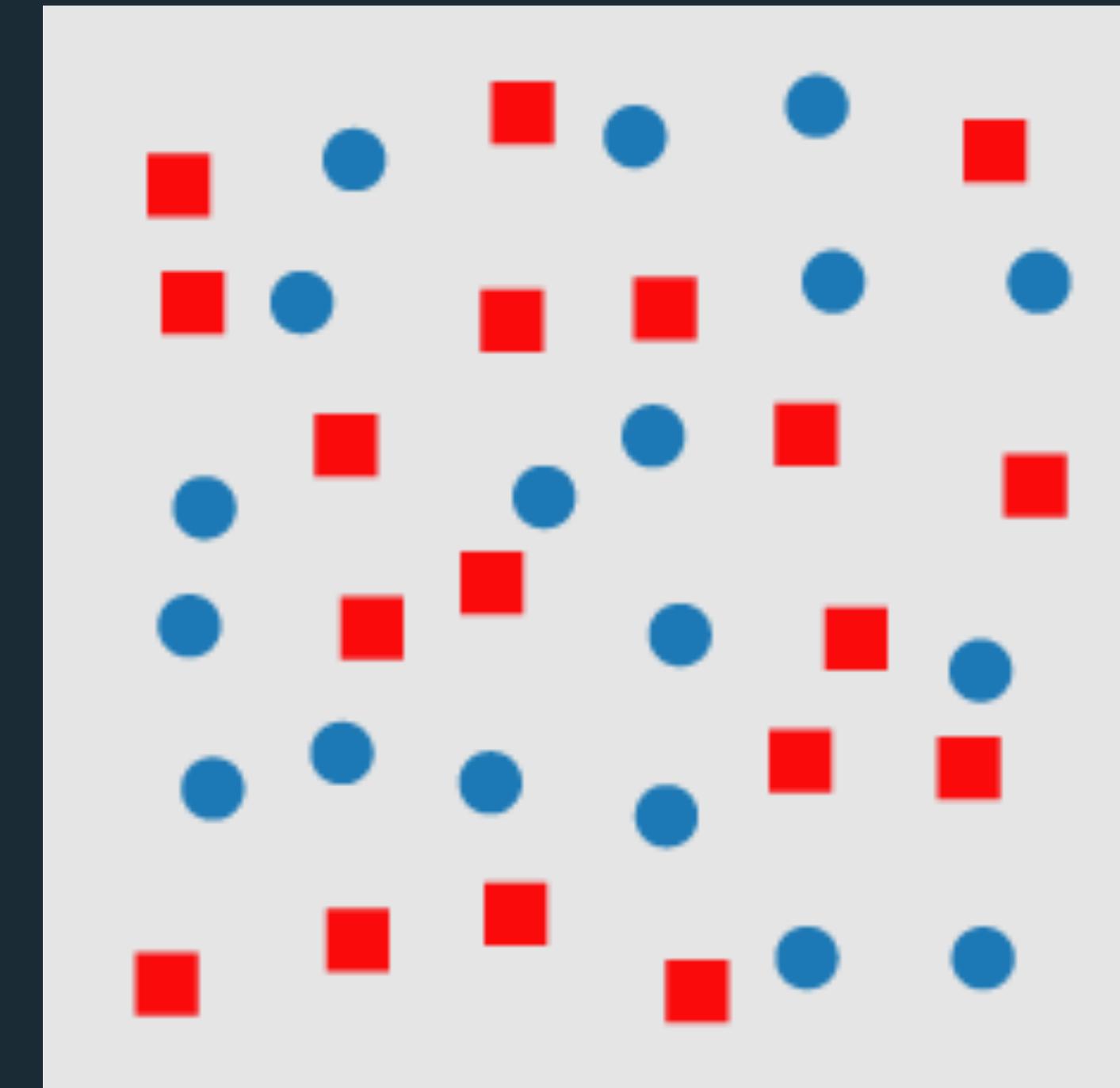
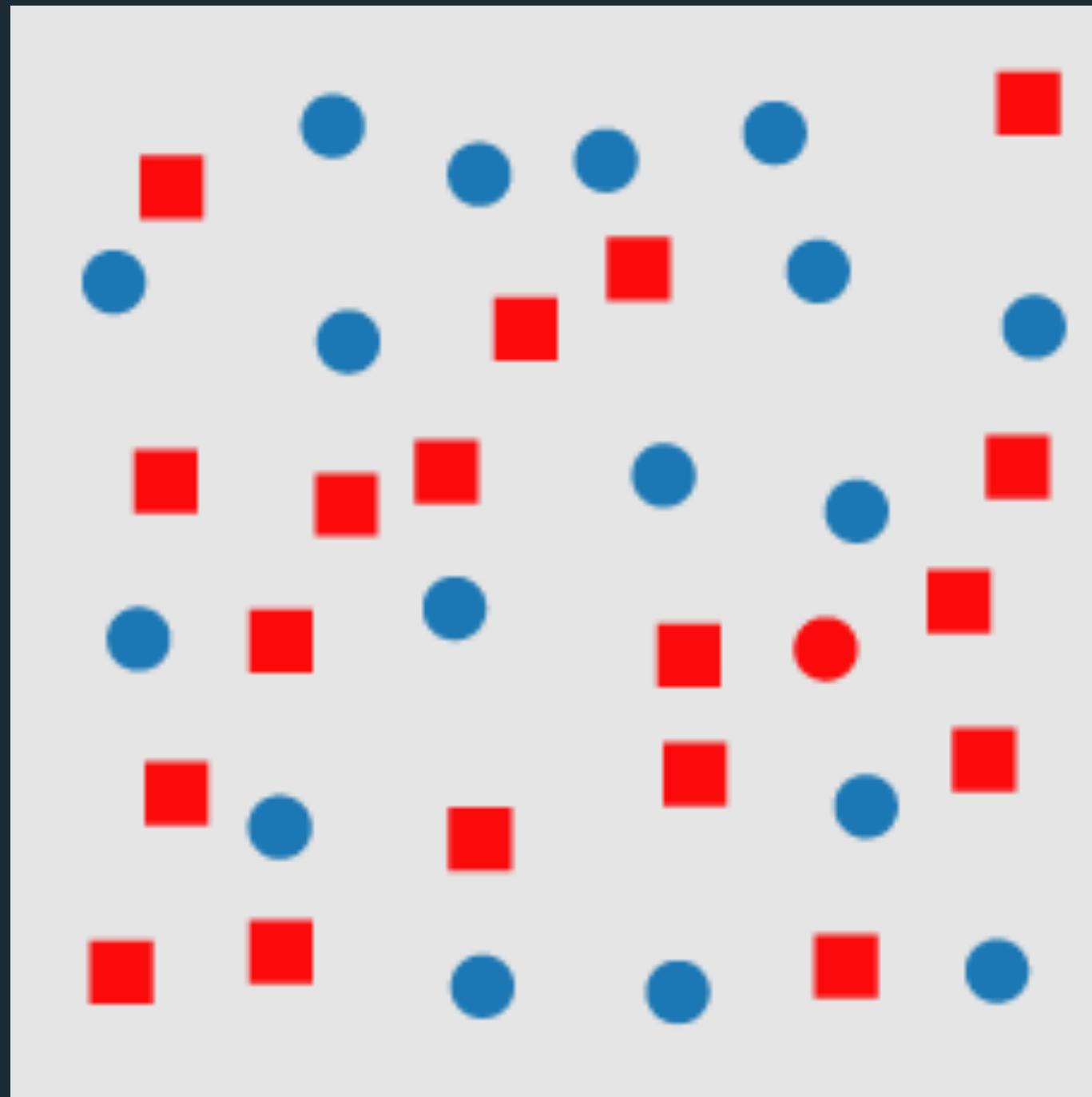
Visual Pop-Out: Color

Visual Pop-Out: Shape

Pre-Attentive: immediately recognize variation with little or no conscious effort (<200–250 ms).

Attentive: Takes some deliberate effort to perceive differences.

Feature Conjunctions



[Healey & Enns 2012]

Pre-Attentive Processing

Visual Pop-Out: Color Visual Pop-Out: Shape Feature Conjunctions

Conjunctions are *not* pre-attentive except for spatial conjunctions:

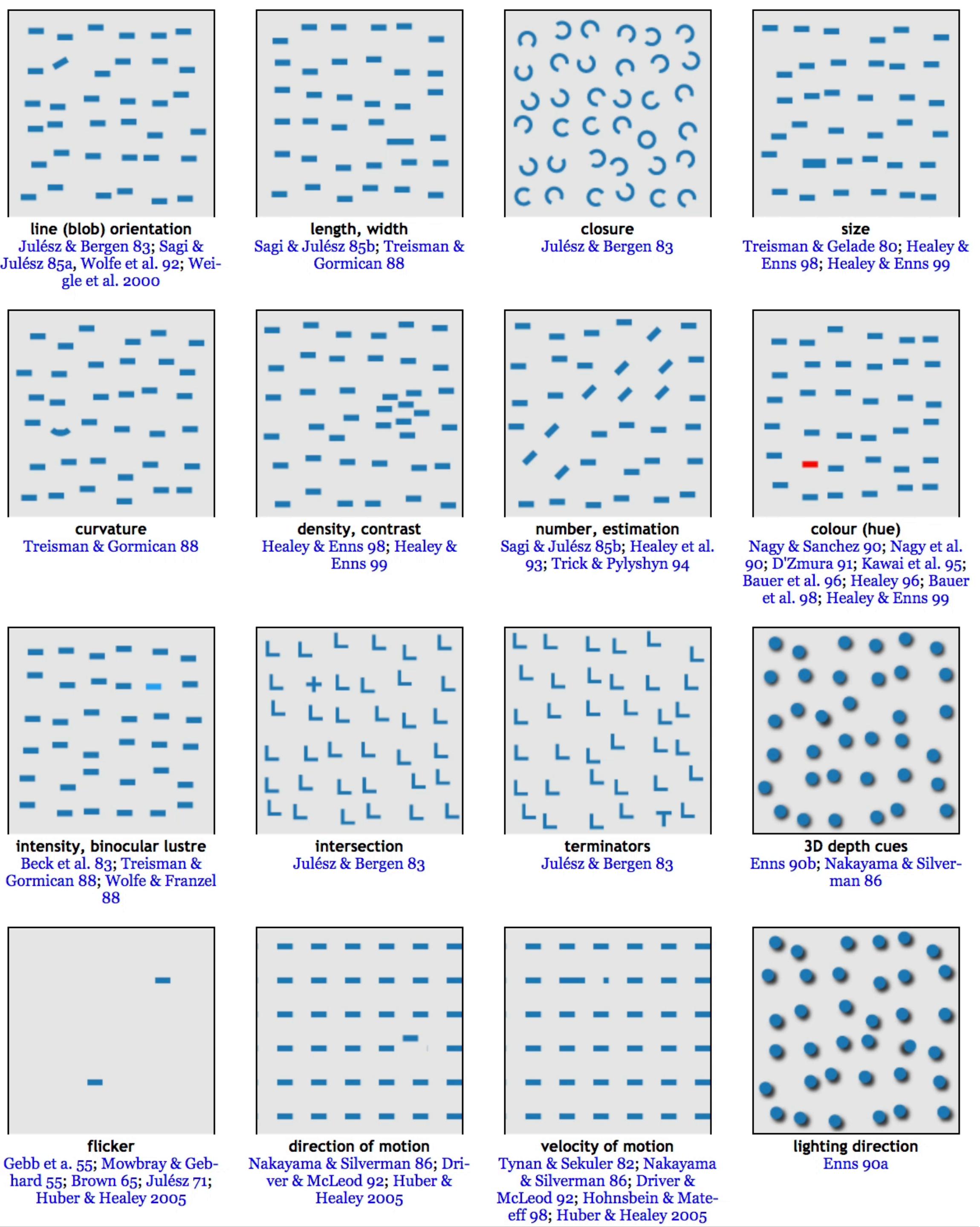
Motion & 3D disparity

Motion & color

Motion & shape

3D disparity & color

3D disparity & shape



[Healey & Enns 2012]

Signal Detection

Magnitude Estimation

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Pop Out: how easy is it to spot some values from the rest?

Signal Detection

Magnitude Estimation

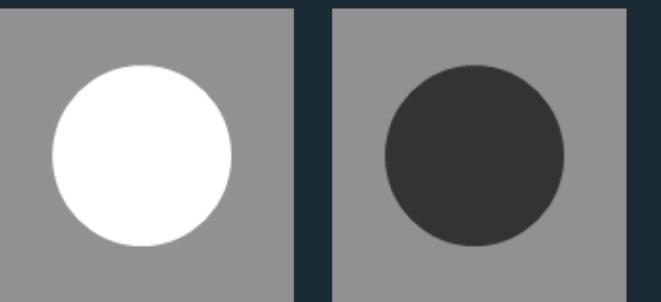
Pre-Attentive Processing

Selective Attention

Gestalt Grouping

Separability: how much interaction occurs between attributes?

One-Dimensional: Lightness



White



White



Black



White



Black



White



Black



Black



White



White

One-Dimensional: Shape



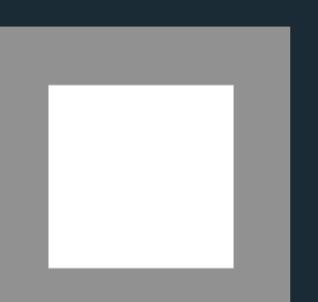
Square



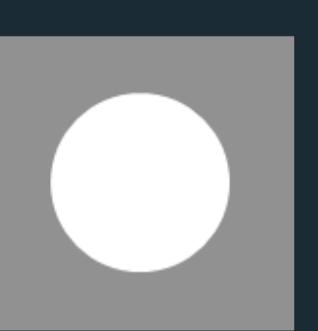
Circle



Circle



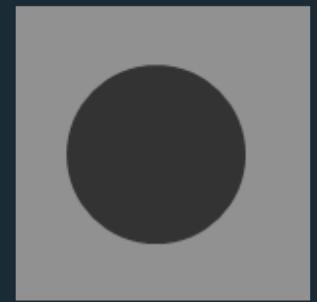
Square



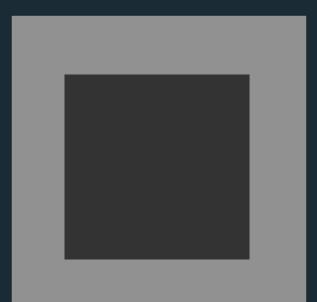
Circle



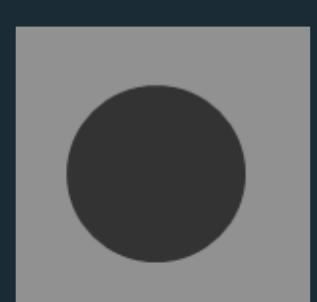
Circle



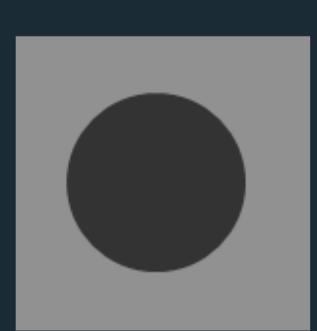
Circle



Square

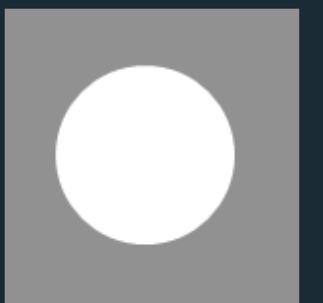


Circle



Circle

Redundant: Shape & Lightness



White



Black



Black



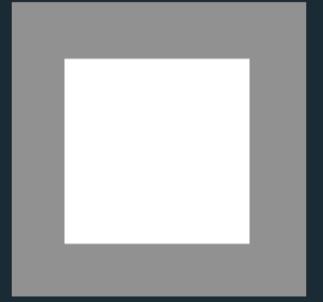
White



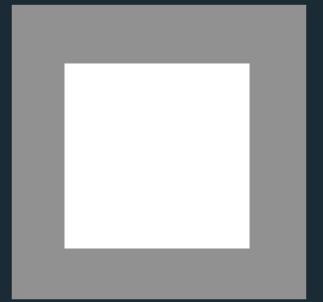
Black



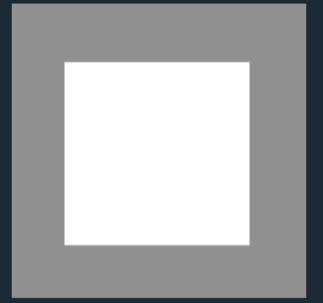
Circle



Square



Square



Square



Circle

Orthogonal: Shape & Lightness



White



Circle

Black



Square

White



Square

Black



Circle

White



Square

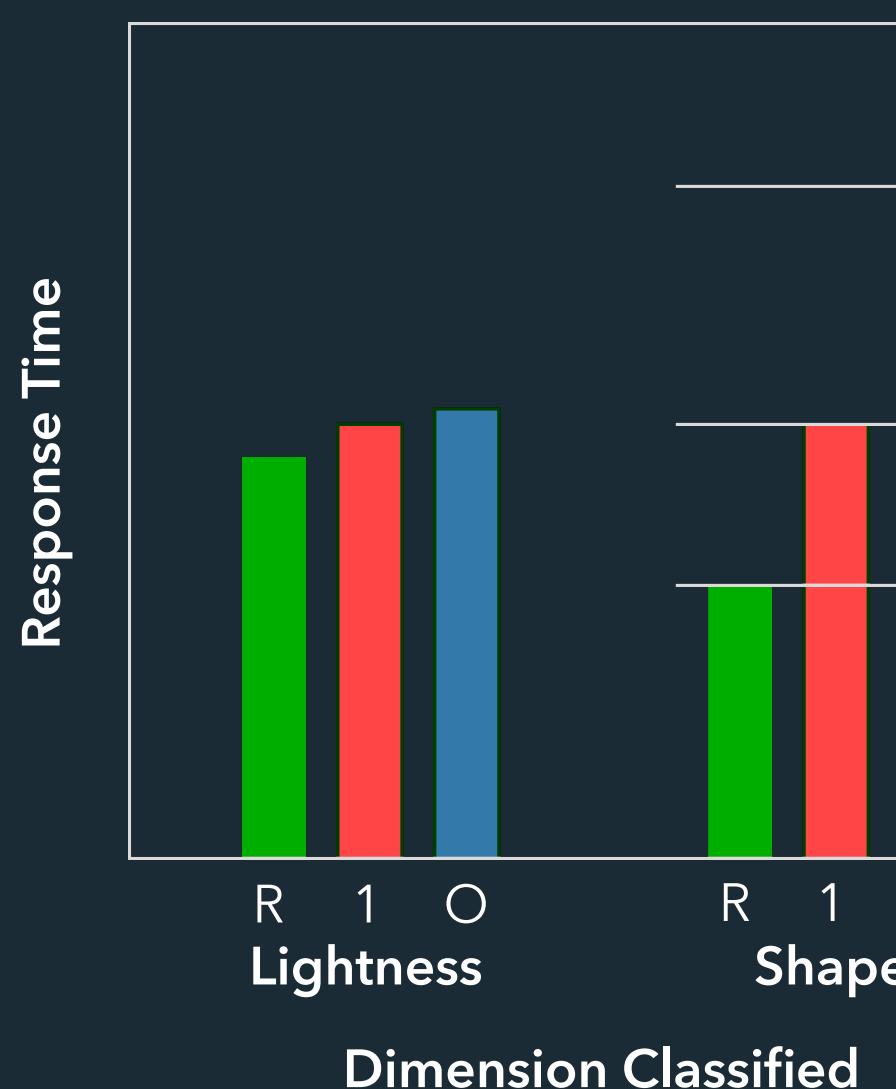
Principles

Redundancy Gain

Improved performance when both dimensions provide the same information.

Filtering Interference

Difficulty in ignoring one dimension while attending to another.



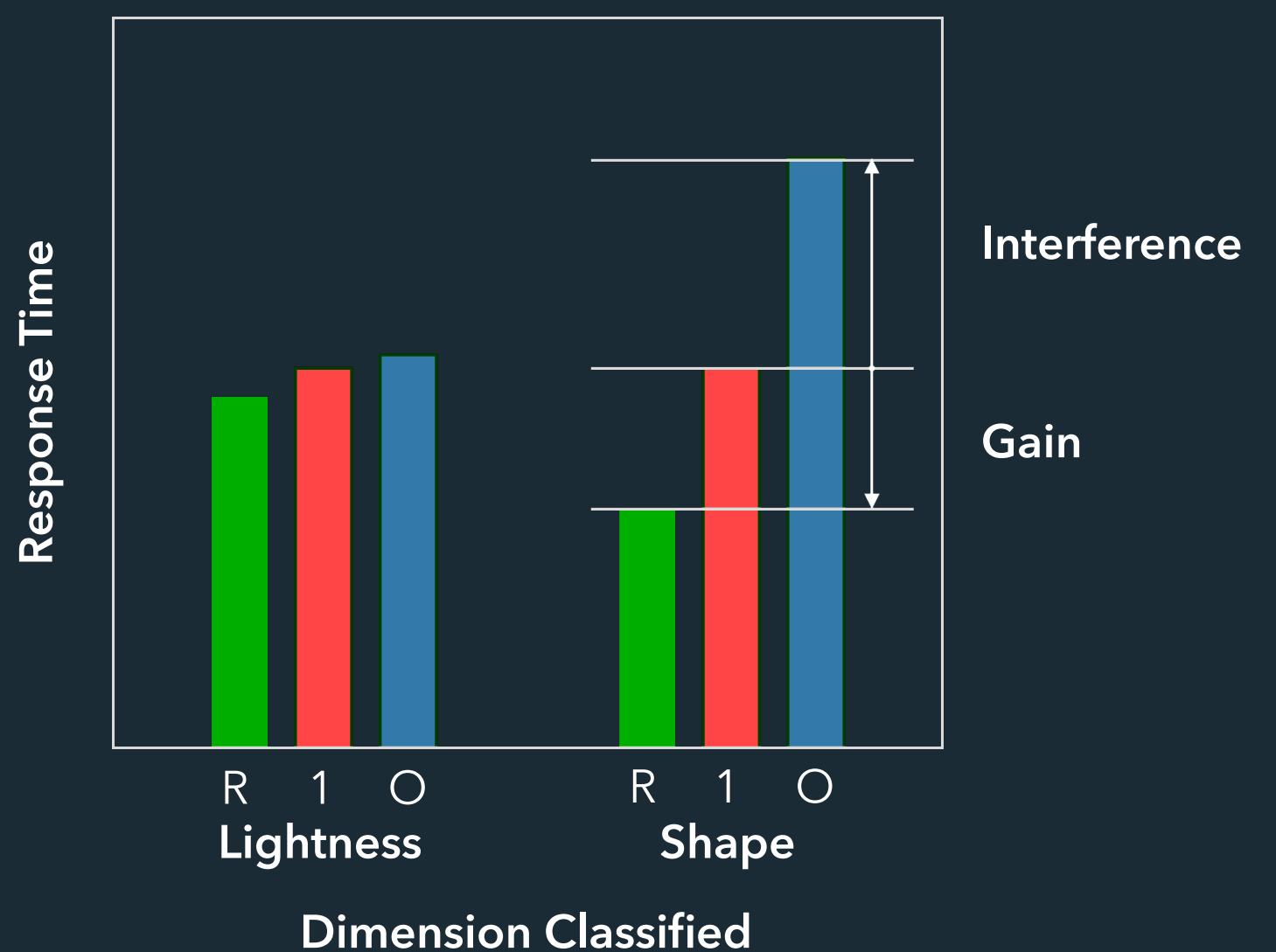
Principles

Redundancy Gain

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Filtering Interference

Difficulty in ignoring one dimension while attending to another.



Types of Dimensions

Separable

No interference or redundancy gain.

Integral

Filtering interference and redundancy gain.

Configural

Only interference. No redundancy gain.

Asymmetric

One dimension is separable from the other, but not vice versa.

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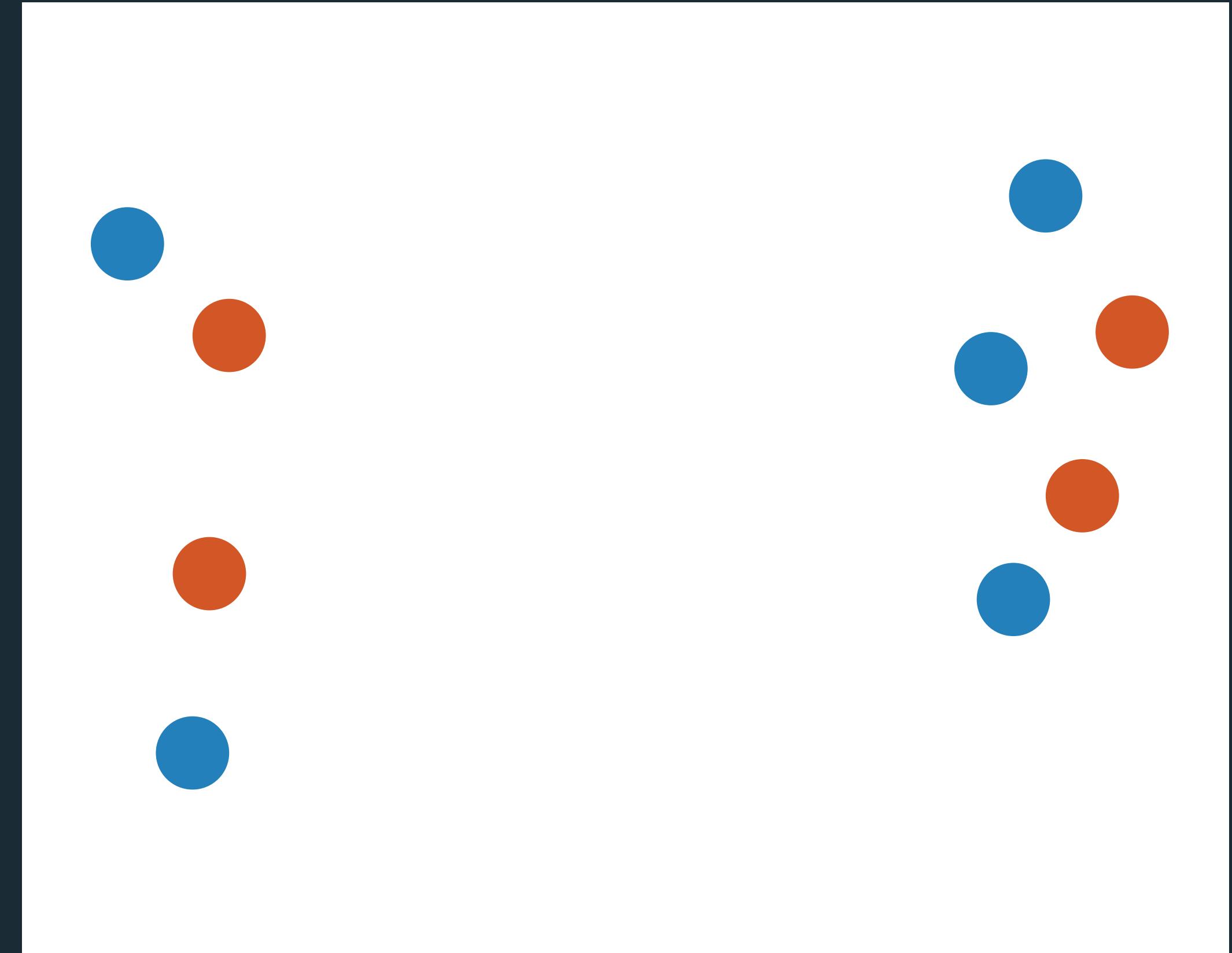
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Position & Hue (Color)?



[Tamara Munzner, *Visualization Analysis and Design* (2014)]

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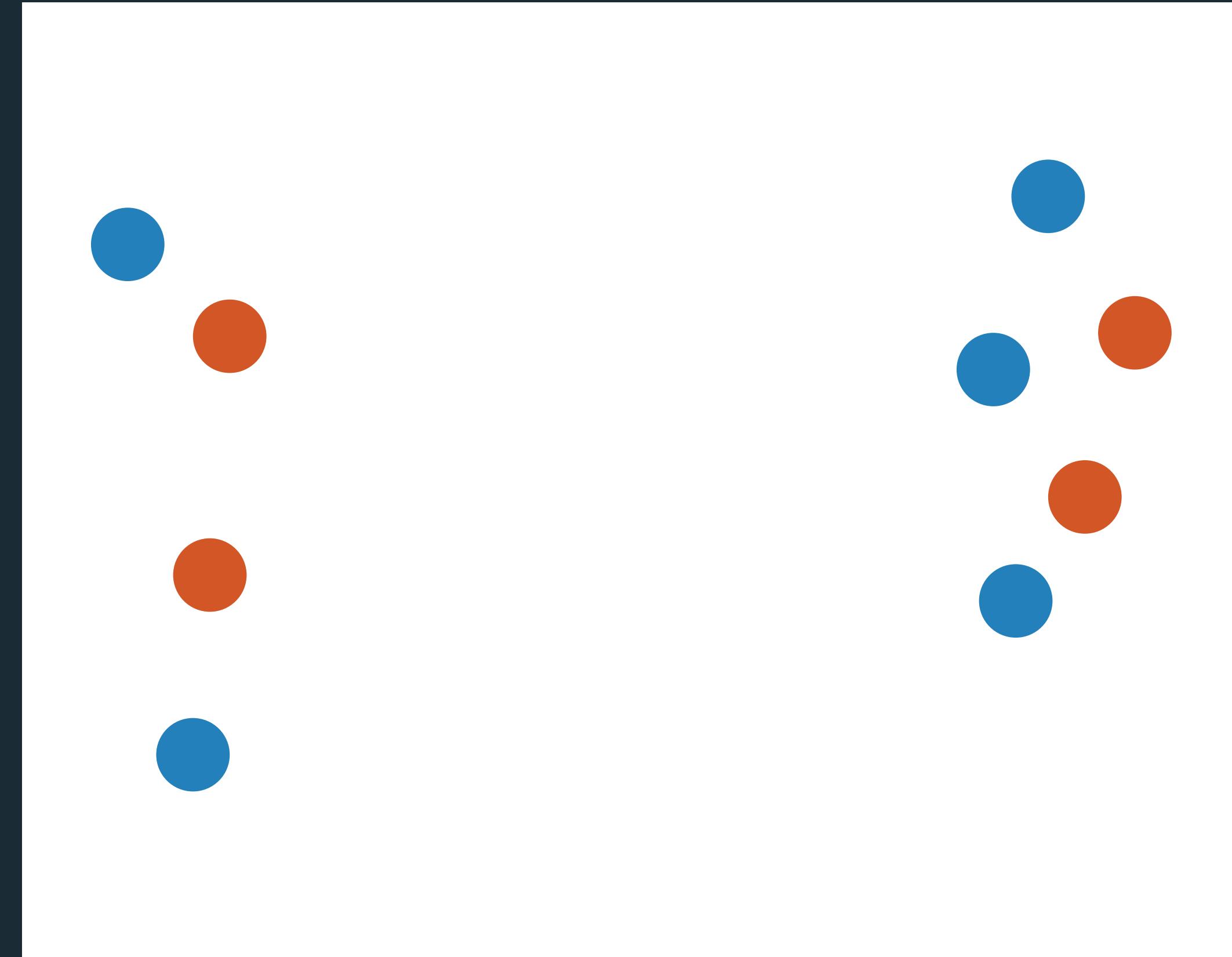
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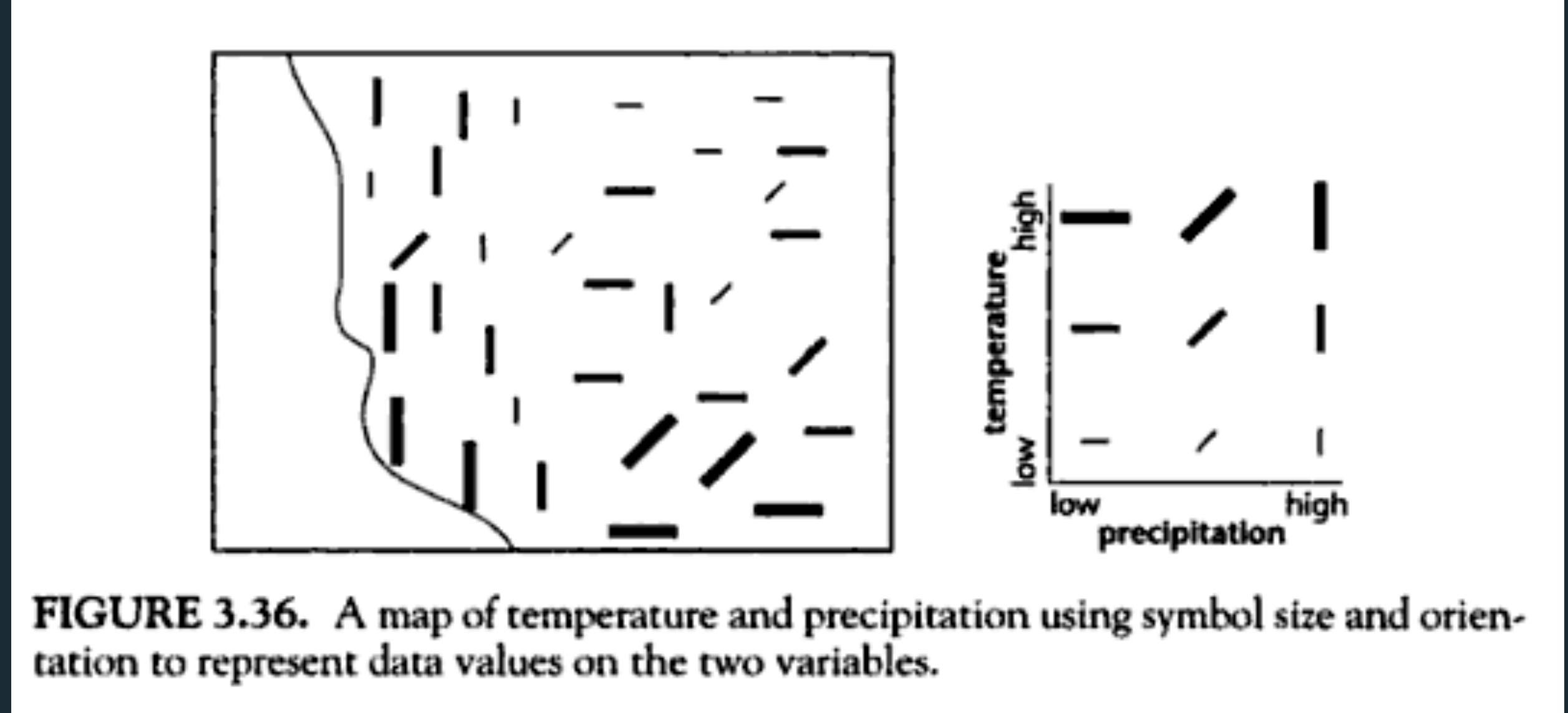


FIGURE 3.36. A map of temperature and precipitation using symbol size and orientation to represent data values on the two variables.

[MacEachren 1995]

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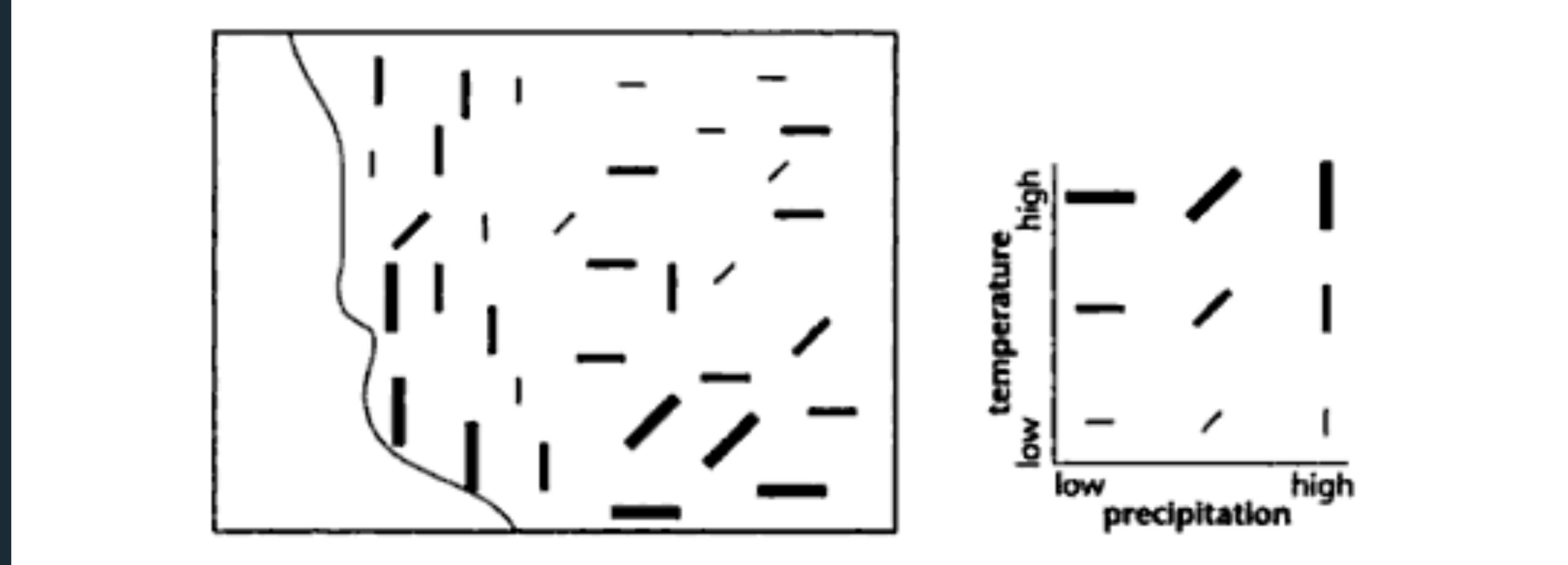


FIGURE 3.36. A map of temperature and precipitation using symbol size and orientation to represent data values on the two variables.

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Types of Dimensions

Size & Value?

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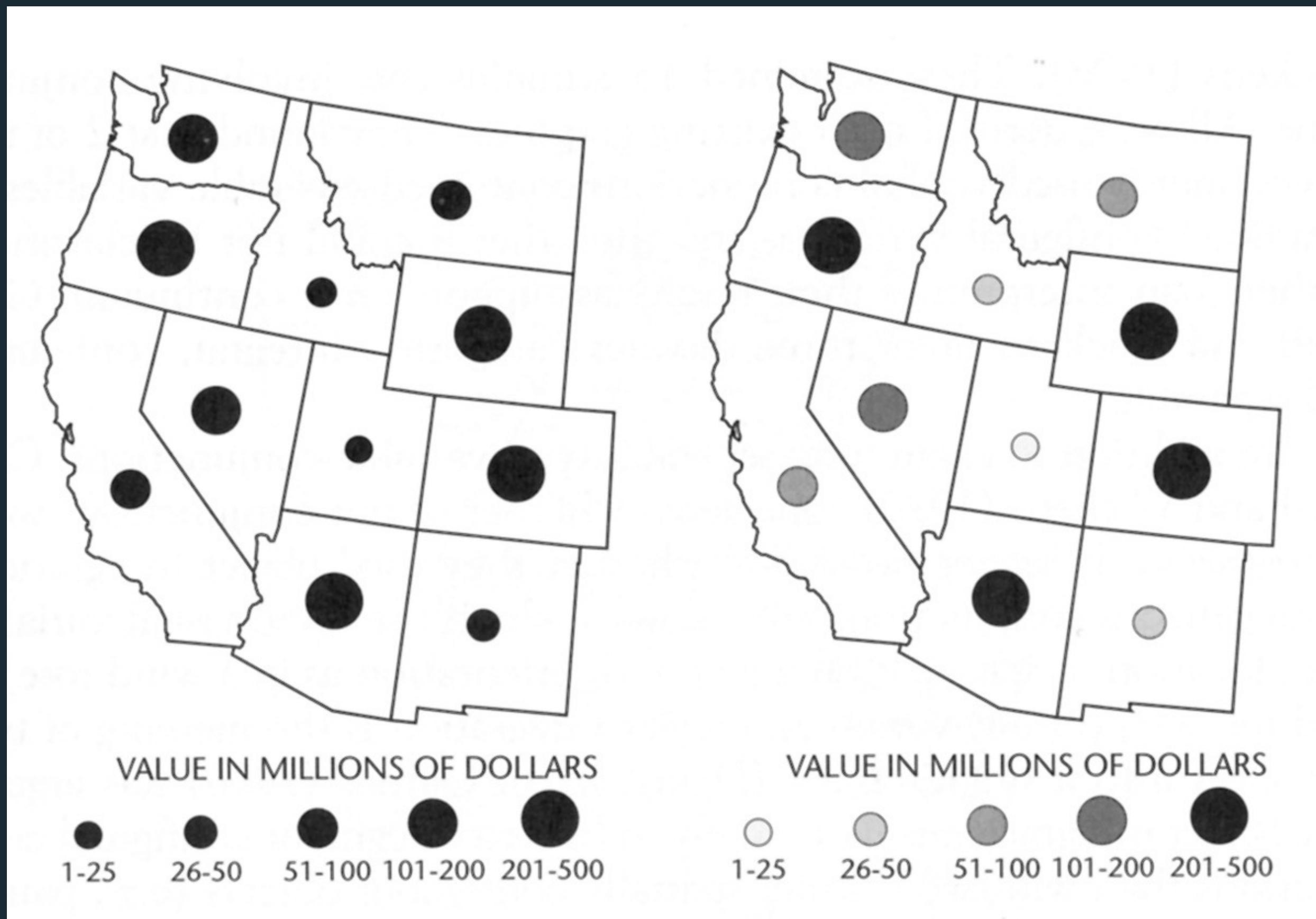
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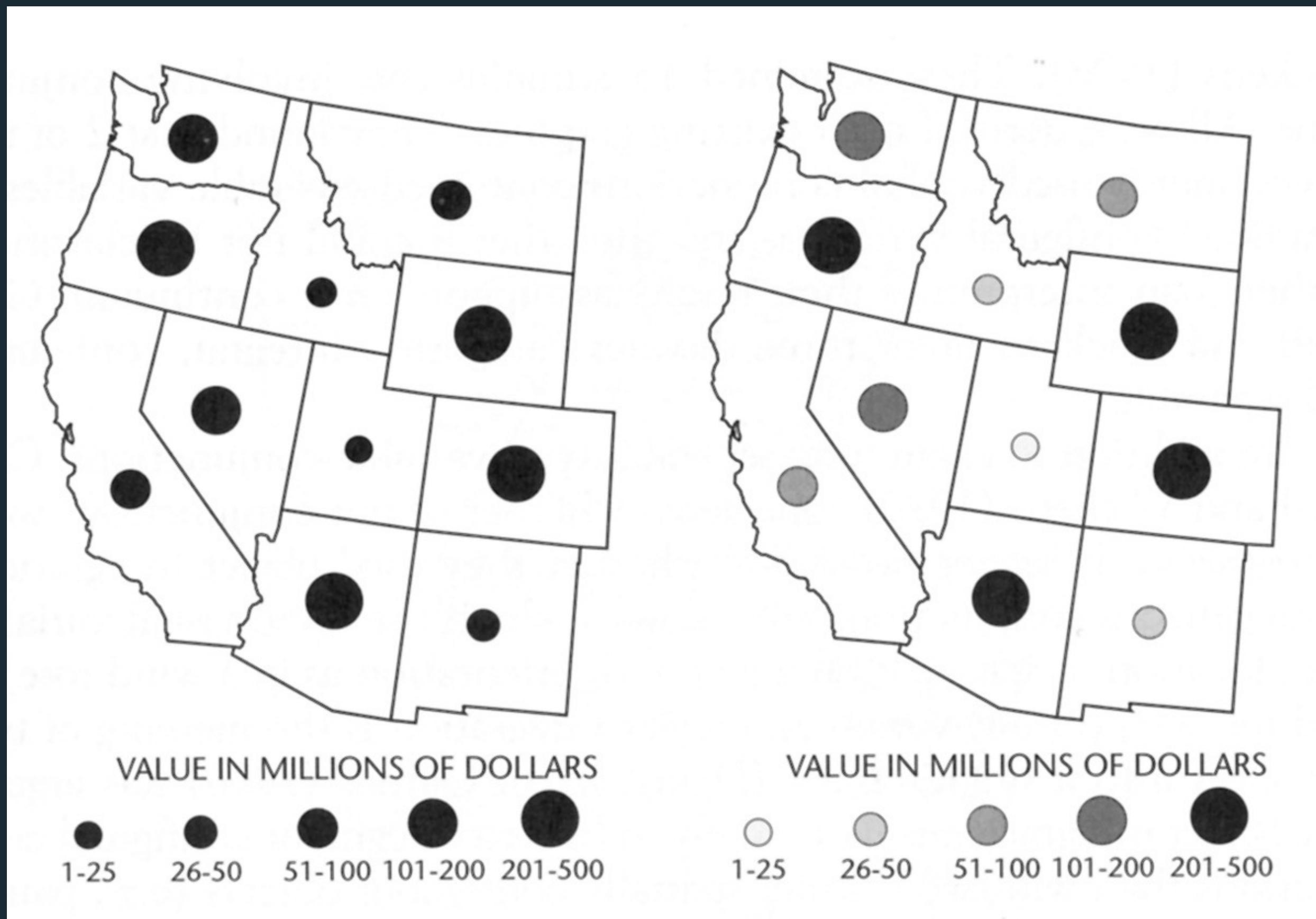
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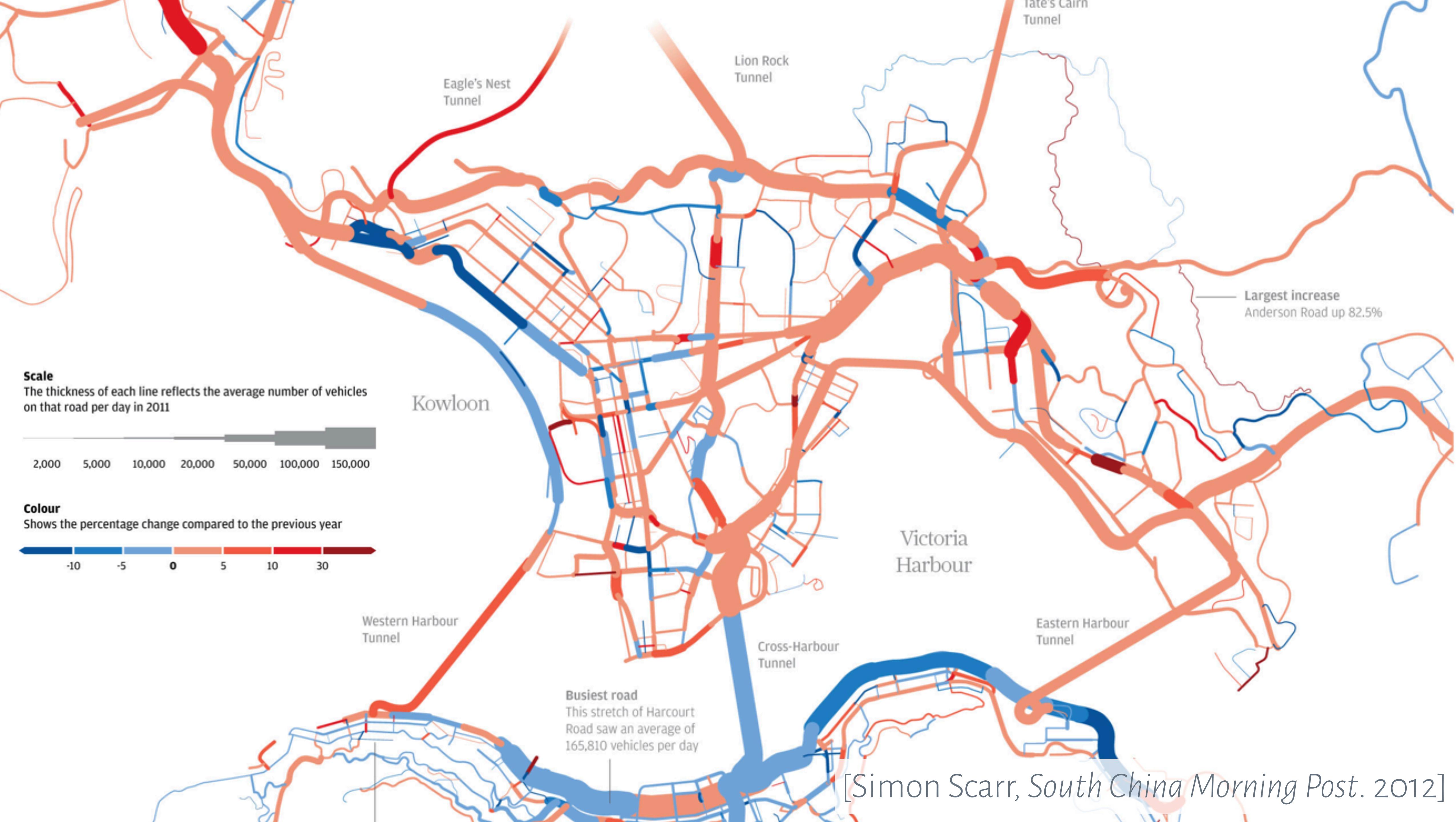
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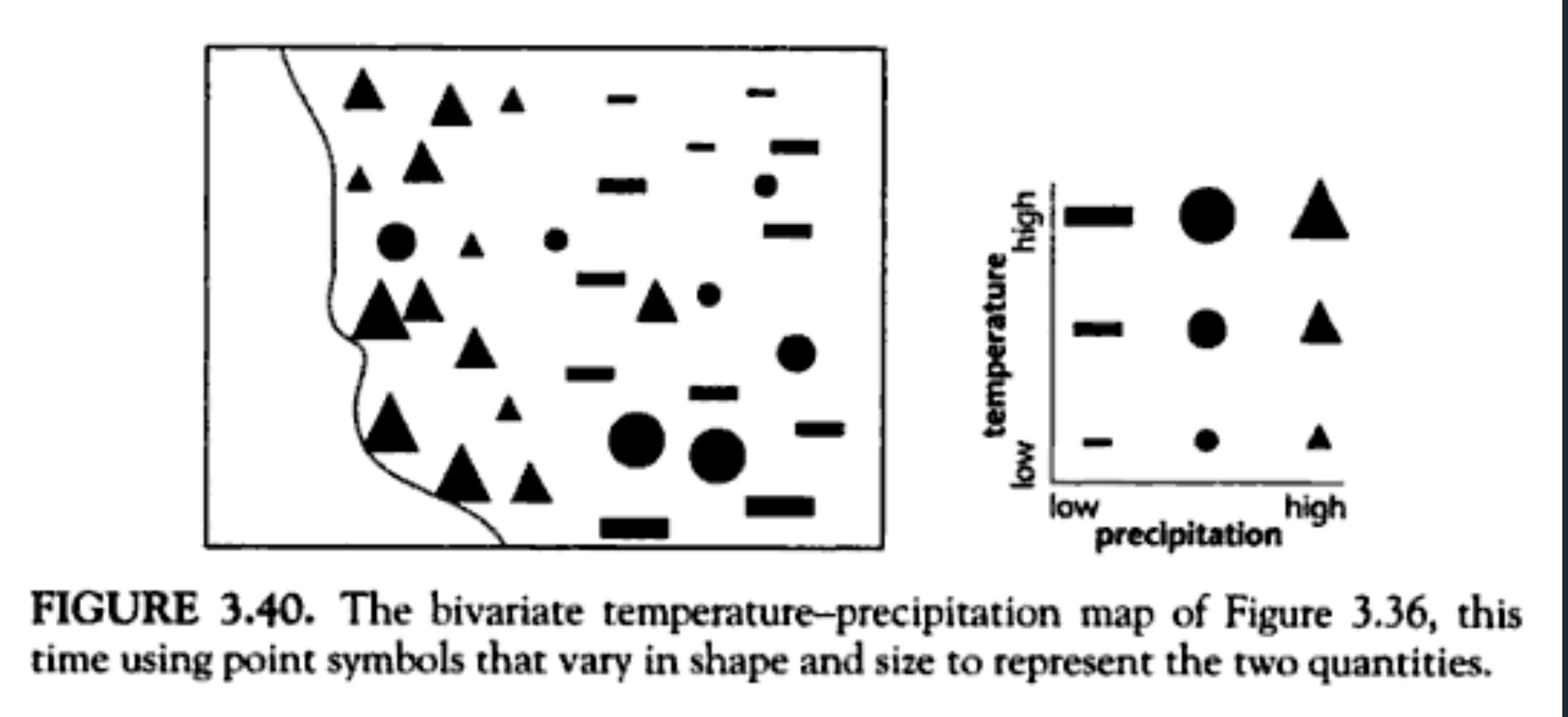


FIGURE 3.40. The bivariate temperature–precipitation map of Figure 3.36, this time using point symbols that vary in shape and size to represent the two quantities.

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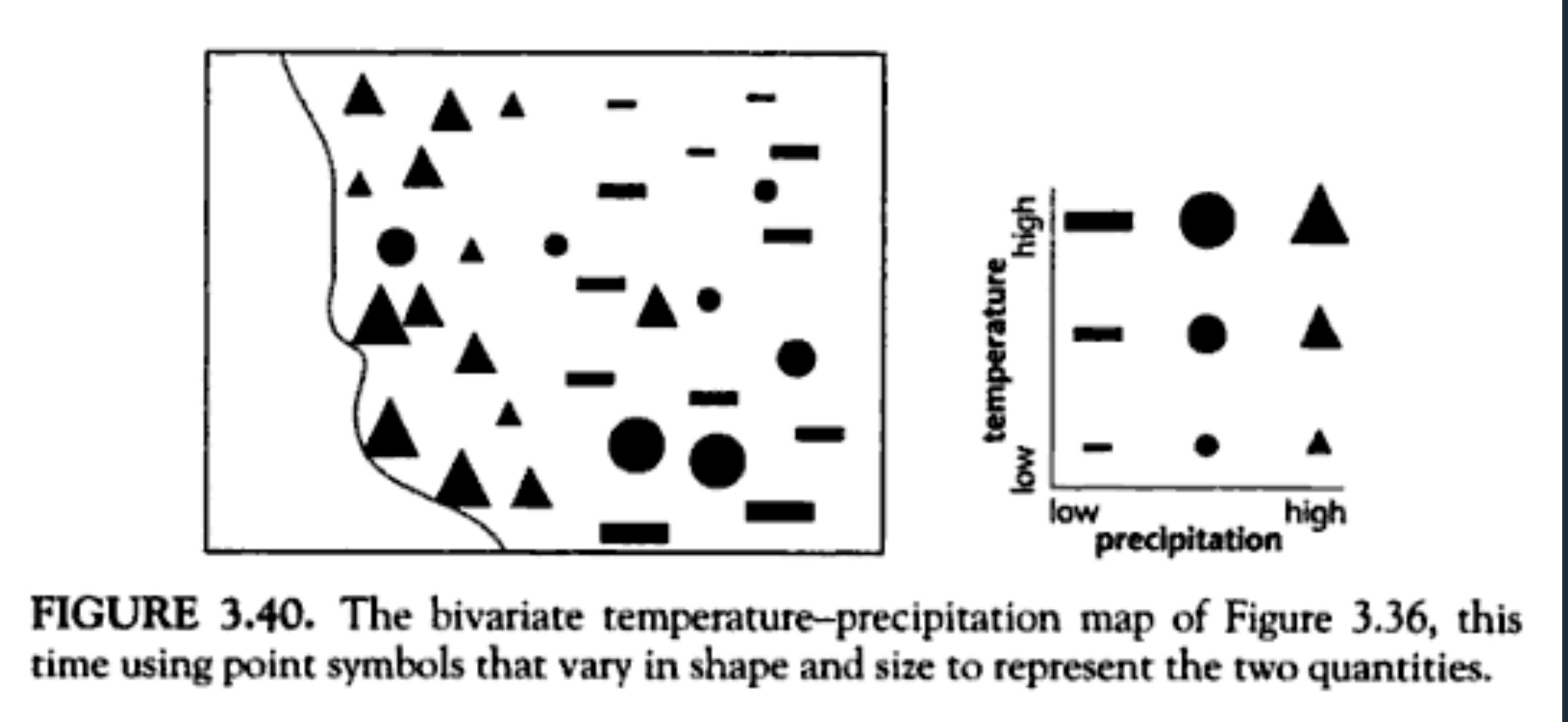


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Width & Height?

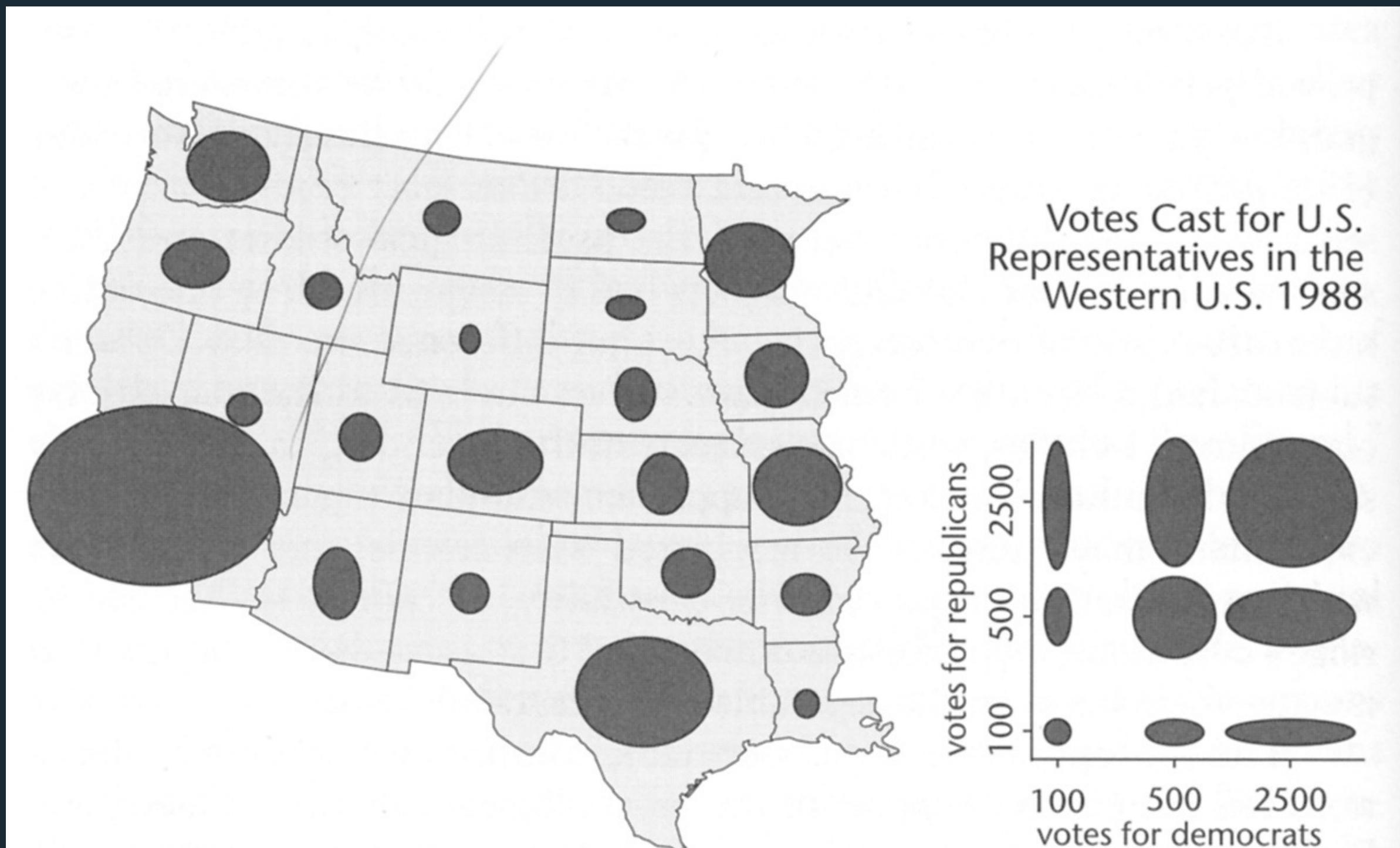


FIGURE 3.38. An example of the use of an ellipse as a map symbol in which the horizontal and vertical axes represent different (but presumably related) variables.

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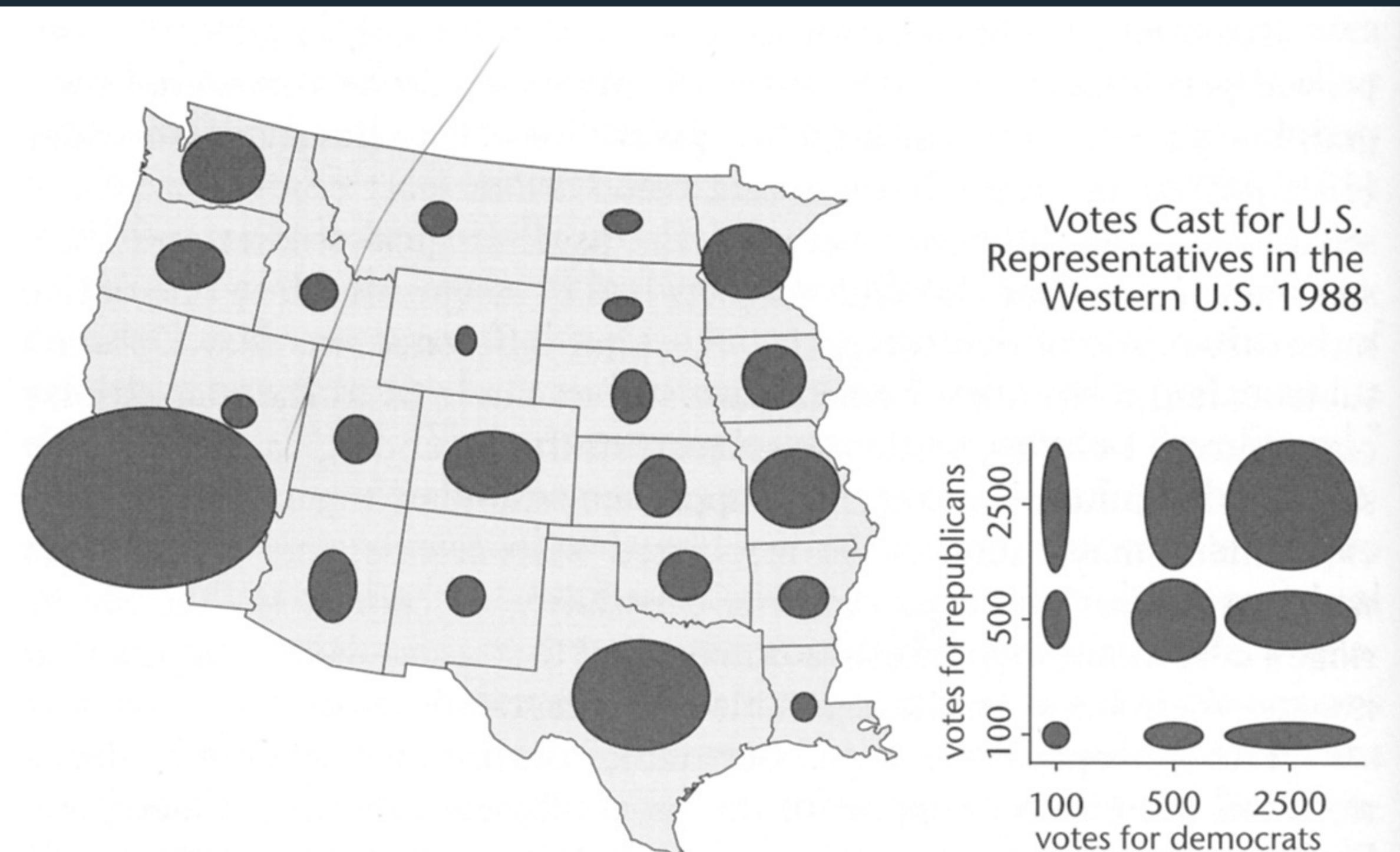


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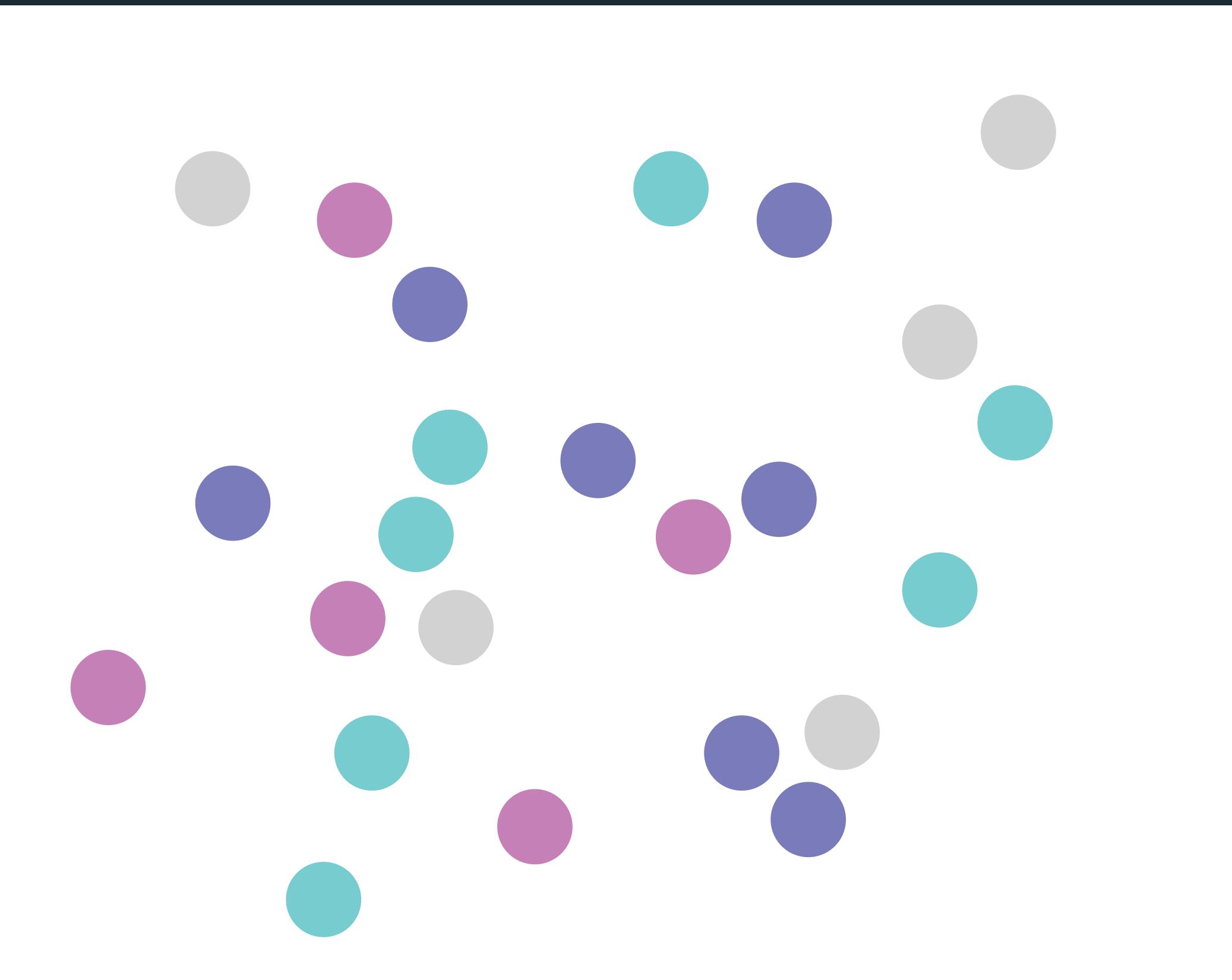
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Red & Green / Yellow & Blue?



[Tamara Munzner, *Visualization Analysis and Design* (2014)]

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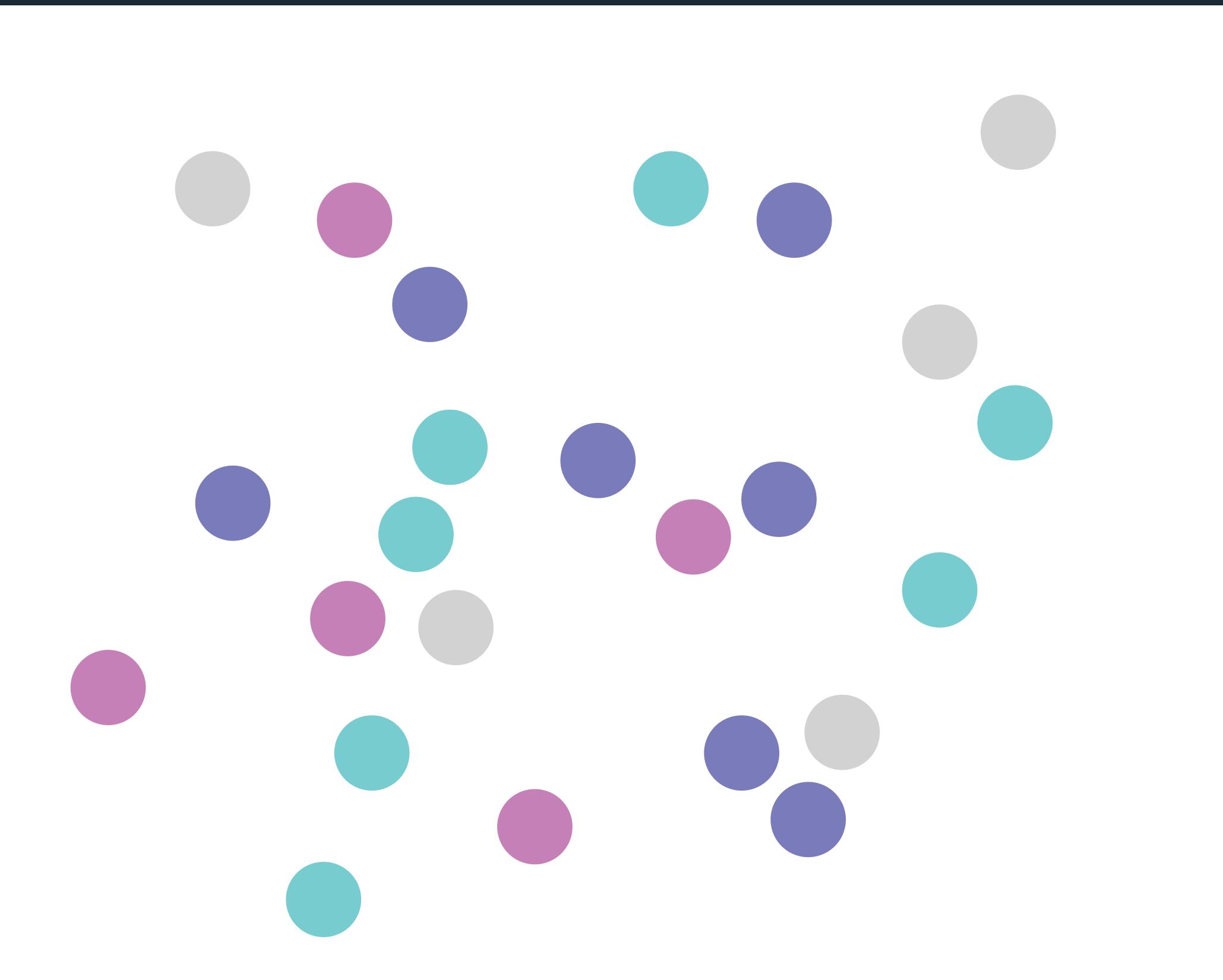
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Red & Green / Yellow & Blue?



[Tamara Munzner, *Visualization Analysis and Design* (2014)]

Types of Dimensions

Separable

No interference or redundancy gain.

blue

Integral

Filtering interference and redundancy gain.

red

Configural

Only interference. No redundancy gain.

green

Asymmetric

One dimension is separable from the other,
but not vice versa.

orange

purple

Types of Dimensions

Separable

No interference or redundancy gain.

blue

Integral

Filtering interference and redundancy gain.

yellow

Configural

Only interference. No redundancy gain.

red

Asymmetric

One dimension is separable from the other, but not vice versa.

green

orange

purple

Types of Dimensions

Separable

No interference or redundancy gain.

blue

Integral

Filtering interference and redundancy gain.

yellow

Configural

Only interference. No redundancy gain.

red

Asymmetric

One dimension is separable from the other, but not vice versa.

green

orange

purple

Signal Detection

Magnitude Estimation

Pre-Attentive Processing

Selective Attention

Gestalt Grouping

Separability: how much interaction occurs between attributes?

Signal Detection

Magnitude Estimation

Pre-Attentive Processing

Selective Attention

Gestalt Grouping

Gestalt Principles

pragnänz: we favor the simplest and most stable interpretations

Figure / Ground

Proximity

Similarity

Symmetry

Connectedness

Continuity

Closure

Common Fate

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Ambiguous – vase or faces?



Unambiguous (?)

Gestalt Principles

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Common Fate



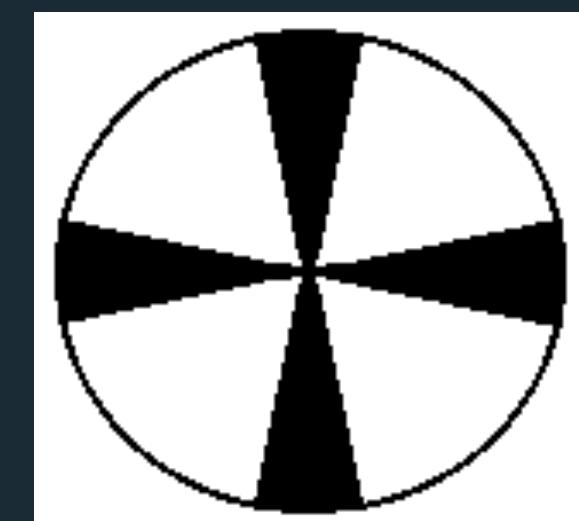
Ambiguous – vase or faces?



Unambiguous (?)



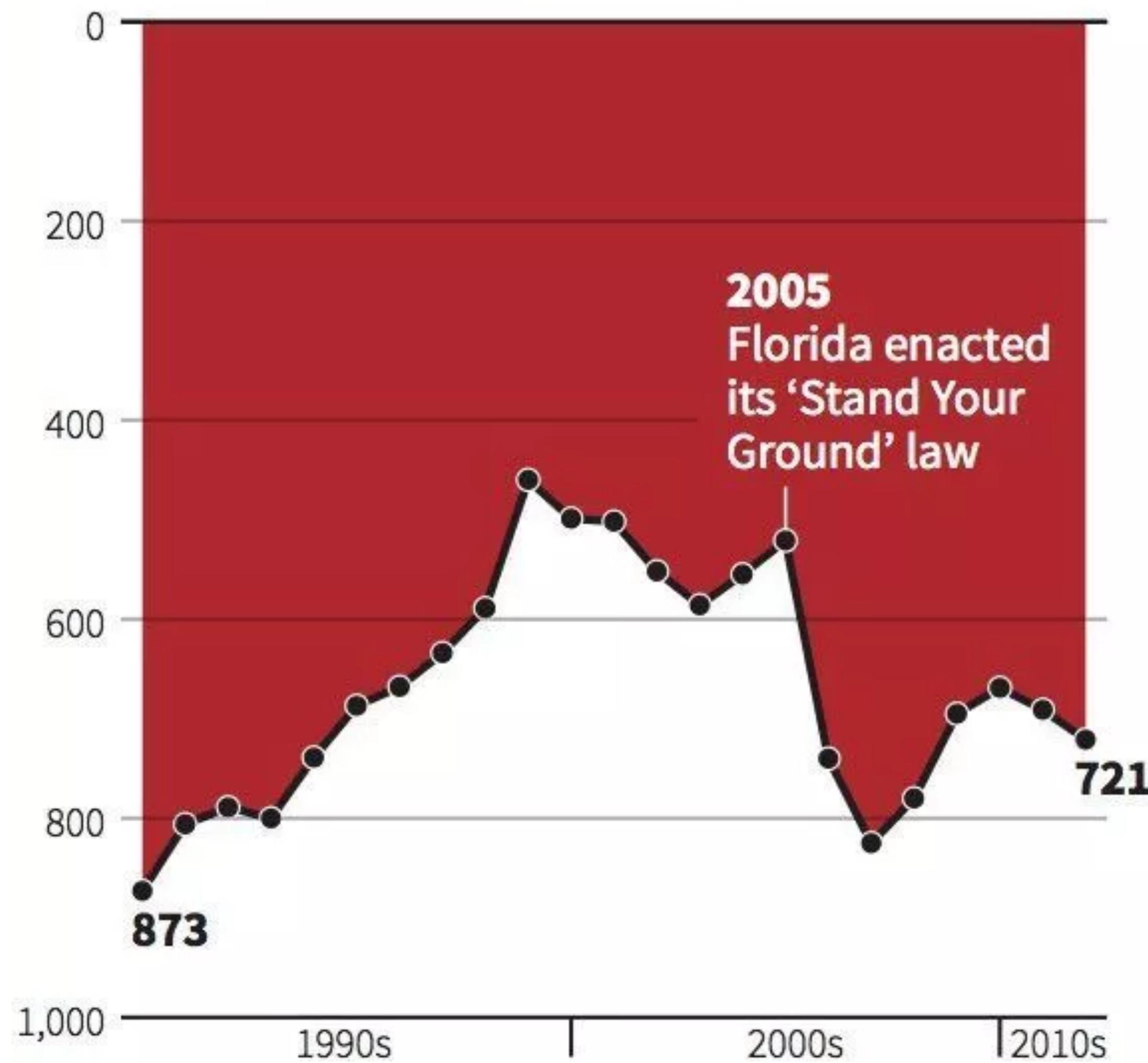
Principle of *surroundedness*.



Principle of *relative size*.

Gun deaths in Florida

Number of murders committed using firearms



Source: Florida Department of Law Enforcement

Gestalt Principles

pragnänz: we favor the simplest and most stable interpretations

Figure / Ground

Proximity

Similarity

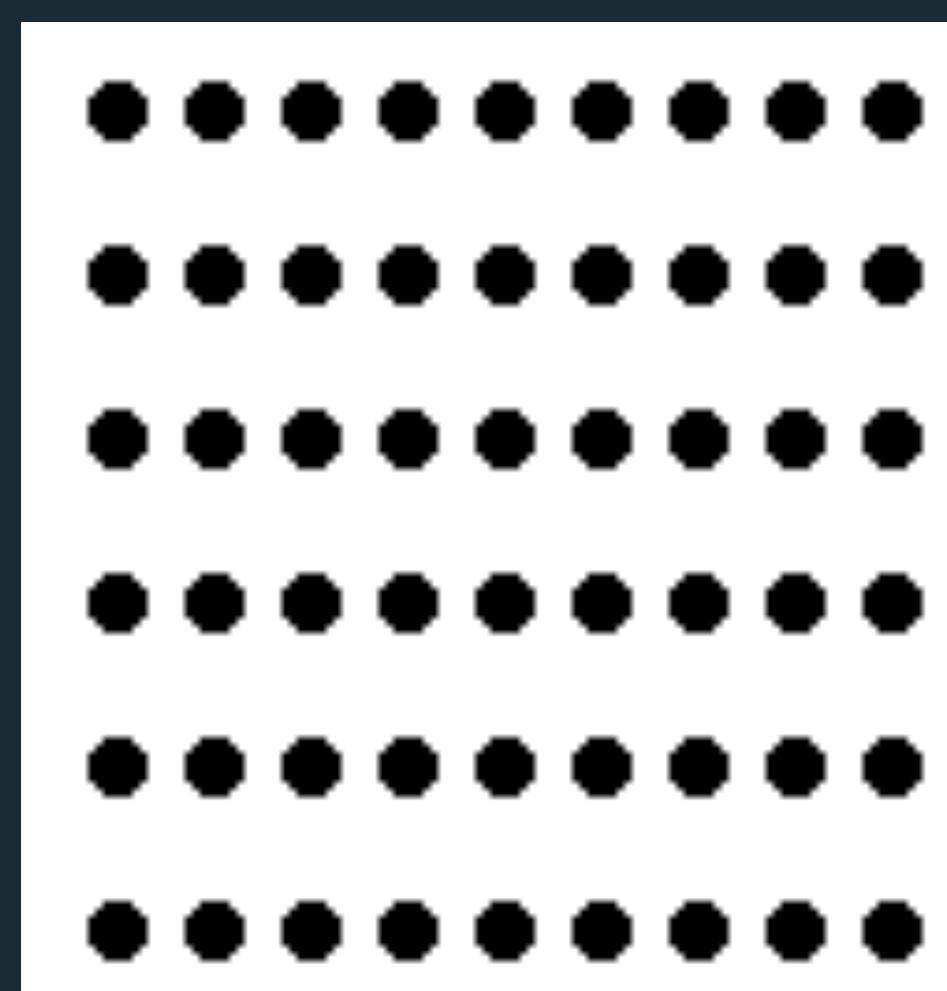
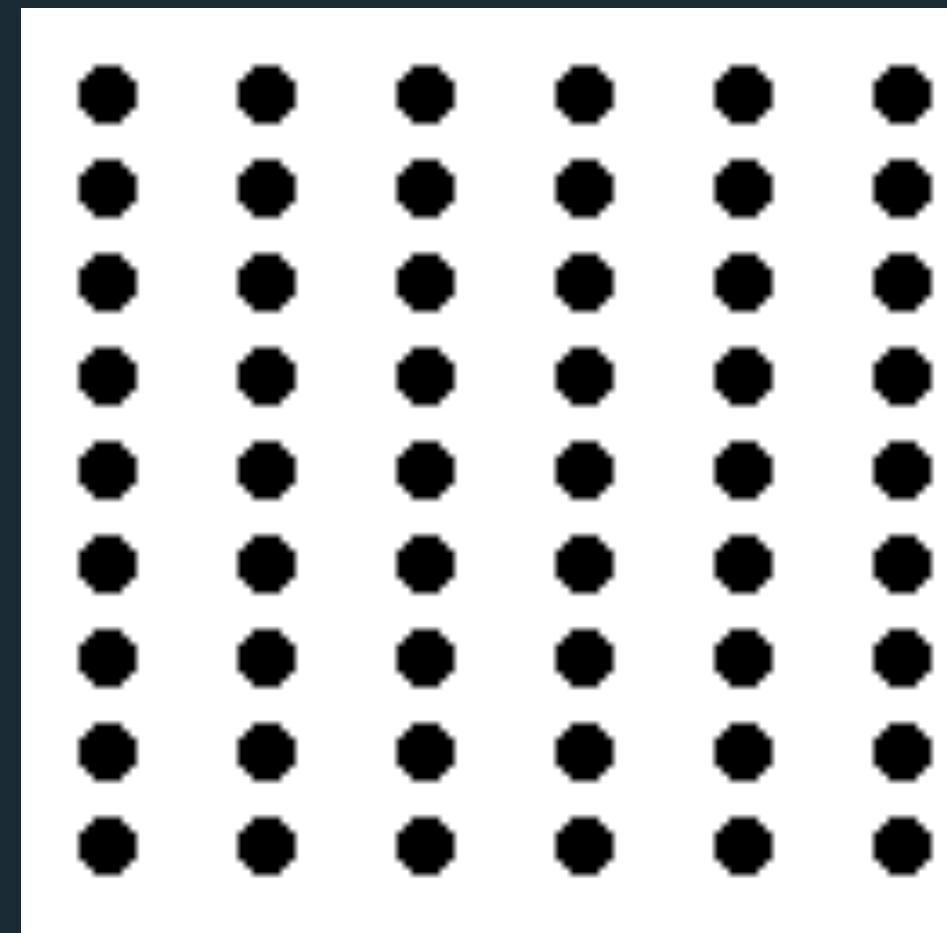
Symmetry

Connectedness

Continuity

Closure

Common Fate



Driving Shifts Into Reverse

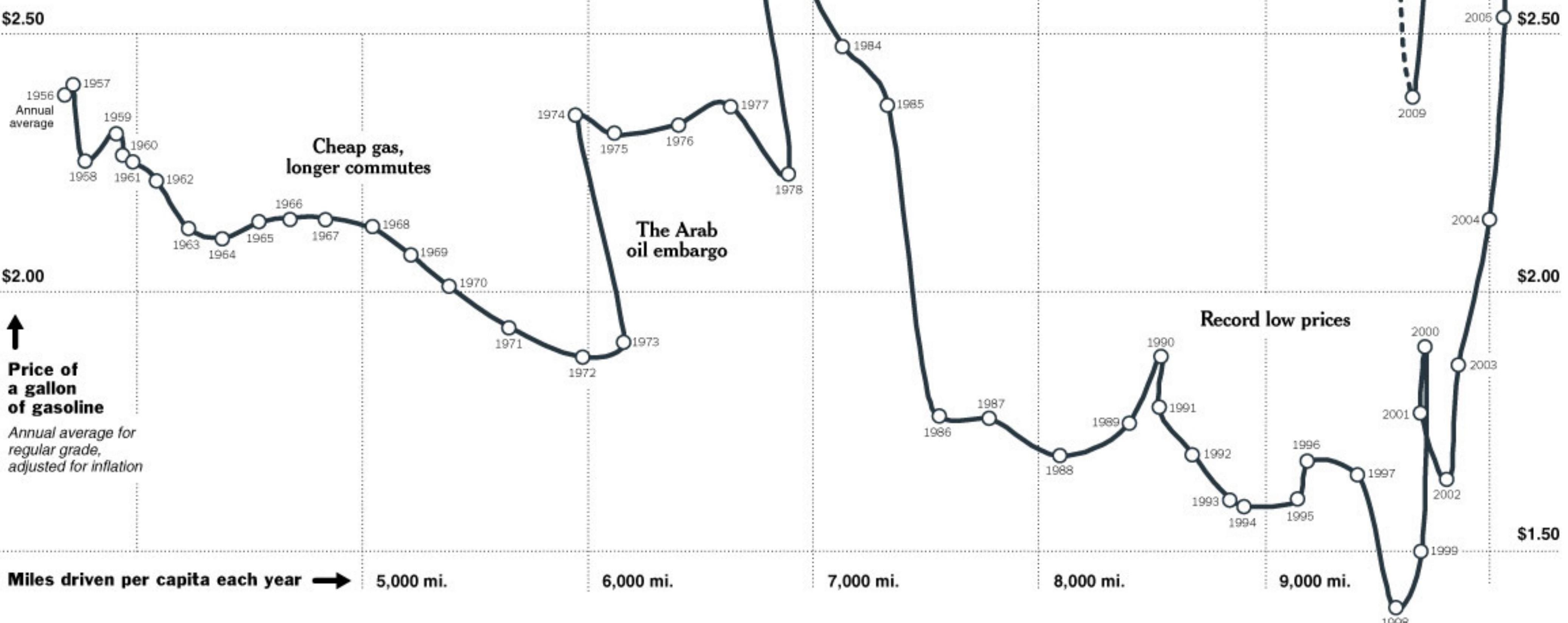
ECONOMISTS have long studied the relationship between driving habits and gasoline prices. Low gas prices can bring periods of profligate driving, and a quick jump in prices can cause many vehicles to languish in garages.

Until recently, Americans have driven more each year than the previous one, with a few brief exceptions. In 1956, Americans of driving age drove about 4,000 miles a year, on average. Fifty years later, that figure had climbed above 10,000.

But the latest recession has caused some big changes. High unemployment meant that fewer people were driving to work, and a slump in consumer spending

meant that less freight needed to be moved around the country. As gas prices soared in 2005, the number of miles driven — including commercial and personal — began to fall, and continued to drop after 2008 even as gasoline became cheaper.

"People were surprised by the very rapid rise in gas prices, and they changed their driving behavior," said Kenneth A. Small, a transportation economist at the University of California, Irvine. "But my suspicion is that it is temporary. As soon as unemployment gets back to pre-recession levels, we will see Americans doing a lot more driving again."



Gestalt Principles

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Proximity

Similarity

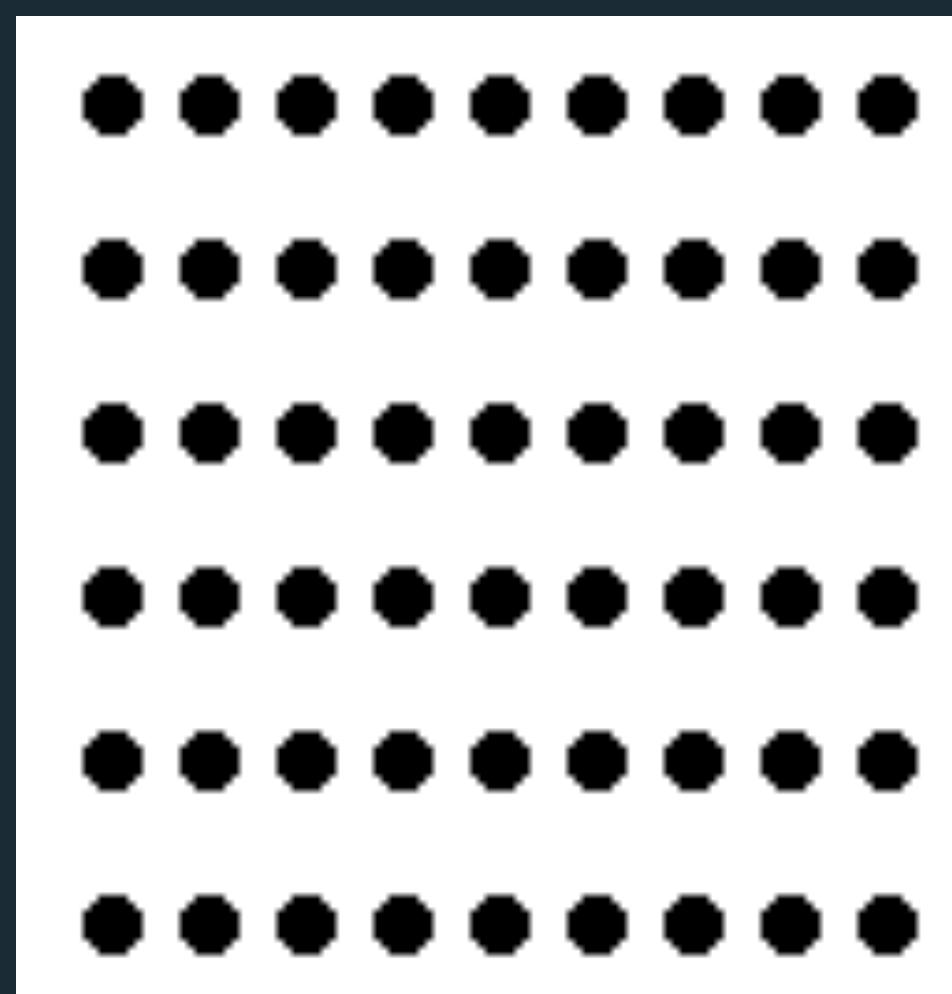
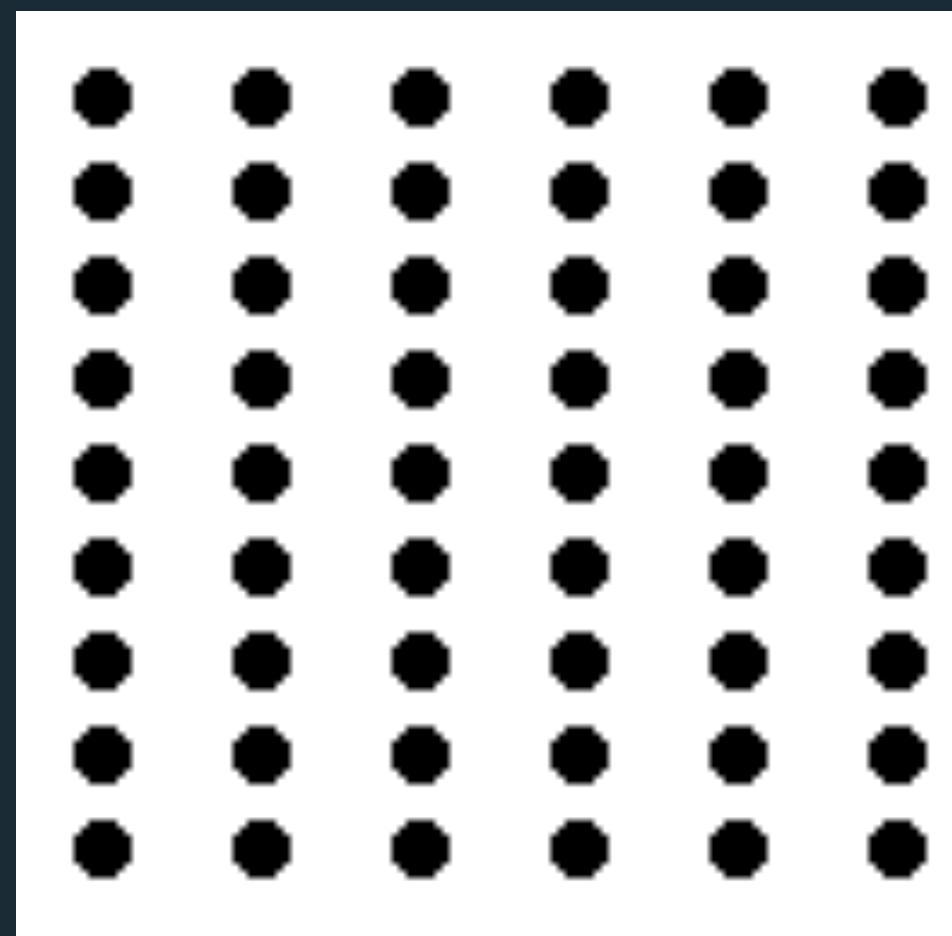
Symmetry

Connectedness

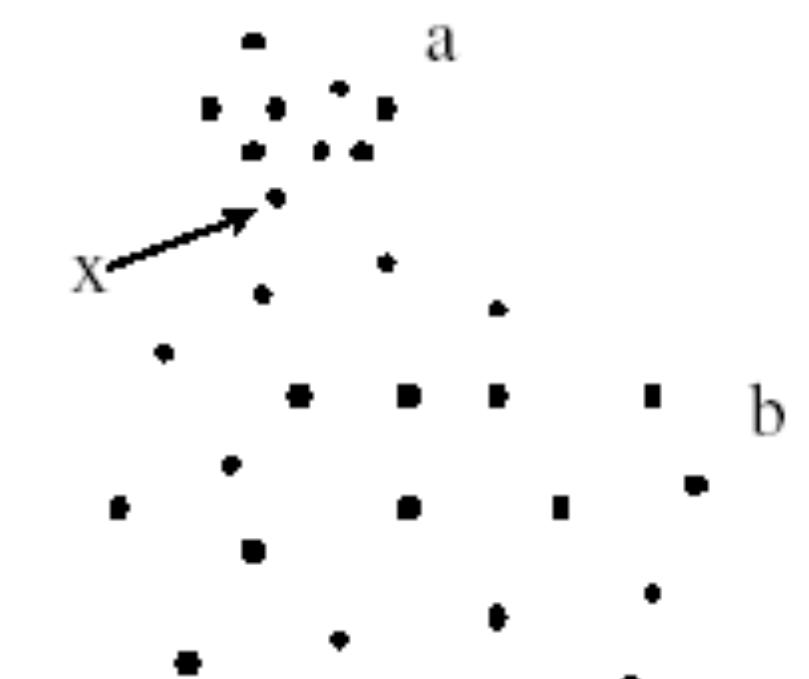
Continuity

Closure

Common Fate



[Ware 2000]



Principle of *concentration*.

Gestalt Principles

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Figure / Ground

Proximity

Similarity

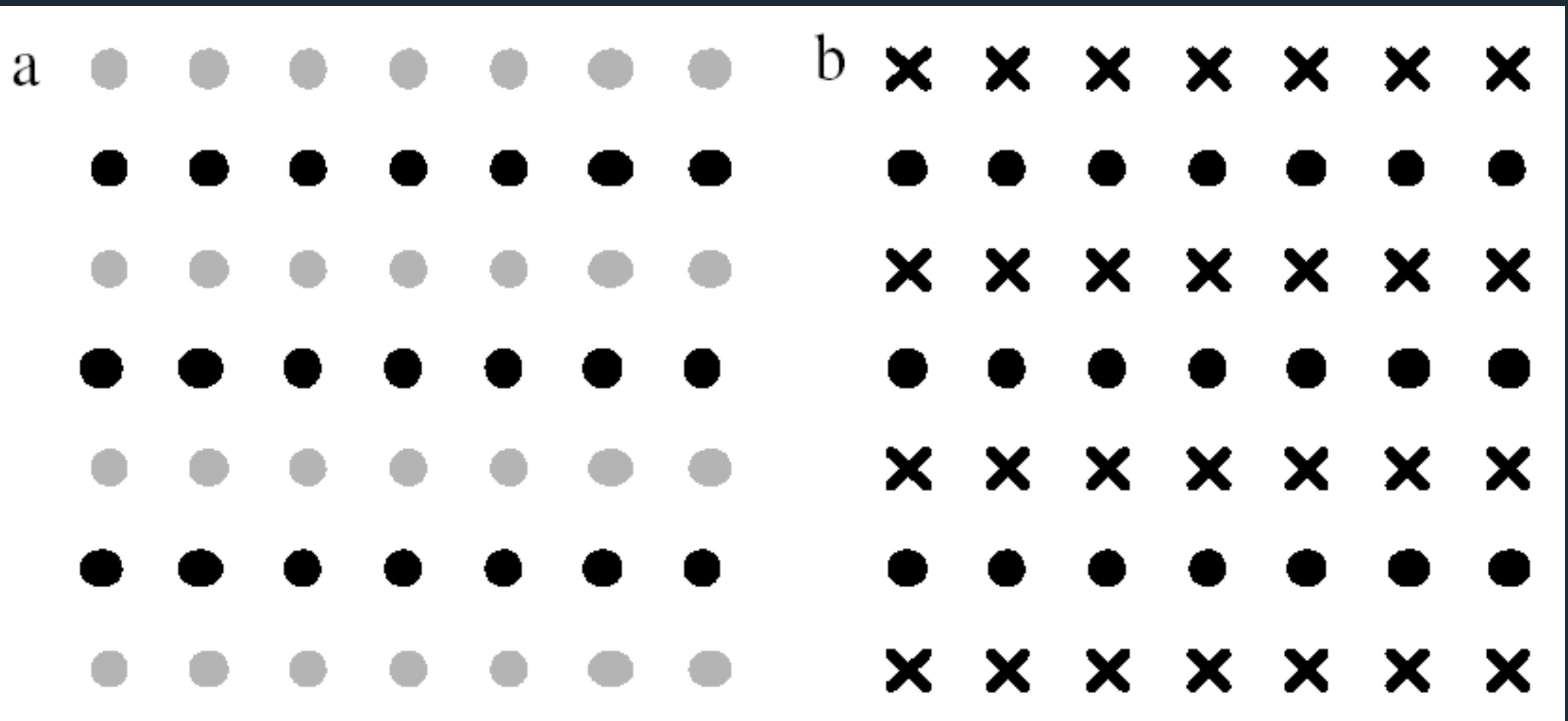
Symmetry

Connectedness

Continuity

Closure

Common Fate



Rows dominate due to similarity.

[Ware 2004]

Gestalt Principles

pragnänz: we favor the simplest and most stable interpretations

Figure / Ground

Proximity

Similarity

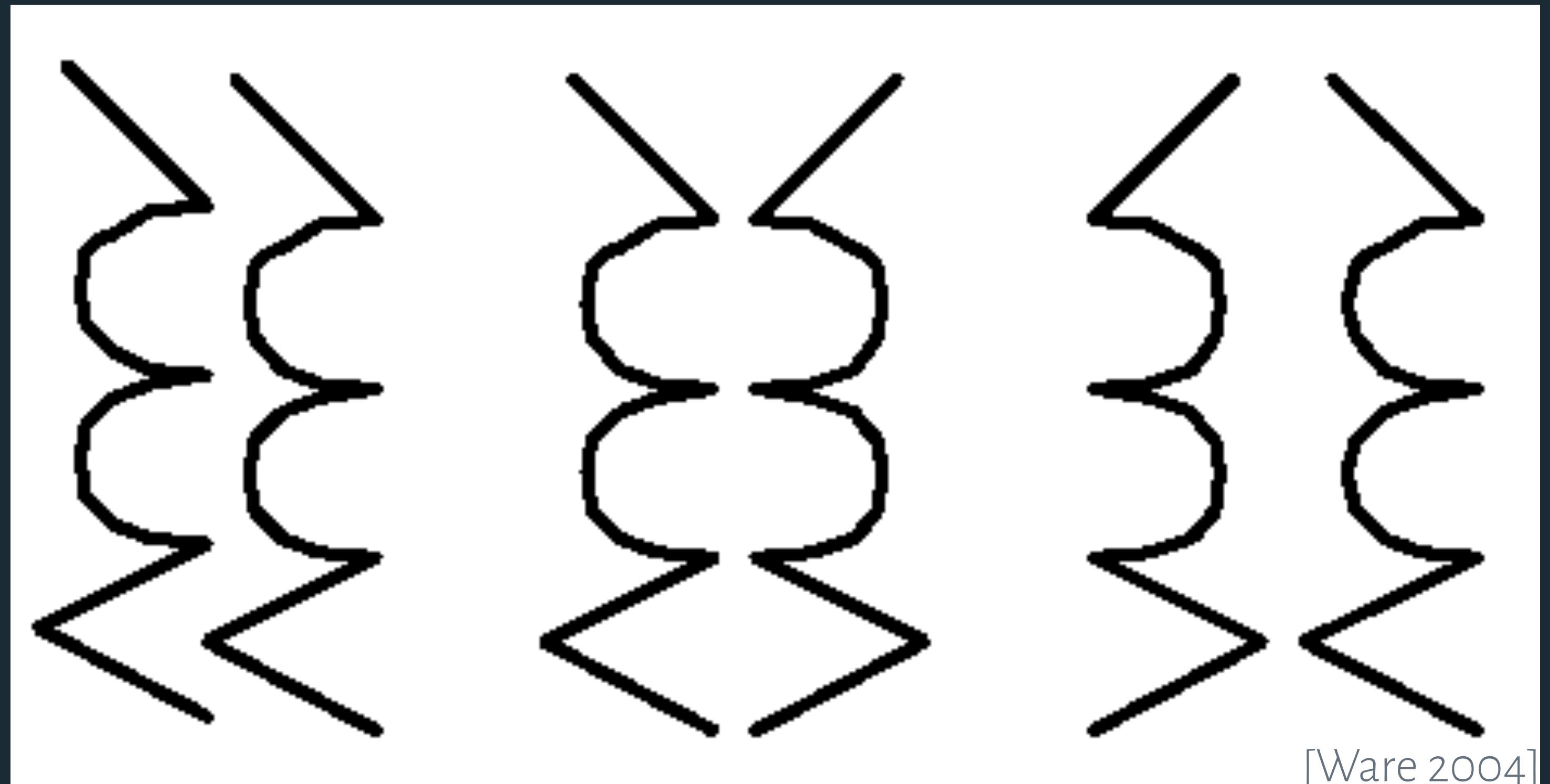
Symmetry

Connectedness

Continuity

Closure

Common Fate



[Ware 2004]

Bilateral symmetry gives the strong sense of a figure.

Gestalt Principles

pragnänz: we favor the simplest and most stable interpretations

Figure / Ground

Proximity

Similarity

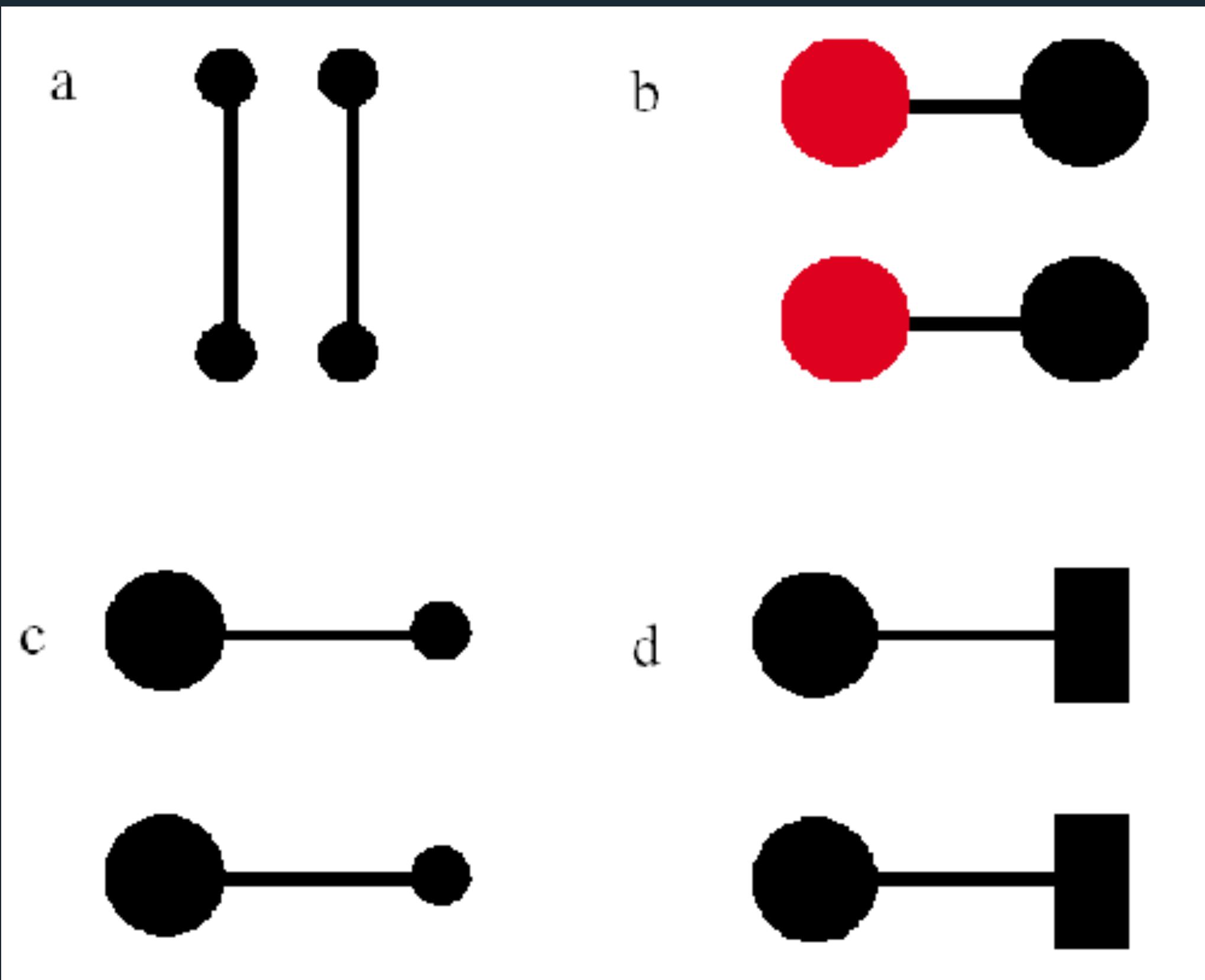
Symmetry

Connectedness

Continuity

Closure

Common Fate



[Ware 2004]

Connectedness overrules proximity, size, color, shape, etc.

Gestalt Principles

Figure / Ground

Proximity

Similarity

Symmetry

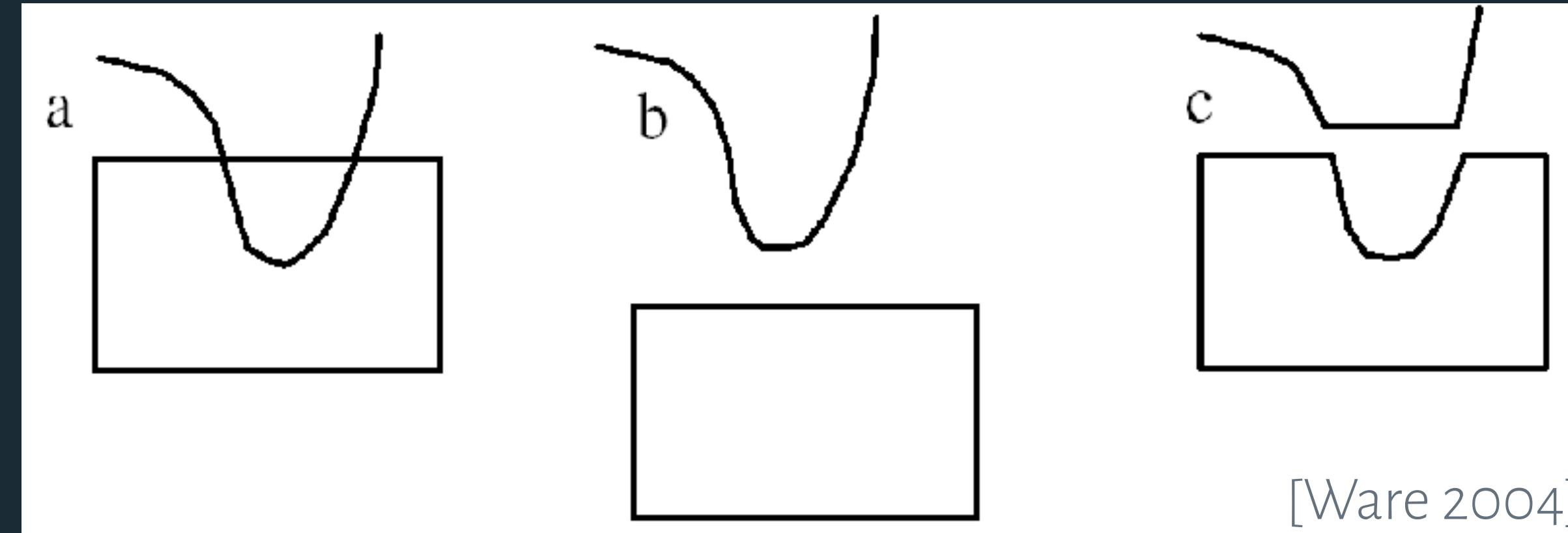
Connectedness

Continuity

Closure

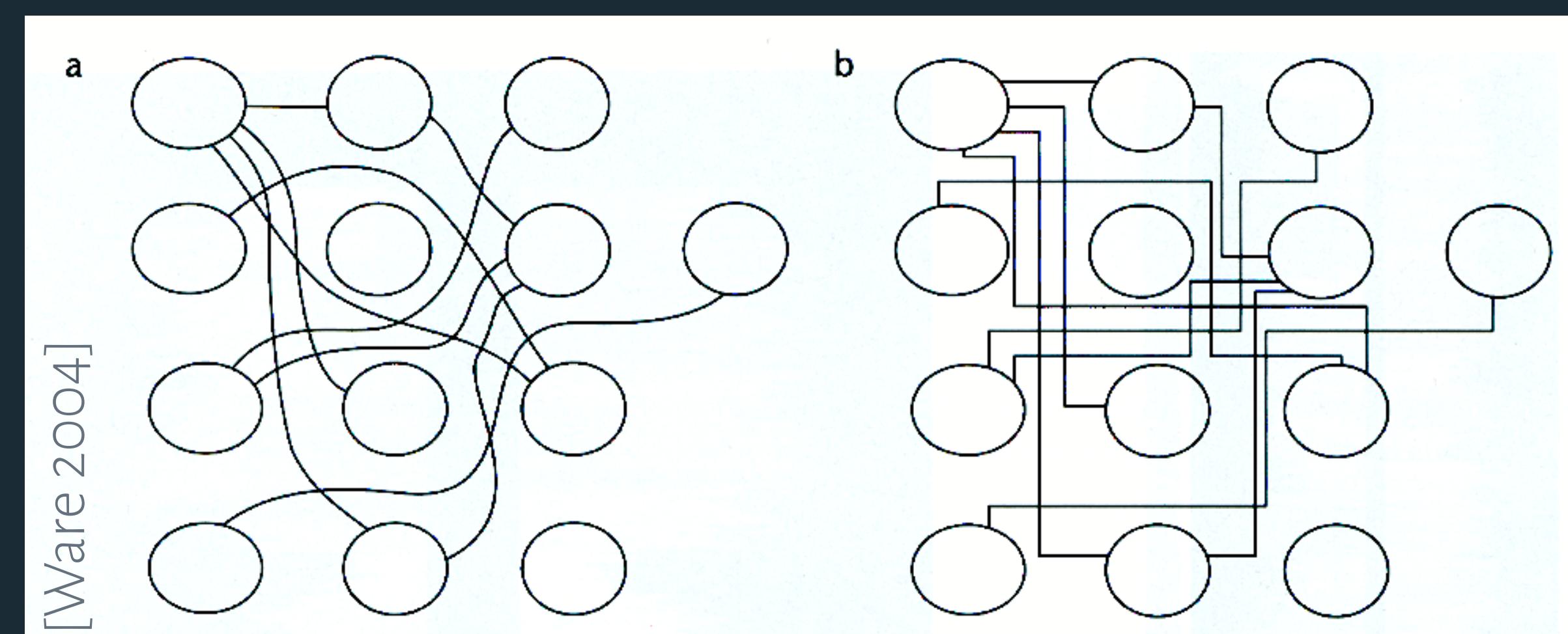
Common Fate

pragnänz: we favor the simplest and most stable interpretations



[Ware 2004]

We prefer smooth, not abrupt, changes.



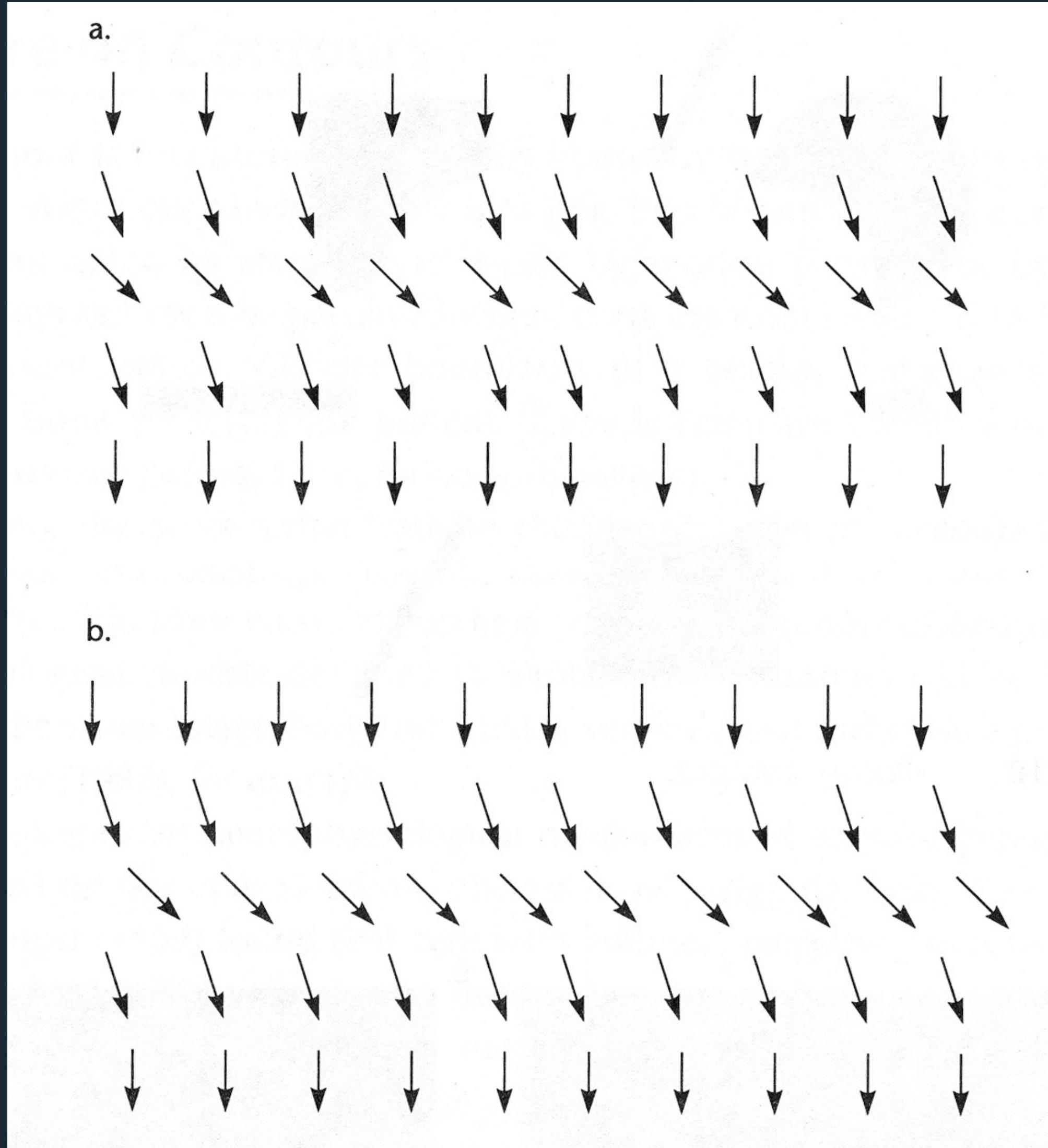
[Ware 2004]

Connections are clearer with smooth contours.

Gestalt Principles

Figure / Ground
Proximity
Similarity
Symmetry
Connectedness
Continuity
Closure
Common Fate

pragnänz: we favor the simplest and most stable interpretations



Prefer field that shows smooth continuous contours

[Ware 2004]

Gestalt Principles

Figure / Ground

Proximity

Similarity

Symmetry

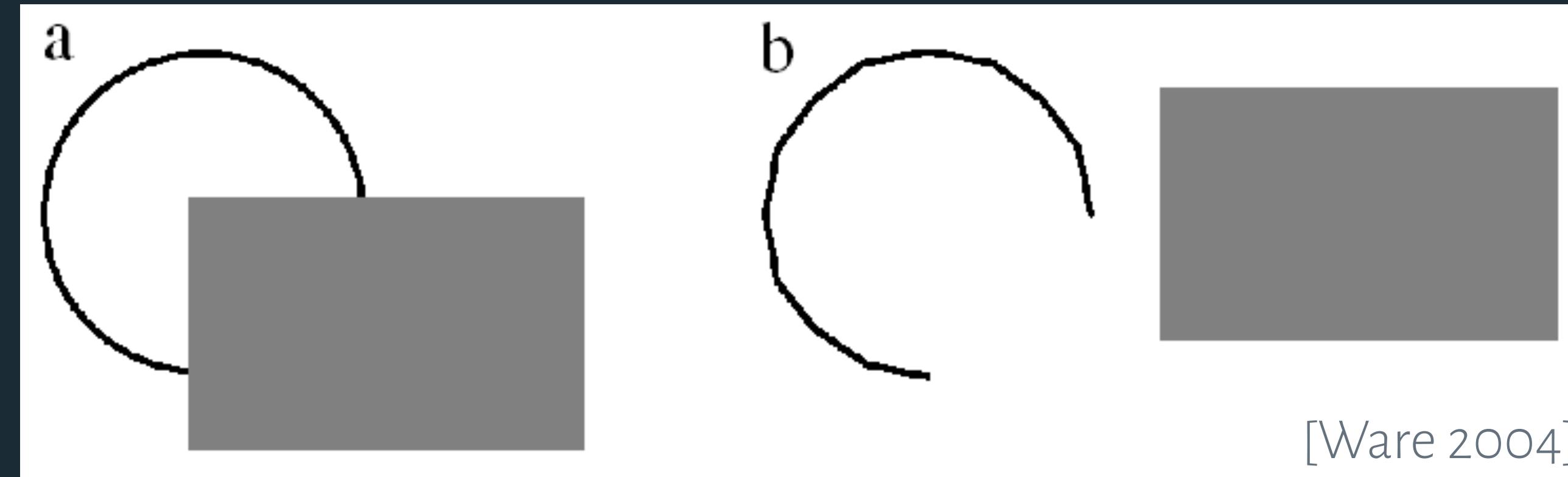
Connectedness

Continuity

Closure

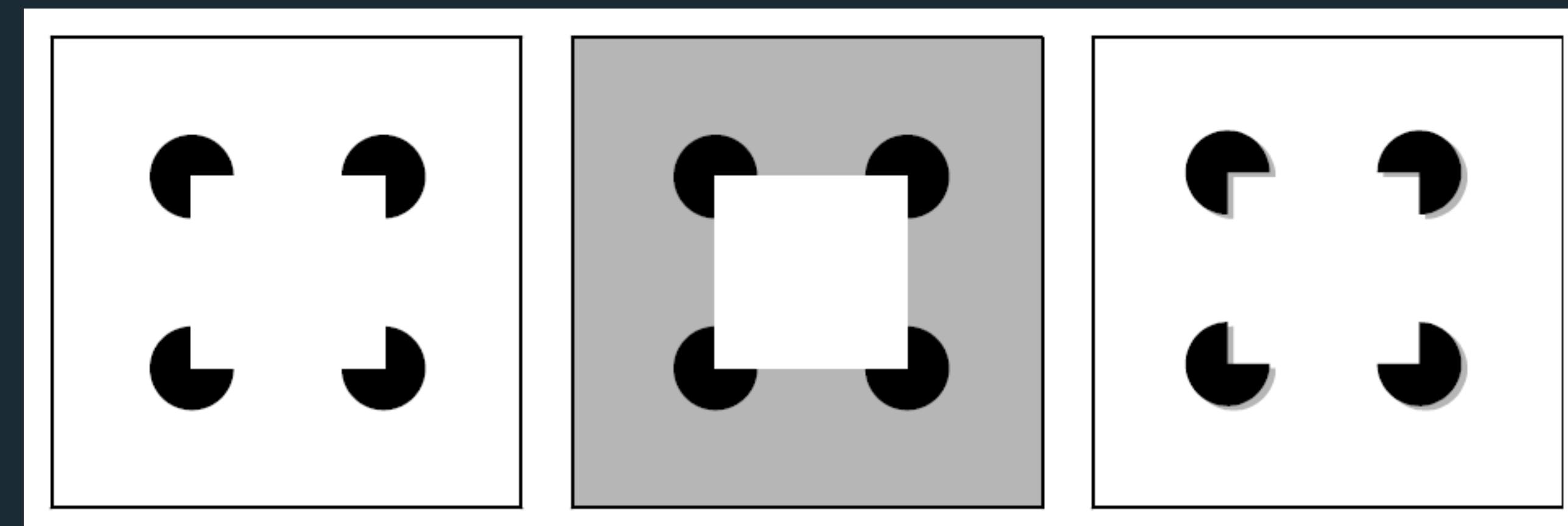
Common Fate

pragnänz: we favor the simplest and most stable interpretations



[Ware 2004]

We see a circle behind a rectangle, not a broken circle.



[Durand 02]

Illusory contours

Gestalt Principles

pragnänz: we favor the simplest and most stable interpretations

Figure / Ground

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Common Fate



Dots moving together are grouped.

Signal Detection

Use 4-5 steps for most channels,
hard for people to distinguish more

Magnitude Estimation

Even a direct map to e.g. area or
brightness won't always work.

Pre-Attentive Processing

Use channels that are pre-attentive
for callouts e.g. color, shape.

Selective Attention

...but be careful with combinations
of channels!

Gestalt Grouping

Use these to improve annotations,
coloring, animations.