

Designing Visualizations

INFO 201

<https://slides.com/joelross/info201w17-visualization/live>

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Final Project Description

Today's Objectives

By the end of class, you should be able to

- Identify the purpose and usage of data visualizations
- Assess the effectiveness of different **visual encodings**
- Organize data in the proper **shape**
- Plot geographic data on **maps**

London, 1854

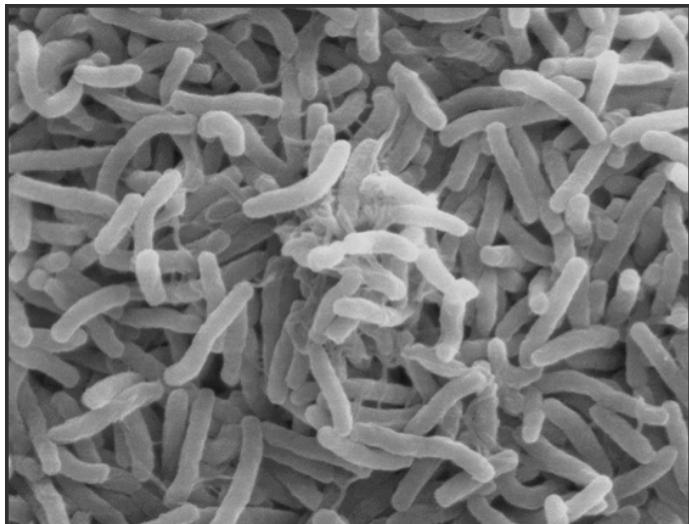
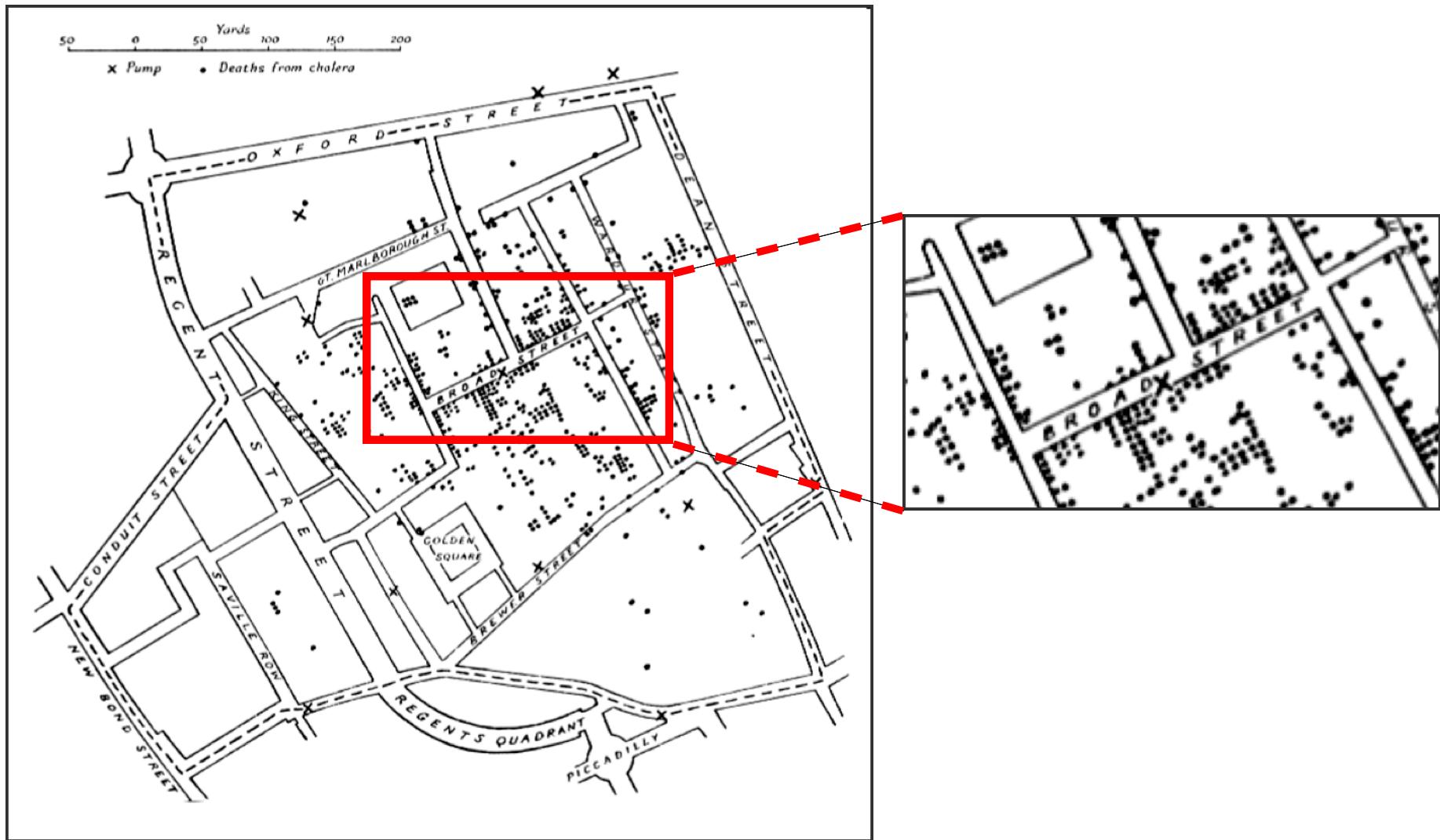


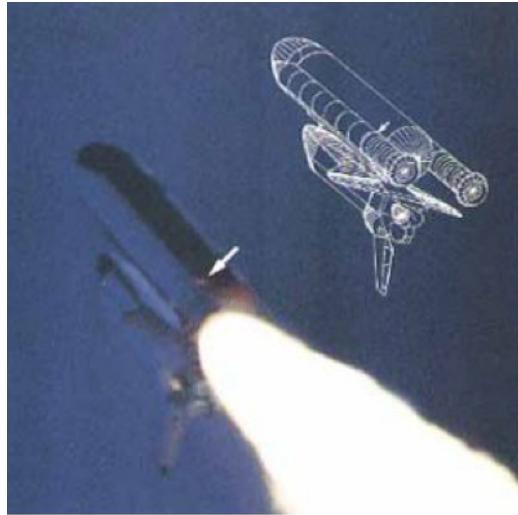
TABLE I

| | Date. | No. of Fatal Attacks. | Deaths |
|-------|-----------|-----------------------|--------|
| | August 19 | 1 | 1 |
| " | 20 | 1 | 0 |
| " | 21 | 1 | 2 |
| " | 22 | 0 | 0 |
| " | 23 | 1 | 0 |
| " | 24 | 1 | 2 |
| " | 25 | 0 | 0 |
| " | 26 | 1 | 0 |
| " | 27 | 1 | 1 |
| " | 28 | 1 | 0 |
| " | 29 | 1 | 1 |
| " | 30 | 8 | 2 |
| " | 31 | 56 | 3 |
| Sept. | 1 | 143 | 70 |
| " | 2 | 116 | 127 |
| " | 3 | 54 | 76 |
| " | 4 | 46 | 71 |
| " | 5 | 36 | 45 |
| " | 6 | 20 | 37 |
| " | 7 | 28 | 32 |

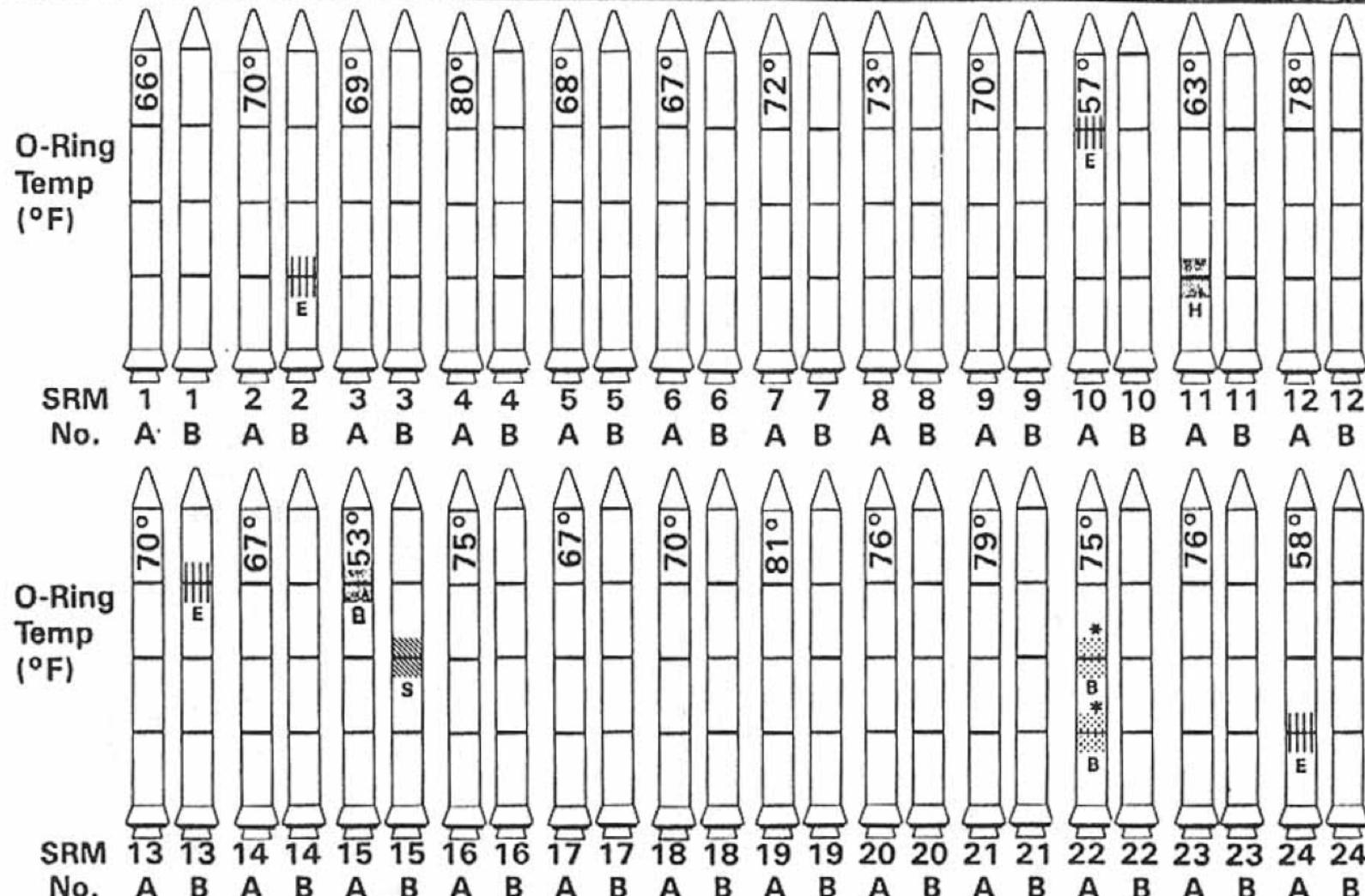
John Snow



Challenger Explosion (1986)



History of O-Ring Damage in Field Joints (Cont)



* No Erosion

86446-1E

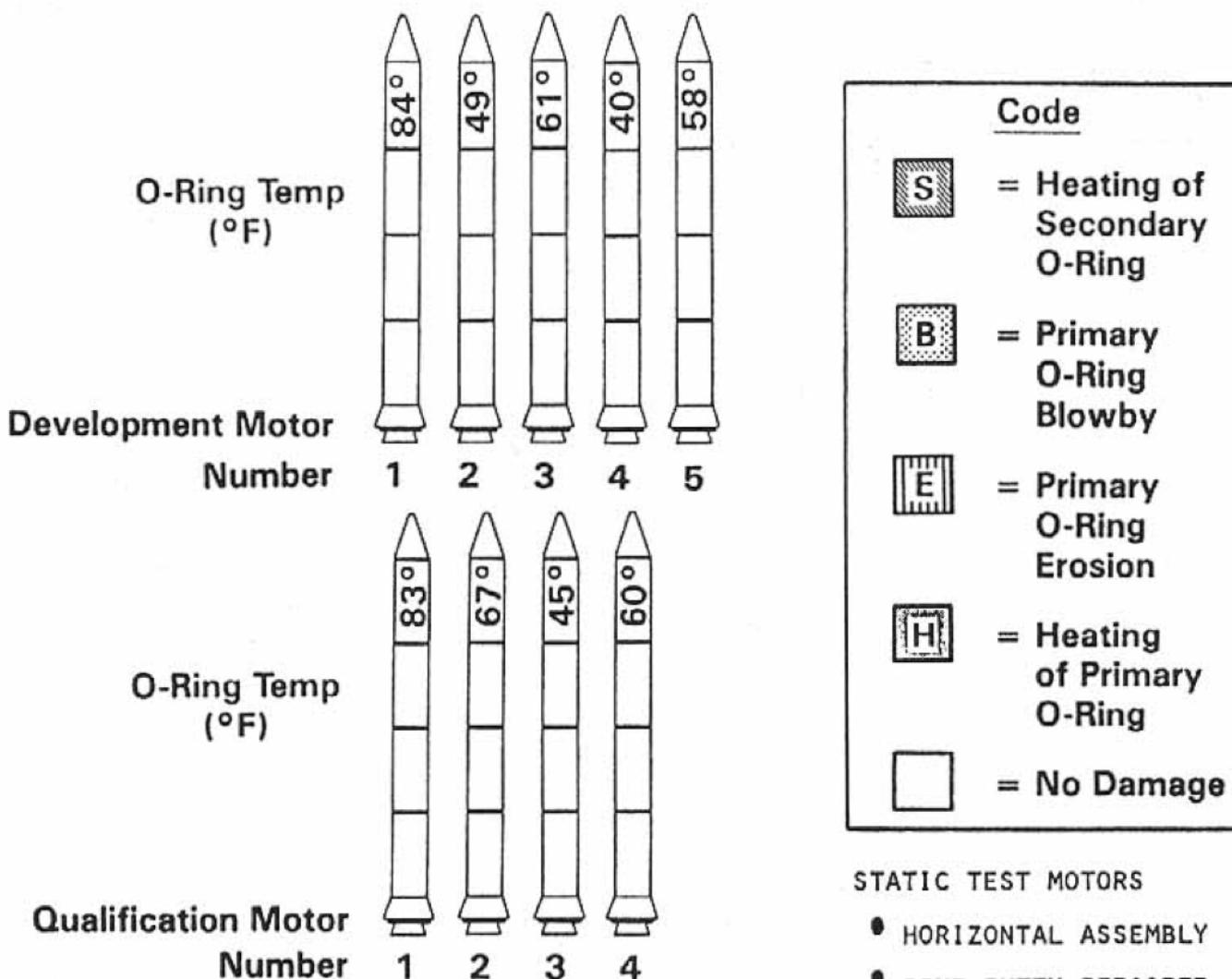
MORTON THIOKOL, INC.

Wasatch Operations

INFORMATION ON THIS PAGE WAS PREPARED TO SUPPORT AN ORAL PRESENTATION
AND CANNOT BE CONSIDERED COMPLETE WITHOUT THE ORAL DISCUSSION

[Ref. 2/26-2 2 of 3]

History of O-Ring Damage in Field Joints



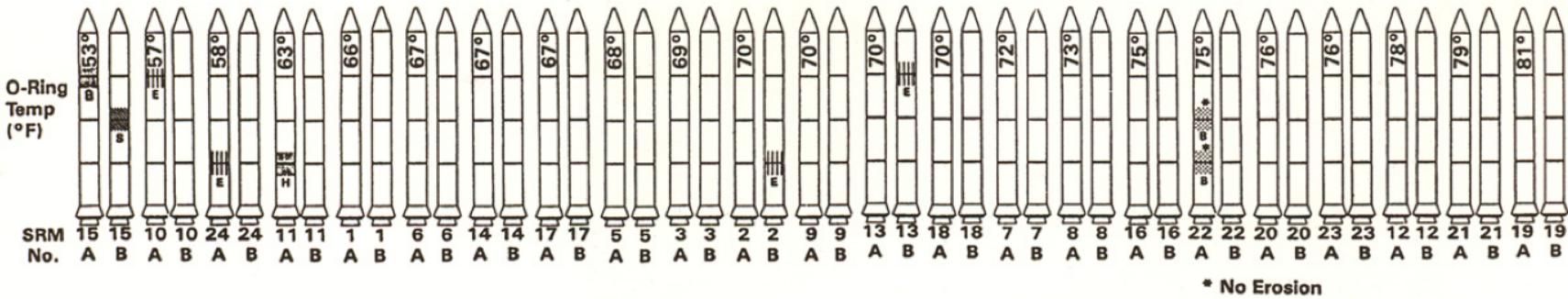
MORTON THIOKOL, INC.

Wasatch Operations

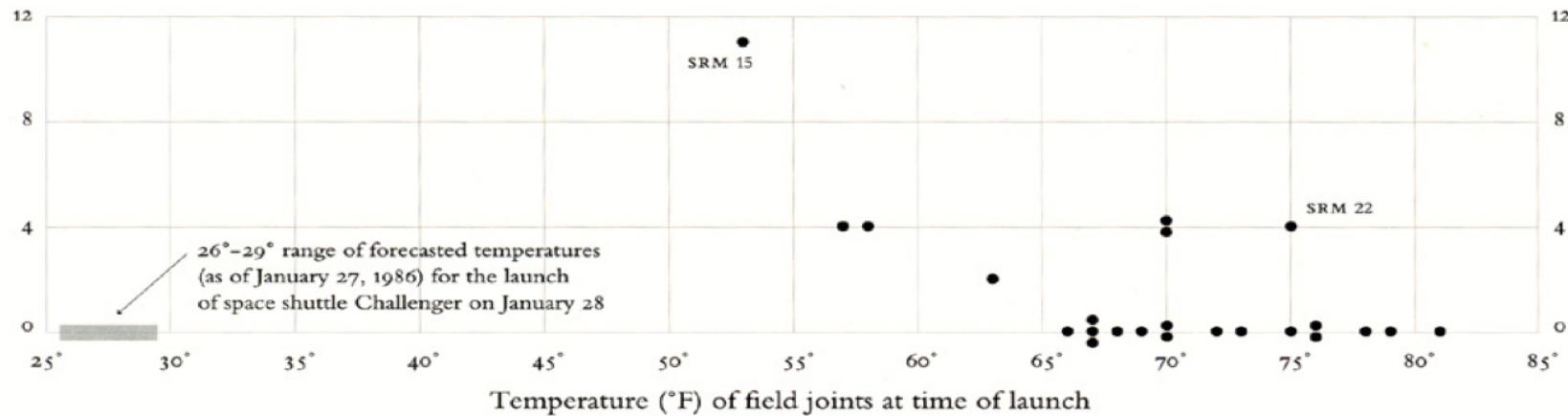
INFORMATION ON THIS PAGE WAS PREPARED TO SUPPORT AN ORAL PRESENTATION
AND CANNOT BE CONSIDERED COMPLETE WITHOUT THE ORAL DISCUSSION

[Ref. 2/26-2 1 of 3]

Tufte's Revision



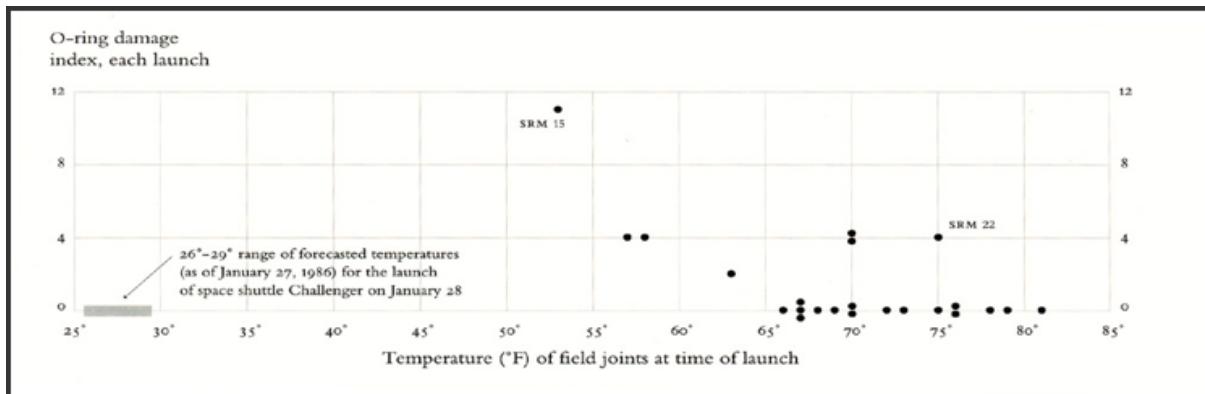
O-ring damage index, each launch



Reasoning (Exploration)



Communication (Explanation)



Impact

Example: how segregated are U.S. cities?

http://vallandingham.me/racial_divide/

Visualizations
encode data to
represent it visually



Ceci n'est pas une pipe.

Magritte

Rene Magritte, *The Treachery of Images*

Importance of Representations

Let's play a game!

- Have 9 digits: **1 2 3 4 5 6 7 8 9**
- Two players take turns selecting a digit
 - Each digit can be selected only once
- First player whose digits sum to **15** wins!

A takes **8**, B takes **2**, A takes **4**, B takes **3**, A takes **5**

*What should **B** do?!*

Importance of Representations

| | | |
|---|---|---|
| 4 | 3 | 8 |
| 9 | 5 | 1 |
| 2 | 7 | 6 |

| | | |
|---|---|---|
| 4 | 3 | 8 |
| 9 | 5 | 1 |
| 2 | 7 | 6 |

| | | |
|---|---|---|
| X | O | X |
| | X | |
| O | | |

*What should **B** do?!*

**"A good representation
captures the essential
elements of the event,
deliberately leaving out
the rest"**

- Donald Norman

Visual Channels

A visual property that can communicate meaning. An **aesthetic** that data can map to.

④ Position

→ Horizontal



→ Vertical



→ Both



④ Color



④ Shape



④ Tilt



④ Size

→ Length



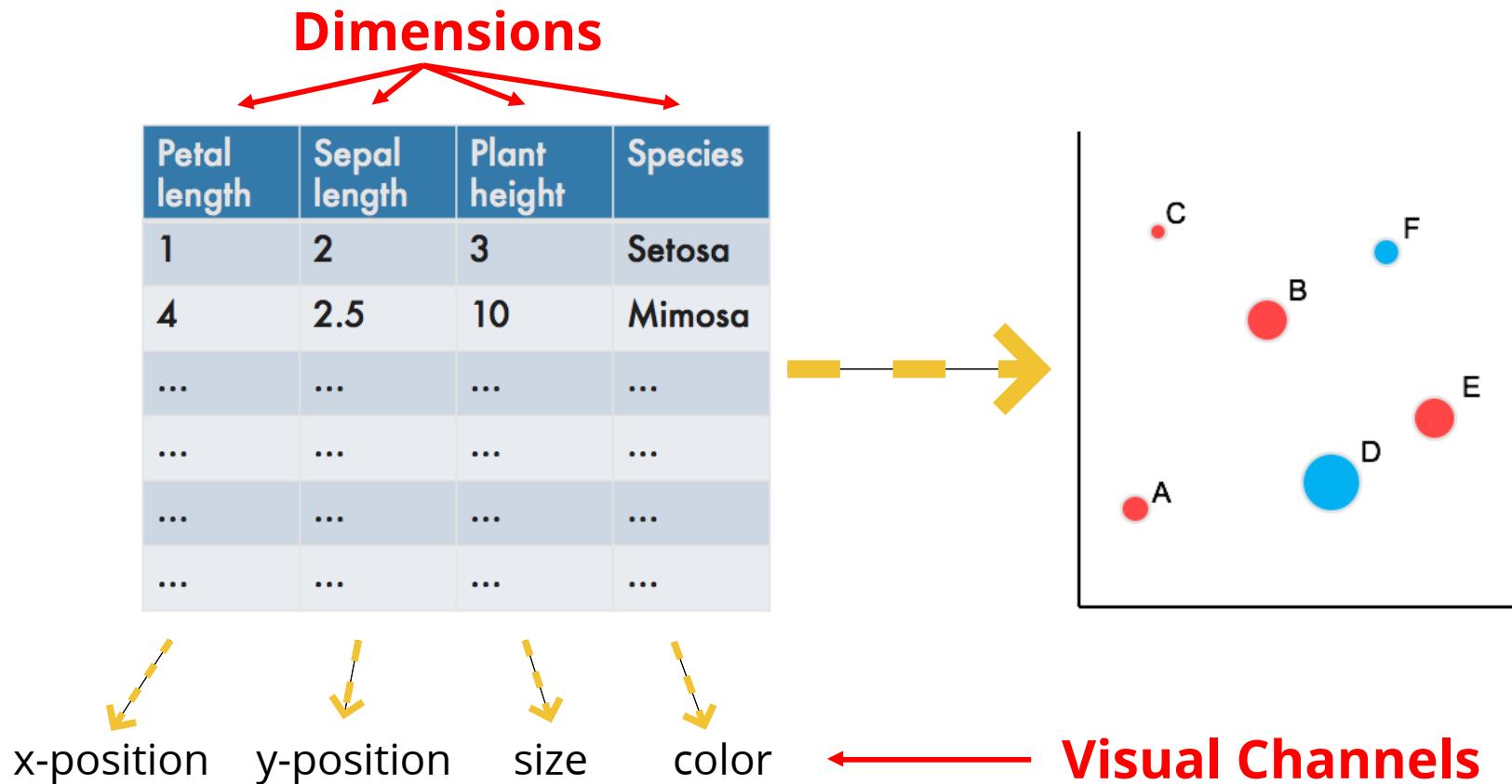
→ Area



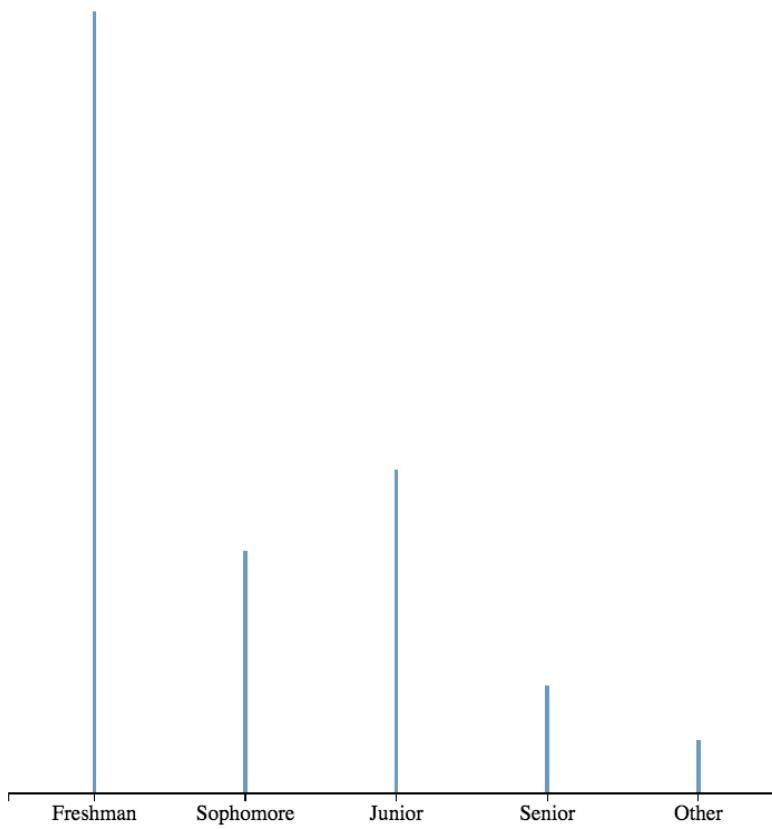
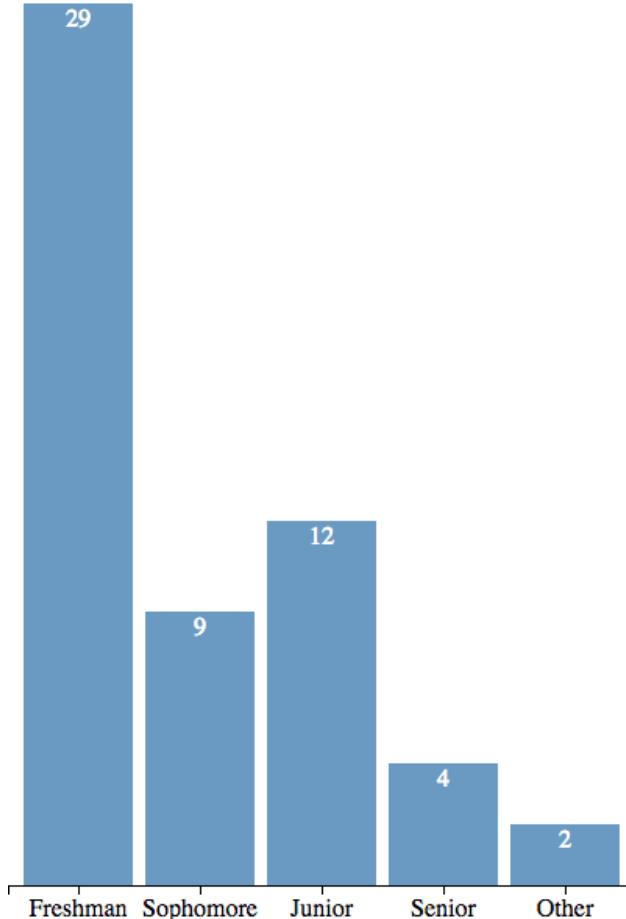
→ Volume



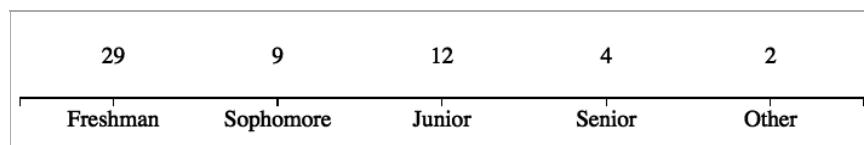
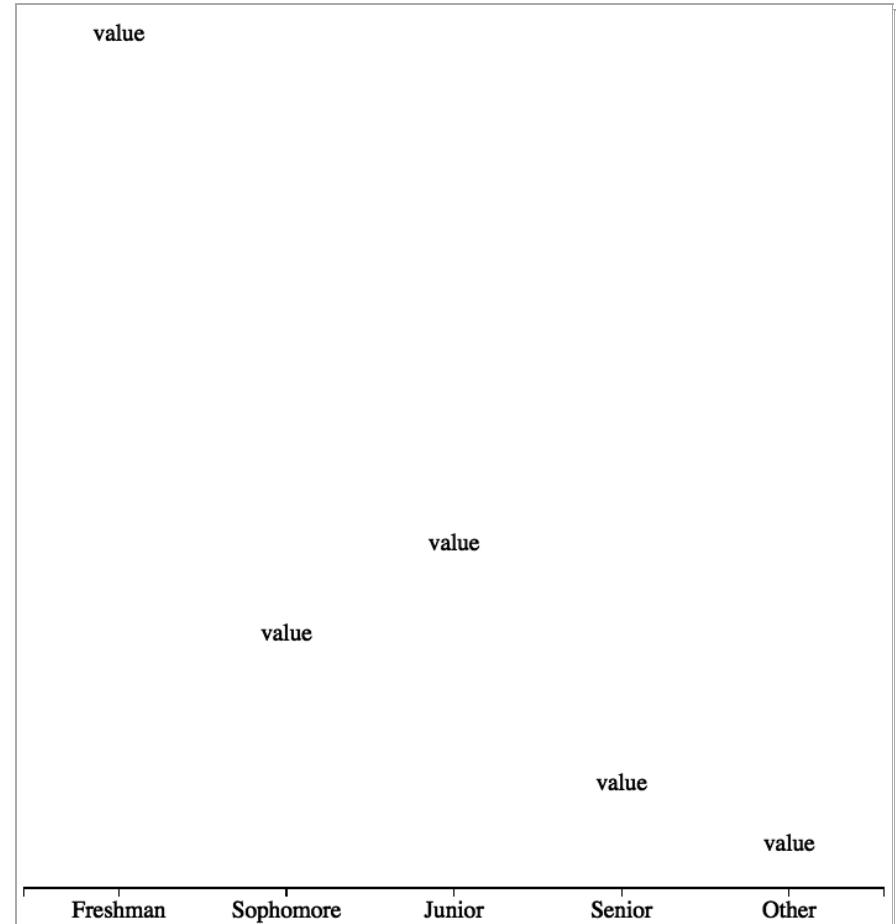
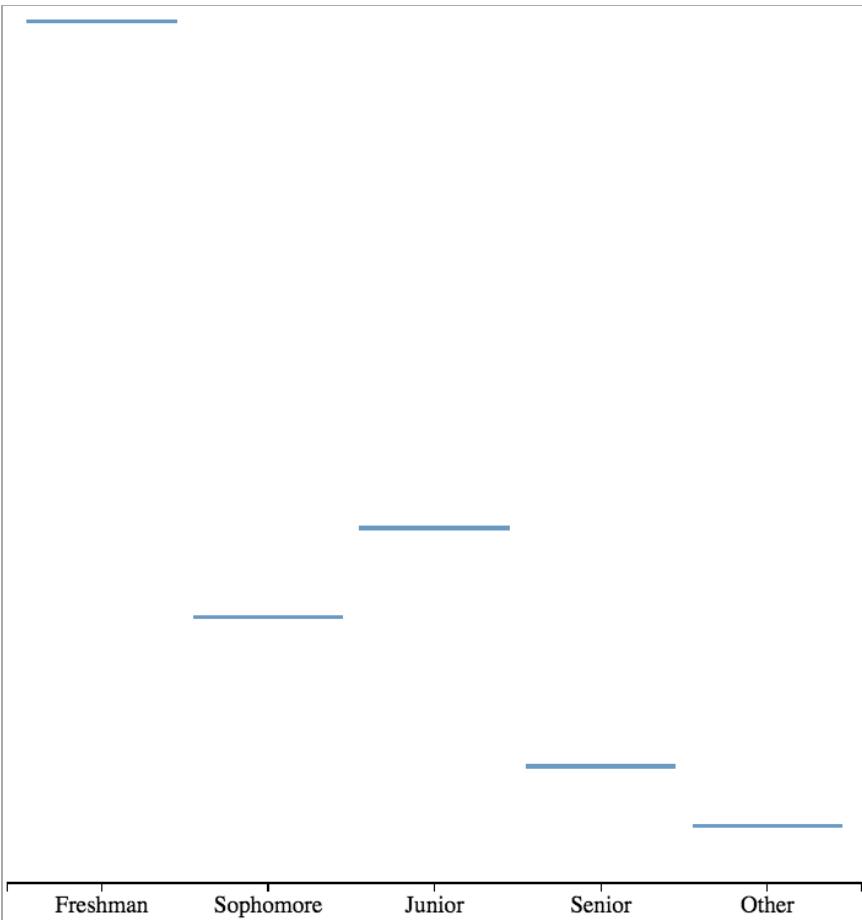
Aesthetic Mappings



Different Encodings



Different Encodings



What makes a good encoding?

Mackinlay's Expressiveness Criteria

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.

Unexpressive Visualizations

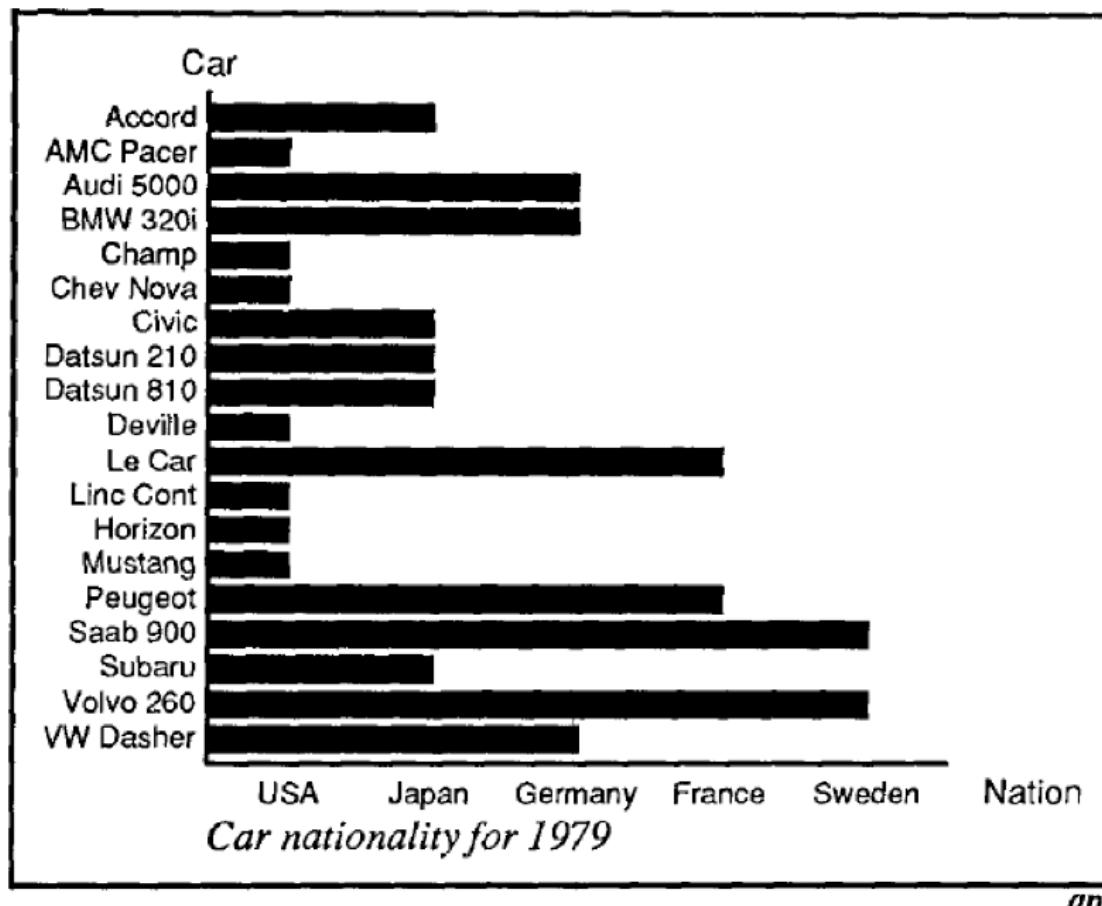


Fig. 11. Incorrect use of a bar chart for the *Nation* relation. The lengths of the bars suggest an ordering on the vertical axis, as if the USA cars were longer or better than the other cars, which is not true for the *Nation* relation.

(Mackinlay, 1986)

Mackinlay's Effectiveness Criteria

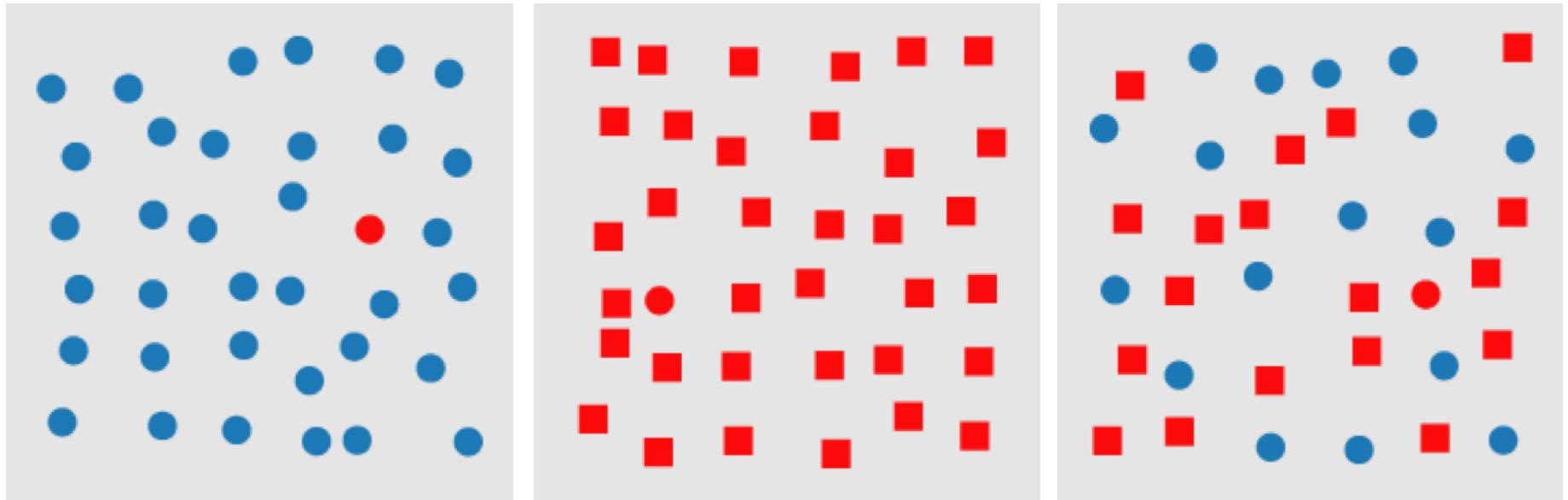
A visualization is more **effective** than another visualization if the information conveyed by one visualization is more readily **perceived** than the information in the other visualization.

28049385628406947862485839220894862089476
90187850988342609284687248598238240985246
87498752208948520298485092485329088452029
88452902884352892842589987458784958784985

How many 3's are there?

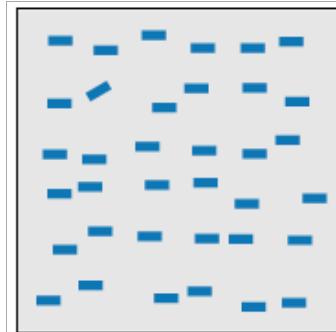
28049385628406947862485839220894862089476
90187850988342609284687248598238240985246
87498752208948520298485092485329088452029
88452902884352892842589987458784958784985

How many 3's are there?

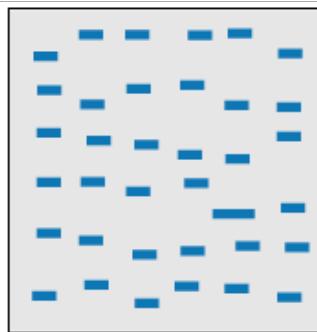


Find the red dot

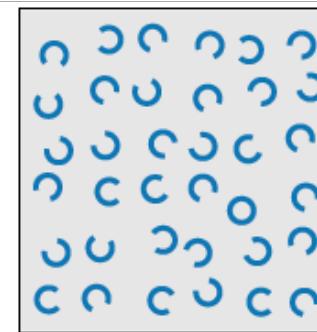
Pre-attentive Visual Perception



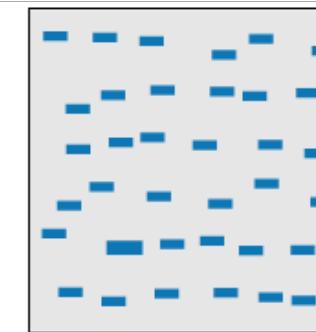
line (blob) orientation
Julész & Bergen 83; Sagi &
Julész 85a, Wolfe et al. 92; Wei-
gle et al. 2000



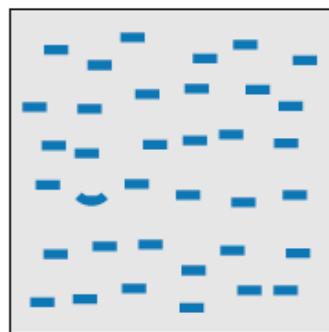
length, width
Sagi & Julész 85b; Treisman &
Gormican 88



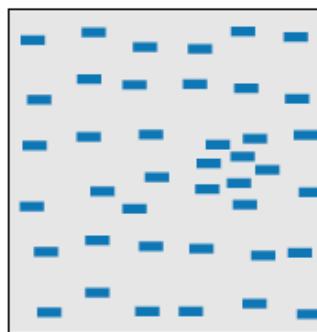
closure
Julész & Bergen 83



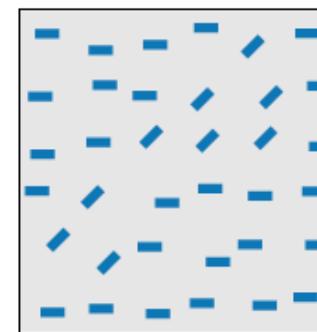
size
Treisman & Gelade 80; Healey &
Enns 98; Healey & Enns 99



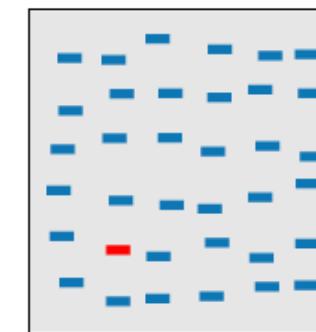
curvature
Treisman & Gormican 88



density, contrast
Healey & Enns 98; Healey &
Enns 99



number, estimation
Sagi & Julész 85b; Healey et al.
93; Trick & Pylyshyn 94



colour (hue)
Nagy & Sanchez 90; Nagy et al.
90; D'Zmura 91; Kawai et al. 95;
Bauer et al. 96; Healey 96; Bauer
et al. 98; Healey & Enns 99



**How do we choose
a visual channel to
map to?**

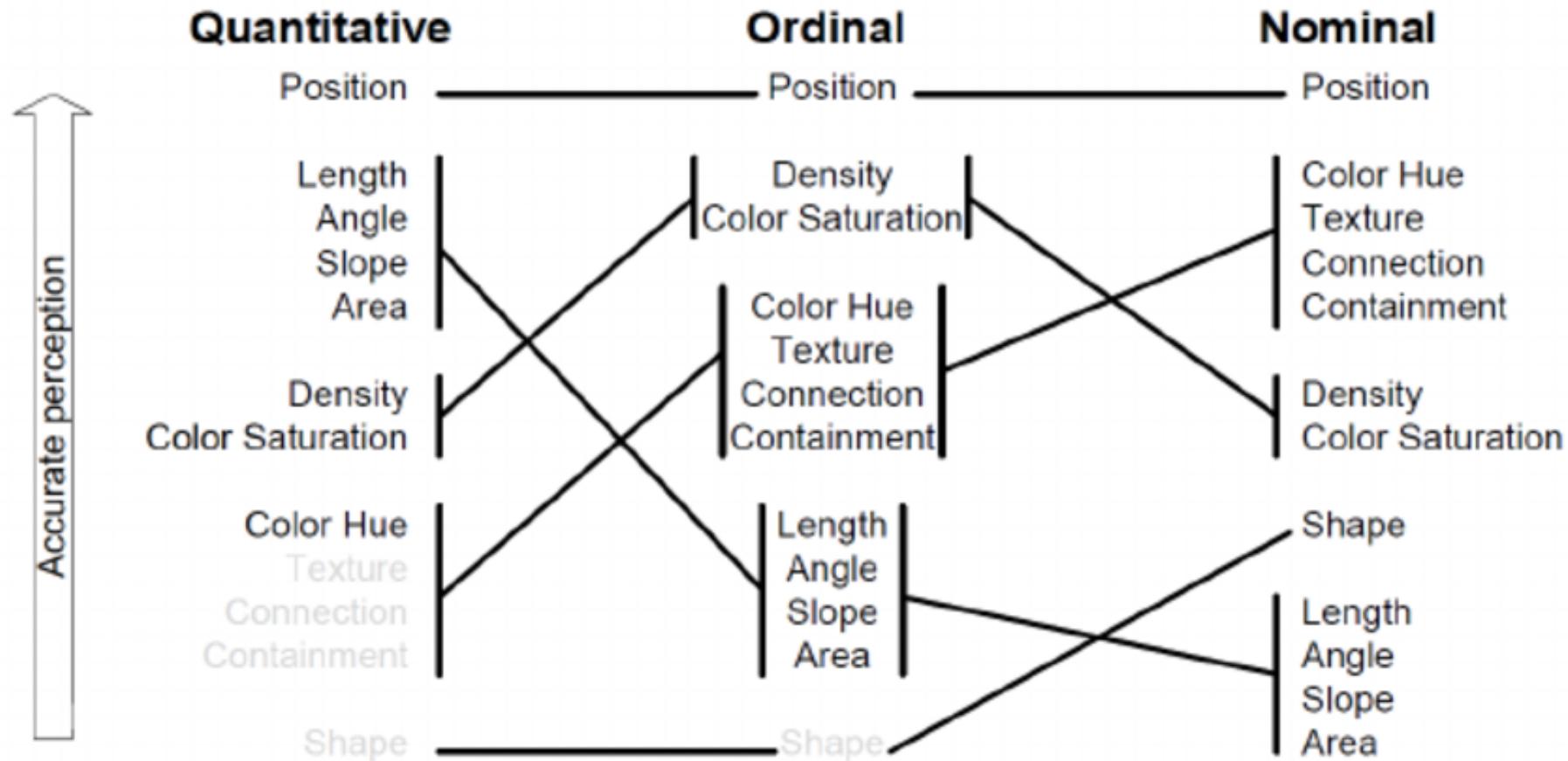
Level of Measurement

RECALL

A way of classifying the nature of data values. Applies to all data analysis, distinct from the R "data type".

| Level | Example | Operations |
|---|--|--|
| ★ Nominal unordered used for classification | <i>Fruits:</i> apples, bananas, oranges, etc. | <code>== !=</code> "same or different" |
| ★ Ordinal ordered can comparison | <i>Grade of meat:</i> Grade A, Grade AA, Grade AAA, etc. | <code>== != < ></code> "bigger or smaller" |
| ★ Interval (Quantitative) ordered, no set "zero" can find difference | <i>Dates:</i> 05/15/2012, 04/17/2015, etc. | <code>== != < ></code> <code>+ -</code> "3 units bigger" |
| Ratio (Quantitative) ordered, fixed "zero" can find magnitude | <i>Lengths:</i> 1 inch, 1.5 inches, 2 inches, etc. | <code>== != < ></code> <code>+ - * /</code> "twice as big" |

Visual Channel Effectiveness



Position

The most accurate visual channel for all data types.

Resemblance (nominal)

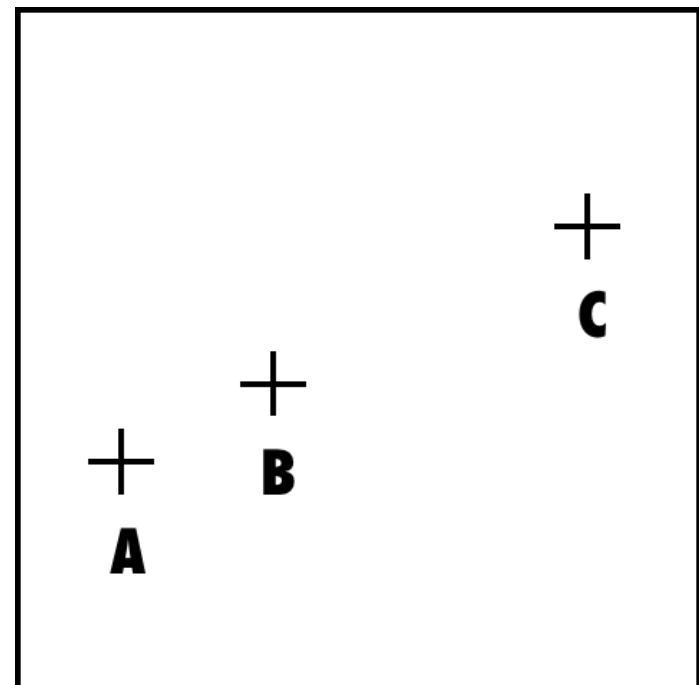
(A != B != C)

Order (ordinal)

(B is between A and C)

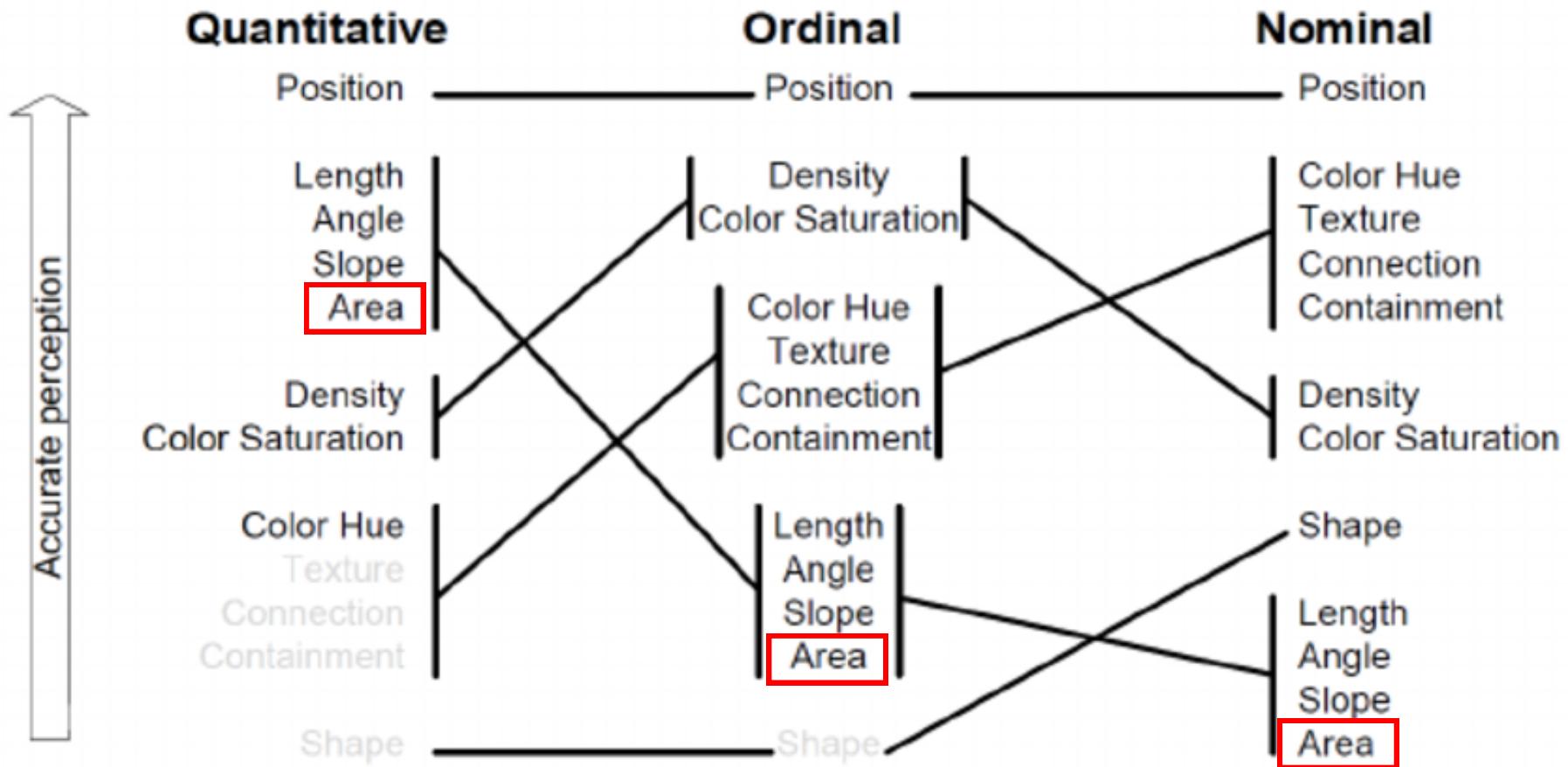
Proportion (quantitative)

(BC is 2x long as AB)



Size

Less accurate as data is less quantitative



Color: Hue, Brightness, Saturation

Hue

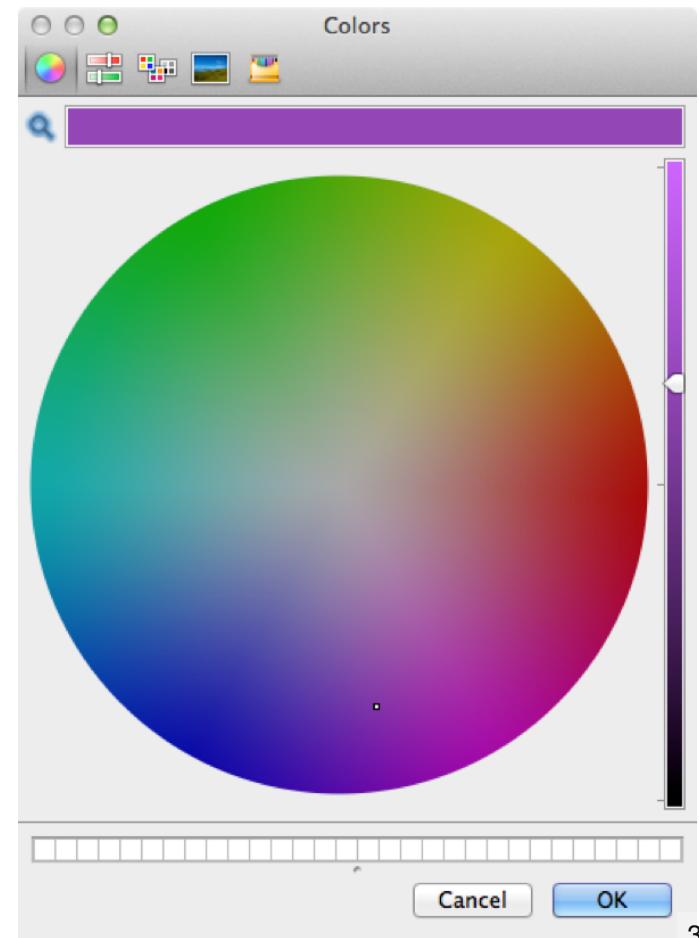
(red, yellow, green, etc)

Saturation

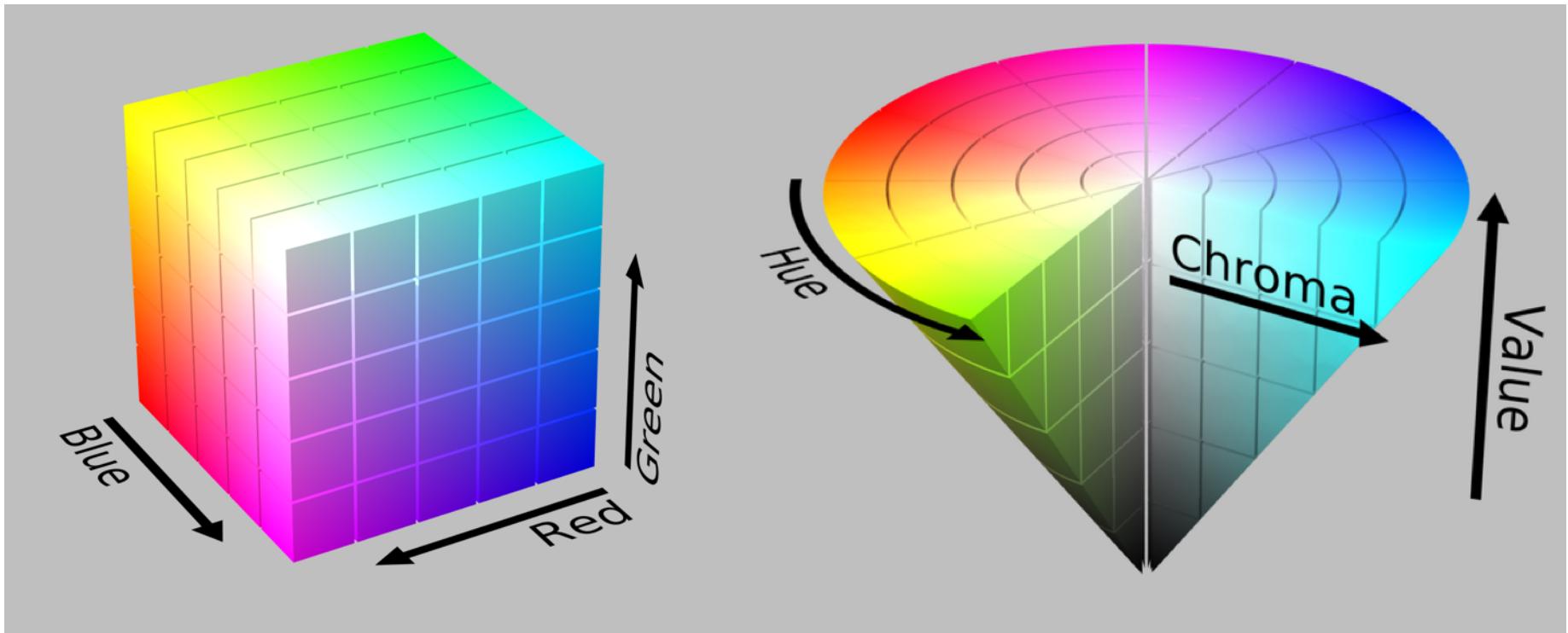
(vivid vs. washed out)

Brightness or Value

(luminance)

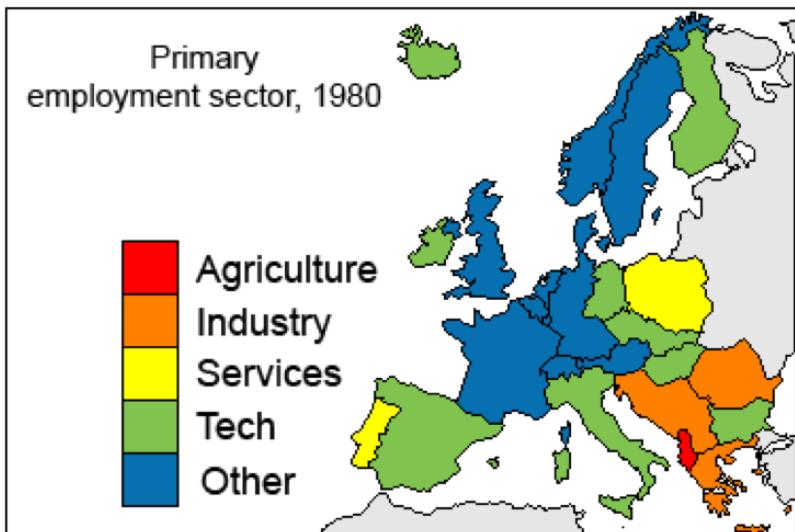


Color: RGB vs. HSB

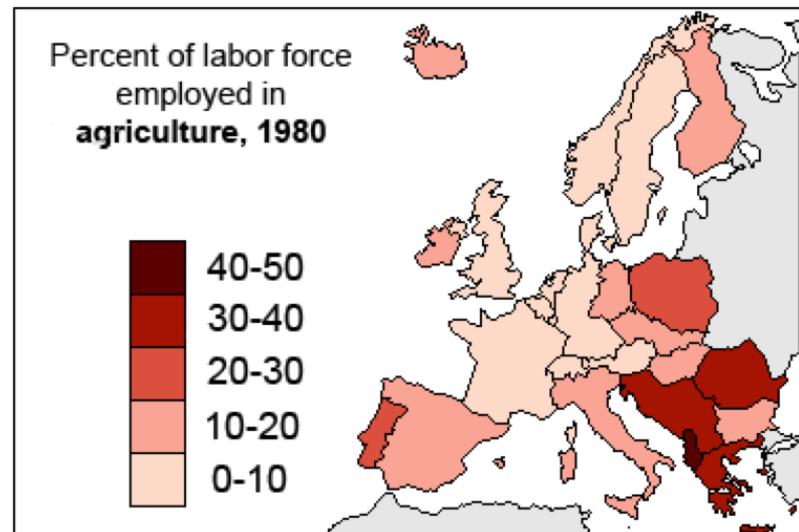


Color

**Hue is good for *nominal* variables,
Saturation is good for *ordinal* variables**



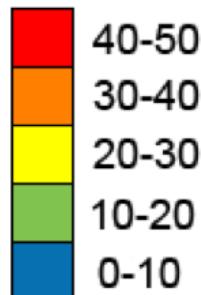
Hue (nominal)



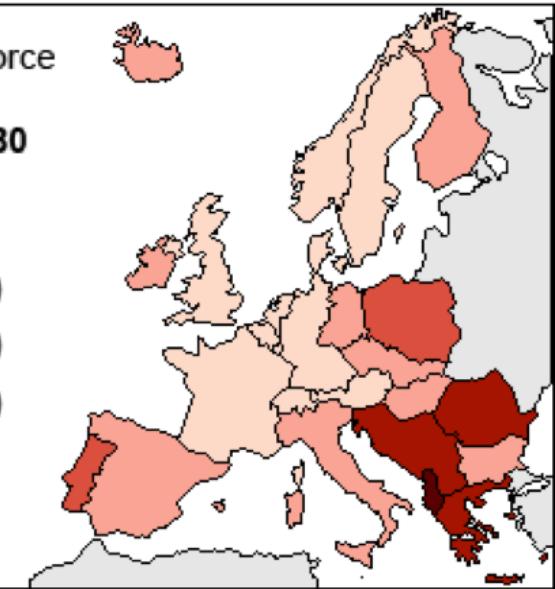
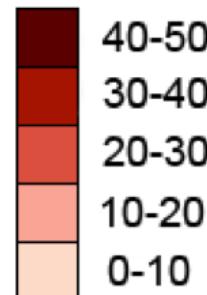
Saturation (ordinal)

Which is more effective?

Percent of labor force
employed in
agriculture, 1980



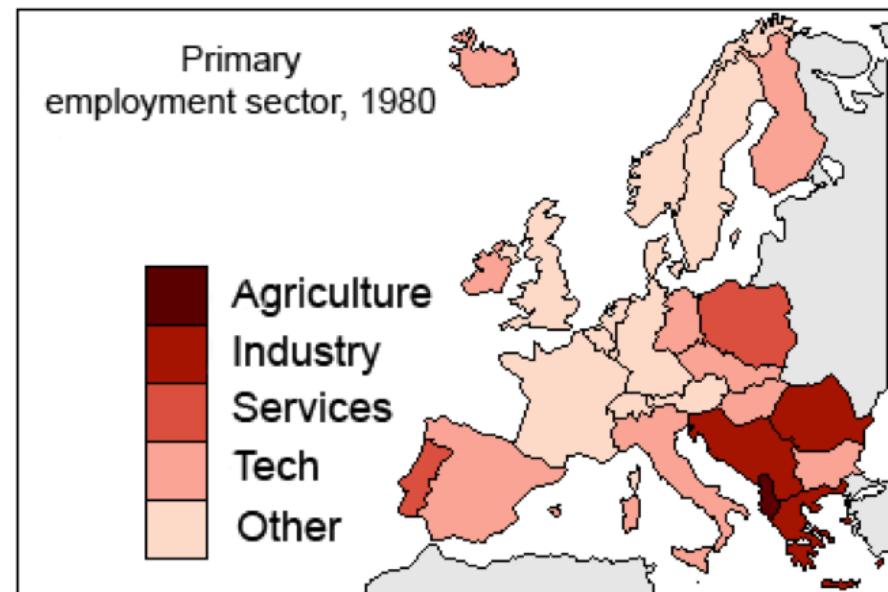
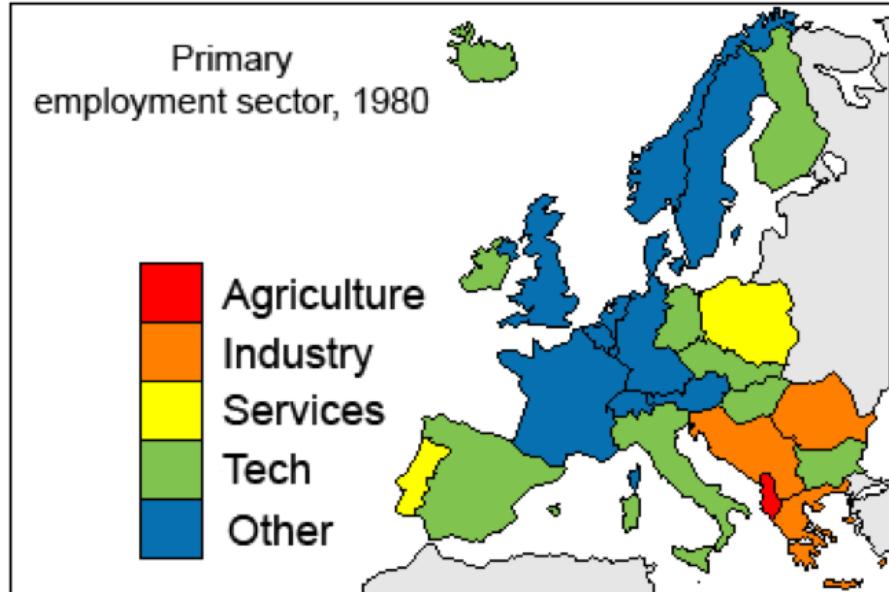
Percent of labor force
employed in
agriculture, 1980



A

B

Which is more effective?



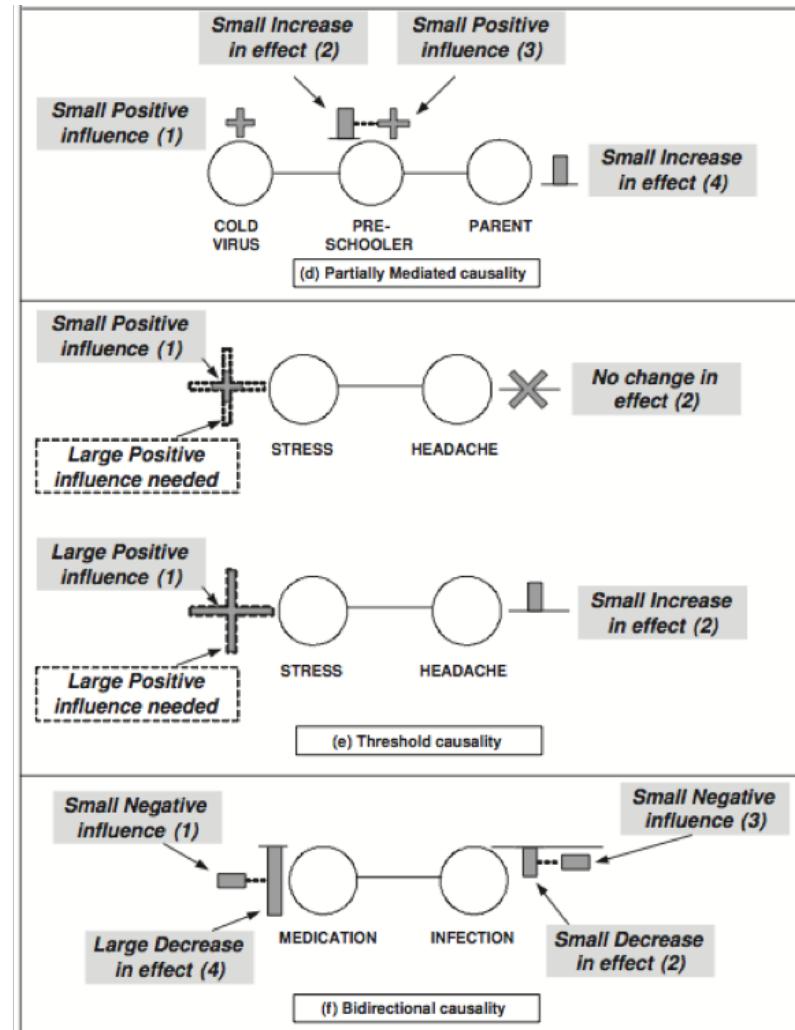
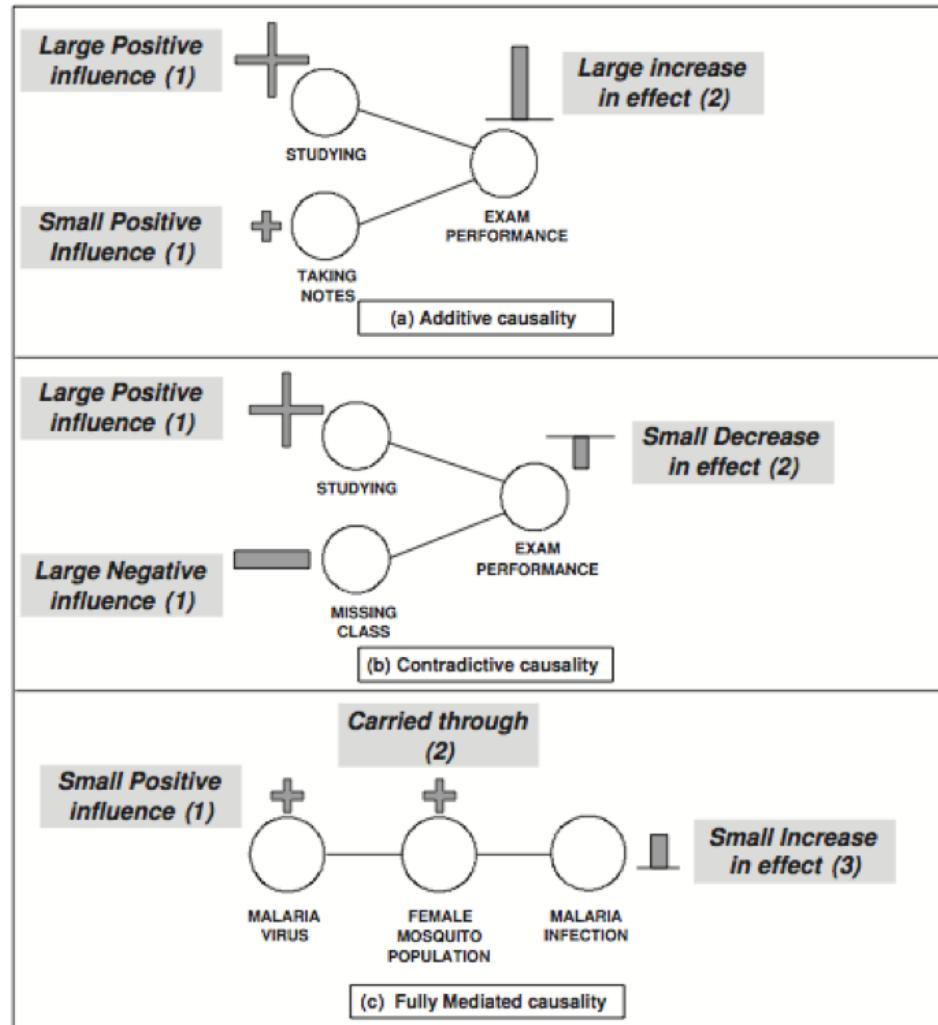
A

B

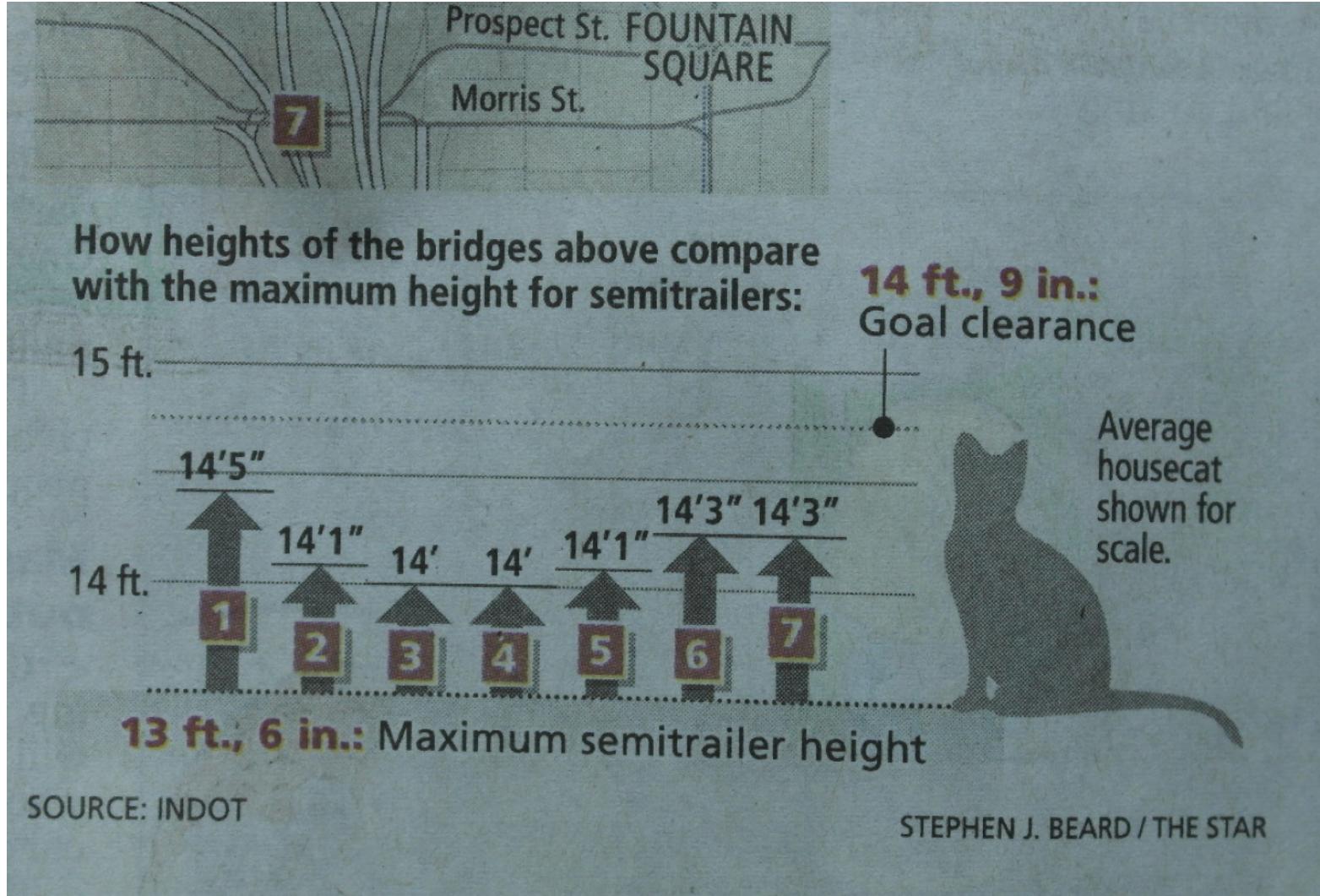
Other Channels: Orientation

<http://hint.fm/wind/>

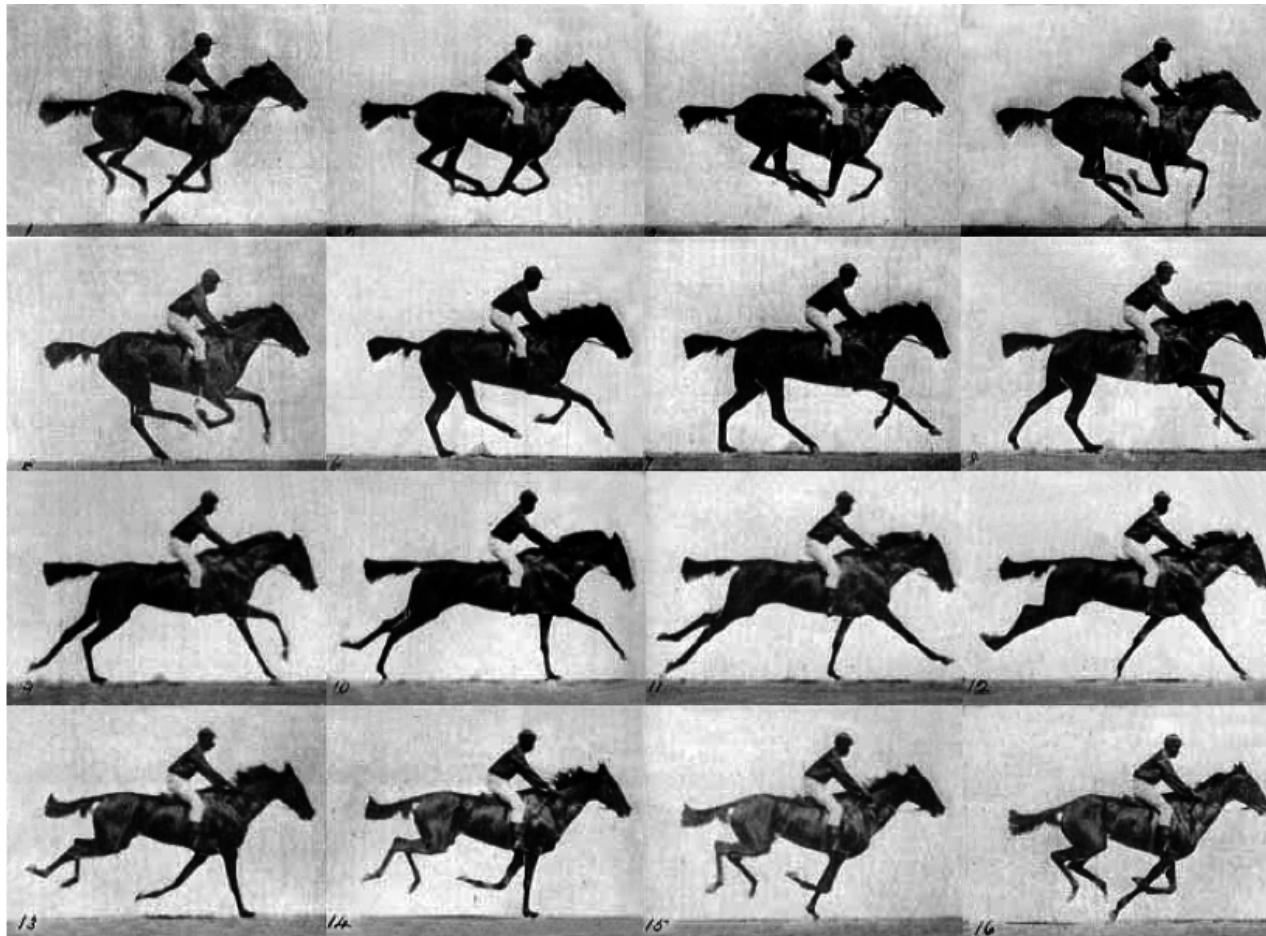
Labels & Glyphs



Poor Labeling



Animation



E.g.: Michotte's Launching Effect (visualizing causality)

<http://cogweb.ucla.edu/Discourse/Narrative/michotte-demo.swf>

**Utilize the channels
that are most effective
for the data that is
most important**

**But you need to know what data is
important to your audience!**

Back to programming....

Grammar of Graphics

RECALL

Words used to describe the visual components and aspects of a graphic.

- **Data** shown in the plot
- **Geometric objects (geoms)** that appear on the plot
- **Aesthetic mappings** from the data to the **geoms**
- **Statistical transformation** used to calculate the data
- **Scales** (range of values) for each **aesthetic**
- **Coordinate system** to organize the **geoms**
- **Facets** or groups of data shown in different plots

RECALL

ggplot2

Use the `ggplot()` function to draw a plot, specifying plot elements via the grammar.

```
# plot the `mpg` data set, with highway milage  
# on the x axis and engine displacement (power)  
# on the y axis:
```

```
ggplot(data = mpg) +  
  geom_point(mapping = aes(x = displ, y = hwy))
```

↑
**geometric objects
(points)**

↑
**aesthetic
mappings**

RECALL

Facets

Break a plot into parts with **facets** (similar to `group_by()` in `dplyr`). Each facet acts like a "level" in a factor, with a plot for each level.

```
# a plot with facets based on vehicle type.  
ggplot(data = mpg) +  
  geom_point(mapping = aes(x = displ, y = hwy)) +  
  facet_wrap(~class)
```



A formula, read as
"as a function of"

Consider a data set...

| | name | section | math_exam1 | math_exam2 | spanish_exam1 | spanish_exam2 |
|---|--------|---------|------------|------------|---------------|---------------|
| 1 | Mason | a | 91 | 88 | 79 | 99 |
| 2 | Tabi | a | 82 | 79 | 88 | 92 |
| 3 | Bryce | a | 93 | 77 | 92 | 92 |
| 4 | Ada | b | 100 | 99 | 83 | 82 |
| 5 | Bob | b | 78 | 88 | 87 | 85 |
| 6 | Filipe | b | 91 | 93 | 77 | 95 |

What if we want to
facet by exam?

Data Shape

Wide Data

| | name | section | math_exam1 | math_exam2 | spanish_exam1 | spanish_exam2 |
|---|--------|---------|------------|------------|---------------|---------------|
| 1 | Mason | a | 91 | 88 | 79 | 99 |
| 2 | Tabi | a | 82 | 79 | 88 | 92 |
| 3 | Bryce | a | 93 | 77 | 92 | 92 |
| 4 | Ada | b | 100 | 99 | 83 | 82 |
| 5 | Bob | b | 78 | 88 | 87 | 85 |
| 6 | Filipe | b | 91 | 93 | 77 | 95 |

6 rows x 4 cols
= 24 scores

Long Data

| | name | section | exam | score |
|----|--------|---------|---------------|-------|
| 1 | Mason | a | math_exam1 | 91 |
| 2 | Tabi | a | math_exam1 | 82 |
| 3 | Bryce | a | math_exam1 | 93 |
| 4 | Ada | b | math_exam1 | 100 |
| 5 | Bob | b | math_exam1 | 78 |
| 6 | Filipe | b | math_exam1 | 91 |
| 7 | Mason | a | math_exam2 | 88 |
| 8 | Tabi | a | math_exam2 | 79 |
| 9 | Bryce | a | math_exam2 | 77 |
| 10 | Ada | b | math_exam2 | 99 |
| 11 | Bob | b | math_exam2 | 88 |
| 12 | Filipe | b | math_exam2 | 93 |
| 13 | Mason | a | spanish_exam1 | 79 |
| 14 | Tabi | a | spanish_exam1 | 88 |

24 rows x 1 col
= 24 scores

Data Shape

We can convert between **wide** and **long** data (and vice versa) using the **tidyverse** package.

```
# Alternatively, install "tidyverse"
install.packages("tidyverse") # once per machine
library("tidyverse")

# Make a data.frame (example)
students <- data.frame(
  name = c('Mason', 'Tabi', 'Bryce', 'Ada', 'Bob', 'Filipe'),
  section = c('a', 'a', 'a', 'b', 'b', 'b'),
  math_exam1 = c(91, 82, 93, 100, 78, 91),
  math_exam2 = c(88, 79, 77, 99, 88, 93),
  spanish_exam1 = c(79, 88, 92, 83, 87, 77),
  spanish_exam2 = c(99, 92, 92, 82, 85, 95)
)
```

<http://tidyverse.org/>

<http://tidyverse.org/>

Data Shape

Convert from **wide** to **long** using `gather()`. The **key** is a new column containing *gathered colnames*, and **value** is a new column with their values.

```
students.long <- gather(students.wide,  
                         key = exam,  
                         value = score,  
                         math_exam1, math_exam2,  
                         spanish_exam1, spanish_exam2  
                         )
```

names for new columns

col data to populate with

Convert from **long** to **wide** using `spread()`. The **key** is where to get the *new colnames*, and **value** is where to get the values

```
# spread by column "exam"  
stu.wide <- spread(students.long, key = exam, value = score)  
  
# spread by column "name"  
stu.wide.name <-  
    spread(students.long, key = name, value = score)
```

Map Visualizations

Polygons

Use **geom_polygon** to draw shapes.

```
rect <- data.frame(x.coords = c(3, 5, 5, 3),  
                    y.coords = c(4, 4, 2, 2))  
  
ggplot(data = rect) +  
  geom_polygon(aes(x = x.coords, y = y.coords))
```

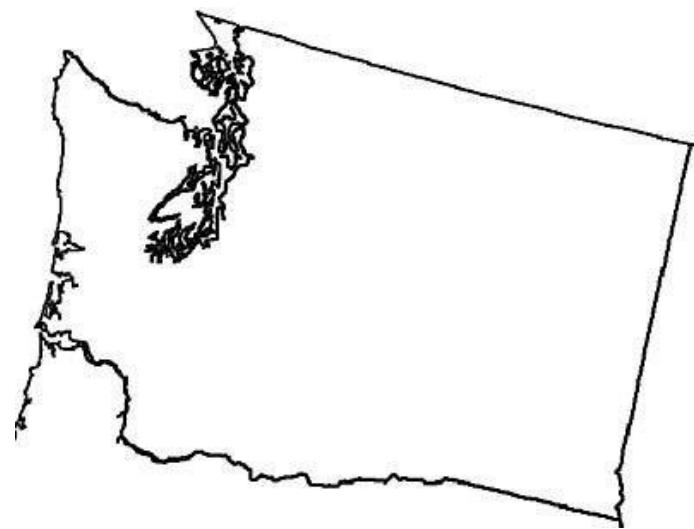
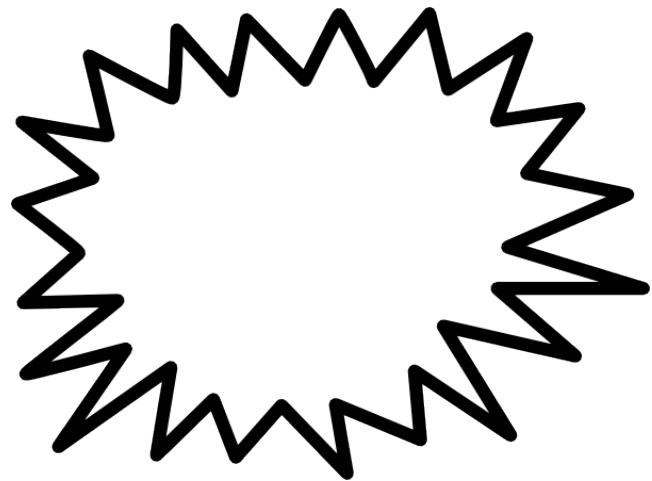
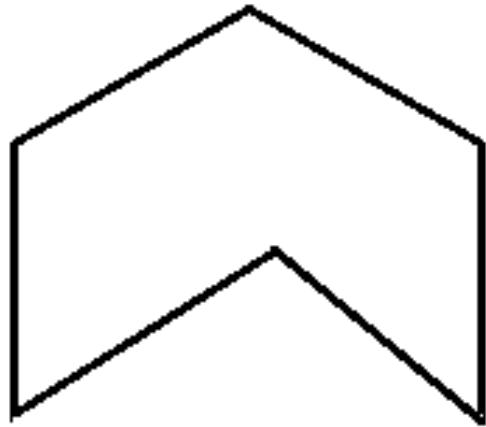
each row is a corner point

Draw multiple polygons by **grouping** points.

```
double.rect <- data.frame(x.coords = c(1,2,2,1, 3,4,4,3),  
                           y.coords = c(2,2,1,1, 2,2,1,1),  
                           rect.num = c(1,1,1,1, 2,2,2,2))  
  
ggplot(data = double.rect) +  
  geom_polygon(aes(x=x.coords, y=y.coords, group = rect.num))
```

which rect the point goes with

Polygons



Map Data

ggplot2 provides a set of data frames (from the `maps` library) which include polygon definitions for different geographic maps.

Access these data frames with the `map_data()` function.

```
# load the data
usa.states <- map_data("state")

# plot the polygons
ggplot(data = usa.states) +
  geom_polygon(aes(x = long, y = lat, group = group)) +
  coord_quickmap() # map coordinate system!
```

Module 13 exercise-3



Clone a fresh copy, or pull changes!

```
# In Terminal (in your repo directory)

# Add a new "bookmark" to remote called "upstream" to the repo
git remote add upstream https://github.com/info201-w17/module13-ggplot2

# download changes from the upstream repo
git pull upstream master
```

Action Items!

- Be comfortable with **module 13**
- Assignment 5 due ***Thursday before class***
 - (Assignment 6 online soon)

Thursday: What makes a good visualization?
Also maps.