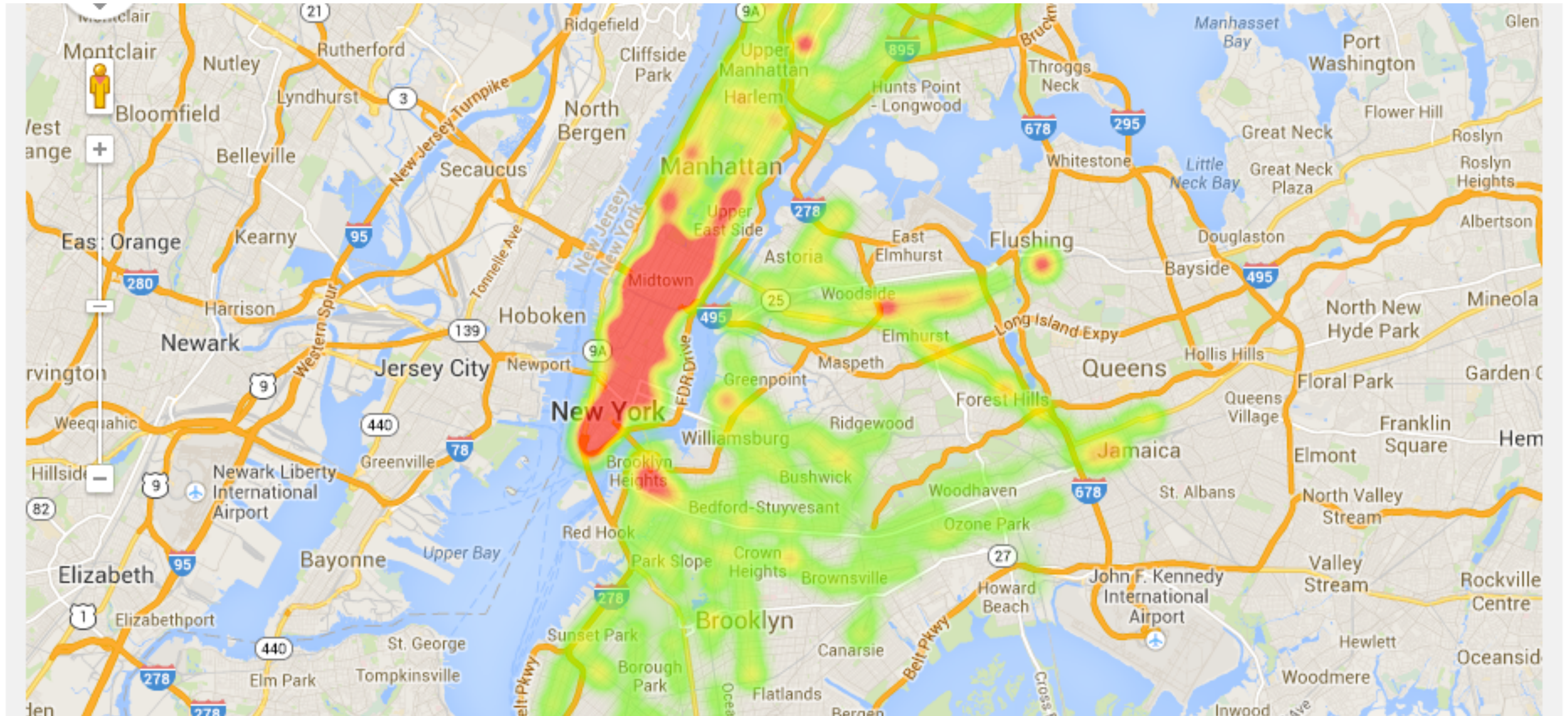


Programming Principles in Python (CSCI 503/490)

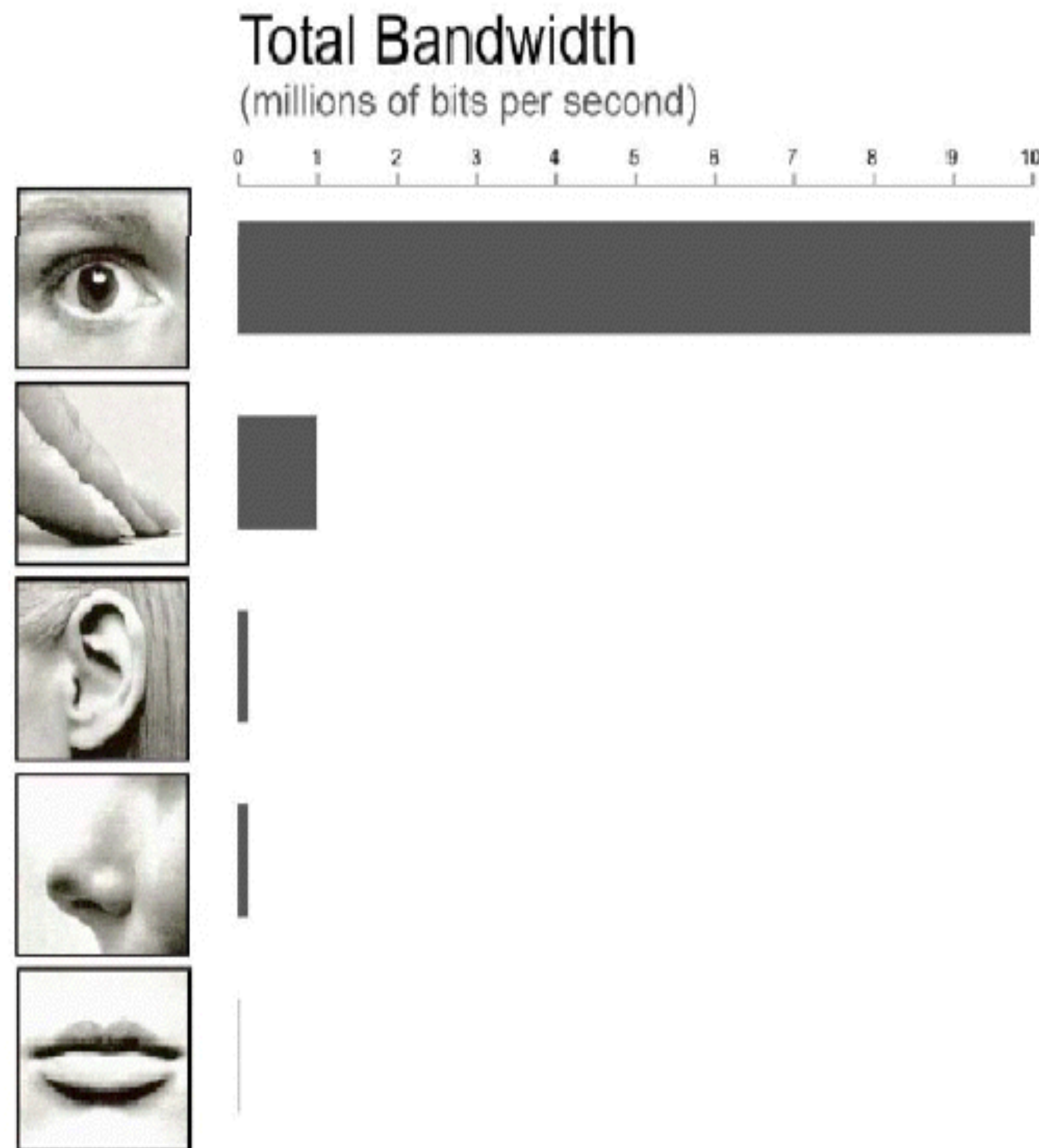
Data Visualization & Machine Learning

Dr. David Koop

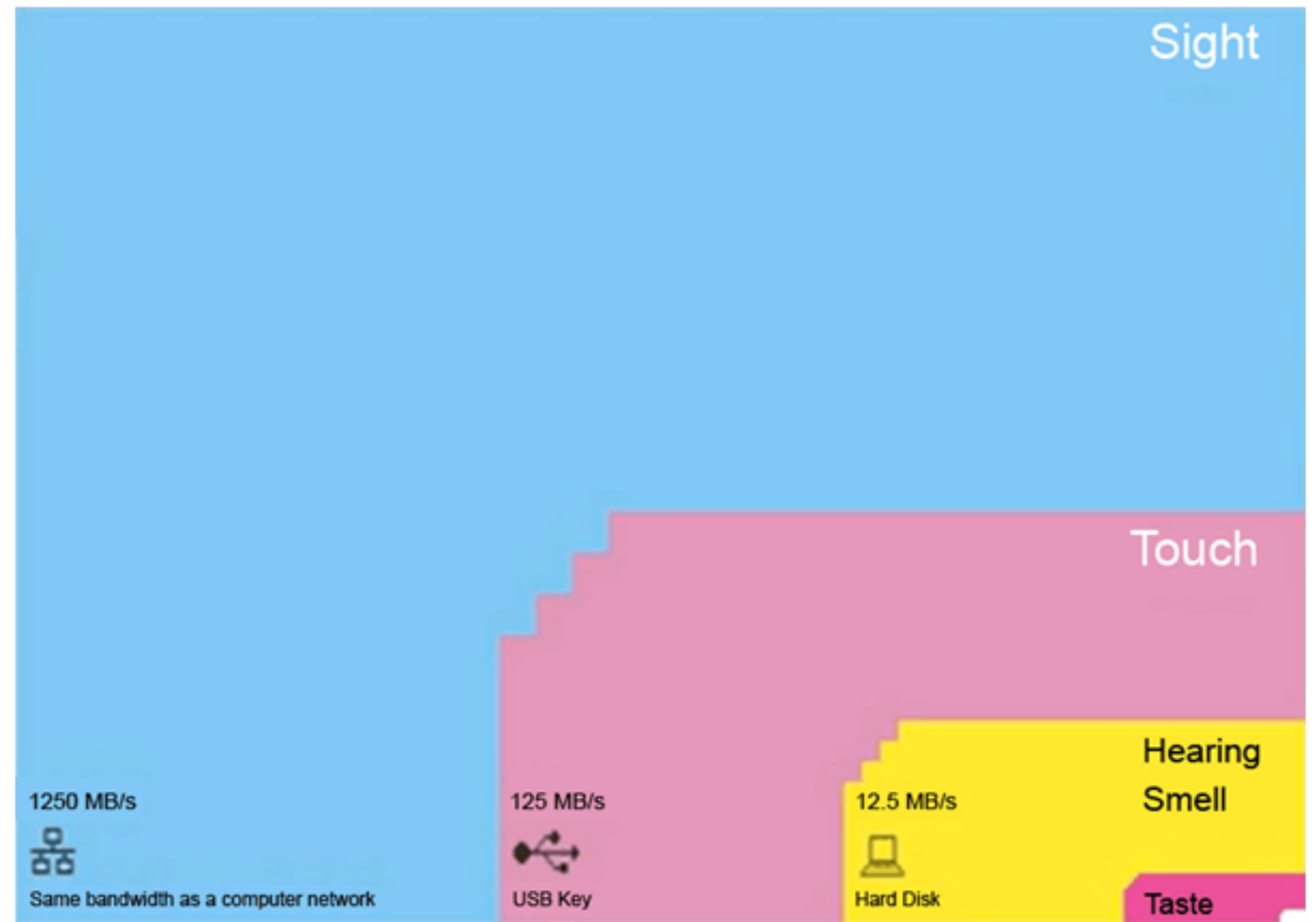
Exploring Data through Visualization



Why do we visualize data?

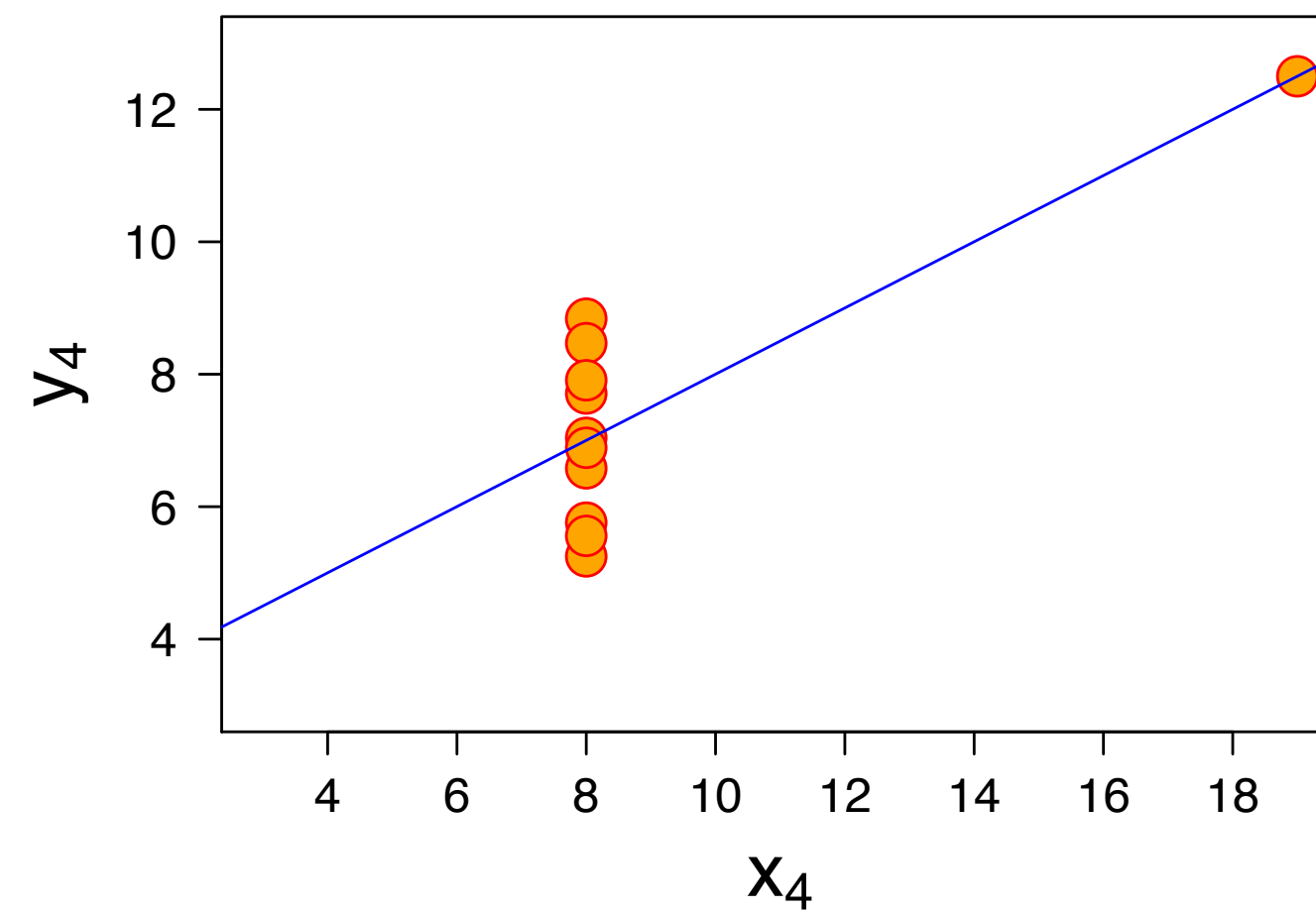
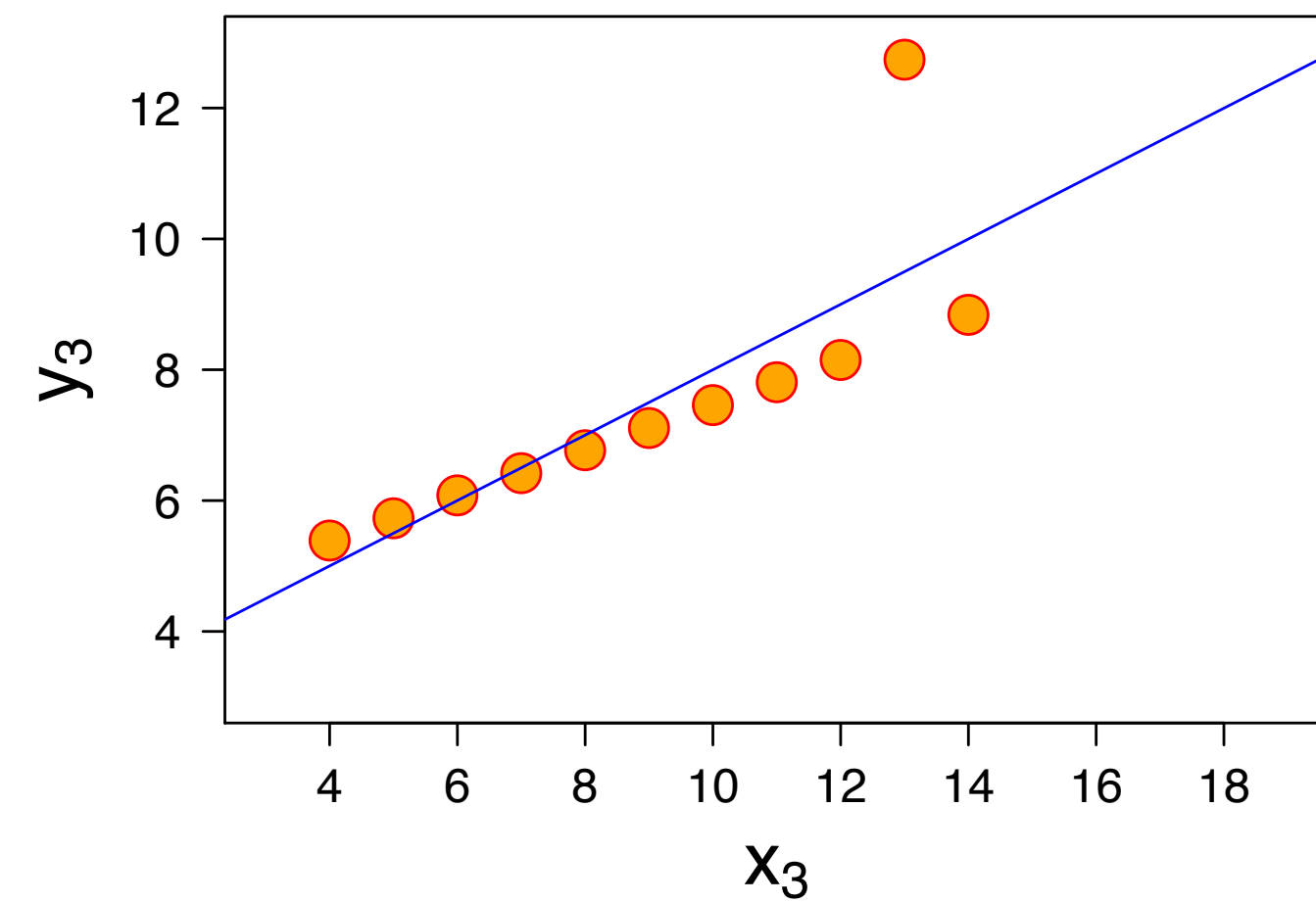
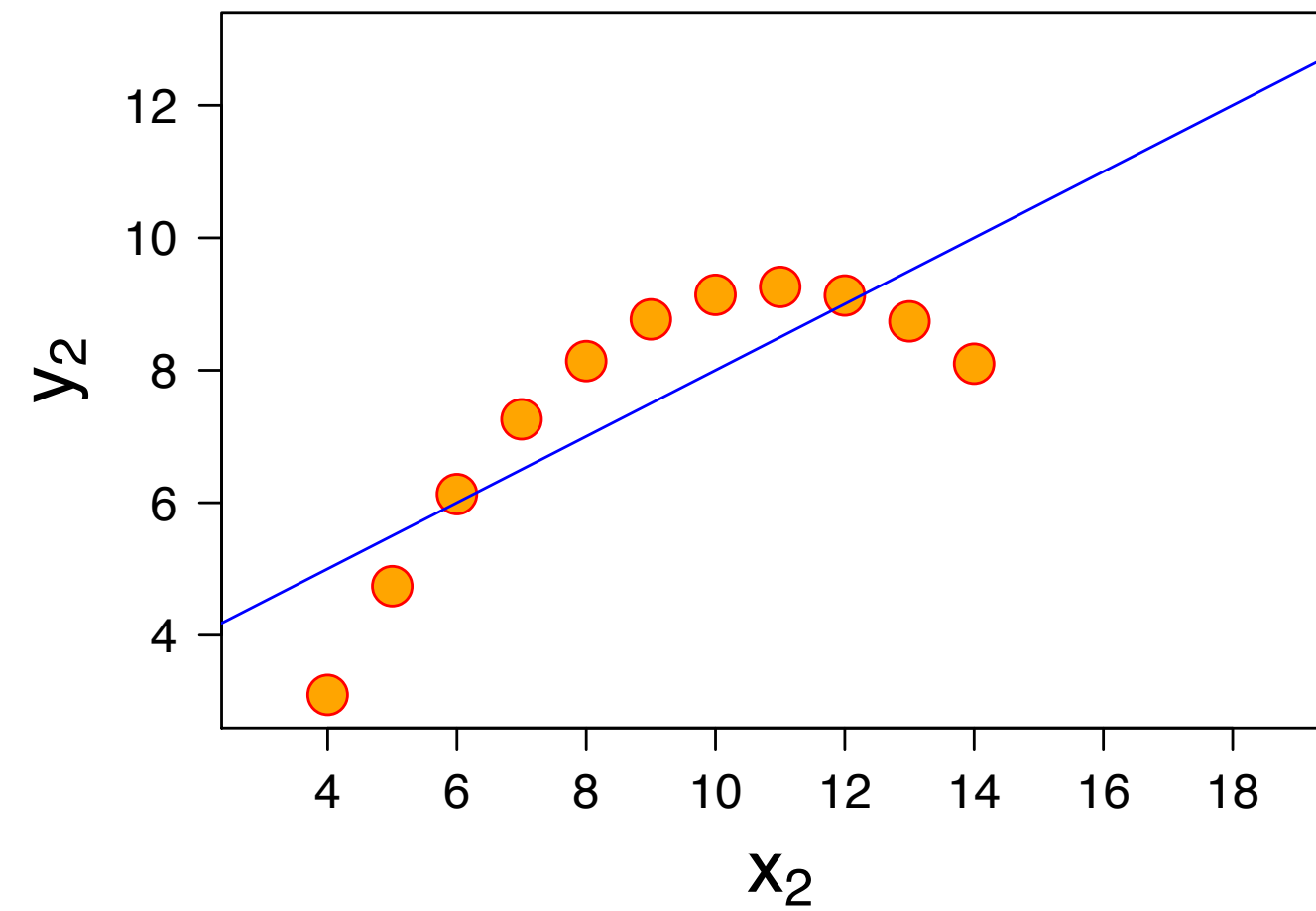
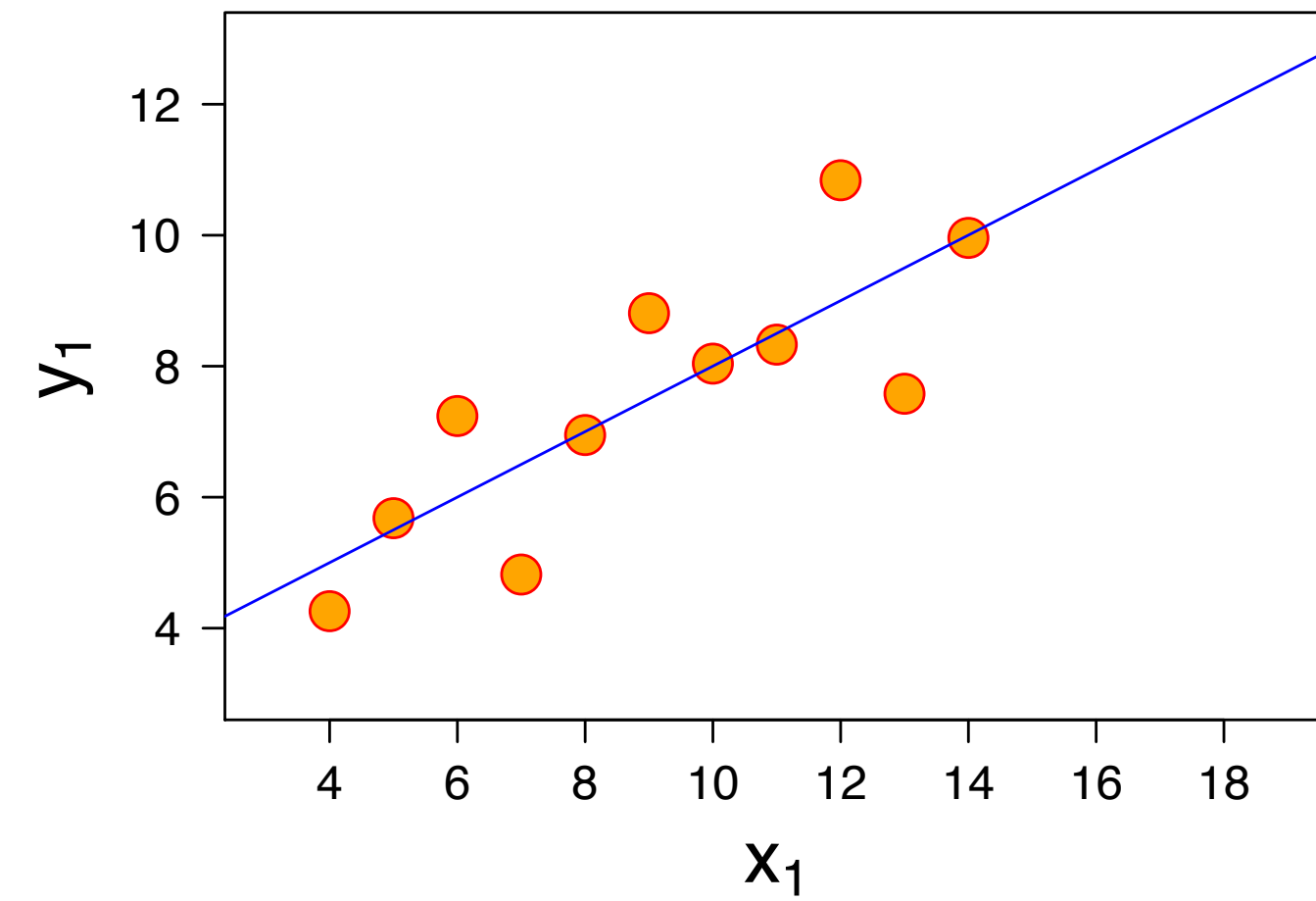


[via A. Lex]



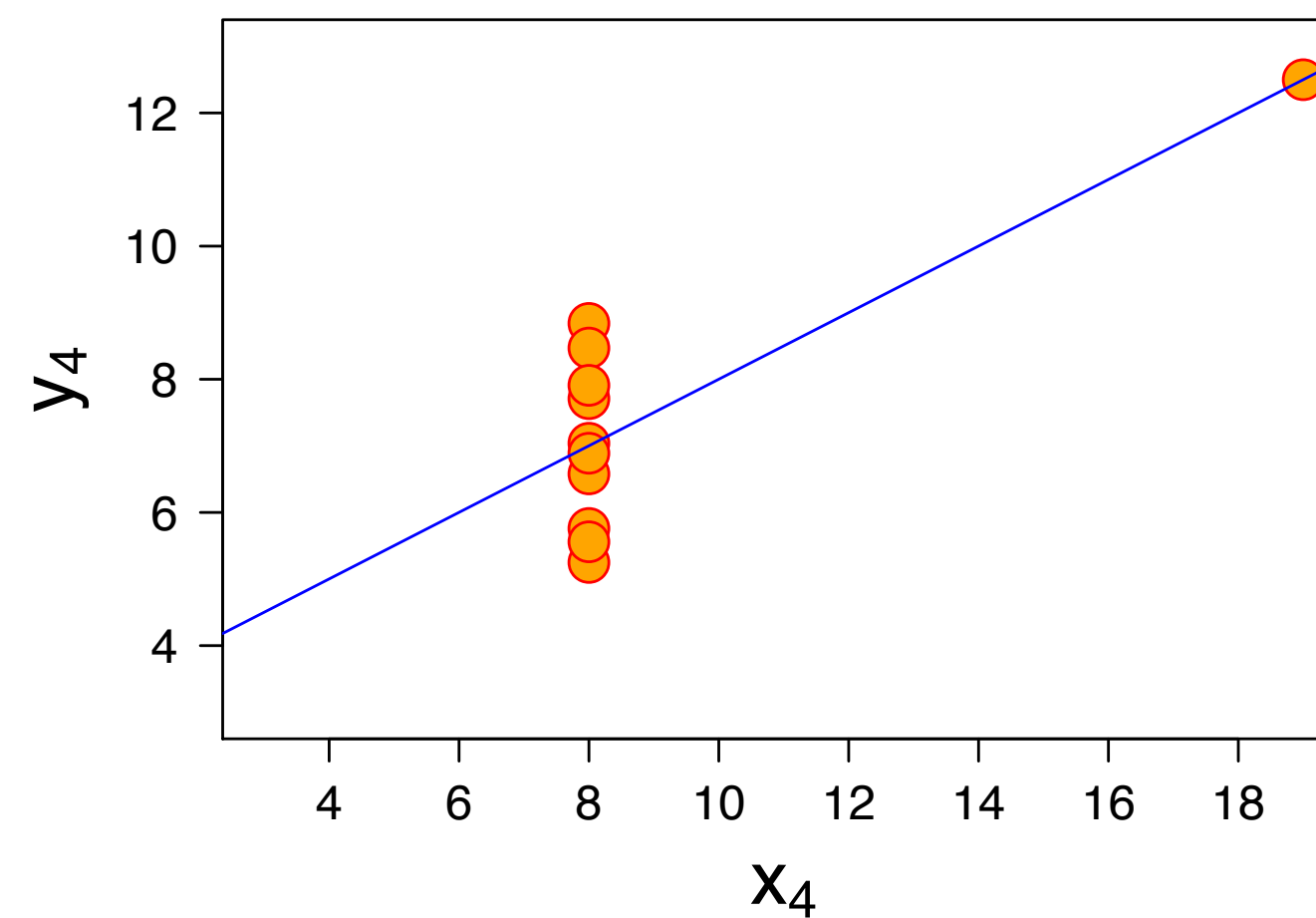
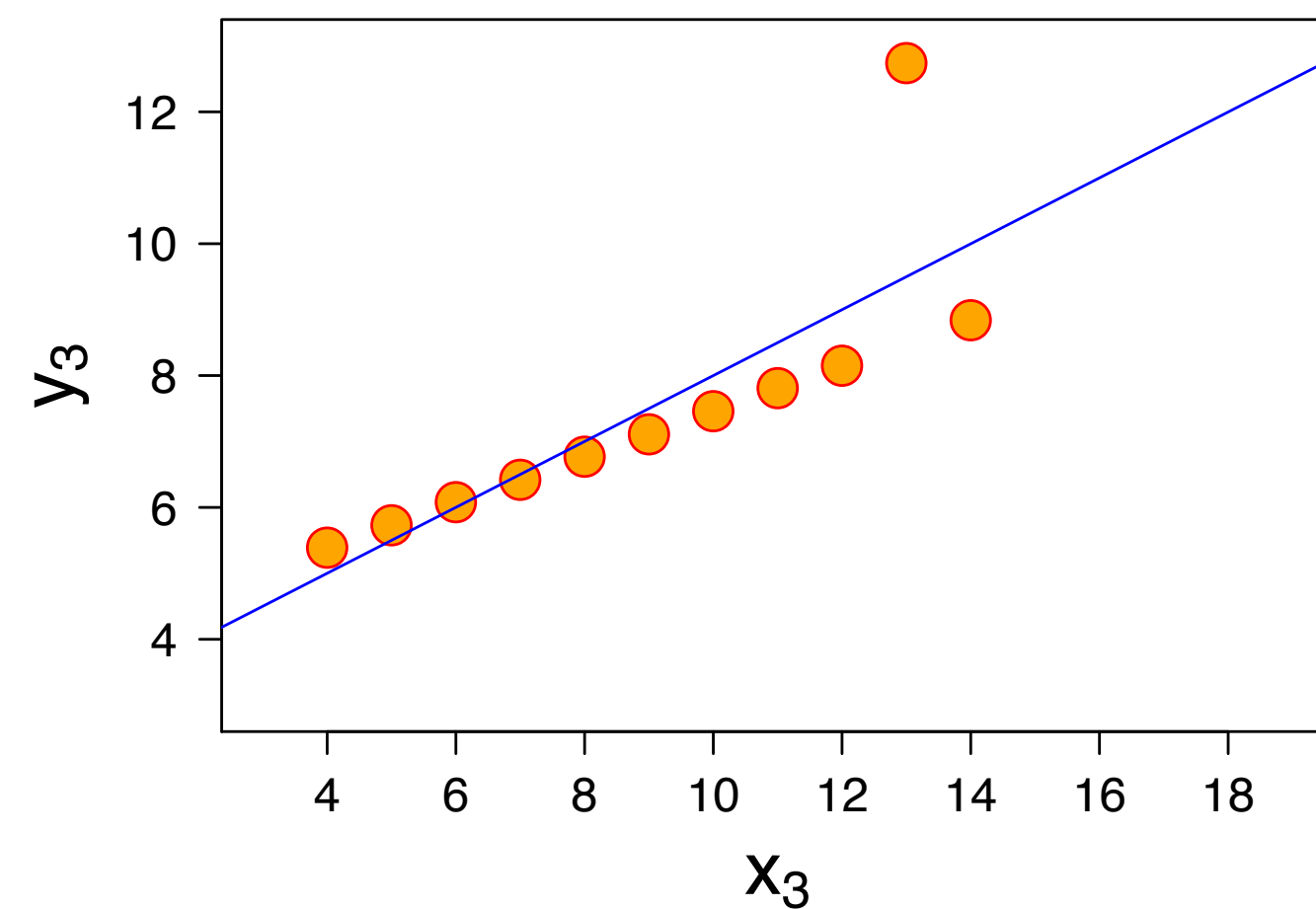
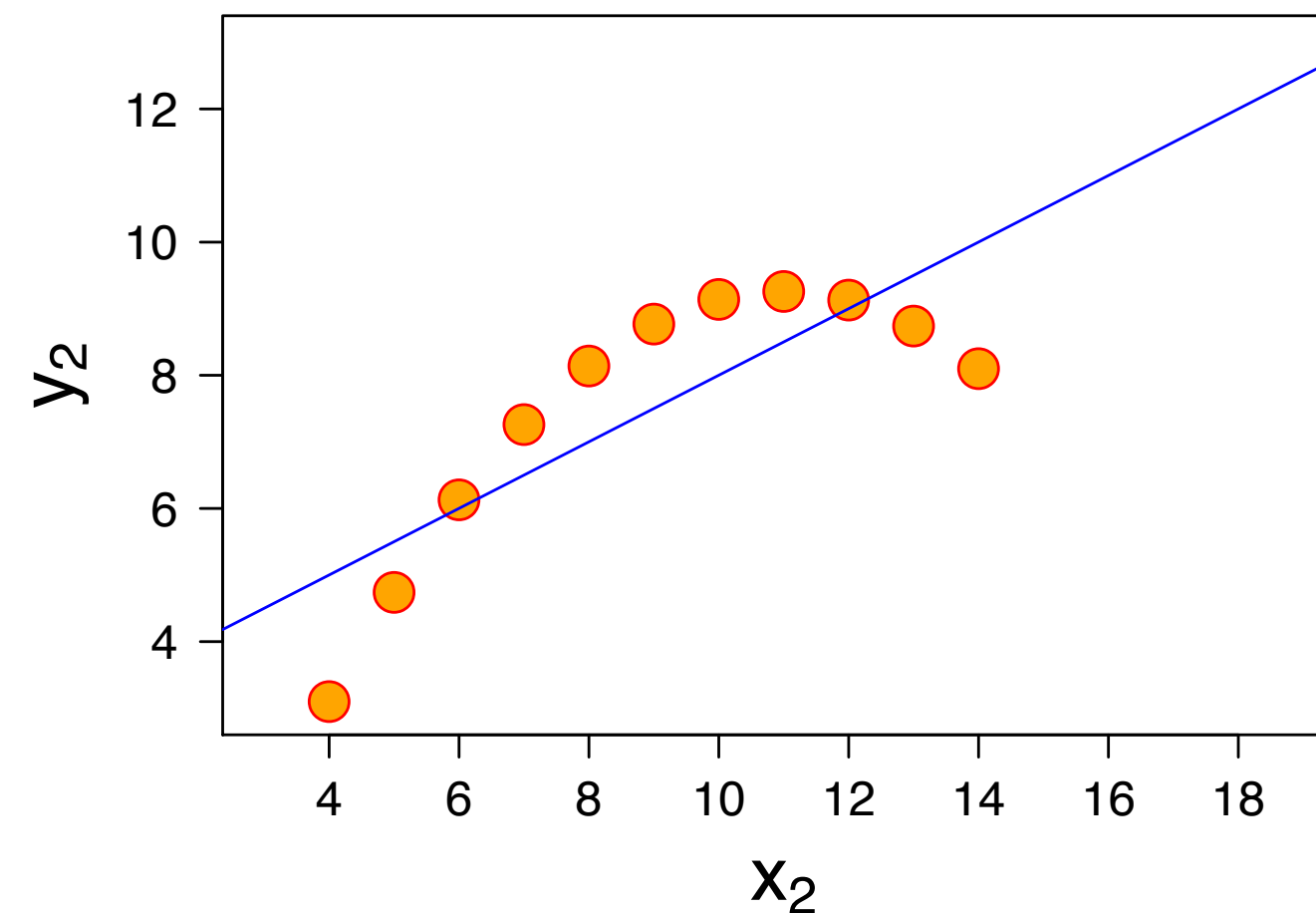
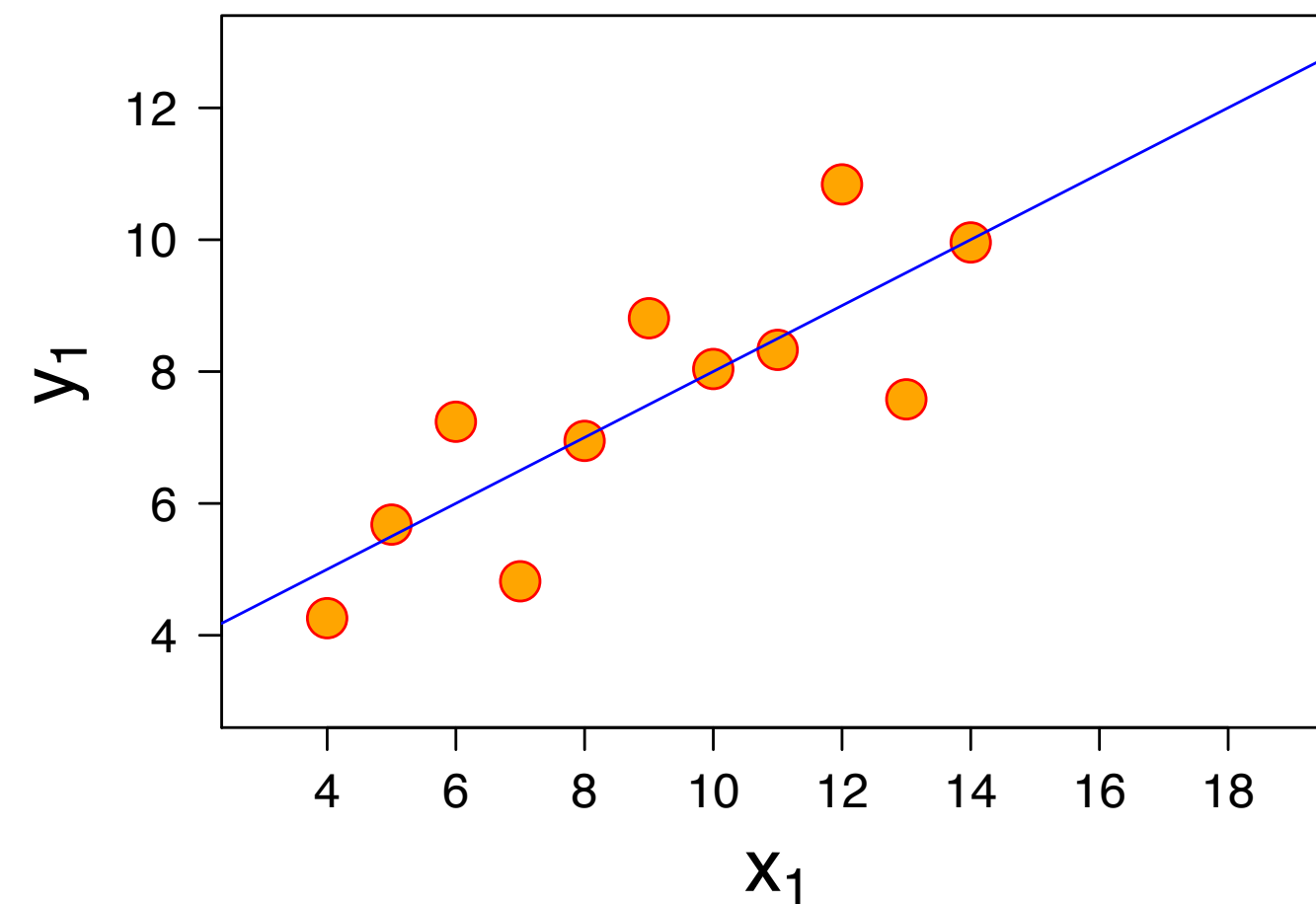
[T. Nørretranders]

Why Visual?



[F. J. Anscombe]

Why Visual?



Mean of x	9
Variance of x	11
Mean of y	7.50
Variance of y	4.122
Correlation	0.816

[F. J. Anscombe]

Visualization Goals

- "The purpose of visualization is **insight**, not pictures" – B. Schneiderman
- Identify patterns, trends
- Spot outliers
- Find similarities, correlation

Data is Encoded via Visual Channels

➔ Position

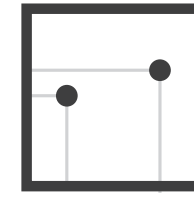
➔ Horizontal



➔ Vertical



➔ Both



➔ Color



➔ Shape



➔ Tilt

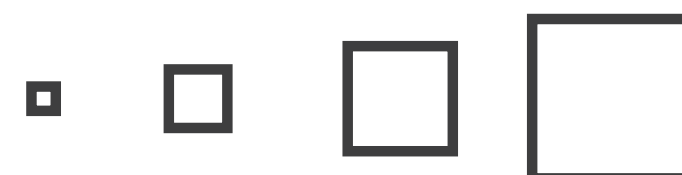


➔ Size

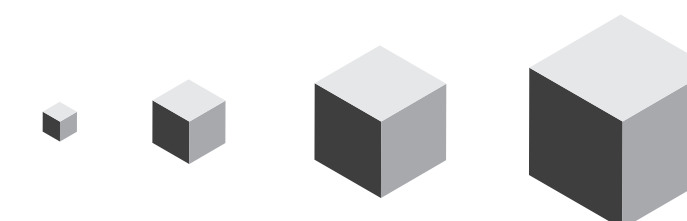
➔ Length



➔ Area



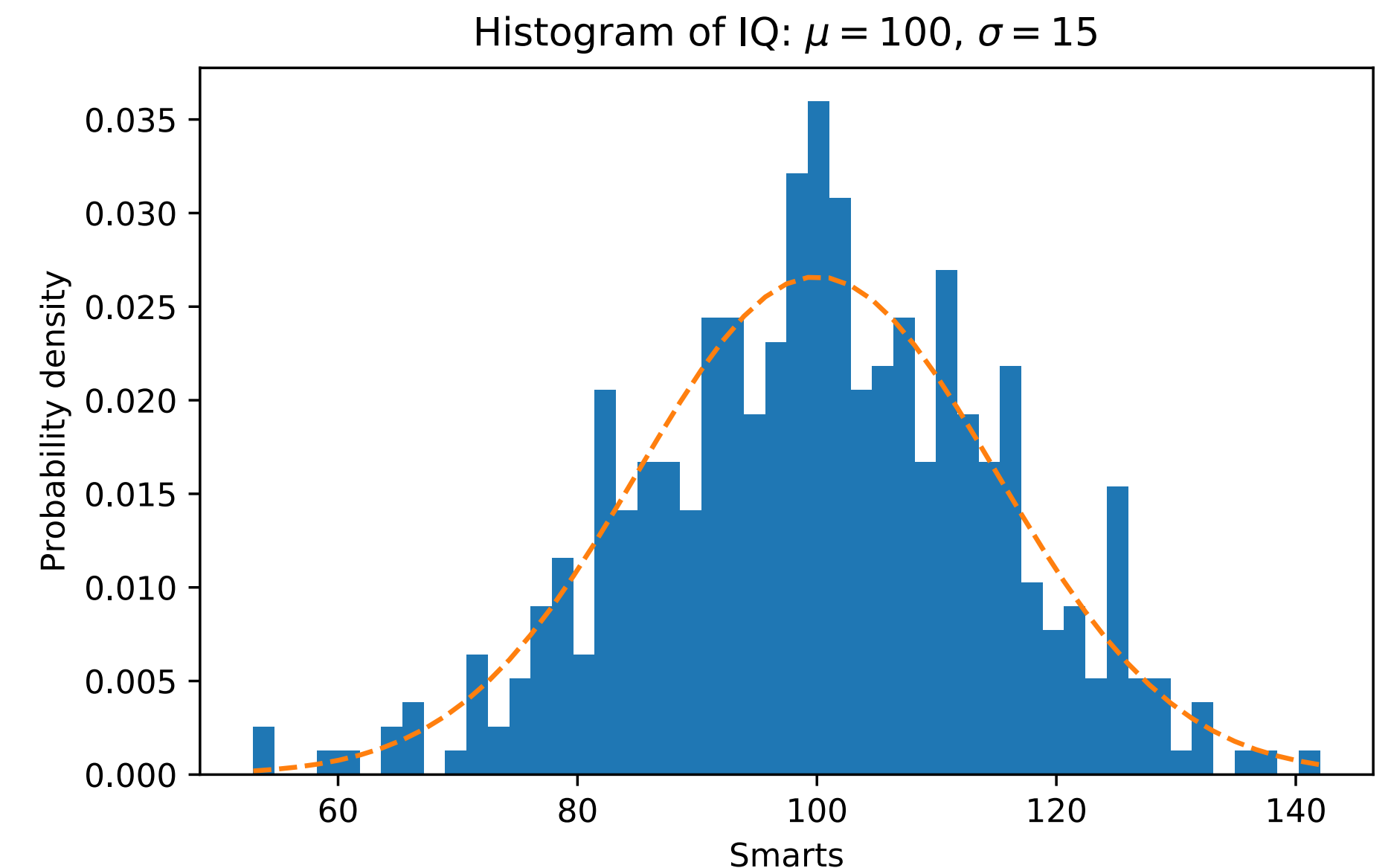
➔ Volume



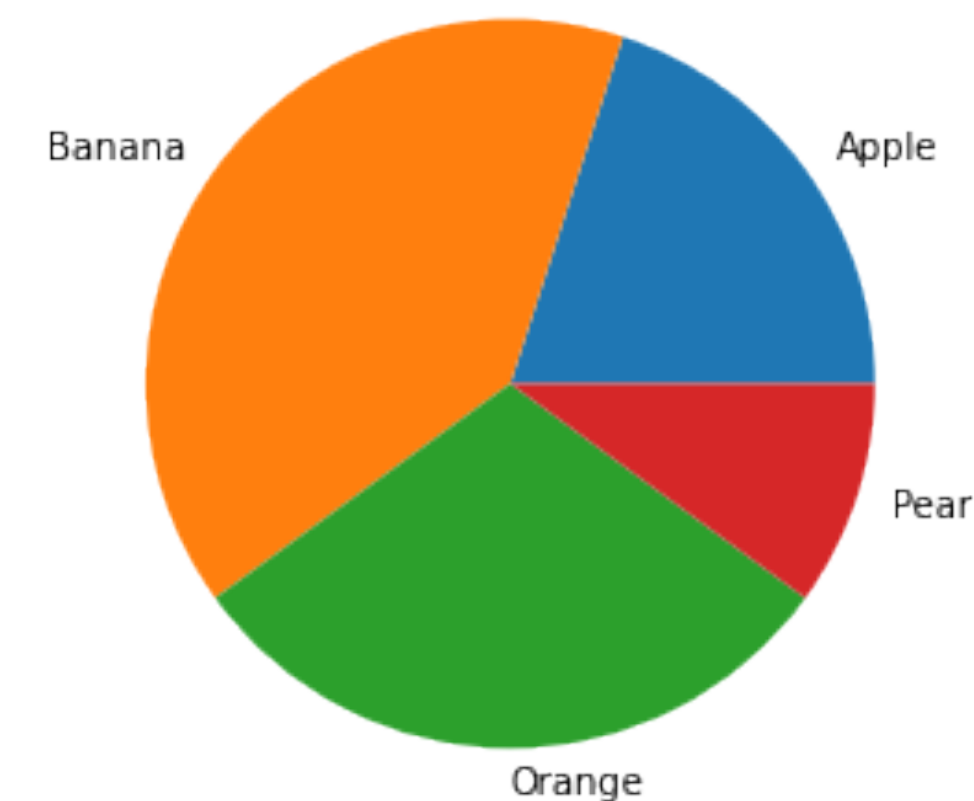
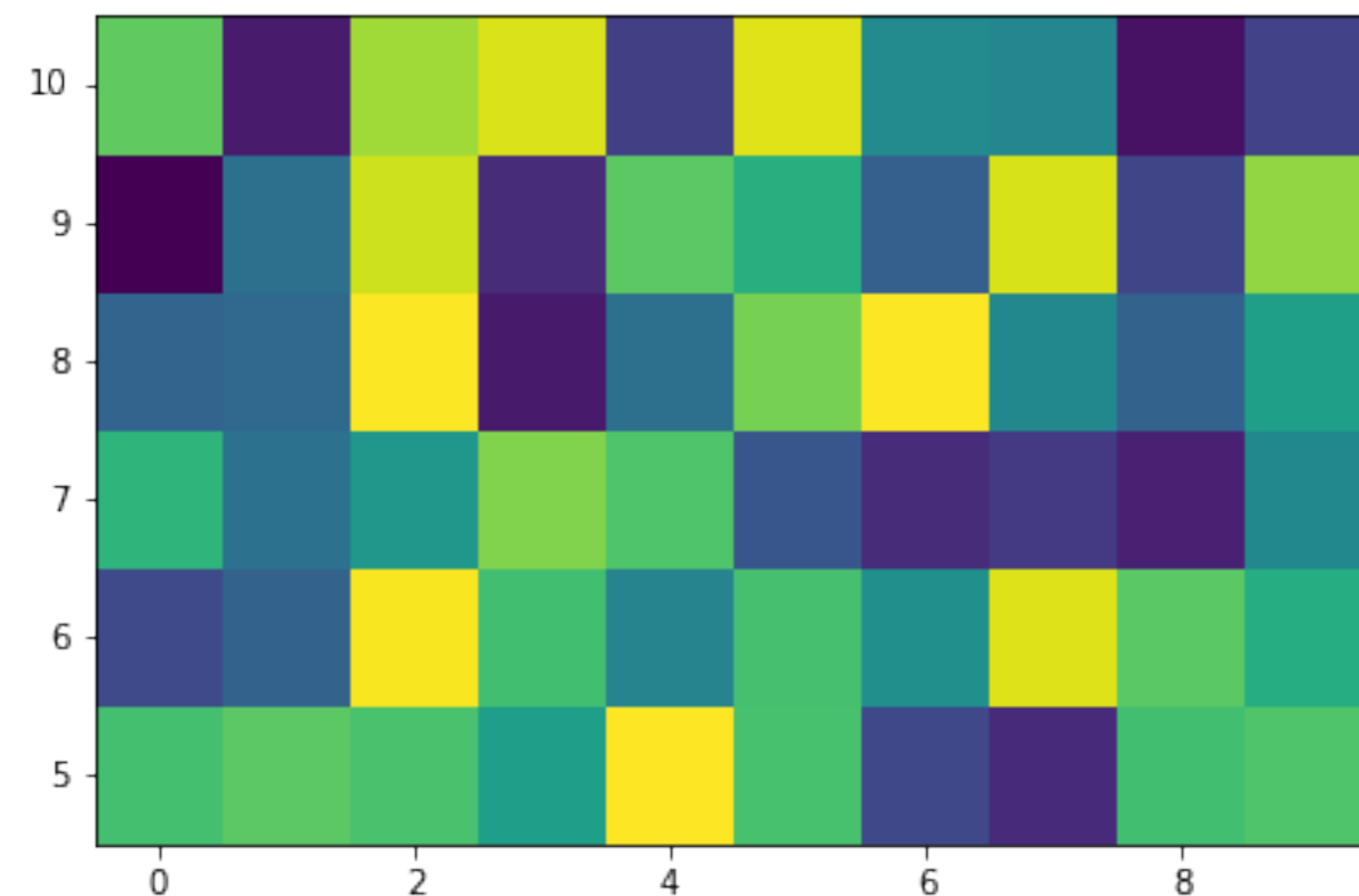
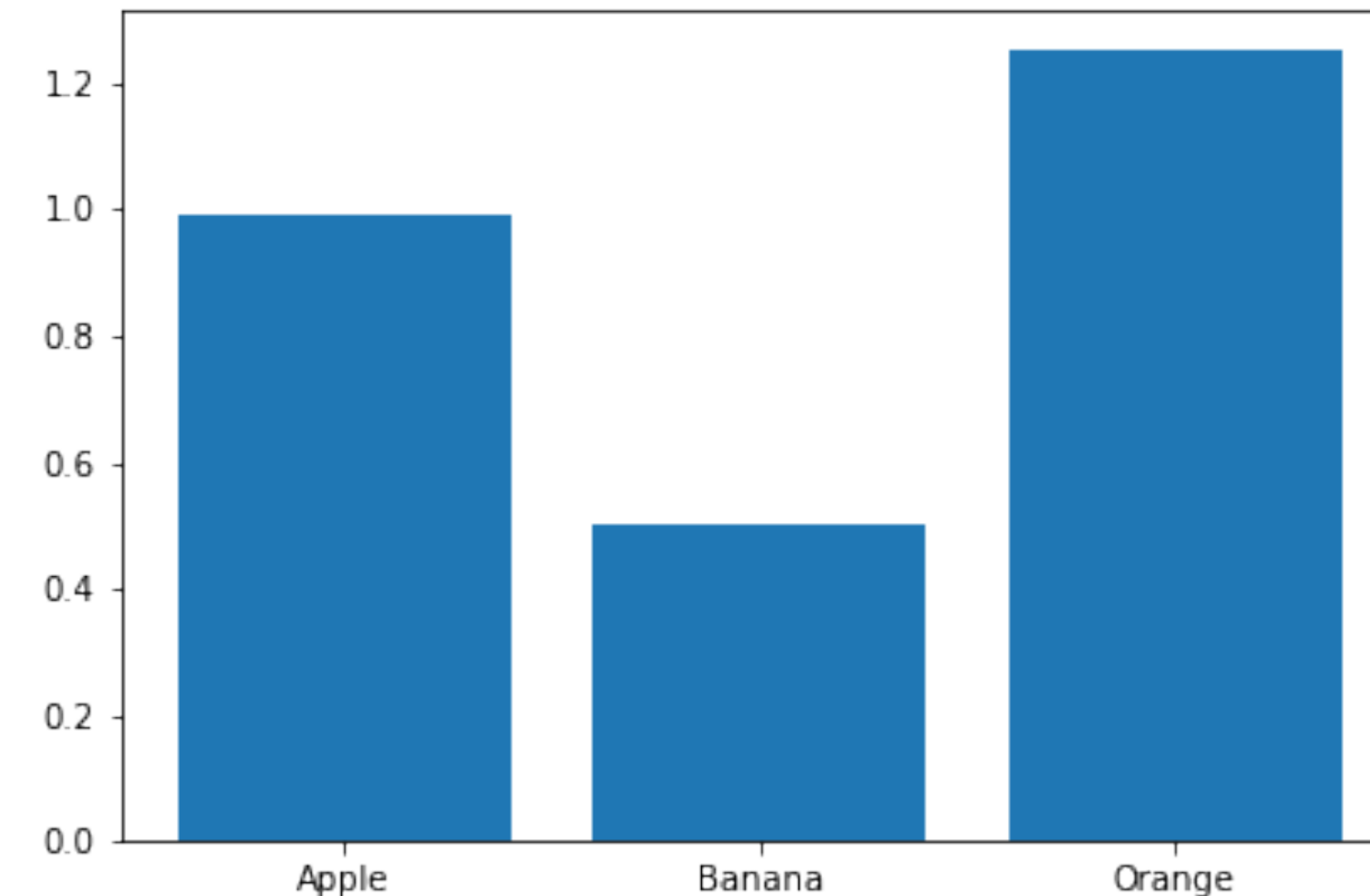
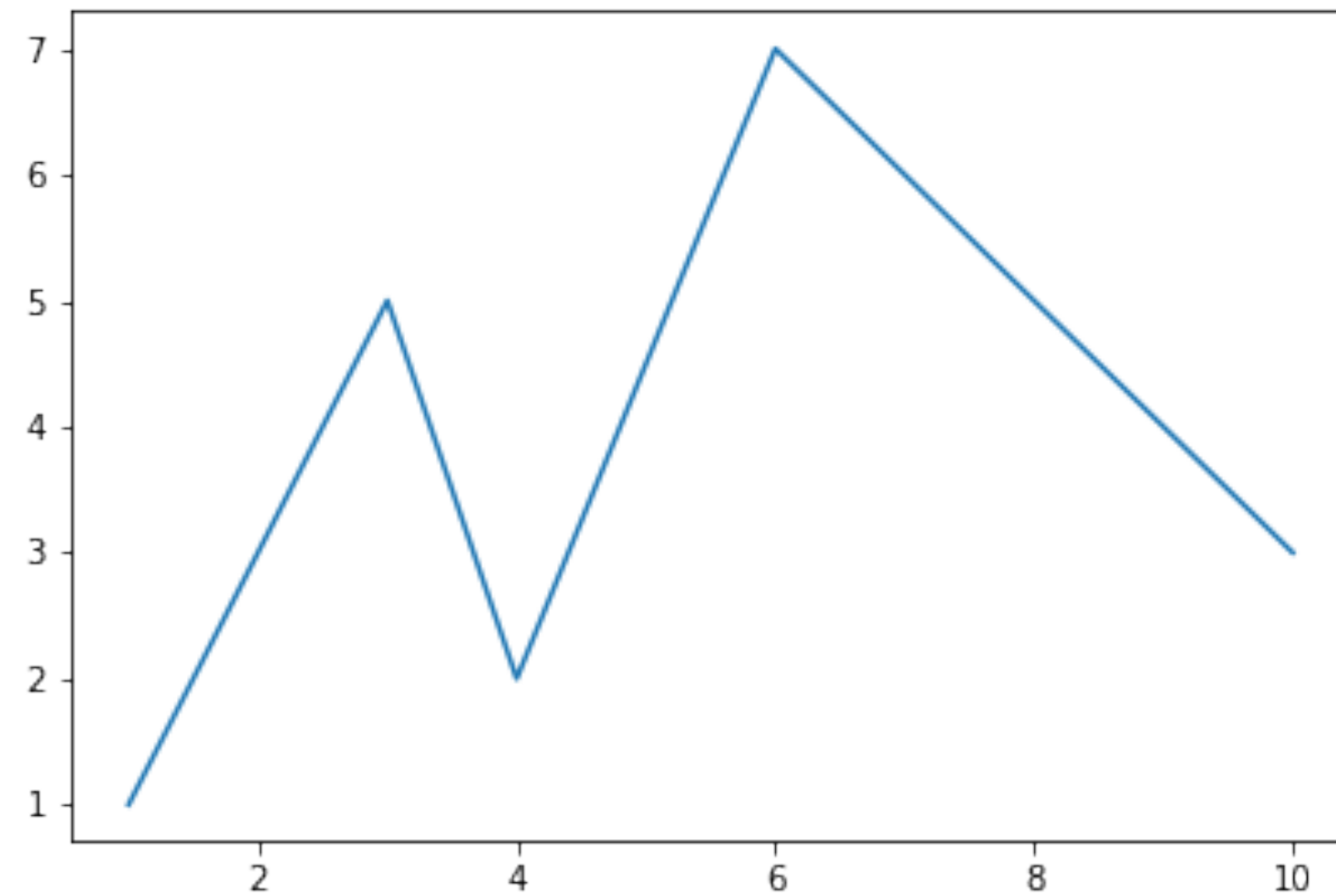
[Munzner (ill. Maguire), 2014]

matplotlib

- Strengths:
 - Designed like Matlab
 - Many rendering backends
 - Can reproduce almost any plot
 - Proven, well-tested
- Weaknesses:
 - API is imperative
 - Not originally designed for the web
 - Dated styles



Many different types of charts



Many different types of charts

- Bar chart

- `plt.bar(['Apple', 'Banana', 'Orange'], [0.99, 0.50, 1.25])`

- Grid Heatmap

- `plt.pcolormesh(x, y, Z)`

- Pie chart:

- `plt.pie([20, 40, 30, 10],
 labels=['Apple', 'Banana', 'Orange', 'Pear'])`

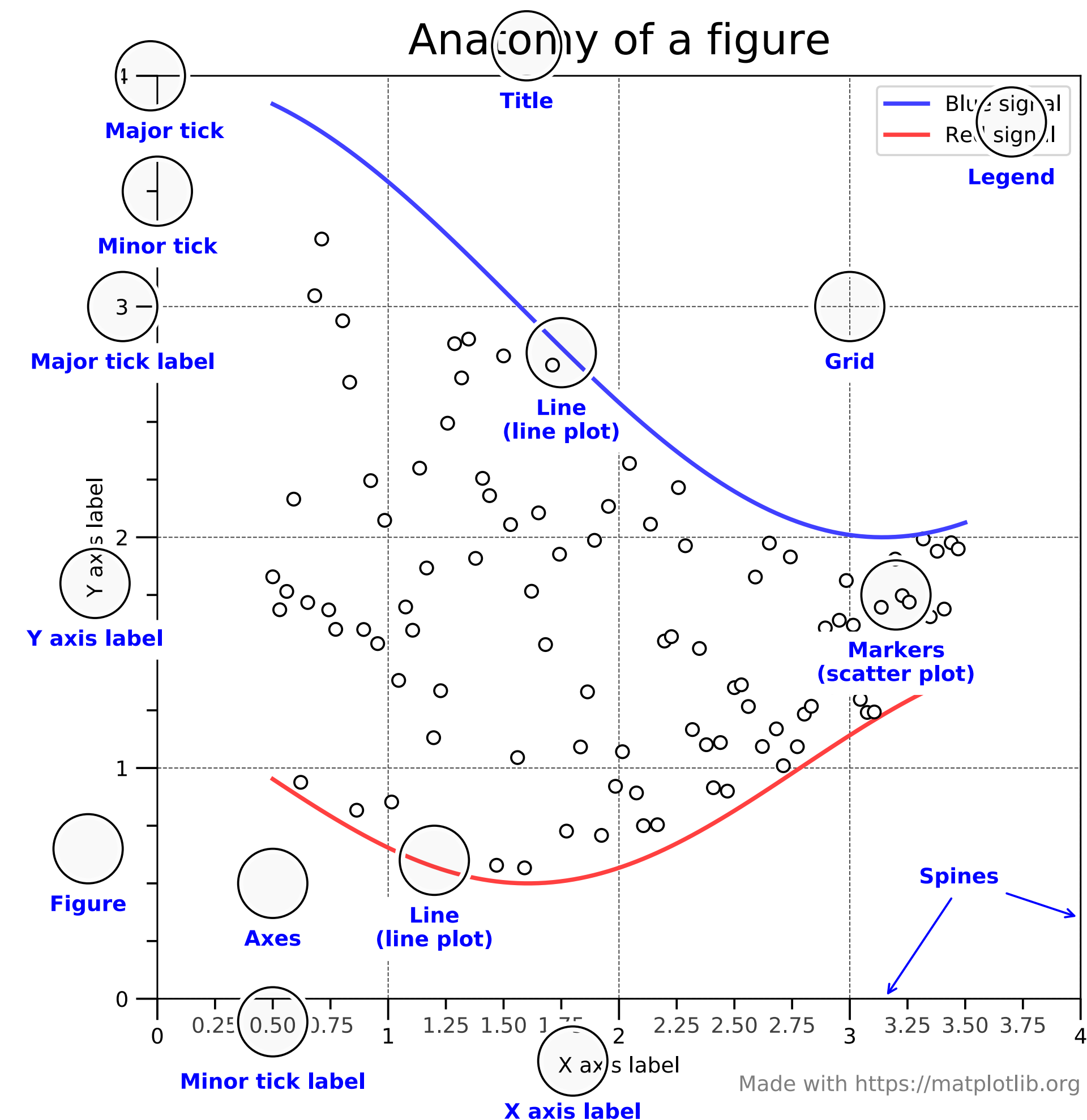
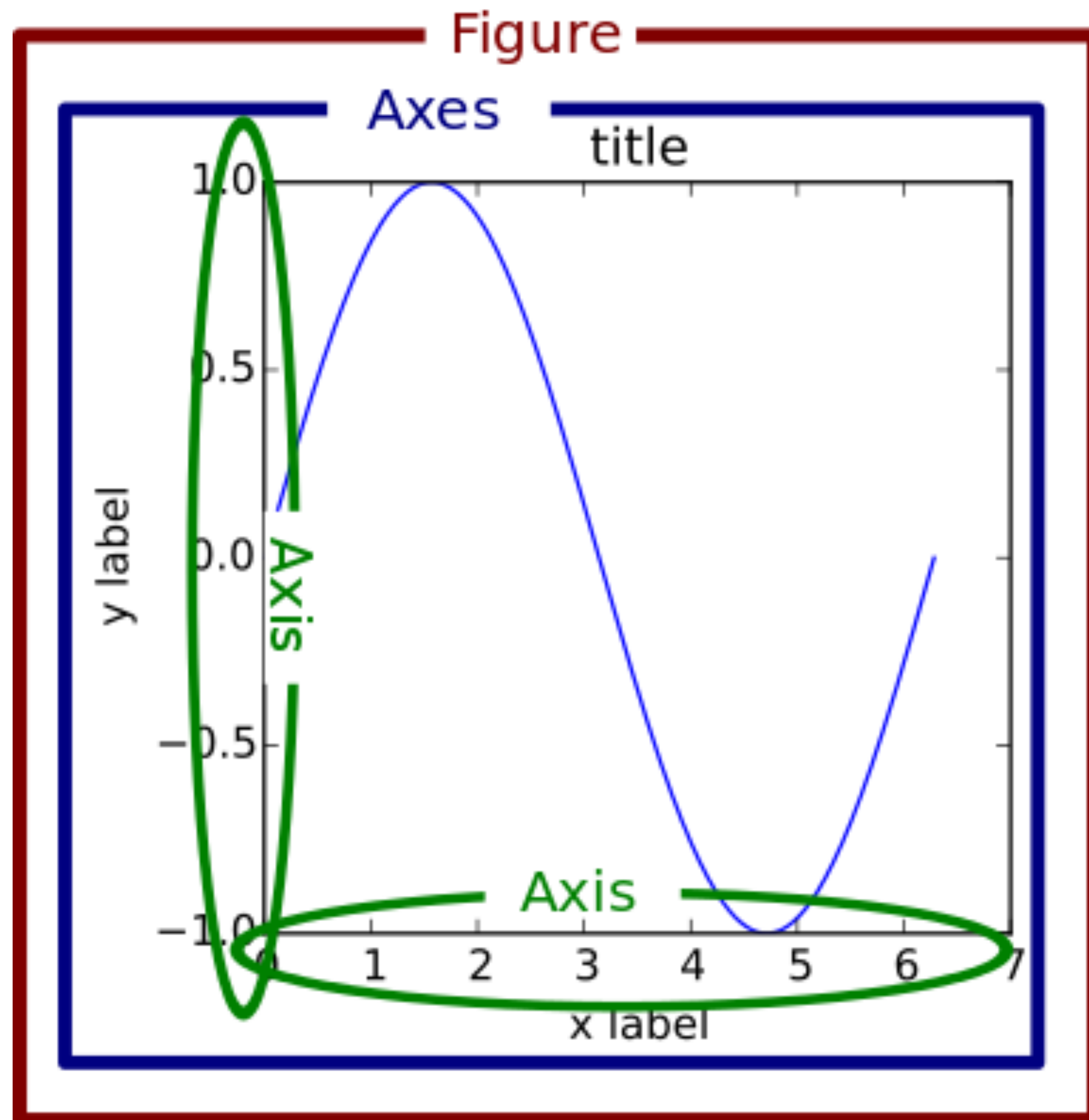
matplotlib tutorials

- <https://matplotlib.org/stable/tutorials/index.html>
- <https://github.com/rougier/matplotlib-tutorial>

Adding Labels

- `plt.xlabel`: set x label
- `plt.ylabel`: set y label
- `plt.title`: set title
- ```
plt.plot([1, 3, 4, 6, 10], [1, 5, 2, 7, 3])
plt.xlabel('Age')
plt.ylabel('Number of Jumps')
plt.title('Kangaroo Jumps Today')
```

# Anatomy of a Figure



[B. Solomon & matplotlib]

# Figure and Axes Objects

---

- pyplot is stateful, functions affect the "current" figure and axes
  - `plt.gcf()`: gets current figure
  - `plt.gca()`: gets current axes
  - Creates one if it doesn't exist!
- This is not aligned with object-based programming ideas
- Most methods in pyplot are translated to methods on the current axes (gca)
- We can instead call these directly, but first need to create them:
  - `fig, ax = plt.subplots()` # "constructor-like" method  
`ax.scatter([1, 3, 4, 6, 10], [1, 5, 2, 7, 3])`

# Object-Based Plotting

---

- `fig, ax = plt.subplots()` # "constructor-like" method  
`ax.scatter([1,3,4,6,10],[1,5,2,7,3])`
- Use getters/setters for labels and title
  - `ax.set_xlabel('Age')`
  - `ax.set_ylabel('Number of Jumps')`
  - `ax.set_title('Kangaroo Jumps Today')`
- We can also call methods on the figure:
  - `fig.tight_layout()` # reduce margins

# Multiple Figures

---

- subplots allows multiple axes in the same figure:
  - `fig, ax = plt.subplots(2, 2, figsize=(10, 10))` # rows, then columns
- `ax` is now a 2x2 numpy array
- Can put any type of visualization on each pair of axes
- `ax[0,0].plot([1,3,4,6,10],[1,5,2,7,3])`  
`ax[0,1].bar(['Apple','Banana','Orange'],[0.99,0.50,1.25])`  
`ax[1,0].pcolormesh(x, y, Z)`  
`ax[1,1].pie([20,40,30,10],`  
`labels=['Apple','Banana','Orange','Pear'])`

# pandas Integration

---

- Can call many of these methods directly from pandas
- Handled through `kind` kwarg or `.plot` accessor
- It will try to guess a reasonable visualization, but may fail:
  - `fruit.plot()`
- Instead, specify `x` and `y` and other parameters:
  - `fruit.plot(kind='bar', x='name', y='price')`
  - `plt.bar(x='name', height='price', data=fruit)` # SIMILAR
  - `fruit.plot.scatter(x='price', y='count', c='name')` # ERROR
  - ```
colors = {'Apple': 'red', 'Orange': 'orange',  
          'Banana': 'yellow', 'Pear': 'green'}  
fruit.plot.scatter(x='price', y='count',  
                  c=fruit['name'].map(colors))
```

Assignment 8

- Due Friday, Dec. 7
- Last Assignment
- Data and Visualization
- Same Illinois Unemployment Data
- Use polars or pandas
- Must use matplotlib or altair where directed

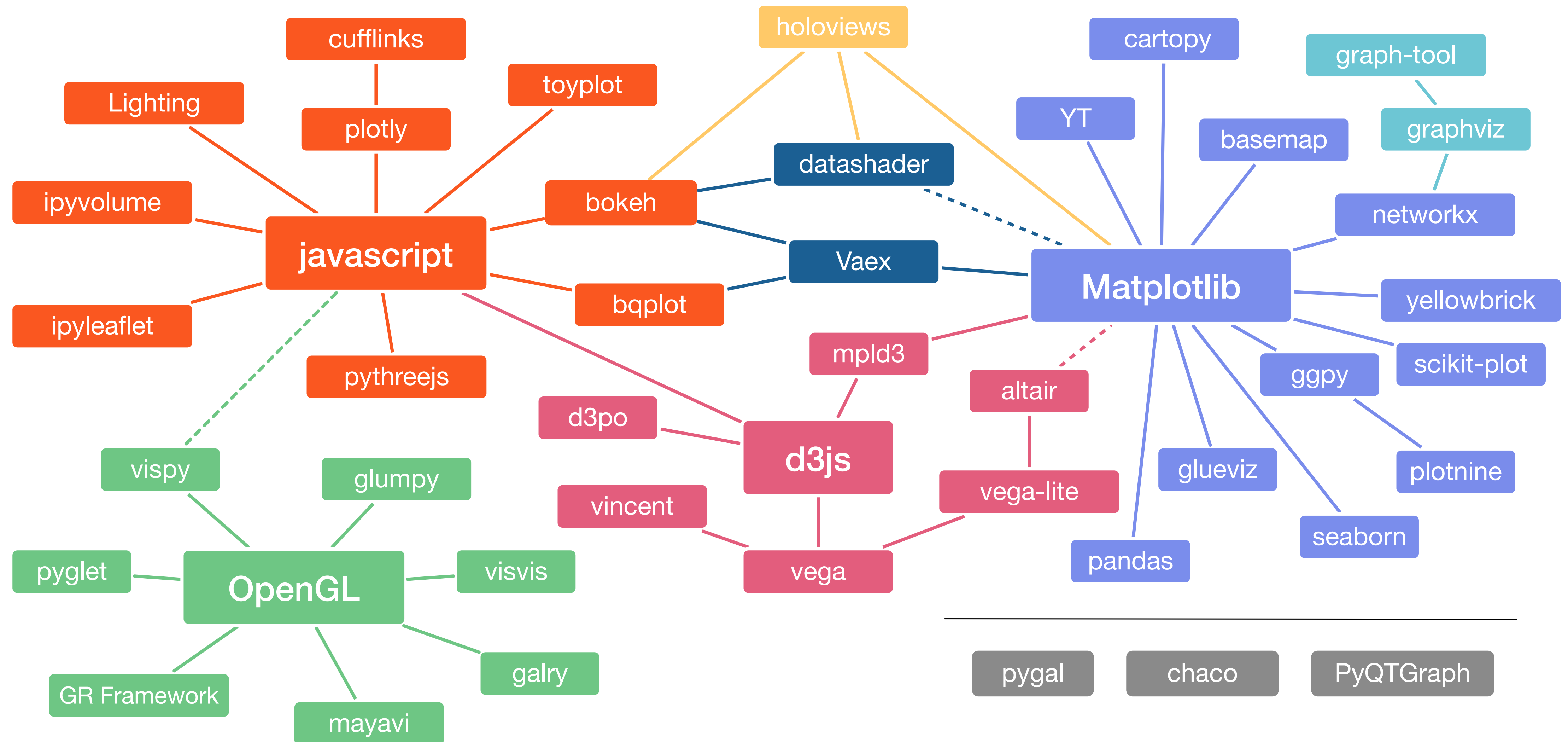
Final Exam

- Wednesday, Dec. 11, **8:00**-9:50am in PM 103
- **More comprehensive** than Test 2
- Expect questions from topics covered on Test 1 and 2
- Expect questions from the last few weeks of class (data, visualization, machine learning)
- Similar format
-

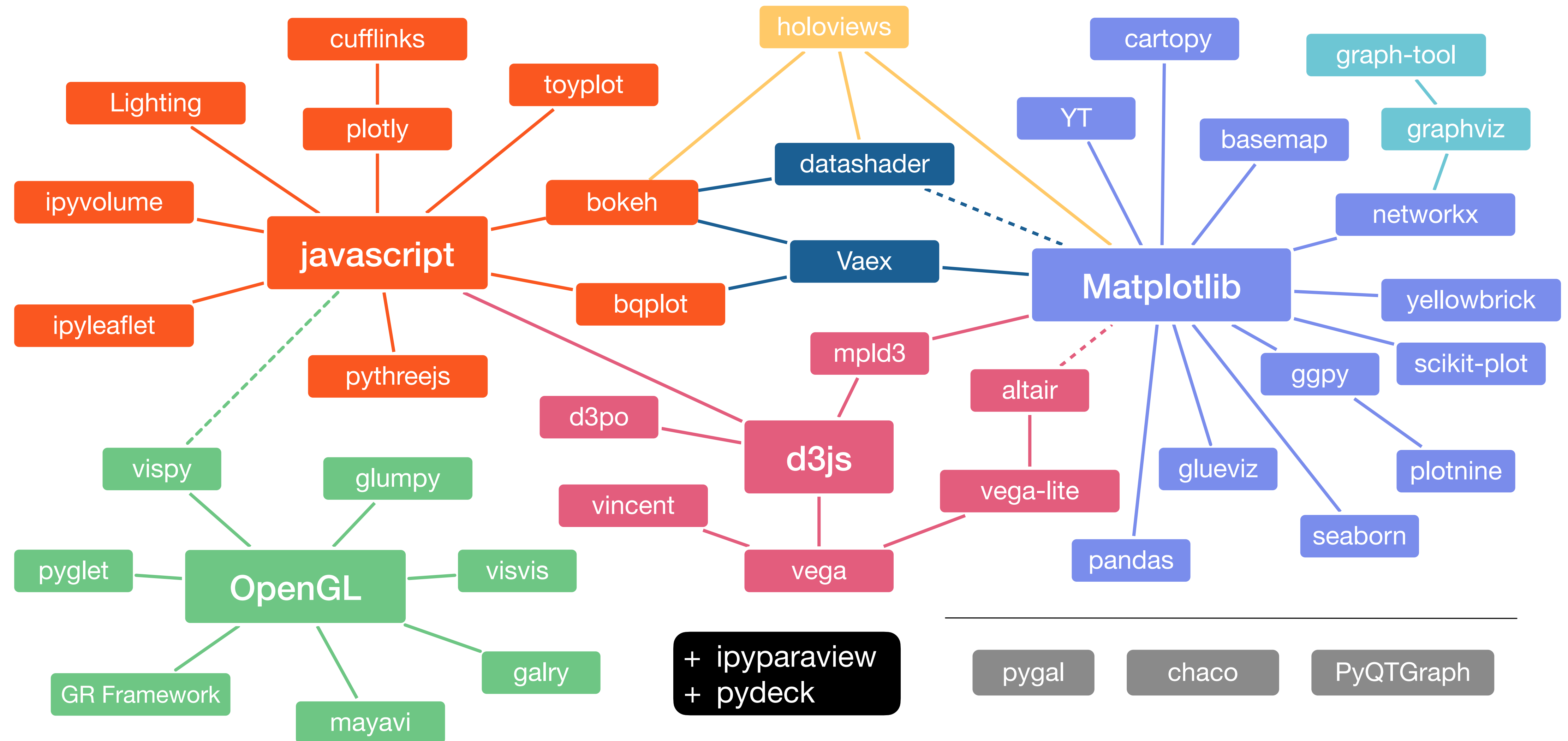
Wednesday

- Wrap up Machine Learning
- Review
 - Bring your questions!
 - Your last chance to ask questions before the final
- Also, teaching evaluations

The Python Visualization Landscape



The Python Visualization Landscape



History of Vega-Lite & Altair

- "Grammar of Graphics", L. Wilkinson
- "A Layered Grammar of Graphics", H. Wickham
- ggplot: plotting library for R
- Vega: similar idea for Javascript/JSON (U. Washington, A. Satyanarayan)
 - "Declarative language for creating, saving, and sharing interactive visualization designs"
 - More focus on interaction and reactive signals
 - Separation between specification and runtime
- Vega-Lite: higher-level language than Vega (U. Washington, D. Moritz)
 - uses carefully designed rules to default settings

History of Vega-Lite & Altair

- Altair: Python interface to Vega-Lite (J. VanderPlas)
 - "spend more time understanding your data and its meaning"
 - Specify the what, minimize the amount of code directing the how
 - Python can write JSON specification just as well as any other language
 - Bindings make it more Python-friendly, integrate with pandas, add support for Jupyter, etc.
- Vega Fusion (J. Mease)
 - Scaling to larger datasets
 - Serverside scaling

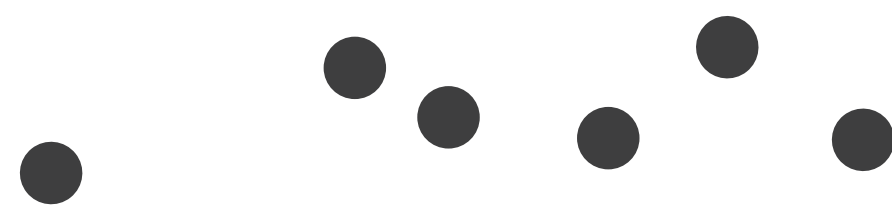
Basic Example

- ```
import altair as alt
import pandas as pd
data = pd.DataFrame({'x': [1, 3, 4, 6, 10], 'y': [1, 5, 2, 7, 3]})
alt.Chart(data).mark_line().encode(x='x', y='y')
```
- Easiest to use data from a pandas data frame
  - Another option is a csv or json file
  - Can support geo\_interface, too
- `Chart` is the basic unit
- Mark: `.mark_*()` indicates the geometry created for each data item
- Encode: `.encode()` allows visual properties to be set to data attributes

# Visual Marks

- **Marks** are the basic graphical elements in a visualization
- Marks classified by dimensionality:

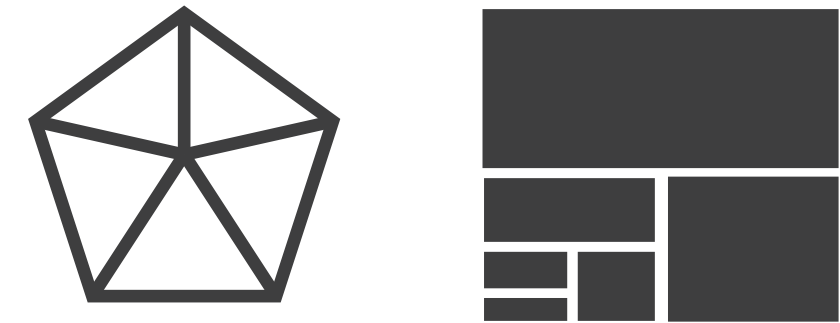
➔ **Points**



➔ **Lines**



➔ **Areas**



- Also can have surfaces, volumes
- Think of marks as a mathematical definition, or if familiar with tools like Adobe Illustrator or Inkscape, the path & point definitions
- Altair: area, bar, circle, geoshape, image, line, point, rect, rule, square, text, tick
  - Also compound marks: boxplot, errorband, errorbar

# Encode via Visual Channels

## ➔ Position

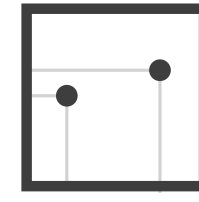
➔ Horizontal



➔ Vertical



➔ Both



## ➔ Color



## ➔ Shape



## ➔ Tilt

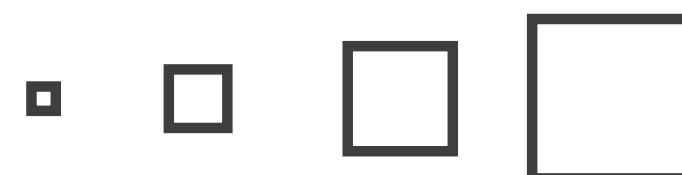


## ➔ Size

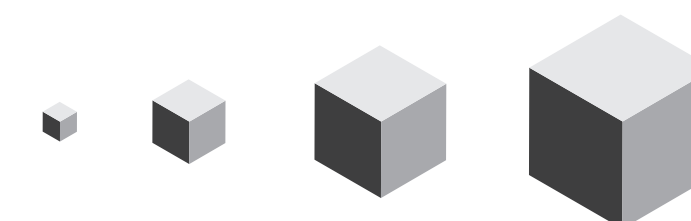
➔ Length



➔ Area



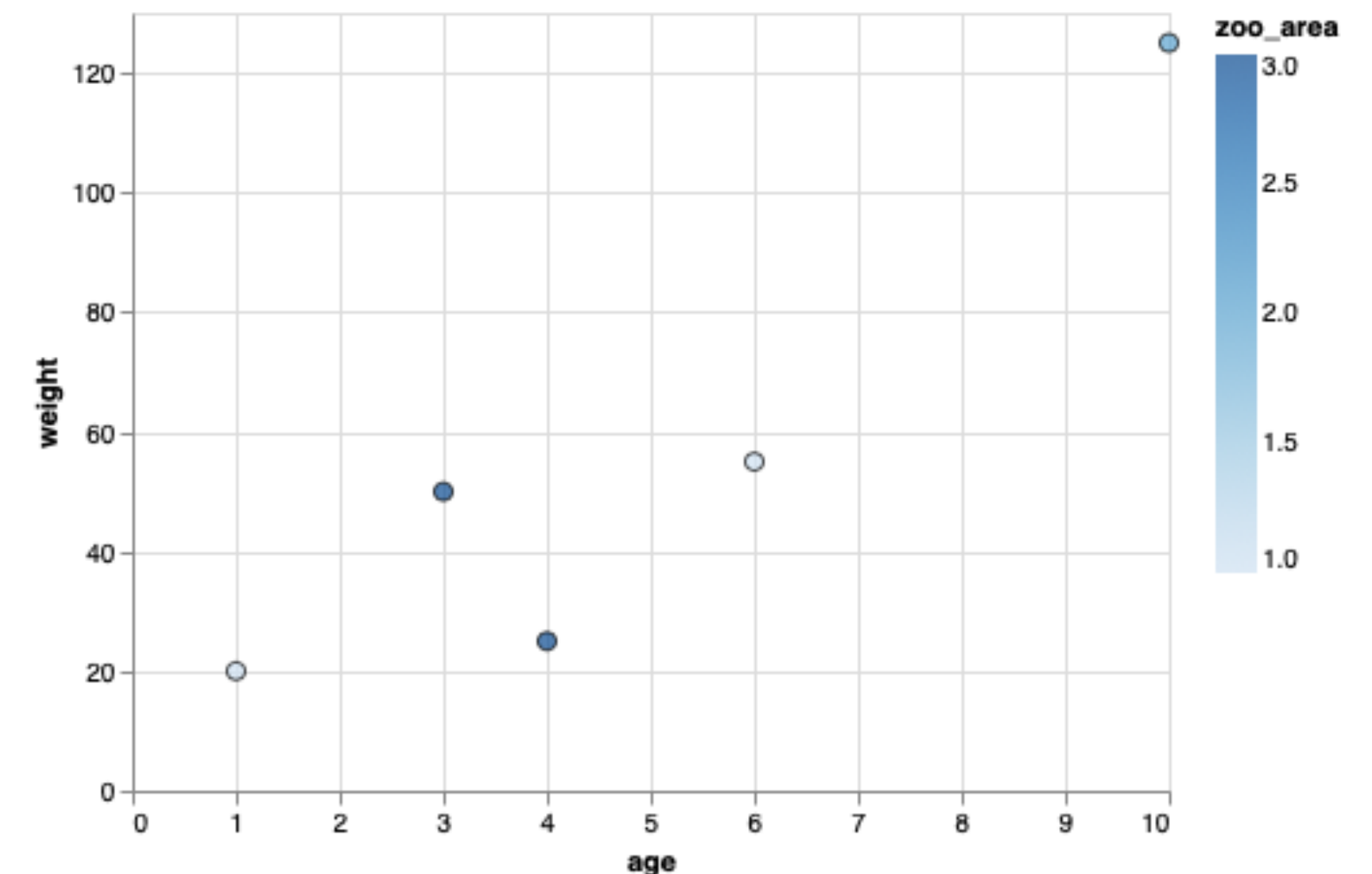
➔ Volume



[Munzner (ill. Maguire), 2014]

# Easily Explore Different Encodings

```
• data = pd.DataFrame({
 'age': [1, 3, 4, 6, 10],
 'weight': [20, 50, 25, 55, 125],
 'zoo_area': [1, 3, 3, 1, 2],
 'num_scoops': [3, 2, 4, 2, 3]
})
alt.Chart(data).mark_point(
 filled=True, size=50,
 stroke='black', strokeWidth=1
) .encode(
 x='age',
 y='weight',
 color='zoo_area'
)
```



Problem: zoo\_area is not a continuous value,  
nor is it ordered in any way!

# Data Attributes and Altair Types

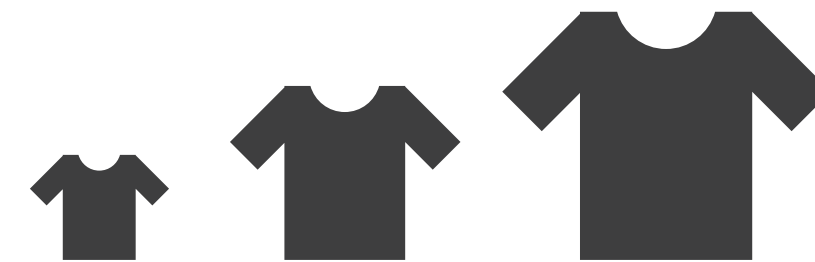
---

→ Categorical

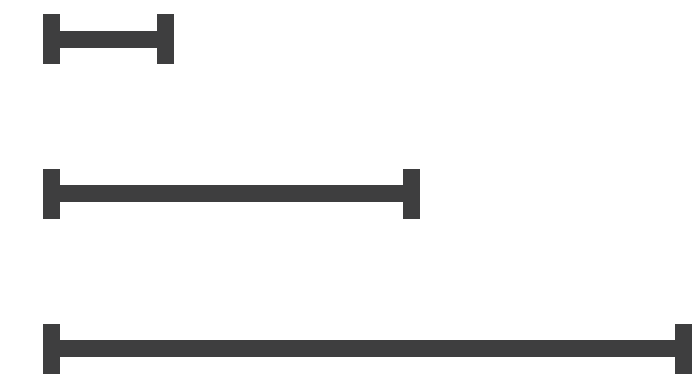


→ Ordered

→ *Ordinal*



→ *Quantitative*



# Data Attributes and Altair Types

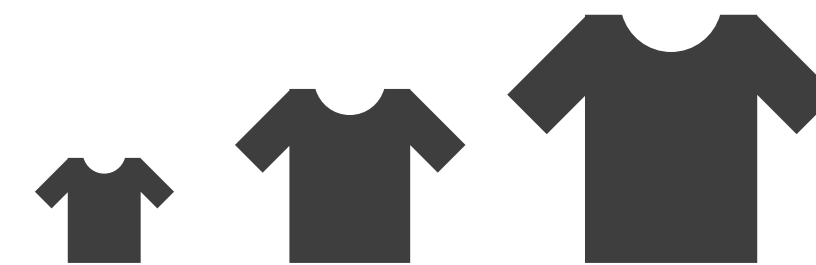
---

→ Categorical

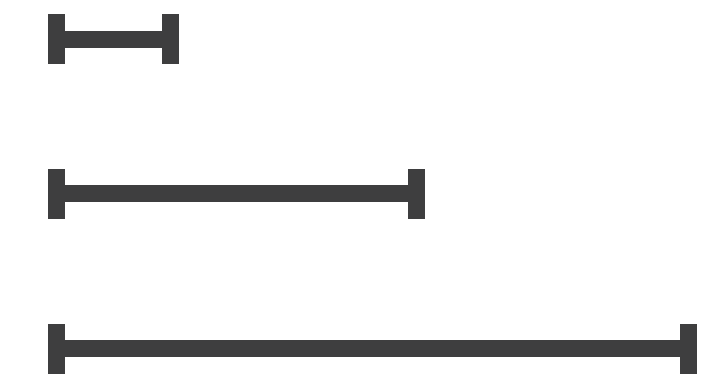


→ Ordered

→ *Ordinal*



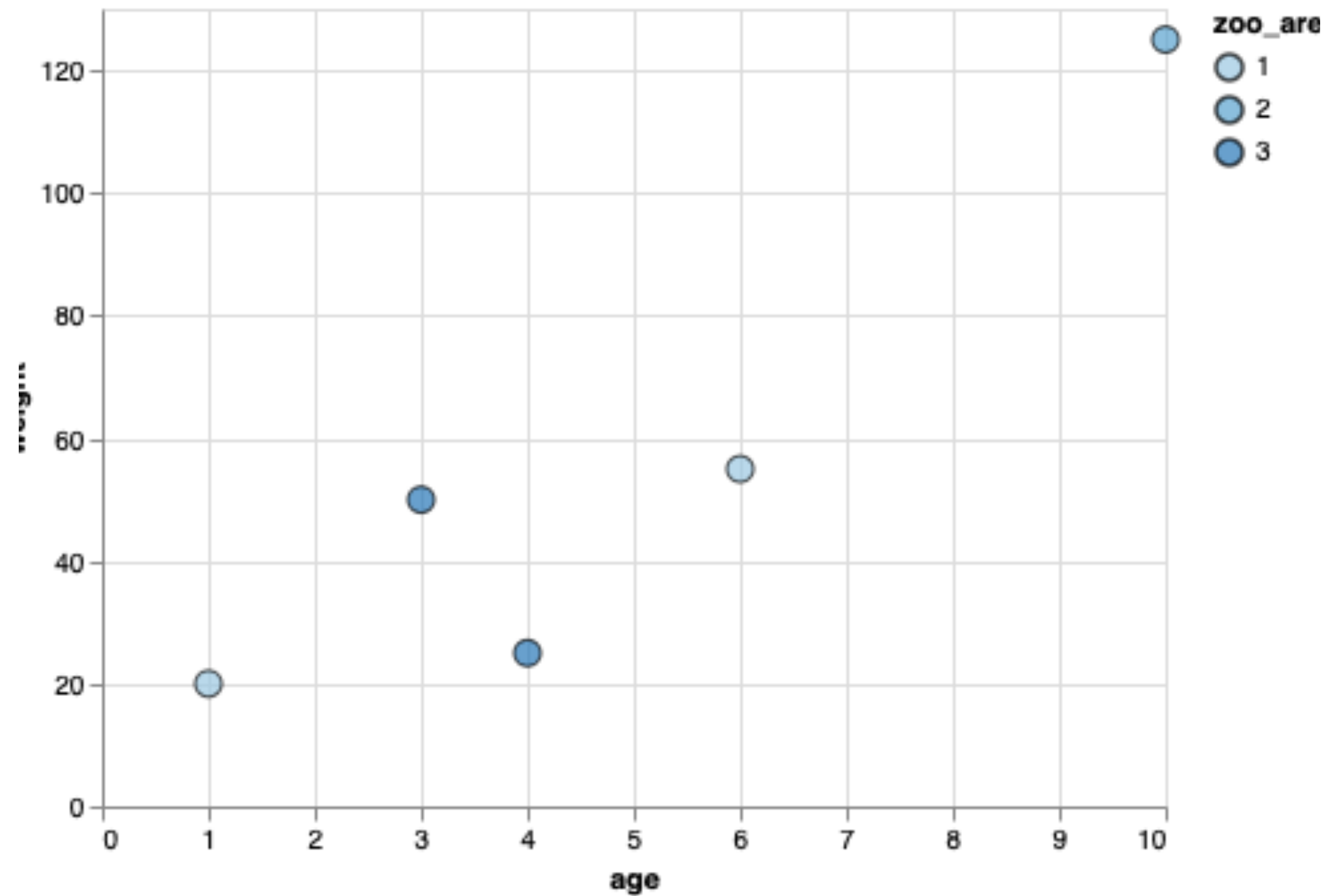
→ *Quantitative*



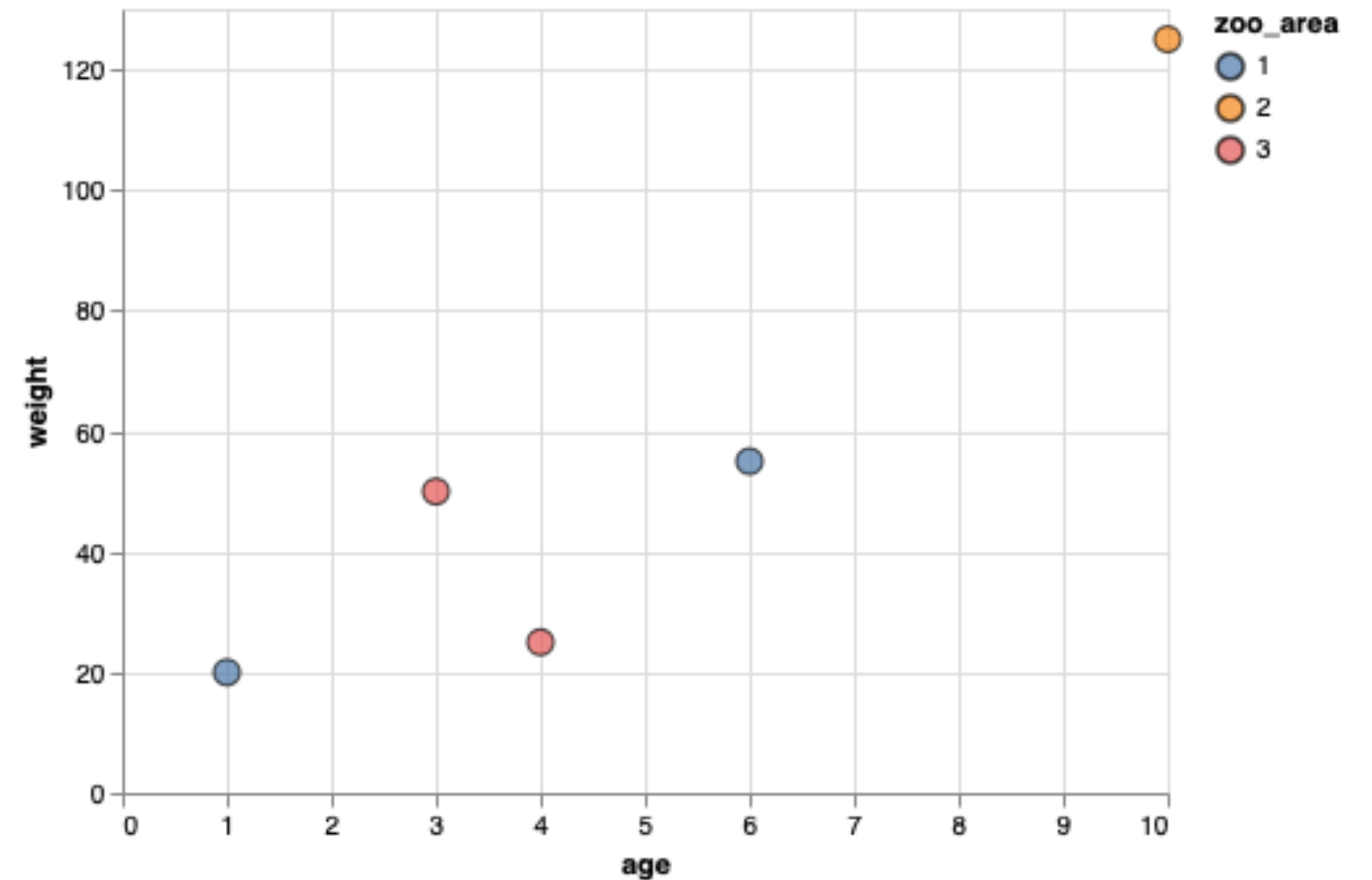
- Categorical data = Nominal (N)
- Ordinal data = Ordinal (O)
- Quantitative data = Quantitative (Q)
- Temporal data = Temporal (T)

[Munzner (ill. Maguire), 2014]

# Specifying the Type



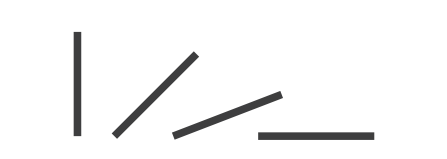
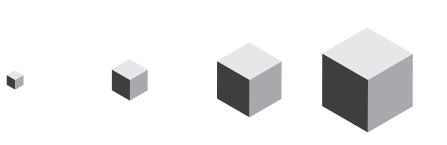
zoo\_area:0



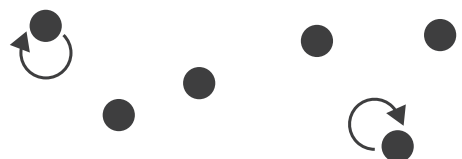
zoo\_area:N

# Different Channels for Different Attribute Types

➔ **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)            |  |

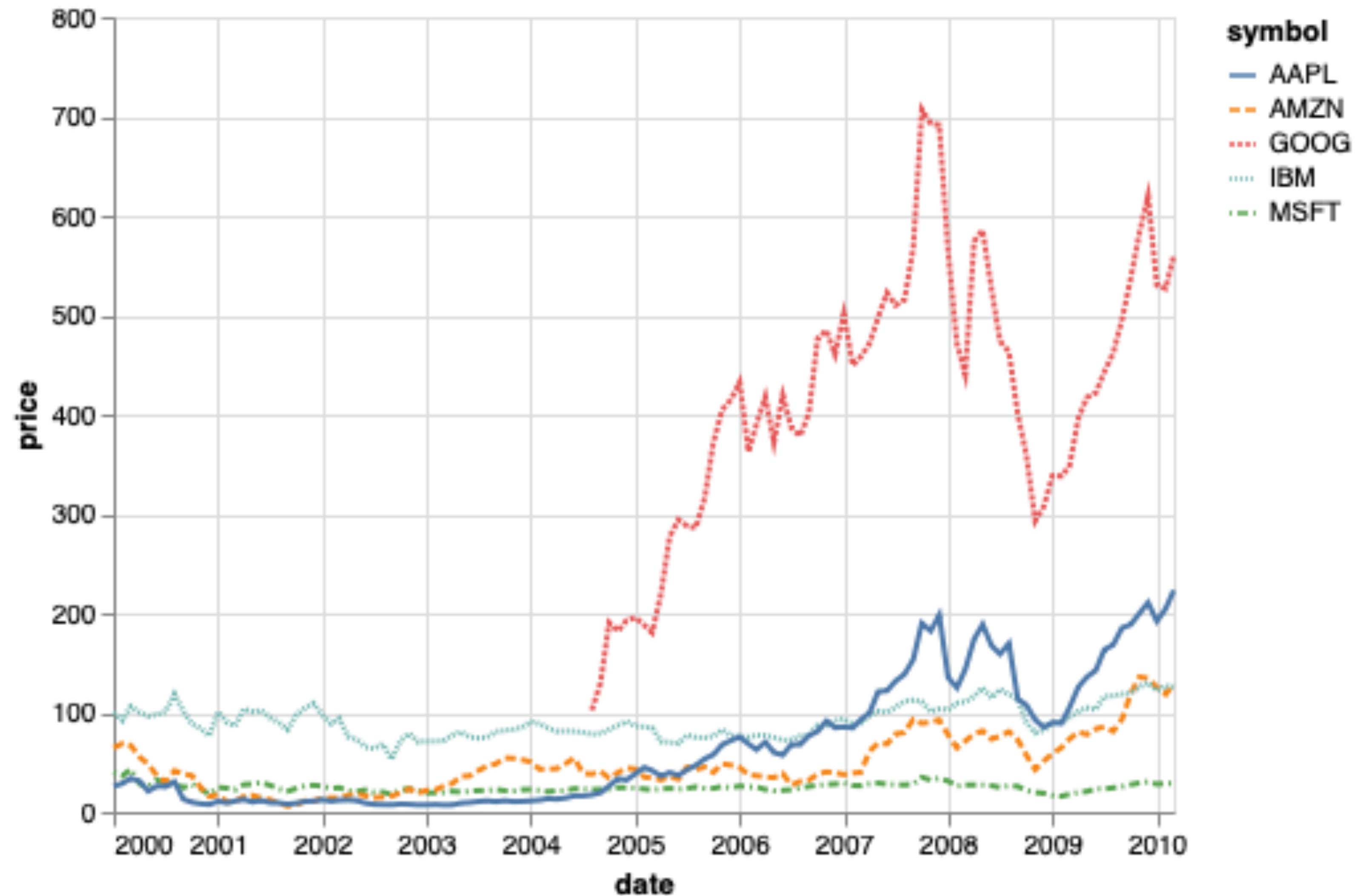
➔ **Identity** Channels: **Categorical** Attributes

|                |                                                                                     |
|----------------|-------------------------------------------------------------------------------------|
| Spatial region |  |
| Color hue      |  |
| Motion         |  |
| Shape          |  |

Altair will use its rules to pick whether to use color hue or saturation based on the type

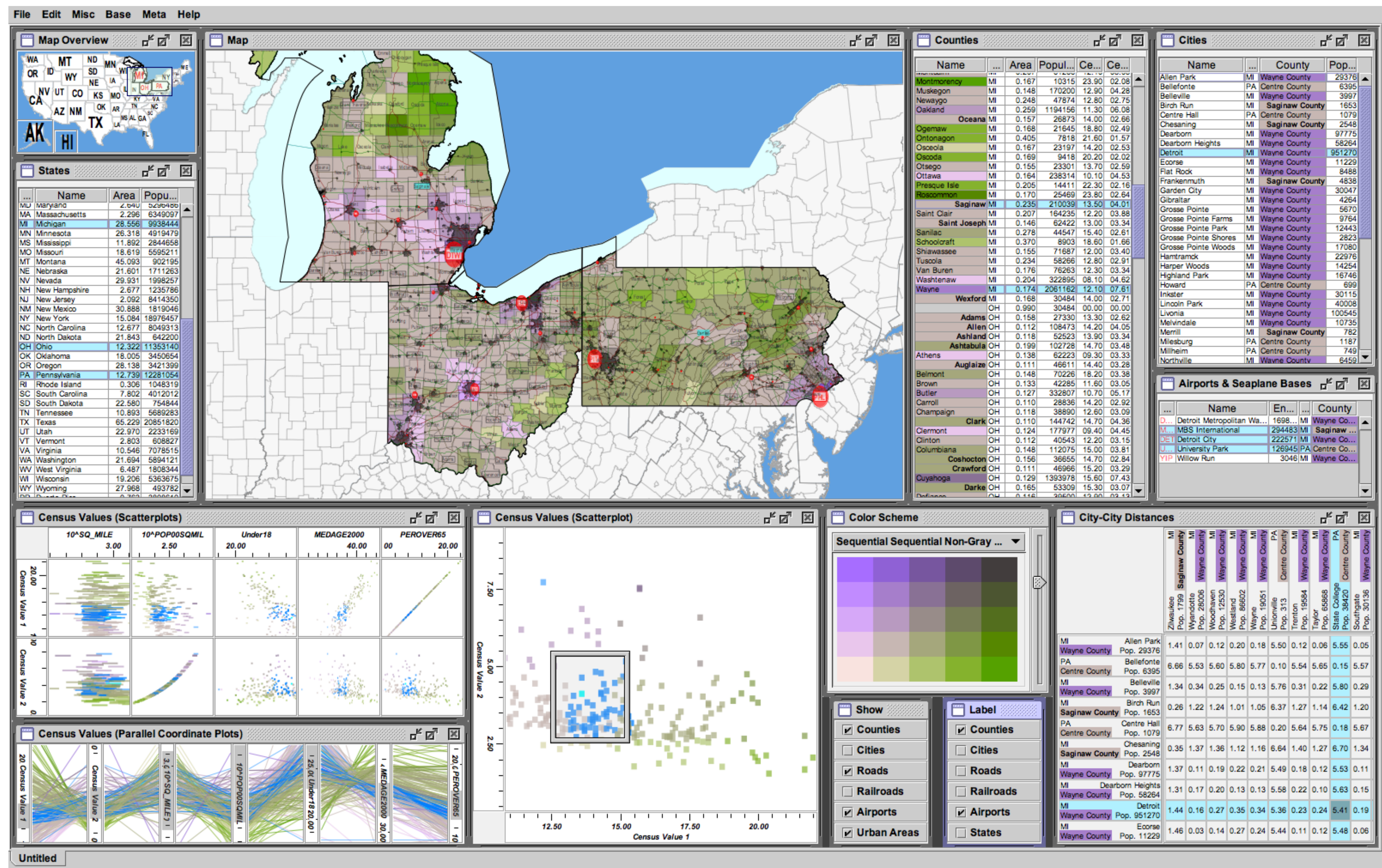
[Munzner (ill. Maguire), 2014]

# Multiple Views in Visualization



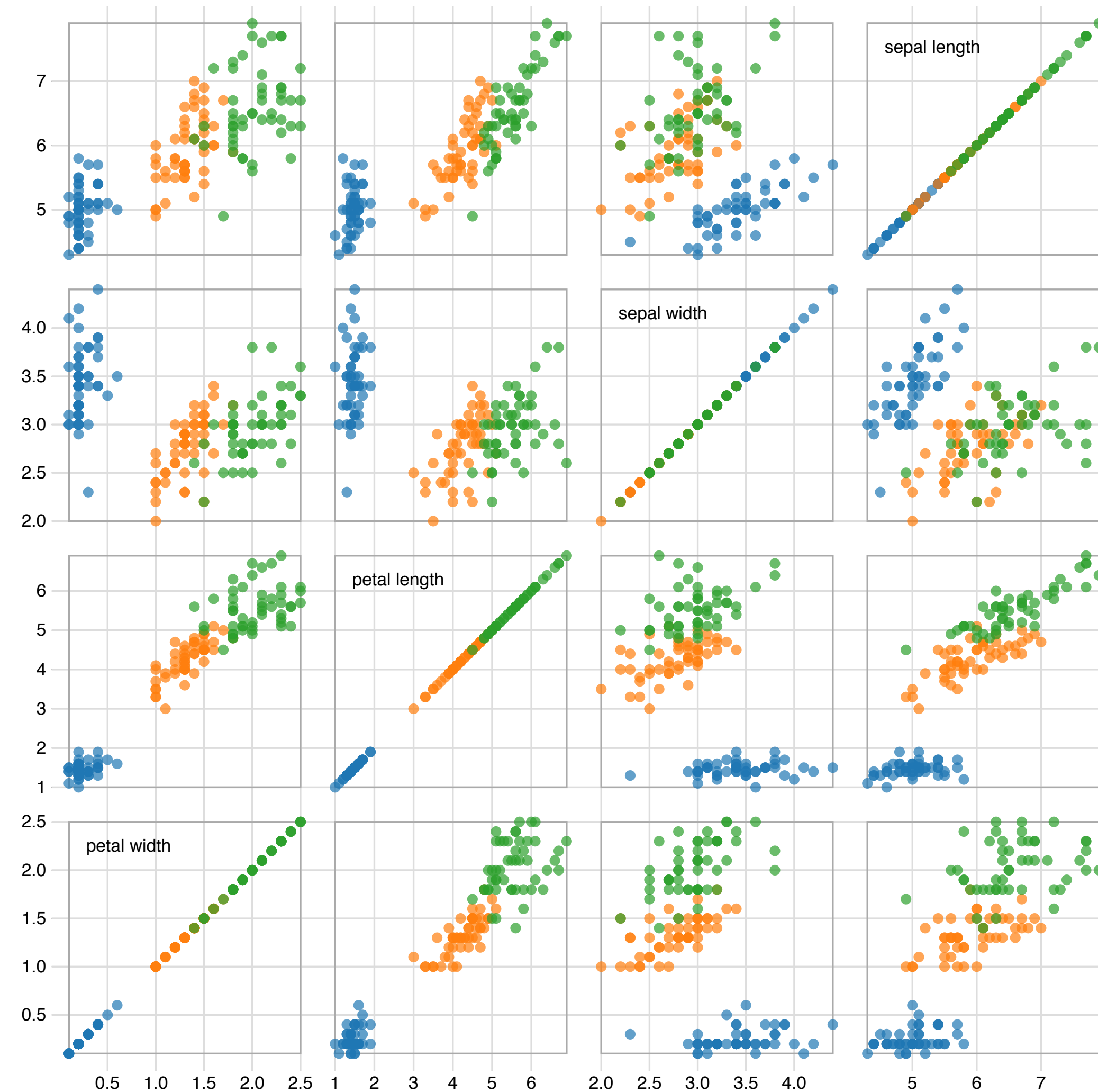
[Altair]

# Multiple Views in Visualization



[Improvise, Weaver, 2004]

# Multiple Views in Visualization



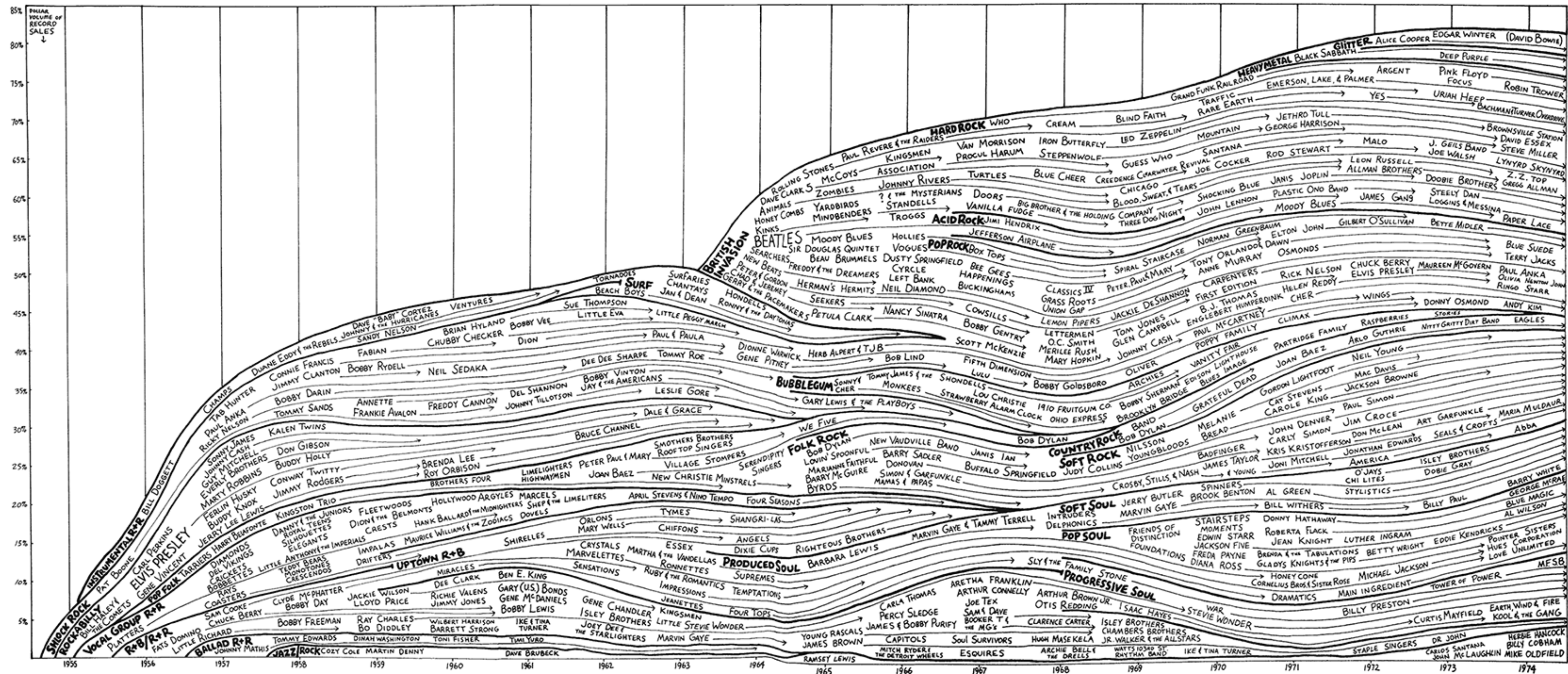
[M. Bostock]

# Altair Supports Concatenation, Layering, & Repetition

---

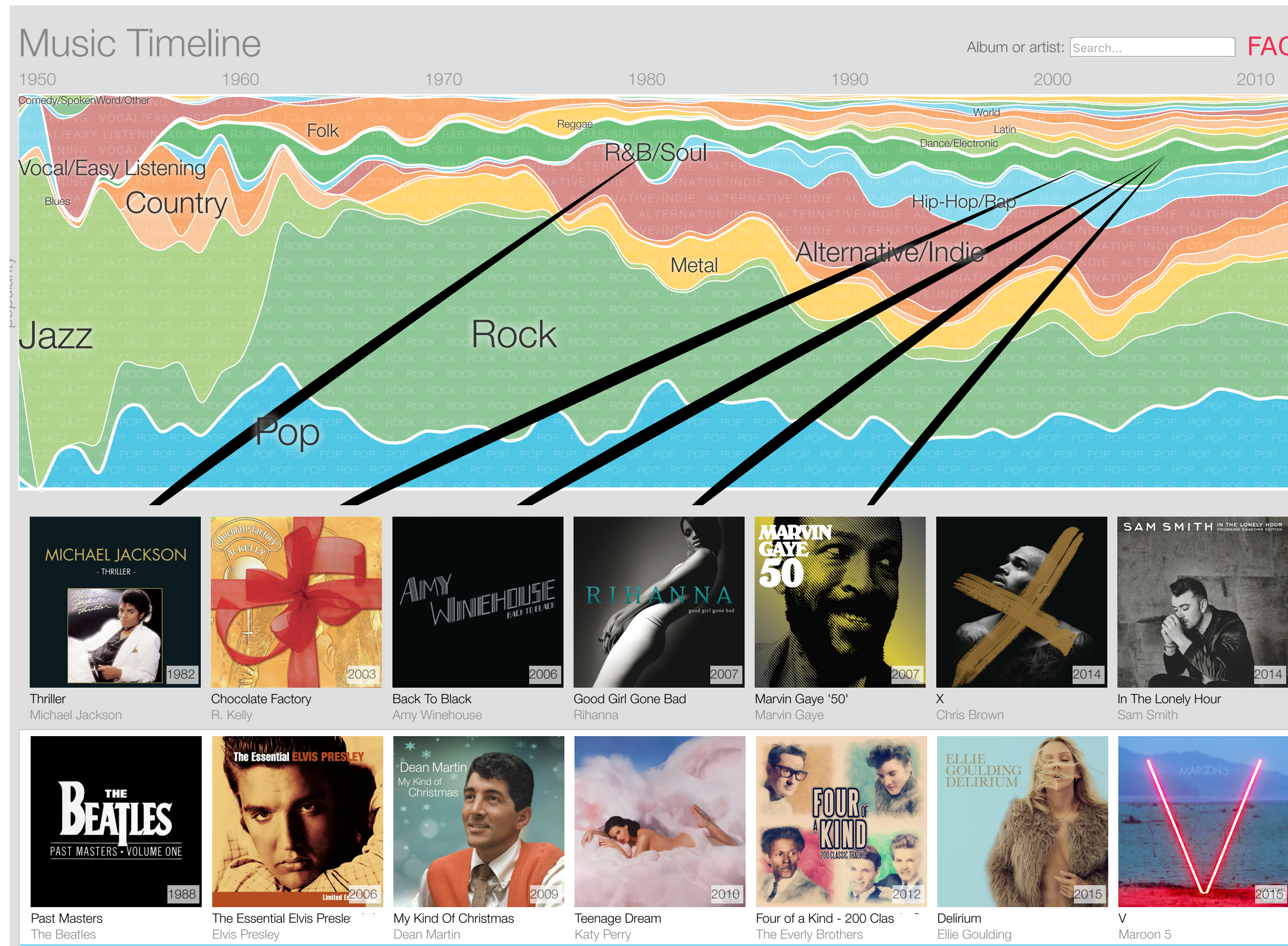
- Layering:
  - + Operator
- Concatenation:
  - Horizontal: | operator
  - Vertical: & operator
- Repetition
  - Use of .repeat for layout
  - Reference repeated variables in the encoding

# Visualization



[Rock 'N' Roll is Here to Pay, R. Garofalo, 1977 (via Tufte)]

# Also Visualization, but with Interaction



[Music Timeline, Google Research (no working version)]

# Interaction

---

- Grammar of Graphics, why not Grammar of Interaction?
- Vega-Lite/Altair is about **interactive** graphics
- Types of Interactions:
  - Selection
  - Zoom
  - Brushing

# Selection

---

- Selection is often used to initiate other changes
- User needs to select something to drive the next change
- What can be a selection target?
  - Items, links, attributes, (views)
- How?
  - mouse click, mouse hover, touch
  - keyboard modifiers, right/left mouse click, force
- Selection modes:
  - Single, multiple
  - Contiguous?

# Highlighting

---

- Selection is the user action
- Feedback is important!
- How? Change selected item's visual encoding
  - Change color: want to achieve visual popout
  - Add outline mark: allows original color to be preserved
  - Change size (line width)
  - Add motion: marching ants



# Highlighting

---

- Selection is the user action
- Feedback is important!
- How? Change selected item's visual encoding
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  - Change size (line width)
  - Add motion: marching ants



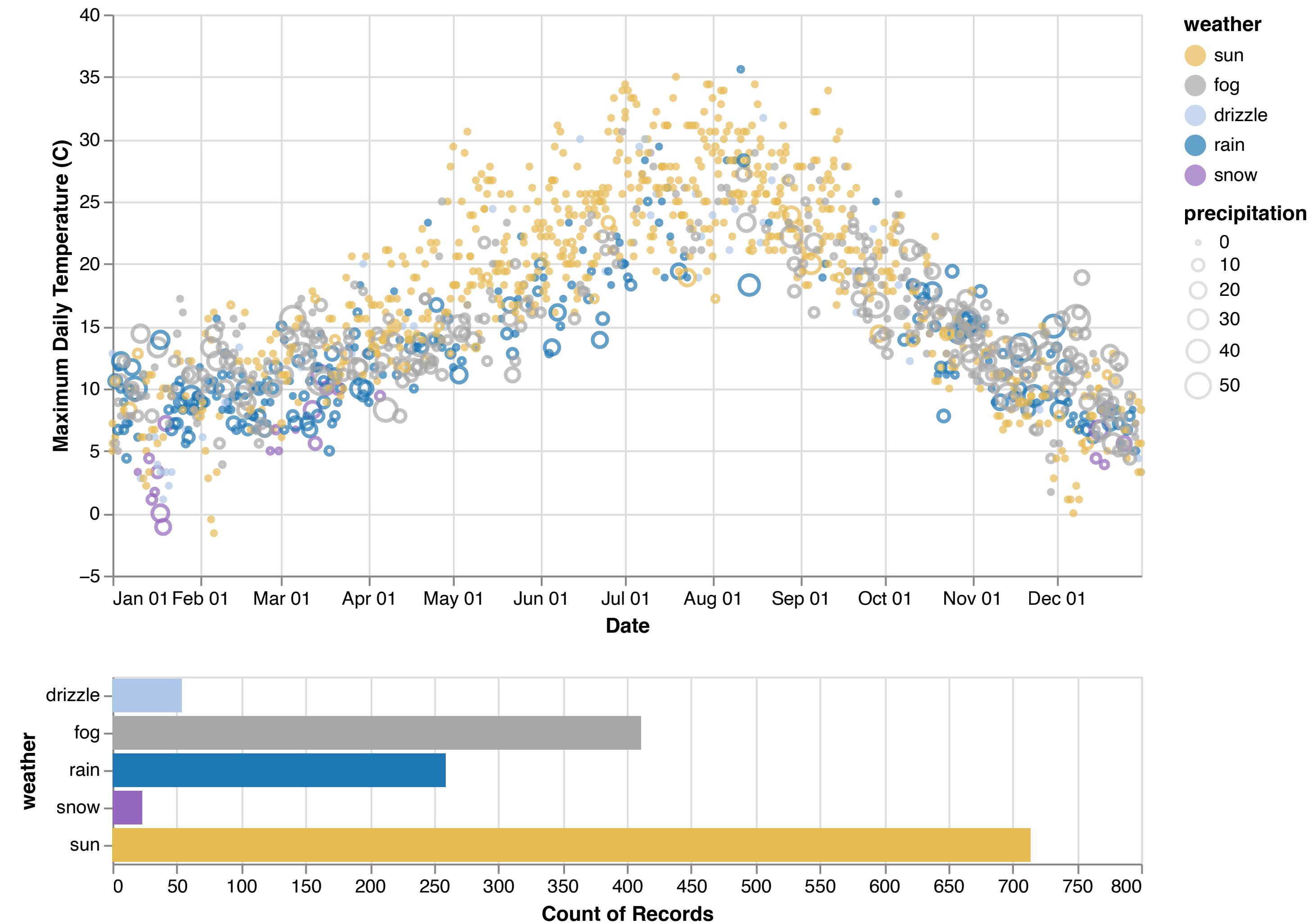
# Altair's Interactive Charts

---

- <https://altair-viz.github.io/gallery/index.html#interactive-charts>

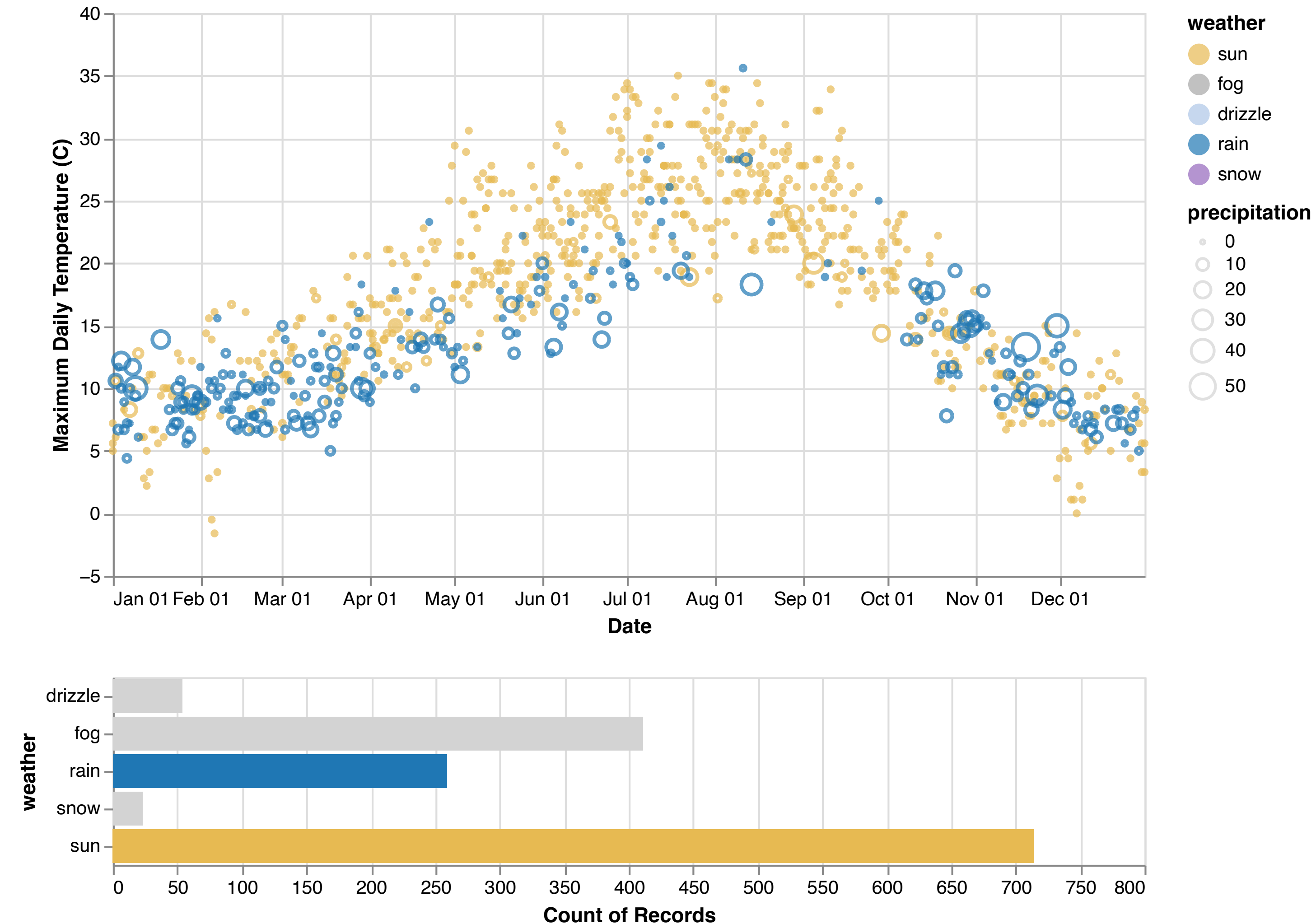
# Interaction

Seattle Weather: 2012-2015



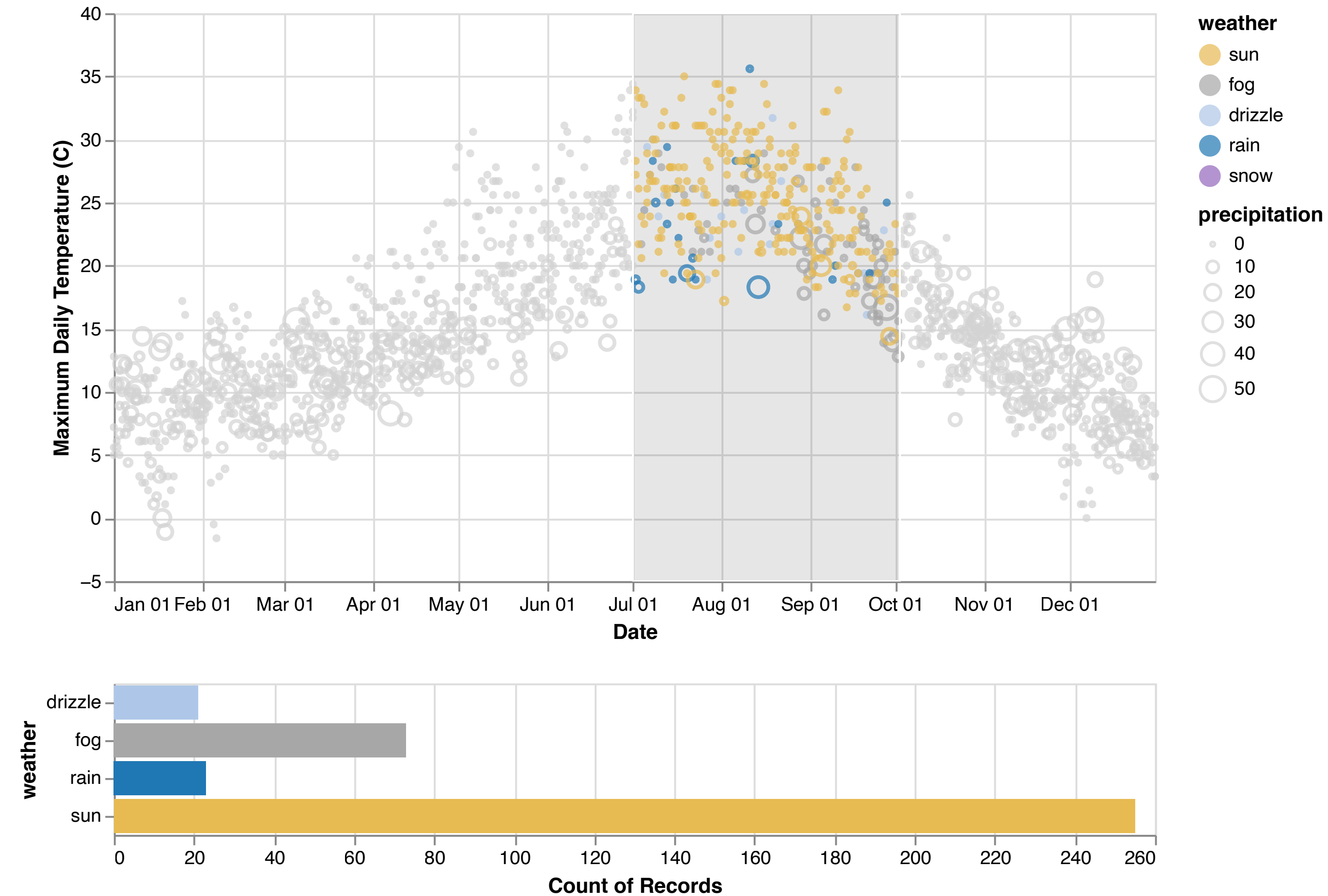
# Weather Selection: Rain vs. Sun

Seattle Weather: 2012-2015



# Date Selection: July-September Sun

Seattle Weather: 2012-2015



# Machine Learning in Python

# Tasks Machine Learning can Help With

- Identifying the zip code from handwritten digits on an envelope



- Detecting fraudulent activity in credit card transactions
- Identifying topics in a set of blog posts
- Grouping customers with similar preferences

[A. Müller & S. Guido, Introduction to Machine Learning with Python, J. Steppan (MNIST image)]

# When to Use Machine Learning?

---

- ML is used when:
  - Human expertise does not exist (navigating on Mars)
  - Humans can't explain their expertise (speech recognition)
  - Models must be customized (personalized medicine)
  - Models are based on huge amounts of data (genomics)
- ML isn't always useful:
  - Calculating payroll...

[E. Alpaydin via [E. Eaton](#)]

# Questions when building a machine learning solution

---

- What question(s) am I trying to answer? Do I think the data collected can answer that question?
- What is the best way to phrase my question(s) as a machine learning problem?
- Have I collected enough data to represent the problem I want to solve?
- What features of the data did I extract, and will these enable the right predictions?
- How will I measure success in my application?

[A. Müller & S. Guido]

# Machine Learning Workflow Overview

---

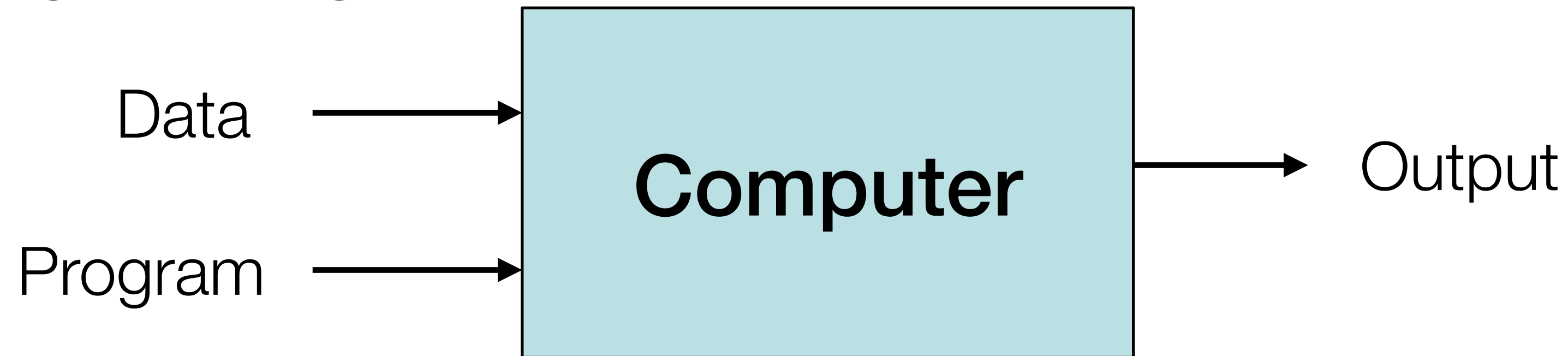
1. Should I use ML on this problem?
  - Is there a pattern to detect? Can I solve it analytically? Do I have data?
2. Gather and organize data.
  - Preprocessing, cleaning, visualizing.
3. Establishing a baseline.
4. Choosing a model, loss, regularization, ...
5. Optimization (could be simple, could be a Phd...).
6. Hyperparameter search.
7. Analyze performance & mistakes, and iterate back to step 4 (or 2).

[R. Grosse et al.]

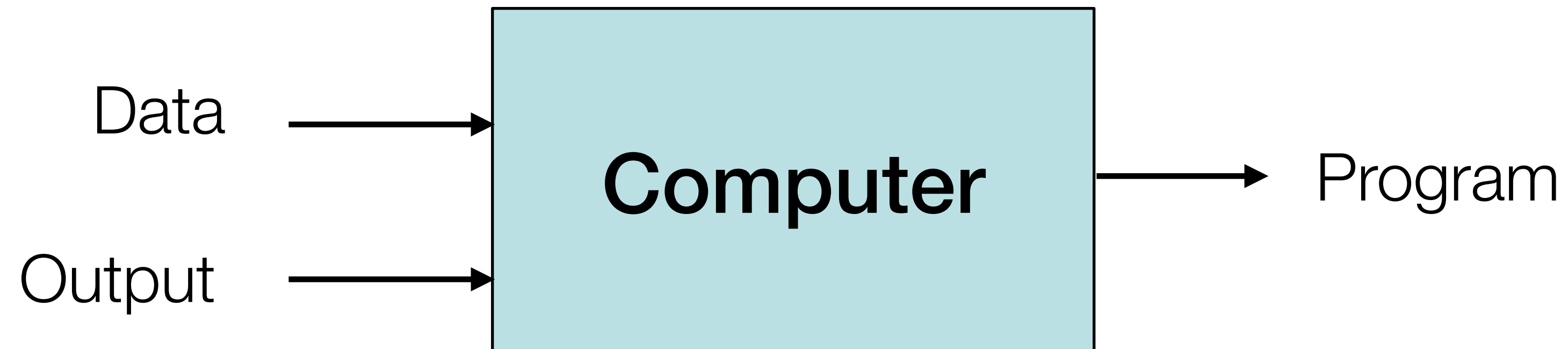
# Machine Learning

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- Traditional Programming



- Machine Learning



[P. Domingos]

# Machine Learning

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- Every machine learning algorithm has three components:
  - Representation
  - Evaluation
  - Optimization

# Representation

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- Decision trees
- Sets of rules / Logic programs
- Instances
- Graphical models (Bayes/Markov nets)
- Neural networks
- Support vector machines
- Model ensembles
- Etc.

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# Evaluation

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- Accuracy
- Precision and recall
- Squared error
- Likelihood
- Posterior probability
- Cost / Utility
- Margin
- Entropy
- K-L divergence
- Etc.

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# Optimization

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- Combinatorial optimization
  - E.g.: Greedy search
- Convex optimization
  - E.g.: Gradient descent
- Constrained optimization
  - E.g.: Linear programming

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# Final Exam

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- Wednesday, Dec. 11, **8:00**-9:50am in PM 103
- **More comprehensive** than Test 2
- Expect questions from topics covered on Test 1 and 2
- Expect questions from the last few weeks of class (data, visualization, machine learning)
- Similar format
  
- Bring Questions on Wednesday!