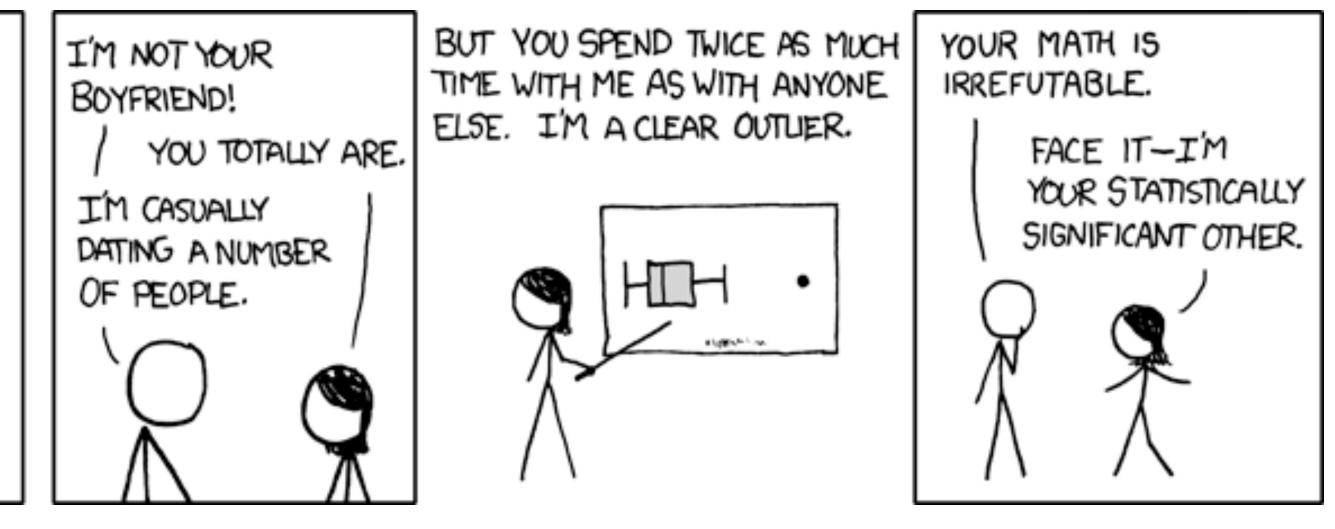
CS-5630 / CS-6630 Uisualization Fitering & Aggregation



CAN MY BOYFRIEND COME ALONG?

Alexander Lex alex@sci.utah.edu



Administrativa

This/Next Week

Tomorrow: Group announcement Saturday: Homework 5 Monday: Project Proposal Put proposal into github repo Submit only link via canvas Tuesday: Peer Feedback (mandatory, graded) Thursday: Class cancelled



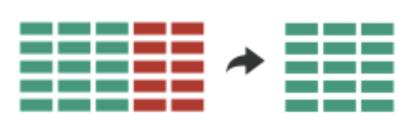
Filter & Flggregate

Reducing Items and Attributes

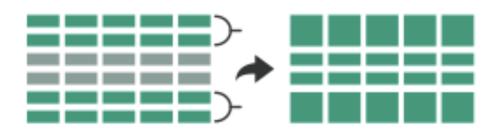
- → Filter
 - → Items



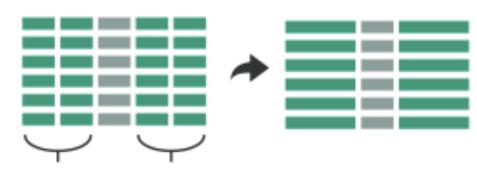
→ Attributes



- → Aggregate
 - → Items



→ Attributes





Filter

elements are eliminated What drives filters?

Any possible function that partitions a dataset into two sets

Bigger/smaller than x

Fold-change

Noisy/insignificant



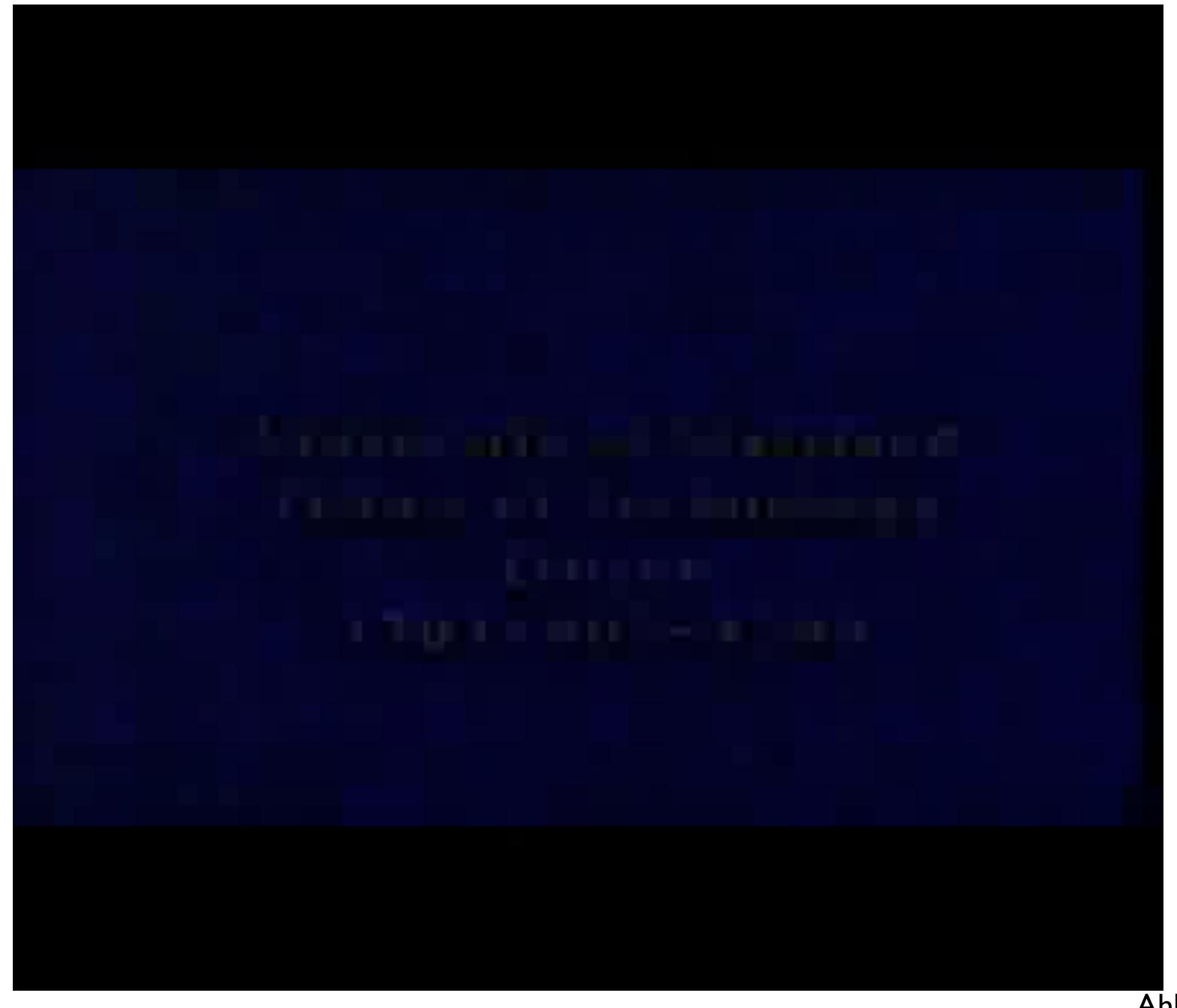
→ Attributes



Dynamic Queries / Filters

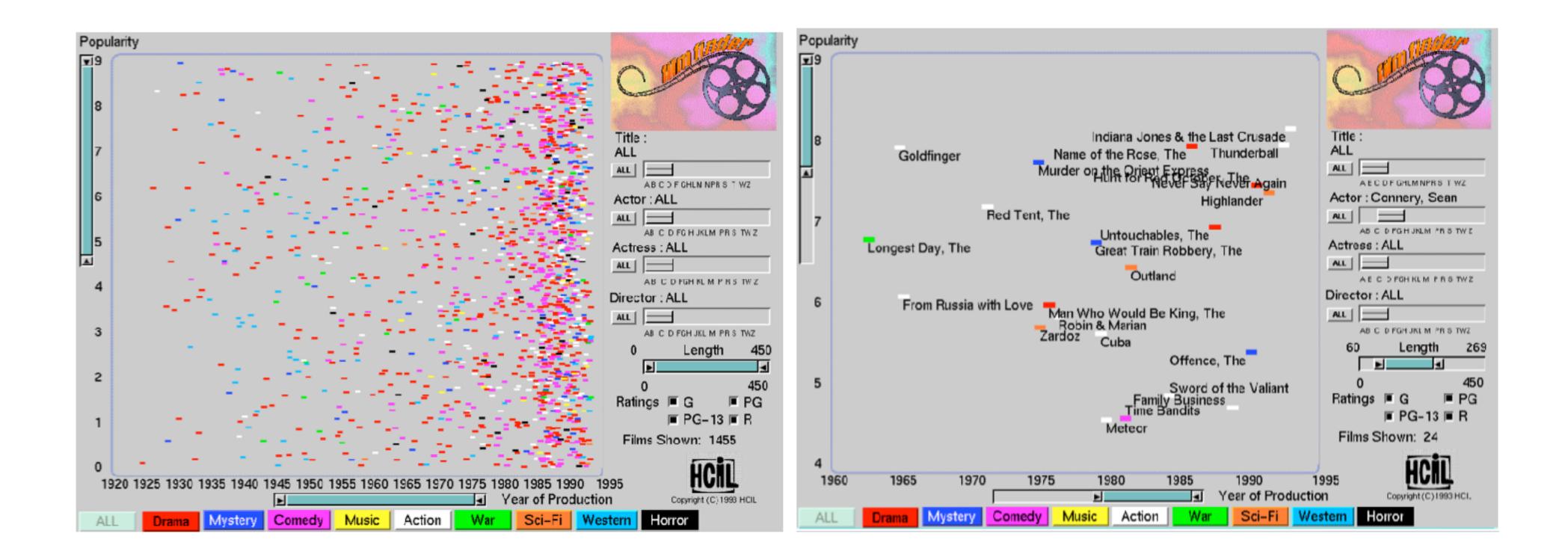
- coupling between encoding and interaction so that user can immediately see the results of an action

Queries: start with 0, add in elements Filters: start with all, remove elements Approach depends on dataset size



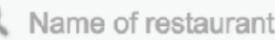
Ahlberg 1994

ITEM FILTERING

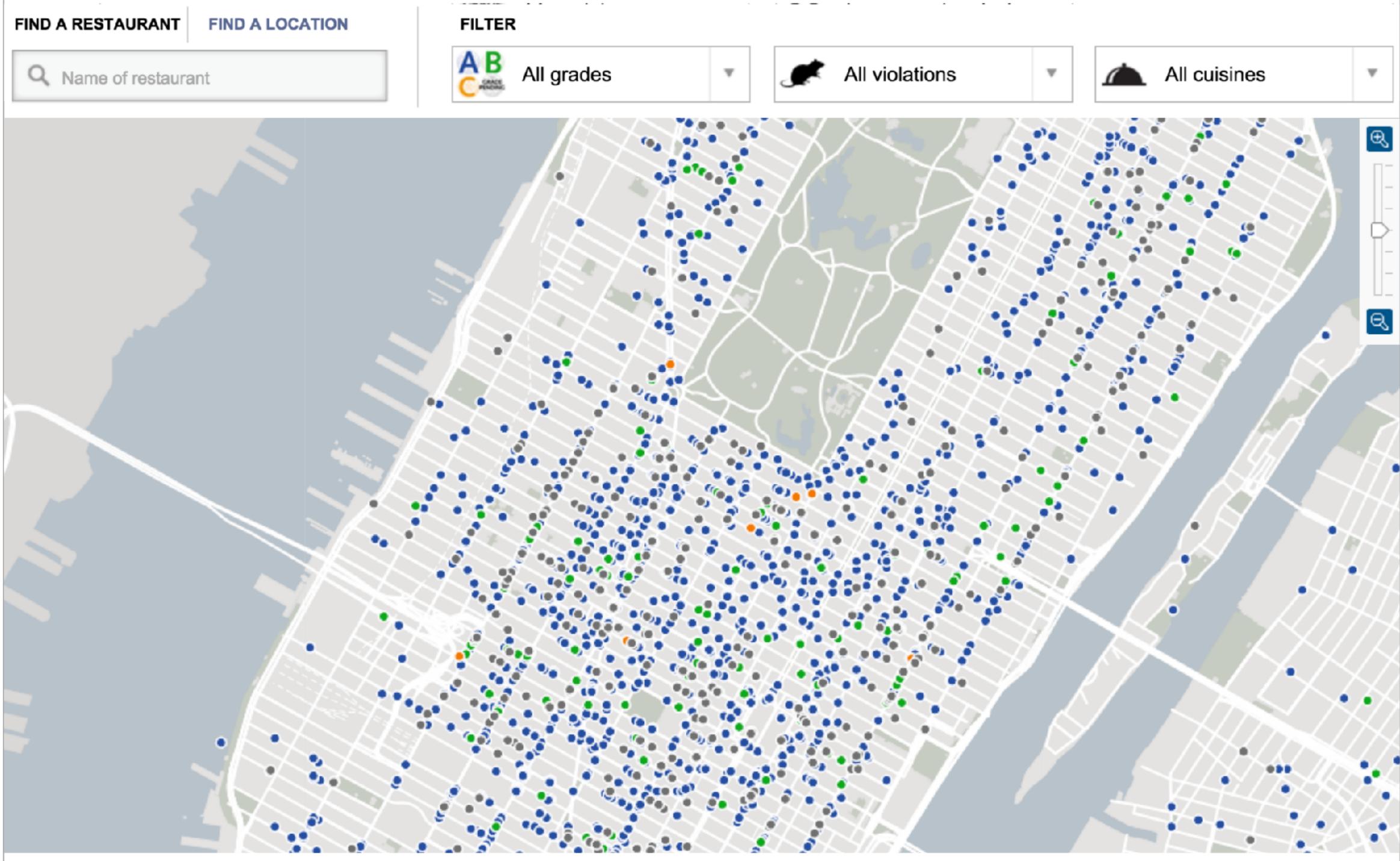




Ahlberg 1994

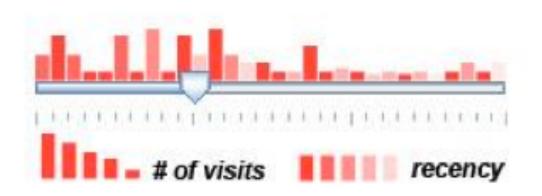




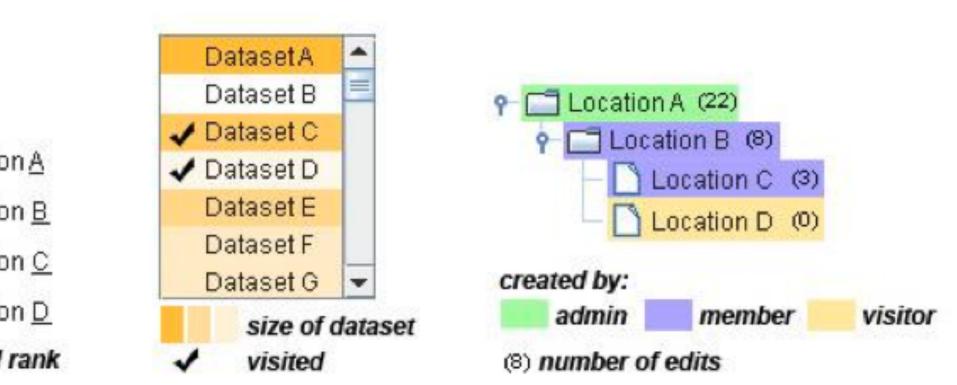


Scented Widgets

information scent: user's (imperfect) perception of data GOAL: lower the cost of information foraging through better cues

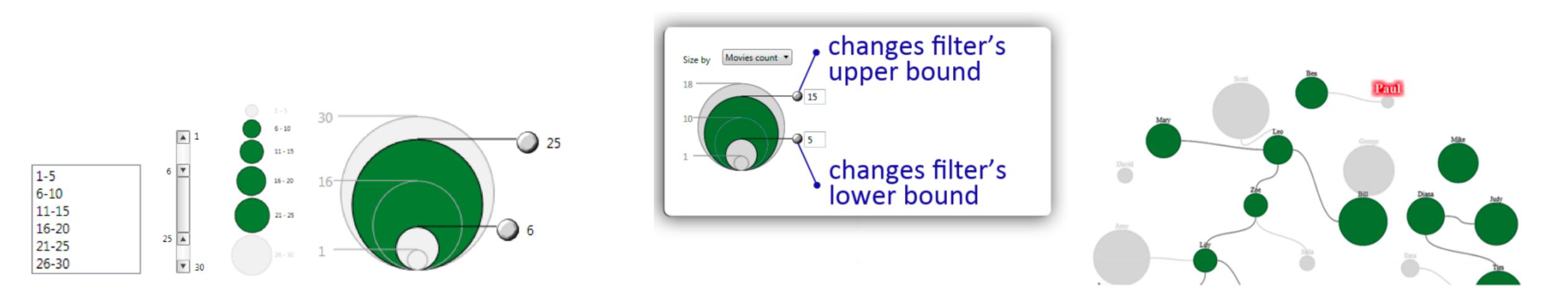






Interactive Legends

Controls combining the visual representation of static legends with interaction mechanisms of widgets Define and control visual display together

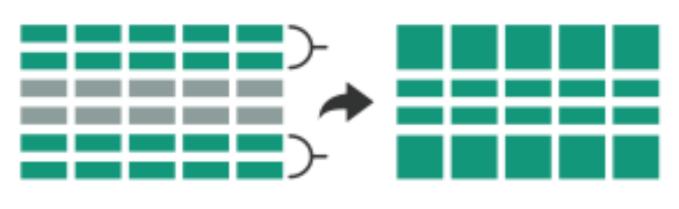


flggregation

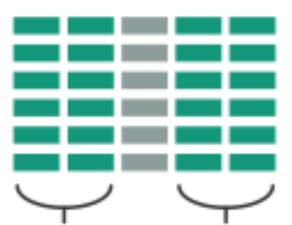
Aggregate

a group of elements is represented by a (typically smaller) number of derived elements

→ Items

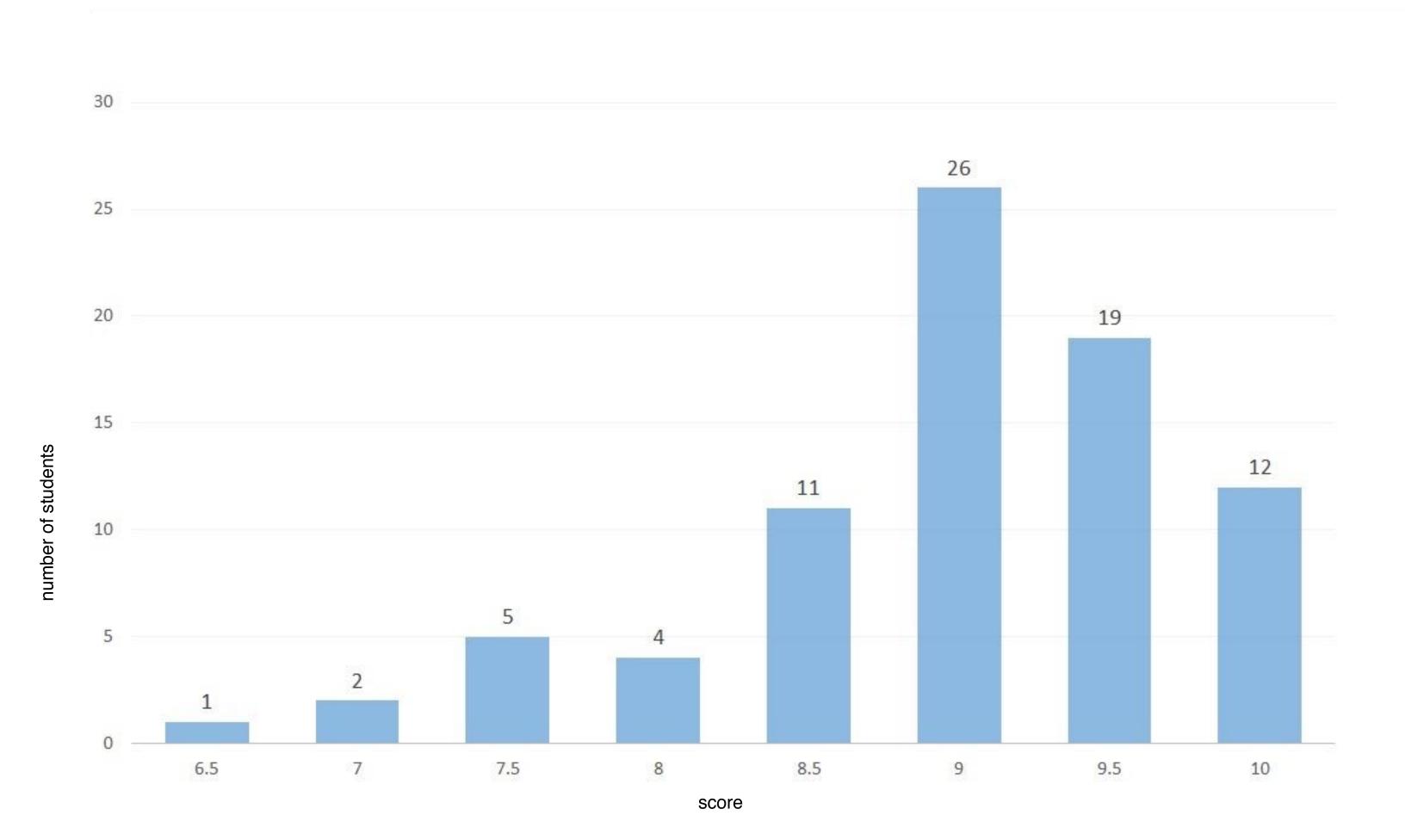


→ Attributes



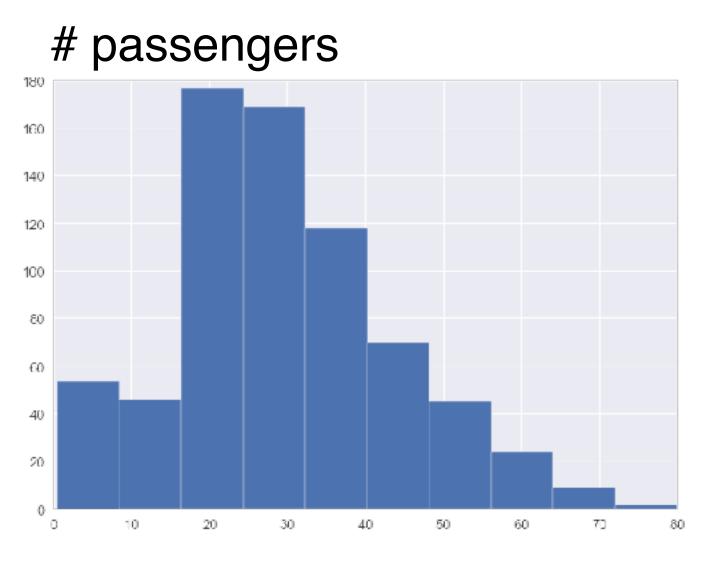
Item Aggregation

Histogram

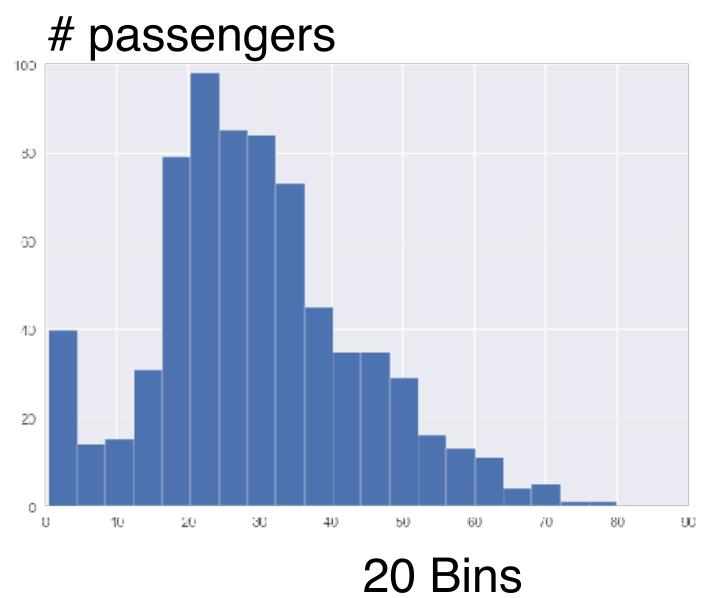


Histogram

Good #bins hard to predict make interactive! rule of thumb: #bins = sqrt(n)



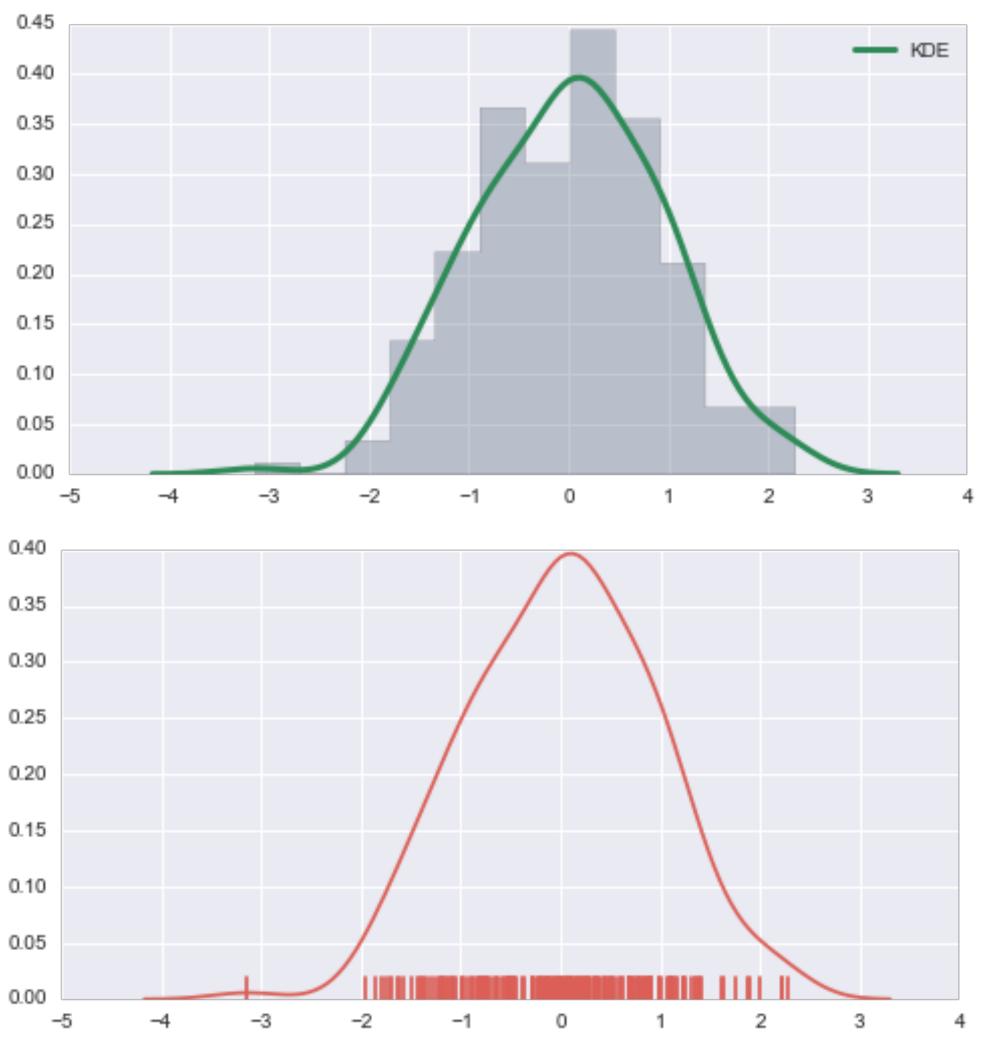




age



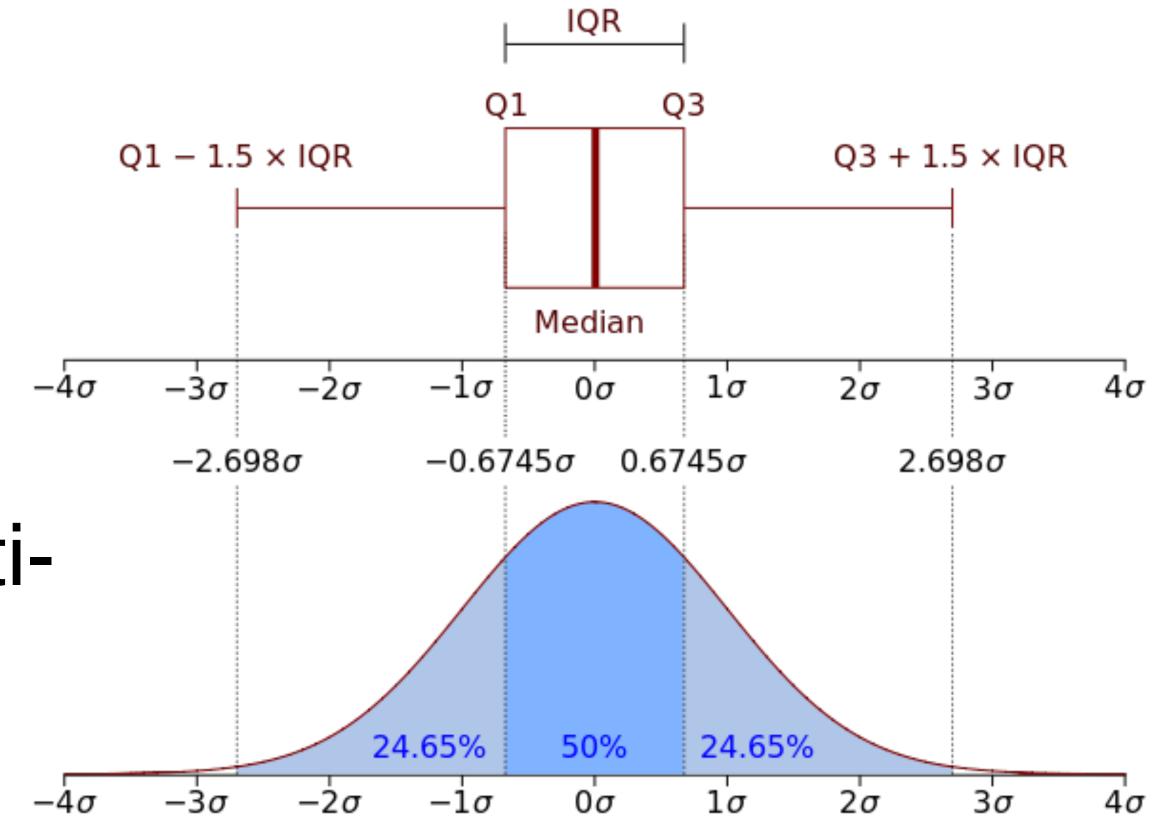
Density Plots



http://web.stanford.edu/~mwaskom/software/seaborn/tutorial/plotting_distributions.html

Box Plots

- aka Box-and-Whisker Plot Show outliers as points!
- Not so great for non-normal distributed data
- Especially bad for bi- or multimodal distributions



Wikipedia

One Boxplot, Four Distributions

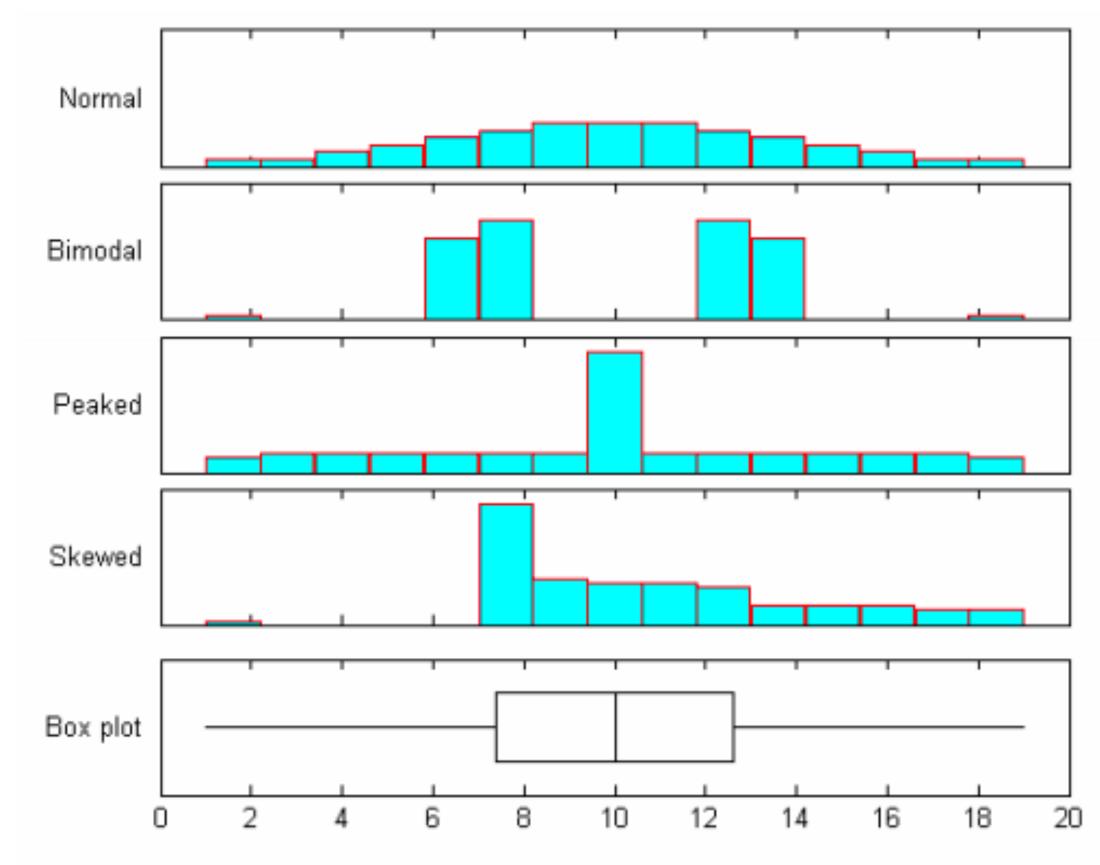


Figure 1: Histograms and box plot: four samples each of size 100

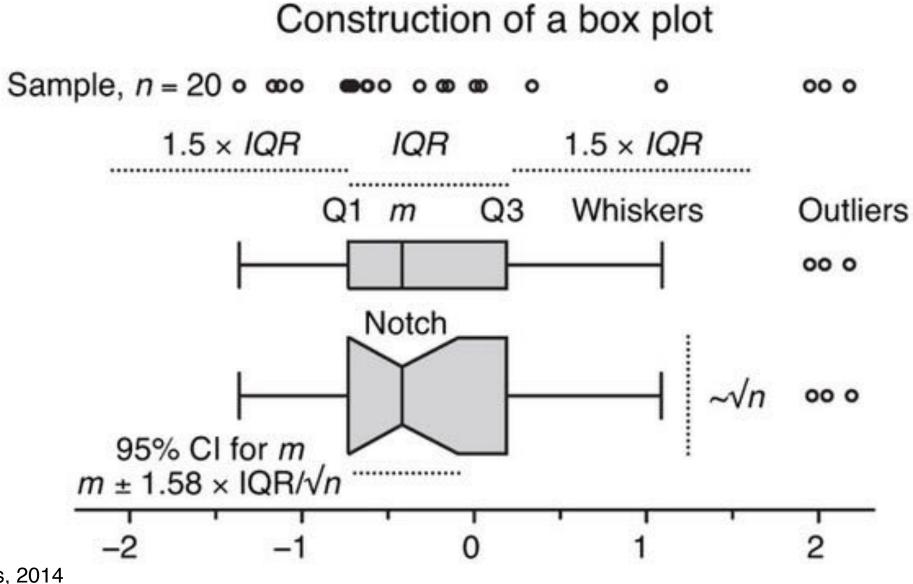
http://stat.mq.edu.au/wp-content/uploads/2014/05/Can_the_Box_Plot_be_Improved.pdf

Notched Box Plots

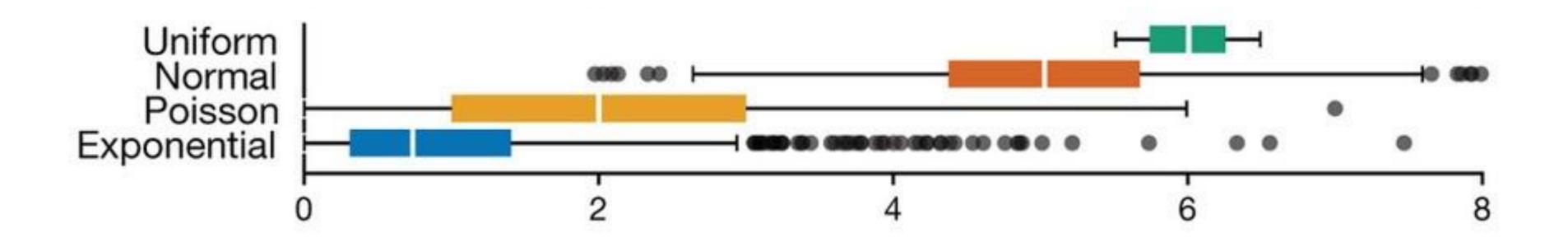
Notch shows m +/- 1.5i x IQR/sqrt(n)

A guide to statistical significance.

Kryzwinski & Altman, PoS, Nature Methods, 2014

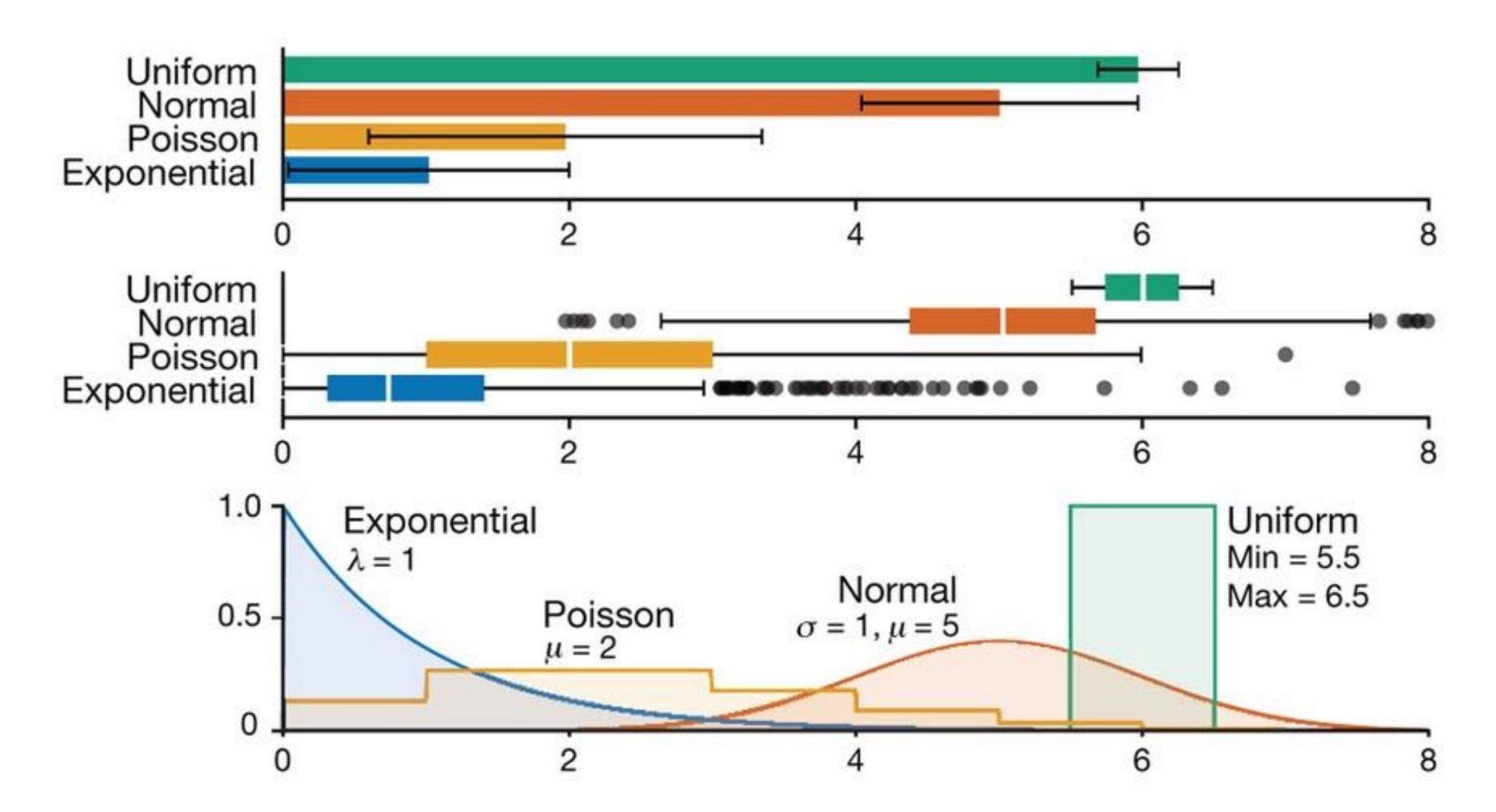


Box(and Whisker) Plots



http://xkcd.com/539/

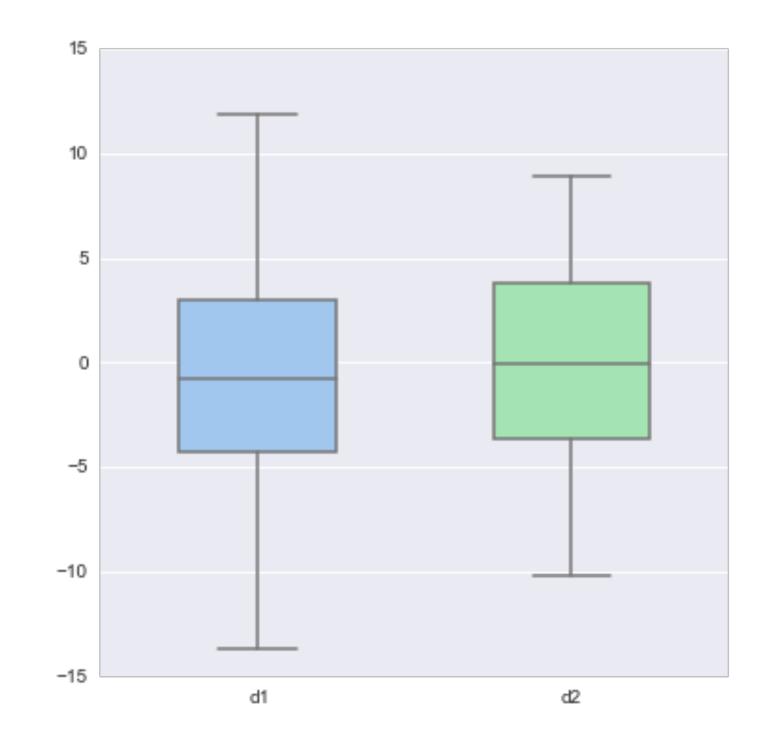
Comparison



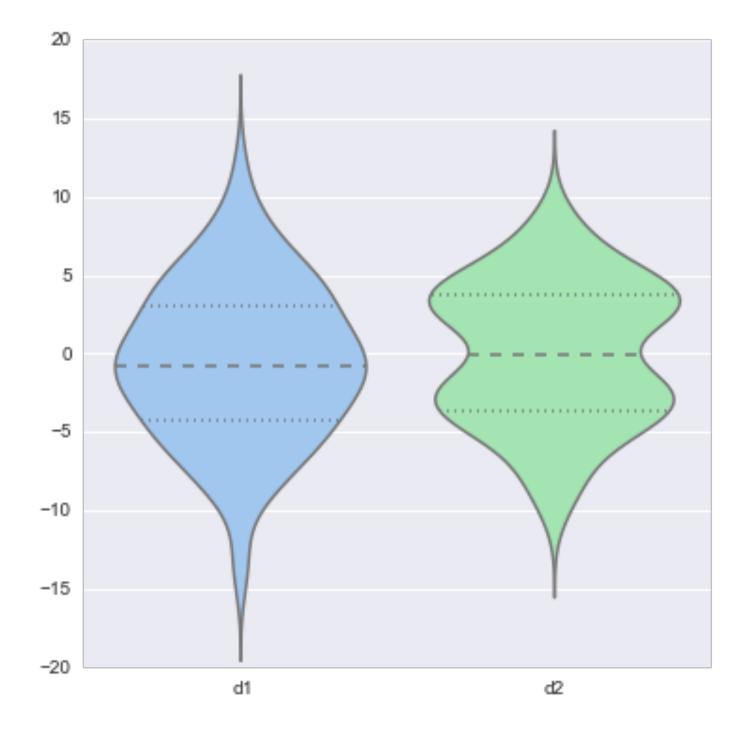
Streit & Gehlenborg, PoV, Nature Methods, 2014

Violin Plot

= Box Plot + Probability Density Function



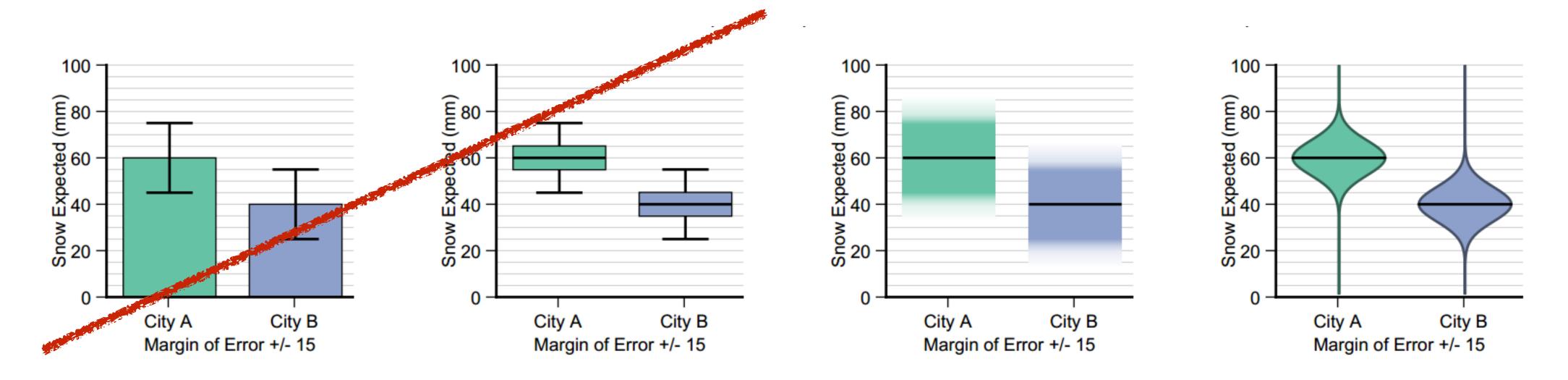




http://web.stanford.edu/~mwaskom/software/seaborn/tutorial/plotting_distributions.html



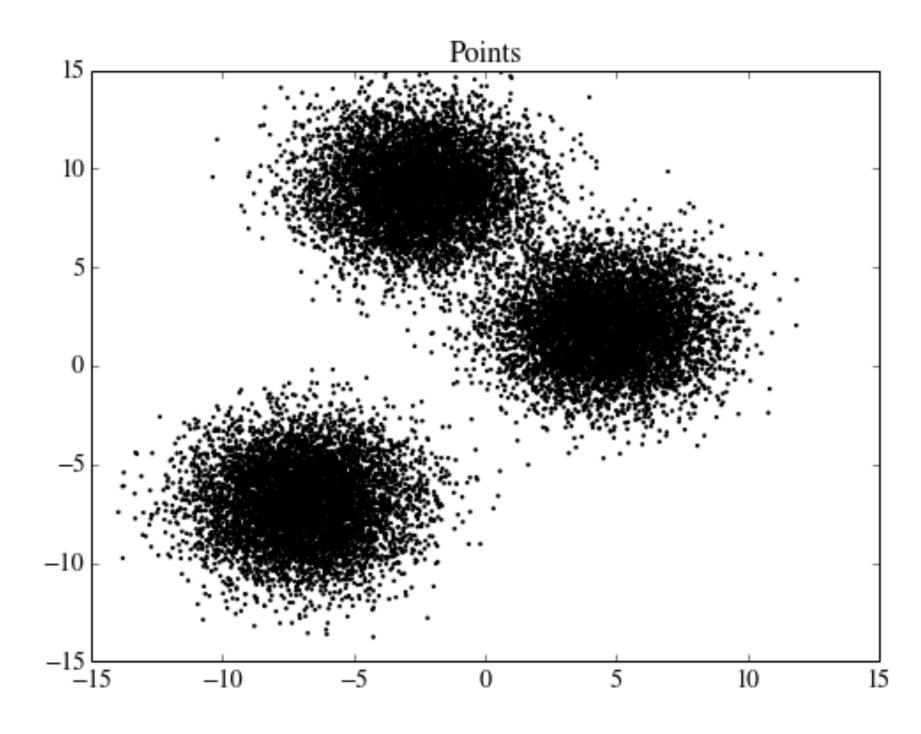
Showing Expected Values & Uncertainty NOT a distribution!

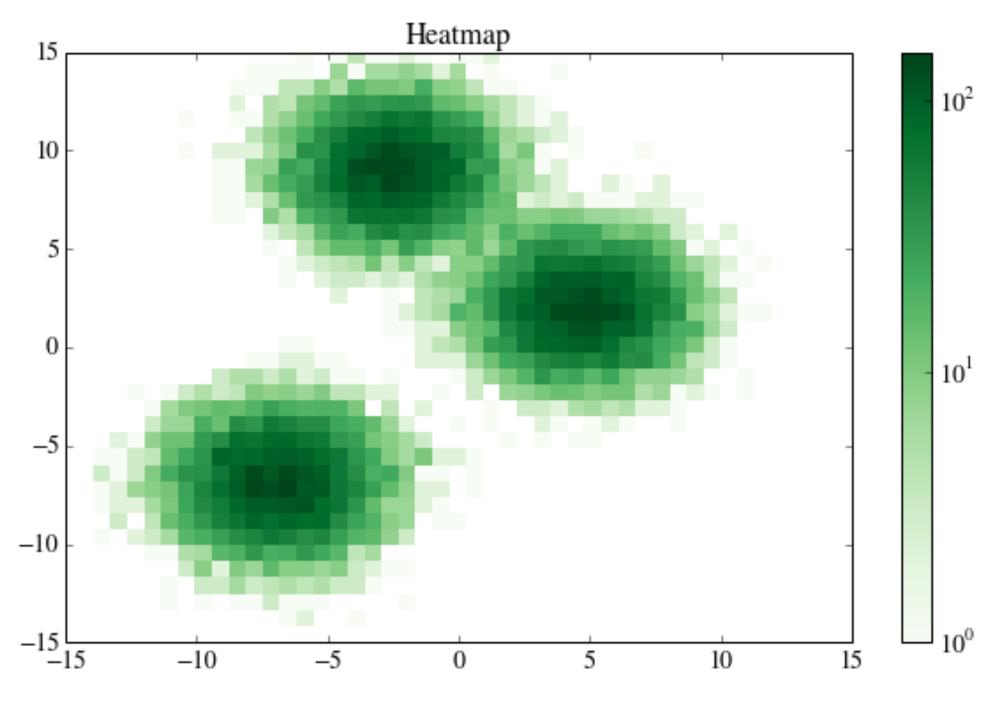


Error Bars Considered Harmful: Exploring Alternate Encodings for Mean and Error Michael Correll, and Michael Gleicher

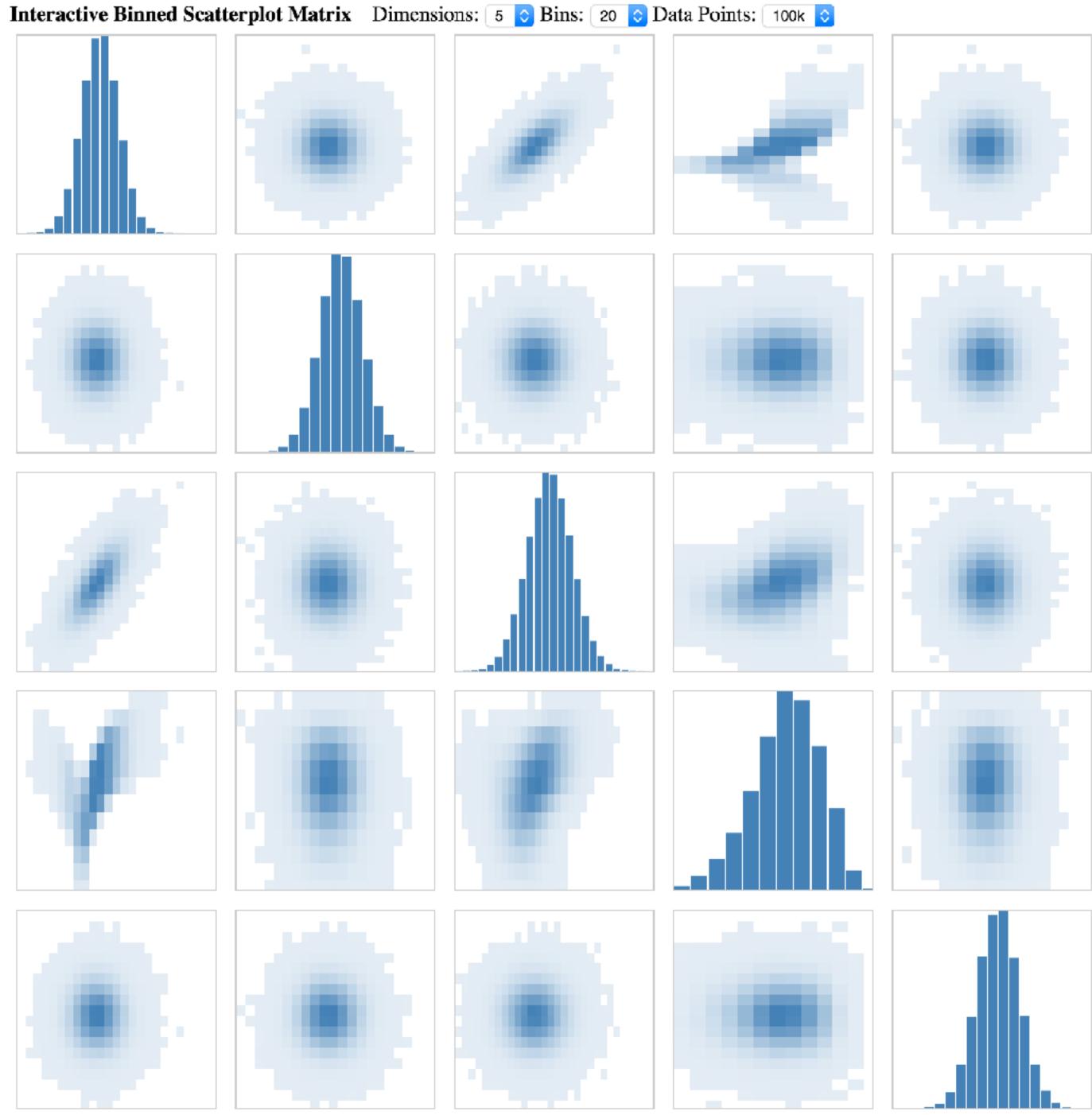
Heat Maps

binning of scatterplots instead of drawing every point, calculate grid and intensities



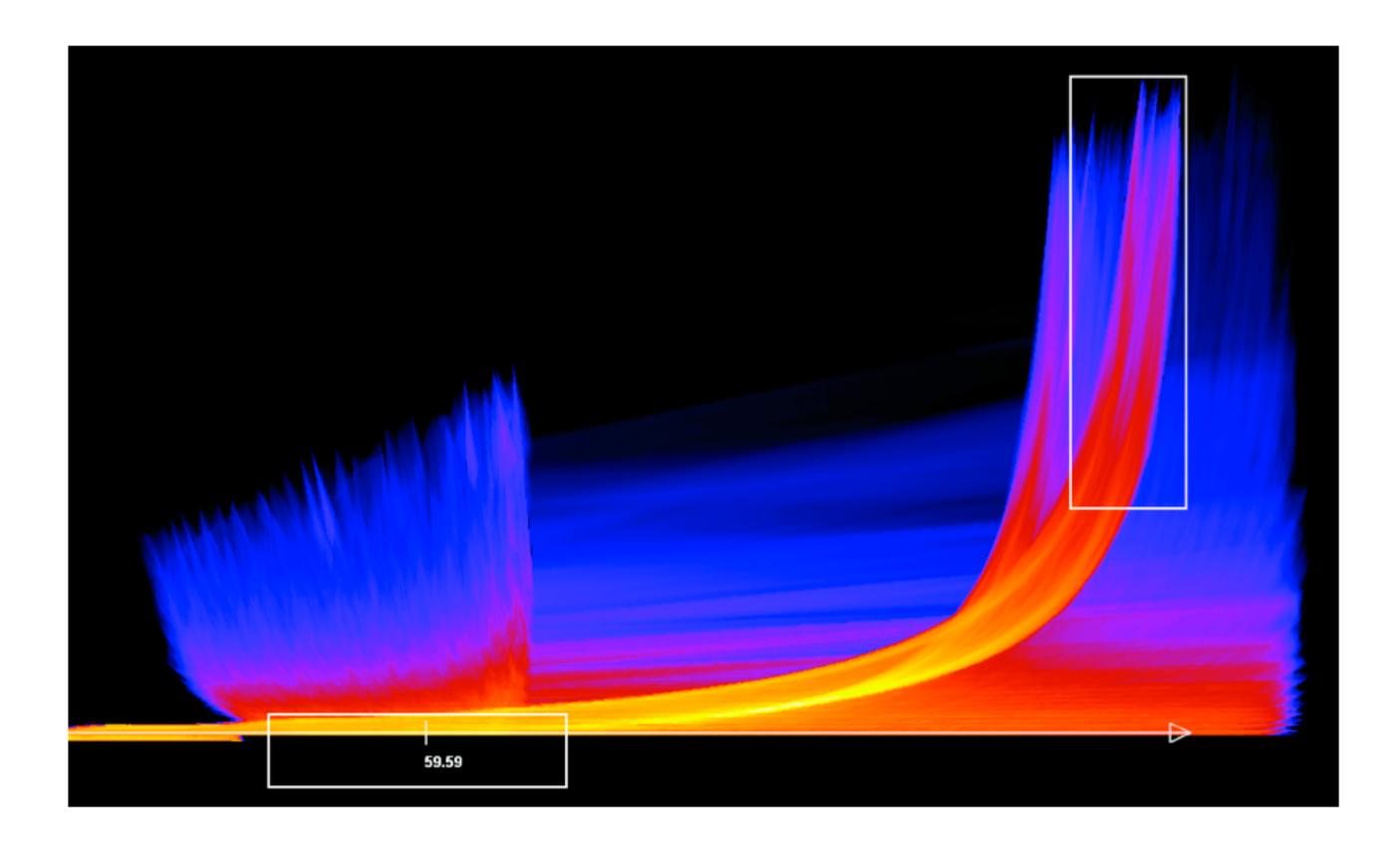


²D Density Plots



Powered by Datavore and D3.

Continuous Scatterplot

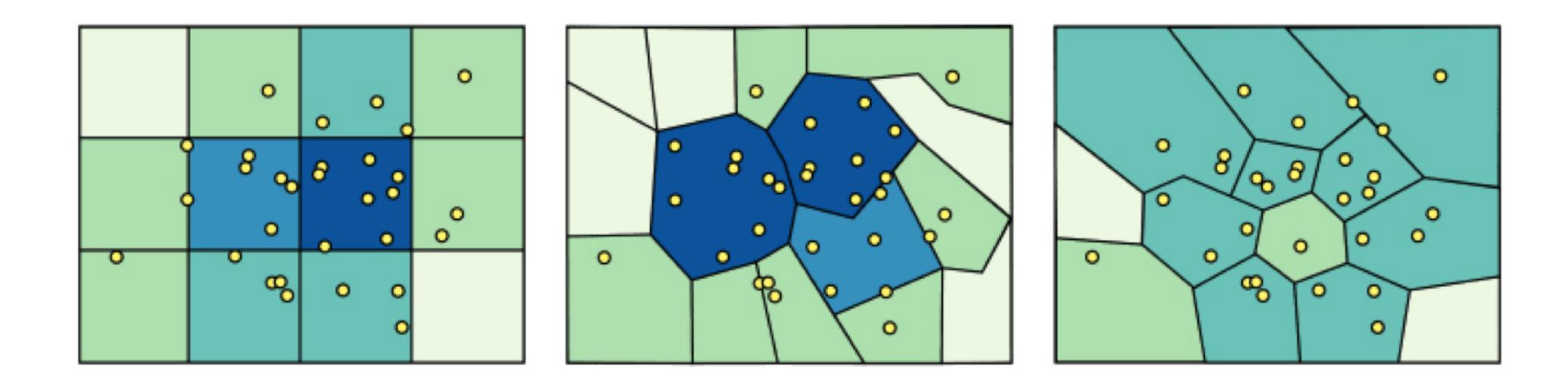


Bachthaler 2008

Spatial Aggregation

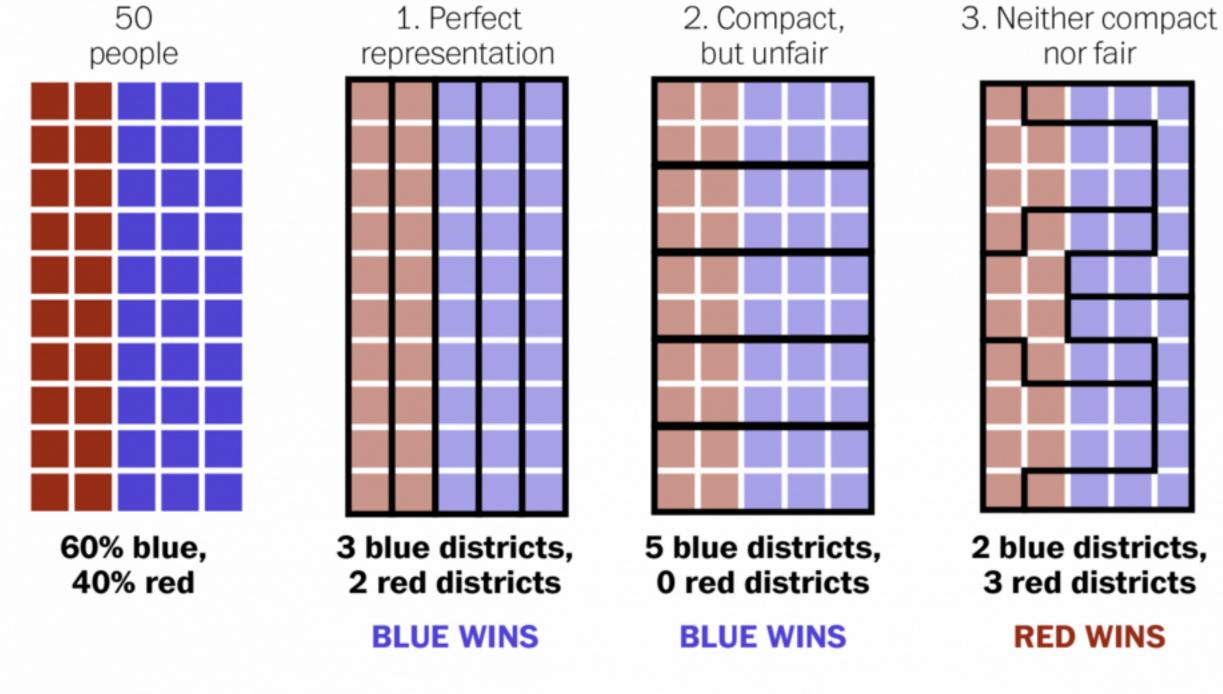
modifiable areal unit problem

in cartography, changing the boundaries of the regions used to analyze data can yield dramatically different results



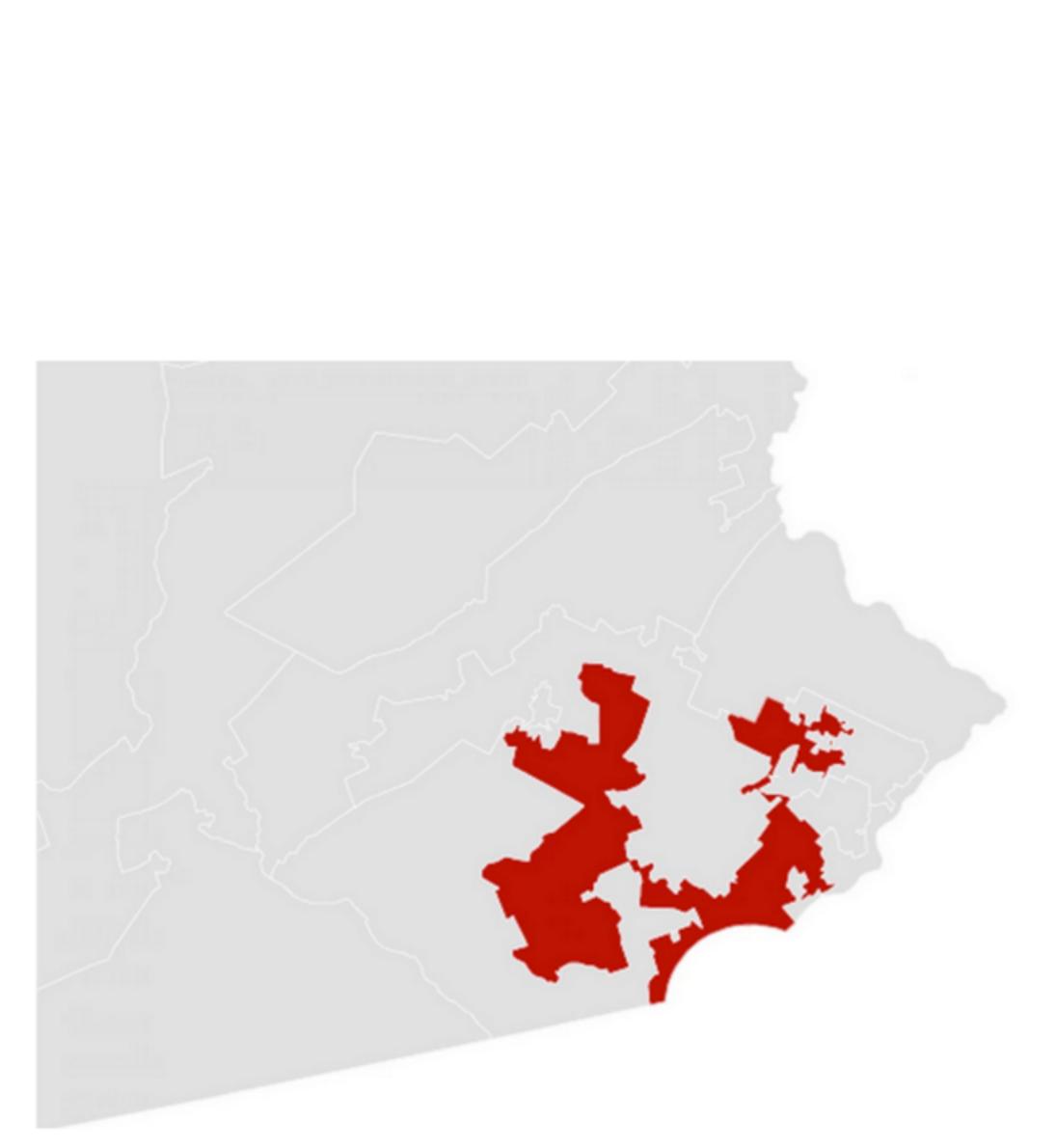
Gerrymandering, explained

Three different ways to divide 50 people into five districts

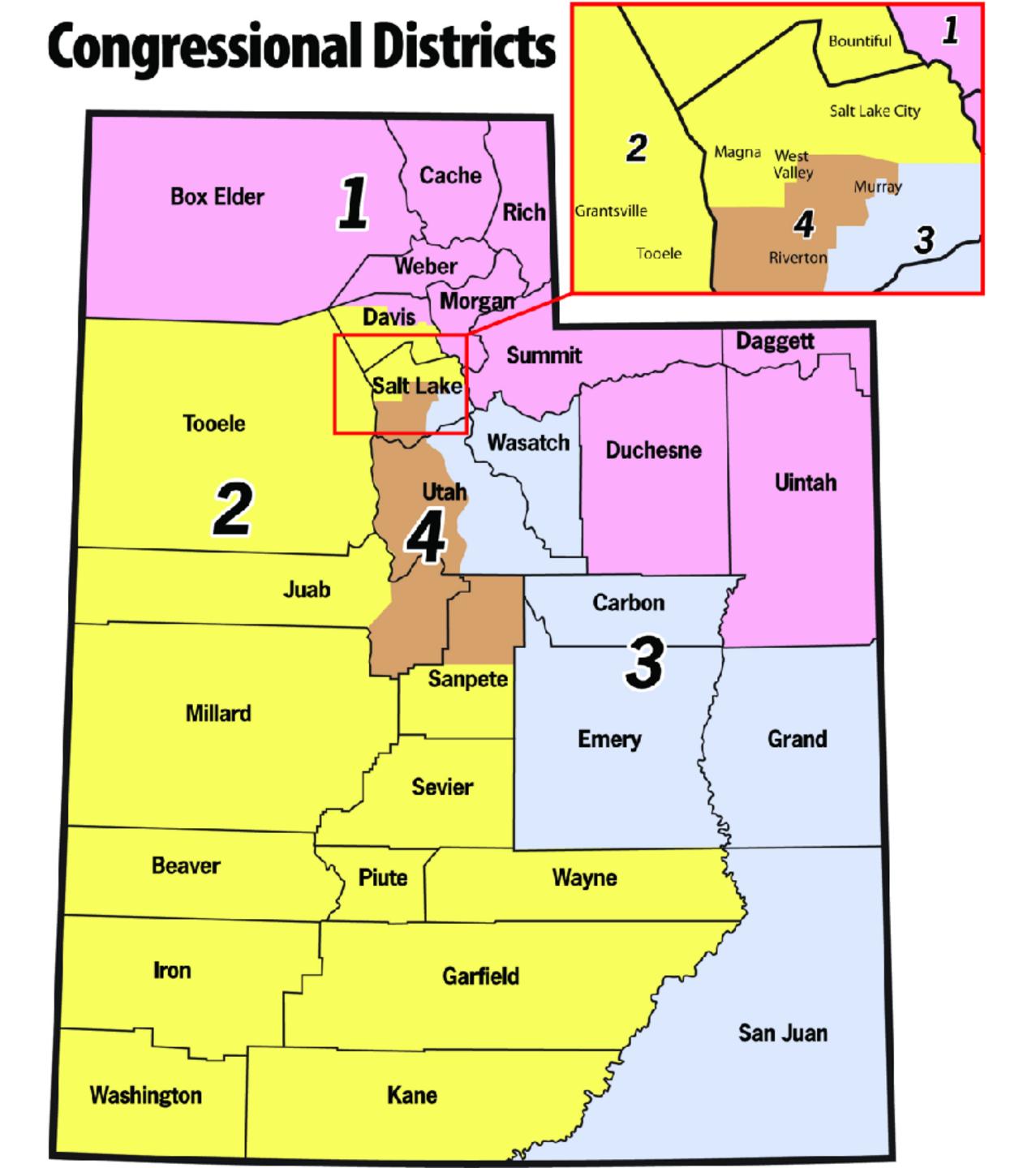


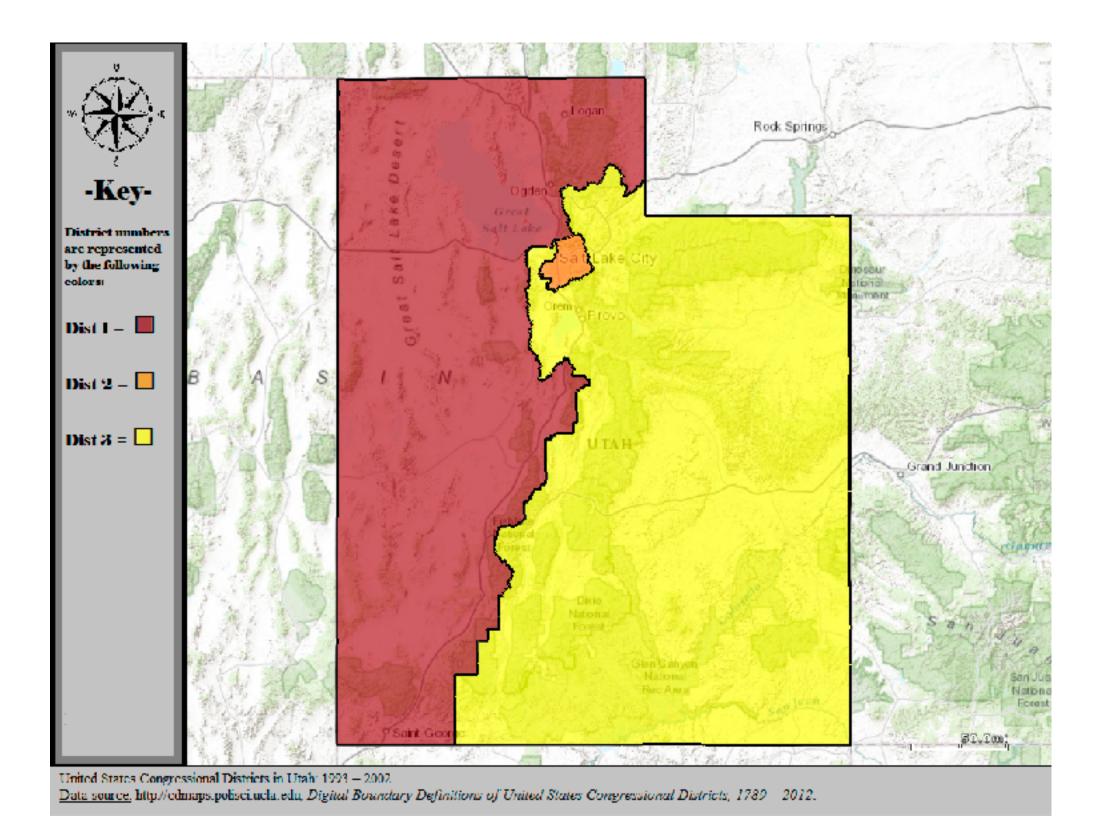
WASHINGTONPOST.COM/WONKBLOG

Adapted from Stephen Nass



A real district in Pennsylvania Democrats won 51% of the vote but only 5 out of 18 house seats





Valid till 2002

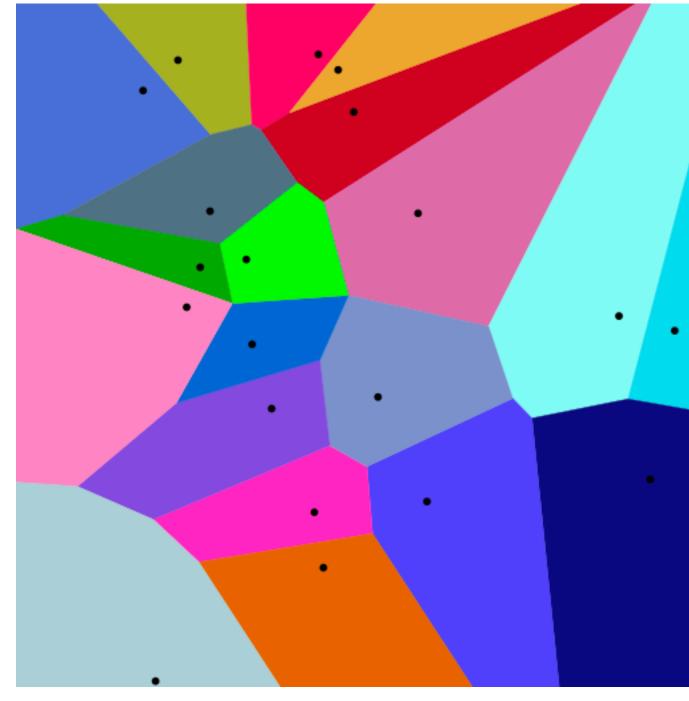
http://www.sltrib.com/opinion/ 1794525-155/lake-salt-republicancounty-http-utah

Voronoi Diagrams

Given a set of locations, for which area is a location n closest?

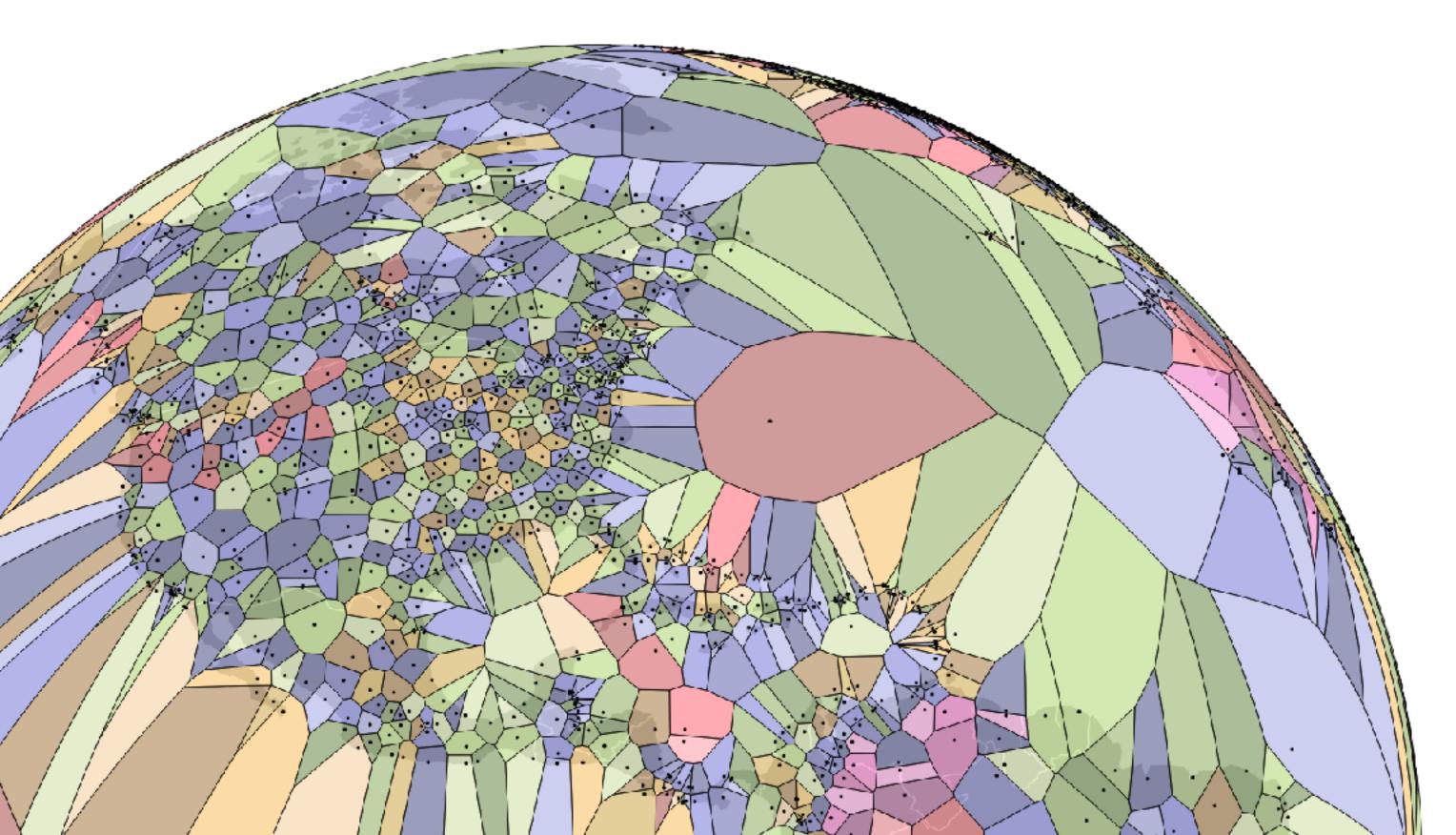
D3 Voronoi Layout:

https://github.com/mbostock/d3/wiki/ Voronoi-Geom



Voronoi Examples

World Airports Voronoi



Sidenote: Voronoi for Interaction

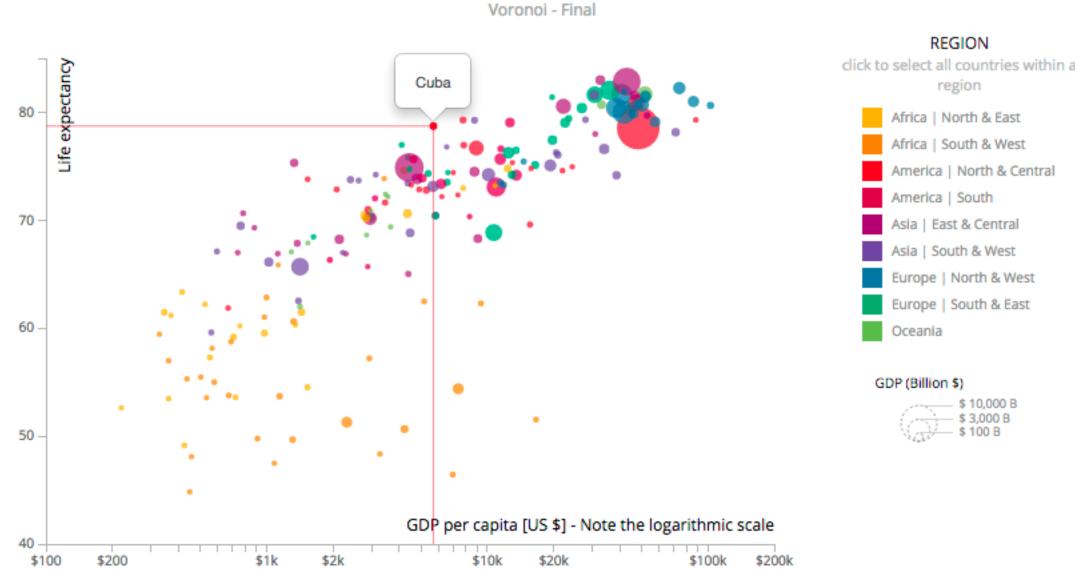


Voronoi for Interaction

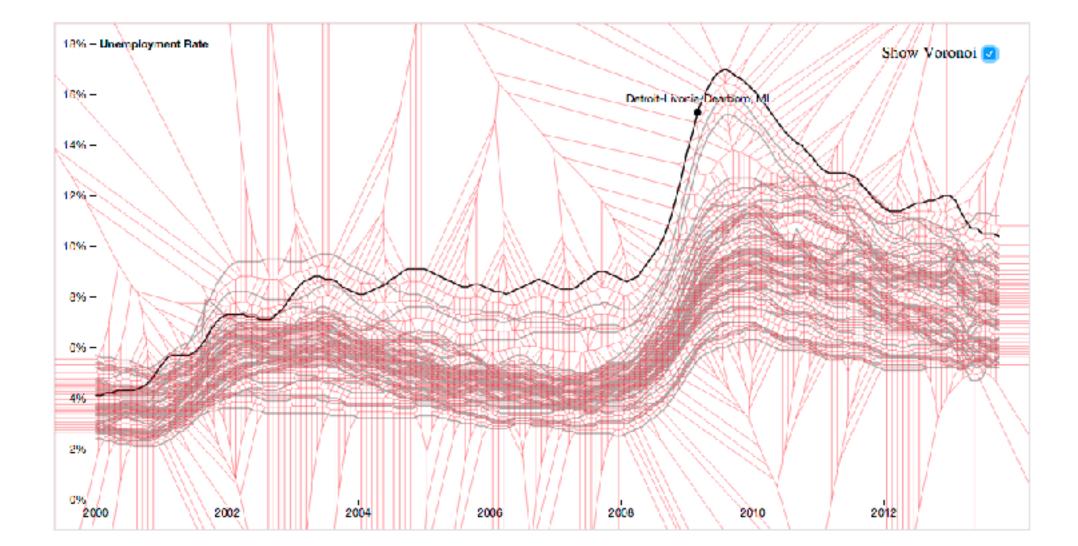
Useful for interaction: Increase size of target area to click/hover

Instead of clicking on point, hover in its region

https://github.com/d3/d3-voronoi/



Life expectancy versus GDP per Capita



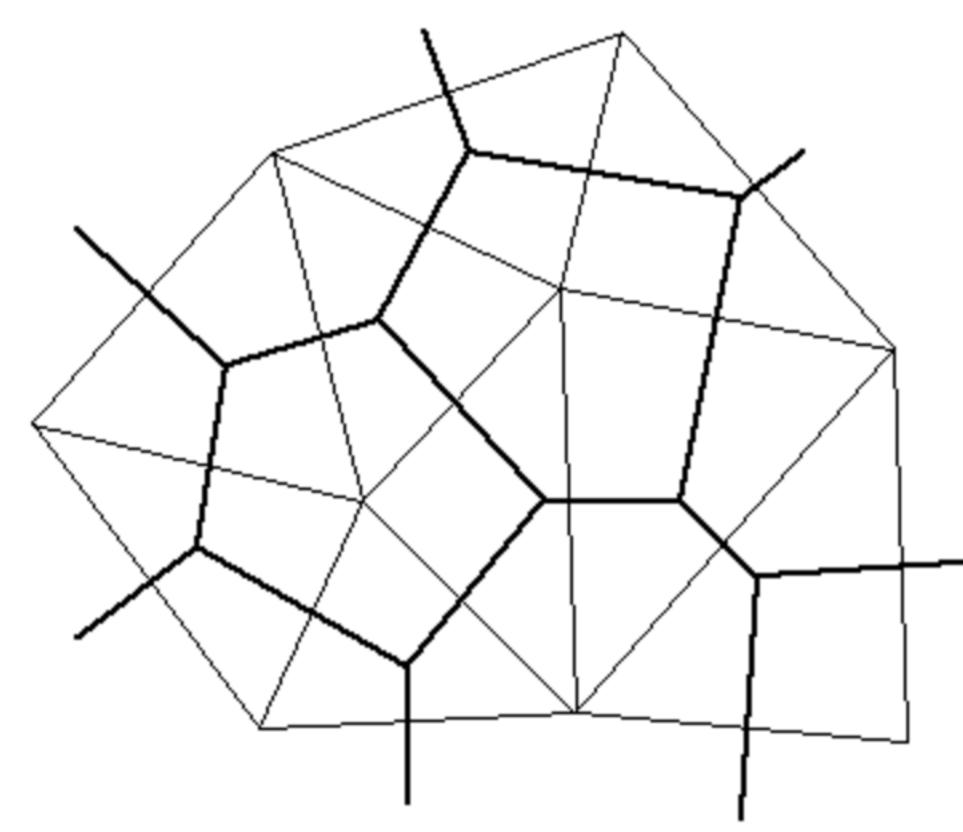
North & West

South & East

Constructing a Voronoi Diagram

Calculate a Delauney triangulation

Voronoi edges are perpendicular to triangle edges.



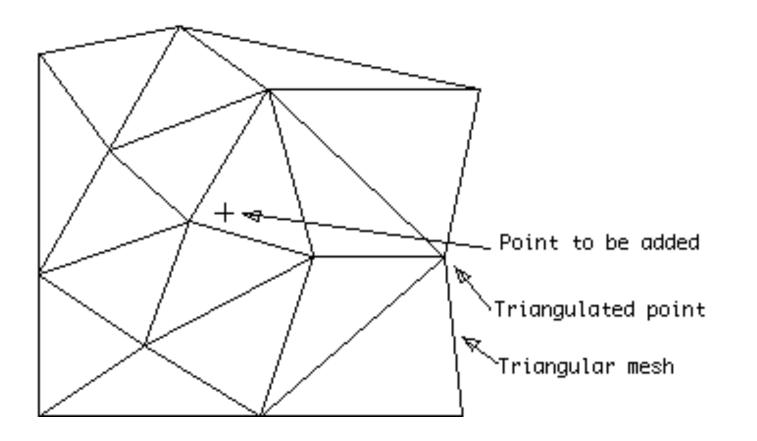
http://paulbourke.net/papers/triangulate/



Delauney Triangulation

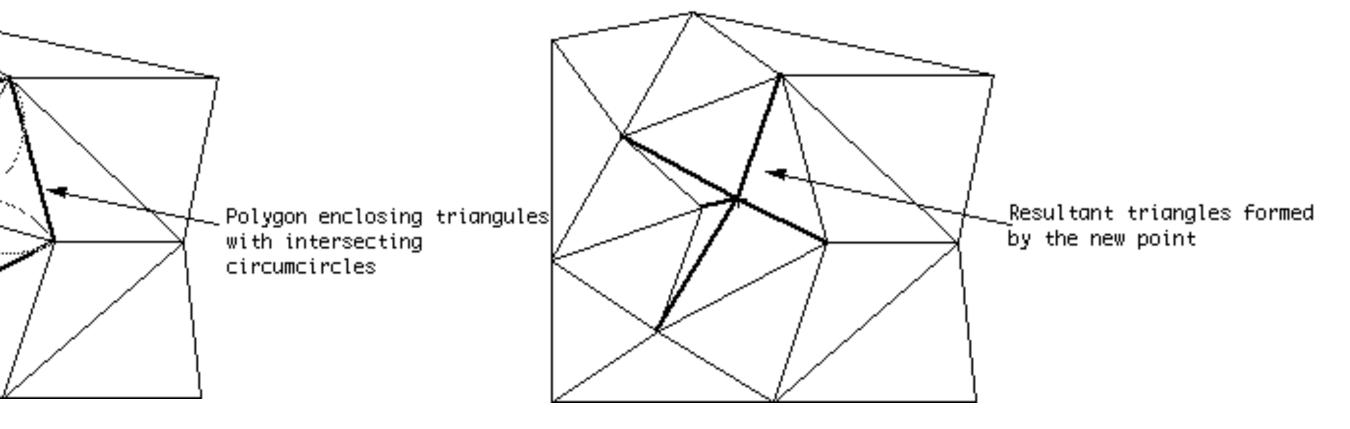
Start with all-encompassing fake triangle

For existing triangles: check if circumcircle contains new point





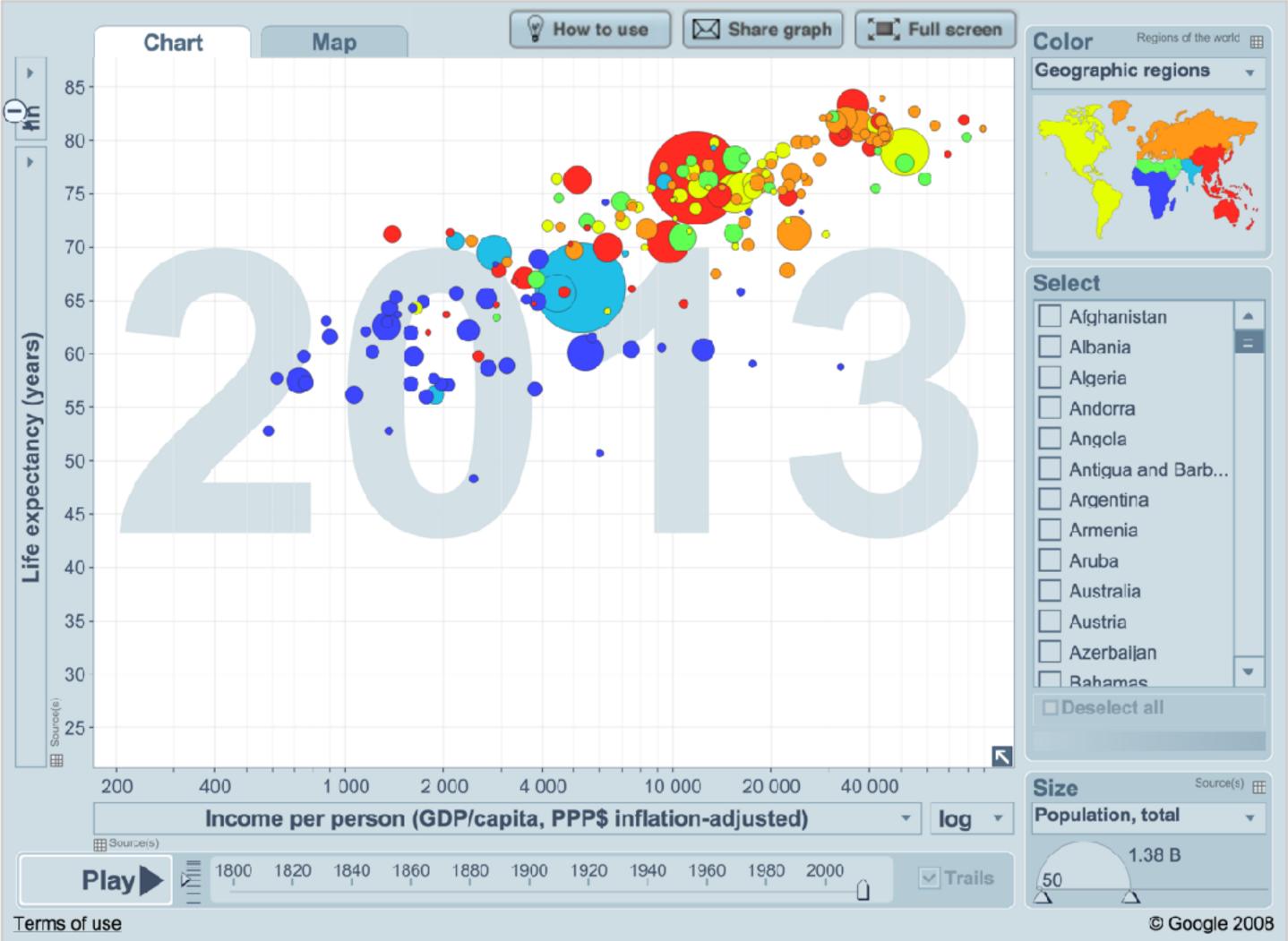
Outer edges of triangles form polygon, delete all inner edges Create triangle connecting all outer edges to new point.



Design Critique

GapMinder

https://goo.gl/Fcx28n Tool: http://goo.gl/jWNOUb



Attribute aggregation

- 1) group attributes and compute a similarity score across the set
- 2) dimensionality reduction, to preserve meaningful structure

Attribute aggregation

- 1) group attributes and compute a similarity score across the set
- 2) dimensionality reduction, to preserve meaningful structure

Clustering

- Classification of items into "similar" bins
- Based on similarity measures
 - Euclidean distance, Pearson correlation, ...
- Partitional Algorithms
 - divide data into set of bins
 - # bins either manually set (e.g., kmeans) or automatically determined (e.g., affinity propagation)

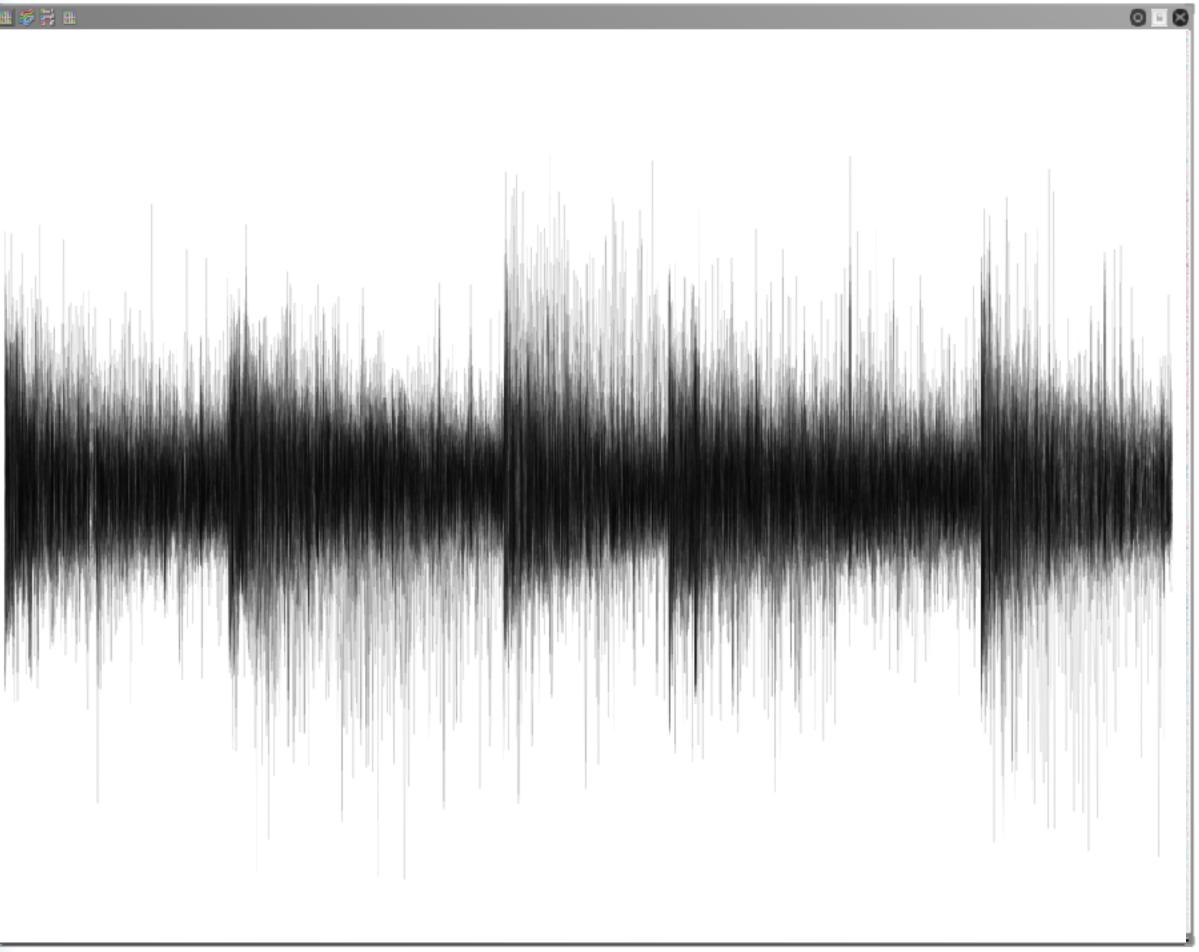
- Hierarchical Algorithms Produce "similarity tree" – dendrogram
- **Bi-Clustering**
- Clusters dimensions & records
- Fuzzy clustering
 - allows occurrence of elements in multiples clusters

Clustering Applications

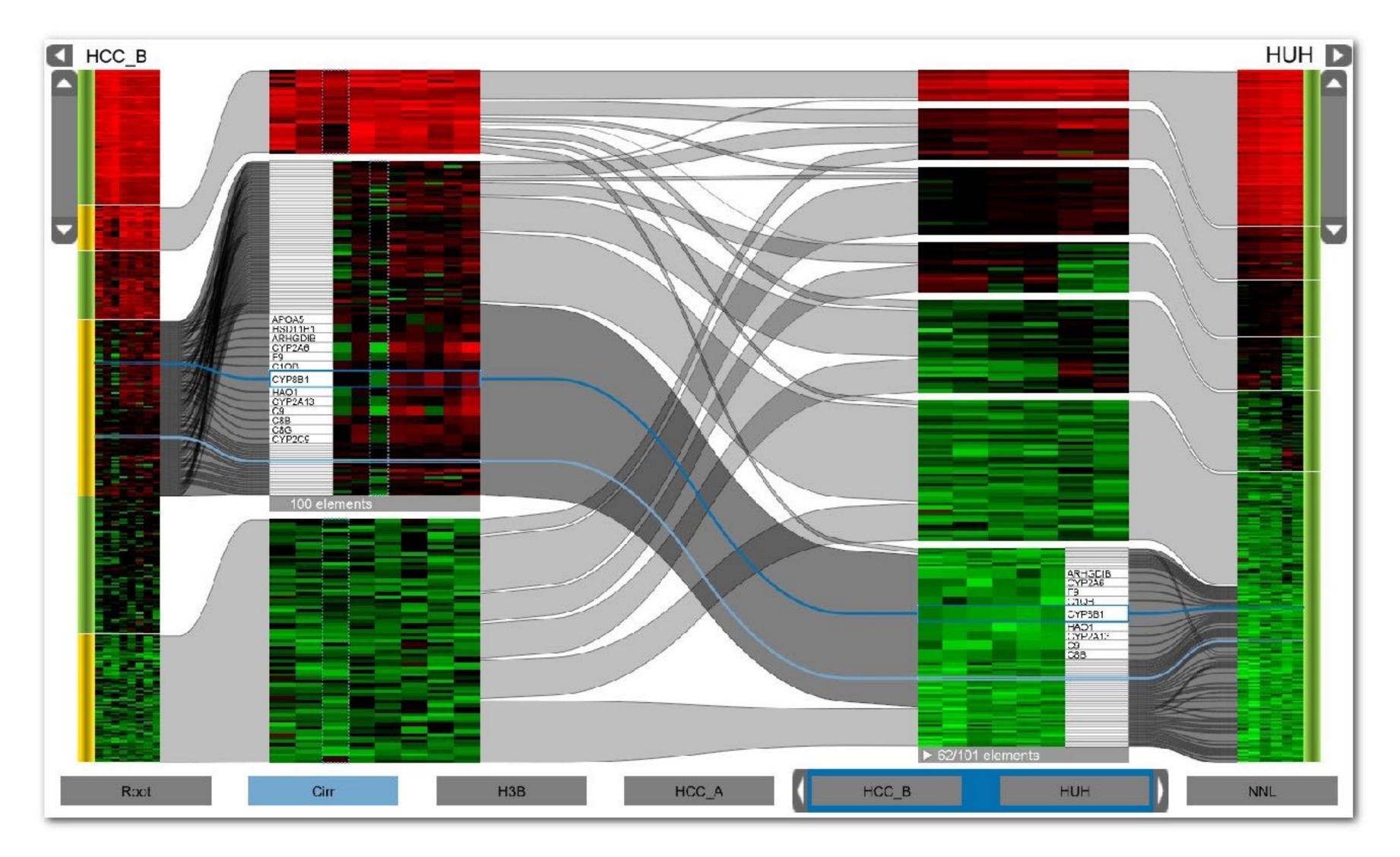
Clusters can be used to order (pixel based techniques) brush (geometric techniques) aggregate Aggregation cluster more homogeneous than whole dataset statistical measures, distributions, etc. more meaningful

Clustered Heat Map

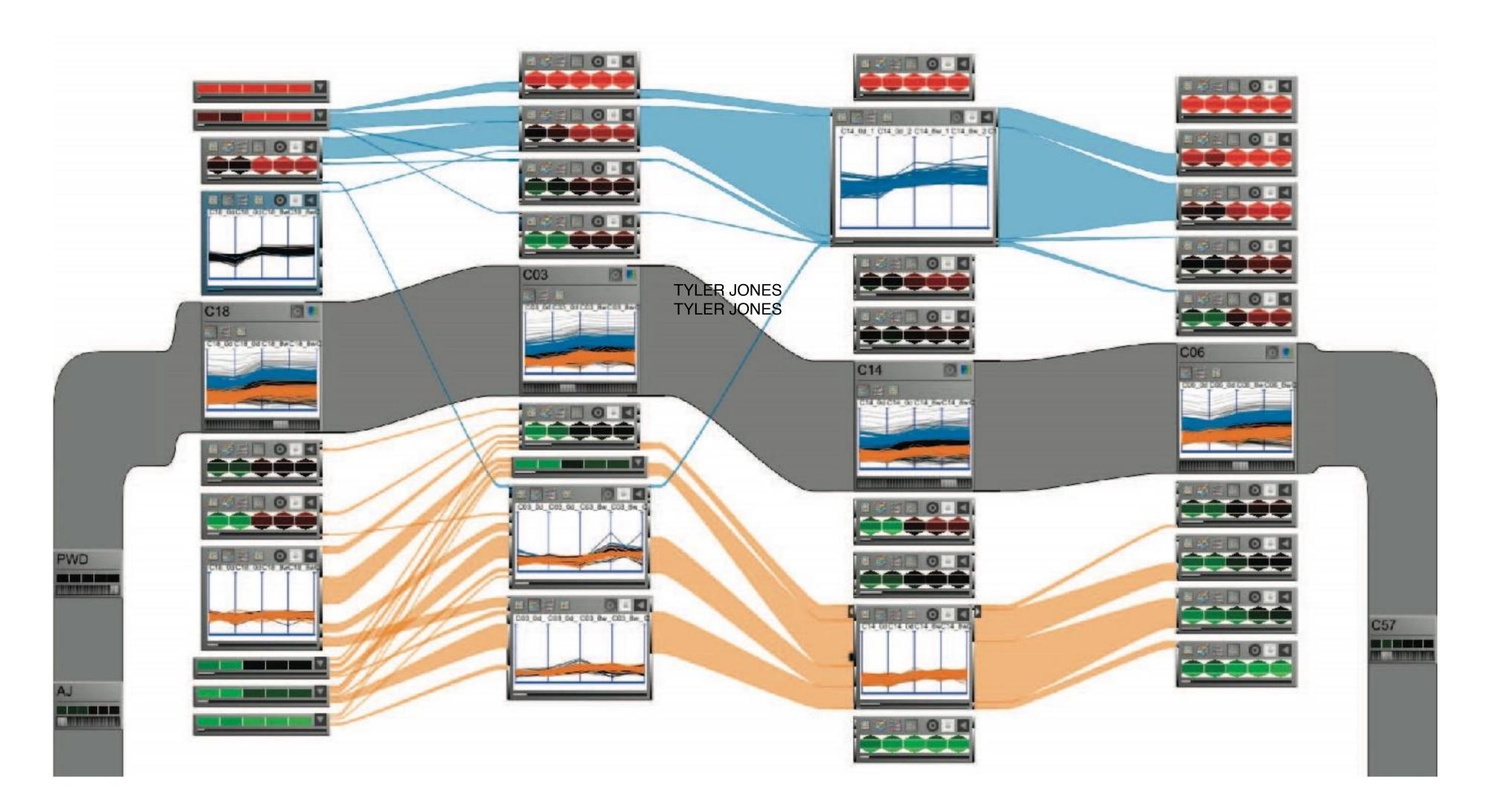
mRNA - Ground Truth/All g	
Neural	
Classical	
Mesenchymal	
Proneural	



Cluster Comparison

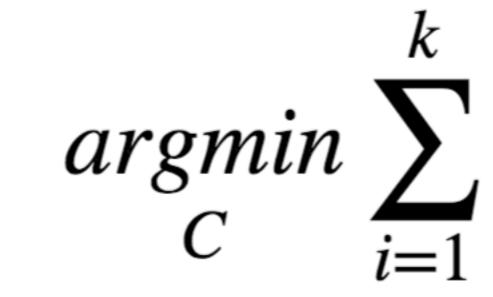


Aggregation



Example: K-Means

Goal: Minimize aggregate intra-custer distance (*inertia*)



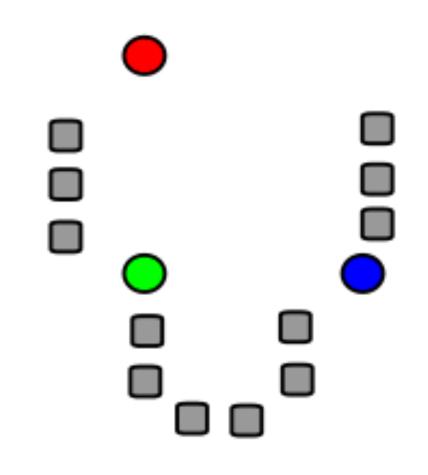
total squared distance from point to center of its cluster for euclidian distance: this is the variance measure of how internally coherent clusters are

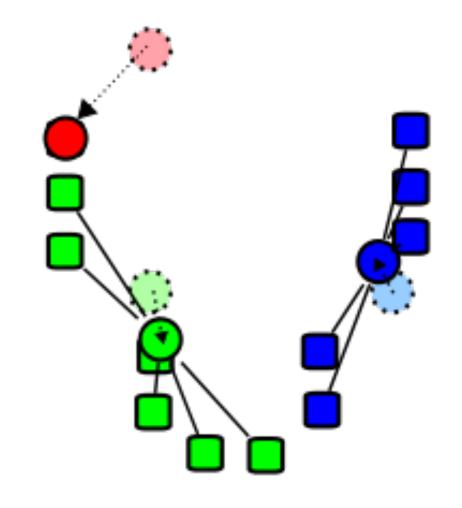
$$\sum_{x \in C_i} \|x - \mu_i\|^2$$

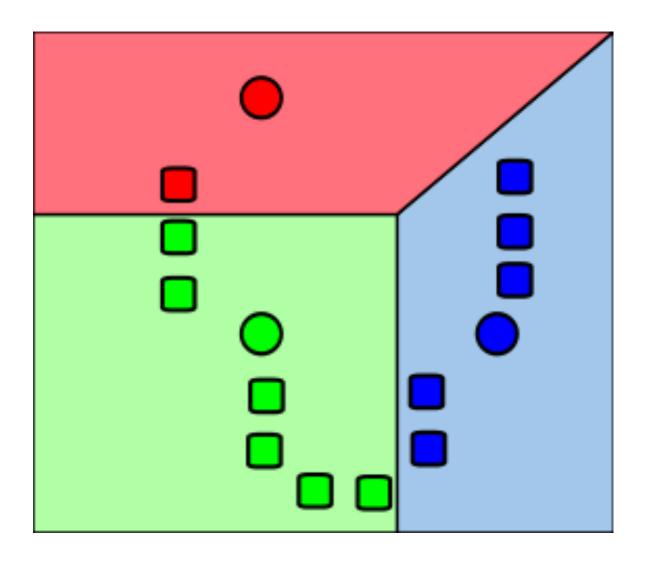
Lloyd's Algorithm

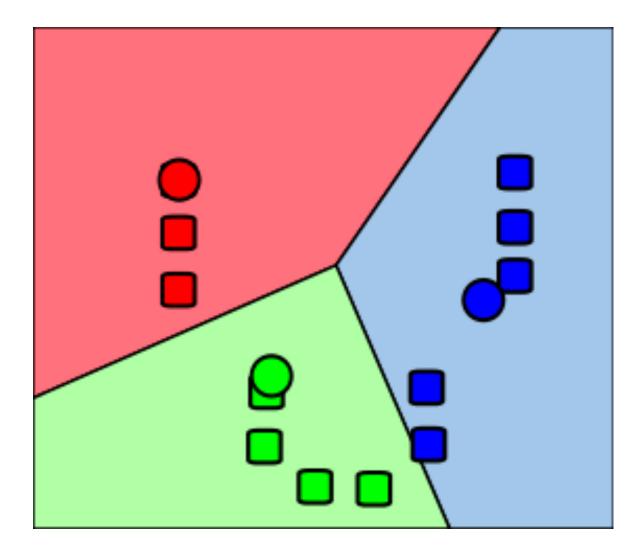
Input: set of records $x_1 \dots x_n$, and k (nr clusters) Pick k starting points as centroids $c_1 \dots c_k$ While not converged:

- 1. for each point x_i find closest centroid c_i
 - for every c_i calculate distance $D(x_i, c_i)$
 - assign x_i to cluster *j* defined by smallest distance
- 2. for each cluster *j*, compute a new centroid c_i by calculating the average of all x_i assigned to cluster j
- Repeat until convergence, e.g.,
 - no point has changed cluster
 - distance between old and new centroid below threshold
 - number of max iterations reached

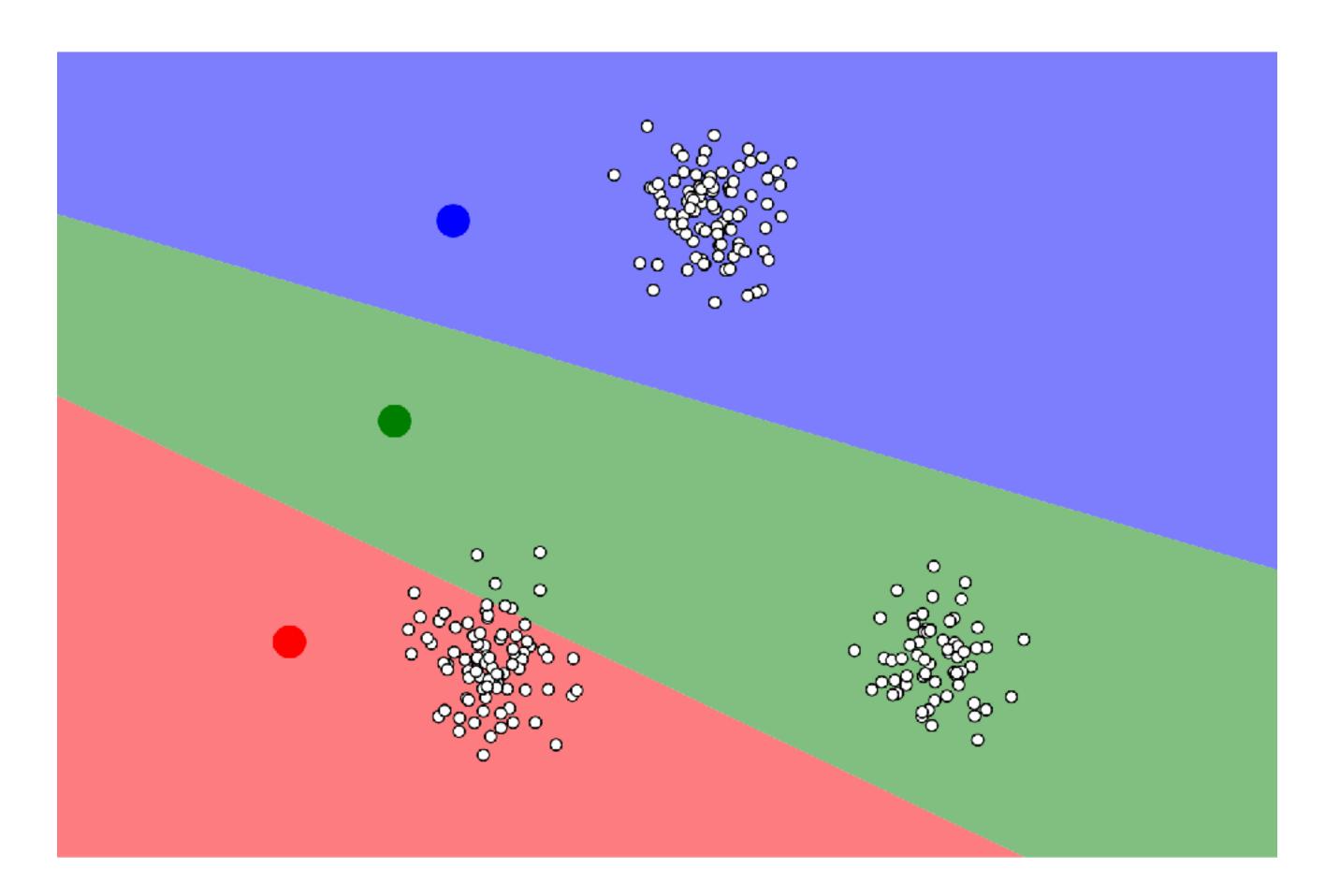






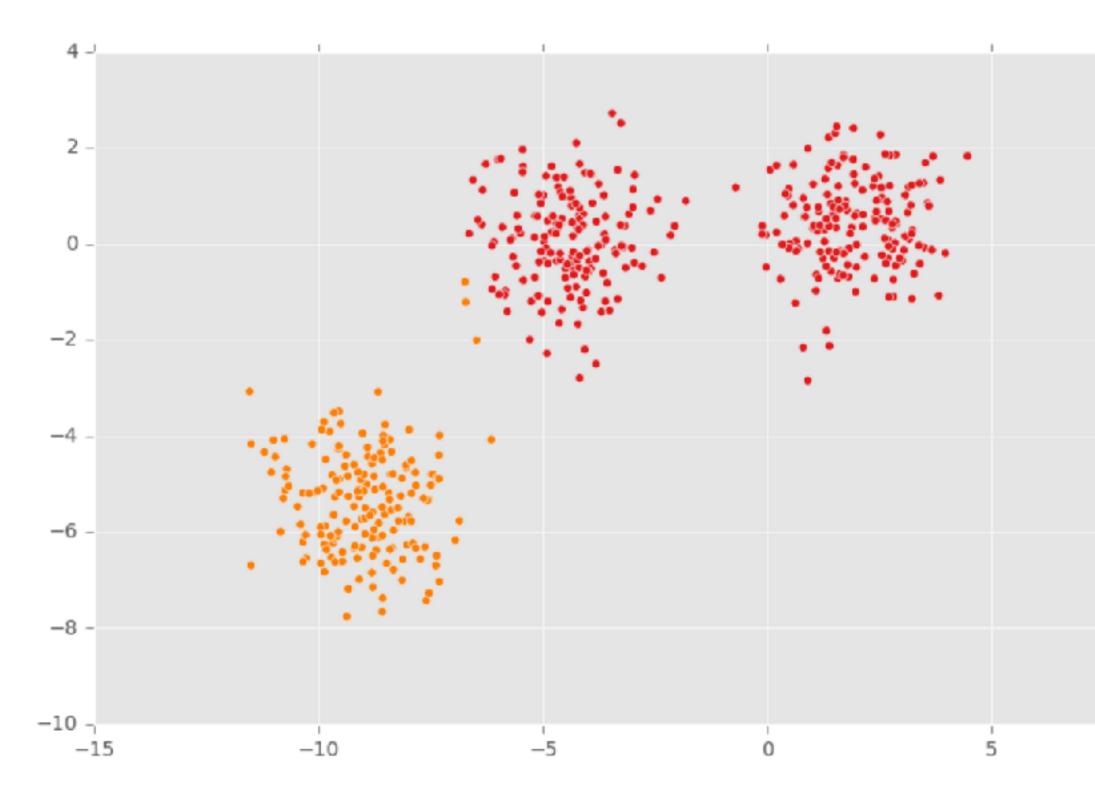


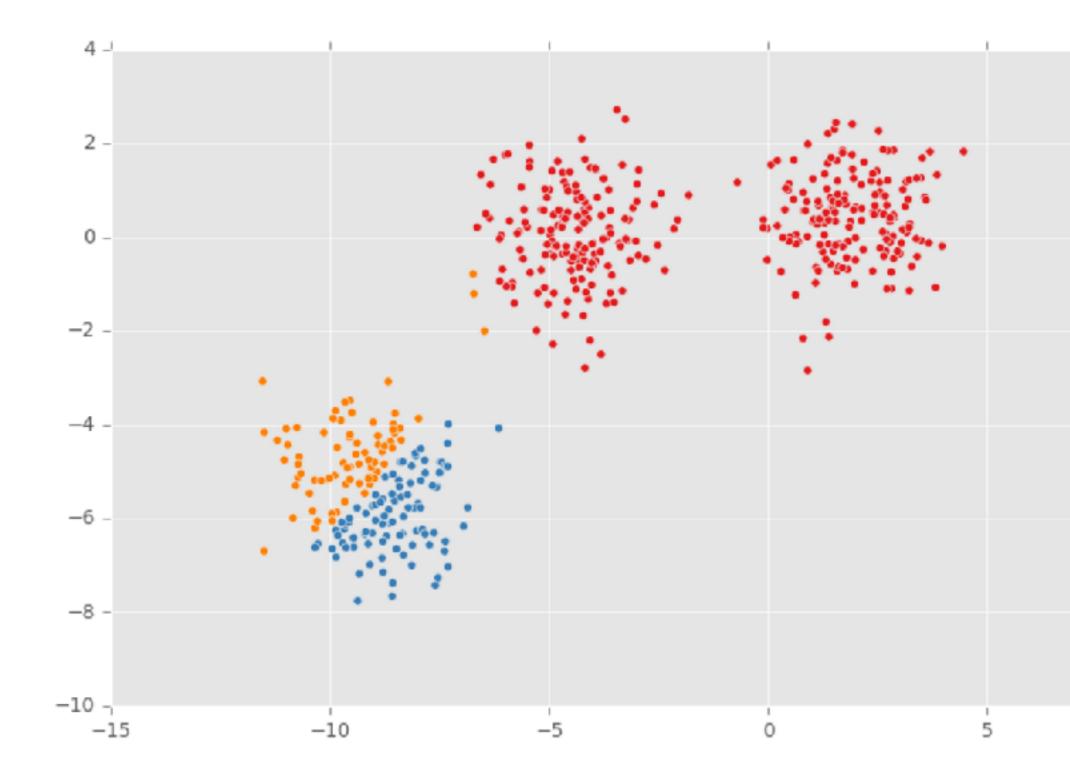
Illustrated



https://www.naftaliharris.com/blog/visualizing-k-means-clustering/

Choosing K





10

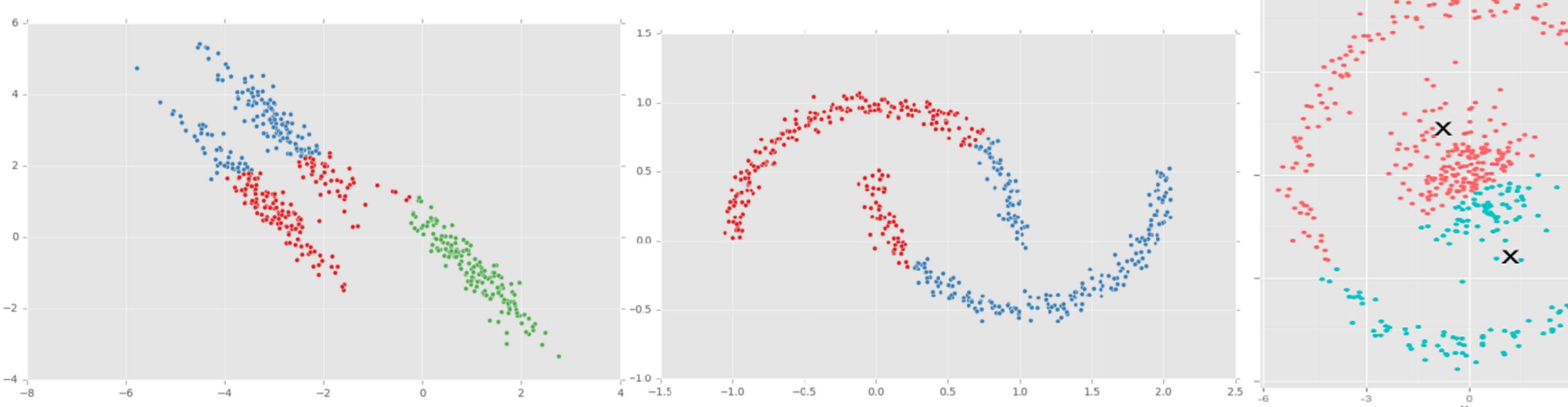


Properties

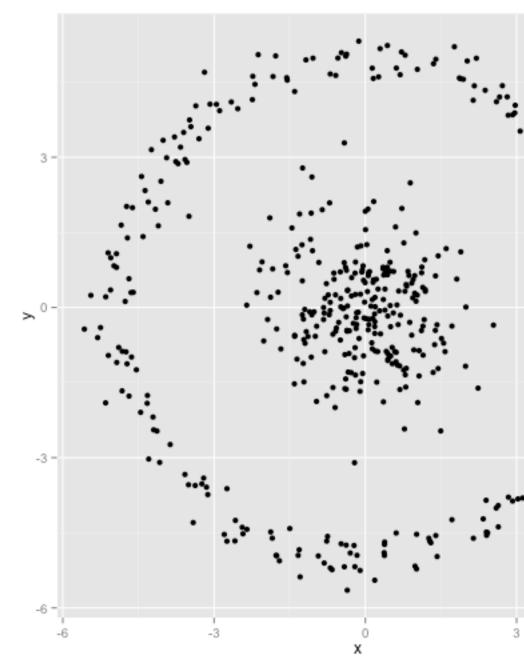
Lloyds algorithm doesn't find a global optimum Instead it finds a local optimum It is very fast: common to run multiple times and pick the solution with the minimum inertia

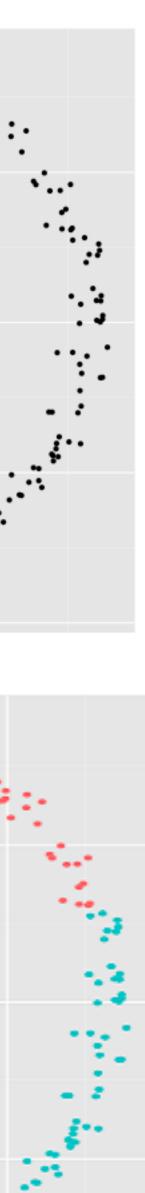
K-Means Properties

Assumptions about data: roughly "circular" clusters of equal size

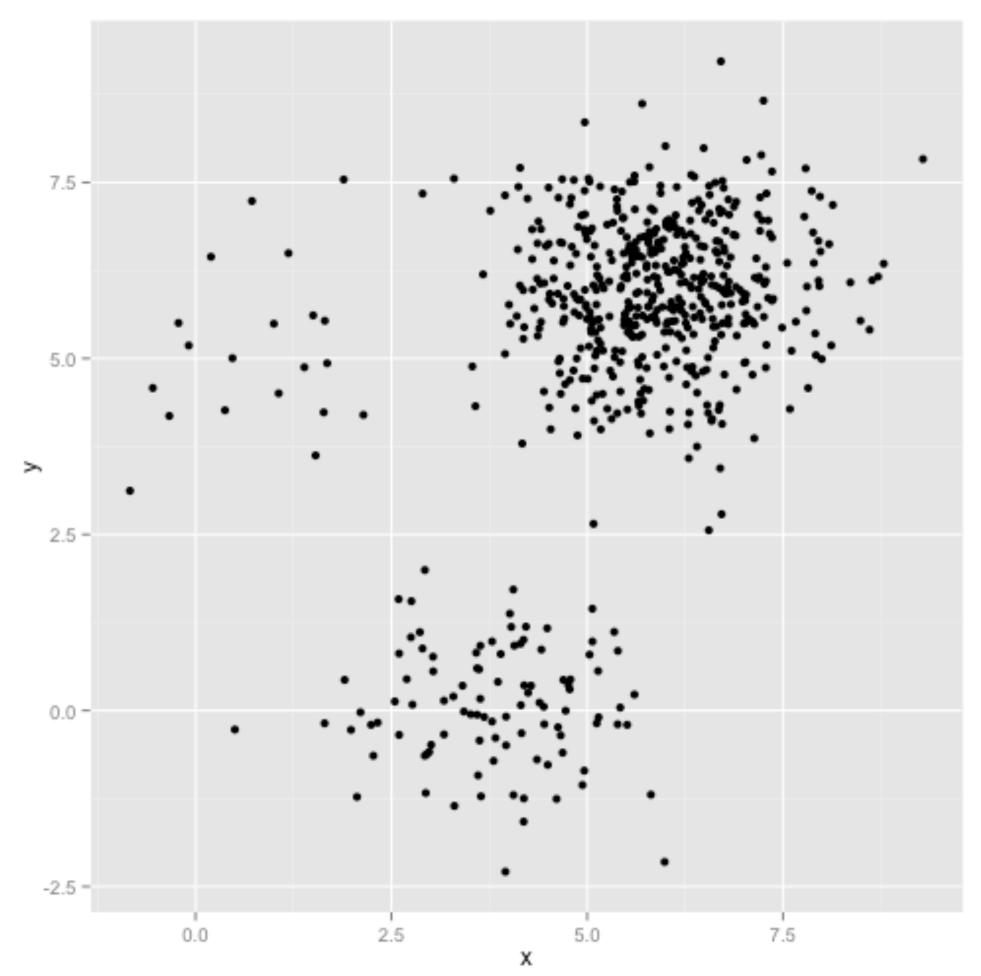




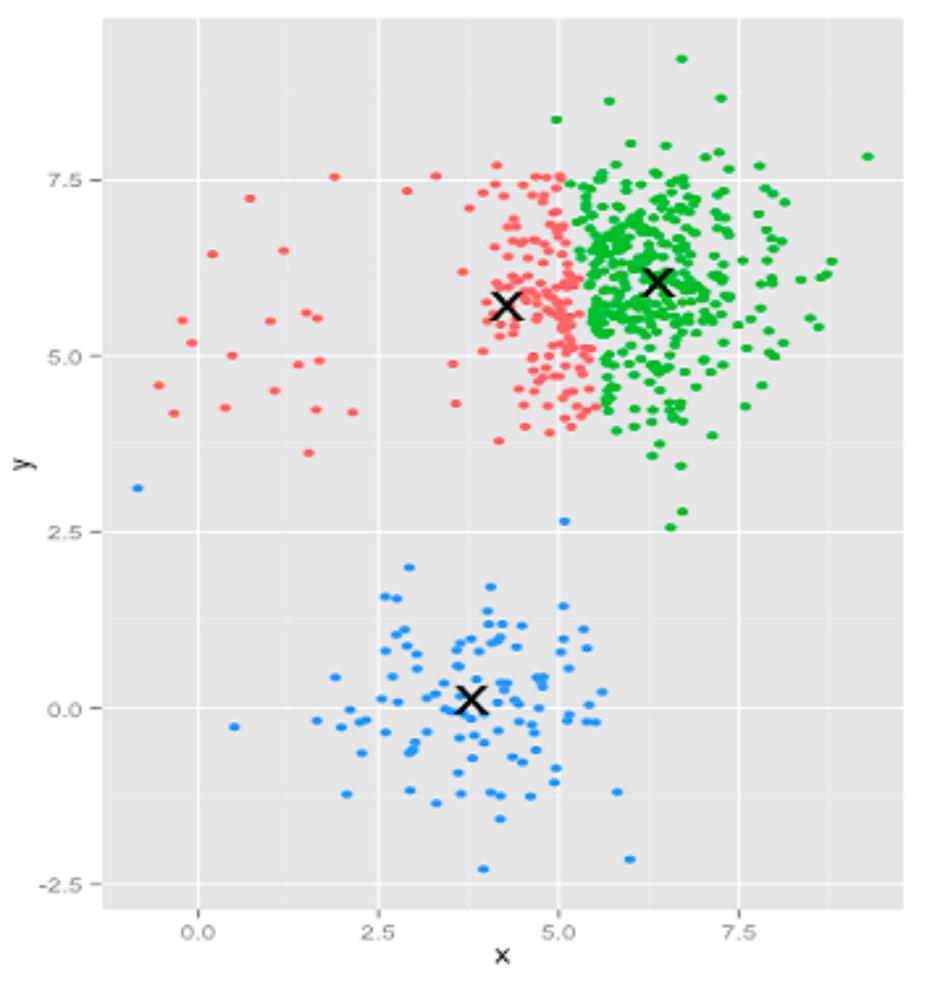




K-Means Unequal Cluster Size



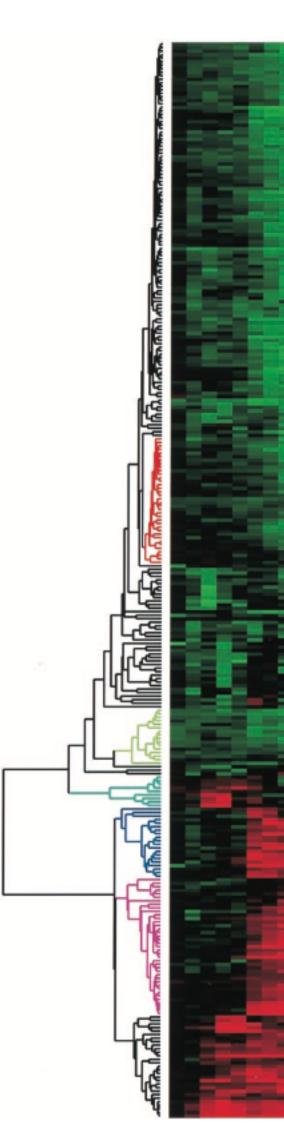
http://stats.stackexchange.com/questions/133656/how-to-understand-the-drawbacks-of-k-means

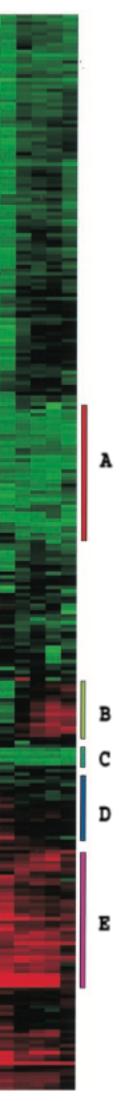


K-means assignments

Hierarchical Clustering

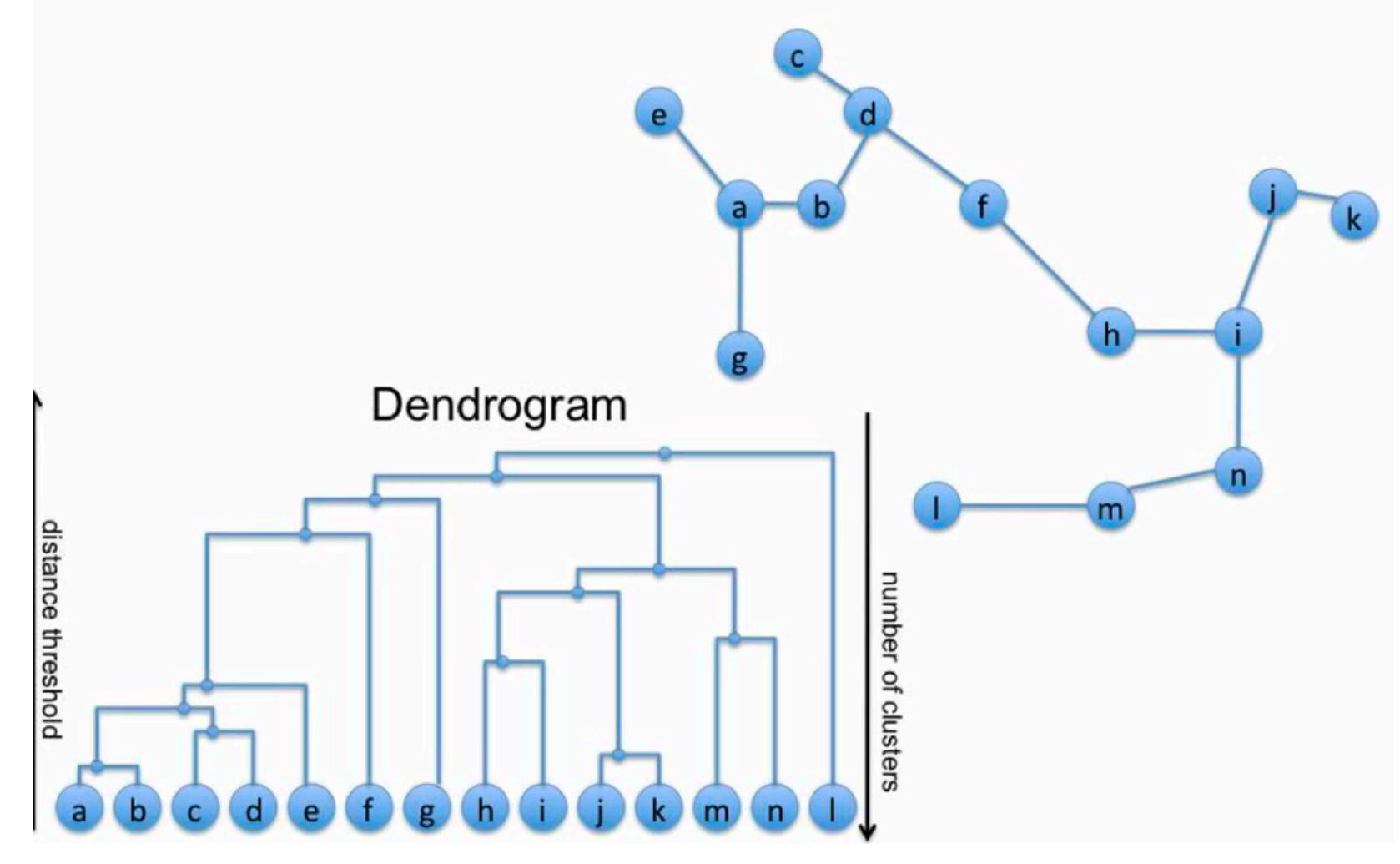
Two types: agglomerative clustering start with each node as a cluster and merge divisive clustering start with one cluster, and split





Agglomerative Clustering Idea

Agglomerative Clustering Idea



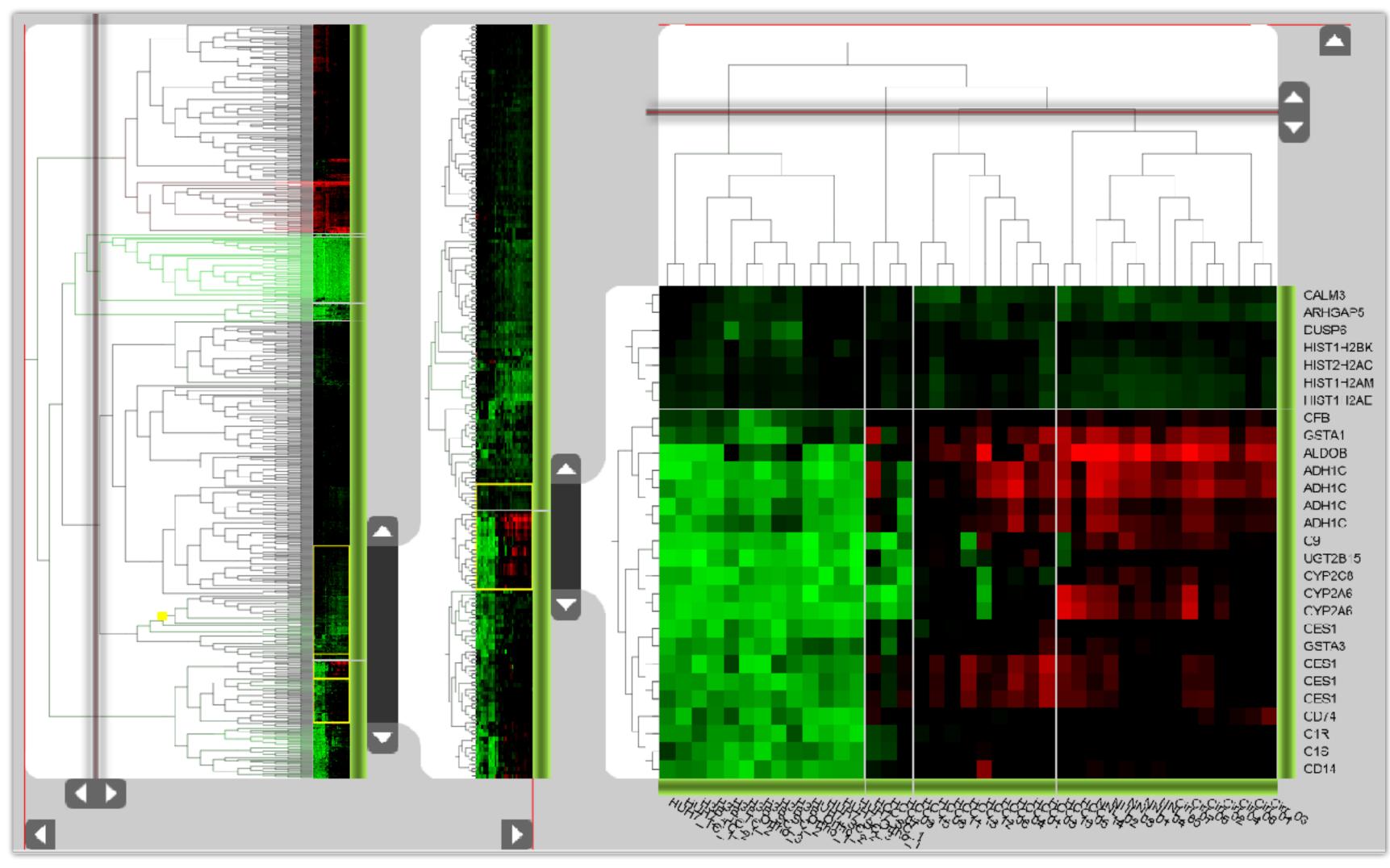
Linkage Criteria

How do you define similarity between two clusters to be merged (A and B)?

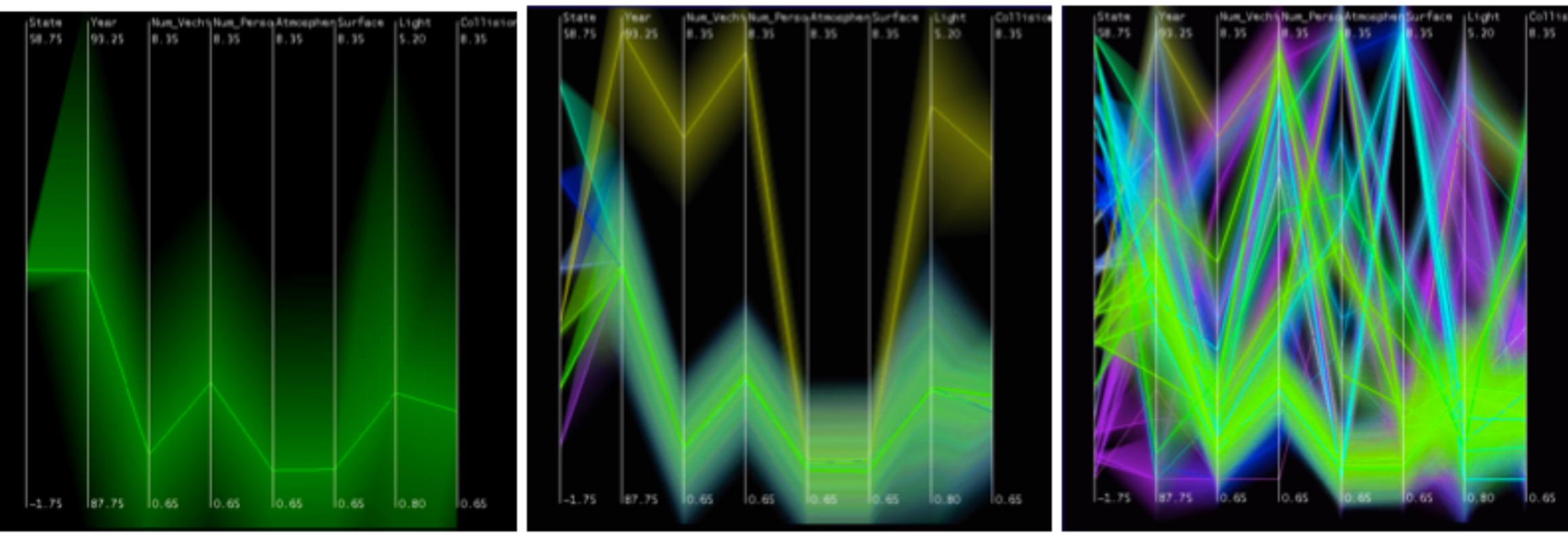
- use maximum linkage distance
- use minimum linkage distance
- use average linkage distance
- use centroid distance

Names	Formula
Maximum or complete-linkage clustering	$\max \left\{ d(a,b) : a \in A, b \in B ight\}.$
Minimum or single-linkage clustering	$\min\{d(a,b):a\in A,b\in B\}.$
Mean or average linkage clustering, or UPGMA	$\frac{1}{ A B }\sum_{a\in A}\sum_{b\in B}d(a,b).$
Centroid linkage clustering, or UPGMC	$\ c_s - c_t\ $ where c_s and c_t are the centroids of clusters s and t , respectively.

F+C Approach, with Dendrograms



Hierarchical Parallel Coordinates



Fua 1999

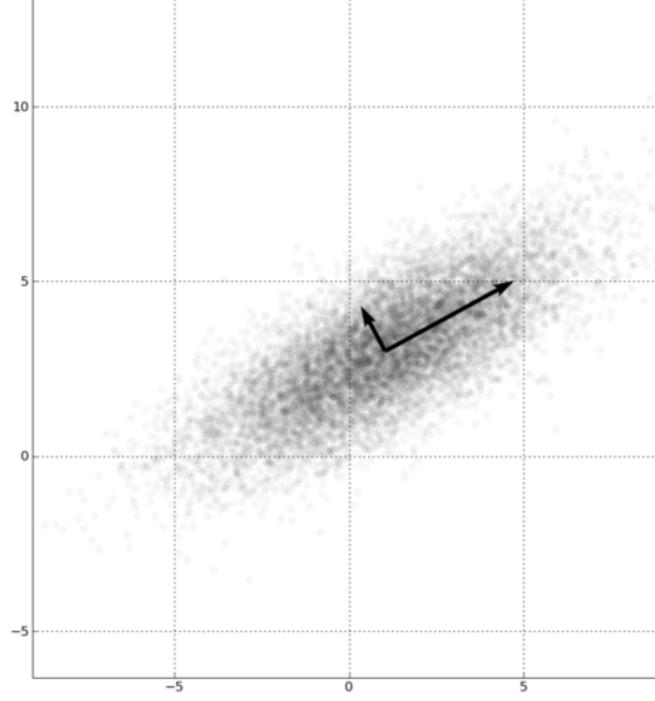


Attribute aggregation

- 1) group attributes and compute a similarity score across the set
- 2) dimensionality reduction, to preserve meaningful structure

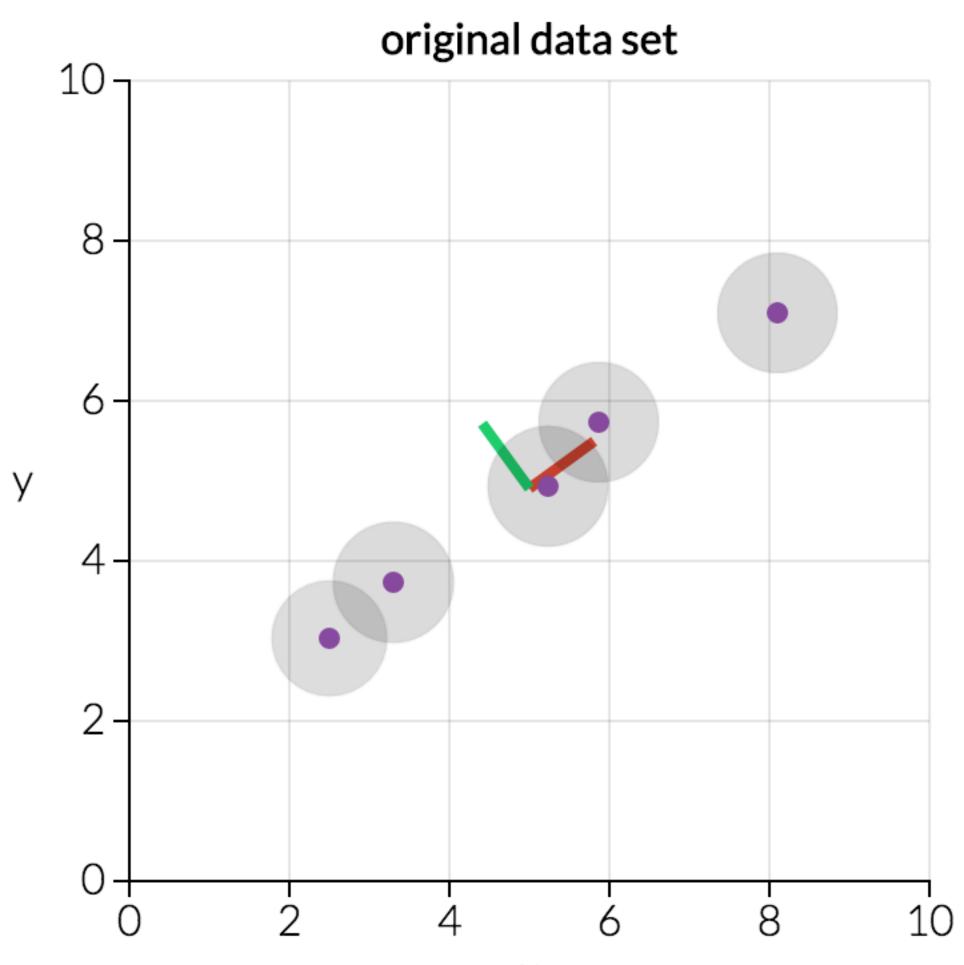
Dimensionality Reduction

- Reduce high dimensional to lower dimensional space
- Preserve as much of variation as possible
- Plot lower dimensional space Principal Component Analysis (PCA)
 - linear mapping, by order of variance

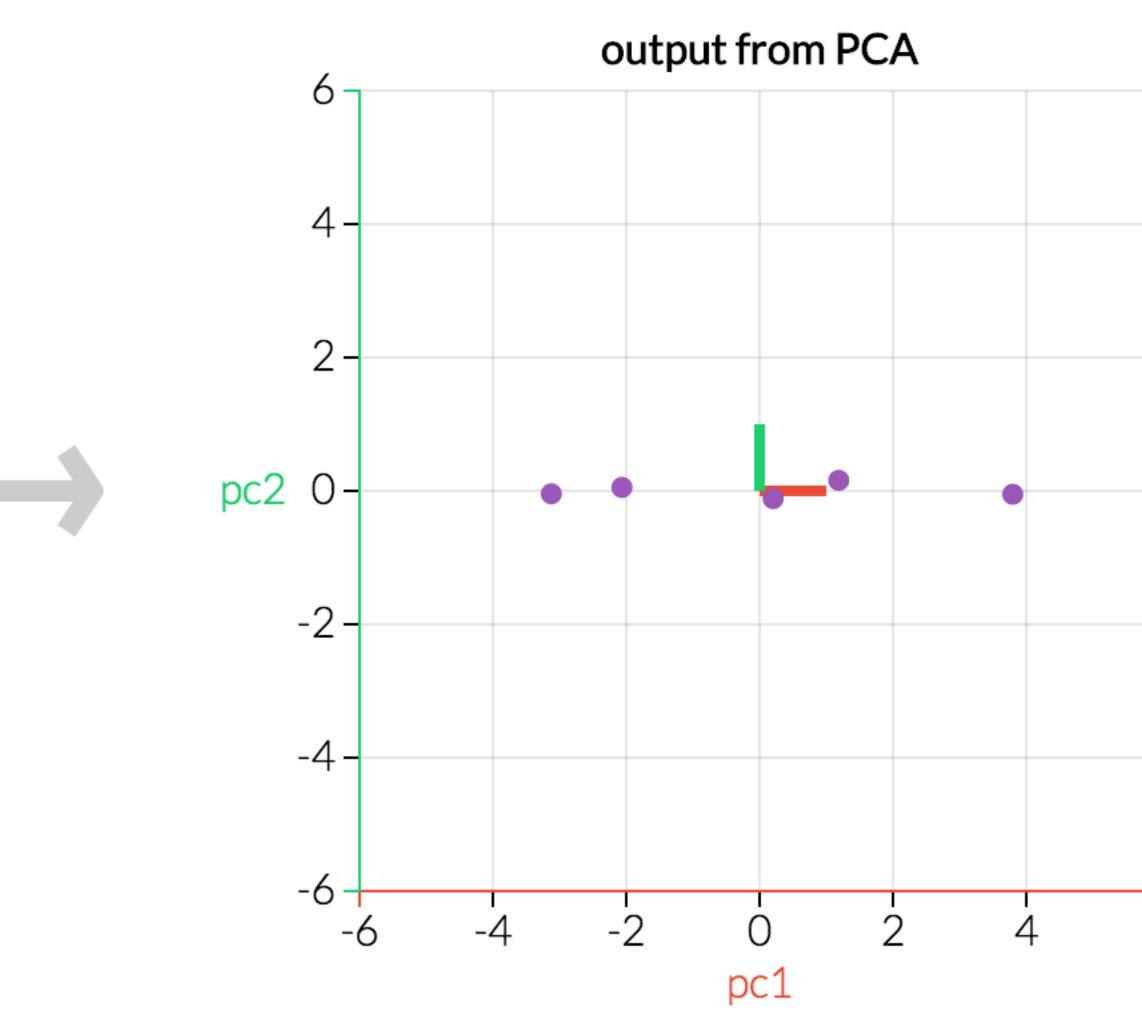






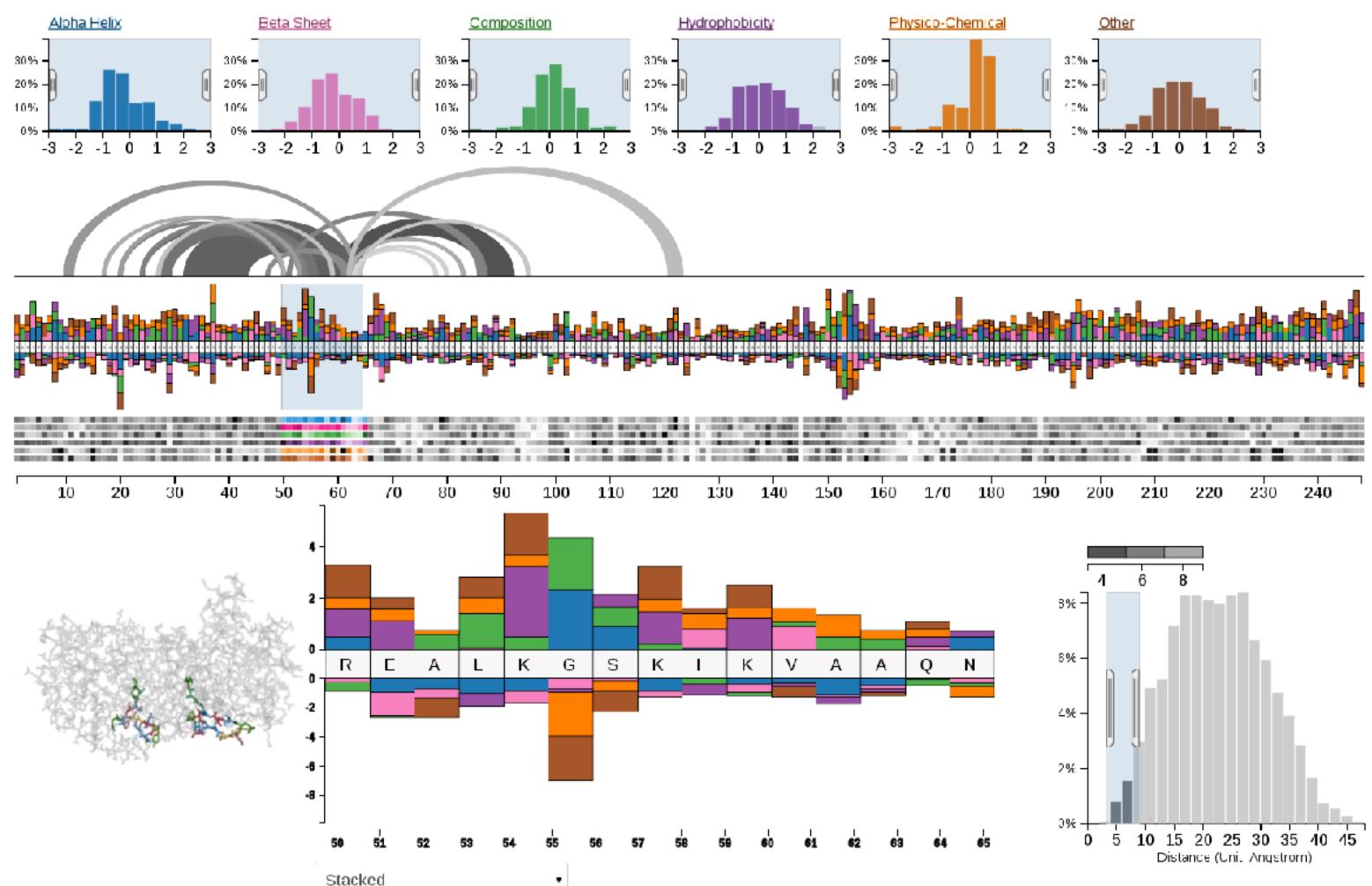


Х





PCA Example – Class Project 2013



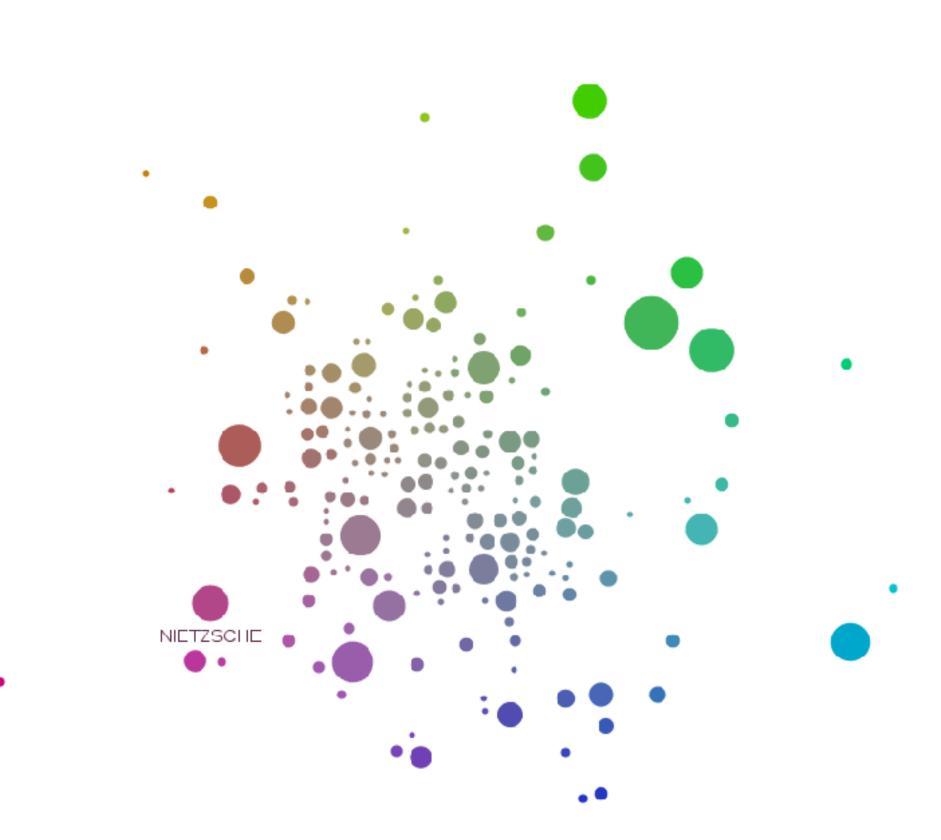
http://mu-8.com/

[Mercer & Pandian]

Multidimensional Scaling

- Nonlinear, better suited for some DS
- Multiple approaches
- Works based on projecting a similarity matrix
 - How do you compute similarity?
 - How do you project the points?
- Popular for text analysis

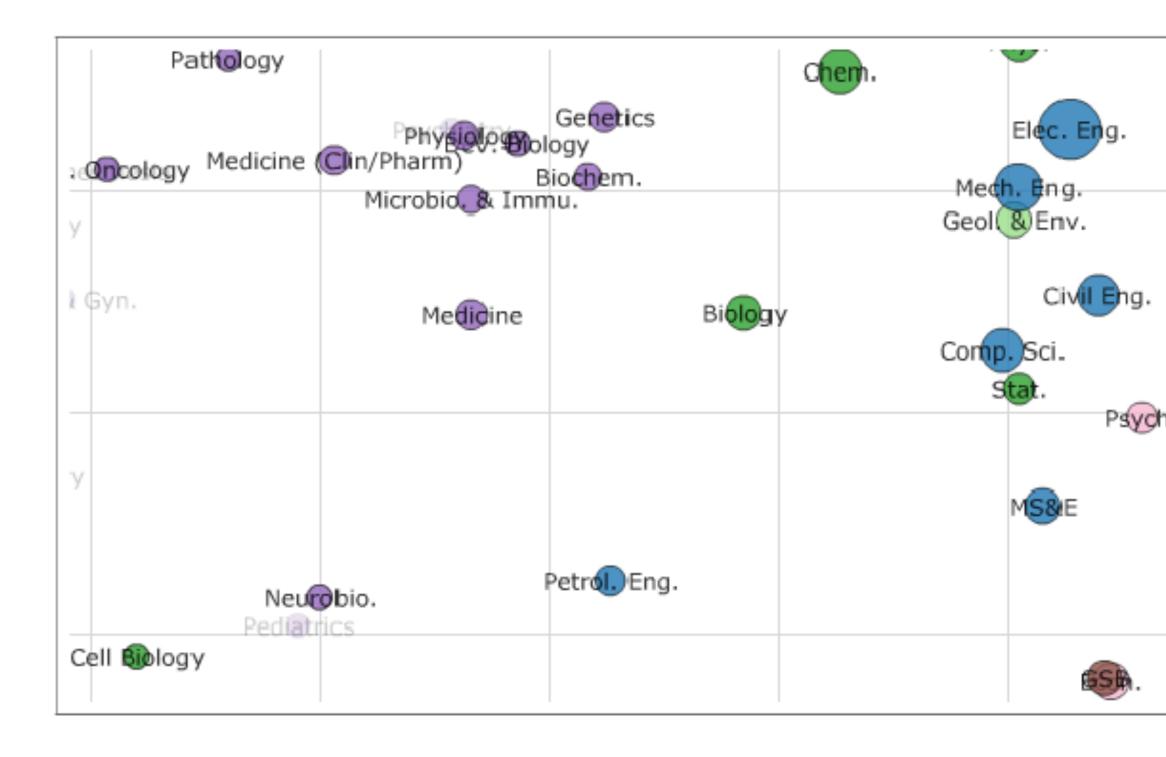




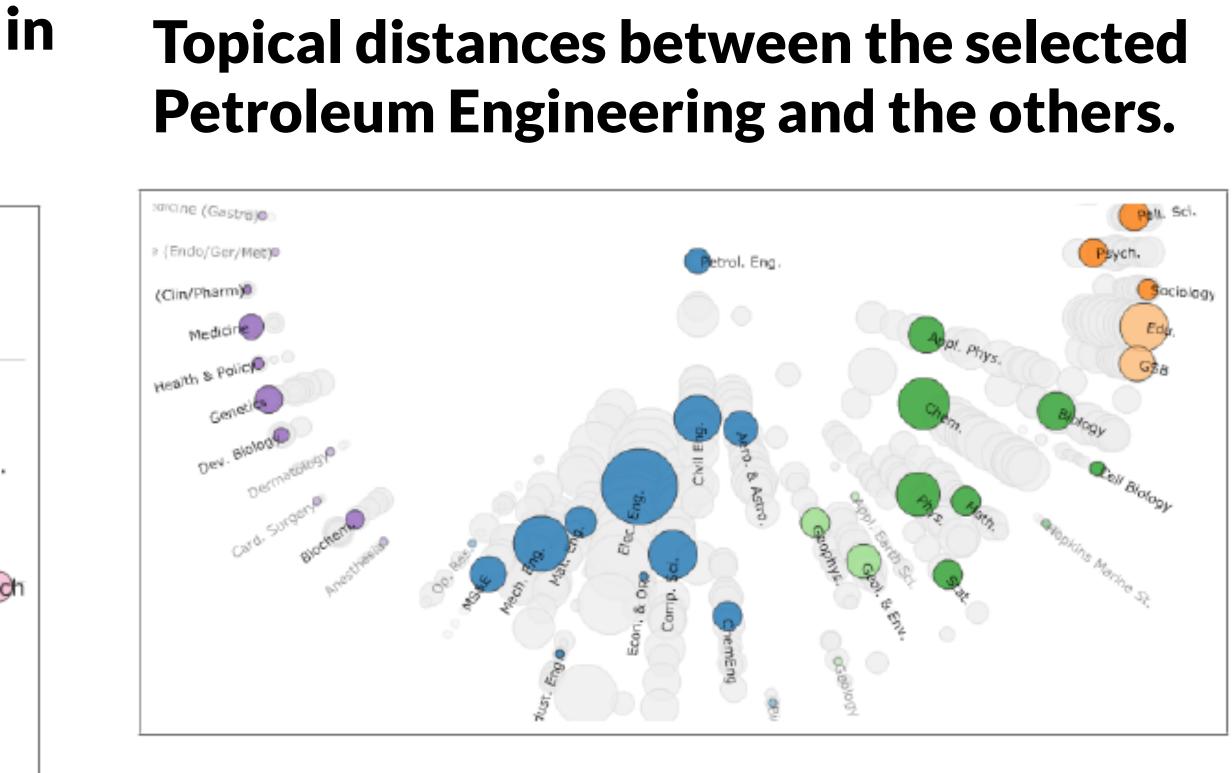
[Doerk 2011]

Can we Trust Dimensionality Reduction?

Topical distances between departments in a 2D projection



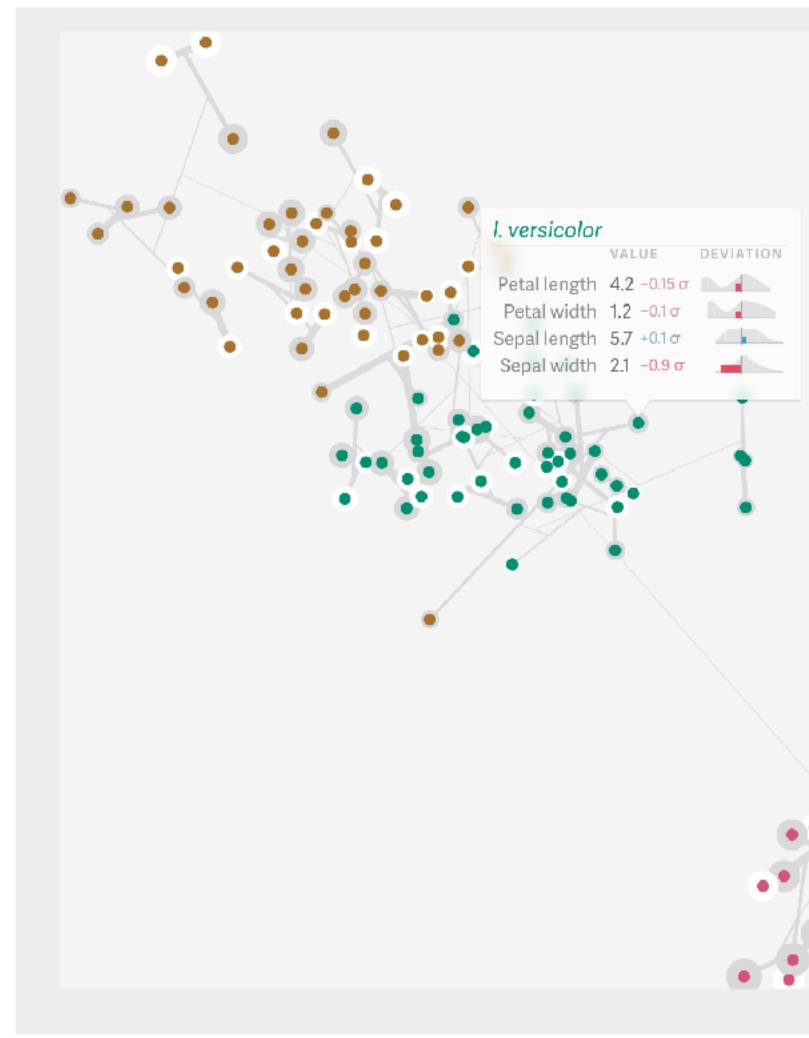
http://www-nlp.stanford.edu/projects/dissertations/browser.html



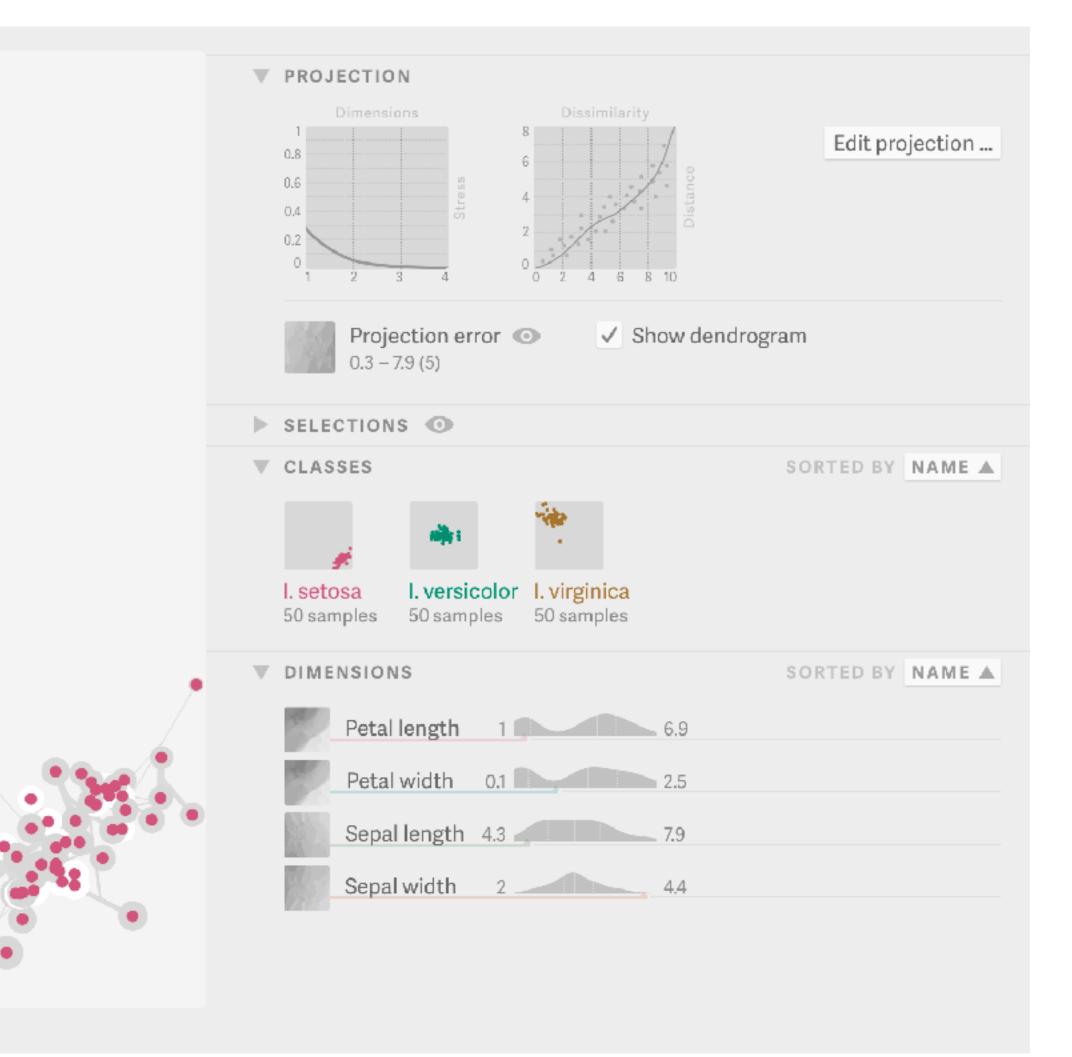
[Chuang et al., 2012]



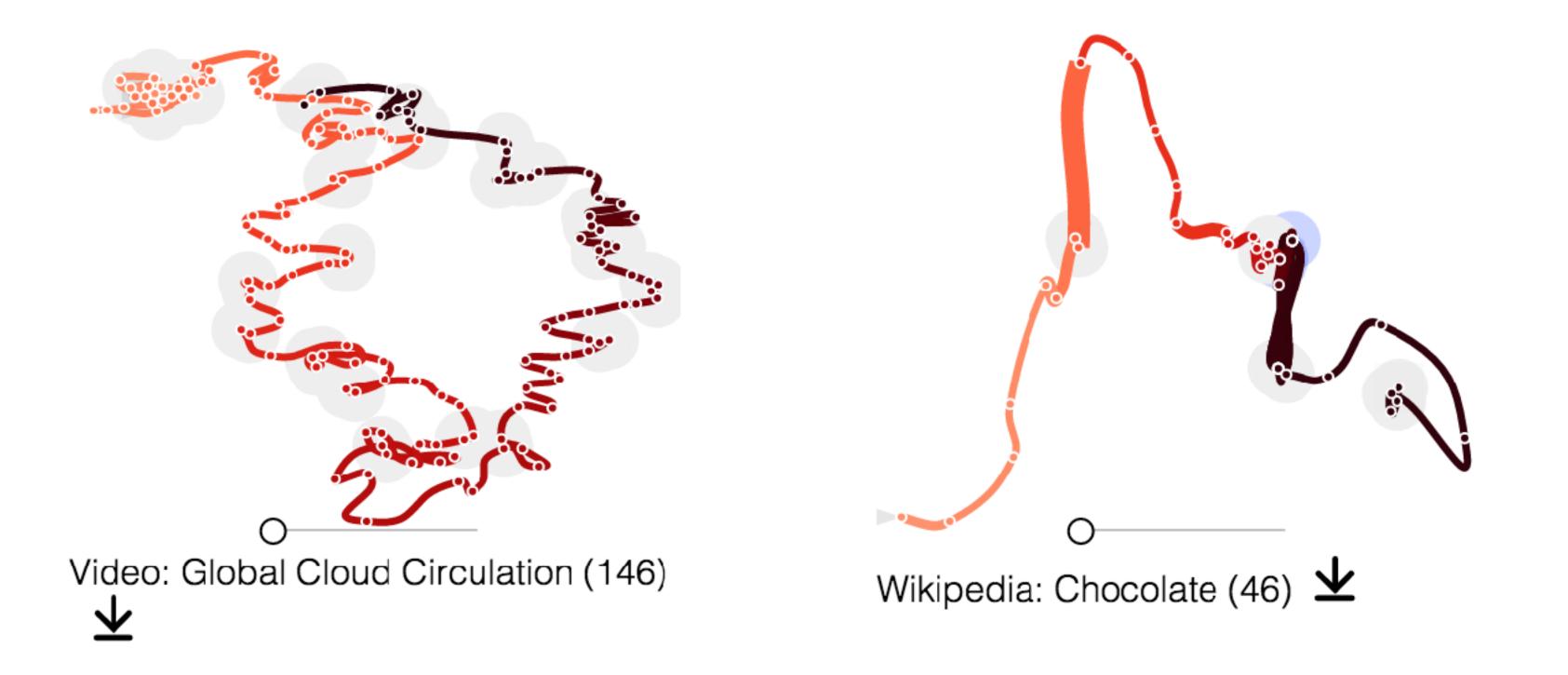
Probing Projections



http://julianstahnke.com/probing-projections/



MDS for Temporal Data: TimeCurves



http://aviz.fr/~bbach/timecurves/



