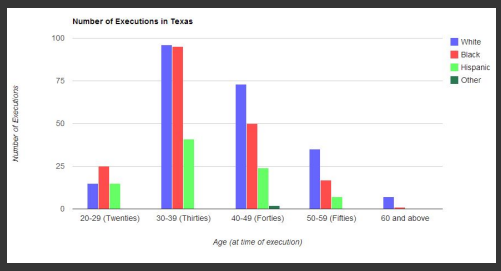
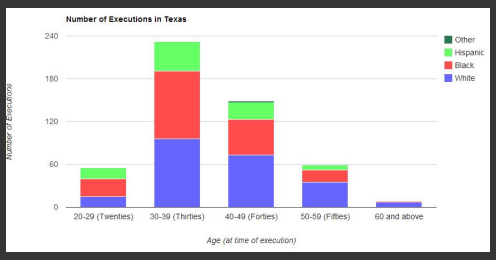
Exploratory Data Analysis

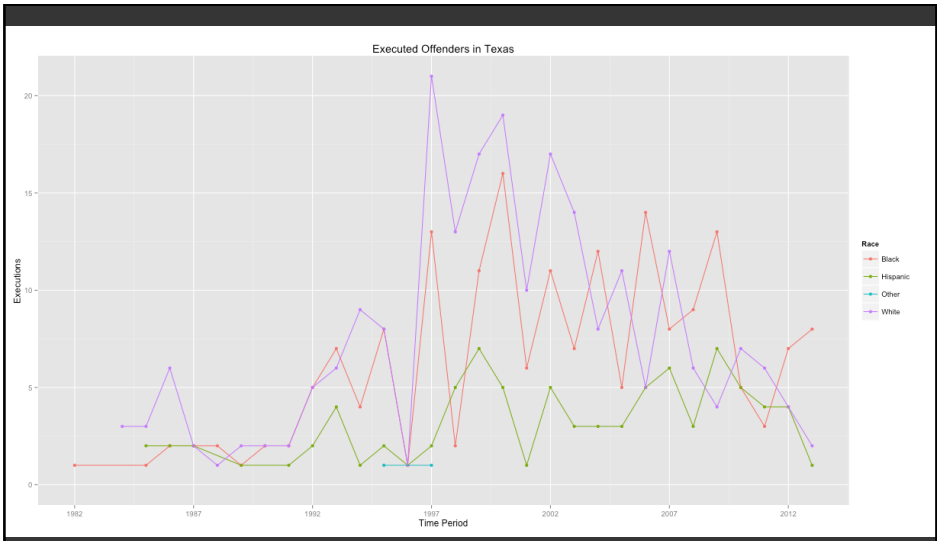
Maneesh Agrawala

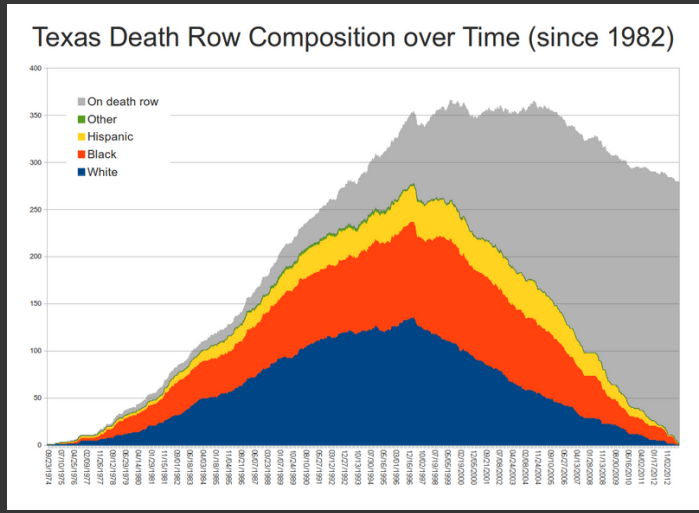
CS 294-10: Visualization
Fall 2013

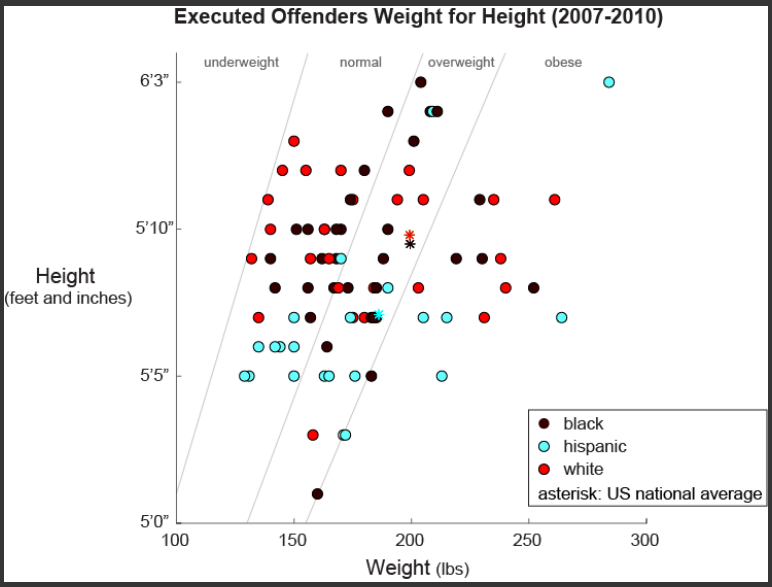
Last Time: Visualization Designs

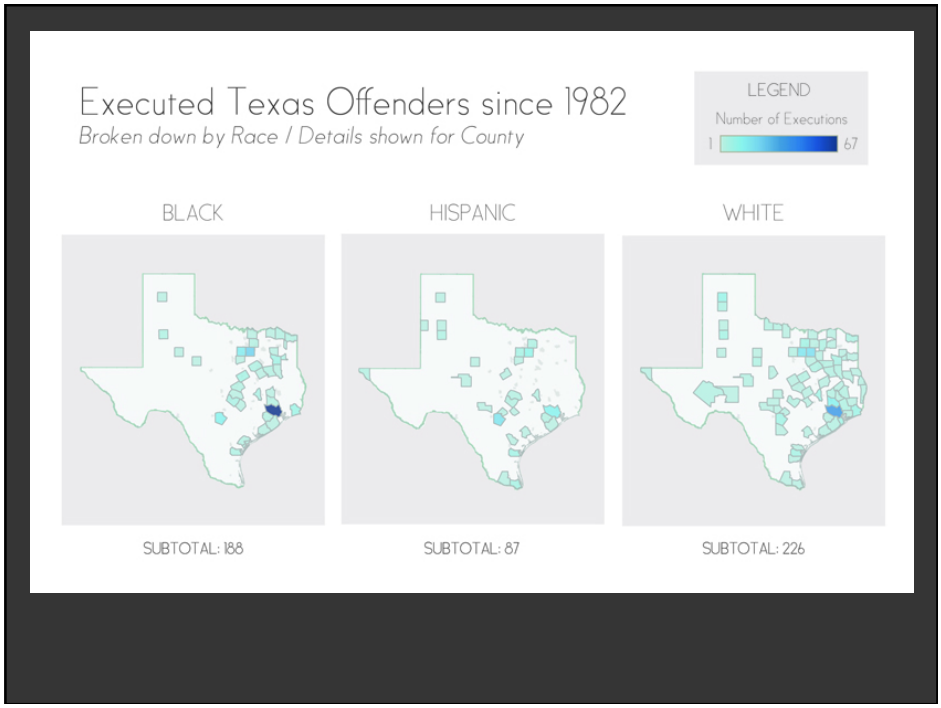

Number of Executions in Texas
100
75
50
25
0
Number of Executions
White
Black
Hispanic
Other
20-29 (Twenties)
30-39 (Thirties)
40-49 (Forties)
50-59 (Fifties)
60 and above
Age (at time of execution)

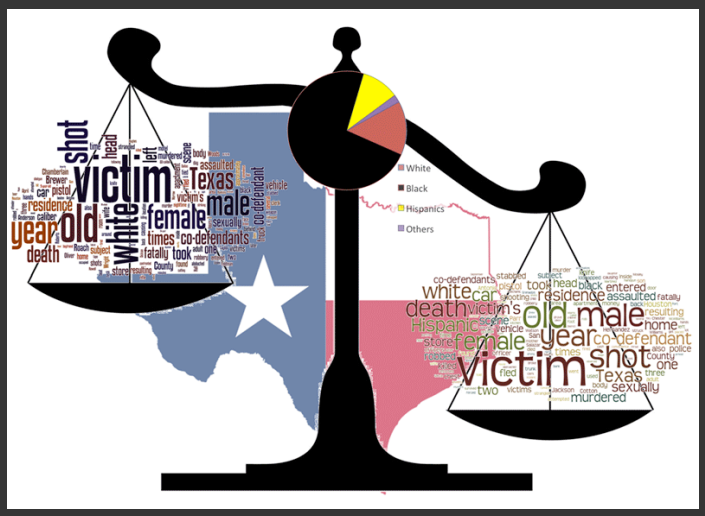

Number of Executions in Texas
240
180
120
60
0
Number of Executions
Other
Hispanic
Black
White
20-29 (Twenties)
30-39 (Thirties)
40-49 (Forties)
50-59 (Fifties)
60 and above
Age (at time of execution)


Executed Offenders in Texas
20
15
10
5
0
Executions
1982
1987
1992
1997
2002
2007
2012
Time Period
Race
Black
Hispanic
Other
White


Texas Death Row Composition over Time (since 1982)
On death row
Other
Hispanic
Black
White


Executed Offenders Weight for Height (2007-2010)
underweight
normal
overweight
obese
6'3"
5'10"
5'5"
5'0"
Height
(feet and inches)
100
150
200
250
300
Weight (lbs)
black
hispanic
white
asterisk: US national average


Executed Texas Offenders since 1982
Broken down by Race / Details shown for County
LEGEND
Number of Executions
1
67
BLACK
HISPANIC
WHITE
SUBTOTAL: 188
SUBTOTAL: 87
SUBTOTAL: 226


White
Black
Hispanics
Others
victim
shot
Texas
male
female
old
white
year
death
Hispanic
car
residence
assaulted
co-defendant
sexually
murdered

Design Considerations

Title, labels, legend, captions, source!

Expressiveness and Effectiveness

- Avoid unexpressive marks (lines? bars? gradients?)
- Use perceptually effective encodings
- Don't distract: faint gridlines, pastel highlights/fills
- The "elimination diet" approach – start minimal

Support comparison and pattern perception

- Between elements, to a reference line, or to counts

Design Considerations

Group / sort data by meaningful dimensions

Transform data (e.g., invert, log, normalize)

- Are model choices (regression lines) appropriate?

Reduce cognitive overhead

- Minimize visual search, minimize ambiguity
  - Avoid legend lookups if direct labeling works
  - Avoid color mappings with indiscernible colors

Be consistent! Visual inferences should consistently support data inferences

In-Class Review

Procedure

Break into groups of 4 (assigned by me)
Appoint a time keeper
Take turns showing your visualization – present findings (~3 min each)
Then critique – rubric on next slide (~5 min each)

- Get feedback from everyone in group
- Author must take notes

Post writeup to assignment 1 page after class

- Include feedback
- Briefly describe how you would re-design the visualization

Write-up of critique will be used in grading

In-Class Review Rubric

Expressiveness

- Prioritizes important information / Avoids false inferences
- Consistent visual mappings (e.g., respect color mappings)
- Make encodings *meaningful* rather than arbitrary

Effectiveness

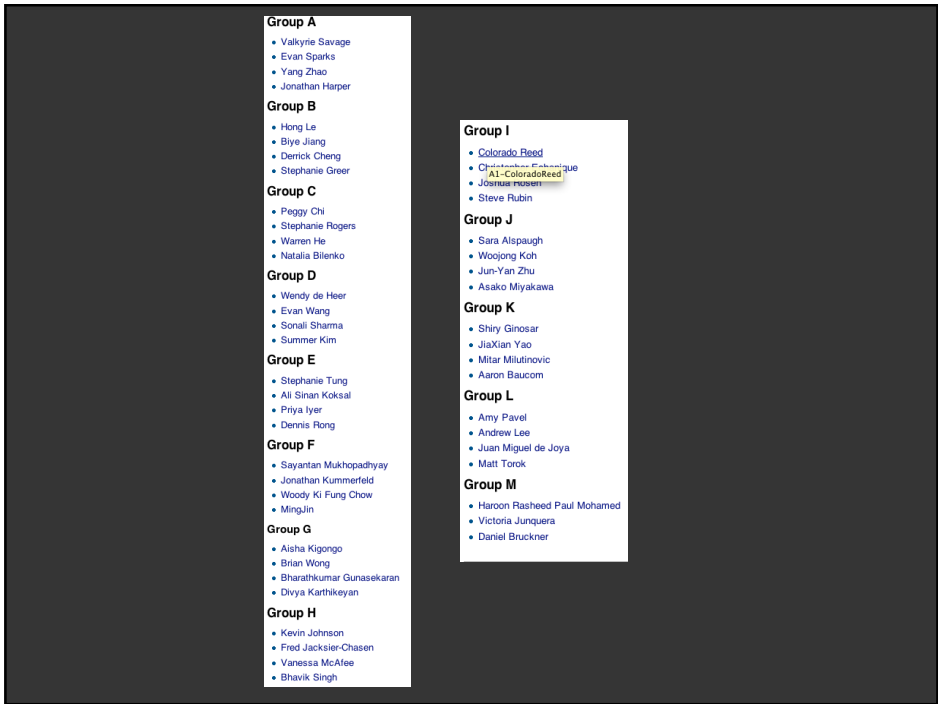
- Facilitates accurate decoding / Minimizes cognitive overhead
- Highlight elements of primary interest

Grouping / Sorting

Data Transformation

Non-Data Elements

- Descriptive: Title, Label, Caption, Data Source, Annotations
- Reference: Gridlines, Legend

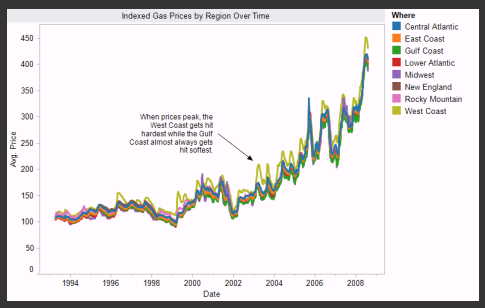

Group A
Valkyrie Savage
Evan Sparks
Yang Zhao
Jonathan Harper
Group B
Hong Le
Biye Jiang
Derrick Cheng
Stephanie Greer
Group C
Peggy Chi
Stephanie Rogers
Warren He
Natalia Bilenko
Group D
Wendy de Heer
Evan Wang
Sonali Sharma
Summer Kim
Group E
Stephanie Tung
Ali Sinan Koksal
Priya Iyer
Dennis Rong
Group F
Sayantan Mukhopadhyay
Jonathan Kummerfeld
Woody Ki Fung Chow
MingJin
Group G
Aisha Kigongo
Brian Wong
Bharathkumar Gunasekaran
Divya Karthikeyan
Group H
Kevin Johnson
Fred Jacksier-Chasen
Vanessa McAfee
Bhavik Singh
Group I
Colorado Reed
A1-ColoradoReed
Steve Rubin
Group J
Sara Alspaugh
Woojong Koh
Jun-Yan Zhu
Asako Miyakawa
Group K
Shiry Ginosar
JiaXian Yao
Mitar Milutinovic
Aaron Baucom
Group L
Amy Pavel
Andrew Lee
Juan Miguel de Joya
Matt Torok
Group M
Haroon Rasheed Paul Mohamed
Victoria Junquera
Daniel Bruckner

Assignment 2: Exploratory Data Analysis

Use existing software to formulate & answer questions

First steps

- Step 1: Pick a domain
- Step 2: Pose questions
- Step 3: Find data
- Iterate



Create visualizations

- Interact with data
- Question will evolve
- Tableau

Make wiki notebook

- Keep record of all steps you took to answer the questions

Due before class on Sep 30, 2013

Exploratory Data Analysis

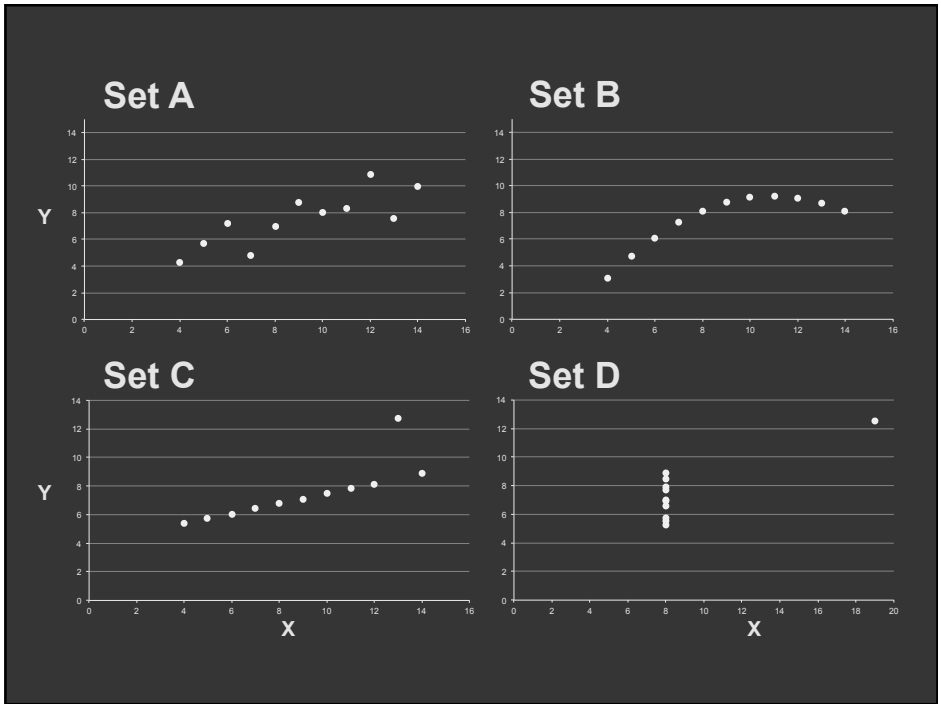
The Future of Data Analysis, John W. Tukey 1962

Set A		Set B		Set C		Set D	
X	Y	X	Y	X	Y	X	Y
10	8.04	10	9.14	10	7.46	8	6.58
8	6.95	8	8.14	8	6.77	8	5.76
13	7.58	13	8.74	13	12.74	8	7.71
9	8.81	9	8.77	9	7.11	8	8.84
11	8.33	11	9.26	11	7.81	8	8.47
14	9.96	14	8.1	14	8.84	8	7.04
6	7.24	6	6.13	6	6.08	8	5.25
4	4.26	4	3.1	4	5.39	19	12.5
12	10.84	12	9.11	12	8.15	8	5.56
7	4.82	7	7.26	7	6.42	8	7.91
5	5.68	5	4.74	5	5.73	8	6.89

Summary Statistics Linear Regression

$u_X = 9.0$ $\sigma_X = 3.317$ $Y = 3 + 0.5\ X$
$u_Y = 7.5$ $\sigma_Y = 2.03$ $R^2 = 0.67$

[Anscombe 73]



Topics

Exploratory Data Analysis

- Data Diagnostics
- Graphical Methods
- Data Transformation

Confirmatory Data Analysis

- Statistical Hypothesis Testing
- Graphical Inference

Data Diagnostics

Bureau of Justice Statistics – Data Online
http://bjs.ojp.usdoj.gov/

Reported crime in Alabama

Year	Population	Property crime rate	Burglary rate	Larceny-theft rate	Motor vehicle theft rate
2004	4525375	4029.3	987	2732.4	309.9
2005	4548327	3900	955.8	2656	289
2006	4599030	3937	968.9	2645.1	322.9
2007	4627851	3974.9	980.2	2687	307.7
2008	4661900	4081.9	1080.7	2712.6	288.6

Reported crime in Alaska

Year	Population	Property crime rate	Burglary rate	Larceny-theft rate	Motor vehicle theft rate
2004	657755	3370.9	573.6	2456.7	340.6
2005	663253	3615	622.8	2601	391
2006	670053	3582	615.2	2588.5	378.3
2007	683478	3373.9	538.9	2480	355.1
2008	686293	2928.3	470.9	2219.9	237.5

Reported crime in Arizona

Year	Population	Property crime rate	Burglary rate	Larceny-theft rate	Motor vehicle theft rate
2004	5739879	5073.3	991	3118.7	963.5
2005	5953007	4827	946.2	2958	922
2006	6166318	4741.6	953	2874.1	914.4
2007	6338755	4502.6	935.4	2780.5	786.7
2008	6500180	4087.3	894.2	2605.3	587.8

Reported crime in Arkansas

Year	Population	Property crime rate	Burglary rate	Larceny-theft rate	Motor vehicle theft rate
2004	2750000	4033.1	1096.4	2699.7	237
2005	2775708	4068	1085.1	2720	262
2006	2810872	4021.6	1154.4	2596.7	270.4
2007	2834797	3945.5	1124.4	2574.6	246.5
2008	2855390	3843.7	1182.7	2433.4	227.6

Reported crime in California

Year	Population	Property crime rate	Burglary rate	Larceny-theft rate	Motor vehicle theft rate
2004	35842038	3423.9	686.1	2033.1	704.8
2005	36154147	3321	692.9	1915	712
2006	36457549	3175.2	676.9	1831.5	666.8
2007	36553215	3032.6	648.4	1784.1	600.2
2008	36756666	2940.3	646.8	1769.8	523.8

Reported crime in Colorado

Year	Population	Property crime rate	Burglary rate	Larceny-theft rate	Motor vehicle theft rate
2004	4601821	3918.5	717.3	2679.5	521.6

Data “Wrangling”

One often needs to manipulate data prior to analysis. Tasks include reformatting, cleaning, quality assessment, and integration

Some approaches:

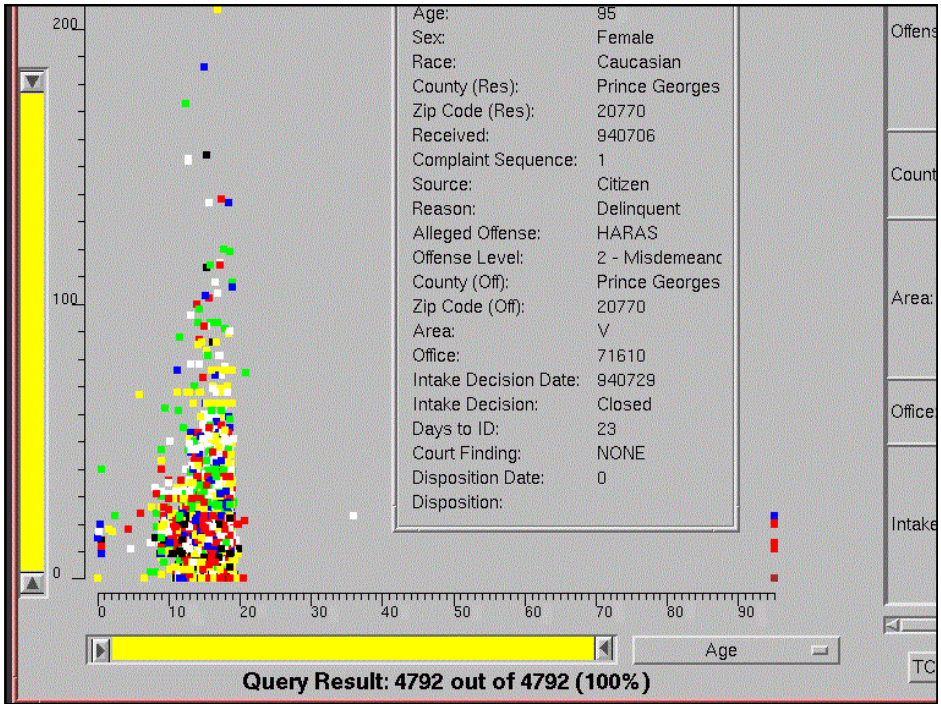
- Writing custom scripts
- Manual manipulation in spreadsheets
- Data Wrangler: http://vis.stanford.edu/wrangler
- Google Refine: http://code.google.com/p/google-refine

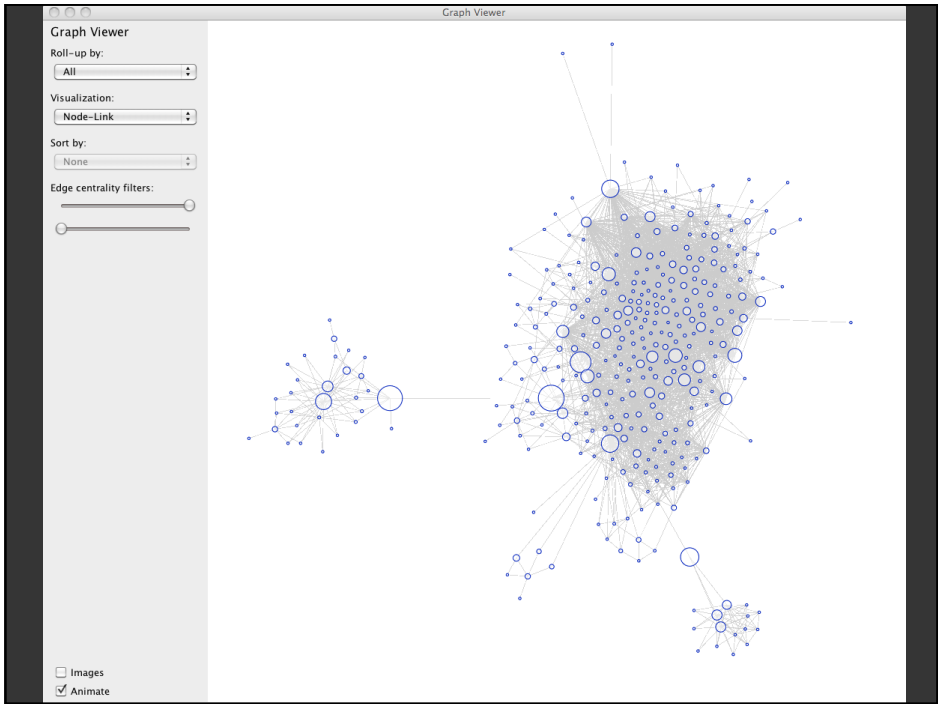
How to gauge the quality of a visualization?

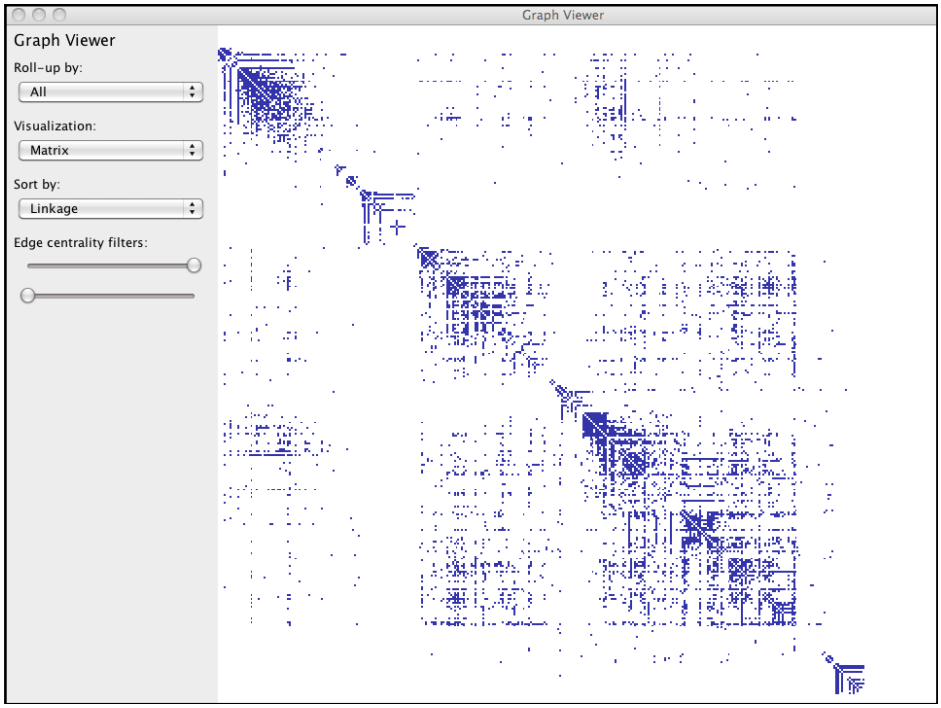
“The first sign that a visualization is good is that it shows you a problem in your data…

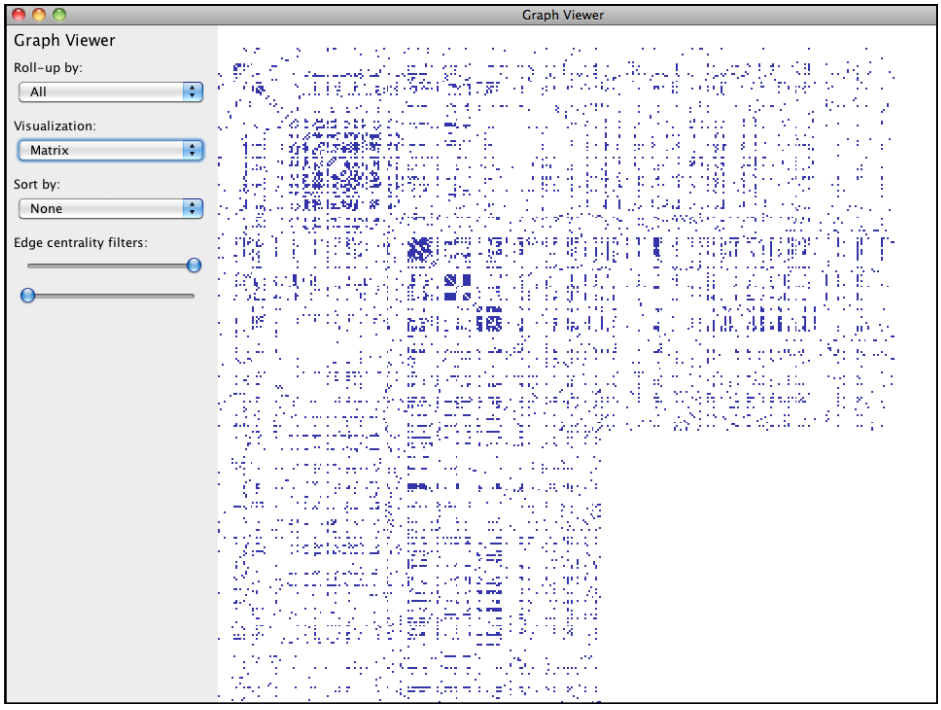
…every successful visualization that I've been involved with has had this stage where you realize, "Oh my God, this data is not what I thought it would be!" So already, you've discovered something.”

- Martin Wattenberg


200
100
0
0
10
20
30
40
50
60
70
80
90
Age:
Sex:
Race:
County (Res):
Zip Code (Res):
Received:
Complaint Sequence:
Source:
Reason:
Alleged Offense:
Offense Level:
County (Off):
Zip Code (Off):
Area:
Office:
Intake Decision Date:
Intake Decision:
Days to ID:
Court Finding:
Disposition Date:
Disposition:
95
Female
Caucasian
Prince Georges
20770
940706
1
Citizen
Delinquent
HARAS
2 - Misdemeanc
Prince Georges
20770
V
71610
940729
Closed
23
NONE
0
Offens
Count
Area:
Office
Intake
Age
TC
Query Result: 4792 out of 4792 (100%)


Graph Viewer
Graph Viewer
Roll-up by:
All
Visualization:
Node-Link
Sort by:
None
Edge centrality filters:
Images
Animate


Graph Viewer
Graph Viewer
Roll-up by:
All
Visualization:
Matrix
Sort by:
Linkage
Edge centrality filters:


Graph Viewer
Graph Viewer
Roll-up by:
All
Visualization:
Matrix
Sort by:
None
Edge centrality filters:

Visualize Friends by School?

Berkeley	|||||||||||||||||||||||||||||||
Cornell	||||
Harvard	||||||||
Harvard University	|||||||
Stanford	||||||||||||||||||
Stanford University	|||||||||
UC Berkeley	||||||||||||||||||
UC Davis	|||||||||
University of California at Berkeley	||||||||||||||
University of California, Berkeley	||||||||||||||||
University of California, Davis	|||

Data Quality & Usability Hurdles

Missing Data	no measurements, redacted, …?
Erroneous Values	misspelling, outliers, …?
Type Conversion	e.g., zip code to lat-lon
Entity Resolution	diff. values for the same thing?
Data Integration	effort/errors when combining data

LESSON: **Anticipate problems with your data.**
Many research problems around these issues!

Exploratory Analysis: Effectiveness of Antibiotics

The Data Set

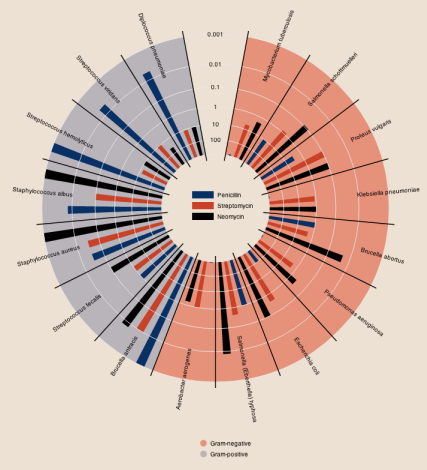
Genus of Bacteria	String
Species of Bacteria	String
Antibiotic Applied	String
Gram-Staining?	Pos / Neg
Min. Inhibitory Concent. (g)	Number

Collected prior to 1951

What questions might we ask?

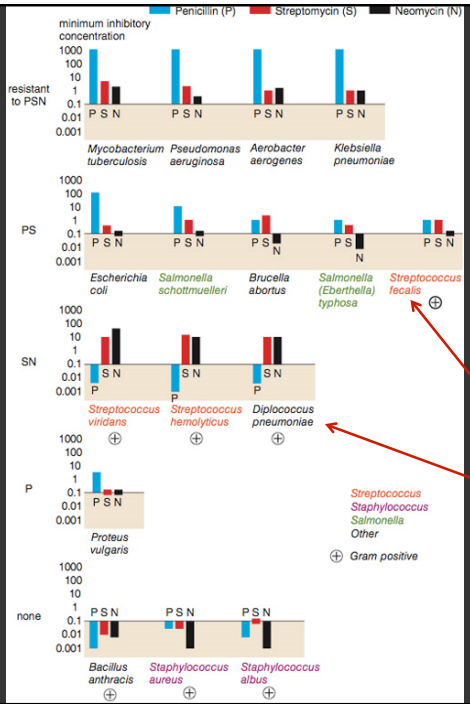
Table 1: Burtin's data.	Antibiotic			
Bacteria	Penicillin	Streptomycin	Neomycin	Gram Staining
Aerobacter *aerogenes*	870	1	1.6	negative
Brucella *abortus*	1	2	0.02	negative
Brucella *anthracis*	0.001	0.01	0.007	positive
Diplococcus *pneumoniae*	0.005	11	10	positive
Escherichia *coli*	100	0.4	0.1	negative
Klebsiella *pneumoniae*	850	1.2	1	negative
Mycobacterium *tuberculosis*	800	5	2	negative
Proteus *vulgaris*	3	0.1	0.1	negative
Pseudomonas *aeruginosa*	850	2	0.4	negative
Salmonella (Eberthella) *typhosa*	1	0.4	0.008	negative
Salmonella *schottmuelleri*	10	0.8	0.09	negative
Staphylococcus *albus*	0.007	0.1	0.001	positive
Staphylococcus *aureus*	0.03	0.03	0.001	positive
Streptococcus *fecalis*	1	1	0.1	positive
Streptococcus *hemolyticus*	0.001	14	10	positive
Streptococcus *viridans*	0.005	10	40	positive

Will Burtin, 1951



Bacteria	Penicillin	Antibiotic Streptomycin	Neomycin	Gram stain
Aerobacter aerogenes	870	1	1.6	–
Brucella abortus	1	2	0.02	–
Bacillus anthracis	0.001	0.01	0.007	+
Diplococcus pneumoniae	0.005	11	10	+
Escherichia coli	100	0.4	0.1	–
Klebsiella pneumoniae	850	1.2	1	–
Mycobacterium tuberculosis	800	5	2	–
Proteus vulgaris	3	0.1	0.1	–
Pseudomonas aeruginosa	850	2	0.4	–
Salmonella (Eberthella) typhosa	1	0.4	0.008	–
Salmonella schottmuelleri	10	0.8	0.09	–
Staphylococcus albus	0.007	0.1	0.001	+
Staphylococcus aureus	0.03	0.03	0.001	+
Streptococcus fecalis	1	1	0.1	+
Streptococcus hemolyticus	0.001	14	10	+
Streptococcus viridans	0.005	10	40	+

How do the drugs compare?



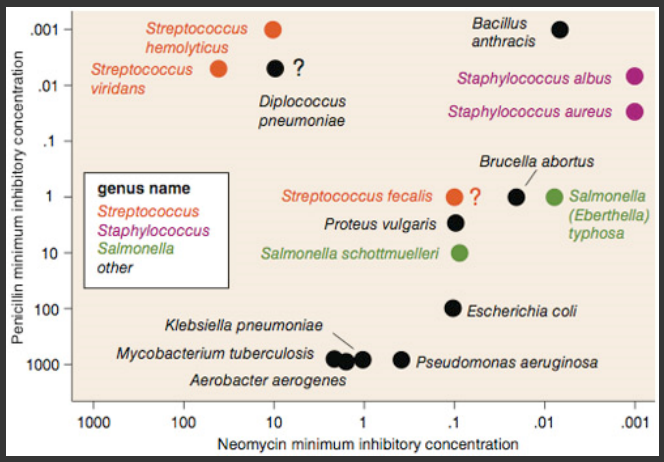
How do the bacteria group with respect to antibiotic resistance?

Not a streptococcus! (realized ~30 yrs later)

Really a streptococcus! (realized ~20 yrs later)

Wainer & Lysen

American Scientist, 2009



How do the bacteria group w.r.t. resistance? Do different drugs correlate?

Wainer & Lysen

American Scientist, 2009

Lessons

Exploratory Process

1. Construct graphics to address questions
2. Inspect "answer" and assess new questions
3. Repeat!

Transform the data appropriately (e.g., invert, log)

"Show data variation, not design variation"

-Tufte