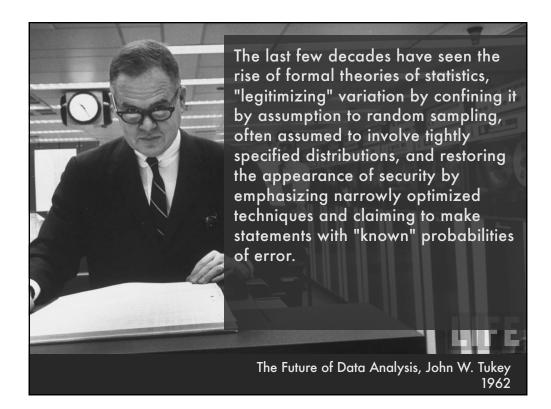
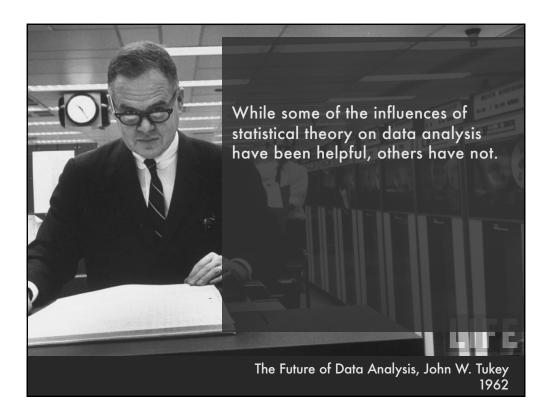


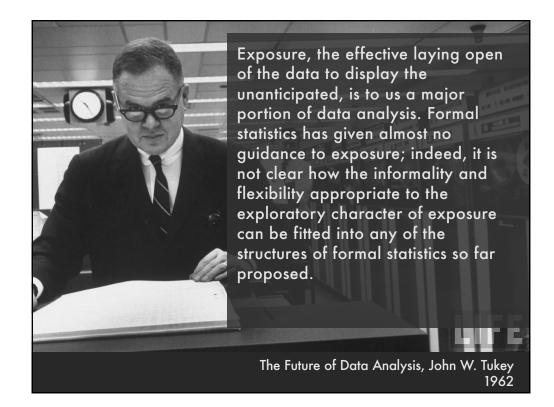
Due before class on Oct 15, 2018





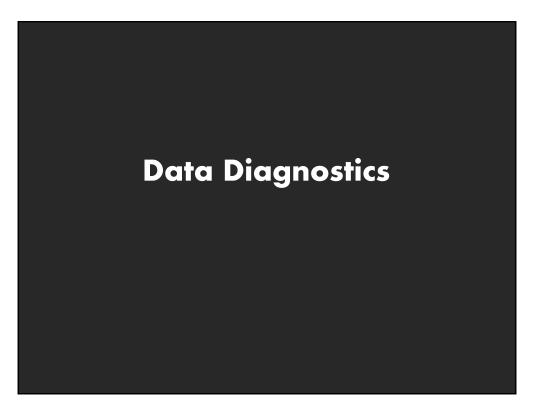






Topics

Data Diagnostics Effectiveness of antibiotics Confirmatory analysis Graphical Inference Intro to Tableau



Bureau of Justice Statistics – Data Online http://bjs.ojp.usdoj.gov/						
Report	Reported crime in Alabama					
Year 2004 2005 2006 2007 2008	Population 4525375 4029.3 4548327 3900 4599030 3937 4627851 3974.9 4661900 4081.9	Property crime rate 987 2732.4 309.9 955.8 2656 289 968.9 2645.1 322.9 980.2 2687 307.7 1080.7 2712.6 288.6	Burglary rate	Larceny-theft rate	Motor vehicle theft rate	
Report	ed crime in Alask	a				
Year 2004 2005 2006 2007 2008	Population 657755 3370.9 663253 3615 670053 3582 683478 3373.9 686293 2928.3	Property crime rate 573.6 2456.7 340.6 622.8 2601 391 615.2 2588.5 378.3 538.9 2480 355.1 470.9 2219.9 237.5	Burglary rate	Larceny-theft rate	Motor vehicle theft rate	
Report	ed crime in Arizo	ina				
Year 2004 2005 2006 2007 2008	Population 5739879 5073.3 5953007 4827 6166318 4741.6 6338755 4502.6 6500180 4087.3	Property crime rate 991 3118.7 963.5 946.2 2958 922 953 2874.1 914.4 935.4 2780.5 786.7 894.2 2605.3 587.8	Burglary rate	Larceny-theft rate	Motor vehicle theft rate	
Report	Reported crime in Arkansas					
Year 2004 2005 2006 2007 2008	Population 2750000 4033.1 2775708 4068 2810872 4021.6 2834797 3945.5 2855390 3843.7	Property crime rate 1096.4 2699.7 237 1085.1 2720 262 1154.4 2596.7 270.4 1124.4 2574.6 246.5 1182.7 2433.4 227.6	Burglary rate	Larceny-theft rate	Motor vehicle theft rate	
Reported crime in California						
Year 2004 2005 2006 2007 2008	Population 35842038 36154147 36457549 36553215 36756666	Property crime rate 3423.9 686.1 2033.1 3321 692.9 1915 3175.2 676.9 1831.5 3032.6 648.4 1784.1 2940.3 646.8 1769.8	Burglary rate 704.8 712 666.8 600.2 523.8	Larceny-theft rate	Motor vehicle theft rate	
Reported crime in Colorado						
Year 2004	Population 4601821 3918.5	Property crime rate 717.3 2679.5 521.6	Burglary rate	Larceny-theft rate	Motor vehicle theft rate	

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: <u>http://vis.stanford.edu/wrangler</u>

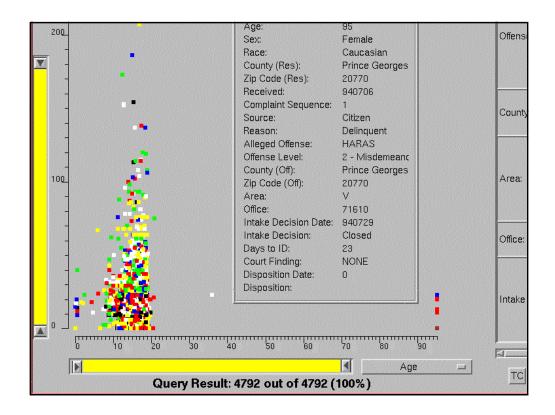
Google Refine: <u>http://code.google.com/p/google-refine</u>

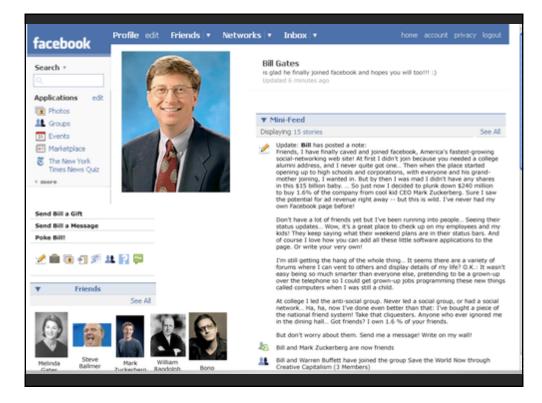
How to gauge the quality of a visualization?

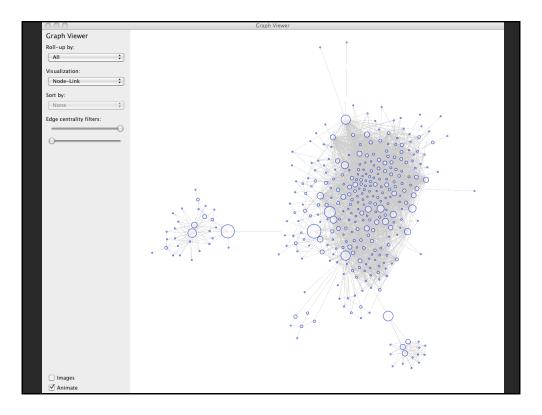
"The first sign that a visualization is good is that it shows you a problem in your data...

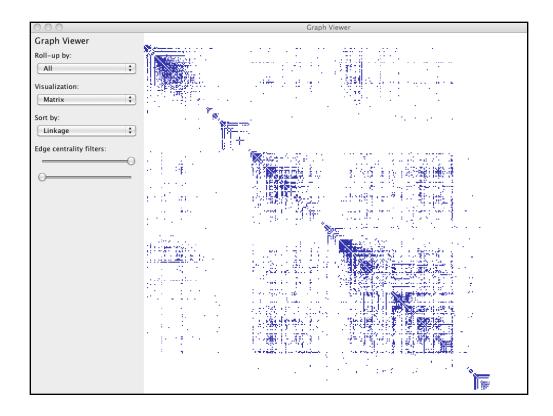
...every successful visualization that I've been involved with has had this stage where you realize, "Oh my God, this data is not what I thought it would be!" So already, you've discovered something."

- Martin Wattenberg









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Graph Viewer					
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	[10] M. M. M. Kamara, J. K. Kawa, Kaw Kawa, Kawa, K				
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Visualize Friends by School?

Berkeley Cornell Harvard Harvard University Stanford Stanford University UC Berkeley UC Davis Univ. of California at Berkeley Univ. of California, Berkeley Univ. of California, Davis

Data Quality & Usability Hurdles

Missing Data Erroneous Values Type Conversion Entity Resolution Data Integration

no measurements, redacted, …? misspelling, outliers, …? e.g., zip code to lat-lon diff. values for the same thing? effort/errors when combining data

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

Exploratory Analysis: Effectiveness of Antibiotics

What questions might we ask?

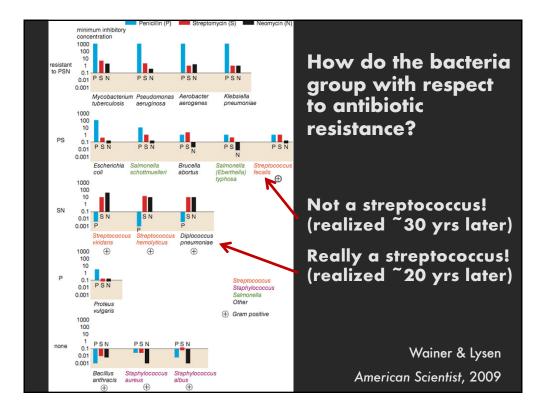
Table 1: Burtin's data.		Antibiotic		
Bacteria	Penicillin	Streptomycin	Neomycin	Gram Staining
Aerobacter aerogenes	870	1	1.6	negative
Brucella abortus	1	2	0.02	negative
Brucella anthracis	0.001	0.01	0.007	positive
Diplococcus pneumoniae	0.005	11	10	positive
Escherichia coli	100	0.4	0.1	negative
Klebsiella 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) <i>typhosa</i>	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 <i>fecalis</i>	1	1	0.1	positive
Streptococcus hemolyticus	0.001	14	10	positive
Streptococcus viridans	0.005	10	40	positive

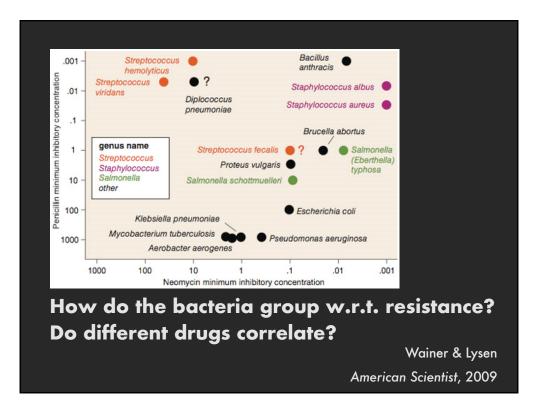
The Data Set

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

Collected prior to 1951

Will Burtin, 1951 Penicillin Antibiotic Gram Streptomycin Neomycin stain Bacteria Aerobacter aerogenes 870 1 1.6 _ Brucella abortus 1 2 0.02 0.001 Bacillus anthracis 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 0.1 + 1 1 Streptococcus hemolyticus 0.001 14 10 + Streptococcus viridans 0.005 10 40 How do the drugs compare?





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



Some Uses of Formal Statistics

What is the probability that the pattern I'm seeing might have arisen by chance?

- With what parameters does the data best fit a given function? What is the goodness of fit?
- How well do one (or more) data variables predict another?

... and many others

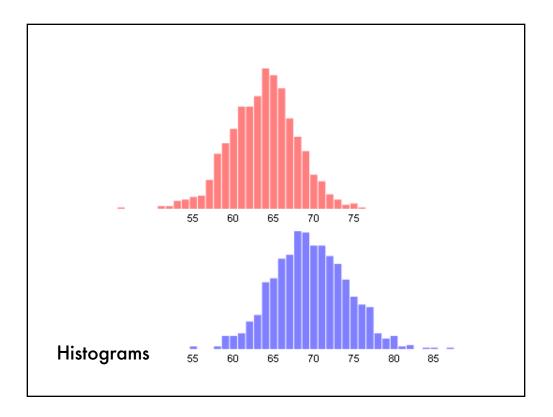
Example: Heights by Gender

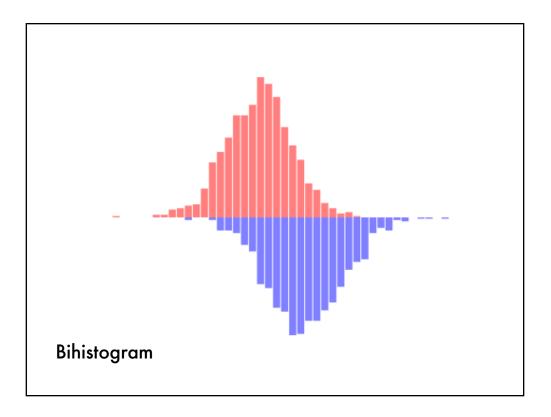
Gender Height (in) Male / Female Number

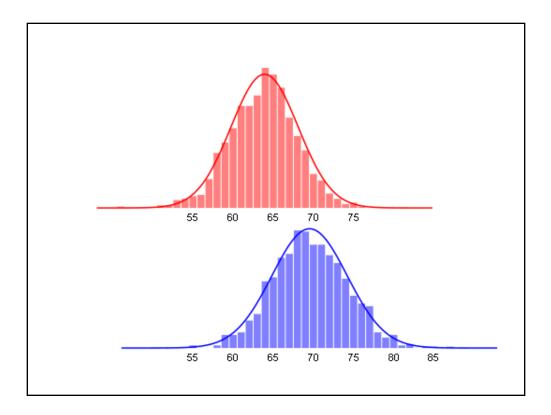
μ _m = 69.4	σ _m = 4.69	N _m = 1000
μ _f = 63.8	σ _f = 4.18	$N_{f} = 1000$

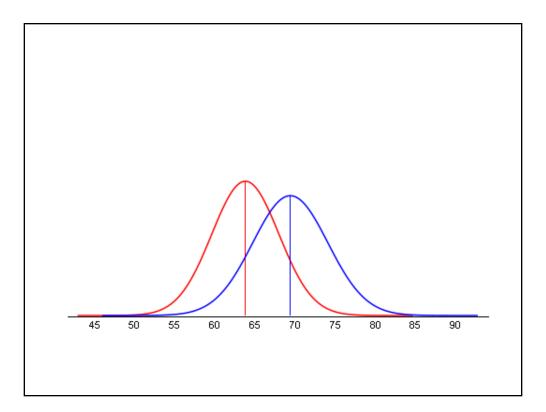
Is this difference in heights significant?

In other words: assuming no true difference, what is the prob. that our data is due to chance?









Formulating a Hypothesis

Null Hypothesis (H₀): (population)

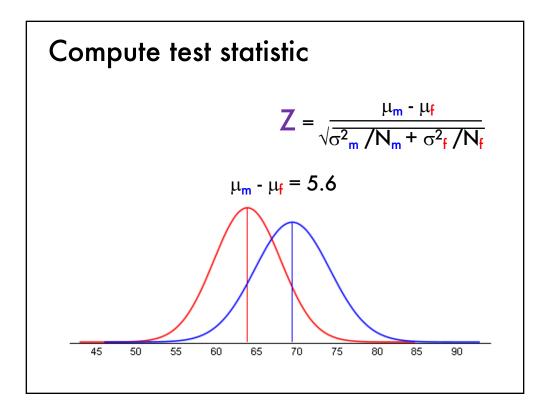
 $\mu_m = \mu_f$

- Alternate Hypothesis (H_a): $\mu_m \neq \mu_f$ (population)
- A statistical hypothesis test assesses the likelihood of the null hypothesis.
- What is the probability of sampling the observed data assuming population means are equal?

This is called the *p* value

Testing Procedure

Compute a test statistic. This is a number that in essence summarizes the difference.

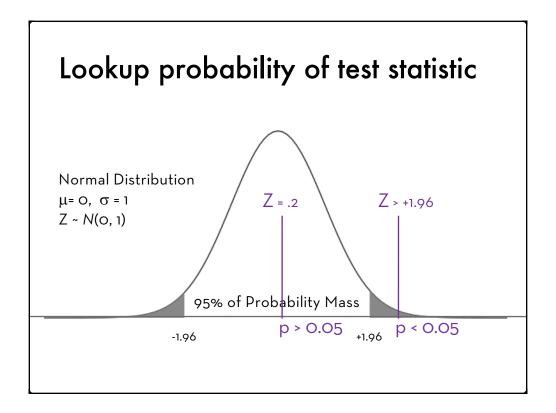


Testing Procedure

Compute a test statistic. This is a number that in essence summarizes the difference.

The possible values of this statistic come from a known probability distribution.

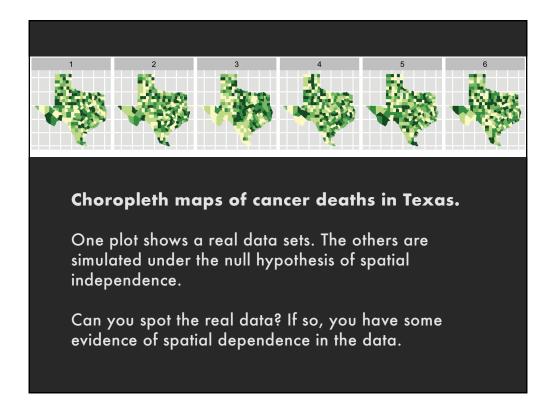
According to this distribution, look up the probability of seeing a value meeting or exceeding the test statistic. This is the *p* value.

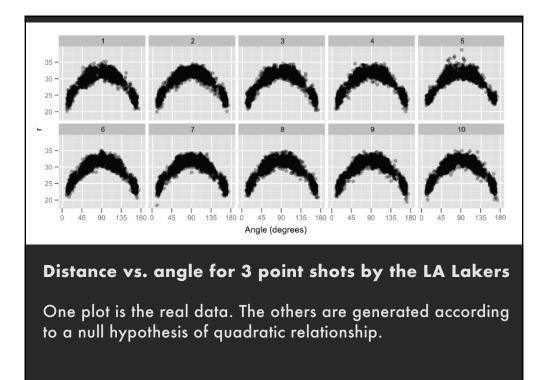


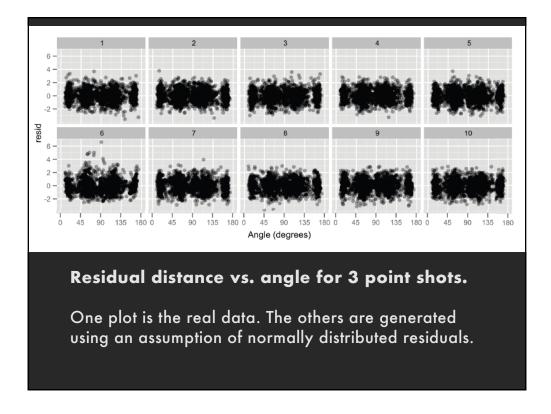
Statistical Significance

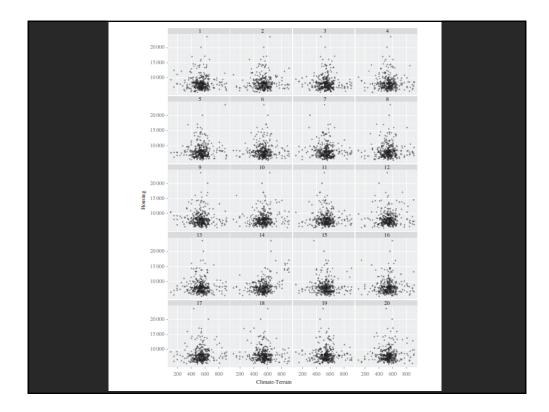
- The threshold at which we consider it safe (or reasonable?) to *reject the null hypothesis*
- If p < 0.05, we typically say that the observed effect or difference is statistically significant
- This means that there is a less than 5% chance that the observed data is due to chance
- Note that the choice of 0.05 is a somewhat arbitrary threshold (chosen by R. A. Fisher)



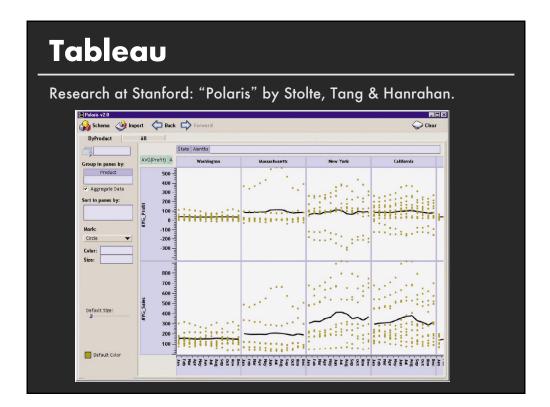












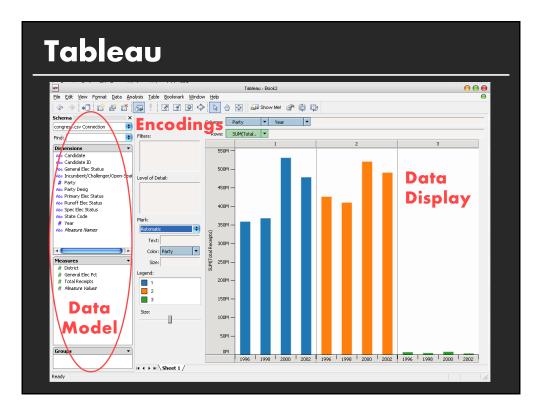


Tableau demo

The dataset:

- Federal Elections Commission Receipts
- Every Congressional Candidate from 1996 to 2002
- **4 Election Cycles**
- 9216 Candidacies

Data Set Schema

- Year (Qi)
- Candidate Code (N)
- Candidate Name (N)
- Incumbent / Challenger / Open-Seat (N)
- Party Code (N) [1=Dem,2=Rep,3=Other]
- Party Name (N)
- Total Receipts (Qr)
- State (N)
- District (N)

 This is a subset of the larger data set available from the FEC, but should be sufficient for the demo

Hypotheses?

What might we learn from this data?

Hypotheses?

What might we learn from this data?

- Has spending increased over time?
- Do democrats or republicans spend more money?
- Candidates from which state spend the most money?

Tableau Demo