Exploratory Data Analysis

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CS 294-10: Visualization Fall 2014

Last Time: Visualization Designs



In-Class Review Rubric

Expressiveness

- Do the mappings show the facts and only the facts?
- Are visual mappings consistent? (e.g., respect color mappings)

Effectiveness

- Are perceptually effective encodings used?
- Are the most important data mapped to the most effective visual variables?

Cognitive Load (Efficiency)

Are there extraneous (unmapped) visual elements?

Data Transformation

Are transformations (filter, sort, derive, aggregate) appropriate?

Guides (Non-Data Elements)

- Descriptive, consistent: Title, Label, Caption, Source, Annotations
- Meaningful references: Gridlines, Legend













Set	tΔ	Set	B	Se	t C	Se	of D
x	Y	×	γ	x	γ	×	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 u _y = 7.5	$\begin{array}{c} 0 \sigma_{\rm X} = 3\\ 5 \sigma_{\rm Y} = 2\end{array}$	8.317 2.03	Y = 3 - R ² = 0	+ 0.5 X .67		[Anscon	nbe 73]



Topics

Exploratory Data Analysis

Data Diagnostics Graphical Methods Data Transformation

Confirmatory Data Analysis

Statistical Hypothesis Testing Graphical Inference

Data Diagnostics

Rumpau of Justice Statistics Data Oplian						
http://bjs.ojp.usdoj.gov/						
Reported	Reported crime in Alabama					
Year F 2004 4 2005 4 2006 4 2007 4 2007 4 2008 4	Population 4525375 4029.3 4548327 3900 4599030 3937 4627851 3974.9 4661900 4081.9	Property crime 987 2732.4 955.8 2656 968.9 2645.1 980.2 2687 1080.7 2712.6	rate 309.9 289 322.9 307.7 288.6	Burglary rate	Larceny-theft rate	Motor vehicle theft rate
Reported	crime in Alaska	a				
Year F 2004 6 2005 6 2006 6 2007 6 2008 6	Population 657755 3370.9 663253 3615 670053 3582 683478 3373.9 686293 2928.3	Property crime 573.6 2456.7 622.8 2601 615.2 2588.5 538.9 2480 470.9 2219.9	rate 340.6 391 378.3 355.1 237.5	Burglary rate	Larceny-theft rate	Motor vehicle theft rate
Reported	crime in Arizo	na				
Year P 2004 9 2005 9 2006 6 2007 6 2008 6	Population 5739879 5073.3 5953007 4827 6166318 4741.6 6338755 4502.6 6500180 4087.3	Property crime 991 3118.7 946.2 2958 953 2874.1 935.4 2780.5 894.2 2605.3	rate 963.5 922 914.4 786.7 587.8	Burglary rate	Larceny-theft rate	Motor vehicle theft rate
Reported	crime in Arkan:	sas				
Year 2004 2005 2006 2007 2007 2008	Population 2750000 4033.1 2775708 4068 2810872 4021.6 2834797 3945.5 2855390 3843.7	Property crime 1096.4 2699.7 1085.1 2720 1154.4 2596.7 1124.4 2574.6 1182.7 2433.4	rate 237 262 270.4 246.5 227.6	Burglary rate	Larceny-theft rate	Motor vehicle theft rate
Reported crime in California						
Year 7 2004 2 2005 2 2006 2 2007 2 2008 2	Population 35842038 36154147 36457549 36553215 36756666	Property crime 3423.9 686.1 3321 692.9 3175.2 676.9 3032.6 648.4 2940.3 646.8	rate 2033.1 1915 1831.5 1784.1 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 P 2004 4	Population 4601821 3918.5	Property crime 717.3 2679.5	rate 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: http://vis.stanford.edu/wrangler

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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	a shekara ta shekara t				

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 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)

Collected prior to 1951

Will Burtin, 1951

Bacteria	Penicillin	Antibiotic Streptomycin	Neomycin	Gran 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	+

String

String

String

Pos / Neg

Number



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

Exploratory Analysis: Participation on Amazon's Mechanical Turk







0 1 1 1 2 2 2 2 3 3 3 3 3 4 4 4 4 4 4 4 4 4 5 6 7 8 8 8 8 8 9 0 0 0 0 1 1 1 1 2 2 3 3 3 3 4 4 4 4 5 5 6 7 7 8 9 9 9 9 9 1 2 0 0 1 1 1 5 7 8 9 0 0 1 2 3 3 3 4 6 6 8 8 3 0 0 1 1 1 1 3 3 4 5 5 5 6 7 8 9 4 0 2 3 5 6 7 7 7 9 5 6 1 2 6 7 8 9 9 9 7 0 0 0 1 6 7 9 8 0 0 1 2 3 4 4 4 4 4 4 4 5 6 7 7 7 9 Stem-and-Leaf Plot











Lessons

Even for "simple" data, a variety of graphics might provide insight. Again, tailor the choice of graphic to the questions being asked, but be open to surprises.

Graphics can be used to understand and help assess the quality of statistical models.

Premature commitment to a model and lack of verification can lead an analysis astray.

Confirmatory Data Analysis

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









Formulating a Hypothesis

Null Hypothesis (Alternate Hypoth	(H _o): Iesis (H	l _a):	μ _m = μ _f μ _m ≠ μ _f	(population) (population)	
A statistical hypo likelihood of t	othesis he null	test a hypot	ssesses 1 hesis.	the	
What is the probability of sampling the observed data assuming population means are equal?					
This is called the	p valu	e			

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)

Common Statistical Methods

Data Type

2 uni. dists

> 2 uni. dists

Question

Do data distributions have different "centers"? (aka "location" tests)

Are observed counts significantly different?

Are two vars related?

? 2 varia

Do 1 (or more) variables predict another?

> 2 multi. dists Counts in categories

2 variables

Continuous Binary t-Test ANOVA MANOVA

Parametric

Mann-Whitney U Kruskal-Wallis Median Test

Non-Parametric

χ² (chi-squared)

Rank correl.

Linear regression Logistic regression

Pearson coeff.

Graphical Inference

Buja Cook, Hoffman, Wickham et al.

Residual distance vs. angle for 3 point shots.

One plot is the real data. The others are generated using an assumption of normally distributed residuals.

Summary

Exploratory analysis may combine graphical methods, data transformations, and statistics

Use questions to uncover more questions

Formal methods may be used to confirm, sometimes on held-out or new data

Visualization can further aid assessment of fitted statistical models