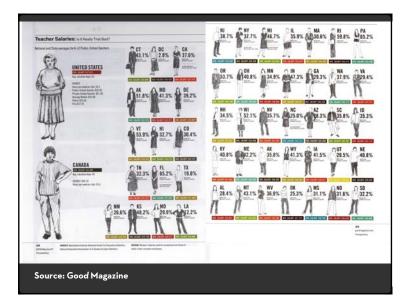
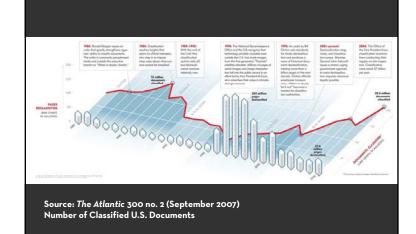
CS448B ::: 30 Sep 2009 Multidimensional Vis

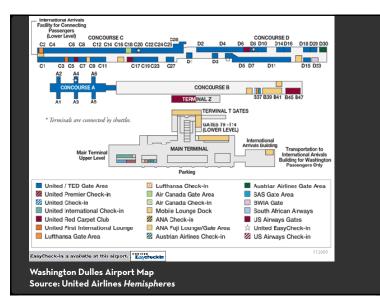


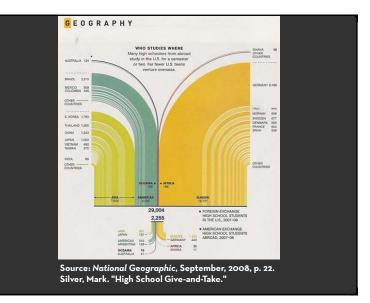
Jeffrey Heer Stanford University

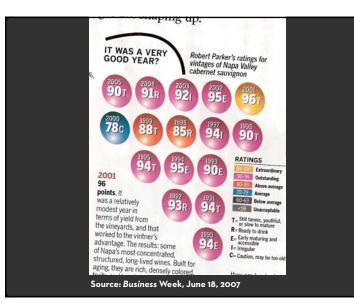


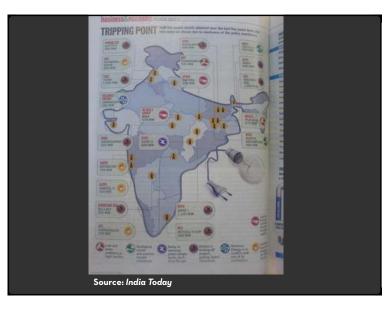


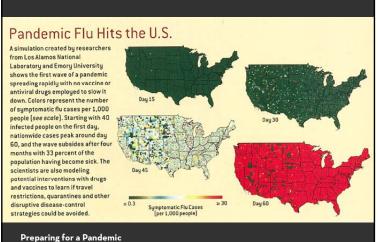




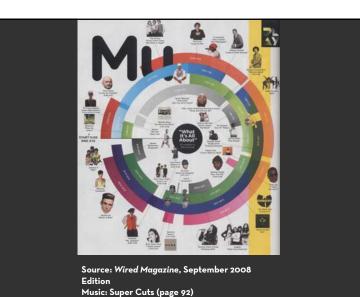












Assignment 2: Visual Data Analysis

Use visualization software (Tableau) to form & answer questions

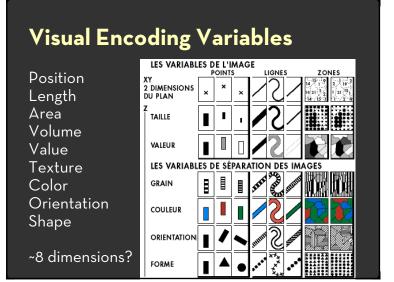
First steps:

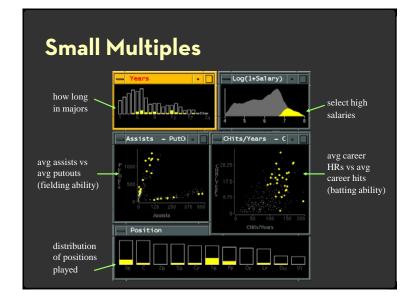
- Step 1: Pick a domain
- Step 2: Pose questions
- Step 3: Find Data
- Iterate
- Create visualizations
- Interact with data
- Refine your questions
- Make wiki notebook
- Keep record of your analysis
- Prepare a final graphic and caption



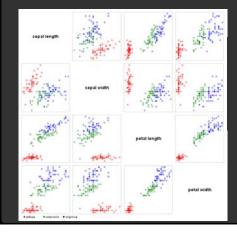
Due by end of day on Monday, October 12

Multidimensional Visualization



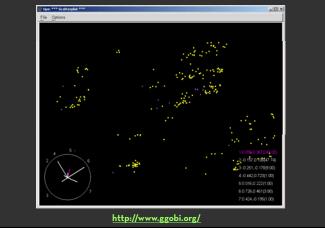


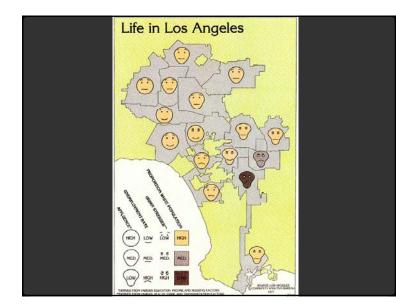
Scatterplot Matrix (SPLOM)



Scatter plots enabling pair-wise comparison of each data dimension.

Dimensional Projection





Chernoff Faces (1973)

Insight: We have evolved a sophisticated ability to interpret facial expression.

Idea: Map data variables to facial features.



Pupil s

Question: Do we process facial features in an uncorrelated way? (i.e., are they separable?)

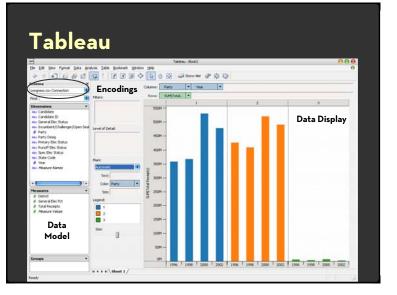
This is just one example of nD "glyphs"

Tableau / Polaris

Polaris

Research at Stanford by Stolte, Tang, and Hanrahan.





Polaris/Tableau Approach

Insight: can simultaneously specify both database queries and visualization

Choose data, then visualization, not vice versa

Use smart defaults for visual encodings

More recently: automate visualization design

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

Hypotheses?

What might we learn from this data? • ??

Hypotheses?

What might we learn from this data? Correlation between receipts and winners? Do receipts increase over time? Which states spend the most? Which party spends the most? Margin of victory vs. amount spent? Amount spent between competitors?

Polaris/Tableau Approach

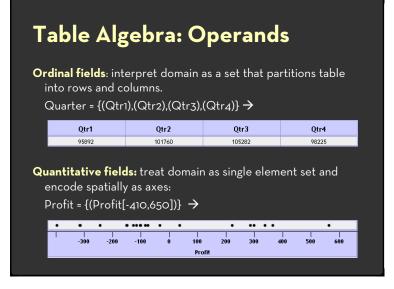
Insight: can simultaneously specify both database queries and visualization Choose data, then visualization, not vice versa Use smart defaults for visual encodings More recently: automate visualization design

Specifying Table Configurations

- Operands are the database fields
- Each operand interpreted as a set {...}
- Quantitative and Ordinal fields treated differently

Three operators:

- concatenation (+)
- cross product (x)
- nest (/)



Concatenation (+) Operator

Ordered union of set interpretations

Quarter + Product Type

- = {(Qtr1),(Qtr2),(Qtr3),(Qtr4)} + {(Coffee), (Espresso)}
- = {(Qtr1),(Qtr2),(Qtr3),(Qtr4),(Coffee),(Espresso)}

Qtr1	Qtr2	Qtr3	Qtr4	Coffee	Espresso						
48	59	57	53	151	21						
Profit + Sales = {(Profit[-310,620]),(Sales[0,1000])}											
			2// 2								
	·				•						
-200	0 200	-400 600	200	400 640	800						
	Profit		Sales								

Cross (x) Operator

Cross-product of set interpretations

Quarter x Product Type

= {(Qtr1,Coffee), (Qtr1, Tea), (Qtr2, Coffee), (Qtr2, Tea), (Qtr3, Coffee), (Qtr3, Tea), (Qtr4, Coffee), (Qtr4, Tea)}



Nest (/) Operator

Cross-product filtered by existing records

Quarter x Month

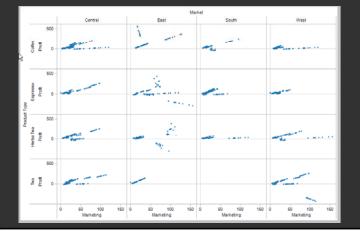
creates twelve entries for each quarter. i.e., (Qtr1, December)

Quarter / Month

creates three entries per quarter based on tuples in database (not semantics)

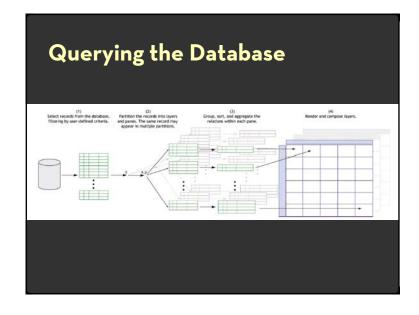
- - .	-	_		
Ordi	nal	- 0	rdi	na
Product Type			t Type	
State	Coffee	Espresso	Herbal Tea	Теа
Colorado	•	٠	•	•
Connecticut	•	•	•	•
Florida	•	•	•	•
Illinois	•		•	•
Iowa	•	•		•
Louisiana	•	•	•	
Massachusetts	•	•	•	•
Missouri	•	•	•	•
Nevada	•	•	•	•
New Hampshire	•	•		•
New Mexico		•	•	
New York	•	•	•	•
Ohio	•	•	•	•
Oklahoma	•	•	•	
Oregon	•	•	•	•
Texas	•	•	•	
Utah	•	•	•	•
Washington	•	•	•	•
Wisconsin	•	•	•	•

Quantitative - Quantitative



Ordinal - Quantitative

Concerning Concerning

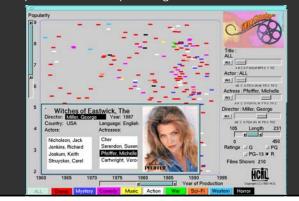


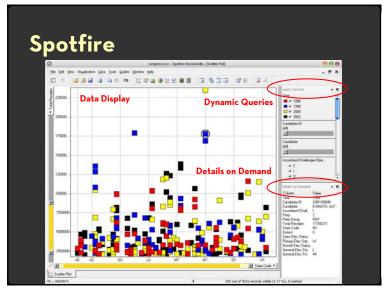
9



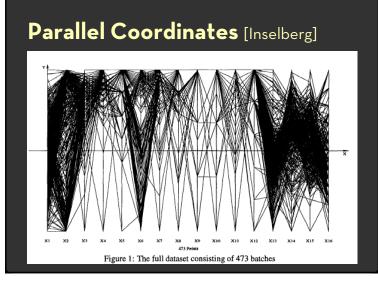
Spotfire

Research at UMD, College Park: "Starfield Displays" and "Dynamic Queries" by Ahlberg and Shneiderman





Parallel Coordinates



The Multidimensional Detective

The Dataset:

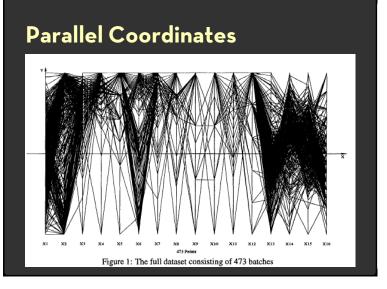
- Production data for 473 batches of a VLSI chip
- 16 process parameters:

X1: The yield: % of produced chips that are useful

- X2: The quality of the produced chips (speed)
- X3 ... X12: 10 types of defects (zero defects shown at top)
- X13 ... X16: 4 physical parameters

The Objective: Raise the yield (X1) and maintain high quality (X2)

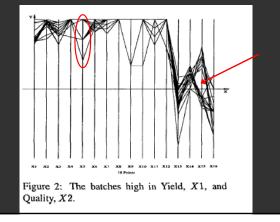
A. Inselberg, Multidimensional Detective, Proceedings of IEEE Symposium on Information Visualization (InfoVis '97), 1997

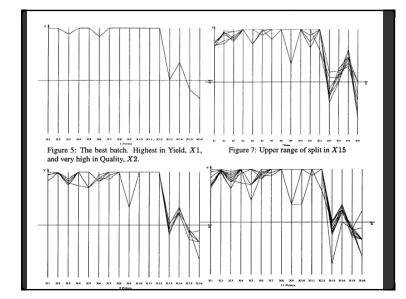


Inselberg's Principles

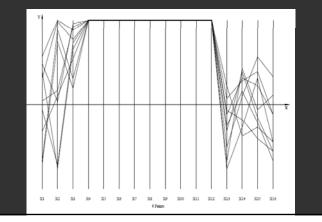
- 1. Do not let the picture scare you
- 2. Understand your objectives
 - Use them to obtain visual cues
- 3. Carefully scrutinize the picture
- 4. Test your assumptions, especially the "I am really sure of's"
- 5. You can't be unlucky all the time!

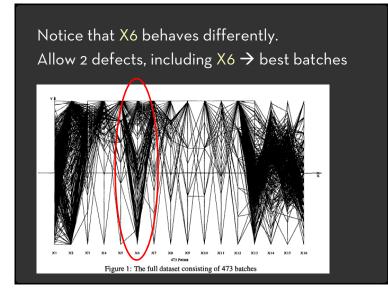
Each line represents a tuple (e.g., VLSI batch) Filtered below for high values of X1 and X2





Look for batches with *nearly* zero defects (9/10) Most of these have low yields \rightarrow defects OK.





Parallel Coordinates

Free implementation: Parvis by Ledermen

• <u>http://home.subnet.at/flo/mv/parvis/</u>

