Information Visualization and Knowledge Management

Dr. Margaret-Anne Storey Dept of Computer Science

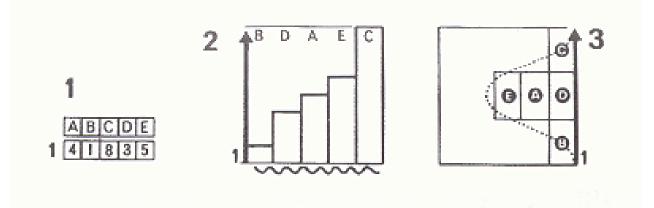
1D, 2D, 3D and > 3D

How much can we see?

- We can perceive only in 3D (4D if we include time)
- But we can see more than 3 or 4 dimensions if we make use of Bertin's retinal variables
- But how many can we effectively use?

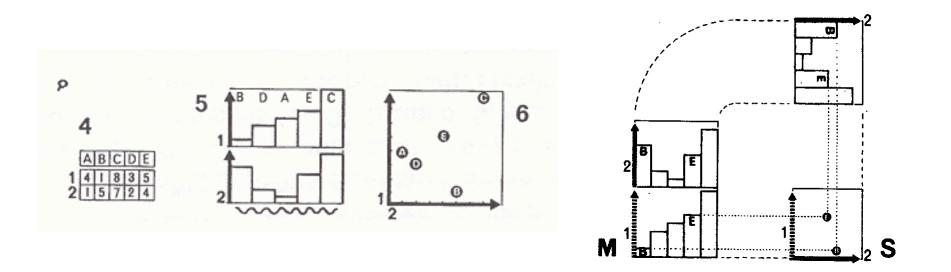
Univariate data, 1D

• We can use bar charts, point plots....



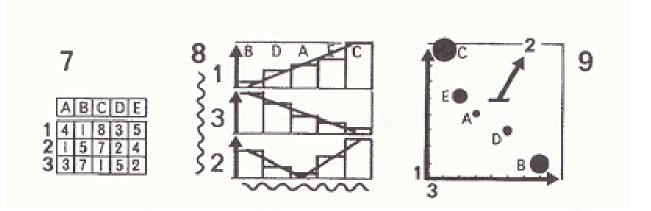
Bivariate data, 2D

- We can use scatterplot (really two barcharts projected onto a single image)
- Two barcharts side by side
- Or perhaps a barchart and one retinal variable

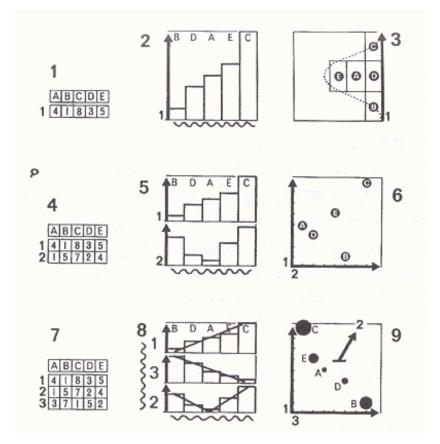


Trivariate data

- Scatter plot + retinal variable
- Scatter plot in 3D (using perspective)
- Or show multiples of 1D drawings

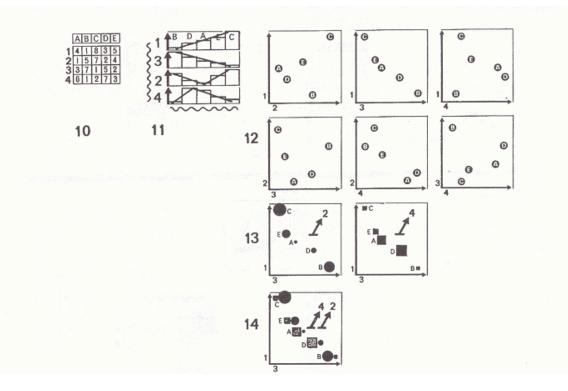


1, 2 and 3 rows



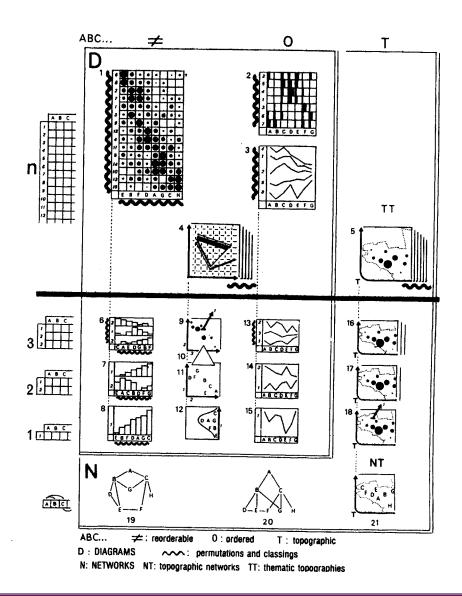
- Axes are the most effective dimension for encoding data in space
- 1, 2, 3 dimensions of data are also special cases of n dimensions

Multivariate data, >3D



"Impassable" barrier of 3 (or 4) dimensions...??A table with more than 3 rows cannot be constructed directly as a single signifying image – we need a series of images

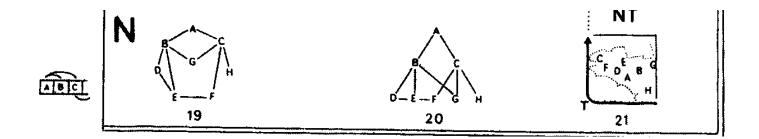
Multivariate data



- We can use a matrix permutation and discover all relationships through permutations (dynamic graphics)
- But we can permute rows only if the data is reorderable
- If the data is ordered, we can create an image file or array of curves (but only if the slopes are meaningful)
- Pick the series of images and select an important relationship to reveal
- We can also use interaction and dynamic manipulation to reveal relationships

Networks

- A Network (N) portrays the relationships which exist among the elements of a single component
- Can be unordered, so the image can be transformed any way in the plane
- If it is ordered, it can't be transformed (a tree is ordered in 1D, a map is ordered in 2D)
- Networks can be represented in matrix form (which can be permuted if it is reorderable)

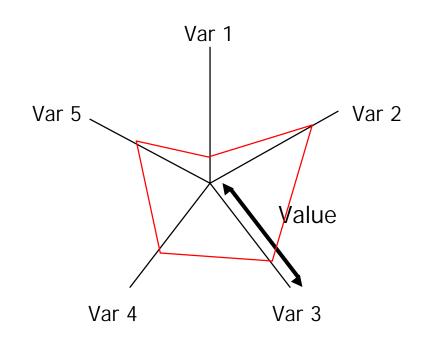


Bertin's synoptic

Synoptic -- affording a general view of the whole

- Start with the # of rows and nature of the components
- Select the image construction corresponding to the structure of the data (ordered, unordered, ordinal, nominal etc)
- Decide which is the important relationship to show first
- Use permutations to disclose other relationships

Star Plots

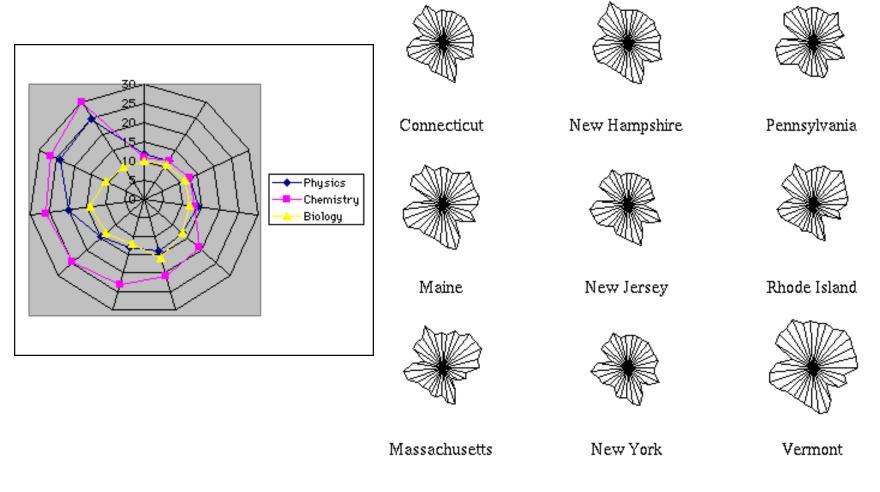


Space out the n variables at equal angles around a circle

Each "spoke" encodes a variable's value

Based on John Stasko slide

Starplots



• <u>http://seamonkey.ed.asu.edu/~behrens/asu/reports/compre/radar_plot.GIF</u>

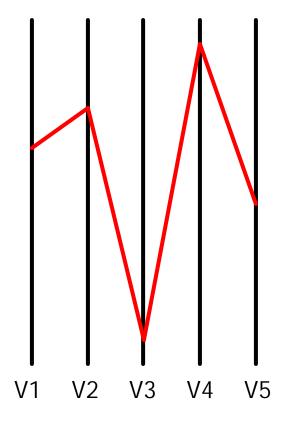
Multidimensional Detective

- System that uses parallel coordinates for information analysis and discovery
- Interactive tool
 - Can focus on certain data items
 - Color

Taken from: A. Inselberg, "Multidimensional Detective" InfoVis '97, 1997.

Based on John Stasko slide

Explicit Variables: Parallel Coordinates (Inselberg)



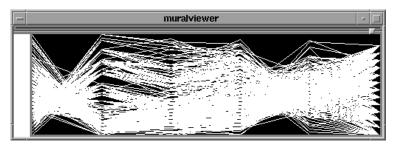
Encode variables along a horizontal row

Vertical line specifies single variable

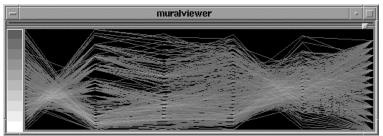
Red line specifies a case

Based on John Stasko slide

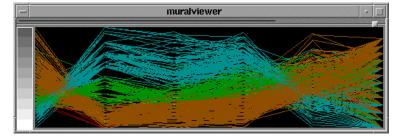
Parallel Coordinates Example



Basic



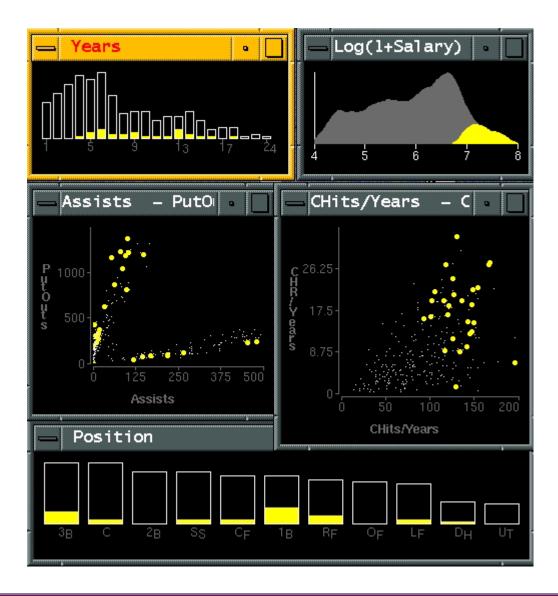
Grayscale



Color

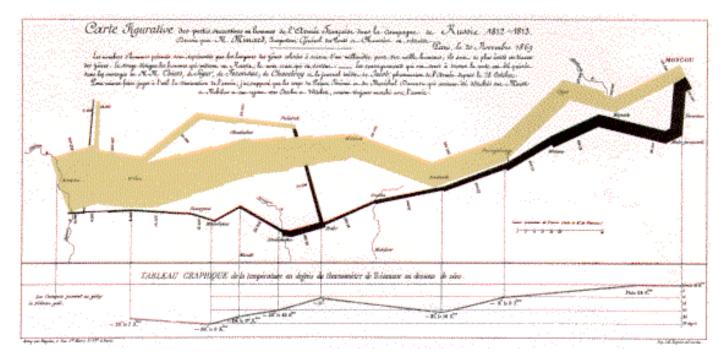
Based on John Stasko slide

Link different types of graphs: Scatterplots and histograms and bars (from Wills 95)

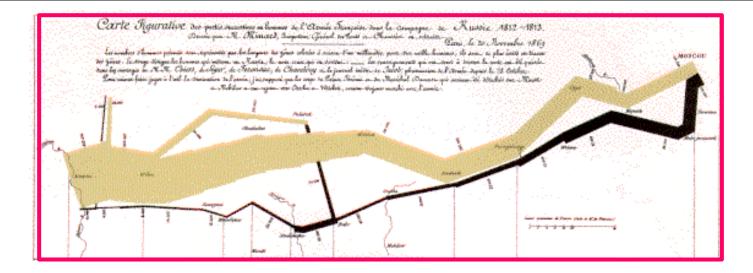


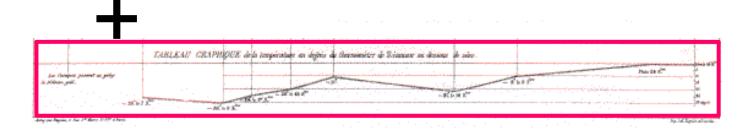
Composition/decomposition

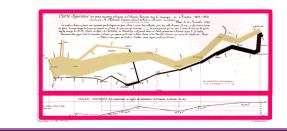
• Minard's 1869 Napoleon's march



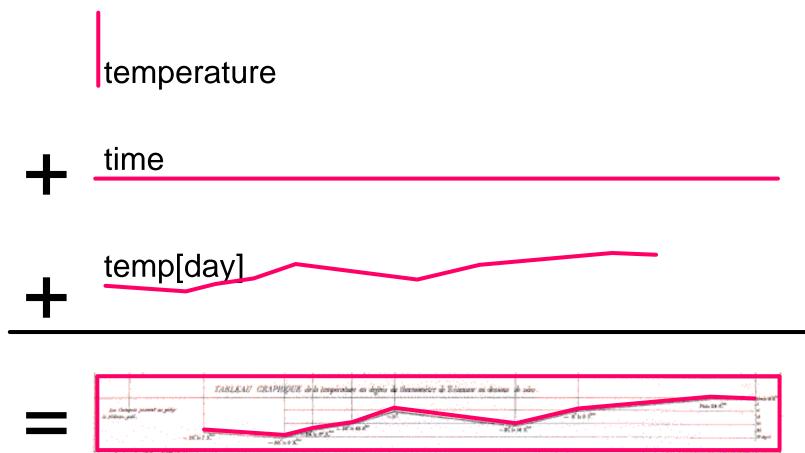
Single axis composition





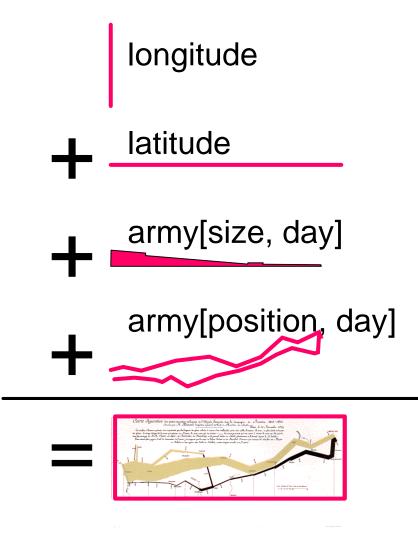


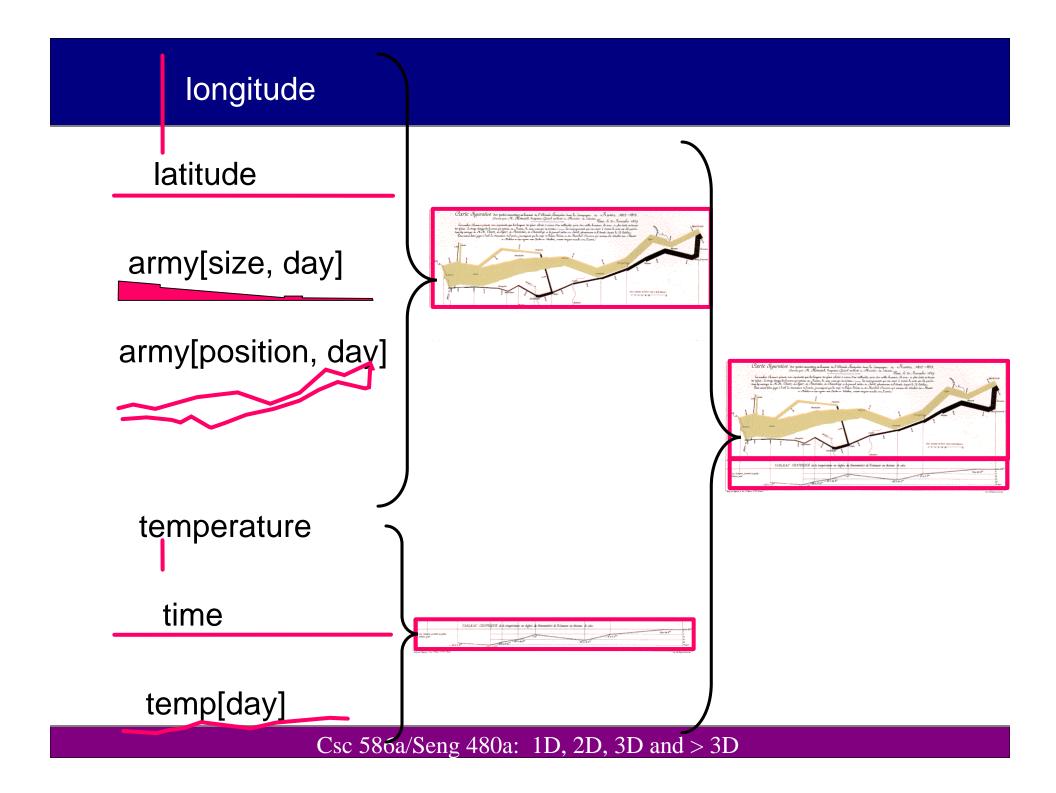
Mark Composition



Ng. Int. Eggint all and

Mark composition





- Chernoff faces: <u>http://people.cs.uchicago.edu/~wiseman/chernoff/</u>
- Preattentive Processing: <u>http://www.csc.ncsu.edu/faculty/healey/PP/PP.html</u>
- Table Lens:

http://www.tablelens.com/

• Discussion point:

– How many retinal variables can we use in a scatterplot to encode information?

Tool support for automatically designing effective graphical presentations

APT toolPolaris

Mackinlay's APT Environment

- Goal: to develop an application independent presentation tool that automatically designs effective graphical presentations (such as bar charts, graphs, scatter plots) of relational information
- Approach: Graphical presentations are "sentences of graphical languages"
- Mackinlay's work was the first attempt on automatic design

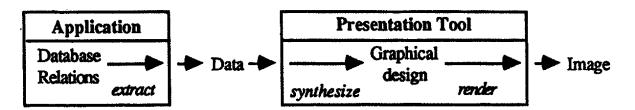


Fig. 1. A linear model for generating presentations. This simplified model, which does not include feedback loops that are required for difficult design problems, describes the fundamental process of generating a graphical presentation. A graphical design synthesized by a presentation tool describes the basic structure and meaning of a graphical presentation. The rendering process fills in the details that are required to form the final image.

Mackinlay's APT Environment (2)

- The tool extracts information from the database, analyzes it and suggests a graphic design
- 'A graphic design' is an abstract description of an image that indicates the graphical techniques (e.g. color, position) that are used to encode information
- But graphic design criteria must be codified before the presentation tool can synthesize effective designs
- The graphical presentation problem is to synthesize a graphical design that expresses a set of relations and their structural properties effectively

Fig. 8. Bertin's graphical objects and graphical relationships.

Marks: Positional: Temporal: Retinal:	Points, lines, and areas 1-D, 2-D, and 3-D Animation
IUGUILAI:	Color, shape, size, saturation, texture, and orientation

• A problem: Present the price and mileage relations, omit the car details

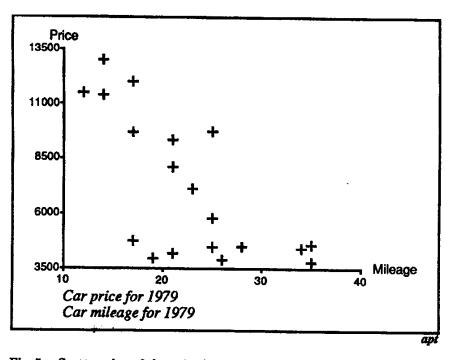


Fig. 5. Scatter plot of the price/mileage input. The graphical design for this image is in Figure 4. The design expresses the relations only if the application permits the details about the cars to be omitted. The *apt* in the lower right corner indicates that APT designed and rendered this diagram.

• But if the car details are needed what are the options?

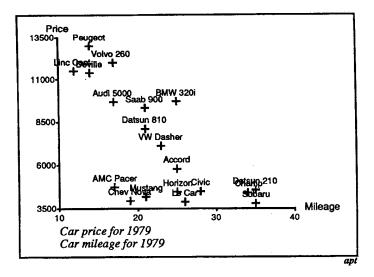


Fig. 6. Labeled scatter plot for the price/mileage input. Although a more sophisticated rendering could avoid the overlapping of the labels, two basic problems of a labeled scatter plot design reduce its effectiveness. First, labels make it difficult to perceive the positions of the points. Second, a given label is difficult to find.

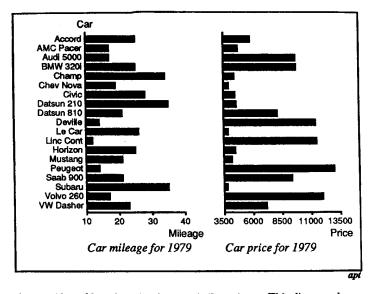


Fig. 7. Aligned bar chart for the price/mileage input. This diagram shows the detailed properties of the cars better than a scatter plot. However, the general relationships are not so easy to see.

Criteria in MacKinlay's framework

- Expressiveness criteria:
 - Determines whether a graphical language can express the desired information and *only* the desired information
- Effectiveness criteria:
 - Determines whether the desired information exploits the capabilities of the output medium and the human visual system

Incorrect use of a bar chart – Violates expressiveness criteria

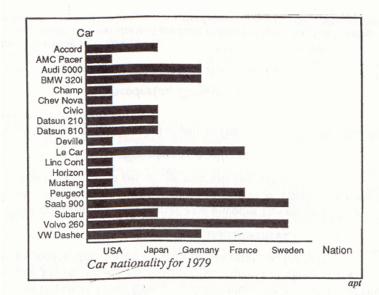


Fig. 11. Incorrect use of a bar chart for the *Nation* relation. The lengths of the bars suggest an ordering on the vertical axis, as if the USA cars were longer or better than the other cars, which is not true for the *Nation* relation.

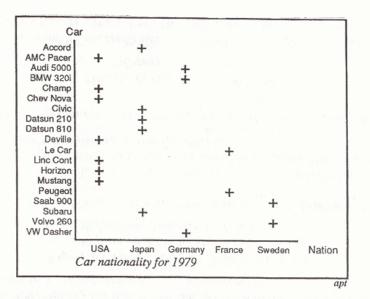


Fig. 12. Correct use of a plot chart for the *Nation* relation. Since bar charts encode ordered domain sets, plot charts are conventionally used to encode nominal domain sets. The ordering of the labels on the axes is ignored.

Is this ok?

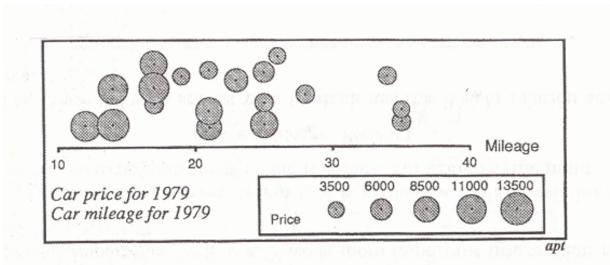
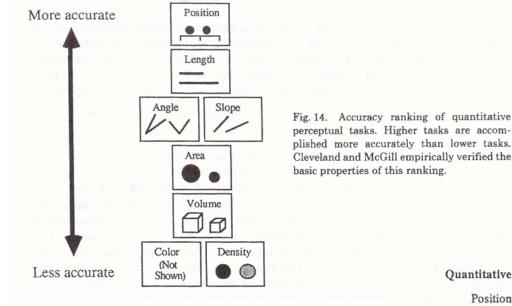


Fig. 13. Area/position presentation of the *Price* and *Mileage* relations. The vertical positioning of the marks reduces the chance that a mark is covered. This technique is called jittering; the vertical positioning does not encode any information.

Effectiveness criteria

- Effectiveness criteria:
 - Determines whether the desired information exploits the capabilities of the output medium and the human visual system
 - But there is no empirically verified theory of human perceptual capabilities... so we build theories based on what is known...

Ranking of perceptual tasks



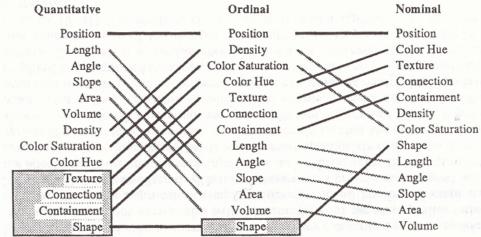


Fig. 15. Ranking of perceptual tasks. The tasks shown in the gray boxes are not relevant to these types of data.

Primitive Graphical Language
Horizontal axis, vertical axis
Line chart, bar chart, plot chart
Color, shape, size, saturation, texture, orientation
Road map, topographic map
Tree, acyclic graph, network
Pie chart, Venn diagram,

Fig. 22. A basis set of primitive graphical languages.

Length and appropriateness of the retinal variables...

- It is often ok to confuse 2 quantitative values close in value, but usually a bigger mistake to misinterpret ordinal or nominal values
- Is area a suitable variable for nominal data?

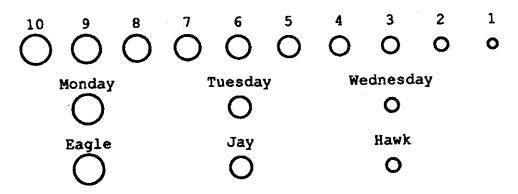


Fig. 16. Analysis of the area task. The top case shows that area is moderately effective for encoding quantitative information. The middle case shows that it is possible to encode ordinal information as long as the step size between areas is large enough so that the values are not confused. The bottom case shows that it is possible to encode nominal information, but people may perceive an ordinal encoding.

Which representations are better?

• Which scatterplot is preferred?

Scatter plot Area/Position	Price position area	<i>Mileage</i> position position
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Fig. 17. Comparison of perceptual tasks for the price/mileage designs.

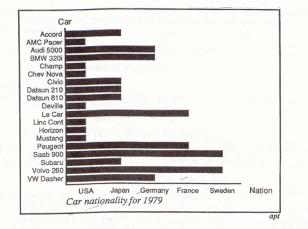


Fig. 11. Incorrect use of a bar chart for the *Nation* relation. The lengths of the bars suggest an ordering on the vertical axis, as if the USA cars were longer or better than the other cars, which is not true for the *Nation* relation.

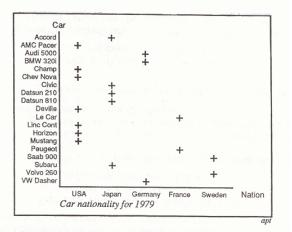


Fig. 12. Correct use of a plot chart for the *Nation* relation. Since bar charts encode ordered domain sets, plot charts are conventionally used to encode nominal domain sets. The ordering of the labels on the axes is ignored.

Which representations is better? (2)

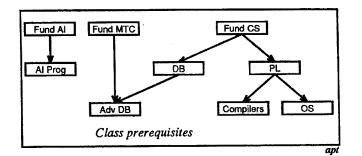
• Which scatterplot is preferred?

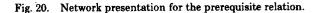
Scatter plot 1	Price position	<i>Mileage</i> position	Weight area
Scatter plot 2	area	position	position

Fig. 18. Example of designs not ordered by the effectiveness ranking.

• General principle: encode more important information more effectively (so if price was more important use position rather than area for price)

Combining position and network representations





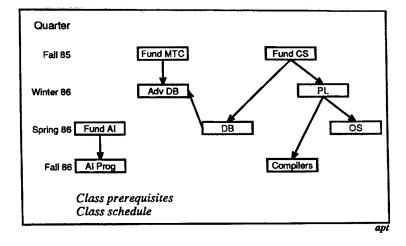


Fig. 19. Composite presentation for the prerequisite and schedule relations. The links encode the prerequisite relationships between computer science classes. The position on the vertical axis encodes the scheduling of the classes. Note that the advanced database class is scheduled before its prerequisite.

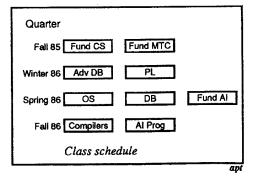


Fig. 21. Vertical-axis presentation for the schedule relation.

Expressiveness of retinal techniques

	Nominal	Ordinal	Quantitative
Size		•	•
Saturation		٠	•
Texture	٠	٠	
Color	٠	*	
Orientation	•	•	
Shape	٠		

Fig. 25. Expressiveness of retinal techniques. The – indicates that size and saturation should not be used for nominal measurements because they will probably be perceived to be ordered. The * indicates that the full color spectrum is not ordered. However, parts of the color spectrum are ordinally perceived [23].

Encoding more than 3 variables....

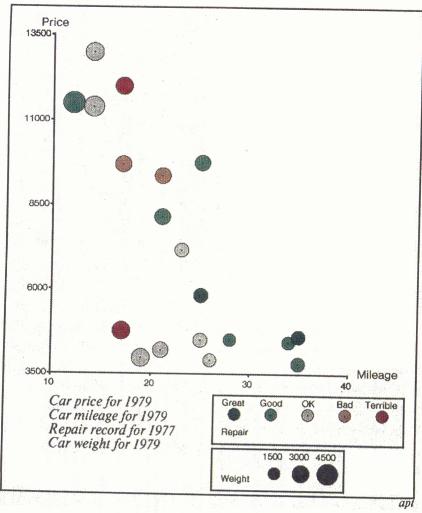


FIGURE 30

Color scatter plot for four automobile relations. The design expresses the relations only if the application permits the details about the cars to be omitted.

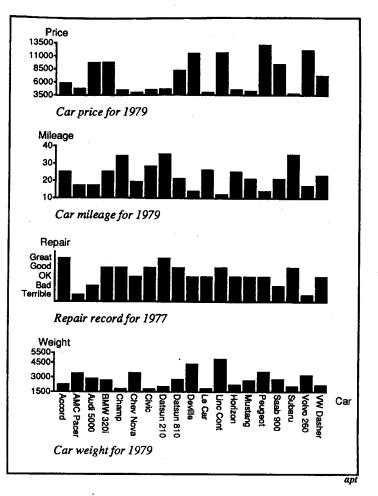


Fig. 31. Aligned bar chart of automobile data. This diagram shows the details about the car domain set. However, the general relationships are not so easy to see as in the scatter plot design.

Polaris system

Visualization of Multi-dimensional Relational Databases

Motivation

- Over the past few years, multi-dimensional large databases have become very common in a variety of domains/applications
- A major challenge is to extract meaning, discover structure, find patterns and derive causal relationships
- Exploratory analysis is one of:
 - Hypothesis
 - Experiment
 - Discovery
 - Iterate analysts need to rapidly change both *what* data they are viewing and *how* they are viewing that data

Polaris

- Polaris creates visualizations using a set of tables based on fields of the database
- The use of tables to organize multiple graphs is a well known technique by statisticians in their analysis of data
- Each table has layers and panes (each pane may have a different graphic)

Relational databases

- In a relational database: Each row in a table corresponds to a basic entity or fact and each column represents a property of that entity
- A row is referred to as a tuple, and a column is a field
- A single relational database will contain many heterogeneous but interrelated tables
- Fields in a database may be nominal, ordinal or quantitative
- Fields in a database can be partitioned in 2 types:
 - Dimensions (e.g. name or type of a product, independent variables)
 - Measures (e.g. price or size, dependent variables)
- Polaris assumed that all nominal fields are dimensions and all quantitative fields are measures

Requirements for visualizing multidimensional databases

- 1. Data dense displays
- 2. Multiple display types
- 3. Exploratory interface
- Polaris addresses these requirements by providing an interface for rapidly and incrementally generating table-based displays
 - Each table axis may have multiple nested dimensions
 - Each table entry (or pane) contains a set of records that are visually encoded as a set of marks to create a graphic

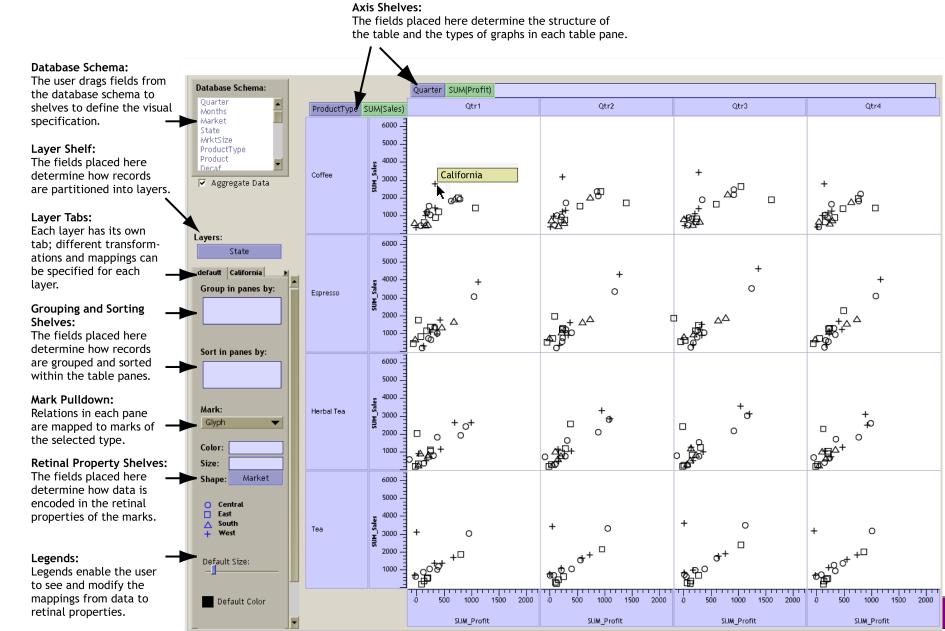


- Multivariate multiple dimensions of the data can be explicitly encoded in the structure of the table
- Comparative tables generate "small multiple displays of information" [Tufte] which can be compared, exposing patterns and trends across dimensions of the data
- Tables are familiar...



- Interface approach is to drag and drop fields from the database schema onto shelves throughout the display
- Interface supports "brushing"
- Some screenshots...

Polaris User Interface



Ordinal-Ordinal

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tabase Schema:		Position									
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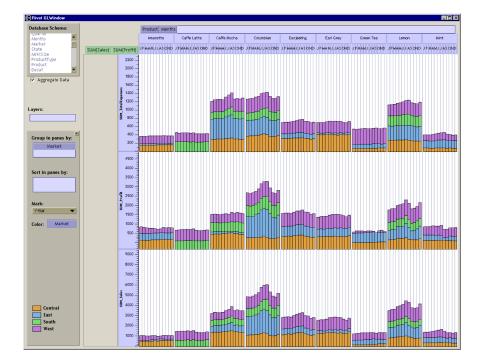
Ordinal-Ordinal

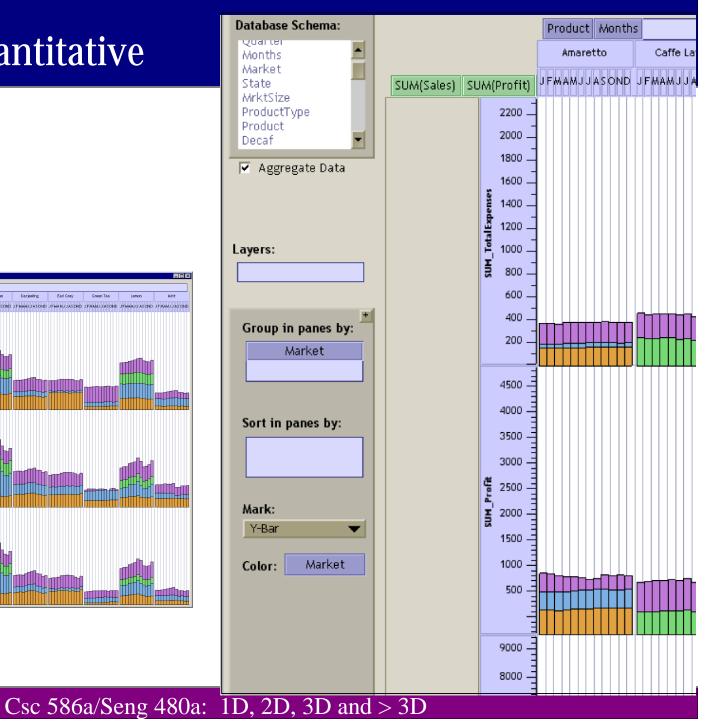
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Database Schema:		Position	
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	S.F.	•	•
ayers:	S.D.	•	•
	Pit.	•	•
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	Oak.	•	•
	N.Y.	٠	•
Sort in panes by:	Mon.	•	•
	Min.	•	•
Mark:	Mil.	•	•
Circle	L.A.	•	•
Color: SUM(Errors) Size: SUM(Errors)	к.с.	•	•
Size: SUM(Errors)	Hou.	٠	•
	Det.	•	•
54.0	Cle.	•	
27.0 , 21, 31, and			

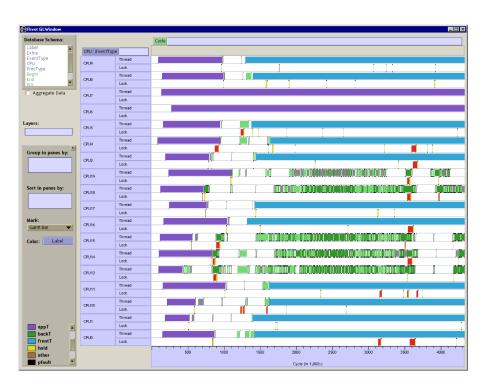
Csc 586a/Seng 480a: 27.0

Ordinal-Quantitative





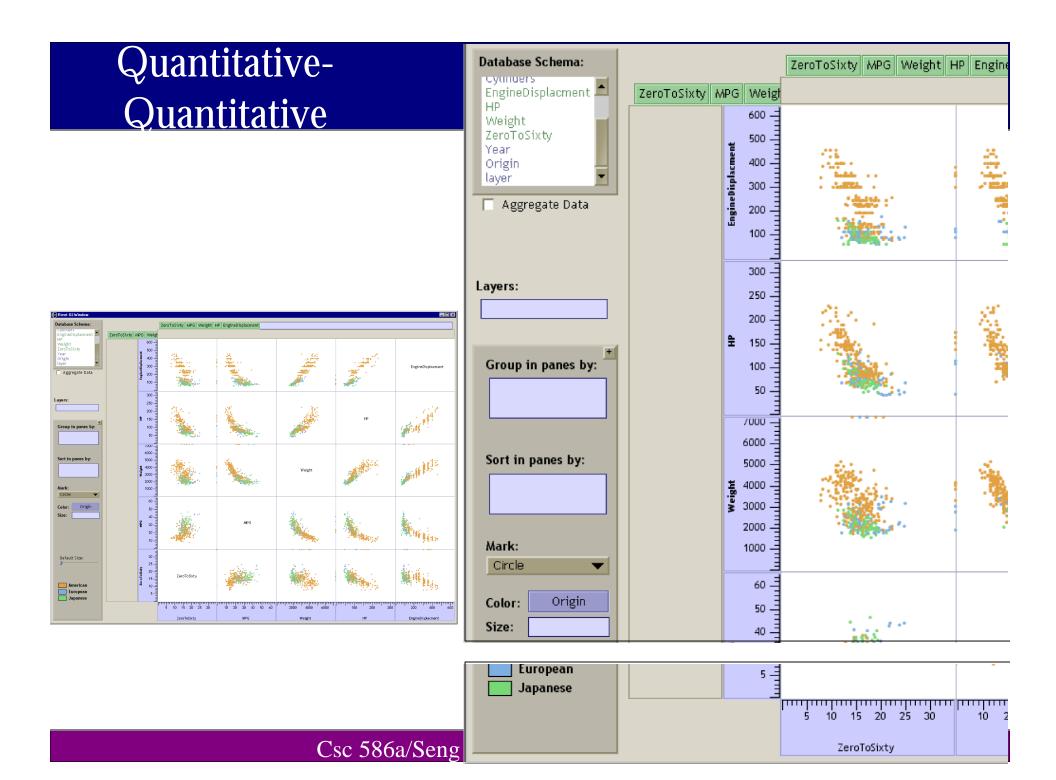
Ordinal-Quantitative



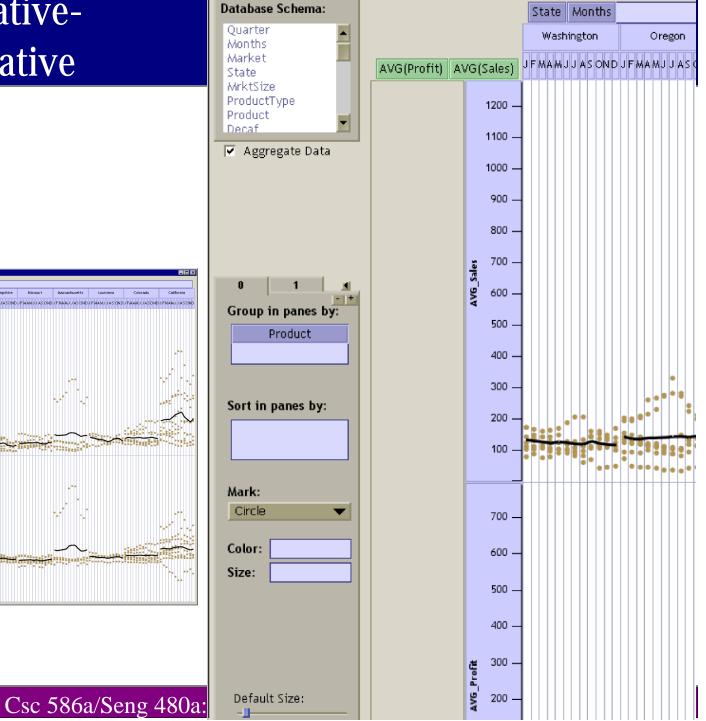
Database Schema:			Cycle
Label Extra	CPU EventType		
EventType CPU	CPU9	Thread	
ProcType	CPUY	Lock	
Begin End	CPU8	Thread	
PID	CF00	Lock	
🗖 Aggregate Data	CPU7	Thread	
	CF07	Lock	
	CPU6	Thread	
	CFOO	Lock	•
Layers:	CPU5	Thread	
		Lock	
	CPU4	Thread	
Group in panes by:		Lock	
Group in panes by:	CPU2	Thread	
		Lock	
	CPU19	Thread	
		Lock	
Sort in panes by:	CPU18	Thread	
	Croit	Lock	
	CPU17	Thread	
		Lock	
Mark:	CPU16	Thread	
Gantt Bar 🔍	51010	Lock	
Color: Label	CPU15	Thread	
	3015	Lock	

appi 🔺		LUCK	
backT 👘	CPU0	Thread	
frontT	CPOU	Lock	
hold		·	
other			500
📕 pfault 📃			

Csc 586a/Seng 480a:



Quantitative-Quantitative



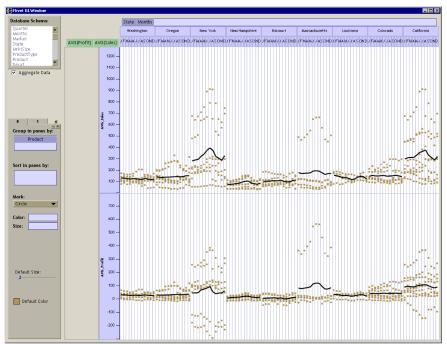


Table Algebra: Ordinal

Ordinal fields: Quarter, Months, Product

Quantitative fields: Profit, Sales

0 = Quarter = {Qtr1, Qtr2, Qtr3, Qtr4} = Qtr1 + Qtr2 + Qtr3 + Qtr4:

Qtr1	Qtr2	Qtr3	Qtr4
------	------	------	------

0 + 0 = Quarter + Product = {Qtr1, Qtr2, Qtr3, Qtr4, Coffee, Espresso, Herbal Tea, Tea}:

O×O = Quarter × Product = {(Qtr1,Coffee), (Qtr1,Espresso), (Qtr1,Hebral Tea), (Qtr1, Tea), (Qtr2, Coffee) ... (Qtr4, Tea)}:

Qtr1			Qtr2				Qtr3				Qtr4				
Coffee	Espresso	Herbal Tea	Теа	Coffee	Espresso	Herbal Tea	Tea	Coffee	Espresso	Herbal Tea	Tea	Coffee	Espresso	Herbal Tea	Tea

0/0 = Quarter / Month = {(Qtr1, Jan), (Qtr1, Feb), (Qtr1, Mar), (Qtr2, Apr), (Qtr2, May) ... (Qtr4, Dec)}:

	Qtr1			Qtr2			Qtr3		Qtr4			
Jan	Feb	Mar	Apr	May	Jun	Jul	Aug	Sep	Oct	Nov	Dec	

+ Concatenate

- x Cross
- / Nest

Table Algebra: Quantitative

Q = Profit = {Profit}:

 Profit (in thousands)

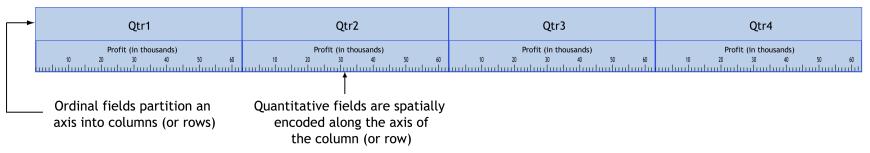
 10
 20
 30
 40
 50
 60

Q + Q = Profit + Sales = {Profit, Sales}:

 Profit (in thousands)
 Sales

 10
 20
 30
 40
 50
 60
 50
 100
 150
 200
 250
 300

O×Q = Quarter × Profit = {(Qtr1,Profit), (Qtr2, Profit), (Qtr3, Profit), (Qtr4, Profit)}:



+ Concatenatex Cross/ Nest



- Visualization barrier on 3 or 4 dimensions of data...
- But interaction and permutations can be used to increase the number of relationships we can see
- Small multiple displays of information that are linked can reveal important relationships
- Appropriateness of the retinal variables to nominal, ordinal and quantitative data
- Some images can be automatically generated using tools such as Polaris and other tools from Visual Inxight
- Example: <u>http://www.inxight.com/products/st_viewer/</u>