

# An Introduction to Information Visualization Techniques for Exploring Large Database

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Fall 2005

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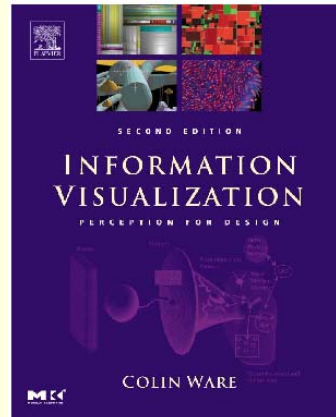
## Visual Perception

Class 2, Part A

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## Book

### ■ Dr. Colin Ware



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## Semiotics of Graphics

- Is visualization a science?
- Sensory vs. Arbitrary symbols
- Sensory representation
  - Understanding without training
  - Sensory immediacy
  - Cross-cultural validity
- Arbitrary representation
  - Hard to learn
  - Easy to forget
  - Embedded in culture and applications
  - Formally powerful
  - Capable of rapid change
- Most visualizations are hybrids!

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# Visual Perception

- What is visual perception?
  - process of knowing or being aware of information through the eyes.
  - process of acquiring, interpreting, selecting, and organizing sensory information.

<http://en.wikipedia.org/wiki/Perception>

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# Related Disciplines

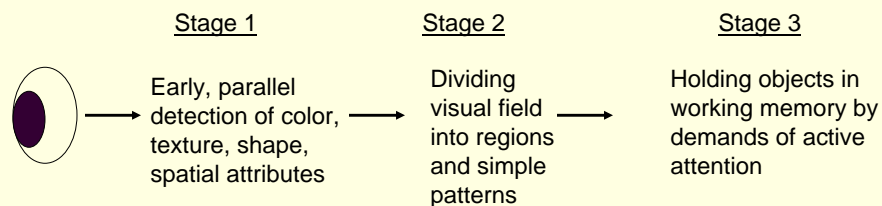
- Psychophysics
  - Applying methods of physics to measuring human perceptual systems
    - How fast must light flicker until we perceive it as constant?
    - What change in brightness can we perceive?
- Cognitive psychology
  - Understanding how people think, here, how it relates to perception

- Dr. John Stasko, Slides of CS7500 at Gatech

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# One Simple Model of Perceptual Processing

- Three stage process
  - Parallel extraction of low-level properties of scene
  - Pattern perception
  - Sequential goal-directed processing



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## Stage 1 - Low-level, Parallel

- Neurons in eye & brain responsible for different kinds of information
  - Orientation, color, texture, movement, etc.
- Arrays of neurons work in parallel
- Occurs “automatically”
- Rapid
- Information is transitory, briefly held in iconic store
- Bottom-up data-driven model of processing
- Often called “pre-attentive” processing

- Dr. John Stasko, Slides of CS7500 at Gatech

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## Stage 2 – Pattern Perception

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- Slow serial processing
- Involves working and long-term memory
- A combination of bottom-up feature processing and top-down attentional mechanisms
- Different visual systems for object recognition and visually guided motion

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## Stage 3 – Sequential Goal-Directed

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- A few objects are constructed from the available patterns to provide answers to visual queries
- Top-down attention-driven model of processing
- Slow serial processing

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## Preattentive Processing

- The most important contribution vision science make to data visualization is that:  
**A limited set of visual properties can be detected very rapidly and accurately by the low-level visual system**
- Tasks that can be performed on large multi-element displays in less than 200 to 250 milliseconds (msec) are considered **preattentive**. (Eye movements: 200 msec)

<http://www.csc.ncsu.edu/faculty/healey/PP/index.html>

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## Count 3s

1281768756138976546984506985604982826762  
9809858458224509856458945098450980943585  
9091030209905959595772564675050678904567  
8845789809821677654876364908560912949686

1281768756138976546984506985604982826762  
9809858458224509856458945098450980943585  
9091030209905959595772564675050678904567  
8845789809821677654876364908560912949686

- Dr. John Stasko, Slides of CS7500 at Gatech

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## Tasks

- Target detection
  - Is something there?
- Boundary detection
  - Can the elements be grouped?
- Counting
  - How many elements of a certain type are present?

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## Preattentive Features

- |   |   |
|---|---|
| <ul style="list-style-type: none"><li>■ Form<ul style="list-style-type: none"><li>■ Line orientation</li><li>■ Line length</li><li>■ Line width</li><li>■ Line collinearity</li><li>■ Size</li><li>■ Curvature</li><li>■ Spatial grouping</li><li>■ Blur</li><li>■ Added marks</li><li>■ Numerosity</li></ul></li></ul> | <ul style="list-style-type: none"><li>■ Color<ul style="list-style-type: none"><li>■ Hue</li><li>■ Intensity</li></ul></li><li>■ Motion<ul style="list-style-type: none"><li>■ Flicker</li><li>■ Direction of motion</li></ul></li><li>■ Spatial Position<ul style="list-style-type: none"><li>■ 2D position</li><li>■ Stereoscopic depth</li><li>■ Convex/concave shape from shading</li></ul></li></ul> |
|---|---|

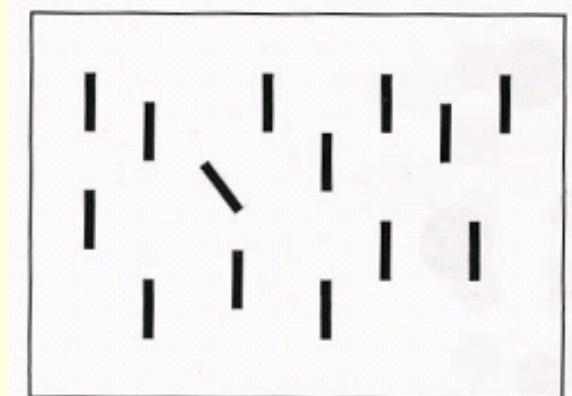
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## Example

- Find the distinct one

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## Orientation

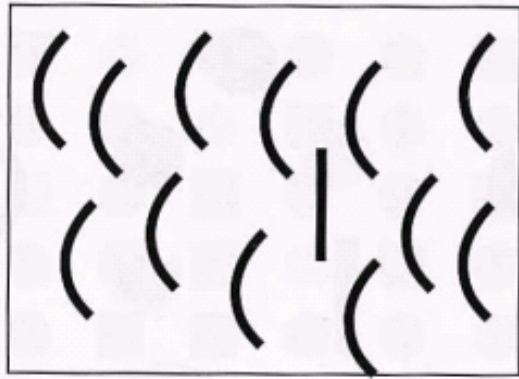


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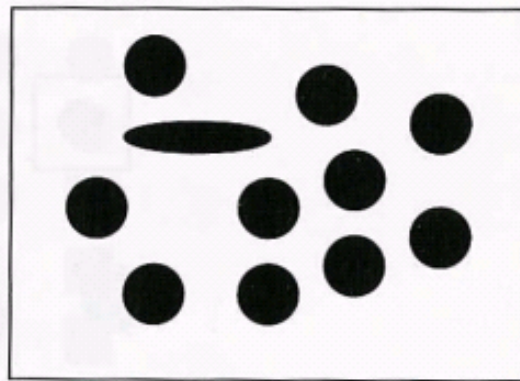
## Curved/Straight



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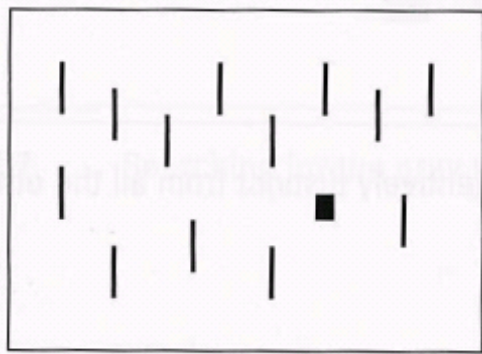
## Shape



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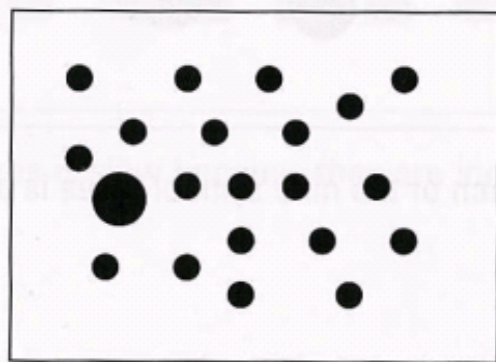
## Shape



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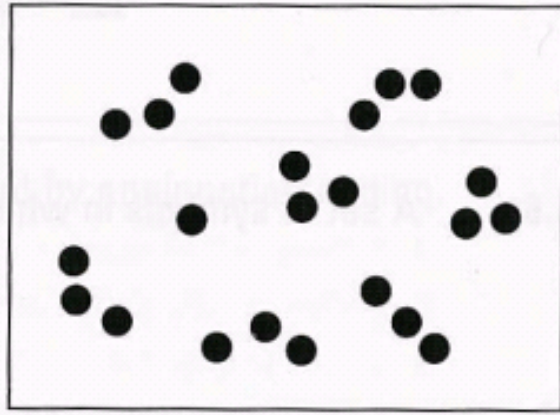
## Size



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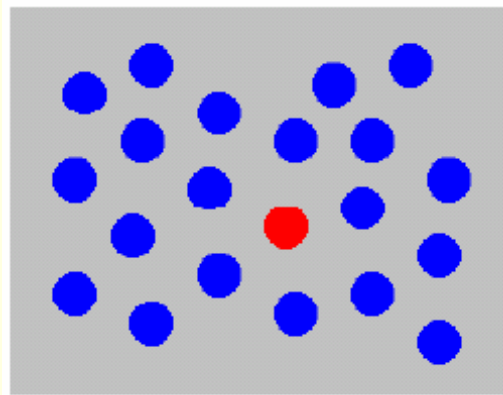
## Number



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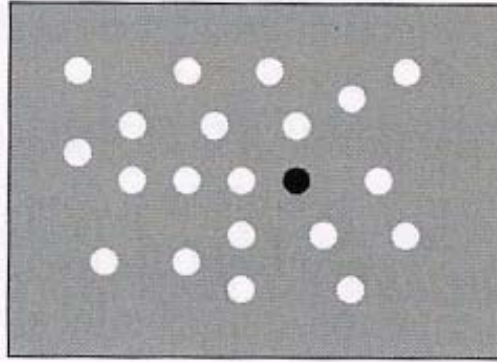
## Hue



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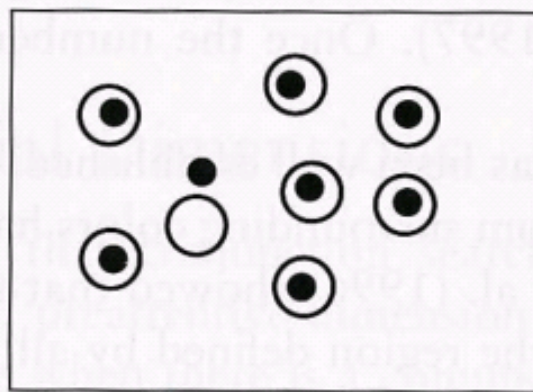
## Gray/Value



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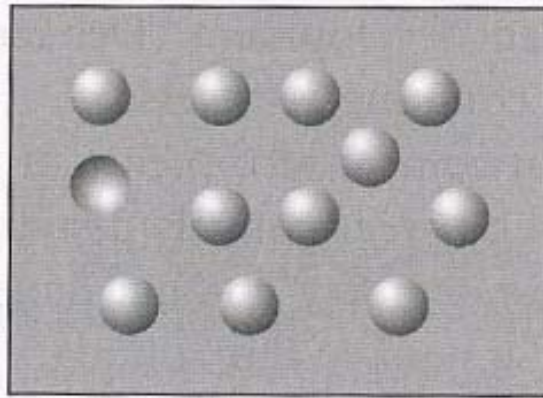
## Enclosure



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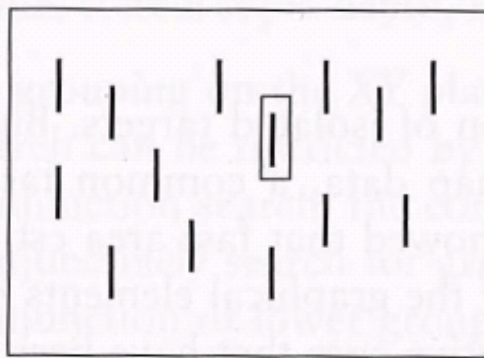
## Covexity/Concavity



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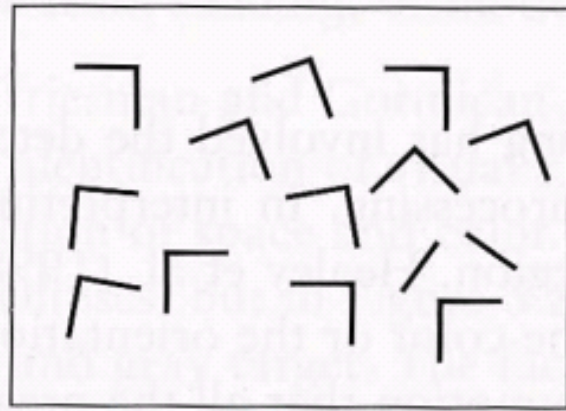
## Addition



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## Juncture

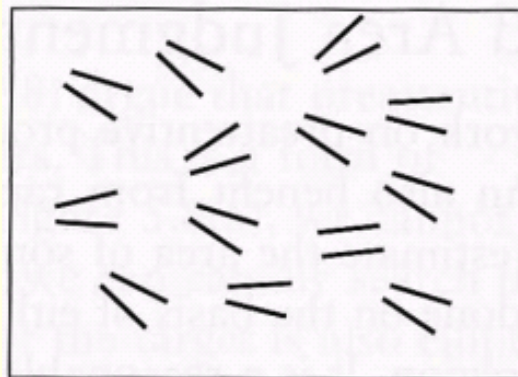


Ware 2004

Not!

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## Parallelism

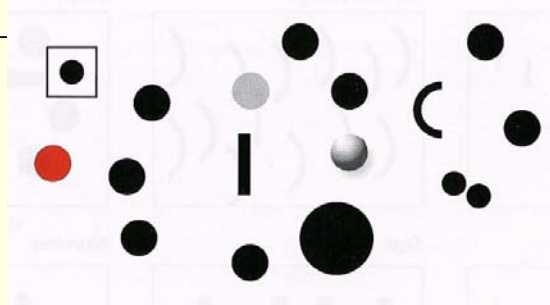


Ware 2004

Not!

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## Multiple Symbol Types



- Preattentive symbols become less distinct as the variety of distractors increase
- Two factors
  - Degree of difference of target from nontargets
  - Degree of difference of nontargets from each other

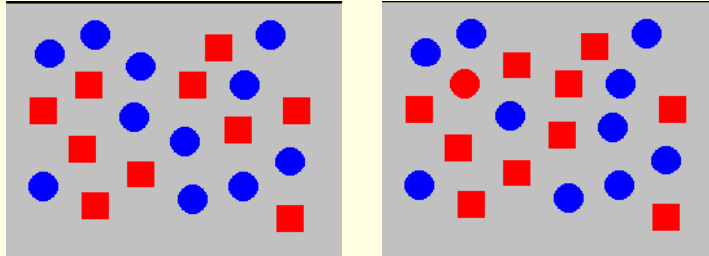
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## Example

- Determine if a red circle is present

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## Hue and Shape



- Cannot be done preattentively
- Must perform a sequential search
- Conjunction of features (shape and hue) causes it

- Dr. John Stasko, Slides of CS7500 at Gatech

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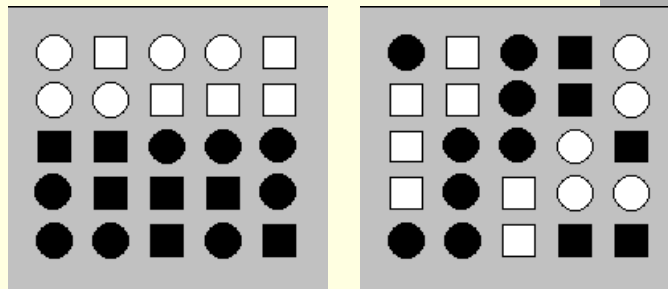
## Example

- Is there a boundary in the display?

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## Fill and Shape



- Left can be done preattentively since each group contains one unique feature
- Right cannot (there is a boundary!) since the two features are mixed (fill and shape)

- Dr. John Stasko, Slides of CS7500 at Gatech

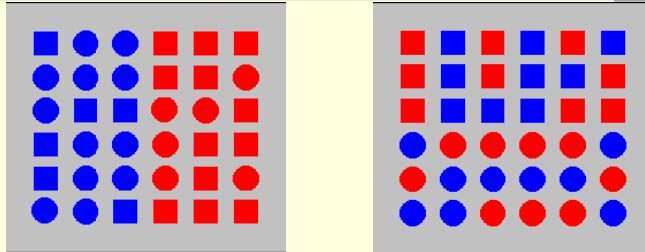
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## Example

- Is there a boundary in the display?

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## Feature Hierarchy



- Left: Boundary detected preattentively based on hue regardless of shape
- Right: a horizontal form boundary cannot be preattentively identified when hue varies randomly in the background
- Visual systems favour color over shape

<http://www.csc.ncsu.edu/faculty/healey/PP/index.html>

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## Discussion

- What can we do using preattentive features?

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# Multi-Dimensional Visualization (con.)

Class 2, Part B

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## Classification

- Geometric techniques
  - *Scatterplot matrices, parallel coordinates, landscapes, ...*
- Hierarchical techniques
  - *Dimensional stacking, worlds-within-worlds, ...*
- Icon-based techniques
  - *Star glyphs, stick figures, shape-coding, color icons, chernoff faces, ...*
- Pixel-oriented techniques
  - *Recursive pattern, circle segments, spiral, axes techniques,...*
- Table-based techniques
  - *HeatMap, tableLens*

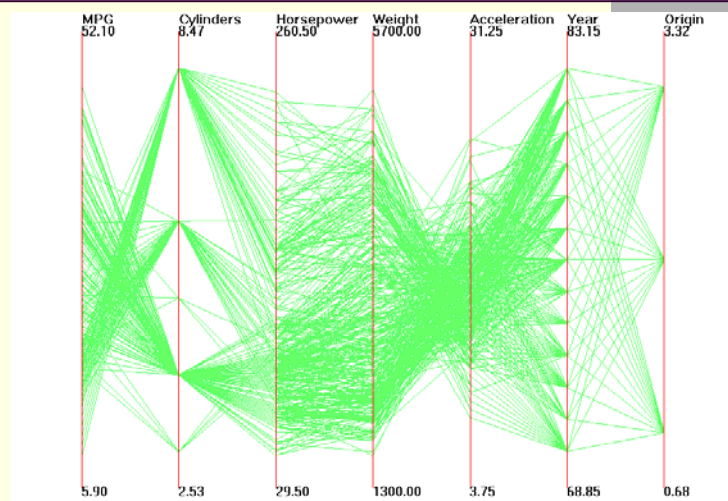
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## Geometric Techniques

- Basic idea: Visualization of geometric transformations and projections of data
  - Scatterplot-matrices [And 72, Cle 93]
  - Parallel coordinates [Ins 85, ID 90]
  - Star coordinates [Kan 2000]
  - Landscapes [Wis 95]
  - Projection Pursuit Techniques [Hub 85]
  - Projection Views [FB 94, STDS 95]
  - Hyperslice [WL 93]

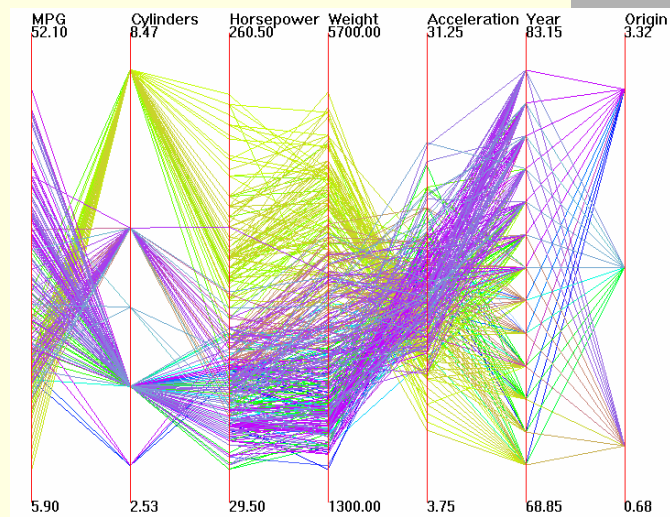
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## Parallel Coordinates



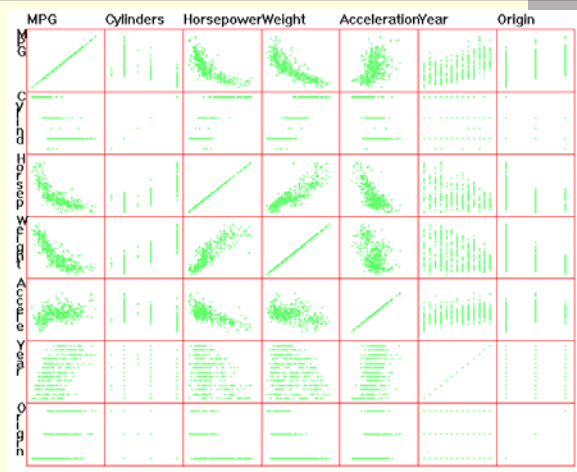
40

## Parallel Coordinates



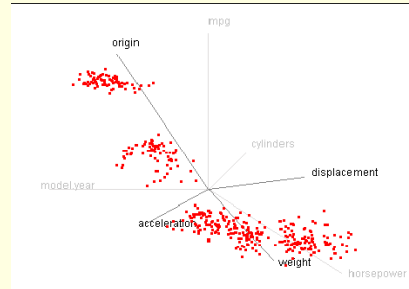
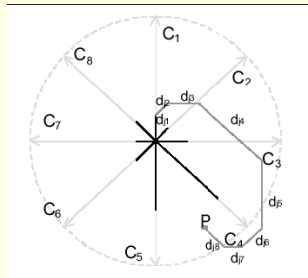
41

## Scatterplot-Matrices



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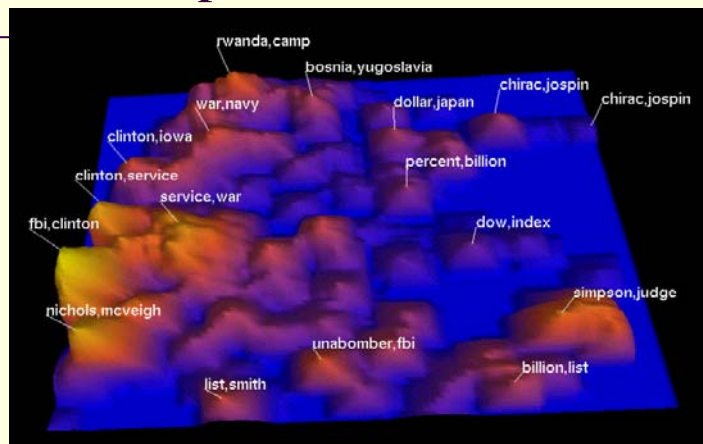
## Star Coordinates



E. Kandogan, "Star Coordinates: A Multi-dimensional Visualization Technique with Uniform Treatment of Dimensions", InfoVis 2000 Late-Breaking Hot Topics, Oct. 2000

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## Landscapes



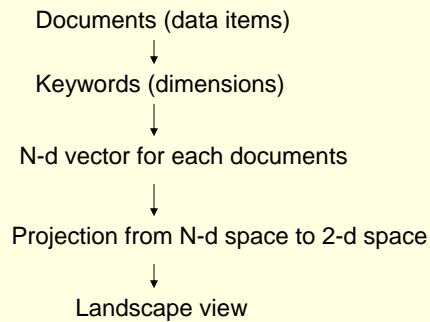
Hot topics of a news collection

L. Nowell, E. Hetzler, and T. Tanasse. "Change Blindness in Information Visualization: A Case Study". Infovis 2001

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# Landscapes

## ■ How was the figure generated?



Wise, J., Thomas, J., et al. "Visualizing the Non-Visual: Spatial Analysis and Interaction with Information from Text Documents", Infovis 95

45

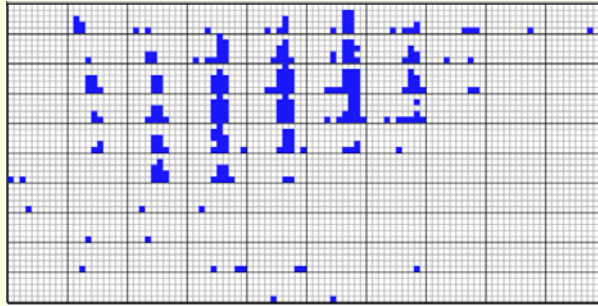
# Hierarchical Techniques

## ■ Basic ideas: Visualization of the data using a hierarchical partitioning into subspaces

- Dimensional Stacking [LWW90]
- Worlds-within-Worlds [FB 90a/b]
- Treemap [Shn 92, Joh 93]
- Cone Trees [RMC 91]
- InfoCube [RG93]

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## Dimensional Stacking

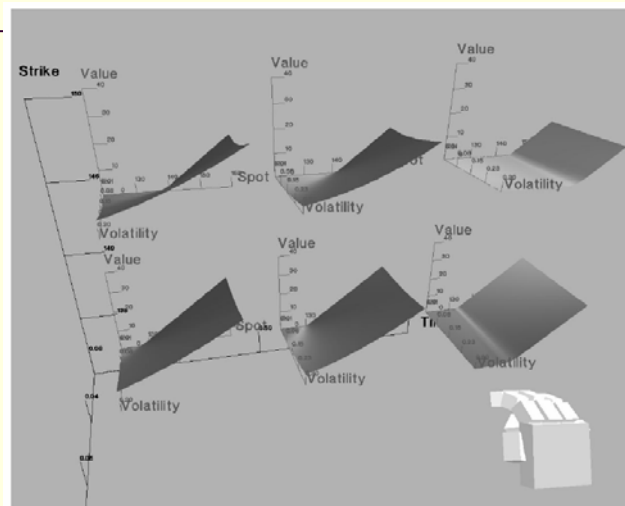


visualization of oil mining data with longitude and latitude mapped to the outer x-, y- axes and ore grade and depth mapped to the inner x-, y- axes

M. Ward, Worcester Polytechnic Institute

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## Worlds within Worlds



Feiner S., Beshers C.: 'World within World: Metaphors for Exploring <sup>48</sup> n-dimensional Virtual Worlds', Proc. UIST, 1990, pp. 76-83.



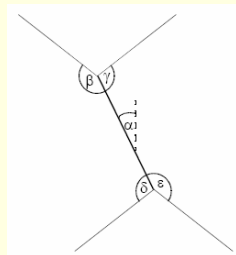
## Icon-Based Techniques

- Basic idea: visualization of the data values as features of icons
  - Chernoff-Faces [Che 73, Tuf 83]
  - Stick Figures [Pic 70, PG88]
  - Shape Coding [Bed 90]
  - Color Icons [Lev 91, KK94]
  - TileBars [Hea 95]

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## Stick Figures

- The mapping
  - Two attributes - display axes
  - Others - angle and/or length of limbs
- Texture patterns in the visualization show certain data characteristics

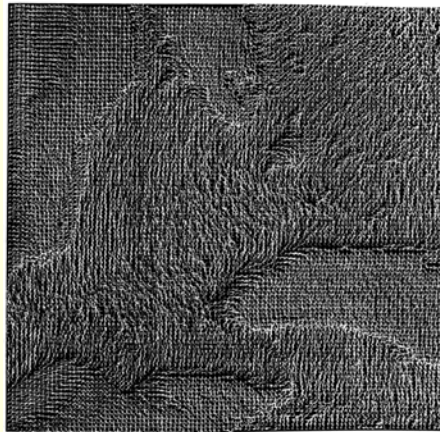


Pickett R. M., Grinstein G. G.:  
'Iconographic Displays for Visualizing  
Multidimensional Data', Proc. IEEE Conf.  
on Systems, Man and Cybernetics, 1988,  
pp. 514-519.

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Stick figure icon

## Stick Figures



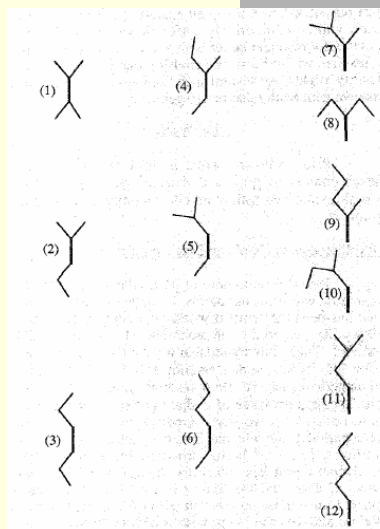
5-dim. image  
data from the  
great lake  
region

<http://ivpr.cs.uml.edu/gallery/>  
G. Grinstein, University of Massachusetts at Lowell

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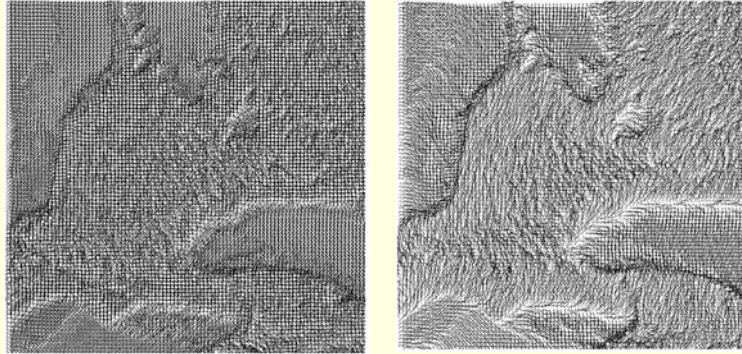
## Stick Figures

- Stick figure icon family
- Try all different mappings
  - $12 \times 5! = 1440$  pictures
  - Movie



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## Stick Figures



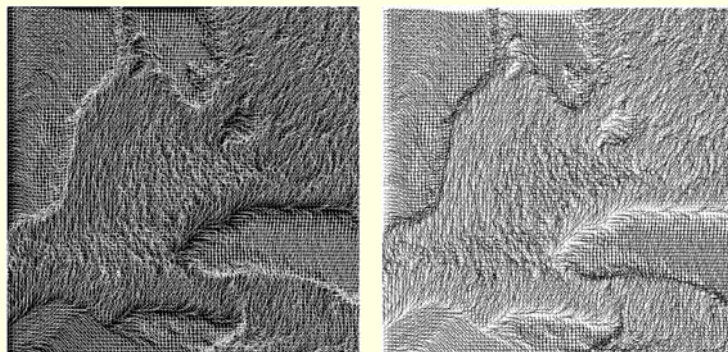
The same dataset, different mapping

<http://ivpr.cs.uml.edu/gallery/>

G. Grinstein, University of Massachusetts at Lowell

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## Stick Figures



Black background

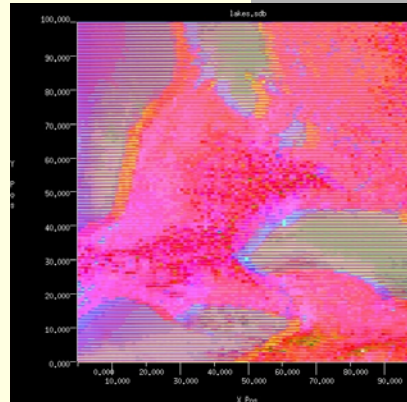
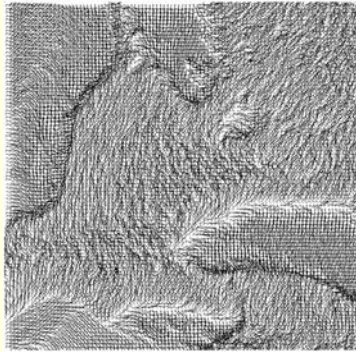
White background

<http://ivpr.cs.uml.edu/gallery/>

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## More

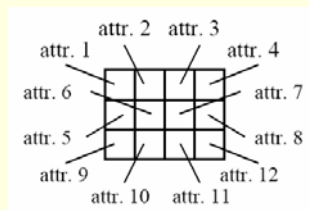


<http://ivpr.cs.uml.edu/gallery/>  
G. Grinstein, University of Massachusetts at Lowell

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## Shape Coding

- Data item - small array of fields
- Each field - one attribute value
- Arrangement of attribute fields (e.g., 12-dimensional data):

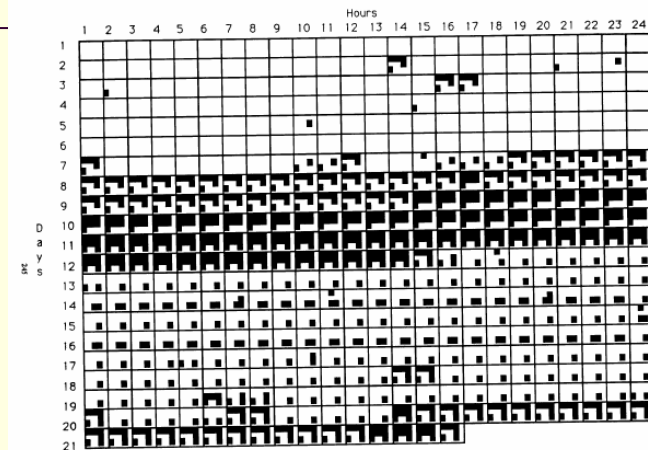


- Sort the arrays

The figure is taken from Dr. D. Keim's tutorial notes in Infovis 00

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## Shape Coding



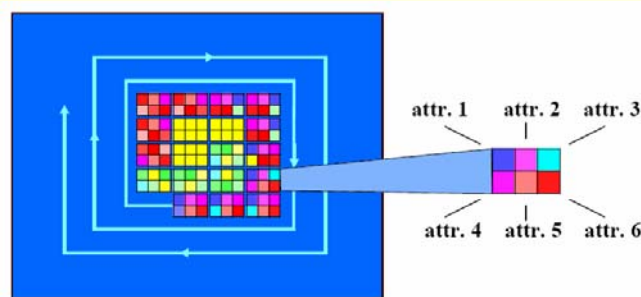
Day by Hour: Thirteen Parameters of Magnetosphere and Solar Wind Data

Beddow J.: 'Shape Coding of Multidimensional Data on a Microcomputer Display', Visualization '90, 1990, pp. 238-246.

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## Color Icons [Lev 91, KK94]

- Data item - color icon (arrays of color fields)
- Each field - one attribute value
- Arrangement is query-dependent (e.g., spiral)

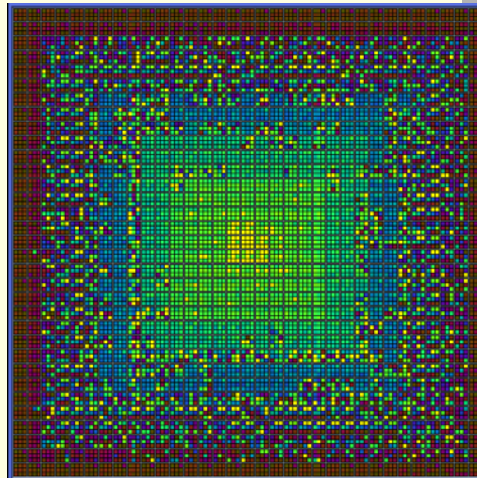


6 dimensional dataset

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The figure is taken from Dr. D. Keim's tutorial notes in Infovis 00.

## Color Icons



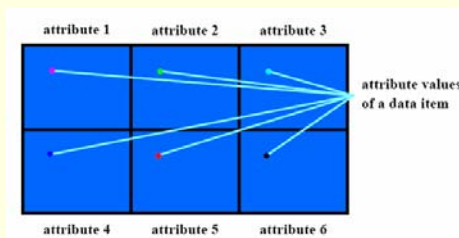
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The figure is taken from Dr. D. Keim's tutorial notes in Infovis 00

## Pixel-Oriented Techniques

### ■ Basic idea

- Each attribute value - one colored pixel (value ranges -> fixed colormap)
- Attribute values for each attribute are presented in separate subwindows

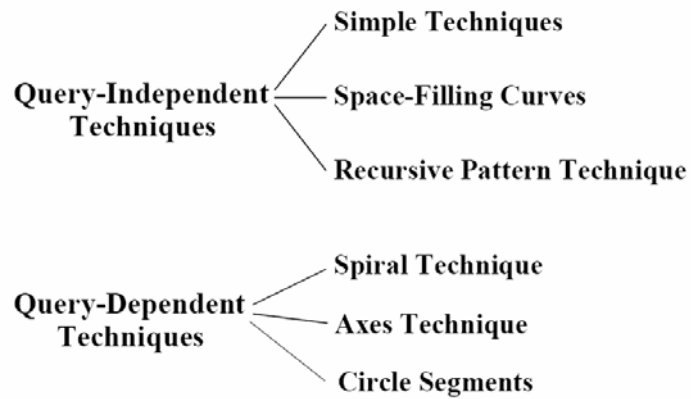


60

The figure is taken from Dr. D. Keim's tutorial notes in Infovis 00

## Pixel-Oriented Techniques

### Overview [Kei 96]

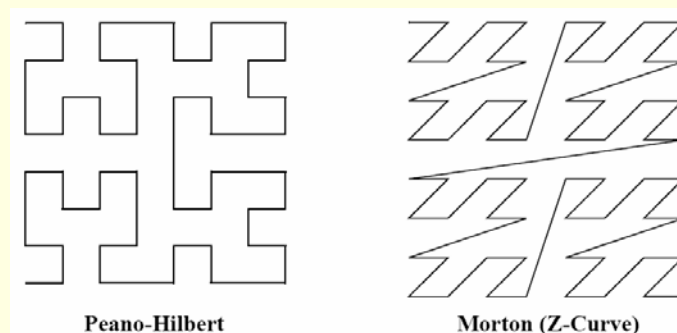


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The figure is taken from Dr. D. Keim's tutorial notes in Infovis 00

## Query-Independent Techniques

### ■ Space-Filling Curve Arrangements



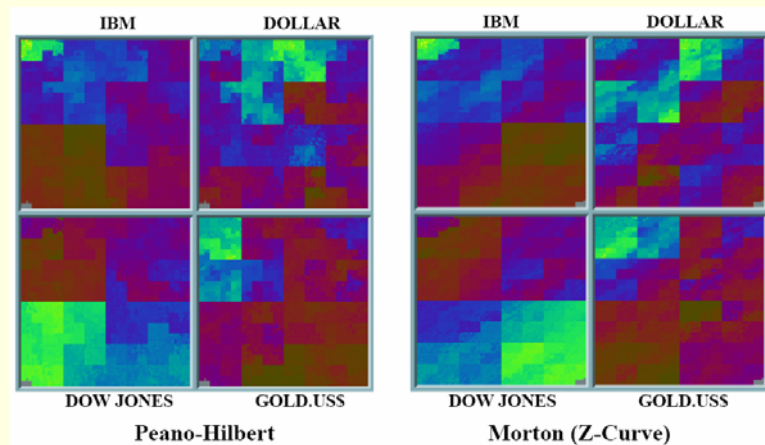
62

The figure is taken from Dr. D. Keim's tutorial notes in Infovis 00



## Query-Independent Techniques

### ■ Space-Filling Curve Arrangements

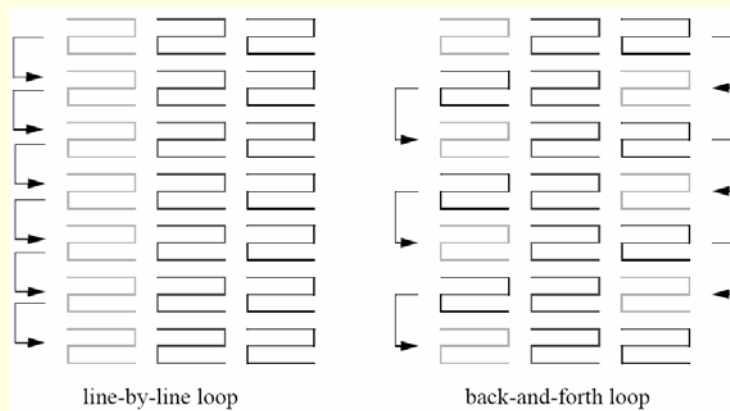


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The figure is taken from Dr. D. Keim's tutorial notes in Infovis 00

## Query-Independent Techniques

### ■ Recursive pattern arrangements



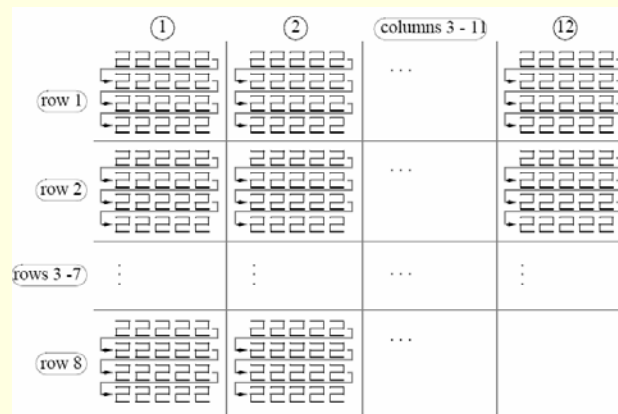
64

The figure is taken from Dr. D. Keim's tutorial notes in Infovis 00



## Query-Independent Techniques

### ■ Recursive pattern arrangements

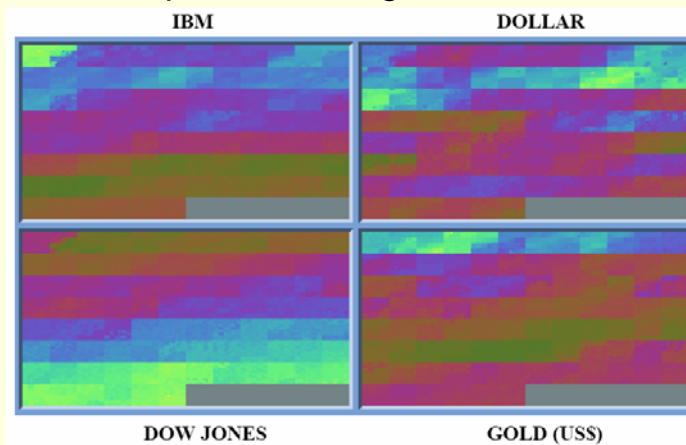


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The figure is taken from Dr. D. Keim's tutorial notes in Infovvis 00

## Query-Independent Techniques

### ■ Recursive pattern arrangements



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The figure is taken from Dr. D. Keim's tutorial notes in Infovvis 00

## Query-Dependent Techniques

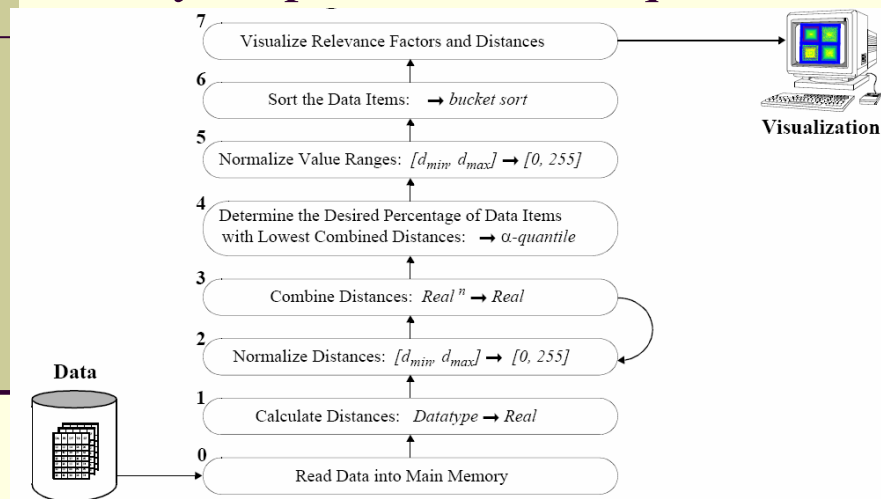
### ■ Basic idea:

- data items ( $a_1, a_2, \dots, a_m$ ) & query ( $q_1, q_2, \dots, q_m$ )
- distances ( $d_1, d_2, \dots, d_m$ )
- extend distances by overall distance ( $d_{m+1}$ )
- determine data items with lowest overall distances
- map distances to color (for each attribute)
- visualize each distance value  $d_i$  by one colored pixel

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The slide is taken from Dr. D. Keim's tutorial notes in Infovis 00

## Query-Dependent Techniques

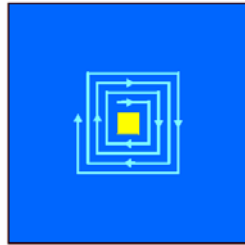


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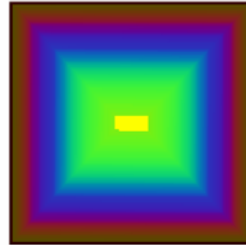
The figure is taken from Dr. D. Keim's tutorial notes in Infovis 00

## Query-Dependent Techniques

### ■ Spiral technique []



Arrangement



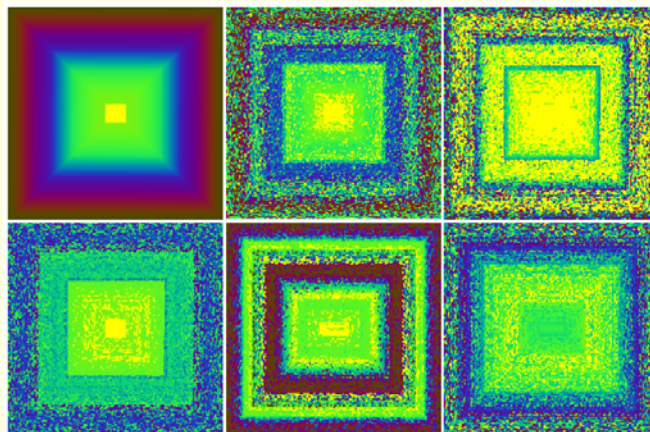
The  $m+1$  dimension  
(overall distance)

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The figure is taken from Dr. D. Keim's tutorial notes in Infovis 00

## Query-Dependent Techniques

### ■ Spiral technique []

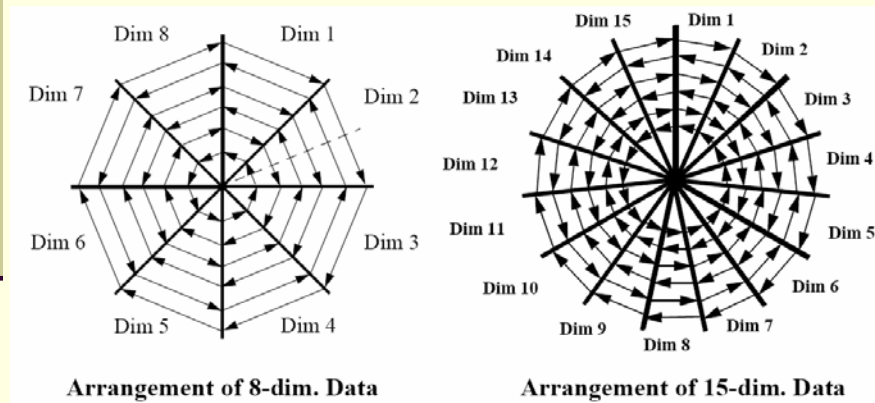


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The figure is taken from Dr. D. Keim's tutorial notes in Infovis 00

## Pixel-Oriented Technique

### ■ Circle segments

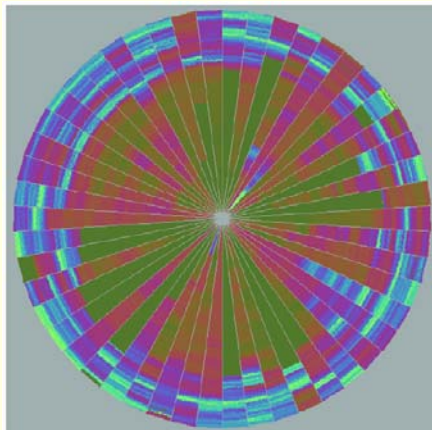


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The figure is taken from Dr. D. Keim's tutorial notes in Infovis 00

## Pixel-Oriented Technique

### ■ Circle segments

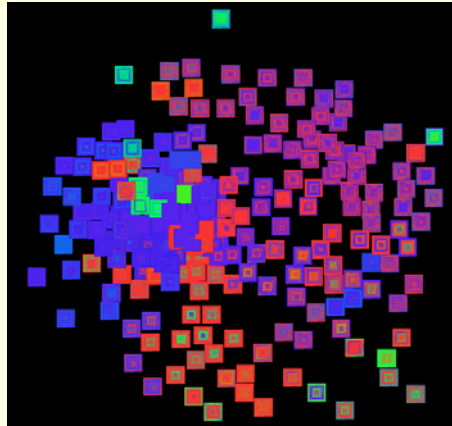


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The figure is taken from Dr. D. Keim's tutorial notes in Infovis 00

## Pixel-Oriented Technique

- Value and Relation displays [yang:2004]



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## Table-Based Techniques

- Basic idea: Visualization that improves the existing spreadsheet table format
  - HeatMap
  - Table Lens [RC 94]

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## Input Data Table

### Example Data

- 16 Variables
- 98 Records

One Record

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## Detail Data Table View

### Example Record

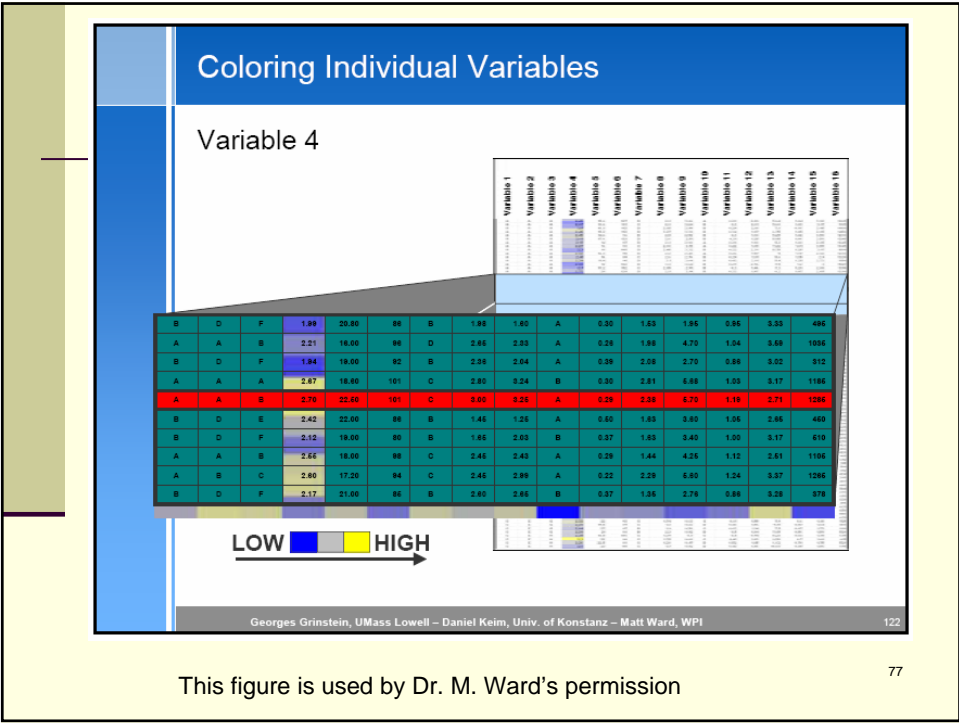
Variable 1	Variable 2	Variable 3	Variable 4	Variable 5	Variable 6	Variable 7	Variable 8	Variable 9	Variable 10	Variable 11	Variable 12	Variable 13	Variable 14	Variable 15	Variable 16
B	D	F	1.99	20.85	96	B	1.88	1.83	A	0.35	1.63	1.95	0.85	0.35	486
A	A	B	2.21	18.65	96	D	2.85	2.83	A	0.25	1.89	4.70	1.04	0.59	1096
B	D	F	1.94	18.65	92	B	2.94	2.84	A	0.39	2.08	2.79	0.88	0.02	912
A	A	A	2.87	18.40	101	C	2.85	0.24	B	0.35	2.81	6.59	1.03	0.17	1185
A	A	B	2.70	22.19	99	C	0.85	3.23	A	0.29	2.39	4.73	1.19	2.71	1285
B	D	C	2.42	22.65	86	B	1.46	1.25	A	0.55	1.65	3.65	1.06	2.46	460
B	D	F	2.12	18.05	86	B	1.65	2.63	B	0.37	1.63	0.45	1.00	0.17	618
A	A	B	2.65	18.05	98	C	2.45	2.43	A	0.28	1.44	4.25	1.12	2.61	1190
A	B	C	2.80	17.25	84	C	2.46	2.89	A	0.22	2.29	6.90	1.24	0.37	1086
B	D	F	3.17	21.00	85	B	2.85	3.65	B	0.37	1.35	2.75	0.89	0.28	979

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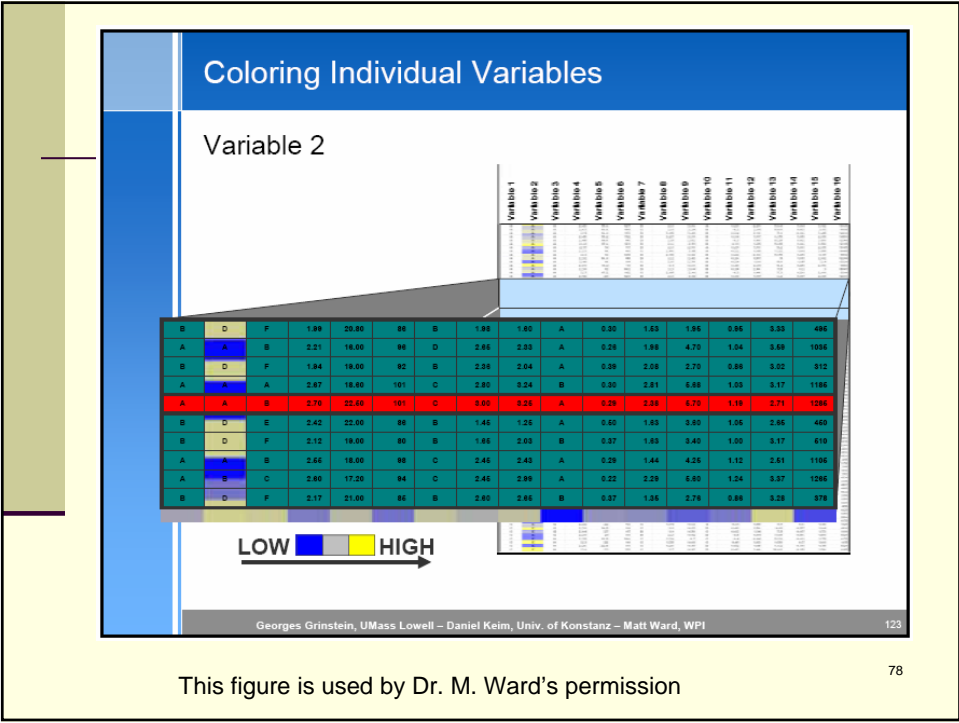
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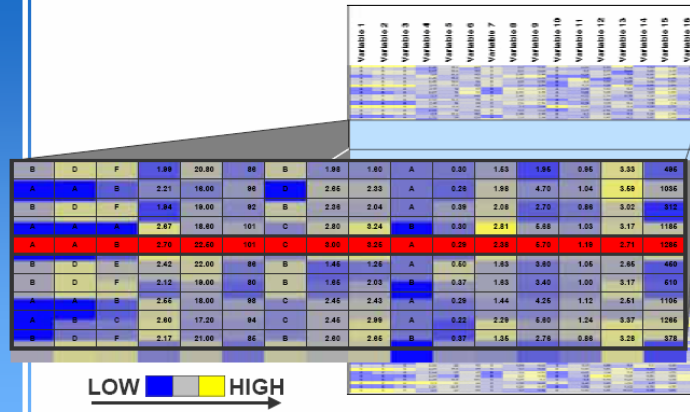


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## Transformation to HeatMap

All Columns



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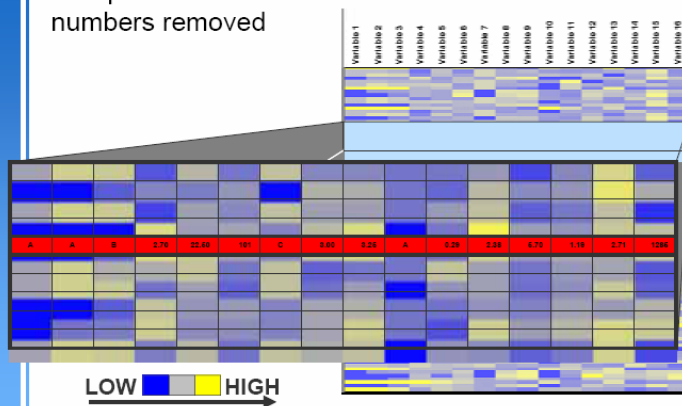
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## Example HeatMap

Completed view with numbers removed



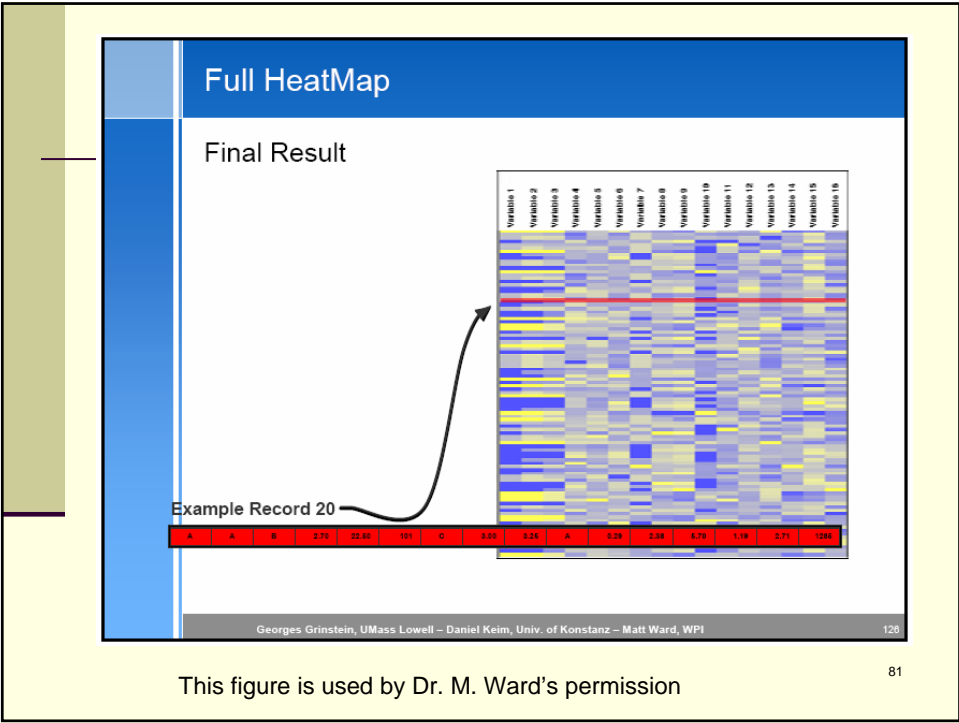
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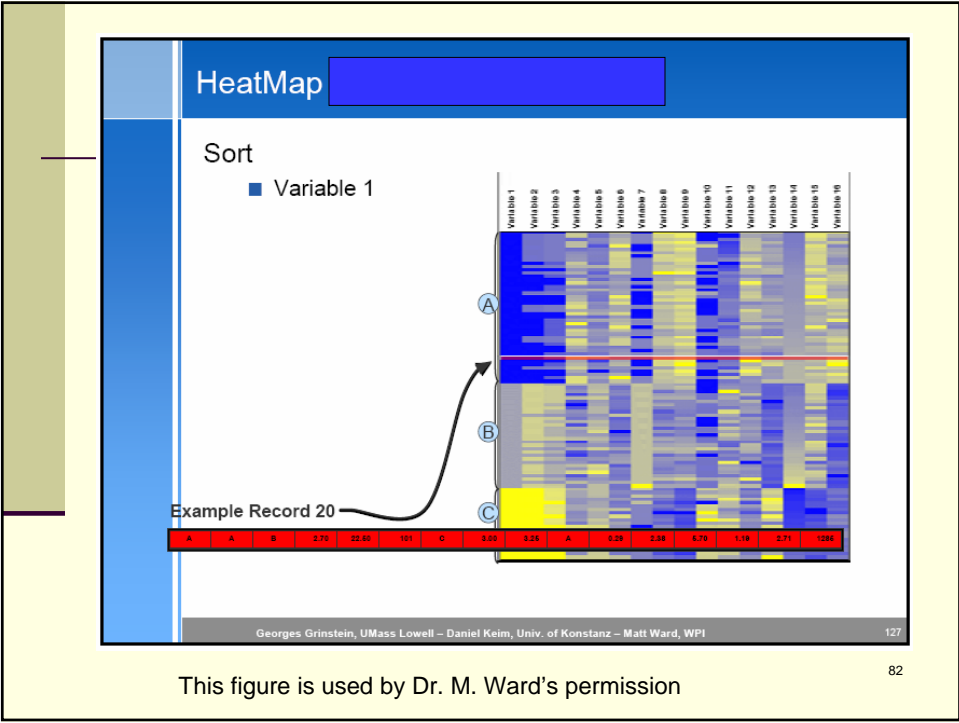
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[illegible]

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	Market	Company	Date	Volume	High	Low	Close
1	New York	Xerox	7/23/1992	79600	71.9	71.3	71.9
2	New York	Xerox	7/24/1992	109000	73	71.3	72.5
3	New York	Xerox	7/27/1992	164000	73.8	72.5	73.5
4	New York	Xerox	7/28/1992	319000	75.1	73.3	75
5	New York	Xerox	7/29/1992	402000	74.6	73	73.6

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## Major References

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- Colin Ware. Information Visualization, 2004
- Daniel Keim. Tutorial note in InfoVis 2000
- John Stasko. Course slides, Fall 2005