

Multivariate Visual Representations 2



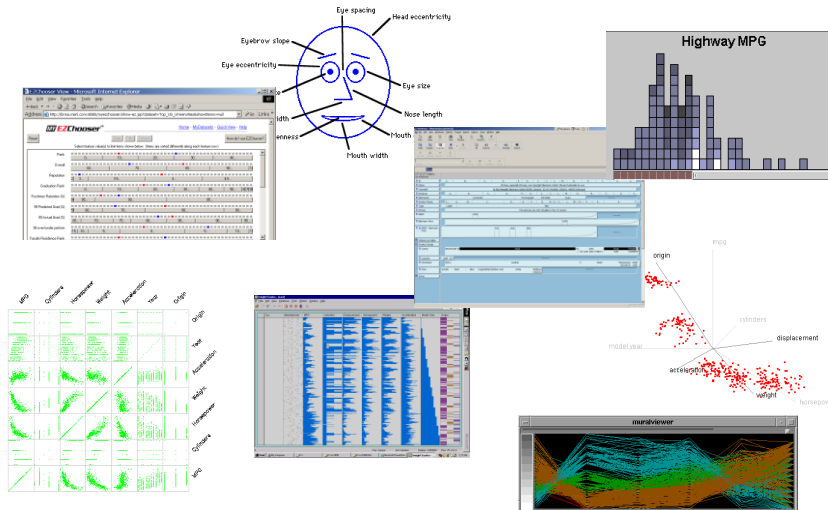
CS 7450 - Information Visualization
Sep. 15, 2011
John Stasko

Recap



- We examined a number of techniques for projecting >2 variables (modest number of dimensions) down onto the 2D plane
 - Scatterplot matrix
 - Table lens
 - Parallel coordinates
 - etc.

Varieties of Techniques



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Dust & Magnet



- Altogether different metaphor
- Data cases represented as small bits of iron dust
- Different attributes given physical manifestation as magnets
- Interact with objects to explore data

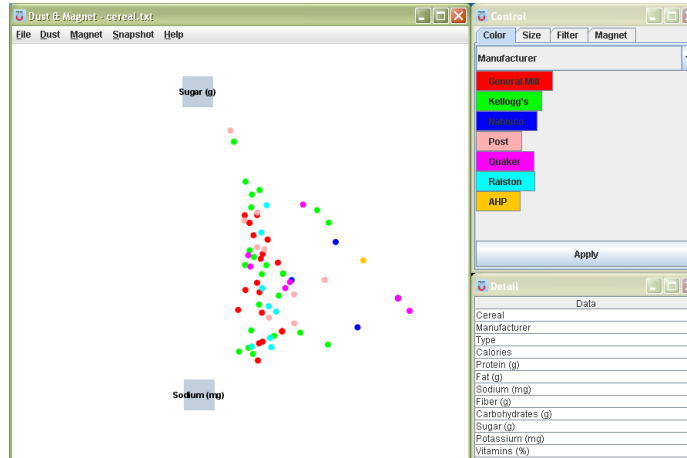
Yi, Melton, Stasko & Jacko
Information Visualization '05

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Interface



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Interaction

- Iron bits (data) are drawn toward magnets (attributes) proportional to that data element's value in that attribute
 - Higher values attracted more strongly
- All magnets present on display affect position of all dust
- Individual power of magnets can be changed
- Dust's color and size can be connected to attributes as well

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Interaction



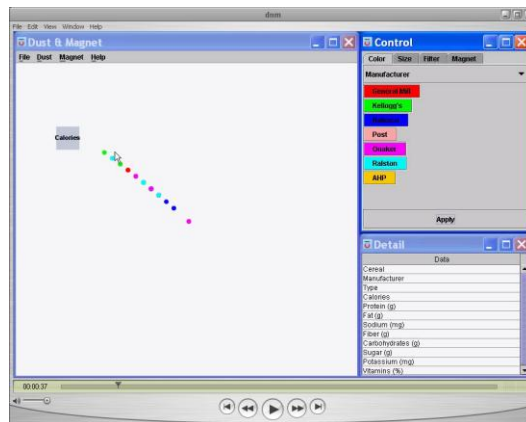
- Moving a magnet makes all the dust move
 - Also command for shaking dust
- Different strategies for how to position magnets in order to explore the data

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See It Live



<ftp://ftp.cc.gatech.edu/pub/people/stasko/movies/dnm.mov>

Video &
Demo

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Set Operations



- Different type of problem
 - Large set of items, each can be in one or more sets
 - How do we visually represent the set membership?

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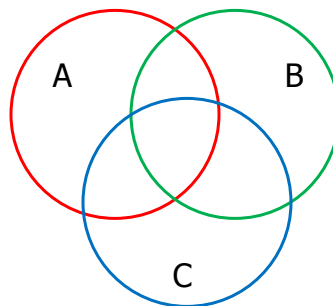
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Standard Technique



Venn
Diagram



Contains all possible zones of overlap

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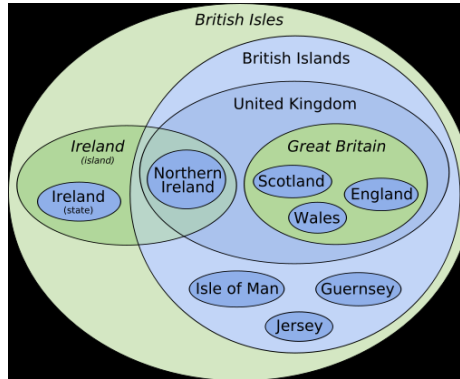
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Alternately



Euler
Diagram

Does not
necessarily
show all
possible
overlap zones



http://en.wikipedia.org/wiki/File:British_Isles_Euler_diagram_15.svg

But what's the problem?

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Bubble Sets



Video

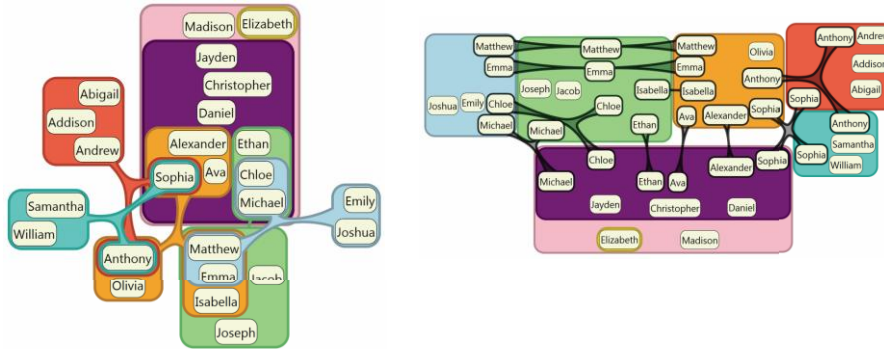
Collins et al
TVCG (InfoVis) '09

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ComED & DupED



Video

Riche & Dwyer
TVCG (InfoVis) '10

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Step Back



- Most of the techniques we've examined work for a modest number of data cases or variables
 - What happens when you have lots and lots of data cases and/or variables?

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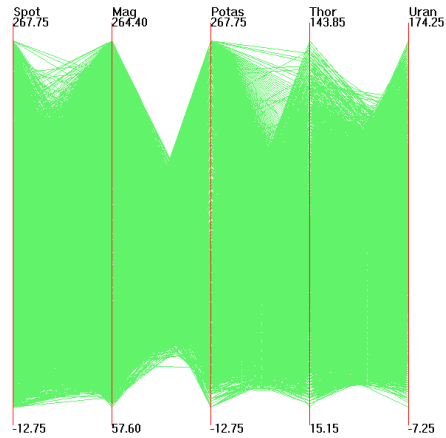
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Many Cases



Recall



Out5d dataset (5 dimensions, 16384 data items)

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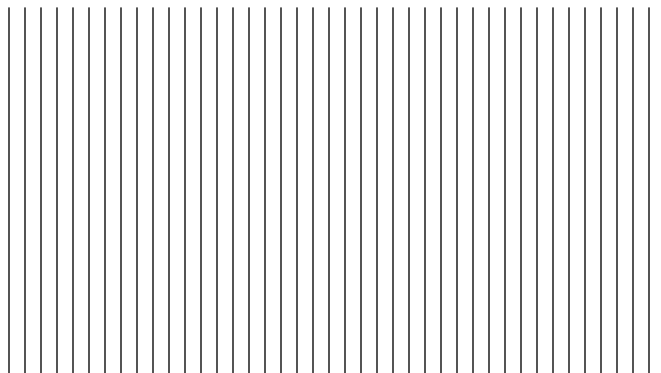
(courtesy of J. Yang)

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Many Variables



Recall



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Strategies



- How are we going to deal with such big datasets with so many variables per case?
- Ideas?

General Notion



- Data that is similar in most dimensions ought to be drawn together
 - Cluster at high dimensions
- Need to project the data down into the plane and give it some ultra-simplified representation
- Or perhaps only look at certain aspects of the data at any one time

Mathematical Assistance 1



- There exist many techniques for clustering high-dimensional data with respect to all those dimensions
 - Affinity propagation
 - k-means
 - Expectation maximization
 - Hierarchical clustering

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Mathematical Assistance 2



- There exist many techniques for projecting n-dimensions down to 2-D (dimensionality reduction)
 - Multi-dimensional scaling (MDS)
 - Principal component analysis
 - Linear discriminant analysis
 - Factor analysis

Comput Sci & Eng courses
Visual Analytics, Prof. Lebanon

Data mining
Knowledge discovery

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Other Techniques



- Other techniques exist to manage scale
 - Sampling – We only include every so many data cases or variables
 - Aggregation – We combine many data cases or variables
 - Interaction (later)
 - Employ user interaction rather than special renderings to help manage scale

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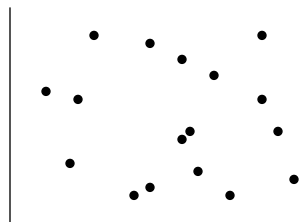
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Our Focus



- Visual techniques
- Many are simply graphic transformations from N-D down to 2-D



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Use?



- What kinds of questions/tasks would you want such techniques to address?
 - Clusters of similar data cases
 - Useless dimensions
 - Dimensions similar to each other
 - Outlier data cases
 - ...
- Think back to our “cognitive tasks” discussion

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Now



- We'll examine a number of other visual techniques intended for larger, higher-dimensional data sets

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Can We Make a Taxonomy?



- D. Keim proposes a taxonomy of techniques
 - Standard 2D/3D display
Bar charts, scatterplots
 - Geometrically transformed display
Parallel coordinates
 - Iconic display
Needle icons, Chernoff faces
 - Dense pixel display
What we're about to see...
 - Stacked display
Treemaps, dimensional stacking

TVCG '02

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Minimum Possible?



- We have data cases with variables
- What's the smallest representation we can use?
 - How?

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Dense Pixel Display



- Represent data case or a variable as a pixel
- Million or more per display
- Seems to rely on use of color
- Can pack lots in
- Challenge: What's the layout?

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One Representation



- Grouping arrangement
- One pixel per variable
- Each data case has its own small rectangular icon
- Plot out variables for data point in that icon using a grid or spiral layout

Uses color scale

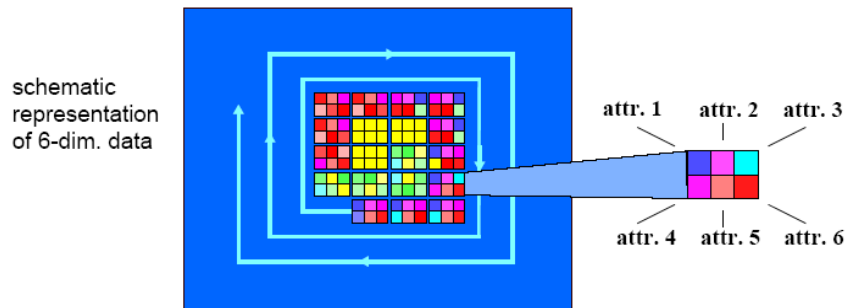


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Illustration



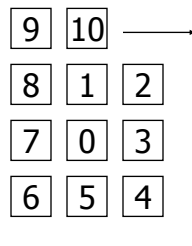
Levkowitz
Vis '91

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Spiral Technique



Dimensions

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Related Idea



- Pixel Bar Chart
- Overload typical bar chart with more information about individual elements

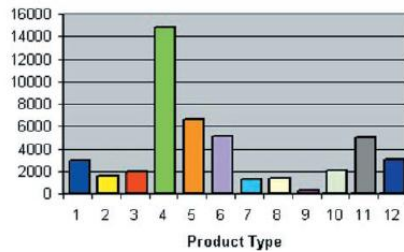
Keim et al
Information Visualization '02

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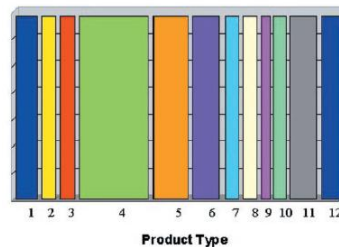
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Idea 1



Height encodes quantity



Width encodes quantity

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Idea 2



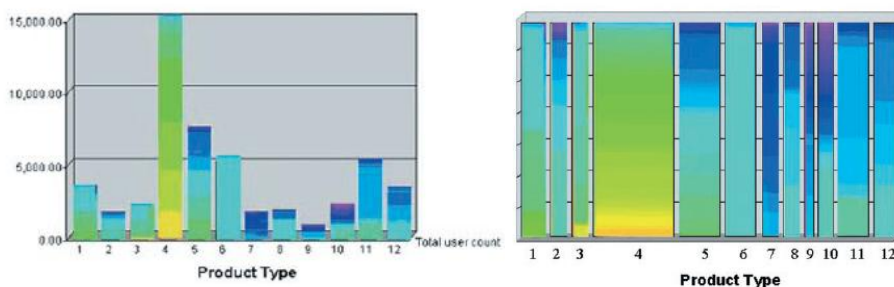
- Make each pixel within a bar correspond to a data point in that group represented by the bar
 - Can do millions that way
- Color the pixel to represent the value of one of the data point's variables

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Idea 3



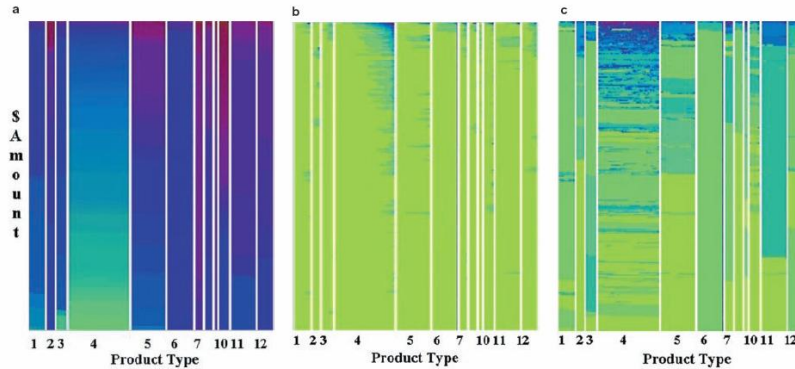
Each pixel is a customer
Color encodes amount spent by that person
High-bright, Low-dark
Ordered by that color attribute too
Right one shows more customers

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Idea 4



Product type is x-axis divider
 Customers ordered by
 y-axis: dollar amount
 x-axis: number of visits
 Color is (a) dollar amount spent, (b) number of visits, (c) sales quantity

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Idea 5

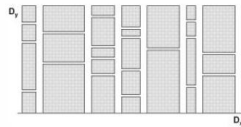


Figure 7 Dividing attributes on x- and y-axis (e.g., D_x =Product Type, D_y =Region).

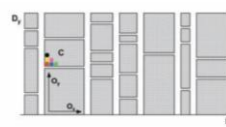


Figure 8 Ordering attributes on x- and y-axis (e.g., O_x =Dollar Amount, O_y =Quantity).

Can divide on two different attributes on x and y

Order items on both x and y

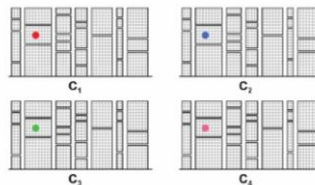


Figure 9 Multiple coloring attributes (e.g., C_1 =dollar amount, C_2 =no. of visits, C_3 =quantity, C_4 =region).

Color maps to some attribute
 (Same item always at same x,y position)

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Idea 6



Mapping specified by 5 tuple $\langle D_x, D_y, O_x, O_y, C \rangle$

D_x – Attribute partitions x axis
 D_y – Attribute partitions y axis
 O_x – Attribute specifies x ordering
 O_y – Attribute specifies y ordering
 C – Attribute specifies color mapping

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Example Application

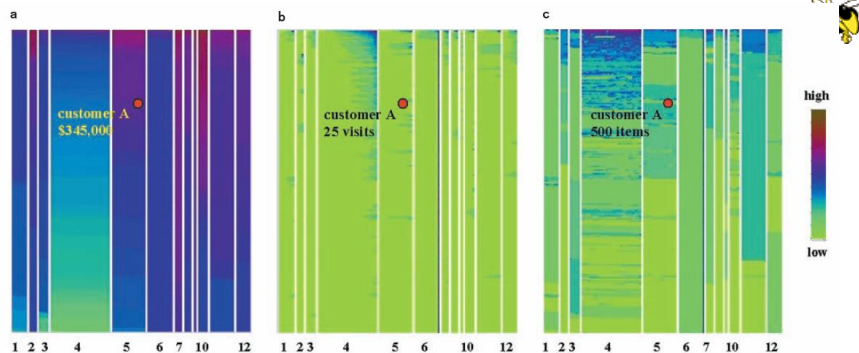


Figure 13 Multi-pixel bar chart for mining 405,000 sales transaction records. ($D_x = \text{Product Type}$, $D_y = \perp$, $O_x = \text{no. of visits}$, $O_y = \text{dollar amount}$, C). (a) Color: dollar amount. (b) Color: no. of visits. (c) Color: quantity.

1. Product type 7 and product type 10 have the top dollar amount customers (dark colors of bar 7 and 10 in Figure 13a)
2. The dollar amount spent and the number of visits are clearly correlated, especially for product type 4 (linear increase of dark colors at the top of bar 4 in Figure 13b)
3. Product types 4 and 11 have the highest quantities sold (dark colors of bar 4 and 11 in Figure 13c)
4. Clicking on pixel A shows details for that customer

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Thoughts?



- Do you think that would be a helpful exploratory tool?

High Dimensions



- Those techniques could show lots of data, but not so many dimensions at once
 - Have to pick and choose

Another Idea



- Use the dense pixel display for showing data and dimensions, but then project into 2D plane to encode more information
- VaR – Value and relation display

Yang et al
InfoVis '04

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Algorithm

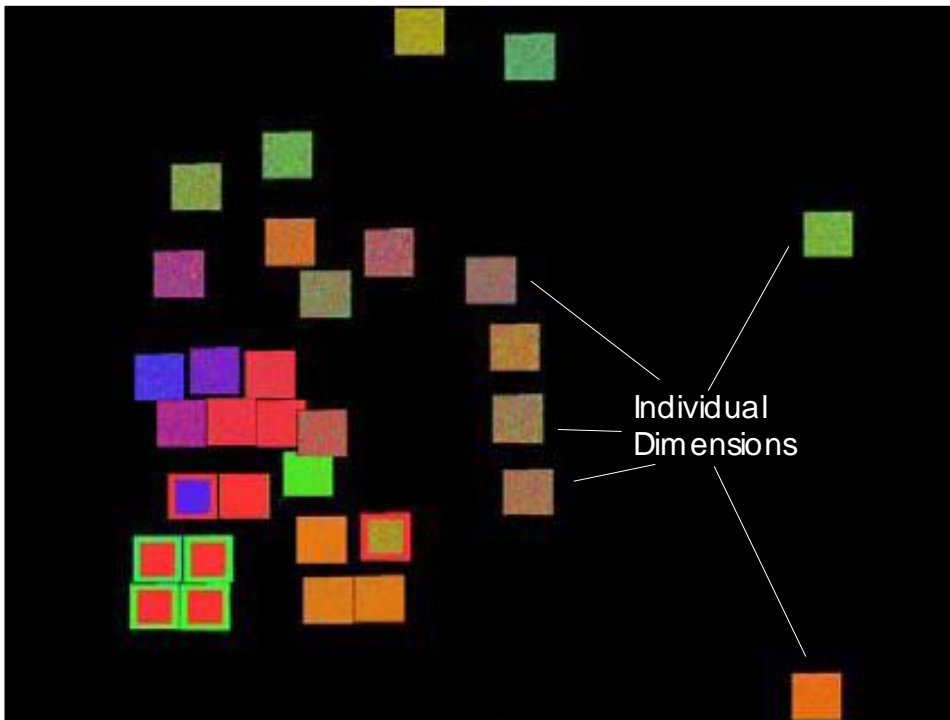


- Find a correlation function for comparing dimensions
- Calculate distances between dimensions (similarities)
- Make each dimension into a dense pixel glyph
- Assign position for each glyph in 2D plane using multi-dimensional scaling

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Questions

- What order are the data cases in each dimension-glyph?
 - Maybe there is a predefined order
 - Choose one dimension as “important” then order data cases by their values in that dimension
 - “Important” one may be the one in which many cases are similar

Alternative



- Instead of each glyph being a dimension, it can be a data case

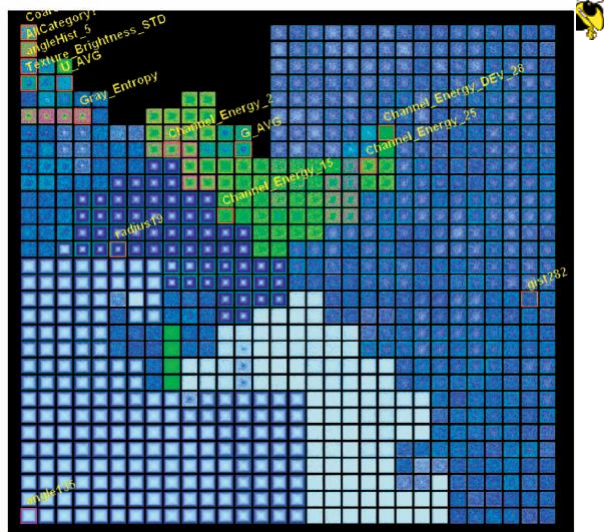
Follow-on Work



- Use alternate positioning strategies other than MDS
- Use Jigsaw map idea (Wattenberg, InfoVis '05) to lay out the dimensions into a grid
 - Removes overlap
 - Limits number that can be plotted

New Layout

Plot the glyphs
into the grid
positions



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HCE

- Hierarchical Clustering Explorer
- Implements "rank by feature" framework
- Help guide user to choose 1D distributions and 2D scatterplots from various dimensions of a data set
- Combine statistical analysis with user-directed exploration

Seo & Shneiderman
Information Visualization '05

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Idea



- Choose a feature detection criterion to rank 1D and 2D projections of a data set
- Use person's perceptual abilities to pick out interesting items from view

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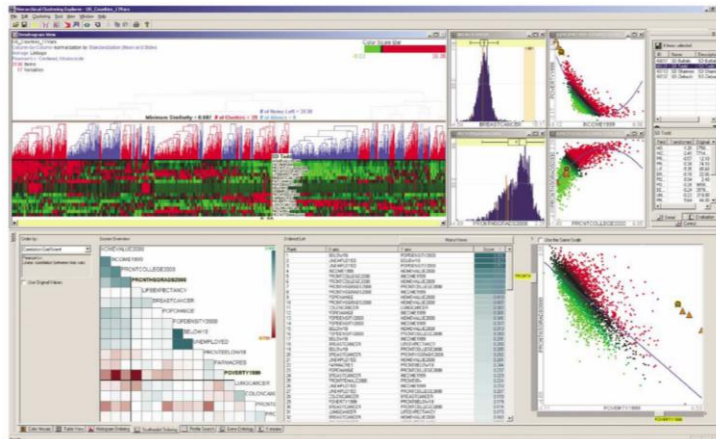
HCE UI



Some chosen distributions and scatterplots

Cases in
columns,
variables
in rows

Group
similar
cases



Seven tabs at bottom to choose from

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Operation



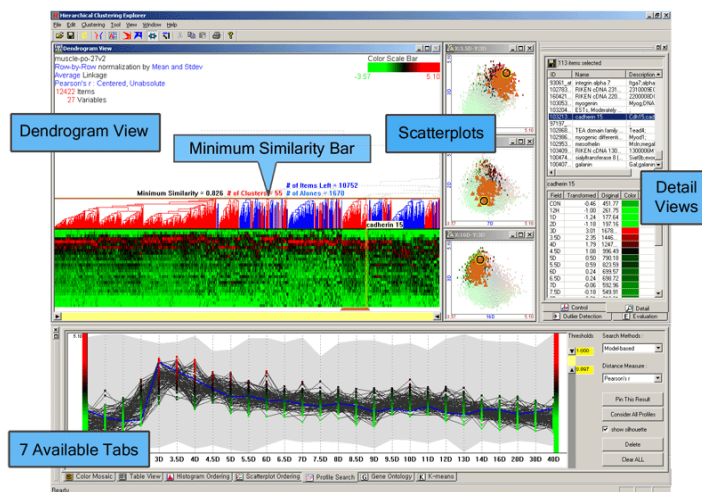
- When you choose the histogram ordering or scatterplot ordering tabs at the bottom left, these give results based on various statistical measures
- You can then choose some of them to visualize

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Demo



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Recap



- We've seen many general techniques for multivariate data these past two days
 - Know strengths and limitations of each
 - Know which ones are good for which circumstances
- We still haven't explored interaction much

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HW 3



- Visualization design
- Due Tuesday
 - Bring two copies
- Questions?

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Upcoming



- Tufte's Design Principles
 - Reading:
Envisioning Information (if you have it)
- Few's Design Guidance
 - Reading
Now You See It chapters 5-12