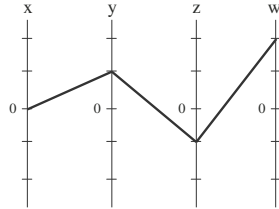




## Parallel Coordinates

- Point → line
- (0,1,-1,2)=



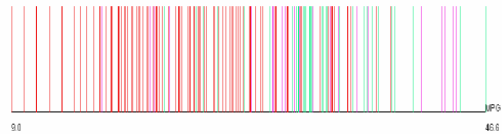
## Base Visualizations

- Scatter Plot
- Parallel Coordinates
- Survey Plot
- Radviz spring visualization

## Parameters of DA

- Nine parameters are selected to describe the graphics properties of each DA:
  - p1: size of the scatter plot points
  - p2: length of the perpendicular lines extending from individual anchorpoints in a scatter plot
  - p3: length of the lines connecting scatter plot points that are associated with the same data point
  - p4: width of the rectangle in a survey plot
  - p5: length of the parallel coordinate lines
  - p6: blocking factor for the parallel coordinate lines
  - p7: size of the radviz plot point
  - p8: length of the "spring" lines extending from individual anchorpoints of a radviz plot
  - p9: the zoom factor for the "spring" constant K

## Basic Single DA

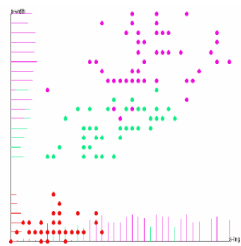


- Dimension – miles per gallon
- Data values are mapped to the axis
- Mapped data points - *anchorpoints*, represent the coord values (points along a DA)
- Lines extended from anchorpoints
- Color – type of car (American – red, Japanese – green, and European – purple)

## Two-DA scatter plot

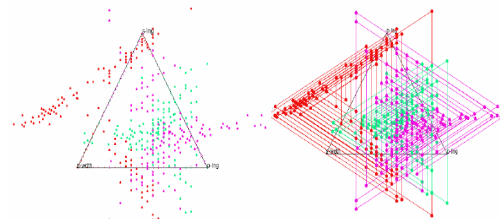
- DA scatter plot using two DAs
- Perpendicular lines extending outward from the anchor points
- If they meet, plot the point at the intersection

- p1: size of the scatter plot points
- p2: length of the perpendicular lines extending from individual anchor points in a scatter plot
- p3: length of the lines connecting scatter plot points that are associated with the same data point



$P = (0.8, .2, 0, 0, 0, 0, 0, 0, 0)$

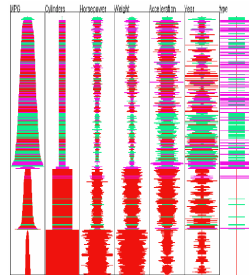
## Three DAs



$P = (0.6, 0, 0, 0, 0, 0, 0, 0, 0)$      $P = (.6, 0, 1.0, 0, 0, 0, 0, 0, 0)$   
 P3: length of lines connecting all displayed points associated with one real data point (record)

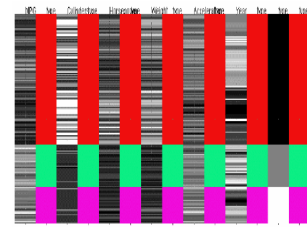
## Seven DA Survey Plot

- 7 vertical DAs in a row
- Rectangle extending from an anchor point
  - size is based on the dimensional value
  - eg. Type- discrete value
  - red < green < purple



## CCCViz - Color Correlated Column

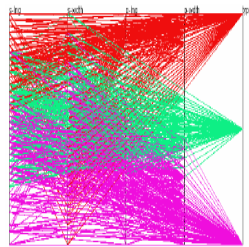
- Does a dimension (gray scales) correlate with a particular classification dimension (color scale)?
- Correlation is seen in mpg, cylinders etc.
- p4: width of the rectangle in a survey plot



CCCViz DAs with  $P = (0, 0, 0, 1.0, 0, 0, 0, 0, 0)$

## DAs in PC configuration

- Line from one DA anchorpoint is drawn to another
  - length of these connecting lines is controlled by p5.
  - p5 = 1.0, fully connected, every anchorpoint connects to all the other (N-1) anchorpoints
- P6 controls how many DAs a p5 connecting line can cross
  - p6 = 0, traditional PC



$P = (0, 0, 0, 0, 1.0, 1.0, 0, 0, 0)$

## DAs in Regular Polygon

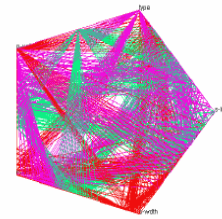
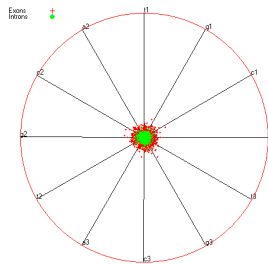


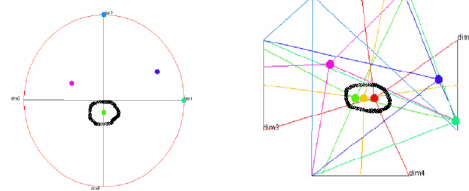
Figure 15 DAs in a regular polygon  $P = (0, 0, 0, 0, 1.0, 0, 0, 0, 0)$  Mesh Plot

## Intro. to RadViz Spring Force

- a radial visualization
- One spring for each dimension.
- One end attached to perimeter point. The other end attached to a data point.
- Each data point is displayed where the sum of the spring forces equals 0.



## DAs RadViz



Original Radviz - 3 overlapping points  
Limitation: data points with different values can overlap

DAs spread polygon  
 $P = (0, 0, 0, 0, 0, 0, .5, 1.0, .5)$

## DA layout

✓ Parameters - Done !

### Layout

- DAs can be arranged with any arbitrary size, shape or position
- Permits a large variety of visualization designs

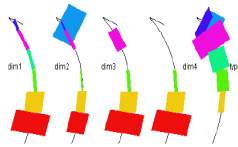
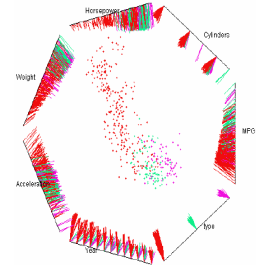


Figure 20 Survey Plot parameters with curved DAs

## Combinations of Visualizations

Can we combine features of two (or more) visualizations?

- Combination of Parallel Coordinates and Radviz



## Visualization Space

- Nine parameters define the size of our visualization space as  $\mathbb{R}^9$
- Include the geometry of the DAs, assuming 3 parameters are used to define the geometry
- The size of our visualization space is  $\mathbb{R}^{12}$
- "Grand Tour" through visualization space is possible
- New visualizations can be created during a tour

## Evaluation

### Strong Points

- ☺ Idea
- ☺ Many examples of visualizations with real data

### Weak Points

- ☹ Not accessible
- ☹ Short explanation of examples
- ☹ Lack of examples for some statement
- ☹ No implementation details

## Where are we

✓ Dimensional Anchors

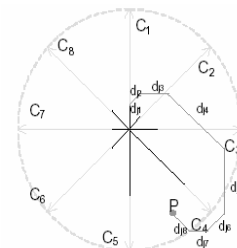
### Star Coordinates

- a new interactive multidimensional technique
- helpful in visualizing multi-dimensional clusters, trends, and outliers

- StarClass - Interactive Visual Classification Using Star Coordinates

## Star Coordinates

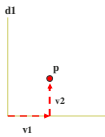
- Each dimension shown as an axis
- Data value in each dimension is represented as a vector.
- Data points are scaled to the length of the axis
  - min mapping to origin
  - max mapping to the end



## Star Coordinates Contd

### Cartesian

$$P=(v1, v2)$$

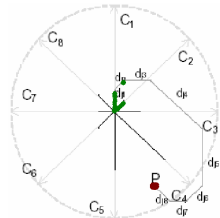


Mapping:

- Items  $\rightarrow$  dots
- $\Sigma$  attribute vectors  $\rightarrow$  position

### Star Coordinates

$$P=(v1, v2, v3, v4, v5, v6, v7, v8)$$



## Interaction Features

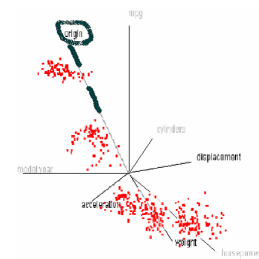
- **Scaling**
  - allows user to change the length of an axis
  - increases or decrease the contribution of a data column
- **Rotation**
  - changes the direction of the unit vector of an axis
  - makes a particular data column more or less correlated with the other columns
- **Marking**
  - selects individual points or all points within a rectangular area and paints them in color
  - makes points easy to follow in the subsequent transformations

## Interaction Features

- **Range Selection**
  - select value ranges on one or more axes, mark and paint them
  - allows users to understand the distribution of particular data value ranges in current layout
- **Histogram**
  - provides data distribution for each dimension
- **Footprints**
  - leave marks of data points on the trail for recent transformations

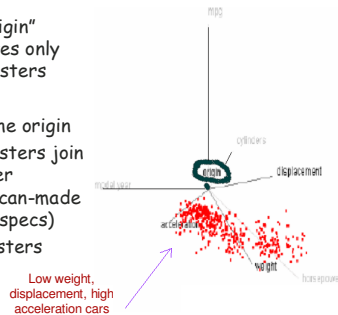
## Applications - Cluster Analysis

- **Playing with the "cars" dataset**
  - scaling, rotating, & turning off some coordinates
- **Four major clusters in the data discovered**



## Applications - Cluster Analysis

- **Scaling the "origin" coordinate moves only the top two clusters**
  - (JP & Euro)
- **Down-scaling the origin**
  - these two clusters join one of the other clusters (American-made cars of similar specs)
- **Result: two clusters**

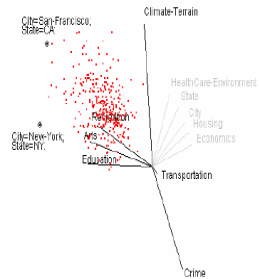


## SC - useful in visualizing clusters

- **Within few minutes users can identify how the data is clustered**
- **Gain an understanding of the basic characteristics of these clusters**

## Multi-factor Analysis

- Dataset - "Places"
  - ratings wrt climate, transportation, housing, education, arts, recreation, crime, health-care, and economics
- Important desirable factors pulled together in one direction and neg. undesirable factors in the opposite

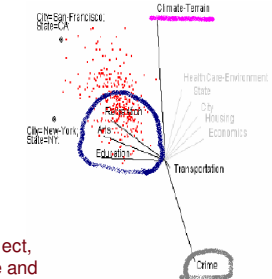


## Mutli-factor Analysis con't

- Desirable factors:
  - recreation, art, & education
  - climate (most)
- Undesirable factor:
  - crime

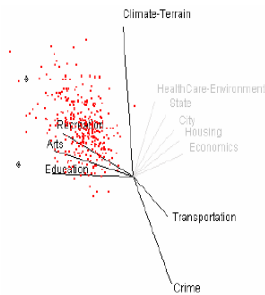
What can you conclude about NY and SF?

- NY – outlier
- SF – comparable arts, ect, but better climate and lower crime



## Multi-factor Analysis contd

- Scale up transportation
  - other cities beat SF in the combined measure



## Evaluation of SC in Multi-factor Analysis

- ⊗ Exact individual contributions of these factors are not immediately clear
- ⊗ The visualization provides users with an overview of how a number of factors affect the overall decision making

## Evaluation

### Strong Points

- ⊗ idea
- ⊗ many concrete examples with full explanations

### Weak points

- ⊗ ugly figures (undistinguishable)

## Where we are

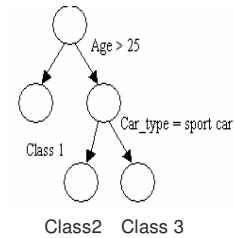
- ✓ Dimensional Anchors
- ✓ Star Coordinates
  - a new interactive multi-D visualization tech.
- StarClass - Interactive Visual Classification Using Star Coordinates

## Classification

- Each object in a dataset belongs to exactly one class among a set of classes.
- Training set data : labeled (class known)
- Build model based on training set
- Classification: use the model to assign a class to each object in the testing set.

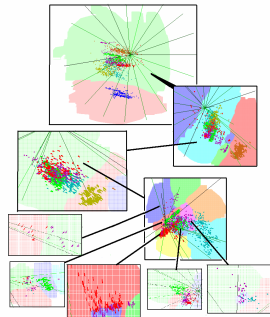
## Classification Method

- Decision trees



## Visual-base DT Construction

- Visual Classification
    - projecting
    - painting
    - region can be re-projected
    - recursively define a decision tree.
    - each project correspond to a node in decision tree
    - Majority class at leaf node determines class assignment
- (the class with the most number of objects mapping to a terminal region is the "expected class")



## Evaluation of the system

Good

Bad

- ⊙ Makes use of human judgment and guides the classification process
  - ⊙ Good accuracy
  - ⊙ Increase in user's understanding of the data
- ⊙ expertise required?

## Evaluation of the Paper

Good

- ⊙ Ideas
- ⊙ Accessible
- ⊙ Concrete examples

Bad

- ⊙ No implementation discussed

## Summary

- Dimensional Anchor
  - unify visualization techniques
- Star Coordinate
  - new interactive visualization techniques
  - Visualizing clusters and outliers
- StarClass
  - interactive classification using star coordinate

## Reference

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- [Dimensional Anchors: a Graphic Primitive for Multidimensional Multivariate Information Visualizations](#), P. Hoffman, G. Grinstein, & D. Prinkney, Proc. Workshop on New Paradigms in Information Visualization and Manipulation, Nov. 1999, pp. 9-16.
- [Visualizing Multi-dimensional Clusters, Trends, and Outliers using Star Coordinates](#), Eser Kandogan, Proc. KDD 2001
- [StarClass: Interactive Visual Classification Using Star Coordinates](#), S. Teoh & K. Ma, Proc. SIAM 2003
- <http://graphics.cs.ucdavis.edu/~steoh/research/classification/SDM03.ppt>