

Part III

Geometrically-transformed displays

7 Introduction

8 Scatter plot matrix

- Interaction

9 Parallel Coordinates

- Overlapping
- Variable order

A class of Visual mapping transformation:

- for table data
- mostly data **independent**
- maps one object to a set of points (and maybe lines) in 2D or in 3D
- examples:
 - scatter plot matrix
 - Andrews' curves
 - parallel coordinates

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Scatter plot matrix

- A matrix of scatter plots

X_1 alone	(X_2, X_1)	(X_3, X_1)
(X_1, X_2)	X_2 alone	(X_3, X_2)
(X_1, X_3)	(X_2, X_3)	X_3 alone

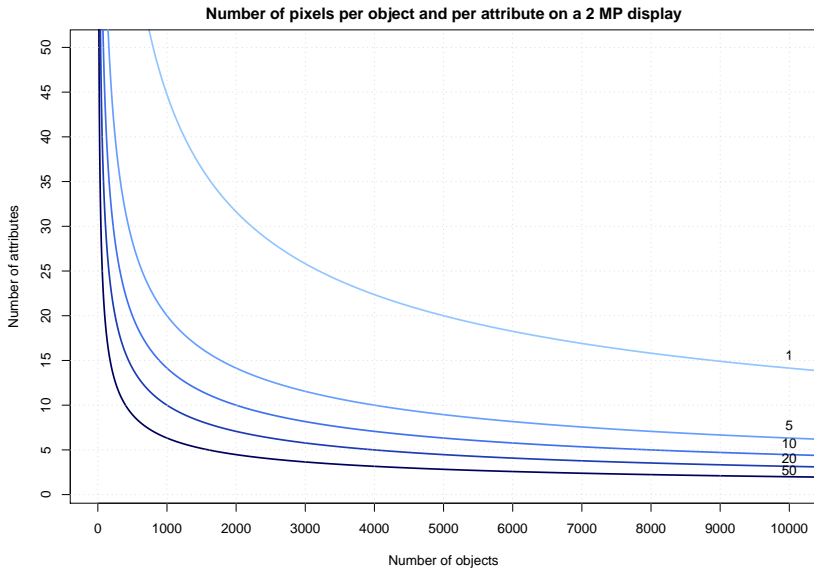
- Pros:

- almost no learning curve
- inherits scatter plot extensions

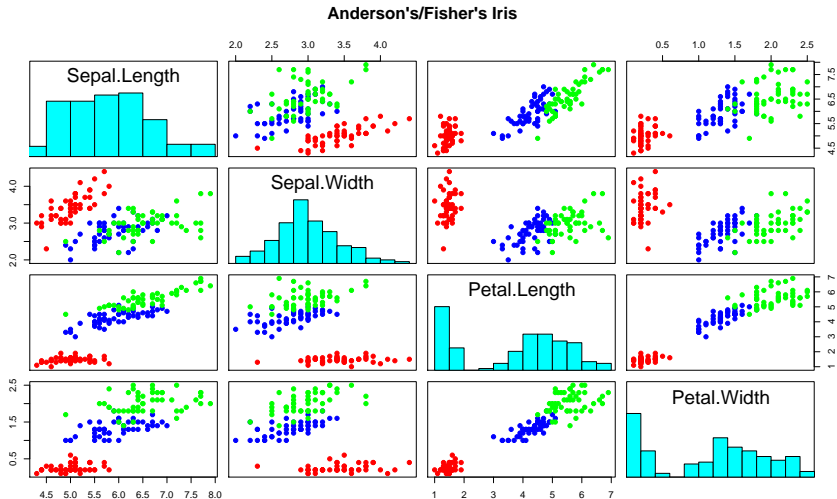
- Cons:

- worsen scatter plot overlapping problem: p variables $\Rightarrow p^2$ scatter plots $\Rightarrow p^2 n$ values (n objects)
- links between plots are difficult to understand

Scalability

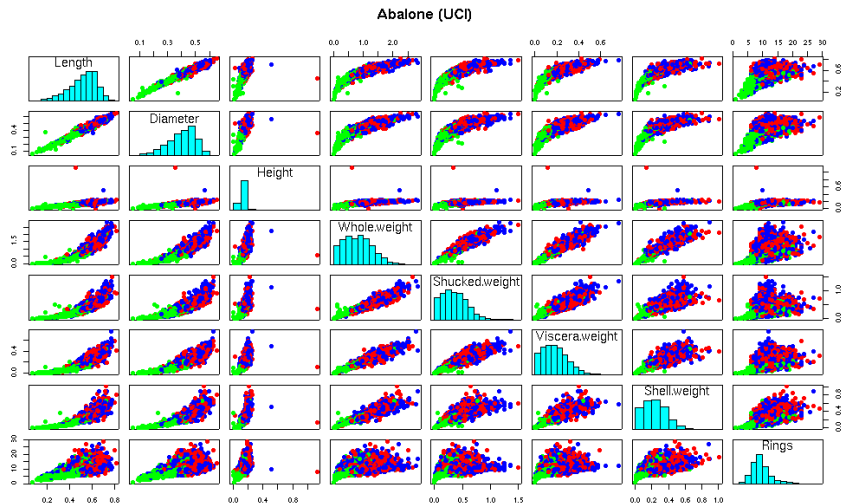


Scalability



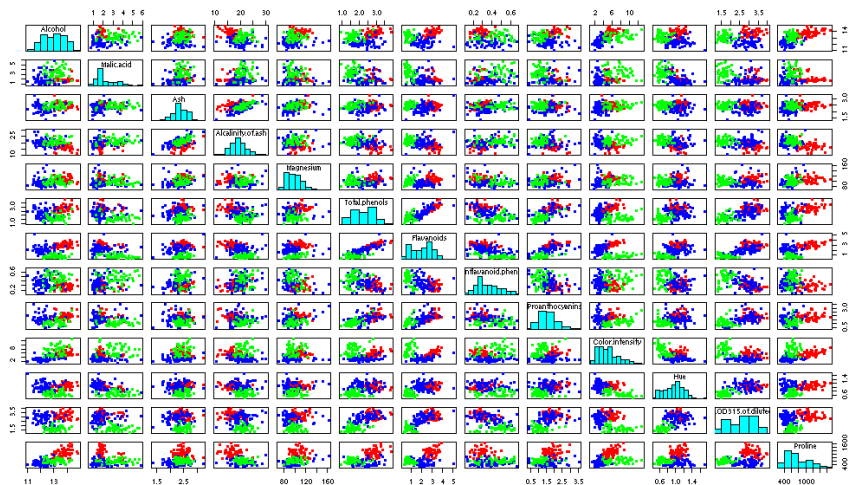
4+1 variables, 150 objects

Scalability



8+1 variables, 4177 objects

Italian Wines



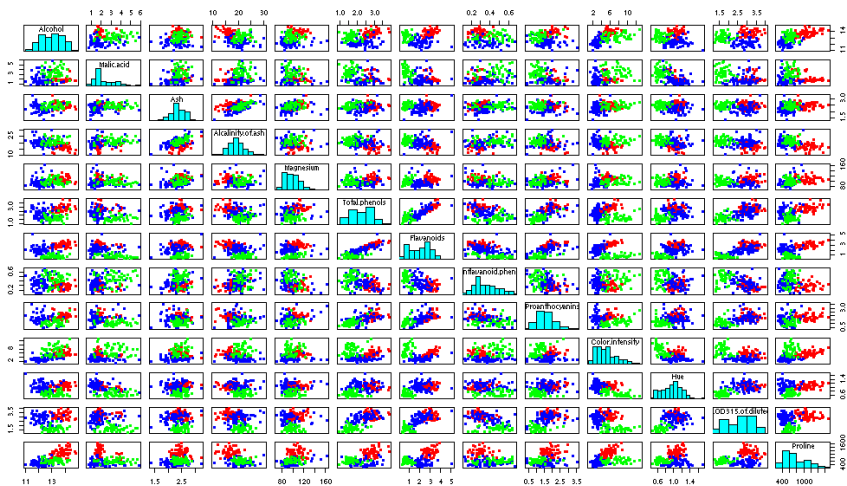
13+1 variables, 178 objects

Scalability via interaction

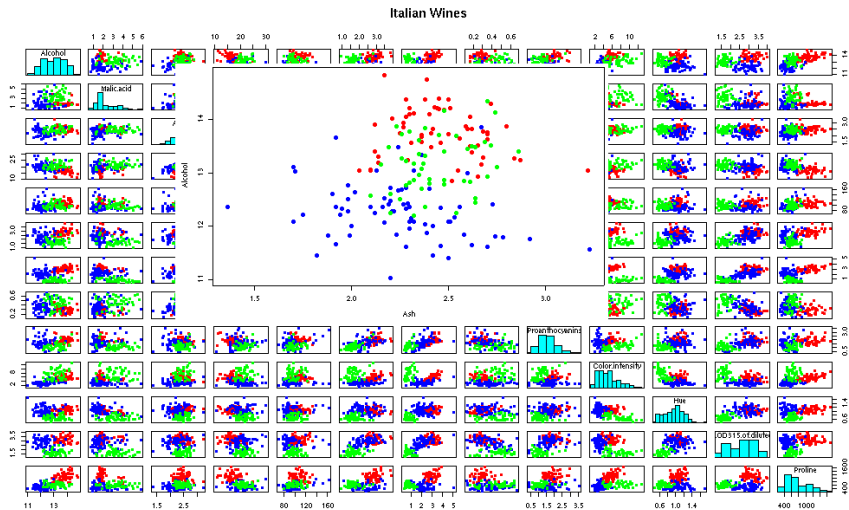
- User choice:
 - variable selection
 - variable ordering
- Instant zoom (popup)
- Linking and brushing:
 - brushing:
 - selection of a subset of the objects
 - display of the selection (hue based)
 - linking:
 - brush in one view
 - display the results in all views
 - views don't have to be scatter plots
 - In previous slides: class information!

Zooming popup

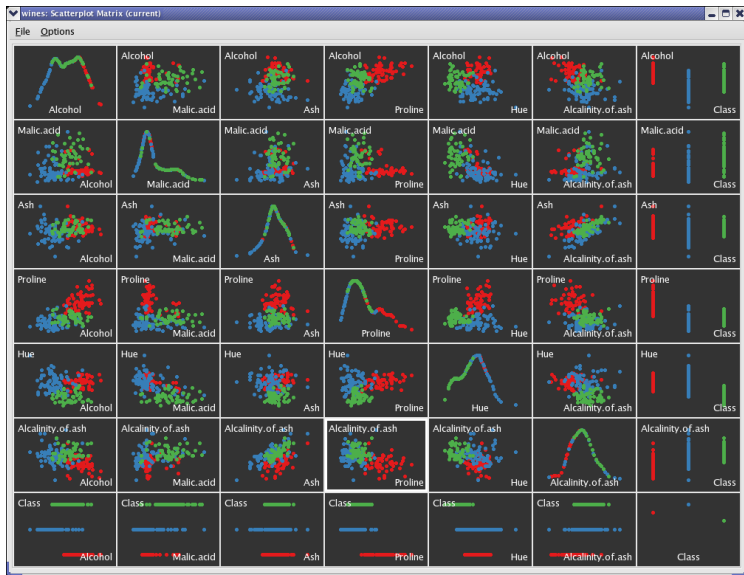
Italian Wines



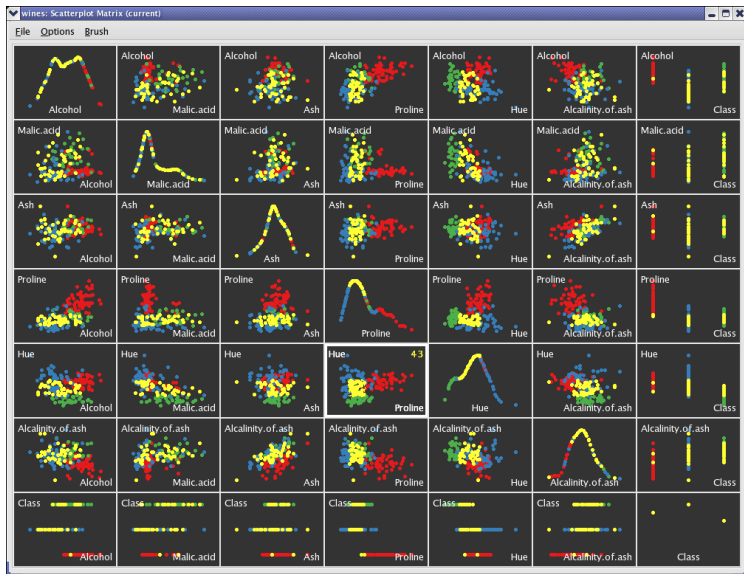
Zooming popup



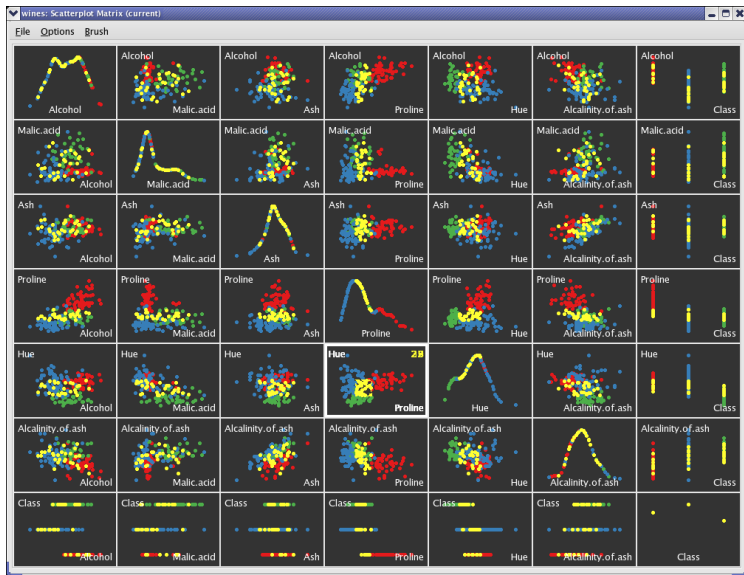
Linking and brushing (Ggobi)



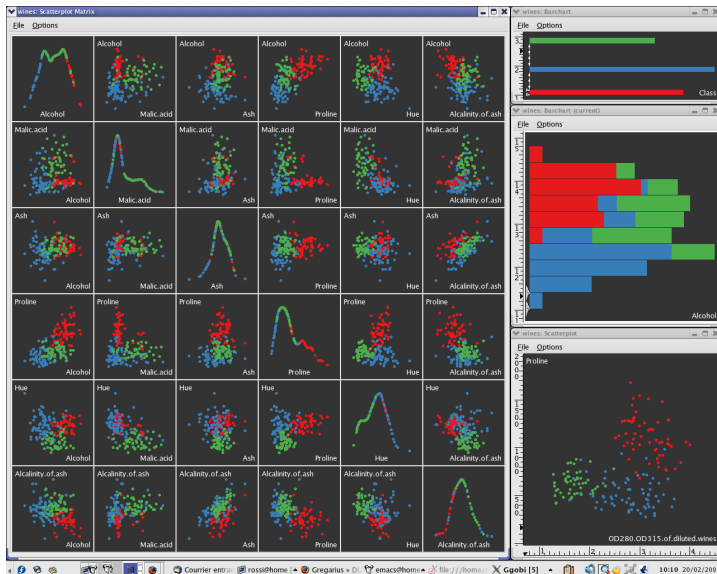
Linking and brushing (Ggobi)



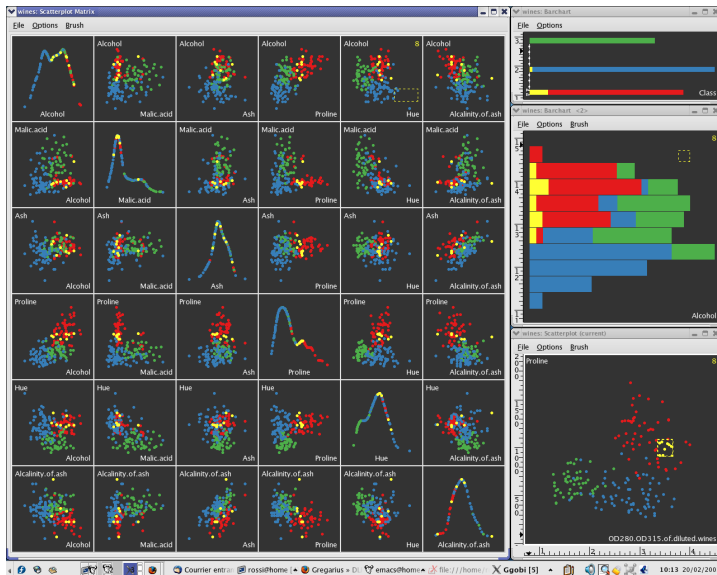
Linking and brushing (Ggobi)



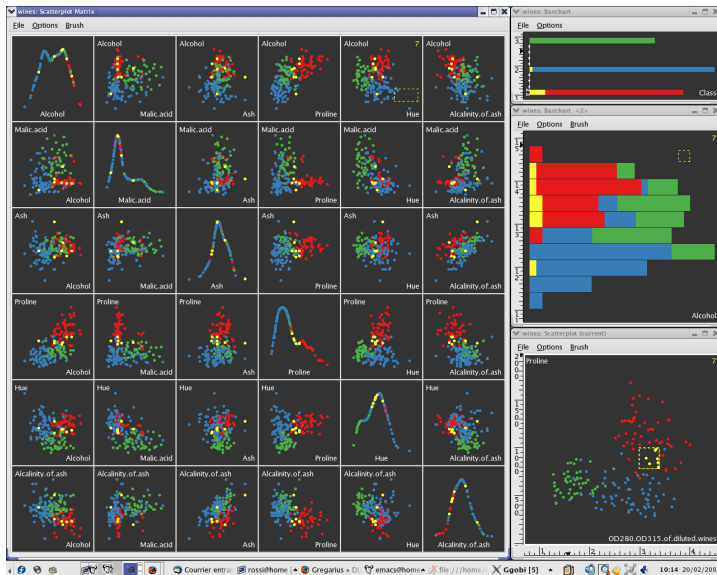
Linking and brushing (Ggobi)



Linking and brushing (Ggobi)



Linking and brushing (Ggobi)



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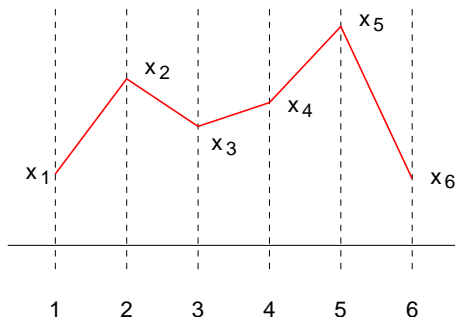
9 **Parallel Coordinates**

- **Overlapping**
- **Variable order**

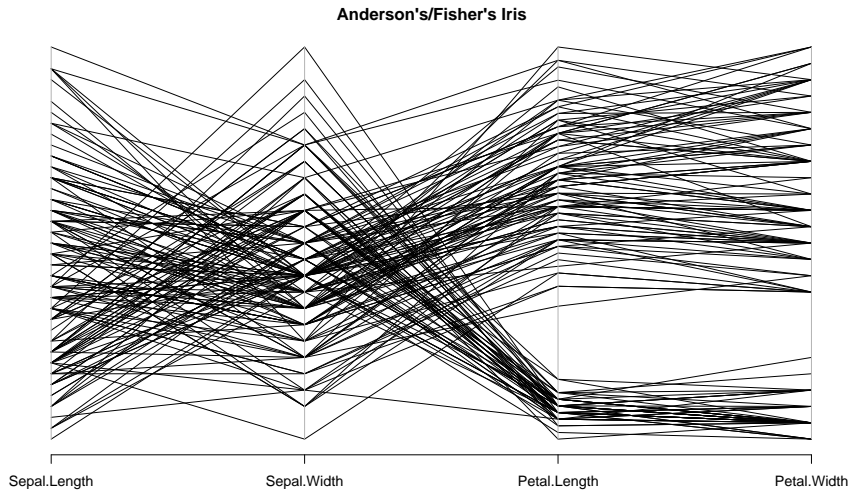
Parallel Coordinates

Proposed in 1985 by A. Inselberg

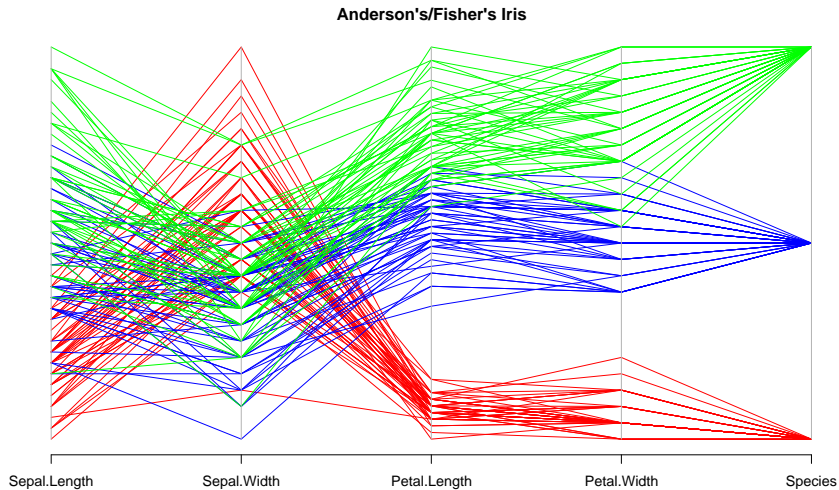
- A variable = a vertical axis
- An object = a piecewise linear curve
- (x_1, \dots, x_p) is mapped to the polyline that joins $(1, x_1), (2, x_2), \dots, (p, x_p)$



Iris dataset

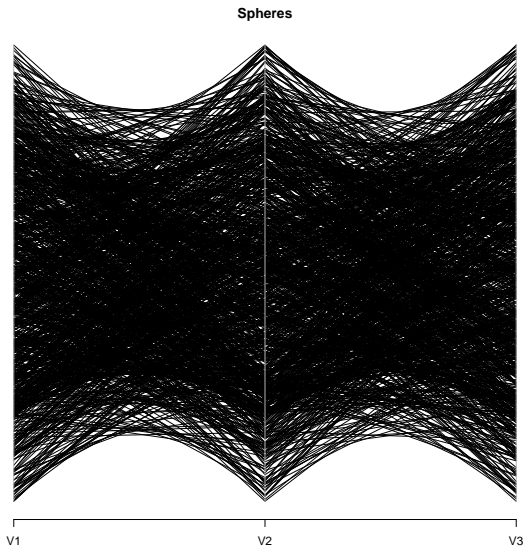


Iris dataset

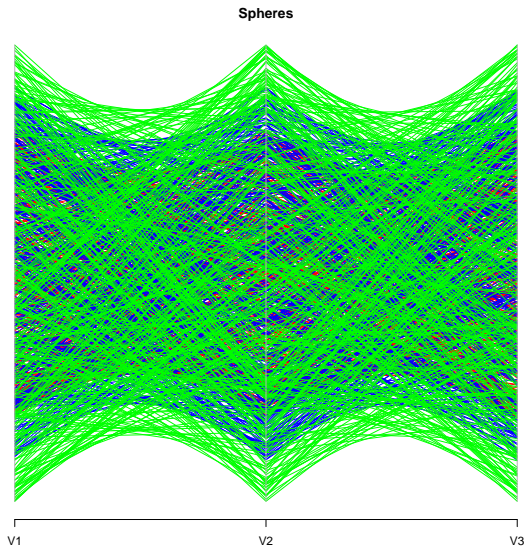


- Pros:
 - reasonable variable scalability
 - powerful after proper training
- Cons:
 - learning curve
 - major overlapping problem
 - variable order
- Reducing the problems:
 - rendering (transparency)
 - interactivity (brushing)
 - simplification (data clustering)
 - variable ordering (user guided and/or **automated**)

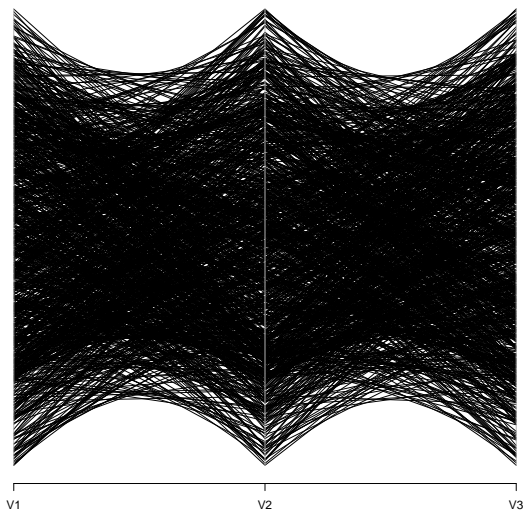
Overlapping



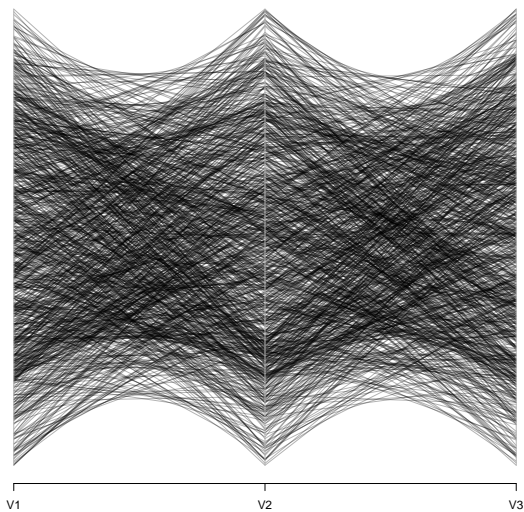
Overlapping

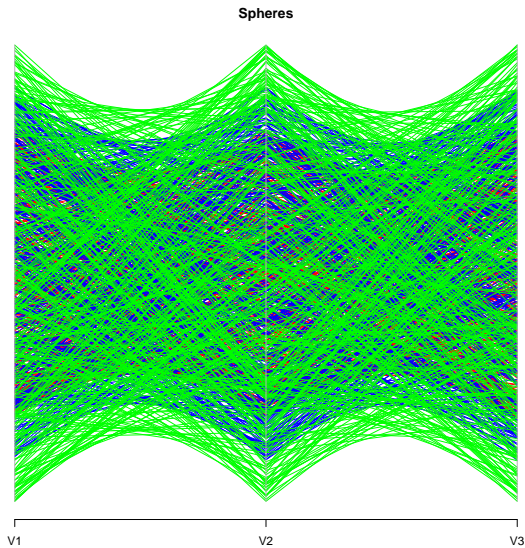


Spheres

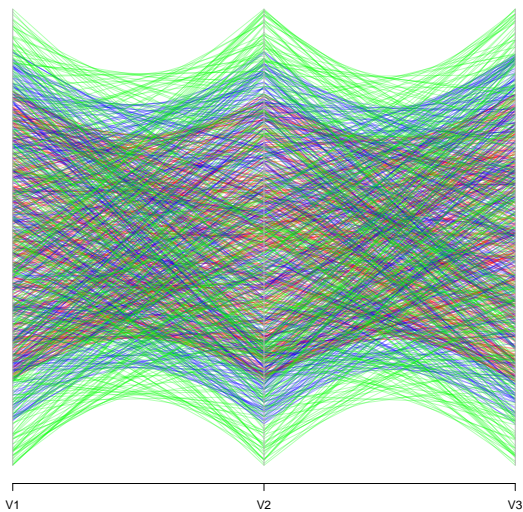


Spheres

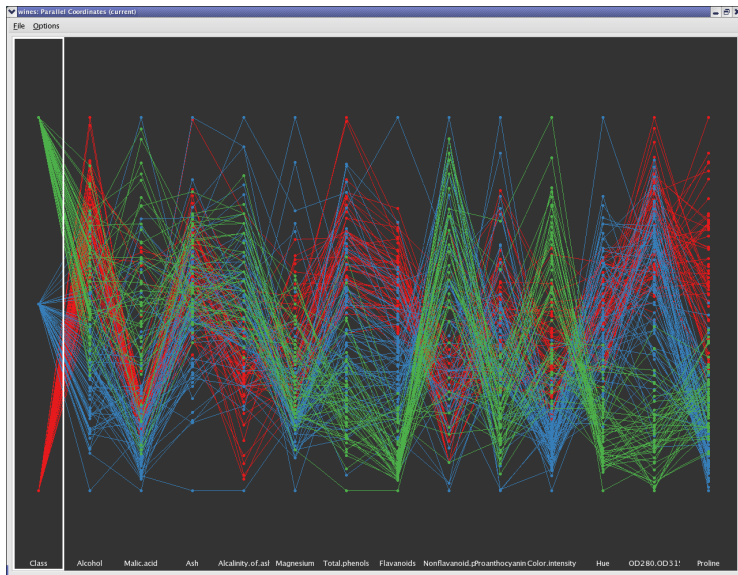




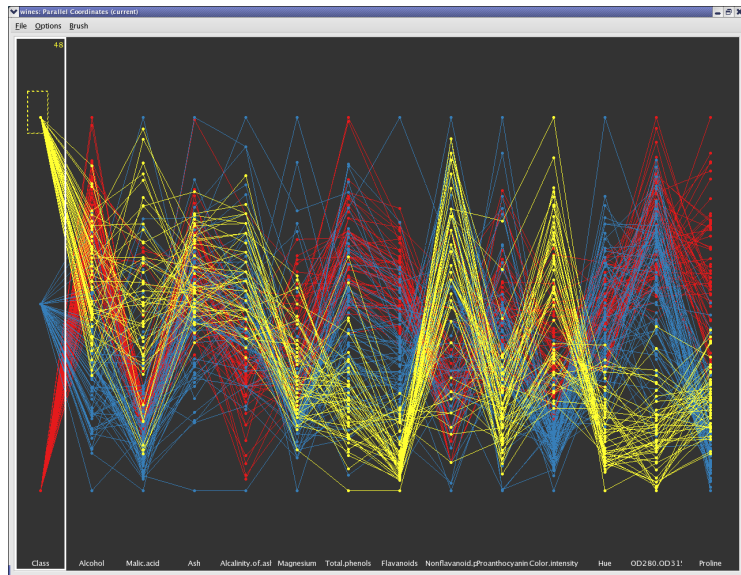
Spheres



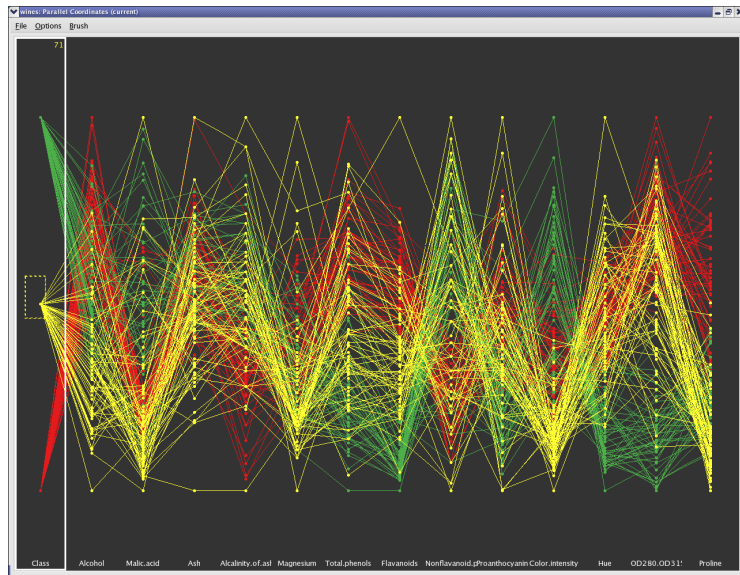
Brushing



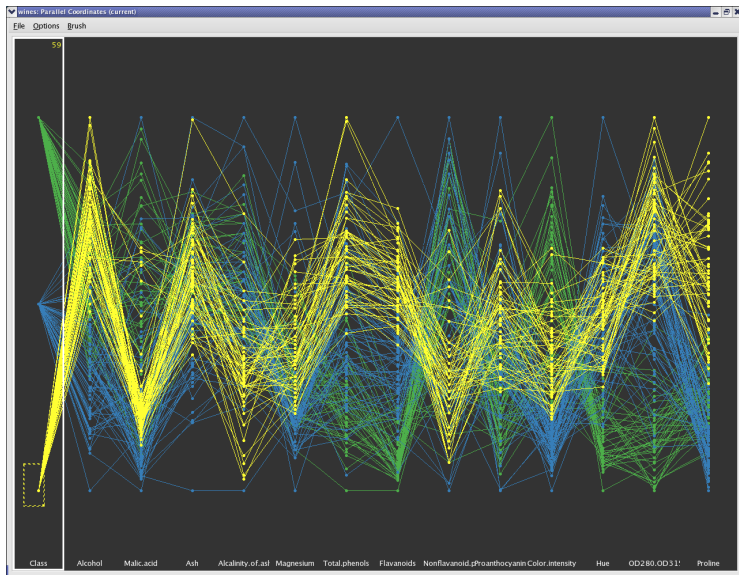
Brushing



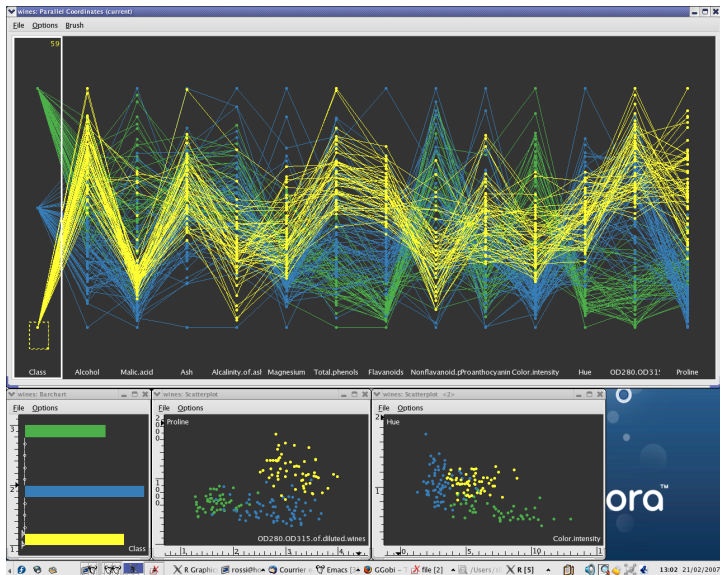
Brushing



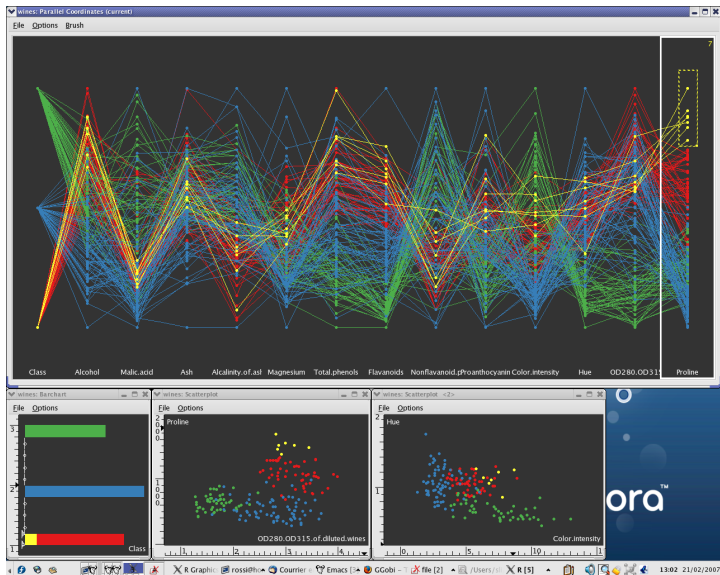
Brushing



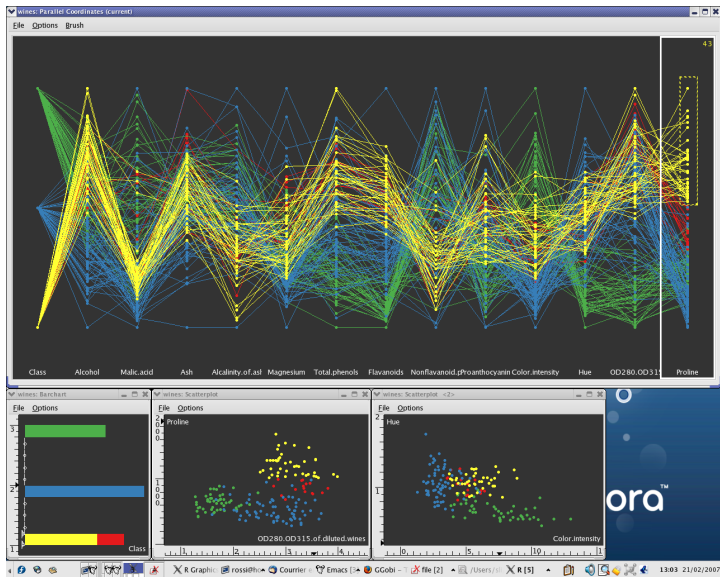
Brushing and linking



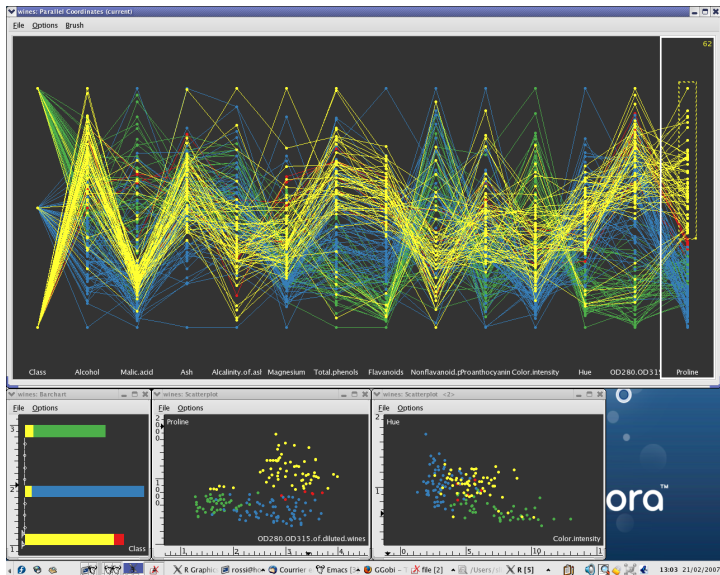
Brushing and linking



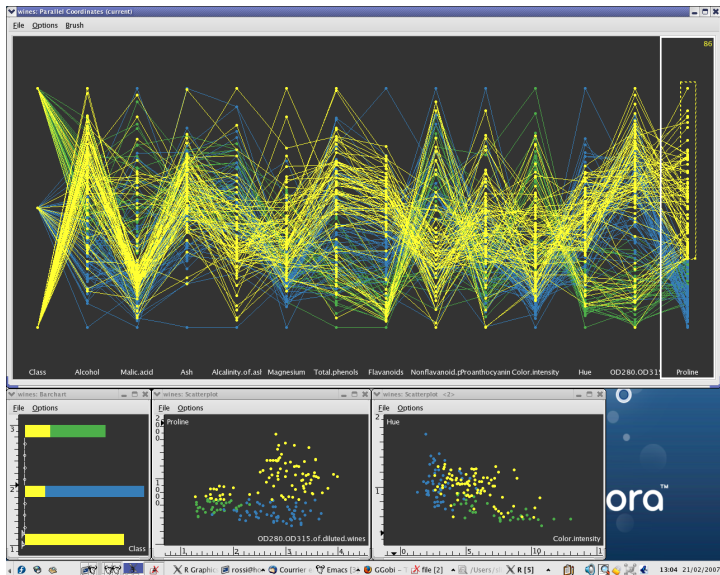
Brushing and linking



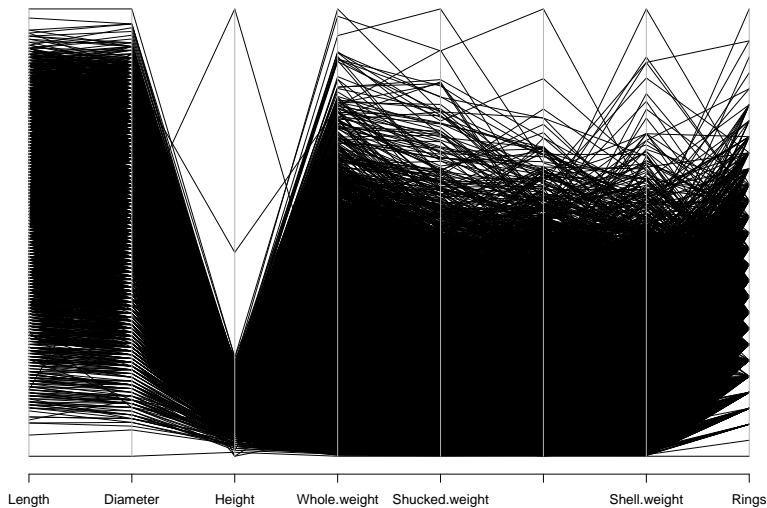
Brushing and linking



Brushing and linking

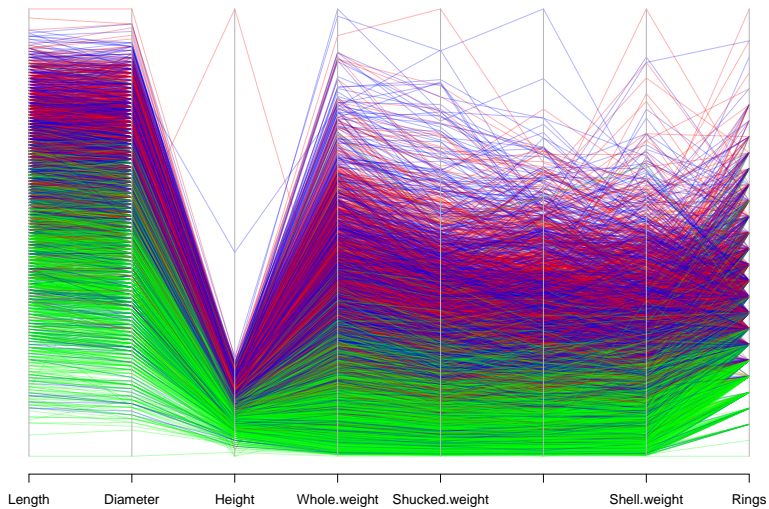


Abalone (UCI)

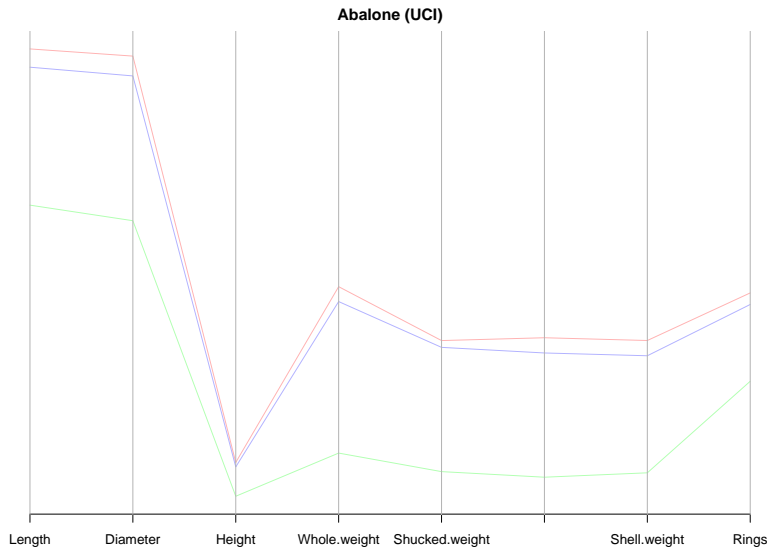


Clustering

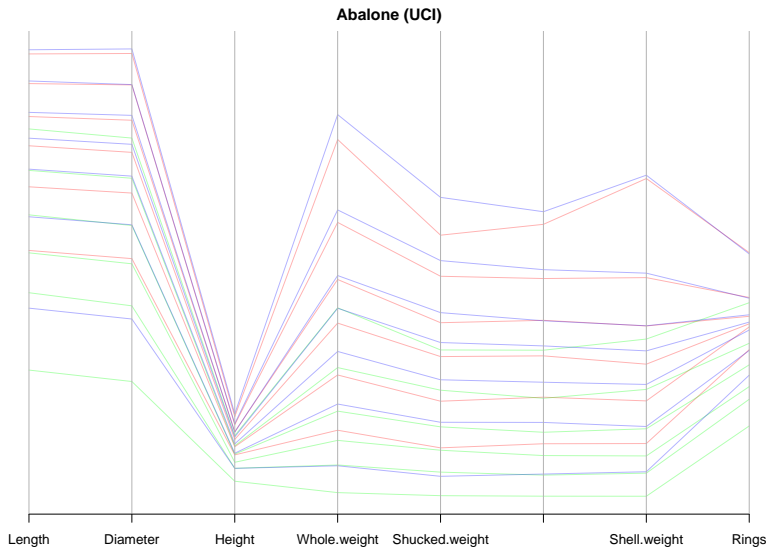
Abalone (UCI)



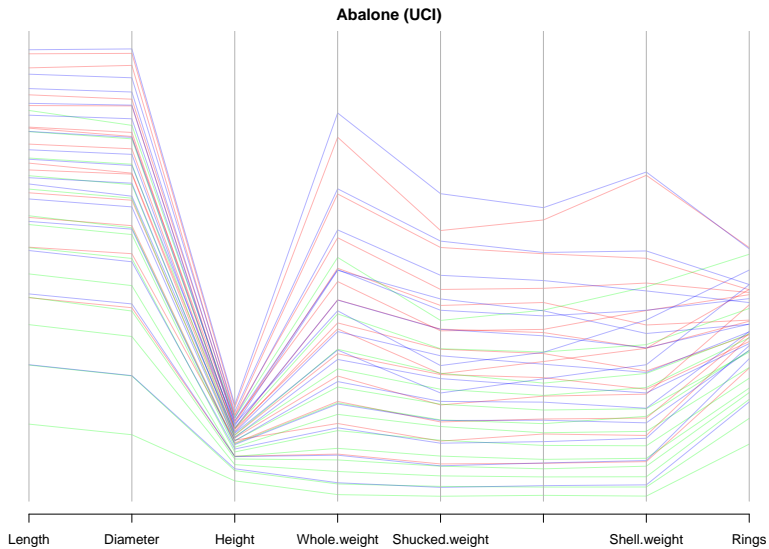
Clustering



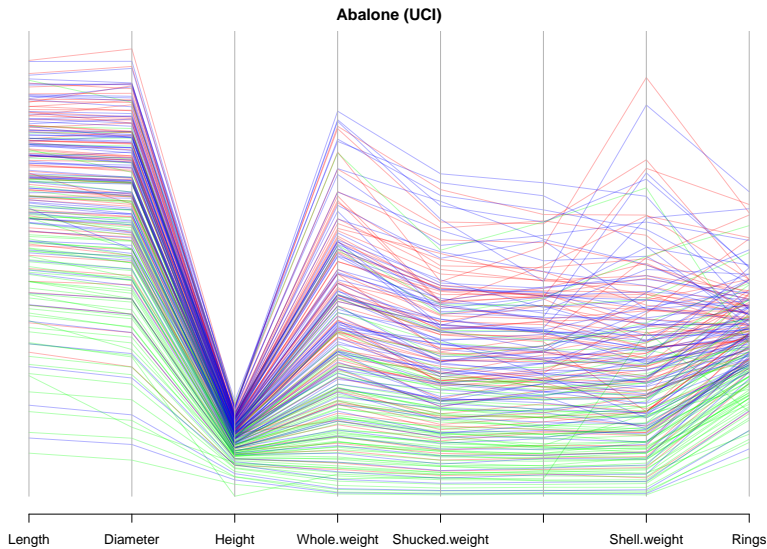
Clustering



Clustering

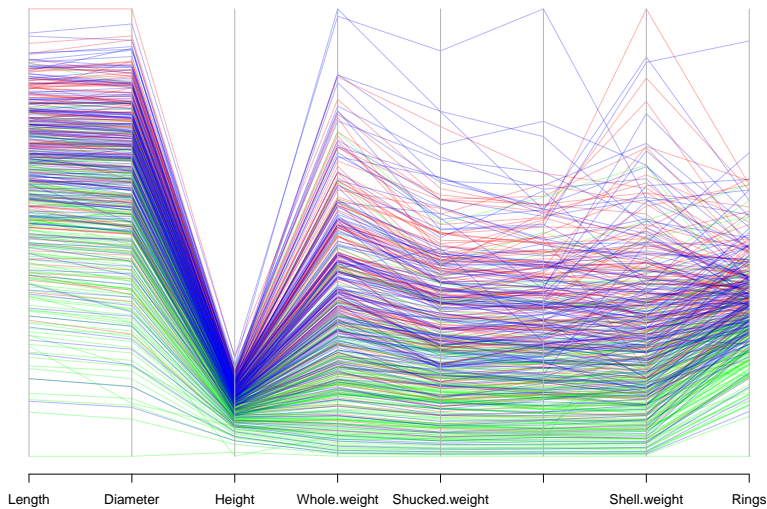


Clustering



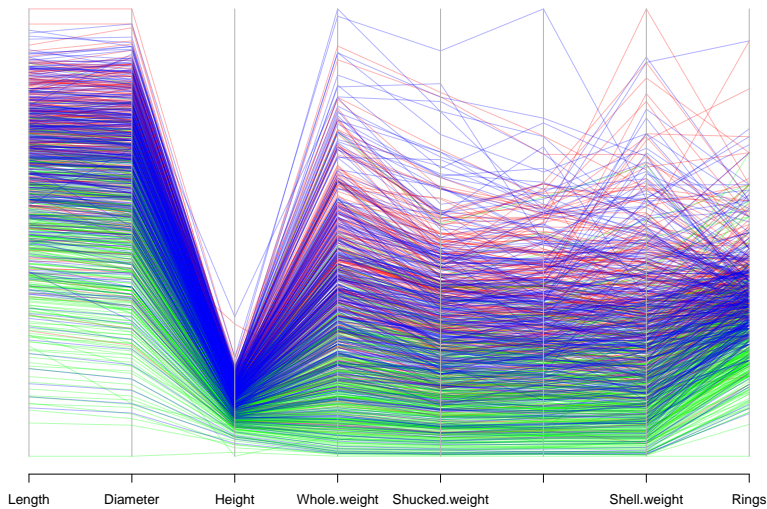
Clustering

Abalone (UCI)



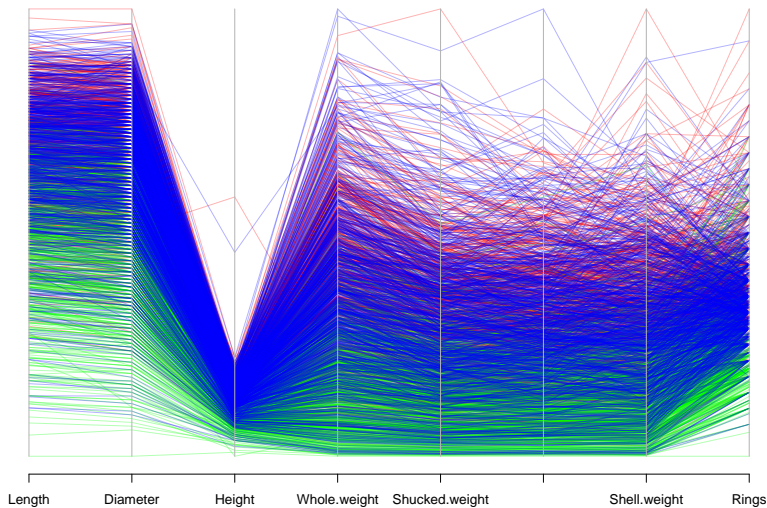
Clustering

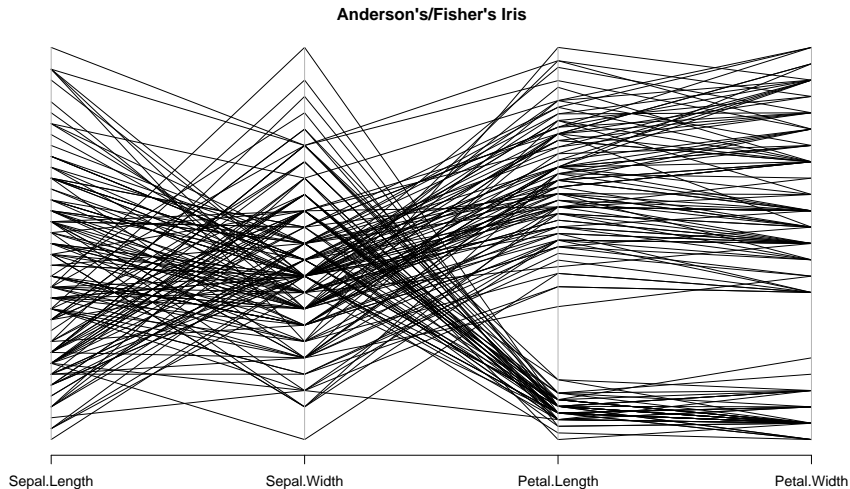
Abalone (UCI)



Clustering

Abalone (UCI)





Are petal variables correlated?



Are petal variables correlated?

Automated ordering

- resemblance measure between variables (e.g., correlation): $s(i, j)$
- variable permutation $\sigma : \{1, \dots, p\} \rightarrow \{1, \dots, p\}$
- optimal σ with respect to:

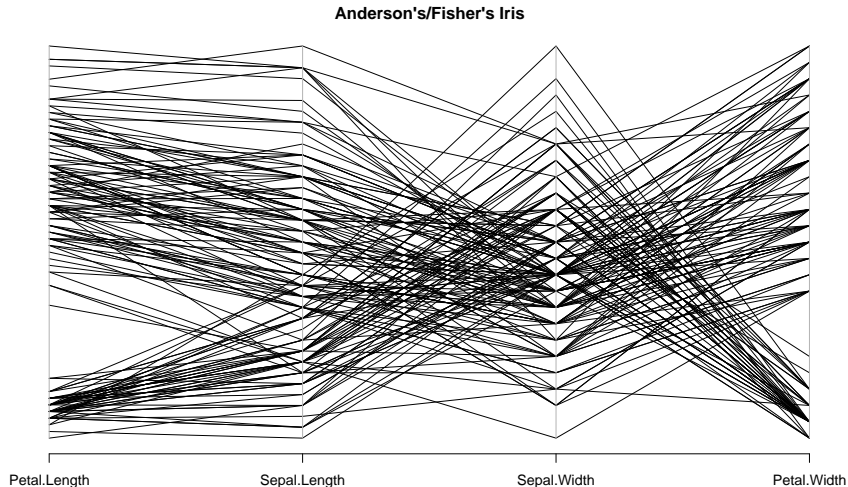
$$Q(\sigma) = \sum_{k=1}^{p-1} s(\sigma(k), \sigma(k+1))$$

- NP-complete (Ankerst, Berchtold & Keim, 1998) \Rightarrow heuristics:
 - traveling salesman analogy: optimization methods (e.g., Ant colony)
 - variable clustering
 - etc.
- an instance of the **seriation** problem: finding an optimal order among objects

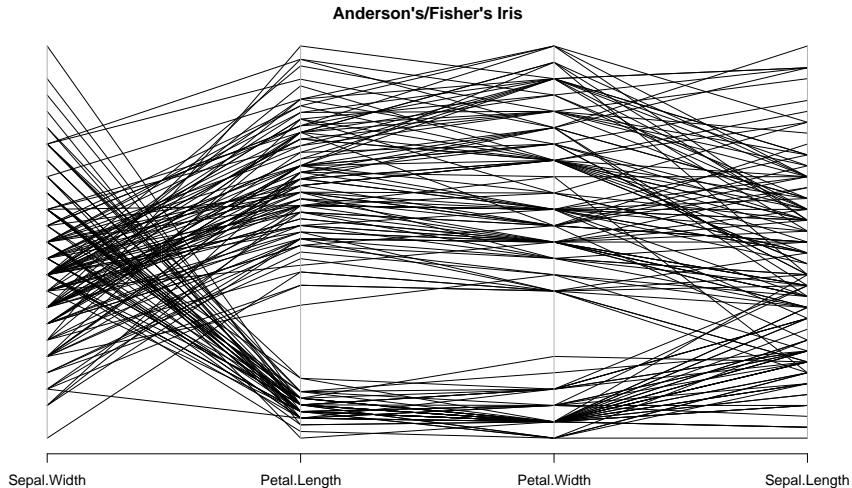
Projection to 1D:

- $d(i, j)$ dissimilarity between variables
- map a variable i to a real number y_i
- choose $(y_i)_{1 \leq i \leq p}$ such that $(y_i - y_j)^2 \simeq d(i, j)^2$:
 - many possibilities
 - see Prof. Lee's lecture
- order variables according to $(y_i)_{1 \leq i \leq p}$
- this is *exactly* a projection problem

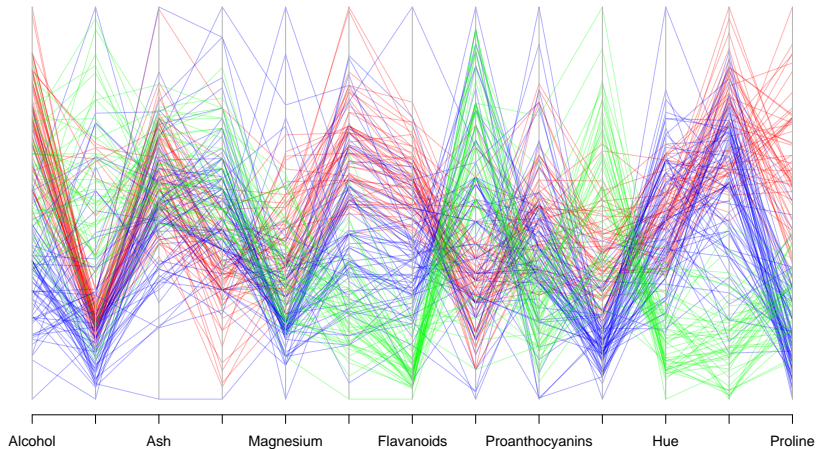
Ordering by Multi Dimensional Scaling



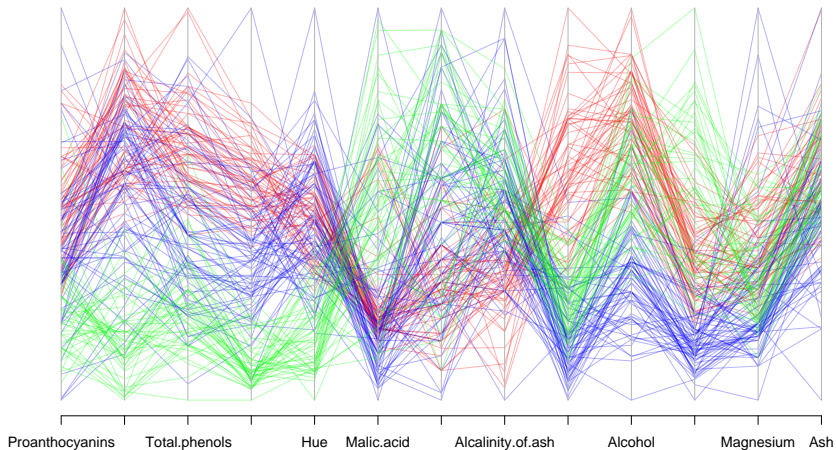
Ordering by Multi Dimensional Scaling



Ordering by Multi Dimensional Scaling



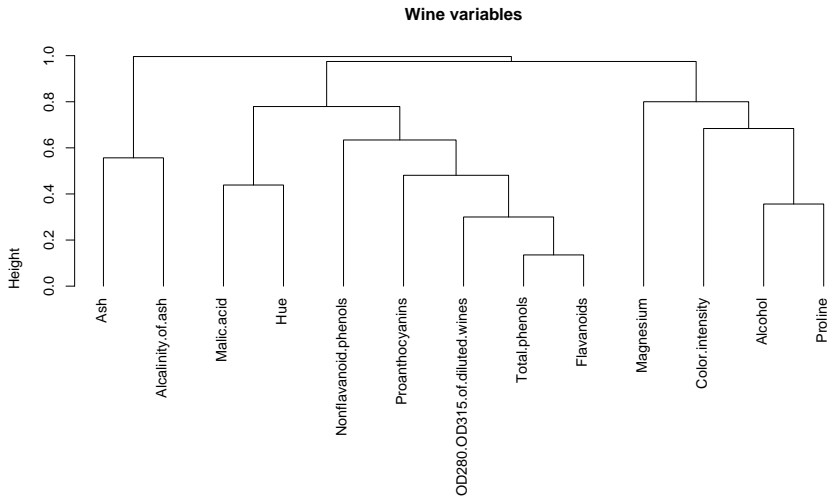
Ordering by Multi Dimensional Scaling



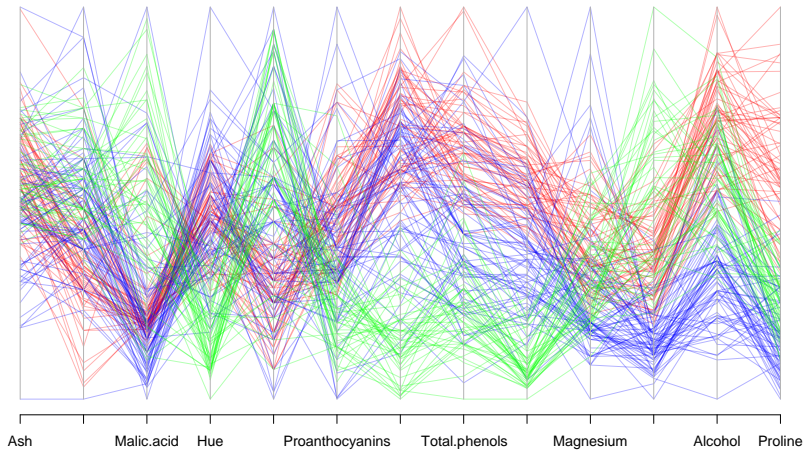
Clustering based approach

- Cluster variables according to resemblance measure
- Use clustering results to produce an order:
 - One dimensional Self Organizing Map
 - Hierarchical clustering:
 - with some heuristics
 - with optimal leaf ordering ($O(p^4)$, Bar-Joseph, Gifford and Jaakkola, 2001)
- Can be used to reduce the number of variables

Hierarchical clustering with heuristics



Hierarchical clustering with heuristics



Hierarchical clustering with heuristics

