VOS: A New Method for Visualizing Similarities between Objects

Nees Jan van Eck Ludo Waltman

Econometric Institute, Erasmus University Rotterdam P.O. Box 1738, 3000 DR Rotterdam, The Netherlands Email: {nvaneck,lwaltman}@few.eur.nl

> 50 Years of Econometrics June 10, 2006

Introduction

- A new method for visualizing similarities between objects is presented.
- The method is called VOS, which is an abbreviation for visualization of similarities.
- The objective of VOS is to provide a low-dimensional visualization in which objects are located in such a way that the distance between any pair of objects reflects their similarity as accurately as possible.
- The relationship between VOS and multidimensional scaling (MDS) is established.

Mathematical notation

- n: number of objects
- ▶ *m*: number of dimensions of the visualization
- ▶ $\mathbf{S} = (s_{ij})$: $n \times n$ similarity matrix satisfying $s_{ij} \ge 0$, $s_{ii} = 0$, and $s_{ij} = s_{ji}$ for all $i, j \in \{1, ..., n\}$
- ▶ X: $n \times m$ matrix containing the coordinates of the objects 1,..., n
- **x**_i: *i*th row of **X**, containing the coordinates of object *i*

Constrained optimization problem

A VOS solution is obtained by solving the following constrained optimization problem:

Minimize

$$E(\mathbf{X}; \mathbf{S}) = \sum_{i < j} s_{ij} \|\mathbf{x}_i - \mathbf{x}_j\|^2$$
(1)

subject to

$$\sum_{i < j} \|\mathbf{x}_i - \mathbf{x}_j\| = 1,$$
(2)

where $\|\cdot\|$ denotes the Euclidean norm.

Ideal coordinates

▶ The ideal coordinates of object *i* are given by

$$c_i(\mathbf{X}, \mathbf{S}) = \frac{\sum_j s_{ij} \mathbf{x}_j}{\sum_j s_{ij}}.$$
 (3)

Suppose the coordinates of all objects except object *i* are fixed. The objective function of VOS then reduces to

$$E_i(\mathbf{x}_i; \mathbf{X}, \mathbf{S}) = \sum_j s_{ij} \|\mathbf{x}_i - \mathbf{x}_j\|^2.$$
(4)

Minimization of (4) results in

$$\mathbf{x}_i = c_i(\mathbf{X}, \mathbf{S}). \tag{5}$$

 Apparently, VOS has the tendency to locate objects close to their ideal coordinates.

A simple example data set

Data set taken from Mardia, Kent, and Bibby (1979)¹:

$$n = 51$$
 $s_{ij} = \begin{cases}
8 & \text{if } 1 \leq |i-j| \leq 3 \\
7 & \text{if } 4 \leq |i-j| \leq 6 \\
\dots & \dots & \dots \\
1 & \text{if } 22 \leq |i-j| \leq 24 \\
0 & \text{otherwise}
 \end{cases}$

¹K.V. Mardia, J.T. Kent, and J.M. Bibby. *Multivariate analysis*. Academic Press, 1979.

Results of VOS and MDS



- The MDS solution demonstrates the horseshoe effect (Mardia et al., 1979).
- VOS seems to take into account indirect similarities via third objects.

Sammon mapping

Objective

To locate objects in a low-dimensional space in such a way that the distance between any pair of objects reflects their dissimilarity as accurately as possible.

Mathematical notation

▶ **D** =
$$(d_{ij})$$
: $n \times n$ dissimilarity matrix satisfying $d_{ij} > 0$ and $d_{ij} = d_{ji}$ for all $i, j \in \{1, ..., n\}$

Optimization problem

 A Sammon mapping solution is obtained by minimizing the following objective function

$$\sigma(\mathbf{X}; \mathbf{D}) = \sum_{i < j} \frac{(d_{ij} - \|\mathbf{x}_i - \mathbf{x}_j\|)^2}{d_{ij}}.$$
 (6)

Relationship between VOS and MDS

Conditional equivalence of VOS and Sammon mapping

Theorem

Let $s_{ij} > 0$ for all *i* and *j* ($i \neq j$), and let similarities be transformed into dissimilarities using $d_{ij} = s_{ij}^{-1}$ ($i \neq j$). VOS and Sammon mapping are then equivalent in the sense that VOS solutions and Sammon mapping solutions differ only by a multiplicative constant. A proof is provided in Van Eck and Waltman (2006)².

Conditional equivalence of VOS and weighted MDS

Sammon mapping is equivalent to weighted MDS where for each pair of objects *i* and *j* the weight equals d_{ij}^{-1} . It therefore follows from the above theorem that there also exists a conditional equivalence between VOS and weighted MDS.

²N.J. van Eck and L. Waltman. VOS: a new method for visualizing similarities between objects. Technical Report ERS-2006-020-LIS, Erasmus University Rotterdam, Erasmus Research Institute of Management, 2006.

Conclusions

VOS has the following properties:

- VOS has the tendency to locate objects close to their ideal coordinates.
- VOS seems to pay more attention to indirect similarities via third objects than MDS.
- VOS is, under certain conditions, equivalent to Sammon mapping and, as a consequence, to weighted MDS.

References

Description and analysis of VOS

 N.J. van Eck and L. Waltman. VOS: a new method for visualizing similarities between objects. Technical Report ERS-2006-020-LIS, Erasmus University Rotterdam, Erasmus Research Institute of Management, 2006.

Application of VOS to the visualization of associations between concepts based on co-occurrence data

N.J. van Eck, L. Waltman, J. van den Berg, and U. Kaymak. Visualizing the WCCI 2006 knowledge domain. In *Proceedings* of the 15th IEEE International Conference on Fuzzy Systems, 2006. Accepted for publication.

Related poster presentation

For a practical application of VOS and an experimental comparison between VOS and MDS, please visit the poster presentation *A Comparison of Knowledge Domain Visualization Approaches* by Nees Jan van Eck.