

## THE METHOD OF COMPUTATION in MDPREF<sup>1</sup>

- Given N subjects and p stimuli, we define a first-score matrix S (N x p)

$$S = \{ s(i,j) \}$$

of N subjects' ratings or rankings of the p stimuli (the data).

- The solution, consists of the two solution matrices

$$X = \{ x(i,a) \} \quad i = 1,2,\dots,N; \quad a = 1,2,\dots,r$$

(a configuration of N subject vectors in an r- dimensional space), and

$$Y = \{ y(j,a) \} \quad J = 1,2,\dots,p; \quad a = 1,2,\dots,r$$

( a configuration of p stimulus points an r-dimensional space).

- The data are related to the solution by means of a (fitted) second-score score matrix, S\* (N x p):

$$S^* = \{ s^*(i,j) \} = X.Y, \text{ and}$$

$$S^* \approx S \quad (\text{i.e. } S^* \text{ is a LS fit to } S)$$

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<sup>1</sup>This is based on Carroll (1964); see also MDS(X) Users' Guide and User Manual

- The solution is obtained by factoring (singular-value decomposition) s.t.

$$\begin{aligned} S^* &= U \beta V' \\ &= U_r \beta_r V_r' \end{aligned}$$

consisting of the first  $r$  columns of  $U$  and of  $V'$  respectively.

The solution-matrices are then given as :

$$X = U_r \beta_r; \text{ and } Y = \beta_r V_r \text{ resp.}$$

- This is an Eckart-Young factorization:

$U$  is a matrix with eigenvectors of  $SS'$  as its cols.

$V$  has as its columns the eigenvectors of  $S'S$

$\beta$  is a diagonal matrix of the corresponding eigenvalues,  $\lambda(j)$ .

- If the eigenvalues are ordered according to decreasing size, then  $X$  and  $Y$  (of rank  $r$ ) give the best LS approximation to  $S^*$ .