

# STAT406- Methods of Statistical Learning

## Lecture 25

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# Multidimensional scaling - Goal

- Given observations  $\mathbf{X}_1, \dots, \mathbf{X}_n$ 
  - find lower dimensional points  $\mathbf{Y}_1, \dots, \mathbf{Y}_n$  (e.g. in 2 or 3 dimensions)
  - that have a similar configuration (pair-wise distances) to that of the  $\mathbf{X}_j$ 's
- This is different from PCA

# Classical (Metric) Solution

- Look at the original pairwise distances

$$\delta_{ij} = \|\mathbf{X}_i - \mathbf{X}_j\|$$

construct  $\mathbf{Y}_1, \dots, \mathbf{Y}_n \in \mathbb{R}^k$  with

$$d_{ij} = \|\mathbf{Y}_i - \mathbf{Y}_j\| \approx \delta_{ij}$$

# Classical (Metric) Solution

- $\delta_{ij}$  are a dissimilarity if

$$\delta_{ii} = 0, \quad \delta_{ij} \geq 0, \quad \delta_{ij} = \delta_{ji}$$

- It is Euclidean if there exist  $\mathbf{w}_1, \dots, \mathbf{w}_n \in \mathbb{R}^r$  such that

$$\delta_{ij} = \|\mathbf{w}_i - \mathbf{w}_j\|$$

# Multidimensional scaling

- Theorem: (Gower, 1966; Mardia, 1979) Let  $a_{ij} = -\delta_{ij}^2/2$  and

$$\mathbf{B} = (\mathbf{I}_n - \mathbf{1}_n \mathbf{1}'_n/n) \mathbf{A} (\mathbf{I}_n - \mathbf{1}_n \mathbf{1}'_n/n)$$

then  $\delta$  is Euclidean iff  $\mathbf{B} \geq \mathbf{0}$

- Theorem is constructive

# Multidimensional scaling

- Finds  $\mathbf{y}_1, \dots, \mathbf{y}_n \in \mathbb{R}^f$  where  $f = \text{rank}(\mathbf{B})$  such that

$$\delta_{ij} = \|\mathbf{y}_i - \mathbf{y}_j\|$$

(the  $\mathbf{y}$ 's are the rows of the  $n \times f$  matrix of eigenvectors of  $\mathbf{B}$  with non-zero eigenvalues, properly standardized)

# MDS

- Do not need  $\mathbf{X}_j$ 's
- When  $\delta$  is not Euclidean some eigenvalues of  $\mathbf{B}$  are negative
- We can still take the  $k$  eigenvectors (properly standardized) associated to the largest positive eigenvalues

# Non-metric MDS

- Focus on the order of the pairwise distances (rather than their magnitudes)

$$\text{minimize STRESS}^2 = \frac{\sum (f(\delta_{ij}) - d_{ij})^2}{\sum d_{ij}^2}$$

where  $f$  is non-decreasing

# Non-metric MDS

- Start with initial  $d_{ij}$
- Find optimal  $\hat{f}(\delta_{ij})$  ( $\hat{f}$  non-decreasing)
- Given  $\hat{f}(\delta_{ij})$  find configuration with optimal  $d_{ij}$  in STRESS
- Iterate

# Languages

TABLE 12.3 NUMERALS IN 11 LANGUAGES

English (E)	Norwegian (N)	Danish (Da)	Dutch (Du)	German (G)	French (Fr)	Spanish (Sp)	Italian (I)	Polish (P)	Hungarian (H)	Finnish (Fi)
one	en	en	een	eins	un	uno	uno	jeden	egy	yksi
two	to	to	twee	zwei	deux	dos	due	dwa	ketto	kaksi
three	tre	tre	drie	drei	trois	tres	tre	trzy	harom	kolme
four	fire	fire	vier	vier	quatre	cuatro	quattro	cztery	negy	neua
five	fem	fem	vijf	funf	cinq	cinco	cinque	piec	ot	viisi
six	seks	seks	zes	sechs	six	scis	sei	szesc	hat	kuusi
seven	sju	syv	zeven	sieben	sept	siete	sette	siedem	het	seitseman
eight	atte	otte	acht	acht	huit	ochos	otto	osiem	nyolc	kahdeksan
nine	ni	ni	negen	neun	neuf	nueve	nove	dziewiec	kilenc	yhdeksan
ten	ti	ti	tien	zehn	dix	diez	dieci	dziesiec	tiz	kymmenen

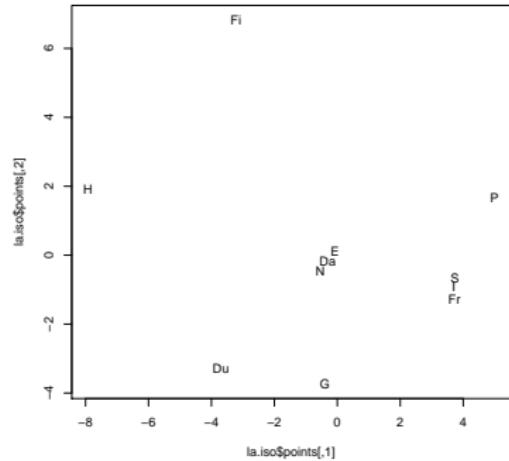
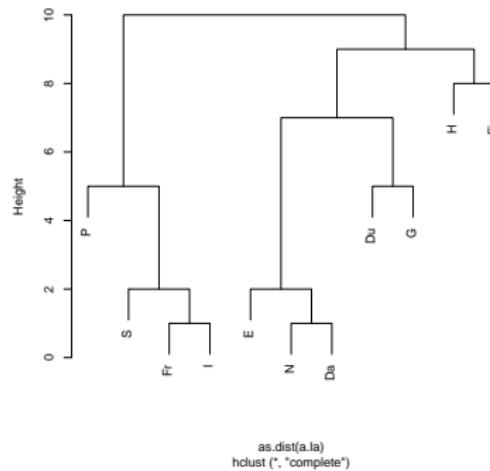
# Languages

## Dissimilarities

	E	N	Da	Du	G	Fr	S	I	P	H	Fi
E											
N	2										
Da	2	1									
Du	7	5	6								
G	6	4	5	5							
Fr	6	6	6	9	7						
S	6	6	5	9	7	2					
I	6	6	5	9	7	1	1				
P	7	7	6	10	8	5	3	4			
H	9	8	8	8	9	10	10	10	10		
Fi	9	9	9	9	9	9	9	9	9	9	8

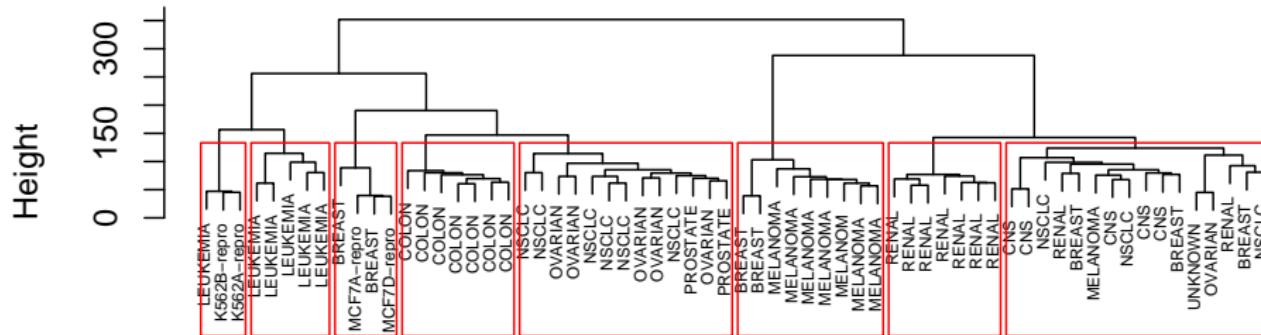
# Complete linkage + MDS

Cluster Dendrogram



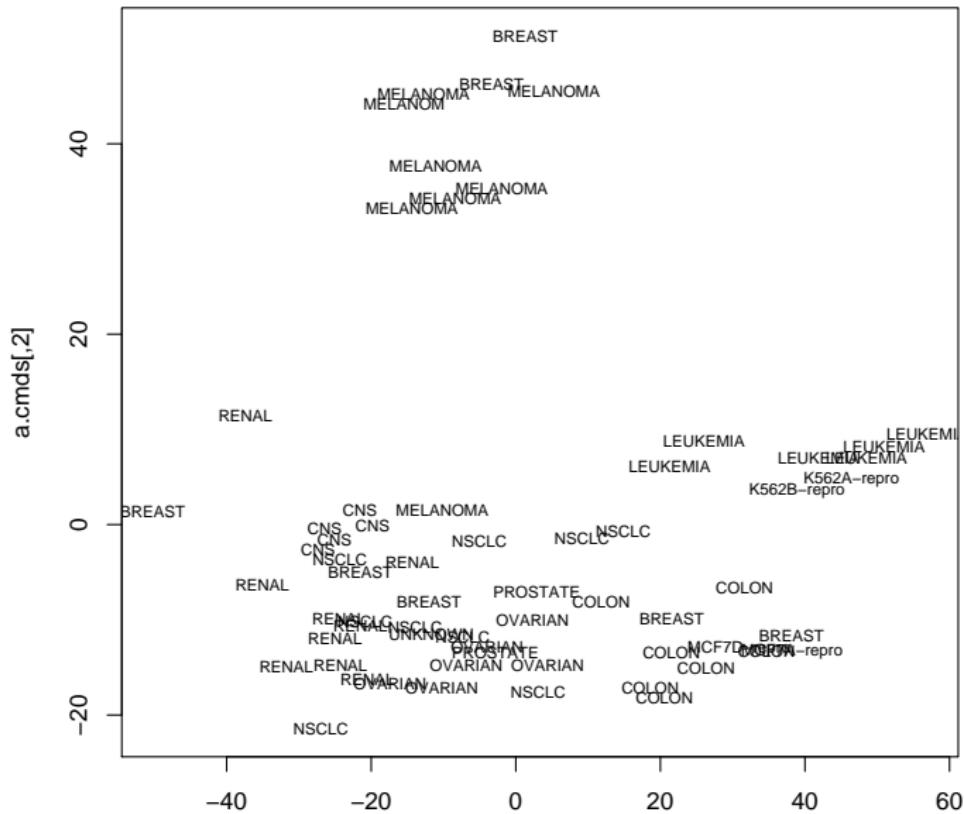
# Cancer

## Cluster Dendrogram

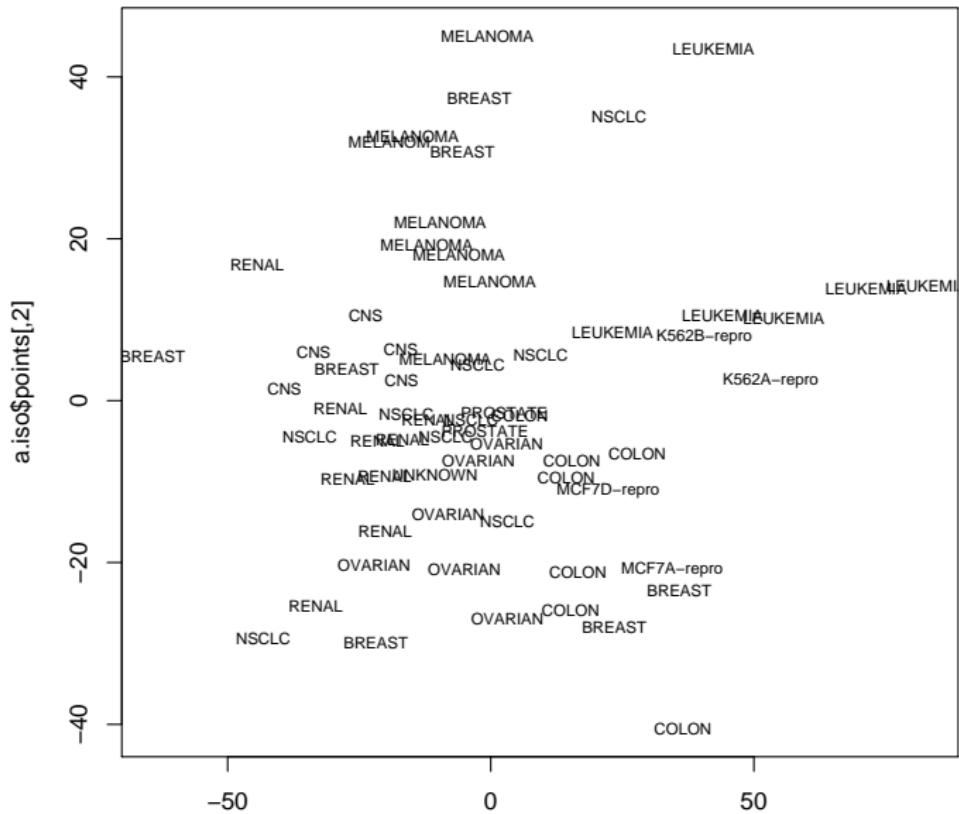


nci.dis  
hclust (\*, "ward")

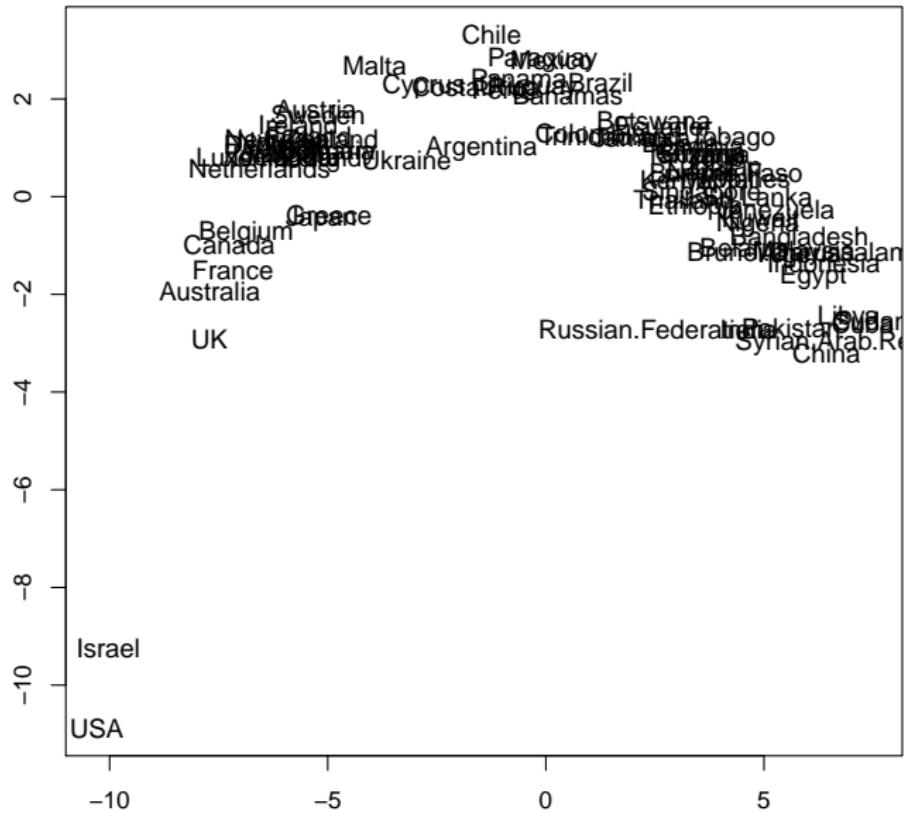
# Cancer



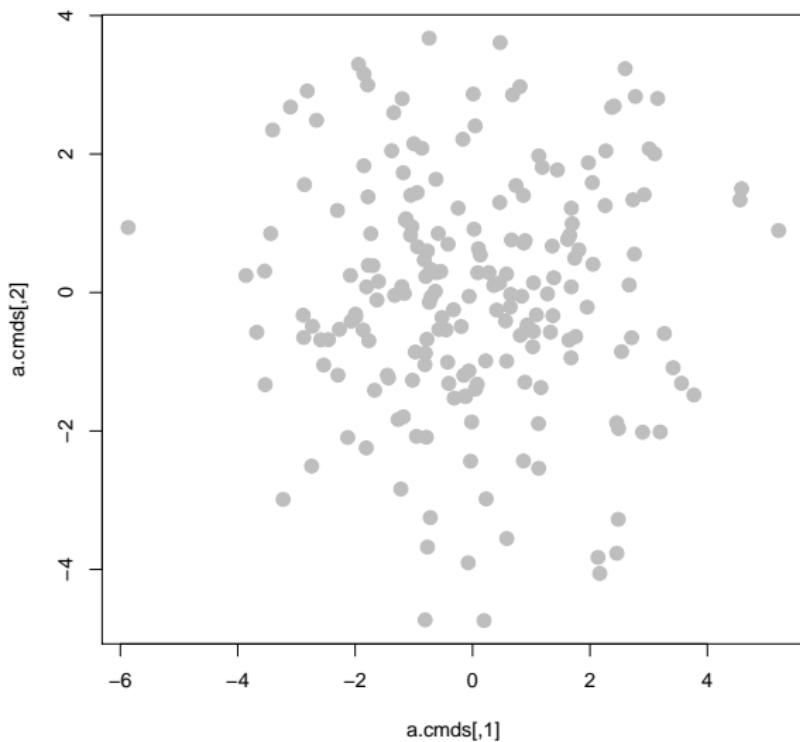
# Cancer



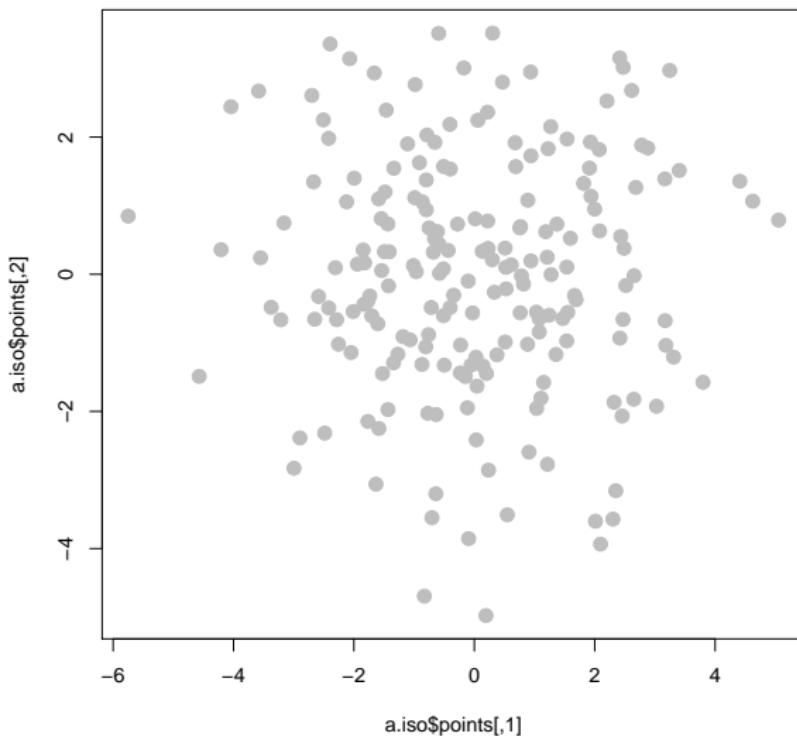
# UN Votes



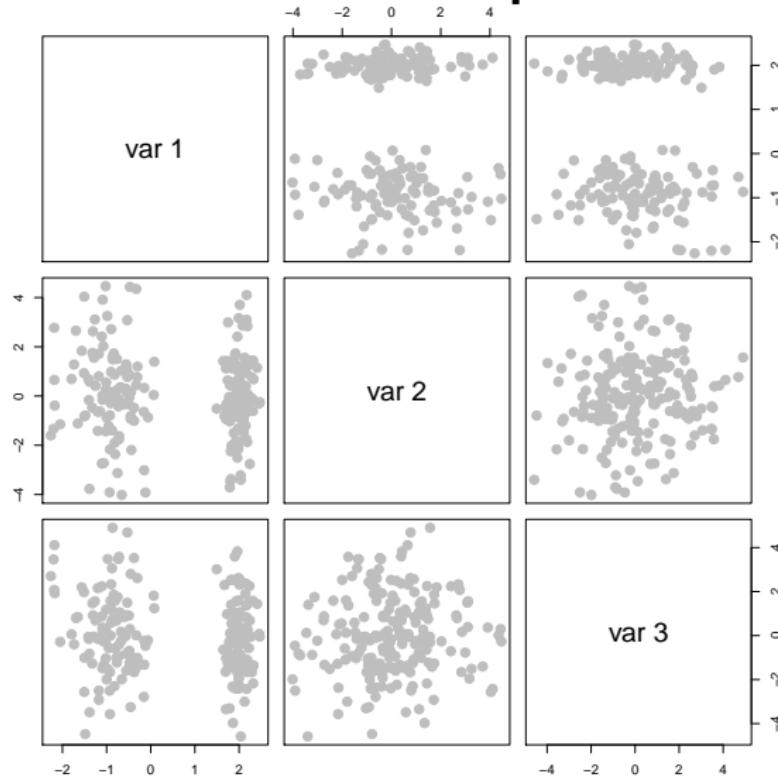
# A word of caution - cmdscale



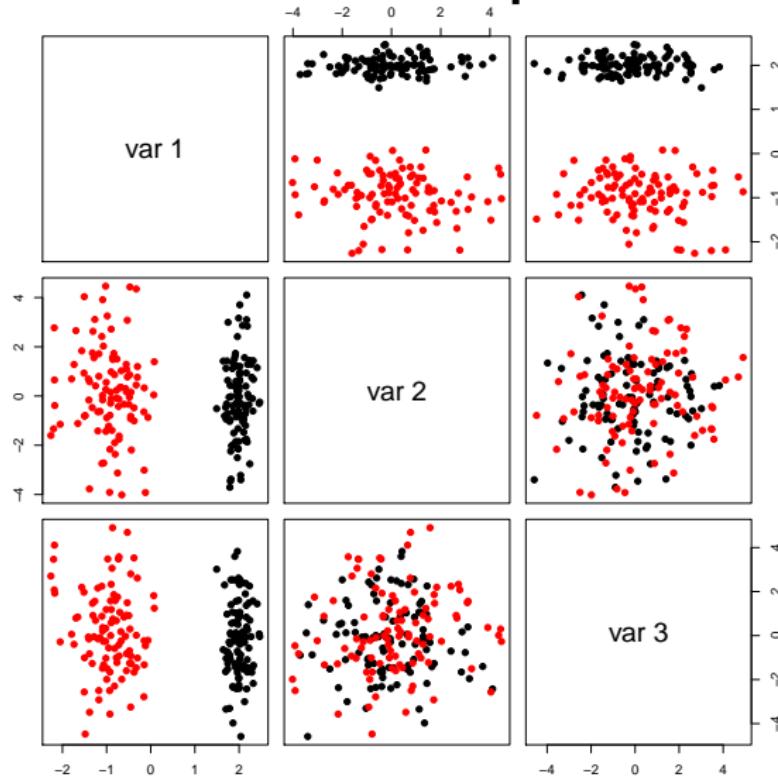
# A word of caution - isoMDS



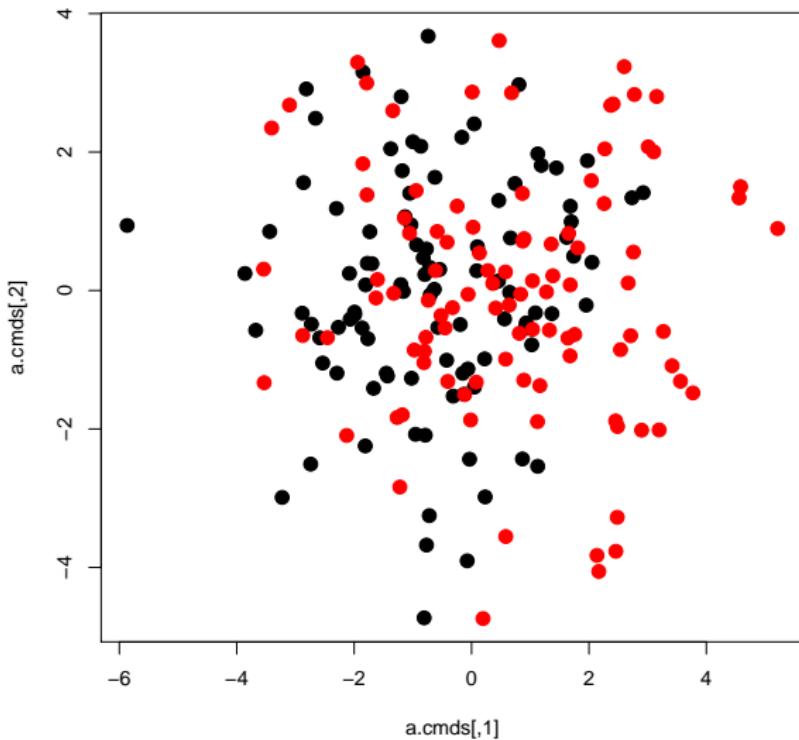
# A word of caution - pairs



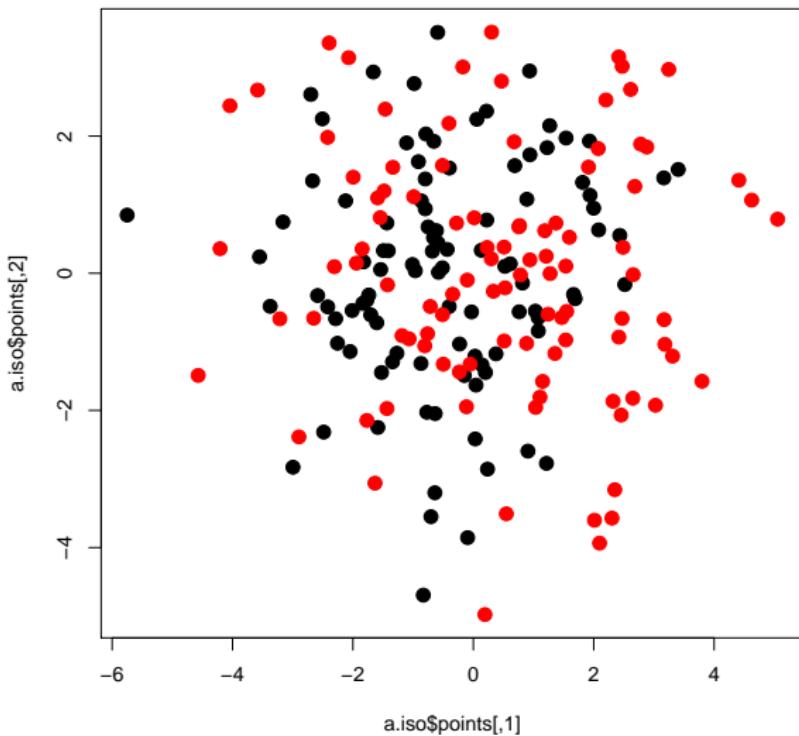
# A word of caution - pairs



# A word of caution - cmdscale



# A word of caution - isoMDS



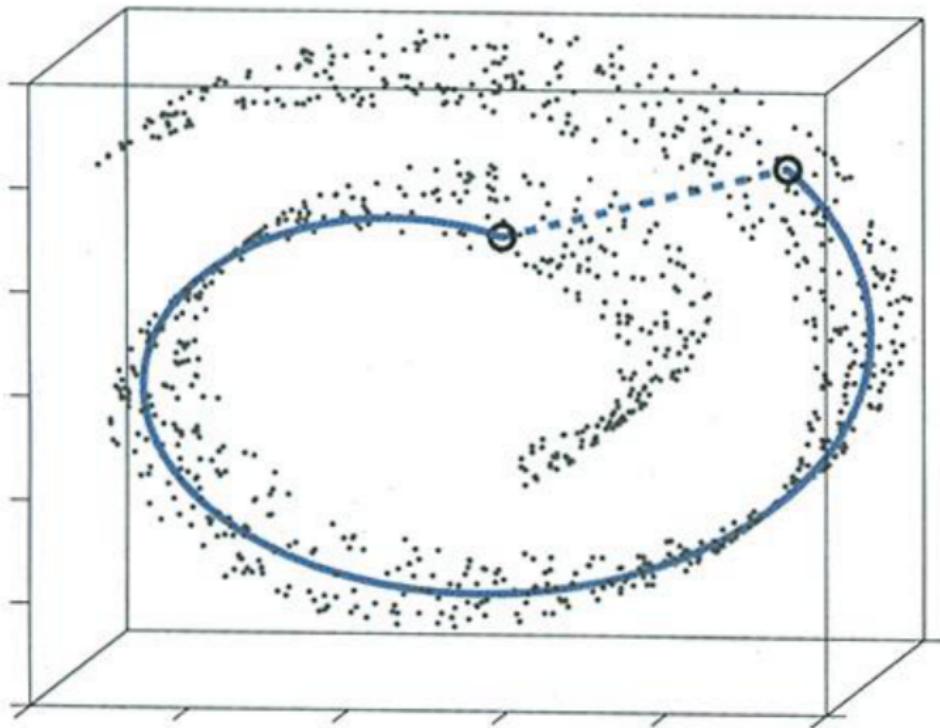
# ISOMAP

- Construct a graph to approximate geodesic distances
- Use local distances (e.g. to closest neighbours)
- Distance on manifold is approximated by distance on the graph (shortest path)
- Apply classical MDS to the graph distances

Tenenbaum, J. B., de Silva, V. and Langford, J. C. (2000). A global geometric framework for nonlinear dimensionality reduction, *Science* 290: 2319-2323

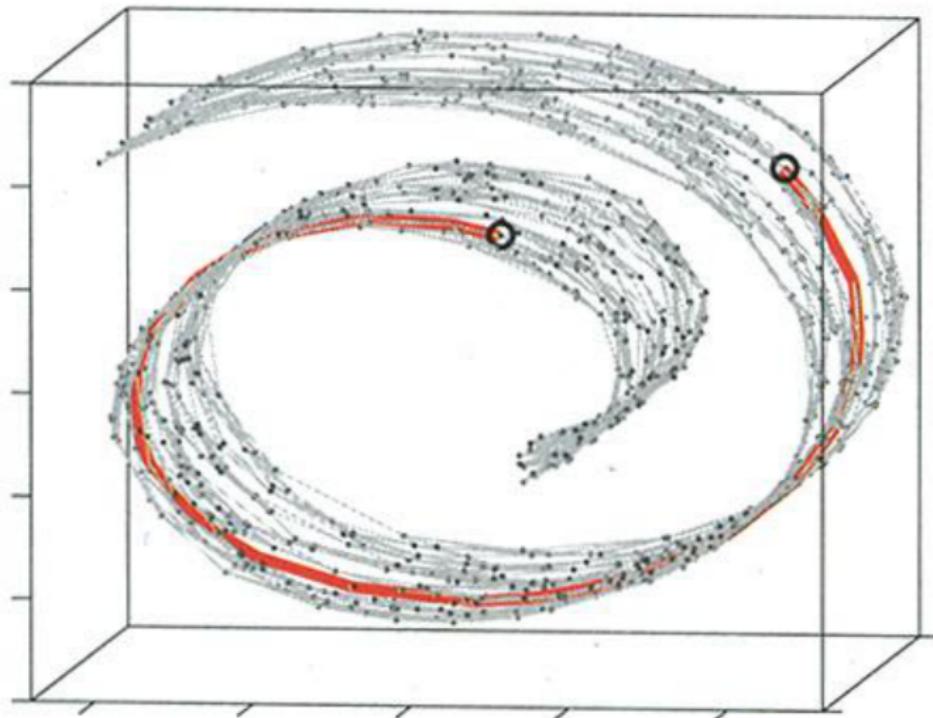
# ISOMAP

A



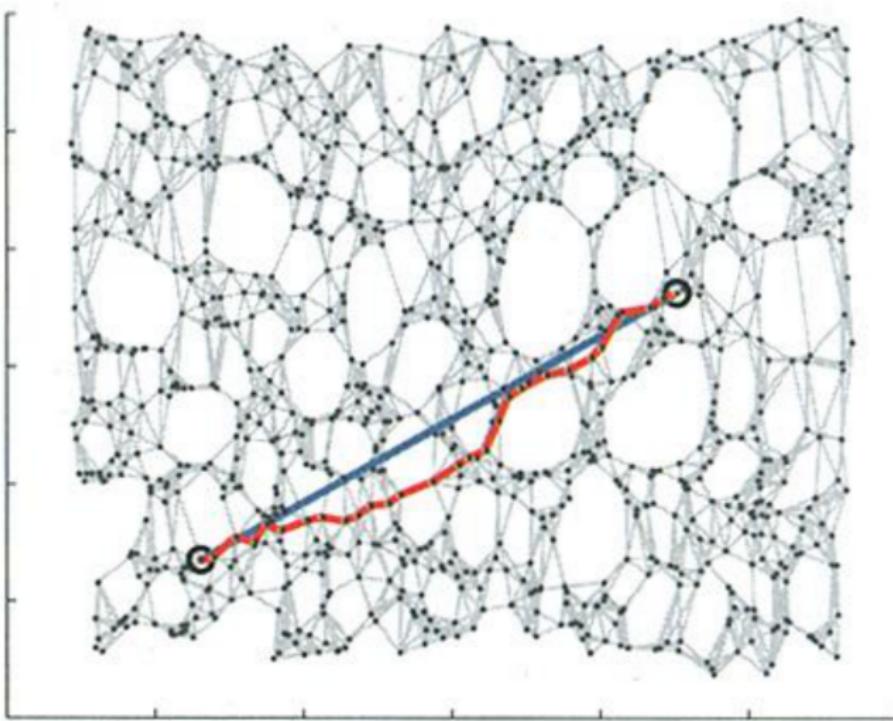
# ISOMAP

B

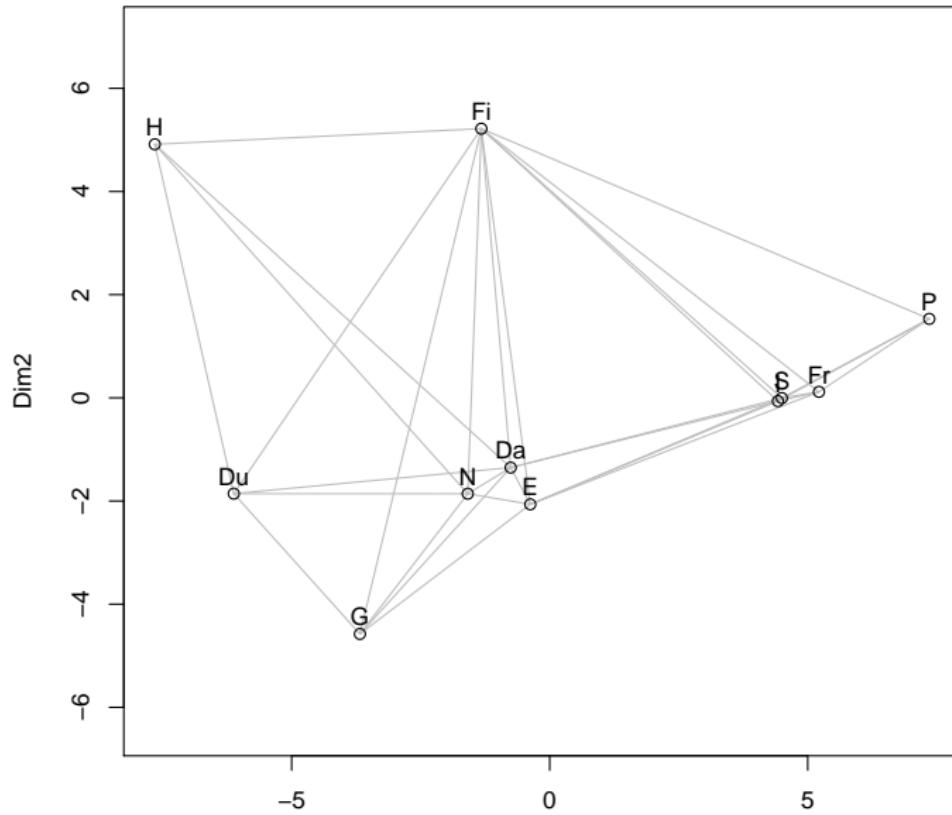


# ISOMAP

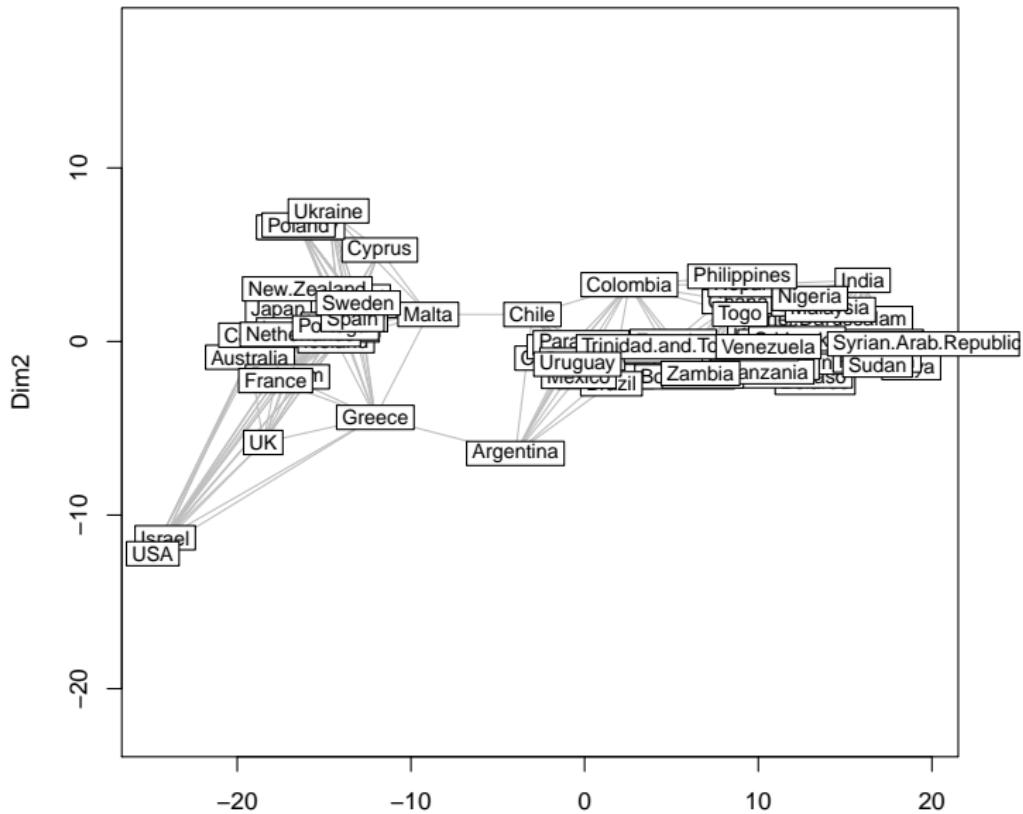
C



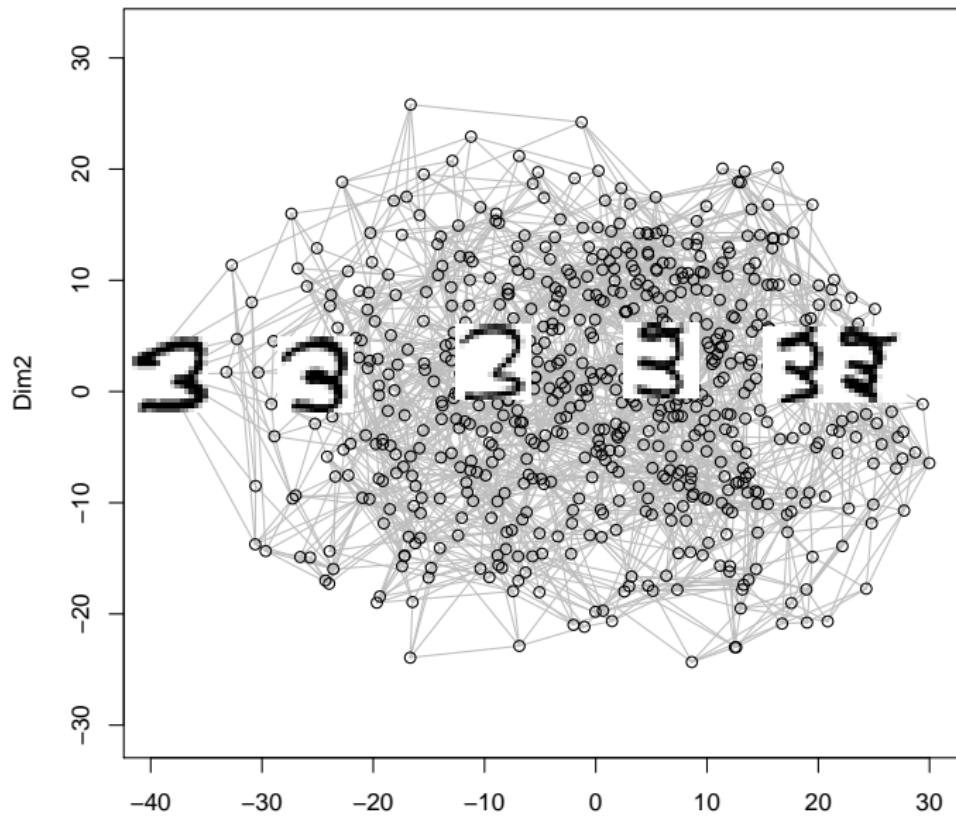
# ISOMAP - Languages



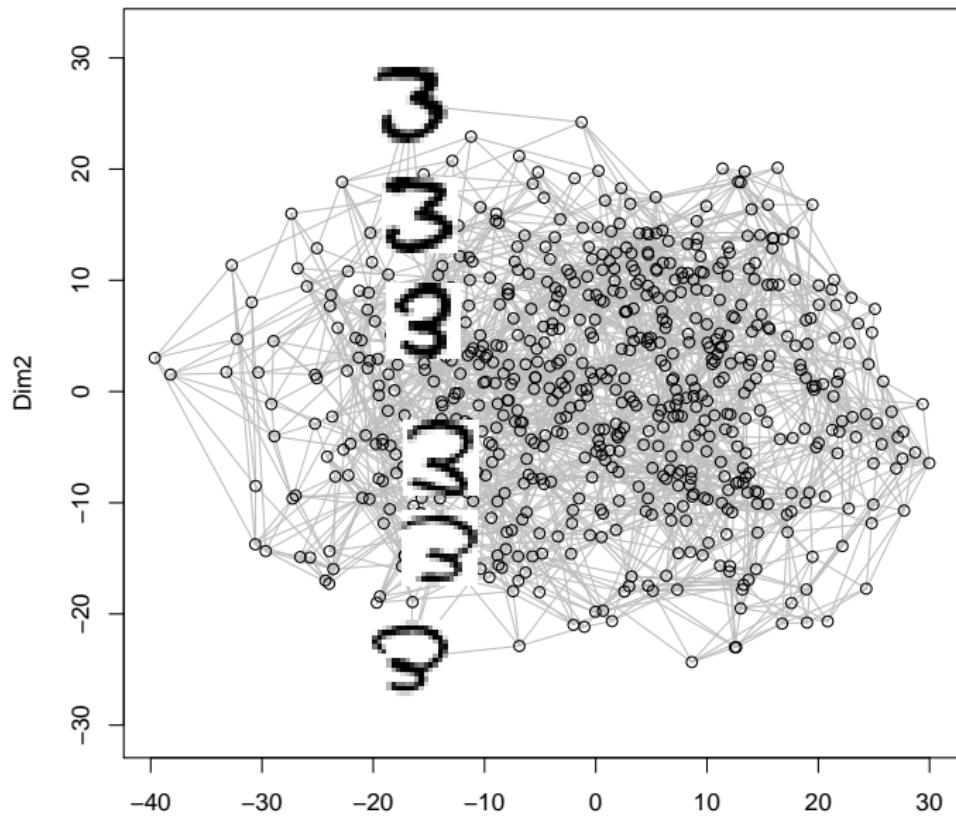
# ISOMAP - UN votes



# ISOMAP - Digits



# ISOMAP - Digits



Original content from Tenenbaum et al (2000), Science

# Multidimensional scaling

- $\delta_{ij}$  are a dissimilarity if

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- It is Euclidean if there exist  $\mathbf{w}_1, \dots, \mathbf{w}_n \in \mathbb{R}^r$  such that

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