

# PROXIMITY/DISTANCE MEASURES- PROBLEM SOLVING

Dr. Umarani Jayaraman  
Assistant Professor



INDIAN INSTITUTE OF INFORMATION TECHNOLOGY,  
DESIGN AND MANUFACTURING,  
KANCHEEPURAM

# Topic:

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- Solved Problems

# Mahalanobis distance

- We know the quadratic form distance

$$d_Q(\mathbf{x}, \mathbf{y}) = \sqrt{(\mathbf{x} - \mathbf{y})^T \mathbf{A} (\mathbf{x} - \mathbf{y})}$$

- Replace  $\mathbf{A}$  in quadratic form distance by inverse of covariance matrix  $\Sigma$  to get **Mahalanobis distance**

$$d_M(\mathbf{x}, \mathbf{y}) = \sqrt{(\mathbf{x} - \mathbf{y})^T \Sigma^{-1} (\mathbf{x} - \mathbf{y})}$$

# Example 1: Mahalanobis Distance

- Suppose you have data for five people, and each person vector has a height, score on some test, and an age.

X	Y	Z
Height	Score	Age
64	580	29
66	570	33
68	590	37
69	660	46
73	600	55

mean

68	600	40
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# Steps to compute mahanalobis distance

- Find the mean vector
- Find the covariance metrics
- Find the inverse of covariance metrics
- Then multiply it with  $(X-Y)$ , let say  $C_1$
- Then  $(X-Y)^T \cdot C_1$

# Example 1: Mahalanobis Distance

- The mean of the data is  $(68.0, 600.0, 40.0)$ .
- Now suppose you want to know how far another person,  $v = (66, 640, 44)$ , is from this data.
- It turns out the Mahalanobis Distance is 5.33 (no units).

# Example 1: Mahalanobis Distance

$$\square Z = \begin{bmatrix} -4 & -20 & -11 \\ -2 & -30 & -7 \\ 0 & -10 & -3 \\ 1 & 60 & 6 \\ 5 & 0 & 15 \end{bmatrix}$$

□ Find the covariance matrix

$$\square \Sigma = \frac{1}{N-1} Z^T Z$$

# Example 1: Mahalanobis Distance

$$\Sigma = \frac{1}{4} \begin{bmatrix} -4 & -2 & 0 & 1 & 5 \\ -20 & -30 & -10 & 60 & 0 \\ -11 & -7 & -3 & 6 & 15 \end{bmatrix} \begin{bmatrix} -4 & -20 & -11 \\ -2 & -30 & -7 \\ 0 & -10 & -3 \\ 1 & 60 & 6 \\ 5 & 0 & 15 \end{bmatrix}$$
$$= \frac{1}{4} \begin{bmatrix} 46 & 200 & 139 \\ 200 & 5000 & 820 \\ 139 & 820 & 440 \end{bmatrix}$$



# Example 1: Mahalanobis Distance

$$\square \Sigma = \begin{bmatrix} 11.50 & 50 & 34.75 \\ 50 & 1250 & 205 \\ 34.75 & 205 & 110 \end{bmatrix}$$

$$\square \Sigma^{-1} = \begin{bmatrix} 3.6885 & 0.0627 & -1.2821 \\ 0.0627 & 0.0022 & -0.0240 \\ -1.2821 & -0.0240 & 0.4588 \end{bmatrix}$$

$$\square D(h_q, h_t) = (h_q - h_t)^T A (h_q - h_t) \text{ --- Eqn(1)}$$

$$\square \text{ Here } A = \Sigma^{-1}$$

# Example 1: Mahalanobis Distance

□  $Eqn(1)$  becomes

$$\begin{aligned} & [-2 \quad 40 \quad 4] \begin{bmatrix} 3.6885 & 0.0627 & -1.2821 \\ 0.0627 & 0.0022 & -0.0240 \\ -1.2821 & -0.0240 & 0.4588 \end{bmatrix} \begin{bmatrix} -2 \\ 40 \\ 4 \end{bmatrix} \\ &= [-9.9964 \quad -0.1325 \quad 3.4413] \begin{bmatrix} -2 \\ 40 \\ 4 \end{bmatrix} \\ &= \sqrt{28.4573} \\ &= 5.33 \end{aligned}$$

# Quadratic form distance

$$d_Q(x, y) = \sqrt{(x - y)^T A (x - y)}$$

- For example  $A_{ij} = 1 - c_{ij}/c_{\max}$  for color histograms
- $c_{ij}$  is bin-to-bin distance and  $c_{\max}$  the maximum distance
- Note
  - ▣ If  $A$  is an identity matrix, then **Euclidean**
  - ▣ If  $A$  is a diagonal matrix, then **weighted Euclidean**
  - ▣ If  $A$  is a inverse of covariance matrix, then **Mahalanobis distance**
  - ▣ **Is it a Metric? Yes , if  $A$  is positive definite**

## Example 2: Quadratic form distance

- Given the histogram of a pure red image

$$h_q = [1, 0, 0]^T$$

and a pure orange image:

$$h_t = [0, 1, 0]^T$$

- The transformation matrix  $A = \begin{bmatrix} 1 & 0.9 & 0 \\ 0.9 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$

- Find the **quadratic form distance** and the Euclidean distance between them.

## Example 2: Quadratic form distance

□  $D(h_q, h_t) = (h_q - h_t)^T A (h_q - h_t) \text{ --- Eqn(1)}$

□  $h_q = [1 \ 0 \ 0]^T$

$$= \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}$$

□  $h_t = [0 \ 1 \ 0]^T$

$$= \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}$$

## Example 2: Quadratic form distance

- $h_q - h_t = \begin{bmatrix} 1 \\ -1 \\ 0 \end{bmatrix}$
- $(h_q - h_t)^T = [1 \quad -1 \quad 0]$  --- Eqn(2)
- Substituting Eqn (2) and value of  $A$  in Eqn (1),
- $D(h_q, h_t) = [1 \quad -1 \quad 0] \begin{bmatrix} 1 & 0.9 & 0 \\ 0.9 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 \\ -1 \\ 0 \end{bmatrix}$

## Example 2: Quadratic form distance

$$\begin{aligned}\square D(h_q, h_t) &= [1 \quad -1 \quad 0] \begin{bmatrix} 1 & 0.9 & 0 \\ 0.9 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 \\ -1 \\ 0 \end{bmatrix} \\ &= [0.1 \quad -0.1 \quad 0] \begin{bmatrix} 1 \\ -1 \\ 0 \end{bmatrix} \\ &= (0.1) + (0.1) + 0 \\ &= 0.2\end{aligned}$$

# Quadratic form distance

$$d_Q(x, y) = \sqrt{(x - y)^T A (x - y)}$$

- For example  $A_{ij} = 1 - c_{ij}/c_{\max}$  for color histograms
- $c_{ij}$  is bin-to-bin distance and  $c_{\max}$  the maximum distance
- Note
  - ▣ If  $A$  is an identity matrix, then **Euclidean**
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  - ▣ **Is it a Metric? Yes , if  $A$  is positive definite**



# Example 3: Euclidean Distance

$$\begin{aligned}\square D^2(h_q, h_t) &= (h_q - h_t)^2 \\ &= [1 \quad -1 \quad 0] \begin{bmatrix} 1 \\ -1 \\ 0 \end{bmatrix} \\ &= [1 + 1 + 0] \\ &= 2\end{aligned}$$
$$\square D(h_q, h_t) = \sqrt{2}$$

# Summary-Metric Similarity Measures

## □ Solved Problems

- Mahalanobis distance
- Quadratic form distance
- Euclidean distance

THANK YOU