

Proximity measures:

In order to classify patterns, they need to be compared against each other and against a standard. When a new pattern is present and it is necessary to classify it.

* Distance measure:

A distance measure is used to find the dissimilarity between pattern representations. Patterns which are more similar should be closer. The distance function could be a metric or a non-metric.

A metric is a measure for which the following properties hold:

i) positive reflexivity $d(x, x) = 0$

1. $d(x, y) \geq 0$ - non-negativity

2. $d(x, y) = 0 \Leftrightarrow x = y$ identity.

ii) symmetry $d(x, y) = d(y, x)$

iii) triangular inequality $d(x, y) \leq d(x, z) + d(z, y)$

The popularly used distance metric called 'Minkowski metric' is of the form

$$L_p(x, y) = \left[\sum_{i=1}^d (|x_i - y_i|)^p \right]^{\frac{1}{p}} \text{ - which is also called as } L_p \text{ Norm.}$$

When p is 1; it is called the Manhattan distance / L1 distance / Taxicab distance / city-block distance

		(i, j+2)		
		(i-1, j+2)	(i, j+1)	(i+1, j+1)
		(i-2, j)	(i-1, j)	(i, j)
		(i-1, j-1)	(i, j-1)	(i+1, j-1)
		(i, j-2)		

$$X = \langle x_1, x_2, \dots, x_d \rangle ;$$

$$Y = \langle y_1, y_2, \dots, y_d \rangle ;$$

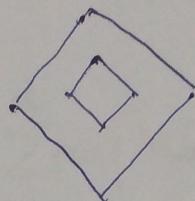
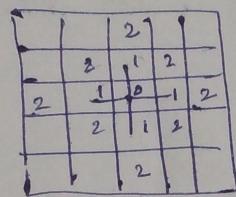
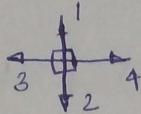
when $d=2$ $X = \langle x_1, x_2 \rangle$
 $Y = \langle y_1, y_2 \rangle$

(2)

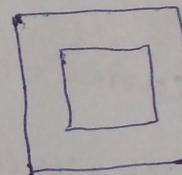
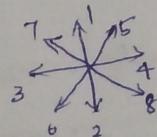
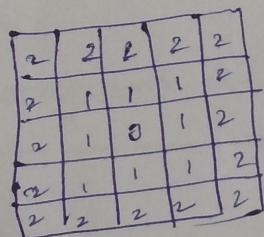
$$011120 \\ No$$

$$L_1(x, y) = |x_1 - y_1| + |x_2 - y_2|$$

Based on 4-connectivity ; the L_1 norm is called
as city block distance



Based on 8-connectivity ; the L_1 norm is called
as chess board distance
max distance



when $p=2$; It is called Euclidean distance

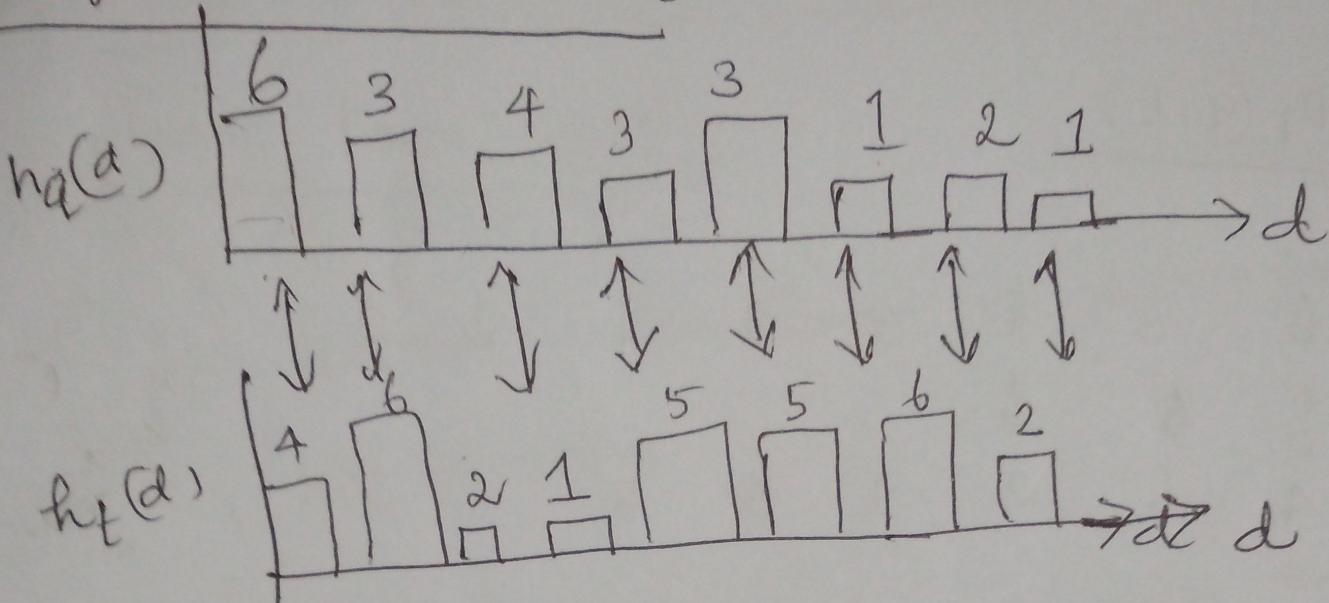
$$L_2(x, y) = \left[(x_1 - y_1)^2 + (x_2 - y_2)^2 \right]^{\frac{1}{2}}$$

$$= \sqrt{(x_1 - y_1)^2 + (x_2 - y_2)^2}$$

L_∞ : max (chess board distance)

$$L_{-\infty} = \min$$

histogram L₁ distance:



$$\sum_{i=1}^d \min[h_q(i), h_t(i)]$$

$$D_1(q, t) = 1 - \frac{\sum_{i=1}^d \min[h_q(i), h_t(i)]}{\sum_{i=1}^d \max[h_q(i), h_t(i)]}$$

If histograms are normalized ~~heights~~

$$D_1(q, t) = \sum_{i=1}^d [h_q(i) - h_t(i)]$$

According to $D_1(q, t)$

$$q = \begin{matrix} 6 & 3 & 4 & 3 & 3 & 1 & 2 & 1 \\ \downarrow & \downarrow & \downarrow & & & & & \\ t = 4 & 6 & 2 & 1 & 5 & 5 & 6 & 2 \end{matrix}$$

$$\min(6, 4) = \dots$$

$$\frac{4+3+2+1+3+1+2+1}{\min(111, 111)} = \frac{17}{1} = \underline{\underline{17}}$$

This captures similarity between two histograms

$$q \text{ and } t \quad D_1(q, t) = \underline{\underline{17}}$$

According to $D_2(q, t)$:

$$D_2(q, t) = (6-4)^2 + (3-6)^2 + (4-2)^2 + (3-1)^2 + (3-5)^2 + (1-5)^2 + (2-6)^2 + (1-2)^2$$

$$= \sqrt{4+9+4+4+4+16+16+1}$$

$$= \sqrt{62}$$

$$= \underline{\underline{7.87}}$$