



# Digital Image Processing

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## Solved Example 1 K-Mean clustering

i	x	y
X1	2	6
X2	3	4
X3	3	8
X4	4	7
X5	6	2
X6	6	4
X7	7	3
X8	7	4
X9	8	5
X10	7	6

- Apply K-Medoid clustering algorithm to form two clusters.
- Use Manhattan distance to find the between data point and medoid.

## Solved Example 1 K-Mean clustering

### Step 1

- Select two medoids
- C1=(3, 4)
- C2=(7, 4)
- *Manhattan Dist* =  $|x_1 - x_2| + |y_1 - y_2|$
- $Mdist[(2, 6), (3, 4)] = |2 - 3| + |6 - 4| = \underline{3}$
- $Mdist[(3, 4), (3, 4)] = |3 - 3| + |4 - 4| = \underline{0}$



i	x	y	C1	C2	Cluster
X1	2	6	3		
X2	<u>3</u>	<u>4</u>	0		
X3	3	8	4		
X4	4	7	4		
X5	6	2	5		
X6	6	4	3		
X7	7	3	5		
X8	7	4	4		
X9	8	5	6		
X10	7	6	6		

## Solved Example 1 K-Mean clustering

### Step 1

- Select two medoids

- $C1=(3, 4)$

- $C2=(7, 4)$

- **Manhattan Dist** =  $|x_1 - x_2| + |y_1 - y_2|$

$$|2-7| + |6-4| = 5 + 2 = 7$$

$$|3-7| + |4-4| = 4 + 0 = 4$$

i	x	y	C1	C2	Cluster
X1	2	6	3	7	
X2	3	4	0	4	
X3	3	8	4	8	
X4	4	7	4	6	
X5	6	2	5	3	
X6	6	4	3	1	
X7	7	3	5	1	
X8	7	4	4	0	
X9	8	5	6	2	
X10	7	6	6	2	

## Solved Example 1 K-Mean clustering

If  $C1 < C2$  then result is C1 otherwise C2

Step 2

- Cluster are
- C1:  $\{(2,6), (3,4), (3,8), (4,7)\}$
- C2:  $\{(6, 2), (6, 4), (7, 3), (7, 4), (8, 5), (7,6)\}$

i	x	y	C1	C2	Cluster
X1	2	6	3	7	C1
X2	3	4	0	4	C1
X3	3	8	4	8	C1
X4	4	7	4	6	C1
X5	6	2	5	3	C2
X6	6	4	3	1	C2
X7	7	3	5	1	C2
X8	7	4	4	0	C2
X9	8	5	6	2	C2
X10	7	6	6	2	C2

## Solved Example 1 K-Mean clustering

- C1: {(2,6), (3,4), (3,8), (4,7)}
- C2: {(6, 2), (6, 4), (7, 3), (7, 4), (8, 5), (7,6)}

- Calculate the Total Cost

- $Cost(c, x) = \sum_i |c_i - x_i|$

$$= |3 - 2| + |4 - 6| = 1 + 2 = 3$$

- $Total Cost = \{Cost((3,4), (2,6)) + Cost((3,4), (3,8)) + Cost((3,4), (4,7)) + Cost((7,4), (6,2)) + Cost((7,4), (6,4)) + Cost((7,4), (7,3)) + Cost((7,4), (8,5)) + Cost((7,4), (7,6))\}$

$$Total Cost = 3 + 4 + 4 + 2 + 3 + 1 + 1 + 2 = 20$$

## Solved Example 1 K-Mean clustering

### Step 3

- Randomly select one non-medoid point and recalculate the cost.
- $C1=(3, 4)$  and  $C2=(7, 4)$
- $O=(7, 3)$
- Swap  $C2$  with  $O$
- **New Medoids**
- $C1=(3, 4)$  and  $O=(7, 3)$

i	x	y	C1	C2	Cluster
X1	2	6			
X2	3	4			
X3	3	8			
X4	4	7			
X5	6	2			
X6	6	4			
X7	7	3			
X8	7	4			
X9	8	5			
X10	7	6			

## Solved Example 1 K-Mean clustering

- **New Medoids**
- $C1=(3, 4)$  and  $O=(7, 3)$
- *Manhattan Dist* =  $|x_1 - x_2| + |y_1 - y_2|$
- $Mdist[(2, 6), (7, 3)] = |2 - 7| + |6 - 3| = 8$

i	x	y	C1	O	Cluster
X1	2	6	3		
X2	3	4	0		
X3	3	8	4		
X4	4	7	4		
X5	6	2	5		
X6	6	4	3		
X7	7	3	5		
X8	7	4	4		
X9	8	5	6		
X10	7	6	6		



## Solved Example 1 K-Mean clustering

- **New Medoids**
- $C1=(3, 4)$  and  $O=(7, 3)$
- *Manhattan Dist* =  $|x_1 - x_2| + |y_1 - y_2|$
- $Mdist[(2, 6), (7, 3)] = |2 - 7| + |6 - 3| = 8$

i	x	y	C1	O	Cluster
X1	2	6	3	8	
X2	3	4	0	5	
X3	3	8	4	9	
X4	4	7	4	7	
X5	6	2	5	2	
X6	6	4	3	2	
X7	7	3	5	0	
X8	7	4	4	1	
X9	8	5	6	3	
X10	7	6	6	3	

## Solved Example 1 K-Mean clustering

- New Cluster are
- C1: {(2,6), (3,4), (3,8), (4,7)}
- O: {(6, 2), (6, 4), (7, 3), (7, 4), (8, 5), (7,6)}

i	x	y	C1	O	Cluster
X1	2	6	3	8	C1
X2	3	4	0	5	C1
X3	3	8	4	9	C1
X4	4	7	4	7	C1
X5	6	2	5	2	O
X6	6	4	3	2	O
X7	7	3	5	0	O
X8	7	4	4	1	O
X9	8	5	6	3	O
X10	7	6	6	3	O

## Solved Example 1 K-Mean clustering

- C1: {(2,6), **(3,4)**, (3,8), (4,7)}
- O: {(6, 2), (6, 4), **(7, 3)**, (7, 4), (8, 5), (7,6)}
- **Calculate the Total Cost**
- $Cost(c, x) = \sum_i |c_i - x_i|$
- $Current\ Total\ Cost = \{Cost((3,4), (2,6)) + Cost((3,4), (3,8)) + Cost((3,4), (4,7)) + Cost((7,3), (6,2)) + Cost((7,3), (6,4)) + Cost((7,3), (7,4)) + Cost((7,3), (8,5)) + Cost((7,3), (7,6))\}$
- $Current\ Total\ Cost = 3 + 4 + 4 + 2 + 2 + 1 + 3 + 3 = 22$

## Solved Example 1 K-Mean clustering

### Step 4

- Cost of Swapping of medoid C2 with O
- $S = \text{Current Total Cost} - \text{Previous Total Cost}$
- $S = 22 - 20 = 2 > 0$
- Hence Swapping C2 with O is not a good Idea.

## Solved Example 2 K-Mean clustering

- Suppose that the data mining task is to cluster points into three clusters,
- where the points are
- $A1(2, 10)$ ,  $A2(2, 5)$ ,  $A3(8, 4)$ ,  $B1(5, 8)$ ,  $B2(7, 5)$ ,  $B3(6, 4)$ ,  $C1(1, 2)$ ,  $C2(4, 9)$ .
- The distance function is Euclidean distance.
- Suppose initially we assign  $A1$ ,  $B1$ , and  $C1$  as the center of each cluster, respectively.

## Solved Example 2 K-Mean clustering

Initial Centroids:

A1: (2, 10)

B1: (5, 8)

C1: (1, 2)

Data Points			Distance to						Cluster	New Cluster
			$x_i$ 2	$y_i$ 10	5	8	1	2		
A1	2	10								
A2	2	5								
A3	8	4								
B1	5	8								
B2	7	5								
B3	6	4								
C1	1	2								
C2	4	9								

EP

$$d(p_1, p_2) = \sqrt{(x_2 - x_1)^2 + (y_2 - y_1)^2}$$

## Solved Example 2 K-Mean clustering

Initial Centroids:

A1: (2, 10)

B1: (5, 8)

C1: (1, 2)

Data Points			Distance to				Cluster	New Cluster
			2	10	5	8	1	2
A1	2	10	0.00					
A2	2	5	5.00					
A3	8	4	8.49					
B1	5	8	3.61					
B2	7	5	7.07					
B3	6	4	7.21					
C1	1	2	8.06					
C2	4	9	2.24					

$$d(p_1, p_2) = \sqrt{(x_2 - x_1)^2 + (y_2 - y_1)^2}$$

## Solved Example 2 K-Mean clustering

Initial Centroids:

A1: (2, 10)

B1: (5, 8)

C1: (1, 2)

Data Points			Distance to						Cluster	New Cluster
			2	10	5	8	1	2		
A1	2	10	0.00		3.61		8.06			
A2	2	5	5.00		4.24		3.16			
A3	8	4	8.49		5.00		7.28			
B1	5	8	3.61		0.00		7.21			
B2	7	5	7.07		3.61		6.71			
B3	6	4	7.21		4.12		5.39			
C1	1	2	8.06		7.21		0.00			
C2	4	9	2.24		1.41		7.62			

$$d(p_1, p_2) = \sqrt{(x_2 - x_1)^2 + (y_2 - y_1)^2}$$



## Solved Example 2 K-Mean clustering

Initial Centroids:

A1: (2, 10)

B1: (5, 8)

C1: (1, 2)

New Centroids:

A1: (2, 10)

B1: (6, 6)

C1: (1.5, 3.5)

Data Points			Distance to						Cluster	New Cluster
			2	10	5	8	1	2		
A1	2	10	0.00		3.61		8.06		1	
A2	2	5	5.00		4.24		3.16		3	
A3	8	4	8.49		5.00		7.28		2	
B1	5	8	3.61		0.00		7.21		2	
B2	7	5	7.07		3.61		6.71		2	
B3	6	4	7.21		4.12		5.39		2	
C1	1	2	8.06		7.21		0.00		3	
C2	4	9	2.24		1.41		7.62		2	

$$\frac{13}{5} = 2.6$$

## Solved Example 2 K-Mean clustering

Current Centroids:

A1: (2, 10)

B1: (6, 6)

C1: (1.5, 3.5)

Data Points			Distance to						Cluster	New Cluster
			2	10	6	6	1.5	3.5		
A1	2	10	0.00		5.66		6.52		1	1
A2	2	5	5.00		4.12		1.58		3	3
A3	8	4	8.49		2.83		6.52		2	2
B1	5	8	3.61		2.24		5.70		2	2
B2	7	5	7.07		1.41		5.70		2	2
B3	6	4	7.21		2.00		4.53		2	2
C1	1	2	8.06		6.40		1.58		3	3
C2	4	9	2.24		3.61		6.04		2	1

$$d(p_1, p_2) = \sqrt{(x_2 - x_1)^2 + (y_2 - y_1)^2}$$

## Solved Example 2 K-Mean clustering

Current Centroids:

A1: (3, 9.5)

B1: (6.5, 5.25)

C1: (1.5, 3.5)

New Centroids:

A1: (3.67, 9)

B1: (7, 4.33)

C1: (1.5, 3.5)

Data Points			Distance to				Cluster	New Cluster
			3	9.5	6.5	5.25	1.5	3.5
A1	2	10	1.12	6.54	6.52	1		
A2	2	5	4.61	4.51	1.58	3		
A3	8	4	7.43	1.95	6.52	2		
B1	5	8	2.50	3.13	5.70	1		
B2	7	5	6.02	0.56	5.70	2		
B3	6	4	6.26	1.35	4.53	2		
C1	1	2	7.76	6.39	1.58	3		
C2	4	9	1.12	4.51	6.04	1		

## Solved Example 2 K-Mean clustering

Current Centroids:

A1: (3.67, 9)

B1: (7, 4.33)

C1: (1.5, 3.5)

Data Points			Distance to					Cluster	New Cluster
			3.67	9	7	4.33	1.5		
A1	2	10	1.94	7.56	6.52	1	1		
A2	2	5	4.33	5.04	1.58	3	3		
A3	8	4	6.62	1.05	6.52	2	2		
B1	5	8	1.67	4.18	5.70	1	1		
B2	7	5	5.21	0.67	5.70	2	2		
B3	6	4	5.52	1.05	4.53	2	2		
C1	1	2	7.49	6.44	1.58	3	3		
C2	4	9	0.33	5.55	6.04	1	1		

$$d(p_1, p_2) = \sqrt{(x_2 - x_1)^2 + (y_2 - y_1)^2}$$