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ENGG 5103: Techniques for data mining

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I. K-MEANS CLUSTERING (25%)

Given the following points, present each iteration of k-means clustering algorithm until convergence. The distance is measured using Euclidean distance.

P1: (1,1.5), P2: (2,3), P3: (4,5), P4: (5,7), P5: (3.5,4.5), P6: (2.5,5.5)

(1) Suppose k=2, initial guess is (1,2) and (3,4).

Initial Center: (1, 2), (3, 4).

After one iteration, we have clustering centers (1.5, 2.25), (3.75, 5.5).

The first clustering contains points P1:(1,1.5), P2:(2,3).

The second clustering contains points P3: (4,5), P4: (5,7), P5: (3.5,4.5), P6: (2.5,5.5).

(2) Suppose k = 2, initial guess is (2,5) and (3,6).

Initial Centers: (2,5) and (3,6).

After one iteration, we have clustering centers (2.25, 3.625), (4.5, 6.0).

The first clustering contains points P1: (1, 1.5), P2: (2, 3), P5: (3.5, 4.5), P6: (2.5, 5.5)

The second Clustering contains points P3:(4,5), P4:(5,7)

(3) Suppose k = 3, initial guess is (1, 2), (3, 4) and (2, 3).

Initial Center: (1, 2), (3, 4) and (2, 3).

After one iteration, we have clustering centers (1.0, 1.5), (2.0, 3.0), (3.75, 5.5).

The first clustering contains point P1:(1,1.5).

The second clustering contains point P2:(2,3).

The third clustering contains points P3: (4,5), P4: (5,7), P5: (3.5,4.5), P6: (2.5,5.5).

Code is as follows: https://github.com/wangwanlitype1/ENGG5103

II. CLUSTER COHESION AND SEPARATION (25%)

Given the following points, answer the following questions.

P1: (1,1.5) P2: (2, 3) P3: (4, 5) P4: (5,7) P5: (3.5, 4.5) P6: (2.5, 5.5)

(1) Suppose they are divided into two clusters. P1, P2 and P3 belong to cluster 1, P4, P5 and P6 belong to cluster 2. Please calculate the SSE, SSB and TSS.

Answer:

The P1, P2 and P3 belong to cluster 1, we therefore have corresponding $c_1:(7/3,9.5/3)$.

The P4, P5 and P6 belong to cluster 2, we therefore have corresponding c_2 : (11/3, 17/3).

The SSE,SSB and TSS are defined as:

$$SSE = \sum_{i=1}^{K} \sum_{x \in C_i} (x - c_i)^2$$
 (1)

$$SSB = \sum_{i=1}^{K} |C_i| (c - c_i)^2$$
 (2)

$$TSS = \sum_{i=1}^{K} \sum_{x \in C_i} (x - c)^2$$
 (3)

We can calculate c as

$$c: (0.5*(7/3+11/3), 0.5*(9.5/3+17/3)) = (3,13.25/3)$$

$$\tag{4}$$

$$SSE = \begin{bmatrix} 1 - 7/3 \\ 1.5 - 9.5/3 \end{bmatrix}^{2} + \begin{bmatrix} 2 - 7/3 \\ 3 - 9.5/3 \end{bmatrix}^{2} + \begin{bmatrix} 4 - 7/3 \\ 5 - 9.5/3 \end{bmatrix}^{2} + \begin{bmatrix} 5 - 11/3 \\ 7 - 17/3 \end{bmatrix}^{2} + \begin{bmatrix} 3.5 - 11/3 \\ 4.5 - 17/3 \end{bmatrix}^{2} + \begin{bmatrix} 2.5 - 11/3 \\ 5.5 - 17/3 \end{bmatrix}^{2} = \begin{bmatrix} 47/6 \\ 28/3 \end{bmatrix}$$
(5)

$$SSB = 3 \begin{bmatrix} 3 - 7/3 \\ 13.25/3 - 9.5/3 \end{bmatrix}^2 + 3 \begin{bmatrix} 3 - 11/3 \\ 13.25/3 - 17/3 \end{bmatrix}^2 = \begin{bmatrix} 8/3 \\ 75/8 \end{bmatrix}$$
 (6)

$$TSS = \begin{bmatrix} 1-3\\1.5-13.25/3 \end{bmatrix}^2 + \begin{bmatrix} 2-3\\3-13.25/3 \end{bmatrix}^2 + \begin{bmatrix} 4-3\\5-13.25/3 \end{bmatrix}^2 + \begin{bmatrix} 5-3\\7-13.25/3 \end{bmatrix}^2 + \begin{bmatrix} 3.5-3\\4.5-13.25/3 \end{bmatrix}^2 + \begin{bmatrix} 2.5-3\\5.5-13.25/3 \end{bmatrix}^2 = \begin{bmatrix} 21/2\\449/24 \end{bmatrix}$$
(7)

(2) Suppose they are divided into three clusters. P1, P2 belongs to cluster 1, P3, P4 belongs to cluster 2, P5 and P6 belongs to cluster 3. Please calculate the SSE, SSB and TSS.

Answer:

The P1, P2 belong to cluster 1, we therefore have corresponding $c_1:(1.5,2.25)$.

The P3, P4 belong to cluster 2, we therefore have corresponding c_2 : (4.5,6).

The P5, P6 belong to cluster 3, we therefore have corresponding c_3 : (3,5).

We can calculate c as

$$c: (1/3*(1.5+4.5+3), 1/3*(2.25+6+5)) = (3, 13.25/3)$$
(8)

SSE =
$$\begin{bmatrix} 1 - 1.5 \\ 1.5 - 2.25 \end{bmatrix}^2 + \begin{bmatrix} 2 - 1.5 \\ 3 - 2.25 \end{bmatrix}^2 + \begin{bmatrix} 4 - 4.5 \\ 5 - 6 \end{bmatrix}^2 + \begin{bmatrix} 5 - 4.5 \\ 7 - 6 \end{bmatrix}^2 + \begin{bmatrix} 3.5 - 3 \\ 4.5 - 5 \end{bmatrix}^2 + \begin{bmatrix} 2.5 - 3 \\ 5.5 - 5 \end{bmatrix}^2 = \begin{bmatrix} 3/2 \\ 29/8 \end{bmatrix}$$
 (9)

$$SSB = 2 \begin{bmatrix} 3 - 1.5 \\ 13.25/3 - 2.25 \end{bmatrix}^{2} + 2 \begin{bmatrix} 3 - 4.5 \\ 13.25/3 - 6 \end{bmatrix}^{2} + 2 \begin{bmatrix} 3 - 3 \\ 13.25/3 - 5 \end{bmatrix}^{2} = \begin{bmatrix} 9 \\ 181/12 \end{bmatrix}$$
 (10)

$$TSS = \begin{bmatrix} 1-3\\1.5-13.25/3 \end{bmatrix}^2 + \begin{bmatrix} 2-3\\3-13.25/3 \end{bmatrix}^2 + \begin{bmatrix} 4-3\\5-13.25/3 \end{bmatrix}^2 + \begin{bmatrix} 5-3\\7-13.25/3 \end{bmatrix}^2 + \begin{bmatrix} 3.5-3\\4.5-13.25/3 \end{bmatrix}^2 + \begin{bmatrix} 2.5-3\\5.5-13.25/3 \end{bmatrix}^2 = \begin{bmatrix} 21/2\\449/24 \end{bmatrix}$$
(11)

(3) Show the relationship among SSE, SSB and TSS.

$$SSE + SSB = TSS \tag{12}$$

Equations (5) and (6) generates

$$SSE + SSB = \begin{bmatrix} 47/6 \\ 28/3 \end{bmatrix} + \begin{bmatrix} 8/3 \\ 75/8 \end{bmatrix} = TSS = \begin{bmatrix} 21/2 \\ 449/24 \end{bmatrix}$$
 (13)

which is equal to equation (7).

Similarly, equations (9) and (10) also generates

$$SSE + SSB = \begin{bmatrix} 3/2 \\ 29/8 \end{bmatrix} + \begin{bmatrix} 9 \\ 181/12 \end{bmatrix} = TSS = \begin{bmatrix} 21/2 \\ 449/24 \end{bmatrix}$$
 (14)

III. HIERARCHICAL CLUSTERING (25%)

Given the following points, answer the following questions.

P1: (1,1.5), P2: (2,3), P3: (4,5), P4: (5,7), P5: (3.5,4.5), P6: (2.5,5.5)

(1) Perform hierarchical clustering using complete linkage with agglomerative algorithm.

- 1.1) The point P5: (3.5, 4.5) is clustered into the point P3: (4,5), namely (P3: (4,5), P5: (3.5, 4.5)).
- 1.2) The point P6:(2.5,5.5) is clustered into (P3:(4,5),P5:(3.5,4.5)), namely (P3:(4,5),P5:(3.5,4.5),P6:(2.5,5.5)).
 - 1.3) The point P2:(2,3) is clustered into P1:(1,1.5), namely (P1:(1,1.5),P2:(2,3)).
- 1.4) The point P4:(5,7) is clustered into (P3:(4,5),P5:(3.5,4.5),P6:(2.5,5.5)), namely (P3:(4,5),P4:(5,7),P5:(3.5,4.5),P6:(2.5,5.5)).

As a result, the first clusters contain points P1:(1,1.5), P2:(2,3). The second clusters contain points P3:(4,5), P4:(5,7), P5:(3.5,4.5), P6:(2.5,5.5).

(2) Perform hierarchical clustering using single linkage with agglomerative algorithm.

- 2.1) The point P5:(3.5,4.5) is clustered into the point P3:(4,5), namely (P3:(4,5),P5:(3.5,4.5)).
- 2.2) The point P6:(2.5,5.5) is clustered into (P3:(4,5),P5:(3.5,4.5)), namely (P3:(4,5),P5:(3.5,4.5),P6:(2.5,5.5)).
 - 2.3) The point P2:(2,3) is clustered into P1:(1,1.5), namely (P1:(1,1.5),P2:(2,3)).
- 2.4) The points (P3:(4,5),P5:(3.5,4.5),P6:(2.5,5.5)) are clustered into (P1:(1,1.5),P2:(2,3)), namely (P1:(1,1.5),P2:(2,3),P3:(4,5),P5:(3.5,4.5),P6:(2.5,5.5)).

As a result, the first clusters contain points P1:(1,1.5), P2:(2,3), P3:(4,5), P5:(3.5,4.5), P6:(2.5,5.5). The second clusters contain points P4:(5,7).

Code is as follows: https://github.com/wangwanlitype1/ENGG5103

IV. SOM CLUSTERING (25%)

X = 1, 1.5, 2.5, 4.8, 2.8, 3.2, 3.8, 4.7, 3.3, 3.1

Y = 2, 2.3, 2.8, 4.5, 5.1, 6.3, 6.7, 6.8, 5.7, 3.9

Suppose that a = 0.2, a(neighbor) = 0.3, the size of neighbor is 5, perform SOM clustering and show the final results.

Answer:

Clustering simulation results for the first 5 iterations are as follows:

The first iteration is as follows:

Cluster1: correponding sample index[1, 2, 5, 6, 7, 8, 9]

Cluster2: correponding sample index[3, 4, 10]

The second iteration is as follows:

Cluster1: correponding sample index[1, 2, 5, 6, 7, 8, 9]

Cluster2: correponding sample index[3, 4, 10]

The third iteration is as follows:

Cluster1: correponding sample index[1, 5, 6, 7, 9]

Cluster2: correponding sample index[2, 8]

Cluster3: correponding sample index[3, 4, 10]

The forth iteration is as follows:

Cluster1: correponding sample index[1, 5, 6, 7, 9]

Cluster2: correponding sample index[2]

Cluster3: correponding sample index[3, 4, 10]

Cluster4: correponding sample index[8]

The fifth iteration is as follows:

Cluster1: correponding sample index[1, 5, 6, 7, 9]

Cluster2: correponding sample index[2]

Cluster3: correponding sample index[3, 4, 10]

Cluster4: correponding sample index[8]

It is necessary to notice that

a): I did not adopt fixed parameters like a=0.2, a(neighbor)=0.3 and the size of neighbor 5 since fixed parameters are not enough for us to do clustering. For example, it is better to set parameter of learning rate as a nonlinear function of the discrete time t and topology distance t. In general, the learning rate of the weight e(t,t) is set as

$$e(t,l) = \eta(t)e^{-l}. (15)$$

In the following code, the learning rate e(t, l) is set as

$$e(t,l) = \frac{e^{-l}}{t+2}. (16)$$

This means that learning rate decreases along with increasing discrete time t and topology distance l.

- b): The SOM clustering result may be not fixed and changed for each run since the weight of the SOM is initialized randomly.
 - c): The output of SOM clustering includes 2×2 neurons.

Code is as follows: https://github.com/wangwanlitype1/ENGG5103