

CSCI3230 / ESTR3108 2021-22 First Term Assignment 2

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University Guideline on Academic Honesty:

<http://www.cuhk.edu.hk/policy/academichonesty/>

Faculty of Engineering Guidelines to Academic Honesty:

http://www.erg.cuhk.edu.hk/erg-intra/upload/documents/ENGG_Discipline.pdf

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1a) The optimization problem is

$$\max_{\alpha} \left[\sum_{i=1}^8 \alpha_i - \frac{1}{2} \sum_{i=1}^8 \sum_{j=1}^8 \alpha_i \alpha_j y_i y_j x_i^T x_j \right]$$

$$s.t. \sum_{i=1}^8 \alpha_i y_i = 0, \alpha_i \geq 0$$

By using scikit-learn library, the optimal α is

$$\alpha_1 = 0.03172284, \alpha_2 = 0, \alpha_3 = 0.1360743, \alpha_4 = 0,$$

$$\alpha_5 = 0, \alpha_6 = 0, \alpha_7 = 0, \alpha_8 = -0.16779714$$

$$\text{where } \alpha_1 + \alpha_2 + \alpha_3 + \alpha_4 + \alpha_5 + \alpha_6 + \alpha_7 + \alpha_8 = 0$$

(The library calculated $\alpha_i \leftarrow y_i \alpha_i$,

α_8 should be 0.16779714 if following $\alpha_i \geq 0$)

The support vector is $\left\{ \begin{bmatrix} 6 \\ -8 \end{bmatrix}, \begin{bmatrix} 9 \\ -8.5 \end{bmatrix}, \begin{bmatrix} 9 \\ -5 \end{bmatrix} \right\}$

$$\mathbf{w} \text{ is } \begin{bmatrix} -0.09516853 \\ -0.57142857 \end{bmatrix}$$

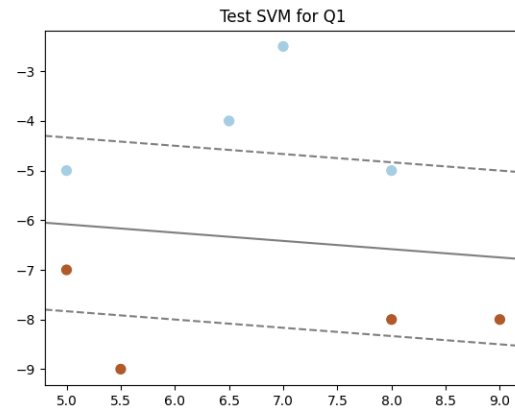
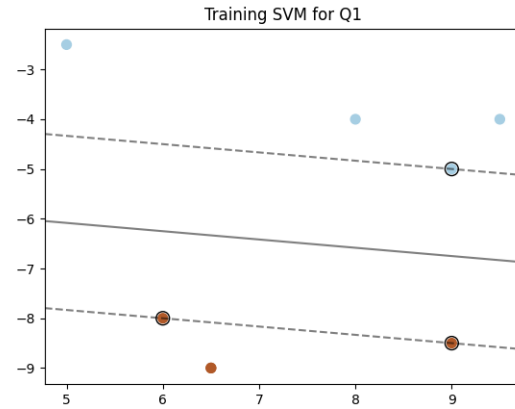
$$b = \frac{1}{|S|} \sum_{s \in S} \left(\frac{1}{y_s} - w^T x_s \right) = -3.000556512$$

We can get the hyperplane:

$$\begin{bmatrix} -0.09516853 \\ -0.57142857 \end{bmatrix}^T x - 3.000556512 = 0$$

b) For all test case predicted result y_i indexed by i :

$$\begin{cases} y_i = -1 & \text{if } \begin{bmatrix} -0.09516853 \\ -0.57142857 \end{bmatrix}^T x_i - 3.000556512 < 0 \\ y_i = 1 & \text{if } \begin{bmatrix} -0.09516853 \\ -0.57142857 \end{bmatrix}^T x_i - 3.000556512 > 0 \end{cases}$$



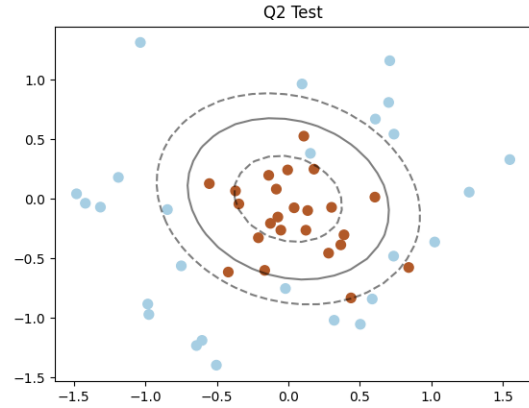
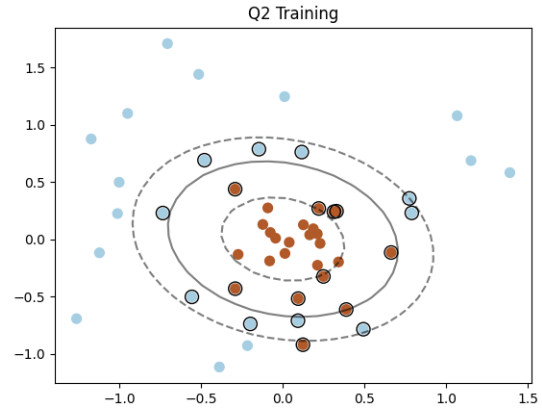
The prediction obtained is:

$$y_1 = 1, y_2 = 1, y_3 = 1, y_4 = 1, y_5 = -1, y_6 = -1, y_7 = -1, y_8 = -1$$

c) Removing the 2nd data point in training set will **NOT** change the optimal hyperplane, but removing the 8th one **WILL** change it. Because 2nd one is **NOT** the support vectors, but 8th one **IS**.

2) Polynomial kernel is used to be the kernel function since the positive cases are centralized like a circle and the negative cases are surrounding the positive result. The degree chosen for the polynomial kernel is 2 but not higher value is to prevent overfitting. (Accuracy of the model is 92%)

See attached file *CSCI3230Asg2Q2.py* for coding details.



3a) By calculating the distance between the initial cluster points and the data given, the data points will be put into the closest cluster centroid by calculating

$$\sqrt{(data_x - centroid_x)^2 + (data_y - centroid_y)^2}$$

So from the Assign Point Cluster Figure,

Cluster center (5, 4): a, b, c, d, e, f, g, h, i, j, k, l

Cluster center (12, 6): m, n, o, p, q, r, s, t, u, v, w, x

New centroid for (5, 4)

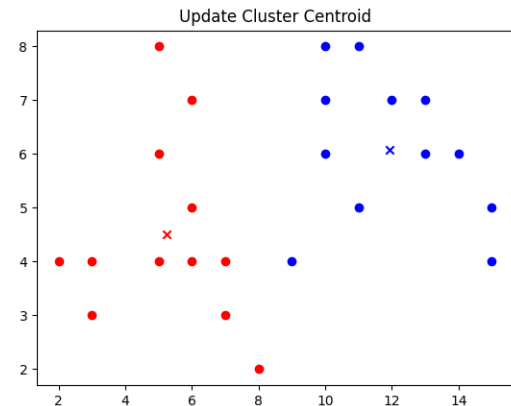
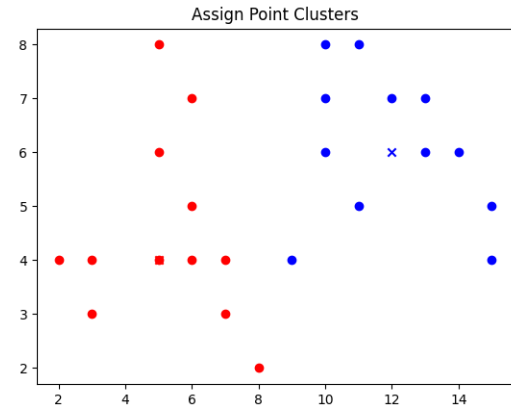
$$= \left(\frac{2+3+3+5+5+5+6+6+6+7+7+8}{12}, \frac{4+4+3+8+6+4+7+5+4+4+3+2}{12} \right)$$

$$= (5.25, 4.5)$$

New centroid for (12, 6)

$$= \left(\frac{10+10+10+11+11+12+13+13+14+15+15+9}{12}, \frac{8+7+6+8+5+7+7+6+6+5+4+4}{12} \right)$$

$$= (11.916667, 6.083333)$$



b) Consider the following table, all the point that having neighbourhood relationship will be marked as -1,

	a	b	c	d	e	f	g	h	i	j	k	l	m	n	o	p	q	r	s	t	u	v	w	x
a	-1	-1	-1																					
b	-1	-1	-1			-1																		
c	-1	-1	-1																					
d				-1	-1		-1																	
e				-1	-1	-1	-1	-1																
f		-1			-1	-1		-1	-1	-1														
g				-1	-1		-1	-1																
h					-1	-1	-1	-1	-1	-1														
i						-1		-1	-1	-1	-1													
j						-1		-1	-1	-1	-1													-1
k									-1	-1	-1	-1												
l											-1	-1												
m													-1	-1	-1	-1								
n													-1	-1	-1	-1		-1						
o													-1	-1	-1		-1							
p													-1	-1		-1		-1						
q															-1		-1							
r														-1		-1		-1	-1	-1				
s																		-1	-1	-1	-1			
t																		-1	-1	-1	-1			
u																			-1	-1	-1	-1		
v																					-1	-1	-1	
w																						-1	-1	
x										-1														-1

By scanning each rows, if Sum of that row ≤ -3 , it will be a core point. Therefore,
Core points: a, b, c, d, e, f, g, h, i, j, k, m, n, o, p, r, s, t, u, v

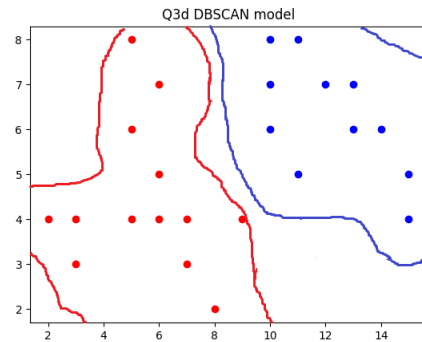
c) Yes, the intermediate points on the chain is $u \leftarrow t \leftarrow r \leftarrow n \leftarrow o$.

d) The right hand side shown the points processed by DBSCAN model, which:

Red color represents the Cluster 1: a, b, c, d, e, f, g, h, i, j, k, l, x;

Blue color represents the Cluster 2: m, n, o, p, q, r, s, t, u, v, w

Reminded that there is **NO noise points**.



4) For **MIN distance**,

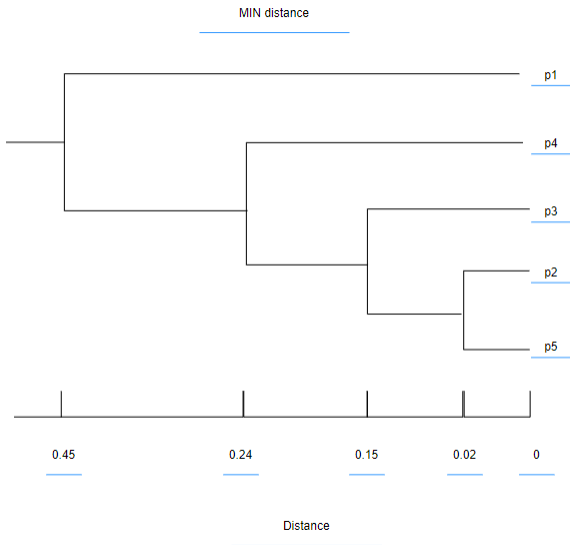
$$M_{min} = \begin{matrix} & p1 & p2 & p3 & p4 & p5 \\ \begin{matrix} p1 \\ p2 \\ p3 \\ p4 \\ p5 \end{matrix} & \begin{pmatrix} 0 & - & - & - & - \\ 0.9 & 0 & - & - & - \\ 0.59 & 0.36 & 0 & - & - \\ 0.45 & 0.53 & 0.56 & 0 & - \\ 0.65 & 0.02 & 0.15 & 0.24 & 0 \end{pmatrix} \end{matrix} \Rightarrow \text{pick } p2, p5 = 0.02 \text{ because they are minimum}$$

$$\Rightarrow \begin{matrix} & p25 & p1 & p3 & p4 \\ \begin{matrix} p25 \\ p1 \\ p3 \\ p4 \end{matrix} & \begin{pmatrix} 0 & - & - & - \\ 0.45 & 0 & - & - \\ 0.15 & 0.59 & 0 & - \\ 0.24 & 0.45 & 0.56 & 0 \end{pmatrix} \end{matrix} \Rightarrow \text{pick } p25, p3 = 0.15 \text{ because they are minimum}$$

$$\Rightarrow \begin{matrix} & p253 & p1 & p4 \\ \begin{matrix} p253 \\ p1 \\ p4 \end{matrix} & \begin{pmatrix} 0 & - & - \\ 0.45 & 0 & - \\ 0.24 & 0.59 & 0 \end{pmatrix} \end{matrix} \Rightarrow \text{pick } p253, p4 = 0.24 \text{ because they are minimum}$$

$$\Rightarrow \begin{matrix} & p2534 & p1 \\ \begin{matrix} p2534 \\ p1 \end{matrix} & \begin{pmatrix} 0 & - \\ 0.45 & 0 \end{pmatrix} \end{matrix} \Rightarrow \text{pick } p2534, p1 = 0.45 \text{ because they are minimum}$$

The dendrogram is shown below:



For **MAX distance**,

$$M_{max} = \begin{matrix} & p1 & p2 & p3 & p4 & p5 \\ \begin{matrix} p1 \\ p2 \\ p3 \\ p4 \\ p5 \end{matrix} & \begin{pmatrix} 0 & - & - & - & - \\ 0.9 & 0 & - & - & - \\ 0.59 & 0.36 & 0 & - & - \\ 0.45 & 0.53 & 0.56 & 0 & - \\ 0.65 & 0.02 & 0.15 & 0.24 & 0 \end{pmatrix} \end{matrix} \Rightarrow \text{pick } p2, p5 = 0.02 \text{ because they are minimum}$$

$$\Rightarrow \begin{matrix} & p25 & p1 & p3 & p4 \\ \begin{matrix} p25 \\ p1 \\ p3 \\ p4 \end{matrix} & \begin{pmatrix} 0 & - & - & - \\ 0.9 & 0 & - & - \\ 0.36 & 0.59 & 0 & - \\ 0.53 & 0.45 & 0.56 & 0 \end{pmatrix} \end{matrix} \Rightarrow \text{pick } p25, p3 = 0.36 \text{ because they are minimum}$$

$$\Rightarrow \begin{matrix} & p253 & p1 & p4 \\ \begin{matrix} p253 \\ p1 \\ p4 \end{matrix} & \begin{pmatrix} 0 & - & - \\ 0.9 & 0 & - \\ 0.56 & 0.45 & 0 \end{pmatrix} \end{matrix} \Rightarrow \text{pick } p1, p4 = 0.45 \text{ because they are minimum}$$

$$\Rightarrow \begin{matrix} & p253 & p14 \\ \begin{matrix} p253 \\ p14 \end{matrix} & \begin{pmatrix} 0 & - \\ 0.9 & 0 \end{pmatrix} \end{matrix} \Rightarrow \text{pick } p253, p14 = 0.9 \text{ because they are minimum}$$

The dendrogram is shown below:

