

Homework-2

Q1(a) Data, we will classify C1 as -1 and C2 as 1, thus we have dataset
 $D = \{(4.8, -10.4)\} \rightarrow -1$

$$D = \frac{1}{2}((4.8, -10.4) \rightarrow -1)$$

$$1 \leq ((12.8, 5.6) \rightarrow p)^{-1}$$

$$(3.2, -7.2) \rightarrow -1$$

$$1 \leq (7, 2, 4) \rightarrow (1)$$

$$1 \leq (0.4, -10) \rightarrow -1$$

18 (9.6, 8.8) \rightarrow 14

$$(1.8, -2.2) \rightarrow$$

(6, 1, 34) \Rightarrow 7

(6.0, 2.0) : 1

→ Plot of $\log \text{Rate of disappearance}$ vs t is attached in plot page - 1 (i.e.) (n + 1) &

Optimization Function $\rightarrow \min_{w,b} \frac{1}{2} \|w\|^2$

$$F_1 = \left(d + \frac{1}{2} \pi - F_{\text{min}}(d, \theta) \right) \cdot \frac{\sin(\theta)}{\sin(\theta) + \cos(\theta)} \quad 2$$

Constraints: $y^{(i)} \cdot (\mathbf{w}^T \mathbf{x}^{(i)} + b) \geq 1$

$$\{x = \cos^2(\theta) + i\sin^2(\theta), \quad \theta \in [0, \pi]\} = \{e^{i\theta}\}$$

For this problem, this translates to:

Optimization function: $\min_{w,b} \frac{1}{2} (w_1^2 + w_2^2)$

Constraints = $\omega_1(x) = 7$

and the other side of the world.

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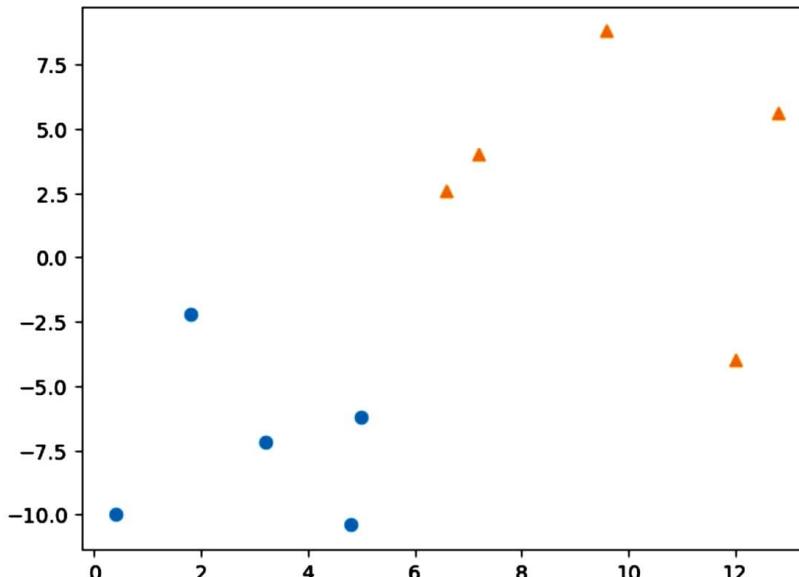
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Python 3 (ipykernel) O



PLOT 1

```
In [26]: 1 X = [4.8,3.2,0.4,1.8,5]
2 Y = [-10.4,-7.2,-10,-2.2,-6.2]
3 plt.scatter(X,Y, marker="o")
4 X = [12.8,7.2,9.6,6.6,12]
5 Y = [5.6,4,8.8,2.6,-4]
6 plt.scatter(X,Y,marker = "^")
7 plt.show()
```



$$\text{constraints} - (4.8\omega_1 - 10.4\omega_2) \geq 1$$

$$(12.8\omega_1 + 5.6\omega_2) \geq 1$$

$$-(3.2\omega_1 - 7.2\omega_2) \geq 1$$

$$(7.2\omega_1 + 4\omega_2) \geq 1$$

$$-(0.4\omega_1 - 10\omega_2) \geq 1$$

$$(9.6\omega_1 + 8.8\omega_2) \geq 1$$

$$-(1.8\omega_1 - 2.2\omega_2) \geq 1$$

$$(6.6\omega_1 + 2.6\omega_2) \geq 1$$

$$-(5\omega_1 - 6.2\omega_2) \geq 1$$

$$(12\omega_1 - 4\omega_2) \geq 1$$

→ The corresponding Lagrangian in general is

$$L(\omega, b, \alpha) = \frac{1}{2} \|\omega\|^2 - \sum_{i=1}^n \alpha_i [y^{(i)}(\omega^T x^{(i)} + b) - 1]$$

For this problem it translates into

$$L(\omega, b, \alpha) = \frac{1}{2} (\omega_1^2 + \omega_2^2) - \alpha_1 [-(4.8\omega_1 - 10.4\omega_2 + b) - 1]$$

$$-\alpha_2 [(12.8\omega_1 + 5.6\omega_2 + b) - 1]$$

$$-\alpha_3 [-(3.2\omega_1 - 7.2\omega_2 + b) - 1]$$

$$-\alpha_4 [(7.2\omega_1 + 4\omega_2 + b) - 1]$$

$$-\alpha_5 [-(0.4\omega_1 - 10\omega_2 + b) - 1]$$

$$-\alpha_6 [(9.6\omega_1 + 8.8\omega_2 + b) - 1]$$

$$-\alpha_7 [-(1.8\omega_1 - 2.2\omega_2 + b) - 1]$$

$$-\alpha_8 [(6.6\omega_1 + 2.6\omega_2 + b) - 1]$$

$$-\alpha_9 [-(5\omega_1 - 6.2\omega_2 + b) - 1]$$

$$-\alpha_{10} [(12\omega_1 - 4\omega_2 + b) - 1]$$

The Lagrangian Dual problem is defined as

Optimization Function

$$\max_{\alpha} \sum_{i=1}^n \alpha_i - \frac{1}{2} \sum_{i,j=1}^n y^{(i)} y^{(j)} \alpha_i \alpha_j \{x^{(i)}, x^{(j)}\}$$

constraint

$$\alpha_i \geq 0$$

$$\sum_{i=1}^n \alpha_i y^{(i)} = 0$$

For the given problem it translates to

Optimization Function

$$\begin{aligned} \max_{\alpha} & \sum_{i=1}^8 \alpha_i - \frac{1}{2} (\alpha_1 (131.2\alpha_1 - 3.2\alpha_2 + 90.24\alpha_3 \\ & + 7.04\alpha_4 + 105.92\alpha_5 + 45.44\alpha_6 + 31.52\alpha_7 - 4.64\alpha_8 \\ & + 88.48\alpha_9 - 99.12\alpha_{10}) \\ & + \alpha_2 (-3.2\alpha_1 + 195.2\alpha_2 - 0.64\alpha_3 + 114.56\alpha_4 + 50.88\alpha_5 + 172.16\alpha_6 \\ & - 10.72\alpha_7 + 99.04\alpha_8 - 29.28\alpha_9 + 131.2\alpha_{10}) \\ & + \alpha_3 (90.24\alpha_1 - 0.64\alpha_2 + 62.08\alpha_3 + 5.76\alpha_4 + 73.28\alpha_5 + 32.64\alpha_6 + 21.6\alpha_7 \\ & - 2.39\alpha_8 + 60.64\alpha_9 - 67.2\alpha_{10}) \\ & + \alpha_4 (7.04\alpha_1 + 114.56\alpha_2 + 5.76\alpha_3 + 67.84\alpha_4 + 37.12\alpha_5 + 104.32\alpha_6 - 4.16\alpha_7 + 57.92\alpha_8 \\ & - 11.2\alpha_9 + 70.4\alpha_{10}) + \alpha_5 (105.92\alpha_1 + 50.88\alpha_2 + 73.28\alpha_3 + 37.12\alpha_4 + 100.16\alpha_5 + \\ & + 84.16\alpha_6 + 22.72\alpha_7 + 23.36\alpha_8 + 64.0\alpha_9 - 44.8\alpha_{10}) + \alpha_6 (45.44\alpha_1 + 172.16\alpha_2 + \\ & 32.64\alpha_3 + 104.32\alpha_4 + 84.16\alpha_5 + 169.6\alpha_6 + 2.08\alpha_7 + 86.24\alpha_8 + 6.56\alpha_9 \\ & + 80\alpha_{10}) + \alpha_7 (31.52\alpha_1 - 10.72\alpha_2 + 21.6\alpha_3 - 4.16\alpha_4 + 22.72\alpha_5 + 2.08\alpha_6 + \\ & 8.08\alpha_7 - 6.16\alpha_8 + 22.64\alpha_9 - 30.40\alpha_{10}) + \alpha_8 (-4.64\alpha_1 + 99.04\alpha_2 - 2.4\alpha_3 + \\ & 57.92\alpha_4 + 23.36\alpha_5 + 86.24\alpha_6 - 6.16\alpha_7 + 50.32\alpha_8 - 16.88\alpha_9 + 68.8\alpha_{10}) \end{aligned}$$

$$\text{Contd} \quad + \alpha_9 (88.48\alpha_1 - 29.28\alpha_2 + 60.64\alpha_3 - 11.2\alpha_4 \\ + 64\alpha_5 + 6.56\alpha_6 + 22.64\alpha_7 - 16.88\alpha_8 + \\ 63.44\alpha_9 - 84.8\alpha_{10})$$

$$+ \alpha_{10} (-99.2\alpha_1 + 131.2\alpha_2 - 67.2\alpha_3 + 70.4\alpha_4 \\ - 44.8\alpha_5 + 80\alpha_6 - 30.4\alpha_7 + 68.8\alpha_8 + \\ - 84.8\alpha_9 + 160\alpha_{10}))$$

Constraints

$$\alpha_i \geq 0 \quad \text{where } i = 1 \dots 10$$

$$-\alpha_1 + \alpha_2 - \alpha_3 + \alpha_4 - \alpha_5 + \alpha_6 - \alpha_7 + \alpha_8 - \alpha_9 + \alpha_{10} = 0$$

(b) Now from this we can perform SMO using a heuristic to pick pairs of α variables and optimizing for them

→ So now we have the least distance b/w the two classification points and they are α_7, α_8

→ For first iteration we will use these two ~~constraint~~ α values

Iteration - 1

→ For iteration first consider the following equations

$$w = \sum_{i=1}^{\infty} \alpha_i y^{(i)} x^{(i)} = 0$$

assuming all the α_i are zero

and the offset b for the primal:

$$b = \max_{i=1}^n y^{(i)} - \mathbf{w}^T \mathbf{x}^{(i)} + \min_{i=1}^n y^{(i)} - \mathbf{w}^T \mathbf{x}^{(i)} = 0$$

Iteration 1 (α_8, α_7)

$$\sum \alpha_i y^{(i)} = 0$$

$i \neq 8, 7$

$$\alpha_8 = \frac{(81 + 8\alpha_7) y^{(7)}}{y^{(7)}} = \alpha_7$$

→ We can rewrite the unclipped optimization

$$\alpha_7^{\text{new, u}} = \operatorname{argmax}_{\alpha_7} (\alpha_7 + \alpha_8)$$

$$\therefore \alpha_7^{\text{new, u}} = \operatorname{argmax}_{\alpha_7} \alpha_7 + \alpha_8 - \frac{1}{2} (8.08\alpha_7 - 6.16\alpha_8)$$

Putting the equation $\alpha_7 = \alpha_8$

$$\begin{aligned} \alpha_7 &= \operatorname{argmax}_{\alpha_7} 2\alpha_7 - \frac{1}{2} [1.92\alpha_7^2 + 44.16\alpha_7] \\ &= \operatorname{argmax}_{\alpha_7} 2\alpha_7 - \frac{1}{2} [46.08\alpha_7^2] \end{aligned}$$

To optimize this we take derivative and set it to 0

$$\therefore \frac{\partial}{\partial \alpha_7} (2\alpha_7 - \frac{1}{2} [46.08\alpha_7^2]) = 0$$

$$\therefore 2 - 46.08\alpha_7 = 0$$

$$\therefore \alpha_7 = 2/46.08 = 0.0434$$

thus we can say
 newclipped $\alpha_7 = 0.0434$; newclipped $\alpha_8 = 0.0434$

Now we can resolve

$$w = \sum_{i=1}^m \alpha_i y^{(i)} x^{(i)} = (0.0434)(-1)(1.8, -2.2)^T + (0.0434)(1)(6.6, 2.6)^T$$

$$\therefore w = \begin{bmatrix} -0.07812 \\ 0.109548 \end{bmatrix} + \begin{bmatrix} 0.28644 \\ 0.11284 \end{bmatrix}$$

$$= \begin{bmatrix} 0.20832 \\ 0.20832 \end{bmatrix}$$

$$b = -0.083328 + 1.667$$

$$b = 0.791$$

The plot for this graph is attached
 as plot-2

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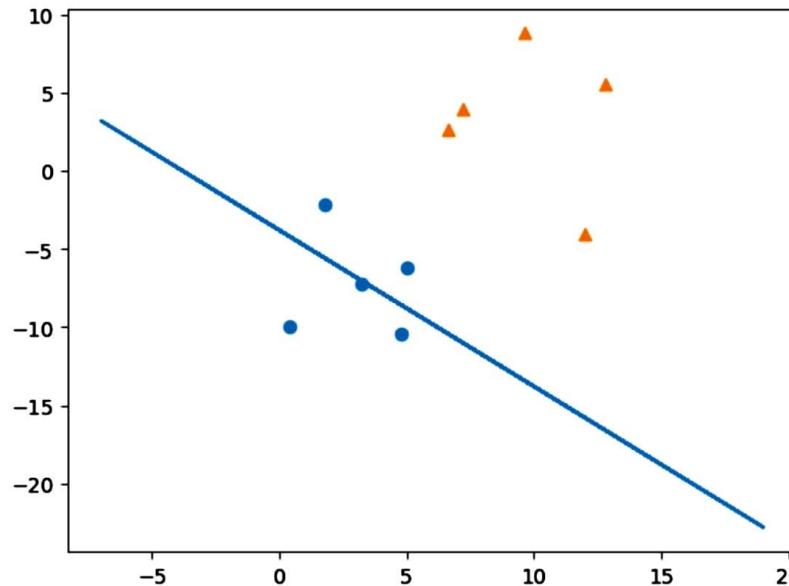
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Python 3 (ipykernel) O



PLOT 2

```
In [27]:  
1 x = [4.8,3.2,0.4,1.8,5]  
2 y = [-10.4,-7.2,-10,-2.2,-6.2]  
3 plt.scatter(x,y, marker="o")  
4 x = [12.8,7.2,9.6,6.6,12]  
5 y = [5.6,4,8.8,2.6,-4]  
6 plt.scatter(x,y,marker = "^")  
7 x = np.random.randint(low=-10, high=20, size=20)  
8 y = -x-3.797  
9 plt.plot(x,y)  
10 plt.show()  
11
```



Iteration 2: α_7 & α_9 are now defying the margin hence we again iterate

$$y = -\sum \alpha_i y^{(i)} = 0$$

α_9, α_7

∴ Rewriting α_7 in terms of α_9

$$\therefore \alpha_7 + \alpha_9 - \alpha_8 = 0$$

$$\therefore \alpha_7 = \alpha_8 - \alpha_9$$

$$= 0.20832 - \alpha_9$$

$$\therefore \alpha_9^{\text{new, unclipped}} = \operatorname{argmax}_{\alpha_9} 0.20832 - \alpha_9 + \alpha_9$$

$$- \frac{1}{2} (\alpha_7 (8.08\alpha_7 - 6.16\alpha_8 + 22.64\alpha_9))$$

$$+ \alpha_8 (-6.16\alpha_7 + 50.32\alpha_8 - 16.88\alpha_9)$$

$$+ \alpha_9 (-30.4\alpha_7 + 68.8\alpha_8 - 84.8\alpha_9)$$

$$\therefore \alpha_9 = \operatorname{argmax}_{\alpha_9} = 0.20832$$

$$- \frac{1}{2} [\alpha_7 (8.08\alpha_7 - 6.16 \times 0.20832) \alpha_9 + 0.20832 (-6.16\alpha_7 + 50.32 \times 0.20832 - 16.88\alpha_9)]$$

$$+ \alpha_9 (-30.4\alpha_7 + 68.8 \times 0.20832 - 84.8\alpha_9)$$

$$= 0.20832$$

$$= 0.20832 - \frac{1}{2} [(0.20832 - \alpha_9)(8.082(0.20832 - \alpha_9))]$$

$$- 0.20832 - \frac{1}{2} \left(-68.96\alpha_9^2 + 8.397\alpha_9 + 2 \right)$$

Now taking derivative for optimization

$$\therefore \frac{\partial}{\partial \alpha_9} \left(0.20832 - \frac{1}{2} (-68.96\alpha_9^2 + 8.397\alpha_9 + 2) \right)$$

$$= 68.96 \alpha_9 - 8.397 = 0$$

$$\therefore \alpha_9 = \frac{8.397}{68.96} = 0.121$$

$$\therefore \alpha_7^{\text{new}} = 0.20832 - 0.121 \\ = 0.0865$$

$$\therefore w = \sum_{i=1}^n \alpha_i y_i^{(i)} x_i^{(i)}$$

$$= (-1) (0.0865) \begin{bmatrix} 1.8 \\ -2.2 \end{bmatrix}$$

$$+ (1) (0.20832) \begin{bmatrix} 6.6 \\ 2.6 \end{bmatrix}$$

$$+ (-1) (0.121) \begin{bmatrix} 5 \\ -6.2 \end{bmatrix}$$

$$\begin{bmatrix} -0.1557 \\ 0.1903 \end{bmatrix} + \begin{bmatrix} \cancel{1.374} \\ 0.541 \end{bmatrix} + \begin{bmatrix} -0.605 \\ 0.7502 \end{bmatrix}$$

$$= \begin{bmatrix} 0.6138 \\ 1.4815 \end{bmatrix}$$

$$b = \frac{-2.15536 + 1.4335}{2} = -0.36093$$

* The plot is attached as plot-3

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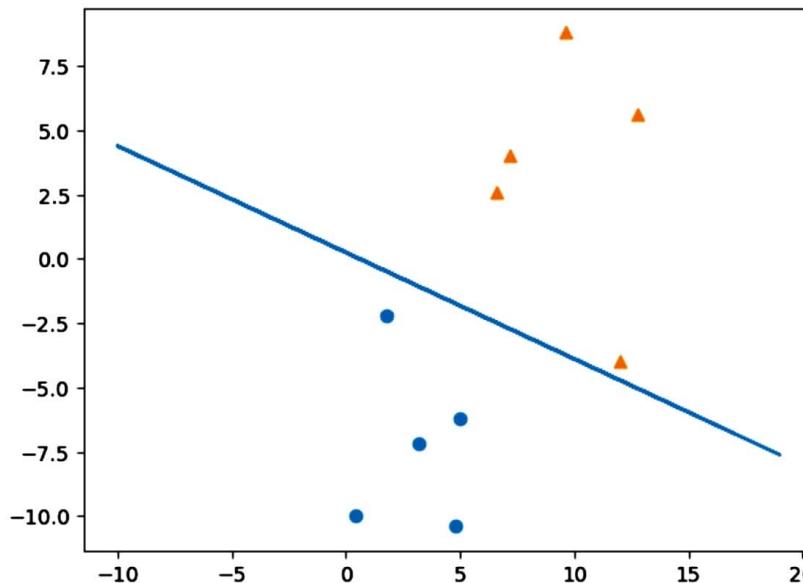
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Python 3 (ipykernel) O



PLOT 3

```
In [30]: 1 X = [4.8,3.2,0.4,1.8,5]
2 Y = [-10.4,-7.2,-10,-2.2,-6.2]
3 plt.scatter(X,Y, marker="o")
4 X = [12.8,7.2,9.6,6.6,12]
5 Y = [5.6,4,8.8,2.6,-4]
6 plt.scatter(X,Y,marker = "^")
7 x = np.random.randint(low=-10, high=20, size=40)
8 y = -0.4139*x+0.2544
9 plt.plot(x,y)
10 plt.show()
11
```



Iteration 3 - Now there are no points which are being violated thus now I pick α for point $(12, -4)$

and $\alpha_1 = (4.8, -10.4)$

$\alpha_{10} =$

$$\rightarrow \therefore -\sum \alpha_i y^{(i)} = 0$$

$$\therefore \cancel{\alpha_1 + \alpha_{10}} \quad \alpha_1 - \alpha_{10} + 0.0865 = -0.20832 \\ + 0.121 = 0$$

$$\therefore \alpha_1 - \alpha_{10} = 0.00082$$

$$\therefore \alpha_1 = 0.00082 + \alpha_{10}$$

$\alpha_{10}^{\text{new}} =$

$$\therefore \underset{\alpha_{10}}{\text{argmax}} \left[\alpha_1 + \alpha_{10} - \frac{1}{2} \left[\alpha_1 (131.2\alpha_1 - 99.2\alpha_{10}) \right] \right]$$

$$= 2\alpha_{10} + 0.00082 - \frac{1}{2} \left[0.00082 + \alpha_{10}(131.2(0.00082) - 99.2\alpha_{10}) + \alpha_{10}(-99.2(0.00082) + 160\alpha_{10}) \right]$$

$$= 2\alpha_{10} + 0.00082 - \frac{1}{2} \left[92.8\alpha_{10}^2 + 0.0449\alpha_{10} \right]$$

for optimization

$$\frac{d}{d\alpha_{10}} (2\alpha_{10} + 0.00082 - \frac{1}{2} [92.8\alpha_{10}^2 + 0.0449\alpha_{10}]) = 0$$

$$\therefore 2 - 92.8\alpha_{10} - \frac{1}{2} \times 0.0449 = 0$$

$$\therefore 2 - 92.8\alpha_{10} - 0.02245 = 0$$

$$\therefore 92.8\alpha_{10} = 1.97755$$

$$\therefore \alpha_{10} = 0.0213$$

$$\therefore \alpha_1^{\text{newclipped}} = 0.0222$$

$$\therefore w = \sum_{i=1}^n d_i y^{(i)} x^{(i)}$$

$$= \begin{bmatrix} 0.6133 \\ 1.4815 \end{bmatrix} + (1)(0.0222) \begin{bmatrix} 4.8 \\ -10.4 \end{bmatrix} + (1)(0.0213) \begin{bmatrix} 12 \\ -4 \end{bmatrix}$$

$$= \begin{bmatrix} 0.6133 - 0.10656 + 0.2556 \\ 1.4815 + 0.23088 - 0.0852 \end{bmatrix}$$

$$= \begin{bmatrix} 0.76234 \\ 1.62718 \end{bmatrix}$$

$$b = \frac{-2.207584 + 2.63936}{2}$$

$$= 0.21588$$

The plot for this iteration is in plot-34 attached.

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Python 3 (ipykernel) O



1 # PLOT 4

```
In [31]: 1 X = [4.8,3.2,0.4,1.8,5]
2 Y = [-10.4,-7.2,-10,-2.2,-6.2]
3 plt.scatter(X,Y, marker="o")
4 X = [12.8,7.2,9.6,6.6,12]
5 Y = [5.6,4,8.8,2.6,-4]
6 plt.scatter(X,Y,marker = "^")
7 x = np.random.randint(low=-10, high=20, size=40)
8 y = -0.4685*x-0.1326
9 plt.plot(x,y)
10 plt.show()
11
```

