

For 1 and 11 see the picture after.

2.  $x_1, x_3$  and  $x_4$  are support vectors.

3.

$$\bar{\alpha} = [\alpha_1 \ \alpha_2 \ \alpha_3 \ \alpha_4]^T$$

$$L(\bar{w}, b, \alpha_j) = \frac{1}{2} \|\bar{w}\|^2 - \sum_j \alpha_j [(\bar{w} \cdot x_j + b)y_j - 1]$$

4. Fix  $\alpha_j$

$$\begin{aligned} \nabla_{\bar{w}} L(\bar{w}, b, \alpha_j) &= \bar{w} - \sum_j \alpha_j y_j x_j = 0 \rightarrow \bar{w} = \sum_j \alpha_j y_j x_j \\ \nabla_b L(\bar{w}, b, \alpha_j) &= -\sum_j \alpha_j y_j = 0 \rightarrow \sum_j \alpha_j y_j = 0 \end{aligned} \quad \left. \vphantom{\begin{aligned} \nabla_{\bar{w}} L(\bar{w}, b, \alpha_j) \\ \nabla_b L(\bar{w}, b, \alpha_j) \end{aligned}} \right\} \text{KKT conditions}$$

5. Plug  $\bar{w} = \sum_j \alpha_j y_j x_j$  back to  $L(\bar{w}, b, \alpha_j)$

$$\begin{aligned} L(\bar{w}, b, \alpha_j) &= \frac{1}{2} (\sum_j \alpha_j y_j x_j)^T (\sum_j \alpha_j y_j x_j) - \sum_j (\alpha_j y_j \bar{w} \cdot x_j + \alpha_j b y_j - \alpha_j) \\ &= \frac{1}{2} (\sum_j \alpha_j y_j x_j)^T (\sum_j \alpha_j y_j x_j) - \sum_j \alpha_j y_j \bar{w} \cdot x_j \underbrace{[-b \sum_j \alpha_j y_j]}_{\rightarrow \text{zero}} + \sum_j \alpha_j \\ &= \frac{1}{2} (\sum_j \alpha_j y_j x_j)^T (\sum_j \alpha_j y_j x_j) - \sum_j \alpha_j y_j (\sum_j \alpha_j y_j x_j) \cdot x_j + \sum_j \alpha_j \\ &= \frac{1}{2} (\sum_j \alpha_j y_j x_j)^T (\sum_j \alpha_j y_j x_j) - \sum_j \alpha_j y_j (\sum_j \alpha_j y_j x_j)^T x_j + \sum_j \alpha_j \\ &= (\sum_j \alpha_j y_j x_j)^T \left[ \frac{1}{2} (\sum_j \alpha_j y_j x_j) - \sum_j \alpha_j y_j x_j \right] + \sum_j \alpha_j \\ &= \sum_j \alpha_j - \frac{1}{2} (\sum_j \alpha_j y_j x_j)^T (\sum_j \alpha_j y_j x_j) \\ &\text{maximize}_{\alpha_j} \sum_j \alpha_j - \frac{1}{2} (\sum_j \alpha_j y_j x_j)^T (\sum_j \alpha_j y_j x_j) \end{aligned}$$

6. We know that the first point  $x_2$  is not the support vector and should be removed from the constraint

$$L(\bar{w}, b, \alpha_j) = \frac{1}{2} \|\bar{w}\|^2 - \sum_{j \neq 1} \alpha_j [(\bar{w} \cdot x_j + b)y_j - 1]$$

7.

$$\sum_{j \neq 1} \alpha_j y_j = 0 \quad \text{complementary slackness}$$

8.

$$Y = [y_1 \ y_2 \ y_3 \ y_4]^T$$

$$X = [x_1 \ x_2 \ x_3 \ x_4]^T$$

$$\sum_j \alpha_j y_j x_j = \alpha_1 y_1 x_1 + \alpha_2 y_2 x_2 + \alpha_3 y_3 x_3 + \alpha_4 y_4 x_4$$

$$= \alpha_1 x_1 - \alpha_2 x_2 - \alpha_3 x_3 - \alpha_4 x_4$$

$$= [\alpha_1 \ \alpha_2 \ \alpha_3 \ \alpha_4] [x_1 \ -x_2 \ -x_3 \ -x_4]^T$$

Therefore

$$\begin{aligned}
(\sum_j \alpha_j y_j x_j)^T (\sum_j \alpha_j y_j x_j) &= [x_1 - x_2 - x_3 - x_4] [\alpha_1 \alpha_2 \alpha_3 \alpha_4]^T [\alpha_1 \alpha_2 \alpha_3 \alpha_4] [x_1 - x_2 - x_3 - x_4]^T \\
&= [\alpha_1 \alpha_2 \alpha_3 \alpha_4] [x_1 - x_2 - x_3 - x_4]^T [x_1 - x_2 - x_3 - x_4] [\alpha_1 \alpha_2 \alpha_3 \alpha_4]^T \\
&= [\alpha_1 \alpha_2 \alpha_3 \alpha_4] \begin{bmatrix} 1 & 1 & -1 & 0 \\ 1 & 1 & 0 & -1 \end{bmatrix}^T \begin{bmatrix} 1 & 1 & -1 & 0 \\ 1 & 1 & 0 & -1 \end{bmatrix} [\alpha_1 \alpha_2 \alpha_3 \alpha_4]^T \\
&= A^T H A
\end{aligned}$$

Where

$$\begin{aligned}
A &= [\alpha_1 \alpha_2 \alpha_3 \alpha_4]^T \\
H &= \begin{bmatrix} 1 & 1 & -1 & 0 \\ 1 & 1 & 0 & -1 \end{bmatrix}^T \begin{bmatrix} 1 & 1 & -1 & 0 \\ 1 & 1 & 0 & -1 \end{bmatrix} = \begin{bmatrix} 2 & 2 & -1 & -1 \\ 2 & 2 & -1 & -1 \\ -1 & -1 & 1 & 0 \\ -1 & -1 & 0 & 1 \end{bmatrix}
\end{aligned}$$

Rewrite the original optimization problem into quadratic programming (Q.P.) form:

$$\begin{aligned}
&\text{minimize}_A \quad \frac{1}{2} A^T H A - 1^T A \\
&\text{subject to} \quad A^T Y = 0 \\
&\quad \quad \quad A \geq 0
\end{aligned}$$

By solving the above optimization problem, it gives

$$A = [4 \ 0 \ 2 \ 2]^T$$

Which means

$$\alpha_1 = 4, \alpha_2 = 0, \alpha_3 = \alpha_4 = 2$$

9.

$$\bar{w} = \sum_j \alpha_j y_j x_j = 4[1 \ 1]^T + 0 - 2[1 \ 0]^T - 2[0 \ 1]^T = [2 \ 2]^T$$

10.

$$(\bar{w} \cdot x_j + b) y_j - 1 = 0$$

Select a support vector, say  $x_1$ , plug it into the above equation:

$$([2 \ 2][1 \ 1]^T + b)(1) = 1 \rightarrow b = -3$$

Therefore:

$$\bar{w} \cdot x + b = 0$$

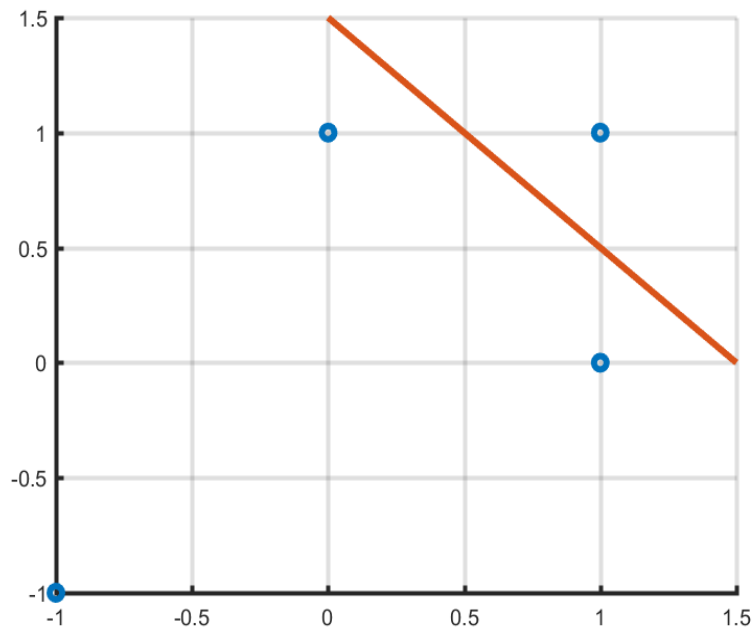
$$[2 \ 2]x - 3 = 0 \rightarrow -[1 \ 1]x + 1.5 = 0$$

Therefore:

$$\bar{w}^* = [-1 \ -1]^T$$

$$b^* = 1.5$$

Which is the same as the initial guess in q1.



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close all
clear all

x1 = [1;1];
x2 = [-1;-1];
x3 = [1;0];
x4 = [0;1];
x = [x1 x2 x3 x4];
j = 4;

Y(:, :, 4) = [1;-1;-1;-1];
m = 0:0.1:1.5;
k = - m + 1.5;

figure(1)
p1 = scatter(x(1,:), x(2,:));
set(p1, 'linewidth', 3)
hold on
p2 = plot(m,k);
set(p2, 'linewidth', 3)
grid on
set(gca, 'linewidth', 2)

% support point: x1, x3 and x4

X = transpose(x);

% Formulating Q.P.
H = [2 2 -1 -1; 2 2 -1 -1; -1 -1 1 0; -1 -1 0 1];
f = [-1; -1; -1; -1];
lb = zeros(4,1);
Aeq = [1 -1 -1 -1];
beq = 0;

[x,fval,exitflag,output,lambda] = ...
    quadprog(H,f,[],[],Aeq, beq, lb, []);

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