For 1 and 11 see the picture after.

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

3.

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

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

4. Fix  $\alpha_i$ 

$$\nabla_{\overline{w}}L(\overline{w},b,\alpha_{j}) = \overline{w} - \Sigma_{j}\alpha_{j}y_{j}x_{j} = 0 \to \overline{w} = \Sigma_{j}\alpha_{j}y_{j}x_{j}$$

$$\nabla_{b}L(\overline{w},b,\alpha_{j}) = -\Sigma_{j}\alpha_{j}by_{j} = 0 \to \Sigma_{j}\alpha_{j}y_{j} = 0$$

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

$$L(\overline{w}, b, \alpha_{j}) = \frac{1}{2} (\Sigma_{j} \alpha_{j} y_{j} x_{j})^{T} (\Sigma_{j} \alpha_{j} y_{j} x_{j}) - \Sigma_{j} (\alpha_{j} y_{j} \overline{w} \cdot x_{j} + \alpha_{j} b y_{j} - \alpha_{j})$$

$$= \frac{1}{2} (\Sigma_{j} \alpha_{j} y_{j} x_{j})^{T} (\Sigma_{j} \alpha_{j} y_{j} x_{j}) - \Sigma_{j} \alpha_{j} y_{j} \overline{w} \cdot x_{j} - b \Sigma_{j} \alpha_{j} y_{j} + \Sigma_{j} \alpha_{j}$$

$$= \frac{1}{2} (\Sigma_{j} \alpha_{j} y_{j} x_{j})^{T} (\Sigma_{j} \alpha_{j} y_{j} x_{j}) - \Sigma_{j} \alpha_{j} y_{j} (\Sigma_{j} \alpha_{j} y_{j} x_{j}) \cdot x_{j} + \Sigma_{j} \alpha_{j}$$

$$= \frac{1}{2} (\Sigma_{j} \alpha_{j} y_{j} x_{j})^{T} (\Sigma_{j} \alpha_{j} y_{j} x_{j}) - \Sigma_{j} \alpha_{j} y_{j} (\Sigma_{j} \alpha_{j} y_{j} x_{j})^{T} x_{j} + \Sigma_{j} \alpha_{j}$$

$$= (\Sigma_{j} \alpha_{j} y_{j} x_{j})^{T} [\frac{1}{2} (\Sigma_{j} \alpha_{j} y_{j} x_{j}) - \Sigma_{j} \alpha_{j} y_{j} x_{j}] + \Sigma_{j} \alpha_{j}$$

$$= \Sigma_{j} \alpha_{j} - \frac{1}{2} (\Sigma_{j} \alpha_{j} y_{j} x_{j})^{T} (\Sigma_{j} \alpha_{j} y_{j} x_{j})$$

$$maximize_{\alpha_{j}} \Sigma_{j} \alpha_{j} - \frac{1}{2} (\Sigma_{j} \alpha_{j} y_{j} x_{j})^{T} (\Sigma_{j} \alpha_{j} y_{j} x_{j})$$

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

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

7.

$$\Sigma_{j\neq j^*}\alpha_j y_j = 0$$

Should have been more specific.

8.

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

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

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

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

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

Therefore

$$\begin{split} \left(\Sigma_{j}\alpha_{j}y_{j}x_{j}\right)^{T}\left(\Sigma_{j}\alpha_{j}y_{j}x_{j}\right) &= [x_{1}-x_{2}-x_{3}-x_{4}][\alpha_{1}\ \alpha_{2}\ a_{3}\ \alpha_{4}]^{T}[\alpha_{1}\ \alpha_{2}\ a_{3}\ \alpha_{4}][x_{1}-x_{2}-x_{3}-x_{4}]^{T} \\ &= [\alpha_{1}\ \alpha_{2}\ a_{3}\ \alpha_{4}][x_{1}-x_{2}-x_{3}-x_{4}]^{T}[x_{1}-x_{2}-x_{3}-x_{4}][\alpha_{1}\ \alpha_{2}\ a_{3}\ \alpha_{4}]^{T} \\ &= [\alpha_{1}\ \alpha_{2}\ a_{3}\ \alpha_{4}]\begin{bmatrix} 1 & 1 & -1 & 0 \\ 1 & 1 & 0 & -1 \end{bmatrix}^{T}\begin{bmatrix} 1 & 1 & -1 & 0 \\ 1 & 1 & 0 & -1 \end{bmatrix}[\alpha_{1}\ \alpha_{2}\ a_{3}\ \alpha_{4}]^{T} \\ &= A^{T}HA \end{split}$$

Where

$$A = \begin{bmatrix} \alpha_1 & \alpha_2 & \alpha_3 & \alpha_4 \end{bmatrix}^T$$

$$H = \begin{bmatrix} 1 & 1 & -1 & 0 \\ 1 & 1 & 0 & -1 \end{bmatrix}^T \begin{bmatrix} 1 & 1 & -1 & 0 \\ 1 & 1 & 0 & -1 \end{bmatrix} = \begin{bmatrix} 2 & 2 & -1 & -1 \\ 2 & 2 & -1 & -1 \\ -1 & -1 & 1 & 0 \\ -1 & -1 & 0 & 1 \end{bmatrix}$$

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

$$minimize_A \frac{1}{2}A^THA - 1^TA$$

$$subject \ to \ A^TY = 0$$

$$A \ge 0$$

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.

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

10.

$$(\overline{w} \cdot x_i + b)y_i - 1 = 0$$

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:

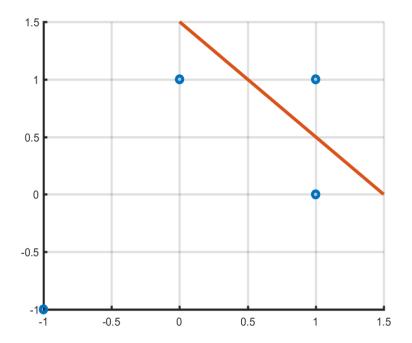
$$\overline{w} \cdot x + b = 0$$
  
[2 2] $x - 3 = 0 \rightarrow -[1 \ 1]x + 1.5 = 0$ 

Therefore:

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

$$b^* = 1.5$$

Which is the same as the initial guess in q1.



```
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, []);
```