Solution Set: Support Vector Classifier

 $h_3 = -\varepsilon_3$ $h_4 = -\varepsilon_4.$

- 1. Our convex optimization problem takes the form: $\underset{(\beta_0,\beta,\varepsilon)\in\mathbb{R}^7}{\text{minimize}} \quad f(\beta_0,\beta_1,\beta_2,\varepsilon_1,\varepsilon_2,\varepsilon_3,\varepsilon_4)$ given the constraint $g_i(\beta_0, \beta, \varepsilon) \leq 0$ for i = 1, 2, 3, 4and $h_i(\beta_0, \beta, \varepsilon) \le 0$ for i = 1, 2, 3, 4where $(\beta_0, \beta, \varepsilon) = \frac{1}{2} \|\beta\|^2 + C \sum_{i=1}^4 \varepsilon_i$, $g_i(\beta_0, \beta, \varepsilon) = 1 - \varepsilon_i - y_i(\beta_0 + \beta_1 x_{i1} + \beta_2 x_{i2})$ for = 1, 2, 3, 4, and $h_i(\beta_0, \beta, \varepsilon) = -\varepsilon_i$ for i = 1, 2, 3, 4 $g_1 = 1 - \varepsilon_1 + (\beta_0)$ So $g_2 = 1 - \varepsilon_2 + (\beta_0 + \beta_2)$ $g_3 = 1 - \varepsilon_3 - (\beta_0 - \beta_1)$ $g_4 = 1 - \varepsilon_4 - (\beta_0 + \beta_1)$ $h_1 = -\varepsilon_1$ $h_2 = -\varepsilon_2$

The dual Lagrangian is given by $L_D(\alpha) = \sum_{i=1}^4 \alpha_i - \frac{1}{2} \sum_{i=1}^4 \sum_{j=1}^4 \alpha_i \alpha_j y_i y_j x_i^T x_j$.

So $L_D(x, \alpha) = (\alpha_1 + \alpha_2 + \alpha_3 + \alpha_4) - \frac{1}{2} [\alpha_2^2 + \alpha_3^2 + \alpha_4^2 - 2\alpha_3\alpha_4]$

We want to maximize $L_D(\alpha)$ subject to the constraints $0 \le \alpha_i \le C \forall i$ and $\alpha_1 y_1 + \alpha_2 y_2 + \alpha_3 y_3 + \alpha_2 y_2 + \alpha_3 y_3 + \alpha_3 y_$ $\alpha_4 y_4 = 0$. That is, we need $0 \le \alpha_i \le C \forall i$ and $-\alpha_1 - \alpha_2 + \alpha_3 + \alpha_4 = 0$. These constraints give us a four-dimensional plane in the positive box $0 \le \alpha_i \le C \forall i$.

Let $H = \{(\alpha_1, \dots, \alpha_4) \in \mathbb{R}^4 | 0 \le \alpha_i \le C \forall i \text{ and } -\alpha_1 - \alpha_2 + \alpha_3 + \alpha_4 = 0\}$. We want to maximize $L_D(\alpha_1, \ldots, \alpha_4)$ on H.

To find the maximum of $L_D(\alpha_1, ..., \alpha_4)$ on H, we can use any computational software.

It turns out that, for C = 2, the maximum value of L_D on H is 6 and it occurs at $(\alpha_1, \dots, \alpha_4) = (2, 2, 2, 2)$.

$$\beta = \sum_{i=1}^{4} \alpha_i y_i x_i$$
$$\implies \beta = \begin{bmatrix} 0\\ -2 \end{bmatrix}.$$

By complementary slackness, we have $\alpha_i (1 - \varepsilon_i - y_i (\beta_0 + \beta_1 x_{i1} + \beta_2 x_{i2})) = 0 \forall i = 1, ..., 4$

This gives us a system of 4 equations and 5 unknowns ε_1 , ε_2 , ε_3 , ε_4 , β_0 . Solving this system gives $\varepsilon_1 = 2$, $\varepsilon_2 = \varepsilon_3 = \varepsilon_4 = 0$, and $\beta_0 = 1$.

The equation of our hyperplane is given by $\beta_0 + \beta_1 X_1 + \beta_2 X_2 = 0$.

So we get $1 - 2X_2 = 0$

$$\implies X_2 = \frac{1}{2}$$

Since $\alpha_i > 0$ for i = 1, ..., 4, we have that each x_i satisfies $y_i(\beta_0 + \beta_1 x_{i1} + \beta_2 x_{i2}) = 1 - \varepsilon_i$. Hence, x_1, x_2, x_3, x_4 are all support vectors.

b) For C = 4, L_D has an absolute max value of 10 and it occurs at $(\alpha_1, ..., \alpha_4) = (4, 2, 3, 3)$.

 $\beta = \begin{bmatrix} 0 \\ -2 \end{bmatrix}$, $\beta_0 = 1$, $\varepsilon_1 = 2$, $\varepsilon_2 = \varepsilon_3 = \varepsilon_4 = 0$, and the hyperplane is $X_2 = \frac{1}{2}$, the same result we got for C = 2.

c) For C = 1, L_D has an absolute max value of $\frac{7}{2}$ and it occurs at $(\alpha_1, \dots, \alpha_4) = (1, 1, 1, 1)$.

 $\beta = \begin{bmatrix} 0 \\ -1 \end{bmatrix}$. The complementary slackness equations $\alpha_i (1 - \varepsilon_i - y_i (\beta_0 + \beta_1 x_{i1} + \beta_2 x_{i2})) = 0 \forall i = 1, ..., 4$

give us a system of 4 equations and 5 unknowns $\varepsilon_1, \varepsilon_2, \varepsilon_3, \varepsilon_4, \beta_0$. This system has more than one solution. One solution is $\beta_0 = 0, \varepsilon_1 = 1, \varepsilon_2 = 0, \varepsilon_3 = 1, \varepsilon_4 = 1$. The hyperplane is $X_2 = 0$. Another solution is $\beta_0 = 1, \varepsilon_1 = 2, \varepsilon_2 = 1, \varepsilon_3 = 0, \varepsilon_4 = 0$. The hyperplane is $X_2 = 1$.

2. Our convex optimization problem takes the form:

 $\begin{array}{l} \underset{(\beta_0,\beta,\varepsilon)\in\mathbb{R}^8}{\text{minimize}} \quad f(\beta_0,\beta_1,\beta_2,\varepsilon_1,\varepsilon_2,\varepsilon_3,\varepsilon_4,\varepsilon_5) & \text{given the constraint} \\ g_i(\beta_0,\beta,\varepsilon) \leq 0 \text{ for } i = 1,2,3,4,5 \\ \text{and } h_i(\beta_0,\beta,\varepsilon) \leq 0 \text{ for } i = 1,2,3,4,5 \\ \text{where } (\beta_0,\beta,\varepsilon) = \frac{1}{2} ||\beta||^2 + C \sum_{i=1}^5 \varepsilon_i , \\ g_i(\beta_0,\beta,\varepsilon) = 1 - \varepsilon_i - y_i(\beta_0 + \beta_1 x_{i1} + \beta_2 x_{i2}) \text{ for } = 1,2,3,4,5, \\ \text{and } h_i(\beta_0,\beta,\varepsilon) = -\varepsilon_i \text{ for } i = 1,2,3,4,5 \end{array}$ So $\begin{array}{c} g_1 = 1 - \varepsilon_1 - (\beta_0 + \beta_2) \\ g_2 = 1 - \varepsilon_2 - (\beta_0 - \beta_2) \\ g_3 = 1 - \varepsilon_3 + (\beta_0) \\ g_4 = 1 - \varepsilon_4 + (\beta_0 + \beta_1 + \beta_2) \\ g_5 = 1 - \varepsilon_5 + (\beta_0 + \beta_1 - \beta_2) \end{array}$

$$g_{1} = 1 - \varepsilon_{1} - (\beta_{0} + \beta_{2})$$

$$g_{2} = 1 - \varepsilon_{2} - (\beta_{0} - \beta_{2})$$

$$g_{3} = 1 - \varepsilon_{3} + (\beta_{0})$$

$$g_{4} = 1 - \varepsilon_{4} + (\beta_{0} + \beta_{1} + \beta_{2})$$

$$g_{5} = 1 - \varepsilon_{5} + (\beta_{0} + \beta_{1} - \beta_{2})$$

$$h_{1} = -\varepsilon_{1}$$

$$h_{2} = -\varepsilon_{2}$$

$$h_{3} = -\varepsilon_{3}$$

$$h_{4} = -\varepsilon_{4}$$

$$h_{5} = -\varepsilon_{5}$$

The dual Lagrangian is given by $L_D(\alpha) = \sum_{i=1}^5 \alpha_i - \frac{1}{2} \sum_{i=1}^5 \sum_{j=1}^5 \alpha_i \alpha_j y_i y_j x_i^T x_j$.

So $L_D(x, \alpha) = (\alpha_1 + \alpha_2 + \alpha_3 + \alpha_4 + \alpha_5) - \frac{1}{2} [\alpha_1^2 + \alpha_2^2 + 2\alpha_4^2 + 2\alpha_5^2 - 2\alpha_1\alpha_2 - 2\alpha_1\alpha_4 + 2\alpha_1\alpha_5 + 2\alpha_2\alpha_4 - 2\alpha_2\alpha_5]$

We want to maximize $L_D(\alpha)$ subject to the constraints $0 \le \alpha_i \le C \forall i$ and $\alpha_1 y_1 + \alpha_2 y_2 + \alpha_3 y_3 + \alpha_4 y_4 + \alpha_5 y_5 = 0$. That is, we need $0 \le \alpha_i \le C \forall i$ and $\alpha_1 + \alpha_2 - \alpha_3 - \alpha_4 - \alpha_5 = 0$. These constraints give us a five-dimensional plane in the positive box $0 \le \alpha_i \le C \forall i$.

Let $H = \{(\alpha_1, ..., \alpha_4, \alpha_5) \in \mathbb{R}^5 | 0 \le \alpha_i \le C \forall i \text{ and } \alpha_1 + \alpha_2 - \alpha_3 - \alpha_4 - \alpha_5 = 0\}$. We want to maximize $L_D(\alpha_1, ..., \alpha_4, \alpha_5)$ on H.

To find the maximum of $L_D(\alpha_1, ..., \alpha_4, \alpha_5)$ on *H*, we can use any computational software.

It turns out that, for C = 2, the maximum value of L_D on H is 6 and it occurs at $(\alpha_1, ..., \alpha_4, \alpha_5) = (2, 2, 2, 1, 1)$.

$$\beta = \sum_{i=1}^{5} \alpha_i y_i x_i$$
$$\implies \qquad \beta = \begin{bmatrix} -2\\ 0 \end{bmatrix}.$$

By complementary slackness, we have $\alpha_i (1 - \varepsilon_i - y_i (\beta_0 + \beta_1 x_{i1} + \beta_2 x_{i2})) = 0 \forall i = 1, ..., 5$ and $\mu_i \varepsilon_i = 0 \forall i$.

Since $\alpha_4, \alpha_5 \neq C$ and $\alpha_i = C - \mu_i \forall i, \mu_4, \mu_5 \neq 0$. Thus, $\varepsilon_4 = \varepsilon_5 = 0$ since $\mu_i \varepsilon_i = 0 \forall i$.

Using $\alpha_i (1 - \varepsilon_i - y_i(\beta_0 + \beta_1 x_{i1} + \beta_2 x_{i2})) = 0 \forall i$, we can solve for β_0 and the remaining ε 's. We get $\beta_0 = 1, \varepsilon_1 = 0, \varepsilon_2 = 0, \varepsilon_3 = 2, \varepsilon_4 = 0, \varepsilon_5 = 0$.

The equation of our hyperplane is given by $\beta_0 + \beta_1 X_1 + \beta_2 X_2 = 0$.

So we get $1 - 2X_1 = 0$

$$\implies X_1 = \frac{1}{2}$$

Since $\alpha_i > 0$ for i = 1, ..., 5, we have that each x_i satisfies $y_i(\beta_0 + \beta_1 x_{i1} + \beta_2 x_{i2}) = 1 - \varepsilon_i$. Hence, x_1, x_2, x_3, x_4, x_5 are all support vectors.

b) For C = 4, L_D has an absolute max value of 10 and it occurs at $(\alpha_1, \dots, \alpha_5) = (4, 2, 4, 2, 0)$.

 $\beta = \begin{bmatrix} -2 \\ 0 \end{bmatrix}$, $\beta_0 = 1$, $\varepsilon_1 = 0$, $\varepsilon_2 = 0$, $\varepsilon_3 = 2$, $\varepsilon_4 = 0$, $\varepsilon_5 = 0$, and the hyperplane is $X_1 = \frac{1}{2}$, the same result we got for C = 2.

c) For C = 1, L_D has an absolute max value of $\frac{7}{2}$ and it occurs at $(\alpha_1, ..., \alpha_5) = (1, 1, 1, \frac{1}{2}, \frac{1}{2})$.

$$\beta = \begin{bmatrix} -1 \\ 0 \end{bmatrix}$$
, $\beta_0 = 0$, $\varepsilon_1 = 1$, $\varepsilon_2 = 1$, $\varepsilon_3 = 1$, $\varepsilon_4 = 0$, $\varepsilon_5 = 0$, and the hyperplane is $X_1 = 0$