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// Problem #3 //

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(a)

Because of linear separability, there exists a hyperplane $\bar{\mathbf{v}}^T \bar{\mathbf{x}} + b$ such that

$$[\min(\bar{\mathbf{v}}^T \bar{\mathbf{x}} + b) \text{ when } y = 1] \geq 0 > [\max(\bar{\mathbf{v}}^T \bar{\mathbf{x}} + b) \text{ when } y = -1]$$

Let \bar{x}_p be the positive data point closest to the hyperplane and \bar{x}_n be the negative data point closest to the hyperplane. Because of linear separability, $p^+ \geq 0 > p^-$. These two points are

$$p^+ = [\min(\bar{\mathbf{v}}^T \bar{\mathbf{x}} + b) \text{ when } y = 1] = \bar{\mathbf{v}}^T \bar{x}_p + b$$

$$p^- = [\max(\bar{\mathbf{v}}^T \bar{\mathbf{x}} + b) \text{ when } y = -1] = \bar{\mathbf{v}}^T \bar{x}_n + b$$

We can shift the hyperplane by a value c , so that the new hyperplane is equidistant from \bar{x}_p and \bar{x}_n . We can find the value of c .

$$\frac{|\bar{\mathbf{v}}^T \bar{x}_p + b - c|}{\|\bar{\mathbf{v}}\|} = \frac{|\bar{\mathbf{v}}^T \bar{x}_n + b - c|}{\|\bar{\mathbf{v}}\|}$$

$$\bar{\mathbf{v}}^T \bar{x}_p + b - c = -(\bar{\mathbf{v}}^T \bar{x}_n + b - c)$$

$$p^+ - c = -p^- + c$$

$$c = \frac{p^+ + p^-}{2}$$

The new hyperplane $\bar{\mathbf{v}}^T \bar{\mathbf{x}} + b - c = 0$ separates the data set D .

$$[\min(\bar{\mathbf{v}}^T \bar{\mathbf{x}} + b - c) \text{ when } y = 1] = \frac{p^+ - p^-}{2}$$

$$[\max(\bar{\mathbf{v}}^T \bar{\mathbf{x}} + b - c) \text{ when } y = -1] = \frac{p^- - p^+}{2}$$

$$\text{Hence, } y(\bar{\mathbf{v}}^T \bar{\mathbf{x}} + b - c) \geq \frac{p^+ - p^-}{2}$$

Since $p^+ > p^-$ and $c = \frac{p^+ + p^-}{2}$. We can set $\bar{\mathbf{w}} = \frac{\bar{\mathbf{v}}}{c}$ and $\theta = \frac{b-c}{c}$ and $\delta = 0$.

Now plugging in $\delta = 0$, we can see that

$$\bar{\mathbf{w}}^T \bar{\mathbf{x}} + \theta \geq 1 \geq 0 \text{ when } y = 1, \text{ and}$$

$$\bar{\mathbf{w}}^T \bar{\mathbf{x}} + \theta \leq -1 < 0 \text{ when } y = -1$$

(b)

When $\delta > 0$, we can split this into a few cases:

When $1 > \delta > 0$, the data set is linearly separable because of the proof in part (a).

When $\delta \geq 1$, the data set can be separable or inseparable.

When $\min \delta \geq 1$, the data set is not separable.

(c)

To satisfy the LP formulation given, the optimal solution would be when $\bar{\mathbf{w}} = 0$, $\theta = 0$, and $\delta = 0$

However, this does not result in a hyperplane, so we cannot use this formulation.