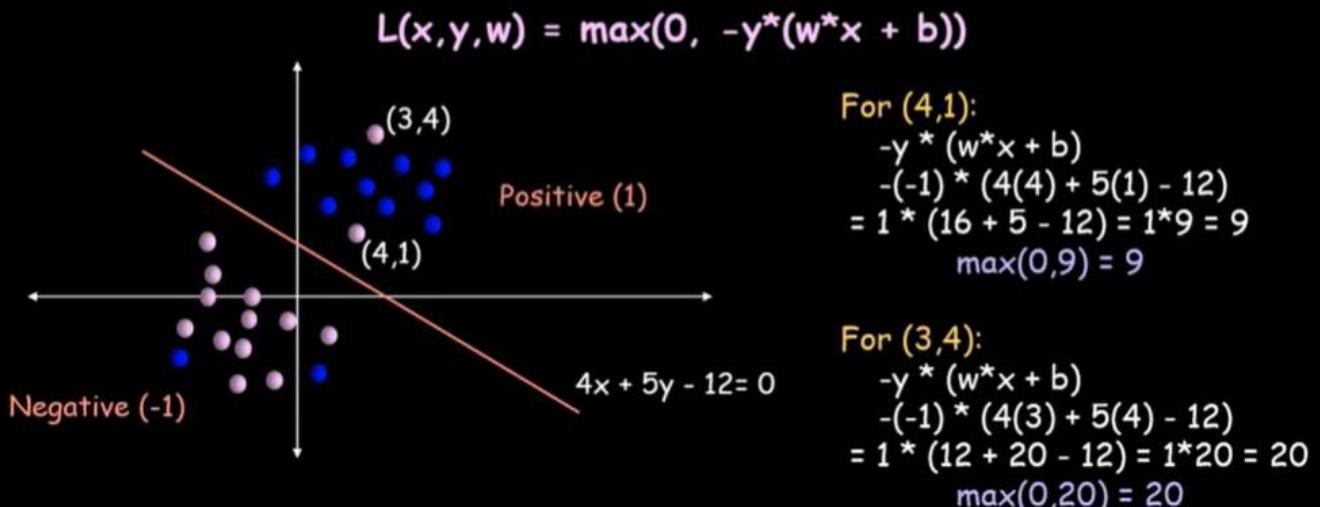
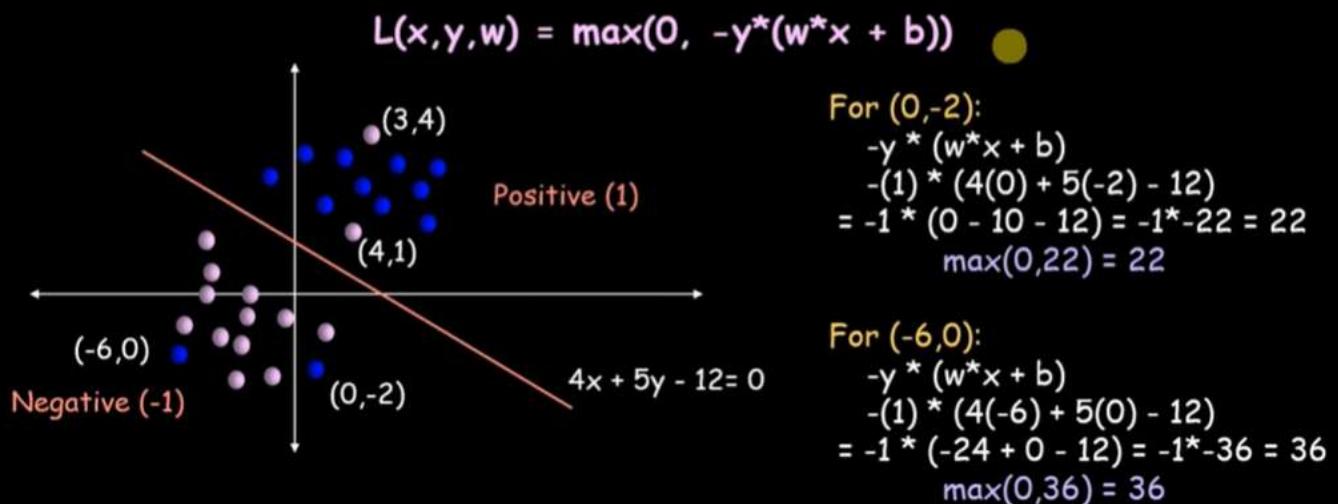


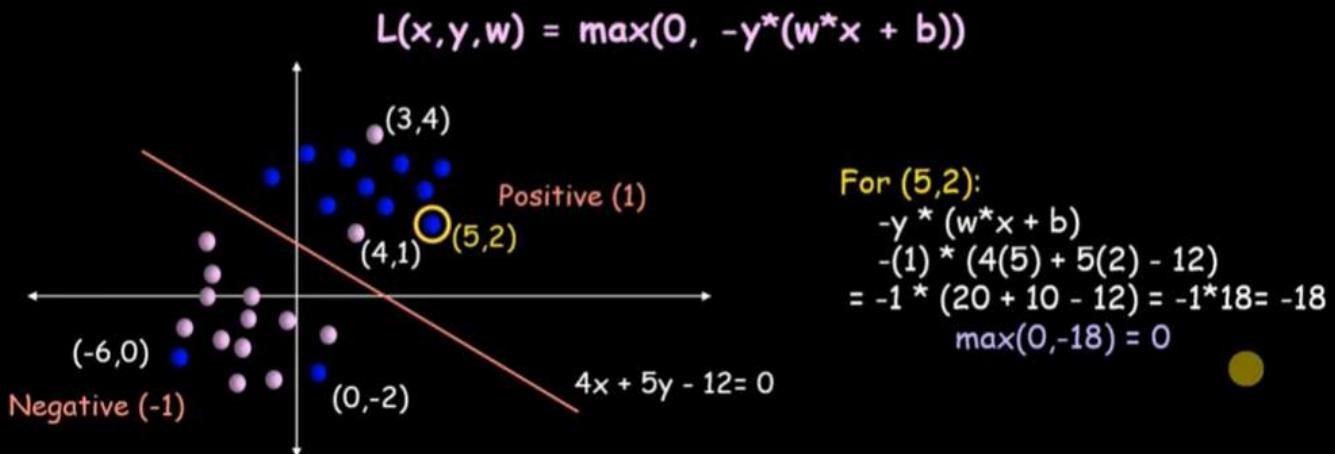
Hinge Loss formulation:



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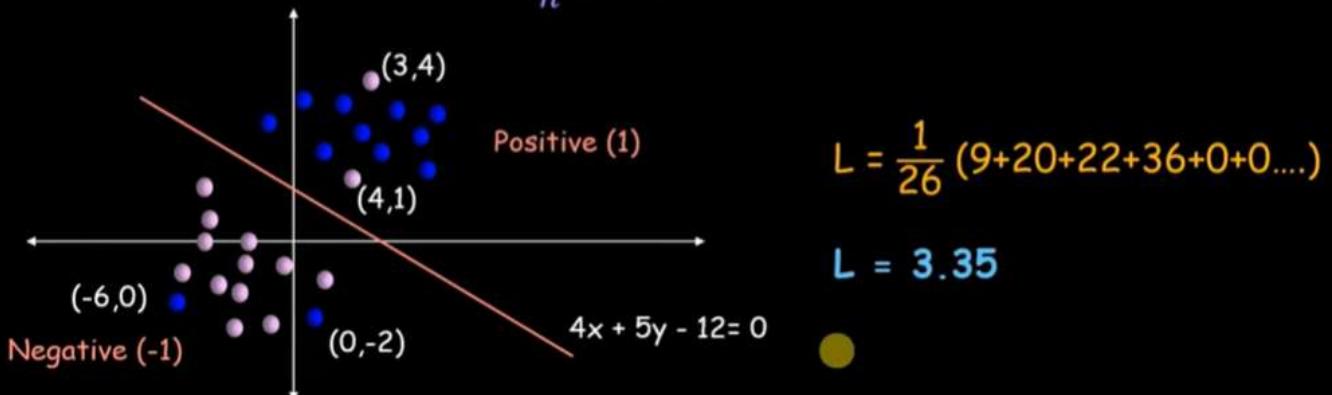


Hinge Loss formulation:



Total Hinge loss for this classifier:

$$L(x, y, w) = \frac{1}{n} \sum \max(0, -y^*(w^*x + b))$$



Loss can be reduced using Gradient Descent

We can find slope for that particular point by differentiating the loss function with weights $\left(\frac{\partial L}{\partial w}\right)$ and then adjust the current parameter using it.

We can update the weights and bias using the following formulation

$$w_{\text{new}} = w_{\text{old}} - \eta \frac{\partial L}{\partial w}$$

$$\text{where, } \frac{\partial L}{\partial w} = \frac{\partial(-y^*(w^*x + b))}{\partial w}$$

$$\frac{\partial L}{\partial w} = \frac{-y^*(\partial(w^*x) + \partial(b))}{\partial w}$$

$$\frac{\partial L}{\partial w} = -y^*(x) + 0 = -yx$$

$$w_{\text{new}} = w_{\text{old}} + \eta * y^* x$$

We can update the weights and bias using the following formulation

$$b_{\text{new}} = b_{\text{old}} - \eta \frac{\partial L}{\partial b}$$

where, $\frac{\partial L}{\partial b} = \frac{\partial(-y^*(w^*x + b))}{\partial b}$

$$\frac{\partial L}{\partial b} = \frac{-y^*(\partial(w^*x) + \partial(b))}{\partial b}$$

$$\frac{\partial L}{\partial b} = -y^*(0) + 1 = -y$$

$$b_{\text{new}} = b_{\text{old}} + \eta * y$$

Updates are made until the loss gets zero or gets nearly equal to zero