

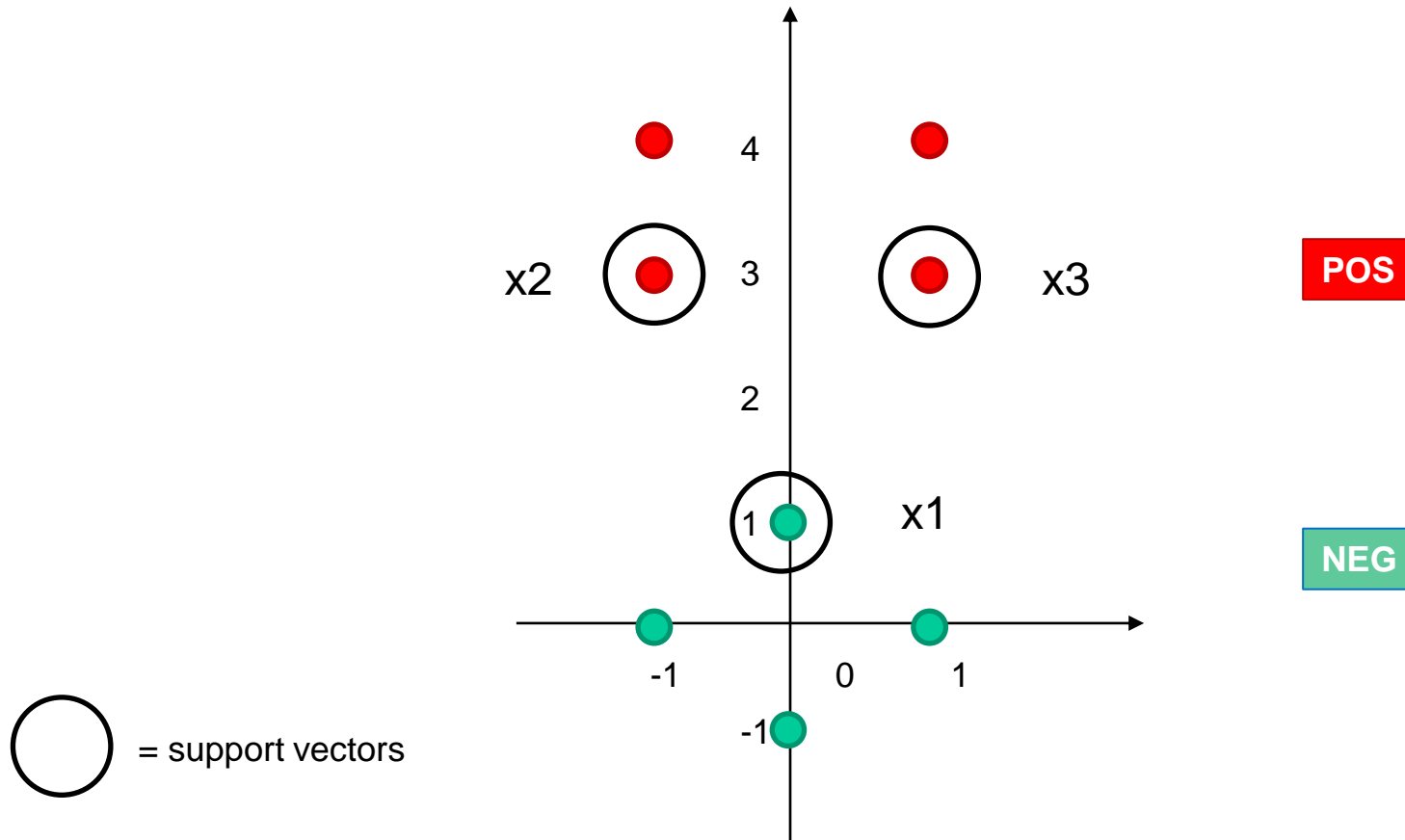
CS 1674: Intro to Computer Vision

Support Vector Machines: Exercise

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Example

Simplified case: Support vectors are given



Solving for the alphas

- We know for the support vectors, $f(x) = 1$ or -1 exactly
- Add a 1 in the feature representation for the bias
- The support vectors have coordinates and labels:
 - $x_1 = [? \ ? \ 1], y_1 = -1$
 - $x_2 = [? \ ? \ 1], y_2 = +1$
 - $x_3 = [? \ ? \ 1], y_3 = +1$

$$\mathbf{w}^T \mathbf{x} + b \quad \text{vs} \quad \mathbf{w}'^T \mathbf{x}' \quad \text{where} \quad \mathbf{w}' = \begin{bmatrix} \mathbf{w}, \\ b \end{bmatrix} \quad \mathbf{x}' = \begin{bmatrix} \mathbf{x} \\ 1 \end{bmatrix}$$

Solving for the alphas

- For support vectors, $w^T x_j = y_j$ so $\sum_i \alpha_i y_i \text{dot}(x_i, x_j) = y_j$
- Thus we can form the following system of linear equations (one for each of three j 's), with $\alpha_1, \alpha_2, \alpha_3$ as the unknowns:
- $\alpha_1 = ?, \alpha_2 = ?, \alpha_3 = ?$

Solving for w , b

We know $w = \alpha_1 y_1 x_1 + \dots + \alpha_N y_N x_N$ where $N = \# \text{ SVs}$

Thus $w = ?$, $b = ?$

Plotting the boundary

For SVMs, we used this eq for a line: $ax + cy + b = 0$
where $w = [a \ c]$

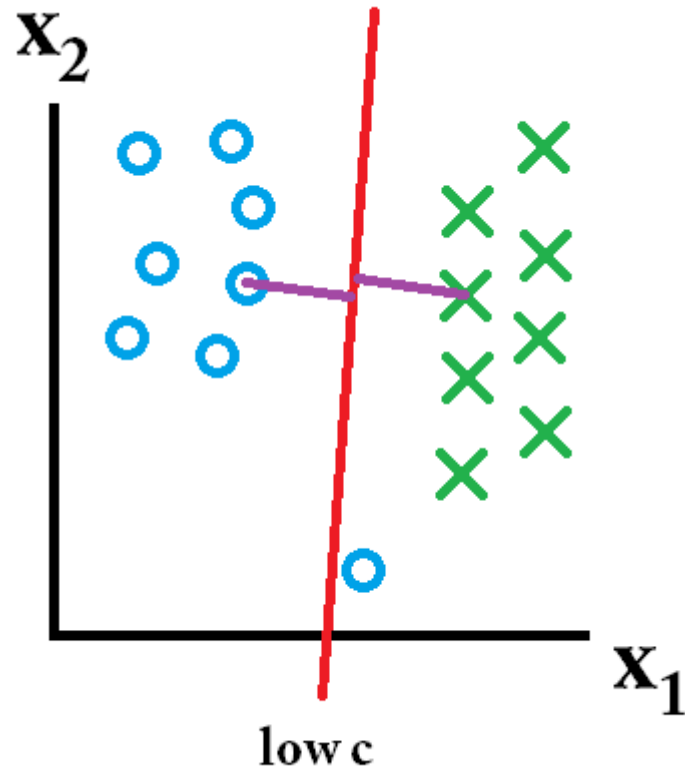
Thus $ax + b = -cy \rightarrow y = (-a/c) x + (-b/c)$

Thus y-intercept is ?

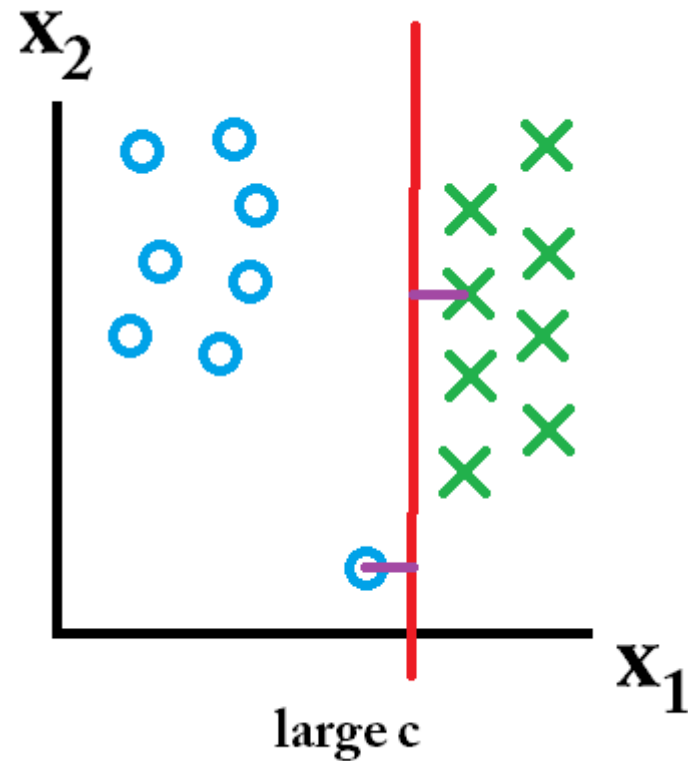
Slope of decision boundary?

Effect of margin size vs miscl. cost (c)

Training set



Misclassification ok, want large margin
(low cost)



Misclassification not ok
(high cost)

Effect of margin size vs miscl. cost (c)

Find a test set A which is better classified using a small / low cost c

Find a test set B which is better classified using a large / high cost c