

SUPPORT VECTOR MACHINES FOR LINEARLY SEPARABLE PROBLEMS

Consider an application of predicting an incoming email as Spam(+1) and Ham(-1). It is a binary classification problem where, objective is to learn $f(x)$ such that:

$$f(x) = \begin{cases} +1, & \text{if } W_1X_1 + W_2X_2 + \dots + W_nX_n \geq \theta \\ -1, & \text{otherwise} \end{cases} \quad (1)$$

where, X_i are variables in the problem.

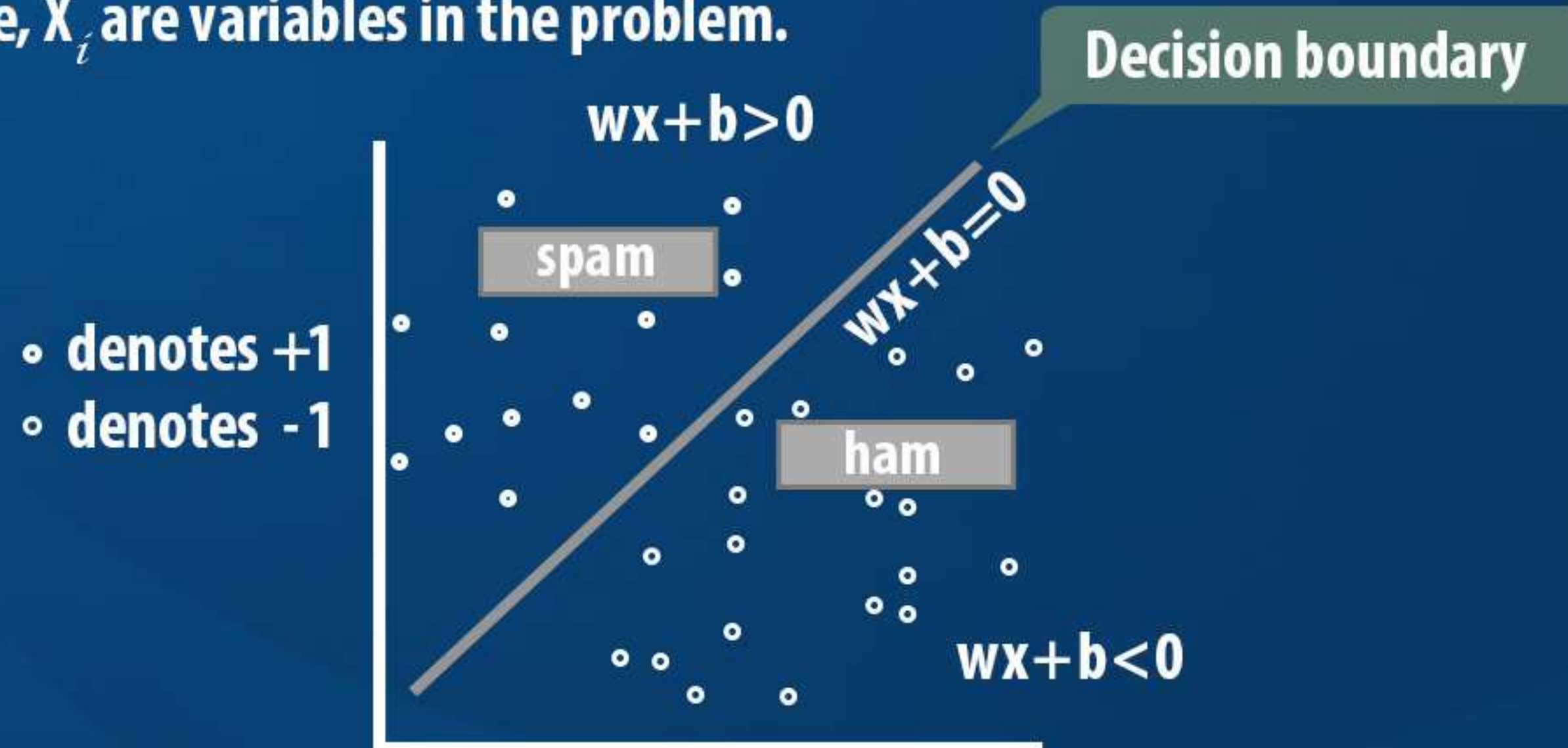


Figure 1: Support Vector Machines for Linearly separable problems

$$W \cdot X + b = 0$$

$$W_0 + W_1X_1 + W_2X_2 = 0$$

$$W_0 + W \cdot X = 0$$

$$W_0 + \sum W_i X_i$$

$$(W_1 \cdot W_2) (X_1 \cdot X_2)$$

The key objective here is:

How to find the line or in other words, how to find W 's?

HOW TO FIND THE BEST LINEAR SEPARABLE PROBLEMS?

There can exist several lines (defined by W 's) as decision boundary, which line to select?

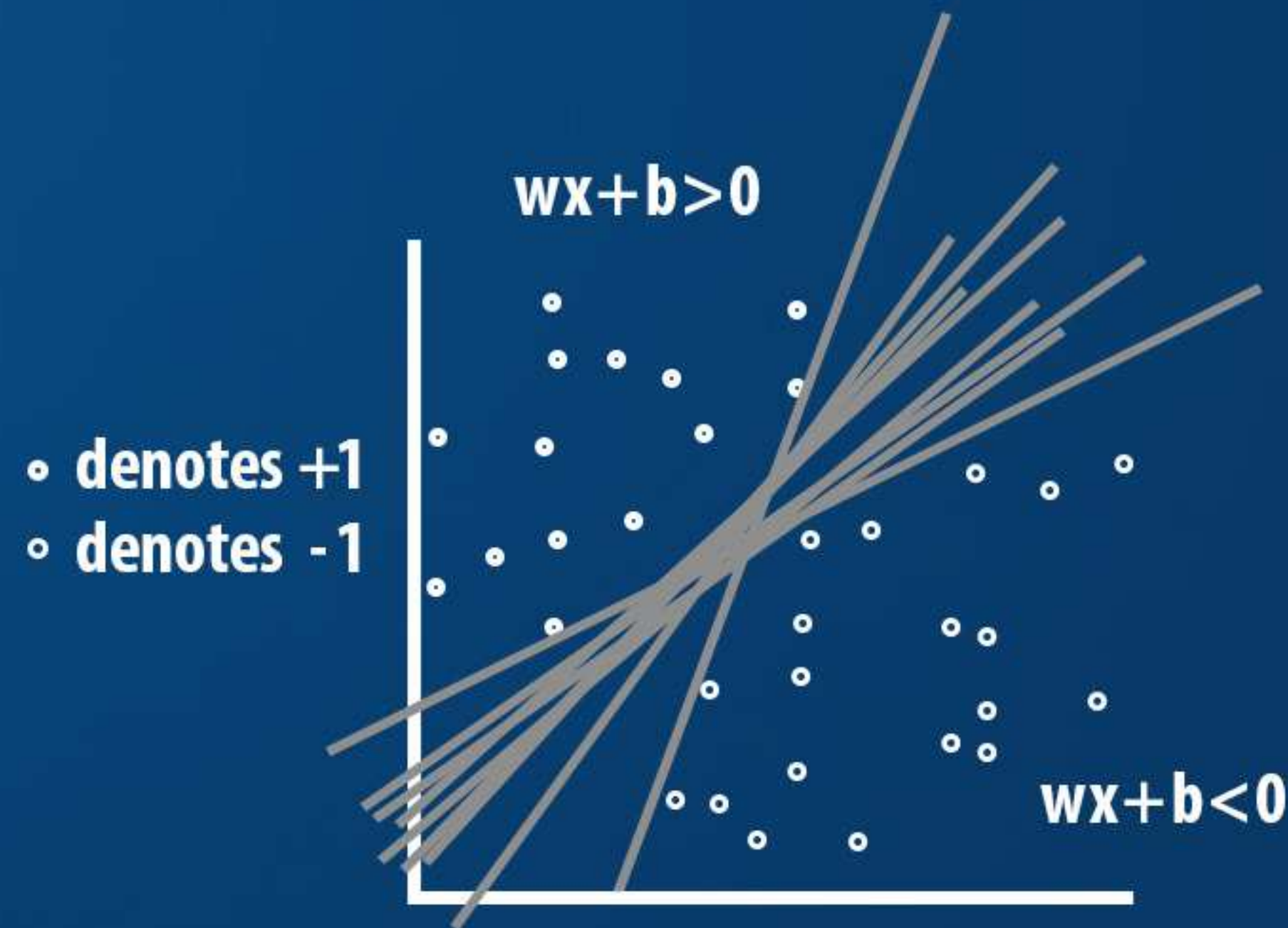


Figure 2: Multiple decision boundaries for some given binary classification problem

The best line is the one that has maximum margin.



FINDING THE BEST DECISION BOUNDARY

Margin: Distance of closest examples from the decision line.

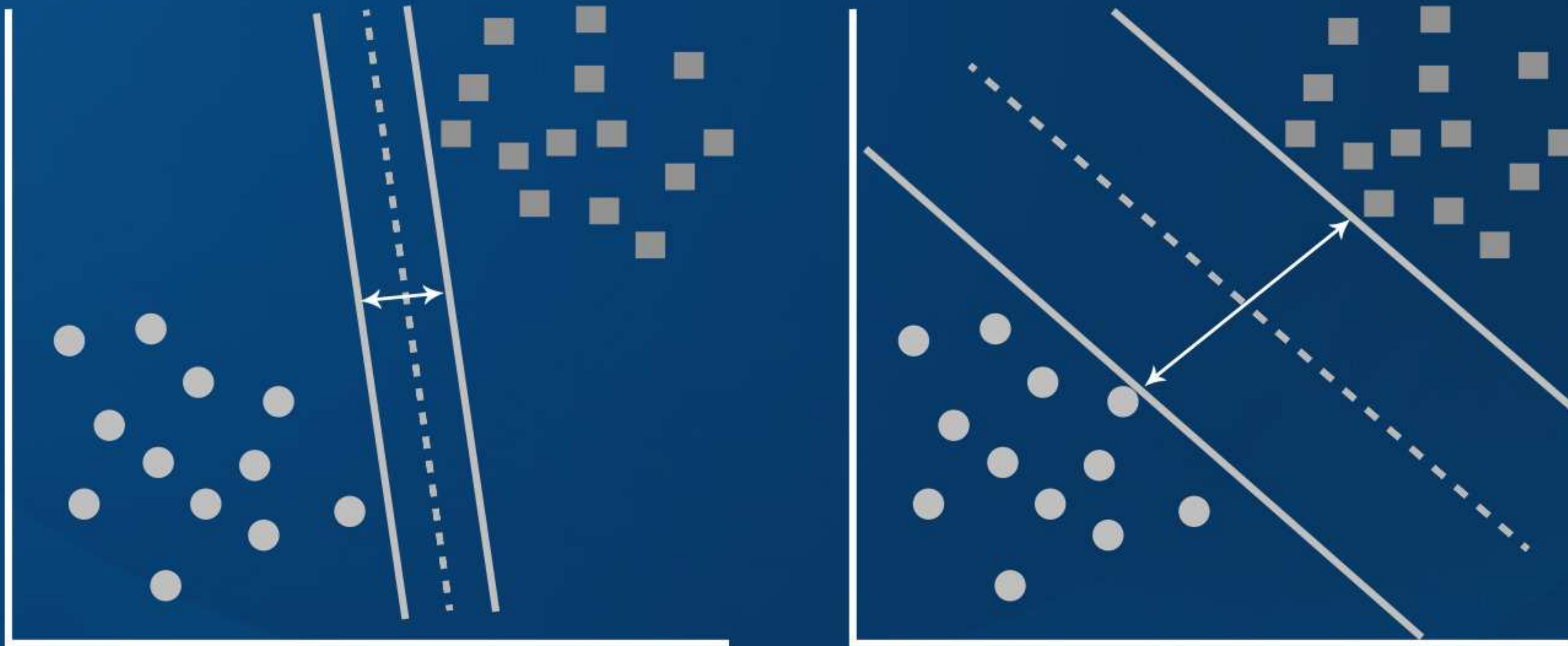


Figure 3: Finding best decision boundary based on margin.

DEFINITION OF SUPPORT VECTORS

The separating decision boundary is defined by support vectors.

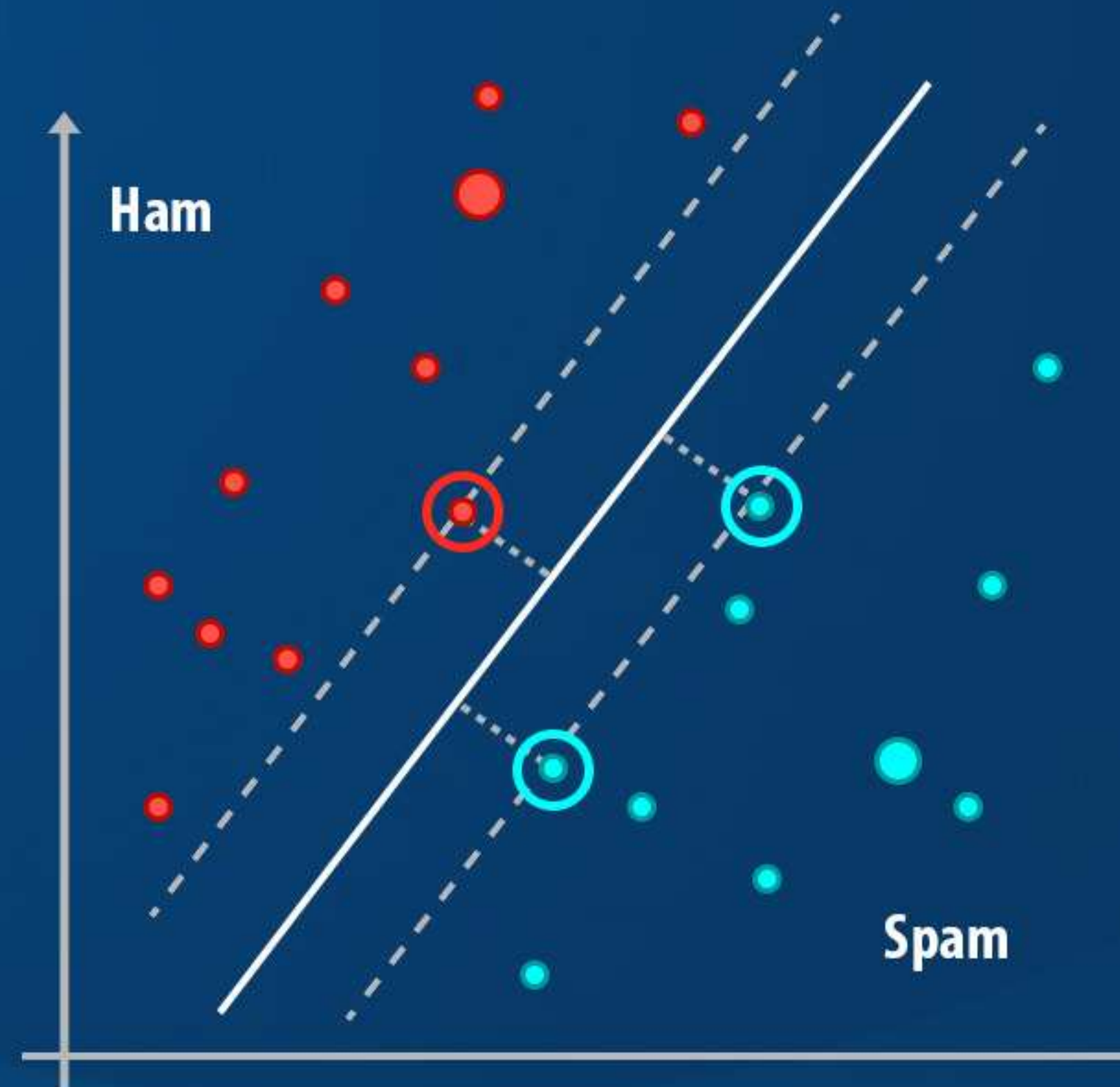
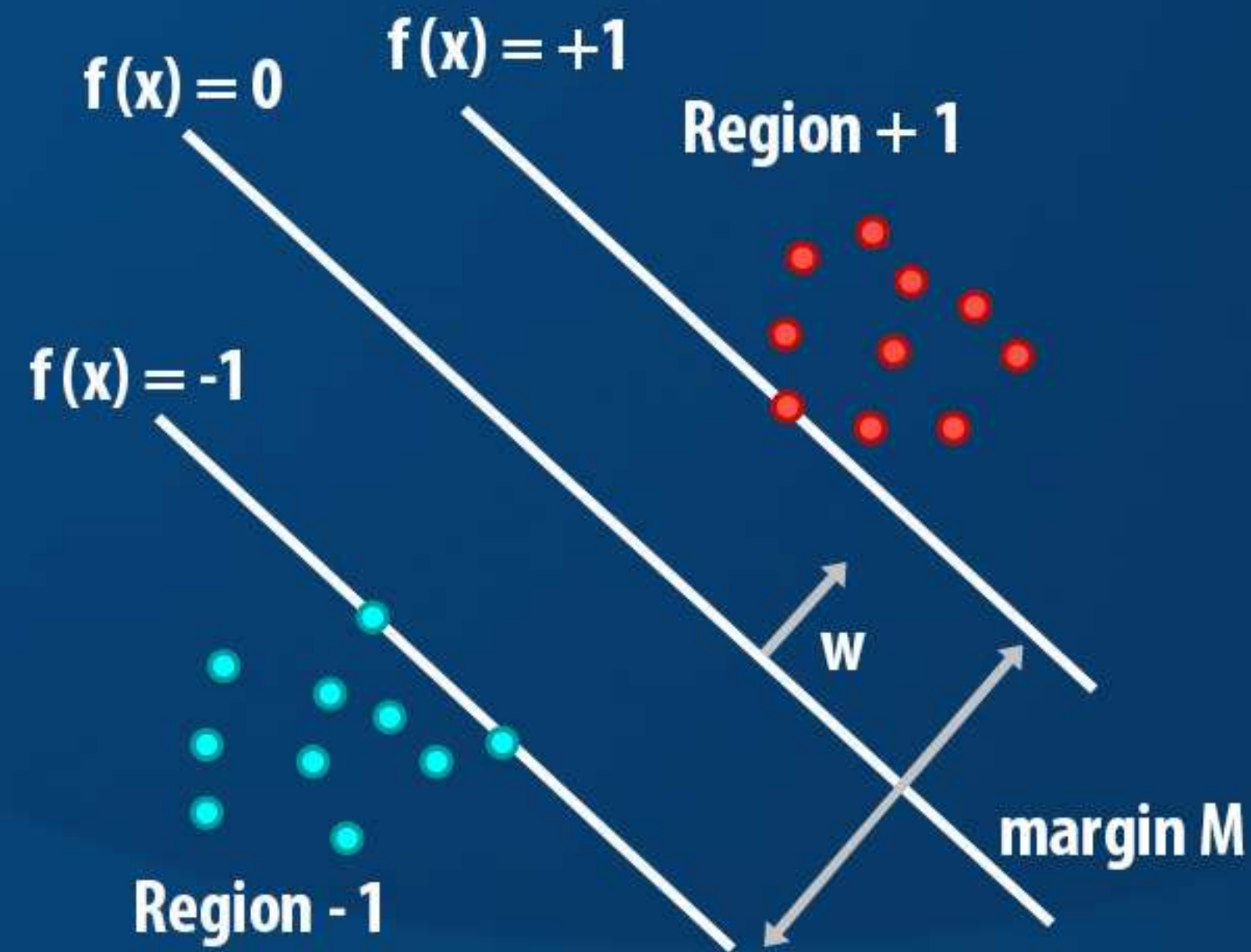


Figure 4: The highlighted data points are called Support Vectors

DEFINING MARGIN

1. Maximise Margin.
2. Define class +1 in one region, and -1 in some another region.
3. Make regions as far apart as possible.



$$f(x) = w \cdot x + b$$

$f(x) > +1$ is region +1

$f(x) < -1$ is region -1

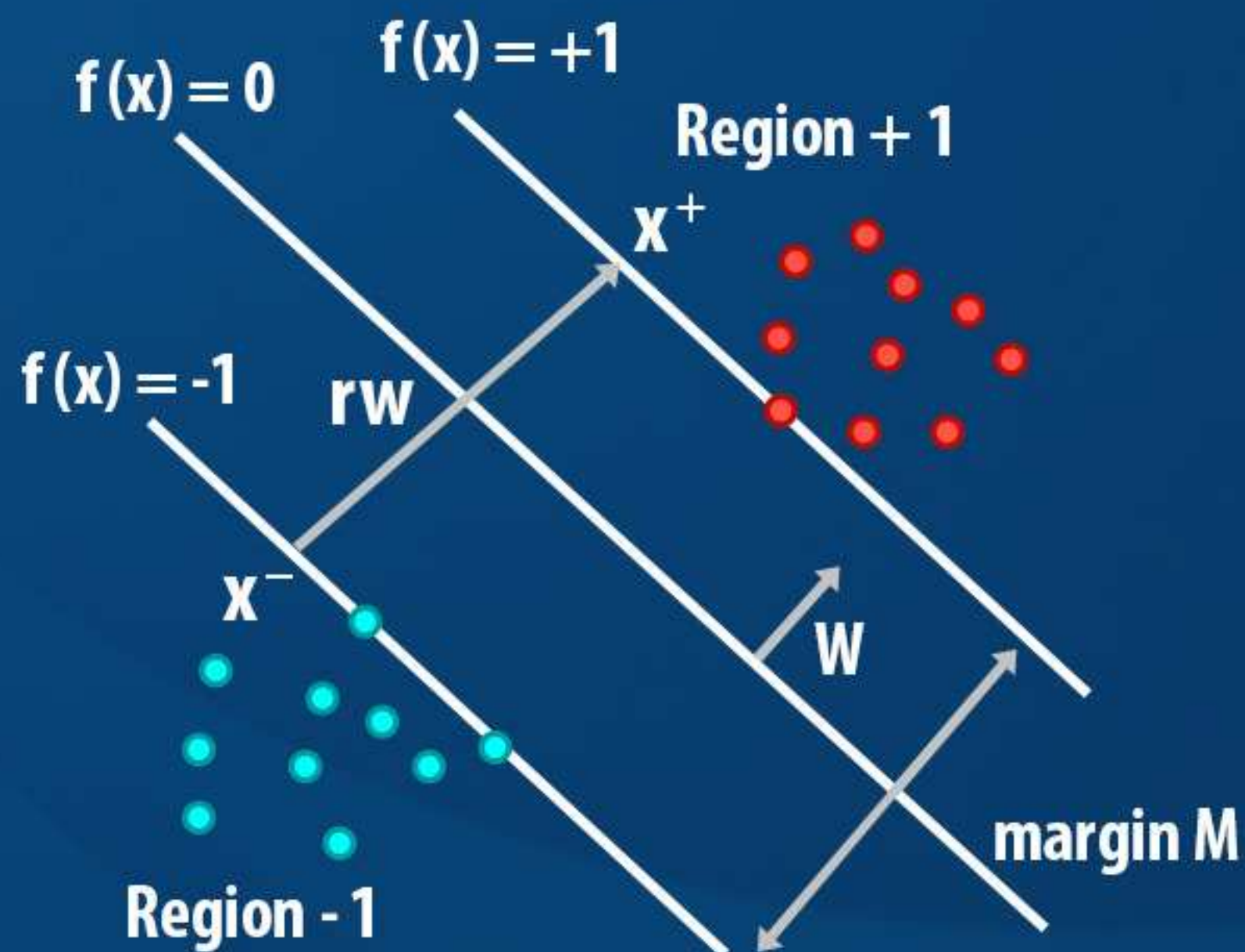
Figure 5: Defining functions for boundaries

HOW TO MAXIMIZE MARGIN?

1. The first thing to note here is that the w 's are perpendicular to the decision boundary.
2. Choose two data points x^+ such that $f(x^+) = +1$ and x^- such that $f(x^-) = -1$.

Therefore,

$$x^+ = x^- + r \cdot w \quad (2)$$



$$f(x) = wx + b$$

$$w \cdot x^- + b = -1$$

$$w \cdot x^+ + b = +1$$

$$w(x^- + rw) + b = +1$$

$$wx^- + r \|w\|^2 + b = +1$$

$$w = (1, 3)$$

$$w \cdot w = 1^2 + 3^2 = 10$$

$$\|w\|^2 = (\sqrt{1^2 + 3^2})^2 = (\sqrt{10})^2 = 10$$

$$r \|w\|^2 - 1 = 1$$

$$r = \frac{2}{\|w\|^2}$$

$$M = \|x^+ - x^-\|^2$$

$$M = \|rw\| = \frac{2}{\|w\|^2} \cdot \|w\|$$

$$M = \frac{2}{\|w\|}$$

$$M = \frac{2}{\sqrt{w^T \cdot w}}$$

MAXIMISING MARGIN IS A OPTIMISATION PROBLEM

1. Classify each data point correctly
2. Maximise the margin with subject to constraints as follows

$$M = \arg \max_w \frac{2}{\sqrt{(w^T \cdot w)}} \quad (3)$$

subject to constraint in Equations below such that:

$$y^i = +1 \longrightarrow w \cdot x^i + b \geq 1 \quad (4)$$

$$y^i = -1 \longrightarrow w \cdot x^i + b \leq 1 \quad (5)$$

where Equation 4 and Equation 5 can written as:

$$y^i (w \cdot x^i + b) \geq +1 \quad (6)$$

It is optimisation problem solved using Lagrange multipliers method.