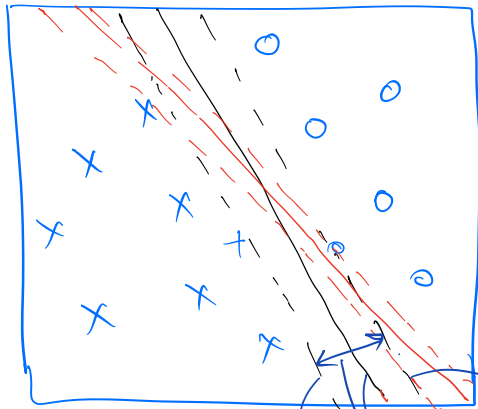


Support vector Machines



If there is a hyperplane that can perfectly classify data \implies data is linearly separable

we want to find a boundary that is a maximum margin hyperplane

\rightarrow A small margin \implies boundary highly sensitive to noise or slightly different dataset. hence worse generalization on unseen dataset.

\rightarrow Bigger margin \rightarrow less complex model & hence better generalization

\implies for a Binary class problem

labels $y_i \in \{-1, 1\}$

let $w^T x + b = 0$: equation of separating hyperplane ($d-1$ dimension) that separates the two classes.

In 2d: It's a line.

In 3d: It's a plane

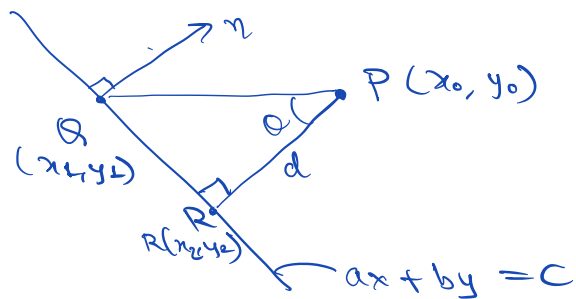
\vdots

This is achieved by placing them on opposite sides. Hence:

$$w^T x_i + b > 0 \quad y_i = 1$$

$$w^T x_i + b < 0 \quad y_i = -1$$

Distance of a point from a line



$$\begin{aligned} d &= \|QP\| \cos \theta \\ &= \frac{\|QP\| \|n\| \cos \theta}{\|n\|} \\ &= \frac{\langle QP, n \rangle}{\|n\|} \\ &= \frac{|(x_0 - x_1, y_0 - y_1) \cdot (a, b)|}{\sqrt{a^2 + b^2}} \\ &= \frac{|a(x_0 - x_1) + b(y_0 - y_1)|}{\sqrt{a^2 + b^2}} \\ &= \frac{|ax_0 - ax_1 + by_0 - by_1|}{\sqrt{a^2 + b^2}} \\ &= \frac{|ax_0 + by_0 + c|}{\sqrt{a^2 + b^2}} \end{aligned}$$

vector orthogonal
to $ax + by + c = 0$

$$Q: ax_1 + by_1 = -c \quad \text{--- (i)}$$

$$R: ax_2 + by_2 = -c \quad \text{--- (ii)}$$

$$\text{(i)} - \text{(ii)}$$

$$a(x_1 - x_2) + b(y_1 - y_2) = 0$$

$$(a, b) \cdot (x_1 - x_2, y_1 - y_2) = 0$$

\vec{RQ} along
direction of
line

$\Rightarrow (a, b)$ is a vector orthogonal
to the line

$$n = (a, b)$$

$$\Rightarrow D(x) = \frac{|w^T x + b|}{\|w\|} \quad \left\{ \begin{array}{l} \text{distance of any point in} \\ \text{the feature space to the} \\ \text{hyperplane} \end{array} \right.$$

Let

$k_+ > 0$: Distance of closest pt with $y = 1$ from hyperplane.

$k_- > 0$: Distance of closest pt with $y = -1$ from hyperplane

\Rightarrow we have constraints

$$\frac{w^T x_i + b}{\|w\|} \geq k_+ \quad \text{if } y_i = 1$$

$$\frac{w^T x_i + b}{\|w\|} \leq -k_- \quad \text{if } y_i = -1$$

\Rightarrow combining the 2 equations & let $M = k_+ = k_-$

\Rightarrow Margin = $2M$ $\left\{ \begin{array}{l} M \text{ distance from either side of the line} \end{array} \right.$

\Rightarrow we want to find maximum margin hyperplane that follows the above constraints:

$$\max_{w, b} M \quad \longrightarrow (2.72)$$

$$\text{subject to: } y_i (w^T x_i + b) \geq M \|w\|$$

If any w, b satisfy above constraints then their scaled versions would also satisfy it.

→ Conveniently choosing $\|w\| = \frac{1}{M}$ so $\max M \Leftrightarrow \min \|w\|$

→ we have equivalent problem:

$$\min \|w\|$$

or

$$\begin{array}{l} \min_{w, b} \frac{\|w\|^2}{2} \\ \text{st: } y_i(w^T x_i + b) \geq 1 \end{array}$$

→ Quadratic Programming Problem