

(Q6)

(Explain hyper plane classification using maximal-margin)
(Illustrated & derive)

① if $w \rightarrow$ weight vector realising a functional margin of 1 on the positive point x^+ and the negative point x^- , compute its Geometric margin.

② functional margin of 1 implies
 $\langle w, x^+ \rangle + b = +1$
 $\langle w, x^- \rangle + b = -1$

③ while to compute Geometric margin, Normalize w , geometric margin $\gamma \rightarrow$ functional margin of resulting classifier

$$\gamma = \frac{1}{2} \left(\left\langle \frac{w}{\|w\|_2}, x^+ \right\rangle - \left\langle \frac{w}{\|w\|_2}, x^- \right\rangle \right)$$

$$= \frac{1}{2\|w\|_2} (\langle w, x^+ \rangle - \langle w, x^- \rangle) = \frac{1}{\|w\|_2}$$

④ Resulting margin = $\frac{1}{\|w\|_2}$

⑤ Given linearly separable training sample

$$S = ((x_1, y_1), (x_2, y_2), \dots, (x_l, y_l))$$

⑥ Hyperplane (w, b) that solves optimization problem

$$\underset{w, b}{\text{minimize}} \quad \langle w, w \rangle$$

$$\text{subject to } y_i (\langle w, x_i \rangle + b) \geq 1, \quad i=1, \dots, l,$$

\rightarrow realises the maximum margin

hyperplane with geometric margin

$$= \left[\gamma = \frac{1}{\|w\|_2} \right]$$

(transform it to dual problem)