

QUESTION - 6

VC dimension of Gaussian Naive Bayes. \rightarrow is atleast three

[NAIVE BAYES ASSUMPTION]
 \downarrow

$$P(Y=1|X) = \frac{P(Y=1) P(X|Y=1)}{P(Y=1) P(X|Y=1) + P(Y=0) P(X|Y=0)}$$

$$= \frac{1}{1 + \exp \left(\ln \left(\frac{P(Y=0) P(X|Y=0)}{P(Y=1) P(X|Y=1)} \right) \right)}$$

$$P(X|Y=0) = \prod_i P(x_i|Y=0)$$

$$P(X|Y=1) = \prod_i P(x_i|Y=1)$$

$$= \frac{1}{1 + \exp \left(\ln \left(\frac{P(Y=0)}{P(Y=1)} \right) + \sum_i \ln \frac{P(x_i|Y=0)}{P(x_i|Y=1)} \right)}$$

$$= \frac{1}{1 + \exp \left(\underbrace{\ln \frac{1-\pi}{\pi}}_b + \sum_i \underbrace{\ln \frac{P(x_i|Y=0)}{P(x_i|Y=1)}}_{\text{if this can be expressed as some } w_i^T x} \right)}$$

[Use assumption of GAUSSIAN]

$$\Rightarrow \sum_i \ln \frac{P(x_i|Y=0)}{P(x_i|Y=1)} = \sum_i \ln \frac{\frac{1}{\sqrt{2\pi\sigma_i^2}} \exp \left(-\frac{(x_i - \mu_{i0})^2}{2\sigma_i^2} \right)}{\frac{1}{\sqrt{2\pi\sigma_i^2}} \exp \left(-\frac{(x_i - \mu_{i1})^2}{2\sigma_i^2} \right)}$$

$$\Rightarrow \sum_i \ln \left(\exp \left(\frac{(x_i - \mu_{i1})^2 - (x_i - \mu_{i0})^2}{2\sigma_i^2} \right) \right)$$

$$\Rightarrow \sum_i \left[\frac{\mu_{i1}^2 - \mu_{i0}^2}{2\sigma_i^2} + x_i \frac{\mu_{i0} - \mu_{i1}}{\sigma_i^2} \right]$$

Substituting back gives $P(Y=1|X) = \frac{1}{1 + \exp(b + w^T x)}$

[SAME FORM AS LOGISTIC REGRESSION]

LINEAR SEPARATOR can

exist $\Rightarrow VC(H) \geq 3$