

Assignment - 1

AI20BTECH11006

1)

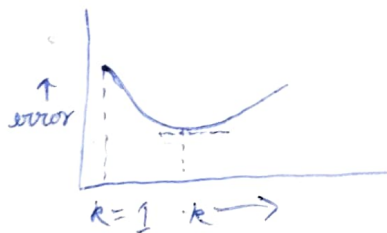
a) $k=1$

The training error will be 0, because the nearest point will be that instance itself.

As k begins to increase from 1 to $n-1$, we will start getting nearest points from other classes because the data is overlapping.

At $k=n$, we would get an error rate of 0.5 because the nearest n -points would consist of both classes of $n/2$ points each. even if we consider weighted

b) For $k=1$ the generalisation error would be very high because of over-fitting. As k grows the error should decrease because in k ~~the~~ nearest neighbour the data point is assumed to be the same label as its neighbour. After a certain value of k , the error would start increasing again because more & more points of the other classes would be in the k -nearest neighbour. The plot would look like



For the given question however the plot can fluctuate a lot because the points of other classes would be in the nearest neighbours even for small k .

c)

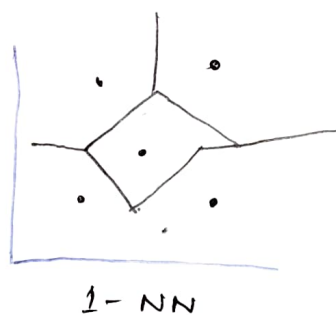
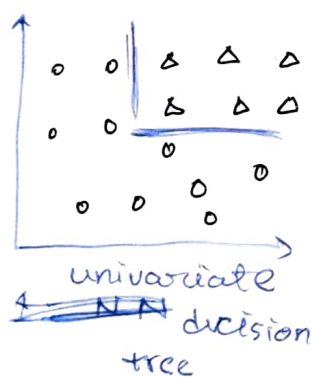
- 1) The ~~test~~ data size must increase a lot to accomodate the higher input dimension, this is because the distance can be similar & large for two different classes.

In certain dimensions, a set of points may be closer while overall ~~a~~ points may be closer to each other. The points may live on a hypersphere.

- 2) The amount of computation would increase linearly with the dimensions.

d) No,

The univariate decision tree can only form perpendicular lines to distinguish between classes, while in the case of 1-NN we have voronoi diagram which may not have only perpendicular lines. A pictorial representation for the same can be found below



2)

a) The PDF of gaussian is ~~is~~ given by

$$f(x) = \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{(x-\mu)^2}{2\sigma^2}}$$

10 We have to maximise $f(x_1) f(x_2) \dots f(x_n)$

or $\log(f(x_1)) + \log(f(x_2)) + \dots + \log(f(x_n))$

The derivative w.r.t x should be 0.

$$\frac{f'(x_1)}{f(x_1)} + \frac{f'(x_2)}{f(x_2)} + \dots + \frac{f'(x_n)}{f(x_n)} = 0$$

$$x_1 - \mu + x_2 - \mu + x_3 - \mu + \dots + x_n - \mu = 0$$

$$\mu = \frac{x_1 + x_2 + \dots + x_n}{n}$$

$$\text{for Class 1} \Rightarrow \mu_1 = \frac{0.5 + 0.1 + 0.2 + 0.4 + 0.3 + 0.2 + 0.1 + 0.35 + 0.25}{10} = 0.26$$

$$\text{for Class 2} \Rightarrow \mu_2 = \frac{0.9 + 0.8 + 0.75 + 1.0}{4} = \frac{3.45}{4} = 0.8625$$

$$\sigma_1^2 = 0.0149$$

$$\sigma_2^2 = 0.0092$$

Class Probability

$$P_1 = \frac{n(\text{class 1})}{\text{Total}} = \frac{5}{7}$$

$$P_2 = \frac{n(\text{class 2})}{\text{Total}} = \frac{2}{7}$$

~~P(class 1)~~

$$P(\text{class 1} | X = 0.6) = \frac{P(X = 0.6 | \text{class 1}) \cdot P(\text{class 1})}{P(X = 0.6)}$$

$$= \frac{P(X = 0.6 | \text{class 1}) p_1}{P(X = 0.6, \text{class 1}) + P(X = 0.6, \text{class 2})}$$

$$= \frac{P(X = 0.6 | \text{class 1}) p_1}{P(X = 0.6 | \text{class 1}) p_1 + P(X = 0.6 | \text{class 2}) p_2}$$

$$= \frac{\frac{1}{\sigma_1} e^{-\frac{(0.6 - 0.26)^2}{2\sigma_1^2}} \frac{5}{7}}{\frac{1}{\sigma_1} e^{-\frac{(0.6 - 0.26)^2}{2\sigma_1^2}} \frac{5}{7} + \frac{1}{\sigma_2} e^{-\frac{(0.6 - 0.8625)^2}{2\sigma_2^2}} \frac{2}{7}}$$

$$= \frac{\frac{1}{\sqrt{0.0149}} e^{-\frac{0.34^2}{2(0.0149)}} \times 5}{\frac{1}{\sqrt{0.0149}} e^{-\frac{0.34^2}{2(0.0149)}} \frac{5}{7} + \frac{1}{\sqrt{0.0092}} e^{-\frac{(0.2625)^2}{2(0.0092)}}$$

$$= 0.6319$$

6)

$$x_{\text{politics}} = \begin{bmatrix} 1 & 0 & 1 & 1 & 1 & 0 & 1 & 1 \\ 0 & 0 & 0 & 1 & 0 & 0 & 1 & 1 \\ 1 & 0 & 0 & 1 & 1 & 0 & 1 & 1 \\ 0 & 1 & 0 & 0 & 1 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 & 1 & 0 & 1 & 1 \\ 0 & 0 & 0 & 1 & 1 & 0 & 0 & 1 \end{bmatrix}$$

$$x_{\text{sports}} = \begin{bmatrix} 1 & 1 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 & 0 & 0 & 0 \\ 1 & 1 & 0 & 1 & 0 & 0 & 0 & 0 \\ 1 & 1 & 0 & 1 & 0 & 0 & 0 & 1 \\ 1 & 1 & 0 & 1 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 & 1 & 0 & 0 \end{bmatrix}$$

$$x = [1 \ 0 \ 0 \ 1 \ 1 \ 1 \ 1 \ 0]$$

The required probability is

$$\frac{1}{3} \times \frac{5}{6} \times \frac{5}{6} \times \frac{5}{6} \times \frac{5}{6} \times \frac{4}{6} \times \frac{1}{6}$$

$$\frac{1}{3} \times \frac{5}{6} \times \frac{5}{6} \times \frac{5}{6} \times \frac{5}{6} \times \frac{4}{6} \times \frac{1}{6} + \frac{4}{6} \times \frac{2}{6} \times \frac{5}{6} \times \frac{4}{6} \times \frac{1}{6} \times \frac{1}{6} \times \frac{1}{6}$$

$$= 1$$