

## MODULE 4

1. Consider a medical diagnosis problem in which there are two alternative hypotheses: 1. that the patient has a particular form of cancer (+) and 2. That the patient does not (-). A patient takes a lab test and the result comes back positive. The test returns a correct positive result in only 98% of the cases in which the disease is actually present, and a correct negative result in only 97% of the cases in which the disease is not present. Furthermore, .008 of the entire population have this cancer. Determine whether the patient has Cancer or not using MAP hypothesis.

Solution :-

\* From the above situation, probabilities can be summarized as follow -

$$P(\text{cancer}) = 0.008 \quad P(\neg \text{cancer}) = 0.992$$

$$P(\oplus | \text{cancer}) = 0.98 \quad P(\ominus | \text{cancer}) = 0.02$$

$$P(\oplus | \neg \text{cancer}) = 0.03 \quad P(\ominus | \neg \text{cancer}) = 0.97$$

Should we diagnose the patient as having cancer or not?

$$P(\oplus | \text{cancer}) P(\text{cancer}) = (0.98) * (0.008) = 0.0078$$

$$P(\oplus | \neg \text{cancer}) P(\neg \text{cancer}) = 0.03 * 0.992 = 0.0298$$

So, 
$$\underline{h_{MAP} = \neg \text{cancer}}$$

\* The exact posterior probabilities can also be determined by normalizing the above quantities so that they sum to 1

$$P(\text{cancer} | \oplus) = \frac{0.0078}{0.0078 + 0.0298} = \underline{\underline{0.21}}$$

$$P(\neg \text{cancer} | \oplus) = \frac{0.0298}{0.0078 + 0.0298} = \underline{\underline{0.79}}$$

2. Apply Naïve Bayes classifier for *PlayTennis* concept learning problem to classify the following novel instance

< Outlook = sunny, Temperature = cool, Humidity = high, Wind = strong >

| Day | Outlook  | Temperature | Humidity | Wind   | PlayTennis |
|-----|----------|-------------|----------|--------|------------|
| D1  | Sunny    | Hot         | High     | Weak   | No         |
| D2  | Sunny    | Hot         | High     | Strong | No         |
| D3  | Overcast | Hot         | High     | Weak   | Yes        |
| D4  | Rain     | Mild        | High     | Weak   | Yes        |
| D5  | Rain     | Cool        | Normal   | Weak   | Yes        |
| D6  | Rain     | Cool        | Normal   | Strong | No         |
| D7  | Overcast | Cool        | Normal   | Strong | Yes        |
| D8  | Sunny    | Mild        | High     | Weak   | No         |
| D9  | Sunny    | Cool        | Normal   | Weak   | Yes        |
| D10 | Rain     | Mild        | Normal   | Weak   | Yes        |
| D11 | Sunny    | Mild        | Normal   | Strong | Yes        |
| D12 | Overcast | Mild        | High     | Strong | Yes        |
| D13 | Overcast | Hot         | Normal   | Weak   | Yes        |
| D14 | Rain     | Mild        | High     | Strong | No         |

Table: Training examples for the target concept *PlayTennis*

⇒ Solution

\* Task is to predict the target value (Yes or No) of the target concept *playTennis* for this new instance.

$$* V_{NB} = \underset{v_j \in \{yes, no\}}{\operatorname{argmax}} P(v_j) \prod_i P(a_i | v_j) \leftarrow \text{equ (1)}$$

$$V_{NB} = \underset{v_j \in \{yes, no\}}{\operatorname{argmax}} P(v_j) \cdot P(\text{outlook} = \text{sunny} | v_j) \cdot$$

$$P(\text{Temp} = \text{cool} | v_j) \cdot P(\text{Humidity} = \text{high} | v_j) \cdot P(\text{Wind} = \text{Strong} | v_j)$$

\* First, the probabilities of the different target values is estimated based on their frequencies over 14 training examples

$$P(\text{playTennis} = \text{yes}) = \frac{9}{14} = 0.64$$

$$P(\text{playTennis} = \text{no}) = \frac{5}{14} = 0.36$$

\* Next, estimate the conditional probabilities

$$P(\text{outlook} = \text{sunny} \mid \text{playTennis} = \text{yes}) = \frac{2}{9} = 0.22$$

$$P(\text{outlook} = \text{sunny} \mid \text{playTennis} = \text{no}) = \frac{3}{5} = 0.60$$

$$P(\text{Temp} = \text{cool} \mid \text{playTennis} = \text{yes}) = \frac{3}{9} = 0.33$$

$$P(\text{Temp} = \text{cool} \mid \text{playTennis} = \text{no}) = \frac{1}{5} = 0.20$$

$$P(\text{Humidity} = \text{high} \mid \text{playTennis} = \text{yes}) = \frac{3}{9} = 0.33$$

$$P(\text{Humidity} = \text{high} \mid \text{playTennis} = \text{no}) = \frac{4}{5} = 0.80$$

$$P(\text{wind} = \text{strong} \mid \text{playTennis} = \text{yes}) = \frac{3}{9} = 0.33$$

$$P(\text{wind} = \text{strong} \mid \text{playTennis} = \text{no}) = \frac{3}{5} = 0.60$$

\* Next calculate VNB according equation ①

$$= P(\text{yes}) \cdot P(\text{sunny} \mid \text{yes}) \cdot P(\text{cool} \mid \text{yes}) \cdot P(\text{high} \mid \text{yes}) \cdot P(\text{strong} \mid \text{yes})$$

$$= 0.64 * 0.22 * 0.33 * 0.33 * 0.33$$

$$= \underline{\underline{0.0053}}$$

$$\begin{aligned}
 &= P(\text{no}) \cdot P(\text{Sunny}|\text{no}) \cdot P(\text{cool}|\text{no}) \cdot P(\text{high}|\text{no}) \cdot P(\text{Strong}|\text{no}) \\
 &= 0.036 * 0.60 * 0.20 * 0.80 * 0.60 \\
 &= \underline{\underline{0.0206}}
 \end{aligned}$$

Thus, the naive Bayes classifier assigns the target value "no" to the new instance.

i.e., playTennis = no

| Outlook | Temp | Humidity | Wind   | playTennis |
|---------|------|----------|--------|------------|
| Sunny   | Cool | high     | Strong | no         |

\* Normalizing the quantities to sum to 1, calculate the conditional probability of target values.

$$\text{Yes} = \frac{0.0053}{0.0053 + 0.0206} = \underline{\underline{0.205}}$$

$$\text{No} = \frac{0.0206}{0.0053 + 0.0206} = \underline{\underline{0.795}}$$