

Homework 4

Q1) Bayes' theorem describes updates to probabilities of the hypotheses, based on given data.

Let M, N denote two events. Now Baye's theorem provides a relⁿ b/w them.

$$P(M \cap N) = P(M) P(N/M) \quad - (1)$$

$$P(M \cap N) = P(N) P(M/N) \quad - (2)$$

Equating (1) and (2), we get

$$P(M/N) = \frac{P(N/M) P(M)}{P(N)}$$

In machine learning, various attributes and features are used for predictions and classifications.

Using the theorem, given probability distributions of various variables, we can calculate the probability of the response variable belonging to a particular value.

Thus our end goal is to find the best possible hypothesis, given initial probabilities. Thus prior prob. and best hypothesis can be found with the Bayes' theorem.

Q2) Wkt, from Bayes' rule,

$$\begin{aligned} P(\text{cancer}/+) &= \frac{P(+/\text{cancer}) \times P(\text{cancer})}{P(+/\text{cancer}) \times P(\text{cancer}) + P(+/\sim\text{cancer}) \times P(\sim\text{cancer})} \\ &= \frac{0.98 \times 0.008}{0.98 \times 0.008 + 0.03 \times 0.992} \\ &= \underline{\underline{0.21}} \end{aligned}$$

Also,

$$\begin{aligned} P(\text{cancer}/++) &= \frac{P(+/\text{cancer}) \times P(\text{cancer}/+)}{P(+/\text{cancer}) \times P(\text{cancer}/+) + P(+/\sim\text{cancer}) \times P(\sim\text{cancer}/+)} \end{aligned}$$

Where

$$P(\text{cancer}/+) = 0.21 \quad \text{and}$$

$$P(\sim\text{cancer}/+) = 0.79$$

$$\begin{aligned} \therefore P(\text{cancer}/++) &= \frac{0.98 \times 0.21}{0.9 \times 0.21 + 0.03 \times 0.79} \\ &= \underline{\underline{0.91}} \end{aligned}$$

$$\begin{aligned} \therefore P(\sim\text{cancer}/++) &= 1 - P(\text{cancer}/++) \\ &= \underline{\underline{0.09}} \end{aligned}$$

Q3)

Given, outlook = sun
Temp = cool
Humidity = high
wind = strong

$$\text{Now, } P(\text{Yes}) = \frac{8}{12} = \underline{\underline{0.67}}$$

$$P(\text{No}) = \frac{4}{12} = \underline{\underline{0.33}}$$

Now,

$$P(\text{strong} / \text{Yes}) = 0.375$$

$$P(\text{high} / \text{Yes}) = 0.375$$

$$P(\text{cool} / \text{Yes}) = 0.375$$

$$P(\text{sun} / \text{Yes}) = 0.25$$

$$P(\text{strong} / \text{No}) = 0.5$$

$$P(\text{high} / \text{No}) = 0.75$$

$$P(\text{cool} / \text{No}) = 0.25$$

$$P(\text{sun} / \text{No}) = 0.75$$

$$\therefore P(\text{Yes}) = 0.67 \times 0.25 \times 0.375 \times 0.375 \times 0.375 \\ = \underline{\underline{0.833 \times 10^{-3}}}$$

$$P(\text{No}) = 0.33 \times 0.75 \times 0.25 \times 0.75 \times 0.5 \\ = \underline{\underline{0.0232}}$$

Hence, Naive Bayes' classifier assigns the value of No

Q4) For the first iteration,

$$\begin{aligned} \text{net}_c &= w_{co} + a \times w_{ca} + b \times w_{cb} \\ &= 0.1 + 0.1 + 0 \\ &= \underline{\underline{0.2}} \end{aligned}$$

$$\therefore O_c = \frac{1}{1 + e^{-\text{net}_c}} = \frac{1}{1 + e^{-0.2}} = \underline{\underline{0.55}}$$

$$\therefore \text{net}_d = w_{do} + O_c \times w_{dc}$$

$$\begin{aligned} &= 0.1 + 0.55 \times 0.1 \\ &= \underline{\underline{0.155}} \end{aligned}$$

Also,

$$\begin{aligned} O_d &= \frac{1}{1 + e^{-\text{net}_d}} = \frac{1}{1 + e^{-0.155}} \\ &= \underline{\underline{0.539}} \end{aligned}$$

Hence using backpropagation,

$$\begin{aligned} \delta_d &= O_d \times (1 - O_d) \times (t_d - O_d) \\ &= 0.539 \times (1 - 0.539) \times (1 - 0.539) \\ &= \underline{\underline{0.115}} \end{aligned}$$

Now,

$$\begin{aligned}\Delta w_{dc} &= \eta * \delta_d * O_c + \alpha * 0 \\ &= 0.3 * 0.115 * 0.55 + 0 \\ &= \underline{\underline{0.019}}\end{aligned}$$

$$\Delta w_{do} = 0.034$$

$$\begin{aligned}\therefore w_{dc} &= w_{dc} + \Delta w_{dc} \\ &= 0.1 + 0.019 = \underline{\underline{0.119}}\end{aligned}$$

$$\begin{aligned}w_{do} &= w_{do} + \Delta w_{do} \\ &= 0.1 + 0.034 = \underline{\underline{0.134}}\end{aligned}$$

$$\begin{aligned}\delta_c &= O_c * (1 - O_c) * (w_c + \delta_d) \\ &= 0.55 * (1 - 0.55) * (0.1 * 0.115) \\ &= \underline{\underline{0.003}}\end{aligned}$$

$$\begin{aligned}\Delta w_{ca} &= \eta * \delta_c * \chi_a + \alpha * 0 \\ &= 0.3 * 0.003 * 1 = \underline{\underline{0.001}}\end{aligned}$$

and

$$\Delta w_{co} = 0.001$$

$$\Delta w_{cb} = 0 //$$

Now weights after 1st iteration,

$$w_{co} = 0.101$$

$$w_{do} = 0.134$$

$$w_{ca} = 0.101$$

$$w_{dc} = \underline{\underline{0.119}}$$

$$\underline{\underline{w_{cb} = 0.100}}$$

* Second iteration

$$\begin{aligned} net_c &= w_{co} + a w_{ca} + b w_{cb} \\ &= 0.101 + 0 + 0.100 = \underline{\underline{0.201}} \end{aligned}$$

$$O_c = \frac{1}{1 + e^{-0.201}} = \underline{\underline{0.55}}$$

$$\begin{aligned} net_d &= w_{do} + O_c * w_{dc} = 0.134 + 0.55 * 0.119 \\ &= \underline{\underline{0.1994}} \end{aligned}$$

$$\text{Similarly, } O_d = \frac{1}{1 + e^{-0.1994}} = \underline{\underline{0.5496}}$$

Now using backpropagation,

$$\begin{aligned} \odot_d &= O_d * (1 - O_d) * (t_d - O_d) \\ &= 0.5496 (1 - 0.5496) (0 - 0.5496) \\ &= \underline{\underline{-0.136}} \end{aligned}$$

$$\begin{aligned}
 \Delta w_{dc} &= \eta \times \delta_d \times O_c + \alpha \times \Delta w_{dc}(\text{old}) \\
 &= 0.3 (-0.136) (0.55) + 0.9 (0.019) \\
 &= \underline{\underline{-0.0053}}
 \end{aligned}$$

$$\text{and } \Delta w_{do} = -0.01$$

$$\begin{aligned}
 \therefore w_{dc} &= w_{dc} + \Delta w_{dc} \\
 &= 0.119 + (-0.003) \\
 &= \underline{\underline{0.113}}
 \end{aligned}$$

$$\begin{aligned}
 w_{do} &= w_{do} + \Delta w_{do} = 0.134 + (-0.01) \\
 &= \underline{\underline{0.124}}
 \end{aligned}$$

$$\begin{aligned}
 \delta_c &= O_c \times (1 - O_c) \times (w_{dc} + \delta_d) \\
 &= 0.55 (1 - 0.55) (0.113) (-0.136) \\
 &= \underline{\underline{-0.0038}}
 \end{aligned}$$

$$\begin{aligned}
 \therefore \Delta w_{ca} &= \eta \times \delta_c \times n_a + \alpha \times w_{ca}(\text{old}) \\
 &= 0.3 \times (-0.0038) \times 0 + 0.9 \times 0.101 \\
 &= \underline{\underline{0.009}}
 \end{aligned}$$

$$\text{and } \Delta w_{co} = 0$$

$$\Delta w_{cb} = \underline{\underline{-0.001}}$$

Here, final weights are given as

$$W_{co} = 0.101$$

$$W_{ca} = 0.102$$

$$W_{cb} = 0.099$$

$$W_{do} = 0.124$$

$$W_{dc} = 0.113$$