

# 'Assignment -03'

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Ans. to the que no-01

(a)

Given,

$$P(A, B, C) = 0.2 \quad , \quad P(\bar{A}, \bar{B}, C) = 0.35$$

$$P(A, \bar{B}, C) = 0.15 \quad , \quad P(A) = 0.5$$

$$P(A, \bar{B}, \bar{C}) = 0.05 \quad , \quad P(\bar{C}) = 0.25$$

$$P(\bar{A}, B, C) = 0.05$$

$$P(\bar{A}, B, \bar{C}) = 0.05$$

Now,

$$P(A) = P(A, B, C) + P(A, \bar{B}, \bar{C}) + P(A, \bar{B}, C) + P(\bar{A}, \bar{B}, \bar{C})$$

$$\Rightarrow 0.5 = 0.2 + P(A, \bar{B}, \bar{C}) + 0.15 + 0.05$$

$$\Rightarrow P(A, B, \bar{C}) = 0.5 - (0.2 + 0.15 + 0.05)$$
$$= 0.1$$

$$\therefore P(A, B, \bar{C}) = 0.1$$

and,

$$P(\bar{A}) = P(\bar{A}, B, C) + P(\bar{A}, \bar{B}, \bar{C}) + P(\bar{A}, \bar{B}, C) + P(\bar{A}, B, \bar{C})$$

$$\Rightarrow 1 - P(A) = 0.05 + 0.05 + 0.35 + P(\bar{A}, \bar{B}, \bar{C})$$

$$\Rightarrow 1 - 0.5 = 0.45 + P(\bar{A}, \bar{B}, \bar{C})$$

$$\Rightarrow P(\bar{A}, \bar{B}, \bar{C}) = 0.05$$

$$\therefore P(\bar{A}, \bar{B}, \bar{C}) = 0.05$$

So, the joint probability distribution table,

		B		$\neg B$	
		C	$\neg C$	C	$\neg C$
A	0.2	0.1	0.15	0.05	
	0.05	0.05	0.35	0.05	

(b)

$$P(\neg B) = P(A, \neg B, C) + P(\neg A, \neg B, C) + P(A, \neg B, \neg C) + P(\neg A, \neg B, \neg C)$$

$$= 0.15 + 0.35 + 0.05 + 0.05$$

$$= 0.6$$

$$\therefore P(\neg B) = 0.6$$

(c)

$$P(A | \neg B, C) = \frac{P(A, \neg B, C)}{P(\neg B, C)}$$

$$= \frac{P(A, \neg B, C)}{P(\neg A, \neg B, C) + P(A, \neg B, C)}$$

$$= \frac{0.15}{0.35 + 0.15}$$

$$= \frac{0.15}{0.5}$$

$$= 0.3$$

$$\therefore P(A | \neg B, C) = 0.3$$

$$\text{and, } P(\neg A | B, C) = \frac{P(\neg A, B, C)}{P(B, C)}$$

$$= \frac{P(\neg A, B, C)}{P(A, B, C) + P(\neg A, B, C)}$$

$$= \frac{0.05}{0.1 + 0.05}$$

$$= \frac{0.05}{0.15}$$

$$= 0.333$$

$$\therefore P(\neg A | B, C) = 0.333 \quad (\underline{\text{Ans.}})$$

(d)

$$P(A) = 0.5$$

$$\text{and, } P(C) = 1 - P(\neg C)$$

$$P(B) = 1 - P(\neg B)$$

$$= 1 - 0.25$$

$$= 0.75$$

$$= 0.4$$

$$= 0.75$$

Now,

$$P(A \cap B \cap C) = 0.2$$

and,

$$\begin{aligned} & P(A) \times P(B) \times P(C) \\ &= 0.5 \times 0.4 \times 0.75 \end{aligned}$$

$$= 0.15$$

Since,  $P(A \cap B \cap C) \neq P(A) \times P(B) \times P(C)$ ,

A, B, C are not independent. (Ans.)

Ans. to the que. no - 02

(a)

Let D = Disease

D' = No Disease

T = Tested Positive

T' = Tested Negative

Now,

$$P(D) = \frac{1}{1000} = 0.001$$

$$\begin{aligned} P(D') &= 1 - P(D) \\ &= 0.999 \end{aligned}$$

$$P(T|D') = \frac{1}{100} = 0.01$$

$$P(T'|D) = \frac{3}{100} = 0.03, \quad P(T|D) = 1 - 0.03 \\ = 0.97$$

Now, using bayes rule,

$$\begin{aligned} P(D|T) &= \frac{P(T|D) \cdot P(D)}{P(T)} \\ &= \frac{P(T|D) \cdot P(D)}{P(T|D) \cdot P(D) + P(T|D') \cdot P(D')} \\ &= \frac{0.97 \times 0.001}{(0.97 \times 0.001) + (0.01 \times 0.999)} \\ &= 0.0885 \end{aligned}$$

So, the probability of having that disease given that patient tested positive once is 0.0885.

(Ans.)

(b)

We got  $P(D|T) = 0.0885$ . We can use this as our new prior,

$$P_n(D) = 0.0885, P_n(D') = 1 - 0.0885 \\ = 0.9115$$

Again using Bayes rule,

$$\begin{aligned} P(D|T_1, T_2) &= \frac{P(T|D) \cdot P_n(D)}{P(T)} \\ &= \frac{P(+|D) \cdot P_n(D)}{P(+|D) \cdot P_n(D) + P(+|D') \cdot P_n(D')} \\ &= \frac{(0.97 \times 0.0885)}{(0.97 \times 0.0885) + (0.01 \times 0.9115)} \\ &= 0.904 \end{aligned}$$

$$\therefore P(D|T_1, T_2) = 0.904$$

So, the probability of the patient having that disease given that patient tested positive twice is 0.904.

(Ans.)

Ans. to the que. no-03

(a)

Given,

$$D_1 = \{(8, 1.2), (9, 1.3), (10, 1.5), (12, 1.8), (15, 2.3)\}$$

$$\eta = \frac{100}{100 * 10t}$$

$$\theta_i = i$$

for iteration 1,

$$\eta = 0.1$$

$$\theta_0 = 0$$

$$\theta_1 = 1$$

$$\begin{aligned} \text{So, } \theta_0 &= \theta_0 + \eta \sum_{i=1}^n (y^{(i)} - h_\theta(x^{(i)})) x_i^{(i)} \\ &= 0 + 0.1 \left[ \{(1.2-8) \cdot 1\} + \{(1.3-9) \cdot 9\} + \{(1.5-10) \cdot 10\} + \right. \\ &\quad \left. \{(1.8-12) \cdot 12\} + \{(2.3-15) \cdot 15\} \right] \\ &= 0.1 \times (-45.9) \\ &= -4.59 \end{aligned}$$

$$\begin{aligned}
 \theta_1 &= \theta_1 + \eta \sum_{i=1}^n (y^{(i)} - h_\theta(x^{(i)})) x_j^{(i)} \\
 &= 1 + 0.1 \left[ \{ (1.2-8) \cdot 8 \} + \{ (1.3-9) \cdot 9 \} + \{ (1.5-10) \cdot 10 \} + \{ (1.8-12) \cdot 12 \} \right. \\
 &\quad \left. + \{ (2.3-15) \cdot 15 \} \right] \\
 &= 1 + 0.1 \times (-521.6) \\
 &= -51.16
 \end{aligned}$$

For Iteration 2,

$$\eta = 0.05$$

$$\theta_0 = -4.59$$

$$\theta_1 = -51.16$$

$$\begin{aligned}
 \text{So, } \theta_0 &= \theta_0 + \eta \sum_{i=1}^n (y^{(i)} - h_\theta(x^{(i)})) \\
 &= -4.59 + 0.05 \left[ \{ 1.2 - (-4.59 + (-51.16 \times 8)) \} + \{ 1.3 - (-4.59 + (-51.16 \times 9)) \} + \right. \\
 &\quad \{ 1.5 - (-4.59 + (-51.16 \times 10)) \} + \{ 1.8 - (-4.59 + (-51.16 \times 12)) \} + \\
 &\quad \left. \{ 2.3 - (-4.59 + (-51.16 \times 15)) \} \right] \\
 &= -4.59 + 0.05 (415.07 + 466.33 + 517.69 + 620.31 + 774.29) \\
 &= 135.0945
 \end{aligned}$$

$$\theta_1 = \theta_1 + \eta \sum_{i=1}^n (y^{(i)} - h_\theta(x^{(i)})) x_j^{(i)}$$

$$= -51.16 + 0.05 \left[ \begin{array}{l} \{ 1.2 - (-4.59 + (-51.16 \times 8)) \} \cdot 8 + \\ \{ 1.3 - (-4.59 + (-51.16 \times 9)) \} \cdot 9 + \\ \{ 1.5 - (-4.59 + (-51.16 \times 10)) \} \cdot 10 + \\ \{ 1.8 - (-4.59 + (-51.16 \times 12)) \} \cdot 12 + \\ \{ 2.3 - (-4.59 + (-51.16 \times 15)) \} \cdot 15 \end{array} \right]$$

$$= -51.16 + 0.05 (3320.56 + 4196.97 + 5176.9 + 7443.72 + 11614.35)$$

$$= 1536.465$$

(Ans.)

(b)

Given,

$$D_2 = \{ (10, 1), (13, 1), (5, 0), (15, 1), (7, 0) \}$$

For iteration 1,

$$\theta_0 = 0$$

$$\theta_1 = 1$$

$$\eta = 0.1$$

$$\theta_0 = \theta_0 + \eta \sum_{i=1}^n \{y^{(i)} - h_\theta(x^{(i)})\}$$

$$= 0 + 0.1 \left[ \{1 - g(10)\} + \{1 - g(13)\} + \{0 - g(5)\} + \{1 - g(15)\} + \{0 - g(7)\} \right]$$

$$= 0.1 \times \left[ (4.54 \times 10^{-5}) + (2.26 \times 10^{-6}) + (-0.9933) + (3.06 \times 10^{-7}) + (-0.999) \right]$$

$$= -0.19922$$

$$\theta_1 = \theta_1 + \eta \sum_{i=1}^n \{y^{(i)} - h_\theta(x^{(i)})\} x_j^{(i)}$$

$$= 1 + 0.1 \left[ \{1 - g(10)\} 10 + \{1 - g(13)\} 13 + \{0 - g(5)\} 5 + \{1 - g(15)\} 15 + \{0 - g(7)\} 7 \right]$$

$$= -0.196$$

For iteration 2,

$$\theta_0 = -0.19922$$

$$\theta_1 = -0.196$$

$$\eta = 0.05$$

$$\begin{aligned}
 \theta_0 &= \theta_0 + \eta \sum_{i=1}^n \{y^{(i)} - h_\theta(x^{(i)})\} \\
 &= -0.19922 + 0.05 \times \left[ \begin{array}{l} \{1 - g(-0.19922 - 0.196 \times 10)\} + \\ \{1 - g(-0.19922 - 0.196 \times 13)\} + \\ \{0 - g(-0.19922 - 0.196 \times 5)\} + \\ \{1 - g(-0.19922 - 0.196 \times 15)\} + \\ \{0 - g(-0.19922 - 0.196 \times 7)\} \end{array} \right] \\
 &= -0.19922 + 0.1194 \\
 &= -0.0798
 \end{aligned}$$

$$\begin{aligned}
 \theta_1 &= \theta_1 + \eta \sum_{i=1}^n \{y^{(i)} - h_\theta(x^{(i)})\} x_i^{(i)} \\
 &= -0.196 + 0.05 \times \left[ \begin{array}{l} \{1 - g(-0.19922 - 0.196 \times 10)\} 10 + \\ \{1 - g(-0.19922 - 0.196 \times 13)\} 13 + \\ \{0 - g(-0.19922 - 0.196 \times 5)\} 5 + \\ \{1 - g(-0.19922 - 0.196 \times 15)\} 15 + \\ \{0 - g(-0.19922 - 0.196 \times 7)\} 7 \end{array} \right] \\
 &= -0.196 + 0.05 \times (33.179) \\
 &= -0.196 + 1.659 \\
 &= 1.463
 \end{aligned}$$

(Ans)

(c)

Let  $X$  be the matrix and  $y$  be the vector. So our loss function (least squared loss function) that is,

$$J(\theta) = \frac{1}{2} \sum_{i=1}^n (\bar{y}^{(i)} - y^{(i)})^2$$

the matrix form for  $J(\theta)$ ,

$$\begin{aligned} J(\theta) &= (X\theta - y)^T (X\theta - y) \\ &= (\theta^T X^T - y^T) (X\theta - y) \\ &= (\theta^T X^T X \theta - \theta^T X^T y - y^T X \theta + y^T y) \\ &= \theta^T X^T X \theta - 2\theta^T X^T y + y^T y \end{aligned}$$

Now,

$$\begin{aligned} \frac{\partial}{\partial \theta} J(\theta) &= \frac{\partial}{\partial \theta} (\theta^T X^T X \theta - 2\theta^T X^T y + y^T y) \\ &= \frac{\partial}{\partial \theta} (\theta^T X^T X \theta) - 2 \frac{\partial}{\partial \theta} (\theta^T X^T y) + \frac{\partial}{\partial \theta} (y^T y) \\ &= 2X^T X \theta - 2X^T y \end{aligned}$$

Now, to minimize, the gradient must be 0,

$$2x^T \phi - 2x^T y = 0$$

$$\Rightarrow x^T x \phi = x^T y$$

$$\Rightarrow (x^T x) \phi = x^T y$$

$$\Rightarrow \phi = (x^T x)^{-1} x^T y$$

So, the closed-form solution for linear regression,

$$\phi = (x^T x)^{-1} \cdot x^T y$$

(Ans.)