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Ans to the ques no:- 01

[a]

For conditional independence, $P(\text{cheat} \cap \text{pass} | \text{study})$ should be equal to $P(\text{cheat} | \text{study}) \cdot P(\text{pass} | \text{study})$.

Hence, $P(\text{cheat} \cap \text{pass} | \text{study}) = \frac{P(\text{cheat} \cap \text{pass} \cap \text{study})}{P(\text{study})}$

$$P(\text{cheat} \cap \text{pass} \cap \text{study}) = (0.25)(0.15) + (0.25)(0.02) = \frac{0.25}{0.25 + 0.15 + 0.02 + 0.03}$$

$$= \frac{0.25(0.15 + 0.02)}{0.25 + 0.15 + 0.02 + 0.03} = \frac{5}{17} \approx 0.28$$

Now, $P(\text{cheat} | \text{study}) \cdot P(\text{Pass} | \text{study}) = \frac{P(\text{study} | \text{cheat}) P(\text{cheat})}{P(\text{study})} \cdot \frac{P(\text{study} | \text{pass}) \cdot P(\text{pass})}{P(\text{study})}$

$$= \frac{(0.25)(0.15 + 0.02)(0.25 + 0.15 + 0.02 + 0.03)}{0.25 + 0.15 + 0.02 + 0.03}$$

$$\times \frac{(0.25 + 0.15)(0.25 + 0.15 + 0.02 + 0.03)}{0.25 + 0.15 + 0.02 + 0.03}$$

$$\approx 0.198$$

$$\text{Ans. } P(\text{cheat} \cap \text{pass} | \text{study}) \neq P(\text{cheat} | \text{study}) \cdot P(\text{pass} | \text{study})$$

Therefore, cheat and pass are not conditionally independent given study.

Ques. Find $P(\text{pass} \cup \text{cheat})$ if the following is known:

[b]

$$P(\text{pass} \cup \text{cheat}) = P(\text{pass}) + P(\text{cheat}) - P(\text{pass} \cap \text{cheat})$$

$$= (0.25 + 0.15 + 0.10 + 0.13) + (0.25 + 0.02 + 0.10 + 0.22) - (0.25 + 0.10)$$

$$= 0.87$$

(Ans)

Ans to the Ques no:- 2

a

$$P(\neg \text{cold}) = 0.12 + 0.04 + 0.10 + 0.07 \\ = 0.33$$

(Ans)

b

$$P(\neg \text{cloudy} \mid \neg \text{Rain} \cap \neg \text{cold}) = \frac{P(\neg \text{cloudy} \cap \neg \text{Rain} \cap \neg \text{cold})}{P(\neg \text{Rain} \cap \neg \text{cold})} \\ = \frac{0.07}{0.04 + 0.07} \\ = \frac{7}{11} \approx 0.636$$

(Ans)

c

$$P(\neg \text{Rain} \mid \neg \text{cloudy}) = \frac{P(\neg \text{Rain} \cap \neg \text{cloudy})}{P(\neg \text{cloudy})} \\ = \frac{0.03 + 0.07}{0.26 + 0.03 + 0.10 + 0.07}$$

$$P(\text{Rain}) = \frac{5}{23} \approx 0.217$$

(Ans)



Ques no 2. If P(A) = 0.32, P(B) = 0.12, P(C) = 0.06, P(D) = 0.04, P(A ∩ B) = 0.03, P(A ∩ C) = 0.07, P(B ∩ C) = 0.08, P(B ∩ D) = 0.07, P(C ∩ D) = 0.09.

[d]

Ans :-

$$P(\neg \text{Rain} \cup \text{cloudy}) = P(\neg \text{Rain}) + P(\text{cloudy}) - P(\neg \text{Rain} \cap \text{cloudy})$$

$$= (0.06 + 0.04 + 0.03 + 0.07) + (0.32 + 0.12 + 0.06 + 0.07) \\ - (0.06 + 0.04)$$

$$= 0.64 \quad \underline{\text{(Ans)}}$$

base — o — x — o —

$\Rightarrow 0.64 \times 100\%$

Ans to the ques no:- 3

[a]

$$P(\text{Football} \cap \text{Left-handed}) = 0.15$$

$$P(\text{Football} | \text{Left-handed}) = \frac{P(\text{Football} \cap \text{Left-handed})}{P(\text{Left-handed})}$$

$$= \frac{0.15}{0.24 + 0.15 + 0.15}$$

$$P(\text{Right-handed} \cap \text{Cricket}) = \frac{5}{18} \approx 0.278$$

(Ans)

b

$$P(\text{Right-handed} \mid \text{Cricket}) = \frac{P(\text{Right-handed} \cap \text{Cricket})}{P(\text{Cricket})}$$

$$\text{Cricket} = P(\text{Football} \cup \text{Cricket}) = 0.1 + 0.24 = 0.34$$

$$= \frac{5}{17} \approx 0.294$$

(Ans)

c

~~$$P(\text{Football} \cup \text{Cricket}) = P(\text{Football}) + P(\text{Cricket}) - P(\text{Football} \cap \text{Cricket})$$

$$= (0.15 + 0.1) + (0.24 + 0.1) - 0$$

$$= 0.59$$~~

(Ans)

$$P(\text{Football} \cap \text{Cricket}) = P(\text{Football}) + P(\text{Cricket}) - P(\text{Football} \cup \text{Cricket})$$

$$= (0.15 + 0.1) + (0.24 + 0.1) - (0.24 + 0.1 + 0.15 + 0.1)$$

$$= 0 \quad (\text{Ans})$$

$P(\text{Right-handed} \cup \text{Left-handed})$

$$= P(\text{Right-handed}) + P(\text{Left-handed}) - P(\text{Right-handed} \cap \text{Left-handed})$$

$$= (0.1 + 0.1 + 0.26) + (0.24 + 0.15 + 0.15) - 0$$

$$= 1 \quad (\text{Ans})$$

Here, $P(\text{Football} \cap \text{Right-handed}) = 0.1$

$$\text{And, } P(\text{Football}) \cdot P(\text{Right-handed}) = (0.15 + 0.1) \times (0.1 + 0.1 + 0.26)$$

$$= 0.115$$

As, $P(\text{Football} \cap \text{Right-handed}) \neq P(\text{Football}) \cdot P(\text{Right-handed})$

Therefore, playing football ~~does not~~ depends on being right-handed.



Ans to the ques no:- 4

Let,

Machine detects lie = D

Person tells lie = P

Hence,

$$P(D|P) = 0.96 \quad \therefore P(\neg D|P) = \frac{1-0.96}{= 0.04}$$

$$P(\neg D|\neg P) = 0.95 \quad \therefore P(D|\neg P) = \frac{1-0.95}{= 0.05}$$

$$P(P) = 0.02 \quad \therefore P(\neg P) = \frac{1-0.02}{= 0.98}$$

Event of machine = D

[a]

$$P(P|D) = \frac{P(D|P) \cdot P(P)}{P(D)} = \frac{0.96 \times 0.02}{P(D)} = \frac{0.0192}{P(D)}$$

$$P(\neg P|D) = \frac{P(D|\neg P) \cdot P(\neg P)}{P(D)} = \frac{0.05 \times 0.98}{P(D)} = \frac{0.049}{P(D)}$$

$$\therefore \frac{0.0192}{P(D)} < \frac{0.049}{P(D)}$$

So, it is more likely that the person is not a liar.

b

$$\text{Hence, } P(\neg P | D) = \frac{P(D | \neg P) \cdot P(\neg P)}{P(D)}$$

$$= \frac{0.05 \times 0.98}{P(D \cap P) + P(D \cap \neg P)}$$

$$= \frac{0.05 \times 0.98}{P(D|P) \cdot P(P) + P(D|\neg P) \cdot P(\neg P)}$$

$$= \frac{0.05 \times 0.98}{0.96 \times 0.02 + 0.05 \times 0.98}$$

$$\approx 0.718$$

(Ans)

Ans to the question no:- 5

Bayes theorem is complicated when the dataset is quite large and we have to find joint probability under some condition. Bayes theorem's this issue is resolved by Naive Bayes by introducing conditional independence among the independent features given the output variable.

For example, let's consider the dataset below:-

Outlook	Wind	Play
Sunny	False	No
Sunny	True	No
Rainy	False	Yes
Sunny	True	Yes

Here, outlook and wind are features and play is output variable.

Now, if our query is $P(\text{Sunny} \cap \text{False} | \text{Yes})$, that is probability of $\text{play} = \text{Yes}$ ~~outlook = sunny~~ and ~~wind = False~~ when $P(\text{play} = \text{Yes})$.

Following Bayes theorem, $P(\text{sunny} \cap \text{False})$

Now, we want to find the probability of
play = Yes when outlook = sunny and
wind = False, that is, $P(\text{Yes} | \text{sunny} \cap \text{False})$.

Applying Naive Bayes, $P(\text{Yes} | \text{sunny} \cap \text{False})$

$$\frac{P(\text{Yes}) \cdot P(\text{sunny} \cap \text{False} | \text{Yes})}{P(\text{sunny} \cap \text{False})}$$

Hence, we come across 0 probability problem,
because, $P(\text{sunny} \cap \text{False} | \text{Yes}) = 0$.

This problem is solved by Naive Bayes by
introducing conditional probability.

$$\therefore P(\text{Yes} | \text{sunny} \cap \text{False}) = \frac{P(\text{sunny} | \text{Yes}) \cdot P(\text{False} | \text{Yes}) \cdot P(\text{Yes})}{P(\text{sunny} \cap \text{False})}$$

$$= \frac{\frac{1}{2} \cdot \frac{1}{2} \cdot \frac{1}{4}}{\frac{1}{4}} = \frac{1}{4}$$

$$= \frac{1}{8} \times \frac{4}{1} = \frac{1}{2}$$

So, we get a value for playing in this

process which solves the problem of Bayes theorem.

probabilistic $\xrightarrow{\text{probabilistic}}$ $\xrightarrow{\text{probabilistic}}$ $\xrightarrow{\text{probabilistic}}$ $\xrightarrow{\text{probabilistic}}$

for each condition results in probability

(Sunny, Mild, Normal, Windy) \rightarrow $\xrightarrow{\text{probabilistic}}$

Ans to the ques. no. 6

[a]

$$P(\text{Yes} | \text{sunny} \wedge \text{mild} \wedge \text{Normal} \wedge \text{windy})$$

$$= \frac{P(\text{sunny} \wedge \text{mild} \wedge \text{Normal} \wedge \text{windy} | \text{Yes}) \cdot P(\text{Yes})}{P(\text{sunny} \wedge \text{mild} \wedge \text{Normal} \wedge \text{windy})}$$

$$= \frac{P(\text{sunny} | \text{Yes}) \cdot P(\text{mild} | \text{Yes}) \cdot P(\text{Normal} | \text{Yes}) \cdot P(\text{windy} | \text{Yes}) \cdot P(\text{Yes})}{P(\text{sunny} \wedge \text{mild} \wedge \text{Normal} \wedge \text{windy})}$$

$$= \frac{P(\text{sunny} | \text{Yes}) \cdot P(\text{mild} | \text{Yes}) \cdot P(\text{Normal} | \text{Yes}) \cdot P(\text{windy} | \text{Yes}) \cdot P(\text{Yes})}{P(\text{sunny} \wedge \text{mild} \wedge \text{Normal} \wedge \text{windy})}$$

Similarly, $P(\text{No} | \text{sunny} \wedge \text{mild} \wedge \text{Normal} \wedge \text{windy})$

$$= \frac{P(\text{sunny} | \text{No}) \cdot P(\text{mild} | \text{No}) \cdot P(\text{Normal} | \text{No}) \cdot P(\text{windy} | \text{No}) \cdot P(\text{No})}{P(\text{sunny} \wedge \text{mild} \wedge \text{Normal} \wedge \text{windy})}$$

Now, the learning phase! -

Outlook	Play Tennis = Yes	Play Tennis = No
Overcast	$\frac{2}{5}$	$\frac{0}{3}$
Sunny	$\frac{2}{5}$	$\frac{3}{3}$
Rainy	$\frac{1}{5}$	$\frac{0}{3}$

Humidity	Yes	No
Cool	$\frac{2}{5}$	$\frac{0}{3}$
Mild	$\frac{3}{5}$	$\frac{2}{3}$
Hot	$\frac{0}{5}$	$\frac{1}{3}$

Temp	Yes	No
Normal	$\frac{4}{5}$	$\frac{0}{3}$
High	$\frac{1}{5}$	$\frac{3}{3}$

Wind	Yes	No
True	$\frac{3}{5}$	$\frac{1}{3}$
False	$\frac{2}{5}$	$\frac{2}{3}$

$$\therefore P(\text{Yes} \mid \text{sunny} \cap \text{mild} \cap \text{Normal} \cap \text{windy})$$

$$= \frac{\frac{2}{5} \cdot \frac{3}{5} \cdot \frac{4}{5} \cdot \frac{3}{5} \cdot \frac{5}{8}}{P(\text{sunny} \cap \text{mild} \cap \text{Normal} \cap \text{windy})}$$

$$= \frac{0.072}{P(\text{sunny} \cap \text{mild} \cap \text{Normal} \cap \text{windy})}$$

$$\text{And, } P(\text{No} \mid \text{sunny} \cap \text{mild} \cap \text{Normal} \cap \text{windy})$$

$$= \frac{\frac{3}{3} \cdot \frac{2}{3} \cdot \frac{0}{3} \cdot \frac{1}{3} \cdot \frac{3}{8}}{P(\text{sunny} \cap \text{mild} \cap \text{Normal} \cap \text{windy})}$$

$$= \frac{0}{P(\text{sunny} \cap \text{mild} \cap \text{Normal} \cap \text{windy})}$$

$$\therefore P(\text{Yes} \mid \text{Sunny} \cap \text{Mild} \cap \text{Normal} \cap \text{Windy}) > P(\text{No} \mid \text{Sunny} \cap \text{Mild} \cap \text{Normal} \cap \text{Windy})$$

So, it is more likely that the player is going to play tennis.

Tb

$$P(\text{Yes} \mid \text{Overcast} \cap \text{Hot} \cap \text{Normal} \cap \text{windy})$$

$$= \frac{P(\text{Overcast} \mid \text{Yes}) \cdot P(\text{Hot} \mid \text{Yes}) \cdot P(\text{Normal} \mid \text{Yes}) \cdot P(\text{windy} \mid \text{Yes}) \cdot P(\text{Yes})}{P(\text{Overcast} \cap \text{Hot} \cap \text{Normal} \cap \text{windy})}$$

$$= \frac{\frac{2}{5} \cdot \frac{0}{5} \cdot \frac{4}{5} \cdot \frac{3}{5} \cdot \frac{5}{8}}{P(\text{Overcast} \cap \text{Hot} \cap \text{Normal} \cap \text{windy})}$$

$$= \frac{0}{P(\text{Overcast} \cap \text{Hot} \cap \text{Normal} \cap \text{windy})}$$

Similarly,

$$P(\text{No} \mid \text{Overcast} \cap \text{Hot} \cap \text{Normal} \cap \text{windy})$$

$$\begin{aligned}
 &= \frac{P(\text{Overcast} | \text{No}) \cdot P(\text{Hot} | \text{No}) \cdot P(\text{Normal} | \text{No}) \cdot P(\text{Windy} | \text{No}) \cdot P(\text{No})}{P(\text{Overcast} \cap \text{Hot} \cap \text{Normal} \cap \text{Windy})} \\
 &= \frac{\frac{0}{3} \cdot \frac{1}{3} \cdot \frac{0}{3} \cdot \frac{1}{3} \cdot \frac{3}{8}}{P(\text{Overcast} \cap \text{Hot} \cap \text{Normal} \cap \text{Windy})} \\
 &= \frac{0}{P(\text{Overcast} \cap \text{Hot} \cap \text{Normal} \cap \text{Windy})}
 \end{aligned}$$

$$\therefore P(\text{Yes} | \text{Overcast} \cap \text{Hot} \cap \text{Normal} \cap \text{Windy}) = P(\text{No} | \text{Overcast} \cap \text{Hot} \cap \text{Normal} \cap \text{Windy})$$

Therefore the result is inconclusive, as both are same.

$$\begin{array}{c}
 \text{Probability of } (\text{Overcast} \cap \text{Hot} \cap \text{Normal} \cap \text{Windy}) \\
 \hline
 0 \quad \times \quad 0
 \end{array}$$

$$\begin{array}{c}
 \text{Probability of } (\text{Overcast} \cap \text{Hot} \cap \text{Normal} \cap \text{Windy}) \\
 \hline
 0 \quad \times \quad 0
 \end{array}$$

$$\begin{array}{c}
 \text{Probability of } (\text{Overcast} \cap \text{Hot} \cap \text{Normal} \cap \text{Windy}) \\
 \hline
 0 \quad \times \quad 0
 \end{array}$$

Ans to the ques no. 7 :-

Ques 7) Ans to the ques no. 7 :-

[a]

$$E(\text{Edible}) = -\frac{9}{16} \log_2 \frac{9}{16} - \frac{7}{16} \log_2 \frac{7}{16}$$

$$= 0.987$$

$$= 0.989$$

(Ans)

[b]

For color :-

$$E(\text{Edible} | \text{color} = \text{Yellow}) = -\frac{8}{13} \log_2 \frac{8}{13} - \frac{5}{13} \log_2 \frac{5}{13}$$

$$= 0.961$$

$$E(\text{Edible} | \text{color} = \text{Green}) = -\frac{1}{3} \log_2 \frac{1}{3} - \frac{2}{3} \log_2 \frac{2}{3}$$

$$= 0.918$$

$$\therefore I_G(\text{Edible} | \text{color}) = 0.989 - \left(\frac{13}{16} \times 0.961 + \frac{3}{16} \times 0.918 \right)$$

$$= 0.036$$

For Size:-

$$E(\text{Edible} \mid \text{Size} = \text{Small}) = -\frac{6}{8} \log_2 \frac{6}{8} - \frac{2}{8} \log_2 \frac{2}{8}$$
$$= 0.811$$

$$E(\text{Edible} \mid \text{Size} = \text{Large}) = -\frac{3}{8} \log_2 \frac{3}{8} - \frac{5}{8} \log_2 \frac{5}{8}$$
$$= 0.954$$

$$\therefore \text{IG}(\text{Edible} \mid \text{size}) = 0.989 - \left(\frac{8}{16} \times 0.811 + \frac{8}{16} \times 0.954 \right)$$
$$= 0.1065$$

For Shape:-

$$E(\text{Edible} \mid \text{Shape} = \text{Round}) = -\frac{6}{12} \log_2 \frac{6}{12} - \frac{6}{12} \log_2 \frac{6}{12}$$
$$= 1$$

$$E(\text{Edible} \mid \text{Shape} = \text{Irregular}) = -\frac{3}{4} \log_2 \frac{3}{4} - \frac{1}{4} \log_2 \frac{1}{4}$$
$$= 0.811$$

$$\therefore \text{IG}(\text{Edible} \mid \text{shape}) = 0.989 - \left(\frac{12}{16} \times 1 + \frac{4}{16} \times 0.811 \right)$$
$$= 0.03625$$

As the information gain of 'size' is largest,
it is the better feature.

