

2.4 Conditional probability :

For any two events A and B with $P(B) > 0$, the conditional probability of A given that B has already occurred is defined by $P(A|B) = \frac{P(A \cap B)}{P(B)}$.

Similarly, $P(B|A) = \frac{P(A \cap B)}{P(A)}$ - $P(A) > 0$.

Multiplication Rule for $P(A \cap B)$

$$P(A \cap B) = P(A|B) \cdot P(B) = P(B|A) \cdot P(A)$$

* For 3 events show that $P(A \cap B \cap C) = P(A) P(B|A) P(C|A \cap B)$

$$\begin{aligned} \text{Proof: R.H.S} &= P(A) P(B|A) P(C|A \cap B) \\ &= P(A) \frac{P(A \cap B)}{P(A)} \frac{P(C \cap A \cap B)}{P(A \cap B)} \\ &= P(A \cap B \cap C) = \text{L.H.S} \quad (\text{proved}) \end{aligned}$$

* For n events $A_1, A_2, A_3, \dots, A_n$

$$P(A_1 \cap A_2 \cap \dots \cap A_n) = P(A_1) P(A_2|A_1) P(A_3|A_1 \cap A_2) P(A_4|A_1 \cap A_2 \cap A_3) \dots P(A_n|A_1 \cap A_2 \cap \dots \cap A_{n-1})$$

$$\begin{aligned} \text{Proof: R.H.S} &= P(A_1) P(A_2|A_1) P(A_3|A_1 \cap A_2) \dots P(A_n|A_1 \cap A_2 \cap \dots \cap A_{n-1}) \\ &= P(A_1) \frac{P(A_1 \cap A_2)}{P(A_1)} \frac{P(A_1 \cap A_2 \cap A_3)}{P(A_1 \cap A_2)} \dots \frac{P(A_1 \cap A_2 \cap \dots \cap A_n)}{P(A_1 \cap A_2 \cap \dots \cap A_{n-1})} \\ &= P(A_1 \cap A_2 \cap \dots \cap A_n) \quad (\text{proved}) \end{aligned}$$

Example:

Suppose that of all individuals buying a certain digital camera, 60% include an optional memory card in their purchase, 40% include an extra battery and 30% include both a card and battery. Consider selecting a buyer and let $A = \{\text{memory card purchased}\}$ and $B = \{\text{Battery purchased}\}$. Then $P(A) = 0.60$, $P(B) = 0.40$, and $P(A \cap B) = 0.30$.

then $P(A|B)$ = the probability that an optional card was also purchased, Given that the selected individual purchased an extra battery

$$= \frac{P(A \cap B)}{P(B)} = \frac{0.30}{0.40} = 0.75$$

similarly $P(\text{battery} / \text{memory card})$

$$= P(B|A) = \frac{P(A \cap B)}{P(A)} = \frac{0.30}{0.60} = 0.50$$

Total probability and Bayes' theorem

Recall that events A_1, A_2, \dots, A_K are mutually exclusive if no two have any common outcomes i.e. if $A_i \cap A_j = \emptyset$, $i \neq j$.

The events A_1, A_2, \dots, A_K are exhaustive if one A_i must occur, so that $A_1 \cup A_2 \cup \dots \cup A_K = S$.

The Law of Total probability

Let A_1, A_2, \dots, A_K be mutually exclusive and exhaustive events. Then for any other event B ,

$$\begin{aligned} P(B) &= P(B|A_1)P(A_1) + P(B|A_2)P(A_2) + \dots + P(B|A_K)P(A_K) \\ &= \sum_{i=1}^K P(B|A_i)P(A_i) \end{aligned}$$

Proof: Since A_1, A_2, \dots, A_K are exhaustive, we have

$$\bigcup_{i=1}^K A_i = S$$

Now, $B = B \cap S$ ($\because B \subseteq S$)

$$= B \cap \bigcup_{i=1}^K A_i$$

$$= B \cap [A_1 \cup A_2 \cup A_3 \cup \dots \cup A_K]$$

$$= (B \cap A_1) \cup (B \cap A_2) \cup (B \cap A_3) \cup \dots \cup (B \cap A_K)$$

$$\begin{aligned} \Rightarrow P(B) &= P(B \cap A_1) + P(B \cap A_2) + \dots + P(B \cap A_K) \quad (\because \text{the events are mutually exclusive}) \\ &= P(A_1)P(B|A_1) + P(A_2)P(B|A_2) + \dots + P(A_K)P(B|A_K) \\ &= \sum_{i=1}^K P(A_i)P(B|A_i) \quad (\text{proved}) \end{aligned}$$

Bayes' theorem;

Let A_1, A_2, \dots, A_K be a collection of K mutually exclusive and ~~each~~ exhaustive events with prior probabilities $P(A_i)$ ($i=1, 2, \dots, K$). Then for any other event B for which $P(B) > 0$, the probability of A_j given that B has occurred is

$$P(A_j|B) = \frac{P(A_j \cap B)}{P(B)}$$

$$= \frac{P(B|A_j)P(A_j)}{\sum_{i=1}^K P(B|A_i)P(A_i)}$$

[using multiplication rule, $j=1, 2, \dots, K$]
[using total probability]

Example (Incidence of rare disease)

only 1 in 1000 adults is afflicted with a rare disease for which a diagnostic test has been developed. The test is such that when an individual actually has the disease, a positive result will occur 99% of the time, whereas an individual without the disease will show a positive test result only 2% of the time. If a randomly selected individual is tested and the result is positive, what is the probability that the individual has the disease?

Ans: Let A_1 = individual has the disease
 A_2 = individual does not have the disease
 B = positive test result.

then,

then, $P(A_1) = \frac{1}{1000} = 0.001$, $P(A_2) = \frac{999}{1000} = 0.999$

$P(B|A_1) = \frac{99}{100} = 0.99$, $P(B|A_2) = \frac{2}{100} = 0.02$.

\therefore The probability that the individual has the disease, given that selected individual is tested positive is

$$P(A_1|B) = \frac{P(B|A_1) P(A_1)}{\sum_{i=1}^2 P(B|A_i) P(A_i)} \quad (\text{By Baye's theorem})$$

$$= \frac{P(B|A_1) P(A_1)}{P(B|A_1) P(A_1) + P(B|A_2) P(A_2)}$$

$$= \frac{0.99 \times 0.001}{0.99 \times 0.001 + 0.02 \times 0.999}$$

$$= \frac{0.00099}{0.02097}$$

$$= 0.047$$

Ans