

Bay's theorem:-

Statement:-

Suppose E_1, E_2, \dots, E_n are mutually exclusive events of a sample space $S \ni P(E_i) > 0$ for $i = 1, 2, \dots, n$ and A is any arbitrary event of $S \ni P(A) > 0$ and $A \subseteq \bigcup_{i=1}^n E_i$.
Then Conditional probability of E_i given A is

$$P(E_i | A) = \frac{P(A | E_i) \cdot P(E_i)}{\sum_{i=1}^n P(A | E_i) \cdot P(E_i)}$$

Proof:-

Given $A \subseteq \bigcup_{i=1}^n E_i$

($A \subseteq B$ then $A \cap B = A$)

$$A = A \cap \bigcup_{i=1}^n E_i$$

$$A = \bigcup_{i=1}^n (A \cap E_i)$$

$$\begin{aligned} & A \cap (E_1 \cup E_2 \cup \dots \cup E_n) \\ &= (A \cap E_1) \cup (A \cap E_2) \cup \dots \cup (A \cap E_n) \end{aligned}$$

E_1, E_2, \dots, E_n are mutually exclusive
then $A \cap E_1, A \cap E_2, \dots, A \cap E_n$ are also mutually ex.

$$(A \cap E_i) \cap (A \cap E_j) = \emptyset \text{ for } i \neq j$$

$$P(A) = P\left[\bigcup_{i=1}^n (A \cap E_i)\right]$$

$$= P\left[(A \cap E_1) \cup (A \cap E_2) \cup \dots \cup (A \cap E_n)\right]$$

$$= P(A \cap E_1) + P(A \cap E_2) + \dots + P(A \cap E_n)$$

$$P(A) = \sum_{i=1}^n P(A \cap E_i)$$

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

$$P(A) = \sum_{i=1}^n P(A|E_i) \cdot P(E_i)$$

$$P(E_i|A) = \frac{P(A \cap E_i)}{P(A)} = \frac{P(A|E_i) P(E_i)}{\sum_{i=1}^n P(A|E_i) \cdot P(E_i)}$$

Bayes' Theorem