

Theory of Probability

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Two events $E, F \subset S$, then the conditional probability of E given F is

$$P(E|F) = \frac{P(EF)}{P(F)}$$

Fix the event F , and consider the function
 $Q(E) = P(E|F)$.

It turns out that $Q(\cdot) = P(\cdot|F)$ satisfies the three axioms of probability:

① $0 \leq Q(E) \leq 1$

PF: $0 \leq P(E|F) \leq 1$

$$0 \leq \frac{P(EF)}{P(F)} \leq 1$$

$$\Rightarrow 0 \leq P(EF) \leq P(F)$$

satisfied because $EF \subseteq F$

② $P(S|F) = 1 = Q(S)$

$$P(S|F) = \frac{P(SF)}{P(F)} = \frac{P(F)}{P(F)} = 1$$

③ Let E_i be mutually exclusive events: The events $E_i F$ are mutually excl.

$$Q\left(\bigcup_i E_i\right) = \sum_i Q(E_i)$$

$$P\left(\bigcup_i E_i | F\right) = \frac{P\left(\bigcup_i E_i \cap F\right)}{P(F)} = \frac{P\left(\bigcup_i E_i F\right)}{P(F)} = \sum_i \frac{P(E_i F)}{P(F)} = Q(E_i) \quad \square$$

$Q(E) = P(E|F)$ define a new probability function.

Consider conditional probabilities under Q :

$$\begin{aligned} Q(E_1 | E_2) &= \frac{Q(E_1, E_2)}{Q(E_2)} = \frac{P(E_1, E_2 | F)}{P(E_2 | F)} \\ &= \frac{P(E_1, E_2 F)}{P(F)} \cdot \frac{P(F)}{P(E_2 F)} = \frac{P(E_1, E_2 F)}{P(E_2 F)} \end{aligned}$$

Conditional Independence

Definition: E_1 and E_2 are conditionally independent with respect to F if:

$$P(E_1 | E_2 F) = P(E_1 | F). \quad \text{Very similar.}$$

Independence means : $P(A|B) = P(A)$.

An equivalent definition of conditional independence is:

$$P(E_1, E_2 | F) = P(E_1 | F) P(E_2 | F)$$

(compare with $P(AB) = P(A) P(B)$ if A, B are independent).

PF of equivalence:

$$\begin{aligned} P(E_1 | E_2 F) &= \frac{P(E_1, E_2 F)}{P(E_2 F)} \\ &= \frac{P(E_1, E_2 | F) P(F)}{P(E_2 F)} \end{aligned}$$

Since $P(E_1 | E_2 F) = P(E_1 | F)$, we have that

$$P(E_1 | F) = \frac{P(E_1, E_2 | F) P(F)}{P(E_2 F)}$$

$$P(E_1, E_2 | F) = P(E_1 | F) \underbrace{\frac{P(E_2 | F)}{P(F)}}_{= P(E_2 | F)}$$

$$\underline{P(E_1, E_2 | F) = P(E_1 | F) P(E_2 | F)}$$