

Conditional Probability

For events A, B , the probability of A , given B has occurred
(denoted as $P(A|B) = \frac{P(A \cap B)}{P(B)}$)
* $P(B)$ must be > 0

Independence

If there are 2 events, A, B , they are (pairwise) independent when
 $P(A \cap B) = P(A)P(B)$

If there are $n \geq 3$ events A_1, \dots, A_n , they are mutually independent if for any arbitrary subcollection of the A_i 's
(A_{i_1}, \dots, A_{i_m})

$$P(A_{i_1} \cap \dots \cap A_{i_m}) = P(A_{i_1}) \dots P(A_{i_m})$$

Bayes' Rule

For events, A, B ,

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

Conditional Independence

Events A, B , are conditionally independent given C , when

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

* implies $P(A|B, C) = P(A|C)$
 $P(B|A, C) = P(B|C)$

REMEMBER: Pairwise independence DOES NOT imply mutual independence

$$\left(\begin{array}{l} P(A \cap B) = P(A)P(B) \\ P(B \cap C) = P(B)P(C) \\ P(A \cap C) = P(A)P(C) \end{array} \right) \not\Rightarrow P(A \cap B \cap C) = P(A)P(B)P(C)$$

* One of the most useful rules for calculating an unknown, when you are given conditional probabilities.

REMEMBER: Conditional independence DOES NOT imply

independence (unless A, B, C are mutually independent)

$$(P(A \cap B|C) = P(A|C)P(B|C)) \not\Rightarrow P(A \cap B) = P(A)P(B)$$

Useful Things to Remember

Law of Total Probability

For partition of $S: A_1, A_2, \dots, A_n$ and event B .

$$P(B) = P(B \cap A_1) + \dots + P(B \cap A_n)$$

From this, we have

$$P(B) = P(B|A_1)P(A_1) + \dots + P(B|A_n)P(A_n)$$

* Very useful when using Bayes' Rule

$$\begin{aligned} P(A|B) &= \frac{P(B|A)P(A)}{P(B)} \\ &= \frac{P(B|A)P(A)}{P(B|A)P(A) + P(B|A^c)P(A^c)} \end{aligned}$$

Multiplication Rule

For events A_1, A_2, \dots, A_n

$$\begin{aligned} P(A_1 \cap A_2 \cap \dots \cap A_n) &= P(A_1) \times \\ &\quad P(A_2|A_1) \times \\ &\quad P(A_3|A_1, A_2) \times \dots \\ &\quad P(A_n|A_1, A_2, \dots, A_{n-1}) \end{aligned}$$

$$5.i) P(A_1 \cap A_2) \neq P(A_1)P(A_2)$$

We want to show this B - the event of biased die

$$\begin{aligned} P(A_1) &= P(A_1|B)P(B) + P(A_1|B^c)P(B^c) \\ &= \frac{2}{6} \cdot \frac{1}{2} + \frac{1}{6} \cdot \frac{1}{2} \\ &= \frac{1}{4} \end{aligned}$$

$$\text{Similarly, } P(A_2) = \frac{1}{4}$$

$$\begin{aligned} P(A_1 \cap A_2) &= P(A_1 \cap A_2|B)P(B) + \\ &\quad P(A_1 \cap A_2|B^c)P(B^c) \end{aligned}$$

Biased

$$\begin{aligned} \{1, 2, 3, 4, 6, 6\} &= \frac{4}{36} \cdot \frac{1}{2} + \frac{1}{36} \cdot \frac{1}{2} \\ &= \frac{5}{72} \end{aligned}$$

$$P(A_1)P(A_2) = \frac{1}{4} \cdot \frac{1}{4} = \frac{1}{16}$$

$$P(A_1 \cap A_2) \neq P(A_1)P(A_2)$$

$$5.ii) a) P(A_1 \cap A_2 | B) = P(A_1 | B) P(A_2 | B)$$

$$P(A_1 \cap A_2 | B) = \frac{4}{36}$$

Die Outcomes: $\{1, 2, 3, 4, 6_a, 6_b\}$

$$A_1 \cap A_2 | B = \{(6_a, 6_a), (6_a, 6_b), (6_b, 6_a), (6_b, 6_b)\}$$

$$P(A_1 | B) = \frac{2}{6}, P(A_2 | B) = \frac{2}{6}$$

$$b) P(A_1 \cap A_2 | B^c) = P(A_1 | B^c) P(A_2 | B^c)$$

$$P(A_1 \cap A_2 | B^c) = \frac{1}{36}$$

$$P(A_1 | B^c) = \frac{1}{6}, P(A_2 | B^c) = \frac{1}{6}$$

11. ii) ^{Show} $P(A|B, C) < P(A|C)$

Given

- $P(A \cap B) > 0$
- $P(A \cup B) < 1$
- $C = A \cup B$
- A, B are independent given C

(Verify: $P(A|B, C) = P(A|B \cap C)$)

$$\begin{aligned}
 P(A|B, C) &= \frac{P(A|B \cap C)}{P(B \cap C)} \\
 &= \frac{P(A \cap B \cap C)}{P(B \cap C)} \\
 &= \frac{P(A \cap B \cap (A \cup B))}{P(B \cap (A \cup B))} \\
 &= \frac{P(A \cap B)}{P(B)} \\
 &= \frac{P(A)P(B)}{P(B)} = P(A)
 \end{aligned}$$

$$\begin{aligned}
 P(A|C) &= \frac{P(A \cap C)}{P(C)} \\
 &= \frac{P(A)}{P(C)}
 \end{aligned}$$

$P(A|C) < P(A|B, C)$ since $0 < P(C) < 1$