

CONDITIONAL PROBABILITY

3.3

Exercise 3.3 A machine reports the true value of a thrown 12-sided die 5 out of 6 times.

a. If the machine reports a 1 has been tossed, what is the probability that it is actually a 1?

b. Now let the machine only report whether a 1 has been tossed or not. Does the probability change?

(a) We will denote the events $T_k = \text{The toss has been } k \quad \forall k \in \{1, \dots, 12\}$ and the event $M_k = \text{the machine reports } k \quad \forall k \in \{1, \dots, 12\}$. Let's figure out the probabilities:

$$P(T_k) = 1/12 \quad \forall k \text{ because is a 12-sided die}$$

We don't know directly $P(M_k)$, but the conditional probability depending of the real toss: $P(M_k | T_{k'}) \quad k, k' \in \{1, \dots, 12\}$

- If $k = k' \Rightarrow P(M_k | T_{k'}) = 5/6$

- If $k \neq k' \Rightarrow P(M_k | T_{k'}) = \frac{1}{6} \cdot \frac{1}{11}$ because $1/6$ is the prob. of giving a false value and $1/11$ is the prob. that the false value is $k \neq k'$ and not other $k_2 \neq k'$.

Now, we can calculate:

$$\begin{aligned} P(T_2 | M_2) &= \frac{P(M_2 | T_2) P(T_2)}{P(M_2)} = \underbrace{\frac{P(M_2 | T_2) P(T_2)}{\sum_{k=1}^{12} P(M_k | T_k) P(T_k)}}_{{}^{\text{5/6} \cdot 1/12}} = \\ &= \frac{\frac{5}{6} \cdot 1/12}{\frac{5}{6} \cdot 1/12 + \sum_{k=1}^{12} P(M_k | T_k) P(T_k)} = \frac{\frac{5}{6} \cdot 1/12}{\frac{5}{6} \cdot 1/12 + 11 \cdot (\frac{1}{6} \cdot \frac{1}{11} \cdot \frac{1}{12})} = \frac{5/6}{28/82} = \boxed{\frac{5}{6}} \end{aligned}$$

(b) In this case, $P(M_k | T_{k'})$ is changing, because we don't count the selection of the false value by the machine. Now, $\forall k, k' \in \{1, \dots, 12\}$

- If $k = k' \Rightarrow P(M_k | T_{k'}) = 5/6$

- If $k \neq k' \Rightarrow P(M_k | T_{k'}) = \frac{1}{6}$ because the machine only reports that k' is not the tossed value.

$$\begin{aligned} P(T_2 | M_2) &= \frac{P(M_2 | T_2) P(T_2)}{P(M_2)} = \underbrace{\frac{P(M_2 | T_2) P(T_2)}{\sum_{k=1}^{12} P(M_k | T_k) P(T_k)}}_{{}^{\text{5/6} \cdot 1/12}} = \\ &= \frac{\frac{5}{6} \cdot 1/12}{\frac{5}{6} \cdot 1/12 + \sum_{k=1}^{12} P(M_k | T_k) P(T_k)} = \frac{\frac{5}{6} \cdot 1/12}{\frac{5}{6} \cdot 1/12 + 11 \cdot (\frac{1}{6} \cdot \frac{1}{12})} = \boxed{\frac{5}{16}} \end{aligned}$$

→ The probability changes !!

3.8

Exercise 3.8 We have two coins of identical appearance. We know that one is a fair coin and the other flips heads 80% of the time. We choose one of the two coins uniformly at random. We discard the coin that was not chosen. We now flip the chosen coin independently 10 times, producing a sequence $Y_1 = y_1, Y_2 = y_2, \dots, Y_{10} = y_{10}$.

- Intuitively, without doing any computation, are these random variables independent?
- Compute the probability $P(Y_1 = 1)$.
- Compute the probabilities $P(Y_2 = 1 | Y_1 = 1)$ and $P(Y_{10} = 1 | Y_1 = 1, \dots, Y_9 = 1)$.
- Given your answers to b) and c), would you now change your answer to a)? If so, discuss why your intuition had failed.

(a) Intuitively, should be independent, because every flip doesn't depend of the other ones

(b) For calculating $P(Y_2 = 1)$, we need to know what coin has been flipped, so we use the law of total probability. If we denote $B = \begin{cases} 0 & \text{fair coin} \\ 1 & \text{biased coin} \end{cases}$

$$P(Y_2 = 1) = \sum_{i=0}^1 P(Y_2 = 1 | B = i) P(B = i) = \frac{1}{2} \sum_{i=0}^1 P(Y_2 = 1 | B = i) = \frac{1}{2} \left(\frac{1}{2} + \frac{4}{5} \right) = \boxed{\frac{13}{20}}$$

(c) We will compute $P(Y_2 = 1 | Y_1 = 1)$ first. By the same reasoning of (b), we need to condition it by the coin chosen, because the prob. of $Y_2 = 1$ depends of the coin chosen (and maybe of Y_1).

$$\begin{aligned} P(Y_2 = 1 | Y_1 = 1) &= \sum_{i=0}^1 \underbrace{P(Y_2 = 1 | B = i, Y_1 = 1)}_{\text{depends only of } B} P(B = i | Y_1 = 1) = \\ &= \frac{4}{5} \cdot \frac{P(Y_2 = 1 | B = 0) P(B = 0)}{P(Y_1 = 1)} + \frac{1}{2} \cdot \frac{P(Y_2 = 1 | B = 1) P(B = 1)}{P(Y_1 = 1)} = \\ &= \frac{4}{5} \cdot \frac{\frac{4}{5} \cdot \frac{1}{2}}{\frac{13}{20}} + \frac{1}{2} \cdot \frac{\frac{1}{2} \cdot \frac{1}{2}}{\frac{13}{20}} \approx 0.68 \quad (\neq 0.65 = \frac{13}{20}) \end{aligned}$$

Bayes Theorem

We will now define $X = 1 \Leftrightarrow Y_1 = 1, \dots, Y_{10} = 1$

$P(X = 1) = \sum_{i=0}^1 P(X = 1 | B = i) P(B = i) = \frac{1}{2} \left[\left(\frac{1}{2}\right)^9 + \left(\frac{4}{5}\right)^9 \right]$ so, with this, we can apply the same computations of before:

$$\begin{aligned} P(Y_2 = 1 | X = 1) &= \sum_{i=0}^1 P(Y_2 = 1 | B = i, X = 1) P(B = i | X = 1) = \\ &= \frac{4}{5} \cdot \frac{P(X = 1 | B = 0) P(B = 0)}{P(X = 1)} + \frac{1}{2} \cdot \frac{P(X = 1 | B = 1) P(B = 1)}{P(X = 1)} = \\ &= \frac{4}{5} \cdot \frac{\frac{4}{5} \cdot \frac{1}{2}}{P(X = 1)} + \frac{1}{2} \cdot \frac{\frac{1}{2} \cdot \frac{1}{2}}{P(X = 1)} \approx 0.8 \quad (\neq 0.68 = P(X = 1)) \end{aligned}$$

(d) I didn't think about how is more probable that the coin is the biased one if we got after a lot of "independent" tosses always head.