

2.1.1

Since we have that $A \subset B$, $A \cap B = A$. Then,

$$\Pr(A | B) = \frac{\Pr(A \cap B)}{\Pr(B)} = \frac{\Pr(A)}{\Pr(B)}$$

2.1.4

Odds are parenthetical. He must start with A ($\frac{1}{2}$), then proceed to stay ($\frac{1}{3}$), switch ($\frac{2}{3}$), and stay ($\frac{1}{3}$). This leaves us with

$$\frac{1}{2} \cdot \frac{1}{3} \cdot \frac{2}{3} \cdot \frac{1}{3} = \frac{1}{27}$$

2.1.11

Note that since $A \cap B$ and $A^C \cap B$ are disjoint, as no x can be both in A and A^C , $\Pr(A \cap B) + \Pr(A^C \cap B) = \Pr((A \cap B) \cup (A^C \cap B)) = \Pr(B)$.

$$\Pr(A^C | B) = \frac{\Pr(A^C \cap B)}{\Pr(B)} = \frac{\Pr(B) - \Pr(A \cap B)}{\Pr(B)} = 1 - \frac{\Pr(A \cap B)}{\Pr(B)} = 1 - \Pr(A | B)$$

2.2.2

We have that $\Pr(A^C \cap B^C) = \Pr((A \cup B)^C) = 1 - \Pr(A \cup B)$ from De Morgan's, which expands to

$$\begin{aligned} 1 - (\Pr(A) + \Pr(B) - \Pr(A \cap B)) &= 1 - \Pr(A) - \Pr(B) + \Pr(A) \Pr(B) \\ &= (1 - \Pr(A))(1 - \Pr(B)) \\ &= \Pr(A^C) \Pr(B^C) = \Pr(A^C \cap B^C) \end{aligned}$$

This shows that A^C, B^C are independent.

2.2.12a

From before, we have that complements are also independent if the original events were independent; similarly, we have that if A, B are independent, as $A \cap B$ and $A \cap B^C$ are disjoint we have $\Pr(A \cap B^C) = \Pr(A) - \Pr(A \cap B) = \Pr(A)(1 - \Pr(B)) = \Pr(A) \Pr(B^C)$, so A, B^C are also independent.

The odds that none of the three occur:

$$(1 - \frac{1}{4})(1 - \frac{1}{3})(1 - \frac{1}{2}) = \frac{1}{4}$$

2.2.12b

The odds that exactly one occurs:

$$\frac{1}{4}(1 - \frac{1}{3})(1 - \frac{1}{2}) + \frac{1}{3}(1 - \frac{1}{4})(1 - \frac{1}{2}) + \frac{1}{2}(1 - \frac{1}{4})(1 - \frac{1}{3}) = \frac{1}{12} + \frac{1}{8} + \frac{1}{4} = \frac{11}{24}$$

2.2.22

Claim. Suppose that A_1, A_2, B are events such that $\Pr(A_1 \cap B) > 0$. Then A_1, A_2 are conditionally independent given $B \iff \Pr(A_2 | A_1 \cap B) = \Pr(A_2 | B)$.

Proof. (\implies) We already have that

$$\Pr(A_1 \cap A_2 | B) = \Pr(A_1 | B) \Pr(A_2 | B)$$

as we assume that A_1, A_2 are conditionally independent.

Expanding,

$$\begin{aligned} \Pr(A_1 \cap A_2 | B) &= \Pr(A_1 | B) \Pr(A_2 | B) \\ \implies \frac{\Pr(A_1 \cap A_2 \cap B)}{\Pr(B)} &= \frac{\Pr(A_1 \cap B)}{\Pr(B)} \cdot \frac{\Pr(A_2 \cap B)}{\Pr(B)} \\ \implies \frac{\Pr(A_1 \cap A_2 \cap B)}{\Pr(A_1 \cap B)} &= \frac{\Pr(A_2 \cap B)}{\Pr(B)} \\ \implies \Pr(A_2 | A_1 \cap B) &= \Pr(A_2 | B) \end{aligned}$$

(\impliedby) We have that $\Pr(A_2 | A_1 \cap B) = \Pr(A_2 | B) \implies \frac{\Pr(A_1 \cap A_2 \cap B)}{\Pr(A_1 \cap B)} = \frac{\Pr(A_2 \cap B)}{\Pr(B)}$.

$$\begin{aligned} \Pr(A_1 | B) \Pr(A_2 | B) &= \frac{\Pr(A_1 \cap B)}{\Pr(B)} \cdot \frac{\Pr(A_1 \cap B)}{\Pr(B)} \\ &= \frac{\Pr(A_1 \cap B)}{\Pr(B)} \cdot \frac{\Pr(A_2 \cap A_2 \cap B)}{\Pr(A_1 \cap B)} \\ &= \frac{\Pr(A_1 \cap A_2 \cap B)}{\Pr(B)} \\ &= \Pr(A_1 \cap A_2 | B) \end{aligned}$$

□

2.3.4

Let E_1 = sick, E_2 = not sick, and A = a positive test.

$$\Pr(E_1 | A) = \frac{\Pr(E_1) \Pr(A | E_1)}{\Pr(E_1) \Pr(A | E_1) + \Pr(E_2) \Pr(A | E_2)} = \frac{0.00001 \cdot 0.95}{0.99999 \cdot 0.05} = 0.00019$$

2.3.8a

The odds of arriving at that configuration with coin n is $p_n(1 - p_n)^3$, and let the event of having coin n have odds $\Pr(n) = \frac{1}{5}$. Then, Bayes' Theorem tells us that the posterior odds of having coin n are

$$\frac{\Pr(n) p_n(1 - p_n)^3}{\sum_{i=1}^5 \Pr(n) p_i(1 - p_i)^3} = \frac{p_n(1 - p_n)^3}{\sum_{i=1}^5 p_i(1 - p_i)^5}$$

Actually computing, we have that the posterior odds are $\Pr(1 | TTTH) = 0$, $\Pr(2 | TTTH) = 0.587$, $\Pr(3 | TTTH) = 0.347$, $\Pr(4 | TTTH) = 0.065$, and $\Pr(5 | TTTH) = 0$.

2.3.8a

This can be computed as $\sum_{i=1}^5 \Pr(i | TTTH) p_i(1 - p_i)^2$, which comes out to be 0.129.

2.5.1

$$\begin{aligned} & \frac{\Pr(A \cap D)}{\Pr(D)} \geq \frac{\Pr(B \cap D)}{\Pr(D)} \\ \Rightarrow & \Pr(A \cap D) \geq \Pr(B \cap D) \\ & \frac{\Pr(A \cap D^C)}{\Pr(D^C)} \geq \frac{\Pr(B \cap D^C)}{\Pr(D^C)} \\ \Rightarrow & \Pr(A \cap D^C) \geq \Pr(B \cap D^C) \\ \Rightarrow & \Pr(A \cap D) + \Pr(A \cap D^C) \geq \Pr(B \cap D) + \Pr(B \cap D^C) \\ \Rightarrow & \Pr(A) \geq \Pr(B) \end{aligned}$$

2.5.23

Let E_1 = statistician, E_2 = economist, A = shy.

$$\Pr(E_1 | A) = \frac{\Pr(E_1) \Pr(A | E_1)}{\Pr(E_1) \Pr(A | E_1) + \Pr(E_2) \Pr(A | E_2)} = \frac{0.8 * 0.1}{0.8 * 0.1 + 0.15 * 0.9} = 0.372$$

2.5.29a

There are $\binom{52}{13}$ ways to choose 13 cards; $\binom{4}{n} \binom{48}{13-n}$ ways to choose 12 cards and n aces.

$$1 - \frac{\binom{4}{1} \binom{48}{12}}{\binom{52}{13} - \binom{48}{13}} = 0.369$$

2.5.29b

Let H = picking the ace of hearts, and A = the number of aces. Then, $\Pr(H) = \frac{13}{52}$.

$$\begin{aligned} \Pr(A \geq 2 | H) &= \frac{\Pr(A \geq 2 \cap H)}{\Pr(H)} \\ &= \frac{\Pr(H) - \Pr(A < 2 \cap H)}{\Pr(H)} \\ &= \frac{\Pr(H) - \Pr(A = 1 \cap H)}{\Pr(H)} \\ &= \frac{\Pr(H) - \frac{\binom{48}{12}}{\binom{52}{13}}}{\Pr(H)} = 0.561 \end{aligned}$$