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Solutions Sheet 1

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Exercise 1

First, we will assign values to $p(M), p(m), p(e), P(n | M), P(n | m), P(n | e)$, so that $p(M) + p(m) + p(e) = 1$:

$$p(M) := 0.01$$

$$p(m) := 0.1$$

$$p(e) := 0.89$$

$$p(n | M) = 0.9$$

$$p(n | m) = 0.08$$

$$p(n | e) = 0.02$$

The next step to apply Bayes' rule is now to calculate the probability of a noise:

$$p(n) = p(M) \cdot p(n | M) + p(m) \cdot p(n | m) + p(e) \cdot p(n | e)$$

$$= 0.01 \cdot 0.9 + 0.1 \cdot 0.08 + 0.89 \cdot 0.02$$

$$= 0.0348$$

We can now calculate the probability of a monster given noise by applying Bayes' rule:

$$p(M | n) = \frac{p(n | M) \cdot p(M)}{p(n)}$$

$$= \frac{0.9 \cdot 0.01}{0.0348}$$

$$\approx 0.26$$

These results got confirmed by the jupyter notebook presented in the lecture:

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In [5]: 1 # what is the probability of having noise
        2 p_n = p_nm + p_nm + p_ne # p(n) = p(n,M) + p(n,m) + p(n,e)
        3 print(f"p(n) = {p_n}")
        p(n) = 0.0348

In [7]: 1 # what is the (posterior) probability of having a ... given that we hear noise
        2 p_M_n = Bayes_rule(p_M, p_n_M, p_n) # p(M|n) = p(M) p(n|M) / p(n)
        3 p_m_n = Bayes_rule(p_m, p_n_m, p_n) # p(m|n) = p(m) p(n|m) / p(n)
        4 p_e_n = Bayes_rule(p_e, p_n_e, p_n) # p(e|n) = p(e) p(n|e) / p(n)
        5 print(f"p(M|n) = {p_M_n}")
        6 print(f"p(m|n) = {p_m_n}")
        7 print(f"p(e|n) = {p_e_n}")

p(M|n) = 0.2586206896551725
p(m|n) = 0.2298585974712644
p(e|n) = 0.5114942528735632


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Exercise 2

Given:

- The plausibility of B is $p(B)$.
- $p(B | A) \geq p(B)$

Show: (a) $p(B | A) \geq p(B)$

Proof. Is equivalent to second assumption. 

□

(b) $p(B | \neg A) \leq p(B)$

Proof. By assumption:

$$\begin{aligned}
 p(B|A) &\geq p(B) \\
 \Leftrightarrow \frac{p(A | B) \cdot p(B)}{p(A)} &\geq p(B) \quad \checkmark \\
 \Leftrightarrow \frac{(1 - p(\neg A | B)) \cdot p(B)}{1 - p(\neg A)} &\geq p(B) \quad \checkmark \\
 \Leftrightarrow (1 - p(\neg A | B)) \cdot p(B) &\geq p(B) \cdot (1 - p(\neg A)) \quad \checkmark \\
 \Leftrightarrow 1 - p(\neg A | B) &\geq 1 - p(\neg A) \quad \checkmark \\
 \Leftrightarrow -p(\neg A | B) &\geq -p(\neg A) \quad \checkmark \\
 \Leftrightarrow p(\neg A | B) &\leq p(\neg A) \quad \checkmark \\
 \Leftrightarrow p(\neg A | B) &\leq p(\neg A) \cdot \frac{p(B)}{p(B)} \\
 \Leftrightarrow \frac{p(\neg A | B) \cdot p(B)}{p(\neg A)} &\leq p(B) \quad \checkmark \\
 \Leftrightarrow p(B | \neg A) &\leq p(B) \quad \checkmark
 \end{aligned}$$

□

(c) $p(A | B) \geq p(A)$

Proof. By assumption:

$$\begin{aligned}
 p(B|A) &\geq p(B) \\
 \Leftrightarrow \frac{p(A|B) \cdot p(B)}{p(A)} &\geq p(B) \quad \checkmark \\
 \Leftrightarrow p(A|B) &\geq p(A) \quad \checkmark
 \end{aligned}$$

□

(d) $p(A | \neg B) \leq p(A)$

Proof. By assumption:

$$\begin{aligned}
 p(B|A) &\geq p(B) \\
 \stackrel{(c)}{\Leftrightarrow} p(A|B) &\geq p(A) \quad \checkmark \\
 \stackrel{(b)}{\Leftrightarrow} p(A|\neg B) &\leq p(A) \quad \checkmark
 \end{aligned}$$

□

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Exercise 3

For this exercise we will formalise the given facts as follows:

1. C = "The Patient has Cancer"
2. R = "A Patient's Mammogram is positive"

We can therefore deduct from the given facts:

1. $P(C) = 0.008$ *✓*
2. $P(R | C) = 0.9$ *✓*
3. $P(R | \neg C) = 0.07$ *✓*
4. $P(\neg C) = 1 - P(C) = 0.992$ *✓*
5. $P(R) = P(C) \cdot P(R | C) + P(\neg C) \cdot P(R | \neg C) = 0.008 \cdot 0.9 + 0.992 \cdot 0.07 = 0.08$ *✓*

Given these deductions, we can now calculate the addressed probability that a patient has breast cancer given a positive Mammogram $P(C | R)$ by applying Bayes' Rule:

$$\begin{aligned}
 P(C | R) &= \frac{P(R | C) \cdot P(C)}{P(R)} \quad \checkmark \\
 &= \frac{0.9 \cdot 0.008}{0.08} \\
 &= 0.09 \quad \checkmark
 \end{aligned}$$

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Exercise 4

For this exercise we will formulate the given facts as follows :

- W_A = "The warden announces A"
- W_B = "The warden announces B"
- W_C = "The warden announces C"
- A = "A is pardoned"
- B = "B is pardoned"
- C = "C is pardoned"

Before any announcement is made by the warden we obviously have :

$$P(A) = P(B) = P(C) = \frac{1}{3} \quad \checkmark$$

No Matter what the warden will either announce A or B , which brings us to the conclusion:

$$P(A) = P(W_B | A) + P(W_C | A) = \frac{1}{3}$$

Because in this case the warden flips a coin we know:

$$P(W_B | A) = P(W_C | A) = \frac{1}{6}$$

We know that $P(W_B | B) = 0$ and we can therefore also deduct that :

$$P(W_B) = P(A) \cdot P(W_B | A) + P(C) \cdot P(W_B | C) = \frac{1}{3} + \frac{1}{6} = \frac{1}{2} \quad \checkmark$$

Now we can apply Bayes' formula to calculate the updated probability after the Warden's announcement:

$$\begin{aligned} P(A | W_B) &= \frac{P(W_B | A) \cdot P(A)}{P(W_B)} \\ &= \frac{\frac{1}{6} \cdot \frac{1}{3}}{\frac{1}{2}} \\ &= \frac{1}{3} \quad \checkmark \end{aligned}$$

We also now that $P(W_B | C) = 1$. It follows:

$$\begin{aligned}
 P(C \mid W_B) &= \frac{P(W_B \mid C) \cdot P(C)}{P(W_B)} \\
 &= \frac{1 \cdot \frac{1}{3}}{\frac{1}{2}} \\
 &= \frac{2}{3} \quad \checkmark
 \end{aligned}$$

Therefore C's argument is right.

✓

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