

1. - $P(E) \geq 0$ for all $E \subset S$ (S is the sample space)
 - $P(S) = 1$
 - Let E_1 and E_2 be mutually exclusive events
 $P(E_1 \cup E_2) = P(E_1) + P(E_2)$

2. a) $A_1 \cup A_2 = A_1 \cup A_2$
 b) $A_1 \cap A_2 = \emptyset$
 c) $P(A_1 \cup A_2) = P(A_1) + P(A_2)$
 d) $P(A_1 \cap A_2) = 0$
 e) Ω is the sample space of flipping a coin
 $\rightarrow A_1 = \text{Heads}$ and $A_2 = \text{Tails}$

3. a) $P(A \cap B) = P(A) \cdot P(B)$
 b) $P(A|B) = \frac{P(A \cap B)}{P(B)} = \frac{P(A) \cdot P(B)}{P(B)} = P(A)$

4. a) $p(x, y) = p(x) \cdot p(y)$
 b) $P(X > 2 | Y < 1)$: these two events don't seem to be dependent on each other so $P(X > 2 | Y < 1) = P(X > 2)$

5. - Discrete: the support consists of a finite number of all possible values of the variable.
 - Continuous: the support is a range / interval

(6)	Type (i)	Support (ii)
$X \sim \text{Beta}(a, b)$	Continuous	$[0, 1]$
$X \sim \text{Bin}(n, p)$	Discrete	$\{0, 1, 2, \dots, n\}$
$X \sim \text{Gamma}(a, b)$	Continuous	$(0, +\infty)$
$X \sim \text{Normal}(\mu, \sigma^2)$	Continuous	$(-\infty, +\infty)$
$X \sim \text{Poisson}(\lambda)$	Discrete	$\{0, 1, 2, \dots\}$
$X \sim \text{Uniform}(a, b)$	Continuous	$[a, b]$

7. PMF:

- $f(x) \geq 0$ for all $x \in S \rightarrow f(x)$ is defined

- $\sum f(x) = 1$ (x is all elements in S)

- $f(x) = P(X=x)$

PDF:

- $f(x) \geq 0 \rightarrow f(x)$ is defined

- The area under the curve totals to 1

- $f(x)$ = the probability of X being in a range.

$$P(a \leq X \leq b) = \int_a^b f(x) dx$$

1.8. a) Let X be the random variable representing the number of defective batteries. X follows a binomial distribution $X \sim \text{Bin}(n, \theta)$

1) $E(X) = n \times \theta = 10 \times 0.05 = 0.5$

b) $P(X=x) = \binom{n}{x} \times 0.05^x \times 0.95^{(n-x)}$

$$\Rightarrow P(X=0) = \binom{10}{0} \times 0.05^0 \times 0.95^{10}$$

$$P(X=1) = \binom{10}{1} \times 0.05^1 \times 0.95^9$$

$$\Rightarrow P(X=0) + P(X=1) \approx 0.91 = 91\%$$

9. $\int_0^\infty \theta^5 e^{-3\theta} d\theta \rightarrow \theta$ is the variable of a PDF $f(\theta)$. Then,

$\int_0^\infty \theta^5 e^{-3\theta} d\theta$ is the probability $P(\theta \geq 0) \geq 0$

$\Rightarrow \int_0^\infty \theta^5 e^{-3\theta} d\theta \geq 0 \Rightarrow \theta^5 e^{-3\theta} \geq 0 \Rightarrow \theta^5 \geq 0 \Rightarrow \theta \geq 0 \Rightarrow [0, \infty) \rightarrow \text{support}$

(Kernel: $e^{-3\theta}$) (because it doesn't involve θ)

\Rightarrow This resembles the Gamma distribution:

$[0, \infty)$ is the support so the function

$\int_0^\infty \theta^5 e^{-3\theta} d\theta$ is covering the entire range of all possible

values of the random variable, so it must equal to 1

II. ELISA test

HEV: the person has HIV

TP: test is positive

$$a) P(TP) = P(TP | HEV) \cdot P(HEV) + P(TP | HEV^c) \cdot P(HEV^c)$$

$$= 0.99 \times \frac{1}{10000} + (1 - P(TP^c | HEV^c)) \times (1 - P(HEV))$$

$$\approx 9.95 \cdot 10^{-5}$$

$$b) P(HEV | TP) = \frac{P(TP | HEV) \cdot P(HEV)}{P(TP)}$$

$$= \frac{P(TP | HEV) \cdot P(HEV)}{(1 - P(TP^c | HEV^c)) \cdot P(HEV^c) + P(TP | HEV) \cdot P(HEV)}$$

$$= \frac{1 - P(TP^c | HEV^c)}{1 - P(TP^c | HEV^c) + P(TP | HEV) / P(HEV^c)}$$

$$= \frac{(1 - 0.01) \times \frac{1}{10000}}{1 - 0.01 + 0.99 \cdot 10^4} \approx 9.9 \cdot 10^{-5}$$

$$c) P(HEV | TP) = \frac{P(TP | HEV) \cdot P(HEV)}{P(TP)}$$

$$\approx \frac{0.99 \times \frac{1}{10000}}{9.95 \cdot 10^{-5}} \approx 1.$$

III. Genetic Status

$$a) P(X = u) = 0.5^u (1 - 0.5)^{1-u}$$

for $x \in \{0, 1\}$. u is equally likely to be either.

$$b) P(y_1 = 0, y_2 = 0 | \theta = 1)$$

$$= P(y_1 = 0 | \theta = 1) \cdot P(y_2 = 0 | \theta = 1)$$

$$= \frac{P(\theta = 1 | y_1 = 0) \cdot P(y_1 = 0)}{P(\theta = 1)} \cdot \frac{P(\theta = 1 | y_2 = 0) \cdot P(y_2 = 0)}{P(\theta = 1)}$$

$$= \frac{P(\theta = 1) \cdot P(y_1 = 0)}{P(\theta = 1)} \cdot \frac{P(\theta = 1) \cdot P(y_2 = 0)}{P(\theta = 1)}$$

$$= 0.5 \cdot 0.5 = 0.25$$

$$P(y_1 = 0, y_2 = 0 | \theta = 0)$$

$$= P(y_1 = 0 | \theta = 0) \cdot P(y_2 = 0 | \theta = 0)$$

$$= 1 \cdot 1$$

$$= 1$$

/ fixed thanks to peer reviews

$$c) \quad p(\theta = 1 | y_1, y_2) = \frac{p(y_1, y_2 | \theta = 1) \cdot p(\theta = 1)}{p(y_1, y_2)}$$

$$= \frac{p(y_1, y_2 | \theta = 1) \cdot p(\theta = 1)}{p(y_1, y_2 | \theta = 1) \cdot p(\theta = 1) + p(y_1, y_2 | \theta = 0) \cdot p(\theta = 0)}$$

$$= \frac{0.25 \times 0.5}{0.25 \times 0.5 + 0.25 \times 0.5} = 0.5$$

$$d) \quad p(\theta = 1 | y_1, y_2, y_3) = \frac{p(y_1, y_2, y_3 | \theta = 1) \cdot p(\theta = 1)}{p(y_3)} \quad \text{prior}$$

$$\text{OR} \quad = \frac{p(y_3 | \theta = 1) \cdot p(\theta = 1 | y_1, y_2)}{p(y_3 | \theta = 1) \cdot p(\theta = 1 | y_1, y_2) + p(y_3 | \theta = 0) \cdot p(\theta = 0 | y_1, y_2)}$$

$$= \frac{0.25 \times 0.5}{0.25 \times 0.5 + 0.25 \times 0.5} = 0.5$$