STAT1301 Advanced Analysis of Scientific Data

Semester 2, 2025, Assignment 2

Caleb Yates s49886350

1 Introduction

Throughout the report, the following syntactical shortcuts and notation will be used.

If X is a random variable of the sample space Ω , an abbreviation of set notation is as follows:

Abbreviate
$$\{d : \forall d \in \Omega \text{ and } X(d) = x\}$$

As $\{X = x\}$

Additionally, when thinking in terms of sets becomes obselete,

Abbreviate
$$P(\{d : \forall d \in \Omega \text{ and } X(d) = x\})$$

As $P(X = x)$

The abbriviation will be used with inequalities as well.

Given some random variable X, there must exist a function mapping from the sample space Ω to the domain of X, which can be at most \mathbb{R} . This function is (intuitively) named X. This function incidentally defines the random variable, which is the motivating reason for using its letter to represent its mapping. The notation Domain[X] will be used throughout this report to indicate the domain of the function mapping X and hence the random variable X itself by definition.

Also, $N(\mu, \sigma)$ indicates that σ is $\sqrt{\text{Var}}$, aka the standard deviation. This is opposed to the syntax of $N(\mu, \sigma^2)$. For clarity, $\sigma = \text{will}$ always be explicitly written to avoid ambiguity.

Various probability (and set) theorems are used throughout this report. For clarity, the following are named:

$$P({X < x}) = P({x > X})\forall x \tag{1}$$

$$P({X < x}) = 1 - P({X > x}) \forall x$$
(2)

Above (1) and (2) are true for any random variable X.

$$X \sim N(\mu = 0, \sigma)$$

$$\Longrightarrow P(\{X < x\}) = P(\{X > -x\})$$

$$\Longleftrightarrow P(\{X < -x\}) = P(\{X > x\})$$
(3)

When X is a symmetrical distribution around 0, for example the standard normal distribution Z, above (3) is true.

To begin, lets define the sample space

$$\Omega = \{(a, b) \in \{1, 2, 3, 4, 5, 6\}\}$$
$$|\Omega| = 36$$

Notice this is uniform, and hence that a and b are independent

$$P(A) = \frac{|A|}{|\Omega|} = \frac{|A|}{36}$$

2.1 Part a)

Let X be a random variable representing the payout of a given dice roll $(a, b) \in \Omega$:

$$X((a,b) \in \Omega) = a \cdot b$$

Let f_X be the PMF of X. Note $f_X(x \in \Omega) = P(\{X = x\})$. By cases, the probability distribution of X can be deduced:

$$f_{X}(1) = \frac{1}{36}$$

$$f_{X}(2) = \frac{2}{36}$$

$$f_{X}(2) = \frac{2}{36}$$

$$f_{X}(3) = \frac{2}{36}$$

$$f_{X}(4) = \frac{3}{36}$$

$$f_{X}(4) = \frac{3}{36}$$

$$f_{X}(5) = \frac{2}{36}$$

$$f_{X}(10) = \frac{2}{36}$$

$$f_{X}(10) = \frac{2}{36}$$

$$f_{X}(24) = \frac{2}{36}$$

$$f_{X}(25) = \frac{1}{36}$$

$$f_{X}(30) = \frac{2}{36}$$

$$f_{X}(30) = \frac{2}{36}$$

$$f_{X}(30) = \frac{2}{36}$$

$$f_{X}(30) = \frac{1}{36}$$

For all other values x, $f_X(x) = 0$

2.2 Part b)

This makes determining the expected value of X trivial:

$$\begin{split} \mathbf{E}(\mathbf{X}) &= \sum_{c \in \Omega} \mathbf{X}(c) \mathbf{P}(c) \\ &= \sum_{x \in \text{Domain}[\mathbf{X}]} x \mathbf{P}(\{\mathbf{X} = x\}) \\ &= 1 \cdot f_{\mathbf{X}}(1) + 2 \cdot f_{\mathbf{X}}(2) + \cdots 30 \cdot f_{\mathbf{X}}(30) + 36 \cdot f_{\mathbf{X}}(36) \\ &= \frac{1}{36} + \frac{4}{36} + \cdots \frac{60}{36} + \frac{36}{36} \\ &= \frac{441}{36} = \frac{49}{4} = 12.25 \end{split}$$

2.3 Part c)

Evaluating Var(X) is similarly trivial

$$\begin{split} \operatorname{Var}(X) &= \operatorname{E}[(X - \operatorname{E}(X))^2] \\ &= \sum_{c \in \Omega} (X(c) - \frac{49}{4})^2 \operatorname{P}(\{c\}) \\ &= \sum_{x \in \operatorname{Domain}[X]} (x - \frac{49}{4})^2 \operatorname{P}(\{X = x\}) \\ &= (1 - \frac{49}{4})^2 \cdot \frac{1}{36} + (2 - \frac{49}{4})^2 \cdot \frac{2}{36} + \cdots (30 - \frac{49}{4})^2 \cdot \frac{2}{36} + (36 - \frac{49}{4})^2 \cdot \frac{1}{36} \\ &= \frac{11515}{144} \approx 79.97 \\ \Longrightarrow \sigma_X &= \sqrt{\operatorname{Var}(X)} = \sqrt{\frac{11515}{144}} \approx 8.942 \end{split}$$

Understanding this question in terms of a sample space isn't very fruitful. Ω is completely unspecified, we can only deduce that $|\Omega| \geq (0, 20)$, which implies it is continuous. $P(A) : \exists A \in \Omega$ is also completely unknown.

3.1 Part a)

Let X be the continuous random variable of algae growth as measured in grams of biomass produced. Note Domain[X] = (0, 20).

Since X is a random variable, its PDF f_X must sum to 1:

$$\begin{split} 1 &= \int_{c \in \Omega} \mathrm{P}(\{c\}) \\ &= \int_{x \in \mathrm{Domain}[X]} \mathrm{P}(\{\mathrm{X} = x\}) \\ &= \int_{0}^{20} c(x^2 - 60x + 800) \mathrm{d}x \\ &= c[\frac{1}{3}x^3 - 30x^2 + 800x]_{x=0}^{x=20} \\ 1/c &= [\frac{1}{3}(20)^3 - 30(20)^2 + 800(20)] - [0 - 0 + 0] \\ 1/c &= \frac{20000}{3} \\ c &= \frac{3}{20000} \end{split}$$

3.2 Part b)

Let F_X be the CDF of X:

$$F_{X} = \int_{-\infty}^{x} f_{X}(x) dx$$

$$= \int_{0}^{x} c(x^{2} - 60x + 800) dx$$

$$= c \left[\frac{1}{3}x^{3} - 30x^{2} + 800x \right]_{x=0}^{x=x}$$

$$\frac{F_{X}}{c} = \left[\frac{1}{3}x^{3} - 30x^{2} + 800x \right] - \left[\frac{1}{3}0^{3} - 30 \cdot 0^{2} + 800 \cdot 0 \right]$$

$$\implies F_{X} = c \left(\frac{1}{3}x^{3} - 30x^{2} + 800x \right) \text{ for } 0 \le x \le 20$$

$$= \frac{1}{20000}x^{3} - \frac{9}{2000}x^{2} + \frac{3}{25}x$$

3.3 Part c)

$$E(X) = \int_{x \in Domain[X]} x f_X dx$$

$$= \int_0^{20} x \cdot c(x^2 - 60x + 800) dx$$

$$\frac{E(X)}{c} = \int_0^{20} x^3 - 60x^2 + 800x dx$$

$$= \left[\frac{1}{4} x^4 - 20x^3 + 400x^2 \right]_{x=0}^{x=20}$$

$$= \left[\frac{1}{4} (20)^4 - 20(20)^3 + 400(20)^2 \right] - [0 - 0 + 0]$$

$$= 40000 - 160000 + 160000$$

$$E(X) = c \cdot 40000$$

$$E(X) = 6 \text{ grams}$$

3.4 Part d)

$$\begin{split} &P(\{X > 10\} | \{X > 2\}) \\ &= \frac{P(\{X > 10\} \cap \{X > 2\})}{\{X > 2\}} \\ &= \frac{P(\{X > 10\})}{P(\{X > 2\})} \end{split}$$

From the CDF definition of X, $P({X < x}) = f_X(x)$

$$\Rightarrow P(\{X > 10\}) = 1 - P(\{X < 10\})$$

$$= 1 - F_X(10)$$

$$= 1 - \frac{4}{5}$$

$$= \frac{1}{5}$$

$$\Rightarrow P(\{X > 2\}) = 1 - P(\{X < 2\})$$

$$= 1 - F_X(2)$$

$$= 1 - \frac{139}{625}$$

$$= \frac{486}{625}$$

$$\Rightarrow \frac{P(\{X > 10\})}{P(\{X > 2\})} = \frac{\frac{1}{5}}{\frac{486}{625}} = \frac{125}{486} \approx 0.2572$$
(4)

Therefore, the probability that the biomass exceeds 10 grams, given that it is detectable, is above in $(4) = \frac{125}{486}$.

Assume that p = 0.25 for all the products, not just the 25 that were sampled.

The sample space for this is again completely unspecified, and the P probability function is practically useless for this question. For convenience, the sample space Ω is therefore defined as the domain of X, representing the number of products passing the specific inspection.

$$\Omega = \{1, 2, 3...24, 25\}$$

This makes the definition of X trivial, and its domain incidentally the entire sample space:

$$X(a \in \Omega) = a$$

4.1 Part a)

Since each product has a p = 0.25 probability of passing inspection, and there are 25 products, and it is assumed each inspection and product is independent of each other, X is a binomial distribution:

$$X \sim Bin(n = 25, p = 0.25)$$

Notes the following theorems about binomial distributions and X:

$$P(\{X = x\}) = \binom{n}{x} p^x (1-p)^{n-x} = \binom{25}{x} 0.25^x \cdot 0.75^{25-x}$$
$$E(X) = np = \frac{25}{4}$$
$$Var(X) = np(1-p) = \frac{75}{16}$$

4.2 Part b)

Let X_2 be the random variable representing the probability distribution of X with an n parameter such that the probability of finding a defect-free product eexceeds 99%:

$$X_2 \sim Bin(n, p = 0.25)$$

$$\begin{split} P(\{X_2 \geq 1\}) &> 0.99 \\ 0.99 &< P(\{X_2 \geq 1\}) \\ 0.99 &< 1 - P(\{X_2 = 0\}) \\ 0.99 - 1 &< -\binom{n}{0}(0.25)^0(0.75)^n \\ 0.01 &> 1 \cdot 1 \cdot 0.75^n \\ \log_{0.75} 0.01 &> n \\ &\implies n < \log_{0.75} 0.01 \approx 16.008 \end{split}$$

Therefore the minimum (integer) sample size is n = 16.

4.3 Part c)

The random variable Y is dependent on X. Given a possibility $a \in \Omega$ from the sample space, Y(a) explicitly depends upon X(a) such that it exactly equals:

$$Y(a \in \Omega) = 3X(a) - (25 - X(a))$$

= $4X(a) - 25$

This allows us to calculate E(X) and Var(X) relatively easily using probability theorems:

$$E(Y) = E(4X - 25)$$

$$= 4E(X) - 25$$

$$= 4 \cdot \frac{25}{4} - 25$$

$$= 0$$

$$Var(Y) = Var(4X - 25)$$
$$= 4^{2}Var(X)$$
$$= 16 \cdot \frac{75}{16}$$
$$= 75$$

4.4 Part d)

Since Y is defined in terms of X, this isn't too difficult to evaluate:

$$P(\{Y \ge 27\}) = P(\{4X - 25 \ge 27\})$$

$$= P(\{4X \ge 52\})$$

$$= P(\{X \ge 13\})$$

$$\approx 0.00337$$

This can be calculated by running 1 - pbinom(12, 25, 0.25) in R

Let $\Omega = (-\infty, +\infty)$ in units °C, representing the continuous range of possible temperatures in the storeroom. An argument could be made to limit this to $(-\infty, 8)$.

Let X be a random variable for the temperature inside the storeroom.

5.1 Part a)

$$X \sim N(\mu = 7.5^{\circ}C, \sigma = 0.3^{\circ}C)$$

$$P(\{7.2 < X < 8\}) = P(\{\frac{7.2 - 7.5}{0.3} < \frac{X - \mu}{\sigma} < \frac{8 - 7.5}{0.3}\})$$

$$= P(\{-\frac{2}{3} < Z < \frac{5}{3}\})$$

$$= P(\{Z < \frac{5}{3}\}) - P(\{-\frac{2}{3} < Z\})$$

$$= P(\{Z < \frac{5}{3}\}) - P(\{Z > -\frac{2}{3}\}) \text{ from (1)}$$

$$= P(\{Z < \frac{5}{3}\}) - (1 - P(\{Z < \frac{2}{3}\})) \text{ from (3)}$$

$$= P(\{Z < \frac{5}{3}\}) + P(\{Z < \frac{2}{3}\}) - 1$$

Using stats tables this equals 0.9515+0.7454-1=0.6969. Using R running pnorm $(\frac{5}{3})$ - pnorm $(-\frac{2}{3})=0.6997$.

5.2 Part b)

$$X \sim N(\mu, \sigma = 0.3^{\circ}C)$$

$$P(\{X > 8^{\circ}C\}) = 1\%$$

$$0.01 = P(\{X > 8\})$$

$$= 1 - P(\{X < 8\}) \text{ from (2)}$$

$$0.99 = P(\{X < 8\})$$

$$= P(\{\frac{X - \mu}{\sigma} = \frac{8 - \mu}{\sigma}\})$$

$$0.99 = P(\{Z = \frac{8 - \mu}{3}\})$$

Let z be the value which satisfies $P({Z < z}) = 0.99$.

$$\implies \frac{8-\mu}{3} = z$$

$$8-\mu = 3z$$

$$-\mu = 3x - 8$$

$$\mu = 8 - 3z$$

Using the stats table, $z\approx 2.33$ which implies $\mu\approx 8-3\cdot 2.33=1.01$ °C. Using R, z=qnorm(0.99) ≈ 2.326348 , which implies $\mu\approx 8-3\cdot 2.326348\approx 1.021$ °C.

5.3 Part c)

We are given no information about the parameters of X

$$X \sim N(\mu, \sigma)$$

$$\begin{split} P(\{\mu - 1^{\circ}C < X < \mu + 1^{\circ}C\}) &= 95\% \\ 0.95 &= P(\{\frac{(\mu - 1) - \mu}{\sigma} < \frac{X - \mu}{\sigma} < \frac{(\mu - 1) - \mu}{\sigma}\}) \\ &= P(\{\frac{-1}{\sigma} < Z < \frac{+1}{\sigma}\}) \\ &= P(\{\frac{-1}{\sigma} < Z\} \cap \{Z < \frac{+1}{\sigma}\}) \end{split}$$

Note that the complement of $\{\frac{-1}{\sigma} < Z\} \cap \{Z < \frac{+1}{\sigma}\}$ is $\{Z < \frac{-1}{\sigma}\} \cup \{Z > \frac{+1}{\sigma}\}$

$$0.95 = 1 - (P(\{Z < \frac{-1}{\sigma}\} \cup \{Z > \frac{+1}{\sigma}\}))$$

$$= 1 - (P(\{Z < \frac{-1}{\sigma}\}) + P(\{Z > \frac{+1}{\sigma}\}))$$

$$= 1 - 2P(\{Z < \frac{-1}{\sigma}\})$$

$$0.05 = 2P(\{Z < \frac{-1}{\sigma}\})$$

$$0.025 = P(\{Z < \frac{-1}{\sigma}\})$$

$$1 - 0.025 = 1 - P(\{Z < \frac{-1}{\sigma}\})$$

$$0.975 = P(\{Z < \frac{+1}{\sigma}\}) \text{ from (3)}$$

Let z be the solution to $0.975 = P(\{Z < z\})$

$$\implies z = \frac{+1}{\sigma}$$

$$\implies \sigma = \frac{1}{z}$$

Using the stats table, $z\approx 1.96$ which implies $\sigma\approx\frac{1}{1.96}\approx 0.510204\approx 0.51$. Using R z=qnorm $(0.975)\approx 1.959964$ which implies $\sigma\approx\frac{1}{1.959964}\approx 0.5102135\approx 0.51$.