

Introduction to Mathematical Statistics I
Extra Credit

Freya Zou

1. Determine whether the following statements are true or false. Justify your answer.

- (i) If A, B are independent, then A and B^c (the complement of B) are also independent.

True

If A, B are independent $P(A \cap B) = P(A)P(B)$

The complement of B is B^c , and we know: $P(A \cap B^c) = P(A) - P(A \cap B)$

we can substitute:

$P(A \cap B^c) = P(A) - P(A)P(B) = P(A)(1 - P(B)) = P(A)P(B^c)$.
Thus we know $P(A \cap B^c) = P(A)P(B^c)$, showing A and B^c are independent

- (ii) For a discrete random variable Y we have that $E(Y^2) \geq [E(Y)]^2$.

True:

According to variance property: $\text{Var}(Y) = E(Y^2) - [E(Y)]^2$

we know Variance is always non-negative ($\text{Var}(Y) \geq 0$).

thus we get $E(Y^2) - [E(Y)]^2 \geq 0$ which implies:

$$E(Y^2) \geq [E(Y)]^2$$

- (iii) If $Y \sim \text{Geometric}(p)$, then the random variable $Y^* = Y - 1$ has mean $\frac{1-p}{p}$.

True:

For $Y \sim \text{Geometric}(p)$, Y represents the number of trials required for the first success, and its mean is $E(Y) = \frac{1}{p}$ the number of failures before the success. Subtracting 1 from Y decreases the mean by 1:

$$E(Y^*) = E(Y) - 1 = \frac{1}{p} - 1 = \frac{1-p}{p}$$

thus, the mean of Y^* is $\frac{1-p}{p}$.

2. A quality-control program at a plastic bottle production line involves inspecting finished bottles for flaws such as microscopic holes. The probability that a bottle has a flaw is 0.002. If a bottle has a flaw, the probability that it will fail the inspection is 0.995. If a bottle does not have a flaw, the probability that it will pass the inspection is 0.990.

(i) If a bottle does not have a flaw, what is the probability that it will fail the inspection?

(ii) What is the probability that a randomly selected bottle has a flaw and fails the inspection?

(iii) If a bottle fails the inspection, what is the probability that it has a flaw?

$$\text{Given: } P(\text{flaw}) = 0.002, P(\text{fail}|\text{flaw}) = 0.995$$

$$P(\text{pass}|\text{no flaw}) = 0.990, P(\text{no flaw}) = 1 - 0.002 = 0.998$$

$$\text{i) } P(\text{fail}|\text{no flaw}) = 1 - P(\text{pass}|\text{no flaw}) = 1 - 0.990 = 0.01$$

$$\text{ii) } P(\text{flaw} \cap \text{fail}) = P(\text{fail}|\text{flaw}) \cdot P(\text{flaw}) = 0.995 * 0.002 = 0.00199$$

$$\begin{aligned} \text{iii) } P(\text{flaw}|\text{fail}) &= \frac{P(\text{fail}|\text{flaw}) P(\text{flaw})}{P(\text{fail}|\text{flaw}) P(\text{flaw}) + P(\text{fail}|\text{no flaw}) P(\text{no flaw})} \\ &= \frac{0.995 * 0.002}{0.995 * 0.002 + 0.01 * 0.998} \\ &= \frac{0.00199}{0.00199 + 0.00998} = \frac{0.00199}{0.01197} \\ &= 0.1662489 \end{aligned}$$

3. Consider an experiment consisting in rolling a fair die twice. We can represent the possible outcomes by ordered pairs. For instance $(1, 3)$ means that we obtained a 1 in the first roll and a 3 in the second roll. Since $(1, 3)$ and $(3, 1)$ are different outcomes we have total of 36 (equally likely) possible outcomes for this experiment. Define the random variable Y to be the sum of the numbers observed in the two rolls.

- (i) Write down all the possible values that the random variable Y can take.
- (ii) Find $P(Y = 6)$.
- (iii) Suppose that somebody tells you that he performed the experiment and observed the value $Y = 4$. What is the probability that he obtained a 3 in the first roll?

i) Possible values that the random variable Y can take: 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12

ii) two rolls have sample space: $6 \cdot 6 = 36$
There are 5 outcomes where $Y=6$: $(3, 3)$ $(2, 4)$ $(4, 2)$ $(1, 5)$ $(5, 1)$

$$\text{thus: } P(Y=6) = \frac{\text{Number of favourable outcomes}}{\text{Total number of outcomes}} = \frac{5}{36} \approx 0.13889$$

iii) when $Y=4$: there are 3 situations are:

$(2, 2)$ $(1, 3)$ $(3, 1)$

The first $Y=3$ has $(1, 2)$

$$3) P(A|B) = \frac{P(A \cap B)}{P(B)} = P(\text{First roll}=3 | Y=4) = \frac{P(\text{First roll}=3) \cap P(Y=4)}{P(Y=4)}$$

when $Y=4$ have 3 outcomes: $(1, 3)$, $(2, 2)$, $(3, 1)$ and we have sample space $6 \cdot 6 = 36$ thus $P(Y=4) = \frac{3}{36}$

we also can find $P(\text{first roll}=3) \cap P(Y=4)$, which means first roll is 3 and the sum is 4, this happens only one outcome $(3, 1)$, thus $P(\text{first roll}=3) \cap P(Y=4) = 1/36$

$$P(\text{first roll}=3 | Y=4) = \frac{P(\text{first roll}=3) \cap P(Y=4)}{P(Y=4)} = \frac{1/36}{3/36} = \frac{1}{3} \approx 0.333$$

4. Let Y be a random variable with pdf

$$f(y) = \begin{cases} 0.2 & , -1 < y < 0 \\ 0.2 + cy & , 0 \leq y \leq 1 \\ 0 & , \text{elsewhere} \end{cases}$$

(i) Find the cdf $F(y)$ (Note: First you need to determine the value of c)

(ii) Find $P(Y < 1/2 | Y > -1/2)$.

(iii) Find the mean and variance of Y .

$$\begin{aligned} 1) \text{ PDF} = \int_{-\infty}^{\infty} f(y) dy &= 1 = \int_{-1}^1 f(y) dy = \int_{-1}^0 0.2 dy + \int_0^1 (0.2 + cy) dy = 1 \\ &\Rightarrow = \int_{-1}^0 0.2 dy + \int_0^1 0.2 dy + \int_0^1 cy dy \\ &\Rightarrow = 0.2 + 0.2 + \frac{c}{2} = 1 \\ &\Rightarrow c = 1.2 \end{aligned}$$

now, the PDF: $f(y) = \begin{cases} 0.2 & , -1 < y < 0 \\ 0.2 + 1.2y & , 0 \leq y \leq 1 \\ 0 & , \text{elsewhere} \end{cases}$

1. For $y < -1$: $F(y) = 0$, since $f(y) = 0$ outside $[-1, 1]$.

2. For $-1 \leq y < 0$: $F(y) = \int_{-1}^y 0.2 dt = 0.2(y+1) = 0.2y + 0.2$

$$\begin{aligned} 3. \text{ For } 0 \leq y \leq 1: F(y) &= F(0) + \int_0^y (0.2 + 1.2t) dt \\ &= 0 + \int_0^y 0.2 dt + \int_0^y 1.2t dt \\ &= 0.2y + 0.6y^2 \end{aligned}$$

Therefore, $F(y) = 0.2 + 0.2y + 0.6y^2 = 0.2 + 0.2y + 0.6y^2$

4. For $y > 1$: $F(y) = 1$. since probability is 1.

thus. CDF for F_y is:

$$F(y) = \begin{cases} 0 & y < -1 \\ 0.2y + 0.2 & -1 \leq y < 0 \\ 0.2 + 0.2y + 0.6y^2 & 0 \leq y \leq 1 \\ 1 & y > 1 \end{cases}$$

$$ii) P(Y < \frac{1}{2} | Y > -\frac{1}{2}) = \frac{P(-\frac{1}{2} < Y < \frac{1}{2})}{P(Y > -\frac{1}{2})}$$

$$= \frac{F(\frac{1}{2}) - F(-\frac{1}{2})}{1 - F(-\frac{1}{2})}$$

no we can substitute, $y = -1/2$, using $F(y) = 0.2y + 0.2(-1 \leq y < 0)$:

$$F(-\frac{1}{2}) = 0.2(-\frac{1}{2}) + 0.2 = 0.1 + 0.2 = 0.1$$

and when $y = \frac{1}{2}$ we can substitute in $F(y) = 0.2 + 0.2y + 0.6y^2 (0 \leq y \leq 1)$

$$\text{then we can get: } F(\frac{1}{2}) = 0.2 + 0.2(\frac{1}{2}) + 0.6(\frac{1}{2})^2 = 0.2 + 0.1 + 0.6(0.25) = 0.45$$

We also need to find $1 - F(-\frac{1}{2}) = 1 - 0.1 = 0.9$ thus we know:

$$P(Y < \frac{1}{2} | Y > -\frac{1}{2}) = \frac{F(\frac{1}{2}) - F(-\frac{1}{2})}{1 - F(-\frac{1}{2})} = \frac{0.45 - 0.1}{0.9} \approx 0.39$$

iii)

$$\begin{aligned} \mu = E(Y) &= \int_{-1}^1 y f(y) dy = \int_{-1}^0 y(0.2) dy + \int_0^1 y(0.2 + 1.2y) dy \\ &= 0.2 \int_{-1}^0 y dy + \int_0^1 0.2y dy + \int_0^1 1.2y^2 dy \\ &= -0.1 + 0.1 + 0.4 \\ &= 0.4 \end{aligned}$$

$$\begin{aligned} E(Y^2) &= \int_{-1}^0 y^2(0.2) dy + \int_0^1 y^2(0.2 + 1.2y) dy \\ &= -0.05 + 0.36666 = 0.31666 \end{aligned}$$

$$\text{var}(Y) = E(Y^2) - [E(Y)]^2 = 0.31666 - (0.4)^2 \approx 0.15666$$

thus we get mean is 0.4 and variance is 0.15666

5. Suppose that $Y \sim \text{Gamma}(\alpha, \beta)$. That is, Y has pdf

$$f(y) = \begin{cases} \frac{y^{\alpha-1} e^{-y/\beta}}{\Gamma(\alpha) \beta^\alpha} & , 0 < y < \infty \\ 0 & , \text{otherwise} \end{cases}$$

where $\alpha, \beta > 0$.

- (i) Verify that $\int_{-\infty}^{\infty} f(y) dy = 1$. (Hint: Recall the definition of a gamma function given in class)
- (ii) Without using moment generating functions, find $E(Y)$ and $\text{Var}(Y)$.
- (iii) Find the moment generating function of Y and use it to verify your results in part (ii).

i) since $f(y)$ is defined as zero for $y \leq 0$, we only need to verify the integral over $(0, \infty)$:

$$\begin{aligned} \int_0^{\infty} f(y) dy &= \int_0^{\infty} \frac{y^{\alpha-1} e^{-y/\beta}}{\Gamma(\alpha) \beta^\alpha} dy \\ &= \frac{1}{\Gamma(\alpha) \beta^\alpha} \int_0^{\infty} y^{\alpha-1} e^{-y/\beta} dy \end{aligned}$$

Now we can substitute $u = \frac{y}{\beta}$, $y = \beta u$ and $dy = \beta du$:

$$\begin{aligned} \int_0^{\infty} y^{\alpha-1} e^{-y/\beta} dy &= \int_0^{\infty} (\beta u)^{\alpha-1} e^{-u} \beta du \\ &= \beta^\alpha \int_0^{\infty} u^{\alpha-1} e^{-u} du \end{aligned}$$

Now we substitute back into original the original expression:

$$\int_0^{\infty} f(y) dy = \frac{1}{\Gamma(\alpha) \beta^\alpha} \cdot \beta^\alpha \int_0^{\infty} u^{\alpha-1} e^{-u} du.$$

$$\begin{aligned} &= \frac{1}{\Gamma(\alpha)} \boxed{\int_0^{\infty} u^{\alpha-1} e^{-u} du} = \frac{\Gamma(\alpha)}{\Gamma(\alpha)} \\ &= \frac{\Gamma(\alpha)}{\Gamma(\alpha)} = 1 \end{aligned}$$

by the definition of the Gamma distribution

Therefore: $\int_0^{\infty} f(y) dy = \frac{\Gamma(\alpha)}{\Gamma(\alpha)} = 1$, this verifies that $f(y)$ is a valid probability density function

ii) Mean $E(Y) = \alpha\beta$:

The expected value of a random variable Y is given by:

$$E(Y) = \int_0^{\infty} y f(y) dy \quad \text{where } f(y) \text{ is Gamma distribn.}$$

$$f(y) = \frac{y^{\alpha-1} e^{-y/\beta}}{\Gamma(\alpha) \beta^{\alpha}}, \quad \text{substituting the PDF into the formula for } E(Y):$$

$$E(Y) = \int_0^{\infty} y \frac{y^{\alpha-1} e^{-y/\beta}}{\Gamma(\alpha) \beta^{\alpha}} dy = \frac{1}{\Gamma(\alpha) \beta^{\alpha}} \int_0^{\infty} y^{\alpha} e^{-y/\beta} dy$$

Change of variable: let $u = \frac{y}{\beta}$ so $y = \beta u$ and $dy = \beta du$:

$$E(Y) = \frac{1}{\Gamma(\alpha) \beta^{\alpha}} \int_0^{\infty} (\beta u)^{\alpha} e^{-u} du$$

$$= \frac{\beta^{\alpha+1}}{\Gamma(\alpha) \beta^{\alpha}} \int_0^{\infty} u^{\alpha} e^{-u} du$$

$$= \frac{\beta}{\Gamma(\alpha)} \int_0^{\infty} u^{\alpha} e^{-u} du$$

Since the integral of $u^{\alpha} e^{-u}$ over u from $[0, \infty]$ equals $\Gamma(\alpha+1)$:

$$E(Y) = \frac{\beta \Gamma(\alpha+1)}{\Gamma(\alpha)} = \alpha\beta$$

variance $\text{Var}(Y)$:

$$E(Y^2) = \int_0^{\infty} y^2 f(y) dy = \int_0^{\infty} y^2 \frac{y^{\alpha-1} e^{-y/\beta}}{\Gamma(\alpha) \beta^{\alpha}} dy$$
$$= \frac{1}{\Gamma(\alpha) \beta^{\alpha}} \int_0^{\infty} y^{\alpha+1} e^{-y/\beta} dy$$

Change of variable, $u = y/\beta$, $y = \beta u$, $dy = \beta du$:

$$E(Y^2) = \frac{\beta^{\alpha+2}}{\Gamma(\alpha)} \int_0^{\infty} (\beta u)^{\alpha+1} e^{-u} \beta du = \frac{\beta^{\alpha+2} \Gamma(\alpha+2)}{\Gamma(\alpha)}$$

$$\Rightarrow \alpha(\alpha+1) \beta^2 \quad (\text{using the property } \Gamma(\alpha+2) = (\alpha+1)\alpha \Gamma(\alpha))$$

$$\text{Var}(Y) = \alpha(\alpha+1) \beta^2 - (\alpha\beta)^2 = \alpha\beta^2$$

$$\text{thus, } \text{Var}(Y) = \alpha\beta^2$$

iii) The moment generating function (MGF) of a Gamma random variable $Y \sim \text{Gamma}(\alpha, \beta)$ is given by $M_Y(t) = E(e^{tY})$

then MGF is: $M_Y(t) = (1 - \beta t)^{-\alpha}$, for $t < \frac{1}{\beta}$

Verifying $E(Y)$ and $\text{Var}(Y)$ using the MGF:

1. Mean $E(Y)$:

The mean $E(Y)$ can be found by differentiating the MGF with respect to t and evaluating at $t=0$:

$$E(Y) = M'_Y(0)$$

differentiate $M_Y(t) = (1 - \beta t)^{-\alpha}$ with respect to t :

$$M'_Y(t) = \alpha \beta (1 - \beta t)^{-\alpha-1}$$

Substitute $t=0$:

$$M'_Y(t) = \alpha \beta$$

so, $E(Y) = \alpha \beta$, which agrees with the result from part (ii)

2. Variance $\text{Var}(Y)$:

To find $\text{Var}(Y)$, we need $E(Y^2) = M''_Y(0)$ and then use

$$\text{Var}(Y) = E(Y^2) - [E(Y)]^2$$

differentiate $M_Y(t) = (1 - \beta t)^{-\alpha}$ with respect to t :

$$M''_Y(t) = \alpha \beta (\alpha + 1) \beta (1 - \beta t)^{-\alpha-2} = \alpha (\alpha + 1) \beta^2 (1 - \beta t)^{-\alpha-2}$$

Substitute $t=0$: $M''_Y(0) = \alpha (\alpha + 1) \beta^2$, thus $E(Y^2) = \alpha (\alpha + 1) \beta^2$

$$\begin{aligned} \text{Now, } \text{Var}(Y) &= E(Y^2) - [E(Y)]^2 = \alpha (\alpha + 1) \beta^2 - (\alpha \beta)^2 \\ &= \alpha \beta^2 \end{aligned}$$

This confirms that $\text{Var}(Y) = \alpha \beta^2$, which matches the result in part (i)