

1. Your Name: Answer Key

Consider the infinitely many independent, identical experiments of throwing a pair of dice and the outcome of each experiment is the sum of the two numbers. Let $N \geq 1$ be the number of experiments such that the number 5 or 7 first appears as the outcome. Let further X be that number at the N -th experiment, i.e. X can be either a 5 or 7.

- a) Find $P(N = n)$ for $n \geq 1$.
 b) Find $P(X = 5)$.
 c) Prove or disprove that N and X are independent.

[Ross, p. 50 #25;
 p. 79 #4h
 p. 104 #76;
 p. 272, #24]

$$(a) P(5) = P((1,4), (2,3), (3,2), (4,1)) = \frac{4}{36} = \frac{1}{9}$$

$$P(7) = P((1,6), (2,5), \dots, (6,1)) = \frac{6}{36} = \frac{1}{6}$$

$$P(\text{no 5, no 7}) = 1 - \frac{1}{9} - \frac{1}{6} = \frac{26}{36}$$

$$P(N=n) = P\left(\underbrace{\text{no 5, no 7}}_{n-1 \text{ times}} \text{ then } \underbrace{5 \text{ or } 7}_n\right)$$

$$= (1 - \frac{1}{9} - \frac{1}{6})^{n-1} (\frac{1}{9} + \frac{1}{6}) = \left(\frac{26}{36}\right)^{n-1} \left(\frac{10}{36}\right)$$

$$(b) P(X=5) = \sum_{n=1}^{\infty} P(N=n, X=5)$$

$$= \sum_{n=1}^{\infty} P\left(\underbrace{\text{no 5, no 7}}_{n-1 \text{ times}} \text{ then } \underbrace{5}_n\right)$$



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$$\begin{aligned}
 &= \sum_{n=1}^{\infty} (1-p_5-p_7)^{n-1} p_5 = \left[\sum_{n=1}^{\infty} (1-p_5-p_7)^{n-1} \right] p_5 \\
 &= p_5 \frac{1}{1-(1-p_5-p_7)} = \boxed{\frac{p_5}{p_5+p_7}}
 \end{aligned}$$

$$(c) \quad P(N=n) P(X=5) = (1-p_5-p_7)^{n-1} (p_5+p_7) \times \frac{p_5}{p_5+p_7}$$

$$\parallel \quad = (1-p_5-p_7)^{n-1} p_5$$

$$P(N=n, X=5) = P\left(\begin{array}{c} n-1 \\ \text{no } 5, \text{ no } 7 \end{array} \mid \begin{array}{c} n \\ 5 \end{array}\right)$$

$$= (1-p_5-p_7)^{n-1} p_5$$

Hence N & X are independent.



2. Your Name: _____

Consider a city in which the male and female drivers occupy α and $1 - \alpha$ fractions of the whole city driver population. In any given year, a male and female driver will have an accident with probability p_M and p_F . Assume that the behavior of each driver is independent from year to year.

Now a driver is randomly chosen. Let A_i be the event that this driver will have an accident in the i -th year. Let M be the event that the randomly chosen driver is a male.

a) Suppose $p_M > p_F$. Show that $P(M|A_1) > p(M)$.

b) Suppose $p_M \neq p_F$. Show that $P(A_2|A_1) > p(A_1)$.

[Ross p.101, #6]

$$(a) \quad P(M) = \alpha$$

$$P(M|A_1) = \frac{P(M \cap A_1)}{P(A_1)} = \frac{P(A_1|M)P(M)}{P(A_1)}$$

$$= \frac{p_M \alpha}{p_M \alpha + p_F (1-\alpha)}$$

$$P(A_1) = P(A_1|M)P(M) + P(A_1|F)P(F)$$

$$P(M|A_1) - P(M)$$

$$= \frac{p_M \alpha}{p_M \alpha + p_F (1-\alpha)} - \alpha = \frac{p_M \alpha - p_M \alpha^2 - p_F \alpha (1-\alpha)}{p_M \alpha + p_F (1-\alpha)}$$

$$= \frac{\alpha (1-\alpha) (p_M - p_F)}{p_M \alpha + p_F (1-\alpha)} > 0 \quad (\because p_M > p_F)$$



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$$(b) \quad P(A_2|A_1) - P(A_1)$$

$$= \frac{P(A_2 \cap A_1) - P(A_1)}{P(A_1)}$$

$$= \frac{P(A_2 \cap A_1|M)P(M) + P(A_2 \cap A_1|F)P(F) - P(A_1)}{P(A_1|M)P(M) + P(A_1|F)P(F)}$$

$$= \frac{P_M^2 \alpha + P_F^2 (1-\alpha) - [P_M \alpha + P_F (1-\alpha)]^2}{P(A_1)}$$

Numerator

$$= P_M^2 \alpha + P_F^2 (1-\alpha) - P_M^2 \alpha^2 - P_F^2 (1-\alpha)^2 - 2P_M P_F \alpha (1-\alpha)$$

$$= P_M^2 (\alpha - \alpha^2) + P_F^2 [(1-\alpha) - (1-\alpha)^2] - 2P_M P_F \alpha (1-\alpha)$$

(1-\alpha)(1-(1-\alpha))

$$= \alpha(1-\alpha) [P_M^2 + P_F^2 - 2P_M P_F]$$

$$= \alpha(1-\alpha) (P_M - P_F)^2 > 0 \quad (P_M \neq P_F)$$



3. Your Name: _____

Let X_1, X_2, \dots, X_n be a collection of iid exponential random variables with parameter λ . Let

$$\begin{aligned} Y_1 &= X_1, \\ Y_2 &= X_1 + X_2, \\ Y_3 &= X_1 + X_2 + X_3, \\ &\dots \\ Y_n &= X_1 + \dots + X_n. \end{aligned}$$

[Ross, p. 265,
example 7e]

Find the joint pdf $p(y_1, y_2, \dots, y_n)$ of Y_1, Y_2, \dots, Y_n .

$$P_Y(y_1, y_2, \dots, y_n) = \frac{P_X(x_1, x_2, \dots, x_n)}{\left| \frac{dy_1 \dots dy_n}{dx_1 \dots dx_n} \right|}$$

$$\begin{aligned} P_X(x_1, x_2, \dots, x_n) &= (\lambda e^{-\lambda x_1})(\lambda e^{-\lambda x_2}) \dots (\lambda e^{-\lambda x_n}) \\ &= \lambda^n e^{-\lambda(x_1 + x_2 + \dots + x_n)} \\ &= \lambda^n e^{-\lambda y_n} \end{aligned}$$

$$\left| \frac{dy_1 \dots dy_n}{dx_1 \dots dx_n} \right| = \det \begin{vmatrix} 1 & 0 & \dots & 0 \\ 1 & 1 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 1 & 1 & \dots & 1 \end{vmatrix} = 1$$

Hence $P_Y(y_1, \dots, y_n) = \lambda^n e^{-\lambda y_n}$



4. Your Name: _____

The pdf $p(x)$ of the Gamma distribution with parameter $(\alpha > 0, \lambda > 0)$ is given by:

$$p(x) = \begin{cases} \frac{\lambda e^{-\lambda x} (\lambda x)^{\alpha-1}}{\Gamma(\alpha)}, & x \geq 0, \\ 0 & x < 0. \end{cases}$$

Let X and Y be independent Gamma distributed random variables with parameters (α, λ) and (β, λ) . Show analytically that $X + Y$ has a Gamma distribution with parameter $(\alpha + \beta, \lambda)$.

Show how as a by-product that the above conclusion leads to the following integration identity for $\alpha, \beta > 0$:

$$\int_0^1 x^{\alpha-1} (1-x)^{\beta-1} dx = \frac{\Gamma(\alpha)\Gamma(\beta)}{\Gamma(\alpha+\beta)}.$$

(You are welcome to prove the above identity by any other method.)

$$Z = X + Y$$

[Ross, p. 242, Proposition 3.1]

$$p_Z(z) = \int_0^z p_\alpha(x) p_\beta(z-x) dx$$

$$= \int_0^z \frac{\lambda e^{-\lambda x} (\lambda x)^{\alpha-1}}{\Gamma(\alpha)} \frac{\lambda e^{-\lambda(z-x)} (\lambda(z-x))^{\beta-1}}{\Gamma(\beta)} dx$$

$$= \frac{\lambda^{\alpha+\beta-1} e^{-\lambda(x+z-x)}}{\Gamma(\alpha)\Gamma(\beta)} \int_0^z x^{\alpha-1} (z-x)^{\beta-1} dx$$

$$\text{let } u = \frac{x}{z}$$



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$$= \frac{\lambda^{\alpha+\beta} e^{-\lambda z}}{\Gamma(\alpha)\Gamma(\beta)} \int_0^1 (zw)^{\alpha-1} z^{\beta-1} (1-w)^{\beta-1} z dw$$

$$= \frac{\lambda^{\alpha+\beta} e^{-\lambda z} z^{\alpha-1+\beta-1}}{\Gamma(\alpha)\Gamma(\beta)} \int_0^1 w^{\alpha-1} (1-w)^{\beta-1} dw$$

$$= \left[\frac{\int_0^1 w^{\alpha-1} (1-w)^{\beta-1} dw}{\Gamma(\alpha)\Gamma(\beta)} \right] \lambda e^{-\lambda z} (\lambda z)^{\alpha+\beta-1}$$

↗
a constant.

↗
in the form of $\text{Gamma}(\alpha+\beta, \lambda)$

Hence in order for $p_Z(z)$ to be a pdf, $\int_0^\infty p_Z(z) dz = 1$

Thus the constant must be $\frac{1}{\Gamma(\alpha+\beta)}$.

i.e. $\int_0^1 w^{\alpha-1} (1-w)^{\beta-1} dw = \frac{\Gamma(\alpha)\Gamma(\beta)}{\Gamma(\alpha+\beta)}$



5. Your Name: _____

Let X_1, \dots, X_n be a collection of iid random variables with expectations and variances equal to μ and σ^2 . Define the "sample mean" \bar{X} and "sample variance" S^2 as

$$\bar{X} = \frac{1}{n} (X_1 + \dots + X_n), \quad S^2 = \frac{1}{n} \sum_{i=1}^n (X_i - \bar{X})^2$$

Compute $\text{Var}(\bar{X})$ and $E(S^2)$.

[Ross, p. 307, Example 4a]

$$\text{Var}(\bar{X}) = \text{Var}\left(\frac{X_1 + \dots + X_n}{n}\right) = \frac{1}{n^2} \text{Var}(X_1 + \dots + X_n)$$

$$= \frac{1}{n^2} [\text{Var}(X_1) + \dots + \text{Var}(X_n)] \quad (\because X_1, X_2, \dots, X_n \text{ independent})$$

$$= \frac{1}{n} \text{Var}(X_1) = \frac{\sigma^2}{n}$$

$$S^2 = \frac{1}{n} \sum_{i=1}^n (X_i - \bar{X})^2$$

~~Method 1~~

$$E(S^2) = \frac{1}{n} \sum_{i=1}^n E(X_i - \bar{X})^2$$

$$= \frac{1}{n} \sum_{i=1}^n E(X_i^2) - 2E$$



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Method 1
$$S^2 = \frac{1}{n} \sum_{i=1}^n (X_i^2 - 2X_i\bar{X} + \bar{X}^2)$$

$$= \frac{1}{n} \left[\sum_{i=1}^n X_i^2 - 2 \left(\sum_{i=1}^n X_i \right) \bar{X} + \sum_{i=1}^n \bar{X}^2 \right]$$

$$= \frac{1}{n} \left[\sum_{i=1}^n X_i^2 - 2 \overset{= (n\bar{X})}{\left(\sum_{i=1}^n X_i \right)} \bar{X} + n \bar{X}^2 \right]$$

$$= \frac{1}{n} \left[\sum_{i=1}^n X_i^2 - n(\bar{X})^2 \right] \quad (\text{Var}(X) = EX^2 - (EX)^2)$$

$$\text{Var}(S^2) = \frac{1}{n} \left[n(\text{Var}(X_1)^2 + \mu^2) - n E[(\bar{X})^2] \right]$$

$$= \sigma^2 + \mu^2 - E[(\bar{X})^2] \quad (E\bar{X} = \mu)$$

$$= \sigma^2 + \mu^2 - \text{Var}(\bar{X}) - (E\bar{X})^2$$

$$= \sigma^2 + \mu^2 - \frac{\sigma^2}{n} - \mu^2 = \sigma^2 - \frac{\sigma^2}{n} = \boxed{\left(\frac{n-1}{n} \right) \sigma^2}$$



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Method 2

$$\begin{aligned} S^2 &= \frac{1}{n} \sum_{i=1}^n (X_i - \bar{X})^2 \\ &= \frac{1}{n} \sum_{i=1}^n \left(X_i - \frac{X_1 + \dots + X_n}{n} \right)^2 \\ &= \frac{1}{n^3} \sum_{i=1}^n \left(nX_i - (X_1 + \dots + X_n) \right)^2 \\ &= \frac{1}{n^3} \sum_{i=1}^n \left((n-1)X_i - \sum_{j \neq i} X_j \right)^2 \\ &= \frac{1}{n^3} \sum_{i=1}^n \left[(n-1)(X_i - \mu) - \sum_{j \neq i}^{n-1 \text{ so many}} (X_j - \mu) \right]^2 \end{aligned}$$

Have

$$\begin{aligned} ES^2 &= \frac{1}{n^3} \sum_{i=1}^n E \left(\left[(n-1)(X_i - \mu) - \sum_{j \neq i} (X_j - \mu) \right]^2 \right) \\ &= \frac{1}{n^3} \sum_{i=1}^n \left\{ (n-1)^2 E[(X_i - \mu)^2] + \sum_{j \neq i} E[(X_j - \mu)^2] \right. \\ &\quad \left. + E(\text{cross terms}) \rightarrow 0 \right\} \end{aligned}$$



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(The cross ~~terms~~ terms are zero as X_i, X_j are ind.)

$$= \frac{1}{n^3} n \left\{ (n-1)^2 \sigma^2 + (n-1) \sigma^2 \right\}$$

$$= \frac{n(n-1)(\cancel{n-1+1})}{n^3} \sigma^2$$

$$= \boxed{\left(\frac{n-1}{n} \right) \sigma^2}$$



6. Your Name: _____

(Estimation of the length of an interval.) Let $L > 0$ be some unknown but fixed length. Let X_1, X_2, \dots be a sequence of iid random variables uniformly distributed on $[0, L]$. The goal is to use the X_i 's to estimate L .

a) Let $A_n = 2 \frac{X_1 + \dots + X_n}{n}$. Show that A_n is an unbiased estimator in the sense that $E(A_n) = L$.

b) Let $B_n = \gamma_n \max\{X_1, X_2, \dots, X_n\}$ where γ_n is some number. Find the correct value of γ_n such that B_n is also an unbiased estimator, i.e. $E(B_n) = L$.

(Hint: find the distribution of B_n first.)

c) Find $\text{Var}(A_n)$ and $\text{Var}(B_n)$.

d) Which estimator is "more superior"?

[Weiman:
Statistics, view toward
Applications,
p. 55]

$$(a) \quad A_n = 2 \left(\frac{X_1 + \dots + X_n}{n} \right)$$

$$E A_n = \frac{2}{n} n E(X_1) = \frac{2}{n} n \frac{L}{2} = L$$

$$(b) \quad B_n = \gamma_n \underbrace{\max(X_1, \dots, X_n)}_{M_n}$$

$$\begin{aligned} P(M_n \leq t) &= P(X_1 \leq t) P(X_2 \leq t) \dots P(X_n \leq t) \\ &= \left(\frac{t}{L} \right)^n \end{aligned}$$

$$P(t) = n t^{n-1} / L^n$$



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$$\begin{aligned}
 E(B_n) &= \gamma_n E(M_n) = \gamma_n \int_0^L t \frac{n}{L^n} t^{n-1} dt \\
 &= \frac{n\gamma_n}{L^n} \int_0^L t^n dt = \frac{n}{n+1} \gamma_n L = L
 \end{aligned}$$

Hence $\gamma_n = \frac{n+1}{n}$

$$(c) \text{Var}(A_n) = \text{Var}\left(\frac{X_1 + \dots + X_n}{n} \times 2\right)$$

$$= \frac{2^2}{n^2} n \text{Var}(X_1) = \frac{4}{n} \frac{L^2}{12} = \frac{L^2}{3n}$$

$$\text{Var}(B_n) = \text{Var}(\gamma_n M_n) = \gamma_n^2 \text{Var}(M_n)$$

$$= \gamma_n^2 [E(M_n^2) - (E M_n)^2]$$



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$$E(M_n^2) = \int_0^L t^2 \frac{n t^{n-1}}{L^n} dt = \frac{n}{L^n} \int_0^L t^{n+1} dt$$

$$= \left(\frac{n}{n+2} \right) L^2$$

$$(E M_n)^2 = \left(\frac{n}{n+1} \right)^2 L^2$$

$$\text{Var}(B_n) = \gamma_n^2 \left[\left(\frac{n}{n+2} \right) - \frac{n^2}{(n+1)^2} \right] L^2$$

$$= L^2 \gamma_n^2 \left[\frac{n(n+1)^2 - n^2(n+2)}{(n+2)(n+1)^2} \right]$$

$$= L^2 \frac{(n+1)^2}{n^2} \left[\frac{\cancel{n^3} + 2\cancel{n^2} + \cancel{n} - \cancel{n^3} - 2\cancel{n^2}}{(n+2)(n+1)^2} \right]$$

$$= \frac{L^2}{n(n+2)}$$



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$$1d) \text{Var}(A_n) = \frac{L^2}{3n} = O\left(\frac{1}{n}\right) \xrightarrow{n \rightarrow \infty} 0$$

$$\text{Var}(B_n) = \frac{L^2}{n(n+2)} = O\left(\frac{1}{n^2}\right) \xrightarrow{n \rightarrow \infty} 0$$

~~Both~~ Both $\text{Var}(A_n)$ & $\text{Var}(B_n)$ go to zero as $n \rightarrow \infty$ but $\text{Var}(B_n)$ goes to zero at a faster rate ($O(\frac{1}{n^2}) \ll O(\frac{1}{n})$)

Hence B_n is more superior.

(Variance measures fluctuations.)