

A.13.1.

$$\text{Cov}(X, a + bX + cX^2) = \text{Cov}(X, a) + \text{Cov}(X, bX) + \text{Cov}(X, cX^2)$$

$$= 0 + b \text{Var}(X) + c \text{Cov}(X, X^2)$$

$$= b(E(X^2) - \underbrace{\{E(X)\}^2}_{=0}) + c(\underbrace{E(X^3)}_{=0} - \underbrace{E(X)E(X^2)}_{=0})$$

$$= b$$

$$\text{Var}(X) = 1.$$

$$\text{Var}(Y) = \text{Var}(a + bX + cX^2) = \text{Var}(bX + cX^2)$$

$$= b^2 \text{Var}(X) + c^2 \text{Var}(X^2) + 2bc \underbrace{\text{Cov}(X, X^2)}_{=0 \text{ from above calculation}}$$

$$= b^2 + c^2(E(X^4) - \{E(X^2)\}^2)$$

$$= b^2 + c^2(3 - 1) = b^2 + 2c^2.$$

$$\rho_{XY} = \frac{\text{Cov}(X, Y)}{\sqrt{\text{Var}(X) \text{Var}(Y)}} = \frac{b}{\sqrt{b^2 + 2c^2}}.$$

A.13.2

Let Y = arrival time of the professor. We are told
(a) $Y \sim \text{uniform}(0, 4)$ where $Y=0$ defines 9 am. and $Y=4$ defines

1 pm. Given $Y=y$, the duration of the task is X where

$X|Y=y \sim \text{exponential}(\frac{1}{5-y})$. Consequently, $E(X|Y=y) = 5-y$

since the mean of an exponential(λ) = $\frac{1}{\lambda}$. Therefore,
the expected duration is (by the law of total expectation)

$$E(X) = E(E(X|Y)) = E(5-Y) = 5 - E(Y) = 5 - 2 = 3,$$

since the mean of a uniform(a, b) is the midpoint of the interval.

A.13.2 (continued)

$$(b) E(X+Y) = E(X) + E(Y) = 3 + 2 = 5.$$

A.13.3

$$M_S(t) = E(e^{tS}) = E(E(e^{tS} | N))$$

$$= \sum_{n=1}^{\infty} E(e^{tS} | N=n) P(N=n) \quad \text{by the Law of total expectation.}$$

Now, given $N=n$, $S = \sum_{i=1}^n X_i$ is independent of N and, therefore,

$$M_S(t) = \sum_{n=1}^{\infty} E(e^{t \sum_{i=1}^n X_i}) \cdot (1-q)^{n-1} q = \sum_{n=1}^{\infty} (M_{X_1}(t))^n (1-q)^{n-1} q$$

$$= \sum_{n=1}^{\infty} \left(\frac{p e^t}{1 - (1-p)e^t} \right)^n (1-q)^{n-1} q = \frac{q}{1-q} \sum_{n=1}^{\infty} \left(\frac{(1-q)p e^t}{1 - (1-p)e^t} \right)^n$$

$$= \frac{q}{1-q} \cdot \frac{\frac{(1-q)p e^t}{1 - (1-p)e^t}}{1 - \frac{(1-q)p e^t}{1 - (1-p)e^t}} = \frac{q}{1-q} \cdot \frac{(1-q)p e^t}{(1 - (1-p)e^t) - (1-q)p e^t}$$

$$= \frac{q p e^t}{1 - (1-qp)e^t} \quad \text{is the mgf of } S, \text{ which is the}$$

mgf of a geometric(qp) and therefore, the (unconditional) distribution of S is geometric(qp).

A.13.4

(a) We know $P(X_1 = i) = \frac{1}{6}$ for each $i = 1, 2, 3, 4, 5, 6$.

and $P(X_2 + X_3 = j) = \frac{6 - |7 - j|}{36}$ for each $j = 2, 3, 4, \dots, 12$.

Therefore, by the law of total probability

$$P(X_1 + X_2 + X_3 = 9) = \sum_{i=1}^6 P(X_1 + X_2 + X_3 = 9 | X_1 = i) P(X_1 = i)$$

$$= \sum_{i=1}^6 P(X_2 + X_3 = 9 - i | X_1 = i) \cdot \frac{1}{6} = \frac{1}{6} \sum_{i=1}^6 P(X_2 + X_3 = 9 - i)$$

Since $X_2 + X_3$ is indep.
of X_1 .

$$= \frac{1}{6} \left(P(X_2 + X_3 = 8) + P(X_2 + X_3 = 7) + P(X_2 + X_3 = 6) + P(X_2 + X_3 = 5) \right. \\ \left. + P(X_2 + X_3 = 4) + P(X_2 + X_3 = 3) \right)$$

$$= \frac{1}{6} \left(\frac{5}{36} + \frac{6}{36} + \frac{5}{36} + \frac{4}{36} + \frac{3}{36} + \frac{2}{36} \right) = \frac{25}{216} \approx .1157.$$

(b) If we naively apply the Central Limit theorem using

$$\mu = E(X_1) = 3.5 \text{ and } \sigma^2 = \text{Var}(X_1) = \frac{35}{12} \Rightarrow \sigma = 1.707825 \dots$$

then

$$P(8.5 \leq X_1 + X_2 + X_3 \leq 9.5) = P\left(\frac{8.5 - 3(3.5)}{1.707825\sqrt{3}} \leq \frac{S_3 - n\mu}{\sigma\sqrt{n}} \leq \frac{9.5 - 3(3.5)}{1.707825\sqrt{3}} \right)$$

$$= P(-.68 \leq \frac{S_n - n\mu}{\sigma\sqrt{n}} \leq .34) \approx \Phi(-.34) - \Phi(-.68) =$$

$$\Phi(.68) - \Phi(.34) = .1186.$$

Note that this is a fairly good approximation!

A.13.5 First of all

$$\mu = E(X_1) = \int_0^1 x(2x^3 - 2x + 1) dx = \int_0^1 2x^4 - 2x^2 + x dx = \frac{2}{5} - \frac{2}{3} + \frac{1}{2} = \frac{7}{30}.$$

$$E(X_1^2) = \int_0^1 x^2(2x^3 - 2x + 1) dx = \int_0^1 2x^5 - 2x^3 + x^2 dx = \frac{2}{6} - \frac{1}{2} + \frac{1}{3} = \frac{1}{6}.$$

$$\sigma^2 = E(X_1^2) - \{E(X_1)\}^2 = \frac{1}{6} - \left(\frac{7}{30}\right)^2 = \frac{101}{900} \Rightarrow \sigma = .334996 \dots$$

$$\begin{aligned} P(\bar{X} > \frac{1}{3}) &= P\left(\frac{S_n}{n} > \frac{1}{3}\right) = P\left(\frac{\frac{S_n}{n} - \mu}{\frac{\sigma}{\sqrt{n}}} > \frac{\frac{1}{3} - \frac{7}{30}}{\frac{.334996}{\sqrt{90}}}\right) \\ &= P\left(\frac{\bar{X} - \mu}{\frac{\sigma}{\sqrt{n}}} > 2.83\right) \approx 1 - \Phi(2.83) = .0023. \end{aligned}$$

A.13.6.

$$\begin{aligned} M_X(t) &= E(e^{tX}) = \int_0^\infty e^{tx} e^{-x} dx = \int_0^\infty e^{-x(1-t)} dx < \infty \text{ when } t < 1. \\ &= \left[-\frac{e^{-x(1-t)}}{1-t} \right]_{x=0}^\infty = \frac{1}{1-t} = (1-t)^{-1}. \end{aligned}$$

Let X_1, X_2, \dots be independent $\text{Exp}(1)$, so that $\mu = 1$ and $\sigma = 1$.

$$\begin{aligned} M_{Y_n}(t) &= E(e^{tY_n}) = E\left(e^{t\left(\frac{S_n - n\mu}{\sigma\sqrt{n}}\right)}\right) = E\left(e^{t\left(\frac{S_n - n}{\sqrt{n}}\right)}\right) = e^{-\frac{tn}{\sqrt{n}}} E\left(e^{\frac{t}{\sqrt{n}} S_n}\right) \\ &= e^{-\frac{tn}{\sqrt{n}}} \left(1 - \frac{t}{\sqrt{n}}\right)^{-n} = \left(e^{\frac{t}{\sqrt{n}}} \left(1 - \frac{t}{\sqrt{n}}\right)\right)^{-n} = \left(\left(1 + \frac{t}{\sqrt{n}} + \frac{t^2}{2n} + \dots\right) \left(1 - \frac{t}{\sqrt{n}}\right)\right)^{-n} \\ &= \left(1 + \cancel{\frac{t}{\sqrt{n}}} + \frac{t^2}{2n} + \dots - \cancel{\frac{t}{\sqrt{n}}} - \frac{t^2}{n} - \dots\right)^{-n} = \left(1 - \frac{t^2}{2n} + \dots\right)^{-n} \quad \text{terms having } n^x \text{ in denominator with } x > 1. \\ &= \left\{ \left(1 - \frac{t^2}{2n} + \dots\right)^n \right\}^{-1} \quad \text{So that } \lim_{n \rightarrow \infty} \left\{ \left(1 - \frac{t^2}{2n} + \dots\right)^n \right\}^{-1} = \left\{ e^{-t^2/2} \right\}^{-1} = e^{t^2/2} \end{aligned}$$

thus, Y_n has a distribution converging to that of a standard normal.

A.13.6 (continued)

$$\begin{aligned} P(S_{100} > 120) &= P\left(\frac{S_{100} - 100}{\sqrt{100}} > \frac{120 - 100}{10}\right) \\ &= P\left(\frac{S_{100} - 100}{10} > 2\right) \approx 1 - \Phi(2) = .0228. \end{aligned}$$