## Solutions 12

1. (a) Since  $X \sim U(0.30)$  we have:

$$F(x) = \int_0^x \frac{1}{30 - 0} ds = \frac{s}{30} \bigg|_0^x = \frac{x}{30} \mathcal{I}_{[0,30]}(x) + \mathcal{I}_{(30,\infty)}(x).$$

(b)

$$P(X > 10) = 1 - P(X \le 10) = 1 - F(10) = 1 - \frac{10}{30} = \frac{2}{3}.$$

(c) Use theorem 1.10 to obtain

$$P(X \ge 20 | X \ge 15) = \frac{P(X \ge 20 \cap X \ge 15)}{P(X \ge 15)} = \frac{P(X \ge 20)}{P(X \ge 15)} = \frac{1 - F(20)}{1 - F(15)}$$
$$= \frac{1 - \frac{20}{30}}{1 - \frac{15}{30}} = \frac{\frac{1}{3}}{\frac{1}{2}} = \frac{2}{3}.$$

- (d)  $E(X) = \frac{0+30}{2} = 15$  minutes.
- 2. (a) We find  $P(S_1) = \frac{1}{6}$ ,  $P(S_2) = \frac{5}{6} \cdot \frac{1}{6}$ ,  $P(S_3) = \left(\frac{5}{6}\right)^2 \frac{1}{6}$ , ...,  $P(S_k) = \left(\frac{5}{6}\right)^{k-1} \frac{1}{6}$ .
  - (b) Note that the events  $S_1, \ldots, S_k$  are mutually disjoint, therefore we get

$$\begin{aligned} \text{P(a 6 will shop up eventually)} &= \lim_{k \to \infty} \text{P}(S_1 \cup \dots \cup S_k) = \lim_{k \to \infty} \sum_{j=1}^k \text{P}(S_j) \\ &= \frac{1}{6} \sum_{j=1}^{\infty} \left(\frac{5}{6}\right)^{j-1} = \frac{1}{6} \sum_{m=0}^{\infty} \left(\frac{5}{6}\right)^m \stackrel{\text{Hint}}{=} \frac{1}{6} \frac{1}{1 - \frac{5}{6}} = 1 \; \checkmark \end{aligned}$$

3. (a) We have

$$\int_{0}^{1} \int_{0}^{1} f(x,y) dx dy = \alpha \int_{0}^{1} \int_{0}^{1} (1 + (2x - 1)(2y - 1)) dx dy = \alpha \left( 1 + (x^{2} - x) \Big|_{0}^{1} (y^{2} - y) \Big|_{0}^{1} \right)$$
$$= \alpha \stackrel{!}{=} 1$$

- (b) Yes, since no parameters are involved!
- (c) For  $x,y \in (0,1)$  we have  $F(x,y) = \int_0^x \int_0^y (1+(2x-1)(2y-1)) dx dy = xy + xy(x-1)(y-1)$ . Therefore,

$$F(x,y) = xy(1 + (x-1)(y-1))\mathbb{I}_{(0,1)}(x)\mathbb{I}_{(0,1)}(y) + x\mathbb{I}_{(0,1)}(x)\mathbb{I}_{[1,\infty)}(y) + y\mathbb{I}_{[1,\infty)}(x)\mathbb{I}_{(0,1)}(y) + \mathbb{I}_{[1,\infty)}(x)\mathbb{I}_{[1,\infty)}(y)$$

(d) 
$$P(X \ge \frac{1}{2}, Y \le 1) = P(X \le 1, Y \le 1) - P(X < \frac{1}{2}, Y \le 1) = F(1, 1) - F(\frac{1}{2}, 1) = 1 - \frac{1}{2} = \frac{1}{2}$$
.

(e) In (c) We found  $F(x) = x\mathbb{I}_{(0,1)}(x)$  and  $F(y) = y\mathbb{I}_{(0,1)}(y)$  implying the marginal pdf's  $f(x) = \mathbb{I}_{(0,1)}(x)$  and  $f(y) = \mathbb{I}_{(0,1)}(y)$ . Since  $f(x,y) \neq f(x) \cdot f(y)$  theorem 2.9 indicates that X and Y are not stochastically independent.

(f)

$$\begin{split} \mathbf{P}(X>Y) &= \int_{x=y}^{1} \int_{y=0}^{1} f(x,y) dx dy = \int_{y=0}^{1} \int_{x=y}^{1} (1 + (2x-1)(2y-1)) dx dy \\ &= \int_{y=0}^{1} [(1-y) - (2y-1)(y^2-y))] dy = \int_{0}^{1} (1-y) - (2y^3 - 3y^2 + y) dy \\ &= \int_{0}^{1} 1 - 2y^3 + 3y^2 - 2y \ dy = \left[ y - \frac{y^4}{2} + y^3 - y^2 \right]_{0}^{1} = \frac{1}{2}. \end{split}$$

4. (a) Using the Lindeberg-Levy-CLT (thm. 5.13):

$$\sqrt{n} \frac{\overline{X}_n - \mu}{\sigma} \stackrel{d}{\to} \mathcal{N}(0,1) \text{ with } \mu = \frac{\alpha}{\alpha + \beta} \text{ and } \sigma^2 = \frac{\alpha\beta}{(\alpha + \beta)^2(\alpha + \beta + 1)}$$

yields

$$\overline{X}_n \stackrel{a}{\sim} \mathcal{N}\left(\frac{\alpha}{\alpha+\beta}, \frac{\alpha\beta}{n(\alpha+\beta)^2(\alpha+\beta+1)}\right)$$

(b) Use the Delta-Method (thm. 5.18) by defining  $Z_n = g(\overline{X}_n) = \exp(-\overline{X}_n^2)$  with  $G = g'(\overline{X}_n)\big|_{\overline{X}_n = \mu} = -2\overline{X}_n \exp(-\overline{X}_n^2)\big|_{\overline{X}_n = \mu} = -2\frac{\alpha}{\alpha + \beta} \exp\left(-\frac{\alpha^2}{(\alpha + \beta)^2}\right)$ . Thus

$$Z_n \stackrel{a}{\sim} \mathcal{N}\left(\exp\left(-\frac{\alpha^2}{(\alpha+\beta)^2}\right), \frac{4\alpha^3\beta}{n(\alpha+\beta)^4(\alpha+\beta+1)}\exp\left(-\frac{2\alpha^2}{(\alpha+\beta)^2}\right)\right)$$

(c) Applying Corollary 5.1 we have

$$E(Z_n) \stackrel{n \to \infty}{\to} \exp\left(-\frac{\alpha^2}{(\alpha + \beta)^2}\right)$$
 and  $Var(Z_n) \stackrel{n \to \infty}{\to} 0$ , which implies

 $Z_n \xrightarrow{m} \exp\left(-\frac{\alpha^2}{(\alpha+\beta)^2}\right)$  which i turn implies by thm. 5.9 that  $Z_n \xrightarrow{p} \exp\left(-\frac{\alpha^2}{(\alpha+\beta)^2}\right) \stackrel{\alpha=\beta}{=} \exp\left(-\frac{1}{4}\right)$ .

- (d) Check the requirements of the change of variables technique (thm. 2.12) and apply said theorem for y = g(x) = 1 x:
  - $q'(x) = -1 \neq 0 \forall x \checkmark$
  - $g^{-1}(y) = 1 y$  exists  $\forall y \in (0,1) \checkmark$ , therefore

$$h(y) = f(g^{-1}(y)) \left| \frac{dg^{-1}(y)}{dy} \right| = \frac{1}{B(\alpha, \beta)} (1 - y)^{\alpha - 1} y^{\beta - 1} \mathbb{I}_{(0, 1)}(y),$$

which is the distribution of a  $Beta(\beta,\alpha)$  because  $B(\alpha,\beta)=B(\beta,\alpha)$ .

(e)

$$M_Z(t) = \mathbf{E}(e^{tZ}) = \mathbf{E}(e^{t(X_1 + X_2 + X_3)}) = \mathbf{E}(e^{tX_1}e^{tX_2}e^{tX_3}) \stackrel{\text{ind.}}{=} \mathbf{E}(e^{tX_1})\mathbf{E}(e^{tX_2})\mathbf{E}(e^{tX_3})$$
$$= M_{X_1}(t)M_{X_2}(t)M_{X_3}(t) = \left(\sum_{r=0}^{\infty} \frac{B(r + \alpha, \beta)}{B(\alpha, \beta)} \frac{t^r}{r!}\right)^3$$