# MH2500 AY19/20

## Solution 1.

(a) 
$$\frac{\binom{3}{3}}{\binom{7}{3}} = \frac{1}{35}.$$

(b) 
$$\begin{split} &P(\text{remaining all blue}|\text{drawn all red}) \\ &= P(i = 3|\text{drawn all red}) \\ &= \frac{P(\text{drawn all red}|i = 3)P(i = 3)}{P(\text{drawn all red}|i = 3)P(i = 3) + P(\text{drawn all red}|i = 2)P(i = 2) + P(\text{drawn all red}|i = 1)P(i = 1)} \\ &= \frac{1/35}{\binom{3}{1}/\binom{7}{1} + \binom{3}{2}/\binom{7}{2} + \binom{3}{3}/\binom{7}{3}} = \frac{1}{21}. \end{split}$$

### Solution 2.

- (a) X is uniformly distributed in (-1,1).  $f_X(x) = \frac{1}{2}$ , if  $x \in (-1,1)$ .
- (b) For a given X = x,

$$|x\cos\theta - y\sin\theta| < 1 \implies \frac{x\cos\theta - 1}{\sin\theta} < y < \frac{x\cos\theta + 1}{\sin\theta}.$$

Since X, Y are uniformly distributed in  $D_{\theta}$ , therefore

$$Y|X = x \sim \text{Unif}\left(\frac{x\cos\theta - 1}{\sin\theta}, \frac{x\cos\theta + 1}{\sin\theta}\right)$$

The conditional density function is

$$f_{Y|X}(y|x) = \frac{1}{\frac{x\cos\theta + 1}{\sin\theta} - \frac{x\cos\theta - 1}{\sin\theta}} = \frac{\sin\theta}{2}, \quad \text{for } y \in \left(\frac{x\cos\theta - 1}{\sin\theta}, \frac{x\cos\theta + 1}{\sin\theta}\right)$$

(c) 
$$\mathbb{E}(Y|X) = \int_{\frac{x \cos \theta + 1}{\sin \theta}}^{\frac{x \cos \theta + 1}{\sin \theta}} y \cdot \frac{\sin \theta}{2} dy = \frac{x}{\tan \theta}.$$

(d) 
$$\mathbb{E}(Y) = \mathbb{E}_X \mathbb{E}(Y|X) = \mathbb{E}\left(\frac{X}{\tan \theta}\right) = \frac{1}{\tan \theta} \mathbb{E}(X) = 0.$$

(e) 
$$\mathbb{E}(Y^2) = \mathbb{E}_X \mathbb{E}_Y(Y^2|X) = \int_{-1}^1 f_X(x) \int_{\frac{x \cos \theta - 1}{\sin \theta}}^{\frac{x \cos \theta + 1}{\sin \theta}} y^2 \frac{\sin \theta}{2} dy dx = \frac{\sin \theta}{4} \int_{-1}^1 \int_{\frac{x \cos \theta - 1}{\sin \theta}}^{\frac{x \cos \theta + 1}{\sin \theta}} y^2 dy dx = \frac{1 + \cos^2 \theta}{3 \sin^2 \theta}.$$

Therefore,  $Var(Y) = \mathbb{E}(Y^2) - \mathbb{E}(Y)^2 = \frac{1 + \cos^2 \theta}{3 \sin^2 \theta}$ .

(f) Since  $\mathbb{E}(Y) = 0$ 

$$Cov(X,Y) = \mathbb{E}(XY) = \mathbb{E}_X \mathbb{E}(XY|X) = \mathbb{E}_X (X\mathbb{E}(Y|X)) = \mathbb{E}_X \left( X \cdot \frac{X}{\tan \theta} \right) = \frac{1}{3 \tan \theta} > 0.$$

X, Y are positively correlated.

### Solution 3.

(a) 
$$P(X_2 = k) = \binom{n}{k} \left(\frac{1}{6}\right)^k \left(\frac{5}{6}\right)^{n-k}.$$

(b) 
$$P(X_2 = k, X_4 = \ell) = \frac{n!}{k!\ell!(n-k-\ell)!} \left(\frac{1}{6}\right)^k \left(\frac{1}{6}\right)^\ell \left(\frac{4}{6}\right)^{n-k-\ell}.$$

(c) 
$$P(X_2 = k | X_4 = \ell) = \frac{(n-\ell)!}{k!(n-k-\ell)!} \left(\frac{1}{5}\right)^k \left(\frac{4}{5}\right)^{n-k-\ell}.$$

(d) 
$$P(X_2 + X_4 = k) = \binom{n}{k} \left(\frac{2}{6}\right)^k \left(\frac{4}{6}\right)^{n-k}, \qquad k = 0, 1, \dots, n.$$

(e) Let  $W_i, Y_i$  be the indicator variables that the *i*-th roll results in number 2 and 4 respectively. Therefore,

$$X_2 = \sum_{i=1}^{n} W_i, \qquad X_4 = \sum_{i=1}^{n} Y_i$$

$$Cov(X_2, X_4) = Cov\left(\sum_{i=1}^n W_i, \sum_{j=1}^n Y_j\right)$$

$$= \sum_{i=1}^n \sum_{j=1}^n Cov(W_i, Y_j)$$

$$= \sum_{i=1}^n Cov(W_i, Y_i) + \sum_{i \neq j} Cov(W_i, Y_j)$$

For  $i \neq j$ ,  $W_i$  and  $Y_j$  are independent, hence  $Cov(W_i, Y_j) = 0$ .

$$Cov(W_i, Y_i) = \mathbb{E}(W_i Y_i) - \mathbb{E}(W_i) \mathbb{E}(Y_i)$$

$$= P(i\text{-th throw} = 2 \cap i\text{-th throw} = 4) - P(i\text{-th throw} = 2)P(i\text{-th throw} = 4) = -\frac{1}{36}$$

Therefore,  $Cov(X_2, X_4) = -\frac{n}{36}$ .

#### Solution 4.

(a) 
$$\mathbb{E}\left(\sum_{i=1}^{2m}b_{m,i}Y_{i}\right) = \mathbb{E}\left(\sum_{i=1}^{m}b_{m,i}Y_{i}\right) + \mathbb{E}\left(\sum_{i=m+1}^{2m}b_{m,i}Y_{i}\right) = \sum_{i=1}^{m}\mathbb{E}\left(Y_{i}\right) - \sum_{i=m+1}^{2m}\mathbb{E}\left(Y_{i}\right) = 0.$$

$$Var\left(\sum_{i=1}^{2m}b_{m,i}Y_{i}\right) = \sum_{i=1}^{m}b_{m,i}^{2}Var\left(Y_{i}\right) + \sum_{i=m+1}^{2m}b_{m,i}^{2}Var\left(Y_{i}\right) = \sum_{i=1}^{2m}Var\left(Y_{i}\right) = 2m \cdot \frac{1}{4} = \frac{m}{2}.$$

By Chebyshev's inequality,

$$P\left(\left|\sum_{i=1}^{2m} b_{m,i} Y_i\right| > c_1 \sqrt{m}\right) \le \frac{m/2}{c_1^2 m} = \frac{1}{2c_1^2} = \frac{1}{10}.$$

We can set  $c_1 \ge \sqrt{5}$ .

(b) Observe that

$$\sum_{i=1}^{2m} b_{m,i} Y_i = \sum_{i=1}^{m} Y_i - \sum_{i=m+1}^{2m} Y_i = \sum_{i=1}^{m} (Y_i - Y_{m+i})$$

Let  $W_i = Y_i - Y_{m+i}$ .  $\mathbb{E}(W_i) = 0$  and  $Var(W_i) = Var(Y_i) + Var(Y_{m+i}) = \frac{1}{2}$ . By central limit theorem,  $\frac{1}{m} \sum_{i=1}^{m} W_i$  converges in distribution to  $N(0, \frac{1}{2m})$ . Therefore,

$$\lim_{m \to \infty} P\left(\left|\sum_{i=1}^{2m} b_{m,i} Y_i\right| > c_2 \sqrt{m}\right) = \lim_{m \to \infty} P\left(\left|\frac{1}{m} \sum_{i=1}^m W_i\right| > \frac{c_2}{\sqrt{m}}\right) = P\left(|Z| > \frac{\frac{c_2}{\sqrt{m}}}{\sqrt{\frac{1}{2m}}}\right) = P(|Z| > \sqrt{2}c_2),$$

where Z is the standard normal distribution. We now find  $c_2$ , such that  $P(|Z| > \sqrt{2}c_2) > 0.9$ . Let  $\Phi(\cdot)$  be the CDF of Z, then

$$P(|Z| > \sqrt{2}c_2) = 2 - 2\Phi(\sqrt{2}c_2) > 0.9 \implies \text{choose } 0 < c_2 < \frac{1}{\sqrt{2}}\Phi^{-1}(0.55).$$