

**EECS 126 — MIDTERM #2 Solutions**

**1a. (i)** Yes,

$$\begin{aligned} E(X + Y) &= \iint (x + y)f_{XY}(x, y) dx dy \\ &= \iint xf_{XY}(x, y)(dx)dy + \iint yf_{XY}(x, y) dx dy \\ &= E(X) + E(Y) \end{aligned}$$

**(ii)** No, true if  $X, Y$  are independent (and more generally, uncorrelated).

**(iii)** No, true if  $Y = c$ , a constant.

**(iv)** No,  $E(XY|Y=3) = 3E(X|Y=3)$ . This equals  $3E(X)$  if, for example,  $X$  and  $Y$  are independent.

**1b. (i)** Yes,

$$\begin{aligned} E(Y|X = x) &= \int yf_{Y|X}(y|x) dy \\ &= \int yf_Y(y) dy = E(Y) \end{aligned}$$

for any  $x$ .

$$\begin{aligned} \text{(ii)} \quad E(Y|X) &= \frac{1}{2}E(X|X) + \frac{1}{2}E(-X|X) \\ &= 0 \\ &= E(Y) \end{aligned}$$

$X$  and  $Y$  are clearly not independent.

**2.** Let  $X_i, Y_i$  be the horizontal and vertical displacement travelled at step  $i$ . Let  $D$  be the squared distance after  $n$  steps.

$$D = \left( \sum_{i=1}^n X_i \right)^2 + \left( \sum_{i=1}^n Y_i \right)^2$$

$X_i, X_j$  are independent,  $E(X_i) = 0$ , and similarly for  $Y_i$ 's.

$$\begin{aligned}
\text{So, } E(D) &= \sum_{i=1}^n E(X_i^2) + \sum_{i \neq j} E(X_i X_j) + \sum_{i=1}^n E(Y_i^2) \\
&\quad + \sum_{i \neq j} E(Y_i Y_j) \\
&= \sum_{i=1}^n E(X_i^2) + \sum_{i=1}^n E(Y_i^2) \\
&= \sum_{i=1}^n E(X_i^2 + Y_i^2) \\
&= n
\end{aligned}$$

**3a.**  $E(N) = E[E(N|M)]$

$$\begin{aligned}
&= \sum_{m=0}^{\infty} E(N|M=m)P(M=m) \\
&= \sum_{m=0}^{\infty} m(1-\varepsilon)P(M=m) \\
&= (1-\varepsilon)E(M) = \frac{(1-\varepsilon)(1-p)}{p}
\end{aligned}$$

**3b.**  $E(N^2) = E[E(N^2|M)]$

$$\begin{aligned}
E(N^2|M) &= \text{Var}(N^2|M) + [E(N|M)]^2 \\
&= M\varepsilon(1-\varepsilon) + [M(1-\varepsilon)]^2 \\
E(N^2) &= E[M\varepsilon(1-\varepsilon) + M^2(1-\varepsilon)^2] \\
&= \frac{(1-p)\varepsilon(1-\varepsilon)}{p} + \left[ \frac{1-p}{p^2} + \frac{(1-p)^2}{p^2} \right] (1-\varepsilon)^2 \\
&= \frac{\varepsilon(1-\varepsilon)(1-p)}{p} + \frac{(1-p)(2-p)}{p^2} (1-\varepsilon)^2
\end{aligned}$$

**3c.**  $P(N=n) = \sum_{m=0}^{\infty} P(M=m)P(N=n|M=m)$

$$P(N=n|M=m) = \begin{cases} \binom{m}{n} (1-\varepsilon)^n \varepsilon^{m-n} & \text{if } n \leq m \\ 0 & \text{otherwise} \end{cases}$$

So,  $P(N=n) = \sum_{m=n}^{\infty} \binom{m}{n} (1-\varepsilon)^n \varepsilon^{m-n} \cdot (1-p)^m p$