

ME424 HW5

Zhuang Yulun 11811126

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1 Probabilistic modeling of uncertainties

Define event $A = \{S_a \text{ is accepted}\}$ and event $B = \{S_b \text{ is accepted}\}$
 $\Omega = \{(A, B), (A, \bar{B}), (\bar{A}, B), (\bar{A}, \bar{B})\}$

1.1

Given

$$P(A) = 0.8, P(B) = 0.6, P(AB) = P(A)P(B)$$

We have

$$P(AB) = P(A)P(B) = 0.48$$

$$P(A\bar{B}) = P(A)P(\bar{B}) = 0.32$$

$$P(\bar{A}B) = P(\bar{A})P(B) = 0.12$$

$$P(\bar{A}\bar{B}) = P(\bar{A})P(\bar{B}) = 0.16$$

1.2

Given

$$P(A|B) = 1, P(A|\bar{B}) = 0.3, P(\bar{A}\bar{B}) = 0.35$$

We have $P(B) = \frac{P(\bar{A}\bar{B})}{P(A|\bar{B})} = 0.5$

Thus,

$$P(AB) = P(A|B)P(B) = 0.5$$

$$P(A\bar{B}) = P(A|\bar{B})P(\bar{B}) = 0.15$$

$$P(\bar{A}B) = P(\bar{A}|B)P(B) = 0$$

$$P(\bar{A}\bar{B}) = 0.35$$

2 Conditional Probability and Expectation

2.1

Given

$$p_X(x) = \frac{1}{n+1} \text{ for } x = 0, 1, \dots, n$$

$$p_{Y|X=x}(y|x) = \frac{1}{x+1} \text{ for } y = 0, 1, \dots, x$$

(a) $E(Y|X = x) = x/2$

(b) $E(Y) = \sum_{i=0}^n E(Y|X = x)p_X(x) = n/4$

(c) $p_{XY}(x, y) = p_X(x)p_{Y|X=x}(y|x) = \frac{1}{(n+1)(x+1)} \text{ for } x \geq y, \text{ otherwise } 0$

(d) $p_Y(y) = \sum_{x=y}^n p_{XY}(x, y) = \sum_{x=y}^n \frac{1}{(n+1)(x+1)}$ for $x \geq y$, otherwise 0

(e)

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n = 100;
E = 0;
for y = 0:n
    p = 0;
    for x = y:n
        p = p+1/(n+1)/(x+1);
    end
    E = E + y*p;
end

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The result $E(Y) = 25$ for $n = 100$, which is the same to (b)

(f) $E(g(X)Y) = 2E(Y|X = 1)p_X(1) + 2E(Y|X = n)p_X(n) = 1$

3 Conditional Density and Expectation

(a) $E(Y|X = x) = \frac{x-6}{2}$

(b) $f_X(x) = \int_0^2 f_{XY}(x, y)dy = x + \frac{1}{2}, x \in [0, 1]$
 $f_Y(y) = \int_0^1 f_{XY}(x, y)dx = \frac{1}{4}(y+1), y \in [0, 2]$
 $f_{X|Y}(x|y) = \frac{f_{XY}(x, y)}{f_Y(y)} = \frac{2x+y}{y+1}, x \in [0, 1]$
 \Rightarrow
 $E(X) = \int_0^1 x(f_X(x))dx = \frac{7}{12}$
 $E(X|Y = \frac{1}{2}) = \int_0^1 xf_{X|Y}(x|\frac{1}{2})dx = \frac{11}{18}$

(c) Given

$$f(x_1|x_2) = f(x_1)$$

$$f(x_1|x_3) = f(x_1)$$

$$f(x_2, x_3) = f(x_2, x_3|x_1) = \frac{f(x_1, x_2, x_3)}{f(x_1)}$$

We have

$$f(x_1, x_2|x_3) = \frac{f(x_1, x_2, x_3)}{f(x_3)}$$

$$= \frac{f(x_1)f(x_2, x_3)}{f(x_3)}$$

$$= f(x_1|x_3)f(x_2|x_3)$$

4 Random Vectors

(a)

$$E(W) = \begin{bmatrix} E(X_1) \\ E(X_3) \end{bmatrix} = \begin{bmatrix} 1 \\ -2 \end{bmatrix}$$

$$Cov(W) = \begin{bmatrix} Cov(X_1, X_1) & Cov(X_1, X_3) \\ Cov(X_3, X_1) & Cov(X_3, X_3) \end{bmatrix} = \begin{bmatrix} 6 & 5 \\ 5 & 6 \end{bmatrix}$$

(b)

$$Cov(W, X_2) = \begin{bmatrix} Cov(X_1, X_2) \\ Cov(X_3, X_2) \end{bmatrix} = \begin{bmatrix} 2 \\ 3 \end{bmatrix}$$

(c)

$$\begin{aligned} Cov(V, V) &= \begin{bmatrix} Cov(X_2 - 1, X_2 - 1) & Cov(X_2 - 1, X_1 + X_3) \\ Cov(X_2 - 1, X_1 + X_3) & Cov(X_1 + X_3, X_1 + X_3) \end{bmatrix} \\ &= \begin{bmatrix} Cov(X_2, X_2) & Cov(X_2, X_1) + Cov(X_2, X_3) \\ Cov(X_2, X_1) + Cov(X_2, X_3) & Cov(X_1) + Cov(X_3) + 2Cov(X_1, X_3) \end{bmatrix} \\ &= \begin{bmatrix} 9 & 5 \\ 5 & 22 \end{bmatrix} \end{aligned}$$