HOMEWORK 1

Nguyễn Thị Minh Ngọc Full name:

Student ID: 11219280 Class: **DSEB 63**

Ex1:

a. The marginal distribution:

•
$$p(X = x_1) = 0.01 + 0.05 + 0.1 = 0.16$$

•
$$p(X = x_2) = 0.02 + 0.1 + 0.05 = 0.17$$

•
$$p(X = x_3) = 0.03 + 0.05 + 0.03 = 0.11$$

•
$$p(X = x_4) = 0.1 + 0.07 + 0.05 = 0.22$$

•
$$p(X = x_5) = 0.1 + 0.2 + 0.04 = 0.34$$

•
$$p(Y = y_1) = 0.01 + 0.02 + 0.03 + 0.1 + 0.1 = 0.26$$

•
$$p(Y = y_2) = 0.05 + 0.1 + 0.05 + 0.07 + 0.2 = 0.47$$

•
$$p(Y = y_3) = 0.1 + 0.05 + 0.03 + 0.05 + 0.04 = 0.27$$

b. The conditional distribution

•
$$p(X = x_1 | Y = y_1) = \frac{0.01}{0.01 + 0.02 + 0.03 + 0.1 + 0.1} = \frac{1}{26} \approx 0.038$$

• $p(X = x_2 | Y = y_1) = \frac{0.02}{0.01 + 0.02 + 0.03 + 0.1 + 0.1} = \frac{1}{13} \approx 0.077$

•
$$p(X = x_2 | Y = y_1) = \frac{0.02}{0.01 + 0.02 + 0.03 + 0.1 + 0.1} = \frac{1}{13} \approx 0.077$$

•
$$p(X = x_3 | Y = y_1) = \frac{0.01 + 0.02 + 0.03 + 0.1 + 0.1}{0.01 + 0.02 + 0.03 + 0.1 + 0.1} = \frac{3}{26} \approx 0.115$$

•
$$p(X = x_4 | Y = y_1) = \frac{0.1}{0.01 + 0.02 + 0.03 + 0.1 + 0.1} = \frac{5}{13} \approx 0.385$$

•
$$p(X = x_4 | Y = y_1) = \frac{0.01 + 0.02 + 0.03 + 0.1 + 0.1}{0.01 + 0.02 + 0.03 + 0.1 + 0.1} = \frac{5}{13} \approx 0.385$$

• $p(X = x_5 | Y = y_1) = \frac{0.1}{0.01 + 0.02 + 0.03 + 0.1 + 0.1} = \frac{5}{13} \approx 0.385$

•
$$p(Y = y_1 | X = x_3) = \frac{0.03}{0.03 + 0.05 + 0.03} = \frac{3}{11} \approx 0.273$$

•
$$p(Y = y_2 | X = x_3) = \frac{0.05}{0.03 + 0.05 + 0.03} = \frac{5}{11} \approx 0.455$$

•
$$p(Y = y_2 | X = x_3) = \frac{0.03 + 0.05 + 0.03}{0.03 + 0.05 + 0.03} = \frac{5}{11} \approx 0.455$$

• $p(Y = y_3 | X = x_3) = \frac{0.03}{0.03 + 0.05 + 0.03} = \frac{3}{11} \approx 0.273$

Ex2:

* Case 1: X and Y are discrete random variables

$$E_x[X|Y] = \sum_{i=1}^n x_i P(X = x_i|Y)$$

$$E_{y}[E_{x}[X|Y]] = \sum_{j=1}^{m} E_{x}[X|Y = y_{j}]P(X|Y = y_{j})$$

$$= \sum_{j=1}^{m} (\sum_{i=1}^{n} x_{i}P(X = x_{i}|Y = y_{j})P(X|Y = y_{j}))$$

$$= \sum_{j=1}^{m} \sum_{i=1}^{n} x_{i}P(X = x_{i}|Y = y_{j})P(X|Y = y_{j})$$

$$= \sum_{i=1}^{n} \sum_{j=1}^{m} x_{i}P(X = x_{i}|Y = y_{j})P(X|Y = y_{j})$$

$$= \sum_{i=1}^{n} \sum_{j=1}^{m} x_{i}P(X = x_{i}, Y = y_{j})$$

$$= \sum_{i=1}^{n} x_{i}P(X = x_{i})$$

$$= E(X)$$

Hence, $E(X) = E_y[E_x[X|Y]]$

* Case 2: X and Y are continuous random variables

$$E_{x}[X|Y] = \int_{-\infty}^{\infty} x f_{x|y}(x,y) dx$$

$$E_{y}[E_{x}[X|Y]] = \int_{-\infty}^{+\infty} (E_{x}[X|Y]) f_{y}(y) dy$$

$$= \int_{-\infty}^{+\infty} \int_{-\infty}^{+\infty} (x f_{x|y}(x,y) dx) f_{y}(y) dy$$

$$= \int_{-\infty}^{+\infty} \int_{-\infty}^{+\infty} x (f_{x|y}(x,y) f_{y}(y)) dx dy$$

$$= \int_{-\infty}^{+\infty} \int_{-\infty}^{+\infty} x f_{x,y}(x,y) dx dy$$

$$= \int_{-\infty}^{+\infty} x \int_{-\infty}^{+\infty} f_{x,y}(x,y) dy dx$$

$$= \int_{-\infty}^{+\infty} x f_{x}(x) dx$$

$$= E(X)$$

Hence, $E(X) = E_{\nu}[E_{x}[X|Y]]$

Ex3:

* Case 1: X and Y are discrete random variables

$$V_{X} = \sum_{i=1}^{n} (x_{i} - \mu)^{2} P(X = x_{i})$$

$$= \sum_{i=1}^{n} (x_{i} - E_{X}(X))^{2} P(X = x_{i})$$

$$= \sum_{i=1}^{n} (x_{i}^{2} - 2x_{i}E_{X}(X) + E_{X}(X)^{2}) P(X = x_{i})$$

$$= \sum_{i=1}^{n} x_{i}^{2} P(X = x_{i}) - \sum_{i=1}^{n} 2x_{i}E_{X}(X) P(X = x_{i}) + \sum_{i=1}^{n} E_{X}(X)^{2} P(X = x_{i})$$

$$= E_{X}(X^{2}) - 2E_{X}(X) \sum_{i=1}^{n} x_{i} P(X = x_{i}) + E_{X}(X)^{2} \sum_{i=1}^{n} P(X = x_{i})$$

$$= E_{X}(X^{2}) - 2E_{X}(X)E_{X}(X) + E_{X}(X)^{2}$$

$$= E_{X}(X^{2}) - E_{X}(X)^{2}$$

Hence, $V_r(X) = E_r(X^2) - E_r(X)^2$

* Case 2: X and Y are continuous random variables

$$V_{x} = \int_{-\infty}^{+\infty} (x - \mu)^{2} f(x) dx$$

$$= \int_{-\infty}^{+\infty} (x^{2} - 2x\mu + \mu^{2}) f(x) dx$$

$$= \int_{-\infty}^{+\infty} x^{2} f(x) dx - \int_{-\infty}^{+\infty} 2x\mu f(x) dx + \int_{-\infty}^{+\infty} \mu^{2} f(x) dx$$

$$= E_{x}(X^{2}) - (2\mu \int_{-\infty}^{+\infty} xf(x) dx) + (\mu^{2} \int_{-\infty}^{+\infty} f(x) dx)$$

$$= E_{x}(X^{2}) - 2\mu^{2} + \mu^{2}$$

$$= E_{x}(X^{2}) - E_{x}(X)^{2}$$

Hence, $V_x(X) = E_x(X^2) - E_x(X)^2$

Ex4:

X: Cancerous Tumors $\rightarrow X^c$: Benign Tumors

Y: Positive Mammogram Result \rightarrow Y^c: Negative Mammogram Result

$$P(x) = 0.01 \rightarrow P(X^c) = 0.99$$

 $P(Y|X) = 0.8 \rightarrow P(Y^c|X) = 0.2$
 $P(Y^c|X^c) = 0.9 \rightarrow P(Y|X^c) = 0.1$

$$P(X|Y) = ?$$

According to Baye's Rule:

$$P(X|Y) = \frac{P(Y|X)P(X)}{P(Y)}$$

$$= \frac{P(Y|X)P(X)}{P(Y,X) + P(Y,X^c)}$$

$$= \frac{P(Y|X)P(X)}{P(Y|X)P(X) + P(Y|X^c)P(X^c)}$$

$$= \frac{0.8 * 0.01}{0.8 * 0.01 + 0.1 * 0.99} = \frac{8}{107} \approx 0.075 \approx 7.5\%$$

The probability of cancer of this women after receiving positive mammogram result is about 7.5%. Hence, I do not agree with the estimated result of those physicians.

Ex5:

X: Wrong first pick $\rightarrow X^c$: Right first pick

Y: Win when switching $\rightarrow Y^c$: Lose when switching

We consider the probability of win if I switch under two cases.

- Case 1: My initial pick was wrong
- The probability that my first pick was wrong is $P(X) = \frac{3}{4}$
- So, in case that my first pick was wrong, the right door will be one of two unopened door, then the probability that I will win when switching is $P(Y|X) = \frac{1}{2}$
- → The probability of win if my first pick was wrong and I choose to switch when being asked is:

$$P(X,Y) = P(Y|X) * P(X) = \frac{1}{2} * \frac{3}{4} = \frac{3}{8}$$

- Case 2: My initial pick was right
- The probability that my first pick was right is $P(X) = \frac{1}{4}$
- So, in case that my first pick was right, I have no chance to win when switching, then the probability that I will win when switching is $P(Y|X^c) = 0$
- → In case of right first pick, the probability of win if I choose to switch is:

$$P(X^c, Y) = P(Y|X^c) * P(X^c) = 0 * \frac{1}{4} = 0$$

Hence, the probability of win when switching is:

$$P(Y) = P(Y,X) + P(Y,X^c) = \frac{3}{8} + 0 = \frac{3}{8}$$