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# AI1103-Assignment 6

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#### Download all latex-tikz codes from

https://github.com/vaishnavi-w/AI1103/blob/main/ Assignment6/latex6.tex

#### QUESTION

Which of the following conditions imply independence of random variables *X* and *Y*?

- 1) Pr(X > a|Y > a) = Pr(X > a) for all  $a \in R$
- 2) Pr(X > a|Y < b) = Pr(X > a) for all  $a, b \in R$
- 3) X and Y are uncorrelated
- 4) E[(X a)(Y b)] = E[X a]E[Y b] for all  $a, b \in R$

#### Solution

1) Two random variables *X* and *Y* are independent when the joint probability distribution of random variables is product of their individual probability distributions i.e for all sets A,B

$$Pr(X \in A, Y \in B) = Pr(X \in A) Pr(Y \in B)$$
(0.0.1)

Consider the CDFs,

$$F_X(a) = \Pr(X < a)$$
 (0.0.2)

$$F_Y(b) = \Pr(Y < b)$$
 (0.0.3)

$$F_{XY}(a, b) = \Pr(X < a, Y < b)$$
 (0.0.4)

Let  $F_{X,Y}(a,b) = F_X(a) F_Y(b)$  be true. Partial derivative w.r.t a.

$$\frac{\partial F_{X,Y}(a,b)}{\partial a} = \frac{\partial F_X(a)}{\partial a} F_Y(b) \qquad (0.0.5)$$

Partial derivative w.r.t b,

$$\frac{\partial^2 F_{X,Y}(a,b)}{\partial b \partial a} = \frac{\partial F_X(a)}{\partial a} \frac{\partial F_Y(b)}{\partial b} \qquad (0.0.6)$$

$$\implies p_{X,Y}(a,b) = p_X(a) p_Y(b)$$
 (0.0.7)

when X,Y are discrete. And,

$$\implies f_{XY}(a,b) = f_X(a) f_Y(b)$$
 (0.0.8)

when X,Y are continuous, for all  $a, b \in R$ .  $\therefore$  Two random variables are independent if the joint CDF can be expressed as the product of individual CDFs i.e for all  $a, b \in R$ 

$$F_{X,Y}(a,b) = F_X(a) F_Y(b)$$
 (0.0.9)

Consider,

$$\Pr(X > a | Y > a) = \frac{\Pr(X > a, Y > a)}{\Pr(Y > a)}$$
(0.0.10)

Given that Pr(X > a|Y > a) = Pr(X > a),

$$\implies \Pr(X > a, Y > a) = \Pr(X > a) \Pr(Y > a)$$
(0.0.11)

for all  $a \in R$ .

$$1 - F_X(a) - F_Y(a) = \Pr(X > a) - \Pr(Y < a)$$

$$= \Pr(X > a, Y > a) + \Pr(X > a, Y < a)$$

$$- \Pr(X > a, Y < a) - \Pr(X < a, Y < a)$$
(0.0.12)

$$1 - F_X(a) - F_Y(a) = \Pr(X > a, Y > a) - F_{X,Y}(a, a) \quad (0.0.13)$$

Substituting in (0.0.11),

$$1 - F_X(a) - F_Y(a) + F_{X,Y}(a, a) =$$

$$(1 - F_X(a))(1 - F_Y(a)) \quad (0.0.14)$$

$$\implies F_{X,Y}(a,a) = F_X(a) F_Y(a)$$
 (0.0.15)

On comparing with (0.0.9) in this case, it is true only under the condition that b = a. It may not be true for other values of b. The spectrum of conditions for independence is underrepresented. Hence, the condition does not imply independence of X and Y.

**Counterexample:** Consider two random variables  $X,Y \in \{0,1,2\}$  with the probabilities of the ordered pairs (X,Y) given in the Table1

X Y	0	1	2
0	0.2	0.1	0.1
1	0.2	0.1	0.05
2	0.1	0.1	0.05

TABLE 1: Pr(X, Y)

Case 1: a < 0

$$Pr(X > a|Y > a) = 1 = Pr(X > a)$$
 (0.0.16)

Case 2:  $0 \le a < 1$ 

$$\Pr(X > a | Y > a) = \frac{\Pr(X, Y > a)}{\Pr(Y > a)} = \frac{0.3}{0.5} = 0.6$$
(0.0.17)

$$Pr(X > a) = Pr(X = 1) + Pr(X = 2) = 0.6$$
(0.0.18)

Case 3:  $1 \le a < 2$ 

$$\Pr(X > a | Y > a) = \frac{\Pr(X, Y > a)}{\Pr(Y > a)} = \frac{0.05}{0.2} = 0.25$$
(0.0.19)

$$Pr(X > a) = Pr(X = 2) = 0.25$$
(0.0.20)

Case 4:  $a \ge 2$ 

$$\Pr(X > a | Y > a) = \frac{\Pr(X, Y > a)}{\Pr(Y > a)} \quad (0.0.21)$$

is not defined as Pr(Y > a) = 0. In all the cases, Pr(X > a|Y > a) = Pr(X > a) is true. Consider,

$$Pr(X = 1, Y = 2) = 0.05$$
 (0.0.22)

$$Pr(X = 1) Pr(Y = 2) = 0.35 \times 0.2 = 0.7$$
  
 $\neq Pr(X = 1, Y = 2) \quad (0.0.23)$ 

Clearly, X and Y are not independent.

### Option 1 is incorrect

2) From Bayes theorem,

$$\Pr(X > a | Y < b) = \frac{\Pr(X > a, Y < b)}{\Pr(Y < b)}$$
(0.0.24)

Given that Pr(X > a | Y < b) = Pr(X > a),

$$\implies \Pr(X > a, Y < b) = \Pr(X > a) \Pr(Y < b)$$
(0.0.25)

for all  $a, b \in R$ . Consider

$$F_Y(b) = \Pr(X > a, Y < b) + \Pr(X < a, Y < b)$$
(0.0.26)

$$\implies F_Y(b) - F_{X,Y}(a,b) = \Pr(X > a, Y < b)$$
(0.0.27)

Substituting in (0.0.25),

$$F_{Y}(b) - F_{X,Y}(a,b) = (1 - F_{X}(a)) F_{Y}(b)$$

$$(0.0.28)$$

$$\implies F_{X,Y}(a,b) = F_{X}(a) F_{Y}(b)$$

$$(0.0.29)$$

for all  $a, b \in R$ . Thus, X and Y are independent. **Option 2 is correct** 

3) Two random variables *X* and *Y* are uncorrelated if their covariance is zero.

$$cov[X, Y] = E[XY] - E[X]E[Y] = 0$$
(0.0.30)

Uncorrelatedness does not imply independence.

**Counterexample:** Let  $X \sim U[-1,1]$  be a uniformly distributed random variable.

$$f_X(x) = \begin{cases} \frac{1}{2} & -1 \le x \le 1\\ 0 & otherwise \end{cases}$$
 (0.0.31)

$$E[X] = \int_{-1}^{1} x f(x) dx = 0 \qquad (0.0.32)$$

Let  $Y = X^2$  be another random variable. X and Y are dependent.

$$cov[X, Y] = E[XY] - E[X]E[Y] (0.0.33)$$

$$= E[X^3] - 0 \times E[Y] (0.0.34)$$

$$= \int_{-1}^{1} x^3 f(x) dx = 0 (0.0.35)$$

X and Y are uncorrelated but not independent. Option 3 is incorrect

4) Given that,

$$E[(X-a)(Y-b)] = E[X-a]E[Y-b]$$
(0.0.36)

$$cov[(X-a), (Y-b)] = E[(X-a)(Y-b)]$$
  
-  $E[X-a]E[Y-b]$  (0.0.37)

$$\implies cov[(X-a)(Y-b)] = 0 = cov[X, Y]$$
(0.0.38)

From option 3, it follows that X and Y are not necessarily independent.

Option 4 is incorrect.