

Independent Random Variables

Definition 1: Let X_1 and X_2 be the random variables having the joint pdf $f(x_1, x_2)$ (joint pmf $p(x_1, x_2)$). Then the random variables X_1 and X_2 are said to be independent if, and only if,

$$f(x_1, x_2) = f_{X_1}(x_1) f_{X_2}(x_2)$$

($p(x_1, x_2) = p_{X_1}(x_1) p_{X_2}(x_2)$), where $f_{X_1}(x_1)$ ($p_{X_1}(x_1)$) and $f_{X_2}(x_2)$ ($p_{X_2}(x_2)$) are the marginal pdfs (pmfs) of X_1 + X_2 respectively.

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Theorem 1 When the conditional pdf $f_{2/1}(x_2/x_1)$ does not depend upon x_1 , we have

$$f_2(x_2) = f_{2/1}(x_2/x_1), \text{ and}$$

$$f(x_1, x_2) = f_1(x_1) f_2(x_2).$$

Sketch: we know that the marginal pdf of X_2 is, for random variables of the continuous type,

$$f_2(x_2) = \int_{-\infty}^{\infty} f(x_1, x_2) dx_1$$

Since, $f_{2/1}(x_2/x_1) = \frac{f(x_1, x_2)}{f_1(x_1)}$, we have

$$f_2(x_2) = \int_{-\infty}^{\infty} f_{2/1}(x_2/x_1) f_1(x_1) dx_1$$

$$= f_{2/1}(x_2/x_1) \int_{-\infty}^{\infty} f_1(x_1) dx_1$$

$$\Rightarrow f_2(x_2) = f_{2/1}(x_2/x_1) \text{ as } \int_{-\infty}^{\infty} f_1(x_1) dx_1 = 1$$

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(2)

Also, $f(x_1, x_2) = f_{2|1}(x_2|x_1) f_1(x_1)$

$$= f_2(x_2) f_1(x_1)$$

Thus, if the conditional distribution of X_2 , given $X_1 = x_1$, is independent of any assumption about x_1 , then $f(x_1, x_2) = f(x_1) f(x_2)$, that is, the random variables X_1 and X_2 are independent. //

Example 8. If the random variables X_1 and X_2 have the joint pdf $f(x_1, x_2) = \begin{cases} 2e^{-x_2-x_1}, & 0 < x_1 < x_2, & 0 < x_2 < \infty \\ 0 & \text{elsewhere,} \end{cases}$ then X_1 and X_2 are dependent.

Solution: The marginal pdfs of X_1 and X_2 are

$$f_1(x_1) = \begin{cases} \int_{-\infty}^{\infty} f(x_1, x_2) dx_2 = 2 \int_{x_1}^{\infty} e^{-x_2-x_1} dx_2 \\ = 2e^{-x_1} e^{-x_1} = 2e^{-2x_1}, & 0 < x_1 < \infty \\ 0 & \text{elsewhere.} \end{cases}$$

$$\text{and } f_2(x_2) = \begin{cases} \int_{-\infty}^{\infty} f(x_1, x_2) dx_1 = 2e^{-x_2} \int_0^{x_2} e^{-x_1} dx_1 \\ = 2e^{-2x_2}, & 0 < x_2 < \infty \\ 0 & \text{elsewhere} \end{cases}$$

$$\text{we have, } f_1(x_1) f_2(x_2) = \begin{cases} 2e^{-2(x_1+x_2)} & 0 < x_1 < x_2, \\ & x_2 < \infty \\ 0 & \text{elsewhere} \end{cases}$$

$\neq f(x_1, x_2)$,
∴ the random variables X_1 and X_2 are dependent. //

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Theorem 2: Let the random variables X_1 and X_2 have supports S_1 and S_2 , respectively, and have the joint pdf $f(x_1, x_2)$. Then X_1 and X_2 are independent if, and only if

$$f(x_1, x_2) = g(x_1) h(x_2), \quad \text{--- (i)}$$

for some nonnegative function, $g(x_1)$, of x_1 , and a nonnegative function, $h(x_2)$, of x_2 , that is,

$g(x_1) > 0$ when $x_1 \in S_1$ and zero elsewhere, and $h(x_2) > 0$ when $x_2 \in S_2$ and zero elsewhere.

Proof. Let X_1 and X_2 are independent random variables. Then, we have

$$f(x_1, x_2) = f_1(x_1) f_2(x_2),$$

where $f_1(x_1)$ and $f_2(x_2)$ are the marginal pdfs of X_1 & X_2 , respectively.

\therefore By the def. of pdf,

$$f_1(x_1) > 0 \text{ when } x_1 \in S_1 \text{ \& zero elsewhere}$$

$$\text{and } f_2(x_2) > 0 \text{ when } x_2 \in S_2 \text{ \& zero elsewhere}$$

Thus, the condition (i) is fulfilled.

Conversely, let the condition (i) is fulfilled,

$$\text{that is, } f(x_1, x_2) = g(x_1) h(x_2)$$

$$\text{Now } f_1(x_1) = \int_{-\infty}^{\infty} f(x_1, x_2) dx_2 = \int_{-\infty}^{\infty} g(x_1) h(x_2) dx_2$$

$$= g(x_1) \int_{-\infty}^{\infty} h(x_2) dx_2 = C_1 g(x_1),$$

where $C_1 = \int_{-\infty}^{\infty} h(x_2) dx_2$ is a constant.

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$$\begin{aligned}
 \text{Here, } f_2(x_2) &= \int_{-\infty}^{\infty} f(x_1, x_2) dx_1 \\
 &= h(x_2) \int_{-\infty}^{\infty} g(x_1) dx_1 \\
 &= c_2 h(x_2), \text{ where } c_2 = \int_{-\infty}^{\infty} g(x_1) dx_1, \\
 &\quad \text{is a constant}
 \end{aligned}$$

By defⁿ of pdf, we have

$$\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(x_1, x_2) dx_1 dx_2 = 1$$

$$\Rightarrow \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} g(x_1) h(x_2) dx_1 dx_2 = 1$$

$$\Rightarrow \left(\int_{-\infty}^{\infty} g(x_1) dx_1 \right) \left(\int_{-\infty}^{\infty} h(x_2) dx_2 \right) = 1$$

$$\Rightarrow c_1 c_2 = 1$$

$$\begin{aligned}
 \text{Thus, } f(x_1, x_2) &= g(x_1) h(x_2) \\
 &= c_1 c_2 g(x_1) h(x_2) \\
 &= c_1 g(x_1) \cdot c_2 h(x_2)
 \end{aligned}$$

$$\Rightarrow f(x_1, x_2) = f_1(x_1) f_2(x_2)$$

$\Rightarrow X_1$ and X_2 are independent random variables.

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(5)

Example 9. Find $P(0 < X_1 < \frac{1}{3}, 0 < X_2 < \frac{1}{3})$ if the random variables X_1 and X_2 have the joint pdf $f(x_1, x_2) = 4x_1(1-x_2), 0 < x_1 < 1, 0 < x_2 < 1$, zero elsewhere.

Solution. We have

$$f(x_1, x_2) = \begin{cases} 4x_1(1-x_2), & 0 < x_1 < 1, \\ & 0 < x_2 < 1 \\ 0 & \text{elsewhere} \end{cases}$$

$$\text{Let } g(x_1) = \begin{cases} 2x_1, & 0 < x_1 < 1 \\ 0 & \text{elsewhere} \end{cases}, \text{ and}$$

$$h(x_2) = \begin{cases} 2(1-x_2), & 0 < x_2 < 1 \\ 0 & \text{elsewhere} \end{cases}$$

Then $f(x_1, x_2) = g(x_1)h(x_2)$, where

$g(x_1) > 0$ if $x_1 \in S_1$ & $h(x_2) > 0$ if $x_2 \in S_2$

Hence, X_1 and X_2 are independent random variables.

$$\Rightarrow P(0 < X_1 < \frac{1}{3}, 0 < X_2 < \frac{1}{3})$$

$$= \left(\int_0^{\frac{1}{3}} 2x_1 dx_1 \right) \left(\int_0^{\frac{1}{3}} 2(1-x_2) dx_2 \right) \quad \left(\begin{smallmatrix} \text{See} \\ \text{Theorem 4} \\ \text{on page 17} \end{smallmatrix} \right)$$

$$= \left(\frac{1}{3} \right)^2 [1 - (2/3)^2] = \frac{5}{18} \quad //$$

Theorem 3. Let $F(x_1, x_2)$ be the joint cdf of random vector (X_1, X_2) , and $F_1(x_1)$ ($F_2(x_2)$) be the marginal cdf of X_1 (X_2), respectively. Then X_1 and X_2 are independent if, and only if

$$F(x_1, x_2) = F_1(x_1) F_2(x_2) \quad \forall (x_1, x_2) \in \mathbb{R}^2$$

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(6)
Proof. Let X_1 and X_2 be independent random variables.

Then $f(x_1, x_2) = f_1(x_1) f_2(x_2)$ ——— (i)

By definition, $F(x_1, x_2) = P(X_1 \leq x_1, X_2 \leq x_2)$

$$= \int_{-\infty}^{x_1} \int_{-\infty}^{x_2} f(w_1, w_2) dw_2 dw_1$$

$$= \int_{-\infty}^{x_1} \int_{-\infty}^{x_2} f_1(w_1) f_2(w_2) dw_2 dw_1 \quad \text{by (i)}$$

$$= \left(\int_{-\infty}^{x_1} f_1(w_1) dw_1 \right) \left(\int_{-\infty}^{x_2} f_2(w_2) dw_2 \right)$$

$$= F_1(x_1) F_2(x_2)$$

$$\therefore F(x_1, x_2) = F_1(x_1) F_2(x_2)$$

Conversely, Suppose

$$F(x_1, x_2) = F_1(x_1) F_2(x_2)$$

Then,

$$f(x_1, x_2) = \frac{\partial^2}{\partial x_1 \partial x_2} F(x_1, x_2)$$

$$= \frac{\partial^2}{\partial x_1 \partial x_2} F_1(x_1) F_2(x_2)$$

$$= \frac{\partial}{\partial x_1} \left(\frac{\partial}{\partial x_2} F_1(x_1) F_2(x_2) \right)$$

$$= \frac{\partial}{\partial x_1} \left(F_1(x_1) \frac{\partial}{\partial x_2} F_2(x_2) \right)$$

$$= \frac{\partial}{\partial x_1} \left(F_1(x_1) f_2(x_2) \right)$$

$$= f_2(x_2) \frac{\partial}{\partial x_1} F_1(x_1) = f_2(x_2) f_1(x_1)$$

$\therefore f(x_1, x_2) = f_1(x_1) f_2(x_2)$, hence X_1
and X_2 are independent. //

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(7)

Theorem 4. The random variables X_1 and X_2 are independent random variables if, and only if the following condition holds:

$$P(a < X_1 \leq b, c < X_2 \leq d) = P(a < X_1 \leq b) P(c < X_2 \leq d) \quad \text{--- (ii)}$$

Proof: Let X_1 and X_2 be independent. Then

$$\begin{aligned} & P(a < X_1 \leq b, c < X_2 \leq d) \\ &= F(b, d) - F(a, d) - F(b, c) + F(a, c) \\ &= F_1(b) F_2(d) - F_1(a) F_2(d) - F_1(b) F_2(c) + F_1(a) F_2(c) \\ &= [F_1(b) - F_1(a)] [F_2(d) - F_2(c)] \quad (\text{using Theorem 3}) \\ &= P(a < X_1 \leq b) P(c < X_2 \leq d) \end{aligned}$$

Conversely, suppose the condition (ii) holds.

Consider $F(x_1, x_2) = P(-\infty < X_1 \leq x_1, -\infty < X_2 \leq x_2)$

$$= P(-\infty < X_1 \leq x_1) P(-\infty < X_2 \leq x_2) \quad \text{using (ii)}$$

$$\Rightarrow F(x_1, x_2) = F_1(x_1) F_2(x_2)$$

Hence, using Theorem 3, it implies that X_1 and X_2 are independent.

Example 10. Let the joint pdf of X_1 and X_2 be

$$f(x_1, x_2) = \begin{cases} x_1 + x_2, & 0 < x_1 < 1, 0 < x_2 < 1 \\ 0 & \text{elsewhere} \end{cases}$$

Show that X_1 and X_2 are dependent.

Solution, we have

$$P(0 < X_1 \leq \frac{1}{2}, 0 < X_2 \leq \frac{1}{2}) = \int_0^{\frac{1}{2}} \int_0^{\frac{1}{2}} (x_1 + x_2) dx_1 dx_2 = \frac{1}{8},$$

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SUSHIL KUMAR

(8)

and the marginal pdfs of X_1 and X_2 are

$$f_1(x_1) = \begin{cases} \int_0^1 (x_1 + x_2) dx_2 = x_1 + \frac{1}{2}, & 0 < x_1 < 1 \\ 0 & \text{elsewhere} \end{cases}$$

$$f_2(x_2) = \begin{cases} \int_0^1 (x_1 + x_2) dx_1 = \frac{1}{2} + x_2, & 0 < x_2 < 1 \\ 0 & \text{elsewhere} \end{cases}$$

$$\begin{aligned} \text{Now, } P(0 < X_1 \leq \frac{1}{2}) &= \int_0^{\frac{1}{2}} f_1(x_1) dx_1 \\ &= \int_0^{\frac{1}{2}} (x_1 + \frac{1}{2}) dx_1 = \frac{3}{8}, \end{aligned}$$

$$\begin{aligned} \text{and } P(0 < X_2 \leq \frac{1}{2}) &= \int_0^{\frac{1}{2}} (\frac{1}{2} + x_2) dx_2 \\ &= \frac{3}{8}. \end{aligned}$$

Since, the condition (ii) does not hold, X_1 and X_2 are dependent. //

Theorem 5, Suppose X_1 and X_2 are independent, and that $E(u(X_1))$ and $E(v(X_2))$ exist. Then

$$E[u(X_1)v(X_2)] = E[u(X_1)] E[v(X_2)]$$

Proof: Since X_1 and X_2 are independent,
 $f(x_1, x_2) = f_1(x_1)f_2(x_2)$.

By definition,

$$\begin{aligned} E[u(X_1)v(X_2)] &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} u(x_1)v(x_2)f_1(x_1)f_2(x_2) dx_1 dx_2 \\ &= \left[\int_{-\infty}^{\infty} u(x_1)f_1(x_1) dx_1 \right] \left[\int_{-\infty}^{\infty} v(x_2)f_2(x_2) dx_2 \right] \\ &= E[u(X_1)] E[v(X_2)]. \quad // \end{aligned}$$

SUSHIL KUMAR AZAD

(9)
 We now prove a very useful theorem about independent random variables.

Theorem 6. Suppose the joint mgf, $M(t_1, t_2)$ exists for the random variables X_1 and X_2 . Then X_1 and X_2 are independent if, and only if $M(t_1, t_2) = M(t_1, 0) M(0, t_2)$ (ii)

Proof: Let X_1 and X_2 be independent.

By definition,

$$M(t_1, t_2) = E[e^{t_1 X_1 + t_2 X_2}]$$

$$= E[e^{t_1 X_1} e^{t_2 X_2}]$$

$$= E[e^{t_1 X_1}] E[e^{t_2 X_2}] \text{, using Theorem 5.}$$

$$\Rightarrow M(t_1, t_2) = M(t_1, 0) M(0, t_2)$$

Conversely, Suppose the condition (ii) holds.

By definition,

$$M(t_1, 0) = E(e^{t_1 X_1})$$

$$= \int_{-\infty}^{\infty} e^{t_1 x_1} f_1(x_1) dx_1, \text{ and}$$

$$M(0, t_2) = E(e^{t_2 X_2})$$

$$= \int_{-\infty}^{\infty} e^{t_2 x_2} f_2(x_2) dx_2$$

Thus, we have

$$M(t_1, 0) M(0, t_2) = \left[\int_{-\infty}^{\infty} e^{t_1 x_1} f_1(x_1) dx_1 \right] \left[\int_{-\infty}^{\infty} e^{t_2 x_2} f_2(x_2) dx_2 \right]$$

$$= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} e^{t_1 x_1 + t_2 x_2} f_1(x_1) f_2(x_2) dx_1 dx_2$$

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SUSHIL KUMAR ARORA

(10)

Now, $M(t_1, t_2) = M(t_1, 0) M(0, t_2)$ by assumption

$$\Rightarrow M(t_1, t_2) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} e^{t_1 x_1 + t_2 x_2} f_1(x_1) f_2(x_2) dx_1 dx_2 \quad \text{--- (i)}$$

but, by defⁿ

$$\begin{aligned} M(t_1, t_2) &= E[e^{t_1 x_1 + t_2 x_2}] \\ &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} e^{t_1 x_1 + t_2 x_2} f(x_1, x_2) dx_1 dx_2 \end{aligned} \quad \text{--- (ii)}$$

Comparing (i) & (ii), we get

$$f(x_1, x_2) = f_1(x_1) f_2(x_2)$$

Since the mgf of $X_1 + X_2$, $M(t_1, t_2)$ is unique.

$\Rightarrow X_1$ and X_2 are independent. //

Example 11. Let X_1 and X_2 be random variables

with joint pmf $p(x_1, x_2) = \frac{1}{2^{x_1 + x_2}}$, for

$1 \leq x_i < \infty$, $i=1, 2$, where x_1 and x_2 are integers, zero elsewhere. Determine the

joint mgf of (X_1, X_2) and, show that

X_1 and X_2 are independent random variables.

Solution. By definition, the mgf of X_1 and X_2 ,

$$\begin{aligned} M(t_1, t_2) &= E[e^{t_1 X_1 + t_2 X_2}] \\ &= \sum_{x_1=1}^{\infty} \sum_{x_2=1}^{\infty} e^{t_1 x_1 + t_2 x_2} \frac{1}{2^{x_1 + x_2}} \end{aligned}$$

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SUSHIL KUMAR A-26D

$$= \sum_{x_1=1}^{\infty} \sum_{x_2=1}^{\infty} \left[e^{t_1 x_1} \left(\frac{1}{2}\right)^{x_1} e^{t_2 x_2} \left(\frac{1}{2}\right)^{x_2} \right] \quad (11)$$

$$= \sum_{x_1=1}^{\infty} \left[\frac{e^{t_1}}{2} \right]^{x_1} \sum_{x_2=1}^{\infty} \left[\frac{e^{t_2}}{2} \right]^{x_2}$$

$$= \left[\frac{\left(\frac{e^{t_1}}{2}\right)}{1 - \left(\frac{e^{t_1}}{2}\right)} \right] \left[\frac{\left(\frac{e^{t_2}}{2}\right)}{1 - \left(\frac{e^{t_2}}{2}\right)} \right]$$

$$M(t_1, t_2) = \left[\frac{e^{t_1}}{2 - e^{t_1}} \right] \left[\frac{e^{t_2}}{2 - e^{t_2}} \right],$$

Now, the marginal pmfs of X_1 and X_2 are,

$$\begin{aligned} p_1(x_1) &= \sum_{x_2=1}^{\infty} p(x_1, x_2) \\ &= \frac{1}{2^{x_1}} \sum_{x_2=1}^{\infty} \left(\frac{1}{2}\right)^{x_2} \\ &= \frac{1}{2^{x_1}}, \quad 1 \leq x_1 < \infty \end{aligned}$$

$$\& \quad p_2(x_2) = \frac{1}{2^{x_2}}, \quad 1 \leq x_2 < \infty.$$

$$\begin{aligned} \text{Thus, } M(t_1, 0) &= E[e^{t_1 X_1}] \\ &= \sum_{x_1=1}^{\infty} e^{t_1 x_1} p_1(x_1) \\ &= \sum_{x_1=1}^{\infty} e^{t_1 x_1} \frac{1}{2^{x_1}} = \sum_{x_1=1}^{\infty} \left(\frac{e^{t_1}}{2}\right)^{x_1} \\ &= \frac{e^{t_1}}{2 - e^{t_1}}, \end{aligned}$$

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(12)

Similarly, $M(0, t_2) = \frac{e^{t_2}}{(2 - e^{t_2})}$

Thus, $M(t_1, 0) M(0, t_2)$
 $= \left(\frac{e^{t_1}}{2 - e^{t_1}} \right) \left(\frac{e^{t_2}}{2 - e^{t_2}} \right)$
 $= M(t_1, t_2)$

$\Rightarrow X_1$ and X_2 are independent.

Example 12. Let X and Y have the joint pdf

$$f(x, y) = \begin{cases} 3x, & 0 < y < x < 1 \\ 0, & \text{elsewhere} \end{cases}$$

Are X and Y independent? If not, find $E(X|Y)$.

Solution. The marginal pdfs of X and Y are

$$f_X(x) = \int_0^x f(x, y) dy$$

$$= \int_0^x 3x dy = 3x^2$$

$$\therefore f_X(x) = \begin{cases} 3x^2, & 0 < x < 1 \\ 0, & \text{elsewhere} \end{cases}$$

and $f_Y(y) = \int_y^1 3x dx = \begin{cases} \frac{3}{2}(1 - y^2), & 0 < y < x < 1 \\ 0, & \text{elsewhere} \end{cases}$

Now, $f_X(x) f_Y(y) = \begin{cases} \frac{9}{2} x^2 (1 - y^2), & 0 < y < x < 1 \\ 0, & \text{elsewhere} \end{cases}$

which is not equal to $f(x, y)$, implies that X and Y are dependent.

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SUSHIL KUMAR 714575

(13)

Next $f_{1/2}(x|y) = \frac{f(x,y)}{f_y(y)}$, $y \in S_y$

$$= \frac{3x}{\left(\frac{3}{2}\right)(1-y^2)} = \frac{2x}{(1-y^2)}$$

$$\Rightarrow E(X|y) = \int_y^1 x f_{1/2}(x|y) dx$$

$$= \int_0^1 \frac{2x^2}{(1-y^2)} dx = \frac{2}{(1-y^2)} \times \frac{1}{3}(1-y^3)$$

$$\Rightarrow E(X|y) = \frac{2(1+y+y^2)}{3(1+y)} \quad //$$

SUNIL KUMAR 1202