

$$\begin{aligned}
 P[X < a] &= \int_0^a \int_0^\infty 2e^{-x} e^{-2y} dy dx \\
 &= \int_0^a 2e^{-x} \int_0^\infty e^{-2y} dy dx \\
 &= \int_0^a 2e^{-x} \left[-\frac{1}{2} e^{-2y} \right]_0^\infty dx \\
 &= \int_0^a e^{-x} (0 + 1) = \int_0^a e^{-x} dx = \left[-e^{-x} \right]_0^a = 1 - e^{-a}
 \end{aligned}$$

$$\begin{aligned}
 P[X < Y] &= \int_0^\infty \int_x^\infty 2e^{-x} e^{-2y} dy dx \\
 &= \int_0^\infty e^{-x} \int_x^\infty 2e^{-2y} dy dx \\
 &= \int_0^\infty e^{-x} \left[-e^{-2y} \right]_x^\infty dx \\
 &= \int_0^\infty e^{-x} (0 + e^{-2x}) dx = \int_0^\infty e^{-3x} dx \\
 &= \left[-\frac{1}{3} e^{-3x} \right]_0^\infty = \frac{1}{3}
 \end{aligned}$$

2.3 Independent Random Variables

X, Y are independent iff

$$P[X \in A, Y \in B] = P[X \in A] \cdot P[Y \in B]$$

for all (reasonable) $A, B \subseteq \mathbb{R}$.

Equivalent (proof!)

$$P[X \leq a, Y \leq b] = P[X \leq a, Y \leq b] \quad \text{f.a. } a, b \in \mathbb{R}$$

that is

$$F(a, b) = F_X(a) \cdot F_Y(b)$$

Equivalent for discrete R.Vs:

$$P(X, Y) = P_X(x) \cdot P_Y(y) \quad \text{f.a. } x, y \in \mathbb{R}$$

for the probability mass functions;

$$f(x, y) = f_X(x) f_Y(y) \quad \text{f.a. } x, y \in \mathbb{R}$$

for the densities of continuous R.Vs:

Example 33. Let X, Y be independent, each with density

$$f(x) = \begin{cases} e^{-x} & x > 0 \\ 0 & \text{otherwise} \end{cases}$$

Note: The probability $P[Y \leq 0] = 0$ because

$$P[Y > 0] = \int_0^\infty e^{-x} dx = 1.$$

What is the density of $\frac{X}{Y}$?

To get the distribution function, we calculate:

$$\begin{aligned}
P[\frac{X}{Y} \leq a] &= P[X \leq aY] \\
&= \int_0^\infty \int_0^{ay} e^{-x} e^{-y} dx dy \\
&= \int_0^\infty e^{-y} \left[-e^{-x} \right]_0^{ay} dy = \int_0^\infty e^{-y} (1 - e^{-ay}) dy \\
&= \int_0^\infty e^{-y} - \int_0^\infty e^{-(1+a)y} dy \\
&= 1 - \left[-\frac{1}{1+a} e^{-(1+a)y} \right]_0^\infty \\
&= 1 + \left(0 - \frac{1}{1+a} \right) = 1 - \frac{1}{1+a}
\end{aligned}$$

The density λ

$$\frac{d}{da} \left(1 - \frac{1}{1+a}\right) = \frac{d}{da} (1 - (1+a)^{-1}) = (1+a)^{-2},$$

that is $f_{\text{density}}(a) = \frac{1}{(1+a)^2}$

Joint probability distributions can be defined for RVs X_1, \dots, X_n . Also

joint mass functions

joint densities

independence.

Example 34 The price of a stock changes daily by amount x with probability $p(x)$, with

$$p(x) = \begin{cases} 0.05 & x \in \{-3, 3\} \\ 0.10 & x \in \{-2, 2\} \\ 0.15 & x \in \{-1, 1\} \\ 0.40 & x = 0 \end{cases}$$

Let X_i be the change on day i , and assume the X_i are independent. Then

$$P[X_1=1, X_2=2, X_3=0] = p(1)p(2)p(0)$$

$$= 0.15 \times 0.10 \times 0.4$$

$$= 0.006$$