

The probability density of the sum of two uncorrelated random variables is not necessarily the convolution of its two marginal densities

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If two random variables X and Y are *independent*, then the probability density of their sum is equal to the convolution of the probability densities of X and Y . With obvious notation, we have

$$p_{X+Y}(z) = \int dx p_X(x) p_Y(z-x) . \quad (1)$$

The proof is simple: Independence of the two random variables implies that

$$p_{X,Y}(x,y) = p_X(x) p_Y(y) . \quad (2)$$

And by the transformation theorem for probability densities we immediately get

$$\begin{aligned} p_{X+Y}(z) &= \int dx \int dy p_{X,Y}(x,y) \delta(x+y-z) \\ &= \int dx p_{X,Y}(x, z-x) \end{aligned} \quad (3)$$

$$\stackrel{(2)}{=} \int dx p_X(x) p_Y(z-x) . \quad (4)$$

We here want to convince ourselves by a counterexample that *uncorrelatedness* of the random variables does *not suffice* for the convolution formula to hold. To see this, let us look at the probability density

$$p_{X,Y}(x,y) = \frac{x^2 + y^2}{4\pi} e^{-\frac{1}{2}(x^2+y^2)} . \quad (5)$$

This probability density evidently does not factorize. Indeed, the marginal densities are given by

$$p_X(x) = \int dy p_{X,Y}(x,y) = \frac{1+x^2}{\sqrt{8\pi}} e^{-\frac{1}{2}x^2} , \quad (6)$$

with the same functional form of course also holding for $p_Y(y)$. Evidently, Eqn. (2) does not hold for this choice of $p_{X,Y}(x,y)$ and its marginal densities $p_X(x)$ and $p_Y(y)$. However, since $p_{X,Y}(x,y)$ is rotationally symmetric about the origin, the covariance of X and Y vanishes, hence X and Y are *uncorrelated, and yet dependent*.

What is now the probability density of $X+Y$? From the transformation theorem we get

$$\begin{aligned} p_{X+Y}(z) &\stackrel{(3)}{=} \int dx p_{X,Y}(x, z-x) \\ &\stackrel{(5)}{=} \int dx \frac{x^2 + (z-x)^2}{4\pi} e^{-\frac{1}{2}[x^2+(z-x)^2]} \\ &= \frac{2+z^2}{8\sqrt{\pi}} e^{-\frac{1}{4}z^2} . \end{aligned} \quad (7)$$

On the other hand, the convolution of p_X and p_Y is

$$\begin{aligned} [p_X * p_Y](z) &= \int dx p_X(x) p_Y(z-x) \\ &\stackrel{(6)}{=} \int dx \frac{1+x^2}{\sqrt{8\pi}} e^{-\frac{1}{2}x^2} \frac{1+(z-x)^2}{\sqrt{8\pi}} e^{-\frac{1}{2}(z-x)^2} \\ &= \frac{z^4 + 4z^2 + 44}{128\sqrt{\pi}} e^{-\frac{1}{4}z^2} , \end{aligned} \quad (8)$$

which differs from the correct answer. Fig. 1 illustrates the difference between these two functions.

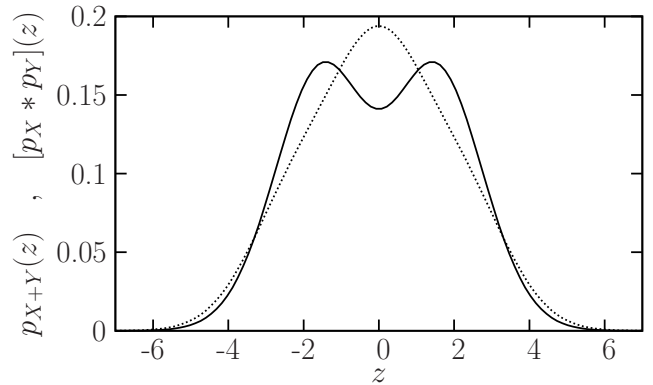


FIG. 1: True probability density of the sum random variable $p_{X+Y}(z)$ (solid line) and convolution of its marginal densities, $[p_X * p_Y](z)$ (dotted line).