

Independent continuous random variables

Several ways to specify whether two random variables X and Y are independent. All of these ways are equivalent to each other.

One way is to show that the joint density (i.e., joint probability density function) of X and Y can be factored into two parts, one of which only has x 's and the other of which only has y 's. With some possible rescaling, this means that the two parts are the densities of X and Y respectively, and it also means that X and Y are independent continuous random variables. In other words, X and Y are independent continuous random variables if and only if their joint density can be factored into a product of their (single variables) densities:

$$f_{X,Y}(x,y) = f_X(x)f_Y(y) \quad \text{for all } x,y.$$

Another possible formulation of independence (which is equivalent to the one above) is that the joint CDF (joint cumulative distribution function) factors into a product of two functions, one of which only has x 's, and the other of which only has y 's. Again, with some possible rescaling, the result is that the joint CDF factors into the CDF of X times the CDF of Y . I.e., X and Y are independent if and only if

$$F_{X,Y}(x,y) = F_X(x)F_Y(y) \quad \text{for all } x,y.$$

One more equivalent interpretation. We have not (yet) said what the conditional density of a continuous random variable is, but once we learn it, we can formulate independence in terms of the conditional density. Namely, X and Y are independent continuous random variables if and only if

$$f_{X|Y}(x|y) = f_X(x) \quad \text{for all } x,y,$$

or equivalently,

$$f_{Y|X}(y|x) = f_Y(y) \quad \text{for all } x,y.$$