

## Result

This result is called sometimes called the *probability inverse transform*.

**Proposition 1.** Let  $(\Omega, \mathcal{F}, \mathbf{P})$  be a probability space and let  $X : \Omega \to \mathbf{R}$  be a random variable with cumulative distribution function  $F_X : \mathbf{R} \to [0,1]$ . Suppose  $F_X^{-1} : [0,1] \to \mathbf{R}$  exists, then  $Y = F_X^{-1} \circ X$  is a random variable with cumulative distribution function  $F_Y : [0,1] \to [0,1]$  satisfying  $F_Y(y) = y$ .

**Remark 1.** The conclusion is equivalent to the following: Y has a density and that density is the standard unform density (see Uniform Densities).

Proof. Express 
$$F_Y(\gamma) = \mathbf{P}[Y \le \gamma] = \mathbf{P}(Y^{-1}([0, \gamma]))$$
 Notice<sup>1</sup>

$$Y^{-1}([0, \gamma]]) = \{\omega \in \Omega \mid Y(\omega) \le \gamma\}$$

$$= \{\omega \in \Omega \mid F_X(X(\omega)) \le \gamma\}$$

$$= \{\omega \in \Omega \mid X(\omega) \le F_X^{-1}(\gamma)\}. = X^{-1}(\cdots).$$

**Remark 2.** Using different notation the above can be expressed succinctly as

$$F_Y(\gamma) = \mathbf{P}[Y \le \gamma] = \mathbf{P}[F_X \circ X \le \gamma]$$
$$= \mathbf{P}[X \le F_X^{-1}(\gamma)] = F_X(F_X^{-1}(\gamma)) = \gamma.$$

Future editions will discuss  $inverse\ transform\ sampling.$ 

<sup>&</sup>lt;sup>1</sup>Future editions will complete.

