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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
$$Y^{-1}([0, \gamma]]) = \{\omega \in \Omega \mid Y(\omega) \le \gamma\}$$
$$= \{\omega \in \Omega \mid F_X(X(\omega)) \le \gamma\}$$
$$= \{\omega \in \Omega \mid X(\omega) < F_Y^{-1}(\gamma)\}. = X^{-1}(\cdots).^2$$

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¹Future editions will include.

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.

