



Why

Given an outcome variable $X : \Omega \rightarrow V$, and some probabilities on a sample space Ω , there is a natural set of probabilities to associate with outcomes in V .

Result

Suppose $X : \Omega \rightarrow V$ is a random variable on a finite sample space Ω . Given $p : \Omega \rightarrow \mathbf{R}$ is a probability distribution inducing probability measure $P : \mathcal{P}(\Omega) \rightarrow \mathbf{R}$, define the function $q : V \rightarrow \mathbf{R}$ by

$$q(x) = P(X = x)$$

It is easy to verify that q is nonnegative and normalized. The latter fact follows from the observation that the sets $\{X^{-1}(x)\}_{x \in V}$ partition the set Ω .

The function q is sometimes called the *distribution of X* (the *induced distribution*, *induced probability distribution*) of the random variable X .

Notation

Given a random variable $X : \Omega \rightarrow V$, and some distribution $p : \Omega \rightarrow \mathbf{R}$, it is common to denote the induced distribution by p_X . Given a distribution $q : V \rightarrow \mathbf{R}$ and a random variable $X : \Omega \rightarrow V$ it is common to see the notation $X \sim q$ as an abbreviation of the sentence “The random variable X has distribution q .”

Computation

Suppose $X : \Omega \rightarrow V$ is a random variable and $f : V \rightarrow U$. Define $Y : \Omega \rightarrow U$ by $y(\omega) \equiv f(x(\omega))$ for every $\omega \in \Omega$. We frequently denote this by $Y = f(X)$. Y is a random variable with induced distribution $p_Y : U \rightarrow \mathbf{R}$ satisfying

$$p_Y(y) = \sum_{\omega \in \Omega | Y(\omega)=y} p(\omega) = \sum_{x \in V | f(x)=y} p_X(x).$$

Consequently, as a matter of practical computation, we can evaluate probabilities having to do with the outcome variable X using p_X instead of p and same with Y .¹

¹Future editions will give an example.

