

OUTCOME VARIABLE EXPECTATION

Why

If we model some measured value as a random variable with induced distribution $p:V\to \mathbf{R}$, then one interpretation of p(v) for $v\in V$ is the proportion of times in a large number of trials that we expect to measure the value $v.^1$

Definition

Given a distribution $p: \Omega \to \mathbf{R}$ and a real-valued outcome variable $x: \Omega \to \mathbf{R}$, the expectation (or mean) of x under p is $\sum_{\omega \in \Omega} p(\omega) x(\omega)$.

Notation

We denote the expectation of x under p by $\mathbf{E}_p(x)$. When there is no chance of ambiguity, we write $\mathbf{E}(x)$.

Properties

Let $x, y: \Omega \to \mathbf{R}$ be two outcome variables and $p: \Omega \to \mathbf{R}$ a distribution. Let $\alpha, \beta \in \mathbf{R}$. Define $z = \alpha x + \beta y$ by $z(\omega) = \alpha x(\omega) + \beta y(\omega)$. Then $\mathbf{E}(z) = \alpha \mathbf{E}(x) + \beta \mathbf{E}(z)$. Many authors refer to this property as the *linearity* of expectation.

Example: expectation

Suppose $\Omega = \{1, 2, 3, 4, 5\}$ with p(1) = 0.1, p(2) = 0.15, p(3) = 0.1, p(5) = 0.25 and p(5) = 0.4. Define $x : \Omega \to \mathbf{R}$ by

$$x(a) = \begin{cases} -1 & \text{if } a = 1 \text{ or } a = 2, \\ 1 & \text{if } a = 3 \text{ or } a = 4, \\ 2 & \text{if } a = 5. \end{cases}$$

 $^{^1\}mathrm{Future}$ editions may modify this explanation, and take a genetic approach via summary statistics.

The expectation of x under p is

$$\mathbf{E}x = -1 - 0.15 + 0.1 + 0.25 + 2(0.4) = 0.9.$$

Two routes for computation

Denote by $p_x:V\to \mathbf{R}$ the induced distribution of $x:\Omega\to V$ (where $V\subset \mathbf{R}$). Then $\mathbf{E}(x)=\sum_{v\in V}p_x(v)v$ since

$$\sum_{\omega \in \Omega} p(\omega)x(\omega) = \sum_{v \in V} \sum_{\omega \in x^{-1}(v)} x(\omega)p(\omega)$$
$$= \sum_{v \in V} v \sum_{\omega \in x^{-1}(v)} p(\omega)$$
$$= \sum_{v \in V} x(v)p_x(v).$$

Interpretations

We interpret the mean as the center of mass of the induced distribution.

