Repetition – conditional probability

Definition

Let A and B be events. The **conditional probability** of A given B is defined by

$$\Pr\{A|B\} = \begin{cases} \frac{\Pr\{A \cap B\}}{\Pr\{B\}}, & \Pr\{B\} > 0, \\ \text{Not defined}, & \Pr\{B\} = 0. \end{cases}$$

Definition

Assume X and Y are jointly distributed random variables. The **conditional PMF** $p_{X|Y}(x|y)$ of X given Y is given by

$$p_{X|Y}(x|y) = \frac{\Pr\{X = x, Y = y\}}{\Pr\{Y = y\}} = \frac{p_{X,Y}(x,y)}{p_Y(y)}, \text{ if } p_Y(y) > 0.$$

Theorem

The conditional PMF is essential to us because we can calculate the joint PMF as

$$p_{X,Y}(x,y) = p_Y(y)p_{X|Y}(x|y).$$

Theorem

Let X and Y be random variables. Then **the law of total probability** gives

$$\begin{split} \Pr\{X = x\} &= \sum_{y} \Pr\{X = x, Y = y\} \\ &= \sum_{y} \Pr\{Y = y\} \Pr\{X = x | Y = y\}, \\ \Pr\{Y = y\} &= \sum_{x} \Pr\{Y = y, X = x\} \\ &= \sum_{x} \Pr\{X = x\} \Pr\{Y = y | X = x\}. \end{split}$$

Definition

Let X and Y be random variables, and g a real function. The **conditional** expected value of g(X) given Y = y is

$$E[g(X)|Y = y] = \sum_{x} g(x) Pr\{X = x|Y = y\}, \text{ if } Pr\{Y = y\} > 0.$$

Theorem

Let X and Y be random variables, then

$$E[X] = E[E[X|Y]],$$

$$Var[X] = E[Var[X|Y]] + Var[E[X|Y]].$$

The former is often called **the law of iterated expectation** and the latter **the law of total variance**.

Note: E[X|Y] is a random variable that takes value E[X|Y=y] with probability $Pr\{Y=y\}$.

Note 2: The conditional variance can be written as

$$Var[X|Y] = E[X^2|Y] - E[X|Y]^2.$$