

#### **EVENT PROBABILITIES**

### Why

Since one and only one outcome occurs, given a distribution on outcomes, we define the probability of a set of outcomes as the sum of their probabilities.

#### Definition

Suppose p is a distribution on  $\Omega$ . For any event  $A \subset \Omega$ , we call the value  $\sum_{a \in A} p(a)$  the *event probability*. We refer to the probability of A. The probability A is the sum of the probabilities of its outcomes.

We can define a function  $\mathbf{P}: \mathcal{P}(\Omega) \to \mathbf{R}$  by  $\mathbf{P}(A) = \sum_{a \in A} p(a)$ . We call  $\mathbf{P}$  the event probability function (or the probability measure) induced by p. Since  $\mathbf{P}$  depends on the sample space  $\Omega$  and the distribution p, we ocassionally denote this dependence by  $\mathbf{P}_{\Omega,p}$  or  $\mathbf{P}_p$ .

### Example: die

Define  $p:\{1,\ldots,6\}\to \mathbf{R}$  by  $p(\omega)=1/6$  for  $\omega=1,\ldots,6$ . Define the event  $E=\{2,4,6\}$ . Then

$$\mathbf{P}(E) = \sum_{\omega \in E} p(\omega) = p(2) + p(4) + p(6) = 1/2.$$

# Properties of P

As a result of the conditions on p,  $\mathbf{P}$  satisfies

- 1.  $\mathbf{P}(A) \geq 0$  for all  $A \subset \Omega$ ;
- 2.  $\mathbf{P}(\Omega) = 1 \text{ (and } \mathbf{P}(\varnothing) = 0);$
- 3.  $\mathbf{P}(A) + \mathbf{P}(B)$  for all  $A, B \subset \Omega$  and  $A \cap B = \emptyset$ . This statement follows from the more general identity

$$\mathbf{P}(A \cup B) = \mathbf{P}(A) + \mathbf{P}(B) - \mathbf{P}(A \cap B)$$

for  $A, B \subset \Omega$ , by using  $P(\emptyset) = 0$  of (2) above.

Do all such **P** satisfying (1)-(3) have a corresponding underlying probability distribution? In other words, suppose  $f: \mathcal{P}(\Omega) \to \mathbf{R}$  satisfies (1)-(3). These three conditions are sometimes called the *axioms of probability* for finite sets.

Define  $q: \Omega \to \mathbf{R}$  by  $q(\omega) = f(\{\omega\})$ . If f satisfies the axioms, then q is a probability distribution. For this reason we call any function satisfying (i)-(iii) an event probability function (or a probability measure).

## Probability by cases

Let **P** be a probability event function. Suppose  $A_1, \ldots, A_n$  partition  $\Omega$ . Then for any  $B \subset \Omega$ ,

$$\mathbf{P}(B) = \sum_{i=1}^{n} \mathbf{P}(A_i \cap B).$$

Some authors call this the *law of total probability*.

