



## Why

We can extend the notion of independence beyond pairs of uncertain events, to sets of events.

## Definition

Suppose  $P$  is an event probability function on a finite sample space  $\Omega$ . The events  $A_1, \dots, A_n$  are *independent* (or *mutually independent*), if for all  $k$  between 1 and  $n$ , and distinct indices  $i_1, \dots, i_k$  between 1 and  $n$ ,

$$P(A_{i_1} \cap A_{i_2} \cap \dots \cap A_{i_k}) = P(A_{i_1})P(A_{i_2}) \dots P(A_{i_k}).$$

Similar to the case of pairs of events, one can show that this condition is equivalent to the statement that for any *distinct* indices  $i_1, \dots, i_k, j_1, \dots, j_l$ ,

$$P(A_{j_1} \cap \dots \cap A_{j_l} \mid A_{i_1} \cap \dots \cap A_{i_k}) = P(A_{j_1} \cap \dots \cap A_{j_l})$$

## Examples

*n* tosses of a coin. As usual, model  $n$  tosses of a coin with  $\{0, 1\}^n$  and put a distribution  $p : \Omega \rightarrow [0, 1]$  so that

$$p(\omega) = 1/2^n \quad \text{for all } \omega \in \Omega$$

Now, for  $i = 1, \dots, n$ , define the event  $A_i$  by

$$A_i = \{\omega \in \Omega \mid \omega(i) = 1\}$$

We claim that the set  $\{A_1, \dots, A_n\}$  is mutually independent. To see this, notice that for any distinct indices  $i_1, \dots, i_k$ ,

$$|A_{i_1} \cap \dots \cap A_{i_k}| = 2^{n-k}$$

This holds because, once  $k$  of the coin flips, there are  $2^{n-k}$  ways for the remaining coins to land (using the fundamental principle of counting). Consequently,

$$P(A_{i_1} \cap \dots \cap A_{i_k}) = \frac{2^{n-k}}{2^n} = 2^{-k}$$

We can use this result with one set  $P(A_i) = 1/2$ , and so we obtain

$$P(A_{i_1} \cap \cdots A_{i_k}) = P(A_{i_1}) \cdots P(A_{i_k}),$$

as desired.

### **Basic implications**

It can be shown<sup>1</sup> that if  $A_1, \dots, A_n$  are independent events, and  $B_1, \dots, B_n$  are events such that  $B_i$  is either  $A_i$  or  $A_i^c$ , then  $B_1, \dots, B_n$  are mutually independent.

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<sup>1</sup>Future editions will.



