

## JOINT PROBABILITY MATRICES

## Why

We can characterize the dependence of two events in terms of the rank of a particular matrix.

## **Definition**

Given a probability measure  $\mathbf{P}: \mathcal{P}(\Omega) \to \mathbf{R}$  on the finite set  $\Omega$  and two events  $A, B \subset \Omega$ , the *joint probability matrix* of A and B is the matrix

$$M = \begin{bmatrix} \mathbf{P}(A \cap B) & \mathbf{P}(A \cap C_{\Omega}(B)) \\ \mathbf{P}(C_{\Omega}(A) \cap B) & \mathbf{P}(C_{\Omega}(A) \cap C_{\Omega}(B)) \end{bmatrix}.$$

## Characterization of independence

If A and B are independent, then so are A and  $C_{\Omega}(B)$ , B and  $C_{\Omega}(A)$ , and  $C_{\Omega}(A)$  and  $C_{\Omega}(B)$ . In other words,

$$M = \begin{bmatrix} \mathbf{P}(A) \\ \mathbf{P}(C_{\Omega}(A)) \end{bmatrix} \begin{bmatrix} \mathbf{P}(B) & \mathbf{P}(C_{\Omega}(B) \end{bmatrix}.$$

In this case, we see that rank(M) = 1.

Conversely, suppose rank(M) = 1. Then, using the law of total probability, each row is a multiple of

$$M1 = \begin{bmatrix} \mathbf{P}(A) \\ \mathbf{P}(C_{\Omega}(A)) \end{bmatrix}.$$

In particular, we have  $\mathbf{P}(A \cap B) = \alpha \mathbf{P}(A)$  and  $\mathbf{P}(C_{\Omega}(A) \cap B) = \alpha P(C_{\Omega}(A))$ . So

$$\mathbf{P}(A \cap B) + \mathbf{P}(C_{\Omega}(A) \cap B) = \alpha(\mathbf{P}(B) + \mathbf{P}(C_{\Omega}(A))),$$

from which we deduce  $\alpha = \mathbf{P}(B)$  Likewise, the multiplier for the second column of M is  $\mathbf{P}(C_{\Omega}(B))$ . In other words, A and B are independent. We conclude that A and B are independent if and only if  $\mathrm{rank}(M) = 1$ .

