## What Is a Stochastic a Matrix?

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A stochastic matrix is an  $n \times n$  matrix with nonnegative entries and unit row sums. If  $A \in \mathbb{R}^{n \times n}$  is stochastic then Ae = e, where  $e = [1, 1, \dots, 1]^T$  is the vector of ones. This means that e is an eigenvector of A corresponding to the eigenvalue 1.

The identity matrix is stochastic, as is any permutation matrix. Here are some other examples of stochastic matrices:

$$A_{n} = n^{-1}ee^{T}, \text{ in particular } A_{3} = \begin{bmatrix} \frac{1}{3} & \frac{1}{3} & \frac{1}{3} \\ \frac{1}{3} & \frac{1}{3} & \frac{1}{3} \\ \frac{1}{3} & \frac{1}{3} & \frac{1}{3} \end{bmatrix}, \qquad (1)$$

$$B_{n} = \frac{1}{n-1}(ee^{T} - I), \text{ in particular } B_{3} = \begin{bmatrix} 0 & \frac{1}{2} & \frac{1}{2} \\ \frac{1}{2} & 0 & \frac{1}{2} \\ \frac{1}{2} & \frac{1}{2} & 0 \end{bmatrix}, \qquad (2)$$

$$C_{n} = \begin{bmatrix} 1 & & & & \\ \frac{1}{2} & \frac{1}{2} & & & \\ \frac{1}{2} & \frac{1}{2} & & & \\ \vdots & \vdots & \ddots & & \\ \frac{1}{n} & \frac{1}{n} & \dots & \frac{1}{n} \end{bmatrix}. \qquad (3)$$

For any matrix A, the spectral radius  $\rho(A) = \max\{|\lambda| : \lambda \text{ is an eigenvalue of } A\}$  is bounded by  $\rho(A) \leq ||A||$  for any norm. For a stochastic matrix, taking the  $\infty$ -norm (the maximum row sum of absolute values) gives  $\rho(A) \leq 1$ , so since we know that 1 is an eigenvalue,  $\rho(A) = 1$ . It can be shown that 1 is a semisimple eigenvalue, that is, if there are k eigenvalues equal to 1 then there are k linearly independent eigenvectors corresponding to 1 (Meyer, 2000, p. 696).

A lower bound on the spectral radius can be obtained from Gershgorin's theorem. The *i*th Gershgorin disc is defined by  $|\lambda - a_{ii}| \leq \sum_{j \neq i} |a_{ij}| = 1 - a_{ii}$ , which implies  $|\lambda| \geq 2a_{ii} - 1$ . Every eigenvalue  $\lambda$  lies in the union of the *n* discs and so must satisfy

$$2\min_{i} a_{ii} - 1 \le |\lambda| \le 1.$$

The lower bound is positive if A is strictly diagonally dominant by rows.

If A and B are stochastic then AB is nonnegative and ABe = Ae = e, so AB is stochastic. In particular, any positive integer power of A is stochastic. Does  $A^k$  converge as  $k \to \infty$ ? The answer is that it does, and the limit is stochastic, as long as 1 is the only

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eigenvalue of modulus 1, and this will be the case if all the elements of A are positive (by Perron's theorem). The simplest example of non-convergence is the stochastic matrix

$$A = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix},$$

which has eigenvalues 1 and -1. Since  $A^2 = I$ , all even powers are equal to I and all odd powers are equal to A. For the matrix (1),  $A_n^k = A_n$  for all k, while for (2),  $B_n^k \to A_n$  as  $k \to \infty$ . For (3),  $C_n^k$  converges to the matrix with 1 in every entry of the first column and zeros everywhere else.

Stochastic matrices arise in Markov chains. A transition matrix for a finite homogeneous Markov chain is a matrix whose (i,j) element is the probability of moving from state i to state j over a time step. It has nonnegative entries and the rows sum to 1, so it is a stochastic matrix. In applications including finance and healthcare, a transition matrix may be estimated for a certain time period, say one year, but a transition matrix for a shorter period, say one month, may be needed. If A is a transition matrix for a time period p then a stochastic pth root of p0, which is a stochastic solution p0 the equation p1 a transition matrix for a time period a factor p2 smaller. Therefore p3 and p4 (weeks to days) are among the values of interest. Unfortunately, a stochastic p4th root may not exist. For example, the matrix

$$\begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 0 & 0 & 1 \end{bmatrix}$$

has no pth roots at all, let alone stochastic ones. Yet many stochastic matrices do have stochastic roots. The matrix (3) has a stochastic pth root for all p, as shown by Higham and Lin (2011), who give a detailed analysis of pth roots of stochastic matrices. For example, to four decimal places,

$$C_6^{1/7} = \begin{bmatrix} 1.000 \\ 0.094 & 0.906 \\ 0.043 & 0.102 & 0.855 \\ 0.027 & 0.050 & 0.103 & 0.820 \\ 0.019 & 0.032 & 0.052 & 0.103 & 0.795 \\ 0.014 & 0.023 & 0.034 & 0.053 & 0.102 & 0.774 \end{bmatrix}$$

A stochastic matrix is sometime called a row-stochastic matrix to distinguish it from a column-stochastic matrix, which is a nonnegative matrix for which  $e^T A = e^T$  (so that  $A^T$  is row-stochastic). A matrix that is both row-stochastic and column-stochastic is called doubly stochastic. A permutation matrix is an example of a doubly stochastic matrix. If U is a unitary matrix then the matrix with  $a_{ij} = |u_{ij}|^2$  is doubly stochastic. A magic square scaled by the magic sum is also doubly stochastic. For example,

```
4
          14
                 15
                        1
A =
   4.7059e-01
                 5.8824e-02
                               8.8235e-02
                                            3.8235e-01
   1.4706e-01
                               2.9412e-01
                                             2.3529e-01
                 3.2353e-01
   2.6471e-01
                               1.7647e-01
                                             3.5294e-01
                 2.0588e-01
                 4.1176e-01
                               4.4118e-01
   1.1765e-01
                                             2.9412e-02
>> [sum(A) sum(A')]
ans =
                                                  1
     1
           1
                  1
                        1
                               1
                                     1
                                            1
>> eig(A)'
ans =
   1.0000e+00
                 2.6307e-01 -2.6307e-01
                                            8.5146e-18
```

## References

- Nicholas J. Higham and Lijing Lin, On pth Roots of Stochastic Matrices, Linear Algebra Appl. 435, 448–463, 2011.
- Carl D. Meyer, Matrix Analysis and Applied Linear Algebra, Society for Industrial and Applied Mathematics, Philadelphia, PA, USA, 2000.

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