

**Important Ideas and Useful Facts:**

- (i) **Diagonal matrices:** A square matrix  $D$  is *diagonal* if all entries off the diagonal are zero. If  $D$  and  $E$  are diagonal then  $DE = ED$  is also diagonal, and its diagonal entries are simply the products of corresponding diagonal entries of  $D$  and  $E$ . Thus the diagonal elements of  $D^n$  are just the  $n$ th powers of the diagonal elements of  $D$ . A *scalar* matrix is a diagonal matrix in which all elements along the diagonal are equal. The scalar matrices commute with all square matrices of the same size.
- (ii) **Diagonalisation:** Let  $M$  be a square  $n \times n$  matrix with eigenvalues  $\lambda_1, \dots, \lambda_n$  and corresponding eigenvectors  $\mathbf{v}_1, \dots, \mathbf{v}_n$ . Then

$$MP = PD$$

where  $D$  is the diagonal matrix with eigenvalues down the diagonal and  $P$  the matrix with corresponding eigenvectors as columns. If  $P$  is invertible then

$$M = PDP^{-1} \quad \text{and} \quad D = P^{-1}MP.$$

In this case we say that  $M$  is *diagonalisable*, in which case powers of  $M$  can be found easily by the formula

$$M^k = PD^kP^{-1}.$$

If the eigenvalues are all different then  $P$  is invertible and  $M$  is diagonalisable.

- (iii) **Similar matrices:** Two matrices  $A$  and  $B$  are said to be *similar* or *conjugate* if there is an invertible matrix  $P$  such that  $B = P^{-1}AP$ . Similarity is an equivalence relation (that is, similarity is reflexive, symmetric and transitive). In particular, a matrix is diagonalisable if and only if it is similar to a diagonal matrix.
- (iv) **Stochastic matrices:** A square matrix  $M$  is *stochastic* if all the entries are nonnegative and the columns add to 1, and *regular* if, further, some positive power of  $M$  has all positive entries. A column matrix  $\mathbf{v}$  is a *probability vector* if all of its entries are nonnegative and add to 1, and, further, becomes a *steady state vector* for  $M$  if  $M\mathbf{v} = \mathbf{v}$ .
- (v) **Existence and uniqueness of a steady state vector:** If  $M$  is a regular stochastic matrix then there exists a unique steady state vector  $\mathbf{v}$  for  $M$ , in which case, for any probability vector  $\mathbf{x}$ ,

$$\lim_{k \rightarrow \infty} M^k \mathbf{x} = \mathbf{v}.$$

- (vi) **Perron's Theorem and existence of dominant eigenvalues:** If  $M$  is a square matrix all of whose entries are positive then  $M$  has a positive real eigenvalue  $\lambda$  such that  $|\mu| \leq \lambda$  for all eigenvalues  $\mu$  of  $M$ , and, furthermore, there exists an eigenvector corresponding to  $\lambda$ , all of whose entries are positive.

Questions labelled with an asterisk are suitable for students aiming for a distinction or higher.

### Tutorial Exercises:

- Working over  $\mathbb{R}$ , diagonalise  $M$  and find  $M^k$  for any positive integer  $k$  where  $M$  is each of the following:

$$\begin{array}{lll}
 \text{(a)} \quad \begin{bmatrix} 2 & 0 \\ 1 & 1 \end{bmatrix} & \text{(b)} \quad \begin{bmatrix} 1 & 1 \\ -2 & 4 \end{bmatrix} & \text{(c)} \quad \begin{bmatrix} 0 & -1 \\ -1 & 0 \end{bmatrix} \\
 \text{(d)} \quad \begin{bmatrix} 1 & 1 & 1 \\ 0 & 2 & -1 \\ 0 & 0 & 3 \end{bmatrix} & \text{(e)} \quad \begin{bmatrix} 0 & 1 & 0 \\ -2 & 3 & 0 \\ 0 & 0 & 2 \end{bmatrix} & \text{(f)}^* \quad \begin{bmatrix} -7 & -2 & 6 \\ -2 & 1 & 2 \\ -10 & -2 & 9 \end{bmatrix}
 \end{array}$$

- Verify that the matrix

$$M = \begin{bmatrix} 1/2 & 2/5 \\ 1/2 & 3/5 \end{bmatrix}$$

is regular stochastic and find its unique steady state vector. Find an expression for  $M^n$  and check the limiting behaviour as  $n \rightarrow \infty$ .

- Consider the real matrix  $A = \begin{bmatrix} 2 & 3 \\ 3 & 4 \end{bmatrix}$ . Put  $\mathbf{v}_0 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$  and  $\mathbf{v}_k = A\mathbf{v}_{k-1}$  for  $k \geq 1$ .

- Compute  $\mathbf{v}_1, \dots, \mathbf{v}_5$  exactly.
- Find the characteristic polynomial of  $A$ . Find its two roots exactly, and also write down their decimal expansions correct to five decimal places.
- Let  $\mathbf{v}$  be the unique scalar multiple of  $\mathbf{v}_5$  with 1 in the first position. Verify that, correct to five decimal places,  $\mathbf{v}$  is an eigenvector of  $A$  corresponding to its largest eigenvalue.

- Consider the matrix  $B = \begin{bmatrix} -4 & 3 \\ 3 & -2 \end{bmatrix}$ . Put  $\mathbf{w}_0 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$  and  $\mathbf{w}_k = B\mathbf{w}_{k-1}$  for  $k \geq 1$ .

- Compute  $\mathbf{w}_1, \dots, \mathbf{w}_5$  exactly.
- Let  $\mathbf{w}$  be the unique scalar multiple of  $\mathbf{w}_5$  with 1 in the first position. Verify that, correct to five decimal places,  $\mathbf{w}$  is an eigenvector of  $B$ .
- \* Check that  $B = A^{-1}$  from the previous exercise. Explain how the phenomenon in part (b) is connected to the phenomenon in part (c) of the previous exercise.

- \* The sequence of *Fibonacci numbers* is obtained by writing down 1 twice, and obtaining each successive number by adding the previous two numbers together:

$$1, 1, 2, 3, 5, 8, 13, 21, 34, 55, 89, 144, \dots$$

If we let  $x_n$  denote the  $n$ th Fibonacci number then

$$x_1 = x_2 = 1, \quad x_n = x_{n-1} + x_{n-2} \quad \text{for } n \geq 3,$$

so that, by a simple induction,

$$\begin{bmatrix} x_n \\ x_{n-1} \end{bmatrix} = \begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} x_{n-1} \\ x_{n-2} \end{bmatrix} = \begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix}^{n-2} \begin{bmatrix} 1 \\ 1 \end{bmatrix}.$$

Diagonalise  $M = \begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix}$  to find a general formula for the  $n$ th Fibonacci number.

### Further Exercises:

6. Consider the matrix  $M = \begin{bmatrix} 0 & 1/2 & 1/2 \\ 1/2 & 1/2 & 0 \\ 1/2 & 0 & 1/2 \end{bmatrix}$ .
- (a) Verify that  $M$  is stochastic and regular.
  - (b) Find the unique steady state vector for  $M$ .
  - (c)\* Find an expression for  $M^n$  and check the limiting behaviour observed in part (b) as  $n \rightarrow \infty$ .
7. Verify that if  $A$  and  $B$  are similar matrices such that  $B$  is a scalar matrix then  $A = B$ .
- 8.\* Prove that if  $A$  is a real square matrix with exactly one eigenvalue  $\lambda \in \mathbb{C}$  then in fact  $\lambda \in \mathbb{R}$  and, further,  $A$  is diagonalisable if and only if  $A = \lambda I$ .
- 9.\* Consider the matrices

$$A = \begin{bmatrix} 3 & -1 \\ 4 & -1 \end{bmatrix} \quad \text{and} \quad B = \begin{bmatrix} 34 & 99 & 0 \\ -11 & -32 & 0 \\ -4 & -12 & 1 \end{bmatrix}$$

Find the characteristic polynomials of  $A$  and  $B$ . Deduce, from the previous exercise, that neither  $A$  nor  $B$  is diagonalisable.

- 10.\* Let  $A$  be a square matrix over a field  $F$ . Prove that  $A\mathbf{v} = \mathbf{0}$  for some nonzero column vector  $\mathbf{v}$  if and only if  $A$  is not invertible.
11. Let  $M$  be an upper or lower triangular  $n \times n$  matrix with diagonal entries  $\lambda_1, \dots, \lambda_n$ . Verify that the characteristic polynomial of  $M$  is

$$\chi(\lambda) = \det(\lambda I - M) = (\lambda - \lambda_1) \dots (\lambda - \lambda_n).$$

Thus the eigenvalues of  $M$  are just its diagonal entries.

- 12.\* Let  $M$  be a square matrix over  $\mathbb{C}$ . Define the *trace* of  $M$ , denoted by  $\text{tr}(M)$  to be the sum of the diagonal elements of  $M$ .
- (a) Prove that if  $A$  and  $B$  are square matrices of the same size then  $\text{tr}(AB) = \text{tr}(BA)$ .
  - (b) Deduce that if  $A$  and  $B$  are similar matrices then  $\text{tr}(A) = \text{tr}(B)$ .
  - (c) It is a theorem that  $M$  is similar to a triangular matrix. Deduce from this and the previous exercise that  $\text{tr}(M)$  is the sum of the eigenvalues of  $M$  (with multiplicity).