Conten₂₂ eigenvalues and eigenvectors

- 1. Basic concepts
- 2. Applications of eigenvalues and eigenvectors
- 3. Repeated eigenvalues and symmetric matrices
- 4. Numerical determination of eigenvalues and eigenvectors

Learning outcomes

Needs doing

Time allocation

You are expected to spend approximately thirteen hours of independent study on the material presented in this workbook. However, depending upon your ability to concentrate and on your previous experience with certain mathematical topics this time may vary considerably.

Basic Concepts





From an applications viewpoint, eigenvalue problems are probably the most important problems that arise in connection with matrix analysis. In this Section, after necessary preliminaries, we discuss the basic concepts. We shall see that eigenvalues and eigenvectors are associated with square matrices of order $n \times n$. If n is small (2 or 3), determining eigenvalues is a fairly straightforward process (requiring the solution of a low order polynomial equation). Obtaining eigenvectors is a little strange initially and it will help if you read this preliminary Section first.

	① have a knowledge of determinants and
Before starting this Section you should	matricesa knowledge of linear first order differential equations
Learning Outcomes	✓ obtain eigenvalues and eigenvectors of 2×2 and 3×3 matrices
After completing this Section you should be able to	 ✓ state certain basic properties of eigenvalues and eigenvectors

1. Basic Concepts

Determinants

A square matrix possesses an associated determinant. Unlike a matrix, which is an array of numbers, a determinant has a single value

A two by two matrix

$$C = \left[\begin{array}{cc} c_{11} & c_{12} \\ c_{21} & c_{22} \end{array} \right]$$

has an associated determinant

det (C) =
$$\begin{vmatrix} c_{11} & c_{12} \\ c_{21} & c_{22} \end{vmatrix} = c_{11} c_{22} - c_{21} c_{12}$$

(Note the use of square or round brackets which indicates a matrix and of straight vertical lines to denote a determinant.)

A three by three matrix has an associated determinant

$$\det(C) = \begin{vmatrix} c_{11} & c_{12} & c_{13} \\ c_{21} & c_{22} & c_{23} \\ c_{31} & c_{32} & c_{33} \end{vmatrix}$$

Among other ways this determinant can be evaluated by an "expansion about the top row."

$$\det(C) = c_{11} \begin{vmatrix} c_{22} & c_{23} \\ c_{32} & c_{33} \end{vmatrix} - c_{12} \begin{vmatrix} c_{21} & c_{23} \\ c_{31} & c_{33} \end{vmatrix} + c_{13} \begin{vmatrix} c_{21} & c_{22} \\ c_{31} & c_{32} \end{vmatrix}$$

Note the minus sign in the second term.



Evaluate the determinants

$$\det(A) = \begin{vmatrix} 4 & 6 \\ 3 & 1 \end{vmatrix} \qquad \det(B) = \begin{vmatrix} 4 & 8 \\ 1 & 2 \end{vmatrix} \qquad \det(C) = \begin{vmatrix} 6 & 5 & 4 \\ 2 & -1 & 7 \\ -3 & 2 & 0 \end{vmatrix}$$

Your solution

$$det A = 4 \times 1 - 6 \times 3 = -14 \qquad det B = 4 \times 2 - 8 \times 1 = 0$$
$$det C = 6 \begin{vmatrix} -1 & 7 \\ 2 & 0 \end{vmatrix} - 5 \begin{vmatrix} 2 & 7 \\ -3 & 0 \end{vmatrix} + 4 \begin{vmatrix} 2 & -1 \\ -3 & 2 \end{vmatrix}$$
$$= 6 \times (-14) - 5(21) + 4(4-3) = -185$$

A matrix such as $B = \begin{bmatrix} 4 & 8 \\ 1 & 2 \end{bmatrix}$ in the previous exercise which has zero determinant is called a **singular** matrix. The other two matrices A and C are **non-singular**. The key factor to be aware of is as follows:



Any non-singular $n \times n$ matrix C, one for which $\det(C) \neq 0$, possesses an inverse C^{-1} i.e. $CC^{-1} = C^{-1}C = I$

where I denotes the $n \times n$ identity matrix. A singular matrix does **not** possess an inverse.

Systems of Linear Equations

We first recall some basic results in linear (matrix) algebra. Consider a system of n equations in n unknowns x_1, x_2, \ldots, x_n :

 $c_{11}x_{1} + c_{12}x_{2} + \dots + c_{1n}x_{n} = k_{1}$ $c_{21}x_{1} + c_{22}x_{2} + \dots + c_{2n}x_{n} = k_{2}$ $\vdots + \vdots + \dots + \vdots = \vdots$ $c_{n1}x_{1} + c_{n2}x_{2} + \dots + c_{nn}x_{n} = k_{n}$

We can write such a system in matrix form:

$$\begin{bmatrix} c_{11} & c_{12} & \dots & c_{1n} \\ c_{21} & c_{22} & \dots & c_{2n} \\ \vdots & \vdots & \dots & \vdots \\ c_{n1} & c_{n2} & \dots & c_{nn} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} = \begin{bmatrix} k_1 \\ k_2 \\ \vdots \\ k_n \end{bmatrix}, \quad \text{or equivalently} \quad CX = K.$$

We see that C is an $n \times n$ matrix (called the coefficient matrix), $X = \{x_1, x_2, \ldots, x_n\}^T$ is the $n \times 1$ column vector of unknowns and K is an $n \times 1$ column vector of given constants.

Basic Results in Linear Algebra

Consider the system of equations CX = K.

• When the inverse C^{-1} exists the equations have a unique solution.

This will occur if $det(C) \neq 0$; then the unique solution is $X = C^{-1}K$.

• If K = 0 the system of equations is called **homogeneous**. In this special case if C^{-1} exists the only solution is X = 0. This is called the **trivial solution** as this could have been deduced by inspection. The only possibility of obtaining non-trivial solutions for a homogeneous

system of equations is when C^{-1} does not exist i.e. when det(C) = 0. It is easy to show that the equations have an infinite number of solutions in this case.

Recall that for a homogeneous system CX = K with $K = \underline{O}$ the only possibility of obtaining a non-zero solution for X is for C to be a singular matrix $(\det(C) = 0)$. There is no unique solution in this case but an infinite number.

Examples

(a) Solve the non-homogeneous system of equations

or CX = K where

$$C = \begin{bmatrix} 1 & 1 \\ 2 & 1 \end{bmatrix} \qquad X = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \qquad K = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$$

Here $det(C) = -1 \neq 0$. The system of equations has the unique solution

$$X = \left[\begin{array}{c} x_1 \\ x_2 \end{array} \right] = \left[\begin{array}{c} 1 \\ 0 \end{array} \right].$$

(b) Solve the homogeneous system

$$\begin{array}{rcl} x_1 - x_2 &=& 0 \\ x_1 + x_2 &=& 0 \end{array}$$

Here $C = \begin{bmatrix} 1 & -1 \\ 1 & 1 \end{bmatrix}$ and $det(C) = 2 \neq 0$. The unique solution is the **trivial** solution

$$X = \left[\begin{array}{c} x_1 \\ x_2 \end{array} \right] = \left[\begin{array}{c} 0 \\ 0 \end{array} \right]$$

(c) Solve the homogeneous system

Here det $C = \begin{vmatrix} 1 & 1 \\ 2 & 2 \end{vmatrix} = 0$ The solutions are **any** numbers such that $x_1 = -x_2$ i.e. $X = \begin{bmatrix} \alpha \\ -\alpha \end{bmatrix} \text{ where } \alpha \text{ is arbitrary.}$

i.e. there are an infinite number of possible solutions in this case.

A Simple Eigenvalue Problem

We shall be interested in simultaneous equations of the form:

$$AX = \lambda X,$$

where A is an $n \times n$ matrix, X is an $n \times 1$ column vector and λ is a scalar (a constant) and, in the first instance, we examine some simple examples to gain experience of solving problems of this type.

Example Consider the following system with n = 2:

$$\begin{array}{rcl} 2x+3y &=& \lambda x\\ 3x+2y &=& \lambda y \end{array}$$

so that

$$A = \begin{bmatrix} 2 & 3 \\ 3 & 2 \end{bmatrix} \quad \text{and} \quad X = \begin{bmatrix} x \\ y \end{bmatrix}.$$

It appears that there are three unknowns x, y, λ . The obvious questions to ask are: can we find x, y? what is λ ?

Solution

To solve this problem we firstly re-arrange the equations (take all unknowns onto one side);

$$(2 - \lambda)x + 3y = 0$$

$$3x + (2 - \lambda)y = 0$$
(1)
(2)

Therefore, from equation (2):

$$x = -\frac{(2-\lambda)}{3}y.$$

Then when we substitute this into (1)

$$-\frac{(2-\lambda)^2}{3}y + 3y = 0$$

which simplifies to

$$[-(2-\lambda)^2 + 9]y = 0.$$

We conclude that either y = 0 or $9 = (2 - \lambda)^2$. There are thus two cases to consider:

Case 1

If y = 0 then x = 0 (from (3)) and we get the **trivial solution**. (We could have guessed this solution at the outset).

(3)

Solution (contd.) Case 2 $9 = (2 - \lambda)^2$ which gives, on taking square roots: $\pm 3 = 2 - \lambda$ giving $\lambda = 2 \pm 3$ so $\lambda = 5$ or $\lambda = -1$. Now, from equation (3), if $\lambda = 5$ then x = +y and if $\lambda = -1$ then x = -y.

We have now completed the analysis. We have found values for λ but we also see that we cannot obtain unique values for x and y: all we can find is the ratio between these quantities. This behaviour is typical, as we shall now see, of an eigenvalue problem.

2. General Eigenvalue Problems

Consider a given square matrix A. If X is a column vector and λ is a scalar (a number) then the relation.

$$AX = \lambda X \tag{4}$$

is called an **eigenvalue problem**. Our purpose is to carry out an analysis of this equation in a manner similar to the example above. However, we will attempt a more general approach which will apply to **all** problems of this kind.

Firstly, we can spot an obvious solution (for X) to these equations. The solution X = 0 is a possibility (for then both sides are zero). We will not be interested in these **trivial solutions** of the eigenvalue problem. Our main interest will be in the occurrence of **non-trivial solutions** for X. These may exist for special values of λ , called the **eigenvalues** of the matrix A. We proceed as in the previous example:

take all unknowns to one side:

$$(A - \lambda I)X = 0\tag{5}$$

where I is a unit matrix with the same dimensions as A. (Note that $AX - \lambda X = 0$ does **not** simplify to $(A - \lambda)X = 0$ as you **cannot** subtract a scalar λ from a matrix A). This equation (5) is a **homogeneous** system of equations. In the notation of the earlier discussion $C \equiv A - \lambda I$ and $K \equiv 0$. For such a system we know that non-trivial solutions will only exist if the determinant of the coefficient matrix is zero:

$$\det(A - \lambda I) = 0 \tag{6}$$

Equation (6) is called the **characteristic equation** of the eigenvalue problem. We see that the characteristic equation only involves the unknown λ . The characteristic equation is generally a polynomial in λ , with degree being the same as the order of A (so if A is 2×2 the characteristic equation is a quadratic, if A is a 3×3 it is a cubic equation and so on). For each value of λ

that is obtained the corresponding value of X is obtained by solving the original equations (4). These X's are called **eigenvectors**.

N.B. We shall see that eigenvectors are only unique up to a multiplicative factor: i.e. if X satisfies $AX = \lambda X$ then so does kX when k is a constant.

Example Find the eigenvalues and eigenvectors of the matrix

$$A = \left[\begin{array}{rr} 1 & 0 \\ 1 & 2 \end{array} \right]$$

Solution

The eigenvalues and eigenvectors are found by solving the eigenvalue probelm

 $AX = \lambda X \qquad X = \left[\begin{array}{c} x \\ y \end{array} \right]$

i.e.

 $(A - \lambda I)X = 0.$

Non-trivial solutions will exist if

$$\det (A - \lambda I) = 0$$

that is,

det
$$\left\{ \begin{bmatrix} 1 & 0 \\ 1 & 2 \end{bmatrix} - \lambda \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \right\} = 0$$

$$\therefore \qquad \begin{vmatrix} 1 - \lambda & 0 \\ 1 & 2 - \lambda \end{vmatrix} = 0,$$

expanding this determinant:

$$(1-\lambda)(2-\lambda) = 0.$$

Hence the solutions for λ are: $\lambda = 1$ and $\lambda = 2$.

So we have found two values of λ for this 2×2 matrix A. Since these are unequal they are said to be **distinct** eigenvalues.

To each value of λ there corresponds an eigenvector. We now proceed to find the eigenvectors.

Case 1

 $\lambda=1$ (smaller eigenvalue). Then our original eigenvalue problem becomes: AX=X.~ In full this is

Solution (contd.)

Simplifying

$$\begin{array}{rcl} x &=& x & (a) \\ x + y &=& 0 & (b) \end{array}$$

All we can deduce here is that x = -y

$$\therefore \qquad X = \left[\begin{array}{c} x \\ -x \end{array} \right] \text{ for any } x \neq 0$$

(We specify $x \neq 0$ as, otherwise, we would have the trivial solution)

So the eigenvectors corresponding to eigenvalue $\lambda = 1$ are all proportional to $\begin{vmatrix} 1 \\ -1 \end{vmatrix}$, e.g.

$$\begin{bmatrix} 2\\ -2 \end{bmatrix}, \begin{bmatrix} -1\\ 1 \end{bmatrix} \text{ etc.}$$

Sometimes we write the eigenvector in **normalised** form that is, with modulus or magnitude 1. Here, the normalised form of X is

$$\frac{1}{\sqrt{2}} \begin{bmatrix} 1\\ -1 \end{bmatrix} \qquad \text{which is unique.}$$

Case 2

Now we consider the larger eigenvalue $\lambda = 2$. Our original eigenvalue problem $AX = \lambda X$ becomes AX = 2X which gives the following equations:

$$\left[\begin{array}{rr}1 & 0\\1 & 2\end{array}\right]\left[\begin{array}{r}x\\y\end{array}\right] = 2\left[\begin{array}{r}x\\y\end{array}\right]$$

i.e.

These equations imply that x = 0 whilst the variable y may take any value whatsoever (except zero as this gives the trivial solution).

Thus the eigenvector corresponding to eigenvalue $\lambda = 2$ has the form $\begin{bmatrix} 0 \\ y \end{bmatrix}$, e.g. $\begin{bmatrix} 0 \\ 1 \end{bmatrix}$, $\begin{bmatrix} 0 \\ 2 \end{bmatrix}$ etc. The normalised eigenvector here is $\begin{bmatrix} 0 \\ 1 \end{bmatrix}$

Solution (contd.) In conclusion: the matrix $A = \begin{bmatrix} 1 & 0 \\ 1 & 2 \end{bmatrix}$ has two eigenvalues and two associated normalised eigenvectors: $\lambda_1 = 1, \quad \lambda_2 = 2$ $X_1 = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ -1 \end{bmatrix} \quad X_2 = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$

Example Find the eigenvalues and eigenvectors of the 3×3 matrix

$$A = \begin{bmatrix} 2 & -1 & 0 \\ -1 & 2 & -1 \\ 0 & -1 & 2 \end{bmatrix}$$

Solution

The eigenvalues and eigenvectors are found by solving the eigenvalue problem

$$AX = \lambda X \qquad X = \begin{bmatrix} x \\ y \\ z \end{bmatrix}$$

Proceeding as in the previous example:

$$(A - \lambda I)X = 0$$

and non-trivial solutions for X will exist if

$$\det (A - \lambda I) = 0$$

that is,

$$\det \left\{ \begin{bmatrix} 2 & -1 & 0 \\ -1 & 2 & -1 \\ 0 & -1 & 2 \end{bmatrix} - \lambda \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \right\} = 0$$

i.e.
$$\begin{vmatrix} 2-\lambda & -1 & 0 \\ -1 & 2-\lambda & -1 \\ 0 & -1 & 2-\lambda \end{vmatrix} = 0.$$

Expanding this determinant we find:
$$(2-\lambda) \begin{vmatrix} 2-\lambda & -1 \\ -1 & 2-\lambda \end{vmatrix} + \begin{vmatrix} -1 & -1 \\ 0 & 2-\lambda \end{vmatrix} = 0$$

that is,

Solution (contd.)

 $(2 - \lambda) \{(2 - \lambda)^2 - 1\} - (2 - \lambda) = 0$ Taking out the common factor $(2 - \lambda)$: $(2 - \lambda) \{4 - 4\lambda + \lambda^2 - 1 - 1\}$ which gives: $(2 - \lambda) [\lambda^2 - 4\lambda + 2] = 0.$ This is easily solved to give: $\lambda = 2 \text{ or } \lambda = \frac{4 \pm \sqrt{16 - 8}}{2} = 2 \pm \sqrt{2}.$ So (typically) we have found three possible values of λ for this 3×3 matrix A. To each value of λ there corresponds an eigenvector.

Case 1

 $\lambda = 2 - \sqrt{2}$ (lowest eigenvalue). Then $AX = (2 - \sqrt{2})X$ implies

$$2x - y = (2 - \sqrt{2})x$$

-x + 2y - z = (2 - \sqrt{2})y
-y + 2z = (2 - \sqrt{2})z

Simplifying

$$\sqrt{2}x - y = 0 \tag{a}$$

$$x + \sqrt{2y} - z = 0 \tag{b}$$

$$-y + \sqrt{2}z = 0 \qquad (c)$$

We conclude the following:

(c) $\Rightarrow y = \sqrt{2}z$ (a) $\Rightarrow y = \sqrt{2}x$ \therefore these two relations give x = zThen (b) $\Rightarrow -x + 2x - x = 0$

The last equation gives us no information; it simply states that 0 = 0.

 $\therefore \qquad X = \begin{bmatrix} x \\ \sqrt{2}x \\ x \end{bmatrix} \text{ for any } x \neq 0 \text{ (otherwise we would have the trivial solution). So the eigenvectors corresponding to eigenvalue } \lambda = 2 - \sqrt{2} \text{ are all proportional to } \begin{bmatrix} 1 \\ \sqrt{2} \\ 1 \end{bmatrix}.$ In normalised form we have an eigenvector $\frac{1}{2} \begin{bmatrix} 1 \\ \sqrt{2} \\ 1 \end{bmatrix}.$

Solution (contd.) Case 2 $\lambda = 2$. Here AX = 2X implies $\begin{bmatrix} 2 & -1 & 0 \\ -1 & 2 & -1 \\ 0 & -1 & 2 \end{bmatrix} \begin{bmatrix} x \\ y \\ z \end{bmatrix} = 2 \begin{bmatrix} x \\ y \\ z \end{bmatrix}$ i.e. 2x - y = 2x-x + 2y - z = 2y-y + 2z = 2zAfter simplifying the equations become: (a)-y = 0-x-z = 0(b)-y = 0(c)(a), (c) imply y = 0: (b) implies x = -zeigenvector has the form $\begin{bmatrix} x \\ 0 \\ -x \end{bmatrix}$ for any $x \neq 0$. That is, eigenvectors corresponding to $\lambda = 2$ are all proportional to $\begin{bmatrix} 1\\0\\-1 \end{bmatrix}$. In normalised form we have an eigenvector $\frac{1}{\sqrt{2}} \begin{bmatrix} 1\\ 0\\ -1 \end{bmatrix}$. Case 3 $\lambda = 2 + \sqrt{2}$ (largest eigenvalue). Proceeding along similar lines to cases 1,2 above we find that the eigenvectors corresponding to $\lambda = 2 + \sqrt{2}$ are each proportional to $\begin{vmatrix} 1 \\ -\sqrt{2} \\ 1 \end{vmatrix}$ with normalised eigenvector $\frac{1}{2} \begin{vmatrix} 1 \\ -\sqrt{2} \\ 1 \end{vmatrix}$.

Solution (contd.) In conclusion the matrix $A = \begin{bmatrix} 2 & -1 & 0 \\ -1 & 2 & -1 \\ 0 & -1 & 2 \end{bmatrix}$ has three distinct eigenvalues and three corresponding normalised eigenvectors: $\lambda_1 = 2 - \sqrt{2}, \qquad \lambda_2 = 2 \qquad \lambda_3 = 2 + \sqrt{2}$ $X_1 = \frac{1}{2} \begin{bmatrix} 1 \\ \sqrt{2} \\ 1 \end{bmatrix}, \qquad X_2 = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ 0 \\ -1 \end{bmatrix}, \qquad X_3 = \frac{1}{2} \begin{bmatrix} 1 \\ -\sqrt{2} \\ 1 \end{bmatrix}$

Exercises

1. Find the eigenvalues and eigenvectors of each of the following matrices A:

Answers (eigenvectors are written in normalised form)
1. (a) 3 and 2;
$$\begin{bmatrix} 2\sqrt{5} \\ 1/\sqrt{5} \end{bmatrix}$$
 and $\begin{bmatrix} 1/\sqrt{2} \\ 1/\sqrt{5} \end{bmatrix}$
(b) 3 and 5; $\frac{1}{\sqrt{5}} \begin{bmatrix} 2 \\ 0 \\ 1 \end{bmatrix}$; $\begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}$; $\frac{1}{\sqrt{5}} \begin{bmatrix} 1 \\ 0 \\ -2 \end{bmatrix}$
(c) 1, 4 and 6; $\frac{1}{\sqrt{5}} \begin{bmatrix} 2 \\ 0 \\ 1 \end{bmatrix}$; $\frac{1}{\sqrt{5}} \begin{bmatrix} 0 \\ 2 \\ 1 \end{bmatrix}$; $\frac{1}{\sqrt{5}} \begin{bmatrix} 1 \\ 0 \\ -2 \end{bmatrix}$
(d) 0, -1 and 2; $\frac{1}{\sqrt{26}} \begin{bmatrix} 1 \\ 5 \\ 0 \end{bmatrix}$; $\frac{1}{\sqrt{5}} \begin{bmatrix} 0 \\ 2 \\ 1 \end{bmatrix}$; $\frac{1}{\sqrt{5}} \begin{bmatrix} 1 \\ 0 \\ -2 \end{bmatrix}$
(e) $\begin{bmatrix} 1 \\ 8^{-1} \\ -2 \end{bmatrix}$
(f) $\begin{bmatrix} 1 \\ 8^{-1} \\ -2 \end{bmatrix}$
(g) $\begin{bmatrix} 1 \\ 8^{-1}$

3. Properties of Eigenvalues and Eigenvectors

There are a number of general properties of eigenvalues and eigenvectors which you should be familiar with. You will be able to use them as a check on some of your calculations.

Property 1

For any square matrix A:

sum of eigenvalues = sum of diagonal terms of A (called the **trace** of A)

Formally, for an $n \times n$ matrix A:

 $\sum_{i=1}^{n} \lambda_i = \operatorname{trace}(A)$

(Repeated eigenvalues must be counted according to their multiplicity. Thus if $\lambda_1 = 4, \lambda_2 = 4, \lambda_3 = 1$ then $\sum_{i=1}^{3} \lambda_i = 9$).

Property 2

For any square matrix A

product of eigenvalues = determinant of A

Formally:

 $\lambda_1 \lambda_2 \lambda_3 \cdots \lambda_n = \prod_{i=1}^n \lambda_i = \det(A)$

The symbol \prod simply denotes multiplication, as \sum denotes summation.

Example Verify properties 1 and 2 for the 3×3 matrix:

$$A = \begin{bmatrix} 2 & -1 & 0 \\ -1 & 2 & -1 \\ 0 & -1 & 2 \end{bmatrix}$$

whose eigenvalues were found earlier.

Solution

The three eigenvalues of this matrix are:

$$\lambda_1 = 2 - \sqrt{2}, \qquad \lambda_2 = 2, \qquad \lambda_3 = 2 + \sqrt{2}$$

Therefore

$$\lambda_1 + \lambda_2 + \lambda_3 = (2 - \sqrt{2}) + 2 + (2 + \sqrt{2}) = 6 = \operatorname{trace}(A)$$

whilst
$$\lambda_1 \lambda_2 \lambda_3 = (2 - \sqrt{2})(2)(2 + \sqrt{2}) = 4 = \det(A)$$

Property 3

Eigenvectors of a matrix A corresponding to distinct eigenvalues are **linearly independent** i.e. one eigenvector **cannot** be written as a linear sum of the other eigenvectors. The proof of this result is omitted but we illustrate this property with two examples.

We saw earlier that the matrix

$$A = \begin{bmatrix} 1 & 0\\ 1 & 2 \end{bmatrix}$$

has distinct eigenvalues $\lambda_1 = 1$ $\lambda_2 = 2$ with associated eigenvectors

$$X^{(1)} = \frac{1}{\sqrt{2}} \begin{bmatrix} 1\\ -1 \end{bmatrix} \qquad X^{(2)} = \begin{bmatrix} 0\\ 1 \end{bmatrix}$$

respectively.

Clearly $X^{(1)}$ is **not** a constant multiple of $X^{(2)}$ and these eigenvectors are said to be **linearly** independent.

We also saw that the 3×3 matrix

$$A = \begin{bmatrix} 2 & -1 & 0 \\ -1 & 2 & -1 \\ 0 & -1 & 2 \end{bmatrix}$$

had the following distinct eigenvalues $\lambda_1 = 2 - \sqrt{2}$, $\lambda_2 = 2$, $\lambda_3 = 2 + \sqrt{2}$ with corresponding eigenvectors of the form shown:

$$X^{(1)} = \begin{bmatrix} 1\\ \sqrt{2}\\ 1 \end{bmatrix}, \qquad X^{(2)} = \begin{bmatrix} 1\\ 0\\ -1 \end{bmatrix}, \qquad X^{(3)} = \begin{bmatrix} 1\\ -\sqrt{2}\\ 1 \end{bmatrix}$$

Clearly none of these eigenvectors is a constant multiple of any other. Nor is any one obtainable as a linear combination of the other two. The 3 eigenvectors are linearly independent.

Property 4: Diagonal Matrices

A 2×2 diagonal matrix D has the form

$$D = \left[\begin{array}{cc} a & 0 \\ 0 & d \end{array} \right]$$

The characteristic equation

$$|D - \lambda I| = 0$$
 is $\begin{vmatrix} a - \lambda & 0 \\ 0 & d - \lambda \end{vmatrix} = 0$

i.e. $(a - \lambda)(d - \lambda) = 0$

So the eigenvalues are simply the diagonal elements a and d.

Similarly a 3×3 diagonal matrix has the form

$$D = \left[\begin{array}{rrr} a & 0 & 0 \\ 0 & b & 0 \\ 0 & 0 & c \end{array} \right]$$

having characteristic equation

$$|D - \lambda I| = (a - \lambda)(b - \lambda)(c - \lambda) = 0$$

so again the diagonal elements **are** the eigenvalues.

We can see that a diagonal matrix is a particularly simple matrix to work with. In addition to the eigenvalues being obtainable immediately by inspection it is exceptionally easy to multiply diagonal matrices.



Obtain the products of D_1D_2 and D_2D_1 the diagonal matrices

	$\begin{bmatrix} a \end{bmatrix}$	0	0		e	0	0]
$D_1 =$	0	b	0	$D_2 =$	0	f	0
$D_1 =$	0	0	c	$D_2 =$	0	0	g

Your solution	
	which of course is also diagonal.
	$D^{\mathrm{I}}D^{\mathrm{S}} = D^{\mathrm{S}}D^{\mathrm{I}} = \begin{bmatrix} 0 & 0 & c\partial \\ 0 & p \downarrow & 0 \\ \alpha \in & 0 & 0 \end{bmatrix}$

Exercises

- 1. If $\lambda_1, \lambda_2, \ldots, \lambda_n$ are the eigenvalues of a matrix A, prove the following:
 - (a) A^T has eigenvalues $\lambda_1, \lambda_2, \ldots, \lambda_n$.
 - (b) If A is upper triangular, then eigenvalues are exactly the main diagonal entries.
 - (c) The inverse matrix A^{-1} has eigenvalues $\frac{1}{\lambda_1}, \ \frac{1}{\lambda_2}, \dots, \frac{1}{\lambda_n}$.
 - (d) The matrix A kI has eigenvalues $\lambda_1 k, \lambda_2 k, \dots, \lambda_n k$.
 - (e) (Harder) The matrix A^2 has eigenvalues $\lambda_1^2, \lambda_2^2, \ldots, \lambda_n^2$.
 - (f) (Harder) The matrix A^k (k is non-negative integer) has eigenvalues $\lambda_1^k, \lambda_2^k, \dots, \lambda_n^k$.

Verify the above results for any 2×2 matrix and any 3×3 matrix found in the previous list of exercises.

N.B. Some of these results are useful in the numerical calculation of eigenvalues which we shall consider later.

sıəmsuA

I. (a) Using the property that for any square matrix det(A) = det(A) is the property that for any square matrix det(A) is (a, b) = det(A).

$$0 = {}^{T}(I \wedge A)$$
təb nəht $0 = (I \wedge A)$ təb

This immediately tells us that det $(A^T - \lambda I) = 0$ which shows that λ is also an eigenvalue of A^T .

(b) Here simply write down a typical upper triangular matrix U which has terms on the leading diagonal $u_{11}, u_{22}, \ldots, u_{nn}$ and above this. Then construct $(U - \lambda I)$. Finally imagine how you would then obtain det $(U - \lambda I) = 0$. You should see that the determinant is obtained by multiplying together those terms on the leading diagonal. Here the characteristic equation is:

$$0 = (\gamma - {}^{uu}n) \cdots (\gamma - {}^{22}n)(\gamma - {}^{11}n)$$

This polynomial has the obvious roots $\lambda_1 = u_{11}$, $\lambda_2 = u_{22}$, ..., $\lambda_n = u_{nn}$.

(c) Here we begin with the usual eigenvalue problem $AX = \lambda X$. If A has an inverse A^{-1} we can multiply both sides by A^{-1} on the left to give

 $X^{I-h}A = X$ səviş dəidw $XA^{I-h} = (Xh)^{I-h}$

(d), (e), and (f) are proved in similar way to the proof outlined in (c).

or, dividing through by the scalar A we get

$$X\frac{\mathrm{I}}{\mathrm{A}} = X^{\mathrm{I}-\mathrm{A}}$$

which shows that if λ and X are respectively eigenvalue and eigenvector of A then λ^{-1} and X has the respectively eigenvalue and eigenvector of A^{-1} .

As an example consider $A = \begin{bmatrix} 2 & 3 \\ 3 & 2 \end{bmatrix}$. This matrix has eigenvalues $\lambda_1 = -1$, $\lambda_2 = 5$ with corresponding eigenvectors $X_1 = \begin{bmatrix} 1 \\ -1 \end{bmatrix}$ and $X_2 = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$. The reader should verify (by direct multiplication) that $A^{-1} = -\frac{1}{5} \begin{bmatrix} 2 & -3 \\ -3 & 2 \end{bmatrix}$ has eigenvalues -1 and $\frac{1}{5}$ with respective eigenvalues $X_1 = \begin{bmatrix} 1 \\ -1 \end{bmatrix}$. eigenvectors $X_1 = \begin{bmatrix} 1 \\ -1 \\ -1 \end{bmatrix}$ and $X_2 = \begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix}$.