

## Overview

Last week introduced the important Diagonalisation Theorem:

*An  $n \times n$  matrix  $A$  is diagonalisable if and only if there is a basis for  $\mathbb{R}^n$  consisting of eigenvectors of  $A$ .*

This week we'll continue our study of eigenvectors and eigenvalues, but instead of focusing just on the matrix, we'll consider the associated linear transformation.

From Lay, §5.4

### Question

*If we always treat a matrix as defining a linear transformation, what role does diagonalisation play?*

(The version of the lecture notes posted online has more examples than will be covered in class.)

## Introduction

We know that a matrix determines a linear transformation, but the converse is also true:

if  $T : \mathbb{R}^n \rightarrow \mathbb{R}^m$  is a linear transformation, then  $T$  can be obtained as a matrix transformation

$$(*) \quad T(\mathbf{x}) = A\mathbf{x} \quad \text{for all } \mathbf{x} \in \mathbb{R}^n$$

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for a unique matrix  $A$ .

To construct this matrix, define

$$A = [T(\mathbf{e}_1) \ T(\mathbf{e}_2) \ \cdots \ T(\mathbf{e}_n)],$$

the  $m \times n$  matrix whose columns are the images via  $T$  of the vectors of the standard basis for  $\mathbb{R}^n$  (notice that  $T(\mathbf{e}_i)$  is a vector in  $\mathbb{R}^m$  for every  $i = 1, \dots, n$ ).

The matrix  $A$  is called the *standard matrix* of  $T$ .

## Example 1

Let  $T : \mathbb{R}^2 \rightarrow \mathbb{R}^3$  be the linear transformation defined by the formula

$$T \left( \begin{bmatrix} x \\ y \end{bmatrix} \right) = \begin{bmatrix} x - y \\ 3x + y \\ x - y \end{bmatrix}.$$

Find the standard matrix of  $T$ .

The standard matrix of  $T$  is the matrix  $[[T(\mathbf{e}_1)]_{\mathcal{E}} \ [T(\mathbf{e}_2)]_{\mathcal{E}}]$ .

Since

$$T(\mathbf{e}_1) = T \left( \begin{bmatrix} 1 \\ 0 \end{bmatrix} \right) = \begin{bmatrix} 1 \\ 3 \\ 1 \end{bmatrix}, \quad T(\mathbf{e}_2) = T \left( \begin{bmatrix} 0 \\ 1 \end{bmatrix} \right) = \begin{bmatrix} -1 \\ 1 \\ -1 \end{bmatrix},$$

the standard matrix of  $T$  is the  $3 \times 2$  matrix

$$\begin{bmatrix} 1 & -1 \\ 3 & 1 \\ 1 & -1 \end{bmatrix}.$$

## Example 2

Let  $A = \begin{bmatrix} 2 & 0 & 1 \\ 0 & -1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$ . What does the linear transformation  $T(\mathbf{x}) = A\mathbf{x}$  do to each of the standard basis vectors?

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- The image of  $\mathbf{e}_1$  is the vector  $\begin{bmatrix} 2 \\ 0 \\ 0 \end{bmatrix} = T(\mathbf{e}_1)$ . Thus, we see that  $T$  multiplies any vector parallel to the  $x$ -axis by the scalar 2.

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- The image of  $\mathbf{e}_3$  is the vector  $\begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix} = T(\mathbf{e}_3)$ . Thus, we see that  $T$  sends a vector parallel to the  $z$ -axis to a vector with equal  $x$  and  $z$  coordinates.



When we introduced the notion of coordinates, we noted that choosing different bases for our vector space gave us different coordinates. For example, suppose

$$\mathcal{E} = \{\mathbf{e}_1, \mathbf{e}_2, \mathbf{e}_3\} \text{ and } \mathcal{B} = \{\mathbf{e}_1, \mathbf{e}_2, -\mathbf{e}_1 + \mathbf{e}_3\}.$$

Then

$$\mathbf{e}_3 = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}_{\mathcal{E}} = \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix}_{\mathcal{B}}.$$

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Instead, it's more precise to write

$$[T(\mathbf{x})]_{\mathcal{E}} = A[\mathbf{x}]_{\mathcal{E}} \quad \text{with} \quad A = [[T(\mathbf{e}_1)]_{\mathcal{E}} \ [T(\mathbf{e}_2)]_{\mathcal{E}} \ \cdots \ [T(\mathbf{e}_n)]_{\mathcal{E}}]$$

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Every linear transformation  $T$  from  $\mathbb{R}^n$  to  $\mathbb{R}^m$  can be described as multiplication by its standard matrix: the standard matrix of  $T$  describes the action of  $T$  in terms of the coordinate systems on  $\mathbb{R}^n$  and  $\mathbb{R}^m$  given by the standard bases of these spaces.

If we start with a vector expressed in  $\mathcal{E}$  coordinates, then it's convenient to represent the linear transformation  $T$  by  $[T(\mathbf{x})]_{\mathcal{E}} = A[\mathbf{x}]_{\mathcal{E}}$ .

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(Note that the domain and codomain can be described using different coordinates! This is obvious when  $A$  is an  $m \times n$  matrix for  $m \neq n$ , but it's also true for linear transformations from  $\mathbb{R}^n$  to itself.)

### Example 3

For  $A = \begin{bmatrix} 2 & 0 & 1 \\ 0 & -1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$ , we saw that  $[T(\mathbf{x})]_{\mathcal{E}} = A[\mathbf{x}]_{\mathcal{E}}$  acted as follows:

- $T$  multiplies any vector parallel to the  $x$ -axis by the scalar 2.
- $T$  multiplies any vector parallel to the  $y$ -axis by the scalar  $-1$ .
- $T$  sends a vector parallel to the  $z$ -axis to a vector with equal  $x$  and  $z$  coordinates.

Describe the matrix  $B$  such that  $[T(\mathbf{x})]_{\mathcal{B}} = A[\mathbf{x}]_{\mathcal{B}}$ , where  $\mathcal{B} = \{\mathbf{e}_1, \mathbf{e}_2, -\mathbf{e}_1 + \mathbf{e}_3\}$ .



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Just as the  $i^{\text{th}}$  column of  $A$  is  $[T(\mathbf{e}_i)]_{\mathcal{E}}$ , the  $i^{\text{th}}$  column of  $B$  will be  $[T(\mathbf{b}_i)]_{\mathcal{B}}$ .

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Since  $\mathbf{e}_1 = \mathbf{b}_1$ ,  $T(\mathbf{b}_1) = 2\mathbf{b}_1$ .

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Since  $\mathbf{e}_1 = \mathbf{b}_1$ ,  $T(\mathbf{b}_1) = 2\mathbf{b}_1$ . Similarly,  $T(\mathbf{b}_2) = -\mathbf{b}_2$ .

Thus we see that  $B = \begin{bmatrix} 2 & 0 & * \\ 0 & -1 & * \\ 0 & 0 & * \end{bmatrix}$ .

The third column is the interesting one. Again, recall

$$\mathcal{B} = \{\mathbf{e}_1, \mathbf{e}_2, -\mathbf{e}_1 + \mathbf{e}_3\}, \text{ and}$$

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$$T(\mathbf{b}_3) = T(-\mathbf{e}_1 + \mathbf{e}_3) = -T(\mathbf{e}_1) + T(\mathbf{e}_3) = -2\mathbf{e}_1 + (\mathbf{e}_1 + \mathbf{e}_3) = -\mathbf{e}_1 + \mathbf{e}_3 = \mathbf{b}_3.$$

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Thus we see that  $B = \begin{bmatrix} 2 & 0 & 0 \\ 0 & -1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$ .

Notice that in  $B$  coordinates, the matrix representing  $T$  is diagonal!

Every linear transformation  $T : V \rightarrow W$  between finite dimensional vector spaces can be represented by a matrix, but the matrix representation of a linear transformation depends on the choice of bases for  $V$  and  $W$  (thus it is not unique).

This allows us to reduce many linear algebra problems concerning abstract vector spaces to linear algebra problems concerning the familiar vector spaces  $\mathbb{R}^n$ . This is important even for linear transformations  $T : \mathbb{R}^n \rightarrow \mathbb{R}^m$  since certain choices of bases for  $\mathbb{R}^n$  and  $\mathbb{R}^m$  can make important properties of  $T$  more evident: to solve certain problems easily, it is important to choose the *right* coordinates.



## Matrices and linear transformations

Let  $T : V \rightarrow W$  be a linear transformation that maps from  $V$  to  $W$ , and suppose that we've fixed a basis  $\mathcal{B} = \{\mathbf{b}_1, \dots, \mathbf{b}_n\}$  for  $V$  and a basis  $\mathcal{C} = \{\mathbf{c}_1, \dots, \mathbf{c}_m\}$  for  $W$ .

For any vector  $\mathbf{x} \in V$ , the coordinate vector  $[\mathbf{x}]_{\mathcal{B}}$  is in  $\mathbb{R}^n$  and the coordinate vector of its image  $[T(\mathbf{x})]_{\mathcal{C}}$  is in  $\mathbb{R}^m$ .

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We want to associate a matrix  $M$  with  $T$  with the property that  $M[\mathbf{x}]_{\mathcal{B}} = [T(\mathbf{x})]_{\mathcal{C}}$ .

It can be helpful to organise this information with a diagram

$$\begin{array}{ccc} V \ni \mathbf{x} & \xrightarrow{T} & T(\mathbf{x}) \in W \\ \downarrow & & \downarrow \\ \mathbb{R}^n \ni [\mathbf{x}]_{\mathcal{B}} & \xrightarrow{\text{multiplication by } M} & [T(\mathbf{x})]_{\mathcal{C}} \in \mathbb{R}^m \end{array}$$

where the vertical arrows represent the coordinate mappings.

Here's an example to illustrate how we might find such a matrix  $M$ :  
Let  $\mathcal{B} = \{\mathbf{b}_1, \mathbf{b}_2\}$  and  $\mathcal{C} = \{\mathbf{c}_1, \mathbf{c}_2\}$  be bases for two vector spaces  $V$  and  $W$ , respectively.

Let  $T : V \rightarrow W$  be the linear transformation defined by

$$T(\mathbf{b}_1) = 2\mathbf{c}_1 - 3\mathbf{c}_2, \quad T(\mathbf{b}_2) = -4\mathbf{c}_1 + 5\mathbf{c}_2.$$

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Why does this define the entire linear transformation? For an arbitrary vector  $\mathbf{v} = x_1\mathbf{b}_1 + x_2\mathbf{b}_2$  in  $V$ , we define its image under  $T$  as

$$T(\mathbf{v}) = x_1 T(\mathbf{b}_1) + x_2 T(\mathbf{b}_2).$$

For example, if  $\mathbf{x}$  is the vector in  $V$  given by  $\mathbf{x} = 3\mathbf{b}_1 + 2\mathbf{b}_2$ , so that

$[\mathbf{x}]_{\mathcal{B}} = \begin{bmatrix} 3 \\ 2 \end{bmatrix}$ , we have

$$\begin{aligned} T(\mathbf{x}) &= T(3\mathbf{b}_1 + 2\mathbf{b}_2) \\ &= 3T(\mathbf{b}_1) + 2T(\mathbf{b}_2) \\ &= 3(2\mathbf{c}_1 - 3\mathbf{c}_2) + 2(-4\mathbf{c}_1 + 5\mathbf{c}_2) \\ &= -2\mathbf{c}_1 + \mathbf{c}_2. \end{aligned}$$

Equivalently, we have

$$\begin{aligned} [T(\mathbf{x})]_C &= [3T(\mathbf{b}_1) + 2T(\mathbf{b}_2)]_C \\ &= 3[T(\mathbf{b}_1)]_C + 2[T(\mathbf{b}_2)]_C \\ &= \begin{bmatrix} [T(\mathbf{b}_1)]_C & [T(\mathbf{b}_2)]_C \end{bmatrix} \begin{bmatrix} 3 \\ 2 \end{bmatrix} \\ &= \begin{bmatrix} [T(\mathbf{b}_1)]_C & [T(\mathbf{b}_2)]_C \end{bmatrix} [\mathbf{x}]_B \end{aligned}$$

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In this case, since  $T(\mathbf{b}_1) = 2\mathbf{c}_1 - 3\mathbf{c}_2$  and  $T(\mathbf{b}_2) = -4\mathbf{c}_1 + 5\mathbf{c}_2$  we have

$$[T(\mathbf{b}_1)]_C = \begin{bmatrix} 2 \\ -3 \end{bmatrix} \quad \text{and} \quad [T(\mathbf{b}_2)]_C = \begin{bmatrix} -4 \\ 5 \end{bmatrix}$$

and so

$$\begin{aligned}[T(\mathbf{x})]_C &= \begin{bmatrix} 2 & -4 \\ -3 & 5 \end{bmatrix} \begin{bmatrix} 3 \\ 2 \end{bmatrix} \\ &= \begin{bmatrix} -2 \\ 1 \end{bmatrix}.\end{aligned}$$



In the last page, we are not so much interested in the actual calculation but in the equation

$$[T(\mathbf{x})]_{\mathcal{C}} = \begin{bmatrix} [T(\mathbf{b}_1)]_{\mathcal{C}} & [T(\mathbf{b}_2)]_{\mathcal{C}} \end{bmatrix} [\mathbf{x}]_{\mathcal{B}}$$

This gives us the matrix  $M$ :

$$M = \begin{bmatrix} [T(\mathbf{b}_1)]_{\mathcal{C}} & [T(\mathbf{b}_2)]_{\mathcal{C}} \end{bmatrix}$$

whose columns consist of the coordinate vectors of  $T(\mathbf{b}_1)$  and  $T(\mathbf{b}_2)$  with respect to the basis  $\mathcal{C}$  in  $W$ .

In general, when  $T$  is a linear transformation that maps from  $V$  to  $W$  where  $\mathcal{B} = \{\mathbf{b}_1, \dots, \mathbf{b}_n\}$  is a basis for  $V$  and  $\mathcal{C} = \{\mathbf{c}_1, \dots, \mathbf{c}_m\}$  is a basis for  $W$  the matrix associated to  $T$  with respect to these bases is

$$M = \left[ [T(\mathbf{b}_1)]_{\mathcal{C}} \quad \cdots \quad [T(\mathbf{b}_n)]_{\mathcal{C}} \right].$$

We write  ${}_{\mathcal{C} \leftarrow \mathcal{B}} T$  for  $M$ , so that  ${}_{\mathcal{C} \leftarrow \mathcal{B}} T$  has the property

$$\begin{aligned} [T(\mathbf{x})]_{\mathcal{C}} &= \left[ [T(\mathbf{b}_1)]_{\mathcal{C}} \quad \cdots \quad [T(\mathbf{b}_n)]_{\mathcal{C}} \right] [\mathbf{x}]_{\mathcal{B}} \\ &= {}_{\mathcal{C} \leftarrow \mathcal{B}} T [\mathbf{x}]_{\mathcal{B}}. \end{aligned}$$

The matrix  ${}_{\mathcal{C} \leftarrow \mathcal{B}} T$  describes how the linear transformation  $T$  operates in terms of the coordinate systems on  $V$  and  $W$  associated to the basis  $\mathcal{B}$  and  $\mathcal{C}$  respectively.

**NB.**  ${}_{\mathcal{C} \leftarrow \mathcal{B}} T$  is the matrix for  $T$  relative to  $\mathcal{B}$  and  $\mathcal{C}$ . It depends on the choice of *both* the bases  $\mathcal{B}, \mathcal{C}$ . The order of  $\mathcal{B}, \mathcal{C}$  is important.

In the case that  $T : V \rightarrow V$  and  $\mathcal{B} = \mathcal{C}$ ,  ${}_{\mathcal{B} \leftarrow \mathcal{B}} T$  is written  $[T]_{\mathcal{B}}$  and is the *matrix for  $T$  relative to  $\mathcal{B}$* , or more shortly the  $\mathcal{B}$ -matrix of  $T$ .

So by taking bases in each space, and writing vectors with respect to these bases,  $T$  can be studied by studying the matrix associated to  $T$  with respect to these bases.

## Algorithm for finding the matrix ${}_{\mathcal{C} \leftarrow \mathcal{B}} T$

To find the matrix  ${}_{\mathcal{C} \leftarrow \mathcal{B}} T$  where  $T : V \rightarrow W$  relative to

a basis  $\mathcal{B} = \{\mathbf{b}_1, \dots, \mathbf{b}_n\}$  of  $V$

a basis  $\mathcal{C} = \{\mathbf{c}_1, \dots, \mathbf{c}_m\}$  of  $W$

- Find  $T(\mathbf{b}_1), T(\mathbf{b}_2), \dots, T(\mathbf{b}_n)$ .
- Find the coordinate vector  $[T(\mathbf{b}_1)]_{\mathcal{C}}$  of  $T(\mathbf{b}_1)$  with respect to the basis  $\mathcal{C}$ . This is a column vector in  $\mathbb{R}^m$ .
- Do this for each  $T(\mathbf{b}_i)$ .
- Make a matrix from these column vectors. This matrix is  ${}_{\mathcal{C} \leftarrow \mathcal{B}} T$ .

N.B. The coordinate vectors  $[T(\mathbf{b}_1)]_{\mathcal{C}}, [T(\mathbf{b}_2)]_{\mathcal{C}}, \dots, [T(\mathbf{b}_n)]_{\mathcal{C}}$  have to be written as columns (not rows!).

# Examples

## Example 4

Let  $\mathcal{B} = \{\mathbf{b}_1, \mathbf{b}_2, \mathbf{b}_3\}$  and  $\mathcal{D} = \{\mathbf{d}_1, \mathbf{d}_2\}$  be bases for vector spaces  $V$  and  $W$  respectively.  $T : V \rightarrow W$  is the linear transformation with the property that

$$T(\mathbf{b}_1) = 3\mathbf{d}_1 - 5\mathbf{d}_2,$$

$$T(\mathbf{b}_2) = -\mathbf{d}_1 + 6\mathbf{d}_2,$$

$$T(\mathbf{b}_3) = 4\mathbf{d}_2$$

We find the matrix  ${}_{\mathcal{D} \leftarrow \mathcal{B}} T$  of  $T$  relative to  $\mathcal{B}$  and  $\mathcal{D}$ .

We have

$$[T(\mathbf{b}_1)]_{\mathcal{D}} = \begin{bmatrix} 3 \\ -5 \end{bmatrix}, [T(\mathbf{b}_2)]_{\mathcal{D}} = \begin{bmatrix} -1 \\ 6 \end{bmatrix}$$

and

$$[T(\mathbf{b}_3)]_{\mathcal{D}} = \begin{bmatrix} 0 \\ 4 \end{bmatrix}$$

This gives

$$\begin{aligned} T_{\mathcal{D} \leftarrow \mathcal{B}} &= \begin{bmatrix} [T(\mathbf{b}_1)]_{\mathcal{D}} & [T(\mathbf{b}_2)]_{\mathcal{D}} & [T(\mathbf{b}_3)]_{\mathcal{D}} \end{bmatrix} \\ &= \begin{bmatrix} 3 & -1 & 0 \\ -5 & 6 & 4 \end{bmatrix}. \end{aligned}$$

### Example 5

Define  $T : \mathbb{P}_2 \rightarrow \mathbb{R}^2$  by

$$T(p(t)) = \begin{bmatrix} p(0) + p(1) \\ p(-1) \end{bmatrix}.$$

- (a) Show that  $T$  is a linear transformation.
- (b) Find the matrix  ${}_{\mathcal{E} \leftarrow \mathcal{B}} T$  of  $T$  relative to the standard bases  $\mathcal{B} = \{1, t, t^2\}$  and  $\mathcal{E} = \{\mathbf{e}_1, \mathbf{e}_2\}$  of  $\mathbb{P}_2$  and  $\mathbb{R}^2$ .

(a) This is an exercise for you.

combinations of the vectors in  $\mathcal{E}$ ).

$$T(1) = \begin{bmatrix} 1+1 \\ 1 \end{bmatrix} = \begin{bmatrix} 2 \\ 1 \end{bmatrix} = 2\mathbf{e}_1 + \mathbf{e}_2$$

$$T(t) = \begin{bmatrix} 0+1 \\ -1 \end{bmatrix} = \begin{bmatrix} 1 \\ -1 \end{bmatrix} = \mathbf{e}_1 - \mathbf{e}_2$$

$$T(t^2) = \begin{bmatrix} 0+1 \\ 1 \end{bmatrix} = \begin{bmatrix} 1 \\ 1 \end{bmatrix} = \mathbf{e}_1 + \mathbf{e}_2.$$

- STEP 2 We find the coordinate vectors of  $T(1)$ ,  $T(t)$ ,  $T(t^2)$  in the basis  $\mathcal{E}$ :

$$[T(1)]_{\mathcal{E}} = \begin{bmatrix} 2 \\ 1 \end{bmatrix}, \quad [T(t)]_{\mathcal{E}} = \begin{bmatrix} 1 \\ -1 \end{bmatrix}, \quad [T(t^2)]_{\mathcal{E}} = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

- STEP 3 We form the matrix whose columns are the coordinate vectors in step 2

$$T_{\mathcal{E} \leftarrow \mathcal{B}} = \begin{bmatrix} 2 & 1 & 1 \\ 1 & -1 & 1 \end{bmatrix}$$



## Example 6

Let  $V = \text{Span}\{\sin t, \cos t\}$ , and  $D : V \rightarrow V$  the linear transformation  $D : f \mapsto f'$ . Let  $\mathbf{b}_1 = \sin t$ ,  $\mathbf{b}_2 = \cos t$ ,  $\mathcal{B} = \{\mathbf{b}_1, \mathbf{b}_2\}$ , a basis for  $V$ . We find the matrix of  $T$  with respect to the basis  $\mathcal{B}$ .

- STEP 1 We have

$$D(\mathbf{b}_1) = \cos t = 0\mathbf{b}_1 + 1\mathbf{b}_2,$$

$$D(\mathbf{b}_2) = -\sin t = -1\mathbf{b}_1 + 0\mathbf{b}_2.$$

- STEP 2 From this we have

$$[D(\mathbf{b}_1)]_{\mathcal{B}} = \begin{bmatrix} 0 \\ 1 \end{bmatrix}, [D(\mathbf{b}_2)]_{\mathcal{B}} = \begin{bmatrix} -1 \\ 0 \end{bmatrix},$$

- STEP 3 So that

$$[D]_{\mathcal{B}} = \left[ [T(\mathbf{b}_1)]_{\mathcal{B}} \quad [T(\mathbf{b}_2)]_{\mathcal{B}} \right] = \begin{bmatrix} 0 & -1 \\ 1 & 0 \end{bmatrix}.$$

Let  $f(t) = 4 \sin t - 6 \cos t$ . We can use the matrix we have just found to get the derivative of  $f(t)$ . Now  $[f(t)]_{\mathcal{B}} = \begin{bmatrix} 4 \\ -6 \end{bmatrix}$ . Then

$$\begin{aligned} [D(f(t))]_{\mathcal{B}} &= [D]_{\mathcal{B}}[f(t)]_{\mathcal{B}} \\ &= \begin{bmatrix} 0 & -1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} 4 \\ -6 \end{bmatrix} \\ &= \begin{bmatrix} 6 \\ 4 \end{bmatrix}. \end{aligned}$$

This, of course gives

$$f'(t) = 6 \sin t + 4 \cos t$$

which is what we would expect.

## Example 7

Let  $M_{2 \times 2}$  be the vector space of  $2 \times 2$  matrixes and let  $\mathbb{P}_2$  be the vector space of polynomials of degree at most 2. Let  $T : M_{2 \times 2} \rightarrow \mathbb{P}_2$  be the linear transformation given by

$$T \left( \begin{bmatrix} a & b \\ c & d \end{bmatrix} \right) = a + b + c + (a - c)x + (a + d)x^2.$$

We find the matrix of  $T$  with respect to the basis

$$\mathcal{B} = \left\{ \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix} \right\} \text{ for } M_{2 \times 2} \text{ and the standard basis}$$
$$\mathcal{C} = \{1, x, x^2\} \text{ for } \mathbb{P}_2.$$

- STEP 1 We find the effect of  $T$  on each of the basis elements:

$$T \left( \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} \right) = 1 + x + x^2,$$

$$T \left( \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} \right) = 1,$$

$$T \left( \begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix} \right) = 1 - x,$$

$$T \left( \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix} \right) = x^2.$$

- STEP 2 The corresponding coordinate vectors are

$$\left[ T \left( \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} \right) \right]_C = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix},$$

$$\left[ T \left( \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} \right) \right]_C = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix},$$

$$\left[ T \left( \begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix} \right) \right]_C = \begin{bmatrix} 1 \\ -1 \\ 0 \end{bmatrix},$$

$$\left[ T \left( \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix} \right) \right]_C = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}.$$

- STEP 3 Hence the matrix for  $T$  relative to the bases  $\mathcal{B}$  and  $\mathcal{C}$  is

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

## Example 8

We consider the linear transformation

$$H : \mathbb{P}_2 \rightarrow M_{2 \times 2}$$

given by

$$H(a + bx + cx^2) = \begin{bmatrix} a + b & a - b \\ c & c - a \end{bmatrix}$$

We find the matrix of  $H$  with respect to the standard basis  $\mathcal{C} = \{1, x, x^2\}$  for  $\mathbb{P}_2$  and

$$\mathcal{B} = \left\{ \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix} \right\} \text{ for } M_{2 \times 2}.$$

- STEP 1 We find the effect of  $H$  on each of the basis elements:

$$H(1) = \begin{bmatrix} 1 & 1 \\ 0 & -1 \end{bmatrix}, \quad H(x) = \begin{bmatrix} 1 & -1 \\ 0 & 0 \end{bmatrix}, \quad H(x^2) = \begin{bmatrix} 0 & 0 \\ 1 & 1 \end{bmatrix}.$$

- STEP 2 The corresponding coordinate vectors are

$$[H(1)]_{\mathcal{B}} = \begin{bmatrix} 1 \\ 1 \\ 0 \\ -1 \end{bmatrix}, \quad [H(x)]_{\mathcal{B}} = \begin{bmatrix} 1 \\ -1 \\ 0 \\ 0 \end{bmatrix}, \quad [H(x^2)]_{\mathcal{B}} = \begin{bmatrix} 0 \\ 0 \\ 1 \\ 1 \end{bmatrix}.$$



- STEP 3 Hence the matrix for  $H$  relative to the bases  $\mathcal{C}$  and  $\mathcal{B}$  is

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

## Linear transformations from $V$ to $V$

The most common case is when  $T : V \rightarrow V$  and  $\mathcal{B} = \mathcal{C}$ . In this case  ${}_{\mathcal{B} \leftarrow \mathcal{B}} T$  is written  $[T]_{\mathcal{B}}$  and is the *matrix for  $T$  relative to  $\mathcal{B}$*  or simply the  *$\mathcal{B}$ -matrix of  $T$* .

The  $\mathcal{B}$ -matrix for  $T : V \rightarrow V$  satisfies

$$[T(\mathbf{x})]_{\mathcal{B}} = [T]_{\mathcal{B}}[\mathbf{x}]_{\mathcal{B}}, \quad \text{for all } \mathbf{x} \in V. \quad (1)$$

$$\begin{array}{ccc} \mathbf{x} & \xrightarrow{T} & T(\mathbf{x}) \\ \downarrow & & \downarrow \\ [\mathbf{x}]_{\mathcal{B}} & \xrightarrow{\text{multiplication by } [T]_{\mathcal{B}}} & [T(\mathbf{x})]_{\mathcal{B}} \end{array}$$

## Examples

### Example 9

Let  $T : \mathbb{P}_2 \rightarrow \mathbb{P}_2$  be the linear transformation defined by

$$T(p(x)) = p(2x - 1).$$

We find the matrix of  $T$  with respect to  $\mathcal{E} = \{1, x, x^2\}$

- STEP 1 It is clear that

$$\begin{aligned}T(1) &= 1, & T(x) &= 2x - 1, \\T(x^2) &= (2x - 1)^2 = 1 - 4x + 4x^2\end{aligned}$$

- STEP 2 So the coordinate vectors are

$$[T(1)]_{\mathcal{E}} = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, [T(x)]_{\mathcal{E}} = \begin{bmatrix} -1 \\ 2 \\ 0 \end{bmatrix}, [T(x^2)]_{\mathcal{E}} = \begin{bmatrix} 1 \\ -4 \\ 4 \end{bmatrix}.$$

- STEP 3 Therefore

$$[T]_{\mathcal{E}} = \begin{bmatrix} 1 & -1 & 1 \\ 0 & 2 & -4 \\ 0 & 0 & 4 \end{bmatrix}$$

### Example 10

We compute  $T(3 + 2x - x^2)$  using part (a).

The coordinate vector of  $p(x) = (3 + 2x - x^2)$  with respect to  $\mathcal{E}$  is given by

$$[p(x)]_{\mathcal{E}} = \begin{bmatrix} 3 \\ 2 \\ -1 \end{bmatrix}.$$

We use the relationship

$$[T(p(x))]_{\mathcal{E}} = [T]_{\mathcal{E}}[p(x)]_{\mathcal{E}}.$$

This gives

$$\begin{aligned} [T(3 + 2x - x^2)]_{\mathcal{E}} &= [T(p(x))]_{\mathcal{E}} \\ &= [T]_{\mathcal{E}}[p(x)]_{\mathcal{E}} \\ &= \begin{bmatrix} 1 & -1 & 1 \\ 0 & 2 & -4 \\ 0 & 0 & 4 \end{bmatrix} \begin{bmatrix} 3 \\ 2 \\ -1 \end{bmatrix} \\ &= \begin{bmatrix} 0 \\ 8 \\ -4 \end{bmatrix} \end{aligned}$$

It follows that  $T(3 + 2x - x^2) = 8x - 4x^2$ .

## Example 11

Consider the linear transformation  $F : M_{2 \times 2} \rightarrow M_{2 \times 2}$  given by

$$F(A) = A + A^T$$

where  $A = \begin{bmatrix} a & b \\ c & d \end{bmatrix}$ .

We use the basis

$\mathcal{B} = \left\{ \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix} \right\}$  for  $M_{2 \times 2}$  to find a matrix representation for  $T$ .

More explicitly  $F$  is given by

$$F\left(\begin{bmatrix} a & b \\ c & d \end{bmatrix}\right) = \begin{bmatrix} a & b \\ c & d \end{bmatrix} + \begin{bmatrix} a & c \\ b & d \end{bmatrix} = \begin{bmatrix} 2a & b+c \\ b+c & 2d \end{bmatrix}$$

- STEP 1 We find the effect of  $F$  on each of the basis elements:

$$F\left(\begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}\right) = \begin{bmatrix} 2 & 0 \\ 0 & 0 \end{bmatrix}, F\left(\begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}\right) = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix},$$

$$F\left(\begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix}\right) = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}, F\left(\begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix}\right) = \begin{bmatrix} 0 & 0 \\ 0 & 2 \end{bmatrix}.$$

- STEP 2 The corresponding coordinate vectors are

$$\left[ F \left( \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} \right) \right]_{\mathcal{B}} = \begin{bmatrix} 2 \\ 0 \\ 0 \\ 0 \end{bmatrix}, \quad \left[ F \left( \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} \right) \right]_{\mathcal{B}} = \begin{bmatrix} 0 \\ 1 \\ 1 \\ 0 \end{bmatrix},$$

$$\left[ F \left( \begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix} \right) \right]_{\mathcal{B}} = \begin{bmatrix} 0 \\ 1 \\ 1 \\ 0 \end{bmatrix}, \quad \left[ F \left( \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix} \right) \right]_{\mathcal{B}} = \begin{bmatrix} 0 \\ 0 \\ 0 \\ 2 \end{bmatrix}.$$



- STEP 3 Hence the matrix representing  $F$  is

$$\begin{bmatrix} 2 & 0 & 0 & 0 \\ 0 & 1 & 1 & 0 \\ 0 & 1 & 1 & 0 \\ 0 & 0 & 0 & 2 \end{bmatrix}.$$

## Example 12

Let  $V = \text{Span} \{e^{2x}, e^{2x} \cos x, e^{2x} \sin x\}$ .

We find the matrix of the differential operator  $D$  with respect to

$$\mathcal{B} = \{e^{2x}, e^{2x} \cos x, e^{2x} \sin x\}.$$

- STEP 1 We see that

$$D(e^{2x}) = 2e^{2x}$$

$$D(e^{2x} \cos x) = 2e^{2x} \cos x - e^{2x} \sin x$$

$$D(e^{2x} \sin x) = 2e^{2x} \sin x + e^{2x} \cos x$$

- STEP 2 So the coordinate vectors are

$$[D(e^{2x})]_{\mathcal{B}} = \begin{bmatrix} 2 \\ 0 \\ 0 \end{bmatrix}, [D(e^{2x} \cos x)]_{\mathcal{B}} = \begin{bmatrix} 0 \\ 2 \\ -1 \end{bmatrix},$$

$$\text{and } [D(e^{2x} \sin x)]_{\mathcal{B}} = \begin{bmatrix} 0 \\ 1 \\ 2 \end{bmatrix}.$$

- STEP 3 Hence

$$[D]_{\mathcal{B}} = \begin{bmatrix} 2 & 0 & 0 \\ 0 & 2 & 1 \\ 0 & -1 & 2 \end{bmatrix}.$$

### Example 13

We use this result to find the derivative of  
 $f(x) = 3e^{2x} - e^{2x} \cos x + 2e^{2x} \sin x$ .

The coordinate vector of  $f(x)$  is given by

$$[f]_{\mathcal{B}} = \begin{bmatrix} 3 \\ -1 \\ 2 \end{bmatrix}.$$

We do this calculation using

$$[D(f)]_{\mathcal{B}} = [D]_{\mathcal{B}}[f]_{\mathcal{B}}.$$

This gives

$$\begin{aligned} [D(f)]_{\mathcal{B}} &= [D]_{\mathcal{B}}[f]_{\mathcal{B}} \\ &= \begin{bmatrix} 2 & 0 & 0 \\ 0 & 2 & 1 \\ 0 & -1 & 2 \end{bmatrix} \begin{bmatrix} 3 \\ -1 \\ 2 \end{bmatrix} \\ &= \begin{bmatrix} 6 \\ 0 \\ 5 \end{bmatrix}. \end{aligned}$$

This indicates that

$$f'(x) = 6e^{2x} + 5e^{2x} \sin x.$$

You should check this result by differentiation.

### Example 14

We use the previous result to find  $\int(4e^{2x} - 3e^{2x} \sin x) dx$

We recall that with the basis  $\mathcal{B} = \{e^{2x}, e^{2x} \cos x, e^{2x} \sin x\}$  the matrix representation of the differential operator  $D$  is given by

$$[D]_{\mathcal{B}} = \begin{bmatrix} 2 & 0 & 0 \\ 0 & 2 & 1 \\ 0 & -1 & 2 \end{bmatrix}.$$

We also notice that  $[D]_{\mathcal{B}}$  is invertible with inverse:

$$[D]_{\mathcal{B}}^{-1} = \begin{bmatrix} 1/2 & 0 & 0 \\ 0 & 2/5 & -1/5 \\ 0 & 1/5 & 2/5 \end{bmatrix}.$$

The coordinate vector of  $4e^{2x} - 3e^{2x} \sin x$  with respect to the basis  $\mathcal{B}$  is given by  $\begin{bmatrix} 4 \\ 0 \\ -3 \end{bmatrix}$ . We use this together with the inverse of  $[D]_{\mathcal{B}}$  to find the antiderivative  $\int(4e^{2x} - 3e^{2x} \sin x) dx$ :

$$[D]_{\mathcal{B}}^{-1}[4e^{2x} - 3e^{2x}]_{\mathcal{B}} = \begin{bmatrix} 1/2 & 0 & 0 \\ 0 & 2/5 & -1/5 \\ 0 & 1/5 & 2/5 \end{bmatrix} \begin{bmatrix} 4 \\ 0 \\ -3 \end{bmatrix} = \begin{bmatrix} 2 \\ 3/5 \\ -6/5 \end{bmatrix}.$$

So the antiderivative of  $4e^{2x} - 3e^{2x} \sin x$  in the vector space  $V$  is  $2e^{2x} + \frac{3}{5}e^{2x} \cos x - \frac{6}{5}e^{2x} \sin x$ , and we can deduce that  $\int(4e^{2x} - 3e^{2x} \sin x) dx = 2e^{2x} + \frac{3}{5}e^{2x} \cos x - \frac{6}{5}e^{2x} \sin x + C$  where  $C$  denotes a constant.

## Linear transformations and diagonalisation

In an applied problem involving  $\mathbb{R}^n$ , a linear transformation  $T$  usually appears as a matrix transformation  $\mathbf{x} \mapsto A\mathbf{x}$ . If  $A$  is diagonalisable, then there is a basis  $\mathcal{B}$  for  $\mathbb{R}^n$  consisting of eigenvectors of  $A$ . In this case the  $\mathcal{B}$ -matrix for  $T$  is diagonal, and diagonalising  $A$  amounts to finding a diagonal matrix representation of  $\mathbf{x} \mapsto A\mathbf{x}$ .

### Theorem

*Suppose  $A = PDP^{-1}$ , where  $D$  is a diagonal  $n \times n$  matrix. If  $\mathcal{B}$  is the basis for  $\mathbb{R}^n$  formed by the columns of  $P$ , then  $D$  is the  $\mathcal{B}$ -matrix for the transformation  $\mathbf{x} \mapsto A\mathbf{x}$ .*



## Proof.

Denote the columns of  $P$  by  $\mathbf{b}_1, \mathbf{b}_2, \dots, \mathbf{b}_n$ , so that  $\mathcal{B} = \{\mathbf{b}_1, \mathbf{b}_2, \dots, \mathbf{b}_n\}$  and

$$P = \begin{bmatrix} \mathbf{b}_1 & \mathbf{b}_2 & \cdots & \mathbf{b}_n \end{bmatrix}.$$

In this case,  $P$  is the change of coordinates matrix  $P_{\mathcal{B}}$  where

$$P[\mathbf{x}]_{\mathcal{B}} = \mathbf{x} \quad \text{and} \quad [\mathbf{x}]_{\mathcal{B}} = P^{-1}\mathbf{x}.$$

If  $T$  is defined by  $T(\mathbf{x}) = A\mathbf{x}$  for  $\mathbf{x}$  in  $\mathbb{R}^n$ , then

$$\begin{aligned} [T]_{\mathcal{B}} &= \begin{bmatrix} [T(\mathbf{b}_1)]_{\mathcal{B}} & \cdots & [T(\mathbf{b}_n)]_{\mathcal{B}} \end{bmatrix} \\ &= \begin{bmatrix} [A\mathbf{b}_1]_{\mathcal{B}} & \cdots & [A\mathbf{b}_n]_{\mathcal{B}} \end{bmatrix} \\ &= \begin{bmatrix} P^{-1}A\mathbf{b}_1 & \cdots & P^{-1}A\mathbf{b}_n \end{bmatrix} \\ &= P^{-1}A \begin{bmatrix} \mathbf{b}_1 & \mathbf{b}_2 & \cdots & \mathbf{b}_n \end{bmatrix} \\ &= P^{-1}AP = D \end{aligned}$$



In the proof of the previous theorem the fact that  $D$  is diagonal is never used. In fact the following more general result holds:

If an  $n \times n$  matrix  $A$  is similar to a matrix  $C$  with  $A = PCP^{-1}$ , then  $C$  is the  $\mathcal{B}$ -matrix of the transformation  $\mathbf{x} \rightarrow A\mathbf{x}$  where  $\mathcal{B}$  is the basis of  $\mathbb{R}^n$  formed by the columns of  $P$ .

## Example

### Example 15

Consider the matrix  $A = \begin{bmatrix} 4 & -2 \\ -1 & 3 \end{bmatrix}$ .  $T$  is the linear transformation  $T : \mathbb{R}^2 \rightarrow \mathbb{R}^2$  defined by  $T(\mathbf{x}) = A\mathbf{x}$ . We find a basis  $\mathcal{B}$  for  $\mathbb{R}^2$  with the property that  $[T]_{\mathcal{B}}$  is diagonal.

The first step is to find the eigenvalues and corresponding eigenspaces for  $A$ :

$$\begin{aligned} \det(A - \lambda I) &= \det \begin{bmatrix} 4 - \lambda & -2 \\ -1 & 3 - \lambda \end{bmatrix} \\ &= (4 - \lambda)(3 - \lambda) - 2 \\ &= \lambda^2 - 7\lambda + 10 \\ &= (\lambda - 2)(\lambda - 5). \end{aligned}$$

The eigenvalues of  $A$  are  $\lambda = 2$  and  $\lambda = 5$ . We need to find a basis vector for each of these eigenspaces.

$$E_2 = \text{Nul} \begin{bmatrix} 2 & -2 \\ -1 & 1 \end{bmatrix}$$

$$= \text{Span} \left\{ \begin{bmatrix} 1 \\ 1 \end{bmatrix} \right\}$$

$$E_5 = \text{Nul} \begin{bmatrix} -1 & -2 \\ -1 & -2 \end{bmatrix}$$

$$= \text{Span} \left\{ \begin{bmatrix} -2 \\ 1 \end{bmatrix} \right\}$$

Put  $\mathcal{B} = \left\{ \begin{bmatrix} 1 \\ 1 \end{bmatrix}, \begin{bmatrix} -2 \\ 1 \end{bmatrix} \right\}$ .

Then  $[T]_{\mathcal{B}} = D = \begin{bmatrix} 2 & 0 \\ 0 & 5 \end{bmatrix}$ , and with  $P = \begin{bmatrix} 1 & -2 \\ 1 & 1 \end{bmatrix}$  and  $P^{-1}AP = D$ , or equivalently,  $A = PDP^{-1}$ .