

# Eigenvalues and Eigenvectors

week 11-12 Fall 2006

## 1 Eigenvalues and eigenvectors

The most simple linear transformation from  $\mathbb{R}^n$  to  $\mathbb{R}^n$  may be the transformation of the form:

$$T(x_1, x_2, \dots, x_n) = (\lambda_1 x_1, \lambda_2 x_2, \dots, \lambda_n x_n),$$
$$T \left( \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} \right) = \begin{bmatrix} \lambda_1 & 0 & \cdots & 0 \\ 0 & \lambda_2 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & \lambda_n \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix}$$

**Example 1.1.** The linear transformation  $T : \mathbb{R}^2 \rightarrow \mathbb{R}^2$ , defined by

$$T(\mathbf{x}) = \begin{bmatrix} 2 & 0 \\ 0 & 3 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 2x_1 \\ 3x_2 \end{bmatrix},$$

is to dilate the first coordinate two times and the second coordinate three times.

**Example 1.2.** Let  $T : \mathbb{R}^2 \rightarrow \mathbb{R}^2$  be a linear transformation defined by

$$T(\mathbf{x}) = \begin{bmatrix} 4 & -2 \\ 3 & -3 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}.$$

What can we say about  $T$  geometrically? Consider the basis  $\mathcal{B} = \{\mathbf{u}_1, \mathbf{u}_2\}$  of  $\mathbb{R}^2$ , where

$$\mathbf{u}_1 = \begin{bmatrix} 2 \\ 1 \end{bmatrix}, \quad \mathbf{u}_2 = \begin{bmatrix} 1 \\ 3 \end{bmatrix}.$$

Then

$$T(\mathbf{u}_1) = \begin{bmatrix} 6 \\ 3 \end{bmatrix} = 3 \begin{bmatrix} 2 \\ 1 \end{bmatrix} = 3\mathbf{u}_1,$$
$$T(\mathbf{u}_2) = \begin{bmatrix} -2 \\ -6 \end{bmatrix} = -2 \begin{bmatrix} 1 \\ 3 \end{bmatrix} = -2\mathbf{u}_2.$$

For any vector  $\mathbf{v} = c_1\mathbf{u}_1 + c_2\mathbf{u}_2$ , we have  $[\mathbf{v}]_{\mathcal{B}} = \begin{bmatrix} c_1 \\ c_2 \end{bmatrix}$ , and

$$T(\mathbf{v}) = c_1T(\mathbf{u}_1) + c_2T(\mathbf{u}_2) = 3c_1\mathbf{u}_1 - 2c_2\mathbf{u}_2,$$

Thus

$$[T(\mathbf{v})]_{\mathcal{B}} = \begin{bmatrix} 3c_1 \\ -2c_2 \end{bmatrix} = \begin{bmatrix} 3 & 0 \\ 0 & -2 \end{bmatrix} \begin{bmatrix} c_1 \\ c_2 \end{bmatrix}.$$

If the one uses the basis  $\mathcal{B}$  to describe vector  $\mathbf{v}$  with coordinate vector  $[\mathbf{v}]_{\mathcal{B}}$ , then the coordinate vector of  $T(\mathbf{v})$  under the basis  $\mathcal{B}$  is simply described as

$$[T(\mathbf{v})]_{\mathcal{B}} = \begin{bmatrix} 3 & 0 \\ 0 & -2 \end{bmatrix} [\mathbf{v}]_{\mathcal{B}}.$$

This means that the matrix of  $T$  relative to the basis  $\mathcal{B}$  is as simple as a diagonal matrix.

The above discussion demonstrates that the nonzero vectors  $\mathbf{v}$  satisfying the condition

$$T(\mathbf{v}) = \lambda \mathbf{v} \tag{1.1}$$

for scalars  $\lambda$  is important to describe a linear transformation  $T$ .

**Definition 1.1.** Given a linear transformation

$$T : \mathbb{R}^n \rightarrow \mathbb{R}^n, \quad T(\mathbf{x}) = A\mathbf{x}.$$

A nonzero vector  $\mathbf{v}$  in  $\mathbb{R}^n$  is called an **eigenvector** of  $T$  (the matrix  $A$ ) if there exists a scalar  $\lambda$  such that

$$T(\mathbf{v}) = A\mathbf{v} = \lambda \mathbf{v}. \tag{1.2}$$

The scalar  $\lambda$  is called an **eigenvalue** of  $T$  (the matrix  $A$ ) and the nonzero vector  $\mathbf{v}$  is called an **eigenvector of  $T$  (of the matrix  $A$ ) corresponding to the eigenvalue  $\lambda$** .

**Example 1.3.** Let  $A = \begin{bmatrix} 1 & 6 \\ 5 & 2 \end{bmatrix}$ . Then  $\mathbf{u} = \begin{bmatrix} 6 \\ -5 \end{bmatrix}$  is an eigenvector of  $A$ . However,  $\mathbf{v} = \begin{bmatrix} 3 \\ -2 \end{bmatrix}$  is not an eigenvector of  $A$ .

**Proposition 1.2.** For any  $n \times n$  matrix  $A$ , the value 0 is an eigenvalue of  $A \iff \det A = 0$ .

*Proof.* Note that the set of eigenvectors of  $A$  corresponding to the zero eigenvalue is the set  $\text{Nul } A - \{\mathbf{0}\}$ ; and  $A$  is invertible if and only if  $\text{Nul } A \neq \{\mathbf{0}\}$ . The theorem follows from the two facts.  $\square$

**Theorem 1.3.** If  $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_p$  be eigenvectors of a matrix  $A$  corresponding to distinct eigenvalues  $\lambda_1, \lambda_2, \dots, \lambda_p$ , respectively, then  $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_p$  are linearly independent.

*Proof.* Let  $k$  be the smallest positive integer such that  $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_k$  are linearly independent. If  $k = p$ , nothing is to be proved. If  $k < p$ , then  $\mathbf{v}_{k+1}$  is a linear combination of  $\mathbf{v}_1, \dots, \mathbf{v}_k$ ; that is, there exist constants  $c_1, c_2, \dots, c_k$  such that

$$\mathbf{v}_{k+1} = c_1 \mathbf{v}_1 + c_2 \mathbf{v}_2 + \dots + c_k \mathbf{v}_k.$$

Applying the matrix  $A$  to both sides, we have

$$\begin{aligned} A\mathbf{v}_{k+1} &= \lambda_{k+1} \mathbf{v}_{k+1} \\ &= \lambda_{k+1} (c_1 \mathbf{v}_1 + c_2 \mathbf{v}_2 + \dots + c_k \mathbf{v}_k) \\ &= c_1 \lambda_{k+1} \mathbf{v}_1 + c_2 \lambda_{k+1} \mathbf{v}_2 + \dots + c_k \lambda_{k+1} \mathbf{v}_k; \end{aligned}$$

$$\begin{aligned} A\mathbf{v}_{k+1} &= A(c_1 \mathbf{v}_1 + c_2 \mathbf{v}_2 + \dots + c_k \mathbf{v}_k) \\ &= c_1 A\mathbf{v}_1 + c_2 A\mathbf{v}_2 + \dots + c_k A\mathbf{v}_k \\ &= c_1 \lambda_1 \mathbf{v}_1 + c_2 \lambda_2 \mathbf{v}_2 + \dots + c_k \lambda_k \mathbf{v}_k. \end{aligned}$$

Thus

$$c_1(\lambda_{k+1} - \lambda_1) \mathbf{v}_1 + c_2(\lambda_{k+1} - \lambda_2) \mathbf{v}_2 + \dots + c_k(\lambda_{k+1} - \lambda_k) \mathbf{v}_k = \mathbf{0}.$$

Since  $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_k$  are linearly independent, we have

$$c_1(\lambda_{k+1} - \lambda_1) = c_2(\lambda_{k+1} - \lambda_2) = \dots = c_k(\lambda_{k+1} - \lambda_k) = 0.$$

Note that the eigenvalues are distinct. Hence

$$c_1 = c_2 = \dots = c_k = 0,$$

which implies that  $\mathbf{v}_{k+1}$  is the zero vector  $\mathbf{0}$ . This is contradictory to that  $\mathbf{v}_{k+1} \neq \mathbf{0}$ .  $\square$

## 2 How to find eigenvectors?

To find eigenvectors, it is meant to find vectors  $\mathbf{x}$  and scalar  $\lambda$  such that

$$A\mathbf{x} = \lambda\mathbf{x}, \quad (2.1)$$

that is,

$$(\lambda I - A)\mathbf{x} = \mathbf{0}. \quad (2.2)$$

Since  $\mathbf{x}$  is required to be nonzero, the system (2.2) is required to have nonzero solutions; we thus have

$$\det(\lambda I - A) = 0. \quad (2.3)$$

Expanding the  $\det(\lambda I - A)$ , we see that

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

is a polynomial of degree  $n$  in  $\lambda$ , called the **characteristic polynomial** of  $A$ . To find eigenvalues of  $A$ , it is meant to find all roots of the polynomial  $p(\lambda)$ . The polynomial equation (2.3) about  $\lambda$  is called the **characteristic equation** of  $A$ . For an eigenvalue  $\lambda$  of  $A$ , the system

$$(\lambda I - A)\mathbf{x} = \mathbf{0}$$

is called the **eigensystem** for the eigenvalue  $\lambda$ ; its solution set  $\text{Nul}(\lambda I - A)$  is called the **eigenspace** corresponding to the eigenvalue  $\lambda$ .

**Theorem 2.1.** *The eigenvalues of a triangular matrix are the entries on its main diagonal.*

**Example 2.1.** The matrix

$$\begin{bmatrix} 2 & -1 & 0 \\ 0 & 5 & 0 \\ 0 & -1 & 2 \end{bmatrix}.$$

has the characteristic polynomial

$$p(\lambda) = \begin{vmatrix} \lambda - 2 & 1 & 0 \\ 0 & \lambda - 5 & 0 \\ 0 & 1 & \lambda - 2 \end{vmatrix} = (\lambda - 2)^2(\lambda - 5).$$

Then there are two eigenvalues  $\lambda_1 = 2$  and  $\lambda_2 = 5$ .

For  $\lambda_1 = 2$ , the eigensystem

$$\begin{bmatrix} \lambda_1 - 2 & 1 & 0 \\ 0 & \lambda_1 - 5 & 0 \\ 0 & 1 & \lambda_1 - 2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 0 & 1 & 0 \\ 0 & -3 & 0 \\ 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

has two linearly independent eigenvectors

$$\mathbf{v}_1 = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \quad \mathbf{v}_2 = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}.$$

For  $\lambda_2 = 5$ , the eigensystem

$$\begin{bmatrix} \lambda_2 - 2 & 1 & 0 \\ 0 & \lambda_2 - 5 & 0 \\ 0 & 1 & \lambda_2 - 2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 3 & 1 & 0 \\ 0 & 0 & 0 \\ 0 & 1 & 3 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

has one linearly independent eigenvector

$$\mathbf{v}_3 = \begin{bmatrix} 1 \\ -3 \\ 1 \end{bmatrix}.$$

**Example 2.2.** The matrix

$$\begin{bmatrix} 2 & -1 & 0 \\ 0 & 5 & 0 \\ -1 & -1 & 2 \end{bmatrix}$$

has the characteristic polynomial

$$p(\lambda) = \begin{vmatrix} \lambda - 2 & 1 & 0 \\ 0 & \lambda - 5 & 0 \\ 1 & 1 & \lambda - 2 \end{vmatrix} = (\lambda - 2)^2(\lambda - 5).$$

We obtain two eigenvalues  $\lambda_1 = 2$  and  $\lambda_2 = 5$ .

For  $\lambda_1 = 2$  (though it is of multiplicity 2), the eigensystem

$$\begin{bmatrix} 0 & 1 & 0 \\ 0 & -3 & 0 \\ 1 & 1 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

has only one linearly independent eigenvector

$$\mathbf{v}_1 = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}.$$

For  $\lambda_2 = 5$ , eigen-system

$$\begin{bmatrix} 3 & 1 & 0 \\ 0 & 0 & 0 \\ 1 & 1 & 3 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

has one linearly independent eigenvector

$$\mathbf{v}_2 = \begin{bmatrix} 3 \\ -9 \\ 2 \end{bmatrix}.$$

**Example 2.3.** Find the eigenvalues and the eigenvectors for the matrix

$$A = \begin{bmatrix} 1 & 3 & 3 \\ 3 & 1 & 3 \\ 3 & 3 & 1 \end{bmatrix}.$$

The characteristic equation of  $A$  is

$$\begin{aligned} \det(\lambda I - A) &= \begin{vmatrix} \lambda - 1 & -3 & -3 \\ -3 & \lambda - 1 & -3 \\ -3 & -3 & \lambda - 1 \end{vmatrix} \quad (R_2 - R_3) \\ &= \begin{vmatrix} \lambda - 1 & -3 & -3 \\ 0 & \lambda + 2 & -(\lambda + 2) \\ -3 & -3 & \lambda - 1 \end{vmatrix} \\ &= (\lambda - 1) \begin{vmatrix} \lambda + 2 & -(\lambda + 2) \\ -3 & \lambda - 1 \end{vmatrix} - 3 \begin{vmatrix} -3 & -3 \\ \lambda + 2 & -(\lambda + 2) \end{vmatrix} \\ &= (\lambda - 1)(\lambda + 2)(\lambda - 4) - 18(\lambda + 2) = (\lambda + 2)^2(\lambda - 7). \end{aligned}$$

Then  $A$  has two eigenvalues  $\lambda_1 = -2$  and  $\lambda_2 = 7$ .

For  $\lambda_1 = -2$  (its multiplicity is 2), the eigen-system

$$\begin{bmatrix} -3 & -3 & -3 \\ -3 & -3 & -3 \\ -3 & -3 & -3 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

has two linearly independent eigenvectors

$$\mathbf{v}_1 = \begin{bmatrix} -1 \\ 1 \\ 0 \end{bmatrix}, \quad \mathbf{v}_2 = \begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix}.$$

For  $\lambda = 7$ , the eigen-system

$$\begin{bmatrix} 6 & -3 & -3 \\ -3 & 6 & -3 \\ -3 & -3 & 6 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

has one linearly independent eigenvector

$$\mathbf{v}_3 = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}.$$

**Theorem 2.2.** Let  $\lambda$ ,  $\mu$  and  $\nu$  be distinct eigenvalues of a matrix  $A$ . Let  $\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_p$  be linearly independent eigenvectors for the eigenvalue  $\lambda$ ;  $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_q$  be linearly independent eigenvectors for the eigenvalue  $\mu$ ; and  $\mathbf{w}_1, \mathbf{w}_2, \dots, \mathbf{w}_r$  be linearly independent eigenvectors for the eigenvalue  $\nu$ . Then the vectors

$$\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_p, \mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_q, \mathbf{w}_1, \mathbf{w}_2, \dots, \mathbf{w}_r$$

together are linearly independent.

*Proof.* Suppose there are scalars  $a_1, \dots, a_p, b_1, \dots, b_q, c_1, \dots, c_r$  such that

$$(a_1\mathbf{u}_1 + \dots + a_p\mathbf{u}_p) + (b_1\mathbf{v}_1 + \dots + b_q\mathbf{v}_q) + (c_1\mathbf{w}_1 + \dots + c_r\mathbf{w}_r) = \mathbf{0}. \quad (2.4)$$

It suffices to show that all the scalars  $a_1, \dots, a_p, b_1, \dots, b_q, c_1, \dots, c_r$  are 0. Set

$$\mathbf{u} = a_1\mathbf{u}_1 + \dots + a_p\mathbf{u}_p, \quad \mathbf{v} = b_1\mathbf{v}_1 + \dots + b_q\mathbf{v}_q, \quad \mathbf{w} = c_1\mathbf{w}_1 + \dots + c_r\mathbf{w}_r.$$

Note that

$$A\mathbf{u} = a_1A\mathbf{u}_1 + \dots + a_pA\mathbf{u}_p = a_1\lambda\mathbf{u}_1 + \dots + a_p\lambda\mathbf{u}_p = \lambda\mathbf{u}.$$

Similarly,  $A\mathbf{v} = \mu\mathbf{v}$  and  $A\mathbf{w} = \nu\mathbf{w}$ . If  $\mathbf{u} = \mathbf{0}$ , then the linear independence of  $\mathbf{u}_1, \dots, \mathbf{u}_p$  implies that

$$a_1 = \dots = a_p = 0.$$

Similarly,  $\mathbf{v} = \mathbf{0}$  implies  $b_1 = \dots = b_q = 0$ , and  $\mathbf{w} = \mathbf{0}$  implies  $c_1 = \dots = c_r = 0$ .

Now we claim that  $\mathbf{u} = \mathbf{v} = \mathbf{w} = \mathbf{0}$ . If not, there are following three types.

**Type 1:**  $\mathbf{u} \neq \mathbf{0}, \mathbf{v} = \mathbf{w} = \mathbf{0}$ . Since  $\mathbf{v} = \mathbf{w} = \mathbf{0}$ , it follows from (2.4) that  $\mathbf{u} = \mathbf{0}$ , a contradiction.

**Type 2:**  $\mathbf{u} \neq \mathbf{0}, \mathbf{v} \neq \mathbf{0}, \mathbf{w} = \mathbf{0}$ . Then  $\mathbf{u}$  is the eigenvector of  $A$  for the eigenvalue  $\lambda$  and  $\mathbf{v}$  the eigenvector of  $A$  for the eigenvalue  $\mu$ ; they are eigenvectors for distinct eigenvalues. So  $\mathbf{u}$  and  $\mathbf{v}$  are linearly independent. But (2.4) shows that  $\mathbf{u} + \mathbf{v} = \mathbf{0}$ , which means that  $\mathbf{u}$  and  $\mathbf{v}$  are linearly dependent, a contradiction.

**Type 3:**  $\mathbf{u} \neq \mathbf{0}, \mathbf{v} \neq \mathbf{0}, \mathbf{w} \neq \mathbf{0}$ . This means that  $\mathbf{u}, \mathbf{v}, \mathbf{w}$  are eigenvectors of  $A$  for distinct eigenvalues  $\lambda, \mu, \rho$  respectively. So they are linearly independent. However, (2.4) shows that  $\mathbf{u} + \mathbf{v} + \mathbf{w} = \mathbf{0}$ , which means that  $\mathbf{u}, \mathbf{v}, \mathbf{w}$  are linearly dependent, a contradiction again.  $\square$

**Note 1.** The above theorem is also true for more than three distinct eigenvalues.

### 3 Diagonalization

**Definition 3.1.** An  $n \times n$  matrix  $A$  is said to be **similar** to an  $n \times n$  matrix  $B$  if there is an invertible matrix  $P$  such that

$$P^{-1}AP = B.$$

**Theorem 3.2.** *Similar matrices have the same characteristic polynomial and hence have the same eigenvalues.*

**Note.** Similar matrices may have different eigenvectors. For instance, the matrices

$$A = \begin{bmatrix} 2 & -1 & 0 \\ 0 & 5 & 0 \\ -1 & -1 & 2 \end{bmatrix} \quad \text{and} \quad B = \begin{bmatrix} 2 & -1 & 0 \\ 0 & 5 & 0 \\ 0 & -1 & 2 \end{bmatrix}$$

have the same eigenvalues  $\lambda_1 = 2$  and  $\lambda_2 = 5$ ; but  $A$  and  $B$  have different eigenvectors.

A square matrix  $A$  is called **diagonal** if all non-diagonal entries are zero, that is,

$$A = \begin{bmatrix} a_1 & 0 & \cdots & 0 \\ 0 & a_2 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & a_n \end{bmatrix}.$$

It is easy to see that for any  $k$ ,

$$A^k = \begin{bmatrix} a_1^k & 0 & \cdots & 0 \\ 0 & a_2^k & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & a_n^k \end{bmatrix}.$$

**Definition 3.3.** A square matrix  $A$  is said to be **diagonalizable** if  $A$  is similar to a diagonal matrix, that is, there exists an invertible matrix  $P$  and a diagonal matrix  $D$  such that

$$P^{-1}AP = D.$$

**Theorem 3.4 (Diagonalization Theorem).** *An  $n \times n$  matrix  $A$  is diagonalizable iff  $A$  has  $n$  linearly independent eigenvectors.*

*Proof.* We demonstrate the proof for the case  $n = 3$ .

If  $A$  is diagonalizable, there exist an invertible matrix  $P$  and a diagonal matrix  $D$  such that  $P^{-1}AP = D$ , where

$$P = [\mathbf{u}, \mathbf{v}, \mathbf{w}] = \begin{bmatrix} u_1 & v_1 & w_1 \\ u_2 & v_2 & w_2 \\ u_3 & v_3 & w_3 \end{bmatrix}, \quad D = \begin{bmatrix} \lambda & 0 & 0 \\ 0 & \mu & 0 \\ 0 & 0 & \nu \end{bmatrix}.$$

Note that  $P^{-1}AP = D$  is equivalent to  $AP = PD$ . Since  $AP = A[\mathbf{u}, \mathbf{v}, \mathbf{w}] = [A\mathbf{u}, A\mathbf{v}, A\mathbf{w}]$  and

$$PD = \begin{bmatrix} u_1 & v_1 & w_1 \\ u_2 & v_2 & w_2 \\ u_3 & v_3 & w_3 \end{bmatrix} \begin{bmatrix} \lambda & 0 & 0 \\ 0 & \mu & 0 \\ 0 & 0 & \nu \end{bmatrix} = \begin{bmatrix} \lambda u_1 & \mu v_1 & \nu w_1 \\ \lambda u_2 & \mu v_2 & \nu w_2 \\ \lambda u_3 & \mu v_3 & \nu w_3 \end{bmatrix} = [\lambda\mathbf{u}, \mu\mathbf{v}, \nu\mathbf{w}],$$

we have  $[A\mathbf{u}, A\mathbf{v}, A\mathbf{w}] = [\lambda\mathbf{u}, \mu\mathbf{v}, \nu\mathbf{w}]$ . Thus

$$A\mathbf{u} = \lambda\mathbf{u}, \quad A\mathbf{v} = \mu\mathbf{v}, \quad A\mathbf{w} = \nu\mathbf{w}.$$

Since  $P$  is invertible, the vectors  $\mathbf{u}, \mathbf{v}, \mathbf{w}$  are linearly independent. It follows that  $\mathbf{u}, \mathbf{v}, \mathbf{w}$  are three linearly independent eigenvectors of  $A$ .

Conversely, if  $A$  has three linear independent eigenvectors  $\mathbf{u}, \mathbf{v}, \mathbf{w}$  corresponding to the eigenvalues  $\lambda, \mu, \nu$  respectively. Then  $A\mathbf{u} = \lambda\mathbf{u}$ ,  $A\mathbf{v} = \mu\mathbf{v}$ ,  $A\mathbf{w} = \nu\mathbf{w}$ . Let

$$P = [\mathbf{u}, \mathbf{v}, \mathbf{w}] = \begin{bmatrix} u_1 & v_1 & w_1 \\ u_2 & v_2 & w_2 \\ u_3 & v_3 & w_3 \end{bmatrix}, \quad D = \begin{bmatrix} \lambda & 0 & 0 \\ 0 & \mu & 0 \\ 0 & 0 & \nu \end{bmatrix}.$$

Then

$$\begin{aligned}
 AP &= A[\mathbf{u}, \mathbf{v}, \mathbf{w}] = [A\mathbf{u}, A\mathbf{v}, A\mathbf{w}] = [\lambda\mathbf{u}, \mu\mathbf{v}, \nu\mathbf{w}] \\
 &= \begin{bmatrix} \lambda u_1 & \mu v_1 & \nu w_1 \\ \lambda u_2 & \mu v_2 & \nu w_2 \\ \lambda u_3 & \mu v_3 & \nu w_3 \end{bmatrix} \\
 &= \begin{bmatrix} u_1 & v_1 & w_1 \\ u_2 & v_2 & w_2 \\ u_3 & v_3 & w_3 \end{bmatrix} \begin{bmatrix} \lambda & 0 & 0 \\ 0 & \mu & 0 \\ 0 & 0 & \nu \end{bmatrix} = PD.
 \end{aligned}$$

Since  $\mathbf{u}, \mathbf{v}, \mathbf{w}$  are linearly independent, thus the matrix  $P$  is invertible. Therefore

$$P^{-1}AP = D.$$

This means that  $A$  is diagonalizable. □

**Example 3.1.** Diagonalize the matrix

$$A = \begin{bmatrix} 3 & 1 & 1 \\ 1 & 3 & 1 \\ -1 & -1 & 1 \end{bmatrix}$$

and compute  $A^8$ .

The characteristic polynomial of  $A$  is

$$\begin{aligned}
 |\lambda I - A| &= \begin{vmatrix} \lambda - 3 & -1 & -1 \\ -1 & \lambda - 3 & -1 \\ 1 & 1 & \lambda - 1 \end{vmatrix} \begin{array}{l} R_2 + R_3 \\ R_1 - (\lambda - 3)R_3 \end{array} \\
 &= \begin{vmatrix} 0 & -(\lambda - 2) & -(\lambda - 2)^2 \\ 0 & \lambda - 2 & \lambda - 2 \\ 1 & 1 & \lambda - 1 \end{vmatrix} = (\lambda - 2)^2(\lambda - 3).
 \end{aligned}$$

There are two eigenvalues  $\lambda_1 = 2$  and  $\lambda_2 = 3$ .

For  $\lambda_1 = 2$ , the eigensystem

$$\begin{bmatrix} -1 & -1 & -1 \\ -1 & -1 & -1 \\ 1 & 1 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

has two linearly independent eigenvectors

$$\mathbf{v}_1 = \begin{bmatrix} -1 \\ 1 \\ 0 \end{bmatrix}, \quad \mathbf{v}_2 = \begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix}.$$

For  $\lambda_2 = 3$ , the eigensystem

$$\begin{bmatrix} 0 & -1 & -1 \\ -1 & 0 & -1 \\ 1 & 1 & 2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

has one linearly independent eigenvector

$$\mathbf{v}_3 = \begin{bmatrix} -1 \\ -1 \\ 1 \end{bmatrix}.$$

Set

$$P = \begin{bmatrix} -1 & -1 & -1 \\ 1 & 0 & -1 \\ 0 & 1 & 1 \end{bmatrix}, \quad D = \begin{bmatrix} 2 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 3 \end{bmatrix}.$$

Then

$$P^{-1}AP = \begin{bmatrix} -1 & 0 & -1 \\ 1 & 1 & 2 \\ -1 & -1 & -1 \end{bmatrix} \begin{bmatrix} 3 & 1 & 1 \\ 1 & 3 & 1 \\ -1 & -1 & 1 \end{bmatrix} \begin{bmatrix} -1 & -1 & -1 \\ 1 & 0 & -1 \\ 0 & 1 & 1 \end{bmatrix} = D,$$

or equivalently,

$$PDP^{-1} = A.$$

Thus

$$\begin{aligned} A^8 &= \underbrace{(PDP^{-1})(PDP^{-1}) \cdots (PDP^{-1})}_8 = PD^8P^{-1} \\ &= \begin{bmatrix} -1 & -1 & -1 \\ 1 & 0 & -1 \\ 0 & 1 & 1 \end{bmatrix} \begin{bmatrix} 2^8 & 0 & 0 \\ 0 & 2^8 & 0 \\ 0 & 0 & 3^8 \end{bmatrix} \begin{bmatrix} -1 & 0 & -1 \\ 1 & 1 & 2 \\ -1 & -1 & -1 \end{bmatrix} \\ &= \begin{bmatrix} 3^8 & 3^8 - 2^8 & 3^8 - 2^8 \\ 3^8 - 2^8 & 3^8 & 3^8 - 2^8 \\ 2^8 - 3^8 & 2^8 - 3^8 & 2^9 - 3^8 \end{bmatrix}. \end{aligned}$$

**Example 3.2.** Compute the matrix  $A^8$ , where

$$A = \begin{bmatrix} 1 & 1 & 1 \\ 0 & 3 & 3 \\ -2 & 1 & 1 \end{bmatrix}.$$

The characteristic polynomial of  $A$  is

$$\begin{aligned} |\lambda I - A| &= \begin{vmatrix} \lambda - 1 & -1 & -1 \\ 0 & \lambda - 3 & -3 \\ 2 & -1 & \lambda - 1 \end{vmatrix} \\ &= (\lambda - 1) \begin{vmatrix} \lambda - 3 & -3 \\ -1 & \lambda - 1 \end{vmatrix} + 2 \begin{vmatrix} -1 & -1 \\ \lambda - 3 & -3 \end{vmatrix} \\ &= (\lambda - 1)(\lambda^2 - 4\lambda) + 2\lambda \\ &= \lambda(\lambda - 2)(\lambda - 3). \end{aligned}$$

We have eigenvalues  $\lambda_1 = 0$ ,  $\lambda_2 = 2$ , and  $\lambda_3 = 3$ .

For  $\lambda_1 = 0$ ,

$$\begin{bmatrix} -1 & -1 & -1 \\ 0 & -3 & -3 \\ 2 & -1 & -1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix}, \quad \mathbf{v}_1 = \begin{bmatrix} 0 \\ -1 \\ 1 \end{bmatrix}.$$

For  $\lambda_2 = 2$ ,

$$\begin{bmatrix} 1 & -1 & -1 \\ 0 & -1 & -3 \\ 2 & -1 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix}, \quad \mathbf{v}_2 = \begin{bmatrix} -2 \\ -3 \\ 1 \end{bmatrix}.$$

For  $\lambda_3 = 3$ ,

$$\begin{bmatrix} 2 & -1 & -1 \\ 0 & 0 & -3 \\ 2 & -1 & 2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix}, \quad \mathbf{v}_3 = \begin{bmatrix} 1/2 \\ 1 \\ 0 \end{bmatrix}.$$

Set

$$P = [\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3] = \begin{bmatrix} 0 & -2 & 1/2 \\ -1 & -3 & 1 \\ 1 & 1 & 0 \end{bmatrix}.$$



Then

$$P^{-1}AP = \begin{bmatrix} 0 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 3 \end{bmatrix} = D.$$

Thus

$$\begin{aligned} A^8 &= \underbrace{(PDP^{-1})(PDP^{-1})\cdots(PDP^{-1})}_8 = P \begin{bmatrix} 0 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 3 \end{bmatrix}^8 P^{-1} \\ &= \begin{bmatrix} 0 & -2 & 1/2 \\ -1 & -3 & 1 \\ 1 & 1 & 0 \end{bmatrix} \begin{bmatrix} 0 & 0 & 0 \\ 0 & 2^8 & 0 \\ 0 & 0 & 3^8 \end{bmatrix} \begin{bmatrix} 1 & -1/2 & 1/2 \\ -1 & 1/2 & 1/2 \\ -2 & 2 & 2 \end{bmatrix} \\ &= \begin{bmatrix} 2^8 & 0 & 0 \\ 2^8 & 2^7 & 2^7 \\ -2^8 & 2^7 & 2^7 \end{bmatrix}. \end{aligned}$$

## 4 Complex eigenvalues

**Theorem 4.1.** For a  $2 \times 2$  matrix, if one of the eigenvalues of  $A$  is not a real number, then the other eigenvalue must be conjugate to this complex eigenvalue. Let

$$\lambda = a - bi \quad \text{with } b \neq 0$$

be a complex eigenvalue of  $A$  and let  $\mathbf{x} = \mathbf{u} + i\mathbf{v}$  be a complex eigenvector of  $A$  for  $\lambda$ , that is,

$$A(\mathbf{u} + i\mathbf{v}) = (a - bi)(\mathbf{u} + i\mathbf{v}).$$

Let  $P = [\mathbf{u}, \mathbf{v}]$ . Then

$$P^{-1}AP = \begin{bmatrix} a & -b \\ b & a \end{bmatrix}.$$

*Proof.*

$$\begin{aligned} A\mathbf{x} &= A\mathbf{u} + iA\mathbf{v}, \\ A\mathbf{x} &= \lambda\mathbf{x} = (a - bi)(\mathbf{u} + i\mathbf{v}) = (a\mathbf{u} + b\mathbf{v}) + i(-b\mathbf{u} + a\mathbf{v}). \end{aligned}$$

It follows that

$$A\mathbf{u} = a\mathbf{u} + b\mathbf{v}, \quad A\mathbf{v} = -b\mathbf{u} + a\mathbf{v}.$$

Thus

$$AP = A[\mathbf{u}, \mathbf{v}] = [\mathbf{u}, \mathbf{v}] \begin{bmatrix} a & -b \\ b & a \end{bmatrix} = P \begin{bmatrix} a & -b \\ b & a \end{bmatrix}.$$

□

**Example 4.1.** Let  $A = \begin{bmatrix} 5 & -2 \\ 1 & 3 \end{bmatrix}$ . Find a matrix  $P$  such that  $P^{-1}AP$  is diagonal or antisymmetric.

*Solution.* Since  $\begin{bmatrix} \lambda - 5 & 2 \\ -1 & \lambda - 3 \end{bmatrix} = \lambda^2 - 8\lambda + 17$ , the eigenvalues of  $A$  are complex numbers

$$\lambda = \frac{-8 \pm \sqrt{64 - 4 \cdot 17}}{2} = 4 \pm i.$$

For  $\lambda = 4 - i$ , we have the eigensystem

$$\begin{bmatrix} -1 - i & 2 \\ -1 & 1 - i \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

Solving the system, we have the eigenvector

$$\begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 1 - \iota \\ 1 \end{bmatrix} = \begin{bmatrix} 1 \\ 1 \end{bmatrix} + \iota \begin{bmatrix} -1 \\ 0 \end{bmatrix}$$

Let  $P = \begin{bmatrix} 1 & -1 \\ 1 & 0 \end{bmatrix}$ . Then  $P^{-1} = \begin{bmatrix} 0 & 1 \\ -1 & 1 \end{bmatrix}$ .

$$\begin{bmatrix} 0 & 1 \\ -1 & 1 \end{bmatrix} \begin{bmatrix} 5 & -2 \\ 1 & 3 \end{bmatrix} \begin{bmatrix} 1 & -1 \\ 1 & 0 \end{bmatrix} = \begin{bmatrix} 4 & -1 \\ 1 & 4 \end{bmatrix}$$