

31 Eigenvalues and eigenvectors

By the end of this section, you should be able to answer the following questions:

- How do you find the eigenvalues and eigenvectors of a given square matrix?
- What are some simple properties of eigenvalues and eigenvectors?
- Prove that the eigenvectors corresponding to distinct eigenvalues are linearly independent.

A great deal of this section should be familiar to you. We start by recalling some results on vector spaces associated with matrices.

31.1 Column space, row space, rank, nullity

For any $m \times n$ real matrix A , the *null space* of A is the vector space

$$N(A) = \{x \in \mathbb{R}^n \mid Ax = 0\}.$$

The dimension of $N(A)$ is called the *nullity* of A , denoted $\text{nullity}(A)$.

The column space of an $m \times n$ matrix A is the space spanned by the column vectors of A ($\subseteq \mathbb{R}^m$). The dimension of the column space of A is called the *rank* of A , denoted $\text{rank}(A)$. This coincides with the number of non-zero rows in the r.e.f. of A .

The row space of an $m \times n$ matrix A is the space spanned by the row vectors of A ($\subseteq \mathbb{R}^n$). A basis is given by the non-zero rows in the r.e.f. of A . The dimension of the row space is also given by the rank of A .

Note that the row space of A^T = column space of A .

For $m \times n$ matrices,

$$\text{rank}(A) + \text{nullity}(A) = n.$$

Defn $\{v_1, v_2, \dots, v_n\}$ is linearly independent
 \iff If $a_1 v_1 + a_2 v_2 + \dots + a_n v_n = 0$
 then $a_1 = a_2 = \dots = a_n = 0$

Review MATH1051
 $V \subseteq \mathbb{R}^n$
 s.t. $0 \in V$
 $v+w \in V$
 for $v, w \in V$
 $k \cdot v \in V, k \in \mathbb{R}$

31.2 Non-singular matrices

For $n \times n$ square matrix A , we have several conditions for the existence of A^{-1} .

For $n \times n$ matrix A , the following are equivalent:

$$AA^{-1} = A^{-1}A = I$$

- * 1. A is non-singular. i.e. A^{-1} exists.
- * 2. $Ax = 0$ has only the trivial solution $x = 0$.
3. If U is a r.e.f. for A , then U has no row of all zeros.
4. $Ax = b$ has a solution for every n -dimensional column vector b .
- * 5. $\det(A) \neq 0$.
6. The columns of A are linearly independent.
7. The rows of A are linearly independent.
8. $\text{nullity}(A) = 0$.
9. $\text{rank}(A) = n$.

31.3 Eigenvalues and eigenvectors

Let A be a square matrix. Then an *eigenvector* of A is a vector $v \neq 0$ such that

$$Av = \lambda v,$$

for some scalar λ .

The scalar λ is called the corresponding *eigenvalue*.

If v is an eigenvector of A , then so is tv for any scalar $t \neq 0$.

Recall if λ is an eigenvalue of A , with corresponding eigenvector v , then $Av = \lambda v = \lambda Iv$, so $(A - \lambda I)v = 0$. Hence $x = v$ is a non-trivial solution to the homogeneous system of equations $(A - \lambda I)x = 0$, and conversely, if there is a non-trivial solution then λ is an eigenvalue of A . Thus:

λ is an eigenvalue of $A \iff Ax = \lambda x$
 if and only if $(A - \lambda I)x = 0$ has a non-trivial solution
 if and only if $A - \lambda I$ is singular
 if and only if $\det(A - \lambda I) = 0$. *

} compare with
1, 2, 5 above

For an $n \times n$ matrix A , $\det(A - \lambda I)$ is a polynomial of degree n in λ , called the *characteristic polynomial* of A .

The equation $\det(A - \lambda I) = 0$ is the characteristic equation of A .

Eigenvalues λ may be complex numbers, and the eigenvectors v may have complex components, even for real matrices A .

To find the eigenvalues and eigenvectors, do the following:

1. Find the roots of the characteristic polynomial, $\det(A - \lambda I) = 0$. These are the eigenvalues.
2. For each eigenvalue λ , find all v satisfying $(A - \lambda I)v = 0$. These are the eigenvectors. The vector space spanned by the eigenvectors corresponding to each eigenvalue is called the *eigenspace* associated to λ .

$$I = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}$$

31.3.1 Example

Find the eigenvalues and eigenvectors of $A = \begin{pmatrix} -3 & 1 & 0 \\ 1 & -2 & 1 \\ 0 & 1 & -3 \end{pmatrix}$.

$$\begin{aligned} \det(A - \lambda I) &= \det \begin{pmatrix} -3-\lambda & 1 & 0 \\ 1 & -2-\lambda & 1 \\ 0 & 1 & -3-\lambda \end{pmatrix} \\ &= -(3+\lambda) \{ (2+\lambda)(3+\lambda) - 1 \} - 1 \{ -(3+\lambda) \} \\ &= -(3+\lambda) \{ 6 + 5\lambda + \lambda^2 - 1, -1 \} \\ &= -(3+\lambda) \{ \lambda^2 + 5\lambda + 4 \} \\ &= -(3+\lambda)(4+\lambda)(1+\lambda) = 0 \\ &\Rightarrow \lambda = -3, -4, -1. \quad (A - \lambda I)v = 0 \end{aligned}$$

$$\begin{aligned} \underline{\lambda = -3}: (A + 3I)v &= 0 \\ \Rightarrow \begin{pmatrix} 0 & 1 & 0 \\ 1 & 1 & 1 \\ 0 & 1 & 0 \end{pmatrix} \begin{pmatrix} a \\ b \\ c \end{pmatrix} &= \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix} \\ \Rightarrow b=0 \text{ \& } a+c=0 &\Rightarrow c=-a \\ \Rightarrow \begin{pmatrix} a \\ 0 \\ -a \end{pmatrix} &\sim a \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix} \end{aligned}$$

$$(A+4I)\underline{v} = \underline{0}$$

$$\underline{\lambda = -4} : \begin{pmatrix} 1 & 1 & 0 \\ 1 & 2 & 1 \\ 0 & 1 & 1 \end{pmatrix} \begin{pmatrix} a \\ b \\ c \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}$$

$$\Rightarrow \begin{cases} a+b=0 \\ b+c=0 \end{cases} \Rightarrow a=-b=c$$

$$\Rightarrow \begin{pmatrix} c \\ -c \\ c \end{pmatrix} \text{ or just } c \begin{pmatrix} 1 \\ -1 \\ 1 \end{pmatrix}$$

check $\lambda = -1 \rightarrow c \begin{pmatrix} 1 \\ 2 \\ 1 \end{pmatrix}$

Note: Can never have $\underline{0}$ as an eigenvector.

For $n = 2, 3$, we can solve the characteristic equation to get eigenvalues. For $n \geq 4$ there are better numerical methods.

31.4 Simple properties

For a square matrix A :

1. A and A^T have the same eigenvalues.

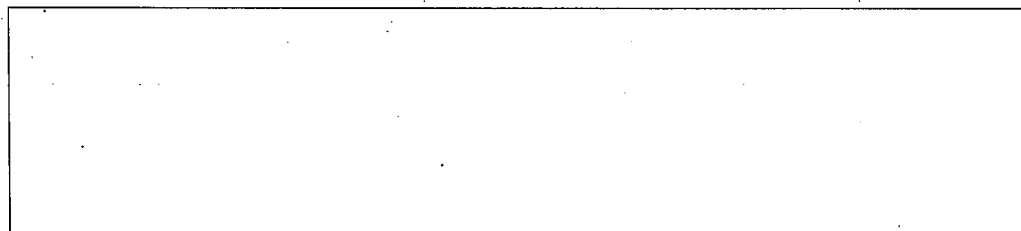
$$\det(A^T - \lambda I) = \det((A - \lambda I)^T) \\ \uparrow \\ \text{diagonal} = \det(A - \lambda I)$$

2. A is singular if and only if $\lambda = 0$ is an eigenvalue of A .

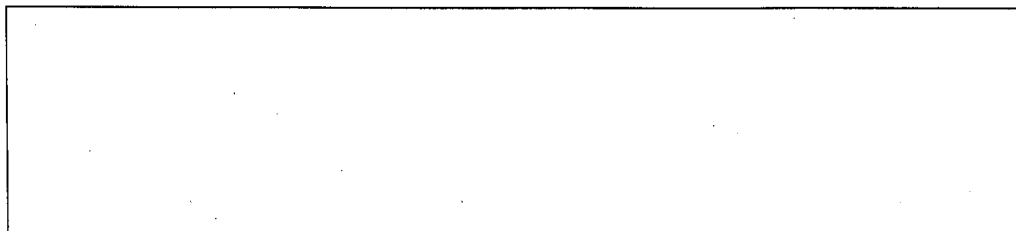
logic from p198

$$A \text{ singular} \Leftrightarrow A\underline{x} = \underline{0} \text{ has nontrivial sol.} \\ \Leftrightarrow \lambda = 0 \text{ is an eigenvalue.}$$

3. If λ is an eigenvalue of A , then λ^2 is an eigenvalue of A^2 , and $1/\lambda$ is an eigenvalue of A^{-1} when A is non-singular.



4. If λ is an eigenvalue of A , then $\lambda - m$ is an eigenvalue of $A - mI$, for any scalar m .



INDUCTION

Statement S_k $k=1, 2, 3, \dots$

Suppose the following hold:

- 1. S_1 is true
- 2. $\forall n \geq 1$, if S_n is true then S_{n+1} is true.

Then the statement

" $\forall k \geq 1, S_k$ " is true.

In proof ... ① Show S_1 is true

② Show that if S_n is true
then S_{n+1} must be true $\forall n \geq 1$

\Rightarrow result.

31.5 Eigenvectors corresponding to distinct eigenvalues are linearly independent

If $\lambda_1, \lambda_2, \dots, \lambda_k$ are distinct eigenvalues of A , with corresponding eigenvectors v_1, v_2, \dots, v_k (such that v_i corresponds to λ_i), then the set of eigenvectors $\{v_1, v_2, \dots, v_k\}$ is linearly independent.

use induction. • $k=1$, $\{v_1\}$ is l.i.
 • assume true for n such that
 $1 \leq n < k$.
 (i.e. assume $\{v_1, v_2, \dots, v_n\}$ is l.i.)
 Set $a_1 v_1 + a_2 v_2 + \dots + a_n v_n + a_{n+1} v_{n+1} = \underline{0}$ (*)
 (mult. (*) by A)
 $\Rightarrow a_1 \lambda_1 v_1 + a_2 \lambda_2 v_2 + \dots + a_n \lambda_n v_n + a_{n+1} \lambda_{n+1} v_{n+1} = \underline{0}$ ①
 (mult. (*) by $\lambda_{n+1} \neq 0$)
 $\Rightarrow a_1 \lambda_{n+1} v_1 + \dots + a_n \lambda_{n+1} v_n + a_{n+1} \lambda_{n+1} v_{n+1} = \underline{0}$ ②
 ② - ① $\Rightarrow a_1 (\lambda_{n+1} - \lambda_1) v_1 + \dots + a_n (\lambda_{n+1} - \lambda_n) v_n = \underline{0}$
 Since $\{v_1, \dots, v_n\}$ is l.i.
 $\Rightarrow a_1 (\lambda_{n+1} - \lambda_1) = \dots = a_n (\lambda_{n+1} - \lambda_n) = 0$
 $\Rightarrow a_1 = a_2 = \dots = a_n = 0$.
 Equ. (*) $\Rightarrow a_{n+1} v_{n+1} = \underline{0} \Rightarrow a_{n+1} = 0$.
 $\Rightarrow \{v_1, v_2, \dots, v_n, v_{n+1}\}$ is l.i.
 \Rightarrow result by induction.