

Eigenvalue

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eigen-vector cannot be 0
but λ can be equal to 0

1 Eigenvalue and eigenvector

An eigenvector of an $n \times n$ matrix A is a nonzero vector \mathbf{x} such that $A\mathbf{x} = \lambda\mathbf{x}$ for some scalar λ . A scalar λ is called an eigenvalue of A if there is a nontrivial solution \mathbf{x} of $A\mathbf{x} = \lambda\mathbf{x}$, such an \mathbf{x} is called an eigenvector corresponding to λ .

λ is an eigenvalue of an $n \times n$ matrix A if and only if the equation

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

has a nontrivial solution. The set of all solutions of (1) is just the null space of the matrix $A - \lambda I$, so this set is a subspace of \mathbb{R}^n . It is called the eigenspace of A corresponding to λ , which consists of the zero vector and all the eigenvectors corresponding to λ .

Remark 1. 0 is an eigenvalue of matrix A if and only if the equation

$$A\mathbf{x} = 0\mathbf{x}$$

has a nontrivial solution. Furthermore, the system has a nontrivial solution if and only if A is not invertible.

$\in \mathbb{R}^{n \times n}$
 $A\mathbf{x} = \lambda\mathbf{x}$
If $\lambda=0$ is an eigenvalue.
 \Rightarrow there exists $\mathbf{x} \neq \mathbf{0}$ e. vector
s.t. $A\mathbf{x} = \mathbf{0}$
(2)
 $\Leftrightarrow \dim(\text{null}(A)) > 0$
 $\Leftrightarrow \text{rank}(A) < n$
 $\Leftrightarrow A$ is non-singular
 $\Leftrightarrow \dots \dots$

2 Characteristic polynomial

The scalar equation

$$\det(A - \lambda I) = 0 \tag{3}$$

is called the characteristic equation of A . A scalar λ is an eigenvalue of an $n \times n$ matrix A if and only if λ satisfies the characteristic equation.

Remark 2. $\det(A - \lambda I)$ is a polynomial in λ . It can be shown that if A is an $n \times n$ matrix, then $\det(A - \lambda I)$ is a polynomial of degree n . This polynomial is called the characteristic polynomial of A . The (algebraic) multiplicity of an eigenvalue λ is its multiplicity as a root of the characteristic polynomial.

Remark 3. Matrix A always has n eigenvalues if count the multiplicity.

Theorem 2.1. If $\mathbf{v}_1, \dots, \mathbf{v}_r$ are eigenvectors that correspond to distinct eigenvalues $\lambda_1, \dots, \lambda_r$ of an $n \times n$ matrix A , then the set $\{\mathbf{v}_1, \dots, \mathbf{v}_r\}$ is linearly independent.

Proof. Let $\lambda_1 \neq \lambda_2$ be two eigenvalues and v_1 and v_2 be the corresponding eigenvectors. Assume v_1 and v_2 are linear dependent. That is, there exists $a \neq 0$ such that,

$$v_1 + av_2 = 0. \tag{4}$$

Multiply A on both sides of the equation,

$$Av_1 + aAv_2 = \lambda_1 v_1 + a\lambda_2 v_2 = 0. \tag{5}$$

Multiply (4) by λ_1 , we have $\lambda_1 v_1 + a\lambda_1 v_2 = 0$. Subtract (4) from (5) it follows that,

$$a(\lambda_2 - \lambda_1)v_2 = 0.$$

This implies $\lambda_1 = \lambda_2$, which is the contradiction. □

A is similar to B, iff $\exists P$ non singular
s.t. $A = P B P^{-1}$

3 Similarity

A and B are $n \times n$ matrices, then A is similar to B if there is an invertible matrix P such that $P^{-1}AP = B$, or, equivalently, $A = PBP^{-1}$. We say that A and B are similar. Changing A into $P^{-1}AP$ is called a similarity transformation.

Theorem 3.1. If $n \times n$ matrices A and B are similar, then they have the same characteristic polynomial and hence the same eigenvalues (with the same multiplicities).

Proof. A and B are similar, i.e., $A = PBP^{-1}$ for some P . Let λ and v be an eigen-pair of A , i.e., $Av = \lambda v$. We now show that λ is also an eigenvalue of B . Substitute $A = PBP^{-1}$ into $Av = \lambda v$, it follows that,

$$BP^{-1}v = \lambda P^{-1}v.$$

Let us denote $w = P^{-1}v$. w is not zero since P is invertible (because the null space of B is zero). This implies that λ is also an eigenvalue of B . □

A is diagonalizable if $A = P D P^{-1}$, where D
is a diagonal matrix, & P is invertible.

4 Diagonalizable

Theorem 4.1. An $n \times n$ matrix A is diagonalizable if and only if A has n linearly independent eigenvectors.

In fact, $A = PDP^{-1}$, with D a diagonal matrix, if and only if

$$P = [P_1 \dots P_n]$$

$$D = \begin{pmatrix} \lambda_1 & & \\ & \dots & \\ & & \lambda_n \end{pmatrix} \Rightarrow AP_i = \lambda_i P_i$$

- the columns of P are n linearly independent eigenvectors of A .
- the diagonal entries of D are eigenvalues of A that correspond, respectively, to the eigenvectors in P .

Remark 4. If $A = PDP^{-1}$ for some invertible P and diagonal D , then A^k is easy to compute.

Thm 3.1.

If A & B are similar, $\exists P$ (invertible)

s.t. $A = PBP^{-1}$

Suppose (λ, v) is an eigen pair of A ,

$$Av = \lambda v \quad (v \neq 0)$$

Replace A by PBP^{-1}

$$PBP^{-1}v = \lambda v$$

$$\Rightarrow \underbrace{BP^{-1}v}_w = \lambda \underbrace{P^{-1}v}_w$$

$$Bw = \lambda w$$

By the Rank theorem

$$\dim(\text{null}(P^{-1})) + \underbrace{\text{rank}(P^{-1})}_n = n$$

$$\Rightarrow \dim(\text{null}(P^{-1})) = 0$$

$$\Rightarrow \text{null}(P^{-1}) = \{0\}$$

Since $v \neq 0$, $\text{rank}(P^{-1}) = n$

$\Leftrightarrow \text{null}(P^{-1}) = \{0\} \Rightarrow P^{-1}v \neq 0 \Rightarrow (P^{-1}v, w)$ is an eigen-pair of B .

A is Unitary Diagonalizable.

if $A = UDU^*$, where U is unitary & D is diagonal.

① $U^* = (\bar{U})^t$

U has orthonormal columns.

② U is unitary, if $U^*U = UU^* = I \rightarrow$

③ D is a diagonal matrix.

We will discuss more about the unitary matrix next time.

5 Schur Factorization

A Schur factorization of a matrix A is a factorization

$$A = QTQ^*, \quad (6)$$

where Q is unitary and T is upper-triangular. A and T are similar, the eigenvalues of A necessarily appear on the diagonal of T .

Theorem 5.1 (Schur). Every square matrix $A \in \mathbb{C}^{n \times n}$ has a Schur factorization.

Proof. Let us prove by induction. Case $n = 1$ is trivial. Let us prove the result for $n > 1$. Assume the result is true for $n - 1$ size square matrix. Let y be any eigenvector of A with eigenvalue λ , define an unitary matrix $U = [x, P_{n \times (n-1)}] \in \mathbb{C}^{n \times n}$ with the first column being normalized y (denoted as x).

$$\begin{bmatrix} x^* \\ P^* \end{bmatrix} A \begin{bmatrix} x & P \end{bmatrix} = \begin{bmatrix} x^* A \\ P^* A \end{bmatrix} \begin{bmatrix} x & P \end{bmatrix} = \begin{bmatrix} x^* Ax & x^* AP \\ P^* Ax & P^* AP \end{bmatrix} = \begin{bmatrix} \lambda & B \\ 0 & C \end{bmatrix}, \quad (7)$$

where $B \in \mathbb{R}^{1 \times (n-1)}$ and $C \in \mathbb{R}^{(n-1) \times (n-1)}$. By the inductive hypothesis, there exists a Schur factorization of VRV^* of C , where V is unitary and R is upper-triangular. Define,

$$Q = U \begin{bmatrix} 1 & 0 \\ 0 & V \end{bmatrix}. \quad (8)$$

It is easy to verify Q is unitary and we have,

$$Q^* A Q = \begin{bmatrix} 1 & 0 \\ 0 & V^* \end{bmatrix} \begin{bmatrix} \lambda & B \\ 0 & C \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & V \end{bmatrix} \quad (9)$$

$$= \begin{bmatrix} \lambda & B \\ 0 & V^* C \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & V \end{bmatrix} = \begin{bmatrix} \lambda & BV \\ 0 & R \end{bmatrix} := T. \quad (10)$$

Since A and T are similar, they have the same eigenvalues. □

Thm 5.1. (Schur)

Let us prove by mathematical induction.

Step 1. when $n=1$, trivial.

Step 2. Assume the result is true for $n-1$ square matrix.

Step 3. Let $A \in \mathbb{C}^{n \times n}$ & (λ, x) be an eigen-pair of A ,

& further assume that $\|x\|^2 = 1$.

unitary matrix:

all cols are orthogonal to each other & $\| \cdot \| = 1$.

Define an unitary matrix $U = \begin{bmatrix} x & \vdots \\ p^* & \vdots \end{bmatrix} \in \mathbb{C}^{n \times n}$

$$U^* A U = \begin{pmatrix} x^* \\ p^* \end{pmatrix} A \underbrace{\begin{pmatrix} x & P \end{pmatrix}}_U$$

$$\downarrow x = \begin{pmatrix} 1 \\ \vdots \\ 0 \end{pmatrix}$$

$$P = \begin{pmatrix} 0 & 0 & \dots & 0 \\ 1 & 0 & \dots & 0 \\ 0 & 1 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & 1 \end{pmatrix}$$

$$= \begin{pmatrix} x^* A & \\ p^* A & \end{pmatrix}_{2,1} \begin{pmatrix} x & P \end{pmatrix}_{1,2}$$

$$= \begin{pmatrix} x^* A x & x^* A P \\ p^* A x & p^* A P \end{pmatrix}$$

$$= x^* \begin{pmatrix} Ax \\ \| \end{pmatrix}$$

$$= \lambda x^* x$$

$$= \lambda \|x\|^2 = \lambda$$

$$= p^* \lambda x$$

$$= \lambda p^* x$$

$$= 0$$

$$\begin{pmatrix} \lambda & B \\ 0 & C \end{pmatrix}$$

$B = x^* A P$
 $C = p^* A P$

$C \in \mathbb{C}^{(n-1) \times (n-1)}$. By the hypothesis, there exist unitary matrix V & an upper triangular matrix R s.t. $C = VRV^*$

Define
$$Q = U_{n,n} \begin{bmatrix} 1 & 0 \\ 0 & V \end{bmatrix}$$

$(n-1)(n-1)$

Now let us verify Q is unitary.

$$\begin{aligned} QQ^* &= U \begin{bmatrix} 1 & 0 \\ 0 & V \end{bmatrix} \underbrace{\begin{bmatrix} 1 & 0 \\ 0 & V^* \end{bmatrix}}_{Q^*} U^* \\ &= U \begin{bmatrix} 1 & 0 \\ 0 & \underbrace{VV^*}_{= I_{n-1}} \end{bmatrix} U^* \\ &= UU^* = I \end{aligned}$$

$$T := Q^* A Q$$

$$= \begin{bmatrix} 1 & 0 \\ 0 & V^* \end{bmatrix} U^* A U \begin{bmatrix} 1 & 0 \\ 0 & V \end{bmatrix}$$

$$= \begin{bmatrix} \lambda & B \\ 0 & V^* C \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & V \end{bmatrix}$$

$$= \begin{bmatrix} \lambda & BV \\ 0 & R \end{bmatrix}$$

① R is an upper triangular matrix

$\Rightarrow T$ is also upper-triangular.

② T & A have the same eigenvalues.

(b/c, T & A are similar & thm 3.1)

③ T is upper triangular

\Rightarrow diagonal entries of T are e-values of T

\Rightarrow diagonal entries of T are e-values of A .