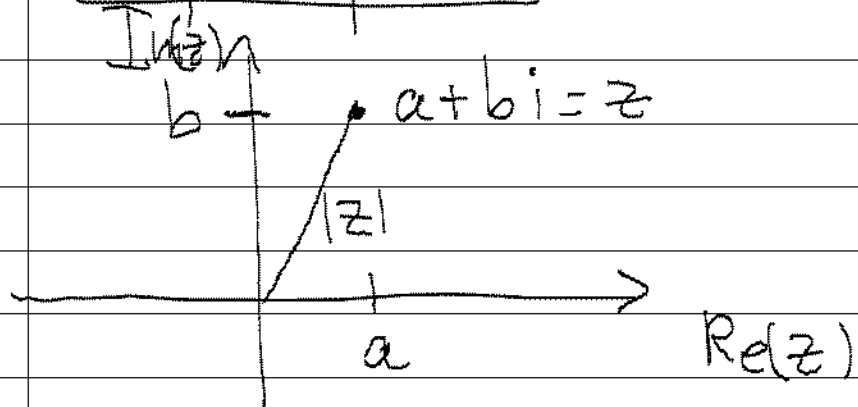


Def: $z = a + bi$ a and $b \in \mathbb{R}$. The complex conjugate of z is $\bar{z} = a - bi$

Example: $z = 1 + 2i$ then $\bar{z} = 1 - 2i$

Obs: $z \bar{z} = (a + bi)(a - bi) = a^2 - (bi)^2 = a^2 + b^2$

Complex plane



Def: The modulus or absolute value of

$z = a + bi$ is $|z| = \sqrt{a^2 + b^2} = \sqrt{z \bar{z}}$

Def: \mathbb{C}^n is the set of all n -vectors whose components are complex numbers $z \in \mathbb{C}^n$, then

$$z = \begin{bmatrix} z_1 \\ z_2 \\ \vdots \\ z_n \end{bmatrix} \quad z_i \in \mathbb{C}$$

Obs: $z = a + bi$ $w = c + di$

$$\frac{z}{w} = \frac{z \overline{w}}{w \overline{w}} = \frac{z \overline{w}}{|w|^2} = \frac{(a+bi)(c-di)}{(c+di)(c-di)} = \frac{(ac+bd) + (bc-ad)i}{(c^2+d^2)}$$

Obs: $A \in \mathbb{R}^{n \times n}$ we can have complex eigenvalues and eigen
vectors

$$Av = \lambda v \quad \lambda \in \mathbb{C} \quad \text{and} \quad v \in \mathbb{C}^n \quad v \neq 0$$

Obs: Let $P(x)$ be a real polynomial. Let $z \in \mathbb{C}$. If $P(z) = 0$ then

$$(*) \quad P(z) = a_n z^n + a_{n-1} z^{n-1} + \dots + a_1 z + a_0 = 0$$

(note) If w_1 and w_2 are two complex numbers, then

$$\overline{w_1 + w_2} = \overline{w_1} + \overline{w_2} \quad \text{and} \quad \overline{w_1 w_2} = \overline{w_1} \overline{w_2}$$

Back to $(*)$. Take complex conjugate of $(*)$

$$\overline{a_n z^n + a_{n-1} z^{n-1} + \dots + a_1 z + a_0} = \overline{0}$$

$$\text{Then } \overline{a_n} \overline{z}^n + \overline{a_{n-1}} \overline{z}^{n-1} + \dots + \overline{a_1} \overline{z} + \overline{a_0} = 0 \quad \text{but } a_i \in \mathbb{R}$$

$$a_n \overline{z}^n + a_{n-1} \overline{z}^{n-1} + \dots + a_1 \overline{z} + a_0 = 0, \quad \text{i.e.}$$

$$P(\bar{z}) = 0$$

We proved that, if $P(x)$ is real, then $P(z) = 0$ if and only if

$$P(\bar{z}) = 0$$

Example: both $-1+i$ and $-1-i$ are roots of $x^2+2x+2=0$

Obs: Let $A \in \mathbb{R}^{n \times n}$. Then

$A v = \lambda v$ if and only if $A \bar{v} = \bar{\lambda} \bar{v}$

$$v = \begin{bmatrix} v_1 \\ \vdots \\ v_n \end{bmatrix}$$

then

$$\bar{v} = \begin{bmatrix} \bar{v}_1 \\ \vdots \\ \bar{v}_n \end{bmatrix}$$

λ is an eigenvalue of A with eigenvector v if and only if $\bar{\lambda}$ is an eigenvalue of A with eigenvector \bar{v} ($A \in \mathbb{R}^{n \times n}$)

Example: Find the eigenvalues and eigenvectors of

$$A = \begin{bmatrix} 6 & -1 \\ 5 & 4 \end{bmatrix}$$

$$P(\lambda) = (6-\lambda)(4-\lambda) + 5 = \lambda^2 - 10\lambda + 29 = 0$$

$$\boxed{\lambda = \frac{10 \pm \sqrt{100 - 116}}{2} = \frac{10 \pm \sqrt{-16}}{2} = \frac{10 \pm i\sqrt{16}}{2} = \boxed{5 \pm 2i}}$$

Eigenvector of $\boxed{\lambda_1 = 5 + 2i}$

$$A - \lambda_1 I = \begin{bmatrix} 1-2i & -1 \\ 5 & -1-2i \end{bmatrix}$$

$$(A - \lambda_1 I) v = 0$$

$(1-2i)v_1 + (-1)v_2 = 0$ to get a solution set $v_1 = 1$

$$v = \begin{bmatrix} 1 \\ 1-2i \end{bmatrix}$$

Eigen vectors	Eigen values
$\begin{bmatrix} 1 \\ 1-2i \end{bmatrix}$	$5+2i$
$\begin{bmatrix} 1 \\ 1+2i \end{bmatrix}$	$5-2i$

Obs: $\lambda = 0$ is an eigenvalue of A if and only if A is singular

Obs: $\lambda \neq 0$, A non-singular, $A v = \lambda v$ $v \neq 0$.

$A^{-1}(A v) = A^{-1}(\lambda v)$ | A non-singular. λ is an eigenvalue of

$(A^{-1}A) v = \lambda A^{-1} v$ | A with eigenvector v if and only if

$I v = \lambda A^{-1} v$ | λ^{-1} is an eigenvalue of A^{-1} with eigen-

$v = \lambda A^{-1} v$ | vector v .

$$\boxed{\lambda^{-1} v = A^{-1} v}$$

Symmetric matrices

Def: A is symmetric if $A^T = A$

Example $\begin{bmatrix} 5 & -1 \\ -1 & 1 \end{bmatrix}$ is symmetric

Obs: $A v = \lambda v$ $A \in \mathbb{R}^{n \times n}$ A symmetric $\lambda \in \mathbb{C}$ $v \in \mathbb{C}^n$

$$\bar{v}^T (A v) = \bar{v}^T (\lambda v)$$

$$\bar{v}^T A v = \lambda \bar{v}^T v$$

transpose

$$(\bar{v}^T A v)^T = \lambda (\bar{v}^T v)^T$$

$$v^T A^T (\bar{v}^T)^T = \lambda v^T (\bar{v}^T)^T$$

then

$$v^T A \bar{v} = \lambda (v^T \bar{v})$$

take complex conjugate

$$\bar{v}^T \bar{A} v = \bar{\lambda} (\bar{v}^T v) \quad \text{but } A \text{ is real}$$

$$\bar{v}^T A v = \bar{\lambda} (\bar{v}^T v)$$

then

$$\lambda (\bar{v}^T v) = \bar{\lambda} (\bar{v}^T v) \quad (*)$$

$$\bar{v}^T v = [\bar{v}_1 \quad \bar{v}_2 \quad \dots \quad \bar{v}_n] \begin{bmatrix} v_1 \\ v_2 \\ \vdots \\ v_n \end{bmatrix} = |v_1|^2 + |v_2|^2 + \dots + |v_n|^2 > 0$$

because $v \neq 0$

Then \otimes implies

$$\lambda = \bar{\lambda}$$

Theorem: All the eigenvalues of a symmetric real matrix are real.

Inner product: $x, y \in \mathbb{R}^n$

$$x \cdot y = x_1 y_1 + x_2 y_2 + \dots + x_n y_n = [x_1 \quad x_2 \quad \dots \quad x_n] \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{bmatrix} = x^T y$$

$$\|x\| = \sqrt{x \cdot x} = \sqrt{x^T x}$$

x and y are orthogonal if $x^T y = 0$

Theorem: $A \in \mathbb{R}^{n \times n}$. A symmetric. Eigenvectors of A of different eigenvalues are orthogonal.

proof: $A v_1 = \lambda_1 v_1$ and $A v_2 = \lambda_2 v_2$ $\lambda_1 \neq \lambda_2$ $v_1 \neq 0$
 $v_2 \neq 0$

$$v_2^T A v_1 = \lambda_1 v_2^T v_1$$

$$v_1^T A v_2 = \lambda_2 v_1^T v_2$$

transpose

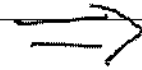
$$v_1^T A^T v_2 = \lambda_1 v_1^T v_2$$

$$v_1^T A v_2 = \lambda_1 v_1^T v_2$$

but $A^T = A$

$$\lambda_1 v_1^T v_2 = \lambda_2 v_1^T v_2 \quad \text{but}$$

$$\lambda_1 \neq \lambda_2$$



$$v_1^T v_2 = 0$$

Obs: $A \in \mathbb{R}^{n \times n}$. $P(\lambda) = \det(A - \lambda I)$

$P(\lambda)$ is a polynomial of degree n . There exists $\lambda_1, \dots, \lambda_r \in \mathbb{C}$ and n_1, \dots, n_r positive integers such that

$$n_1 + n_2 + \dots + n_r = n \quad \text{and}$$

$$P(\lambda) = (-1)^n (\lambda - \lambda_1)^{n_1} \dots (\lambda - \lambda_r)^{n_r}$$

$$\lambda_i \neq \lambda_j \quad \text{if } i \neq j$$

$\lambda_1, \lambda_2, \dots, \lambda_r$ are the different eigenvalues

n_i is called the algebraic multiplicity of λ_i

The geometric multiplicity of λ_i is the number of linear-

ly independent eigenvectors of A with eigenvalue λ_i = number of free variables in the system $(A - \lambda_i I)x = 0$ (same as number of parameters t_1, t_2, \dots in the solution).

Obs: $\text{geometric multiplicity of } \lambda_i \leq \text{algebraic multiplicity of } \lambda_i$

Obs: If A is symmetric,

$\text{geometric multiplicity of } \lambda_i = \text{algebraic multiplicity of } \lambda_i$

Theorem: If A is symmetric, then A has n linearly independent eigenvectors.

Orthogonal matrices: A is orthogonal if $A^{-1} = A^T$, i.e.

$$A \in \mathbb{R}^{n \times n} \quad \text{and} \quad A^T A = I$$

Example: $A = \begin{bmatrix} 1/\sqrt{2} & -1/\sqrt{2} \\ 1/\sqrt{2} & 1/\sqrt{2} \end{bmatrix}$ is orthogonal

check

$$A^T A = \begin{bmatrix} 1/\sqrt{2} & 1/\sqrt{2} \\ -1/\sqrt{2} & 1/\sqrt{2} \end{bmatrix} \begin{bmatrix} 1/\sqrt{2} & -1/\sqrt{2} \\ 1/\sqrt{2} & 1/\sqrt{2} \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

Obs: $B = \begin{bmatrix} b_1 & b_2 & \dots & b_n \end{bmatrix}$ $A = \begin{bmatrix} a_1 & a_2 & \dots & a_n \end{bmatrix}$ $a_i \in \mathbb{R}^n$
 $b_i \in \mathbb{R}^n$

$$A, B \in \mathbb{R}^{n \times n}$$

$$A^T B = \begin{bmatrix} a_1^T \\ a_2^T \\ \vdots \\ a_n^T \end{bmatrix} \begin{bmatrix} b_1 & b_2 & \dots & b_n \end{bmatrix} = \begin{bmatrix} a_1^T b_1 & a_1^T b_2 & \dots & a_1^T b_n \\ a_2^T b_1 & a_2^T b_2 & \dots & a_2^T b_n \\ \vdots & \vdots & \ddots & \vdots \\ a_n^T b_1 & a_n^T b_2 & \dots & a_n^T b_n \end{bmatrix}$$

$$C = A^T B \quad c_{ij} = a_i \cdot b_j$$

Obs: $A \in \mathbb{R}^{n \times n}$ $B = A^T A$ $b_{ij} =$ i^{th} column of A inner product j^{th} column of A

Theorem: $A \in \mathbb{R}^{n \times n}$. A is orthogonal if and only if

$$\begin{matrix} i^{\text{th}} \text{ column of } A \text{ inner} \\ \text{product } j^{\text{th}} \text{ column of } A \end{matrix} = \begin{cases} 0 & \text{if } i \neq j \\ 1 & \text{if } i = j \end{cases}$$

$$A = [a_1, a_2, \dots, a_n] \quad a_i = i^{\text{th}} \text{ column of } A$$

A is orthogonal if and only if the set a_1, a_2, \dots, a_n is orthonormal.

Diagonalization

Def: A matrix A is diagonalizable if $A = PDP^{-1}$ for some diagonal matrix D . A, D and $P \in \mathbb{R}^{n \times n}$

Fact: $P \in \mathbb{R}^{n \times n}$ has an inverse if and only if the columns of P are linearly independent.

Obs: $A = P D P^{-1}$ then $A P = P D$

$P = [p_1 \ p_2 \ \dots \ p_n]$ p_j is the j 'th column of P

$$D = \begin{bmatrix} \lambda_1 & 0 & \dots & 0 \\ 0 & \dots & \dots & 0 \\ \vdots & \vdots & \vdots & \vdots \\ 0 & \dots & 0 & \lambda_n \end{bmatrix}$$

$$\begin{aligned} A P &= A [p_1 \ p_2 \ \dots \ p_n] = \\ &= [A p_1 \ A p_2 \ \dots \ A p_n] \end{aligned}$$

$$P D = P \begin{bmatrix} \lambda_1 & & & 0 \\ & \lambda_2 & & \\ & & \dots & \\ 0 & & & \lambda_n \end{bmatrix} = \begin{bmatrix} P \begin{bmatrix} \lambda_1 \\ 0 \\ \vdots \\ 0 \end{bmatrix} & P \begin{bmatrix} 0 \\ \lambda_2 \\ \vdots \\ 0 \end{bmatrix} & \dots & P \begin{bmatrix} 0 \\ 0 \\ \vdots \\ \lambda_n \end{bmatrix} \end{bmatrix}$$

$$j^{\text{th}} \text{ column of } AP = A p_j$$

$$j^{\text{th}} \text{ column of } PD = P \begin{bmatrix} 0 \\ \vdots \\ \lambda_j \\ \vdots \\ 0 \end{bmatrix} = \lambda_j p_j$$

$$A p_i = \lambda_i p_i$$

Obs: A is diagonalizable if and only if A has n linearly independent eigenvectors.