

Examples (Diagonalization)

ex) The Rotation Matrix in $\mathbb{R}^2$: Let

$$A = \begin{bmatrix} 0 & -1 \\ 1 & 0 \end{bmatrix}. \text{ (1) compute \begin{cases} \text{e-values} \\ \text{vectors} \end{cases}}$$

$$\det(A - \lambda I) = 0 \Rightarrow p(\lambda) = \lambda^2 + 1 = 0$$

$\lambda_{1,2} = \pm i$ are the e-values.

$$(A - \lambda_1 I) x_1 = \begin{bmatrix} -i & -1 \\ 1 & -i \end{bmatrix} x_1 = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$$x_1 = \begin{bmatrix} 1 \\ -i \end{bmatrix}$$

$$(A - \lambda_2 I) x_2 = \begin{bmatrix} i & -1 \\ 1 & i \end{bmatrix} x_2 = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$$x_2 = \begin{bmatrix} 1 \\ +i \end{bmatrix} = \bar{x}_1.$$

If $z = \alpha + i\beta$
 $\bar{z} = \alpha - i\beta$ (complex conjugate)

There are 2 distinct e-vectors, x_1 & x_2 .

$$A = SDS^{-1} \quad \text{For } S = \begin{bmatrix} x_1 & x_2 \end{bmatrix}, \text{ i.e.}$$

$$S = \begin{bmatrix} 1 & 1 \\ -i & i \end{bmatrix}. \text{ Form } D = \begin{bmatrix} i & 0 \\ 0 & -i \end{bmatrix}.$$

Check on your own: $AS = SD$ //

ex) $A = \begin{bmatrix} 2 & 0 & 0 \\ 1 & 2 & 1 \\ -1 & 0 & 1 \end{bmatrix}$. Will have 2 distinct e'values, but 3
l.l. eigenvectors $\therefore$ diagonalization is possible

(1) Compute e'values:

$$\det(A - \lambda I) = 0 = (2 - \lambda)^2 (1 - \lambda) = 0$$

$\lambda = 2$, repeated root,

$\lambda = 1$, simple root.

(2) Compute eigenvectors:

$$\text{For } \lambda_1: (A - \lambda_1 I)x_1 = 0 \text{ or } x_1 = \begin{bmatrix} 0 \\ 1 \\ 1 \end{bmatrix}.$$

for $\lambda_2 = \lambda_3 = 2$

$$(A - \lambda_2 I)x_j = 0 \quad j=2,3$$

$$(A - 2I)x_j = \begin{bmatrix} 0 & 0 & 0 \\ 1 & 0 & 1 \\ -1 & 0 & -1 \end{bmatrix} x_j \sim \begin{bmatrix} 1 & 0 & 1 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = 0$$

$$x_2 = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} \text{ and } \begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix} = x_3 \text{ work and are l.l.}$$

so

$$S = \begin{bmatrix} 0 & 0 & -1 \\ -1 & 1 & 0 \\ 1 & 0 & 1 \end{bmatrix} \quad D = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 2 \end{bmatrix}$$
 //

SOME REMARKS ON $A = SDS^{-1}$

(1) If A has no repeated e'values $\Rightarrow$ it will have the required number of LI e'vectors.

(2) In $A = SDS^{-1}$, the S is not unique, but D is.

(3) $AS = SD$ holds if columns of S are e'vectors of A and not otherwise. Other matrices will not produce the D : this has to do with the rules of matrix multiplication: to see this, suppose first column of S is $v_1 \Rightarrow SD = \lambda_1 v_1$.
If this is to agree with first column of $AS = Av_1$
 $\Rightarrow v_1$ has to be the e'vector associated with λ_1 .

(4) Since not all matrices have LI e'vectors
not all matrices are diagonalizable. //

An example of a non-diagonalizable matrix:

ex) $A = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}$ $\lambda_1 = \lambda_2 = 0$ ($\det(A - \lambda I) = 0$)

so $(A - \lambda I)x = 0 = Ax = 0$, has an eigenvector $v = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$

That's the only e'vector of A .

hence A is not diagonalizable //

Thus far we covered addition and multiplication (algebra).
The next topic brings in the geometric side:

Inner Product Spaces

LADR Ch 7

Vector spaces equipped with an inner product.

Some topics we need to cover first for along the way):

- ① Cauchy-Schwarz Inequality
- ② Gram-Schmidt orthogonalization procedure.
- ③ Linear functionals on inner product (vector) spaces
- ④ Calculating distances (norm) between subspaces.

What's an inner product?

Inner Product & Norm

Spce $x \in \mathbb{R}^n$

$\|x\|_2 = \sqrt{x_1^2 + x_2^2 + x_3^2 + \dots + x_n^2}$ is the L_2 norm.

Spce that $x = \begin{pmatrix} x_1 + iy_1 \\ x_2 + iy_2 \\ \vdots \\ x_n + iy_n \end{pmatrix}$ i.e. $x \in \mathbb{C}^n$
How to compute the L_2 norm?

The complex conjugate of the vector x is

$$\bar{x} = \begin{pmatrix} x_1 - iy_1 \\ x_2 - iy_2 \\ \vdots \\ x_n - iy_n \end{pmatrix}$$

so we use an overbar to indicate complex conjugate.

L_2 norm for $x \in \mathbb{C}^n$

$$\|x\|_2 = \sqrt{x^T \bar{x}} = \sqrt{\bar{x}^T x}$$

↑
transpose

We also introduce the Hermitian of a matrix:

Hermitian

For $A \in \mathbb{C}^{n \times m}$, A^* is the Hermitian matrix.

$$A^* = \bar{A}^T. \text{ If } A \in \mathbb{R}^{n \times m} \Rightarrow A^* = A^T //$$

The dot product for two vector $v, w \in \mathbb{R}^n$

$$v \cdot w \equiv v_1 w_1 + v_2 w_2 + \dots + v_n w_n$$

$$v \cdot w = v^T w$$

Properties of the dot product:

$$\begin{cases} x \cdot x \geq 0 & \forall x \in \mathbb{R}^n \\ x \cdot x = 0 & \text{iff } x = \emptyset \\ x \cdot y = y \cdot x \end{cases} //$$

Inner Product on V is a function that takes an ordered pair (u, v) of elements of V to a number

$$\langle u, v \rangle \in \mathbb{F}. \text{ It has}$$

Properties of the inner product:

(c) Positivity: $\langle u, v \rangle \geq 0 \quad \forall u, v \in V$

(b) Definiteness: $\langle v, v \rangle = 0$ iff $v = \emptyset$.

(c) Additivity (in the n^{th} slot). For the first slot:

$$\langle u+v, w \rangle = \langle u, w \rangle + \langle v, w \rangle$$

(d) Homogeneity (in the n^{th} slot)
first slot:

$$\langle \lambda u, v \rangle = \lambda \langle u, v \rangle \quad \lambda \in \mathbb{F}$$

(e) Conjugate Symmetry

$$\langle u, v \rangle = \overline{\langle v, u \rangle} \quad \forall u, v \in V$$

Ex) Euclidean inner product on $\mathbb{R}^n$:

let $w, v \in \mathbb{R}^n$, the inner product is

$$\langle w, v \rangle = w_1 \bar{v}_1 + w_2 \bar{v}_2 + \dots + w_n \bar{v}_n$$

Ex) Inner product on vector space of (possibly complex)

continuous functions on the interval $[1, 1]$:

$$\langle f, g \rangle = \int_1^1 f(x) \bar{g}(x) dx \quad \text{if } g \in \mathbb{R}, \text{ then } \bar{g} = g$$

ex) Inner product on $\mathcal{P}(\mathbb{R}^+ \setminus 0)$
(weighted). let $p, q \in \mathcal{P}(\mathbb{R}^+ \setminus 0)$

$$\langle p, q \rangle = \int_0^\infty p(x) q(x) e^{-x} dx //$$

We're ready to define an

Inner Product space: a vector space
equipped with an inner product. //