

Orthogonality :

- Orthogonal vectors & Subspaces
- Projections
- Orthonormal Basis
- Gram-Schmidt

v_1, v_2, \dots, v_n $\xrightarrow{\text{Gram-Schmidt}}$ e_1, e_2, \dots, e_n (Examples)
 linearly independent Orthonormal

$$A_{m \times n} = QR_{n \times n}$$

R upper-triangular + invertible
diagonal entries > 0 .

The columns of Q are orthonormal

$$\text{rank}(A) = r.$$

Determinants :

- Definition and Examples

$A_{n \times n}$

$|A|$ or $\det A$

- Properties 1-10.
- Formulas : (a) LDU(?)
(b) Cofactor expansion:

Important:

(a) $\det(AB) = \det(A) \det(B)$;

(b) $\det(A^T) = \det(A)$;

(c) $\det A = \sum_{\text{All } P's} a_{1\alpha} a_{2\beta} \dots a_{n\gamma} \det P.$

$$\det(A) = a_{11}C_{11} + a_{12}C_{12} + \dots + a_{1n}C_{1n}$$

(c) Techniques: Recursion Relation,
Vandermonde Matrix.

(d) Applications: Inverse / Volume / Cramer's Rule /
Pivots.

Eigenvalues & Eigenvectors

 $A_{n \times n}$ Square matrix

$$Ax = \lambda x$$

 λ : eigenvalue $v (\neq 0)$: eigenvector

$$\Leftrightarrow (A - \lambda I)x = 0$$

$$\Leftrightarrow x \in N(A - \lambda I)$$

 $P(\lambda) = \det(A - \lambda I)$: characteristic polynomial

$$\deg P(\lambda) = n$$

Computation:

 $\lambda = \lambda_1$: λ_1 eigenvalue

$$P(\lambda) = (\lambda_1 - \lambda)(\lambda_2 - \lambda) \dots (\lambda_n - \lambda)$$

$$= (-\lambda)^n + (\lambda_1 + \dots + \lambda_n)(-\lambda)^{n-1} + \dots + \underbrace{\lambda_1 \dots \lambda_n}_{\det A}$$

Solve $Ax = \lambda x$ for x

$$= a_{11} + \dots + a_{nn}$$

 $x \rightarrow$ eigenvector.Diagonalizability : $A_{n \times n}$ diagonalizable \Leftrightarrow There exist n linearly independent eigenvectors:

$$v_1, v_2, \dots, v_n$$

$$S = \begin{bmatrix} v_1 & v_2 & \dots & v_n \end{bmatrix} \rightarrow \text{eigenvector matrix}$$

$$\begin{matrix} \downarrow & \downarrow & & \downarrow \\ \lambda_1 & \lambda_2 & \dots & \lambda_n \end{matrix} \rightarrow \text{eigenvalues}$$

$$\Lambda = \begin{bmatrix} \lambda_1 & & 0 \\ & \lambda_2 & \\ 0 & & \ddots \\ & & & \lambda_n \end{bmatrix} \rightarrow \text{eigenvalue matrix}$$

$$S^{-1}AS = \Lambda \quad (\text{putting together})$$

$$\Leftrightarrow A = S\Lambda S^{-1}$$

$$A^k = S\Lambda^k S^{-1}, \quad k=1,2,\dots$$

Table on Page 306. Properties of Eigenvalues and Eigenvectors.

Complex number / vector / matrix.

- $z = a + bi : a, b \in \mathbb{R}$

- $\vec{z} = \begin{bmatrix} z_1 \\ z_2 \\ \vdots \\ z_n \end{bmatrix} \in \mathbb{C}^n, \quad \|\vec{x}\|^2 = \vec{x}^H \vec{x}$

- $A^H = (\bar{A})^T.$

$$A^H = A$$

Hermitian matrix : Properties.

- Unitary Matrix : $U^H U = U U^H = I_n$

Properties.

- Similarity Transformations:

$$A \longrightarrow M^{-1} A M = B$$

Same eigenvalues

 x eigenvector of A $\Rightarrow M^{-1}x$ eigenvector of B .

Change of Basis

- Spectral Theorem / Real : $Q^T A Q = \Lambda$ given that $A = A^T$

(Schur's Lemma) \ Complex : $U^H A U = \Lambda$ given that $A^H = A$.

- Normal Matrix

- Jordan Forms ($n=2, 3$)

- Definitions: Positive Definiteness + Quadratic Forms

$$x^T A x = f(x_1, \dots, x_n)$$

- Tests: $G B + G C \leftrightarrow$ Positive Definiteness.
 $G.D \leftrightarrow$ Positive Semidefiniteness.

- Quadratic Surfaces.

- Law of Inertia.

- Quadratic Forms.

- Singular Value Decomposition.

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

$$= \begin{bmatrix} -\frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \end{bmatrix} \begin{bmatrix} \sqrt{3} & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} \frac{1}{\sqrt{6}} & \frac{2}{\sqrt{6}} & \frac{1}{\sqrt{6}} \\ -\frac{1}{\sqrt{2}} & 0 & \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{3}} & \frac{1}{\sqrt{3}} & \frac{1}{\sqrt{3}} \end{bmatrix} = U \Sigma V^T$$

- Applications.

Additional Page.

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$$Ax = b \quad \text{Nonhomogeneous} \qquad Ax = 0 \quad \text{Homogeneous}$$

$$A_{m \times n} : \begin{array}{l} m = n, \\ m < n, \\ m > n. \end{array} \quad \begin{array}{l} \text{Special Solution(s)} \\ \text{Complete Solution(s)} \end{array}$$

Matrix :

- Invertible matrix
- rank
- LU Factorization / QR / SVD.
- Four Fundamental Subspaces
- Eigenvalues and Eigenvectors
- Determinants.

Special Matrices:

- Projection matrix.
- Rotation matrix.
- Orthogonal matrix.
- Positive definite matrix.
- Hermitian Matrix.
- Unitary Matrix.

Example: $A_{n \times m}, B_{m \times n}$

$$(a) \begin{vmatrix} I_m & B \\ A & I_n \end{vmatrix} = |I_n - AB| = |I_m - BA|$$

$$\begin{bmatrix} I_m & 0 \\ -A & I_n \end{bmatrix} \begin{bmatrix} I_m & B \\ A & I_n \end{bmatrix} = \begin{bmatrix} I_m & B \\ 0 & I_n - AB \end{bmatrix}$$

$$\Rightarrow |I_n - AB| = \begin{vmatrix} I_m & B \\ A & I_n \end{vmatrix} \dots (1) \quad \text{Meanwhile,}$$

$$\begin{bmatrix} I_m & B \\ A & I_n \end{bmatrix} \begin{bmatrix} I_m & 0 \\ -A & I_n \end{bmatrix} = \begin{bmatrix} I_m - BA & B \\ 0 & I_n \end{bmatrix}$$

$$\Rightarrow |I_m - BA| = \begin{vmatrix} I_m & B \\ A & I_n \end{vmatrix} \dots (2) \quad (1) + (2) \Rightarrow \text{Done! \#}$$

$$\lambda \neq 0: |\lambda I_n - AB| = \lambda^{n-m} |\lambda I_m - BA|$$

$$\left| I_n - A \left(\frac{B}{\lambda} \right) \right| = \left| I_m - \left(\frac{B}{\lambda} \right) A \right| \Rightarrow \frac{1}{\lambda^n} |\lambda I_n - AB| = \frac{1}{\lambda^m} |\lambda I_m - BA|$$

$$\Rightarrow |\lambda I_n - AB| = \lambda^{n-m} |\lambda I_m - BA|$$