

Linear Algebra cheat sheet

Vectors

dot product: $u \cdot v = \|u\| \cdot \|v\| \cdot \cos(\phi) = u_x v_x + u_y v_y$

cross product: $u \times v = \begin{pmatrix} u_y v_z - u_z v_y \\ u_z v_x - u_x v_z \\ u_x v_y - u_y v_x \end{pmatrix}$

enclosed angle:

$$\cos \phi = \frac{u \cdot v}{\|u\| \cdot \|v\|}$$
$$\|u\| \cdot \|v\| = \sqrt{(u_x^2 + u_y^2)(v_x^2 + v_y^2)}$$

Matrices

basic operations

transpose: $[A^T]_{ij} = [A]_{ji}$: "mirror over main diagonal"

multiply: $A_{N \times M} \cdot B_{M \times K} = M_{N \times K}$

invert: $\begin{bmatrix} a & b \\ c & d \end{bmatrix}^{-1} = \frac{1}{\det(A)} \begin{bmatrix} d & -b \\ -c & a \end{bmatrix} = \frac{1}{ad-bc} \begin{bmatrix} d & -b \\ -c & a \end{bmatrix}$

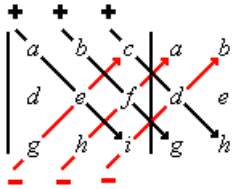
norm:

condition:

determinants

$\det(A) = \sum_{\sigma \in S_n} \text{sgn}(\sigma) \prod_{i=1}^n A_{i, \sigma_i}$

For 3x3 matrices (Sarrus rule):



arithmetic rules:

$\det(A \cdot B) = \det(A) \cdot \det(B)$

$$\det(A^{-1}) = \frac{1}{\det(A)}$$

$\det(rA) = r^n \det A$, for all $A^{n \times n}$ and scalars r

eigenvalues, eigenvectors, eigenspace

1. Calculate **eigenvalues** by solving $\det(A - \lambda I) = 0$
2. Any vector x that satisfies $(A - \lambda_i I)x = 0$ is **eigenvector** for λ_i .
3. $\text{Eig}_A(\lambda_i) = \{x \in \mathbb{C}^n : (A - \lambda_i)x = 0\}$ is **eigenspace** for λ_i .

definiteness

obvious properties

square: $N \times N$

symmetric: $A = A^T$

diagonal: 0 except a_{kk}

\Rightarrow implies triangular (eigenvalues on main diagonale)

orthogonal

$$A^T = A^{-1}$$

nonsingular

A is nonsingular = invertible = regular iff:

There is a matrix $B := A^{-1}$ such that $AB = I = BA$.

$$\det(A) \neq 0$$

$TAx = b$ has exactly one solution for each b.

The columns of A are linearly independent.

$$\Rightarrow \det(A)^{-1} = \det(A^{-1})$$

$$\Rightarrow (A^{-1})^{-1} = A$$

$$\Rightarrow (A^T)^{-1} = (A^{-1})^T$$

diagonalizable

$A^{N \times N}$ can be diagonalized iff it has n linear independent eigenvectors. Then, there is an invertible T , such that:

$$D := T^{-1}AT = \begin{pmatrix} \lambda_1 & & \\ & \ddots & \\ & & \lambda_n \end{pmatrix}$$

$$A = T^{-1}DT \quad \text{and} \quad AT = T \begin{pmatrix} \lambda_1 & & \\ & \ddots & \\ & & \lambda_n \end{pmatrix}$$

T can be created with eigenvectors of A.

diagonally dominant matrix

$$\forall i, |a_{ii}| \geq \sum_{j \neq i} |a_{ij}|$$

Hermitian

triangular

\Rightarrow Eigenvalues on main diagonale

unitary

$$\Rightarrow \text{cond}_2(A) = 1$$

idempotent

adjugate

block matrices

$$\det \begin{pmatrix} A & 0 \\ C & D \end{pmatrix} = \det \begin{pmatrix} A & B \\ 0 & D \end{pmatrix} = \det(A) \det(D)$$

rank

kernel

trace

defined on $n \times n$ square matrices: $\text{tr}(A) = a_{11} + a_{22} + \dots + a_{nn}$
(sum of the elements on the main diagonal)

span

spectrum

$$\sigma(A) = \{\lambda \in \mathbb{C} : \lambda \text{ is eigenvalue of } A\}$$