

Def: A vector is a tuple of numbers.

Def: The dimension is the size of this tuple.

Def: A matrix is a 2-dimensional grid of numbers.

Def: $\mathbb{R}^{m \times n}$ denotes all $m \times n$ matrices with field \mathbb{R} .

Def: $\mathbb{C}^{m \times n}$ denotes all $m \times n$ matrices with field \mathbb{C} .

Def: vector-matrix product: $\begin{bmatrix} a & b \\ c & d \end{bmatrix} \cdot \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} ax + by \\ cx + dy \end{bmatrix}$ (Inner dimensions must agree. Outer dimensions remain.)

$\begin{bmatrix} ax + by \\ cx + dy \end{bmatrix} = x \begin{bmatrix} a \\ c \end{bmatrix} + y \begin{bmatrix} b \\ d \end{bmatrix}$: Express matrix-vector multiplication as a linear combination of the matrix.

Def: Inner product $< \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix}, \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{bmatrix} > = [x_1 \ x_2 \ \cdots \ x_n] \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{bmatrix} = x_1 y_1 +$

$x_2 y_2 + \cdots + x_n y_n$

Def: Matrix Multiplication: $A \in \mathbb{R}^{m \times k}, B \in \mathbb{R}^{k \times n}, C = AB \in \mathbb{R}^{m \times n}$ and $C_{ij} = \sum_{k=1}^n A_{ik} B_{kj}$.

Def: span of vectors v_1, v_2, \dots, v_n is the set of vectors that can be obtained as linear combination of the vectors v_1, v_2, \dots, v_n :

$\text{span}\{v_1, v_2, \dots, v_n\} = a_1 v_1 + a_2 v_2 + \cdots + a_n v_n$ for a_1, a_2, \dots, a_n are scalars.

Def: A set of vector v_1, v_2, \dots, v_n is said to be linearly independent if none of these vectors can be expressed as a linear combination of others.

Solving a system of linear equations

Def: Augmented matrix - append \vec{b} to A , where $A \in \mathbb{R}^{m \times n}$ and $b \in \mathbb{R}^n$.

Def: A pivot in a row is its leftmost non-zero element.

Note: if the column space of A , $\mathcal{C}(A)$ is LD, there is no soln or infinite number solns.

Matrix Multiplication: $AB = C$ iff $C_{ik} = \langle A_i^T, B_k \rangle$.

Application: find the walk of length n from node i to j in a graph. The answer will be A_{ij}^n , where A is the adjacency matrix of the graph.

Thm: If $A^{-1}A = I$ and $A\tilde{A}^{-1} = I$, then $A^{-1} = \tilde{A}^{-1}$

Def: A is symmetric iff $A^T = A$.

Def: A matrix whose transpose is also its inverse is called orthogonal. $A^T A = I \implies \langle \text{col } i, \text{col } j \rangle = \delta_{ij}$.

Def: length of a vector. $\|v\| = \sqrt{\langle v, v \rangle}$. (Also known as L2 norm).

Def: Angle between two vectors. $\frac{\langle v, w \rangle}{\|v\| \|w\|} = \cos \theta$.

Def: Permutation is the operation where in each element in the pre-image appears exactly once in the image.

Properties of a permutation matrix:

- square matrix
- there is a one '1' per every row/col

- all rows are distinct, all columns are distinct.

Suppose A and B are both permutation matrices of the same dimension. AB is another permutation.

Permutations don't change the length of the vector and the between vectors.

Def: Transpose $A_{ij}^T = A_{ji}$. A does NOT need to be square.

Thm: If A is a permutation matrix, then its transpose is its inverse. Moreover, A is also an orthogonal matrix.

Rotation: matrix $\begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix}$, one can show that apply a rotation matrix to a vector does NOT alter its length. Rotation matrices are orthogonal.