



# Linear Algebra Refresher Part 1

An introduction of Vector and Matrix



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## Topics To Be Covered:

- A** Scalars, Vectors and Matrices
- B** Operations on Vectors and Matrices
- C** Special Matrices
- D** Determinants and Inverse of Matrix
- E** Solving Simultaneous Equations

# A Scalars



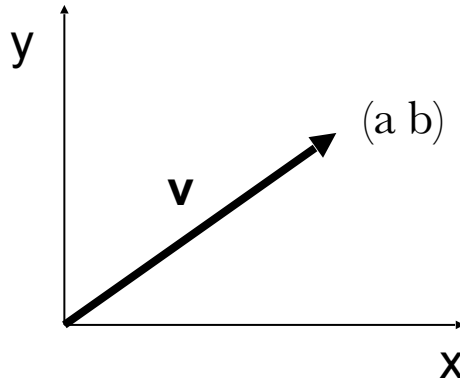
Any real number, or any quantity that can be measured using a single real number. Temperature, length, and mass are all scalars. A scalar has magnitude but no direction.



# A What is a Vector?

- Column of numbers e.g. height, weight and age of a person.
- Think of a vector as a directed line segment in N-dimensions! (has “length” and “direction”).

$$\mathbf{x} = \begin{pmatrix} \text{height} \\ \text{weight} \\ \text{age} \end{pmatrix}$$

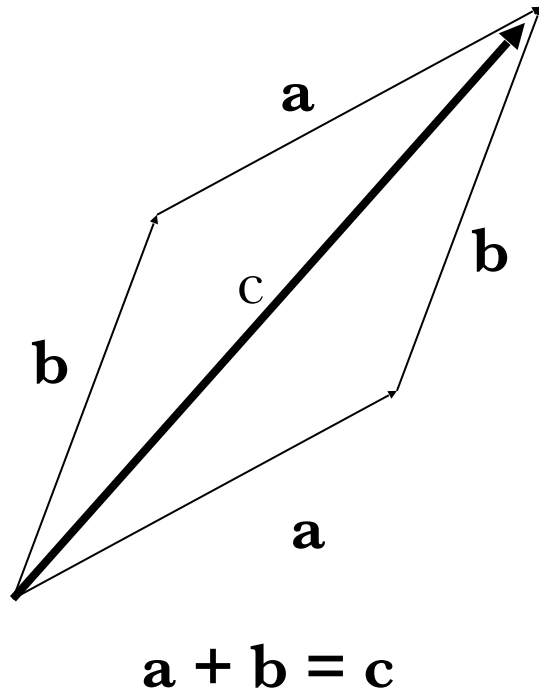


$$\mathbf{x}_n = \begin{pmatrix} x_1 \\ x_2 \\ x_3 \\ \dots \\ x_n \end{pmatrix}$$

# B Vector Addition



$$\mathbf{x} + \mathbf{y} = (x_1, x_2) + (y_1, y_2) = (x_1 + y_1, x_2 + y_2)$$



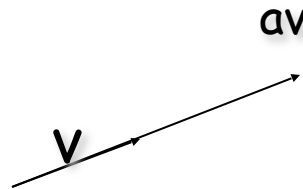
The Statement of Parallelogram law of vector addition is, If two vectors are considered to be the adjacent sides of a parallelogram, then the resultant of two vectors is given by the vector that is a diagonal passing through the point of contact of two vectors.



# Scalar Product



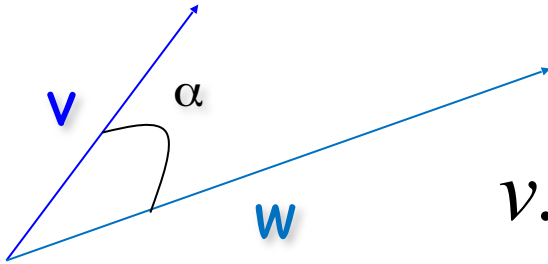
$$a\mathbf{v} = a(x_1, x_2) = (ax_1, ax_2)$$



- Multiplying a vector by scalar changes its magnitude but keeps the direction fixed.
- This operation is also known as scaling.



## B Inner (dot) Product



$$\mathbf{v} \cdot \mathbf{w} = (x_1, x_2) \cdot (y_1, y_2) = x_1 y_1 + x_2 y_2$$

$$\mathbf{v} \cdot \mathbf{w} = (x_1, x_2) \cdot (y_1, y_2) = \|\mathbf{v}\| \cdot \|\mathbf{w}\| \cos \alpha$$

$$\mathbf{v} \cdot \mathbf{w} = 0 \Leftrightarrow \mathbf{v} \perp \mathbf{w}$$

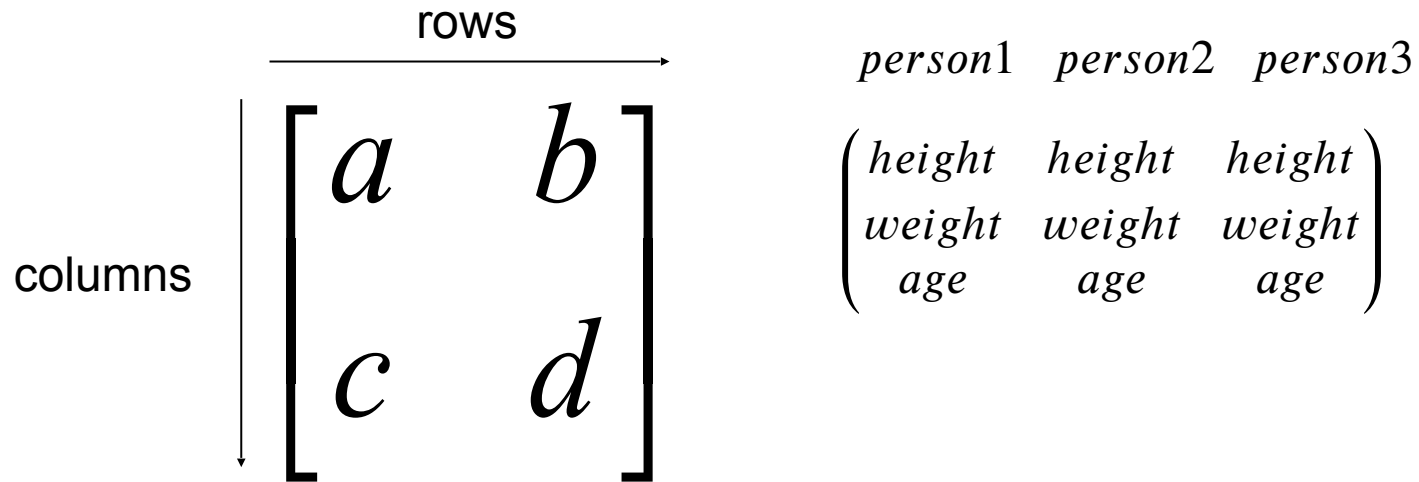
- When two vectors are orthogonal (or perpendicular to each other) their inner product is zero.
- The  $L_2$  norm of a vector  $\mathbf{v}$  is the square root of self dot product. That is square root of  $(\mathbf{v} \cdot \mathbf{v}) = \text{square root of } (x_1^2 + x_2^2)$ .





# A What is a Matrix?

- A matrix is a set of elements, organized into rows and columns.
- A matrix is an array of vectors.



A matrix of order  $r \times c$  means there are  $r$  rows and  $c$  columns.  
Vector is a  $n \times 1$  matrix.

# A Matrix



$$\mathbf{A} = \begin{bmatrix} 1 & 2 & 3 \\ 5 & 4 & 1 \\ 6 & 7 & 4 \end{bmatrix}$$

Square (3 x 3)

$$\mathbf{C} = \begin{bmatrix} 1 & 4 \\ 2 & 7 \\ 3 & 8 \end{bmatrix}$$

Rectangular (3 x 2)

$$\mathbf{D} = \begin{bmatrix} d_{11} & d_{12} & d_{13} \\ d_{21} & d_{22} & d_{23} \\ d_{31} & d_{32} & d_{33} \end{bmatrix}$$

$d_{ij}$  :  $i^{\text{th}}$  row,  $j^{\text{th}}$  column



# B Matrix Operations

$$\begin{bmatrix} a & b \\ c & d \end{bmatrix} + \begin{bmatrix} e & f \\ g & h \end{bmatrix} = \begin{bmatrix} a+e & b+f \\ c+g & d+h \end{bmatrix}$$

Just add elements

Commutative:  **$\mathbf{A+B=B+A}$**

Associative:  **$(\mathbf{A+B})+\mathbf{C=A+(B+C)}$**

$$\begin{bmatrix} a & b \\ c & d \end{bmatrix} - \begin{bmatrix} e & f \\ g & h \end{bmatrix} = \begin{bmatrix} a-e & b-f \\ c-g & d-h \end{bmatrix}$$

Just subtract elements



# B Matrix Operations

- Scalar multiplication

$$\alpha A = \alpha \begin{pmatrix} a & b & c \\ d & e & f \\ g & h & i \end{pmatrix} = \begin{pmatrix} \alpha a & \alpha b & \alpha c \\ \alpha d & \alpha e & \alpha f \\ \alpha g & \alpha h & \alpha i \end{pmatrix}$$

- Transpose

$$A^T = \begin{pmatrix} a & b & c \\ d & e & f \\ g & h & i \end{pmatrix}^T = \begin{pmatrix} a & d & g \\ b & e & h \\ c & f & i \end{pmatrix}$$

(i,j)<sup>th</sup> element of a matrix becomes (j,i)<sup>th</sup> element of its transpose matrix.

A matrix is said to be symmetric matrix if it is equal to its own transpose.



# B Matrix Operations

$$\begin{bmatrix} a & b \\ c & d \end{bmatrix} \begin{bmatrix} e & f \\ g & h \end{bmatrix} = \begin{bmatrix} ae + bg & af + bh \\ ce + dg & cf + dh \end{bmatrix}$$

Multiply each row  
by each column

$$\begin{bmatrix} l_{11} & \textcircled{l_{12}} & l_{13} \\ l_{21} & l_{22} & l_{23} \\ l_{31} & l_{32} & l_{33} \end{bmatrix} = \begin{bmatrix} \cancel{m_{11}} & \cancel{m_{12}} & \cancel{m_{13}} \\ m_{21} & m_{22} & m_{23} \\ m_{31} & m_{32} & m_{33} \end{bmatrix} \cdot \begin{bmatrix} n_{11} & n_{12} & n_{13} \\ n_{21} & n_{22} & n_{23} \\ n_{31} & n_{32} & n_{33} \end{bmatrix}$$

$$l_{12} = m_{11}n_{12} + m_{12}n_{22} + m_{13}n_{32}$$

$$C^{m \times p} = A^{m \times n} \times B^{n \times p} \qquad c_{i,j} = \sum_{k=1}^n a_{i,k} \times b_{k,j}$$



# B Matrix multiplication

- Matrix multiplication is NOT commutative.  
 $AB \neq BA$
- Matrix multiplication is associative.  
 $A(BC) = (AB)C$
- Matrix multiplication is distributive.  
 $A(B+C) = AB+AC$   
 $(A+B)C = AC+BC$



# B Vector Products

Two vectors:

$$\mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} \quad \mathbf{y} = \begin{bmatrix} y_1 \\ y_2 \\ y_3 \end{bmatrix}$$

Inner product = scalar

Inner product  $\mathbf{X}^T \mathbf{Y}$  is a scalar  
 $(1 \times n) (n \times 1) = (1 \times 1)$

$$\mathbf{x}^T \mathbf{y} = \begin{bmatrix} x_1 & x_2 & x_3 \end{bmatrix} \begin{bmatrix} y_1 \\ y_2 \\ y_3 \end{bmatrix} = x_1 y_1 + x_2 y_2 + x_3 y_3 = \sum_{i=1}^3 x_i y_i$$

Outer product = matrix

$$\mathbf{xy}^T = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} \begin{bmatrix} y_1 & y_2 & y_3 \end{bmatrix} = \begin{bmatrix} x_1 y_1 & x_1 y_2 & x_1 y_3 \\ x_2 y_1 & x_2 y_2 & x_2 y_3 \\ x_3 y_1 & x_3 y_2 & x_3 y_3 \end{bmatrix}$$

Outer product  $\mathbf{XY}^T$  is a matrix  
 $(n \times 1) (1 \times n) = (n \times n)$



# Identity matrix



Consider the 3x3 matrix:

$$I_3 = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}$$

For any nxn matrix  $\mathbf{A}$ , we have  $\mathbf{A} I_n = I_n \mathbf{A} = \mathbf{A}$

For any nxm matrix  $\mathbf{A}$ , we have  $I_n \mathbf{A} = \mathbf{A}$ , and  $\mathbf{A} I_m = \mathbf{A}$

Worked example

$$\mathbf{A} I_3 = \mathbf{A}$$

for a 3x3 matrix:

$$\begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{bmatrix} \times \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} = \begin{bmatrix} 1+0+0 & 0+2+0 & 0+0+3 \\ 4+0+0 & 0+5+0 & 0+0+6 \\ 7+0+0 & 0+8+0 & 0+0+9 \end{bmatrix}$$





# C Special types of matrix

- Null matrix

$$\begin{pmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix}$$

- Diagonal matrix

$$\begin{pmatrix} 1 & 0 & 0 \\ 0 & 3 & 0 \\ 0 & 0 & 4 \end{pmatrix}$$

- Lower triangular matrix

$$\begin{pmatrix} 1 & 0 & 0 \\ 2 & 3 & 0 \\ 8 & 7 & 4 \end{pmatrix}$$

- Upper triangular matrix

$$\begin{pmatrix} 1 & 2 & 8 \\ 0 & 3 & 7 \\ 0 & 0 & 4 \end{pmatrix}$$

- Orthogonal matrix

- An orthogonal matrix is a square matrix whose rows and columns are orthogonal and they have unit length.

$$\begin{pmatrix} 0 & -0.8 & -0.6 \\ 0.8 & -0.36 & 0.48 \\ 0.6 & -0.48 & -0.64 \end{pmatrix}$$



# D Matrix inverse

- A square matrix  $A$  is called nonsingular or invertible if there exists a matrix  $B$  such that:

$$AB = BA = I_n \quad \begin{bmatrix} 1 & 1 \\ -1 & 2 \end{bmatrix} \times \begin{bmatrix} \underline{2} & \underline{3} & \underline{-1} \\ & & \underline{3} \end{bmatrix} = \begin{bmatrix} \underline{2} + \underline{1} & \underline{3} & \underline{-1} + \underline{1} \\ & & \underline{3} & \underline{3} \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

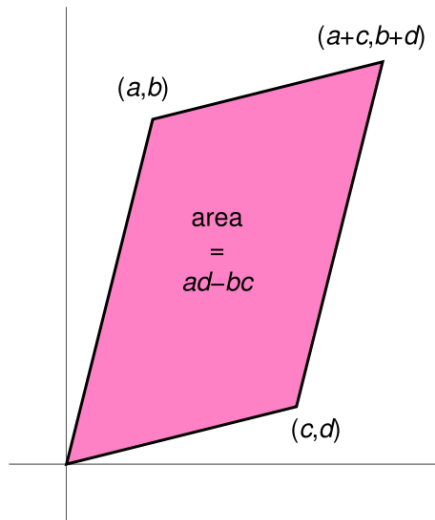
- A common notation for the inverse of a matrix  $A$  is  $A^{-1}$ . So:  $AA^{-1} = A^{-1}A = I_n$
- The inverse matrix is unique when it exists.
- So if  $A$  is invertible, then  $A^T$  is also invertible and then  $(A^T)^{-1} = (A^{-1})^T$



# Determinant



- Given a square matrix  $A$ , its determinant is a real number associated with the matrix.
- The determinant of  $A$  is written as  $\det(A)$  or  $|A|$ .



For a 2x2 matrix, the definition is

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

The area of the parallelogram is the absolute value of the determinant of the matrix formed by the vectors representing the parallelogram's sides.

# D Determinant: 2x2 examples



$$\det \begin{pmatrix} 1 & 1 \\ 3 & 4 \end{pmatrix} = \begin{vmatrix} 1 & 1 \\ 3 & 4 \end{vmatrix} = (1)(4) - (1)(3) = 1$$

$$\det \begin{pmatrix} 1 & 2 \\ 2 & 4 \end{pmatrix} = \begin{vmatrix} 1 & 2 \\ 2 & 4 \end{vmatrix} = (1)(4) - (2)(2) = 0$$



# D Determinant

- To define  $\det(A)$  for larger matrices, we will need the definition of a minor  $M_{ij}$ .
- The minor  $M_{ij}$  of a matrix  $A$  is the matrix formed by removing the  $i^{\text{th}}$  row and the  $j^{\text{th}}$  column of  $A$ .

$$\underline{A} = \begin{pmatrix} 1 & 1 & -2 \\ -1 & 2 & 3 \\ 2 & 7 & 0 \end{pmatrix}$$

$$\underline{A} = \begin{pmatrix} 1 & 1 & -2 \\ -1 & 2 & 3 \\ 2 & 7 & 0 \end{pmatrix}$$

$M_{11}$  : remove row 1, col 1

$$M_{11} = \begin{pmatrix} 2 & 3 \\ 7 & 0 \end{pmatrix}$$

$M_{12}$  : remove row 1, col 2

$$M_{12} = \begin{pmatrix} -1 & 3 \\ 2 & 0 \end{pmatrix}$$



# Determinant



For a matrix

$$A = \begin{pmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{pmatrix}$$

- Its determinant is given by

$$|A| = a_{11}|M_{11}| - a_{12}|M_{12}| + a_{13}|M_{13}|$$

- From the formula for a 2x2 matrix:

$$|M_{11}| = \begin{vmatrix} a_{22} & a_{23} \\ a_{32} & a_{33} \end{vmatrix} = a_{22}a_{33} - a_{23}a_{32}$$



# Determinant



$$\underline{A} = \begin{pmatrix} 1 & 1 & -2 \\ -1 & 2 & 3 \\ 2 & 7 & 0 \end{pmatrix}$$

$$|A| = 1 \times |M_{11}| - 1 \times |M_{12}| + (-2) \times |M_{13}|$$

$$|A| = 1 \times \begin{vmatrix} 2 & 3 \\ 7 & 0 \end{vmatrix} - 1 \times \begin{vmatrix} -1 & 3 \\ 2 & 0 \end{vmatrix} + (-2) \times \begin{vmatrix} -1 & 2 \\ 2 & 7 \end{vmatrix}$$

$$= 1 \times (-21) - 1 \times (-6) + (-2) \times (-11) = 7$$



# D A general formula for determinants

- For a  $n \times n$  matrix  $A=(a_{ij})$  the **co-factors** of  $A$  are defined by

$$C_{ij} = (-1)^{i+j} |M_{ij}|$$

- The determinant of  $A$  is given by the formula

$$|A| = \sum_{i=1}^n a_{ij} C_{ij} \quad \text{for any } j=1,2,\dots,n$$





# Matrix Inverse



$$A^{-1} = \frac{1}{\det(A)} \begin{pmatrix} C_{11} & C_{12} & \cdots & C_{1n} \\ C_{21} & C_{22} & \cdots & C_{2n} \\ \cdots & \cdots & \cdots & \cdots \\ C_{n1} & C_{n2} & \cdots & C_{nn} \end{pmatrix}^T$$

# E Solving simultaneous equations



- For one linear equation  $ax=b$  where the unknown is  $x$  and  $a$  and  $b$  are constants,
- 3 possibilities:

If  $a \neq 0$  then  $x = \frac{b}{a} \equiv a^{-1}b$  thus there is single solution

If  $a = 0, b = 0$  then the equation  $ax = b$  becomes  $0 = 0$  and any value of  $x$  will do

If  $a = 0, b \neq 0$  then  $ax = b$  becomes  $0 = b$  which is a contradiction

# E Solving simultaneous equations



Let's use solution  $x = a^{-1}b$  from the single equation to solve

For example

$$2x_1 + 3x_2 = 5$$

$$x_1 - 2x_2 = -1$$

In matrix notation,

$$\begin{pmatrix} 2 & 3 \\ 1 & -2 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 5 \\ -1 \end{pmatrix}$$

$$A\mathbf{x} = \mathbf{b}$$

$$\mathbf{x} = A^{-1}\mathbf{b}$$

# E Solving simultaneous equations



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

$$x = \frac{1}{\det \begin{pmatrix} 2 & 3 \\ 1 & -2 \end{pmatrix}} \begin{pmatrix} -2 & -1 \\ -3 & 2 \end{pmatrix}^T \begin{pmatrix} 5 \\ -1 \end{pmatrix}$$

$$x = \frac{1}{-7} \begin{pmatrix} -2 & -3 \\ -1 & 2 \end{pmatrix} \begin{pmatrix} 5 \\ -1 \end{pmatrix} = \frac{1}{-7} \begin{pmatrix} -10 + 3 \\ -5 - 2 \end{pmatrix}$$

$$\begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \frac{1}{-7} \begin{pmatrix} -7 \\ -7 \end{pmatrix} = \begin{pmatrix} 1 \\ 1 \end{pmatrix} \Rightarrow \begin{cases} x_1 = 1 \\ x_2 = 1 \end{cases}$$