

# Vectors II

## Basis Sets

Vectors as Matrices

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

$\underline{a}$  and  $\underline{b}$  are vectors but obey the rules of matrices.

Q. Can all vector operations be done by matrices?

★ Addition and scalar multiplication can be.

★ Multiplication?  $\underline{a} = \begin{pmatrix} a_1 \\ a_2 \\ a_3 \end{pmatrix}$   $\underline{b} = \begin{pmatrix} b_1 \\ b_2 \\ b_3 \end{pmatrix}$

if we try  $\underline{a}\underline{b}$ , it is not possible.

But we can do  $\underline{a}^T \underline{b}$  which is just the dot product of  $\underline{a}$  &  $\underline{b}$ .

What if we try  $\underline{a}\underline{b}^T = \begin{pmatrix} a_1 \\ a_2 \\ a_3 \end{pmatrix} (b_1, b_2, b_3)$

$$\underline{a}\underline{b}^T = \begin{pmatrix} a_1 b_1 & a_1 b_2 & a_1 b_3 \\ a_2 b_1 & a_2 b_2 & a_2 b_3 \\ a_3 b_1 & a_3 b_2 & a_3 b_3 \end{pmatrix} \text{ diag. is dot product terms.}$$

It's quite hard to get the cross product terms from  $\underline{a}\underline{b}^T$ .

## Basis Sets

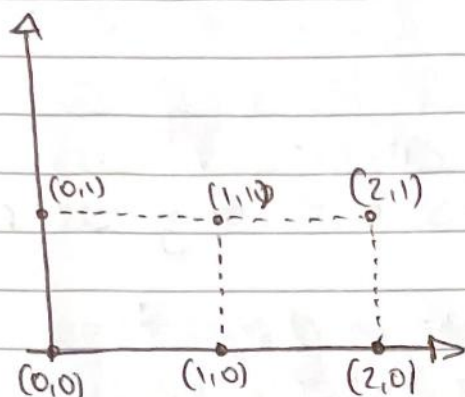
$$A\underline{x} = \underline{b}$$

say we want to operate A on many  $\underline{x}_i$ .

If we first form the matrix

$$X = (\underline{x}_1, \underline{x}_2, \underline{x}_3, \underline{x}_4, \dots)$$

We can then do ~~AX~~  $AX = B$



let  $x_1^T = (0,0)$ ,  $x_2^T = (1,0)$ ,  
 $x_3^T = (0,1)$ ,  $x_4^T = (1,1)$

then  $X = \begin{pmatrix} 0 & 1 & 0 & 1 \\ 0 & 0 & 1 & 1 \end{pmatrix}$

if we let  $A = \begin{pmatrix} 2 & 0 \\ 0 & 1 \end{pmatrix}$

$$AX = \begin{pmatrix} 2 & 0 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} 0 & 1 & 0 & 1 \\ 0 & 0 & 1 & 1 \end{pmatrix} \\ = \begin{pmatrix} 0 & 2 & 0 & 2 \\ 0 & 0 & 1 & 1 \end{pmatrix}$$

$$\underline{A} = \begin{pmatrix} a_{11} \\ a_{21} \end{pmatrix} \quad \underline{A}_2 = \begin{pmatrix} a_{12} \\ a_{22} \end{pmatrix}$$

We can write  $A$  as  $A = (\underline{A}_1, \underline{A}_2)$

So  $A\underline{x} = \underline{b} = \begin{pmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} b_1 \\ b_2 \end{pmatrix}$

$$\Rightarrow \begin{cases} a_{11}x + a_{12}y = b_1 \\ a_{21}x + a_{22}y = b_2 \end{cases} \Rightarrow \underline{A}_1 x + \underline{A}_2 y = \underline{b}$$

N.B. only a solution if  $\det(A) \neq 0$ .

if  $\det(A) = 0$  then  $\underline{A}_1 = \lambda \underline{A}_2$ ,  $\begin{pmatrix} a_{11} \\ a_{21} \end{pmatrix} = \lambda \begin{pmatrix} a_{12} \\ a_{22} \end{pmatrix}$

### Orthormal Basis Sets

\* Non-zero vectors form an orthogonal set if they are orthogonal to each other. i.e. their inner product is zero.

\* In addition, if all vectors are of unit norm  $\|\underline{x}\| = 1$ , then it is called an orthonormal set.



We can use the gram-Schmidt process to produce an orthonormal basis set.

Example if we start with a valid basis in  $\mathbb{R}^3$  of  $\underline{v}_1, \underline{v}_2, \underline{v}_3$

N.B.  $\underline{v}_1' = \underline{v}_1$

$\hat{e}_1 = \frac{\underline{v}_1}{\|\underline{v}_1\|}$   $\leftarrow \hat{e}_1$  has same direction as  $\underline{v}_1$ , but of unit length.

$\underline{v}_2' = \underline{v}_2 - \frac{\underline{v}_2 \cdot \underline{v}_1'}{\|\underline{v}_1'\|} \underline{v}_1'$   $\leftarrow \frac{\underline{v}_2 \cdot \underline{v}_1'}{\|\underline{v}_1'\|} = \text{projection of } \underline{v}_2 \text{ onto } \underline{v}_1$

aka.  $\underline{v}_2' = \underline{v}_2$  - the projection of  $\underline{v}_2$  onto  $\underline{v}_1$  in the direction of  $\underline{v}_1$   $\leftarrow \frac{\underline{v}_1}{\|\underline{v}_1\|} \underline{v}_1 = 1$ , has normalised length of 1, has direction.

$\hat{e}_2 = \frac{\underline{v}_2'}{\|\underline{v}_2'\|}$   $\leftarrow$  transform into a unit vector.

$\underline{v}_3' = \underline{v}_3 - \frac{\underline{v}_3 \cdot \underline{v}_1'}{\|\underline{v}_1'\|} \underline{v}_1' - \frac{\underline{v}_3 \cdot \underline{v}_2'}{\|\underline{v}_2'\|} \underline{v}_2'$

$\uparrow$   
projection of  $\underline{v}_3$  onto  $\underline{v}_1$

$\uparrow$   
projection of  $\underline{v}_3$  onto  $\underline{v}_2$

$\hat{e}_3 = \frac{\underline{v}_3'}{\|\underline{v}_3'\|}$   $\leftarrow$  make unit length.

$\hat{e}_1, \hat{e}_2, \hat{e}_3$  are an orthonormal basis set.  
 $\underline{v}_1', \underline{v}_2', \underline{v}_3'$  are an orthogonal basis set.

Homogeneous Equations

consider

$$A\underline{x} = \underline{0} \leftarrow \text{vector}$$

this is called a homogeneous equation.

$$\begin{pmatrix} a & b & c \\ d & e & f \\ g & h & i \end{pmatrix} = 0$$

from cramer's rule,  $x_i = \frac{\Delta_i}{\Delta} = \frac{0}{\Delta}$

only when  $\Delta = 0$  do we get non-trivial solutions.

We know we can write a matrix as a combination of vectors.

$$\underline{A} = \underline{A}_1 + \underline{A}_2 + \underline{A}_3$$

$$\underline{b} = x_1 \underline{A}_1 + x_2 \underline{A}_2 + x_3 \underline{A}_3$$

but now  $0 = x_1 \underline{A}_1 + x_2 \underline{A}_2 + \dots$

$$\underline{A}_1 = \frac{-1}{x_1} [x_2 \underline{A}_2 + x_3 \underline{A}_3 + \dots]$$

So if  $x_1 \neq 0$ , then  $\underline{A}_1$  is linearly dependent upon  $\underline{A}_2, \underline{A}_3, \dots$

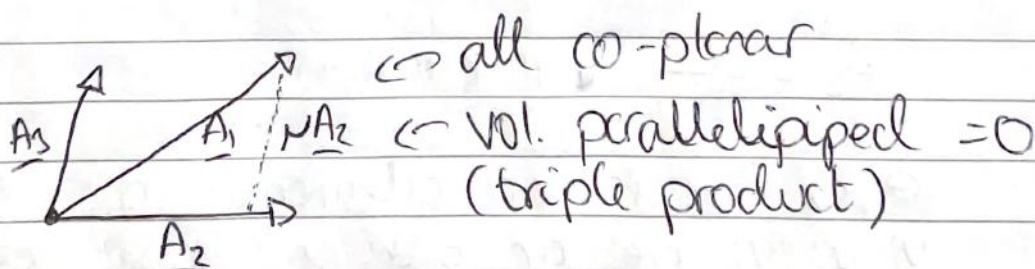
Example

3D vector space.

$$0 = x_1 \underline{A}_1 + x_2 \underline{A}_2 + x_3 \underline{A}_3$$

$$\underline{A}_1 = \lambda \underline{A}_2 + \mu \underline{A}_3$$

$$\det(\underline{A}) = |\underline{A}_1 \underline{A}_2 \underline{A}_3| = |\lambda \underline{A}_2 + \mu \underline{A}_3 \underline{A}_2 \underline{A}_3| = 0$$



Example

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

$$\underline{A} = \begin{pmatrix} 1 & 0 & 1 \\ 1 & 1 & 0 \\ 0 & 1 & -1 \end{pmatrix} \quad \underline{b} = \begin{pmatrix} 0 \\ 1 \\ 1 \end{pmatrix}$$



to solve for  $\underline{x}$ , we will use gaussian elimination.

$$\left(\begin{array}{ccc|c} 1 & 0 & 1 & 0 \\ 1 & 1 & 0 & 1 \\ 0 & 1 & -1 & 1 \end{array}\right) \xRightarrow{R_2 - R_1} \left(\begin{array}{ccc|c} 1 & 0 & 1 & 0 \\ 0 & 1 & -1 & 1 \\ 0 & 1 & -1 & 1 \end{array}\right) \xRightarrow{R_3 - R_2} \left(\begin{array}{ccc|c} 1 & 0 & 1 & 0 \\ 0 & 1 & -1 & 1 \\ 0 & 0 & 0 & 0 \end{array}\right)$$

We cannot proceed any further using gaussian elimination. let's make  $z$  arbitrary.

$$z = \lambda.$$

$$\begin{aligned} x + z &= 0 & \Rightarrow & x = -\lambda \\ y - z &= 1 & & y = \lambda \end{aligned}$$

$$\begin{pmatrix} x \\ y \\ z \end{pmatrix} = \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix} + \lambda \begin{pmatrix} -1 \\ 1 \\ 1 \end{pmatrix}$$

$$\underline{x} = \underline{x}_0 + \lambda \underline{x}_1$$

particular solution

$$A \underline{x}_0 = \left(\begin{array}{ccc|c} 1 & 0 & 1 & 0 \\ 1 & 1 & 0 & 1 \\ 0 & 1 & -1 & 0 \end{array}\right) = \begin{pmatrix} 0 \\ 1 \\ 1 \end{pmatrix} = \underline{b}$$

$$A \underline{x}_1 = \left(\begin{array}{ccc|c} 1 & 0 & 1 & -1 \\ 1 & 1 & 0 & 1 \\ 0 & 1 & -1 & 1 \end{array}\right) = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix} = \underline{0}$$

$\therefore \underline{x}_1$  is general solution to the homogeneous solution.

Example

$$A = \begin{pmatrix} 1 & -2 & 1 \\ -2 & 4 & -2 \\ 3 & -6 & 3 \end{pmatrix} \quad \underline{b} = \begin{pmatrix} 2 \\ -4 \\ 6 \end{pmatrix}$$

$$\left( \begin{array}{ccc|c} 1 & -2 & 1 & 2 \\ -2 & 4 & -2 & -4 \\ 3 & -6 & 3 & 6 \end{array} \right) \Rightarrow \left( \begin{array}{ccc|c} 1 & -2 & 1 & 2 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{array} \right)$$

$R_2/-2, R_3/3$   
 $R_2-R_1, R_3-R_1$

let  $y = \lambda$        $z = \mu$

$$x - 2\lambda + \mu = 2$$

$$x = 2 + 2\lambda - \mu$$

$$\underline{x} = \begin{pmatrix} 2 \\ 0 \\ 0 \end{pmatrix} + \lambda \begin{pmatrix} 2 \\ 1 \\ 0 \end{pmatrix} + \mu \begin{pmatrix} -1 \\ 0 \\ 1 \end{pmatrix} = \begin{pmatrix} x \\ y \\ z \end{pmatrix}$$

$$\underline{x} = \underline{x}_0 + \lambda \underline{x}_1 + \mu \underline{x}_2$$

$$A \underline{x}_0 = \begin{pmatrix} 1 & -2 & 1 \\ -2 & 4 & -2 \\ 3 & -6 & 3 \end{pmatrix} \begin{pmatrix} 2 \\ 0 \\ 0 \end{pmatrix} = \begin{pmatrix} 2 \\ -4 \\ 6 \end{pmatrix} = \underline{b} \quad \text{particular solution}$$

$$A \underline{x}_1 = \begin{pmatrix} 1 & -2 & 1 \\ -2 & 4 & -2 \\ 3 & -6 & 3 \end{pmatrix} \begin{pmatrix} 2 \\ 1 \\ 0 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}$$

$$A \underline{x}_2 = \begin{pmatrix} 1 & -2 & 1 \\ -2 & 4 & -2 \\ 3 & -6 & 3 \end{pmatrix} \begin{pmatrix} -1 \\ 0 \\ 1 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}$$

In general,  $n$  linearly dependent vectors  
 span a space  $\mathbb{R}^{m-n}$  in  $\mathbb{R}^m$ .  
 Span a