

## Examples of Vector Spaces:

JAN. 22 / 18  
APPLIED ANAL.

$\mathbb{R}$ ?

P - the set of polynomials

$$(x^3 + 2x + 1) + (-x^3 + x^2) = x^2 + 2x + 1$$

$P_n$  - the set of polynomials of degree at most  $n$   
(including the zero polynomial)

Let  $V$  be a vector space. The subset  $W$  of  $V$  is a Subspace if it is a vector space using the same addition and scalar multiplication.

- e.g.  $P_n$  is a subspace of  $P$
- e.g.  $P_3$  is a subspace of  $P_5$
- e.g.  $\{\emptyset\}$  and  $V$  are subspaces of  $V$

THM: Let  $W$  be a subset of a vector space  $V$ .

Then  $W$  is a subspace of  $V$  if and only if

(i)  $0 \in W$

(ii) if  $w_1, w_2 \in W$  then  $w_1 + w_2 \in W$  (closure under addition)

(iii) if  $w \in W$ ,  $\lambda \in \mathbb{R}$ , then  $\lambda w \in W$  (closure under scalar mult.)

- e.g. let  $V = \mathbb{R}^3$ ,  $W = \{(a, b, 2a+3b) : a, b \in \mathbb{R}\}$   
is  $W$  a subspace of  $V$ ?

(i) let  $a = b = 0$ . Then  $(0, 0, 0) \in W$

(ii) Take  $(a_1, b_1, 2a_1 + 3b_1), (a_2, b_2, 2a_2 + 3b_2) \in W$   
 $(a_1 + a_2, b_1 + b_2, 2(a_1 + a_2) + 3(b_1 + b_2)) \in W$

(iii) Take  $(a, b, 2a+3b) \in W$ ,  $\lambda \in \mathbb{R}$

$$\lambda(a, b, 2a+3b) = (\lambda a, \lambda b, 2\lambda a + 3\lambda b) \in W$$

It is a Subspace

- e.g. let  $V = \mathbb{R}^2$ ,  $W = \{(a, a^2) : a \in \mathbb{R}\}$

(i)  $(0, 0) \in W$

(ii) Take  $(a_1, a_1^2), (a_2, a_2^2) \in W$

$$(a_1, a_1^2) + (a_2, a_2^2) = (a_1 + a_2, a_1^2 + a_2^2)$$

As  $a_1^2 + a_2^2 \neq (a_1 + a_2)^2$ , this will not be in  $W$  in general

$$(1, 1) \in W, (2, 4) \in W \quad (1, 1) + (2, 4) = (3, 5) \in W$$

$W$  is not closed under addition, so not a subspace

Let  $V$  be a vector space,  $v_1, \dots, v_k \in V$

A vector  $v \in V$  is a linear combination of  $v_1, \dots, v_k$

$$\text{if } v = \lambda_1 v_1 + \lambda_2 v_2 + \dots + \lambda_k v_k \text{ for some } \lambda_1, \dots, \lambda_k \in \mathbb{R}$$

- e.g. in  $\mathbb{R}^2$ ,  $3\langle 1, 2 \rangle - 4\langle 2, 3 \rangle = \langle -5, -6 \rangle$   
 $\langle -5, -6 \rangle$  is a linear combination of  $\langle 1, 2 \rangle$  and  $\langle 2, 3 \rangle$
- e.g. Is  $\langle 1, 2, 3 \rangle$  a linear combination of  $\langle -1, 0, 1 \rangle$  and  $\langle 2, 0, 6 \rangle$ ?  
NO!:  $\lambda_1\langle -1, 0, 1 \rangle + \lambda_2\langle 2, 0, 6 \rangle = \langle -\lambda_1 + 2\lambda_2, 0, \lambda_1 + 6\lambda_2 \rangle$   
 $\neq \langle 1, 2, 3 \rangle$

Let  $V$  be a vector space,  $v_1, \dots, v_k \in V$ . These vectors are linearly dependent if there exist  $\lambda_1, \dots, \lambda_k \in \mathbb{R}$ , not all zero, so that  $\lambda_1v_1 + \lambda_2v_2 + \dots + \lambda_kv_k = \mathbf{0}$ . If not they are linearly independent.

- e.g.  $\vec{i}, \vec{j}, \vec{k}$  in  $\mathbb{R}^3$  are linearly independent

Suppose  $\lambda_1\vec{i} + \lambda_2\vec{j} + \lambda_3\vec{k} = \vec{0}$

$$\langle \lambda_1, \lambda_2, \lambda_3 \rangle = \langle 0, 0, 0 \rangle. \text{ So } \lambda_1 = \lambda_2 = \lambda_3 = 0$$

- e.g. in  $P$ , the vectors  $1, x, x^2, x^3$  are linearly dependent

Suppose  $\lambda_1(1) + \lambda_2x + \lambda_3x^2 + \lambda_4x^3$  is the zero polynomial

$$\text{Then } \lambda_1 = \lambda_2 = \lambda_3 = \lambda_4 = 0$$

- e.g. are  $\langle 1, 0, 2 \rangle, \langle 3, 0, 0 \rangle, \langle 2, -1, 8 \rangle$  linearly independent in  $\mathbb{R}^3$ ?

Suppose  $\lambda_1\langle 1, 0, 2 \rangle + \lambda_2\langle 3, 0, 0 \rangle + \lambda_3\langle 2, -1, 8 \rangle = \langle 0, 0, 0 \rangle$

$$\lambda_1 + 2\lambda_1 + 3\lambda_2 + 2\lambda_3 - \lambda_3 + 8\lambda_3 = \langle 0, 0, 0 \rangle$$

$$\lambda_1 = 0, \lambda_2 = 0, \lambda_3 = 0. \text{ They are linearly independent.}$$

- e.g. in  $\mathbb{R}^2$ ,  $\langle 1, 2 \rangle, \langle 2, 3 \rangle, \langle -5, -6 \rangle$  are linearly dep.

$$3\langle 1, 2 \rangle - 4\langle 2, 3 \rangle = \langle -5, -6 \rangle$$

$$3\langle 1, 2 \rangle - 4\langle 2, 3 \rangle - 1\langle -5, -6 \rangle = \langle 0, 0 \rangle$$

Let  $V$  be a vector space,  $v_1, \dots, v_k \in V$  such that  $v_1, \dots, v_k$  span  $V$  if every vector in  $V$  is a linear combination of  $v_1, \dots, v_k$ .

- e.g.  $\vec{i}, \vec{j}, \vec{k}$  span  $\mathbb{R}^3$

$$\langle a_1, a_2, a_3 \rangle = a_1\vec{i} + a_2\vec{j} + a_3\vec{k}$$

- e.g.  $1, x, x^2, x^3, \dots, x^n$  span  $P_n$

$$a_0 + a_1x + a_2x^2 + a_3x^3 + \dots + a_nx^n = a_01 + a_1x + a_2x^2 + \dots + a_nx^n$$

- e.g. there is no finite set of vectors that spans  $P$

If  $v_1, \dots, v_k \in V$ , then the Span of  $v_1, \dots, v_k$  is the set of all linear combinations of  $v_1, \dots, v_k$ . It is a subspace of  $V$ .

- (i)  $\emptyset = \emptyset v_1 + \emptyset v_2 + \dots + \emptyset v_k$  is in the span
- (ii)  $(\lambda_1 v_1 + \dots + \lambda_k v_k) + (\mu_1 v_1 + \dots + \mu_k v_k)$   
 $= (\lambda_1 + \mu_1) v_1 + \dots + (\lambda_k + \mu_k) v_k$  is in the span.
- (iii)  $\mu(\lambda_1 v_1 + \dots + \lambda_k v_k) = \mu\lambda_1 v_1 + \dots + \mu\lambda_k v_k$  is in the span.

Let  $V$  be a vector space, then  $v_1, \dots, v_k \in V$  are said to be a basis for  $V$  if they are linearly independent and span  $V$ .

- e.g.  $\vec{i}, \vec{j}, \vec{k}$  form a basis for  $\mathbb{R}^3$
- e.g. in  $\mathbb{R}^n$ , let  $e_1 = \langle 1, 0, 0, 0, \dots, 0 \rangle, e_2 = \langle 0, 1, 0, 0, \dots, 0 \rangle, \dots, e_n = \langle 0, 0, 0, 0, \dots, 1 \rangle$ .

Then  $e_1, \dots, e_n$  form the standard basis for  $\mathbb{R}^n$ .

- e.g. in  $P_n$ ,  $1, x, x^2, x^3, \dots, x^n$  is the standard basis
- e.g.  $P$  has no simple basis

If  $V$  is a vector space, then the number of vectors in a basis for  $V$  is fixed, and is called the dimension of  $V$ ,  $\dim V$ . If  $V$  has no simple basis, it is infinite-dimensional.

- e.g.  $\dim \mathbb{R}^n = n$
- e.g.  $\dim P_n = n+1$
- e.g.  $P$  is infinite-dimensional
- e.g.  $W = \{ \langle a, b, 2a+3b \rangle = a, b \in \mathbb{R} \}$

$$\langle a, b, 2a+3b \rangle = a \langle 1, 0, 2 \rangle + b \langle 0, 1, 3 \rangle$$

$$\langle 1, 0, 2 \rangle, \langle 0, 1, 3 \rangle \text{ span } W$$

They are linearly independent and  $\therefore \dim W = 2$

In any vector space  $V$ ,  $v_1, v_2$  are linearly independent provided neither is a scalar multiple of the other.

[ONLY use with vectors]

Suppose  $\lambda_1 v_1 + \lambda_2 v_2 = \emptyset$ . If  $\lambda_1 = \emptyset$ ,

$$\lambda_1 v_1 = -\lambda_2 v_2, \text{ so } v_1 = -\frac{\lambda_2}{\lambda_1} v_2$$

If  $\lambda_2 \neq \emptyset$ ,  $\lambda_2 v_2 = -\lambda_1 v_1$ .

$$v_2 = -\frac{\lambda_1}{\lambda_2} v_1$$

Last time:

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Applied Anal.

- Subspace, subspace test
- linear combination
- linear dependence/independence
- span
- basis
- dimension

Let  $\vec{v}_1, \dots, \vec{v}_n \in \mathbb{R}^n$ . Then  $\vec{v}_1, \dots, \vec{v}_n$  are said to be an orthogonal set of vectors if  $\vec{v}_i \cdot \vec{v}_j = 0$  whenever  $i \neq j$ .

- e.g.  $\langle 1, 1, 0 \rangle, \langle 1, -1, 0 \rangle, \langle 0, 0, 3 \rangle$   
hence that  $\vec{v}_1, \dots, \vec{v}_n \in \mathbb{R}^n$  are an orthogonal set of vectors if they are an orthogonal set and each is a unit vector.

We can normalize an orthogonal set of nonzero vectors to obtain an orthogonal set.

- e.g. above example  $(\frac{1}{\sqrt{2}})\langle 1, 1, 0 \rangle, (\frac{1}{\sqrt{2}})\langle 1, -1, 0 \rangle$   
 $(\frac{1}{\sqrt{3}})\langle 0, 0, 3 \rangle$  is an orthogonal set.

Orthogonal means  $\vec{v}_i \cdot \vec{v}_j = \delta_{ij} = \begin{cases} 1, & i=j \\ 0, & i \neq j \end{cases}$   
"Kronecker delta"

We know that if  $V$  is a subspace of  $\mathbb{R}^n$  and  $\vec{v}_1, \dots, \vec{v}_n$  is a basis for  $V$ , then for any  $\vec{v} \in V$

there exist  $\lambda_1, \dots, \lambda_n \in \mathbb{R}$  so that  $\vec{v} = \lambda_1 \vec{v}_1 + \dots + \lambda_n \vec{v}_n$   
but finding  $\lambda_1, \dots, \lambda_n$  can be a lot of work.

But if  $\vec{v}_1, \dots, \vec{v}_n$  is an orthonormal basis, then

$$\vec{v} = (\vec{v} \cdot \vec{v}_1) \vec{v}_1 + (\vec{v} \cdot \vec{v}_2) \vec{v}_2 + \dots + (\vec{v} \cdot \vec{v}_n) \vec{v}_n$$

Suppose  $\vec{v} = \lambda_1 \vec{v}_1 + \dots + \lambda_n \vec{v}_n$

$$\text{Then } \vec{v} \cdot \vec{v}_i = \lambda_1 \vec{v}_1 \cdot \vec{v}_i + \lambda_2 \vec{v}_2 \cdot \vec{v}_i + \dots + \lambda_n \vec{v}_n \cdot \vec{v}_i = \lambda_i$$

- e.g.  $\vec{v}_1 = \langle \frac{1}{\sqrt{2}}, \frac{1}{\sqrt{2}}, 0 \rangle, \vec{v}_2 = \langle \frac{1}{\sqrt{2}}, -\frac{1}{\sqrt{2}}, 0 \rangle$   
 $\vec{v}_3 = \langle 0, 0, 1 \rangle$

write  $\langle 3, 5, 2 \rangle$  as a linear combination

$$\langle 3, 5, 2 \rangle \cdot \langle \frac{1}{\sqrt{2}}, \frac{1}{\sqrt{2}}, 0 \rangle = \frac{8}{\sqrt{2}}$$

$$\langle 3, 5, 2 \rangle \cdot \langle \frac{1}{\sqrt{2}}, -\frac{1}{\sqrt{2}}, 0 \rangle = -\frac{2}{\sqrt{2}}$$

$$\langle 3, 5, 2 \rangle \cdot \langle 0, 0, 1 \rangle = 2$$

$$\langle 3, 5, 2 \rangle \cdot \frac{8}{\sqrt{2}} \vec{v}_1 - \frac{2}{\sqrt{2}} \vec{v}_2 + 2 \vec{v}_3$$

... (2)

② ... Suppose we have a basis  $\vec{U}_1, \dots, \vec{U}_n$  for a V or  $\mathbb{R}^n$ ?

We use the Gram-Schmidt algorithm to obtain an orthonormal basis. We will obtain an orthonormal basis  $\vec{V}_1, \dots, \vec{V}_n$  for V, then normalize we obtain an orthonormal basis  $\vec{W}_1, \dots, \vec{W}_n$ .

$$\vec{V}_1 = \vec{U}_1$$

$$\vec{V}_2 = \vec{U}_2 - \frac{\vec{U}_2 \cdot \vec{V}_1}{\vec{V}_1 \cdot \vec{V}_1} \vec{V}_1 - \frac{\vec{U}_2 \cdot \vec{V}_2}{\vec{V}_2 \cdot \vec{V}_2} \vec{V}_2$$

$$\vdots$$

$$\vec{V}_k = \vec{U}_k - \frac{\vec{U}_k \cdot \vec{V}_1}{\vec{V}_1 \cdot \vec{V}_1} \vec{V}_1 - \frac{\vec{U}_k \cdot \vec{V}_2}{\vec{V}_2 \cdot \vec{V}_2} \vec{V}_2 - \dots - \frac{\vec{U}_k \cdot \vec{V}_{k-1}}{\vec{V}_{k-1} \cdot \vec{V}_{k-1}} \vec{V}_{k-1}$$

At any stage, we can multiply  $\vec{V}_i$  by a non-zero scalar.

- e.g. V is a subspace of  $\mathbb{R}^3$  with basis  $\vec{U}_1 = \langle 1, 3, 2 \rangle$ ,  $\vec{U}_2 = \langle 3, 2, 0 \rangle$ . Find orthonormal basis for V.

$$\vec{V}_1 = \vec{U}_1 = \langle 1, 3, 2 \rangle$$

$$\vec{V}_2 = \vec{U}_2 - \frac{\vec{U}_2 \cdot \vec{V}_1}{\vec{V}_1 \cdot \vec{V}_1} \vec{V}_1 = \langle 3, 2, 0 \rangle - \left(\frac{9}{14}\right) \langle 1, 3, 2 \rangle$$

Replace  $\vec{V}_2$  with  $14\vec{V}_2$

$$\vec{V}_2 = 14 \langle 3, 2, 0 \rangle - 9 \langle 1, 3, 2 \rangle = \langle 33, 1, -18 \rangle$$

$$\text{Normalize: } \vec{W}_1 = \frac{1}{\sqrt{14}} \langle 1, 3, 2 \rangle, \quad \vec{W}_2 = \frac{1}{\sqrt{1089+1+324}} \langle 33, 1, -18 \rangle$$

- e.g. Let V be a subspace of  $\mathbb{R}^4$  with the series

$$\vec{U}_1 = \langle 1, 1, 0, 0 \rangle, \quad \vec{U}_2 = \langle 2, -1, 0, 0 \rangle, \quad \vec{U}_3 = \langle 1, 1, -1, 0 \rangle$$

Find an orthonormal basis

$$\vec{V}_1 = \vec{U}_1 = \langle 1, 1, 0, 0 \rangle$$

$$\vec{V}_2 = \vec{U}_2 - \frac{\vec{U}_2 \cdot \vec{V}_1}{\vec{V}_1 \cdot \vec{V}_1} \vec{V}_1 = \langle 2, -1, 0, 0 \rangle - \left(\frac{1}{2}\right) \langle 1, 1, 0, 0 \rangle$$

Replace  $\vec{V}_2$  with  $2\vec{V}_2$

$$\vec{V}_2 = 2 \langle 2, -1, 0, 0 \rangle - 1 \langle 1, 1, 0, 0 \rangle = \langle 3, -3, 0, 0 \rangle$$

$$\vec{V}_3 = \vec{U}_3 - \frac{\vec{U}_3 \cdot \vec{V}_1}{\vec{V}_1 \cdot \vec{V}_1} \vec{V}_1 - \frac{\vec{U}_3 \cdot \vec{V}_2}{\vec{V}_2 \cdot \vec{V}_2} \vec{V}_2 = \langle 1, 1, -1, 0 \rangle - \left(\frac{1}{2}\right) \langle 1, 1, 0, 0 \rangle - \dots - \left(\frac{1}{2}\right) \langle 3, -3, 0, 0 \rangle = \langle 0, 0, -1, 0 \rangle$$

$$\text{Normalize } \vec{W}_1 = \frac{1}{\sqrt{2}} \langle 1, 1, 0, 0 \rangle, \quad \vec{W}_2 = \left(\frac{1}{\sqrt{18}}\right) \langle 3, -3, 0, 0 \rangle, \\ \vec{W}_3 = \langle 0, 0, -1, 0 \rangle$$

## Chapter 8 - Matrices (8.1 - 8.10, 8.12, 8.14, 8.15)

Let  $m$  and  $n$  be positive integers. Then an  $m \times n$

matrix is a rectangular array of numbers with  $m$  rows,  $n$  columns

- e.g.  $2 \times 3$  matrix:  $\begin{pmatrix} 1 & 2 & 3 \\ 4 & 12 & 0 \end{pmatrix}$        $3 \times 2$  matrix  $\begin{pmatrix} -2 & 7 \\ 6 & 5 \\ 18 & ? \end{pmatrix}$

We use capital letters to denote matrices.

If  $A$  is our matrix, then we write  $a_{ij}$  for the  $(ij)$ -entry that is, the entry in row  $i$ , column  $j$

- e.g.  $A = \begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{pmatrix}$        $a_{23} = 6$ ,  $a_{31} = 7$

The diagonal entries are  $a_{ii}$

- e.g. above, 1, 5, 9

A Square matrix has the same number of rows as columns

A Column vector is a matrix with 1 column

A Row vector is a matrix with 1 row

- e.g. Column vector:  $\begin{pmatrix} 1 \\ 3 \\ 7 \end{pmatrix}$ , Row vector  $(-2, 1, 0, 4, 7)$

The  $m \times n$  zero matrix has zeroes for all entries.

Denote the matrix  $0$

We write  $A = B$  if  $A$  and  $B$  are both  $m \times n$  matrices and  $a_{ij} = b_{ij}$  for all  $i, j$

- e.g.  $\begin{pmatrix} 1 & 2 \\ 3 & 5 \end{pmatrix} \neq \begin{pmatrix} 5 & 3 \\ 1 & 2 \end{pmatrix}$

Let  $A$  and  $B$  be  $m \times n$  matrices then their sum  $A + B$  is the  $m \times n$  matrix  $C$  so that  $c_{ij} = a_{ij} + b_{ij}$  for all  $i, j$

- e.g.  $\begin{pmatrix} 1 & 2 & 3 \\ 7 & 2 & -1 \end{pmatrix} + \begin{pmatrix} -3 & 4 & 5 \\ 2 & 0 & 4 \end{pmatrix} = \begin{pmatrix} 1+(-3) & 2+4 & 3+5 \\ 7+2 & 0+2 & -1+4 \end{pmatrix} = \begin{pmatrix} -2 & 6 & 8 \\ 9 & 2 & 3 \end{pmatrix}$

If  $A$  and  $B$  do not have the same size their sum is undefined.

If  $A$  is an  $m \times n$  matrix, its negative  $-A$  is found by taking the negative of each entry.

- e.g.  $- \begin{pmatrix} 2 & 1 & -2 \\ 3 & 2 & 6 \end{pmatrix} \Rightarrow \begin{pmatrix} -2 & -1 & 2 \\ -3 & -2 & -6 \end{pmatrix}$

## Properties of Matrix addition :

Let  $A, B, C$  be  $m \times n$  matrices. Then

(i)  $A + B$  is an  $m \times n$  matrix

(ii)  $(A + B) + C = A + (B + C)$  (associativity)

(iii)  $A + B = B + A$  (commutativity)

(iv)  $A + O = A$  (additive identity)

(v)  $A + (-A) = O$  (additive inverse)

→ Find all 3-forms of the line through  $(1, 3, 5)$  and  $(2, 1, 7)$

Vector:  $(x, y, z) = (1, 3, 5) + t(1, -4, 2)$

Parametric:  $x = 1+t$ ,  $y = 3-4t$ ,  $z = 5+2t$

Symmetric:  $\frac{x-1}{1} = \frac{y-3}{-4} = \frac{z-5}{2}$

→ A triangle has vertices  $(1, 2, 0)$ ,  $(2, 1, -1)$ ,  $(4, 3, 7)$

Find the area.

Side vectors:  $\vec{a} = \langle 1, -1, -1 \rangle$

$\vec{b} = \langle 2, 2, 8 \rangle$

$$\vec{a} \times \vec{b} = \begin{vmatrix} \vec{i} & \vec{j} & \vec{k} \\ 1 & -1 & -1 \\ 2 & 2 & 8 \end{vmatrix} \Rightarrow \begin{vmatrix} -1 & -1 \\ 2 & 8 \end{vmatrix} \vec{i} - \begin{vmatrix} 1 & -1 \\ 2 & 8 \end{vmatrix} \vec{j} + \begin{vmatrix} 1 & -1 \\ 2 & 2 \end{vmatrix} \vec{k}$$

$$\Rightarrow \langle -6, -10, 4 \rangle$$

$$\text{Area} = \frac{1}{2} \sqrt{(-6)^2 + (-10)^2 + (4)^2} = \frac{1}{2} \sqrt{150} \text{ish}$$

↪ cross products only exist in  $\mathbb{R}^3$  space.

(the same question with

→ Let  $V$  be the set of all polynomials in the form

$$a + bx + (a+b)x^2, \quad a, b \in \mathbb{R}$$

Is  $V$  a subspace of  $P$ ?

$$\text{Let } a = b = 0 : 0 \in V$$

$$(a_1 + b_1 x + (a_1 + b_1)x^2) + (a_2 + b_2 x + (a_2 + b_2)x^2)$$

$$= (a_1 + a_2) + (b_1 + b_2)x + (a_1 + b_1 + a_2 + b_2)x^2$$

$$= (a_1 + a_2) + (b_1 + b_2)x + ((a_1 + a_2) + (b_1 + b_2))x^2 \in V$$

$$\lambda(a + bx + (ab)x^2)$$

$$= \lambda a + \lambda b x + (\lambda a + \lambda b)x^2 \in V$$

$V$  is a subspace

→ e.g. Let  $V = \{ \langle a, b, \|a+b\| \rangle, a, b \in \mathbb{R}^3 \}$

is  $V$  a subspace of  $\mathbb{R}^3$

Let  $a = b = \emptyset = \langle 0, 0, 0 \rangle \in V$

$$\langle a_1, b_1, \|a_1+b_1\| \rangle + \langle a_2, b_2, \|a_2+b_2\| \rangle$$

$$= \langle a_1 + a_2, b_1 + b_2, \|a_1+b_1\| + \|a_2+b_2\| \rangle$$

$$\langle 1, 1, 2 \rangle + \langle -1, -1, 2 \rangle = \langle 0, 0, 4 \rangle \notin V$$

Not a subspace, not closed under addition

→ Are these polynomials linearly dependent or independent?

$$3 + 4x^2, 2 - 7x + 2x^2, 6x^2$$

$$\lambda_1(3 + 4x^2) + \lambda_2(2 - 7x + 2x^2) + \lambda_3(6x^2) = 0 + 0x + 0x^2$$

$$(3\lambda_1 + 2\lambda_2) - (7\lambda_2)x + (4\lambda_1 + 2\lambda_2 + 6\lambda_3)x^2 = 0 + 0x + 0x^2$$

$$3\lambda_1 + 2\lambda_2 = 0 \quad / \quad -7\lambda_2 = 0 \quad / \quad 4\lambda_1 + 2\lambda_2 + 6\lambda_3 = 0$$

$$\lambda_1 = 0 \quad / \quad \lambda_2 = 0 \quad / \quad \lambda_3 = 0$$

∴ linearly independent

→ Let  $V$  be the subspace of  $\mathbb{R}^3$  with basis

$$\vec{U}_1 = \langle 1, 1, 1 \rangle, \vec{U}_2 = \langle 1, -3, 1 \rangle, \vec{U}_3 = \langle 1, 0, 2 \rangle$$

Find an orthogonal set for  $V$

$$\text{Graham-Schmidt: } \vec{V}_1 = \vec{U}_1 = \langle 1, 1, 1 \rangle$$

$$\vec{V}_2 = \vec{U}_2 - \frac{\vec{U}_2 \cdot \vec{V}_1}{\vec{V}_1 \cdot \vec{V}_1} \vec{V}_1 = \langle 1, -3, 1 \rangle - \left(-\frac{1}{3}\right) \langle 1, 1, 1 \rangle$$

$$\text{Mult by 3: } \vec{V}_2 = 3\langle 1, -3, 1 \rangle + \langle 1, 1, 1 \rangle = \langle 4, -8, 4 \rangle$$

$$\text{Mult by } 1/4: \vec{V}_2 = \langle 1, -2, 1 \rangle$$

$$\vec{V}_3 = \vec{U}_3 - \frac{\vec{U}_3 \cdot \vec{V}_1}{\vec{V}_1 \cdot \vec{V}_1} \vec{V}_1 - \frac{\vec{U}_3 \cdot \vec{V}_2}{\vec{V}_2 \cdot \vec{V}_2} \vec{V}_2$$

$$\Rightarrow \langle 1, 0, 2 \rangle - \left(\frac{2}{3}\right) \langle 1, 1, 1 \rangle - \left(\frac{3}{6}\right) \langle 1, -2, 1 \rangle$$

$$\text{Mult by 2: } \vec{V}_3 = 2 \langle 1, 0, 2 \rangle - 2 \langle 1, 1, 1 \rangle - \langle 1, -2, 1 \rangle \\ : \langle -1, 0, 1 \rangle$$

$$\text{Normalize: } \left(\frac{1}{\sqrt{3}}\right) \langle 1, 1, 1 \rangle, \left(\frac{1}{\sqrt{6}}\right) \langle 1, -2, 1 \rangle, \left(\frac{1}{\sqrt{2}}\right) \langle -1, 0, 1 \rangle$$

- Last time: Orthogonal / Orthonormal set

JAN. 26/18  
APPLIED ANAL.

- Orthonormal set
- Gram-Schmidt
- $M \times N$  matrix
- Square matrix ( $m = n$ )
  - row vector, column vector
- zero vector
- Matrix addition

Let  $A$  be an  $m \times n$  matrix,  $\lambda \in \mathbb{R}$ , then the scalar multiple  $\lambda A$  is the  $m \times n$  matrix  $B$  so that  $b_{ij} = \lambda a_{ij}$  for all  $i, j$

$$\text{e.g. } A = \begin{pmatrix} 1 & 2 \\ 3 & -1 \\ 0 & 6 \end{pmatrix}, \quad \lambda A = \begin{pmatrix} 4 & 8 \\ 12 & -4 \\ 0 & 24 \end{pmatrix}$$

Properties: let  $A$  and  $B$  be  $m \times n$  matrices,  $\lambda, \mu \in \mathbb{R}$

(i)  $\lambda A$  is an  $m \times n$  matrix (closure under scalar mult.)

(ii)  $\lambda(A+B) = \lambda A + \lambda B$  (distributive law)

(iii)  $(\lambda + \mu)A = \lambda A + \mu A$  ("")

(iv)  $\lambda(\mu A) = (\lambda\mu)A$

(v)  $1A = A$

The  $m \times n$  matrices form a vector space

Let  $A = (a_{11}, a_{12}, \dots, a_{1n})$  be a row vector ( $1 \times n$ )

Let  $B = \begin{pmatrix} b_{11} \\ b_{21} \\ \vdots \\ b_{n1} \end{pmatrix}$  be a column vector ( $n \times 1$ )

Then  $AB = (a_{11}b_{11} + a_{12}b_{12} + \dots + a_{1n}b_{1n})$  ( $a 1 \times 1$  matrix)

$$\text{- e.g. } (1, 3, 2) \begin{pmatrix} 4 \\ 5 \\ 3 \end{pmatrix} = (1)(4) + (3)(5) + (2)(3) = 33$$

Let  $A$  be an  $m \times n$  matrix and  $B$  be an  $N \times R$  matrix

Then the product  $AB$  is the  $M \times R$  matrix  $C$  so that

$$c_{ij} = a_{i1}b_{1j} + a_{i2}b_{2j} + \dots + a_{in}b_{nj}$$

$$\text{- e.g. } \begin{pmatrix} 1 & 3 \\ 2 & 5 \\ 4 & -3 \end{pmatrix} \begin{pmatrix} 2 & 1 & 0 \\ 1 & 0 & -3 \end{pmatrix} \Rightarrow \begin{pmatrix} (1)(2) + (3)(1) & (1)(1) + (3)(-3) & (1)(0) + (3)(-3) \\ (2)(2) + (5)(1) & (2)(1) + (5)(-3) & (2)(0) + (5)(-3) \\ (4)(2) + (-3)(1) & (4)(1) + (-3)(-3) & (4)(0) + (-3)(-3) \end{pmatrix}$$

$$\underbrace{\begin{matrix} 2 \times 2 \\ 2 \times 3 \end{matrix}}_{\rightarrow} \underbrace{\begin{pmatrix} 14 & -8 & 6 \\ 24 & -13 & 10 \end{pmatrix}}_{2 \times 3}$$

$$\text{- e.g. } \begin{pmatrix} 1 & -1 & 2 \\ 3 & 1 & 4 \end{pmatrix} \begin{pmatrix} 2 & 4 \\ 0 & 1 \\ 1 & 2 \end{pmatrix} = \begin{pmatrix} 4 & 7 \\ 10 & 21 \end{pmatrix}$$

$$\underbrace{\begin{matrix} 2 \times 3 \\ 3 \times 2 \end{matrix}}_{\rightarrow} \underbrace{\begin{matrix} 2 \times 2 \\ 3 \times 2 \end{matrix}}_{\rightarrow}$$

- Properties : (i)  $(AB)C = A(BC)$  - (associativity)  
(ii)  $A(B+C) = AB+AC$  - (distributive law)  
(iii)  $(B+C)A = BA+CA$  - ("")

Matrix multiplication is not commutative!

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

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

$$AB = \emptyset \neq A = \emptyset \text{ or } B = \emptyset$$

The  $m \times n$  identity matrix  $I_n$  has all of the diagonal entries as "1", everything else is 0

- e.g.  $I_3 = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}$

If  $A$  is an  $m \times n$  matrix then  $I_m A \times A = A I_n$

- Let  $A$  be an  $m \times n$  matrix, then its transpose  $A^T$  is the  $n \times m$  matrix  $B$  so that  $b_{ij} = a_{ji}$  for all  $i, j$   
- e.g.  $A = \begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \end{pmatrix}$ ,  $A^T = \begin{pmatrix} 1 & 4 \\ 2 & 5 \\ 3 & 6 \end{pmatrix}$

Properties : Let  $A, B$  be  $m \times n$  matrices,  $C$  is  $m \times 2 \in \mathbb{R}$

$$(i) (A+B)^T = A^T + B^T$$

$$(ii) (\lambda A)^T = \lambda (A^T)$$

$$(iii) (A^T)^T = A$$

$$(iv) (AC)^T = C^T A^T$$

Let  $A$  be a square matrix. Then  $A$  is the upper triangular matrix if all the entries below are zero.

- e.g.  $\begin{pmatrix} 1 & 2 & 3 \\ 0 & 3 & 5 \\ 0 & 0 & 1 \end{pmatrix}, \begin{pmatrix} 0 & 2 & -2 \\ 0 & 1 & 0 \\ 0 & 0 & 4 \end{pmatrix}$

$A$  is lower triangular matrix if all entries above are zero.

- e.g.  $\begin{pmatrix} 3 & 0 \\ 1 & 4 \end{pmatrix}$

In a diagonal matrix, all entries off the diagonal are zero

- e.g.  $\begin{pmatrix} 1 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 3 \end{pmatrix}$

A System of linear equations has the form

$$\begin{aligned}a_{11}x_1 + a_{12}x_2 + \dots + a_{1n}x_n &= b_1 \\a_{21}x_1 + a_{22}x_2 + \dots + a_{2n}x_n &= b_2 \\&\vdots \\a_{m1}x_1 + a_{m2}x_2 + \dots + a_{mn}x_n &= b_m\end{aligned}$$

This is a system of  $m$  equations in  $n$  unknowns.  
The  $x_i$  are the variables, the  $a_{ij}$  are the coefficients  
and the  $b_i$  are the constants.

- e.g.  $2x_1 + 5x_2 = 0$       2 eqs, 2 unknowns  
 $-7x_1 + 2x_2 = 14$

A solution is a system is an  $n$ -tuple  $(c_1, \dots, c_n)$  so that letting  $x_1 = c_1, x_2 = c_2, \dots, x_n = c_n$ , all equations are satisfied simultaneously.

To solve a system means to find all possible solutions.

The system is inconsistent if there are no solutions.

Otherwise, it is consistent, and there may be one solution or infinitely many.

The following elementary row operations can be performed without changing the solution to the system.

- 1) Multiply an equation by a non-zero constant
- 2) Add a multiple of one equation to another
- 3) Swap row equations

- e.g. solve  $2x_1 + 6x_2 = 8$  ①  
 $4x_1 + 11x_2 = 1$  ②

Multiply ① by  $(-\frac{1}{2})$      $x_1 + 3x_2 = 4$  ①  
 $4x_1 + 11x_2 = 1$  ②

Add  $-4\textcircled{1}$  to ②     $x_1 + 3x_2 = 4$  ①  
 $-x_2 = -15$  ③

Multiply ③ by  $-1$      $x_1 + 3x_2 = 4$  ①  
 $x_2 = 15$  ③

Add  $-3\textcircled{3}$  to ①     $x_1 = -41$  ①  
 $x_2 = 15$  ③

$x_1 = -41, x_2 = 15$  (unique solution)

If our system of equations is:

$$\begin{aligned}a_{11}x_1 + \dots + a_{1n}x_n &= b_1 \\&\vdots \\a_{m1}x_1 + \dots + a_{mn}x_n &= b_m\end{aligned}$$

③ ... then the coefficient matrix is  

$$\begin{pmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ \vdots & \vdots & & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mn} \end{pmatrix}$$

And the augmented matrix is

$$\left( \begin{array}{cccc|c} a_{11} & a_{12} & \dots & a_{1n} & b_1 \\ \vdots & \vdots & & \vdots & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mn} & b_m \end{array} \right)$$

The augmented matrix conveys all the information about the system. we can perform elementary row operations that do not change the solution to the system.

- 1) Multiply a row by a nonzero constant
- 2) Add a multiple of one row to another
- 3) swap two rows

- e.g.  $2x_1 + 6x_2 = 8$  Form the augmented matrix:  
 $4x_1 + 11x_2 = 1$

$$\left( \begin{array}{cc|c} 2 & 6 & 8 \\ 4 & 11 & 1 \end{array} \right) \xrightarrow{\text{mult. } \textcircled{1} \text{ by } \frac{1}{2}} \left( \begin{array}{cc|c} 1 & 3 & 4 \\ 4 & 11 & 1 \end{array} \right) \xrightarrow[\text{to } \textcircled{2}]{\text{add } -4 \textcircled{1}} \left( \begin{array}{cc|c} 1 & 3 & 4 \\ 0 & -1 & -15 \end{array} \right)$$

$$\xrightarrow[\text{by } -1]{\text{mult. } \textcircled{2}} \left( \begin{array}{cc|c} 1 & 3 & 4 \\ 0 & 1 & 15 \end{array} \right) \xrightarrow[\text{to } \textcircled{1}]{\text{add } -3 \textcircled{2}} \left( \begin{array}{cc|c} 1 & 0 & -41 \\ 0 & 1 & 15 \end{array} \right).$$

$$\therefore \begin{aligned} x_1 &= -41 \\ x_2 &= 15 \end{aligned}$$