

**Aim lecture:** We examine the kernel which measures the failure of uniqueness of solns to linear eqns. The concept of linear independence naturally arises. This in turn gives the concept of a basis which allows us to construct co-ordinate systems.

**Recall** For a fn  $f : X \rightarrow Y$  and subset  $Y' \subseteq Y$ , the *inverse image* of  $Y'$  is the set  $f^{-1}(Y') = \{x \in X | f(x) \in Y'\}$ .

## Prop-Defn

Let  $T : V \rightarrow W$  be a linear map.

- 1 For any subspace  $W' \subseteq W$ , the inverse image  $T^{-1}(W')$  is a subspace of  $V$ .
- 2 In particular, the *kernel* of  $T$ , defined to be  $\ker T = T^{-1}(\mathbf{0})$  is a subspace of  $V$ .

**Proof.** Just check closure axioms noting  $T\mathbf{0}_V = \mathbf{0}_W \implies \mathbf{0} \in T^{-1}(W')$  and for  $\mathbf{v}, \mathbf{v}' \in T^{-1}(W'), \beta \in F$ , we know  $\beta\mathbf{v} + \mathbf{v}' \in T^{-1}(W')$  since

$$T(\beta\mathbf{v} + \mathbf{v}') = \beta T\mathbf{v} + T\mathbf{v}' \in W'$$

by closure axioms for  $W'$ .

# Examples

**E.g. 1** We saw last lecture that the set of solns  $V$  to the DE  $\frac{d^2y}{dx^2} + y = 0$  is a subspace of  $C^\infty(\mathbb{R})$  by showing it is the span of some set of vectors. We can also see immediately it is a subspace by recognising it as the kernel of the linear map  $\frac{d^2}{dx^2} + \text{id} : C^\infty(\mathbb{R}) \rightarrow C^\infty(\mathbb{R})$ .

In fact, many subspaces naturally arise or can be described this way.

**E.g. 2** Let  $\mathbf{v} \in \mathbb{R}^n$ . Then the hyperplane  $\mathbf{v}^\perp$  is a subspace of  $\mathbb{R}^n$  since it is the kernel of the following map,

# Kernel measures failure of injectivity

You should already know the following.

## Prop

Let  $T : V \rightarrow W$  be a linear map &  $\mathbf{w} \in W$ .

- 1 Given a particular soln  $\mathbf{v} = \mathbf{v}_p$  to the eqn  $T\mathbf{v} = \mathbf{w}$ , we obtain the complete set of solns as  $T^{-1}(\mathbf{w}) = \{\mathbf{v}_p + \mathbf{v}_h \mid \mathbf{v}_h \in \ker T\}$ .
- 2 In particular, the soln to  $T\mathbf{v} = \mathbf{w}$  is unique (assuming it exists) iff  $\ker T = \mathbf{0}$ .
- 3  $T$  is injective iff  $\ker T = \mathbf{0}$ .

**Proof.** 2)& 3) immediately follow from 1) which we now prove.

Given  $\mathbf{v}_h \in \ker T$  note that

$$T(\mathbf{v}_p + \mathbf{v}_h) = T\mathbf{v}_p + T\mathbf{v}_h = \mathbf{w} + \mathbf{0} = \mathbf{w}$$

so  $\mathbf{v}_p + \mathbf{v}_h \in T^{-1}(\mathbf{w})$ . Hence it remains to show that any soln  $\mathbf{v} \in T^{-1}(\mathbf{w})$  has the form  $\mathbf{v}_p + \mathbf{v}_h$  for some  $\mathbf{v}_h \in \ker T$ .

Now  $\mathbf{v}_h = \mathbf{v} - \mathbf{v}_p \in \ker T$  since

$$T\mathbf{v}_h = T(\mathbf{v} - \mathbf{v}_p) = T\mathbf{v} - T\mathbf{v}_p = \mathbf{w} - \mathbf{w} = \mathbf{0}.$$

Thus  $\mathbf{v} = \mathbf{v}_p + \mathbf{v}_h$  has the desired form.

# Injectivity of linear maps $C : \mathbb{F}^n \longrightarrow V$

We have seen that any linear map  $C : \mathbb{F}^n \longrightarrow V$  determined by the row matrix  $(\mathbf{v}_1 \dots \mathbf{v}_n) \in V^n$  is surjective iff  $V = \text{Span}(\mathbf{v}_1, \dots, \mathbf{v}_n)$ . We determine the condition for injectivity below.

## Prop-Defn

With the above notn, the map  $C$  is injective iff for any  $\beta_1, \dots, \beta_n \in \mathbb{F}$ ,

$$\beta_1 \mathbf{v}_1 + \dots + \beta_n \mathbf{v}_n = \mathbf{0} \implies 0 = \beta_1 = \dots = \beta_n.$$

In other words, the only linear combn of the  $\mathbf{v}_i$ 's which is zero, is the trivial linear combn. In this case, we say that the set  $\{\mathbf{v}_1, \dots, \mathbf{v}_n\}$  is *linearly independent*. Otherwise, we say it is *linearly dependent*.

**Proof.** Clear.

# Ordered bases & co-ordinate systems

We now give a way of constructing co-ordinate systems.

## Defn

Let  $V = \mathbb{F}$ -space. A *basis* for  $V$  is a linearly independent spanning set for  $V$ .

From the results we have already proved, we immediately see

## Prop-Defn

Let  $C : \mathbb{F}^n \rightarrow V$  be the linear map given by the row matrix  $(\mathbf{v}_1 \dots \mathbf{v}_n) \in V^n$ . Then  $C$  defined a co-ordinate system iff  $\{\mathbf{v}_1, \dots, \mathbf{v}_n\}$  is a basis for  $V$ . In other words, co-ordinate systems correspond to *ordered bases* for  $V$ . The *co-ordinates* of  $\mathbf{v} \in V$  wrt  $C$  is  $C^{-1}\mathbf{v} = (x_1, \dots, x_n)^T$ . In this case,

$$\mathbf{v} = x_1\mathbf{v}_1 + \dots + x_n\mathbf{v}_n$$

is the unique way of writing  $\mathbf{v}$  as a linear combn of  $\{\mathbf{v}_1, \dots, \mathbf{v}_n\}$ .

Note that for a co-ord system  $C$  as above, the corresp basis is its sets of entries which are given by  $C\mathbf{e}_j, j = 1, \dots, n$  i.e. it's the image of the standard basis on  $\mathbb{F}^n$ .

# Example of co-ordinate systems

**E.g.** Show that the matrix  $C = \begin{pmatrix} 1+x & 2+3x \end{pmatrix} : \mathbb{C}^2 \rightarrow \mathbb{C}[x]_{\leq 1}$  defines a co-ordinate system on  $\mathbb{C}[x]_{\leq 1}$ . Find the co-ordinates of  $3+4x$  wrt  $C$ .

# Computing bases/ co-ord systems for matrices

Recall from MATH1241/1251 how to find bases for kernels of matrices.

**E.g.** Find a basis and hence co-ord system for the kernel of the matrix

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

# Co-ordinate systems of direct sums.

## Prop

- 1 For  $i = 1, \dots, r$ , let  $T_i : V_i \rightarrow W_i$  be isomorphisms. Then  $\oplus_i T_i : \oplus_i V_i \rightarrow \oplus_i W_i$  is also an isomorphism.
- 2 In particular, given co-ordinate systems  $C_i : \mathbb{F}^{n_i} \rightarrow W_i$ , we obtain a co-ordinate system  $\oplus_i C_i : \mathbb{F}^n \rightarrow \oplus_i W_i$  where  $n = \sum_i n_i$ .
- 3 Consider an internal direct sum  $W = \oplus_i W_i$  & bases  $B_i \subset W_i$  for  $W_i$ . Then  $\cup_i B_i$  is a basis for  $W$ .

**Proof.** 1) We need only show  $\oplus_i T_i^{-1}$  is the inverse for  $\oplus_i T_i$ . For ease of writing we do this when  $r = 2$ .

$$(T_1 \oplus T_2) \circ (T_1^{-1} \oplus T_2^{-1}) = \begin{pmatrix} T_1 & 0 \\ 0 & T_2 \end{pmatrix} \begin{pmatrix} T_1^{-1} & 0 \\ 0 & T_2^{-1} \end{pmatrix} =$$

Sim,  $(T_1^{-1} \oplus T_2^{-1}) \circ (T_1 \oplus T_2) = \text{id}$  so 1) holds.

It's clear 1)  $\implies$  2). Also 3) follows from 2) as the basis corresp to  $(\oplus C_i)$  consist of the  $(\oplus C_i)\mathbf{e}_j, j = 1, \dots, n$ .



# A matrix example

**E.g.** We have seen already that  $V = M_{22}(\mathbb{R})$  is the internal direct sum of the subspaces  $V^+$ ,  $V^-$  of symmetric & anti-symmetric matrices resp.

Now any symmetric (resp anti-symmetric) matrix can be written uniquely in the form

$$\begin{pmatrix} a & b \\ b & c \end{pmatrix}, \quad \text{resp} \begin{pmatrix} 0 & d \\ -d & 0 \end{pmatrix}$$

so we obtain the following bases for  $V^+$ ,  $V^-$ ,  $V$ .

# Alternate characterisation of linear dependence

## Prop

A finite set  $S = \{\mathbf{v}_1, \dots, \mathbf{v}_n\}$  of vectors is linearly dependent iff we can write one of them as a linear combn of the others.

**Proof.** Suppose that  $S$  is lin dependent so there is a non-trivial linear relation say  $\beta_1\mathbf{v}_1 + \dots + \beta_n\mathbf{v}_n = \mathbf{0}$  with say  $\beta_i \neq 0$ . Then  $\mathbf{v}_i$  is a linear combn of the others as we may re-write

$$\mathbf{v}_i = - \sum_{j \neq i} \beta_i^{-1} \beta_j \mathbf{v}_j.$$

Reversing the above computation gives the converse.

# Alternate characterisation of bases

## Theorem

Let  $V = \mathbb{F}$ -space &  $B \subseteq V$  be finite. The following are equiv conds on  $B$ .

- 1)  $B$  is a basis for  $V$ .
- 2)  $B$  is a minimal spanning set for  $V$ .
- 3)  $B$  is a maximal linearly independent set in the sense that,  $B$  is linearly independent but  $B'$  is linearly dependent for any set  $B'$  strictly containing  $B$ .

**Proof.** The equivalence of 1) & 2) follows from the alternate characterisation of linear dependence. The proof of 1)  $\iff$  3) follows easily (ex) from

## Lemma

Suppose  $B = \{\mathbf{v}_1, \dots, \mathbf{v}_n\}$  is lin independent & let  $\mathbf{v} \in V$ . Then  $B \cup \{\mathbf{v}\}$  is lin indep iff  $\mathbf{v} \notin \text{Span}(B)$ .

## proof continued

To see ( $\implies$ ), suppose  $B \cup \{\mathbf{v}\}$  is lin indep but  $\mathbf{v} \in \text{Span}(B)$ . Then  $B \cup \{\mathbf{v}\}$  must be linearly dependent by our alternate charn of linear dependence and we obtain a contradiction. Hence  $\mathbf{v} \notin \text{Span}(B)$ .

For ( $\impliedby$ ), suppose that  $\mathbf{v} \notin \text{Span}(B)$  but  $B \cup \{\mathbf{v}\}$  is lin dependent. Hence there is a non-trivial linear relation of form

$$\beta \mathbf{v} + \beta_1 \mathbf{v}_1 + \dots + \beta_n \mathbf{v}_n = \mathbf{0}.$$

Now  $\beta \neq 0$  as  $B$  is linearly independent. Hence we may solve to see  $\mathbf{v}$  is a linear combn of the  $\mathbf{v}_i$  so lies in  $\text{Span}(B)$ . Again we obtain a contradiction so  $B \cup \{\mathbf{v}\}$  is lin indep.

### Corollary

*Any vector space that is spanned by a finite set  $S$  has a basis consisting of a subset of  $S$ .*