

# Math 2270-004 Week 8 notes

We will not necessarily finish the material from a given day's notes on that day. We may also add or subtract some material as the week progresses, but these notes represent an in-depth outline of what we plan to cover. These notes cover material in 4.2-4.6.

Mon Feb 26

• 4.2 - 4.3 bases for vector spaces and subspaces;  $Nul A$  and  $Col A$ ; generalization to linear transformations.

Announcements: I'll post fft soltns tonight

'til for  $\mathbb{R}^n$  our favorite basis is  $\{\vec{e}_1, \vec{e}_2, \dots, \vec{e}_n\}$ . But for a subspace of  $\mathbb{R}^n$ , what's a "best" basis?  
no row operations required

Warm-up Exercise:

a) check that  $\left\{ \begin{bmatrix} 2 \\ 1 \\ 0 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 0 \\ -1 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} -1 \\ 0 \\ -1 \\ 0 \\ 1 \end{bmatrix} \right\}$  is linearly independent.  
Which entries do you focus on?

b) express  $\begin{bmatrix} 8 \\ 2 \\ -2 \\ 3 \\ -1 \end{bmatrix}$  as a linear combination of the vectors above.  
Which entries do you focus on?

$$a) c_1 \begin{bmatrix} 2 \\ 1 \\ 0 \\ 0 \\ 0 \end{bmatrix} + c_2 \begin{bmatrix} 1 \\ 0 \\ -1 \\ 1 \\ 0 \end{bmatrix} + c_3 \begin{bmatrix} -1 \\ 0 \\ -1 \\ 0 \\ 1 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix} \begin{matrix} \leftarrow 2^{nd} \\ \leftarrow 4^{th} \\ \leftarrow 5^{th} \end{matrix}$$

$$b) \begin{matrix} 2 & & 3 \\ || & & || \\ c_1 & \begin{bmatrix} 2 \\ 1 \\ 0 \\ 0 \\ 0 \end{bmatrix} & + c_2 & \begin{bmatrix} 1 \\ 0 \\ -1 \\ 1 \\ 0 \end{bmatrix} & + c_3 & \begin{bmatrix} -1 \\ 0 \\ -1 \\ 0 \\ 1 \end{bmatrix} & = & \begin{bmatrix} 8 \\ 2 \\ -2 \\ 3 \\ -1 \end{bmatrix} \end{matrix}$$

2<sup>nd</sup> entries  
4<sup>th</sup>  
5<sup>th</sup>

$$\begin{matrix} c_1 = 0 \\ c_2 = 0 \\ c_3 = 0 \end{matrix}$$

almost as good  
as standard  
basis vectors.

$$\begin{matrix} 2^{nd} \text{ entry} & c_1 = 2 \\ 4^{th} \text{ entry} & c_2 = 3 \\ 5^{th} & c_3 = -1 \end{matrix}$$

$$2 \begin{bmatrix} 2 \\ 1 \\ 0 \\ 0 \\ 0 \end{bmatrix} + 3 \begin{bmatrix} 1 \\ 0 \\ -1 \\ 1 \\ 0 \end{bmatrix} - \begin{bmatrix} -1 \\ 0 \\ -1 \\ 0 \\ 1 \end{bmatrix} = \begin{bmatrix} 8 \\ 2 \\ -2 \\ 3 \\ -1 \end{bmatrix} \checkmark$$

## Monday Review!

We've been discussing vector spaces, which are a generalization of  $\mathbb{R}^n$ : Namely, a *vector space* is a nonempty set  $V$  of objects, called *vectors*, on which are defined two operations, called *addition* and *scalar multiplication*, so that ten natural axioms about vector addition and scalar multiplication hold (along with three additional useful consequences that we often use, and that you thought about on your food for thought).

Last week we discovered that certain subsets of vector spaces are also vector spaces (with the same addition and scalar multiplication as in the larger space) - namely subspaces of a vector space  $V$ : these are subsets  $H$  of  $V$  that satisfy

*sub vector spaces.*

- a) The zero vector of  $V$  is in  $H$
- b)  $H$  is closed under vector addition, i.e. for each  $\underline{u} \in H, \underline{v} \in H$  then  $\underline{u} + \underline{v} \in H$ .
- c)  $H$  is closed under scalar multiplication, i.e. for each  $\underline{u} \in H, c \in \mathbb{R}$ , then also  $c\underline{u} \in H$ .

We defined linear dependence and linear independence for sets of vectors  $\{\underline{v}_1, \underline{v}_2, \dots, \underline{v}_p\}$  in a vector space  $V$ .

A basis for a vector space  $V$  is a set of vectors  $\{\underline{v}_1, \underline{v}_2, \dots, \underline{v}_p\}$  that span  $V$  and that is also linearly independent.

The *dimension* of a vector space  $V$  is the number of vectors in any basis for  $V$ . (We'll show why every basis for a fixed vector space  $V$  - no matter how weird  $V$  may seem - has the same number of vectors, later this week.)

We showed that one way subspaces arise is as  $H = \text{span}\{\underline{v}_1, \underline{v}_2, \dots, \underline{v}_p\}$  for sets of vectors  $\{\underline{v}_1, \underline{v}_2, \dots, \underline{v}_p\}$  in a vector space  $V$ . This is an explicit way to describe  $H$  because you are saying exactly which vectors are in it. If the vectors in the spanning set  $\{\underline{v}_1, \underline{v}_2, \dots, \underline{v}_p\}$  are not already independent, we illustrated how to remove extraneous dependent vectors without shrinking the span, until we were left with a basis for the subspace  $H$ . (We'll return to this today .... it was an example with  $H = \text{Col } A$ )

We discovered that the only subsets of  $\mathbb{R}^3$  that succeed at being subspaces of  $\mathbb{R}^3$  are

- $\{\underline{0}\}$   $0$ -dim'l, by def.
- $\text{span}\{\underline{u}\}$  for some  $\underline{u} \neq \underline{0}$  (a line thru the origin) 1 - dimensional subspaces
- $\text{span}\{\underline{u}, \underline{v}\}$  for some  $\{\underline{u}, \underline{v}\}$  linearly independent 2 - dimensional subspaces
- $\text{span}\{\underline{u}, \underline{v}, \underline{w}\} = \mathbb{R}^3$  for  $\{\underline{u}, \underline{v}, \underline{w}\}$  linearly independent 3 - dimensional (sub)space.

- We realized that what happens in  $\mathbb{R}^3$  with respect to subspaces, generalizes to  $\mathbb{R}^n$ .

Towards the end of class on Friday we realized that for an  $m \times n$  matrix  $A$ ,

$$\text{Nul } A := \{\underline{x} \in \mathbb{R}^n \text{ for which } A\underline{x} = \underline{0}\}$$

is a subspace. This is an implicit way to specify a subspace, because you're prescribing equations which the elements  $\underline{x}$  must satisfy, but not explicitly saying what the elements are.

Picking up where we left off ....

Note  
 $\mathbb{R}^2 = \left\{ \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} : x_1, x_2 \in \mathbb{R} \right\}$   
 a 2-dim'l  
 plane in  $\mathbb{R}^3$ ,  
 thru  $\tilde{0}$ , i.e.  
 $\text{span}\{\underline{u}, \underline{v}\}$   
 $\underline{u}, \underline{v}$  ind.  
 is not  $\mathbb{R}^2$

Monday Feb. 26  
finish from Friday

Nul A

Exercise 1a) For the same matrix  $A$  as in Exercise 2 from Wednesday's notes, express the vectors in  $\text{Nul}(A)$  explicitly, using the methods of Chapters 1-2. Notice these are vectors in the domain of the associated linear transformation  $T: \mathbb{R}^5 \rightarrow \mathbb{R}^3$  given by  $T(\underline{x}) = A\underline{x}$ , so are a subspace of  $\mathbb{R}^5$ .

$$A\underline{x} = \underline{0}$$

$$A = \begin{bmatrix} 1 & -2 & 0 & -1 & 1 \\ -2 & 4 & 0 & 2 & -2 \\ 3 & -6 & 4 & 1 & 7 \end{bmatrix} \begin{matrix} 0 \\ 0 \\ 0 \end{matrix} \text{ reduces to } \begin{bmatrix} 1 & -2 & 0 & -1 & 1 \\ 0 & 0 & 1 & 1 & 1 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix} \begin{matrix} 0 \\ 0 \\ 0 \end{matrix} \leftarrow$$

$$\{\underline{x} : A\underline{x} = \underline{0}\}$$

implicit description

$$x_1 = 2t_2 + t_4 - t_5$$

$$x_2 = t_2 \text{ free}$$

$$x_3 = -t_4 - t_5$$

$$x_4 = t_4 \text{ free}$$

$$x_5 = t_5 \text{ free}$$

explicit  
description

$$\begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \end{bmatrix} = t_2 \begin{bmatrix} 2 \\ 1 \\ 0 \\ 0 \\ 0 \end{bmatrix} + t_4 \begin{bmatrix} 1 \\ 0 \\ -1 \\ 1 \\ 0 \end{bmatrix} + t_5 \begin{bmatrix} -1 \\ 0 \\ -1 \\ 0 \\ 1 \end{bmatrix}$$

1b) Exhibit a basis for  $\text{Nul}(A)$ .

free params came from  
non-pivot columns, e.g.  
 $x_2 = t_2$  free.  
 $x_4 = t_4$  free  
 $x_5 = t_5$  free

as consequence,

2<sup>nd</sup> entry of dep.  
eqn says  $c_1 = 0$

4<sup>th</sup> entry says  
 $c_2 = 0$

5<sup>th</sup> entry says  
 $c_3 = 0$

where  
we  
ended.  
needed  
to check

actually  
a basis

independence. that was warmup prob.

$$\text{if } c_1 \begin{bmatrix} 2 \\ 1 \\ 0 \\ 0 \\ 0 \end{bmatrix} + c_2 \begin{bmatrix} 1 \\ 0 \\ -1 \\ 1 \\ 0 \end{bmatrix} + c_3 \begin{bmatrix} -1 \\ 0 \\ -1 \\ 0 \\ 1 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}$$

OR

$$t_2 \begin{bmatrix} 2 \\ 1 \\ 0 \\ 0 \\ 0 \end{bmatrix} + t_4 \begin{bmatrix} 1 \\ 0 \\ -1 \\ 1 \\ 0 \end{bmatrix} + t_5 \begin{bmatrix} -1 \\ 0 \\ -1 \\ 0 \\ 1 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}$$

$$\text{Nul } A = \text{span} \left\{ \begin{bmatrix} 2 \\ 1 \\ 0 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 0 \\ -1 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} -1 \\ 0 \\ -1 \\ 0 \\ 1 \end{bmatrix} \right\}$$



Monday Feb 26  
Review

from Wed. notes, covered Friday.

subspace as  
a span

4.2 Null spaces, column spaces, and linear transformations from  $\mathbb{R}^n$  to  $\mathbb{R}^m$ .

**Definition** Let  $A$  be an  $m \times n$  matrix, expressed in column form as  $A = [\mathbf{a}_1 \ \mathbf{a}_2 \ \mathbf{a}_3 \ \dots \ \mathbf{a}_n]$ . The *column space* of  $A$ , written as  $\text{Col } A$ , is the span of the columns:

$$\text{Col } A = \text{span}\{\mathbf{a}_1 \ \mathbf{a}_2 \ \mathbf{a}_3 \ \dots \ \mathbf{a}_n\}.$$

Equivalently, since

$$A\mathbf{x} = x_1\mathbf{a}_1 + x_2\mathbf{a}_2 + \dots + x_n\mathbf{a}_n$$

we see that  $\text{Col } A$  is also the range of the linear transformation  $T: \mathbb{R}^n \rightarrow \mathbb{R}^m$  given by  $T(\mathbf{x}) = A\mathbf{x}$ , i.e.

$$\text{Col } A = \{\mathbf{b} \in \mathbb{R}^m \text{ such that } \mathbf{b} = A\mathbf{x} \text{ for some } \mathbf{x} \in \mathbb{R}^n\}.$$

**Theorem** By the "spans are subspaces" theorem,  $\text{Col}(A)$  is always a subspace of  $\mathbb{R}^m$ .

**Exercise 2a)** Consider

$$A = \begin{bmatrix} 1 & -2 & 0 & -1 & 1 \\ -2 & 4 & 0 & 2 & -2 \\ 3 & -6 & 4 & 1 & 7 \end{bmatrix}.$$

By the Theorem,  $\text{col}(A)$  is a subspace of  $\mathbb{R}^3$ . Which is it:  $\{\mathbf{0}\}$ , a line thru the origin, a plane thru the origin, or all of  $\mathbb{R}^3$ . Hint:

$$\begin{array}{c} \vec{a}_1 \quad \vec{a}_2 \quad \vec{a}_3 \quad \vec{a}_4 \quad \vec{a}_5 \\ \begin{bmatrix} 1 & -2 & 0 & -1 & 1 \\ -2 & 4 & 0 & 2 & -2 \\ 3 & -6 & 4 & 1 & 7 \end{bmatrix} \end{array} \quad \text{reduces to} \quad \begin{array}{c} \vec{r}_1 \quad \vec{r}_2 \quad \vec{r}_3 \quad \vec{r}_4 \quad \vec{r}_5 \\ \begin{bmatrix} 1 & -2 & 0 & -1 & 1 \\ 0 & 0 & 1 & 1 & 1 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix} \end{array}$$

$$\begin{array}{l} \vec{a}_1 \neq \vec{0} \\ \vec{a}_2 = -2\vec{a}_1 \\ \vec{a}_3 \text{ ind. of } \vec{a}_1 \\ \vec{a}_4 = -\vec{a}_1 + \vec{a}_3 \\ \vec{a}_5 = \vec{a}_1 + \vec{a}_3 \end{array}$$

$$\begin{array}{l} \vec{r}_1 \neq \vec{0} \\ \vec{r}_2 = -2\vec{r}_1 \\ \vec{r}_3 \text{ ind. of } \vec{r}_1 \\ \vec{r}_4 = -\vec{r}_1 + \vec{r}_3 \\ \vec{r}_5 = \vec{r}_1 + \vec{r}_3 \end{array}$$

**2b)** Is there a more efficient way to express  $\text{Col } A$  as a span that doesn't require all five column vectors?

$$\begin{aligned} & c_1\vec{a}_1 + c_2\vec{a}_2 + c_3\vec{a}_3 + c_4\vec{a}_4 + c_5\vec{a}_5 \\ &= c_1\vec{a}_1 + c_2(-2\vec{a}_1) + c_3\vec{a}_3 + c_4(-\vec{a}_1 + \vec{a}_3) + c_5(\vec{a}_1 + \vec{a}_3) \\ &= d_1\vec{a}_1 + d_3\vec{a}_3 \end{aligned}$$

$$\begin{aligned} \text{Col } A &= \text{span}\{\vec{a}_1, \vec{a}_2, \dots, \vec{a}_5\} \\ &= \text{span}\{\vec{a}_1, \vec{a}_3\} \end{aligned}$$

YES.

so,  
 $\text{col } A$  was  
just a plane  
thru origin

Not all bases are created equal!

Theorem: Let  $\text{span}\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_p\} = H$  be a subspace. The following *elementary operations* do not effect the span of the resulting ordered set:

(i) swap two of the vectors in the set, i.e. replace  $\mathbf{v}_j$  with  $\mathbf{v}_k$ , and replace  $\mathbf{v}_k$  with  $\mathbf{v}_j$ .

$$\text{span}\{\tilde{\mathbf{v}}_1, \dots, \tilde{\mathbf{v}}_j, \dots, \tilde{\mathbf{v}}_k, \dots, \tilde{\mathbf{v}}_p\} = \text{span}\{\tilde{\mathbf{v}}_1, \dots, \tilde{\mathbf{v}}_k, \dots, \tilde{\mathbf{v}}_j, \dots, \tilde{\mathbf{v}}_p\}$$

(ii) replace  $\mathbf{v}_j$  with  $c\mathbf{v}_j$ , for  $c \neq 0$ .

$$\text{span}\{\tilde{\mathbf{v}}_1, \dots, \tilde{\mathbf{v}}_j, \dots, \tilde{\mathbf{v}}_p\} = \text{span}\{\tilde{\mathbf{v}}_1, \dots, c\tilde{\mathbf{v}}_j, \dots, \tilde{\mathbf{v}}_p\}$$

$$c_1\tilde{\mathbf{v}}_1 + c_2\tilde{\mathbf{v}}_2 + \dots + c_j\tilde{\mathbf{v}}_j + \dots + c_p\tilde{\mathbf{v}}_p = c_1\tilde{\mathbf{v}}_1 + c_2\tilde{\mathbf{v}}_2 + \dots + \frac{c_j}{c}c\tilde{\mathbf{v}}_j + \dots + c_p\tilde{\mathbf{v}}_p$$

(iii) for  $j \neq k$ , replace  $\mathbf{v}_k$  with  $\mathbf{v}_k + c\mathbf{v}_j$ .

$$\text{span}\{\tilde{\mathbf{v}}_1, \dots, \tilde{\mathbf{v}}_j, \dots, \tilde{\mathbf{v}}_k, \dots, \tilde{\mathbf{v}}_p\} = \text{span}\{\tilde{\mathbf{v}}_1, \dots, \tilde{\mathbf{v}}_j, \dots, (\tilde{\mathbf{v}}_k + c\tilde{\mathbf{v}}_j), \dots, \tilde{\mathbf{v}}_p\}$$

$$c_1\tilde{\mathbf{v}}_1 + c_2\tilde{\mathbf{v}}_2 + \dots + c_k\tilde{\mathbf{v}}_k + \dots + c_p\tilde{\mathbf{v}}_p = c_1\tilde{\mathbf{v}}_1 + \dots + (c_j + c_k)c\tilde{\mathbf{v}}_j + \dots + c_k(\tilde{\mathbf{v}}_k + c\tilde{\mathbf{v}}_j) + \dots + c_p\tilde{\mathbf{v}}_p$$

Exercise 1) Use the "change of spanning set" theorem above, to find a better basis for  $\text{Col } A$  than the one we came up with by culling dependent vectors, on Friday. Hint: Use elementary column operations to compute the reduced column echelon form of  $A$ . Illustrate why this new basis is a better basis for  $\text{Col } A$  by seeing how easy it is to express any one of the original column vectors in terms of this improved basis.

In this example,  $A = [\mathbf{a}_1 \ \mathbf{a}_2 \ \mathbf{a}_3 \ \mathbf{a}_4 \ \mathbf{a}_5]$  and  $\text{Col } A = \text{span}\{\mathbf{a}_1, \mathbf{a}_2, \mathbf{a}_3, \mathbf{a}_4, \mathbf{a}_5\} = \text{span}\{\tilde{\mathbf{a}}_1, \tilde{\mathbf{a}}_3\}$ .

$$A = \begin{bmatrix} 1 & -2 & 0 & -1 & 1 \\ -2 & 4 & 0 & 2 & -2 \\ 3 & -6 & 4 & 1 & 7 \end{bmatrix}$$

As we just reviewed, on Friday we realized that a pretty good basis for  $\text{Col } A$  is  $\{\mathbf{a}_1, \mathbf{a}_3\}$ .

$$\begin{bmatrix} 1 & -2 & 0 & -1 & 1 \\ -2 & 4 & 0 & 2 & -2 \\ 3 & -6 & 4 & 1 & 7 \end{bmatrix} \quad \text{row reduces to} \quad \begin{bmatrix} 1 & -2 & 0 & -1 & 1 \\ 0 & 0 & 1 & 1 & 1 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix}$$

Now column reduce  $A$  to get a basis for  $\text{Col } A$  that's as good as you could hope for....and show this by expressing each of the original columns in terms of this basis.

$$A = \begin{bmatrix} \vec{a}_1 & \vec{a}_2 & \vec{a}_3 & \vec{a}_4 & \vec{a}_5 \\ 1 & -2 & 0 & -1 & 1 \\ -2 & 4 & 0 & 2 & -2 \\ 3 & -6 & 4 & 1 & 7 \end{bmatrix}$$

better  
Col A basis

rref!

$$\begin{aligned} 2\vec{a}_1 + \vec{a}_2 &\rightarrow \vec{a}_2 \\ \vec{a}_1 + \vec{a}_4 &\rightarrow \vec{a}_4 \\ -\vec{a}_1 + \vec{a}_5 &\rightarrow \vec{a}_5 \end{aligned}$$

$$\begin{bmatrix} 1 & 0 & 0 & 0 & 0 \\ -2 & 0 & 0 & 0 & 0 \\ 3 & 0 & 4 & 4 & 4 \end{bmatrix}$$

$$\begin{aligned} a_2 &\rightarrow a_5 \\ a_5 &\rightarrow a_2 \end{aligned}$$

$$\begin{bmatrix} 1 & 0 & 0 & 0 & 0 \\ -2 & 0 & 0 & 0 & 0 \\ 3 & 4 & 4 & 4 & 0 \end{bmatrix}$$

$$\frac{1}{4}a_2 \rightarrow a_2$$

$$\begin{bmatrix} 1 & 0 & 0 & 0 & 0 \\ -2 & 0 & 0 & 0 & 0 \\ 3 & 1 & 4 & 4 & 0 \end{bmatrix}$$

$$\begin{aligned} -4\vec{a}_2 + \vec{a}_3 &\rightarrow \vec{a}_3 \\ -4\vec{a}_2 + \vec{a}_4 &\rightarrow \vec{a}_4 \end{aligned}$$

$$\begin{bmatrix} 1 & 0 & 0 & 0 & 0 \\ -2 & 0 & 0 & 0 & 0 \\ 3 & 1 & 0 & 0 & 0 \end{bmatrix}$$

$$-3\vec{a}_2 + \vec{a}_1 \rightarrow \vec{a}_1$$

$$\begin{bmatrix} 1 & 0 & 0 & 0 & 0 \\ -2 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 \end{bmatrix}$$

$$= \left\{ \begin{bmatrix} 1 \\ -2 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} \right\}$$

"best" basis for Col A

Tuesday:

$$\vec{a}_5 = \begin{bmatrix} 1 \\ -2 \\ 7 \end{bmatrix} = 1 \begin{bmatrix} 1 \\ -2 \\ 0 \end{bmatrix} + 7 \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$$

$$\vec{a}_4 = \begin{bmatrix} -1 \\ 2 \\ 1 \end{bmatrix} = -1 \begin{bmatrix} 1 \\ -2 \\ 0 \end{bmatrix} + 1 \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$$

general linear transformations.

The ideas of nullspace and column space generalize to arbitrary linear transformations between vector spaces - with slightly more general terminology.

**Definition** Let  $V$  and  $W$  be vector spaces. A function  $T: V \rightarrow W$  is called a *linear transformation* if for each  $\mathbf{x} \in V$  there is a unique vector  $T(\mathbf{x}) \in W$  and so that

$$(i) \quad T(\mathbf{u} + \mathbf{v}) = T(\mathbf{u}) + T(\mathbf{v}) \quad \text{for all } \mathbf{u}, \mathbf{v} \in V$$

$$(ii) \quad T(c\mathbf{u}) = cT(\mathbf{u}) \quad \text{for all } \mathbf{u} \in V, c \in \mathbb{R}$$

$T: \mathbb{R}^n \rightarrow \mathbb{R}^m$   
linear  
same def.  
(we showed  $T(\vec{x}) = A\vec{x}$ )  
Nul  $A$   
Col  $A$

**Definition** The *kernel* (or *nullspace*) of  $T$  is defined to be  $\{\mathbf{u} \in V: T(\mathbf{u}) = \mathbf{0}\}$ .

**Definition** The *range* of  $T$  is  $\{\mathbf{w} \in W: \mathbf{w} = T(\mathbf{v}) \text{ for some } \mathbf{v} \in V\}$ .

**Theorem** Let  $T: V \rightarrow W$  be a linear transformation. Then the kernel of  $T$  is a subspace of  $V$ . The range of  $T$  is a subspace of  $W$ .

$Z = \text{kernel } T \text{ in } V, \text{ is subspace}$

$$1) \quad \vec{0} \in Z$$

$$2) \quad \vec{u}, \vec{v} \in Z \Rightarrow \vec{u} + \vec{v} \in Z$$

$$3) \quad \vec{u} \in Z \Rightarrow c\vec{u} \in Z.$$

$$1) \text{ is } T(\vec{0}) = \vec{0} ? \quad \underline{\text{yes}}. \quad T(\vec{0} + \vec{0}) = T(\vec{0}) + T(\vec{0})$$

$$2) \text{ if } T(\vec{u}) = \vec{0}, T(\vec{v}) = \vec{0} \quad \begin{matrix} T(\vec{0}) = T(\vec{0}) + T(\vec{0}) \\ -T(\vec{0}) \quad \quad -T(\vec{0}) \end{matrix}$$

$$\text{then } T(\vec{u} + \vec{v}) \quad \vec{0} = T(\vec{0}) !!$$

$$(i) \quad = T(\vec{u}) + T(\vec{v}) \\ = \vec{0} + \vec{0} = \vec{0}. \quad \text{so } \vec{u} + \vec{v} \in Z.$$

$$3) \text{ if } T(\vec{u}) = \vec{0} \\ \text{then } T(c\vec{u}) = cT(\vec{u}) = c\vec{0} = \vec{0}.$$

(ii)

so  $c\vec{u} \in Z$

Range  $T$  is subspace of  $W$ :  
show 1), 2), 3).

1) is  $\vec{0} \in \text{Range } T$ ? yes:  $T(\vec{0}) = \vec{0}$   
2) is Range  $T$  closed under addition?

let  $\vec{w}_1, \vec{w}_2 \in \text{range } T$   
then  $\exists \vec{u}_1, \vec{u}_2 \in V$  s.t.  $T(\vec{u}_1) = \vec{w}_1$   
 $T(\vec{u}_2) = \vec{w}_2$   
so  $T(\vec{u}_1 + \vec{u}_2) = T(\vec{u}_1) + T(\vec{u}_2)$   
 $= \vec{w}_1 + \vec{w}_2 \in \text{range } T!$

3) is Range  $T$  closed under scalar mult  
let  $\vec{w} \in \text{Range of } T$   
(is  $c\vec{w} \in \text{Range } T$ ?)

$$\text{so } \vec{w} = T(\vec{u})$$

$$\text{so } T(c\vec{u}) = cT(\vec{u}) = c\vec{w} \quad !!!$$

Exercise 2 Let  $V$  be the vector space  $C^1[a, b]$  of real-valued functions  $f$  defined on an interval  $[a, b]$  with the property that they are differentiable and that their derivatives are continuous functions on  $[a, b]$ . Let  $W$  be the vector space  $C[a, b]$  of all continuous functions on the interval  $[a, b]$ . Let  $D : V \rightarrow W$  be the derivative transformation

$$D(f) = f'.$$

$$f(x) = e^{3x}$$

$$D(f)_{(x)} = 3e^{3x}$$

2a) What Calculus differentiation rules tell you that  $D$  is a linear transformation?

$$T(\vec{u} + \vec{v}) = T(\vec{u}) + T(\vec{v})$$

$$T(c\vec{u}) = cT(\vec{u})$$

$$(f + g)' = f' + g'$$

$$D(f + g) = D(f) + D(g)$$

$$(cf)' = cf' \quad c \text{ const}$$

$$D(cf) = cD(f)$$

2b) What subspace is the kernel of  $D$ ?

$$\text{kernel } D = \{ \text{constant functions} \} = \{ c, c \in \mathbb{R} \}$$

Mean Value Theorem.

$$= \text{span} \{ 1 \}$$

$$1(x) = 1$$

2c) What is the range of  $D$ ?  $= C[a, b]$ .

because if  $f(x)$  is continuous on  $[a, b]$

then  $\int_a^x f(t) dt$  is an antiderivative  $\in C^1[a, b]$ .

$$F(x) = \int_a^x f(t) dt$$

$$D(F) = f !$$

Fundamental Theorem of Calculus, part I

Tues Feb 27

- 4.4 Coordinate systems for finite dimensional vector spaces

Announcements:

Let  $V$  be a vector space with basis  $\beta = \{\vec{v}_1, \vec{v}_2, \dots, \vec{v}_n\}$   
This basis gives us a coordinate system for  $V$ :  
for  $\vec{v} = c_1 \vec{v}_1 + c_2 \vec{v}_2 + \dots + c_n \vec{v}_n$   
 $c_1, c_2, \dots, c_n$  are the coordinates of  $\vec{v}$  with respect to  $\beta$

$$\begin{bmatrix} c_1 \\ c_2 \\ \vdots \\ c_n \end{bmatrix} \in \mathbb{R}^n$$

'til 12:57

Warm-up Exercise:

consider  $\text{span} \left\{ \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ -2 \end{bmatrix} \right\}$  which is a plane in  $\mathbb{R}^3$  with  
implicit equation

$$0x + 2y + z = 0.$$

the point  $\begin{bmatrix} 1.5 \\ .5 \\ -1 \end{bmatrix}$  lies on this plane  
 $(2y + z = 0)$

Find  $\begin{bmatrix} c_1 \\ c_2 \end{bmatrix}$  so that

$$1.5 \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} + .5 \begin{bmatrix} 0 \\ 1 \\ -2 \end{bmatrix} = \begin{bmatrix} 1.5 \\ .5 \\ -1 \end{bmatrix}.$$

Sketch.

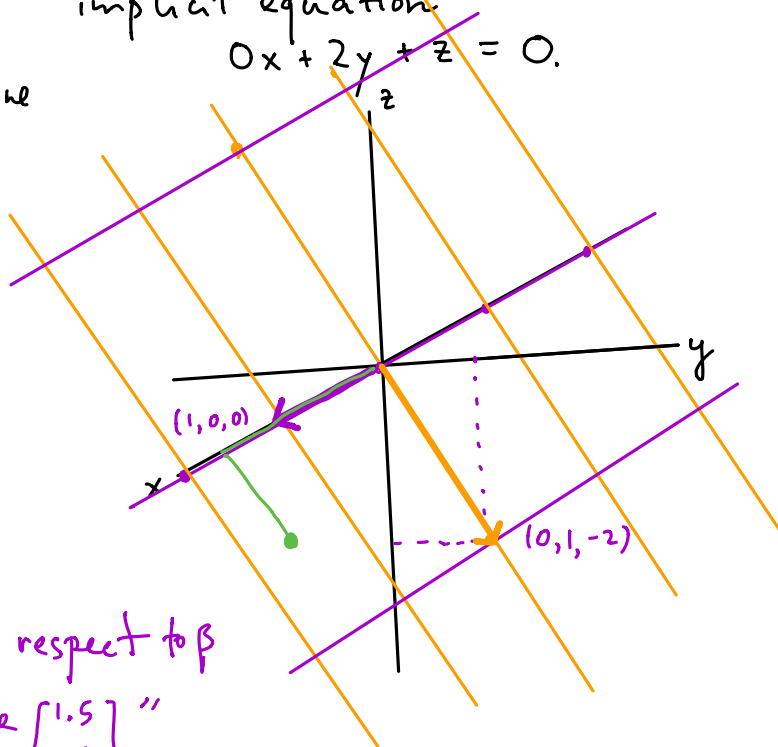
$$1.5 \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} + .5 \begin{bmatrix} 0 \\ 1 \\ -2 \end{bmatrix}$$

"coordinates of  $\begin{bmatrix} 1.5 \\ .5 \\ -1 \end{bmatrix}$  with respect to  $\beta$   
are  $\begin{bmatrix} 1.5 \\ .5 \end{bmatrix}$ "

$\nearrow$   
 $\mathbb{R}^3$

$\nearrow$   
 $\mathbb{R}^2$

this is the way that a plane thru the  
origin in  $\mathbb{R}^3$  is "like"  $\mathbb{R}^2$ , but not  $\mathbb{R}^2$ .



Theorem Let  $V$  be a vector space, and let  $\{\underline{b}_1, \underline{b}_2, \dots, \underline{b}_p\}$  be a basis for  $V$ . Then for each  $\underline{v} \in V$  there is a unique set of scalars  $c_1, c_2, \dots, c_p$  so that

$$\underline{v} = c_1 \underline{b}_1 + c_2 \underline{b}_2 + \dots + c_p \underline{b}_p. \quad E_1$$

proof:

Suppose

we could also write  $\underline{v} = d_1 \underline{b}_1 + d_2 \underline{b}_2 + \dots + d_p \underline{b}_p \quad E_2$

subtract  $E_2$   
from  $E_1$   
(and use vector  
space  
axioms)

commutativity  
associativity

$$\begin{aligned} \underline{0} &= c_1 \underline{b}_1 + c_2 \underline{b}_2 + \dots + c_p \underline{b}_p - (d_1 \underline{b}_1 + d_2 \underline{b}_2 + \dots + d_p \underline{b}_p) \\ &= (c_1 \underline{b}_1 - d_1 \underline{b}_1) + (c_2 \underline{b}_2 - d_2 \underline{b}_2) + \dots + (c_p \underline{b}_p - d_p \underline{b}_p) \end{aligned}$$

$$\underline{0} = (c_1 - d_1) \underline{b}_1 + (c_2 - d_2) \underline{b}_2 + \dots + (c_p - d_p) \underline{b}_p$$

because  $\{\underline{b}_1, \underline{b}_2, \dots, \underline{b}_p\}$  in lin. indep.

know

$$c_1 - d_1 = 0$$

$$c_2 - d_2 = 0$$

$$\vdots$$

$$c_p - d_p = 0$$

i.e. each  $c_j = d_j$



Definition (Each basis gives us a coordinate system). Let  $V$  be a vector space, and let  $\beta = \{\underline{b}_1, \underline{b}_2, \dots, \underline{b}_p\}$  be a basis for  $V$ . For each  $\underline{v} \in V$  we say that *the coordinates of  $\underline{v}$  with respect to  $\beta$*  are  $c_1, c_2, \dots, c_p$  if

$$\underline{v} = c_1 \underline{b}_1 + c_2 \underline{b}_2 + \dots + c_p \underline{b}_p.$$

And, we write the vector of the coordinates of  $\underline{v}$  with respect to  $\beta$  as:

$$[\underline{v}]_{\beta} = \begin{bmatrix} c_1 \\ c_2 \\ \vdots \\ c_p \end{bmatrix} \in \mathbb{R}^p.$$

coord.  
vector of  $\underline{v}$   
with respect  
to the basis  $\beta$ .

Example: For the vector space

$$P_3 = \{p(t) = a_0 + a_1 t + a_2 t^2 + a_3 t^3 \text{ such that } a_0, a_1, a_2, a_3 \in \mathbb{R}\}$$

we've checked that

$$\beta = \{1, t, t^2, t^3\}$$

is a basis. So the coordinate vector of

$$p(t) = 3 - 4t^2 + t^3$$

with respect to  $\beta$  is

$$[p]_{\beta} = \begin{bmatrix} 3 \\ 0 \\ -4 \\ 1 \end{bmatrix}$$



And, if  $q \in \mathcal{P}_3$ , with

$$[q]_{\beta} = \begin{bmatrix} -2 \\ 1 \\ 7 \\ 0 \end{bmatrix}$$

then

$$q(t) = -2 + t + 7t^2.$$

It turns out that we can understand pretty much any vector space question about  $\mathcal{P}_3$  by interpreting the question in terms of the coordinates with respect to  $\beta$ , which lets us work in  $\mathbb{R}^4$  in lieu of  $\mathcal{P}_3$ . That's what coordinates with respect to a basis are good for, when you're working with a non-standard vector space.

Exercise 1) Let

$$\beta = \left\{ \begin{bmatrix} 1 \\ -1 \end{bmatrix}, \begin{bmatrix} 1 \\ 3 \end{bmatrix} \right\} = \{\underline{u}, \underline{v}\}$$

be a non-standard basis of  $\mathbb{R}^2$ .

1a) Suppose  $\underline{x}$  is a vector in  $\mathbb{R}^2$ , and

$$[\underline{x}]_{\beta} = \begin{bmatrix} 2 \\ 1 \end{bmatrix}.$$

Find the standard coordinates for  $\underline{x}$ , i.e. its coordinates with respect to the standard basis  $E = \{\underline{e}_1, \underline{e}_2\}$ .

$$\Rightarrow \underline{x} = 2 \begin{bmatrix} 1 \\ -1 \end{bmatrix} + 1 \begin{bmatrix} 1 \\ 3 \end{bmatrix} = \begin{bmatrix} 3 \\ 1 \end{bmatrix} = [\underline{x}]_E \quad \left\{ \begin{array}{l} [\underline{x}]_E = \begin{bmatrix} 1 & 1 \\ -1 & 3 \end{bmatrix} [\underline{x}]_{\beta} \\ \begin{array}{l} P_{\beta \leftarrow E} \\ P_{E \leftarrow \beta} \end{array} \end{array} \right.$$

1b) Find the  $\beta$ -coordinates for the vector  $\underline{b} = \begin{bmatrix} -2 \\ 8 \end{bmatrix}$ . (The math may seem familiar.)

1c) Interpret your work in 1ab geometrically, in terms of the coordinate system generated by  $\beta$ .

$$1b). \quad \underline{b} = \begin{bmatrix} -2 \\ 8 \end{bmatrix}$$

$$[\underline{b}]_{\beta}: \quad c_1 \begin{bmatrix} 1 \\ -1 \end{bmatrix} + c_2 \begin{bmatrix} 1 \\ 3 \end{bmatrix} = \begin{bmatrix} -2 \\ 8 \end{bmatrix}$$

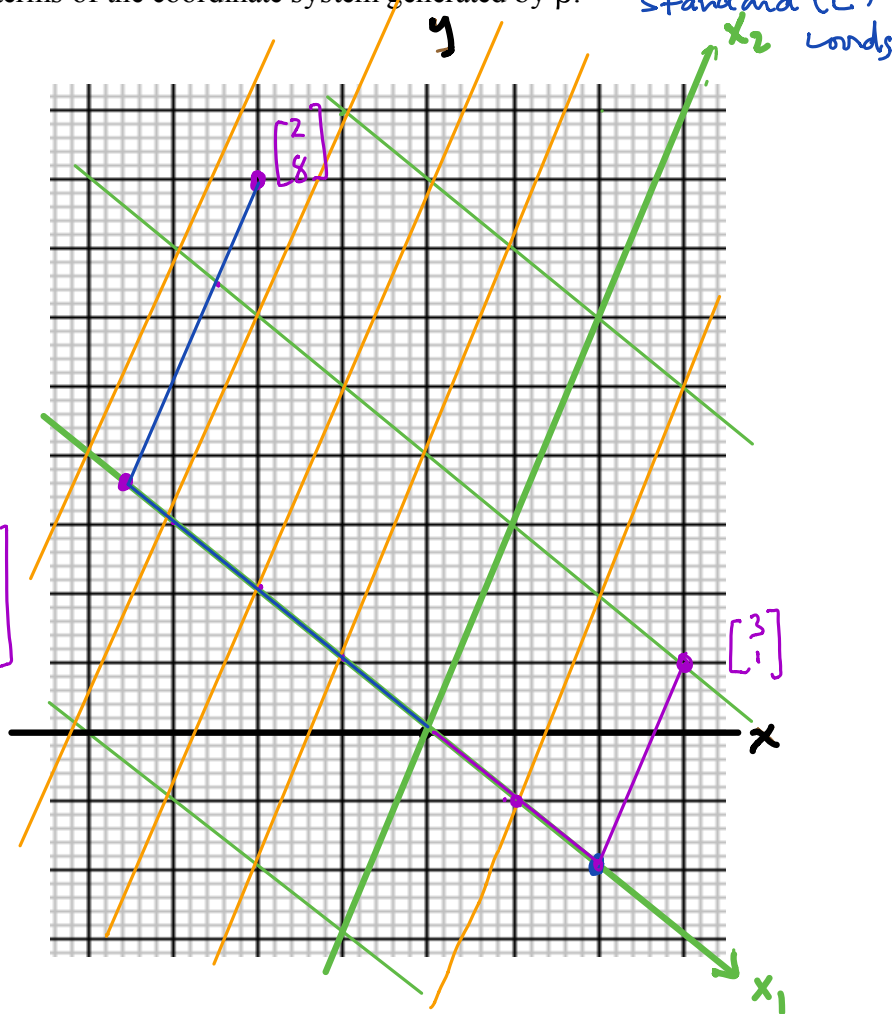
$$\begin{bmatrix} 1 & 1 \\ -1 & 3 \end{bmatrix} \begin{bmatrix} c_1 \\ c_2 \end{bmatrix} = \begin{bmatrix} -2 \\ 8 \end{bmatrix}$$

$$\begin{bmatrix} c_1 \\ c_2 \end{bmatrix} = \frac{1}{4} \begin{bmatrix} 3 & -1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} -2 \\ 8 \end{bmatrix}$$

$$= \frac{1}{4} \begin{bmatrix} -14 \\ 6 \end{bmatrix}$$

$$= \begin{bmatrix} -3.5 \\ 1.5 \end{bmatrix}$$

$$\text{So } \begin{bmatrix} \underline{b} \end{bmatrix}_{\beta} = \begin{bmatrix} -3.5 \\ 1.5 \end{bmatrix}$$



Theorem Let  $V, W$  be vector spaces, and  $T: V \rightarrow W$  a linear transformation. If  $T$  is 1-1 and onto, then the inverse function  $T^{-1}$  is also a linear transformation,  $T^{-1}: W \rightarrow V$ . In this case, we call  $T$  an isomorphism.

proof: We have to check that for all  $\underline{u}, \underline{w} \in W$  and all  $c \in \mathbb{R}$ ,

check

$$T^{-1}(\underline{u} + \underline{w}) = T^{-1}(\underline{u}) + T^{-1}(\underline{w})$$

$$T^{-1}(c \underline{u}) = c T^{-1}(\underline{u}).$$

Since  $T$  is 1-1

it suffices to check that  $T$  of LHS's =  $T$  RHS's

$$T \text{ of RHS } T(T^{-1}(\underline{u}) + T^{-1}(\underline{w})) \stackrel{T \text{ linear}}{=} T(T^{-1}(\underline{u})) + T(T^{-1}(\underline{w}))$$

$$T \text{ of LHS: } T(T^{-1}(\underline{u} + \underline{w})) = \underline{u} + \underline{w}$$

$T$  is linear

for scalar multiplication:  $T(T^{-1}(c \underline{u})) = c \underline{u}$ ,  $T(c T^{-1}(\underline{u})) \stackrel{T \text{ linear}}{=} c T(T^{-1}(\underline{u})) = c \underline{u}$

Theorem Let  $V$  be a vector space, with basis  $\beta = \{\underline{b}_1, \underline{b}_2, \dots, \underline{b}_n\}$ . Then the coordinate transform  $T: V \rightarrow \mathbb{R}^n$  defined by

$$T(\underline{v}) = [\underline{v}]_{\beta}$$

is linear, and it is an isomorphism.

Friday!!

$$\underline{v} = c_1 \underline{b}_1 + c_2 \underline{b}_2 + \dots + c_n \underline{b}_n$$

$$\underline{w} = d_1 \underline{b}_1 + d_2 \underline{b}_2 + \dots + d_n \underline{b}_n$$

$$\text{so } [\underline{v}]_{\beta} = \begin{bmatrix} c_1 \\ c_2 \\ \vdots \\ c_n \end{bmatrix}$$

$$[\underline{w}]_{\beta} = \begin{bmatrix} d_1 \\ d_2 \\ \vdots \\ d_n \end{bmatrix}$$

$$\underline{v} + \underline{w} = (c_1 + d_1) \underline{b}_1 + (c_2 + d_2) \underline{b}_2 + \dots + (c_n + d_n) \underline{b}_n$$

$$(1) \quad T(\underline{v} + \underline{w}) = T(\underline{v}) + T(\underline{w}) \quad [\underline{v} + \underline{w}]_{\beta} = \begin{bmatrix} c_1 + d_1 \\ c_2 + d_2 \\ \vdots \\ c_n + d_n \end{bmatrix} = [\underline{v}]_{\beta} + [\underline{w}]_{\beta}$$

$$c \underline{v} = c(c_1 \underline{b}_1 + c_2 \underline{b}_2 + \dots + c_n \underline{b}_n) = c c_1 \underline{b}_1 + c c_2 \underline{b}_2 + \dots$$

$$(2) \quad T(c \underline{v}) = c T(\underline{v})$$

$$[c \underline{v}]_{\beta} = \begin{bmatrix} c c_1 \\ c c_2 \\ \vdots \\ c c_n \end{bmatrix} = c [\underline{v}]_{\beta}$$

I know  $T^{-1}: \mathbb{R}^n \rightarrow V$

$$T^{-1} \begin{bmatrix} e_1 \\ e_2 \\ \vdots \\ e_n \end{bmatrix} = e_1 \underline{v}_1 + e_2 \underline{v}_2 + \dots + e_n \underline{v}_n$$

'til 12:57

# Friday warm-up (Exercise 2 in Tuesday notes)

Exercise 2: Use coordinates with respect to the basis  $\{1, t, t^2\} = \beta$ , to check whether or not the set of polynomials  $\{p_1(t), p_2(t), p_3(t)\}$  is a basis for  $P_2$ , where

$$\begin{aligned} p_1(t) &= 1 + t^2 \\ p_2(t) &= 2 + 3t + t^2 \\ p_3(t) &= -3t + t^2. \end{aligned} \quad [p_i]_\beta = \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix} \text{ etc.}$$

i.e. check whether the coord vectors in  $\mathbb{R}^3$  are independent & span  $\mathbb{R}^3$ . Then make conclusions about  $\{p_1, p_2, p_3\}$

$$\begin{aligned} p_2 &= 2p_1 - p_3 \\ 2p_1 - p_2 - p_3 &= 0 \\ 2(1+t^2) - (2+3t+t^2) - (-3t+t^2) &= 0! \end{aligned}$$

how, with coord vectors instead the "old way"

$$c_1 \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix} + c_2 \begin{bmatrix} 2 \\ 3 \\ 1 \end{bmatrix} + c_3 \begin{bmatrix} 0 \\ -3 \\ 1 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

$p_1(t) = 1 + 0t + 1t^2$        $p_2(t) = 2 + 3t + t^2$

$$\begin{array}{ccc|ccc} 1 & 2 & 0 & 0 & 1 & 2 & 0 \\ 0 & 3 & -3 & 0 & 0 & -1 & 0 \\ 1 & 1 & 1 & 0 & 0 & 0 & 0 \end{array} \xrightarrow{\text{rref}} \begin{array}{ccc|ccc} 1 & 0 & 2 & 0 & 0 & 2 & 0 \\ 0 & 1 & -1 & 0 & 0 & -1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 \end{array}$$

$c_1 = -2c_3$   
 $c_2 = c_3$   
 $c_3 = \text{free}$

Theorem: If  $T: V \rightarrow W$  is an isomorphism, then

$\{\vec{v}_1, \dots, \vec{v}_p\}$  is (in)dependent in  $V$  if and only if

$\{T\vec{v}_1, T\vec{v}_2, \dots, T\vec{v}_p\}$  are (in)dependent in  $W$

$$c_3 = 1: -2 \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix} + 1 \begin{bmatrix} 2 \\ 3 \\ 1 \end{bmatrix} - 1 \begin{bmatrix} 0 \\ -3 \\ 1 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

proof: If  $c_1 \vec{v}_1 + c_2 \vec{v}_2 + \dots + c_p \vec{v}_p = \vec{0}$   
 then  $T(c_1 \vec{v}_1 + c_2 \vec{v}_2 + \dots + c_p \vec{v}_p) = T(\vec{0}) = \vec{0}$   
 linearity  $\longrightarrow$   $c_1 T(\vec{v}_1) + c_2 T(\vec{v}_2) + \dots + c_p T(\vec{v}_p) = \vec{0}$

So, if  $\{\vec{v}_1, \dots, \vec{v}_p\}$  are dependent then  $\{T(\vec{v}_1), T(\vec{v}_2), \dots, T(\vec{v}_p)\}$  are also dependent, with the same weights  
 apply same reasoning in reverse, using  $T^{-1}$ , to show that if  $\{T\vec{v}_1, T\vec{v}_2, \dots, T\vec{v}_p\}$  are dependent in  $W$ , then  $\{\vec{v}_1, \vec{v}_2, \dots, \vec{v}_p\}$  are dependent in  $V$ .  $\square$

by logic:  
same fact  
for "independent"

Exercise 3 Generalize the example of Exercise 1: Suppose  $\beta = \{\underline{b}_1, \underline{b}_2, \dots, \underline{b}_n\}$  is a non-standard basis of  $\mathbb{R}^n$ . And let  $E = \{\underline{e}_1, \underline{e}_2, \dots, \underline{e}_n\}$  be the standard basis of  $\mathbb{R}^n$ . For  $\underline{x} \in \mathbb{R}^n$ , how do you convert between  $\underline{x} = [\underline{x}]_E$ , and  $[\underline{x}]_\beta$ , and vice-versa?

$$* \quad \text{if } \underline{x} = [\underline{x}]_E = c_1 \underline{b}_1 + c_2 \underline{b}_2 + \dots + c_n \underline{b}_n = \begin{bmatrix} \underline{b}_1 & \underline{b}_2 & \dots & \underline{b}_n \end{bmatrix} \begin{bmatrix} c_1 \\ c_2 \\ \vdots \\ c_n \end{bmatrix}$$

that means  $[\underline{x}]_\beta = \begin{bmatrix} c_1 \\ c_2 \\ \vdots \\ c_n \end{bmatrix}$

$$* \quad [\underline{x}]_E = B [\underline{x}]_\beta \quad B \text{ is called } P_{E \leftarrow \beta}$$

$$** \quad B^{-1} [\underline{x}]_E = [\underline{x}]_\beta \quad B^{-1} \text{ is called } P_{\beta \leftarrow E}$$

Wed Feb 28

- 4.5 dimension of a vector space, and related facts about span and linear independence.

Monday!

Announcements:

- Tuesday's notes today.
- Quiz

'til 10:57

Warm-up Exercise:

$$\vec{a}_1 \quad \vec{a}_2 \quad \vec{a}_3 \quad \vec{a}_4 \quad \text{col } A = \text{span} \{ \vec{a}_1, \vec{a}_2, \vec{a}_3, \vec{a}_4 \}$$
$$A = \begin{bmatrix} 1 & 3 & 0 & 4 \\ 2 & -2 & -8 & 0 \\ -1 & 2 & 5 & 1 \\ 0 & 4 & 4 & 4 \end{bmatrix} \quad \text{rref}(A) = \begin{bmatrix} 1 & 0 & -3 & 1 \\ 0 & 1 & 1 & 1 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

what is the dimension of  $\text{Col } A$ ?  $= 2$  (# of vectors in a basis)  
a basis for  $\text{Col } A$ ?  $\{ \vec{a}_1, \vec{a}_2 \}$

what is dimension of  $\text{Nul } A$ ?

could you find a basis of  $\text{Nul } A$ ?

$\dim \text{Nul } A = 2 = \# \text{ of free variables when we solve}$   
 $A\vec{x} = \vec{0}$

$= \# \text{ columns of } \text{rref}(A) \text{ without pivots.}$

There is a circle of ideas related to linear independence, span, and bases for vector spaces, which it is good to try and understand carefully. That's what we'll do today. These ideas generalize (and use) ideas we've already explored more concretely, and facts we already know to be true for the vector spaces  $\mathbb{R}^n$ .

Theorem 1 (constructing a basis from a spanning set): Let  $V$  be a vector space of dimension at least one, and let  $\text{span}\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_p\} = V$ .

Then a subset of the spanning set is a basis for  $V$ . (We followed a procedure like this to extract bases for  $\text{Col } A$ .)

Theorem 2 Let  $V$  be a vector space, with basis  $\beta = \{\mathbf{b}_1, \mathbf{b}_2, \dots, \mathbf{b}_n\}$ . Then any set in  $V$  containing more than  $n$  elements must be linearly dependent. (We used reduced row echelon form to understand this in  $\mathbb{R}^n$ .)

Theorem 4 Let  $V$  be a vector space, with basis  $\beta = \{\underline{b}_1, \underline{b}_2, \dots, \underline{b}_n\}$ . Let  $\alpha = \{\underline{a}_1, \underline{a}_2, \dots, \underline{a}_p\}$  be a set of independent vectors that don't span  $V$ . Then  $p < n$ , and additional vectors can be added to the set  $\alpha$  to create a basis  $\{\underline{a}_1, \underline{a}_2, \dots, \underline{a}_p, \dots, \underline{a}_n\}$  (We followed a procedure like this when we figured out all the subspaces of  $\mathbb{R}^3$ .)

Theorem 3 Let  $V$  be a vector space, with basis  $\beta = \{\underline{b}_1, \underline{b}_2, \dots, \underline{b}_n\}$ . Then no set  $\alpha = \{\underline{a}_1, \underline{a}_2, \dots, \underline{a}_p\}$  with  $p < n$  vectors can span  $V$ . (We know this for  $\mathbb{R}^n$ .)



Theorem 5 Let  $V$  be a vector space, with basis  $\beta = \{\underline{b}_1, \underline{b}_2, \dots, \underline{b}_n\}$ . Then every basis for  $V$  has exactly  $n$  vectors. Furthermore, if  $\alpha = \{\underline{a}_1, \underline{a}_2, \dots, \underline{a}_n\}$  is another collection of exactly  $n$  vectors in  $V$ , and if  $\text{span}\{\underline{a}_1, \underline{a}_2, \dots, \underline{a}_n\} = V$ , then the set  $\alpha$  is automatically linearly independent and a basis. Conversely, if the set  $\{\underline{a}_1, \underline{a}_2, \dots, \underline{a}_n\}$  is linearly independent, then  $\text{span}\{\underline{a}_1, \underline{a}_2, \dots, \underline{a}_n\} = V$  is guaranteed, and  $\alpha$  is a basis. (We know all these facts for  $\mathbb{R}^n$  from reduced row echelon form considerations.)

Corollary Let  $V$  be a vector space of dimension  $n$ . Then the subspaces of  $V$  have dimensions  $0, 1, 2, \dots, n-1, n$ . (We know this for  $\mathbb{R}^n$ .)

Fri Mar 2

- 4.6 The four subspaces associated with a matrix. the rank of a matrix.

Announcements: Finish Tuesday notes  
I think we'll finish Wed, but if we don't, that's fine  
• fft

Warm-up Exercise: See Tuesday notes

From Tuesday notes, finished on Friday

Theorem Let  $V, W$  be vector spaces, and  $T: V \rightarrow W$  a linear transformation. If  $T$  is 1-1 and onto, then the inverse function  $T^{-1}$  is also a linear transformation,  $T^{-1}: W \rightarrow V$ . In this case, we call  $T$  an isomorphism.

proof: We have to check that for all  $\underline{u}, \underline{w} \in W$  and all  $c \in \mathbb{R}$ ,

check

$$T^{-1}(\underline{u} + \underline{w}) = T^{-1}(\underline{u}) + T^{-1}(\underline{w})$$

$$T^{-1}(c\underline{u}) = c T^{-1}(\underline{u}).$$

Since  $T$  is 1-1

it suffices to check that  $T$  of LHS's =  $T$  RHS's

$$T \text{ of RHS } T(T^{-1}(\underline{u}) + T^{-1}(\underline{w})) \stackrel{T \text{ linear}}{=} T(T^{-1}(\underline{u})) + T(T^{-1}(\underline{w}))$$

$$T \text{ of LHS: } T(T^{-1}(\underline{u} + \underline{w})) = \underline{u} + \underline{w}$$

$T$  is linear

$$\text{for scalar multiplication: } T(T^{-1}(c\underline{u})) = c\underline{u}, \quad T(c T^{-1}(\underline{u})) \stackrel{T \text{ linear}}{=} c T T^{-1}(\underline{u}) = c\underline{u}$$

Theorem Let  $V$  be a vector space, with basis  $\beta = \{\underline{b}_1, \underline{b}_2, \dots, \underline{b}_n\}$ . Then the coordinate transform  $T: V \rightarrow \mathbb{R}^n$  defined by

$$T(\underline{v}) = [\underline{v}]_{\beta}$$

is linear, and it is an isomorphism.

Friday!!

$$\underline{v} = c_1 \underline{b}_1 + c_2 \underline{b}_2 + \dots + c_n \underline{b}_n$$

$$\underline{w} = d_1 \underline{b}_1 + d_2 \underline{b}_2 + \dots + d_n \underline{b}_n$$

$$\text{so } [\underline{v}]_{\beta} = \begin{bmatrix} c_1 \\ c_2 \\ \vdots \\ c_n \end{bmatrix}$$

$$[\underline{w}]_{\beta} = \begin{bmatrix} d_1 \\ d_2 \\ \vdots \\ d_n \end{bmatrix}$$

$$\underline{v} + \underline{w} = (c_1 + d_1) \underline{b}_1 + (c_2 + d_2) \underline{b}_2 + \dots + (c_n + d_n) \underline{b}_n$$

$$(1) \quad T(\underline{v} + \underline{w}) = T(\underline{v}) + T(\underline{w}) \quad [\underline{v} + \underline{w}]_{\beta} = \begin{bmatrix} c_1 + d_1 \\ c_2 + d_2 \\ \vdots \\ c_n + d_n \end{bmatrix} = [\underline{v}]_{\beta} + [\underline{w}]_{\beta}$$

$$c\underline{v} = c(c_1 \underline{b}_1 + c_2 \underline{b}_2 + \dots + c_n \underline{b}_n) = cc_1 \underline{b}_1 + cc_2 \underline{b}_2 + \dots$$

$$(2) \quad T(c\underline{v}) = c T(\underline{v})$$

$$[c\underline{v}]_{\beta} = \begin{bmatrix} cc_1 \\ cc_2 \\ \vdots \\ cc_n \end{bmatrix} = c [\underline{v}]_{\beta}$$

I know  $T^{-1}: \mathbb{R}^n \rightarrow V$

$$T^{-1} \begin{bmatrix} e_1 \\ e_2 \\ \vdots \\ e_n \end{bmatrix} = e_1 \underline{v}_1 + e_2 \underline{v}_2 + \dots + e_n \underline{v}_n$$

'til 12:57

# Friday warm-up (Exercise 2 in Tuesday notes)

Exercise 2: Use coordinates with respect to the basis  $\{1, t, t^2\} = \beta$ , to check whether or not the set of polynomials  $\{p_1(t), p_2(t), p_3(t)\}$  is a basis for  $P_2$ , where

$$\begin{aligned} p_1(t) &= 1 + t^2 \\ p_2(t) &= 2 + 3t + t^2 \\ p_3(t) &= -3t + t^2. \end{aligned} \quad [p_i]_\beta = \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix} \text{ etc.}$$

i.e. check whether the coord vectors in  $\mathbb{R}^3$  are independent & span  $\mathbb{R}^3$ . Then make conclusions about  $\{p_1, p_2, p_3\}$

$$\begin{aligned} p_2 &= 2p_1 - p_3 \\ 2p_1 - p_2 - p_3 &= 0 \\ 2(1+t^2) - (2+3t+t^2) - (-3t+t^2) &= 0! \end{aligned}$$

how, with coord vectors instead the "old way"

$$c_1 \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix} + c_2 \begin{bmatrix} 2 \\ 3 \\ 1 \end{bmatrix} + c_3 \begin{bmatrix} 0 \\ -3 \\ 1 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

$p_1(t) = 1 + 0t + 1t^2$        $p_2(t) = 2 + 3t + t^2$

$$\begin{array}{ccc|ccc} 1 & 2 & 0 & 0 & 1 & 2 & 0 \\ 0 & 3 & -3 & 0 & 0 & -1 & 0 \\ 1 & 1 & 1 & 0 & 0 & 0 & 0 \end{array} \xrightarrow{\text{rref}} \begin{array}{ccc|ccc} 1 & 0 & 2 & 0 & 0 & 2 & 0 \\ 0 & 1 & -1 & 0 & 0 & -1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 \end{array}$$

$c_1 = -2c_3$   
 $c_2 = c_3$   
 $c_3 = \text{free}$

Theorem: If  $T: V \rightarrow W$  is an isomorphism, then

$\{\vec{v}_1, \dots, \vec{v}_p\}$  is (in)dependent in  $V$  if and only if

$\{T\vec{v}_1, T\vec{v}_2, \dots, T\vec{v}_p\}$  are (in)dependent in  $W$

$$c_3 = 1: -2 \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix} + 1 \begin{bmatrix} 2 \\ 3 \\ 1 \end{bmatrix} - 1 \begin{bmatrix} 0 \\ -3 \\ 1 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

proof: If  $c_1 \vec{v}_1 + c_2 \vec{v}_2 + \dots + c_p \vec{v}_p = \vec{0}$   
 then  $T(c_1 \vec{v}_1 + c_2 \vec{v}_2 + \dots + c_p \vec{v}_p) = T(\vec{0}) = \vec{0}$   
 linearity  $\longrightarrow$   $c_1 T(\vec{v}_1) + c_2 T(\vec{v}_2) + \dots + c_p T(\vec{v}_p) = \vec{0}$

So, if  $\{\vec{v}_1, \dots, \vec{v}_p\}$  are dependent then  $\{T(\vec{v}_1), T(\vec{v}_2), \dots, T(\vec{v}_p)\}$  are also dependent, with the same weights  
 apply same reasoning in reverse, using  $T^{-1}$ , to show that if  $\{T\vec{v}_1, T\vec{v}_2, \dots, T\vec{v}_p\}$  are dependent in  $W$ , then  $\{\vec{v}_1, \vec{v}_2, \dots, \vec{v}_p\}$  are dependent in  $V$ .  $\square$

by logic:  
same fact  
for "independent"

Exercise 3 Generalize the example of Exercise 1: Suppose  $\beta = \{\underline{b}_1, \underline{b}_2, \dots, \underline{b}_n\}$  is a non-standard basis of  $\mathbb{R}^n$ . And let  $E = \{\underline{e}_1, \underline{e}_2, \dots, \underline{e}_n\}$  be the standard basis of  $\mathbb{R}^n$ . For  $\underline{x} \in \mathbb{R}^n$ , how do you convert between  $\underline{x} = [\underline{x}]_E$ , and  $[\underline{x}]_\beta$ , and vice-versa?

$$* \quad \text{if } \underline{x} = [\underline{x}]_E = c_1 \underline{b}_1 + c_2 \underline{b}_2 + \dots + c_n \underline{b}_n = [\underline{b}_1 \ \underline{b}_2 \ \dots \ \underline{b}_n] \begin{bmatrix} c_1 \\ c_2 \\ \vdots \\ c_n \end{bmatrix}$$

that means  $[\underline{x}]_\beta = \begin{bmatrix} c_1 \\ c_2 \\ \vdots \\ c_n \end{bmatrix}$

$$* \quad [\underline{x}]_E = B [\underline{x}]_\beta \quad B \text{ is called } P_{E \leftarrow \beta}$$

$$** \quad B^{-1} [\underline{x}]_E = [\underline{x}]_\beta \quad B^{-1} \text{ is called } P_{\beta \leftarrow E}$$