

(1)

Math 2250-1

Fri 10/7 Finish 4.1-4.4

Finish Wed. notes, p. 2-3, discuss. pages 1-2 in today's notes [p. 3-5 are a summary of the rest of the past week, included in case you might find the summary helpful.]

Related to exercise 17, it is useful to know that if

$\tilde{v}_1, \tilde{v}_2, \dots, \tilde{v}_k$  are linearly independent vectors in  $V$

but they don't span  $V$ ,

then if you pick any  $\tilde{w} \in V$ ,  $\tilde{w} \notin \text{span}\{\tilde{v}_1, \dots, \tilde{v}_k\}$   
the larger collection

$\tilde{v}_1, \tilde{v}_2, \dots, \tilde{v}_k, \tilde{w}$  will be linearly independent.

$$\text{check: } c_1\tilde{v}_1 + c_2\tilde{v}_2 + \dots + c_k\tilde{v}_k + d\tilde{w} = \tilde{0}$$

$$\text{if } d = 0, \text{ then } c_1\tilde{v}_1 + c_2\tilde{v}_2 + \dots + c_k\tilde{v}_k = \tilde{0}$$

$$\text{so } c_1 = c_2 = \dots = c_k = 0 \text{ also.}$$

and we can't have  $d \neq 0$ , since

$$\text{then } \tilde{w} = -\frac{c_1}{d}\tilde{v}_1 - \frac{c_2}{d}\tilde{v}_2 - \dots - \frac{c_k}{d}\tilde{v}_k$$

and  $\tilde{w} \notin \text{span}\{\tilde{v}_1, \dots, \tilde{v}_k\}$ . ■

in this way you can successively build up to a basis for  $V$ , containing the original independent vectors  $\tilde{v}_1, \tilde{v}_2, \dots, \tilde{v}_k$ .

the explanation of the facts about dimension & basis on page 3 of Wed. notes is on the next page of today's notes.

(2)

Key facts about independence, span, basis, dimension (§4.4)

① Primary fact: If a finite collection of vectors spans a vector space  $V$ , then any collection having a greater number of vectors is dependent

reason: Let  $\{\tilde{v}_1, \tilde{v}_2, \dots, \tilde{v}_k\}$  span  $V$ , let  $\{\tilde{w}_1, \tilde{w}_2, \dots, \tilde{w}_{k+l}\}$  ( $l > 0$ ) be a larger collection

We search for lin. comb. coeff's  $c_j$  s.t.

$$c_1 \tilde{w}_1 + c_2 \tilde{w}_2 + \dots + c_{k+l} \tilde{w}_{k+l} = \vec{0}$$

Express each  $\tilde{w}_j$  as a linear combo of the spanning set:

$$c_1 \begin{bmatrix} q_1 \tilde{v}_1 \\ q_2 \tilde{v}_1 \\ \vdots \\ q_{k+l} \tilde{v}_1 \end{bmatrix} + c_2 \begin{bmatrix} q_1 \tilde{v}_2 \\ q_2 \tilde{v}_2 \\ \vdots \\ q_{k+l} \tilde{v}_2 \end{bmatrix} + \dots + c_{k+l} \begin{bmatrix} q_1 \tilde{v}_k \\ q_2 \tilde{v}_k \\ \vdots \\ q_{k+l} \tilde{v}_k \end{bmatrix} = \begin{bmatrix} 0 \tilde{v}_1 \\ 0 \tilde{v}_2 \\ \vdots \\ 0 \tilde{v}_k \end{bmatrix}$$

equating coeff's for each  $\tilde{v}_i$ , we get a dependency if we can find a  $\vec{c} \neq \vec{0}$  so that

$$[A] \begin{bmatrix} \vec{c} \end{bmatrix} = \begin{bmatrix} \vec{0} \end{bmatrix}$$

But the solution space to this homogeneous equation is at least  $l$ -dimensional, so non-zero solns  $\vec{c}$  exist. ■

Logical consequences of ①:

② If a finite collection of vectors in  $V$  is independent, no collection with fewer vectors can span  
reason: logic! If a smaller set did span, the larger set would've been dependent! ■

③ If  $\{\tilde{v}_1, \tilde{v}_2, \dots, \tilde{v}_k\}$  is a basis for  $V$ , then every other basis also consists of exactly  $k$  vectors  
reason: fewer vectors can't span, more vectors would be dependent. ■

④ If  $\dim V = k$  and if  $\{\tilde{w}_1, \tilde{w}_2, \dots, \tilde{w}_k\}$  are independent, they span! (and are a basis)  
reason: if they didn't span there would be a  $\tilde{v} \in V$  not in their span, so that the set  $\{\tilde{w}_1, \tilde{w}_2, \dots, \tilde{w}_k, \tilde{v}\}$  would still be independent. This would violate ①! ■

⑤ If  $\dim V = k$  and if  $\{\tilde{w}_1, \tilde{w}_2, \dots, \tilde{w}_k\}$  span, then they're independent! (so are a basis)  
reason: if they were dependent we could throw one of them away without shrinking the span, so we would have  $(k-1)$  vectors spanning  $V$ , in violation of ②! ■

### 4.1 - 4.4 summary (for reference)

(also include the facts about dimension &  
bases on the previous page.)

Since last Friday (and in HW) we've been studying problems and concepts related to

linear combinations  $c_1\vec{v}_1 + c_2\vec{v}_2 + \dots + c_k\vec{v}_k$  of vectors  $\vec{v}_1, \vec{v}_2, \dots, \vec{v}_k$ .

if we're working in  $\mathbb{R}^m$  we answer all such questions using the matrix theory from Chapter 3, because

$$c_1\vec{v}_1 + c_2\vec{v}_2 + \dots + c_k\vec{v}_k = \underbrace{\begin{bmatrix} \vec{v}_1 & \vec{v}_2 & \dots & \vec{v}_k \end{bmatrix}}_k \begin{bmatrix} c_1 \\ c_2 \\ \vdots \\ c_k \end{bmatrix} = B\vec{c}$$

linear independence / dependence for  $\vec{v}_1, \vec{v}_2, \dots, \vec{v}_k$ :

$$\begin{array}{l} c_1\vec{v}_1 + c_2\vec{v}_2 + \dots + c_k\vec{v}_k = \vec{0} \\ \Rightarrow c_1 = c_2 = \dots = c_k = 0 \end{array} \quad \begin{array}{l} c_1\vec{v}_1 + c_2\vec{v}_2 + \dots + c_k\vec{v}_k = \vec{0} \text{ for some choice of} \\ c_1, c_2, \dots, c_k \text{ with not all of} \\ c_1, c_2, \dots, c_k = 0. \end{array}$$

test:

$$B\vec{c} = \vec{0} \rightarrow B \left| \begin{array}{c} \vec{0} \\ \vdots \\ \vec{0} \end{array} \right.$$

$$\downarrow \\ \text{rref}(B) \left| \begin{array}{c} \vec{0} \\ \vdots \\ \vec{0} \end{array} \right.$$

if every column of  $\text{rref}(B)$  has a leading 1,  
deduce  $\vec{c} = \vec{0}$

Independent

if some column of  $\text{rref}(B)$  does not have a leading 1  $\Rightarrow$  free parameters

$\Rightarrow$  only many dependencies

dependent

$$\begin{bmatrix} & B \\ m & \vdots \\ & k \end{bmatrix}$$

this is guaranteed to happen if the number of vectors  $k$ , is greater than the  $m$  in  $\mathbb{R}^m$

④

span of  $\vec{v}_1, \vec{v}_2, \dots, \vec{v}_k$  is the collection of all their linear combinations:

$$\text{span} \{ \vec{v}_1, \vec{v}_2, \dots, \vec{v}_k \} = \{ \vec{w} = c_1 \vec{v}_1 + c_2 \vec{v}_2 + \dots + c_k \vec{v}_k, c_j \in \mathbb{R}, 1 \leq j \leq k \}.$$

We checked that the linear combination coefficients for each  $\vec{w} \in \text{span} \{ \vec{v}_1, \vec{v}_2, \dots, \vec{v}_k \}$  are unique if and only if  $\vec{v}_1, \vec{v}_2, \dots, \vec{v}_k$  are linearly independent. This is the value of lin. ind. vectors - there is no redundancy in how we express vectors in their span, using linear combinations.

Do  $\vec{v}_1, \vec{v}_2, \dots, \vec{v}_k \in \mathbb{R}^m$ : solve  $c_1 \vec{v}_1 + c_2 \vec{v}_2 + \dots + c_k \vec{v}_k = \vec{x}$  for  $c_1, c_2, \dots, c_k$ , whenever  $\vec{x}$  is given.  
solve for  $\vec{c}$ !

test:  $B \mid \vec{x}$



rref( $B$ )

if rref( $B$ ) has no zero rows, backsolve to find  $\vec{c}$ 's, regardless of  $\vec{x}$ , so  
 $\vec{v}_1, \vec{v}_2, \dots, \vec{v}_k$  span  $\mathbb{R}^m$

if rref( $B$ ) has a row of zeroes, then there are conditions  $\vec{x}$  must satisfy (to make the  $z_j$ 's = 0 in each row of rref( $B$ ) that = 0) so that  $\vec{x}$  is in the span.

Thus  $\vec{v}_1, \vec{v}_2, \dots, \vec{v}_k$  do NOT span  $\mathbb{R}^m$

this is guaranteed to happen if the number of vectors  $k$  is less than the  $m$  in  $\mathbb{R}^m$

$$m \left[ \begin{array}{c} \\ \\ \vdots \\ \\ k \end{array} \right] B$$

(5)

a subspace  $W$  of a vector space  $V$  is a special subset of  $V$  that is closed under addition ( $\alpha$ ) and closed under scalar multiplication ( $\beta$ ).

This means that  $W$  is itself a vector space.  
The two ways we have seen subspaces arise are

$$1) \quad W = \text{span} \{ \vec{v}_1, \vec{v}_2, \dots, \vec{v}_k \}$$

$$2) \quad W = \text{solution space to } A_{m \times n} \vec{x} = \vec{0}, \text{ i.e. } \{ \vec{x} \in \mathbb{R}^n \text{ s.t. } A\vec{x} = \vec{0} \}.$$

a basis of a vector space  $V$  (which could be a subspace of a larger vector space) is a linearly independent collection of vectors  $\vec{v}_1, \vec{v}_2, \dots, \vec{v}_k$

- 1) that  $\text{span } V$  and
- 2) are linearly independent.

the dimension of  $V$  is the number of vectors in a basis for  $V$

in other contexts, dimension may also be called "degrees of freedom". It is the number of scalar parameters <sup>needed</sup> to describe all elements of  $V$  in a non-redundant (i.e. unique) way.

$$\text{If } W = \text{span} \{ \vec{v}_1, \vec{v}_2, \dots, \vec{v}_k \}$$

eliminate dependent vectors  
to create a basis

in  $\mathbb{R}^m$ :

$$\left[ \begin{array}{c|c|c|c} \vec{v}_1 & \vec{v}_2 & \dots & \vec{v}_k \\ \hline 0 & 0 & \dots & 0 \end{array} \right]$$

$\downarrow$  rref

$$\left[ \begin{array}{c|c|c|c} \vec{w}_1 & \vec{w}_2 & \dots & \vec{w}_k \\ \hline 0 & 0 & \dots & 0 \end{array} \right]$$

read dependencies for  
 $\vec{v}_1, \dots, \vec{v}_k$  from those for

$$\vec{w}_1, \dots, \vec{w}_k$$

If  $W = \text{solution space to } A\vec{x} = \vec{0}$

$$A \mid \vec{0}$$

$\downarrow$  rref

$$\text{rref}(A) \mid \vec{0}$$

backsolve;  
put in linear combination form

$$\vec{x} = t_1 \vec{w}_1 + \dots + t_k \vec{w}_k \quad t_1, t_2, \dots, t_k \in \mathbb{R}$$

$\vec{w}_1, \vec{w}_2, \dots, \vec{w}_k$  will always be  
a basis for the solution space.