

## Announcements

October 5

- ▶ The midterm will be returned in recitation on Friday.
- ▶ Homeworks 2.2 and 2.3 are due Friday at 6am.
- ▶ Quiz on Friday: 2.1, 2.2, 2.3.
- ▶ Midterm 2 will take place in Recitation on Friday, 10/28.
- ▶ Office hours: today 1–2pm, tomorrow 3:30–4:30pm, and by appointment, in Skiles 221.
  - ▶ As always, TAs' office hours are posted on the website.
  - ▶ Math Lab is also a good place to visit.

## Section 2.9

### Dimension and Rank

## Coefficients of Basis Vectors

**Recall:** a **basis** of a subspace  $V$  is a set of vectors that *spans*  $V$  and is *linearly independent*.

**Lemma** ← like a theorem, but less important

If  $\mathcal{B} = \{v_1, v_2, \dots, v_m\}$  is a basis for a subspace  $V$ , then any vector  $x$  in  $V$  can be written as a linear combination

$$x = c_1 v_1 + c_2 v_2 + \dots + c_m v_m$$

for *unique* coefficients  $c_1, c_2, \dots, c_m$ .

We know  $x$  is a linear combination of the  $v_i$  because they span  $V$ . Suppose that we can write  $x$  as a linear combination with different coefficients:

$$x = c_1 v_1 + c_2 v_2 + \dots + c_m v_m$$

$$x = c'_1 v_1 + c'_2 v_2 + \dots + c'_m v_m$$

Subtracting:

$$0 = x - x = (c_1 - c'_1)v_1 + (c_2 - c'_2)v_2 + \dots + (c_m - c'_m)v_m$$

Since  $v_1, v_2, \dots, v_m$  are linearly independent, they only have the trivial linear dependence relation. That means each  $c_i - c'_i = 0$ , or  $c_i = c'_i$ .

## Bases as Coordinate Systems

The unit coordinate vectors  $e_1, e_2, \dots, e_n$  form a basis for  $\mathbf{R}^n$ . Any vector is a unique linear combination of the  $e_i$ :

$$v = \begin{pmatrix} 3 \\ 5 \\ -2 \end{pmatrix} = 3 \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} + 5 \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix} - 2 \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix} = 3e_1 + 5e_2 - 2e_3.$$

Observe: the *coordinates* of  $v$  are exactly the *coefficients* of  $e_1, e_2, e_3$ .

We can go backwards: given any basis  $\mathcal{B}$ , we interpret the coefficients of a linear combination as “coordinates” with respect to  $\mathcal{B}$ .

### Definition

Let  $\mathcal{B} = \{v_1, v_2, \dots, v_m\}$  be a basis of a subspace  $V$ . Any vector  $x$  in  $V$  can be written uniquely as a linear combination  $x = c_1 v_1 + c_2 v_2 + \dots + c_m v_m$ . The coefficients  $c_1, c_2, \dots, c_m$  are the **coordinates of  $x$  with respect to  $\mathcal{B}$** . The  **$\mathcal{B}$ -coordinate vector of  $x$**  is the vector

$$[x]_{\mathcal{B}} = \begin{pmatrix} c_1 \\ c_2 \\ \vdots \\ c_m \end{pmatrix} \quad \text{in } \mathbf{R}^m.$$

## Bases as Coordinate Systems

### Example 1

Let  $v_1 = \begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix}$ ,  $v_2 = \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}$ ,  $\mathcal{B} = \{v_1, v_2\}$ ,  $V = \text{Span}\{v_1, v_2\}$ .

*Verify* that  $\mathcal{B}$  is a basis:

*Span*: by definition  $V = \text{Span}\{v_1, v_2\}$ .

*Linearly independent*: because they are not multiples of each other.

**Question:** If  $[x]_{\mathcal{B}} = \begin{pmatrix} 5 \\ 2 \end{pmatrix}$ , then what is  $x$ ?

$$[x]_{\mathcal{B}} = \begin{pmatrix} 5 \\ 2 \end{pmatrix} \quad \text{means} \quad x = 5v_1 + 2v_2 = 5 \begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix} + 2 \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} = \begin{pmatrix} 7 \\ 2 \\ 7 \end{pmatrix}.$$

**Question:** Find the  $\mathcal{B}$ -coordinates of  $x = \begin{pmatrix} 5 \\ 3 \\ 5 \end{pmatrix}$ .

We have to solve the vector equation  $x = c_1 v_1 + c_2 v_2$  in the unknowns  $c_1, c_2$ .

$$\left( \begin{array}{cc|c} 1 & 1 & 5 \\ 0 & 1 & 3 \\ 1 & 1 & 5 \end{array} \right) \rightsquigarrow \left( \begin{array}{cc|c} 1 & 1 & 5 \\ 0 & 1 & 3 \\ 0 & 0 & 0 \end{array} \right) \rightsquigarrow \left( \begin{array}{cc|c} 1 & 0 & 2 \\ 0 & 1 & 3 \\ 0 & 0 & 0 \end{array} \right)$$

So  $x = 2v_1 + 3v_2$  and  $[x]_{\mathcal{B}} = \begin{pmatrix} 2 \\ 3 \end{pmatrix}$ .

## Bases as Coordinate Systems

### Example 2

Let  $v_1 = \begin{pmatrix} 2 \\ 3 \\ 2 \end{pmatrix}$ ,  $v_2 = \begin{pmatrix} -1 \\ 1 \\ 1 \end{pmatrix}$ ,  $v_3 = \begin{pmatrix} 2 \\ 8 \\ 6 \end{pmatrix}$ ,  $V = \text{Span}\{v_1, v_2, v_3\}$ .

**Question:** find a basis for  $V$ .

$V$  is the column span of the matrix

$$A = \begin{pmatrix} 2 & -1 & 2 \\ 3 & 1 & 8 \\ 2 & 1 & 6 \end{pmatrix} \xrightarrow{\text{row reduce}} \begin{pmatrix} 1 & 0 & 2 \\ 0 & 1 & 2 \\ 0 & 0 & 0 \end{pmatrix}.$$

A basis for the column span is formed by the pivot columns:  $\mathcal{B} = \{v_1, v_2\}$ .

**Question:** Find the  $\mathcal{B}$ -coordinates of  $x = \begin{pmatrix} 4 \\ 11 \\ 8 \end{pmatrix}$ .

We have to solve  $x = c_1 v_1 + c_2 v_2$ .

$$\left( \begin{array}{cc|c} 2 & -1 & 4 \\ 3 & 1 & 11 \\ 2 & 1 & 8 \end{array} \right) \xrightarrow{\text{row reduce}} \left( \begin{array}{cc|c} 1 & 0 & 3 \\ 0 & 1 & 2 \\ 0 & 0 & 0 \end{array} \right)$$

So  $x = 3v_1 + 2v_2$  and  $[x]_{\mathcal{B}} = \begin{pmatrix} 3 \\ 2 \end{pmatrix}$ .

## Bases as Coordinate Systems

### Summary

If  $\mathcal{B} = \{v_1, v_2, \dots, v_m\}$  is a basis for a subspace  $V$  and  $x$  is in  $V$ , then finding the  $\mathcal{B}$ -coordinates for  $x$  means solving the vector equation

$$x = c_1 v_1 + c_2 v_2 + \cdots + c_m v_m$$

in the unknowns  $c_1, c_2, \dots, c_m$ . These are the  $\mathcal{B}$ -coordinates. This (usually) means row reducing the augmented matrix

$$\left( \begin{array}{ccccc|c} | & | & & | & | \\ v_1 & v_2 & \cdots & v_m & x \\ | & | & & | & | \end{array} \right).$$

**Question:** what happens if you try to find the  $\mathcal{B}$ -coordinates of  $x$  *not* in  $V$ ? You end up with an inconsistent system:  $V$  is the span of  $v_1, v_2, \dots, v_m$ , and if  $x$  is not in the span, then  $x = c_1 v_1 + c_2 v_2 + \cdots + c_m v_m$  has no solution.

## Bases as Coordinate Systems

Picture

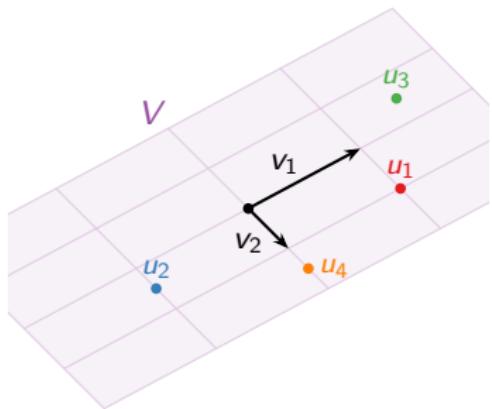
Let

$$v_1 = \begin{pmatrix} 2 \\ -1 \\ 1 \end{pmatrix} \quad v_2 = \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix}$$

These form a basis  $\mathcal{B}$  for the plane

$$V = \text{Span}\{v_1, v_2\}$$

in  $\mathbf{R}^3$ .



Question: estimate the  $\mathcal{B}$ -coordinates of these vectors:

$$[\textcolor{red}{u_1}]_{\mathcal{B}} = \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix} \quad [\textcolor{blue}{u_2}]_{\mathcal{B}} = \begin{pmatrix} -1 \\ \frac{1}{2} \\ 0 \end{pmatrix} \quad [\textcolor{green}{u_3}]_{\mathcal{B}} = \begin{pmatrix} \frac{3}{2} \\ -\frac{1}{2} \\ 0 \end{pmatrix} \quad [\textcolor{orange}{u_4}]_{\mathcal{B}} = \begin{pmatrix} 0 \\ \frac{3}{2} \\ 0 \end{pmatrix}$$

Remark

Many of you want to think of a plane in  $\mathbf{R}^3$  as "being"  $\mathbf{R}^2$ . Choosing a basis  $\mathcal{B}$  and using  $\mathcal{B}$ -coordinates is one way to make sense of that. But remember that the coordinates are the coefficients of a linear combination of the basis vectors.

## The Rank Theorem

Recall:

- ▶ The **dimension** of a subspace  $V$  is the number of vectors in a basis for  $V$ .
- ▶ A basis for the column space of a matrix  $A$  is given by the pivot columns.
- ▶ A basis for the null space of  $A$  is given by the vectors attached to the free variables in the parametric vector form.

### Definition

The **rank** of a matrix  $A$ , written  $\text{rank } A$ , is the dimension of the column space  $\text{Col } A$ .

Observe:

$\text{rank } A = \dim \text{Col } A = \text{the number of columns with pivots}$

$\dim \text{Nul } A = \text{the number of free variables}$

$= \text{the number of columns without pivots.}$

### Rank Theorem

If  $A$  is an  $m \times n$  matrix, then

$\text{rank } A + \dim \text{Nul } A = n = \text{the number of columns of } A.$

## The Rank Theorem

Continued

### Rank Theorem

If  $A$  is an  $m \times n$  matrix, then

$$\text{rank } A + \dim \text{Nul } A = n = \text{the number of columns of } A.$$

What does this mean? In the equation  $Ax = b$ ,

- ▶ You have some number of degrees of freedom in choosing  $b$  for which  $Ax = b$  is consistent (the column span).
- ▶ For a given  $b$  in the column span, you have some number of degrees of freedom in choosing  $x$  (the solution set).
- ▶ These two numbers always sum to  $n$ .

This is a nontrivial relationship between the *solution set of  $Ax = b$*  and the *space of all  $b$  such that  $Ax = b$  is consistent*.

### Example

If  $A = \begin{pmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{pmatrix}$ , then  $\text{rank } A = 1$  and  $\dim \text{Nul } A = 2 = 3 - 1$ .

## The Rank Theorem

### Example

Since the first two columns are a basis for Col A, the rank is 2, and any  $b$  in Col A can be written uniquely as

$$b = c_1 \begin{pmatrix} 1 \\ -2 \\ 2 \end{pmatrix} + c_2 \begin{pmatrix} 2 \\ -3 \\ 4 \end{pmatrix}.$$

So there are two degrees of freedom in choosing the possible  $b$ 's.

Since there are two free variables  $x_3, x_4$ , any solution to  $Ax = b$  (for  $b$  in  $\text{Col } A$ ) can be written uniquely in vector parametric form as

$$x = p + x_3 \begin{pmatrix} 8 \\ -4 \\ 1 \\ 0 \end{pmatrix} + x_4 \begin{pmatrix} 7 \\ -3 \\ 0 \\ 1 \end{pmatrix},$$

where  $p$  is a particular solution. There are two degrees of freedom in choosing  $x$ . The Rank Theorem says  $2 + 2 = 4$ .

Poll

Let  $A$  and  $B$  be  $3 \times 3$  matrices. Suppose that  $\text{rank}(A) = 2$  and  $\text{rank}(B) = 2$ . Is it possible that  $AB = 0$ ? Why or why not?

If  $AB = 0$ , then  $ABx = 0$  for every  $x$  in  $\mathbf{R}^3$ .

This means  $A(Bx) = 0$ , so  $Bx$  is in  $\text{Nul } A$ .

This is true for every  $x$ , so  $\text{Col } B$  is contained in  $\text{Nul } A$ .

But  $\dim \text{Nul } A = 1$  and  $\dim \text{Col } B = 2$ , and a 1-dimensional space can't contain a 2-dimensional space.

Hence it can't happen.

# The Basis Theorem

## Basis Theorem

Let  $V$  be a subspace of dimension  $m$ . Then:

- ▶ Any  $m$  linearly independent vectors in  $V$  form a basis for  $V$ .
- ▶ Any  $m$  vectors that span  $V$  form a basis for  $V$ .

In other words, if you *already* know that  $\dim V = m$ , then any  $m$  linearly independent vectors in  $V$  automatically span  $V$ , and any  $m$  vectors that span  $V$  are automatically linearly independent.

## Why?

- ▶ If you had  $m$  linearly independent vectors that don't form a basis, then they don't span. Hence you can find another vector in  $V$  but not in the span of these  $m$ , to get  $m + 1$  linearly independent vectors. The span of these has dimension  $m + 1$ . But a subspace of dimension  $m$  can't contain a subspace of larger dimension.
- ▶ If you had  $m$  vectors that span but don't form a basis, they're linearly dependent. This means you can remove a vector to get  $m - 1$  vectors that span  $V$ . This means  $\dim V < m$ .

# The Invertible Matrix Theorem

## Addenda

### The Invertible Matrix Theorem

Let  $A$  be a square  $n \times n$  matrix, and let  $T: \mathbf{R}^n \rightarrow \mathbf{R}^n$  be the linear transformation  $T(x) = Ax$ . The following statements are equivalent.

1.  $A$  is invertible.
2.  $T$  is invertible.
3.  $A$  is row equivalent to  $I_n$ .
4.  $A$  has  $n$  pivots.
5.  $Ax = 0$  has only the trivial solution.
6. The columns of  $A$  are linearly independent.
7.  $T$  is one-to-one.
8.  $Ax = b$  is consistent for all  $b$  in  $\mathbf{R}^n$ .
9. The columns of  $A$  span  $\mathbf{R}^n$ .
10.  $T$  is onto.
11.  $A$  has a left inverse (there exists  $B$  such that  $BA = I_n$ ).
12.  $A$  has a right inverse (there exists  $B$  such that  $AB = I_n$ ).
13.  $A^T$  is invertible.
14. The columns of  $A$  form a basis for  $\mathbf{R}^n$ .
15.  $\text{Col } A = \mathbf{R}^n$ .
16.  $\dim \text{Col } A = n$ .
17.  $\text{rank } A = n$ .
18.  $\text{Nul } A = \{0\}$ .
19.  $\dim \text{Nul } A = 0$ .

These are equivalent to the previous conditions by the Rank Theorem and the Basis Theorem. For instance, if the columns of  $A$  span  $\mathbf{R}^n$ , then because there are  $n$  columns and  $\dim \mathbf{R}^n = n$ , they form a basis. Hence  $\dim \text{Nul } A = 0$ .