

Math 1600B Lecture 20, Section 2, 26 Feb 2014

Announcements:

Read Section 3.6 for next class. Work through recommended [homework questions](#).

Extra Midterm Review: Friday, 4:30-6:00pm, MC110. Bring questions.

Midterm location is based on first letter of last name: HSB35 A-H, HSB236 I-Q, HSB240 R-Z. Be sure to write in the correct room!

Midterm: Saturday, March 1, 6:30pm-9:30pm. It will cover the material up to and including last lecture (not today's lecture), but not electrical networks. **Practice midterms** are on the website.

Tutorials: Midterm review this week. No quiz.

Help Centers: Monday-Friday 2:30-6:30 in MC106.

Intent to Register Open House for Actuarial & Statistical Sciences, Applied Math, Mathematics, Computer Science: Thursday, Feb 27, 3:30-5pm, MC320. Free pizza!

Partial review of Lectures 18 and 19:

Subspaces

Definition: A **subspace** of \mathbb{R}^n is any collection S of vectors in \mathbb{R}^n such that:

1. The zero vector $\vec{0}$ is in S .
2. S is **closed under addition**: If \vec{u} and \vec{v} are in S , then $\vec{u} + \vec{v}$ is in S .
3. S is **closed under scalar multiplication**: If \vec{u} is in S and c is any scalar, then $c\vec{u}$ is in S .

Basis

Definition: A **basis** for a subspace S of \mathbb{R}^n is a set of vectors $\vec{v}_1, \dots, \vec{v}_k$ such that:

1. $S = \text{span}(\vec{v}_1, \dots, \vec{v}_k)$, and
2. $\vec{v}_1, \dots, \vec{v}_k$ are linearly independent.

Subspaces associated with matrices

Definition: Let A be an $m \times n$ matrix.

1. The **row space** of A is the subspace $\text{row}(A)$ of \mathbb{R}^n spanned by the rows of A .
2. The **column space** of A is the subspace $\text{col}(A)$ of \mathbb{R}^m spanned by the columns of A .
3. The **null space** of A is the subspace $\text{null}(A)$ of \mathbb{R}^n consisting of the solutions to the system $A\vec{x} = \vec{0}$.

Theorem 3.20: Let A and R be row equivalent matrices. Then $\text{row}(A) = \text{row}(R)$.

Also, $\text{null}(A) = \text{null}(R)$. But **elementary row operations change the column space!** So $\text{col}(A) \neq \text{col}(R)$.

Theorem: If R is a matrix in row echelon form, then the nonzero rows of R form a basis for $\text{row}(R)$.

So if R is a row echelon form of A , then a basis for $\text{row}(A)$ is given by the nonzero rows of R .

Now, since $\text{null}(A) = \text{null}(R)$, the columns of R have the same dependency relationships as the columns of A .

It is easy to see that the pivot columns of R form a basis for $\text{col}(R)$, so the corresponding columns of A form a basis for $\text{col}(A)$.

We learned in Chapter 2 how to use R to find a basis for the **null space** of a matrix A , even though we didn't use this terminology.

Summary

Finding bases for $\text{row}(A)$, $\text{col}(A)$ and $\text{null}(A)$:

1. Find the reduced row echelon form R of A .
2. The nonzero rows of R form a basis for $\text{row}(A) = \text{row}(R)$.
3. The columns of A that correspond to the columns of R with leading 1's form a basis for $\text{col}(A)$.
4. Use back substitution to solve $R\vec{x} = \vec{0}$; the vectors that arise are a basis for $\text{null}(A) = \text{null}(R)$.

Row echelon form is in fact enough. Then you look at the columns with leading nonzero entries (the pivot columns).

These methods can be used to compute a basis for a subspace S spanned by some vectors $\vec{v}_1, \dots, \vec{v}_k$.

The **row method**:

1. Form the matrix A whose rows are $\vec{v}_1, \dots, \vec{v}_k$, so $S = \text{row}(A)$.
2. Reduce A to row echelon form R .
3. The nonzero rows of R will be a basis of $S = \text{row}(A) = \text{row}(R)$.

The **column method**:

1. Form the matrix A whose columns are $\vec{v}_1, \dots, \vec{v}_k$, so $S = \text{col}(A)$.
2. Reduce A to row echelon form R .
3. The columns of A that correspond to the columns of R with leading entries form a basis for $S = \text{col}(A)$.

New material

Dimension and Rank

We have seen that a subspace has many bases. [Have you noticed anything about the number of vectors in each basis?](#)

Theorem 3.23: Let S be a subspace of \mathbb{R}^n . Then any two bases for S have the same number of vectors.

Idea of proof:

Suppose that $\{\vec{u}_1, \vec{u}_2\}$ and $\{\vec{v}_1, \vec{v}_2, \vec{v}_3\}$ were both bases for S . We'll show that this is impossible, by showing that $\{\vec{v}_1, \vec{v}_2, \vec{v}_3\}$ is linearly dependent. Since $\{\vec{u}_1, \vec{u}_2\}$ is a basis, we can express each \vec{v}_i in terms of the \vec{u}_j 's:

$$\vec{v}_1 = a_{11}\vec{u}_1 + a_{21}\vec{u}_2$$

$$\vec{v}_2 = a_{12}\vec{u}_1 + a_{22}\vec{u}_2$$

$$\vec{v}_3 = a_{13}\vec{u}_1 + a_{23}\vec{u}_2$$

Then

$$\begin{aligned} & c_1\vec{v}_1 + c_2\vec{v}_2 + c_3\vec{v}_3 \\ &= c_1(a_{11}\vec{u}_1 + a_{21}\vec{u}_2) + c_2(a_{12}\vec{u}_1 + a_{22}\vec{u}_2) + c_3(a_{13}\vec{u}_1 + a_{23}\vec{u}_2) \\ &= (c_1a_{11} + c_2a_{12} + c_3a_{13})\vec{u}_1 + (c_1a_{21} + c_2a_{22} + c_3a_{23})\vec{u}_2 \end{aligned}$$

But the homogenous system

$$c_1a_{11} + c_2a_{12} + c_3a_{13} = 0$$

$$c_1a_{21} + c_2a_{22} + c_3a_{23} = 0$$

has nontrivial solutions! (Why?) Therefore, we can find nontrivial c_1, c_2, c_3 such that

$$c_1\vec{v}_1 + c_2\vec{v}_2 + c_3\vec{v}_3 = \vec{0} \quad \square$$

A very similar argument works for the general case.

Definition: The number of vectors in a basis for a subspace S is called the **dimension** of S , denoted $\dim S$.

Example: $\dim \mathbb{R}^n = n$

Example: If S is a line through the origin in \mathbb{R}^2 or \mathbb{R}^3 , then $\dim S = 1$

Example: If S is a plane through the origin in \mathbb{R}^3 , then $\dim S = 2$

Example: If $S = \text{span}\left(\begin{bmatrix} 3 \\ 0 \\ 2 \end{bmatrix}, \begin{bmatrix} -2 \\ 1 \\ 1 \end{bmatrix}, \begin{bmatrix} 1 \\ 1 \\ 3 \end{bmatrix}\right)$, then $\dim S = 2$

Example: Let A be the matrix from last class whose reduced row echelon form is

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

Then: $\dim \text{row}(A) = 3$ $\dim \text{col}(A) = 3$ $\dim \text{null}(A) = 2$

Note that $\dim \text{row}(A) = \text{rank}(A)$, since we defined the rank of A to be the number of nonzero rows in R . The above theorem shows that this number doesn't depend on how you row reduce A .

We call the dimension of the null space the **nullity** of A and write $\text{nullity}(A) = \dim \text{null}(A)$. This is what we called the "number of free variables" in Chapter 2.

From the way we find the basis for $\text{row}(A)$, $\text{col}(A)$ and $\text{null}(A)$, can you deduce any relationships between their dimensions?

Theorems 3.24 and 3.26: Let A be an $m \times n$ matrix. Then

$$\dim \text{row}(A) = \dim \text{col}(A) = \text{rank}(A)$$

and

$$\text{rank}(A) + \text{nullity}(A) = n.$$

Very important!

Questions:

True/false: for any A , $\text{rank}(A) = \text{rank}(A^T)$. True, since $\text{rank}(A) = \dim \text{row}(A) = \dim \text{col}(A^T) = \text{rank}(A^T)$.

True/false: if A is 2×5 , then the nullity of A is 3. False. We know that $\text{rank}(A) \leq 2$ and $\text{rank}(A) + \text{nullity}(A) = 5$, so $\text{nullity}(A) \geq 3$ (and ≤ 5).

True/false: if A is 5×2 , then $\text{nullity}(A) \geq 3$. False.
 $\text{rank}(A) + \text{nullity}(A) = 2$, so $\text{nullity}(A) = 0, 1$ or 2 .

Example: Find the nullity of

$$M = \begin{bmatrix} 1 & 2 & 3 & 4 & 5 & 6 & 7 \\ 8 & 9 & 10 & 11 & 12 & 13 & 14 \end{bmatrix}$$

and of M^T . **Any guesses?** The rows of M are linearly independent, so the rank is 2, so the nullity is $7 - 2 = 5$. The rank of M^T is also 2, so the nullity of M^T is $2 - 2 = 0$.

For larger matrices, you would compute the rank by row reduction.

Fundamental Theorem of Invertible Matrices, Version 2

Theorem 3.27: Let A be an $n \times n$ matrix. The following are equivalent:

- A is invertible.
- $A\vec{x} = \vec{b}$ has a unique solution for every $\vec{b} \in \mathbb{R}^n$.
- $A\vec{x} = \vec{0}$ has only the trivial (zero) solution.
- The reduced row echelon form of A is I_n .
- $\text{rank}(A) = n$
- $\text{nullity}(A) = 0$
- The columns of A are linearly independent.
- The columns of A span \mathbb{R}^n .
- The columns of A are a basis for \mathbb{R}^n .

Proof: We saw that (a), (b), (c) and (d) are equivalent in [Theorem 3.12](#). The new ones are easier:

(d) \iff (f): the only $n \times n$ matrix in row echelon form with n nonzero rows is I_n .

(f) \iff (g): follows from $\text{rank}(A) + \text{nullity}(A) = n$.

(c) \iff (h): easy.

(i) \implies (f) \implies (d) \implies (b) \implies (i): Explain.

(h) and (i) \iff (j): Clear.

In fact, since $\text{rank}(A) = \text{rank}(A^T)$, **all** of the statements are also equivalent to the statements with A replaced by A^T . In particular, we can add the following:

k. The rows of A are linearly independent.

l. The rows of A span \mathbb{R}^n .

m. The rows of A are a basis for \mathbb{R}^n .

Example 3.52: Show that the vectors $\begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}$, $\begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix}$ and $\begin{bmatrix} 4 \\ 9 \\ 7 \end{bmatrix}$ form a basis for \mathbb{R}^3 .

Solution: Show that matrix A with these vectors as the columns has rank 3. On board.

Not covering Theorem 3.28.

Questions

These questions weren't covered in lecture, but I leave them here for you to think about.

True/false: Every subspace of \mathbb{R}^3 has dimension 0, 1, 2 or 3.

True. A set of four or more vectors in \mathbb{R}^3 is always linearly dependent (why?), and so every basis for a subspace of \mathbb{R}^3 has at most three vectors.

True/false: If a matrix A has row echelon form

$$R = \begin{bmatrix} 0 & 2 & 3 \\ 0 & 0 & 4 \end{bmatrix}$$

then a basis for $\text{col}(A)$ is given by $\begin{bmatrix} 2 \\ 0 \end{bmatrix}$ and $\begin{bmatrix} 3 \\ 4 \end{bmatrix}$.

False. Those vectors are a basis for $\text{col}(R)$. To get a basis for $\text{col}(A)$, you take the second and third columns of A .

Question: What's a basis for $\text{row}(A)$?