

# Solving Linear Systems, Continued and The Inverse of a Matrix

Math 240 — Calculus III

Summer 2013, Session II

Monday, July 15, 2013



1. Solving Linear Systems
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**Gaussian elimination** solves a linear system by reducing to REF via elementary row ops and then using back substitution.

### Example

$$\begin{array}{r}
 3x_1 - 2x_2 + 2x_3 = 9 \\
 x_1 - 2x_2 + x_3 = 5 \\
 2x_1 - x_2 - 2x_3 = -1
 \end{array}
 \rightsquigarrow
 \begin{bmatrix}
 3 & -2 & 2 & 9 \\
 1 & -2 & 1 & 5 \\
 2 & -1 & -2 & -1
 \end{bmatrix}$$

$$\rightarrow
 \begin{bmatrix}
 1 & -2 & 1 & 5 \\
 0 & 1 & 3 & 5 \\
 0 & 0 & 1 & 2
 \end{bmatrix}
 \rightsquigarrow
 \begin{array}{r}
 x_1 - 2x_2 + x_3 = 5 \\
 x_2 + 3x_3 = 5 \\
 x_3 = 2
 \end{array}$$

### Steps

1.  $P_{12}$
2.  $A_{12}(-3)$
3.  $A_{13}(-2)$
4.  $A_{32}(-1)$
5.  $A_{23}(-3)$
6.  $M_3\left(\frac{-1}{13}\right)$

Back substitution gives the solution  $(1, -1, 2)$ .



Reducing the augmented matrix to RREF makes the system even easier to solve.

### Example

$$\begin{bmatrix} 1 & -2 & 1 & 5 \\ 0 & 1 & 3 & 5 \\ 0 & 0 & 1 & 2 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 0 & 0 & 1 \\ 0 & 1 & 0 & -1 \\ 0 & 0 & 1 & 2 \end{bmatrix} \rightsquigarrow \begin{array}{rcl} x_1 & = & 1 \\ x_2 & = & -1 \\ x_3 & = & 2 \end{array}$$

### Steps

1.  $A_{32}(-3)$
2.  $A_{31}(-1)$
3.  $A_{21}(2)$

Now, without any back substitution, we can see that the solution is  $(1, -1, 2)$ .

The method of solving a linear system by reducing its augmented matrix to RREF is called **Gauss-Jordan elimination**.



## Definition

The **rank** of a matrix,  $A$ , is the number of nonzero rows it has after reduction to REF. It is denoted by  $\text{rank}(A)$ .

If  $A$  the coefficient matrix of an  $m \times n$  linear system and  $\text{rank}(A^\#) = \text{rank}(A) = n$  then the REF looks like

$$\left[ \begin{array}{cccccc} 1 & * & * & \cdots & * & \\ & 1 & * & \cdots & * & \\ & & \ddots & & \vdots & \\ 0 & & & & 1 & * \\ 0 & \dots\dots\dots & & & & 0 \end{array} \right] \rightsquigarrow \begin{array}{l} x_1 = * \\ x_2 = * \\ \vdots \\ x_n = * \end{array}$$

## Lemma

Suppose  $Ax = \mathbf{b}$  is an  $m \times n$  linear system with augmented matrix  $A^\#$ . If  $\text{rank}(A^\#) = \text{rank}(A) = n$  then the system has a unique solution.



## Example

Determine the solution set of the linear system

$$\begin{aligned}x_1 + x_2 - x_3 + x_4 &= 1, \\2x_1 + 3x_2 + x_3 &= 4, \\3x_1 + 5x_2 + 3x_3 - x_4 &= 5.\end{aligned}$$

Reduce the augmented matrix.

$$\left[ \begin{array}{ccccc|c} 1 & 1 & -1 & 1 & 1 & 1 \\ 2 & 3 & 1 & 0 & 4 & 4 \\ 3 & 5 & 3 & -1 & 5 & 5 \end{array} \right] \xrightarrow{\substack{A_{12}(-2) \\ A_{13}(-3) \\ A_{23}(-2)}} \left[ \begin{array}{ccccc|c} 1 & 1 & -1 & 1 & 1 & 1 \\ 0 & 1 & 3 & -2 & 2 & 2 \\ 0 & 0 & 0 & 0 & -2 & -2 \end{array} \right]$$

The last row says  $0 = -2$ ; the system is inconsistent.

## Lemma

Suppose  $Ax = \mathbf{b}$  is a linear system with augmented matrix  $A^\#$ . If  $\text{rank}(A^\#) > \text{rank}(A)$  then the system is inconsistent.



## Example

Determine the solution set of the linear system

$$5x_1 - 6x_2 + x_3 = 4,$$

$$2x_1 - 3x_2 + x_3 = 1,$$

$$4x_1 - 3x_2 - x_3 = 5.$$

Reduce the augmented matrix.

$$\begin{bmatrix} 5 & -6 & 1 & 4 \\ 2 & -3 & 1 & 1 \\ 4 & -3 & -1 & 5 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 0 & -1 & 2 \\ 0 & 1 & -1 & 1 \\ 0 & 0 & 0 & 0 \end{bmatrix} \rightsquigarrow \begin{array}{r} x_1 - x_3 = 2 \\ x_2 - x_3 = 1 \end{array}$$

The unknown  $x_3$  can assume any value. Let  $x_3 = t$ . Then by back substitution we get  $x_2 = t + 1$  and  $x_1 = t + 2$ . Thus, the solution set is the line

$$\{(t + 2, t + 1, t) : t \in \mathbb{R}\}.$$



## Definition

When an unknown variable in a linear system is free to assume any value, we call it a **free variable**. Variables that are not free are called **bound variables**.

The value of a bound variable is uniquely determined by a choice of values for all of the free variables in the system.

## Lemma

*Suppose  $A\mathbf{x} = \mathbf{b}$  is an  $m \times n$  linear system with augmented matrix  $A^\#$ . If  $\text{rank}(A^\#) = \text{rank}(A) < n$  then the system has an infinite number of solutions. Such a system will have  $n - \text{rank}(A^\#)$  free variables.*





# Solving linear systems with free variables

Solving Linear  
Systems

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Gauss-Jordan  
elimination

Rank

Inverse  
matrices

Definition

Computing  
inverses

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inverses

Using inverse  
matrices

Conclusion

## Example

Use Gaussian elimination to solve

$$\begin{aligned}x_1 + 2x_2 - 2x_3 - x_4 &= 3, \\3x_1 + 6x_2 + x_3 + 11x_4 &= 16, \\2x_1 + 4x_2 - x_3 + 4x_4 &= 9.\end{aligned}$$

Reducing to row-echelon form yields

$$\begin{aligned}x_1 + 2x_2 - 2x_3 - x_4 &= 3, \\x_3 + 2x_4 &= 1.\end{aligned}$$

Choose as free variables those variables that **do not** have a pivot in their column.

In this case, our free variables will be  $x_2$  and  $x_4$ . The solution set is the plane

$$\{(5 - 2s - 3t, s, 1 - 2t, t) : s, t \in \mathbb{R}\}.$$



# The inverse of a square matrix

Can we divide by a matrix? What properties should the inverse matrix have?

## Definition

Suppose  $A$  is a square,  $n \times n$  matrix. An **inverse matrix** for  $A$  is an  $n \times n$  matrix,  $B$ , such that

$$AB = I_n \quad \text{and} \quad BA = I_n.$$

If  $A$  has such an inverse then we say that it is **invertible** or **nonsingular**. Otherwise, we say that  $A$  is **singular**.

## Remark

**Not every matrix is invertible.**

If you have a linear system  $A\mathbf{x} = \mathbf{b}$  and  $B$  is an inverse matrix for  $A$  then the linear system has the unique solution

$$\mathbf{x} = B\mathbf{b}.$$



## The inverse of a square matrix

## Example

If

$$A = \begin{bmatrix} 1 & -1 & 2 \\ 2 & -3 & 3 \\ 1 & -1 & 1 \end{bmatrix} \quad \text{and} \quad B = \begin{bmatrix} 0 & -1 & 3 \\ 1 & -1 & 1 \\ 1 & 0 & -1 \end{bmatrix} = A^{-1}$$

then  $B$  is *the* inverse of  $A$ .

## Theorem (Matrix inverses are well-defined)

Suppose  $A$  is an  $n \times n$  matrix. If  $B$  and  $C$  are two inverses of  $A$  then  $B = C$ .

Thus, we can write  $A^{-1}$  for *the* inverse of  $A$  with no ambiguity.

## Useful Example

$$\text{If } A = \begin{bmatrix} a & b \\ c & d \end{bmatrix} \text{ and } ad - bc \neq 0 \text{ then } A^{-1} = \frac{1}{ad-bc} \begin{bmatrix} d & -b \\ -c & a \end{bmatrix}.$$



## Finding the inverse of a matrix

Inverse matrices sound great! How do I find one?

Suppose  $A$  is a  $3 \times 3$  invertible matrix. If  $A^{-1} = [\mathbf{x}_1 \quad \mathbf{x}_2 \quad \mathbf{x}_3]$  then

$$A\mathbf{x}_1 = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \quad A\mathbf{x}_2 = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}, \quad \text{and} \quad A\mathbf{x}_3 = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}.$$

We can find  $A^{-1}$  by solving 3 linear systems at once!

In general, form the augmented matrix and reduce to RREF.  
You end up with  $A^{-1}$  on the right.

$$\left[ A \mid I_n \right] \rightsquigarrow \left[ I_n \mid A^{-1} \right]$$



## Finding the inverse of a matrix

## Example

Let's find the inverse of  $A = \begin{bmatrix} 1 & -1 & 2 \\ 2 & -3 & 3 \\ 1 & -1 & 1 \end{bmatrix}$ .

Take the augmented matrix and row reduce.

$$\left[ \begin{array}{ccc|ccc} 1 & -1 & 2 & 1 & 0 & 0 \\ 2 & -3 & 3 & 0 & 1 & 0 \\ 1 & -1 & 1 & 0 & 0 & 1 \end{array} \right] \rightsquigarrow \left[ \begin{array}{ccc|ccc} 1 & 0 & 0 & 0 & -1 & 3 \\ 0 & 1 & 0 & 1 & -1 & 1 \\ 0 & 0 & 1 & 1 & 0 & -1 \end{array} \right]$$

$\underbrace{\hspace{10em}}_{A^{-1}}$

## Steps

1.  $A_{12}(-2)$
2.  $A_{13}(-1)$
3.  $M_2(-1)$
4.  $M_3(-1)$
5.  $A_{32}(-1)$
6.  $A_{31}(-2)$
7.  $A_{21}(1)$



## Finding the inverse of a matrix

In order to find the inverse of a matrix,  $A$ , we row reduced an augmented matrix with  $A$  on the left. What if we don't end up with  $I_n$  on the left?

## Theorem

*An  $n \times n$  matrix,  $A$ , is invertible if and only if  $\text{rank}(A) = n$ .*

## Example

Find the inverse of the matrix  $A = \begin{bmatrix} 1 & 3 \\ 2 & 6 \end{bmatrix}$ .

Try to reduce the matrix to RREF.

$$\begin{bmatrix} 1 & 3 \\ 2 & 6 \end{bmatrix} \xrightarrow{A_{12}(-2)} \begin{bmatrix} 1 & 3 \\ 0 & 0 \end{bmatrix}$$

Since  $\text{rank}(A) < 2$ , we conclude that  $A$  is not invertible.  
Notice that  $(1)(6) - (3)(2) = 0$ .



## Finding the inverse of a matrix

Diagonal matrices have simple inverses.

### Proposition

*The inverse of a diagonal matrix is the diagonal matrix with reciprocal entries.*

$$\begin{bmatrix} a_{11} & & 0 \\ & \ddots & \\ 0 & & a_{nn} \end{bmatrix}^{-1} = \begin{bmatrix} a_{11}^{-1} & & 0 \\ & \ddots & \\ 0 & & a_{nn}^{-1} \end{bmatrix}$$

Upper and lower triangular matrices have inverses of the same form.

### Proposition

*The inverse of an upper triangular matrix is upper triangular.  
The inverse of a lower triangular matrix is lower triangular.*



Suppose  $A$  and  $B$  are  $n \times n$  invertible matrices.

- ▶  $A^{-1}$  is invertible and  $(A^{-1})^{-1} = A$ .
- ▶  $AB$  is invertible and  $(AB)^{-1} = B^{-1}A^{-1}$ .
- ▶  $A^T$  is invertible and  $(A^T)^{-1} = (A^{-1})^T$ .

### Corollary

Suppose  $A_1, A_2, \dots, A_k$  are invertible  $n \times n$  matrices. Then their product,  $A_1A_2 \cdots A_k$  is invertible, and

$$(A_1A_2 \cdots A_k)^{-1} = A_k^{-1}A_{k-1}^{-1} \cdots A_1^{-1}.$$





Recall that if  $A$  is an invertible matrix then the linear system  $A\mathbf{x} = \mathbf{b}$  has the unique solution  $\mathbf{x} = A^{-1}\mathbf{b}$ .

### Example

Solve the linear system

$$\begin{aligned}x_1 + 3x_2 &= 1, \\2x_1 + 5x_2 &= 3.\end{aligned}$$

The coefficient matrix is  $A = \begin{bmatrix} 1 & 3 \\ 2 & 5 \end{bmatrix}$ , so  $A^{-1} = \begin{bmatrix} -5 & 3 \\ 2 & -1 \end{bmatrix}$ .

The inverse of a  $2 \times 2$  matrix is

$$\begin{bmatrix} a & b \\ c & d \end{bmatrix}^{-1} = \frac{1}{ad - bc} \begin{bmatrix} d & -b \\ -c & a \end{bmatrix} \text{ when } ad - bc \neq 0.$$

$$\text{Hence, } \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} -5 & 3 \\ 2 & -1 \end{bmatrix} \begin{bmatrix} 1 \\ 3 \end{bmatrix} = \begin{bmatrix} 4 \\ -1 \end{bmatrix}.$$



Inverse matrices are an elegant way of solving linear systems. They do have some drawbacks:

- ▶ They are only applicable when the coefficient matrix is square.
- ▶ Even in the case of a square matrix, an inverse may not exist.
- ▶ They are hard to compute, at least as complicated as doing Gauss-Jordan elimination.

However, they can be useful if

- ▶ the coefficient matrix has an obvious inverse,
- ▶ you need to solve multiple linear systems with the same coefficients.

