

Lecture 11 : The Fundamental Theorem of Invertible Matrices

Math 3013
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Agenda:

1. Matrix Inverses
2. The Calculation of a Matrix Inverse
3. The Fundamental Theorem of Invertible Matrices

Matrix Inverses

Definition

Suppose \mathbf{A} is an $n \times n$ matrix. Then any matrix \mathbf{B} such that

$$\mathbf{BA} = \mathbf{I} \quad \text{and} \quad \mathbf{AB} = \mathbf{I}$$

is called a **matrix inverse** of \mathbf{A} .

Theorem

If a square matrix \mathbf{A} has an inverse, it is unique.

Notation: If \mathbf{A} is invertible, we will denote its (unique) inverse by \mathbf{A}^{-1}

Calculating Matrix Inverses - the underlying idea

- ▶ Uniqueness of \mathbf{A}^{-1} :
If we can find a matrix \mathbf{C} such that $\mathbf{CA} = \mathbf{I}$, then $\mathbf{C} = \mathbf{A}^{-1}$
- ▶ **Elementary matrices** allow one to implement elementary row operations via matrix multiplication

$$\text{If } \mathbf{E}_{\mathcal{R}} := \mathcal{R}(\mathbf{I}), \text{ then } \mathcal{R}(\mathbf{A}) = \mathbf{E}_{\mathcal{R}}\mathbf{A}$$

- ▶ If \mathbf{A} can be row reduced to \mathbf{I} , then there will be a sequence of elementary row operations $\mathcal{R}_1, \mathcal{R}_2, \dots, \mathcal{R}_k$ that transform \mathbf{A} to \mathbf{I} :

$$\begin{aligned}\mathbf{I} &= \mathcal{R}_k(\cdots \mathcal{R}_2(\mathcal{R}_1(\mathbf{A}))) \\ &= \mathbf{E}_{\mathcal{R}_k} \cdots \mathbf{E}_{\mathcal{R}_2} \mathbf{E}_{\mathcal{R}_1} \mathbf{A}\end{aligned}$$

$$\Rightarrow \quad \mathbf{A}^{-1} = \mathbf{E}_{\mathcal{R}_k} \cdots \mathbf{E}_{\mathcal{R}_2} \mathbf{E}_{\mathcal{R}_1}$$

So all we need a procedure for directly calculating

$$\mathbf{A}^{-1} = \mathbf{E}_{\mathcal{R}_k} \cdots \mathbf{E}_{\mathcal{R}_2} \mathbf{E}_{\mathcal{R}_1}$$

Algorithm for Calculating \mathbf{A}^{-1}

Theorem

Suppose \mathbf{A} is an $n \times n$ matrix and let $[\mathbf{A}''|\mathbf{I}''] = R.R.E.F.([\mathbf{A}|\mathbf{I}])$.

- ▶ If $\mathbf{A}'' = \mathbf{I}$, then $\mathbf{A}^{-1} = \mathbf{I}''$.
- ▶ If $\mathbf{A}'' \neq \mathbf{I}$, then \mathbf{A} has no inverse.

Calculating Inverse Matrices: Example

Find the inverse of

$$\mathbf{A} = \begin{bmatrix} 2 & 1 & 4 \\ 3 & 2 & 5 \\ 0 & -1 & 1 \end{bmatrix}$$

We have

$$[\mathbf{A}|\mathbf{I}] = \left[\begin{array}{ccc|ccc} 2 & 1 & 4 & 1 & 0 & 0 \\ 3 & 2 & 5 & 0 & 1 & 0 \\ 0 & -1 & 1 & 0 & 0 & 1 \end{array} \right]$$

$$\downarrow \quad R_2 \rightarrow R_2 - \frac{3}{2}R_1$$

$$\left[\begin{array}{ccc|ccc} 2 & 1 & 4 & 1 & 0 & 0 \\ 0 & \frac{1}{2} & -1 & -\frac{3}{2} & 1 & 0 \\ 0 & -1 & 1 & 0 & 0 & 1 \end{array} \right]$$

$$\left[\begin{array}{ccc|ccc} 2 & 1 & 4 & 1 & 0 & 0 \\ 0 & \frac{1}{2} & -1 & -\frac{3}{2} & 1 & 0 \\ 0 & -1 & 1 & 0 & 0 & 1 \end{array} \right]$$

$$\downarrow \quad R_2 \rightarrow 2R_2$$

$$\left[\begin{array}{ccc|ccc} 2 & 1 & 4 & 1 & 0 & 0 \\ 0 & 1 & -2 & -3 & 2 & 0 \\ 0 & -1 & 1 & 0 & 0 & 1 \end{array} \right]$$

$$\downarrow \quad R_3 \rightarrow R_3 + R_2$$

$$\left[\begin{array}{ccc|ccc} 2 & 1 & 4 & 1 & 0 & 0 \\ 0 & 1 & -2 & -3 & 2 & 0 \\ 0 & 0 & -1 & -3 & 2 & 1 \end{array} \right] = R.E.F. ([\mathbf{A}|\mathbf{I}])$$

$$\left[\begin{array}{ccc|ccc} 2 & 1 & 4 & 1 & 0 & 0 \\ 0 & 1 & -2 & -3 & 2 & 0 \\ 0 & 0 & -1 & -3 & 2 & 1 \end{array} \right]$$

$$\downarrow \quad R_3 \rightarrow -R_3$$

$$\left[\begin{array}{ccc|ccc} 2 & 1 & 4 & 1 & 0 & 0 \\ 0 & 1 & -2 & -3 & 2 & 0 \\ 0 & 0 & 1 & 3 & -2 & -1 \end{array} \right]$$

$$\downarrow \quad R_2 + 2R_3$$

$$\left[\begin{array}{ccc|ccc} 2 & 1 & 4 & 1 & 0 & 0 \\ 0 & 1 & 0 & 3 & -2 & -2 \\ 0 & 0 & 1 & 3 & -2 & -1 \end{array} \right]$$

$$\left[\begin{array}{ccc|ccc} 2 & 1 & 4 & 1 & 0 & 0 \\ 0 & 1 & 0 & 3 & -2 & -2 \\ 0 & 0 & 1 & 3 & -2 & -1 \end{array} \right]$$

$$\downarrow \quad R_1 \rightarrow R_1 - 4R_3$$

$$\left[\begin{array}{ccc|ccc} 2 & 1 & 0 & -11 & 8 & 4 \\ 0 & 1 & 0 & 3 & -2 & -2 \\ 0 & 0 & 1 & 3 & -2 & -1 \end{array} \right]$$

$$\downarrow \quad R_1 \rightarrow R_1 - R_2$$

$$\left[\begin{array}{ccc|ccc} 2 & 0 & 0 & -14 & 10 & 6 \\ 0 & 1 & 0 & 3 & -2 & -2 \\ 0 & 0 & 1 & 3 & -2 & -1 \end{array} \right]$$

$$\left[\begin{array}{ccc|ccc} 2 & 0 & 0 & -14 & 10 & 6 \\ 0 & 1 & 0 & 3 & -2 & -2 \\ 0 & 0 & 1 & 3 & -2 & -1 \end{array} \right]$$

$$\downarrow \quad R_1 \rightarrow \frac{1}{2}R_1$$

$$\left[\begin{array}{ccc|ccc} 1 & 0 & 0 & -7 & 5 & 3 \\ 0 & 1 & 0 & 3 & -2 & -2 \\ 0 & 0 & 1 & 3 & -2 & -1 \end{array} \right]$$

$$R.R.E.F. ([\mathbf{A}|\mathbf{I}]) = \left[\begin{array}{ccc|ccc} 1 & 0 & 0 & -7 & 5 & 3 \\ 0 & 1 & 0 & 3 & -2 & -2 \\ 0 & 0 & 1 & 3 & -2 & -1 \end{array} \right] = [\mathbf{I}|\mathbf{A}^{-1}]$$

and so

$$\mathbf{A}^{-1} = \begin{bmatrix} -7 & 5 & 3 \\ 3 & -2 & -2 \\ 3 & -2 & -1 \end{bmatrix}$$

The Fundamental Theorem of Invertible Matrices

The following theorem shows the close connections between various problems we have considered.

Theorem

Let \mathbf{A} be an $n \times n$ matrix. The following statements are equivalent (i.e., if any one of these statements is true, then all are true).

- (a) \mathbf{A}^{-1} exists.
- (b) Every linear system $\mathbf{Ax} = \mathbf{b}$ has a unique solution.
- (c) $\mathbf{Ax} = \mathbf{0}$ has only the trivial solution $\mathbf{x} = \mathbf{0}$.
- (d) The R.R.E.F. of \mathbf{A} is the $n \times n$ identity matrix.
- (e) \mathbf{A} is a product of elementary matrices.

Proof: We'll demonstrate the following chain of implications

$$(a) \Rightarrow (b) \Rightarrow (c) \Rightarrow (d) \Rightarrow (e) \Rightarrow (a)$$

(a) \Rightarrow (b)

To show : if \mathbf{A}^{-1} exists, then $\mathbf{Ax} = \mathbf{b}$ has a unique solution

Suppose \mathbf{A} is invertible, with inverse \mathbf{A}^{-1} .

I claim $\mathbf{x} = \mathbf{A}^{-1}\mathbf{b}$ is a solution of $\mathbf{Ax} = \mathbf{b}$. Indeed,

$$\mathbf{Ax} = \mathbf{A}(\mathbf{A}^{-1}\mathbf{b}) = (\mathbf{AA}^{-1})\mathbf{b} = \mathbf{Ib} = \mathbf{b}$$

and so $\mathbf{x} = \mathbf{A}^{-1}\mathbf{b}$ is a solution.

Now suppose \mathbf{y} is any other solution of $\mathbf{Ax} = \mathbf{b}$. Then

$$\begin{aligned}\mathbf{Ay} = \mathbf{b} &\quad \Rightarrow \quad \mathbf{A}^{-1}\mathbf{Ay} = \mathbf{A}^{-1}\mathbf{b} \\ &\quad \Rightarrow \quad \mathbf{Iy} = \mathbf{A}^{-1}\mathbf{b} \\ &\quad \Rightarrow \quad \mathbf{y} = \mathbf{A}^{-1}\mathbf{b}\end{aligned}$$

(so \mathbf{y} is the same solution as we had before.)



(b) \Rightarrow (c)

To show : if linear systems $\mathbf{Ax} = \mathbf{b}$ have unique solutions for each \mathbf{b} , then the only solution of $\mathbf{Ax} = \mathbf{0}$ is $\mathbf{x} = \mathbf{0}$.

Suppose every linear system $\mathbf{Ax} = \mathbf{b}$ has a unique solution.

Then if we choose $\mathbf{b} = \mathbf{0}$, then $\mathbf{Ax} = \mathbf{0}$ has a unique solution.

On the other hand, $\mathbf{x} = \mathbf{0}$ is obviously a solution of $\mathbf{Ax} = \mathbf{0}$.

Since by hypothesis the solution of such an equation is unique, we conclude that $\mathbf{Ax} = \mathbf{0}$ has $\mathbf{x} = \mathbf{0}$ as its unique solution.



(c) \Rightarrow (d)

To show : if $\mathbf{x} = \mathbf{0}$ is the only solution of $\mathbf{Ax} = \mathbf{0}$, then \mathbf{A} is row reducible to the identity matrix

This follows from our row reduction algorithm for solving linear systems.

Assume $\mathbf{x} = \mathbf{0}$ is the unique solution to $\mathbf{Ax} = \mathbf{0}$

Then the solution equation for \mathbf{x} corresponds to the augmented matrix

$$\left[\begin{array}{ccccc|c} 1 & 0 & \cdots & \cdots & 0 & 0 \\ 0 & 1 & \cdots & \cdots & 0 & 0 \\ \vdots & & \ddots & & \vdots & \vdots \\ 0 & 0 & \cdots & 1 & 0 & 0 \\ 0 & 0 & \cdots & 0 & 1 & 0 \end{array} \right]$$

for which the coefficient part (the “ \mathbf{A} -part” of the augmented matrix) is the identity matrix \mathbf{I} .

(c) \Rightarrow (d) , Cont'd

On the other hand, since all solutions to a linear system $\mathbf{Ax} = \mathbf{0}$ are obtainable by row reducing the augmented matrix $[\mathbf{A} \mid \mathbf{0}]$ to its unique Reduced Row Echelon Form, we can conclude that \mathbf{A} must be row reducible to the identity matrix.



(d) \Rightarrow (e)

To show : if \mathbf{A} can be row reduced to \mathbf{I} , then \mathbf{A} is a product of elementary matrices

Assume there is a sequence $\mathcal{R}_1, \mathcal{R}_2, \dots, \mathcal{R}_k$ of elementary row operations that systematically converts \mathbf{A} to the identity matrix \mathbf{I} . Say,

$$\mathcal{R}_k (\mathcal{R}_{k-1} (\cdots (\mathcal{R}_1 (\mathbf{A})))) = \mathbf{I}$$

or, implementing the row operations via matrix multiplication by elementary matrices,

$$\mathbf{E}_{\mathcal{R}_k} \mathbf{E}_{\mathcal{R}_{k-1}} \cdots \mathbf{E}_{\mathcal{R}_1} \mathbf{A} = \mathbf{I}$$

This tells us that

$$\mathbf{A}^{-1} = \mathbf{E}_{\mathcal{R}_k} \mathbf{E}_{\mathcal{R}_{k-1}} \cdots \mathbf{E}_{\mathcal{R}_1}$$

So \mathbf{A}^{-1} is a product of elementary matrices

(d) \Rightarrow (e), Cont'd

Next we use two facts

Lemma

If $\mathbf{E}_{\mathcal{R}}$ is an elementary matrix, then $\mathbf{E}_{\mathcal{R}}$ is invertible and $(\mathbf{E}_{\mathcal{R}})^{-1}$ is another elementary matrix.

Lemma

If $\mathbf{A}_1, \mathbf{A}_2, \dots, \mathbf{A}_{\ell}$ are invertible matrices, then the matrix product $\mathbf{A}_1 \mathbf{A}_2 \cdots \mathbf{A}_{\ell}$ is invertible and

$$(\mathbf{A}_1 \mathbf{A}_2 \cdots \mathbf{A}_{\ell})^{-1} = (\mathbf{A}_{\ell})^{-1} \cdots (\mathbf{A}_2)^{-1} (\mathbf{A}_1)^{-1}$$

Thus,

$$\begin{aligned} \mathbf{A} &= (\mathbf{A}^{-1})^{-1} \\ &= (\mathbf{E}_{\mathcal{R}_k} \mathbf{E}_{\mathcal{R}_{k-1}} \cdots \mathbf{E}_{\mathcal{R}_1})^{-1} \\ &= \mathbf{E}_{\mathcal{R}_1}^{-1} \mathbf{E}_{\mathcal{R}_2}^{-1} \cdots \mathbf{E}_{\mathcal{R}_k}^{-1} \end{aligned}$$

and so \mathbf{A} is a product of elementary matrices.

(e) \Rightarrow (a)

To show: if \mathbf{A} is a product of elementary matrices, then \mathbf{A} is invertible.

Suppose

$$\mathbf{A} = \mathbf{E}_{\mathcal{R}_1} \cdots \mathbf{E}_{\mathcal{R}_k}$$

Then because each of the matrix factor $\mathbf{E}_{\mathcal{R}_i}$ is invertible, \mathbf{A} is also invertible and, moreover,

$$\mathbf{A}^{-1} = (\mathbf{E}_{\mathcal{R}_1} \mathbf{E}_{\mathcal{R}_2} \cdots \mathbf{E}_{\mathcal{R}_k})^{-1} = (\mathbf{E}_{\mathcal{R}_k})^{-1} \cdots (\mathbf{E}_{\mathcal{R}_2})^{-1} (\mathbf{E}_{\mathcal{R}_1})^{-1}$$

Having completed the chain of implications

$$(a) \Rightarrow (b) \Rightarrow (c) \Rightarrow (d) \Rightarrow (e) \Rightarrow (a)$$

the theorem is now proved. □

Overview of Material to be Covered on First Exam

- ▶ Vectors as elements of \mathbb{R}^n
 - ▶ Vector addition
 - ▶ Scalar multiplication
 - ▶ The dot product
- ▶ Simple Geometric Constructs: Points, Lines, Planes and Hyperplanes as Sets of Vectors

$$\mathcal{H} = \{\mathbf{x} \in \mathbb{R}^n \mid \mathbf{x} = \mathbf{p}_0 + t_1\mathbf{v}_1 + t_2\mathbf{v}_2 \cdots t_k\mathbf{v}_k; t_1, t_2, \dots, t_k \in \mathbb{R}\}$$

- ▶ Matrices and Matrix Algebra
 - ▶ Matrix Addition
 - ▶ Scalar Multiplication of Matrices
 - ▶ Matrix Multiplication
 - ▶ Matrix Transposes

Overview, Cont'd

► Linear Systems

- The geometry of solution sets of linear systems
 - solution set of single linear equation in m unknowns is a $(m - 1)$ -dimensional hyperplane in \mathbb{R}^m
 - solution set of n equations in m -unknowns is the intersection of n hyperplanes in \mathbb{R}^m
- Naive Expectation: Given n linear equations in m unknowns, one expects the solutions to form a $(m - n)$ -dimensional hyperplane in \mathbb{R}^m

Exceptions:

- linear systems with redundant equations
 - linear systems with internal contradictions
- Matrix formulation: **$\mathbf{Ax} = \mathbf{b}$**
- Solving linear systems via row reduction
 - Augmented Matrices **$[\mathbf{A} \mid \mathbf{b}]$**
 - Row Echelon Form
 - Reduced Row Echelon Form
 - Fixed Variables and Free Variables
 - Solution Set as a Hyperplane

Overview, Cont'd

► Matrix Inverses

► Calculating \mathbf{A}^{-1}

► The Fundamental Theorem of Matrix Inverses

Theorem: Let \mathbf{A} be an $n \times n$ matrix. The following statements are equivalent

- (a) \mathbf{A} is invertible.
- (b) Every linear system $\mathbf{Ax} = \mathbf{b}$ has a unique solution.
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