

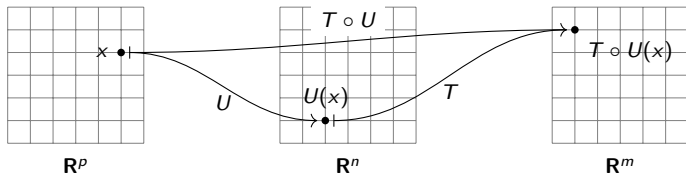
Review for Chapter 2

Selected Topics

Matrix Multiplication/Inversion and Linear Transformations

Let $T: \mathbf{R}^n \rightarrow \mathbf{R}^m$ and $U: \mathbf{R}^p \rightarrow \mathbf{R}^n$ be linear transformations with matrices A and B . The **composition** is the linear transformation

$$T \circ U: \mathbf{R}^p \rightarrow \mathbf{R}^m \quad \text{defined by} \quad T \circ U(x) = T(U(x)).$$



Fact: The matrix for $T \circ U$ is AB .

Now let $T: \mathbf{R}^n \rightarrow \mathbf{R}^n$ be an *invertible* linear transformation. This means there is a linear transformation $T^{-1}: \mathbf{R}^n \rightarrow \mathbf{R}^n$ such that $T \circ T^{-1}(x) = x$ for all x in \mathbf{R}^n . Equivalently, it means T is one-to-one and onto.

Fact: If A is the matrix for T , then A^{-1} is the matrix for T^{-1} .

Matrix Multiplication/Inversion and Linear Transformations

Example

Let $T: \mathbf{R}^2 \rightarrow \mathbf{R}^2$ scale the x -axis by 2, and let $U: \mathbf{R}^2 \rightarrow \mathbf{R}^2$ be counterclockwise rotation by 90° .

Their matrices are:

$$A = \begin{pmatrix} 2 & 0 \\ 0 & 1 \end{pmatrix} \quad B = \begin{pmatrix} 0 & -1 \\ 1 & 0 \end{pmatrix}.$$

The composition $T \circ U$ is: first rotate counterclockwise by 90° , then scale the x -axis by 2. The matrix for $T \circ U$ is

$$AB = \begin{pmatrix} 2 & 0 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} 0 & -1 \\ 1 & 0 \end{pmatrix} = \begin{pmatrix} 0 & -2 \\ 1 & 0 \end{pmatrix}.$$

The inverse of U rotates *clockwise* by 90° . The matrix for U^{-1} is

$$B^{-1} = \begin{pmatrix} 0 & 1 \\ -1 & 0 \end{pmatrix}.$$

Matrix Inverses

The **inverse** of an $n \times n$ matrix A is a matrix A^{-1} such that $AA^{-1} = I_n$ (equivalently, $A^{-1}A = I_n$).

2×2 case:

$$A = \begin{pmatrix} a & b \\ c & d \end{pmatrix} \implies A^{-1} = \frac{1}{ad - bc} \begin{pmatrix} d & -b \\ -c & a \end{pmatrix}.$$

$n \times n$ case: Row reduce the augmented matrix $(A \mid I_n)$. If you get $(I_n \mid B)$, then $B = A^{-1}$. Otherwise, A is not invertible.

Solving linear systems by “dividing by A ”: If A is invertible, then

$$Ax = b \iff x = A^{-1}b.$$

Important

If A is invertible, then $Ax = b$ has exactly one solution for any b , namely, $x = A^{-1}b$.

Solving Linear Systems by Inverting Matrices

Example

Important

If A is invertible, then $Ax = b$ has exactly one solution for any b , namely, $x = A^{-1}b$.

Example

$$\text{Solve } \begin{pmatrix} 2 & 1 \\ 1 & 3 \end{pmatrix} x = \begin{pmatrix} b_1 \\ b_2 \end{pmatrix}.$$

Answer:

$$x = \begin{pmatrix} 2 & 1 \\ 1 & 3 \end{pmatrix}^{-1} \begin{pmatrix} b_1 \\ b_2 \end{pmatrix} = \frac{1}{2 \cdot 3 - 1 \cdot 1} \begin{pmatrix} 3 & -1 \\ -1 & 2 \end{pmatrix} \begin{pmatrix} b_1 \\ b_2 \end{pmatrix} = \frac{1}{5} \begin{pmatrix} 3b_1 - b_2 \\ -b_1 + 2b_2 \end{pmatrix}$$

Elementary Matrices

Definition

An **elementary matrix** is a square matrix E which differs from I_n by one row operation.

There are three kinds:

$$\begin{array}{c} \text{scaling} \\ (R_2 = 2R_2) \end{array}$$

$$\begin{pmatrix} 1 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 1 \end{pmatrix}$$

$$\begin{array}{c} \text{row replacement} \\ (R_2 = R_2 + 2R_1) \end{array}$$

$$\begin{pmatrix} 1 & 0 & 0 \\ 2 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}$$

$$\begin{array}{c} \text{swap} \\ (R_1 \leftrightarrow R_2) \end{array}$$

$$\begin{pmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{pmatrix}$$

Fact: if E is the elementary matrix for a row operation, then EA differs from A by the same row operation.

$$A = \begin{pmatrix} 1 & 0 & 0 \\ 2 & 3 & 4 \end{pmatrix} \rightsquigarrow B = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 3 & 4 \end{pmatrix}$$

You get B by subtracting $2 \times$ the first row of A from the second row.

$$B = EA \quad \text{where} \quad E = \begin{pmatrix} 1 & 0 \\ -2 & 1 \end{pmatrix} \quad \left(\begin{array}{l} \text{subtract } 2 \times \text{ the first row} \\ \text{of } I_2 \text{ from the second row} \end{array} \right).$$

The Inverse of an Elementary Matrix

Fact: the inverse of an elementary matrix E is the elementary matrix obtained by doing the opposite row operation to I_n .

$$\begin{array}{cccc} R_2 = R_2 \times 2 & R_2 = R_2 \div 2 & R_2 = R_2 + 2R_1 & R_2 = R_2 - 2R_1 \\ \begin{pmatrix} 1 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 1 \end{pmatrix}^{-1} & = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1/2 & 0 \\ 0 & 0 & 1 \end{pmatrix} & \begin{pmatrix} 1 & 0 & 0 \\ 2 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}^{-1} & = \begin{pmatrix} 1 & 0 & 0 \\ -2 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \end{array}$$

$$\begin{array}{ccc} R_1 \longleftrightarrow R_2 & R_1 \longleftrightarrow R_2 & \\ \begin{pmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{pmatrix}^{-1} & = & \begin{pmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{pmatrix} \end{array}$$

If A is invertible, then there are a sequence of row operations taking A to I_n :

$$E_r E_{r-1} \cdots E_2 E_1 A = I_n.$$

Taking inverses (note the order!):

$$A = E_1^{-1} E_2^{-1} \cdots E_r^{-1} I_n = E_1^{-1} E_2^{-1} \cdots E_r^{-1}.$$

The Invertible Matrix Theorem

For reference

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. $\dim \text{Nul } A = 0$.

Learn it!

LU factorization

A guru provides, for (suitable) $m \times n$ matrix A , matrices L and U such that

- ▶ L is lower triangular $m \times m$ matrix (with ones on the diagonal)
- ▶ U is an $m \times n$ row echelon form (not necessary reduced)
- ▶ $A = LU$

E.g.

$$A = \begin{pmatrix} 2 & 4 & -1 & 5 & -1 \\ -4 & -5 & 3 & -8 & 1 \\ 2 & -5 & -4 & 1 & 8 \\ -6 & 0 & 7 & -3 & 1 \end{pmatrix} = \begin{pmatrix} 1 & 0 & 0 & 0 \\ -2 & 1 & 0 & 0 \\ 1 & -3 & 1 & 0 \\ -3 & 4 & 2 & 1 \end{pmatrix} \begin{pmatrix} 2 & 4 & -1 & 5 & -2 \\ 0 & 3 & 1 & 2 & -3 \\ 0 & 0 & 0 & 2 & 1 \\ 0 & 0 & 0 & 0 & 5 \end{pmatrix}$$

If $A = LU$ and want to solve $Ax = b$

1. Solve $Ly = b$,
2. Solve $Ux = y$,
3. Now, x is a solution to $Ax = b$.

Construction of LU factorizations

Our assumption: Row reduction of A requires *no row swaps*

1. *Row reduce* the matrix A , but *do not normalize pivots* to being one's; e.g.
2. Separate the row reduction according to 'clearing' pivot columns
3. From row reduction, gather 'vectors' below pivots and normalize. These form the 'vectors' in L .
4. If there are more columns in L than pivots, then leave rest of entries blank.

1.-2.

$$A = \begin{bmatrix} 2 & 4 & -1 & 5 & -2 \\ -4 & -5 & 3 & -8 & 1 \\ 2 & -5 & -4 & 1 & 8 \\ -6 & 0 & 7 & -3 & 1 \end{bmatrix} \sim \begin{bmatrix} 2 & 4 & -1 & 5 & -2 \\ 0 & 3 & 1 & 2 & -3 \\ 0 & -9 & -3 & -4 & 10 \\ 0 & 12 & 4 & 12 & -5 \end{bmatrix} = A_1$$
$$\sim A_2 = \begin{bmatrix} 2 & 4 & -1 & 5 & -2 \\ 0 & 3 & 1 & 2 & -3 \\ 0 & 0 & 0 & 2 & 1 \\ 0 & 0 & 0 & 4 & 7 \end{bmatrix} \sim \begin{bmatrix} 2 & 4 & -1 & 5 & -2 \\ 0 & 3 & 1 & 2 & -3 \\ 0 & 0 & 0 & 2 & 1 \\ 0 & 0 & 0 & 0 & 5 \end{bmatrix} = U$$

3.

$$L = \begin{bmatrix} 2 & & & \\ -4 & 3 & & \\ 2 & -9 & 2 & \\ -6 & 12 & 4 & 5 \end{bmatrix} \begin{matrix} \div 2 \\ \div 3 \\ \div 2 \\ \div 5 \end{matrix} \begin{matrix} \downarrow \\ \downarrow \\ \downarrow \\ \downarrow \end{matrix} \begin{bmatrix} 1 & & & \\ -2 & 1 & & \\ 1 & -3 & 1 & \\ -3 & 4 & 2 & 1 \end{bmatrix}$$

A second example

Find the LU factorization of A :

$$A = \begin{bmatrix} 2 & -4 & -2 & 3 \\ 6 & -9 & -5 & 8 \\ 2 & -7 & -3 & 9 \\ 4 & -2 & -2 & -1 \\ -6 & 3 & 3 & 4 \end{bmatrix} \sim \begin{bmatrix} 2 & -4 & -2 & 3 \\ 0 & 3 & 1 & -1 \\ 0 & -3 & -1 & 6 \\ 0 & 6 & 2 & -7 \\ 0 & -9 & -3 & 13 \end{bmatrix} \rightarrow \begin{bmatrix} 2 \\ 6 \\ 2 \\ 4 \\ -6 \end{bmatrix} \begin{bmatrix} 3 \\ -3 \\ 6 \\ -9 \end{bmatrix} \begin{bmatrix} 5 \\ -5 \\ 10 \end{bmatrix}$$

$\div 2$ $\div 3$ $\div 5$
 \downarrow \downarrow \downarrow

$$\sim \begin{bmatrix} 2 & -4 & -2 & 3 \\ 0 & 3 & 1 & -1 \\ 0 & 0 & 0 & 5 \\ 0 & 0 & 0 & -5 \\ 0 & 0 & 0 & 10 \end{bmatrix} \sim \begin{bmatrix} 2 & -4 & -2 & 3 \\ 0 & 3 & 1 & -1 \\ 0 & 0 & 0 & 5 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix} = U$$

$$\begin{bmatrix} 1 & & & & \\ 3 & 1 & & & \\ 1 & -1 & 1 & & \dots \\ 2 & 2 & -1 & 1 & \\ -3 & -3 & 2 & 2 & \end{bmatrix},$$

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