



Linear Alegbra MathNoteBook

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Tuesday

Introduction
Gaussian Elimination
Complexity Analysis

Friday

Matrix Multiplication
Elementary Matrix
Properties of Matrix
Permutation Matrix
LU decomposition
LDU decomposition
LU Decomposition with row exchanges

Tuesday

Review
Special matrix multiplication case
Inverse

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Review of 2 weeks
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How to solve a general rectangular

7 — Week3

7.1 Tuesday

7.1.1 Introduction

Why do we learn Linear Algebra?

So, we raise the question again, why do we learn LA?

- Basis of AI/ML/SP/etc.
In information age, *artificial intelligence, machine learning, structured programming*, and otherwise gains great popularity among researchers. LA is the basis of them, so in order to explore science in modern age, you should learn LA well.
- Solving linear system of equations.
How to solve linear system of equations efficiently and correctly is the **key** question for mathematicians.
- Internal grace.
LA is very beautiful, hope you enjoy the beauty of math.
- Interview questions.
LA is often used for interview questions for phd. Because the upper bound of difficulty for LA is **infinity**, interviewer often choose LA to question phd.

What is LA?

The main part of Mathematics is given below:

$$\text{mathematics} \left\{ \begin{array}{l} \text{Analysis+Calculus} \\ \text{Algebra:foucs on structure} \\ \text{Geometry} \end{array} \right.$$

All parts of math are based on **axiom systems**. And **LA** is the significant part of *Algebra*, which focus on the linear structure.

7.1.2 Review of 2 weeks

Motivating question: How to solve linear system equations?

The basic method is **Gaussian Elimination** (To make equations *simpler*. The main idea is *induction*)

Given one equation $ax = b$, you can easily solve it:

$$\implies \text{"if } a > 0, \text{ no solution." or } x = \frac{b}{a}$$

By induction, *if you can solve $n \times n$ systems, can you solve $(n+1) \times (n+1)$ systems?*

In this process, math notations is needed:

- matrix multiplication
- matrix inverse
- transpose, symmetric matrices

So in first two weeks, we just learn two things:

- *linear system could be solved almost by G.E.*
- *Furthermore, Gaussian Elimination is (almost) LU decomposition.*

But there is a question remained to be solved:

For **singular** system, How to solve it?

- When will it has no solution, when it has infinitely many solutions? (Note that singular system don't have unique solution.)
- If it has infinitely many solutions, how to find and express these solutions?

If we express system as matrix, we only to answer the question: **How to solve rectangular?**

7.1.3 Examples of solving equations

- For square case, we often convert the system into $\mathbf{R}\mathbf{x} = \mathbf{c}$, where \mathbf{R} is of *row echelon form*.
- But for rectangular case, *row echelon form*(ref) is not enough, we must convert it into **reduced row echelon form**(rref):

$$\mathbf{U}(\text{ref}) = \begin{bmatrix} 1 & 0 & \times & \times & \times & 0 & \times \\ 0 & 1 & \times & \times & \times & 0 & \times \\ 0 & 0 & 0 & 0 & 0 & 1 & \times \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 \end{bmatrix} \implies \mathbf{R}(\text{rref}) = \begin{bmatrix} 1 & 0 & \times & \times & \times & 0 & \times \\ 0 & 1 & \times & \times & \times & 0 & \times \\ 0 & 0 & 0 & 0 & 0 & 1 & \times \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 \end{bmatrix}$$

■ **Example 7.1** We discuss how to solve square matrix of **rref**: If all rows have nonzero entry, we have:

$$\begin{bmatrix} 1 & & 0 \\ & 1 & \\ & & 1 \\ 0 & & & 1 \end{bmatrix} \mathbf{x} = \mathbf{c} \implies \mathbf{x} = \mathbf{c}$$

We already solved this system, but note that *the last row could be all zero*:

$$\begin{bmatrix} 1 & & & \\ & 1 & & \\ & & 1 & \\ & & & 0 \end{bmatrix} \mathbf{x} = \mathbf{c} \implies \begin{cases} x_1 = c_1 \\ x_2 = c_2 \\ x_3 = c_3 \\ 0 = c_4 \end{cases}$$

So the result has two cases:

- If $c_4 \neq 0$, we have no solution of this system.
- If $c_4 = 0$, we have infinitely many solutions, which can be expressed as:

$$x_{\text{complete}} = \begin{pmatrix} c_1 \\ c_2 \\ c_3 \\ x_4 \end{pmatrix} = \begin{pmatrix} c_1 \\ c_2 \\ c_3 \\ 0 \end{pmatrix} + x_4 \begin{pmatrix} 0 \\ 0 \\ 0 \\ 1 \end{pmatrix}$$

where x_4 could be arbitrary number.

Hence for the $n \times n$ systems, does Gaussian Elimination work?

Answer: Almost, except “pivot=0” case.

- All pivots $\neq 0 \implies$ system has unique solution.
- Some pivots = 0 (The matrix is singular)
 1. No solution.
 2. Infinitely many solution.

What is G.E. doing? (Nonsingular case.)

Abstraction: We use matrix to represent system of equations (Chinese mathematicians fail to do this.):

$$\begin{cases} a_{11}x_1 + a_{12}x_2 + \cdots + a_{1n}x_n = b_1 \\ a_{21}x_1 + a_{22}x_2 + \cdots + a_{2n}x_n = b_2 \\ \dots \\ a_{m1}x_1 + a_{m2}x_2 + \cdots + a_{mn}x_n = b_m \end{cases} \implies \mathbf{Ax} = \mathbf{b}$$

By postmultiplying \mathbf{E}_{ij} and \mathbf{P}_{ij} we do one step of elimination:

$$\mathbf{E}_{ij}\mathbf{Ax} = \mathbf{b} \quad \mathbf{P}_{ij}\mathbf{Ax} = \mathbf{b}$$

By several steps of elimination, we obtain the final result:

$$\hat{\mathbf{L}}\mathbf{PAx} = \hat{\mathbf{L}}\mathbf{Pb}$$

where $\hat{\mathbf{L}}\mathbf{PA}$ represents a upper triangular matrix \mathbf{U} , $\hat{\mathbf{L}}$ is the lower triangular matrix.

$$\implies \hat{\mathbf{L}}\mathbf{PA} = \mathbf{U} \implies \mathbf{PA} = \hat{\mathbf{L}}^{-1}\mathbf{U} \triangleq \mathbf{LU}$$

So Gaussian Elimination is almost the \mathbf{LU} decomposition.

Example for solving rectangular system of rref

Recall the definition for rref:

Definition 7.1 — reduced row echelon form. Suppose a matrix has r nonzero rows, each row has leading 1 as pivots. If all columns with pivots (call it pivot column) are all zero entries apart from the pivot in this column, then this matrix is said to be **reduced row echelon form (rref)**.

Next we want to show an example for how to solve non-square system of rref, note that in last lecture we know the solution is given by:

$$\mathbf{x}_{\text{complete}} = \mathbf{x}_p + \mathbf{x}_{\text{special}}$$

■ **Example 7.2** We try to solve the system $\begin{bmatrix} 1 & 3 & 0 & -1 \\ 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 \end{bmatrix} \mathbf{x} = \mathbf{c}$.

- **step 1:** Find null space. Thus we only need to solve $\mathbf{R}\mathbf{x} = \mathbf{0}$.

$$\Rightarrow \begin{bmatrix} 1 & 3 & 0 & -1 \\ 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} \Rightarrow \begin{cases} x_1 + 3x_2 + 0x_3 - x_4 = 0 \\ x_3 + x_4 = 0 \end{cases}$$

What should we do next? We want to express the **pivot variable** as the form of **free variable**.

Note that the pivot columns in \mathbf{R} are column 1 and 3, so the pivot variable is x_1 and x_3 . The free variable is the remaining variable, say, x_2 and x_4 .

Hence the expression for x_1 and x_3 is given by:

$$\begin{cases} x_1 = -3x_2 + x_4 \\ x_2 = -x_4 \end{cases}$$

Hence all solutions to $\mathbf{R}\mathbf{x} = \mathbf{0}$ are

$$\mathbf{x}_{\text{special}} = \begin{bmatrix} -3x_2 + x_4 \\ x_2 \\ -x_4 \\ x_4 \end{bmatrix} = x_2 \begin{bmatrix} -3 \\ 1 \\ 0 \\ 0 \end{bmatrix} + x_4 \begin{bmatrix} 1 \\ 0 \\ -1 \\ 1 \end{bmatrix}$$

where x_2 and x_4 can be taken arbitrarily.

- **step 2:** find particular solution to $\mathbf{R}\mathbf{x} = \mathbf{c}$.

The trick for this step is to set $x_2 = x_4 = 0$. (set free variable to be zero and then derive the pivot variable.)

$$\Rightarrow \begin{bmatrix} 1 & 3 & 0 & -1 \\ 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ 0 \\ x_3 \\ 0 \end{bmatrix} = \begin{bmatrix} c_1 \\ c_2 \\ c_3 \end{bmatrix} \Rightarrow \begin{cases} x_1 = c_1 \\ x_3 = c_2 \\ 0 = c_3 \end{cases}$$

\Rightarrow

$$\text{– if } c_3 = 0, \text{ then exists particular solution } \mathbf{x}_p = \begin{bmatrix} c_1 \\ 0 \\ c_2 \\ 0 \end{bmatrix};$$

– if $c_3 \neq 0$, then $\mathbf{R}\mathbf{x} = \mathbf{c}$ has no solution.

- **Final solution:** Assume $c_3 = 0$, then all solution to $\mathbf{R}\mathbf{x} = \mathbf{c}$ is given by:

$$\mathbf{x}_{\text{complete}} = \mathbf{x}_p + \mathbf{x}_{\text{special}} = \begin{bmatrix} c_1 \\ 0 \\ c_2 \\ 0 \end{bmatrix} + x_2 \begin{bmatrix} -3 \\ 1 \\ 0 \\ 0 \end{bmatrix} + x_4 \begin{bmatrix} 1 \\ 0 \\ -1 \\ 1 \end{bmatrix}$$

Next we show how to solve a general rectangular:

7.1.4 How to solve a general rectangular

For linear system $\mathbf{Ax} = \mathbf{b}$, where \mathbf{A} is rectangular, we can solve this system as follows:

- **step1:** Gaussian Elimination.

With proper rows permutation (postmultiply \mathbf{P}_{ij}) and row transformation (postmultiply \mathbf{E}_{ij}) we convert \mathbf{A} into $\mathbf{R}(\text{rref})$, then we only need to solve $\mathbf{Rx} = \mathbf{c}$.

■ **Example 7.3** The first example is a 3×4 matrix with two pivots:

$$\mathbf{A} = \begin{bmatrix} 1 & 1 & 2 & 3 \\ 2 & 2 & 8 & 10 \\ 3 & 3 & 10 & 13 \end{bmatrix}$$

clearly $a_{11} = 1$ is the first pivot, clear row 2 and row 3 of this matrix:

$$\begin{aligned} \mathbf{A} &\xrightarrow[\begin{matrix} \mathbf{E}_{21} = \begin{bmatrix} 1 & 0 & 0 \\ -2 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \\ \mathbf{E}_{31} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ -3 & 0 & 1 \end{bmatrix} \end{matrix}]{\begin{bmatrix} 1 & 1 & 2 & 3 \\ 0 & 0 & 4 & 4 \\ 0 & 0 & 4 & 4 \end{bmatrix}} \xrightarrow[\begin{matrix} \mathbf{E}_{12} = \begin{bmatrix} 1 & -\frac{1}{2} & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \\ \mathbf{E}_{32} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & -1 & 1 \end{bmatrix} \end{matrix}]{\begin{bmatrix} 1 & 1 & 0 & 1 \\ 0 & 0 & 4 & 4 \\ 0 & 0 & 0 & 0 \end{bmatrix}} \\ &\Rightarrow \begin{bmatrix} 1 & 1 & 0 & 1 \\ 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 \end{bmatrix} \end{aligned}$$

If we want to solve $\mathbf{Ax} = \mathbf{b}$, firstly we should convert \mathbf{A} into $\begin{bmatrix} 1 & 1 & 0 & 1 \\ 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 \end{bmatrix} (\text{rref})$. ■

Then we should identify **pivot variables** and **free variables**. we can follow the direction to derive these:

pivot \Rightarrow pivot columns \Rightarrow pivot columns \Rightarrow pivot variable

■ **Example 7.4** we want to identify **pivot variables** and **free variables** of \mathbf{R} :

$$\mathbf{R} = \begin{bmatrix} 1 & 0 & \times & \times & \times & 0 & \times \\ 0 & 1 & \times & \times & \times & 0 & \times \\ 0 & 0 & 0 & 0 & 0 & 1 & \times \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 \end{bmatrix}$$

The pivot are r_{11}, r_{22}, r_{36} . So the pivot columns are column 1, 2, 6. So the *pivot variables* are x_1, x_2, x_6 ; the *free variables* are x_3, x_4, x_5, x_7 . ■

- **step2:** Compute null space $N(\mathbf{A})$. In order to find $N(\mathbf{A})$, we only need to compute $N(\mathbf{R})$.
 - For each of $(n - r)$ free variables,
 - set value of **it** to be 1.
 - set other **free variables** to be 0.
 - Then solve $\mathbf{Rx} = \mathbf{0}$ to get special solution \mathbf{y}_j for $j = 1, 2, \dots, n - r$.

■ **Example 7.5** continue with 3×4 matrix example:

$$R = \begin{bmatrix} 1 & 1 & 0 & 1 \\ 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

We want to find special solutions to $R\mathbf{x} = \mathbf{0}$:

1. Set $x_2 = 1$ and $x_4 = 0$. Solve $R\mathbf{x} = \mathbf{0}$, then $x_1 = -1$ and $x_3 = 0$.

Hence one special solution is $y_1 = \begin{bmatrix} -1 \\ 1 \\ 0 \\ 0 \end{bmatrix}$.

2. Set $x_2 = 0$ and $x_4 = 1$. Solve $R\mathbf{x} = \mathbf{0}$, then $x_1 = -1$ and $x_3 = -1$.

Then another special solution is $y_2 = \begin{bmatrix} -1 \\ 0 \\ -1 \\ 1 \end{bmatrix}$.

- When we get $(n - r)$ special solutions to $R\mathbf{x} = \mathbf{0}$: y_1, y_2, \dots, y_{n-r} .
Then $N(\mathbf{A}) = \text{span}(y_1, y_2, \dots, y_{n-r})$.

■ **Example 7.6** We continue the example above, when we get all special solutions

$y_1 = \begin{bmatrix} -1 \\ 1 \\ 0 \\ 0 \end{bmatrix}, y_2 = \begin{bmatrix} -1 \\ 0 \\ -1 \\ 1 \end{bmatrix}$, **the null space contains all linear combinations of the special solutions.**

$$\mathbf{x}_{\text{special}} = \text{span}\left(\begin{bmatrix} -1 \\ 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} -1 \\ 0 \\ -1 \\ 1 \end{bmatrix}\right) = x_2 \begin{bmatrix} -1 \\ 1 \\ 0 \\ 0 \end{bmatrix} + x_4 \begin{bmatrix} -1 \\ 0 \\ -1 \\ 1 \end{bmatrix}$$

where x_2, x_4 here could be arbitrary.

- **step3:** Compute a particular solution \mathbf{x}_p .

The easiest way is to “read” from $R\mathbf{x} = \mathbf{c} = \begin{bmatrix} c_1 \\ c_2 \\ \vdots \\ c_n \end{bmatrix}$:

Suppose $R \in \mathbb{R}^{m \times n}$ has $r (\leq m)$ pivot variables, then it has $(m - r)$ zero rows and $(n - r)$ free variables. In order to have solution, we must have $c_{r+1} = \dots = c_n = 0$. In other words, **For a solution to exist, zero rows in R must also be zero in \mathbf{c} .**

■ **Example 7.7** If $R\mathbf{x} = \begin{bmatrix} 1 & 3 & 0 & 2 \\ 0 & 0 & 1 & 4 \\ 0 & 0 & 0 & 0 \end{bmatrix} \mathbf{x} = \mathbf{c} = \begin{bmatrix} c_1 \\ c_2 \\ c_3 \end{bmatrix}$, then in order to have a solution, we must let $c_3 \neq 0$.

So we have to discuss the particular solution case by case:

- **case1:** one of c_{r+1}, \dots, c_n is nonzero, then the system has **no** solution.
- **case2:** $c_{r+1} = \dots = c_n$, then a particular solution exists:

$$\mathbf{x}_p = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix}$$

We set all **free variables** to be zero, and pivot variables are from \mathbf{c} . More specifically, the first entry in \mathbf{c} is exactly the value for the first pivot variable; the second entry in \mathbf{c} is exactly the value for the second pivot variable.

■ **Example 7.8** If $\mathbf{R}\mathbf{x} = \begin{bmatrix} 1 & 3 & 0 & 2 \\ 0 & 0 & 1 & 4 \\ 0 & 0 & 0 & 0 \end{bmatrix} \mathbf{x} = \mathbf{c} = \begin{bmatrix} c_1 \\ c_2 \\ 0 \end{bmatrix}$, we want to compute particular solution

$$\mathbf{x}_p = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix}$$

Then we know x_2, x_4 are free variable, so $x_2 = x_4 = 0$; x_1, x_3 are pivot variable, so we have $\begin{pmatrix} x_1 \\ x_3 \end{pmatrix} = \begin{pmatrix} c_1 \\ c_2 \end{pmatrix}$.

Hence the solution for $\mathbf{R}\mathbf{x} = \mathbf{c}$ is $\begin{bmatrix} c_1 \\ 0 \\ c_2 \\ 0 \end{bmatrix}$. ■

- **Final step:** All solution of $\mathbf{A}\mathbf{x} = \mathbf{b}$ are $\mathbf{x}_{\text{complete}} = \mathbf{x}_p + \mathbf{x}_{\text{special}}$, where $\mathbf{x}_{\text{special}} \in N(\mathbf{A})$. \mathbf{x}_p is defined in step3, $\mathbf{x}_{\text{special}}$ is defined in step2.

However, where does the number r come? r denotes the **rank** of a matrix, which will be discussed next lecture.

