

Linear Systems of Equations

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i.e. things like **Reduced Row Echelon Form (RREF)** and **Gaussian Elimination**

Linear Systems

$$\text{Ex: } \begin{array}{l} 3x + 4y = 7 \\ -2x + 5y = 0 \end{array} \quad \text{NOT} \quad \begin{array}{l} 3x^2 + 4 \cdot \sin(y) = 7 \\ x \cdot y = 2 \end{array}$$

$\underbrace{\qquad\qquad\qquad}_{A}$

#1 Skill: be able to write in matrix form

$$\underbrace{\begin{bmatrix} 3 & 4 \\ -2 & 5 \end{bmatrix}}_A \cdot \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} 7 \\ 0 \end{bmatrix}$$

More generally

$$\begin{aligned} a_{11}x_1 + a_{12}x_2 + a_{13}x_3 + \dots + a_{1n}x_n &= b_1 \\ a_{21}x_1 + a_{22}x_2 + a_{23}x_3 + \dots + a_{2n}x_n &= b_2 \\ &\vdots \\ a_{m1}x_1 + a_{m2}x_2 + a_{m3}x_3 + \dots + a_{mn}x_n &= b_m \end{aligned}$$

$$\underbrace{\begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \ddots & \vdots \\ \vdots & & & a_{mn} \end{bmatrix}}_A \cdot \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} = \begin{bmatrix} b_1 \\ \vdots \\ b_m \end{bmatrix}$$

$m \times n$

• Underdetermined (more unknowns than equations)

$$m < n \quad \begin{array}{c|c|c} m & \begin{array}{c} n \\ \hline A \end{array} & \begin{array}{c} \vec{x} \\ \hline \end{array} = \begin{array}{c} m \\ \hline \vec{b} \end{array} \end{array}$$

Facts: $\text{null}(A)$ is non-trivial

There need not be any solutions
(ex: $A=0, \vec{b} \neq 0$)

but if there is a solution,
it is **not unique**

• Overdetermined (more equations than unknowns)

$$m > n \quad \begin{array}{c|c|c} m & \begin{array}{c} n \\ \hline A \end{array} & \begin{array}{c} \vec{x} \\ \hline \end{array} = \begin{array}{c} m \\ \hline \vec{b} \end{array} \end{array}$$

Facts: There need not be any solutions.

If there are solutions, then some equations are redundant.

The solution is **unique** (if it exists)
if $\text{rank}(A) = n$ "full rank"

- Exactly determined (square): one equation per unknown
We'll focus on this case in our class — simplest

$$m=n \quad \begin{matrix} n \\ | \\ A \\ | \\ m \end{matrix} \quad \begin{matrix} x \\ | \\ = \\ | \\ b \end{matrix}$$

Facts: There need not be a solution, nor does it need to be unique.

If A is invertible, $\vec{x} = A^{-1}\vec{b}$.

The solution (exists and is unique) occurs
iff A is full rank ($\text{rank}(A) = n$)
iff A is invertible.

 Computationally we don't do this

Solving Square Systems of equations

Observations: ① solving $3x + 4y = 2$ is the same as solving $5x - 3y = 9$

① Re-arrange / swap $5x - 3y = 9$ $3x + 4y = 2$

② $3x + 4y = 2 \iff \alpha(3x + 4y) = \alpha(2)$ for some $\alpha \neq 0$ precise mathematical meaning: iff

② Scale (if $\alpha=0$ then \Rightarrow but not \Leftarrow) means "implies"

③ $3x + 4y = 2 \iff 3x + 4y + \beta = 2 + \beta$ for some β

③ Add

These are the tricks behind Gaussian Elimination

Linear Algebra 101 method to solve a ^{square} linear system of equations

$$3x_1 + 4x_2 + 5x_3 = 6$$

$$6x_2 + 0x_3 = 4$$

$$9x_3 + 6x_2 + 4x_3 = -2$$

① Form the augmented matrix $[A : b]$

$$\left[\begin{array}{ccc|c} 3 & 4 & 5 & 6 \\ 0 & 6 & 0 & 4 \\ 9 & 6 & 4 & -2 \end{array} \right]$$

② Apply elementary row operations to the augmented matrix
essentially swap, scale, add

Start in upper left

① Scale it to a 1

(2) add a multiple of top row to other rows in order to make their 1st entry a 0

(3) move-on: repeat (1) and (2) on all but 1st row + 1st column
(only swap if step (1) impossible due to 0 entry)

$$\begin{array}{l} R_1 \\ R_2 \\ R_3 \end{array} \left[\begin{array}{ccc|c} 3 & 4 & 5 & 6 \\ 6 & 0 & -3 & 4 \\ 9 & 6 & 4 & -2 \end{array} \right] \xrightarrow{R_1 \leftarrow R_1/3} \left[\begin{array}{ccc|c} 1 & 4/3 & 5/3 & 2 \\ 6 & 0 & -3 & 4 \\ 9 & 6 & 4 & -2 \end{array} \right]$$

$$R_2 \leftarrow R_2 - 6 \cdot R_1$$

$$\left[\begin{array}{ccc|c} 1 & 4/3 & 5/3 & 2 \\ 0 & -8 & -3-10 & 4-12 \\ 9 & 6 & 4 & -2 \end{array} \right] \xrightarrow{-8/(-8)} \left[\begin{array}{ccc|c} 1 & 4/3 & 5/3 & 2 \\ 0 & 1 & 13/8 & 1 \\ 0 & 6-12 & 4-15 & -2-18 \end{array} \right]$$

$$R_2 \leftarrow -R_2/8$$

$$\left[\begin{array}{ccc|c} 1 & 4/3 & 5/3 & 2 \\ 0 & 1 & 13/8 & 1 \\ 0 & -6 & -11 & -20 \end{array} \right] \xrightarrow{R_3 \leftarrow R_3 + 6 \cdot R_2} \left[\begin{array}{ccc|c} 1 & 4/3 & 5/3 & 2 \\ 0 & 1 & 13/8 & 1 \\ 0 & 0 & -11+6 \cdot \frac{13}{8} & -20+6 \cdot 1 \end{array} \right]$$

$$R_3 \leftarrow -8/19 \cdot R_3$$

$$\left[\begin{array}{ccc|c} 1 & 4/3 & 5/3 & 2 \\ 0 & 1 & 13/8 & 1 \\ 0 & 0 & 1 & -14 - \frac{5}{4} \end{array} \right] \xrightarrow{35/2 \text{ or } 17.5}$$

STOP

This is row echelon form
and what we did was Gaussian Elimination

You might recall reduced row echelon form (RREF) (and its procedure is called Gauss-Jordan Elimination)

which continues and gets it to

$$\left[\begin{array}{ccc|c} 1 & 0 & 0 & \tilde{b}_1 \\ 0 & 1 & 0 & \tilde{b}_2 \\ 0 & 0 & 1 & \tilde{b}_3 \end{array} \right]$$

$$\begin{aligned} 1 \cdot x_1 + 0 \cdot x_2 + 0 \cdot x_3 &= \tilde{b}_1 \\ 0 \cdot x_1 + 1 \cdot x_2 + 0 \cdot x_3 &= \tilde{b}_2 \\ 0 \cdot x_1 + 0 \cdot x_2 + 1 \cdot x_3 &= \tilde{b}_3 \end{aligned}$$

$$\Rightarrow \begin{aligned} x_1 &= \tilde{b}_1 \\ x_2 &= \tilde{b}_2 \\ x_3 &= \tilde{b}_3 \end{aligned}$$

Instead, do BACK SUBSTITUTION

note: we didn't have to make the pivot a 1

$$\left[\begin{array}{ccc|cc} 1 & 4/3 & 5/3 & 1 & 2 \\ 0 & 1 & 13/8 & 1 & 1 \\ 0 & 0 & 1 & | & -14/19 \\ & & & | & 112/19 \end{array} \right] \quad \text{means} \quad \begin{aligned} 1 \cdot x_1 + 4/3 x_2 + 5/3 x_3 &= 2 \\ 1 \cdot x_2 + 13/8 x_3 &= 1 \\ 1 \cdot x_3 &= 112/19 \end{aligned}$$

Observe this is upper triangular

$$\textcircled{1} \quad x_3 = 112/19$$

\textcircled{2} plug x_3 into equation 2

$$1 \cdot x_2 + 13/8 \cdot (112/19) = 1$$

$$\text{so } x_2 = 1 - \frac{13 \cdot 112}{8 \cdot 19} \approx 8.58$$

\textcircled{3} plug x_2 and x_3 into equation 1

$$1 \cdot x_1 + 4/3 (8.58) + 5/3 (112/19) = 2$$

so now we easily solve for x_1 .



That's Gaussian Elimination with Back Substitution

Ch 6.1 in the book gives a more formal algorithm, but most important is to understand what/why.

FLUP counts

\textcircled{1} For Gaussian Elimination

ignoring possible swaps.

not really necessary
but won't affect big-O notation

1. For 1st column, we scale entire row, and then add to rest of the rows

$$\begin{matrix} n & | & \text{* * * *} \\ & | & \text{* * * *} \\ & | & \text{* * * *} \\ & | & \text{* * * *} \end{matrix}$$

$O(n^2)$ operations

2. For 2nd column, same thing but exclude 1st row and 1st column

$$\begin{matrix} & | & \text{* * * *} \\ & | & \text{* * * *} \\ & | & \text{* * * *} \\ & | & \text{* * * *} \end{matrix}$$

$O((n-1)^2)$ operations

3. etc.

So total # operations is $O\left(\sum_{j=0}^{n-1} (n-j)^2\right) = O\left(\sum_{i=1}^n i^2\right)$ via change-of-variables
 $i = n-j$

Recall $\sum_{j=1}^n j = \frac{n(n+1)}{2} = O(n^2)$

$$\sum_{j=1}^n j^2 = \frac{n(n+1)(2n+1)}{6} = O(n^3)$$

$$= O\left(\frac{n(n+1)(2n+1)}{6}\right)$$

$$= O(n^3)$$

(book does it
much more
carefully)

② For back-substitution:

last equation: $O(1)$ work

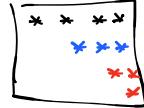
2nd-to-last equation: $O(2)$ work

⋮

k^{th} -to-last eq'n $O(k)$ work

⋮

$n^{\text{th}}\text{-to-last eq'n}$ aka 1st equation $O(n)$ work



$$\text{So } \sum_{j=1}^n O(j) \text{ flops, } = O(n^2) \text{ flops}$$

Message

① Gaussian Elimination on a $n \times n$ matrix takes $O(n^3)$ flops

② Back substitution (on an upper triangular matrix) takes $O(n^2)$ flops

Solving n linear equations in n unknowns takes $O(n^3)$ time

but solving a triangular system takes just $O(n^2)$ time

(and solving a diagonal system takes $O(n)$ time)

Relatively slow, which is why we pay attention carefully!

Multiple right-hand-sides

Solve $A\vec{x}_1 = \vec{b}_1$ and $A\vec{x}_2 = \vec{b}_2$

① make augmented matrix

$$\left[\begin{array}{c|c} A & B \end{array} \right]$$

↑
or, generally,
 K

$$\begin{array}{c|c} A & -K \\ \hline X & B \end{array} = \begin{array}{c|c} -K & \\ \hline B & \end{array}$$

$$B = \begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_n \end{bmatrix}$$

② Apply Gaussian elimination

$$O(n^2(n+K))$$

③ Do back substitution

$$O(n(n+K))$$

Finding A^{-1}

Set $B = I$, so $AX = I$ solve for X , $X = A^{-1} \cdot I = A^{-1}$

So to solve $A\vec{x} = \vec{b}$, if we did $\vec{x} = A^{-1} \vec{b}$, that involved
a lot of unnecessary work, and it's less accurate

* Don't use A^{-1} unless you really really need to
Standard use cases when you need A^{-1} : (none)