

# Linear Systems of Equations

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

## Linear Systems

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

#1 skill: be able to write in matrix form

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

More generally

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

$$\underbrace{\begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \ddots & \\ \vdots & & & \\ a_{m1} & & \dots & a_{mn} \end{bmatrix}}_A \cdot \underbrace{\begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix}}_{\vec{x}} = \underbrace{\begin{bmatrix} b_1 \\ \vdots \\ b_m \end{bmatrix}}_{\vec{b}}$$

$m \times n$

• Underdetermined (more unknowns than equations)

$$m < n \quad \boxed{m \times n} \quad \boxed{A} \quad \boxed{n} \quad \boxed{\vec{x}} = \boxed{m} \quad \boxed{\vec{b}}$$

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)

$$m > n \quad \boxed{m \times n} \quad \boxed{A} \quad \boxed{n} \quad \boxed{\vec{x}} = \boxed{m} \quad \boxed{\vec{b}}$$

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

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

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  $\beta$

③ Add

These are the tricks behind Gaussian Elimination

Linear Algebra 101 method to solve a <sup>square</sup> linear system of equations

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

$$6x_2 + 0x_3 - 3x_3 = 4$$

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

① Form the augmented matrix  $[A : b]$

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

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 1<sup>st</sup> entry a 0

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

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

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

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

$$R_2 \leftarrow -R_2/8$$

$$\left[ \begin{array}{ccc|c} 1 & 4/3 & 5/3 & 2 \\ 0 & 1 & 13/8 & 1 \\ 0 & -6 & -11 & -20 \end{array} \right] \xrightarrow{R_3 \leftarrow R_3 + 6 \cdot R_2} \left[ \begin{array}{ccc|c} 1 & 4/3 & 5/3 & 2 \\ 0 & 1 & 13/8 & 1 \\ 0 & 0 & -11 + \frac{69}{8} & -20 + 6 \end{array} \right]$$

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

$$\left[ \begin{array}{ccc|c} 1 & 4/3 & 5/3 & 2 \\ 0 & 1 & 13/8 & 1 \\ 0 & 0 & 1 & -14 - \frac{8}{19} \end{array} \right]$$

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

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

$$\text{meaning } 1 \cdot x_1 + 0 \cdot x_2 + 0 \cdot x_3 = \tilde{b}_1$$

$$0 \cdot x_1 + 1 \cdot x_2 + 0 \cdot x_3 = \tilde{b}_2$$

$$0 \cdot x_1 + 0 \cdot x_2 + 1 \cdot x_3 = \tilde{b}_3$$

$$x_1 = \tilde{b}_1$$

$$x_2 = \tilde{b}_2$$

$$x_3 = \tilde{b}_3$$

Instead, do BACK SUBSTITUTION

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

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

observe this is upper triangular

Solve last equation first

①  $x_3 = \frac{112}{19}$

② plug  $x_3$  into equation 2

$$1 \cdot x_2 + \frac{13}{8} \cdot \left( \frac{112}{19} \right) = 1$$

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

③ plug  $x_2$  and  $x_3$  into equation 1

$$1 \cdot x_1 + \frac{4}{3} (8.58) + \frac{5}{3} \left( \frac{112}{19} \right) = 2$$

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.

### FLOP counts

① For Gaussian Elimination

ignoring possible swaps.

- For 1<sup>st</sup> column, we scale entire row, and then add to rest of the rows

$$\begin{array}{c|ccc} n & & & \\ \hline & * & * & * \\ & * & * & * \\ & * & * & * \\ & * & * & * \end{array} \quad O(n^2) \text{ operations}$$

- For 2<sup>nd</sup> column, same thing but exclude 1<sup>st</sup> row and 1<sup>st</sup> column

$$\begin{array}{c|ccc} & * & * & * \\ \hline & * & * & * \\ & * & * & * \\ & * & * & * \end{array} \quad O((n-1)^2) \text{ operations}$$

3. etc.

$$\text{So total # operations is } O\left(\sum_{j=1}^n (n-j)^2\right) = O\left(\sum_{j=1}^n (n^2 - 2nj + j^2)\right)$$

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

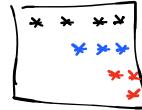
$$\sum_{j=1}^n j^2 = \frac{n(n+1)(2n+1)}{6} = O(n^3)$$

$$\begin{aligned} &= O(n^3) - 2nO(n^2) + O(n^3) \\ &= O(n^3) \end{aligned}$$

$$= \underline{\underline{O(n^3)}} \quad (\text{book does it much more carefully})$$

(2) For back-substitution:

last equation:  $O(1)$  work  
 2nd-to-last equation:  $O(2)$  work  
 $\vdots$   
 $k^{\text{th}}\text{-to-last eq'n}$ :  $O(k)$  work  
 $\vdots$   
 $n^{\text{th}}\text{-to-last eq'n}$ :  $O(n)$  work  
 aka 1st equation



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

### 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

$$\left[ \begin{array}{c|c} A & B \\ \hline \vec{b}_1 & \vec{b}_2 \end{array} \right]$$

↑  
or, generally,  
 $K$

$$\begin{array}{c|c} A & -K \\ \hline \vec{x} & B \end{array} = \begin{array}{c|c} \vec{b}_1 & \vec{b}_2 \\ \hline I & I \end{array}$$

$$B = \begin{bmatrix} \vec{b}_1 & \vec{b}_2 \\ I & I \end{bmatrix}$$

(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)