

# Gaussian Elimination

This is how a computer solves systems of linear equations using elimination. Almost all questions in this class will reduce to this procedure! (The interesting part is how they do so.)

**Def:** Two matrices are row equivalent if you can get from one to the other using row operations.

**NB:** If augmented matrices are row equivalent then they have the same solution sets.

**Algorithm (Gaussian Elimination/row reduction):**

**Input:** Any matrix

**Output:** A row-equivalent matrix in REF.

**Procedure:**

(1a) If the first nonzero column has a zero entry at the top, row swap so that the top entry is nonzero.

$$\begin{bmatrix} 0 & 4 & 3 & 3 \\ 1 & 1 & -1 & 3 \\ 5 & -3 & -6 & -6 \end{bmatrix} \xrightarrow{R_1 \leftrightarrow R_2} \begin{bmatrix} 1 & 1 & -1 & 3 \\ 0 & 4 & 3 & 3 \\ 5 & -3 & -6 & -6 \end{bmatrix}$$

This is now the first pivot position.

(1b) Perform row replacements to clear all entries below the first pivot.

$$\begin{bmatrix} 1 & 1 & -1 & 3 \\ 0 & 4 & 3 & 3 \\ 5 & -3 & -6 & -6 \end{bmatrix} \xrightarrow{R_3 - 5R_1} \begin{bmatrix} 1 & 1 & -1 & 3 \\ 0 & 4 & 3 & 3 \\ 0 & -8 & -1 & -21 \end{bmatrix}$$

Now ignore the row & column with the first pivot and recurse into the submatrix below and to the right:

$$\begin{bmatrix} 1 & 1 & -1 & 3 \\ 0 & 4 & 3 & 3 \\ 0 & -8 & -1 & -21 \end{bmatrix}$$

(2a) If the first nonzero column has a zero entry at the top, row swap so that the top entry is nonzero.

$$\begin{bmatrix} 1 & 1 & -1 & 3 \\ 0 & 4 & 3 & 3 \\ 0 & -8 & -1 & -21 \end{bmatrix} \quad \text{(Not applicable to this matrix)}$$

second pivot

(2b) Perform row replacements to clear all entries below the second pivot.

$$\begin{bmatrix} 1 & 1 & -1 & 3 \\ 0 & 4 & 3 & 3 \\ 0 & -8 & -1 & -21 \end{bmatrix} \xrightarrow{R_3 + 2R_2} \begin{bmatrix} 1 & 1 & -1 & 3 \\ 0 & 4 & 3 & 3 \\ 0 & 0 & 5 & -15 \end{bmatrix}$$

etc. (recurse)

Doesn't mess up the 1<sup>st</sup> column!

In our example, the recursion has terminated:

$$\begin{bmatrix} 1 & 1 & -1 & 3 \\ 0 & 4 & 3 & 3 \\ 0 & 0 & 5 & -15 \end{bmatrix} \text{ is in REF!}$$

**Important:** If you want to apply this algorithm to an augmented matrix, just **delete** the augmentation line (pretend it's not augmented).

**Demo:** Gauss-Jordan slideshow

Use **Rabinoff's Reliable Row Reducer** on the HW!  
↳ use the Sage cell when this gets easy

What about other row operations?

There are usually other (more clever!) ways to use row operations to put a matrix in REF.

These will yield the same solutions, of course!

But for some things, you have to apply the algorithm **as written** - eg. for LU decompositions (next time).

# Jordan Substitution

This is the **back-substitution** procedure

It is necessary when you have **∞ solutions**.

It puts a matrix into the following form:

**Def:** A matrix is in **reduced row echelon form**

**(RREF)** if:

(1-2) It is in **REF**

(so RREF implies REF)

(3) All pivots are equal to 1.

(4) A pivot is the only nonzero entry in its column.

REF

$$\begin{bmatrix} \bullet & \bullet & \bullet & \bullet \\ 0 & 0 & \bullet & \bullet \\ 0 & 0 & 0 & \bullet \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

● = nonzero (pivot)

RREF

$$\begin{bmatrix} 1 & \bullet & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

● = any number

**Eg:**

$$\left[ \begin{array}{ccc|c} 1 & 2 & 3 & 6 \\ 0 & -5 & -10 & -20 \\ 0 & 0 & 0 & 30 \end{array} \right]$$

is in REF. How to put into RREF?  
Do **back substitution!**

## Row Operations

$$\left[ \begin{array}{ccc|c} 1 & 2 & 3 & 6 \\ 0 & -5 & -10 & -20 \\ 0 & 0 & 10 & 30 \end{array} \right]$$

(scale so this is 1)

$$R_3 \div 10 \left\{ \begin{array}{l} \text{solve for} \\ x_3 \end{array} \right.$$

$$\left[ \begin{array}{ccc|c} 1 & 2 & 3 & 6 \\ 0 & -5 & -10 & -20 \\ 0 & 0 & 1 & 3 \end{array} \right]$$

(kill these)

$$\begin{array}{l} R_1 \div 3R_3 \\ R_2 \div 10R_3 \end{array} \left\{ \begin{array}{l} \text{substitute } x_3=3 \text{ into } R_1 \& R_2 \\ \text{then move the constants to} \\ \text{the RHS} \end{array} \right.$$

$$\left[ \begin{array}{ccc|c} 1 & 2 & 0 & -3 \\ 0 & -5 & 0 & 10 \\ 0 & 0 & 1 & 3 \end{array} \right]$$

$$\begin{array}{rcl} x_1 + 2x_2 & = & -3 \\ -5x_2 & = & 10 \\ x_3 & = & 3 \end{array}$$

(scale so this is 1)

$$R_2 \div -5 \left\{ \begin{array}{l} \text{solve} \\ \text{for} \\ x_2 \end{array} \right.$$

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

(kill this)

$$\begin{array}{rcl} x_1 + 2x_2 & = & -3 \\ x_2 & = & -2 \\ x_3 & = & 3 \end{array}$$

## Back-Substitution

$$\begin{array}{rcl} x_1 + 2x_2 + 3x_3 & = & 6 \\ -5x_2 - 10x_3 & = & -20 \\ 10x_3 & = & 30 \end{array}$$

$$\begin{array}{rcl} x_1 + 2x_2 + 3x_3 & = & 6 \\ -5x_2 - 10x_3 & = & -20 \\ x_3 & = & 3 \end{array}$$

$R_1 \rightarrow 2R_2$   $\left\{ \begin{array}{l} \text{substitute } x_2 = -2 \text{ into } R_1 \\ \text{then move the constants to} \\ \text{the RHS} \end{array} \right.$

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

$$\begin{array}{rcl} x_1 & & = 1 \\ & x_2 & = -2 \\ & & x_3 = 3 \end{array}$$

This is in RREF:

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

$$\rightsquigarrow \begin{array}{l} x_1 = 1 \\ x_2 = -2 \\ x_3 = 3 \end{array}$$

Solved ✓

Upshot: Jordan substitution is exactly back-substitution.

Demo: Gauss-Jordan slideshow, cont'd

## Algorithm (Jordan Substitution):

Input: A matrix in REF

Output: The row-equivalent matrix in RREF.

Procedure:

Loop, starting at the last pivot:

(a) Scale the pivot row so the pivot = 1.

(b) Use row replacements to kill the entries above that pivot.

"theorem"

Thm: The RREF of a matrix is unique.

In other words, if you start with a matrix, do any legal row operations at all, and end with a matrix in RREF, then it's the same matrix that Gauss-Jordan will produce.

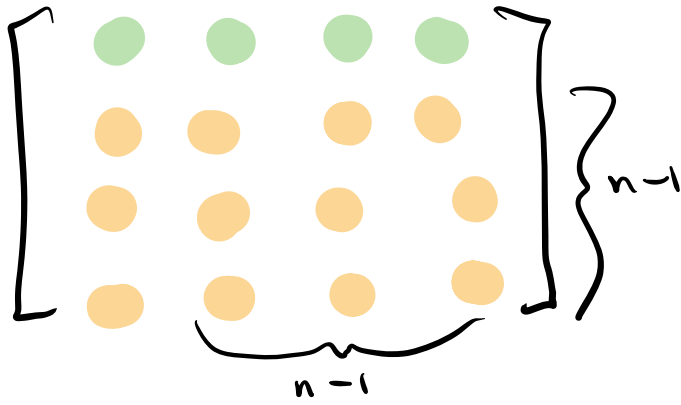
↳ Gaussian elimination + Jordan substitution.

NB: Jordan substitution gives you a RREF matrix with the same pivots. So uniqueness of RREF implies uniqueness of pivot positions.

# Computational Complexity

How much computer time does Gauss-Jordan take?

Gaussian Elimination on an  $n \times n$  matrix takes:



$$(n-1)(n-1) \text{ mult}$$

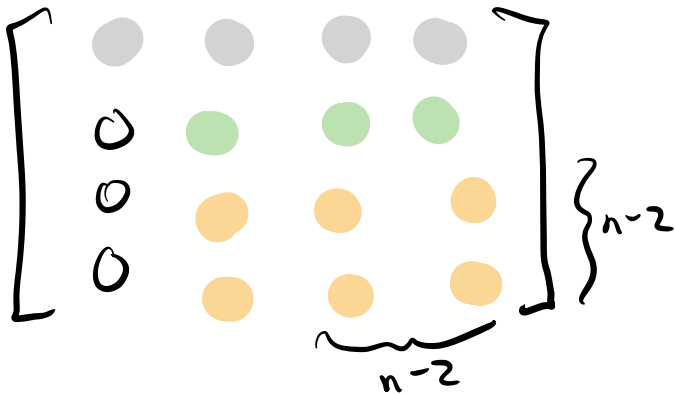
$$(n-1)(n-1) \text{ add}$$

$$\frac{2(n-1)^2}{2(n-1)^2}$$

flops = floating point operations

Step 1: Each row replacement requires  $n-1$  multiplications &  $n-1$  additions. (no computation in 1<sup>st</sup> col: just write "0")

Do for  $n-1$  lower rows:



$$(n-2)(n-2) \text{ mult}$$

$$+ (n-2)(n-2) \text{ add}$$

$$\frac{2(n-2)^2}{2(n-2)^2} \text{ flops}$$

etc.

Step 2: Each row replacement requires  $n-2$  multiplications &  $n-2$  additions. Must do this for  $n-2$  remaining rows



pyramidal number

$$\text{Total: } 2 \left[ (n-1)^2 + (n-2)^2 + \dots + 1^2 \right]$$

$$= 2 \cdot \frac{n(n-1)(2n-1)}{6} \approx \frac{2}{3} n^3 \text{ flops}$$

## Back-Substitution

$x_n = \dots$  1 mult = 1 flop

$x_{n-1} + x_n = \dots$  2 mult, 1 add = 3 flops

(substitute  $x_n$ ,  $\times$ , subtract,  $\div$ )

$x_{n-2} + x_{n-1} + x_n = \dots$  3 mult, 2 add = 5 flops

(substitute  $x_n$  &  $x_{n-1}$ ,  $\times$ ,  $\times$ , subtract,  $\div$ )

$x_1 + \dots + x_n = \dots$  n mult, (n-1) add = 2n-1 flops

Total:  $1+3+5+\dots+(2n-1) = n^2$  flops

NB:  $\frac{2}{3}n^3$  is a lot more than  $n^2$ !

For a  $1000 \times 1000$  matrix,  $\frac{2}{3}n^3 \approx \frac{2}{3}$  gigaflops

but  $n^2 = 1$  megaflop. If we want to solve

$Ax=b$  for 1000 values of  $b$ , doing elimination each time takes  $\frac{2}{3}$  teraflops!

I won't expect you to know a lot about computational complexity. Just learn which procedures are  $O(n^3)$  and which are  $O(n^2)$ .

# Inverse Matrices

Question: When solving  $Ax=b$ , when can we "divide by A"?

If " $x = \frac{b}{A}$ " makes sense, then  $Ax=b$  has exactly one solution " $x = \frac{b}{A}$ " for every  $b$ .

This means RREF( $A|b$ ) looks like this:

$$\left( \begin{array}{ccc|c} 1 & 0 & 0 & b_1 \\ 0 & 1 & 0 & b_2 \\ 0 & 0 & 1 & b_3 \end{array} \right)$$

Def: An  $n \times n$  (square!) matrix  $A$  is invertible if there exists another  $n \times n$  matrix  $B$  such that  $AB = I_n = BA$ .  $I_n = \begin{bmatrix} 1 & & 0 \\ & \ddots & \\ 0 & & 1 \end{bmatrix}$   $n \times n$  identity matrix  
Otherwise it's called singular.

Note:  $B = A^{-1}$ , called the inverse of  $A$ .

Eg:  $A = \begin{bmatrix} 2 & 3 \\ 1 & 2 \end{bmatrix}$   $B = \begin{bmatrix} 2 & -3 \\ -1 & 2 \end{bmatrix}$

$$AB = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \quad BA = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

$$\Rightarrow B = A^{-1}$$

Eg:  $A = \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}$   $\begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} \overbrace{\begin{bmatrix} a & b \\ c & d \end{bmatrix}}^{\text{"B"}} = \begin{bmatrix} a & b \\ 0 & 0 \end{bmatrix} \neq I_2$

so  $A$  is singular (non-invertible).

**Remark:** Since  $AB \neq BA$  in general, you have to require  $AB = I_n = BA$  a priori. **But:**

**Fact:** If  $A$  and  $B$  are  $n \times n$  matrices and  $AB = I_n$  or  $BA = I_n$ , then  $B = A^{-1}$ .

So the definition above is a bit pedantic...

**Remark:** A non-square matrix does not admit both a left- and right-inverse, so not invertible. (Can't solve  $AB = I_m$  and  $CA = I_n$  unless  $A$  is square.) This is why we only treat invertibility of **square** matrices.

**Fact:**  $(A^{-1})^{-1} = A$

because  $AB = I_n$  means  $B = A^{-1}$  and  $A = B^{-1}$

**Fact:** If  $A$  &  $B$  are invertible, then so is  $AB$ , and  $(AB)^{-1} = \cancel{A^{-1}B^{-1}} B^{-1}A^{-1}$ .

**Check:**  $(AB)(B^{-1}A^{-1}) = A(BB^{-1})A^{-1} = A(I_n)A^{-1} = AA^{-1} = I_n$  ✓

**NB:** Why not  $(AB)^{-1} = A^{-1}B^{-1}$ ? Let's check:

$$(AB)(A^{-1}B^{-1}) = ABA^{-1}B^{-1} = ???$$

(no cancellation - can't re-order the terms!)

Thm<sup>2</sup>: Let  $A$  be an  $n \times n$  matrix. either all are true or all are false  
The following are equivalent: (TFAE)

- (1)  $A$  is invertible
- (2) The RREF of  $A$  is  $I_n$  ← coefficient matrix!
- (3)  $A$  has a pivot in every row/every column.  
( $A$  has  $n$  pivots)

We'll see why a bit later.

Eg:  $A = \begin{bmatrix} 2 & 3 \\ 1 & 2 \end{bmatrix} \xrightarrow{\text{RREF}} \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} = I_2$  invertible  
● = pivots

Eg:  $A = \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}$  is in RREF,  $\neq I_2$  singular

How do you compute the inverse?

Algorithm (Matrix Inversion):

Input: A square matrix.

Output: The inverse matrix, or "singular"

Procedure:

- (a) Form the augmented matrix  $[A | I_n]$
- (b) Run Gauss-Jordan on  $[A | I_n]$ .
- (c) If the output is  $[I_n | B]$  then  $B = A^{-1}$ .  
Otherwise  $A$  is singular.

← why does this work?  
Next time.

Eg: Compute  $\begin{bmatrix} 2 & 3 \\ 1 & 2 \end{bmatrix}^{-1}$ .

$$\begin{aligned} \left[ \begin{array}{cc|cc} 2 & 3 & 1 & 0 \\ 1 & 2 & 0 & 1 \end{array} \right] & \xrightarrow{R_2 \leftrightarrow R_1} \left[ \begin{array}{cc|cc} 1 & 2 & 0 & 1 \\ 2 & 3 & 1 & 0 \end{array} \right] \\ & \xrightarrow{R_2 - 2R_1} \left[ \begin{array}{cc|cc} 1 & 2 & 0 & 1 \\ 0 & -1 & 1 & -2 \end{array} \right] \\ & \xrightarrow{R_1 \leftrightarrow R_2} \left[ \begin{array}{cc|cc} 0 & -1 & 1 & -2 \\ 1 & 2 & 0 & 1 \end{array} \right] \\ & \xrightarrow{R_1 \times (-1)} \left[ \begin{array}{cc|cc} 0 & 1 & -1 & 2 \\ 1 & 2 & 0 & 1 \end{array} \right] \\ & \xrightarrow{R_1 - 2R_2} \left[ \begin{array}{cc|cc} 0 & 1 & -1 & 2 \\ 1 & 0 & 2 & -3 \end{array} \right] \\ & \xrightarrow{R_1 \leftrightarrow R_2} \left[ \begin{array}{cc|cc} 1 & 0 & 2 & -3 \\ 0 & 1 & -1 & 2 \end{array} \right] \end{aligned}$$

This has the form  $\left[ I_2 \mid \begin{matrix} 2 & -3 \\ -1 & 2 \end{matrix} \right]$

$$\text{So } \begin{bmatrix} 2 & 3 \\ 1 & 2 \end{bmatrix}^{-1} = \begin{bmatrix} 2 & -3 \\ -1 & 2 \end{bmatrix}.$$

Eg: Compute  $\begin{bmatrix} 1 & 3 \\ -1 & -3 \end{bmatrix}^{-1}$ .

$$\begin{aligned} \left[ \begin{array}{cc|cc} 1 & 3 & 1 & 0 \\ -1 & -3 & 0 & 1 \end{array} \right] & \xrightarrow{R_2 + R_1} \left[ \begin{array}{cc|cc} 1 & 3 & 1 & 0 \\ 0 & 0 & 1 & 1 \end{array} \right] \\ & \xrightarrow{R_1 - R_2} \left[ \begin{array}{cc|cc} 1 & 3 & 0 & -1 \\ 0 & 0 & 1 & 1 \end{array} \right] \end{aligned}$$

But this is in RREF and it does not have the form  $[I_2 \mid B] \Rightarrow$  singular.

NB: We knew this after the first step: no pivot in the second column.

Actually there's a **shortcut** for **2x2** matrices:

**Fact:**  $\begin{bmatrix} a & b \\ c & d \end{bmatrix}$  is invertible  $\iff ad-bc \neq 0$ ,  
in which case

$$\begin{bmatrix} a & b \\ c & d \end{bmatrix}^{-1} = \frac{1}{ad-bc} \begin{bmatrix} d & -b \\ -c & a \end{bmatrix}$$

**Eg:**  $\begin{bmatrix} 2 & 3 \\ 1 & 2 \end{bmatrix}^{-1} = \frac{1}{2 \cdot 2 - 3 \cdot 1} \begin{bmatrix} 2 & -3 \\ -1 & 2 \end{bmatrix}$

**Check:**  $\frac{1}{ad-bc} \begin{bmatrix} d & -b \\ -c & a \end{bmatrix} \begin{bmatrix} a & b \\ c & d \end{bmatrix}$   
 $= \frac{1}{ad-bc} \begin{bmatrix} ad-bc & bd-bd \\ ac-ac & ad-bc \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$  ✓

What is this good for?

Suppose  $A$  is invertible. Let's solve  $Ax=b$ .

$$Ax=b \iff A^{-1}(Ax) = A^{-1}b$$

$$\iff (A^{-1}A)x = A^{-1}b$$

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

For invertible  $A$ :

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

In particular,  $Ax=b$  has exactly one solution for any  $b$ , and we have an expression for  $b$  in terms of  $x$

Eg: Solve 
$$\begin{aligned} 2x_1 + 3x_2 &= b_1 \\ x_1 + 2x_2 &= b_2 \end{aligned}$$

$$A = \begin{bmatrix} 2 & 3 \\ 1 & 2 \end{bmatrix} \quad A^{-1} = \begin{bmatrix} 2 & -3 \\ -1 & 2 \end{bmatrix}$$

$$A \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} b_1 \\ b_2 \end{bmatrix} \iff \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = A^{-1} \begin{bmatrix} b_1 \\ b_2 \end{bmatrix} = \begin{bmatrix} 2 & -3 \\ -1 & 2 \end{bmatrix} \begin{bmatrix} b_1 \\ b_2 \end{bmatrix}$$

$$\iff \begin{aligned} x_1 &= 2b_1 - 3b_2 \\ x_2 &= -b_1 + 2b_2 \end{aligned}$$

So if you want to solve 
$$\begin{aligned} 2x_1 + 3x_2 &= 3 \\ x_1 + 2x_2 &= 4 \end{aligned}$$

$$\Rightarrow x_1 = 2(3) - 3(4) = -6$$

$$x_2 = -(3) + 2(4) = 5$$