

Gaussian Elimination – Example 1

Consider a full 2×2 linear system to start with:

$$\begin{array}{rcl} x_1 + 3x_2 & = & 5 \\ \textcolor{red}{2}x_1 + 4x_2 & = & 6 \end{array} \quad (Ax = b)$$

Step 1: Reduce $Ax = b$ to an upper triangular system $Ux = z$.

- Subtract 2 times equation 1 from equation 2: $r_2 - 2r_1 \rightarrow r_2$ ($r_i = \text{“row } i\text{”}$)

$$r_2 - 2r_1 \Rightarrow r_2 \Rightarrow 0 - 2x_2 = -4 \\ x_2 = 2$$

- Notice that only the second equation changed.
- Forget what you learned in other courses about Gaussian Elimination:
 - * **DO NOT** eliminate the upper triangular part!
 - * **DO NOT** scale the last equation by $-\frac{1}{2}$!
 - * **DO NOT** perform any other row/column ops! (even if you know they're valid)

Step 2: Once system is in upper triangular form, solve using backward substitution:

$$x_2 = \textcolor{blue}{2}$$

$$x_1 = \textcolor{blue}{-1}$$

Gaussian Elimination – Example 2

This time, use augmented matrix notation to solve $Ax = b$ where $A = \begin{bmatrix} -3 & 2 & -1 \\ 6 & -6 & 7 \\ 3 & -4 & 4 \end{bmatrix}$ and $b = [-1, -7, -6]^T$.

$$m_{21}r_1 \rightarrow \begin{array}{cccc} 6 & -4 & 2 & 1 \end{array} \quad 2$$

Step 1: Reduce to upper triangular form:

- Write as an equivalent augmented system:

$$[A | b] = \left[\begin{array}{ccc|c} -3 & 2 & -1 & -1 \\ 6 & -6 & 7 & -7 \\ 3 & -4 & 4 & -6 \end{array} \right] \quad (-3 \text{ is called the } \underline{\text{pivot element}})$$

- Use row $i = 1$ to eliminate (zero out) the sub-diagonal entries in column $j = 1$:

(a) compute multipliers $m_{ij} = \frac{a_{ij}}{a_{jj}}$: $m_{21} = \frac{6}{-3} = -2$, $m_{31} = \frac{3}{-3} = -1$

(b) perform elementary row operations: $r_i - m_{ij}r_j \rightarrow r_i$ (for $j = i + 1, \dots, n$)

$$\begin{array}{l} r_2 - m_{21}r_1 \rightarrow r_2 \\ r_3 - m_{31}r_1 \rightarrow r_3 \end{array} \quad \left[\begin{array}{ccc|c} -3 & 2 & -1 & -1 \\ 0 & -2 & 5 & -9 \\ 0 & -2 & 3 & -7 \end{array} \right]$$

- For column $j = 2$, use row 2 and multiplier $m_{32} = \frac{-2}{-2} = 1$ to zero out the last remaining sub-diagonal entry:

$$r_3 - m_{32}r_2 \rightarrow r_3 \quad \left[\begin{array}{ccc|c} -3 & 2 & -1 & -1 \\ 0 & -2 & 5 & -9 \\ 0 & 0 & -2 & 2 \end{array} \right] \Rightarrow \text{upper triangle Matrix}$$

Step 2: Perform back substitution to solve $Ux = z$:

$$x_3 = \frac{2}{-2} = -1$$

$$x_2 = \frac{[-9 - 5(-1)]}{-2} = 2$$

$$x_1 = \frac{-1 + (-1) - 2(2)}{-3} = 2$$

$$\Rightarrow x = \begin{bmatrix} 2 \\ 2 \\ -1 \end{bmatrix} \quad \text{check } Ax=b \text{ to verify solution}$$

Recall: Outcomes of Row Reduction

- When row-reducing an augmented matrix there are three possible outcomes:

1.
$$\left[\begin{array}{ccc|c} -2 & -3 & 1 & -1 \\ 0 & 1 & 0 & 3 \\ 0 & 0 & 8 & -5 \end{array} \right]$$

 $\det = -16 \neq 0$

Original system is solvable

\Rightarrow **unique solution**

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

 $\det = 0$
 Last row: “0 = nonzero”

Matrix is singular and
RHS is inconsistent

\Rightarrow **no solution**

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

 $\det = 0$
 Last row: “0 = 0”

Matrix is singular and
RHS is consistent

\Rightarrow **infinitely many solutions**

- In MACM 316, we’re mainly interested in “uniquely solvable” problems (like 1).
- But any GE algorithm should still “gracefully” handle special cases (like 2,3).

Gaussian Elimination – Example 3

Another 3×3 example with $A = \begin{bmatrix} 2 & 1 & 4 \\ 3 & 2 & 1 \\ -1 & 4 & -2 \end{bmatrix}$ and $b = \begin{bmatrix} 7 \\ 1 \\ 1 \end{bmatrix}$

Step 1: Row-reduce to upper triangular form.

- Rewrite as an augmented system:

$$[A \mid b] = \left[\begin{array}{ccc|c} 2 & 1 & 4 & 7 \\ 3 & 2 & 1 & 1 \\ -1 & 4 & -2 & 1 \end{array} \right]$$

Pivot

$m_{21} = \frac{3}{2}$
 $m_{31} = -\frac{1}{2}$

- Eliminate column 1 entries **BUT** this time store multipliers in place of the sub-diagonal zeroes (why? later ...)

$$\begin{array}{l} r_2 - \frac{3}{2}r_1 \rightarrow r_2 \\ r_3 - (-\frac{1}{2})r_1 \rightarrow r_3 \end{array} \quad \left[\begin{array}{ccc|c} 2 & 1 & 4 & 7 \\ (\frac{3}{2}) & \frac{1}{2} & -5 & -\frac{19}{2} \\ (-\frac{1}{2}) & \frac{9}{2} & 0 & \frac{9}{2} \end{array} \right]$$

Do Not Scale r_2 by 2

$m_{32} = \frac{9/2}{1/2} = 9$

- Storing the multiplier m_{ij} for later, do not use for back substitution.

- Eliminate remaining entry in column 2:

$$r_3 - 9r_2 \rightarrow r_3 \quad \left[\begin{array}{ccc|c} 2 & 1 & 4 & 7 \\ (\frac{3}{2}) & \frac{1}{2} & -5 & -\frac{19}{2} \\ (-\frac{1}{2}) & (9) & 45 & 90 \end{array} \right]$$

Do Not Scale r_3 by $\frac{1}{45}$

Step 2: Finish with backward substitution, ignoring the (m_{ij})

$$x_3 = 2$$

$$x_2 = 2 \left(-\frac{19}{2} + 5(2) \right) = 1$$

$$x_1 = -1$$

$$x = \begin{bmatrix} -1 \\ 1 \\ 2 \end{bmatrix}$$

... this is looking more and more like an algorithm!

Gaussian Elimination (GE) Algorithm

Pseudo-code for row reduction step:

```

for j = 1 : n-1 do
    if A(j,j) = 0 then
        ERROR: Division by zero!
    endif

    % Zero out sub-diagonal entries in col j
    for i = j+1 : n do
        m(i,j) = A(i,j) / A(j,j)    % multiplier

        % Row operation r_i - m_ij*r_j -> r_i
        A(i,j) = 0
        for p = j+1 : n do
            A(i,p) = A(i,p) - m(i,j) * A(j,p)
        enddo
        b(i) = b(i) - m(i,j) * b(j)
    enddo
enddo
if A(n,n) = 0 then
    ERROR: Division by zero!
endif

```

Input:

n : number of unknowns
A(1:n,1:n) : matrix entries
b(1:n) : RHS vector

Output:

x(1:n) : solution vector

Backward substitution step:

```

x(n) = b(n) / A(n,n)
for i = n-1 : -1 : 1 do
    x(i) = b(i)
    for j = i+1 : n do
        x(i) = x(i) - A(i,j) * x(j)
    enddo
    x(i) = x(i) / A(i,i)
enddo
return x

```

Note: Could replace $m(i,j)$ with $A(i,j)$! (store multipliers in A)

Operation Count for GE

Count total number of multiply / divide operations (neglect $+$ / $-$):

[Note: Textbook counts **all** operations: $+$, $-$, \times , \div]

- Cost of row reduction phase, $\mathcal{R}(n)$:

$$\begin{aligned}\mathcal{R}(n) &= 2 \sum_{j=1}^{n-1} j + \sum_{j=1}^{n-1} j^2 \\ &= \frac{2n(n-1)}{2} + \frac{n(n-1)(2n-1)}{6} \\ &= \frac{2n^3 + 3n^2 - 5n}{6} = O(n^3)\end{aligned}$$

Column j	Multipliers and updating RHS	Updating row entries
1	$n-1$	$(n-1)^2$
2	$n-2$	$(n-2)^2$
\vdots	\vdots	\vdots
$n-2$	2	2^2
$n-1$	1	1^2
Total	$2 \sum_{j=1}^{n-1} j$	$\sum_{j=1}^{n-1} j^2$

- Cost of backward substitution, $\mathcal{B}(n)$:

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

- Total operation count:

$$\mathcal{T}(n) = \mathcal{R}(n) + \mathcal{B}(n) = \frac{n^3 + 3n^2 - n}{3} = O(n^3)$$

Question: What fraction of total time is spent on row reduction?

ASIDE: Big-O Notation and Computational Complexity

- When comparing algorithms, big-O notation becomes extremely useful
- Suppose a problem has some size n (for n large) then the total cost or **complexity** of an algorithm may be nicely represented by an expression like:

$$O(n), \quad O(n^p), \quad O(n^2 \log(n)), \quad \text{etc.}$$

- These rates of growth in cost are studied in the next set of lectures (**3c**)
- Here are a few references on big-O notation for large- n growth:

* http://youtu.be/___vX2sj1pXU :
a 5-minute introduction to $O(n^p)$ type costs for simple algorithm components (warning: n and N are often swapped)

* <http://bigocheatsheet.com> :
common CS algorithms compared

