

Lecture 20 GAUSSIAN ELIMINATION (GE)

Solve where $A \in \mathbb{C}^{m \times m}$, nonsingular
 $Ax = b$ $x \in \mathbb{C}^m$, $b \in \mathbb{C}^m$

GE transforms problem into an upper triangular equivalent problem, which can be solved by back substitution.

$$\begin{pmatrix} \text{---} & \text{---} \\ A & \end{pmatrix} \rightarrow \begin{pmatrix} \text{---} & \text{---} \\ U & \end{pmatrix}$$
$$L_{m-1} \cdots L_2 L_1 A = U$$
$$\underbrace{L^{-1}}_{\text{---}}$$
$$\therefore A = LU$$

We use the familiar ROW ECHELON operations to find U (and L)

As far as solving $x = A^{-1}b$, we note that

$$LUx = b \Rightarrow \begin{cases} y = Ux \\ Ly = b \end{cases}$$

Algorithm: (w/o pivoting)

$$\begin{array}{c}
 4 \times 4 \\
 \left(\begin{array}{|c|c|c|c|} \hline & & & \\ \hline \end{array} \right) \xrightarrow{L_1} \left(\begin{array}{|c|c|c|c|} \hline X & & & \\ \hline 0 & & & \\ \hline 0 & & & \\ \hline 0 & & & \\ \hline \end{array} \right) \xrightarrow{L_2} \left(\begin{array}{|c|c|c|c|} \hline X & X & & \\ \hline 0 & X & & \\ \hline 0 & 0 & & \\ \hline 0 & 0 & & \\ \hline \end{array} \right) \xrightarrow{L_3} \left(\begin{array}{|c|c|c|c|} \hline X & X & X & \\ \hline 0 & X & X & \\ \hline 0 & 0 & X & \\ \hline 0 & 0 & 0 & \\ \hline \end{array} \right) \\
 A \qquad \qquad L_1 A \qquad \qquad L_2 L_1 A \qquad \qquad L_3 L_2 L_1 A
 \end{array}$$

$$\xrightarrow{L_4} \left(\begin{array}{|c|c|c|c|} \hline X & X & X & X \\ \hline X & X & X & \\ \hline X & X & & \\ \hline 0 & & & X \\ \hline \end{array} \right) = U$$

$L_4 L_3 L_2 L_1 A$

Rules:

GS

$$A = QR$$

triangular orthogonalization

Householder

$$A = QR$$

orthogonal triangularization

LU

$$A = LV$$

triangular triangularization

$$\text{ex) } \left| \begin{array}{cccc|c} E_1 & \left(\begin{array}{cccc|c} 1 & -1 & 2 & -1 & -8 \\ E_2 & \left(\begin{array}{cccc|c} 0 & 0 & -1 & -1 & -4 \\ E_3 & \left(\begin{array}{cccc|c} 0 & 0 & 1 & -1 & 6 \\ E_4 & \left(\begin{array}{cccc|c} 0 & 0 & 0 & 1 & 4 \end{array} \right) \end{array} \right) \end{array} \right) \end{array} \right) \xrightarrow{E_2 = E_2 - 2E_1} \left(\begin{array}{cccc|c} 1 & -1 & 2 & -1 & -8 \\ 0 & 0 & -1 & -1 & -4 \\ 0 & 0 & 1 & -1 & 6 \\ 0 & 0 & 0 & 1 & 4 \end{array} \right) \xrightarrow{E_3 = E_3 - E_1} \left(\begin{array}{cccc|c} 1 & -1 & 2 & -1 & -8 \\ 0 & 0 & -1 & 0 & -2 \\ 0 & 0 & 1 & 0 & 4 \\ 0 & 0 & 0 & 1 & 4 \end{array} \right)$$

$$\left(\begin{array}{cccc|c} 1 & -1 & 2 & -1 & -8 \\ 0 & 0 & -1 & -1 & -4 \\ 0 & 0 & 1 & 1 & 6 \\ 0 & 0 & 0 & 1 & 4 \end{array} \right) \xrightarrow{E_3 \rightarrow E_2} \left(\begin{array}{cccc|c} 1 & -1 & 2 & -1 & -8 \\ 0 & 0 & -1 & 1 & 6 \\ 0 & 0 & 1 & -1 & -4 \\ 0 & 0 & 0 & 1 & 4 \end{array} \right) \xrightarrow{E_4 = E_4 - E_1} \left(\begin{array}{cccc|c} 1 & -1 & 2 & -1 & -8 \\ 0 & 0 & -1 & 1 & 6 \\ 0 & 0 & 1 & -1 & -4 \\ 0 & 0 & 0 & 0 & 0 \end{array} \right)$$

$$\left(\begin{array}{cccc|c} 1 & -1 & 2 & -1 & -8 \\ 0 & 0 & -1 & 1 & 6 \\ 0 & 0 & 1 & -1 & -4 \\ 0 & 0 & 0 & 0 & 0 \end{array} \right) \xrightarrow{(A|b) \rightarrow (L_1 A | L_1 b)} \left(\begin{array}{cccc|c} 1 & -1 & 2 & -1 & -8 \\ 0 & 0 & -1 & 1 & 6 \\ 0 & 0 & 1 & -1 & -4 \\ 0 & 0 & 0 & 0 & 0 \end{array} \right) \xrightarrow{E_4 = E_4 + 2E_3} \left(\begin{array}{cccc|c} 1 & -1 & 2 & -1 & -8 \\ 0 & 0 & -1 & 1 & 6 \\ 0 & 0 & 1 & -1 & -4 \\ 0 & 0 & 0 & 0 & 12 \end{array} \right)$$

$$\left(\begin{array}{cccc|c} 1 & -1 & 2 & -1 & -8 \\ 0 & 0 & -1 & 1 & 6 \\ 0 & 0 & 1 & -1 & -4 \\ 0 & 0 & 0 & 0 & 12 \end{array} \right) \xrightarrow[L^{-1}]{(L_2 L_1 A | L_2 L_1 b)} \left(\begin{array}{cccc|c} 1 & -1 & 2 & -1 & -8 \\ 0 & 0 & -1 & 1 & 6 \\ 0 & 0 & 1 & -1 & -4 \\ 0 & 0 & 0 & 0 & 12 \end{array} \right) \xrightarrow[L^{-1}]{\text{so } L^{-1}A = U}$$

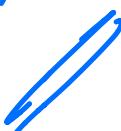
Then we can back substitute
from bottom to top:

$$2x_4 = 4 \Rightarrow x_4 = 2$$

$$-x_3 = x_4 - 4 \Rightarrow x_3 = 2$$

$$2x_2 = 6 + x_3 - x_4 \Rightarrow x_2 = 3$$

$$x_1 = x_2 - 2x_3 + x_4 - 8 \Rightarrow x_1 = -7$$



COMPUTATIONAL COMPLEXITY $Ax=b \Rightarrow (A|b)$

Storage 2D array $(\mathbb{C}^{m,m+1})$

(since the "lower" portion of the array, where zeros go,

can provide all the storage you need for the factors that compute the L matrix

Speed (measured in FLOPS) $\propto O(m^3)$

To see this:

GAUSSIAN ELIMINATION ALGORITHM

Solves $Ax=b$ $m \times m$

[Input $(A|b) = (a_{ij})$ $1 \leq i, j \leq m$ AUGMENTED MATRIX
Output x_i $1 \leq i \leq m$, or error message
% elimination process

- (1) for $i=1:m-1$
 - (2) let p be smallest integer with $i \leq p \leq m$ & $a_{pi} \neq 0$
if no integer p can be found: output('no unique solution')
stop
 - (3) if $p \neq i$ $E_p \leftrightarrow E_i$
 - (4) for $j=i+1:m$
 - (5) $w_{ji} = a_{ji}/a_{ii}$
 - (6) $E_j - w_{ji}E_i \leftrightarrow E_j$
- (7) if $a_{mm} = 0$: output('no unique solution'), stop
- (8) $x_m = a_{m,m+1}/a_{mm}$ % begin backsubstitute
- (9) for $i=m-1:-1$
 $x_i = (a_{i,m+1} - \sum_{j=i+1}^m a_{ij}x_j)/a_{ii}$
- (10) output(x)
end

OPERATION COUNT

In pink are lines of interest.

Look at (5) & (6):

(5) $(m-i)$ divides

(6) multiplies results in $(m-i)(m-i+1)$

then $(m-i)(m-i+1)$ subtracts

So for (5) & (6), for each $i=1, 2, \dots, m-1$

$$\text{Mult/div } (m-i) + (m-i)(m-i+1) = (m-i)(m-i+2)$$

adds/sub $(m-i)(m-i+1)$

$$\therefore \text{Mults/DVs} : \sum_{i=1}^{m-1} (m-i)(m-i+2) = (m^2 + 2m) \sum_{i=1}^{m-1} 1$$

$$- 2(m+1) \sum_{i=1}^{m-1} i + \sum_{i=1}^{m-1} i^2 = \frac{2m^3 + 3m^2 - 5m}{6}$$

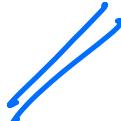
$$\text{Sub/Add} : \sum_{i=1}^{m-1} (m-i)(m-i+1) = (m^2 + m) \sum_{i=1}^{m-1} 1$$

$$-(2m+1) \sum_{i=1}^{m-1} i + \sum_{i=1}^{m-1} i^2 = \frac{m^3 - m}{3}$$

(8) & (9): $1 + (m-i)$ mults, $(m-i+1)$ adds for each summation term, then 1 sub 8 1 divide

$$\therefore \frac{m^2 + m}{2} \text{ (mults)} + \frac{m^2 - m}{2} \text{ (adds)}$$

$$\therefore \underbrace{\frac{m^3 + m^2 - m}{3}}_{\text{mults/dvs}} + \underbrace{\frac{m^3 + 3m^2 - 5m}{6}}_{\text{add/sub}} = \mathcal{O}(m^3)$$



PIVOTING & SCALING: how to minimize the effect
of round off errors in an ill-conditioned matrix problem?

Partial (row) Pivoting: row exchanges

Full (column/row) Pivoting: row & col exchanges

Motivating Example:

$$\text{ex}) \quad \begin{aligned} 0.001x + 1.00y &= 1.00 \\ 1.00x + 1.00y &= 2.00 \end{aligned} \quad \left\{ \begin{array}{l} \text{(f)} \\ \text{(g)} \end{array} \right.$$

True solution (to 5 significant digits)

$$x = 1.00010 \quad y = 0.99990$$

We used exact arithmetic & clamped to 5 sig digits

Solution of (f) to 5 significant digits will be

$$-10000y = -10000 \Rightarrow y = 1.00$$

$$\text{and } x = 0.00 \quad \text{Pretty bad!}$$

We did Gaussian elimination

We will do partial pivoting: all we do here is
switch the rows. Proceed with Gaussian
elimination:

$$1.00x + 1.00y = 2.00$$

$$0.0001x + 1.00y = 1.00$$

Perform GE (3 significant digit)

$$x = 1.00$$

$$y = 1.00$$

'Great Improvement'

ANALYSIS OF PROBLEM w/ ROUND OFF:

$$\text{Ex) } E_1 : 0.003000x_1 + 59.14x_2 = 59.17$$

$$E_2 : 5.291x_1 - 6.130x_2 = 46.78$$

has exact solution to 4-digit rounding:

$$x_1 = 10.00 \quad x_2 = 1.00$$

$a_{11} = 0.003000$ is pivot & is small

$$m_{21} = \frac{5.291}{0.003} = 1763.66 \text{ rounds } 1763.$$

$$E_2 - m_{21}E_1 \rightarrow E_2$$

$$0.003000x_1 + 59.14x_2 = 59.17$$

$$- 104300x_2 = -104400$$

Instead of the more precise

$$0.003000x_1 + 59.14x_2 = 59.17$$

$$- 104309.376x_2 = -104309.376$$

∴ disparity in magnitudes $M_{21}b_1$ and b_2
 has introduced round off, but error has not
 propagated, yet.

Backward solution

$$x_2 = 1.001 \text{ close to } x_2 = 1.000$$

but

$$x_1 = \frac{59.17 - (59.14)(1.001)}{0.003000} = -10.00$$

