

Pitfalls of Gaussian Elimination:-

1. Division by zero

$$\begin{aligned}x_2 + x_3 &= 13 \\5x_1 + 3x_2 + 2x_3 &= 4 \\6x_1 + 2x_2 + 8x_3 &= 9\end{aligned}$$

Solution: Pivoting \rightarrow exchange the rows so that the pivot elements is non-zero

2. Round off errors

computers have a finite precision. Elimination has many divisions \rightarrow has to be rounded off

In substitution \rightarrow round off errors can propagate

Solution (sort of) :- use more significant figures.

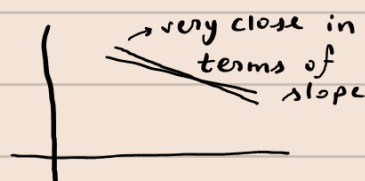
Always check the solution by substituting in original equations.

3. 122? - conditioned systems

$$\left. \begin{aligned}x_1 + 2x_2 &= 10 \\1.1x_1 + 2x_2 &= 10.4\end{aligned} \right\} \begin{array}{l} \text{solution} \\ x_1 = 4, x_2 = 3 \end{array}$$

\swarrow

$$1.05x_1 + 2x_2 = 10.4 \quad \text{Solution}$$



If we change coefficient $x_1 = 8, x_2 = 1$ slightly, solution changes a lot

$$a_{11}x_1 + a_{12}x_2 = b_1$$

$$a_{21}x_1 + a_{22}x_2 = b_2$$

$$x_2 = \frac{-a_{11}}{a_{12}}x_1 + \frac{b_1}{a_{12}}$$

$$x_2 = \frac{-a_{21}}{a_{22}}x_1 + \frac{b_2}{a_{22}}$$

$$\frac{-a_{11}}{a_{12}} \cong \frac{-a_{21}}{a_{22}}$$

$$\underbrace{a_{11}a_{22} - a_{12}a_{21}}_{\substack{\uparrow \\ \text{Determinant}}} \cong 0$$

Multiply by constant

→ system still remains ill-conditioned but determinant changes.

4. Singular Systems Rows are LI

determinant is zero

Output of Gaussian elimination - upper triangular matrix

Determinant of upper triangular matrix - product of diagonals

If during the elimination process, any diagonal entry is 0 → output that system is singular.

Improving Solutions :-

- ① Use more significant figures
- ② Pivoting we don't want to divide by numbers close to zero entries

look for the largest element in the column →
switch that row with the current one to make the
largest element the pivot
→ partial pivoting

Full pivoting { Look for the largest element in the matrix
make that the pivot
→ not used in practice

- ③ Scaling Required when some equations have much larger coefficients than others

$$2x_1 + 100000x_2 = 100002$$

$$x_1 + x_2 = 2$$

② { Correct : $x_1 = 1.00002$ $x_1 = x_2 = 1$
 $x_2 = 0.99998$

Scaling → look at a column, normalise all the entries to maximum of 1.

Gauss - Jordan!

Instead of eliminating unknown from all subsequent equations, eliminate from all (including previous)

Also normalise so that leading coefficients is 1.

$$\begin{bmatrix} a_{11} & 0 & \dots & 0 \\ 0 & a_{22} & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & \dots & \dots & a_{nn} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} = \begin{bmatrix} b_1' \\ b_2' \\ \vdots \\ b_n' \end{bmatrix}$$

$$a_{ii} = 1 \quad \forall i \in [n]$$

$$\text{operations} \rightarrow n^3 + O(n^2)$$

Gauss-Jordan (arg, n, m, x)

for k = 1 to m

d = arg(k, k)

for j = 1 to n

arg(k, j) = arg(k, j) / d

end for

for i = 1 to m

if i ≠ k

d = arg(i, k)

for j = k to n

arg(i, j) = arg(i, j) - d * arg(k, j)

end for

end if

end for

end for

n = m + 1

n = no. of columns

m = no. of rows

Exactly as
gaussian

?

for k = 1 to m

(k, k) = arg(k, n)

end for

