CE 311K: Linear System of Equations

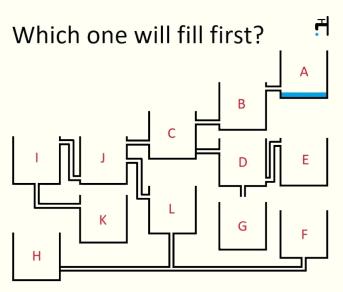
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Linear System of Equations

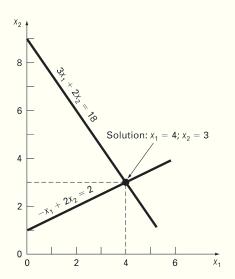
Linear System of Equations



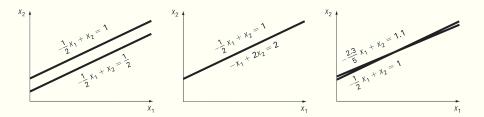
Solving Linear System of Equations

$$3x_1 + 2x_2 = 18$$
$$-x_1 + 2x_2 = 2$$

Solving Linear System of Equations



Singularity and III-conditioned



Linear System of Equations

-Singularity and III-conditioned

A singular matrix is a square matrix which is not invertible. Alternatively, a matrix is singular if and only if it has a determinant of 0.

Solving Linear System of Equations

- Direct Methods
 - Gauss Elimination
 - @ Gauss-Jordan Elimination
 - U decomposition
- Iterative Methods
 - Jacobi iterative
 - @ Gauss-Seidel

Direct methods

Consider a system of 3 linear equations for simplicity:

$$a_{11}x_1 + a_{12}x_2 + a_{13}x_3 = b_1$$

 $a_{21}x_1 + a_{22}x_2 + a_{23}x_3 = b_2$
 $a_{31}x_1 + a_{32}x_2 + a_{33}x_3 = b_3$

Matrix form is:

$$\begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} b_1 \\ b_2 \\ b_3 \end{bmatrix}$$

Concise form: Ax = b

Systems that can be solved easily

$$\begin{bmatrix} a_{11} & a_{12} & a_{13} \\ 0 & a_{22} & a_{23} \\ 0 & 0 & a_{33} \end{bmatrix} \quad \begin{bmatrix} a_{11} & 0 & 0 \\ a_{21} & a_{22} & 0 \\ a_{31} & a_{32} & a_{33} \end{bmatrix} \quad \begin{bmatrix} a_{11} & 0 & 0 \\ 0 & a_{22} & 0 \\ 0 & 0 & a_{33} \end{bmatrix}$$

Solve by "back substitution' Upper triangle system (U)

$$\begin{bmatrix} a_{11} & a_{12} & a_{13} \\ 0 & a_{22} & a_{23} \\ 0 & 0 & a_{33} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} \begin{bmatrix} b_1 \\ b_2 \\ b_3 \end{bmatrix}$$

Start with the last equation 3: $x_3 = b_3/a_{33}$

Equation 2: $a_{22}x_2 + a_{23}x_3 = b_2$ so

$$x_2 = \frac{b_2 - a_{23}x_3}{a_{22}}$$

Equation 1: $a_{11}x_1 + a_{12}x_2 + a_{13}x_3 = b_1$ so

$$x_1 = \frac{b_1 - (a_{12}x_2 + a_{13}x_3)}{a_{11}}$$

General for 'n' systems: $x_n = b_n/a_{nn} \ x_i = \frac{b_i - \sum_{j=i+1}^n a_{ij} x_j}{a_{ii}}$

Gauss Elimination

Consider a system of 3 linear equations:

$$\begin{bmatrix} 2 & 4 & 6 \\ 4 & 11 & 21 \\ 6 & 21 & 52 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} \begin{bmatrix} 24 \\ 72 \\ 158 \end{bmatrix}$$

Divide row 2 by -2 and row 3 by -3

$$\begin{bmatrix} 2 & 4 & 6 \\ 0 & 3 & 9 \\ 0 & 9 & 34 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} \begin{bmatrix} 24 \\ 24 \\ 86 \end{bmatrix}$$

Second reduction:

$$\begin{bmatrix} 2 & 4 & 6 \\ 0 & 3 & 9 \\ 0 & 0 & 7 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} \begin{bmatrix} 24 \\ 24 \\ 14 \end{bmatrix}$$

Linear System of Equations

-Gauss Elimination

```
\begin{bmatrix} 2 & 4 & 6 \\ 4 & 11 & 21 \\ 6 & 21 & 52 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} \begin{bmatrix} 24 \\ 72 \\ 158 \end{bmatrix}
                                                                                                                          \begin{bmatrix} 2 & 4 & 6 \\ 0 & 3 & 9 \\ 0 & 9 & 34 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} \begin{bmatrix} 24 \\ 24 \\ 86 \end{bmatrix}
Second reduction
                                                                                                                             \begin{bmatrix} 2 & 4 & 6 \\ 0 & 3 & 9 \\ 0 & 0 & 7 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} \begin{bmatrix} 24 \\ 24 \\ 14 \end{bmatrix}
x<sub>1</sub> = 2, x<sub>2</sub> = 2, x<sub>3</sub> = 2
```

```
a = np.array([[2,4,6], [4,11,21], [6, 21, 52]])
b = np.array([24, 72, 158])
x = np.linalg.solve(a, b)
```

Gauss Elimination: Limitations

- lacktriangle Prone to round off errors, when we have many (>100) equations.
- ② If coefficient matrix is sparse (lots of zeros), elimination methods are very inefficient.

Gauss Seidel Iterative approach

For conciseness, we limit to 3×3 equations. If diagonal elements are all non-zero, then the equations can be solved as:

$$\begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} b_1 \\ b_2 \\ b_3 \end{bmatrix}$$

$$x_1 = \frac{b_1 - a_{12}x_2 - a_{13}x_3}{a_{11}}$$

$$x_2 = \frac{b_2 - a_{21}x_1 - a_{23}x_3}{a_{22}}$$

$$x_3 = \frac{b_3 - a_{31}x_1 - a_{32}x_2}{a_{33}}$$

Gauss Seidel Iterative approach

Using Gauss-Seidel solve for [x]

$$4x_1 + x_2 + 2x_3 = 4$$
$$3x_1 + 5x_2 + x_3 = 7$$
$$x_1 + x_2 + 3x_3 = 3$$

Assuming an initial guess of $x_1, x_2, x_3 = 0$. End of first iteration.

$$x_1 = \frac{4 - x_2 - 2x_3}{4} = \frac{4 - 0 - 0}{4} = 1$$

$$x_2 = \frac{7 - 3x_1 - x_3}{5} = \frac{7 - 3 * 1 - 0}{5} = 0.8$$

$$x_3 = \frac{3 - x_1 - x_2}{3} = \frac{3 - 1 - 0.8}{3} = 0.4$$

Gauss Seidel Convergence criteria

Convergence can be checked using the relative error.

$$|\varepsilon_{\mathsf{a},i}| = \left| \frac{x_i^k - x_i^{k-1}}{x_i^k} * 100\% \right| < \varepsilon_{\mathsf{s}}$$

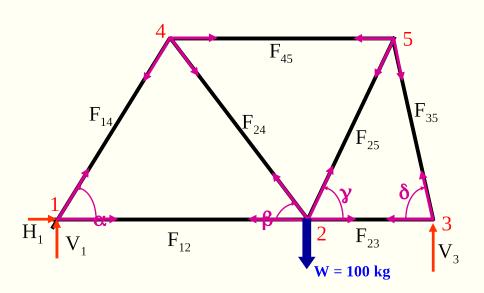
where k, and k-1 represents the current and previous iterations

Convergence can be checked using the relative error.

-Gauss Seidel Convergence criteria

Gauss-Seidel only works for diagonally dominant matrix

Truss analysis



Truss analysis: Force balance

Node 1
$$\begin{cases} \sum F_{y,1} = V_1 + F_{14} \sin \alpha = 0 \\ \sum F_{x,1} = H_1 + F_{12} + F_{14} \sin \alpha = 0 \end{cases}$$
Node 2
$$\begin{cases} \sum F_{y,2} = F_{24} \sin \beta + F_{25} \sin \gamma = 100 \\ \sum F_{x,2} = -F_{12} + F_{23} - F_{24} \cos \beta + F_{25} \cos \gamma = 0 \end{cases}$$
Node 3
$$\begin{cases} \sum F_{y,3} = V_3 + F_{35} \sin \delta = 0 \\ \sum F_{y,3} = -F_{23} - F_{35} \cos \delta = 0 \end{cases}$$
Node 4
$$\begin{cases} \sum F_{y,4} = -F_{14} \sin \alpha - F_{24} \sin \beta = 0 \\ \sum F_{x,4} = -F_{14} \cos \alpha + F_{24} \cos \beta + F_{45} = 0 \end{cases}$$
Node 5
$$\begin{cases} \sum F_{y,5} = -F_{25} \cos \gamma + F_{35} \sin \delta = 0 \\ \sum F_{y,5} = -F_{25} \cos \gamma + F_{35} \cos \delta - F_{45} = 0 \end{cases}$$

Truss analysis: Matrix formulation