Gaussian Elimination

MATLAB Implementation

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Contents

gaussian_elimination	4
Syntax	4
Description	4
Examples	
Links	5
Gaussian Elimination	6
References	8

gaussian elimination

Solves the linear system $A\mathbf{x} = \mathbf{b}$ for \mathbf{x} using Gaussian elimination with partial pivoting.

Syntax

```
x = gaussian_elimination(A,b)
```

Description

 $\mathbf{x} = \mathtt{gaussian_elimination}(\mathbf{A}, \mathbf{b})$ solves the linear system $A\mathbf{x} = \mathbf{b}$ for the vector $\mathbf{x} \in \mathbb{R}^n$, where $A \in \mathbb{R}^{n \times n}$ is a tridiagonal matrix and $\mathbf{b} \in \mathbb{R}^n$ is a vector.

Examples

Example 1

Solve the linear system $A\mathbf{x} = \mathbf{b}$ for \mathbf{x} , where

$$A = \begin{bmatrix} 2 & -1 & 5 \\ 1 & 1 & -3 \\ 2 & 4 & 1 \end{bmatrix}, \quad \mathbf{b} = \begin{bmatrix} 10 \\ -2 \\ 1 \end{bmatrix}$$

■ SOLUTION

Entering A and b into MATLAB,

```
% defines matrix A
A = [2,-1, 5;
          1, 1,-3;
          2, 4, 1];
% defines vector b
b = [10;
          -2;
          1];
```

To solve the linear system for x,

```
x = gaussian elimination(A,b)
```

This yields the result

Links

MATLAB® Central's File Exchange:

https://www.mathworks.com/matlabcentral/fileexchange/89306-gaussian-elimination-gaussian_elimination

GitHub®:

https://github.com/tamaskis/gaussian_elimination-MATLAB

Gaussian Elimination

Gaussian elimination can be used to solve the linear system

$$A\mathbf{x} = \mathbf{b}$$

where $A \in \mathbb{R}^{n \times n}$ and $\mathbf{x}, \mathbf{b} \in \mathbb{R}^{n \times 1}$. In Algorithms 1 and 2 below, we will be referring to the rows and columns of A as well as to the elements of \mathbf{x} at different points. Here are the conventions we use:

$$A = \begin{bmatrix} \mathbf{a}_1 & \mathbf{a}_2 & \dots & \mathbf{a}_j & \dots & \mathbf{a}_{n-1} & \mathbf{a}_n \end{bmatrix} = \begin{bmatrix} \bar{a}_1 \\ \bar{a}_2 \\ \vdots \\ \bar{a}_i \\ \vdots \\ \bar{a}_{n-1} \\ \bar{a}_n \end{bmatrix}, \qquad \mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_i \\ \vdots \\ x_{n-1} \\ x_n \end{bmatrix}$$

Algorithms 1 and 2^1 below implement Gaussian elimination with partial pivoting and are adapted from Algorithm 6.2 in [1, pp. 374–375]. Note that ε is used to denote the machine epsilon.

Note that together, these algorithms represent a single algorithm. However, this single algorithm is split into two due to its length and the limitations of typesetting algorithms in LATEX.

Algorithm 1: Gaussian elimination with partial pivoting (part 1).

```
1 Given: A, x, b
   // determines n (where A \in \mathbb{R}^{n \times n})
 n = \operatorname{size}(A, 1)
   // redefines A by augmenting the original A with b
 A = [A \ b]
    // initializes boolean variable to keep track if matrix is singular
 4 \text{ singular} = \text{false}
    // elimination process
 \mathbf{5} \ \mathbf{for} \ i = 1 \ \mathbf{to} \ n - 1 \ \mathbf{do}
        // determines pivot row
        Define the vector \mathbf{p} = (i, i + 1, ..., n - 1, n)^T.
 6
         Remove the elements p_k of \mathbf{p} where A_{k,i} = 0.
        p = \min(\mathbf{p})
        // exit the loop if all possible pivots in the i^{th} column are zero (to machine
           precision) because that would result in a singular matrix
        if \max |\mathbf{a}_i| \leq \varepsilon then
 9
             singular = true
10
             Exit loop – the matrix is singular and no unique solution exists.
11
         end
12
        // switches pivot row with the i^{th} row if p \neq i
        if p \neq i then
13
             Store the i^{\text{th}} row of A as \bar{a}_i.
14
             Store the p^{th} row of A as \bar{a}_p.
15
             Set the i^{th} row of A to \bar{a}_p.
16
             Set the p^{\text{th}} row of A to \bar{a}_i.
17
         end
18
        // elementary row operation
         for j = i + 1 to n do
19
             \bar{a}_j = \bar{a}_j - \left(\frac{A_{j,i}}{A_{i,i}}\right) \bar{a}_i
20
         end
21
22 end
```

Algorithm 2: Gaussian elimination with partial pivoting (part 2).

// determines if A is singular (if the bottom right element of A is 0 (to within machine precision), then the entire bottom row is 0 and A is singular)

- $\begin{array}{ll} \mathbf{1} \ \ \mathbf{if} \ |A_{n,n}| \leq \varepsilon \ \mathbf{then} \\ \mathbf{2} \ \ | \ \ \mathrm{singular} = \mathrm{true} \\ \mathbf{3} \ \ \mathbf{end} \end{array}$
- 4 Preallocate/initialize the solution vector $\mathbf{x} \in \mathbb{R}^{n \times 1}$.

// perform backward substitution to solve $A\mathbf{x} = \mathbf{b}$ if A is nonsingular

$$\begin{array}{c|c} \mathbf{5} \ \ \mathbf{if} \ ! (\text{singular}) \ \mathbf{then} \\ \mathbf{6} \\ \\ \mathbf{6} \\ \\ \mathbf{6} \\ \\ \mathbf{7} \\ \\ \mathbf{for} \ i = \frac{A_{n,n+1}}{A_{n,n}} \\ \mathbf{7} \\ \\ \mathbf{for} \ i = n-1 \ \mathbf{to} \ 1 \ \mathbf{by} -1 \ \mathbf{do} \\ \mathbf{8} \\ \mathbf{9} \\ \\ \\ \mathbf{S} = 0 \ \mathbf{for} \ i+1 \ \mathbf{to} \ n \ \mathbf{do} \\ \\ \\ \\ \mathbf{S} = S + A_{i,j} x_j \\ \\ \mathbf{end} \\ \mathbf{11} \\ \\ \\ \mathbf{x}_i = \frac{A_{i,n+1} - S}{A_{i,i}} \\ \\ \mathbf{end} \\ \mathbf{12} \\ \\ \mathbf{end} \\ \mathbf{13} \ \mathbf{end} \\ \\ \mathbf{13} \ \mathbf{end} \\ \end{array}$$

References

[1] Richard L. Burden and J. Douglas Faires. *Numerical Analysis*. 9th. Boston, MA: Brooks/Cole, Cengage Learning, 2011.