Solving Linear Systems by using Jacobi, GS, SOR

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Review

Jacobi, GS, SOR

Part 1 Review: Jacobi Method & Gauss – Seidel Method

Jacobi Method

$$A = M - N$$

$$M = D$$

$$N = L + U$$

$$\mathsf{A} = \mathsf{M} - \mathsf{N}$$

$$\mathsf{M} = \mathsf{D}$$

$$\mathsf{N} = \mathsf{L} + \mathsf{U}$$

$$M = \begin{bmatrix} a_{11} & & & \\ & \ddots & & \\ & & a_{nn} \end{bmatrix}, \quad N = -\begin{bmatrix} 0 & a_{12} & \cdots & a_{1n} \\ a_{21} & \ddots & & \vdots \\ \vdots & & \ddots & \vdots \\ a_{n1} & \cdots & a_{n,n-1} & 0 \end{bmatrix}.$$

Gauss – Seidel Method

$$A = M - N$$

$$M = D - L$$

$$N = U$$

$$M\mathbf{x}^{(k+1)} = N\mathbf{x}^{(k)} + \mathbf{b}.$$

$$(D - L)\mathbf{x}^{(k+1)} = \mathbf{b} + U\mathbf{x}^{(k)}$$

Part 1 Review

Jacobi Method

$$x_i^{(k+1)} = \frac{1}{a_{ii}} \left[b_i - \sum_{j \neq i} a_{ij} x^{(k)} \right]$$

Gauss – Seidel Method

$$x_i^{(k+1)} = \frac{1}{a_{ii}} \left[b_i - \sum_{j=1}^{i-1} a_{ij} x_j^{(k+1)} - \sum_{j=i+1}^{n} a_{ij} x_j^{(k)} \right]$$

Successive Over - Relaxation Method

$$x_i^{(k+1)} = \frac{\omega}{a_{ii}} \left[b_i - \sum_{j=1}^{i-1} a_{ij} x_j^{(k+1)} - \sum_{j=i+1}^{N} a_{ij} x_j^{(k)} \right] + (1 - \omega) x_i^{(k)}$$

Problem #1

Part 2 Problem #1

설명: 푸아송 방정식의 블록 삼중대각 행렬(희소 행렬)

구문:

● A = gallery('poisson',n)은 n×n 메시에 5점 연산자를 사용한 푸아송 방정식의 이산화에서 얻은 차수 n^2의 블록 삼중대각 희소 행렬을 반환합니다.

$$A = \begin{pmatrix} 4 & -1 & 0 & 0 & 0 & -1 & 0 & 0 & \cdots & 0 \\ -1 & 4 & -1 & 0 & 0 & 0 & -1 & 0 & \cdots & 0 \\ 0 & -1 & 4 & -1 & 0 & 0 & 0 & -1 & \cdots & 0 \\ 0 & 0 & -1 & 4 & -1 & 0 & 0 & 0 & \cdots & 0 \\ 0 & 0 & 0 & -1 & 4 & -1 & 0 & 0 & \cdots & 0 \\ -1 & 0 & 0 & 0 & -1 & 4 & -1 & 0 & \cdots & 0 \\ 0 & -1 & 0 & 0 & 0 & -1 & 4 & -1 & \cdots & 0 \\ \vdots & \vdots \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & \cdots & 4 \end{pmatrix}$$

$$\chi = \begin{pmatrix} 1 \\ 1 \\ 1 \\ \cdot \\ \cdot \\ 1 \\ 1 \end{pmatrix}$$

$$\Rightarrow Ax = b$$

Part 3 Code: Jacobi Method

```
%% Algorithm: Jacobi Method
 2 -
       function [x,itr] = jacobi(A,b,x,tol)
       normVal=Inf;
       itr = 0;
       n=size(x,1);
       while normVal>tol
           xold=x;
 8
           for i=1:n
 9 🖹
10
11
               sigma=0;
12
13
               for j=1:n
14
                   if j~=i
15
                       sigma=sigma+A(i,j)*xold(j);
16
17
                   end
18
19
               end
20
               x(i)=(b(i)-sigma)/A(i,i);
22
           end
23
           itr=itr+1;
24
25
           normVal=norm(xold-x);
26
       end
27
```

Part 3 Code: Gauss – Seidel Method

```
1
       %% Algorithm: Gauss Seidel Method
       function [x,itr] = gs(A,b,x,tol)
 2 -
       normVal=1;
       itr = 0;
       n=size(x,1);
       while normVal>tol
           xold=x;
 8
9 =
           for i=1:n
10
11
               sigma=0;
12
13 🗀
               for j=1:i-1
14
                       sigma=sigma+A(i,j)*x(j);
15
               end
16
17 🗀
               for j=i+1:n
18
                       sigma=sigma+A(i,j)*xold(j);
19
               end
20
21
               x(i)=(1/A(i,i))*(b(i)-sigma);
22
           end
23
24
           itr=itr+1;
25
           normVal=norm(xold-x);
26
       end
27
```

Part 3

Code: Successive Over - Relaxation Method

```
%% Algorithm: SOR Method
       function [x,itr] = sor(A,b,x,w,tol)
       normVal=Inf;
 3
       itr = 0;
       n=size(x,1);
       while normVal>tol
           xold=x;
 8
           for i=1:n
 9
10
               sigma=0;
11
12 E
              for j=1:i-1
                       sigma=sigma+A(i,j)*x(j);
13
14
               end
15
               for j=i+1:n
16
                       sigma=sigma+A(i,j)*xold(j);
17
18
               end
19
20
               x(i)=w*(1/A(i,i))*(b(i)-sigma)+(1-w)*xold(i);
           end
22
23
           itr=itr+1;
           normVal=norm(xold-x);
24
25
       end
26
27
```

```
\omega = \frac{2}{1 + \sqrt{1 - [\rho(T_j)]^2}}.
```

Part 4 Conclusion

Iteration number	n = 5	n = 10
Jacobi	78	255
Gauss-Seidel	43	137
SOR	17	31
	$\omega = 1.33$	$\omega = 1.56$