MATH 105LA FINAL PROJECT

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1 Introduction

In this project, we develop a method of calculating the error of the estimated solution $\mathbf{u}^{(k)}$ to the linear system

corresponding to the finite difference approximation for the differential equation

$$\frac{d^2 u(x)}{dx^2} = f(x), \qquad x \in [0, \pi], \qquad f(x) = \sin(x), \qquad u(0) = u(\pi) = 0.$$

We implement the symmetric power method and compute the largest eigenvalue of T_j (as well as T_g and T_s) to show that the norm of the error will decay to zero as k approaches infinity and therefore the Jacobi method (and the Gauss-Seidel and SOR methods) converges to the exact solution $\mathbf{u}_{\text{exact}}$ of the linear system. Finally, we compare the speed of convergence of each method and the dependence of the convergence rate on the number of iterations.

2 Algorithm

Links to the MATLAB files: symmjacobi.m symmgauss.m symmSOR.m

The symmjacobi function takes the inputs n (the dimension of A), tol (tolerance), and N (the maximum number of iterations) and outputs mu (the approximate dominant eigenvalue) and v (the approximate eigenvector). First, it determines the matrix A based on n and sets the initial eigenvector guess to be the vector of dimension n where each component is 1. Then it obtains D, L, U, and T_j from A and normalizes the initial vector guess. On each iteration, it finds y=T_j*v, mu=transpose(v)*y, the error, and v=y/norm(y), the kth eigenvector.

If norm(y) is 0, it states T_j has the eigenvalue 0 and ends the algorithm. When the error is less than the given tolerance, it ends the algorithm. Otherwise, it sets k=k+1 for the next iteration and stops when it reaches the maximum number of iterations.

The symmgauss and symmSOR functions work similarly, but symmgauss obtains T_g instead of T_j and symmSOR obtains T_s from symmjacobi by setting $w=2/(1+sqrt(1-(symmjacobi(n,1e-6,300))^2))$.

3 Answers

a) Let $\mathbf{u}_{exact} = \mathbf{u}$

$$\begin{split} D &= \begin{bmatrix} -2 & \cdots & 0 \\ \vdots & \ddots & \vdots \\ 0 & \cdots & -2 \end{bmatrix} \Rightarrow D^{-1} = \begin{bmatrix} -\frac{1}{2} & \cdots & 0 \\ \vdots & \ddots & \vdots \\ 0 & \cdots & -\frac{1}{2} \end{bmatrix} \\ (L+U) &= \begin{bmatrix} 0 & 1 & \cdots & 0 \\ \vdots & \ddots & \ddots & \vdots \\ 1 & \ddots & \ddots & 1 \\ 0 & \cdots & 1 & 0 \end{bmatrix} \Rightarrow (L+U)\mathbf{u}_{exact} = \begin{bmatrix} u_2 \\ u_1 + u_3 \\ \vdots \\ u_{n-2} + u_n \\ u_{n-1} \end{bmatrix} \\ &\Rightarrow -D^{-1}(L+U)\mathbf{u}_{exact} = \frac{1}{2} \begin{bmatrix} u_2 \\ u_1 + u_3 \\ \vdots \\ u_{n-2} + u_n \\ u_{n-1} \end{bmatrix} \\ D^{-1}\mathbf{f} &= -\frac{1}{2}h^2 \begin{bmatrix} f_1 \\ \vdots \\ f_n \end{bmatrix} = \begin{bmatrix} u_1 - \frac{1}{2}u_2 \\ -\frac{1}{2}u_1 + u_2 - \frac{1}{2}u_3 \\ \vdots \\ -\frac{1}{2}u_{n-2} + u_{n-1} - \frac{1}{2}u_n \\ -\frac{1}{2}u_{n-1} + u_n \end{bmatrix} \\ &\Rightarrow -D^{-1}(L+U)\mathbf{u}_{exact} + D^{-1}\mathbf{f} = \begin{bmatrix} \frac{1}{2}u_2 + u_1 - \frac{1}{2}u_2 \\ \frac{1}{2}u_1 + \frac{1}{2}u_3 - \frac{1}{2}u_1 + u_2 - \frac{1}{2}u_3 \\ \vdots \\ \frac{1}{2}u_{n-2} + \frac{1}{2}u_{n-1} + u_n \end{bmatrix} = \begin{bmatrix} u_1 \\ u_2 \\ \vdots \\ u_{n-1} \\ u_n \end{bmatrix} = \mathbf{u}_{exact} \end{split}$$

 \Rightarrow \mathbf{u}_{exact} is a fixed point for the iteration

b)
$$\mathbf{e}^{(k+1)} = \mathbf{u}^{(k+1)} - \mathbf{u}_{exact}$$

 $= -D^{-1}(L+U)\mathbf{u}^{(k)} + D^{-1}\mathbf{f} + D^{-1}(L+U)\mathbf{u}_{exact} - D^{-1}\mathbf{f}$
 $= -D^{-1}(L+U)\mathbf{u}^{(k)} + D^{-1}(L+U)\mathbf{u}_{exact}$
 $= -D^{-1}(L+U)(\mathbf{u}^{(k)} - \mathbf{u}_{exact})$
 $= -D^{-1}(L+U)\mathbf{e}^{(k)}$

c)
$$e^{(k+1)} = T_j e^k = T_j^{k+1} e^0$$
 (*)

 $T_j \mathbf{v} = \lambda \mathbf{v} \Rightarrow T_j^{k+1} \mathbf{v} = \lambda^{k+1} \mathbf{v}$ $\Rightarrow \mathbf{e}^{(k+1)} = \lambda^{k+1} \mathbf{v}$ is a nontrivial solution to (*) if and only if λ is an eigenvalue of T_j and \mathbf{v} is the associated eigenvector.

Suppose that $|\lambda| \ge 1$. Then $\rho(T_j) \ge |\lambda| \ge 1 \Rightarrow ||\mathbf{e}^{k+1}||_2 = ||\lambda^{k+1}\mathbf{v}||_2 = |\lambda|^{k+1}||\mathbf{v}||_2 \ne 0$ as $k \to \infty$. Suppose that $\rho(T_j) < 1$. Then $|\lambda| \le \rho(T_j) < 1 \Rightarrow ||\mathbf{e}^{(k+1)}||_2 = |\lambda|^{k+1}||\mathbf{v}||_2 \to 0$ as $k \to \infty$.

d) Jacobi

Using tol=1e-6, and N=300:

```
n=3: For N=10, 100, and 1000, mu=0.6667 \Rightarrow \rho(T_i)=0.6667 < 1 \Rightarrow ||e^{(k+1)}||_2 \rightarrow 0 \text{ as } k \rightarrow 0
 \infty
  >> [mu, v] = symmjacobi(3, 1e-6, 300)
  The maximum number of iterations exceeded
  mu =
         0.6667
  v =
        0.5774
        0.5774
        0.5774
n=5: For N=10, 100, and 1000, mu=0.8571 \Rightarrow \rho(T_i)=0.8571<1 \Rightarrow ||\mathbf{e}^{(k+1)}||_2 \rightarrow 0 \text{ as } k \rightarrow 0
 \infty
  >> [mu, v] = symmjacobi (5, 1e-6, 300)
  The maximum number of iterations exceeded
  mu =
        0.8571
        0.3086
        0.4629
        0.6172
        0.4629
        0.3086
```

- n=7: For N=10, 100, and 1000, mu=0.9210 \Rightarrow $\rho(T_j)$ =0.9210<1 \Rightarrow $||\mathbf{e}^{(k+1)}||_2 \rightarrow 0$ as $k \rightarrow \infty$

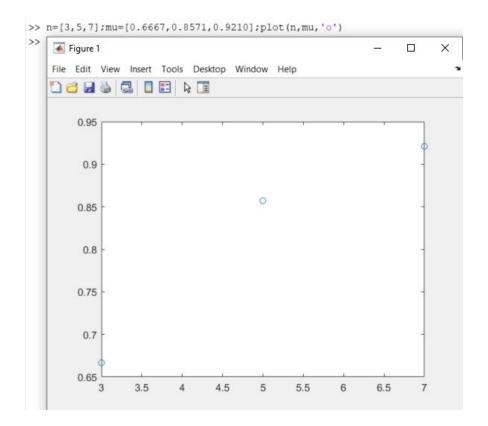
```
>> [mu,v]=symmjacobi(7,1e-6,300)
The maximum number of iterations exceeded
mu =

0.9210

v =

0.1988
0.3393
0.4798
0.4798
0.4798
0.3393
0.1988
```

- Plotting largest eigenvalue against n:



e) Gauss-Seidel

Using tol=1e-6, and N=300:

```
n=3: For N=10, 100, and 1000, mu=0.5000 \Rightarrow \rho(T_g)=0.5000<1 \Rightarrow ||e^{(k+1)}||_2 \rightarrow 0 as k \rightarrow \infty

>> [mu,v]=symmgauss(3,1e-6,300)
The maximum number of iterations exceeded mu = 0.5000

v = 0.6667
0.6667
0.3333
```

- n=5: For N=10, 100, and 1000, mu=0.7500 \Rightarrow $\rho(T_g)$ =0.7500<1 \Rightarrow $\|e^{(k+1)}\|_2 \rightarrow 0$ as $k \rightarrow \infty$

- n=7: For N=10, 100, and 1000, mu=0.8536 \Rightarrow $\rho(T_g)$ =0.8536<1 \Rightarrow $\|\mathbf{e}^{(k+1)}\|_2 \rightarrow 0$ as $k \rightarrow \infty$

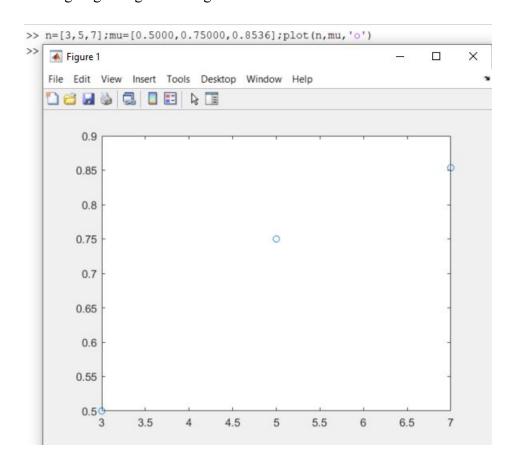
```
>> [mu,v]=symmgauss(7,1e-6,300)
The maximum number of iterations exceeded
mu =

0.8536

v =

0.2395
0.4088
0.4935
0.4935
0.4212
0.2979
0.1489
```

- Plotting largest eigenvalue against n:



Using tol=1e-6, and N=300:

- n=3: For N=10, mu=0.2921 and for N=100, N=1000, mu=0.2918 $\Rightarrow \rho(T_s)=0.2918<1$ $\Rightarrow \|\mathbf{e}^{(k+1)}\|_2 \to 0 \text{ as } k \to \infty$

- n=5: For N=10, mu=0.4522 and for N=100, N=1000, mu=0.4268 $\Rightarrow \rho(T_s)=0.4268<1$

$$\Rightarrow \|\mathbf{e}^{(k+1)}\|_2 \rightarrow 0 \text{ as } k \rightarrow \infty$$

```
>> w=2/(1+sqrt(1-(symmjacobi(5,1e-6,300))^2)); [mu,v]=symmSOR(w,5,1e-6,300)
The maximum number of iterations exceededThe maximum number of iterations exceeded
mu =

0.4268

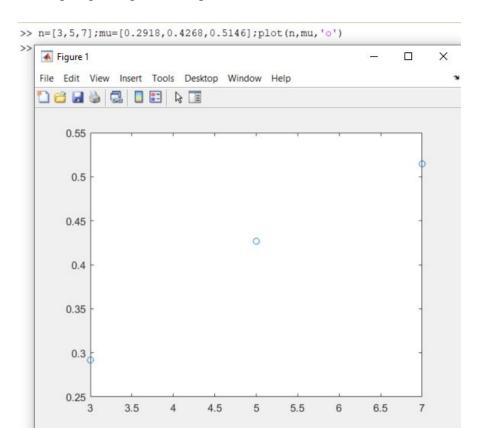
v =

0.5526
0.6252
0.4716
0.2668
0.1006
```

- n=7: For N=10, mu=0.5255 and for N=100, N=1000, mu=0.5146 $\Rightarrow \rho(T_s)=0.5146<1$ $\Rightarrow \|\mathbf{e}^{(k+1)}\|_2 \rightarrow 0 \text{ as } k \rightarrow \infty$

- Plotting largest eigenvalue against n:

0.1460



f) The smaller $\rho(T)$ is, the faster $\rho(T)^k$ converges to 0 and thus the faster λ^k and $\|\mathbf{e}^{(k+1)}\|_2 = \|\lambda^k \mathbf{e}^{(0)}\|_2$ converge to 0. Since the estimated $\rho(T_j)$, $\rho(T_g)$, and $\rho(T_s)$ satisfy $\rho(T_s) \leq \rho(T_g) \leq \rho(T_j)$

for all n, the SOR method converges the fastest. For all methods, $\,\rho(T)$ increases as n increases.

Therefore, as n increases, the rate of convergence decreases.