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MATH 4441

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Exam 2

1.

$$\begin{bmatrix} x(x_0) & \log(x_0) & \cos(x_0) \\ x(x_1) & \log(x_1) & \cos(x_1) \\ x(x_2) & \log(x_2) & \cos(x_2) \\ \dots & \dots & \dots \\ x(x_n) & \log(x_n) & \cos(x_n) \end{bmatrix} * \begin{bmatrix} c_0 \\ c_1 \\ c_2 \end{bmatrix} = \begin{bmatrix} y_0 \\ y_1 \\ y_2 \\ \dots \\ y_n \end{bmatrix}$$

2.

See code

3.

For Jacobi: For the matrix where $a_{11}, a_{22} \neq 0$, the way to determine if it converges or diverges is through the special radius of the B matrix $\rho(B) : B = D^{-1}(L + U)x_n$.

$$\begin{aligned} D &= \begin{bmatrix} a_{11} & 0 \\ 0 & a_{22} \end{bmatrix} \Rightarrow D^{-1} = \begin{bmatrix} \frac{1}{a_{11}} & 0 \\ 0 & \frac{1}{a_{22}} \end{bmatrix} \\ U &= \begin{bmatrix} 0 & a_{12} \\ 0 & 0 \end{bmatrix} \text{ and } L = \begin{bmatrix} 0 & 0 \\ a_{21} & 0 \end{bmatrix} \Rightarrow L + U = \begin{bmatrix} 0 & a_{12} \\ a_{21} & 0 \end{bmatrix} \\ \Rightarrow B &= \begin{bmatrix} \frac{1}{a_{11}} & 0 \\ 0 & \frac{1}{a_{22}} \end{bmatrix} * \begin{bmatrix} 0 & a_{12} \\ a_{21} & 0 \end{bmatrix} = \begin{bmatrix} 0 & \frac{a_{12}}{a_{11}} \\ \frac{a_{21}}{a_{22}} & 0 \end{bmatrix} \\ \lambda(B) &= \left[-\frac{\sqrt{a_{12}\sqrt{a_{22}}}}{\sqrt{a_{11}\sqrt{a_{21}}}}, \frac{\sqrt{a_{12}\sqrt{a_{22}}}}{\sqrt{a_{11}\sqrt{a_{21}}}} \right] \\ \rho(B) &= \max(| \pm \frac{\sqrt{a_{12}\sqrt{a_{22}}}}{\sqrt{a_{11}\sqrt{a_{21}}}} |) \end{aligned}$$

For Gauss-Seidel: For the matrix where $a_{11}, a_{22} \neq 0$, the way to determine if it converges or diverges is through the special radius of the B matrix $\rho(B) : B = (D - L)^{-1}(U)$.

$$\begin{aligned} (D - L) &= \begin{bmatrix} a_{11} & 0 \\ -a_{21} & a_{22} \end{bmatrix} \Rightarrow (D - L)^{-1} = \begin{bmatrix} \frac{1}{a_{11}} & 0 \\ \frac{a_{12}}{a_{11}a_{22}} & \frac{1}{a_{22}} \end{bmatrix} \\ (D - L)^{-1} * U &= \begin{bmatrix} 0 & \frac{a_{12}}{a_{11}} \\ 0 & \frac{a_{12} * a_{21}}{a_{11} * a_{22}} \end{bmatrix} \end{aligned}$$

$$\lambda(B) = [0, \frac{a_{12}a_{21}}{a_{11}a_{22}}]$$

$$\rho(B) = \max(|0, \frac{a_{12}a_{21}}{a_{11}a_{22}}|)$$

The speed of convergence really depends on the values for $a_{11}, a_{12}, a_{21}, a_{22}$ quite a bit. For the algorithms to converge, the matrix A must be row diagonally-dominant, which would imply that for Jacobi or Gauss-Seidel iterative methods to work, $a_{11} > a_{12}$ and $a_{21} < a_{22}$. Therefore, $\max(|\pm \frac{\sqrt{a_{12}}\sqrt{a_{22}}}{\sqrt{a_{11}}\sqrt{a_{21}}}|) > \max(|\pm \frac{\sqrt{a_{12}}\sqrt{a_{22}}}{\sqrt{a_{11}}\sqrt{a_{21}}}|)$, as the latter holds the maximum of 0 and $\frac{a_{12}a_{21}}{a_{11}a_{22}}$, which is smaller than the maximum of $\max(|\pm \frac{\sqrt{a_{12}}\sqrt{a_{22}}}{\sqrt{a_{11}}\sqrt{a_{21}}}|)$ due to the relative sizes of a_{11} and a_{22} to a_{12} and a_{21} . Therefore, the Gauss-Seidel iterative method converges faster.

4.

4.a:

$$\rho(B) : B = D^{-1}(L + U)x_n :$$

$$D^{-1} = \begin{bmatrix} 1/2 & 0 & 0 \\ 0 & 1/2 & 0 \\ 0 & 0 & 1/2 \end{bmatrix}$$

$$D^{-1}(L + U) = \begin{bmatrix} 0 & -\frac{1}{2} & \frac{1}{2} \\ 1 & 0 & 1 \\ -\frac{1}{2} & \frac{1}{2} & 0 \end{bmatrix}$$

$$\lambda(B) = [\frac{1}{2}]$$

$$\rho(B) = \frac{1}{2}$$

4.b.

See code.

The code ran for 58 iterations before MATLAB finally rounded to the exact values of 1, 2, and -1 for x. However, it became accurate to 5 decimals (the default display truncation) after 17 iterations.

$$4.c. \rho(B) : B = (D - L)^{-1}(U):$$

$$(D - L) = \begin{bmatrix} 1/2 & 0 & 0 \\ 0 & 1/2 & 0 \\ 0 & 0 & 1/2 \end{bmatrix} \Rightarrow (D - L)^{-1} = \begin{bmatrix} 1/2 & 0 & 0 \\ 1/2 & 1/2 & 0 \\ -1/2 & -1/4 & 1/2 \end{bmatrix}$$

$$(D - L)^{-1}(U) = \begin{bmatrix} 0 & -\frac{1}{2} & \frac{1}{2} \\ 0 & -\frac{1}{2} & \frac{3}{2} \\ 0 & \frac{1}{2} & -1 \end{bmatrix}$$

$$\lambda(B) = [0]$$

$$\rho(B) = 0$$

4.d.

See Code

For my code, it took 27 iterations to get the ∞ -norm $< 1 * 10^{-6}$

5.

5.a

eigenvalues: 0, 1, 3

I have a function in the MATLAB file, *GaussSeidel_relaxation*, which solves for the eigenvalues.

eigenvector for $\lambda = 0$: $(1, 1, 1)^T$

$$Ax = \lambda x \Rightarrow \begin{bmatrix} 1 & -1 & 0 \\ -1 & 2 & -1 \\ 0 & -1 & 1 \end{bmatrix} * x = \text{lambda} * x = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

$$\Rightarrow x = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$$

5.b

see code.

ϵ	iterations
$\epsilon = 1$:	2
$\epsilon = 10^{-1}$:	183
$\epsilon = 10^{-2}$:	1908
$\epsilon = 10^{-3}$:	19164
$\epsilon = 10^{-4}$:	191718
$\epsilon = 0$:	(it is still running)

It is interesting to see how quickly an efficient algorithm can become inefficient. I did not realize how quickly the SOR algorithm becomes inefficient with ϵ of small values, though I suppose from looking at the iterative equation it should not be unexpected, with ϵ being multiplied across the iterative algorithm.