Large Sparse Matrix Computations: Homework 03

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Exercise 1.

 $Proof. \ 1. \ (a-1)$ $M = \begin{bmatrix} 2 & & & \\ & \cdot & & \\ & & \cdot & \\ & & & 2 \end{bmatrix} \implies M^{-1} = \begin{bmatrix} \frac{1}{2} & & & \\ & \cdot & & \\ & & \cdot & \\ & & \frac{1}{2} \end{bmatrix}.$ $N = \begin{bmatrix} 0 & 1 & & \\ 1 & \cdot & \cdot & \\ & & \cdot & \cdot \\ & & & \cdot & 1 \\ & & & 0 \end{bmatrix} \implies M^{-1}N = \begin{bmatrix} 0 & \frac{1}{2} & & \\ \frac{1}{2} & \cdot & \cdot & \\ & & \cdot & \cdot & \frac{1}{2} \\ & & & \frac{1}{2} & 0 \end{bmatrix} \equiv A$

Since $A^k \longrightarrow 0$ when $k \longrightarrow \infty$, then $\rho(A) < 1 \Longrightarrow$ Jacobi iteration of (a) converges.

Since $A^k \longrightarrow 0$ when $k \longrightarrow \infty$, then $\rho(A) < 1 \Longrightarrow GS$ iteration of (a) converges.

$$M = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} \implies M^{-1} = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}.$$

$$N = \begin{bmatrix} 0 & -2 & 2 \\ -1 & 0 & -1 \\ -2 & -2 & 0 \end{bmatrix} \implies M^{-1}N = \begin{bmatrix} 0 & -2 & 2 \\ -1 & 0 & -1 \\ -2 & -2 & 0 \end{bmatrix} \equiv A.$$

Since $\rho(A) = 0 < 1$, then Jacobi iteration of (b) converges.

$$M = \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 2 & 2 & 1 \end{bmatrix} \implies M^{-1} = \begin{bmatrix} 1 & 1 & 1 \\ -1 & 1 & 1 \\ -2 & 1 \end{bmatrix}.$$

$$N = \begin{bmatrix} 0 & -2 & 2 \\ 0 & 0 & -1 \\ 0 & 0 & 0 \end{bmatrix} \implies M^{-1}N = \begin{bmatrix} 0 & -2 & 2 \\ 0 & 2 & -3 \\ 0 & 0 & 2 \end{bmatrix} \equiv A.$$

Since $\rho(A) = 2 > 1$, then GS iteration of (b) does not converge.

 $M = \begin{bmatrix} 2 \\ 2 \\ 2 \end{bmatrix} \implies M^{-1} = \begin{bmatrix} \frac{1}{2} \\ \frac{1}{2} \\ \frac{1}{2} \end{bmatrix}.$

$$N = \begin{bmatrix} 0 & 1 & -1 \\ -2 & 0 & -2 \\ 1 & 1 & 0 \end{bmatrix} \implies M^{-1}N = \begin{bmatrix} 0 & \frac{1}{2} & \frac{-1}{2} \\ -1 & 0 & -1 \\ \frac{1}{2} & \frac{1}{2} & 0 \end{bmatrix} \equiv A.$$

Since $\rho(A) = \frac{\sqrt{5}}{2} > 1$, then Jacobi iteration of (c) does not converge. (c-2)

$$M = \begin{bmatrix} 2 \\ 2 & 2 \\ -1 & -1 & 2 \end{bmatrix} \implies M^{-1} = \begin{bmatrix} \frac{1}{2} \\ \frac{-1}{2} & \frac{1}{2} \\ \frac{1}{4} & \frac{1}{2} \end{bmatrix}.$$

$$N = \begin{bmatrix} 0 & 1 & -1 \\ 0 & 0 & -2 \\ 0 & 0 & 0 \end{bmatrix} \implies M^{-1}N = \begin{bmatrix} 0 & 2 & -2 \\ 0 & 2 & -6 \\ 0 & -1 & 3 \end{bmatrix} \equiv A.$$

Since $\rho(A) = 5 > 1$, then GS iteration of (c) does not converge

Exercise 2.

Proof. (a) \Longrightarrow (b)

By theorem: A is M-matrix \iff There exists v > 0 such that Av > 0.

(A+D)v = Av + Dv > 0. (Since Av > 0 and Dv > 0).

And $A+D \in \mathbb{Z}^{n \times n} \implies A+D$ is M-matrix $\implies A+D$ is invertible.

Exercise 3.

Proof. (a)

Since $A \ge 0$ irreducible, $(I+A)^{n-1}$ is positive.

$$(I + A^T)^{n-1} = ((I + A)^{n-1})^T$$

is also positive.

By Perron Lemma there is an y > 0 such that

$$y^{T}(I+A)^{n-1} = \rho((I+A)^{n-1})y^{T}$$

Let λ be the eigenvalue satisfying $|\lambda| = \rho(A)$ and $Ax = \lambda x$, $x \neq 0$. Further,

$$\rho^{2}(A) |x| \le \rho(A)A |x| = A\rho(A) |x| \le A^{2} |x|.$$

and in general,

$$\rho^k(A)\left|x\right| \leq A^k\left|x\right|, \text{ for } k=1,2,\ldots$$

Hence

$$(1 + \rho(A))^{n-1} |x| \le (I + A)^{n-1} |x|.$$

Multiplying y^T from left it implies

$$(1 + \rho(A))^{n-1}(y^T | x|) \le y^T (I + A)^{n-1} | x| = \rho((I + A)^{n-1})y^T | x|.$$

Since $y^T |x| > 0$, it implies

$$(1 + \rho(A))^{n-1} \le \rho((I+A)^{n-1}).$$

The eigenvalue of $(I+A)^{n-1}$ are of the form $(1+\alpha)^{n-1}$, where α is an eigenvalue of A. Hence there is an eighenvalue μ of A such that

$$|(1+\mu)^{n-1}| = \rho((I+A)^{n-1}).$$

On the other hand, we have $|\mu| \leq \rho(A)$. We have

$$(1 + \rho(A))^{n-1} \le |(1 + \mu)^{n-1}|$$

and further

$$1 + \rho(A) \le |1 + \mu| \le 1 + |\mu| \le 1 + \rho(A).$$

Thus $\mu \geq 0$ and hence $\mu = \rho(A)$.

For k = 1, it follows

$$A|x| = \rho(A)|x| \text{ or } A|x| = \mu|x|.$$

and further

$$(I+A)^{n-1}|x| = |(1+\mu)^{n-1}||x| = \rho((I+A)^{n-1})|x|.$$

Using Perron's Lemma, we get |x| > 0.

 \implies There is only one linearly independent eigenvector belonging to eigenvalue μ .

Moreover $\rho(A) > 0$ as A is distinct from the null matrix.

We want to claim: $\rho(A)$ is a simple eigenvalue of A if and only if:

(i) there is a unique linearly independent eigenvector of A to λ , say μ and also only one linearly independent eigenvector of A^T belonging to λ , say v.

 $(ii)v^Tu \neq 0.$

Only one linearly independent eigenvector of A, say u, belongs to $\rho(A)$. Moreover u > 0.

Similarly, $A^T \ge 0$ irreducible. The respective eigenvector v of A^T (to $\rho(A)$) can be chosen positive as well v > 0.

Therefore $v^T u > 0$ and $\rho(A)$ is simple.

Suppose $Az = \xi z, z \ge 0$ and $\xi \ne \rho(A)$. We have shown that A^T has a positive eigenvector, say w > 0. Then

$$A^T w = \rho(A)w.$$

But

$$w^T A z = w^T \xi z = \xi(w^T z),$$

i.e.,

$$w^T A z = \rho(A)(w^T z),$$

which is a contradiction in view of $\rho(A) - \xi \neq 0$ and $w^T z > 0$.

(b)

Suppose x > 0, such that $Ax = 0 = 0 \cdot x$.

So x is a positive eigenvector.

By (a)
$$\implies Ax = \rho(A) \cdot x$$
.

Since $\rho(A) > 0$, then $\rho(A) \cdot x > 0 = Ax$. It is a contradiction.

$$\implies x = 0.$$

Exercise 4.

Exercise 5.

Exercise 6.

Proof.