

**Exercise 1.** 30pts + 5pts

- a) Let  $f(x) = x^2 - x - 1$  continuous function. Take  $a = 1$  and  $b = 2$ . We have  $f(a)f(b) = -1 < 0$ . Thus, we can apply the bisection method to find a zero  $x_*$  of  $f$  on  $[a, b]$ . In order to have an accuracy  $10^{-5}$ , a sufficient condition is to have the number of iteration  $n$  satisfying:

2pts3pts

$$\frac{1}{2^{n+1}}|b - a| \leq 10^{-5} \quad \Rightarrow \quad -(n+1) \cdot \log 2 \leq -5 \cdot \log 10 \quad \Rightarrow \quad n \geq 5 \frac{\log 10}{\log 2} \approx 16.6.$$

4+2pts

Thus,  $n = 17$  iterations are sufficient.

- b) We use the Newton method starting at  $x_0 = 1$ . The first iteration gives:

$$x_1 = x_0 - \frac{f(x_0)}{f'(x_0)} = 1 - \frac{-1}{1} = 2.$$

5+3pts

- c) We show that  $g(x) = \sqrt{1+x}$  is a contraction on  $I = [0, \infty)$ .

– “ $g(I) \subset I$ ”.

For  $x \geq 0$ ,  $g(x) = \sqrt{1+x} \geq 0$ . Thus  $g(x) \in I$ .

4pts

– “ $|g(x) - g(y)| \leq k|x - y|$ ”.

We estimate the maximum slope of  $g$ :

$$k = \max_{x \in I} |g'(x)| = \max_{x \geq 0} \frac{1}{2\sqrt{1+x}} = \frac{1}{2} < 1.$$

4pts

We deduce by the mean-value theorem that  $g$  is a contraction on  $I$ . Thus, by the contracting mapping theorem, the sequence  $x_{n+1} = g(x_n)$  with  $x_0 \in I$  converges to the unique fixed point  $x_*$  of  $g$  on  $I$ :

$$g(x_*) = x_* \quad \Rightarrow \quad \sqrt{1+x_*} = x_* \quad \Rightarrow \quad 1 + x_* = x_*^2.$$

3pts

Therefore  $f(x_*) = 0$ . Since  $x_* \geq 1$ ,  $x_* = \varphi$ .

*Extra:* Assume  $g$  contraction on  $I$ . First,  $g(g(I)) \subset g(I) \subset I$ . Moreover, for any  $x, y$  in  $I$ :

2pts

$$|g(g(x)) - g(g(y))| \leq k|g(x) - g(y)| \leq k^2|x - y|$$

3pts

with  $k^2 < 1$  (since  $0 \leq k < 1$ ). Therefore  $x \mapsto g(g(x))$  is also a contraction on  $I$ .

**Remark.** Notice that the reverse is false:  $g(x) = e^{-x}$  is not a contraction on  $I = \mathbb{R}$ , whereas  $g(g(x)) = e^{-e^{-x}}$  is.

**Exercise 2.** 25pts + 5pts

First, we perform Gauss elimination on  $A$ :

$$\left[ \begin{array}{ccc|ccc} 2 & 2 & 2 & 1 & 0 & 0 \\ 2 & 4 & 4 & 0 & 1 & 0 \\ 2 & 4 & 6 & 0 & 0 & 1 \end{array} \right] \Rightarrow \left[ \begin{array}{ccc|ccc} 2 & 2 & 2 & 1 & 0 & 0 \\ 0 & 2 & 2 & -1 & 1 & 0 \\ 0 & 2 & 4 & -1 & 0 & 1 \end{array} \right] \Rightarrow \left[ \begin{array}{ccc|ccc} 2 & 2 & 2 & 1 & 0 & 0 \\ 0 & 2 & 2 & -1 & 1 & 0 \\ 0 & 0 & 2 & -1 & -1 & 1 \end{array} \right] \quad \boxed{10\text{pts}}$$

Thus,  $MA = U$  with:

$$U = \begin{bmatrix} 2 & 2 & 2 \\ 0 & 2 & 2 \\ 0 & 0 & 2 \end{bmatrix} \quad \text{and} \quad M = \begin{bmatrix} 1 & 0 & 0 \\ -1 & 1 & 0 \\ -1 & -1 & 1 \end{bmatrix}$$

We then compute the inverse of  $M$ :

$$\left[ \begin{array}{ccc|ccc} 1 & 0 & 0 & 1 & 0 & 0 \\ -1 & 1 & 0 & 0 & 1 & 0 \\ -1 & -1 & 1 & 0 & 0 & 1 \end{array} \right] \Rightarrow \left[ \begin{array}{ccc|ccc} 1 & 0 & 0 & 1 & 0 & 0 \\ 0 & 1 & 0 & 1 & 1 & 0 \\ 0 & -1 & 1 & 1 & 0 & 1 \end{array} \right] \Rightarrow \left[ \begin{array}{ccc|ccc} 1 & 0 & 0 & 1 & 0 & 0 \\ 0 & 1 & 0 & 1 & 1 & 0 \\ 0 & 0 & 1 & 1 & 1 & 1 \end{array} \right] \quad \boxed{8\text{pts}}$$

Thus,  $A = LU$  with  $L = \begin{bmatrix} 1 & 0 & 0 \\ 1 & 1 & 0 \\ 1 & 1 & 1 \end{bmatrix}$ .

To solve  $A\mathbf{x} = \mathbf{b}$  with  $\mathbf{b} = (1, 0, 0)$ , we first solve:  $L\mathbf{y} = \mathbf{b}$  which gives:

$$y_1 = 1, \quad y_2 = -1, \quad y_3 = 0.$$

Then,  $U\mathbf{x} = \mathbf{y}$  which gives:

$$x_3 = 0, \quad x_2 = -\frac{1}{2}, \quad x_1 = 1.$$

Thus,  $\mathbf{x} = (1, -3\frac{1}{2}, 0)^T$ .

*Extra:* to find the Choleski decomposition of  $A$ , we only need to change  $L$  and  $U$  by the factor  $\sqrt{2}$  since  $U = 2 \cdot L^T$ :

$$A = LU = L \cdot 2 \cdot L^T = (\sqrt{2}L) \cdot (\sqrt{2}L)^T.$$

Thus,  $A = CC^T$  with:

$$C = \sqrt{2}L = \begin{bmatrix} \sqrt{2} & 0 & 0 \\ \sqrt{2} & \sqrt{2} & 0 \\ \sqrt{2} & \sqrt{2} & \sqrt{2} \end{bmatrix}. \quad \boxed{5\text{pts}}$$

**Exercise 3.** 20pts

- Taking a vector  $\mathbf{x}$  of unit length (i.e.  $\|\mathbf{x}\| = 1$ ), we have:

$$\|AB\mathbf{x}\| = \|A(B\mathbf{x})\| \leq \|A\| \cdot \|B\mathbf{x}\| \quad \text{3pts}$$

using the inequality  $\|A\mathbf{z}\| \leq \|A\| \cdot \|\mathbf{z}\|$  for any vector  $\mathbf{z}$  (notice that  $B\mathbf{x}$  is not necessarily of unit length). Applying this inequality once more leads to:

$$\|AB\mathbf{x}\| \leq \|A\| \cdot \|B\mathbf{x}\| \leq \|A\| \cdot \|B\| \cdot \|\mathbf{x}\| = \|A\| \cdot \|B\|, \quad \text{3pts}$$

since  $\|\mathbf{x}\| = 1$ . We conclude:

$$\|AB\| = \max_{\|\mathbf{x}\|=1} \|AB\mathbf{x}\| \leq \max_{\|\mathbf{x}\|=1} \|A\| \cdot \|B\| = \|A\| \cdot \|B\|.$$

- Taking  $A = \begin{bmatrix} 1 & 0 \\ a & 1 \end{bmatrix}$  and  $B = \begin{bmatrix} 1 & b \\ 0 & 1 \end{bmatrix}$ , we find  $\rho(A) = \rho(B) = 1$ . However, we also have:

$$AB = \begin{bmatrix} 1 & b \\ a & ab + 1 \end{bmatrix}.$$

with the characteristic polynomial:  $P(\lambda) = \lambda^2 - (ab + 2)\lambda + 1$ . Taking for example  $ab = 1$ , we obtain the eigenvalues:  $\lambda_1 = \frac{3 - \sqrt{3^2 - 4}}{2}$ ,  $\lambda_2 = \frac{3 + \sqrt{3^2 - 4}}{2} > 1$ . Thus, here,  $\rho(AB) > \rho(A) \cdot \rho(B)$ .

6pts3pts5pts**Exercise 4.** 25pts

- The iteration matrix for the Jacobi method is given by:

$$J_j = D^{-1}(L + U) = \begin{bmatrix} 1 & 0 \\ 0 & 5 \end{bmatrix}^{-1} \begin{bmatrix} 0 & 2 \\ 2 & 0 \end{bmatrix} = \begin{bmatrix} 0 & 2 \\ 2/5 & 0 \end{bmatrix}. \quad \text{3+3pts}$$

The characteristic polynomial of  $J_j$  is given by:  $P_{J_j}(\lambda) = \lambda^2 - 4/5$ . Thus, the eigenvalues are  $\lambda = \pm\sqrt{4/5}$ . Since  $\rho(J_j) = \sqrt{4/5} < 1$ , the Jacobi method converges.

4pts

- The iteration matrix for the Gauss-Seidel method is given by:

$$J_g = (D - L)^{-1}U = \begin{bmatrix} 1 & 0 \\ -2 & 5 \end{bmatrix}^{-1} \begin{bmatrix} 0 & 2 \\ 0 & 0 \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 2/5 & 1/5 \end{bmatrix} \begin{bmatrix} 0 & 2 \\ 0 & 0 \end{bmatrix} = \begin{bmatrix} 0 & 2 \\ 0 & 4/5 \end{bmatrix}. \quad \text{3+4pts}$$

The eigenvalues are  $\lambda = 0, \frac{4}{5}$ , thus  $\rho(J_g) = \frac{4}{5} < 1$ , the Gauss-Seidel method converges.

4pts

Moreover, since  $\rho(J_g) < \rho(J_j)$ , we expect the Gauss-Seidel method to converge faster than the Jacobi method.

4pts