

EXERCISE 5(*) Let A be a n -by- n symmetric positive definite matrix, and define the function $\|\mathbf{x}\|_A : \mathbb{R}^n \rightarrow \mathbb{R}$ by $\|\mathbf{x}\|_A = \sqrt{\mathbf{x}^T A \mathbf{x}}$.

- (a) Let $A^{1/2}$ denote the unique symmetric positive definite square root of A , which satisfies $(A^{1/2})^2 = A$ (you may assume without proof that $A^{1/2}$ exists). Show that $\|\mathbf{x}\|_A = \|A^{1/2}\mathbf{x}\|_2$, and use this fact to check that the function $\|\mathbf{x}\|_A$ defines a norm on \mathbb{R}^n . Determine constants $0 < c \leq C$ such that

$$c\|\mathbf{x}\|_2 \leq \|\mathbf{x}\|_A \leq C\|\mathbf{x}\|_2, \quad \forall \mathbf{x} \in \mathbb{R}^n.$$

- (b) Now consider the stationary Richardson method for the solution of the linear system $A\mathbf{x} = \mathbf{b}$, with iteration matrix $B_\alpha = I - \alpha A$. Show that $A^{1/2}B_\alpha = B_\alpha A^{1/2}$. Use this to prove that

$$\|\mathbf{e}^{k+1}\|_A \leq \rho(B_\alpha)\|\mathbf{e}^k\|_A,$$

where $\mathbf{e}^k = \mathbf{x} - \mathbf{x}^k$ denotes the solution error after the k th iteration.

Sol :

a), Firstly, we should show that $\|\mathbf{x}\|_A = \|A^{\frac{1}{2}}\mathbf{x}\|_2$.

$$\text{Since } \|A^{\frac{1}{2}}\mathbf{x}\|_2 = \sqrt{(A^{\frac{1}{2}}\mathbf{x}) \cdot (A^{\frac{1}{2}}\mathbf{x})^T},$$

$$\begin{aligned} \text{and } \|\mathbf{x}\|_A &= \sqrt{\mathbf{x}^T A \mathbf{x}} = \sqrt{\mathbf{x}^T [(A^{\frac{1}{2}})^T A^{\frac{1}{2}}] \mathbf{x}} \\ &= \sqrt{(A^{\frac{1}{2}} \cdot \mathbf{x})^T \cdot (A^{\frac{1}{2}} \cdot \mathbf{x})} = \|A^{\frac{1}{2}}\mathbf{x}\|_2 \end{aligned}$$

Next, we should check the function $\|\mathbf{x}\|_A$ defines a norm on \mathbb{R}^n .

If $\|\mathbf{x}\|_A : \mathbb{R}^n \rightarrow \mathbb{R}$ is a norm on \mathbb{R}^n , it must satisfy the following conditions:

- 1, (i) $\|\mathbf{x}\|_A \geq 0, \forall \mathbf{x} \in \mathbb{R}^n$, and (ii) $\|\mathbf{x}\|_A = 0$ if and only if $\mathbf{x} = 0$;

Check: since $A^{\frac{1}{2}}$ is symmetric positive matrix, so if $\mathbf{x} \in \mathbb{R}^n$ and $\mathbf{x} \neq 0$, we must have $A^{\frac{1}{2}}\mathbf{x} \neq 0$, and $(A^{\frac{1}{2}}\mathbf{x})^T \cdot (A^{\frac{1}{2}}\mathbf{x}) > 0$; And only $\mathbf{x} = 0, A^{\frac{1}{2}}\mathbf{x} = 0$.

- 2, $\|\alpha \mathbf{x}\|_A = |\alpha| \|\mathbf{x}\|_A \quad \forall \alpha \in \mathbb{R} \text{ and } \forall \mathbf{x} \in \mathbb{R}^n$;

$$\text{Check: } \|\alpha \mathbf{x}\|_A = \|\alpha \cdot A^{\frac{1}{2}}\mathbf{x}\|_2 = |\alpha| \|A^{\frac{1}{2}}\mathbf{x}\|_2 = |\alpha| \|\mathbf{x}\|_A$$

- 3, $\|\mathbf{x} + \mathbf{w}\|_A \leq \|\mathbf{x}\|_A + \|\mathbf{w}\|_A, \forall \mathbf{x}, \mathbf{w} \in \mathbb{R}^n$;

$$\text{Check: } \|\mathbf{x} + \mathbf{w}\|_A = \|A^{\frac{1}{2}}(\mathbf{x} + \mathbf{w})\|_2 \leq \|A^{\frac{1}{2}}\mathbf{x}\|_2 + \|A^{\frac{1}{2}}\mathbf{w}\|_2 = \|\mathbf{x}\|_A + \|\mathbf{w}\|_A$$

Therefore, the function $\|x\|_A$ defines a norm on \mathbb{R}^n .

Finally, the constants $0 < c \leq C$ such that

$$c\|x\|_2 \leq \|x\|_A \leq C\|x\|_2, \quad \forall x \in \mathbb{R}^n$$

Since the function $\|x\|_A$ defines a norm, and $\|x\|_A = \|A^{\frac{1}{2}}x\|_2$, we can get that:

$$\|cx\|_2 \leq \|A^{\frac{1}{2}}x\|_2 \leq \|Cx\|_2$$

Because A is a n -by- n symmetric positive definite matrix, so $A^{\frac{1}{2}} > 0$.

Therefore, $0 < c = C = A^{\frac{1}{2}}$.

b), Firstly, we should to prove that $A^{\frac{1}{2}}B_\alpha = B_\alpha A^{\frac{1}{2}}$,

Since $B_\alpha = I - \alpha A$, so we can get:

$$\begin{aligned} A^{\frac{1}{2}}B_\alpha &= B_\alpha A^{\frac{1}{2}} \Rightarrow A^{\frac{1}{2}}(I - \alpha A) = (I - \alpha A) \cdot A^{\frac{1}{2}} \\ &\Rightarrow A^{\frac{1}{2}} - A^{\frac{1}{2}}\alpha A = A^{\frac{1}{2}} - \alpha A A^{\frac{1}{2}} \end{aligned}$$

Because relaxation parameter $\alpha \in \mathbb{R}$, so,

$$\begin{aligned} &\Rightarrow A^{\frac{1}{2}} - \alpha A^{\frac{1}{2}}A = A^{\frac{1}{2}} - \alpha A A^{\frac{1}{2}} \\ &\Rightarrow A^{\frac{1}{2}} - \alpha I = A^{\frac{1}{2}} - \alpha I \end{aligned}$$

Therefore, $A^{\frac{1}{2}}B_\alpha = B_\alpha A^{\frac{1}{2}}$.

Next, we need to prove that $\|e^{k+1}\|_A \leq \rho(B_\alpha)\|e^k\|_A$.

According to the question, we can get that:

$$\|e^{k+1}\|_2 = \|B_2 \cdot e^k\|_2 \leq \|B_2\|_2 \cdot \|e^k\|_2$$

Then, we multiply $\|A^{\frac{1}{2}}\|_2$ to both sides.

$$\|A^{\frac{1}{2}}\|_2 \|e^{k+1}\|_2 \leq \|A^{\frac{1}{2}}\|_2 \cdot \|B_2\|_2 \cdot \|e^k\|_2$$

Because $A^{\frac{1}{2}} B_2 = B_2 A^{\frac{1}{2}}$, Next, we can get:

$$\|A^{\frac{1}{2}} e^{k+1}\|_2 \leq \|B_2\|_2 \cdot \|A^{\frac{1}{2}} \cdot e^k\|_2$$

From Lemma 4.4.3, we can know that $\|B_2\|_2 = \rho(B_2)$.

$$\text{So, } \|e^{k+1}\|_A \leq \rho(B_2) \|e^k\|_A,$$