

Homework 6

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- **Collaborators:** I finish this homework by myself. However, I ask matlab for help when I solve the Problem 5.
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Problem 1.

$$\begin{aligned}\hat{r} - r &= Ax - A\hat{x} \\ &= A(A^T A)^{-1} A^T b - A(A^T A + F)^{-1} A^T b \\ &= A(A^T A + F)^{-1} ((A^T A + F) - (A^T A)) (A^T A)^{-1} A^T b \\ &= A(A^T A + F)^{-1} F x\end{aligned}$$

We know from the lecture that

$$\mathcal{K}_2(A)^2 = \text{cond}_2(A^T A) = \frac{|\rho(A^T A)|}{|\sigma_n(A)|^2} = \frac{\|A\|_2^2}{|\sigma_n(A)|^2}$$

Noticed that $\|(A^T A)^{-1}\|_2 = \|A^+\|_2^2 = \frac{1}{|\sigma_n(A)|^2} \leq \frac{1}{2\|F\|_2}$, hence

$$\|(A^T A + F)^{-1}\|_2 \leq \frac{\|(A^T A)^{-1}\|_2}{1 - \|F\|_2 \cdot \|(A^T A)^{-1}\|_2} \leq \frac{\|A^+\|_2^2}{\frac{1}{2}}$$

Therefore

$$\begin{aligned}\|\hat{r} - r\|_2 &\leq \|A\|_2 \cdot \|(A^T A + F)^{-1}\|_2 \cdot \|F\|_2 \cdot \|x\|_2 \\ &\leq 2\|A\|_2 \cdot \|A^+\|_2^2 \cdot \|F\|_2 \cdot \|x\|_2 \\ &= 2\mathcal{K}_2(A)^2 \frac{\|F\|_2}{\|A\|_2} \|x\|_2\end{aligned}$$

Problem 2.

$$\begin{aligned}
 \|x - \hat{x}\|_2 &= \|(A^T A)^{-1} f\|_2 \\
 &\leq \|(A^T A)^{-1}\| \cdot \|f\|_2 \\
 &\leq \|A^+\|_2^2 \cdot \epsilon \|A\|_2 \|b\|_2 \\
 &= \epsilon \mathcal{K}_2(A)^2 \frac{\|A\|_2 \|b\|_2}{\|A\|_2^2} \\
 &= \epsilon \mathcal{K}_2(A)^2 \frac{\|A\|_2 \|b\|_2}{\|A^T A\|_2}
 \end{aligned}$$

Then

$$\frac{\|x - \hat{x}\|_2}{\|x\|_2} \leq \epsilon \mathcal{K}_2(A)^2 \frac{\|A\|_2 \|b\|_2}{\|A^T A\|_2 \|x\|_2} \leq \epsilon \mathcal{K}_2(A)^2 \frac{\|A\|_2 \|b\|_2}{\|A^T A x\|_2} = \epsilon \mathcal{K}_2(A)^2 \frac{\|A\|_2 \|b\|_2}{\|A^T b\|_2}$$

Problem 3. We already have method to compute $A^T A$, $A^T b$, and we do Gauss elimination to it to get an equation:

$$Ux = b'$$

where U is upper triangular matrix, with

$$U = \begin{bmatrix} U' & B \\ 0 & 0 \end{bmatrix}, b' = \begin{bmatrix} \hat{b}' \\ 0 \end{bmatrix}$$

U' is the upper triangular matrix with all diagonal elements 1.

Then solutions of $A^T A x = A^T b$ are

$$x = \begin{bmatrix} U^{-1}(\hat{b}' - Bx') \\ x' \end{bmatrix} = \begin{bmatrix} U^{-1}\hat{b}' \\ 0 \end{bmatrix} - \begin{bmatrix} U^{-1}B \\ -I \end{bmatrix} x'$$

Therefore the solution of least norm is exactly the least square solution of

$$\begin{bmatrix} U^{-1}B \\ -I \end{bmatrix} x' = \begin{bmatrix} U^{-1}\hat{b}' \\ 0 \end{bmatrix}$$