

Condition number of matrix inversion [Turing/von-Neumann]

Consider the function

$$f(A) = A^{-1}.$$

We will analyze the condition number of this function. It will turn out to have several applications; most notably the analyses of the **gradient descent** and the **conjugate gradient descent** methods.

We first need to fix what norm we will use to measure errors when the underlying objects are matrices. The right answer turns out to be the **operator norm**, defined for a matrix A as follows:

$$\|A\| = \|A\|_{2 \rightarrow 2} := \sup_{\|x\|_2 \leq 1} \|Ax\|_2$$

Ex: Verify that this is a norm: [convex,

linear wr.t. multiplication by constant, positive semi-definite, and attaining the value 0 only when $A=0$].

The operator norm is also submultiplicative

$$\|AB\| \leq \|A\| \cdot \|B\| \quad (\text{why?}).$$

Now we define the condition no. as.

$$\kappa_A := \text{cond}^f(A) = \lim_{\delta \downarrow 0} \sup_{\substack{E: \|E\| \leq \delta \\ \|A\|}} \frac{\frac{\|(A+E)^{-1} - A^{-1}\|}{\|A^{-1}\|}}{\frac{\|A+E - A\|}{\|A\|}}.$$

Thm:- $\kappa_A = \|A\| \cdot \|A^{-1}\|.$

Pf:- We first show that $\kappa_A \leq \|A\| \cdot \|A^{-1}\|.$

We first note

$$\begin{aligned} (A+E)^{-1} &= ((I + EA^{-1})A)^{-1} = A^{-1}(I + EA^{-1})^{-1} \\ &= A^{-1}(I - EA^{-1}) + D \end{aligned}$$

where $\|D\| = O(\|E\|^2).$

Thus, $\|(A+E)^{-1} - A^{-1}\| = \|-A^{-1}EA^{-1} + D\|$
 $\leq \|A^{-1}\|^2 \|E\| + O(\|E\|^2). \quad \text{--- (1)}$

$$\delta_7 \quad \kappa_A \leq \lim_{\delta \rightarrow 0} \sup_{E: \|E\| \leq \delta} \frac{\|A^{-1}\|^2 \|E\| + O(\|E\|^2)}{\frac{\|A^{-1}\|}{\|A\|}}$$

$$= \lim_{\delta \rightarrow 0} (\|A\| \|A^{-1}\| + O(\delta^2))$$

$$= \|A\| \|A^{-1}\|.$$

$$\text{Thus, } \kappa_A \leq \|A\| \cdot \|A^{-1}\|.$$

We now show that this is an equality.

Claim:- If E is s.t. $\|A^{-1} E A^{-1}\| = \|A^{-1}\|^2 \|E\|$, then ① holds with equality. [Ex.]

Now, let y be s.t.

$$\|A^{-1} y\| = \|A^{-1}\|; \|y\| = 1$$

(Such a y exists, [Ex.])

$$\text{Let } x = \frac{\|A^{-1} y\|}{\|A^{-1}\|}, \text{ and } B = y x^T.$$

Claim:- Setting $E = \delta B$ is sufficient. [Why?]

Condition number of lin. system solving

Consider

$$Ax = b. \quad \underline{\varphi(A, b) = A^{-1}b.}$$

$$A \in \mathbb{R}^{n \times n}, \quad b \in \mathbb{R}^n.$$

For perturbed versions \tilde{A}, \tilde{b} of A, b , we set

$$R := \text{RelError}(A, b) = \max \left\{ \frac{\|A - \tilde{A}\|}{\|A\|}, \frac{\|b - \tilde{b}\|}{\|b\|} \right\}.$$

We want to estimate

$\text{cond}_{\varphi}(A, b):$

$$\text{Set } x = \varphi(A, b) = A^{-1}b, \quad \tilde{x} = \tilde{A}^{-1}\tilde{b}$$

$$\text{RelError}(\varphi(A, b)) = \frac{\|\tilde{x} - x\|}{\|x\|}.$$

$$\text{Let } \tilde{A} = A + E, \quad \tilde{b} = b + e \text{ so that} \\ \|E\| \leq R\|A\|, \quad \|e\| \leq R\|b\|.$$

Now

$$\begin{aligned} \tilde{x} &= (A + E)^{-1}(b + e) \\ &= ((I + EA^{-1})A)^{-1}(b + e) \\ &= A^{-1}(I + EA^{-1})^{-1}(b + e) \end{aligned}$$

Now when $\|EA^{-1}\| < 1$,

$$\begin{aligned} I + EA^{-1} &= I - EA^{-1} + (EA^{-1})^2 - (EA^{-1})^3 + \dots \\ &= I - EA^{-1} + D \\ &\text{where } \|D\| = O(R^2). \end{aligned}$$

$$\begin{aligned} \text{So, } \tilde{x} &= (A^{-1} - A^{-1}EA^{-1})(b + e) + O(R^2) \\ &= x + A^{-1}e - A^{-1}EA^{-1}b + A^{-1}EA^{-1}e + O(R^2) \end{aligned}$$

But $\|A^{-1}EA^{-1}e\| = O(R^2)$, so we have,

$$\|\tilde{x} - x\| \leq \|A^{-1}e - A^{-1}EA^{-1}b\| + O(R^2) \quad \text{--- } \textcircled{x}$$

$$\textcircled{1} \leq \|A^{-1}e\| + \|A^{-1}EA^{-1}b\| + O(R^2)$$

$$\textcircled{2} \leq \|A^{-1}\| \|e\| + \|A^{-1}\| \|E\| \|b\| + O(R^2)$$

$$\textcircled{3} \leq R \|A^{-1}\| \|b\| + R \|A^{-1}\| \|A\| \|x\| + O(R^2)$$

$$\Rightarrow \frac{\|\tilde{x} - x\|}{\|x\|} \leq R \left(\kappa_A + \frac{\|A^{-1}\| \cdot \|b\|}{\|A^{-1}b\|} \right) + O(R^2)$$

Now, by definition of R

$$\text{Cond}_\psi(A, b) = \lim_{R \downarrow 0} \frac{\frac{\|\tilde{x} - x\|}{\|x\|}}{R} \leq \kappa_A + \frac{\|A^{-1}\| \cdot \|b\|}{\|A^{-1}b\|}$$

To prove equality, we need to choose the perturbations E and e carefully so that inequalities

①, ② and ③ above hold with equality.

Let v be a unit vector such that $\|A^{-1}v\| = \|A^{-1}\|$.
 Let $y = \frac{A^{-1}v}{\|A^{-1}\|}$ be the unit vector in the direction of $A^{-1}v$, and $z = \frac{x}{\|x\|}$ the unit vector in the direction of x . Let B be the matrix $-vz^T$. Now note that

$$A^{-1}v = \|A^{-1}\| y$$

$$A^{-1}Bx = -\|A^{-1}\| \|x\| y$$

Let $e = R\|b\|v$, $E = R\|A\|B$.

Then, from \textcircled{x} ,

$$\begin{aligned} \|\tilde{x} - x\| &\geq \|A^{-1}e - A^{-1}Ex\| - O(R^2) \\ &= R\left(\|A^{-1}\|\|b\|\|y\| + \|A\|\|A^{-1}\|\|x\|\|y\|\right) - O(R^2) \\ &= R\left(\|A^{-1}\|\|b\| + \|A\|\|A^{-1}\|\|x\|\right) - O(R^2) \end{aligned}$$

since $\|y\| = 1$.

$$\Rightarrow \frac{\|\tilde{x} - x\|}{\|x\|} \geq R\left(\kappa_A + \frac{\|A^{-1}\|\|b\|}{\|A^{-1}b\|}\right) - O(R^2)$$

so that

$$\text{cond}_\varphi(A, \hat{b}) \geq \kappa_A + \frac{\|A^{-1}\|\|b\|}{\|A^{-1}b\|}.$$

□