Mathematics, Pusan National University

Numerical Linear Algebra

Lecture 12. Conditioning and Condition Numbers

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Condition of a Problem



We can view a **problem** as a function $f: X \to Y$ from a normed vector space X of data to a normed vector space Y of solutions. **well-conditioned** problem : **small** change of x leads to **small** change in f(x). **ill-conditioned** problem : **small** change of x leads to **large** change in f(x)

Absolute Condition Number



Let δx : small change of x, write $\delta f = f(x + \delta x) - f(x)$. Absolute Condition Number $\hat{\kappa} = \hat{\kappa}(x)$ of the problem f at x is defined as

$$\hat{\kappa} = \lim_{\delta \to 0} \sup_{\|\delta x\| \le \delta} \frac{\|\delta f\|}{\|\delta x\|},\tag{1}$$

simply,

$$\hat{\kappa} = \sup_{\delta x} \frac{\|\delta f\|}{\|\delta x\|}.\tag{2}$$

If f is differentiable, let J(x) be a **Jacobian** of f at x. Then the absolute condition number becomes

$$\hat{\kappa} = ||J(x)||. \tag{3}$$

Relative Condition Number



The relative condition number $\kappa = \kappa(x)$ is defined by

$$\kappa = \lim_{\delta \to 0} \sup_{\|\delta x\| \le \delta} \left(\frac{\|\delta f\|}{\|f(x)\|} / \frac{\|\delta x\|}{\|x\|} \right),\tag{4}$$

or, assuminf δx and δf are infinitesimal,

$$\kappa = \sup_{\|\delta x\| \le \delta} \left(\frac{\|\delta f\|}{\|f(x)\|} / \frac{\|\delta x\|}{\|x\|} \right). \tag{5}$$

If f is differentiable, we can express this quantity in terms of the Jacobian:

$$\kappa = \frac{\|J(x)\|}{\|f(x)\|/\|x\|}.$$
 (6)

Examples



Example

Let $f(x) = \frac{x}{2}$ where $x \in \mathbb{C}$. The Jacobian of the f is $J = f' = \frac{1}{2}$, so,

$$\kappa = \frac{\|J\|}{\|f(x)\|/\|x\|} = \frac{1/2}{(x/2)/x} = 1. \tag{7}$$

This problem is well-conditioned.

Examples



Example

Consider the problem $f(x) = x_1 - x_2$ where the vector $x \in \mathbb{C}^2$. For simplicity, we use the ∞ -norm on the data space \mathbb{C}^2 . The Jacobian of f is

$$J = \begin{bmatrix} \frac{\delta f}{\delta x_1} & \frac{\delta f}{\delta x_2} \end{bmatrix} = \begin{bmatrix} 1 & -1 \end{bmatrix}$$
 (8)

with $||J||_{\infty} = 2$. The condition number is

$$\kappa = \frac{\|J\|_{\infty}}{\|f(x)\|/\|x\|} = \frac{2}{|x_1 - x_2|/\max\{|x_1|, |x_2|\}}.$$
 (9)

As $|x_1 - x_2| = 0$, the problem is **ill-conditioned**.

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Condition of Matrix-Vector Multiplication



Fixed $A \in \mathbb{C}^{m \times n}$ and consider the problem of computing Ax from input x. Then,

$$\kappa = \sup_{\delta x} \left(\frac{\|A(x + \delta x) - Ax\|}{\|Ax\|} / \frac{\|\delta x\|}{\|x\|} \right) = \sup_{\delta x} \frac{\|A\delta x\|}{\|\delta x\|} / \frac{\|Ax\|}{\|x\|} = \|A\| \frac{\|x\|}{\|Ax\|}$$
(10)

We can use the fact that $\frac{\|x\|}{\|Ax\|} \le \|A^{-1}\|$ to lossen (10) to a bound independent of x:

$$\kappa \le \|A\| \|A^{-1}\| \tag{11}$$

or

$$\kappa = \alpha ||A|| ||A^{-1}|| \text{ where } \alpha = \frac{||x||}{||Ax||} / ||A^{-1}||$$
(12)

Condition of Matrix-Vector Multiplication



Theorem

Let $A \in \mathbb{C}^{m \times m}$ be nonsingular and consider the equation Ax = b. The problem of computing b, given x, has condition number

$$\kappa = \|A\| \frac{\|x\|}{\|b\|} \le \|A\| \|A^{-1}\| \tag{13}$$

with respect ro perturbation of x. The problem of computing x, given b, has condition number

$$\kappa = \|A^{-1}\| \frac{\|b\|}{\|x\|} \le \|A\| \|A^{-1}\| \tag{14}$$

with respect to perturbation of b. If $\|\cdot\| = \|\cdot\|_2$, then equality holds in (13) if x is a multiple of a right singular vector of A corresponding to the minimal singular value σ_m , and equality holds in (14) if b is a multiple of a left singular vector of A corresponding to the maximal singular value σ_1 .

Condition Number of a Matrix



Condition number of $A \in \mathbb{C}^{m \times m}$, denoted by $\kappa(A)$:

$$\kappa(A) = ||A|| ||A^{-1}|| = \frac{\sigma_1}{\sigma_m}$$
 (15)

For a rectangular matrix $A \in \mathbb{C}^{m \times n}$ of full rank, $m \ge n$, the condition number is defined in terms of the pseudoinverse: $\kappa(A) = \|A\| \|A^+\|$. So,

$$\kappa(A) = \frac{\sigma_1}{\sigma_n} \tag{16}$$

Condition of a System of Equations



In theorem, we held A fixed and perturbed x or b. What happens if we perturb A? Specifically, ler us hold b fixed and consider the vehavior of the problem $A \mapsto x = A^{-1}b$ when A is perturbed by δA . Then x must change bt δx , where

$$(A + \delta A)(x + \delta x) = b \Rightarrow Ax + A(\delta x) + (\delta A)x + (\delta A)(\delta x) = b$$

$$\Rightarrow A(\delta x) + (\delta A)x + (\delta A)(\delta x) = 0$$

$$\Rightarrow A(\delta x) + (\delta A)x = 0$$

$$\Rightarrow \delta x = -A^{-1}(\delta A)x$$

$$\Rightarrow \|\delta x\| \le \|A^{-1}\| \|\delta A\| \|x\|$$

or equivalently,

$$\frac{\|\delta x\|}{\|x\|} / \frac{\|\delta A\|}{\|A\|} \le \|A^{-1}\| \|A\| = \kappa(A).$$

Condition of a System of Equations



Theorem

Let b be fixed and consider the problem of computing $x = A^{-1}b$, where A is square and nonsingular. The condition number of this problem with respect ro perturbations in A is

$$\kappa = ||A|| ||A^{-1}|| = \kappa(A). \tag{17}$$

Two thorems in this lecture are of fundamental importance in numerical algebra, for ther determine how accurately one can solve systems of equations. If a problem Ax = b contains an **ill-conditioned** matrix A, one must always expect to "loss $\log_{10} \kappa(A)$ digits" in computing the solution, except under very special circumstances.

Exercise



Exercise 12.1

Suppose A is a 202×202 matrix with $||A||_2 = 100$ and $||A||_F = 101$. Give the sharpest possible lower bound on the 2-norm condition number $\kappa(A)$.

Solution

Let $\sigma_1 \ge \cdots \ge \sigma_{202}$ be the singular values of A. We know that $\sigma_1 = ||A||_2 = 100$. If $\sigma_{202} > 1$, then $\sigma_i > 1$ for all i, and so

$$||A||_F = \sqrt{\sum_{i=1}^{202} \sigma_i^2} = \sqrt{\sigma_1^2 + \sum_{i=2}^{202} \sigma_i^2} > \sqrt{100^2 + 201} = \sqrt{10201} = 101.$$

Since $||A||_F = 101$, it cannot be the case that $\sigma_{202} > 1$: $\sigma_{202} \le 1$.

Exercise



Exercise 12.1

Suppose A is a 202×202 matrix with $||A||_2 = 100$ and $||A||_F = 101$. Give the sharpest possible lower bound on the 2-norm condition number $\kappa(A)$.

Solution

Furthermore, this upper bound is attainable. If $\sigma_2 = \cdots = \sigma_{202} = 1$, then

$$||A||_F = \sqrt{\sum_{i=1}^{202} \sigma_i^2} = \sqrt{100^2 + 201} = 101.$$

This leads to the sharpest possible lower bound on $\kappa(A)$, which is given by

$$\kappa(A) = \frac{\sigma_1}{\sigma_{202}} \ge \frac{100}{1} = 100.$$

