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Note Title

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Definition

For an $n \times n$ matrix, we define the matrix norm $\|A\|$ subordinate to a given vector norm to be

$$\|A\| = \sup \{ \|Ax\| : x \in \mathbb{R}^n \text{ and } \|x\| = 1 \}$$

Sup - Supremum

$$= \sup_{\|x\|=1} \|Ax\|$$

Least upper bound

The Supremum is taken over all n -dimensional

Vectors with unit norm.

Lemma

A subordinate matrix norm satisfies the following properties, for any matrices A & B .

$$\textcircled{i} \quad \|A\| > 0 \quad \text{if } A \neq 0$$

$$\textcircled{ii} \quad \|\alpha A\| = |\alpha| \|A\| \quad \text{for any scalar } \alpha.$$

$$\textcircled{iii} \quad \|A + B\| \leq \|A\| + \|B\| \quad \text{triangular inequality}$$

We also have the additional properties

$$(iv) \|I\| = 1$$

$$I = \begin{bmatrix} 1 & & 0 \\ & \ddots & \\ 0 & & 1 \end{bmatrix}$$

$$(v) \|Ax\| \leq \|A\| \|x\|$$

$$(vi) \|AB\| \leq \|A\| \|B\|$$

The matrix norms subordinate to the vector

norms $\|\cdot\|_1$, $\|\cdot\|_2$ & $\|\cdot\|_\infty$ are

$$\|A\|_1 = \max_{1 \leq j \leq n} \sum_{i=1}^n |a_{ij}| \quad \text{column sum norm}$$

Sum over the columns.

$$\|A\|_{\infty} = \max_{1 \leq i \leq n} \sum_{j=1}^n |a_{ij}| \quad \text{L}_{\infty} \text{-matrix norm}$$

$$\|A\|_2 = \max_{1 \leq i \leq n} \sqrt{|\sigma_i|} \quad \text{spectral / L}_2 \text{-matrix norm}$$

the σ_i are the eigenvalues of $A^T A$ (called the singular values of A).

The largest σ_{\max} in absolute value is known as the spectral radius of A .

Example

Let

$$A = \begin{bmatrix} 1 & 2 & -1 \\ 0 & 3 & -1 \\ 5 & -1 & 1 \end{bmatrix}$$

To compute ∞ -matrix norm we sum over the rows.

$$\text{row 1: } \sum_{j=1}^3 |a_{1j}| = |1| + |2| + |-1| = 4$$

$$\text{row 2: } \sum_{j=1}^3 |a_{2j}| = |0| + |3| + |-1| = 4$$

$$\text{row 3: } \sum_{j=1}^3 |a_{3j}| = |5| + |-1| + |1| = 7$$

$$\|A\|_{\infty} = \max \{4, 4, 7\} = 7$$

To compute λ_1 -maxima norm sum over columns

$$\sum_{i=1}^3 |a_{i1}| = |1| + |6| + |5| = 6$$

$$\sum_{i=1}^3 |a_{i2}| = |2| + |3| + |-1| = 6$$

$$\sum_{i=1}^3 |a_{i3}| = |-1| + |-1| + |1| = 3$$

$$\|A\|_1 = \max \{6, 6, 3\} = 6$$

Condition Number and $\|A\|$ -Conditioning

Definition

We define the condition number of an $n \times n$ non singular matrix A for the norm $\|\cdot\|_p$ to be

$$\kappa_p(A) = \|A\|_p \|A^{-1}\|_p$$

The condition number of a matrix A gives a measure of how sensitive systems of equations, with coefficient matrix A , are to small perturbations.

If a system $Ax = b$ is sensitive to perturbations then the matrix A is said to be ill-conditioned and the condition number is large.

Suppose we want to solve $Ax = b$ where A is non singular.

Suppose the right side is perturbed by an amount δb and the solution is perturbed by an amount δx

Then

$$\begin{aligned} A(x + \delta x) &= Ax + A\delta x \\ &= \underline{\underline{b + A\delta x}} \end{aligned}$$

$$= b + \underline{\underline{Sb}} \quad \text{where } ASx = Sb$$

Recall that for $Ax = b$ we have

$$\|Ax\| = \|b\|$$

So

$$\|b\| = \|Ax\| \leq \|A\| \|x\|$$

i.e.

$$\|b\| \leq \|A\| \|x\|$$

$$\Rightarrow \frac{1}{\|x\|} \leq \frac{\|A\|}{\|b\|} \quad (*)$$

For the perturbed linear problem

$$A \delta x = \delta b$$

$$A \delta x = \delta b$$

$$\underbrace{A^{-1}A}_{I} \delta x = A^{-1} \delta b$$

$$\delta x = A^{-1} \delta b$$

We have

$$\delta x = A^{-1} \delta b$$

because A is non-singular
 $\therefore A$ is invertible

So that

$$\|\delta x\| = \|A^{-1} \delta b\|$$

$$\leq \|A^{-1}\| \|\delta b\|$$

i.e.

$$\|\delta x\| \leq \|A^{-1}\| \|\delta b\| \quad (**)$$

Combining (*) and (**)

$$\frac{1}{\|x\|} \leq \frac{\|A\|}{\|b\|}$$

$$\|x\| \leq \|A^{-1}\| \|b\|$$

We get $\frac{\|s x\|}{\|x\|} \leq \|A\| \|A^{-1}\| \frac{\|s b\|}{\|b\|}$

i.e.

$$\frac{\|s x\|}{\|x\|} \leq \kappa(A) \frac{\|s b\|}{\|b\|}$$

where $\kappa(A)$ is the condition number of A