

Lecture #14

02/10/2023

Important result from linear algebra:

Any induced matrix norm is sub-multiplicative.

$$\|AB\|_p \leq \|A\|_p \|B\|_p$$

for all $1 \leq p \leq \infty$

For example, when $p=2$, then

$$\|AB\|_2 \leq \|A\|_2 \|B\|_2 \text{ etc.}$$

Let us specialize the above result for nonsingular A , and $B = A^{-1}$

Then,

$$\underbrace{\| \underbrace{A A^{-1}}_I \|_p}_{\|I\|_p = 1} \leq \underbrace{\|A\|_p \|A^{-1}\|_p}_{\substack{K_p(A) \\ (\text{p-norm condition number})}}$$

$$\Leftrightarrow 1 \leq K_p(A) \text{ for all } 1 \leq p \leq \infty.$$

$\therefore \text{In general: } 1 \leq K_p(A) \leq \infty$

The lower bound is achieved e.g., by $A = I$
(1) or other orthogonal matrices.

The upper bound (∞) is achieved by singular matrices.

For any nonsingular/invertible matrix A :

$$1 \leq \kappa_p(A) < \infty.$$

If $\kappa_*(A) \approx 1$, then we say the matrix A is "well-conditioned".

Then, small changes in \underline{b} vector will cause small changes in the solution \underline{x} for $A\underline{x} = \underline{b}$

If $\kappa(A) \gg 1$, then we say the matrix A is "ill-conditioned"

Example:

$$A = \begin{pmatrix} 5 & 7 & 6 & 5 \\ 7 & 10 & 8 & 7 \\ 6 & 8 & 10 & 9 \\ 5 & 7 & 9 & 10 \end{pmatrix}.$$

$\det(A) = 1 \neq 0$ (so there is unique solution for $A\underline{x} = \underline{b}$ for any $\underline{b} \neq \underline{0}$)

If we want to solve for \underline{x} in the square linear system: $A\underline{x} = \underline{b}$ where

$$\underline{b} = \begin{pmatrix} 2 \\ 3 \\ 3 \\ 3 \end{pmatrix}.$$

→ You can do: $A \backslash b$
in MATLAB

Check that $\underline{x} = \begin{pmatrix} 1 \\ 1 \\ 1 \\ 1 \end{pmatrix}.$

what if we try to compute \underline{x} with

$$\underline{b} = \begin{pmatrix} 22.9 \\ 32.1 \\ 32.9 \\ 31.1 \end{pmatrix}$$

Check that $\underline{x} = \begin{pmatrix} -7.2 \\ 6 \\ 2.9 \\ -0.1 \end{pmatrix}$.

Try another:

$$\underline{b} = \begin{pmatrix} 22.99 \\ 32.01 \\ 32.99 \\ 31.01 \end{pmatrix} \Rightarrow \underline{x} = \begin{pmatrix} 0.18 \\ 1.5 \\ 1.19 \\ 0.89 \end{pmatrix}.$$

Check for example:

$$\begin{aligned}k_{\infty}(A) &= \|A\|_{\infty} \|A^{-1}\|_{\infty} \\&= 33 \times 136 \\&= 4488 \gg 1.\end{aligned}$$

i.e., matrix A is ill-conditioned.

Related MATLAB commands:

$\gg \text{norm}(x, p) \leftarrow$ for computing p -norm of a vector x , for $0 < p \leq \infty$

$\gg \text{norm}(X, p) \leftarrow$ for computing induced p -norm of any rectangular matrix X , for $p = 1, 2, \text{Inf}, \text{"fro"}$,

$$\gg \text{cond}(A, p) \leftarrow \kappa_p(A) := \|A\|_p \|A^{-1}\|_p$$

Interpolation problem

Fit a curve passing through ALL given datapoints

x	y
x_1	y_1
x_2	y_2
\vdots	
x_n	y_n

Find $y(x)$

satisfying $y_i = y(x_i)$

for all $i = 1, 2, \dots, n$

A common choice is :

$$y = p(x)$$

$$= c_1 + c_2 x + c_3 x^2 + c_4 x^3 + \dots + c_n x^{n-1}$$

i.e., $y = p(x)$ is a univariate polynomial in x
of degree $(n-1)$.

where $n = \#$ of datapoints/samples

$$\therefore y_i = p(x_i) \text{ for all } i=1, \dots, n$$

\therefore we get n equations in $\underbrace{n \text{ unknowns}}$

$$\underbrace{c_1, c_2, \dots, c_n}_{\begin{pmatrix} c_1 \\ \vdots \\ c_n \end{pmatrix}}$$

In fact, we get n linear equations in n unknowns: c_1, c_2, \dots, c_n

$$\begin{pmatrix} 1 & x_1 & x_1^2 & \dots & x_1^{n-1} \\ 1 & x_2 & x_2^2 & \dots & x_2^{n-1} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & x_{n-1} & x_{n-1}^2 & \dots & x_{n-1}^{n-1} \\ 1 & x_n & x_n^2 & \dots & x_n^{n-1} \end{pmatrix} \begin{pmatrix} c_1 \\ \vdots \\ c_{n-1} \\ c_n \end{pmatrix} = \begin{pmatrix} y_1 \\ \vdots \\ y_{n-1} \\ y_n \end{pmatrix}$$

$\underbrace{\hspace{15em}}_{n \times n \text{ Known matrix}} \quad \underbrace{\hspace{10em}}_{n \times 1 \text{ Unknown vector}} = \underbrace{\hspace{10em}}_{n \times 1 \text{ Known vector}}$

$\Leftrightarrow \quad \underbrace{\text{Known matrix}}_{\text{Known}} \times \underline{\underline{c}} = \underline{\underline{y}} \quad \underbrace{\text{Unknown}}_{\text{Unknown}} \quad \left. \vphantom{\underline{\underline{c}}} \right\} \text{Square linear system in unknown } \underline{\underline{c}}$

This matrix

$$X := \begin{pmatrix} 1 & x_1 & x_1^2 & \dots & x_1^{n-1} \\ 1 & x_2 & x_2^2 & \dots & x_2^{n-1} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & x_{n-1} & x_{n-1}^2 & \dots & x_{n-1}^{n-1} \\ 1 & x_n & x_n^2 & \dots & x_n^{n-1} \end{pmatrix}$$

is called
Vandermonde
matrix.

Theorem:

det of Vandermonde
matrix

$$\det(\underset{\substack{\uparrow \\ \text{Vandermonde} \\ \text{matrix}}}{X}) = \prod_{1 \leq i < j \leq n} (x_j - x_i)$$

← named after
the French
mathematician
Vandermonde

Consequence:

Unique solution vector $\underline{\leq}$

if and only if $\underbrace{\det(X) \neq 0}_{\Leftrightarrow}$

$x_i \neq x_j$ for all $j \neq i$