

Exam 1: content thru end lec 8

HW4 (short to allow exam prep)

Goals:

- Frobenius

- ~~Solve~~ Solve

$$\|A\|_2 = \sigma_{\max}$$

SVD? expensive \uparrow

$$\kappa(A) = \|A\| \|A^{-1}\| < 1?$$

$$\geq \|AA^{-1}\| = \|I\| = 1$$

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

Computing the 2-Norm

Using the SVD of A , identify the 2-norm.

$$\sigma_{\max}$$

Express the matrix condition number $\text{cond}_2(A)$ in terms of the SVD:

$$\sigma_1 / \sigma_n$$

Not a matrix norm: Frobenius

The 2-norm is very costly to compute. Can we make something simpler?

$$\|A\|_F := \sqrt{\sum_{i,j} |A_{i,j}|^2}$$

What about its properties?

- definiteness
- $\| \alpha A \|_F = |\alpha| \|A\|_F$
- triangle inequality
- $\|AB\|_F \leq \|A\|_F \|B\|_F$ (Cauchy-Schwarz)

Frobenius Norm: Properties

$$\|A\|_x = \max_{\|x\|_x=1} \|Ax\|_x$$

Is the Frobenius norm induced by any vector norm?

$$\|I\|_F = n \quad \text{for } I \in \mathbb{R}^{n \times n} \quad \rightarrow \text{no, for induced n:}$$

How does it relate to the SVD?

$$\|I\| = 1$$

$$A = U \Sigma V^T$$
$$\|A\|_F = \|\Sigma\|_F = \sqrt{\sum_{i=1}^n \sigma_i^2}$$
$$\|QA\|_2 = \|A\|_2 \quad \|QA\|_F = \|A\|_F$$

la. norm $(A) \rightarrow$ Fro
 $(A, \text{"Fro"})$

Solving Systems: Simple cases

$$3n^5 - 15n^2 + \dots$$

Solve $D\mathbf{x} = \mathbf{b}$ if D is diagonal. (Computational cost?) $\rightarrow = O(n^5)$

$$x_i = b_i / d_{ii} \quad O(n) \quad \text{as } n \rightarrow \infty$$

Solve $Q\mathbf{x} = \mathbf{b}$ if Q is orthogonal. (Computational cost?) $Q \in \mathbb{R}^{n \times n}$

$$Q\vec{x} = \vec{b} \mid Q^T \cdot \quad \cancel{Q^T Q \vec{x} = Q^T \vec{b}} \quad O(n^2)$$

Given SVD $A = U\Sigma V^T$, solve $A\mathbf{x} = \mathbf{b}$. (Computational cost?)

$$\begin{aligned} U \Sigma V^T \vec{x} &= \vec{b} \mid U^T \cdot & V^T \vec{x} &= \vec{y} \\ \Leftrightarrow \Sigma V^T \vec{x} &= U^T \vec{b} & \Leftrightarrow \vec{x} &= V \vec{y} \\ \Leftrightarrow \Sigma \vec{y} &= U^T \vec{b} & & O(n^2) \end{aligned}$$

$$(A \vec{x})_i = \sum_j A_{ij} x_j \quad i \in \{1, \dots, n\}$$

n multiplications

$n-1$ additions

n outputs

$\hookrightarrow n(n+n-1)$

$$n(2n-1) = O(n^2)$$

Solving Systems: Triangular matrices

Solve

$$\begin{bmatrix} a_{11} & a_{12} & a_{13} & a_{14} \\ a_{22} & a_{23} & a_{24} & \\ & a_{33} & a_{34} & \\ & & a_{44} & \end{bmatrix} \begin{bmatrix} x \\ y \\ z \\ w \end{bmatrix} = \begin{bmatrix} b_1 \\ b_2 \\ b_3 \\ b_4 \end{bmatrix}$$

Handwritten annotations: Blue arrows point to the diagonal elements $a_{11}, a_{22}, a_{33}, a_{44}$. Blue circles highlight a_{22} and a_{34} . A blue arrow points upwards from the right side of the equation.

$$\begin{aligned} a_{33} \cdot z + a_{34} \cdot w &= b_3 \\ a_{44} \cdot w &= b_4 \end{aligned}$$

Handwritten annotations: A blue circle around a_{34} and a green checkmark next to w in the second equation.

Demo: Coding back-substitution [cleared]

What about non-triangular matrices?

Gaussian elimination

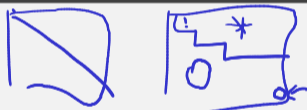
Gaussian Elimination

Demo: Vanilla Gaussian Elimination [cleared]

What do we get by doing Gaussian Elimination?

REF

How is that different from being upper triangular?



What if we do not just eliminate downward but also upward?

Gauss-Jordan \rightarrow won't consider

if inv: GE gives ∇ : can use BS

LU Factorization

What is the LU factorization?

from GE

$$A = LU$$

E? \rightarrow later

Solving $Ax = b$

Does LU help solve $Ax = b$?

$$\begin{array}{l} LU \vec{x} = b \\ \quad \downarrow \\ \text{solve } L\vec{y} = b \quad \text{for } \vec{y} \quad (\text{using } \neq S) \\ \text{solve } U\vec{x} = \vec{y} \quad \text{for } \vec{x} \quad (\text{using } BS) \end{array}$$

Determining an LU factorization

$$\begin{bmatrix} \boxed{A} & \boxed{B} \\ \hline & \end{bmatrix} \leftarrow AB$$

$$A = \left(\begin{array}{c|c} \begin{matrix} a_{11} & a_{12} \\ \hline a_{21} & A_{22} \end{matrix} & \begin{matrix} \xrightarrow{u} \\ \xrightarrow{u} \end{matrix} \\ \hline \end{array} \right) = \left(\begin{array}{c|c} \begin{matrix} l_{11} & 0 \\ \hline l_{21} & L_{22} \end{matrix} & \begin{matrix} \xrightarrow{u} \\ \xrightarrow{u} \end{matrix} \\ \hline \end{array} \right) \left(\begin{array}{c|c} \begin{matrix} u_{11} & u_{12} \\ \hline 0 & U_{22} \end{matrix} & \begin{matrix} \xrightarrow{u} \\ \xrightarrow{u} \end{matrix} \\ \hline \end{array} \right)$$

$$\left(\begin{array}{c|c} \begin{matrix} 1 & 0 \\ \hline l_{21} & L_{22} \end{matrix} & \begin{matrix} \xrightarrow{u} \\ \xrightarrow{u} \end{matrix} \\ \hline \end{array} \right) \left(\begin{array}{c|c} \begin{matrix} u_{11} & u_{12} \\ \hline 0 & U_{22} \end{matrix} & \begin{matrix} \xrightarrow{u} \\ \xrightarrow{u} \end{matrix} \\ \hline \end{array} \right) \left(\begin{array}{c|c} \begin{matrix} a_{11} & a_{12} \\ \hline a_{21} & A_{22} \end{matrix} & \begin{matrix} \xrightarrow{u} \\ \xrightarrow{u} \end{matrix} \\ \hline \end{array} \right)$$

$a_{11} = l_{11} \cdot u_{11}$ ← impose $R_{ii} = 1$
 $\hookrightarrow u_{11} = a_{11}$

$a_{21} = l_{21} \cdot u_{11}$
 $\hookrightarrow l_{21} = a_{21} / u_{11}$

$a_{12} = u_{11} \cdot u_{12} + l_{21} \cdot u_{12}$
 $\hookrightarrow u_{12} = a_{12} / u_{11}$

$A_{22} = l_{21} \cdot u_{12} + L_{22} U_{22}$
 ↑ ↑ ↑
 order prod!

Demo: LU Factorization [cleared]

$$L_{22}U_{22} = A_{22} - \vec{L}_{21}\overset{\leftarrow}{U}_{12} \quad \checkmark \text{ RHS}$$

↑

'recurse' to LU fact of
size $(n-1) \times (n-1)$

Computational Cost



What is the computational cost of multiplying two $n \times n$ matrices?

$$O(n^3)$$

▶ $u_{11} = a_{11}, \mathbf{u}_{12}^T = \mathbf{a}_{12}^T.$

▶ $l_{21} = \mathbf{a}_{21}/u_{11}.$

▶ $L_{22}U_{22} = A_{22} - l_{21}\mathbf{u}_{12}^T.$

$\left. \begin{array}{l} \leftarrow O(n) \\ \leftarrow O(n) \\ \leftarrow O(n^2) \end{array} \right\} n \text{ times}$

What is the computational cost of carrying out LU factorization on an $n \times n$ matrix?

$$O(n^3)$$

Demo: Complexity of Mat-Mat multiplication and LU [cleared]

LU: Failure Cases?

Is LU/Gaussian Elimination bulletproof?

