

Numerical Linear Algebra

Basically, solving $\underline{A} \underline{x} = \underline{b}$ on the computer.

↑ ↘ ↑
Matrix Vector

Please review:

- 1) Row reduction
- 2) Inner products
- 3) Gram-Schmidt
- 4) Eigen values

Why? It's how things are actually calculated in real life.

So far in this class:

- Fixed pt : $x_{n+1} = f(x_n)$
- Polynomial Interpolation : $y(x) = \sum L_i(x) y_i$
- Numerical Diff. / Int. : $f'(x_0) / \int_0^1 f(x) dx$

\Rightarrow All functions of one variable.

In real life, functions have many variables.

eg. Newton's method

$$\underline{\underline{1D}} \quad x_{n+1} = x_n - \frac{f(x_n)}{f'(x_n)}$$

nD

$$\underline{x}_{n+1} = \underline{x}_n - J(x_n)^{-1} f(\underline{x}_n)$$

↑
vector

↑
Matrix inverse
- Jacobian

eg. $f \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} \sin(x_1 + x_2) \\ \cos(x_2) \end{pmatrix}$

$$J = \begin{pmatrix} \frac{\partial f_1}{\partial x_1} & \frac{\partial f_1}{\partial x_2} \\ \frac{\partial f_2}{\partial x_1} & \frac{\partial f_2}{\partial x_2} \end{pmatrix} = \begin{pmatrix} \cos(x_1 + x_2) & \cos(x_1 + x_2) \\ 0 & -\sin(x_2) \end{pmatrix}$$

- In practice $n \sim 10^3 - 10^6$,
you can't compute this analytically
or by hand
-

What do we care about?

- Cost

Focus

- Stability

Won't really
cover in
this class

Basic operations:


1) Dot product

$$\underline{x} \cdot \underline{y} = \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix} \cdot \begin{pmatrix} y_1 \\ \vdots \\ y_n \end{pmatrix}$$

n multiplications, $n-1$ additions
- $O(n)$ cost

2) Matrix-vector multiplication

$$\underline{\underline{A}} \underline{x} = \begin{pmatrix} a_{11} & a_{12} & \dots \\ a_{21} & a_{22} & \dots \\ \vdots & \vdots & \ddots \end{pmatrix} \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix} = \begin{pmatrix} y_1 \\ \vdots \\ y_m \end{pmatrix}$$


 $m \times n$ $n \times 1$ $m \times 1$

\Rightarrow m dot products of size n

$$\Rightarrow O(mn)$$

3) Matrix-matrix multiplication

$$\underline{\underline{A}} \underline{\underline{B}} = \begin{pmatrix} a_{11} & \dots & \dots \\ \vdots & \ddots & \vdots \\ \vdots & \vdots & \ddots \end{pmatrix} \begin{pmatrix} b_{11} & \dots & \dots \\ \vdots & \ddots & \vdots \\ \vdots & \vdots & \ddots \end{pmatrix}$$

$m \times n$ $n \times p$

$O(mnp)$. Can you see why?

4) Solution. $\underline{A}\underline{x}=\underline{b}$, given \underline{A} , \underline{b} ,
want to find \underline{x} .

Idea: 1) Markov inverse **X NO**

2) Gauss-Elimination **✓ YES**

- Row reduction

ex.
$$\begin{pmatrix} 1 & 2 \\ 3 & 4 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 2 \\ 3 \end{pmatrix}$$

① \downarrow row 2 $-= 3 \times$ row 1

$$\begin{pmatrix} 1 & 0 \\ -3 & 1 \end{pmatrix} \begin{pmatrix} 1 & 2 \\ 3 & 4 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 1 & 0 \\ -3 & 1 \end{pmatrix} \begin{pmatrix} 2 \\ 3 \end{pmatrix}$$



$$\begin{pmatrix} 1 & 2 \\ 0 & -2 \end{pmatrix}$$

$$= \begin{pmatrix} 2 \\ -3 \end{pmatrix}$$

$$\Rightarrow \begin{pmatrix} 1 & 2 \\ 0 & -2 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 2 \\ -3 \end{pmatrix}$$



②

Now can solve

- back-substitution

$$x_1 + 2x_2 = 2$$

$$-2x_2 = -3$$

\Rightarrow

$$x_1 = -1$$

$$x_2 = \frac{3}{2}$$

Overall:

Step 1: $\begin{pmatrix} 1 & 0 \\ -3 & 1 \end{pmatrix} A = \begin{pmatrix} 1 & 2 \\ 0 & -2 \end{pmatrix}$

$\underbrace{\hspace{10em}}$

lower triangular upper triangular

$\Rightarrow \underline{A} = \underline{L} \underline{U}$ factorize

- Called LU factorization

Step 2: Back substitution

$$\underline{U} \underline{x} = \underline{L} \underline{b}$$

→ solve x_n , then solve x_{n-1}, \dots, x_1
in order

What is the cost?

① When you row reduce :

eg. $\begin{pmatrix} 1 & \dots & \dots \\ 2 & \dots & \dots \\ 3 & \dots & \dots \\ 4 & \dots & \dots \end{pmatrix}$

you do:

row 2 - 2 × row 1

row 3 - 3 × row 1

row 4 - 4 × row 1

so $n-1$ rows where you subtract
a row

$\Rightarrow O(n^2)$ for this part,

→ BUT, have to do this for every row,

$\Rightarrow O(n^3)$ cost overall

(2) Backsubstitution:

$$a_{11}x_1 + a_{12}x_2 + \dots + a_{1n}x_n = b_1$$

$$a_{22}x_2 + \dots + a_{2n}x_n = b_2$$

⋮

$$a_{n-1,n-1}x_{n-1} + a_{n-1,n}x_n = b_{n-1}$$

$$a_{nn}x_n = b_n$$

To solve: $a_{nn}x_n = b_n$

$$\rightarrow x_n = b_n / a_{nn},$$

| divide.

$$a_{n-1,n-1}x_{n-1} + a_{n-1,n}x_n = b_{n-1}$$

$$\Rightarrow x_{n-1} = \frac{b_{n-1} - a_{n-1,n}x_n}{a_{n-1,n-1}}$$

$a_{n-1,n-1}$

| multiply, | division, | subtraction

⋮

$$a_{kk}x_k + \dots + a_{kn}x_n = b_k$$

$$\Rightarrow x_k = \frac{b_k - a_{k,k+1}x_{k+1} - \dots - a_{kn}x_n}{a_{kk}}$$

$n-k$ multiply, $n-k$ adds,

1 division

$$\Rightarrow O(n-k)$$

operations

\Rightarrow Overall cost for back-substitution

$$O(n^2)$$