

Lecture 10: Recall:

Fast Fourier Transform (FFT) (Cooley and Tukey, 1965)

Let $F_n = \begin{pmatrix} 1 & 1 & \dots & 1 \\ 1 & \omega_n & \dots & \omega_n^{n-1} \\ \vdots & \vdots & & \vdots \\ 1 & \omega_n^{n-1} & \dots & \omega_n^{(n-1)^2} \end{pmatrix}$ where $\omega_n = e^{i(\frac{2\pi}{n})}$

Let $\vec{y} = F_n \vec{x}$, where $\vec{y} = \begin{pmatrix} y_0 \\ y_1 \\ \vdots \\ y_{n-1} \end{pmatrix}$ and $\vec{x} = \begin{pmatrix} x_0 \\ x_1 \\ \vdots \\ x_{n-1} \end{pmatrix}$. Suppose $n=2m$.

Then, for each $0 \leq j \leq n-1$,

$$y_j = \sum_{k=0}^{n-1} \omega_n^{jk} x_k = \sum_{k=0}^{2m-1} \omega_{2m}^{kj} x_k.$$

Divide $k = 0, 1, 2, \dots, 2m-1$ into two parts:

Part 1: $0, 2, 4, 6, \dots, 2(m-1)$ (Even)

Part 2: $1, 3, 5, 7, \dots, 2m-1$ (Odd)

Then:
$$y_j = \underbrace{\sum_{k=0}^{m-1} \omega_n^{2kj} x_{2k}}_{\text{Part 1}} + \underbrace{\sum_{k=0}^{m-1} \omega_n^{(2k+1)j} x_{2k+1}}_{\text{Part 2}}$$

Denote $\vec{x}' = \begin{pmatrix} x_0 \\ x_2 \\ \vdots \\ x_{2m-2} \end{pmatrix}$, $\vec{x}'' = \begin{pmatrix} x_1 \\ x_3 \\ \vdots \\ x_{2m-1} \end{pmatrix}$. Let $\vec{y}' = F_m \vec{x}'$ and $\vec{y}'' = F_m \vec{x}''$.

For $j = 0, 1, 2, \dots, m-1$, we have:

$$y_j = \overbrace{(F_m \vec{x}')_j}^{m \times m} + \omega_n^j \overbrace{(F_m \vec{x}'')_j}^{m \times m} = \underbrace{(\vec{y}')_j}_{\substack{\uparrow \\ j\text{-th entry of } \vec{y}'}} + \omega_n^j \underbrace{(\vec{y}'')_j}_{\substack{\downarrow \\ j\text{-th entry of } \vec{y}''}}$$

For $j = 0, 1, 2, \dots, m-1$, we have

$$y_{j+m} = \underbrace{(F_m \vec{x}')_j}_{m \times m} - \omega_n^j \underbrace{(F_m \vec{x}'')_j}_{m \times m}$$

Note: $n \times n$ matrix multiplication becomes $\frac{n}{2} \times \frac{n}{2} = m \times m$ matrix multiplication.

For simplicity, we denote: $\begin{pmatrix} v_1 \\ v_2 \\ \vdots \\ v_n \end{pmatrix} \otimes \begin{pmatrix} w_1 \\ w_2 \\ \vdots \\ w_n \end{pmatrix} = \begin{pmatrix} v_1 w_1 \\ v_2 w_2 \\ \vdots \\ v_n w_n \end{pmatrix}$

Summary of FFT

Step 1: Split $\vec{x}$ into $\vec{x}' = \begin{pmatrix} x_0 \\ x_2 \\ \vdots \\ x_{2(m-1)} \end{pmatrix}$ and $\vec{x}'' = \begin{pmatrix} x_1 \\ x_3 \\ \vdots \\ x_{2m-1} \end{pmatrix}$

Step 2: Compute $\vec{y}' = F_m \vec{x}'$ and $\vec{y}'' = F_m \vec{x}''$, where $F_m = \frac{n}{2} \times \frac{n}{2}$ matrix $\text{m} \times \text{m}$
 $\mathcal{O}(m^2)$ $\mathcal{O}(m^2)$

Step 3: Compute: $\vec{\omega}_n$

$$\begin{pmatrix} y_0 \\ y_1 \\ \vdots \\ y_{m-1} \end{pmatrix} = \vec{y}' + \begin{pmatrix} \omega_n^0 \\ \omega_n^1 \\ \vdots \\ \omega_n^{m-1} \end{pmatrix} \otimes \vec{y}'' \quad \text{and} \quad \begin{pmatrix} y_m \\ y_{m+1} \\ \vdots \\ y_{2m-1} \end{pmatrix} = \vec{y}' - \begin{pmatrix} \omega_n^0 \\ \omega_n^1 \\ \vdots \\ \omega_n^{m-1} \end{pmatrix} \otimes \vec{y}''$$

$\mathcal{O}(m)$ $\mathcal{O}(m)$

Remark: Computational cost : $O(m^2) + O(m)$ (multiplication + addition)
ss
 $O(m^2)$

Computational cost for FFT: Assume $n = 2^L$.

Let C_m = computational cost of F_m . Then $C_1 = 1$.

Claim: $C_{2m} = 2C_m + 3m$

Proof: Step 2: $\vec{y}' = F_m \vec{x}'$, $\vec{y}'' = F_m \vec{x}''$ ($= 2C_m$)

Step 3: $y_j = \vec{y}_j' + \omega_n^j \vec{y}_j''$
 $y_{j+m} = \vec{y}_j' - \omega_n^j \vec{y}_j''$
for $j = 0, 1, 2, \dots, m-1$

(= 1 multiplication
+
1 addition
+
1 subtraction)

↓
Total : $3m$

$\therefore C_{2m} = 2C_m + 3m$.

Now, $n = 2^l$.

$$\therefore C_{2^l} = 2C_{2^{l-1}} + 3 \cdot 2^{l-1}.$$

$$\begin{aligned}\therefore 2^{-l}C_{2^l} &= 2^{-(l-1)}C_{2^{l-1}} + \frac{3}{2} = 2^{-(l-2)}C_{2^{l-2}} + 2\left(\frac{3}{2}\right) \\ &= \vdots\end{aligned}$$

$$\begin{aligned}\therefore C_{2^l} &= \underbrace{2^l}_n + \frac{3}{2} l \underbrace{2^l}_n \quad = 2^0 C_{2^0} + l\left(\frac{3}{2}\right) = 1 + \frac{3}{2}l \\ &= n + \frac{3}{2}n \log_2 n = \mathcal{O}(n \log_2 n)\end{aligned}$$

Butterfly diagram (Algorithmic visualization)

Consider F_4 (4×4 matrix).

[Recall: $\vec{y}_e = \vec{y}' = F_m \vec{x}'$ Denote $\vec{x}' := \vec{x}_e$, $\vec{x}'' = \vec{x}_o$]
 $\vec{y}_o = \vec{y}'' = F_m \vec{x}''$

$$\begin{array}{lcl} \vec{y}_e := F_2 \vec{x}_e = \begin{pmatrix} x_0 \\ x_2 \end{pmatrix} = \begin{pmatrix} \vdots \end{pmatrix} & \xrightarrow{\quad} & \begin{pmatrix} y_0 \\ y_1 \end{pmatrix} \\ \vec{y}_o := F_2 \vec{x}_o = \begin{pmatrix} x_1 \\ x_3 \end{pmatrix} = \begin{pmatrix} \vdots \end{pmatrix} & \xrightarrow[\text{---}\vec{\omega}_4]{\vec{\omega}_4} & \begin{pmatrix} y_2 \\ y_3 \end{pmatrix} \end{array}$$

$\vec{\omega}_4 = \begin{pmatrix} \omega_4^0 \\ \omega_4^1 \end{pmatrix}$

Diagram means:

$$\left. \begin{array}{l} \begin{pmatrix} y_0 \\ y_1 \end{pmatrix} = \vec{y}_e + \vec{\omega}_4 \otimes \vec{y}_o \\ \begin{pmatrix} y_2 \\ y_3 \end{pmatrix} = \vec{y}_o - \vec{\omega}_4 \otimes \vec{y}_o \end{array} \right\} \text{ Depend on } \vec{y}_o \text{ and } \vec{y}_e.$$

For $F_2 \vec{x}_e$: $\overset{1}{\leftarrow} F_1 \vec{x}_{ee} = (x_0)$
 $\overset{1}{\leftarrow} F_1 \vec{x}_{eo} = (x_2)$

$$\begin{pmatrix} x_0 \\ x_2 \end{pmatrix} \xrightarrow{\begin{pmatrix} \vec{W}_2^0 \\ -\vec{W}_2^0 \end{pmatrix}} \vec{y}_e = F_2 \vec{x}_e = \vec{y}_e$$

For $F_2 \vec{x}_o$: $\overset{1}{\leftarrow} F_1 \vec{x}_{oe} = (x_1)$
 $\overset{1}{\leftarrow} F_1 \vec{x}_{oo} = (x_3)$

$$\begin{pmatrix} x_1 \\ x_3 \end{pmatrix} \xrightarrow{\begin{pmatrix} \vec{W}_2^0 \\ -\vec{W}_2^0 \end{pmatrix}} \vec{y}_o = F_2 \vec{x}_o = \vec{y}_o$$

Remark:

$$\vec{x}_e = \begin{pmatrix} x_0 \\ x_2 \end{pmatrix}, \quad \vec{x}_o = \begin{pmatrix} x_1 \\ x_3 \end{pmatrix}$$

$\downarrow$
 $\vec{x}_{ee} = (x_0)$

$\downarrow$
 $\vec{x}_{oe} = (x_1)$

$\vec{x}_{eo} = (x_2)$

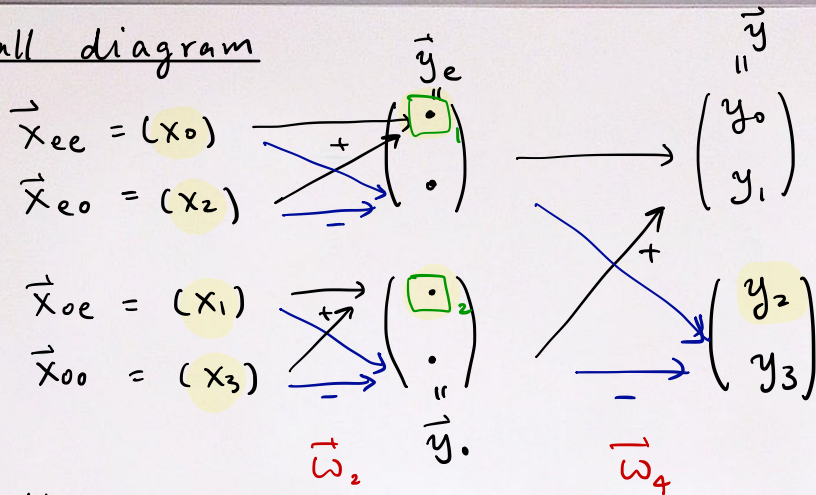
$\vec{x}_{oo} = (x_3)$

$$\vec{W}_2^0 = (W_2^0) = (1)$$

$$\vec{y}_e = \begin{pmatrix} x_0 + \vec{W}_2^0 \otimes (x_2) \\ x_0 - \vec{W}_2^0 \otimes (x_2) \end{pmatrix} = \begin{pmatrix} x_0 + x_2 \\ x_0 - x_2 \end{pmatrix}$$

$$\vec{y}_o = \begin{pmatrix} x_1 + \vec{W}_2^0 \otimes (x_3) \\ x_1 - \vec{W}_2^0 \otimes (x_3) \end{pmatrix} = \begin{pmatrix} x_1 + x_3 \\ x_1 - x_3 \end{pmatrix}$$

Overall diagram



(Butterfly diagram)

Using the diagram, find y_2 .

$$y_2 = \square_1 - (\vec{\omega}_4)_0 \cdot \square_2 \quad ; \quad \square_1 = x_0 + x_2$$

$$= \square_1 - \square_2 \quad \square_2 = x_1 + x_3$$

$$\therefore y_2 = x_0 + x_2 - (x_1 + x_3)$$

Iterative method to solve huge linear system

Recall: Numerical spectral method handles periodic functions.

Consider: (*) $\frac{d^2 u}{dx^2} = f$, $u(0) = A$, $u(1) = B$
 $x \in [0, 1]$.

Partition $[0, 1]$ into $x_j = jh$ where $h = \frac{1}{n+1}$

Then: (*) is discretized as:

$$\frac{u_{i+1} - 2u_i + u_{i-1}}{h^2} = f(x_i) \quad \text{or}$$

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

$$\tilde{A} \begin{pmatrix} u_1 \\ u_2 \\ \vdots \\ u_n \end{pmatrix} = \begin{pmatrix} f(x_1) - \frac{A}{h^2} \\ f(x_2) \\ \vdots \\ f(x_n) - \frac{B}{h^2} \end{pmatrix}$$

$u_i \stackrel{\text{def}}{=} u(x_i)$

$D = \begin{pmatrix} -2 & 1 & 0 & \dots & 1 \\ 1 & -2 & 1 & \dots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & \dots & 1 & -2 \end{pmatrix}$ has eigenvector e^{itx}

$$\tilde{A} = \frac{1}{h^2} \begin{pmatrix} -2 & 1 & \dots & 0 \\ 1 & -2 & \dots & 1 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & \dots & 1 & -2 \end{pmatrix}$$

Question: How to solve BIG linear system?

Method 1: Gaussian elimination

Comp. cost : $\mathcal{O}(n^3)$

Sol : exact.

Method 2: LU factorization.

Decompose $A = LU$
 $\uparrow \quad \nwarrow$
 lower upper

(If A is SPD, then $A = LL^T$
by Cholesky decomposition)

$L \begin{pmatrix} \vec{y} \\ \vec{x} \end{pmatrix} = \vec{b}$ by solving $\begin{cases} L\vec{y} = \vec{b} \\ U\vec{x} = \vec{y} \end{cases}$ — easy

Comp. cost : $\mathcal{O}(n^3)$

Sol : exact.

Goal: Develop iterative method: find a sequence $\vec{x}_0, \vec{x}_1, \vec{x}_2, \dots$
such that $\vec{x}_k \rightarrow \vec{x}^* = \text{sol. of } A\vec{x} = \vec{f} \text{ as } k \rightarrow \infty$.

Remark: We can stop when error is small enough.

Method: Splitting method

Consider a linear system $A\vec{x} = \vec{f}$ where $A \in M_{n \times n}$ (n is BIG)

Split A as follows: $A = N + (A - N) = N - \underbrace{(N - A)}_P$

Then: $A\vec{x} = \vec{f} \Leftrightarrow (N - P)\vec{x} = \vec{f} \Leftrightarrow N\vec{x} = P\vec{x} + \vec{f}$

Develop an iterative scheme as follows:

$$(\star) \quad N\vec{x}^{n+1} = P\vec{x}^n + \vec{f}$$

If $\{\vec{x}^n\}_{n=1}^{\infty}$ converges, then it converges to the sol $\vec{x}^*$ of $A\vec{x} = \vec{f}$

- Remark:
- N should be simple = easy to find inverse.
 - N should have an inverse
 - N should be "related to" A .
 - N should be chosen such that $\{\vec{x}^n\}_{n=1}^{\infty}$ converges.

Splitting choice 1: Jacobi method

Take $N = D$ = diagonal part of A

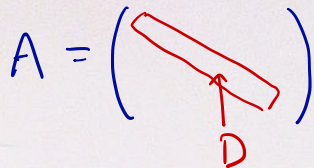
Split A as $A = D - (D - A)$

$$\begin{aligned}\text{Then: } A\vec{x} &= \vec{f} \Leftrightarrow D\vec{x} - (D - A)\vec{x} = \vec{f} \\ &\Leftrightarrow D\vec{x} = (D - A)\vec{x} + \vec{f}\end{aligned}$$

We can consider an iterative scheme such that:

$$\begin{aligned}D\vec{x}^{k+1} &= (D - A)\vec{x}^k + \vec{f} \\ \Leftrightarrow \vec{x}^{k+1} &= D^{-1}(D - A)\vec{x}^k + D^{-1}\vec{f}\end{aligned}$$

Remark: All diagonal entries of A must be non-zero, such that D is non-singular.

$$A = \begin{pmatrix} & & \\ & \color{red}{\boxed{}} & \\ & & \color{red}{\boxed{}} \end{pmatrix}$$
A diagram of a 3x3 matrix A. The diagonal elements are enclosed in red boxes, and a red arrow points from the label 'D' below to the middle diagonal box, indicating that D is the diagonal part of A.

This is equivalent to solving: (assume $A = (a_{ij})_{1 \leq i, j \leq n}$) $\vec{x}^k = \begin{pmatrix} x_1^k \\ \vdots \\ x_n^k \end{pmatrix}$

$$\begin{cases} a_{11}x_1^{k+1} + a_{12}x_2^k + a_{13}x_3^k + \dots + a_{1n}x_n^k = f_1 & (\text{for } x_1^{k+1}) \\ a_{21}x_1^k + a_{22}x_2^{k+1} + a_{23}x_3^k + \dots + a_{2n}x_n^k = f_2 & (\text{for } x_2^{k+1}) \\ \vdots \\ a_{n1}x_1^k + a_{n2}x_2^k + a_{n3}x_3^k + \dots + a_{nn}x_n^{k+1} = f_n & (\text{for } x_n^{k+1}) \end{cases}$$

Example: Consider: $\begin{pmatrix} 5 & -2 & 3 \\ -3 & 9 & 1 \\ 2 & -1 & -7 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} -1 \\ 2 \\ 3 \end{pmatrix}$

Then: Jacobi method gives:

$$\vec{x}^{k+1} = \begin{pmatrix} 5 & 0 & 0 \\ 0 & 9 & 0 \\ 0 & 0 & -7 \end{pmatrix}^{-1} \begin{pmatrix} 0 & 2 & -3 \\ 3 & 0 & -1 \\ -2 & 1 & 0 \end{pmatrix} \vec{x}^k + \begin{pmatrix} 5 & 0 & 0 \\ 0 & 9 & 0 \\ 0 & 0 & -7 \end{pmatrix}^{-1} \begin{pmatrix} -1 \\ 2 \\ 3 \end{pmatrix}$$

Start with $\vec{x}^0 = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}$. The sequence almost converge in 7

iteration to get: $\vec{x}_7 = \begin{pmatrix} -0.186 \\ 0.331 \\ -0.423 \end{pmatrix}$

Question:

- Does it always converge?
- Is the initialization important?