

明天 7:30 - 9:30 pm 期中考试.

线上 9:35 pm 提交.

DFT 高能 Fourier 变换

FFT Fast Fourier Transform.

Cooley - Tukey 1965 Gauss 1805
IBM Princeton.

多项式乘法.

$$f(x) = a_0 + a_1 x + \dots + a_d x^d \quad \deg \leq d$$

$$g(x) = b_0 + b_1 x + \dots + b_d x^d \quad \deg \leq d.$$

$$\begin{aligned} f \cdot g &= a_0 b_0 + (a_0 b_1 + b_0 a_1) x \\ &\quad + \left(\sum_l \underline{a_l} \underline{b_{k-l}} \right) x^k + \dots \end{aligned}$$

“复杂度” 乘法 次数. $C \cdot d^2$
方法 $\underline{O(d^2)}$

FFT 降到 $O(d \log d)$

f 由 x_0, x_1, \dots, x_d 的取值唯一确定.

$$\underline{x_i \neq x_j, \forall i \neq j}$$

$$\begin{pmatrix} f(x_0) \\ f(x_1) \\ \vdots \\ f(x_d) \end{pmatrix} = \underbrace{\begin{pmatrix} 1 & x_0 & x_0^2 & \cdots & x_0^d \\ 1 & x_1 & x_1^2 & \cdots & x_1^d \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & x_d & x_d^2 & \cdots & x_d^d \end{pmatrix}}_{矩阵 M} \begin{pmatrix} a_0 \\ a_1 \\ \vdots \\ a_d \end{pmatrix}$$

矩阵 M 为 微量矩阵.

$$|M| = \prod_{i < j} (x_i - x_j) \neq 0. \quad M \text{ 可逆.}$$

有

$$\begin{pmatrix} a_0 \\ \vdots \\ a_d \end{pmatrix} = M^{-1} \cdot \begin{pmatrix} f(x_0) \\ \vdots \\ f(x_d) \end{pmatrix}$$

$$h(x) = \underline{f(x) \cdot g(x)} \quad \deg \leq 2d.$$

$$(x_0, \dots, x_{2d})$$

$$h(x_i) = \underline{f(x_i) \cdot g(x_i)}$$

$O(d)$ 次运算.

i) 例：代入值 (Evaluation)

左乘 $M \cdot \begin{pmatrix} \end{pmatrix}$

$\frac{(2^{1+1})}{n} \frac{(2^{d+1})}{n} \begin{pmatrix} a_0 \\ \vdots \\ a_d \\ 0 \\ \vdots \\ 0 \end{pmatrix}$

$O(d^2)$ $O(dn)$

恢复 (recover) h 系数

左乘 $M^{-1} \cdot \begin{pmatrix} \end{pmatrix}_{n \times n}$

$O(d^2)$

(improve) $O(d^2)$ $2/n$

取 $x_1, \dots, \underbrace{x_{n/2}}, \underbrace{-x_1, \dots, -x_{n/2}}$

$f(x)$ even function (偶函数)

$$f(x_i) = f(-x_i)$$

odd function (奇函数)

$$f(x_i) = -f(-x_i)$$

$$f(x) = x^4 + 3x^3 + 2x^2 + x + 1$$

$$= \frac{(x^4 + 2x^2 + 1)}{f_e(x^2)} + \frac{(3x^3 + x)}{x \cdot f_o(x^2)}$$

deg f = d

$$\deg f_e, \deg f_o \leq \frac{d}{2}$$

$$\underline{f(x_i)} = \underline{f_e(x_i^2)} + x_i \underline{f_o(x_i^2)}$$

$$\underline{f(-x_i)} = \underline{f_e(x_i^2)} - x_i \underline{f_o(x_i^2)}$$

$$\frac{f_e(x_i^2)}{x_1^2 \cdots x_{\frac{n}{2}}^2}$$

$$\frac{f_o(x_i^2)}{x_1^2 \cdots x_{\frac{n}{2}}^2}$$

$$f(x_i)$$

$\pm x_1, \pm x_2 \dots \pm x_{n/2}$ 不是 \pm 配对
 $x_1^2, \dots, x_{\frac{n}{2}}^2$ 是 \pm 配对

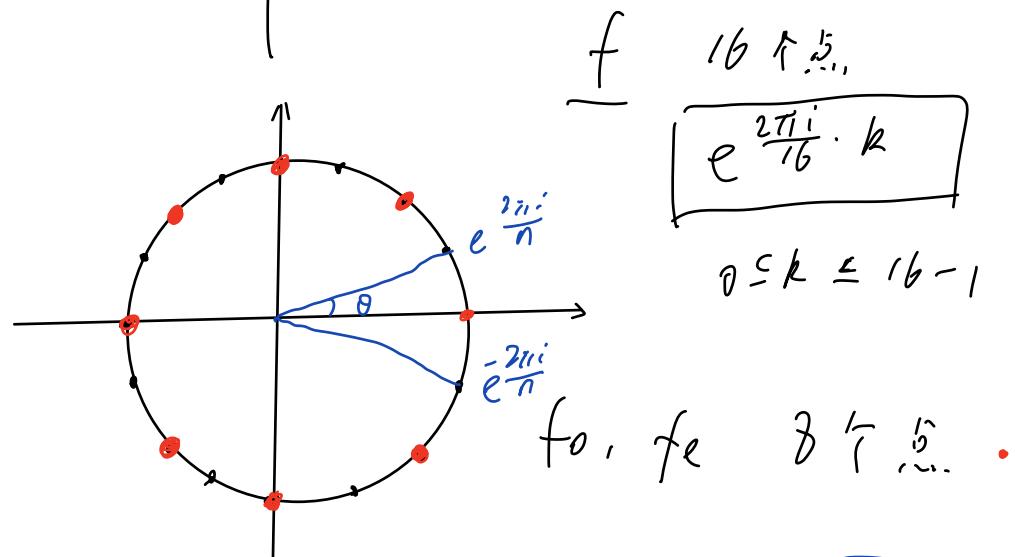
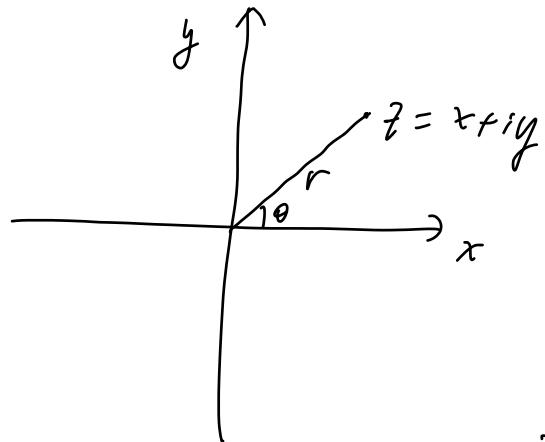
$$\begin{array}{c}
 f \\
 | \qquad | \\
 x_1 \quad -x_1 \\
 \backslash \qquad / \\
 f_0, \quad f_L \\
 \hline
 \underline{x_1^2 = 1} \qquad \qquad \qquad \underline{x_2^2 = -x_1^2 = -1}
 \end{array}$$

$\exists | \lambda$ 复数 \square

$$\boxed{x_1^4} = 1.$$

$$\begin{array}{ccccccccc} & -1 & \sqrt{-1} & -\sqrt{-1} & e^{\frac{2\pi i x_1}{3}} & -e^{\frac{2\pi i x_1}{3}} & e^{-\frac{2\pi i x_1}{3}} & -e^{-\frac{2\pi i x_1}{3}} \\ | & \diagdown & & \diagdown & \diagdown & \diagdown & \diagdown & \diagdown \\ & & -1 & & \sqrt{-1} & & -\sqrt{-1} & \\ | & \diagdown & & \diagdown & \diagdown & & \diagdown & \\ & & -1 & & & & & \\ | & \diagdown & & & \diagdown & & & \\ & & & & -1 & & & \end{array}$$

複數乘法. $z = r \cdot e^{i\theta} = r(\cos\theta + i\sin\theta)$



1) 級數算法, $n = 2^k$ $w = e^{-\frac{2\pi i}{n}}$

$$\begin{pmatrix} f(1) \\ f(w) \\ f(w^2) \\ \vdots \\ f(w^{n-1}) \end{pmatrix} = \underbrace{\begin{pmatrix} 1 & 1 & \cdots & 1 \\ 1 & w & \cdots & w^{n-1} \\ 1 & w^2 & w^4 & \cdots & w^{2(n-1)} \\ \vdots & \vdots & & & \vdots \\ 1 & & & & w^{(n-1)n} \end{pmatrix}}_{F_n} \begin{pmatrix} a_0 \\ \vdots \\ a_{n-1} \end{pmatrix}$$

F_n 左乘 $\begin{pmatrix} a_0 \\ \vdots \\ a_{n-1} \end{pmatrix}$ 有 recursive 算法.
 “复杂度” $O(n \log n)$ 相对于直接算法
 $O(n^2)$

“相似” 矩阵
 F_n^{-1} 左乘 $\begin{pmatrix} f(1) \\ \vdots \\ f(w^{n-1}) \end{pmatrix}$ 是否可以简化.

定理: $(\bar{F}_n)^T \cdot F_n = n \cdot I_n$.

\bar{F}_n 是 F_n 中每个元素作 $\sqrt[n]{\text{复数}}$

证明: $F_n = (v_0, \dots, v_{n-1})$

$$(\bar{F}_n)^T \cdot F_n = \left(\begin{array}{c} \bar{v}_0^T \\ \bar{v}_1^T \\ \vdots \\ \bar{v}_{n-1}^T \end{array} \right) \left(\begin{array}{c} v_0 \dots v_{n-1} \end{array} \right)$$

$$= \begin{pmatrix} & \\ & V_i^T \cdot v_j \\ & \end{pmatrix}$$

$$\bar{V}_i^T v_j = \sum_{k=0}^{n-1} (\bar{w}^i)^k \cdot (w^j)^k$$

$$= \sum_{k=0}^{n-1} (w^{j-i} \cdot w^i)^k$$

$(\bar{w} = w^{-1})$ $= \sum_{k=0}^{n-1} (w^{j-i})^k$

$|w|^2 = \bar{w} \cdot w = 1$

① $j=i$, $\bar{V}_i^T \cdot v_j = n$.

② $j \neq i$. $\underbrace{(j-i, n)}_{\text{最大公约数}} = 1$. $j-1, n$ 互素.

$$(w^{j-i})^k, \quad k=0, \dots, n-1, \text{ 互不相同.}$$

是 $x^n - 1 = 0$ 有根.

$$\sum_{k=0}^{n-1} (w^{j-i})^k = 0.$$

w^{j-i} 是 $\chi^{\frac{n}{(j-i,n)}} - 1$ 的根.

$$\sum_{k=0}^n (w^{j-i})^k = 0.$$

因此: $F_n^{-1} = \frac{1}{n} (\bar{F}_n^\top)$

$$F_n^{-1} \cdot \begin{pmatrix} f^{(1)} \\ \vdots \\ f^{(n)} \end{pmatrix} = \frac{1}{n} \begin{pmatrix} 1 & \cdots & 1 \\ 1/\bar{w}, (\bar{w})^2, \dots, (\bar{w})^n \\ 1/(\bar{w})^2 \\ \vdots \\ \vdots \\ 1 \end{pmatrix}$$

$\bar{w} = e^{\frac{2\pi i}{n}}$ Recursive 1) 由 \bar{w} 算 \bar{F}_n

$O(n \log n)$

矩陣角度.

$$F_{2n} \begin{pmatrix} a_0 \\ \vdots \\ a_{2n-1} \end{pmatrix} = \underbrace{\begin{pmatrix} I_n & D_n \\ - & I_n - D_n \end{pmatrix}}_{O(n)} \begin{pmatrix} \bar{F}_n & \\ & \bar{F}_n \end{pmatrix} \begin{pmatrix} f_{even} \\ f_{odd} \end{pmatrix}$$

$$D_n = \begin{pmatrix} 1 & & \\ w_{2n} & \ddots & \\ & \ddots & w_{2n}^{n-1} \end{pmatrix}$$

$$w_{2n} = e^{-\frac{2\pi i}{n}}$$

$$f_m = \underbrace{\begin{pmatrix} I_n & D_n \\ I_n - D_n \end{pmatrix}}_{O(n)} \underbrace{\begin{pmatrix} F_n & \\ & F_n \end{pmatrix}}_{\frac{1}{2} \sum n} \underbrace{\begin{pmatrix} P_{2n} \end{pmatrix}}_{2 O\left(\frac{n}{2}\right)}$$

\downarrow

$O(n)$

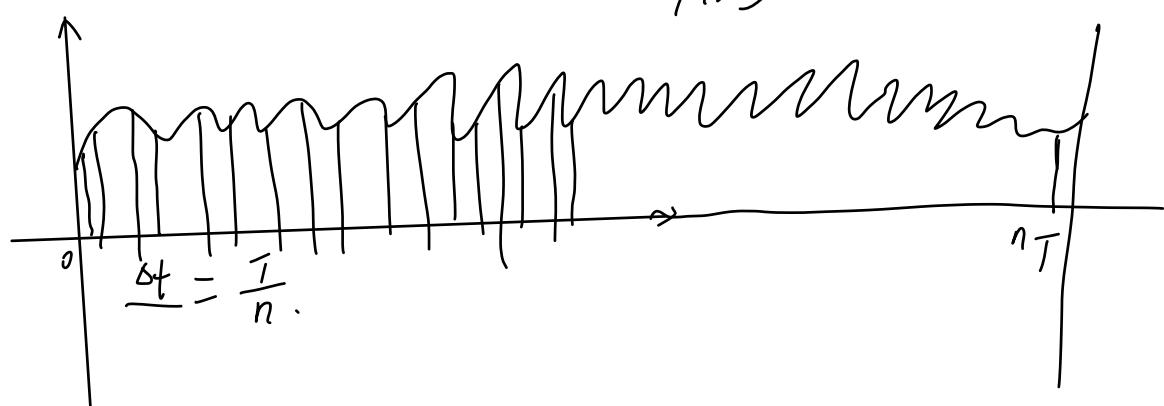
$O(n \log n)$

通常 DFT.

Discrete Fourier Transform

$$W_n = e^{-\frac{2\pi i}{n}}$$

$f(x)$



$$\text{取样} \quad f(x_k), \quad x_k = \frac{k \cdot T}{n}$$

$$k = 0, \dots, n-1.$$

$$\begin{pmatrix} f_0 \\ \vdots \\ f_{n-1} \end{pmatrix} \in \mathbb{R}^n, (\mathbb{C}^n)$$

$$DFT: \mathbb{C}^n \rightarrow \mathbb{C}^n$$

$$\begin{pmatrix} f_0 \\ \vdots \\ f_{n-1} \end{pmatrix} \mapsto F_n \cdot \begin{pmatrix} f_0 \\ \vdots \\ f_{n-1} \end{pmatrix} = \begin{pmatrix} \hat{f}_0 \\ \vdots \\ \hat{f}_{n-1} \end{pmatrix}$$

$$\text{Inverse. iDFT: } \mathbb{C}^n \rightarrow \mathbb{C}^n$$

$$\begin{pmatrix} \hat{f}_0 \\ \vdots \\ \hat{f}_{n-1} \end{pmatrix} \mapsto (F_n^{-1}) \begin{pmatrix} \hat{f}_0 \\ \vdots \\ \hat{f}_{n-1} \end{pmatrix}$$

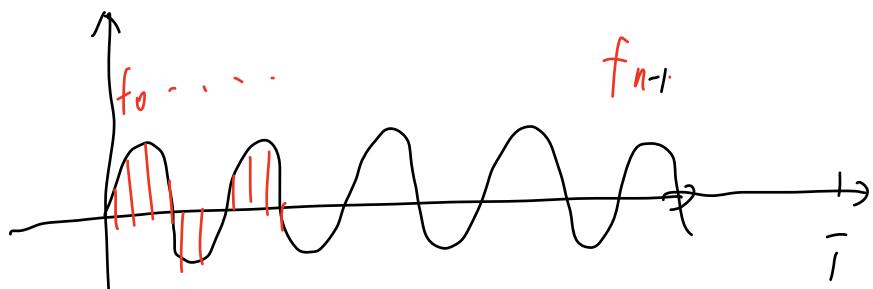
↓

$$\frac{1}{\sqrt{n}} (\bar{F}_n)^T$$

$$(f = \sin(2\pi a t))$$

$$= \frac{e^{2\pi i a t} - e^{-2\pi i a t}}{2i}$$

$a \neq \frac{\pi}{N}$ 时.



$$\underline{DFT} (f_0 \dots f_{n-1})^T$$

Definitie $e^{2\pi i \alpha t} = f(x)$

$$f_k = e^{2\pi i \alpha \left(k \frac{T}{n} \right)}$$

$$DFT (f_0, \dots f_{n-1})^T$$

$$= (\hat{f}_0 \dots \hat{f}_{n-1})$$

$$\hat{f}_k = \sum_{l=0}^{n-1} (w^k)^l \cdot e^{2\pi i \left(\frac{\alpha T}{n} \cdot l \right)}$$

$$= \sum_{l=0}^{n-1} e^{2\pi i \left(\frac{\alpha T}{n} - \frac{k}{n} \right) \cdot l}$$

$$= \sum_{l=0}^{n-1} e^{2\pi i \cdot \frac{(aT-k)l}{n}}$$

aT 整數.

$\frac{k = aT \pmod n}{k \neq aT \pmod n} = n$

$\lceil f_n \rceil$ 大於 b .

$$\frac{k}{T} = a$$

$$a > 0$$

$$f(t) = e^{-2\pi i at}$$

$$k = n - aT$$

$$\text{如果 } f(t) = \overline{f(t)},$$

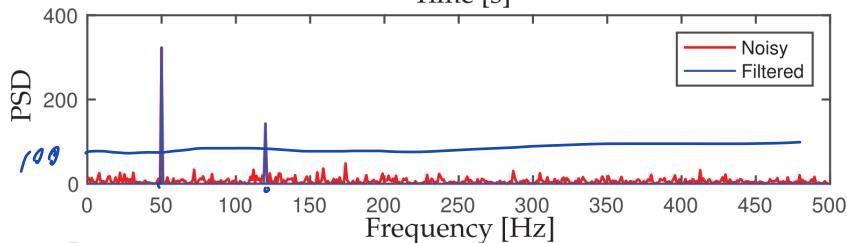
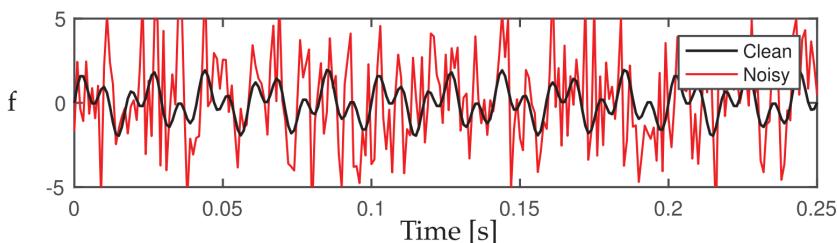
$\hat{f}_0 \dots \hat{f}_{n-1}$ 有 "對稱性"

$$\hat{f}_n = \overline{\hat{f}_{n-n}}$$

$$f = \sin(2\pi f_1 t) + \sin(2\pi f_2 t)$$

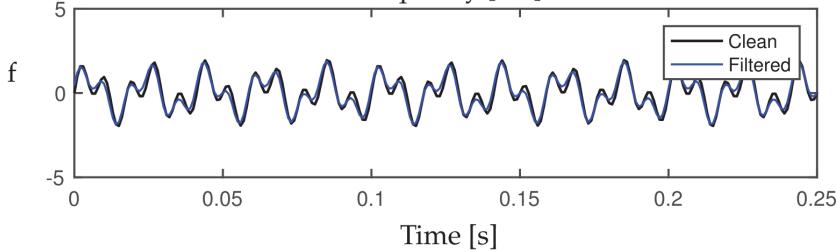
$$f_1 = 50, \quad f_2 = 120$$

$$f + \text{Noise}, \quad T=1, \quad n=1000$$



$$\frac{k}{T}$$

$S-1$

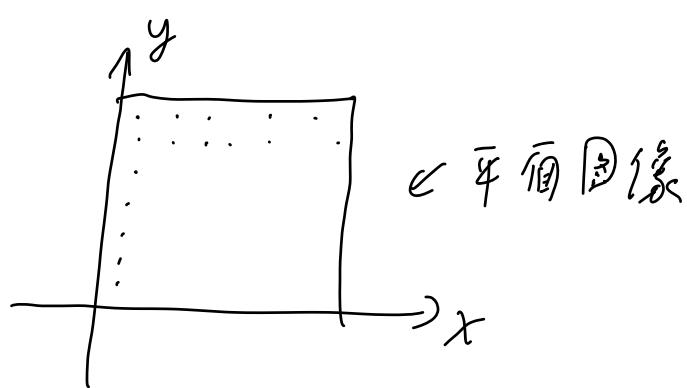


$$PSD = \frac{|\hat{f}_k|^2}{n} \quad \leftarrow \begin{array}{l} f \text{ 不加 noise } \rightarrow 0 \text{ 值} \\ |\hat{f}_k|^2 \text{ 在 } 50 \text{ 到 } 120. \end{array}$$

$$\underbrace{\text{filtered } f}_{=} \underbrace{i DFT}_{=} \underbrace{(\hat{f}_k \cdot \underbrace{\{PSD > 100\}}_{\text{只保留值}})}_{\downarrow \text{只保留值}}$$

① De noise

② compress



(x_k, y_l) 为 $f(x_k, y_l)$ 的 DFT.

若非 $|\hat{f}(k, l)|$ 小的值.