

宽带信号: 10Hz 100Hz 50Hz

① 分帧 ② 加窗 ③ fft .

$X[n] \rightarrow 16000 = N = \text{length } X$

$i \quad 2 \quad 3 \quad 4 \quad \dots \quad N-2 \quad N-1 \quad N$

=维 Y $\underbrace{\left(2 \lfloor \frac{N}{512} \rfloor + 1\right)}_{\text{行}}, \quad \underline{512}$ 列的二维数组

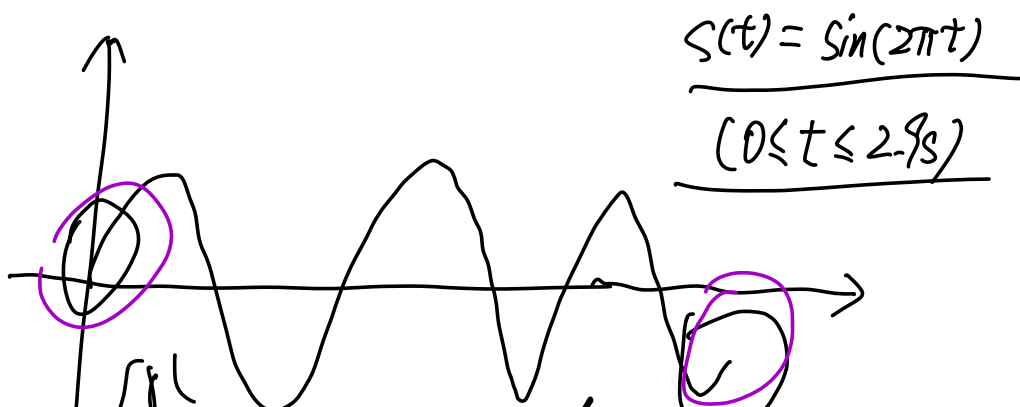
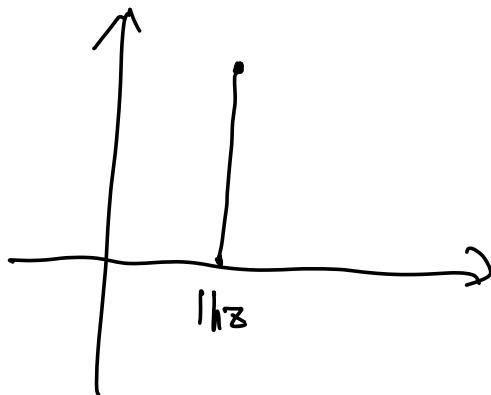
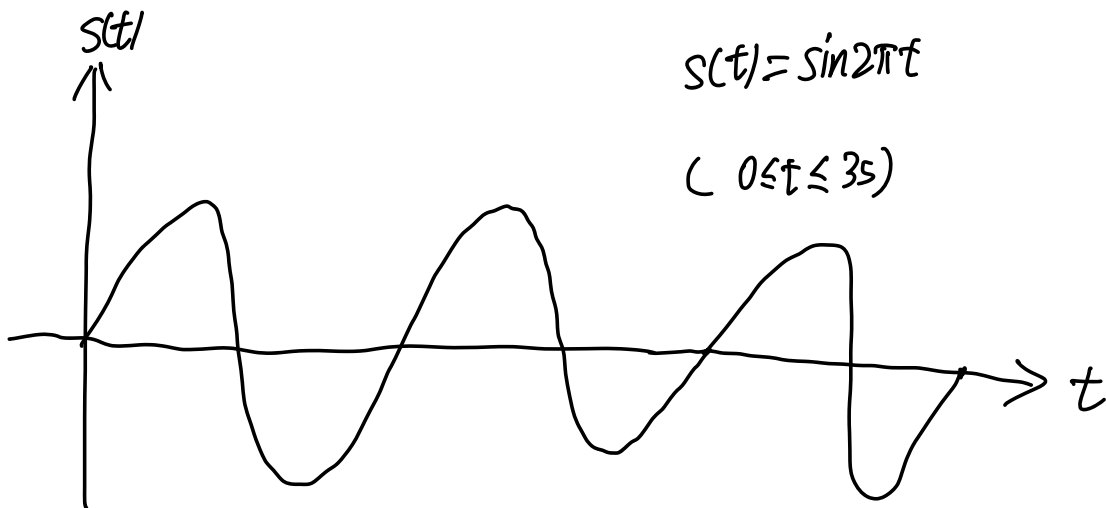
具体:

第一帧: 将 $x[0] \sim x[512]$ 存在二维数组 Y 的第一行.

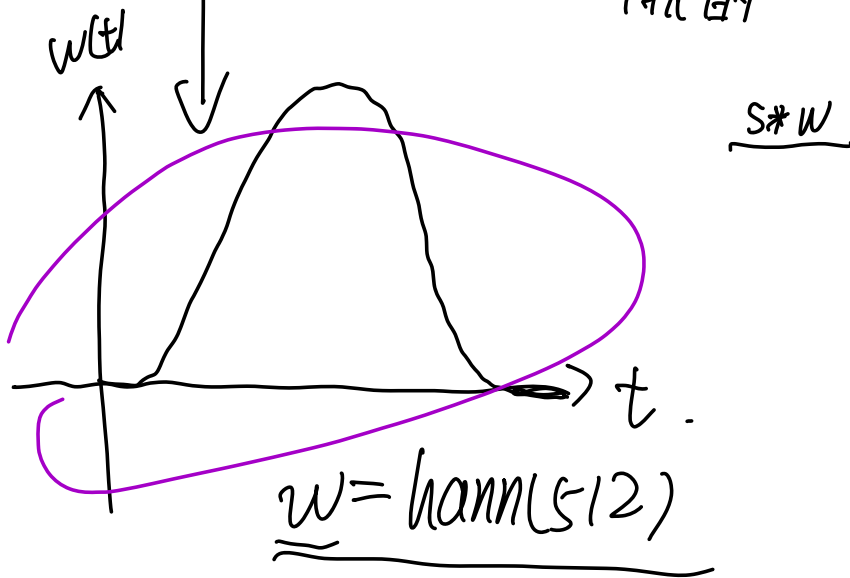
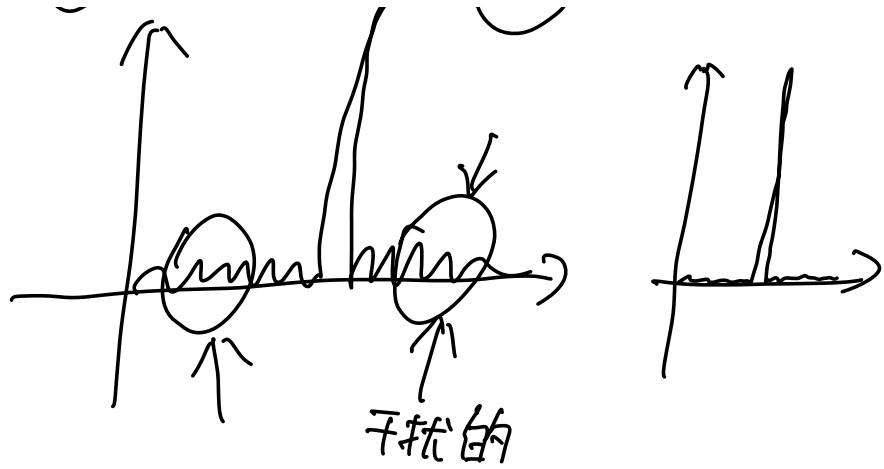
第二帧: 将 $x[257] \sim x[769]$ 存在二维数组 Y 的第二行

$\equiv \quad \underline{x[513] \sim x[1024]} \quad \equiv$

$$y = \begin{bmatrix} x[1] & x[2] & x[3] & \dots & x[512] \\ x[257] & x[258] & \dots & \dots & x[768] \\ x[513] & x[514] & \dots & \dots & x[1024] \\ x[769] & x[770] & \dots & \dots & \dots \end{bmatrix}$$



512



$$y = \begin{bmatrix} w[0]x[1] & w[2]x[2] & w[3]x[3] & \dots & w[512]x[512] \\ w[1]x[257] & w[3]x[258] & \dots & \dots & w[512]x[768] \\ w[2]x[513] & w[4]x[514] & \dots & \dots & w[512]x[1024] \\ w[5]x[769] & w[6]x[770] & \dots & \dots & \dots \end{bmatrix}$$

fft

回到第二题:

4个麦克风信号, 4个信号分别做短时傅里叶变换.

对于每个信号的每一帧你都会得到一个长度为512的数组.

$$f_s = 44100 \text{ Hz}$$

$$\text{fft}(\boxed{}_1 \quad \boxed{}_2 \quad \boxed{}_3 \quad \dots \quad \boxed{}_{512})$$

$f[k]$

$$\begin{array}{ccccccc} \boxed{} & \boxed{} & \boxed{} & - & - & - & \boxed{} & 512 \\ \uparrow & & & & & & & \\ & & \frac{(k-1)}{512} f_s = t & & & & \frac{(K-1)}{N} f_s \end{array}$$

回到music 算法

9: -90; 90

Multiple Signal Classification algorithm (MUSIC algorithm)

Let \mathbf{U}_n denote the $J \times (J-P)$ matrix containing the $J-P$ eigenvectors corresponding to all zero eigenvalues.

$$P_{music}(\theta) = \frac{1}{\sum_{i=1}^{J-P} |\mathbf{a}^H(\theta) \mathbf{u}_i|^2} = \frac{1}{\mathbf{a}^H(\theta) \mathbf{U}_n \mathbf{U}_n^H \mathbf{a}(\theta)}$$

$$\mathbf{a}(\theta) = \begin{bmatrix} e^{-j2\pi d \sin \theta / \lambda} \\ e^{-j4\pi d \sin \theta / \lambda} \\ \vdots \\ e^{-j(M-1)2\pi d \sin \theta / \lambda} \end{bmatrix}$$

The P largest peaks of $P_{music}(\theta)$ provide the source DOA, but the EVD of $\mathbf{A} \mathbf{R}_x \mathbf{A}^H$ is not available in practice.

$$\mathbf{A} \mathbf{R}_x \mathbf{A}^H \mathbf{u}_i = \lambda \cdot \mathbf{u}_i \Rightarrow \mathbf{R}_x \mathbf{u}_i = (\mathbf{A} \mathbf{R}_s \mathbf{A}^H + \mathbf{R}_n) \mathbf{u}_i = (\lambda + \sigma_n^2) \mathbf{u}_i$$

So now \mathbf{u}_i is eigenvector of the smallest eigenvalues of \mathbf{R}_x which means \mathbf{U}_n contains the eigenvectors corresponding to the $J-P$ smallest eigenvalues of \mathbf{R}_x .

$$\text{In practice, } \hat{\mathbf{R}}_x = \frac{1}{N} \sum_{k=1}^N \mathbf{x}[k] \mathbf{x}^H[k].$$

① $\underline{\mathbf{R}}_x = \frac{1}{N} \mathbf{X} \mathbf{X}^H$

② 求 \mathbf{R}_x 的最小特征值以及它对应

的特征向量，
 $\begin{pmatrix} \alpha_1 \\ \alpha_2 \\ \alpha_3 \\ \alpha_4 \end{pmatrix} \quad \begin{pmatrix} \beta_1 \\ \beta_2 \\ \beta_3 \\ \beta_4 \end{pmatrix}$
 ↖ ↗

$$\mathbf{U} = \begin{pmatrix} \alpha_1 & \beta_1 \\ \alpha_2 & \beta_2 \\ \alpha_3 & \beta_3 \\ \alpha_4 & \beta_4 \end{pmatrix}$$

$$f_0 = \frac{f_s}{512}$$

对于每一个 $k f_0$ ($k=1, 2, \dots, 256$)

我们都可以算出对应的 $\hat{\mathbf{R}}_x^{(k f_0)}$

$$h[k+1] \rightarrow \frac{k f_s}{512}$$



$$\begin{array}{l}
 \text{第-帧fft} \quad \text{第=帧fft} \quad \text{三} \quad \text{四} \\
 \text{50 } k f_0 \rightarrow (k+1)\text{项} \quad \text{50 } k f_0 \rightarrow \text{第}(k+1)\text{项} \\
 \text{第-帧fft} \\
 \text{完的第}(k+1)\text{项, } k f_0 \quad \sim \quad \sim \quad \sim \quad \sim \\
 \sim \quad \sim \quad \sim \quad \sim \quad \sim \\
 \sim \quad \sim \quad \sim \quad \sim \quad \sim \\
 \sim \quad \sim \quad \sim \quad \sim \quad \sim
 \end{array}
 \begin{array}{l}
 \leftarrow \text{对于第一个麦麦风} \\
 \leftarrow \text{第=个} \\
 \leftarrow \text{第三} \\
 \leftarrow \text{四}
 \end{array}$$

$\chi^{(k f_0)}$
 U_x 帧数

对于每一个 $k f_0$ ($k=0,1,2,\dots$)

都可以算出 R_x

两个特征值对应的特征向量

$U^{(k f_0)}$

$$P(\theta) = \frac{1}{\sum_{k=1}^{256} \underline{a^h(\theta)} U_n^{(k f_0)} U_n^{h(k f_0)} \underline{a^h(\theta)}}$$

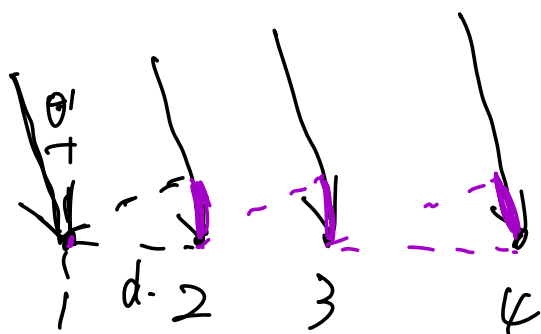
$$a^h(\theta) = \begin{pmatrix} 1 \\ e^{-2\pi j \frac{d \sin \theta}{\lambda}} \\ e^{-4\pi j \frac{d \sin \theta}{\lambda}} \\ e^{-6\pi j \frac{d \sin \theta}{\lambda}} \end{pmatrix}$$

MUSIC 算法的证明!

考虑 4 个麦克风, 2 个声源的情况.

$$a(\theta) = \begin{bmatrix} e^{-j2\pi \frac{ds \sin \theta}{\lambda}} \\ e^{-j4\pi \frac{ds \sin \theta}{\lambda}} \\ e^{-j6\pi \frac{ds \sin \theta}{\lambda}} \end{bmatrix}$$

source.



$$v = \lambda f.$$

$$f = \frac{\omega}{2\pi}$$

$$\omega = 2\pi f$$

$$\tau = \frac{ds \sin \theta}{v}$$

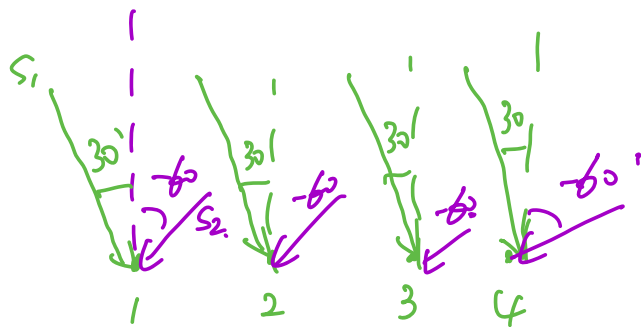
$$\lambda = \frac{v}{f}$$

$$\Delta \phi = \omega \tau.$$

$$= 2\pi f \tau.$$

$$e^{-j\Delta\phi} = e^{-j \cdot 2\pi f \tau} = e^{-j 2\pi f \frac{d \sin\theta}{v.}} \\ = e^{-j 2\pi \frac{d \sin\theta}{\lambda.}}$$

有两个声源 30° , -60° .



对于 S_1 : $a_1(\theta) = \begin{bmatrix} 1 \\ e^{j2\pi \frac{d \sin 30^\circ}{\lambda}} \\ e^{-j4\pi \frac{d \sin 30^\circ}{\lambda}} \\ e^{-j6\pi \frac{d \sin 30^\circ}{\lambda}} \end{bmatrix}$

$$\downarrow \\ x[k] = a_1(\theta) \cdot s_1[k]$$

对于 source 2:

$$a_2(\theta) = \begin{bmatrix} 1 \\ e^{-j2\pi \frac{d \sin(-60^\circ)}{\lambda}} \\ e^{-j4\pi \frac{d \sin(-60^\circ)}{\lambda}} \\ e^{-j6\pi \frac{d \sin(-60^\circ)}{\lambda}} \end{bmatrix}$$

$$\downarrow \\ x[k] = a_2(\theta) \cdot s_2[k]$$

由于 2 个，环境噪声

$$x = x_1 + x_2 + \underset{\uparrow}{n[k]} = A \cdot s[k] + n[k]$$

$$A = [a_1(\theta) \quad a_2(\theta)] = \begin{bmatrix} e^{j2\pi \frac{d \sin(-60^\circ)}{\lambda}} & e^{-j2\pi \frac{d \sin 30^\circ}{\lambda}} \\ e^{-j2\pi \frac{d \sin(-60^\circ)}{\lambda}} & e^{-j4\pi \frac{d \sin 30^\circ}{\lambda}} \\ e^{-j6\pi \frac{d \sin(-60^\circ)}{\lambda}} & e^{-j6\pi \frac{d \sin 30^\circ}{\lambda}} \end{bmatrix}$$

$$s[k] = \begin{bmatrix} \frac{s_1[k]}{\dots} \\ s_2[k] \end{bmatrix} \leftarrow \text{复指数信号}$$

下面证明 $\hat{R}_x = \frac{1}{N} \cdot x \cdot x^H$ 存在两个很小的特征值。

在这里 期望 已求平均。

$$\underline{R_x = \frac{1}{N} (A \cdot s + n) (A \cdot s + n)^H}$$

$$R_x = \frac{1}{N} (As + n) (s^H A^H + n^H)$$

$$= \frac{1}{N} (A s \cdot s^H \cdot A^H + A s \cdot n^H + n \cdot s^H \cdot A^H + n \cdot n^H)$$

$$= \frac{1}{N} (\underline{A \cdot s \cdot s^H \cdot A^H} + n \cdot n^H)$$

$$\text{关于 } \frac{1}{N} A \cdot s \cdot n^H \text{ 和 } \frac{1}{N} n \cdot s^H \cdot A^H \approx 0$$

n 高斯白噪声

$$\underline{n \ll s} \quad E[n] \approx 0.$$

$$\underline{\sum s[k] \approx 0.}$$

$$\lim_{T \rightarrow +\infty} \frac{1}{T} \int_0^T e^{j\omega t} dt \approx 0.$$

$$\lim_{T \rightarrow +\infty} \frac{1}{N} \sum_{n=0}^{+\infty} e^{j\omega n} \approx 0.$$

$$\underline{s_1(t) = e^{j(\omega_1 t + \phi_1)}}$$

$$\underline{s_2(t) = e^{j(\omega_1 t + \phi_2)}}$$

f_s .

$$s_1 = \left[e^{j(\omega_1 \frac{1}{f_s} + \phi_1)} \quad e^{j(\omega_1 \frac{2}{f_s} + \phi_1)} \quad \dots \quad e^{j(\omega_1 \frac{k}{f_s} + \phi_1)} \quad \dots \right]$$

$$s_2 = \left[e^{j(\omega_1 \frac{1}{f_s} + \phi_2)} \quad e^{j(\omega_1 \frac{2}{f_s} + \phi_2)} \quad \dots \quad e^{j(\omega_1 \frac{k}{f_s} + \phi_2)} \quad \dots \right]$$

$$f_s = 44100 \text{ Hz}$$

$$\begin{bmatrix} s_1 \\ s_2 \end{bmatrix} [s_1^h \mid s_2^h] = \begin{bmatrix} L & L e^{j(\phi_2 - \phi_1)} \\ L e^{j(\phi_1 - \phi_2)} & L \end{bmatrix}$$

$$A \cdot S \cdot S^h \cdot A^h = U_i^h A \begin{bmatrix} L & L e^{j(\phi_1 - \phi_2)} \\ L e^{j(\phi_2 - \phi_1)} & L \end{bmatrix} (U_i A^h)^h = 0.$$

↑

$$\Rightarrow U_i^h A = 0$$

不存在 U_i st. $A^h U_i \neq 0$

$$U_i^H A = 0$$

$$A^H U_i = 0$$

a_i

$$\begin{bmatrix} 1 & e^{+j2\pi \frac{d \sin(-60)}{\lambda}} & e^{+j4\pi \frac{d \sin(-60)}{\lambda}} & e^{+j6\pi \frac{d \sin(-60)}{\lambda}} \\ 1 & e^{+j2\pi \frac{d \sin 30}{\lambda}} & e^{+j4\pi \frac{d \sin 30}{\lambda}} & e^{+j6\pi \frac{d \sin 30}{\lambda}} \end{bmatrix} \begin{bmatrix} \alpha_1 \\ \alpha_2 \\ \alpha_3 \\ \alpha_4 \end{bmatrix} \approx \mathbf{0}$$

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Let U_n denote the $J \times (J-P)$ matrix containing the $J-P$ eigenvectors corresponding to all zero eigenvalues.

$$P_{music}(\theta) = \frac{1}{\sum_{i=1}^{J-P} |\mathbf{a}^H(\theta) \mathbf{u}_i|^2} = \frac{1}{\mathbf{a}^H(\theta) U_n U_n^H \mathbf{a}(\theta)}$$

$$\theta = -60$$

$$\theta = 30$$

$$\frac{1}{\sigma_n^2 U_i}$$

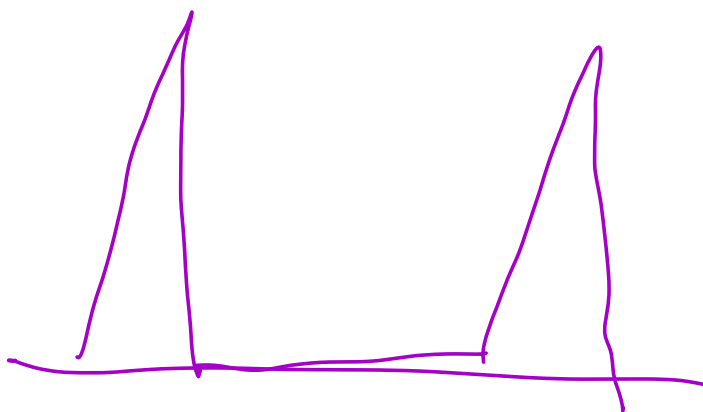
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So now \mathbf{u}_i is eigenvector of the smallest eigenvalues of R_x which means U_n contains the eigenvectors corresponding to the $J-P$ smallest eigenvalues of R_x .

$$\text{In practice, } \hat{R}_x = \frac{1}{N} \sum_{k=1}^N \mathbf{x}[k] \mathbf{x}^H[k].$$

$$R_x(\mathbf{u}_i) = \underset{\lambda=0}{(\lambda + \sigma_n^2)} \mathbf{u}_i = \sigma_n^2 \mathbf{u}_i$$



$$S_m(x) = \frac{a_0}{2} + \sum_{n=1}^m (a_n \cos nx + b_n \sin nx)$$

$$a_n = \frac{1}{\pi} \int_{-\pi}^{\pi} f(t) \cos nt \, dt$$

$$b_n = \frac{1}{\pi} \int_{-\pi}^{\pi} f(t) \sin nt \, dt$$

$$S_m(x) = \frac{1}{\pi} \int_0^{\pi} f(t) \left[\frac{1}{2} + \sum_{n=1}^m \cos n(t-x) \right] dt.$$

$$\lim_{m \rightarrow +\infty} S_m(x) = f(x)$$

Riemann 引理:

Dirichlet 引理:

\Rightarrow Dirichlet - Jordan 判别法

离散时间傅里叶级数

$$\underline{x(t)} \xrightarrow{\text{sampling}} \underline{x[n]} \quad f_s.$$

N 为整数.

$$x(t) = \sin 10\pi t \\ 0 \leq t \leq 5s$$

$$f_s = 10\text{Hz}$$

$$x[n] = \sin \pi n$$

$$x[k] = x\left(\frac{k}{f_s}\right)$$

$$x[n].$$

$$x[n] = \sum_{k=1}^N a_k e^{jk\left(\frac{2\pi}{N}\right)n} = \sum_{k=r+1}^{N+r} a_k e^{jk\left(\frac{2\pi}{N}\right)n}$$

r

$$e^{jk\left(\frac{2\pi}{N}\right)(n+N)} = e^{jk\left(\frac{2\pi}{N}\right) \cdot n} \quad \boxed{1}.$$

$$\sum_{n=1}^N e^{jk(2\pi/N)n} = \begin{cases} N, & k=0, \pm N, \pm 2N, \dots \\ 0 & \text{其它.} \end{cases}$$

$$\sum_{n=1}^N x[n] e^{-jr(2\pi/N)n} = \sum_{n=1}^N \sum_{k=1}^N a_k e^{j(k-r)2\pi/N \cdot n} \\ = Na_r.$$

$$a_r = \frac{1}{N} \sum_{n=1}^N x[n] e^{-jr(2\pi/N)n}.$$

$$\left\{ \begin{aligned} x[n] &= \sum_{k=1}^N a_k e^{jk(2\pi/N)n} \\ a_k &= \frac{1}{N} \sum_{n=1}^N x[n] e^{-jk(2\pi/N)n} \end{aligned} \right.$$

$$a_k = a_{k+rN} \quad (r=0, \pm 1, \pm 2, \dots)$$

$x[n]$ 实信号

$$a_{+k} = a_{-k}^*$$

$$a_{-k} = \frac{1}{N} \sum_{n=1}^N x[n] e^{jk \left(\frac{2\pi}{N}\right) n}$$

$$a_{+k} = \frac{1}{N} \sum_{n=1}^N x[n] e^{-jk \left(\frac{2\pi}{N}\right) n}$$

$$a_{+k} = a_{-k}^*$$

$$a_{+k} = a_{N-k}^*$$

812.

256

257