

1 Basic of Fourier Transform

If $f(x) = f(x + T)$ then $f(x)$ can be written as

$$f(x) = \sum_{-\infty}^{+\infty} c_k e^{\frac{2\pi i k x}{T}}$$

because

$$e^{\frac{2\pi i k x}{T}} = e^{\frac{2\pi i k (x+T)}{T}}$$

Based on orthogonality,

$$c_k = \frac{1}{T} \int_0^T f(x) e^{-i \frac{2\pi k x}{T}} dx$$

The above is the Fourier transform in continuous case, in discrete case If $x = n\Delta t$, where $n = 1 \dots N$, and $T = N\Delta t$,

$$c_k = \frac{1}{N\Delta t} \sum_{n=1}^N f(n\Delta t) e^{-i 2\pi k \frac{1}{N\Delta t} n\Delta t} d(n\Delta t) = \frac{1}{N} \sum_{n=1}^N f(n) e^{-i 2\pi k \frac{n}{N}}$$

This is the discrete Fourier transform.

$$\Delta f = \frac{1}{T} = \frac{1}{N\Delta t} = \frac{F_s}{N}$$

Where F_s , N are the sample frequency and number of samples.

Properties

1) To be eligible, $f(x)$ has to be a period function with time T . This leads to uniform sampling theorem used in signal processing. The uniform sampling theorem states w

2) If $f(x)$ is real, which means $f(x) = f^*(x)$.

$$\sum_{-\infty}^{+\infty} c_k e^{2\pi i \frac{1}{T} k x} = \sum_{-\infty}^{+\infty} c_k^* e^{-2\pi i \frac{1}{T} k x} = \sum_{-\infty}^{+\infty} c_{-k} e^{2\pi i \frac{1}{T} k x}$$

so $c_k = c_{-k}^*$, $||c_k|| = ||c_{-k}||$.

3) $c_k = c_{k+N}$. So when a signal contains frequency component no larger than B , in other words, the bandwidth of the signal is $2B(-B \text{ to } B)$, then in order to capture the whole bandwidth of the signal, $N\Delta f > 2B$. This leads to Nyquist sampling theorem $F_s > 2B(\text{bandwidth})$.

2 Fast Fourier Transform

$$X_k = \sum_{n=0}^{N-1} x_n e^{-i 2\pi k \frac{n}{N}}$$

let

$$u_k = e^{-i2\pi k \frac{p}{N}}$$

then we have the basis orthogonality

$$u_{k1}^T u_{k2} = N \delta_{k1, k2}$$

We recognize we can write X_k with even index terms and odd index terms

$$X_k = \text{Even index parts} + \text{Odd index parts}$$

$$\begin{aligned} &= \sum_{m=0}^{N/2-1} x_{2m} e^{-\frac{2\pi i}{N} 2mk} + \sum_{m=0}^{N/2-1} x_{2m+1} e^{-\frac{2\pi i}{N} (2m+1)k} \\ &= \sum_{m=0}^{N/2-1} x_{2m} e^{-\frac{2\pi i}{N/2} mk} \end{aligned}$$

(We can view this as Fourier Transform of $N/2$ even indexed points, where k is $0, 1, \dots, N/2$)

$$+ e^{-\frac{2\pi i}{N} k}$$

$$\sum_{m=0}^{N/2-1} x_{2m+1} e^{-\frac{2\pi i}{N/2} mk}$$

(We can view this as Fourier Transform of $N/2$ odd indexed points, where k is $0, 1, \dots, N/2$)

(Since each part is a Fourier transform of $N/2$ points, k has to be smaller than $N/2$)

$$= E_k + e^{-\frac{2\pi i}{N} k} O_k$$

As noted, the above derivation is for $k < N/2$, a very similar derivation for $N/2 \leq k < N$ leads to

$$X_{k+N/2} = E_k - e^{-\frac{2\pi i}{N} k} O_k$$

Now we have divided the FFT of N points to two FFT with $N/2$ points. Keep going till we reach the size to one, then combine together recursively.