

(Continuous) Fourier Transform (FT)

continuous $\xrightarrow{\quad}$ $X(\omega) = \int_{-\infty}^{\infty} x(t) e^{-j\omega t} dt$

$$X(f) = \int_{-\infty}^{\infty} x(t) e^{-j2\pi f t} dt$$

continuous $\xrightarrow{\quad}$ $x(t) = \frac{1}{2\pi} \int_{-\infty}^{\infty} X(\omega) e^{j\omega t} d\omega$

$$x(t) = \int_{-\infty}^{\infty} X(f) e^{j2\pi f t} df$$

Discrete Time Fourier Transform (DTFT)

periodic with period 2π $\xrightarrow{\quad}$ continuous $\xrightarrow{\quad}$ $X(\Omega) = \sum_{n=-\infty}^{\infty} x[n] e^{-j\Omega n}$

discrete $\xrightarrow{\quad}$ $x[n] = \frac{1}{2\pi} \int_0^{2\pi} X(\Omega) e^{j\Omega n} d\Omega$

Note: Ω is different from ω .

Question: 1) if we view $x[n]$ as $x(n)$ why not taking continuous Fourier transform?

2) why ~~take~~ only integrate over only one period in inverse transform

Fourier Series

discrete $\xrightarrow{\quad}$ $X[k] = \frac{1}{T} \int_0^T x(t) e^{-j\omega_k t} dt, \quad \omega_k = \frac{2\pi k}{T}$

Question: 1) why integrate over only one period in forward transform

periodic with period T. $\xrightarrow{\quad}$ continuous $\xrightarrow{\quad}$ $x(t) = \sum_{k=-\infty}^{\infty} X[k] e^{j\omega_k t}$

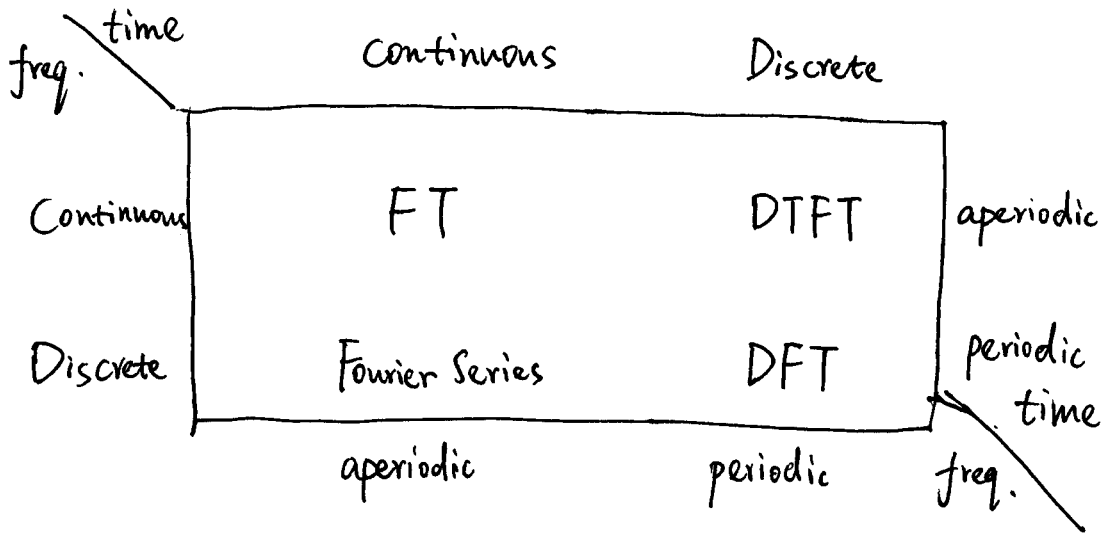
Discrete Fourier Transform (DFT)

periodic $\xrightarrow{\quad}$ discrete $\xrightarrow{\quad}$ $X[k] = \sum_{n=0}^{N-1} x[n] e^{-j\omega_k n}$

$$\omega_k = \frac{2\pi k}{N}$$

1) why sum over only one period in both forward and inverse transforms?

periodic $\xrightarrow{\quad}$ discrete $\xrightarrow{\quad}$ $x[n] = \frac{1}{N} \sum_{k=0}^{N-1} X[k] e^{j\omega_k n}$



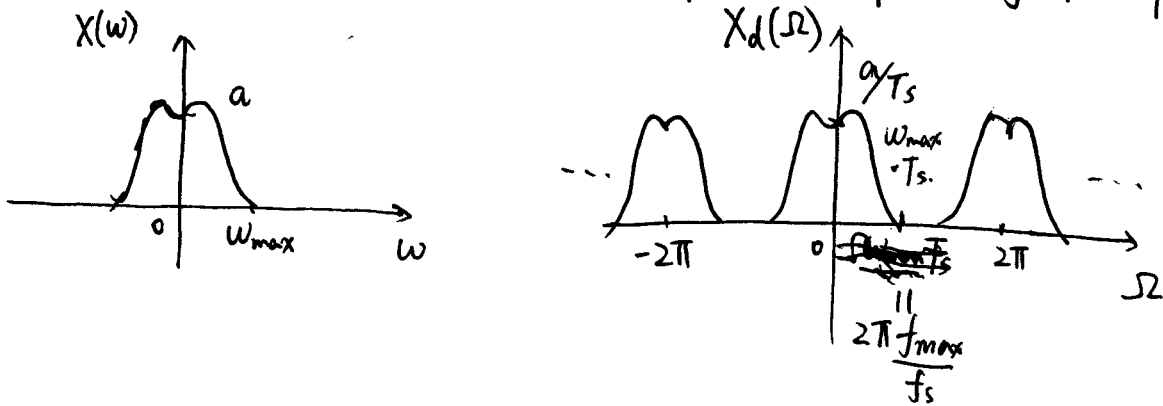
Relation between FT and DTFT (Sampling Theorem Revisit)

Let $x[n] = x(nT_s)$, $f_s = \frac{1}{T_s}$ Sampling rate.

$$\text{then } X_d(\Omega) = \frac{1}{T_s} \sum_{m=-\infty}^{\infty} X\left(\frac{\Omega}{T_s} + m \frac{2\pi}{T_s}\right)$$

↑
↑
 DTFT spectrum FT spectrum

i.e. DTFT spectrum of $x[n]$ is periodic replicas of FT spectrum of $x(t)$.
(x-axis scaled, y-axis scaled)



$\therefore f_{max}$ should be less than $f_s/2$ to prevent aliasing.

Proof: Consider inverse transforms.

(3)

inverse FT:
$$x(t) = \frac{1}{2\pi} \int_{-\infty}^{\infty} X(\omega) e^{j\omega t} d\omega$$

$$= \frac{1}{2\pi} \int_0^{2\pi} \sum_{m=-\infty}^{\infty} \int_{m \cdot \frac{2\pi}{T_s}}^{(m+1) \cdot \frac{2\pi}{T_s}} X(\omega) e^{j\omega t} d\omega$$

$$= \frac{1}{2\pi} \sum_{m=-\infty}^{\infty} \int_0^{\frac{2\pi}{T_s}} X(\omega + m \frac{2\pi}{T_s}) e^{j(\omega + m \frac{2\pi}{T_s})t} d\omega$$

$$= \frac{1}{2\pi} \int_0^{\frac{2\pi}{T_s}} e^{j\omega t} \sum_{m=-\infty}^{\infty} X(\omega + m \frac{2\pi}{T_s}) e^{jm \frac{2\pi}{T_s} t} d\omega$$

$$\therefore x[n] = x(nT_s) = \frac{1}{2\pi} \int_0^{\frac{2\pi}{T_s}} e^{j\omega nT_s} \sum_{m=-\infty}^{\infty} X(\omega + m \frac{2\pi}{T_s}) e^{jm \frac{2\pi}{T_s} nT_s} d\omega$$

Let $\Omega = \omega \cdot T_s$

$$= \frac{1}{2\pi} \int_0^{2\pi} e^{j\Omega n} \sum_{m=-\infty}^{\infty} X\left(\frac{\Omega}{T_s} + m \frac{2\pi}{T_s}\right) \cdot \frac{1}{T_s} d\Omega$$

By definition,
$$x_d[n] = \frac{1}{2\pi} \int_0^{2\pi} X_d(\Omega) e^{j\Omega n} d\Omega$$

Comparing the above two equations, we have

$$X_d(\Omega) = \frac{1}{T_s} \sum_{m=-\infty}^{\infty} X\left(\frac{\Omega}{T_s} + m \frac{2\pi}{T_s}\right) \quad \square$$

Therefore, for simplicity, we only specify one period of $X_d(\Omega)$, i.e., $-\pi \leq \Omega < \pi$, or $0 \leq \Omega < 2\pi$.

In computers, everything is discrete, including Fourier spectra. ④

How to discretize DTFT spectrum $X_d(\Omega)$?

Suppose we have $x[n]$ that only has non-zero values between for $0 \leq n < N$

Taking DTFT, we have

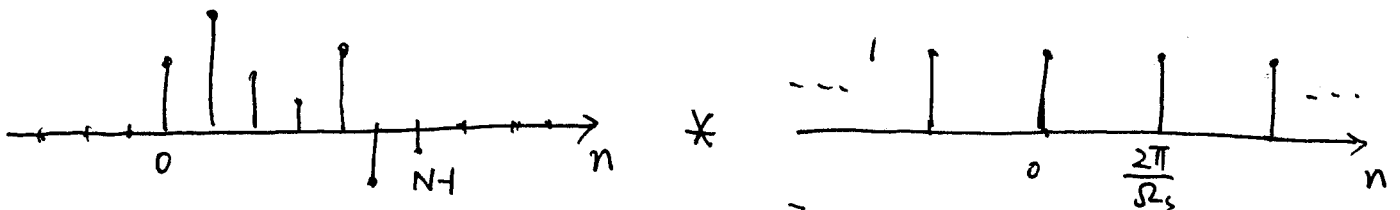
$$X_d(\Omega) = \sum_{n=-\infty}^{\infty} x[n] e^{-j\Omega n} = \sum_{n=0}^{N-1} x[n] e^{-j\Omega n}, \text{ periodic } 2\pi$$

How should we sample $X_d(\Omega)$ so that $x[n]$ can be recovered from the samples? Another application of Sampling Theorem!

Sampling $X_d(\Omega) \Leftrightarrow X_d(\Omega) \cdot \delta(\Omega - k\Omega_s)$
with period Ω_s

\Downarrow inverse DTFT

$$\begin{aligned} & \cancel{x[n]} * \delta\left[n - m \frac{2\pi}{\Omega_s}\right] \\ & \cancel{x[nT]} * \delta(\dots) \end{aligned}$$



To prevent aliasing, we need $\Omega_s \geq \frac{2\pi}{N}$, i.e., $\Omega_s \leq \frac{2\pi}{N}$, i.e., we should use at least N samples to represent $X_d(\Omega)$ for $0 \leq \Omega < 2\pi$

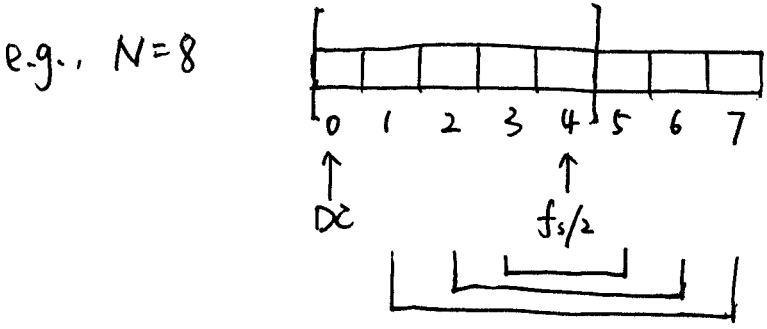
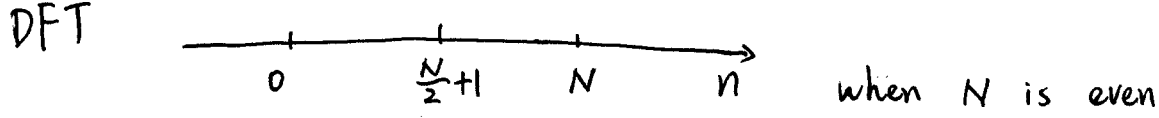
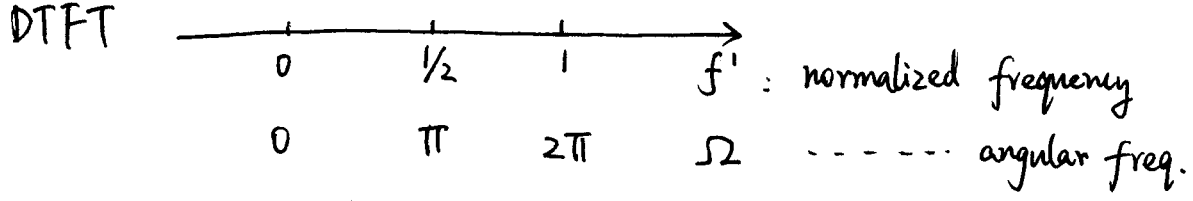
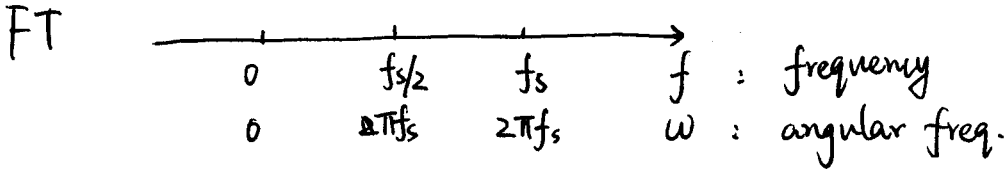
If we use N samples, then the frequencies of these samples are $\frac{2\pi}{N} \cdot k$ for $0 \leq k \leq N-1$. We can define discrete spectrum

DFT: $X[k] = X_d(k\Omega_s) = \sum_{n=0}^{N-1} x[n] e^{-jk\Omega_s n} = \sum_{n=0}^{N-1} x[n] e^{-j\frac{2\pi kn}{N}}$ (periodic, N).

IDFT: $\hat{x}[n] = \frac{1}{N} \sum_{k=0}^{N-1} X[k] e^{j\frac{2\pi kn}{N}}$ (periodic, N)

when $0 \leq n \leq N-1$, $\hat{x}[n] = x[n]$. And we do not use $\hat{x}[n]$ for $n \notin [0, N-1]$ anyway, so DFT pair is defined for $0 \leq n, k \leq N-1$.

We have discussed the relation between FT, DTFT, DFT, and Fourier Series



conjugate symmetric since time-domain signal is real.
 We only need half-spectrum, i.e., $0 \leq n \leq \frac{N}{2} + 1$, for audio signals.

Fast Fourier Transform (FFT) is an efficient algorithm to calculate DFT

Complexity : DFT : $O(N^2)$ N^2 complex multiplications

FFT : $O(N \log N)$, "divide-and-conquer" Advanced Algorithms

Applications. convolution: naive calculation : $O(N^2)$
 $x[n] * y[n]$

first discovered by (Gauss, 1805) using FFT: $IFFT(FFT(x) \times FFT(y))$
 re-discovered by (Cooley, Tukey) 1965. $\therefore 3N \log N + N$ complex multiplications. $O(N \log N)$.

\uparrow \uparrow \uparrow \uparrow
 $N \log N$ $N \log N$ N $N \log N$

Correlation, Auto-correlation, etc.

~~No. 1~~ of Top 10 Algorithms in 20th century!
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