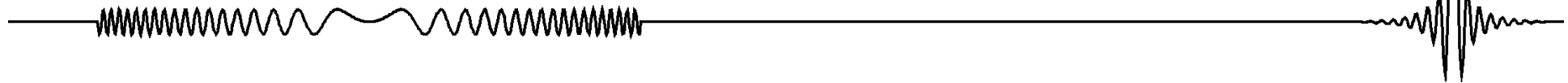


Applied Signal Processing & Computer Science



Chapter 5: Frequency Domain Analysis

Xiaoxiang Zhu

Xiao.zhu@dlr.de

Remote Sensing Technology

TU München

&

Remote Sensing Technology Institute

German Aerospace Center (DLR)

Oberpfaffenhofen

Chapter 5: Frequency Domain Analysis

5.1 Fourier Transform

5.2 Discrete-Time Fourier Transform(DTFT)

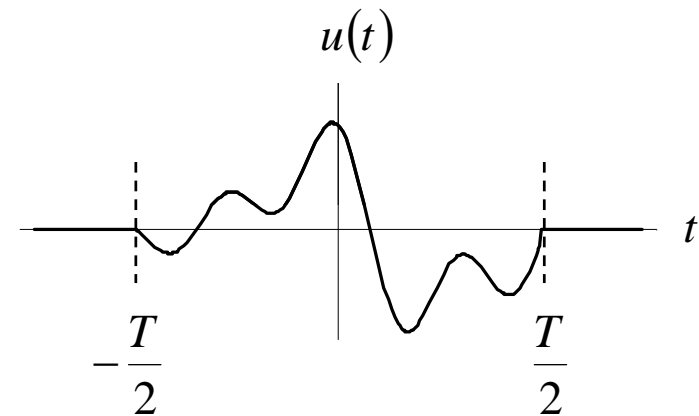
5.3 Properties of the Fourier Transform

5.4 Discrete Fourier Transform (DFT)

Fourier Series

- Expansion of energy- and time-limited signals in series of orthogonal harmonic basis functions

- Without loss of generality: $-\frac{T}{2} < t < +\frac{T}{2}$



- Fourier series: $u(t) = \sum_{n=-\infty}^{+\infty} c_n \cdot \Psi_n(t) \quad \forall \quad |t| < \frac{T}{2}$ with $\Psi_n(t) = \exp\left(j 2 \pi \frac{n}{T} t\right)$
- \uparrow
 Fourier
coefficient of
order n

\uparrow
 basis function
(harmonic oscillation)

\uparrow
 frequency $n \cdot f_0$
with $f_0 = 1/T$

Continuous-time Fourier Transform

$$U(f) = \int_{-\infty}^{+\infty} u(t) \exp(-j 2\pi f t) dt \quad \text{Fourier Transform}$$

$$u(t) = \int_{-\infty}^{+\infty} U(f) \exp(j 2\pi f t) df \quad \text{inverse Fourier Transform}$$

with $u(t)$ absolute integrable: $\int_{-\infty}^{+\infty} |u(t)| dt < \infty$ (sufficient condition)

Some power-limited signals can also be Fourier transformed :

$$U(f) = \lim_{\varepsilon \rightarrow 0} \int_{-\infty}^{+\infty} u(t) \exp(-\varepsilon |t|) \exp(-j 2\pi f t) dt$$

Common symbols :

$$\begin{aligned} u(t) &\circ \longrightarrow \bullet U(f) \\ u(t) &\rightarrow U(f) \\ u(t) &\leftrightarrow U(f) \end{aligned}$$

Discrete-Time and Discrete-Frequency Signals

$u[n]$ time-limited, e.g. for $n = 0, 1, 2, \dots, N - 1$

\Rightarrow spectrum $U_d(f)$ can be sampled
sampling interval $\leq 1/(NT)$

$\Rightarrow u[n]$ continues periodically with period $\geq N$

Caution: A time-limited signal can *never* be *exactly* frequency band-limited.
Therefore, $u[n]$ is only an *approximation* for a physical time-limited signal. In general the approximation gets better for large N .

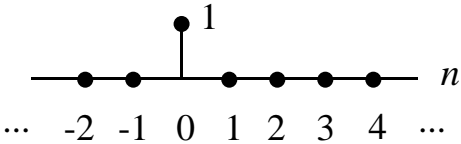
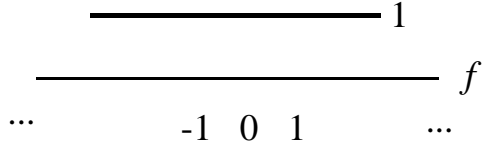
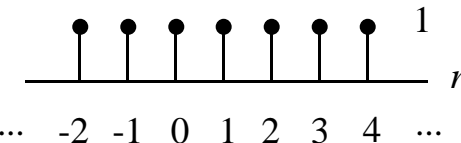
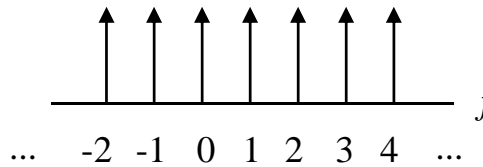
Discrete-Time Fourier Transform (DTFT)

$$U_d(f) = \sum_{n=-\infty}^{+\infty} u[n] \exp(-j 2\pi f n T)$$

$$u[n] = T \int_{-1/(2T)}^{+1/(2T)} U_d(f) \exp(j 2\pi f n T) df$$

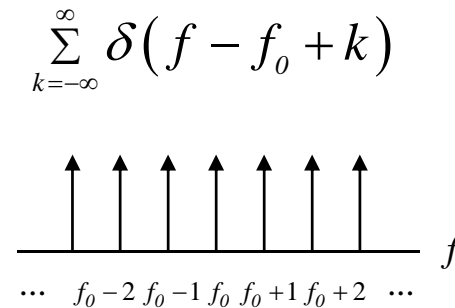
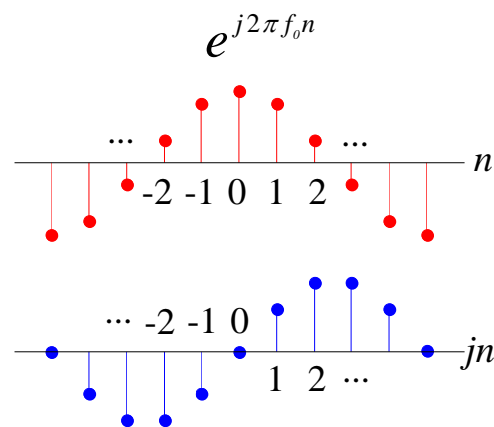
often used: normalized angular frequency: $\Omega = 2\pi f T \Rightarrow -\pi < \Omega \leq \pi$

Commonly Used DTFT Pairs

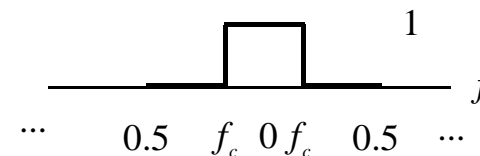
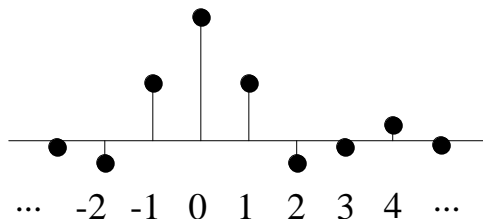
Sequence	Discrete-Time Fourier Transform
$\delta[n]$ 	1 
$1, (-\infty < n < \infty)$ 	$\sum_{k=-\infty}^{+\infty} \delta(f + k)$ 

Commonly Used DTFT Pairs

Sequence	Discrete-Time Fourier Transform
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$$h_{LP}[n] = \frac{\sin 2\pi f_c n}{\pi n}, \quad (-\infty < n < \infty) \quad H_{LP}(e^{j2\pi f}) = \begin{cases} 1, & 0 \leq |f| \leq f_c \\ 0, & f_c < |f| < \frac{1}{2} \end{cases}$$



Discrete-time Fourier Transform Theorems

$$u[n] \leftrightarrow U_d(f)$$

$$h[n] \leftrightarrow H_d(f)$$

Linearity ($\alpha, \beta \in \mathbb{C}$): $\alpha u[n] + \beta h[n] \leftrightarrow \alpha U_d(f) + \beta H_d(f)$

Time-reversal: $u[-n] \leftrightarrow U_d^*(f)$

Shift: $u[n - n_0] \leftrightarrow U_d(f) \exp(-j 2\pi n_0 T f)$

$$u[n] \exp(j 2\pi f_0 n T) \leftrightarrow U_d(f - f_0)$$

Convolution: $u[n] * h[n] \leftrightarrow U_d(f) H_d(f)$

Modulation: $u[n] h[n] \leftrightarrow \frac{1}{T} \int_{-1/(2T)}^{1/(2T)} U_d(f') * H_d(f - f') df'$

Correlation: $u[n] \otimes h[n] \leftrightarrow U_d(f) H_d^*(f)$

Derivatives: $-j2\pi nT u[n] \leftrightarrow \frac{d}{df} U_d(f)$

Parseval's theorem: $\sum_{n=-\infty}^{+\infty} u[n] h^*[n] = \frac{1}{T} \int_{-1/(2T)}^{1/(2T)} U_d(f) H_d^*(f) df$

energy preservation: $\sum_{n=-\infty}^{+\infty} |u[n]|^2 = \frac{1}{T} \int_{-1/(2T)}^{1/(2T)} |U_d(f)|^2 df$

symmetry:

index: e: even part
 o: odd part

$$\operatorname{Re}\{u_e[n]\} \leftrightarrow \operatorname{Re}\{U_{d,e}(f)\}$$

$$\operatorname{Re}\{u_o[n]\} \leftrightarrow j \operatorname{Im}\{U_{d,o}(f)\}$$

$$j \operatorname{Im}\{u_e[n]\} \leftrightarrow j \operatorname{Im}\{U_{d,e}(f)\}$$

$$j \operatorname{Im}\{u_o[n]\} \leftrightarrow \operatorname{Re}\{U_{d,o}(f)\}$$

real-valued signals:

$$u[n] \in \mathbb{R} \quad \leftrightarrow \quad U_d(f) = \operatorname{Re}\{U_d(f)\} + j \operatorname{Im}\{U_d(f)\} = |U_d(f)| \exp(j\phi_{U_d}(f))$$

$$U_d(-f) = U_d^*(f)$$

$$\operatorname{Re}\{U_d(f)\}: \text{even} \qquad |U_d(f)| : \text{even}$$

$$\operatorname{Im}\{U_d(f)\}: \text{odd} \qquad \phi_{U_d}(f) : \text{odd}$$

Discrete Fourier Transform (DFT)

$$U[k] = \sum_{n=0}^{N-1} u[n] \exp\left(-j \frac{2\pi}{N} n k\right)$$

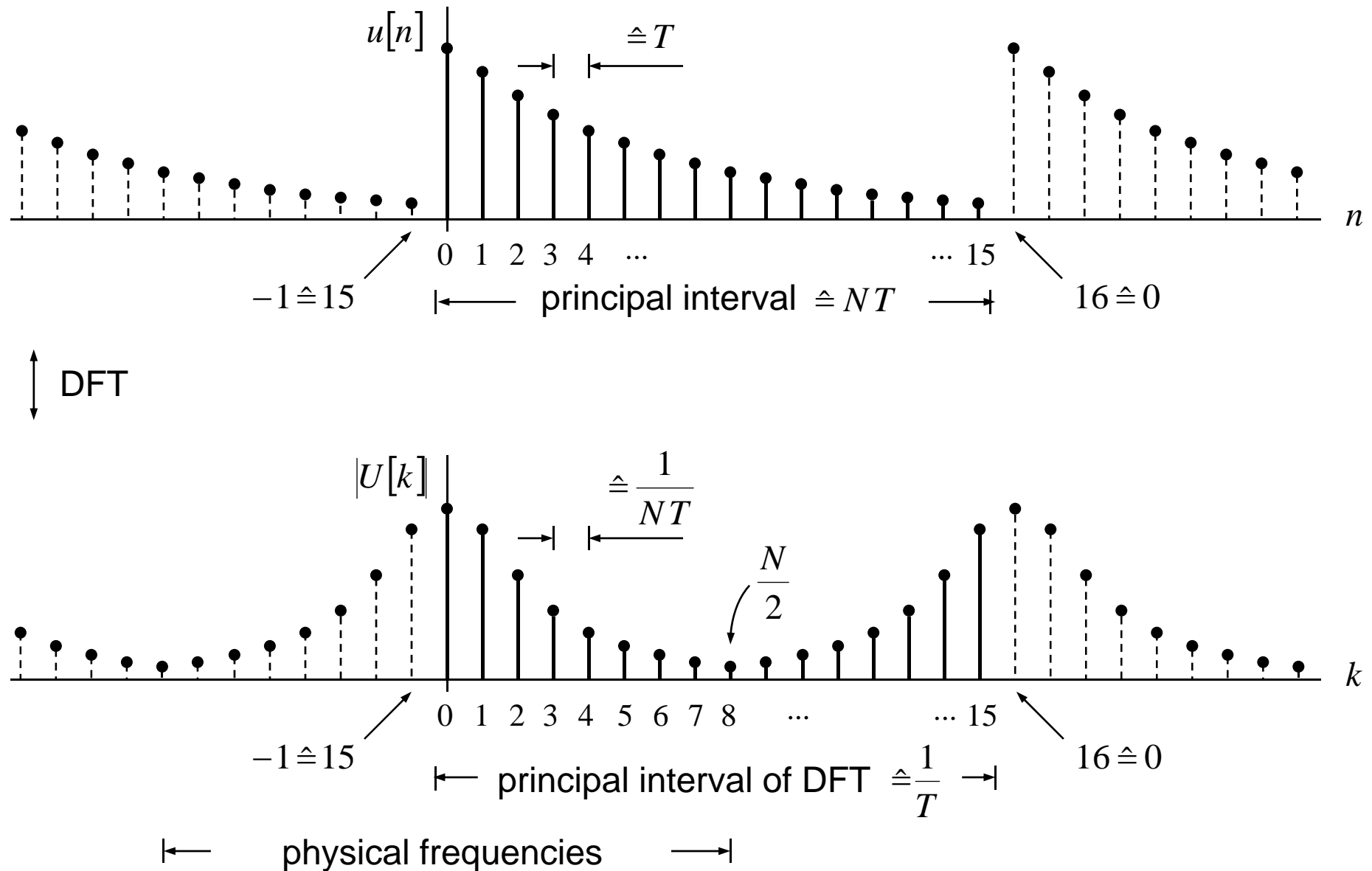
$$u[n] = \frac{1}{N} \sum_{k=0}^{N-1} U[k] \exp\left(j \frac{2\pi}{N} n k\right)$$

⇒ Both signal and spectrum consist of a finite number of samples (“Finite Signals”)

Parseval's Theorem:

$$\sum_{n=0}^{N-1} |u[n]|^2 = \frac{1}{N} \sum_{k=0}^{N-1} |U[k]|^2$$

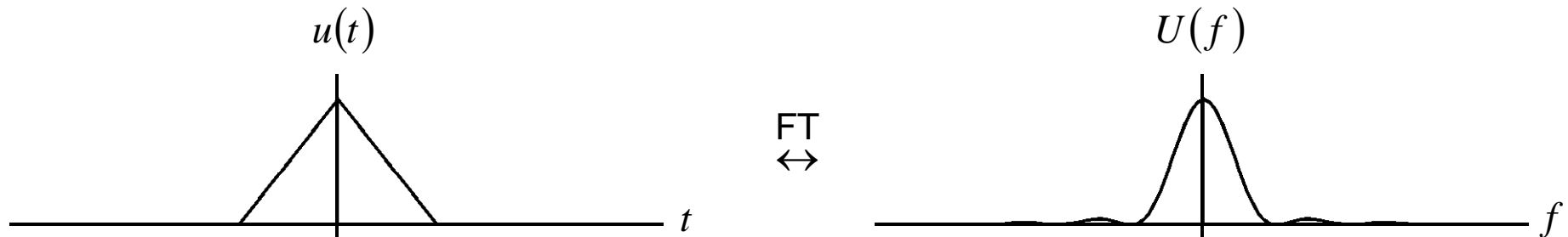
Example DFT with $N = 16$



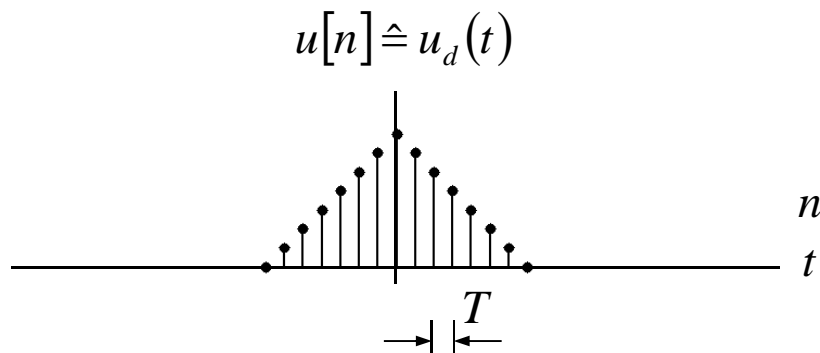
Summary:

discrete-time, discrete-frequency and finite signals

continuous signal (energy-limited):

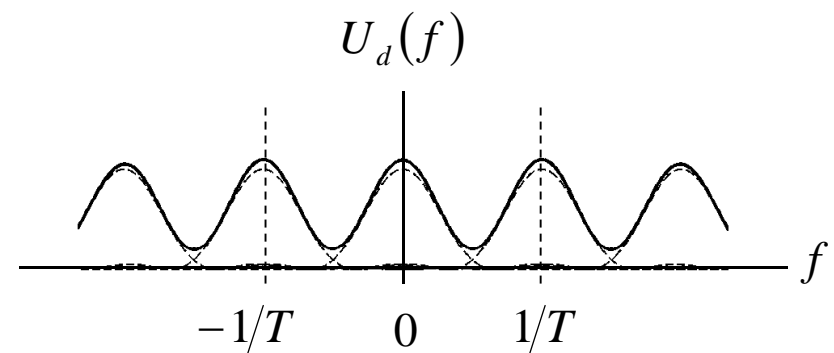


discrete-time signal

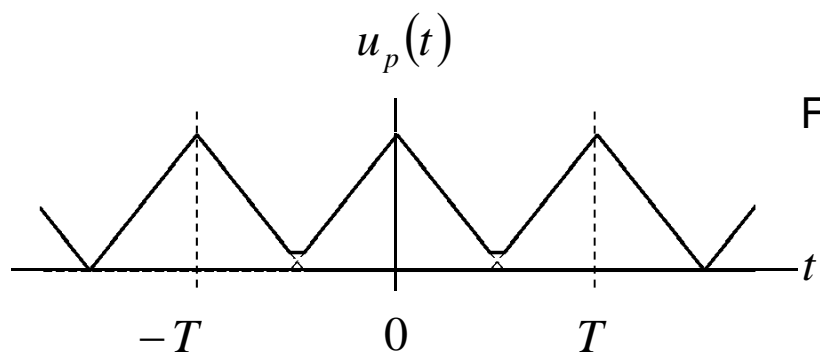


DTFT
 \leftrightarrow

periodic spectrum

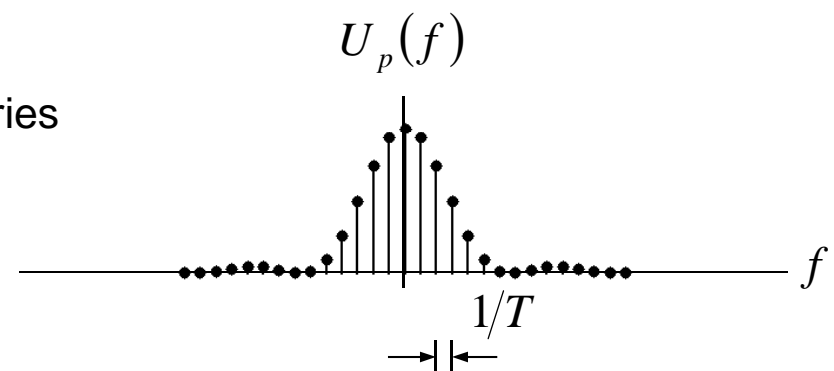


periodic signal

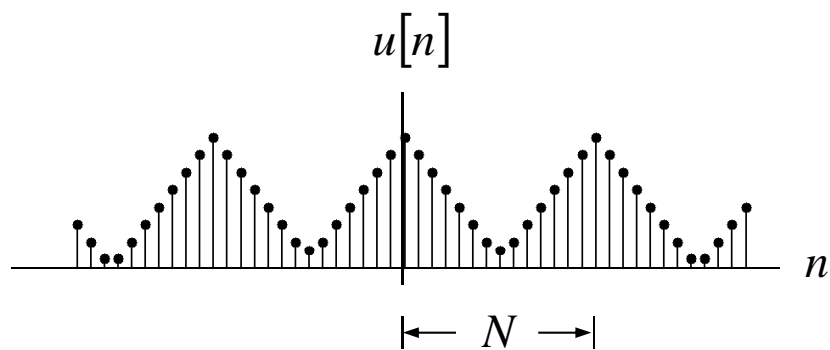


Fourier Series
 \leftrightarrow

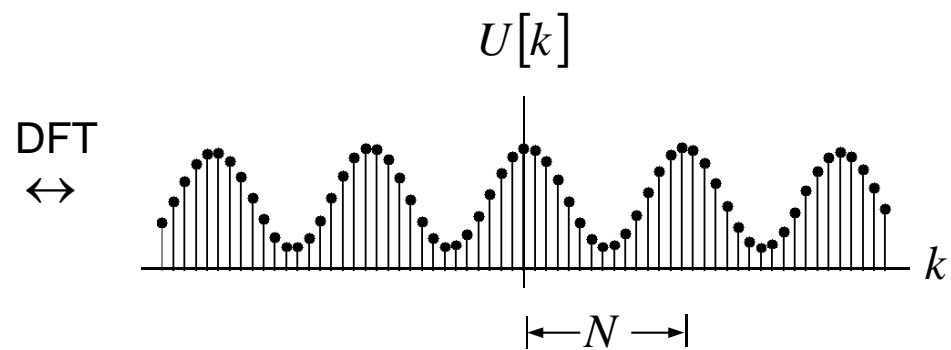
discrete-frequency spectrum



finite signal



finite spectrum



DFT
 \leftrightarrow