# Spectral analysis

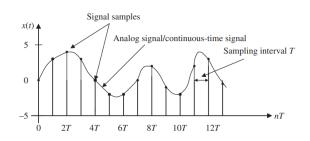
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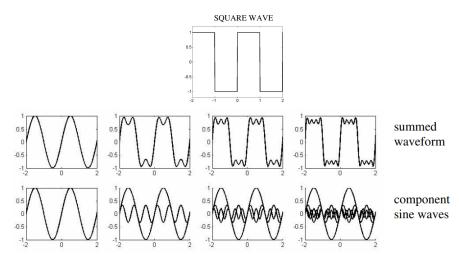
#### Motivation

- We are monitoring **continuous** signal.
- We are saving **discrete** samples of this signal in our computer.
- Under what conditions, and how, can we recover the continuous signal x(t) from the saved samples?



#### The Fourier Transform

The Fourier transform allows to represent a continuous signal as a linear combination of sinusoids.



#### The Fourier Transform

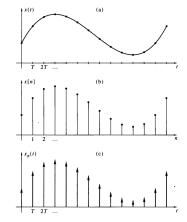
$$x(t) = \frac{1}{2\pi} \int_{-\infty}^{\infty} X(j\Omega) e^{j\Omega t} \ d\Omega = \frac{1}{2\pi} \int_{-\infty}^{\infty} X(j\Omega) (\cos(\Omega t) + j\sin(\Omega t)) \ d\Omega$$
$$= \frac{1}{2\pi} \int_{-\infty}^{\infty} X(j\Omega) \cos(\Omega t) d\Omega + j\frac{1}{2\pi} \int_{-\infty}^{\infty} X(j\Omega) \sin(\Omega t) \ d\Omega$$

where

$$x(j\Omega) = \int_{-\infty}^{\infty} x(t)e^{-j\Omega t} dt$$

# Sampled signal

We sample values of a continuous function x(t) at regular times nT, where T is called the **sampling interval** and its inverse 1/T is called the **sampling frequency**.



**Figure 3.1** The sampling operation: (a) the continuous-time signal x(t); (b) the point-sampled sequence x[n]; (c) the impulse-sampled signal  $x_n(t)$ .

# Continuous representation of a sampled signal

• Dirac delta function  $(\delta(t))$ :

$$\delta(t-t_0) = 0 \quad t \neq t_0$$
 
$$\int_{t_1}^{t_2} f(t) \delta(t-t_0) dt = f(t_0) \quad t_1 < t_0 < t_2$$

 Thus, a sampled signal (with sample interval T) can be represented with the function:

$$x_c(t) = \sum_{n=-\infty}^{\infty} x(t) \, \delta(t - nT)$$

- $x_c(t) = 0$   $t \neq nT$   $\int_{-T}^{nT+\Delta} x_c(t) dt = x(nT)$

### Sampling theorem

#### Theorem (Sampling theorem (Shannon, 1948))

Let  $x_s(t)$  be a sampled signal, with sampling period T, of a continuous signal x(t).

Let  $X(j\Omega)$  be the Fourier transform of x(t).

Then the Fourier transform of  $x_s(t)$  is  $X_s(j\Omega) = \frac{1}{T} \sum_{k=-\infty}^{\infty} X(j(\Omega - \frac{2\pi k}{T}))$ .

# Sampling theorem

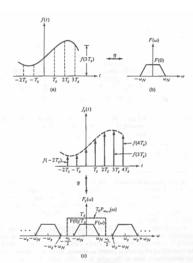


Figure 6.4 Illustrations of the Sampling Theorem

$$\omega_s = \frac{2\pi}{T_s}$$

Poularikas and Seely, 1994

### Reconstruction procedure

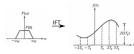
**1** Fourier transform the sampled signal  $x_s(t)$ , yielding  $X_s(j\Omega)$ .



② low-pass filter  $X_s(j\Omega)$  between frequencies  $\Omega = -\frac{\pi}{T}$  and  $\Omega = \frac{\pi}{T}$ , yielding  $X_s^{LP}(j\Omega)$ .

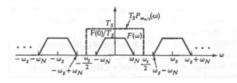


**1** the desired continuous signal is the inverse Fourier transform of  $X_s^{LP}(j\Omega)$  (i.e.,  $x(t) = \mathcal{IFT}\{X_s^{LP}(j\Omega)\}\$ ).

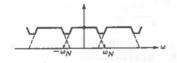


# Reconstruction requirements

we want



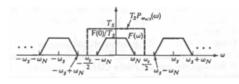
we do NOT want



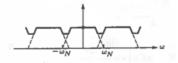
• we want  $\frac{\omega_s}{2} > \omega_N$ , or  $\frac{2\pi}{T} \frac{1}{2} = \frac{\omega_s}{2} > \omega_N = 2\pi f_N$ , or  $\frac{1}{T} \frac{1}{2} > f_N$ , or  $f_s = \frac{1}{T} > 2f_N$ , or  $f_s > 2f_N$ .

# Reconstruction requirements

we want



we do NOT want.



• we want  $\frac{\omega_s}{2} > \omega_N$ , or  $\frac{2\pi}{T} \frac{1}{2} = \frac{\omega_s}{2} > \omega_N = 2\pi f_N$ , or  $\frac{1}{T} \frac{1}{2} > f_N$ , or  $f_s = \frac{1}{T} > 2f_N$ , or  $f_s > 2f_N$ .

To reconstruct without error a continuous signal from its samples we need to sample it at a frequency larger than twice its maximal frequency.

### Reconstruction example

continuous signal in time domain

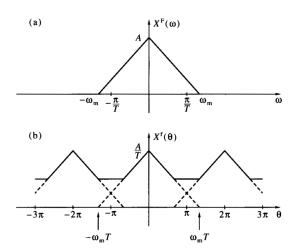
$$egin{aligned} x(t) &= \left( ext{sinc}(f_0 t) 
ight)^2 \ ext{sinc}(t) &= \left\{ egin{array}{c} rac{ ext{sin}(\pi t)}{\pi t} & t 
eq 0 \ 1 & t = 0 \end{array} 
ight. \end{aligned}$$

continuous signal in frequency domain

$$x(j\Omega) = \begin{cases} \frac{1}{f_0} \left( 1 - \frac{|\Omega|}{2\pi f_0} \right) & \Omega \le 2\pi f_0 \\ 0 & \Omega > \pi f_0 \end{cases}$$

 $\omega_N = 2\pi f_0$ ,  $f_N = f_0$  and by the sampling theorem  $f_s > 2f_0$ .

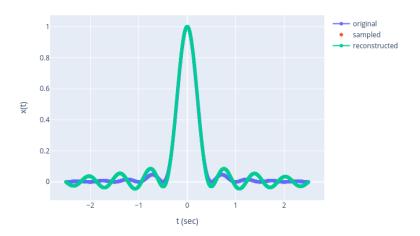
### Reconstruction example: sampling below the Nyquist rate



**Figure 3.4** Sampling of a band-limited signal below the Nyquist rate: (a) Fourier transform of the continuous-time signal; (b) Fourier transform of the sampled signal.

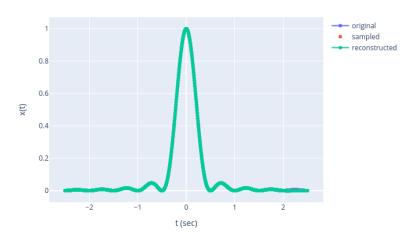
# Reconstruction example: sampling below the Nyquist rate

fs=0.8\*Nyquist rate



# Reconstruction example: sampling at the Nyquist rate

fs=1.0\*Nyquist rate



# Summary

# **Bibliography**

Shannon, C. E. (1948). A mathematical theory of communication. *The Bell system technical journal*, 27(3):379–423.