



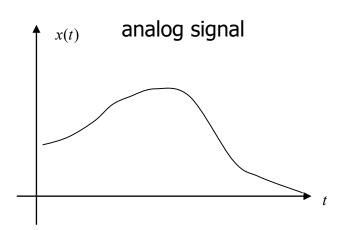
Discrete-Time Signals

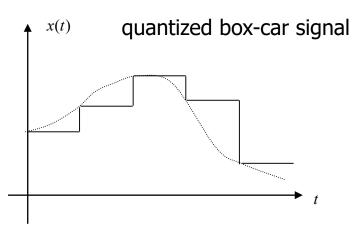


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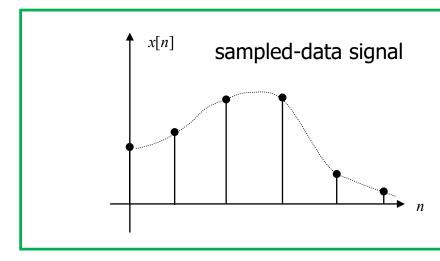


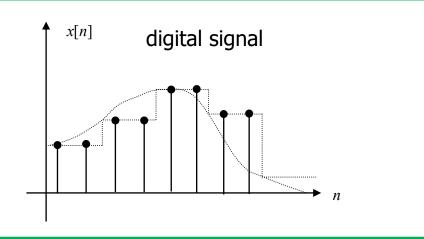
4.1 Overview





Discrete-Time Signals are suited for computer-based processing

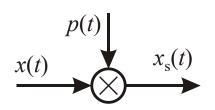






4.2 Sampling and Reconstruction

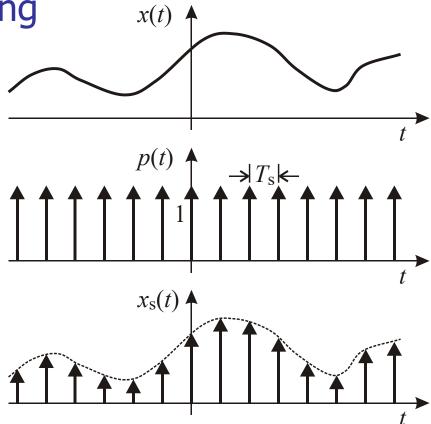
4.2.1 Impulse-Train Sampling



$$p(t) = \sum_{n = -\infty}^{\infty} \delta(t - nT_{\rm S})$$

$$x_{s}(t) = x(t) \cdot p(t)$$

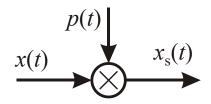
$$= \sum_{n = -\infty}^{\infty} x(nT_{s}) \cdot \delta(t - nT_{s})$$



 $T_{\rm S}$: sampling period



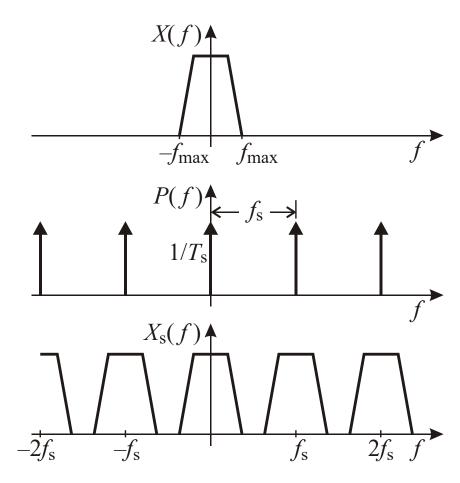
Frequency domain



$$P(f) = \frac{1}{T_{\rm s}} \sum_{k=-\infty}^{\infty} \delta(f - k \cdot f_{\rm s})$$

$$X_{s}(f) = X(f) * P(f)$$

$$= \frac{1}{T_{s}} \sum_{k=-\infty}^{\infty} X(f - k \cdot f_{s})$$



 $f_{\rm S} = 1/T_{\rm S}$: sampling frequency



Sampling theorem

If x(t) is a band-limited signal with $X(\omega) = 0$ for $|f| > f_{\text{max}}$, then x(t) is uniquely determined by its samples $x(nT_{\text{S}})$, if $f_{\text{S}} > 2f_{\text{max}}$, where $f_{\text{S}} = 1/T_{\text{S}}$.

NYQUIST rate

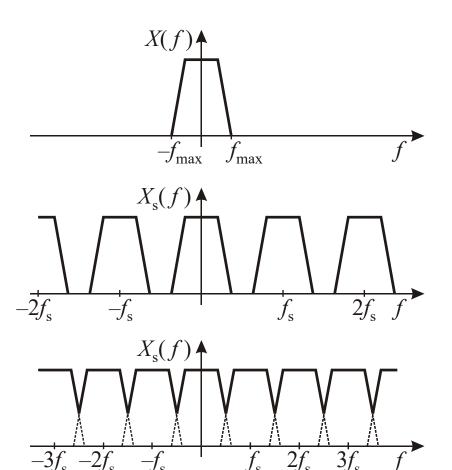
The frequency $2f_{\text{max}}$ is referred to as NYQUIST rate. This is the minimal required sampling rate.



HARRY NYQUIST 1889 - 1976



Undersampling vs. oversampling



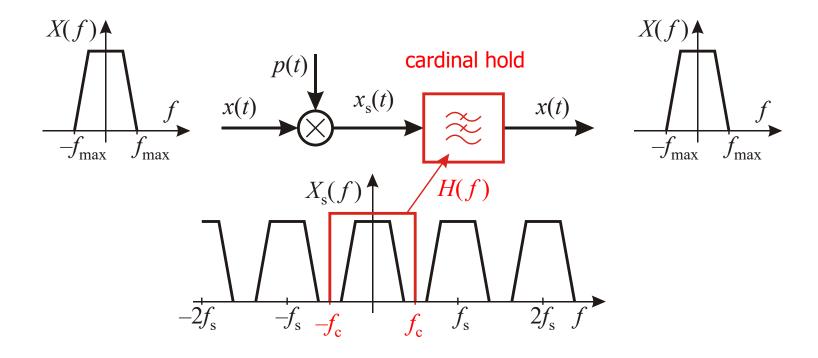
Oversampling: $f_S > 2f_{\text{max}}$

Undersampling: $f_S < 2f_{max}$ \rightarrow aliasing error



Reconstruction

Given an oversampled signal with $f_{\rm S} > 2f_{\rm max}$ the original signal x(t) can be reconstructed by an ideal low-pass filter with gain $T_{\rm S}$ and cutoff frequency greater than $f_{\rm max}$ and less than $f_{\rm S} - f_{\rm max}$.

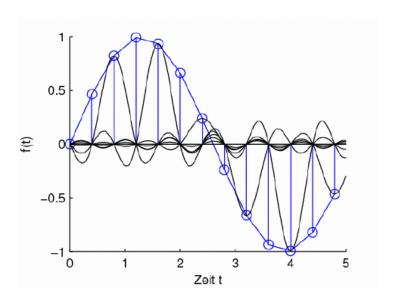


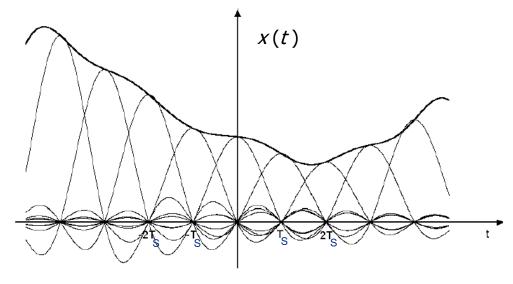


Reconstruction

$$X(f) = X_{s}(f) \cdot H(f) \qquad H(f) = T_{s} \cdot \text{rect}\left(\frac{f}{f_{s}}\right) \quad \bullet - \circ \quad h(t) = \text{si}\left(\pi f_{s}t\right)$$

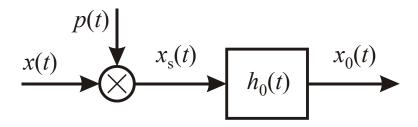
$$x(t) = x_{s}(t) * h(t) = \sum_{n = -\infty}^{\infty} x(nT_{s}) \cdot \text{si}\left[\pi f_{s}(t - nT_{s})\right]$$







4.2.2 Sampling with a Zero-Order Hold

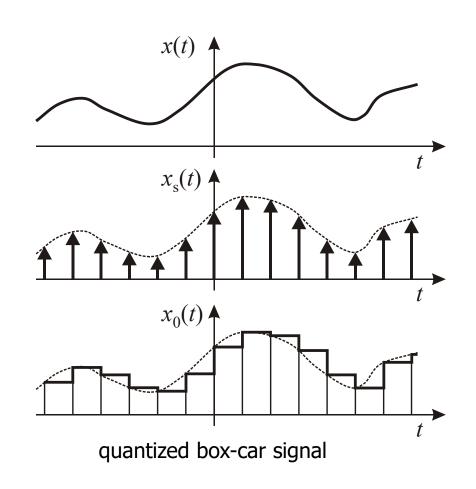


box-car circuit, sample- and-hold circuit, data clamp

$$h_0(t) = \text{rect}\left(\frac{t - T_s / 2}{T_s}\right)$$

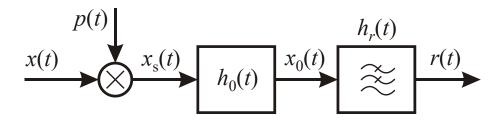
$$x_0(t) = x_s(t) * h_0(t) =$$

$$= \sum_{n = -\infty}^{\infty} x(nT_s) \cdot h_0(t - nT_s)$$



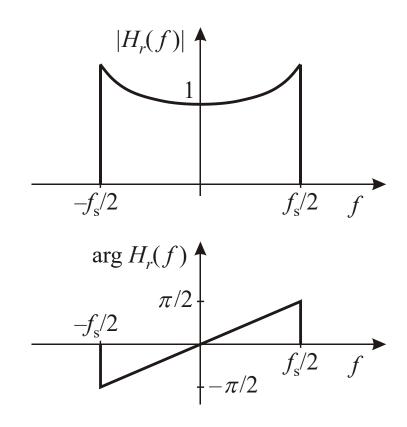


Reconstruction



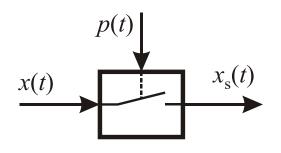
$$H_r(f) = \frac{H(f) \cdot e^{j\omega T_s/2}}{T_s \cdot si(\pi f T_s)}$$

In many situations the output $x_0(t)$ of the zero-order hold can be considered an adequate approximation of the original signal x(t) by itself.



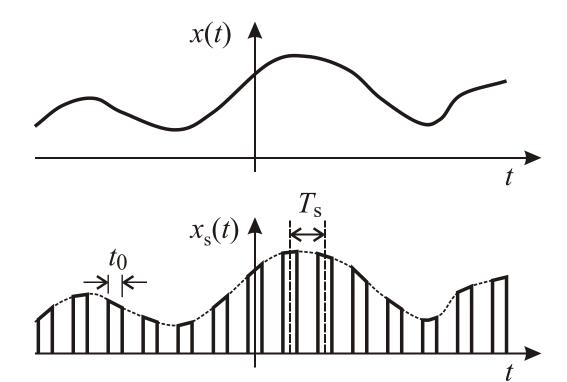


4.2.3 Sampling with a Linear Gate



$$x_{\rm s}(t) = x(t) \cdot p(t)$$

$$p(t) = \sum_{n=-\infty}^{\infty} \text{rect}\left(\frac{t - nT_{s}}{t_{0}}\right)$$





4.3 Signal Notation and Properties

Discrete signal

$$x[n] = x_{cont}(t = n \cdot T_s)$$
 T_s : sampling period; n : integer $\{x[n]\} = \{..., x[-1], x[0], x[1], x[2],...\} = \{..., 0.9, 0.1, 2.2, 3.7, ...\}$

- sequence of data
- MATLAB: n = 1, ..., N

$$\vec{x} = [x[0], x[1], x[2], ..., x[N-1]]^{T}$$

vector representation

Digital signal

$$\{\hat{x}[n]\} = \{..., \hat{x}[-1], \hat{x}[0], \hat{x}[1], \hat{x}[2],...\} = \{..., 1, 0, 2, 4, ...\}$$



Finite length sequence

$$n_1 \le n \le n_2$$
; $N = n_2 - n_1 + 1$; N : length, duration

Causal signal

$$x[n] = 0$$
 for $n < N_1$ with $N_1 \ge 0$

Periodic sequence

$$x[n] = x[n \pm k \cdot N]$$
; $k = 1, 2, 3, ...$; N : period



Signal energy

$$E_{\infty} = \lim_{N \to \infty} \sum_{n=-N}^{N} |x[n]|^2 = \sum_{n=-\infty}^{\infty} |x[n]|^2$$

Signal power

$$P_{\infty} = \lim_{N \to \infty} \frac{1}{2N+1} \sum_{n=-N}^{N} |x[n]|^2$$

Energy signal: $E_{\infty} < \infty$, $P_{\infty} = 0$

Power signal: $P_{\infty} < \infty$, $E_{\infty} \rightarrow \infty$

Strength of a signal: The signal strength is given by its norm.

$$||x||_p = \left(\sum_{n=-\infty}^{\infty} |x[n]|^p\right)^{1/p}$$
; $p = 1, 2, ...$



Symmetry Relations

- Conjugate symmetric sequence: $x_{CS}[n] = x_{CS}^*[-n]$
 - Real sequence \rightarrow even: $x_e[n] = x_e[-n]$
- Conjugate antisymmetric sequence: $x_{ca}[n] = -x_{ca}^*[-n]$
 - Real sequence \rightarrow odd: $x_0[n] = -x_0[-n]$
- Any complex sequence x[n] can be composed as:

$$x[n] = x_{cs}[n] + x_{ca}[n]$$

$$x_{cs}[n] = \frac{x[n] + x^*[-n]}{2}$$
, $x_{ca}[n] = \frac{x[n] - x^*[-n]}{2}$

$$x_{cs}[0] = real ; x_{ca}[0] = imaginary ; x_{o}[0] = 0$$



4.4 Test Sequences

Unit sample sequence

$$\delta[n] = \begin{cases} 1 & , & n = 0 \\ 0 & , & n \neq 0 \end{cases}$$

Unit step sequence

$$u[n] = \begin{cases} 1, & n \ge 0 \\ 0, & n < 0 \end{cases}$$

Rectangular sequence

$$\operatorname{rect}_{N}[n] = \begin{cases} 1 & \text{for } |n| \le (N-1)/2 \\ 0 & \text{elsewhere} \end{cases}; N \text{ odd}$$

Sinusoidal sequence

$$x[n] = \cos(2\pi f \cdot n \cdot T_s) = \cos(n \cdot \Omega)$$

 $\Omega = 2\pi f \cdot T_s = 2\pi f / f_s$ Ω : normalized angular frequency

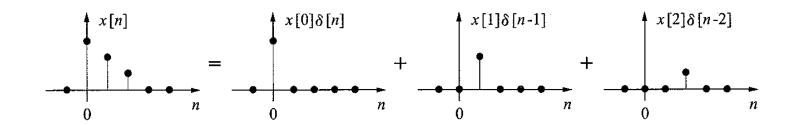


4.5 Signal Operations

4.5.1 Elementary Operations

Superposition

$$x[n] = \sum_{i=-\infty}^{\infty} x[i] \delta[n-i]$$



Sifting equation

$$x[n]*\delta[n] = x[n] ; x[n]*\delta[n-k] = x[n-k]$$



4.5.2 Convolution

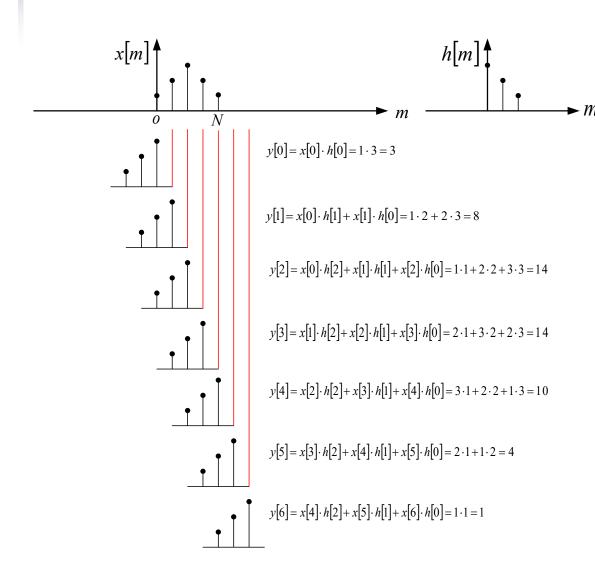
Generic

$$y[n] = \sum_{k=-\infty}^{\infty} x_1[k]x_2[n-k] = x_1[n] * x_2[n]$$

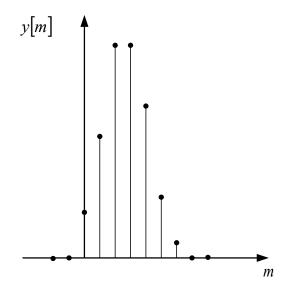
Convolution of periodic sequences

$$y[n] = \frac{1}{N} \sum_{k=0}^{N-1} x_1[k] \cdot x_2[n-k] = \frac{1}{N} \sum_{k=0}^{N-1} x_2[k] \cdot x_1[n-k]$$





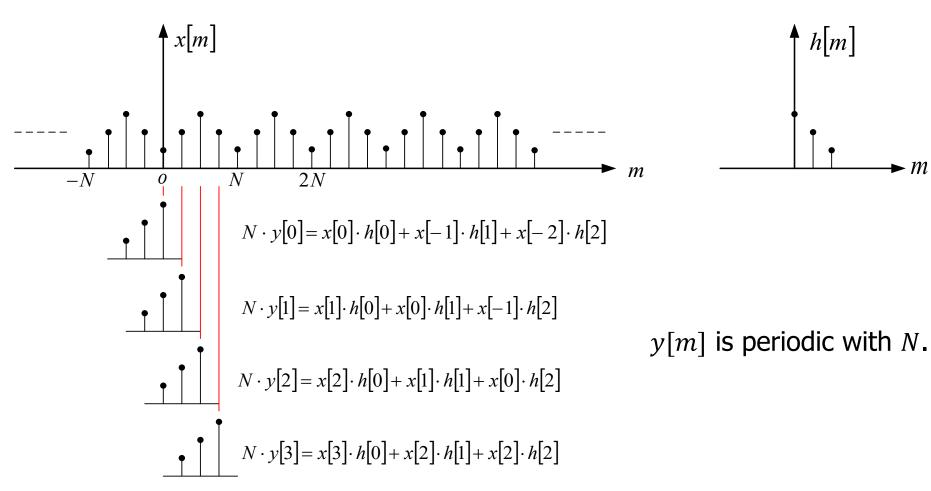
$$y[m] = x[m] * h[m]$$



$$Length = N_1 + N_2 - 1$$



$$y[m] = x[m] * h[m]$$

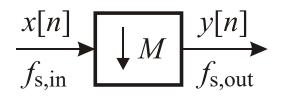




4.5.3 Sampling Rate Alteration

Down-sampler, sampling rate compressor

$$y[n] = x[n \cdot M]$$
; M : positive integer $f_{s,out} = f_{s,in} / M$



M-1 consecutive samples are removed from the input sequence.

Up-sampler, sampling rate expander

$$y[n] = \begin{cases} x[n/L] &, n = 0, \pm L, \pm 2L, \dots \\ \text{interpolation} &, \text{ otherwise} \end{cases}$$
; L:positive integer

$$f_{\text{s,out}} = f_{\text{s,in}} \cdot L$$

L-1 new samples are inserted between two consecutive samples of the input sequence.

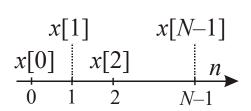
$$\frac{x[n]}{f_{s,in}} \qquad \uparrow L \qquad \frac{y[n]}{f_{s,out}}$$

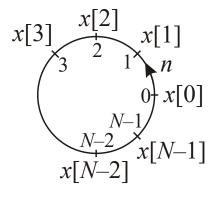


4.5.4 Operations on Finite-Length Sequences

Length-N sequence

$$x[n], 0 \le n \le N-1$$

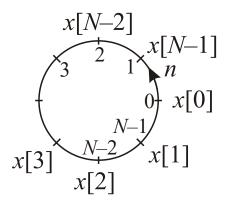




Circular time reversal

$$y[n] = x[\langle -n \rangle_N]$$

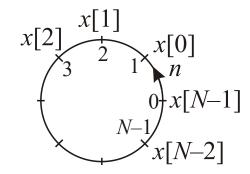
 $\langle m \rangle_N = m \text{ modulo } N = r \text{ : residue}$
 $r = m + l \cdot N \text{ ; } 0 \le r \le N - 1$





Circular shift

$$x_{\rm C}[n] = x[\langle n - n_0 \rangle_N]$$



$$x_{\mathbf{C}}[n] = x[\langle n-1 \rangle_N]$$



Symmetry Relations of Finite-Length Sequences

- Circular conjugate symmetric sequence: $x_{CCS}[n] = x_{CCS}^*[\langle -n \rangle_N]$
 - Real sequence \rightarrow circular even: $x_{ce}[n] = x_{ce}[\langle -n \rangle_N]$
- Circular conjugate antisymmetric sequence: $x_{cca}[n] = -x_{cca}^*[\langle -n \rangle_N]$
 - Real sequence \rightarrow circular odd: $x_{CO}[n] = -x_{CO}[\langle -n \rangle_N]$
- Any finite-length complex sequence x[n] can be composed as:

$$x[n] = x_{\text{ccs}}[n] + x_{\text{cca}}[n] ; 0 \le n \le N-1$$

$$x_{\text{ccs}}[n] = \frac{x[n] + x^*[\langle -n \rangle_N]}{2} , \quad x_{\text{cca}}[n] = \frac{x[n] - x^*[\langle -n \rangle_N]}{2}$$

$$x_{ccs}[0] = real$$
; $x_{cca}[0] = imaginary$

N even:
$$x_{ccs}[N/2] = real$$
; $x_{cca}[N/2] = imaginary$



Convolution

$$x_1[n], x_2[n], 0 \le n \le N-1$$

Linear convolution

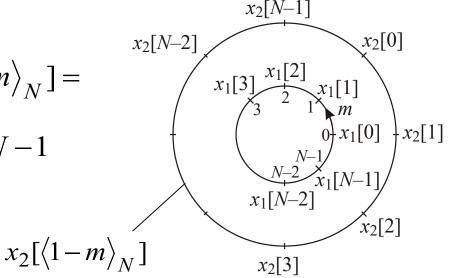
$$y_{L}[n] = \sum_{m=0}^{N-1} x_{1}[m]x_{2}[n-m] = x_{1}[n] * x_{2}[n] ; 0 \le n \le 2N-2$$

Both length-N sequences are zero-padded to length 2N-1.

Circular convolution

$$y_{C}[n] = \sum_{m=0}^{N-1} x_{1}[m]x_{2}[\langle n-m \rangle_{N}] =$$

$$= x_{1}[n] * x_{2}[n] ; \quad 0 \le n \le N-1$$





4.6 Discrete-Time Fourier Transform (DTFT)

• FOURIER transform of x[n]

$$X(e^{j\Omega}) = \sum_{n=-\infty}^{+\infty} x[n] \cdot e^{-jn\Omega} \; ; \quad \Omega = \omega \cdot T_s = 2\pi f / f_s$$

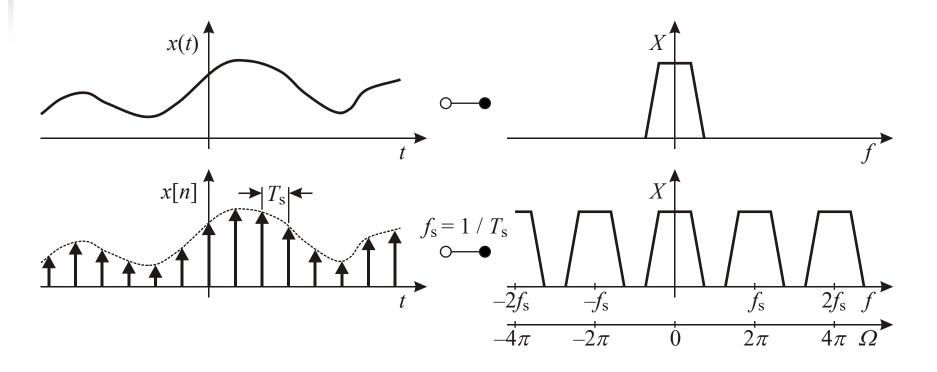
$$\Omega : \text{ normalized angular frequency}$$

$$x[n] \circ - \bullet \; X(e^{j\Omega}) = \text{DTFT}\{x[n]\}$$

■ Inverse Fourier transform of $X(e^{j\Omega})$

$$x[n] = \frac{1}{2\pi} \cdot \int_{2\pi} X(e^{j\Omega}) \cdot e^{jn\Omega} d\Omega$$
$$X(e^{j\Omega}) \bullet - \circ x[n] = IDTFT \{X(e^{j\Omega})\}$$





- The Fourier transform of the discrete signal x[n] is periodic with $\Delta f = f_{\rm S}$ or $\Delta \Omega = 2\pi$, respectively.
- The FOURIER transform of most practical sequences can be expressed in terms of a sum of a convergent geometric series, which can be summed in a simple closed form.



Some Properties of the Fourier transform

$$x[n] \circ - \bullet \ X(e^{j\Omega})$$
 $y[n] \circ - \bullet \ Y(e^{j\Omega})$

$$y[n] \circ - \bullet Y(e^{j\Omega})$$

Time shifting

$$x[n-n_0] \circ - \bullet e^{-jn_0\Omega} \cdot X(e^{j\Omega})$$

Frequency shifting

$$x[n] \cdot e^{jn\Omega_0} \circ - \bullet X(e^{j(\Omega - \Omega_0)})$$

Multiplication

$$x[n] \cdot y[n] \circ - \bullet \frac{1}{2\pi} \cdot X(e^{j\Omega}) * Y(e^{j\Omega})$$

Convolution

$$x[n] * y[n] \circ - \bullet X(e^{j\Omega}) \cdot Y(e^{j\Omega})$$



4.7 Discrete and Fast Fourier Transform (DFT, FFT)

- DFT and FFT are orthogonal finite-length transforms of length N
- FOURIER transform (DFT, FFT) of x[n]

$$X[k] = \sum_{n=0}^{N-1} x[n] \cdot e^{-j2\pi kn/N} = \sum_{n=0}^{N-1} x[n] \cdot W_N^{kn} \quad ; \quad 0 \le k \le N-1 \qquad x[n] \circ - \bullet \quad X[k]$$

• Inverse Fourier transform (IDFT, IFFT) of X[k]

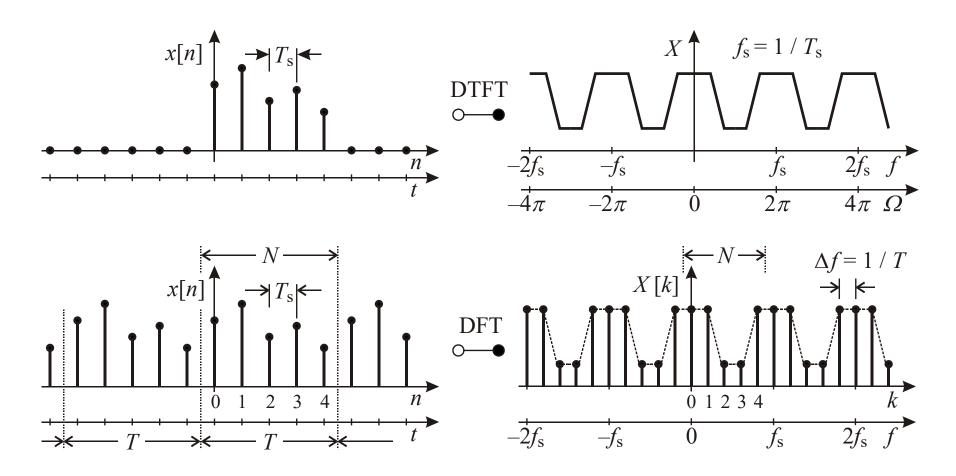
$$x[n] = \frac{1}{N} \cdot \sum_{k=0}^{N-1} X[k] \cdot e^{j2\pi kn/N} = \frac{1}{N} \cdot \sum_{k=0}^{N-1} X[k] \cdot W_N^{-kn} \; ; \quad 0 \le n \le N-1$$
$$X[k] \bullet - \circ \; x[n]$$

Relation:

$$W_N = e^{-j2\pi/N}$$
 ; $N = \frac{T}{T_s} = \frac{f_s}{\Delta f}$



DTFT versus DFT, FFT





Temporal and Spectral Resolution

- The sample period T_S and the sampling frequency f_S , respectively, determine the **temporal resolution**.
- The number of samples N together with the sampling frequency f_S determine the observation period T and thus the **spectral** resolution Δf .



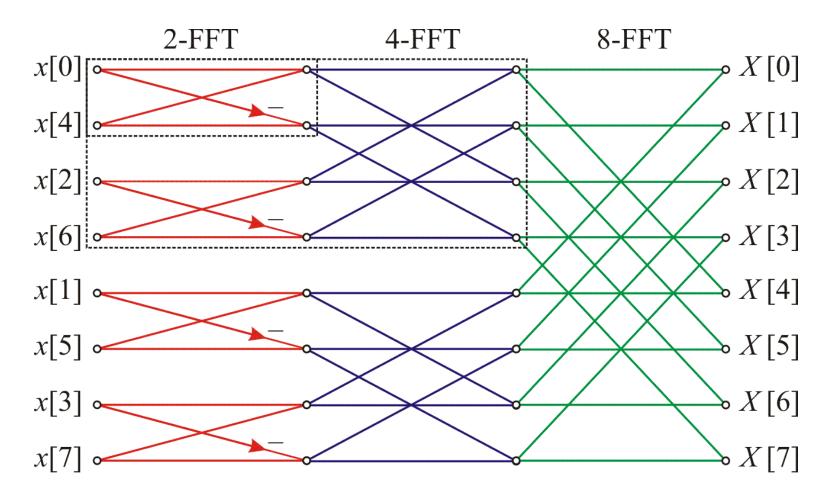
Computational Complexity Issues

- DFT requires N^2 complex multiplications and N(N-1) complex additions. This results in $N \cdot (2N-1)$ complex operations.
- FFT requires $N = 2^n$. This results in $N \cdot ldN = n \cdot 2^n$ complex operations.
- Complex operations:

N	DFT	FFT
4	28	8
8	120	24
16	496	64
32	2,016	160
64	8,128	384
128	32,640	896
256	130,816	2,048



Flow-graph of FFT algorithm





Properties of DFT and FFT

$$x[n], y[n], n = 0, 1, ..., N-1 \circ - \bullet X[k], Y[k], k = 0, 1, ..., N-1$$

Linearity

$$a \cdot x[n] + b \cdot y[n] \circ - \bullet \quad a \cdot X[k] + b \cdot Y[k]$$

• **Symmetry** properties of a complex sequence x[n]

$$x[n] = x_{Re}[n] + j \cdot x_{Im}[n] \circ - \bullet \quad X[k] = X_{Re}[k] + j X_{Im}[k]$$

$$x_{\text{ccs}}[n] \circ - \bullet \operatorname{Re}\{X[k]\}$$
 $x_{\text{cca}}[n] \circ - \bullet \operatorname{j:Im}\{X[k]\}$



Symmetry properties of a real sequence $x[n] = x_{Ce}[n] + x_{CO}[n]$

$$x_{\rm ce}[n] \circ - \bullet \operatorname{Re}\{X[k]\}$$

$$x_{\text{ce}}[n] \circ - \bullet \operatorname{Re}\{X[k]\}$$
 $x_{\text{co}}[n] \circ - \bullet \operatorname{j-Im}\{X[k]\}$

ce: circular even

co: circular odd

 $X[k] = X^* [\langle -k \rangle_{N}]$; circular conjugate symmetric

- ce: Re $\{X[k]\}$ = Re $\{X[\langle -k \rangle_N]\}$; co: Im $\{X[k]\}$ = -Im $\{X[\langle -k \rangle_N]\}$
- $\operatorname{ce}:|X[k]| = |X[\langle -k \rangle_N]; \quad \operatorname{co}: \arg\{X[k]\} = -\arg\{X[\langle -k \rangle_N]\}$

Duality

$$X[n] \circ - \bullet \ N \cdot x \left| \left\langle -k \right\rangle_N \right|$$

PARSEVAL's theorem

$$\sum_{n=0}^{N-1} |x[n]|^2 = \frac{1}{N} \cdot \sum_{n=0}^{N-1} |X[k]|^2$$



Circular time shifting

$$x \lceil \langle n - n_0 \rangle_N \rceil \circ - \bullet e^{-j2\pi k n_0/N} \cdot X[k]$$

Circular frequency shifting

$$e^{j2\pi k_0 n/N} \cdot x[n] \circ - \bullet X \left[\left\langle k - k_0 \right\rangle_N \right]$$

N-point circular convolution

$$\sum_{m=0}^{N-1} x[m] y \left[\left\langle n - m \right\rangle_N \right] \circ - \bullet \ X[k] \cdot Y[k]$$

Multiplication

$$x[n] \cdot y[n] \circ - \bullet \frac{1}{N} \cdot \sum_{m=0}^{N-1} X[m] Y[\langle k - m \rangle_N]$$



Leakage Effect

Periodic signal:

$$x(t) = e^{j2\pi f_0 t}$$
 $\circ - \bullet$ $X(f) = \delta(f - f_0)$

Discrete length-N signal:

$$x[n] = e^{j2\pi f_0 nT_s}$$
; $0 \le n \le N-1$; $T = N \cdot T_s$

$$X[k] = e^{j\pi(f_0NT_s - k)\frac{N-1}{N}} \cdot \frac{\sin(\pi(f_0NT_s - k))}{\sin(\frac{\pi}{N}(f_0NT_s - k))} ; \quad 0 \le k \le N - 1$$

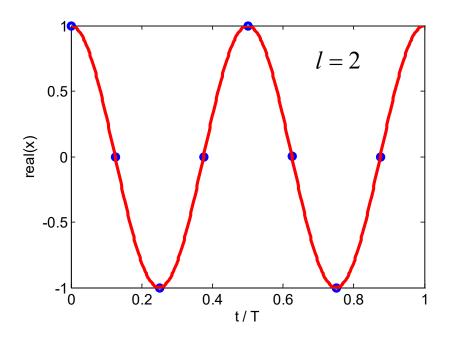


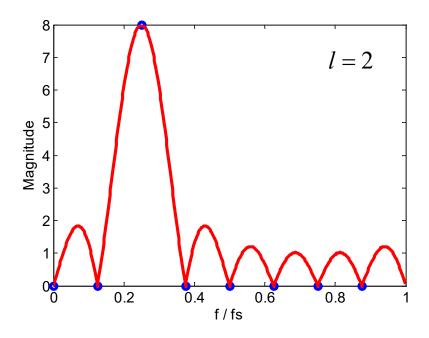
No leakage, if

$$f_0 NT_s = l = f_0 T = \frac{f_0}{\Delta f} = \frac{T}{1/f_0}$$
; l:positive integer

$$X[k] = N \cdot \delta[k-l]$$

$$0 \le k \le N-1$$

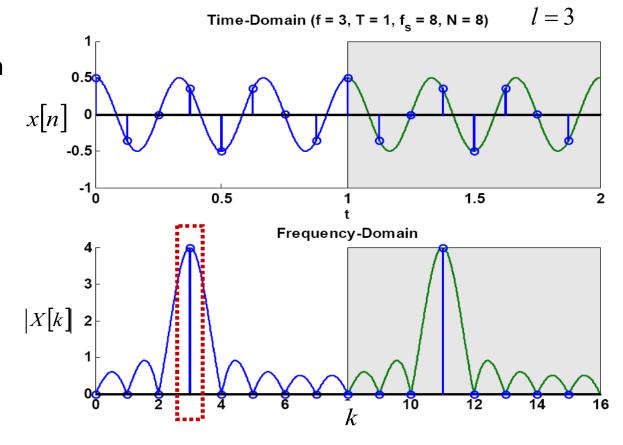






No leakage:

- The periodicity of x(t) is conserved.
- Ideal spectral analysis: one main spectral line

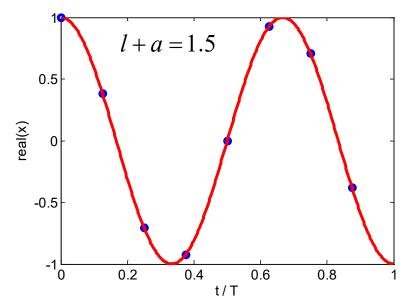


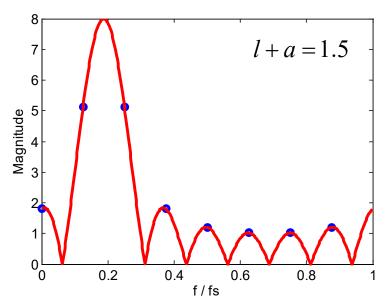


Leakage effect, if

$$f_0 NT_s = l + a = f_0 T = \frac{f_0}{\Delta f} = \frac{T}{1/f_0}$$
; l:pos.integer; $-0.5 < a \le 0.5$

$$X[k] = e^{j\pi(l-k+a)\frac{N-1}{N}} \cdot \frac{\sin(\pi(l-k+a))}{\sin(\frac{\pi}{N}(l-k+a))} ; \quad 0 \le k \le N-1$$





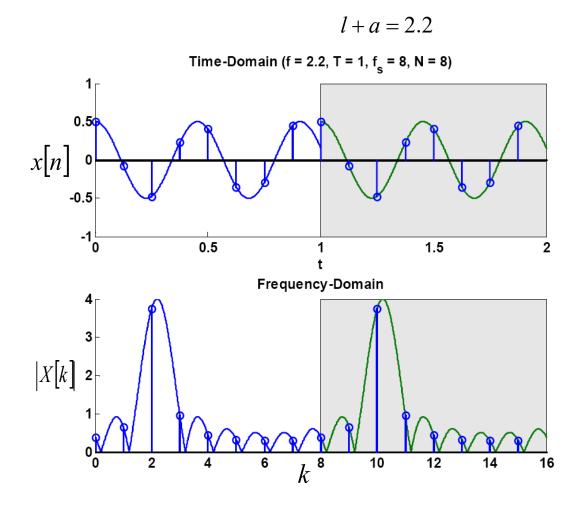
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Prof. Dr.-Ing. Uwe Meier



- The periodicity of x(t) is violated.
- Nonideal spectral analysis: reduced main spectral line + several additional lines

→ leakage (Lattenzaun-Effekt)





Zero Padding

• Increasing the observation time $T = NT_S$ by increasing N improves the spectral resolution Δf .

