

EDISP

(English) Digital Signal Processing

Jacek Misiurewicz

e-mail: jmisiure@elka.pw.edu.pl

Institute of Electronic Systemss
Warsaw University of Technology
Warsaw, Poland

General information

“Credits” 2h/week lecture + 2h/week $\approx 7 \cdot 4h$ **laboratory** exercises in groups ≤ 12 persons.

Next lecture Mon, 10:15-12 \longrightarrow *the lab schedule will be announced then!!*

Textbook A. V. Oppenheim, R. W. Schaffer, *Discrete-Time Signal Processing*, Prentice-Hall 1989 (or II ed, 1999; also acceptable previous editions entitled *Digital Signal Processing*).

Contact J. Misiurewicz, (jmisiure@elka.pw.edu.pl) room 453. A web page is under construction (<http://staff.elka.pw.edu.pl/~jmisiure/>)

Homeworks Announced as a preparation for the labs.

Exams Two short tests within lecture hours (TBA with the lab schedule) and a final exam during the winter exam session (TBA).

Scoring:		2x10%	=	20%	tests
		6x5%	=	30%	lab + homework (lab 0 – not scored)
				50%	final exam

Signal classification

Continuous or Discrete **amplitude** and **time**.

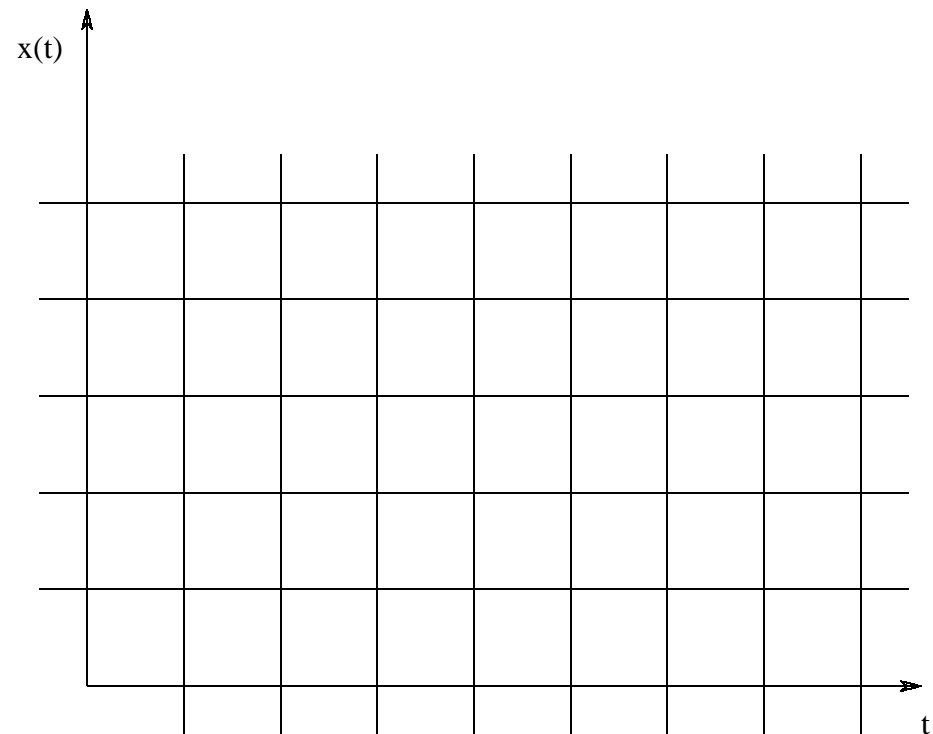
CA-CT → “analog” signals

DA-CT →

CA-DT → CCD, SC, SAW devices

DA-DT → digital devices

We’ll speak mainly about DT properties; only in some subject DA will be of importance.



DT signal representations

DT signal \longleftrightarrow a number sequence

$$x[n] = \{x(n)\}$$

$x[n]$ is a number sequence (or ...)

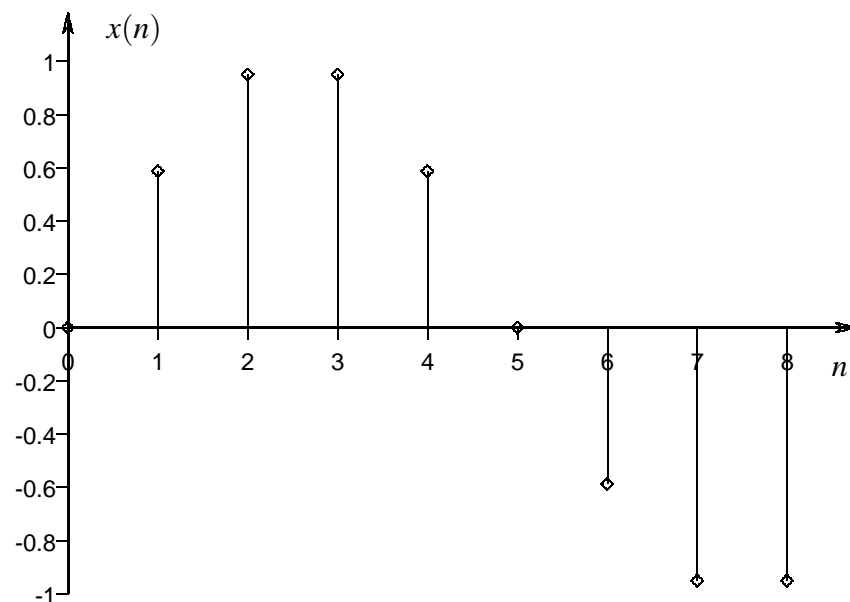
$x(n)$ is a n -th sample

\longrightarrow $x(n)$ is *undefined* for $n \notin I$

- it *may* come from sampling of analog signal
- but it may also be inherently discrete
- n may correspond to: time, space,

...

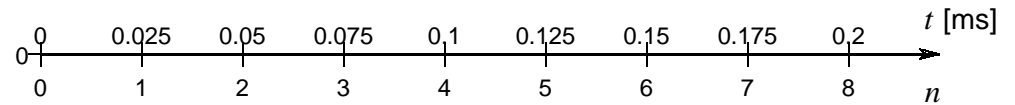
However, the most popular interpretation is: periodic sampling in time.



Periodic sampling

$$n \longleftarrow \longrightarrow n \cdot T_s$$

$$x(n) = x_a(nT_s)$$

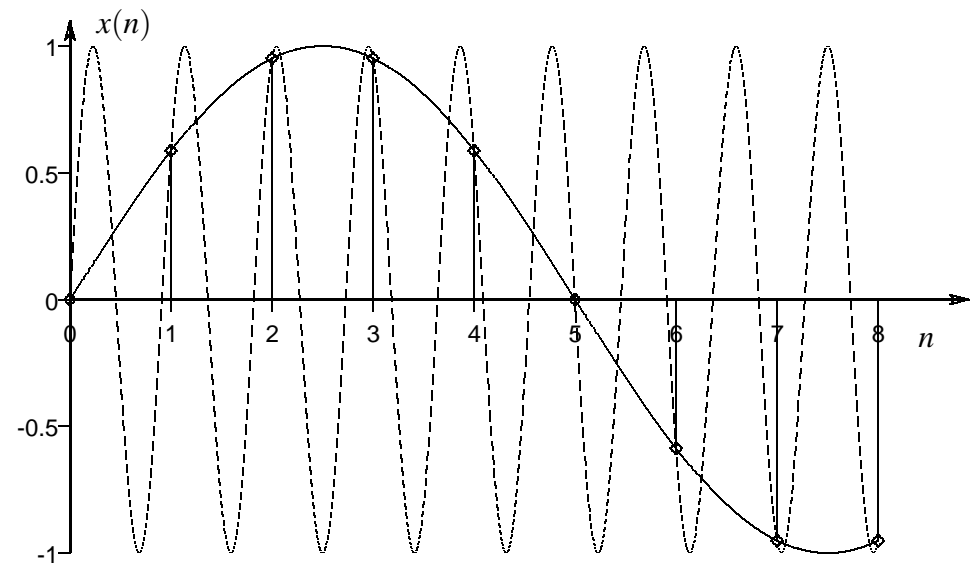


$$n = t / T_s, \quad T_s = 0.025 \text{ [ms]}$$

Misinterpretations

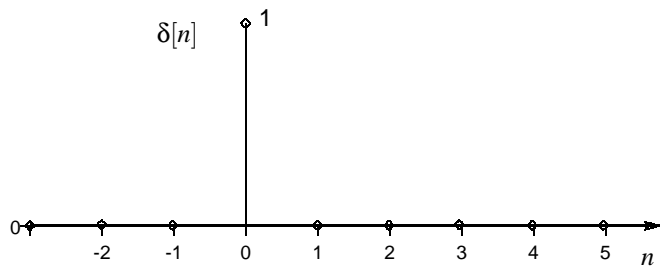
→ we do not know what is between points

- a) $\sin(n \cdot (1/5) \cdot \pi)$ or
- b) $\sin(n \cdot (2 + 1/5) \cdot \pi)$?



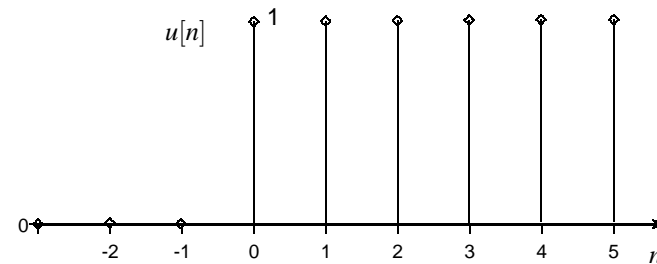
Number sequence (or DT signal) operations; basic sequences

operation	notation	definition
sum	$z[n] = x[n] + y[n]$	$\forall n \ z(n) = x(n) + y(n)$
scale	$z[n] = \alpha \cdot y[n]$	$\forall n \ z(n) = \alpha \cdot y(n)$
shift	$z[n] = x[n - n_0]$	$\forall n \ z(n) = x(n - n_0)$
difference	$z[n] = x[n] - y[n]$	$\forall n \ z(n) = x(n) - y(n)$
product	$z[n] = x[n] \cdot y[n]$	$\forall n \ z(n) = x(n) \cdot y(n)$



Unit sample sequence (DT impulse)

$$\delta[n] = u[n] - u[n - 1]$$



Unit step sequence

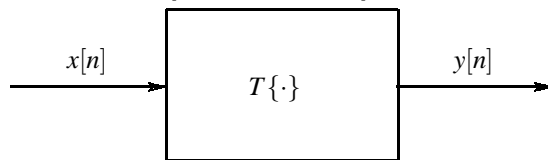
$$u[n] = \sum_{k=0}^{\infty} \delta[n - k]$$

DT systems

A DT system: an operator mapping an input sequence $x[n]$ into an output sequence $y[n]$.

$$y[n] = T\{x[n]\}$$

→ A rule (formula) for computing output sequence values $y(n)$ from the input sequence values $x(n)$.



Implementations:

- PC program
- matlab m-file
- custom VLSI or FPGA
- programmable digital signal processor

Examples:

$$y(n) = 3 \cdot x(n)$$

$$y(n) = \frac{x(n) + x(n-1)}{2}$$

$$y(n) = \frac{1}{M} \sum_{k=0}^{M-1} x(n-k)$$

$$y(n) = \sum_{k=-\infty}^{\infty} h(k) \cdot x(n-k)$$

Linear & time-invariant DT systems

Linearity property

$$T\{\alpha_1 x_1[n] + \alpha_2 x_2[n]\} = \alpha_1 T\{x_1[n]\} + \alpha_2 T\{x_2[n]\}$$

in other words:

if

$$x_1[n] \longrightarrow y_1[n]$$

$$x_2[n] \longrightarrow y_2[n]$$

then

$$\alpha x_1[n] \longrightarrow \alpha y_1[n] \quad (\text{scaling, homogeneity})$$

$$x_1[n] + x_2[n] \longrightarrow y_1[n] + y_2[n] \quad (\text{additivity})$$

Time invariance (shift invariance)

If

$$T\{x[n]\} = y[n]$$

then

$$\forall n_0, T\{x[n - n_0]\} = y[n - n_0]$$

Linear systems - examples

- $y(n) = 3 \cdot x(n)$ – is linear; it is also *memoryless*
- $y(n) = \frac{x(n)+x(n-1)}{2}$ (not memoryless):

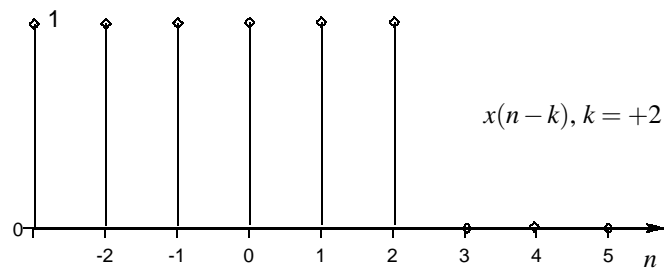
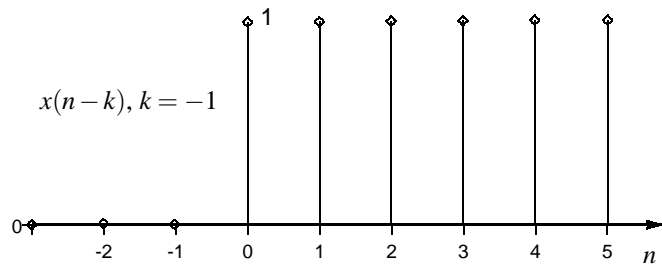
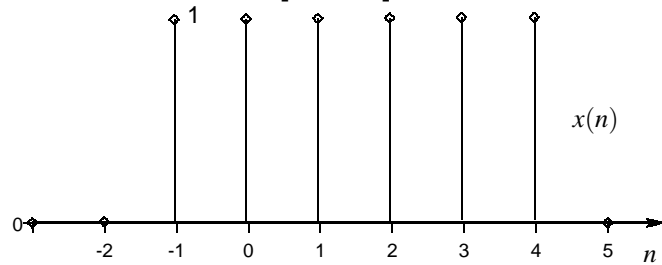
$$\begin{aligned}
 T\{\alpha_1 x_1(n) + \alpha_2 x_2(n)\} &= \frac{[\alpha_1 x_1(n) + \alpha_2 x_2(n)] + [\alpha_1 x_1(n-1) + \alpha_2 x_2(n-1)]}{2} = \\
 &= \frac{\alpha_1 x_1(n) + \alpha_1 x_1(n-1)}{2} + \frac{\alpha_2 x_2(n) + \alpha_2 x_2(n-1)}{2} = \\
 &= \alpha_1 \frac{x_1(n) + x_1(n-1)}{2} + \alpha_2 \frac{x_2(n) + x_2(n-1)}{2} = \alpha_1 y_1(n) + \alpha_2 y_2(n) \quad \text{end}
 \end{aligned}$$

(not L) $y(n) = (x(n))^2$ because

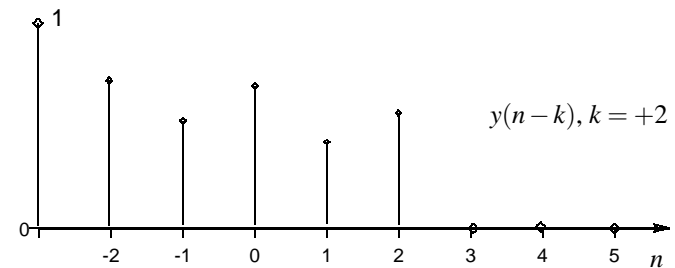
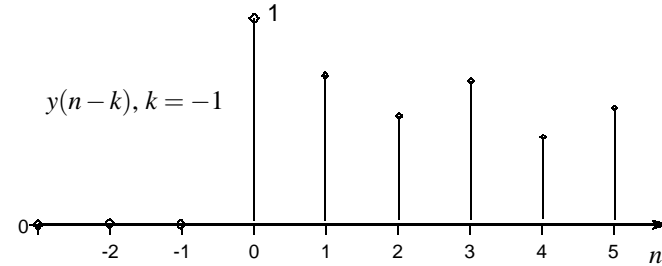
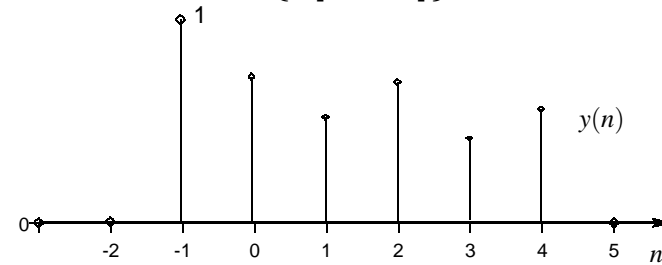
$$T\{x_1(n) + x_2(n)\} = (x_1(n) + x_2(n))^2 = (x_1(n))^2 + (x_2(n))^2 + [2 \cdot x_1(n)x_2(n)]$$

shift example

Input signals $x[n-k]$.



Responses $T\{x[n-k]\}$ of TI system $T\{.\}$



Other properties: causality, stability

causality

→ $y(n_0)$ depends only on $x(n)$, $n \leq n_0$ (*usually less important in DT implementations*)

stability

→ bounded input causes bounded output [BIBO]

bounded → $\exists B_x : \forall n \ |x(n)| \leq B_x < \infty$

Examples

Decimator (compressor)

$$y(n) = x(Mn)$$

→ L, but not TI (*prove it!*)

1-st order difference

forward: $y(n) = x(n+1) - x(n)$ → noncausal

backward: $y(n) = x(n) - x(n-1)$ → causal

Accumulator

$$y(n) = \sum_{k=-\infty}^n x(k)$$

→ unstable; (*hint: feed it with $u[n]$*)

LTI systems: impulse response

$h[n] = T\{\delta[n]\}$ \longrightarrow impulse response of $T\{.\}$

$h[n]$ characterizes completely system $T\{.\}$ – we may compute its response for any input $x[n]$.

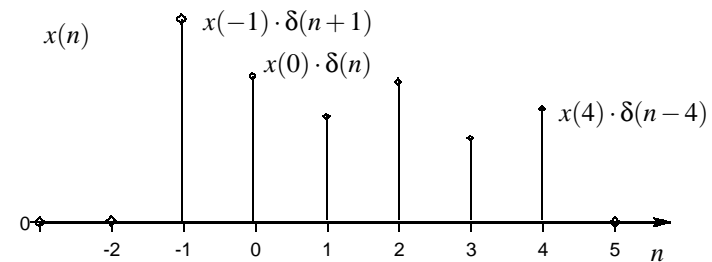
- Decompose $x[n]$ into weighted sum of impulses $\delta[n - k]$

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

- Superpose responses (use LTI properties)

$$y[n] = \sum_{k=-\infty}^{\infty} x[k]h[n - k]$$

\longrightarrow this is a **convolution sum**



Convolution example

Convolution properties

$$y[n] = \sum_{k=-\infty}^{\infty} x[k]h[n-k]$$

we denote as

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

Properties of “*”

“*” is commutative: $x[n] * h[n] = h[n] * x[n]$

“*” distributes over addition $x[n] * (h_1[n] + h_2[n]) = x[n] * h_1[n] + x[n] * (h_2[n])$

System and $h[n]$

- causality $\Leftrightarrow h[n] = 0, n < 0$. A hint: $y(n) = \sum_{k=-\infty}^{\infty} h(k)x(n-k)$
 - stability $\Leftrightarrow S = \sum_{k=-\infty}^{\infty} |h(k)| < \infty$
-

Linear difference equations

... describe an important class of LTI systems.

$$\sum_{k=0}^N a_k y(n-k) = \sum_{k=0}^M b_k x(n-k), \quad a_0 = 1 \text{ (traditionally)}$$

or

$$\begin{aligned} y(n) &= a_1 \cdot y(n-1) + a_2 \cdot y(n-2) + \dots + a_n \cdot y(n-N) + \\ &+ b_0 \cdot x(n) + b_1 \cdot x(n-1) + b_2 \cdot x(n-2) + \dots + b_n \cdot x(n-M) \end{aligned}$$

Note: if, for a given input $x_p[n]$, an output sequence $y_p[n]$ satisfies given differential equation,

$$y[n] = y_p[n] + y_h[n]$$

will also satisfy the equation, if $y_h[n]$ is a solution to $\sum_{k=0}^N a_k y(n-k) = 0$ (homogenous equation).

Differential equation – example

An equation: $y(n) = a \cdot y(n-1) + x(n)$

with input

$$x(n) = 0, n < 0$$

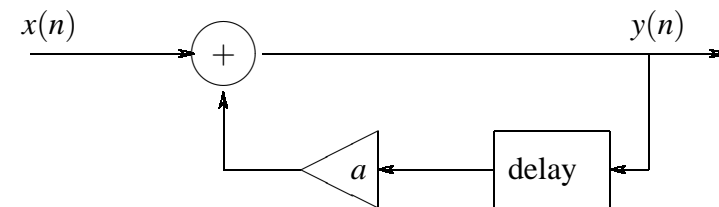
$$x(n) \neq 0, n > 0.$$

$$y(0) = a \cdot y(-1) + x(0)$$

$$y(1) = a \cdot y(0) + x(1)$$

$$y(2) = a \cdot y(1) + x(2)$$

...



Initial condition: $y(-1) = \alpha$

Let $x[n] = \delta[n]$

$$y(0) = a \cdot \alpha + 1$$

$$y(1) = a(a \cdot \alpha + 1) = a^2 \alpha + a$$

$$y(2) = a^3 \alpha + a^2$$

...

$$y(n) = a^{n+1} \alpha + a^n$$

example – continued

