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# 1 Week 1 Module 1: Basics of Digital Signal Processing

# 1.1 Introduction to digital signal processing

# 1.1.1 Signal

• Description of the evolution of a physical phenomenon

phenomenon signal
weather temperature
sound pressure
sound magnetic deviation
light intensity gray level on paper

# 1.1.2 Processing

- Analysis: Understanding the information carried by the signal
- Synthesis: Creating a signal to contain the given information

## 1.1.3 Digital

- Discrete Time
  - Splice up time into a series of descrete instance without loosing information
  - Harry Nyquist and Claude Shannon state with the blueSampling Theorem that continuous time representation and discrete time representation are equivalent.
  - The Sampling Theorem: Under appropriate "slowness" conditions for  $\mathbf{x}(t)$  we have

$$x(t) = \sum_{n = -\infty}^{\infty} x[n] \operatorname{sinc}(\frac{t - nT_s}{T_s})$$
 (1)

- The condition under which the Sampling Theorem holds was given by Fourier and it's blueFourier Analysis.
- The fouriere transform will give us a quantitive measure how fast a signal moves

## • Discrete Amplitude

- Through discretisation of amplitudes only a set of predefined values are possible.
- The set of levels is countable i.e. we can always map the level of a sample to an integer. If our data is represented by integer it becomes complete abstract and general which has very importand consequences in the following three domains:
  - \* Storage special devices for recoding needed
  - \* Processing General purpose microprocessor is sufficient
  - \* **Transmission** Reproduction of the original signal and therefore eliminating nois is easy

#### 1.1.4 From Analog to Digital Signal Processing

- Analog asks for  $f_{(t)} = ?$
- Digital represents data as a sequence of numbers (scaled with a factor of 1000)

# 1.2 Discrete-Time Signals

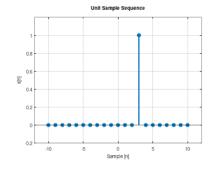
#### 1.2.1 Basic Definitions

- Sequence: defined as bluecomplex-valued function
- Discrete-Time Signal: a sequence of complex numbers
  - one dimension (for now)
  - notation: x[n]
  - two-sides sequencies: x:  $\mathbb{Z} \to \mathbb{C}$
  - n is a-dimensional "time", sets an order on the sequence of samples
  - analysis: periodic measurement
  - synthesis: stream of generated samples, reproduce a physical phenomenon

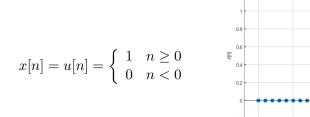
# 1.2.2 Octave Algorithm for some basic Signals

## Unit Impulse

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



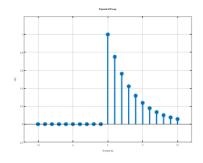
# Unit Step



# Real-valued exponential Sequence

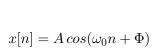
0 Sample [n]

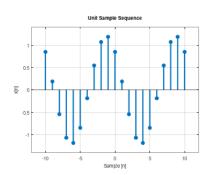
$$x[n] = a^n, \, \forall n \ a \in \mathbb{R}$$



# Sinusoidal Sequence

```
function [x,n] = cosseq(n1,n2,A, omega, phi)
% Generates x(n) = A*cos(2*pi*omega*n + phi); n1 <= n2
% ------
% [x,n] = cosseq(n1,n2,A,omega,phi)
%
    n = [n1:n2]; x = A*cos(2*pi*omega*n + phi);
end</pre>
```





#### 1.2.3 Classes of Discrete-Time signals

- 1. Finite-Length
  - indicate notation: x[n], n = 0.1.2....N 1
  - vector notation:  $x = [x_0, x_1, ...x_{N-1}]^T$
  - practical entities, good for numerical packages (e.g. numpy)
- 2. Infinte-Length
  - sequence notation: x[n],  $n \in \mathbb{Z}$
  - abstraction, good for theorems

#### 3. Periodic

- N-periodic sequence:  $\tilde{x}[n] = \tilde{x}[n+kN]$ , n,k,N  $\in \mathbb{Z}$
- same information as in finite-length of length N
- bluenatural bridge between finite and infinite length
- 4. Finite-Support blueFinite-support sequence black

$$\overline{x}[n] = \begin{cases} x[n] & if 0 \le n < N, n \in \mathbb{Z} \\ 0 & otherwise \end{cases}$$
 (2)

- same information as in finite-length of length N
- another bridge between finite and infinite lengths

## 5. Elementary Operations

Scaling

$$y[n] = ax[n] \rightarrow \begin{cases} a > 0 & amplification \\ a < 0 & attenuation \end{cases}$$
 (3)

Sum

$$y[n] = x[n] + z[n] \tag{4}$$

**Product** 

$$y[n] = x[n] * z[n]$$
 (5)

Shift

$$y[n] = x[n-k] \rightarrow \begin{cases} k > 0 & deleay \\ k < 0 & anticipate \end{cases}$$
 (6)

Integration

$$y[n] = \sum_{k=-\infty}^{n} x[k] \tag{7}$$

Differentation

$$y[n] = x[n] - x[n-1]$$
 (8)

 $[logo=, couleur=yellow!10, barre=snake, arrondi=0.1] Relation\ Operator\ and\ Signals$ 

- The blueunit step can be optained by applying the blueintegration operator to the bluediscrete time pulse.
- The blueunit impulse can be optained by applying the bluedifferentation operator to the blueunit step.

#### 1.2.4 Energy and Power

**Energy** Many sequencies have an infinity amount of energy e.g. the unit step u[n],

$$E_x = ||x||_2^2 = \sum_{k=-\infty}^{\infty} |x[n]|^2$$
 (9)

**Power** To describe the energetic properties of the sequencies we use the concept of power

$$P_x = ||x||_2^2 = \frac{1}{N} \sum_{n=0}^{N-1} |x[n]|^2$$
 (10)

Many signals have infi

N = 33

# 1.3 Basic signal processing

#### 1.3.1 How a PC plays discrete-time sounds

1. The discrete-time sinusoid

$$x[n] = sin(\omega_0 t + \Theta)$$

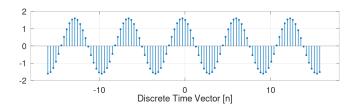
# Vector lenght

```
n=-(N-1)/2:pi/10:(N-1)/2; # Discrete Time Vector
omega0 = pi/10;
theta = pi/2

f = 1.6*sin(omega0+n + theta); # The sinusoid

# Do not open the graphic window in org
figure( 1, "visible", "off");

stem(n,f, "filled", "linewidth", 2, "markersize", 6);
axis([-(N-1+4)/2 (N-1+4)/2 -2 2])
set(gca, "fontsize", 24);
grid on;
xlabel("Discrete Time Vector [n]");
print -dpng "-S1400,350" ./image/sin.png;
# Org-Mode specific output
ans = "./image/sin.png";
```



## 2. Digital vs physical frequency

- Discrete Time:
  - Periodicity: how many samples before the pattern repeats (M)
  - n: no physical dimension
- Physical World:
  - Periodicity: hog many seconds before the pattern repeats
  - frequency measured in Hz
- Soundcard T<sub>s</sub> System Clock
  - A sound card takes ever  $\rm T_s$  an new sample from the discrete-time sequence.
  - periodicity of M samples  $\rightarrow$  periodicity of M  $T_s$  seconds
  - real world frequency

$$f = \frac{1}{M T_s} Hz \tag{11}$$

- Example
  - usually we choose  $F_s$  the number of samples per seconds
  - $-\ T_{\rm s}=1/F_{\rm s}$

$$F_s = 48000$$
e.g. a typical value

$$T_s = 20.8 \mu \ s$$

$$f = 440Hz$$
, with M = 110

# 1.3.2 The Karplus Strong Algorithm

- 1. The Moving Average
  - simple average (2 point average)

$$m = \frac{a+b}{2} \tag{12}$$

• moving average: take a "local" average

$$y[n] = \frac{x[n] + x[n-1]}{2} \tag{13}$$

• Average a sinusoid

$$x[n] = cos(\omega n)$$

$$y[n] = \frac{cos(\omega n) - cos(\omega (n - 1))}{2}$$

$$y[n] = cos(\omega n + \theta)$$

[couleur=yellow!10, arrondi = 0.1, logo=, ombre=true]Linear Transformation Applying a linear transformation to a sinusoidal input results in a sinusoidal output of the same frequency with a phase shift.

2. Reversing the loop

$$y[n] = x[n] + \alpha y[n-1] \rightarrow$$
 The Karplus Strong Algorithm (14)

- Zero Initial Conditions:
  - set a start time (usually  $n_0 = 0$ )
  - assume input and output are zero for all time before  $N_0$

## 1.4 Digital Frequency

[logo=,couleur=yellow!10,barre=snake]Digital Frequency

$$\sin\left(n(\omega + 2k\pi)\right) = \sin\left(n\omega + \phi\right), \text{ k in } \mathbb{Z}$$

$$= e^{i(\phi + n*2\pi\omega)}$$
(15)

[logo=,couleur=yellow!10, barre=snake]Complex Exponential

$$\omega = \frac{M}{N} \times 2 \times \pi \tag{16}$$

# 1.5 The Reproduction Formula

 $[logo=, couleur=yellow!10, \, barre=snake, \, arrondi=0.1] \\ Reproduction \, Formula$ 

$$x[n] = \sum_{k=-\infty}^{\infty} x[k]\delta[n-k]$$
 (17)

Any bluesignal can be expressed as a linear combination of wighted and shifted pulses.