Digital Signal Processing

Sampling of Analog Signals

#### Signals

- Signal = a measurable quantity which varies in time, space or some other variable
- ► Example: a voltage signal
- ▶ Represented as a mathematical function, e.g. v(t).
- Off Topic: Glossary
  - "e.g." = "exampli gratia" (lat.) = "for example" (eng.) = "de exemplu" (rom.)
  - "i.e." = "id est" (lat) = "that is" (eng.) = "adică" (rom.)

# Signal dimension

- ▶ Unidimensional (1D) signal: = a function of a signal variable
  - ▶ Example: v(t) = a function of time
- ► Multidimensional (2D, 3D, ...) signal = a function of multiple variables
  - Example: intensity of an image across a photo, I(x, y), depends on the two coordinates x and y that specify the position

#### Continuous and discrete signals

- ► Continuous (analog) signal = function of a continuous variable
  - ► Signal has a value for possible value of the variable in the defined range
  - ▶ The variable may be defined only in a certain range (e.g.  $t \in [0, 100]$ ), but it is a compact range
- ▶ Discrete signal = function of a discrete variable
  - ► Signal has values only at certain discrete values (samples)
  - ▶ Indexed with natural numbers: x[-1], x[0], x[1] etc.
  - Outside the samples, the signal is not defined

#### Notation

- We use the following notation throughout these lectures
- Continuous signals
  - ▶ Has **round parantheses**, e.g.  $x_a(t)$
  - Sometimes has the a subscript
  - ► The variable is usually t (time)
  - x(2.3) = the value of the signal a(t) at t = 2.3
- Discrete signals
  - ▶ Has square brackets, e.g. x[n]
  - ▶ The variables are denoted as n or k (suggest natural numbers)
  - $\triangleright$  x[3] =the value of the signal x[n] for n = 3
  - x[1.5] = does not exist

#### Signals with continuous and discrete values

- ▶ The signal values can be continuous or discrete
  - ► Example: signal values stored as 8-bit or 16-bit values
- On digital systems, signals always have discrete values due to finite number precision

## Discrete frequency

- A signal is **periodic** if the values repeat themselves after a certain time (**period**)
- Frequency = inverse of period
- Continuous signals:

$$x_a(t) = x(t+T)$$

- ► *T* is usually measured in seconds (or some other unit)
- $F = \frac{1}{T}$  is measured in Hz =  $\frac{1}{s}$  (Hertz)
- Discrete signals:

$$x[n] = x[n+N]$$

- N has no unit, because it is just a number
- $f = \frac{1}{N}$  has no unit also

# Domain of existence of frequency

- Continuous signals
  - ▶ Period T can be as small as possible  $T \rightarrow 0$
  - ▶ Therefore F could go up to  $\infty$
- Analog signals
  - ▶ Smallest period is N = 2 (excluding N = 1, constant signals)
  - ► Largest possible frequency is  $f_{max} = \frac{1}{2}$
  - ► Consequence of using natural numbers to index the samples (x[0], x[1], x[2]...), without any physical unit attached

## Sampling

- Taking the values from an analog signal at certain discrete moments of time, usually periodic
- ▶ Distance between two samples = sampling period  $T_s$
- **Sampling frequency**  $F_s = \frac{1}{T_s}$
- Why sampling?
  - ► Converts continuous signals to discrete
  - Processing of continuous signals is expensive
  - Processing of discrete signals is cheap (digital devices)
  - Sometimes nothing is lost due to sampling

## Sampling equation

▶ Sampling of the continuous signal  $x_a$ :

$$x[n] = x_a(n \cdot T_s)$$

▶ The *n*-th value of the discrete signal x[n] is the value of the analog signal  $x_a(t)$  taken after *n* sampling periods, at  $t = n \cdot T_s$ 

# Sampling of harmonic signals

Let's sample a cosine:  $x_a(t) = cos(2\pi Ft)$ 

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

$$= cos(2\pi F nT_s)$$

$$= cos(2\pi F n \frac{1}{F_s})$$

$$= cos(2\pi \frac{F}{F_s} n)$$

- Sampling a continuous cosine (or sine) produces a discrete cosine (or sine)
- ▶ The discrete frequency is  $f = \frac{F}{F_s}$

# Sampling theorem (Nyquist-Shannon)

▶ If a signal that has maximum frequency  $F_{max}$  is sampled with a a sampling frequency

$$F_s \geq 2F_{max}$$
,

▶ then it can be perfectly reconstructed from its samples using the formula:

$$x_a(t) = \sum_{n=-\infty}^{+\infty} x[n] \cdot \frac{\sin(\pi(F_s t - n))}{\pi(F_s t - n)}.$$

#### Comments on the sampling theorem

- ► All the information in the original signal is contained in the samples, provided the sampling frequency is high enough
- We can process discrete samples instead of the original analog signals
- ▶ Sampling with  $F_s \ge 2F_{max}$  makes the discrete frequency smaller than 1/2

$$f = \frac{F}{F_s} \le \frac{F_{max}}{F_s} \le \frac{1}{2}$$

# Aliasing

- http://www.dictionary.com/browse/alias:
  - ▶ "alias": a false name used to conceal one's identity; an assumed name
- What happens when the sampling frequency is not high enough?
- Every discrete frequency that exceeds  $f_{max}=\frac{1}{2}$  is **identical** (an alias) to a frequency that is lower than  $f_{max}=\frac{1}{2}$
- Proof:
  - Consider  $x[n] = cos(2\pi fn)$ ,  $f > \frac{1}{2}$
  - We can always subtract  $2\pi n$  since cos() is periodical
  - ▶ This means reducing f with 1
  - ▶ Thus we can always end up a frequency  $f' \in [-1/2, 1/2]$  (up to a sign change)

# Aliasing (continued)

$$cos(2\pi(\frac{1}{2}+\epsilon)n) = cos(2\pi(\frac{1}{2}-\epsilon)n)$$
  $sin(2\pi(\frac{1}{2}+\epsilon)n) = -sin(2\pi(\frac{1}{2}-\epsilon)n)$ 

- Aliasing only affects digital signals
- ▶ Sampling with  $F_s \ge 2F_{max}$  ensures  $f \le \frac{1}{2}$ , so no aliasing

## The problem of aliasing

- ► Sampling different analog signals, with different frequencies, leads to exactly the same samples
- How can be know from what signal did the samples come from? Impossible.
- ▶ Better remove from the signal the frequencies larger than  $\frac{F_s}{2}$ , otherwise they will be similar to a different frequency and create confusion
- Anti-alias filter: a low-pass filter situated before a sampling circuit, rejecting all frequencies  $F > \frac{F_s}{2}$  from the signal before sampling
  - Standard practice in the design of processing systems

# Signal reconstruction from samples

▶ A discrete frequency  $f \in \left[-\frac{1}{2}, \frac{1}{2}\right]$  will be reconstructed as follows:

$$x_r(t) = x[\frac{t}{T_s}] = x[t \cdot F_s]$$

- ▶ For a discrete frequency outside the  $\left[-\frac{1}{2},\frac{1}{2}\right]$  interval
  - ▶ Reconstruction of the original frequency is impossible
  - ► The frequency is replaced with the aliased frequency f' from the interval  $\left[-\frac{1}{2}, \frac{1}{2}\right]$
- lacktriangle Reconstruction always produces signals with frequencies in  $\left[-\frac{Fs}{2},\frac{Fs}{2}\right]$
- Only signals sampled according to the sampling theorem will be reconstructed identically

## Signal quantization and coding

- ► In practice, the values of the samples are rounded to fixed levels, e.g. 8-bit, 16-bit values.
- ► This "rounding" is known as quantization
- The "rounding error" is known as quantization error
- Converting the value in binary form is known as coding

# A/D and D/A conversion

- ▶ Sampling + quantization + coding is usually done by an Analog to Digital Converter (ADC)
  - ▶ It takes an analog signal and produces a sequence of binary-coded values
- Reconstructing an analog signal from numeric samples is done by a Digital to Analog Converter (DAC)
  - ▶ Usually reconstruction is not based on sampling theorem equation, which is too complex, but with simpler empiric approaches.