

# Digital Signal Processing

# Implementation

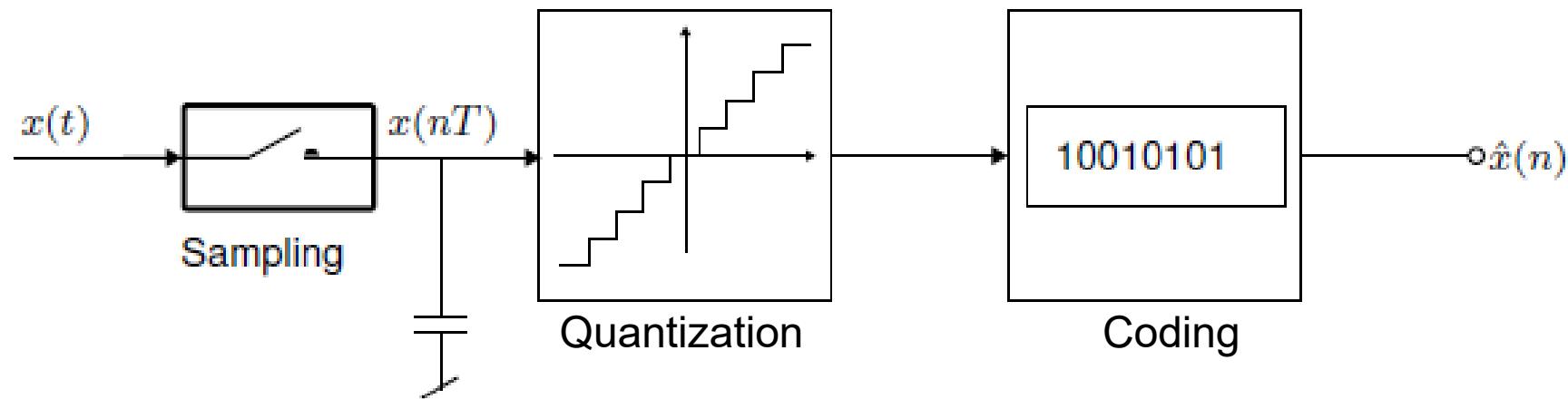
Zhuoqi Cheng

[zch@mimi.sdu.dk](mailto:zch@mimi.sdu.dk)

SDU Robotics

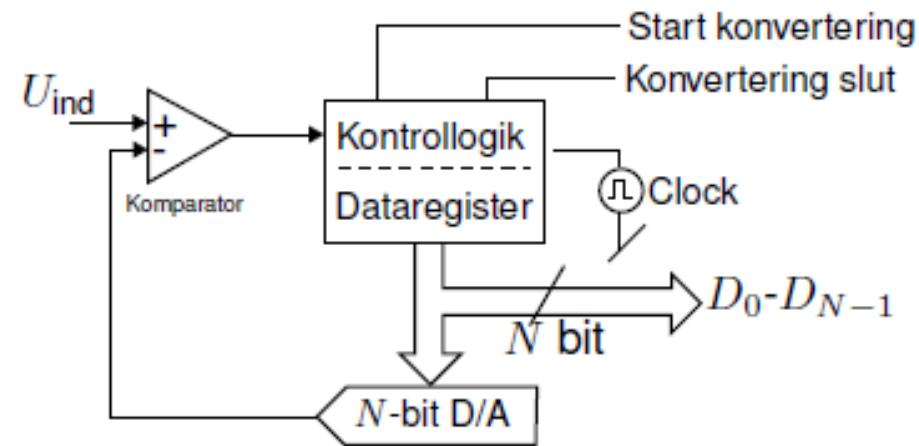
# Quantization

In analog-digital conversion, the input signal  $x(t)$  is transformed into a sequence with finite resolution as shown below.



# Analog-Digital converter principle

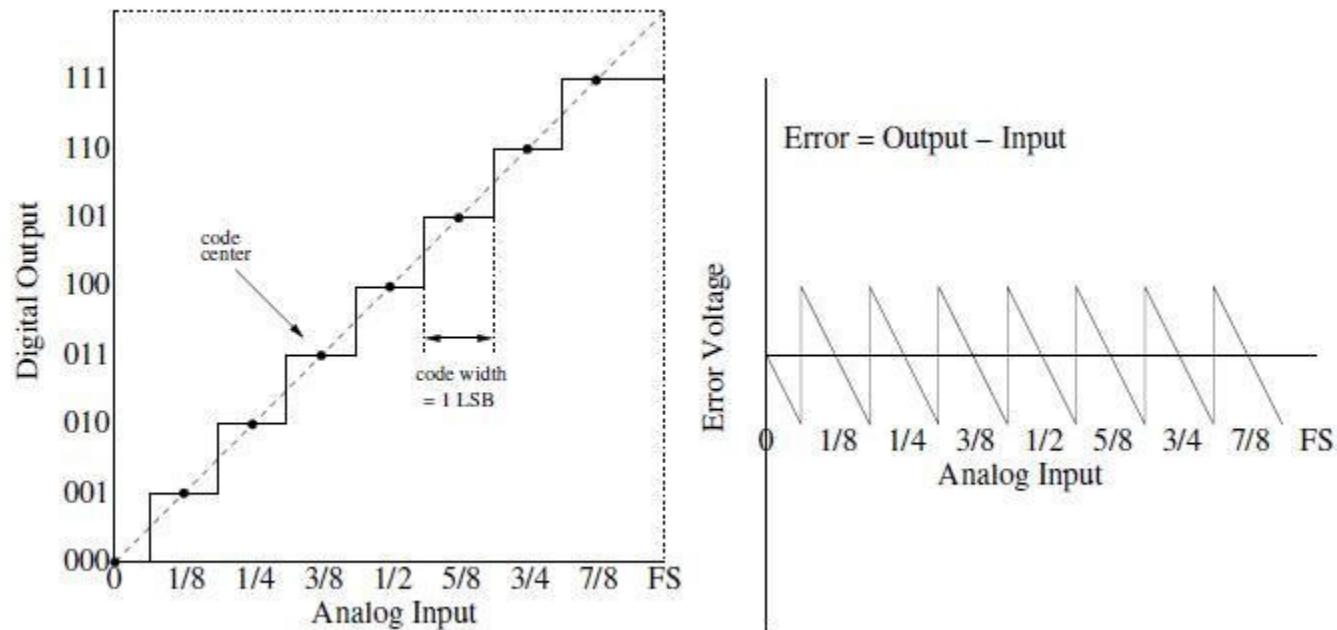
A Successive Approximation A/D converter is built according to the principle shown below, where the output of the A/D converter is compared with the input sequence through a comparator.



If the conversion is to take place faster (in the MHz range), then a Flash A/D converter is used.

# Quantization

During A/D conversion, the input signal is quantized so that it has a word length of  $N$  bits, i.e. the representation has  $2^N$  different quantization levels.



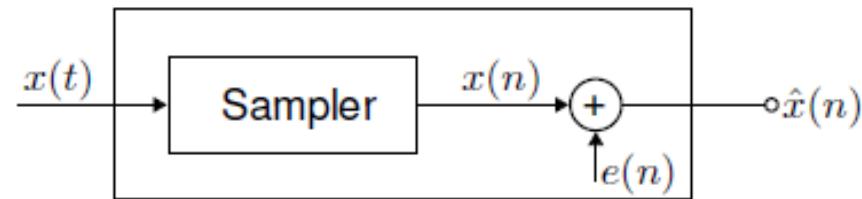
# Quantization error

The quantization error means that the value of the input sequence  $x(n)$  and the quantized input sequence  $\hat{x}(n)$  are different

$$\hat{x}(n) = x(n) + e(n)$$

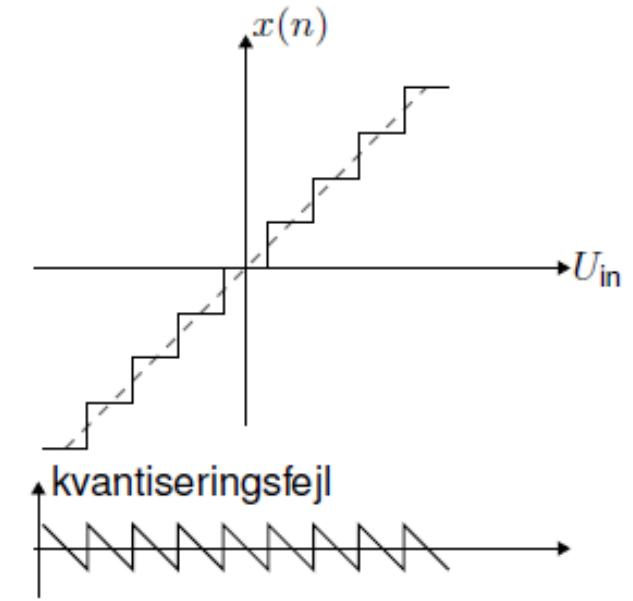
where  $e(n)$  is the quantization error.

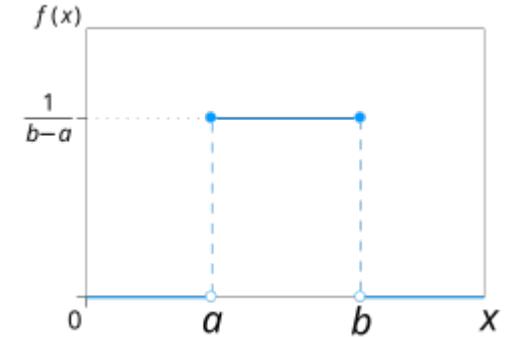
The quantization error can be thought of as a noise sequence as shown below.



The magnitude of the quantization error is bounded as

$$-\frac{\Delta V}{2} \leq e_q \leq \frac{\Delta V}{2}$$





# Quantization error

The signal varies from  $-V$  to  $+V$ , and has  $2^N$  quantization levels.

$$2V = 2^N \Delta V$$

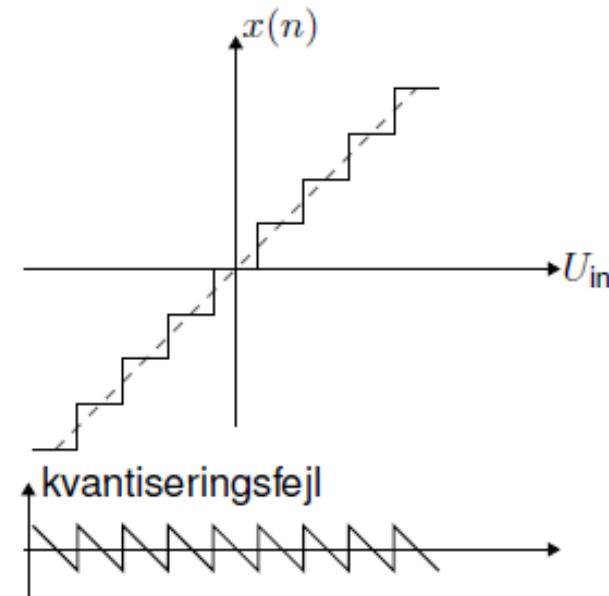
The quantization error is modeled as a uniform distribution in  $-\frac{\Delta V}{2} \leq e_q \leq \frac{\Delta V}{2}$ . According to the theory of probability and random variables, the quantization noise is calculated as

$$E(e_q^2) = \frac{\Delta V^2}{12}$$

Signal-to-Quantization Noise Ratio (SQNR) can be calculated as

$$SQNR = \frac{E((V/\sqrt{2})^2)}{E(e_q^2)} = \frac{\left(\frac{2^N \Delta V}{2\sqrt{2}}\right)^2}{\frac{\Delta V^2}{12}} = (2^N \cdot \frac{3}{2})^2$$

$$SQNR_{dB} = 20 \log_{10}(2^N \cdot \frac{3}{2}) \approx 6N \text{ [dB]}$$



# Multirate sampling

Multirate sampling is used, for example, if data is to be processed at a different sample frequency than it was recorded or if several signals sampled with different frequencies are to be put together.

This procedure is also known as **resampling**. It can be

- **Upsampling**
- **Downsampling**

Matlab function:  
 $y = \text{downsample}(x,n)$   
&  
 $y = \text{upsample}(x,n)$

Please check them by your own.

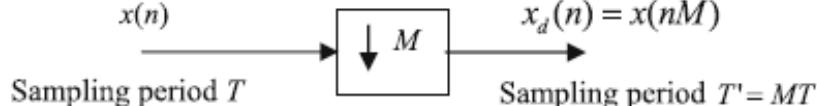
# Downsampling

Let  $x(n)$  be a sequence obtained by sampling with the sample time  $T$ . When downsampling, a sequence with sample interval  $T' > T$  is desired. Specifically desired  $T' = MT$  and thus have

$$x_d(n) = x(nM)$$

where **M is an integer**.

A block diagram for a downsampling is shown here.



```
% Matlab
fs_original = 16000; % Original sampling rate: 16 kHz
t = 0:1/fs_original:0.05;

f1 = 500; % 0.5 kHz
f2 = 1000; % 1 kHz
f3 = 2000; % 2 kHz

% original signal
x_mixed = sin(2*pi*f1*t) + sin(2*pi*f2*t) + sin(2*pi*f3*t);
figure;
plot(t, x_mixed);

% Frequency spectrum
N = length(x_mixed);
f = (-N/2:N/2-1)*(fs_original/N);
X_mixed = 1/N*fftshift(abs(fft(x_mixed)));
figure;
plot(f, X_mixed);
```

```
% downsampling
downsample_factor = 2;
fs_down = fs_original / downsample_factor;

x_down = x_mixed(1:downsample_factor:end);
t_down = t(1:downsample_factor:end);
figure;
plot(t_down, x_down);

% Frequency spectrum of downsampled
N_down = length(x_down);
f_down = (-N_down/2:N_down/2-1)*(fs_down/N_down);
X_down = 1/N_down*fftshift(abs(fft(x_down)));
figure;
plot(f_down, X_down);
```

# Why downsampling?

- Signal length reduces (less memory)
- Faster processing time
- If you are using FIR filter, its order (or length) can be reduced.

$$\text{FIR half length } M = \frac{B_n f_s}{2\Delta f}$$

# Downsampling spectrum

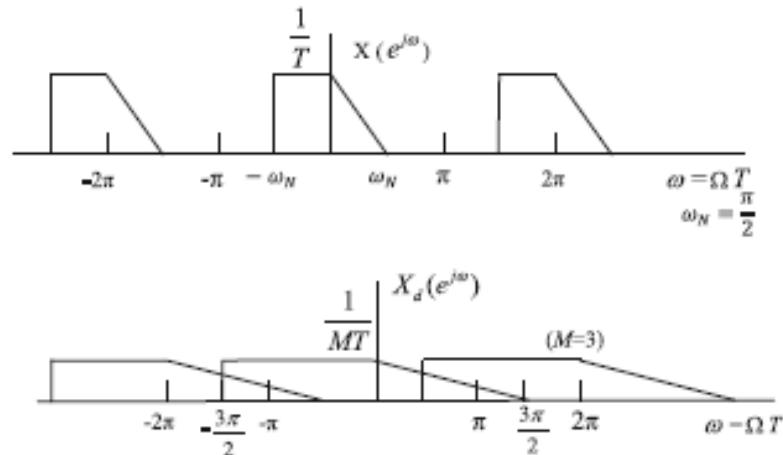
By Fourier transform of  $x_d$ , the following spectrum function is obtained

$$X_d(e^{j\omega}) = \frac{1}{M} \sum_{k=0}^{M-1} X(e^{\frac{j(\omega-2\pi k)}{M}})$$

Recap: DFT

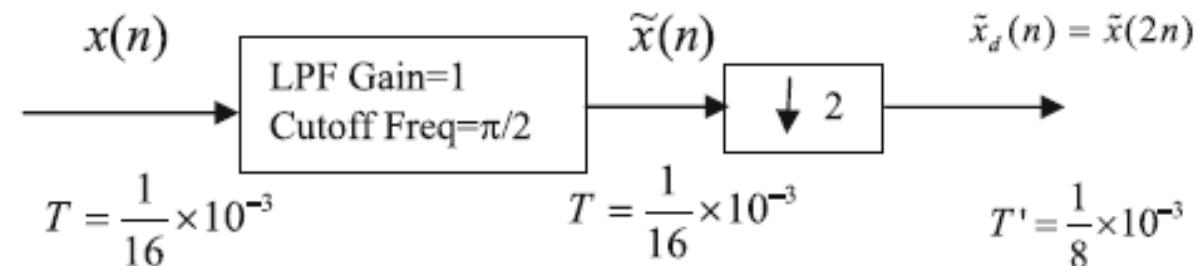
$$X(m) = \frac{1}{N} \sum_{n=0}^{N-1} x(n)e^{-j2\pi mn/N}$$

Example here showing:  
 $\omega_N = \frac{\pi}{2}$   
and  $M = 3$



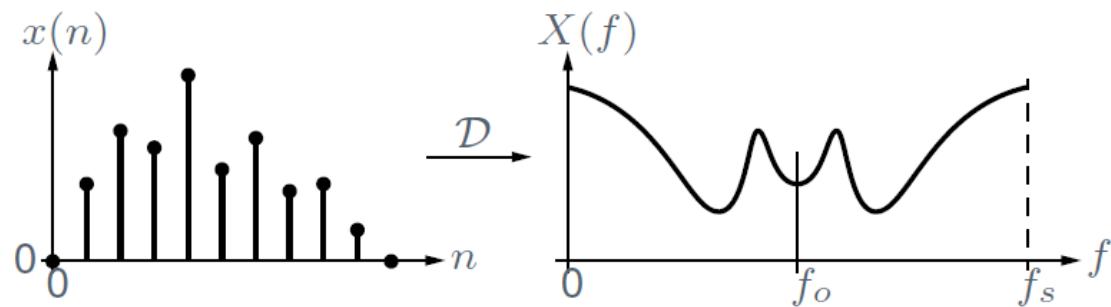
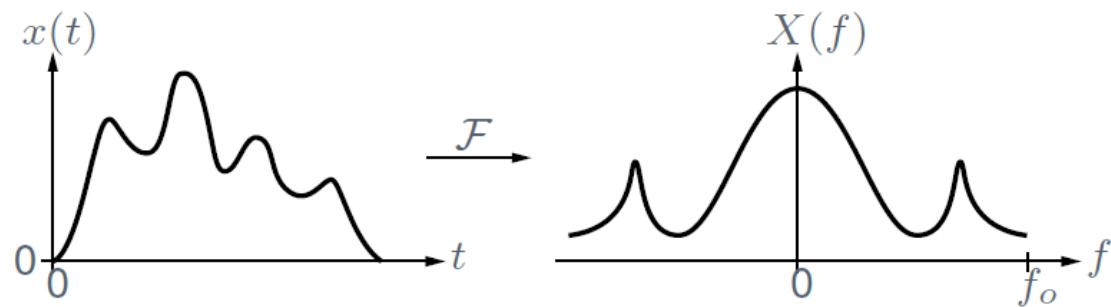
# Downsampling anti-aliasing

To avoid aliasing, an anti-aliasing filter  $H(z)$  (low-pass filter) must be added before downsampling.



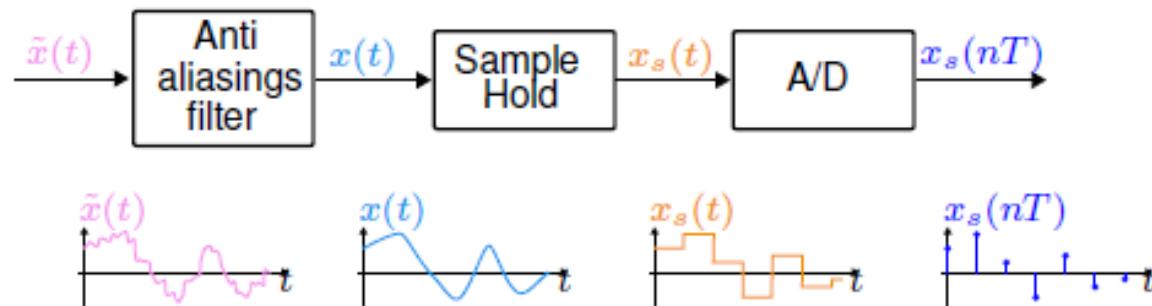
# Recap: Aliasing

When performing an FFT, it is important to limit the frequency content of the input signal  $x(t)$  is analyzed. If  $x(t)$  has frequency components above the convolution frequency  $f_o = f_s/2$ , then aliasing occurs.



# Anti-aliasing filter

To avoid aliasing, a low-pass filter must be inserted before sampling the signal.  
This filter must remove frequencies above  $f_o$ .



The cutoff frequency of the anti-aliasing filter must be smaller than  $f_o$ .  
In practice, the filter is better to limit the analysis area to approximately  $f_s/3$ .

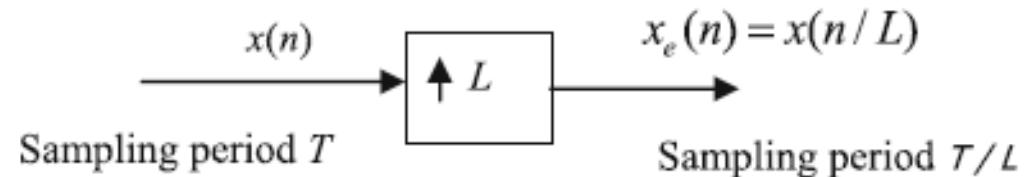
# Upsampling

Let  $x(n)$  be a sequence obtained by sampling with the sample time  $T$ . When upsampling, a sequence with sample interval  $T' < T$  is desired. Specifically desired  $T' = T/L$  and thus have

$$x_u(n) = \begin{cases} x(n/L) & \text{hvis } n/L \in \mathbb{Z} \\ 0 & \text{ellers} \end{cases}$$

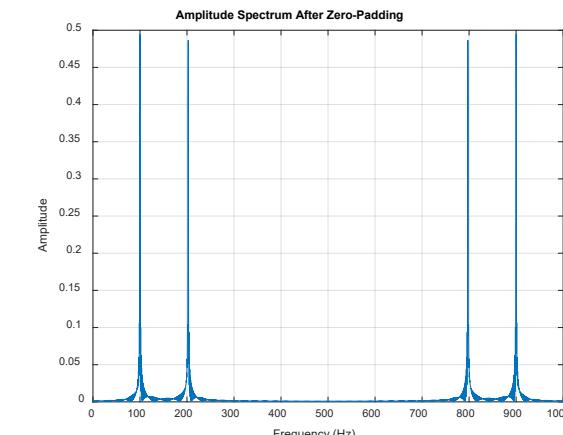
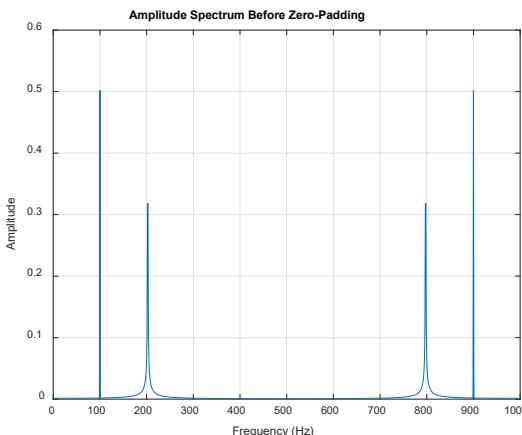
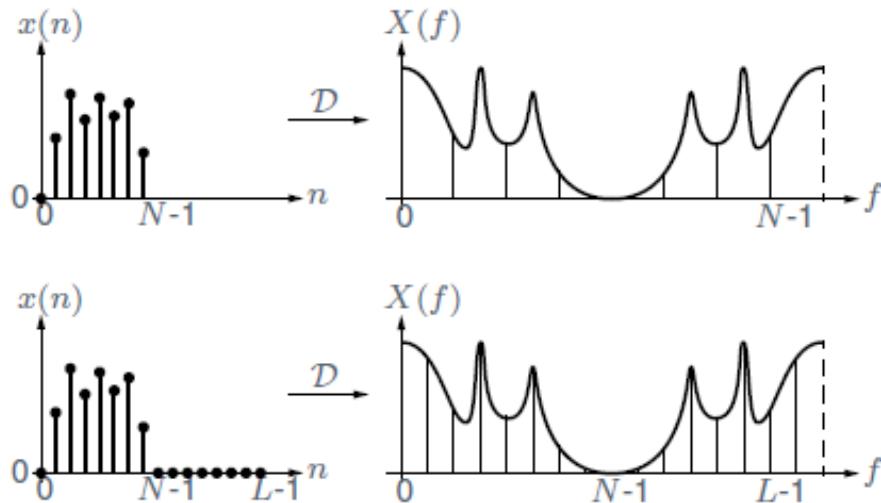
where  **$L$  is an integer**

A block diagram for an upsampling is shown here.



# Zero-padding

When an N-point Fourier Transform is performed, the frequency step is  $F = f_s/N$ . If a higher resolution of the frequency spectrum is desired, then zero filling can be used so that the signal is increased to L samples.



```
fs = 1e3; % [Hz] sampling frequency
t = 0:0.001:1-0.001;
x = cos(200*pi*t)+sin(405*pi*t); % Signal

% Plot the time domain signal
figure;
plot(t, x);
title('Original Signal in Time Domain');

% FFT Analyse
N = length(x); % Original signal length
X = fft(x);
f = fs*(0:N-1)/N;

figure;
plot(f, abs(X)/N);
title('Amplitude Spectrum Before Zero-Padding');

% Zero-padding
N_padded = 2048;
X_padded = fft(x, N_padded); % FFT with zero-padding

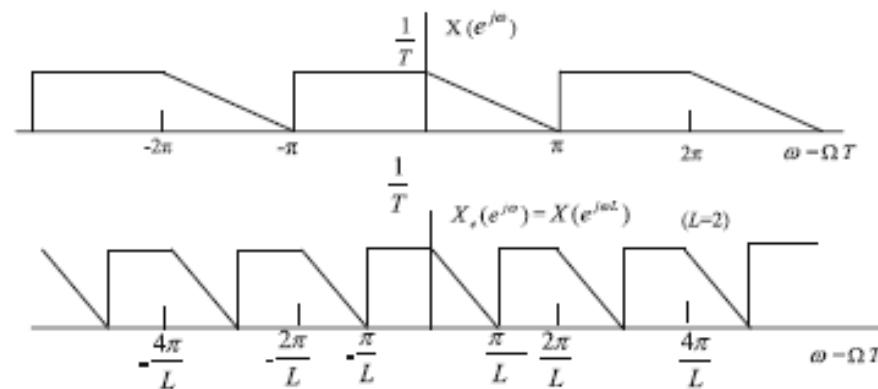
% Frequency axis for padded signal
f_padded = fs*(0:N_padded-1)/N_padded;

figure;
plot(f_padded, abs(X_padded)/N);
title('Amplitude Spectrum After Zero-Padding');
```

# Upsampling spectrum

By Fourier transform of  $x_u$ , the following spectrum function is obtained

$$X_u(e^{j\omega}) = \sum_{n=Lk} x_u(n)e^{-j\omega n} = \sum_k x_u(Lk)e^{-jL\omega k} = \sum_k x(k) e^{-jL\omega k} = X(e^{jL\omega})$$



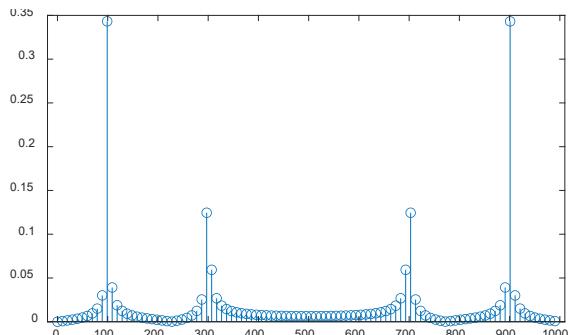
# Matlab – upsampling example

```
fs_original = 1000; % Original sampling rate: 1 kHz
t_original = 0:1/fs_original:0.1;

f1 = 100; % 100 Hz
f2 = 300; % 300 Hz

x_original = 0.7 * sin(2*pi*f1*t_original) + 0.3 * sin(2*pi*f2*t_original);
figure;
plot(t_original, x_original);

% Compute FFT
N_original = length(x_original);
f_original = (0:N_original-1) * (fs_original/N_original);
X_original = fft(x_original);
figure;
stem(f_original, abs(X_original)/N_original);
```



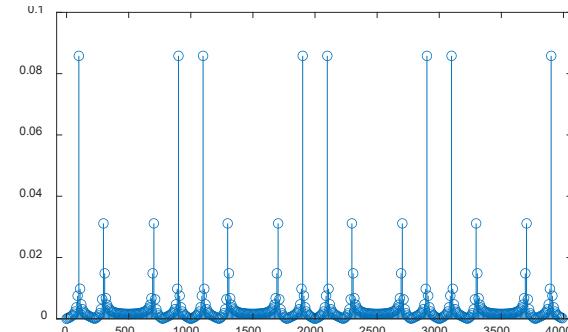
```
upsample_factor = 4;
fs_upsampled = fs_original * upsample_factor; % New sampling rate: 4 kHz

% Upsampling: Insert zeros between samples
x_upsampled_zeros = zeros(1, length(x_original) * upsample_factor);
x_upsampled_zeros(1:upsample_factor:end) = x_original;

% Time vector for upsampled signal
t_upsampled = (0:length(x_upsampled_zeros)-1) / fs_upsampled;
figure;
plot(t_upsampled, x_upsampled_zeros);

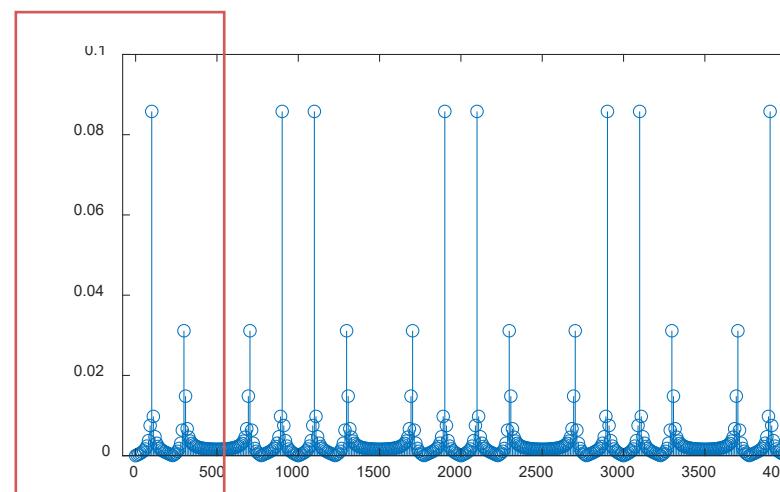
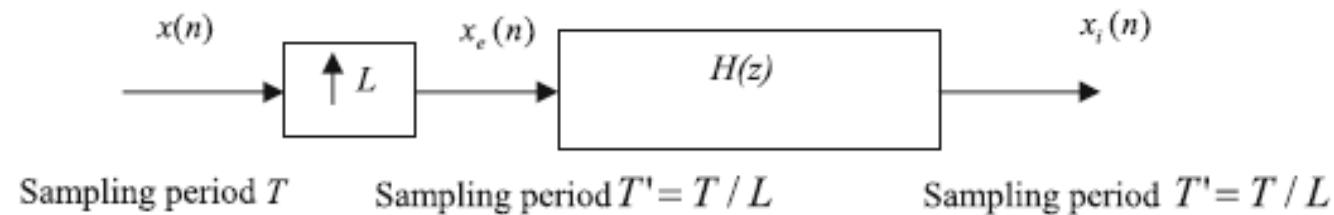
% Compute FFT
N_upsampled = length(x_upsampled_zeros);
f_upsampled = (0:N_upsampled-1) * (fs_upsampled/N_upsampled);

X_upsampled_zeros = fft(x_upsampled_zeros);
figure;
stem(f_upsampled, abs(X_upsampled_zeros)/N_upsampled);
```



# Correction of spectrum

To avoid the repeated spectrum, a low-pass filter  $H(z)$  must be added after the upsampling.



# FIR lowpass filter is used

```
cutoff = fs_original/2; % Cutoff at original Nyquist frequency  
filter_order = 100;  
b = fir1(filter_order, cutoff/(fs_upsampled/2));  
  
% Apply filter to remove spectral images  
x_upsampled_filtered = conv(x_upsampled_zeros, b * upsample_factor, 'same');  
  
t_upsampled_filtered = (0:length(x_upsampled_filtered)-1) / fs_upsampled;  
  
% Plot interpolated signal  
plot(t_upsampled_filtered, x_upsampled_filtered);  
  
% Compute FFT  
N_filtered = length(x_upsampled_filtered);  
f_filtered = (0:N_filtered-1) * (fs_upsampled/N_filtered);  
  
X_upsampled_filtered = fft(x_upsampled_filtered);  
  
stem(f_filtered, abs(X_upsampled_filtered)/N_filtered);
```

