

Assignment 1

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Applied Signal Processing Laboratory 2023



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Introduction

A number of exercises have been performed with the use of MATLAB as an introduction to signal processing. In the following slides a detailed analysis of each of the exercises have been shown with plots, explanations and values

List of exercises:

- Exercise 1.1 : Energy of a rectangular pulse
- Exercise 1.2 : Filtering of 3 sines on the frequency axis
- Exercise 1.3 : Quantization

Exercise 1.1 - Energy of a rectangular pulse

- Choose the integer values of the amplitude A and the duration T of a rectangular pulse
- Fix the total snapshot time: $T_{max} = 10$
- Choose a very high sampling frequency, e.g. $f_s = \frac{1000}{T}$
- Generate and plot the rectangular pulse on the time axis

$$s(t) = AP_T(t) \quad 0 \leq t < T_{max}$$

- Compute analytically the pulse energy.
- Compute the energy on the time axis.

Exercise 1.1 – Plot the pulse on the time axis

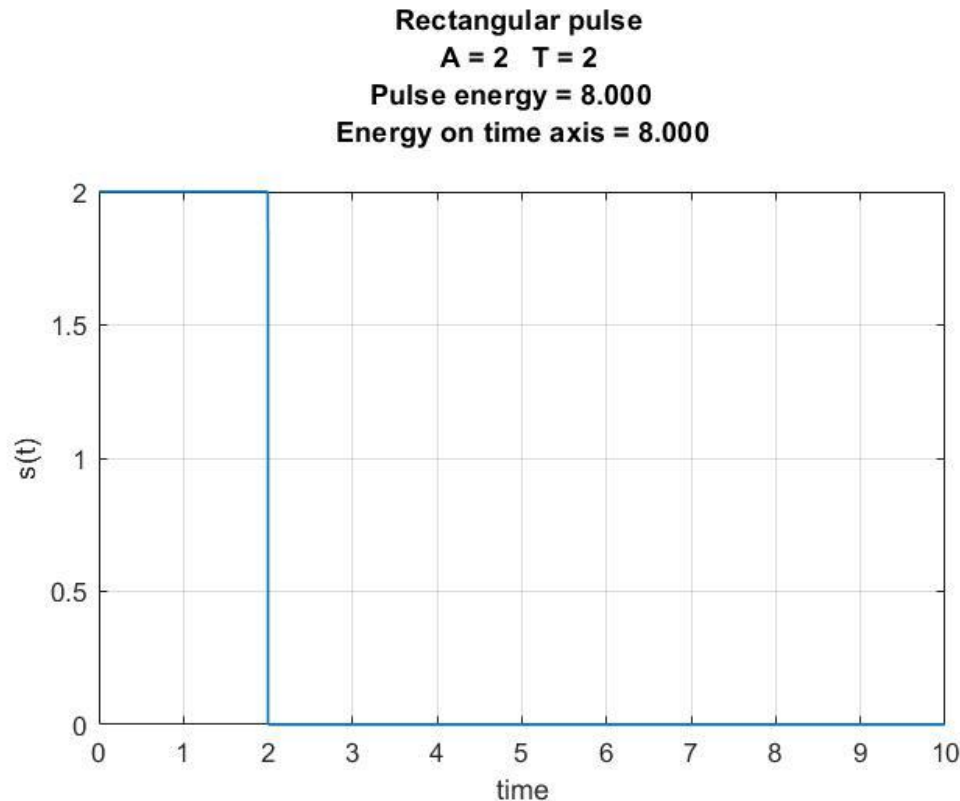


Figure 1.1: Plot of the rectangular pulse with $A = 2$ and $T = 2$, including analytical and true energy values calculated in time domain

The method used to generate the rectangular pulse is shown in the following code:

```
s = zeros(1,N_total);  
s(1:N_pulse) = A;
```

where N_pulse is the number of samples of the rectangular pulse where the signal value will be equal to the amplitude A of the signal assigned and zero elsewhere

Energy of the signal has been calculated using the following formulae:

$$E_{rectangularPulse} = A^2 * T$$

Formula for calculating the energy of a rectangular pulse

```
e_analitical = A^2*T;
```

$$\sum_{t=-\infty}^{+\infty} |s(t)|^2$$

Formula for calculating the energy of a discrete time signal sampled on the time axis

```
e_time = sum(s.^2)*T_s;
```

where T_s is the sampling period

Exercise 1.1 – Plot the pulse on the frequency axis

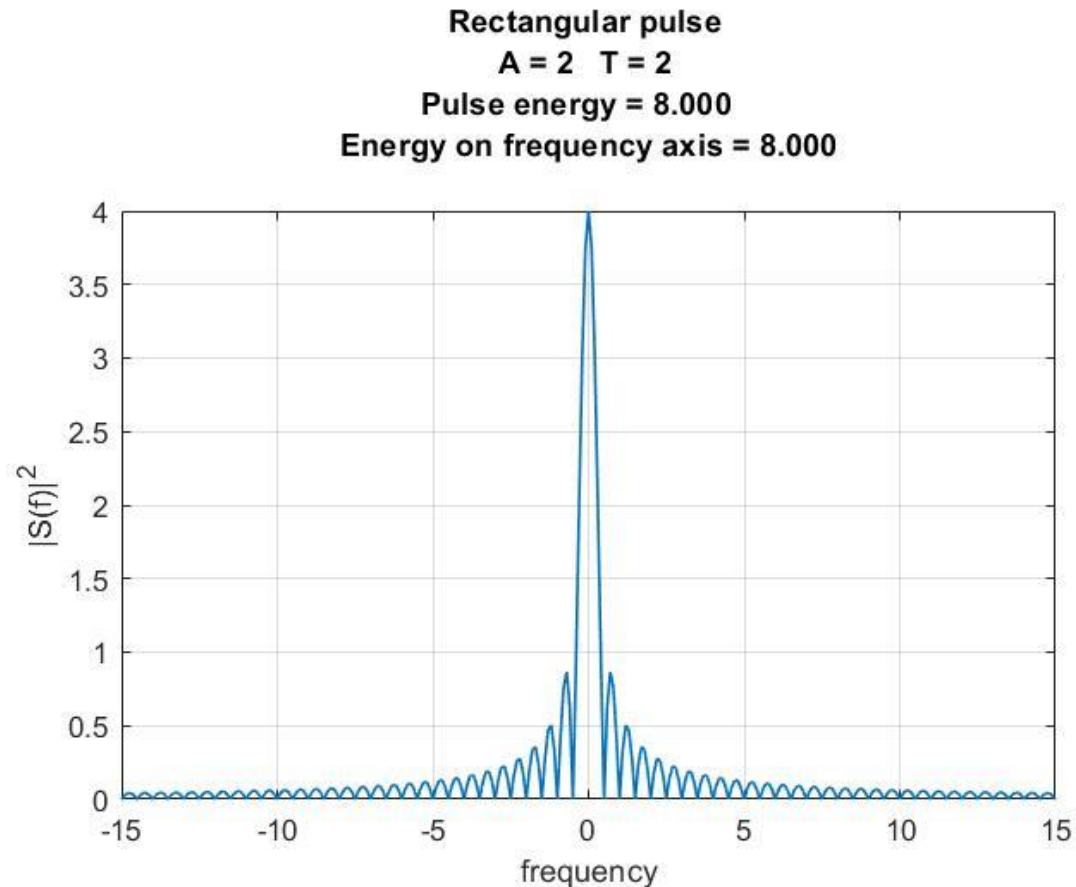


Figure 1.2: Plot of the rectangular pulse in frequency domain with A = 2 and T = 2, including analytical and true energy values calculated in frequency domain

The plot in Figure 1.2 shows the rectangular signal $s(t)$ transformed from time domain into frequency domain using the Fourier Transform, in MATLAB implemented using the `fft` function as:

```
S = fft(s)*T_s;
```

Due to the Parseval's theorem, the energy of a time domain signal equals the energy of a frequency domain signal which has been shown on the plot

$$\sum_{n=-\infty}^{+\infty} |S(f)|^2$$

Formula for calculating the energy in frequency domain

```
e_frequency = sum(M.^2)*f_res;
```

where f_res is the resolution frequency

Exercise 1.1 – Examples with different values of A and T

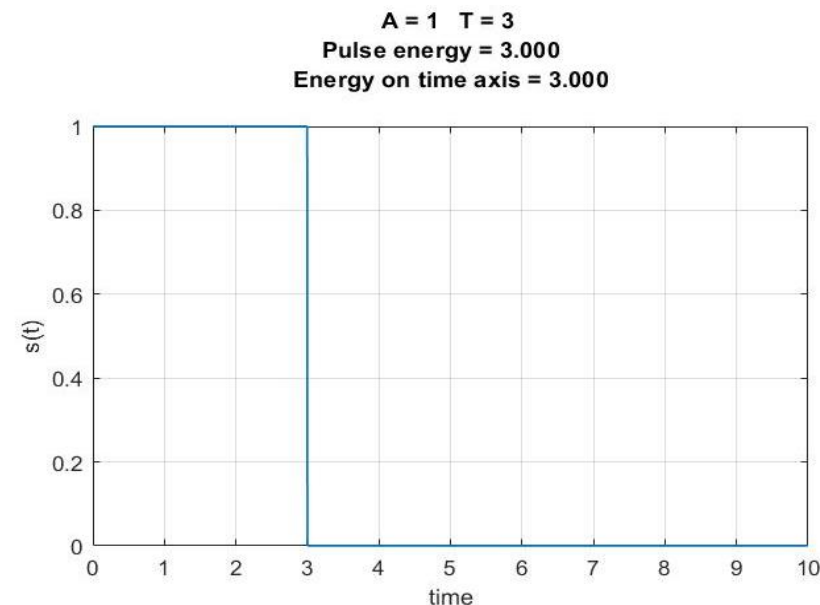
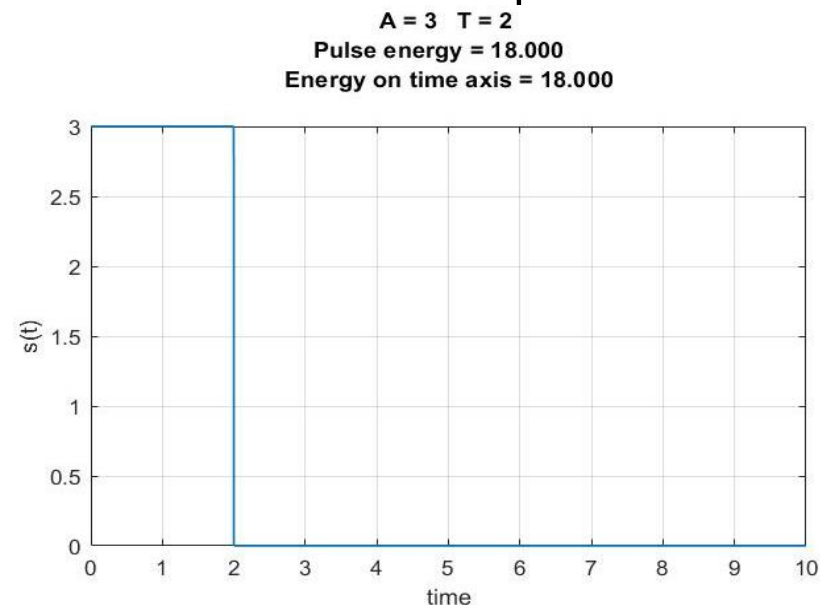
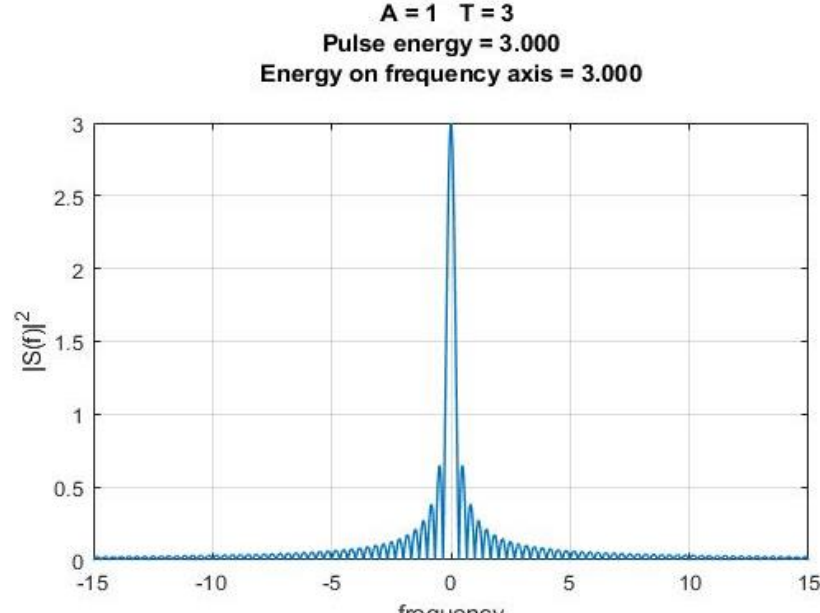
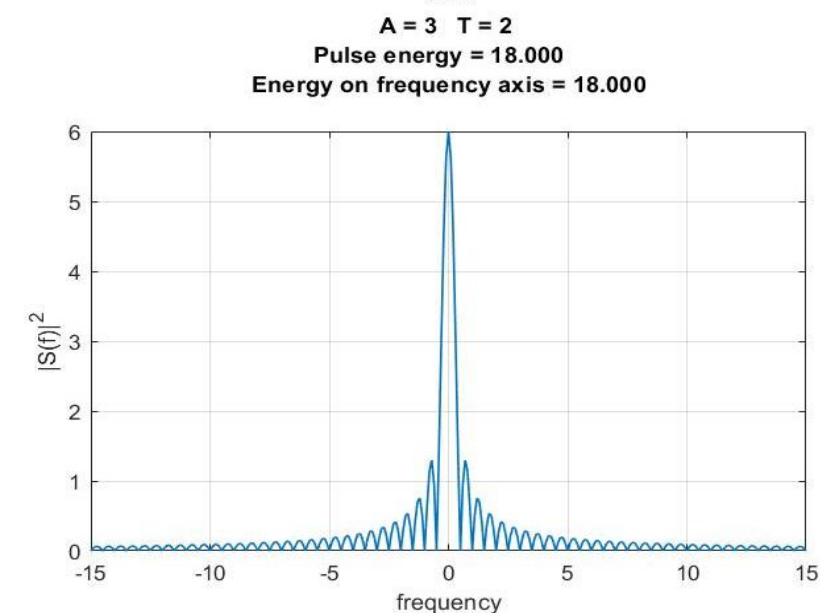


Figure 1.3: plots of the rectangular signal in time and frequency domain with different A and T values

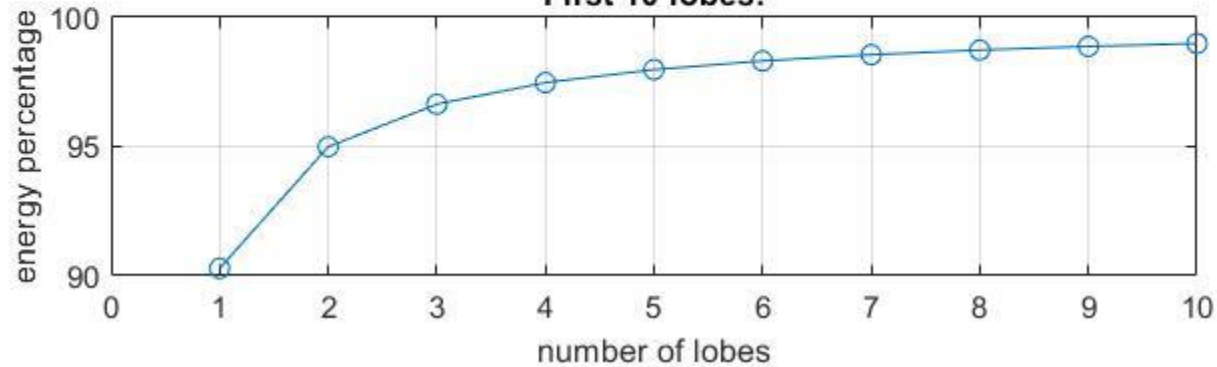
In each of the examples, the energy calculated in the time domain is equal to the energy calculated in frequency domain



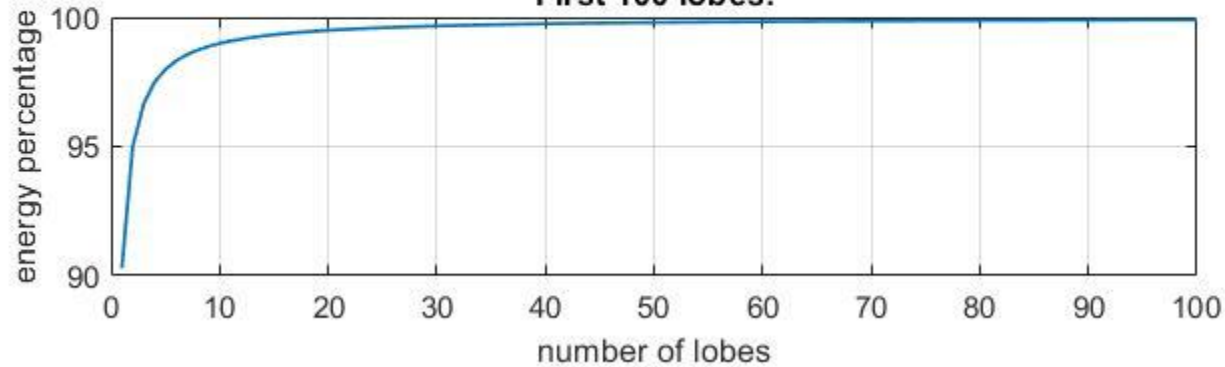
Exercise 1.1 – Percentage of energy contained in the first 10 and 100 lobes

$$A = 2$$
$$T = 2$$

First 10 lobes:



First 100 lobes:



The plots in Figure 1.4 show that approximately 90.3% of the total energy of the rectangular signal is contained in the first lobe:

Example of energy contained in first lobe wrt total energy given $A=2$ and $T=2$:

Total energy: 8

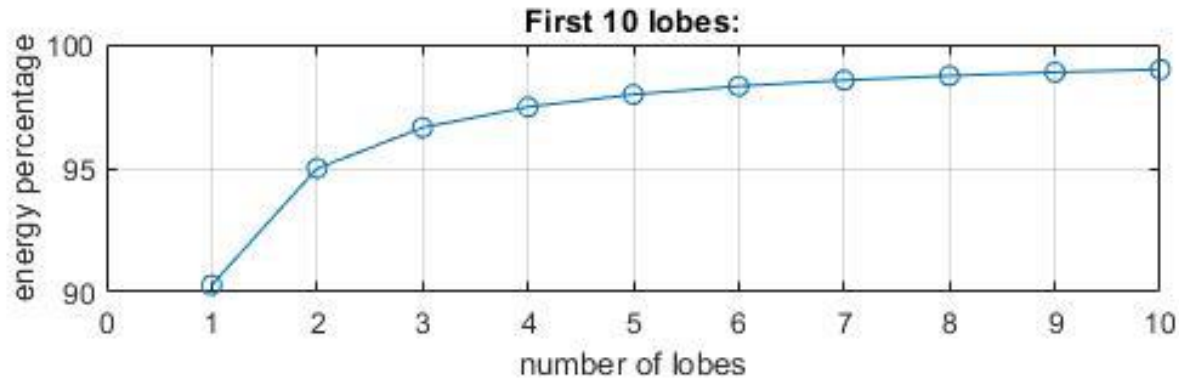
Energy in first lobe: 7.2226

In the first 10 lobes approximately 98.9% of the total energy is contained, whereas in the first 100 lobes 99.9%

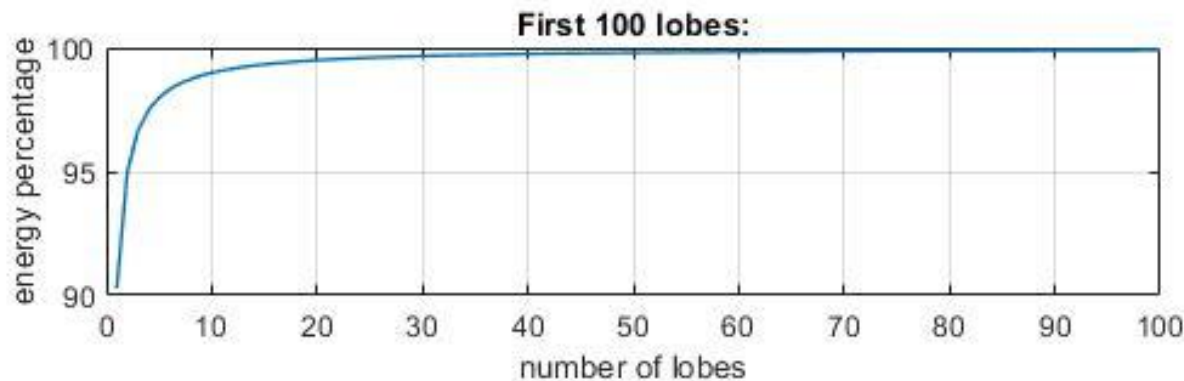
Figure 1.4: energy percentage of the first 10 and the first 100 lobes

Exercise 1.1 – Percentage of energy contained in the first 10 and 100 lobes – different values of A and T

$$A = 1$$
$$T = 3$$



In the Figure 1.5 an example of the percentage of energy contained in the first 10 and 100 lobes with different A and T values: in this case $A=1$ and $T=3$ is shown.



The same trend as in the previous example is followed: regardless of the A and T values chosen, the percentage of energy in the lobes will not change!

Figure 1.5: energy percentage of the first 10 and the first 100 lobes with $A=1$ and $T=3$

Exercise 1.2 – Filtering of 3 sines on the frequency axis

- Consider a signal x obtained as the sum of three sinusoidal signals:

$$x(t) = s_A + s_B + s_C = \sin(2\pi f_A t + \varphi_A) + \sin(2\pi f_B t + \varphi_B) + \sin(2\pi f_C t + \varphi_C)$$

where: $f_A = 1$, $f_B = 2$ and $f_C = 3$ and φ_A , φ_B and φ_C are three random phases

- Use $T_{max} = 10$ and $fsam = 100$

Exercise 1.2 – Plot of $x(t)$ and $|X(f)|$

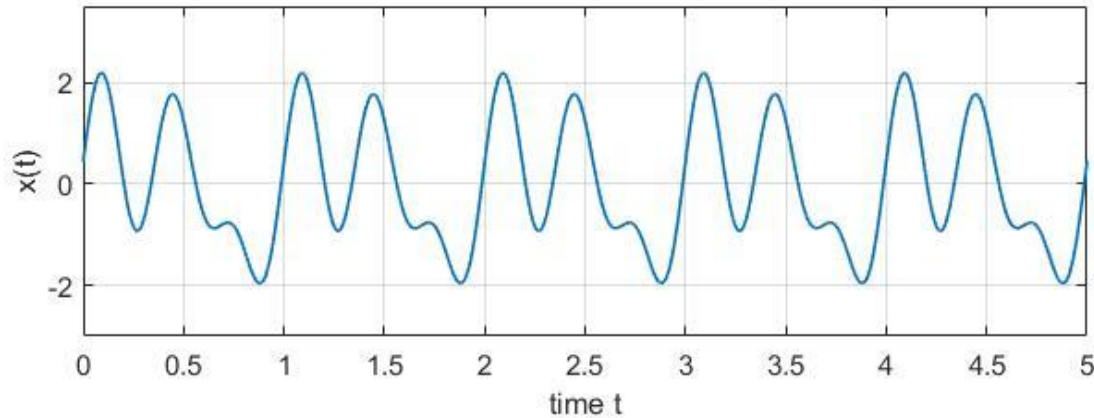


Figure 2.1.1: plot of $x(t)$ wrt. the time axis

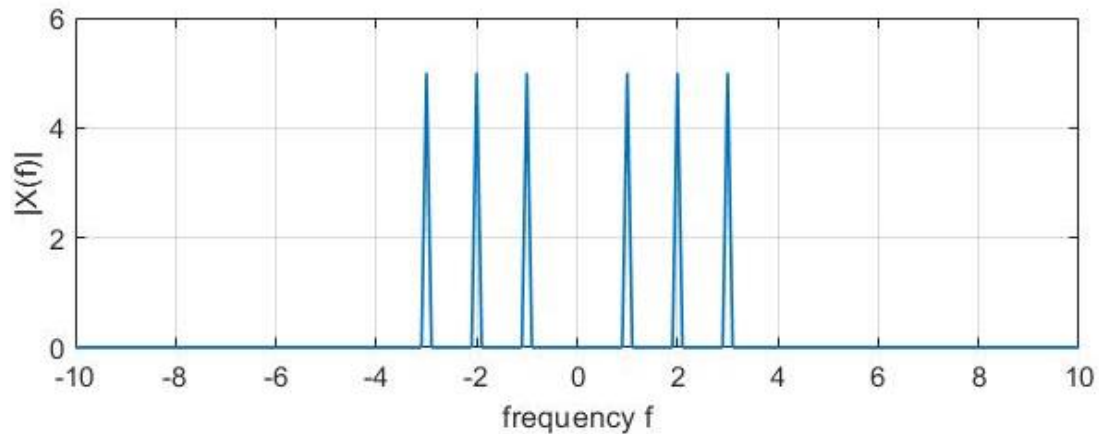


Figure 2.1.2: plot of $|X(f)|$ wrt. the frequency axis

The plot in Figure 2.1.1 shows the signal $x(t)$ obtained by taking the sum of the three sinusoidal signals with three random phase shifts:

Example in Figure 2.1:

$$\begin{aligned}\varphi_A &= 6.0626 \\ \varphi_B &= 0.9903 \\ \varphi_C &= 6.0984\end{aligned}$$

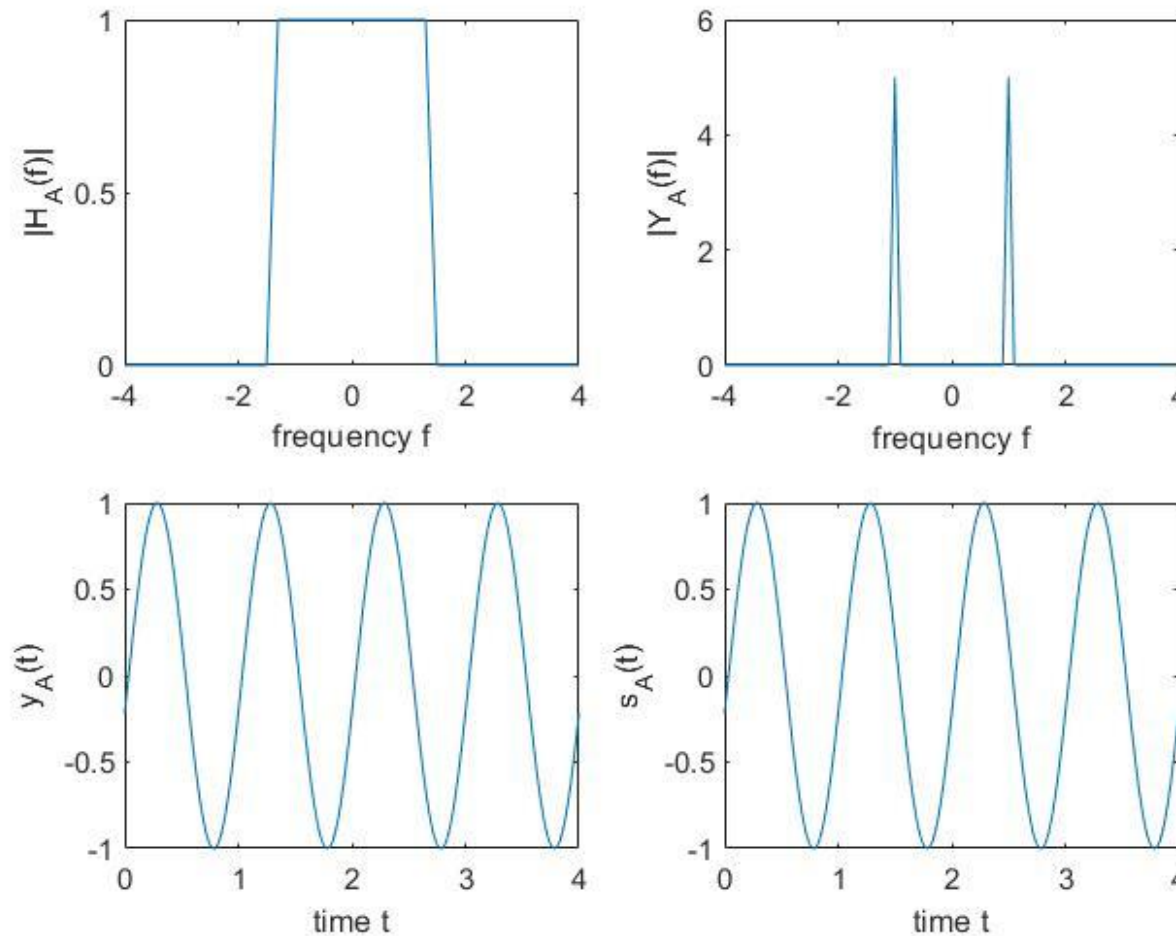
The plot in Figure 2.1.2 shows the magnitude of the frequency components obtained by performing a Fourier transform on the signal $x(t)$, taking its magnitude and shifting it as to be symmetric wrt. the frequency axis.

Since the frequency components for s_A , s_B , s_C are equal to 1, 2, 3 respectively, a pulse can be seen at each of the corresponding frequencies.

Exercise 1.2 – Filtering of 3 sines on the frequency axis

- Design a low-pass filter with frequency response $H_A(f)$ to isolate the sinusoidal signal s_A with frequency f_A
- Design a band-pass filter with frequency response $H_B(f)$ to isolate the sinusoidal signal s_B with frequency f_B
- Design a high-pass filter with frequency response $H_C(f)$ to isolate the sinusoidal signal s_C with frequency f_C
- Multiply $X(f)$ by $H_A(f)$ to obtain $Y_A(f)$
- Compute the ifft for each to obtain $y_A(t)$, $y_B(t)$, $y_C(t)$

Exercise 1.2 – Low-pass filter design and plot of $H_A(f)$, $Y_A(f)$, $y_A(t)$, $s_A(t)$



The low-pass filter has been designed by generating a rectangular signal with width 2.8 and centered about zero.

In MATLAB it has been implemented as:

```
H_A = rectangularPulse(-1.4,1.4,f_symm);
```

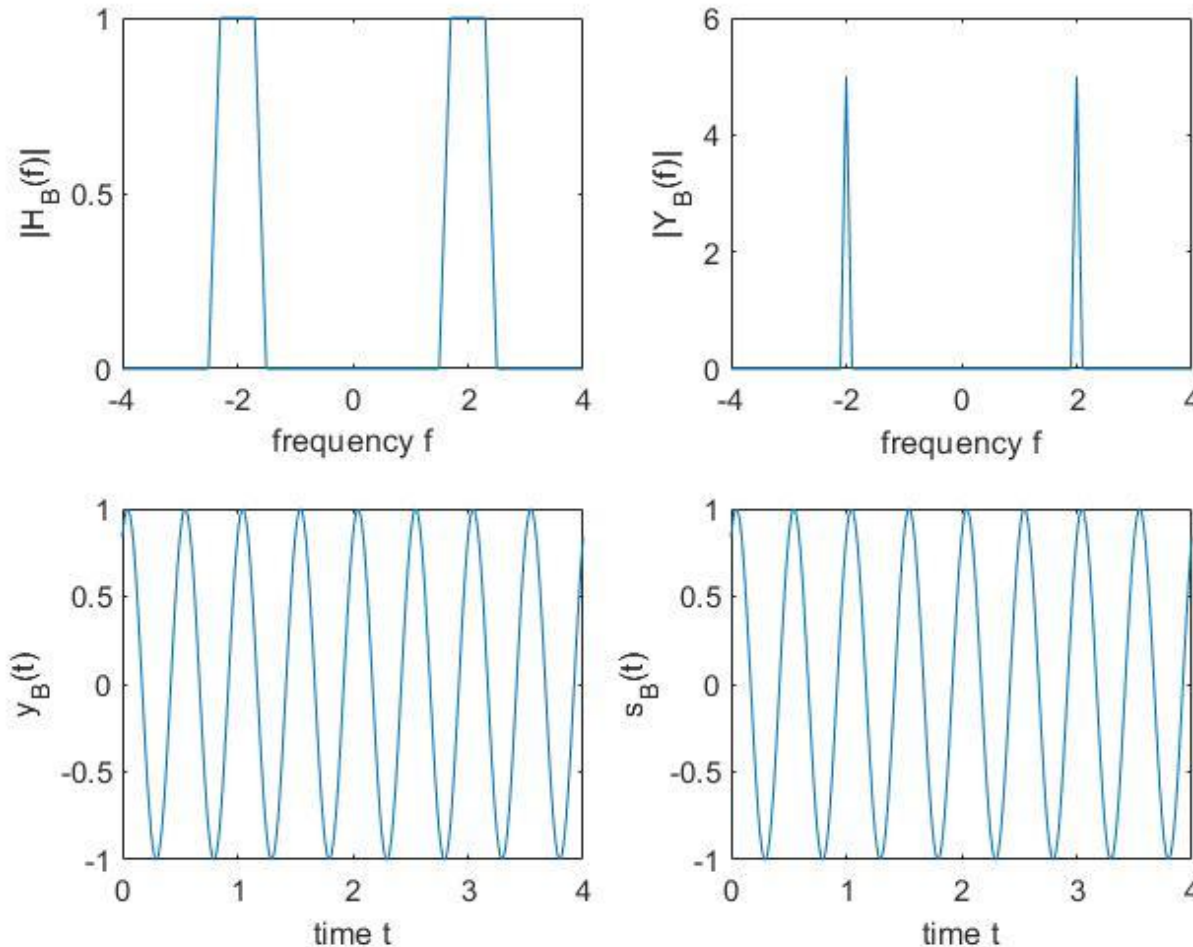
where f_symm is the symmetric frequency axis (centered at zero).

Multiplying $X(f)$ by $H_A(f)$, we obtained $Y_A(f)$.

By performing an ifft on $Y_A(f)$, we obtain the signal component $y_A(t)$ with frequency equal to 1, which is identical to the original signal $s_A(t)$

Figure 2.2: low-pass filter with frequency response $H_A(f)$, isolated frequency component of $Y_A(f)$, signal $y_A(t)$ obtained by ifft and the original signal $s_A(t)$

Exercise 1.2 – Band-pass filter design and plot of $H_B(f)$, $Y_B(f)$, $y_B(t)$, $s_B(t)$



The band-pass filter has been designed by generating and taking a sum of two rectangular signals with width 0.8 each, centered at -2 and 2 respectively.

In MATLAB it has been implemented as:

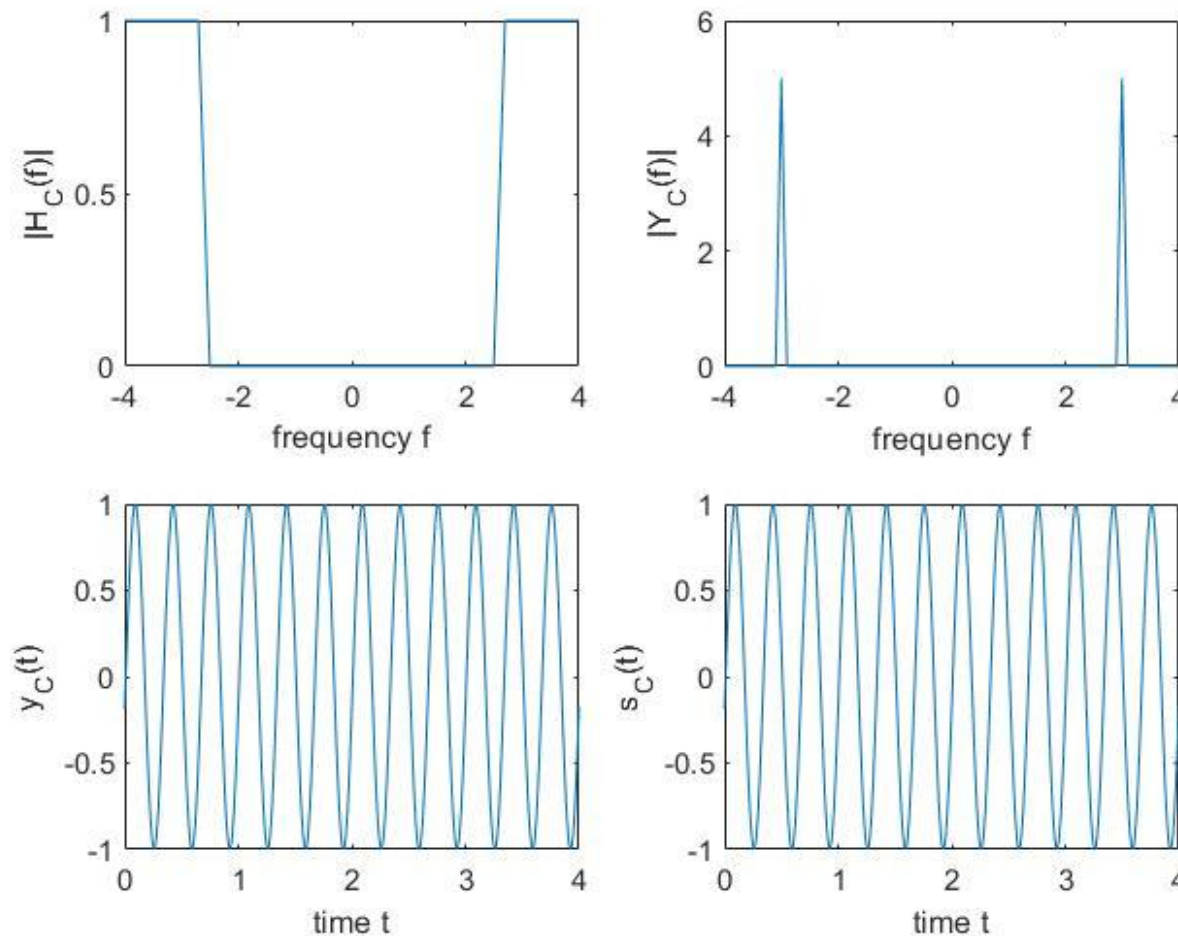
```
H_B =  
rectangularPulse(-2.4,-1.6,f_symm) +  
rectangularPulse(1.6,2.4,f_symm);
```

Multiplying $X(f)$ by $H_B(f)$, we obtained $Y_B(f)$.

By performing an ifft on $Y_B(f)$, we obtain the signal component $y_B(t)$ with frequency equal to 2, which is identical to the original signal $s_B(t)$

Figure 2.3: low-pass filter with frequency response $H_B(f)$, isolated frequency component of $Y_B(f)$, signal $y_B(t)$ obtained by ifft and the original signal $s_B(t)$

Exercise 1.2 – High-pass filter design and plot of $H_C(f)$, $Y_C(f)$, $y_C(t)$, $s_C(t)$



The high-pass filter has been designed by generating a rectangular signal with width 5.2, centered about zero and subtracted from 1.

In MATLAB it has been implemented as:

```
H_C =  
1-rectangularPulse(-2.6,2.6,f_symm);
```

Multiplying $X(f)$ by $H_C(f)$, we obtained $Y_C(f)$.

By performing an ifft on $Y_C(f)$, we obtain the signal component $y_C(t)$ with frequency equal to 3, which is identical to the original signal $s_C(t)$

Figure 2.4: low-pass filter with frequency response $H_C(f)$, isolated frequency component of $Y_C(f)$, signal $y_C(t)$ obtained by ifft and the original signal $s_C(t)$

Exercise 1.3 – Quantization

- Consider a time axis $0 \leq t < T_{max} = 1$
- Consider a sinusoidal signal with $A = 1$ and a given frequency value f_0 (e.g., $f_0 = 1$):

$$s(t) = A \cos(2\pi f_0 t)$$

- Choose a very high sampling frequency above the Nyquist frequency (e.g., $f_{sam} = 100$).
- Fix the number of quantization bits m .
- Divide the amplitude domain into 2^m segments. Use the value in the middle of the segment to represent it by using the Matlab **quantiz** function:

$$[\text{index}, \text{sq}] = \text{quantiz}(s, \text{partition}, \text{codebook})$$

where:

partition is a vector with $2^m - 1$ values

codebook is a vector with 2^m values

Exercise 1.3 – Plot of the original signal $s(t)$ vs. quantized sinusoidal signal $s_q(t)$

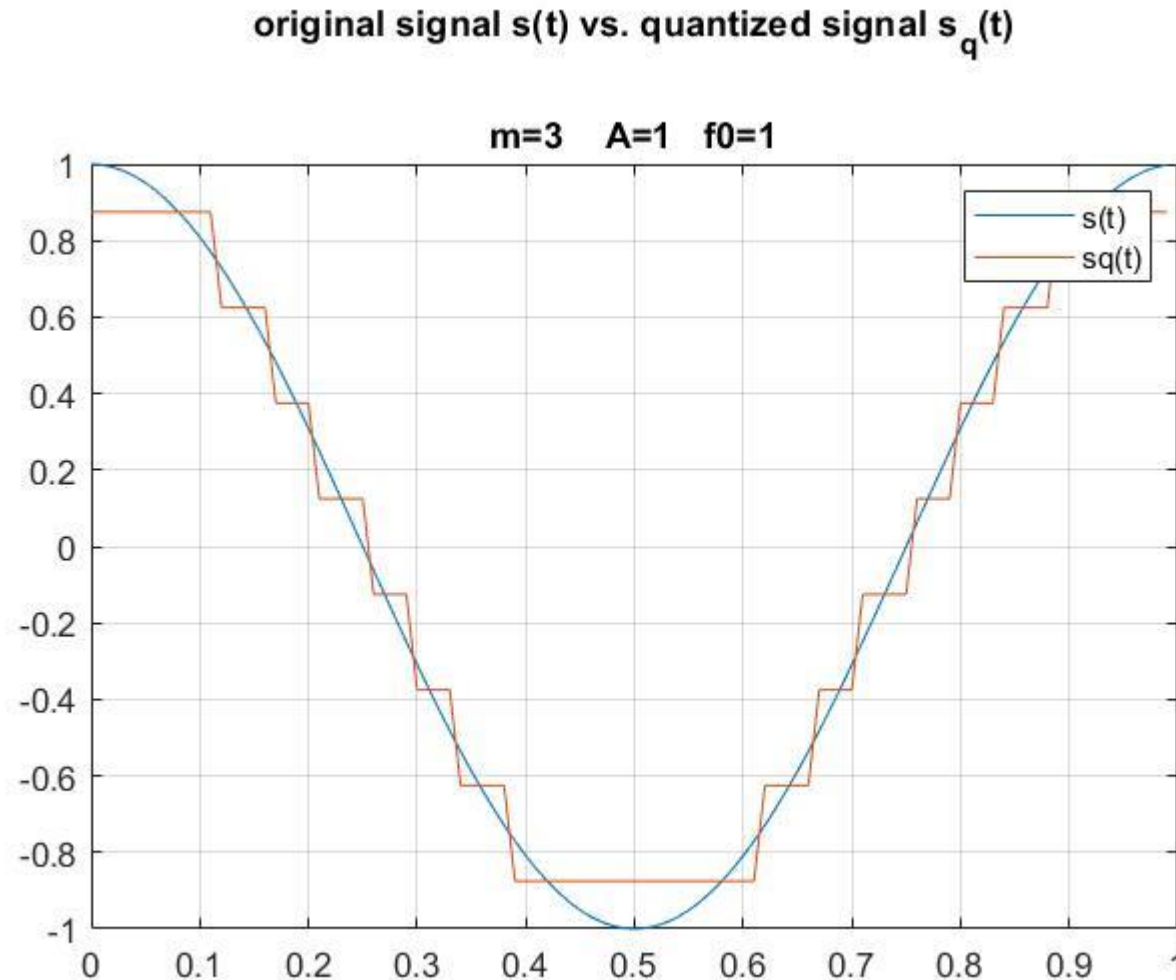


Figure 3.1: the original signal $s(t)$ vs. quantized sinusoidal signal $s_q(t)$ with the quantization bits m equal to 3, $A=1$ and $f_0=1$

The following parameters have been chosen in generating $s(t)$ and $s_q(t)$:

Quantization bits $m = 3$

Signal amplitude $A = 1$

Signal frequency $f_0 = 1$

To calculate the length of each segment using MATLAB, we introduce a value δ such that:

$$\delta = (\max(s) - \min(s)) / (2^m);$$

The partition and codebook are then calculated as:

```
partition=[-((2^m)/2-1)*delta:delta:((2^m)/2-1)*delta];
codebook=[-((2^m)/2-1)*delta-delta/2
          :delta:
          ((2^m)/2-1)*delta+delta/2];

[index,sq] = quantiz(s,partition,codebook);
```


Exercise 1.3 – Comparing the spectrum of the original and the quantized signal

In Figure 3.2 it can be seen that in the spectrum of the quantized signal $|S_q(f)|^2$ in dB there are additional frequency components that do not appear in the spectrum of the original signal $|S(f)|^2$ in dB.

Those frequencies are present due to noise produced by the quantization process of the original signal.

An important quantity to introduce is the SNR (signal-to-noise ratio) which compares the levels of the desired signal vs. the level of noise.

The expected value for SNR is $\approx 6m$ [dB]

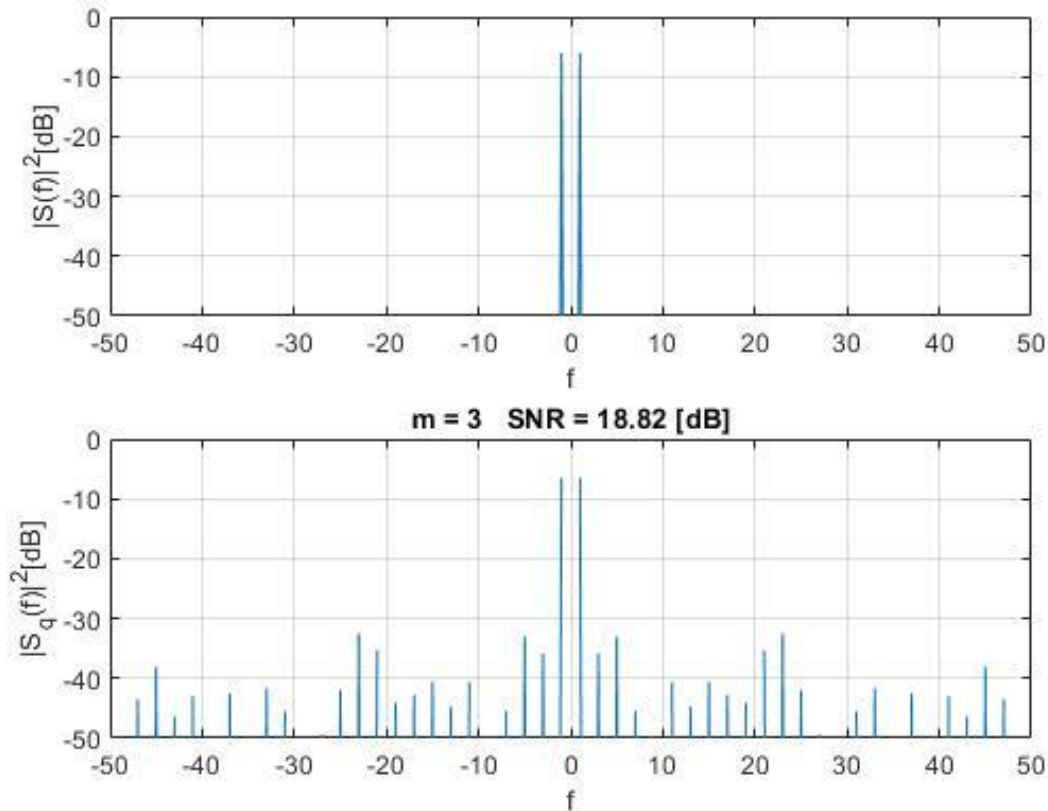


Figure 3.2: the spectrum original signal vs the spectrum of the quantized sinusoidal signal with $m = 3$ and SNR = 18.82 [dB] $\approx 6m$ [dB]

Exercise 1.3 – Computing the SNR

- Given the quantized signal, compute the energy E_s of the frequency component f_0 , the energy E_n of the noise components at frequencies $f \neq f_0$ and their ratio in dB:

$$SNR = 10 \log_{10} \frac{E_s}{E_n} [dB]$$

The following MATLAB code performs the calculation:

```
E_s = max(SMQ);  
New_s = SM - SMQ;  
E_n = sum(New_s.^2)*f_res;  
SNR1 = 10*log10(E_s/E_n);
```

- The SNR can also be computed using the formula: $SNR = 10 \log_{10}(1.5 * 4^m) [dB]$

Exercise 1.3 – SNR for different values of m

In the following plots examples for $m = 4$ and $m = 5$ are shown.

For each, the SNR is indeed $\simeq 6m$ [dB]: $m = 4$, SNR = 24.92 [dB]

$m = 5$, SNR = 30.06 [dB]

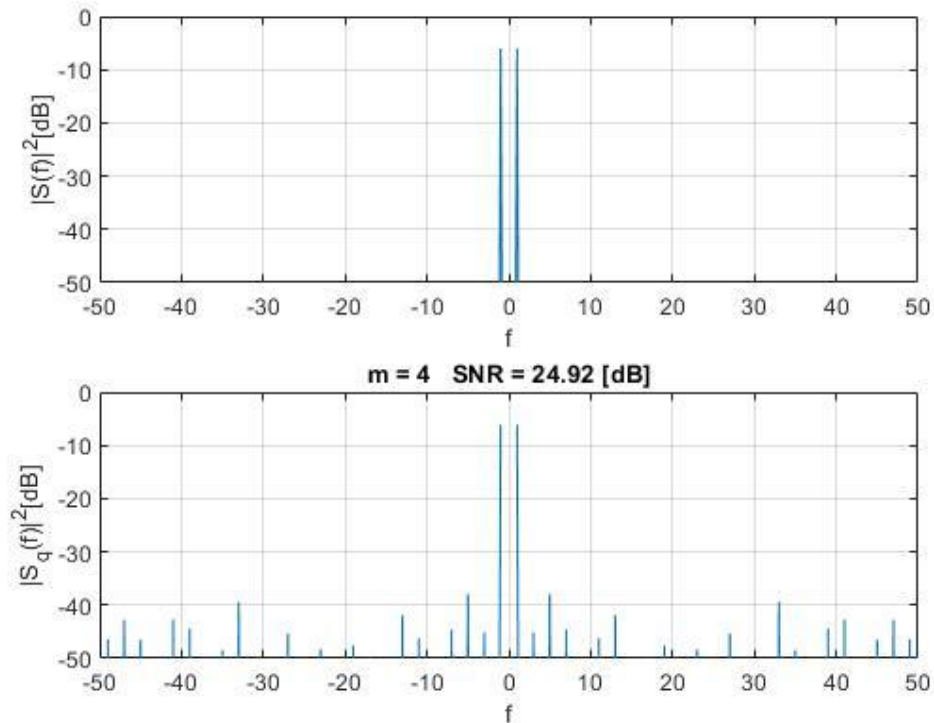


Figure 3.3: the spectrum original signal vs the spectrum of the quantized sinusoidal signal with $m = 4$

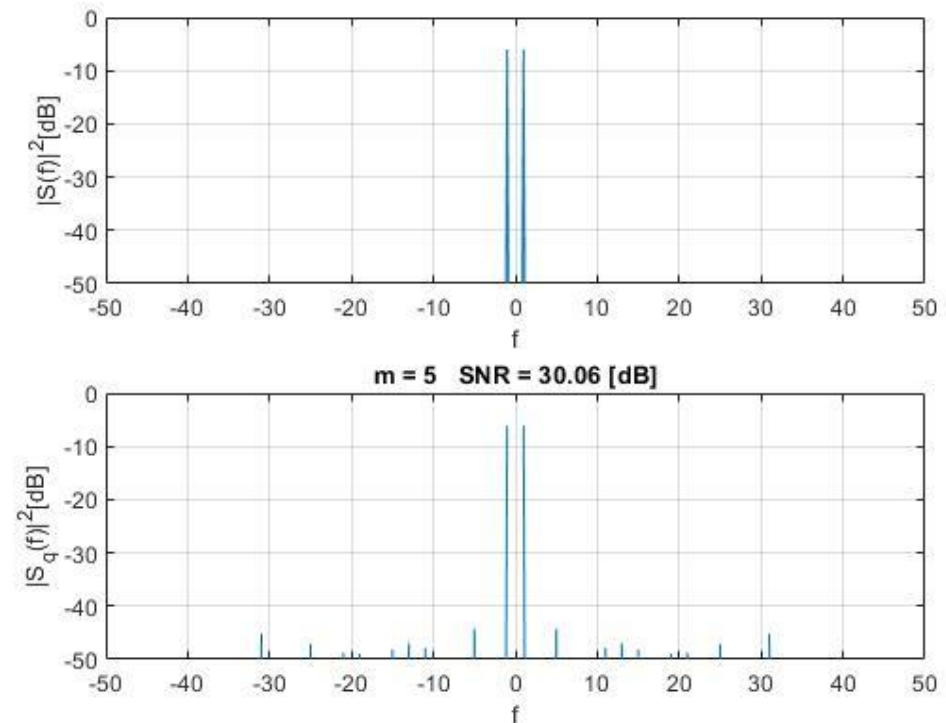


Figure 3.4: the spectrum original signal vs the spectrum of the quantized sinusoidal signal with $m = 5$