

In [1]:

```
import numpy as np
import matplotlib.pyplot as plt
%matplotlib inline
```

Task 1.a

Compute the Fourier transform for the function: $f(x) = xe^{-\alpha|x|}, \alpha > 0$

$$F(\omega) = \int_{-\infty}^{+\infty} f(x)e^{-j\omega x} dx = \int_{-\infty}^0 xe^{\alpha x}e^{-j\omega x} dx + \int_0^{+\infty} xe^{-\alpha x}e^{-j\omega x} dx = I_1 + I_2$$

$$\begin{aligned} I_1 &= \int_{-\infty}^0 xe^{(\alpha-j\omega)x} dx = \frac{1}{\alpha-j\omega} \int_{-\infty}^0 xde^{(\alpha-j\omega)x} = \\ &= \frac{1}{\alpha-j\omega} \left[xe^{(\alpha-j\omega)x} \Big|_{-\infty}^0 - \frac{1}{\alpha-j\omega} \int_{-\infty}^0 e^{(\alpha-j\omega)x} d((\alpha-j\omega)x) \right] = \\ &= \frac{1}{\alpha-j\omega} \left[0 - \frac{1}{\alpha-j\omega} xe^{(\alpha-j\omega)x} \Big|_{-\infty}^0 \right] = \\ &= \frac{1}{\alpha-j\omega} \left[-\frac{1}{\alpha-j\omega} (1-0) \right] = \frac{-1}{(\alpha-j\omega)^2} \end{aligned}$$

Note:

$$\begin{aligned} xe^{(\alpha-j\omega)x} \Big|_{-\infty}^0 &= 0 - \lim_{x \rightarrow -\infty} xe^{(\alpha-j\omega)x} = \\ &= - \lim_{x \rightarrow -\infty} \frac{x}{e^{-(\alpha-j\omega)x}} = [\frac{\infty}{\infty}] = \text{by L'Hopital's rule} = \\ &= - \lim_{x \rightarrow -\infty} \frac{1}{-(\alpha-j\omega)e^{-(\alpha-j\omega)x}} = 0 \end{aligned}$$

$$\begin{aligned} I_2 &= \int_0^{+\infty} xe^{-(\alpha+j\omega)x} dx = -\frac{1}{\alpha+j\omega} \int_0^{+\infty} xde^{-(\alpha+j\omega)x} = \\ &= -\frac{1}{\alpha+j\omega} \left[xe^{-(\alpha+j\omega)x} \Big|_0^{+\infty} + \frac{1}{\alpha+j\omega} \int_0^{+\infty} e^{-(\alpha+j\omega)x} d(-(\alpha+j\omega)x) \right] = \\ &= -\frac{1}{(\alpha+j\omega)^2} (0-1) = \frac{1}{(\alpha+j\omega)^2} \end{aligned}$$

Note:

$$\begin{aligned} xe^{-(\alpha+j\omega)x} \Big|_0^{+\infty} &= \lim_{x \rightarrow +\infty} xe^{-(\alpha+j\omega)x} - 0 = \\ &= \lim_{x \rightarrow +\infty} \frac{x}{e^{(\alpha+j\omega)x}} = [\frac{\infty}{\infty}] = \text{by L'Hopital's rule} = \\ &= \lim_{x \rightarrow +\infty} \frac{1}{(\alpha+j\omega)e^{(\alpha+j\omega)x}} = 0 \end{aligned}$$

Finally:

$$\begin{aligned} I &= I_2 + I_1 = \frac{1}{(\alpha + j\omega)^2} + \frac{-1}{(\alpha - j\omega)^2} = \\ &= \frac{\alpha^2 - 2j\omega\alpha - \omega^2 - \alpha^2 - 2j\omega + \omega^2}{(\alpha + j\omega)(\alpha + j\omega)(\alpha - j\omega)(\alpha - j\omega)} = \frac{-4j\omega\alpha}{(\alpha^2 + \omega^2)^2} \end{aligned}$$

Answer:

$$F(\omega) = \frac{-4j\omega\alpha}{(\alpha^2 + \omega^2)^2}$$

Task 1.b

Compute the Fourier transform for the function: $f(x) = e^{-a^2x^2} \cos(bx)$

Here I will use the shifting property of FT. For this reason I will start my calculations from computing the FT of function: $f(x) = e^{-a^2x^2}$ and then will apply shift to the result.

$$\begin{aligned} F_1(\omega) &= \int_{-\infty}^{+\infty} e^{-a^2x^2} e^{-j\omega x} dx = \\ &= \int_{-\infty}^{+\infty} e^{-(a^2x^2 + j\omega x - \frac{\omega^2}{4a^2} + \frac{\omega^2}{4a^2})} dx = \\ &= e^{-\frac{\omega^2}{4a^2}} \int_{-\infty}^{+\infty} e^{-\left(|a|x + \frac{j\omega}{2|a|}\right)^2} dx = \\ &= e^{-\frac{\omega^2}{4a^2}} \frac{1}{|a|} \int_{-\infty}^{+\infty} e^{-\left(|a|x + \frac{j\omega}{2|a|}\right)^2} d\left(|a|x + \frac{j\omega}{2|a|}\right) = \frac{\sqrt{\pi}}{|a|} e^{-\frac{\omega^2}{4a^2}} \end{aligned}$$

Applying shifting property:

$$\begin{aligned} F(\omega) &= \int_{-\infty}^{+\infty} e^{-a^2x^2} \cos(bx) e^{-j\omega x} dx = \frac{1}{2} \int_{-\infty}^{+\infty} e^{-a^2x^2} (e^{jbx} + e^{-jbx}) e^{-j\omega x} dx = \\ &= \frac{1}{2} \left(\int_{-\infty}^{+\infty} e^{-a^2x^2} e^{-j(\omega-b)x} dx + \int_{-\infty}^{+\infty} e^{-a^2x^2} e^{-j(\omega+b)x} dx \right) = \\ &= \frac{1}{2} \left(\frac{\sqrt{\pi}}{|a|} e^{-\frac{(\omega-b)^2}{4a^2}} + \frac{\sqrt{\pi}}{|a|} e^{-\frac{(\omega+b)^2}{4a^2}} \right) = \frac{\sqrt{\pi}}{2|a|} \left(e^{-\frac{(\omega-b)^2}{4a^2}} + e^{-\frac{(\omega+b)^2}{4a^2}} \right) \end{aligned}$$

Answer:

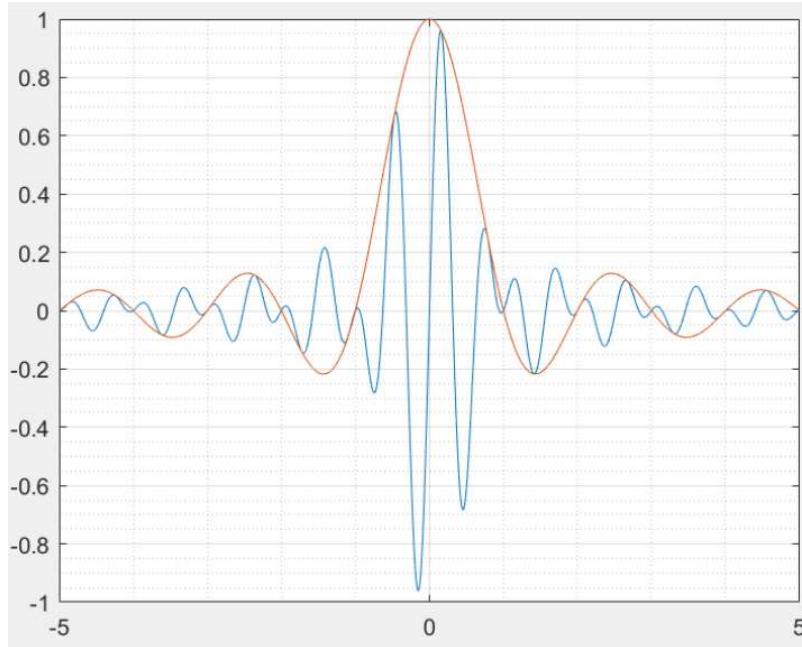
$$F(\omega) = \frac{\sqrt{\pi}}{2|a|} \left(e^{-\frac{(\omega-b)^2}{4a^2}} + e^{-\frac{(\omega+b)^2}{4a^2}} \right)$$

Task 2

Compute the spectrum of the cosine-filled sinc function:

In [149...]

```
#plot picture from HW file
from IPython.display import Image
display(Image(filename='sinc.png', width = 400))
```



First of all, I defined the cosine-filled sinc function: $f(x) = \frac{\sin(\pi x)}{\pi x} \cos(\omega_0 x + \varphi_0)$. I write ω_0, φ_0 because I want to derive the representation of cosine-filled sinc function in the frequency domain in general case.

By the way, it is obvious that for our case $\varphi_0 = -\frac{\pi}{2}$ (because cosine is delayed ($f(0) = 0$)) and $\omega_0 = 3\pi$

Proof: $f(x) = \frac{\sin(\pi x)}{\pi x} \cos(\omega_0 x - \frac{\pi}{2}) = 0$,

then

$$\begin{cases} \text{sinc}(x) = 0 \\ \cos(\omega_0 x - \frac{\pi}{2}) = 0 \end{cases}$$

from the second equation we get

$$\begin{aligned} \omega_0 x - \frac{\pi}{2} &= \frac{\pi}{2} + \pi k, k \in N \\ \omega_0 x &= \pi + \pi k \end{aligned}$$

Then according to the graph above I can say that when $x = \frac{1}{3}$ cosine-filled function equals to zero, so the first root (when $k = 0$) of the equation above can be find in this way:

$$\begin{aligned} \frac{\omega_0}{3} &= \pi \\ \omega_0 &= 3\pi \end{aligned}$$

$$\begin{aligned} F(\omega) &= \int_{-\infty}^{+\infty} \text{sinc}(x) \cos(\omega_0 x + \varphi_0) e^{-j\omega x} dx = \\ &= \int_{-\infty}^{+\infty} \text{sinc}(x) \left(\frac{e^{j(\omega_0 x + \varphi_0)} + e^{-j(\omega_0 x + \varphi_0)}}{2} \right) \cdot e^{-j\omega x} dx = \\ &= \frac{1}{2} \left(\int_{-\infty}^{+\infty} e^{j\varphi_0} \text{sinc}(x) e^{-j(\omega - \omega_0)x} dx + \int_{-\infty}^{+\infty} e^{-j\varphi_0} \text{sinc}(x) e^{-j(\omega + \omega_0)x} dx \right) = I_1 + I_2 \end{aligned}$$

$$I_1 = \frac{1}{2} e^{j\varphi_0} \int_{-\infty}^{+\infty} \text{sinc}(x) e^{-j(\omega - \omega_0)x} dx = \begin{cases} \frac{1}{2} e^{j\varphi_0}, |\omega - \omega_0| \leq \pi \\ 0, \text{otherwise} \end{cases}$$

$$I_2 = \frac{1}{2} e^{-j\varphi_0} \int_{-\infty}^{+\infty} \text{sinc}(x) e^{-j(\omega + \omega_0)x} dx = \begin{cases} \frac{1}{2} e^{-j\varphi_0}, |\omega + \omega_0| \leq \pi \\ 0, \text{otherwise} \end{cases}$$

For our case $\omega_0 = 3\pi$, $\varphi_0 = -\frac{\pi}{2}$ we get:

$$I_1 = \begin{cases} \frac{j}{2}, 2\pi \leq \omega \leq 4\pi \\ 0, \text{otherwise} \end{cases}$$

$$I_2 = \begin{cases} -\frac{j}{2}, -4\pi \leq \omega \leq -2\pi \\ 0, \text{otherwise} \end{cases}$$

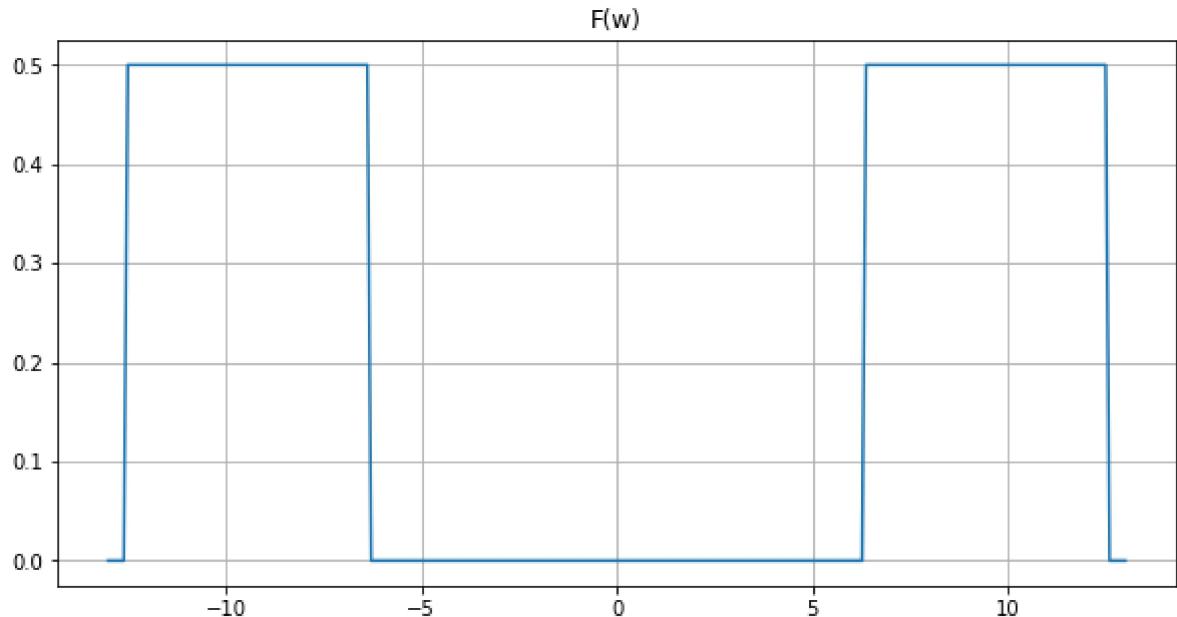
Finally:

$$I = \begin{cases} -\frac{j}{2}, -4\pi \leq \omega \leq -2\pi \\ \frac{j}{2}, 2\pi \leq \omega \leq 4\pi \\ 0, \text{otherwise} \end{cases}$$

Now let's plot the graph. Firstly, we need to find the magnitude of the signal in frequency domain. In our case it is obvious that $A = \frac{1}{2}$ (by definition $A(\omega) = \sqrt{F(\omega)F^*(\omega)}$)

```
In [3]: x = np.linspace(-13, 13, 260)
y = []
for point in x:
    if (-4 * np.pi) <= point <= -2 * np.pi:
        y.append(0.5)
    elif (2 * np.pi) <= point <= 4 * np.pi: # of course, I can fill y
                                                # List in one if clause
        y.append(0.5)                         #but I did it in this
                                                #way for the better understanding
    else:
        y.append(0)
```

```
In [4]: fig, ax = plt.subplots(1, 1, figsize=(10,5))
ax.plot(x, y)
ax.grid()
ax.set_title('F(w)');
```



We received the rectangular window in frequency domain shifted by $\omega_0 = 3\pi$ with two times decreased magnitude.

Answer:

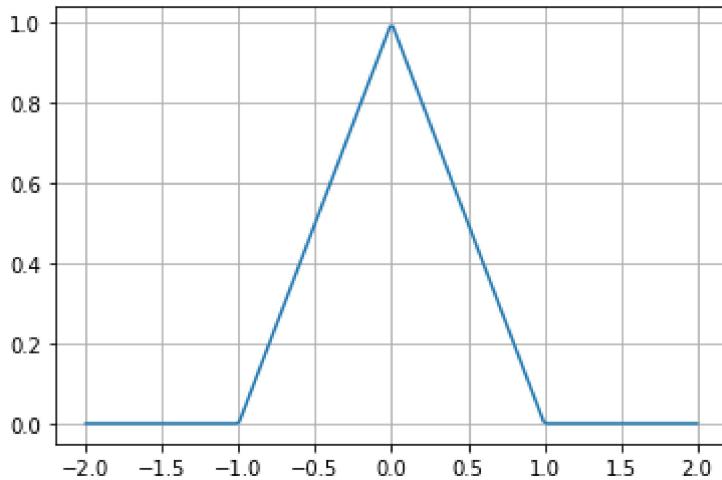
$$I = \begin{cases} -\frac{j}{2}, -4\pi \leq \omega \leq -2\pi \\ \frac{j}{2}, 2\pi \leq \omega \leq 4\pi \\ 0, \text{otherwise} \end{cases}$$

Task 3

Compute the auto-correlation of the triangle pulse:

```
In [5]: x = np.linspace(-2, 2, 200)
y = []
for point in x:
    if (-1 <= point <= 0):
        y.append(point + 1)
    elif (0 < point <= 1):
        y.append(-point + 1)
    else:
        y.append(0)
```

```
In [6]: plt.plot(x, y) # just check that everything is ok
plt.grid()
```



```
In [7]: autocorr = np.correlate(y, y, 'full') # calculate an autocorrelation of the given fu
```

```
In [8]: print(f'length x {len(x)}')
print(f'length autocorr {len(autocorr)}')
```

length x 200
length autocorr 399

```
In [9]: ac = autocorr[100:300] # cut zeros from both sides
#(because the Length is 2 * Len(x) - 1 and I want Len(autocorr) = Len(x))
```

```
In [10]: len(ac)
```

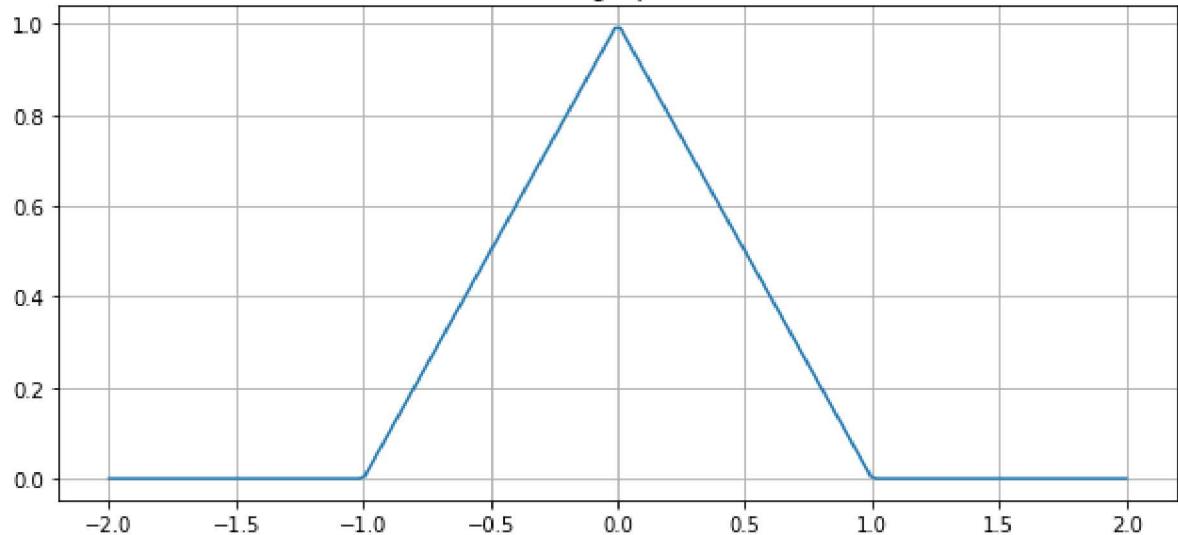
Out[10]: 200

```
In [11]: ac_norm = ac / max(ac) # normalize the autocorrelation
```

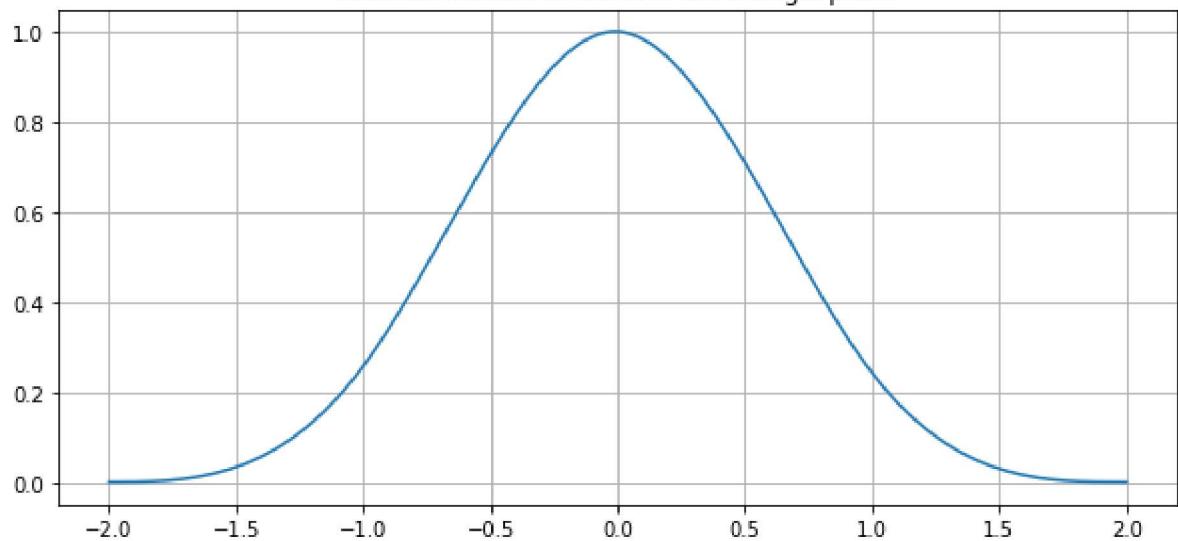
Answer

```
In [12]: fig, ax = plt.subplots(2, 1, figsize=(10,10))
ax[0].plot(x, y)
ax[0].grid(True)
ax[0].set_title('Triangle pulse')
ax[1].plot(x, ac_norm)
ax[1].grid(True)
ax[1].set_title('Auto-correlation function of the triangle pulse');
```

Triangle pulse



Auto-correlation function of the triangle pulse



In []:

In []:

Task 4.a

Compute (by hands) the convolution of the following signals:

$$\mathbf{h} = [2, 3, 6, 8], \mathbf{x} = [1, 2, 10, 1]$$

By definition of the discrete convolution:

$$(f * g)[n] = \sum_{m=-M}^{M} f[n-m]g[m]$$

Firstly I flip h signal:

$$h_{flipped} = [8, 6, 3, 2]$$

Then calculate the convolution:

1) Step 1:

$$\begin{array}{r} 8632 \\ \cdot \quad \quad \quad 12101 \\ 2 \cdot 1 = 2 \end{array}$$

2) Step 2:

$$\begin{array}{r}
 8632 \\
 \cdot\cdot\cdot\cdot \\
 12101 \\
 3 \cdot 1 + 2 \cdot 2 = 3 + 4 = 7
 \end{array}$$

3) Step 3:

$$\begin{array}{r}
 & 8 & 6 & 3 & 2 \\
 & \dots & 1 & 2 & 10 & 1 \\
 6 \cdot 1 + 3 \cdot 2 + 2 \cdot 10 = 6 + 6 + 20 = 32
 \end{array}$$

4) Step 4:

$$\begin{array}{r}
 & 8 & 6 & 3 & 2 \\
 & 1 & 2 & 10 & 1 \\
 8 \cdot 1 + 6 \cdot 2 + 3 \cdot 10 + 2 \cdot 1 & = 8 + 12 + 30 + 2 = 52
 \end{array}$$

5) Step 5:

$$\begin{array}{r} \dots\dots 8\ 6\ 3\ 2 \\ \quad\quad\quad 1\ 2\ 10\ 1 \\ 8 \cdot 2 + 6 \cdot 10 + 3 \cdot 1 = 16 + 60 + 3 = 79 \end{array}$$

6) Step 6:

$$\begin{array}{r} & 8 & 6 & 3 & 2 \\ \cdots & & & & \\ & 1 & 2 & 10 & 1 \\ 8 \cdot 10 + 6 \cdot 11 + 3 \cdot 1 & = 80 + 6 & = 86 \end{array}$$

7) Step 7:

$$\begin{array}{r} \dots \quad 8 \ 6 \ 3 \ 2 \\ \quad 1 \ 2 \ 10 \ 1 \\ 8 \cdot 1 \equiv 8 \end{array}$$

Answer: $\text{conv} = [2, 7, 32, 52, 79, 86, 8]$

Task 4.b

Compute (by hands) the convolution of the following signals:

h = [5, 1, 3, 10], x =[9, 6, 10, 1]

By definition of the discrete convolution:

$$(f * g)[n] = \sum_{m=-M}^M f[n-m]g[m]$$

Firstly I flip h signal:

$$h_{flipped} = [10, 3, 1, 5]$$

Then calculate the convolution:

1) Step 1:

$$\begin{array}{r} 10 \ 3 \ 1 \ 5 \\ \cdots \cdots \cdots \cdots \\ 9 \ 6 \ 10 \ 1 \\ 5 \cdot 9 = 45 \end{array}$$

2) Step 2:

$$\begin{array}{r} 10 \ 3 \ 1 \ 5 \\ \cdots \cdots \cdots \cdots \\ 9 \ 6 \ 10 \ 1 \\ 5 \cdot 6 + 1 \cdot 9 = 39 \end{array}$$

3) Step 3:

$$\begin{array}{r} 10 \ 3 \ 1 \ 5 \\ \cdots \cdots \cdots \cdots \\ 9 \ 6 \ 10 \ 1 \\ 3 \cdot 9 + 1 \cdot 6 + 5 \cdot 10 = 27 + 6 + 50 = 83 \end{array}$$

4) Step 4:

$$\begin{array}{r} \cdots \cdots 10 \ 3 \ 1 \ 5 \\ \cdots \cdots 9 \ 6 \ 10 \ 1 \\ 10 \cdot 9 + 3 \cdot 6 + 1 \cdot 10 + 5 \cdot 1 = 90 + 18 + 10 + 5 = 123 \end{array}$$

5) Step 5:

$$\begin{array}{r} \cdots \cdots 10 \ 3 \ 1 \ 5 \\ \cdots \cdots 9 \ 6 \ 10 \ 1 \\ 10 \cdot 6 + 3 \cdot 10 + 1 \cdot 1 = 60 + 30 + 1 = 91 \end{array}$$

6) Step 6:

$$\begin{array}{r} \cdots \cdots 10 \ 3 \ 1 \ 5 \\ \cdots \cdots 9 \ 6 \ 10 \ 1 \\ 10 \cdot 10 + 3 \cdot 1 = 100 + 3 = 103 \end{array}$$

7) Step 7:

$$\begin{array}{r} \cdots \cdots 10 \ 3 \ 1 \ 5 \\ \cdots \cdots 9 \ 6 \ 10 \ 1 \\ 10 \cdot 1 = 10 \end{array}$$

Answer: $conv = [45, 39, 83, 123, 91, 103, 10]$

Task 5.a

The sinusoidal signal with the frequency $f = 6\text{kHz}$ is sampled with frequency $f_s = 10\text{kHz}$. Compute the apparent frequency after the signal reconstruction.

For simplicity assume that the magnitude of signal is 1, the phase is 0. Then
 $f(x) = \sin(2 \cdot 6 \cdot 10^3 \pi x)$

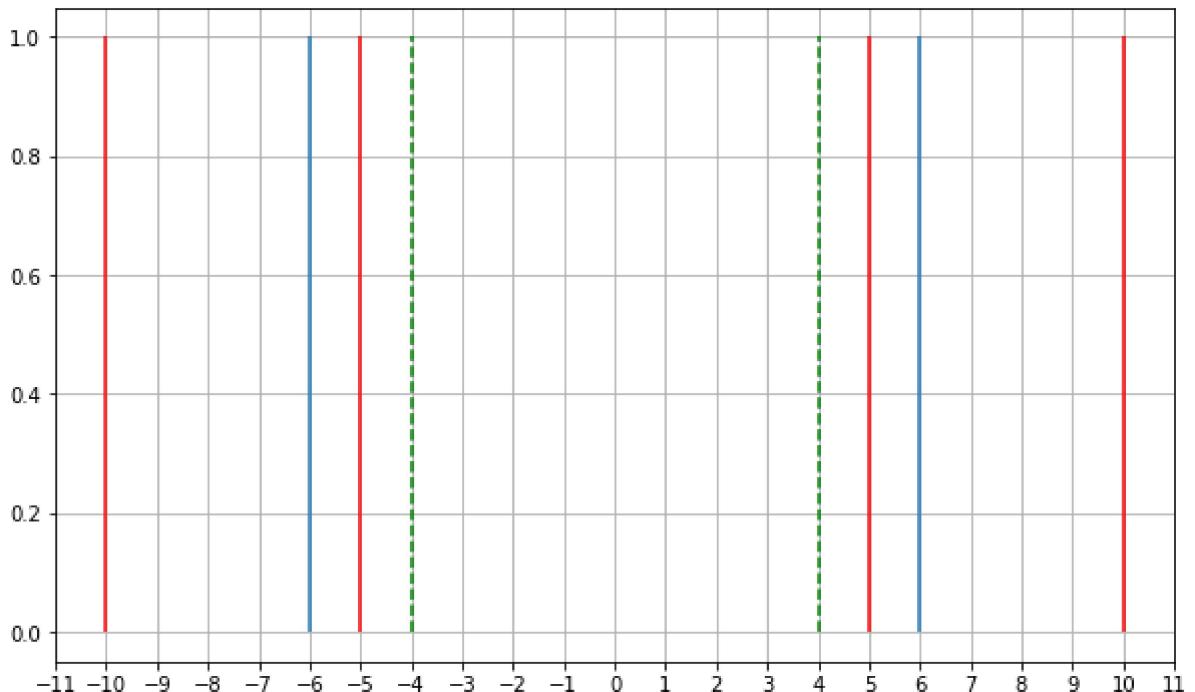
The FT of this function will give two delta functions in frequency domain located on $f = \pm 6\text{kHz}$. Let's plot it.

In [28]:

```
x = list(range(-12, 12 ,1))
y = []
for point in x:
    if point == -6:
        y.append(1)
    elif point == 6:
        y.append(1)
    else:
        y.append(0)
```

In [46]:

```
# red corresponds to the sampling frequency
# blue corresponds to the initial signal frequency
# green dash corresponds to the aliasing frequency
fig, ax = plt.subplots(figsize=(10,6))
major_ticks = list(range(-16, 16, 1))
ax.vlines(6, 0, 1)
ax.vlines(-6, 0, 1)
ax.grid()
ax.set_xticks(major_ticks);
# Then I plot our sampling frequency
ax.vlines(5, 0, 1, color = 'r')
ax.vlines(-5, 0, 1, color = 'r')
# It is obvious that frequency of the initial signal lies
# outside the fundamental frequency range ( $|F| \leq F_s/2$ ) so we will achieve
# aliasing here.
# Then I will plot the copies of initial signal near the multiples of  $F_s$ 
ax.vlines(10, 0, 1, color = 'r')
ax.vlines(-10, 0, 1, color = 'r')
ax.vlines(4, 0, 1, color = 'g', linestyle = '--')
ax.vlines(-4, 0, 1, color = 'g', linestyle = '--');
# As you can see I obtain frequency = +4 kHz inside the fundamental frequency range
# It is the aliasing frequency
```



Answer: after the signal reconstruction we will get the sinusoidal signal with the frequency $f = 4\text{kHz}$

Task 5.b

The sinusoidal signal with the frequency $f = 6\text{kHz}$ is represented by 4 signals periods sampled with the sample clock frequency $f_s = 15\text{kHz}$. Are these samples enough to reconstruct the initial signal correctly.

I think that we can not reconstruct signal correctly based on these samples. Because when we sample function at frequency f_s the sinusoidal functions yields the same set of samples $\{\sin(2\pi(f + Nf_s)x + \varphi, N = 0, \pm 1, \pm 2, \dots\}$ So without additional information we can't distinguish the frequency of the original function. I will try to give an example.

In [142...]

```
# define the frequency of the initial signal and sampling frequency
f = 6
f_s = 15
T = 1 / f
T_s = 1 / f_s
```

In [143...]

```
x = np.arange(0, 4 * T, T_s)
```

In [144...]

```
y = np.sin(2 * np.pi * f * x) # initial signal
```

In [145...]

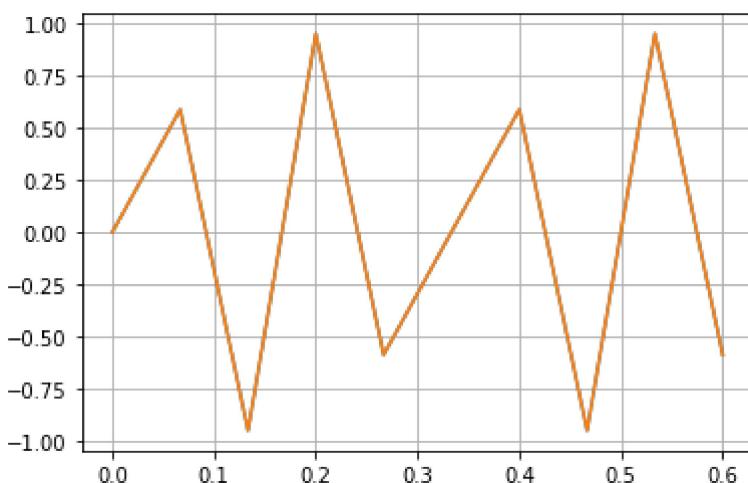
```
N = 1
```

In [147...]

```
y_2 = np.sin(2 * np.pi * (f + N * f_s) * x) # aliased sinusoidal signal
```

In [148...]

```
plt.plot(x, y)
plt.plot(x, y_2);
plt.grid()
```



According to the graph above, both sinusoidal functions with different carrier frequencies has the same set of samples, that's why their graphs coincides.

Answer: we can't reconstruct the initial signal correctly without additional information about this signal.

You can download this notebook and run locally:

https://github.com/dmasny99/DSP_skoltech/blob/main/HW1_Masnyi_Dmitrii.ipynb

