

### 1: The First Problem

$$\begin{aligned}
 x(t) &: \text{non-periodic.} \\
 CTFT &: x(jw) = \int_{-\infty}^{\infty} x(t)e^{-jwt} dt \\
 &: x(t) = \frac{1}{2\pi} \int_{-\infty}^{\infty} x(jw)e^{jwt} dw
 \end{aligned}$$

### 2: The Second Problem

### 3: The Third Problem

**Unit Impulse** is a sequence which has only one nonzero value, that occurs at  $n = 0$ .

$$\delta[n] = \begin{cases} 1, & \text{if } n = 0, \\ 0, & \text{if } n \neq 0. \end{cases}$$

**Unit Impulse Response** is the output sequence, when the input to the FIR filter is a unit impulse sequence.

$$h[n] = \sum_{k=0}^M b_k \delta[n - k] = \begin{cases} 1, & \text{if } n = 0, \\ 0, & \text{if } n \neq 0. \end{cases}$$

### 4: The Fourth Problem

There are two general LTI system properties: **time invariant** and **linearity**.

**Time Invariant:**

$$\begin{aligned}
 x[n] &\mapsto y[n] \\
 x[n - n_0] &\mapsto y[n - n_0]
 \end{aligned}$$

**Linearity:**

$$\begin{aligned}
 x_1[n] &\mapsto y_1[n] \\
 x_2[n] &\mapsto y_2[n] \\
 x[n] = \alpha x_1[n] + \beta x_2[n] &\mapsto y[n] = \alpha y_1[n] + \beta y_2[n]
 \end{aligned}$$

**Derivation of Convolutional Sum:**

We can represent any signal as sum of impulses:

$$\begin{aligned} x[n] &\mapsto \boxed{S} \mapsto y[n] \\ x[n] &= \sum_{k=-\infty}^{\infty} x[n] \delta[n-k] \end{aligned}$$

The output will be:

$$\begin{aligned} y[n] &= S[x[n]] \\ y[n] &= S\left[\sum_{k=-\infty}^{\infty} x[n] \delta[n-k]\right] \\ y[n] &= \sum_{k=-\infty}^{\infty} x[n] S[\delta[n-k]] \quad (\text{linearity}) \end{aligned}$$

Now, we need to know  $S[\delta[n-k]]$ , we can assume that this equation can be written safely:

$$\begin{aligned} h[n] &= S[\delta[n]] \\ h[n-k] &= S[\delta[n-k]] \quad (\text{time invariant}) \end{aligned}$$

All these equations above give us the convolutional sum formula:

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

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**5: The Fifth Problem**


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- (a)
  - Not casual:** The function is dependent on future
  - Stable:** x and y can be bounded.
- (b)
  - Casual:** The function is independent on future
  - Not stable:** x and y cannot be bounded.
- (c)
  - Casual:** The function is independent on future
  - Stable:** The function cannot exceed 1.
- (d)
  - Casual:** The function is independent on future
  - Stable:** The function cannot exceed 1.

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## 6: The Sixth Problem

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I designed a function named **my\_conv** which takes two parameters, signal array ( $x[n]$ ) and impulse response ( $h[n]$ ), and returns the convolution.

```
[Emre:Homework2 KE0$ cat my_conv.py

def my_conv(x, h):
    m = [[0] * (len(x) + len(h) - 1) for _ in range(len(h))]
    y = [0] * (len(x) + len(h) - 1)

    for i in range(len(h)):
        for j in range(len(x)):
            m[i][i+j] = h[i] * x[j]

    for i in range(len(x) + len(h) - 1):
        for j in range(len(h)):
            y[i] += m[j][i]

    return y

if __name__ == '__main__':
    X = [2, 4, 6, 4, 2]
    H = [3, -1, 2, 1]

    print(my_conv(X, H))

[Emre:Homework2 KE0$ python my_conv.py
[6, 10, 18, 16, 18, 12, 8, 2]
```

Figure 1: The output of my\_conv function for given parameters in question

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## 7: The Seventh Problem

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$$\begin{aligned}
 y[n] &= x[n] * h[n] \\
 y[n] &= [2 \ 4 \ 6 \ 4 \ 2] * \begin{bmatrix} 3 & -1 & 2 & 1 & 0 & 0 & 0 & 0 \\ 0 & 3 & -1 & 2 & 1 & 0 & 0 & 0 \\ 0 & 0 & 3 & -1 & 2 & 1 & 0 & 0 \\ 0 & 0 & 0 & 3 & -1 & 2 & 1 & 0 \\ 0 & 0 & 0 & 0 & 3 & -1 & 2 & 1 \end{bmatrix} \\
 y[n] &= [6 \ 10 \ 18 \ 16 \ 18 \ 12 \ 8 \ 2]
 \end{aligned}$$

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## 8: The Eighth problem

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The `convolve2d` method in the `scipy` library should take at least two parameter, array likes inputs and convolve this two 2-dimensional arrays.

When the program executed, the resulting image, called "result.png" will be created in the very same folder.

```
[Emre:Homework2 KE0$ cat convolution.py
```

```
import numpy
import scipy.signal
import matplotlib.image as mpimg
from skimage.color import rgb2gray
```

```
K = numpy.array([[1 / 9.] * 3 for i in range(3)])
img = mpimg.imread('noisyCameraman.png')
```

```
convolved = scipy.signal.convolve2d(rgb2gray(img), K, mode='same', boundary='fill', fillvalue=0)
scipy.misc.imsave('result.png', convolved)
```

```
[Emre:Homework2 KE0$ python3 convolution.py
```

Figure 2: The usage of written python program



Figure 3: The resulting image

As we can see the resulting image in Figure 3, convolved image is smoother.