```
In [1]:
                                                                                           H
# Implementation of sigmoid function
import numpy as np
def sigmoid(z):
 return 1 / (1 + np.exp(-z))
print("Sigmoid of 4 is:", sigmoid (4))
Sigmoid of 4 is: 0.9820137900379085
In [2]:
                                                                                           H
nrint("Sigmoid of nositive number(5) is." sigmoid(5))
Sigmoid of positive number(5) is: 0.9933071490757153
In [3]:
                                                                                           H
print("Sigmoid of negative number(-5) is:".sigmoid(-5))
Sigmoid of negative number(-5) is: 0.0066928509242848554
In [4]:
print("Difference between Derivative of Sigmoid (5) and (-5) is:", sigmoid(5)*(1- sigmoid(5)
Difference between Derivative of Sigmoid (5) and (-5) is: -1.214306433183765
e-16
In [6]:
                                                                                           H
# vanishing gradient
Difference between sigmoid of 14 and 15: 5.256258007735326e-07
In [7]:
                                                                                           H
# Implementation of tanh function
def tanh(z):
 return np.tanh(z)
tanh of 4 is: 0.999329299739067
In [8]:
# zero centric
print("tanh of positive number(15) is:",tanh(15))
tanh of positive number(15) is: 0.999999999998128
```

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In [9]:
print("tanh of positive number(-15) is:",tanh(-15))
tanh of positive number(-15) is: -0.999999999998128
In [10]:
# vanishing gradient
Difference between tanh of 14 and 15: 1.1957101975212936e-12
In [12]:
                                                                                           H
# Implementation of ReLU(Rectified Linear Unit) function
def relu(z):
 return max(0, z)
ReLU of 10 is: 10
In [13]:
                                                                                           H
# dying neuron( all neurons from (0,-inf)=0, this causes problem in backpropagation)
print("ReLU of "+str(z)+" is :",relu(z))
z = -50
print("ReLU of "+str(z)+" is :",z * (z > 0))
ReLU of -0.4 is: 0
ReLU of -50 is : 0
In [14]:
                                                                                           M
# Implementation of leaky relu function
def leakyrelu(z):
 return np.maximum(0.01 * z, z)
print("ReLU of "+str(z)+" is :",leakyrelu(z))
ReLU of 10 is: 10.0
In [15]:
                                                                                           H
z = -1
print("ReLU of "+str(z)+" is :",leakyrelu(z))
ReLU of -1 is : -0.01
```

```
In [16]:
```

```
# Implementation of Exponential Relu function

def erelu(z,alpha):
    return z if z >= 0 else alpha*(np.exp(z) -1)
print("Exponential Relu for 10 is :",erelu(10,3))
print("Exponential Relu for -10 is :",erelu(-10,3))
```

```
Exponential ReLu for 10 is : 10 Exponential ReLu for -10 is : -2.9998638002107123
```

In [21]:

```
#Implementation of filter (of size 3*3 and 5*5) on an image
import cv2
import matplotlib.pyplot as plt
def main():
   path = "D:\\Youtube Code\\Python\\Python OpenCV3\\Python-OpenCV3\\Dataset\\"
    imgpath = path + "4.2.07.tiff"
   img = cv2.imread(imgpath, 1)
   img = cv2.cvtColor(img, cv2.COLOR_BGR2RGB)
   box = cv2.boxFilter(img, -1, (3, 3))
   blur = cv2.blur(img, (5, 5))
   gaussian = cv2.GaussianBlur(img, (3, 3), 0)
   titles = ['Original Image', 'Box Filter',
              'Blur', 'Gaussian Blur']
   outputs = [img, box, blur, gaussian]
   for i in range(4):
        plt.subplot(2, 2, i+1)
        plt.imshow(outputs[i])
        plt.title(titles[i])
        plt.xticks([])
        plt.yticks([])
   plt.show()
```

In [23]:

```
# Zero padding and stride = 1
def convolve2D(image, kernel, padding=0, strides=1):
   # Cross Correlation
   kernel = np.flipud(np.fliplr(kernel))
   # Gather Shapes of Kernel + Image + Padding
   xKernShape = kernel.shape[0]
   yKernShape = kernel.shape[1]
   xImgShape = image.shape[0]
   yImgShape = image.shape[1]
   # Shape of Output Convolution
   xOutput = int(((xImgShape - xKernShape + 2 * padding) / strides) + 1)
   yOutput = int(((yImgShape - yKernShape + 2 * padding) / strides) + 1)
   output = np.zeros((xOutput, yOutput))
   # Apply Equal Padding to All Sides
   if padding != 0:
        imagePadded = np.zeros((image.shape[0] + padding*2, image.shape[1] + padding*2))
        imagePadded[int(padding):int(-1 * padding), int(padding):int(-1 * padding)] = image
        print(imagePadded)
   else:
        imagePadded = image
   # Iterate through image
   for y in range(image.shape[1]):
        # Exit Convolution
        if y > image.shape[1] - yKernShape:
        # Only Convolve if y has gone down by the specified Strides
        if y % strides == 0:
            for x in range(image.shape[0]):
                # Go to next row once kernel is out of bounds
                if x > image.shape[0] - xKernShape:
                    break
                try:
                    \# Only Convolve if x has moved by the specified Strides
                        output[x, y] = (kernel * imagePadded[x: x + xKernShape, y: y + yKer
                except:
                    break
   return output
```

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In []:
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