```
Q
       {x}
                                  sum = 0
mid = int(mimg[i][j])
for p in range(-kernel_size, kernel_size+1):
    for q in range(-kernel_size, kernel_size+1):
        distance = int(ming[p+i][q+j])-mid
        range_filter[p-kernel_size][q-kernel_size]=np.exp(-distance**2/(sigma**2))/(sigma**2)
range_riter[p-kernel_size][q-kernel_size]=np.exp(-distance**2/(sigma**2))/(sigma**2)
filter = np.multiply(gf, range_filter)
norm = np.sum(filter)
for p in range(-kernel_size, kernel_size+1):
    for q in range(-kernel_size, kernel_size+1):
        sum+=mimg[i-(p+kernel_size)][j-(q+kernel_size)]*filter[p+kernel_size][q+kernel_size]
output[i-kernel_size][j-kernel_size] = sum/norm
                                                                                                                                                                                                                                                          ↑ ↓ © 目 ‡ 🖟 🗎 :
            bilFilter = cv2.bilateralFilter(img, kernel_size*2+1, sigma, sigma)
                    figure , (axis_1, axis_2,axis_3) = plt.subplots(1,3, figsize=(12,12))
axis_1.imshow(img, "gray")
axis_1.set_title("Input Image")
                    axis_2.imshow(output, "gray")
axis_2.set_title("Output Image")
axis_3.set_title("BilateralFilter")
                     axis_3.imshow(bilFilter, "gray")
                     <matplotlib.image.AxesImage at 0x7fbe9a1cd950>
                                                                                                                                                   BilateralFilter
                                          Input Image
                                                                                             Output Image
                                                                           100
                      100
                                                                                                                                100
                       200
                                                                            200
                                                                                                                                200
                       300
                       400
                                                                           400
                                                                                                                                400
<>
                                  100 200 300 400
                                                                     500
                                                                                        100 200 300 400
                                                                                                                         500
                                                                                                                                            100 200 300 400
                                                                                                                                                                              500
\equiv
```

```
Mean
Q
                                                                                                                                                                                                                                            ↑ ↓ © 目 ‡ 🖟 🖥 :
            kernel_size = int(input("Enter Kernel Size: "))
kernel = np.ones((kernel_size, kernel_size), np.float32)
{x}
                    kernel_size=kernel_size//2
output = np.zeros((img.shape[0], img.shape[1]), np.float32)
norm = kernel.sum()
                    norm = kernel.sum()
mimg = cv2.copyMakeBorder(img,kernel_size, kernel_size, kernel_size, kernel_size, cv2.BORDER_REPLICATE)
for i in range(2*kernel_size, mimg.shape[0]):
    for j in range(2*kernel_size,mimg.shape[1]):
    sum = 0
    for p in range(-kernel_size, kernel_size+1):
                                 for q in range(-kernel_size, kernel_size+1):
    sum+=mimg[i-(p+kernel_size)][j-(q+kernel_size)]*kernel[p+kernel_size][q+kernel_size]
    output[i-(2*kernel_size)][j-(2*kernel_size)] = sum
                    output/=norm
                    figure, (axis_1, axis_2) = plt.subplots(1,2, figsize=(12,12))
                    axis_1.imshow(img, "gray")
axis_1.set_title("Input Image")
axis_2.imshow(output, "gray")
                    axis_2.set_title("Output Image")
             Enter Kernel Size: 5
                    Text(0.5, 1.0, 'Output Image')
                                                    Input Image
                     100
                                                                                                   100
                     150
                                                                                                   150
                      250
                                                                                                   250 -
```

```
Q
            Gaushian
\{x\}
       v [108] def conv(kernel_size,filter):
    output = np.zeros((img.shape[0], img.shape[1]), np.float32)
    mimg = cv2.copyMakeBorder(img,kernel_size, kernel_size, kernel_size, kernel_size, cv2.BORDER_REPLICATE)
    for i in range(2*kernel_size, mimg.shape[0]):
        for j in range(2*kernel_size,mimg.shape[1]):
        sum = 0
sum = 0
for p in range(-kernel_size, kernel_size+1):
    for q in range(-kernel_size, kernel_size+1):
        sum+=mimg[i-(p+kernel_size)][j-(q+kernel_size)]*filter[p+kernel_size][q+kernel_size]
    output[i-(2*kernel_size)][j-(2*kernel_size)] = sum
                        return output
       gf = np.zeros((2*kernel_size+1, 2*kernel_size+1), np.float32)
                           for i in range(-kernel_size, kernel_size+1):
    for j in range (-kernel_size, kernel_size+1):
        gf[i+kernel_size][j+kernel_size]=np.exp(-(pow(i,2)+pow(j,2))/(2*pow(sigma,2)))
return gf, kernel_size
<>
                                                                                                                                                                                                                                                      ↑ ↓ ⊕ 目 ‡ 🖟 🖥 :
            kernel, kernel_size = gaussian_kernel()
output = conv(kernel_size, kernel)
figure, (axis_1, axis_2) = plt.subplots(1,2, figsize=(12,12))
axis_1.imshow(dimg, "gray")
axis_1.set_title("Input Image")
\{x\}
axis_2.imshow(output, "gray")
axis_2.set_title("Output Image")
                    Enter kernel Size: 5
Text(0.5, 1.0, 'Output Image')
                                                      Input Image
                                                                                                                                     Output Image
                      100
                      150
                       200
<>
>_
Q
            Laplacian
{x}
                                                                                                                                                                                                                                                       ↑ ↓ © 目 $ 🖟 🗎 📋 :
                   def laplacian_kernelp():
    lf = [[0,-1,0],[-1,4,-1],[0,-1,0]]
    return lf
def laplacian_kerneln():
                           lf = [[0,1,0],[1,-4,1],[0,1,0]]
return lf
        kernel = laplacian_kernelp()
kernel = np.array(kernel)
                     outputp = conv(kernel.shape[0]//2, kernel)
                    kernel = laplacian_kerneln()
kernel = np.array(kernel)
outputn=conv(kernel.shape[0]//2, kernel)
                     figure, (axis_1, axis_2, axis_3) = plt.subplots(1,3, figsize=(12,12))
                     axis_1.imshow(img, "gray")
axis_1.set_title("Input Image")
                    axis_2.inshow(outputp, "gray")
axis_2.set_title("Output with positive center")
axis_3.inshow(outputn, "gray")
axis_3.set_title("Output with negative center")
<>
Text(0.5, 1.0, 'Output with negative center')
                                         Input Image
                                                                                   Output with positive center
                                                                                                                                       Output with negative center
                                                                                                                                50
                      100
                                                                           100
                                                                                                                               100
                      150
                                                                           150
                                                                                                                               150
                                                                                                                               200
\equiv
>_
                                                                                                       150
```

```
↑ ♥ ♥ ♥ ₽ ₽ :
               Sobel
\{x\} \bigvee_{0s} [89] def horizontal_sobel_kernel(kernel_size: int):
                                  row_value = [[0 for i in range(kernel_size)] for j in range(1)]
col_value = [[1 for i in range(1)] for j in range(kernel_size)]
center = kernel_size//2
col_value[center][0] = 2
for i in range(1):
    for j in range(kernel_size):
        row_value[iff]=center
row_value[i][j]=center

center-=1

center-=1

horizontal_sobel_kernel = [[0 for i in range(kernel_size)] for j in range(kernel_size)]

horizontal_sobel_kernel = np.matmul(col_value, row_value)

return horizontal_sobel_kernel
                          def vertical_sobel_kernel(kernel_size: int):
                                 col_value = [[0 for i in range(1)] for j in range(kernel_size)]
row_value = [[1 for i in range(kernel_size)] for j in range(1)]
center = kernel size//2
row_value[0][center]=2
for i in range(kernel_size):
    for j in range(1):
                                 for j in range(1):
    col_value[i][j]=center
    center-=1
vertical_sobel_kernel = [[0 for i in range(kernel_size)] for j in range(kernel_size)]
vertical_sobel_kernel = np.matmul(col_value, row_value)
return vertical_sobel_kernel
\equiv
           kernel_size = int(input("Enter Kernel Size"))
kernel = horizontal_sobel_kernel(kernel_size)
kernel = np.array(kernel)
 Q
{x}
                          h = conv(kernel_size//2, kernel)
kernel = vertical_sobel_kernel(kernel_size)
kernel = np.array(kernel)
                          v = conv(kernel_size//2, kernel)
                          \texttt{output = np.zeros((v.shape[0], v.shape[1]), np.float32)}
                          #gradient Magnitude
for i in range(v.shape[0]):
                                  for j in range(v.shape[1]):
   output[i][j] = np.sqrt(pow(h[i][j],2)+pow(v[i][j],2))
                          figure, (axis_1, axis_2, axis_3, axis_4) = plt.subplots(1,4, figsize=(12,12))
                        figure, (axis_1, axis_2, axis_3, axis_4) =
axis_1.inshow(img, "gray")
axis_1.set_title("Input Image")
axis_2.set_title("Whorizontally Filtered")
axis_2.set_title("Whorizontally Filtered")
axis_3.imshow(v, "gray")
axis_3.set_title("Vertically Filtered")
axis_4.imshow(output, "gray")
axis_4.set_title("Output Image(Merged)")
                          Enter Kernel Size5
Text(0.5, 1.0, 'Output Image(Merged)')
                                                                                      Horizontally Filtered
                                                                                                                              50
                                                                                                                              100
                            100
 <>
\equiv
```

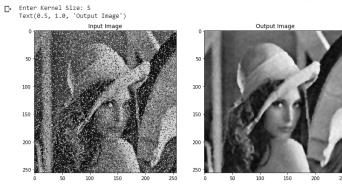
```
Q
             Schar
def vertical_scharr_kernel():
    vertical_schar_kernel = [[3,10,3],[0,0,0],[-3,-10,-3]]
    return vertical_schar_kernel
         7s [92] kernel = vertical_scharr_kernel()
 Q
                      kernel = np.array(kernel)
v = conv(kernel.shape[0]//2, kernel)
{x}
                      output = np.zeros((v.shape[0], v.shape[1]), np.float32)
#gradient Magnitude
for i in range(v.shape[0]):
    for j in range(v.shape[1]):
        output[i][j] = np.sqrt(pow(h[i][j],2)+pow(v[i][j],2))
figure, (axis_1, axis_2, axis_3, axis_4) = plt.subplots(1,4, figsize=(12,12))
axis_1.imshow(img, "gray")
axis_1.set_title("Input Image")
axis_2.imshow(h, "gray")
axis_2.imshow(h, "gray")
axis_2.set_title("Horizontally Filtered")
                      axis_3.imshow(v, "gray")
axis_3.set_title("Vertically Filtered")
axis_4.imshow(output, "gray")
axis_4.set_title("Output Image(Merged)")
                      Text(0.5, 1.0, 'Output Image(Merged)')
                                                                          Horizontally Filtered
                                                                                                                     Vertically Filtered
                                                                                                           100
                        100
                                                                                                           150
 <>
\equiv
```

```
Prewitt
\{x\}
         \begin{tabular}{ll} \checkmark & [93] & def horizontal\_prewitt\_kernel(): \\ & horizontal\_prewitt\_kernel = [[1,0,-1],[1,0,-1],[1,0,-1]] \\ & return horizontal\_prewitt\_kernel \end{tabular} 
def vertical_prewitt_kernel():
    vertical_prewitt_kernel = [[1,1,1],[0,0,0],[-1,-1,-1]]
    return vertical_prewitt_kernel
         kernel = horizontal_prewitt_kernel()
kernel = np.array(kernel)
h = conv(kernel.shape[0]//2, kernel)
 Q
                       kernel = vertical_prewitt_kernel()
                      kernel = np.array(kernel)
v = conv(kernel.shape[0]//2, kernel)
{x}
                       output = np.zeros((v.shape[0], v.shape[1]), np.float32)
#gradient Magnitude
                      for i in range(v.shape[0]):
    for j in range(v.shape[1]):
        output[i][j] = np.sqrt(pow(h[i][j],2)+pow(v[i][j],2))
                       figure, (axis_1, axis_2, axis_3, axis_4) = plt.subplots(1,4, figsize=(12,12))
                      rigure, (axis_i, axis_, axis_i, axis_i) =
axis_1.inshow(img, "gray")
axis_1.set_title("Input Image")
axis_2.imshow(h, "gray")
axis_2.set_title("Horizontally Filtered")
axis_3.imshow(v, "gray")
axis_3.set_title("Vertically Filtered")
axis_4.imshow(t, "gray")
                      axis_4.imshow(output, "gray")
axis_4.set_title("Output Image(Merged)")
                       Text(0.5, 1.0, 'Output Image(Merged)')
                                                                            Horizontally Filtered
                                        Input Image
                                                                                                                         Vertically Filtered
                                                                                                                                                                Output Image(Me
                                                                                                               100
                                                                                                                                                         100
                         100
                                                                                                               150
                         150
\equiv
>_
```

```
Robert
\{x\}
                    \begin{tabular}{ll} $\swarrow$ [95] $ $def $ horizontal\_roberts\_kernel(): \\ $horizontal\_roberts\_kernel = [[\theta,\theta,\theta],[\theta,1,\theta],[\theta,\theta,-1]] $ \end{tabular} 
return horizontal_roberts_kernel
                                                       def vertical_roberts_kernel():
    vertical_roberts_kernel = [[0,0,0],[0,0,1],[0,-1,0]]
    return vertical_roberts_kernel
                 √
7s [96]
 Q
                                                       kernel = vertical_roberts_kernel()
kernel = np.array(kernel)
v = conv(kernel.shape[0]//2, kernel)
{x}
                                                        \texttt{output = np.zeros((v.shape[0], v.shape[1]), np.float32)}
#gradient Magnitude
for i in range(v.shape[0]):
    for j in range(v.shape[1]):
        output[i][j] = np.sqrt(pow(h[i][j],2)+pow(v[i][j],2))
                                                         figure, (axis_1, axis_2, axis_3, axis_4) = plt.subplots(1,4, figsize=(12,12))
                                                       ragure, (axis_i, axis_c, axis_s, axis_i) = axis_1.imshow(img, "gray") axis_1.set_title("Input Image") axis_2.simshow(h, "gray") axis_2.set_title("Horizontally Filtered") axis_3.imshow(v, "gray") axis_3.set_title("Vertically Filtered") axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_1.axis_
                                                        axis_4.imshow(output, "gray")
axis_4.set_title("Output Image(Merged)")
                                                        Text(0.5, 1.0, 'Output Image(Merged)')
                                                                                                 Input Image
                                                                                                                                                                                           Horizontally Filtered
                                                                                                                                                                                                                                                                                 50
                                                                                                                                                                                                                                                                               100
                                                                                                                                                                                                                                                                                                                                                                                        100
                                                            100
                                                                                                                                                                                                                                                                                150
 <>
                                                              200
                                                                                                                                                                                                                                                                                 200
\equiv
>_
```

```
Midean

| Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Midean | Mi
```



<> =