1. To implement the imageGradientX() function I used numpy.ndenumerate() function to iterate over the image. This function returns an iterator with pairs of array indices and values. I used the indices to capture the value of the next pixel in the x direction, which I then used to calculate the change. I set the pixels along the edges to 0 indicating that there is no change because there are no more pixels to compare. After calculating the gradient I used numpy.absolute() function to convert that to the magnitude of the gradient.

```
image (numpy.ndarray): A grayscale image represented in a numpy array.
          output (numpy.ndarray): The image gradient in the X direction.
     # WRITE YOUR CODE HERE.
     \begin{array}{ll} \text{output} = \overline{\textbf{n}} \textbf{p.zeros} \text{(image.shape)} \\ \text{image} = \text{image.astype} \text{(np.float64)} \end{array}
     for index, pixel in np.ndenumerate(image):
          # Edge case
if index[x] == image.shape[x] - 1:
    output[index[y], index[x]] = 0
                 output[index[y],\ index[x]] = image[index[y],\ index[x] + 1] - pixel
     return np.obsolute output)
# END OF FUNCTION.
def imageGradientY(image):
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```

2. The imageGradientY() function was almost exactly the same except I used the indices to capture the value of the next pixel in the y direction, which I then used to calculate the change. I found I needed to convert the image array to float64 when my initial attempts resulted in noise. This was due to overflow errors as the input array was uint8.

```
Args:
    image (numpy.ndarray): A grayscale image represented in a numpy array.

Returns:
    output (numpy.ndarray): The image gradient in the Y direction.

"""

# WRITE YOUR CODE HERE.
y = 0
x = 1

output = np.zeros image.shape)
image = image.astype np.float64)

for index, pixel in np.ndenumerate image):
    # Edge case
    if index[y] == image.shape[y] = 1:
        output[index[y], index[x]] = 0
    else:
        output[index[y], index[x]] = image[index[y] + 1, index[x]] - pixel

return np.absolute output)
# END OF FUNCTION.
```

3. To implement the computeGradient() function I again used the numpy.ndenumerate() function to iterate through the image as well as the kernel. I first checked the bounds of the image to ensure the cross-correlation would take place in the bounds of the image. I then looped through the kernel using the image and kernel indices to capture the proper pixels to perform the crosscorrelation on.

```
output (numpy.ndarray): The computed gradient for the input image.
     output = np.zeros (image.shape)
image = image.astype np.float64)
     for imageIndex, imagePixel in np.ndenumerate(image):
           if imageIndex[x] > 0 and imageIndex[y] > 0 and imageIndex[x] < image.shape[x] = 1 and imageIndex[y] < image.shape[y] = 1:
                 sum = 0
                 for kernelIndex, kernelValue in np.ndenumerate kernel):
                      u = kernelIndex[x]
v = kernelIndex[y]
                      sum += image[imageIndex[y] + v, imageIndex[x] + u] * kernelValue
                 output[imageIndex[y], imageIndex[x]] = sum
     return output
     # END OF FUNCTION.
                          ^O WriteOut
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```

4. I started with the following image while experimenting with edge detection:



After experimenting with several kernels, left/right average derivative, Prewitt, Sobel, Gaussian derivative, etc. I found the Sobel kernel provided the most accurate depiction of the edges. After taking the X and Y gradients I converted them to black and white. I used a threshold of 96 when converting the gradient in the X direction and used a threshold of 128 when converting the gradient in the Y direction. To create the final image shown below I averaged the X and Y gradients after applying the threshold.



For reference I have included the cv2.Canny() result of my original image:

