

lab3-21035

November 19, 2023

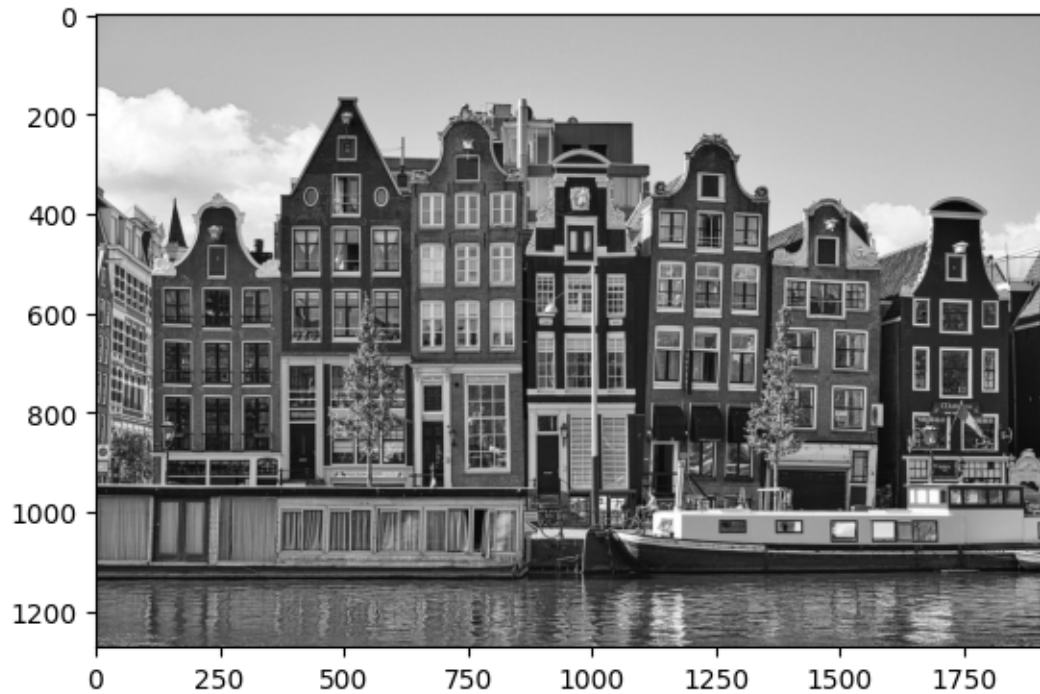
```
[34]: import cv2
import numpy as np
import matplotlib.pyplot as plt
import math
```

```
[35]: import cv2
import numpy as np
import matplotlib.pyplot as plt
import matplotlib.image as mpimg

# Read the image
img = cv2.imread('buildings.jpg')

# Convert to grayscale
gray = cv2.cvtColor(img, cv2.COLOR_BGR2GRAY)

# Show the image with matplotlib
plt.imshow(gray, cmap='gray')
plt.show()
```

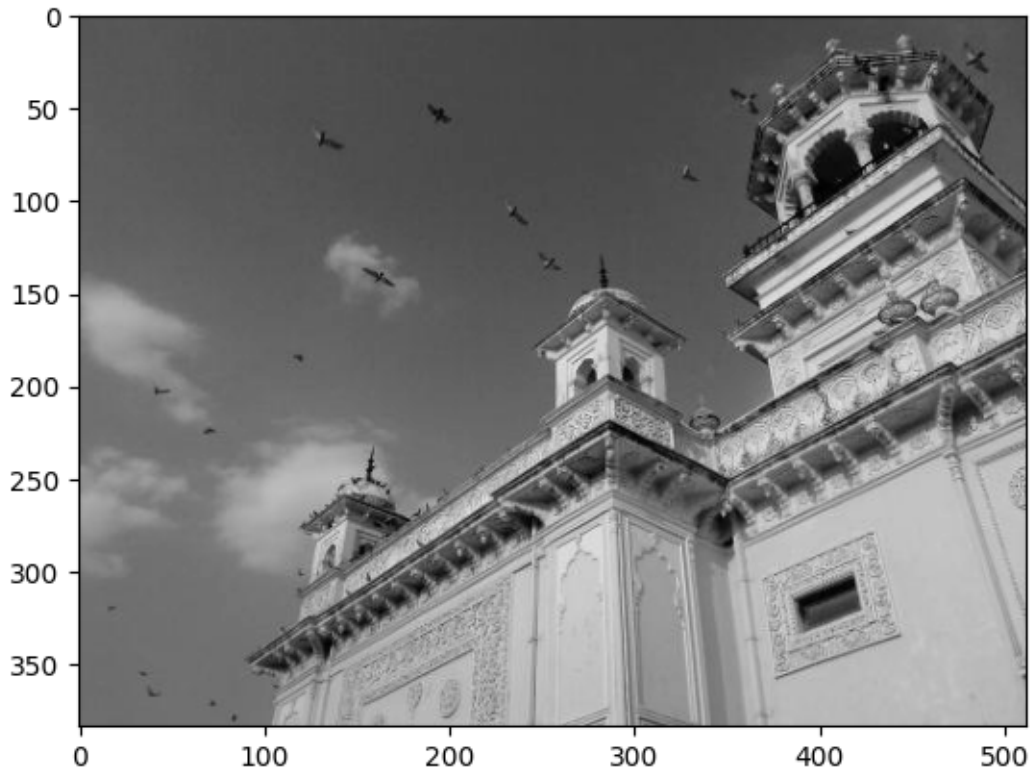


```
[36]: import cv2
import numpy as np
import matplotlib.pyplot as plt
import matplotlib.image as mpimg

# Read the image
img = cv2.imread('home.jpg')

# Convert to grayscale
gray = cv2.cvtColor(img, cv2.COLOR_BGR2GRAY)

# Show the image with matplotlib
plt.imshow(gray, cmap='gray')
plt.show()
```



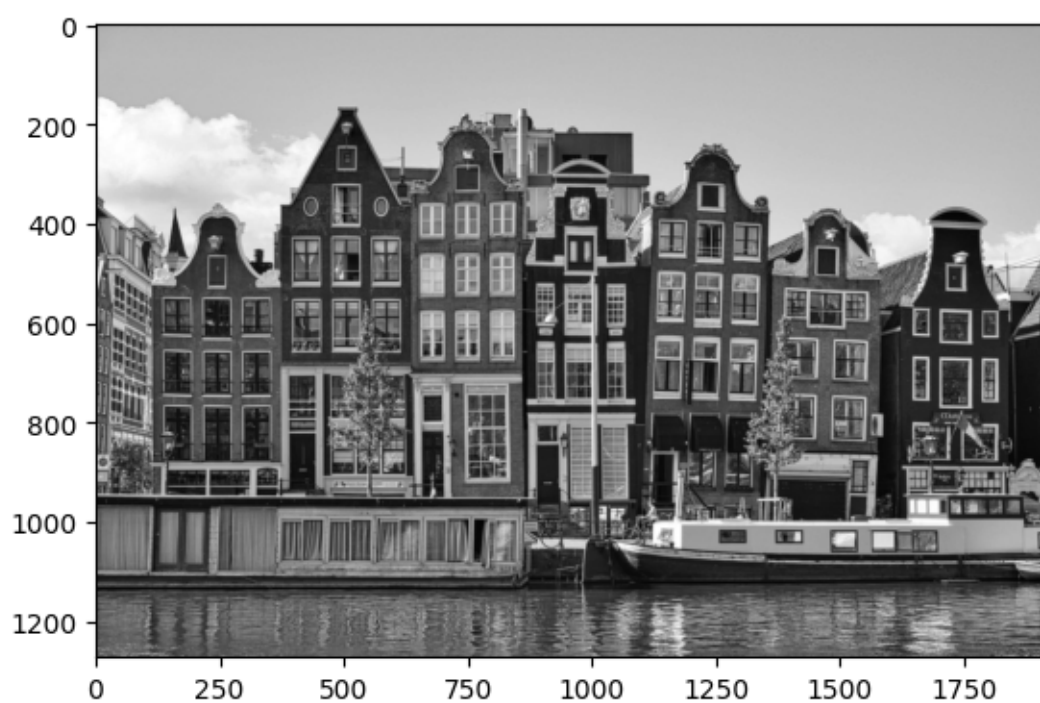
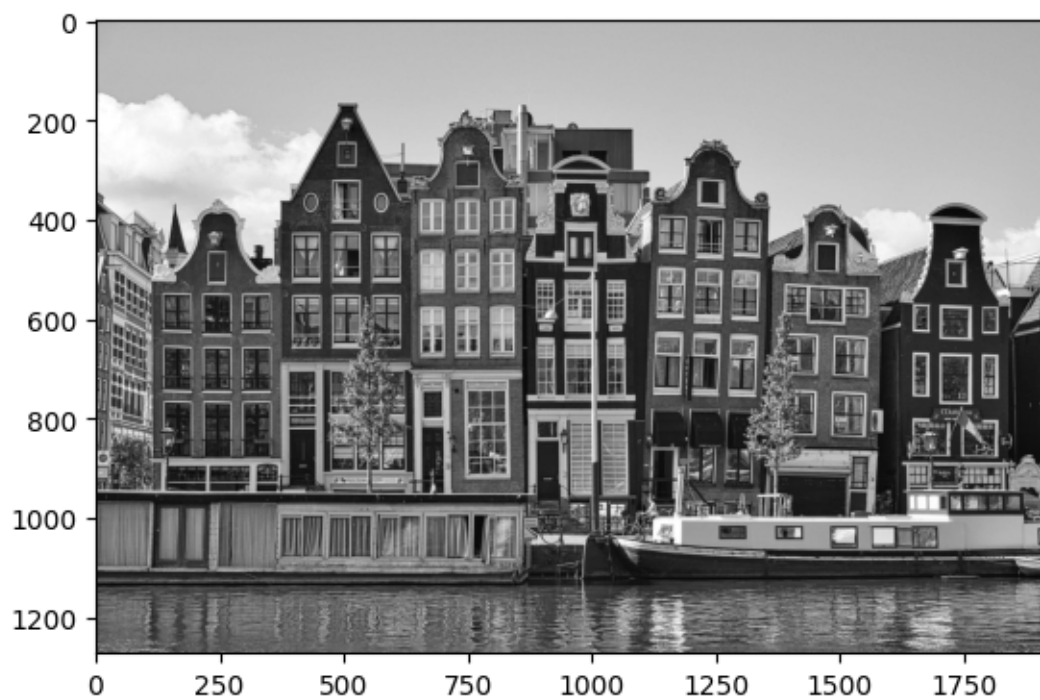
```
[37]: import cv2
import numpy as np
import matplotlib.pyplot as plt
import matplotlib.image as mpimg

# Read the image
img = cv2.imread('buildings.jpg')

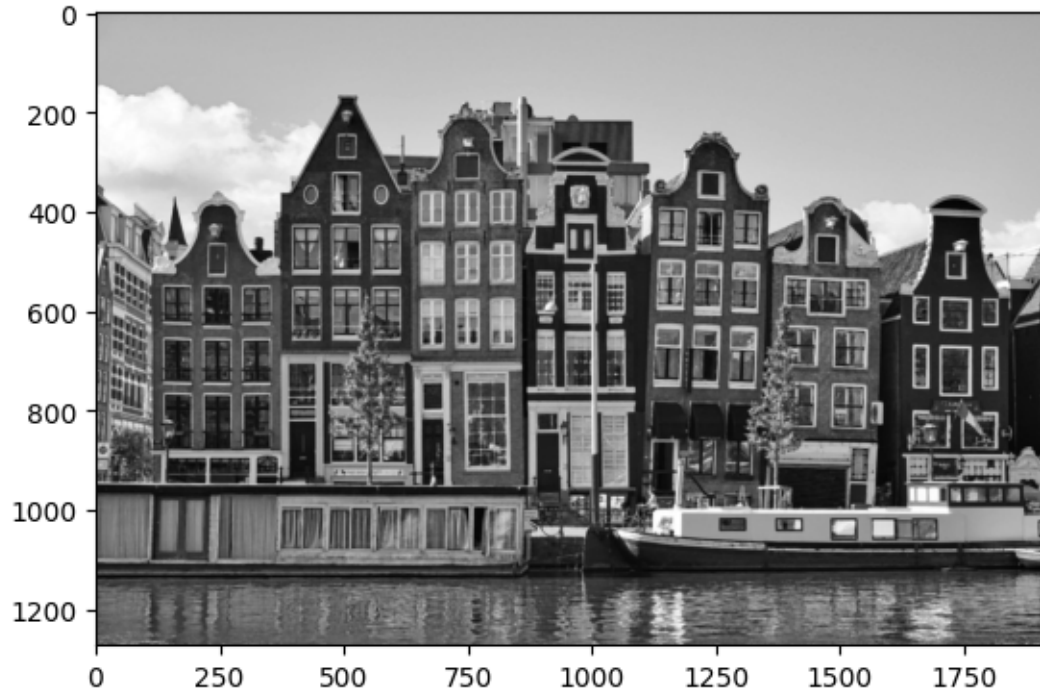
# Convert to grayscale
gray = cv2.cvtColor(img, cv2.COLOR_BGR2GRAY)

# Show the image with matplotlib
plt.imshow(gray, cmap='gray')
plt.show()

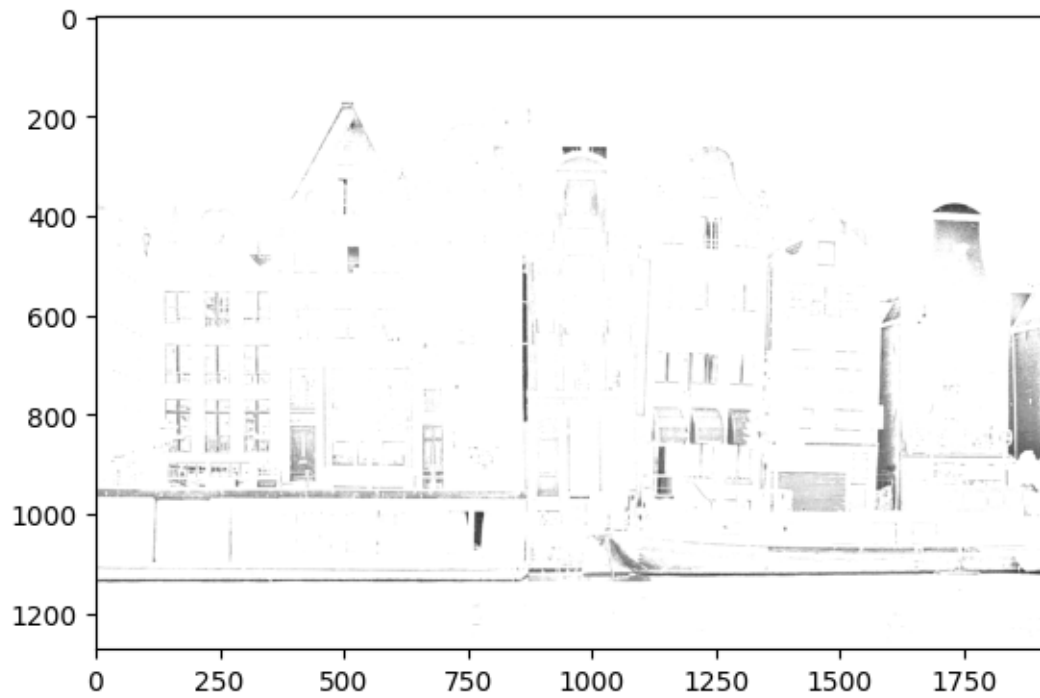
# Mean filter
kernel = np.ones((3,3),np.float32)/9
dst = cv2.filter2D(gray,-1,kernel)
plt.imshow(dst, cmap='gray')
plt.show()
```



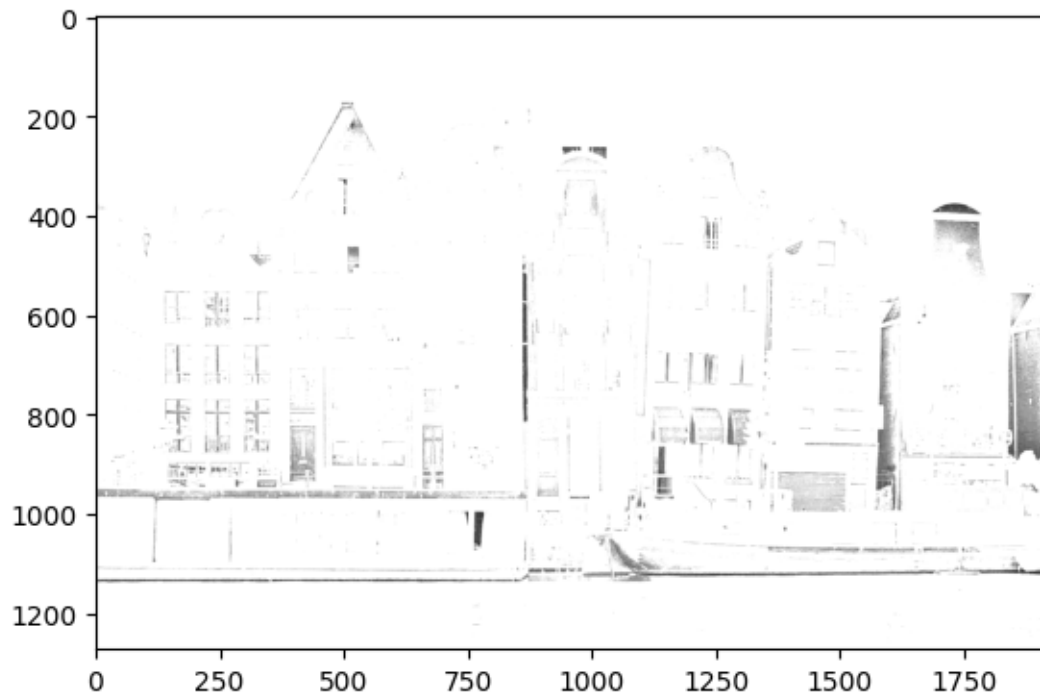
```
[38]: #Median filter
median = cv2.medianBlur(gray,5)
plt.imshow(median, cmap='gray')
plt.show()
```



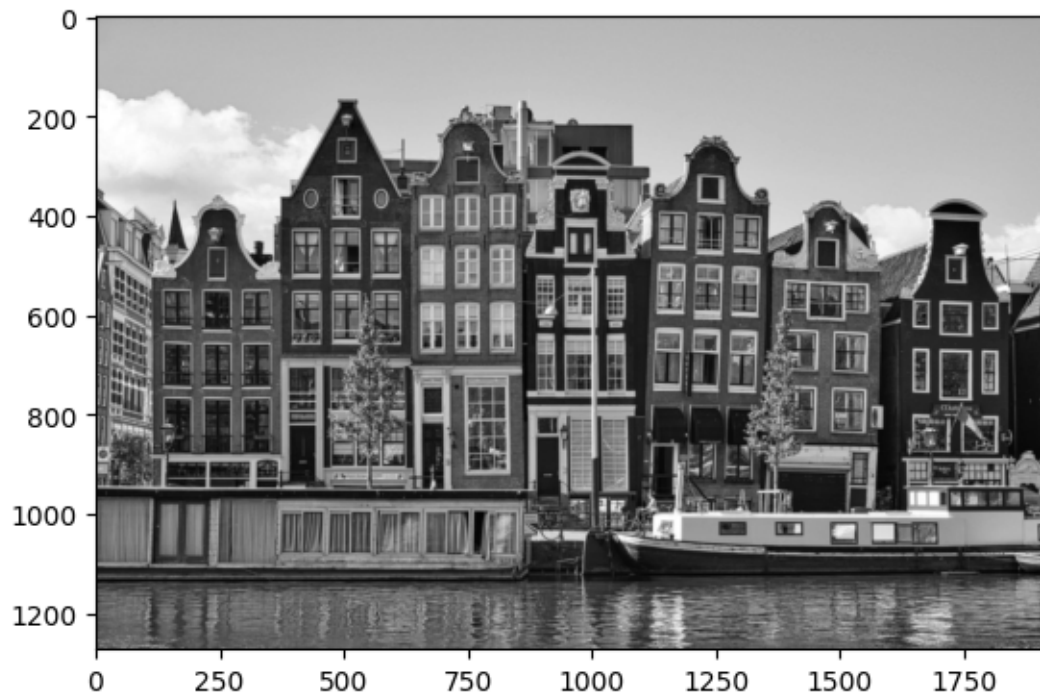
```
[39]: #max filter
kernel = np.ones((3,3),np.float32)
dst = cv2.filter2D(gray,-1,kernel)
plt.imshow(dst, cmap='gray')
plt.show()
```



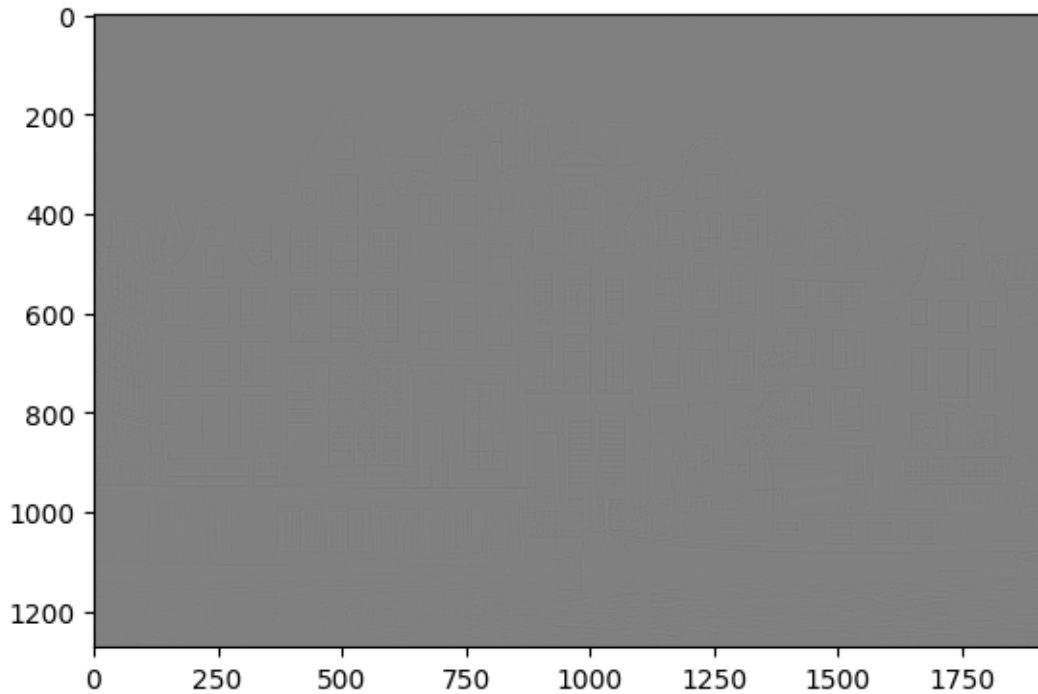
```
[40]: #min filter  
kernel = np.ones((3,3),np.float32)  
dst = cv2.filter2D(gray,-1,kernel)  
plt.imshow(dst, cmap='gray')  
plt.show()
```



```
[41]: #Guassian filter  
blur = cv2.GaussianBlur(gray,(5,5),0)  
plt.imshow(blur, cmap='gray')  
plt.show()
```



```
[42]: #laplacian filter  
laplacian = cv2.Laplacian(gray,cv2.CV_64F)  
plt.imshow(laplacian, cmap='gray')  
plt.show()
```

3.a. Apply an edge detector to the image and observe the intensity of the resulting output. Subsequently, employ a 3×3 mean filter on the initial image and reapply the edge detector. Comment on the difference. Investigate the effects of using a 5×5 or a 7×7 filter in this context

```
[43]: import cv2
import numpy as np
import matplotlib.pyplot as plt
import matplotlib.image as mpimg

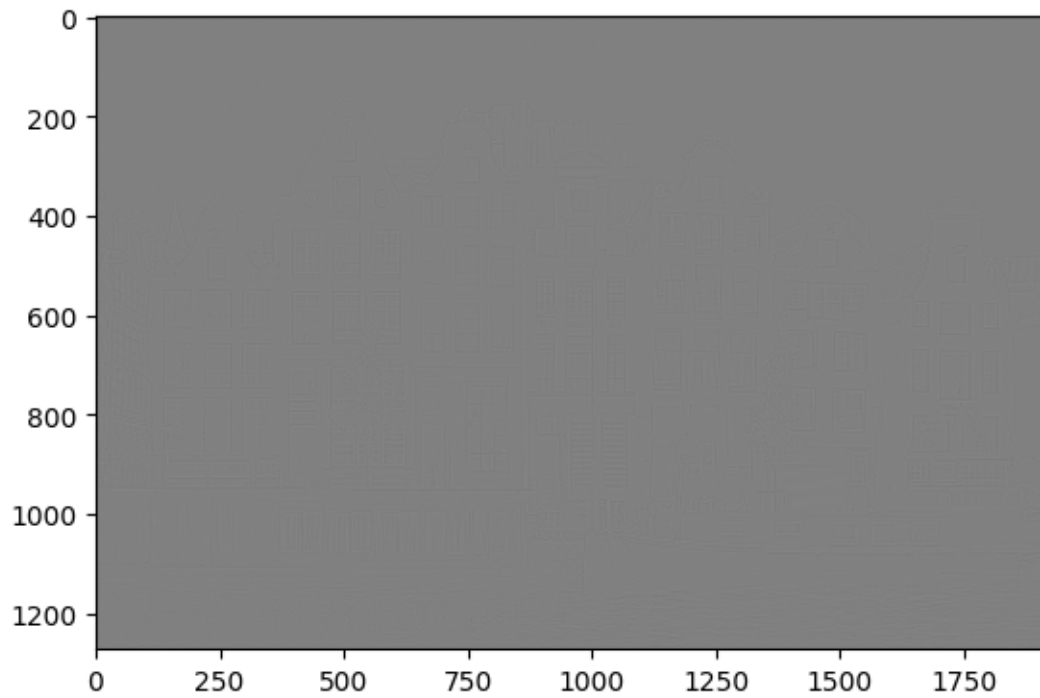
# Read the image
img = cv2.imread('buildings.jpg')

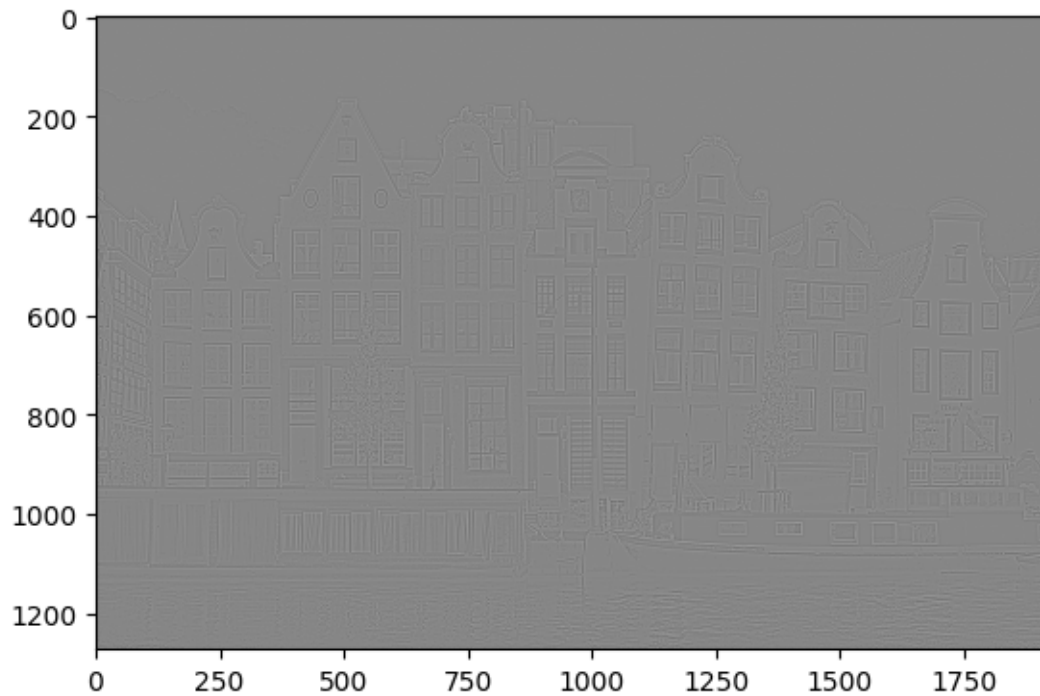
# Convert to grayscale
gray = cv2.cvtColor(img, cv2.COLOR_BGR2GRAY)

#laplacian filter
laplacian = cv2.Laplacian(gray, cv2.CV_64F)
plt.imshow(laplacian, cmap='gray')
plt.show()

#Mean filter
kernel = np.ones((3,3), np.float32)/9
dst = cv2.filter2D(gray, -1, kernel)
plt.imshow(dst, cmap='gray')
```

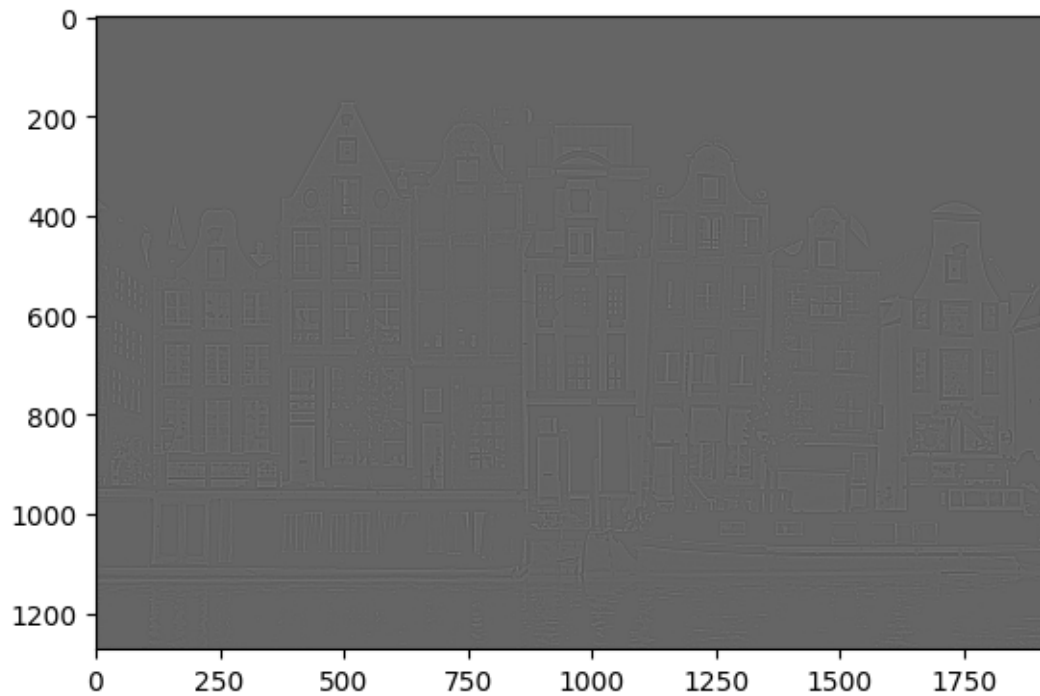
```
#laplacian filter  
laplacian = cv2.Laplacian(dst,cv2.CV_64F)  
plt.imshow(laplacian, cmap='gray')  
plt.show()
```





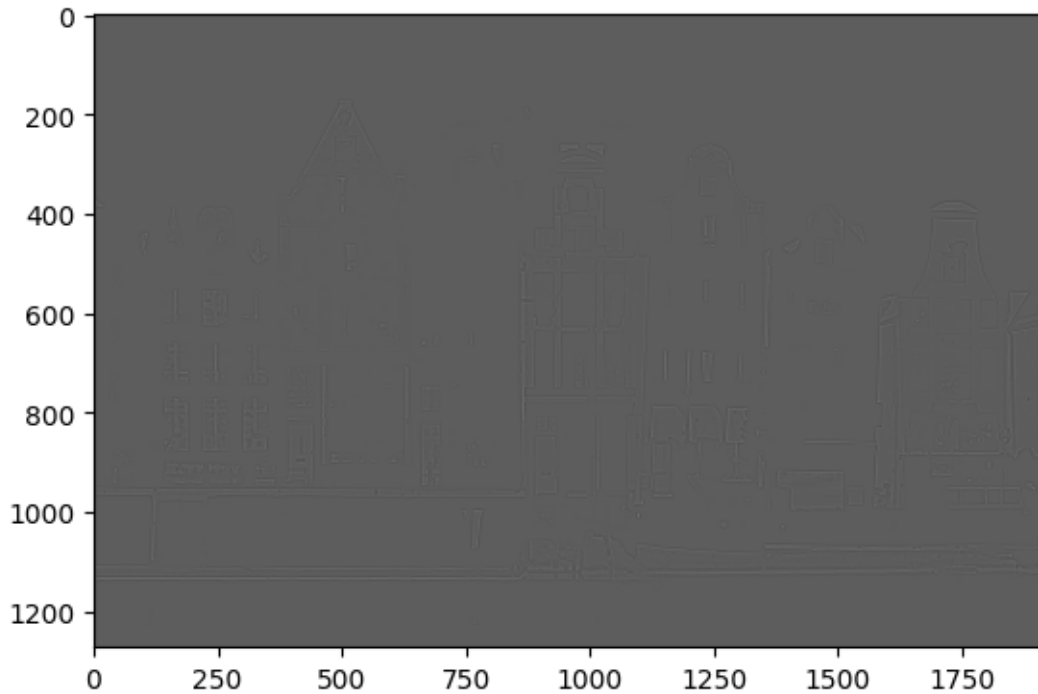
```
[44]: #Mean filter
kernel = np.ones((5,5),np.float32)/9
dst = cv2.filter2D(gray,-1,kernel)
plt.imshow(dst, cmap='gray')

#laplacian filter
laplacian = cv2.Laplacian(dst,cv2.CV_64F)
plt.imshow(laplacian, cmap='gray')
plt.show()
```



```
[45]: #Mean filter
kernel = np.ones((7,7),np.float32)/9
dst = cv2.filter2D(gray,-1,kernel)
plt.imshow(dst, cmap='gray')

#laplacian filter
laplacian = cv2.Laplacian(dst,cv2.CV_64F)
plt.imshow(laplacian, cmap='gray')
plt.show()
```



#The difference between above three is :

#The edges are more visible in the first image than the second and third image. #The edges are more visible in the second image than the third image.

3.b.Assess the comparative speed of mean and median filters when employing identical-sized neighbourhoods and images. Analyse how the efficiency of each filter is influenced by variations in the image size and the size of the neighbourhood

```
[46]: import cv2
import numpy as np
import matplotlib.pyplot as plt
import matplotlib.image as mpimg
import time

# Read the image
img = cv2.imread('home.jpg')
# Convert to grayscale
gray = cv2.cvtColor(img, cv2.COLOR_BGR2GRAY)

#Mean filter
start = time.time()
kernel = np.ones((3,3),np.float32)/9
dst = cv2.filter2D(gray,-1,kernel)
end = time.time()
```

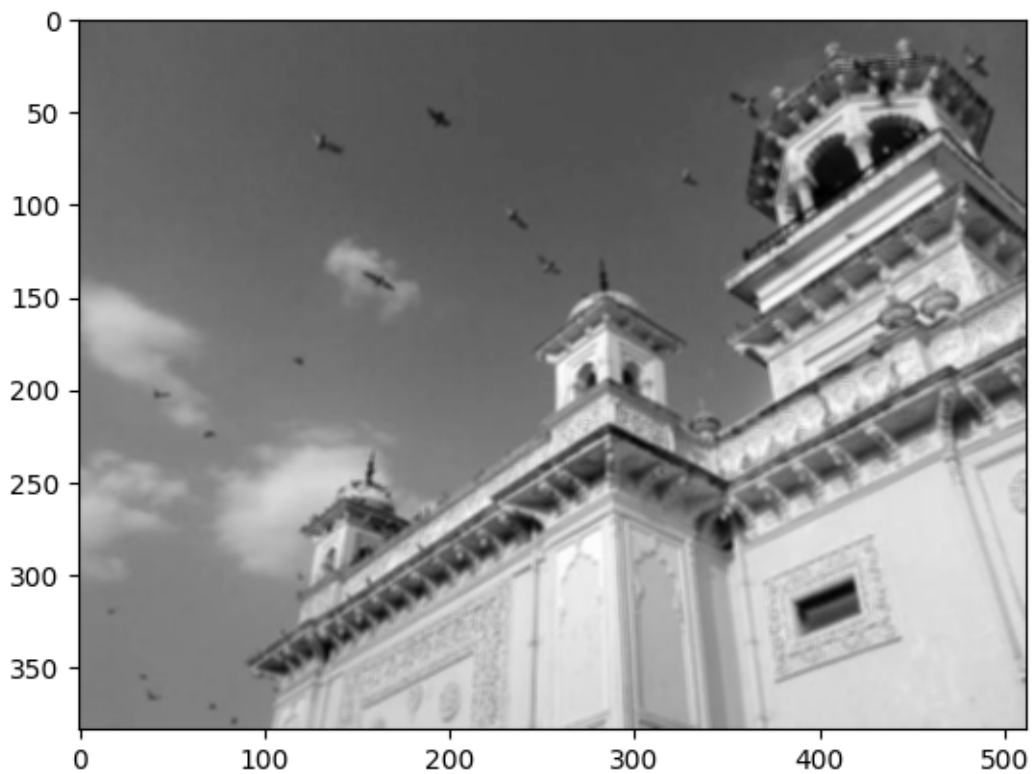
```

print(end - start)
plt.imshow(dst, cmap='gray')
plt.show()

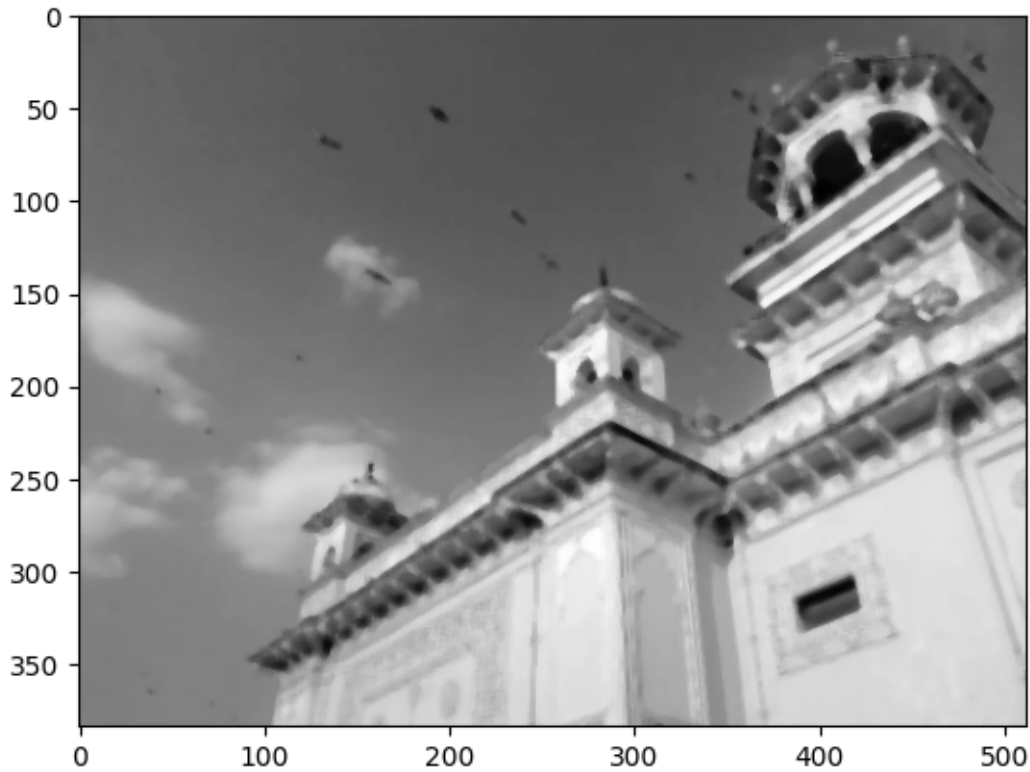
#Median filter
start = time.time()
median = cv2.medianBlur(gray,5)
end = time.time()
print(end - start)
plt.imshow(median, cmap='gray')
plt.show()

```

0.0003190040588378906



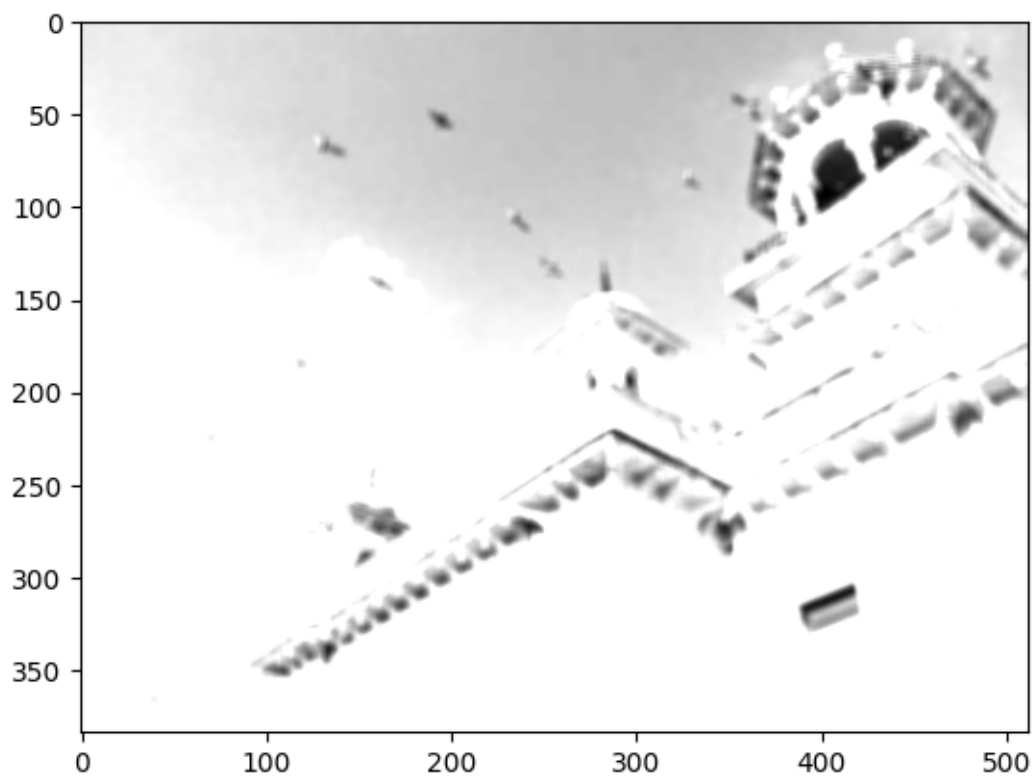
0.0006730556488037109



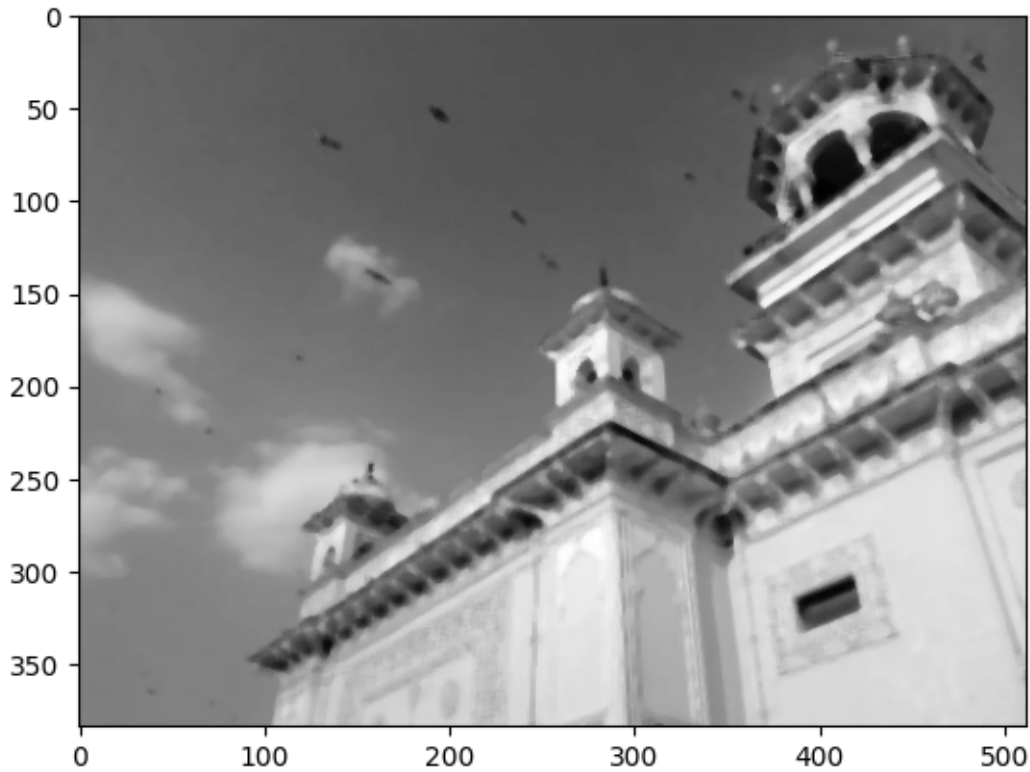
```
[47]: #Mean filter
start = time.time()
kernel = np.ones((5,5),np.float32)/9
dst = cv2.filter2D(gray,-1,kernel)
end = time.time()
print(end - start)
plt.imshow(dst, cmap='gray')
plt.show()

#Median filter
start = time.time()
median = cv2.medianBlur(gray,5)
end = time.time()
print(end - start)
plt.imshow(median, cmap='gray')
plt.show()
```

0.0005409717559814453



0.0008330345153808594



3.c.Experiment with the Gaussian filter using various sigma values and evaluate each one on the basis of noise removal and preservation of image details.

```
[48]: import cv2
import numpy as np
import matplotlib.pyplot as plt
import matplotlib.image as mpimg

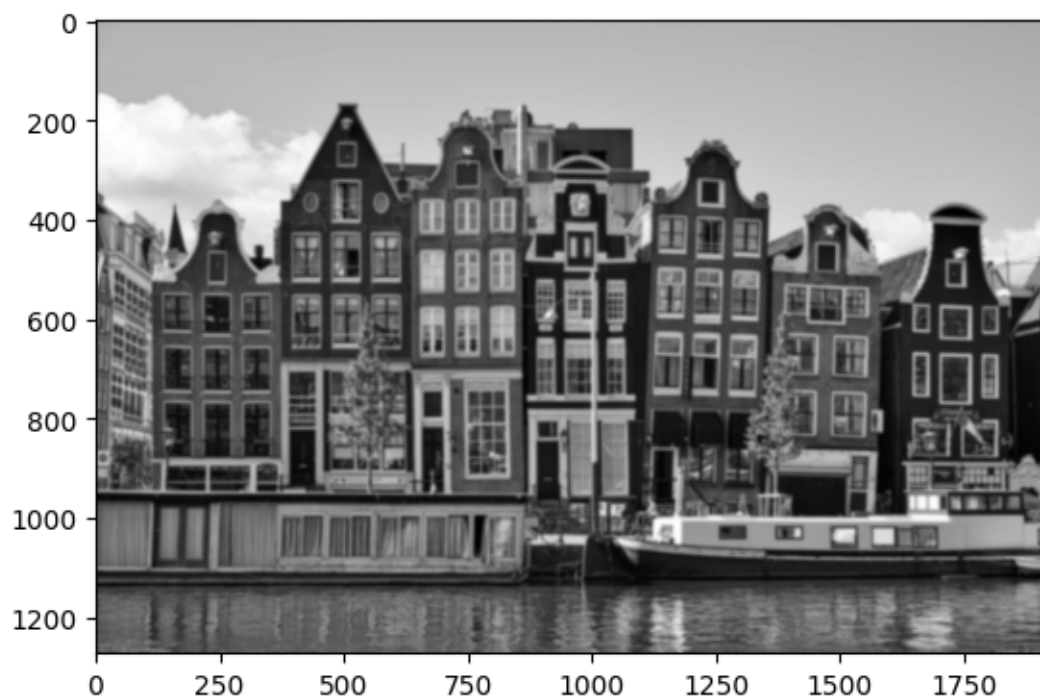
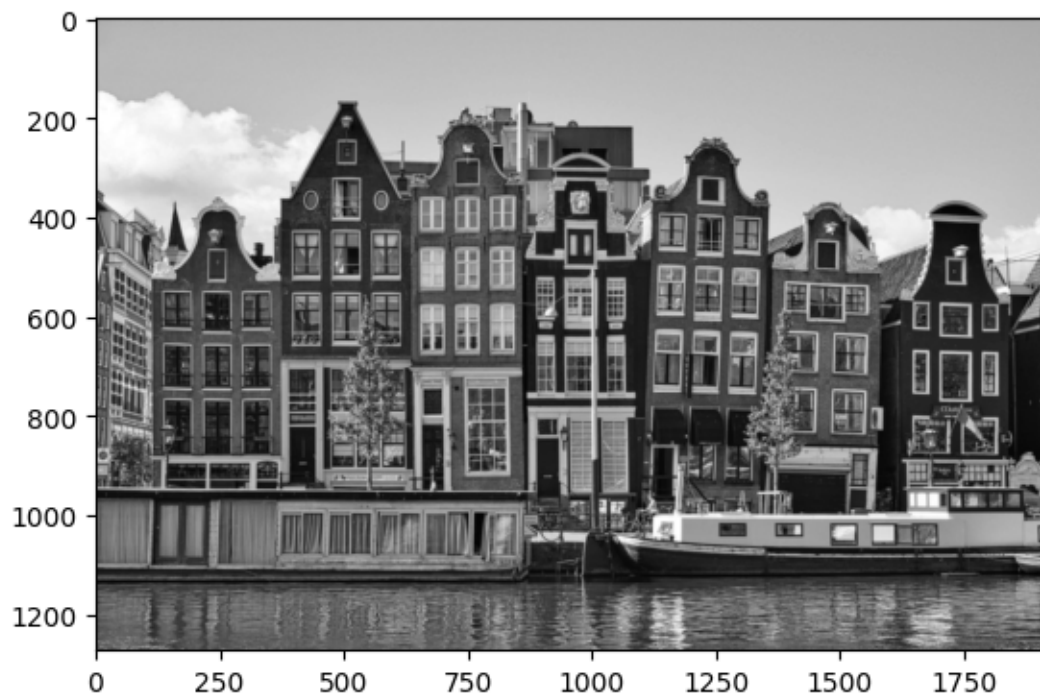
# Read the image
img = cv2.imread('buildings.jpg')

# Convert to grayscale
gray = cv2.cvtColor(img, cv2.COLOR_BGR2GRAY)

#Gaussian filter
blur = cv2.GaussianBlur(gray,(5,5),0)
plt.imshow(blur, cmap='gray')
plt.show()

#Gaussian filter
blur = cv2.GaussianBlur(gray,(15,15),0)
plt.imshow(blur, cmap='gray')
```

```
plt.show()
```



3.d. Apply following Laplacian filters over the image and compare the results

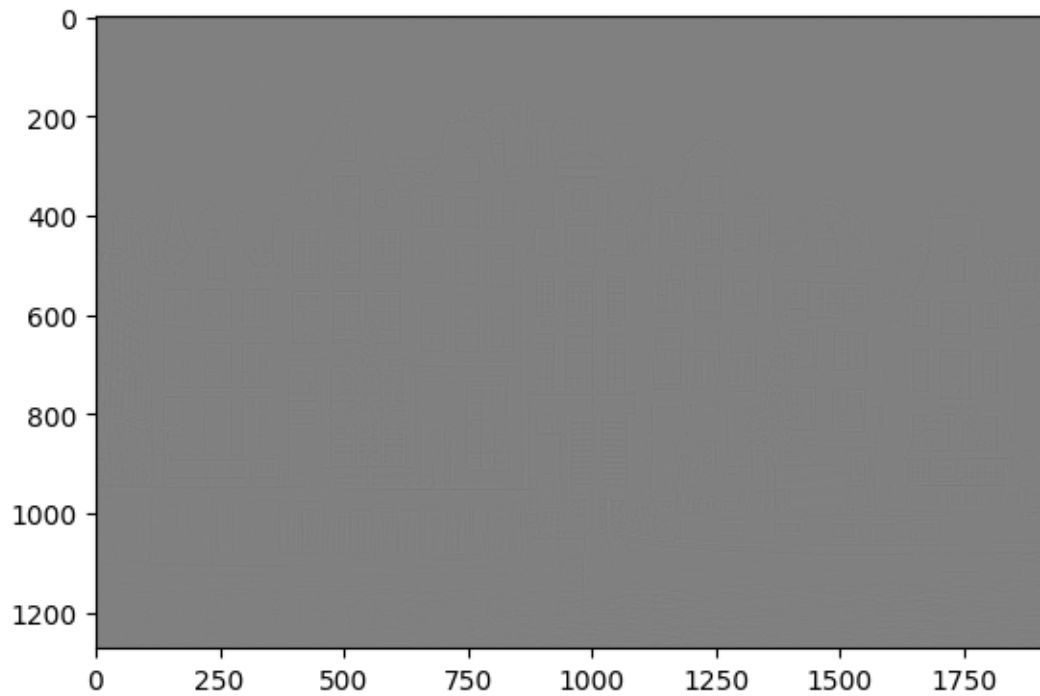
```
[49]: #Apply following Laplacian filters over the image
```

```
import cv2
import numpy as np
import matplotlib.pyplot as plt
import matplotlib.image as mpimg

# Read the image
img = cv2.imread('buildings.jpg')

# Convert to grayscale
gray = cv2.cvtColor(img, cv2.COLOR_BGR2GRAY)

#laplacian filter
laplacian = cv2.Laplacian(gray,cv2.CV_64F)
plt.imshow(laplacian, cmap='gray')
plt.show()
```



#Results:

The edges are more visible in the first image than the second image.

1 what will happen if the kernel weights are negated ?

If you negate the kernel weights, you change the way the convolution operation works. Instead of emphasizing certain features, the convolution operation will now suppress them. This is because convolution is a linear operation, and flipping the sign of the kernel weights is equivalent to changing the phase of the operation.