Task_3

April 26, 2023

0.1 Task 3 – Morphological Operators

Use morphological filtering on the mask obtained from Task 2.

a) Successively apply morphological opening and closing on the mask (imopen, imclose).

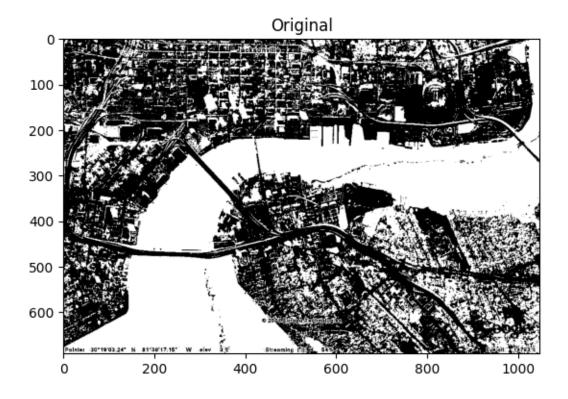
```
[]: from matplotlib import pyplot as plt import cv2 import numpy as np
```

```
[]: # import the binarized image
img = cv2.imread("Binary_Image.png",cv2.IMREAD_GRAYSCALE)
plt.imshow(img,cmap="gray")
plt.show()
```

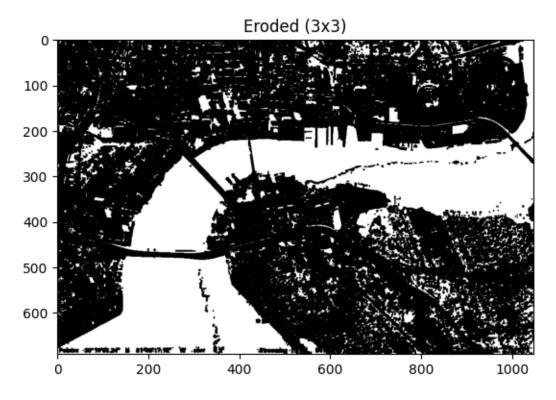


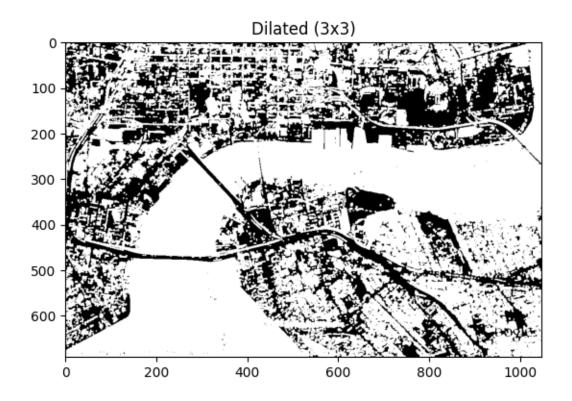
```
[]: # Show eroded and dilated images
plt.plot()
plt.imshow(img, cmap='gray')
plt.title('Original')
```

[]: Text(0.5, 1.0, 'Original')



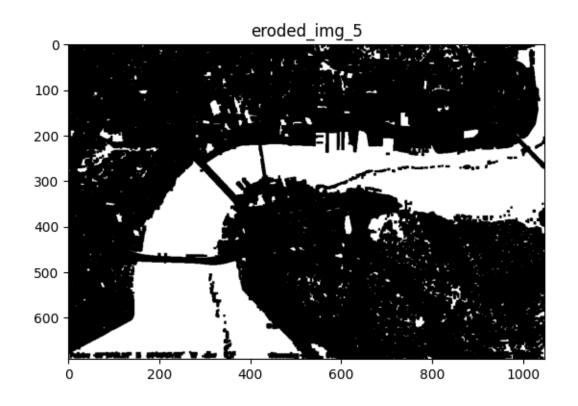
```
[]: # Show eroded and dilated images
plt.plot()
plt.imshow(eroded_img_3, cmap='gray')
plt.title('Eroded (3x3)')
plt.show()
plt.plot()
plt.imshow(dilated_img_3, cmap='gray')
plt.title('Dilated (3x3)')
plt.show()
```

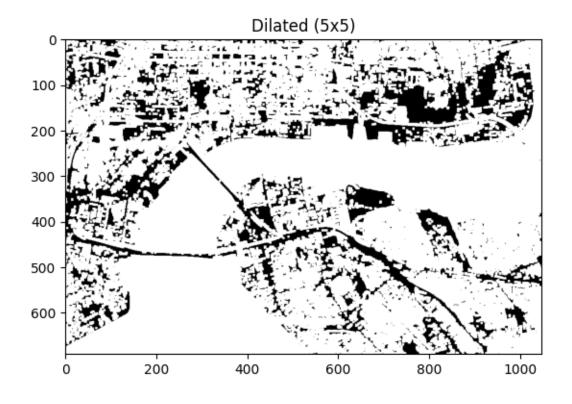




```
[]: plt.plot()
  plt.imshow(eroded_img_5, cmap='gray')
  plt.title('eroded_img_5')
  plt.show()

plt.plot()
  plt.imshow(dilated_img_5, cmap='gray')
  plt.title('Dilated (5x5)')
  plt.show()
```





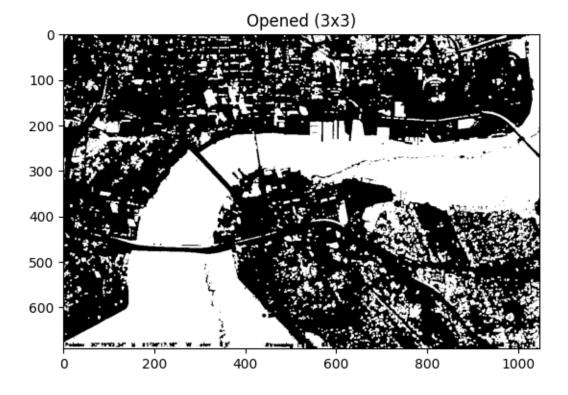
```
[]: # Apply opening and closing with 3x3 kernel
  opened_img_3 = cv2.morphologyEx(img, cv2.MORPH_OPEN, kernel_3)
  closed_img_3 = cv2.morphologyEx(img, cv2.MORPH_CLOSE, kernel_3)

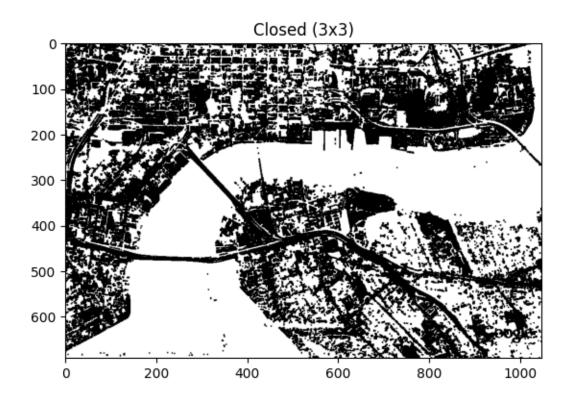
# Apply opening and closing with 5x5 kernel
  opened_img_5 = cv2.morphologyEx(img, cv2.MORPH_OPEN, kernel_5)
  closed_img_5 = cv2.morphologyEx(img, cv2.MORPH_CLOSE, kernel_5)

cv2.imwrite("Filtered_Image.png",eroded_img_5)
```

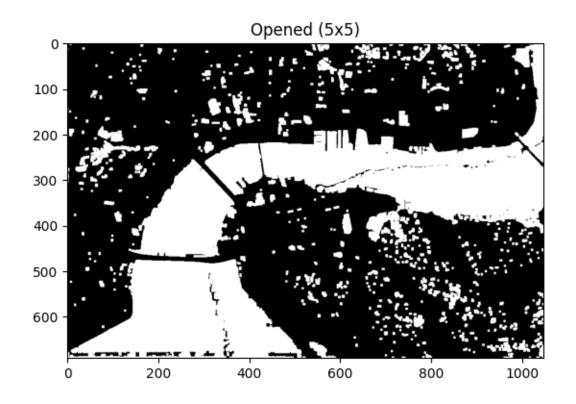
[]: True

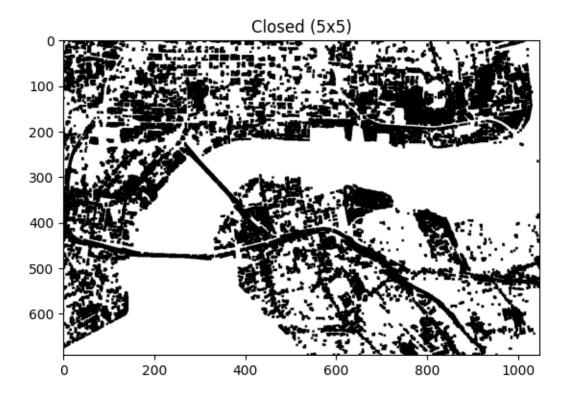
```
[]: # Show opened and closed images
plt.plot()
plt.imshow(opened_img_3, cmap='gray')
plt.title('Opened (3x3)')
plt.show()
plt.plot()
plt.imshow(closed_img_3, cmap='gray')
plt.title('Closed (3x3)')
plt.show()
```





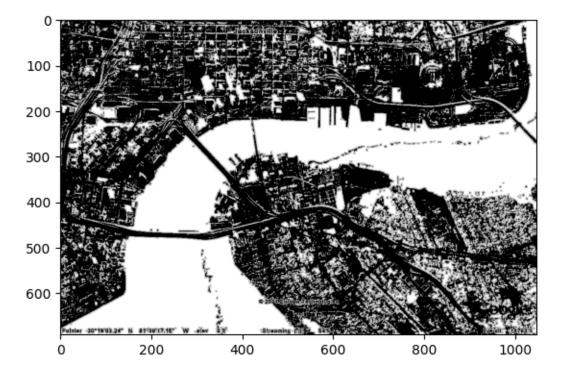
```
[]: plt.plot()
   plt.imshow(opened_img_5, cmap='gray')
   plt.title('Opened (5x5)')
   plt.show()
   plt.plot()
   plt.imshow(closed_img_5, cmap='gray')
   plt.title('Closed (5x5)')
   plt.show()
```





b) Visualize an overlay of the enhanced image from Task 1 and the morphologically filtered mask.

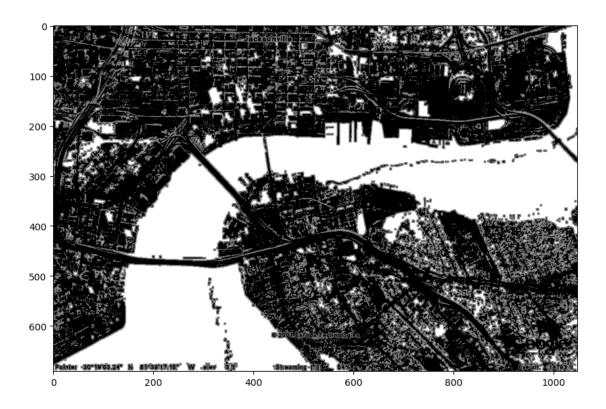
```
[]: Overlay_enhanced_image = cv2.add(img, eroded_img_5)
   plt.figure()
   plt.imshow(Overlay_enhanced_image,cmap="gray")
   cv2.imwrite("Overlayed_Enhanced_Image.png", Overlay_enhanced_image)
   plt.show()
```



```
[]: # Blend the images using addWeighted function
blended = cv2.addWeighted(img, 0.5, eroded_img_5, 0.5, 0)

# Display the blended image
plt.figure(figsize= (10,10))
plt.imshow(blended,cmap="gray")
plt.show()

# Save the blended image
cv2.imwrite("Overlayed_Image.png", blended)
```



[]: True

d) Are the results satisfactory? What are the limitations of this approach for foreground-background separation?

We can see that the non water surfaces also consists of white pixels. This completely gives the dilemma on identifying the water bodies instintisly. So, due to this I can conclude that the results were satisfactory. Limitation of this approach is that it relies on the assumption that the foreground and background regions can be separated by a threshold value in the original grayscale image. This may not always be the case, especially if the foreground and background regions have similar intensity values or if there are regions of intermediate intensity that are difficult to classify. erosion adn dilation leads to loss of the identifying data in the image.

e) Test your main function with a different input image of your choice. Do you notice a significant difference for the chosen input image?

I had tested the main function with different image from the google. The things I observed were:

1. Histogram diffece from the assignment Image. This leads the difficulties in selection of threshold values for the binarizatuion.

2. Still the background whater bodies and some parts in tyhe images are the white. It may be because of the low pixels in the background image.