

CS-663 Assignment 3 Q1

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October 18, 2020

1 (30 points) Harris Corner Detection

Input Image :

- 1/data/boat.mat

Assume the pixel dimensions to be equal along both axes, i.e., assume an aspect ratio of 1:1 for the axes.

Shift and rescale the intensities in the image to lie within the range $[0, 1]$.

Implement the Harris corner detector algorithm. The parameters underlying this algorithm are: two Gaussian smoothing levels involved in computing the structure tensor (the first Gaussian to smooth the image before computing the gradient, the second Gaussian for the weighted averaging to compute the structure tensor), the constant k in the corner-ness measure. Tune these three parameters to get the best results.

- (16 points) Write a function `myHarrisCornerDetector.m` to implement this.
- (4 points) Display the derivative images, corresponding to the partial derivatives along the X and Y axes.
- (4 points) Display an image (along with a colormap) of the principal eigenvalue of the structure tensor evaluated at each pixel. Display another image (along with a colormap) of the other eigenvalue of the structure tensor evaluated at each pixel.
- (6 points) Display the image (along with a colormap) of the Harris corner-ness measure. Tune the free parameters such that positive values in this image should correspond to “corner” structures in the image. Report all three parameter values used.

```
1  import numpy as np
2  import cv2
3  import matplotlib.pyplot as plt
4  import h5py
5  from tqdm import tqdm
6
7  def normalize_image(image):
8      """Normalize the image to remain between 0-1
9      Take the image and use the Min-Max normalization criterion to normalize the images
10
11      inputs: image(input image to be normalized)
12      outputs: normalized(normalized image)
13      """
14      out = image.copy()
15      normalized = (out-np.min(out))/(np.max(out)-np.min(out))
16      return normalized
17
18  def get_image(filename):
19      """Extract the image from mat file
20      inputs: filename(filepath of the .mat file)
21      outputs: the image as numpy array after normalizing
22      """
23      f = h5py.File(filename,"r")
24      out = f.get('imageOrig')
25      out = np.array(out)
```

```

26     return normalize_image(out)
27
28 def dnorm(x,sigma,mu=0):
29     """Calculates the 1D Gaussian kernel
30     inputs: x(kernel position),sigma(standard deviation of the kernel),mu(optional, default=0)
31     outputs: 1D kernel
32     """
33     return (1.0/(np.sqrt(2*np.pi)*sigma))*np.e**(-(((x - mu)/sigma)**2) / 2)
34
35 def gaussian_kernel(ksize,sigma):
36     """Calculates and returns 2D the Gaussian Kernel
37     inputs: ksize(Gaussian Kernel Size),sigma(standard deviation)
38     outputs: kernel_2D (2D gaussian kernel)
39     """
40     kernel_1D = np.linspace(-(ksize // 2), ksize // 2, ksize)
41     for i in range(ksize):
42         kernel_1D[i] = dnorm(kernel_1D[i], sigma=sigma)
43
44     # computers outer product of two 1-D gaussian kernels
45     # to produce a 2D Gaussian Kernel
46     kernel_2D = np.outer(kernel_1D.T, kernel_1D.T)
47     kernel_2D *= 1.0/np.sum(kernel_2D)
48     return kernel_2D
49
50 def convolution(image,kernel_size,sigma):
51     """Performs Convolution of the image with a Gaussian kernel
52     inputs: image(input image),kernel_size(Gaussian kernel size),
53            sigma(Gaussian Kernel Standard deviation)
54     outputs: output (convoluted image with the kernel)
55     """
56     kernel = gaussian_kernel(kernel_size,sigma=sigma)
57     image_row, image_col = image.shape
58     kernel_row, kernel_col = kernel.shape
59     output = np.zeros(image.shape)
60
61     # CREATING ZERO-PADDED IMAGE
62     pad_height = int((kernel_row - 1) / 2)
63     pad_width = int((kernel_col - 1) / 2)
64     padded_image = np.zeros((image_row + (2 * pad_height), image_col + (2 * pad_width)))
65     padded_image[pad_height:padded_image.shape[0] - pad_height, pad_width:padded_image.shape[1]
66                 -pad_width] = image
67
68     # CONVOLUTION OPERATION DONE HERE
69     for row in tqdm(range(image_row),desc="Gaussian Convolution"):
70         for col in range(image_col):
71             output[row, col] = np.sum(kernel * padded_image[row:row + kernel_row,
72             col:col + kernel_col])
73     output = (output/np.max(output))
74
75     return output
76
77 def calculate_gradient(image):
78     """Calculate the image gradient
79     inputs: image (input image)
80     outputs: dy(gradient along Y),dx(gradient along X),
81            Ixx(square of gradient along X),Iyy(square of gradient along Y),
82            Ixy(product of the gradient along X and Y)
83     """
84     dy, dx = np.gradient(image)
85     Ixx = dx**2
86     Ixy = dy*dx
87     Iyy = dy**2
88     return (dy,dx,Ixx,Iyy,Ixy)
89
90 def cornernessMeasure(Sxx,Syy,Sxy,k):

```

```

91     """Calculate the cornerness of the window
92     inputs: Sxx,Syy,Sxy,k(cornerness constant)
93     outputs: det(determinant of matrix),trace(trace of matrix),r(cornerness measure)
94     """
95     det = (Sxx * Syy) - (Sxy**2)
96     trace = Sxx + Syy
97     r = det - k*(trace**2)
98     return r,det,trace
99
100 def findCorners(filename, window_size_blur, sigma_weights, k, thresh):
101     """
102     Finds and returns list of corners and new image with corners drawn
103     :param img: The original image
104     :param window_size: The size (side length) of the sliding window
105     :param k: Harris corner constant. Usually 0.04 - 0.06
106     :param thresh: The threshold above which a corner is counted
107     :return:
108     """
109
110     img = get_image(filename)
111     sigma_window = 0.3*((window_size_blur-1)*0.5 - 1) + 0.8
112     img = convolution(img,window_size_blur,sigma_window)
113     #img = img*255.0
114     img = np.rot90(img)
115
116     #Find x and y derivatives
117     (dy,dx,Ixx,Iyy,Ixy) = calculate_gradient(img)
118     height,width = img.shape
119
120     plt.figure()
121     plt.imshow(dy,cmap='inferno',origin="lower")
122     plt.title("Derivative along Y")
123     plt.colorbar()
124     plt.savefig('../images/y_derivative.png',cmap='inferno',bbox_inches="tight")
125
126     plt.figure()
127     plt.imshow(dx,cmap='inferno',origin="lower")
128     plt.title("Derivative along X")
129     plt.colorbar()
130     plt.savefig('../images/x_derivative.png',cmap='inferno',bbox_inches="tight")
131
132     cornerList = []
133     offset = int(window_size_blur/2)
134
135     eigvalues = np.zeros((height,width,2))
136     corner_img =np.zeros((height,width))
137
138     #Loop through image and find our corners
139     min_r = 1000000
140     max_r = 0
141     for y in tqdm(range(offset, height-offset),desc="Finding Corners..."):
142         for x in range(offset, width-offset):
143             #Calculate sum of squares
144             weight_kernel = gaussian_kernel(2*offset+1,sigma=sigma_weights)
145             windowIxx = Ixx[y-offset:y+offset+1, x-offset:x+offset+1]*weight_kernel
146             windowIxy = Ixy[y-offset:y+offset+1, x-offset:x+offset+1]*weight_kernel
147             windowIyy = Iyy[y-offset:y+offset+1, x-offset:x+offset+1]*weight_kernel
148             Sxx = windowIxx.sum()
149             Sxy = windowIxy.sum()
150             Syy = windowIyy.sum()
151
152             #Find determinant and trace, use to get corner response
153             r,det,trace = cornernessMeasure(Sxx,Syy,Sxy,k)
154             eigvalues[x,y,0] = (trace + np.sqrt(trace**2 - 4*det))/2
155             eigvalues[x,y,1] = (trace - np.sqrt(trace**2 - 4*det))/2

```

```

156         corner_img[x,y]=r
157
158         if(r>max_r):
159             max_r = r
160         if (r<min_r):
161             min_r = r
162         #If corner response is over threshold, color the point and add to corner list
163         if r > thresh:
164             cornerList.append([x, y, r])
165
166     print("Minimum Cornerness Value :",min_r)
167     print("Maximum Cornerness Value :",max_r)
168     return img, cornerList,corner_img ,eigvalues
169
170 def plotCorners(finalImg,cornerList):
171     plt.figure()
172     plt.imshow(finalImg,cmap="gray",origin="lower")
173     plt.colorbar()
174     plt.title("Red Corners in Image")
175     for i in cornerList:
176         k,j,l = i
177         plt.scatter(k,j,color="r",s=0.3,marker="*")
178     plt.savefig('../images/Harris.png',bbox_inches='tight')
179
180 def plotEigenValues(eigvalues):
181     plt.figure()
182     plt.imshow(np.rot90(eigvalues[:, :,0]),cmap='inferno')
183     plt.title("Eigen Value 1")
184     plt.colorbar()
185     plt.savefig('../images/Eigen_value1.png',cmap='inferno',bbox_inches='tight')
186     plt.figure()
187     plt.imshow(np.rot90(eigvalues[:, :,1]),cmap='inferno')
188     plt.title("Eigen Value 2")
189     plt.colorbar()
190     plt.savefig('../images/Eigen_value2.png',cmap='inferno',bbox_inches='tight')
191
192 def plotCornernessMeasure(corner_img):
193     plt.figure()
194     plt.imshow(np.rot90(corner_img),cmap='hot')
195     plt.title("Cornerness Measure Plot")
196     plt.colorbar()
197     #plt.show()
198     plt.savefig('../images/Cornerness.png',cmap='hot')
199
200 def HarrisCornerDetector(filename>window_size_blur, sigma_weights, k, thresh):
201     img, cornerList,corner_img ,eigvalues = findCorners(filename>window_size_blur, sigma_weights, k, thresh)
202     plotCorners(img,cornerList)
203     plotEigenValues(eigvalues)
204     plotCornernessMeasure(corner_img)
205
206
207
208 if __name__=="__main__":
209     filename="../data/boat.mat"
210     window_size = 7
211     k = 0.06
212     thresh = 0.8*10**(-5)
213     sigma_weights = 1.2
214     print("Window Size: " + str(window_size))
215     print("K-value for Cornerness measure: " + str(k))
216     print("Corner Response Threshold:" + str(thresh))
217     HarrisCornerDetector(filename, int(window_size),sigma_weights, float(k), thresh)

```

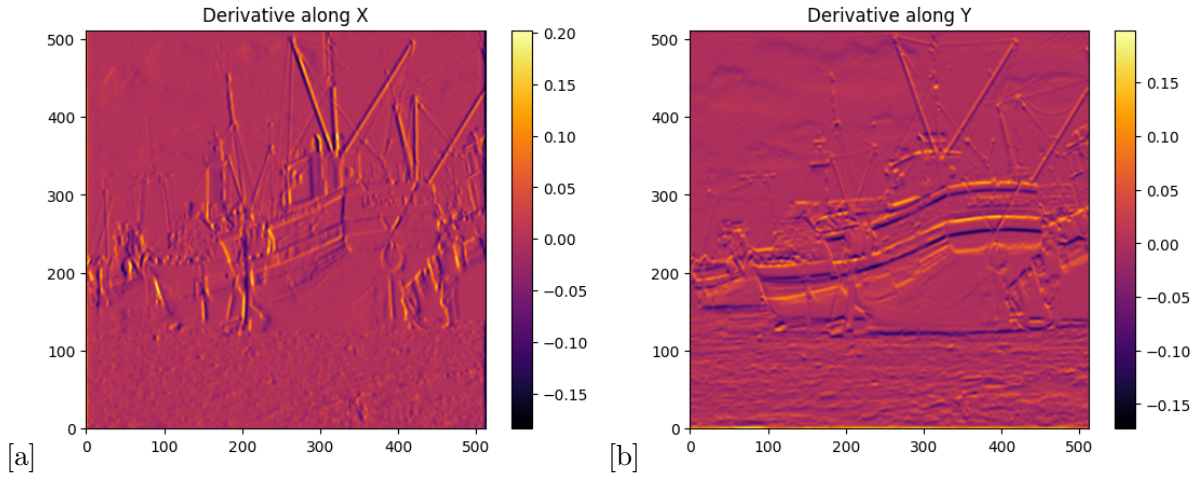


Figure 1: (a) X-derivative of 1/data/boat.mat (b) Y-derivative of 1/data/boat.mat

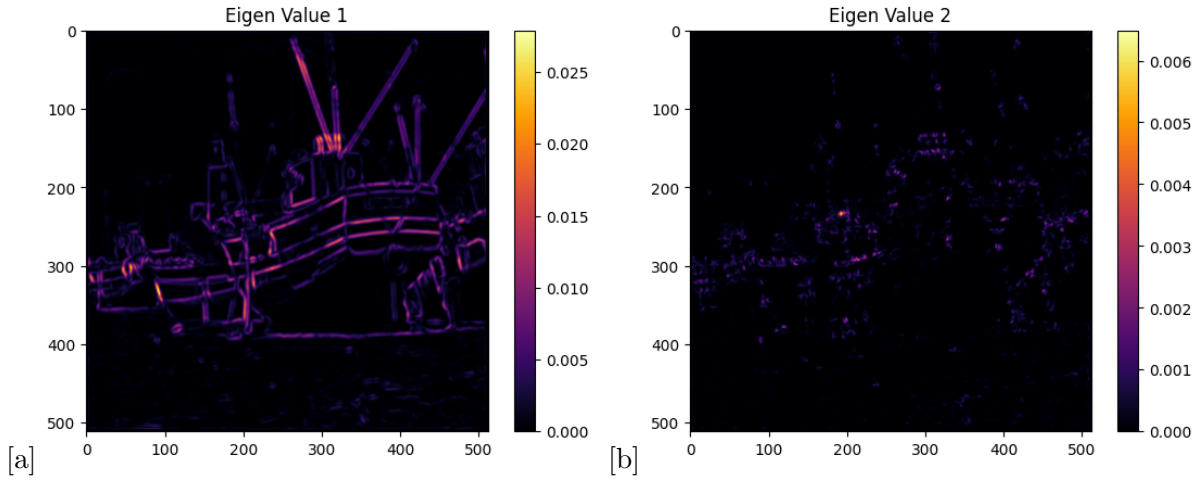


Figure 2: (a) Eigen Value 1 of 1/data/boat.mat (b) Eigen Value 2 of 1/data/boat.mat

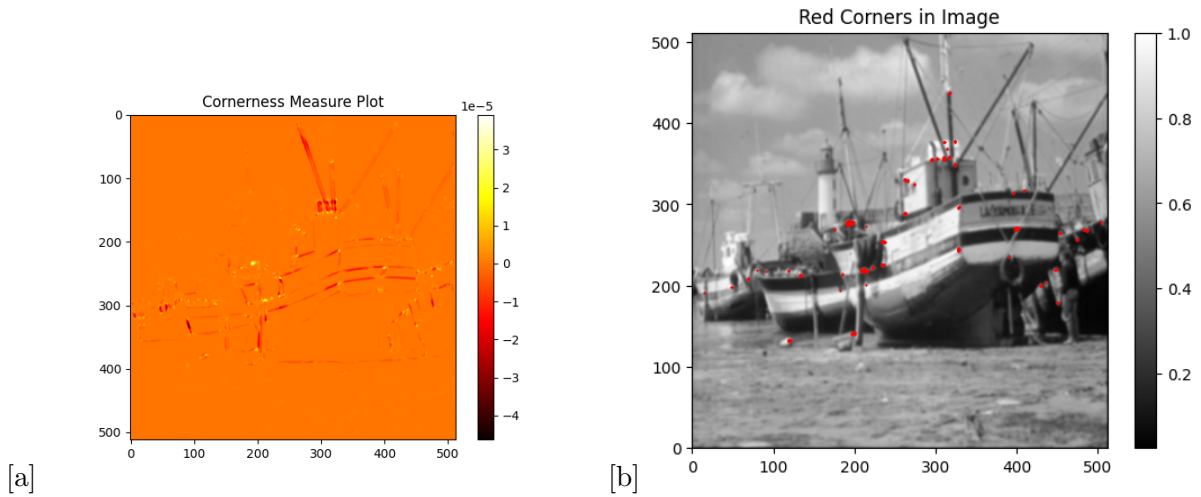


Figure 3: (a) Cornerness Measure Image of 1/data/boat.mat (b) Corner Plot on image of 1/data/boat.mat

Observation

The three parameters tuned are:

- Window Size of Gaussian to smooth the image : 7
- Gaussian Kernel Standard Deviation for weights : 1.2
- Cornerness constant (k): 0.06
- Threshold value for calculating corners : 0.8×10^{-5}

myMainScript.py

```
1 from myHarrisCornerDetection import HarrisCornerDetector
2
3 filename="./data/boat.mat"
4 window_size = 7
5 k = 0.06
6 thresh = 0.8*10**(-5)
7 sigma_weights = 1.2
8 print("Window Size: " + str(window_size))
9 print("K-value for Cornerness measure: " + str(k))
10 print("Corner Response Threshold:" + str(thresh))
11 HarrisCornerDetector(filename, int(window_size),sigma_weights, float(k), thresh)
```