

CANNY EDGE DETECTOR

DEEPAK PANT
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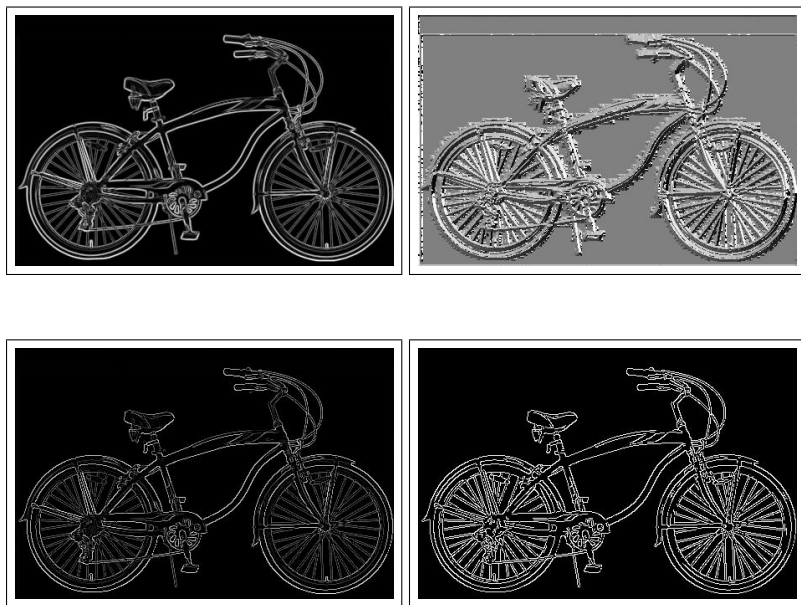
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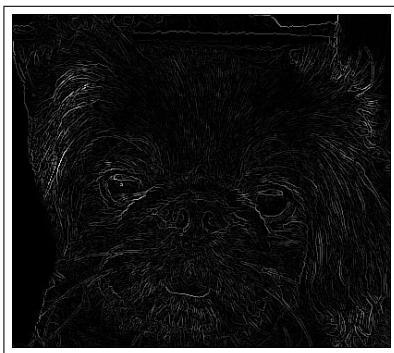
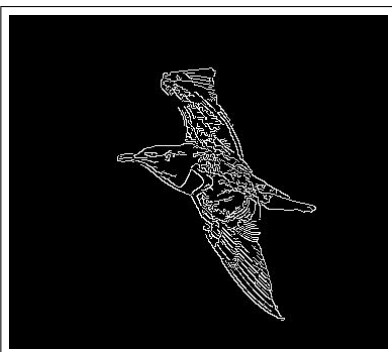
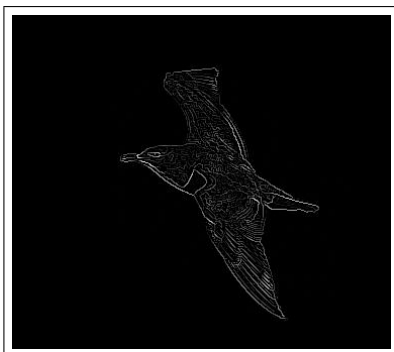
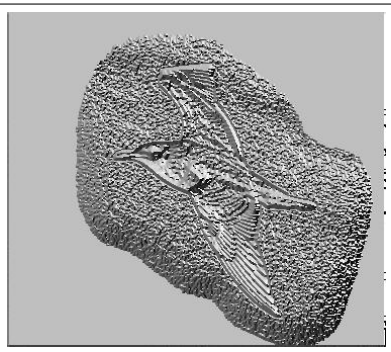
Abstract

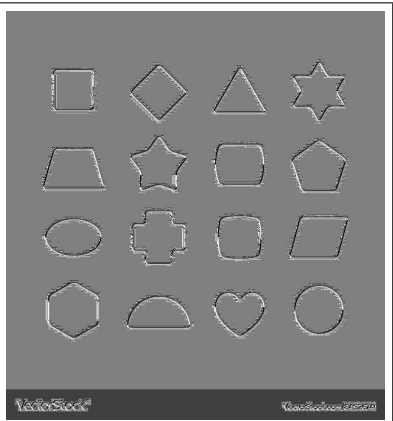
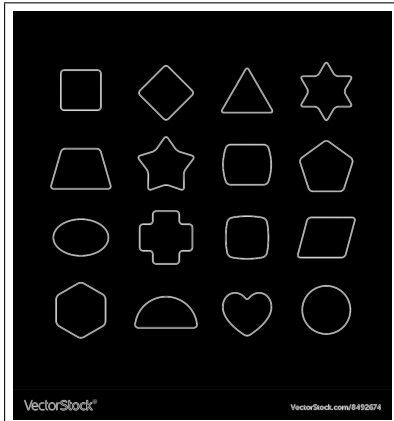
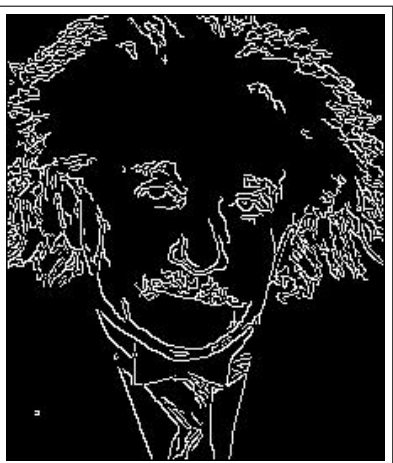
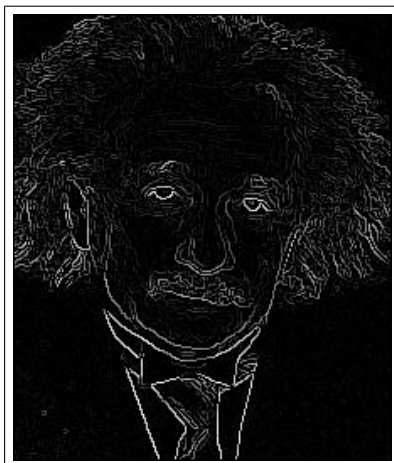
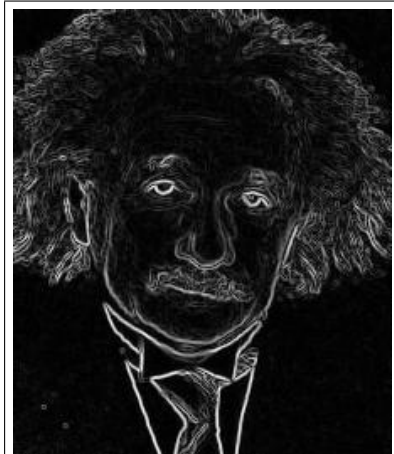
Canny edge detection is an algorithm to find edges in a image. Steps are taken to reduce noise. It involves computing smooth gradient of images, then applying non-maximal suppression to thin the images. To reduce noisy edges hysteresis thresholding is applied.

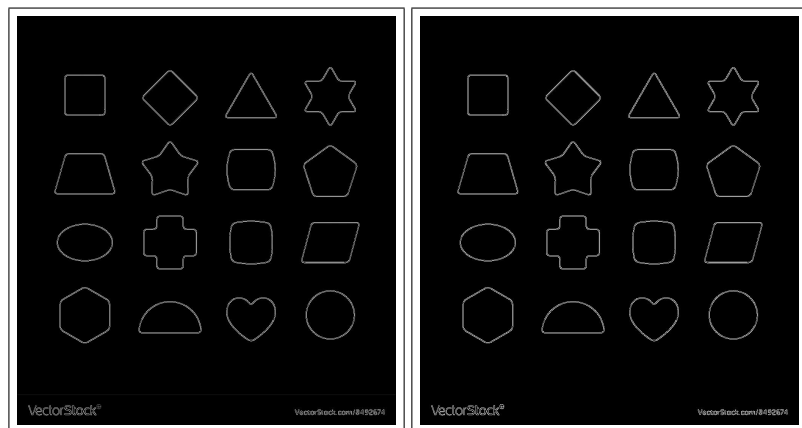
Firstly, I computed the luminance of given RGB image and then convolved it with sobel filter(which acts as a convolution of derivative operator with gaussian filter). Now the image has thick edges. Non-maximal suppression makes these edges one pixel thick. Edges which are less connected are then removed by hysteresis thresholding.

1 Output Images









2 Algorithms

Algorithm 1 NonMaximalSuppressor(gradstr, graddir)

```

1: thinimage = np.zeros(gradstr.shape)
2:
3: while  $i < \text{height}$  do
4:   while  $j < \text{width}$  do
5:
6:     if  $\text{closestdir}[i, j] == 0$  then
7:       if  $(\text{gradstr}[i, j] > \text{gradstr}[i, j + 1])$  and  $(\text{gradstr}[i, j] > \text{gradstr}[i, j - 1])$  then
8:          $\text{thinimage}[i, j] = \text{gradstr}[i, j]$ 
9:       else
10:         $\text{thinimage}[i, j] = 0$ 
11:      end if
12:    end if
13:
14:    if  $\text{closestdir}[i, j] == 45$  then
15:      if  $(\text{gradstr}[i, j] > \text{gradstr}[i + 1, j + 1])$  and  $(\text{gradstr}[i, j] > \text{gradstr}[i - 1, j - 1])$ 
then
16:         $\text{thinimage}[i, j] = \text{gradstr}[i, j]$ 
17:      else
18:         $\text{thinimage}[i, j] = 0$ 
19:      end if
20:    end if
21:
22:    if  $\text{closestdir}[i, j] == 90$  then
23:      if  $(\text{gradstr}[i, j] > \text{gradstr}[i + 1, j])$  and  $(\text{gradstr}[i, j] > \text{gradstr}[i - 1, j])$  then
24:         $\text{thinimage}[i, j] = \text{gradstr}[i, j]$ 
25:      else
26:         $\text{thinimage}[i, j] = 0$ 
27:      end if
28:    end if
29:
30:    if  $\text{closestdir}[i, j] == 135$  then
31:      if  $(\text{gradstr}[i, j] > \text{gradstr}[i + 1, j - 1])$  and  $(\text{gradstr}[i, j] > \text{gradstr}[i - 1, j +$ 
1]) then
32:         $\text{thinimage}[i, j] = \text{gradstr}[i, j]$ 
33:      else
34:         $\text{thinimage}[i, j] = 0$ 
35:      end if
36:    end if
37:
38:  end while
39: end while
40: return thinimage

```

Algorithm 2 HysteresisThresholding(img)

```

1: lowr = 0.15
2: hir = 0.35
3: tlow = max(img)*lowr
4: thigh = max(img)*hir
5: hysimg = np.copy(img)
6:
7: while i < height do
8:   while j < width do
9:     if img[i, j] > thigh then
10:      hysimg[i, j] = 1
11:     end if
12:     if img[i, j] < tlow then
13:      hysimg[i, j] = -1
14:     else
15:      hysimg[i, j] = 0
16:     end if
17:
18:     Apply BFS to find those weak pixels that are 8-connected to strong pixels
19:     Classify these pixels as strong pixels
20:   end while
21: end while
22: return hysimg

```

3 Observations

Canny edge detector showed correct output only if it was given good threshold values. Each image had to be given different pair of thresholds as some had strong edges and other had weak edges. I found that we could keep a low and high ratios of the maximum value of a pixel intensity of the image. The ratio between high and low was around 2.5-3.5.

In hysteresis thresholding, I suppressed some edges with very low intensities. Of the remaining some were weak and some were strong. Those pixels which are 8-connected to the strong pixels are then classified strong. In the end I suppressed the remaining weak pixels. For traversal through the image I used BFS. I used a multi source BFS which starts at each strong pixels and then propagates outward till it cannot reclassify weak pixels.

The convolution function I used in this assignment was the one built in scipy. When I ran the same code on the same image, I found the convolve I had developed earlier was very slow compared to the one already present in library.

4 Conclusion

The canny edge detector showed good output in case of tuning the thresholds. For each image different pair of thresholds were required. The high ratio was found to be around 0.35 and the lower one around 0.10. Hysteresis thresholding removed unnecessary edges. This canny edge detector was still found somewhat inferior to the one present in OpenCV library.