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BLG 453E  
Computer Vision  
Homework IV

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# 1 Usage

The code depends on numpy, opencv and PyQt5. If all requirements are satisfied, then code can be executed.

---

```
$ python3 filename
```

---

## 2 Corner Detection

In first part of homework, Harris Corner Detection[1] algorithm is implemented. In brief, the algorithm checks the gradient changes in windows, if the change satisfies threshold, we mark this point as a corner. It is a very simple definition of the algorithm. Before the calculating gradients, I convoluted the image with a Gaussian filter which has kernel size is three and  $\sigma = 1$ , in order to remove noise from image. Then we need to calculate image gradients( $I_x$  and  $I_y$ ), then we can compute G matrix.

$$G = \begin{bmatrix} \sum I_x^2 & \sum I_x I_y \\ \sum I_x I_y & \sum I_y^2 \end{bmatrix};$$

We choose a proper W (window size), and calculate eigenvalues of matrix, then  $\lambda_2$  computed and compared to an empiric threshold. If the  $\lambda_2$  is bigger than threshold point is marked as corner.

### 2.1 Implementation

I calculated the gradients with cconvolution filters. I tried Sobel [2] filter and Prewitt [3] filter. I decided to use Prewitt after several tries.

---

```
def imgConv(img, kernel, threshold=10):
    C = img.shape[2]
    res = np.ones(img.shape)
    for i in range(C):
        res[:, :, i] = signal.convolve2d(img[:, :, i], kernel, mode="same")
    res = np.abs(res)
    res[res < 15] = 0
    res = res.astype(np.float)
    res = 255 * (res - np.min(res)) / (np.max(res) - np.min(res))
    return res
```

```
smoothImage = cv2.GaussianBlur(img, (ksize, ksize), sigma)
Ix = imgConv(smoothImage, filter_x)
Iy = imgConv(smoothImage, filter_y)

# Calculate Ix2 and Iy2 and Ixy
Ixx = Ix**2
Iyy = Iy**2
Ixy = Iy*Ix
```

---

After calculating the gradients, sliding window is implemented. I choose threshold as 700. Output example is added.

---

```
for y in range(half, h-half):
    for x in range(half, w-half):
        # Normalize sums respect number of elements in window
        Sumxx = Ixx[y-half:y+half+1, x -
                    half:x+half+1].sum() / wsize**2
        Sumxy = Ixy[y-half:y+half+1, x -
                    half:x+half+1].sum() / wsize**2
        Sumyy = Iyy[y-half:y+half+1, x -
                    half:x+half+1].sum() / wsize**2

        G = np.array([[Sumxx, Sumxy], [Sumxy, Sumyy]])

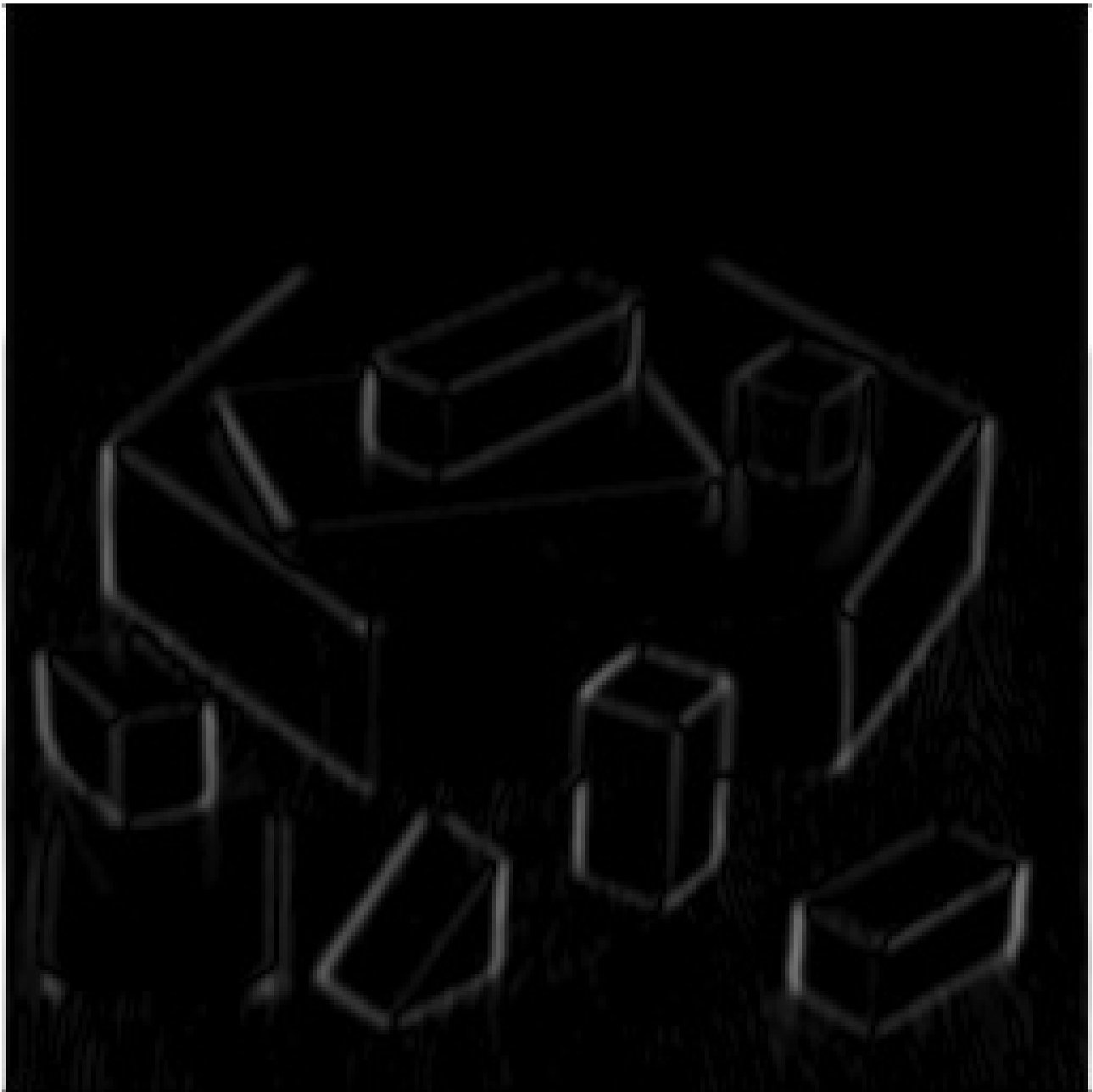
        lamdas, vector = np.linalg.eig(G)
        r = lamdas.min()

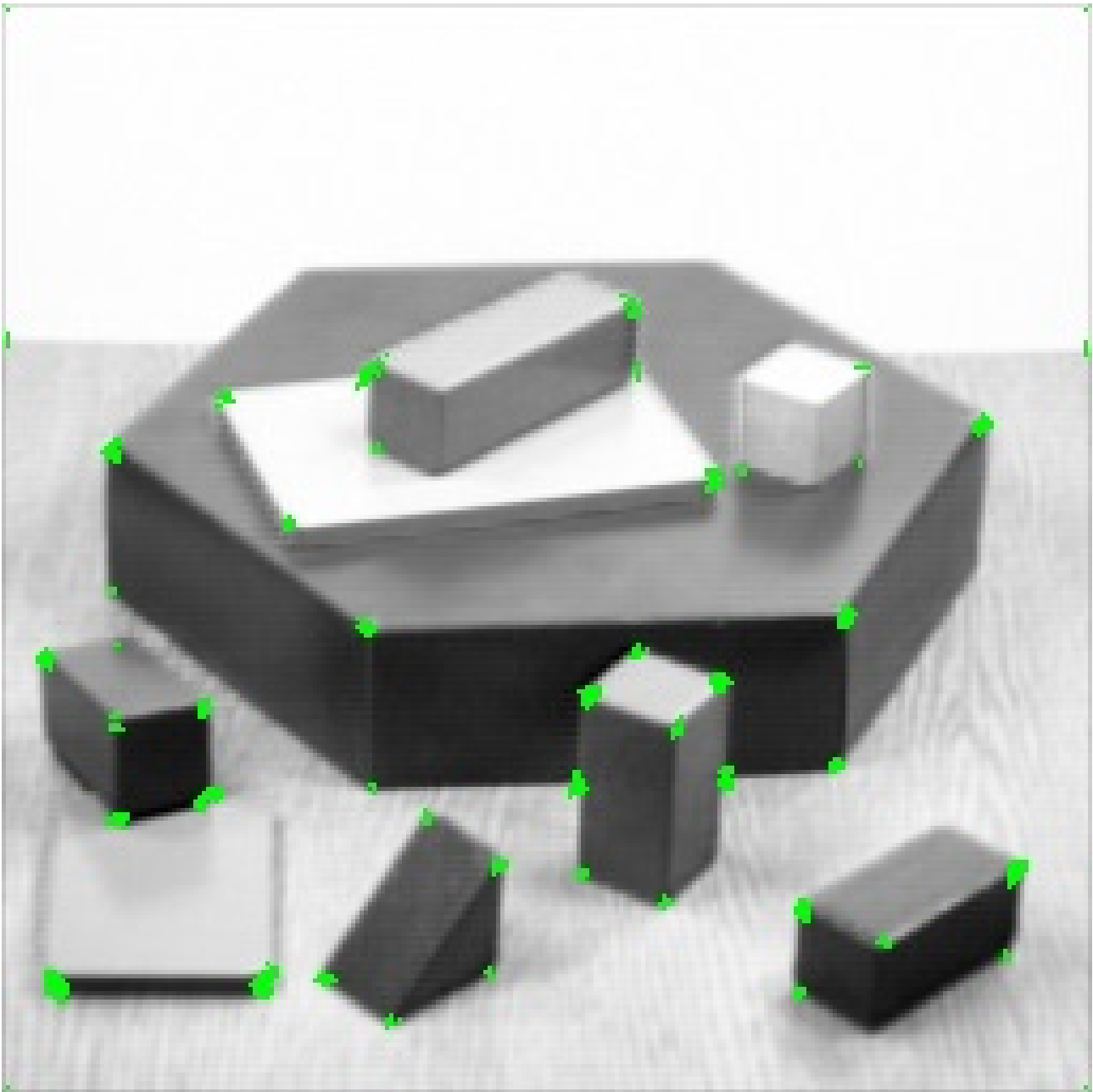
        if r > thresh:
            corner_x.append(x)
            corner_y.append(y)

color_img[corner_y, corner_x] = (0, 255, 0)
cv2.imwrite("corners.jpg", color_img.astype(np.uint8))
return color_img
```

---

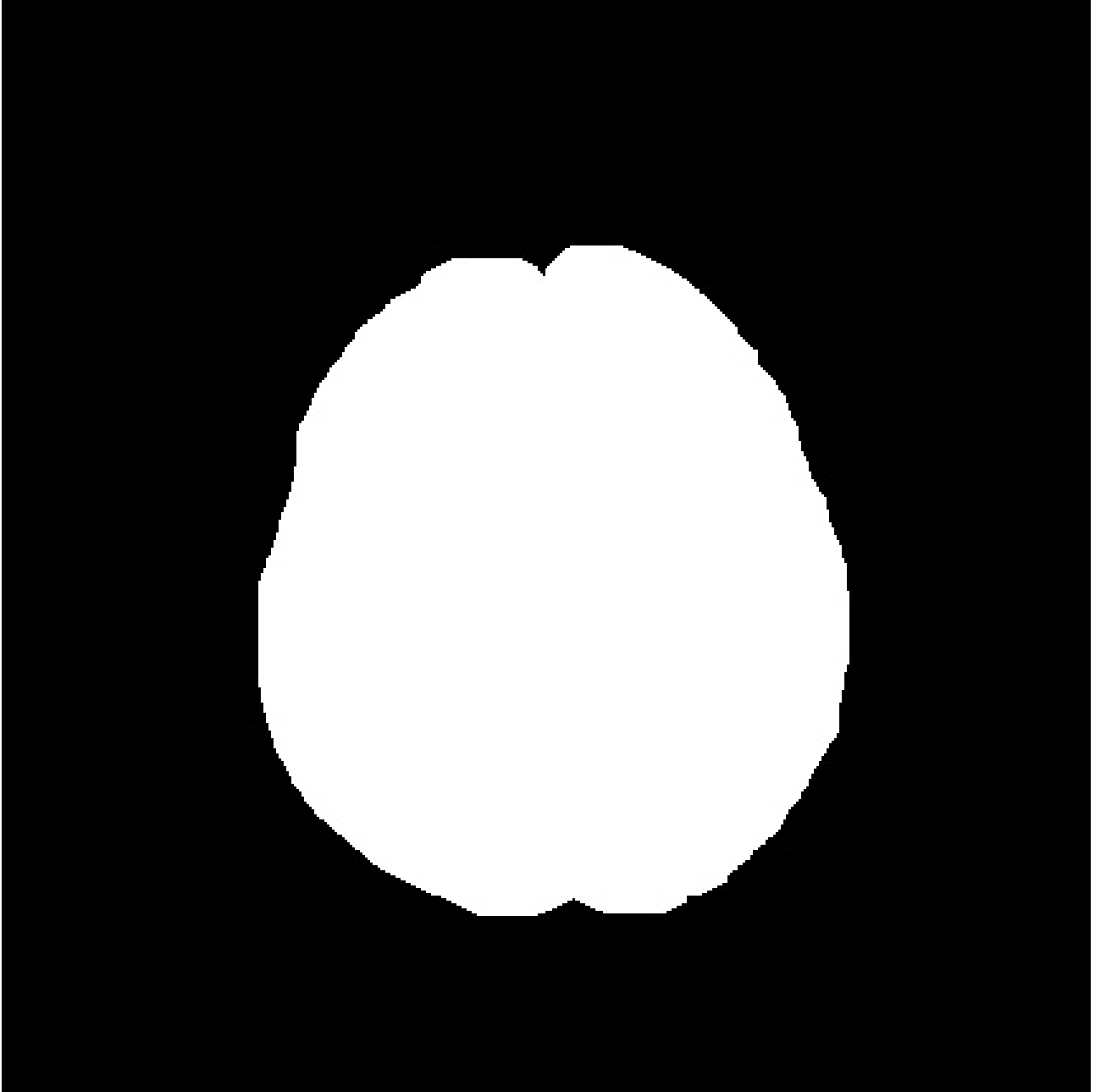






### 3 Segmentation

In this part homework, we implemented K-means [4] algorithm to MR image of Brain. Objective was detecting the tumor area and finding its boundaries. First I converted image to gray scale using opencv method. I used 55 as the threshold. Then I used "morphologyEx" method in opencv with two different shaped kernel in order get the mask given in the homework PDF.





### 3.1 Segmentation with K means

I select non-zero intensity valued points to cluster.

---

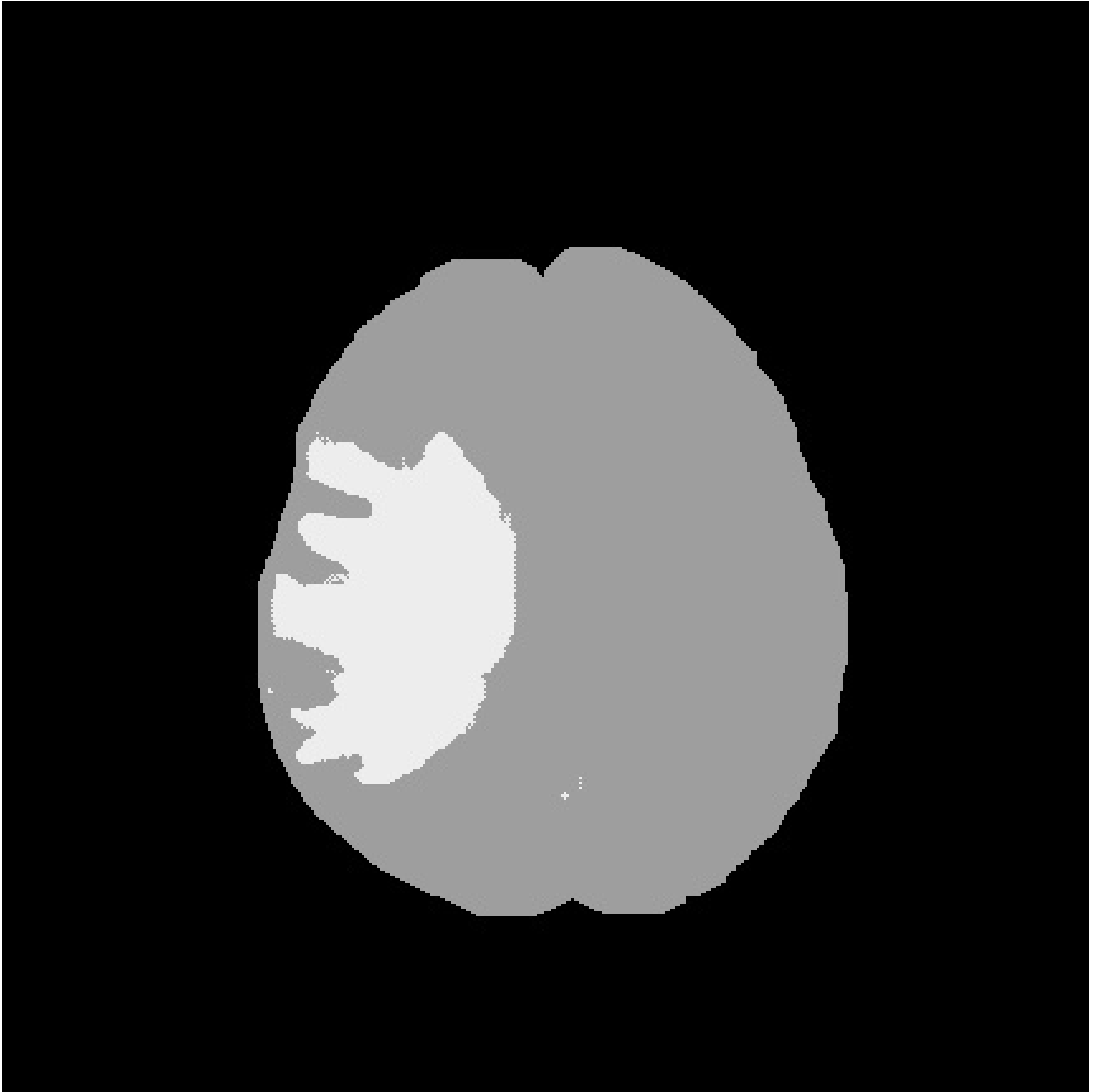
```
def Kmeans(img, K=2, n_iters=20):
    np.random.seed(1)
    shape = img.shape
    x, y = np.where(img > 0)
    centers = [randPoint(img, x, y) for i in range(K)]
    clusters = np.zeros(shape)
    # Converge logic should be different
    for i in range(n_iters):

        for i in range(len(x)):
            clusters[x[i], y[i]] = closest(
                img[x[i], y[i]], (x[i], y[i]), centers)
        for i in range(K):
            mean = np.sum(img[clusters == i+1]) / \
                len(np.where(clusters == i+1)[0])
            centers[i] = mean

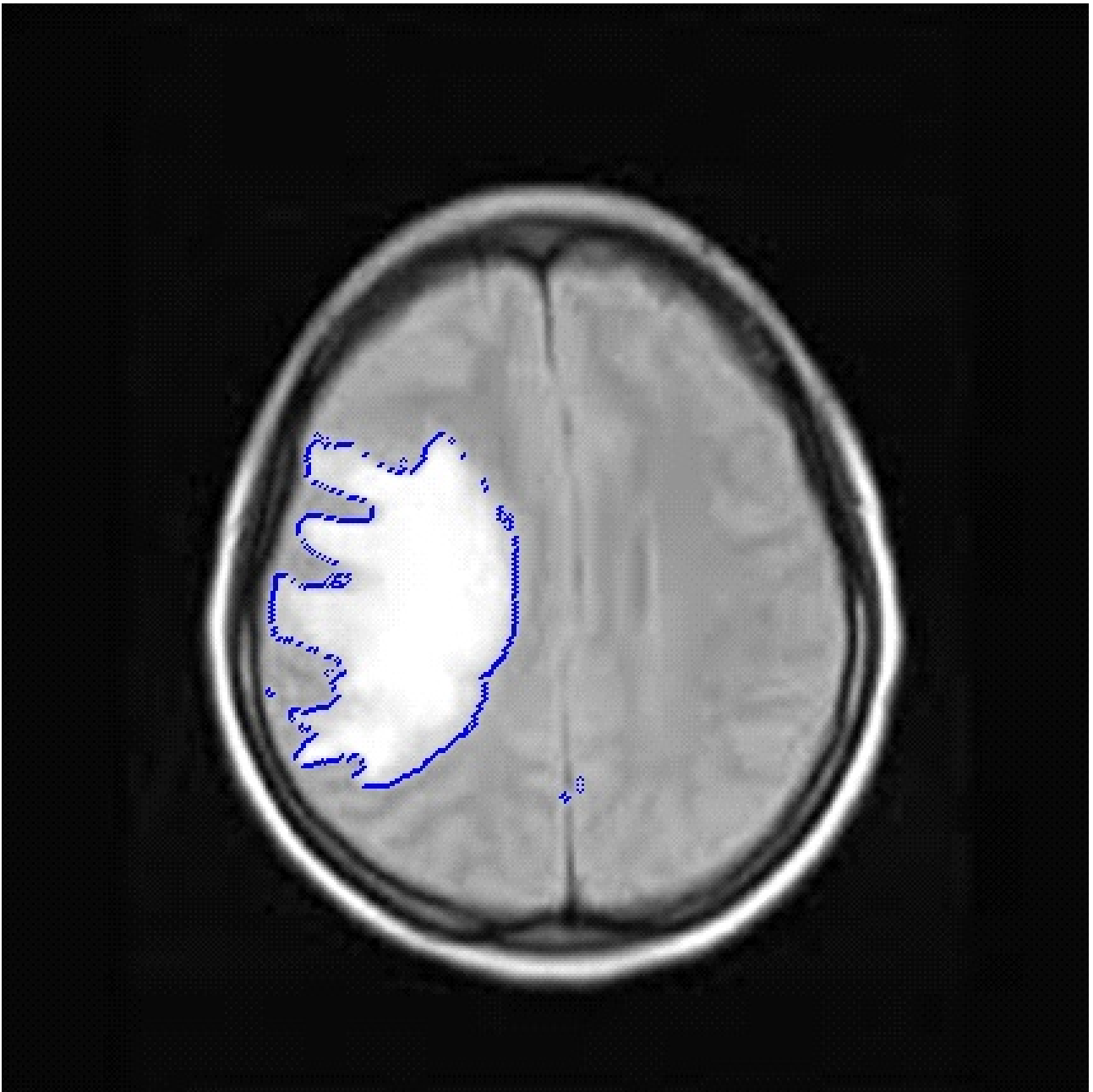
    clusters[clusters == 1] = round(centers[0])
    clusters[clusters == 2] = round(centers[1])
```

---

Then I painted clustered pixels to same color.



In order to get boundaries, I made an assumption that number of tumor pixels are lesser than number of normal brain pixels. Then I painted normal brain pixels to black, I get gradients from this image and used it as mask. Then I get the pixels that greater than zero to make them blue in original image.



## References

- [1] [Harris Corner Detection](#)
- [2] [Sobel Filter](#)
- [3] [Prewitt Filter](#)
- [4] [K Means](#)