# IMAGE SEGMENTATION ACTIVITY 7

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## Original Image



A Vulture

#### Segmentation using various methods:

- Thresholding
- Parametric segmentation
- Non-Parametric segmentation

#### Done using:

- Jupyter Notebook (Python)
  - Packages:

import matplotlib.pyplot as plt

import numpy as np

import cv2

from scipy.signal import find\_peaks

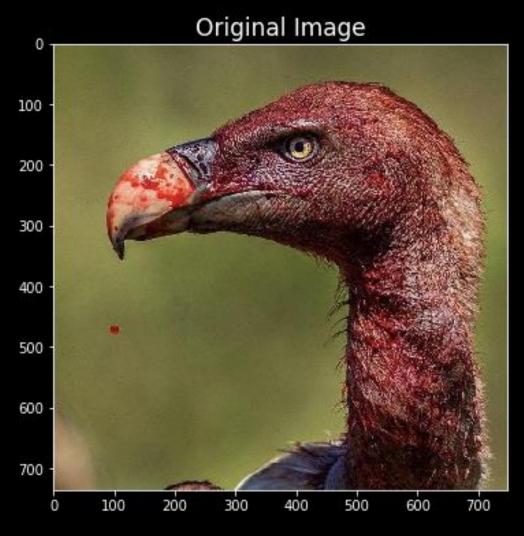
from skimage import img\_as\_float

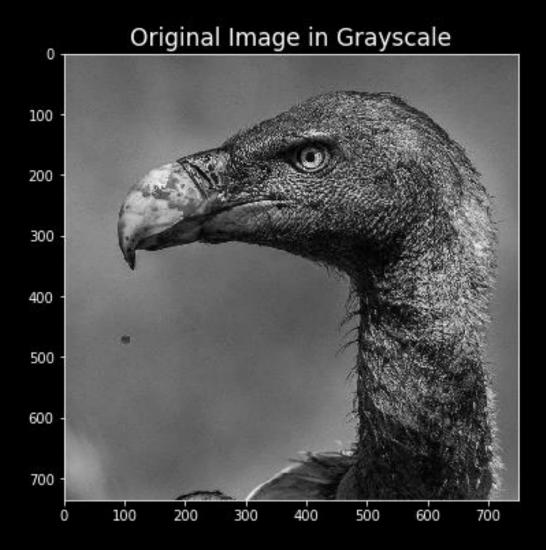
\* Photos taken from Google Images and Pinterest

## Thresholding

1

## Step 1 To Grayscale



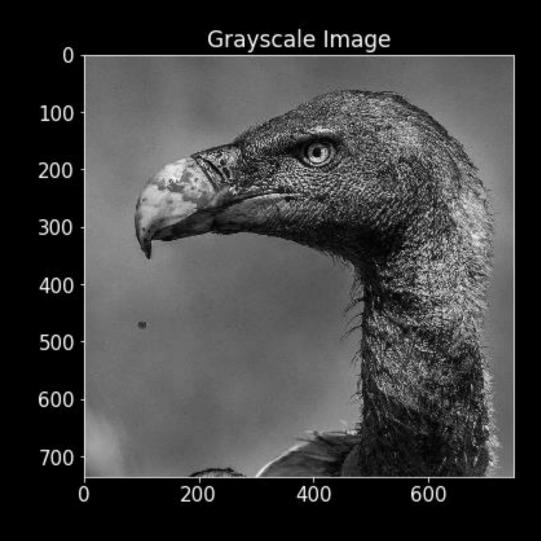


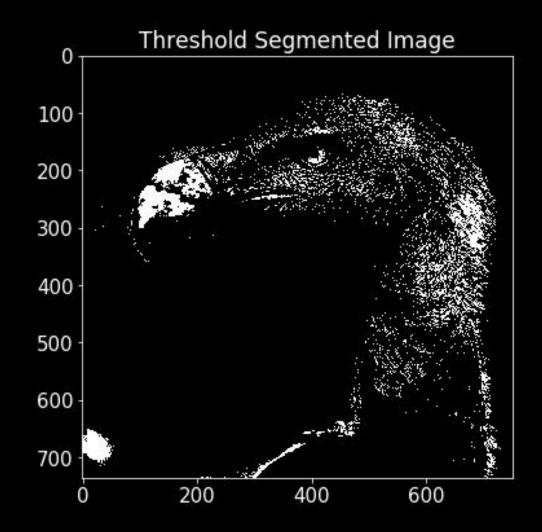
## Step 2 Thresholding

```
def thresh(gray):
    BW = gray > 145
    return BW
```

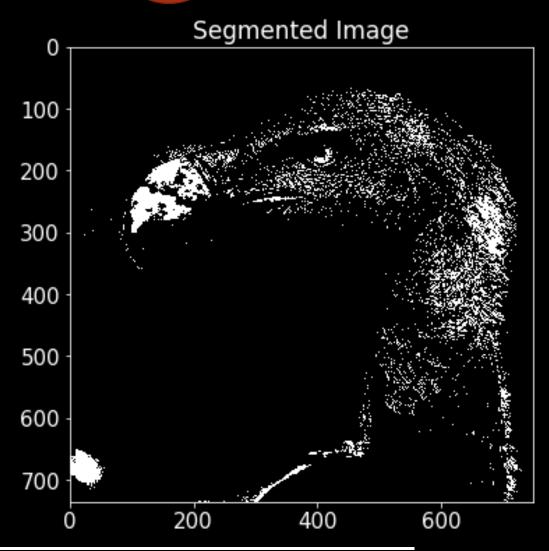
- 'gray' is the grayscale of the original image
- A threshold is set at 145
- The threshold value is adjustable in order to attain the best segmentation

### Step 2 Thresholding (Result)





## Step 2 Thresholding (Result)



- 'This segmentation method was able to produce the outline of the Vulture.
- However, these lines aren't definite.
   After all, this method has its limit since
  the threshold value may account for
  other parts of the picture like the
  white pigments outside the Vulture.

### Parametric Segmentation

2

## Step 1 NCC coordinates

- With M as the original image, the NCC coordinates were taken by using these equations on the left.
- 'blue' is now dependent on r and g.
- The image was converted to float for the calculation to avoid errors such as 'True divide by zero'.

```
M_f = img_as_float(M)
R,G,B = cv2.split(M_f)

# Overall pixels
I = R + G + B
r = R/I
g = G/I
b = 1-r-g
```

### Parametric Segmentation

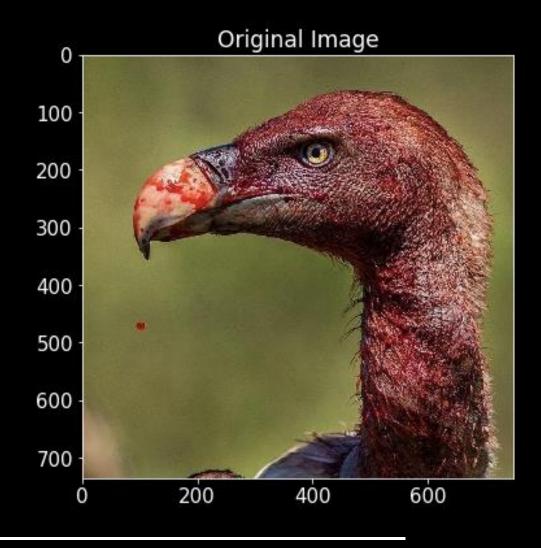
- Gaussian distribution was taken for both r and g.
- 'joint' is the product of the r and g's respective Gaussian distribution.
- The resulting image now has the 'joint' histogram.

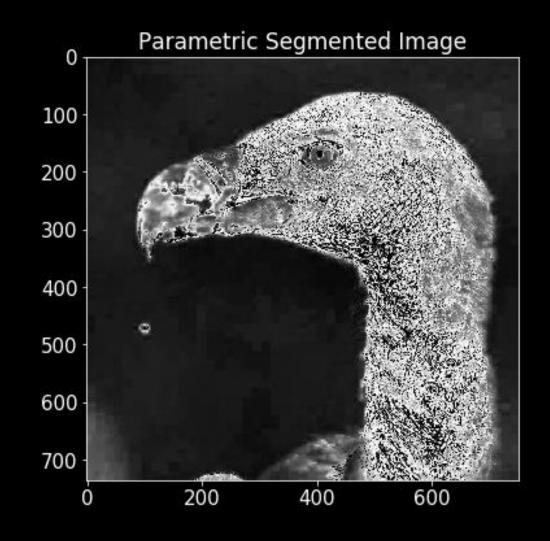
```
def Gauss_dist(c,c_s):
    cc = 1/(np.std(c_s)*np.sqrt(2*np.pi))
    cep = -(c-np.mean(c_s))**2/(2*np.std(c_s)**2)
    pc = cc*np.exp(cep)
    return pc
```

```
pr = Gauss_dist(r,r_s)
pg = Gauss_dist(g,g_s)
joint = pr*pg
```

## Step 2

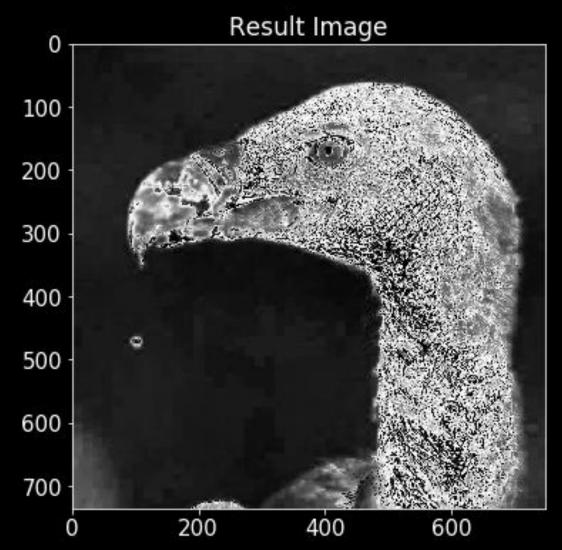
#### 2 Parametric Segmentation (Results)





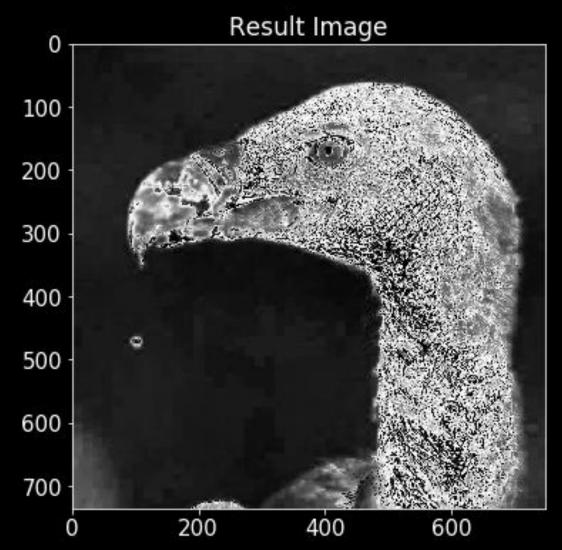
## Step 2 Parametric Segmentation (Results)

- 'This segmentation method was able to produce the outline of the Vulture accurately.
- This method also captured the blood droplet on the left side of the image.

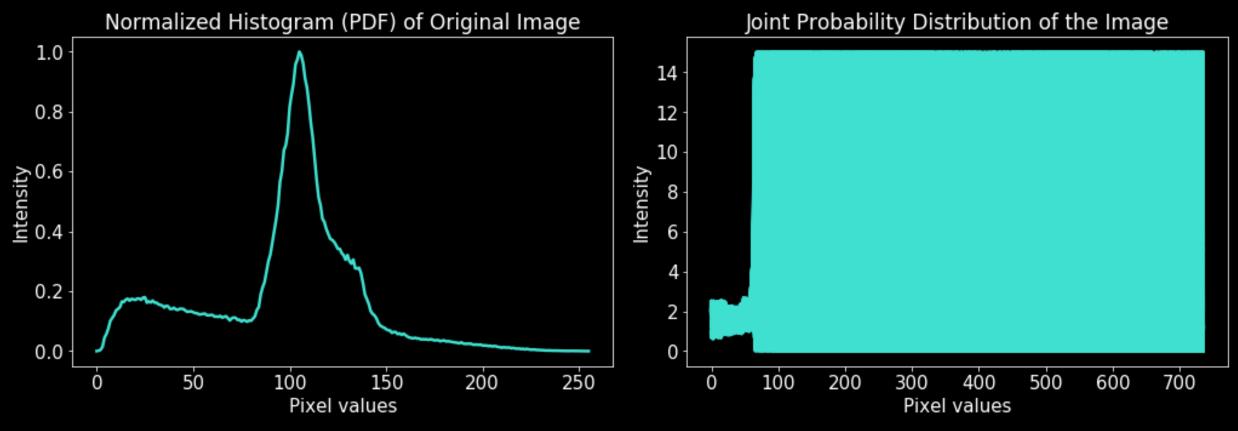


## Step 2 Parametric Segmentation (Results)

- 'This segmentation method was able to produce the outline of the Vulture accurately.
- This method also captured the blood droplet on the left side of the image.



## Step 2 Parametric Segmentation (Results)



- The difference in apparent in these histograms
- Most of the pixel values had the same max peak values
- Unknowingly, the x-axis reached 700 px. This prompted a new shape for the image

#### Non-Parametric Segmentation

3

#### Step 1 Non-Parametric Segmentation

```
def Non_parametric(image,patch,hist_sub):
    hsv = cv2.cvtColor(patch,cv2.COLOR_BGR2HSV)
    hsvt = cv2.cvtColor(image,cv2.COLOR_BGR2HSV)
    roihist = cv2.calcHist([hsv],[0, 1], None, [180, 256], [0, 180, 0, 256])
    cv2.normalize(roihist,roihist,0,255,cv2.NORM_MINMAX)
    dst = cv2.calcBackProject([image],[0,1],hist_sub,[0,180,0,256],1)
    # Now convolute with circular disc
    dst = cv2.calcBackProject([hsvt],[0,1],roihist,[0,180,0,256],1)
    disc = cv2.getStructuringElement(cv2.MORPH_ELLIPSE,(5,5))
    cv2.filter2D(dst,-1,disc,dst)
    # threshold and binary AND
    ret,thresh = cv2.threshold(dst,50,255,0)
    thresh = cv2.merge((thresh,thresh,thresh))
    res = cv2.bitwise_and(image,thresh)
    return res
```

- This was a code I found online:
   <a href="https://docs.opencv.org/master/dc/df6/tutorial\_py\_histogram\_backprojection.html?fbcl">https://docs.opencv.org/master/dc/df6/tutorial\_py\_histogram\_backprojection.html?fbcl</a>
  - id=lwAR1p80GPxC0mRmZoeJldTYYyu1xuq47la4o98vnb6qofh36EfWbo2pfiaTg
- I wasn't able to do the indexing algorithm for this.

#### Step 1 Non-Parametric Segmentation

```
def Non_parametric(image,patch,hist_sub):
    hsv = cv2.cvtColor(patch,cv2.COLOR_BGR2HSV)
    hsvt = cv2.cvtColor(image,cv2.COLOR_BGR2HSV)
    roihist = cv2.calcHist([hsv],[0, 1], None, [180, 256], [0, 180, 0, 256])
    cv2.normalize(roihist,roihist,0,255,cv2.NORM_MINMAX)
    dst = cv2.calcBackProject([image],[0,1],hist_sub,[0,180,0,256],1)
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    res = cv2.bitwise_and(image,thresh)
    return res
```

- This code uses a lot of packages from cv2.
- The histograms for the image and patch was produced using calcHist in cv2
- A disc was used as a filter mask
- A threshold was set in order to produce a fully contrasting image

#### Step 1 Non-Parametric Segmentation

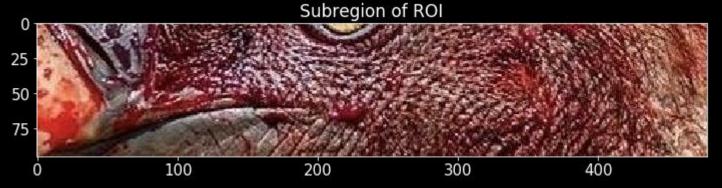
```
def Non_parametric(image,patch,hist_sub):
    hsv = cv2.cvtColor(patch,cv2.COLOR_BGR2HSV)
    hsvt = cv2.cvtColor(image,cv2.COLOR_BGR2HSV)
    roihist = cv2.calcHist([hsv],[0, 1], None, [180, 256], [0, 180, 0, 256])
    cv2.normalize(roihist,roihist,0,255,cv2.NORM_MINMAX)
    dst = cv2.calcBackProject([image],[0,1],hist_sub,[0,180,0,256],1)
    # Now convolute with circular disc
    dst = cv2.calcBackProject([hsvt],[0,1],roihist,[0,180,0,256],1)
    disc = cv2.getStructuringElement(cv2.MORPH_ELLIPSE,(5,5))
    cv2.filter2D(dst,-1,disc,dst)
    # threshold and binary AND
    ret,thresh = cv2.threshold(dst,50,255,0)
    thresh = cv2.merge((thresh,thresh,thresh))
    res = cv2.bitwise_and(image,thresh)
    return res
```

 The images were converted to HSV in order to produce a colored segmented image

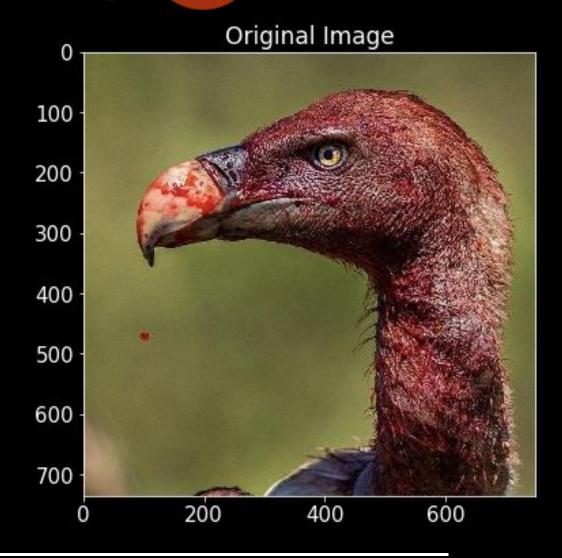
## Step 2 N

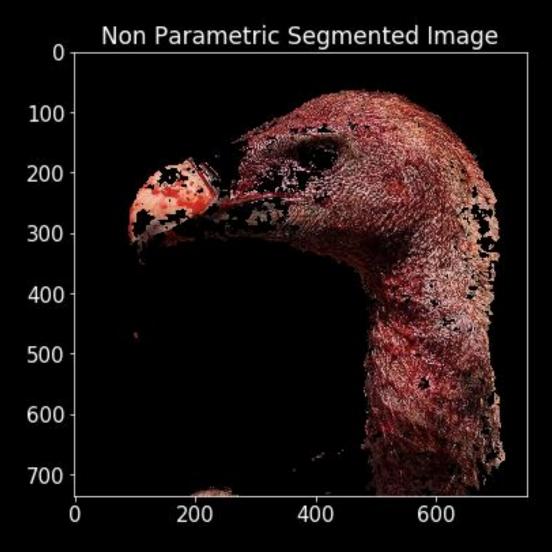
#### Non-Parametric Segmentation

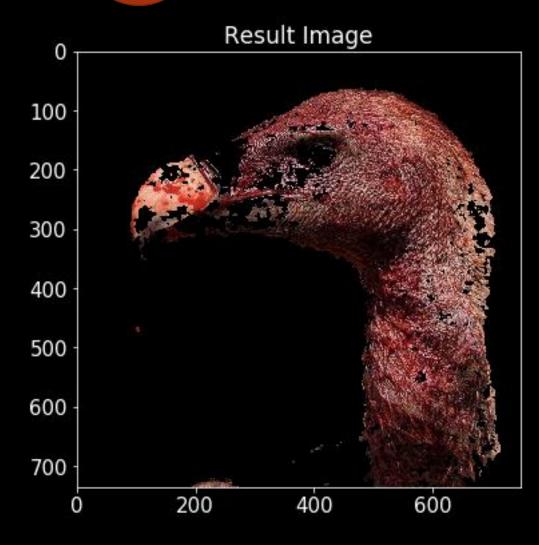




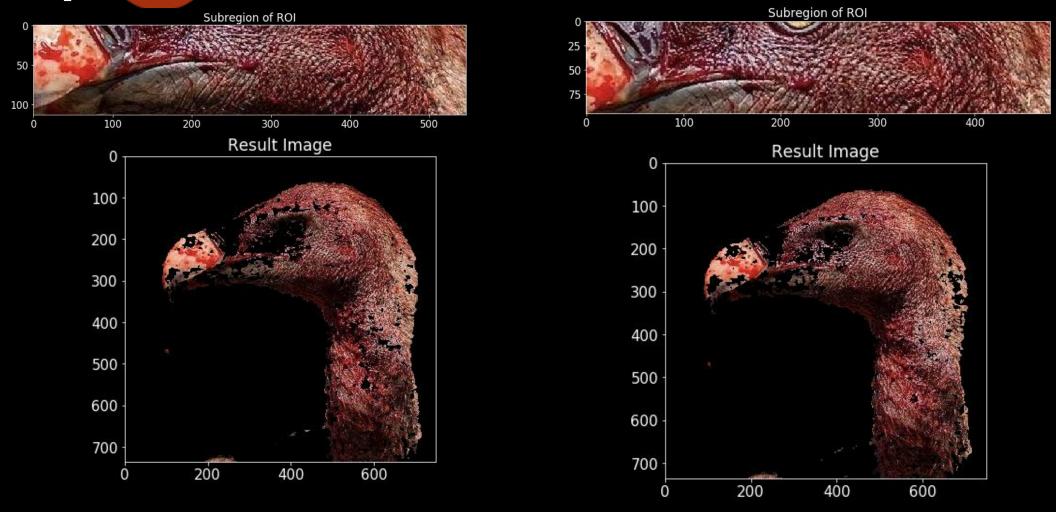
 This was the patch or subregion used in order to segment the Vulture



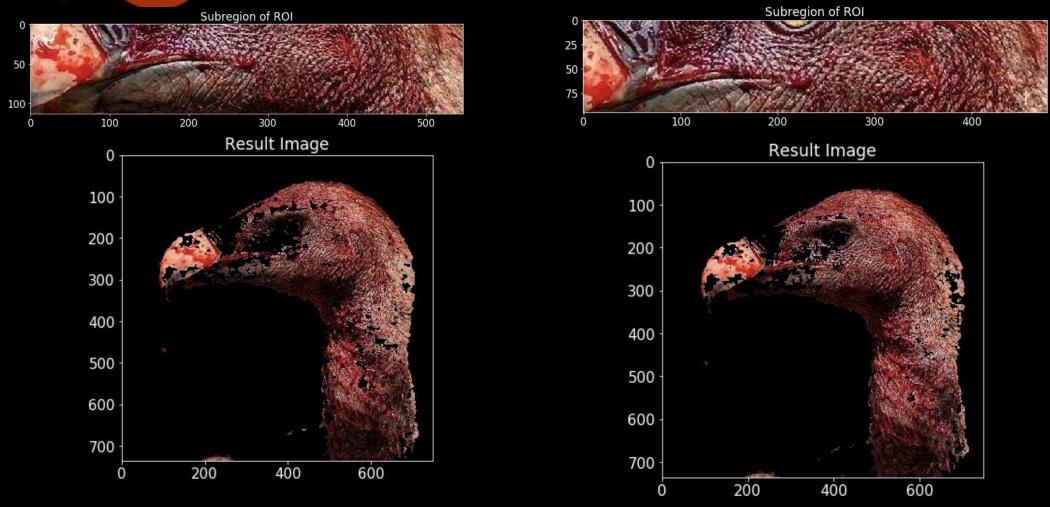




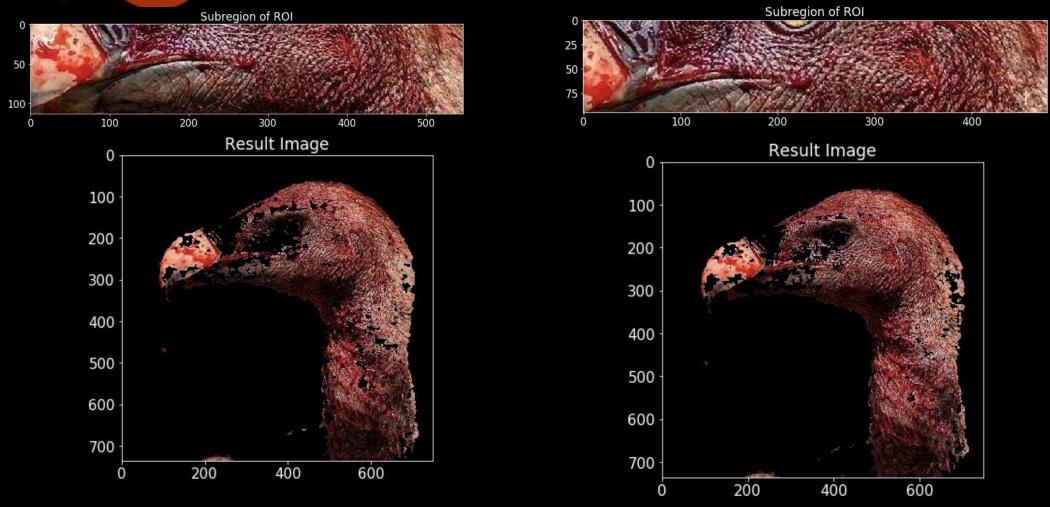
- 'This segmentation method was able to produce the outline of the Vulture
- However, there are missing parts on the Vulture's features that weren't captured. This may be due to the patch used



The difference between these two patches was the vertical shift



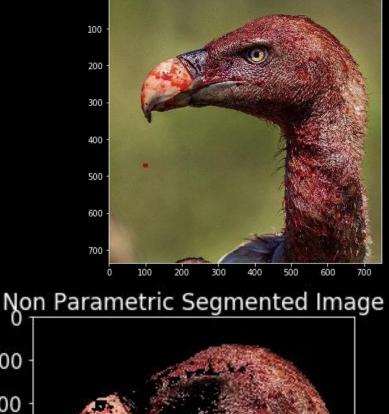
• There's a difference in the resulting image. The Vulture had lesser features on the left since the bottom of the eye was excluded in the patch



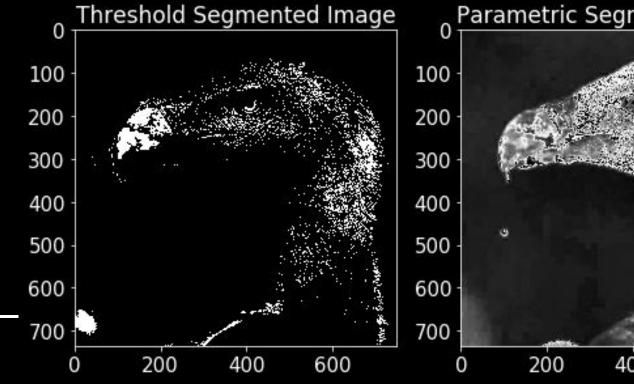
• This shows that the Non-parametric segmentation relies on the patch used- which can be a weakness and a strength. However, in this case, it is its weakness.

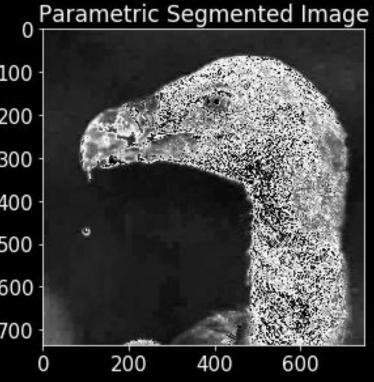
Best segmented image: Parametric

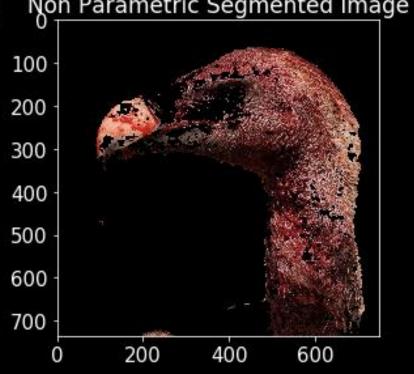
\* This may not be the case for other images. Let us try other colorful images as well.



Original Image

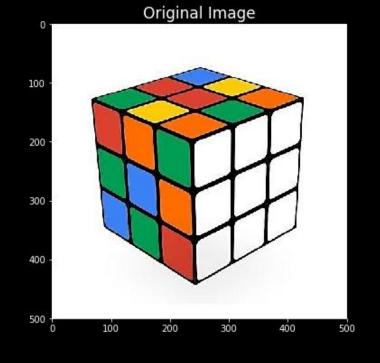


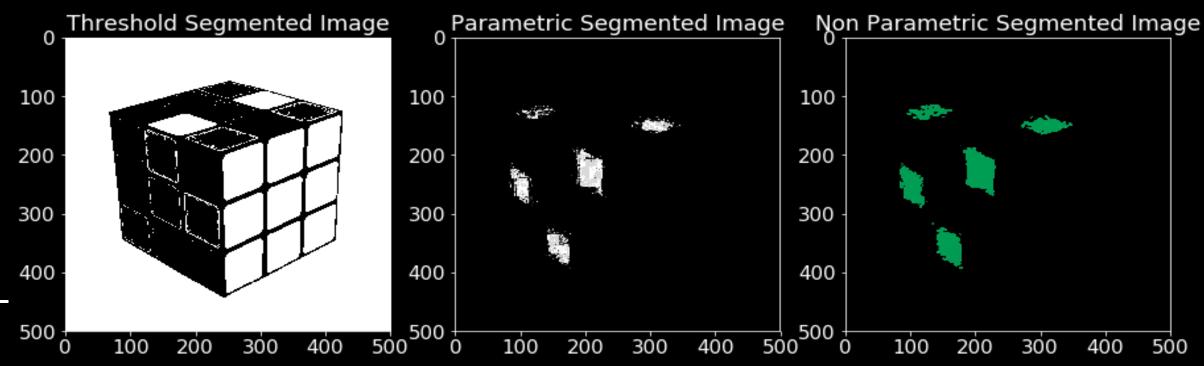






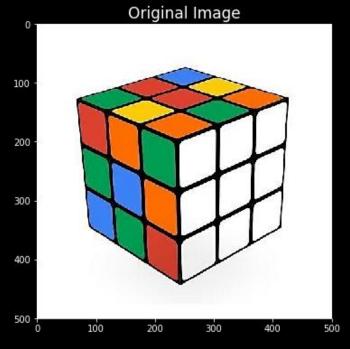
- Green cells:
- Best segmented image: Non parametric
- \* Threshold can't be used for this since it only works for grayscale

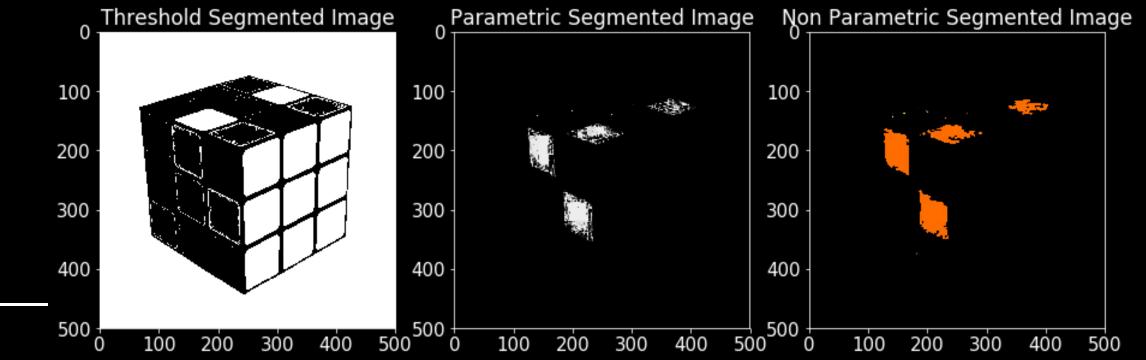






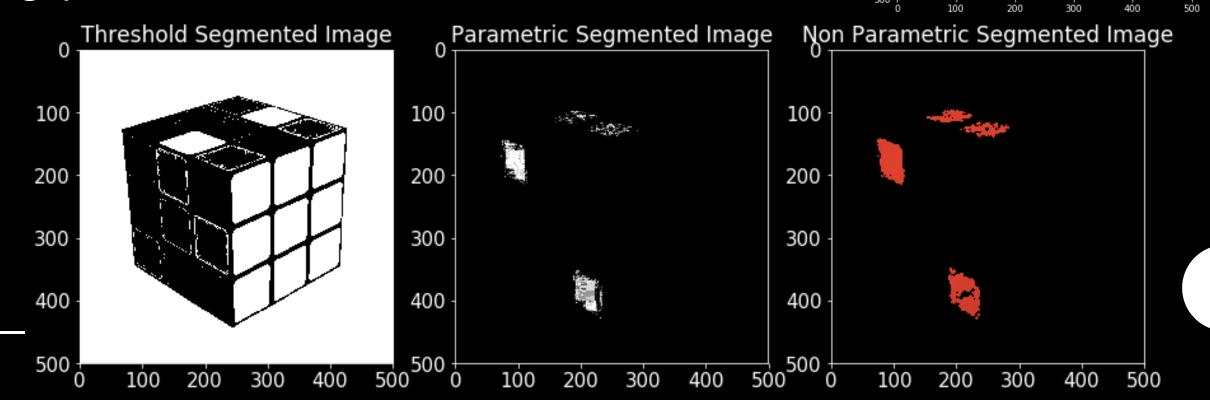
- Orange cells:
- Best segmented image: Non parametric
- \* Threshold can't be used for this since it only works for grayscale







- Red cells:
- Best segmented image: Non parametric
- \* Threshold can't be used for this since it only works for grayscale

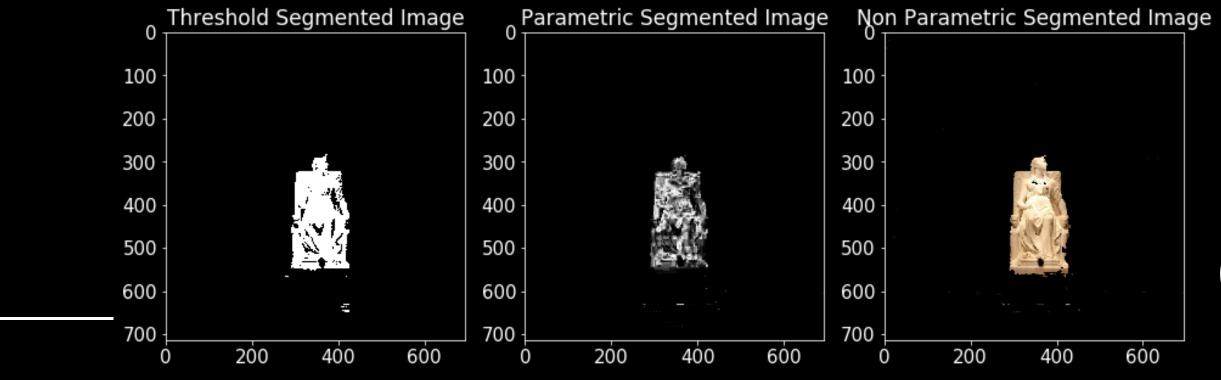


Original Image



Best segmented image: Non - parametric





#### \*

#### Other pictures

- Concluding from other pictures used, the preferred segmentation method differ depending on the original image.
- Preference lies on the colors or pixel values the ROI has.
- Threshold works best for grayscale images (with contrasting backgrounds
- Parametric works best for detailed images whose histogram is similar to a Gaussian distribution
- Non-Parametric works better for segmenting certain colored object because of the extraction of desired patch.

## \* Pointssss

- TC:5
- QP:4
- IN: 1....??

• This was so much fun @! Thank you!