

Image Processing Applications in FinTech



Bhavesh Bhatt



Agenda for the Talk

- What problems are we trying to solve?
- Introduction to OpenCV with Python
- Image processing applications in Flexiloans
- Separating foreground from background using Otsu's Binarization

He is the victim of a Jaan
Pehchaan model ...

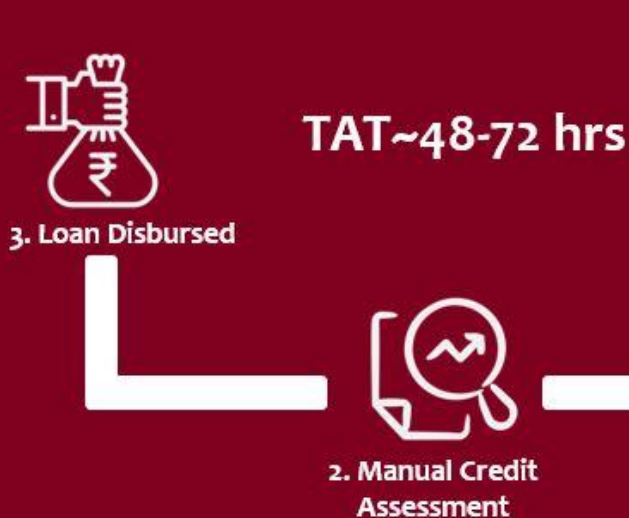
... and we are solving his
problem

- Runs a small business
- Based in Udaipur
- No CIBIL record, but occasionally needs financing
- **'Phones a friend'** when he needs money because it is quick (not cheap)

Realizing Our Dream driven by Data-driven decision making

PRESENT

Manual intervention is needed
for credit assessment



FUTURE

Automated Credit Assessment enabling
“Loans at a click”



Image Basics I

- How do we represent an image?
- Images are stored in forms of a 2-D matrix. For a colour image (RGB) we store 3 similar matrices stacked over each other.
- Why RGB - $R + G + B$ combines to give White

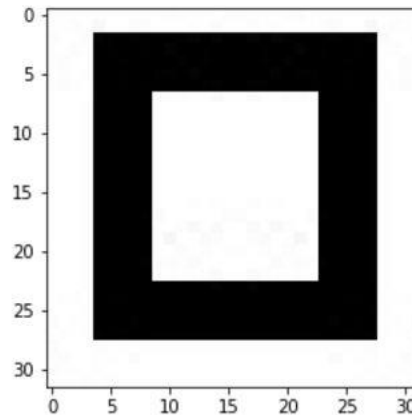
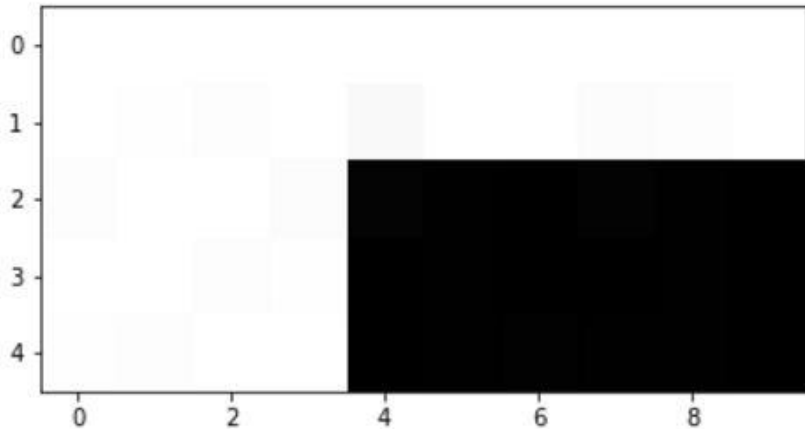


Image Basics II

```
plt.imshow(img[0:5,0:10], cmap='gray')  
plt.show()
```



```
print img[0:5,0:10]
```

```
[[255 255 255 255 255 255 255 255 255 255]  
 [255 254 253 255 249 255 255 252 253 255]  
 [253 255 255 252  4  1  0  3  1  0]  
 [255 255 253 254  0  1  0  0  1  0]  
 [254 253 255 255  0  1  2  1  1  0]]
```

- Considering the top left portion of the image.
- White pixels represented by values close to 255 and black pixels represented by values close to 0.

Image Basics III

- Colour images can be represented by three matrices.
- Each matrix specifies the amount of Red, Green and Blue that makes up the image

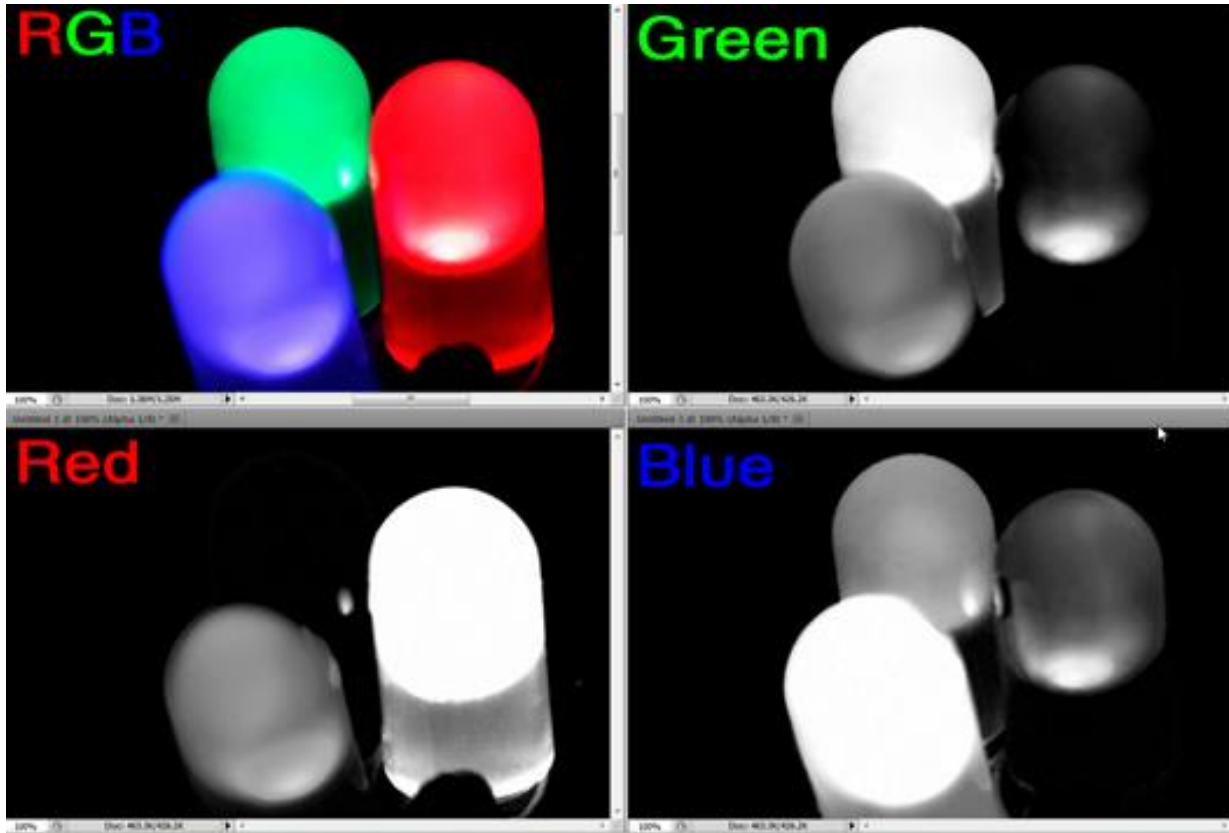


Image Processing using OpenCV I

- OpenCV is a library of programming functions mainly aimed at real-time computer vision.
- Lets learn how we can use OpenCV for Image processing.

Loading an image

```
image = cv2.imread("pydata_mumbai.jpg")
```

Saving the image from OpenCV

```
cv2.imwrite("newimage.jpg", image)
```

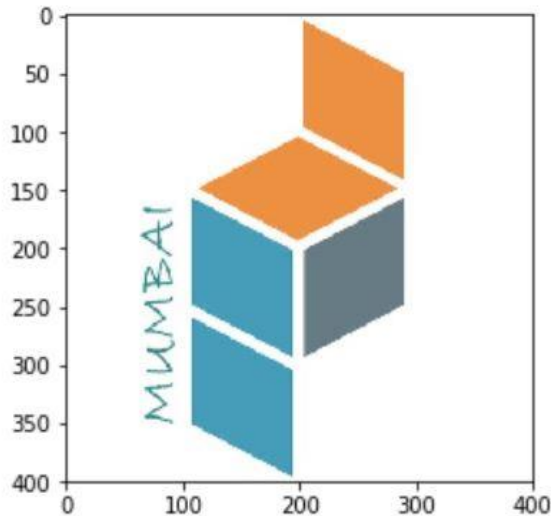
True



Image Processing using OpenCV II

Displaying the inputted image

```
show_RGBImage(image)
```



```
def show_RGBImage(image):  
    '''  
    OpenCV represents RGB images as  
    multi-dimensional NumPy array  
    but in reverse order.  
    This means that images are actually  
    represented in BGR order rather than RGB.  
    '''  
    plt.imshow(cv2.cvtColor(image, cv2.COLOR_BGR2RGB))  
    plt.axis("on")  
    return plt.show()
```

```
print "width: %d pixels" % (image.shape[1])  
print "height: %d pixels" % (image.shape[0])  
print "channels: %d" % (image.shape[2])
```

```
width: 400 pixels  
height: 400 pixels  
channels: 3
```

```
type(image)
```

```
numpy.ndarray
```



Image Processing using OpenCV III

Accessing & Manipulating pixels

```
(b, g, r) = image[0, 0]  
print "Pixel at (0, 0) - Red: %d, Green: %d, Blue: %d" % (r, g, b)
```

Pixel at (0, 0) - Red: 255, Green: 255, Blue: 255

Creating a black image

```
blank_image = np.zeros((400,400,3))  
type(blank_image[0,0,0])
```

numpy.float64

```
black_image = np.zeros((400,400,3), np.uint8)  
show_RGBimage(black_image)
```

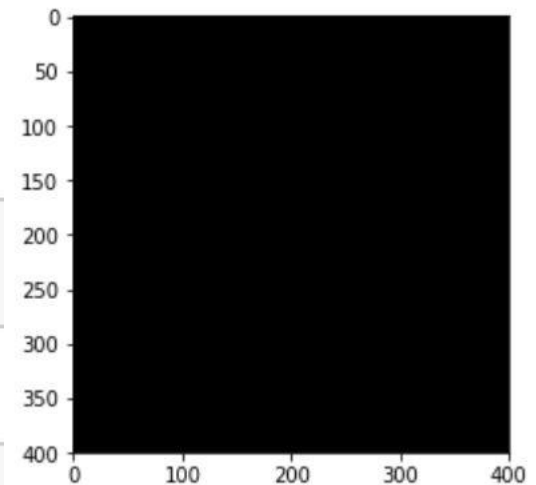


Image Processing using OpenCV IV

Creating a white image

```
white_image = 255*np.ones((400,400,3), np.uint8)  
show_RGBimage(white_image)
```

```
image_ref1 = image
```

```
image_ref1[0:100, 0:100] = (0, 0, 0)  
show_RGBimage(image_ref1)
```

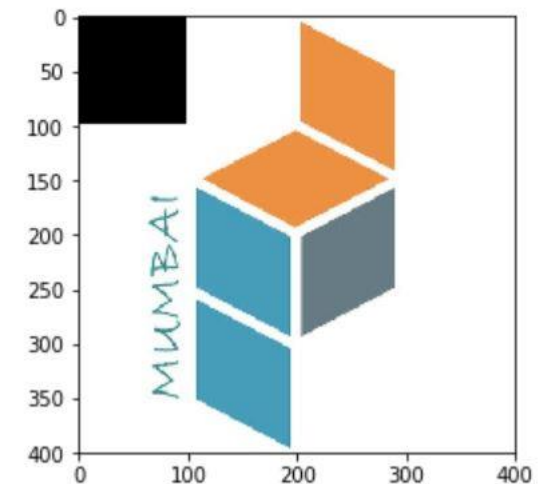
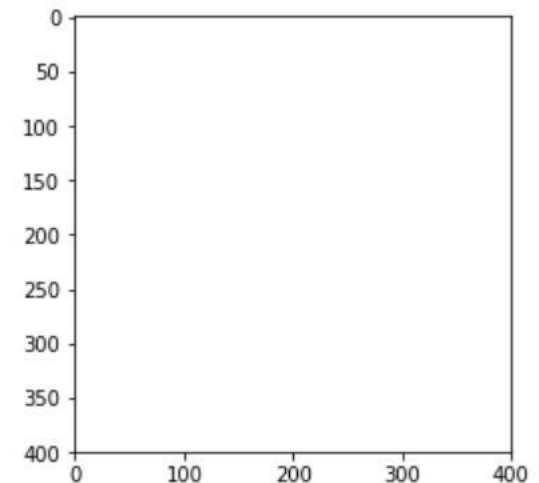


Image Processing using OpenCV V

■ Drawing shapes using OpenCV

```
canvas = 255*np.ones((300, 300, 3), dtype = "uint8")
red = (0, 0, 255)
# OpenCV representation of Red is based on BGR
# and not RGB representation

cv2.rectangle(canvas, (50, 50), (150,250), red, 5);

# Draw a rectangle starting from point (50,50)
# till (150,250) and the
# points are represented as (x,y)
```

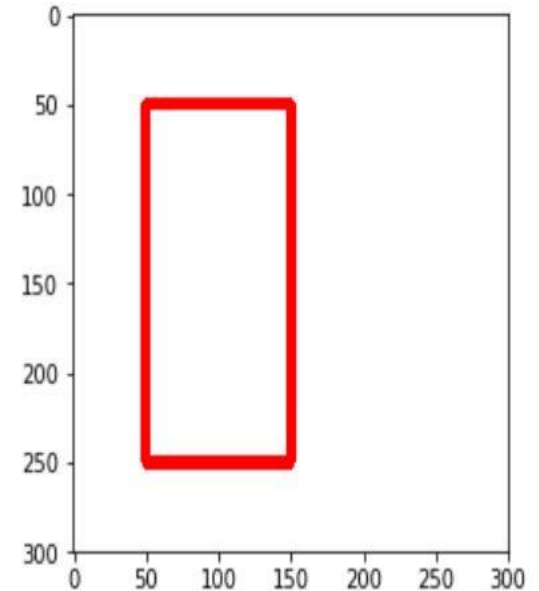


Image Processing using OpenCV VI

- **Histogram** - Displays the frequency distribution of the intensity values of an image

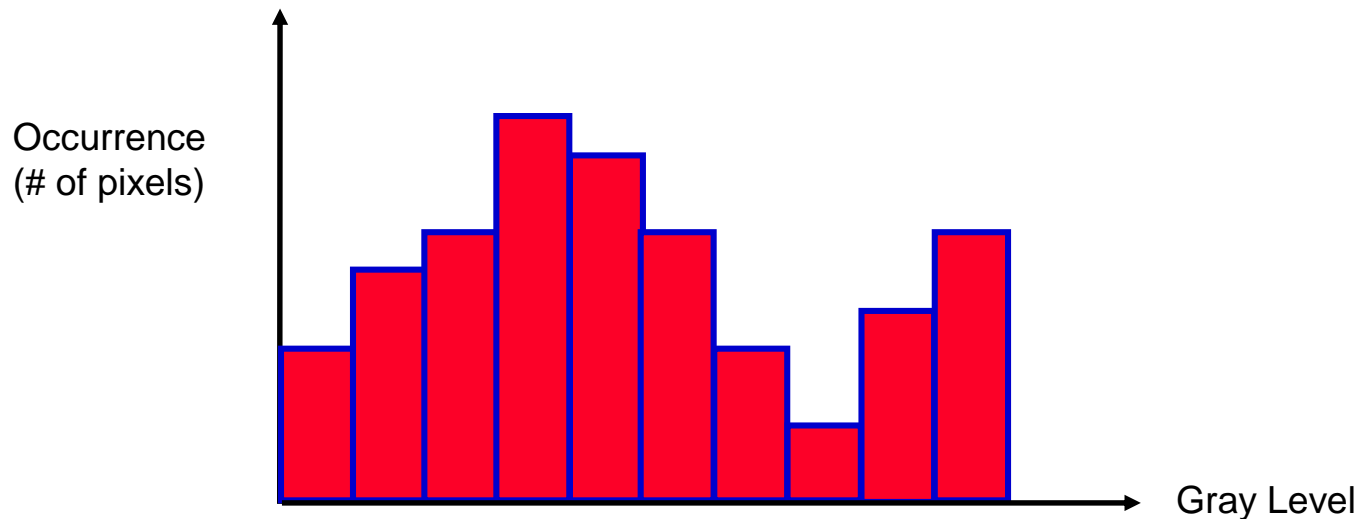


Image Processing using OpenCV VII

■ Histogram of Black Image

```
image_black_gray = cv2.cvtColor(black_image, cv2.COLOR_BGR2GRAY)  
plot_histogram(image_black_gray)
```

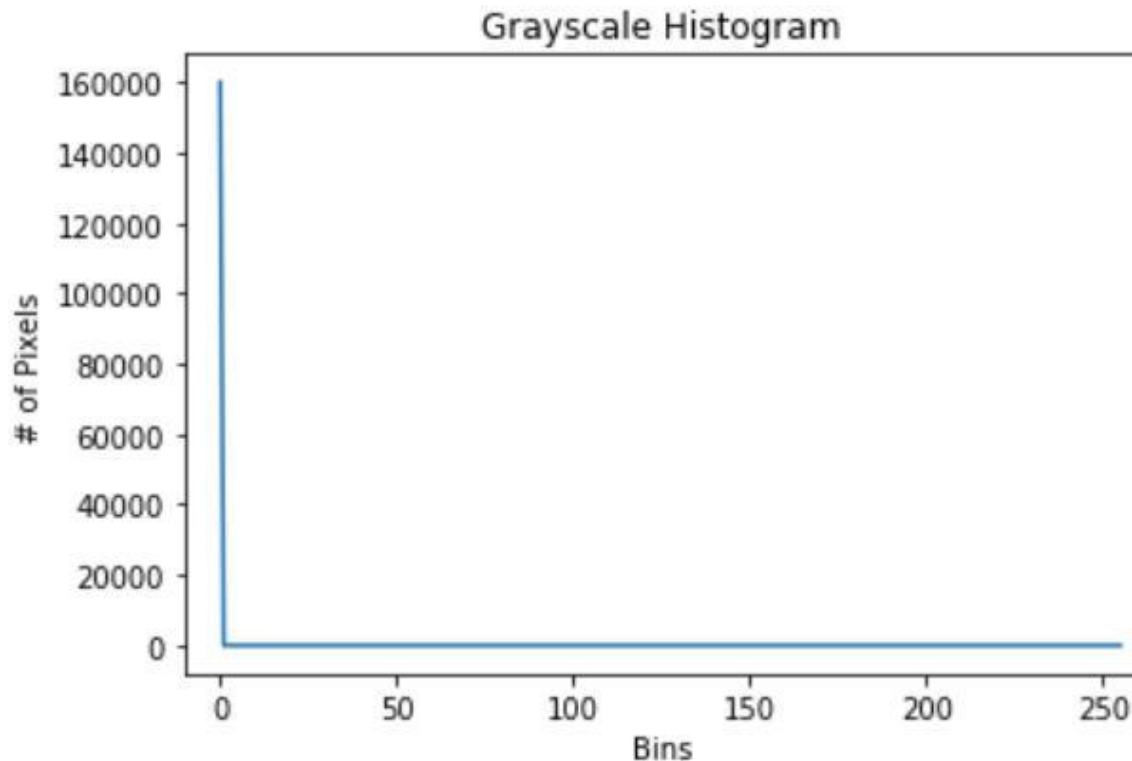


Image Processing using OpenCV VIII

■ Histogram of White Image

```
image_white_gray = cv2.cvtColor(white_image, cv2.COLOR_BGR2GRAY)  
plot_histogram(image_white_gray)
```

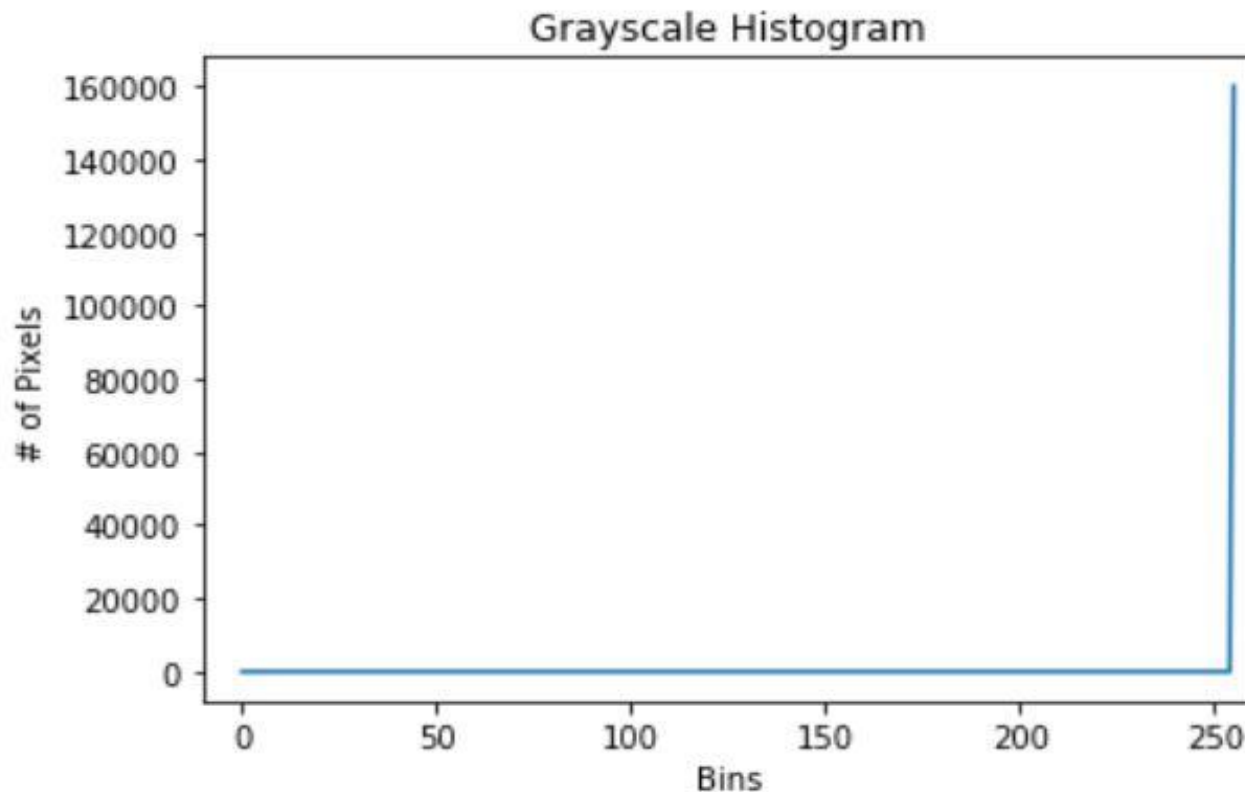


Image Processing at Flexiloans

- Business document classification
- Fraud document detection
- Automated bank cheque processing



Business Document Classification I

Business Challenge :

- An average human takes approximately 10-15 minutes to classify and digitize business documents.
- In a real-time lending environment, automated document classification systems should be deployed to reduce the human effort from minutes to microseconds without compromising on the accuracy

Solutions :

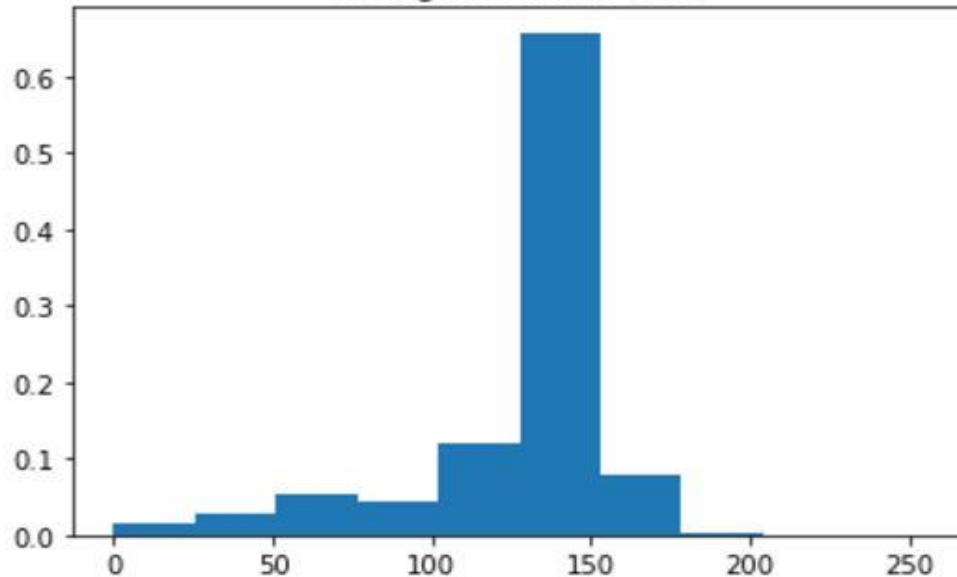
- Histogram intersection techniques
- OCR
- Pretrained models (Inception, VGG16)
- CNN (convolutional neural network)



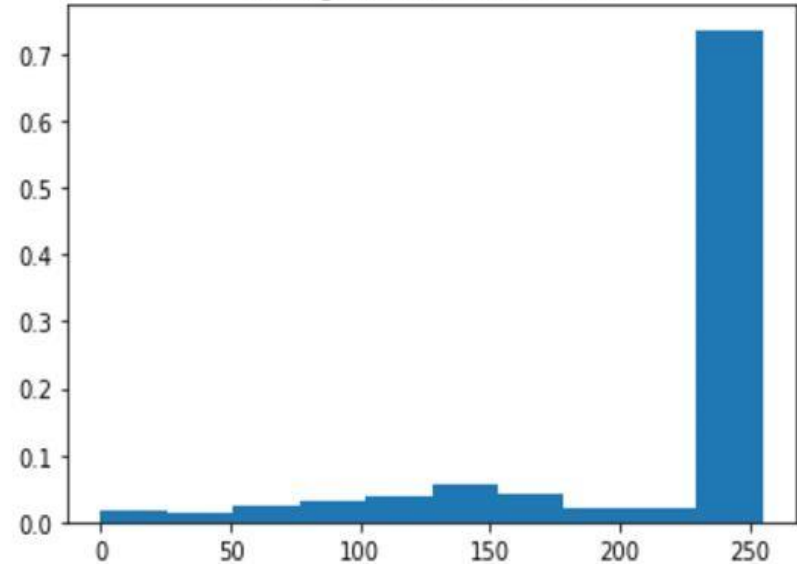
Business Document Classification II

■ Histogram intersection techniques

Histogram of PAN Card



Histogram of Aadhaar Card



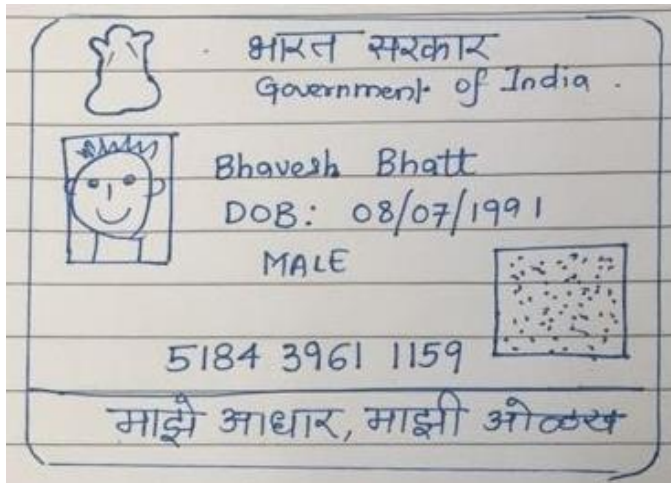
आधार - आदमी का अधिकार

Business Document Classification III

■ Optical Character Recognition

```
In [37]: img_loc = '/home/user/Desktop/Test_Aadhar/WhatsApp Image 2018-01-19 at 12.52.44 PM.jpeg'
```

```
In [38]: predict_aadhar_or_not_aadhar(img_loc)
```

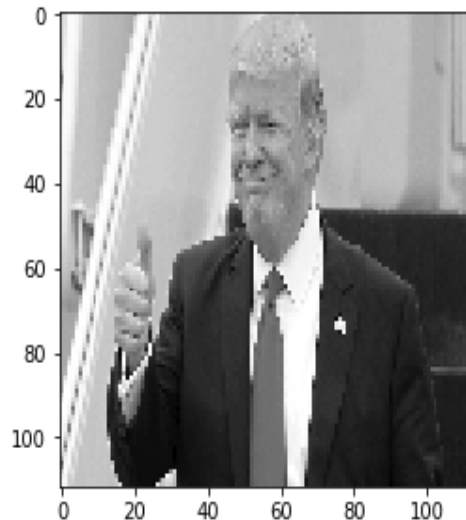


INFO:tensorflow:Restoring parameters from alexnet_pan_aadhar_with_background.cpkt
The given image is not aadhaar card

Business Document Classification IV

```
In [33]: img_loc = '/home/user/Desktop/Test_Aadhar/Donald_Trump.jpg'
```

```
In [34]: predict_aadhar_or_not_aadhar(img_loc)
```



```
INFO:tensorflow:Restoring parameters from alexnet_pan_aadhar_with_background.cpkt  
The given image is not aadhaar card
```

Business Document Classification V

```
In [23]: img_loc = '/home/user/Desktop/Test_Aadhar/Aadhar_Card.jpg'
```

```
In [24]: predict_aadhar_or_not_aadhar(img_loc)
```



INFO:tensorflow:Restoring parameters from alexnet_pan_aadhar_with_background.cpkt
The given image is aadhaar card

Binarizing an Image I

- Why do we binarize an image?
Converting to binary is often used in order to find a Region Of Interest - a portion of the image that is of interest for further processing.
- Global Thresholding - If pixel value is greater than a threshold value, it is assigned '1' - white else it is assigned '0' - black.
- The OpenCV function used for this task is `cv2.threshold`

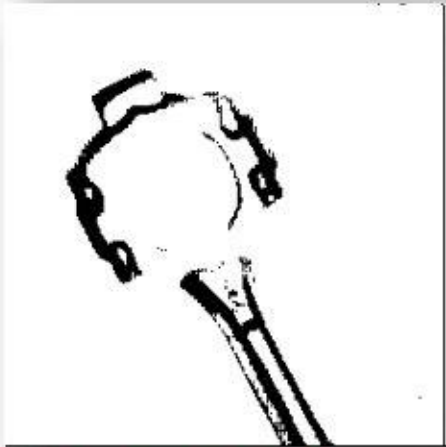


Binarizing an Image II

Original Image



Binary Image



Threshold too low



Threshold too high

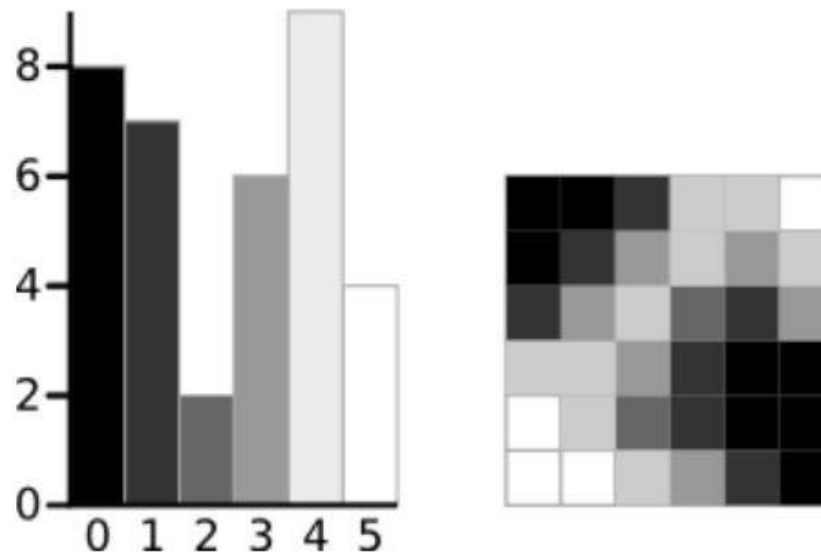
Otsu Thresholding I

- Otsu's thresholding method involves iterating through all the possible threshold values and calculating a measure of spread for the pixel levels.
- The 2 clusters should be as tight as possible for which the **within class variance** should be minimum yet the distance separating them should be maximum i.e. inter-class or **between class variance**.
- In simple terms, I want to find a threshold value which best divides my histogram into 2 halves.
- Reference
<http://www.labbookpages.co.uk/software/imgProc/otsuThreshold.html>



Otsu Thresholding II

- Consider a 6x6 image which has 6 grayscale levels.
- 0 represents black and 5 represents white.

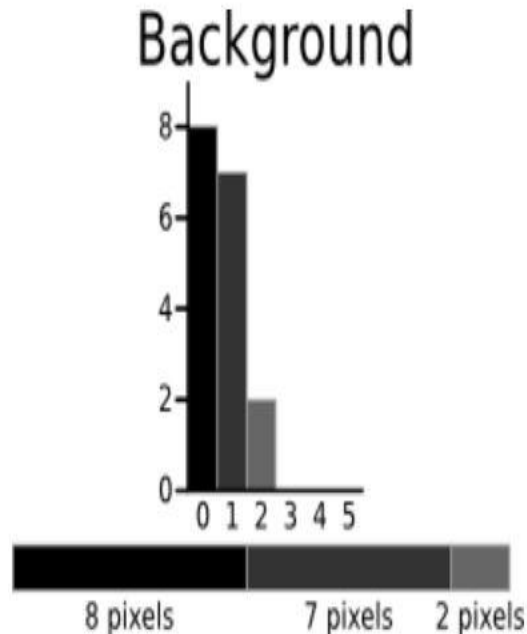


- **Reference**

<http://www.labbookpages.co.uk/software/imgProc/otsuThreshold.html>

Otsu Thresholding III

- Assuming the threshold is 3



$$\text{Weight } W_b = \frac{8 + 7 + 2}{36} = 0.4722$$

$$\text{Mean } \mu_b = \frac{(0 \times 8) + (1 \times 7) + (2 \times 2)}{17} = 0.6471$$

$$\begin{aligned} \text{Variance } \sigma_b^2 &= \frac{((0 - 0.6471)^2 \times 8) + ((1 - 0.6471)^2 \times 7) + ((2 - 0.6471)^2 \times 2)}{17} \\ &= \frac{(0.4187 \times 8) + (0.1246 \times 7) + (1.8304 \times 2)}{17} \\ &= 0.4637 \end{aligned}$$

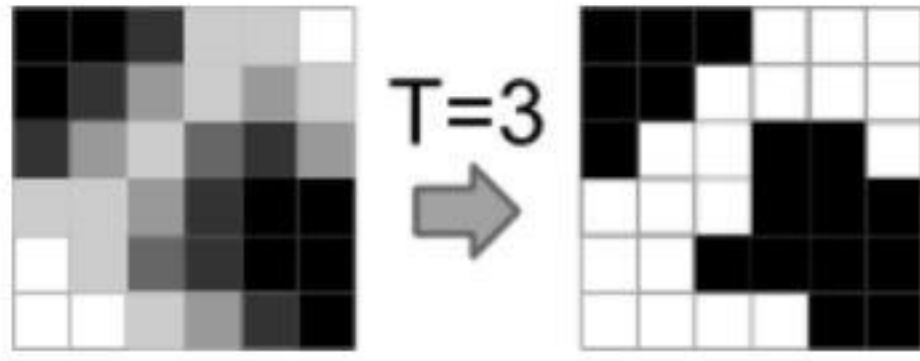
$$\begin{aligned} \text{Within Class Variance } \sigma_W^2 &= W_b \sigma_b^2 + W_f \sigma_f^2 = 0.4722 \times 0.4637 + 0.5278 \times 0.5152 \\ &= 0.4909 \end{aligned}$$

- Reference

<http://www.labbookpages.co.uk/software/imgProc/otsuThreshold.html>

Otsu Thresholding IV

Within Class Variance	$\sigma_W^2 = 3.1196$	$\sigma_W^2 = 1.5268$	$\sigma_W^2 = 0.5561$	$\sigma_W^2 = 0.4909$	$\sigma_W^2 = 0.9779$	$\sigma_W^2 = 2.2491$
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- Otsu's thresholding is a simple method which works really well when the image histogram has 2 distinct peaks



- Reference

<http://www.labbookpages.co.uk/software/imgProc/otsuThreshold.html>

Thank You

- Link to the code and presentation -
https://github.com/bhattbhavesh91/PyDataMumbai-21April_2018