

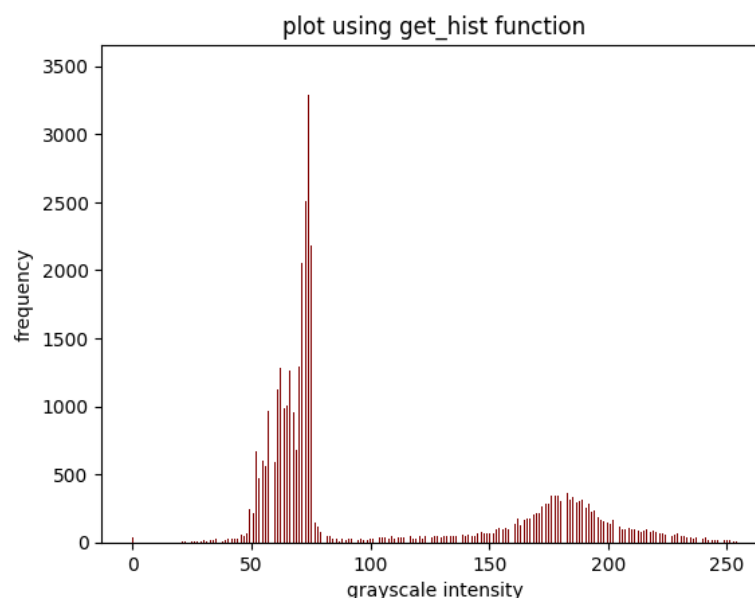
# Report- Assignment 1

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1. **Histogram Computation:** Compute the histogram of the image coins.png, by finding the frequency of pixels for each intensity level  $\{0, 1, \dots, 255\}$ . Show the histogram by plotting frequencies w.r.t. intensity levels. Comment on what you observe. Also, find the average intensity of the image using this histogram. Verify the result with the actual average intensity

## RESULTS:

```
time taken= 0.05159139633178711
Average calculated from histogram = 103.30500158906722
Average calculated over all pixel = 103.30500158906722
```



## INFERENCES:

- The majority of the intensities lie between 50 and 80 with peak near 75. Therefore it is the background colour of the image
- The background is black and the coins are of intensity around 175.

2. **Otsu's Binarization:** In the class, we showed that  $\sigma^2 w(t) + \sigma^2 b(t) = \sigma^2 T$ , where  $t$  is the threshold for binarization. Binarize the image coins.png by finding the optimal threshold  $t$  by:

(a) Minimizing the within class variance  $\sigma^2 w(t)$  over  $t$ .

(b) Maximizing the between class variance  $\sigma^2 b(t)$  over  $t$ .

Verify that both methods are equivalent. Compare the time taken by each of the approaches.

## RESULTS:

Verifying that maxima of  $\sigma^2_b(t)$  and minima of  $\sigma^2_w(t)$  occur at same value of  $t$ .

time taken for between variance calculation = 0.3152885437011719

time taken for within variance calculation = 0.23804402351379395

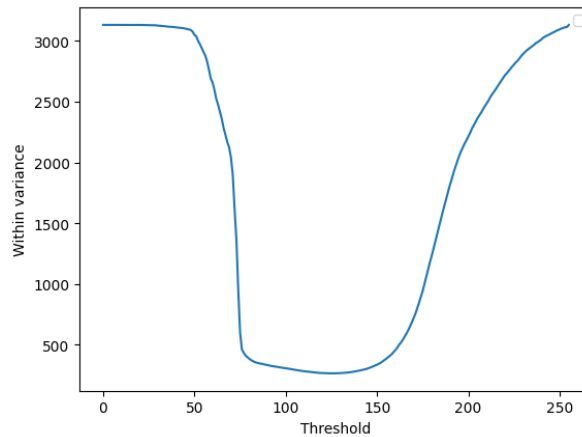
variance between class = 2865.7017638569228

theshold for max between class variance = 125

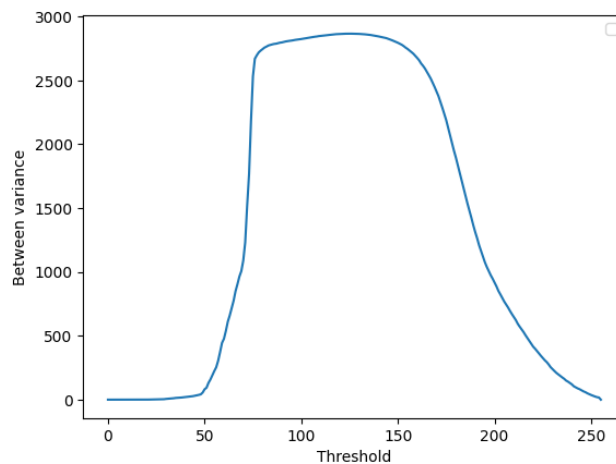
variance within class = 265.1024571148032

theshold for min within class variance = 125

Plot of  $\sigma^2_w(t)$  :



Plot of  $\sigma^2_b(t)$  :

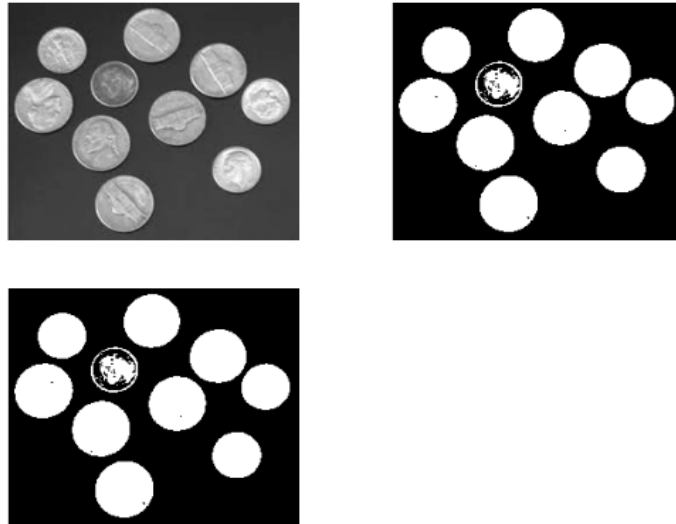


## INFERENCES:

- We can see the thresholds found by minimizing within class variance and maximizing between class variance are identical and thus it is evident that they must be identical procedures.

$$\sigma_w^2(t) + \sigma_b^2(t) = \sigma_T^2$$

- It is intuitive from above equation, we know the total variance is independent of the threshold, if we maximize one of the functions, the other in that case would be the minimum. Thus it is mathematically consistent
- The time taken by both the processes are identical.



3. **Depth based Extraction:** The image IIScTextDepth.png is an inverse depth map of IIScText.png. A depth map indicates the depth of an object from the camera for each pixel. Particularly, an inverse depth map has a higher value when the object is nearer to the camera and a lower value when it is farther apart. Binarize the inverse depth map IIScTextDepth.png and use that information to extract the text in IIScText.png and display it over the background image IIScMainBuilding.png. The expected image is shown below.

**REPORT:**



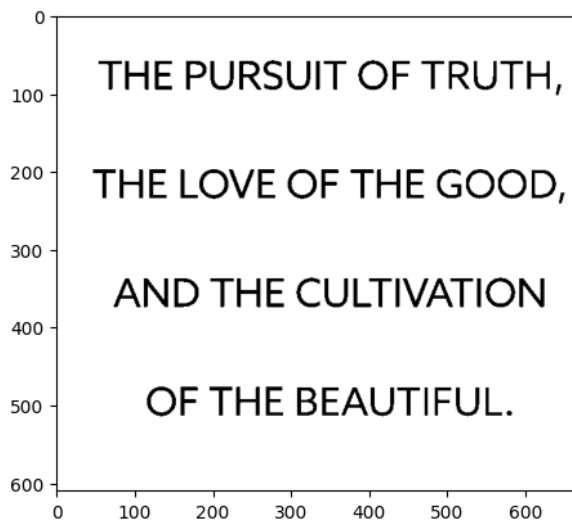
## INFERENCES:

- In this process first the IIScTextDepth.png is Binarized using otsu's binarization algorithm. The binarized image then is used to extract the information from the IIScText.png and paste on the IIScMainBuilding.png
- The question helps us to understand intuitively what a depth map is used for. We can use depth map to basically extract information of the background and foreground (in this case) using elementary methods like binarization. Depth map is a method to decipher depth in a 2D image.

## 4. Connected Components: Binarize the image quote.png and count the total number of characters excluding punctuations using connected component analysis.

## REPORT:

Binarized image:



Region matrix:



Details of sets of regions:

```
length of list = 67
height of all regions = [36 36 36 36 36 36 36 36 35 36 35 35 36 35 36 37 37 13 36 36 36 36 36 36
36 36 36 36 36 37 37 37 37 35 13 36 36 36 36 36 36 36 36 36 36 36 37
36 37 36 36 36 36 36 36 35 36 35 35 35 36 35 37 35 5 35]
Average height of pixels = 34.80597014925373
```

Result:

```
no. of letters in Image = 64
time taken= 12.13597583770752
```

#### INFERENCES:

- This question helped to get an idea of how regions can be made in an image. This knowledge opened up new doors to what and how an image can be manipulated.
- In this approach height to the regions was used to differentiate between letter and comma but, more sophisticated method can be devised for an image where varieties of symbols are present.

5. **(Optional Bonus Question) - MSER: Maximally Stable Extremal Regions (MSER)** correspond to regions of connected components which, when thresholded around a certain threshold, are stable in terms of the size of the component. Determine the number of characters in Characters.png based on MSER.

Think about why finding connected components over an Otsu binarized image will not work well in this scenario.

#### REPORT:

Characteristic features of all regions formed with their info:

```

info of image- format:-[[jth element of marix, ith element of marix, height, width]]
[[1029 350 187 157]
 [ 209 350 187 56]
 [ 613 350 187 203]
 [1250 529 187 204]
 [1218 308 189 158]
 [1244 338 187 204]
 [1240 307 187 204]
 [1149 322 189 26]
 [1215 328 189 159]
 [1149 346 189 38]
 [ 259 350 189 159]
 [1052 349 187 204]
 [ 202 350 189 44]
 [1220 421 187 204]
 [1132 340 189 57]
 [ 282 349 187 204]
 [1119 308 189 57]
 [1188 333 187 204]
 [1097 324 189 57]
 [1182 400 188 204]
 [1005 319 189 57]
 [1000 308 189 58]
 [ 157 308 189 58]
 [1143 309 188 204]
 [1146 345 188 204]
 [ 241 325 188 204]
 [1052 308 189 58]
 [ 282 291 71 204]
 [ 230 313 198 204]
 [1027 309 188 204]
 [ 230 269 111 204]
 [ 230 289 151 204]
 [ 230 282 136 204]
 [ 708 331 189 212]
 [ 737 346 194 211]
 [ 385 350 189 211]
 [ 731 375 195 105]
 [ 711 385 195 105]
 [ 723 404 195 104]
 [ 727 417 195 102]
 [ 773 350 195 101]
 [ 699 432 188 212]
 [ 730 440 188 212]
 [ 702 350 194 119]]

```

Information about their stability of regions and more:

stability array

```

[111 99 19 0 1 1 0 0 0 0 0 0 0 1 0 4 1 0
 0 0 0 0 2 0 0 0 0 0 1 1 0 0 0 1 0 69
 0 1 1 0 5 0 0 42]

```

increments in binarization threshold value = 1

Number of stable regions = 5

time taken = 854.0407636165619

## INFERENCES:

- This question helps us to utilize the region forming algorithm and use it to find regions which might be not possible to find using otsu's binarization
- Otsu's binarization will not be able to binarize the image. Because the letters have both in black and white pixels, they both cannot showed using a single threshold value.
- In this question, (a) and (b) steps are done

```

# (a) Sweep over all thresholds.
for t in range (0, 255, inc):
    # (b) For each threshold, determine connected components in the image.
    img_m = np.round(io.imread(img))
    img_info = find_region_MSER(img_m, t)

```

```
def find_region_MSER(img, th):
    bin_img = binarize_img_MSER(img, th)      #bin_img[0]->binarized image
                                              #bin_img[1]->majority pixel black or white
    img = bin_img[0]
    blk = bin_img[1]
    x = nos_MSER(img, blk)
    return x
```

In next part regions are formed,

```
# Making region matrix r[], and taking account of connected regions in r_lst = [{},{},...]
mn = num_img.shape
for i in range (int(0.3*mn[0]), int(0.7*mn[0]), 1):
    for j in range (int(0.1*mn[1]), int(0.9*mn[1]), 1):
        if (num_img[i][j]== colr and i>0 and j>0):
            if (num_img[i-1][j]== colr):
                r[i][j] = r[i-1][j]
            if (num_img[i][j-1]==colr):
                if(r[i-1][j]!=r[i][j-1]):
                    # print("log: found 2 existing region")
                    update_region_MSER(0, r_lst, i, j, r[i][j-1], r[i-1][j], True)
                    r[i][j-1] = r[i-1][j]
            if (num_img[i][j-1]==colr and num_img[i-1][j]!=colr):
                r[i][j] = r[i][j-1]
            if (num_img[i][j-1]!=colr and num_img[i-1][j]!=colr):
                update_region_MSER(count, r_lst, i, j, 0, 0, False)
            r[i][j] = count
            count += 1
```

In this set of codes the geometry of the regions are determined,

```
for i in range (int(0.3*mn[0]), int(0.7*mn[0]), 1):
    for j in range (int(0.1*mn[1]), int(0.9*mn[1]), 1):
        ptr=0
        for x in r_lst:
            if r[i][j] in x:
                pxls[ptr] +=1
                if (wd_f[ptr]>j): wd_f[ptr]=j
                pxl_wd[ptr] = j - wd_f[ptr]

                if (ht_f[ptr]>i): ht_f[ptr]=i
                pxl_ht[ptr] = i -ht_f[ptr]
            if pxl_ht[ptr]>200 or pxl_wd[ptr]>250:
                r_lst = np.delete(r_lst, ptr, axis=0)
                pxls = np.delete(pxls, ptr)
                pxl_ht = np.delete(pxl_ht, ptr)
                pxl_wd = np.delete(pxl_wd, ptr)
                continue
        ptr+=1
```

In this step the region geometries are compiled in array and others discarded.

The array is then returned

```
for x in r_info:
    if pxls[ptr]<10000 or pxl_ht[ptr]<50 or pxl_ht[ptr]>200 or pxl_wd[ptr]<25 or pxl_wd[ptr]>250:
        r_info = np.delete(r_info, ptr, axis=0)
        pxls = np.delete(pxls, ptr)
        pxl_ht = np.delete(pxl_ht, ptr)
        pxl_wd = np.delete(pxl_wd, ptr)
        continue
    codx = wd_f[ptr]+int(pxl_wd[ptr]/2)
    cody = ht_f[ptr]+int(pxl_ht[ptr]/2)
    r_info[ptr][0] = codx
    r_info[ptr][1] = cody
    r_info[ptr][2] = pxl_ht[ptr]
    r_info[ptr][3] = pxl_wd[ptr]
    ptr+=1

return r_info
```

The returned information is used to “measure stability” of the regions and update count[] of all the region info in img\_info[] keeping record of all previous regions found.

```
# (b) For each threshold, determine connected components in the image.
img_m = np.round(io.imread(img))
img_info = find_region_MSER(img_m, t)

# (c) A connected component is termed an MSER if the size of the component does not change
#      much (within "tol= 10") for a small perturbation "inc= (5-10)" in the choice of the threshold.
#      Determination of the stable threshold for each connected component.
if(isinstance(img_info, np.ndarray)):
    for x in img_info:
        prsnt = False
        ptr = 0
        for y in info:
            if(np.allclose(x, y, atol=tol)):
                count[ptr]+=inc
                prsnt = True
                break
            if(prsnt): break
        ptr+=1
    if(prsnt==False):
        info = np.append(info, [x], axis = 0)
        count = np.append(count, [0])

return info, count, inc
```

- We have scanned the image 3 times for binarization, region formation and geometry calculation. We only scan the matrix column from 0.3 to 0.7 times column length and row from 0.1 to 0.9 times row length, these lengths were chosen by using observation. This was done to improve runtime of the program.
- The limit for height, width and no. of pixels to be cutoff and removed were also selected by observation.
- This question helped to broaden the knowledge of how an image matrix can be manipulated and information be extracted like height, width and center points of region which defines the characteristics of that region.