

# Image Segmentation via Thresholding

```
In [1]: import numpy as np
import math
import cv2
import matplotlib.pyplot as plt
```

## Algorithm Basic Global Thresholding

```
In [4]: from IPython.display import Image
Image("midean.png")
```

Out[4]:

1. **Input:**
  - Grayscale image (a 2D array of pixel intensities).
  - Initial guess for the threshold value,  $T_0$ .
  - Tolerance value to determine convergence.
2. **Initialization:**
  - Set  $T$  to the initial guess value  $T_0$ .
3. **Iterative Thresholding:**
  - **Repeat until convergence:**
    - a. **Partition the Image:**
      - Divide the image into two sets based on the current threshold  $T$ :
        - Foreground (F): Pixels with intensities greater than  $T$ .
        - Background (B): Pixels with intensities less than or equal to  $T$ .
    - b. **Calculate Mean Intensities:**
      - Compute the mean intensities of pixels in F and B:
        - Calculate the average intensity of pixels in F (denoted as  $\mu_F$ ).
        - Calculate the average intensity of pixels in B (denoted as  $\mu_B$ ).
    - c. **Update Threshold:**
      - Compute a new threshold value  $T_{\text{new}}$  using:
$$T_{\text{new}} = \frac{\mu_F + \mu_B}{2}$$
    - d. **Check Convergence:**
      - Calculate the absolute difference between  $T_{\text{new}}$  and  $T$ :
        - If the absolute difference  $|T_{\text{new}} - T|$  is less than the tolerance value, stop iterating.
    - e. **Update Threshold:**
      - Update  $T$  to  $T_{\text{new}}$ .
4. **Output:**
  - Use the final threshold  $T$  to binarize the image:
    - Set pixel intensity to 0 (representing background) if the pixel intensity is less than or equal to  $T$ .
    - Set pixel intensity to 255 (representing foreground) if the pixel intensity is greater than  $T$ .

## Apply Basic Global Thresholding

```
In [2]: def basic_global_thresholding(img, deltheshold):
        initial_mean = np.mean(img)
        while(True):
            pre_mean = initial_mean
            background = []
            foreground = []
            for i in range(img.shape[0]):
                for j in range(img.shape[1]):
                    if(img[i][j] > initial_mean):
                        foreground.append(img[i][j])
                    else:
                        background.append(img[i][j])
            mean_back = np.mean(background)
            mean_fore = np.mean(foreground)
            initial_mean = (mean_back + mean_fore)/2
            if(abs(initial_mean - pre_mean) <= deltheshold):
                break
        return math.ceil(initial_mean)
```

## Algorithm Otsu Thresholding

```
In [6]: from IPython.display import Image
        Image("otsu.png")
```

Out[6]:

1. **Compute Histogram:**
  - Calculate the histogram of the input grayscale image. The histogram represents the frequency of each pixel intensity level.
2. **Normalize Histogram:**
  - Normalize the histogram so that each bin value represents the probability density function of the pixel intensities.
3. **Compute Cumulative Sum:**
  - Compute the cumulative sum  $\text{cumsum}(i)$  of the normalized histogram up to intensity level  $i$ .
4. **Compute Cumulative Mean:**
  - Compute the cumulative mean  $\text{cum\_mean}(i)$  of the pixel intensities up to intensity level  $i$ :  

$$\text{cum\_mean}(i) = \sum_{k=0}^i k \times \text{hist}(k)$$
5. **Global Mean:**
  - Calculate the global mean intensity  $\mu$  of the entire image:  

$$\mu = \sum_{i=0}^{L-1} i \times \text{hist}(i)$$
where  $L$  is the number of intensity levels (typically 256 for an 8-bit grayscale image).
6. **Compute Between-Class Variance:**
  - For each possible threshold  $t$  (from 0 to  $L - 1$ ):
    - a. Compute the weight  $w_0(t)$  of the background (pixels with intensity  $< t$ ):  

$$w_0(t) = \text{cumsum}(t)$$
    - b. Compute the weight  $w_1(t)$  of the foreground (pixels with intensity  $\geq t$ ):  

$$w_1(t) = 1 - w_0(t)$$
    - c. Compute the mean intensity  $\mu_0(t)$  of the background:  

$$\mu_0(t) = \frac{\text{cum\_mean}(t)}{w_0(t)}$$
    - d. Compute the mean intensity  $\mu_1(t)$  of the foreground:  

$$\mu_1(t) = \frac{\mu - \text{cum\_mean}(t)}{w_1(t)}$$
    - e. Compute the between-class variance  $\sigma_b^2(t)$ :  

$$\sigma_b^2(t) = w_0(t) \times w_1(t) \times (\mu_0(t) - \mu_1(t))^2$$
7. **Find Optimal Threshold:**
  - The optimal threshold  $T$  is the one that maximizes the between-class variance  $\sigma_b^2(t)$ :  

$$T = \arg \max_t \sigma_b^2(t)$$
8. **Threshold the Image:**
  - Use the optimal threshold  $T$  to binarize the image:
    - Set pixel intensity to 0 (background) if intensity  $< T$ .
    - Set pixel intensity to 255 (foreground) if intensity  $\geq T$ .

## Apply Otsu Thresholding

In [3]:

```
def otsu_thresholding(image):
    histogram, _ = np.histogram(image, bins=256, range=(0, 256))

    # Normalize histogram
    histogram = histogram / float(image.size)

    max_variance = 0
    threshold = 0

    for i in range(1, 256):
        w0 = np.sum(histogram[:i])
        w1 = np.sum(histogram[i:])

        if w0 == 0:
            mu0 = 0
        else:
            mu0 = np.sum(np.arange(i) * histogram[:i]) / w0

        if w1 == 0:
```

```

        mu1 = 0
    else:
        mu1 = np.sum(np.arange(i, 256) * histogram[i:]) / w1

    variance = w0 * w1 * (mu0 - mu1) ** 2

    if variance > max_variance:
        max_variance = variance
        threshold = i

    return threshold

```

```

In [4]: def segmented_image(img, threshold):
        bin_image = np.zeros_like(img)
        for i in range(img.shape[0]):
            for j in range(img.shape[1]):
                if threshold < img[i][j]:
                    bin_image[i][j] = 1

        return bin_image

```

```

In [5]: img = cv2.imread('CoverImages/lena.tiff', 0)
        plt.imshow(img, cmap='gray')
        plt.title("Real Image")
        plt.xticks([])
        plt.yticks([])
        plt.show()

```

Real Image



```

In [6]: threshold_value = otsu_thresholding(img)
        threshold_value

```

Out[6]: 118

```
In [7]: seg = segmented_image(img,threshold_value)
```

```
In [8]: plt.imshow(seg, cmap='gray')  
plt.title("Apply Otsu Thresholding")  
plt.xticks([])  
plt.yticks([])  
plt.show()
```



```
In [9]: threshold_value = basic_global_thresholding(img,0.01)  
seg1 = segmented_image(img,threshold_value)
```

```
In [10]: threshold_value
```

Out[10]: 118

```
In [11]: plt.imshow(seg1, cmap='gray')  
plt.title("Apply Basic Global Thresholding")  
plt.xticks([])  
plt.yticks([])  
plt.show()
```

## Apply Basic Global Thresholding



In [ ]: