

# Lecture 5 – Image Segmentation (图像分割)

This lecture will cover:

- Morphological Image Processing (形态学图像处理)
  - Morphological operation
  - Morphological algorithm
- Image Segmentation (图像分割)
  - Point, Line and Edge Detection (点、线和边缘检测)
  - **Thresholding (阈值处理)**
  - **Region-based Segmentation (区域分割)**
  - Segmentation using Morphological Watersheds (形态学分水岭分割)

# Thresholding

## ➤ Global thresholding (全局阈值处理)

- Basic global thresholding
- Optimum global thresholding using Otsu's method
- Improve global thresholding by using image smoothing
- Improve global thresholding by using edges
- Multiple thresholds

## ➤ Variable thresholding (可变阈值处理)

- Image partitioning (图像分块)
- Variable thresholding based on local image properties
- Using moving average (移动平均)

# Foundation

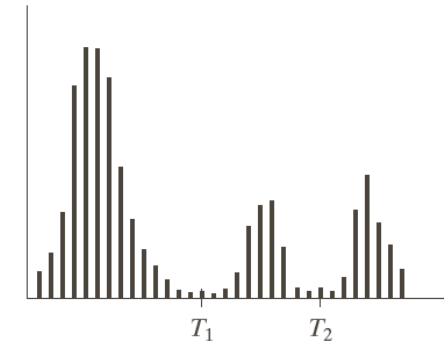
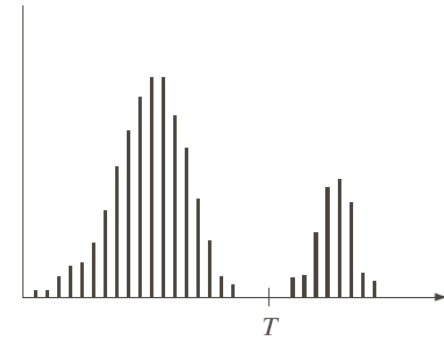
## ➤ Definition

$$g(x, y) = \begin{cases} 1, & f(x, y) > T \text{ (object points)} \\ 0, & f(x, y) \leq T \text{ (background points)} \end{cases}$$

Where

- *Global Thresholding (全局阈值处理)* if  $T$  is constant over an entire image
- *Variable/Local/Regional Thresholding (可变/局域/区域阈值处理)* if  $T$  changes over an image
- *Dynamic/Adaptive Thresholding (动态/自适应阈值处理)* if  $T$  depends on spatial coordinates  $(x, y)$
- *Multiple Thresholding (多阈值处理)*

$$g(x, y) = \begin{cases} a, & f(x, y) > T_2 \\ b, & T_1 < f(x, y) \leq T_2 \\ c, & f(x, y) \leq T_1 \end{cases}$$

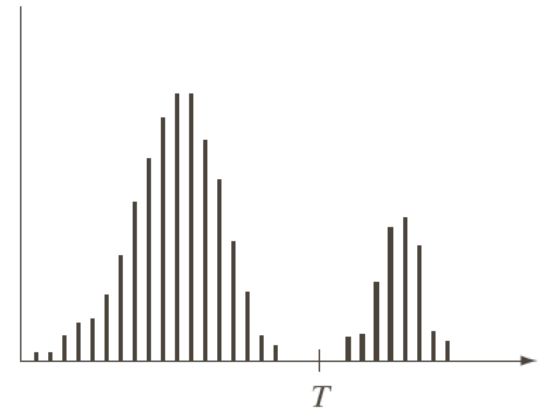


➤ **Matlab function:** `BW = im2bw(I,level)`

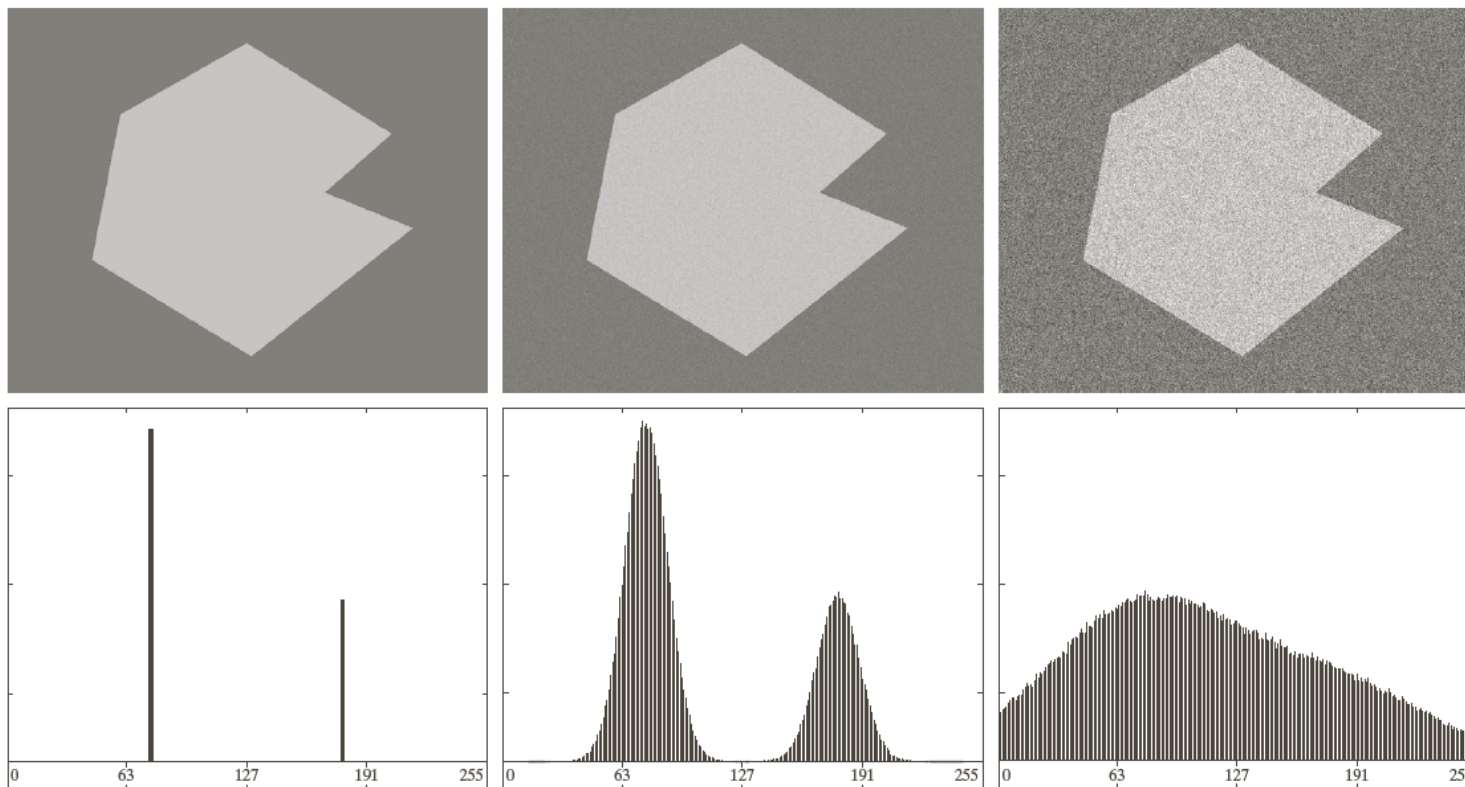
# Intensity Valley

**Key factors affecting the properties of the valley which separate the histogram modes**

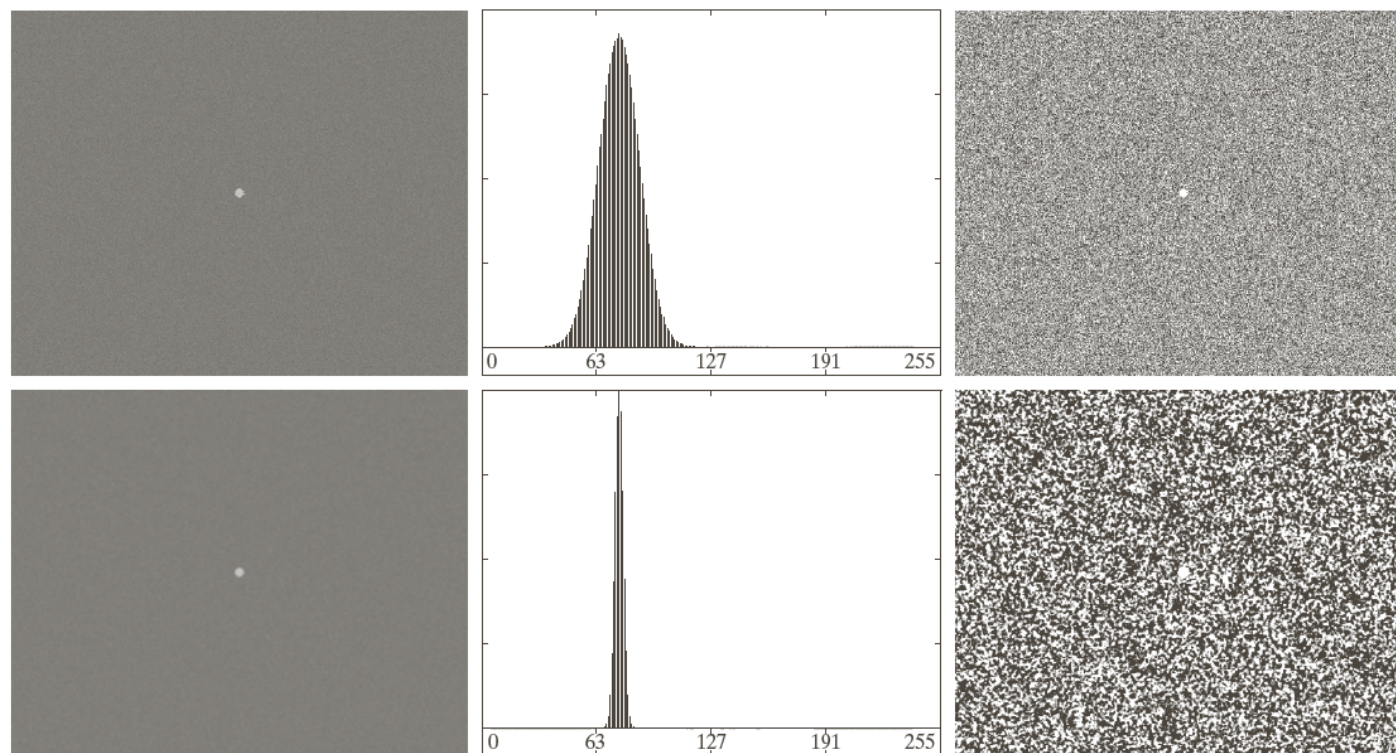
- The separation between peaks
- The noise content in the image
- The relative sizes of objects and background
- The uniformity of illumination source
- The uniformity of reflectance properties of the image



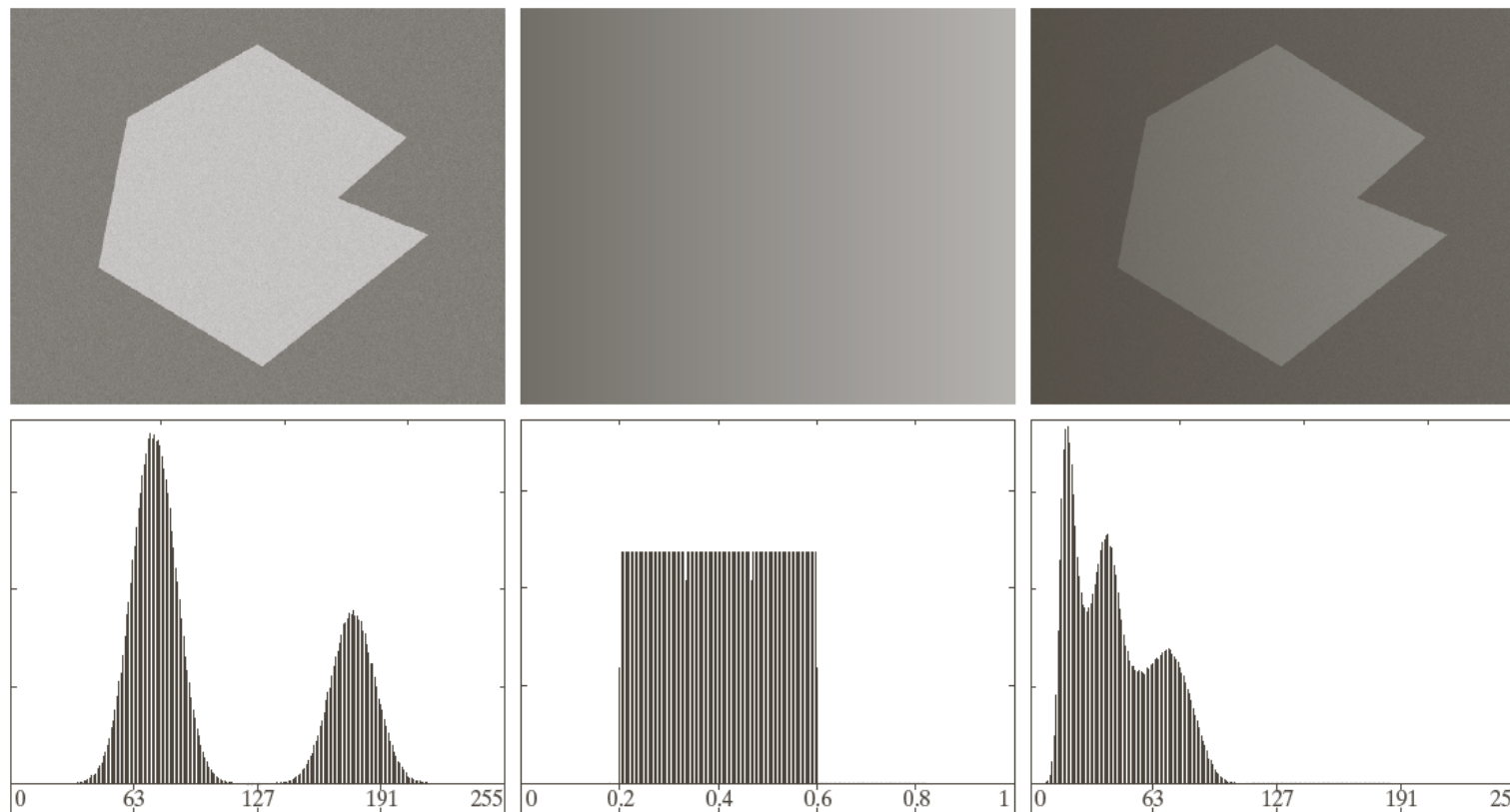
# Influence of Noise



# Influence of objects and background sizes



# Influence of Illumination and Reflection

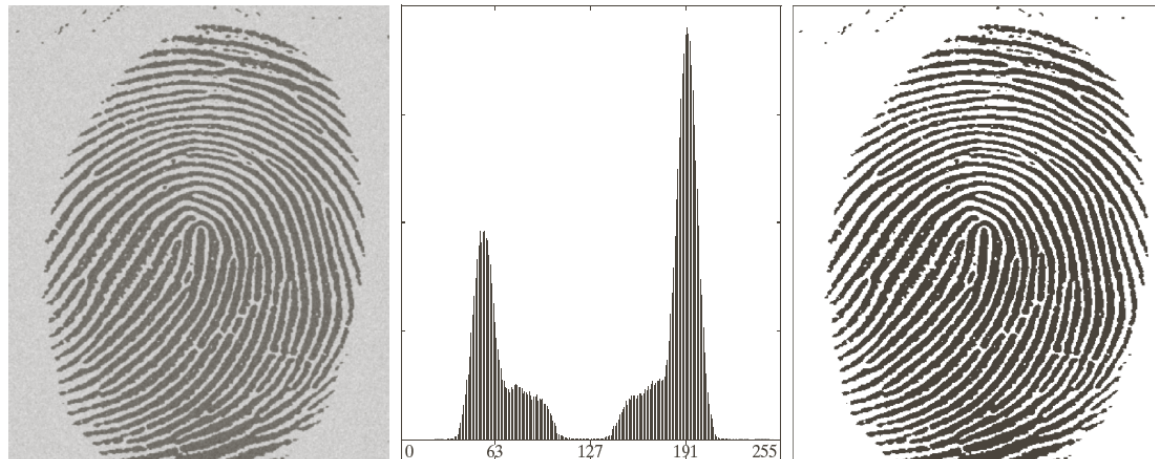




# Basic Global Thresholding

## ➤ Steps:

1. Select an initial estimate of the global threshold  $\mathbf{T}$ ;
2. Segment the image using  $\mathbf{T}$  to two groups  $G_1(>T)$  and  $G_2(\leq T)$  ;
3. Compute average intensity  $m_1$  and  $m_2$  for  $G_1$  and  $G_2$  respectively;
4. Compute new threshold  $\mathbf{T}=(m_1 + m_2)/2$ ;
5. Repeat 2-4 until the difference between  $\mathbf{T}$  in successive iteration is smaller than requirement.





# Otsu's Method

## ➤ Between-class variance (类间方差):

$$\sigma_B^2 = \frac{(m_G P_1 - m)^2}{P_1(1 - P_1)}$$

Where

$m_G$ : average intensity of entire image

$P_1$ : the cumulative probability of all pixels in the intensity range  $[0, k]$

$m$ : the average intensity up to level  $k$

## ➤ Matlab function:

`[level, EM] = graythresh(I)`

# Otsu's Method

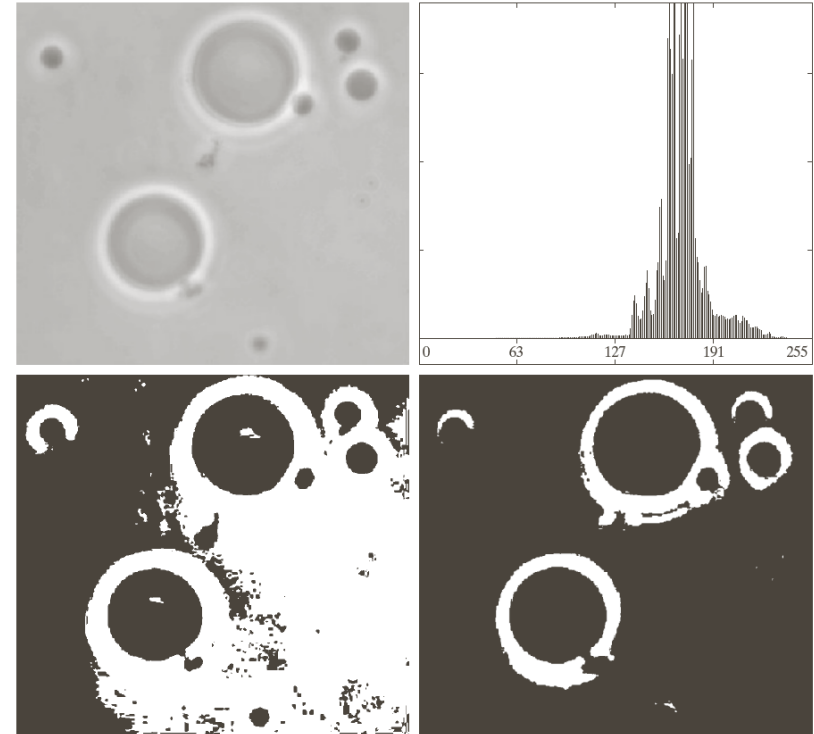
## ➤ Algorithm summary:

1. compute the normalized histogram of the input image  $p_i$ ;
2. compute the cumulative sums  $P_1(k) = \sum_{i=0}^k p_i$ ;
3. compute the cumulative means  $m(k) = \sum_{i=0}^k ip_i$
4. compute the global intensity mean  $m_G = \sum_{i=0}^{L-1} ip_i$
5. compute between-class variance  $\sigma_B^2$  for  $k = 0, 1, \dots, L - 1$ ;

$$\sigma_B^2 = \frac{(m_G P_1 - m)^2}{P_1(1 - P_1)}$$

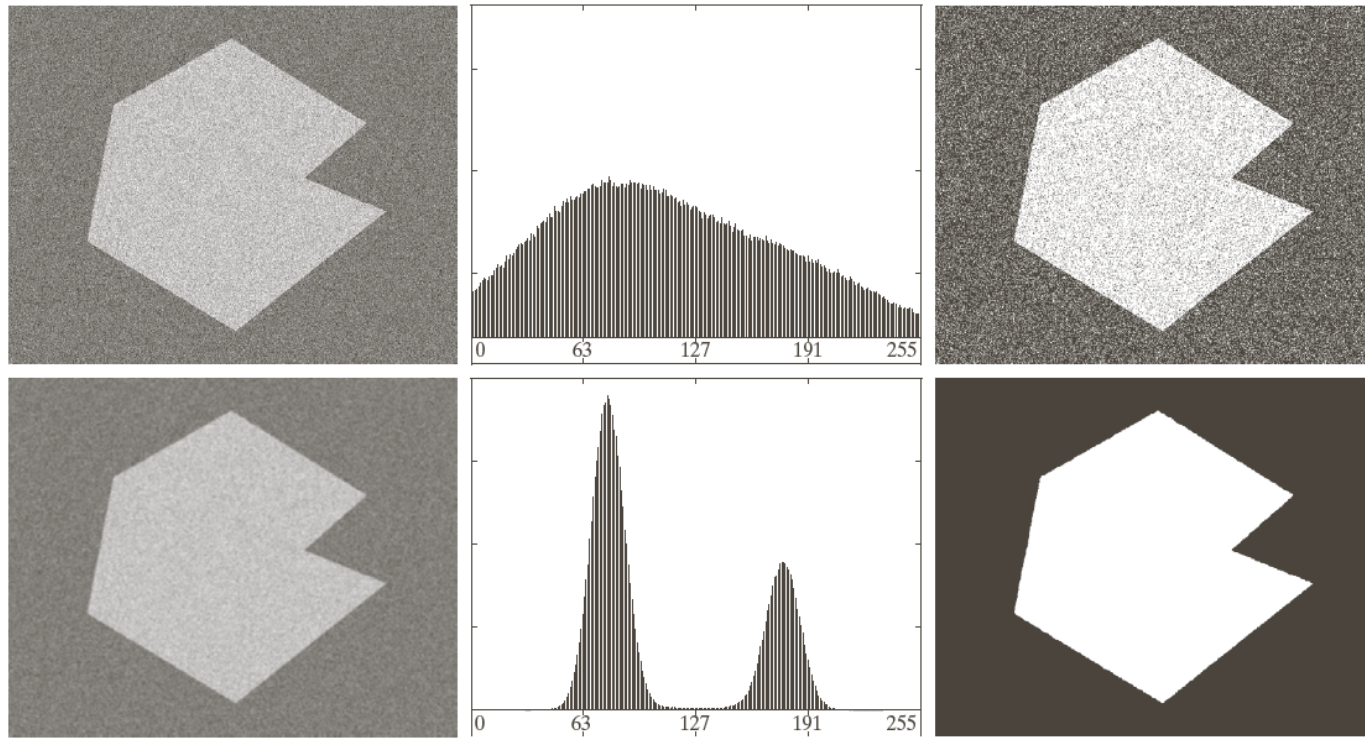
6. Obtain the Otsu threshold  $k^*$  when  $\sigma_B^2(k^*)$  is the maximum of all  $k$  value

7. Obtain the separability measure  $\eta^* = \frac{\sigma_B^2(k^*)}{\sigma_G^2}$



# Improve Global Thresholding

➤ Using image smoothing:



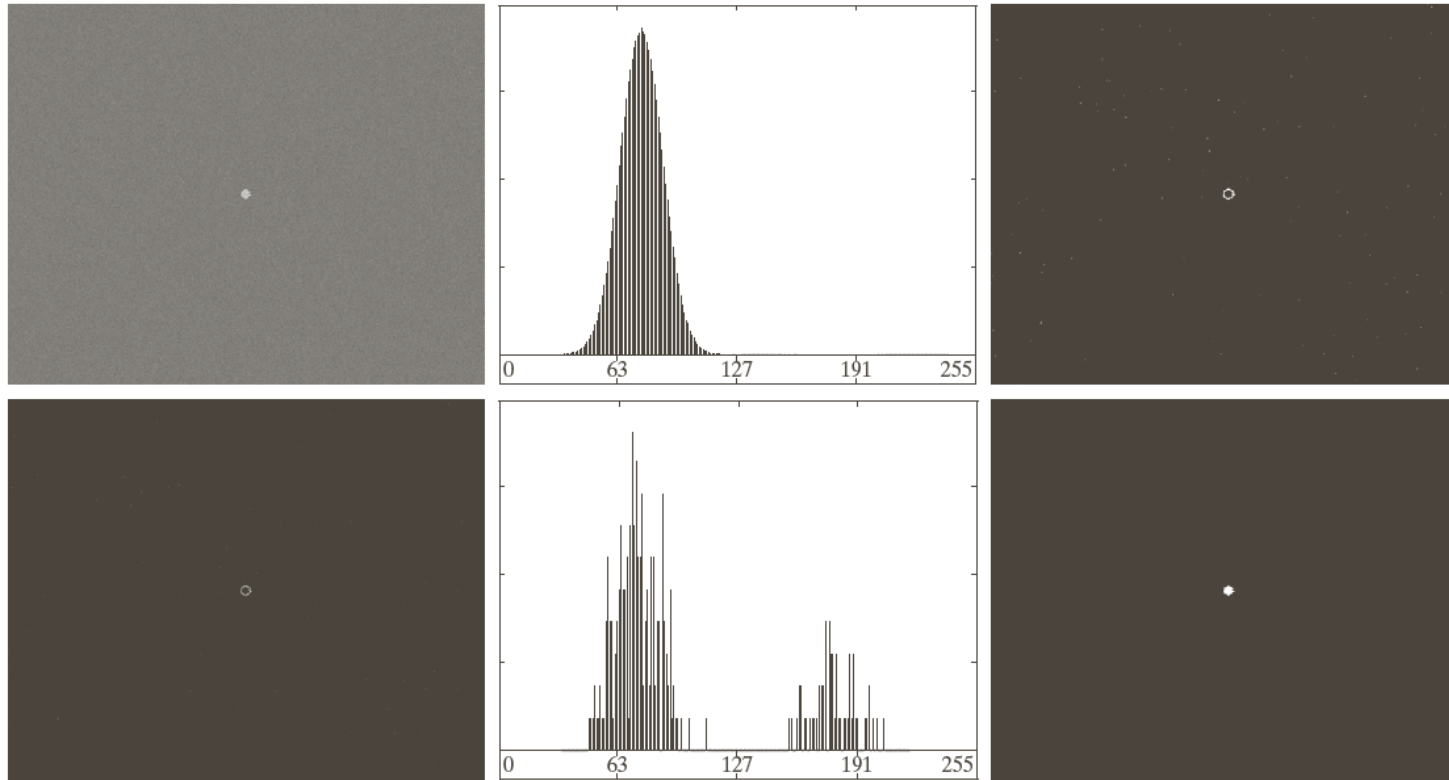
# Improve Global Thresholding

## ➤ Using edges:

1. compute an edge image from the input image  $f(x, y)$  using any edge detector;
2. specify a threshold value  $T$ ;
3. Threshold the edge image using  $T$  to produce a binary image  $g_T(x, y)$
4. compute a histogram using only the pixels in  $f(x, y)$  that correspond to the locations of the 1-valued pixels in  $g_T(x, y)$
5. use the histogram to segment  $f(x, y)$ ;

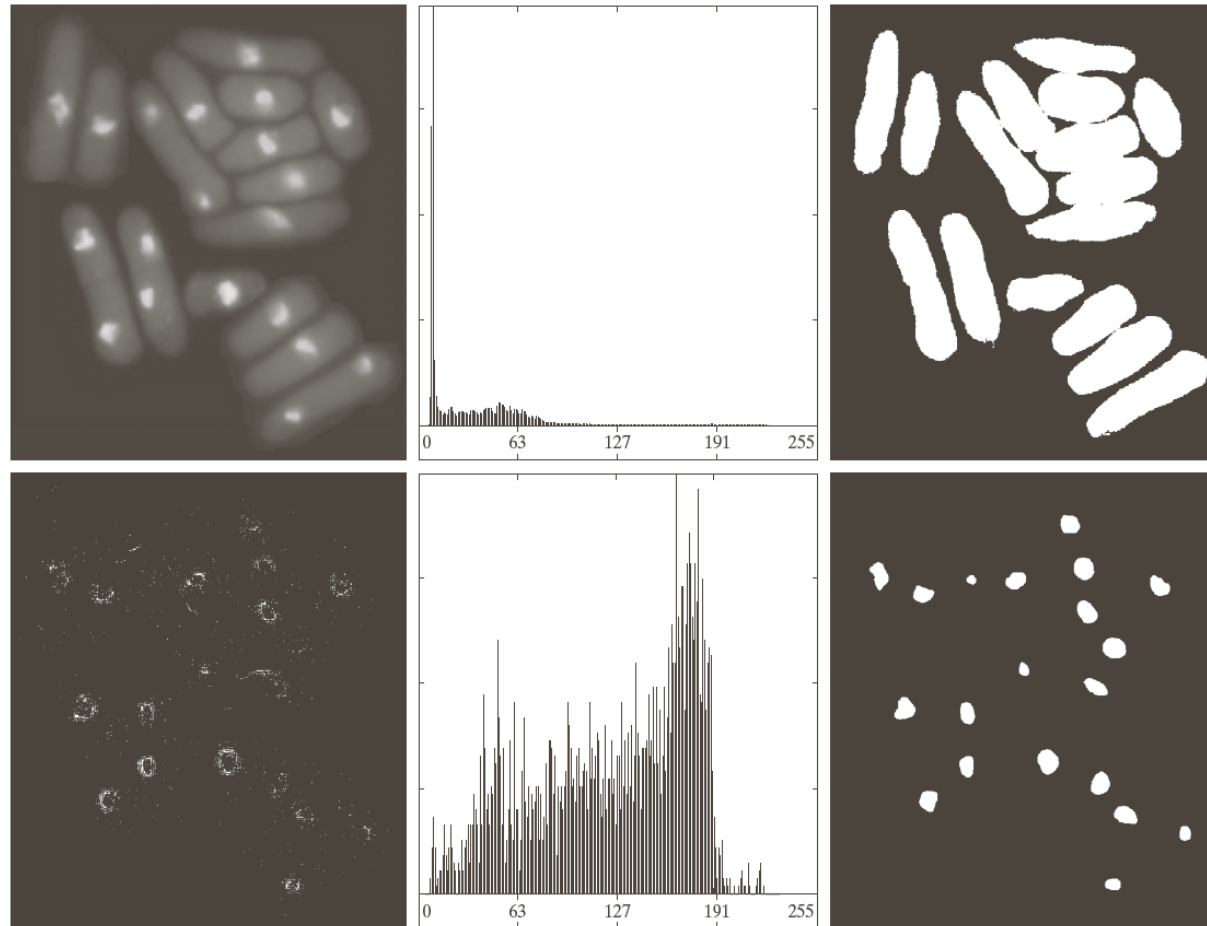
# Improve Global Thresholding

➤ Using edges:



# Improve Global Thresholding

## ➤ Using edges:



# Multiple thresholds

## ➤ Between-class variance (类间方差):

$$\sigma_B^2 = P_1(m_1 - m_G)^2 + P_2(m_2 - m_G)^2 + P_3(m_3 - m_G)^2$$

Where

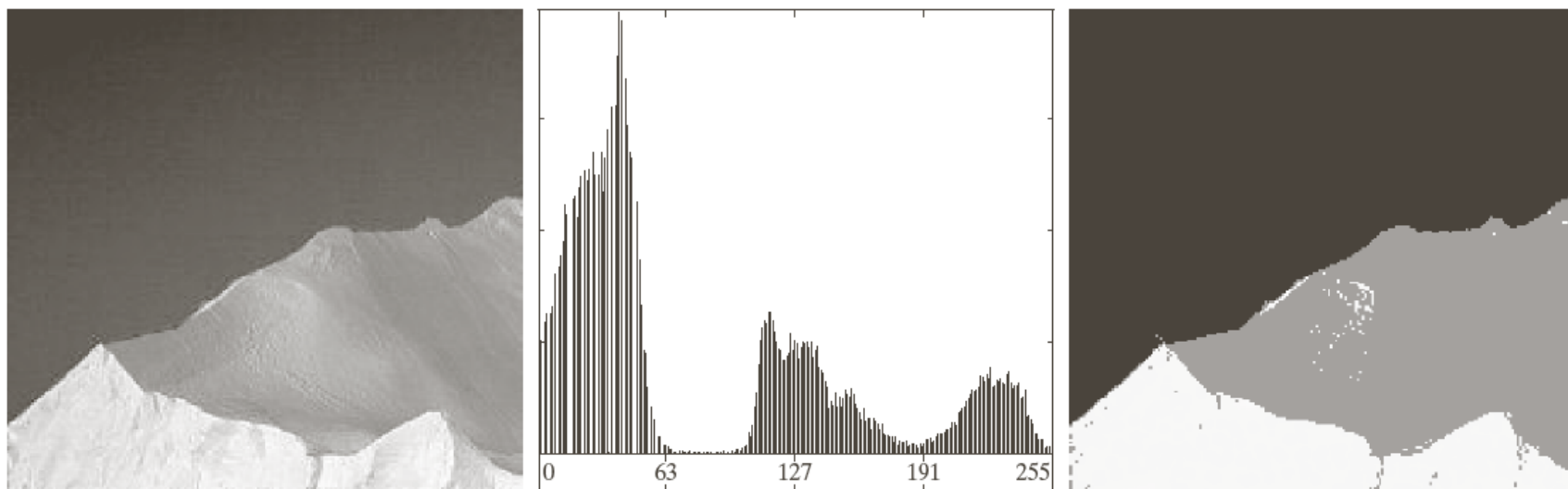
$$\begin{aligned} P_1 &= \sum_{i=0}^{k_1} p_i & P_2 &= \sum_{i=k_1+1}^{k_2} p_i & P_3 &= \sum_{i=k_2+1}^{L-1} p_i \\ m_1 &= \sum_{i=0}^{k_1} ip_i & m_2 &= \sum_{i=k_1+1}^{k_2} ip_i & m_3 &= \sum_{i=k_2+1}^{L-1} ip_i \\ P_1 m_1 + P_2 m_2 + P_3 m_3 &= m_G & P_1 + P_2 + P_3 &= 1 \end{aligned}$$

The two optimum thresholds  $k_1^*$  and  $k_2^*$  are the values that maximize  $\sigma_B^2(k_1, k_2)$ , then

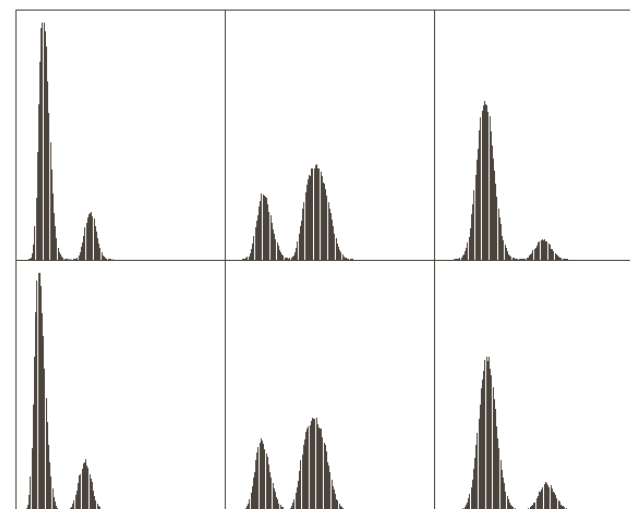
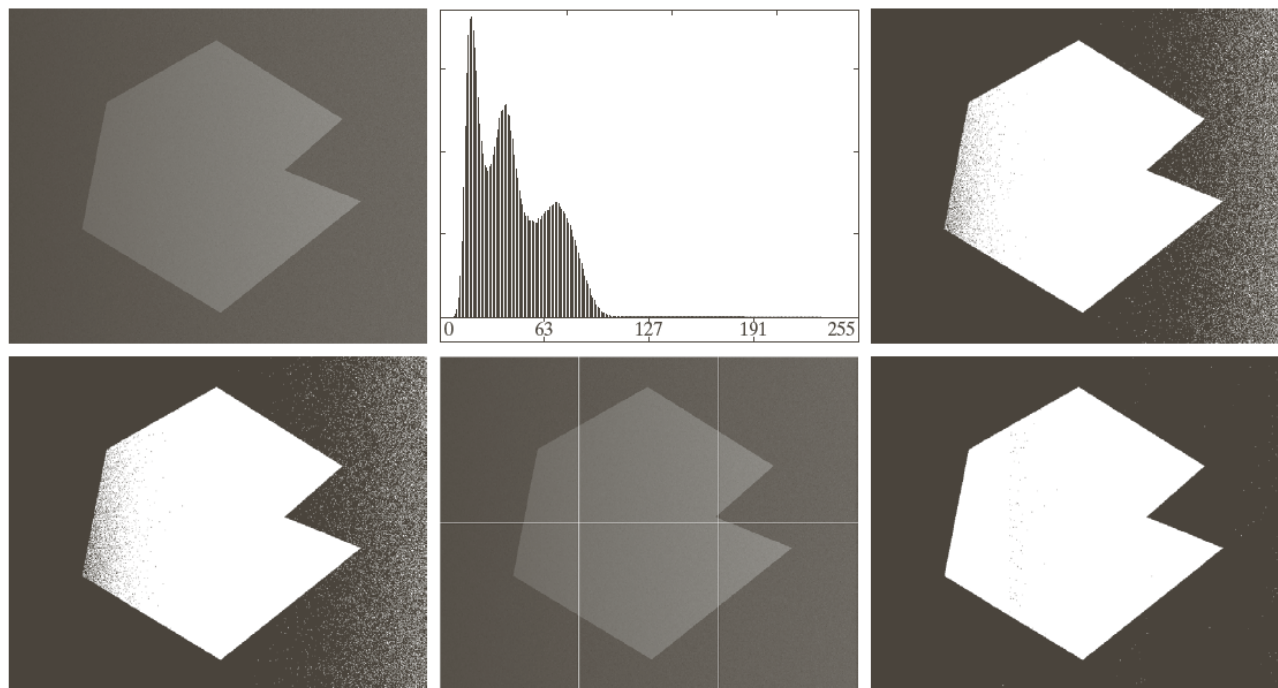
$$g(x, y) = \begin{cases} a, & f(x, y) \leq k_1^* \\ b, & k_1^* < f(x, y) \leq k_2^* \\ c, & f(x, y) > k_2^* \end{cases} \quad \text{and} \quad \eta(k_1^*, k_2^*) = \frac{\sigma_B^2(k_1^*, k_2^*)}{\sigma_G^2}$$



# Multiple thresholds



# Image partitioning (图像分块)



# Variable thresholding based on local image properties

## ➤ Algorithm:

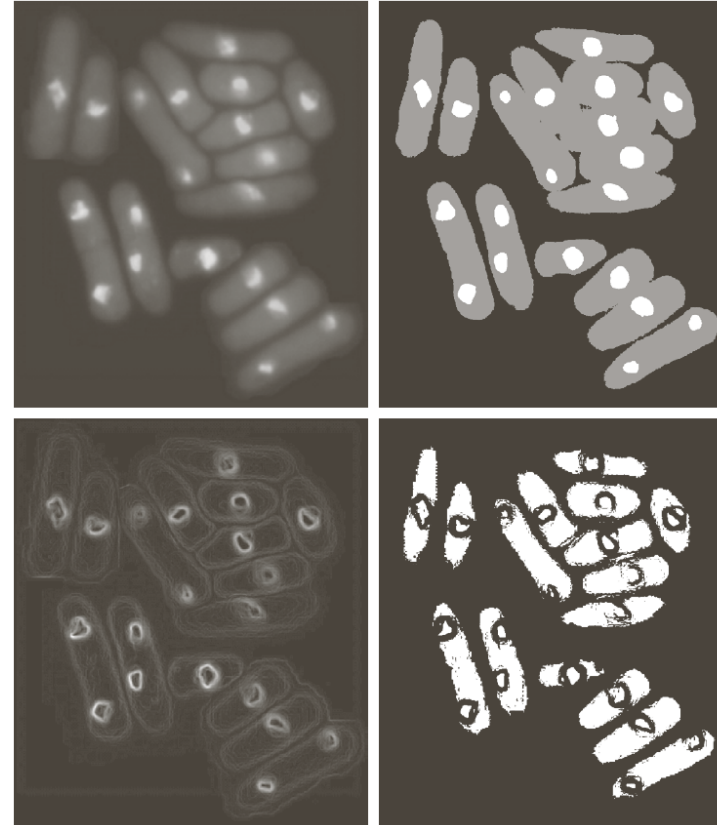
$$T_{xy} = a\sigma_{xy} + bm_{xy}$$

or

$$T_{xy} = a\sigma_{xy} + bm_G$$

## ➤ Matlab function:

```
g = stdfilt(f, nhood);
```



# Moving average (移动平均)

$$m(k+1) = \frac{1}{n} \sum_{i=k+2-n}^{k+1} z_i = m(k) + \frac{1}{n} (z_{k+1} - z_{k-n})$$



# Region Growing (区域生长)

## ➤ Algorithm based on 8-connectivity:

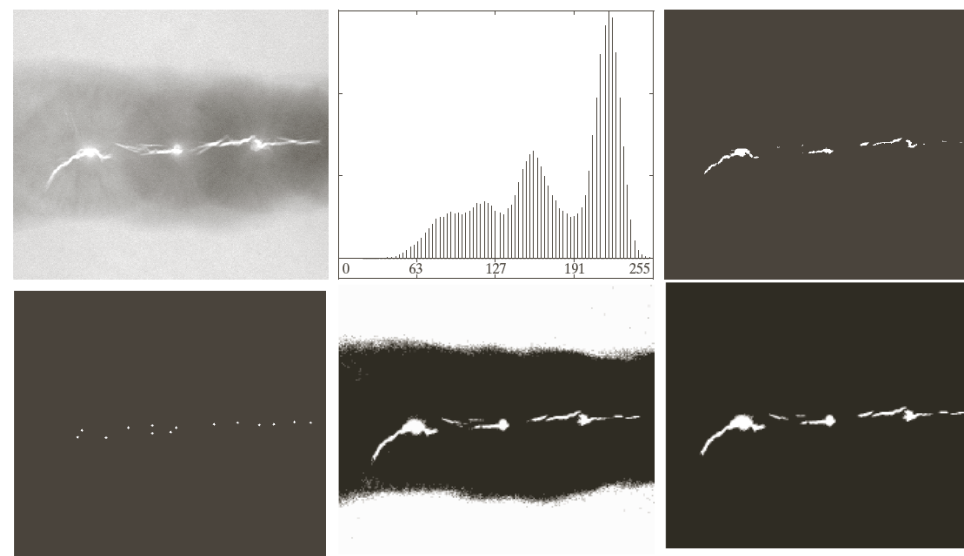
Where

$f(x, y)$ : input image

$S(x, y)$ : a seed array

$Q$ : a predicate to be applied at each location

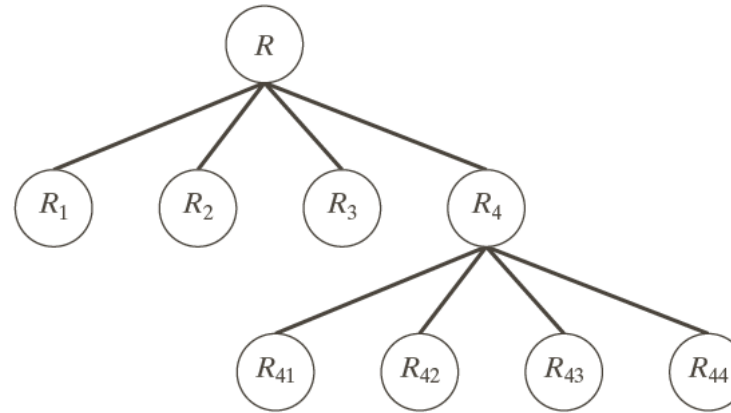
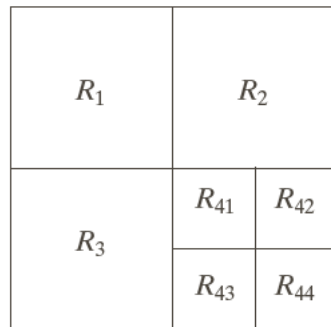
1. Find all connected components in  $S(x, y)$  and erode each component to one pixel;
2. Form an image  $f_Q$  based on if satisfying  $Q$
3. In  $f_Q$ , find all the 1-valued points which 8-connected to each seed point in  $S$ , and form an image  $g$ ;
4. Label each connected component in  $g$ , and this is the segmented image obtained by region growing.



# Region Splitting and Merging(区域分裂与聚合)

## ➤ Steps

1. Split into four disjoint quadrants any region  $R_i$  for which  $Q(R_i) = \text{False}$  (need to specify a minimum quadregion size beyond which no further splitting is carried out);
2. Merge any adjacent regions  $R_j$  and  $R_k$  for which  $Q(R_j \cup R_k) = \text{True}$ ;
3. Stop when no further merging is possible.



# Region Splitting and Merging(区域分裂与聚合)

