Image Processing 실습 12.

2022. 05. 23.

실습 수업 소개

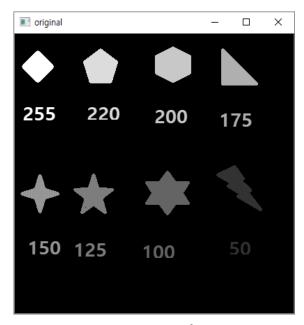
- 과목 홈페이지
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- 실습 중 질문사항
 - 수업중 질문 or 메일을 통한 질문
 - 메일로 질문할 때 [IP] 를 제목에 붙여주세요

목 차

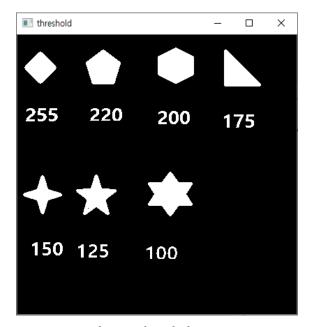
- 실습
 - Threshold
 - Adaptive Threshold
- 과제
 - Threshold: Otsu's method

Threshold

- ▶ 영상을 흑/백으로 분류하여 처리하는 것을 이진화 라고 한다.
- ▶ 이때 기준이 되는 임계값을 threshold value라고 한다

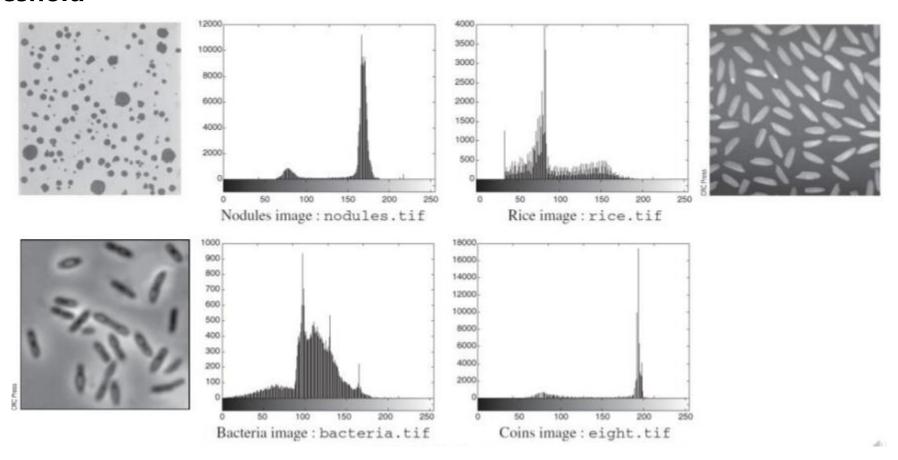


original

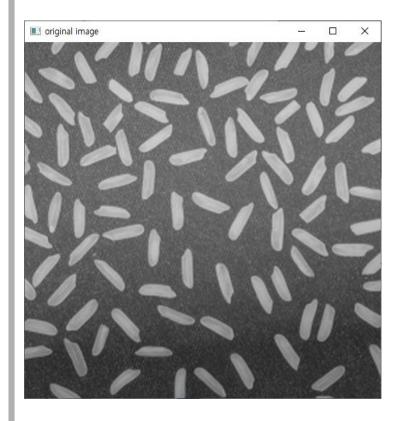


thresholding threshold value = 85

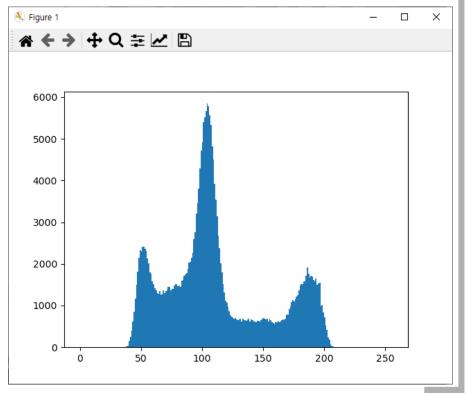
Threshold

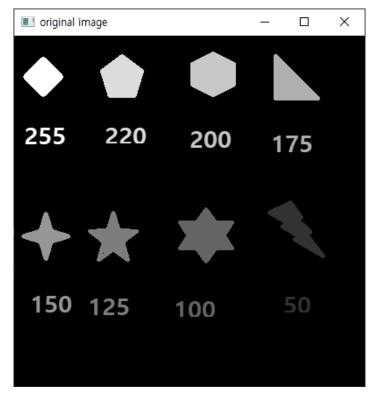


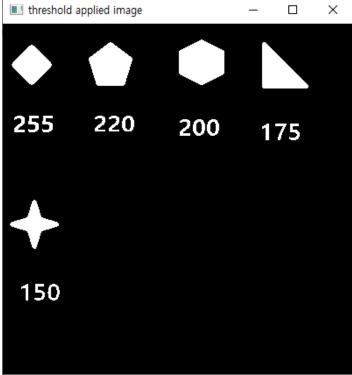
```
def apply_threshold(img, th=120):
    dst = np.zeros(img.shape, img.dtype)
    # TODO
    # Apply threshold
    return dst
jif __name__ == '__main__':
    img = cv2.imread('circles_adaptive_threshold.png', cv2.IMREAD_GRAYSCALE)
    img_th = apply_threshold(img, th=120)
    cv2.imshow('original image', img)
    cv2.imshow('threshold applied image', img_th)
    plt.hist(img.ravel(), 256, [0, 256])
    plt.show()
    cv2.waitKey(0)
    cv2.destroyAllWindows()
```

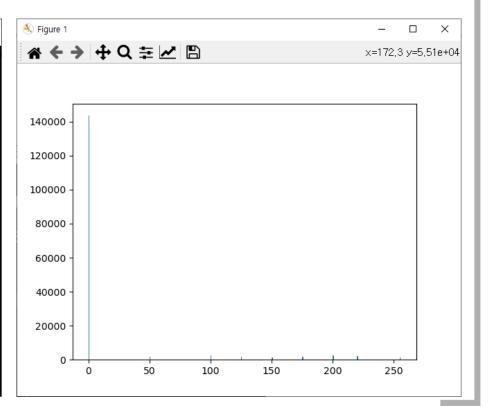


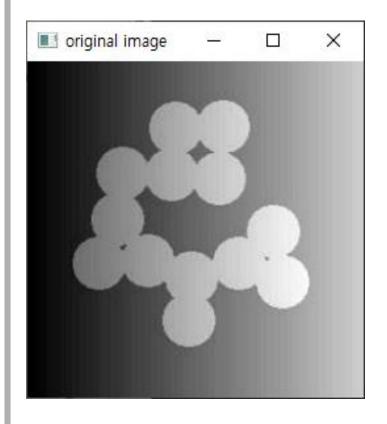


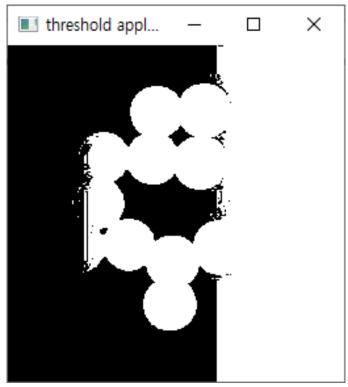


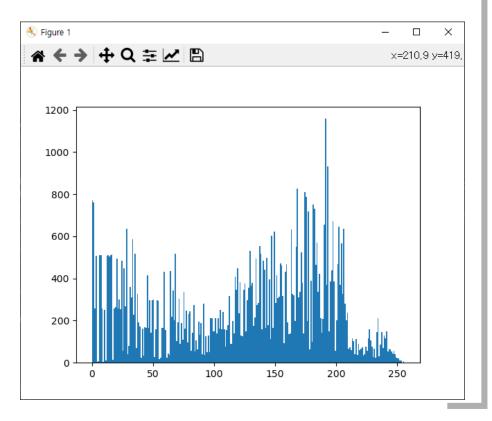




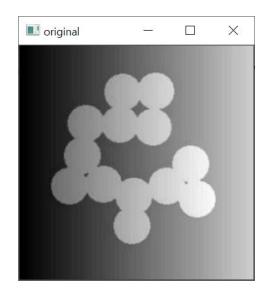


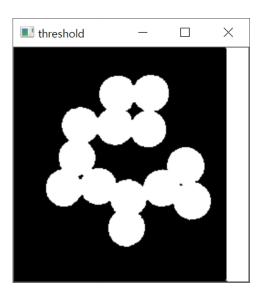






• Adaptive Threshold 실습

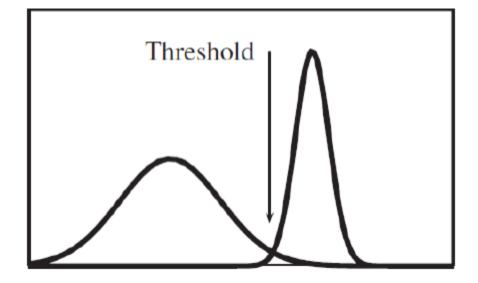


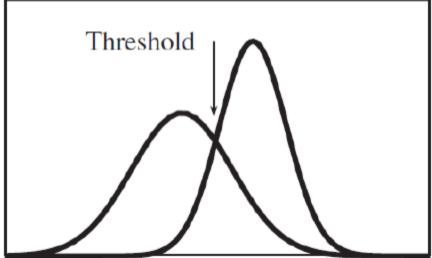


Adaptive Threshold 실습

```
def adaptive_threshold(img, group_num=4):
   img_split = np.hsplit(img, group_num)
    thresholds = list()
   dst = list()
   # TODO
   # 구분된 이미지마다 cv2의 otsu 방법으로 이진화.
   # 이후, 하나의 이미지로 다시 합친다.
   return dst, thresholds
if __name__ == '__main__':
   img = cv2.imread('circles_adaptive_threshold.png', cv2.IMREAD_GRAYSCALE)
   dst, val = adaptive_threshold(img, group_num=4)
   print('Threshold: ', val)
   cv2.imshow('original', img)
   cv2.imshow('adaptive threshold', dst)
   cv2.waitKey(0)
   cv2.destroyAllWindows()
```

• Threshold: Otsu's method



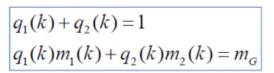


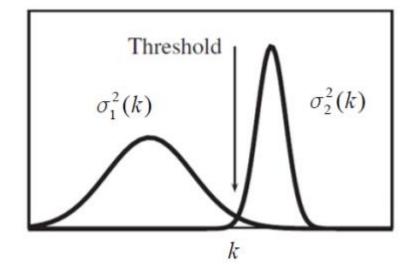
• Threshold: Otsu's method

$$\begin{aligned} p_i &= n_i / MN \quad \text{and} \quad \sum_{i=0}^{L-1} p_i = 1 \\ q_1(k) &= \sum_{i=0}^k p(i) \\ m_1(k) &= \frac{\sum_{i=0}^k ip(i)}{\sum_{k=0}^k p(i)} = \frac{1}{q_1(k)} \sum_{i=0}^k ip(i) \\ m_2(k) &= \frac{\sum_{i=k+1}^{L-1} ip(i)}{\sum_{i=k+1}^{L-1} p(i)} = \frac{1}{q_2(k)} \sum_{i=k+1}^{L-1} ip(i) \\ \sigma_1^2(k) &= \frac{1}{q_1(k)} \sum_{i=0}^k [i - m_1(k)]^2 p(i) \\ &= \frac{1}{q_1(k)} \sum_{i=0}^k i^2 p_i - m_1^2(k) \end{aligned} \qquad \begin{aligned} \sigma_2^2(k) &= \frac{1}{q_2(k)} \sum_{i=k+1}^{L-1} [i - m_2(k)]^2 p(i) \\ &= \frac{1}{q_2(k)} \sum_{i=k+1}^{L-1} i^2 p(i) - m_2^2(k) \end{aligned}$$

For an entire image,

$$m_G = \sum_{i=0}^{L-1} i p(i), \qquad \sigma_G^2 = \sum_{i=0}^{L-1} [i - m_G]^2 p(i)$$





$$\sigma_{W}^{2}(k) = q_{1}(k)\sigma_{1}^{2}(k) + q_{2}(k)\sigma_{2}^{2}(k)$$

$$\sigma_{B}^{2}(k) = q_{1}(k)[m_{1}(k) - m_{G}]^{2} + q_{2}(k)[m_{2}(k) - m_{G}]^{2}$$

$$= q_{1}(k)q_{2}(k)[m_{1}(k) - m_{2}(k)]^{2}$$

$$k_{opt} = \arg\min_{k} \sigma_{W}^{2}(k)$$

$$k^{*} = \arg\max_{k} \sigma_{B}^{2}(k)$$

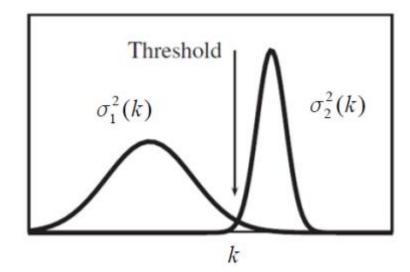
- Threshold: Otsu's method
 - Iterative하게 구할 수 있음

$$q_1(0) = p_0, m_1(0) = 0$$

$$q_1(k+1) = q_1(k) + p(k+1)$$

$$m_1(k+1) = \frac{q_1(k)m_1(k) + (k+1)p(k+1)}{q_1(k+1)}$$

$$m_2(k+1) = \frac{q_2(k)m_2(k) - (k+1)p(k+1)}{q_2(k+1)}$$



Threshold: Otsu's method

Threshold from cv2: 88.0

Threshold from my: 88



original



cv2



과제

• 제출 방법

- 코드 파일
 - 구현 결과가 포함된 python 파일(.py)
- 보고서
 - [IP]201900000_홍길동_11주차_과제.pdf
 - 보고서 양식 사용
 - PDF 파일 형식으로 제출(pdf가 아닌 다른 양식으로 제출시 감점)
- 제출 파일
 - [IP]201900000_홍길동_11주차_과제.zip
 - .py 파일과 pdf 보고서를 하나의 파일로 압축한 후, 양식에 맞는 이름으로 제출

QnA