Practice: Image Processing II

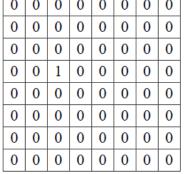
COMPUTER VISION (COURSE-HY24011)
Q YOUN HONG

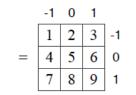
Content

- Linear Filtering
- Non-linear filtering
 - Adding random noise to an image
 - Median filtering
 - Bilateral filtering
- Integral Image
- Image Thresholding
 - Global Thresholding
 - Adaptive Thresholding
 - Otsu's Algorithm

 Convolution: computing a pixel value from a collection of neighboring pixels (area-based operation)

- Convolution applies the flipped window, whereas correlation does not
- A window u is also called a kernel or a filter

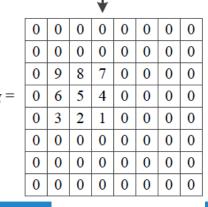


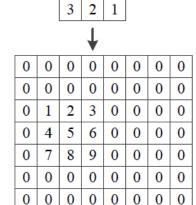


Output

Input







8

5 4

0

0

| | | _ | 1 | 2 | 3 | 4 | 5 | 6 | / | 8 | 9 | |
|--------|--|---|---|---|----|-----|-----|----|-----|----|---|--|
| | $f = \int_{0}^{\infty} $ | 0 | 0 | 1 | 0 | 2 | 2 | 4 | 3 | 1 | 0 | |
| | | | | Χ | Χ | X | | | | | | |
| Window | = 2 4 3 | | | 2 | 4 | 3 | | | | | | |
| u | | | | | + | | • | | | | | |
| | <i>g</i> = | - | 3 | 4 | 8 | 14 | 24 | 29 | 23 | 10 | - | |
| | | | | | Co | ori | rel | at | ioi | n | | |

| 0 | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 |
|---|---|---|---|----|----|----|----|----|---|
| 0 | 0 | 1 | 0 | 2 | 2 | 4 | 3 | 1 | 0 |
| | | X | X | X | | | | | |
| | | 3 | 4 | 2 | | | | | |
| | | | + | | • | | | | |
| - | 2 | 4 | 7 | 12 | 22 | 28 | 26 | 13 | - |
| | | | | | | | | | |

Convolution

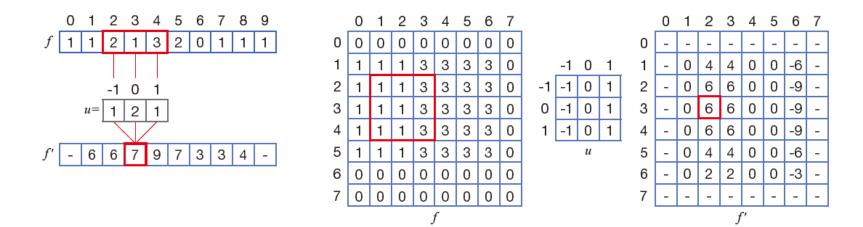
Window u

0 0

0

2D Convolution in OpenCV

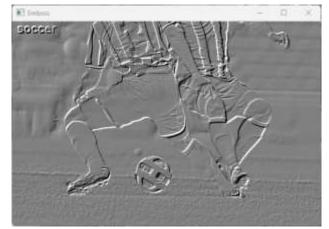
- Some well-known filters are already implemented as built-in functions in OpenCV, e.g.) Gaussian filter
- Custom filters can be also applied
- Handling boundary pixels 0 padding or copy padding
- Cannot be implemented in-place

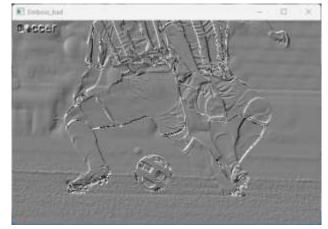


```
import cv2 as cv
       import numpy as np
       img=cv.imread('soccer.jpg')
       img=cv.resize(img,dsize=(0,0),fx=0.4,fy=0.4)
       gray=cv.cvtColor(img,cv.COLOR BGR2GRAY)
       cv.putText(gray, 'soccer',(10,20),cv.FONT HERSHEY SIMPLEX,0.7,(255,255,255),2)
       cv.imshow('Original',gray)
10
       smooth=np.hstack((cv.GaussianBlur(gray,(5,5),0.0),
11
                        cv.GaussianBlur(gray,(9,9),0.0),
12
                         cv.GaussianBlur(gray,(15,15),0.0)))
13
       cv.imshow('Smooth',smooth)
14
15
       femboss=np.array([[-1.0, 0.0, 0.0],
16
                         [ 0.0, 0.0, 0.0],
17
                         [ 0.0, 0.0, 1.0]])
18
       gray16=np.int16(gray)
19
       emboss=np.uint8(np.clip(cv.filter2D(gray16,-1,femboss)+128,0,255))
20
      emboss bad=np.uint8(cv.filter2D(gray16,-1,femboss)+128)
21
       emboss worse=cv.filter2D(gray,-1,femboss)
22
23
24
       cv.imshow('Emboss',emboss)
25
       cv.imshow('Emboss bad',emboss bad)
       cv.imshow('Emboss worse',emboss worse)
26
28
       cv.waitKey()
       cv.destroyAllWindows()
```









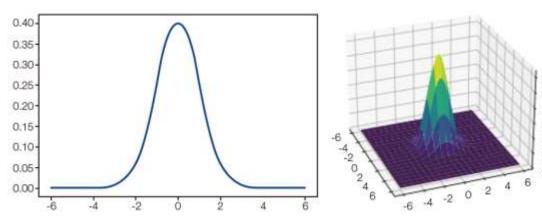


Execution results

```
import cv2 as cv
       import numpy as np
       img=cv.imread('soccer.jpg')
       img=cv.resize(img,dsize=(0,0),fx=0.4,fy=0.4)
       gray=cv.cvtColor(img,cv.COLOR BGR2GRAY)
       cv.putText(gray, 'soccer', (10,20), cv.FONT HERSHEY SIMPLEX, 0.7, (255, 255, 255), 2)
       cv.imshow('Original',gray)
9
10
       smooth=np.hstack((cv.GaussianBlur(gray,(5,5),0.0),
11
                         cv.GaussianBlur(gray,(9,9),0.0),
12
                         cv.GaussianBlur(gray,(15,15),0.0)))
13
       cv.imshow('Smooth',smooth)
14
15
       femboss=np.array([[-1.0, 0.0, 0.0],
16
                           0.0, 0.0, 0.0],
17
                         [ 0.0, 0.0, 1.0]])
18
       gray16=np.int16(gray)
19
       emboss=np.uint8(np.clip(cv.filter2D(gray16,-1,femboss)+128,0,255))
20
       emboss bad=np.uint8(cv.filter2D(gray16,-1,femboss)+128)
21
       emboss worse=cv.filter2D(gray,-1,femboss)
22
23
24
       cv.imshow('Emboss',emboss)
25
       cv.imshow('Emboss bad',emboss bad)
       cv.imshow('Emboss worse',emboss worse)
26
28
       cv.waitKey()
       cv.destroyAllWindows()
```



(L10-L12) Apply Gaussian filters of different sizes to img



1D Gaussian

$$g(x) = \frac{1}{\sigma\sqrt{2\pi}}e^{-\frac{x^2}{2\sigma^2}}$$

2D Gaussian

$$g(y,x) = \frac{1}{\sigma^2 2\pi} e^{-\frac{y^2 + x^2}{2\sigma^2}}$$

0.150

0.125

0.100

0.075

0.050

0.025

```
import cv2 as cv
      import numpy as np
      img=cv.imread('soccer.jpg')
      img=cv.resize(img,dsize=(0,0),fx=0.4,fy=0.4)
      gray=cv.cvtColor(img,cv.COLOR BGR2GRAY)
      cv.putText(gray, 'soccer', (10,20), cv.FONT HERSHEY SIMPLEX, 0.7, (255, 255, 255), 2)
      cv.imshow('Original',gray)
9
10
      smooth=np.hstack((cv.GaussianBlur(gray,(5,5),0.0),
                                                                  (L10-L12) cv.GaussianBlur(): Apply Gaussian filters of different sizes to img
11
                      cv.GaussianBlur(gray,(9,9),0.0),
12
                      cv.GaussianBlur(gray,(15,15),0.0)))
                                                                   - The size of window is given as an input
13
      cv.imshow('Smooth',smooth)
14
                                                                   - 3<sup>rd</sup> parameter: Gaussian standard deviation (\sigma), If 0, computed from
      femboss=np.array([[-1.0, 0.0, 0.0],
15
16
                        0.0, 0.0, 0.0],
                                                                   the size of window as sigma = 0.3*((ksize-1)*0.5 - 1) + 0.8
17
                       [ 0.0, 0.0, 1.0]])
18
                                                                   - Gaussian filters are used to smooth the input image
19
      gray16=np.int16(gray)
      emboss=np.uint8(np.clip(cv.filter2D(gray16,-1,femboss)+128,0,255)) (The larger the window, the smoother the resulting image)
20
      emboss bad=np.uint8(cv.filter2D(gray16,-1,femboss)+128)
21
      emboss worse=cv.filter2D(gray,-1,femboss)
22
23
24
      cv.imshow('Emboss',emboss)
                                                                  (*) Other built-in filters: cv.blur(), cv.boxFilter(), cv.medianFilter(),
      cv.imshow('Emboss bad',emboss bad)
25
      cv.imshow('Emboss worse',emboss worse)
                                                                   cv.bilateralFilter()
26
28
      cv.waitKey()
      cv.destroyAllWindows()
```

```
import cv2 as cv
       import numpy as np
       img=cv.imread('soccer.jpg')
       img=cv.resize(img,dsize=(0,0),fx=0.4,fy=0.4)
       gray=cv.cvtColor(img,cv.COLOR BGR2GRAY)
       cv.putText(gray, 'soccer', (10,20), cv.FONT HERSHEY SIMPLEX, 0.
       cv.imshow('Original',gray)
10
       smooth=np.hstack((cv.GaussianBlur(gray,(5,5),0.0),
11
                         cv.GaussianBlur(gray,(9,9),0.0),
12
                         cv.GaussianBlur(gray,(15,15),0.0)))
13
       cv.imshow('Smooth',smooth)
14
       femboss=np.array([[-1.0, 0.0, 0.0],
15
16
                           0.0, 0.0, 0.0],
17
                         [ 0.0, 0.0, 1.0]])
18
19
       gray16=np.int16(gray)
       emboss=np.uint8(np.clip(cv.filter2D(gray16,-1,femboss)+128,0,255))
20
       emboss bad=np.uint8(cv.filter2D(gray16,-1,femboss)+128)
21
       emboss worse=cv.filter2D(gray,-1,femboss)
22
23
24
       cv.imshow('Emboss',emboss)
       cv.imshow('Emboss bad',emboss bad)
25
       cv.imshow('Emboss worse',emboss worse)
26
28
       cv.waitKey()
       cv.destroyAllWindows()
```



(L15-L22) Design a custom emboss filter and apply to img

- Filter is defined as a 2D array
- cv.filter2D(): apply the filter to the input image
- (*) Users have to take care of type conversions (to avoid overflows)

Content

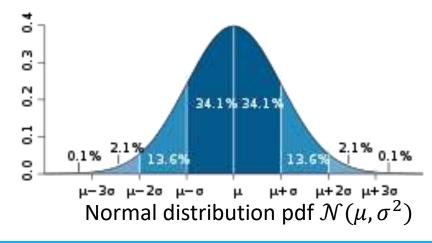
- Linear Filtering
- Non-linear filtering
 - Adding random noise to an image
 - Median filtering
 - Bilateral filtering
- Integral Image
- Image Thresholding
 - Global Thresholding
 - Adaptive Thresholding
 - Otsu's Algorithm

Noise Model

- Noise in digital images arises during image acquisition and transmission
- Imaging sensor creates undesirable noise due to many environmental and mechanical factors
- Noise model:

$$\underline{f(y,x)} = \underline{s(y,x)} + \underline{n(y,x)}$$
Digital image Input signal Noise function

- Gaussian noise model
 - Widely used to model noise (white noise)
 - n(y,x) has the normal distribution (the Gaussian distribution) pdf $\mathcal{N}(0,\sigma^2)$



```
import cv2 as cv
       import numpy as np
       # import matplotlib.pyplot as plt
       img = cv.imread('lenna.png', cv.IMREAD GRAYSCALE)
       if img is None:
           print('File not found')
       img32 = np.float32(img)
       noise = np.zeros((img.shape[0], img.shape[1]), np.float32)
10
11
12
       cv.randn(noise, 0, 10)
       noiseimg1 = np.uint8(np.clip(cv.add(img32, noise), 0, 255))
13
14
15
       cv.randn(noise, 0, 20)
       noiseimg2 = np.uint8(np.clip(cv.add(img32, noise), 0, 255))
16
17
18
       cv.randn(noise, 0, 30)
       noiseimg3 = np.uint8(np.clip(cv.add(img32, noise), 0, 255))
19
20
21
       titles = ['original', 'stddev10', 'stddev20', 'stddev30']
       images = [img, noiseimg1, noiseimg2, noiseimg3]
22
23
24
       # Drawing images with matplotlib
25
       #for i in range(4):
            plt.subplot(2,2,i+1), plt.imshow(images[i], 'gray')
26
27
            plt.title(titles[i])
           #plt.xticks([]),plt.yticks([])
28
29
       #plt.show()
30
31
       for i in range(4):
           cv.imshow(titles[i], images[i])
32
           cv.imwrite(titles[i] +'.jpg', images[i])
33
34
       cv.waitKey()
       cv.destroyAllWindows()
```



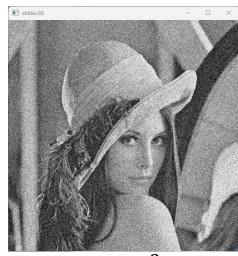
Original Image



With $\mathcal{N}(0,20^2)$ noise



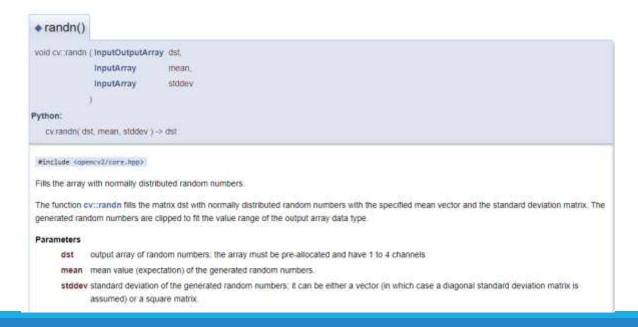
With $\mathcal{N}(0,10^2)$ noise



With $\mathcal{N}(0,30^2)$ noise

```
import cv2 as cv
       import numpy as np
      # import matplotlib.pyplot as plt
      img = cv.imread('lenna.png', cv.IMREAD GRAYSCALE)
       if img is None:
           print('File not found')
       img32 = np.float32(img)
      noise = np.zeros((img.shape[0], img.shape[1]), np.float32)
10
11
12
       cv.randn(noise, 0, 10)
      noiseimg1 = np.uint8(np.clip(cv.add(img32, noise), 0, 255))
13
14
15
       cv.randn(noise, 0, 20)
      noiseimg2 = np.uint8(np.clip(cv.add(img32, noise), 0, 255))
16
17
18
       cv.randn(noise, 0, 30)
19
      noiseimg3 = np.uint8(np.clip(cv.add(img32, noise), 0, 255))
20
21
      titles = ['original', 'stddev10', 'stddev20', 'stddev30']
      images = [img, noiseimg1, noiseimg2, noiseimg3]
22
23
24
       # Drawing images with matplotlib
25
       #for i in range(4):
           plt.subplot(2,2,i+1), plt.imshow(images[i], 'gray')
26
27
           plt.title(titles[i])
           #plt.xticks([]),plt.yticks([])
28
29
       #plt.show()
30
31
       for i in range(4):
32
           cv.imshow(titles[i], images[i])
          cv.imwrite(titles[i] +'.jpg', images[i])
33
34
       cv.waitKev()
35
      cv.destroyAllWindows()
```

- (L10) creates a 2D array of the same size as 'img' to save a noise value at each pixel
- (L12) cv.randn(): fills the array with normally distributed random numbers
- Mean(μ) = 0, the standard deviation(σ) = 10
- Since negative numbers can be created, 'noise' should save signed numbers (see L9,L10)



```
import cv2 as cv
       import numpy as np
      # import matplotlib.pyplot as plt
      img = cv.imread('lenna.png', cv.IMREAD GRAYSCALE)
       if img is None:
           print('File not found')
       img32 = np.float32(img)
10
      noise = np.zeros((img.shape[0], img.shape[1]), np.float32)
11
12
       cv.randn(noise, 0, 10)
      noiseimg1 = np.uint8(np.clip(cv.add(img32, noise), 0, 255))
13
14
      cv.randn(noise, 0, 20)
15
      noiseimg2 = np.uint8(np.clip(cv.add(img32, noise), 0, 255))
16
17
18
       cv.randn(noise, 0, 30)
      noiseimg3 = np.uint8(np.clip(cv.add(img32, noise), 0, 255))
19
20
21
      titles = ['original', 'stddev10', 'stddev20', 'stddev30']
22
      images = [img, noiseimg1, noiseimg2, noiseimg3]
23
24
      # Drawing images with matplotlib
25
       #for i in range(4):
           plt.subplot(2,2,i+1), plt.imshow(images[i], 'gray')
26
           plt.title(titles[i])
27
           #plt.xticks([]),plt.yticks([])
28
29
       #plt.show()
30
31
       for i in range(4):
32
           cv.imshow(titles[i], images[i])
          cv.imwrite(titles[i] +'.jpg', images[i])
33
34
       cv.waitKev()
      cv.destroyAllWindows()
35
```

- (L13) cv.add(): adds the Gaussian noise to input image (unsigned 'img' has been converted to 'img32')
- (L13) cv.clip(): clips the range of resulting image to [0,255]

(*) Drawing images

- You can draw images either using OpenCV or Matplotlib
- (L24~L29): draw images using drawing functions in matplotlib
- plt.subplot(): sets the layout of images with the given row and column numbers

- Images disturb more for larger σ (Compare original and $\mathcal{N}(0,30^2)$ -images)
- Consistent noises are allocated with a Gaussian noise model

(Question) Increase the standard deviation(σ) of the noise.

(Question) Change the mean(μ) of the noise.



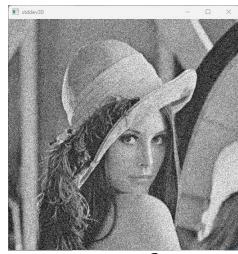
Original Image



With $\mathcal{N}(0,20^2)$ noise



With $\mathcal{N}(0,10^2)$ noise



With $\mathcal{N}(0,30^2)$ noise

Example 2. Bilateral Filtering

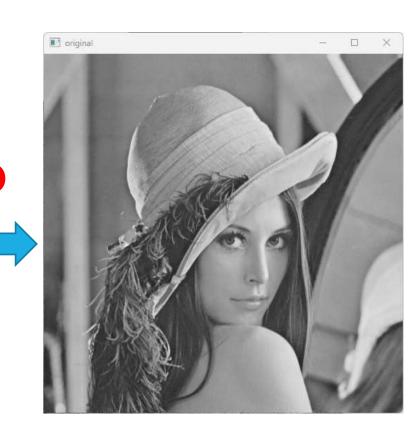
Can we remove noises from the image and restore the image?



Original Image



With a Gaussian noise



Example 2-1. Noise Removal by Gaussian Filter

Let's apply the Gaussian filter to an image with noise



Original Image



With a Gaussian noise

```
import cv2 as cv
       img = cv.imread('stddev30.jpg', cv.IMREAD GRAYSCALE)
       if img is None:
           print('File not found')
       gImg1 =
       gImg2
       gImg3 =
       gImg4 =
       titles = ['original', 'GaussianBlur5', 'GaussianBlur9',
                 'GaussianBlur15', 'GaussianBlur27']
13
       images = [img, gImg1, gImg2, gImg3, gImg4]
15
       for i in range(5):
           cv.imshow(titles[i], images[i])
       cv.waitKev()
       cv.destroyAllWindows()
```

Noise Removal by Gaussian Filter





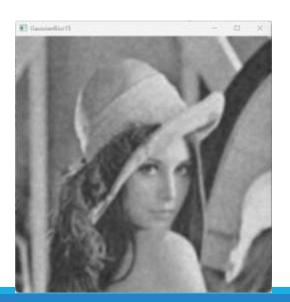


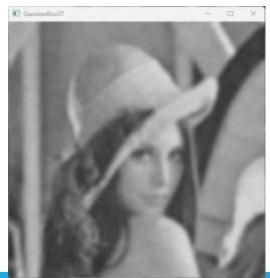


(b) Image with noise



- Gaussian filter is used in noise removal:
- When applying to large, flat areas, noises are blurred with the surrounding pixels
- However, the Gaussian filter also blurs edges of the image (Compare (a) with 'GaussianBlur27')





Applying Gaussian filters to (b) (window size k=5, 9, 15, 27)

Example 2-2. Bilateral Filter

```
import cv2 as cv
       import numpy as np
      img = cv.imread('lenna.png', cv.IMREAD_GRAYSCALE)
      if img is None:
           print('File not found')
 7
8
      # add noise to ima
      img32 = np.float32(img)
10
      noise = np.zeros((img.shape[0], img.shape[1]), dtype=np.float32)
11
12
      cv.randn(noise, 0, 5)
13
      noiseimg = np.uint8(np.clip(cv.add(img32, noise), 0, 255))
14
15
      # apply gaussian filter
      gImg = cv.GaussianBlur(noiseimg, (5,5), 0.0)
16
17
18
      # apply bilateral filter
19
      bImg = cv.bilateralFilter(noiseimg, -1, 10, 5)
20
21
      titles = ['original', 'noise', 'GaussianBlur',
22
                 'BilateralFilter'l
23
      images = [img, noiseimg, gImg, bImg]
24
25
      for i in range(4):
26
           cv.imshow(titles[i], images[i])
27
       cv.waitKey()
      cv.destroyAllWindows()
```



Original image



Filtered with Gaussian filter



Image with noise



Filtered with bilateral filter

Example 2-2. Bilateral Filter

- Bilateral filter: removing noise while preserving edges of the image [Tomasi98]
- $g(j,i) = \frac{\sum_{k,l} f(k,l)w(j,i,k,l)}{\sum_{k,l} w(j,i,k,l)}$, where $w(j,i,k,l) = \exp(-\frac{(j-k)^2 + (i-l)^2}{2\sigma_d^2} - \frac{\|f(j,i) - f(k,l)\|^2}{2\sigma_r^2})$ $(\Rightarrow w(p,q) = \alpha G_{\sigma_d}(\|p-q\|)G_{\sigma_r}(|f(p)-f(q)|)$
- G_{σ_d} : Gaussian blurring based on pixel distance
- G_{σ_r} : Gaussian blurring based on pixel values
- If *p*, *q* are across the edge,
- $\Rightarrow |f(\mathbf{p}) f(\mathbf{q})|$ is large
- $\Rightarrow G_{\sigma_r}(|f(p) f(q)|)$ approaches to zero
- $\Rightarrow w(p,q)$ becomes small
- \Rightarrow q does not contribute to p's pixel value



Original image



Filtered with Gaussian filter



Image with noise

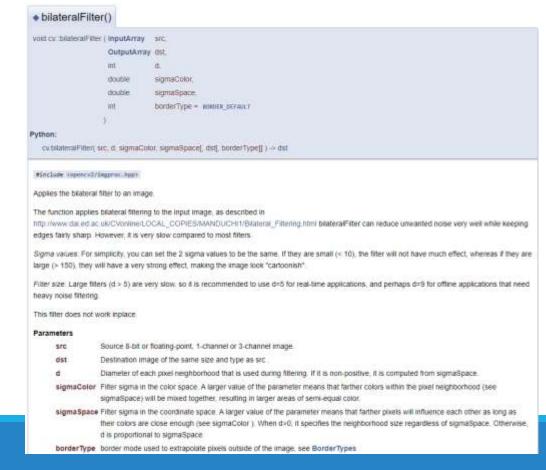


Filtered with bilateral filter

Example 2-2. Bilateral Filter

```
import cv2 as cv
       import numpy as np
      img = cv.imread('lenna.png', cv.IMREAD_GRAYSCALE)
       if img is None:
           print('File not found')
 7
 8
       # add noise to ima
       img32 = np.float32(img)
10
      noise = np.zeros((img.shape[0], img.shape[1]), dtype=np.float32)
11
12
       cv.randn(noise, 0, 5)
13
      noiseimg = np.uint8(np.clip(cv.add(img32, noise), 0, 255))
14
15
       # apply gaussian filter
16
       gImg = cv.GaussianBlur(noiseimg, (5,5), 0.0)
17
18
       # apply bilateral filter
19
      bImg = cv.bilateralFilter(noiseimg, -1, 10, 5)
20
21
      titles = ['original', 'noise', 'GaussianBlur',
22
                 'BilateralFilter'l
23
      images = [img, noiseimg, gImg, bImg]
24
25
       for i in range(4):
26
           cv.imshow(titles[i], images[i])
27
       cv.waitKey()
       cv.destroyAllWindows()
```

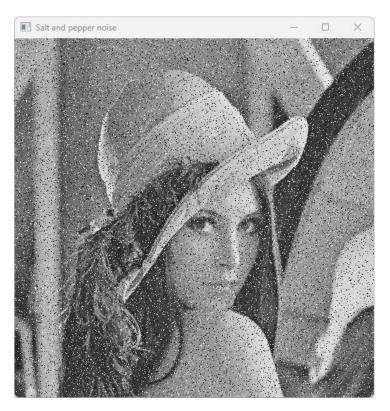
- (L19) cv.bilateralFilter(): applies the bilateral filter to an image
- Set 2 sigma values: σ_r (sigmaColor), σ_d (sigmaSpace)
- (Question) Increase σ_r and σ_d to large numbers



- Salt & Pepper noise (Impulse noise)
 - Caused by sharp and sudden disturbances in signals
 - Shown as random white and black pixels in the image

```
import cv2 as cv
       import random
       img = cv.imread('lenna.png', cv.IMREAD_GRAYSCALE)
       if img is None:
          print('File not found')
       # add salt and pepper noise to ima
       noiseNum = img.size//10
10
11
       for i in range(noiseNum):
           row = random.randrange(img.shape[0])
12
13
           col = random.randrange(img.shape[1])
          img[row,col] = (i % 2) * 255
14
15
16
       cv.imshow('Salt and pepper noise', img)
17
       # Apply Gaussian Filter
18
       # Apply Median Filter
19
21
       cv.waitKey()
       cv.destroyAllWindows()
```





• (Question) Apply a 3x3 Gaussian Filter and a 3x3 median filter to the image with salt and pepper noise and compare the filtering results

(Refer to https://docs.opencv.org/4.x/d4/d86/group__imgproc__filter.html)

```
import cv2 as cv
       import random
       img = cv.imread('lenna.png', cv.IMREAD_GRAYSCALE)
       if img is None:
           print('File not found')
       # add salt and pepper noise to ima
       noiseNum = img.size//10
10
11
       for i in range(noiseNum):
12
           row = random.randrange(img.shape[0])
           col = random.randrange(img.shape[1])
13
           img[row,col] = (i % 2) * 255
14
15
       cv.imshow('Salt and pepper noise', img)
16
18
       # Apply Write codes here!
# Apply Median Filter
19
       cv.waitKey()
       cv.destroyAllWindows()
```



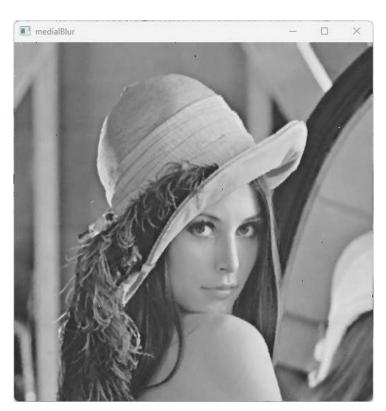


(Answer)

```
import cv2 as cv
       import random
       img = cv.imread('lenna.png', cv.IMREAD_GRAYSCALE)
       if img is None:
           print('File not found')
       # add salt and pepper noise to img
       noiseNum = img.size//10
10
11
       for i in range(noiseNum):
           row = random.randrange(img.shape[0])
12
           col = random.randrange(img.shape[1])
13
           img[row,col] = (i % 2) * 255
14
15
       cv.imshow('Salt and pepper noise', img)
16
17
18
       # Apply Gaussian Filter
19
20
21
       # Apply Median Filter
22
23
24
25
       cv.waitKey()
       cv.destroyAllWindows()
27
```

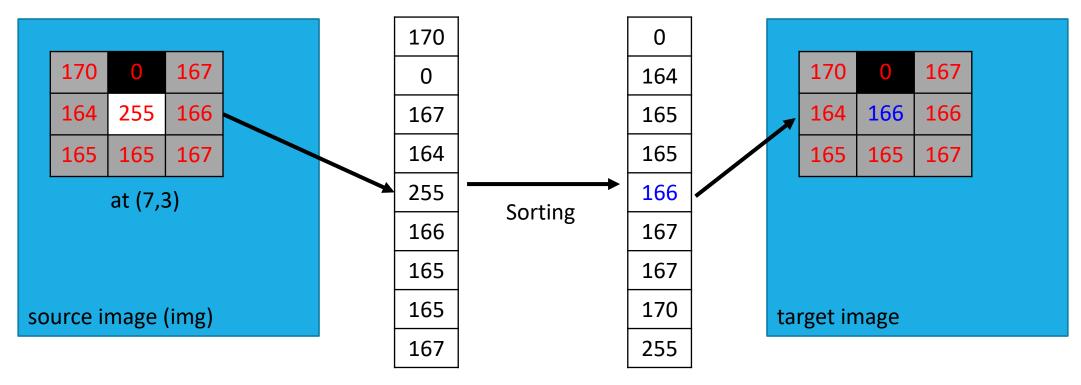


Gaussian filtering result



Median filtering result

Median filter selects the median pixel value within a window



⇒ Removes salt and pepper noise easily

Content

- Non-linear filtering
 - Adding random noise to an image
 - Median filtering
 - Bilateral filtering
- Integral Image
- Image Thresholding
 - Global Thresholding
 - Adaptive Thresholding
 - Otsu's Algorithm

Summed Area Table (Integral Image)

- Accelerate convolution when box filters of different sizes are used
- Summed area table: $s(j,i) = \sum_{k=0}^{j} \sum_{l=0}^{i} f(k,l)$ or s(j,i) = s(j-1,i) + s(j,i-1) s(j-1,i-1) + f(j,i)
- s(j,i) is also called an integral image
- To find the summed area in $[j_0, j_1] \times [i_0, i_1]$, we need 4 samples:

$$S([j_0, j_1], [i_0, i_1]) = s(j_1, i_1) - s(j_1, i_0 - 1) - s(j_0 - 1, i_1) + s(j_0 - 1, i_0 - 1)$$

| 3 | 2 | 7 | 2 | 3 |
|---|---|---|---|---|
| 1 | 5 | 1 | 3 | 4 |
| 5 | 1 | 3 | 5 | 1 |
| 4 | 3 | 2 | 1 | 6 |
| 2 | 4 | 1 | 4 | 8 |

| 3 | 5 | 12 | 14 | 17 |
|----|----|----|----|----|
| 4 | 11 | 19 | 24 | 31 |
| 9 | 17 | 28 | 38 | 46 |
| 13 | 24 | 37 | 48 | 62 |
| 15 | 30 | 44 | 59 | 81 |

| 10 | 3 | 5 | 12 | 14 | 17 |
|----|----|----|----|----|----|
| | 4 | 11 | 19 | 24 | 31 |
| | 9 | 17 | 28 | 38 | 46 |
| | 13 | 24 | 37 | 48 | 62 |
| | 15 | 30 | 44 | 59 | 81 |

- (a) Original image
- (b) Summed area table
- (c) Computation of area sum

(a)
$$S = 24$$

(b)
$$s = 28$$

(c)
$$S = 24$$

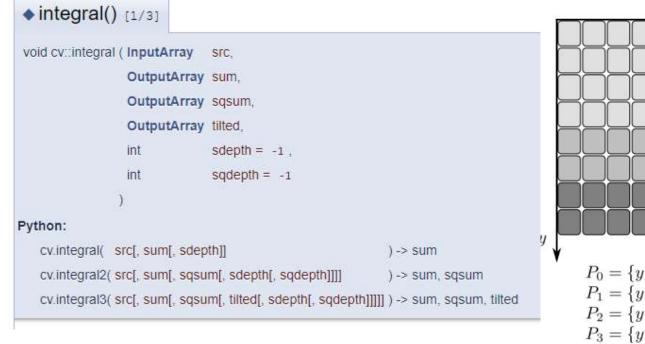
Integral Image

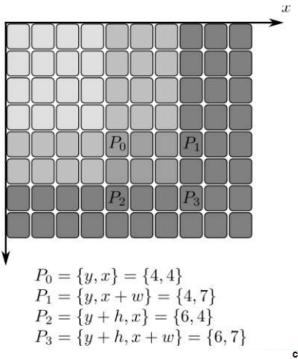
- cv.integral(): computes the integral images of the source image
 - Useful for fast blurring or block correlation computation with a variable window size

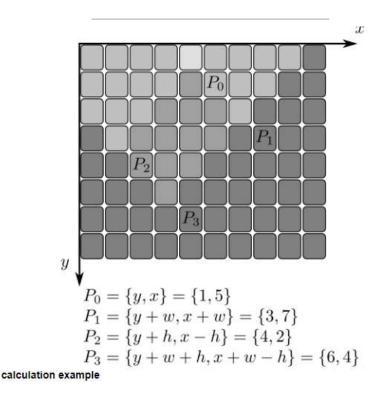
The function calculates one or more integral images for the source image as follows:

```
	extstyle{sum}(X,Y) = \sum_{x < X, y < Y} 	extstyle{image}(x,y)
 ♦ integral() [1/3]
                                                                                                                                        \operatorname{\mathsf{sqsum}}(X,Y) = \sum_{x \in Y, y \in V} \operatorname{\mathsf{image}}(x,y)^2
 void cv::integral ( InputArray src,
                      OutputArray sum.
                                                                                                                               \mathtt{tilted}(X,Y) = \sum_{y < Y.abs(x-X+1) \leq Y-y-1} \mathtt{image}(x,y)
                      OutputArray sqsum,
                      OutputArray tilted,
                                                                      Using these integral images, you can calculate sum, mean, and standard deviation over a specific up-right or rotated rectangular region of the image in a
                                      sdepth = -1,
                                                                      constant time, for example:
                             sqdepth = -1
                      int
                                                                                                        \sum_{x_1 \leq x < x_2, \, y_1 \leq y < y_2} \mathtt{image}(x,y) = \mathtt{sum}(x_2,y_2) - \mathtt{sum}(x_1,y_2) - \mathtt{sum}(x_2,y_1) + \mathtt{sum}(x_1,y_1)
Python:
    cv.integral( src[, sum[, sdepth]]
                                                                           ) -> sum
    cv.integral2( src[, sum[, sqsum[, sdepth[, sqdepth]]]]
                                                                          ) -> sum, sqsum
    cv.integral3( src[, sum[, sqsum[, tilted[, sdepth[, sqdepth]]]]] ) -> sum, sqsum, tilted
```

Integral Image







Parameters

src input image as $W \times H$, 8-bit or floating-point (32f or 64f).

sum integral image as $(W+1) \times (H+1)$, 32-bit integer or floating-point (32f or 64f).

sqsum integral image for squared pixel values; it is (W+1) imes (H+1), double-precision floating-point (64f) array.

tilted integral for the image rotated by 45 degrees; it is (W+1) imes (H+1) array with the same data type as sum.

sdepth desired depth of the integral and the tilted integral images, CV_32S, CV_32F, or CV_64F.

sqdepth desired depth of the integral image of squared pixel values, CV_32F or CV_64F.

Integral Image

(Question) Let's implement a 5x5 box filtering using integral

image

```
import cv2 as cv
       import numpy as np
       img = cv.imread('lenna.png', cv.IMREAD_GRAYSCALE)
       if img is None:
           print("file not found")
       bImg = cv.blur(img, (5,5))
 9
       sumimg = cv.integral(img)
10
       bImg2 = np.zeros((img.shape[0], img.shape[1]))
12
       # write filtering with a 5x5 using suming Write code here!
13
14
15
       titles = ['Original Image', 'Blurred', 'With IntegralImg']
16
       images = [img, bImg, bImg2]
17
18
       for i in range(3):
19
20
           cv.imshow(titles[i], images[i])
21
       cv.waitKey()
       cv.destroyAllWindows()
```

Content

- Non-linear filtering
 - Adding random noise to an image
 - Median filtering
 - Bilateral filtering
- Integral Image
- Image Thresholding
 - Global Thresholding
 - Adaptive Thresholding
 - Otsu's Algorithm

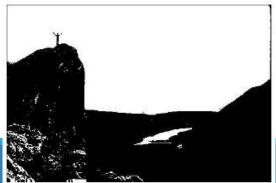
Binarization by image thresholding

• Binarization: convert a grayscale image to a binary image with a threshold T,

$$b(j,i) = \begin{cases} 1, & \text{if } f(j,i) > T \\ 0, & \text{if } f(j,i) \le T \end{cases}$$

- In OpenCV, a binary image has either 0 or 255 as pixel values
- By binarization, an image is divided into region of interest (ROI) and non-ROI
 (E.g. foreground objects and background scene)







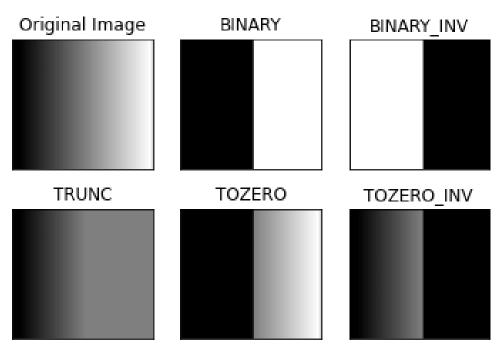






Example 5-1. Thresholding in OpenCV

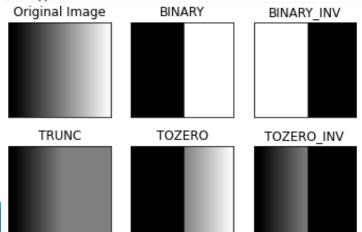
```
import cv2 as cv
    import matplotlib.pyplot as plt
    img = cv.imread('gradient.png', cv.IMREAD_GRAYSCALE)
    if img is None:
        print("file not found")
    ret, thresh1 = cv.threshold(img, 127, 255, cv.THRESH_BINARY)
    ret, thresh2 = cv.threshold(img, 127, 255, cv.THRESH_BINARY_INV)
    ret, thresh3 = cv.threshold(img, 127, 255, cv. THRESH_TRUNC)
    ret, thresh4 = cv.threshold(img, 127, 255, cv. THRESH TOZERO)
    ret, thresh5 = cv.threshold(img, 127, 255, cv.THRESH_TOZERO_INV)
    titles = ['Original Image', 'BINARY', 'BINARY_INV', 'TRUNC',
               'TOZERO', 'TOZERO_INV']
14
    images = [img, thresh1, thresh2, thresh3, thresh4, thresh5]
    for i in range(6):
17
        plt.subplot(2,3,i+1),
        plt.imshow(images[i], 'gray', vmin=0, vmax=255)
18
        plt.title(titles[i])
19
        plt.xticks([]),plt.yticks([])
    plt.show()
```



Execution results

Example 5-1. Thresholding in OpenCV

```
import cv2 as cv
    import matplotlib.pyplot as plt
    img = cv.imread('gradient.png', cv.IMREAD_GRAYSCALE)
    if img is None:
         print("file not found")
    ret, thresh1 = cv.threshold(img, 127, 255, cv. THRESH_BINARY)
    ret, thresh2 = cv.threshold(img, 127, 255, cv. THRESH BINARY INV)
    ret, thresh3 = cv.threshold(img, 127, 255, cv.THRESH_TRUNC)
    ret, thresh4 = cv.threshold(img, 127, 255, cv. THRESH_TOZERO)
    ret, thresh5 = cv.threshold(img, 127, 255, cv. THRESH_TOZERO_INV)
    titles = ['Original Image', 'BINARY', 'BINARY_INV', 'TRUNC',
               'TOZERO', 'TOZERO INV']
    images = [img, thresh1, thresh2, thresh3, thresh4, thresh5]
    for i in range(6):
         plt.subplot(2,3,i+1),
17
         plt.imshow(images[i], 'gray', vmin=0, vmax=255)
18
19
         plt.title(titles[i])
         plt.xticks([]),plt.yticks([])
    plt.show()
```

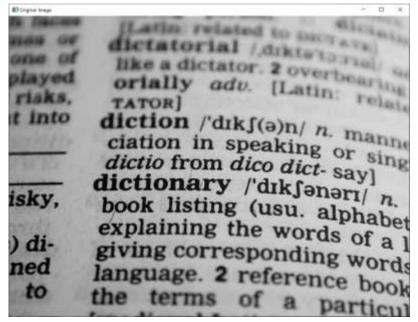


- (L8-L11) cv.threshold(): apply the thresholding to the input grayscale image
- Threshold and maximum value are passed as parameters

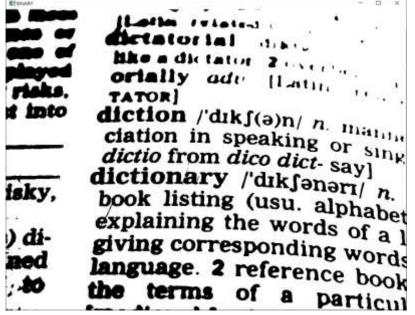
| Enumerator | |
|----------------------|--|
| cv.THRESH_BINARY | $dst(y,x) = \begin{cases} maxval & if \ src(y,x) > thresh \\ 0 & otherwise \end{cases}$ |
| cv.THRESH_BINARY_INV | $dst(y,x) = \begin{cases} 0 & if \ src(y,x) > thresh \\ maxval & otherwise \end{cases}$ |
| cv.THRESH_TRUNC | $dst(y,x) = \begin{cases} thresh & if \ src(y,x) > thresh \\ src(y,x) & otherwise \end{cases}$ |
| cv.THRESH_TOZERO | $dst(y,x) = \begin{cases} src(y,x) & if \ src(y,x) > thresh \\ 0 & otherwise \end{cases}$ |
| cv.THRESH_TOZERO_INV | $dst(y,x) = \begin{cases} 0 & if \ src(y,x) > thresh \\ src(y,x) & otherwise \end{cases}$ |
| cv.THRESH_OTSU | Use Otsu algorithm to choose optimal T |
| cv.THRESH_TRIANGLE | Use Triangle algorithm to choose optimal T |

Example 5-1. Thresholding in OpenCV

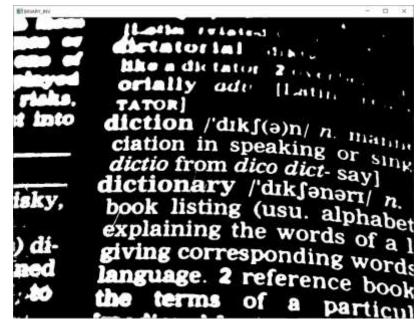
 Comparison between CV.THRESH_BINARY and CV.THRESH_BINARY_INV



Original Image



Thresholding with CV.THRESH_BINARY (T=128)



Thresholding with CV.THRESH_BINARY_INV (T=128)

- ⇒ In OpenCV, objects are often labelled as white, where the background as black
- ⇒ Use CV.THRESH_BINARY or CV.THRESH_BINARY_INV based on images or applications

- cv.threshold() performs global binarization
 - Use the same threshold value T for full image
 - If an image has the different lighting conditions in different areas, a single threshold T would not perform well





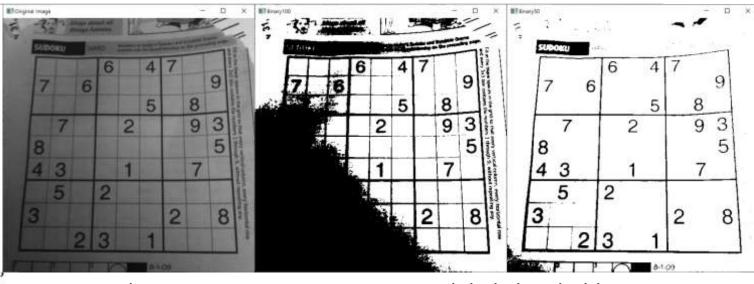
(Left) Original image (Right) Binarized image with a global threshold (T = 100)

⇒ The lower left region of the right image has not been separated well as the area has small pixel values Overall

(Question) Compare with T=50

- Adaptive thresholding: apply a different threshold value for each pixel
 - Set a window around the pixel and determine T from the image histogram within the window
 - At a pixel (y, x), a threshold T(y, x) is determined by: $T(y, x) = \mu(y, x) C$, where $\mu(y, x)$ is the mean of the neighborhood and C is the constant
 - Performs better in binarizing an image having different lighting conditions

```
import cv2 as cv
    img = cv.imread('sudoku.jpg', cv.IMREAD GRAYSCALE)
    if img is None:
        print("file not found")
    ret, thresh1 = cv.threshold(img, 100, 255, cv.THRESH_BINARY)
    ret, thresh2 = cv.threshold(img, 50, 255, cv. THRESH BINARY)
    thresh3 = cv.adaptiveThreshold(img, 255,
10
                                    cv.ADAPTIVE_THRESH_MEAN_C,
11
12
                                    cv.THRESH_BINARY, 11, 2)
13
    thresh4 = cv.adaptiveThreshold(img, 255,
14
                                    CV. ADAPTIVE THRESH GAUSSIAN C.
15
                                    cv. THRESH BINARY, 11, 2)
16
    titles = ['Original Image', 'Binary100', 'Binary50',
               'Adaptive mean', 'Adaptive Gaussian']
18
    images = [img, thresh1, thresh2, thresh3, thresh4]
19
20
21
    for i in range(5):
        cv.imshow(titles[i], images[i])
    cv.waitKey()
    cv.destroyAllWindows()
```



Original Image



Global Thresholding



Adaptive Thresholding

```
import cv2 as cv
    img = cv.imread('sudoku.jpg', cv.IMREAD GRAYSCALE)
    if img is None:
        print("file not found")
    ret, thresh1 = cv.threshold(img, 100, 255, cv.THRESH_BINARY)
    ret, thresh2 = cv.threshold(img, 50, 255, cv. THRESH BINARY)
    thresh3 = cv.adaptiveThreshold(img, 255,
10
11
                                     cv.ADAPTIVE_THRESH_MEAN_C,
12
                                    cv.THRESH_BINARY, 11, 2)
13
    thresh4 = cv.adaptiveThreshold(img, 255,
14
                                     CV. ADAPTIVE THRESH GAUSSIAN C,
                                    cv. THRESH BINARY, 11, 2)
15
16
    titles = ['Original Image', 'Binary100', 'Binary50',
               'Adaptive mean', 'Adaptive Gaussian']
18
    images = [img, thresh1, thresh2, thresh3, thresh4]
19
20
21
    for i in range(5):
        cv.imshow(titles[i], images[i])
    cv.waitKey()
    cv.destroyAllWindows()
```

- (L10-L13) cv.adaptiveThreshold(): apply adaptive thresholding
- cv.ADAPTIVE_THRESH_MEAN_C: the threshold value is the mean of neighborhood minus constant C ($T(y,x) = \mu(y,x) C$)
- cv.ADAPTIVE_THRESH_GAUSSIAN_C: the threshold value is the Gaussian-weighted sum of neighborhood minus constant C

 ${\it cv.} adaptive Threshold (\, {\it src, maxValue, adaptive Method, threshold Type, block Size, C[, dst]}\,) > dst$

```
src Source 8-bit single-channel image.

dst Destination image of the same size and the same type as src.

maxValue Non-zero value assigned to the pixels for which the condition is satisfied

adaptiveMethod Adaptive thresholding algorithm to use, see AdaptiveThresholdTypes. The BORDER_REPLICATE | BORDER_ISOLATED is used to process boundaries.

thresholdType Thresholding type that must be either THRESH_BINARY or THRESH_BINARY_INV, see ThresholdTypes.

blockSize Size of a pixel neighborhood that is used to calculate a threshold value for the pixel: 3, 5, 7, and so on.

C Constant subtracted from the mean or weighted mean (see the details below). Normally, it is positive but may be zero or negative as well.
```

Example 5-3. Otsu Binarization Algorithm

 In global thresholding, find the optimal threshold T that minimizes the weighted within-class variance:

$$T = \underset{t \in \{0,1,\dots,L-1\}}{\operatorname{argmin}} v_{within}(t)$$

, where
$$v_{within}(t) = w_0(t)v_0(t) + w_1(t)v_1(t)$$
,

$$w_0(t) = \sum_{i=0}^t \hat{h}(i),$$
 $w_1(t) = \sum_{i=t+1}^{L-1} \hat{h}(i),$

$$\mu_0(t) = \frac{1}{w_0(t)} \sum_{i=0}^t i\hat{h}(i), \qquad \qquad \mu_1(t) = \frac{1}{w_1(t)} \sum_{i=t+1}^{L-1} i\hat{h}(i),$$

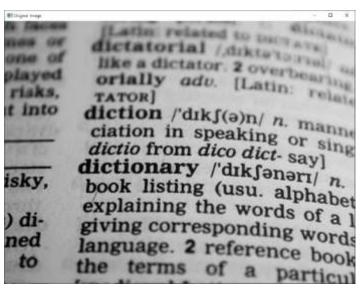
$$v_0(t) = \frac{1}{w_0(t)} \sum_{i=0}^t \hat{h}(i) (i - \mu_0(t))^2, \quad v_1(t) = \frac{1}{w_1(t)} \sum_{i=t+1}^{L-1} \hat{h}(i) (i - \mu_1(t))^2$$

Example 5-3. Otsu Binarization Algorithm

 Otsu binarization: use a flag cv.THRESH_OTSU in cv.threshold()

```
import cv2 as cv
       img = cv.imread('diction.jpg', cv.IMREAD_GRAYSCALE)
       if img is None:
           print("file not found")
       ret,thresh1 = cv.threshold(img,128,255,cv.THRESH BINARY)
       ret,thresh2 = cv.threshold(img,0, 255,cv.THRESH BINARY+cv.THRESH OTSU)
       print('Otsu Threshold',ret)
10
11
       titles = ['Original Image', 'Binary128', 'Otsu']
12
       images = [img, thresh1, thresh2]
13
14
15
       for i in range(3):
16
           cv.imshow(titles[i], images[i])
17
       cv.waitKey()
       cv.destroyAllWindows()
18
```

Original image



Binarized image with Otsu Algorithm (Optimal threshold = 125.0)

```
ciation in speaking or sing diction ary /diksənəri/ n. book listing (usu. alphabet explaining the words of a language. 2 reference book the terms of a particul
```

ReviewTask 0321

(Integral Image) Let's implement a 5x5 box filtering using integral image

```
import cv2 as cv
       import numpy as np
       img = cv.imread('lenna.png', cv.IMREAD_GRAYSCALE)
       if img is None:
           print("file not found")
       bImg = cv.blur(img, (5,5))
10
       sumimg = cv.integral(img)
       bImg2 = np.zeros((img.shape[0], img.shape[1]))
12
       # write filtering with a 5x5 using suming Write code here!
13
14
15
       titles = ['Original Image', 'Blurred', 'With IntegralImg']
16
       images = [img, bImg, bImg2]
17
18
       for i in range(3):
19
20
           cv.imshow(titles[i], images[i])
21
       cv.waitKey()
       cv.destroyAllWindows()
```

ReviewTask 0321

(Otsu Binarization) Binarize the following images using Otsu Algorithm

And print out the optimal threshold values for each image



(soccer.jpg – red channel only)



(rose.png – compare the binarization results of gray-scaled image and red channel only-image)