Chapter 3 Basic Image Filtering Operations

Image Denoising

- 1. Box filter
- 2. Gaussian filter
- 3. Median filter
- 4. Bilateral filter

Image Enhancement

1. Sharp-unsharp masking

```
In [1]: import cv2
import matplotlib.pyplot as plt
%matplotlib inline

image = cv2.imread('./test_data/lenna.png')
plt.imshow(image[:,:,[2,1,0]])
plt.title('original image')
plt.axis('off')
plt.show()
```

original image



Box filtering

```
In [2]: plt.figure(figsize=(16,4))
    plt.subplot(1,5,1)
    plt.title('original')
    plt.axis('off')
    plt.imshow(image[:,:,[2,1,0]])
    for i, k in enumerate([3,5,7,9]):
        result = cv2.blur(image,(k,k))
        plt.subplot(1,5,i+2)
        plt.title('kernel size='+str(k))
        plt.axis('off')
        plt.imshow(result[:,:,[2,1,0]])
    plt.show()
```











Gaussian filtering

```
In [3]: import numpy as np
        noisy_image_3 = (image.astype(np.float) + np.random.randn(*image.shape)*3).astype(dtype=np.uint8)
        noisy image 9 = (image.astype(np.float) + np.random.randn(*image.shape)*9).astype(dtype=np.uint8)
        plt.figure(figsize=(16,8))
        plt.subplot(2,5,1)
        plt.title('noisy image ($\sigma=3$)')
        plt.axis('off')
        plt.imshow(noisy image 3[:,:,[2,1,0]])
        plt.subplot(2,5,6)
        plt.title('noisy image ($\sigma=9$)')
        plt.axis('off')
        plt.imshow(noisy image 9[:,:,[2,1,0]])
        for i, sigma in enumerate([3,5,7,9]):
            result = cv2.GaussianBlur(noisy image 3,(0,0),sigma) # (0,0) means that the kernel size is determined by sigma x and s
            plt.subplot(2,5,i+2)
            plt.title('$\sigma$='+str(sigma))
            plt.axis('off')
            plt.imshow(result[:,:,[2,1,0]])
            result = cv2. Gaussian Blur (noisy image 9, (0,0), sigma) # (0,0) means that the kernel size is determined by sigma x and s
            plt.subplot(2,5,i+7)
            plt.title('$\sigma$='+str(sigma))
            plt.axis('off')
            plt.imshow(result[:,:,[2,1,0]])
        plt.show()
```



Median filtering

```
In [4]: import numpy as np
        noise = np.random.randint(0,80,image.shape[0:2])
        noise1= np.nonzero(noise >= 70)
        noise0= np.nonzero(noise < 10)</pre>
        image salt and pepper = image.copy()
        image salt and pepper[noise1[0],noise1[1],:] = (255,255,255)
        image salt and pepper[noise0[0],noise0[1],:] = (0,0,0)
        plt.figure(figsize=(16,8))
        plt.subplot(2,5,1)
        plt.title('original')
        plt.axis('off')
        plt.imshow(image salt and pepper[:,:,[2,1,0]])
        for i, k in enumerate([3,5,7,9]):
            result = cv2.medianBlur(image salt and pepper,k) # (0,0) means that the kernel size is determined by sigma x and sigma
            plt.subplot(2,5,i+2)
            plt.title('kernel size='+str(k))
            plt.axis('off')
            plt.imshow(result[:,:,[2,1,0]])
            result = cv2. Gaussian Blur (image salt and pepper, (0,0), k) # (0,0) means that the kernel size is determined by sigma x d
            plt.subplot(2,5,7+i)
            plt.title('GaussianBlur $\sigma$='+str(k))
            plt.axis('off')
            plt.imshow(result[:,:,[2,1,0]])
        plt.show()
```



Bilateral filtering

```
In [5]: plt.figure(figsize=(16,4))
         plt.subplot(1,5,1)
         plt.title('original')
         plt.axis('off')
         plt.imshow(image[:,:,[2,1,0]])
         for i, (sigmaColor, sigmaSpace) in enumerate(zip([8,16,32,64],[3,7,11,15])):
              result = cv2.bilateralFilter(image, -1, sigmaColor, sigmaSpace) #d=-1, kernel size is determined by sigmaSpace
              plt.subplot(1,5,i+2)
              plt.title('$\sigma {C}$='+str(sigmaColor)+',$\sigma {S}$='+str(sigmaSpace))
              plt.axis('off')
              plt.imshow(result[:,:,[2,1,0]])
         plt.show()
                                                                                                        \sigma_C = 32, \sigma_S = 11
                                                                                                                                   \sigma_C = 64, \sigma_S = 15
                                                   \sigma_C = 8, \sigma_S = 3
                                                                              \sigma_C = 16, \sigma_S = 7
                         original
```











Image enhancement by sharp-unsharp masking

```
In [6]: image = cv2.imread('./test data/lenna.png')
        plt.figure(figsize=(14,7))
        plt.subplot(2,4,1)
        plt.imshow(image[:,:,[2,1,0]])
        plt.axis('off')
        plt.title('original')
        for sidx,sigma in enumerate([3,9]):
            delta img=image.astype(np.int) - cv2.GaussianBlur(image,(0,0),sigma).astype(np.int)
            for idx,k in enumerate([1,3,5]):
                result = image.astype(np.int) + k*delta img
                result = np.where(result < 0,0,result) #result[result<0] = 0</pre>
                result = np.where(result > 255,255,result) #result[result>255]=255
                enhanced = result.astype(np.uint8) #enhanced = cv2.convertScaleAbs(result)
                plt.subplot(2,4,idx+2+sidx*4)
                plt.axis('off')
                plt.imshow(enhanced[:,:,[2,1,0]])
                plt.title('k={},$\sigma={}$'.format(k,sigma))
        plt.show()
```



Image Interpolation

Flag

nearest neighbor interpolation	cv2.INTER_NEAREST
bilinear interpolation	cv2.INTER_LINEAR
bicubic interpolation	cv2.INTER_CUBIC
resampling using pixel area relation. It may be a preferred method for image decimation, as it gives moire'-free results. But when the image is zoomed, it is similar to the INTER_NEAREST method.	cv2.INTER_AREA
Lanczos interpolation over 8x8 neighborhood	cv2.INTER_LANCZOS4
Bit exact bilinear interpolation	cv2.INTER_LINEAR_EXACT

Description

```
In [7]: import numpy as np
        import matplotlib.pyplot as plt
        import cv2
        image = cv2.imread('./test data/lena-s.png')
        plt.figure(figsize=(14,7))
        plt.subplot(2,4,1)
        plt.imshow(image[:,:,[2,1,0]])
        plt.axis('off')
        plt.title('orginal')
        for idx,interp in enumerate([(cv2.INTER_NEAREST, 'nearest'), (cv2.INTER_LINEAR, 'bilinear'), (cv2.INTER_CUBIC, 'cubic'), (cv2.IN
            dst = cv2.resize(image, None, fx=3.3, fy=3.3, interpolation = interp[0])
            plt.subplot(2,4,idx+2)
            plt.imshow(dst[:,:,[2,1,0]])
            plt.axis('off')
            plt.title(interp[1])
        plt.tight layout()
        plt.show()
```

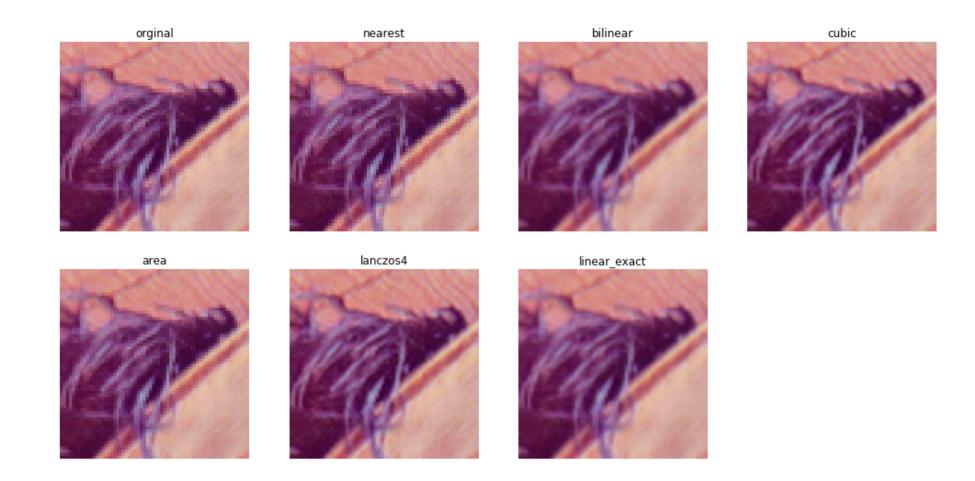
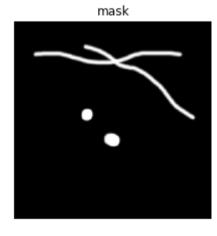


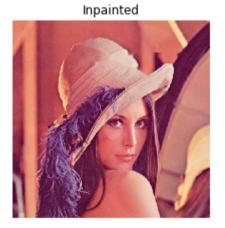
Image Inpainting

```
▶ In [8]: import numpy as np
           import matplotlib.pyplot as plt
           import cv2
           image = cv2.imread('./test_data/lenna.png')
           mask = cv2.imread('./test_data/mask.png',cv2.IMREAD_GRAYSCALE)
           dst = cv2.inpaint(image,mask,3,cv2.INPAINT TELEA)
           plt.figure(figsize=(12,4))
           plt.subplot(1,3,1)
           plt.imshow(image[:,:,[2,1,0]])
           plt.axis('off')
           plt.title('original')
           plt.subplot(1,3,2)
           plt.imshow(mask,cmap='gray')
           plt.axis('off')
           plt.title('mask')
           plt.subplot(1,3,3)
           plt.imshow(dst[:,:,[2,1,0]])
           plt.axis('off')
           plt.title('Inpainted')
```

Out[8]: Text(0.5, 1.0, 'Inpainted')







In []:	:		