

assignment3

September 9, 2018

1 Machine Learning and Computer Vision

1.1 Assignment 3

This assignment contains 2 programming exercises.

1.2 Problem 1: Order-statistic filtering

Order-statistic filters (OSF) are local filters that are only based on the ranking of pixel values inside a sliding window. 1. Create in `imstack(img,s1,s2)` function that creates a stack `xstack` of size `n1` \times `n2` \times `s`, which $s = (2s_1 + 1)(2s_2 + 1)$ from the `n1` \times `n2` image `x`, such that `xstack(i,j,:)` contains all the values of `x` in the neighborhood $(s_1, s_1) \times (s_2, s_2)$. This function should take into account the four possible boundary conditions.

Hint: you can use `imshift`, which we implemented in assignment 1, and only two loops for $s_1 \leq 1$

2. Create in `imosf()` function `imosf(x, type, s1, s2)` that implements order-statistic filters, returns `xosf`. `imosf` should first call `imstack`, next sort the entries of the stack with respect to the third dimension, and create the suitable output `xosf` according to the string `type` as follows:
 - 'median': select the median value,
 - 'erode': select the min value,
 - 'dilate': select the max value,
 - 'trimmed': take the mean after excluding at least 25% of the extreme values on each side.
3. Create in `imopening()` and `imclosing()` function that performs the opening and closing by the means of OSF filters.
4. Load `castle.png`. Write a script to test `imosf()` that loads the image `x = castle` and create a corrupted version of image `x` with 10% of impulse noise (salt and pepper)

Apply your OSF filters and zoom on the results to check that your results are consistent with the following ones:

```

In [1]: import random
import numpy as np
from scipy.misc import imread
import matplotlib.pyplot as plt

def imstack(img, s1, s2):
    n1, n2 = img.shape
    s = (2*s1+1)*(2*s2+1)
    xstack = np.empty((n1, n2, 0))
    for k in range(-s1,s1+1):
        for l in range(-s2,s2+1):
            xshift = np.expand_dims(imshift(img, -k, -l), axis=-1)
            xstack = np.append(xshift,xstack, axis=-1)
    return xstack

def imshift(x, k, l):
    h,w = x.shape
    xshifted = np.zeros((h,w))
    if k == 0:
        k = h
    if l == 0:
        l = w
    xshifted[-k:,-1:] = x[:k,:l]
    xshifted[:-k,:-l] = x[k:,l:]
    xshifted[:-k,-1:] = x[k:,:l]
    xshifted[-k,:-l] = x[:k,l:]
    return xshifted

def sort(simg):
    n1, n2 = simg.shape
    for h in range(n1):
        for w in range(n2):
            simg[h, w].sort()
    return simg

def imosf(x, typ, s1, s2):
    n1, n2 = x.shape
    xosf = np.zeros((n1, n2))
    xstack = imstack(x, s1, s2)
    if typ == "median":
        xosf = np.median(xstack, axis=-1)
    elif typ == "erode":
        xosf = np.min(xstack, axis=-1)
    elif typ == "dilate":
        xosf = np.max(xstack, axis=-1)

```

```

elif typ == "trimmed":
    xosf = np.sort(xstack, axis=-1)
    s = 2*(s1+1)*2*(s2+1)
    xosf = np.mean(xosf[:, :, int(s*0.25):int(s*0.75)], axis=-1)
elif typ == "close":
    tem = np.max(xstack, axis=-1)
    tem = imstack(tem, s1, s2)
    xosf = np.min(tem, axis=-1)
elif typ == "open":
    tem = np.min(xstack, axis=-1)
    tem = imstack(tem, s1, s2)
    xosf = np.max(tem, axis=-1)
return xosf

def pepper(x, rate):
    n1, n2 = x.shape
    noise_p = int(n1*n2*rate)
    for r in range(noise_p):
        x[random.randint(0, n1-1), random.randint(0, n2-1)] = random.choice([0, 255])
    return x

s1 = 2
s2 = 2
img = imread('castle.png')
img_noise = pepper(img, 0.05)

plt.subplot(1, 5, 1)
plt.title("noise")
plt.axis('off')
plt.imshow(img_noise, cmap = plt.get_cmap('gray'))

img1 = imosf(img_noise, "median", s1, s2)
plt.subplot(1, 5, 2)
plt.title("median")
plt.axis('off')
plt.imshow(img1, cmap = plt.get_cmap('gray'))

img2 = imosf(img_noise, "trimmed", s1, s2)
plt.subplot(1, 5, 3)
plt.title("trimmed")
plt.axis('off')
plt.imshow(img2, cmap = plt.get_cmap('gray'))

img11 = imosf(img_noise, "close", s1, s2)
plt.subplot(1, 5, 4)

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plt.title("closing")
plt.axis("off")
plt.imshow(img11 ,cmap = plt.get_cmap('gray'))

img22 = imosf(img_noise,"open",s1,s2)
plt.subplot(1, 5, 5)
plt.title("opening")
plt.axis("off")
plt.imshow(img22 ,cmap = plt.get_cmap('gray'))

plt.show()

```

C:\study\anaconda\lib\site-packages\ipykernel_launcher.py:74: DeprecationWarning: `imread` is deprecated in SciPy 1.0.0, and will be removed in 1.2.0. Use ``imageio.imread`` instead.

<matplotlib.figure.Figure at 0x1eeddb6d2b0>

1.3 Problem 2: Bilateral filter

Now, we will discuss a non-local filter, Bilateral filter.

The bilateral filter is a denoising algorithm that reads as:

1. Create a test_imbilateral(img, sigma) function that loads the image x = castle and adds additive white Gaussian noise of standard deviation = 10
2. Create in imbilateral_naive(img, sigma, s1, s2, h), a function that implements the bilateral filter (except around boundaries) with four loops.
3. test your function on y with s1 = s2 = 10 and h = 1. Zoom on the results to check that your functions are consistent with the following ones:
4. Create function imbilateral(y, sigma, s1, s2, h) that implements the bilateral filter including around boundaries. The idea is again to switch the k, l loops with the i, j loops, and then make use of imshift. The final code should read with only two loops and deal with boundary conditions.
5. Compare the computation times.
6. Increase the noise level, and play with the search window sizes s1 and s2 and filtering parameter h.

```

In [2]: import random
import numpy as np
from scipy.misc import imread
import matplotlib.pyplot as plt
import time

```

```

def GaussianNoise(x, means, sigma):
    n1, n2 = x.shape
    noise_img = np.zeros((n1,n2))
    for i in range(n1):
        for j in range(n2):
            noise_img[i, j] = x[i, j] + random.gauss(means, sigma)
            if noise_img[i, j] < 0:
                noise_img[i, j] = 0
            elif noise_img[i, j] > 255:
                noise_img[i, j] = 255
    return noise_img

def test_imbilateral():
    sigma = 10
    H = 1
    s1 = s2 = 10
    img = imread('castle.png')

    plt.subplot(1, 3, 1)
    plt.title("original")
    plt.axis('off')
    plt.imshow(img, cmap=plt.get_cmap('gray'))

    img_noise = GaussianNoise(img, 0, 10)
    plt.subplot(1, 3, 2)
    plt.title("noise")
    plt.axis('off')
    plt.imshow(img_noise, cmap=plt.get_cmap('gray'))

    t0 = time.time()
    img1 = imbilateral(img_noise, s1, s2, sigma, H)
    print("time cost:", time.time()-t0)
    plt.subplot(1, 3, 3)
    plt.title("bilateral")
    plt.axis('off')
    plt.imshow(img1, cmap=plt.get_cmap('gray'))
    plt.show()

def imbialteral_navie(img, s1, s2, sigma, H):
    n1, n2 = img.shape
    phi = lambda a: np.exp(-(max(a - 2 * sigma ** 2, 0) / (16 * H * sigma ** 2)))
    z = np.zeros((n1, n2))
    x = np.zeros((n1, n2))
    for i in range(n1):

```

```

        for j in range(n2):
            for k in range(-s1, s1 + 1):
                for l in range(-s2, s2 + 1):
                    if (i+k)>(n1-1) or (j+l)>(n2-1):
                        continue
                    x[i, j] += phi((img[(i + k), (j + 1)] - img[i, j]) ** 2) * img[(i + k), (j + 1)]
                    z[i, j] += phi((img[(i + k), (j + 1)] - img[i, j]) ** 2)

x = x / z
return x

def imbilateral(img, s1, s2, sigma, H):
    n1, n2 = img.shape
    phi = lambda a: np.exp(-(np.maximum(a - 2 * sigma ** 2, 0) / (16 * H * sigma ** 2)))
    z = np.zeros((n1, n2))
    x = np.zeros((n1, n2))
    for k in range(-s1, s1 + 1):
        for l in range(-s2, s2 + 1):
            xshift = imshift(img, -k, -l)
            x += phi((xshift - img) ** 2) * xshift
            z += phi((xshift - img) ** 2)
    x = x / z
    return x

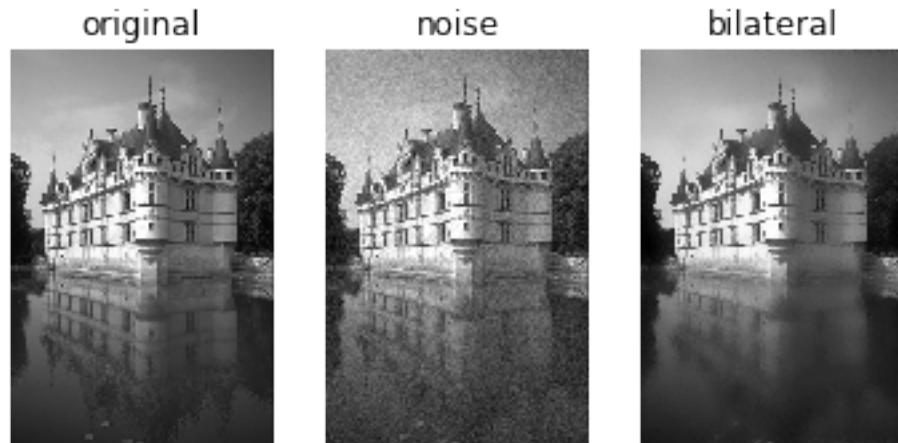
def imshift(x, k, l):
    h, w = x.shape
    xshifted = np.zeros((h, w))
    if k == 0:
        k = h
    if l == 0:
        l = w
    xshifted[-k:, -l:] = x[:k, :l]
    xshifted[:-k, :-l] = x[k:, l:]
    xshifted[:-k, -l:] = x[k:, :l]
    xshifted[-k:, :-l] = x[:k, l:]
    return xshifted

```

```
test_imbilateral()
```

C:\study\anaconda\lib\site-packages\ipykernel_launcher.py:25: DeprecationWarning: `imread` is deprecated in SciPy 1.0.0, and will be removed in 1.2.0.
Use ``imageio.imread`` instead.

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1.4 Conclusion

Have you accomplished all parts of your assignment? What concepts did you use or learn in this assignment? What difficulties have you encountered? Explain your result for each section. Please write one or two short paragraphs in the below Markdown window (double click to edit).

**** Your Conclusion: ****

-- For this time, I learned many things, such as OSF and some other filter such as Bilateral filter. The first time I just didn't know how to deal with the matrix transmission. Then I searched Google and finally found some useful methods in np.

**** Submission Instructions****

Remember to submit your PDF version of this notebook to Gradescope. You can find the export option at File → Download as → PDF via LaTeX