

# 2D\_Gaussian\_filter\_test

February 28, 2019

## 0.1 ASSIGNMENT 1

### 0.2 CS5187 VISION AND IMAGE

You are required to complete both written and programming tasks in this assignment.

#### 0.2.1 PART-1: PROOF

A. Show that a 2D Gaussian filter is separable into two 1D Gaussian filters. (5%)

## 0.3 ASSIGNMENT 1

### 0.4 CS5187 VISION AND IMAGE

You are required to complete both written and programming tasks in this assignment. ### PART-1: PROOF A. Show that a 2D Gaussian filter is separable into two 1D Gaussian filters. (5%) B. Derive the 1st derivative of 2D Gaussian filter. (5%) C. Derive the 2nd derivative of 2D Gaussian filter. (5%) D. Derive the Laplacian of Gaussian (LoG) filter.

```
In [2]: import cv2
import numpy as np
import matplotlib.pyplot as plt
import scipy.stats as st
import math
from scipy import ndimage
from skimage import io
import os
import heapq
import scipy
from scipy.misc import imread
import cPickle as pickle
import matplotlib.pyplot as plt
```

```
In [6]: imgPath = 'Assignment1_data/data/0000.jpg'
```

```
In [254]: def showImg(row, col, title, kernel, index):
plt.subplot(row, col, index)
#plt.subplots_adjust(left=0.0, bottom=0.0, top=1, right=1)
#plt.subplots_adjust(wspace=0, hspace=0)#
```

```

plt.subplots_adjust(top = 1, bottom = 0, right = 1, left = 0, hspace = 0, wspace = 0)
plt.margins(0,0)
plt.imshow(kernel, interpolation='none')
plt.title(title)
plt.xticks([])
plt.yticks([])

```

1D Gaussian filters:

```

In [29]: img = cv2.imread(imgPath)
        blur = cv2.GaussianBlur(img, (5, 5), 0)
        blur1 = cv2.GaussianBlur(blur, (5, 5), 0)
        plt.figure(figsize=(8, 8))
        showImg(1, 2, 'Original', img, 1)
        showImg(1, 2, 'Blurred twice', blur1, 2)
        plt.show()

```

Original



Blurred twice



2D Gaussian filters:

```

In [30]: img = cv2.imread(imgPath)
        kernel = np.ones((5, 5), np.float32)/25
        dst = cv2.filter2D(img, -1, kernel)
        plt.figure(figsize=(8, 8))
        showImg(1, 2, 'Original', img, 1)
        showImg(1, 2, 'Averaging', dst, 2)
        plt.show()

```

Original



Averaging



As you can see, the results are almost the same. These two methods perform consistently.

**B. Derive the 1st derivative of 2D Gaussian filter. (5%)**

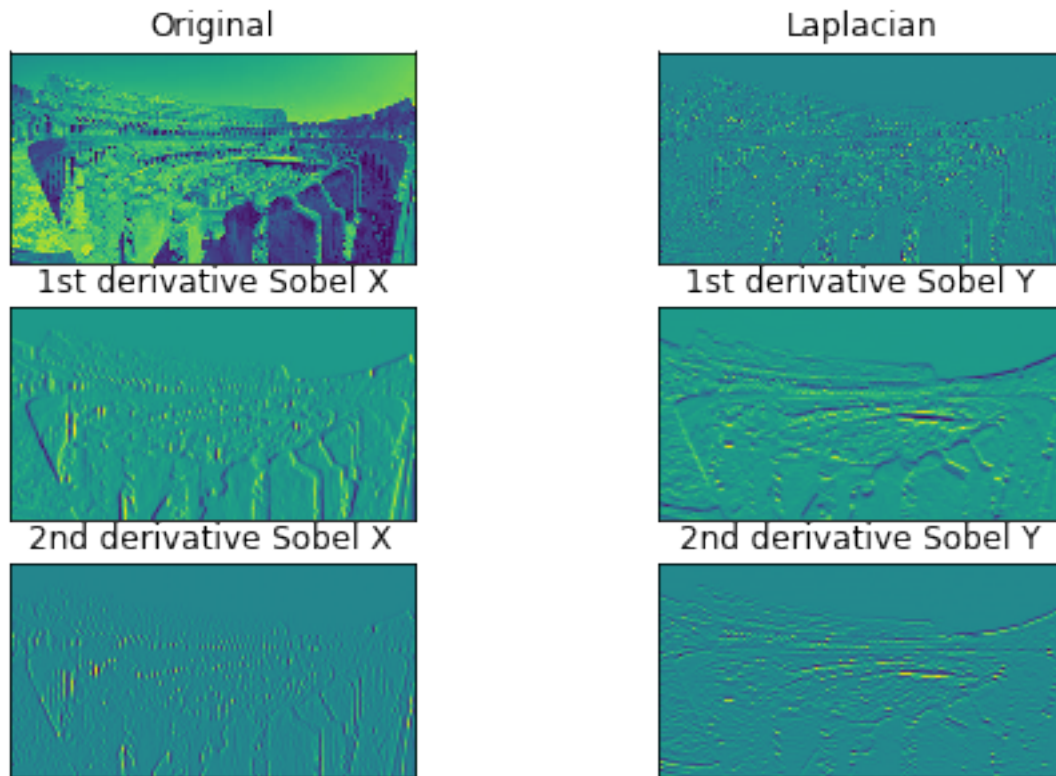
**C. Derive the 2nd derivative of 2D Gaussian filter. (5%)**

**D. Derive the Laplacian of Gaussian (LoG) filter. (5%)**

```
In [31]: img = cv2.imread(imgPath, 0)

laplacian = cv2.Laplacian(img, cv2.CV_64F)
sobelx_1 = cv2.Sobel(img, cv2.CV_64F, 1, 0, ksize=5)
sobely_1 = cv2.Sobel(img, cv2.CV_64F, 0, 1, ksize=5)
sobelx_2 = cv2.Sobel(img, cv2.CV_64F, 2, 0, ksize=5)
sobely_2 = cv2.Sobel(img, cv2.CV_64F, 0, 2, ksize=5)
plt.figure(figsize=(8, 5))
showImg(3, 2, 'Original', img, 1)
showImg(3, 2, 'Laplacian', laplacian, 2)
showImg(3, 2, '1st derivative Sobel X', sobelx_1, 3)
showImg(3, 2, '1st derivative Sobel Y', sobely_1, 4)
showImg(3, 2, '2nd derivative Sobel X', sobelx_2, 5)
showImg(3, 2, '2nd derivative Sobel Y', sobely_2, 6)

plt.show()
```



#### 0.4.1 PART-2: CONVOLUTION

Following below to generate a filter bank of 48 image filters, each with size  $32 \times 32$  pixels: - 4 Gaussian filters with  $\sigma = \{1, 2, 2, 22\}$ . - 8 LOG (Laplacian of Gaussian) filters with  $\sigma = \{2, 2, 22, 4, 32, 6, 62, 12\}$ . - 18 x-directional first derivation of Gaussian filters with  $\sigma =$  and  $\sigma = 3$  in three different scales  $\sigma = \{1, 2, 2\}$  and six rotation orientations  $\theta = \{0, 1, 1, 1, 41, 51\}$ . 23432 - 18 x-directional second derivation of Gaussian filters with  $\sigma =$  and  $\sigma = 3$  in three different scales  $\sigma = \{1, 2, 2\}$  and six rotation orientations  $\theta = \{0, 1, 1, 1, 41, 51\}$ . 23432

##### A. Display the 48 image filters in the report. (5%)

```
In [4]: # generate Gaussian Kernel
def gGaussianKernel(sig):
    getGakernel = cv2.getGaussianKernel(ksize=32, sigma=sig)
    kernel = getGakernel * getGakernel.T
    return kernel

dst = cv2.Sobel(src, ddepth, dx, dy[, dst[, ksize[, scale[, delta[, borderType]]]])

In [5]: # generate LOG Kernel
def gLOGKernel(sig):
    kernel = gGaussianKernel(sig)
```

```

    d1x = np.gradient(kernel, axis=0)
    d2x = np.gradient(d1x, axis=0)
    d1y = np.gradient(kernel, axis=1)
    d2y = np.gradient(d1y, axis=1)
    dst = d2x + d2y
    return dst

In [6]: # generate rotated Kernel
def gRotation(kernel, angle):
    M = cv2.getRotationMatrix2D((16, 16), angle, 1)
    dst = cv2.warpAffine(kernel, M, (32, 32))
    return dst

In [7]: # generate derivation Kernel
def gDeriKernel(xsig, ysig, angle, derivation):
    getxGakernel = cv2.getGaussianKernel(ksize=32, sigma=xsig)
    getyGakernel = cv2.getGaussianKernel(ksize=32, sigma=ysig)
    kernel = getyGakernel * getxGakernel.T
    sobel = np.gradient(kernel, axis=1)
    if derivation == 2:
        sobel = np.gradient(sobel, axis=1)
    dst = gRotation(sobel, angle)
    return dst

In [253]: kernel_data = []
G_sigs = [1, math.sqrt(2), 2, 2 * math.sqrt(2)]
i = 0
for sig in G_sigs:
    i = i + 1
    result = gGaussianKernel(sig)
    kernel_data.append(result)
    showImg1(6, 8, '', result, i)

LOG_sigs = [np.sqrt(2), 2, np.sqrt(8), 4, np.sqrt(18), 6, np.sqrt(72), 12]
for sig in LOG_sigs:
    i = i + 1
    result = gLOGKernel(sig)
    kernel_data.append(result)
    showImg1(6, 8, '', result, i)

x = [1, math.sqrt(2), 2]
y = [3, 3 * math.sqrt(2), 6]
pi = 180
angles = [0, pi/6, pi/3, pi/2, 2*pi/3, 5*pi/6]

for (xsig, ysig) in zip(x, y):
    for angle in angles:
        i = i + 1

```

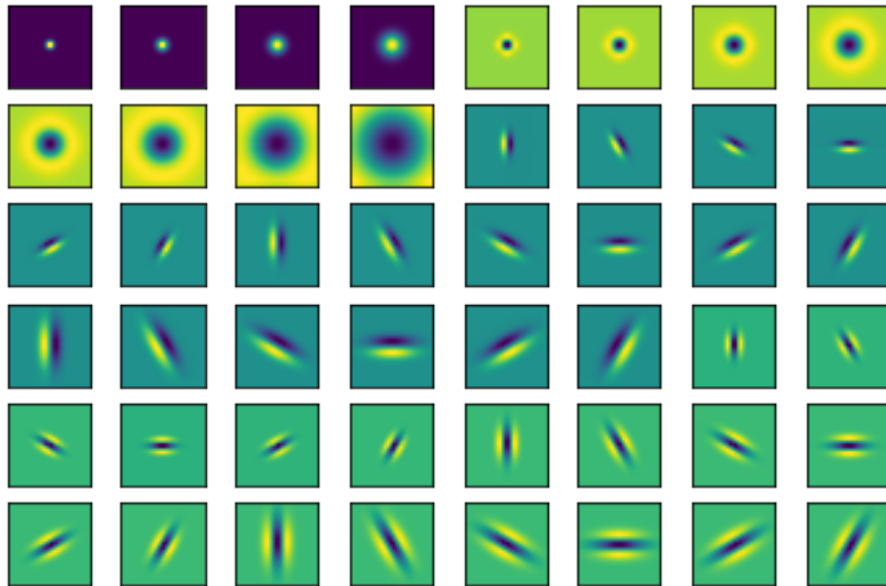
```

        result = gDeriKernel(xsig, ysig, angle, 1)
        kernel_data.append(result)
        showImg1(6, 8, '', result, i)

for (xsig, ysig) in zip(x, y):
    for angle in angles:
        i = i + 1
        result = gDeriKernel(xsig, ysig, angle, 2)
        kernel_data.append(result)
        showImg1(6, 8, '', result, i)
print len(kernel_data)
plt.savefig("Assignment1_data/result/48filter.png",bbox_inches = 'tight')
plt.show()

```

48



**B. Display the 48 image responses of the images “leopard.jpg” and “panda.jpg” after performing convolution with the filter bank. (5%)**

```

In [2]: #image convolute with 48 filters, and display the result by figures
def convolution(img, row, col):
    grayImg = cv2.cvtColor(img, cv2.COLOR_BGR2GRAY)
    i = 0
    results = []
    for fil in kernel_data:
        i = i + 1

```

```

        res = cv2.filter2D(grayImg, -1, fil)
        results.append(res)
        showImg(row, col, '', res, i)
    return results

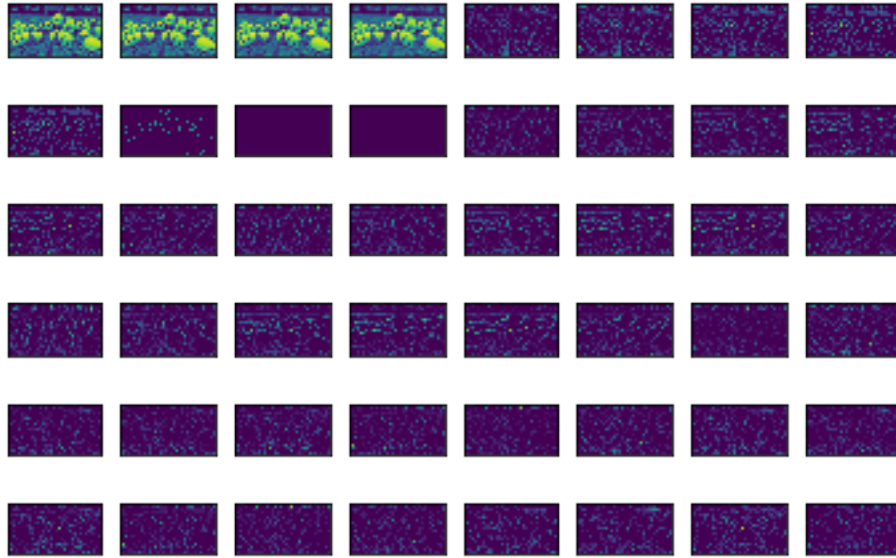
In [256]: def convolution1(img, row, col):
    i = 0
    results = []
    for fil in kernel_data:
        i = i + 1
        res = cv2.filter2D(img, -1, fil/1024)
        results.append(res)
        showImg(row, col, '', res, i)
    return results

In [3]: #image convolute with 48 filters but not to show them
def convolution_no_print(img):
    grayImg = cv2.cvtColor(img, cv2.COLOR_BGR2GRAY)
    results = []
    for fil in kernel_data:
        res = cv2.filter2D(grayImg, -1, fil)
        results.append(res)
    return results

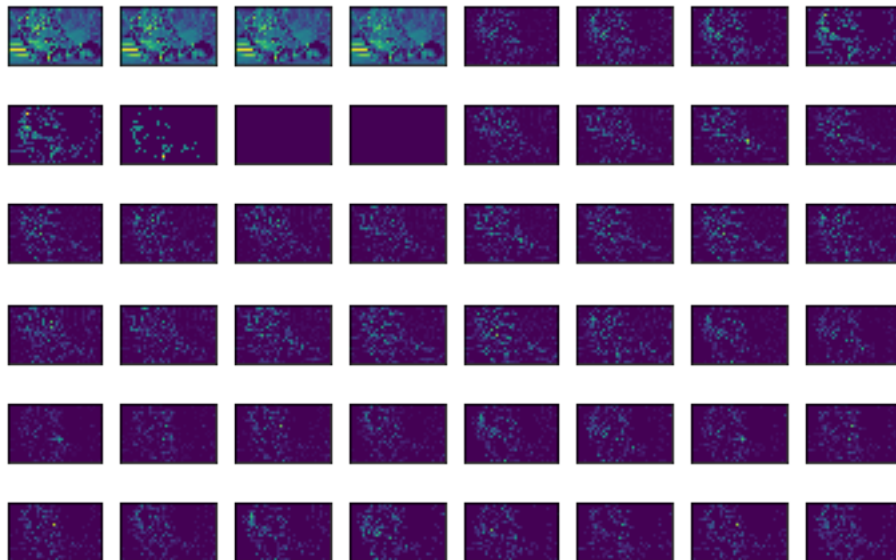
In [4]: #Calculate the mean and var
def meanAndstddev(kernels):
    verix = []
    i = 0
    for res in kernels:
        mean = np.mean(res)
        stddev = np.var(res)
        verix.insert(i, mean)
        verix.insert(i + 48, stddev)
        i = i + 1
    return verix

In [259]: img = cv2.imread("Assignment1_data/panda.jpg")
    pandaRes = convolution(img, 6, 8)
    plt.savefig("Assignment1_data/result/pandafilter.png", pad_inches = 0, bbox_inches = 'tight')
    plt.show()

```



```
In [260]: img = plt.imread("Assignment1_data/leopard.jpg")
leopardRes = convolution(img, 6, 8)
plt.savefig("Assignment1_data/result/leopardfilter.png",pad_inches = 0,bbox_inches =
plt.show()
```





**0.4.2 C. Compute the mean and variance of each image response to form a vector of length 96 elements. Write down the filter that gives the largest value of mean and the filter that gives the largest value of variance for “leopard.jpg” and “panda.jpg” in the report.**

(5%)

```
In [261]: pandaVer = meanAndstddev(pandaRes)
         leopardVer = meanAndstddev(leopardRes)
```

The filters that gives the largest value of mean and the filter that gives the largest value of variance for these two images are both the first filter which is gaussian filter with sigma = 1.

```
In [262]: pandaVerIndex = pandaVer.index(max(pandaVer[:48]))
         print pandaVerIndex+1
         pandaVerIndex = pandaVer.index(max(pandaVer[47:95]))
         print pandaVerIndex-47
         leopardVerIndex = leopardVer.index(max(leopardVer[:48]))
         print leopardVerIndex+1
         leopardVerIndex = leopardVer.index(max(leopardVer[47:95]))
         print leopardVerIndex-47
```

2  
1  
1  
1

## 0.5 PART-3: IMAGE RANKING

**0.5.1 You are given a collection of 2,000 images and 5 query images (in the canvas). You need to extract visual features from these images by performing convolution with the 48 filters Your task is: For each query, retrieve the five most similar images from the collection of 2,000 images. Show the five most similar images of each query in the report.**

```
In [137]: dataImages = []
         dataBeforeImg = []
         queryBeforeImg = []
         path_list = os.listdir("Assignment1_data/data")
         path_list.sort()
         i = 0
         for filename in path_list:
             img = cv2.imread('Assignment1_data/data/%s' %filename)
             dataBeforeImg.append(cv2.imread('Assignment1_data/data/%s' %filename))
             dataImages.append(cv2.resize(img, (200, 200)))
             i = i + 1
         queryImages = []
         path_list = os.listdir("Assignment1_data/query")
         path_list.sort()
         for filename in path_list:
```





## 0.6 PART-4: METHOD COMPARISON

0.6.1 A. Implement any two feature extraction methods that you know (e.g., color histogram, LBP, SIFT, deep learning) to extract features for 2,000 images in Part-3. Show the five most similar images of each query for each method. (10%)

Deep Learning:

```
In [19]: import sys
import matplotlib.pyplot as plt
sys.path.append("..") # Adds higher directory to python modules path.
from img_to_vec import Img2Vec
from PIL import Image
from sklearn.metrics.pairwise import cosine_similarity
```

```
input_path = 'Assignment1_data/query'
input1_path = 'Assignment1_data/data'
```

```
img2vec = Img2Vec()
```

```
In [20]: # For each test image, we store the filename and vector as key, value in a dictionary
pics = {}
for file in os.listdir(input_path):
    img = Image.open('Assignment1_data/query/%s'%file)
    vec = img2vec.get_vec(img)
    pics[file] = vec

datapics = {}
for file in os.listdir(input1_path):
    img = Image.open('Assignment1_data/data/%s'%file)
    vec = img2vec.get_vec(img)
    datapics[file] = vec
```

```
In [104]: import collections
```

```

In [127]: datapics = collections.OrderedDict(sorted(datapics.items()))
          pics = collections.OrderedDict(sorted(pics.items()))

In [128]: datapics.keys()
          pics.keys()

Out[128]: ['0000.jpg', '0001.jpg', '0002.jpg', '0003.jpg', '0004.jpg']

In [129]: dp_result = []
          j = 1
          #f = open("/Users/ponta/PycharmProjects/vision_1/Assignment1_data/result/result.txt")
          for query in list(pics.keys()):
              results = []
              #f.write('Q%d:' %j)
              i = 1
              showImg(1, 6, '', cv2.imread('Assignment1_data/query/%s'%query), i)
              sims = {}
              for data in list(datapics.keys()):
                  res = cosine_similarity(pics[query].reshape((1, -1)), datapics[data].reshape((1, -1)))
                  sims[data] = res
                  results.append(res)
              dp_result.append(results)
              d_view = [(v, k) for k, v in sims.items()]
              #dp_result = [(v, k) for k, v in sims.items()]
              d_view.sort(reverse=True)

              #for v, k in d_view:
              #    #f.write(k[:4])
              #    # f.write(' ')
              #f.write('\n')
              for v, k in d_view[:5]:
                  i = i + 1
                  showImg(1, 6, '', cv2.imread('Assignment1_data/data/%s'%k), i)
                  #print(v, k)
                  plt.savefig("Assignment1_data/result/deeplearning1Q%d.png"%j, pad_inches = 0, format='png')

              #i = i + 1
              j = j + 1
          plt.savefig("Assignment1_data/result/deeplearning.png")
          plt.show()

```



## Color histogram:

```
In [160]: import argparse
import glob
from scipy.spatial.distance import euclidean

In [177]: histdes = []
for data in dataBeforeImg:
    histdes.append(describe(data))

In [179]: queryhistdes = []
for data in queryBeforeImg:
    queryhistdes.append(describe(data))

In [207]: def feature_extraction(dataset, addstr):
    features = {}
    #descriptor = RGBHistogram(bins=[8, 8, 8])

    #for filename in glob.glob(os.path.join(dataset, '*.jpg/png$')):
    for filename in dataset:
        # e.g. places/eiffel_tower.jpg => eiffel_tower
        img_name = os.path.basename(filename).split('.')[0]
        #print img_name
        dataset = addstr + filename
        image = cv2.imread(dataset)
        feature = describe(image)
        # key - image name, value - feature vector
        features[img_name] = feature
    return features

In [208]: path_list = os.listdir("Assignment1_data/data")
path_list.sort()
addstr = 'Assignment1_data/data/'
datafeature = feature_extraction(path_list, addstr)

In [209]: path_list = os.listdir("Assignment1_data/query")
path_list.sort()
addstr = 'Assignment1_data/query/'
queryfeature = feature_extraction(path_list, addstr)

In [230]: def search():
    results = {}
    for queryname, querydes in queryfeature.items():
        result = {}
        for dataname, datades in datafeature.items():
            dist = euclidean(querydes, datades)
            result[dataname] = dist
        result = sorted([(d, n) for n, d in result.items()])
        results[queryname] = result
    return results
```

```
In [231]: hisresults = search()
```

```
In [238]: for queryname, diswithdata in hisresults.items():
           #print query
           i = 1
           showImg(1, 6, '', cv2.imread('Assignment1_data/query/%s.jpg'%queryname), i)
           for dis, name in diswithdata[:5]:
               i = i + 1
               print name
               showImg(1, 6, '', cv2.imread('Assignment1_data/data/%s.jpg'%name), i)
           plt.savefig("Assignment1_data/result/hisQ%s.png"%queryname, pad_inches=0,
                       bbox_inches='tight')
```

0967

0759

0127

1589

1516

0029

0364

0038

0702

1420

1308

1439

1230

0632

1393

1337

0923

0868

1883

0967

0721

0456

0601

0908

1264

**0.6.2 C. Propose a method to fuse (or combine) the results in Part-3 and Part-4. Show the five most similar images of each query.**

```
In [148]: def combiner():
           data = {}
           n = 0
           for (res1,res2) in zip(euclidean_res, dp_result):
```

```

sortRes1 = sorted(res1)
sortRes2 = sorted(res2, reverse=True)
for index in range(0, len(res2)):
    sum = 0.01 * sortRes1.index(res1[index]) + sortRes2.index(res2[index])
    data[index] = sum
data_view = [(v, k) for k, v in data.items()]
data_view.sort()
m = 1
showImg(1, 6, '', queryBeforeImg[n], m)
for v, k in data_view[:5]:
    m = m + 1
    print m
    showImg(1, 6, '', dataBeforeImg[k], m)
plt.savefig("Assignment1_data/result/CombineQ%d.png"%n, pad_inches=0,
            bbox_inches='tight')
n = n + 1

```

In [149]: len(dp\_result[0])

Out[149]: 2000

In [150]: combiner()

2  
3  
4  
5  
6  
2  
3  
4  
5  
6  
2  
3  
4  
5  
6  
2  
3  
4  
5  
6  
2  
3  
4  
5

6

```
In [ ]:
```