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1 Basic Test Results

```
Archive: /tmp/bodek.8PP2i9/impr/ex2/liran.drory/presubmission/submission
      inflating: current/answer_q2.txt
      inflating: current/answer_q3.txt
3
      inflating: current/README
4
      inflating: current/sol2.py
      inflating: current/answer_q1.txt
    /usr/lib/python3/dist-packages/matplotlib/font_manager.py:280: UserWarning: Matplotlib is building the font cache using fc-1
8
      'Matplotlib is building the font cache using fc-list.
9
    ex2 presubmission script
10
      Disclaimer
11
12
      The purpose of this script is to make sure that your code is compliant
      with the exercise API and some of the requirements
14
      The script does not test the quality of your results.
15
      Don't assume that passing this script will guarantee that you will get
16
      a high grade in the exercise
17
18
    login: liran.drory
19
20
21
    submitted files:
    sol2.py
22
23
24
    answer_q1.txt
25
26
    answer_q2.txt
27
    answer_q3.txt
28
29
30
31
32
    The difference of the magnitude happens according to the boundaries of the image
33
34
    In the Fourier is cyclic
35
36
37
    In the Convolution is not affected by the boundaries
    just the frame (outter)
38
39
    answer to q2:
    when the gaussian not in the center the results of the blur image according to the boundry of fourier
    will translate cyclic way, so that the image center will be where the center of the gaussian where.
41
42
43
    because the gaussian is not in the center each pixel will be calculated from different region
44
45
    answer to q3:
    the difference between both of the blurring is the boundry conditions due to
46
47
    fourier cyclic colculation (but it is very small)
    section 1.1
49
50
   DFT and IDFT
    section 1.2
51
    2D DFT and IDFT
52
53
    section 2.1
   derivative using convolution
54
55
   Section 2.2
   derivative using convolution
57 Section 3.1
58 blur spatial
   Section 3.1
```

```
60 blur fourier
61 all tests Passed.
62 - Pre-submission script done.
63
64 Please go over the output and verify that there are no failures/warnings.
65 Remember that this script tested only some basic technical aspects of your implementation
66 It is your responsibility to make sure your results are actually correct and not only
67 technically valid.
```

2 README

- liran.drory
 sol2.py
 answer_q1.txt
 answer_q2.txt
 answer_q3.txt

3 answer q1.txt

```
The difference of the magnitude happens according to the boundaries of the image

In the Fourier is cyclic

In the Convolution is not affected by the boundaries

just the frame (outter)
```

4 answer q2.txt

- when the gaussian not in the center the results of the blur image according to the boundry of fourier
- 2 will translate cyclic way, so that the image center will be where the center of the gaussian where.
- 4 because the gaussian is not in the center each pixel will be calculated from different region

5 answer q3.txt

- the difference between both of the blurring is the boundry conditions due to fourier cyclic colculation (but it is very small)

6 sol2.py

```
__author__ = "Liran Drory"
1
2
    import numpy as np
3
    from scipy.signal import convolve2d as conv
4
    import matplotlib.pylab as plt
   from scipy.misc import imread as imread
   from skimage.color import rgb2gray
    GRAYSCAL = 1
9
   GRAYSCAL\_MATRIX = 2
    RGB = 2
11
    RGB_MATRIX = 3
12
   HIEGTH = 0
    WIDTH = 1
14
   SHADES_OF_GRAY = 255
15
    COLORFULL = 3
17
18
19
    def read_image(filename, representation):
20
21
        Read Image and return matrix [0,1] float64
        Gray scale - 2D
22
        RGB - 3D
23
24
        Parameters
25
26
27
        :param filename: str
           string containing the image filename to read (PATH)
28
29
        :param representation: int
30
           either 1 or 2 defining whether the output
31
            should be a greyscale image (1) or an RGB image (2).
33
34
        Returns
35
        :return numpy array with either 2D matrix or 3D matrix
36
37
                describing the pixels of the image
38
39
40
        # loads the image
41
42
        im = imread(filename)
43
        if representation == RGB:
44
                                                 # pixels to float
45
            im_float = im.astype(np.float64)
            im_float /= 255
46
                                                 # pixels [0,1]
            return im_float
47
        if representation == GRAYSCAL:
49
                                                # pixels to float
50
            im_g = im.astype(np.float64)
            im_g = rgb2gray(im_g)
                                                   # turn to grey
51
            return im_g
52
53
54
    def DFT(signal):
55
56
        Function that return DFT of signal
57
        if matrix is input: return every row DFT
58
```

```
60
         Parameters
 61
 62
          :param signal
 63
 64
         Returns
 65
          -----
 66
         :return complex_fourier_signal
 67
 68
 69
         # find the length of the signal
 70
 71
         N = signal.shape[0]
 72
         if signal.ndim == 2:
             M, N = signal.shape
 73
 74
         # calculate DFT matrix
 75
 76
         u, v = np.meshgrid(np.arange(N), np.arange(N))
         omega = np.exp(-2 * np.pi * 1j / N)
 77
         dft_matrix = np.power(omega, u*v)
 78
 79
          # if it is a matrix of signals
 80
         if signal.ndim == 2:
 81
              # calculate the Fourier Transform
 82
              complex_fourier_signal = np.dot(dft_matrix, signal.transpose())
 83
 84
              return complex_fourier_signal.transpose()
 85
          # calculate the Fourier Transform
 86
          complex_fourier_signal = np.dot(dft_matrix, signal)
 87
         return complex_fourier_signal
 88
 89
 90
     def IDFT(fourier_signal):
 91
 92
 93
          Function that return IDFT of fourier signal
          if matrix is input: return every row IDFT
 94
 95
 96
         Parameters
 97
         :param fourier_signal
 98
 99
100
         Returns
101
         :return 1/N * signal
102
103
          11 11 11
104
105
106
          # find the length of the signal
         N = fourier_signal.shape[0]
107
108
         if fourier_signal.ndim == 2:
              M, N = fourier_signal.shape
109
110
          # calculate IDFT matrix
111
112
         u, v = np.meshgrid(np.arange(N), np.arange(N))
         omega = np.exp(2 * np.pi * 1j / N)
113
         idft_matrix = np.power(omega, u*v)
114
115
          # if it is a matrix of fourier signals
116
         if fourier_signal.ndim == 2:
117
              # calculate the Fourier Transform
118
119
              signal = np.dot(idft_matrix, fourier_signal.transpose())
              return 1/N * signal.transpose()
120
121
122
          # calculate the inverse Fourier Transform
         signal = np.dot(idft_matrix, fourier_signal)
123
         return 1/N * signal
124
125
126
    def DFT2(image):
127
```

```
128
129
         Function that return 2D DFT of image
130
131
         Parameters
132
133
          :param image (matrix)
134
         Returns
135
136
137
         :return fourier_image
138
139
140
         M, N = image.shape
141
142
         # build the dft2_matrix transform
143
         omega_y = np.exp(-2 * np.pi * 1j / M)
144
         u, v = np.meshgrid(np.arange(M), np.arange(M))
145
         dft2_matrix = np.power(omega_y, u*v)
146
147
148
          # calculate the 2D fourier transform
         fourier_image = np.dot(dft2_matrix, DFT(image))
149
150
         return fourier_image
151
152
153
     def IDFT2(fourier_image):
154
155
          Function that return 2D IDFT of an image
156
157
158
         Parameters
159
160
         :param fourier_image
161
         Returns
162
163
164
         :return image
165
         11 11 11
166
167
         M, N = fourier_image.shape
168
         # build the idft2_matrix transform
169
         omega_y = np.exp(2 * np.pi * 1j / M)
170
171
         u, v = np.meshgrid(np.arange(M), np.arange(M))
         idft2_matrix = np.power(omega_y, u*v)
172
173
174
          # calculate the 2D inverse fourier transform
         return 1/M * np.dot(idft2_matrix, IDFT(fourier_image))
175
176
177
     def conv_der(im):
178
179
180
          derivative of an image using convolution
181
182
         Parameters
183
          :param im
184
185
         Returns
186
187
         :return magnitude of the derivative
188
189
190
191
         # set der x/y matrix
192
193
         der_x = np.array([[1, 0, -1]])
         der_y = np.array(der_x.transpose())
194
          \# calculate the derivative to x and y
195
```

```
196
         dx = conv(im, der_x, mode='same')
197
         dy = conv(im, der_y, mode='same')
198
199
         return np.sqrt(np.abs(dx)**2 + np.abs(dy)**2) # = magnitude
200
201
     def fourier_der(im):
202
203
204
          derivative of an image using fourier transform
205
         Parameters
206
207
208
          :param im
209
210
         Returns
211
212
          :return magnitude of the derivative
213
214
215
         # constants
216
         M, N = im.shape
217
         u = np.meshgrid(np.arange(N), np.arange(M))[0] - N//2
218
219
         v = np.meshgrid(np.arange(N), np.arange(M))[1] - M//2
         u_der, v_der = (2 * np.pi * 1j / N), (2 * np.pi * 1j / M)
220
221
         \# calculate dx, dy
222
223
         dx = u_der * IDFT2(np.fft.fftshift(u) * DFT2(im))
         dy = v_der * IDFT2(np.fft.fftshift(v) * DFT2(im))
224
225
226
         return np.sqrt(np.abs(dx)**2 + np.abs(dy)**2) # = magnitude
227
228
229
     def gaussian_kernel_factory(kernel_size):
230
231
          create gaussian matrix
232
233
         Parameters
234
         :param kernel_size
235
236
237
         Returns
238
239
          :return gaussian matrix
240
241
242
         gaussian = binomial_ker = np.array([[1, 1]])
243
244
         while gaussian.shape[1] < kernel_size: gaussian = conv(gaussian, binomial_ker)</pre>
         gaussian_kernel = np.ones((kernel_size, kernel_size)) * gaussian * gaussian.transpose()
245
246
^{247}
         return 1 / gaussian_kernel.sum() * gaussian_kernel
248
249
250
     def blur_spatial(im, kernel_size):
251
252
          blur image using gaussian convolution
253
         Parameters
254
255
^{256}
         :param im
257
258
         :param kernel_size
259
260
         Returns
261
          :return blur image
262
263
```

```
^{264}
          return conv(im, gaussian_kernel_factory(kernel_size), mode='same')
265
266
267
     def blur_fourier(im, kernel_size):
268
269
270
          blur image with DFT multiply
271
272
          Fourier of im \ensuremath{\mathfrak{G}} Fourier of gaussian
          and multiply wisely
273
274
275
          Parameters
276
          :param im
277
278
          :param kernel_size
279
280
          Returns
281
          :return blur image
282
283
          11 11 11
284
285
286
          # build the kernel with zero padding
          kernel_base = gaussian_kernel_factory(kernel_size)
287
          window = np.zeros_like(im).astype(np.float64)
288
          M, N = im.shape
289
          dx, dy = kernel_base.shape
290
          x_{middle}, y_{middle} = N//2, M//2
291
292
          \label{lem:window} window \hbox{\tt [(y_middle-dy//2):(y_middle+dy//2+1), (x_middle-dx//2):(x_middle+dx//2+1)] = kernel\_base}
293
294
          # multiply in the freq domain
295
          return IDFT2(DFT2(im) * DFT2(np.fft.ifftshift(window))).real
^{296}
```