

Cool Images: HW 2

Problem #3: Programming Exercise

Part A: Morphing Lena + Man

For the first part of the programming exercise, we were tasked to create a new image using the phase and magnitude spectrum of the lena and man respectively. The original figures and fourier spectra are as shown below. The last row corresponds to the new image using the new phase and magnitude. We can see the importance of the phase because even if we have the magnitude of the lena fourier spectrum, the figure looks pretty messy and still somewhat cool.

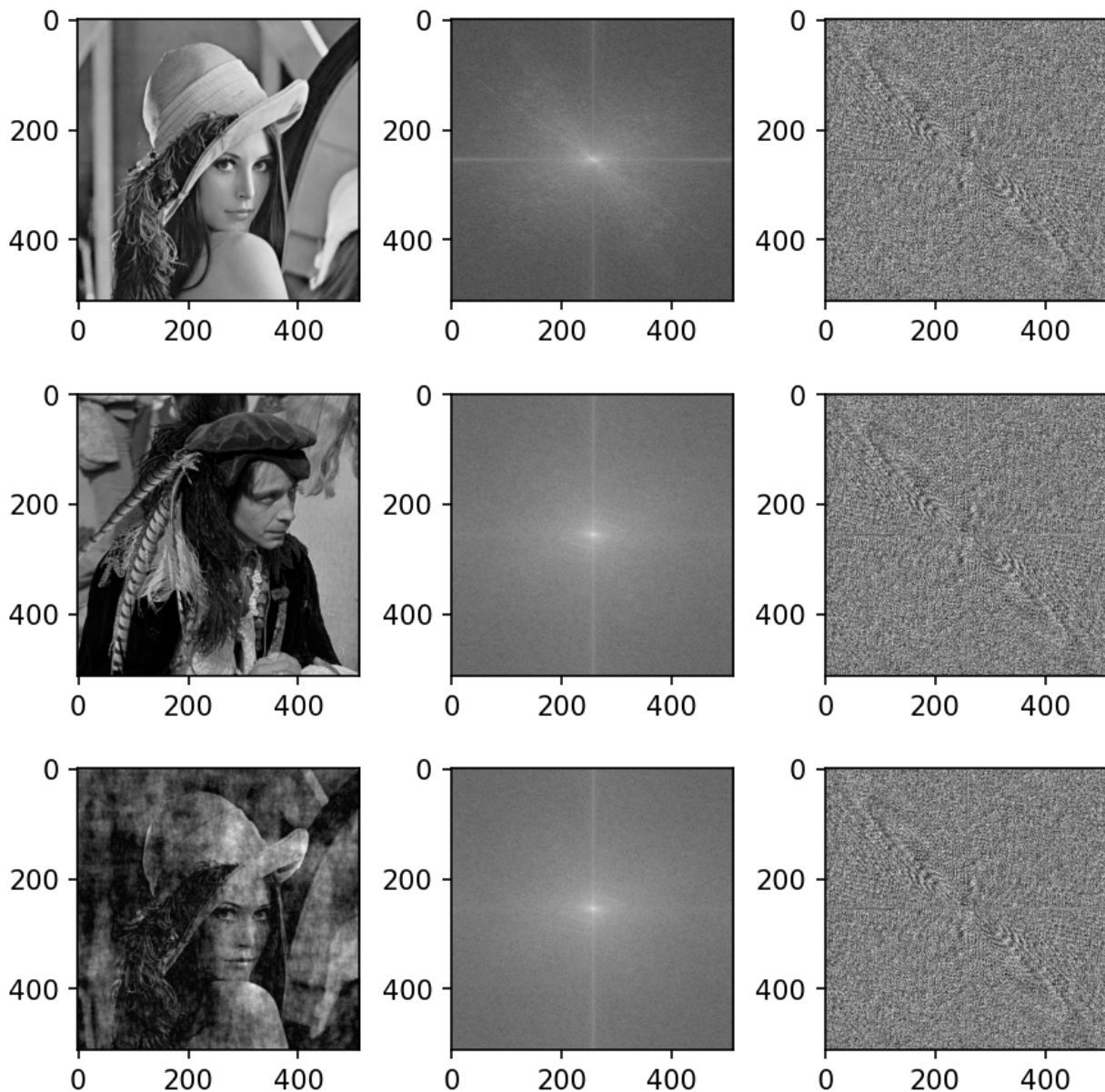


Figure 1: Adam and Eve merge together

Part B: DCT and DFT

For the second part of the programming exercise, we will try to compress the images by using DCT and DFT methods. The following figures show the compressed version of the lena image. It can be seen that the DCT method makes it look better while also using lower space for storage which is perfect.

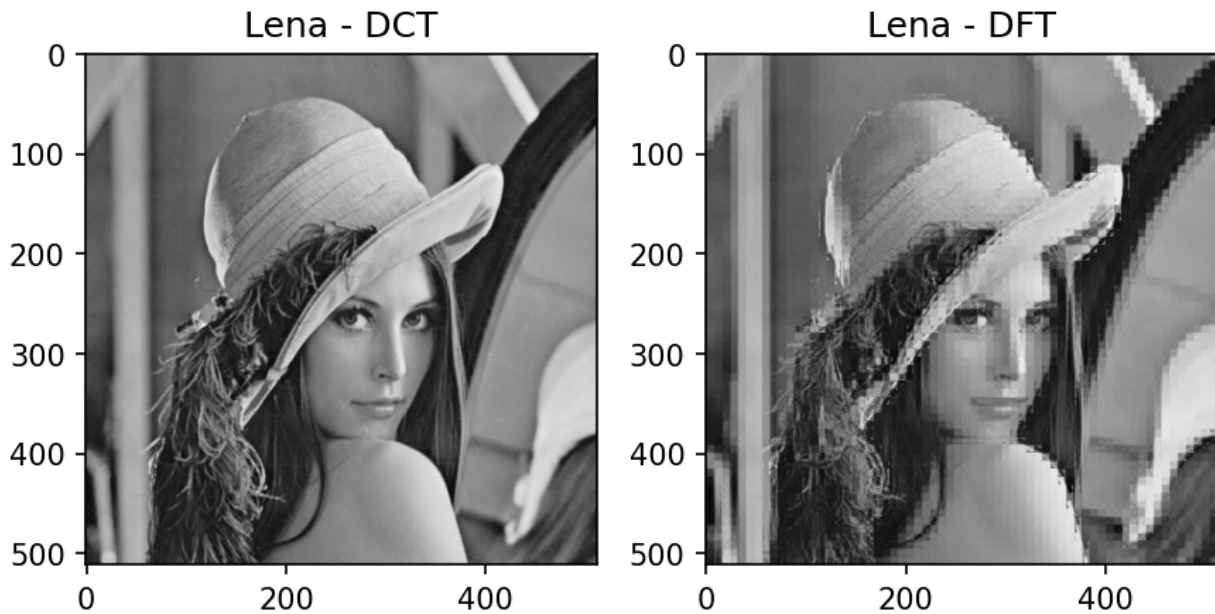


Figure 2: Lena cosine and lena fourier

Part C: Sharpening

For the last part of the programming exercise, the images are sharpened by using a high pass filter. This is pretty cool because it is actually pretty simple but gives out a decent sharpened image. Coins are usually dull but I guess this is a way to make it sharper.

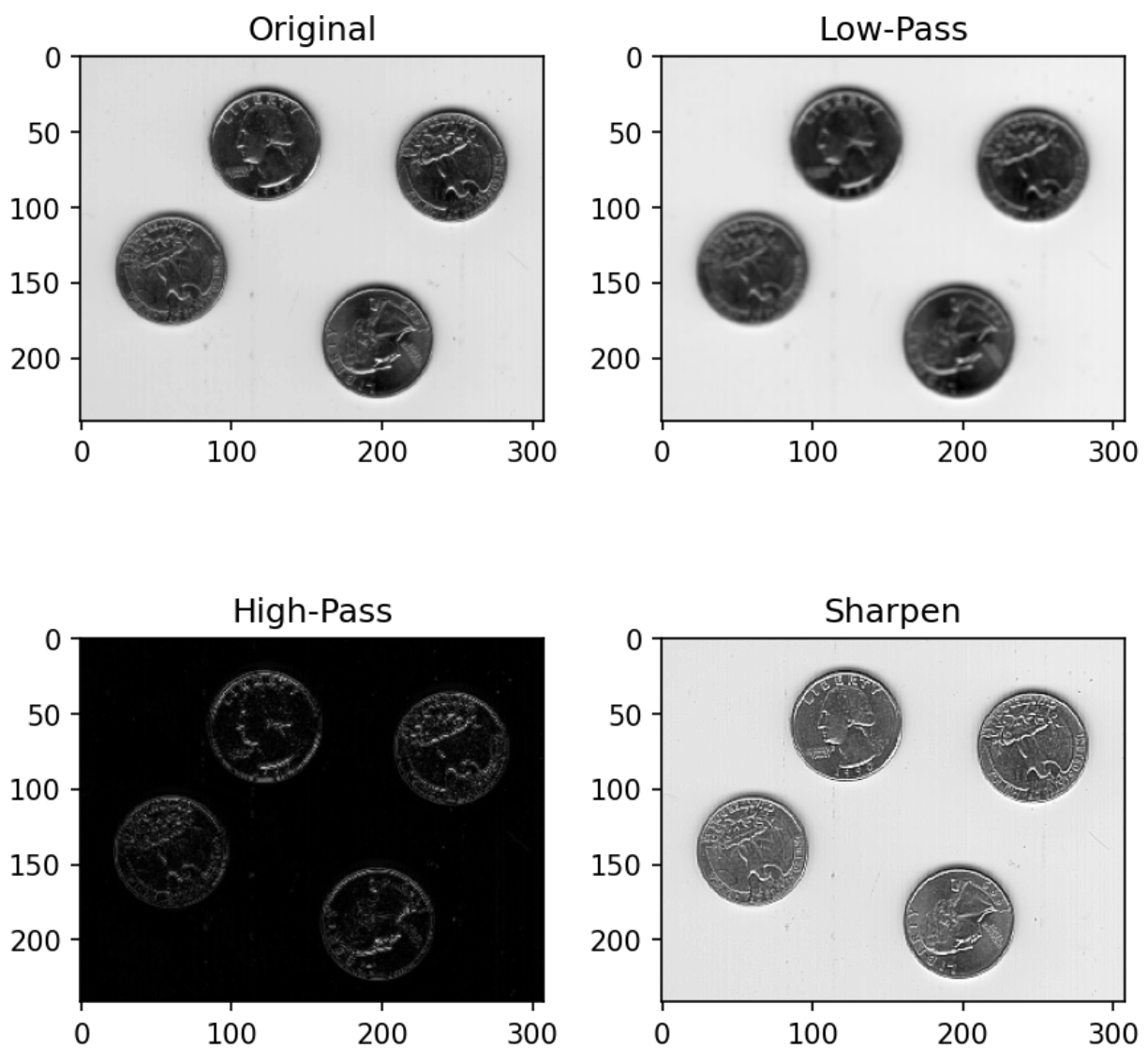


Figure 3: Sharp Coins ???

Appendix

Python Code

```
1 '''
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3 EE 5561 - Image Processing
4 Problem Set 2
5 October 16, 2023
6 '''
7
8 # %% Problem Statement
9 """
10     a) [3 pts] Use the phase of the Fourier spectrum of Lena image, and the magnitude of the
11         Fourier spectrum of the kneeling man image (provided to you), to generate a combined
12         image
13         with the corresponding phase and magnitude Fourier spectra. Display the input images and
14         their relevant Fourier spectra, and the final output image.
15
16     b) [6 pts] Take the DCT of each distinct (i.e. not sliding) 8 x 8. Keep the largest (in
17         magnitude) 10 DCT coefficients, and set the rest to zero. Take the inverse DCT of each
18         block to generate a new image. Do the same with DFT and inverse DFT (i.e. FFT). Display
19         the images.
20         (Hint: In Python, it will be helpful to define a function to extract/process non-
21         overlapping
22         sliding blocks and store the results in a new array. For MATLAB, the built-in function
23         blkproc may be helpful. Also defining a function that performs the given transform, then
24         keeps the largest 10 coefficients and then does the inverse transform will also be
25         useful.)
26
27     c) [9 pts] Perform the sharpening exercise with the coins image ( coins.tif ).
28         Read-in the image, then perform lowpass filtering with an averaging filter (use imfilter
29         /signal.convolve2d).
30         Generate the high-pass image by subtracting this low-pass image from the original.
31         Generate the sharpened image as original image plus 2 times the high-pass image. Display
32         the
33         original, low-pass, high-pass and sharpened images.
34 """
35 # %%
36
37 import numpy as np
38 import matplotlib.pyplot as plt
39 import matplotlib.image as img
40 import scipy
41 import imageio
42 from scipy.fftpack import dct, idct
43 from scipy.signal import convolve2d
44
45 def fft2c(img):
46     return np.fft.fftshift(np.fft.fft2(np.fft.ifftshift(img), norm = 'ortho'))
47
48 def ifft2c(freq):
49     return np.fft.ifftshift(np.fft.ifft2(np.fft.fftshift(freq), norm = 'ortho'))
50
51 def dct2(img):
52     return dct(dct(img.T, norm='ortho').T, norm='ortho')
53
54 def idct2(coeff):
55     return idct(idct(coeff.T, norm='ortho').T, norm='ortho')
56
57 def moving_average_filter(img, kernel_size):
58     kernel = np.ones((kernel_size[0], kernel_size[1])) / (np.product(kernel_size))
59     return convolve2d(img, kernel, mode='same', boundary='wrap')
60
61 # %% Problem 3 Part A
62 def prob3a():
63     # Load different images as an ndarray
64     lena_img = np.asarray(img.imread('lena512.bmp'))
65     man_img = imageio.v2.imread("man.png")
66
67     # Take the fourier transform of the lena and man image
```

```

63 lena_freq = fft2c(lena_img)
64 lena_phase = np.angle(lena_freq)
65 man_freq = fft2c(man_img)
66 man_phase = np.angle(lena_freq)
67
68 # Morph using magnitude of the man and the phase of lena
69 man_mag = np.abs(man_freq)
70 morph_img = ifft2c(man_mag * np.exp(1j * lena_phase))
71
72 # Plot the figures
73 fig, ax = plt.subplots(3,3, figsize=(6,6), dpi=150)
74 ax[0,0].imshow(lena_img, cmap="gray")
75 ax[0,1].imshow(np.log(np.abs(lena_freq)), cmap="gray")
76 ax[0,2].imshow(lena_phase, cmap="gray")
77
78 ax[1,0].imshow(man_img, cmap="gray")
79 ax[1,1].imshow(np.log(np.abs(man_freq)), cmap="gray")
80 ax[1,2].imshow(man_phase, cmap="gray")
81
82 ax[2,0].imshow(np.abs(morph_img), cmap="gray")
83 ax[2,1].imshow(np.log(np.abs(man_freq)), cmap="gray")
84 ax[2,2].imshow(lena_phase, cmap="gray")
85 fig.tight_layout()
86
87
88 # %% Problem 3 Part B
89 def prob3b():
90     lena_img = np.asarray(img.imread('lena512.bmp'))
91     lena_DCT = np.zeros(lena_img.shape)
92     lena_DFT = np.zeros(lena_img.shape)
93     #lena_DCT = dct2(lena_img)
94     #fig, ax = plt.subplots()
95     #ax.imshow(lena_DCT)
96
97     size = 8
98     for row in range(int(round(lena_img.shape[0]/size))):
99         for col in range(int(round(lena_img.shape[1]/size))):
100             temp_img = dct2(lena_img[row*size:(row+1)*size,col*size:(col+1)*size])
101             g = np.unravel_index(np.argsort(np.abs(temp_img.ravel()))[::-1][:10], [size,size])
102         TEMP_IMG = np.zeros((size,size))
103         for index in g:
104             TEMP_IMG[index] = temp_img[index]
105             lena_DCT[row*size:(row+1)*size,col*size:(col+1)*size] = idct2(TEMP_IMG)
106
107             temp_img = fft2c(lena_img[row*size:(row+1)*size,col*size:(col+1)*size])
108             g = np.unravel_index(np.argsort(np.abs(temp_img.ravel()))[::-1][:10], [size,size])
109         TEMP_IMG = np.zeros((size,size))
110         for index in g:
111             TEMP_IMG[index] = temp_img[index]
112             lena_DFT[row*size:(row+1)*size,col*size:(col+1)*size] = ifft2c(TEMP_IMG)
113
114             #if (row+1) % 16 == 0 and (col+1) % 16 == 0:
115             #fig, ax = plt.subplots()
116             #ax.imshow(lena_DCT)
117
118     fig, ax = plt.subplots(1,2, figsize=(6,3), dpi=150)
119     ax[0].imshow(lena_DCT, cmap="gray")
120     ax[0].set_title("Lena - DCT")
121
122     ax[1].imshow(lena_DFT, cmap="gray")
123     ax[1].set_title("Lena - DFT")
124     fig.tight_layout()
125
126     pass
127
128
129 # %% Problem 3 Part C
130 def prob3c():
131     # Apple low_pass filter using convolution
132     coins_img = np.complex64(plt.imread('eight.tif'))

```

```

133 coins_img = np.asarray(img.imread('eight.tif'))
134 #print(coins_img[200,100])
135 coins_low_pass = moving_average_filter(coins_img, [3,3])
136 #print(coins_low_pass[200,100])
137 coins_high_pass = coins_img - coins_low_pass # max is 104 which is unusually high
138 #print(coins_high_pass[200,100])
139 #coins_sharp = np.abs(coins_img + 2 * coins_high_pass)
140 coins_sharp = np.clip(np.abs(coins_img + 2 * coins_high_pass), 0, 255)
141 #print(coins_sharp[200,100])
142
143 # Plot the figures
144 fig, ax = plt.subplots(2,2, figsize=(6,6), dpi=150)
145 ax[0,0].imshow(np.abs(coins_img), cmap="gray")
146 ax[0,0].set_title("Original")
147
148 ax[0,1].imshow(np.abs(coins_low_pass), cmap="gray")
149 ax[0,1].set_title("Low-Pass")
150
151 ax[1,0].imshow(np.abs(coins_high_pass), cmap="gray")
152 ax[1,0].set_title("High-Pass")
153 print(np.max(coins_high_pass))
154
155 ax[1,1].imshow(np.abs(coins_sharp), cmap="gray")
156 ax[1,1].set_title("Sharpen")
157 fig.tight_layout()
158 print(np.max(coins_sharp))
159
160 # %% Main Function
161 if __name__ == "__main__":
162     prob3a()
163     prob3b()
164     prob3c()

```