

CS-663 Assignment 5 Q4

Soham Naha (193079003)
Akshay Bajpai (193079002)
Mohit Agarwala (19307R004)

November 15, 2020

4

Consider the `barbara256.png` image from the homework folder. Implement the following in MATLAB: (a) an ideal low pass filter with cutoff frequency $D \in \{40, 80\}$, (b) a Gaussian low pass filter with $\sigma \in \{40, 80\}$. Show the effect of these on the image, and display all images in your report. Display the frequency response (in log Fourier format) of all filters in your report as well. Comment on the differences in the outputs. Make sure you perform appropriate zero-padding! [20 points]

D_0 is the cut-off frequency.

The frequency domain Ideal Low Pass filter at frequency (u,v) is given by:

$$H(u, v) = 1, \sqrt{u^2 + v^2} \leq D_0 \quad (1)$$

The frequency domain Gaussian Low Pass filter at frequency (u,v) is given by:

$$H(u, v) = \exp\left(-\frac{u^2 + v^2}{2\sigma^2}\right), \quad (2)$$

where $\sigma = D_0$

```
1  import numpy as np
2  import matplotlib.pyplot as plt
3  import cv2
4
5  def read_image(fname, verbose=True, is_rgb=False):
6      """Read the image and return as an array
7      :param fname : the path to the input image
8      :param verbose : if True saves the read image
9      :param is_rgb : if true means the read image is RGB
10     :output image : the image as a numpy array
11     """
12     params = {"fname": fname.split("/")[-1].split(".")[0], "attrib": "Original Image"}
13     if is_rgb:
14         image = cv2.imread(fname)
15     else:
16         image = cv2.imread(fname,0)
17     if verbose:
18         save_images(image, params)
19     return image
20
21 def distance(point1, point2):
22     """Returns euclidean distance between 2 2D points
23     :param point1, point2 : the two 2D-points under consideration
24     :output : the euclidean distance
25     """
26     return np.sqrt(np.sum(np.square(point1-point2)))
27
28 def idealFilterLP(D0, r, c, verbose= True):
29     """Create an ideal Low-Pass filter with a given cutoff frequency and filter shape
30     :param D0 : the cut-off frequency of the low pass filter
31     :param r,c : the shape of the filter (r: rows, c: cols)
32     :param verbose : if True saves the filter constructed as an image
```

```

33     :output base : the constructed filter
34     """
35     params={"filter": "IdealLP", "D0": D0, "r": r, "c": c,
36             "attrib": "Ideal Low Pass Filter with D0 : "+str(D0)}
37     base = np.zeros((r,c))
38     center = np.array([r//2,c//2])
39     for x in range(c):
40         for y in range(r):
41             if distance(np.array([y,x]),center) < D0:
42                 base[y,x] = 1
43
44     if verbose:
45         save_images(base, params)
46
47     return base
48
49 def gaussianLP(D0, r, c, verbose=True):
50     """Create a Gaussian LowPass filter with a given cut-off frequency and filter size
51     :param D0 : the cut-off frequency for the gaussian LP filter
52     :param r,c : the shape of the Gaussian Filter
53     :param verbose : if True saves the filter constructed as an image
54     :output base : the constructed Gaussian Filter
55     """
56     params={"filter": "GaussianLP","D0": D0, "r": r, "c": c,
57             "attrib": "Gaussian Low Pass Filter with D0 : "+str(D0)}
58     base = np.zeros((r,c))
59     center = np.array([r//2,c//2])
60     for x in range(c):
61         for y in range(r):
62             base[y,x] = np.exp(((distance(np.array([y,x]),center)**2)/(2*(D0**2))))
63
64     if verbose:
65         save_images(base, params)
66
67     return base
68
69 def multiply(fname, image, filterKernel, ktype, D ,verbose=True):
70     """Filters the image in frequency domain and returns in spatial domain
71     :param image: the input image in spatial domain
72     :param filterKernel : the LP filter kernel in frequency domain
73     :output inverse_img : the filtered image in spatial domain
74     """
75     name = fname.split("/")[-1].split(".")[0]
76     fft_image = np.fft.fftshift(np.fft.fft2(image))
77     mult_full = fft_image * filterKernel
78     r,c = mult_full.shape
79     #print("Shape Before : ",mult_full.shape)
80     center = (r//2, c//2)
81     mult = mult_full[center[0] - r//4:center[0] + r//4, center[1]-c//4: center[1]+c//4]
82     #print("Shape After : ",mult.shape)
83     inverse_img = np.abs(np.fft.ifft2(np.fft.ifftshift(mult)))
84
85     if verbose:
86         save_images(np.log(1+np.abs(fft_image)), {"attrib": "FFT Image", "fname": name,
87             "type": "FFT_Image","kernel_tyoe": ktype, "D0": str(D) })
88         save_images(np.log(1+np.abs(mult)), {"attrib": "Fourier Filtered Image",
89             "fname": name, "type": "Fourier_Filtered","kernel_tyoe": ktype, "D0": str(D) })
90         save_images(inverse_img, {"attrib": "Filtered Image","type": "Reconstructed",
91             "fname": name, "kernel_type": ktype, "D0": str(D)})
92
93     return inverse_img
94
95 def save_images(image, params, is_rgb=False):
96     """Saves the images based on the parameters passed
97     :param image : the image to be saved

```

```

98     :param params : the parameters of the image on construction
99     """
100     keys = params.keys()
101     path = "../images/"
102     filename = ""
103     title = ""
104     for key in params:
105         if key!="attrib":
106             filename += str(params[key]) + "_"
107
108     for key in params:
109         if key in ["fname","attrib"]:
110             title += str(params[key]) + "_"
111
112     filename = path + filename + ".png"
113     plt.figure()
114     plt.title(title)
115     if is_rgb:
116         plt.imshow(image)
117         plt.colorbar()
118         plt.savefig(filename, bbox_inches="tight")
119     else:
120         plt.imshow(image, cmap="gray")
121         plt.colorbar()
122         plt.savefig(filename,cmap="gray", bbox_inches="tight")
123
124 def pad_image(image):
125     """Padd the input image to twice its size
126     :param image: input image
127     :output out : the double uniformly padded image
128     """
129     r,c = image.shape[:2]
130     out = np.zeros((2*r, 2*c))
131     for i in range(r):
132         for j in range(c):
133             out[2*i,2*j] = image[i,j]
134
135     return out
136
137 def main(fname,D,types):
138     kern = {"ideallP": idealFilterLP, "gaussLP": gaussianLP}
139     in_image = read_image(fname)
140     image = pad_image(in_image)
141     r,c = image.shape[:2]
142     for t in types:
143         for d in D:
144             kernel = kern[t](d,r,c)
145             out = multiply(fname, image, kernel, t, d )
146
147 if __name__=="__main__":
148     filename = "../data/barbara256.png"
149     D = [40,80]
150     types=["ideallP","gaussLP"]
151     main(filename, D, types)

```

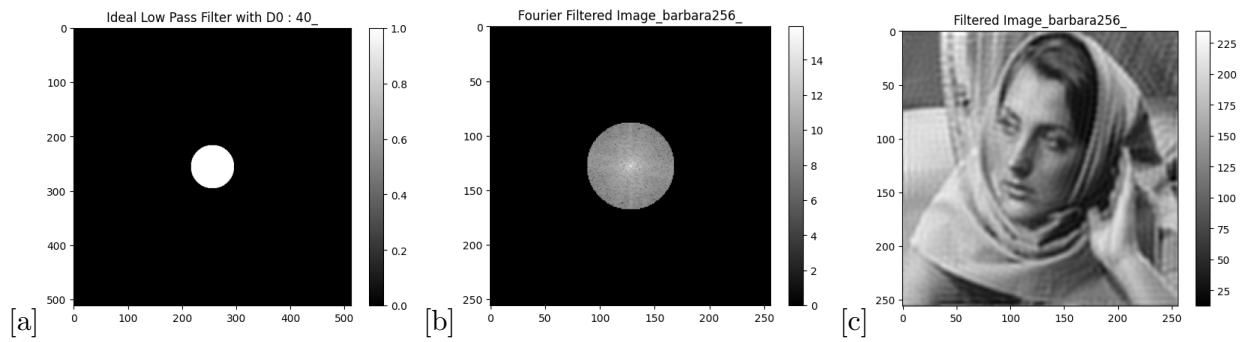


Figure 1: (a) Ideal Low Pass Filter with $D_0 = 40$ (b) Frequency domain filtered output (c) Spatial domain Output Image

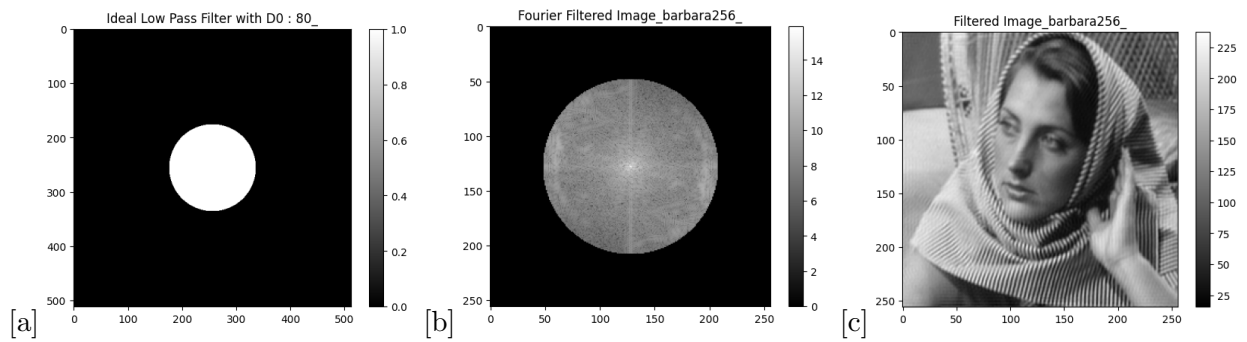


Figure 2: (a) Ideal Low Pass Filter with $D_0 = 80$ (b) Frequency domain filtered output (c) Spatial domain Output Image

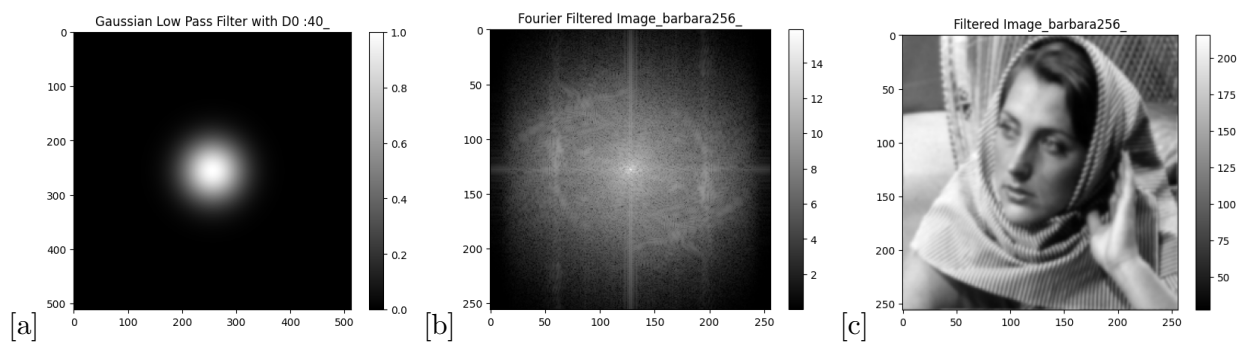


Figure 3: (a) Gaussian Low Pass Filter with $D_0 = 40$ (b) Frequency domain filtered output (c) Spatial domain Output Image

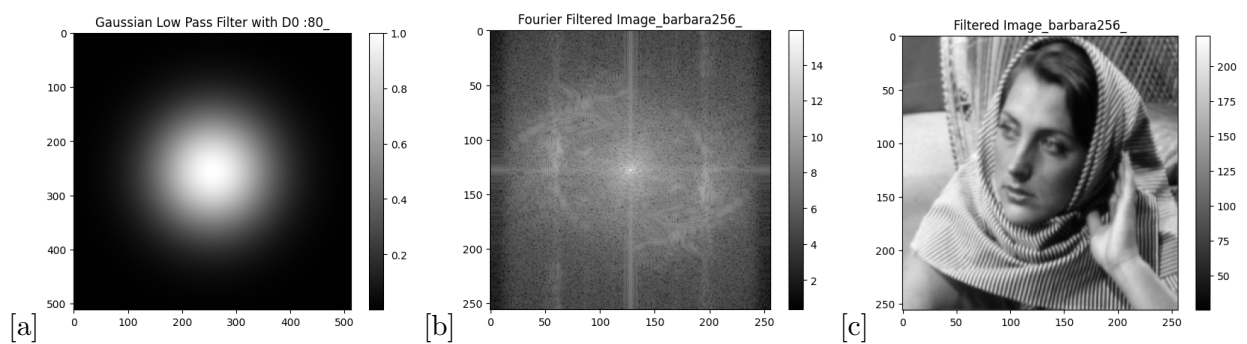


Figure 4: (a) Gaussian Low Pass Filter with $D_0 = 80$ (b) Frequency domain filtered output (c) Spatial domain Output Image

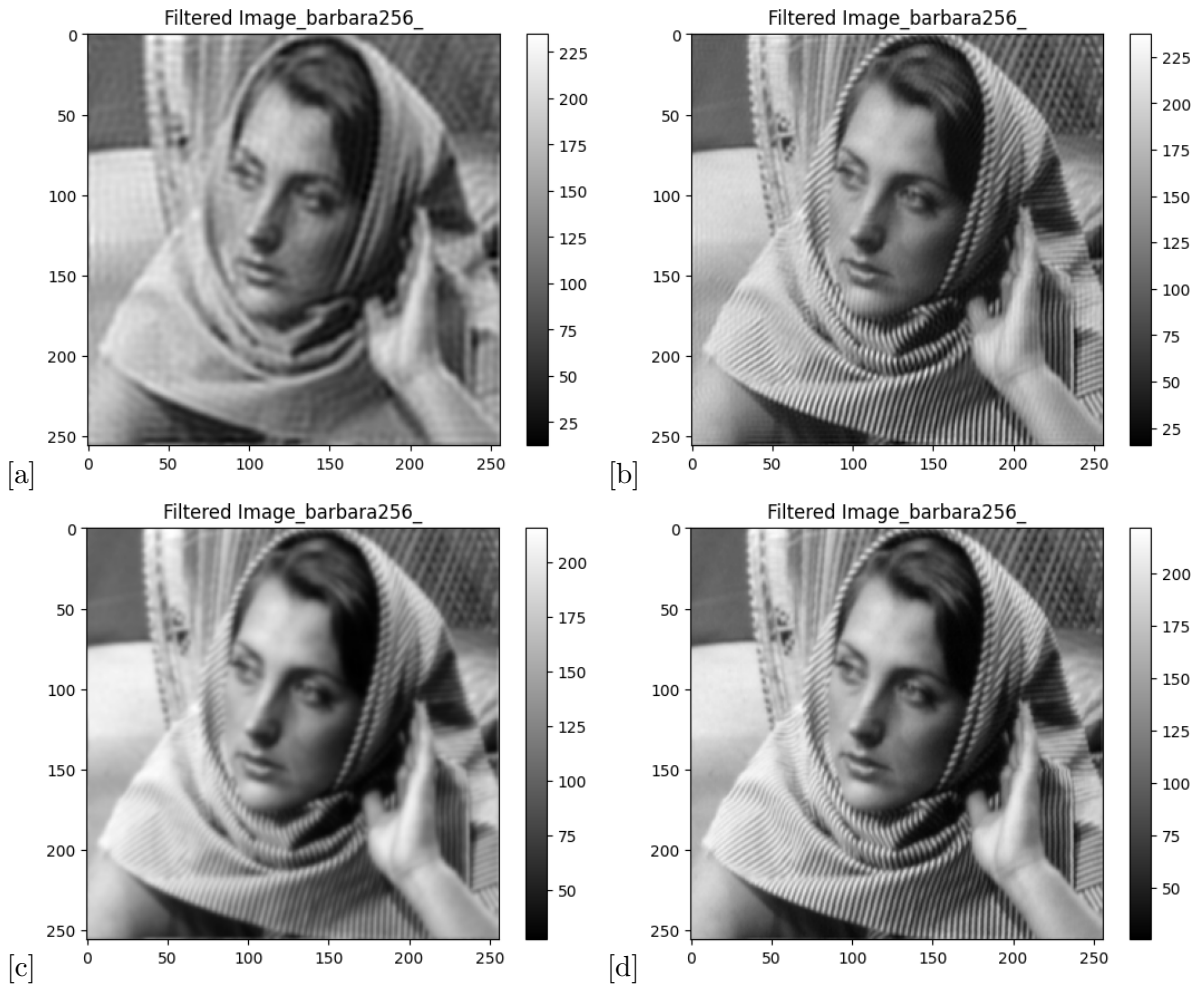


Figure 5: (a) Ideal LP $D_0 = 40$ (b) Ideal LP $D_0 = 80$ (c) Gaussian LP $D_0 = 40$ (d) Gaussian LP $D_0 = 80$

Discussion :

- The low pass filters in general are such that they allow low-frequency components to pass and reject the high frequency components. All the four filters constructed, follow the same rule.
- A higher cut-off frequencies allows more high frequency components to be present in the image, so blurring or smoothing is more effective at low cut-off frequencies.
- The cut-off frequencies of the low-pass filter determines the frequency of transition from passing frequencies to stopping frequencies. In case of ideal filters, the transition is quite fast, but in case of gaussian the transition is smooth.
- The images filtered by the Ideal Low Pass filters have some prominent **ringing artifacts**, as the sharp transitions of the Ideal Low Pass filters introduced jump discontinuity. At $D_0 = 40$, the effect is more prominent then at $D_0 = 80$.
- For the gaussian filter, as the transition is smooth, ringing effects are not present. But the smoothing is not as good as the ideal filter.