Computer Vision: Assignment 1 (Image Filtering)

An image filtering is a technique through which size, colors, shading and other characteristics of an image are altered. An image filter is used to transform the image using different graphical editing techniques. Image filters are usually done through graphic design and editing software.

Image filtering is useful for many applications, including smoothing, sharpening, removing noise, and edge detection. A filter is defined by a kernel, which is a small array applied to each pixel and its neighbors within an image. In most applications, the center of the kernel is aligned with the current pixel, and is a square with an odd number (3, 5, 7, etc.) of elements in each dimension. The process used to apply filters to an image is known as convolution, and may be applied in either the spatial or frequency domain.

Requirements

- Python3
- OpenCV
- Numpy
- Matplotlib

Importing required Libraries

In [0]:

import cv2
import matplotlib.pyplot as plt
import numpy as np
import math
import matplotlib.image as mpimg

Load image from drive and shows the original image.

imshow:

Function that handles both rgb and grayscale image and shows them accordingly.

Parameters:

• img: input img

Return: None

```
In [32]:
```

```
img = mpimg.imread("original.jpg")

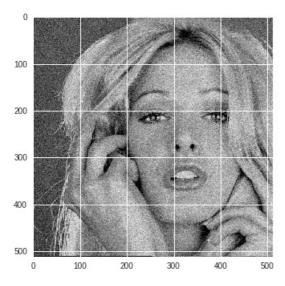
def imshow(img):
    if len(img.shape) < 3:
        plt.imshow(img, cmap="gray")
    elif img.shape[2] == 1:
        plt.imshow(np.resize(img,(img.shape[0],img.shape[1])), cmap="gray")
    else:
        plt.imshow(img)

if len(img.shape) < 3:
    img.resize(*img.shape,1)

imshow(img)

print(img.shape)</pre>
```

(512, 512, 1)



Gamma Correction

gammaCorrection

Function that apply gamma correction to input image

Parameters:

- img(2d/3d matrix): input image
- gamma(double): gamma contant

Return:

• Gamma Corrected image

```
In [33]:
```

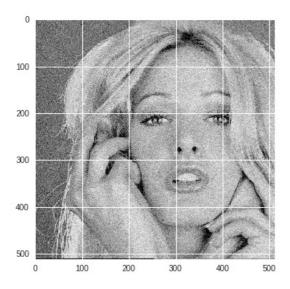
```
# Gamma Coorection
def gammaCorrection(img, gamma):
    out = np.power(img.copy(), gamma)
    return out.astype(np.uint8)

imshow(gammaCorrection(img,0.7))

cv2.imwrite("gamma.jpg",img)
```

Out[33]:

True



Low Pass, High Pass and Gaussian Filter

gfunc:

Implementation of gaussian function

Parameters:

• x, y and sigma(double): gaussian function parameters

Return:

• Scaler value at x,y for sigma of gaussian function

gaussFilter:

Function that applies gaussian filter to an image

Parameters:

- size(tuple): size of gaussian filter
- **sigma**(double): parameter of gaussian function

Return:

• Gaussian Filter(2d matrix)

```
In [34]:
```

```
# low pass filter
(lpfw, lpfh) = (3,3)
lowPassFilter = np.ones((lpfw,lpfh))*1/(lpfw*lpfh)
# high pass filter
(hpfw,hpfh) = (3,3)
highPassFilter = -1*np.ones((hpfw,hpfh))
highPassFilter[hpfw//2,hpfh//2] = -np.sum(highPassFilter)-1
# gaussian filter
def gfunc(x,y,sigma):
    return (math.exp(-(x**2 + y**2)/(2*(sigma**2))))/(2*3.14*(sigma**2))
def gaussFilter(size, sigma):
    out = np.zeros(size)
    for i in range(size[0]):
        for j in range(size[1]):
            out[i,j] = gfunc(i-size[0]//2,j-size[1]//2, sigma)
    return out/np.sum(out)
(gfw,gfh) = (3,3)
gaussianFilter = gaussFilter((gfw,gfh),1)
print("Low Pass Filter")
print(lowPassFilter)
print("High Pass Filter")
print(highPassFilter)
print("Gaussian Filter")
print(gaussianFilter)
Low Pass Filter
```

```
Low Pass Filter
[[0.11111111 0.11111111 0.11111111]
[0.11111111 0.11111111 0.11111111]
[0.11111111 0.11111111 0.11111111]
High Pass Filter
[[-1. -1. -1.]
[-1. 8. -1.]
[-1. -1. -1.]]
Gaussian Filter
[[0.07511361 0.1238414 0.07511361]
[0.1238414 0.20417996 0.1238414]
[0.07511361 0.1238414 0.07511361]]
```

Example of Low Pass and Gaussian Filter

conv

Function that applies convolution to an 2d/3d matrix or numpy array on the given filter

Parameters:

- image (2d/3d matrix): image on which convolution will be applied with given filter
- filter (2d matrix): filter which will applied to image

Return:

• filtered image(2d/3d matrix)

```
In [35]:
```

```
def conv(image, filter):
   iw,ih,id = image.shape
   fw,fh = filter.shape
   out = np.zeros((iw-fw+1,ih-fh+1,id))
   for d in range(id):
      for w in range(ih-fh+1):
         for h in range(iw-fw+1):
             out[w,h,d] = np.sum(filter*image[w:w+fh , h:h+fw , d])
   if id == 1:
      return np.resize(out, (out.shape[0], out.shape[1])).astype(np.uint8)
   else:
      return out.astype(np.uint8)
fig = plt.figure(1, figsize=(18, 16))
for i in range(3,8,2):
   splot = plt.subplot(130-1+((i+1)//2))
   splot.set xlabel(str(i)+"x"+str(i))
   (lpfw, lpfh) = (i,i)
   lowPassFilter = np.ones((lpfw,lpfh))*1/(lpfw*lpfh)
   low image = conv(img, lowPassFilter)
   imshow(low image)
   cv2.imwrite("low"+str(i)+"x"+str(i)+".jpg",low image)
plt.show()
print()
fig = plt.figure(2, figsize=(18, 16))
for i in range(3,8,2):
   splot = plt.subplot(130-1+((i+1)//2))
   splot.set_xlabel(str(i)+"x"+str(i))
   (gfw,gfh) = (i,i)
   gaussianFilter = gaussFilter((gfw,gfh),4)
   gaussian_image = conv(img, gaussianFilter)
   imshow(gaussian image)
   cv2.imwrite("gaussian"+str(i)+"x"+str(i)+".jpg",gaussian_image)
plt.show()
```

Bilateral Filter and its example

bfunc:

Function that create Bilateral Filter according to image portion and give pixel value after applying it to image part

Parameters:

- i,j: pixel value for which bilateral filter will be applied
- fw,fh: Filter width and filter height
- image: input image on which pixel (i,j) exist
- sigma1, sigma2: bilateral function parameters
- bilaterealWFilter: weighted bilateral Filteral

bilateralFilterConv:

Function that apply convolution on given image with the bilateral filter of given size

Parameters:

- image: input image on which bilateral filter will be applied
- fw,fh: widht and height if bilateral filter

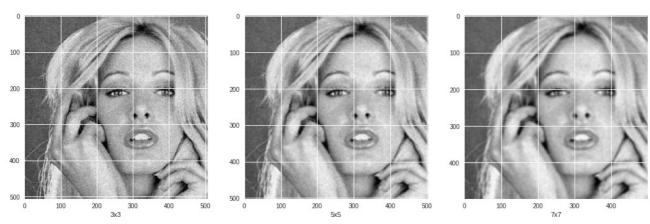
Return:

• returns a bilateral filtered image (2d/3d matrix)

```
In [36]:
```

```
# bilateral filter
def bfunc(i,j,fw,fh,image,sigma1, sigma2, bilateralWFilter):
   imgwork = image[i - fh//2:i+1 + fh//2, j - fw//2:j+1 + fw//2, :]
   bilateralIFilter = ((imgwork - image[i, j,:])**2 )/(2*(sigma1**2))
   bilateralFilter = np.exp(-1*bilateralIFilter)*bilateralWFilter
   bilateralFilter = bilateralFilter/np.sum(bilateralFilter,axis=(0,1))
   return np.sum(np.multiply(imgwork, bilateralFilter),axis=(0,1))
def bilateralFilterConv(image, fw,fh):
   size = image.shape
   sigma1 = 40
   sigma2 = 40
   bilateral1 = 2*3.14*sigma2*sigma2*gaussFilter((fw,fh), sigma2)
   if len(image.shape) < 3 or image.shape[2] == 1:</pre>
       bilateralWFilter = np.resize(bilateral1,(*bilateral1.shape,1))
   else:
       bilateralWFilter = np.stack([bilateral1, bilateral1, bilateral1], axis=2)
   out = np.zeros((size[0]-2*fw +1,size[1]-2*fh +1,size[2]))
   for i in range(size[0]-2*fh +1):
       for j in range(size[1]-2*fw +1):
           out[i,j,:] = bfunc(i+fw-1, j+fh-1, fw, fh, image, sigma1, sigma2, bilateralWFilter)
   if id == 1:
       return np.resize(out, (out.shape[0], out.shape[1])).astype(np.uint8)
   else:
       return out.astype(np.uint8)
fig = plt.figure(1, figsize=(18, 16))
for i in range(3,8,2):
   splot = plt.subplot(130-1+((i+1)//2))
   splot.set_xlabel(str(i)+"x"+str(i))
   bfw,bfh = (i,i)
bilateral_image = bilateralFilterConv(img, bfw,bfh)
   imshow(bilateral image)
   cv2.imwrite("bilateral"+str(i)+"x"+str(i)+".jpg",bilateral image)
plt.show()
```


=====



Non Local Mean Filter and its example

nlmfunc:

Function that create Non Local Mean Filter according to image portion and give pixel value after applying it to image part.

Parameters:

• i,j: pixel value for which NLM filter will be applied

• **fw,fh**: Filter width and filter height

• **nw,nh**: neighbour width and height

• **image**: input image on which pixel (i,j) exist

• sigma1, sigma2: bilateral function parameters

• nlmWFilter: bilateral Filteral

nlmFilterConv:

Function that apply convolution on given image with the NLM filter of given size.

Parameters:

• image: input image on which NLM filter will be applied

nw,nh: neighbour width and heightfw,fh: widht and height if NLM filter

Return:

• returns a NLM filtered image (2d/3d matrix)

```
In [43]:
```

```
# Non Local Mean filter
def nlmfunc(i, j, fw, fh, nw, nh ,image,sigma1, sigma2, nlmWFilter):
    imgmain = image[i - fh//2:i+1 + fh//2, j - fw//2:j+1 + fw//2, :]
    nlmFilter = 0
    for p in range(-(nh//2), 1+(nh//2)):
        for q in range(-(nw//2), 1+(nw//2)):
            imgneighbour = image[i + p - fh//2: i+1+p + fh//2, j+q - fw//2:j+1+q + fw//2, :]
            nlmIFilter = ((imgmain - imgneighbour)**2 )/(2*(sigma1**2))
            nlmFilter += np.exp(-1*nlmIFilter)
    nlmFilter = nlmFilter/np.sum(nlmFilter,axis=(0,1))
    nlmFilter = nlmFilter*nlmWFilter
   nlmFilter = nlmFilter/np.sum(nlmFilter,axis=(0,1))
    return np.sum(np.multiply(imgmain, nlmFilter),axis=(0,1))
def nlmFilterConv(image, fw,fh, nw, nh):
    size = image.shape
    sigmal = 20
    sigma2 = 20
    nlmWFilter1 = 2*3.14*sigma2*sigma2*gaussFilter((fw,fh), sigma2)
   if len(image.shape) < 3 or image.shape[2] == 1:</pre>
        nlmWFilter = np.resize(nlmWFilter1,(*nlmWFilter1.shape,1))
    else:
        nlmWFilter = np.stack([nlmWFilter1, nlmWFilter1, nlmWFilter1], axis=2)
    out = np.zeros((size[0]-2*fw +1-nw//2, size[1]-2*fh +1-nh//2, size[2]))
    for i in range(nh//2, size[0]-2*fh +1-nh//2):
        for j in range(nw//2, size[1]-2*fw +1- nw//2):
            out[i,j,:] = nlmfunc(i+fw-1, j+fh-1, fw, fh, nw, nh, image, sigma1, sigma2, nlmWFilter)
    out[0:nh//2,:,:] = out[nh//2,:,:]
    out[:,0:nw//2,:] = out[:,nw//2,:,np.newaxis]
    if id == 1:
        return np.resize(out, (out.shape[0], out.shape[1])).astype(np.uint8)
   else:
        return out.astype(np.uint8)
fig = plt.figure(1, figsize=(18, 16))
                                        ======= NLM Filter ============")
print(" =====
for i in range(3,8,2):
    splot = plt.subplot(130-1+((i+1)//2))
    splot.set xlabel(str(i)+"x"+str(i))
    nlmfw, nlmfh = (i,i)
    nlm_image = nlmFilterConv(img, nlmfw,nlmfh, i, i)
    imshow(nlm image)
    cv2.imwrite("nlm"+str(i)+"x"+str(i)+".jpg",nlm_image)
plt.show()
                                        == NLM Filter ==
100
200
                                 200
```

Comparing results of all filters

300

400

500

300

400

500

All Comparisons were done using "No Reference Image Quality Assessment". **Non Local Mean** was found best among all filters. However, all filters give the good results and scores given by them are close to each others scores.

```
In [44]:
print("Original Image Score of the given image: 61.09182487181366")
print("3x3: Score of the given image: 17.65453329005433")
print("5x5: Score of the given image: 36.57977626774763")
print("7x7: Score of the given image: 43.03394739384518")
print("9x9: Score of the given image: 49.367030925596936")
print("11x11: Score of the given image: 57.75450589281954")
print("=========== Gaussian Filter========
print("3x3: Score of the given image: 17.598815117503335")
print("5x5: Score of the given image: 40.17641998372548")
print("7x7: Score of the given image: 53.51036518030489")
print("9x9: Score of the given image: 62.92074436030131")
print("11x11: Score of the given image: 69.7025214296061")
print("3x3: Score of the given image: 16.584317547327117")
print("5x5: Score of the given image: 38.824409207202365")
print("7x7: Score of the given image: 44.1374802117526")
print("9x9: Score of the given image: 50.30464610594149")
print("11x11: Score of the given image: 56.92915541182239")
print("3x3: Score of the given image: 16.369365771703586")
print("5x5: Score of the given image: 36.408236332713244")
print("7x7: Score of the given image: 43.01151838326959")
Original Image Score of the given image: 61.09182487181366
=========== Low Pass Filter=============
3x3: Score of the given image: 17.65453329005433
5x5: Score of the given image: 36.57977626774763
7x7: Score of the given image: 43.03394739384518
9x9: Score of the given image: 49.367030925596936
11x11: Score of the given image: 57.75450589281954
     ======= Gaussian Filter===
```

In [48]:

```
print("Garphs of Brisque Scores of Different Filters with there size")

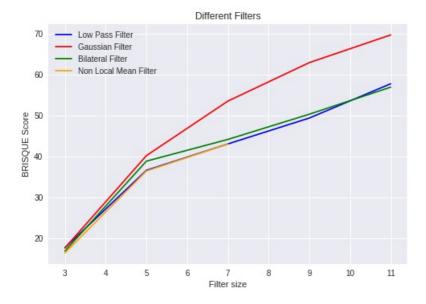
x = [3,5,7,9,11]
y1 = [17.65453329005433, 36.57977626774763, 43.03394739384518, 49.367030925596936, 57.75450589281954]
y2 = [17.598815117503335, 40.17641998372548, 53.51036518030489, 62.92074436030131, 69.7025214296061]
y3 = [16.83974876871534, 38.824409207202365, 44.1374802117526, 50.30464610594149, 56.92915541182239]

x4 = [3, 5, 7]
y4 = [16.369365771703586, 36.408236332713244, 43.01151838326959]

plt.title("Different Filters")
plt.ylabel("Filter size")
plt.ylabel("BRISQUE Score")
plt.ylabel("BRISQUE Score")
plt.plot(x,y1, color="blue", label="Low Pass Filter")
plt.plot(x,y2, color="red", label="Gaussian Filter")
plt.plot(x,y3, color="green", label="Bilateral Filter")
plt.plot(x,y4, color="orange", label="Non Local Mean Filter")
plt.legend(loc='upper left')
```

Garphs of Brisque Scores of Different Filters with there size Out[48]:

<matplotlib.legend.Legend at 0x7fe762e29e10>



References

• Brisque Scores was predicted using python repo which is implementation of paper "No Reference Image Quality Assessment in the Spatial Domain". "https://github.com/krshrimali/No-Reference-Image-Quality-Assessment-using-BRISQUE-Model (https://github.com/krshrimali/No-Reference-Image-Quality-Assessment-using-BRISQUE-Model)"