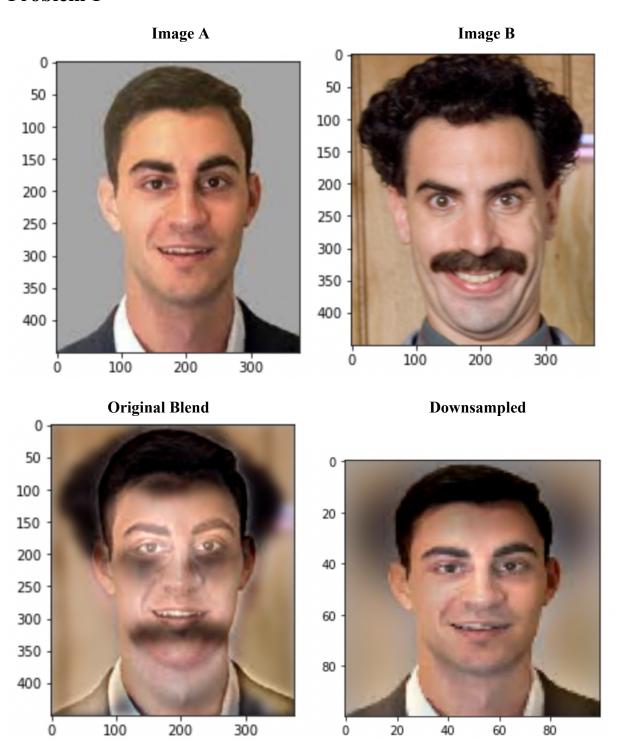
Problem Set 3 6.689 - Advances in Computer Vision

Problem 1

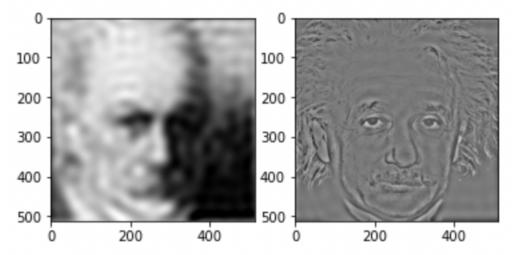


The image above used a Gaussian kernel with dimensions 128x128, mu=64, and sigma=32. The Gaussian seemed to work marginally better, and the value of sigma seemed somewhat irrelevant. The more blurring, the more the lines/edges and high special frequency of image A, and less of B will be prominent.

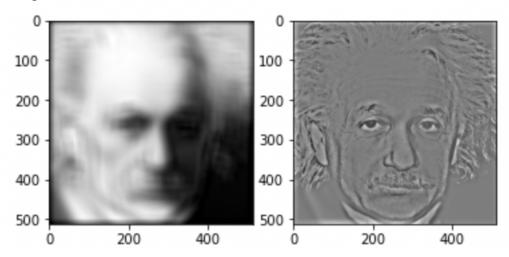
```
#creating a Gaussian filter
def Gauss kernel filter(kern dim, mu, sigma):
  # creating the kernel
  x, y = np.meshgrid(np.arange(kern dim), np.arange(kern dim))
  #creating the Gaussian filter with parameters mu, sigma, and coordinates
  Gauss filt = \exp(-1*((x-mu)**2+(y-mu)**2)/(2*sigma**2))
  return Gauss filt
def box kernel filter(kern xdim, kern ydim):
  return np.ones((kern ydim, kern xdim))/(kern xdim*kern ydim)
#blurring
def blurring(filter):
  blur = filter/np.sum(filter)
  return blur
#Convolving with the color channels
def color convolve(blur w filter, img):
  new img = img.copy()
  for color in range(3):
    new img[:,:,color] = conv2d(img[:,:,color], Gauss blur, mode='same')
    new img = new img.astype('int')
  return new img
GF = Gauss kernel filter(128,64,640)
\#BF = box kernel filter(3,3)
#Gauss blur = blurring(GF)
#Box blur = blurring(BF)
def hybrid(img1, img2, Filter):
  FB = blurring(Filter)
  blurry img1, blurry img2 = color convolve(FB, img1), color convolve(FB, img2)
  first term = blurry img2
  second term = img1-blurry img1
  hybrid img = first term+second term
  return hybrid img
test = hybrid(iggy, borat, GF)
plt.imshow(test)
```

Problem 2:

Using a circle mask with radius 13:



Using a Box Filter with dimensions 30x30:

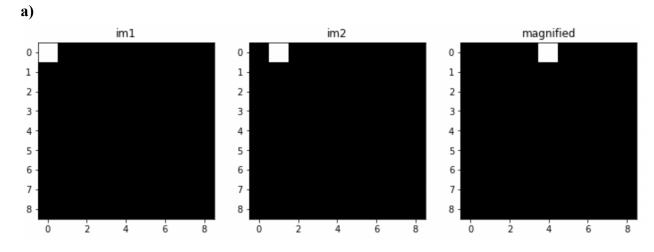


Is the hidden picture Gauss?

```
def circular mask(dimx, dimy, radius):
  x, y = np.meshgrid(np.arange(dimx), np.arange(dimy))
  mask = (sqrt((x-dimx/2)**2+(y-dimy/2)**2) <= radius)*1.0
  return mask
#circle mask = circular mask(stein.shape[0], stein.shape[1], 13)
def dehybrid(img, mask):
  low freq = fftshift(fft2(stein))*mask
  high freq = fftshift(fft2(stein))-low freq
  img1 = intensityscale(real(ifft2(ifftshift(low freq))))
  img2 = intensityscale(real(ifft2(ifftshift(high freq))))
  plt.subplot(1, 2, 1)
  imshow(img1)
```

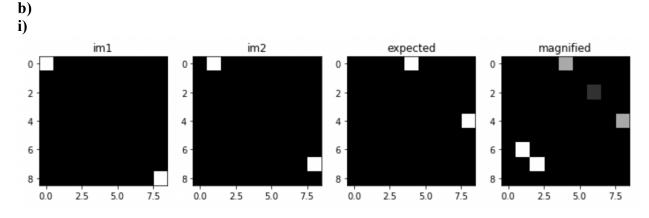
```
plt.subplot(1, 2, 2)
  imshow(img2)
for i in range(2,100):
  circle_mask = circular_mask(stein.shape[0], stein.shape[1], i)
  plt.figure()
  dehybrid(stein, circle mask)
for i in range(5,50,5):
  low_freq = conv2d(stein, box_kernel_filter(i,i), mode='same')
  high freq = stein - conv2d(stein, box kernel filter(i,i), mode='same')
  plt.figure()
  plt.subplot(1, 2, 1)
  imshow(intensityscale(low freq))
  plt.subplot(1, 2, 2)
  imshow(intensityscale(high freq))
```

Problem 3



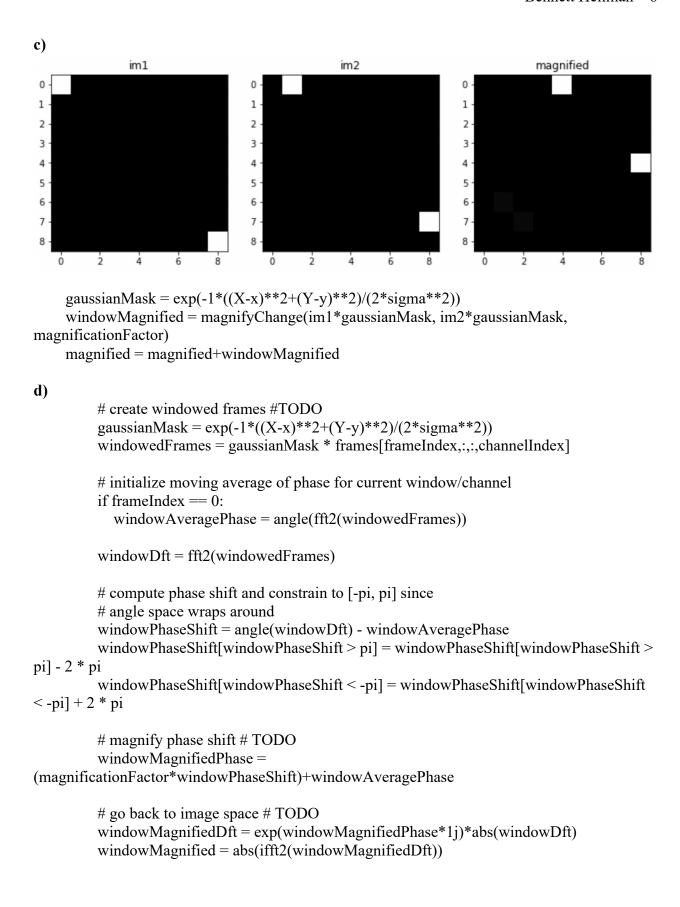
phaseShift = angle(im2Dft)-angle(im1Dft)

magnifiedDft = exp((magnificationFactor*phaseShift+angle(im1Dft))*1j)*abs(im2Dft)



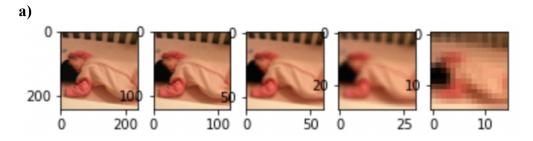
expected = np.zeros([imSize, imSize]) expected[0,1*magnificationFactor] = 1expected[1*magnificationFactor,8] = 1

ii) The key differences in the expected and magnified two squares in the bottom left and the darkest square around (x=6, y=2). The erroneous two white squares are caused by the cyclic nature of a Fourier. The square at (x=1, y=6) is caused by the top left square in image 1 moving vertically and wrapping around into the bottom of the second column. The square at (x=2, y=7) is a similar phenomenon but with the bottom right square moving horizontally and wrapping into the left side of the second to bottom row.

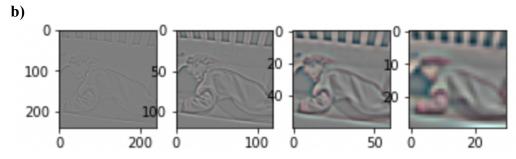


Problem 4)

c)



```
def create gaussian pyramid(vid, num levels=4):
 ### TODO: ENTER YOUR CODE BELOW
 ### return a list with the gaussian pyramid of the video.
 ### consider using the cv2.pyrDown function to create each level of the pyramid.
  pyr list = []
  pyr list.append(vid)
  for i in range(num levels):
    pyr list.append(np.array([cv2.pyrDown(pyr list[-1][j]) for j in range(frames.shape[0])]))
  return pyr list
```



```
def create laplacian pyramid(gaussian pyramid):
 ### TODO: ENTER YOUR CODE BELOW
 ### use the gaussian pyramid to create the laplacian pyramid for the video.
 ### You might find cv2.pyrUp function useful.
  lap pyr = []
  for i in range(len(gaussian_pyramid)-1):
    prev, next1 = gaussian pyramid[i], gaussian pyramid[i+1]
    next1 = np.array([cv2.pyrUp(j) for j in next1])
    diff = prev-next1
    lap pyr.append(diff)
  return lap pyr
```

```
b, a = signal.butter(filter order, [low, high], btype='band')
# filter the laplcian of the video using the signal.lfilter
y = signal.lfilter(b, a, laplace video, axis=0)
```

```
d)
bandpass filtered copy = bandpass filtered.copy()
for i in range(1,len(bandpass_filtered)):
   lvl = bandpass_filtered[-i]
   for j in range(lvl.shape[0]):
bandpass_filtered_copy[-i-1][j,:,:,:]+=cv2.pyrUp(lvl[j,:,:,:])
baby_euler_magnification = frames+bandpass_filtered_copy[0]
```