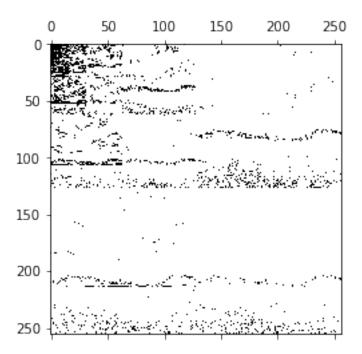
## **Uncompress Script**

## February 16, 2018

```
In [1]: # import the packages
        import numpy as np
        from scipy.misc import imread, imresize, imsave
        import matplotlib.pyplot as plt
        from numpy import matlib
        import math
        from scipy import stats
        import imageio
        from skimage.transform import resize
        import skimage
        import zlib, sys
        import gzip
        import matplotlib
        import scipy
        import copy
        import random
        import io
        import sys
In [2]: # define a function to covert the image to a gray scale image
        def rgb2gray(rgb):
            return np.dot(rgb[...,:3], [0.299, 0.587, 0.114])
        # define a function to get the proper Haar matrix and permutation matrix
        def GetHaarMatrices(N):
            Q = np.matrix("[1,1;1,-1]")
            M = int(N/2)
            T = np.kron(matlib.eye(M),Q)/np.sqrt(2)
            P = np.vstack((matlib.eye(N)[::2,:],matlib.eye(N)[1::2,:]))
            return T,P
In [3]: # use zlib to uncompress the compressed_data
        compressed_data = gzip.open('compressed_data.txt.gz', 'rb').read()
        decompress_data = zlib.decompress(compressed_data)
        # convert the byte-like object to numpy array
        decompress_data = np.frombuffer(decompress_data, dtype=int)
```

```
# reshape the data to 2D
indices = np.reshape(decompress_data, (256, 256))
# show the image before reverse log quantization
plt.spy(indices)
plt.show()
print(indices)
```



```
50 ...
[[127 31
                              0]
[ 64
      63
           46 ...
                              0]
[ 71
       81
           26 ...
                              0]
                              0]
        0
            0 ...
                     0
        0
            0 ...
                     0
                              0]
                              0]]
            0 ...
```

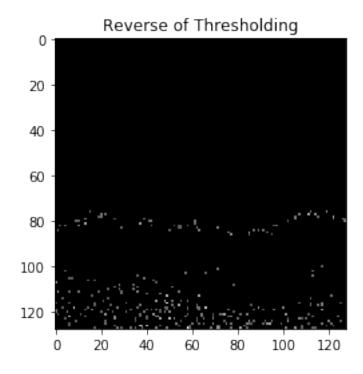
```
In [5]: # using codebook and indices to recover the quanta data
       quanta = np.empty((256, 256))
       for i in range(quanta.shape[0]):
           for j in range(quanta.shape[1]):
              quanta[i][j] = codebook[indices[i][j]]
       print(quanta)
[[130.59818677
               0.79415368
                           2.19811157 ...
                                           0.
                                                       0.
   0.
                           1.76914495 ...
[ 4.63455098
               4.45363012
                                                       0.
                                           0.
 [ 6.71842264 11.68084684
                           0.62617807 ...
                                           0.
                                                       0.
            ]
   0.
Γ 0.
               0.
                           0.
                                           0.
                                                       0.
                                      . . .
            1
   0.
               0.
[ 0.16562914
                           0.
                                           0.
                                                       0.
   0.
            ]
Γ 0.
               0.
                           0.
                                           0.
                                                       0.
   0.
            ]]
In [6]: # reverse threshold to F
       # make a deep copy of F as G
       G = copy.deepcopy(quanta)
       # read in F row by row, find the min nonzero pixel
       # put the number from data codebook before apply thresholding function
       # in order to put the data back to nonzero
       def reverse_thresholding(source):
           index = 0
           for i in range(source.shape[0]):
              for j in range(source.shape[1]):
                  if source[i][j] == 0:
                      index += 1
                  else:
                      continue
```

print(codebook)

## # Apply reverse thresholding function to M reverse\_thresholding(G)

```
# show the image after apply to reverse threshold
plt.imshow(G[128:256,128:256], cmap = plt.get_cmap('gray'))
plt.title("Reverse of Thresholding")
plt.show()
```

## print(G)



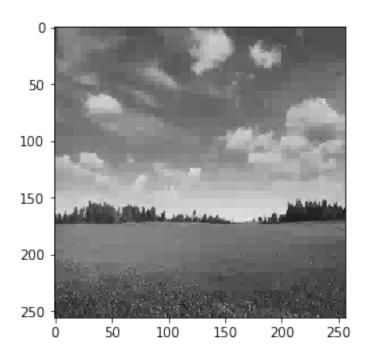
[[130.59818677		0.79415368	2.19811157		0.	0.
	0. ]					
[	4.63455098	4.45363012	1.76914495		0.	0.
	0. ]					
[	6.71842264	11.68084684	0.62617807		0.	0.
	0. ]					
•••						
[	0.	0.	0.		0.	0.
	0. ]					
[	0.16562914	0.	0.		0.	0.
	0. ]					
[	0.	0.	0.		0.	0.
	0. ]	]				

```
In [7]: # open the sign.txt to put back the sign
       sign = open('sign.txt', 'rb').read()
        # convert the byte-like object to numpy array
       sign = np.frombuffer(sign)
        # reshape the sign to 2D numpy array
        sign = np.reshape(sign, (256, 256))
       print(sign)
[[ 1. 1. 1. ... 1. -1. -1.]
 [ 1. -1. -1. ... 1. -1. -1.]
[-1. -1. -1. ... 1. -1. -1.]
 [ 1. -1. -1. ... 1. 1. -1.]
 [ 1. -1. -1. ... 1. 1. -1.]
 [-1. -1. -1. ... -1. 1. -1.]]
In [8]: # put the nagative sign back to the correct position
       G = G * sign
       print(G)
ΓΓ130.59818677
                0.79415368 2.19811157 ... 0.
                                                          -0.
 [ 4.63455098 -4.45363012 -1.76914495 ...
                                              0.
                                                          -0.
             1
 [ -6.71842264 -11.68084684 -0.62617807 ...
                                                          -0.
                                              0.
  -0.
            1
 [ 0.
               -0.
                           -0.
                                        . . .
                                              0.
                                                           0.
  -0.
             ]
 [ 0.16562914 -0.
                           -0.
                                                           0.
                                         ... 0.
  -0.
            ]
 [ -0.
               -0.
                           -0.
                                         ... -0.
                                                           0.
  -0.
             ]]
In [9]: # make a deep copy of G
       J = copy.deepcopy(G)
        # get number of times of decoding and the starting point
       N = len(J)
       times = int(np.log2(N))
       start = 2
        # Doing full-level decoding (Backward Haar Transform)
       for i in range(times):
           T,P = GetHaarMatrices(start)
```

```
J[0:start, 0:start] = T.T*P.T*J[0:start, 0:start]*P*T
    start = 2 * start

# show the result of full-level decoding
plt.figure()
plt.imshow(J, cmap = plt.get_cmap('gray'))
plt.show()

# print the info of J
print(J)
```



```
[[0.31332279 0.31332279 0.43884974 ... 0.31523406 0.31523406 0.31523406]
[0.31332279 0.31332279 0.43884974 ... 0.31523406 0.31523406 0.31523406]
[0.31332279 0.31332279 0.43884974 ... 0.31523406 0.31523406 0.31523406]
...
[0.17002039 0.17002039 0.25283496 ... 0.28278207 0.1971067 0.1971067]
[0.25283496 0.25283496 0.25283496 ... 0.28278207 0.1971067 0.1971067]
[0.25283496 0.25283496 0.25283496 ... 0.28278207 0.1971067 0.1971067]
```

# get the information from the original image(before full-level encoding)
# reads in the original image

```
A = imageio.imread('image.jpg')
         # resize the image(before apply gray scale function) as a 256 by 256 matrix
         A = skimage.transform.resize(A, [256, 256], mode='constant')
         # Apply the rgb2gray function to the image
         A = rgb2gray(A)
         print(A)
 \begin{bmatrix} [0.31905515 \ 0.3648937 \ 0.42410076 \ \dots \ 0.35208627 \ 0.35567083 \ 0.35600784] \\ \end{bmatrix} 
 [0.31661375 0.36307557 0.43115786 ... 0.35208627 0.35567083 0.35600784]
 [0.31623509 0.3615686 0.43780567 ... 0.35208627 0.35567083 0.35600784]
 [0.23155268 0.1562254 0.29268602 ... 0.31176704 0.14906503 0.21840392]
  [0.23837193 \ 0.19653385 \ 0.30720091 \ \dots \ 0.28660407 \ 0.13619611 \ 0.25433132] 
 [0.31247219 0.25910593 0.29068716 ... 0.39854037 0.252657 0.21340572]]
In [11]: # get the maximum value of the original image
         maxValue = np.amax(A)
         print(maxValue)
0.999999999999999
In [12]: # get the 2D info of original image
         print(A)
[[0.31905515 0.3648937 0.42410076 ... 0.35208627 0.35567083 0.35600784]
 [0.31661375 \ 0.36307557 \ 0.43115786 \ \dots \ 0.35208627 \ 0.35567083 \ 0.35600784]
 [0.31623509 0.3615686 0.43780567 ... 0.35208627 0.35567083 0.35600784]
 [0.23155268 \ 0.1562254 \ 0.29268602 \dots 0.31176704 \ 0.14906503 \ 0.21840392]
 [0.23837193 0.19653385 0.30720091 ... 0.28660407 0.13619611 0.25433132]
 [0.31247219 0.25910593 0.29068716 ... 0.39854037 0.252657 0.21340572]]
In [13]: # get the 2D info of the reconstructed image
         print(J)
[[0.31332279 0.31332279 0.43884974 ... 0.31523406 0.31523406 0.31523406]
 [0.31332279 \ 0.31332279 \ 0.43884974 \ \dots \ 0.31523406 \ 0.31523406 \ 0.31523406]
 [0.31332279 \ 0.31332279 \ 0.43884974 \ \dots \ 0.31523406 \ 0.31523406 \ 0.31523406]
 [0.17002039 0.17002039 0.25283496 ... 0.28278207 0.1971067 0.1971067 ]
 [0.25283496 0.25283496 0.25283496 ... 0.28278207 0.1971067 0.1971067 ]
 [0.25283496 0.25283496 0.25283496 ... 0.28278207 0.1971067 0.1971067 ]]
```

```
In [14]: # calculate the MSE
         MSE_arr = np.empty([J.shape[0], J.shape[1]])
         for i in range(J.shape[0]):
             for j in range(J.shape[1]):
                 MSE_arr[i][j] = ((A[i][j] - J[i][j])**2)
         MSE = 0
         for a in range(MSE_arr.shape[0]):
             for b in range(MSE_arr.shape[1]):
                 MSE += MSE_arr[a][b]
         MSE = MSE/(MSE_arr.shape[0]*MSE_arr.shape[1])
         print(MSE)
0.0024329983400938992
In [15]: # calculate the PSNR
         PSNR = 20*math.log10(maxValue) - 10*math.log10(MSE)
         print(PSNR)
26.138581873652385
```