



















1. Average accuracy on the first 500 images: 99.48571%

```
: #Report the fraction of all pixels that are correct in the 500 images.  
print("The accuracy between true images and denoised image :", (np.sum(pic==denoised_images)/(784*500))*100, "%")  
The accuracy between true images and denoised image : 99.4857142857143 %
```

2. One set of sample images for each digit -- For each digit you should put up a row of a (sample image, noisy version, denoised version via MFI). This means should have a total of 30 images (10 rows, 3 columns).

The Image is reconstructed with $\theta_{ij}(H_i, H_j)=0.2$ and $\theta_{ij}(H_i, X_j)=0.2$

I took the best reconstructed images.

True Image	Noisy Image	Reconstructed Image
		
		
		
		
		
		

True Image



Noisy Image



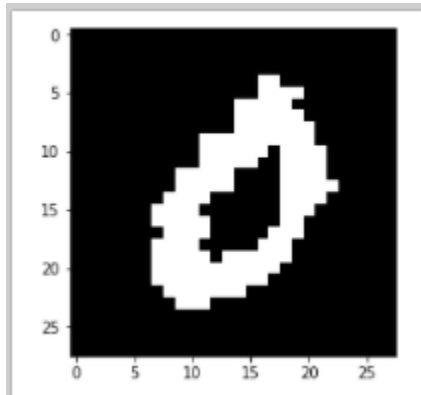
Reconstructed Image



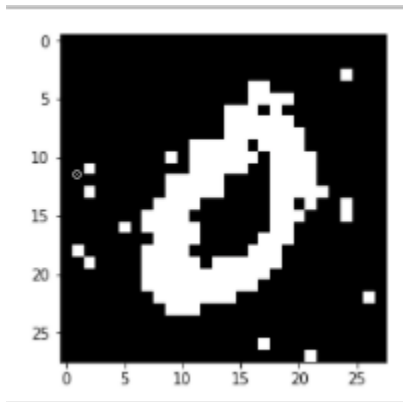
3. Best reconstruction (original, noisy, denoised):

The Image is reconstructed with $\theta_{ij}(H_i, H_j)=0.2$ and $\theta_{ij}(H_i, X_j)=0.2$

Original iMAGE

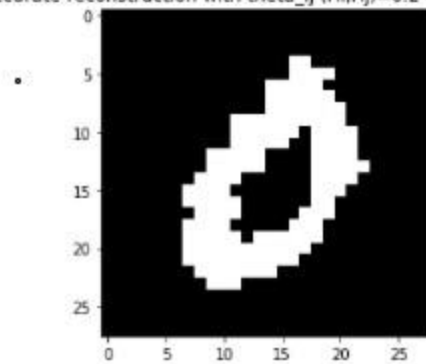


Noisy



Reconstructed

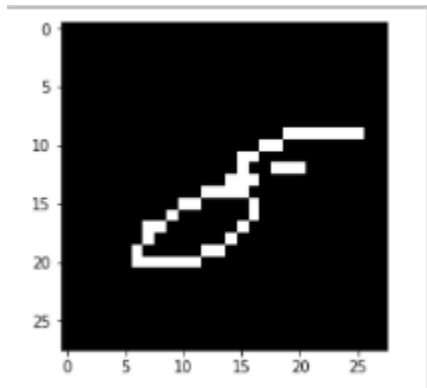
The most accurate reconstruction with $\theta_{ij}(H_i, H_j)=0.2$ and $\theta_{ij}(H_i, X_j)=0.2$



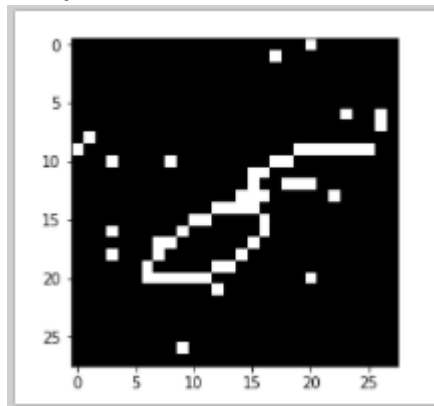
4 . Worst reconstruction (original, noisy, denoised)

The Image is reconstructed with $\theta_{ij}(H_i, H_j)=0.2$ and $\theta_{ij}(H_i, X_j)=0.2$

Original

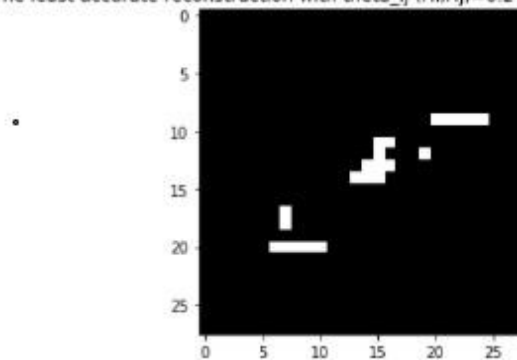


Noisy



Reconstructed

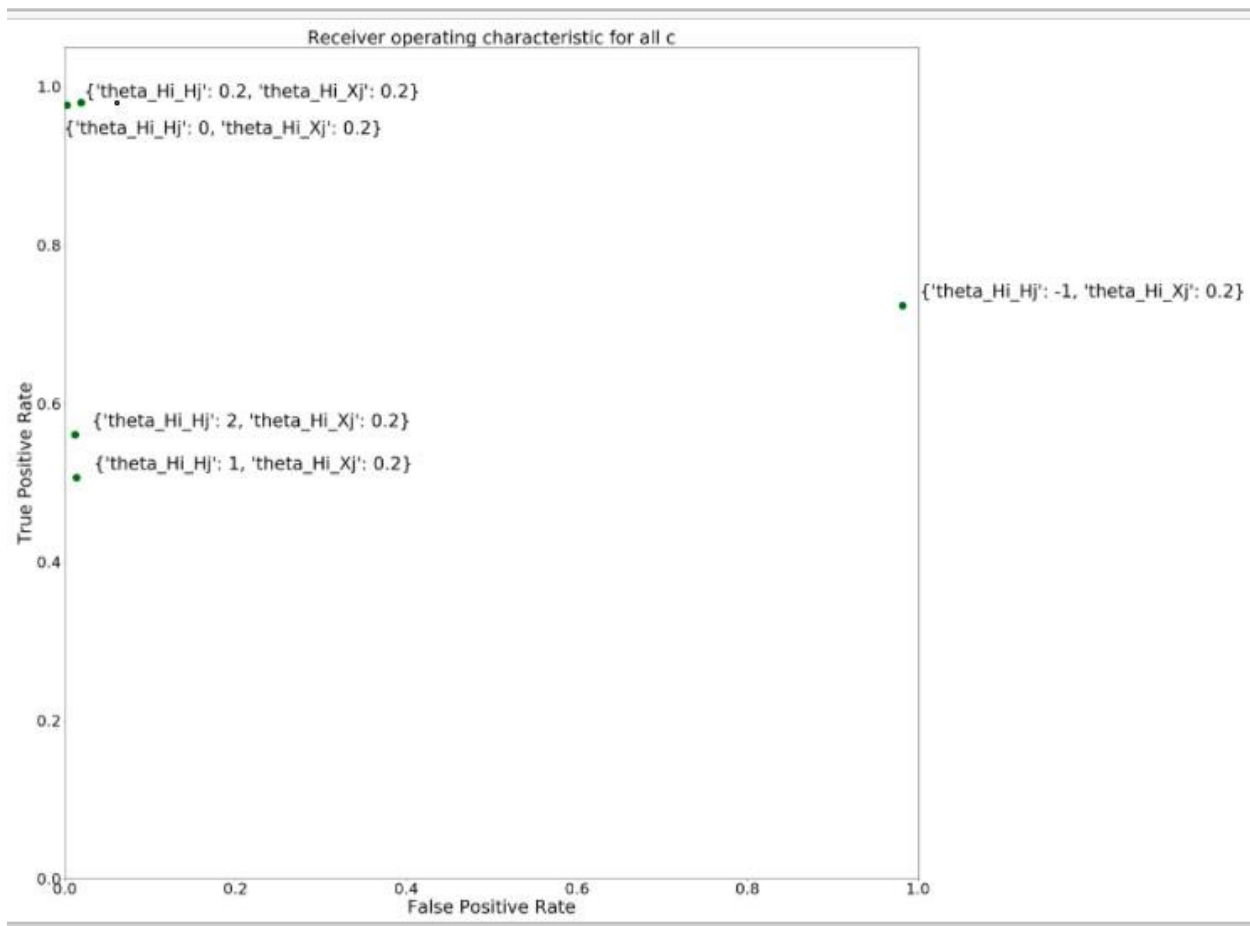
The least accurate reconstruction with $\theta_{ij}(H_i, H_j)=0.2$ and $\theta_{ij}(H_i, X_j)=0.2$



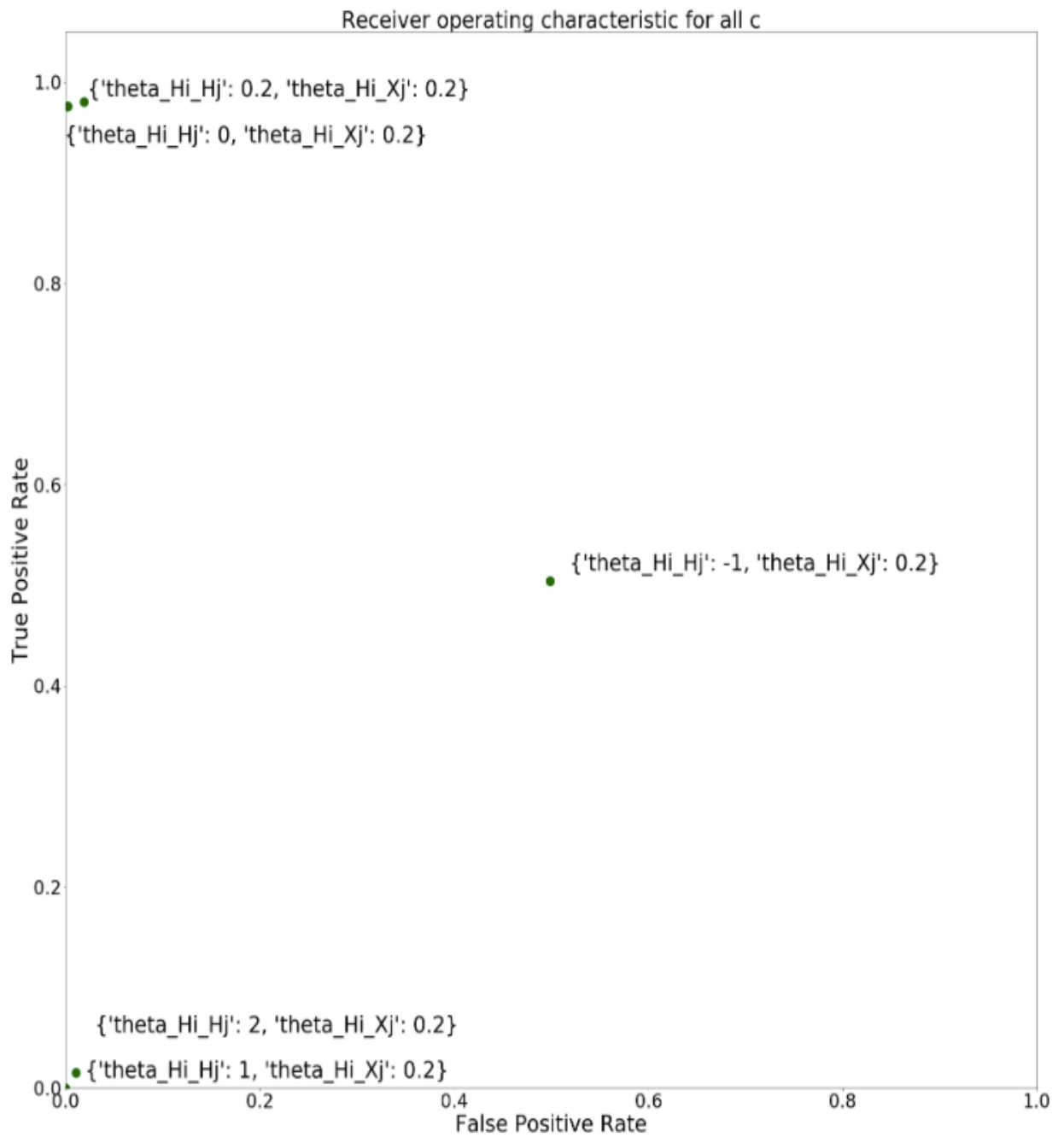
5. ROC curve:

Two Instances;

1. When I don't use updated value of π (While updating π for every pixel in a particular iteration)



2. When I use the updated value of pi ((While updating pi for every pixel in a particular iteration))



6. Code snippet:

Importing Libraries

```
In [1]: import os
import struct
import numpy as np
import pandas as pd
%matplotlib inline
import matplotlib.pyplot as plt
#import tensorflow as tf
import numpy as np
from sklearn.metrics import confusion_matrix
import matplotlib.gridspec as gridspec
|
```

Binarize the Data set

```
def binarize_dataset(X):
    for i in range(0,X.shape[0]):
        for j in range(0,X.shape[1]):
            if(X[i,j]>127):
                X[i,j]=1
            else:
                X[i,j]=-1
    return X
```

Adding random Noise to the data set

```
In [11]: #Create Noisy Data (Randomly)

X_train_noisy=np.copy(Train[:,1:785]).values
y_train_noisy=np.copy(Train[:,0]).values

In [12]: for i in range(0,X_train.shape[0]):
bits_to_flip = np.random.choice(784,15,replace=False)
for j in range(len(bits_to_flip)):
    X_train_noisy[i,bits_to_flip[j]] = (X_train_noisy[i,bits_to_flip[j]])*-1
```


Boltzmann Denoising Calculation (When I don't use the updated value of pi)

```
#Denoising the image
def boltzman(pic,theta_H,theta_X):

    pi_output=[] for k in range(500)]

    min_error=0.1

    for i in range(len(pic)):
        pi=np.ones((28,28))*0.5
        pi_prev=np.ones((28,28))*0.5

        for l in range(30):
            for j in range(pic[i].shape[0]):
                for k in range(pic[i].shape[0]):
                    z=0
                    if(j>0):
                        z=z+((theta_H)*((2*pi_prev[j-1,k])-1))

                    if(j<27):
                        z=z+((theta_H)*((2*pi_prev[j+1,k])-1))

                    if(k>0):
                        z=z+((theta_H)*((2*pi_prev[j,k-1])-1))

                    if(k<27):
                        z=z+((theta_H)*((2*pi_prev[j,k+1])-1))

                    z=z+(pic[i,j,k]*(theta_X))

                    pi[j,k]=(np.exp(z))/((np.exp(z)+np.exp(-z)))

                pi_prev=np.copy(pi)
            pi_output[i]=pi

    return pi_output
```

Boltzamn Denoising Calculation (When I use the updated value of pi)

```

1: #Denoising the image
def boltzman(pic,theta_H,theta_X):

    pi_output=[] for k in range(500)]

    min_error=0.1

    for i in range(len(pic)):
        pi=np.ones((28,28))*0.5
        pi_prev=np.ones((28,28))*0.5

        for l in range(50):
            for j in range(pic[i].shape[0]):
                for k in range(pic[i].shape[0]):
                    z=0
                    if(j>0):

                        z=z+((theta_H)*((2*pi[j-1,k])-1))

                    if(j<27):

                        z=z+((theta_H)*((2*pi[j+1,k])-1))

                    if(k>0):

                        z=z+((theta_H)*((2*pi[j,k-1])-1))

                    if(k<27):

                        z=z+((theta_H)*((2*pi[j,k+1])-1))

                    z=z+(pic[i,j,k]*(theta_X))

                    pi[j,k]=(np.exp(z))/((np.exp(z)+np.exp(-z)))

                pi_prev=pi.copy(pi)
            pi_output[i]=pi

    return pi_output

```

Boltzman Output

```
: #output->Boltzman
def result(pi_updated):
    for i in range(500):
        for j in range(28):
            for k in range(28):
                if(pi_updated[i,j,k]>0.5):
                    pi[i,j,k]=1
                else:
                    pi_updated[i,j,k]=-1
    return pi
```

Creating the ROC Curves

```
#ROC Curve

# theta_ij (Hi,Hj)(Xi,Xj)[(-1,0.2),(0,0.2),(0.2,0.2),(1,0.2),(2,0.2)]

R_P=[]
R_F=[]

for theta in [(-1,0.2),(0,0.2),(0.2,0.2),(1,0.2),(2,0.2)]:
    tp=0
    fp=0
    labels=[]
    TPR=np.sum(pic==1)
    FPR=np.sum(pic==1)
    pi=np.array(boltzman(pic_noisy,theta[0],theta[1]))
    denoised_images=np.array(result(pi))
    for i in range(500):
        for j in range(28):
            for k in range(28):
                if(denoised_images[i,j,k]==1 and pic[i,j,k]==1):
                    tp=tp+1
                elif(denoised_images[i,j,k]==1 and pic[i,j,k]==-1):
                    fp=fp+1

    R_P.append(tp/TPR)
    print("True Positive",tp)
    R_F.append(fp/FPR)
    print("False Positive",fp)
```

```
plt.figure()
lw = 2
plt.rcParams['figure.figsize'] = [30,30]
types = [{"theta_Hi_Hj":-1,"theta_Hi_Xj":0.2}, {"theta_Hi_Hj":0,"theta_Hi_Xj":0.2}, {"theta_Hi_Hj":0.2,"theta_Hi_Xj":0.2}, {"theta_Hi_Hj":1,"theta_Hi_Xj":0.2}, {"theta_Hi_Hj":2,"theta_Hi_Xj":0.2}]
plt.plot([0, 1], [0, 1], color='navy', lw=lw, linestyle='--')
plt.scatter(R_F,R_P, color='darkgreen',s=200)
plt.rc('xtick', labels=30)
plt.rc('ytick', labels=30)
plt.xlim([0.0, 1.0])
plt.ylim([0.0, 1.05])
plt.xlabel('False Positive Rate',fontSize=35)
plt.ylabel('True Positive Rate',fontSize=35)
plt.title('Receiver operating characteristic for all c',fontSize=35)

for i in range(len(R_P)):
    if i==1:
        plt.text(R_F[i]-0.02, R_P[i]-0.04, types[i], fontsize=35)
        continue
    plt.text(R_F[i]+0.02, R_P[i]+0.01, types[i], fontsize=35)

plt.show()
```