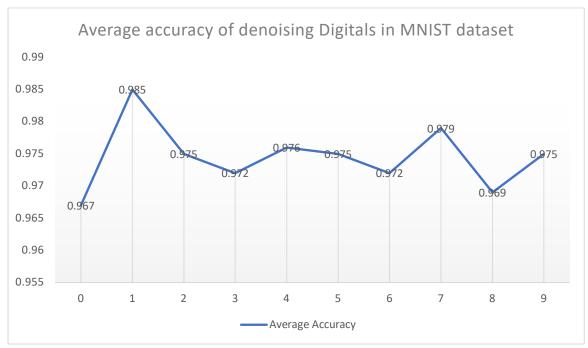
# **CS498: AML HOMEWORK 9**

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# 1. Average accuracy on the first 500 images

(Table 1) Average accuracy of denoising Digitals in MINST dataset

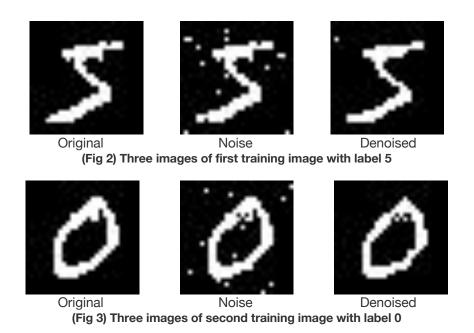
| Original<br>Digital | 0     | 1     | 2     | 3     | 4     | 5     | 6     | 7     | 8     | 9     |
|---------------------|-------|-------|-------|-------|-------|-------|-------|-------|-------|-------|
| Average<br>Accuracy | 0.967 | 0.985 | 0.975 | 0.972 | 0.976 | 0.975 | 0.972 | 0.979 | 0.969 | 0.975 |



(Fig 1) Average accuracy of denoising Digitals in MINST dataset

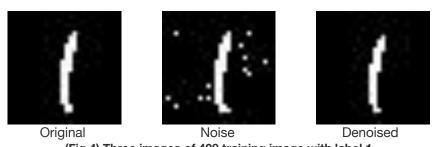
After analyzing this result, we can find it is easier to denoise easy-writing digitals, such as '1' or '7', which only have vertical line or horizontal line or both. Since the flipped points won't be likely to influence the true points.

## 2. Sample images (original, noisy, denoised)



Each image with noise has 20 random flipped points.

# **3.** Best reconstruction (original, noisy, denoised)



(Fig 4) Three images of 408 training image with label 1

After constructing mean field inference, our denoised image has 0.994 of accuracy.

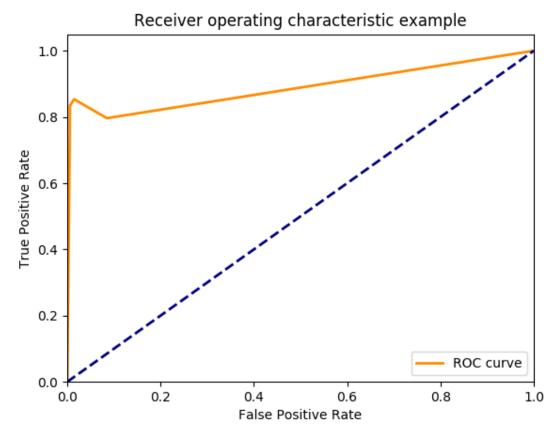
## 4. Worst reconstruction (original, noisy, denoised)



(Fig 5) Three images of 404 training image with label 8

After constructing mean field inference, our denoised image has 0.954 of accuracy.

#### **5.** ROC curve



(Fig 6) ROC of denoising flipped MINST dataset

In the experiment, we get the TPR and FPR via different c is shown below.

(Table 2) TPR and FPR of different c

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|-----|-------|----------------------|---------------------|-------|-------|
| c   | -1    | 0                    | 0.2                 | 1     | 2     |
| TPR | 0.744 | 0.836                | 0.836               | 0.854 | 0.796 |
| FPR | 0.005 | 0.006                | 0.007               | 0.015 | 0.085 |

#### 7. Code snippet

```
import imageio
import matplotlib.pyplot as plt
def getData():
    (x_train, y_train),(x_test, y_test) = mnist.load_data()
          for i in range(len(x)):
              for j in range(len(x[0])):
    x[i][j] = 1 if x[i][j]>=0.5 else -1
     return x_train[:NUM], y_train[:NUM]
def addNoise(data):
     for i in range(len(data)):
          flippings = [np.random.randint(points) for i in range(int(points*0.02))]
          for pos in flippings:
    global rows, columns, points
pos = row*columns+column
    neighbors = list()
    if left[1]>=0: neighbors.append(pos-1)
    if up[0]>=0: neighbors.append(pos-columns)
     if right[1]<=columns-1: neighbors.append(pos+1)</pre>
    if down[0]<=rows-1: neighbors.append(pos+column)</pre>
     for neighbor in neighbors:
         j_row, j_column = divmod(neighbor, columns)
A += thetaHH*(1-2*pai[neighbor]) + thetaHX*(-1)*pic[j_row][j_column]
B += -(thetaHH*(1-2*pai[neighbor]) + thetaHX*(-1)*pic[j_row][j_column])
     return (math.exp(A)/(math.exp(A)+math.exp(B)))
def getTprFpr(predict, origin):
     for i in range(len(predict)):
          if predict[i]==1 and origin[i]==-1:
     return TP/(TP+FN), FP/(FP+TN)
```

```
global rows, columns, points, data, labels, accuracy_list, best_accuracy, worst_accuracy, best_pic, worst_pic
    data recoverd = list()
   count = 0
    for pic in data flipped:
        pai = np.full((1, points),.5)[0]
        prev_sum_pai = float('inf')
        times = 0
        while abs(prev_sum_pai-pai.sum()) >= threshold and times<10:</pre>
            prev_sum_pai = pai.sum()
             for i in range(points):
                row, column = divmod(i, columns)
                pai[i] = update(row, column, pic, pai, thetaHH, thetaHX)
        for i in range(rows):
                 pic_recoverd[i][j] = -1 if pai[pos]>=.5 else 1
        if accuracy >= best_accuracy:
            best_accuracy = accuracy
            best_pic = count
        if accuracy <= worst_accuracy:</pre>
           worst_accuracy = accuracy
           worst_pic = count
    for i in range(len(a)):
                 count += 1
    return count
def saveImages(data, name):
        imageio.imwrite('./result/%s_%s.jpg'%(i,name), img)
def drawPicture(tpr, fpr):
   plt.figure()
    lw=lw, label='ROC curve')
plt.plot([0, 1], [0, 1], color='navy', lw=lw, linestyle='--')
plt.xlim([0.0, 1.0])
    plt.show()
        tpr = (tpr*count + cur_tpr)/(count+1)
fpr = (fpr*count + cur_fpr)/(count+1)
        count_list[label] += 1
        data_recoverd.append(pic_recoverd)
        count += 1
```

```
if __name__ == "__main__":
    thetaHHS = [.2]
    # thetaHHS = [.7], 0, 0.2, 1, 2]
    thetaHKS = 2.
    accuracy_lst = [0.]*10
    count_list = [0]*10
    best_accuracy, worst_accuracy = float('-inf'), float('inf')
    best_accuracy_list = [0.], [0.]

data, labels = getData()
    saveImages(data, 'original')
    rows, columns, points = len(data[0]), len(data[0])*len(data[0])*len(data[0])*len(data[0])0
    data_flipped = addWoise(data)
    saveImages(data_flipped, 'flipped')

for thetaHH in thetaHHS:
    print('='*40*'\nthetaHHS:
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    print('s-*40*'\nthetaHHS:
    print('s-*40*'\nthetaHHS:
    print('Nost_accuracy_list = [accuracy_list[i]/count_list[i] for in range(10)]

for in range(10):
    print('Num: %s\tCount: %s\tAccuracy: %s'%(i,count_list[i],average_accuracy_list[i]))

print('Num: %s\tCount: %s\tBest_accuract:%s'%(best_pic, best_accuracy))

tpr_list.append(1.)
    tpr_list.append(1.)
```