Problem 4. Nearest Neighbor Classification

Data Initialization

```
import numpy as np
from keras.datasets import mnist
from matplotlib import pyplot
from numpy.linalg import norm
(X train, y train), (X test,y test) = mnist.load data()
X train = X train[:600,:,:].astype("float32")/255 # Scale images to
the [0,1] range
y_train = y_train[:600]
N train = X train.shape[0]
X train = X train.reshape((N train, 28*28))
X test = X test[:100,:,:].astype("float32")/255 # Scale images to the
[0,1] range
y_test = y_test[:100]
N_{\text{test}} = X_{\text{test.shape}}[0]
X test = X test.reshape((N_test, 28*28))
print("Trainset X shape: " + str(X_train.shape))
print("train label y shape: " + str(y_train.shape))
print("Testset X shape: " + str(X test.shape))
print("test label y shape: " + str(y_test.shape))
Trainset X shape: (600, 784)
train label y shape: (600,)
Testset X shape: (100, 784)
test label y shape: (100,)
```

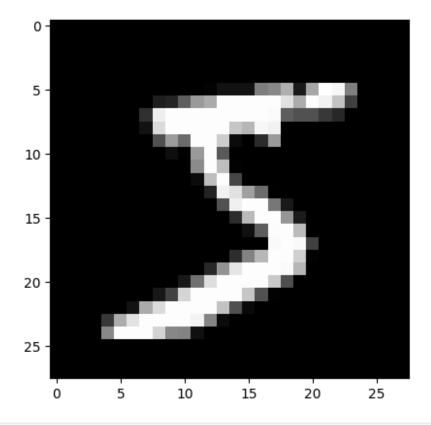
4.1 (1/5) Data Visualization Please use this provided function to solve 4.1

```
def visualization(idx_lst,data,label):
    This function is used to visualize original dataset X or cluster
    representative matrix Z

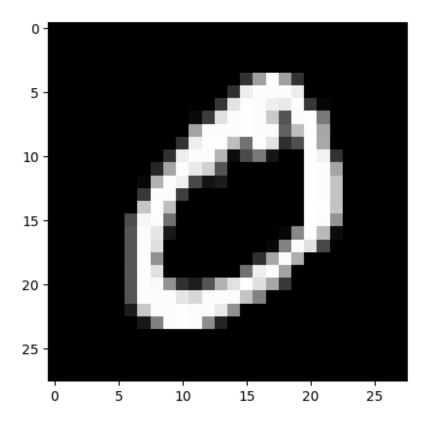
    N = data.shape[0]
    for idx in idx_lst:
        pyplot.imshow(data.reshape((N,28,28))[idx],
    cmap=pyplot.get_cmap('gray'))
        pyplot.show()
        print("The corresponding label of this image is
{}".format(label[idx]))
```

Example: visualize the first two images in the trainset:

$visualization(np.arange({\color{red}2})\,,X_train,y_train)$



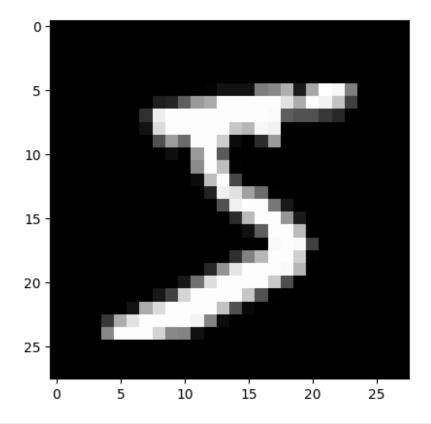
The corresponding label of this image is 5



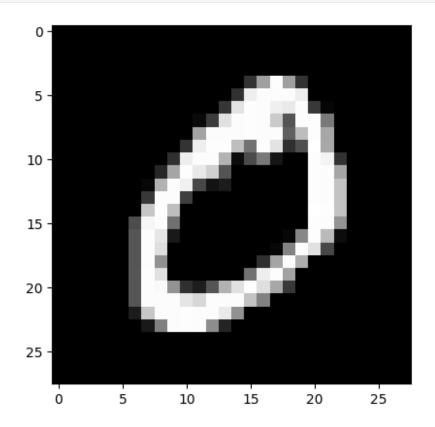
The corresponding label of this image is $\boldsymbol{\theta}$

Alternative way for the above example:

visualization([0,1],X_train,y_train)

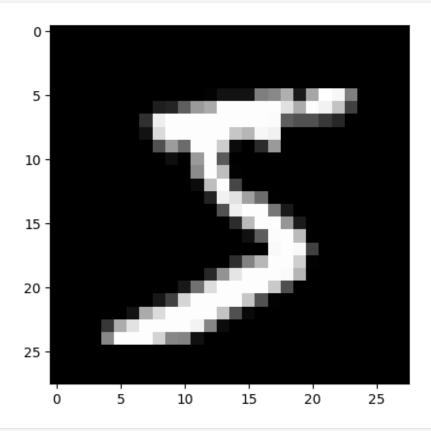


The corresponding label of this image is 5

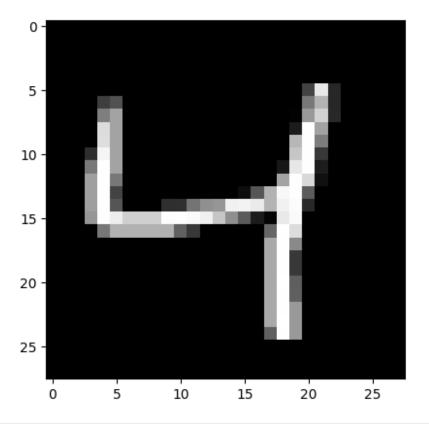


The corresponding label of this image is ${\tt 0}$

Visualize the first and third images (row vectors) in X_train: TODO

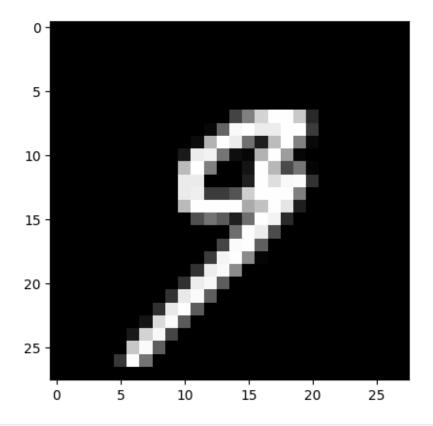


The corresponding label of this image is 5

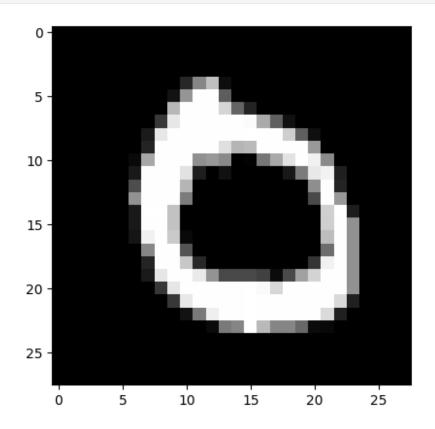


The corresponding label of this image is 4

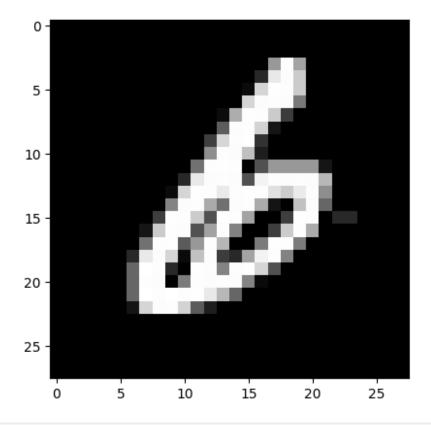
Visualize the last 5 images (row vectors) in X_test: TODO



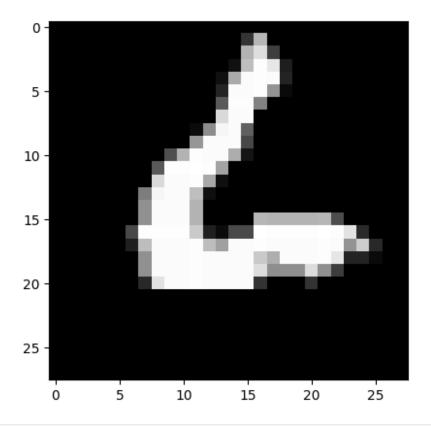
The corresponding label of this image is 9



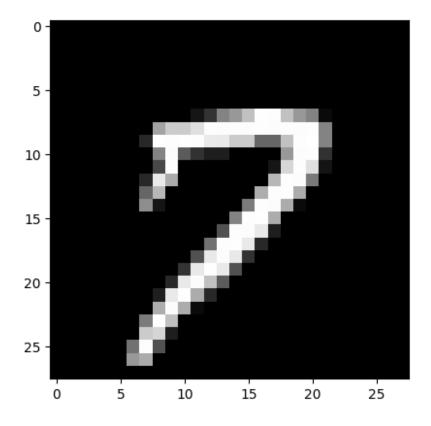
The corresponding label of this image is $\boldsymbol{\theta}$



The corresponding label of this image is 6



The corresponding label of this image is 6



The corresponding label of this image is 7

Write down their corresponding labels here: TODO

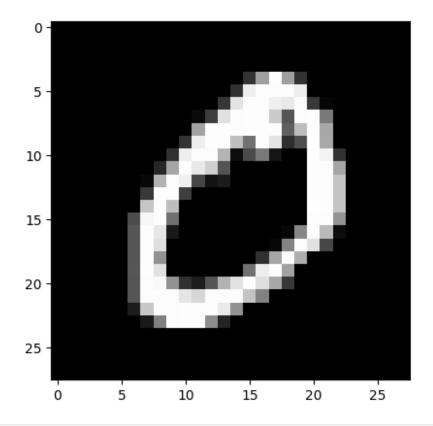
1st image: 5; 3rd image: 4; Last 5 images: 9,0,6,6,7

4.2 (1.5/5) Distance Calculation These three vectors represents three images

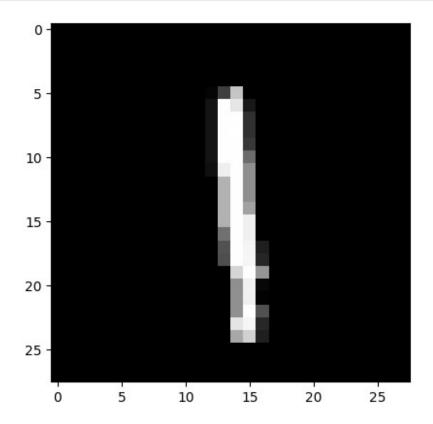
```
v1 = X train[1]
v2 = X train[8]
w = X \text{ test}[5]
distance v1 w = np.linalg.norm(v1 - w)
distance v2 w = np.linalg.norm(v2 - w)
print(f"This is distance from v1 to w = {distance v1 w}")
print(f"This is the distance from v2 to w = {distance_v2_w}")
# ANSWER: The meaning of distance in the problem is how closely
related v1 and v2 are to w.
  # The smaller distance means that it's closely related to w, and not
closely related if the distance is large.
print(v1.shape)
print(w.shape)
This is distance from v1 to w = 10.470195770263672
This is the distance from v2 to w = 7.066826820373535
(784,)
(784,)
```

Visualization them (in the order of v1, v2, w):

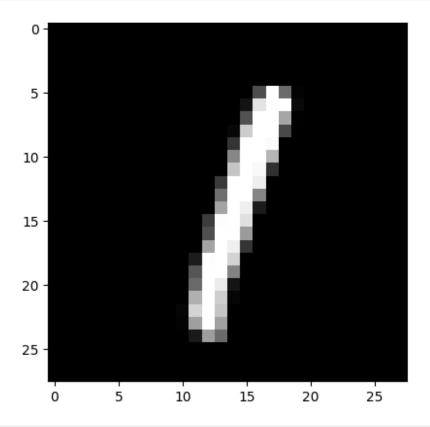
```
visualization([1,8],X_train,y_train)
visualization([5],X_test,y_test)
```



The corresponding label of this image is $\boldsymbol{\theta}$



The corresponding label of this image is 1



The corresponding label of this image is 1

Calculate distances between (1) v1 and w; (2) v2 and w: TODO

```
###### YOUR CODE STARTS HERE ######
dist1 = np.linalg.norm(v1 - w)
dist2 = np.linalg.norm(v2 - w)
###########################
print("dist between v1 and w = {}; \ndist between v2 and w =
{}".format(dist1,dist2))

dist between v1 and w = 10.470195770263672;
dist between v2 and w = 7.066826820373535
```

What does the comparison of distances tell us? Which two images are likely to be more similar to each other as told by the distances? You can upload the photo of your handwritten answer, or type your answer here directly: TODO

The comparision tells us how closely related v1 and v2 are to w. Image v2 is more likely to be more similar to w than v1 because of the smaller distance.

4.3 (2.5/5) KNN Implementation

Write your code: TODO

Validation code

```
# Updated here v1.6
y_predict = knn(X_train,y_train,X_test,y_test,7)
y_predict = y_predict.flatten()
predict_acc = np.sum((y_predict-y_test)==0)/N_test
print("The prediction accuracy = {}%, which should be greater than 80%".format(predict_acc*100))
The prediction accuracy = 83.0%, which should be greater than 80%
```