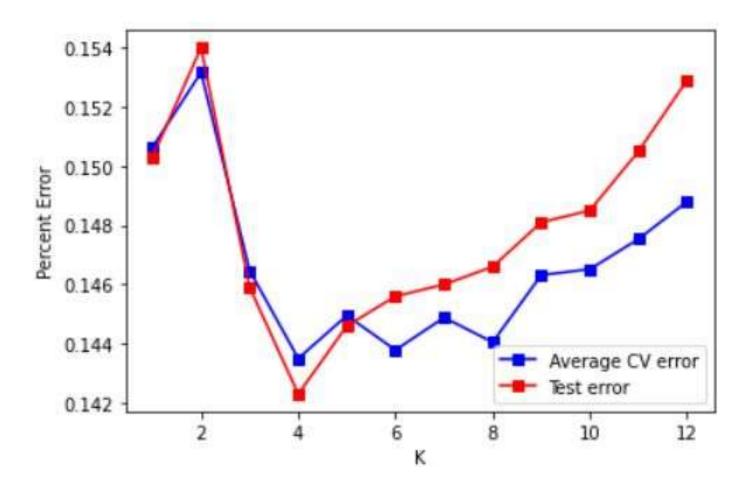
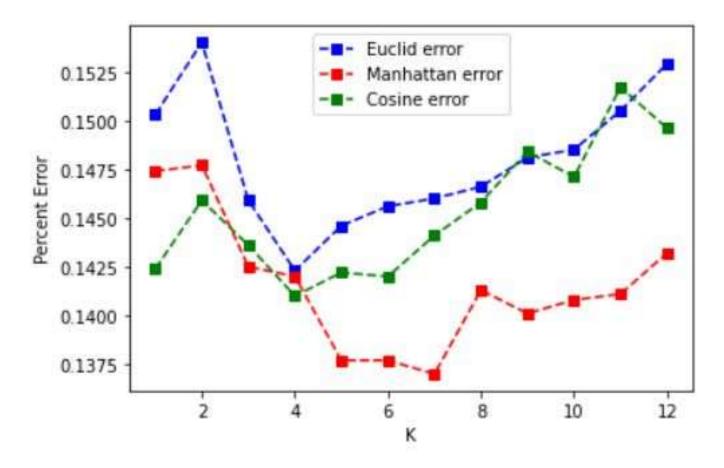
Cory Sweet

Math 251 Homework 1



Both curves have the lowest error rate at k=4. The error in terms is consistent on which k is best.

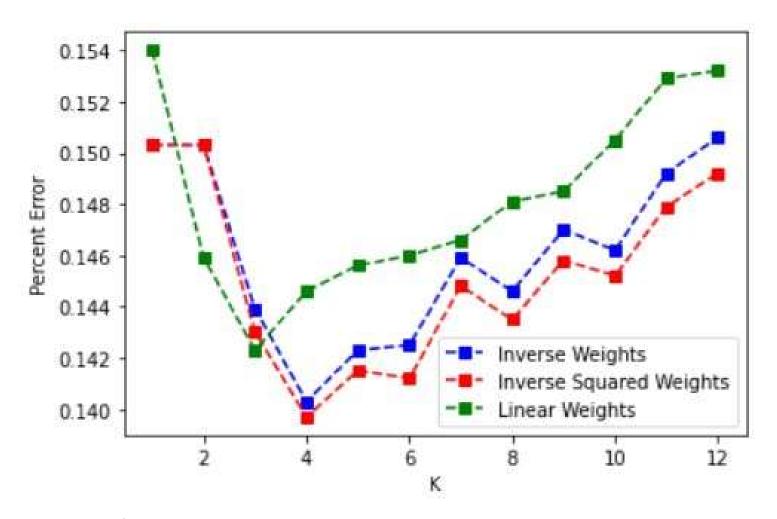
The smallest error rate for the test values was: 0.1423



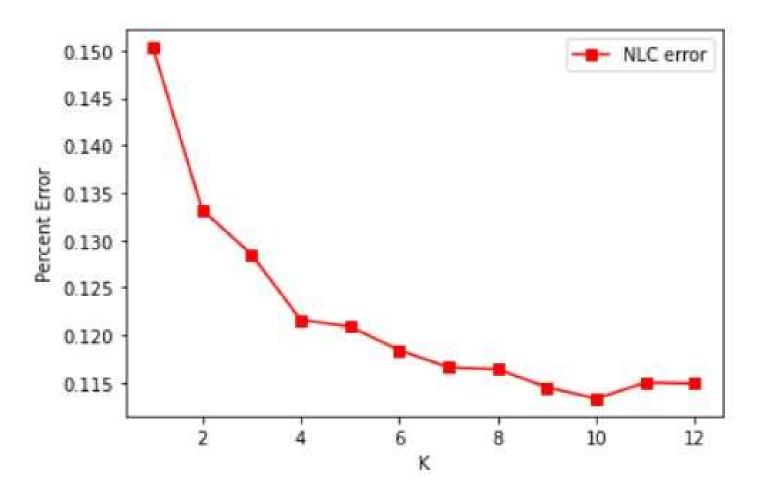
Manhattan distance produced the lowest error rate, and thus is best.

The lowest error rate was: 0.137

The lowest error rate was achieved at k=7.



Inverse Squared weights provided the smallest error. The smallest error was 0.1397 at k = 4



The smallest error was 0.1133 which was achieved at k=10

Question 4 - Confusion Matrix

Predictions

	[[830	1	12	23	4	2	120	0	8	0]
Actual	[1	977	0	16	2	0	3	0	1	0]
	[21	1	821	9	85	0	63	0	0	0]
	[25	4	15	896	35	0	23	0	2	0]
	[1	0	108	30	811	0	47	0	3	0]
	[0	0	0	0	0	959	0	20	0	21]
	[124	1	97	24	84	0	663	0	7	0]
	[0	0	0	0	0	16	0	960	0	24]
	[2	0	1	4	3	2	6	5	977	0]
	[0	0	0	0	0	6	0	20	1	973]]

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
from tensorflow import keras
from sklearn.neighbors import KNeighborsClassifier
from sklearn.model_selection import cross_val_score
from sklearn.metrics import accuracy score
from sklearn.metrics import pairwise_distances
from scipy.spatial.distance import cosine
fashion_mnist = keras.datasets.fashion_mnist
(train_images, train_labels), (test_images, test_labels) = fashion_mnist.load_data()
0
        T-shirt/top
1
        Trouser
2
        Pullover
3
        Dress
4
        Coat
5
        Sandal
6
        Shirt
7
        Sneaker
8
        Bag
9
        Ankle boot
label_names=['T-shirt/top','Trouser','Pullover','Dress','Coat','Sandal',
       'Shirt', 'Sneaker', 'Bag', 'Ankle boot']
X train = np.zeros([60000,784])
for i in range(60000):
  img=train_images[i,:,:]
  X_{train}[i,:] = img.reshape([784])
X_{\text{test}} = \text{np.zeros}([10000,784])
for i in range(10000):
  img=test images[i,:,:]
  X_test[i,:] = img.reshape([784])
X \text{ sub} = X \text{ train}[:600,:]
y sub = train labels[:600]
#cross val
cv acc=[]
for k in range(1,13):
  model = KNeighborsClassifier(n neighbors=k)
  scores = cross_val_score(model, X_train, train_labels, scoring='accuracy', cv=6)
  cv_acc.append(scores.mean())
  print('K is',k,'\n acc is:', scores.mean())
#on testing data
test_acc=[]
```

```
for k in range(1,13):
  model = KNeighborsClassifier(n neighbors=k)
  model.fit(X_train, train_labels)
  y preds = model.predict(X test)
  score=accuracy score(y true=test labels, y pred=y preds)
  test acc.append(score)
  print('K is',k,'\n acc is:', score)
#cv error = 1-np.array(cv acc)
#test_error = 1-np.array(test_acc)
"""My Kernel would often crash while doing these, so I recorded these
cv acc = [.84936666666666666666, .8468333333333332, .85353333333334, .8564999999999999,
     .85505, .8562166666666667, .8551166666666666, .85595, .8536833333333333,
     .853483333333335,.852466666666666, .8512]
test acc=[0.8497,0.846,0.8541,0.8577,0.8554,0.8544,
     0.854,0.8534,0.8519,0.8515,0.8495,0.8471]
cv_error = 1-np.array(cv_acc)
test_error = 1-np.array(test_acc)
#plots
plt.plot(range(1,13), cv error, '-bs', label='Average CV error')
plt.plot(range(1,13), test_error,'-rs', label = 'Test error')
plt.legend()
plt.xlabel('K')
plt.ylabel('Percent Error')
#euclid
euclid_acc=[]
for k in range(1,13):
  model = KNeighborsClassifier(n_neighbors=k,metric='euclidean')
  model.fit(X train, train labels)
  y preds = model.predict(X test)
  score=accuracy score(y true=test labels, y pred=y preds)
  euclid acc.append(score)
  print('K:',k,'\n Accuracy:', score)
#city-block
man_acc=[] #Manhattan dist
for k in range(1,13):
  model = KNeighborsClassifier(n neighbors=k, metric='manhattan')
  model.fit(X_train, train_labels)
  y preds = model.predict(X test)
  score = accuracy_score(y_true=test_labels, y_pred=y_preds)
  man acc.append(score)
  print('K:',k,'\n Accuracy:', score)
#cosine
cos_acc=[]
```

```
for k in range(1,13):
  model = KNeighborsClassifier(n neighbors=k, metric='cosine')
  model.fit(X_train, train_labels)
  y preds = model.predict(X test)
  score=accuracy score(y true=test labels, y pred=y preds)
  cos_acc.append(accuracy_score(y_true=test_labels, y_pred=y_preds))
  print('K:',k,'\n Accuracy:', score)
euclid_acc=[0.8497, 0.846, 0.8541, 0.8577, 0.8554, 0.8544,
      0.854, 0.8534, 0.8519, 0.8515, 0.8495, 0.8471
man acc=[0.8526, 0.8523, 0.8575, 0.858, 0.8623, 0.8623,
     0.863, 0.8587, 0.8599, 0.8592, 0.8589, 0.8568]
cos acc=[0.8576,0.8541,0.8564,0.859,0.8578,0.858,
     0.8559,0.8542,0.8516,0.8529,0.8483,0.8504]
euclid_error = 1-np.array(euclid_acc)
man error = 1-np.array(man acc)
cos_error = 1-np.array(cos_acc)
plt.plot(range(1,13), euclid_error, '--bs', label='Euclid error')
plt.plot(range(1,13), man_error,'--rs', label = 'Manhattan error')
plt.plot(range(1,13), cos_error,'--gs', label = 'Cosine error')
plt.legend()
plt.xlabel('K')
plt.ylabel('Percent Error')
#Inverse
inv_acc=[]
for k in range(1,13):
  model = KNeighborsClassifier(n_neighbors=k,weights='distance')
  model.fit(X train, train labels)
  y preds = model.predict(X test)
  score=accuracy score(y true=test labels, y pred=y preds)
  inv acc.append(score)
  print('K:',k,'\n Accuracy:', score)
#inverse squared
def dist inv sq(v):
  return 1/(v**2)
inv_sq_acc=[]
for k in range(1,13):
  model = KNeighborsClassifier(n_neighbors=k,weights=dist_inv_sq)
  model.fit(X_train, train_labels)
  y preds = model.predict(X test)
  score=accuracy_score(y_true=test_labels, y_pred=y_preds)
  inv sq acc.append(score)
  print('K:',k,'\n Accuracy:', score)
```

```
#linear weights
def f2(v):
  dk=v[-1] #dist to k+1 point
  d1=v[0]
  w=(dk-v)/(dk-d1)
  return w
lin acc=[]
for k in range(1,13):
  model = KNeighborsClassifier(n_neighbors=k)
  model.fit(X_train, train_labels)
  y preds = model.predict(X test)
  score=accuracy_score(y_true=test_labels, y_pred=y_preds)
  lin_acc.append(score)
  print('K:',k,'\n Accuracy:', score)
111111
inv_acc=[0.8497,0.8497,0.8561,0.8597,0.8577,0.8575,
    0.8541,0.8554,0.853,0.8538,0.8508,0.8494]
inv_sq_acc = [0.8497,0.8497,0.857,0.8603,0.8585,0.8588,
       0.8552,0.8565,0.8542,0.8548,0.8521,0.8508]
lin_acc=[0.8497,0.846,0.8541,0.8577,0.8554,0.8544,
     0.854,0.8534,0.8519,0.8515,0.8495,0.8471]
inv error = 1-np.array(inv acc)
inv_sq_error = 1-np.array(inv_sq_acc)
lin error = 1-np.array(lin acc)
plt.plot(range(1,13), inv_error, '--bs', label='Inverse Weights')
plt.plot(range(1,13), inv_sq_error,'--rs', label = 'Inverse Squared Weights')
plt.plot(range(1,13), lin_error,'--gs', label = 'Linear Weights')
plt.legend()
plt.xlabel('K')
plt.ylabel('Percent Error')
G0 = X train[train labels==0,:]
G1 = X_train[train_labels==1,:]
G2 = X_train[train_labels==2,:]
G3 = X train[train labels==3,:]
G4 = X_train[train_labels==4,:]
G5 = X_train[train_labels==5,:]
G6 = X_train[train_labels==6,:]
G7 = X train[train labels==7,:]
G8 = X train[train labels==8,:]
G9 = X_train[train_labels==9,:]
#let k=1
#find dist from k to all train points
#find clostsest 1 from each group
```

```
#find centroid
#find dist from k to each centroid
#find smallest distance
#arg min to find prediction
#k=2
#find next closest
#find centroid
#find dist from test point to each centroid
#find smallest distance
#arg min to find prediction
#... repeat for k 3:12
#
k1 preds=[]
k2 preds=[]
k3 preds=[]
k4 preds=[]
k5_preds=[]
k6 preds=[]
k7_preds=[]
k8_preds=[]
k9 preds=[]
k10_preds=[]
k11 preds=[]
k12_preds=[]
for r in range(X test.shape[0]): #later replace with X test.shape[0]
  G0 dists, G1 dists, G2_dists, G3_dists, G4_dists = [],[],[],[],[]
  G5_dists, G6_dists, G7_dists, G8_dists, G9_dists = [],[],[],[],[]
  closest_img_index_G0, closest_img_index_G1, closest_img_index_G2 = [],[],[]
  closest_img_index_G3, closest_img_index_G4, closest_img_index_G5= [],[],[]
  closest_img_index_G6, closest_img_index_G7, closest_img_index_G8 = [],[],[]
  closest_img_index_G9 = []
  test point = X test[r,:]
  #finds all distances, only run once
  for i in range(G0.shape[0]):
    GO dists.append( np.sqrt(np.sum(( test_point - GO[i,:] )**2)))
    G1_dists.append( np.sqrt(np.sum(( test_point - G1[i,:] )**2)))
    G2 dists.append( np.sqrt(np.sum(( test_point - G2[i,:] )**2)))
    G3_dists.append( np.sqrt(np.sum(( test_point - G3[i,:] )**2)))
    G4_dists.append( np.sqrt(np.sum(( test_point - G4[i,:] )**2)))
    G5_dists.append( np.sqrt(np.sum(( test_point - G5[i,:] )**2)))
    G6 dists.append( np.sqrt(np.sum(( test point - G6[i,:] )**2)))
    G7 dists.append( np.sqrt(np.sum(( test point - G7[i,:] )**2)) )
    G8_dists.append( np.sqrt(np.sum(( test_point - G8[i,:] )**2)))
    G9 dists.append( np.sqrt(np.sum(( test point - G9[i,:] )**2)))
```

```
i0=np.argmin(G0_dists)
i1=np.argmin(G1 dists)
i2=np.argmin(G2_dists)
i3=np.argmin(G3 dists)
i4=np.argmin(G4 dists)
i5=np.argmin(G5 dists)
i6=np.argmin(G6_dists)
i7=np.argmin(G7_dists)
i8=np.argmin(G8 dists)
i9=np.argmin(G9_dists)
closest img index G0.append(i0)
closest img index G1.append(i1)
closest_img_index_G2.append(i2)
closest img index G3.append(i3)
closest img index G4.append(i4)
closest img index G5.append(i5)
closest_img_index_G6.append(i6)
closest img index G7.append(i7)
closest_img_index_G8.append(i8)
closest_img_index_G9.append(i9)
#set that dist to inf so it's not found on next loop
G0_dists[i0] = np.Inf
G1 dists[i1] = np.Inf
G2_dists[i2] = np.Inf
G3 dists[i3] = np.Inf
G4 	ext{ dists[i4]} = np.Inf
G5 dists[i5] = np.Inf
G6_dists[i6] = np.Inf
G7_dists[i7] = np.Inf
G8 dists[i8] = np.Inf
G9_dists[i9] = np.Inf
#find centroid but doesn't matter for k=1
c0 = np.mean(G0[closest img index G0,:],axis=0)
c1 = np.mean(G1[closest img index G1,:],axis=0)
c2 = np.mean(G2[closest img index G2,:],axis=0)
c3 = np.mean(G3[closest img index G3,:],axis=0)
c4 = np.mean(G4[closest img index G4,:],axis=0)
c5 = np.mean(G5[closest_img_index_G5,:],axis=0)
c6 = np.mean(G6[closest_img_index_G6,:],axis=0)
c7 = np.mean(G7[closest img index G7,:],axis=0)
c8 = np.mean(G8[closest_img_index_G8,:],axis=0)
c9 = np.mean(G9[closest img index G9,:],axis=0)
#find dist from test point to each centroid
test to centroids dists = []
test to centroids dists.append(np.sqrt(np.sum((test point - c0)**2)))
test_to_centroids_dists.append(np.sqrt( np.sum( (test_point - c1)**2) ) )
test to centroids dists.append(np.sqrt(np.sum((test point - c2)**2)))
test_to_centroids_dists.append(np.sqrt( np.sum( (test_point - c3)**2) ) )
```

```
test_to_centroids_dists.append(np.sqrt( np.sum( (test_point - c4)**2) ) )
test to centroids dists.append(np.sqrt(np.sum((test point - c5)**2)))
test_to_centroids_dists.append(np.sqrt( np.sum( (test_point - c6)**2) ) )
test to centroids dists.append(np.sqrt(np.sum((test point - c7)**2)))
test to centroids dists.append(np.sqrt(np.sum((test point - c8)**2)))
test to centroids dists.append(np.sqrt(np.sum((test point - c9)**2)))
#actually finds the prediction
k1_preds.append(np.argmin(test_to_centroids_dists))
#find next index
i0=np.argmin(G0 dists)
i1=np.argmin(G1 dists)
i2=np.argmin(G2_dists)
i3=np.argmin(G3 dists)
i4=np.argmin(G4 dists)
i5=np.argmin(G5 dists)
i6=np.argmin(G6_dists)
i7=np.argmin(G7 dists)
i8=np.argmin(G8 dists)
i9=np.argmin(G9_dists)
closest_img_index_G0.append(i0)
closest img index G1.append(i1)
closest_img_index_G2.append(i2)
closest img index G3.append(i3)
closest img index G4.append(i4)
closest img index G5.append(i5)
closest_img_index_G6.append(i6)
closest_img_index_G7.append(i7)
closest img index G8.append(i8)
closest_img_index_G9.append(i9)
#set that dist to inf so it's not found on next loop
G0 dists[i0] = np.Inf
G1 dists[i1] = np.Inf
G2 dists[i2] = np.Inf
G3 dists[i3] = np.Inf
G4 	ext{ dists[i4]} = np.Inf
G5 dists[i5] = np.Inf
G6 \text{ dists[i6]} = np.Inf
G7_dists[i7] = np.Inf
G8 dists[i8] = np.Inf
G9_dists[i9] = np.Inf
#find centroids
c0 = np.mean(G0[closest img index G0,:],axis=0)
c1 = np.mean(G1[closest img index G1,:],axis=0)
c2 = np.mean(G2[closest_img_index_G2,:],axis=0)
c3 = np.mean(G3[closest img index G3,:],axis=0)
c4 = np.mean(G4[closest_img_index_G4,:],axis=0)
c5 = np.mean(G5[closest img index G5,:],axis=0)
c6 = np.mean(G6[closest_img_index_G6,:],axis=0)
```

```
c7 = np.mean(G7[closest_img_index_G7,:],axis=0)
c8 = np.mean(G8[closest img index G8,:],axis=0)
c9 = np.mean(G9[closest_img_index_G9,:],axis=0)
#find dist from test point to each centroid
test to centroids dists = []
test to centroids dists.append(np.sqrt(np.sum((test point - c0)**2)))
test_to_centroids_dists.append(np.sqrt( np.sum( (test_point - c1)**2) ) )
test_to_centroids_dists.append(np.sqrt( np.sum( (test_point - c2)**2) ) )
test_to_centroids_dists.append(np.sqrt( np.sum( (test_point - c3)**2) ) )
test to centroids dists.append(np.sqrt(np.sum((test point - c4)**2)))
test_to_centroids_dists.append(np.sqrt( np.sum( (test_point - c5)**2) ) )
test to centroids dists.append(np.sqrt(np.sum((test point - c6)**2)))
test to centroids dists.append(np.sqrt(np.sum((test point - c7)**2)))
test_to_centroids_dists.append(np.sqrt( np.sum( (test_point - c8)**2) ) )
test to centroids dists.append(np.sqrt(np.sum((test point - c9)**2)))
#actually finds the prediction
k2_preds.append(np.argmin(test_to_centroids_dists))
#find next index
i0=np.argmin(G0_dists)
i1=np.argmin(G1_dists)
i2=np.argmin(G2 dists)
i3=np.argmin(G3_dists)
i4=np.argmin(G4 dists)
i5=np.argmin(G5 dists)
i6=np.argmin(G6 dists)
i7=np.argmin(G7 dists)
i8=np.argmin(G8_dists)
i9=np.argmin(G9 dists)
closest img index G0.append(i0)
closest_img_index_G1.append(i1)
closest img index G2.append(i2)
closest img index G3.append(i3)
closest img index G4.append(i4)
closest img index G5.append(i5)
closest img index G6.append(i6)
closest_img_index_G7.append(i7)
closest_img_index_G8.append(i8)
closest img index G9.append(i9)
#set that dist to inf so it's not found on next loop
G0 dists[i0] = np.Inf
G1 dists[i1] = np.Inf
G2 dists[i2] = np.Inf
G3 dists[i3] = np.Inf
G4 dists[i4] = np.Inf
G5 dists[i5] = np.Inf
G6_dists[i6] = np.Inf
G7 \text{ dists}[i7] = np.Inf
G8_dists[i8] = np.Inf
```

```
G9_dists[i9] = np.Inf
#find centroids
c0 = np.mean(G0[closest img index G0,:],axis=0)
c1 = np.mean(G1[closest img index G1,:],axis=0)
c2 = np.mean(G2[closest img index G2,:],axis=0)
c3 = np.mean(G3[closest img index G3,:],axis=0)
c4 = np.mean(G4[closest_img_index_G4,:],axis=0)
c5 = np.mean(G5[closest img index G5,:],axis=0)
c6 = np.mean(G6[closest_img_index_G6,:],axis=0)
c7 = np.mean(G7[closest_img_index_G7,:],axis=0)
c8 = np.mean(G8[closest_img_index_G8,:],axis=0)
c9 = np.mean(G9[closest img index G9,:],axis=0)
#find dist from test point to each centroid
test to centroids dists = []
test to centroids dists.append(np.sqrt(np.sum((test point - c0)**2)))
test to centroids dists.append(np.sqrt(np.sum((test point - c1)**2)))
test_to_centroids_dists.append(np.sqrt( np.sum( (test_point - c2)**2) ) )
test to centroids dists.append(np.sqrt(np.sum((test point - c3)**2)))
test_to_centroids_dists.append(np.sqrt( np.sum( (test_point - c4)**2) ) )
test_to_centroids_dists.append(np.sqrt( np.sum( (test_point - c5)**2) ) )
test_to_centroids_dists.append(np.sqrt( np.sum( (test_point - c6)**2) ) )
test_to_centroids_dists.append(np.sqrt( np.sum( (test_point - c7)**2) ) )
test to centroids dists.append(np.sqrt(np.sum((test point - c8)**2)))
test_to_centroids_dists.append(np.sqrt( np.sum( (test_point - c9)**2) ) )
#actually finds the prediction
k3 preds.append(np.argmin(test to centroids dists))
#k=4
#find next index
i0=np.argmin(G0_dists)
i1=np.argmin(G1 dists)
i2=np.argmin(G2_dists)
i3=np.argmin(G3 dists)
i4=np.argmin(G4 dists)
i5=np.argmin(G5 dists)
i6=np.argmin(G6 dists)
i7=np.argmin(G7 dists)
i8=np.argmin(G8 dists)
i9=np.argmin(G9_dists)
closest_img_index_G0.append(i0)
closest img index G1.append(i1)
closest_img_index_G2.append(i2)
closest img index G3.append(i3)
closest img index G4.append(i4)
closest img index G5.append(i5)
closest img index G6.append(i6)
closest_img_index_G7.append(i7)
closest img index G8.append(i8)
```

closest_img_index_G9.append(i9)

```
#set that dist to inf so it's not found on next loop
GO dists[i0] = np.Inf
G1 dists[i1] = np.lnf
G2 dists[i2] = np.Inf
G3 dists[i3] = np.Inf
G4 dists[i4] = np.Inf
G5 \text{ dists[i5]} = np.Inf
G6_dists[i6] = np.Inf
G7 \text{ dists}[i7] = np.Inf
G8 dists[i8] = np.Inf
G9 dists[i9] = np.Inf
#find centroids
c0 = np.mean(G0[closest img index G0,:],axis=0)
c1 = np.mean(G1[closest_img_index_G1,:],axis=0)
c2 = np.mean(G2[closest img index G2,:],axis=0)
c3 = np.mean(G3[closest img index G3,:],axis=0)
c4 = np.mean(G4[closest img index G4,:],axis=0)
c5 = np.mean(G5[closest_img_index_G5,:],axis=0)
c6 = np.mean(G6[closest img index G6,:],axis=0)
c7 = np.mean(G7[closest_img_index_G7,:],axis=0)
c8 = np.mean(G8[closest_img_index_G8,:],axis=0)
c9 = np.mean(G9[closest_img_index_G9,:],axis=0)
#find dist from test point to each centroid
test to centroids dists = []
test to centroids dists.append(np.sqrt(np.sum((test point - c0)**2)))
test to centroids dists.append(np.sqrt(np.sum((test point - c1)**2)))
test to centroids dists.append(np.sqrt(np.sum((test point - c2)**2)))
test_to_centroids_dists.append(np.sqrt( np.sum( (test_point - c3)**2) ) )
test_to_centroids_dists.append(np.sqrt( np.sum( (test_point - c4)**2) ) )
test to centroids dists.append(np.sqrt(np.sum((test point - c5)**2)))
test_to_centroids_dists.append(np.sqrt( np.sum( (test_point - c6)**2) ) )
test to centroids dists.append(np.sqrt(np.sum((test point - c7)**2)))
test_to_centroids_dists.append(np.sqrt( np.sum( (test_point - c8)**2) ) )
test to centroids dists.append(np.sqrt(np.sum((test point - c9)**2)))
#actually finds the prediction
k4 preds.append(np.argmin(test to centroids dists))
#find next index
i0=np.argmin(G0 dists)
i1=np.argmin(G1_dists)
i2=np.argmin(G2 dists)
i3=np.argmin(G3_dists)
i4=np.argmin(G4 dists)
i5=np.argmin(G5 dists)
i6=np.argmin(G6 dists)
i7=np.argmin(G7 dists)
i8=np.argmin(G8_dists)
i9=np.argmin(G9 dists)
```

```
closest_img_index_G0.append(i0)
closest img index G1.append(i1)
closest_img_index_G2.append(i2)
closest img index G3.append(i3)
closest img index G4.append(i4)
closest img index G5.append(i5)
closest_img_index_G6.append(i6)
closest_img_index_G7.append(i7)
closest img index G8.append(i8)
closest_img_index_G9.append(i9)
#set that dist to inf so it's not found on next loop
GO dists[i0] = np.Inf
G1 dists[i1] = np.lnf
G2 dists[i2] = np.Inf
G3_dists[i3] = np.Inf
G4 	ext{ dists[i4]} = np.Inf
G5 dists[i5] = np.Inf
G6 \text{ dists[i6]} = np.Inf
G7_dists[i7] = np.Inf
G8 dists[i8] = np.Inf
G9_dists[i9] = np.Inf
#find centroids
c0 = np.mean(G0[closest_img_index_G0,:],axis=0)
c1 = np.mean(G1[closest img index G1,:],axis=0)
c2 = np.mean(G2[closest_img_index_G2,:],axis=0)
c3 = np.mean(G3[closest img index G3,:],axis=0)
c4 = np.mean(G4[closest img index G4,:],axis=0)
c5 = np.mean(G5[closest img index G5,:],axis=0)
c6 = np.mean(G6[closest_img_index_G6,:],axis=0)
c7 = np.mean(G7[closest_img_index_G7,:],axis=0)
c8 = np.mean(G8[closest img index G8,:],axis=0)
c9 = np.mean(G9[closest_img_index_G9,:],axis=0)
#find dist from test point to each centroid
test to centroids dists = []
test to centroids dists.append(np.sqrt(np.sum((test point - c0)**2)))
test to centroids dists.append(np.sqrt(np.sum((test point - c1)**2)))
test to centroids dists.append(np.sqrt(np.sum((test point - c2)**2)))
test to centroids dists.append(np.sqrt(np.sum((test point - c3)**2)))
test_to_centroids_dists.append(np.sqrt( np.sum( (test_point - c4)**2) ) )
test_to_centroids_dists.append(np.sqrt( np.sum( (test_point - c5)**2) ) )
test to centroids dists.append(np.sqrt(np.sum((test point - c6)**2)))
test_to_centroids_dists.append(np.sqrt( np.sum( (test_point - c7)**2) ) )
test to centroids dists.append(np.sqrt(np.sum((test point - c8)**2)))
test to centroids dists.append(np.sqrt(np.sum((test point - c9)**2)))
#actually finds the prediction
k5_preds.append(np.argmin(test_to_centroids_dists))
#find next index
i0=np.argmin(G0 dists)
```

```
i1=np.argmin(G1_dists)
i2=np.argmin(G2 dists)
i3=np.argmin(G3_dists)
i4=np.argmin(G4 dists)
i5=np.argmin(G5 dists)
i6=np.argmin(G6 dists)
i7=np.argmin(G7 dists)
i8=np.argmin(G8_dists)
i9=np.argmin(G9 dists)
closest_img_index_G0.append(i0)
closest_img_index_G1.append(i1)
closest img index G2.append(i2)
closest img index G3.append(i3)
closest_img_index_G4.append(i4)
closest img index G5.append(i5)
closest img index G6.append(i6)
closest img index G7.append(i7)
closest_img_index_G8.append(i8)
closest img index G9.append(i9)
#set that dist to inf so it's not found on next loop
G0_dists[i0] = np.Inf
G1 dists[i1] = np.Inf
G2_dists[i2] = np.Inf
G3 dists[i3] = np.Inf
G4 dists[i4] = np.Inf
G5 dists[i5] = np.Inf
G6 dists[i6] = np.Inf
G7 \text{ dists[i7]} = np.Inf
G8 dists[i8] = np.Inf
G9_dists[i9] = np.Inf
#find centroids
c0 = np.mean(G0[closest_img_index_G0,:],axis=0)
c1 = np.mean(G1[closest_img_index_G1,:],axis=0)
c2 = np.mean(G2[closest img index G2,:],axis=0)
c3 = np.mean(G3[closest_img_index_G3,:],axis=0)
c4 = np.mean(G4[closest img index G4,:],axis=0)
c5 = np.mean(G5[closest img index G5,:],axis=0)
c6 = np.mean(G6[closest img index G6,:],axis=0)
c7 = np.mean(G7[closest_img_index_G7,:],axis=0)
c8 = np.mean(G8[closest_img_index_G8,:],axis=0)
c9 = np.mean(G9[closest img index G9,:],axis=0)
#find dist from test point to each centroid
test to centroids dists = []
test to centroids dists.append(np.sqrt(np.sum((test point - c0)**2)))
test to centroids dists.append(np.sqrt(np.sum((test point - c1)**2)))
test_to_centroids_dists.append(np.sqrt( np.sum( (test_point - c2)**2) ) )
test to centroids dists.append(np.sqrt(np.sum((test point - c3)**2)))
test_to_centroids_dists.append(np.sqrt( np.sum( (test_point - c4)**2) ) )
test to centroids dists.append(np.sqrt(np.sum((test point - c5)**2)))
test_to_centroids_dists.append(np.sqrt( np.sum( (test_point - c6)**2) ) )
```

```
test_to_centroids_dists.append(np.sqrt( np.sum( (test_point - c7)**2) ) )
test to centroids dists.append(np.sqrt(np.sum((test point - c8)**2)))
test_to_centroids_dists.append(np.sqrt( np.sum( (test_point - c9)**2) ) )
#actually finds the prediction
k6_preds.append(np.argmin(test_to_centroids_dists))
#find next index
i0=np.argmin(G0_dists)
i1=np.argmin(G1_dists)
i2=np.argmin(G2_dists)
i3=np.argmin(G3 dists)
i4=np.argmin(G4 dists)
i5=np.argmin(G5_dists)
i6=np.argmin(G6 dists)
i7=np.argmin(G7 dists)
i8=np.argmin(G8 dists)
i9=np.argmin(G9_dists)
closest_img_index_G0.append(i0)
closest_img_index_G1.append(i1)
closest_img_index_G2.append(i2)
closest_img_index_G3.append(i3)
closest img index G4.append(i4)
closest_img_index_G5.append(i5)
closest img index G6.append(i6)
closest img index G7.append(i7)
closest img index G8.append(i8)
closest_img_index_G9.append(i9)
#set that dist to inf so it's not found on next loop
GO dists[i0] = np.Inf
G1_dists[i1] = np.Inf
G2 	ext{ dists[i2]} = np.Inf
G3_dists[i3] = np.Inf
G4 dists[i4] = np.Inf
G5 dists[i5] = np.Inf
G6 dists[i6] = np.Inf
G7 	ext{ dists[i7]} = np.Inf
G8 dists[i8] = np.Inf
G9_dists[i9] = np.Inf
#find centroids
c0 = np.mean(G0[closest_img_index_G0,:],axis=0)
c1 = np.mean(G1[closest img index G1,:],axis=0)
c2 = np.mean(G2[closest_img_index_G2,:],axis=0)
c3 = np.mean(G3[closest img index G3,:],axis=0)
c4 = np.mean(G4[closest_img_index_G4,:],axis=0)
c5 = np.mean(G5[closest_img_index_G5,:],axis=0)
c6 = np.mean(G6[closest img index G6,:],axis=0)
c7 = np.mean(G7[closest_img_index_G7,:],axis=0)
c8 = np.mean(G8[closest img index G8,:],axis=0)
c9 = np.mean(G9[closest_img_index_G9,:],axis=0)
```

```
#find dist from test point to each centroid
test_to_centroids_dists = []
test to centroids dists.append(np.sqrt(np.sum((test point - c0)**2)))
test to centroids dists.append(np.sqrt(np.sum((test point - c1)**2)))
test to centroids dists.append(np.sqrt(np.sum((test point - c2)**2)))
test to centroids dists.append(np.sqrt(np.sum((test point - c3)**2)))
test_to_centroids_dists.append(np.sqrt( np.sum( (test_point - c4)**2) ) )
test to centroids dists.append(np.sqrt(np.sum((test point - c5)**2)))
test_to_centroids_dists.append(np.sqrt( np.sum( (test_point - c6)**2) ) )
test to centroids dists.append(np.sqrt(np.sum((test point - c7)**2)))
test_to_centroids_dists.append(np.sqrt( np.sum( (test_point - c8)**2) ) )
test to centroids dists.append(np.sqrt(np.sum((test point - c9)**2)))
#actually finds the prediction
k7 preds.append(np.argmin(test to centroids dists))
#find next index
i0=np.argmin(G0 dists)
i1=np.argmin(G1 dists)
i2=np.argmin(G2_dists)
i3=np.argmin(G3_dists)
i4=np.argmin(G4_dists)
i5=np.argmin(G5 dists)
i6=np.argmin(G6 dists)
i7=np.argmin(G7 dists)
i8=np.argmin(G8 dists)
i9=np.argmin(G9 dists)
closest_img_index_G0.append(i0)
closest img index G1.append(i1)
closest_img_index_G2.append(i2)
closest img index G3.append(i3)
closest_img_index_G4.append(i4)
closest img index G5.append(i5)
closest img index G6.append(i6)
closest img index G7.append(i7)
closest img index G8.append(i8)
closest img index G9.append(i9)
#set that dist to inf so it's not found on next loop
GO dists[i0] = np.Inf
G1 dists[i1] = np.lnf
G2 dists[i2] = np.Inf
G3 dists[i3] = np.Inf
G4 	ext{ dists[i4]} = np.Inf
G5 dists[i5] = np.Inf
G6 dists[i6] = np.Inf
G7 dists[i7] = np.Inf
G8 dists[i8] = np.Inf
G9_dists[i9] = np.Inf
```

#find centroids

```
c0 = np.mean(G0[closest_img_index_G0,:],axis=0)
c1 = np.mean(G1[closest img index G1,:],axis=0)
c2 = np.mean(G2[closest img index G2,:],axis=0)
c3 = np.mean(G3[closest img index G3,:],axis=0)
c4 = np.mean(G4[closest img index G4,:],axis=0)
c5 = np.mean(G5[closest img index G5,:],axis=0)
c6 = np.mean(G6[closest img index G6,:],axis=0)
c7 = np.mean(G7[closest_img_index_G7,:],axis=0)
c8 = np.mean(G8[closest img index G8,:],axis=0)
c9 = np.mean(G9[closest_img_index_G9,:],axis=0)
#find dist from test point to each centroid
test to centroids dists = []
test to centroids dists.append(np.sqrt(np.sum((test point - c0)**2)))
test_to_centroids_dists.append(np.sqrt( np.sum( (test_point - c1)**2) ) )
test to centroids dists.append(np.sqrt(np.sum((test point - c2)**2)))
test to centroids dists.append(np.sqrt(np.sum((test point - c3)**2)))
test to centroids dists.append(np.sqrt(np.sum((test point - c4)**2)))
test_to_centroids_dists.append(np.sqrt( np.sum( (test_point - c5)**2) ) )
test to centroids dists.append(np.sqrt(np.sum((test point - c6)**2)))
test to centroids dists.append(np.sqrt(np.sum((test point - c7)**2)))
test_to_centroids_dists.append(np.sqrt( np.sum( (test_point - c8)**2) ) )
test_to_centroids_dists.append(np.sqrt( np.sum( (test_point - c9)**2) ) )
#actually finds the prediction
k8_preds.append(np.argmin(test_to_centroids_dists))
#find next index
i0=np.argmin(G0 dists)
i1=np.argmin(G1_dists)
i2=np.argmin(G2 dists)
i3=np.argmin(G3_dists)
i4=np.argmin(G4 dists)
i5=np.argmin(G5_dists)
i6=np.argmin(G6 dists)
i7=np.argmin(G7 dists)
i8=np.argmin(G8 dists)
i9=np.argmin(G9 dists)
closest img index G0.append(i0)
closest_img_index_G1.append(i1)
closest img index G2.append(i2)
closest_img_index_G3.append(i3)
closest img index G4.append(i4)
closest_img_index_G5.append(i5)
closest img index G6.append(i6)
closest img index G7.append(i7)
closest_img_index_G8.append(i8)
closest img index G9.append(i9)
#set that dist to inf so it's not found on next loop
G0 dists[i0] = np.Inf
G1_dists[i1] = np.Inf
```

```
G2_dists[i2] = np.Inf
G3 dists[i3] = np.Inf
G4 \text{ dists}[i4] = np.Inf
G5 dists[i5] = np.Inf
G6 dists[i6] = np.Inf
G7 dists[i7] = np.Inf
G8 dists[i8] = np.Inf
G9_dists[i9] = np.Inf
#find centroids
c0 = np.mean(G0[closest_img_index_G0,:],axis=0)
c1 = np.mean(G1[closest_img_index_G1,:],axis=0)
c2 = np.mean(G2[closest img index G2,:],axis=0)
c3 = np.mean(G3[closest img index G3,:],axis=0)
c4 = np.mean(G4[closest_img_index_G4,:],axis=0)
c5 = np.mean(G5[closest img index G5,:],axis=0)
c6 = np.mean(G6[closest img index G6,:],axis=0)
c7 = np.mean(G7[closest img index G7,:],axis=0)
c8 = np.mean(G8[closest_img_index_G8,:],axis=0)
c9 = np.mean(G9[closest img index G9,:],axis=0)
#find dist from test point to each centroid
test to centroids dists = []
test_to_centroids_dists.append(np.sqrt( np.sum( (test_point - c0)**2) ) )
test to centroids dists.append(np.sqrt(np.sum((test point - c1)**2)))
test_to_centroids_dists.append(np.sqrt( np.sum( (test_point - c2)**2) ) )
test to centroids dists.append(np.sqrt(np.sum((test point - c3)**2)))
test to centroids dists.append(np.sqrt(np.sum((test point - c4)**2)))
test to centroids dists.append(np.sqrt(np.sum((test point - c5)**2)))
test_to_centroids_dists.append(np.sqrt( np.sum( (test_point - c6)**2) ) )
test_to_centroids_dists.append(np.sqrt( np.sum( (test_point - c7)**2) ) )
test to centroids dists.append(np.sqrt(np.sum((test point - c8)**2)))
test_to_centroids_dists.append(np.sqrt( np.sum( (test_point - c9)**2) ) )
#actually finds the prediction
k9 preds.append(np.argmin(test to centroids dists))
#find next index
i0=np.argmin(G0 dists)
i1=np.argmin(G1 dists)
i2=np.argmin(G2_dists)
i3=np.argmin(G3 dists)
i4=np.argmin(G4_dists)
i5=np.argmin(G5 dists)
i6=np.argmin(G6_dists)
i7=np.argmin(G7 dists)
i8=np.argmin(G8 dists)
i9=np.argmin(G9_dists)
closest_img_index_G0.append(i0)
closest img index G1.append(i1)
closest_img_index_G2.append(i2)
```

```
closest_img_index_G3.append(i3)
closest img index G4.append(i4)
closest_img_index_G5.append(i5)
closest img index G6.append(i6)
closest img index G7.append(i7)
closest img index G8.append(i8)
closest img index G9.append(i9)
#set that dist to inf so it's not found on next loop
GO dists[i0] = np.Inf
G1 dists[i1] = np.lnf
G2 dists[i2] = np.Inf
G3 dists[i3] = np.Inf
G4 \text{ dists[i4]} = np.Inf
G5 dists[i5] = np.Inf
G6_dists[i6] = np.Inf
G7 	ext{ dists[i7]} = np.Inf
G8 dists[i8] = np.Inf
G9 dists[i9] = np.Inf
#find centroids
c0 = np.mean(G0[closest_img_index_G0,:],axis=0)
c1 = np.mean(G1[closest_img_index_G1,:],axis=0)
c2 = np.mean(G2[closest_img_index_G2,:],axis=0)
c3 = np.mean(G3[closest_img_index_G3,:],axis=0)
c4 = np.mean(G4[closest img index G4,:],axis=0)
c5 = np.mean(G5[closest_img_index_G5,:],axis=0)
c6 = np.mean(G6[closest img index G6,:],axis=0)
c7 = np.mean(G7[closest img index G7,:],axis=0)
c8 = np.mean(G8[closest img index G8,:],axis=0)
c9 = np.mean(G9[closest_img_index_G9,:],axis=0)
#find dist from test point to each centroid
test_to_centroids_dists = []
test_to_centroids_dists.append(np.sqrt( np.sum( (test_point - c0)**2) ) )
test_to_centroids_dists.append(np.sqrt( np.sum( (test_point - c1)**2) ) )
test to centroids dists.append(np.sqrt(np.sum((test point - c2)**2)))
test to centroids dists.append(np.sqrt(np.sum((test point - c3)**2)))
test to centroids dists.append(np.sqrt(np.sum((test point - c4)**2)))
test to centroids dists.append(np.sqrt(np.sum((test point - c5)**2)))
test to centroids dists.append(np.sqrt(np.sum((test point - c6)**2)))
test to centroids dists.append(np.sqrt(np.sum((test point - c7)**2)))
test_to_centroids_dists.append(np.sqrt( np.sum( (test_point - c8)**2) ) )
test to centroids dists.append(np.sqrt(np.sum((test point - c9)**2)))
#actually finds the prediction
k10 preds.append(np.argmin(test to centroids dists))
#find next index
i0=np.argmin(G0 dists)
i1=np.argmin(G1_dists)
i2=np.argmin(G2 dists)
i3=np.argmin(G3_dists)
```

```
i4=np.argmin(G4_dists)
i5=np.argmin(G5 dists)
i6=np.argmin(G6 dists)
i7=np.argmin(G7 dists)
i8=np.argmin(G8 dists)
i9=np.argmin(G9 dists)
closest_img_index_G0.append(i0)
closest img index G1.append(i1)
closest_img_index_G2.append(i2)
closest_img_index_G3.append(i3)
closest_img_index_G4.append(i4)
closest img index G5.append(i5)
closest img index G6.append(i6)
closest_img_index_G7.append(i7)
closest img index G8.append(i8)
closest img index G9.append(i9)
#set that dist to inf so it's not found on next loop
GO dists[i0] = np.Inf
G1 dists[i1] = np.Inf
G2 	ext{ dists[i2]} = np.Inf
G3_dists[i3] = np.Inf
G4 \text{ dists[i4]} = np.Inf
G5_dists[i5] = np.Inf
G6 dists[i6] = np.Inf
G7_dists[i7] = np.Inf
G8 dists[i8] = np.Inf
G9 dists[i9] = np.Inf
#find centroids
c0 = np.mean(G0[closest_img_index_G0,:],axis=0)
c1 = np.mean(G1[closest img index G1,:],axis=0)
c2 = np.mean(G2[closest_img_index_G2,:],axis=0)
c3 = np.mean(G3[closest img index G3,:],axis=0)
c4 = np.mean(G4[closest_img_index_G4,:],axis=0)
c5 = np.mean(G5[closest img index G5,:],axis=0)
c6 = np.mean(G6[closest img index G6,:],axis=0)
c7 = np.mean(G7[closest img index G7,:],axis=0)
c8 = np.mean(G8[closest img index G8,:],axis=0)
c9 = np.mean(G9[closest img index G9,:],axis=0)
#find dist from test point to each centroid
test to centroids dists = []
test_to_centroids_dists.append(np.sqrt( np.sum( (test_point - c0)**2) ) )
test to centroids dists.append(np.sqrt(np.sum((test point - c1)**2)))
test_to_centroids_dists.append(np.sqrt( np.sum( (test_point - c2)**2) ) )
test to centroids dists.append(np.sqrt(np.sum((test point - c3)**2)))
test to centroids dists.append(np.sqrt(np.sum((test point - c4)**2)))
test to centroids dists.append(np.sqrt(np.sum((test point - c5)**2)))
test to centroids dists.append(np.sqrt(np.sum((test point - c6)**2)))
test_to_centroids_dists.append(np.sqrt( np.sum( (test_point - c7)**2) ) )
test to centroids dists.append(np.sqrt(np.sum((test point - c8)**2)))
test_to_centroids_dists.append(np.sqrt( np.sum( (test_point - c9)**2) ) )
```

```
#actually finds the prediction
k11_preds.append(np.argmin(test_to_centroids_dists))
```

#find next index i0=np.argmin(G0 dists) i1=np.argmin(G1_dists) i2=np.argmin(G2 dists) i3=np.argmin(G3_dists) i4=np.argmin(G4_dists) i5=np.argmin(G5_dists) i6=np.argmin(G6 dists) i7=np.argmin(G7 dists) i8=np.argmin(G8_dists) i9=np.argmin(G9 dists) closest img index G0.append(i0) closest_img_index_G1.append(i1) closest img index G2.append(i2) closest_img_index_G3.append(i3) closest_img_index_G4.append(i4) closest_img_index_G5.append(i5) closest_img_index_G6.append(i6) closest img index G7.append(i7) closest_img_index_G8.append(i8) closest img index G9.append(i9) #set that dist to inf so it's not found on next loop GO dists[i0] = np.Inf $G1_dists[i1] = np.Inf$ G2_dists[i2] = np.Inf G3 dists[i3] = np.Inf $G4_dists[i4] = np.Inf$ G5 dists[i5] = np.InfG6_dists[i6] = np.Inf $G7 ext{ dists[i7]} = np.Inf$ G8 dists[i8] = np.Inf G9 dists[i9] = np.Inf #find centroids c0 = np.mean(G0[closest img index G0,:],axis=0) c1 = np.mean(G1[closest_img_index_G1,:],axis=0) c2 = np.mean(G2[closest img index G2,:],axis=0) c3 = np.mean(G3[closest_img_index_G3,:],axis=0) c4 = np.mean(G4[closest img index G4,:],axis=0) c5 = np.mean(G5[closest_img_index_G5,:],axis=0) c6 = np.mean(G6[closest img index G6,:],axis=0) c7 = np.mean(G7[closest_img_index_G7,:],axis=0) c8 = np.mean(G8[closest_img_index_G8,:],axis=0) c9 = np.mean(G9[closest img index G9,:],axis=0) #find dist from test point to each centroid

test_to_centroids_dists = []

```
test_to_centroids_dists.append(np.sqrt( np.sum( (test_point - c0)**2) ) )
  test to centroids dists.append(np.sqrt(np.sum((test point - c1)**2)))
  test_to_centroids_dists.append(np.sqrt( np.sum( (test_point - c2)**2) ) )
  test to centroids dists.append(np.sqrt(np.sum((test point - c3)**2)))
  test to centroids dists.append(np.sqrt(np.sum((test point - c4)**2)))
  test to centroids dists.append(np.sqrt(np.sum((test point - c5)**2)))
  test to centroids dists.append(np.sqrt(np.sum((test point - c6)**2)))
  test_to_centroids_dists.append(np.sqrt( np.sum( (test_point - c7)**2) ) )
  test_to_centroids_dists.append(np.sqrt( np.sum( (test_point - c8)**2) ) )
  test_to_centroids_dists.append(np.sqrt( np.sum( (test_point - c9)**2) ) )
  #actually finds the prediction
  k12 preds.append(np.argmin(test to centroids dists))
scorek1=accuracy_score(y_true=test_labels, y_pred=k1_preds)
scorek2=accuracy score(y true=test labels, y pred=k2 preds)
scorek3=accuracy_score(y_true=test_labels, y_pred=k3_preds)
scorek4=accuracy score(y true=test labels, y pred=k4 preds)
scorek5=accuracy_score(y_true=test_labels, y_pred=k5_preds)
scorek6=accuracy_score(y_true=test_labels, y_pred=k6_preds)
scorek7=accuracy_score(y_true=test_labels, y_pred=k7_preds)
scorek8=accuracy_score(y_true=test_labels, y_pred=k8_preds)
scorek9=accuracy_score(y_true=test_labels, y_pred=k9_preds)
scorek10=accuracy_score(y_true=test_labels, y_pred=k10_preds)
scorek11=accuracy_score(y_true=test_labels, y_pred=k11_preds)
scorek12=accuracy_score(y_true=test_labels, y_pred=k12_preds)
111111
NLC acc=[0.8497,0.8668,0.8715,0.8784,0.8791,0.8816,
    0.8834,0.8836,0.8855,0.8867,0.885,0.8851]
np.argmax(NLC_acc) #9 is the best
NLC_error = 1-np.array(NLC_acc)
plt.plot(range(1,13), NLC error,'-rs', label = 'NLC error')
plt.legend()
plt.xlabel('K')
plt.ylabel('Percent Error')
#let k=1
#find dist from k to all train points
#find clostsest 1 from each group
```

#find dist from k to all train points
#find clostsest 1 from each group
#find centroid
#find dist from k to each centroid
#find smallest distance
#arg min to find prediction

```
#k=2
#find next closest
#find centroid
#find dist from test point to each centroid
k1_preds=[]
k2 preds=[]
k3_preds=[]
k4 preds=[]
k5_preds=[]
k6 preds=[]
k7 preds=[]
k8_preds=[]
k9 preds=[]
k10 preds=[]
k11 preds=[]
k12_preds=[]
for r in range(X_test.shape[0]):
  G0_dists, G1_dists, G2_dists, G3_dists, G4_dists = [],[],[],[],[]
  G5_dists, G6_dists, G7_dists, G8_dists, G9_dists = [],[],[],[],[]
  closest img index G0, closest img index G1, closest img index G2 = [],[],[]
  closest_img_index_G3, closest_img_index_G4, closest_img_index_G5= [],[],[]
  closest img index G6, closest img index G7, closest img index G8 = [],[],[]
  closest img index G9 = []
  test_point = X_test[r,:]
  #finds all distances, only run once
  for i in range(G0.shape[0]):
    GO_dists.append( np.sqrt(np.sum(( test_point - G0[i,:] )**2)))
    G1_dists.append( np.sqrt(np.sum(( test_point - G1[i,:] )**2)))
    G2 dists.append( np.sqrt(np.sum(( test point - G2[i,:] )**2)))
    G3_dists.append( np.sqrt(np.sum(( test_point - G3[i,:] )**2)))
    G4 dists.append( np.sqrt(np.sum(( test point - G4[i,:] )**2)))
    G5 dists.append( np.sqrt(np.sum(( test point - G5[i,:] )**2)))
    G6 dists.append( np.sqrt(np.sum(( test point - G6[i,:] )**2)))
    G7_dists.append( np.sqrt(np.sum(( test_point - G7[i,:] )**2)))
    G8_dists.append( np.sqrt(np.sum(( test_point - G8[i,:] )**2)))
    G9_dists.append( np.sqrt(np.sum(( test_point - G9[i,:] )**2)))
  #find index of shortest dist
  i0=np.argmin(G0 dists)
  i1=np.argmin(G1 dists)
  i2=np.argmin(G2_dists)
  i3=np.argmin(G3 dists)
  i4=np.argmin(G4_dists)
  i5=np.argmin(G5 dists)
  i6=np.argmin(G6_dists)
```

```
i7=np.argmin(G7_dists)
i8=np.argmin(G8 dists)
i9=np.argmin(G9_dists)
closest img index G0.append(i0)
closest_img_index_G1.append(i1)
closest_img_index_G2.append(i2)
closest img index G3.append(i3)
closest_img_index_G4.append(i4)
closest_img_index_G5.append(i5)
closest_img_index_G6.append(i6)
closest img index G7.append(i7)
closest img index G8.append(i8)
closest_img_index_G9.append(i9)
#set that dist to inf so it's not found on next loop
G0 dists[i0] = np.Inf
G1 dists[i1] = np.Inf
G2_dists[i2] = np.Inf
G3 dists[i3] = np.Inf
G4 \text{ dists}[i4] = np.Inf
G5_dists[i5] = np.Inf
G6 dists[i6] = np.Inf
G7_dists[i7] = np.Inf
G8 dists[i8] = np.Inf
G9_dists[i9] = np.Inf
#find next index
i0=np.argmin(G0 dists)
i1=np.argmin(G1_dists)
i2=np.argmin(G2 dists)
i3=np.argmin(G3_dists)
i4=np.argmin(G4_dists)
i5=np.argmin(G5_dists)
i6=np.argmin(G6 dists)
i7=np.argmin(G7_dists)
i8=np.argmin(G8 dists)
i9=np.argmin(G9 dists)
closest img index G0.append(i0)
closest_img_index_G1.append(i1)
closest img index G2.append(i2)
closest_img_index_G3.append(i3)
closest img index G4.append(i4)
closest_img_index_G5.append(i5)
closest img index G6.append(i6)
closest img index G7.append(i7)
closest_img_index_G8.append(i8)
closest img index G9.append(i9)
#set that dist to inf so it's not found on next loop
G0 dists[i0] = np.Inf
G1_dists[i1] = np.Inf
```

```
G2_dists[i2] = np.Inf
G3 dists[i3] = np.Inf
G4_dists[i4] = np.Inf
G5 dists[i5] = np.Inf
G6 dists[i6] = np.Inf
G7 dists[i7] = np.lnf
G8 dists[i8] = np.Inf
G9 dists[i9] = np.Inf
#find next index
i0=np.argmin(G0_dists)
i1=np.argmin(G1_dists)
i2=np.argmin(G2 dists)
i3=np.argmin(G3 dists)
i4=np.argmin(G4_dists)
i5=np.argmin(G5 dists)
i6=np.argmin(G6 dists)
i7=np.argmin(G7_dists)
i8=np.argmin(G8_dists)
i9=np.argmin(G9 dists)
closest_img_index_G0.append(i0)
closest_img_index_G1.append(i1)
closest_img_index_G2.append(i2)
closest img index G3.append(i3)
closest_img_index_G4.append(i4)
closest img index G5.append(i5)
closest img index G6.append(i6)
closest img index G7.append(i7)
closest_img_index_G8.append(i8)
closest_img_index_G9.append(i9)
#set that dist to inf so it's not found on next loop
G0_dists[i0] = np.Inf
G1 dists[i1] = np.Inf
G2_dists[i2] = np.Inf
G3 dists[i3] = np.Inf
G4_dists[i4] = np.Inf
G5 dists[i5] = np.Inf
G6 dists[i6] = np.Inf
G7 dists[i7] = np.Inf
G8 dists[i8] = np.Inf
G9_dists[i9] = np.Inf
#k=4
#find next index
i0=np.argmin(G0 dists)
i1=np.argmin(G1 dists)
i2=np.argmin(G2 dists)
i3=np.argmin(G3 dists)
i4=np.argmin(G4_dists)
i5=np.argmin(G5 dists)
```

i6=np.argmin(G6_dists)

```
i7=np.argmin(G7_dists)
i8=np.argmin(G8 dists)
i9=np.argmin(G9_dists)
closest img index G0.append(i0)
closest img index G1.append(i1)
closest img index G2.append(i2)
closest_img_index_G3.append(i3)
closest_img_index_G4.append(i4)
closest_img_index_G5.append(i5)
closest_img_index_G6.append(i6)
closest_img_index_G7.append(i7)
closest img index G8.append(i8)
closest img index G9.append(i9)
#set that dist to inf so it's not found on next loop
G0 dists[i0] = np.Inf
G1 dists[i1] = np.Inf
G2 \text{ dists[i2]} = np.Inf
G3_dists[i3] = np.Inf
G4 \text{ dists[i4]} = np.Inf
G5 dists[i5] = np.Inf
G6_dists[i6] = np.Inf
G7 \text{ dists}[i7] = np.Inf
G8_dists[i8] = np.Inf
G9 dists[i9] = np.Inf
#find next index
i0=np.argmin(G0 dists)
i1=np.argmin(G1_dists)
i2=np.argmin(G2_dists)
i3=np.argmin(G3 dists)
i4=np.argmin(G4_dists)
i5=np.argmin(G5_dists)
i6=np.argmin(G6_dists)
i7=np.argmin(G7 dists)
i8=np.argmin(G8 dists)
i9=np.argmin(G9 dists)
closest img index G0.append(i0)
closest_img_index_G1.append(i1)
closest_img_index_G2.append(i2)
closest img index G3.append(i3)
closest_img_index_G4.append(i4)
closest img index G5.append(i5)
closest_img_index_G6.append(i6)
closest img index G7.append(i7)
closest img index G8.append(i8)
closest_img_index_G9.append(i9)
#set that dist to inf so it's not found on next loop
G0_dists[i0] = np.Inf
G1 dists[i1] = np.lnf
G2_dists[i2] = np.Inf
```

```
G3_dists[i3] = np.lnf
G4 	ext{ dists[i4]} = np.Inf
G5_dists[i5] = np.Inf
G6 dists[i6] = np.Inf
G7 dists[i7] = np.Inf
G8 dists[i8] = np.Inf
G9 dists[i9] = np.Inf
#find next index
i0=np.argmin(G0_dists)
i1=np.argmin(G1_dists)
i2=np.argmin(G2 dists)
i3=np.argmin(G3 dists)
i4=np.argmin(G4_dists)
i5=np.argmin(G5 dists)
i6=np.argmin(G6 dists)
i7=np.argmin(G7 dists)
i8=np.argmin(G8_dists)
i9=np.argmin(G9 dists)
closest_img_index_G0.append(i0)
closest_img_index_G1.append(i1)
closest_img_index_G2.append(i2)
closest img index G3.append(i3)
closest_img_index_G4.append(i4)
closest img index G5.append(i5)
closest img index G6.append(i6)
closest img index G7.append(i7)
closest_img_index_G8.append(i8)
closest_img_index_G9.append(i9)
#set that dist to inf so it's not found on next loop
G0_dists[i0] = np.Inf
G1 dists[i1] = np.Inf
G2_dists[i2] = np.Inf
G3 dists[i3] = np.Inf
G4 	ext{ dists[i4]} = np.Inf
G5 dists[i5] = np.Inf
G6 dists[i6] = np.Inf
G7 dists[i7] = np.Inf
G8_dists[i8] = np.Inf
G9_dists[i9] = np.Inf
#find next index
i0=np.argmin(G0 dists)
i1=np.argmin(G1 dists)
i2=np.argmin(G2 dists)
i3=np.argmin(G3_dists)
i4=np.argmin(G4 dists)
i5=np.argmin(G5_dists)
i6=np.argmin(G6 dists)
i7=np.argmin(G7_dists)
```

```
i8=np.argmin(G8_dists)
i9=np.argmin(G9_dists)
closest img index G0.append(i0)
closest img index G1.append(i1)
closest img index G2.append(i2)
closest img index G3.append(i3)
closest_img_index_G4.append(i4)
closest_img_index_G5.append(i5)
closest_img_index_G6.append(i6)
closest_img_index_G7.append(i7)
closest_img_index_G8.append(i8)
closest img index G9.append(i9)
#set that dist to inf so it's not found on next loop
G0_dists[i0] = np.Inf
G1 dists[i1] = np.Inf
G2 dists[i2] = np.Inf
G3 dists[i3] = np.Inf
G4_dists[i4] = np.Inf
G5 dists[i5] = np.Inf
G6_dists[i6] = np.Inf
G7_dists[i7] = np.Inf
G8_dists[i8] = np.Inf
G9_dists[i9] = np.Inf
#find next index
i0=np.argmin(G0 dists)
i1=np.argmin(G1 dists)
i2=np.argmin(G2_dists)
i3=np.argmin(G3_dists)
i4=np.argmin(G4 dists)
i5=np.argmin(G5_dists)
i6=np.argmin(G6_dists)
i7=np.argmin(G7_dists)
i8=np.argmin(G8 dists)
i9=np.argmin(G9 dists)
closest img index G0.append(i0)
closest img index G1.append(i1)
closest_img_index_G2.append(i2)
closest_img_index_G3.append(i3)
closest_img_index_G4.append(i4)
closest_img_index_G5.append(i5)
closest img index G6.append(i6)
closest_img_index_G7.append(i7)
closest img index G8.append(i8)
closest img index G9.append(i9)
#set that dist to inf so it's not found on next loop
G0 dists[i0] = np.Inf
G1_dists[i1] = np.Inf
G2 dists[i2] = np.Inf
G3_dists[i3] = np.Inf
```

```
G4_dists[i4] = np.Inf
G5 dists[i5] = np.Inf
G6_dists[i6] = np.Inf
G7 dists[i7] = np.lnf
G8 dists[i8] = np.Inf
G9 dists[i9] = np.Inf
#find next index
i0=np.argmin(G0_dists)
i1=np.argmin(G1_dists)
i2=np.argmin(G2_dists)
i3=np.argmin(G3 dists)
i4=np.argmin(G4 dists)
i5=np.argmin(G5_dists)
i6=np.argmin(G6 dists)
i7=np.argmin(G7 dists)
i8=np.argmin(G8 dists)
i9=np.argmin(G9_dists)
closest_img_index_G0.append(i0)
closest_img_index_G1.append(i1)
closest_img_index_G2.append(i2)
closest_img_index_G3.append(i3)
closest img index G4.append(i4)
closest_img_index_G5.append(i5)
closest img index G6.append(i6)
closest img index G7.append(i7)
closest img index G8.append(i8)
closest_img_index_G9.append(i9)
#set that dist to inf so it's not found on next loop
GO dists[i0] = np.Inf
G1_dists[i1] = np.Inf
G2 	ext{ dists[i2]} = np.Inf
G3_dists[i3] = np.Inf
G4 dists[i4] = np.Inf
G5 dists[i5] = np.Inf
G6 dists[i6] = np.Inf
G7 dists[i7] = np.Inf
G8 dists[i8] = np.Inf
G9_dists[i9] = np.Inf
#find next index
i0=np.argmin(G0 dists)
i1=np.argmin(G1 dists)
i2=np.argmin(G2 dists)
i3=np.argmin(G3_dists)
i4=np.argmin(G4 dists)
i5=np.argmin(G5_dists)
i6=np.argmin(G6 dists)
i7=np.argmin(G7_dists)
```

```
i8=np.argmin(G8_dists)
i9=np.argmin(G9 dists)
closest img index G0.append(i0)
closest img index G1.append(i1)
closest img index G2.append(i2)
closest img index G3.append(i3)
closest_img_index_G4.append(i4)
closest_img_index_G5.append(i5)
closest_img_index_G6.append(i6)
closest_img_index_G7.append(i7)
closest_img_index_G8.append(i8)
closest img index G9.append(i9)
#set that dist to inf so it's not found on next loop
G0 dists[i0] = np.Inf
G1 dists[i1] = np.Inf
G2 dists[i2] = np.Inf
G3 dists[i3] = np.Inf
G4_dists[i4] = np.Inf
G5 dists[i5] = np.Inf
G6 \text{ dists[i6]} = np.Inf
G7_dists[i7] = np.Inf
G8 dists[i8] = np.Inf
G9_dists[i9] = np.Inf
#find centroids
c0 = np.mean(G0[closest img index G0,:],axis=0)
c1 = np.mean(G1[closest img index G1,:],axis=0)
c2 = np.mean(G2[closest img index G2,:],axis=0)
c3 = np.mean(G3[closest_img_index_G3,:],axis=0)
c4 = np.mean(G4[closest_img_index_G4,:],axis=0)
c5 = np.mean(G5[closest img index G5,:],axis=0)
c6 = np.mean(G6[closest_img_index_G6,:],axis=0)
c7 = np.mean(G7[closest img index G7,:],axis=0)
c8 = np.mean(G8[closest_img_index_G8,:],axis=0)
c9 = np.mean(G9[closest img index G9,:],axis=0)
#find dist from test point to each centroid
test to centroids dists = []
test to centroids dists.append(np.sqrt(np.sum((test point - c0)**2)))
test to centroids dists.append(np.sqrt(np.sum((test point - c1)**2)))
test_to_centroids_dists.append(np.sqrt( np.sum( (test_point - c2)**2) ) )
test to centroids dists.append(np.sqrt(np.sum((test point - c3)**2)))
test_to_centroids_dists.append(np.sqrt( np.sum( (test_point - c4)**2) ) )
test to centroids dists.append(np.sqrt(np.sum((test point - c5)**2)))
test_to_centroids_dists.append(np.sqrt( np.sum( (test_point - c6)**2) ) )
test to centroids dists.append(np.sqrt(np.sum((test point - c7)**2)))
test to centroids dists.append(np.sqrt(np.sum((test point - c8)**2)))
test_to_centroids_dists.append(np.sqrt( np.sum( (test_point - c9)**2) ) )
#actually finds the prediction
k10 preds.append(np.argmin(test to centroids dists))
```

from sklearn.metrics import confusion_matrix

print(confusion_matrix(test_labels, k10_preds, labels=[0,1,2,3,4,5,6,7,8,9]))

```
""" for k=10
[[830 1 12 23 4 2 120 0 8 0]
[ 1977 0 16 2 0 3 0 1 0]
[ 21 1821 9 85 0 63 0 0 0]
[ 25 4 15 896 35 0 23 0 2 0]
[ 1 0 108 30 811 0 47 0 3 0]
[ 0 0 0 0 0 959 0 20 0 21]
[124 1 97 24 84 0 663 0 7 0]
[ 0 0 0 0 0 16 0 960 0 24]
[ 2 0 1 4 3 2 6 5 977 0]
[ 0 0 0 0 0 6 0 20 1 973]]
"""
```