

fsum0np6d

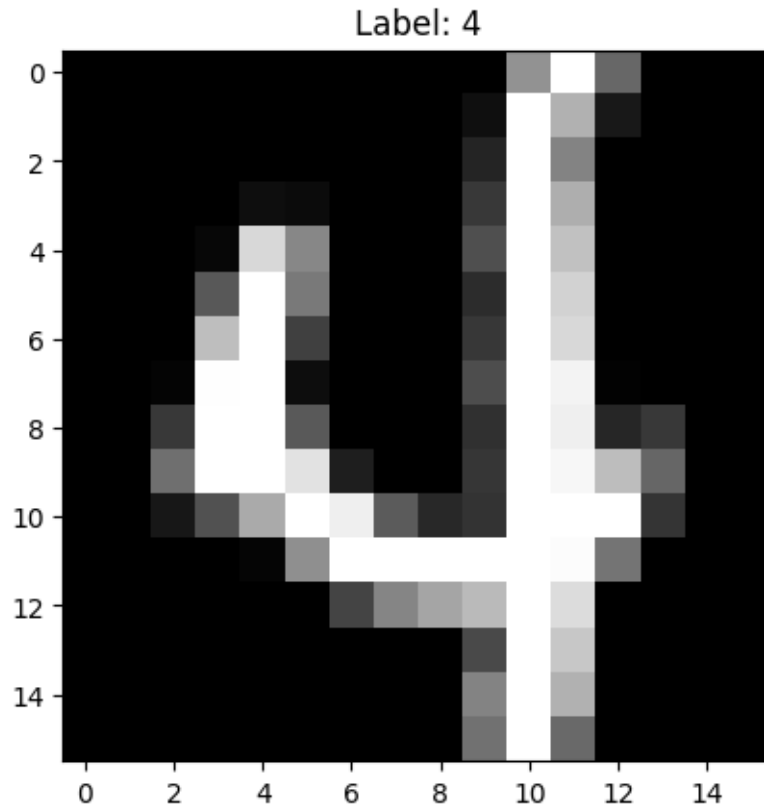
February 14, 2025

```
[487]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
import os
```

```
[488]: path = "/content/usps.h5"
```

```
[489]: import h5py
with h5py.File(path, 'r') as hf:
    train = hf.get('train')
    X_train = train.get('data')[:]
    y_train = train.get('target')[:]
    test = hf.get('test')
    X_test = test.get('data')[:]
    y_test = test.get('target')[:]
```

```
[490]: plt.imshow(X_train[2].reshape(16,16), cmap='gray')
plt.title(f'Label: {y_train[2]}')
plt.show()
```



```
[491]: print("Shape of the Train data features")
print(X_train.shape)
print("Shape of the Train data target")
print(y_train.shape)
print("Shape of the Test data features")
print(X_test.shape)
print("Shape of the Test data target")

print(y_test.shape)
```

```
Shape of the Train data features
(7291, 256)
Shape of the Train data target
(7291,)
Shape of the Test data features
(2007, 256)
Shape of the Test data target
(2007,)
```

```
[492]: print(np.unique(y_train))
```

```
[0 1 2 3 4 5 6 7 8 9]
```

Using PCA and Standard

```
[493]: from sklearn.decomposition import PCA
pca = PCA(n_components=50)
X_train = pca.fit_transform(X_train)
X_test = pca.transform(X_test)
```

```
[494]: from sklearn.preprocessing import StandardScaler
scaler = StandardScaler()
X_train_scaled = scaler.fit_transform(X_train)
X_test_scaled = scaler.transform(X_test)
```

```
[494]:
```

KNN Models

```
[495]: from sklearn.model_selection import KFold
from sklearn.neighbors import KNeighborsClassifier
```

```
[496]: from sklearn.model_selection import cross_val_score
clf = KNeighborsClassifier(n_neighbors=3)
scores = cross_val_score(clf, X_train, y_train, cv=10)
```

```
[497]: print("Train accuracy: ", scores.mean())
```

Train accuracy: 0.9729802506717778

```
[498]: clf.fit(X_train, y_train)
```

```
[498]: KNeighborsClassifier(n_neighbors=3)
```

```
[499]: # Evaluate on test set
from sklearn.metrics import accuracy_score
y_pred = clf.predict(X_test)
test_accuracy = accuracy_score(y_test, y_pred)
```

```
[500]: print("Test accuracy :",test_accuracy)
```

Test accuracy : 0.9486796213253612

```
[501]: from sklearn.metrics import confusion_matrix, classification_report
```

```
[502]: print(confusion_matrix(y_true=y_test, y_pred= y_pred))
```

```
[[355  0  3  0  0  0  0  0  0  1]
 [  0 257  0  0  4  0  2  1  0  0]
 [  7  0 185  1  1  0  0  1  3  0]]
```

```
[ 3  0  1 153  0  6  0  0  1  2]
[ 0  1  2  0 183  1  2  2  1  8]
[ 3  1  2  2  0 148  0  0  1  3]
[ 3  1  1  0  2  0 163  0  0  0]
[ 0  1  1  1  3  0  0 140  0  1]
[ 4  0  2  1  0  2  1  1 153  2]
[ 1  1  1  0  3  0  0  3  1 167]]
```

```
[503]: from sklearn.metrics import precision_score, recall_score

print("Precison score")
print(precision_score(y_true=y_test, y_pred=y_pred, average="macro"))
print("Recall score")
print(recall_score(y_true=y_test, y_pred=y_pred, average="macro"))
```

```
Precison score
0.9484154196887384
Recall score
0.9434767022174325
```

```
[504]: train_scores = []
test_scores = []

for i in range(1,11):
    print("For k value ", i, " :")
    clf = KNeighborsClassifier(n_neighbors=i)
    scores = cross_val_score(clf, X_train, y_train, cv=10)
    score = scores.mean()
    test_score = clf.fit(X_train, y_train).score(X_test, y_test)
    train_scores.append(score)
    test_scores.append(test_score)
    print(f"Neighbors: {i}, Train accuracy: {score:.4f}, Test accuracy: {test_score:.4f}")
    print("Confusion matrix :")
    print(confusion_matrix(y_true=y_test, y_pred= clf.predict(X_test)))
    print("Precision score : ")
    print(precision_score(y_true=y_test, y_pred=clf.predict(X_test),
    ↪average="macro"))
    print("Recall score : ")
    print(recall_score(y_true=y_test, y_pred=clf.predict(X_test),
    ↪average="macro"))
    print("-----")
```

```
For k value 1 :
Neighbors: 1, Train accuracy: 0.9733, Test accuracy: 0.9482
Confusion matrix :
[[355  0  2  0  0  0  0  1  0  1]
 [ 0 255  0  0  6  0  2  1  0  0]
```

```

[ 4  0 186  2  1  0  0  1  4  0]
[ 2  0  1 153  0  9  0  0  0  1]
[ 0  2  0  0 184  2  2  2  1  7]
[ 2  1  2  2  0 149  0  0  3  1]
[ 0  0  1  0  2  4 163  0  0  0]
[ 0  1  1  1  4  0  0 136  2  2]
[ 2  0  2  5  0  2  0  0 151  4]
[ 0  0  1  0  1  0  0  3  1 171]]

```

Precision score :
0.9438701530347882
Recall score :
0.9426831562580743

For k value 2 :
Neighbors: 2, Train accuracy: 0.9667, Test accuracy: 0.9482

Confusion matrix :
[[355 0 3 0 0 0 0 0 0 1]
[0 260 0 0 3 0 1 0 0 0]
[9 0 185 1 1 0 0 1 1 0]
[3 0 1 155 0 6 0 0 0 1]
[1 2 2 0 187 1 0 2 0 5]
[3 1 2 8 0 146 0 0 0 0]
[3 0 1 0 2 3 161 0 0 0]
[0 2 1 1 4 0 0 139 0 0]
[4 0 4 3 0 2 0 1 150 2]
[1 1 1 0 4 1 0 4 0 165]]

Precision score :
0.9491384108099208
Recall score :
0.9417739700153381

For k value 3 :
Neighbors: 3, Train accuracy: 0.9730, Test accuracy: 0.9487

Confusion matrix :
[[355 0 3 0 0 0 0 0 0 1]
[0 257 0 0 4 0 2 1 0 0]
[7 0 185 1 1 0 0 1 3 0]
[3 0 1 153 0 6 0 0 1 2]
[0 1 2 0 183 1 2 2 1 8]
[3 1 2 2 0 148 0 0 1 3]
[3 1 1 0 2 0 163 0 0 0]
[0 1 1 1 3 0 0 140 0 1]
[4 0 2 1 0 2 1 1 153 2]
[1 1 1 0 3 0 0 3 1 167]]

Precision score :
0.9484154196887384
Recall score :
0.9434767022174325

For k value 4 :
Neighbors: 4, Train accuracy: 0.9700, Test accuracy: 0.9447

Confusion matrix :

```
[[355  0  3  0  0  0  0  0  0  1]
 [  0 258  0  0  4  0  2  0  0  0]
 [  6  0 183  1  1  0  2  2  3  0]
 [  2  0  2 155  0  4  0  1  0  2]
 [  0  3  3  0 185  1  2  2  0  4]
 [  5  0  2  4  0 144  0  0  1  4]
 [  3  1  2  0  2  0 162  0  0  0]
 [  0  1  1  1  4  1  0 138  0  1]
 [  5  0  1  3  0  3  2  1 149  2]
 [  1  1  1  0  2  1  0  4  0 167]]
```

Precision score :

0.9439571786232257

Recall score :

0.9381917902972061

For k value 5 :

Neighbors: 5, Train accuracy: 0.9685, Test accuracy: 0.9477

Confusion matrix :

```
[[355  0  2  0  1  0  0  0  0  1]
 [  0 258  0  0  4  0  2  0  0  0]
 [  6  0 183  1  1  0  1  2  4  0]
 [  2  0  2 154  0  5  0  1  0  2]
 [  0  3  2  0 184  1  2  2  0  6]
 [  4  0  2  2  0 147  0  0  1  4]
 [  3  0  2  0  2  0 163  0  0  0]
 [  0  1  1  1  4  1  0 137  0  2]
 [  2  0  1  4  0  3  1  1 152  2]
 [  1  0  1  0  1  1  0  4  0 169]]
```

Precision score :

0.9458380951546841

Recall score :

0.9418095162624136

For k value 6 :

Neighbors: 6, Train accuracy: 0.9669, Test accuracy: 0.9432

Confusion matrix :

```
[[355  0  2  0  1  0  0  0  0  1]
 [  0 258  0  0  4  0  2  0  0  0]
 [  6  0 183  1  1  0  1  2  4  0]
 [  2  0  2 154  0  5  0  1  0  2]
 [  0  3  4  0 184  0  2  2  0  5]
 [  4  0  2  4  0 144  0  0  1  5]
 [  3  0  3  0  2  2 160  0  0  0]
 [  0  1  1  1  4  1  0 139  0  0]]
```

```
[ 4  2  1  2  0  4  2  2 148  1]
[ 1  0  1  0  2  1  0  4  0 168]]
```

Precision score :
0.9415792732878817
Recall score :
0.9365557442921183

For k value 7 :
Neighbors: 7, Train accuracy: 0.9665, Test accuracy: 0.9412

Confusion matrix :
[[355 0 2 0 1 0 0 0 0 1]
[0 258 0 0 4 0 2 0 0 0]
[7 0 182 1 1 0 1 2 4 0]
[3 0 1 154 0 5 0 1 0 2]
[0 3 4 0 182 0 2 2 0 7]
[6 0 2 2 0 144 0 0 1 5]
[3 1 2 0 3 2 159 0 0 0]
[0 1 1 1 4 1 0 138 0 1]
[4 1 1 3 0 5 0 0 149 3]
[1 0 0 0 2 1 0 4 1 168]]

Precision score :
0.9407027284393328
Recall score :
0.9343845960226608

For k value 8 :
Neighbors: 8, Train accuracy: 0.9660, Test accuracy: 0.9397

Confusion matrix :
[[355 0 2 0 1 0 0 0 0 1]
[0 258 0 0 4 0 2 0 0 0]
[7 0 182 1 1 0 2 2 3 0]
[2 0 2 154 0 5 0 1 0 2]
[0 3 4 0 184 0 2 2 0 5]
[6 0 3 4 0 141 0 0 1 5]
[3 0 3 0 2 2 159 0 1 0]
[0 1 1 1 4 1 0 137 1 1]
[4 2 0 3 0 5 1 0 148 3]
[1 0 0 0 2 1 0 4 1 168]]

Precision score :
0.9383173966638152
Recall score :
0.9322269142752629

For k value 9 :
Neighbors: 9, Train accuracy: 0.9649, Test accuracy: 0.9367

Confusion matrix :
[[354 0 2 0 1 0 1 0 0 1]
[0 257 0 0 4 0 3 0 0 0]

```

[ 7  0 181  1  1  0  2  2  4  0]
[ 3  0  1 154  0  6  0  1  0  1]
[ 0  3  4  0 181  0  2  2  0  8]
[ 5  0  3  4  0 143  0  0  1  4]
[ 3  0  3  0  2  2 159  0  1  0]
[ 0  1  1  1  4  1  0 136  1  2]
[ 5  2  0  5  0  3  1  0 147  3]
[ 1  0  0  0  2  1  0  4  1 168]]

```

Precision score :

0.9351246929315924

Recall score :

0.9295318426119934

For k value 10 :

Neighbors: 10, Train accuracy: 0.9645, Test accuracy: 0.9382

Confusion matrix :

```

[[354  0  2  0  1  0  1  0  0  1]
 [ 0 257  0  0  4  0  3  0  0  0]
 [ 7  0 182  1  1  0  2  2  3  0]
 [ 3  0  2 154  0  5  0  1  0  1]
 [ 0  3  5  0 181  0  2  2  0  7]
 [ 4  0  2  5  0 144  0  0  1  4]
 [ 3  0  3  0  2  2 159  0  1  0]
 [ 0  2  1  1  3  1  0 137  1  1]
 [ 4  2  0  4  0  5  1  2 147  1]
 [ 1  0  0  0  1  1  0  5  1 168]]

```

Precision score :

0.935985155090839

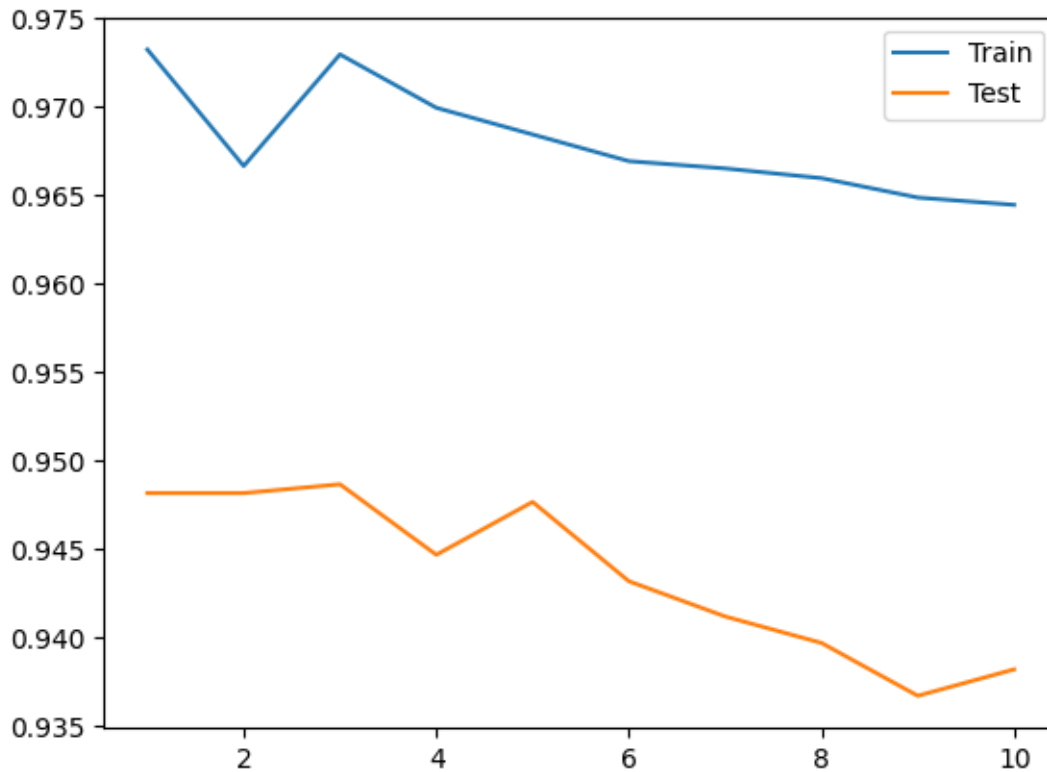
Recall score :

0.9313421652258874

```

[505]: plt.plot(range(1,11), train_scores, label="Train")
plt.plot(range(1,11), test_scores, label="Test")
plt.legend()
plt.show()

```

[505]:

Naive Bayes Model

[506]: `from scipy.ndimage.interpolation import shift`
`from sklearn.metrics import accuracy_score`

<ipython-input-506-7b5c807e8aaf>:1: DeprecationWarning: Please import `shift` from the `scipy.ndimage` namespace; the `scipy.ndimage.interpolation` namespace is deprecated and will be removed in SciPy 2.0.0.

`from scipy.ndimage.interpolation import shift`

[506]:

[507]: `from sklearn.naive_bayes import GaussianNB`

`nb = GaussianNB()`

`scores = cross_val_score(nb, X_train, y_train, cv=10)`

`print("Train accuracy: ", scores.mean())`

`nb.fit(X_train, y_train)`

```

# Predict on test set
y_pred_nb = nb.predict(X_test)

# Compute metrics
accuracy_nb = accuracy_score(y_test, y_pred_nb)
precision_nb = precision_score(y_test, y_pred_nb, average='macro')
recall_nb = recall_score(y_test, y_pred_nb, average='macro')

# Print results
print(f"Naïve Bayes Test Accuracy: {accuracy_nb:.4f}")
print(f"Naïve Bayes Test Precision: {precision_nb:.4f}")
print(f"Naïve Bayes Test Recall: {recall_nb:.4f}")

print("Confusion matrix naive bayes")
print(confusion_matrix(y_true=y_test, y_pred= y_pred_nb))

```

Train accuracy: 0.9009720578010786

Naïve Bayes Test Accuracy: 0.8670

Naïve Bayes Test Precision: 0.8603

Naïve Bayes Test Recall: 0.8568

Confusion matrix naive bayes

```

[[338  0  6  1  2  6  6  0  0  0]
 [ 1 233  0  2  5  3  9  1  3  7]
 [ 1  0 171  5  8  1  2  2  8  0]
 [ 1  0  5 138  1 16  1  1  2  1]
 [ 1  1  9  0 177  1  1  1  0  9]
 [ 3  0  2 10  4 136  0  0  2  3]
 [ 4  0  8  0  5  13 139  0  1  0]
 [ 0  0  2  0  8  3  0 124  2  8]
 [ 2  0  6 10  5 12  1  0 128  2]
 [ 2  0  2  1  9  2  0  2  3 156]]

```

[507]:

[507]:

[507]: