

Παραδοτέο ενδιάμεσης εργασίας Νευρωνικά

Χρυσικός Χρήστος AEM:9432 THMMY

Παρακάτω είναι τα απαραίτητα modules που χρειάζονται για να τρέξουν οι αλγόριθμοι. Χρησιμοποιούμε την βιβλιοθήκη sklearn για τον KNeighborsClassifier και NearestCentroid

```
In [ ]: from sklearn import metrics
import matplotlib.pyplot as plt
from keras.datasets import mnist
from sklearn.neighbors import KNeighborsClassifier, NearestCentroid
from time import time
```

Τα δεδομένα τα φορτώνουμε απο την Database MNIST χρησιμοποιώντας το module keras. Όμως επειδή η fit δέχεται είσοδο 2D Array και τα μοντέλα είναι σε μορφή (SAMPLES_NUM, 28, 28) τα μετασχηματίζουμε σε (SAMPLES_NUM, 784)

Data Transform

```
In [ ]: (train_X, train_y), (test_X, test_y) = mnist.load_data()

train_X = train_X.reshape(train_X.shape[0], train_X.shape[1] ** 2)
test_X = test_X.reshape(test_X.shape[0], test_X.shape[1] ** 2)
```

KNN Classifier

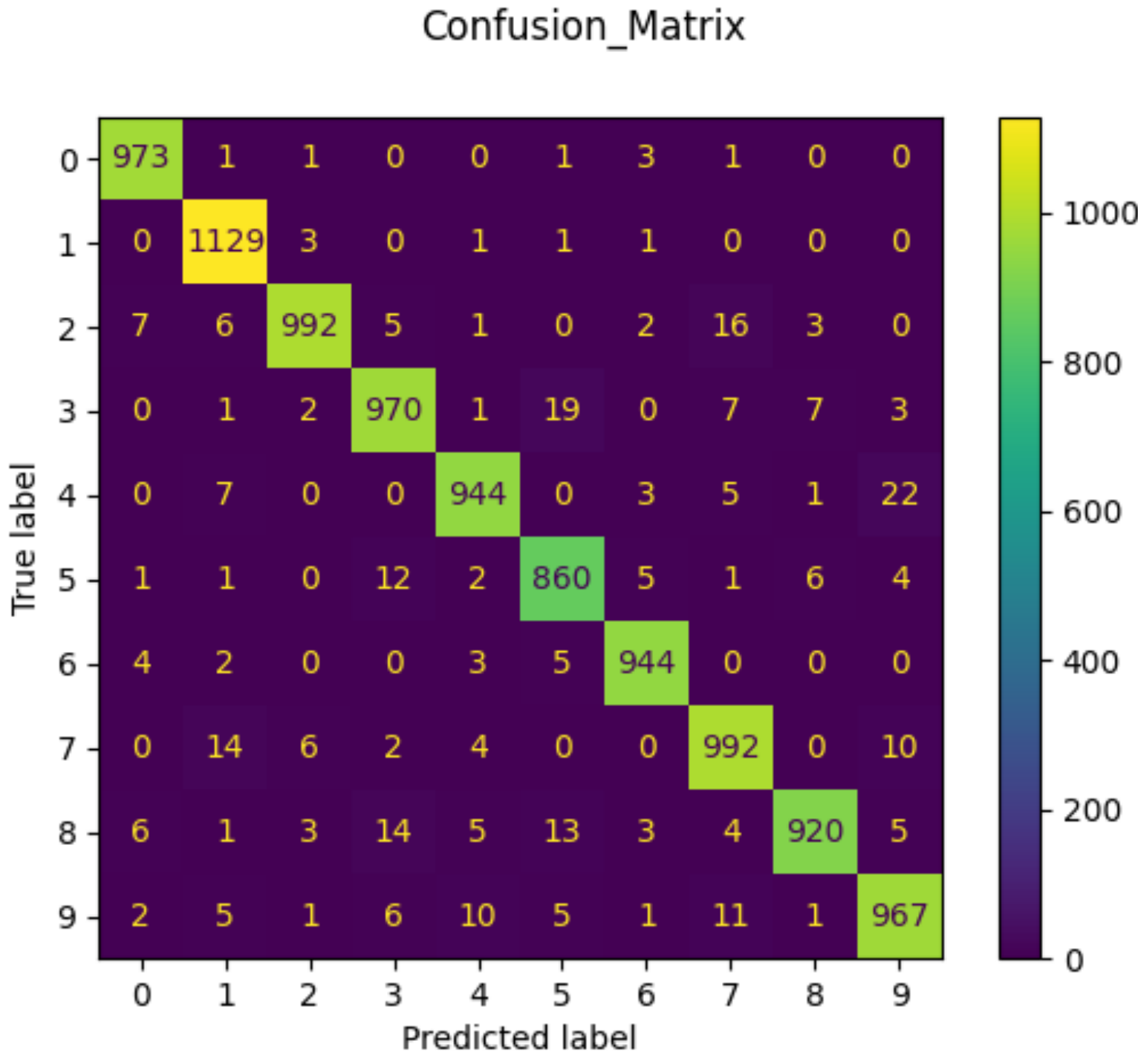
Με μια συνάρτηση επανάληψης τρέχουμε 2 φορές τον classifier για K = 1 και K = 3

```
In [ ]: neighbors = [1, 3]
for k in neighbors:
    #Fitting the KNearest Neighbor Classifier
    knn = KNeighborsClassifier(k)
    # Starting timer
    t = time()
    # Fitting the training data to the model
    knn.fit(train_X, train_y)
    # Creating first time stamp for time elapsed to fit the data
    time_fit = time() - t
    t = time()
    # Using testing data to evaluate the model
    predicted = knn.predict(test_X)
    # Creating second time stamp for time elapsed to predict
    time_predict = time() - t
    print(f"Analysis for K Nearest Neighbor = {k}")
    # Displaying of the time it took to fit, predict the data and the total runtime of the process.
    print(f"Time to fit: {time_fit:.2f}s - Time to predict: {time_predict:.2f}s\nTotal time: "
          f"{(time_fit + time_predict):.2f}s")
    # Displaying the scores for each classification of the model.
    print(
        f"Classification report for classifier {knn}:\n"
        f"{metrics.classification_report(test_y, predicted)}\n"
    )
    # Using Confusion Matrix to display the Model's performance
    display = metrics.ConfusionMatrixDisplay.from_predictions(test_y, predicted)
    display.figure_.suptitle("Confusion_Matrix")
    plt.show()
```

KNN (K = 1) Classifier Scores

```
In [ ]: Analysis for K Nearest Neighbor = 1
Time to fit: 0.01s - Time to predict: 17.34s
Total time: 17.35s
Classification report for classifier KNeighborsClassifier(n_neighbors=1):
```

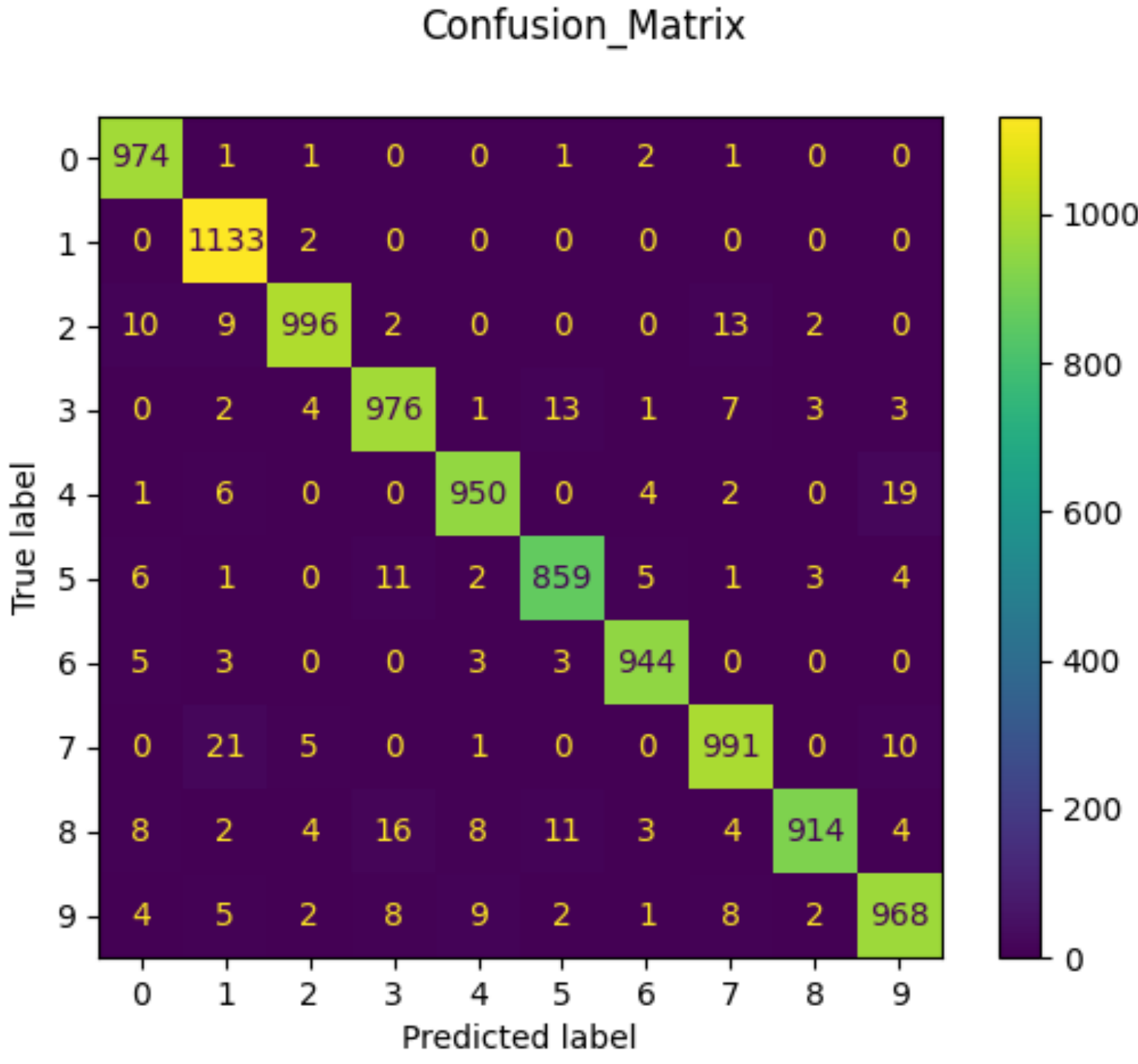
| | precision | recall | f1-score | support |
|--------------|-----------|--------|----------|---------|
| 0 | 0.98 | 0.99 | 0.99 | 980 |
| 1 | 0.97 | 0.99 | 0.98 | 1135 |
| 2 | 0.98 | 0.96 | 0.97 | 1032 |
| 3 | 0.96 | 0.96 | 0.96 | 1010 |
| 4 | 0.97 | 0.96 | 0.97 | 982 |
| 5 | 0.95 | 0.96 | 0.96 | 892 |
| 6 | 0.98 | 0.99 | 0.98 | 958 |
| 7 | 0.96 | 0.96 | 0.96 | 1028 |
| 8 | 0.98 | 0.94 | 0.96 | 974 |
| 9 | 0.96 | 0.96 | 0.96 | 1009 |
| accuracy | | | 0.97 | 10000 |
| macro avg | 0.97 | 0.97 | 0.97 | 10000 |
| weighted avg | 0.97 | 0.97 | 0.97 | 10000 |



KNN (K = 3) Classifier Scores

```
In [ ]: Analysis for K Nearest Neighbor = 3
Time to fit: 0.00s - Time to predict: 18.88s
Total time: 18.89s
Classification report for classifier KNeighborsClassifier(n_neighbors=3):
```

| | precision | recall | f1-score | support |
|--------------|-----------|--------|----------|---------|
| 0 | 0.97 | 0.99 | 0.98 | 980 |
| 1 | 0.96 | 1.00 | 0.98 | 1135 |
| 2 | 0.98 | 0.97 | 0.97 | 1032 |
| 3 | 0.96 | 0.97 | 0.96 | 1010 |
| 4 | 0.98 | 0.97 | 0.97 | 982 |
| 5 | 0.97 | 0.96 | 0.96 | 892 |
| 6 | 0.98 | 0.99 | 0.98 | 958 |
| 7 | 0.96 | 0.96 | 0.96 | 1028 |
| 8 | 0.99 | 0.94 | 0.96 | 974 |
| 9 | 0.96 | 0.96 | 0.96 | 1009 |
| accuracy | | | 0.97 | 10000 |
| macro avg | 0.97 | 0.97 | 0.97 | 10000 |
| weighted avg | 0.97 | 0.97 | 0.97 | 10000 |



Nearest Centroid

Αλλάζουμε τον Classifier και ακολουθούμε ακριβώς την ίδια διαδικασία

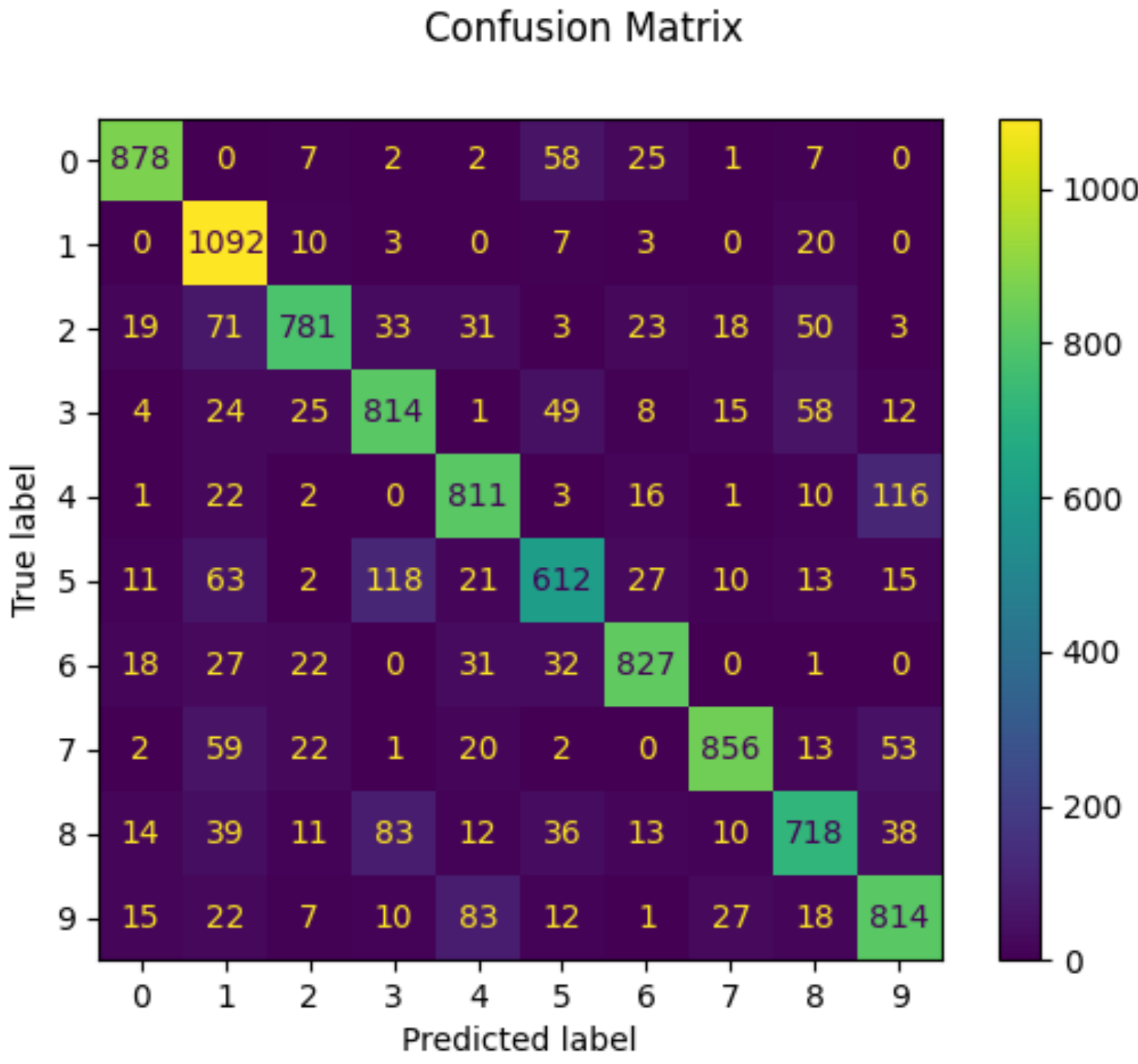
```
In [ ]: #Fitting the Nearest Centroid Classifier
NCentroid = NearestCentroid()
# Starting timer
t = time()
NCentroid.fit(train_X, train_y)
# Creating first time stamp for time elapsed to fit the data
time_fit = time() - t
t = time()
# Using testing data to evaluate the model
predicted = NCentroid.predict(test_X)
# Creating second time stamp for time elapsed to predict
predict_time = time() - t

print(f"Analysis for Nearest Centroid")
# Displaying of the time it took to fit, predict the data and the total runtime of the process.
print(f"Time to fit: {time_fit:.2f}s - Time to predict: {time_predict:.2f}s\nTotal time: "
      f"{(time_fit + time_predict):.2f}s")

# Displaying the scores for each classification of the model.
print(
    f"Classification report for classifier {NCentroid}:\n"
    f"{metrics.classification_report(test_y, predicted)}\n"
)
# Using Confusion Matrix to display the Model's performance
display = metrics.ConfusionMatrixDisplay.from_predictions(test_y, predicted)
display.figure_.suptitle("Confusion_Matrix")
plt.show()
```

```
In [ ]: Analysis for Nearest Centroid
Time to fit: 0.05s - Time to predict: 18.88s
Total time: 18.93s
Classification report for classifier NearestCentroid():
```

| | precision | recall | f1-score | support |
|--------------|-----------|--------|----------|---------|
| 0 | 0.91 | 0.90 | 0.90 | 980 |
| 1 | 0.77 | 0.96 | 0.86 | 1135 |
| 2 | 0.88 | 0.76 | 0.81 | 1032 |
| 3 | 0.77 | 0.81 | 0.78 | 1010 |
| 4 | 0.80 | 0.83 | 0.81 | 982 |
| 5 | 0.75 | 0.69 | 0.72 | 892 |
| 6 | 0.88 | 0.86 | 0.87 | 958 |
| 7 | 0.91 | 0.83 | 0.87 | 1028 |
| 8 | 0.79 | 0.74 | 0.76 | 974 |
| 9 | 0.77 | 0.81 | 0.79 | 1009 |
| accuracy | | | 0.82 | 10000 |
| macro avg | 0.82 | 0.82 | 0.82 | 10000 |
| weighted avg | 0.82 | 0.82 | 0.82 | 10000 |



Conclusion

Είναι πλέον προφανές οτι για το MNIST DataSet ο KNN classifier είναι ιδανικότερος αφού πετυχαίνουμε μεγαλύτερη ακρίβεια και K = 1 και K = 3. Πιο συγκεκριμένα για K = 1 πετυχαίνουμε την ίδια ακρίβεια και σε χαμηλότερο χρόνο.

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