

## Imports

```
import numpy as np
from sklearn.cluster import KMeans
from sklearn.metrics import accuracy_score, confusion_matrix
from sklearn.neighbors import NearestNeighbors
from keras.datasets import mnist
```

## Definitons

```
def label_clusters(cluster_counts):
    num_clusters, num_labels = cluster_counts.shape
    cluster_labels = np.zeros(num_clusters, dtype=int)
    label_counts = np.sum(cluster_counts, axis=0) # How many data points belong to each label

    # Iterate through elements in descending order based on the number of data points per label
    for ni_j in np.argsort(label_counts)[::-1]:
        # Find clusters with the maximum number of data points for this label
        candidate_clusters = np.where(cluster_counts[:, ni_j] == cluster_counts[:, ni_j].max())[0]
        # Assign the label to the first cluster that doesn't have a label or the label isn't assigned yet
        for cluster in candidate_clusters:
            if cluster_labels[cluster] == 0 or ni_j not in cluster_labels:
                cluster_labels[cluster] = ni_j
                break

    return cluster_labels
```

Labels clusters based on the maximum number of training data points with a specific label belonging to each cluster.

Args: cluster\_counts: A numpy array where each row represents a cluster and each column represents a label, with the value at each position indicating how many training data points with that label belong to that cluster.

Returns: A numpy array where each element represents the assigned label for the corresponding cluster.

```
def calculate_accuracy(true_labels, predicted_labels):
    return accuracy_score(true_labels, predicted_labels)
```

## Load Data

```
(X_train, y_train), (X_test, y_test) = mnist.load_data()
```

```
X_train = X_train.reshape(X_train.shape[0], X_train.shape[1] * X_train.shape[2])
X_test = X_test.reshape(X_test.shape[0], X_test.shape[1] * X_test.shape[2])
```

For using method like prediction, They want to 2d not 3d

```
distance_metrics = ['euclidean', 'manhattan', 'cosine']
```

```
for distance_metric in distance_metrics:
    print("Distance Metric:", distance_metric)
    accuracy_scores = []
    cluster_counts = np.zeros((10, 10), dtype=int) # Initialize a table to store cluster counts

    for i in range(5):
        kmeans = KMeans(n_clusters=10, init="random", n_init=10, random_state=i, algorithm="lloyd")
        kmeans.fit(X_train)

        y_pred = kmeans.predict(X_test)
        accuracy = accuracy_score(y_test, y_pred)
        accuracy_scores.append(accuracy)

    # Count how many data points from each label belong to each cluster
    for j in range(len(y_train)):
        cluster_counts[kmeans.labels_[j], y_train[j]] += 1
```

```

# Assign labels to clusters based on cluster_counts
cluster_labels = label_clusters(cluster_counts)

# Training Error
y_train_pred = np.zeros_like(y_train)
for j in range(len(y_train)):
    # Use 1-NN to assign training data point to a cluster based on labeled kmeans centers
    nbrs = NearestNeighbors(n_neighbors=1, algorithm='brute')
    nbrs.fit(kmeans.cluster_centers_)
    distances, indices = nbrs.kneighbors(X_train[j].reshape(1, -1))
    y_train_pred[j] = cluster_labels[indices[0][0]]

train_confusion_matrix = confusion_matrix(y_train, y_train_pred)
train_accuracy = calculate_accuracy(y_train, y_train_pred)
print("Iteration", i+1, "- Accuracy:", accuracy)

y_test_pred = np.zeros_like(y_test)

for j in range(len(y_test)):
    nbrs = NearestNeighbors(n_neighbors=1, algorithm='brute')
    nbrs.fit(kmeans.cluster_centers_)
    distances, indices = nbrs.kneighbors(X_test[j].reshape(1, -1))
    y_test_pred[j] = cluster_labels[indices[0][0]]

test_confusion_matrix = confusion_matrix(y_test, y_test_pred)
test_accuracy = calculate_accuracy(y_test, y_test_pred)

print("Training Accuracy:", train_accuracy)
print("Training Confusion Matrix:\n", train_confusion_matrix)
print("Test Accuracy:", test_accuracy)
print("Test Confusion Matrix:\n", test_confusion_matrix)

print("Mean Accuracy:", np.mean(accuracy_scores))
print()

```

Distance Metric: euclidean

Iteration 1 - Accuracy: 0.1103

Training Accuracy: 0.5057666666666667

Training Confusion Matrix:

```

[[5330  3  17  0  38 162 182  14 177  0]
 [2978 3717  9  0  6  5  8  9 10  0]
 [ 462 362 4193  0 174 329 211 69 158  0]
 [ 248 425 216  0 175 3921 57 48 1041  0]
 [ 331 159 37  0 3189  1 165 1941 19  0]
 [1131 166 15  0 378 1769 122 358 1482  0]
 [ 447 271 86  0 82 28 4913  1 90  0]
 [ 311 327 39  0 1796  5 4 3773 10  0]
 [ 490 296 54  0 193 1130 47 179 3462  0]
 [ 172 239 13  0 2903 85 8 2460 69  0]]

```

Test Accuracy: 0.5074

Test Confusion Matrix:

```

[[876  0  2  0  4 48 30  3 17  0]
 [469 660  1  0  0  2  2  0  1  0]
 [107 59 707  0 30 65 25 11 28  0]
 [ 28 73 40  0 15 696  7  7 144  0]
 [ 38 30  5  0 559  0 36 310  4  0]
 [155 23  4  0 55 287 20 72 276  0]
 [ 74 28 18  0 22  2 795  1 18  0]
 [ 59 59 13  0 292  0  1 603  1  0]
 [ 63 34  7  0 31 207 10 35 587  0]
 [ 25 29  3  0 541  7  4 387 13  0]]

```

Iteration 2 - Accuracy: 0.1929

Training Accuracy: 0.3615333333333333

Training Confusion Matrix:

```

[[5326  21  14  0  38 161 181  0 182  0]
 [3727 2977  9  0  6  5  8  0 10  0]
 [4660 355  70  0 173 330 212  0 158  0]
 [ 779 106 48  0 175 3920 58  0 1045  0]
 [ 217 314 1936  0 3188  1 167  0 19  0]
 [ 486 814 337  0 376 1763 123  0 1522  0]
 [ 569 230  1  0 83 28 4915  0 92  0]
 [ 393 286 3774  0 1793  5  4  0 10  0]
 [ 422 422 181  0 192 1134 47  0 3453  0]
 [ 308 118 2462  0 2899 85  8  0 69  0]]

```

Test Accuracy: 0.358

Test Confusion Matrix:

```

[[873  5  3  0  4 48 30  0 17  0]
 [662 468  0  0  0  2  2  0  1  0]
 [789 84 11  0 30 65 25  0 28  0]

```

```
[133  7  7  0 15 694  7  0 147  0]
[ 36 38 309  0 559  0 35  0  5  0]
[ 76 108 68  0 55 287 19  0 279  0]
[ 89 31  1  0 22  2 794  0 19  0]
[ 74 58 602  0 292  0  1  0  1  0]
[ 58 45 35  0 31 207 10  0 588  0]
[ 46 11 387  0 541  7  4  0 13  0]]
```

Iteration 3 - Accuracy: 0.0965

Training Accuracy: 0.35541666666666666

Training Confusion Matrix:

```
[[5528  21  17  0  14 162 181  0  0  0]
 [3733 2978   9  0   9  5  8  0  0  0]
 [ 801  352 4196  0  69 328 212  0  0  0]
 [1781 106  219  0  48 3922  55  0  0  0]]
```