```
# Importing the dataset from keras
import keras
from keras.datasets import mnist
(x_train, y_train), (x_test, y_test) = mnist.load_data()
# Checking the 'type'
print(type(x_train))
print(type(x_test))
print(type(y_train))
print(type(y_test))
     <class 'numpy.ndarray'>
     <class 'numpy.ndarray'>
     <class 'numpy.ndarray'>
     <class 'numpy.ndarray'>
# Checking the shape
print(x_train.shape)
print(x_test.shape)
print(y_train.shape)
print(y_test.shape)
     (60000, 28, 28)
     (10000, 28, 28)
     (60000,)
     (10000,)
import matplotlib.pyplot as plt
plt.gray() # B/W Images
plt.figure(figsize = (10,9)) # Adjusting figure size
# Displaying a grid of 3x3 images
for i in range(9):
    plt.subplot(3,3,i+1)
    plt.imshow(x train[i])
```

```
<Figure size 432x288 with 0 Axes>
       5
                                   5
                                                               5
      10
                                  10
                                                              10
      15
                                  15
                                                              15
      20
                                  20
                                                              20
      25
                                  25
                                                              25
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                                                                        10
                                                                               20
       0
                                   0
                                                               0
       5
                                                               5
                                   5 -
      10
                                  10
                                                              10
      15
                                  15
# Printing examples in 'y_train'
for i in range(5):
  print(y_train[i])
     5
     0
     4
     1
     9
\# Checking the minimum and maximum values of x_{train}
print(x_train.min())
print(x_train.max())
     0
     255
# Data Normalization
# Conversion to float
x_train = x_train.astype('float32')
x_test = x_test.astype('float32')
# Normalization
x_{train} = x_{train}/255.0
x_test = x_test/255.0
\# Checking the minimum and maximum values of x_{train}
print(x_train.min())
print(x_train.max())
     0.0
     1.0
x_train = x_train.reshape(len(x_train),-1)
```

```
y_train = y_train
import numpy as np
from sklearn.cluster import MiniBatchKMeans
total_clusters = len(np.unique(y_test))
# Initialize the K-Means model
kmeans = MiniBatchKMeans(n clusters = total clusters)
# Fitting the model to training set
kmeans.fit(x train)
     MiniBatchKMeans(n_clusters=10)
kmeans.labels_
     array([4, 8, 2, ..., 6, 0, 9], dtype=int32)
def retrieve_info(cluster_labels,Y):
    Associates most probable label with each cluster in KMeans model
    returns: dictionary of clusters assigned to each label
    .....
  # Initializing
    reference labels = {}
  # For loop to run through each label of cluster label
    for i in range(len(np.unique(kmeans.labels_))):
      index = np.where(cluster labels == i,1,0)
      num = np.bincount(Y[index==1]).argmax()
      reference labels[i] = num
    return reference labels
reference labels = retrieve info(kmeans.labels ,y train)
number labels = np.random.rand(len(kmeans.labels ))
for i in range(len(kmeans.labels_)):
    number_labels[i] = reference_labels[kmeans.labels_[i]]
print(reference_labels)
     \{0: 6, 1: 6, 2: 4, 3: 1, 4: 3, 5: 7, 6: 8, 7: 2, 8: 0, 9: 4\}
# Comparing Predicted values and Actual values
print(number_labels[:20].astype('int'))
print(y_train[:20])
     [3 0 4 1 4 2 1 3 1 4 3 1 3 6 1 7 2 4 6 4]
     [5 0 4 1 9 2 1 3 1 4 3 5 3 6 1 7 2 8 6 9]
```

## **Optimizing the Algorithm**

```
# Function to calculate metrics for the model
def Calmetrics(model.out):
   print('Number of clusters is {}'.format(model.n_clusters))
   print('Inertia : {}'.format(model.inertia_))
   print('Homogeneity : {}'.format(metrics.homogeneity_score(out,model.labels_)))
from sklearn import metrics
cluster_number = [20,45,75,150,200,256]
for cnum in cluster_number:
   total = len(np.unique(y test))
   # Initialize the K-Means model
    kmeans = MiniBatchKMeans(n clusters = cnum)
   # Fitting the model to training set
   kmeans.fit(x train)
   # Calculating the metrics
   Calmetrics(kmeans,y train)
   # Calculating reference labels
   ref = retrieve info(kmeans.labels ,y train)
    # 'number labels' is a list which denotes the number displayed in image
   num = np.random.rand(len(kmeans.labels ))
    for i in range(len(kmeans.labels )):
        num[i] = ref[kmeans.labels_[i]]
    print('Accuracy score : {}'.format(accuracy score(num ,y train)))
   print('\n')

    Number of clusters is 20

    Inertia: 2143990.5
    Homogeneity: 0.6084031883120046
    Accuracy score: 0.705
    Number of clusters is 45
     Inertia: 1899376.5
    Homogeneity: 0.7050452099753474
    Accuracy score: 0.76985
    Number of clusters is 75
    Inertia: 1768545.25
    Homogeneity: 0.7640691660609584
```

Accuracy score : 0.836716666666667

Number of clusters is 150 Inertia: 1615368.5 Homogeneity: 0.8076315310563515 Accuracy score : 0.871 Number of clusters is 200 Inertia: 1556468.5 Homogeneity: 0.8279225976613481 Accuracy score: 0.886816666666667 Number of clusters is 256 Inertia: 1504386.375 Homogeneity: 0.84419418139323 Accuracy score: 0.899216666666667 # Testing model on Testing set # Initialize the K-Means model x test = x test.reshape(len(x test),-1) kmeans = MiniBatchKMeans(n clusters = 256) # Fitting the model to testing set kmeans.fit(x\_test) # Calculating the metrics Calmetrics(kmeans,y\_test) # Calculating the reference labels ref = retrieve info(kmeans.labels , y test) # 'number labels' is a list which denotes the number displayed in image

num = np.random.rand(len(kmeans.labels ))

for i in range(len(kmeans.labels )): num[i] = ref[kmeans.labels\_[i]]

print('Accuracy score : {}'.format(accuracy score(num ,y test))) print('\n')

Number of clusters is 256

Inertia: 240210.5

Homogeneity: 0.8575811778759966

Accuracy score: 0.8937

- Task 1: Run the above code by solving all issues
- Task 2: Optimize the code to improve the accuracy using given tutorial
- Task 3: Try to run the code by using differnet dataset(Other than digits)

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