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## Code:

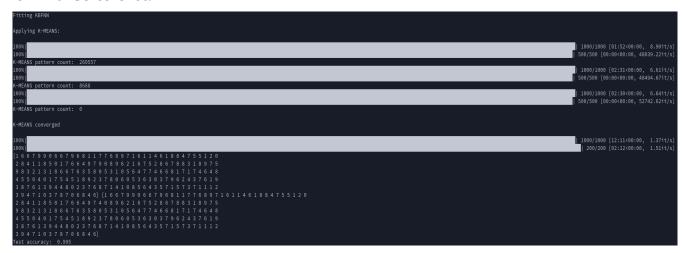
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import numpy as np
from tqdm import tqdm
import seaborn as sn
import matplotlib.pyplot as plt
import pandas as pd
def get_distance(x1, x2):
  sum = 0
  for i in range(len(x1)):
     sum += (x1[i] - x2[i]) ** 2
  return np.sqrt(sum)
def kmeans(X, k, max_iters):
  print('Applying K-MEANS:\n')
  centroids = X[np.random.choice(range(len(X)), k, replace=False)]
  converged = False
  current_iter = 0
  while (not converged) and (current iter < max iters):
     cluster_list = [[] for _ in range(len(centroids))]
    for x in tqdm(X):
       distances_list = []
       for c in centroids:
          distances_list.append(get_distance(c, x))
       cluster list[int(np.argmin(distances list))].append(x)
     cluster_list = list((filter(None, cluster_list)))
     prev_centroids = centroids.copy()
    centroids = []
    for j in tqdm(range(len(cluster list))):
       centroids.append(np.mean(cluster_list[j], axis=0))
    pattern = np.abs(np.sum(prev_centroids) - np.sum(centroids))
     print('K-MEANS pattern count: ', int(pattern))
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converged = (pattern == 0)
     current iter += 1
  print('\nK-MEANS converged\n')
  return np.array(centroids), [np.std(x)] for x in cluster list
class RBF:
  def __init__(self, X, y, tX, ty, num_of_classes,
       k, std from clusters=True):
     self.X = X
     self.y = y
    self.tX = tX
    self.ty = ty
     self.number of classes = num of classes
     self.k = k
     self.std from clusters = std from clusters
  def convert_to_one_hot(self, x, num_of_classes):
     arr = np.zeros((len(x), num_of_classes))
    for i in range(len(x)):
       c = int(x[i])
       arr[i][c] = 1
    return arr
  def rbf(self, x, c, s):
     distance = get_distance(x, c)
    return 1 / np.exp(-distance / s ** 2)
  def rbf_list(self, X, centroids, std_list):
     RBF_list = []
    for x in tqdm(X):
       RBF_list.append([self.rbf(x, c, s) for (c, s) in zip(centroids, std_list)])
    return np.array(RBF_list)
  def plot_confusion_matrix(self):
     mat = [[0 \text{ for in range}(10)] \text{ for in range}(10)]
     print(self.ty, self.pred ty)
    for idx, val in enumerate(self.ty):
       mat[int(val)][int(self.pred_ty[idx])] += 1
    plt.figure(figsize = (10,7))
     sn.heatmap(mat, annot=True)
```

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plt.show()
  def fit(self):
     print('Fitting KBFNN\n')
     self.centroids, self.std list = kmeans(self.X, self.k, max iters=1000)
    if not self.std from clusters:
       dMax = np.max([get_distance(c1, c2) for c1 in self.centroids for c2 in self.centroids])
       self.std list = np.repeat(dMax / np.sqrt(2 * self.k), self.k)
     RBF_X = self.rbf_list(self.X, self.centroids, self.std_list)
     self.w = np.linalg.pinv(RBF X.T @ RBF X) @ RBF X.T @ self.convert to one hot(self.y,
self.number of classes)
     RBF list tst = self.rbf list(self.tX, self.centroids, self.std list)
     self.pred_ty = RBF_list_tst @ self.w
     self.pred_ty = np.array([np.argmax(x) for x in self.pred_ty])
     diff = self.pred ty - self.ty
     self.plot confusion matrix()
    print('Test accuracy: ', len(np.where(diff == 0)[0]) / len(diff))
if __name__ == '__main__':
  df = pd.read csv('mnist train.csv')
  df = df.sample(frac=1)
  train_y = df.iloc[0:1000, 0].to_numpy()
  train_x = df.iloc[0:1000, 1:].to_numpy()
  test_y = df.iloc[0:200, 0].to_numpy()
  test_x = df.iloc[0:200, 1:].to_numpy()
  RBF CLASSIFIER = RBF(train x, train y, test x, test y, num of classes=10,
       k=500, std from clusters=False)
  RBF CLASSIFIER.fit()
```

## **Output:**

## **Terminal Screenshot:**



Accuracy: 0.995 (99.5%)

## **Confusion matrix:**

