Handwritten Digits Recognition using Gaussian Naive Bayes Model (Scikit-Learn Dataset)

```
In [ ]: | # import theScikit-learn dataset
         #1797 samples ans 8*8 grid of pixels
         from sklearn.datasets import load digits
         digits = load digits()
         digits.images.shape
In [21]: #Visualize the dataset
         import matplotlib.pyplot as plt
         fig, axes = plt.subplots(10,10, figsize=(8,8),
                                subplot_kw={'xticks':[], 'yticks':[]},
                                 gridspec_kw=dict(hspace=0.1, wspace=0.1))
         for i, ax in enumerate(axes.flat):
             ax.imshow(digits.images[i], cmap='binary', interpolation = 'nearest'),
             ax.text(0.05,0.05,str(digits.target[i]),
                    transform=ax.transAxes, color = 'blue')
In [22]: #feature matrix, two dimentional [n_samples,n_features]
         #1797 samples
         #64 features
         X = digits.data
         X.shape
Out[22]: (1797, 64)
In [23]: #lebel or target array, one dimentional
         y=digits.target
         y.shape
Out[23]: (1797,)
In [24]: | #Using Manifold learning algorithm or Isomap to reduce the dimention of fetures from 64 to 2
         from sklearn.manifold import Isomap
         iso = Isomap(n_components =2)
         iso.fit(X)
         data_projected = iso.transform(X)
         data_projected.shape
Out[24]: (1797, 2)
In [34]: #now plot the data
         plt.scatter(data_projected[:,0], data_projected[:,1], c = y,
                    edgecolor='none', alpha=10,
                    cmap = plt.cm.get_cmap('Spectral',10))
         plt.colorbar(ticks = range(10))
         plt.clim(-0.5, 9.5)
           200
           150
           100
            50
            0
           -50
          -100
                   -100
In [37]: #splitting train test set
         from sklearn.model_selection import train_test_split
         X_train, X_test, y_train, y_test = train_test_split(X, y, random_state=0)
In [39]: | #fit into Gaussian Naive Bayes model
         from sklearn.naive_bayes import GaussianNB
         GNB= GaussianNB()
         GNB.fit(X train, y train)
         y_GNB = GNB.predict(X_test)
In [40]: #test score
         from sklearn.metrics import accuracy_score
         accuracy_score(y_test,y_GNB)
Out[40]: 0.83333333333333333
In [45]: | #making sklearn confusion matrix using seaborn heatmap
         import seaborn as sns
         from sklearn.metrics import confusion matrix
         matrix=confusion matrix(y test,y GNB)
         sns.heatmap(matrix, square = True, annot = True, cbar=False)
         plt.xlabel('Predicted Value')
         plt.ylabel('True Value')
Out[45]: Text(91.68, 0.5, 'True Value')
                      0 0 0 0 0 4 0
                           0 0 0 15 0
              0 1 0 1 0 43
              0 0 1 0 0 0 51
              0 0 0 0 1 0 0 47
            ∞ - 0 6 0 1 0 1 0 2 38
              0 2 0 4 1 0 0 3 7 30
              0 1 2 3 4 5 6 7 8 9
                      Predicted Value
In [46]: #visualize data with predicted label
         #green for correct prediction
         #red for incorrect prediction
         fig, axes = plt.subplots(10,10, figsize=(8,8),
                                subplot kw={'xticks':[], 'yticks':[]},
                                 gridspec kw=dict(hspace=0.1, wspace=0.1))
         for i, ax in enumerate(axes.flat):
             ax.imshow(digits.images[i], cmap='binary', interpolation = 'nearest'),
             ax.text(0.05, 0.05, str(y_GNB[i]),
                    transform=ax.transAxes,
                     color = 'green' if (y_test[i]==y_GNB[i])
                                         else 'red')
```