CS353 Machine Learning Lab

Image classification & SVM (12/03/21)

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TASK

Write a program to demonstrate Image Recognition. Classify the data using svm and try to identify the images present in the data set.

Dataset

We are using digits dataset from sklearn.datasets

https://scikit-learn.org/stable/modules/generated/sklearn.datasets.load_digits.html

We've implemented 3 types of kernels:

- 1. Linear Kernel
- 2. Polynomial Kernel
- 3. Radial Basis Function (RBF) kernel

Imports

```
In [1]:
        import numpy as np
        import pandas as pd
        from sklearn import datasets, svm, metrics
        from sklearn.model_selection import train_test_split
        from sklearn.metrics import mean squared error, mean absolute error, explained variance score, r2 sc
        from sklearn.metrics import confusion_matrix, accuracy_score,classification_report
```

Loading dataset

```
In [2]:
        digits = datasets.load_digits()
        images = digits.images
        print('Digits dataset keys \n{}'.format(digits.keys()))
        Digits dataset keys
        dict_keys(['data', 'target', 'target_names', 'images', 'DESCR'])
In [3]:
        print('shape of datset: {} \nand target: {}'.format(digits.data.shape, digits.target.shape))
        print('shape of the images: {}'.format(digits.images.shape))
        shape of datset: (1797, 64)
        and target: (1797,)
        shape of the images: (1797, 8, 8)
```

Splitting data

```
In [4]: n_samples = len(digits.images)
        data_images = digits.images.reshape((n_samples, -1))
        X_train, X_test, y_train, y_test = train_test_split(data_images, digits.target)
In [5]: print('Training data and target sizes: \n{}, {}'.format(X_train.shape,y_train.shape))
        print('Test data and target sizes: \n{}, {}'.format(X_test.shape, y_test.shape))
        Training data and target sizes:
```

```
(1347, 64), (1347,)
Test data and target sizes:
(450, 64), (450,)
```

Implementing SVM

clf.fit(X_train,y_train)

differentiate between them.

SVM: Support Vector Machine is a supervised classification algorithm where we draw a line between two different categories to

In [6]: | clf = svm.SVC(kernel="linear")

Linear Kernel

```
Out[6]: SVC(C=1.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0,
            decision_function_shape='ovr', degree=3, gamma='scale', kernel='linear',
            max_iter=-1, probability=False, random_state=None, shrinking=True,
            tol=0.001, verbose=False)
In [7]: y_pred = clf.predict(X_test)
        print(accuracy_score(y_test, y_pred))
        0.98222222222222
```

In [8]: clf_poly = svm.SVC(kernel="poly", degree = 5)

Polynomial Kernel

```
clf_poly.fit(X_train,y_train)
Out[8]: SVC(C=1.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0,
            decision_function_shape='ovr', degree=5, gamma='scale', kernel='poly',
            max_iter=-1, probability=False, random_state=None, shrinking=True,
            tol=0.001, verbose=False)
In [9]: y_pred2 = clf_poly.predict(X_test)
        print(accuracy_score(y_test, y_pred2))
        0.98444444444445
```

In [10]: clf_rbf = svm.SVC(kernel="rbf", gamma=0.001) clf_rbf.fit(X_train,y_train)

RBG Kernel

```
Out[10]: SVC(C=1.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0,
             decision_function_shape='ovr', degree=3, gamma=0.001, kernel='rbf',
             max_iter=-1, probability=False, random_state=None, shrinking=True,
             tol=0.001, verbose=False)
In [11]: y_pred3 = clf_rbf.predict(X_test)
         print(accuracy_score(y_test, y_pred3))
         0.98888888888888
```

1

original predicted

4

4

Out[12]:

```
result = pd.DataFrame({'original' : y_test, 'predicted' : y_pred})
In [12]:
```

```
0
   3
            0
                       0
  ...
                       0
 445
 446
            2
                       2
                       7
 447
                       6
 448
            6
 449
450 rows × 2 columns
```

[[48 0 0

0 0 0 0

```
In [13]: print("Accuracy = {} % ".format(accuracy_score(y_test, y_pred)*100))
         Accuracy = 98.22222222223 %
In [14]:
         print(metrics.confusion_matrix(y_test, y_pred))
```

```
[ 0 48
       0
           0
              0
    0 44
                             0]
    0
        0 46
           0 54
       0
 0
    0
           0
             0 43
        0
           0
              0
                 0
                             1]
 0
    0
       0
           0
              0
                 0
                    0
                       0 35
                             0]
 0
    1
       0
           0
              0
                 0
                   0 1 1 36]]
```

```
In [15]: | print(classification_report(y_test, y_pred))
                         precision
                                       recall f1-score
                                                           support
                     0
                              1.00
                                         1.00
                                                   1.00
                                                                48
                     1
                              0.96
                                         1.00
                                                   0.98
                                                                48
                     2
                              1.00
                                         1.00
                                                   1.00
                                                                44
```

```
3
                     1.00
                                0.98
                                           0.99
                                                         47
            4
                     1.00
                                1.00
                                           1.00
                                                         54
            5
                     0.98
                                0.96
                                           0.97
                                                         45
            6
                     1.00
                                0.98
                                           0.99
                                                         45
            7
                     0.98
                                0.98
                                           0.98
                                                         45
            8
                     0.97
                                0.97
                                           0.97
                                                         36
                     0.92
                                0.95
                                                         38
                                           0.94
                                           0.98
                                                        450
    accuracy
                     0.98
                                0.98
                                           0.98
                                                        450
   macro avg
weighted avg
                     0.98
                                0.98
                                           0.98
                                                        450
```