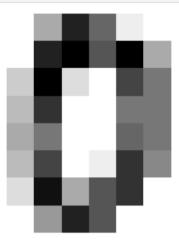


import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
from sklearn import model_selection
from sklearn import cluster
from sklearn import metrics
import seaborn as sns
import warnings
warnings.filterwarnings('ignore')

In [2]: from sklearn.datasets import load_digits
 data, labels = load_digits(return_X_y=True)

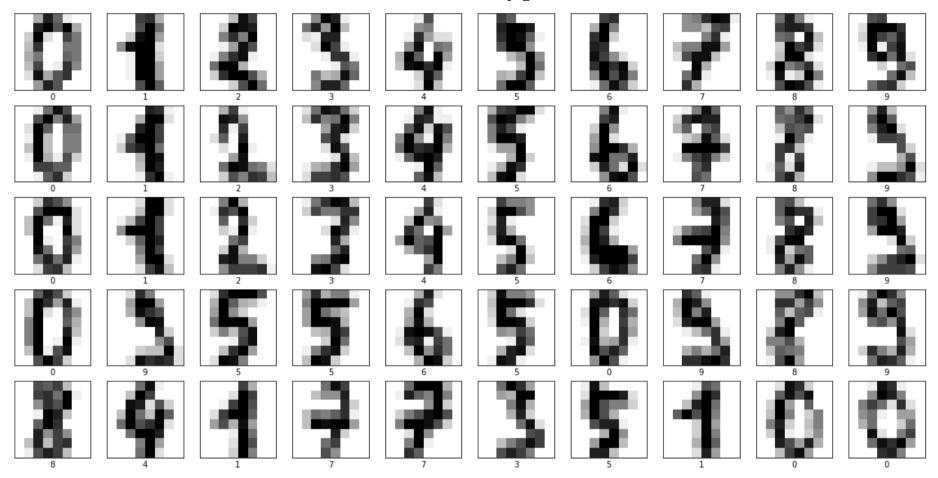
```
In [3]: data[0]
Out[3]: array([0., 0., 5., 13., 9., 1., 0., 0., 0., 0., 13., 15., 10.,
              15., 5., 0., 0., 3., 15., 2., 0., 11., 8., 0., 0., 4.,
              12., 0., 0., 8., 8., 0., 0., 5., 8., 0., 0., 9., 8.,
               0., 0., 4., 11., 0., 1., 12., 7., 0., 0., 2., 14., 5.,
              10., 12., 0., 0., 0., 6., 13., 10., 0., 0., 0.])
In [4]: labels[0]
Out[4]: 0
In [5]: b=data[0]
        a= b.reshape(8,8)
        plt.imshow(a, cmap="binary")
        <matplotlib.image.AxesImage at 0x19e7ea7d570>
Out[5]:
        1
        2
        3 -
        5
        6
In [3]: digits=np.unique(labels)
        digits
Out[3]: array([0, 1, 2, 3, 4, 5, 6, 7, 8, 9])
In [4]: n_samples, n_features = data.shape
        n_samples, n_features
```

```
Out[4]: (1797, 64)
In [5]: first_digit=data[0]
first_digit_image= first_digit.reshape(8,8)
plt.imshow(first_digit_image, cmap="binary")
plt.axis("off")
plt.show()
```



```
In [6]: labels[0]
Out[6]:

In [7]: plt.figure(figsize=(20,10))
    for i in range(50):
        plt.subplot(5,10,i+1)
        plt.yticks([])
        plt.yticks([])
        plt.grid(True)
        x=data[i].reshape(8,8)
        plt.imshow(x, cmap=plt.cm.binary)
        plt.xlabel(labels[i])
    plt.show()
```



```
In [8]: from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test = train_test_split(data, labels, train_size=0.8, random_state=2)
```

```
In [9]: X_train.shape, X_test.shape, y_train.shape, y_test.shape
```

Out[9]: ((1437, 64), (360, 64), (1437,), (360,))

```
In [10]: unique,count = np.unique(y_train,return_counts=True)
unique,count/y_train.shape
```

Out[10]: (array([0, 1, 2, 3, 4, 5, 6, 7, 8, 9]), array([0.10160056, 0.0960334, 0.10160056, 0.10229645, 0.10160056, 0.0967293, 0.10160056, 0.0967293, 0.0960334, 0.10577592]))

```
In [11]:
         unique,count=np.unique(y test,return counts=True)
         unique,count/y test.shape
         (array([0, 1, 2, 3, 4, 5, 6, 7, 8, 9]),
Out[11]:
          array([0.08888889, 0.12222222, 0.08611111, 0.1
                                                               , 0.09722222,
                 0.11944444, 0.09722222, 0.11111111, 0.1
                                                               , 0.077777781))
In [12]: from imblearn.over_sampling import SMOTE
         smote = SMOTE()
In [13]: #x train, y train = smote.fit resample(X train.astype('float'),y train)
         #unique,count = np.unique(y train,return counts=True)
In [14]:
         #unique,count/y train.shape
In [15]: from sklearn.linear model import LogisticRegression
         log reg = LogisticRegression()
In [16]: log_reg.fit(X_train, y_train)
         ▼ LogisticRegression
Out[16]:
         LogisticRegression()
         log reg.intercept
In [17]:
         array([ 0.00032198, -0.07937238, -0.00385451, 0.01053846, 0.00497534,
                -0.00169985, -0.00646291, 0.0135647, 0.09386533, -0.03187616])
         log_reg.coef_[:1]
In [18]:
```

```
array([[ 0.00000000e+00, -8.15705766e-03, -2.32205836e-02,
                  4.61992144e-02, 2.20580221e-02, -9.17586844e-02,
                 -1.35182879e-01, -1.25730676e-02, -1.33547647e-04,
                 -2.76676448e-02, 1.32057172e-02, 8.80933587e-02,
                  4.58700949e-02, 1.25854630e-01, -5.14246170e-02,
                 -1.36075029e-02, -3.68904186e-04, 1.51055813e-02,
                  1.68600194e-01, 3.08444242e-03, -2.76660795e-01,
                  1.34407265e-01, 8.32086378e-02, -4.42461295e-03,
                 -1.77917038e-04, 9.16759492e-02, 1.50048819e-01,
                 -1.39245474e-01, -4.38755662e-01, -2.01798036e-02,
                  1.85175261e-01, -9.23918256e-05, 0.00000000e+00,
                  1.81283421e-01, 1.06883578e-01, -2.44530303e-01,
                 -3.99549904e-01, 3.30743059e-02, 1.29451516e-01,
                  0.00000000e+00, -3.41447798e-04, 2.35667731e-02,
                  1.77988133e-01, -6.96931855e-02, -1.35088531e-01,
                  1.38911449e-01, 3.90999609e-02, -4.18488588e-04,
                 -6.90084203e-04, -5.73505700e-02, 9.71085744e-02,
                  1.16909493e-02, 1.40547033e-01, 7.68943581e-02,
                 -4.26785128e-02, -9.11547234e-03, -1.43540636e-05,
                 -1.36853632e-02, -6.09164000e-02, 9.76872006e-02,
                 -2.79778888e-02, -3.99407490e-02, -4.30172735e-02,
                 -1.57015176e-02]])
In [19]: y predict=log reg.predict(X test)
         y predict[:1]
         array([4])
Out[19]:
In [20]: log reg.score(X test, y test)
         0.94444444444444
Out[20]:
In [21]: from sklearn.metrics import accuracy score
         accuracy score(y test,y predict)
         0.94444444444444
Out[21]:
         plt.figure(figsize=(20,10))
In [22]:
         for i in range(50):
             plt.subplot(5,10,i+1)
             plt.xticks([])
             plt.yticks([])
```

```
plt.grid(True)
   x=X_test[i].reshape(8,8)
    plt.imshow(x, cmap=plt.cm.binary)
    plt.xlabel(y_predict[i])
plt.show()
```

In [23]: from sklearn.metrics import classification_report
 print(classification_report(y_test, y_predict))

In [29]:

Out[29]:

```
0.97
                                                 0.98
                             1.00
                                                             32
                             0.95
                                       0.93
                                                 0.94
                                                             44
                             1.00
                                       1.00
                                                 1.00
                                                             31
                     3
                             0.94
                                       0.92
                                                 0.93
                                                             36
                             0.94
                                       0.89
                                                 0.91
                                                             35
                                                 0.96
                             0.98
                                       0.95
                                                             43
                                                 0.97
                             1.00
                                       0.94
                                                             35
                                       0.97
                             0.97
                                                 0.97
                                                             40
                    8
                             0.81
                                       0.97
                                                 0.89
                                                             36
                             0.86
                                       0.89
                                                 0.88
                                                             28
                                                 0.94
                                                            360
             accuracy
                             0.95
                                       0.94
                                                 0.94
                                                            360
            macro avg
                             0.95
         weighted avg
                                       0.94
                                                 0.95
                                                             360
         from sklearn.preprocessing import QuantileTransformer
In [24]:
         quantile transformer = QuantileTransformer(output distribution='normal', random state=2)
In [25]: X_train_trans = quantile_transformer.fit_transform(X_train)
         X_test_trans = quantile_transformer.transform(X_test)
In [26]:
         lreg = LogisticRegression()
         lreg.fit(X_train_trans,y_train)
In [28]:
Out[28]:
         ▼ LogisticRegression
         LogisticRegression()
         lreg.score(X_test, y_test)
         0.84722222222222
In [30]: y_pred=lreg.predict(X_test_trans)
In [31]:
         accuracy_score(y_test, y_pred)
```

precision

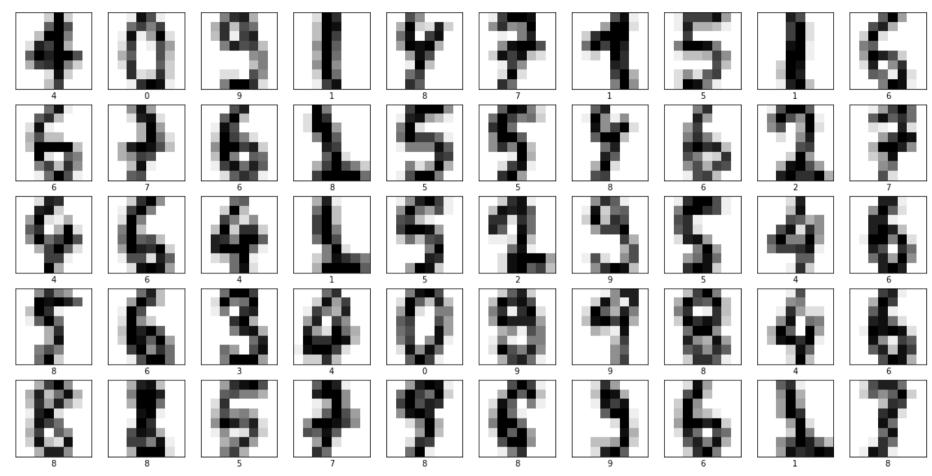
recall f1-score

support

```
0.9333333333333333
Out[31]:
         plt.figure(figsize=(20,10))
In [32]:
         for i in range(50):
              plt.subplot(5,10,i+1)
              plt.xticks([])
             plt.yticks([])
             plt.grid(True)
             x=X_test[i].reshape(8,8)
              plt.imshow(x, cmap=plt.cm.binary)
             plt.xlabel(y pred[i])
         plt.show()
```

```
print(classification report(y test, y pred))
                       precision
                                     recall f1-score
                                                        support
                     0
                             0.97
                                       0.97
                                                 0.97
                                                             32
                             0.98
                                       0.93
                                                 0.95
                                                             44
                     1
                                                 0.95
                     2
                             0.91
                                       1.00
                                                             31
                                       0.89
                                                 0.91
                     3
                             0.94
                                                             36
                                                 0.91
                                                             35
                             0.94
                                       0.89
                                                 0.94
                                                             43
                     5
                             0.93
                                       0.95
                             1.00
                                       0.94
                                                 0.97
                                                             35
                             0.97
                                       0.97
                                                             40
                     7
                                                 0.97
                                                 0.88
                             0.85
                                       0.92
                                                             36
                             0.83
                                       0.86
                                                 0.84
                     9
                                                             28
                                                 0.93
                                                            360
             accuracy
                                                 0.93
            macro avg
                             0.93
                                       0.93
                                                            360
         weighted avg
                             0.94
                                                 0.93
                                       0.93
                                                            360
In [34]: from sklearn.preprocessing import Normalizer
          normalizer = Normalizer().fit(X train) # fit does nothing
          normalizer
Out[34]:
         ▼ Normalizer
         Normalizer()
In [35]: X_train_normalized = normalizer.transform(X_train)
          X test normalized = normalizer.transform(X test)
         X train normalized
```

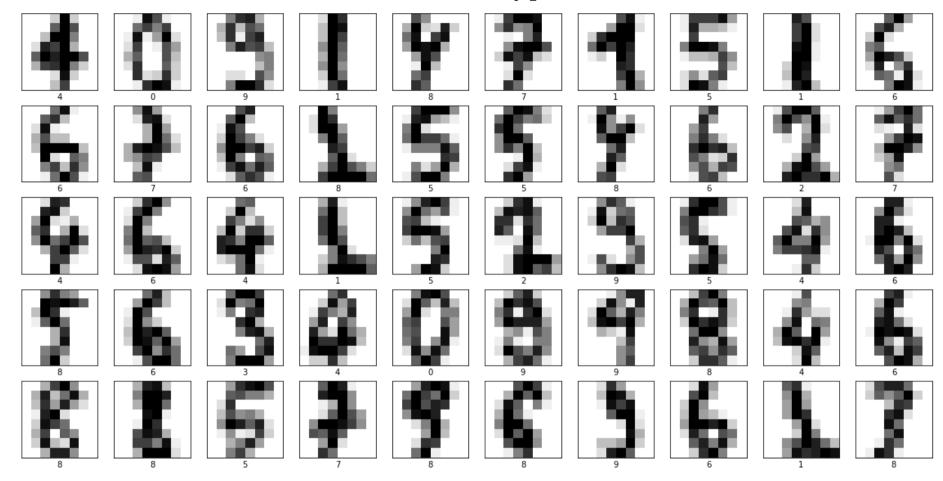
```
array([[0.
                           , 0.
                                       , 0.16224277, ..., 0.25958843, 0.12979422,
Out[35]:
                [0.
                           , 0.01663666, 0.19963986, ..., 0.1497299 , 0.06654662,
                 0.
                [0.
                           , 0.0625229 , 0.2500916 , ..., 0.2500916 , 0.1250458 ,
                 0.
                . . . ,
                           , 0.
                                       , 0. , ..., 0.20416196, 0.05444319,
                [0.
                 0.
                                       , 0.10519842, ..., 0.28052912, 0.28052912,
                [0.
                           , 0.
                 0.21039684],
                [0.
                           , 0.05035088, 0.23497078, ..., 0. , 0.
                 0.
                           11)
In [36]: lreg=LogisticRegression()
         lreg.fit(X train normalized, y train)
Out[36]: ▼ LogisticRegression
         LogisticRegression()
In [37]: lreg.score(X test, y test)
         0.8444444444444444
Out[37]:
In [38]: y_pred=lreg.predict(X_test)
In [39]: plt.figure(figsize=(20,10))
         for i in range(50):
             plt.subplot(5,10,i+1)
             plt.xticks([])
             plt.yticks([])
             plt.grid(True)
             x=X test[i].reshape(8,8)
             plt.imshow(x, cmap=plt.cm.binary)
             plt.xlabel(y pred[i])
         plt.show()
```



In [40]: print(classification_report(y_test, y_pred))

```
precision
                           recall f1-score
                                            support
                  1.00
                            1.00
                                       1.00
                                                   32
           0
           1
                  0.96
                             0.61
                                       0.75
                                                   44
                  1.00
                            0.81
                                       0.89
                                                   31
           2
           3
                   0.94
                            0.81
                                       0.87
                                                   36
                            0.83
                                       0.89
           4
                   0.97
                                                   35
           5
                  1.00
                            0.86
                                       0.92
                                                   43
                  0.92
                            0.94
                                       0.93
                                                   35
                            0.88
                                       0.93
                                                   40
           7
                   1.00
           8
                   0.45
                             0.97
                                       0.61
                                                   36
                  0.79
                                       0.79
                                                   28
           9
                             0.79
                                       0.84
                                                  360
   accuracy
                   0.90
                                       0.86
                                                  360
  macro avg
                             0.85
weighted avg
                  0.91
                             0.84
                                       0.86
                                                  360
```

```
In [41]:
    plt.figure(figsize=(20,10))
    for i in range(50):
        plt.subplot(5,10,i+1)
        plt.yticks([])
        plt.grid(True)
        x=X_test[i].reshape(8,8)
        plt.imshow(x, cmap=plt.cm.binary)
        plt.xlabel(y_pred[i])
    plt.show()
```



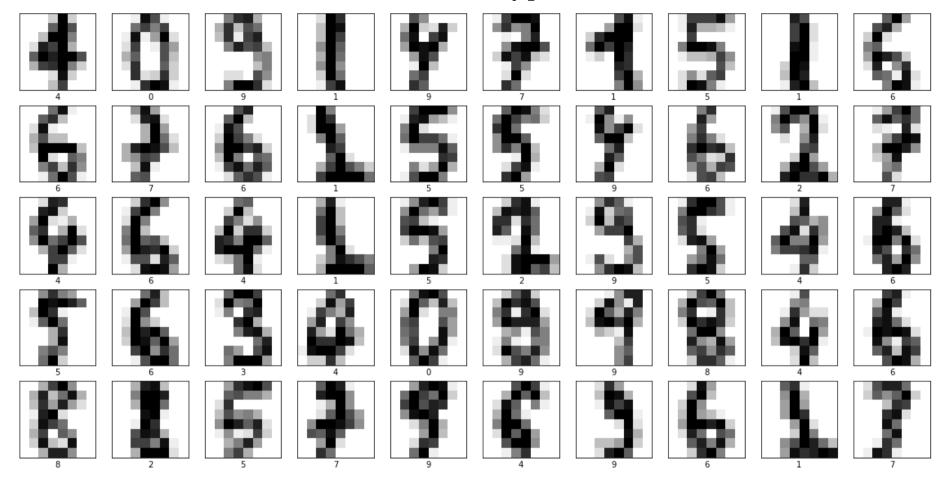
• Without Normalization giving better accuracy

Pipeline

```
Out[43]:
                  ▶ KMeans
           ▶ LogisticRegression
         pipeline.score(X test, y test)
In [44]:
         0.94722222222222
Out[44]:
In [45]: y pred=pipeline.predict(X test)
         y_pred
         array([4, 0, 9, 1, 9, 7, 1, 5, 1, 6, 6, 7, 6, 1, 5, 5, 9, 6, 2, 7, 4, 6,
Out[45]:
                4, 1, 5, 2, 9, 5, 4, 6, 5, 6, 3, 4, 0, 9, 9, 8, 4, 6, 8, 2, 5, 7,
                9, 4, 9, 6, 1, 7, 0, 1, 9, 7, 3, 3, 1, 8, 8, 8, 9, 8, 5, 9, 4, 8,
                3, 5, 8, 4, 3, 9, 3, 8, 7, 3, 3, 0, 8, 7, 2, 8, 5, 3, 8, 7, 6, 4,
                6, 2, 2, 0, 1, 1, 5, 3, 5, 7, 1, 8, 2, 2, 6, 4, 6, 7, 3, 7, 3, 9,
                4, 7, 0, 3, 5, 1, 5, 0, 3, 9, 2, 7, 3, 2, 0, 8, 1, 9, 2, 1, 5, 1,
                0, 3, 4, 3, 0, 9, 3, 2, 2, 7, 3, 1, 6, 7, 2, 8, 3, 1, 1, 6, 4, 8,
                2, 1, 8, 4, 1, 3, 1, 1, 9, 5, 4, 8, 7, 4, 8, 9, 5, 7, 6, 9, 4, 0,
                4, 0, 0, 9, 0, 6, 5, 8, 8, 3, 7, 9, 2, 0, 3, 2, 7, 3, 0, 2, 6, 5,
                2, 7, 0, 6, 9, 3, 1, 1, 3, 5, 2, 3, 5, 2, 1, 2, 9, 4, 6, 5, 5, 5,
                9, 7, 1, 5, 9, 6, 3, 7, 1, 7, 5, 1, 7, 2, 7, 5, 5, 4, 8, 6, 6, 2,
                8, 7, 3, 7, 8, 0, 3, 5, 7, 4, 3, 4, 1, 0, 3, 3, 5, 4, 1, 3, 1, 2,
                5, 1, 4, 0, 3, 1, 5, 5, 7, 4, 0, 1, 0, 8, 5, 5, 5, 4, 0, 1, 8, 6,
                2, 1, 1, 1, 7, 9, 6, 7, 9, 7, 0, 4, 8, 6, 9, 2, 7, 2, 1, 0, 8, 2,
                8, 6, 5, 7, 8, 4, 5, 7, 8, 6, 4, 2, 6, 9, 3, 0, 0, 8, 0, 6, 6, 7,
                1, 4, 5, 6, 9, 7, 2, 8, 5, 1, 2, 4, 1, 8, 8, 3, 6, 0, 8, 0, 6, 5,
                5, 7, 8, 0, 4, 1, 4, 5])
         accuracy score(y test, y pred)
In [46]:
          0.94722222222222
Out[46]:
         print(classification report(y test, y pred))
```

```
recall f1-score
             precision
                                             support
                   1.00
                             0.97
                                       0.98
                                                   32
           0
           1
                   0.95
                             0.93
                                       0.94
                                                   44
                   0.97
                                       0.98
                                                   31
           2
                             1.00
           3
                   0.95
                             0.97
                                       0.96
                                                   36
                                       0.90
           4
                   0.91
                             0.89
                                                   35
           5
                   0.98
                             0.98
                                       0.98
                                                   43
                   0.97
                             0.94
                                       0.96
                                                   35
                   0.97
                             0.95
                                       0.96
                                                   40
           7
           8
                   0.92
                             0.94
                                       0.93
                                                   36
                   0.83
                                       0.86
                                                   28
           9
                             0.89
                                       0.95
                                                  360
   accuracy
                   0.95
                                       0.95
                                                  360
  macro avg
                             0.95
weighted avg
                   0.95
                                       0.95
                             0.95
                                                  360
```

```
In [48]: plt.figure(figsize=(20,10))
    for i in range(50):
        plt.subplot(5,10,i+1)
        plt.yticks([])
        plt.grid(True)
        x=X_test[i].reshape(8,8)
        plt.imshow(x, cmap=plt.cm.binary)
        plt.xlabel(y_pred[i])
    plt.show()
```



from sklearn.model_selection import GridSearchCV
param_grid = dict(kmeans__n_clusters=range(2, 100))
grid_clf = GridSearchCV(pipeline, param_grid, cv=3, verbose=2)
grid_clf.fit(X_train, y_train)

Fitting 3 folds for each of 98 candidates, totalling 294 fits		
[CV] ENDkmeansn_clusters=2;		0.0s
[CV] ENDkmeans_n_clusters=2;		0.0s
[CV] ENDkmeans_n_clusters=2;	total time=	0.0s
[CV] ENDkmeans_n_clusters=3;	total time=	0.0s
[CV] ENDkmeans_n_clusters=3;	total time=	0.0s
[CV] ENDkmeans_n_clusters=3;	total time=	0.0s
[CV] ENDkmeans_n_clusters=4;	total time=	0.0s
[CV] ENDkmeans_n_clusters=4;	total time=	0.0s
[CV] ENDkmeans_n_clusters=4;	total time=	0.0s
[CV] ENDkmeans_n_clusters=5;	total time=	0.0s
[CV] ENDkmeans_n_clusters=5;	total time=	0.0s
[CV] ENDkmeans_n_clusters=5;	total time=	0.0s
[CV] ENDkmeans_n_clusters=6;	total time=	0.0s
[CV] ENDkmeans_n_clusters=6;	total time=	0.0s
[CV] ENDkmeans_n_clusters=6;	total time=	0.0s
[CV] ENDkmeans_n_clusters=7;	total time=	0.0s
[CV] ENDkmeans_n_clusters=7;	total time=	0.0s
[CV] ENDkmeans_n_clusters=7;	total time=	0.0s
[CV] ENDkmeans_n_clusters=8;	total time=	0.0s
[CV] ENDkmeans_n_clusters=8;	total time=	0.0s
[CV] ENDkmeans_n_clusters=8;	total time=	0.0s
[CV] ENDkmeans_n_clusters=9;	total time=	0.0s
[CV] ENDkmeans_n_clusters=9;	total time=	0.0s
[CV] ENDkmeansn_clusters=9;	total time=	0.0s
[CV] ENDkmeansn_clusters=10;	total time=	0.0s
[CV] ENDkmeansn_clusters=10;	total time=	0.0s
[CV] ENDkmeansn_clusters=10;	total time=	0.0s
[CV] ENDkmeansn_clusters=11;	total time=	0.0s
[CV] ENDkmeansn_clusters=11;	total time=	0.0s
[CV] ENDkmeansn_clusters=11;	total time=	0.0s
[CV] ENDkmeansn_clusters=12;		0.0s
[CV] ENDkmeansn_clusters=12;		0.0s
[CV] ENDkmeansn_clusters=12;	total time=	0.0s
[CV] ENDkmeansn_clusters=13;	total time=	0.0s
[CV] ENDkmeansn_clusters=13;		0.0s
[CV] ENDkmeansn_clusters=13;	total time=	0.0s
[CV] ENDkmeansn_clusters=14;	total time=	0.0s
[CV] ENDkmeansn_clusters=14;	total time=	0.0s
[CV] ENDkmeansn_clusters=14;	total time=	0.0s
[CV] ENDkmeansn_clusters=15;		0.0s
[CV] ENDkmeansn_clusters=15;	total time=	0.0s
[CV] ENDkmeansn_clusters=15;	total time=	0.0s
[CV] ENDkmeansn_clusters=16;	total time=	0.0s

			g	
[CV] ENDkmeansn_o	clusters=16;	total	time=	0.0s
[CV] ENDkmeansn_o	clusters=16;	total	time=	0.0s
[CV] ENDkmeansn_o	clusters=17;	total	time=	0.0s
[CV] ENDkmeansn_o	clusters=17;	total	time=	0.0s
[CV] ENDkmeansn_o	clusters=17;	total	time=	0.0s
[CV] ENDkmeansn_o	clusters=18;	total	time=	0.0s
[CV] ENDkmeansn_c	clusters=18;	total	time=	0.0s
[CV] ENDkmeansn_c	clusters=18;	total	time=	0.0s
[CV] ENDkmeansn_o	clusters=19;	total	time=	0.0s
[CV] ENDkmeansn_o	clusters=19;	total	time=	0.0s
[CV] ENDkmeansn_o	clusters=19;	total	time=	0.0s
[CV] ENDkmeansn_o	clusters=20;	total	time=	0.0s
[CV] ENDkmeansn_o				0.0s
[CV] ENDkmeansn_o				0.0s
[CV] ENDkmeansn_o				0.0s
[CV] ENDkmeansn_c				0.0s
[CV] ENDkmeansn_c	•			0.1s
[CV] ENDkmeansn_o				0.0s
[CV] ENDkmeansn_c	•			0.0s
[CV] ENDkmeansn_c				0.0s
[CV] ENDkmeansn_o				0.0s
[CV] ENDkmeansn_o				0.0s
[CV] ENDkmeansn_o				0.0s
[CV] ENDkmeansn_o				0.0s
[CV] ENDkmeansn_o				0.0s
[CV] ENDkmeansn_c				0.0s
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[CV] ENDkmeansn_o				0.0s
[CV] ENDkmeansn_c				0.1s
[CV] ENDkmeansn_c				0.1s
[CV] ENDkmeansn_c				0.1s
[CV] ENDkmeans_n_c				0.1s
[CV] ENDkmeans_n_c				0.1s
[CV] ENDkmeansn_c				0.1s
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[CV] ENDkmeansn_c				0.1s
[CV] ENDkmeans_n_c	•			0.1s
[CV] ENDkmeans_n_c				0.1s
[CV] ENDkmeansn_c				0.1s
[CV] ENDkmeans_n_c	•			0.1s
[CV] ENDkmeansn_c				0.1s
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		g	
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[CV] ENDkmeans_n_clusters=31	total	time=	0.1s
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[CV] ENDkmeansn_clusters=43			0.2s
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[CV] ENDkmeans_n_clusters=43			0.2s
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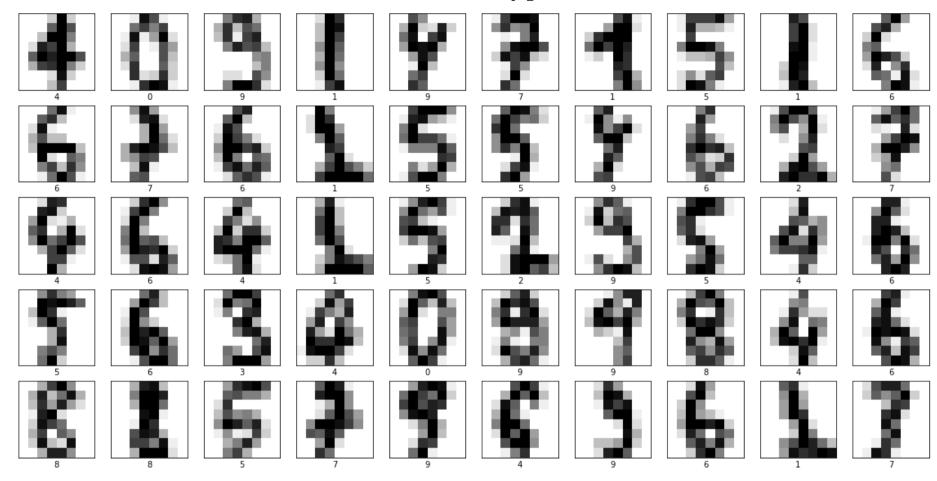
			g	
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[CV] ENDkmeansn_6				0.3s
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[CV] ENDkmeansn_6	clusters=65;	total	time=	0.3s
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[CV] ENDkmeansn_6				0.3s
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[CV] ENDkmeansn_6	clusters=70;	total	time=	0.3s
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[CV] ENDkmeansn_6				0.3s
[CV] ENDkmeansn_6				0.4s
[CV] ENDkmeansn_6				0.4s
[CV] ENDkmeansn_6	•			0.4s
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[CV] ENDkmeansn_6	clusters=72;	total	time=	0.4s
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[CV] ENDkmeansn_	clusters=73;	total	time=	0.5s
[CV] ENDkmeansn_6				0.5s
[CV] ENDkmeansn_6				0.5s
$ [CV] \ END \ \dots$				0.5s
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		g	_
[CV] ENDkmeansn_clusters=75	; total	time=	0.4s
[CV] ENDkmeansn_clusters=75	; total	time=	0.4s
[CV] ENDkmeansn_clusters=75	; total	time=	0.6s
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[CV] ENDkmeansn_clusters=76	; total	time=	0.5s
[CV] ENDkmeansn_clusters=76	; total	time=	0.5s
[CV] ENDkmeansn_clusters=77	; total	time=	0.5s
[CV] ENDkmeansn_clusters=77	; total	time=	0.6s
[CV] ENDkmeansn_clusters=77	; total	time=	1.1s
[CV] ENDkmeansn_clusters=78	; total	time=	0.9s
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[CV] ENDkmeansn_clusters=78	; total	time=	0.5s
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[CV] ENDkmeansn_clusters=79			0.4s
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[CV] ENDkmeansn_clusters=80			0.4s
[CV] ENDkmeansn_clusters=80			0.4s
[CV] ENDkmeansn_clusters=81	-		0.4s
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[CV] ENDkmeansn_clusters=83			0.4s
[CV] ENDkmeansn_clusters=83			0.4s
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[CV] ENDkmeansn_clusters=88			0.4s
[CV] ENDkmeansn_clusters=89			0.4s
[CV] ENDkmeansn_clusters=89	; total	time=	0.4s

```
[CV] END .....kmeans n clusters=89; total time=
                                                            0.4s
      [CV] END ......kmeans n clusters=90; total time=
                                                            0.4s
      [CV] END ......kmeans n clusters=90; total time=
                                                            0.4s
      [CV] END ......kmeans n clusters=90; total time=
                                                            0.4s
      [CV] END ......kmeans _n_clusters=91; total time=
                                                            0.4s
      [CV] END ......kmeans n clusters=91; total time=
                                                            0.45
      [CV] END ......kmeans n clusters=91; total time=
                                                            0.45
      [CV] END ......kmeans n clusters=92; total time=
                                                            0.45
      [CV] END ......kmeans n clusters=92; total time=
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      [CV] END ......kmeans n clusters=92; total time=
                                                            0.4s
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                                                            0.4s
      [CV] END .....kmeans n clusters=93; total time=
                                                            0.45
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      [CV] END ......kmeans n clusters=94; total time=
                                                            0.4s
      [CV] END ......kmeans n clusters=94; total time=
                                                            0.4s
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                                                            0.5s
      [CV] END .....kmeans n clusters=95; total time=
                                                            0.4s
      [CV] END ......kmeans n clusters=95; total time=
                                                            0.4s
      [CV] END ......kmeans n clusters=95; total time=
                                                            0.5s
      [CV] END ......kmeans n clusters=96; total time=
                                                            0.7s
      [CV] END ......kmeans n clusters=96; total time=
                                                            0.8s
      [CV] END ......kmeans n clusters=96; total time=
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                                                            0.8s
      [CV] END ......kmeans n clusters=97; total time=
                                                            0.7s
      [CV] END .....kmeans n clusters=97; total time=
                                                            0.7s
      [CV] END ......kmeans n clusters=98; total time=
                                                            0.5s
      [CV] END ......kmeans n clusters=98; total time=
                                                            0.5s
      [CV] END ......kmeans__n_clusters=98; total time=
                                                            0.5s
      [CV] END ......kmeans n clusters=99; total time=
                                                            0.5s
      [CV] END ......kmeans n clusters=99; total time=
                                                            0.5s
      [CV] END .....kmeans n clusters=99; total time=
                                                            0.5s
           GridSearchCV
Out[49]:
       • estimator: Pipeline
            ▶ KMeans
        ▶ LogisticRegression
```

```
In [50]: grid_clf.best_params_
```

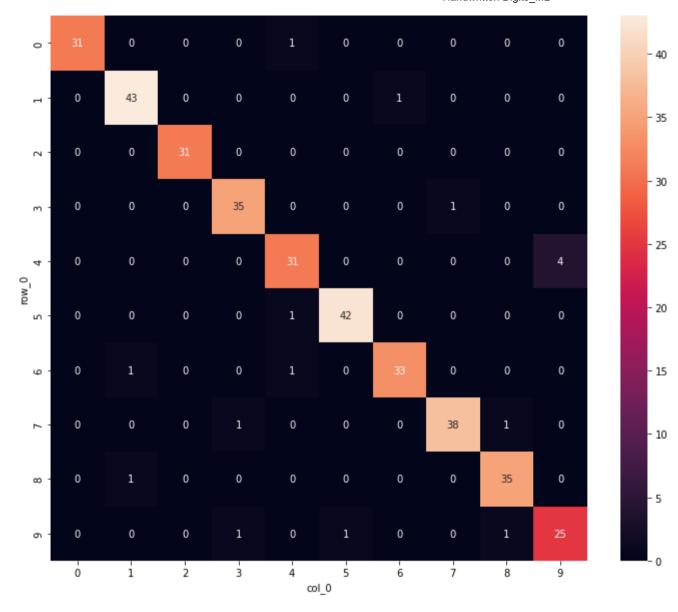
```
{'kmeans__n_clusters': 55}
         grid clf.score(X test, y test)
In [51]:
          0.95555555555556
Out[51]:
         y pred = grid clf.predict(X test)
In [52]:
In [53]: print(classification_report(y_test, y_pred))
                        precision
                                     recall f1-score
                                                        support
                     0
                             1.00
                                       0.97
                                                 0.98
                                                             32
                                                 0.97
                                                             44
                             0.96
                                       0.98
                     2
                             1.00
                                       1.00
                                                 1.00
                                                              31
                                                 0.96
                     3
                             0.95
                                       0.97
                                                              36
                                                 0.90
                     4
                             0.91
                                       0.89
                                                              35
                     5
                             0.98
                                       0.98
                                                 0.98
                                                              43
                             0.97
                                       0.94
                                                 0.96
                                                              35
                                       0.95
                                                 0.96
                     7
                             0.97
                                                              40
                     8
                             0.95
                                       0.97
                                                 0.96
                                                             36
                                                 0.88
                     9
                             0.86
                                       0.89
                                                              28
                                                 0.96
                                                             360
              accuracy
                             0.95
                                       0.95
                                                 0.95
                                                             360
             macro avg
         weighted avg
                             0.96
                                       0.96
                                                 0.96
                                                             360
         plt.figure(figsize=(20,10))
In [54]:
          for i in range(50):
              plt.subplot(5,10,i+1)
              plt.xticks([])
              plt.yticks([])
              plt.grid(True)
             x=X test[i].reshape(8,8)
              plt.imshow(x, cmap=plt.cm.binary)
              plt.xlabel(y_pred[i])
          plt.show()
```



In [55]: conf_matrix = pd.crosstab(y_test, y_pred)
 conf_matrix

```
Out[55]: col_0 0 1 2 3 4 5 6 7 8 9
      row_0
         0 31 0
                              0 0
                       0 0 0
             0 31 0
                0 35
                       0 0 1
                                0
                0 0 31
                       0
                         0 0
                0 0 1 42 0 0
             1 0 0 1 0 33
         7 0 0
                0
                       0 0 38 1 0
                    0
                       0 0 0 35 0
         9 0 0 0 1 0 1 0 0 1 25
```

```
In [56]: plt.figure(figsize=(12,10))
    sns.heatmap(conf_matrix, annot=True, fmt ='')
    plt.show()
```



In []: