Practical 5

Aim: To implement a Machine Learning Classification model using a K Nearest Neighbors Classifier algorithm and enhance the model by K Fold and GridSearchCV cross-validation.

```
In [66]: import numpy as np
    import pandas as pd
    from sklearn.neighbors import KNeighborsClassifier
    from sklearn.model_selection import train_test_split
    from sklearn.model_selection import train_test_split
    from sklearn.model_selection import train_test_split
    from sklearn.datasets import load_iris
    from sklearn.metrics import confusion_matrix,accuracy_score
    from sklearn.metrics import precision_recall_fscore_support
    from sklearn import metrics
    import matplotlib.pyplot as plt
    import seaborn as sns
    import plotly.express as px
```

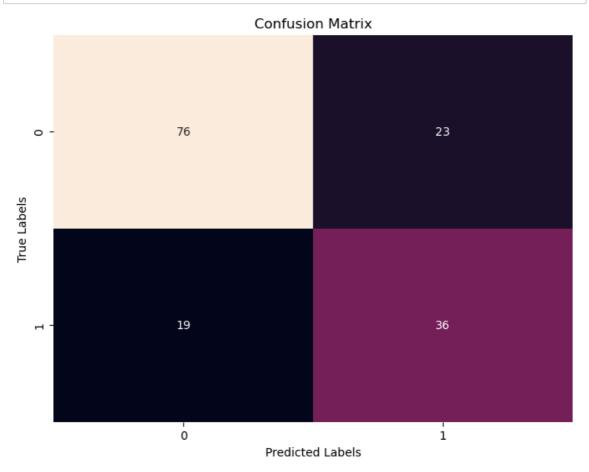
Loading dataset

Out[71]: KNeighborsClassifier(n_neighbors=7)

In a Jupyter environment, please rerun this cell to show the HTML representation or trust the notebook.

On GitHub, the HTML representation is unable to render, please try loading this page with nbviewer.org.

```
y_pred=knn.predict(X_test)
In [72]:
     print(knn.predict(X_test))
     1 1 0 0 0 0]
In [73]: from sklearn.metrics import confusion_matrix
     cm= confusion_matrix(y_test,y_pred)
     print(cm)
     [[76 23]
      [19 36]]
In [74]: import seaborn as sns
     plt.figure(figsize=(8, 6))
     sns.heatmap(cm, annot=True, fmt='d', cbar=False)
     plt.title('Confusion Matrix')
     plt.xlabel('Predicted Labels')
     plt.ylabel('True Labels')
     plt.show()
```



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In [76]: from sklearn.model_selection import KFold cv = KFold(n_splits=10) # perform cross-validation procedure for train_ix, test_ix in cv.split(X): # split data X_train, X_test = X[train_ix, :], X[test_ix, :] y_train, y_test = y[train_ix], y[test_ix] # fit and evaluate a model knn = KNeighborsClassifier(n_neighbors=7) knn.fit(X_train, y_train) y_pred=knn.predict(X_test) print(knn.predict(X_test)) from sklearn.metrics import confusion_matrix cm= confusion_matrix(y_test,y_pred) print(cm) from sklearn.metrics import precision_recall_fscore_support precision, recall, f1_score,_ = precision_recall_fscore_support(y_test, print("Precision:", precision) print("Recall:", recall) print("F1 Score:", f1_score)

```
0 0 0]
[[35 10]
[16 16]]
Precision: [0.68627451 0.61538462]
Recall: [0.7777778 0.5
F1 Score: [0.72916667 0.55172414]
0 1 1]
[[50 5]
[ 9 13]]
Precision: [0.84745763 0.72222222]
Recall: [0.90909091 0.59090909]
F1 Score: [0.87719298 0.65
                 ]
[1\ 1\ 0\ 0\ 0\ 1\ 0\ 0\ 0\ 0\ 0\ 1\ 0\ 0\ 0\ 0\ 0\ 1\ 1\ 1\ 0\ 0\ 0\ 0\ 0\ 1\ 1\ 0\ 0\ 1\ 0
1 0 0]
[[37 6]
[17 17]]
Precision: [0.68518519 0.73913043]
Recall: [0.86046512 0.5
F1 Score: [0.7628866 0.59649123]
[0\;0\;0\;0\;1\;1\;1\;1\;0\;0\;0\;0\;0\;1\;1\;1\;0\;0\;1\;0\;0\;0\;0\;0\;0\;1\;1\;1\;1\;0\;0\;1\;0\;1\;0
0 1 01
[[36 11]
[17 13]]
Precision: [0.67924528 0.54166667]
Recall: [0.76595745 0.43333333]
F1 Score: [0.72
            0.48148148]
0 0 0]
[[42 8]
[12 15]]
Precision: [0.7777778 0.65217391]
Recall: [0.84
           0.5555556]
F1 Score: [0.80769231 0.6
                 ]
[0\ 1\ 1\ 0\ 0\ 0\ 1\ 0\ 0\ 0\ 0\ 1\ 0\ 0\ 0\ 1\ 1\ 0\ 0\ 0\ 0\ 1\ 1\ 0\ 0\ 0\ 0\ 1\ 0
0 0 01
[[43 4]
[12 18]]
Precision: [0.78181818 0.81818182]
Recall: [0.91489362 0.6
F1 Score: [0.84313725 0.69230769]
0 0 01
[[53 10]
[ 8 6]]
Precision: [0.86885246 0.375
                  ]
Recall: [0.84126984 0.42857143]
F1 Score: [0.85483871 0.4
                 ]
0 1 0]
[[45 7]
[ 6 19]]
```

```
Precision: [0.88235294 0.73076923]
Recall: [0.86538462 0.76
F1 Score: [0.87378641 0.74509804]
[0\;0\;0\;1\;0\;0\;1\;0\;0\;0\;0\;1\;0\;0\;0\;0\;0\;1\;0\;0\;0\;0\;0\;0\;0\;1\;1\;1\;1\;0\;0\;0\;0
0 1]
[[43 9]
[10 14]]
Precision: [0.81132075 0.60869565]
Recall: [0.82692308 0.58333333]
F1 Score: [0.81904762 0.59574468]
1 0]
[[38 8]
[12 18]]
Precision: [0.76
               0.69230769]
Recall: [0.82608696 0.6
F1 Score: [0.79166667 0.64285714]
```

```
from sklearn.model_selection import KFold, GridSearchCV
In [77]:
         from sklearn.neighbors import KNeighborsClassifier
         from sklearn.metrics import confusion_matrix, precision_recall_fscore_support
         # Define the KFold cross-validation
         cv = KFold(n_splits=9)
         param_grid = {'n_neighbors': list(range(1, 51, 2))}
         # Initialize the KNN classifier
         knn = KNeighborsClassifier()
         # Initialize GridSearchCV
         grid_search = GridSearchCV(estimator=knn, param_grid=param_grid, cv=cv, sco
         # Perform cross-validation procedure with GridSearchCV
         for train_ix, test_ix in cv.split(X):
             # Split data
             X_train, X_test = X[train_ix, :], X[test_ix, :]
             y_train, y_test = y[train_ix], y[test_ix]
             # Fit model using GridSearchCV
             grid_search.fit(X_train, y_train)
             # Get the best KNN model found by GridSearchCV
             best_knn = grid_search.best_estimator_
             # Predict
             y_pred = best_knn.predict(X_test)
             # Evaluate
             cm = confusion_matrix(y_test, y_pred)
             precision, recall, f1_score, _ = precision_recall_fscore_support(y_test
             # Print results
             print("Confusion Matrix:")
             print(cm)
             print("Precision:", precision)
             print("Recall:", recall)
             print("F1 Score:", f1_score)
             # You can also access the best hyperparameters found
             print("Best parameters found by GridSearchCV:", grid_search.best_params
```

```
Confusion Matrix:
[[45 7]
[16 18]]
Precision: [0.73770492 0.72
Recall: [0.86538462 0.52941176]
F1 Score: [0.79646018 0.61016949]
Best parameters found by GridSearchCV: {'n_neighbors': 13}
Confusion Matrix:
[[49 10]
[14 13]]
Precision: [0.77777778 0.56521739]
Recall: [0.83050847 0.48148148]
F1 Score: [0.80327869 0.52
Best parameters found by GridSearchCV: {'n_neighbors': 21}
Confusion Matrix:
[[39 10]
[20 17]]
Precision: [0.66101695 0.62962963]
Recall: [0.79591837 0.45945946]
F1 Score: [0.72222222 0.53125
Best parameters found by GridSearchCV: {'n_neighbors': 11}
Confusion Matrix:
[[38 12]
[19 16]]
Precision: [0.66666667 0.57142857]
Recall: [0.76
                    0.45714286]
F1 Score: [0.71028037 0.50793651]
Best parameters found by GridSearchCV: {'n_neighbors': 13}
Confusion Matrix:
[[51 2]
[15 17]]
Precision: [0.77272727 0.89473684]
Recall: [0.96226415 0.53125
F1 Score: [0.85714286 0.66666667]
Best parameters found by GridSearchCV: {'n_neighbors': 25}
Confusion Matrix:
[[56 9]
[12 8]]
Precision: [0.82352941 0.47058824]
Recall: [0.86153846 0.4
F1 Score: [0.84210526 0.43243243]
Best parameters found by GridSearchCV: {'n_neighbors': 17}
Confusion Matrix:
[[56 7]
[ 9 13]]
Precision: [0.86153846 0.65
Recall: [0.88888889 0.59090909]
                      0.61904762]
F1 Score: [0.875
Best parameters found by GridSearchCV: {'n neighbors': 13}
Confusion Matrix:
[[48 9]
[13 15]]
Precision: [0.78688525 0.625
Recall: [0.84210526 0.53571429]
F1 Score: [0.81355932 0.57692308]
Best parameters found by GridSearchCV: {'n neighbors': 11}
Confusion Matrix:
[[44 8]
[13 20]]
Precision: [0.77192982 0.71428571]
Recall: [0.84615385 0.60606061]
```

F1 Score: [0.80733945 0.6557377]

Best parameters found by GridSearchCV: {'n_neighbors': 15}