```
ROLL NO:- 58
PRACTICAL NO.:-
PRACTICAL NAME: -. IMPLEMENT SIMPLE KNN USING EUCLIDEAN DISTANCE IN PYTHON.
from pandas import DataFrame
from sklearn.datasets import load_iris
data b = load iris()
df= DataFrame(data_b.data, columns=data_b.feature_names)
df['target'] = data_b.target
#print(df)
#print(data_b.DESCR)
print("Dataset Labels=",data_b.target_names)
from sklearn.neighbors import KNeighborsClassifier
from sklearn import metrics
from sklearn.metrics import confusion_matrix
from sklearn.model_selection import train_test_split
X_train, X_test, Y_train, y_test = train_test_split(df[data_b.feature_names], df['target'],
random_state=1)
print(X_train.head(6))
print(Y_train.head(6))
print(X_test.head())
clf = KNeighborsClassifier(n_neighbors=6)
clf.fit(X_train, Y_train) # model is trained
y_pred=clf.predict(X_test)
#print(y_test, y_pred)
print("Accuracy:",metrics.accuracy_score(y_test, y_pred))
cm = confusion_matrix(y_test, y_pred)
print("Confusion Matrix:")
print(cm)
OUTPUT:
Dataset Labels= ['setosa' 'versicolor' 'virginica']
  sepal length (cm) sepal width (cm) petal length (cm) petal width (cm)
54
         6.5
                 2.8
                           4.6
                                   1.5
108
         6.7
                  2.5
                           5.8
                                   1.8
112
        6.8
                 3.0
                          5.5
                                   2.1
17
                3.5
        5.1
                         1.4
                                  0.3
119
         6.0
                 2.2
                          5.0
                                   1.5
                  2.9
103
         6.3
                           5.6
                                   1.8
54 1
108 2
112 2
17 0
```

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103 2

Name: target, dtype: int32

sepal length (cm) sepal width (cm) petal length (cm) petal width (cm)

14	5.8	4.0	1.2	0.2
98	5.1	2.5	3.0	1.1
75	6.6	3.0	4.4	1.4
16	5.4	3.9	1.3	0.4
131	7 9	3.8	64	2 N

Accuracy: 1.0

Confusion Matrix:

[[13 0 0]

[0160]

[0 0 9]]

CODE: FOR BREAST CANCER DATA SET

```
from pandas import DataFrame
# from sklearn.datasets import load_iris
from sklearn.datasets import load_breast_cancer
from sklearn.neighbors import KNeighborsClassifier
from sklearn import metrics
from sklearn.metrics import confusion_matrix
data_b = load_breast_cancer()
df= DataFrame(data_b.data, columns=data_b.feature_names)
df['target'] = data_b.target
#print(df)
#print(data_b.DESCR)
print("Dataset Labels=",data_b.target_names)
from sklearn.model_selection import train_test_split
X_train, X_test, Y_train, y_test = train_test_split(df[data_b.feature_names], df['target'],
random_state=1)
print(X_train.head(6))
print(Y_train.head(6))
print(X_test.head())
clf = KNeighborsClassifier(n_neighbors=6)
clf.fit(X_train, Y_train) # model is trained
y_pred=clf.predict(X_test)
#print(y_test, y_pred)
print("Accuracy:",metrics.accuracy_score(y_test, y_pred))
cm = confusion_matrix(y_test, y_pred)
print("Confusion Matrix:")
print(cm)
OUTPUT:
Dataset Labels= ['malignant' 'benign']
 mean radius mean texture ... worst symmetry worst fractal dimension
562
     15.22
              30.62 ...
                          0.4089
                                       0.14090
              19.10 ...
                        0.2962
291
    14.96
                                      0.08472
16
     14.68
             20.13 ...
                        0.3029
                                      0.08216
546 10.32
             16.35 ...
                         0.2681
                                      0.07399
293 11.85
              17.46 ...
                         0.3101
                                      0.07007
350 11.66
              17.07 ...
                         0.2731
                                      0.06825
[6 rows x 30 columns]
562 0
291 1
16 0
546 1
```

350 1

Name: target, dtype: int32

mean radius mean texture ... worst symmetry worst fractal dimension

421	14.69	13.98	0.2827	0.09208
47	13.17	18.66	0.3900	0.11790
292	12.95	16.02	0.3380	0.09584
186	18.31	18.58	0.3206	0.06938
414	15.13	29.81	0.3233	0.06165

[5 rows x 30 columns]

Accuracy: 0.9370629370629371

Confusion Matrix:

[[51 4]

[583]]