Artificial Intelligence and Data Science Department

Machine Learning Lab/Even Sem 2023-23/Experiment 1b

Grade:

Class/Roll No.:

Name:

Description / Theory:

Dyotak Kachare	D11AD/26	
Title of Experiment: Introduction to scikit learn, matp	olotlib, seaborn library	
Objective of Experiment: To introduce platforms such as A	Anaconda, COLAB suitable to M	achine learning.
Outcome of Experiment :		
Implement various Machine lear	ning models	
Problem Statement: Introduction to scikit learn, matp	olotlib, seaborn library	



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	Introduction to suited learn / Matplot lib seaborn library	/
	Theory.	
	Mat platlih	1
	Mod platlib Matplotlib is a comprehensive library	
-	for creating static, animated, and	
1	interactive visualizations in Pythan Matplot lib makes things eary	
1	masplot lib makes things eary	-
1	and hard thing possible	
	Plot types offered by matplotlis	
1	1) Pairwise data	
1	2) statistical distribution	
+	3) yridded dala	
+	3) Gridded dala 4) 30 and valumetric data.	4
+	Seaborn	-
	seaborn is a python data visuolizat	ian
1.	library based on metplotlib.	
1	It provides a high level interface	100
1	It provides a high level interface drawing attractive and informative	0~
-	statistical graphics.	-

1 Introduction to Scikit Learn

```
[1]: # Importing required libraries
     import pandas as pd
     import numpy as np
     import matplotlib.pyplot as plt
     import sklearn
     import seaborn as sns
[2]: # Reading data
     heart_disease = pd.read_csv('.../data/cleaned/heart.csv')
     heart_disease.head()
       age sex cp trestbps chol fbs restecg thalach exang oldpeak [2]
[2]:
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[3]: # Create X (all the feature columns)
     X = heart_disease.drop("target", axis=1)
     # Create y (the target column)
     y = heart_disease["target"]
     # Check the head of the features DataFrame
    X.head()
[3]:
       age sex cp trestbps chol fbs restecg thalach exang oldpeak 2

slope \
```

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     4
         3
               2
[4]: # Check the head and the value counts of the labels
     y.head(), y.value_counts()
[4]: (0
           0
           0
      1
      2
           0
      3
           0
      Name: target, dtype: int64,
      target
           526
           499
      Name: count, dtype: int64)
[5]: # Split the data into training and test sets
     from sklearn.model_selection import train_test_split
     X_train, X_test, y_train, y_test = train_test_split(X,
                                                           у,
                                                           test size=0.25) # by₽
      →default train_test_split uses 25% of the data for the test set
     X_train.shape, X_test.shape, y_train.shape, y_test.shape
[5]: ((768, 13), (257, 13), (768,), (257,))
[6]: from sklearn.ensemble import RandomForestClassifier
     clf = RandomForestClassifier()
```

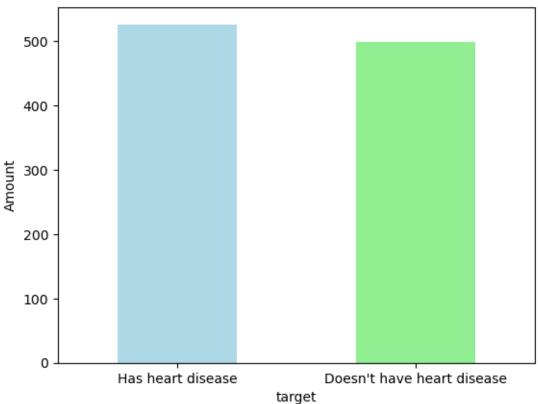
```
[7]: clf.get_params()
 [7]: {'bootstrap': True,
       'ccp alpha': 0.0,
       'class weight': None,
       'criterion': 'gini',
       'max depth': None,
       'max_features': 'sqrt',
       'max leaf nodes': None,
       'max_samples': None,
       'min_impurity_decrease': 0.0,
       'min samples leaf': 1,
       'min_samples_split': 2,
       'min weight fraction leaf': 0.0,
       'n_estimators': 100,
       'n_jobs': None,
       'oob_score': False,
       'random state': None,
       'verbose': 0,
       'warm start': False}
 [8]: clf.fit(X=X_train, y=y_train)
 [8]: RandomForestClassifier()
 [9]: y_preds = clf.predict(X=X_test)
      y_preds
 [9]: array([1, 0, 1, 1, 1, 0, 1, 1, 1, 1, 1, 1, 0, 0, 0, 1, 0, 1, 1, 1, 1,
             0, 0, 1, 0, 0, 0, 1, 1, 0, 1, 0, 0, 1, 1, 1, 1, 1, 1, 1, 0, 0, 0,
             1, 1, 1, 0, 1, 1, 0, 0, 1, 1, 1, 1, 1, 0, 1, 0, 1, 0, 1, 1, 0, 0,
             0, 1, 1, 1, 1, 0, 1, 0, 0, 1, 1, 0, 0, 0, 0, 0, 1, 1, 1, 0, 0, 0,
             1, 0, 1, 1, 1, 1, 0, 0, 0, 1, 0, 1, 1, 0, 0, 0, 1, 0, 0, 0, 1,
             0, 1, 0, 0, 1, 1, 1, 1, 0, 0, 0, 1, 0, 1, 1, 1, 1, 1, 1, 1, 0, 0,
             0, 1, 0, 1, 1, 0, 1, 0, 0, 0, 1, 0, 0, 1, 0, 0, 0, 1, 1, 0, 0, 0,
             1, 1, 1, 0, 0, 1, 1, 1, 0, 1, 0, 1, 1, 1, 0, 0, 0, 0, 0, 1, 0, 0,
             1, 0, 0, 0, 0, 1, 1, 0, 0, 0, 0, 0, 0, 1, 1, 0, 1, 1, 1, 0, 1, 1,
             1, 1, 1, 0, 1, 0, 0, 1, 0, 1, 1, 0, 1, 0, 1, 0, 1, 1, 0, 0, 1, 1,
             0, 1, 0, 0, 0, 1, 0, 1, 1, 1, 0, 0, 1, 1, 0, 0, 0, 0, 1, 1, 0, 1,
             1, 0, 0, 1, 1, 0, 1, 0, 1, 0, 0, 0, 0, 0], dtype=int64)
[10]: from sklearn.metrics import classification report, confusion matrix, [2]
       →accuracy_score
      # Create a classification report
      print(classification_report(y_test, y_preds))
```

precision recall f1-score support

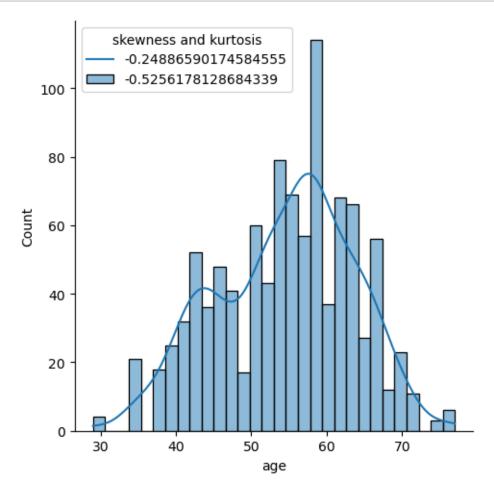
```
0
                    1.00
                              1.00
                                         1.00
                                                     124
           1
                    1.00
                              1.00
                                         1.00
                                                     133
                                         1.00
                                                     257
    accuracy
                                         1.00
                                                     257
   macro avg
                    1.00
                               1.00
weighted avg
                    1.00
                              1.00
                                         1.00
                                                     257
```

```
[11]: # Create a confusion matrix
conf_mat = confusion_matrix(y_test, y_preds)
conf_mat
```

Heart Disease values



```
[13]: #create a distribution plot with normal distribution curve
sns.displot( x = 'age', data = heart_disease, bins = 30, kde = True)
skewness=str(heart_disease["age"].skew())
kurtosis=str(heart_disease["age"].kurt())
plt.legend([skewness,kurtosis],title=("skewness and kurtosis"))
plt.show();
```



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Result and Discussion:

Thus we have successfully used the python libraries.