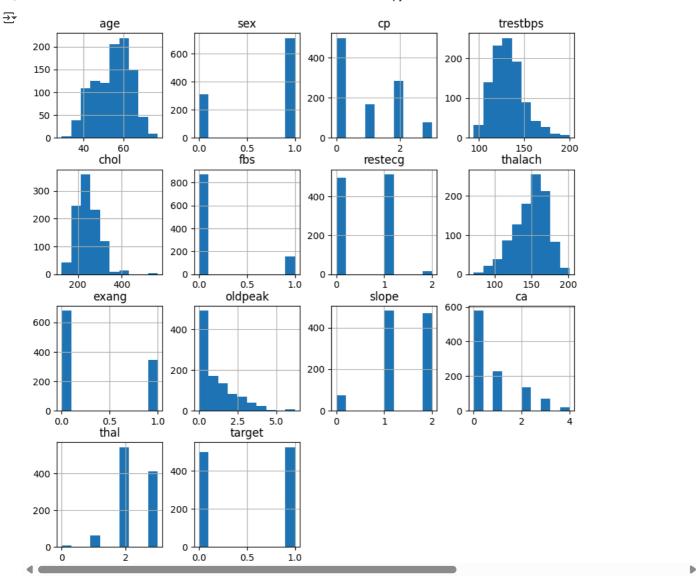
```
!pip install fuzzy_c_means
→ Collecting fuzzy_c_means
          Downloading fuzzy_c_means-1.7.0-py3-none-any.whl (9.0 kB)
       Requirement already satisfied: joblib<2.0.0,>=1.2.0 in /usr/local/lib/python3.10/dist-packages (from fuzzy_c_means) (1.3.2)
       Requirement already satisfied: numpy<2.0.0,>=1.21.1 in /usr/local/lib/python3.10/dist-packages (from fuzzy_c_means) (1.23.5)
       Collecting pydantic<2.0.0,>=1.9.0 (from fuzzy_c_means)
          Downloading pydantic-1.10.14-cp310-cp310-manylinux_2_17_x86_64.manylinux2014_x86_64.whl (3.1 MB)
                                                                        - 3.1/3.1 MB 13.7 MB/s eta 0:00:00
       Collecting tabulate<0.9.0,>=0.8.9 (from fuzzy_c_means)
          Downloading tabulate-0.8.10-py3-none-any.whl (29 kB)
       Requirement already satisfied: tqdm<5.0.0,>=4.64.1 in /usr/local/lib/python3.10/dist-packages (from fuzzy_c_means) (4.66.1)
       Collecting typer<0.5.0,>=0.4.0 (from fuzzy_c_means)
          Downloading typer-0.4.2-py3-none-any.whl (27 kB)
       Requirement already satisfied: typing-extensions>=4.2.0 in /usr/local/lib/python3.10/dist-packages (from pydantic<2.0.0,>=1.9.0->fuz
       Requirement already satisfied: click9.0.0,>=7.1.1 in /usr/local/lib/python3.10/dist-packages (from typer<0.5.0,>=0.4.0->fuzzy_c_mearing (from typer<0.5.0,>=0.4.0->fuzzy_c_mearing (from typer<0.5.0)
       Installing collected packages: typer, tabulate, pydantic, fuzzy\_c\_means
          Attempting uninstall: typer
             Found existing installation: typer 0.9.0
             Uninstalling typer-0.9.0:
                Successfully uninstalled typer-0.9.0
          Attempting uninstall: tabulate
             Found existing installation: tabulate 0.9.0
             Uninstalling tabulate-0.9.0:
               Successfully uninstalled tabulate-0.9.0
          Attempting uninstall: pydantic
             Found existing installation: pydantic 2.6.1
             Uninstalling pydantic-2.6.1:
               Successfully uninstalled pydantic-2.6.1
       ERROR: pip's dependency resolver does not currently take into account all the packages that are installed. This behaviour is the sou
       lida 0.0.10 requires fastapi, which is not installed.
       lida 0.0.10 requires kaleido, which is not installed.
       lida 0.0.10 requires python-multipart, which is not installed.
       lida 0.0.10 requires uvicorn, which is not installed.
       llmx 0.0.15a0 requires cohere, which is not installed.
       llmx 0.0.15a0 requires openai, which is not installed.
       llmx 0.0.15a0 requires tiktoken, which is not installed.
       bigframes 0.20.1 requires tabulate>=0.9, but you have tabulate 0.8.10 which is incompatible.
       Successfully installed fuzzy_c_means-1.7.0 pydantic-1.10.14 tabulate-0.8.10 typer-0.4.2
!pip install scipy
       Requirement already satisfied: scipy in /usr/local/lib/python3.10/dist-packages (1.11.4)
       Requirement already satisfied: numpy<1.28.0,>=1.21.6 in /usr/local/lib/python3.10/dist-packages (from scipy) (1.23.5)
!pip install -U scikit-learn
Requirement already satisfied: scikit-learn in /usr/local/lib/python3.10/dist-packages (1.2.2)
       Collecting scikit-learn
          Downloading \ scikit\_learn-1.4.0-1-cp310-cp310-manylinux\_2\_17\_x86\_64.manylinux\\ 2014\_x86\_64.whl \ (12.1 \ MB) - 
                                                                         12.1/12.1 MB 41.1 MB/s eta 0:00:00
       Requirement already satisfied: numpy<2.0,>=1.19.5 in /usr/local/lib/python3.10/dist-packages (from scikit-learn) (1.23.5)
       Requirement already satisfied: scipy>=1.6.0 in /usr/local/lib/python3.10/dist-packages (from scikit-learn) (1.11.4)
       Requirement already satisfied: joblib>=1.2.0 in /usr/local/lib/python3.10/dist-packages (from scikit-learn) (1.3.2)
       Requirement already satisfied: threadpoolctl>=2.0.0 in /usr/local/lib/python3.10/dist-packages (from scikit-learn) (3.2.0)
       Installing collected packages: scikit-learn
          Attempting uninstall: scikit-learn
             Found existing installation: scikit-learn 1.2.2
             Uninstalling scikit-learn-1.2.2:
               Successfully uninstalled scikit-learn-1.2.2
       ERROR: pip's dependency resolver does not currently take into account all the packages that are installed. This behaviour is the sou
       bigframes 0.20.1 requires tabulate>=0.9, but you have tabulate 0.8.10 which is incompatible.
       Successfully installed scikit-learn-1.4.0
import pandas as pd
import numpy as np
import sklearn as sk
from sklearn.tree import DecisionTreeClassifier
from sklearn.model selection import cross val score
from sklearn.preprocessing import StandardScaler
from sklearn.datasets import make classification
from sklearn.ensemble import BaggingClassifier
from sklearn.ensemble import GradientBoostingClassifier
from fcmeans import FCM
from sklearn.model_selection import train_test_split
from sklearn import metrics
from sklearn.metrics import confusion_matrix
from sklearn.metrics import roc_curve, auc
from sklearn.metrics import mean_squared_error
from sklearn.metrics import mean_absolute_error
import matplotlib.pyplot as plt
import seaborn as sns
```

```
from \ sklearn.metrics \ import \ mean\_squared\_error \ , \ precision\_score, accuracy\_score, f1\_score, recall\_score
from six import StringIO
from IPython.display import Image
from sklearn.tree import export_graphviz
import pydotplus
# load your data
data = pd.read_csv("heart.csv")
data.head(8)
             sex cp trestbps
                                 chol fbs restecg thalach exang oldpeak slope ca thal target
      n
          52
                    0
                            125
                                  212
                                          0
                                                           168
                                                                    0
                                                                            1.0
                                                                                     2
                                                                                        2
                                                                                               3
                                                                                                       0
                1
          53
                    0
                            140
                                   203
                                                   0
                                                           155
                                                                            3.1
                                                                                    0
                                                                                        0
                                                                                               3
                                                                                                       0
      1
                1
                                          1
                                                                    1
      2
          70
                1
                    0
                            145
                                   174
                                          0
                                                           125
                                                                    1
                                                                            2.6
                                                                                    0
                                                                                        0
                                                                                               3
                                                                                                       0
      3
          61
                1
                    0
                            148
                                  203
                                          0
                                                           161
                                                                    0
                                                                            0.0
                                                                                    2
                                                                                               3
                                                                                                       0
                                                                                        1
      4
                0
                            138
                                   294
                                                           106
                                                                    0
                                                                            1.9
                                                                                               2
                                                                                                       0
                                          1
      5
          58
                0
                    0
                            100
                                  248
                                          0
                                                   0
                                                           122
                                                                    0
                                                                            10
                                                                                     1
                                                                                        0
                                                                                               2
                                                                                                       1
      6
          58
                    0
                            114
                                   318
                                          0
                                                   2
                                                           140
                                                                    0
                                                                            4.4
                                                                                    0
                                                                                        3
                                                                                               1
                                                                                                       0
                1
      7
                                                   O
          55
                1
                    0
                            160
                                   289
                                          0
                                                           145
                                                                    1
                                                                            0.8
                                                                                     1
                                                                                        1
                                                                                               3
                                                                                                       0
# Preprocess the data
data.replace('?',-99999, inplace=True)
print(data.axes)
[RangeIndex(start=0, stop=1025, step=1), Index(['age', 'sex', 'cp', 'trestbps', 'chol', 'fbs', 'restecg', 'thalach',
             'exang', 'oldpeak', 'slope', 'ca', 'thal', 'target'],
           dtype='object')]
data.isnull().sum()
\overline{\mathbf{x}}
    age
                  a
     sex
                  0
     ср
     trestbps
                  0
     chol
                  0
     fbs
                  0
     restecg
                  0
     thalach
                  0
     exang
                  a
     oldpeak
                  0
     slope
                  0
                  0
     thal
                  0
     target
     dtype: int64
# Let explore the dataset and do a few visualizations
print(data.loc[10])
₹
    age
                   71.0
     sex
                    0.0
                    0.0
     ср
     trestbps
                  112.0
     chol
                  149.0
     fbs
                    0.0
     restecg
                    1.0
     thalach
                  125.0
     exang
                    0.0
     oldpeak
                    1.6
     slope
                    1.0
                    0.0
     thal
                    2.0
                    1.0
     target
     Name: 10, dtype: float64
data['target'].unique()
\rightarrow array([0, 1])
# Plot histograms for each variable
data.hist(figsize = (10, 10))
plt.show()
```



Describe the dataset
print(data.describe())
print(data.info())

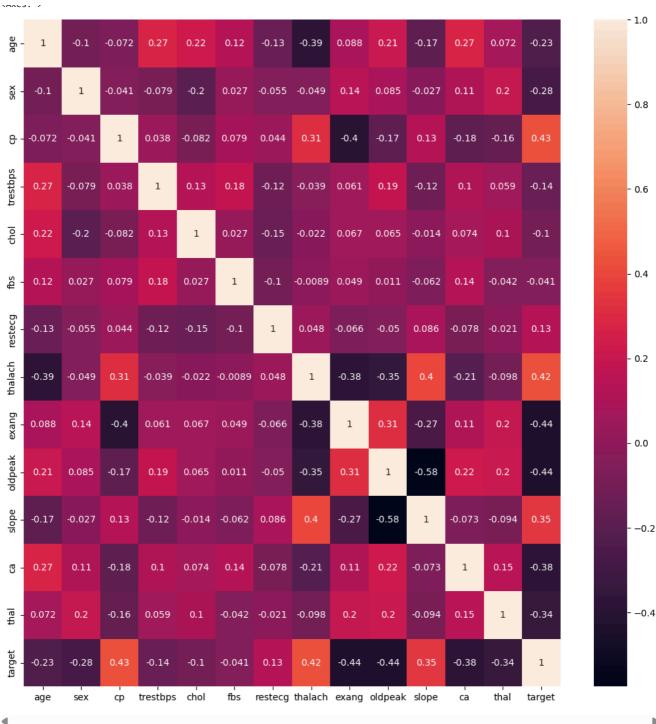
₹		age	sex	ср	trestbps	chol	\	
	count	1025.000000	1025.000000	1025.000000	1025.000000	1025.00000		
	mean	54.434146	0.695610	0.942439	131.611707	246.00000		
	std	9.072290	0.460373	1.029641	17.516718	51.59251		
	min	29.000000	0.000000	0.000000	94.000000	126.00000		
	25%	48.000000	0.000000	0.000000	120.000000	211.00000		
	50%	56.000000	1.000000	1.000000	130.000000	240.00000		
	75%	61.000000	1.000000	2.000000	140.000000	275.00000		
	max	77.000000	1.000000	3.000000	200.000000	564.00000		
		fbs	restecg	thalach	exang	oldpeak	\	
	count	1025.000000	1025.000000	1025.000000	1025.000000	1025.000000		
	mean	0.149268	0.529756	149.114146	0.336585	1.071512		
	std	0.356527	0.527878	23.005724	0.472772	1.175053		
	min	0.000000	0.000000	71.000000	0.000000	0.000000		
	25%	0.000000	0.000000	132.000000	0.000000	0.000000		
	50%	0.000000	1.000000	152.000000	0.000000	0.800000		
	75%	0.000000	1.000000	166.000000	1.000000	1.800000		
	max	1.000000	2.000000	202.000000	1.000000	6.200000		
		slope	ca	thal	target			
	count	1025.000000	1025.000000	1025.000000	1025.000000			
	mean	1.385366	0.754146	2.323902	0.513171			
	std	0.617755	1.030798	0.620660	0.500070			
	min	0.000000	0.000000	0.000000	0.000000			
	25%	1.000000	0.000000	2.000000	0.000000			
	50%	1.000000	0.000000	2.000000	1.000000			
	75%	2.000000	1.000000	3.000000	1.000000			
	max	2.000000	4.000000	3.000000	1.000000			
	<clas< td=""><td>s 'pandas.core</td><td>.frame.DataFr</td><td>ame'></td><td></td><td></td><td></td></clas<>	s 'pandas.core	.frame.DataFr	ame'>				
	Range	RangeIndex: 1025 entries, 0 to 1024						
	Data columns (total 14 columns):							
	#	Column Non-	Null Count D	type				
	0	age 1025	non-null i	.nt64				
		-						

```
sex
              1025 non-null
                               int64
2
    ср
              1025 non-null
                               int64
3
    trestbps
              1025 non-null
                               int64
              1025 non-null
    chol
                               int64
    fbs
               1025 non-null
                               int64
    restecg
              1025 non-null
                               int64
              1025 non-null
                               int64
    thalach
8
              1025 non-null
                               int64
    exang
                               float64
    oldpeak
              1025 non-null
10
   slope
              1025 non-null
                               int64
11
   ca
              1025 non-null
                               int64
12
   thal
              1025 non-null
                               int64
13
   target
              1025 non-null
                               int64
```

dtypes: float64(1), int64(13) memory usage: 112.2 KB

#plt.figure(figsize=(100,81)) #sns.pairplot(data,palette="Set2",kind='hist') plt.figure(figsize=(13,13)) sns.heatmap(data.corr(),annot=True)

→ \makes. /



X = data.drop("target", axis=1)

y = data["target"]

```
# split data
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.33)
sc = StandardScaler()
sc.fit(X_train)
X_{train_std} = sc.transform(X_{train})
X_test_std = sc.transform(X_test)
#Decision Tree Classifier
DT_model = DecisionTreeClassifier(max_depth=3)
DT_model.fit(X_train_std,y_train)
DecisionTreeClassifier ① ?
     DecisionTreeClassifier(max depth=3)
regressor = GradientBoostingClassifier(
    max_depth=3,
    n_estimators=3,
    learning_rate=1.0
regressor.fit(X_train, y_train)
errors = [mean_squared_error(y_test,y_pred) for y_pred in regressor.staged_predict(X_test)]
best_n_estimators = np.argmin(errors)
GBDT_best_regressor = GradientBoostingClassifier(
   max depth = 3,
    n_estimators=best_n_estimators,
   learning_rate = 1.0
GBDT_best_regressor.fit(X_train,y_train)
<del>____</del>
                      GradientBoostingClassifier
     GradientBoostingClassifier(learning_rate=1.0, n_estimators=2)
# Bagging
gbdt = GradientBoostingClassifier()
Bag_GBDT = BaggingClassifier(base_estimator=gbdt, n_estimators=5, max_samples=0.8)
Bag_GBDT.fit(X_train, y_train)
→
     TypeError
                                               Traceback (most recent call last)
     <ipython-input-20-58b0080d037d> in <cell line: 3>()
           1 # Bagging
          2 gbdt = GradientBoostingClassifier()
     ----> 3 Bag_GBDT = BaggingClassifier(base_estimator=gbdt, n_estimators=5, max_samples=0.8)
           4 Bag_GBDT.fit(X_train, y_train)
     TypeError: BaggingClassifier.__init__() got an unexpected keyword argument 'base_estimator'
     4
df tk=data
def lower_limit(data):
 col = data.columns
 data_lower_limit = []
 for i in col:
   df = data[i]
    x = (1*df[0] + 1*df[1] + 2*df[2] + 3*df[3] + 4*df[4]) / sum(df)
   data_lower_limit.append(x)
 return(data_lower_limit)
def middle_value(data):
 col = data.columns
 data_middle_value = []
 for i in col:
   df = data[i]
    x = (1*df[0] + 2*df[1] + 3*df[2] + 4*df[3] + 5*df[4]) / sum(df)
    data_middle_value.append(x)
 return(data_middle_value)
def upper_limit(data):
 col = data.columns
 data_upper_limit = []
 for i in col:
   df = data[i]
    x = (2*df[0] + 3*df[1] + 4*df[2] + 5*df[3] + 5*df[4]) / sum(df)
```

```
data_upper_limit.append(x)
 return(data_upper_limit)
def fuzzi_work(data):
 clower_limit = lower_limit(data)
 dmiddle_value = middle_value(data)
 eupper_limit = upper_limit(data)
 df = pd.DataFrame(list(zip(clower_limit, dmiddle_value, eupper_limit)),columns =['batas bawah', 'nilai tengah', 'batas atas'])
fuzz_tk = fuzzi_work(df_tk)
regressor = GradientBoostingClassifier(
   max depth=2,
    n_estimators=3,
   learning_rate=1.0
regressor.fit(X_train, y_train)
errors = [mean_squared_error(y_test,y_pred) for y_pred in regressor.staged_predict(X_test)]
best_n_estimators = np.argmin(errors)
Fuzzy_best_regressor = GradientBoostingClassifier(
   \max depth = 2,
    n_{estimators=best_n_{estimators}}
   learning_rate = 0.5
Fuzzy_best_regressor.fit(X_train,y_train)
<del>_</del>
                                                                            (i) (?)
                             GradientBoostingClassifier
     GradientBoostingClassifier(learning_rate=0.5, max_depth=2, n_estimators=2)
regressor = GradientBoostingClassifier(
   max_depth=5,
    n_estimators=2,
   learning_rate=1.0
regressor.fit(X_train, y_train)
errors = [mean_squared_error(y_test,y_pred) for y_pred in regressor.staged_predict(X_test)]
best_n_estimators = np.argmin(errors)
best_regressor = GradientBoostingClassifier(
   max_depth =5,
    n_estimators=best_n_estimators,
   learning_rate = 1.0
best regressor.fit(X train,y train)
Bag_Fuzzy_GBDT = BaggingClassifier(
   base_estimator= best_regressor,
    n_estimators=100,
   max_samples = 0.8
   bootstrap=True,
    oob_score = True,
   random state = 0
Bag_Fuzzy_GBDT.fit(X_train,y_train)
     TypeError
                                               Traceback (most recent call last)
     <ipython-input-24-5956afa5d073> in <cell line: 18>()
         16 best_regressor.fit(X_train,y_train)
          17
     ---> 18 Bag_Fuzzy_GBDT = BaggingClassifier(
                 base_estimator= best_regressor,
         19
          20
                 n estimators=100,
     TypeError: BaggingClassifier.__init__() got an unexpected keyword argument 'base_estimator'
#Fuzzy-C-Means Clustering
fcm = FCM(n_clusters=8)
fcm.fit(X_train_std)
FX_train = fcm.predict(X_train_std)
FX_test = fcm.predict(X_test_std)
FX_train= FX_train.reshape(-1, 1)
FX test = FX test.reshape(-1, 1)
```

```
# Bagging
gbdt = GradientBoostingClassifier()
bag_fcm_gbdt = BaggingClassifier(base_estimator=gbdt, n_estimators=10, max_samples=0.8)
bag_fcm_gbdt.fit(FX_train, y_train)
🧺 /usr/local/lib/python3.10/dist-packages/sklearn/ensemble/_base.py:156: FutureWarning: `base_estimator` was renamed to `estimator` ir
   warnings.warn(
           BaggingClassifier
   ▶ base_estimator: GradientBoostingClassifier
        ▶ GradientBoostingClassifier
   4
# Predict
DT_y_pred = DT_model.predict(X_test_std)
print(DT y pred)
→ [0 0 0 ... 0 0 0]
GBDT_y_pred = GBDT_best_regressor.predict(X_test)
print(GBDT_y_pred)
0 0 1 0 1 0 1 0 1 0 0 1 1 0 1 0 1 1 0 0 1 1 1 0 1 0 0 0 0 0 1 1 1 0 0 1 0
   1110000111110001111000110010100111101
   00111000110010011100101111110111010
   1011000111111110101011101100000110011
   1 1 0 1 0 11
# Predict
Bag GBDT y pred = Bag GBDT.predict(X test)
print(Bag_GBDT_y_pred)
  [0 0 1 0 1 0 0 1 0 0 1 0 0 1 1 1 0 1 0 0 1 0 0 1 0 0 1 0 1 1 1 1 1 0 0 1 0 1
   0111110100011000101111100001110111011
   0 1 1 0 1 0 1 0 1 0 0 0 1 0 1 0 1 0 1 1 0 0 1 1 0 0 1 0 0 0 0 0 1 0 1 1 0 1 0
   101010111001110011111001111111110000001
   1001010101010000010111101000110000110100
   1 1 0 1 0 1]
Fuzzy_GBDT_y_pred = Fuzzy_best_regressor.predict(X_test)
print(Fuzzy_GBDT_y_pred)
F [0 0 1 1 1 0 0 0 0 0 1 0 0 0 1 0 1 0 0 0 1 0 0 1 0 0 1 0 0 1 0 0 1 0 0 1 0 0 0 0 0 0 0 0 1
   01011101000110001111111000010101111001
   0\;0\;1\;0\;1\;1\;1\;0\;1\;1\;1\;0\;1\;0\;0\;0\;1\;1\;0\;1\;0\;1\;1\;1\;1\;0\;1\;1\;0\;0\;0\;0
   1000100010001100110010101011110000001
   1 1 0 0 0 1]
Bag_Fuzzy_GBDT_y_pred = Bag_Fuzzy_GBDT.predict(X_test)
print(Bag_Fuzzy_GBDT_y_pred)
1 1 0 1 0 1 1 1 1 1 1 1 0 1 0 1 0 0 0 1 1 1 0 1 0 0 0 1 1 1 1 0 0 0 0 0 1 1 1
   1 1 1 1 0 0 0 0 0 1 1 1 1 1 0 0 0 0 1 1 1 1 1 0 0 0 0 1 1 0 1 1 0 1 0 0 0 1 1 1 1 1 0 1
   1 0 0 1 0 1 0 1 1 1 0 0 0 0 0 1 0 1 1 1 1 0 1 0 0 1 1 0 0 0 0 0 1 0 1 0 0
   1 1 1 1 0 0 0 1 1 1 1 1 1 1 1 0 1 1 0 1 1 1 1 0 0 1 0 0 0 0 0 1 1 0 0 1 1
   1 1 0 1 0 11
bag_fcm_gbdt_y_pred = bag_fcm_gbdt.predict(FX_test)
print(bag_fcm_gbdt_y_pred)
  [0 0 1 1 1 0 0 0 0 0 1 0 0 1 1 1 0 1 0 0 1 0 0 1 0 0 1 0 1 1 1 1 0 1 1 1
   01111011001100111011111100001110111111
```

```
0 1 1 1 1 0 1 0 1 0 0 0 0 0 0 1 0 1 1 1 0 1 1 0 0 1 0 0 0 0 0 1 0 1 1 1 1 0
          1 1 0 1 0 1 1 1 1 1 0 1 0 1 1 1 1 0 0 0 0 1 1 0 1 1 0 0 1 1 1 1 1 0 1 0 0 1 0 1
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          101110011110110101010111100001110011
          1 1 0 0 0 11
i = 0
print ("\n-----")
print ('%-25s %-25s' % ('Original Label', 'Predicted Label', 'Correct/Wrong'))
print ("-----
for label in y_test:
      print ('%-25s %-25s' % (label, bag_fcm_gbdt_y_pred[i]), end="")
       if (label == bag_fcm_gbdt_y_pred[i]):
             print (' %-25s' % ('Correct'))
            print (' %-25s' % ('Wrong'))
      i = i + 1
print ("-----")
print("\nConfusion Matrix:\n",metrics.confusion_matrix(y_test, bag_fcm_gbdt_y_pred))
print ("-----")
\label{lem:print} print("\nClassification Report:\n",metrics.classification\_report(y\_test, bag\_fcm\_gbdt\_y\_pred))
print('Accuracy of the classifer is %0.2f' % metrics.accuracy_score(y_test, bag_fcm_gbdt_y_pred))
print ("----")
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        0
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        0
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        Confusion Matrix:
          [[114 40]
          [ 25 160]]
        Classification Report:
                                precision recall f1-score support
                                                   0.74
                                                                       0.78
                           0
                                        0.82
                                                                                           154
                          1
                                        0.80
                                                   0.86
                                                                      0.83
                                                                                          185
              accuracy
                                                                         0.81
                                                                                            339
                                        0.81
                                                        0.80
                                                                         0.80
                                                                                            339
             macro avg
                                       0.81
                                                        0.81
                                                                         0.81
                                                                                           339
        weighted avg
        Accuracy of the classifer is 0.81
```

```
i = 0
print ("\n----")
print ('%-25s %-25s' % ('Original Label', 'Predicted Label', 'Correct/Wrong'))
for label in y_test:
   print ('%-25s %-25s' % (label, DT_y_pred[i]), end="")
   if (label == DT_y_pred[i]):
      print (' %-25s' % ('Correct'))
      print (' %-25s' % ('Wrong'))
\label{lem:print("\nConfusion Matrix:\n",metrics.confusion\_matrix(y\_test, \ DT\_y\_pred))} \\
print ("-----
print("\nClassification Report:\n",metrics.classification\_report(y\_test, \ DT\_y\_pred))
print ("----")
print('Accuracy of the classifer is %0.2f' % metrics.accuracy_score(y_test,DT_y_pred))
print ("-----")
    ______
    Original Label Predicted Label Correct/Wrong
                                             Correct
    0
                        0
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                                              Correct
                                              Correct
print ('%-25s %-25s %-25s' % ('Original Label', 'Predicted Label', 'Correct/Wrong'))
print ("-----")
for label in y_test:
   print ('%-25s %-25s' % (label, GBDT_y_pred[i]), end="")
   if (label == GBDT_y_pred[i]):
      print (' %-25s' % ('Correct'))
```

```
print (' %-25s' % ('Wrong'))
    i = i + 1
print ("-----")
print("\nConfusion Matrix:\n",metrics.confusion_matrix(y_test, GBDT_y_pred))
print ("----")
print("\nClassification Report:\n",metrics.classification_report(y_test, GBDT_y_pred))
print ("-----")
print('Accuracy of the classifer is %0.2f' % metrics.accuracy_score(y_test, GBDT_y_pred))
print ("-----")
```

	_

Original Label	Predicted Label	Correct/Wrong
)	0	Correct
1	0	Wrong
9	1	Wrong
)	1	Wrong
L	1	Correct
)	0	Correct
)	0	Correct
	0	Wrong
	0	Wrong
)	1	Wrong
	1	Correct
)	0	Correct
1	0	Correct
	1	Correct
	1	Correct
	1	Correct
)	0	Correct
	1	Correct
)	0	Correct
)	0	Correct
	1	Correct
)	0	Correct
)	0	Correct
	1	Correct
)	0	Correct
)	0	Correct
	1	Correct
)	0	Correct
	1	Correct
)	1	Wrong
)	0	Correct
	1	Correct
)	1	Wrong
	1	Correct
)	0	Correct
	1	Correct
	0	Wrong
	1	Correct
	1	Correct
	1	Correct
)	0	Correct
	1	Correct
)	0	Correct
)	0	Correct
	1	Wrong
	1	Correct
•	1	Correct
)	0	Correct
)	0	Correct
)	0	Correct
	1	Correct

```
i = 0
print ("\n----")
print ('%-25s %-25s' % ('Original Label', 'Predicted Label', 'Correct/Wrong'))
for label in y_test:
   print ('%-25s %-25s' % (label, Bag_GBDT_y_pred[i]), end="")
   if (label == Bag_GBDT_y_pred[i]):
      print (' %-25s' % ('Correct'))
      print (' %-25s' % ('Wrong'))
   i = i + 1
\label{lem:print("\nConfusion Matrix:\n",metrics.confusion\_matrix(y\_test, Bag\_GBDT\_y\_pred))} \\
print ("-----")
print("\nClassification Report:\n",metrics.classification_report(y_test, Bag_GBDT_y_pred))
print ("----")
print('Accuracy of the classifer is %0.2f' % metrics.accuracy_score(y_test, Bag_GBDT_y_pred))
print ("-----")
    Original Label Predicted Label Correct/Wrong
                                             Correct
                                             Wrong
    1
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    0
                                             Wrong
                        1
    0
                        a
                                             Correct
    1
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                                              Correct
    0
                                             Correct
    0
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                                              Wrong
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    0
                                              Correct
                                              Correct
print ("\n----")
print ('%-25s %-25s %-25s' % ('Original Label', 'Predicted Label', 'Correct/Wrong'))
print ("-----")
for label in y_test:
   print ('%-25s %-25s' % (label, Fuzzy_GBDT_y_pred[i]), end="")
   if (label == Fuzzy_GBDT_y_pred[i]):
      print (' %-25s' % ('Correct'))
```

```
print (' %-25s' % ('Wrong'))
    i = i + 1
print ("-----")
print("\nConfusion Matrix:\n",metrics.confusion_matrix(y_test, Fuzzy_GBDT_y_pred))
print ("-----")
print("\nClassification Report:\n",metrics.classification_report(y_test, Fuzzy_GBDT_y_pred))
print ("-----")
print('Accuracy of the classifer is %0.2f' % metrics.accuracy_score(y_test, Fuzzy_GBDT_y_pred))
print ("-----")
```

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	_	-

original Label		Correct/Wrong
	0	 Correct
Ĺ	0	Wrong
)	1	Wrong
)	1	Wrong
, L	1	Correct
)	0	Correct
,)	0	Correct
	0	Wrong
	0	Wrong
	0	Correct
	1	Correct
	0	Correct
	0	Correct
	0	Wrong
	1	Correct
	0	Wrong
	1	wrong Wrong
	0	
	0	Wrong Correct
	0	Correct
	1	Correct
	0	Correct
	0	Correct
	1	Correct
	0 0	Correct Correct
	1	Correct
	0	Correct
	1	Correct
	0	Wrong
	0	Wrong
	0	Wrong
	0	Correct
	0	Correct
	0	Wrong
	0	Correct
	1	Correct
	0	Correct
	1	Correct
	0	Wrong
	1	Correct
	1	Correct
	1	Correct
	0	Correct
	1	Correct
	0	Correct
	0	Correct
	0	Correct
	1	Correct
	1	Correct
	0	Correct
	0	Correct
	0	Correct
	1	Correct

→

Original Label	Predicted Label	Correct/Wrong
0	0	Correct
1	0	Wrong
0	1	Wrong
0	0	Correct
1	1	Correct
0	0	Correct
0	0	Correct
1	1	Correct
1	0	Wrong
0	1	Wrong
1	1	Correct
0	0	Correct
0	0	Correct
1	1	Correct
1	1	Correct
1	1	Correct
0	0	Correct
1	1	Correct
0	0	Correct
0	0	Correct
1	1	Correct
0	0	Correct
0	0	Correct
1	1	Correct
0	0	Correct
0	0	Correct
1	1	Correct
0	0	Correct
1	1	Correct
0	1	Wrong
0	0	Correct
1	1	Correct
0	1	Wrong
1	1	Correct
0	0	Correct
1	1	Correct
0	0	Correct
1	1	Correct
0	0	Correct
0	0	Correct
0	0	Correct
1	1	Correct
1	1	Correct
0	0	Correct
0	0	Correct
0	0 1	Correct
1	1	Correct

```
# Calculate the bias error
DT_bias_error = 1 - accuracy_score(y_test, DT_y_pred)
# Calculate the variance error
DT_y_pred_train = DT_model.predict(X_train)
\label{eq:def:def:DT_variance}  \mbox{DT\_variance\_error = accuracy\_score(y\_train, DT\_y\_pred\_train) - accuracy\_score(y\_test, DT\_y\_pred)} 
# Calculate the bias error
GBDT_bias_error = 1 - accuracy_score(y_test, GBDT_y_pred)
# Calculate the variance error
GBDT_y_pred_train = GBDT_best_regressor.predict(X_train)
{\tt GBDT\_variance\_error = accuracy\_score(y\_train, \ {\tt GBDT\_y\_pred\_train}) - accuracy\_score(y\_test, \ {\tt GBDT\_y\_pred})}
print("Bias error: ", GBDT_bias_error)
print("Variance error: ", GBDT_variance_error)
→ Bias error: 0.16519174041297935
     Variance error: 0.05294684245379566
# Calculate the bias error
Bag_GBDT_bias_error = 1 - accuracy_score(y_test, Bag_GBDT_y_pred)
# Calculate the variance error
Bag_GBDT_y_pred_train = Bag_GBDT.predict(X_train)
Bag_GBDT_variance_error = accuracy_score(y_train, Bag_GBDT_y_pred_train) - accuracy_score(y_test, Bag_GBDT_y_pred)
print("Bias error: ", Bag_GBDT_bias_error)
print("Variance error: ", Bag_GBDT_variance_error)
→ Bias error: 0.0471976401179941
     Variance error: 0.03845128443286294
# Calculate the bias error
Fuzzy_GBDT_bias_error = 1 - accuracy_score(y_test, Fuzzy_GBDT_y_pred)
# Calculate the variance error
```