## ML - EXP - 7

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# Dataset



## Code

# → Method - 1 : Bagging ensemble method.

data.head(5)

	age	sex	ср	trestbps	chol	fbs	restecg	thalach	exang	oldpeak	slope	ca	thal	targ€
0	63	1	3	145	233	1	0	150	0	2.3	0	0	1	
1	37	1	2	130	250	0	1	187	0	3.5	0	0	2	
2	41	0	1	130	204	0	0	172	0	1.4	2	0	2	
3	56	1	1	120	236	0	1	178	0	8.0	2	0	2	
4	57	0	0	120	354	0	1	163	1	0.6	2	0	2	

data.tail(5)

	age	sex	ср	trestbps	chol	fbs	restecg	thalach	exang	oldpeak	slope	ca	thal	tar
298	57	0	0	140	241	0	1	123	1	0.2	1	0	3	
299	45	1	3	110	264	0	1	132	0	1.2	1	0	3	
300	68	1	0	144	193	1	1	141	0	3.4	1	2	3	
301	57	1	0	130	131	0	1	115	1	1.2	1	1	3	
302	57	0	1	130	236	0	0	174	0	0.0	1	1	2	

data.describe()

	age	sex	ср	trestbps	chol	fbs	restecg	th
cou	nt 303.000000	303.000000	303.000000	303.000000	303.000000	303.000000	303.000000	303.0
mea	n 54.366337	0.683168	0.966997	131.623762	246.264026	0.148515	0.528053	149.6
sto	9.082101	0.466011	1.032052	17.538143	51.830751	0.356198	0.525860	22.9
mir	29.000000	0.000000	0.000000	94.000000	126.000000	0.000000	0.000000	71.0
25%	47.500000	0.000000	0.000000	120.000000	211.000000	0.000000	0.000000	133.5
50%	55.000000	1.000000	1.000000	130.000000	240.000000	0.000000	1.000000	153.0
75%	61.000000	1.000000	2.000000	140.000000	274.500000	0.000000	1.000000	166.0
ma	<b>x</b> 77.000000	1.000000	3.000000	200.000000	564.000000	1.000000	2.000000	202.0

data.info()

```
<class 'pandas.core.frame.DataFrame'>
     RangeIndex: 303 entries, 0 to 302
     Data columns (total 14 columns):
                    Non-Null Count Dtype
          Column
     ---
      0
                    303 non-null
      1
                    303 non-null
                                    int64
                    303 non-null
                                    int64
          ср
          trestbps 303 non-null
                    303 non-null
         chol
                                    int64
          fbs
                    303 non-null
                                    int64
          restecg
                    303 non-null
                                    int64
                    303 non-null
          thalach
                                    int64
                    303 non-null
      8
                                    int64
          exang
          oldpeak
                                    float64
                    303 non-null
      10 slope
                    303 non-null
                                    int64
                    303 non-null
                                    int64
      12 thal
                    303 non-null
                                    int64
      13 target
                    303 non-null
     dtypes: float64(1), int64(13)
     memory usage: 33.3 KB
# Split the data into features (X) and target (y)
X = data.drop("target", axis=1)
y = data["target"]
# Split the data into training and testing sets
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
# Create a Random Forest classifier
rf_classifier = RandomForestClassifier(n_estimators=100, random_state=42)
# Train the classifier on the training data
rf_classifier.fit(X_train, y_train)
               RandomForestClassifier
     RandomForestClassifier(random_state=42)
# Make predictions on the test data
y_pred = rf_classifier.predict(X_test)
# Evaluate the classifier
accuracy = accuracy_score(y_test, y_pred)
conf_matrix = confusion_matrix(y_test, y_pred)
class_report = classification_report(y_test, y_pred)
# Print the evaluation results
print("Accuracy:", round(accuracy * 100, 2) , "\n\n")
print("Confusion Matrix:\n", conf_matrix, "\n\n")
print("Classification \ Report: \ \ ", \ class\_report, \ \ "\ \ "\ \ ")
     Accuracy: 83.61
     Confusion Matrix:
      [[24 5]
[ 5 27]]
     Classification Report:
                    precision
                                 recall f1-score
                0
                                  0.83
                                            0.83
                                                         29
                        0.84
                                  0.84
                                            0.84
                                                         32
         accuracy
                                            0.84
                                                         61
                        0.84
                                  0.84
                                            0.84
        macro avg
                                                         61
     weighted avg
                                  0.84
                                            0.84
                        0.84
                                                         61
```

### Save the Trained Model

```
import joblib

# Save the trained Random Forest classifier model to a file
model_filename = "random_forest_model.joblib"
joblib.dump(rf_classifier, model_filename)
```

```
print(f"Model saved as '{model_filename}'")

Model saved as 'random_forest_model.joblib'
```

### Load the Save Model

```
[ ] L, 1 cell hidden
```

#### Inference

### - Class 0:

Usually represents the absence of a heart disease condition. In other words, it indicates that the individual does not have heart disease or is classified as "healthy" or "no heart disease."

### Class 1:

Typically represents the presence of a heart disease condition. This indicates that the individual is diagnosed with heart disease or is classified as "having heart disease."

```
import joblib
import pandas as pd
# Load the saved model from the file
model_filename = "random_forest_model.joblib"
loaded_model = joblib.load(model_filename)
# Load new data for inference (replace this with your own data)
new_data = pd.DataFrame({
    'age': [55],
    'sex': [1],
    'cp': [2],
    'trestbps': [130],
    'chol': [250],
    'fbs': [0],
    'restecg': [1],
    'thalach': [155],
    'exang': [0],
    'oldpeak': [1.2],
    'slope': [2],
    'ca': [1],
    'thal': [3]
})
# Use the loaded model to make predictions on new data
new_predictions = loaded_model.predict(new_data)
# Display the predictions
print("Predicted Class for New Data:", new_predictions[0])
     Predicted Class for New Data: 1
if new_predictions[0] == 0:
 print("Heart Disease : Absence ")
else:
 print("Heart Disease : Present ")
```

# Method - 2 : Boosting ensemble method

```
import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn.ensemble import AdaBoostClassifier
from sklearn.tree import DecisionTreeClassifier # Import DecisionTreeClassifier
from sklearn.metrics import accuracy_score, classification_report, confusion_matrix
# Load the dataset
data = pd.read_csv("heart.csv")
```

```
# Split the data into features (X) and target (y)
X = data.drop("target", axis=1)
y = data["target"]
# Split the data into training and testing sets
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
# Create an AdaBoost classifier with a base estimator (e.g., Decision Tree)
base_estimator = DecisionTreeClassifier(max_depth=1) # You can choose a different base estimator if needed
adaboost\_classifier = AdaBoostClassifier (base\_estimator=base\_estimator, n\_estimators=50, random\_state=42)
# Train the AdaBoost classifier on the training data
adaboost_classifier.fit(X_train, y_train)
     /usr/local/lib/python3.10/dist-packages/sklearn/ensemble/_base.py:166: FutureWarning: `base_
       warnings.warn(
                  AdaBoostClassifier
      ▶ base_estimator: DecisionTreeClassifier
             ▶ DecisionTreeClassifier
# Make predictions on the test data
y_pred = adaboost_classifier.predict(X_test)
# Evaluate the classifier
accuracy = accuracy_score(y_test, y_pred)
conf_matrix = confusion_matrix(y_test, y_pred)
class_report = classification_report(y_test, y_pred)
# Print the evaluation results
print("Accuracy:", round(accuracy * 100, 2), "\n\n")
print("Confusion Matrix:\n", conf_matrix, "\n\n")
print("Classification Report:\n", class_report, "\n\n")
     Accuracy: 80.33
     Confusion Matrix:
      [[25 4]
      [ 8 24]]
     Classification Report:
                                recall f1-score support
                    precision
                                 0.86
                0
                        0.76
                                           0.81
                                                        29
                1
                        0.86
                                  0.75
                                           0.80
                                                        32
         accuracy
                                            0.80
                                                        61
                        0.81
                                  0.81
                                            0.80
        macro avg
     weighted avg
                       0.81
                                  0.80
                                            0.80
```

### Save the Trained Model

```
import joblib
# Save the trained Random Forest classifier model to a file
model filename = "adaboost_classifier_model.joblib"
joblib.dump(adaboost_classifier, model_filename)
print(f"Model saved as '{model_filename}'")
     Model saved as 'adaboost_classifier_model.joblib'
```

#### Inference

#### Class 0:

Usually represents the absence of a heart disease condition. In other words, it indicates that the individual does not have heart disease or is classified as "healthy" or "no heart disease."

### Class 1:

Typically represents the presence of a heart disease condition. This indicates that the individual is diagnosed with heart disease or is classified as "having heart disease."

```
# Load the saved AdaBoost classifier model from a file
model_filename = "adaboost_classifier_model.joblib"
loaded_adaboost_model = joblib.load(model_filename)
# Load new data for inference (replace this with your own data)
new_data = pd.DataFrame({
    'age': [55],
    'sex': [1],
    'cp': [2],
   'trestbps': [130],
    'chol': [250],
    'fbs': [0],
   'restecg': [1],
    'thalach': [155],
    'exang': [0],
    'oldpeak': [1.2],
    'slope': [2],
    'ca': [1],
    'thal': [3]
})
# Use the loaded AdaBoost model to make predictions on new data
new_predictions = loaded_adaboost_model.predict(new_data)
# Display the predictions
print("Predicted Class for New Data:", new_predictions[0])
     Predicted Class for New Data: 0
if new_predictions[0] == 0:
 print("Heart Disease : Absence ")
else:
 print("Heart Disease : Present ")
     Heart Disease : Absence
```

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