## Imports

(237, 13) (237,) (60, 13) (60,)

```
from sklearn.model_selection import train_test_split import pandas as pd

from sklearn.ensemble import AdaBoostClassifier from sklearn.linear_model import Perceptron from sklearn.metrics import accuracy_score from sklearn.metrics import confusion_matrix import matplotlib.pyplot as plt import seaborn as sns

from sklearn.base import BaseEstimator, ClassifierMixin from sklearn.utils import resample import numpy as np from sklearn.neural_network import MLPClassifier
```

#### Part1 - Hearth Disease Dataset

```
#Load the dataset from a CSV file
url = 'https://archive.ics.uci.edu/ml/machine-learning-databases/heart-disease/processed.cleveland.data'
column_names = ['age', 'sex', 'cp', 'trestbps', 'chol', 'fbs', 'restecg', 'thalach', 'exang', 'oldpeak', 'slope', 'ca', 'thal', 'target']
heart_data = pd.read_csv(url, names=column_names)
# Preprocess the data
heart_data = heart_data.replace('?', pd.NA).dropna() # Handling missing values
X = heart_data.drop('target', axis=1)
y = heart_data['target']
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     297 rows × 13 columns
 Sonraki adımlar:
                   Xile kod oluşturun

    Önerilen grafikleri göster

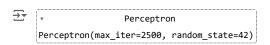
# Split the data into training and testing sets
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
print(X_train.shape)
print(y_train.shape)
print(X_test.shape)
print(y_test.shape)
```

## > EDA



# Part2 - Multi Layer Perceptron

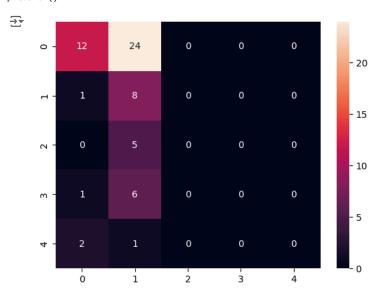
#Perceptron definition to base\_clf and fitting
base\_clf = Perceptron(random\_state=42, max\_iter=2500)
base\_clf.fit(X\_train, y\_train)



#Prediction to preceptron fitting element
y\_pred = base\_clf.predict(X\_test)

#Print accuracy
accuracy = accuracy\_score(y\_test, y\_pred)
print(f"Perceptron: Accuracy = {accuracy}")

#Plot confusion matrix with heatmap
cm = confusion\_matrix(y\_test, y\_pred)
sns.heatmap(cm, annot=True, fmt='d')
plt.show()



#Adaboost classifier, that use Perseptron that we fitted and Samme algorithm. And adaboots fitting
adaBoost = AdaBoostClassifier(estimator=base\_clf, random\_state=42, algorithm='SAMME')
adaBoost.fit(X\_train, y\_train)

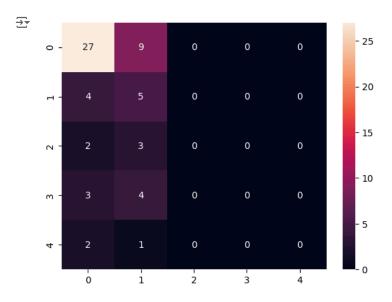
```
AdaBoostClassifier

• estimator: Perceptron

• Perceptron
```

#Adaboost fitting predict
y\_pred = adaBoost.predict(X\_test)

#Print accuracy
accuracy = accuracy\_score(y\_test, y\_pred)
print(f"AdaBoost with MLP as base classifier: Accuracy = {accuracy}")



#### Part3 - Random Decision Forest

```
#Implementation of perceptron
class TrainablePerceptron(BaseEstimator, ClassifierMixin):
    def __init__(self, max_iterations=100):
        self.max iterations = max iterations
        self.perceptron = MLPClassifier(hidden_layer_sizes=(10,), random_state=42, max_iter=2500)
    def fit(self, X, y):
        self.perceptron.fit(X, y)
    def predict(self, X):
        return self.perceptron.predict(X)
#Implementation of random decision forest with Perceptron
class RandomDecisionForest(BaseEstimator, ClassifierMixin):
    def __init__(self, n_estimators=10, max_iterations=100, random_state=None):
        self.n_estimators = n_estimators
        self.max_iterations = max_iterations
        self.random_state = random_state
        self.estimators = []
    def fit(self, X, y):
        for _ in range(self.n_estimators):
            X_resampled, y_resampled = resample(X, y, random_state=self.random_state)
            perceptron = TrainablePerceptron(max_iterations=self.max_iterations)
            \verb"perceptron.fit(X_resampled, y_resampled)"
            self.estimators.append(perceptron)
    def predict(self, X):
       predictions = []
       for estimator in self.estimators:
            predictions.append(estimator.predict(X))
        return np.mean(predictions, axis=0)
#Calling random decision forest class and fitting this
random_forest = RandomDecisionForest(n_estimators=20, max_iterations=2500, random_state=42)
random_forest.fit(X_train, y_train)
```

#Predict fitted random forest
y\_pred = random\_forest.predict(X\_test)

#Print accuracy value
accuracy = accuracy\_score(y\_test, y\_pred)
print(f"Random Decision Forest with Trainable Perceptrons: Accuracy = {accuracy}")

Random Decision Forest with Trainable Perceptrons: Accuracy = 0.5666666666666667

#Plot confusion matrix with heatmap
cm = confusion\_matrix(y\_test, y\_pred)
sns.heatmap(cm, annot=True, fmt='d')
plt.show()

