```
In [5]: import pandas as pd
        from sklearn.model_selection import train_test_split
        from sklearn.metrics import mean_squared_error
        # importing machine learning models for prediction
        import xgboost as xgb
        # importing bagging module
        from sklearn.ensemble import BaggingRegressor
        # loading Iris dataset
        from sklearn.datasets import load_iris
        iris = load_iris()
        # getting target data from the dataset
        target = iris.target
        # getting train data from the dataset
        train = pd.DataFrame(iris.data, columns=iris.feature_names)
        # Splitting between train data into training and validation dataset
        X_train, X_test, y_train, y_test = train_test_split(
        train, target, test_size=0.20)
        # initializing the bagging model using XGBoost as base model with defaul
        model = BaggingRegressor(estimator=xgb.XGBRegressor())
        # training model
        model.fit(X_train, y_train)
        # predicting the output on the test dataset
        pred = model.predict(X_test)
        # printing the mean squared error between real value and predicted value
        print(mean_squared_error(y_test, pred))
        0.03864110131248733
```

```
In [2]: import pandas as pd
        from sklearn.datasets import load_iris
        from sklearn.model_selection import train_test_split
        from sklearn.metrics import accuracy_score
        # importing machine learning models for prediction
        from sklearn.ensemble import GradientBoostingClassifier
        # loading iris dataset
        iris = load_iris()
        # getting feature data from the iris dataset
        features = iris.data
        # getting target data from the iris dataset
        target = iris.target
        # Splitting between train data into training and validation dataset
        X_train, X_test, y_train, y_test = train_test_split(features, target, te
        # initializing the boosting module with default parameters
        model = GradientBoostingClassifier()
        # training the model on the train dataset
        model.fit(X_train, y_train)
        # predicting the output on the test dataset
        pred_final = model.predict(X_test)
        # printing the accuracy score between real value and predicted value
        print(accuracy_score(y_test, pred_final))
```

0.9

```
In [2]:

from sklearn,cluster import Means
from sklearn,cluster import load_iris
import matplotlib.pyplot as plt

# Load the Iris dataset
iris = load_iris()

# Extract the data and target values

X = Iris, data
y = Iris, target

# Create a Moleans object with 3 clusters

kmeans = Moleans (n_clusters=3, n_init=10)

# Fit the Moleans object to the data

kmeans, Irit(X)

# Get the predicted cluster labels

labels = kmeans.predict(X)

# Plot the data points and centroids
plt.scatter(Xi:, 0], Xi:, 1], c=labels, cmap='viridis')
plt.scatter(Means.cluster_centers_[:, 0], kmeans.cluster_centers_[:, 1], marker='*', s=200, c='red')
plt.ylabel(iris.feature_names[0])
plt.ylabel(iris.feature_names[1])

4.0

()

4.5

4.0

2.5

2.0
```

6.5

7.0

7.5

8.0

6.0

4.5

5.0

5.5

```
Iteration 1...
Iteration 2...
Iteration 3...
Iteration 4...
Iteration 5...
Iteration 6...
Iteration 7...
Iteration 8...
Iteration 9...
Iteration 10...
Learned Parameters:
CPT for Burglary:
[0.5 \ 0.5]
CPT for Earthquake:
[0.5 \ 0.5]
CPT for Alarm given Burglary and Earthquake:
[[[0.00000000e+00 1.00000000e+00]
  [9.66383974e-04 9.99033616e-01]]
 [[0.00000000e+00 1.00000000e+00]
  [6.40576290e-03 9.93594237e-01]]]
CPT for JohnCalls given Alarm:
[[0.06051854 0.93948146]
 [0.27711179 0.72288821]]
CPT for MaryCalls given Alarm:
[[0.06051854 0.93948146]
 [0.27711179 0.72288821]]
```

```
x = np.array([[0,0], [0,1], [1,0], [1,1]])
y = np.array([[0], [1], [1], [0]])
                                                                  ↑ ↓ ⊝ 🗏 💠 🗓 📋 :
# Define the network model and its arguments.
# Set the number of neurons/nodes for each layer:
model = Sequential()
model.add(Dense(2, input_shape=(2,)))
model.add(Activation('sigmoid'))
model.add(Dense(1))
model.add(Activation('sigmoid'))
model.compile(loss='mean_squared_error', optimizer='sgd', metrics=['accuracy']) # Print
model.summary()
Model: "sequential_2"
                               Output Shape
 Layer (type)
                                                          Param #
 dense_4 (Dense)
                               (None, 2)
                                                          6
 activation (Activation)
                               (None, 2)
                                                          0
 dense_5 (Dense)
                               (None, 1)
                                                          3
 activation_1 (Activation)
                               (None, 1)
                                                          0
Total params: 9
Trainable params: 9
Non-trainable params: 0
```

Ex_12: