In the model the building part, you can use the cancer dataset, which is a very famous multi-class classificati image of a fine needle aspirate (FNA) of a breast mass. They describe characteristics of the cell nuclei prese

The dataset comprises 30 features (mean radius, mean texture, mean perimeter, mean area, mean smoothne concave points, mean symmetry, mean fractal dimension, radius error, texture error, perimeter error, area error concave points error, symmetry error, fractal dimension error, worst radius, worst texture, worst perimeter, wo concavity, worst concave points, worst symmetry, and worst fractal dimension) and a target (type of cancer). (harmful) and benign (not harmful). Here, you can build a model to classify the type of cancer. The dataset is download it from the UCI Machine Learning Library.

```
import numpy as np #linear algebra library of Python
from sklearn import datasets
cancer = datasets.load_breast_cancer()
cancer.data.shape
\Box
print(cancer.data[0:5])
from sklearn.model selection import train test split #method to split training and tes
X train, X test, y train, y test=train test split(cancer.data, cancer.target, test siz
import xgboost as xgb
from xgboost.sklearn import XGBClassifier
import matplotlib.pyplot as plt
pip install bayesian-optimization
import xgboost as xgb
from xgboost.sklearn import XGBClassifier
import bayes opt
from bayes opt import BayesianOptimization
from sklearn.model_selection import cross_val_score
pbounds = {'n estimators': (50, 1000), 'eta': (0.01, 3), 'max depth': (1,32), 'gamma':
model_tuning = XGBClassifier(n jobs=-1)
def xgboostcv(eta, n_estimators, max_depth, gamma, min_child_weight, subsample, colsam
    return np.mean(cross_val_score(model_tuning, X_train, y_train, cv=5, scoring='accu
optimizer = BayesianOptimization(
    f=xgboostcv,
    pbounds=pbounds,
```

ndom

```
n_iter=3)
print(optimizer.max)
\Box
model = XGBClassifier(eta=2, n_estimators=137, max_depth=10, min_child_weight=5, gamma
model.fit(X_train, y_train)
\Box
y_pred = model.predict(X_test)
from sklearn import metrics
print("Accuracy:", metrics.accuracy_score(y_test, y_pred))
\Box
from sklearn.metrics import confusion_matrix
y_pred=model.predict(X_test)
confusion_matrix(y_test,y_pred)
\Box
```

optimizer.maximize(
 init\_points=2,