

In the model building part, you can use the cancer dataset, which is a very famous multi-class classification image of a fine needle aspirate (FNA) of a breast mass. They describe characteristics of the cell nuclei present in the image. The dataset comprises 30 features (mean radius, mean texture, mean perimeter, mean area, mean smoothness, mean 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, worst concavity, worst concave points, worst symmetry, and worst fractal dimension) and a target (type of cancer). The target can be malignant (harmful) and benign (not harmful). Here, you can build a model to classify the type of cancer. The dataset is available for download 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
```



```
print(cancer.data[0:5])
```

```
from sklearn.model_selection import train_test_split #method to split training and testing data
X_train, X_test, y_train, y_test=train_test_split(cancer.data, cancer.target, test_size=0.2)
```

```
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': (0, 1)}
```

```
model_tuning = XGBClassifier(n_jobs=-1)
```

```
def xgboostcv(eta, n_estimators, max_depth, gamma, min_child_weight, subsample, colsample_bytree):
    return np.mean(cross_val_score(model_tuning, X_train, y_train, cv=5, scoring='accuracy'))
```

```
optimizer = BayesianOptimization(
    f=xgboostcv,
    pbounds=pbounds,
    random_state=1)
```

```
random_state=1),  
    init_points=2,  
    n_iter=3)  
print(optimizer.max)
```



```
model = XGBClassifier(eta=2, n_estimators=137, max_depth=10, min_child_weight=5, gamma=0.1)  
model.fit(X_train, y_train)
```



```
y_pred = model.predict(X_test)  
from sklearn import metrics  
print("Accuracy:", metrics.accuracy_score(y_test, y_pred))
```



```
from sklearn.metrics import confusion_matrix  
y_pred=model.predict(X_test)  
confusion_matrix(y_test,y_pred)
```



