## GRADIENT BOOSTING FRAMEWORK

```
import warnings
warnings.filterwarnings('ignore')
```

## Load datset

```
import pandas as pd
from \ sklearn.datasets \ import \ load\_breast\_cancer, fetch\_california\_housing
california_housing = fetch_california_housing(as_frame=True)
housing_df = california_housing.frame
print("REGRESSION")
display(housing_df.head())
# cls - dataset
breast cancer = load breast cancer(as frame=True)
Bcancer_df = breast_cancer.frame
print("\nCLASSIFICATION")
display(Bcancer_df.head())
REGRESSION
                                                                                                          \overline{\Box}
   MedInc HouseAge AveRooms AveBedrms Population AveOccup Latitude Longitude MedHouseVal
0 8.3252
                      6.984127
                                   1.023810
                                                   322.0
                                                          2.555556
                                                                        37.88
                                                                                  -122.23
                 41.0
                                                                                                  4.526
   8.3014
                      6.238137
                                  0.971880
                                                  2401.0
                                                         2.109842
                                                                        37.86
                                                                                  -122 22
                 21.0
                                                                                                  3.585
                       8.288136
                                                   496.0
2 7.2574
                 52.0
                                   1.073446
                                                          2.802260
                                                                        37.85
                                                                                  -122.24
                                                                                                  3.521
   5 6431
                 52.0
                       5 817352
                                   1 073059
                                                   558.0
                                                          2.547945
                                                                        37.85
                                                                                  -122.25
                                                                                                  3 413
   3.8462
                 52.0
                      6.281853
                                   1.081081
                                                   565.0
                                                         2.181467
                                                                        37.85
                                                                                  -122.25
                                                                                                  3.422
CLASSIFICATION
     mean
               mean
                           mean
                                   mean
                                                mean
                                                             mean
                                                                         mean
                                                                                             mean
                                                                                                                       worst
                                                                                                                                  worst
                                                                                                      fractal
                                                                               concave
                                                                    concavity
   radius
           texture
                     perimeter
                                   area
                                         smoothness
                                                      compactness
                                                                                         symmetry
                                                                                                                    texture
                                                                                                                              perimeter
                                                                                 points
                                                                                                   dimension
     17.99
               10.38
                         122.80 1001.0
                                             0.11840
                                                           0.27760
                                                                       0.3001
                                                                                0.14710
                                                                                            0.2419
                                                                                                      0.07871
                                                                                                                       17.33
                                                                                                                                  184.60 201
     20.57
                                1326 0
                                             0.08474
                                                           0.07864
                                                                       0.0869
                                                                                0.07017
                                                                                            0.1812
                                                                                                      0.05667
1
               17 77
                         132 90
                                                                                                                       23 41
                                                                                                                                  158 80
                                                                                                                                         195
                                                                                0.12790
               21.25
                         130.00 1203.0
                                             0.10960
                                                           0.15990
                                                                       0.1974
                                                                                            0.2069
                                                                                                      0.05999
                                                                                                                                         170
2
     19.69
                                                                                                                       25.53
                                                                                                                                  152.50
               20.38
                          77.58
                                  386.1
                                             0.14250
                                                           0.28390
                                                                                0.10520
                                                                                            0.2597
                                                                                                      0.09744
                                                                                                                       26.50
                                                                                                                                   98.87
                                                                                                                                           56
3
     11.42
                                                                       0.2414
               14.34
                         135.10 1297.0
                                             0.10030
                                                           0.13280
                                                                                            0.1809
     20.29
                                                                       0.1980 0.10430
                                                                                                      0.05883
                                                                                                                       16.67
                                                                                                                                  152.20 157
5 rows × 31 columns
```

# shape

```
print("reg",housing_df.shape)
print("cls",Bcancer_df.shape)

reg (20640, 9)
cls (569, 31)
```

## info

```
housing_df.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 20640 entries, 0 to 20639
Data columns (total 9 columns):
```

```
Column
                Non-Null Count Dtype
    MedInc
                20640 non-null float64
1
    HouseAge
                20640 non-null float64
    AveRooms
                20640 non-null float64
                20640 non-null float64
    AveBedrms
    Population 20640 non-null float64
                20640 non-null float64
    AveOccup
    Latitude
                20640 non-null float64
    Longitude
                20640 non-null float64
   MedHouseVal 20640 non-null float64
dtypes: float64(9)
memory usage: 1.4 MB
```

```
Bcancer_df.info()
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 569 entries, 0 to 568
Data columns (total 31 columns):
                                Non-Null Count Dtype
    Column
    -----
a
    mean radius
                               569 non-null
                                                  float64
    mean texture
                                569 non-null
                                                  float64
                               569 non-null
    mean perimeter
                                                 float64
3
    mean area
                               569 non-null
                                                 float64
                                569 non-null
    mean smoothness
                                                  float64
                             569 non-null
    mean compactness
                                                 float64
6
    mean concavity
                               569 non-null
                                                 float64
    mean concavity
mean concave points
                                569 non-null
                                                  float64
                                569 non-null
    mean symmetry
                                                  float64
    mean fractal dimension 569 non-null
9
                                                  float64
10 radius 1
11 texture error
12 perimeter error 569 non-null 569 non-null 569 non-null 2 200-null
                                                  float64
                                                  float64
                                                  float64
                                                  float64
                                                  float64
14 smoothness error 569 non-null 569 non-null
                                                  float64
                                                  float64
17 concave points error
                                569 non-null
                                                  float64
18 symmetry error
                                569 non-null
                                                  float64
19 fractal dimension error 569 non-null
                                                  float64
20 worst radius 569 non-null
                                                  float64
21 worst texture
                                569 non-null
                                                  float64
22 worst perimeter 569 non-null 23 worst area 569 non-null 24 worst smoothness 569 non-null
                                                  float64
24 worst smoothness 569 non-null
25 worst compactness 569 non-null
26 worst concavity 569 non-null
27 worst concavity
                                                  float64
                                                  float64
                                                  float64
                                                  float64
27 worst concave points
                                569 non-null
                                                  float64
                                569 non-null
28 worst symmetry
                                                  float64
29 worst fractal dimension 569 non-null
                                                  float64
30 target
                                569 non-null
                                                  int64
dtypes: float64(30), int64(1)
memory usage: 137.9 KB
```

```
housing_df["MedHouseVal"].median()

1.797
```

## remove noise columns

```
print("reg",housing_df.shape)
print("cls",Bcancer_df.shape)

reg (20640, 6)
cls (569, 21)
```

# train test split

```
from sklearn.model_selection import train_test_split

# For housing_df
X_housing = housing_df.drop(columns=['MedHouseVal'])
y_housing = housing_df['MedHouseVal']
X_train_housing, X_test_housing, y_train_housing, y_test_housing = train_test_split(X_housing, y_housing, test_size=0.2, random_st

# For Bcancer_df
X_bcancer = Bcancer_df.drop(columns=['target'])
y_bcancer = Bcancer_df['target']
X_train_bcancer, X_test_bcancer, y_train_bcancer, y_test_bcancer = train_test_split(X_bcancer, y_bcancer, test_size=0.2, random_st
```

## XGBoost

#### REGRESSION

```
import xgboost as xgb
from sklearn.metrics import mean_squared_error, r2_score

# Model
xgb_reg = xgb.XGBRegressor(n_estimators=100, learning_rate=0.1, max_depth=3, random_state=42)
xgb_reg.fit(X_train_housing, y_train_housing)

# Predict & Evaluate
y_pred = xgb_reg.predict(X_test_housing)
print("XGBoost Regression MSE:", mean_squared_error(y_test_housing, y_pred))
print("R2:", r2_score(y_test_housing, y_pred))

XGBoost Regression MSE: 0.4411640910853581
R2: 0.6633386229712962
```

```
import matplotlib.pyplot as plt
import numpy as np
import seaborn as sns
from xgboost import plot_tree

plt.figure(figsize=(20,10))
plot_tree(xgb_reg, num_trees=0)
plt.title("XGBoost Regression Tree (Housing)")
plt.show()
```

<Figure size 2000x1000 with 0 Axes>

#### XGBoost Regression Tree (Housing)



### CLASSIFICATION

```
from sklearn.metrics import accuracy_score

# Model
xgb_clf = xgb.XGBClassifier(n_estimators=100, learning_rate=0.1, max_depth=3, random_state=42)
xgb_clf.fit(X_train_bcancer, y_train_bcancer)

# Predict & Evaluate
```

```
y_pred_c = xgb_clf.predict(X_test_bcancer)
print("XGBoost Classification Accuracy:", accuracy_score(y_test_bcancer, y_pred_c))

XGBoost Classification Accuracy: 0.956140350877193
```

```
plt.figure(figsize=(15,8))
plot_tree(xgb_clf, num_trees=0)
plt.title("XGBoost Classification Tree (Breast Cancer)")
plt.show()

<Figure size 1500x800 with 0 Axes>

XGBoost Classification Tree (Breast Cancer)

Word concer points 0.14209997

Word concer points 0.14209997
```

```
from sklearn.metrics import confusion_matrix

cm = confusion_matrix(y_test_bcancer, y_pred_c)
sns.heatmap(cm, annot=True, fmt="d", cmap="Blues")
plt.title("X6Boost Classification Confusion Matrix")
plt.ylabel("Predicted")
plt.ylabel("Actual")
plt.show()

XGBoost Classification Confusion Matrix

-60
-30
-30
-10
```

# LightGBM

## REGRESSION

0

Predicted

```
import lightgbm as lgb
from sklearn.metrics import mean_squared_error, r2_score

lgb_reg = lgb.LGBMRegressor(
    n_estimators=200,
    learning_rate=0.05,
    max_depth=-1,
    num_leaves=31,
    min_child_samples=5,
    random_state=42,
    verbose=-1)
```

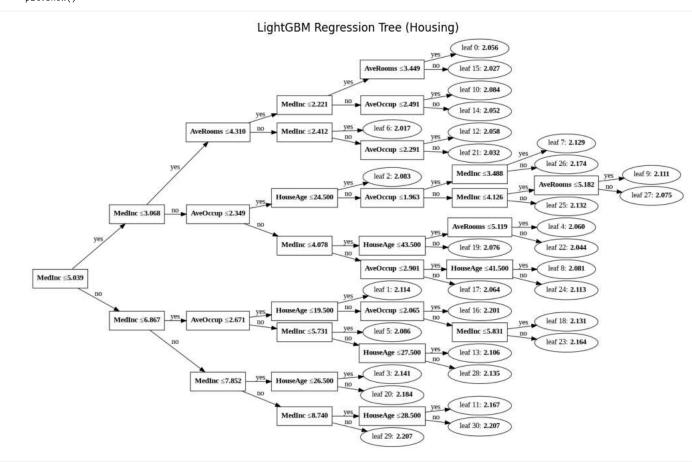
i

```
lgb_reg.fit(X_train_housing, y_train_housing)
y_pred_L = lgb_reg.predict(X_test_housing)
print("LightGBM Regression MSE:", mean_squared_error(y_test_housing, y_pred_L))
print("R2:", r2_score(y_test_housing, y_pred_L))

LightGBM Regression MSE: 0.4189918915761474
R2: 0.6802586837136873
```

```
import matplotlib.pyplot as plt
import lightgbm as lgb
from lightgbm import plot_tree

ax = plot_tree(lgb_reg, tree_index=0, figsize=(15,8))
plt.title("LightGBM Regression Tree (Housing)")
plt.show()
```



### CLASSIFICATION

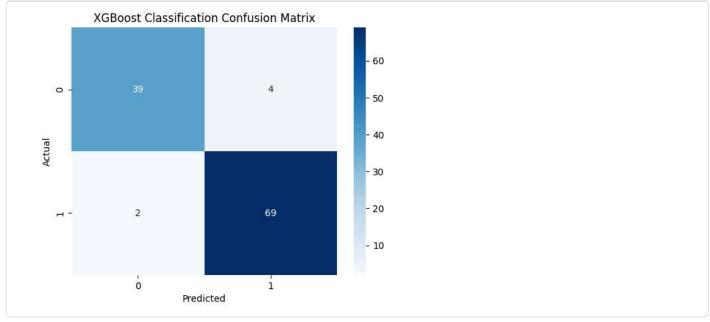
```
lgb_clf = lgb.LGBMClassifier(n_estimators=100, learning_rate=0.1, max_depth=3, random_state=42)
lgb_clf.fit(X_train_bcancer, y_train_bcancer)

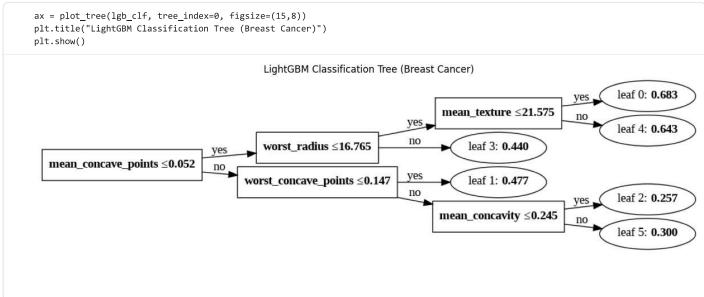
y_pred_LC = lgb_clf.predict(X_test_bcancer)
print("LightGBM Classification Accuracy:", accuracy_score(y_test_bcancer, y_pred_LC))

LightGBM Classification Accuracy: 0.9473684210526315
```

```
from sklearn.metrics import confusion_matrix

cm = confusion_matrix(y_test_bcancer, y_pred_LC)
sns.heatmap(cm, annot=True, fmt="d", cmap="Blues")
plt.title("XGBoost Classification Confusion Matrix")
plt.xlabel("Predicted")
plt.ylabel("Actual")
plt.show()
```





### CatBoost

```
pip install catboost
Requirement already satisfied: catboost in /usr/local/lib/python3.12/dist-packages (1.2.8)
Requirement already satisfied: graphviz in /usr/local/lib/python3.12/dist-packages (from catboost) (0.21)
Requirement already satisfied: matplotlib in /usr/local/lib/python3.12/dist-packages (from catboost) (3.10.0)
Requirement already satisfied: numpy<3.0,>=1.16.0 in /usr/local/lib/python3.12/dist-packages (from catboost) (2.0.2)
Requirement already satisfied: pandas>=0.24 in /usr/local/lib/python3.12/dist-packages (from catboost) (2.2.2)
Requirement already satisfied: scipy in /usr/local/lib/python3.12/dist-packages (from catboost) (1.16.2)
Requirement already satisfied: plotly in /usr/local/lib/python3.12/dist-packages (from catboost) (5.24.1)
Requirement already satisfied: six in /usr/local/lib/python3.12/dist-packages (from catboost) (1.17.0)
Requirement already satisfied: python-dateutil>=2.8.2 in /usr/local/lib/python3.12/dist-packages (from pandas>=0.24->catboost) (2.9
Requirement already satisfied: pytz>=2020.1 in /usr/local/lib/python3.12/dist-packages (from pandas>=0.24->catboost) (2025.2)
Requirement already satisfied: tzdata>=2022.7 in /usr/local/lib/python3.12/dist-packages (from pandas>=0.24->catboost) (2025.2)
Requirement already satisfied: contourpy>=1.0.1 in /usr/local/lib/python3.12/dist-packages (from matplotlib->catboost) (1.3.3)
Requirement already satisfied: cycler>=0.10 in /usr/local/lib/python3.12/dist-packages (from matplotlib->catboost) (0.12.1)
Requirement already satisfied: fonttools>=4.22.0 in /usr/local/lib/python3.12/dist-packages (from matplotlib->catboost) (4.60.0)
Requirement already satisfied: kiwisolver>=1.3.1 in /usr/local/lib/python3.12/dist-packages (from matplotlib->catboost) (1.4.9)
Requirement already satisfied: packaging>=20.0 in /usr/local/lib/python3.12/dist-packages (from matplotlib->catboost) (25.0)
Requirement already satisfied: pillow>=8 in /usr/local/lib/python3.12/dist-packages (from matplotlib->catboost) (11.3.0)
Requirement already satisfied: pyparsing>=2.3.1 in /usr/local/lib/python3.12/dist-packages (from matplotlib->catboost) (3.2.4)
Requirement already satisfied: tenacity>=6.2.0 in /usr/local/lib/python3.12/dist-packages (from plotly->catboost) (8.5.0)
```

```
from catboost import CatBoostRegressor

cat_reg = CatBoostRegressor(iterations=100, learning_rate=0.1, depth=3, verbose=0, random_state=42)

cat_reg.fit(X_train_housing, y_train_housing)

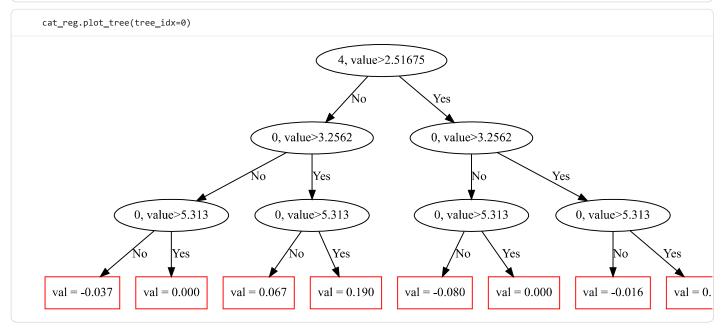
y_pred_cat = cat_reg.predict(X_test_housing)

print("CatBoost Regression MSE:", mean_squared_error(y_test_housing, y_pred_cat))

print("R2:", r2_score(y_test_housing, y_pred_cat))

CatBoost Regression MSE: 0.4504888076209366

R2: 0.6562227403038132
```



### CLASSIFICATION

```
from catboost import CatBoostClassifier

cat_clf = CatBoostClassifier(iterations=100, learning_rate=0.1, depth=3, verbose=0, random_state=42)
cat_clf.fit(X_train_bcancer, y_train_bcancer)

y_pred_catc = cat_clf.predict(X_test_bcancer)
print("CatBoost Classification Accuracy:", accuracy_score(y_test_bcancer, y_pred_catc))

CatBoost Classification Accuracy: 0.9649122807017544
```

```
CatBoost Classification Accuracy: 0.9649122807017544

from sklearn.metrics import confusion_matrix

cm = confusion_matrix(y_test_bcancer, y_pred_catc)
sns.heatmap(cm, annot=True, fmt="d", cmap="Blues")
plt.xitle("XGBoost Classification Confusion Matrix")
plt.xlabel("Predicted")
plt.ylabel("Actual")
plt.show()
```

