## Gradient Boosting

```
# from google.colab import files
 # up = files.upload()
import dataset
  import pandas as pd
 df = pd.read csv('df.csv')
 df.head(3)
             f2
     0 16.5 202.0 865.500000 1880.0 50.000000
    1 18.0 204.0 688.000000 1738.5 44.000000
     2 18.0 203.0 583.666667 1470.0 66.666667
cleaning
 # clean the data
encoding
# encode the data
define x, y
import numpy as np
x = df[['f1', 'f2', 'f3']].values
y = df['T'].values
spliting
### finding best random state
# from sklearn.model_selection import train_test_split
# from sklearn.ensemble import GradientBoostingRegressor
# from sklearn.metrics import r2_score
# import time
# t1 = time.time()
# lst = []
# for i in range(1,10):
      x_train, x_test, y_train, y_test = train_test_split(x, y, test_size=0.2, random_state=i)
      gbr = GradientBoostingRegressor(n_estimators=100, learning_rate=0.1, max_depth=3, random_state=1)
#
#
      gbr.fit(x_train, y_train)
      yhat_test = gbr.predict(x_test)
#
      r2 = r2_score(y_test, yhat_test)
      lst.append(r2)
# t2 = time.time()
# print(f"run time: {round((t2 - t1) / 60 , 0)} min")
# print(f"R2_score = {round(max(lst),2)}")
# print(f"random_state = {np.argmax(lst) + 1}")
from sklearn.model_selection import train_test_split
x_train, X_test, y_train, y_test = train_test_split(x, y, test_size=0.25, random_state=42)
```

```
scaling
```

# XGBoost Regression doesn't need scaling

MAE (test data): 7.943147131116684

## fit the model

```
### K-fold cross validation
# from sklearn.ensemble import GradientBoostingRegressor
# from sklearn.model selection import GridSearchCV
# parameters = {
#
      '': [],
      '': []
#
# }
# gb = GradientBoostingRegressor(random_state=1)
# gs = GridSearchCV(estimator=gb, param grid=parameters, cv=5)
# gs.fit(x_train, y_train)
# best_params = gs.best_params_
# print(best_params)
# def param
# loss='squared_error', learning_rate=0.1, n_estimators=100, subsample=1.0,
# criterion='friedman_mse', min_samples_split=2, min_samples_leaf=1,
# min_weight_fraction_leaf=0.0, max_depth=3, min_impurity_decrease=0.0,
# init=None, random_state=None, max_features=None, alpha=0.9, verbose=0,
# max_leaf_nodes=None, warm_start=False, validation_fraction=0.1,
# n_iter_no_change=None, tol=0.0001, ccp_alpha=0.0
from sklearn.ensemble import GradientBoostingRegressor
gbr = GradientBoostingRegressor(n_estimators=100, learning_rate=0.1, max_depth=3, random_state=42)
gbr.fit(x train, y train)
          {\tt GradientBoostingRegressor}
    GradientBoostingRegressor(random_state=42)
predict test data
yhat_test = gbr.predict(X_test)

    Fvaluate the model

from sklearn.metrics import r2_score, mean_squared_error, mean_absolute_error
print("r2-score (train data): %0.4f" % r2_score(y_train, gbr.predict(x_train)))
print("r2-score (test data): %0.4f" % r2_score(y_test, yhat_test))
→ r2-score (train data): 0.8758
    r2-score (test data): 0.3618
print(f"MSE (train data): {mean_squared_error(y_train, gbr.predict(x_train))}")
print(f"MAE (train data): {mean absolute error(y train, gbr.predict(x train))}")
print(f"MSE (test data): {mean_squared_error(y_test, yhat_test)}")
print(f"MAE (test data): {mean_absolute_error(y_test, yhat_test)}")
→ MSE (train data): 23.444831592628613
    MAE (train data): 3.7765383463895206
    MSE (test data): 93.44204851804126
```

## save the model

```
# import joblib
# joblib.dump(gbr, 'gbr_model.pkl')
```

## load the model

```
# import joblib
# gbr = joblib.load('gbr_model.pkl')
```