## Adaboost Regression

```
# from google.colab import files
 # up = files.upload()
import dataset
  import pandas as pd
 df = pd.read_csv('df.csv')
 df.head(3)
     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
 # df.info()
cleaning
# clean the data
encoding
# encode the data
define x, y
import numpy as np
x = df[['f1', 'f2', 'f3', 'f4']].values
y = df['T'].values
spliting
### finding best random state
# from sklearn.model_selection import train_test_split
# from sklearn.ensemble import AdaBoostRegressor
# from sklearn.tree import DecisionTreeRegressor
# 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=0.25, random_state=i)
#
      base_estimator = DecisionTreeRegressor(max_depth=3)
#
      abr = AdaBoostRegressor(estimator=base_estimator, n_estimators=100, random_state=42)
#
      abr.fit(x_train, y_train)
#
      yhat_test = abr.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
# decision tree based model do not need scaling
fit the model
### K-fold cross validation
# from sklearn.ensemble import AdaBoostRegressor
# from sklearn.tree import DecisionTreeRegressor
# from sklearn.model_selection import GridSearchCV
# parameters = {
     '': [],
#
      '': []
#
# }
# ab = AdaBoostRegressor(random state=42)
# gs = GridSearchCV(estimator=ab, param grid=parameters, cv=5)
# gs.fit(x_train, y_train)
# best_params = gs.best_params_
# print(best_params)# from sklearn.model_selection import GridSearchCV
# def param
# estimator=None, n_estimators=50, random_state=None
# loss='linear', learning rate=1.0
from sklearn.ensemble import AdaBoostRegressor
from sklearn.tree import DecisionTreeRegressor
base_estimator = DecisionTreeRegressor(max_depth=3)
abr = AdaBoostRegressor(estimator=base_estimator, n_estimators=100, random_state=42)
abr.fit(x_train, y_train)
          AdaBoostRegressor
              estimator:
          DecisionTreeRegressor
       ▶ DecisionTreeRegressor ??
predict test data
yhat_test = abr.predict(x_test)
v evaluate the model
from sklearn.metrics import r2_score
print("r2-score (train data): %0.4f" % r2_score(y_train, abr.predict(x_train)))
print("r2-score (test data): %0.4f" % r2_score(y_test, yhat_test))
→ r2-score (train data): 0.6443
    r2-score (test data): 0.3732
```

```
from sklearn.metrics import mean_squared_error
from sklearn.metrics import mean_absolute_error
print(f"MSE (train data): {mean_squared_error(y_train, abr.predict(x_train))}")
print(f"MAE (train data): {mean_absolute_error(y_train, abr.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): 67.16895591769736
MAE (train data): 7.084704259047143
MSE (test data): 91.77392597007267
MAE (test data): 7.828614789194458

v save the model

# import joblib
# joblib.dump(adr, 'abr_model.pkl')

v load the model
```

# import joblib

# abr = joblib.load('abr\_model.pkl')