IMPORT DATASETS

```
In [8]:
         import pandas as pd
         import numpy as np
         from matplotlib import pyplot as plt
         import seaborn as sns
         from tgdm import tgdm
         import warnings
         warnings.filterwarnings('iqnore')
         import time
         from sklearn.ensemble import RandomForestRegressor, AdaBoostRegressor, GradientBoostingRegressor, StackingRegressor
         from sklearn.feature selection import SelectFromModel
         from sklearn.model selection import train test split, GridSearchCV, RandomizedSearchCV
         from sklearn.linear model import LinearRegression
         from sklearn.metrics import r2 score, mean absolute error, make scorer
         from sklearn.model selection import KFold, cross val score, RepeatedKFold
         from sklearn.preprocessing import LabelEncoder, OneHotEncoder, PowerTransformer, StandardScaler
         from scipy.stats import kurtosis, skew
         from scipy.special import boxcox, inv boxcox
         from xqboost import XGBRFRegressor, Booster
         from sklearn.tree import DecisionTreeRegressor
         from sklearn.linear model import Lasso, Ridge, ElasticNet
         from sklearn.svm import SVR
         from lightgbm import LGBMRegressor
         from sklearn.neighbors import KNeighborsRegressor
         from prettytable import PrettyTable
         import joblib
         from sklearn.pipeline import Pipeline
In [ ]:
         from google.colab import drive
         drive.mount('/content/drive')
        Mounted at /content/drive
In [4]:
        def r2 score transform(train pred, test pred, train label, test label):
```

```
This function convert data into inverse yeo-johnson transformation and give a r2 score
Input - train and test predication array, train and test label
Output - inverse transform the array and compute the R2 score

# Inverse transform the arrays
a, b = pt.inverse_transform(train_pred.reshape(-1,1)), pt.inverse_transform(test_pred.reshape(-1,1))
x, y = pt.inverse_transform(train_label.reshape(-1,1)), pt.inverse_transform(test_label.reshape(-1,1))
# get r2 score of train and test data
train = r2_score(x,a)
test = r2_score(y,b)

return train, test
```

```
In [5]:
         def kfold grid search(clf, params, train, label, fold, kfold, search = 'grid'):
           1.1.1
           This function compute the grid or random search cross validation and taken best estimator compute
           R^2 score with kfold validation.
           Input:
             clf = model
             params = dict of parameters for ranfom or grid search
             train = train dataset
             label = train data label
             fold = grid or random cross validation folds
             kfold = fold for kfold CV
             search = grid or random
           1.1.1
           start = time.time()
           # declare the grid search
           if search == 'grid' :
             clf grid = GridSearchCV(estimator = clf,
                                 param grid = params,
                                 n jobs = -1,
                                 scoring = 'r2',
                                 cv = fold,
```

```
verbose = 1)
  # fit data into the grid model
  clf grid.fit(train, label)
  # best estimator for the model
  print('\n')
  print(f'Best Estimator : {clf grid.best estimator }')
  print('\n')
  # Model as best estimator
  clf = clf grid.best estimator
# declre the random search
elif search == 'random' :
  clf grid = RandomizedSearchCV(estimator = clf,
                      param distributions = params,
                      n jobs = -1,
                      scoring = 'r2',
                      cv = fold,
                      verbose = 1,
  # fit data into the grid model
  clf grid.fit(train, label)
  # best estimator for the model
  print('\n')
  print(f'Best Estimator : {clf grid.best estimator }')
  print('\n')
  # Model as best estimator
  clf = clf grid.best estimator
else:
  clf = clf
# genrate folds
kfold = KFold(n splits= kfold, random state= 42)
tr = [ ]
te = [ ]
i = 0
```

```
print('R2 metric : ')
for train in, test in in kfold.split(train):
 # set indices of train and test data
 xtrain, xtest = train.iloc[train in], train.iloc[test in]
 ytrain, ytest = label[train in], label[test in]
 # set best estimator as model
 model = clf
 # fit data into the best estimator
 model.fit(xtrain, ytrain)
 # predict in train and test data
 train pred = model.predict(xtrain)
 test pred = model.predict(xtest)
  # get r2 score of train and test data
 tr pre = r2 score(ytrain, train pred)
 te pre = r2 score(ytest, test pred)
 tr.append(tr pre)
 te.append(te pre)
  print(f'{i} r2 score of the train data {tr pre} and test data {te pre}')
 i += 1
print(f'\nAvg r2 score of train {np.mean(tr)} and test {np.mean(te)}')
end = time.time()
print(f'Time taken - {(end-start)/60} min')
```

Models with PCA + Synthetic and Binary features

• This datasets contain PCA (5) features, Binary features, label encoded features and created synthetic features.

```
In [ ]: # Import the train and test csv files
        train = pd.read csv(r'/content/drive/MyDrive/Datasets/pca+feature train.csv')
        test = pd.read csv(r'/content/drive/MyDrive/Datasets/pca+feature test.csv')
        train = train.drop(['Unnamed: 0'], axis = 1)
        test = test.drop(['Unnamed: 0'], axis = 1)
         train.head()
                 y X0 X1 X2 X3 X4 X5 X6 X8 X10 X11 X12 X13 X14 X15 X16 X17 X18 X19 X20 X21 X22 X23 X24 X26 X27 X28
Out[ ]:
        0 0 130.81
                                                    0
                                                         0
                                                                 0
                                                                     0
                                                                          0
                                                                                                                        0
              88.53
                                                                                                                        0
              76.26 az w
                                                             0
                                                                 0
                                                                     0
                                                                          0
                                                                                  0
                                                                                           0
                                                                                                       0
                                                                                                            0
                          n
                             c d x j x
                                                                                                                        1
                                                                 0
              80.62 az
                       t n f d x l e
                                                             0
                                                                          0
        4 13 78.02 az v n f d h d n
                                                0
                                                    0
                                                             0
                                                                 0
                                                                              0
                                                                                  0
       5 rows × 404 columns
In [ ]:
        #Check the Shape of the test and train data
         print('train data shape - ', train.shape)
         print('test data shape - ', test.shape)
        train data shape - (4209, 404)
        test data shape - (4209, 403)
In [ ]:
         # declare the train set and label
        y2 = train.y
        x2 = train.iloc[:, 10:]
In [ ]:
         # Split train data into train-set and test-set
        x train2, x test2, y train2, y test2 = train test split(x2,y2, test size = 0.33, random state= 42)
         print(x train2.shape, y train2.shape)
         print(x test2.shape, y test2.shape)
```

```
(2820, 394) (2820,)
(1389, 394) (1389,)
```

Decision Tree

```
In [ ]:
         clf = DecisionTreeRearessor()
         params = \{ \text{'max depth'} : [2,3,4,8,10,15], 
                   'max features' : ['auto', 'sqrt', 'log2'],
                   'random state' : [5,10,20,30],
         kfold grid search(clf, params, x2, y2, 10, kfold = 15, search= 'random') \#(0.5780)
        Fitting 10 folds for each of 10 candidates, totalling 100 fits
        [Parallel(n jobs=-1)]: Using backend LokyBackend with 2 concurrent workers.
        [Parallel(n jobs=-1)]: Done 100 out of 100 | elapsed: 3.9s finished
        Best Estimator: DecisionTreeRegressor(ccp alpha=0.0, criterion='mse', max depth=8,
                              max features='auto', max leaf nodes=None,
                              min impurity decrease=0.0, min impurity split=None,
                              min samples leaf=1, min samples split=2,
                              min weight fraction leaf=0.0, presort='deprecated',
                              random state=10, splitter='best')
        R2 metric :
        0 r2 score of the train data 0.6813107472320485 and test data 0.5767082371099517
        1 r2 score of the train data 0.621586039510241 and test data 0.708513052109607
        2 r2 score of the train data 0.630831290790349 and test data 0.5631604647843083
        3 r2 score of the train data 0.6668606818390898 and test data 0.2787127830840411
        4 r2 score of the train data 0.6380355155912681 and test data 0.06995160992884453
        5 r2 score of the train data 0.6311910610649891 and test data 0.5245457377028628
        6 r2 score of the train data 0.63061835338783 and test data 0.6205419720909853
        7 r2 score of the train data 0.6221887781115217 and test data 0.6688339348247292
        8 r2 score of the train data 0.6419099223480337 and test data 0.46515590899698256
        9 r2 score of the train data 0.6279626162710936 and test data 0.5227980632203781
        10 r2 score of the train data 0.6308182124521897 and test data 0.4505968373538033
        11 r2 score of the train data 0.6259084033271662 and test data 0.5516506258243561
        12 r2 score of the train data 0.622938628458103 and test data 0.5444741383056872
        13 r2 score of the train data 0.6294581983558882 and test data 0.5987995140027604
```

R^2 Score 0.5734756571589295 data

Random Forest

```
In [ ]:
         clf = RandomForestRegressor()
         params = \{'n \text{ estimators'}: [40,50,60,70,100],
                      'max depth':[3,5,6,7,8],
                      'min samples split':[2,3,4,5,6,7,8,9,10],
                      'max features': [0.80,.95, 1.0],
                      'min samples leaf': [1, 2,3,4,5,6,7,8,9],
                      'min impurity decrease': [1e-5,1e-4,1e-3,1e-2,1e-1,0,1,10]}
         kfold grid search(clf, params, x2, y2, fold = 10, kfold=15, search= 'random')
        Fitting 10 folds for each of 10 candidates, totalling 100 fits
        [Parallel(n iobs=-1)]: Using backend LokyBackend with 2 concurrent workers.
        [Parallel(n jobs=-1)]: Done 46 tasks | elapsed: 1.4min
        [Parallel(n jobs=-1)]: Done 100 out of 100 | elapsed: 2.8min finished
        Best Estimator: RandomForestRegressor(bootstrap=True, ccp alpha=0.0, criterion='mse',
                              max depth=3, max features=0.8, max leaf nodes=None,
                              max samples=None, min impurity decrease=1e-05,
                              min impurity split=None, min samples leaf=5,
                              min samples split=9, min weight fraction leaf=0.0,
```

```
n_estimators=40, n_jobs=None, oob_score=False,
random_state=None, verbose=0, warm_start=False)
```

```
R2 metric :
        0 r2 score of the train data 0.5715153440871907 and test data 0.6075821902602618
        1 r2 score of the train data 0.5674883076551979 and test data 0.6511201254823615
        2 r2 score of the train data 0.5730553993723171 and test data 0.6021445241149124
        3 r2 score of the train data 0.60601763083342 and test data 0.30708047504635183
        4 r2 score of the train data 0.5773519833442724 and test data 0.5444356633636224
        5 r2 score of the train data 0.575621620235263 and test data 0.5670822703371299
        6 r2 score of the train data 0.5696064626862604 and test data 0.6596220214589319
        7 r2 score of the train data 0.5671942422662717 and test data 0.7064270195874032
        8 r2 score of the train data 0.5826081962660211 and test data 0.46894952095071496
        9 r2 score of the train data 0.5748447938417933 and test data 0.5578146764787673
        10 r2 score of the train data 0.5786768813257471 and test data 0.4979097062115938
        11 r2 score of the train data 0.5727663404360401 and test data 0.5977121928418523
        12 r2 score of the train data 0.569489839201403 and test data 0.6548764845232085
        13 r2 score of the train data 0.5702566085082328 and test data 0.6372309483627063
        14 r2 score of the train data 0.5687708169055417 and test data 0.6442986162389157
        Avg r2 score of train 0.5750176311309981 and test 0.5802857623505823
        Time taken - 3.083255358537038 min
In [ ]:
         rf = RandomForestRegressor(bootstrap=True, ccp alpha=0.0, criterion='mse',
                               max depth=3, max features=0.8, max leaf nodes=None,
                               max samples=None, min impurity decrease=1e-05,
                               min impurity split=None, min samples leaf=5,
                               min samples split=9, min weight fraction leaf=0.0,
                               n estimators=40, n jobs=None, oob score=False,
                               random state=None, verbose=0, warm start=False)
         a = cross val score(rf, x2, y2, scoring= 'r2', cv= 10)
         print(f'R^2 Score {np.mean(a)} data')
```

R^2 Score 0.5745664501511195 data

XGBoost-Regressor

```
In [ ]: clf = XGBRFRegressor(silent=True)
    xparams = {'learning_rate':[0.1,0.5,0.8,1],
```

```
'n estimators':[70,80,100],
              'max depth':[2,3,4],
              'colsample bytree': [0.1,0.5,0.7,0.9,1],
              'subsample': [0.2,0.3,0.5,1],
              'gamma': [0.0001,0.001,0,0.1,0.01,0.5,1],
              'reg alpha':[0.00001,0.0001,0.001,0.01,0.1]}
kfold grid search(clf, params, x2, y2, fold = 10, kfold=15, search= 'random')
Fitting 10 folds for each of 10 candidates, totalling 100 fits
[Parallel(n jobs=-1)]: Using backend LokyBackend with 2 concurrent workers.
[Parallel(n jobs=-1)]: Done 46 tasks | elapsed: 1.6min
[Parallel(n jobs=-1)]: Done 100 out of 100 | elapsed: 3.5min finished
Best Estimator: XGBRFRegressor(base score=0.5, colsample bylevel=1, colsample bynode=0.8,
               colsample bytree=1, gamma=0, learning rate=1, max delta step=0,
               max depth=15, max features='log2', min child weight=1,
               missing=None, n estimators=100, n jobs=1, nthread=None,
               objective='reg:linear', random state=5, reg alpha=0,
               reg lambda=1, scale pos weight=1, seed=None, silent=True,
               subsample=0.8, verbosity=1)
R2 metric :
0 r2 score of the train data 0.5591106706797955 and test data 0.5969348853865986
1 r2 score of the train data 0.5567761302990156 and test data 0.6392482807896993
2 r2 score of the train data 0.5589944292139439 and test data 0.6035569639748404
3 r2 score of the train data 0.5925375912978497 and test data 0.3071454665475717
4 r2 score of the train data 0.5640973355720406 and test data 0.5338131420792587
5 r2 score of the train data 0.5621640135494697 and test data 0.5595743834445313
6 r2 score of the train data 0.5563947957633792 and test data 0.6555906505571116
7 r2 score of the train data 0.553735574181428 and test data 0.7032605078583825
8 r2 score of the train data 0.5690729048532857 and test data 0.4700063619141459
9 r2 score of the train data 0.5620211189566908 and test data 0.5543351501847651
10 r2 score of the train data 0.5666264217016058 and test data 0.48986189262393154
11 r2 score of the train data 0.560688699269444 and test data 0.5770509865715794
12 r2 score of the train data 0.5561268382448219 and test data 0.6574833593687027
13 r2 score of the train data 0.5575377447394365 and test data 0.6320507247078625
14 r2 score of the train data 0.5558804535718433 and test data 0.6482600205565894
Avg r2 score of train 0.56211764812627 and test 0.5752115184377047
Time taken - 4.294117295742035 min
```

```
In [ ]: xg = XGBRFRegressor(colsample bylevel=1,
                        colsample bynode=0.8, colsample bytree=1, gamma=0,
                        learning rate=1, max delta step=0, max_depth=5,
                        max features=0.95, min child weight=1, min impurity decrease=1,
                        min samples leaf=1, min samples split=5, missing=None,
                        n estimators=100, n jobs=1, nthread=None, objective='reg:linear',
                        random state=0, reg alpha=0, reg lambda=1, scale pos weight=1,
                        seed=None. silent=True. subsample=0.8. verbositv=1)
         a = cross \ val \ score(xq, x2, y2, scoring= 'r2', cv= 10)
         print(f'R^2 Score {np.mean(a)} data' )
        R^2 Score 0.5718237374738755 data
       AdaBoost-Regressor
In [ ]:
         clf = AdaBoostRegressor()
         params = {'n estimators' : [100, 150, 200, ],
                   'learning rate' :[0.0001,0.001,0.01, 0.1],
                   'loss' : [ 'linear', 'square', 'exponential'],
                   'random state' : [10,20,30]
         kfold grid search(clf, params, x2, y2, fold = 10, kfold = 15, search = 'random')
        Fitting 10 folds for each of 10 candidates, totalling 100 fits
        [Parallel(n jobs=-1)]: Using backend LokyBackend with 2 concurrent workers.
        [Parallel(n jobs=-1)]: Done 46 tasks | elapsed: 6.3min
        [Parallel(n jobs=-1)]: Done 100 out of 100 | elapsed: 13.6min finished
        Best Estimator: AdaBoostRegressor(base estimator=None, learning rate=0.0001, loss='linear',
                          n estimators=100, random state=10)
        R2 metric :
        0 r2 score of the train data 0.5720113699608151 and test data 0.6158808626834753
        1 r2 score of the train data 0.5619763801409192 and test data 0.6355339181615801
        2 r2 score of the train data 0.5707625343770122 and test data 0.606265905792721
        3 r2 score of the train data 0.6058312584998451 and test data 0.3073106403664668
        4 r2 score of the train data 0.5721886323759073 and test data 0.5351559426964696
        5 r2 score of the train data 0.5744676443771899 and test data 0.5623683551296568
```

```
6 r2 score of the train data 0.5648681504239315 and test data 0.6552271007623669
        7 r2 score of the train data 0.5655959785551095 and test data 0.7061852214705656
        8 r2 score of the train data 0.580944156330872 and test data 0.4719427448074006
        9 r2 score of the train data 0.5714446221740298 and test data 0.5538494689733773
        10 r2 score of the train data 0.5756288720379016 and test data 0.49051296386692245
        11 r2 score of the train data 0.5737776034123998 and test data 0.5774840019937164
        12 r2 score of the train data 0.5684756314832162 and test data 0.6583991657352518
        13 r2 score of the train data 0.5664297487945196 and test data 0.6061219537121953
        14 r2 score of the train data 0.5647298013360682 and test data 0.6489853522620401
        Avg r2 score of train 0.5726088256186492 and test 0.5754149065609471
        Time taken - 15.350130430857341 min
In [ ]:
          ab = AdaBoostRegressor(base estimator=None, learning rate=0.0001, loss='linear',
                           n estimators=100, random state=20)
         a = cross val score(ab, x2, y2, scoring= 'r2', cv= 10)
         print(f'R^2 Score {np.mean(a)} data')
```

R^2 Score 0.5703529803771688 data

Gradient Boosting Regressor

init=None, learning rate=0.01, loss='huber',

```
min impurity decrease=0.0, min impurity split=None,
                                  min samples leaf=1, min samples split=2,
                                  min weight fraction leaf=0.0, n estimators=2000,
                                  n iter no change=11, presort='deprecated',
                                  random state=10, subsample=1.0, tol=0.0001,
                                  validation fraction=0.1, verbose=0, warm start=False)
        R2 metric :
        0 r2 score of the train data 0.5640270026713142 and test data 0.5844962160053929
        1 r2 score of the train data 0.5545558017178335 and test data 0.6263858956242915
        2 r2 score of the train data 0.5683858422857833 and test data 0.5882478867422942
        3 r2 score of the train data 0.6032325417831674 and test data 0.28827689355115715
        4 r2 score of the train data 0.5693710703909289 and test data 0.5268253421455653
        5 r2 score of the train data 0.5753977828991846 and test data 0.5509074520288433
        6 r2 score of the train data 0.5659724999522501 and test data 0.6529337483797
        7 r2 score of the train data 0.5692350047381489 and test data 0.7043729009907196
        8 r2 score of the train data 0.5827182335013722 and test data 0.4542972897883143
        9 r2 score of the train data 0.574540245256353 and test data 0.5319726953736295
        10 r2 score of the train data 0.5814156987703438 and test data 0.5028486395211187
        11 r2 score of the train data 0.5719466824608403 and test data 0.5601058443767275
        12 r2 score of the train data 0.56953918599406 and test data 0.6726943220669401
        13 r2 score of the train data 0.5654950604450828 and test data 0.6548945771237917
        14 r2 score of the train data 0.5668172693254028 and test data 0.6872046861726407
        Avg r2 score of train 0.572176661479471 and test 0.5724309593260751
        Time taken - 11.144120848178863 min
In [ ]:
         gb = GradientBoostingRegressor(alpha=0.9, ccp alpha=0.0, criterion='friedman mse',
                                   init=None, learning rate=0.01, loss='huber',
                                   max depth=3, max features=None, max leaf nodes=None,
                                   min impurity decrease=0.0, min impurity split=None,
                                   min samples leaf=1, min samples split=2,
                                   min weight fraction leaf=0.0, n estimators=2000,
                                   n iter no change=11, presort='deprecated',
                                   random state=10, subsample=1.0, tol=0.0001,
                                   validation fraction=0.1, verbose=0, warm start=False)
         a = cross val score(gb, x2, y2, scoring= 'r2', cv= 10)
         print(f'R^2 Score {np.mean(a)} data')
        R^2 Score 0.5696356812530452 data
```

max depth=3, max features=None, max leaf nodes=None,

LGBM Regressor

```
In [ ]:
         clf = LGBMRegressor()
         params = {'min child samples' : [10, 20,50],
                   'num leaves' : [5,6],
                   'max depth' : [2, 3, 5],
                   'n estimators' : [1000,2000,4000,5000],
                   'learning rate' : [0.0001,0.001,0.01,0.1]
         kfold grid search(clf, params, x2, y2.ravel(), fold = 10, kfold = 15, search = \frac{1}{2} random') #0.6920, 0.6622
        Fitting 10 folds for each of 10 candidates, totalling 100 fits
        [Parallel(n jobs=-1)]: Using backend LokyBackend with 2 concurrent workers.
        [Parallel(n jobs=-1)]: Done 46 tasks
                                                  | elapsed: 4.7min
        [Parallel(n jobs=-1)]: Done 100 out of 100 | elapsed: 8.9min finished
        Best Estimator : LGBMRegressor(boosting type='gbdt', class weight=None, colsample bytree=1.0,
                      importance type='split', learning rate=0.01, max depth=3,
                      min child samples=50, min child weight=0.001, min split gain=0.0,
                      n estimators=4000, n jobs=-1, num leaves=5, objective=None,
                      random state=None, reg alpha=0.0, reg lambda=0.0, silent=True,
                      subsample=1.0, subsample for bin=200000, subsample freq=0)
        R2 metric :
        0 r2 score of the train data 0.6539376054229651 and test data 0.5745765027557168
        1 r2 score of the train data 0.6479659676917198 and test data 0.6541898772014111
        2 r2 score of the train data 0.6553573129657073 and test data 0.5877012528118586
        3 r2 score of the train data 0.6804781717894042 and test data 0.2926826947679162
        4 r2 score of the train data 0.6577511726844416 and test data 0.5400561398326873
        5 r2 score of the train data 0.6563520755321524 and test data 0.5388823520975055
        6 r2 score of the train data 0.6502514789337726 and test data 0.6091735373791863
        7 r2 score of the train data 0.6476857317677991 and test data 0.6896003172617456
        8 r2 score of the train data 0.6622299194857179 and test data 0.47845027538518325
        9 r2 score of the train data 0.6614143575698203 and test data 0.5262909902666517
        10 r2 score of the train data 0.6610274429566736 and test data 0.5078239590939534
        11 r2 score of the train data 0.6547838989872861 and test data 0.5718646213126627
        12 r2 score of the train data 0.6524788377378758 and test data 0.6331363491047368
        13 r2 score of the train data 0.6494901652393692 and test data 0.6363914114588893
        14 r2 score of the train data 0.6552123606829265 and test data 0.6434023316558577
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```

```
max depth=5, max features=0.95, max leaf nodes=None,
                      max samples=None, min impurity decrease=0.001,
                      min impurity split=None, min samples leaf=2,
                      min samples split=8, min weight fraction leaf=0.0,
                      n estimators=70, n jobs=None, oob score=False,
                      random state=None, verbose=0, warm start=False)),
                ('xg', XGBRFRegressor(colsample bylevel=1,
               colsample bynode=0.8, colsample bytree=1, gamma=0,
               learning rate=1, max delta step=0, max depth=5,
               max features=0.95, min child weight=1, min impurity decrease=1,
               min samples leaf=1, min samples split=5, missing=None,
               n estimators=100, n jobs=1, nthread=None, objective='req:linear',
               random state=0, reg alpha=0, reg lambda=1, scale pos weight=1,
               seed=None, silent=True, subsample=0.8, verbosity=1)),
              ('gb', GradientBoostingRegressor(max depth= 2, learning rate= 0.1, random state= 10,n estimators=5000,
                                n iter no change = 11))
stack = StackingRegressor(estimators= estimators,
                          final estimator= Ridge(),
```

```
cv_score = cross_val_score(stack, x2, y2.ravel(), scoring='r2', cv=10, verbose=5, n_jobs=-1)
print('Mean Score:',cv_score.mean())
print('Standard Deviation:',cv_score.std())

[Parallel(n_jobs=-1)]: Using backend LokyBackend with 2 concurrent workers.
Mean Score: 0.5762229576080763
Standard Deviation: 0.08756604747415982

[Parallel(n_jobs=-1)]: Done 10 out of 10 | elapsed: 12.2min remaining: 0.0s
[Parallel(n_jobs=-1)]: Done 10 out of 10 | elapsed: 12.2min finished
```

Models fitting

R² Score

```
'PCA(5) + BINARY + LABEL + SYNTHETIC FEATURES'])
tb.add_column(col[2], ['0.5211', '0.5808', '0.5752', '0.5754', '0.5724', '0.5661', '0.5762'])
tb.add column(col[3], ['0.55187', '0.55327', '0.54654', '0.54911', '0.54427', '0.54420', '0.55610'])
tb.add column(col[4], ['0.54823', '0.54488', '0.54390', '0.54551', '0.53709', '0.52651', '0.54816'])
print(tb)
      MODELS
                                  FEATURES
                                                          | CROSS-VALIDATION | PUBLIC SCORE | PRIVATE SCORE
                   | PCA(5) + BINARY + LABEL + SYNTHETIC FEATURES |
  DECISION TREE
                                                                 0.5211
                                                                               0.55187
                                                                                             0.54823
  RANDOM FOREST
               | PCA(5) + BINARY + LABEL + SYNTHETIC FEATURES |
                                                                 0.5808
                                                                               0.55327
                                                                                             0.54488
     XGB00ST
                 | PCA(5) + BINARY + LABEL + SYNTHETIC FEATURES |
                                                                 0.5752
                                                                               0.54654
                                                                                             0.54390
                 | PCA(5) + BINARY + LABEL + SYNTHETIC FEATURES |
    ADA-BOOST
                                                                 0.5754
                                                                               0.54911
                                                                                             0.54551
 GRADIENT BOOSTING | PCA(5) + BINARY + LABEL + SYNTHETIC FEATURES |
                                                                 0.5724
                                                                               0.54427
                                                                                             0.53709
    LIGHT GBM
                 | PCA(5) + BINARY + LABEL + SYNTHETIC FEATURES |
                                                                 0.5661
                                                                               0.54420
                                                                                             0.52651
 STCKED-ENSEMBLE | PCA(5) + BINARY + LABEL + SYNTHETIC FEATURES |
                                                                0.5762
                                                                               0.55610
                                                                                             0.54816
```

OBSERVATION

- This data set created like using PCA method and synthetic features.
- Using PCA method created 5 features and created 22 synthetic features like difference between twi feature and ratio of another features.
- In these models use 10 k fold cross validation and grid or random search cross validation for the best parameters.

- We can see that above table random forest well performe in the cross validation and also good perform in kaggle public score.
- Stacked ensemble got higher score in public and private score.

Models with PCA + synthetic + Binary features and y target value clip 150sec

```
In [9]:
          # Import the train and test csv files
          # Import the train and test csv files
          train = pd.read csv('C:/Users/Dell/Python/Python/Python AAIC File/Assignments/CASE-STUDY ML/mercedes-benz-greener-mar
          test = pd.read csv('C:/Users/Dell/Python/Python/Python AAIC File/Assignments/CASE-STUDY ML/mercedes-benz-greener-manu
          train = train.drop(['Unnamed: 0'], axis = 1)
          test = test.drop(['Unnamed: 0'], axis = 1)
          train.head()
                   y X0 X1 X2 X3 X4 X5 X6 X8 ... X46_263_119_261 X136_118_136_60 qua_encode_1 qua_encode_2 qua_encode_3 qua_encode
Out[9]:
           0 130.81
                                                                               1.0
                                                                                       6.901179
                                                                                                   80.729628
                                                                                                               14.457824
                       k v at a d
                                                                1.0
                                                                                      28.816796
                                                                0.5
                                                                               1.0
                                                                                                   7.221336
                          t av
                                                                                                               13.855619
                                                                                     567.801379
                                                                                                   9.717606
            7 76.26 az w
                            n c d x j x ...
                                                                1.0
                                                                               0.0
                                                                                                              935.797419
                80.62 az
                                                                1.0
                                                                               0.0
                                                                                     674.842909
                                                                                                   10.394944
                                                                                                              930.330382
         4 13 78.02 az v n f d h d n ...
                                                                1.0
                                                                               0.0
                                                                                     642.841100
                                                                                                  250.973811
                                                                                                              963.246231
        5 rows × 404 columns
In [10]:
          #Check the Shape of the test and train data
          print('train data shape - ', train.shape)
          print('test data shape - ', test.shape)
         train data shape - (4209, 404)
```

```
test data shape - (4209, 403)
In [11]:
          train.loc[train['y'] > 150] = 150
In [12]:
          v2 = train.v
          x2 = train.iloc[:, 10:]
          # Split train data into train-set and test-set
          x train2, x test2, y train2, y test2 = train test split(x2,y2, test size = 0.33, random state= 42)
          print(x train2.shape, y train2.shape)
          print(x test2.shape, y test2.shape)
         (2820, 394) (2820,)
         (1389, 394) (1389,)
        Decision Tree
In [ ]:
          clf = DecisionTreeRegressor(ccp alpha=0.0, criterion='mse', max depth=3,
                                      max features='auto', max leaf nodes=None,
                                      min impurity decrease=0.0, min impurity split=None,
                                      min samples leaf=1, min samples split=2,
                                      min weight fraction leaf=0.0, presort='deprecated',
                                      random state=None, splitter='best')
          params = \{ \text{'max depth'} : [2,3,4,8,10,15], 
                    'max features' : ['auto', 'sqrt', 'log2'],
                    'random state' : [5,10,20,30],
          kfold grid search(clf, params, x2, y2, 10, kfold = 15, search= 'random')
         Fitting 10 folds for each of 10 candidates, totalling 100 fits
         [Parallel(n jobs=-1)]: Using backend LokyBackend with 2 concurrent workers.
         [Parallel(n jobs=-1)]: Done 62 tasks
                                                     | elapsed:
                                                                  4.3s
         [Parallel(n jobs=-1)]: Done 100 out of 100 | elapsed: 5.6s finished
         Best Estimator: DecisionTreeRegressor(ccp alpha=0.0, criterion='mse', max depth=3,
                               max features='auto', max leaf nodes=None,
```

```
min weight fraction leaf=0.0, presort='deprecated',
                              random state=5, splitter='best')
        R2 metric :
        0 r2 score of the train data 0.6480538816132388 and test data 0.6693714310012511
        1 r2 score of the train data 0.642465953902116 and test data 0.7702358085844839
        2 r2 score of the train data 0.6533336168366211 and test data 0.6034552624278658
        3 r2 score of the train data 0.650727064754295 and test data 0.626995114554626
        4 r2 score of the train data 0.6481815912325416 and test data 0.6753606349895027
        5 r2 score of the train data 0.6495238656607814 and test data 0.6594321202166429
        6 r2 score of the train data 0.6496479097970977 and test data 0.6594673475081894
        7 r2 score of the train data 0.6440153039529424 and test data 0.752330200943577
        8 r2 score of the train data 0.6627475307237207 and test data 0.4674067355568491
        9 r2 score of the train data 0.6516848915900538 and test data 0.6280380207465941
        10 r2 score of the train data 0.6597386061652126 and test data 0.5142825155435712
        11 r2 score of the train data 0.6504902324792055 and test data 0.6472108509331767
        12 r2 score of the train data 0.649356510657001 and test data 0.6486385218504416
        13 r2 score of the train data 0.6473059031423953 and test data 0.669704349641284
        14 r2 score of the train data 0.6490679486603913 and test data 0.6551988155854951
        Avg r2 score of train 0.6504227207445076 and test 0.6431418486722366
        Time taken - 0.11364267269770305 min
In [ ]:
         dt = DecisionTreeRegressor(ccp alpha=0.0, criterion='mse', max depth=3,
                               max features='auto', max leaf nodes=None,
                               min impurity decrease=0.0, min impurity split=None,
                               min samples leaf=1, min samples split=2,
                               min weight fraction leaf=0.0, presort='deprecated',
                               random state=5, splitter='best')
         a = cross \ val \ score(dt, x2, y2, scoring= 'r2', cv= 5)
         print(f'R^2 Score {np.mean(a)} data' )
        R^2 Score 0.6355296996079519 data
```

min impurity decrease=0.0, min impurity split=None,

min samples leaf=1, min samples split=2,

AdaBoost-Regressor

```
In [ ]: clf = AdaBoostRegressor(base_estimator=None, learning_rate=0.0001,
```

```
n estimators=300, random state=None, loss= 'linear')
         params = {'n_estimators' : [100, 150, 200, ],
                   'learning rate' :[0.0001,0.001,0.01, 0.1],
                   'loss' : [ 'linear', 'square', 'exponential'],
                   'random state' : [10,20,30]
         kfold grid search(clf, params, x2, y2, fold = 10, kfold = 15, search = 'random')
        Fitting 10 folds for each of 10 candidates, totalling 100 fits
        [Parallel(n jobs=-1)]: Using backend LokyBackend with 2 concurrent workers.
        [Parallel(n jobs=-1)]: Done 46 tasks | elapsed: 6.5min
        [Parallel(n jobs=-1)]: Done 100 out of 100 | elapsed: 14.6min finished
        Best Estimator: AdaBoostRegressor(base estimator=None, learning rate=0.0001, loss='exponential',
                          n estimators=200, random state=10)
        R2 metric :
        0 r2 score of the train data 0.6477109410342181 and test data 0.6667582180352138
        1 r2 score of the train data 0.6411183093664439 and test data 0.769834809588402
        2 r2 score of the train data 0.6519445325653355 and test data 0.6035245529110724
        3 r2 score of the train data 0.6501761962040864 and test data 0.6403407813302944
        4 r2 score of the train data 0.6467394871648264 and test data 0.6752756035497736
        5 r2 score of the train data 0.6482150084579972 and test data 0.6593962447784378
        6 r2 score of the train data 0.6483409079900484 and test data 0.6595427982210769
        7 r2 score of the train data 0.6427737983809947 and test data 0.7524045226988414
        8 r2 score of the train data 0.6612296144447891 and test data 0.4693381380158933
        9 r2 score of the train data 0.6503698925698844 and test data 0.6281263795271057
        10 r2 score of the train data 0.656219322638109 and test data 0.489240343280833
        11 r2 score of the train data 0.649137793025413 and test data 0.6470255138363956
        12 r2 score of the train data 0.6479060792497031 and test data 0.6619152500286267
        13 r2 score of the train data 0.6458712207158706 and test data 0.66957122946993
        14 r2 score of the train data 0.6476381814013082 and test data 0.6553517810957608
        Avg r2 score of train 0.6490260856806018 and test 0.6431764110911772
        Time taken - 18.583150271574656 min
In [ ]:
          ab = AdaBoostRegressor(base_estimator=None, learning_rate=0.0001, loss='exponential',
                           n estimators=200, random state=10)
```

```
a = cross_val_score(ab, x2, y2, scoring= 'r2', cv= 10)
print(f'R^2 Score {np.mean(a)} data' )
```

R^2 Score 0.6424381050598855 data

Gradient Boosting Regressor

```
In [ ]:
         clf = GradientBoostingRegressor(alpha=0.9, ccp alpha=0.0, criterion='friedman mse',
                                   init=None, learning rate=0.01, loss='huber',
                                   max depth=3, max features=None, max leaf nodes=None,
                                   min impurity decrease=0.0, min impurity split=None,
                                   min samples leaf=1, min samples split=2,
                                   min weight fraction leaf=0.0, n estimators=800,
                                   n iter no change=11, presort='deprecated',
                                   random state=10, subsample=1.0, tol=0.0001,
                                   validation fraction=0.1, verbose=0, warm start=False)
         params = {'n estimators' : [500,800,1000, 1500, 2000],
                   'loss' : [ 'huber', 'exponential'],
                   'learning rate' : [0.01, 0.01, 0.1],
                   'max depth' : [3,4,5,7]
         kfold grid search(clf, params, x2, y2, fold = 10, kfold = 15, search = 'random')
        Fitting 10 folds for each of 10 candidates, totalling 100 fits
        [Parallel(n jobs=-1)]: Using backend LokyBackend with 2 concurrent workers.
        [Parallel(n jobs=-1)]: Done 74 tasks
                                                | elapsed: 12.7min
        [Parallel(n jobs=-1)]: Done 100 out of 100 | elapsed: 16.7min finished
        Best Estimator: GradientBoostingRegressor(alpha=0.9, ccp alpha=0.0, criterion='friedman mse',
                                  init=None, learning rate=0.01, loss='huber',
                                  max depth=3, max features=None, max leaf nodes=None,
                                  min impurity decrease=0.0, min impurity split=None,
                                  min samples leaf=1, min samples split=2,
                                  min weight fraction leaf=0.0, n estimators=1000,
                                  n iter no change=11, presort='deprecated',
                                  random state=10, subsample=1.0, tol=0.0001,
                                  validation fraction=0.1, verbose=0, warm start=False)
```

```
R2 metric :
        0 r2 score of the train data 0.6440326836974183 and test data 0.6493814982039965
        1 r2 score of the train data 0.6447564492798792 and test data 0.7714889267681285
        2 r2 score of the train data 0.6556808940577739 and test data 0.5873595530545905
        3 r2 score of the train data 0.6602751070405202 and test data 0.6296922521241193
        4 r2 score of the train data 0.6567517205050433 and test data 0.6810093083229571
        5 r2 score of the train data 0.6438252314922404 and test data 0.641228026796587
        6 r2 score of the train data 0.6528491710955487 and test data 0.6555477858191263
        7 r2 score of the train data 0.643880177692111 and test data 0.7560991766099738
        8 r2 score of the train data 0.6759464362526924 and test data 0.4576462792867908
        9 r2 score of the train data 0.6641764492261998 and test data 0.6133198511826266
        10 r2 score of the train data 0.6675695377306117 and test data 0.49966612497125706
        11 r2 score of the train data 0.6580900059344754 and test data 0.6316905816914543
        12 r2 score of the train data 0.6539165032379398 and test data 0.6742449715204522
        13 r2 score of the train data 0.6514083980245011 and test data 0.7091407057621535
        14 r2 score of the train data 0.6594997815908921 and test data 0.6877275468300248
        Avg r2 score of train 0.6555105697905231 and test 0.6430161725962826
        Time taken - 21.6081263701121 min
In [ ]:
         gb = GradientBoostingRegressor(alpha=0.9, ccp alpha=0.0, criterion='friedman mse',
                                   init=None, learning rate=0.01, loss='huber',
                                   max depth=3, max features=None, max leaf nodes=None,
                                   min impurity decrease=0.0, min impurity split=None,
                                   min samples leaf=1, min samples split=2,
                                   min weight fraction leaf=0.0, n estimators=1000,
                                   n iter no change=11, presort='deprecated',
                                   random state=10, subsample=1.0, tol=0.0001,
                                   validation fraction=0.1, verbose=0, warm start=False)
         a = cross val score(qb, x2, y2, scoring= 'r2', cv= 10)
         print(f'R^2 Score {np.mean(a)} data')
```

R^2 Score 0.640877235785154 data

Random Forest

```
min samples split=8, min weight fraction leaf=0.0,
                       n estimators=70, n jobs=None, oob score=False,
                       random state=None, verbose=0, warm start=False)
 params = \{'n \text{ estimators}': [40,50,60,70,100],
              'max depth':[3,5,6,7,8],
              'min samples split':[2,3,4,5,6,7,8,9,10],
              'max features': [0.80,.95, 1.0],
              'min samples leaf': [1, 2,3,4,5,6,7,8,9],
              'min impurity decrease': [1e-5,1e-4,1e-3,1e-2,1e-1,0,1,10]}
 kfold grid search(clf, params, x2, y2, fold = 10, kfold=15, search= 'random')
Fitting 10 folds for each of 10 candidates, totalling 100 fits
[Parallel(n jobs=-1)]: Using backend LokyBackend with 2 concurrent workers.
[Parallel(n jobs=-1)]: Done 46 tasks | elapsed: 1.9min
[Parallel(n jobs=-1)]: Done 100 out of 100 | elapsed: 3.6min finished
Best Estimator: RandomForestRegressor(bootstrap=True, ccp alpha=0.0, criterion='mse',
                      max depth=6, max features=1.0, max leaf nodes=None,
                      max samples=None, min impurity decrease=0.1,
                      min impurity split=None, min samples leaf=6,
                      min samples split=2, min weight fraction leaf=0.0,
                      n estimators=70, n jobs=None, oob score=False,
                      random state=None, verbose=0, warm start=False)
R2 metric :
0 r2 score of the train data 0.6761806107128834 and test data 0.6573744354284433
1 r2 score of the train data 0.6690398809859387 and test data 0.7684369053571742
2 r2 score of the train data 0.6804516359221022 and test data 0.5979577019930118
3 r2 score of the train data 0.6738827442338753 and test data 0.6428903435953389
4 r2 score of the train data 0.6723676756508512 and test data 0.6868064575030692
5 r2 score of the train data 0.6752737867274246 and test data 0.6627912033311976
6 r2 score of the train data 0.6748901863798457 and test data 0.6620421469359128
7 r2 score of the train data 0.670449835669485 and test data 0.7515754444268572
8 r2 score of the train data 0.6894411286450259 and test data 0.47322601756498917
9 r2 score of the train data 0.6776469544688636 and test data 0.6343698165475739
10 r2 score of the train data 0.6835132636421446 and test data 0.516588880256428
11 r2 score of the train data 0.676175147732486 and test data 0.6506305811277528
12 r2 score of the train data 0.673793912539155 and test data 0.6604770020924078
13 r2 score of the train data 0.6715570812360592 and test data 0.705112780339884
14 r2 score of the train data 0.6744616810048314 and test data 0.6628970220046523
```

Avg r2_score of train 0.6759417017033982 and test 0.648878449233646 Time taken - 4.554728877544403 min

R^2 Score 0.6457039283917678 data

LGBM Regressor

```
In [ ]:
         clf = LGBMRegressor(boosting type='gbdt', class weight=None, colsample_bytree=1.0,
                       importance type='split', learning rate=0.01, max depth=5,
                       min child samples=50, min child weight=0.001, min split gain=0.0,
                       n estimators=1000, n jobs=-1, num leaves=5, objective=None,
                       random state=None, reg alpha=0.0, reg lambda=0.0, silent=True,
                       subsample=1.0, subsample for bin=200000, subsample freq=0)
         params = {'min child samples' : [10, 20,50],
                   'num leaves' : [5,6],
                   'max depth' : [2, 3, 5],
                   'n estimators' : [1000,2000,4000,5000],
                   'learning rate' : [0.0001,0.001,0.01,0.1]
         kfold grid search(clf, params, x2, y2.ravel(), fold = 10, kfold = 15, search = 'random')
        Fitting 10 folds for each of 10 candidates, totalling 100 fits
        [Parallel(n jobs=-1)]: Using backend LokyBackend with 2 concurrent workers.
        [Parallel(n jobs=-1)]: Done 46 tasks
                                                  | elapsed: 3.9min
        [Parallel(n jobs=-1)]: Done 100 out of 100 | elapsed: 8.6min finished
```

```
Best Estimator : LGBMRegressor(boosting type='gbdt', class weight=None, colsample bytree=1.0,
                      importance type='split', learning rate=0.001, max depth=3,
                      min child samples=20, min child weight=0.001, min split gain=0.0,
                      n estimators=5000, n jobs=-1, num leaves=6, objective=None,
                      random state=None, reg alpha=0.0, reg lambda=0.0, silent=True,
                      subsample=1.0, subsample for bin=200000, subsample freq=0)
        R2 metric :
        0 r2 score of the train data 0.6646104803919637 and test data 0.6630022586838633
        1 r2 score of the train data 0.6583630541223624 and test data 0.7675817081799015
        2 r2 score of the train data 0.6690928334118285 and test data 0.6058607904596616
        3 r2 score of the train data 0.6649248016084909 and test data 0.6436142891192629
        4 r2 score of the train data 0.6624097369466116 and test data 0.6936298300139243
        5 r2 score of the train data 0.6646374197084737 and test data 0.665528759256929
        6 r2 score of the train data 0.6650009587077517 and test data 0.6631473971131017
        7 r2 score of the train data 0.6604176474827126 and test data 0.7501455280855138
        8 r2 score of the train data 0.6798367687788365 and test data 0.47817342733889345
        9 r2 score of the train data 0.666810659145471 and test data 0.6324631406650509
        10 r2 score of the train data 0.6751084633746813 and test data 0.5094524762619161
        11 r2 score of the train data 0.6656215129264706 and test data 0.6408139966621724
        12 r2 score of the train data 0.6642275397074902 and test data 0.6602457548831437
        13 r2 score of the train data 0.6622259856150257 and test data 0.7103376271877867
        14 r2 score of the train data 0.6632183999773167 and test data 0.6660837115764423
        Avg r2 score of train 0.6657670841270324 and test 0.650005379699171
        Time taken - 13.400730057557423 min
In [ ]:
         lb = LGBMRegressor(boosting type='gbdt', class weight=None, colsample bytree=1.0,
                       importance type='split', learning rate=0.001, max depth=3,
                       min child samples=20, min child weight=0.001, min split gain=0.0,
                       n estimators=5000, n jobs=-1, num leaves=6, objective=None,
                       random state=None, reg alpha=0.0, reg lambda=0.0, silent=True,
                       subsample=1.0, subsample for bin=200000, subsample freq=0)
         a = cross val score(lb, x2, y2, scoring= 'r2', cv= 10)
         print(f'R^2 Score {np.mean(a)} data')
```

R^2 Score 0.6468964578401072 data

XGBoost-Regressor

```
In [ ]: clf = XGBRFRegressor(colsample bylevel=1,colsample bynode=0.8, colsample bytree=1, gamma=0,
                        learning rate=1, max delta step=0, max depth=5,
                        max features=0.95, min child weight=1, min impurity decrease=1,
                        min samples leaf=1, min samples split=5, missing=None,
                        n estimators=100, n jobs=1, nthread=None, objective='reg:linear',
                        random state=0, reg alpha=0, reg lambda=1, scale pos weight=1,
                        seed=None, silent=True, subsample=0.8, verbosity=1)
         xparams = {'learning rate':[0.1,0.5,0.8,1],
                      'n estimators':[70,80,100],
                      'max depth':[2,3,4],
                      'colsample bytree': [0.1,0.5,0.7,0.9,1],
                      'subsample': [0.2,0.3,0.5,1],
                      'gamma': [0.0001,0.001,0,0.1,0.01,0.5,1],
                      'reg alpha':[0.00001,0.0001,0.001,0.01,0.1]}
         kfold grid search(clf, params, x2, y2, fold = 10, kfold=15, search= False)
        R2 metric :
        0 r2 score of the train data 0.6066547132057518 and test data 0.63525224317498
        1 r2 score of the train data 0.6034804178965002 and test data 0.6963600576690997
        2 r2 score of the train data 0.6105771133880324 and test data 0.601203463972948
        3 r2 score of the train data 0.6128226383721311 and test data 0.5564458465086671
        4 r2 score of the train data 0.6111184302393099 and test data 0.5830068184166133
        5 r2 score of the train data 0.6082304797363157 and test data 0.6281456387239466
        6 r2 score of the train data 0.6089669154327122 and test data 0.6561857533878476
        7 r2 score of the train data 0.6035336362444628 and test data 0.7099815395322746
        8 r2 score of the train data 0.6238022287495313 and test data 0.4687577983273177
        9 r2 score of the train data 0.6125290949699321 and test data 0.5628813136358713
        10 r2 score of the train data 0.6200341125828022 and test data 0.48810031559351164
        11 r2 score of the train data 0.6088743143552269 and test data 0.6168210561799822
        12 r2 score of the train data 0.6083074897370442 and test data 0.658639519541895
        13 r2 score of the train data 0.6073444564835199 and test data 0.6416319396499979
        14 r2 score of the train data 0.607225555256518 and test data 0.6514214996312307
        Avg r2 score of train 0.6102334397766528 and test 0.6103223202630789
        Time taken - 0.7519578218460083 min
In [ ]:
          xg = XGBRFRegressor(colsample bylevel=1,
                        colsample bynode=0.8, colsample bytree=1, gamma=0,
                        learning rate=1, max delta step=0, max depth=5,
```

```
max_features=0.95, min_child_weight=1, min_impurity_decrease=1,
    min_samples_leaf=1, min_samples_split=5, missing=None,
    n_estimators=100, n_jobs=1, nthread=None, objective='reg:linear',
    random_state=0, reg_alpha=0, reg_lambda=1, scale_pos_weight=1,
    seed=None, silent=True, subsample=0.8, verbosity=1)

a = cross_val_score(xg, x2, y2, scoring= 'r2', cv= 10)
    print(f'R^2 Score {np.mean(a)} data')
```

R^2 Score 0.6084653791278974 data

Stack ensemble

```
In [14]:
          estimators = [ ('rf', RandomForestRegressor(bootstrap=True, ccp alpha=0.0, criterion='mse',
                                max depth=5, max features=0.95, max leaf nodes=None,
                                max samples=None, min impurity decrease=0.001,
                                min impurity split=None, min samples leaf=2,
                                min samples split=8, min weight fraction leaf=0.0,
                                n estimators=70, n jobs=None, oob score=False,
                                random state=None, verbose=0, warm start=False)),
                          ('xg', XGBRFRegressor(colsample bylevel=1,
                         colsample bynode=0.8, colsample bytree=1, gamma=0,
                         learning rate=1, max delta step=0, max depth=5,
                         max features=0.95, min child weight=1, min_impurity_decrease=1,
                         min samples leaf=1, min samples split=5, missing=None,
                         n estimators=100, n jobs=1, nthread=None, objective='reg:linear',
                         random state=0, reg alpha=0, reg lambda=1, scale pos weight=1,
                         seed=None, silent=True, subsample=0.8, verbosity=0)),
                        ('lg', LGBMRegressor(boosting type='gbdt', class weight=None, colsample bytree=1.0,
                        importance type='split', learning rate=0.001, max depth=3,
                        min child samples=20, min child weight=0.001, min split gain=0.0,
                        n estimators=5000, n jobs=-1, num leaves=6, objective=None,
                        random state=None, reg alpha=0.0, reg lambda=0.0, silent=True,
                        subsample=1.0, subsample for bin=200000, subsample freq=0))
```

```
stack = StackingRegressor(estimators= estimators,
                                    final estimator= Ridge(),
In [ ]:
          cv score = cross val score(stack, x2, y2.ravel(), scoring='r2', cv= 5, verbose=5, n jobs=-1)
          print('Mean Score:',cv score.mean())
          print('Standard Deviation:',cv score.std()) #0.6404, 0.64200,
         [Parallel(n jobs=-1)]: Using backend LokyBackend with 2 concurrent workers.
         Mean Score: 0.6412507934357022
         Standard Deviation: 0.030613234929429335
         [Parallel(n jobs=-1)]: Done 5 out of 5 | elapsed: 22.7min finished
In [17]:
          stack = stack.fit(x2.values, y2)
In [18]:
          joblib.dump(stack, 'stack model.joblib.dat')
          print('Model saved in disk!')
         Model saved in disk!
```

Models fitting

R^2 Score

```
In [56]:
          col = ['MODELS', 'FEATURES', 'CROSS-VALIDATION', 'PUBLIC SCORE', 'PRIVATE SCORE']
          tb = PrettyTable()
          tb.add column(col[0], ['DECISION TREE', 'RANDOM FOREST', 'XGBOOST', 'ADA-BOOST', 'GRADIENT BOOSTING', 'LIGHT GBM', 'S
          tb.add column(col[1], ['PCA(5) + BINARY + LABEL + Y 150 CLIP +SYNTHETIC FEATURES', 'PCA(5) + BINARY + LABEL + Y 150 CL
                                 'PCA(5) + BINARY + LABEL + Y 150 CLIP +SYNTHETIC FEATURES', 'PCA(5) + BINARY + LABEL + Y 150 CL
                                 'PCA(5) + BINARY + LABEL + Y 150 CLIP +SYNTHETIC FEATURES', 'PCA(5) + BINARY + LABEL + Y 150 CL
                                 'PCA(5) + BINARY + LABEL + Y 150 CLIP +SYNTHETIC FEATURES',])
          tb.add column(col[2], ['0.6451', '0.6488', '0.6102', '0.6431', '0.6430', '0.6500', '0.6412'])
          tb.add column(col[3], ['0.53938', '0.55102', '0.54616', '0.53998', '0.54294', '0.55541', '0.55514'])
          tb.add column(col[4], ['0.53968', '0.54857', '0.53822', '0.53963', '0.53637', '0.54895', '0.54951'])
          print(tb)
                 MODELS
                                                       FEATURES
                                                                                         | CROSS-VALIDATION | PUBLIC SCORE | PR
         IVATE SCORE |
             DECISION TREE
                             | PCA(5) + BINARY + LABEL + Y 150 CLIP +SYNTHETIC FEATURES |
                                                                                                0.6451
                                                                                                                0.53938
         0.53968
                             | PCA(5) + BINARY + LABEL + Y 150 CLIP +SYNTHETIC FEATURES |
                                                                                                0.6488
             RANDOM FOREST
                                                                                                                0.55102
         0.54857
                XGB00ST
                             | PCA(5) + BINARY + LABEL + Y 150 CLIP +SYNTHETIC FEATURES |
                                                                                                0.6102
                                                                                                                0.54616
         0.53822
               ADA-BOOST
                             | PCA(5) + BINARY + LABEL + Y 150 CLIP +SYNTHETIC FEATURES |
                                                                                                0.6431
                                                                                                                0.53998
         0.53963
           GRADIENT BOOSTING | PCA(5) + BINARY + LABEL + Y 150 CLIP +SYNTHETIC FEATURES |
                                                                                                0.6430
                                                                                                                0.54294
         0.53637
               LIGHT GBM
                              PCA(5) + BINARY + LABEL + Y 150 CLIP +SYNTHETIC FEATURES |
                                                                                                0.6500
                                                                                                                0.55541
         0.54895
           STCKED-ENSEMBLE
                             | PCA(5) + BINARY + LABEL + Y 150 CLIP +SYNTHETIC FEATURES |
                                                                                                0.6412
                                                                                                                0.55514
         0.54951
```

+-----+

OBSERVATION

- This dataset contain 5 PCA feature, binary features, label encoded features and synthetic features.
- This dataset with some experiment found that when we clip our y target with threshold 150secs then we got good score.
- We got good score in this data higher than another datasets, we see that in above table stacked ensemble get best score in public 0.55514 and private 0.54951 score.
- All models well perform in cross validation sets with k fold method.

Model with important feature of models (synethic features + 100 PCA + Binary)

```
In [11]:
          # Import the train and test csv files
          train = pd.read csv(r'/content/drive/MyDrive/top feature train.csv')
          test = pd.read csv(r'/content/drive/MyDrive/top feature test.csv')
          train = train.drop(['Unnamed: 0'], axis = 1)
          test = test.drop(['Unnamed: 0'], axis = 1)
          train.head()
Out[11]:
            ID X0 X1 X2 X3 X5 X6 X8 X10 X12 X13 X14 X19 X20 X22 X23 X27 X28 X29 X31 X32 X35 X37 X38 X41 X43 X44 X45 X
                                                                                                                           0
                      at
                                                      0
                                                          0
                                                                                                                           0
                                                                                                                           0
                                                      0
                                                          0
```

```
In [12]:
         #Check the Shape of the test and train data
         print('train data shape - ', train.shape)
         print('test data shape - ', test.shape)
         train data shape - (4209, 350)
         test data shape - (4209, 349)
In [13]:
         train.loc[train['y'] > 150] = 150
         y2 = train.y
                        'ID', 'X0', 'X1', 'X2', 'X3','X5', 'X6', 'X8', 'y']
         move = [
         x2 = train.drop(move, axis= 1)
In [14]:
         # Split train data into train-set and test-set
         x train2, x test2, y train2, y test2 = train test split(x2,y2, test size = 0.33, random state= 42)
         print(x train2.shape, y train2.shape)
         print(x test2.shape, y test2.shape)
         (2820, 341) (2820,)
         (1389, 341) (1389,)
        Decision Tree
```

```
kfold grid search(clf, params, x2, y2, 10, kfold = 15, search= 'grid')
         Fitting 10 folds for each of 72 candidates, totalling 720 fits
         [Parallel(n jobs=-1)]: Using backend LokyBackend with 2 concurrent workers.
         [Parallel(n jobs=-1)]: Done 46 tasks
                                                    | elapsed:
                                                                  4.0s
         [Parallel(n jobs=-1)]: Done 466 tasks
                                                      elapsed:
                                                                 33.95
         [Parallel(n jobs=-1)]: Done 717 out of 720 | elapsed: 1.3min remaining:
                                                                                     0.3s
         [Parallel(n jobs=-1)]: Done 720 out of 720 | elapsed: 1.3min finished
         Best Estimator: DecisionTreeRegressor(ccp alpha=0.0, criterion='mse', max depth=3,
                               max features='auto', max leaf nodes=None,
                               min impurity decrease=0.0, min impurity split=None,
                               min samples leaf=1, min samples split=2,
                               min weight fraction leaf=0.0, presort='deprecated',
                               random state=20, splitter='best')
         R2 metric :
         0 r2 score of the train data 0.5727993448261308 and test data 0.5517333068615851
         1 r2 score of the train data 0.5653722857340067 and test data 0.6649931070125705
         2 r2 score of the train data 0.5734600812347554 and test data 0.6037041847200402
         3 r2 score of the train data 0.6079480564620057 and test data 0.295841387440966
         4 r2 score of the train data 0.5780829128488911 and test data 0.5094437738972812
         5 r2 score of the train data 0.5770455134950616 and test data 0.5546179692984541
         6 r2 score of the train data 0.5714621388760321 and test data 0.63931768707822
         7 r2 score of the train data 0.5679556024559522 and test data 0.7047013902313545
         8 r2 score of the train data 0.58480623748573 and test data 0.4574489093196963
         9 r2 score of the train data 0.57678165997314 and test data 0.5543289851177138
         10 r2 score of the train data 0.5838871469899598 and test data 0.40937726622385506
         11 r2 score of the train data 0.5742070049919803 and test data 0.5752652423899043
         12 r2 score of the train data 0.5703728938595942 and test data 0.6575806943704936
         13 r2 score of the train data 0.5716939319770864 and test data 0.598721533902179
         14 r2 score of the train data 0.5701840060357541 and test data 0.6472752206392525
         Avg r2 score of train 0.5764039211497387 and test 0.5616233772335711
         Time taken - 1.3359219153722128 min
In [22]:
          dt = DecisionTreeRegressor(ccp alpha=0.0, criterion='mse', max depth=3,
                                max features='auto', max leaf nodes=None,
                                min impurity decrease=0.0, min impurity split=None,
                                min samples leaf=1, min samples split=2,
```

R^2 Score 0.5645815440213546 data

Random Forest

```
In [16]:
          clf = RandomForestRegressor()
          params = \{'n \text{ estimators}': [40,50,60,70,100],
                       'max depth':[3,5,6,7,8],
                       'min samples split':[2,3,4,5,6,7,8,9,10],
                       'max features': [0.80,.95, 1.0],
                       'min samples leaf': [1, 2,3,4,5,6,7,8,9],
                       'min impurity decrease':[1e-5,1e-4,1e-3,1e-2,1e-1,0,1,10]}
          kfold grid search(clf, params, x2, y2, fold = 10, kfold=15, search= 'random')
         Fitting 10 folds for each of 10 candidates, totalling 100 fits
         [Parallel(n jobs=-1)]: Using backend LokyBackend with 2 concurrent workers.
         [Parallel(n jobs=-1)]: Done 46 tasks
                                                   | elapsed: 5.3min
         [Parallel(n jobs=-1)]: Done 100 out of 100 | elapsed: 13.0min finished
         Best Estimator: RandomForestRegressor(bootstrap=True, ccp alpha=0.0, criterion='mse',
                               max depth=3, max features=0.95, max leaf nodes=None,
                               max samples=None, min impurity decrease=1e-05,
                               min impurity split=None, min samples leaf=1.
                               min samples split=3, min weight fraction leaf=0.0,
                               n estimators=100, n jobs=None, oob score=False,
                               random state=None, verbose=0, warm start=False)
         R2 metric :
         0 r2 score of the train data 0.577087291813353 and test data 0.6132711188897396
         1 r2 score of the train data 0.5703996644280093 and test data 0.6609684185922715
         2 r2 score of the train data 0.578452544680822 and test data 0.6028701420087133
         3 r2 score of the train data 0.6125421985716001 and test data 0.30357217989243657
         4 r2 score of the train data 0.582456005055181 and test data 0.5311574076974697
         5 r2 score of the train data 0.581347147499635 and test data 0.558037383796935
```

```
6 r2 score of the train data 0.5755225465600952 and test data 0.653998063252534
         7 r2 score of the train data 0.571482877918104 and test data 0.704416810834859
         8 r2 score of the train data 0.5880711478936098 and test data 0.4654762165866627
         9 r2 score of the train data 0.581164795256482 and test data 0.554894060385233
         10 r2 score of the train data 0.5867241938482605 and test data 0.46223717679878484
         11 r2 score of the train data 0.5771074180696558 and test data 0.5786484609591314
         12 r2 score of the train data 0.5749890540584395 and test data 0.6536878481220965
         13 r2 score of the train data 0.5753089240842089 and test data 0.6261519508819551
         14 r2 score of the train data 0.5744542619612625 and test data 0.645344399612696
         Avg r2 score of train 0.5804740047799146 and test 0.5743154425541012
         Time taken - 15.681973775227865 min
In [23]:
          rf = RandomForestRegressor(bootstrap=True, ccp alpha=0.0, criterion='mse',
                                max depth=3, max features=0.95, max leaf nodes=None,
                                max samples=None, min impurity decrease=1e-05,
                                min impurity split=None, min samples leaf=1,
                                min samples split=3, min weight fraction leaf=0.0,
                                n estimators=100, n jobs=None, oob score=False,
                                random state=None, verbose=0, warm start=False)
          a = cross val score(rf, x2, y2, scoring= 'r2', cv= 10)
          print(f'R^2 Score {np.mean(a)} data' )
```

R^2 Score 0.5722703062634319 data

XGBoost-Regressor

```
Fitting 10 folds for each of 10 candidates, totalling 100 fits
         [Parallel(n jobs=-1)]: Using backend LokyBackend with 2 concurrent workers.
         [Parallel(n jobs=-1)]: Done 46 tasks
                                                  | elapsed: 1.5min
         [Parallel(n jobs=-1)]: Done 100 out of 100 | elapsed: 3.3min finished
         Best Estimator: XGBRFRegressor(base score=0.5, colsample bylevel=1, colsample bynode=0.8,
                        colsample bytree=1, gamma=0, learning rate=1, max delta step=0,
                        max depth=7, max features=0.8, min child weight=1,
                        min impurity decrease=0.001, min samples leaf=4,
                        min samples split=2, missing=None, n estimators=40, n jobs=1,
                        nthread=None, objective='reg:linear', random state=0,
                        reg alpha=0, reg lambda=1, scale pos weight=1, seed=None,
                        silent=True, subsample=0.8, verbosity=1)
         R2 metric :
         0 r2 score of the train data 0.5607519129595249 and test data 0.612350904071567
         1 r2 score of the train data 0.5577277386771264 and test data 0.6554340512207613
         2 r2 score of the train data 0.5617422277259452 and test data 0.6032007707051259
         3 r2 score of the train data 0.5954512272952628 and test data 0.306385454262185
         4 r2 score of the train data 0.567338555840642 and test data 0.5264216134296651
         5 r2 score of the train data 0.5652550760126759 and test data 0.554397766212482
         6 r2 score of the train data 0.5588987037502464 and test data 0.6583819998690197
         7 r2 score of the train data 0.556428402163345 and test data 0.703099768991829
         8 r2 score of the train data 0.5716366105241955 and test data 0.47008749323113264
         9 r2 score of the train data 0.5650894144901676 and test data 0.5520598424562043
         10 r2 score of the train data 0.5698685506395984 and test data 0.486998617080977
         11 r2 score of the train data 0.5633495452041231 and test data 0.5760128251942953
         12 r2 score of the train data 0.558672400266895 and test data 0.6593011513004952
         13 r2 score of the train data 0.5600793827235035 and test data 0.63677973271229
         14 r2 score of the train data 0.5586598565803855 and test data 0.6518594742101987
         Avg r2 score of train 0.5647299736569091 and test 0.5768514309965485
         Time taken - 3.8507372379302978 min
In [24]:
         xg = XGBRFRegressor(base score=0.5, colsample bylevel=1, colsample bynode=0.8,
                         colsample bytree=1, gamma=0, learning rate=1, max delta step=0,
                         max depth=7, max features=0.8, min child weight=1,
                         min impurity decrease=0.001, min samples leaf=4,
                         min samples split=2, missing=None, n estimators=40, n jobs=1,
                         nthread=None, objective='reg:linear', random state=0,
                         reg alpha=0, reg lambda=1, scale pos weight=1, seed=None,
                         silent=True, subsample=0.8, verbosity=1)
```

```
a = cross_val_score(xg, x2, y2, scoring= 'r2', cv= 10)
print(f'R^2 Score {np.mean(a)} data' )
```

R^2 Score 0.5730571735111031 data

AdaBoost-Regressor

```
In [18]:
         clf = AdaBoostRegressor(base estimator=None, learning rate=0.0001,
                                  n estimators=300, random state=None, loss= 'linear')
          params = {'n estimators' : [100, 150, 200, ],
                    'learning rate' :[0.0001,0.001,0.01, 0.1],
                    'loss' : [ 'linear', 'square', 'exponential'],
                    'random state' : [10,20,30]
          kfold grid search(clf, params, x2, y2, fold = 10, kfold = 15, search = 'random')
         Fitting 10 folds for each of 10 candidates, totalling 100 fits
         [Parallel(n jobs=-1)]: Using backend LokyBackend with 2 concurrent workers.
         [Parallel(n jobs=-1)]: Done 46 tasks | elapsed: 14.2min
         [Parallel(n jobs=-1)]: Done 100 out of 100 | elapsed: 32.0min finished
         Best Estimator: AdaBoostRegressor(base estimator=None, learning rate=0.0001, loss='square',
                           n estimators=150, random state=30)
         R2 metric :
         0 r2 score of the train data 0.5646822713230479 and test data 0.6066783130115726
         1 r2 score of the train data 0.5577141547036 and test data 0.6649903462149656
         2 r2 score of the train data 0.5651393370662766 and test data 0.6033862254239895
         3 r2 score of the train data 0.5989253378984944 and test data 0.3064867419902134
         4 r2 score of the train data 0.5703661713907457 and test data 0.532300026745163
         5 r2 score of the train data 0.5687084265678528 and test data 0.5544782697611486
         6 r2 score of the train data 0.5626966400650278 and test data 0.6543380452537592
         7 r2 score of the train data 0.5566910900648376 and test data 0.7050220948487183
         8 r2 score of the train data 0.5756621801071371 and test data 0.46433697036193844
         9 r2 score of the train data 0.5682103360047548 and test data 0.5530743664423573
         10 r2 score of the train data 0.5787962790410337 and test data 0.40115919393392063
         11 r2 score of the train data 0.5669787210234464 and test data 0.5789391829398253
```

Gradient Boosting Regressor

```
In [19]:
         clf = GradientBoostingRegressor(max depth= 2, learning rate= 0.1, random_state= 10,n_estimators=5000,
                                         n iter no change = 11)
          params = {'n estimators' : [500,800,1000, 1500, 2000],
                    'loss' : [ 'huber', 'exponential'],
                    'learning rate' : [0.01, 0.01, 0.1],
                    'max depth' : [3,4,5,7]
          kfold grid search(clf, params, x2, y2, fold = 10, kfold = 15, search = 'random')
         Fitting 10 folds for each of 10 candidates, totalling 100 fits
         [Parallel(n jobs=-1)]: Using backend LokyBackend with 2 concurrent workers.
         [Parallel(n jobs=-1)]: Done 62 tasks | elapsed: 26.1min
         [Parallel(n jobs=-1)]: Done 100 out of 100 | elapsed: 48.8min finished
         Best Estimator : GradientBoostingRegressor(alpha=0.9, ccp alpha=0.0, criterion='friedman mse',
                                   init=None, learning rate=0.01, loss='huber',
                                   max depth=3, max features=None, max leaf nodes=None,
                                   min impurity decrease=0.0, min impurity split=None,
                                   min samples leaf=1, min samples split=2,
                                   min weight fraction leaf=0.0, n estimators=1500,
                                   n iter no change=11, presort='deprecated',
```

```
R2 metric :
         0 r2 score of the train data 0.572231628764966 and test data 0.6037034985933161
         1 r2 score of the train data 0.5621641874751212 and test data 0.6414798160235033
         2 r2 score of the train data 0.5867592358384793 and test data 0.5801527896688994
         3 r2 score of the train data 0.6211634297112336 and test data 0.2921090496165797
         4 r2 score of the train data 0.5748619162004004 and test data 0.5269244313269791
         5 r2 score of the train data 0.5872705137166562 and test data 0.549299214543042
         6 r2 score of the train data 0.5855561004342713 and test data 0.6578667585099021
         7 r2 score of the train data 0.570752093827166 and test data 0.702863121942162
         8 r2 score of the train data 0.5842016164058297 and test data 0.4572405045212359
         9 r2 score of the train data 0.5798527918648275 and test data 0.5332054489958042
         10 r2 score of the train data 0.5820903105918269 and test data 0.4970823405618191
         11 r2 score of the train data 0.5844905309440279 and test data 0.5477665372635308
         12 r2 score of the train data 0.5732064802854702 and test data 0.6722729775204286
         13 r2 score of the train data 0.5794606697712166 and test data 0.6724396565967647
         14 r2 score of the train data 0.5778799612181722 and test data 0.6866115001225563
         Avg r2 score of train 0.5814627644699777 and test 0.5747345097204349
         Time taken - 64.8858946720759 min
In [26]:
           gb = GradientBoostingRegressor(alpha=0.9, ccp alpha=0.0, criterion='friedman mse',
                                    init=None, learning rate=0.01, loss='huber',
                                    max depth=3, max features=None, max leaf nodes=None,
                                    min impurity decrease=0.0, min impurity split=None,
                                    min samples leaf=1, min samples split=2,
                                    min weight fraction leaf=0.0, n estimators=1500,
                                    n iter no change=11, presort='deprecated',
                                    random state=10, subsample=1.0, tol=0.0001,
                                    validation fraction=0.1, verbose=0, warm start=False)
          a = cross val score(gb, x2, y2, scoring= 'r2', cv= 10)
          print(f'R^2 Score {np.mean(a)} data')
         R^2 Score 0.5731033817060677 data
```

random state=10, subsample=1.0, tol=0.0001,

validation fraction=0.1, verbose=0, warm start=False)

LGBM Regressor

In [21]:

```
LGBMRegressor(boosting type='gbdt', class weight=None, colsample bytree=1.0,
                        importance type='split', learning rate=0.001, max depth=3,
                        min child samples=20, min child weight=0.001, min split gain=0.0,
                        n estimators=5000, n jobs=-1, num leaves=6, objective=None,
                        random state=None, reg alpha=0.0, reg lambda=0.0, silent=True,
                        subsample=1.0, subsample for bin=200000, subsample freq=0)
          params = {'min child samples' : [10, 20,50],
                    'num leaves' : [5,6],
                    'max depth' : [2, 3, 5],
                    'n estimators' : [1000,2000,4000,5000],
                    'learning rate' : [0.0001,0.001,0.01,0.1]
          kfold grid search(clf, params, x2, y2.ravel(), fold = 10, kfold = 15, search = False)
         R2 metric :
         0 r2 score of the train data 0.5997441404083625 and test data 0.6096233813618482
         1 r2 score of the train data 0.5958881929472761 and test data 0.6445588724260033
         2 r2 score of the train data 0.602526807697845 and test data 0.6007657087156126
         3 r2 score of the train data 0.6327607569536936 and test data 0.3070436943331817
         4 r2 score of the train data 0.6063989312693985 and test data 0.5403439972577462
         5 r2 score of the train data 0.6060046445533136 and test data 0.5558067093017696
         6 r2 score of the train data 0.5993389114393196 and test data 0.6537402142783227
         7 r2 score of the train data 0.5958897819905952 and test data 0.6964613240982211
         8 r2 score of the train data 0.6112823739876911 and test data 0.48196247922911584
         9 r2 score of the train data 0.6066215294634596 and test data 0.5536277630993041
         10 r2 score of the train data 0.607835933029663 and test data 0.5027366167589261
         11 r2 score of the train data 0.6068813272907176 and test data 0.5586564302402194
         12 r2 score of the train data 0.5974123440555079 and test data 0.6520383073567042
         13 r2 score of the train data 0.5979811420951997 and test data 0.6558092545600763
         14 r2 score of the train data 0.5989684485447244 and test data 0.6523321985442174
         Avg r2 score of train 0.6043690177151179 and test 0.577700463437418
         Time taken - 9.296628991762796 min
In [28]:
          lg = LGBMRegressor(boosting type='gbdt', class weight=None, colsample bytree=1.0,
                        importance type='split', learning rate=0.001, max depth=3,
                        min child samples=20, min child weight=0.001, min split gain=0.0,
                        n estimators=5000, n jobs=-1, num leaves=6, objective=None,
                        random state=None, reg alpha=0.0, reg lambda=0.0, silent=True,
                        subsample=1.0, subsample for bin=200000, subsample freq=0)
```

```
a = cross val score(lg, x2, y2, scoring= 'r2', cv= 10)
          print(f'R^2 Score {np.mean(a)} data')
         R^2 Score 0.5736623038048128 data
In [29]:
          estimators = [ ('rf', RandomForestRegressor(bootstrap=True, ccp alpha=0.0, criterion='mse',
                                max depth=6, max features=1.0, max leaf nodes=None,
                                max samples=None, min impurity decrease=0.0001,
                                min impurity split=None, min samples leaf=4,
                                min samples split=10, min weight fraction leaf=0.0,
                                n estimators=100, n jobs=None, oob score=False,
                                random state=None, verbose=0, warm start=False)),
                        ('xgb', XGBRFRegressor(base score=0.5, colsample bylevel=1, colsample bynode=0.8,
                         colsample bytree=1, gamma=0, learning rate=1, max delta step=0,
                         max depth=5, max features=0.95, min child weight=1,
                         min impurity decrease=0.0001, min samples leaf=2,
                         min samples split=4, missing=None, n estimators=50, n jobs=1,
                         nthread=None, objective='reg:linear', random state=0,
                         reg alpha=0, reg lambda=1, scale pos weight=1, seed=None,
                         silent=True, subsample=0.8, verbosity=1)),
                        ('lb', LGBMRegressor(boosting type='gbdt', class weight=None, colsample bytree=1.0,
                        importance type='split', learning rate=0.001, max depth=3,
                        min child samples=20, min child weight=0.001, min split gain=0.0,
                        n estimators=5000, n jobs=-1, num leaves=6, objective=None,
                        random state=None, reg alpha=0.0, reg lambda=0.0, silent=True,
                        subsample=1.0, subsample for bin=200000, subsample freq=0))
In [30]:
          stack = StackingRegressor(estimators= estimators,
                                    final estimator= Ridge(),
          cv score = cross val score(stack, x2, y2.ravel(), scoring='r2', cv= 10, verbose=5, n jobs=-1)
          print('Mean Score:',cv score.mean())
          print('Standard Deviation:',cv score.std())
```

```
[Parallel(n_jobs=-1)]: Using backend LokyBackend with 2 concurrent workers.
Mean Score: 0.575714581019241
Standard Deviation: 0.08823219159340108
[Parallel(n_jobs=-1)]: Done 10 out of 10 | elapsed: 39.0min remaining: 0.0s
[Parallel(n_jobs=-1)]: Done 10 out of 10 | elapsed: 39.0min finished
```

Models fitting

```
In [36]:
          for model, label in zip([dt, rf, xg, ab, gb, lg, stack], ['decision tree', 'random forest', 'xgboost',
                                               'adaboost', 'gradient boost', 'lighgbm', 'stack ensemble']):
           model.fit(x2, y2)
           predict = model.predict(test.iloc[:, 8:])
           df = pd.DataFrame()
            df['ID'] = test.ID
            df['v'] = predict
           df.to csv(label + ' top features(pca-100) with clip.csv', index= False)
            print('csv file compeleted ' + label)
         csv file compeleted decision tree
         csv file compeleted random forest
         csv file compeleted xgboost
         csv file compeleted adaboost
         csv file compeleted gradient boost
         csv file compeleted lighgbm
         csv file compeleted stack ensemble
In [57]:
          col = ['MODELS', 'FEATURES', 'CROSS-VALIDATION' , 'PUBLIC SCORE', 'PRIVATE SCORE']
          tb = PrettyTable()
          tb.add column(col[0], ['DECISION TREE', 'RANDOM FOREST', 'XGBOOST', 'ADA-BOOST', 'GRADIENT BOOSTING', 'LIGHT GBM', 'S
          tb.add column(col[1], ['PCA(100) + BINARY + LABEL + Y 150 CLIP +SYNTHETIC FEATURES', 'PCA(100) + BINARY + LABEL + Y 15
```

```
'PCA(100) + BINARY + LABEL + Y 150 CLIP +SYNTHETIC FEATURES','PCA(100) + BINARY + LABEL + Y 15
                        'PCA(100) + BINARY + LABEL + Y 150 CLIP +SYNTHETIC FEATURES', 'PCA(100) + BINARY + LABEL + Y 15
                        'PCA(100) + BINARY + LABEL + Y 150 CLIP +SYNTHETIC FEATURES',])
tb.add column(col[2], ['0.5616', '0.5743', '0.5768', '0.5715', '0.5847', '0.5775', '0.5775'])
tb.add column(col[3], ['0.54541', '0.54556', '0.54345', '0.54539', '0.54321', '0.55793', '0.55442'])
tb.add column(col[4], ['0.52796', '0.53966', '0.53673', '0.53996', '0.53868', '0.54630', '0.54515'])
print(tb)
       MODELS
                                               FEATURES
                                                                                 | CROSS-VALIDATION | PUBLIC SCORE
PRIVATE SCORE |
                                                                                        0.5616
                                                                                                        0.54541
   DECISION TREE
                    | PCA(100) + BINARY + LABEL + Y 150 CLIP +SYNTHETIC FEATURES |
  0.52796
   RANDOM FOREST
                  | PCA(100) + BINARY + LABEL + Y 150 CLIP +SYNTHETIC FEATURES |
                                                                                                        0.54556
                                                                                        0.5743
   0.53966
      XGB00ST
                    | PCA(100) + BINARY + LABEL + Y 150 CLIP +SYNTHETIC FEATURES |
                                                                                        0.5768
                                                                                                        0.54345
   0.53673
      ADA - BOOST
                    | PCA(100) + BINARY + LABEL + Y 150 CLIP +SYNTHETIC FEATURES |
                                                                                        0.5715
                                                                                                        0.54539
  0.53996
 GRADIENT BOOSTING | PCA(100) + BINARY + LABEL + Y 150 CLIP +SYNTHETIC FEATURES |
                                                                                                        0.54321
                                                                                        0.5847
  0.53868
     LIGHT GBM
                    | PCA(100) + BINARY + LABEL + Y 150 CLIP +SYNTHETIC FEATURES |
                                                                                        0.5775
                                                                                                        0.55793
  0.54630
  STCKED-ENSEMBLE
                  | PCA(100) + BINARY + LABEL + Y 150 CLIP +SYNTHETIC FEATURES |
                                                                                        0.5757
                                                                                                        0.55442
   0.54515
```

OBSERVATION

- This dataset contain 100 PCA features from binary and label encode features. With target y clip at 150secs
- In this dataset light gradient boosting perform well than other. LGB got 0.55793 in public and 0.54630 private score.

Models with selected top features using selectkbest method

```
In [37]:
          # Import the train and test csv files
          train = pd.read csv(r'/content/drive/MyDrive/Datasets/selectK train feature.csv')
          test = pd.read csv(r'/content/drive/MyDrive/Datasets/selectK test feature.csv')
          train = train.drop(['Unnamed: 0'], axis = 1)
          test = test.drop(['Unnamed: 0'], axis = 1)
          train.head()
            X10 X12 X13 X14 X19 X20 X22 X23 X27 X28 X29 X31 X35 X37 X43 X44 X45 X46 X47 X48 X50 X51 X52 X54 X56 X57 X6
Out[37]:
                                                                                                                      0
                  0
                                                     0
                                                         0
                                                                               0
                                                                                                0
                  0
                               0
                                   0
                                        0
                                                                               0
                                                                                            0
                                                                                                             0
                               0
                                   0
                                        0
                                                                               0
                                                                                   0
                                                                                                0
                                                                                                                      0
                                                                                                                          0
        5 rows × 252 columns
In [38]:
          #Check the Shape of the test and train data
          print('train data shape - ', train.shape)
          print('test data shape - ', test.shape)
         train data shape - (4209, 252)
         test data shape - (4209, 251)
In [39]:
          # declare the train set and label
          train.loc[train['y'] > 150] = 150
          y2 = train.y
          x2 = train.drop(['y', 'ID'], axis= 1)
```

```
In [40]: # Split train data into train-set and test-set
    x_train2, x_test2, y_train2, y_test2 = train_test_split(x2,y2, test_size = 0.33,random_state= 42)
    print(x_train2.shape, y_train2.shape)
    print(x_test2.shape, y_test2.shape)

(2820, 250) (2820,)
    (1389, 250) (1389,)
```

Decision Tree

```
In [ ]:
         clf = DecisionTreeRegressor(ccp alpha=0.0, criterion='mse', max depth=8,
                               max features='auto', max_leaf_nodes=None,
                               min impurity decrease=0.0, min impurity split=None,
                               min samples leaf=1, min samples split=2,
                               min weight fraction leaf=0.0, presort='deprecated',
                               random state=10, splitter='best')
         params = \{\text{'max depth'}: [2,3,4,8,10,15],
                   'max features' : ['auto', 'sqrt', 'log2'],
                   'random state' : [5,10,20,30],
         kfold grid search(clf, params, x2, y2, 10, kfold = 15, search= False)
        R2 metric :
        0 r2 score of the train data 0.7003943217596944 and test data 0.5912088567314289
        1 r2 score of the train data 0.684772877128098 and test data 0.7461049970562634
        2 r2 score of the train data 0.7024668433891952 and test data 0.5695775837295918
        3 r2 score of the train data 0.6955558898354829 and test data 0.6247256843562159
        4 r2 score of the train data 0.6943306686873321 and test data 0.6938806609949865
        5 r2 score of the train data 0.6917795390935819 and test data 0.5715765496011846
```

6 r2 score of the train data 0.6950587650032459 and test data 0.6254903902405267 7 r2 score of the train data 0.6899252956182615 and test data 0.7482496095026618 8 r2 score of the train data 0.7066464250517441 and test data 0.4451958898852242 9 r2 score of the train data 0.699592124849862 and test data 0.5409245741315845 10 r2 score of the train data 0.7090500610772879 and test data 0.3688190526192887 11 r2 score of the train data 0.6980781796780453 and test data 0.632043078306469 12 r2 score of the train data 0.6959999890239887 and test data 0.6536206550723016 13 r2 score of the train data 0.6938200362681952 and test data 0.6973450444266769 14 r2 score of the train data 0.6936533776743699 and test data 0.5191145080931587

Avg r2_score of train 0.6967416262758924 and test 0.6018584756498376 Time taken - 0.0696996808052063 min

R^2 Score 0.6350922926910991 data

Random Forest

```
clf = RandomForestRegressor(bootstrap=True, ccp alpha=0.0, criterion='mse',
                       max depth=3, max features=0.8, max leaf nodes=None,
                       max samples=None, min impurity decrease=1e-05,
                       min impurity split=None, min samples leaf=5,
                       min samples split=9, min weight fraction leaf=0.0,
                       n estimators=40, n jobs=None, oob score=False,
                       random state=None, verbose=0, warm start=False)
params = \{'n \text{ estimators}': [40,50,60,70,100],
              'max depth':[3,5,6,7,8],
              'min samples split':[2,3,4,5,6,7,8,9,10],
              'max features': [0.80,.95, 1.0],
              'min samples leaf': [1, 2,3,4,5,6,7,8,9],
              'min impurity decrease': [1e-5,1e-4,1e-3,1e-2,1e-1,0,1,10]}
kfold grid search(clf, params, x2, y2, fold = 10, kfold=15, search= False)
R2 metric :
0 r2 score of the train data 0.6491851497180579 and test data 0.6722657818421716
1 r2 score of the train data 0.6428875293997556 and test data 0.7650168179513238
2 r2 score of the train data 0.6543022991441494 and test data 0.6009548522547989
3 r2 score of the train data 0.651093238820595 and test data 0.6413429640388058
4 r2 score of the train data 0.6486153400378359 and test data 0.6732278735764214
```

```
5 r2 score of the train data 0.6500914264009571 and test data 0.6524032274886278
         6 r2 score of the train data 0.650940549896302 and test data 0.6542854706156918
         7 r2 score of the train data 0.6447753756785972 and test data 0.7542422390060914
         8 r2 score of the train data 0.6659204711262184 and test data 0.46907503683814333
         9 r2 score of the train data 0.6517637548732664 and test data 0.6296769532934854
         10 r2 score of the train data 0.6605605207274278 and test data 0.49104247104708254
         11 r2 score of the train data 0.6507621664241039 and test data 0.6474150149129719
         12 r2 score of the train data 0.6492585389063131 and test data 0.659040789441843
         13 r2 score of the train data 0.6478413435266211 and test data 0.6878571251898438
         14 r2 score of the train data 0.6498242678129074 and test data 0.6548025156628345
         Avg r2 score of train 0.6511881314995407 and test 0.6435099422106758
         Time taken - 0.5376898010571798 min
In [42]:
          rf = RandomForestRegressor(bootstrap=True, ccp alpha=0.0, criterion='mse',
                                max depth=3, max features=0.8, max leaf nodes=None,
                                max samples=None, min impurity decrease=1e-05,
                                min impurity split=None, min samples leaf=5,
                                min samples split=9, min weight fraction leaf=0.0,
                                n estimators=40, n jobs=None, oob score=False,
                                random state=None, verbose=0, warm start=False)
          a = cross val score(rf, x2, y2, scoring= 'r2', cv= 10)
          print(f'R^2 Score {np.mean(a)} data' )
```

R^2 Score 0.6411188337826735 data

XGBoost-Regressor

```
'subsample': [0.2,0.3,0.5,1],
                       'gamma': [0.0001,0.001,0,0.1,0.01,0.5,1],
                       'reg alpha': [0.00001,0.0001,0.001,0.01,0.1]}
          kfold grid search(clf, params, x2, y2, fold = 10, kfold=15, search= False)
         R2 metric :
         0 r2 score of the train data 0.6067502601868848 and test data 0.6374083132665591
         1 r2 score of the train data 0.6037579887656499 and test data 0.6953451163213371
         2 r2 score of the train data 0.6121444549439352 and test data 0.6011016381585581
         3 r2 score of the train data 0.6129187708599382 and test data 0.5565082186028582
         4 r2 score of the train data 0.6113841767426726 and test data 0.582448928624506
         5 r2 score of the train data 0.6078223907106232 and test data 0.6276658565426563
         6 r2 score of the train data 0.6084467931525492 and test data 0.6559734999802659
         7 r2 score of the train data 0.603205051745912 and test data 0.7092297719647569
         8 r2 score of the train data 0.6226020717930794 and test data 0.46860069407268745
         9 r2 score of the train data 0.6124662945950516 and test data 0.5626195379397068
         10 r2 score of the train data 0.6185700634680764 and test data 0.48692268287664975
         11 r2 score of the train data 0.608856439703718 and test data 0.6154646014250558
         12 r2 score of the train data 0.6061588075086194 and test data 0.6581597523169269
         13 r2 score of the train data 0.6072817595908033 and test data 0.6407171369439164
         14 r2 score of the train data 0.6060176585456536 and test data 0.6507545607215774
         Avg r2 score of train 0.6098921988208779 and test 0.6099280206505344
         Time taken - 0.6381794929504394 min
In [43]:
           xq = XGBRFRegressor(colsample bylevel=1,
                         colsample bynode=0.8, colsample bytree=1, gamma=0,
                         learning rate=1, max delta step=0, max depth=5,
                         max features=0.95, min child weight=1, min impurity decrease=1,
                         min samples leaf=1, min samples split=5, missing=None,
                         n estimators=100, n jobs=1, nthread=None, objective='req:linear',
                         random state=0, reg alpha=0, reg lambda=1, scale pos weight=1,
                         seed=None, silent=True, subsample=0.8, verbositv=1)
          a = cross val score(xg, x2, y2, scoring= 'r2', cv= 10)
          print(f'R^2 Score {np.mean(a)} data')
```

R^2 Score 0.6096710572369947 data

AdaBoost-Regressor

```
In []: clf = AdaBoostRegressor(base estimator=None, learning rate=0.0001, loss='linear',
                            n estimators=100, random state=10)
          params = {'n estimators' : [100, 150, 200, ],
                    'learning rate' :[0.0001,0.001,0.01, 0.1],
                    'loss' : [ 'linear', 'square', 'exponential'],
                    'random state' : [10.20.30]
          kfold grid search(clf, params, x2, y2, fold = 10, kfold = 15, search = False)
         R2 metric :
         0 r2 score of the train data 0.6406019024644523 and test data 0.6641913063532133
         1 r2 score of the train data 0.6341117815098307 and test data 0.7675518739100321
         2 r2 score of the train data 0.6448193857404521 and test data 0.601364590744293
         3 r2 score of the train data 0.6420025644845055 and test data 0.6401058176407162
         4 r2 score of the train data 0.6426822251515699 and test data 0.6709984599032441
         5 r2 score of the train data 0.6412649710061986 and test data 0.6561873993797418
         6 r2 score of the train data 0.6414199311868302 and test data 0.6550053083325136
         7 r2 score of the train data 0.635692428613147 and test data 0.750110156147811
         8 r2 score of the train data 0.6566658201904927 and test data 0.4667308312720433
         9 r2 score of the train data 0.6433253839092598 and test data 0.6252818172649386
         10 r2 score of the train data 0.6530673875304205 and test data 0.4887717640412784
         11 r2 score of the train data 0.6424865119330938 and test data 0.6470462212100361
         12 r2 score of the train data 0.6408561893286236 and test data 0.6616694666615984
         13 r2 score of the train data 0.6397242178987328 and test data 0.6801148107924511
         14 r2 score of the train data 0.6405964615369728 and test data 0.6550850005146253
         Avg r2 score of train 0.6426211441656389 and test 0.6420143216112356
         Time taken - 2.685768520832062 min
In [44]:
                 AdaBoostRegressor(base estimator=None, learning rate=0.0001, loss='linear',
                            n estimators=100, random state=10)
          a = cross val score(ab, x2, y2, scoring= 'r2', cv= 10)
          print(f'R^2 Score {np.mean(a)} data')
```

R^2 Score 0.6395565714731158 data

Gradient Boosting Regressor

```
In [ ]:
          clf = GradientBoostingRegressor(alpha=0.9, ccp alpha=0.0, criterion='friedman mse',
                                    init=None, learning rate=0.01, loss='huber',
                                    max depth=3, max features=None, max leaf nodes=None,
                                    min impurity decrease=0.0, min impurity_split=None,
                                    min samples_leaf=1, min_samples_split=2,
                                    min weight fraction leaf=0.0, n estimators=2000,
                                    n iter no change=11, presort='deprecated',
                                    random state=10, subsample=1.0, tol=0.0001,
                                    validation fraction=0.1, verbose=0, warm start=False)
          params = {'n estimators' : [800,1000, 1500, 2000, 2500],
                    'loss' : [ 'huber', 'exponential'],
                    'learning rate' : [0.01, 0.01, 0.1],
                    'max depth' : [3,4,5,7]
          kfold grid search(clf, params, x2, y2, fold = 10, kfold = 15, search = False)
         R2 metric :
         0 r2 score of the train data 0.646898352654415 and test data 0.6463232721630487
         1 r2 score of the train data 0.6445815700448241 and test data 0.7630028652459149
         2 r2 score of the train data 0.6638708224569014 and test data 0.5820050454376444
         3 r2 score of the train data 0.6643729193689359 and test data 0.6298306073660422
         4 r2 score of the train data 0.6452857437229575 and test data 0.6703858624196001
         5 r2 score of the train data 0.6467163192346321 and test data 0.6466422501820477
         6 r2 score of the train data 0.6520411010671734 and test data 0.6518687281616424
         7 r2 score of the train data 0.6511587363040845 and test data 0.7573584179324647
         8 r2 score of the train data 0.6688157545177016 and test data 0.453579856520536
         9 r2 score of the train data 0.6498989292469194 and test data 0.6088874846193633
         10 r2 score of the train data 0.6721223434084025 and test data 0.49579802946005624
         11 r2 score of the train data 0.6556792100878011 and test data 0.6329712960936003
         12 r2 score of the train data 0.6613676334095353 and test data 0.6671419303636512
         13 r2 score of the train data 0.6624697875464138 and test data 0.7195585718225759
         14 r2 score of the train data 0.6567325364921577 and test data 0.6894664334886403
         Avg r2 score of train 0.6561341173041902 and test 0.6409880434184552
         Time taken - 8.37012676000595 min
In [45]:
          gb = GradientBoostingRegressor(alpha=0.9, ccp alpha=0.0, criterion='friedman mse',
                                    init=None, learning rate=0.01, loss='huber',
```

R^2 Score 0.6401325221008258 data

LGBM Regressor

```
In [ ]:
         clf = LGBMRegressor(boosting type='gbdt', class weight=None, colsample bytree=1.0,
                       importance type='split', learning rate=0.01, max depth=3,
                       min child samples=50, min child weight=0.001, min split gain=0.0,
                       n estimators=4000, n jobs=-1, num leaves=5, objective=None,
                       random state=None, reg alpha=0.0, reg lambda=0.0, silent=True,
                       subsample=1.0, subsample for bin=200000, subsample freq=0)
         params = {'min child samples' : [10, 20,50],
                   'num leaves' : [5,6],
                   'max depth' : [2, 3, 5],
                   'n estimators' : [1000,2000,4000,5000],
                   'learning rate' : [0.0001,0.001,0.01,0.1]
         kfold grid search(clf, params, x2, y2.ravel(), fold = 10, kfold = 15, search = False)
        R2 metric :
        0 r2 score of the train data 0.7572398922837325 and test data 0.648629927585285
        1 r2 score of the train data 0.7559447382612352 and test data 0.7524994967038349
        2 r2 score of the train data 0.7635946493631556 and test data 0.5796051680603062
        3 r2 score of the train data 0.7616028837108777 and test data 0.6283320314911397
        4 r2 score of the train data 0.756111951588603 and test data 0.6669960134479473
        5 r2 score of the train data 0.7592501493147876 and test data 0.6344725589160572
        6 r2 score of the train data 0.757831426306504 and test data 0.6637281324013
        7 r2 score of the train data 0.7549205437978412 and test data 0.7264772198521539
```

```
8 r2 score of the train data 0.7704056362229155 and test data 0.47265909739244094
         9 r2 score of the train data 0.7606748853035004 and test data 0.629765276411993
         10 r2 score of the train data 0.7635879477252655 and test data 0.49182201718973395
         11 r2 score of the train data 0.7608877785525503 and test data 0.6206099747367849
         12 r2 score of the train data 0.7592847885219717 and test data 0.6312332206937824
         13 r2 score of the train data 0.7559390326059556 and test data 0.7038297346424771
         14 r2 score of the train data 0.755889171858096 and test data 0.6467577178304664
         Avg r2 score of train 0.7595443650277994 and test 0.6331611724903802
         Time taken - 4.181476438045502 min
In [46]:
          lb = LGBMRegressor(boosting type='gbdt', class weight=None, colsample bytree=1.0,
                        importance type='split', learning rate=0.01, max depth=3,
                        min child samples=50, min child weight=0.001, min split gain=0.0,
                        n estimators=4000, n jobs=-1, num leaves=5, objective=None,
                        random state=None, reg alpha=0.0, reg lambda=0.0, silent=True,
                        subsample=1.0, subsample for bin=200000, subsample freq=0)
          a = cross val score(lb, x2, y2, scoring= 'r2', cv= 10)
          print(f'R^2 Score {np.mean(a)} data')
         R^2 Score 0.6280864312506385 data
In [ ]:
          estimators = [ ('rf', RandomForestRegressor(bootstrap=True, ccp alpha=0.0, criterion='mse',
                                max depth=5, max features=0.95, max leaf nodes=None,
                                max samples=None, min impurity decrease=0.001,
                                min impurity split=None, min samples leaf=2,
                                min samples split=8, min weight fraction leaf=0.0,
                                n estimators=70, n jobs=None, oob score=False,
                                random state=None, verbose=0, warm start=False)),
                          ('xg', XGBRFRegressor(colsample bylevel=1,
                         colsample bynode=0.8, colsample bytree=1, gamma=0,
                         learning rate=1, max delta step=0, max depth=5,
                         max features=0.95, min child weight=1, min impurity decrease=1,
                         min samples leaf=1, min samples split=5, missing=None,
                         n estimators=100, n jobs=1, nthread=None, objective='reg:linear',
                         random state=0, reg alpha=0, reg lambda=1, scale pos weight=1,
                         seed=None, silent=True, subsample=0.8, verbosity=1)),
```

Models fitting

```
csv file done ! decision_tree
csv file done ! random_forest
csv file done ! xgboost
csv file done ! adaboost
csv file done ! gradient_boost
csv file done ! lighgbm
csv file done ! stack-ensemble
```

R² Score

```
In [55]:
    col = ['MODELS', 'FEATURES', 'CROSS-VALIDATION' , 'PUBLIC SCORE', 'PRIVATE SCORE']
    tb = PrettyTable()
    tb.add_column(col[0], ['DECISION TREE', 'RANDOM FOREST', 'XGBOOST', 'ADA-BOOST', 'GRADIENT BOOSTING', 'LIGHT GBM', 'S
    tb.add_column(col[1], ['TOP FEATURES USING SELECT K BEST METHOD', 'TOP FEATURES USING SELECT K BEST METHOD', 'TOP FEATURES USING SELECT K BEST METHOD', 'TOP FEATURES USING SELECT K BEST METHOD'])
    tb.add_column(col[2], ['0.6018', '0.6435', '0.6100', '0.6420', '0.6409', '0.6331', '0.5762'])
    tb.add_column(col[3], ['0.54021', '0.54543', '0.54561', '0.53577', '0.53822', '0.53822', '0.55039'])
    tb.add_column(col[4], ['0.52598', '0.54326', '0.53864', '0.53955', '0.53663', '0.54449', '0.54710'])
    print(tb)
```

MODELS	FEATURES	CROSS-VALIDATION	PUBLIC SCORE	PRIVATE SCORE
DECISION TREE RANDOM FOREST XGB00ST ADA-B00ST GRADIENT B00STING LIGHT GBM STCKED-ENSEMBLE	TOP FEATURES USING SELECT K BEST METHOD TOP FEATURES USING SELECT K BEST METHOD	0.6018 0.6435 0.6100 0.6420 0.6409 0.6331 0.5762	0.54021 0.54543 0.54561 0.53577 0.53822 0.53822	0.52598 0.54326 0.53864 0.53955 0.53663 0.54449 0.54710

OBSERVATION

- This dataset created with help of SelectKBest feature selection method, we selected top 250 most important features. But we can see that these features are not get good score than other.
- In this sets stacked ensembles perform best and also another datasets, stacked model got 0.55039 in public and 0.54710 in private leaderboard.