In [74]:

```
#import all necessary Libraries

import pandas as pd
import numpy as np
import sklearn as sl
from sklearn.decomposition import PCA
from sklearn import metrics
import seaborn as sns
import warnings
warnings.filterwarnings('ignore')
from sklearn import preprocessing
```

In [18]:

```
pip install xgboost
```

```
Requirement already satisfied: xgboost in c:\programdata\ssh\lib\site-pack ages (1.2.0)
Requirement already satisfied: scipy in c:\programdata\ssh\lib\site-packag es (from xgboost) (1.5.0)
Requirement already satisfied: numpy in c:\programdata\ssh\lib\site-packag es (from xgboost) (1.18.5)
Note: you may need to restart the kernel to use updated packages.
```

In [19]:

```
import xgboost as xgb
```

In [20]:

```
#reading the datasets and explore it
#train dataset

df_train=pd.read_csv("train.csv")

df_test=pd.read_csv("test.csv")

print('train shape',df_train.shape)

print('test shape',df_test.shape)
```

```
train shape (4209, 378) test shape (4209, 377)
```

```
In [21]:
```

```
df_train.head()
```

Out[21]:

	ID	У	X0	X1	X2	Х3	X4	X5	X6	X8	 X375	X376	X377	X378	X379	X380
0	0	130.81	k	٧	at	а	d	u	j	0	 0	0	1	0	0	0
1	6	88.53	k	t	av	е	d	У	I	0	 1	0	0	0	0	0
2	7	76.26	az	w	n	С	d	x	j	х	 0	0	0	0	0	0
3	9	80.62	az	t	n	f	d	x	I	е	 0	0	0	0	0	0
4	13	78.02	az	٧	n	f	d	h	d	n	 0	0	0	0	0	0

5 rows × 378 columns

In [22]:

```
df_test.head()
```

Out[22]:

	ID	X0	X1	X2	Х3	X4	X5	X6	X8	X10	 X375	X376	X377	X378	X379	X380	ХЗ
0	1	az	٧	n	f	d	t	а	W	0	 0	0	0	1	0	0	
1	2	t	b	ai	а	d	b	g	у	0	 0	0	1	0	0	0	
2	3	az	٧	as	f	d	а	j	j	0	 0	0	0	1	0	0	
3	4	az	- 1	n	f	d	z	- 1	n	0	 0	0	0	1	0	0	
4	5	W	s	as	С	d	У	i	m	0	 1	0	0	0	0	0	

5 rows × 377 columns

→

In [23]:

```
df_train.dtypes
```

Out[23]:

```
int64
ID
        float64
Χ0
         object
Х1
         object
Χ2
         object
X380
          int64
          int64
X382
X383
          int64
X384
          int64
X385
          int64
Length: 378, dtype: object
```

```
In [24]:
```

```
df_test.dtypes
Out[24]:
ID
         int64
        object
X0
Х1
        object
X2
        object
Х3
        object
          . . .
X380
         int64
X382
         int64
X383
         int64
         int64
X384
X385
         int64
Length: 377, dtype: object
In [25]:
#q2checking for the null value
df_train.isnull().sum()
Out[25]:
ID
        0
        0
Χ0
        0
Х1
        0
X2
        0
X380
X382
        0
X383
        0
X384
        0
X385
Length: 378, dtype: int64
In [26]:
df_test.isnull().sum()
Out[26]:
ΙD
        0
X0
        0
X1
        0
X2
        0
Х3
        0
X380
        0
X382
        0
X383
        0
X384
        0
X385
Length: 377, dtype: int64
```

In [27]:

```
dtype_df = df_train.dtypes.reset_index()
dtype_df.columns = ["Count", "Column Type"]
dtype_df.groupby("Column Type").aggregate('count').reset_index()
```

Out[27]:

	Column Type	Count
0	int64	369
1	float64	1
2	object	8

In [28]:

```
dtype_df = df_test.dtypes.reset_index()
dtype_df.columns = ["Count", "Column Type"]
dtype_df.groupby("Column Type").aggregate('count').reset_index()
```

Out[28]:

	Column Typ	e Count
0	int6	4 369
1	obje	ct 8

In [29]:

```
#datatype of columns
df_train.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 4209 entries, 0 to 4208
Columns: 378 entries, ID to X385
dtypes: float64(1) int64(369) enjoyt
```

dtypes: float64(1), int64(369), object(8)

memory usage: 12.1+ MB

In [31]:

```
#datatype of columns
df_test.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 4209 entries, 0 to 4208
Columns: 377 entries, ID to X385
dtypes: int64(369), object(8)
memory usage: 12.1+ MB
```

In [33]:

```
#splitting the dataset into X and y variable
from sklearn.model_selection import train_test_split

X_train = df_train.drop(['y','ID'],axis = 1)
y_train = df_train['y']
X=X_train
y=y_train
```

In [34]:

```
print('Train dataset shape = ', X_train.shape)
print('Train labels shape = ', y_train.shape)
```

```
Train dataset shape = (4209, 376)
Train labels shape = (4209,)
```

In [35]:

```
X_test = df_test.drop(['ID'], axis = 1)
y_test = df_test['ID']
```

In [37]:

```
print('Test dataset shape = ', X_test.shape)
print('Test labels shape = ', y_test.shape)
```

```
Test dataset shape = (4209, 376)
Test labels shape = (4209,)
```

In [43]:

X_train.corr()

Out[43]:

	X10	X12	X13	X14	X15	X16	X17	X18		
X10	1.000000	-0.033084	-0.028806	-0.100474	-0.002532	-0.005944	-0.010164	-0.010323		
X12	-0.033084	1.000000	0.214825	-0.246513	-0.006212	-0.014584	-0.024937	-0.025327		
X13	-0.028806	0.214825	1.000000	-0.083141	-0.005409	-0.012698	-0.021713	-0.010525		
X14	-0.100474	-0.246513	-0.083141	1.000000	-0.018865	-0.044291	0.012713	-0.076916		
X15	-0.002532	-0.006212	-0.005409	-0.018865	1.000000	-0.001116	-0.001908	-0.001938		
X380	-0.010479	-0.005566	0.023045	0.007743	-0.001968	-0.004619	-0.007899	-0.008022		
X382	-0.010164	-0.024937	-0.021713	0.012713	-0.001908	-0.004480	1.000000	0.085256		
X383	-0.004740	-0.011628	-0.010125	0.023604	-0.000890	-0.002089	-0.003572	0.062481		
X384	-0.002532	-0.006212	0.041242	0.025199	-0.000475	-0.001116	-0.001908	-0.001938		
X385	-0.004387	-0.010765	-0.009373	0.043667	-0.000824	-0.001934	-0.003307	-0.003359		
356 ro	356 rows × 356 columns									
4										

In [44]:

X_test.corr()

Out[44]:

	X10	X11	X12	X13	X14	X15	X16	X17	
X10	1.000000	-0.002146	-0.039453	-0.035496	-0.120379	-0.003717	-0.007125	-0.013108	
X11	-0.002146	1.000000	-0.004369	-0.003931	0.017825	-0.000412	-0.000789	-0.001452	
X12	-0.039453	-0.004369	1.000000	0.283228	-0.245127	-0.007570	-0.014509	-0.016991	
X13	-0.035496	-0.003931	0.283228	1.000000	-0.076145	-0.006811	-0.013054	-0.024015	
X14	-0.120379	0.017825	-0.245127	-0.076145	1.000000	-0.023097	-0.044269	0.000864	
X380	-0.012561	-0.001391	-0.025578	0.054582	0.007787	-0.002410	-0.004619	-0.008498	
X382	-0.013108	-0.001452	-0.016991	-0.024015	0.000864	-0.002515	-0.004821	1.000000	
X383	-0.003035	-0.000336	-0.006180	-0.005560	0.025212	-0.000582	-0.001116	-0.002053	
X384	-0.003717	-0.000412	-0.007570	-0.006811	0.030881	-0.000713	-0.001367	-0.002515	
X385	-0.005681	-0.000629	-0.011569	-0.010408	0.047195	-0.001090	-0.002089	-0.003844	
368 ro	368 rows × 368 columns								

localhost:8888/nbconvert/html/Mercedes Benz Greener Manufacturing Project 2020.ipynb?download=false

In [47]:

```
print(X_train.shape)
print (X_test.shape)
print (y_train.shape)
print(y_test.shape)
```

(4209, 364) (4209, 376) (4209,) (4209,)

In [48]:

#label encodcer

from sklearn.preprocessing import LabelEncoder
le=LabelEncoder()

In [77]:

df_train.apply(LabelEncoder().fit_transform)

Out[77]:

	ID	У	X0	X1	X2	Х3	X4	X5	X6	X8	 X375	X376	X377	X378	X379	Х3
0	0	2466	32	23	17	0	3	24	9	14	 0	0	1	0	0	
1	1	366	32	21	19	4	3	28	11	14	 1	0	0	0	0	
2	2	69	20	24	34	2	3	27	9	23	 0	0	0	0	0	
3	3	133	20	21	34	5	3	27	11	4	 0	0	0	0	0	
4	4	106	20	23	34	5	3	12	3	13	 0	0	0	0	0	
4204	4204	1657	8	20	16	2	3	0	3	16	 1	0	0	0	0	
4205	4205	1766	31	16	40	3	3	0	7	7	 0	1	0	0	0	
4206	4206	1801	8	23	38	0	3	0	6	4	 0	0	1	0	0	
4207	4207	280	9	19	25	5	3	0	11	20	 0	0	0	0	0	
4208	4208	1921	46	19	3	2	3	0	6	22	 1	0	0	0	0	

4209 rows × 378 columns

In [50]:

df_test.apply(LabelEncoder().fit_transform)

Out[50]:

	ID	X0	X1	X2	Х3	X4	X5	X6	X8	X10	 X375	X376	X377	X378	X379	X38
0	0	21	23	34	5	3	26	0	22	0	 0	0	0	1	0	
1	1	42	3	8	0	3	9	6	24	0	 0	0	1	0	0	
2	2	21	23	17	5	3	0	9	9	0	 0	0	0	1	0	
3	3	21	13	34	5	3	31	11	13	0	 0	0	0	1	0	
4	4	45	20	17	2	3	30	8	12	0	 1	0	0	0	0	
4204	4204	6	9	17	5	3	1	9	4	0	 0	0	0	0	0	
4205	4205	42	1	8	3	3	1	9	24	0	 0	1	0	0	0	
4206	4206	47	23	17	5	3	1	3	22	0	 0	0	0	0	0	
4207	4207	7	23	17	0	3	1	2	16	0	 0	0	1	0	0	
4208	4208	42	1	8	2	3	1	6	17	0	 1	0	0	0	0	

4209 rows × 377 columns

1	•	
In []:		
In []:		

#Dta preprocessing

In [72]:

```
def one hot(train data,test data,columns):
    for i,column in enumerate(columns):
        Xtrain = train data[str(column)].T
        Xtest = test data[str(column)].T
        # train df
        lb=preprocessing.LabelBinarizer()
        lb.fit(Xtrain)
        X classes = len(lb.classes )
        Xenc = lb.transform(Xtrain)
        Xtrain enc = pd.DataFrame(data = Xenc, columns = lb.classes )
        train_data.drop([str(column)], axis =1, inplace=True)
        # test_df
        Xenc = lb.transform(Xtest)
        Xtest_enc = pd.DataFrame(data = Xenc, columns = lb.classes_)
        test_data.drop([str(column)], axis =1, inplace=True)
        print('Number of classes in '+str(column)+ ' = '+ str(X_classes))
        train_data = pd.concat((train_data,Xtrain_enc),axis=1)
        test_data = pd.concat((test_data, Xtest_enc), axis=1)
    return train data, test data
In [75]:
train_data , test_data = one_hot(X_train,X_test,['X0', 'X1', 'X2', 'X3', 'X4', 'X5', 'X
6', 'X8'])
Number of classes in X0 = 47
Number of classes in X1 = 27
Number of classes in X2 = 44
Number of classes in X3 = 7
Number of classes in X4 = 4
Number of classes in X5 = 29
Number of classes in X6 = 12
Number of classes in X8 = 25
In [83]:
pca = PCA(random state=21)
pca.fit(train data)
pca train = pca.transform(train data)
In [84]:
X = pca train
y = y_train
X.shape, y.shape
```

```
Out[84]:
```

```
((4209, 551), (4209,))
```

In [98]:

```
#building model

from sklearn.linear_model import LinearRegression

model_lr=LinearRegression()
model_lr=LinearRegression.fit(model_lr,X,y)
model_lr.fit(X,y)
```

Out[98]:

LinearRegression()

from sklearn.linear_model import LinearRegression

model Ir=LinearRegression() model Ir=LinearRegression.fit(model Ir,X,y) model Ir.fit()

In [99]:

```
print ("The intercept of the LR model is: ", model_lr.intercept_ )
```

The intercept of the LR model is: 100.67076162730552

```
In [100]:
```

print ("The regression coefficient value for all the features are : ", model_lr.coef_)

```
The regression coefficient value for all the features are : [-1.19791113e
+00 2.80540466e+00 1.42630589e+00 8.42887878e-01
 2.61870015e+00 -3.08058000e+00 3.02907097e+00 -8.22961450e-01
 2.71383882e+00 -4.27813202e+00 2.54132225e+00 1.28410299e+00
 1.01693833e+00 1.99650437e-01 1.33830146e+00 -1.32973552e-01
 6.56755276e-01 2.26468833e-01 -1.16097629e+00 -1.53115573e+00
 1.29875479e+00 8.32181424e-02 7.35318359e-01 -6.12625271e-01
 -7.30581310e-01 -7.11957425e-01 -5.14735196e-01 -2.95661896e-01
 -1.42307436e+00 8.36950533e-01 -9.24282908e-01 -5.40952764e-01
 -2.33931690e-01 7.01770373e-02 9.60962921e-02 -8.35632108e-01
 9.37783778e-01 -1.44660750e+00 3.01958621e-03 1.79151818e-01
 1.02480870e+00 1.63362779e+00 -2.09326654e+00 -1.08415608e+00
 -1.23397484e-01 3.42372015e-01 3.09955835e-01 -1.21379673e+00
 -1.63088235e+00 -2.83592343e-02 2.73846179e+00 3.50950524e-01
 1.52844778e+00 -4.53091599e-01 1.55177417e+00 -3.42069902e-01
 1.36710104e+00 2.41039617e+00 -5.07413931e-02 1.59504043e+00
 -1.29199222e-01 1.36340420e+00 -1.81057096e-01 1.02164443e+00
 -1.62224542e+00 3.37068520e-01 -1.06990264e+00 2.37138027e-01
 1.85196322e-01 -1.61215179e-02 -3.60828090e-01 1.88732010e+00
 -1.79782967e-01 4.12670020e-01 -2.53285836e-01 7.24768807e-01
 1.94126035e+00 -2.18742859e+00 -1.08629860e+00 2.12512471e+00
 1.55364723e+00 1.03775845e+00 2.78958406e-01 -8.00408661e-01
 8.40508571e-01 -1.08156480e+00 5.50154392e-01 2.16198257e+00
 -2.32791573e-01 3.06322062e+00 6.47464639e-01 -1.00441024e+00
 2.80751541e-01 -1.17119125e+00 2.47721713e-01 -8.97610545e-01
 -3.23539402e+00 -1.20605617e+00 -2.40303870e-01 -6.81703314e-02
 1.82063641e+00 1.06197495e+00 1.93358825e+00 1.25684143e+00
 2.38713011e-01 2.51662797e+00 -1.19384849e+00 6.98761073e-01
 6.43716402e-01 7.41222233e-01 -1.21191700e+00 -3.80670905e-01
 -1.04730299e+00 1.32933420e+00 -2.74220653e+00 1.44471893e+00
 -1.70490892e+00 -1.59384197e+00 1.81662606e+00 1.58768694e+00
 6.56452965e-01 -3.13625486e+00 2.01773310e+00 8.16467851e-01
 1.30450267e+00 1.22217549e+00 1.60097610e+00 -1.33897626e+00
 -1.59502731e+00 -5.22662401e-01 2.33198291e+00 -5.47145605e-01
 -1.53972930e+00 1.45429276e-01 -2.29708426e-01 -2.05565691e+00
 -1.61065027e+00 -1.42469552e+00 -2.87517115e-01 -1.97023511e-01
 -1.27771953e+00 5.65739125e-02 -7.30202816e-01 2.16371375e+00
 -1.50691003e+00 -1.15528275e+00 -1.18045390e+00 -2.53295102e-02
 6.32417172e-01 -2.37735528e+00 8.95146258e-01 -6.03295028e-01
 -1.93691450e+00 -9.36172264e-01 3.01427381e+00 1.06890815e+00
 -1.31942675e-01 -1.19868035e-01 -2.46737283e-01 3.44126076e-01
 -3.42703946e-01 1.74482211e-01 1.32662182e+00 4.04559418e+00
 3.03557156e+00 1.40743700e+00 -1.39618292e-02 3.70901644e-01
 -8.54687855e-01 -1.59320951e-01 1.02932158e+00 -4.68323682e-01
 7.80954808e-01 -8.20936352e-01 8.13281029e-01 -9.16840270e-01
 -4.64952044e-01 -1.54617330e+00 -3.77222925e-01 -1.80948744e-01
 8.20128962e-01 -3.45088924e+00 -1.59114963e+00 -2.68932020e+00
 -1.68145949e+00 4.07977611e-01 1.34790567e+00 -9.47992206e-02
 1.54268998e+00 -1.59677309e+00 -6.49481609e-01 2.35050168e+00
 -1.40591367e+00 -5.43945760e-01 -1.13485505e+00 -1.14969183e+00
 1.02162373e+00 1.58740067e+00 1.11032015e+00 3.52728063e+00
 -1.23566420e+00 1.26533225e+00 -6.36682376e-01 -8.13181758e-01
 2.78104162e+00 4.94690489e-01 -1.30501702e+00 -7.12802559e-01
 3.32903787e+00 -1.20390844e+00 -2.35618837e-01 -4.48440795e+00
 -1.91257552e+00 -1.87197629e+00 4.80822623e-01 -4.64672387e-01
 1.64412731e+00 4.06077254e+00 4.40325354e+00 1.07774422e+00
 -5.12655698e-01 -8.65945637e-01 8.40271264e-01 -3.41412866e+00
 -4.99867272e-01 -6.91178218e-01 1.86030968e+00 -1.07217279e+00
 3.17163109e+00 -2.57625908e+00 -6.53686172e+00 3.78948643e+00
 2.33433023e-02 4.50386670e+00 2.39071301e+00 -4.79748015e+00
 1.55308500e+00 -2.55765507e+00 -6.60296461e+00 -4.80899102e+00
```

```
-6.21957477e+00 -1.47079585e+00
                                2.16042948e+00 -1.36476350e+00
3.96613565e+00 -2.59618655e+00
                                2.32210360e+00 -5.41820094e-01
1.33399767e+00 -1.14755282e+00
                                3.21068585e+00 -4.08883527e+00
-1.86363447e+00 -3.94388474e+00
                                3.20821227e+00 1.40307081e+00
8.19641262e-01 -3.38603672e+00 -2.22831190e-01 -3.92824650e+00
-1.48885414e+00
                2.09384993e+00
                               3.81948087e+00
                                               2.08750119e+00
                8.23322460e-01 -5.26706934e-01 -4.46854739e+00
-3.65762046e+00
-3.48423302e+00 5.13791330e+00 9.59837094e-01
                                               5.19175007e+00
-9.49122578e-01
                2.97482193e-01 4.80868736e+00 4.62334736e+00
-8.22674949e-01
                1.28595321e+00 6.32141329e+00 -4.63373111e+00
2.53350756e+00
                4.73380201e+00
                                1.80059344e-01
                                               4.53987479e-01
2.22159477e+00
                9.37635258e-01 -2.01101375e+00
                                               1.60566569e+00
4.18354884e+00 -3.40217448e+00 -2.16314679e+00 -3.88354734e+00
                6.79544166e-01 1.80983918e+00
2.22832598e+00
                                               4.59174821e+00
1.15714635e+01 -2.23505629e+00 4.39157748e+00 -5.43631221e+00
-8.83365501e+00 1.38521753e+00 7.93257652e+00 4.59383732e+00
-1.90007402e+00 6.08095878e+00 -4.39958085e+00 -3.24487975e+00
8.04973948e+00
                4.07853729e+00 -2.14321952e+00
                                                2.12288254e+00
-1.62044692e+00 1.17287415e+00 -1.72821850e+00 6.61434725e-01
                1.05900749e+01 1.40923712e+00 -3.71034936e-01
-1.67033479e+00
                6.02726701e+00 -4.34990180e+00 -3.34662411e+00
-7.98241258e+00
7.77704047e+00 -6.73281066e-01 2.98816143e+00 -4.30591932e+00
               5.28998487e+00 1.52100410e+01 -2.93376287e+00
7.42114252e+00
-6.34294893e+00
                7.37053864e-01 -1.01063713e+00 2.49707443e+00
                3.92134992e+00 -4.92231211e+00 -3.60965999e+00
-3.14743311e+00
                9.57282065e+00 1.54959997e+00 -6.26024120e-01
-7.55576031e+00
-8.86012056e+00 -4.25985927e+00 1.17879420e+01 -3.73873290e+00
-1.05374908e+01 4.37304385e+00 1.47181139e+00 -1.14553122e+01
9.62784166e+00
                1.18250773e+00 8.19524495e+00 -1.46877369e+01
-2.84240710e+00 -5.25622947e+00 -5.60639134e+00
                                               1.57834619e+00
                2.27652445e+01 -1.71299726e+00 4.07017475e+00
-8.57029893e-02
2.61078352e+00
                7.91817948e+00 3.20857935e+00 -8.21022280e+00
-9.86444704e+00 -4.13923772e+00 9.94087586e+00 -1.82374024e+00
-1.19536734e+01 -3.07280464e+01 -1.13374414e+01 9.07592492e+00
-1.02274664e+01 4.57934022e+01 -1.54188892e+01
                                               1.14666146e+00
-1.68025633e+12
                1.02080765e+12 -2.45548652e+12 -3.95641690e+11
9.35046259e+10 -1.26487175e+12 -1.73031524e+12
                                               1.99378419e+11
8.33895871e+11 2.17105050e+12 -6.29412997e+12 4.59217610e+12
-1.01163852e+13 3.21471266e+11 1.04248142e+13 4.98218021e+12
2.51482044e+12 -1.22656550e+12 -1.92292312e+11 -2.89657221e+11
5.73875790e+12 2.94224144e+12 -8.72572367e+11
                                                5.44486693e+11
1.28018435e+12
                1.63301696e+12 2.34082711e+12
                                                1.30885387e+12
                3.26148479e+11 9.67879404e+11
-1.10211251e+12
                                                1.80340473e+12
-2.22092978e+11 -8.71174788e+11
                               7.81783674e+11 -6.18245421e+11
-4.43119314e+11 -4.18471186e+11 -1.01390580e+12 -1.61212802e+12
1.08970759e+11 7.66451214e+11 -1.91719916e+11 -1.38409982e+12
                3.65523086e+11 4.88176456e+12 -5.02429878e+11
1.32983710e+11
                5.91912591e+11 7.68206266e+11 -8.04202204e+11
5.01961966e+11
-1.15243363e+12
                3.81356195e+11 6.62941512e+11 -4.82458947e+11
3.81130262e+11 -5.49480606e+11 9.39175426e+11
                                               1.01456547e+11
5.67712519e+11 -4.61588203e+11
                               8.56949944e+11 -5.42750425e+11
-5.87913109e+11 -2.97510521e+11
                                5.98304175e+11
                                               1.35214491e+12
-1.15656456e+11 -9.54035080e+11
                                7.76742802e+11
                                                4.75683827e+11
6.70457208e+11 -3.68254436e+11
                                5.09673055e+11
                                                9.20778805e+11
3.88794919e+12
                3.37859270e+11
                                3.82002689e+11
                                                5.67637951e+11
7.38715310e+11
                2.01598598e+10
                                8.89480317e+11 -4.79769081e+11
-7.43410638e+11
                1.32017987e+12 -4.20189685e+11
                                                4.33243406e+11
-3.74956884e+11 -1.13794407e+11
                               8.04155506e+10
                                                4.33560041e+11
-7.07134401e+10 -2.57934669e+11
                                7.30354920e+11
                                                4.84078245e+11
1.96934635e+12 -1.56223892e+12
                                5.88199795e+10
                                                2.20638690e+11
-1.42193150e+11 1.64642584e+11 6.49114868e+11
                                               1.12686224e+12
```

```
1.01238619e+12 -6.32501653e+11 -5.63564976e+11 2.75881181e+11
2.54778454e+10 6.49168580e+11 -6.34285829e+11 4.84079943e+11
6.13763203e+11 -2.85237091e+11 1.06914817e+12 -3.14052948e+11
4.68947111e+11 2.97884616e+11 3.95760315e+11 3.77570777e+11
-8.20788045e+11 1.04222508e+12 1.63161778e+11 4.85059525e+10
2.82303169e+11 -1.27841621e+11 4.63923449e+11 -6.06685609e+11
-9.92305221e+10 -1.29838348e+10 1.06797038e+12
                                               1.23034916e+11
-2.84265529e+11 2.89368209e+11 -3.81063180e+09 2.61104090e+11
-1.30671533e+11 3.22083272e+11 -1.32358010e+11 5.02536997e+11
-7.01286354e+10 4.84552818e+11 8.62692010e+11 -3.16324822e+10
3.61337559e+11 2.41445491e+11 6.00544721e+11 -6.24770220e+11
-7.50406137e+11 -2.43275686e+11 -5.12939675e+11 -2.47046754e+11
4.61493319e+11 -1.13878501e+11 8.93694917e+11 -3.18432879e+10
6.37367452e+11 9.59351506e+11 9.47609257e+11 1.17859581e+12
-5.98153270e+11 -5.17609845e+11 4.39661275e+11 -8.95032814e+11
-6.52475477e+11 -2.44781726e+12 8.91263044e+11 3.09405734e+12
4.09873042e+11 -3.40864925e+10 -1.29387836e+12]
```

In [102]:

```
#eveluate the regression modcel
y_pred_test= model_lr.predict(X)
```

In [103]:

```
#compare preditced value by actual
pd.DataFrame({'Actual y_test': y_test, 'Predicted y_test': y_pred_test})
```

Out[103]:

	Actual y_test	Predicted y_test
0	1	130.804612
1	2	88.524436
2	3	74.658690
3	4	82.216194
4	5	78.016202
4204	8410	109.012347
4205	8411	107.187696
4206	8413	110.938100
4207	8414	89.036535
4208	8416	95.319629

In []:

4209 rows × 2 columns

In [105]:

```
from sklearn.metrics import mean_squared_error
import numpy as np
print('RMSE value of testion dataset')
print (np.sqrt(mean_squared_error(y_test,y_pred_test)))
```

RMSE value of testion dataset 4771.643063124692

In [145]:

#XGBoost

xg_reg=xgb.XGBRegressor(objective='reg:linear', colsample_bytree=0.3,learning_rate=0.1,
max_depth=5,alpha=10,n_estimator=10)

In [109]:

```
xg_reg.fit(X,y)
```

[18:59:28] WARNING: C:/Users/Administrator/workspace/xgboost-win64_release _1.2.0/src/objective/regression_obj.cu:174: reg:linear is now deprecated in favor of reg:squarederror.

[18:59:29] WARNING: C:\Users\Administrator\workspace\xgboost-win64_release _1.2.0\src\learner.cc:516:

Parameters: { n_estimator } might not be used.

This may not be accurate due to some parameters are only used in languag e bindings but

passed down to XGBoost core. Or some parameters are not used but slip t hrough this

verification. Please open an issue if you find above cases.

[18:59:34] WARNING: C:/Users/Administrator/workspace/xgboost-win64_release _1.2.0/src/objective/regression_obj.cu:174: reg:linear is now deprecated in favor of reg:squarederror.

Out[109]:

In []:

In []:		
T []		
In []:		
In []:		