

```
In [1]: import numpy as np
import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn import preprocessing
```

```
In [2]: train_df = pd.read_csv('C:\\Users\\my\\Desktop\\train_test mercedes\\train.csv')
test_df = pd.read_csv('C:\\Users\\my\\Desktop\\train_test mercedes\\test.csv')
```

```
In [3]: print(train_df.shape)
print(train_df.columns)

(4209, 378)
Index(['ID', 'y', 'X0', 'X1', 'X2', 'X3', 'X4', 'X5', 'X6', 'X8',
      ...,
      'X375', 'X376', 'X377', 'X378', 'X379', 'X380', 'X382', 'X383', 'X384',
      'X385'],
      dtype='object', length=378)
```

```
In [4]: print(test_df.shape)
print(test_df.columns)

(4209, 377)
Index(['ID', 'X0', 'X1', 'X2', 'X3', 'X4', 'X5', 'X6', 'X8', 'X10',
      ...,
      'X375', 'X376', 'X377', 'X378', 'X379', 'X380', 'X382', 'X383', 'X384',
      'X385'],
      dtype='object', length=377)
```

```
In [6]: train_df.head()
```

Out[6]:

	ID	y	X0	X1	X2	X3	X4	X5	X6	X8	...	X375	X376	X377	X378	X379	X380	X382	X383	X384	X385
0	0	130.81	k	v	at	a	d	u	j	o	...	0	0	1	0	0	0	0	0	0	0
1	6	88.53	k	t	av	e	d	y	l	o	...	1	0	0	0	0	0	0	0	0	0
2	7	76.26	az	w	n	c	d	x	j	x	...	0	0	0	0	0	0	1	0	0	0
3	9	80.62	az	t	n	f	d	x	l	e	...	0	0	0	0	0	0	0	0	0	0
4	13	78.02	az	v	n	f	d	h	d	n	...	0	0	0	0	0	0	0	0	0	0

5 rows × 378 columns

```
In [7]: test_df.head()
```

Out[7]:

	ID	X0	X1	X2	X3	X4	X5	X6	X8	X10	...	X375	X376	X377	X378	X379	X380	X382	X383	X384	X385
0	1	az	v	n	f	d	t	a	w	0	...	0	0	0	1	0	0	0	0	0	0
1	2	t	b	ai	a	d	b	g	y	0	...	0	0	1	0	0	0	0	0	0	0
2	3	az	v	as	f	d	a	j	j	0	...	0	0	0	1	0	0	0	0	0	0
3	4	az	l	n	f	d	z	l	n	0	...	0	0	0	1	0	0	0	0	0	0
4	5	w	s	as	c	d	y	i	m	0	...	1	0	0	0	0	0	0	0	0	0

5 rows × 377 columns

```
In [9]: train_df.describe()
```

Out[9]:

	ID	y	X10	X11	X12	X13	X14	X15	X16	X17	...
count	4209.000000	4209.000000	4209.000000	4209.0	4209.000000	4209.000000	4209.000000	4209.000000	4209.000000	4209.000000	...
mean	4205.960798	100.669318	0.013305	0.0	0.075077	0.057971	0.428130	0.000475	0.002613	0.007603	...
std	2437.608688	12.679381	0.114590	0.0	0.263547	0.233716	0.494867	0.021796	0.051061	0.086872	...
min	0.000000	72.110000	0.000000	0.0	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	...
25%	2095.000000	90.820000	0.000000	0.0	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	...
50%	4220.000000	99.150000	0.000000	0.0	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	...
75%	6314.000000	109.010000	0.000000	0.0	0.000000	0.000000	1.000000	0.000000	0.000000	0.000000	...
max	8417.000000	265.320000	1.000000	0.0	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	...

8 rows × 370 columns

```
In [10]: test_df.describe()
```

Out[10]:

	ID	X10	X11	X12	X13	X14	X15	X16	X17	X18	..
count	4209.000000	4209.000000	4209.000000	4209.000000	4209.000000	4209.000000	4209.000000	4209.000000	4209.000000	4209.000000	..
mean	4211.039202	0.019007	0.000238	0.074364	0.061060	0.427893	0.000713	0.002613	0.008791	0.010216	..
std	2423.078926	0.136565	0.015414	0.262394	0.239468	0.494832	0.026691	0.051061	0.093357	0.100570	..
min	1.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	..
25%	2115.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	..
50%	4202.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	..
75%	6310.000000	0.000000	0.000000	0.000000	0.000000	1.000000	0.000000	0.000000	0.000000	0.000000	..
max	8416.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	..

8 rows × 369 columns

In [12]:

train_df.var()

C:\Users\my\AppData\Local\Temp\ipykernel_3416\57518514.py:1: FutureWarning: The default value of numeric_only in DataFrame.var is deprecated. In a future version, it will default to False. In addition, specifying 'numeric_only=None' is deprecated. Select only valid columns or specify the value of numeric_only to silence this warning.
train_df.var()

Out[12]:

ID5.941936e+06
y1.607667e+02
X101.313092e-02
X110.000000e+00
X126.945713e-02
...
X3808.014579e-03
X3827.546747e-03
X3831.660732e-03
X3844.750593e-04
X3851.423823e-03
Length: 370, dtype: float64

In [14]:

train_df.var()== 0

C:\Users\my\AppData\Local\Temp\ipykernel_3416\2706582300.py:1: FutureWarning: The default value of numeric_only in DataFrame.var is deprecated. In a future version, it will default to False. In addition, specifying 'numeric_only=None' is deprecated. Select only valid columns or specify the value of numeric_only to silence this warning.
train_df.var()== 0

Out[14]:

IDFalse
yFalse
X10False
X11True
X12False
...
X380False
X382False
X383False
X384False
X385False
Length: 370, dtype: bool

In [15]:

(train_df.var()== 0).values

C:\Users\my\AppData\Local\Temp\ipykernel_3416\3623485227.py:1: FutureWarning: The default value of numeric_only in DataFrame.var is deprecated. In a future version, it will default to False. In addition, specifying 'numeric_only=None' is deprecated. Select only valid columns or specify the value of numeric_only to silence this warning.
(train_df.var()== 0).values

[illegible]

```
In [28]: train_df.isnull().any()
```

```
Out[28]: y      False
          X0      False
          X1      False
          X2      False
          X3      False
          ...
          X380    False
          X382    False
          X383    False
          X384    False
          X385    False
          Length: 365, dtype: bool
```

```
In [29]: train_df.nunique()
```

```
Out[29]: y      2545
         X0       47
         X1       27
         X2       44
         X3        7
         ...
         X380      2
         X382      2
         X383      2
         X384      2
         X385      2
         Length: 365, dtype: int64
```

```
In [30]: obj_dtype = train_df.select_dtypes(include=[object])
obj_dtype
```

Out[30]:	X0	X1	X2	X3	X4	X5	X6	X8
0	k	v	at	a	d	u	j	o
1	k	t	av	e	d	y	l	o
2	az	w	n	c	d	x	j	x
3	az	t	n	f	d	x	l	e
4	az	v	n	f	d	h	d	n
...
4204	ak	s	as	c	d	aa	d	q
4205	j	o	t	d	d	aa	h	h
4206	ak	v	r	a	d	aa	g	e
4207	al	r	e	f	d	aa	l	u
4208	z	r	ae	c	d	aa	g	w

4209 rows × 8 columns

```
In [33]: label_encoder = preprocessing.LabelEncoder()
train df['X0'].unique()
```

```
Out[33]: array(['k', 'az', 't', 'al', 'o', 'w', 'j', 'h', 's', 'n', 'ay', 'f', 'x',  
               'y', 'aj', 'ak', 'am', 'z', 'q', 'at', 'ap', 'v', 'af', 'a', 'e',  
               'ai', 'd', 'aq', 'c', 'aa', 'ba', 'as', 'i', 'r', 'b', 'ax', 'bc',  
               'u', 'ad', 'au', 'm', 'l', 'aw', 'ao', 'ac', 'g', 'ab'],  
              dtype=object)
```

```
In [34]: train_df['X0'] = label_encoder.fit_transform(train_df['X0'])
train_df['X0'].unique()
```

```
Out[34]: array([32, 20, 40,  9, 36, 43, 31, 29, 39, 35, 19, 27, 44, 45,  7,  8, 10,
        46, 37, 15, 12, 42,  5,  0, 26,  6, 25, 13, 24,  1, 22, 14, 30, 38,
        21, 18, 23, 41,  4, 16, 34, 33, 17, 11,  3, 28,  2])
```

```
In [37]: train_df['X1'] = label_encoder.fit_transform(train_df['X1'])
train_df['X2'] = label_encoder.fit_transform(train_df['X2'])
train_df['X3'] = label_encoder.fit_transform(train_df['X3'])
train_df['X4'] = label_encoder.fit_transform(train_df['X4'])
train_df['X5'] = label_encoder.fit_transform(train_df['X5'])
train_df['X6'] = label_encoder.fit_transform(train_df['X6'])
train_df['X8'] = label_encoder.fit_transform(train_df['X8'])
```

```
In [38]: train_df.head()
```

```
Out[38]:
```

	y	X0	X1	X2	X3	X4	X5	X6	X8	X10	...	X375	X376	X377	X378	X379	X380	X382	X383	X384	X385
0	130.81	32	23	17	0	3	24	9	14	0	...	0	0	1	0	0	0	0	0	0	0
1	88.53	32	21	19	4	3	28	11	14	0	...	1	0	0	0	0	0	0	0	0	0
2	76.26	20	24	34	2	3	27	9	23	0	...	0	0	0	0	0	0	1	0	0	0
3	80.62	20	21	34	5	3	27	11	4	0	...	0	0	0	0	0	0	0	0	0	0
4	78.02	20	23	34	5	3	12	3	13	0	...	0	0	0	0	0	0	0	0	0	0

5 rows × 365 columns

```
In [39]: from sklearn.decomposition import PCA
```

```
In [41]: ## PCA with 95%
skl_pca = PCA(n_components= 0.95)
```

```
In [42]: skl_pca.fit(train_df)
```

```
Out[42]:
```

PCA

PCA(n_components=0.95)

```
In [50]: x_train_trans = skl_pca.transform(train_df)
```

```
In [51]: x_train_trans.shape
```

```
Out[51]: (4209, 6)
```

```
In [49]: ### pca with 98%
pca_98 = PCA(n_components = 0.98)
```

```
In [52]: pca_98.fit(train_df)
```

```
Out[52]:
```

PCA

PCA(n_components=0.98)

```
In [53]: x_pca_98 = pca_98.transform(train_df)
print(x_pca_98.shape)
(4209, 12)
```

```
In [55]: train_df.y
```

```
Out[55]:
```

0	130.81
1	88.53
2	76.26
3	80.62
4	78.02
...	
4204	107.39
4205	108.77
4206	109.22
4207	87.48
4208	110.85

Name: y, Length: 4209, dtype: float64

```
In [57]: x= train_df.drop('y',axis = 1)
y = train_df.y
```

```
xtrain,xtest,ytrain,ytest= train_test_split(x,y,test_size=0.3,random_state = 42)
```

```
In [58]: print(xtrain)
print(xtrain.shape)
```

	X0	X1	X2	X3	X4	X5	X6	X8	X10	X12	...	X375	X376	X377	X378	\
370	35	13	16	1	3	9	6	19	0	0	...	0	0	0	0	
3392	15	10	16	2	3	23	9	16	0	0	...	0	0	1	0	
2208	31	3	16	2	3	15	2	21	0	0	...	0	0	1	0	
3942	35	20	8	6	3	26	6	14	0	1	...	1	0	0	0	
1105	36	13	16	5	3	1	6	0	0	0	...	0	0	0	0	
...	
3444	31	10	16	2	3	22	11	17	0	0	...	0	0	1	0	
466	20	25	25	2	3	9	9	9	0	0	...	0	0	0	0	
3092	45	24	3	2	3	21	8	2	0	0	...	1	0	0	0	
3772	45	19	8	5	3	25	8	1	0	1	...	0	0	0	0	
860	22	1	7	2	3	5	9	17	0	0	...	1	0	0	0	
	X379	X380	X382	X383	X384	X385										
370	0	0	0	0	0	0										
3392	0	0	0	0	0	0										
2208	0	0	0	0	0	0										
3942	0	0	0	0	0	0										
1105	0	0	0	0	0	0										
...										
3444	0	0	0	0	0	0										
466	0	0	1	0	0	0										
3092	0	0	0	0	0	0										
3772	0	0	0	0	0	0										
860	0	0	0	0	0	0										

```
[2946 rows x 364 columns]
(2946, 364)
```

```
In [59]: print(ytrain)
print(ytest.shape)
```

370	95.13
3392	117.36
2208	109.01
3942	93.77
1105	103.41
...	
3444	109.42
466	78.25
3092	92.18
3772	91.92
860	87.71

```
Name: y, Length: 2946, dtype: float64
(1263,)
```

```
In [60]: print(xtest)
print(xtest.shape)
```

	X0	X1	X2	X3	X4	X5	X6	X8	X10	X12	...	X375	X376	X377	X378	\
1073	9	16	7	5	3	6	9	11	0	0	...	0	0	0	0	
144	27	13	3	5	3	13	8	22	0	0	...	0	0	0	0	
2380	31	1	21	2	3	18	11	14	1	0	...	1	0	0	0	
184	20	25	22	2	3	13	9	11	0	0	...	0	0	0	0	
2587	8	23	8	3	3	17	8	17	0	0	...	0	0	0	0	
...	
2493	27	20	16	2	3	18	10	5	0	0	...	0	0	1	0	
3388	40	19	24	5	3	23	3	19	0	0	...	0	0	0	0	
3997	22	3	7	0	3	26	6	18	0	0	...	0	0	1	0	
383	40	1	16	6	3	9	8	0	0	0	...	1	0	0	0	
3364	27	4	33	2	3	23	6	24	0	0	...	0	0	1	0	
	X379	X380	X382	X383	X384	X385										
1073	0	0	0	0	0	0										
144	0	0	0	0	0	0										
2380	0	0	0	0	0	0										
184	0	0	1	0	0	0										
2587	0	0	0	0	0	0										
...										
2493	0	0	0	0	0	0										
3388	0	0	0	0	0	0										
3997	0	0	0	0	0	0										
383	0	0	0	0	0	0										
3364	0	0	0	0	0	0										

```
[1263 rows x 364 columns]
(1263, 364)
```

```
In [64]: pca_xtrain= PCA(n_components =0.95)
```

```
pca_xtrain.fit(xtrain)
```

Out[64]:

```
PCA
PCA(n_components=0.95)
```

```
In [65]: pca_xt_trans=pca_xtrain.transform(xtrain)
print(pca_xt_trans.shape)
```

(2946, 6)

```
In [67]: pca_xtest= PCA(n_components =0.95)
pca_xtest.fit(xtest)
```

Out[67]:

```
PCA
PCA(n_components=0.95)
```

```
In [69]: pca_xtest_trans= pca_xtest.transform(xtest)
print(pca_xtest_trans.shape)
```

(1263, 6)

```
In [71]: print(pca_xtest.explained_variance_)
print(pca_xtest.explained_variance_ratio_)
```

```
[206.79524961 120.24273955  67.64680756  61.94375666  48.08214872
  8.7271811 ]
[0.38517942 0.22396563 0.12599979 0.11537722 0.08955841 0.01625536]
```

```
In [72]: test_df
```

Out[72]:

	ID	X0	X1	X2	X3	X4	X5	X6	X8	X10	...	X375	X376	X377	X378	X379	X380	X382	X383	X384	X385
0	1	az	v	n	f	d	t	a	w	0	...	0	0	0	1	0	0	0	0	0	0
1	2	t	b	ai	a	d	b	g	y	0	...	0	0	1	0	0	0	0	0	0	0
2	3	az	v	as	f	d	a	j	j	0	...	0	0	0	1	0	0	0	0	0	0
3	4	az	l	n	f	d	z	l	n	0	...	0	0	0	1	0	0	0	0	0	0
4	5	w	s	as	c	d	y	i	m	0	...	1	0	0	0	0	0	0	0	0	0
...
4204	8410	aj	h	as	f	d	aa	j	e	0	...	0	0	0	0	0	0	0	0	0	0
4205	8411	t	aa	ai	d	d	aa	j	y	0	...	0	1	0	0	0	0	0	0	0	0
4206	8413	y	v	as	f	d	aa	d	w	0	...	0	0	0	0	0	0	0	0	0	0
4207	8414	ak	v	as	a	d	aa	c	q	0	...	0	0	1	0	0	0	0	0	0	0
4208	8416	t	aa	ai	c	d	aa	g	r	0	...	1	0	0	0	0	0	0	0	0	0

4209 rows × 377 columns

```
In [73]: test_obj_dtype = test_df.select_dtypes(include = [object])
test_obj_dtype
```

Out[73]:

	X0	X1	X2	X3	X4	X5	X6	X8
0	az	v	n	f	d	t	a	w
1	t	b	ai	a	d	b	g	y
2	az	v	as	f	d	a	j	j
3	az	l	n	f	d	z	l	n
4	w	s	as	c	d	y	i	m
...
4204	aj	h	as	f	d	aa	j	e
4205	t	aa	ai	d	d	aa	j	y
4206	y	v	as	f	d	aa	d	w
4207	ak	v	as	a	d	aa	c	q
4208	t	aa	ai	c	d	aa	g	r

4209 rows × 8 columns

```
In [75]: test_df['X0']=label_encoder.fit_transform(test_df['X0'])
test_df['X1']=label_encoder.fit_transform(test_df['X1'])
test_df['X2']=label_encoder.fit_transform(test_df['X2'])
test_df['X3']=label_encoder.fit_transform(test_df['X3'])
```

```
test_df['X4']=label_encoder.fit_transform(test_df['X4'])
test_df['X5']=label_encoder.fit_transform(test_df['X5'])
test_df['X6']=label_encoder.fit_transform(test_df['X6'])
test_df['X8']=label_encoder.fit_transform(test_df['X8'])
```

```
In [76]: print(test_df)
print(test_df.shape)
```

```

      ID  X0  X1  X2  X3  X4  X5  X6  X8  X10  ...  X375  X376  X377  X378  \
0      1  21  23  34   5   3  26   0  22   0  ...   0     0     0     1
1      2  42   3   8   0   3   9   6  24   0  ...   0     0     1     0
2      3  21  23  17   5   3   0   9   9   0  ...   0     0     0     1
3      4  21  13  34   5   3  31  11  13   0  ...   0     0     0     1
4      5  45  20  17   2   3  30   8  12   0  ...   1     0     0     0
...
4204  8410   6   9  17   5   3   1   9   4   0  ...   0     0     0     0
4205  8411  42   1   8   3   3   1   9  24   0  ...   0     1     0     0
4206  8413  47  23  17   5   3   1   3  22   0  ...   0     0     0     0
4207  8414   7  23  17   0   3   1   2  16   0  ...   0     0     1     0
4208  8416  42   1   8   2   3   1   6  17   0  ...   1     0     0     0

```

```

      X379  X380  X382  X383  X384  X385
0         0     0     0     0     0     0
1         0     0     0     0     0     0
2         0     0     0     0     0     0
3         0     0     0     0     0     0
4         0     0     0     0     0     0
...
4204     0     0     0     0     0     0
4205     0     0     0     0     0     0
4206     0     0     0     0     0     0
4207     0     0     0     0     0     0
4208     0     0     0     0     0     0

```

```
[4209 rows x 377 columns]
(4209, 377)
```

```
In [79]: test_df = test_df.drop('ID',axis = 1)
```

```
In [80]: pca_test_df = PCA(n_components = 0.95)
pca_test_df.fit(test_df)
```

```
Out[80]: PCA
PCA(n_components=0.95)
```

```
In [81]: pca_tst_df_trans = pca_test_df.transform(test_df)
print(pca_tst_df_trans.shape)
```

```
(4209, 6)
```

```
In [82]: print(pca_test_df.explained_variance_)
print(pca_test_df.explained_variance_ratio_)
```

```
[247.07875325 100.33535335 77.48364816 62.33258307 48.95689653
 8.14203723]
[0.43515102 0.17670897 0.13646292 0.10977912 0.08622208 0.01433962]
```

```
In [83]: y
```

```
Out[83]: 0      130.81
1       88.53
2       76.26
3       80.62
4       78.02
...
4204    107.39
4205    108.77
4206    109.22
4207     87.48
4208    110.85
Name: y, Length: 4209, dtype: float64
```

```
In [86]: from sklearn import svm
from sklearn import model_selection
import xgboost as xgb
```

```
In [ ]: model = xgb.XGBRegressor(objective="reg:linear".learning_rate=0.20)
model.fit(pca_xtrain,ytrain)
y_pred=model.predict(pca_xtest)
y_pred
model.predict(pca_test_df)
```


