

```
In [47]: import pandas as pd
...:
...: df=pd.read_csv('G:\Data Analysis\output.csv')
...: df=df.dropna()
...: X = df.iloc[:,[8,11,14,17,22,23,24,25,26,27,28,29,30,31,32,33,34,35,36,37]].values
...: y = df.iloc[:, 1].values
```

```
In [48]: from sklearn.model_selection import train_test_split
...: X_train, X_test, y_train, y_test = train_test_split(X, y, test_size = 0.3,
random_state = 0)
```

```
In [49]: from sklearn.linear_model import LinearRegression
...: regressor = LinearRegression()
...: regressor.fit(X_train, y_train)
```

```
Out[49]:
LinearRegression(copy_X=True, fit_intercept=True, n_jobs=None,
normalize=False)
```

```
In [50]: import statsmodels.formula.api as sm
...:
...: model1=sm.OLS(y_train,X_train)
...: result=model1.fit()
...: print(result.summary())
```

OLS Regression Results

```
=====
Dep. Variable:          y      R-squared:          0.718
Model:                OLS      Adj. R-squared:      0.716
Method:             Least Squares      F-statistic:      288.2
Date:                Sat, 06 Jul 2019      Prob (F-statistic):      0.00
Time:                  19:08:44      Log-Likelihood:      -4783.3
No. Observations:      2279      AIC:          9607.
Df Residuals:          2259      BIC:          9721.
Df Model:              20
Covariance Type:      nonrobust
=====
```

	coef	std err	t	P> t	[0.025	0.975]
x1	0.0043	0.006	0.759	0.448	-0.007	0.016
x2	0.0153	0.005	2.946	0.003	0.005	0.026
x3	-0.6731	0.127	-5.304	0.000	-0.922	-0.424
x4	0.0153	0.005	3.024	0.003	0.005	0.025
x5	-0.1794	0.049	-3.658	0.000	-0.276	-0.083
x6	0.1038	0.063	1.651	0.099	-0.019	0.227
x7	0.0761	0.051	1.482	0.138	-0.025	0.177
x8	-0.1075	0.045	-2.415	0.016	-0.195	-0.020
x9	-0.1070	0.054	-1.984	0.047	-0.213	-0.001
x10	0.0012	0.000	2.694	0.007	0.000	0.002
x11	3.384e-05	0.001	0.058	0.954	-0.001	0.001
x12	0.0018	0.000	4.370	0.000	0.001	0.003
x13	0.2305	0.102	2.258	0.024	0.030	0.431
x14	0.0007	0.000	2.683	0.007	0.000	0.001
x15	-0.0005	0.000	-1.460	0.145	-0.001	0.000
x16	-0.0017	0.000	-6.445	0.000	-0.002	-0.001
x17	0.0004	0.000	0.836	0.403	-0.001	0.001
x18	-0.0017	0.001	-3.310	0.001	-0.003	-0.001
x19	-0.0001	0.000	-0.277	0.782	-0.001	0.001
x20	0.0011	0.001	0.860	0.390	-0.001	0.003

```
=====
Omnibus:                3072.586    Durbin-Watson:                2.018
Prob(Omnibus):           0.000    Jarque-Bera (JB):           1000006.122
Skew:                    7.374    Prob(JB):                   0.00
Kurtosis:                104.555    Cond. No.                   2.99e+03
=====
```

Warnings:

[1] Standard Errors assume that the covariance matrix of the errors is correctly specified.
[2] The condition number is large, 2.99e+03. This might indicate that there are strong multicollinearity or other numerical problems.

```
In [51]: c=0
...: y_pred = regressor.predict(X_test)
...: for i in range(len(y_pred)):
...:     y_pred[i]=(y_pred[i]<=y_test[i]+1.5 and y_pred[i]>=y_test[i]-1.5)
...:     if(y_pred[i]):
...:         c+=1
...:
...: acc=float(c/len(y_test))
```

```
In [52]: acc
Out[52]: 0.7656090071647902
```

```
In [53]: def backwardElimination(x, sl):
...:     numVars = len(x[0])
...:     for i in range(0, numVars):
...:         regressor_OLS = sm.OLS(y, x).fit()
...:         maxVar = max(regressor_OLS.pvalues).astype(float)
...:         if maxVar > sl:
...:             for j in range(0, numVars - i):
...:                 if (regressor_OLS.pvalues[j].astype(float) == maxVar):
...:                     x = np.delete(x, j, 1)
...:             regressor_OLS.summary()
...:         return x
...:
...:
...: SL = 0.05
...: import numpy as np
...: X=np.append(arr=np.ones((3256,1)).astype(int),values =X ,axis=1)
...: X_opt = X[:,[0,1,2,3,4,5,6,7,8,9,10,11,12,13,14,15,16,17,18,19,20]]
...: X_Modeled = backwardElimination(X_opt, SL)
```

```
In [54]: from sklearn.model_selection import train_test_split
...: X_train, X_test, y_train, y_test = train_test_split(X_Modeled, y, test_size = 0.3,
random_state = 0)
...:
...: from sklearn.linear_model import LinearRegression
...: regressor = LinearRegression()
...: regressor.fit(X_train, y_train)
```

```
Out[54]:
LinearRegression(copy_X=True, fit_intercept=True, n_jobs=None,
normalize=False)
```

```
In [55]: model1=sm.OLS(y_train,X_train)
...: result=model1.fit()
...: print(result.summary())
OLS Regression Results
```

```

Dep. Variable:          y      R-squared:          0.717
Model:                OLS      Adj. R-squared:       0.716
Method:              Least Squares      F-statistic:       442.7
Date:                Sat, 06 Jul 2019      Prob (F-statistic):    0.00
Time:                19:09:15      Log-Likelihood:      -4787.0
No. Observations:      2279      AIC:                9600.
Df Residuals:          2266      BIC:                9675.
Df Model:              13
Covariance Type:      nonrobust

```

```

=====
              coef      std err          t      P>|t|      [0.025      0.975]
-----
x1              0.0183      0.004      4.358      0.000      0.010      0.027
x2             -0.6494      0.123     -5.265      0.000     -0.891     -0.408
x3              0.0162      0.005      3.364      0.001      0.007      0.026
x4             -0.1856      0.048     -3.881      0.000     -0.279     -0.092
x5              0.0700      0.053      1.333      0.183     -0.033      0.173
x6             -0.1175      0.044     -2.667      0.008     -0.204     -0.031
x7              0.0014      0.000      3.238      0.001      0.001      0.002
x8              0.0019      0.000      4.839      0.000      0.001      0.003
x9              0.2500      0.099      2.524      0.012      0.056      0.444
x10             0.0006      0.000      2.576      0.010      0.000      0.001
x11            -0.0006      0.000     -2.055      0.040     -0.001     -2.85e-05
x12            -0.0017      0.000     -6.624      0.000     -0.002     -0.001
x13            -0.0017      0.000     -3.391      0.001     -0.003     -0.001
=====

```

```

Omnibus:              3070.587      Durbin-Watson:          2.022
Prob(Omnibus):         0.000      Jarque-Bera (JB):       992958.038
Skew:                  7.368      Prob(JB):               0.00
Kurtosis:              104.191      Cond. No.               2.48e+03
=====

```

Warnings:

[1] Standard Errors assume that the covariance matrix of the errors is correctly specified.
[2] The condition number is large, 2.48e+03. This might indicate that there are strong multicollinearity or other numerical problems.

```

In [56]: c=0
...: y_pred = regressor.predict(X_test)
...: for i in range(len(y_pred)):
...:     y_pred[i]=(y_pred[i]<=y_test[i]+2 and y_pred[i]>=y_test[i]-2)
...:     if(y_pred[i]):
...:         c+=1
...:
...: accAfterBackElimination=float(c/len(y_test))

```

In [57]: accAfterBackElimination

Out[57]: 0.9068577277379734

In [58]: