# multiple-linear-regression

## November 20, 2024

```
[1]: import pandas as pd
     import numpy as np
     import matplotlib.pyplot as plt
     import seaborn as sns
     import statsmodels.formula.api as smf
     from sklearn.preprocessing import LabelEncoder
     from sklearn.model_selection import train_test_split
     from sklearn.linear_model import LinearRegression
     import warnings
     warnings.filterwarnings('ignore')
     df=pd.read_csv('/content/ToyotaCorolla - MLR.csv')
     df
[1]:
                                                      Automatic
           Price
                  Age_08_04
                                 KM Fuel_Type
                                                                    СС
                                                                        Doors
     0
           13500
                          23
                             46986
                                        Diesel
                                                 90
                                                                 2000
                                                                            3
     1
           13750
                          23
                              72937
                                        Diesel
                                                 90
                                                              0 2000
                                                                            3
     2
                          24
                                                              0 2000
                                                                            3
           13950
                              41711
                                        Diesel
                                                 90
     3
           14950
                          26
                              48000
                                        Diesel
                                                  90
                                                                 2000
                                                                            3
     4
           13750
                          30
                              38500
                                        Diesel
                                                  90
                                                                 2000
                                                                            3
                          •••
     1431
            7500
                          69
                              20544
                                        Petrol
                                                  86
                                                              0 1300
                                                                            3
     1432
           10845
                          72
                              19000
                                        Petrol
                                                  86
                                                              0 1300
                                                                            3
     1433
            8500
                          71
                              17016
                                        Petrol
                                                  86
                                                              0 1300
                                                                            3
     1434
            7250
                          70
                              16916
                                        Petrol
                                                  86
                                                              0 1300
                                                                            3
     1435
            6950
                                                                 1600
                          76
                                   1
                                        Petrol
                                                110
                                                                            5
                       Gears
           Cylinders
                              Weight
     0
                           5
                                 1165
                           5
     1
                    4
                                1165
                    4
                           5
     2
                                1165
     3
                    4
                           5
                                1165
     4
                    4
                           5
                                1170
     1431
                           5
                                1025
                    4
     1432
                    4
                           5
                                1015
                           5
     1433
                                1015
     1434
                           5
                                 1015
```

1435 4 5 1114

[1436 rows x 11 columns]

```
[2]: df.head()
```

```
[2]:
        Price Age_08_04
                             KM Fuel_Type HP
                                               Automatic
                                                             cc Doors Cylinders
     0 13500
                      23
                          46986
                                   Diesel
                                           90
                                                        0
                                                           2000
                                                                     3
                                                                                4
                          72937
                                                           2000
                                                                                4
     1 13750
                      23
                                   Diesel
                                           90
                                                        0
                                                                     3
     2 13950
                      24
                          41711
                                   Diesel
                                           90
                                                        0 2000
                                                                     3
                                                                                4
                                                           2000
                                                                     3
                                                                                4
     3 14950
                      26
                          48000
                                   Diesel
                                                        0
                                           90
                                                                     3
                                                                                4
     4 13750
                      30
                          38500
                                   Diesel
                                           90
                                                           2000
```

```
Gears
           Weight
0
        5
              1165
             1165
1
        5
2
        5
             1165
3
        5
              1165
4
        5
              1170
```

- [3]: df.shape
- [3]: (1436, 11)
- [4]: df.size
- [4]: 15796
- [5]: df.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1436 entries, 0 to 1435
Data columns (total 11 columns):

#	Column	Non-Null Count	Dtype
0	Price	1436 non-null	int64
1	Age_08_04	1436 non-null	int64
2	KM	1436 non-null	int64
3	Fuel_Type	1436 non-null	object
4	HP	1436 non-null	int64
5	Automatic	1436 non-null	int64
6	СС	1436 non-null	int64
7	Doors	1436 non-null	int64
8	Cylinders	1436 non-null	int64
9	Gears	1436 non-null	int64
10	Weight	1436 non-null	int64
dtyp	es: int64(1	0), object(1)	

memory usage: 123.5+ KB

df1

```
[6]: df.describe()
[6]:
                              Age_08_04
                                                                   HP
                                                                          Automatic
                    Price
                                                     KM
              1436.000000
                            1436.000000
                                            1436.000000
                                                          1436.000000
                                                                        1436.000000
     count
             10730.824513
                              55.947075
                                           68533.259749
                                                           101.502089
                                                                           0.055710
     mean
     std
              3626.964585
                              18.599988
                                           37506.448872
                                                            14.981080
                                                                           0.229441
     min
              4350.000000
                               1.000000
                                                            69.000000
                                                                           0.00000
                                               1.000000
     25%
                              44.000000
                                                            90.000000
             8450.000000
                                           43000.000000
                                                                           0.000000
     50%
             9900.000000
                              61.000000
                                           63389.500000
                                                           110.000000
                                                                           0.000000
     75%
             11950.000000
                              70.000000
                                           87020.750000
                                                           110.000000
                                                                           0.00000
             32500.000000
                              80.000000
                                          243000.000000
                                                           192.000000
     max
                                                                           1.000000
                                 Doors
                                        Cylinders
                                                           Gears
                                                                       Weight
                      СС
              1436.00000
                           1436.000000
                                            1436.0
                                                    1436.000000
                                                                  1436.00000
     count
              1576.85585
                              4.033426
                                               4.0
                                                        5.026462
                                                                  1072.45961
     mean
                                               0.0
     std
              424.38677
                              0.952677
                                                        0.188510
                                                                     52.64112
     min
              1300.00000
                              2.000000
                                               4.0
                                                        3.000000
                                                                  1000.00000
     25%
              1400.00000
                              3.000000
                                               4.0
                                                        5.000000
                                                                  1040.00000
     50%
                                               4.0
              1600.00000
                              4.000000
                                                        5.000000
                                                                  1070.00000
     75%
                                               4.0
                                                        5.000000
              1600.00000
                              5.000000
                                                                  1085.00000
             16000.00000
                                               4.0
                                                        6.000000
                                                                  1615.00000
     max
                              5.000000
[7]:
     df.isnull().sum()
[7]: Price
                   0
     Age_08_04
                   0
                   0
     KM
     Fuel_Type
                   0
     HP
                   0
                   0
     Automatic
     СС
                   0
                   0
     Doors
     Cylinders
                   0
     Gears
                   0
     Weight
                   0
     dtype: int64
[8]: df.duplicated().sum()
[8]: 1
[9]: df1=df.drop_duplicates(ignore_index=True)
```

```
[9]:
           Price
                   Age_08_04
                                  KM Fuel_Type
                                                   ΗP
                                                        Automatic
                                                                          Doors
                                                                      СС
            13500
                                         Diesel
                                                                    2000
     0
                           23
                               46986
                                                   90
                                                                 0
                                                                               3
                                                                 0 2000
     1
           13750
                           23
                               72937
                                         Diesel
                                                   90
                                                                               3
     2
            13950
                           24
                               41711
                                         Diesel
                                                   90
                                                                 0 2000
                                                                               3
     3
            14950
                               48000
                                         Diesel
                                                                   2000
                                                                               3
                           26
                                                   90
     4
            13750
                           30
                               38500
                                         Diesel
                                                   90
                                                                 0 2000
                                                                               3
                           •••
     1430
                                                                  1300
             7500
                           69
                               20544
                                         Petrol
                                                   86
                                                                               3
     1431
           10845
                               19000
                                         Petrol
                                                                  1300
                                                                               3
                           72
                                                   86
                                                                0 1300
     1432
             8500
                           71
                               17016
                                         Petrol
                                                   86
                                                                               3
     1433
             7250
                           70
                               16916
                                         Petrol
                                                                 0 1300
                                                                               3
                                                   86
     1434
             6950
                           76
                                         Petrol
                                                                 0 1600
                                                                               5
                                    1
                                                  110
            Cylinders
                       Gears
                               Weight
     0
                            5
                                  1165
                    4
                            5
     1
                    4
                                  1165
     2
                    4
                            5
                                  1165
     3
                    4
                            5
                                  1165
     4
                    4
                            5
                                  1170
                            •••
     1430
                    4
                            5
                                  1025
     1431
                            5
                                  1015
                    4
     1432
                    4
                            5
                                  1015
     1433
                    4
                            5
                                  1015
     1434
                    4
                            5
                                  1114
```

[1435 rows x 11 columns]

# [12]:

```
Traceback (most recent call last)
ValueError
<ipython-input-12-49b3fcfeb4d1> in <cell line: 1>()
---> 1 df1.corr()
/usr/local/lib/python3.10/dist-packages/pandas/core/frame.py in corr(self, u
 →method, min_periods, numeric_only)
  11047
                cols = data.columns
  11048
                idx = cols.copy()
> 11049
                mat = data.to_numpy(dtype=float, na_value=np.nan, copy=False)
  11050
  11051
                if method == "pearson":
/usr/local/lib/python3.10/dist-packages/pandas/core/frame.py in to_numpy(self,__

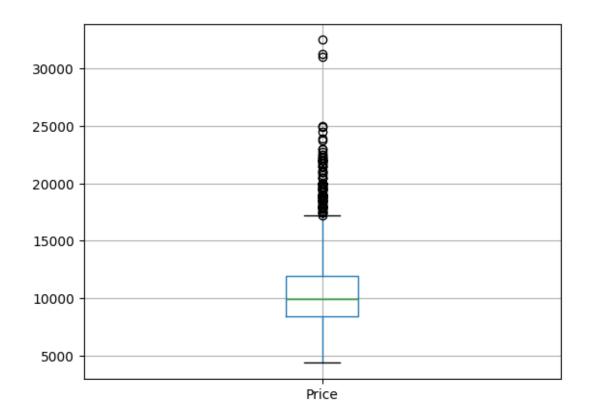
dtype, copy, na_value)

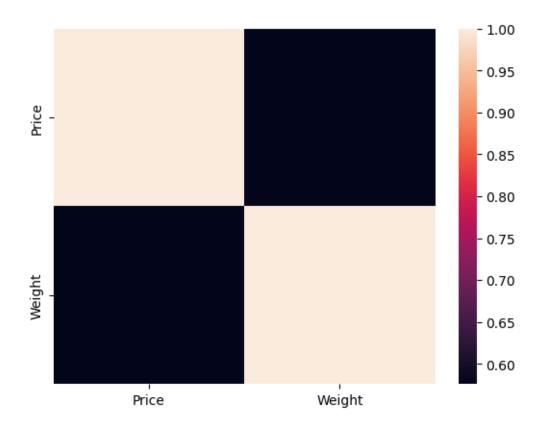
   1991
                if dtype is not None:
   1992
                    dtype = np.dtype(dtype)
```

```
-> 1993
                 result = self._mgr.as_array(dtype=dtype, copy=copy,_
  →na_value=na_value)
    1994
                 if result.dtype is not dtype:
    1995
                     result = np.asarray(result, dtype=dtype)
 /usr/local/lib/python3.10/dist-packages/pandas/core/internals/managers.py in_u
  →as_array(self, dtype, copy, na_value)
    1692
                         arr.flags.writeable = False
    1693
                 else:
 -> 1694
                     arr = self._interleave(dtype=dtype, na_value=na_value)
    1695
                     # The underlying data was copied within _interleave, so no__
  ⇔need
    1696
                     # to further copy if copy=True or setting na_value
 /usr/local/lib/python3.10/dist-packages/pandas/core/internals/managers.py in_
  →_interleave(self, dtype, na_value)
    1751
                     else:
    1752
                         arr = blk.get_values(dtype)
 -> 1753
                     result[rl.indexer] = arr
    1754
                     itemmask[rl.indexer] = 1
    1755
 ValueError: could not convert string to float: 'Diesel'
```

```
[11]: df1.boxplot('Price')
```

[11]: <Axes: >





```
[15]: lab_encoder=LabelEncoder()
[16]: df1['Fuel_Type']=lab_encoder.fit_transform(df1['Fuel_Type'])
[17]: df1
[17]:
             Price
                     Age_08_04
                                                                             Doors
                                                                                    \
                                    KM
                                        Fuel_Type
                                                      HP
                                                          Automatic
                                                                         СС
      0
             13500
                            23
                                 46986
                                                 1
                                                      90
                                                                   0
                                                                       2000
                                                                                  3
      1
             13750
                            23
                                 72937
                                                 1
                                                      90
                                                                   0
                                                                       2000
                                                                                  3
      2
                            24
                                                                       2000
                                                                                  3
             13950
                                 41711
                                                  1
                                                      90
                                                                   0
      3
             14950
                            26
                                 48000
                                                  1
                                                      90
                                                                   0
                                                                       2000
                                                                                  3
      4
                                                                       2000
                                                                                  3
             13750
                            30
                                 38500
                                                 1
                                                      90
                                                                   0
                            •••
                                                                                  3
      1430
              7500
                            69
                                 20544
                                                 2
                                                      86
                                                                   0
                                                                       1300
                                                 2
                                                                                  3
      1431
             10845
                            72
                                 19000
                                                                   0
                                                                       1300
                                                      86
      1432
              8500
                            71
                                 17016
                                                 2
                                                      86
                                                                   0
                                                                       1300
                                                                                  3
      1433
              7250
                            70
                                 16916
                                                 2
                                                      86
                                                                   0
                                                                       1300
                                                                                  3
                                                 2
                                                                                  5
      1434
              6950
                            76
                                     1
                                                     110
                                                                   0
                                                                       1600
             Cylinders
                         Gears
                                 Weight
      0
                              5
                                   1165
      1
                              5
                                   1165
```

```
2
                                 1165
                    4
                            5
      3
                            5
                                 1165
      4
                    4
                            5
                                 1170
      1430
                    4
                            5
                                 1025
      1431
                            5
                                 1015
                    4
      1432
                    4
                            5
                                 1015
      1433
                    4
                            5
                                 1015
      1434
                            5
                                 1114
      [1435 rows x 11 columns]
[19]: target=df1[['Price']]
      features=df1.drop('Price',axis=1)
      x_train,x_test,y_train,y_test=train_test_split(features,target,train_size=0.
       ⇔8,random_state=100)
      print(x_train.shape)
      print(y_train.shape)
      print(x_test.shape)
      print(y_test.shape)
     (1148, 10)
     (1148, 1)
     (287, 10)
     (287, 1)
[20]: linear_model=LinearRegression()
      linear_model.fit(x_train,y_train)
[20]: LinearRegression()
[21]: y_pred=linear_model.predict(x_test)
      y_pred
[21]: array([[15542.62555409],
             [10795.00465001],
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             [8025.28303298],
             [26136.19057591],
             [12656.08053759],
             [16262.88652499],
             [11705.43298529],
             [11427.04450319]])
[22]: linear_model.score(x_train,y_train)
[22]: 0.8626130385692874
[23]: linear_model.score(x_test,y_test)
[23]: 0.8671580177926699
```

[ 9170.92887028],

```
[24]: linear_model.intercept_
[24]: array([-8689.50713129])
[25]: linear_model.coef_
[25]: array([[-1.24537995e+02, -1.76276977e-02, 3.95421517e+02,
             2.62373819e+01, 3.94335343e+02, -3.00403510e-02,
             4.30425267e+00, 3.06954462e-12, 7.33844905e+02,
             1.91383362e+01]])
[26]: from sklearn.metrics import r2_score
     r2_score(y_test,y_pred)
[26]: 0.8671580177926699
[27]: sns.regplot(data=df1,x=y_test,y=y_pred,scatter_kws={'color':
      [27]: <Axes: xlabel='Price'>
          30000
          25000
          20000
           15000
           10000
            5000
```

15000

20000

Price

25000

30000

5000

10000

```
[28]: import statsmodels.formula.api as smf
model1=smf.

→ols('Price~Age_08_04+KM+Fuel_Type+HP+cc+Doors+Gears+Weight',data=df1).fit()
```

[29]: model1.summary()

[29]:

Dep. Variable:	Price	R-squared:	0.863
Model:	OLS	Adj. R-squared:	0.863
Method:	Least Squares	F-statistic:	1127.
Date:	Wed, 20 Nov 2024	Prob (F-statistic):	0.00
Time:	11:23:44	Log-Likelihood:	-12362.
No. Observations:	1435	AIC:	2.474e + 04
Df Residuals:	1426	BIC:	2.479e + 04
Df Model:	8		

Covariance Type: nonrobust

	$\mathbf{coef}$	$\operatorname{std}$ $\operatorname{err}$	$\mathbf{t}$	$\mathbf{P} >  \mathbf{t} $	[0.025	0.975]
Intercept	-9852.8426	1478.988	-6.662	0.000	-1.28e+04	-6951.617
${\bf Age\_08\_04}$	-123.2082	2.608	-47.237	0.000	-128.325	-118.092
$\mathbf{K}\mathbf{M}$	-0.0177	0.001	-13.494	0.000	-0.020	-0.015
$Fuel\_Type$	617.0386	145.592	4.238	0.000	331.441	902.637
$_{ m HP}$	23.2994	2.872	8.114	0.000	17.666	28.932
$\mathbf{cc}$	-0.0418	0.090	-0.465	0.642	-0.218	0.134
$\mathbf{Doors}$	-35.7894	40.222	-0.890	0.374	-114.690	43.112
Gears	582.2973	196.381	2.965	0.003	197.070	967.525
$\mathbf{Weight}$	20.9344	1.014	20.653	0.000	18.946	22.923

Omnibus:	270.225	Durbin-Watson:	1.608
Prob(Omnibus):	0.000	Jarque-Bera (JB):	2870.669
Skew:	-0.545	Prob(JB):	0.00
Kurtosis:	9.843	Cond. No.	3.29e + 06

## Notes:

- [1] Standard Errors assume that the covariance matrix of the errors is correctly specified.
- [2] The condition number is large, 3.29e+06. This might indicate that there are strong multicollinearity or other numerical problems.

[30]: model2=smf.ols('Price~cc',data=df1).fit()

[31]: model2.summary()

[31]:

Dep. Variab	le:	Price		R-square	ed:	0.015
Model:		OLS		Adj. R-s	quared:	0.015
Method:	I	Least Squar	res	F-statisti	ic:	22.52
Date:	We	ed, 20 Nov 2024		Prob (F-	2.29e-06	
Time:		11:23:58		Log-Like	lihood:	-13779.
No. Observa	tions:	1435		AIC:		2.756e + 04
Df Residuals	3:	1433		BIC:		2.757e + 04
Df Model:		1				
Covariance 7	Γype:	nonrobust	_			
	coef	std err	t	$\mathbf{P} >  \mathbf{t} $	[0.025]	0.975]
Intercept	9053.5368	363.894	24.880	0.000	8339.715	9767.359
$\mathbf{cc}$	1.0576	0.223	4.745	0.000	0.620	1.495
Omnib	us:	463.846	Durbi	n-Watso	n: 0.	.269
Prob(Omnibus):		0.000	Jarqu	e-Bera (	<b>JB):</b> 138	86.822
Skew:		1.645	<b>Prob(JB):</b> 7.17e-30			7e-302
Kurtos	is:	6.518	Cond	No.	6.28	8e + 03

#### Notes:

- [1] Standard Errors assume that the covariance matrix of the errors is correctly specified.
- [2] The condition number is large, 6.28e+03. This might indicate that there are strong multicollinearity or other numerical problems.

[32]: model3=smf.ols('Price~Doors',data=df1).fit()
model3.summary()

[32]:

Dep. Variable:		Price	R-squared:		ed:	0.034
Model:		OLS		Adj. R-s	0.033	
Method: Le		Least Squares		F-statisti	49.99	
Date: Wed		d, 20 Nov 2024		Prob (F-	2.40e-12	
Time:		11:24:10		Log-Like	-13765.	
No. Observa	tions:	1435	AIC:			2.753e + 04
Df Residuals	S:	1433	1	BIC:		2.755e + 04
Df Model:		1				
Covariance Type:		nonrobust				
	coef	std err	t	$\mathbf{P}$ > $ \mathbf{t} $	[0.025]	0.975]
Intercept	<b>coef</b> 7916.1452	std err 407.596	t 19.422	P>  t  0.000	[ <b>0.025</b> 7116.596	<b>0.975</b> ] 8715.694
Intercept Doors						
-	7916.1452 695.4978	407.596	19.422 7.071	0.000	7116.596 502.541	8715.694
Doors Omnib	7916.1452 695.4978	407.596 98.366	19.422 7.071 <b>Durbi</b>	0.000	7116.596 502.541 n: 0	8715.694 888.454
Doors Omnib	7916.1452 695.4978 us:	407.596 98.366 465.543	19.422 7.071 <b>Durbi</b>	0.000 0.000 n-Watso	7116.596 502.541 n: 0 JB): 140	8715.694 888.454 .289
Doors Omnib Prob(0	7916.1452 695.4978 us: Omnibus):	407.596 98.366 465.543 0.000	19.422 7.071 <b>Durbit</b> <b>Jarque</b>	0.000 0.000 n-Watso e-Bera (JB):	7116.596 502.541 n: 0 JB): 140	8715.694 888.454 .289 03.980

## Notes:

[1] Standard Errors assume that the covariance matrix of the errors is correctly specified.

[33]: model4=smf.ols('Price~Doors+cc',data=df1).fit() model4.summary()

[33]:

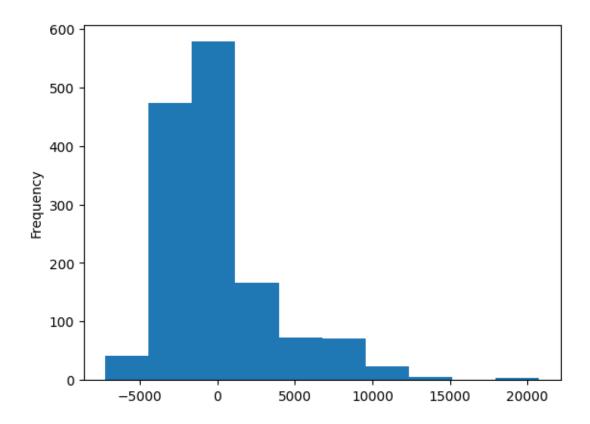
Dep. Variable:	Price	R-squared:	0.046
Model:	OLS	Adj. R-squared:	0.045
Method:	Least Squares	F-statistic:	34.40
Date:	Wed, 20 Nov 2024	Prob (F-statistic):	2.55e-15
Time:	11:24:21	Log-Likelihood:	-13756.
No. Observations:	1435	AIC:	2.752e + 04
Df Residuals:	1432	BIC:	2.753e + 04
Df Model:	2		
Correniance True	nonnobust		

Covariance Type: nonrobust

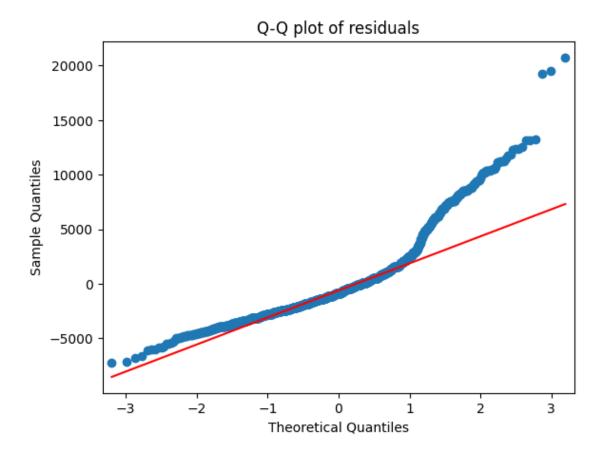
	coef	std err	t	$\mathbf{P}$ > $ \mathbf{t} $	[0.025]	0.975]
Intercept	6568.3395	513.700	12.786	0.000	5560.655	7576.024
Doors	662.3187	98.089	6.752	0.000	469.906	854.732
cc	0.9398	0.220	4.268	0.000	0.508	1.372
Omnibus:		448.494	Durbi	n-Watso	<b>n:</b> 0	.291
$\operatorname{Prob}(C)$	$\mathbf{Omnibus}):$	0.000	Jarque	e-Bera (	<b>JB):</b> 129	97.612
Skew:		1.602	<b>Prob(JB):</b> 1.69e-2		9e-282	
Kurtosis:		6.382	Cond.	No.	9.0	9e + 03

## Notes:

- [1] Standard Errors assume that the covariance matrix of the errors is correctly specified.
- [2] The condition number is large, 9.09e+03. This might indicate that there are strong multicollinearity or other numerical problems.
- [34]: model4.resid.mean()
- [34]: -3.019497058117431e-10
- [35]: model4.resid\_pearson
- [35]: array([ 0.86893322, 0.93980501, 0.99650243, ..., -0.36201304, -0.71637196, -1.25685978])
- [36]: model4.resid.plot(kind='hist')
- [36]: <Axes: ylabel='Frequency'>



```
[37]: import statsmodels.api as sm
    qqplot=sm.qqplot(model4.resid,line='q')
    plt.title("Q-Q plot of residuals")
    plt.show()
```



```
[38]: from sklearn.model_selection import train_test_split,GridSearchCV
     from sklearn.linear_model import Lasso,Ridge,ElasticNet
     from sklearn.metrics import r2_score
     lasso=Lasso(alpha=2)
[39]: lasso.fit(x_train,y_train)
     y_pred=lasso.predict(x_test)
     y_pred
[39]: array([15535.33357727, 10798.60278209, 13128.65182423, 10472.98259672,
             8061.21665822, 9451.62149365, 13895.88023501, 6388.12445434,
            13277.77884749, 8485.9943513, 13260.72638017, 7699.81763217,
            11528.34608186, 4984.8824338, 18084.63657131, 8308.82969744,
            13565.8663738 , 15560.49497551, 16637.24873992, 9382.75511014,
            12445.05192929, 10780.37675906, 10879.21582621, 11884.32081269,
             8321.49461064, 8283.93939524, 9824.64107373, 8456.4439936,
             9460.65953027, 8081.38716205, 9085.4409467, 12026.70510086,
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             7682.66711741, 8011.8436919 , 9803.82916223, 13882.9614978 ,
```

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10044.38684522, 18906.47778314, 9324.19465321, 8564.41080782,
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16643.75514318, 12980.61818731, 9036.08813068, 8798.08441101,
11734.28850085, 7864.72961904, 6963.87921441, 8533.02804597,
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8056.24623983, 8152.4196945, 8477.52824505, 8365.70565543,
7303.07649649, 8200.07213468, 16134.54856246, 8306.01346864,
8497.86050736, 10985.68619737, 8426.57108717, 15200.97754743,
7971.27908647, 12973.39153959, 6508.2368678, 8452.55057639,
7647.08857916, 18233.65414239, 18122.96307466, 16675.41489239,
7072.75800713, 10601.73435566, 19455.66360318, 8442.20374617,
9393.21918963, 16455.80752147, 12769.21946363, 9317.0920238,
16517.62340117, 14117.16648786, 9842.13653046, 9620.62117504,
7491.20803401, 10481.43106302, 11195.17074993, 8941.57886035,
9662.443577 , 7617.86229495, 10318.95396932, 16409.71279849,
10888.1557966 , 11417.51097904, 6441.71673965, 10603.45798808,
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14082.14172982, 8009.07732399, 11606.41185889, 6490.33588876,
6038.79749898, 9265.32154266, 19064.5527119 , 7639.61879951,
15980.3424046 , 7610.33131774 , 9373.34158068 , 9603.28446129 ,
13331.71784876, 12382.27950156, 11079.96570318, 7980.35981034,
11624.70398339, 8605.72324001, 10482.00349334, 11256.60065229,
13587.32944133, 17049.19774515, 9001.86875281, 8586.06782125,
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5418.35829305, 12701.64395161, 10038.22365028, 10492.84917152,
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10112.7218778 , 10066.09641866, 9617.12447517, 13452.41913818,
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11373.21258253, 13585.03130491, 13080.61875016, 11565.52333491,
9375.99447715, 14361.0643934 , 8730.13329557, 9640.36687091,
10873.07188772, 14312.51695312, 13060.53225545, 12248.01370974,
7373.88561326, 7435.97809123, 9602.4249373, 12927.38665251,
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```

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             8722.9911962 ,
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             7477.61364294, 9187.59278733, 17708.84868889, 11932.27741441,
             6271.42962048, 8037.89246158, 26147.71166947, 12697.73890157,
             16256.41194267, 11701.70913111, 11383.18244336])
[40]: r2_score(y_test,y_pred)
[40]: 0.8672128573315887
[41]: lasso.intercept
[41]: array([-8286.44378335])
[42]: lasso.coef_
[42]: array([-1.24427888e+02, -1.77275250e-02,
                                               3.71285103e+02,
                                                                 2.66341196e+01,
             3.51346244e+02, -2.99874423e-02,
                                                0.00000000e+00,
                                                                 0.0000000e+00,
             6.66925493e+02, 1.90995146e+01])
[43]: params={'alpha':[1,2,3,4,5,6]}
[44]: grid_search=GridSearchCV(lasso,params)
      grid_search.fit(x_train,y_train)
[44]: GridSearchCV(estimator=Lasso(alpha=2), param_grid={'alpha': [1, 2, 3, 4, 5, 6]})
[45]: grid_search.best_params_
[45]: {'alpha': 6}
[46]: lasso1=Lasso(alpha=6)
[47]: lasso1.fit(x_train,y_train)
[47]: Lasso(alpha=6)
[48]: y_pred1=lasso1.predict(x_test)
```

## [49]: y\_pred1

```
[49]: array([15526.24858617, 10813.39447402, 13137.41411112, 10483.8982614,
             8072.33117303, 9386.19857596, 13904.73978644, 6386.68226839,
            13282.48758289, 8489.83956192, 13272.73795677, 7617.33258868,
            11539.69551447, 4960.90233284, 18095.64508333, 8323.4773102,
            13572.2550817 , 15560.67421343, 16633.07425123, 9401.76798123,
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             8323.94032091, 8281.00193877, 9840.29837353, 8477.91752653,
             9471.87052615, 8095.00710298, 9086.10459869, 12039.90687935,
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            11628.38844095, 8615.14242747, 10495.18671915, 11270.39095602,
            13588.91840233, 17046.55410428, 9019.32357216, 8615.68820426,
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```

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             12355.48383276, 15274.08601093, 10596.66159685, 13018.76052184,
             14014.84549624, 9920.4246808, 12385.90363941, 6985.7106815,
             10197.20072397, 10072.4483247, 9626.23707057, 13466.26406225,
             8316.89832456, 8148.7987114 , 11515.89552451, 7994.18539806,
             11382.3257689 , 13593.59502075, 13090.02352163, 11569.28502984,
             9378.35508418, 14370.25272808, 8790.79211732, 9651.51409645,
             10885.37352503, 14238.19589507, 13065.98691281, 12258.3618226,
             7366.83400464, 7430.39663818, 9613.61699547, 12936.36529471,
             7376.03607699, 11273.51900219, 9291.69719712, 15584.91954464,
             8442.9509746 , 10035.90910169 , 10782.93425633 , 8721.25538669 ,
             6042.57154054, 10703.46711444, 14901.53029841, 11031.41800624,
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