

In [27]:

```
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from statsmodels.graphics.regressionplots import influence_plot
import statsmodels.formula.api as smf
import numpy as np
import statsmodels.api as sm
from statsmodels.stats.outliers_influence import variance_inflation_factor
```

executed in 15ms, finished 09:45:26 2021-11-26

In [2]:

```
data = pd.read_csv("50_Startups.csv")
data.head()
```

executed in 79ms, finished 09:24:30 2021-11-26

Out[2]:

	R&D Spend	Administration	Marketing Spend	State	Profit
0	165349.20	136897.80	471784.10	New York	192261.83
1	162597.70	151377.59	443898.53	California	191792.06
2	153441.51	101145.55	407934.54	Florida	191050.39
3	144372.41	118671.85	383199.62	New York	182901.99
4	142107.34	91391.77	366168.42	Florida	166187.94

In [3]:

```
data.info()
```

executed in 43ms, finished 09:24:50 2021-11-26

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 50 entries, 0 to 49
Data columns (total 5 columns):
#   Column                Non-Null Count  Dtype
---  -
0   R&D Spend              50 non-null    float64
1   Administration         50 non-null    float64
2   Marketing Spend        50 non-null    float64
3   State                  50 non-null    object
4   Profit                 50 non-null    float64
dtypes: float64(4), object(1)
memory usage: 2.1+ KB
```

In [5]:

```
data.isna().sum()
```

executed in 30ms, finished 09:25:12 2021-11-26

Out[5]:

```
R&D Spend      0
Administration 0
Marketing Spend 0
State          0
Profit         0
dtype: int64
```

In [6]:

```
data.corr()
```

executed in 34ms, finished 09:25:39 2021-11-26

Out[6]:

	R&D Spend	Administration	Marketing Spend	Profit
R&D Spend	1.000000	0.241955	0.724248	0.972900
Administration	0.241955	1.000000	-0.032154	0.200717
Marketing Spend	0.724248	-0.032154	1.000000	0.747766
Profit	0.972900	0.200717	0.747766	1.000000

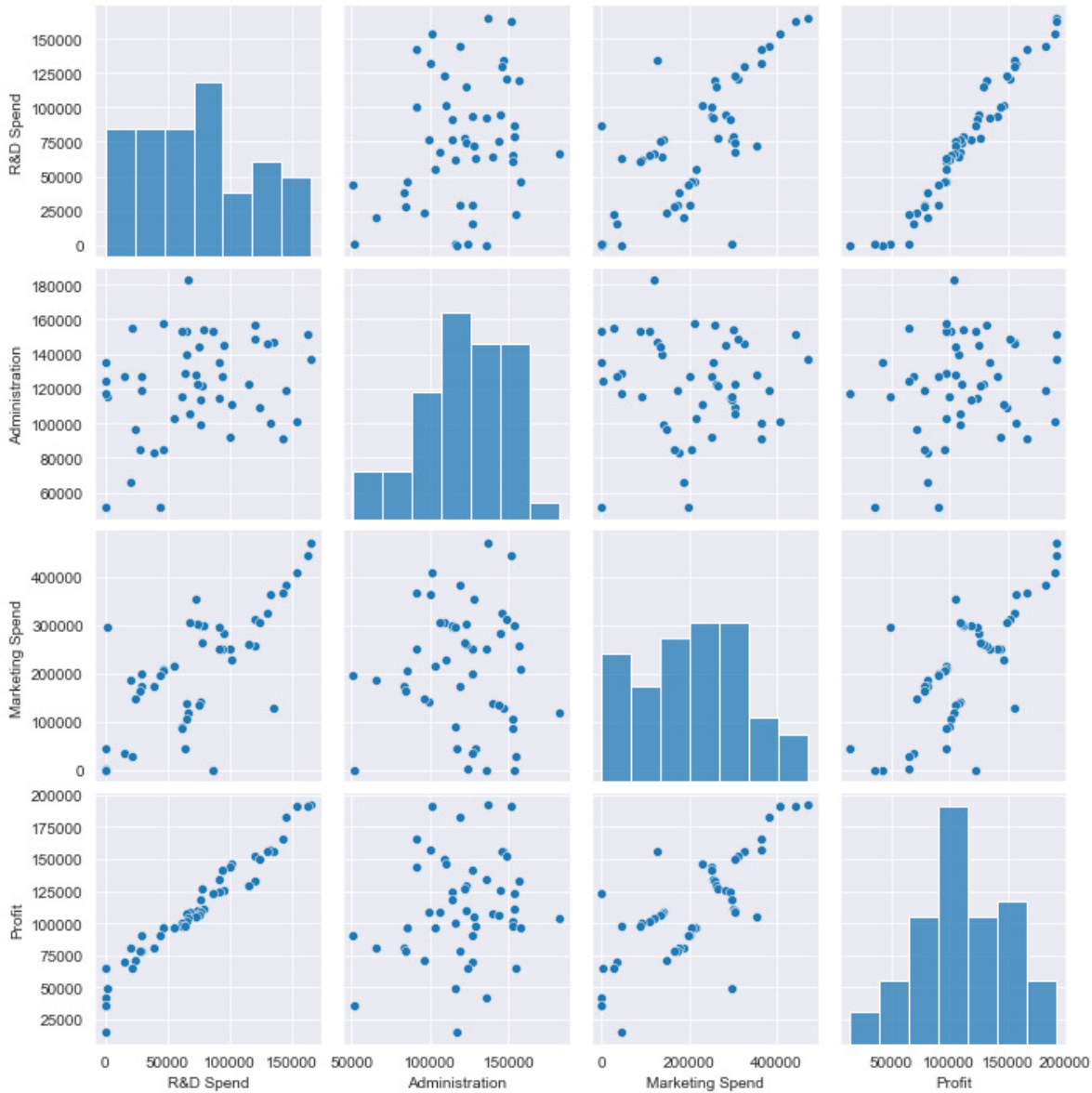
In [7]:

```
sns.set_style(style='darkgrid')  
sns.pairplot(data)
```

executed in 5.33s, finished 09:26:19 2021-11-26

Out[7]:

<seaborn.axisgrid.PairGrid at 0x29acda0e610>



In [12]:

```
corrMatrix = data.corr()
```

executed in 22ms, finished 09:39:52 2021-11-26

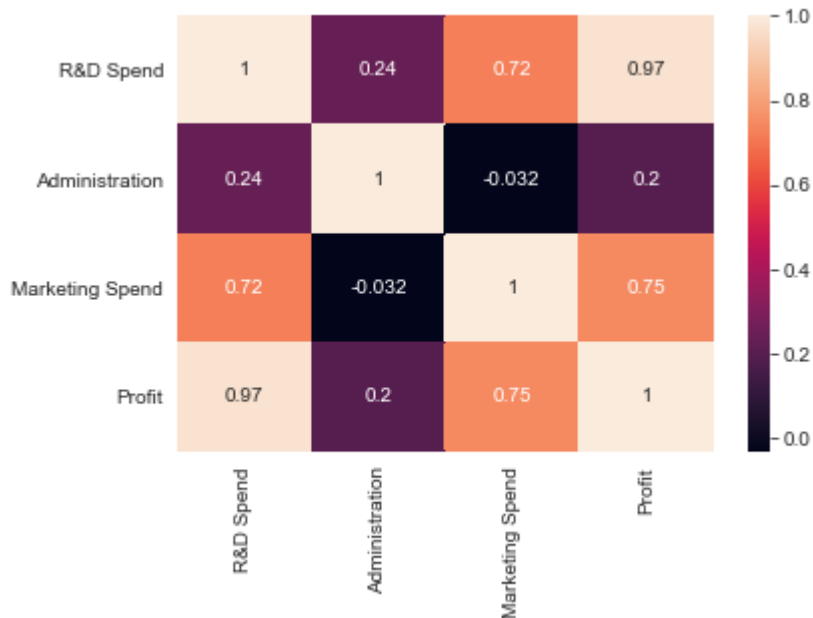
In [13]:

```
sns.heatmap(corrMatrix, annot=True)
```

executed in 577ms, finished 09:40:08 2021-11-26

Out[13]:

<AxesSubplot:>



In [14]:

```
data = pd.get_dummies(data, columns=['State'])
```

executed in 17ms, finished 09:40:37 2021-11-26

In [15]:

```
X = data[['R&D Spend', 'Administration', 'Marketing Spend', 'State_California', 'State_Flori  
Y = data[['Profit']]
```

executed in 23ms, finished 09:41:18 2021-11-26

In [19]:

```
model = sm.OLS(Y, X).fit()  
predictions = model.predict(X)
```

executed in 22ms, finished 09:42:37 2021-11-26

In [20]:

```
model.summary()
```

executed in 48ms, finished 09:42:54 2021-11-26

Out[20]:

OLS Regression Results

Dep. Variable:	Profit	R-squared:	0.951
Model:	OLS	Adj. R-squared:	0.945
Method:	Least Squares	F-statistic:	169.9
Date:	Fri, 26 Nov 2021	Prob (F-statistic):	1.34e-27
Time:	09:42:54	Log-Likelihood:	-525.38
No. Observations:	50	AIC:	1063.
Df Residuals:	44	BIC:	1074.
Df Model:	5		
Covariance Type:	nonrobust		

	coef	std err	t	P> t	[0.025	0.975]
R&D Spend	0.8060	0.046	17.369	0.000	0.712	0.900
Administration	-0.0270	0.052	-0.517	0.608	-0.132	0.078
Marketing Spend	0.0270	0.017	1.574	0.123	-0.008	0.062
State_California	5.013e+04	6884.820	7.281	0.000	3.62e+04	6.4e+04
State_Florida	5.032e+04	7251.767	6.940	0.000	3.57e+04	6.49e+04
State_New York	5.008e+04	6952.587	7.204	0.000	3.61e+04	6.41e+04

Omnibus:	14.782	Durbin-Watson:	1.283
Prob(Omnibus):	0.001	Jarque-Bera (JB):	21.266
Skew:	-0.948	Prob(JB):	2.41e-05
Kurtosis:	5.572	Cond. No.	2.45e+06

Notes:

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

[2] The condition number is large, 2.45e+06. This might indicate that there are strong multicollinearity or other numerical problems.

In [21]:

```
infl = model.get_influence()
```

executed in 22ms, finished 09:43:20 2021-11-26

In [22]:

```
summ_data = infl.summary_frame()
```

executed in 96ms, finished 09:43:40 2021-11-26

In [23]:

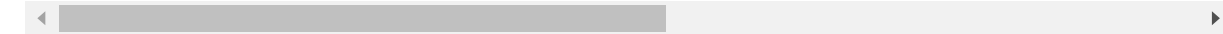
```
summ_data.sort_values('cooks_d', ascending=False)
```

executed in 100ms, finished 09:44:03 2021-11-26

Out[23]:

	dfb_R&D Spend	dfb_Administration	dfb_Marketing Spend	dfb_State_California	dfb_State_Florida	dfb_St
49	0.578956	-0.114232	0.080954	-0.566028	-0.246221	-
48	-0.112734	0.701599	0.418630	-0.783828	-0.801849	-1
45	-0.212843	0.091394	-0.189969	0.095382	0.140857	1
14	-0.221204	-0.257240	0.142195	0.267421	0.086725	1
36	-0.379353	0.189523	0.218405	-0.107545	0.053174	-1
38	-0.189819	-0.313449	0.109261	0.320201	0.309091	1
15	-0.208289	0.066627	0.071114	-0.002577	0.007587	-1
46	0.434369	-0.142646	-0.364064	0.106828	0.034265	1
19	0.252210	0.039342	-0.342025	0.009492	0.035168	1
27	0.271462	-0.146112	-0.339679	0.169919	0.186504	1
2	0.197811	-0.174765	-0.013702	0.080109	0.147293	1
3	0.110000	-0.049701	0.073783	-0.035947	-0.056370	1
43	-0.090858	0.058906	-0.085052	0.024294	0.044997	1
10	0.186811	-0.137400	-0.159583	0.116978	0.216711	1
12	0.069420	0.010085	-0.048411	-0.017490	0.098185	-1
34	-0.196836	0.242310	0.173390	-0.138400	-0.212207	-1
11	0.152695	-0.197500	-0.063320	0.226735	0.131358	1
16	-0.055403	0.029857	0.116725	0.037230	-0.065807	-1
4	-0.153851	0.163418	0.047737	-0.104342	-0.152925	-1
5	-0.081405	0.105665	-0.033055	-0.046292	-0.031296	-1
21	0.134008	-0.163215	-0.156986	0.162458	0.165701	1
35	-0.049889	-0.102221	0.034443	0.099727	0.095288	1
13	0.005568	0.054245	0.040526	-0.007538	-0.074145	-1
9	-0.087676	0.064819	0.009105	-0.073333	-0.016881	-1
26	-0.047078	-0.040934	0.086363	0.018763	-0.038037	1
24	-0.077455	0.089257	0.087987	-0.087996	-0.089711	-1
47	0.046774	-0.046937	0.031389	-0.029301	-0.001347	1
17	0.035255	-0.075381	-0.061166	0.081487	0.083661	1
7	-0.045130	-0.041183	0.001538	0.056695	0.017003	1
25	0.012139	0.046367	-0.028528	0.012690	-0.030258	-1
1	0.020841	0.046268	0.058280	-0.049608	-0.085419	-1
22	0.057093	-0.025665	-0.065599	0.028443	-0.010225	1
18	-0.003027	0.019777	-0.017913	-0.008655	-0.050050	-1
6	-0.095187	0.004676	0.082074	-0.021299	-0.002912	1

	dfb_R&D Spend	dfb_Administration	dfb_Marketing Spend	dfb_State_California	dfb_State_Florida	dfb_St
40	-0.063137	0.020142	0.044268	0.021070	-0.010930	-l
41	-0.022659	-0.051009	-0.006525	0.059567	0.087625	l
39	0.011576	0.050925	-0.005414	-0.074335	-0.047479	-l
28	-0.004892	0.054512	-0.014043	-0.041421	-0.021855	-l
20	-0.018937	0.001269	0.034093	0.009204	-0.011262	-l
32	-0.029851	0.005201	0.042128	-0.025767	-0.013770	-l
23	0.017345	0.004924	-0.019954	-0.003001	-0.014564	-l
33	0.005720	0.010764	-0.000214	-0.012256	-0.024085	-l
42	-0.010791	-0.008202	0.004902	0.018523	0.009901	l
44	-0.005257	0.013394	-0.004732	-0.001883	-0.006265	-l
29	-0.001210	-0.011523	0.007045	0.007881	0.006612	l
8	0.000992	0.008446	0.005091	-0.010483	-0.010897	-l
37	0.001322	-0.016087	-0.000850	0.018397	0.013923	l
30	0.004911	-0.003183	-0.008237	0.004519	0.008048	l
0	-0.000680	-0.001084	-0.002943	0.002627	0.003023	l
31	-0.000298	-0.002100	0.001610	0.001322	0.001054	l

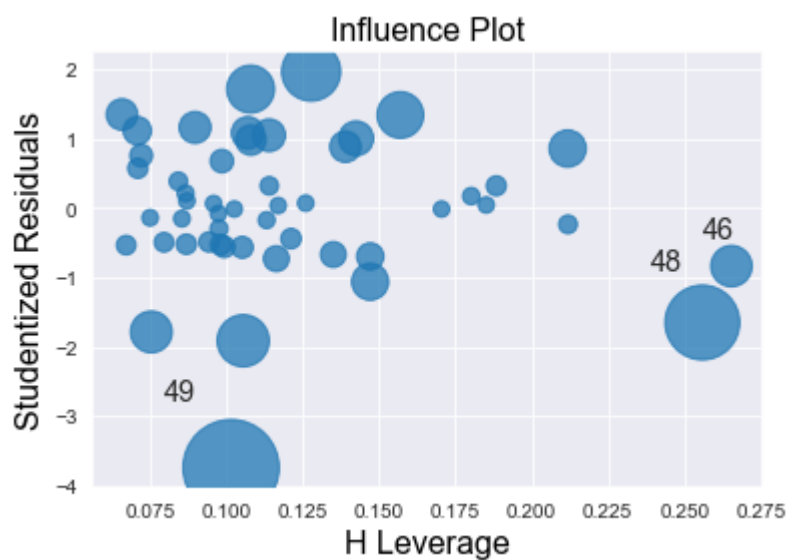
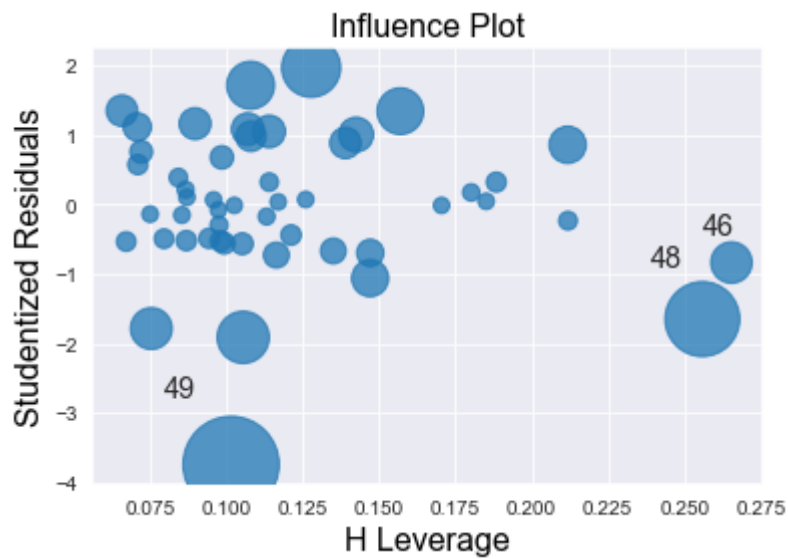


In [24]:

```
infl.plot_influence()
```

executed in 933ms, finished 09:44:22 2021-11-26

Out[24]:



In [25]:

```
vif = pd.DataFrame()
```

executed in 7ms, finished 09:44:41 2021-11-26

In [28]:

```
vif["VIF Factor"] = [variance_inflation_factor(X.values, i) for i in range(X.shape[1])]
```

executed in 22ms, finished 09:45:32 2021-11-26

In [29]:

```
vif["features"] = X.columns
```

executed in 15ms, finished 09:45:49 2021-11-26

In [30]:

```
vif.round(1)
```

executed in 35ms, finished 09:46:03 2021-11-26

Out[30]:

	VIF Factor	features
0	2.5	R&D Spend
1	1.2	Administration
2	2.4	Marketing Spend
3	9.0	State_California
4	9.4	State_Florida
5	9.2	State_New York

In [32]:

```
new_X = data[['R&D Spend', 'Marketing Spend', 'State_California', 'State_Florida', 'State_N
```

executed in 19ms, finished 09:46:43 2021-11-26

In [33]:

```
new_model = sm.OLS(Y, new_X).fit()  
new_predictions = new_model.predict(new_X)
```

executed in 20ms, finished 09:46:57 2021-11-26

In [34]:

```
new_model.summary()
```

executed in 63ms, finished 09:47:07 2021-11-26

Out[34]:

OLS Regression Results

Dep. Variable:	Profit	R-squared:	0.950
Model:	OLS	Adj. R-squared:	0.946
Method:	Least Squares	F-statistic:	215.8
Date:	Fri, 26 Nov 2021	Prob (F-statistic):	9.72e-29
Time:	09:47:07	Log-Likelihood:	-525.53
No. Observations:	50	AIC:	1061.
Df Residuals:	45	BIC:	1071.
Df Model:	4		
Covariance Type:	nonrobust		

	coef	std err	t	P> t	[0.025	0.975]
R&D Spend	0.7967	0.042	18.771	0.000	0.711	0.882
Marketing Spend	0.0298	0.016	1.842	0.072	-0.003	0.062
State_California	4.696e+04	3119.471	15.053	0.000	4.07e+04	5.32e+04
State_Florida	4.71e+04	3670.129	12.833	0.000	3.97e+04	5.45e+04
State_New York	4.694e+04	3342.591	14.043	0.000	4.02e+04	5.37e+04

Omnibus:	14.640	Durbin-Watson:	1.257
Prob(Omnibus):	0.001	Jarque-Bera (JB):	21.037
Skew:	-0.938	Prob(JB):	2.70e-05
Kurtosis:	5.565	Cond. No.	9.45e+05

Notes:

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

[2] The condition number is large, 9.45e+05. This might indicate that there are strong multicollinearity or other numerical problems.

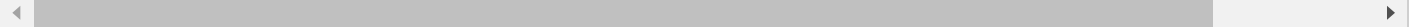
In [35]:

```
new_df = data.drop(data.index[[49,48]])
```

executed in 11ms, finished 09:47:55 2021-11-26

In [37]:

```
new_X = new_df[['R&D Spend', 'Marketing Spend', 'State_California', 'State_Florida', 'State  
new_Y = new_df[['Profit']]
```



executed in 10ms, finished 09:48:34 2021-11-26

In [38]:

```
final_model = sm.OLS(new_Y, new_X).fit()  
predictions = final_model.predict(new_X)
```

executed in 23ms, finished 09:48:52 2021-11-26

In [39]:

```
final_model.summary()
```

executed in 58ms, finished 09:49:03 2021-11-26

Out[39]:

OLS Regression Results

Dep. Variable:	Profit		R-squared:	0.961		
Model:	OLS		Adj. R-squared:	0.958		
Method:	Least Squares		F-statistic:	265.9		
Date:	Fri, 26 Nov 2021		Prob (F-statistic):	1.02e-29		
Time:	09:49:03		Log-Likelihood:	-494.30		
No. Observations:	48		AIC:	998.6		
Df Residuals:	43		BIC:	1008.		
Df Model:	4					
Covariance Type:	nonrobust					
	coef	std err	t	P> t 	[0.025	0.975]
R&D Spend	0.7692	0.035	22.072	0.000	0.699	0.840
Marketing Spend	0.0251	0.013	1.908	0.063	-0.001	0.052
State_California	5.183e+04	2710.866	19.120	0.000	4.64e+04	5.73e+04
State_Florida	5.046e+04	3078.590	16.391	0.000	4.43e+04	5.67e+04
State_New York	5.09e+04	2936.767	17.333	0.000	4.5e+04	5.68e+04
Omnibus:	0.133	Durbin-Watson:	1.645			
Prob(Omnibus):	0.936	Jarque-Bera (JB):	0.304			
Skew:	0.097	Prob(JB):	0.859			
Kurtosis:	2.661	Cond. No.	1.02e+06			

Notes:

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

[2] The condition number is large, 1.02e+06. This might indicate that there are strong multicollinearity or other numerical problems.

In [40]:

```
X_sqrt = np.sqrt(new_df[['R&D Spend', 'Marketing Spend', 'State_California', 'State_Florida
```

executed in 19ms, finished 09:49:31 2021-11-26

In [42]:

```
model3 = sm.OLS(new_Y, X_sqrt).fit()
predictions3 = model3.predict(X_sqrt)
```

executed in 16ms, finished 09:49:57 2021-11-26

In [43]:

```
model3.summary()
```

```
executed in 43ms, finished 09:49:59 2021-11-26
```

Out[43]:

OLS Regression Results

Dep. Variable:	Profit	R-squared:	0.887
Model:	OLS	Adj. R-squared:	0.877
Method:	Least Squares	F-statistic:	84.44
Date:	Fri, 26 Nov 2021	Prob (F-statistic):	8.67e-20
Time:	09:49:59	Log-Likelihood:	-519.91
No. Observations:	48	AIC:	1050.
Df Residuals:	43	BIC:	1059.
Df Model:	4		
Covariance Type:	nonrobust		

	coef	std err	t	P> t	[0.025	0.975]
R&D Spend	340.5455	25.777	13.211	0.000	288.560	392.531
Marketing Spend	20.0497	15.481	1.295	0.202	-11.170	51.270
State_California	1.836e+04	6267.224	2.930	0.005	5724.219	3.1e+04
State_Florida	1.692e+04	7013.669	2.413	0.020	2779.320	3.11e+04
State_New York	1.908e+04	6591.247	2.894	0.006	5782.772	3.24e+04

Omnibus:	7.588	Durbin-Watson:	0.777
Prob(Omnibus):	0.023	Jarque-Bera (JB):	7.161
Skew:	0.941	Prob(JB):	0.0279
Kurtosis:	3.197	Cond. No.	3.04e+03

Notes:

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

[2] The condition number is large, 3.04e+03. This might indicate that there are strong multicollinearity or other numerical problems.

In [44]:

```
Y_sqrt = np.sqrt(new_df['Profit'])
```

```
executed in 18ms, finished 09:50:17 2021-11-26
```

In [45]:

```
model4 = sm.OLS(Y_sqrt, new_X).fit()
predictions4 = model4.predict(new_X)
```

```
executed in 13ms, finished 09:50:30 2021-11-26
```

In [46]:

```
model14.summary()
```

executed in 45ms, finished 09:50:42 2021-11-26

Out[46]:

OLS Regression Results

Dep. Variable:	Profit	R-squared:	0.954
Model:	OLS	Adj. R-squared:	0.950
Method:	Least Squares	F-statistic:	223.3
Date:	Fri, 26 Nov 2021	Prob (F-statistic):	3.68e-28
Time:	09:50:42	Log-Likelihood:	-185.87
No. Observations:	48	AIC:	381.7
Df Residuals:	43	BIC:	391.1
Df Model:	4		
Covariance Type:	nonrobust		

	coef	std err	t	P> t	[0.025	0.975]
R&D Spend	0.0012	5.64e-05	20.622	0.000	0.001	0.001
Marketing Spend	2.473e-05	2.13e-05	1.159	0.253	-1.83e-05	6.78e-05
State_California	241.0032	4.390	54.894	0.000	232.149	249.857
State_Florida	240.7325	4.986	48.283	0.000	230.678	250.787
State_New York	240.9886	4.756	50.669	0.000	231.397	250.580

Omnibus:	4.530	Durbin-Watson:	1.406
Prob(Omnibus):	0.104	Jarque-Bera (JB):	3.371
Skew:	-0.532	Prob(JB):	0.185
Kurtosis:	3.745	Cond. No.	1.02e+06

Notes:

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

[2] The condition number is large, 1.02e+06. This might indicate that there are strong multicollinearity or other numerical problems.

In [47]:

```
model15 = sm.OLS(Y_sqrt, X_sqrt).fit()
predictions5 = model15.predict(X_sqrt)
```

executed in 14ms, finished 09:51:00 2021-11-26

In [48]:

model15.summary()

executed in 54ms, finished 09:51:11 2021-11-26

Out[48]:

OLS Regression Results

Dep. Variable:	Profit	R-squared:	0.929
Model:	OLS	Adj. R-squared:	0.923
Method:	Least Squares	F-statistic:	141.7
Date:	Fri, 26 Nov 2021	Prob (F-statistic):	3.64e-24
Time:	09:51:11	Log-Likelihood:	-196.16
No. Observations:	48	AIC:	402.3
Df Residuals:	43	BIC:	411.7
Df Model:	4		
Covariance Type:	nonrobust		

	coef	std err	t	P> t	[0.025	0.975]
R&D Spend	0.5271	0.030	17.371	0.000	0.466	0.588
Marketing Spend	0.0231	0.018	1.270	0.211	-0.014	0.060
State_California	187.8689	7.377	25.465	0.000	172.991	202.747
State_Florida	187.0162	8.256	22.652	0.000	170.366	203.666
State_New York	189.8076	7.759	24.463	0.000	174.160	205.455

Omnibus:	7.976	Durbin-Watson:	1.243
Prob(Omnibus):	0.019	Jarque-Bera (JB):	7.007
Skew:	0.870	Prob(JB):	0.0301
Kurtosis:	3.692	Cond. No.	3.04e+03

Notes:

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

[2] The condition number is large, 3.04e+03. This might indicate that there are strong multicollinearity or other numerical problems.