```
import pandas as pd

dataset = pd.read_csv("50_Startups.csv")
#dataset = pd.read_csv("50_Startups_withYear.csv")

dataset
```

	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
5	131876.90	99814.71	362861.36	New York	156991.12
6	134615.46	147198.87	127716.82	California	156122.51
7	130298.13	145530.06	323876.68	Florida	155752.60
8	120542.52	148718.95	311613.29	New York	152211.77
9	123334.88	108679.17	304981.62	California	149759.96
10	101913.08	110594.11	229160.95	Florida	146121.95
11	100671.96	91790.61	249744.55	California	144259.40
12	93863.75	127320.38	249839.44	Florida	141585.52
13	91992.39	135495.07	252664.93	California	134307.35
14	119943.24	156547.42	256512.92	Florida	132602.65
15	114523.61	122616.84	261776.23	New York	129917.04
16	78013.11	121597.55	264346.06	California	126992.93
17	94657.16	145077.58	282574.31	New York	125370.37
18	91749.16	114175.79	294919.57	Florida	124266.90
19	86419.70	153514.11	0.00	New York	122776.86
20	76253.86	113867.30	298664.47	California	118474.03
21	78389.47	153773.43	299737.29	New York	111313.02
22	73994.56	122782.75	303319.26	Florida	110352.25
23	67532.53	105751.03	304768.73	Florida	108733.99
24	77044.01	99281.34	140574.81	New York	108552.04
25	64664.71	139553.16	137962.62	California	107404.34
26	75328.87	144135.98	134050.07	Florida	105733.54
27	72107.60	127864.55	353183.81	New York	105008.31
28	66051.52	182645.56	118148.20	Florida	103282.38
29	65605.48	153032.06	107138.38	New York	101004.64
30	61994.48	115641.28	91131.24	Florida	99937.59
31	61136.38	152701.92	88218.23	New York	97483.56
32	63408.86	129219.61	46085.25	California	97427.84
33	55493.95	103057.49	214634.81	Florida	96778.92
34	46426.07	157693.92	210797.67	California	96712.80
35	46014.02	85047.44	205517.64	New York	96479.51
36	28663.76	127056.21	201126.82	Florida	90708.19
37	44069.95	51283.14	197029.42	California	89949.14
38	20229.59	65947.93	185265.10	New York	81229.06
39	38558.51	82982.09	174999.30	California	81005.76
40	28754.33	118546.05	172795.67	California	78239.91
41	27892.92	84710.77	164470.71	Florida	77798.83

```
42   23640.93        96189.63        148001.11  California   71498.49
43   15505.73       127382.30         35534.17    New York   69758.98
44   22177.74       154806.14         28334.72  California   65200.33
45    1000.23       124153.04          1903.93    New York   64926.08
46    1315.46       115816.21        297114.46     Florida   49490.75
47       0.00       135426.92             0.00  California   42559.73
48     542.05        51743.15             0.00    New York   35673.41
49       0.00       116983.80         45173.06  California   14681.40
```

```
datase=pd.get_dummies(dataset,drop_first=True)
```

```
datase
```

```
    R&D Spend  Administration  Marketing Spend     Profit
State_Florida  \
0   165349.20       136897.80        471784.10  192261.83
False
1   162597.70       151377.59        443898.53  191792.06
False
2   153441.51       101145.55        407934.54  191050.39
True
3   144372.41       118671.85        383199.62  182901.99
False
4   142107.34        91391.77        366168.42  166187.94
True
5   131876.90        99814.71        362861.36  156991.12
False
6   134615.46       147198.87        127716.82  156122.51
False
7   130298.13       145530.06        323876.68  155752.60
True
8   120542.52       148718.95        311613.29  152211.77
False
9   123334.88       108679.17        304981.62  149759.96
False
10  101913.08       110594.11        229160.95  146121.95
True
11  100671.96        91790.61        249744.55  144259.40
False
12   93863.75       127320.38        249839.44  141585.52
True
13   91992.39       135495.07        252664.93  134307.35
False
14  119943.24       156547.42        256512.92  132602.65
True
15  114523.61       122616.84        261776.23  129917.04
False
16   78013.11       121597.55        264346.06  126992.93
False
17   94657.16       145077.58        282574.31  125370.37
```

```
False
18   91749.16       114175.79        294919.57  124266.90
True
19   86419.70       153514.11             0.00  122776.86
False
20   76253.86       113867.30        298664.47  118474.03
False
21   78389.47       153773.43        299737.29  111313.02
False
22   73994.56       122782.75        303319.26  110352.25
True
23   67532.53       105751.03        304768.73  108733.99
True
24   77044.01        99281.34        140574.81  108552.04
False
25   64664.71       139553.16        137962.62  107404.34
False
26   75328.87       144135.98        134050.07  105733.54
True
27   72107.60       127864.55        353183.81  105008.31
False
28   66051.52       182645.56        118148.20  103282.38
True
29   65605.48       153032.06        107138.38  101004.64
False
30   61994.48       115641.28         91131.24   99937.59
True
31   61136.38       152701.92         88218.23   97483.56
False
32   63408.86       129219.61         46085.25   97427.84
False
33   55493.95       103057.49        214634.81   96778.92
True
34   46426.07       157693.92        210797.67   96712.80
False
35   46014.02        85047.44        205517.64   96479.51
False
36   28663.76       127056.21        201126.82   90708.19
True
37   44069.95        51283.14        197029.42   89949.14
False
38   20229.59        65947.93        185265.10   81229.06
False
39   38558.51        82982.09        174999.30   81005.76
False
40   28754.33       118546.05        172795.67   78239.91
False
41   27892.92        84710.77        164470.71   77798.83
True
```

```
42   23640.93        96189.63        148001.11   71498.49
False
43   15505.73       127382.30         35534.17   69758.98
False
44   22177.74       154806.14         28334.72   65200.33
False
45    1000.23       124153.04          1903.93   64926.08
False
46    1315.46       115816.21        297114.46   49490.75
True
47       0.00       135426.92             0.00   42559.73
False
48     542.05        51743.15             0.00   35673.41
False
49       0.00       116983.80         45173.06   14681.40
False
```

```
    State_New York
0             True
1            False
2            False
3             True
4            False
5             True
6            False
7            False
8             True
9            False
10           False
11           False
12           False
13           False
14           False
15            True
16           False
17            True
18           False
19            True
20           False
21            True
22           False
23           False
24            True
25           False
26           False
27            True
28           False
29            True
30           False
```

```
31            True
32           False
33           False
34           False
35            True
36           False
37           False
38            True
39           False
40           False
41           False
42           False
43            True
44           False
45            True
46           False
47           False
48            True
49           False
```

```
dataset=datase.replace({True: 1,False: 0})
```

```
dataset
```

```
    R&D Spend  Administration  Marketing Spend     Profit
State_Florida  \
0   165349.20       136897.80        471784.10  192261.83
0
1   162597.70       151377.59        443898.53  191792.06
0
2   153441.51       101145.55        407934.54  191050.39
1
3   144372.41       118671.85        383199.62  182901.99
0
4   142107.34        91391.77        366168.42  166187.94
1
5   131876.90        99814.71        362861.36  156991.12
0
6   134615.46       147198.87        127716.82  156122.51
0
7   130298.13       145530.06        323876.68  155752.60
1
8   120542.52       148718.95        311613.29  152211.77
0
9   123334.88       108679.17        304981.62  149759.96
0
10  101913.08       110594.11        229160.95  146121.95
1
11  100671.96        91790.61        249744.55  144259.40
0
```

```
12   93863.75       127320.38       249839.44  141585.52
1
13   91992.39       135495.07       252664.93  134307.35
0
14  119943.24       156547.42       256512.92  132602.65
1
15  114523.61       122616.84       261776.23  129917.04
0
16   78013.11       121597.55       264346.06  126992.93
0
17   94657.16       145077.58       282574.31  125370.37
0
18   91749.16       114175.79       294919.57  124266.90
1
19   86419.70       153514.11            0.00  122776.86
0
20   76253.86       113867.30       298664.47  118474.03
0
21   78389.47       153773.43       299737.29  111313.02
0
22   73994.56       122782.75       303319.26  110352.25
1
23   67532.53       105751.03       304768.73  108733.99
1
24   77044.01        99281.34       140574.81  108552.04
0
25   64664.71       139553.16       137962.62  107404.34
0
26   75328.87       144135.98       134050.07  105733.54
1
27   72107.60       127864.55       353183.81  105008.31
0
28   66051.52       182645.56       118148.20  103282.38
1
29   65605.48       153032.06       107138.38  101004.64
0
30   61994.48       115641.28        91131.24   99937.59
1
31   61136.38       152701.92        88218.23   97483.56
0
32   63408.86       129219.61        46085.25   97427.84
0
33   55493.95       103057.49       214634.81   96778.92
1
34   46426.07       157693.92       210797.67   96712.80
0
35   46014.02        85047.44       205517.64   96479.51
0
36   28663.76       127056.21       201126.82   90708.19
```

```
1
37   44069.95         51283.14        197029.42   89949.14
0
38   20229.59         65947.93        185265.10   81229.06
0
39   38558.51         82982.09        174999.30   81005.76
0
40   28754.33        118546.05        172795.67   78239.91
0
41   27892.92         84710.77        164470.71   77798.83
1
42   23640.93         96189.63        148001.11   71498.49
0
43   15505.73        127382.30         35534.17   69758.98
0
44   22177.74        154806.14         28334.72   65200.33
0
45    1000.23        124153.04          1903.93   64926.08
0
46    1315.46        115816.21        297114.46   49490.75
1
47       0.00        135426.92             0.00   42559.73
0
48     542.05         51743.15             0.00   35673.41
0
49       0.00        116983.80         45173.06   14681.40
0

    State_New York
0                1
1                0
2                0
3                1
4                0
5                1
6                0
7                0
8                1
9                0
10               0
11               0
12               0
13               0
14               0
15               1
16               0
17               1
18               0
19               1
```

```
20                0
21                1
22                0
23                0
24                1
25                0
26                0
27                1
28                0
29                1
30                0
31                1
32                0
33                0
34                0
35                1
36                0
37                0
38                1
39                0
40                0
41                0
42                0
43                1
44                0
45                1
46                0
47                0
48                1
49                0
```

```
dataset.columns
```

```
Index(['R&D Spend', 'Administration', 'Marketing Spend', 'Profit',
       'State_Florida', 'State_New York'],
      dtype='object')
```

```
independent=dataset[['R&D Spend', 'Administration', 'Marketing
Spend','State_Florida', 'State_New York']]
```

```
dependent=dataset[['Profit']]
```

```
#Training code
#Now we need to split the Training and Test data set from parant
dataset. To split, we are using sklearn as class and call function as
model_selection-> train_test_split
from sklearn.model_selection import train_test_split
#input parameters are passed in function train_test_split with
x,y,size for test data(30 percent from x & y) and random state then
assign to variable x_train,x_test,y_train and y_test
```

```
x_train,x_test,y_train,y_test=train_test_split(independent, dependent,
test_size=0.30, random_state=0)

from sklearn.preprocessing import StandardScaler
sc=StandardScaler()
x_train=sc.fit_transform(x_train)
x_test=sc.transform(x_test)

x_train
```

```
array([[ 1.17644103,  0.84515251,  0.94354978,  2.        , -
0.76870611],
       [ 0.96420324,  1.27283565,  0.42738817,  2.        , -
0.76870611],
       [-1.47369826,  0.0153175 , -1.52350329, -0.5       ,
1.30088727],
       [-1.48308929, -2.79556363, -1.53809178, -0.5       ,
1.30088727],
       [-0.14952431,  1.13637282, -0.71716495, -0.5       ,
1.30088727],
       [ 0.85312042, -0.04431628,  0.46771725, -0.5       ,
1.30088727],
       [-0.22353674, -0.3151007 , -0.83981652,  2.        , -
0.76870611],
       [-0.19454707,  0.21199679, -1.18497259, -0.5       , -
0.76870611],
       [ 0.10478723, -0.08388412,  0.48740807, -0.5       , -
0.76870611],
       [-1.0096458 , -1.07019473, -0.4040623 , -0.5       , -
0.76870611],
       [ 0.06872897, -0.38396487,  0.75036616, -0.5       , -
0.76870611],
       [-1.17638797,  0.14067421, -1.26581817, -0.5       ,
1.30088727],
       [ 0.97648631,  0.9689421 ,  0.84958395, -0.5       ,
1.30088727],
       [ 0.39131191,  0.45560401,  0.3979037 , -0.5       , -
0.76870611],
       [-0.16880669,  0.6131351 , -0.48098026, -0.5       , -
0.76870611],
       [ 1.2088001 , -0.92947267,  1.24226224, -0.5       ,
1.30088727],
       [ 0.44593006,  0.82758768,  0.62707846, -0.5       ,
1.30088727],
       [-0.90483959, -0.20234037, -0.21407884, -0.5       , -
0.76870611],
       [-1.49419935, -0.26298539, -1.19196207, -0.5       , -
0.76870611],
       [ 1.83846539,  1.07214791,  1.86319367, -0.5       , -
0.76870611],
```

```
       [ 0.42966802,  0.13827054,  0.37625394,  2.        , -
0.76870611],
       [-0.59092478, -2.81342077, -0.02839248, -0.5       , -
0.76870611],
       [ 0.08492419, -0.95017758, -0.46096486, -0.5       ,
1.30088727],
       [ 1.26493068,  0.90993408, -0.55948669, -0.5       , -
0.76870611],
       [-0.11002675, -0.69903054,  0.79713886,  2.        , -
0.76870611],
       [-0.90669595,  0.12801572,  0.00300304,  2.        , -
0.76870611],
       [ 0.11250125,  1.16515207,  0.75858643, -0.5       ,
1.30088727],
       [ 0.27709192,  1.15508553, -1.53809178, -0.5       ,
1.30088727],
       [ 1.03371959, -0.585363  ,  0.79877008, -0.5       , -
0.76870611],
       [-0.70388936, -1.58289852, -0.19719396, -0.5       , -
0.76870611],
       [-1.46723718, -0.30831009,  0.73848951,  2.        , -
0.76870611],
       [ 1.46491286, -0.19745694,  1.39810017, -0.5       ,
1.30088727],
       [ 1.89486118,  0.51005662,  2.07686138, -0.5       ,
1.30088727],
       [-1.49419935,  0.4529585 , -1.53809178, -0.5       , -
0.76870611],
       [-1.0396359 ,  1.20524087, -1.32098255, -0.5       , -
0.76870611]])
```

```
#Model Creation by using Support Vector regression algorithm, here
importing SVR algorithm from sklearn.svm
from sklearn.svm import SVR
#regressor=SVR(kernel="linear")#linear
#regressor=SVR(kernel="linear",C=1000, gamma="auto")#linear
#regressor=SVR(kernel="rbf")#non linear
#regressor=SVR(kernel="rbf",C=0.01)#non linear
#regressor=SVR(kernel="rbf",C=10)#non linear
#regressor=SVR(kernel="rbf",C=100)#non linear
regressor=SVR(kernel="rbf",C=1000, gamma=0.1, epsilon=0.1)#non linear
#regressor=SVR(kernel="poly",C=100, gamma="auto", degree=3,
epsilon=0.1, coef0=1)#non linear
#regressor=SVR(kernel="sigmoid")#non linear
regressor.fit(x_train,y_train)
```

```
C:\Users\Maheshwaran\anaconda3\Lib\site-packages\sklearn\utils\
validation.py:1184: DataConversionWarning: A column-vector y was
passed when a 1d array was expected. Please change the shape of y to
```

```
(n_samples, ), for example using ravel().
  y = column_or_1d(y, warn=True)
```

```
SVR(C=1000, gamma=0.1)
```

```
#from where the orgin start
regressor.intercept_
```

```
array([108527.21813668])
```

```
#how many support vector / data points taken
regressor.n_support_
```

```
array([35])
```

```
regressor.support_
```

```
array([ 0,  1,  2,  3,  4,  5,  6,  7,  8,  9, 10, 11, 12, 13, 14, 15,
16,
       17, 18, 19, 20, 21, 22, 23, 24, 25, 26, 27, 28, 29, 30, 31, 32,
33,
       34])
```

```
y_pred=regressor.predict(x_test)
```

```
#import r2_score from sklean.metrics functions
from sklearn.metrics import r2_score
#compare actual test (y_test) vs predicted value on x_test which is
assign in y_pred
r_score=r2_score(y_test,y_pred)
```

```
r_score
```

```
0.029274514540696406
```