Multiple Linear Regression

Importing the libraries

```
In [10]:
         import numpy as np
         import matplotlib.pyplot as plt
         import pandas as pd
```

Importing the dataset

```
dataset = pd.read_csv('50_Startups.csv')
In [11]:
         X = dataset.iloc[:, :-1].values
         y = dataset.iloc[:, -1].values
In [12]: print(x)
```

```
[[165349.2 136897.8 471784.1 'New York']
          [162597.7 151377.59 443898.53 'California']
          [153441.51 101145.55 407934.54 'Florida']
          [144372.41 118671.85 383199.62 'New York']
          [142107.34 91391.77 366168.42 'Florida']
          [131876.9 99814.71 362861.36 'New York']
          [134615.46 147198.87 127716.82 'California']
          [130298.13 145530.06 323876.68 'Florida']
          [120542.52 148718.95 311613.29 'New York']
          [123334.88 108679.17 304981.62 'California']
          [101913.08 110594.11 229160.95 'Florida']
          [100671.96 91790.61 249744.55 'California']
          [93863.75 127320.38 249839.44 'Florida']
          [91992.39 135495.07 252664.93 'California']
          [119943.24 156547.42 256512.92 'Florida']
          [114523.61 122616.84 261776.23 'New York']
          [78013.11 121597.55 264346.06 'California']
          [94657.16 145077.58 282574.31 'New York']
          [91749.16 114175.79 294919.57 'Florida']
          [86419.7 153514.11 0.0 'New York']
          [76253.86 113867.3 298664.47 'California']
          [78389.47 153773.43 299737.29 'New York']
          [73994.56 122782.75 303319.26 'Florida']
          [67532.53 105751.03 304768.73 'Florida']
          [77044.01 99281.34 140574.81 'New York']
          [64664.71 139553.16 137962.62 'California']
          [75328.87 144135.98 134050.07 'Florida']
          [72107.6 127864.55 353183.81 'New York']
          [66051.52 182645.56 118148.2 'Florida']
          [65605.48 153032.06 107138.38 'New York']
          [61994.48 115641.28 91131.24 'Florida']
          [61136.38 152701.92 88218.23 'New York']
          [63408.86 129219.61 46085.25 'California']
          [55493.95 103057.49 214634.81 'Florida']
          [46426.07 157693.92 210797.67 'California']
          [46014.02 85047.44 205517.64 'New York']
          [28663.76 127056.21 201126.82 'Florida']
          [44069.95 51283.14 197029.42 'California']
          [20229.59 65947.93 185265.1 'New York']
          [38558.51 82982.09 174999.3 'California']
          [28754.33 118546.05 172795.67 'California']
          [27892.92 84710.77 164470.71 'Florida']
          [23640.93 96189.63 148001.11 'California']
          [15505.73 127382.3 35534.17 'New York']
          [22177.74 154806.14 28334.72 'California']
          [1000.23 124153.04 1903.93 'New York']
          [1315.46 115816.21 297114.46 'Florida']
          [0.0 135426.92 0.0 'California']
          [542.05 51743.15 0.0 'New York']
          [0.0 116983.8 45173.06 'California']]
In [14]:
         print(y)
         [192261.83 191792.06 191050.39 182901.99 166187.94 156991.12 156122.51
          155752.6 152211.77 149759.96 146121.95 144259.4 141585.52 134307.35
          132602.65 129917.04 126992.93 125370.37 124266.9 122776.86 118474.03
          111313.02 110352.25 108733.99 108552.04 107404.34 105733.54 105008.31
          103282.38 101004.64 99937.59 97483.56 97427.84 96778.92
                                                                        96712.8
           96479.51 90708.19 89949.14 81229.06 81005.76 78239.91
                                                                        77798.83
           71498.49 69758.98 65200.33 64926.08 49490.75 42559.73 35673.41
           14681.4 ]
```

Encoding categorical data

In [16]:

```
ct = ColumnTransformer(transformers=[('encoder', OneHotEncoder(), [3])], remainder='passthrou
         x = np.array(ct.fit transform(x))
In [17]: print(x)
         [[0.0 0.0 1.0 165349.2 136897.8 471784.1]
          [1.0 0.0 0.0 162597.7 151377.59 443898.53]
          [0.0 1.0 0.0 153441.51 101145.55 407934.54]
          [0.0 0.0 1.0 144372.41 118671.85 383199.62]
          [0.0 1.0 0.0 142107.34 91391.77 366168.42]
          [0.0 0.0 1.0 131876.9 99814.71 362861.36]
          [1.0 0.0 0.0 134615.46 147198.87 127716.82]
          [0.0 1.0 0.0 130298.13 145530.06 323876.68]
          [0.0 0.0 1.0 120542.52 148718.95 311613.29]
          [1.0 0.0 0.0 123334.88 108679.17 304981.62]
          [0.0 1.0 0.0 101913.08 110594.11 229160.95]
          [1.0 0.0 0.0 100671.96 91790.61 249744.55]
          [0.0 1.0 0.0 93863.75 127320.38 249839.44]
           [1.0 0.0 0.0 91992.39 135495.07 252664.93]
          [0.0 1.0 0.0 119943.24 156547.42 256512.92]
          [0.0 0.0 1.0 114523.61 122616.84 261776.23]
          [1.0 0.0 0.0 78013.11 121597.55 264346.06]
          [0.0 0.0 1.0 94657.16 145077.58 282574.31]
          [0.0 1.0 0.0 91749.16 114175.79 294919.57]
          [0.0 0.0 1.0 86419.7 153514.11 0.0]
          [1.0 0.0 0.0 76253.86 113867.3 298664.47]
          [0.0 0.0 1.0 78389.47 153773.43 299737.29]
          [0.0 1.0 0.0 73994.56 122782.75 303319.26]
          [0.0 1.0 0.0 67532.53 105751.03 304768.73]
           [0.0 0.0 1.0 77044.01 99281.34 140574.81]
          [1.0 0.0 0.0 64664.71 139553.16 137962.62]
          [0.0 1.0 0.0 75328.87 144135.98 134050.07]
          [0.0 0.0 1.0 72107.6 127864.55 353183.81]
          [0.0 1.0 0.0 66051.52 182645.56 118148.2]
          [0.0 0.0 1.0 65605.48 153032.06 107138.38]
           [0.0 1.0 0.0 61994.48 115641.28 91131.24]
          [0.0 0.0 1.0 61136.38 152701.92 88218.23]
          [1.0 0.0 0.0 63408.86 129219.61 46085.25]
          [0.0 1.0 0.0 55493.95 103057.49 214634.81]
          [1.0 0.0 0.0 46426.07 157693.92 210797.67]
          [0.0 0.0 1.0 46014.02 85047.44 205517.64]
          [0.0 1.0 0.0 28663.76 127056.21 201126.82]
          [1.0 0.0 0.0 44069.95 51283.14 197029.42]
          [0.0 0.0 1.0 20229.59 65947.93 185265.1]
          [1.0 0.0 0.0 38558.51 82982.09 174999.3]
          [1.0 0.0 0.0 28754.33 118546.05 172795.67]
           [0.0 1.0 0.0 27892.92 84710.77 164470.71]
          [1.0 0.0 0.0 23640.93 96189.63 148001.11]
          [0.0 0.0 1.0 15505.73 127382.3 35534.17]
          [1.0 0.0 0.0 22177.74 154806.14 28334.72]
          [0.0 0.0 1.0 1000.23 124153.04 1903.93]
          [0.0 1.0 0.0 1315.46 115816.21 297114.46]
           [1.0 0.0 0.0 0.0 135426.92 0.0]
          [0.0 0.0 1.0 542.05 51743.15 0.0]
          [1.0 0.0 0.0 0.0 116983.8 45173.06]]
```

from sklearn.preprocessing import OneHotEncoder

Splitting the dataset into the Training set and Test set

```
In [18]: from sklearn.model_selection import train_test_split
x_train, x_test, y_train, y_test = train_test_split(x, y, test_size=0.2, random_state=0)
```

```
In [19]: print(x_train)
```

```
[[0.0 1.0 0.0 55493.95 103057.49 214634.81]
          [0.0 0.0 1.0 46014.02 85047.44 205517.64]
          [0.0 1.0 0.0 75328.87 144135.98 134050.07]
           [1.0 0.0 0.0 46426.07 157693.92 210797.67]
          [0.0 1.0 0.0 91749.16 114175.79 294919.57]
          [0.0 1.0 0.0 130298.13 145530.06 323876.68]
          [0.0 1.0 0.0 119943.24 156547.42 256512.92]
          [0.0 0.0 1.0 1000.23 124153.04 1903.93]
           [0.0 0.0 1.0 542.05 51743.15 0.0]
          [0.0 0.0 1.0 65605.48 153032.06 107138.38]
          [0.0 0.0 1.0 114523.61 122616.84 261776.23]
          [0.0 1.0 0.0 61994.48 115641.28 91131.24]
          [1.0 0.0 0.0 63408.86 129219.61 46085.25]
          [1.0 0.0 0.0 78013.11 121597.55 264346.06]
          [1.0 0.0 0.0 23640.93 96189.63 148001.11]
          [1.0 0.0 0.0 76253.86 113867.3 298664.47]
          [0.0 0.0 1.0 15505.73 127382.3 35534.17]
          [0.0 0.0 1.0 120542.52 148718.95 311613.29]
          [1.0 0.0 0.0 91992.39 135495.07 252664.93]
          [1.0 0.0 0.0 64664.71 139553.16 137962.62]
          [0.0 0.0 1.0 131876.9 99814.71 362861.36]
          [0.0 0.0 1.0 94657.16 145077.58 282574.31]
          [1.0 0.0 0.0 28754.33 118546.05 172795.67]
          [1.0 0.0 0.0 0.0 116983.8 45173.06]
          [1.0 0.0 0.0 162597.7 151377.59 443898.53]
          [0.0 1.0 0.0 93863.75 127320.38 249839.44]
          [1.0 0.0 0.0 44069.95 51283.14 197029.42]
          [0.0 0.0 1.0 77044.01 99281.34 140574.81]
          [1.0 0.0 0.0 134615.46 147198.87 127716.82]
          [0.0 1.0 0.0 67532.53 105751.03 304768.73]
          [0.0 1.0 0.0 28663.76 127056.21 201126.82]
           [0.0 0.0 1.0 78389.47 153773.43 299737.29]
          [0.0 0.0 1.0 86419.7 153514.11 0.0]
          [1.0 0.0 0.0 123334.88 108679.17 304981.62]
          [1.0 0.0 0.0 38558.51 82982.09 174999.3]
          [0.0 1.0 0.0 1315.46 115816.21 297114.46]
          [0.0 0.0 1.0 144372.41 118671.85 383199.62]
          [0.0 0.0 1.0 165349.2 136897.8 471784.1]
          [1.0 0.0 0.0 0.0 135426.92 0.0]
          [1.0 0.0 0.0 22177.74 154806.14 28334.72]]
In [20]: print(x_test)
         [[0.0 1.0 0.0 66051.52 182645.56 118148.2]
          [1.0 0.0 0.0 100671.96 91790.61 249744.55]
          [0.0 1.0 0.0 101913.08 110594.11 229160.95]
           [0.0 1.0 0.0 27892.92 84710.77 164470.71]
          [0.0 1.0 0.0 153441.51 101145.55 407934.54]
          [0.0 0.0 1.0 72107.6 127864.55 353183.81]
          [0.0 0.0 1.0 20229.59 65947.93 185265.1]
          [0.0 0.0 1.0 61136.38 152701.92 88218.23]
          [0.0 1.0 0.0 73994.56 122782.75 303319.26]
          [0.0 1.0 0.0 142107.34 91391.77 366168.42]]
In [21]: print(y_train)
         [ 96778.92 96479.51 105733.54 96712.8 124266.9 155752.6 132602.65
           64926.08 35673.41 101004.64 129917.04 99937.59 97427.84 126992.93
           71498.49 118474.03 69758.98 152211.77 134307.35 107404.34 156991.12
          125370.37 78239.91 14681.4 191792.06 141585.52 89949.14 108552.04
          156122.51 108733.99 90708.19 111313.02 122776.86 149759.96 81005.76
           49490.75 182901.99 192261.83 42559.73 65200.33]
In [22]: print(y_test)
         [103282.38 144259.4 146121.95 77798.83 191050.39 105008.31 81229.06
```

97483.56 110352.25 166187.94]

Training the Multiple Linear Regression model on the Training set

```
In [26]: from sklearn.linear_model import LinearRegression
    model = LinearRegression()
    model.fit(x_train, y_train)

Out[26]: LinearRegression()
```

Predicting the Test set results