Multiple Linear Regression

October 17, 2021

Multiple Linear Regression

0.1 Importing the Libraries

```
[1]: import numpy as np
     import pandas as pd
     import matplotlib.pyplot as plt
[2]: from sklearn.compose import ColumnTransformer
     ## encoding categorical data
     from sklearn.preprocessing import OneHotEncoder
[3]: from sklearn.model_selection import train_test_split
[4]: ## Training and model selection
     from sklearn.linear_model import LinearRegression
[5]: import statsmodels.api as sm
    0.2 Importing the Data set
[6]: data_set = pd.read_csv("50_startups.csv")
[7]: x = data_set.iloc[:,:-1].values
     y = data set.iloc[:,-1].values
[8]: 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']
```

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[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']]
[9]: 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 ]
```

0.3 Encoding categorical data

```
[10]: ct = ColumnTransformer(transformers = [('encoder', OneHotEncoder(), [3])],
      →remainder = 'passthrough')
      x = ct.fit transform(x)
      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]]
     0.4 Spliting dataset into training set and test set
[11]: x_train,x_test,y_train,y_test = train_test_split(x,y,test_size=0.2,random_state_
       →= 0)
     0.5 Training Multiple linear Regression Model on Training set
[12]: Regressor = LinearRegression()
[13]: Regressor.fit(x_train,y_train)
[13]: LinearRegression()
     0.6 Predict Test Result
[14]: y_predict = Regressor.predict(x_test)
      y_predict
[14]: array([103015.20159796, 132582.27760816, 132447.73845174, 71976.09851258,
             178537.48221055, 116161.24230166,
                                                67851.69209676, 98791.73374686,
             113969.43533013, 167921.06569551])
[15]: | np.set_printoptions(precision = 2)
      print(np.concatenate((y predict.reshape(len(y predict),1), y test.
       \rightarrowreshape(len(y_test),1)),1))
     [[103015.2 103282.38]
      [132582.28 144259.4 ]
      [132447.74 146121.95]
      [ 71976.1
                  77798.831
      [178537.48 191050.39]
      [116161.24 105008.31]
      [ 67851.69 81229.06]
      [ 98791.73 97483.56]
```

[113969.44 110352.25] [167921.07 166187.94]]