

MReg

April 14, 2022

```
[1]: # Multiple Linear Regression
# Importing the libraries

import numpy as np
import matplotlib.pyplot as plt
import pandas as pd
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[4]: # Importing the dataset

dataset = pd.read_csv('D:/50_Startups.csv')
X = dataset.iloc[:, :-1].values
y = dataset.iloc[:, -1].values
print(X)
```

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[[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']
```

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[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']]
```

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[5]: # Encoding categorical data
from sklearn.compose import ColumnTransformer
from sklearn.preprocessing import OneHotEncoder
ct = ColumnTransformer(transformers=[('encoder', OneHotEncoder(), [3])],
    ↳remainder='passthrough')
X = np.array(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]]
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[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]]

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[6]: *# Splitting the dataset into the Training set and Test set*

```

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)

```

[7]: *# Training the Multiple Linear Regression model on the Training set*

```
from sklearn.linear_model import LinearRegression
regressor = LinearRegression()
regressor.fit(X_train, y_train)
```

[7]: LinearRegression(copy_X=True, fit_intercept=True, n_jobs=None, normalize=False)

```
[8]: # Predicting the Test set results
y_pred = regressor.predict(X_test)
np.set_printoptions(precision=2)
print(np.concatenate((y_pred.reshape(len(y_pred),1), y_test.
    ↳reshape(len(y_test),1)),1))
```

```
[[103015.2  103282.38]
 [132582.28 144259.4 ]
 [132447.74 146121.95]
 [ 71976.1   77798.83]
 [178537.48 191050.39]
 [116161.24 105008.31]
 [ 67851.69  81229.06]
 [ 98791.73  97483.56]
 [113969.44 110352.25]
 [167921.07 166187.94]]
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