# Predicting a Startups Profit/Success Rate using Multiple Linear Regression in Python

Here 50 startups dataset containing 5 columns like "R&D Spend", "Administration", "Marketing Spend", "State", "Profit".

In this dataset first 3 columns provides you spending on Research , Administration and Marketing respectively. State indicates startup based on that state. Profit indicates how much profits earned by a startup.

Clearly, we can understand that it is a multiple linear regression problem, as the independent variables are more than one.

Prepare a prediction model for profit of 50\_Startups data in Python

```
In [62]: import pandas as pd
import numpy as np
from sklearn.model_selection import train_test_split

startup_data = pd.read_csv('50Startups.csv')

print(startup_data.shape)
startup_data.head()

(50, 5)
```

#### Out[62]:

|   | 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 |

### **Pre-processing**

```
In [49]: #One hot encoding using dummies. Change the state column into dummies
    state_dummies = pd.get_dummies(startup_data['State'])

#Add the state_dummies into the data along the columns
    merged_df = pd.concat([startup_data, state_dummies], axis=1)

#drop the State column
    final_data = merged_df.drop(['State'], axis=1)

#Drop one of the dummies columns to avoid dummy variable trap
    final_data = final_data.drop(['New York'], axis=1)
    final_data.head()
```

#### Out[49]:

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

```
In [50]: X = final_data.drop('Profit', axis=1)
y = final_data['Profit']

X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.25, random_print(X_train.shape)
print(X_test.shape)
```

(37, 5) (13, 5)

## In [51]: #Create the Linear regression model from sklearn.linear model import Li

from sklearn.linear\_model import LinearRegression
lr\_model = LinearRegression()
#Train the model

lr\_model.fit(X\_train, y\_train)

#predict using the testing set
y\_pred = lr\_model.predict(X\_test)

#### In [65]: #Check model performance

```
from sklearn.metrics import r2_score, mean_squared_error
score = r2_score(y_pred, y_test)
MSE = mean_squared_error(y_test, y_pred)
print(f'Coefficient of Determination, R {score}')
```

Coefficient of Determination, R 0.9477890107662057

```
In [61]: #Do prediction using new data

new_pred = lr_model.predict([[165350, 136898, 471784, 0, 0]])
print(f' Profit for new data is {new_pred}')
```

Profit for new data is [191997.78365278]