## **Practical 3**

# Aim: To build the model for prediction of profit

#### Importing the libraries

```
import pandas as pd
import numpy as np
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LinearRegression
from sklearn.metrics import mean_squared_error, r2_score
```

#### Reading the dataset and taking into dataframe

```
In [13]: file_path = 'practical2.csv'
In [14]: df = pd.read_csv(file_path)
```

#### reading head of data

```
In [15]: print(df.head())
            R&D Spend Administration Marketing Spend
                                                           State
                                                                    Profit
          165349.20
                           136897.80
                                           471784.10
                                                        New York 192261.83
         1 162597.70
                                           443898.53 California 191792.06
                           151377.59
         2 153441.51
                           101145.55
                                           407934.54
                                                        Florida 191050.39
           144372.41
                           118671.85
                                           383199.62
                                                        New York 182901.99
         4 142107.34
                            91391.77
                                           366168.42
                                                       Florida 166187.94
```

# Selecting relevant columns for the model

As correlation coeafficient with R&D Spend is more than any other attribute so we will selecting it for our prediction model as Independent variable

```
In [16]: x = df[['R&D Spend']]
In [17]: X=x.to_numpy()
```

#### **Target variable**

```
In [18]: y = df['Profit']
In [19]: Y=y.to_numpy()
```

#### Splitting the data into training and testing sets

```
In [20]: x_train, x_test, y_train, y_test = train_test_split(X, Y, test_size=0.2, ra
In [21]: print(x_train.shape)
    print(x_test.shape)
    print(y_train.shape)
    print(y_test.shape)

    (40, 1)
    (10, 1)
    (40,)
    (10,)
```

### Training the model

```
In [22]: from sklearn.linear_model import LinearRegression
    regressor=LinearRegression()
    regressor.fit(x_train,y_train) # Training the algorithm
```

Out[22]: LinearRegression()

In a Jupyter environment, please rerun this cell to show the HTML representation or trust the notebook.

On GitHub, the HTML representation is unable to render, please try loading this page with nbviewer.org.

#### Predictions on the test set

#### **Evaluating the model**

```
In [34]:
        print(f'The Error y-y`: {y_test-y_pred}')
         The Error y-y`: [-1384.89805998 10108.56589422 10914.14980483 5628.285711
         44
          11959.80397492 -4816.46386586 15584.78226243 -2997.87277139
          -1079.50202432 -3250.20843539]
In [42]: e_array=(y_test-y_pred)**2
         print(f'The Error y-y`: {e_array}')
         The Error y-y`: [1.91794264e+06 1.02183104e+08 1.19118666e+08 3.16776000e+
          1.43036911e+08 2.31983242e+07 2.42885438e+08 8.98724115e+06
          1.16532462e+06 1.05638549e+07]
In [44]: # Calculate the sum of elements
         array_sum = np.sum(e_array)
         # Display the result
         print("Sum of array elements:", array_sum)
         # Taking the Mean
         print("Mean of array elements:", array_sum/10)
         Sum of array elements: 684734407.1905932
         Mean of array elements: 68473440.71905932
```

# Above are the actual calculation and below using the predefined functions