

Assignment 01: Evaluate the Ad Budget Dataset of XYZ Firm

The comments/sections provided are your cues to perform the assignment. You don't need to limit yourself to the number of rows/cells provided. You can add additional rows in each section to add more lines of code.

If at any point in time you need help on solving this assignment, view our demo video to understand the different steps of the code.

Happy coding!

1: Import the dataset

```
In [143]:
#Import the required libraries
import numpy as np
import pandas as pd
```

In [144]: #Import the advertising dataset advertising = pd.read csv('Advertising Budget and Sales.csv',index col= 0)

2: Analyze the dataset

In [145]: #View the initial few records of the dataset advertising.head()

Out[145]:

2

45.9 69.3 9.3 17.2 4 151.5 41.3 58.5 18.5 180.8 58.4 12.9 In [146]: #Check the total number of elements in the dataset

TV Ad Budget (\$) Radio Ad Budget (\$) Newspaper Ad Budget (\$) Sales (\$)

37.8

39.3

print(advertising.size) print(advertising.shape[0]*advertising.shape[1]) 800 800

45.1

10.4

In [147]: | #Check the number of observations (rows) and attributes (columns) in the dataset advertising.shape

> 'Sales (\$)'], dtype='object')

3: Find the features or media channels used by the firm

44.5

Out[147]: (200, 4)

In [148]: | #View the names of each of the attributes advertising.columns Out[148]: Index(['TV Ad Budget (\$)', 'Radio Ad Budget (\$)', 'Newspaper Ad Budget (\$)',

4: Create objects to train and test the model; find the sales figures for each channel

TV Ad Budget (\$) Radio Ad Budget (\$) Newspaper Ad Budget (\$)

X = advertising.drop(columns='Sales (\$)') In [150]: | #View the feature object X.head()

69.2

1 230.1 37.8 2 44.5 39.3

Y = advertising[['Sales (\$)']]

In [149]: | #Create a feature object from the columns

45.1 3 17.2 45.9 69.3 4 151.5 41.3 58.5 5 180.8 10.8 58.4 In [151]: | #Create a target object (Hint: use the sales column as it is the response of the dataset)

In [152]: #View the target object Y.head()

Sales (\$) 22.1

2

3

4

5

Y.shape

Out[154]: (200, 1)

10.4

9.3

18.5

12.9

Out[152]:

Out[150]:

In [153]: | #Verify if all the observations have been captured in the feature object X.shape Out[153]: (200, 3)

5: Split the original dataset into training and testing datasets for the model

print(x_train.shape , x_test.shape , y_train.shape , y_test.shape)

In [154]: | #Verify if all the observations have been captured in the target object

In [155]: | #Split the dataset (by default, 75% is the training data and 25% is the testing data) from sklearn.model_selection import train_test_split x_train, x_test , y_train , y_test = train_test_split(X,Y,test_size=0.25, random_state=1) In [156]: | #Verify if the training and testing datasets are split correctly (Hint: use the shape() method)

6: Create a model to predict the sales outcome In [157]: #Create a linear regression model from sklearn.linear model import LinearRegression

lr.intercept_, lr.coef_ Out[158]: (array([2.87696662]), array([[0.04656457, 0.17915812, 0.00345046]]))

In [159]: #Predict the outcome for the testing dataset

In [158]: #Print the intercept and coefficients

[18.6146359], [23.83573998],

[9.83511973], [17.18797614], [16.73086831], [15.05529391], [15.61434433], [12.42541574], [17.17716376],

lr.predict(x_test)

lr = LinearRegression().fit(x train, y train)

(150, 3) (50, 3) (150, 1) (50, 1)

Out[159]: array([[21.70910292], [16.41055243], [7.60955058], [17.80769552],

[16.32488681], [13.43225536], [9.17173403], [17.333853], [14.44479482],

[11.08827566], [18.00537501], [9.28438889], [12.98458458], [8.79950614], [10.42382499], [11.3846456], [14.98082512], [9.78853268], [19.39643187], [18.18099936], [17.12807566], [21.54670213], [14.69809481], [16.24641438], [12.32114579], [19.92422501], [15.32498602], [13.88726522], [10.03162255], [20.93105915], [7.44936831], [3.64695761], [7.22020178], [5.9962782], [18.43381853], [8.39408045], [14.08371047], [15.02195699],

7: Calculate the Mean Square Error (MSE)

[20.35836418], [20.57036347], [19.60636679]])

In [160]: | y_hat = lr.predict(x_test)

 $y = y_test$

In [177]: | #Calculate the MSE

```
In [175]: | #Import required libraries for calculating MSE (mean square error)
from sklearn import metrics
```

mean squared error = np.sqrt(metrics.mean squared error(y,y hat)) In [178]: print('the mean_squared_error is {}'.format(mean_squared_error))

the mean squared error is 1.404651423032895 In []: