

# Analysing Ad Budgets for different media channels

April 24, 2022

## 1 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!**

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### 1: Import the dataset

```
[1]: #Import the required libraries
import pandas as pd
```

```
[2]: #Import the advertising dataset
df_adv_data = pd.read_csv('Advertising Budget and Sales.csv', index_col=0)
```

### 2: Analyze the dataset

```
[3]: #View the initial few records of the dataset
df_adv_data.head()
```

```
[3]:
```

	TV Ad Budget (\$)	Radio Ad Budget (\$)	Newspaper Ad Budget (\$)	Sales (\$)
1	230.1	37.8	69.2	22.1
2	44.5	39.3	45.1	10.4
3	17.2	45.9	69.3	9.3
4	151.5	41.3	58.5	18.5
5	180.8	10.8	58.4	12.9

```
[4]: #Check the total number of elements in the dataset
df_adv_data.size
```

```
[4]: 800
```

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

```
[5]: #Check the number of observations (rows) and attributes (columns) in the dataset
df_adv_data.shape
```

```
[5]: (200, 4)
```

```
[6]: #View the names of each of the attributes
df_adv_data.columns
```

```
[6]: Index(['TV Ad Budget ($)', 'Radio Ad Budget ($)', 'Newspaper Ad Budget ($)',
         'Sales ($)'],
         dtype='object')
```

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

```
[7]: #Create a feature object from the columns
X_feature = df_adv_data[['Newspaper Ad Budget ($)', 'Radio Ad Budget ($)', 'TV Ad_Budget ($)']]
```

```
[8]: #View the feature object
X_feature.head()
```

```
[8]:
```

	Newspaper Ad Budget (\$)	Radio Ad Budget (\$)	TV Ad Budget (\$)
1	69.2	37.8	230.1
2	45.1	39.3	44.5
3	69.3	45.9	17.2
4	58.5	41.3	151.5
5	58.4	10.8	180.8

```
[9]: #Create a target object (Hint: use the sales column as it is the response of_
     ↪ the dataset)
Y_target = df_adv_data[['Sales ($)']]
```

```
[10]: #View the target object
Y_target.head()
```

```
[10]:
```

	Sales (\$)
1	22.1
2	10.4
3	9.3
4	18.5
5	12.9

```
[11]: #Verify if all the observations have been captured in the feature object
X_feature.shape
```

```
[11]: (200, 3)
```

```
[12]: #Verify if all the observations have been captured in the target object
Y_target.shape
```

```
[12]: (200, 1)
```

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

```
[13]: #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_feature,Y_target,random_state=1)
```

```
[14]: #Verify if the training and testing datasets are split correctly (Hint: use the
      ↪shape() method)
      print(x_train.shape)
      print(x_test.shape)
      print(y_train.shape)
      print(y_test.shape)
```

(150, 3)

(50, 3)

(150, 1)

(50, 1)

## 6: Create a model to predict the sales outcome

```
[15]: #Create a linear regression model
      from sklearn.linear_model import LinearRegression
      linreg = LinearRegression()
      linreg.fit(x_train,y_train)
```

```
[15]: LinearRegression()
```

```
[16]: #Print the intercept and coefficients
      print(linreg.intercept_)
      print(linreg.coef_)
```

[2.87696662]

[[0.00345046 0.17915812 0.04656457]]

```
[17]: #Predict the outcome for the testing dataset
      y_pred = linreg.predict(x_test)
      y_pred
```

```
[17]: array([[21.70910292],
      [16.41055243],
      [ 7.60955058],
      [17.80769552],
      [18.6146359 ],
      [23.83573998],
      [16.32488681],
      [13.43225536],
      [ 9.17173403],
      [17.333853  ],
```

```
[14.44479482],
[ 9.83511973],
[17.18797614],
[16.73086831],
[15.05529391],
[15.61434433],
[12.42541574],
[17.17716376],
[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],
[20.35836418],
[20.57036347],
[19.60636679]])
```

## 7: Calculate the Mean Square Error (MSE)

```
[18]: #Import required libraries for calculating MSE (mean square error)
from sklearn import metrics
import numpy as np
```

```
[19]: #Calculate the MSE
print(np.sqrt(metrics.mean_squared_error(y_test,y_pred)))
```

1.4046514230328957

```
[20]: print('True' , y_test.values[0:10])
print()
print('Pred' , y_pred[0:10])
```

True [[23.8]

[16.6]

[ 9.5]

[14.8]

[17.6]

[25.5]

[16.9]

[12.9]

[10.5]

[17.1]]

Pred [[21.70910292]

[16.41055243]

[ 7.60955058]

[17.80769552]

[18.6146359 ]

[23.83573998]

[16.32488681]

[13.43225536]

[ 9.17173403]

[17.333853 ]]

```
[ ]:
```