

Our goal is to predict sales on the basis of the three media budgets TV, Radio and Newspaper.



Machine Learning Model Deployment

Dataset Information

The Advertising dataset consists of the sales of that product in 200 different markets, along with advertising budgets for the product in each of those markets for three different media: TV, radio and newspaper.

We try to determine that there is an association between advertising and sales, so that we can instruct our clients to adjust advertising budgets, thereby indirectly increasing sales.

This dataset shared by Marshall School of Business at the University of Southern California. If you want, you can <u>download</u>

Load Data

Output:

U	nnamed: 0	TV	radio	newspaper	sales
0	1	230.1	37.8	69.2	22.1
1	2	44.5	39.3	45.1	10.4
2	3	17.2	45.9	69.3	9.3
3	4	151.5	41.3	5 <mark>8.</mark> 5	18.5
4	5	180.8	10.8	58.4	12.9

df.columns

Output:

```
Index(['Unnamed: 0', 'TV', 'radio', 'newspaper', 'sales'], dtype='object')
```

df.info()

Output:

Drop unnamed column

```
df = df.drop(['Unnamed: 0'], axis=1)
df.head()
```

Output:

	TV	radio	newspaper	sales
0	230.1	37.8	69.2	22.1
1	44.5	39.3	45.1	10.4
2	17.2	45.9	69.3	9.3
3	151.5	41.3	58.5	18.5
4	180.8	10.8	58.4	12.9

Split data into feature and target

```
X = df.loc[:, df.columns != 'sales']
X.head()
```

Output:

	TV	radio	newspaper
0	230.1	37.8	69.2
1	44.5	39.3	45.1
2	17.2	45.9	69.3
3	151.5	41.3	58.5
4	180.8	10.8	58.4

```
y = df['sales']
y.head()
```

Output:

```
0 22.1
1 10.4
2 9.3
3 18.5
4 12.9
Name: sales, dtype: float64
```

Split data into train and test

```
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)
print(X_train.shape)
print(X_train.head())
```

Output:

```
(150, 3)

TV radio newspaper
155 4.1 11.6 5.7
174 222.4 3.4 13.1
184 253.8 21.3 30.0
23 228.3 16.9 26.2
133 219.8 33.5 45.1
```

```
print(X_test.shape)
print(X_test.head())
```

Output:

```
TV radio newspaper
122 224.0 2.4 15.6
186 139.5 2.1 26.6
5 8.7 48.9 75.0
55 198.9 49.4 60.0
194 149.7 35.6 6.0
```

```
print(y_train.shape)
print(y_train.head())
```

Output:

```
(150,)

155 3.2

174 11.5

184 17.6

23 15.5

133 19.6

Name: sales, dtype: float64
```

```
print(y_test.shape)
print(y_test.head())
```

Output:

Model creation

```
from sklearn.linear_model import LinearRegression
model = LinearRegression()
```

Training

```
model.fit(X_train, y_train)
```

Predict

```
predictions = model.predict(X_test)
predictions[:10]
```

Output:

```
array([13.63521962, 9.78796421, 12.43510316, 21.10986043, 16.14591182, 22.97996462, 12.24553359, 10.14917755, 15.21272804, 18.44335937])
```

Save the model

joblib module

We are using joblib library to save our model.

joblib.dump() and joblib.load() provide a replacement for pickle to work efficiently on arbitrary Python objects containing large data, in particular large numpy arrays.

joblib.dump(value, filename, compress=0, protocol=None, cache_size=None): Persist an arbitrary Python object into one file.

```
import joblib
joblib.dump(model,'output/lr_model.pkl')
```

Predict with new data

```
import numpy as np
test_data = [159.1, 60.2, 90]
#Convert into numpy array and reshape
test_data = np.array(test_data).astype(np.float)
test_data = test_data.reshape(1,-1)
test_data
```

Load saved model

```
filePath = 'output/lr_model.pkl'
file = open(filePath, "rb")
trained_model = joblib.load(file)
```

Predict

```
prediction = trained_model.predict(test_data)
print(prediction)
```

Output:

[21.37257034]

MI Model Deployment

Model Deployment Flask

MI Deployment On Gcp

Google Cloud Platform

MI Flask Deployment

About Help Terms Privacy