

```

1 import pandas as pd
2
3 # Read the dataset into a Pandas DataFrame
4 data = pd.read_csv('https://raw.githubusercontent.com/shakil1819/CSE442-Machine-Learning-Sessional/main/week%203%20-4/Advertis
5
6 # Print the first few rows of the DataFrame
7 print(data.head())
8
9 # Get descriptive statistics of the DataFrame
10 print(data.describe())
11
12 # Print information about the DataFrame
13 print(data.info())
14

```

```

0   TV  radio  newspaper  sales
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

count    200.000000    200.000000    200.000000    200.000000
mean     147.042500     23.264000     30.554000     14.022500
std       85.854236     14.846809     21.778621      5.217457
min        0.700000        0.000000        0.300000        1.600000
25%       74.375000     9.975000     12.750000     10.375000
50%      149.750000     22.900000     25.750000     12.900000
75%      218.825000     36.525000     45.100000     17.400000
max      296.400000     49.600000    114.000000     27.000000
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 200 entries, 0 to 199
Data columns (total 4 columns):
#   Column      Non-Null Count  Dtype
---  -
0   TV          200 non-null      float64
1   radio       200 non-null      float64
2   newspaper   200 non-null      float64
3   sales       200 non-null      float64
dtypes: float64(4)
memory usage: 6.4 KB
None

```

### 1.7.3 Import Libraries in Google Colab

```

1 import numpy as np
2 import pandas as pd
3 import matplotlib.pyplot as plt
4 import seaborn as sns
5 from sklearn.model_selection import train_test_split
6 from sklearn.linear_model import LinearRegression
7 from sklearn.metrics import mean_squared_error, r2_score

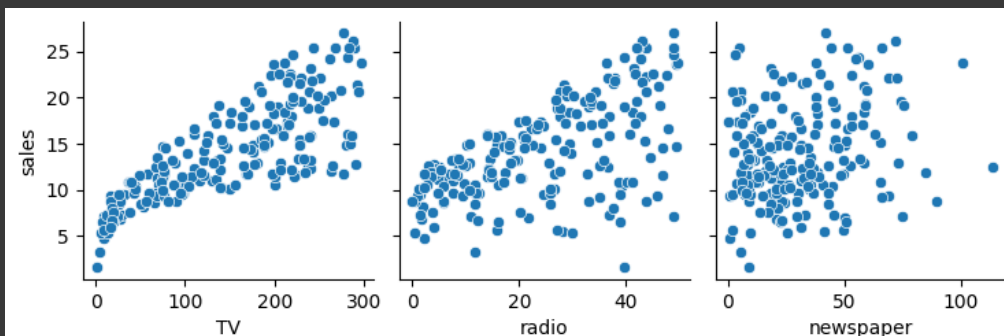
```

### 1.7.5 Plotting the data for understanding

```

1 sns.pairplot(data, x_vars=['TV', 'radio', 'newspaper'], y_vars='sales', kind='scatter')
2 plt.show()

```



### 1.7.6 Plotting the data in different fashion

```
1 # Histograms
2 sns.histplot(data=data, x="TV", bins=20, kde=True)
3 plt.title('TV Advertisement Histogram')
4 plt.xlabel('TV Advertisement')
5 plt.ylabel('Frequency')
6 plt.show()
7
8 # Histogram of radio advertisement
9 sns.histplot(data=data, x='radio', bins=20, kde=True)
10 plt.title('Radio Advertisement Histogram')
11 plt.xlabel('Radio Advertisement')
12 plt.ylabel('Frequency')
13 plt.show()
14
15 # Histogram of newspaper advertisement
16 sns.histplot(data=data, x='newspaper', bins=20, kde=True)
17 plt.title('Newspaper Advertisement Histogram')
18 plt.xlabel('Newspaper Advertisement')
19 plt.ylabel('Frequency')
20 plt.show()
21
22 # Box plots of advertising channels
23 sns.boxplot(data=data[['TV', 'radio', 'newspaper']], orient='horizontal')
24 plt.ylabel('Expenditure')
25 plt.title('Advertisement Channels')
26 plt.show()
27
28 # Correlation matrix
29 correlation_matrix = data[['TV', 'radio', 'newspaper', 'sales']].corr()
30 plt.figure(figsize=(8, 6))
31 sns.heatmap(correlation_matrix, annot=True, cmap='coolwarm', center=8)
32 plt.title('Correlation Matrix')
33 plt.show()
34
35
```



## 1.7.7 Data Pre-processing

### 1.7.7.1 Checking the NULL values

```
1 null_counts = data.isnull().sum()
2 data_cleaned = data.dropna()
3 null_counts
```

```
TV          0
radio       0
newspaper   0
sales       0
dtype: int64
```

### 1.7.7.2 Define dependent (target) and independent (predictor) features

```
1 # Create the independent variable
2 X = data_cleaned[['TV', 'radio', 'newspaper']]
3
4 # Create the dependent variable
5 y = data_cleaned['sales']
```

### 1.7.7.3 Splitting Train and Test data

```
1 X_train,X_test,y_train,y_test = train_test_split(X,y,test_size=0.2,random_state=42)
```

### 1.7.7.4 Feature Scaling for Multiple Linear Regression

```
1 from sklearn.preprocessing import StandardScaler
2 scaler = StandardScaler()
3 X_train_scaled = scaler.fit_transform(X_train)
4 X_test_scaled = scaler.transform(X_test)
```

## 1.7.8 Model Building and Training

### 1.7.8.1 Training the model

```
1 model = LinearRegression()
2 model.fit(X_train_scaled, y_train)
```

```
LinearRegression
LinearRegression()
```

### 1.7.8.2 Predicting for Test Data

```
1 y_pred = model.predict(X_test_scaled)
2
```

## 1.7.9 Model Evaluation

### 1.7.9.1 Calculating Metrics

```

1 mse = mean_squared_error(y_test, y_pred)
2 r2 = r2_score(y_test, y_pred)
3 print("MSE", mse)
4 print("R2", r2)

```

```

MSE 3.174097353976106
R2 0.8994380241009119

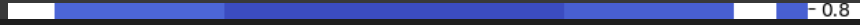
```



## 1.7.10 Visualizing Predictions



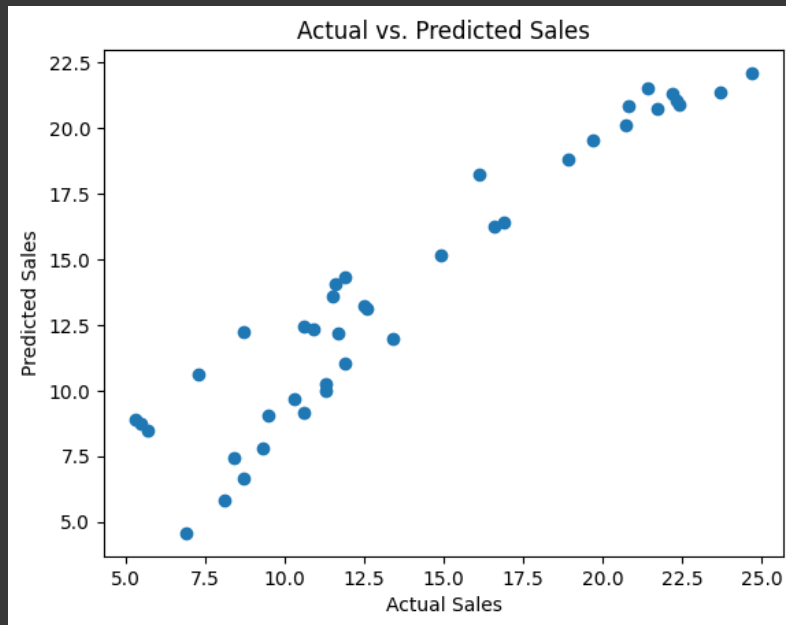
### 1.7.10.1 Visualize predicted vs. actual values



```

1 plt.scatter(y_test, y_pred)
2 plt.xlabel('Actual Sales')
3 plt.ylabel('Predicted Sales')
4 plt.title('Actual vs. Predicted Sales')
5 plt.show()
6

```

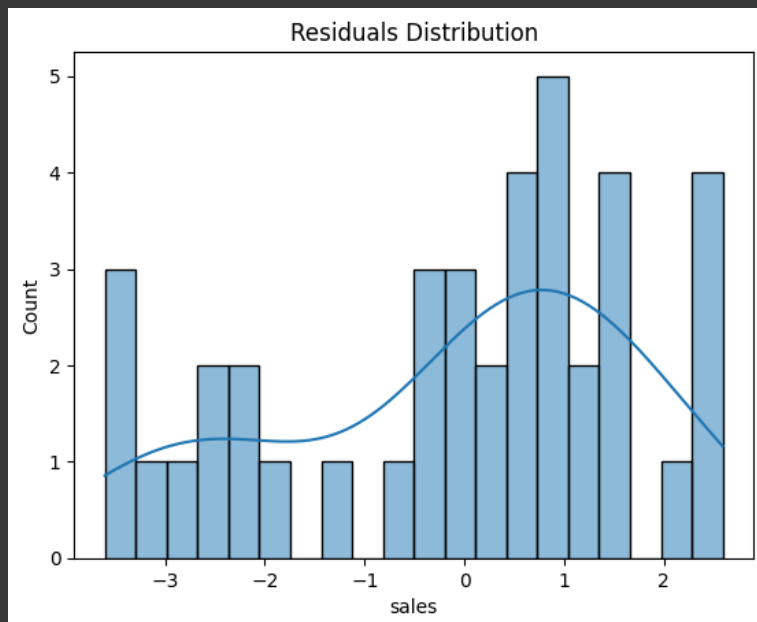


### 1.7.10.2 Visualizing Residuals

```

1 residuals = y_test - y_pred
2 sns.histplot(residuals, bins=20, kde=True)
3 plt.title('Residuals Distribution')
4 plt.show()

```



### 1.7.10.3 Visualizing Coefficients

```
1 coef_df = pd.DataFrame({'feature': X.columns, 'coefficient': model.coef_})
2 sns.barplot(x='coefficient', y='feature', data=coef_df)
3 plt.title('Feature Coefficients')
4 plt.show()
```

