

# sales-prediction

September 27, 2023

## 1 Task - 5 Sales Prediction

```
[2]: import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LinearRegression
from sklearn.metrics import mean_absolute_error, mean_squared_error, r2_score
```

```
[3]: df = pd.read_csv("Advertising.csv")
```

```
[4]: df
```

```
[4]:
```

	Unnamed: 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	58.5	18.5
4	5	180.8	10.8	58.4	12.9
..	...	...	...	...	...
195	196	38.2	3.7	13.8	7.6
196	197	94.2	4.9	8.1	9.7
197	198	177.0	9.3	6.4	12.8
198	199	283.6	42.0	66.2	25.5
199	200	232.1	8.6	8.7	13.4

[200 rows x 5 columns]

```
[5]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 200 entries, 0 to 199
Data columns (total 5 columns):
#   Column      Non-Null Count  Dtype
---  -
0   Unnamed: 0  200 non-null   int64
1   TV          200 non-null   float64
```

```
2   Radio      200 non-null   float64
3   Newspaper  200 non-null   float64
4   Sales      200 non-null   float64
dtypes: float64(4), int64(1)
memory usage: 7.9 KB
```

```
[6]: df.drop("Unnamed: 0",axis=1,inplace=True)
```

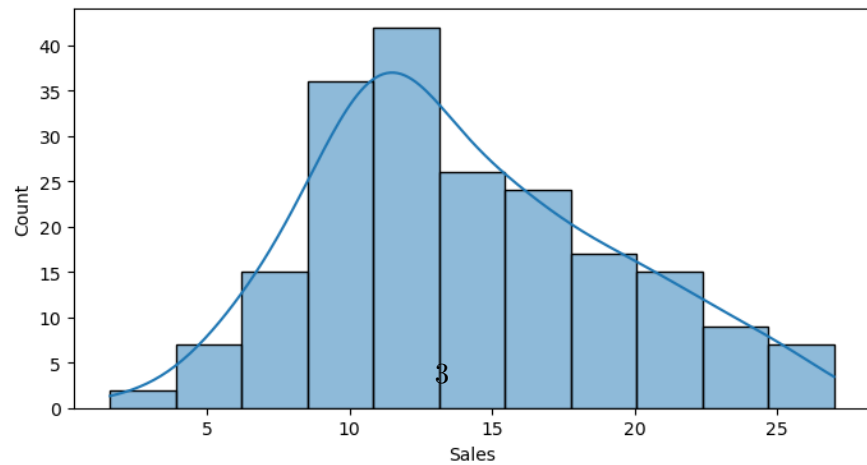
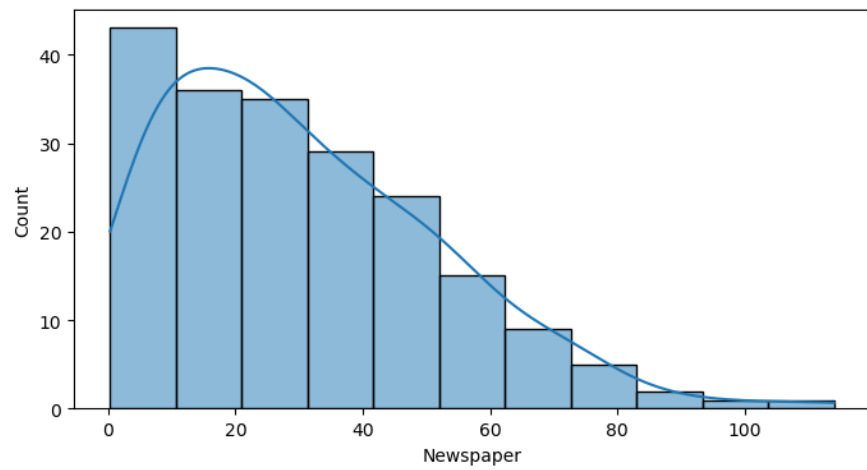
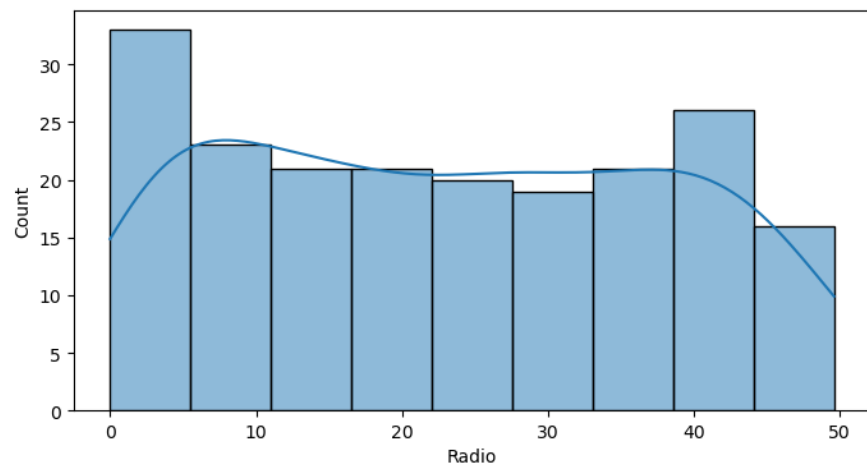
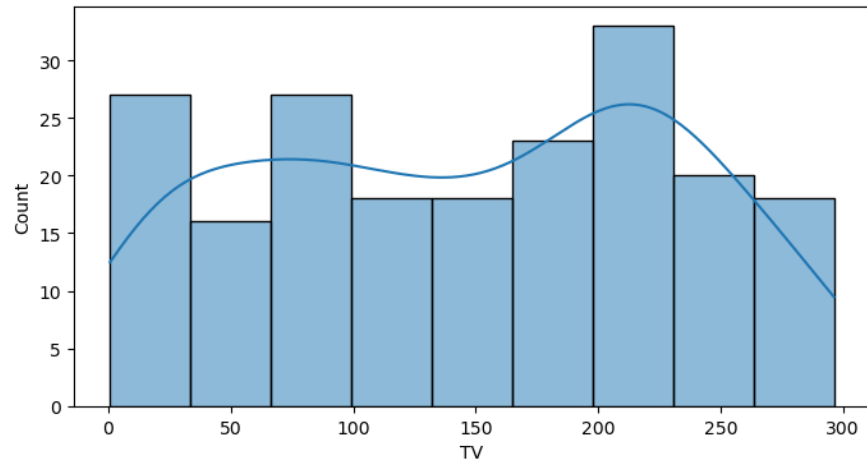
```
[7]: df.isnull().sum()
```

```
[7]: TV          0
Radio         0
Newspaper     0
Sales         0
dtype: int64
```

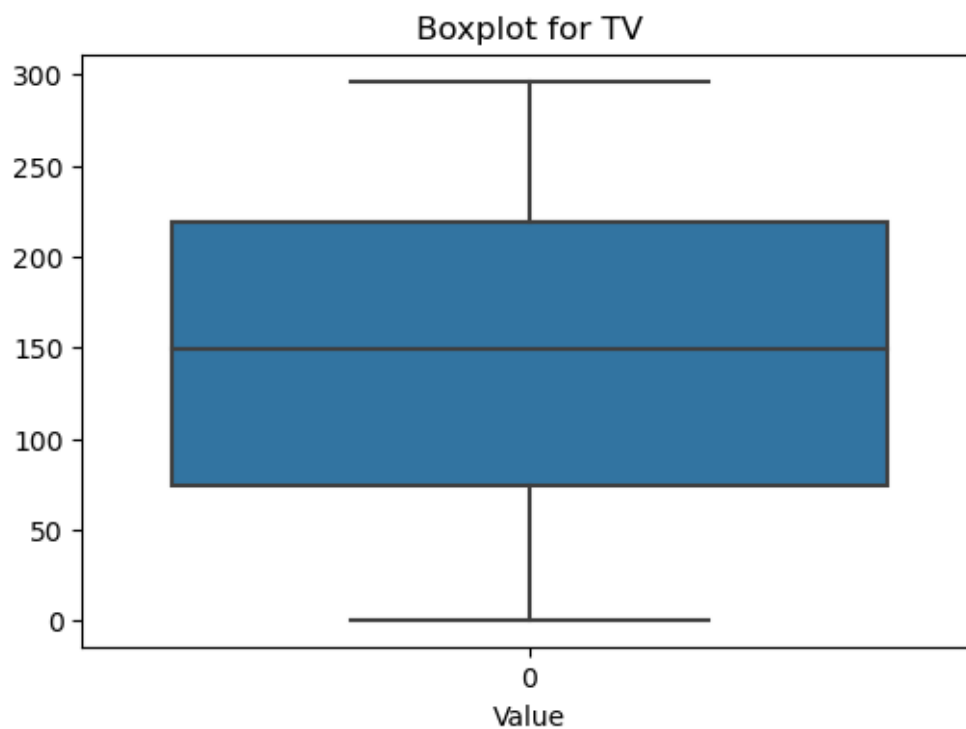
## 2 Performing EDA

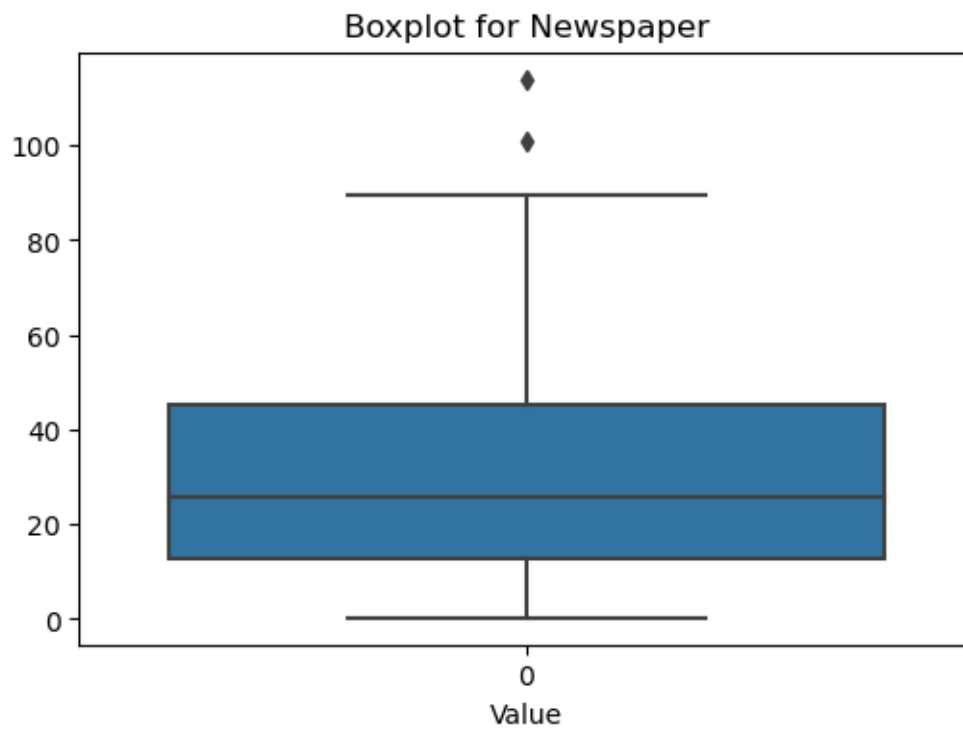
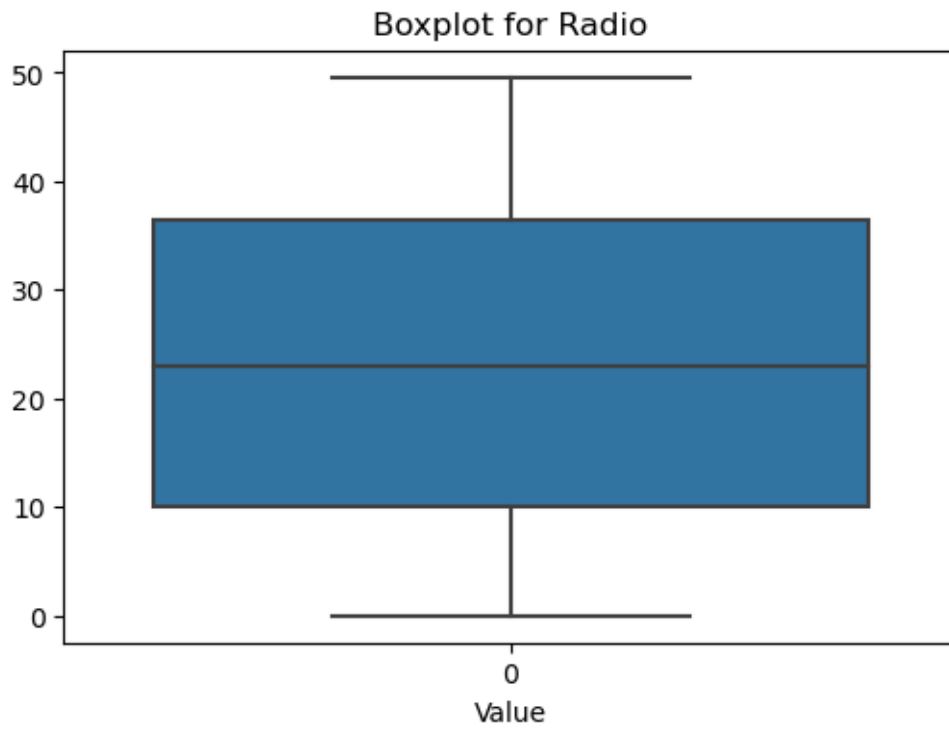
Univarite Analysis

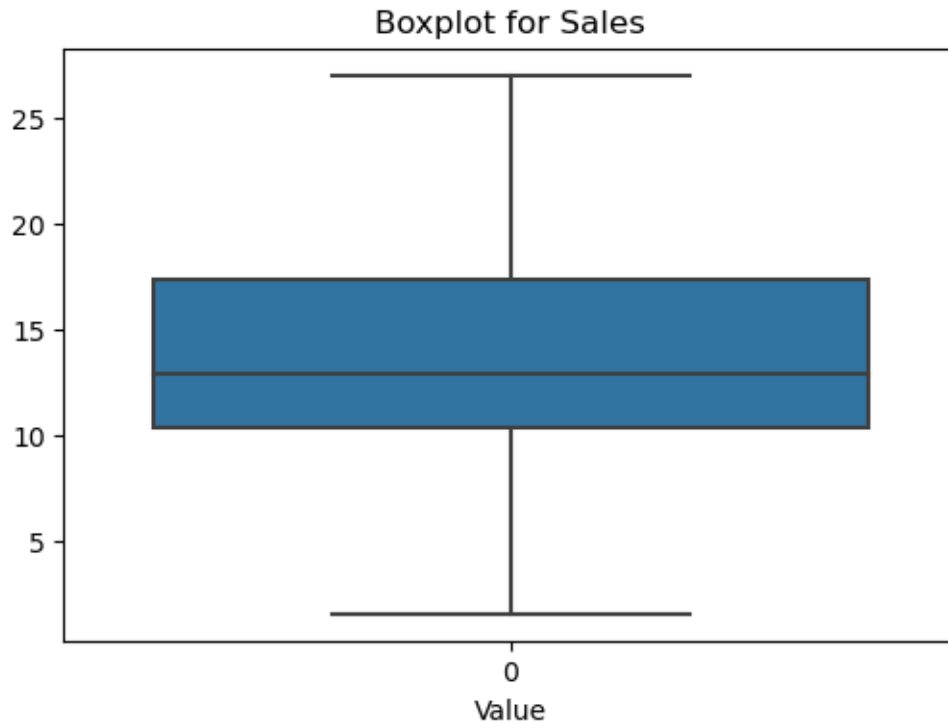
```
[8]: #creating subplot
columns=['TV','Radio','Newspaper','Sales']
plt.figure(figsize=(7,15)) # Adjust the figure size as needed
for i in columns:
    plt.subplot(4,1, columns.index(i) + 1)
    sns.histplot(df[i],kde=True)
plt.tight_layout()
```



```
[25]: #Creating Box-Plot
for column in df.columns:
    plt.figure(figsize=(6, 4)) # Adjust the figure size as needed
    sns.boxplot(data=df[column])
    plt.title(f'Boxplot for {column}')
    plt.xlabel('Value')
    plt.show()
```







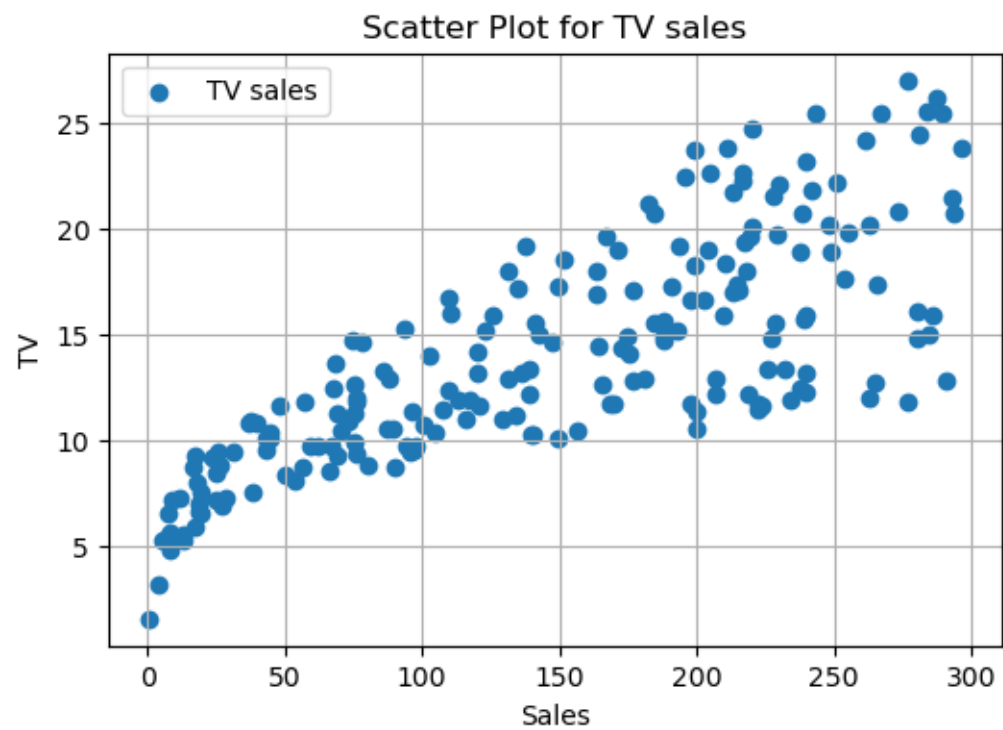
### 3 Bivarite Analysis

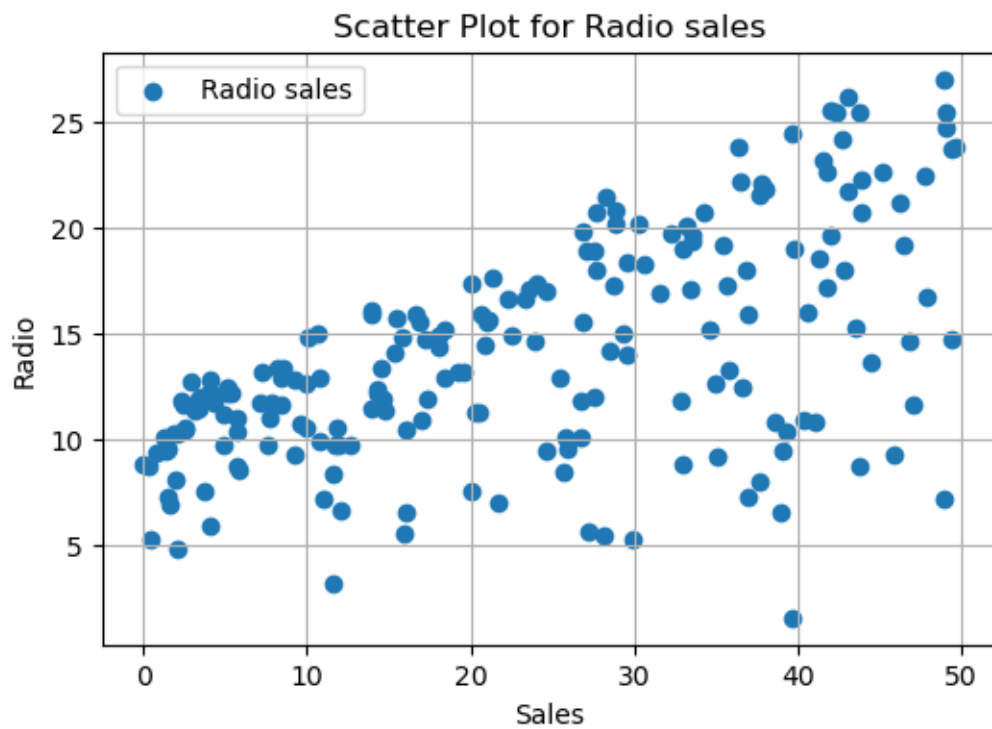
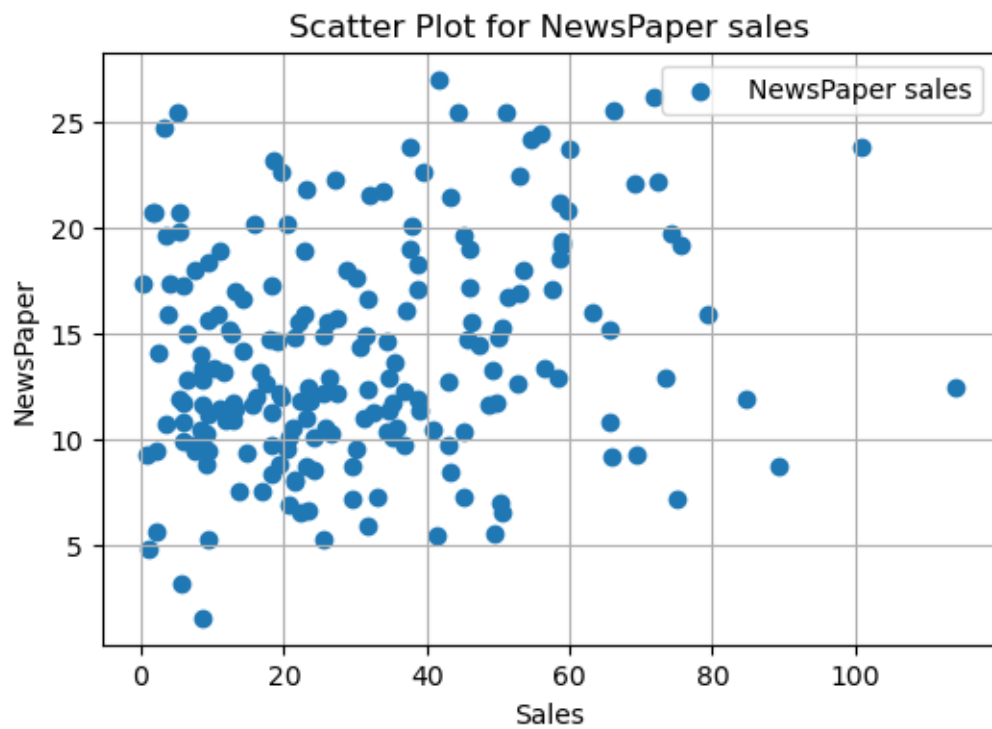
[12]: *#Creating Scatter Plot*

```
data_pairs = [
    (df["TV"], df["Sales"], 'TV sales', 'Sales', 'TV'),
    (df["Newspaper"], df["Sales"], 'NewsPaper sales', 'Sales', 'NewsPaper'),
    (df["Radio"], df["Sales"], 'Radio sales', 'Sales', 'Radio')
]

# Create scatter plots using a for loop
for i, (x, y, label, x_label, y_label) in enumerate(data_pairs):
    plt.figure(figsize=(6, 4)) # Adjust figure size as needed
    plt.scatter(x, y, label=label)
    plt.xlabel(x_label)
    plt.ylabel(y_label)
    plt.title(f'Scatter Plot for {label}')
    plt.legend()
    plt.grid(True)
    plt.show()
```

*# Create sample data (replace this with your own data)*



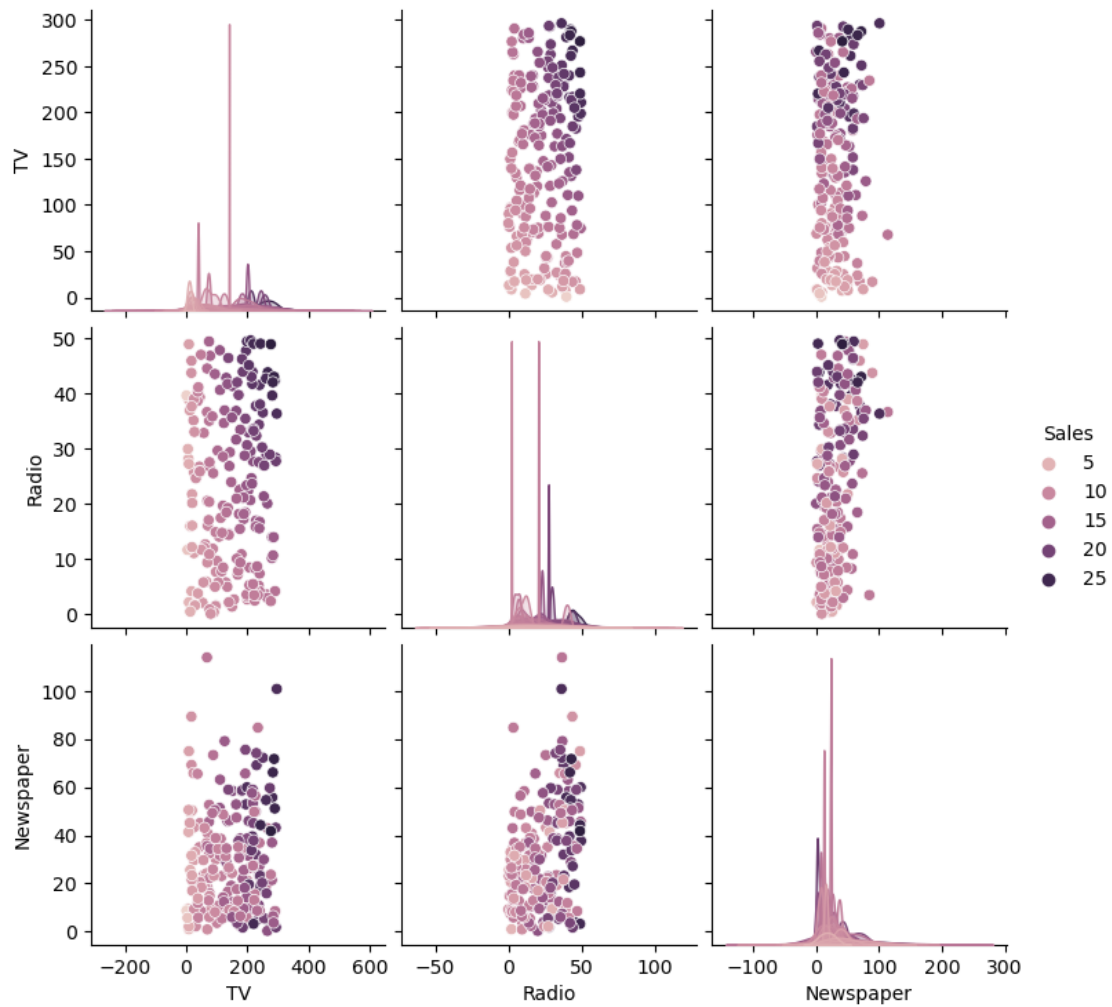




```
[14]: #Create PairPlot
```

```
sns.pairplot(df,hue='Sales')
```

```
[14]: <seaborn.axisgrid.PairGrid at 0x1e1dfb5b050>
```



```
[15]: # Calculate the correlation matrix
```

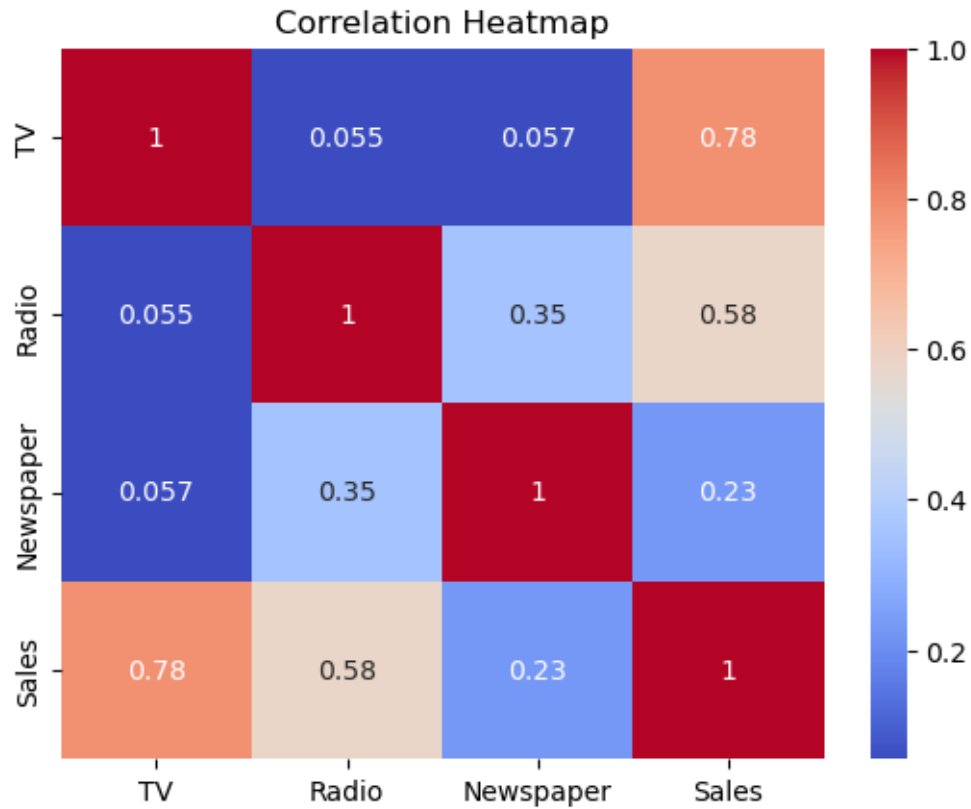
```
correlation_matrix = df.corr()
```

```
# Create a heatmap of the correlation matrix
```

```
sns.heatmap(correlation_matrix, annot=True, cmap='coolwarm')
```

```
plt.title('Correlation Heatmap')
```

```
plt.show()
```

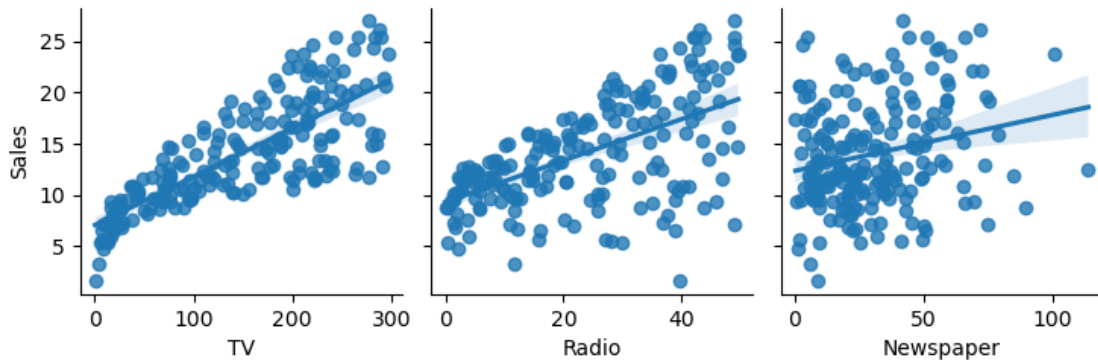


```
[16]: df.corr()*100
```

```
[16]:
```

	TV	Radio	Newspaper	Sales
TV	100.000000	5.480866	5.664787	78.222442
Radio	5.480866	100.000000	35.410375	57.622257
Newspaper	5.664787	35.410375	100.000000	22.829903
Sales	78.222442	57.622257	22.829903	100.000000

```
[17]: sns.pairplot(data = df,x_vars=["TV", "Radio", "Newspaper"], y_vars="Sales",
    ↪kind="reg",palette="pastel")
plt.show()
```



## 4 Splitting Data Into Train and Test

```
[18]: x=df[['TV', 'Radio', 'Newspaper']]
      y=df['Sales']
      x_train,x_test,y_train,y_test=train_test_split(x,y,train_size=0.7,
      ↪random_state=10)
```

## 5 Applying Model for Prediction

```
[19]: model=LinearRegression()
```

```
[20]: model.fit(x_train,y_train)

      y_predict=model.predict(x_test)
```

```
[21]: print(model.intercept_)
      print(model.coef_)
```

```
3.410641588611826
[ 0.04303172  0.19352212 -0.00386729]
```

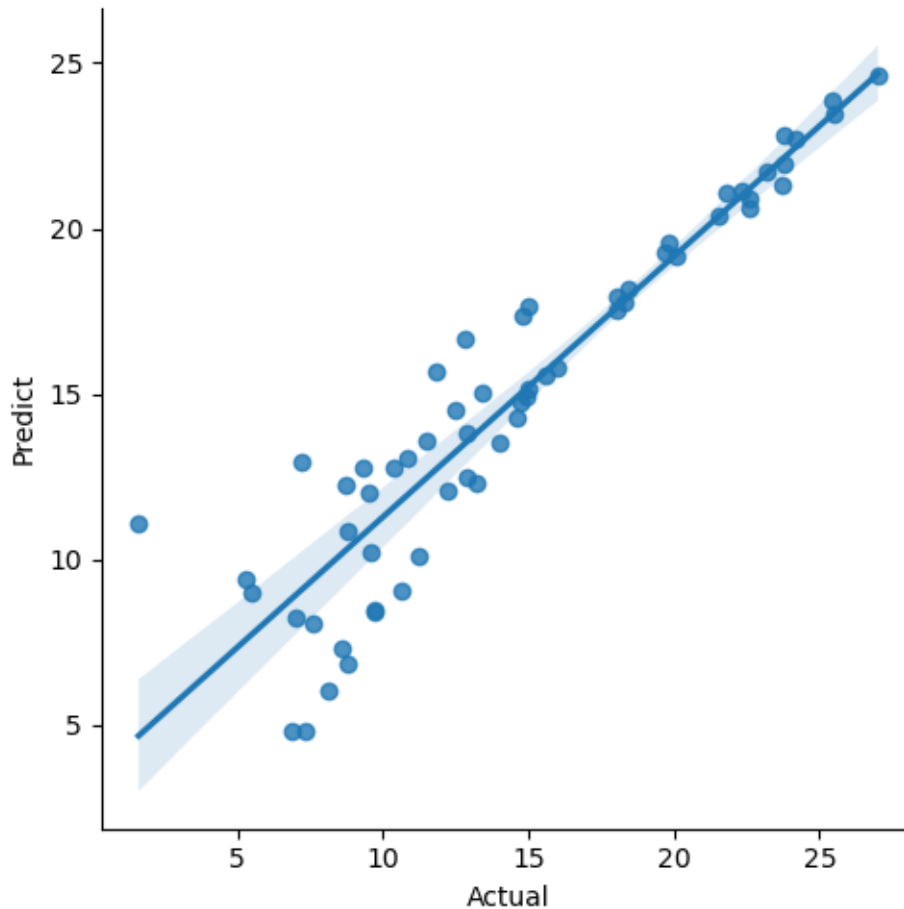
```
[22]: act_predict=pd.DataFrame({
      'Actual':y_test.values.flatten(),
      'Predict':y_predict.flatten()
      })
      act_predict.sample(20)
```

```
[22]:      Actual    Predict
51      12.9    13.837394
35      11.8    15.670965
0       18.4    18.150362
```

28	23.2	21.693574
55	8.1	6.017123
45	8.8	10.875483
24	14.8	17.339943
5	14.6	14.300527
47	13.4	15.028949
8	22.3	21.130349
48	5.5	9.002639
41	13.2	12.331227
2	18.0	17.962819
42	15.0	15.181345
31	21.5	20.393904
15	14.0	13.525769
7	23.7	21.297606
44	23.8	21.934629
17	5.3	9.392972
58	11.2	10.102094

```
[23]: sns.lmplot(data=act_predict,x='Actual',y="Predict")
```

```
[23]: <seaborn.axisgrid.FacetGrid at 0x1e1e07f0110>
```



```
[24]: print("Mean_absolute_error:",mean_absolute_error(y_test,y_predict))
      print("Mean_squared_error:",mean_squared_error(y_test,y_predict))
      print("Squire_Mean_absolute_error:",np.
            ↪sqrt(mean_absolute_error(y_test,y_predict)))
      print("r2_score:",r2_score(y_test,y_predict)*100)
```

```
Mean_absolute_error: 1.7091210770106202
Mean_squared_error: 5.374051588095294
Squire_Mean_absolute_error: 1.3073335752632609
r2_score: 85.67790678044409
```

## 6 ThankYou

```
[ ]:
```