5 180.8 10.8 58.4 12.9 df.isnull().sum() Unnamed: 0 0 Out[9]: TV 0 Radio 0 Newspaper 0 Sales dtype: int64 In [10]: df.info() <class 'pandas.core.frame.DataFrame'> RangeIndex: 200 entries, 0 to 199 Data columns (total 5 columns): Non-Null Count Dtype Column -----Unnamed: 0 200 non-null 0 int64 200 non-null 1 TV float64 2 Radio 200 non-null float64 200 non-null float64 Newspaper float64 Sales 200 non-null dtypes: float64(4), int64(1)memory usage: 7.9 KB In [11]: print('Duplicate_values=',df.duplicated().sum()) Duplicate_values= 0 df.head() In [12]: Unnamed: 0 TV Radio Newspaper Sales Out[12]: 1 230.1 37.8 69.2 22.1 39.3 2 44.5 45.1 10.4 3 17.2 69.3 45.9 9.3 4 151.5 5 180.8 10.8 58.4 12.9 1. Visualizing targate variable 'sales' In [14]: plt.figure(figsize=(20,8)) plt.subplot(1,2,1) plt.title('Sales Distribution') sns.histplot(data=df, x='Sales', kde=True, color='red') plt.subplot(1,2,2) plt.title('spread of sale') sns.boxenplot(data=df, x='Sales', color='purple') plt.show() Sales Distribution spread of sale 40 35 30 25 20 15 10 10 15 20 15 20 25 Sales Sales insights most of the value 8-29 formed bell shape curve In [15]: #2.Visualize Tv vs Sales features plt.figure(figsize=(15,8)) In [18]: plt.title("TV vs SALES") sns.scatterplot(x=df['TV'], y=df['Sales'], color='red') plt.show() TV vs SALES 25 20

In [4]: **import** pandas **as** pd

import pylab

df.head()

0

Unnamed: 0

In [8]:

Out[8]:

import warnings

import numpy as np import seaborn as sns

from scipy import stats

import matplotlib.pyplot as plt

warnings.filterwarnings('ignore')

from sklearn.metrics import r2_score

df = pd.read_csv('sales.csv')

1 230.1

2 44.5

3 17.2

4 151.5

pd.set_option("display.max_columns", None) pd.set_option("display.max_rows", None)

from sklearn.linear_model import LinearRegression from sklearn.tree import DecisionTreeRegressor

from sklearn.model_selection import train_test_split

TV Radio Newspaper Sales

69.2

45.1

69.3

58.5

22.1

10.4

9.3

18.5

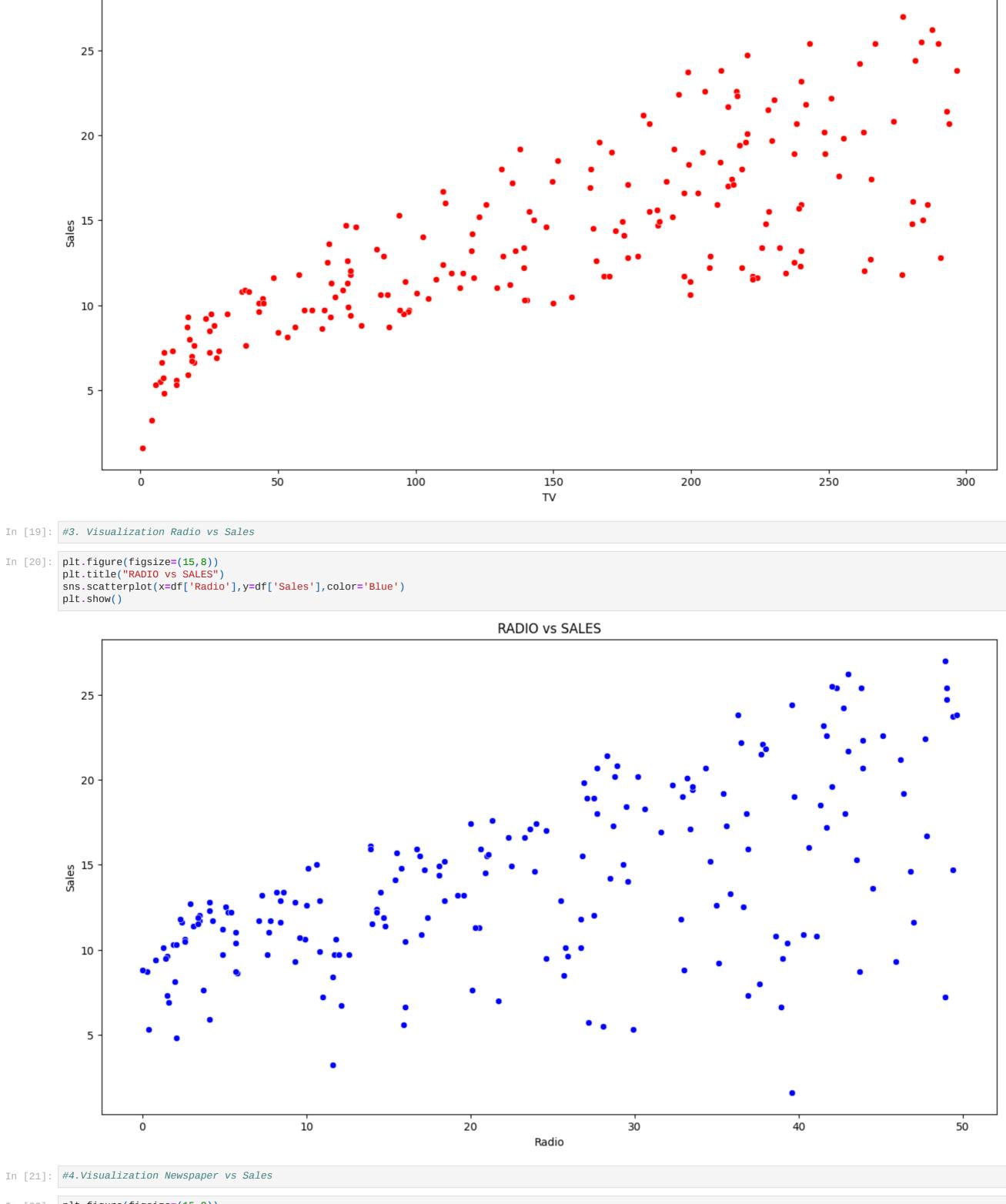
37.8

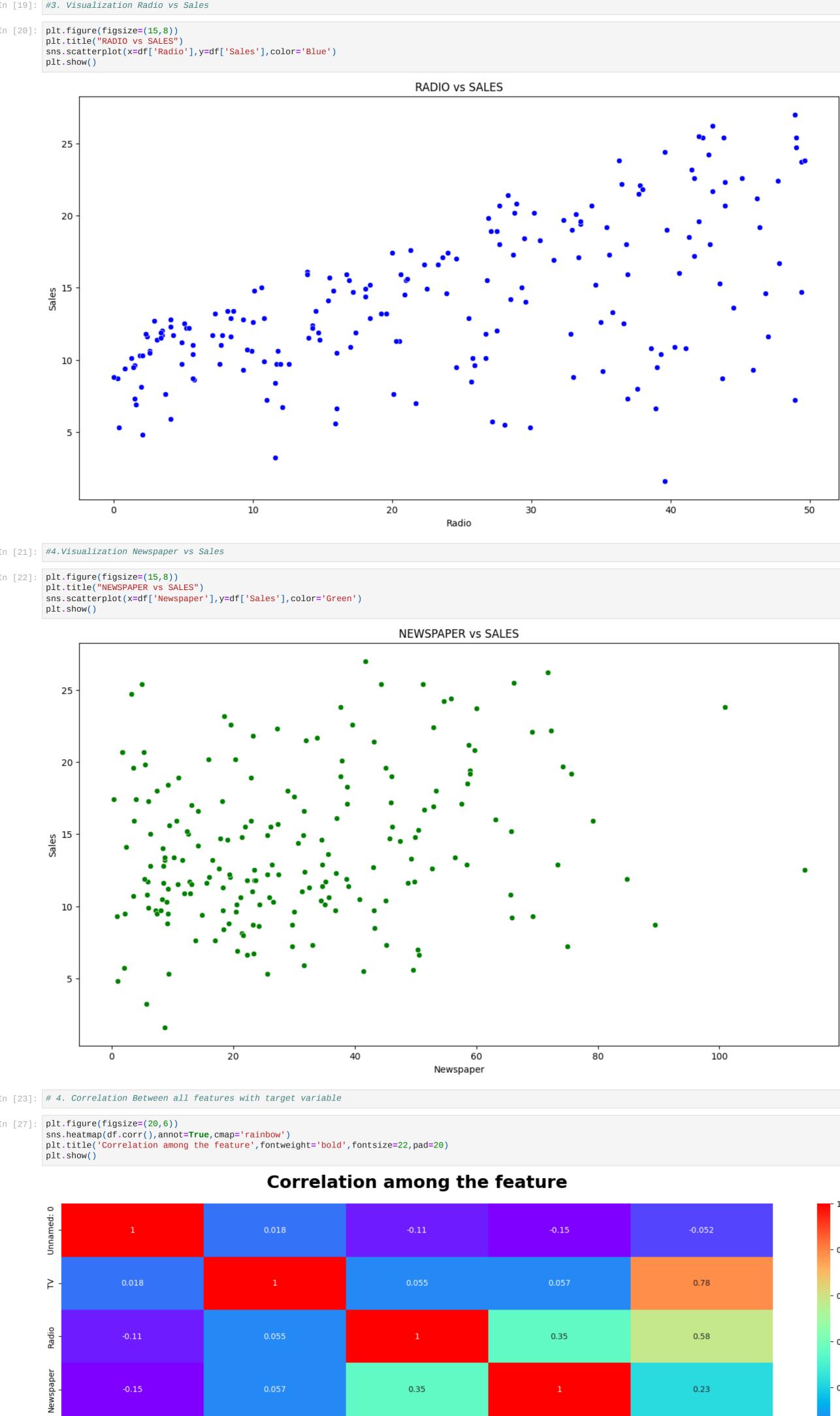
39.3

45.9

41.3

from sklearn.ensemble import RandomForestRegressor,AdaBoostRegressor





In [23]: # 4. Correlation Between all features with target variable 1.0 - 0.8 - 0.6 - 0.4 0.2 - 0.0 -0.052 0.78 0.58 0.23

Radio

Newspaper

Sales

TV

x_train, x_test, y_train, y_test = train_test_split(x, y, test_size=0.2, random_state=42)

Unnamed: 0

1. Selecting label and target for model trainning

Data Preprocessing

y = df[['Sales']]

(160, 3) (160, 1) (40, 3) (40, 1)

In [41]: def model_pred(model):

Model Creation

In [42]: model_pred(LinearRegression())

In [43]: model_pred(DecisionTreeRegressor())

In [44]: model_pred(RandomForestRegressor())

In [45]: model_pred(AdaBoostRegressor())

#Insights

In []:

In []:

In [38]:

In [34]: x = df.drop(columns=['Sales', 'Newspaper'])

print(x_train.shape,y_train.shape)

model.fit(x_train,y_train)

x_train_pred = model.predict(x_train) x_test_pred = model.predict(x_test) a=r2_score(y_train, x_train_pred)*100 b=r2_score(y_test,x_test_pred)*100

print(f"r2_score of {model} model on training data:",a) print(f"r2_score of {model} model on testing data:",b)

r2_score of DecisionTreeRegressor() model on trainning data: 100.0

Highest performance of random forestmodel around 98%

r2_score of LinearRegression() model on trainning data: 89.56328908500461 r2_score of LinearRegression() model on testing data: 90.00623280485392

r2_score of DecisionTreeRegressor() model on testing data: 95.95974396926839

r2_score of RandomForestRegressor() model on trainning data: 99.56601744662393 r2_score of RandomForestRegressor() model on testing data: 98.26262693867045

r2_score of AdaBoostRegressor() model on trainning data: 97.58581309615802 r2_score of AdaBoostRegressor() model on testing data: 96.31275984101697

print(x_test.shape, y_test.shape)