Linear_Regression_feb14

August 19, 2022

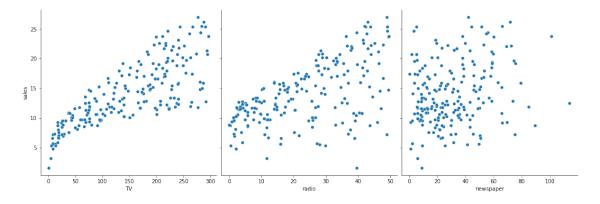
1 Linear Regression

```
[1]: import pandas as pd
    import numpy as np
    import matplotlib.pyplot as plt
    import seaborn as sns
[2]: data=pd.read_csv("Advertising.csv")
    data.head(5)
[2]:
          TV
              radio newspaper
                                sales
       230.1
               37.8
                          69.2
                                  22.1
    1
        44.5
               39.3
                          45.1
                                 10.4
        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
[3]: data.shape
[3]: (200, 4)
[4]: data.info()
    <class 'pandas.core.frame.DataFrame'>
    RangeIndex: 200 entries, 0 to 199
    Data columns (total 4 columns):
                    Non-Null Count Dtype
         Column
         _____
                    _____
     0
         TV
                    200 non-null
                                    float64
     1
                    200 non-null
                                    float64
         radio
     2
         newspaper 200 non-null
                                    float64
     3
                    200 non-null
                                    float64
         sales
    dtypes: float64(4)
    memory usage: 6.4 KB
[5]: #sns.pairplot(data)# all analysis
```

```
sns. 
 \hookrightarrow pairplot(x_vars=["TV", "radio", "newspaper"], y_vars=["sales"], data=data, size=5)
```

/usr/local/lib/python3.7/site-packages/seaborn/axisgrid.py:2076: UserWarning: The `size` parameter has been renamed to `height`; please update your code. warnings.warn(msg, UserWarning)

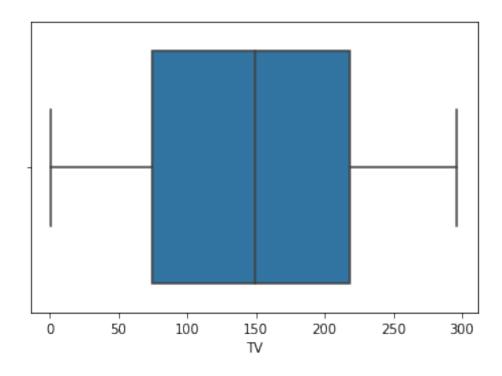
[5]: <seaborn.axisgrid.PairGrid at 0x7f03641f6510>



```
[6]: #box plot for all features
for i in data.columns:
    sns.boxplot(data[i])
    plt.show()
```

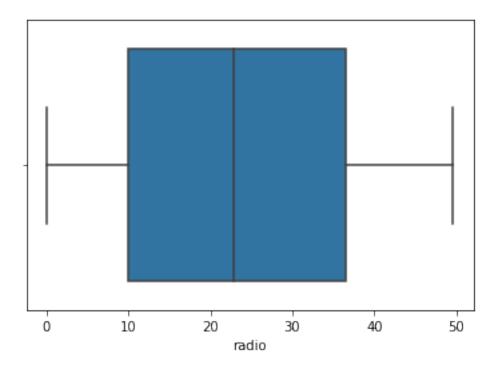
/usr/local/lib/python3.7/site-packages/seaborn/_decorators.py:43: FutureWarning: Pass the following variable as a keyword arg: x. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation.

FutureWarning



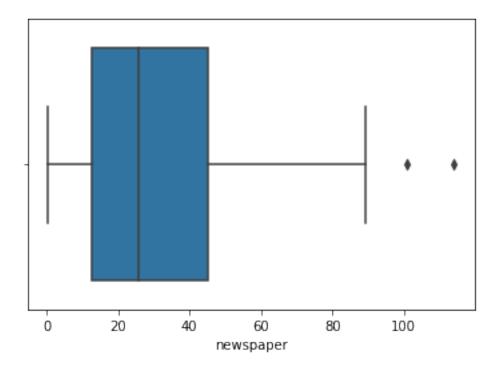
/usr/local/lib/python3.7/site-packages/seaborn/_decorators.py:43: FutureWarning: Pass the following variable as a keyword arg: x. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation.

FutureWarning



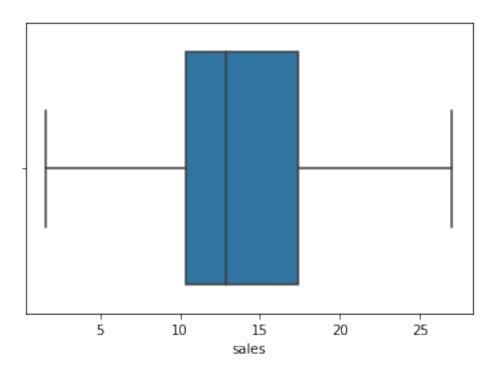
/usr/local/lib/python3.7/site-packages/seaborn/_decorators.py:43: FutureWarning: Pass the following variable as a keyword arg: x. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation.

FutureWarning



/usr/local/lib/python3.7/site-packages/seaborn/_decorators.py:43: FutureWarning: Pass the following variable as a keyword arg: x. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation.

FutureWarning



```
[7]: # outlier Treatement
    from scipy import stats
    iqr=stats.iqr(data["newspaper"])
    q1=stats.scoreatpercentile(data["newspaper"],25)
    q3=stats.scoreatpercentile(data["newspaper"],75)
    lowerbound=q1 -(1.5*iqr)
    upperbound=q3+(1.5*iqr)
    print(lowerbound)
    print(upperbound)
    outliers=data["newspaper"] [((data["newspaper"] < lowerbound) | __
     print('outliers are',outliers)
    -35.775000000000006
    93.625
    outliers are 16
                       114.0
           100.9
    101
    Name: newspaper, dtype: float64
[8]: median = np.median(data["newspaper"])# Replace with median
    for i in outliers:
        data["newspaper"] = np.where(data["newspaper"] == i, median, u

data["newspaper"])
```

```
[9]: # Simple Linear Regression
     # consider one features
     # Tv as feature and sale as a label
[10]: # split input and output features
     X=data.loc[:,["TV"]]
     y=data.loc[:,["sales"]]
[11]: X.head()
[11]:
           TV
     0 230.1
     1 44.5
     2 17.2
     3 151.5
     4 180.8
     2 split the data for training and Testing
[31]: # Scikit learn----- it contain all machine learning realted libraries
     from sklearn.model_selection import train_test_split
     xtrain,xtest,ytrain,ytest=train_test_split(X,y,test_size=0.3,random_state=10)
[32]: print(xtrain.shape)
     print(xtest.shape)
     print(ytrain.shape)
     print(ytest.shape)
     (140, 1)
     (60, 1)
     (140, 1)
     (60, 1)
[33]: # Build the Model
      #process to import the linear regression model
     from sklearn.linear_model import LinearRegression
     # instantiate the obect
     lr=LinearRegression()
```

```
[34]: # fit the data or train the data
      lr.fit(xtrain,ytrain)
[34]: LinearRegression()
[35]: \# y=mx+c
      lr.intercept_
[35]: array([7.30826074])
[36]: lr.coef_
[36]: array([[0.04474675]])
[37]: # Sales = 0.048 *[TV] + 6.917
[38]: #training performance
      ypredtrain=lr.predict(xtrain)
 []: #ypredtrain
[39]: ypredtrain.shape
[39]: (140, 1)
[40]: ypredtrain
[40]: array([[ 8.17634775],
             [17.41207755],
             [14.6154055],
             [16.56188925],
             [14.77201913],
             [17.93113989],
             [17.8013743],
             [10.80745682],
             [13.54148342],
             [8.71778346],
             [13.63097693],
             [ 9.23684579],
             [11.52340487],
             [11.14305747],
             [13.35802174],
             [16.15021912],
             [10.36893864],
             [8.10922762],
             [10.72691267],
```

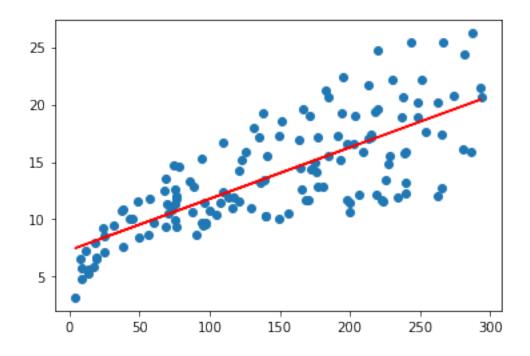
```
[15.22843601],
```

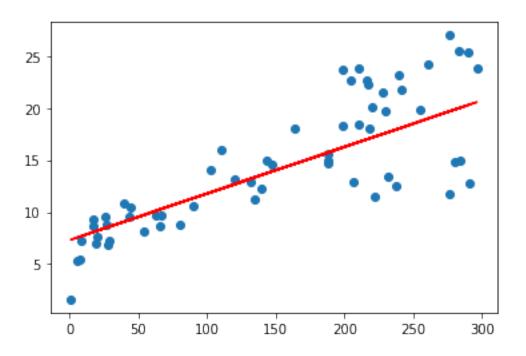
- [10.72691267],
- [18.03853209],
- [11.65764513],
- [9.54559839],
- [17.47472301],
- [19.07218209],
- [15.47901783],
- [13.47883797],
- [10.47000707]
- [19.89999702],
- [14.00684966],
- [11.80083474],
- [16.44107301],
- [10.67769124],
- [16.86169249],
- [18.66498664],
- [10.4673815],
- [14.08739381],
- [17.52394444],
- [18.19067105],
- [20.10583208],
- [10.65084319],
- [8.14502502],
- [13.55043278],
- [15.85936523],
- [15.14341718],
- [7.68413347],
- [15.58193536],
- [14.71832303],
- [10.68664059],
- [20.4459074],
- [9.97516722],
- [19.17509962],
- [12.49888408],
- [15.02707562],
- [14.92415809],
- [11.9887711],
- [16.94671132],
- [8.37323346],
- [15.17026523],
- [16.9153886],
- [7.89891788],
- [8.42692957],
- [15.95333341],
- [18.05195612],
- [9.88119904],
- [17.16597041],

```
[ 9.3084406 ],
```

- [7.49172243],
- [9.00863735],
- [17.96693729],
- [11.590525],
- [7.89444321],
- [16.24866198],
- [8.95941593],
- [11.61289837],
- [13.58623018],
- [15.39847367],
- [17.14359704],
- [8.43140424],
- [16.15021912],
- [18.42335417],
- [18.44125287],
- [9.46952891],
- [14.6691016],
- [14.31560225],
- [12.22145421],
- [12.2210121]
- [12.93292759],
- [8.07790489],
- [15.22843601],
- [19.55544702],
- [16.85721782],
- [9.82302826],
- [12.11406201],
- [12.72261785],
- [19.86867429],
- [14.84361394],
- [18.53522105],
- [10.40473604],
- [18.05195612],
- [19.2511691],
- [13.40276849],
- [12.36016915],
- [10.39578669],
- [10.59267241],
- [12.22145421],
- [7.83179775],
- [20.41458467],
- [10.72243799],
- [17.04962885],
- [15.97570678],
- [7.65728542],
- [14.97337952],
- [11.67106915],

```
[12.70024447],
             [17.08542626],
             [20.17742688],
             [19.19299832],
             [18.01615872],
             [17.3315334],
             [11.35336721],
             [11.25939903],
             [19.06323274],
             [15.58193536],
             [17.25993859],
             [10.34209059],
             [10.66874189],
             [12.55258019],
             [16.36947821],
             [17.60448859],
             [13.09849057],
             [7.69308282],
             [ 9.0175867 ],
             [14.01132433],
             [11.50998084],
             [12.81658603],
             [16.68718016],
             [13.17456005],
             [16.05177626],
             [11.2101776],
             [16.24866198]])
[41]: from sklearn.metrics import mean_squared_error
      print("The Rms value is training data")
      print("----")
      print(np.sqrt(mean_squared_error(ytrain,ypredtrain)))
     The Rms value is training data
     3.0681028951223004
[42]: plt.scatter(xtrain,ytrain)
      plt.plot(xtrain,ypredtrain,color="red")
[42]: [<matplotlib.lines.Line2D at 0x7f031403e310>]
```





```
[47]: from sklearn.metrics import r2_score
#r2score close to 1 is a good value
print(r2_score(ytest,ypredtest))

0.6447879078216139
[]: # End of Simple Linear Regression
```

3 Multiple Linear Regression

```
[48]: data.head(2)
[48]:
               radio newspaper
                                  sales
      0 230.1
                 37.8
                            69.2
                                   22.1
         44.5
                 39.3
                            45.1
                                   10.4
[49]: # splitted i/p and o/p
      xnew=data.drop(["sales"],axis=1) #drop the column
     ynew=data["sales"]
 []: xnew.head(3)
```

```
[50]: from sklearn.model_selection import train_test_split
      xtrain, xtest, ytrain, ytest=train_test_split(xnew, ynew, test_size=0.
       \rightarrow3, random_state=10)
[51]: print(xtrain.shape)
      print(xtest.shape)
     (140, 3)
     (60, 3)
[52]: # build a mlr model
      from sklearn.linear_model import LinearRegression
      mlr=LinearRegression()
      mlr.fit(xtrain,ytrain)
[52]: LinearRegression()
[53]: print("intercept value",mlr.intercept_)
      print("coefficent value",mlr.coef_)
     intercept value 3.392597098985595
     coefficent value [ 0.04304513  0.19298948 -0.00301967]
 []: # sales= 0.043 [TV]+[0.193][radio]-[0.0038]newspaper+ 3.41
[54]: # check the train performance
      ypredtraining_mlr=mlr.predict(xtrain)
[55]: df=pd.DataFrame({"Actual":ytrain,"predicted":ypredtraining_mlr})
      df
[55]:
           Actual predicted
              6.6
                   7.248166
      119
      110
             13.4 14.524089
      95
             16.9 16.360594
      167
             12.2 13.239294
             19.6 18.667212
      193
             15.9 16.358129
      113
      64
             18.0 17.208495
      15
             22.4 20.849473
      125
             10.6 9.345199
             10.6 12.430770
```

```
[140 rows x 2 columns]
[56]: from sklearn.metrics import mean_squared_error
     print("The Rms value is training data")
     print("----")
     print(np.sqrt(mean_squared_error(ytrain,ypredtraining_mlr)))
     The Rms value is training data
     1.3332818533291317
[57]: from sklearn.metrics import r2_score
      #r2score close to 1 is a good value
     print(r2_score(ytrain,ypredtraining_mlr))
     0.920909492761702
[58]: # Test performace
     ypredtest_mlr=mlr.predict(xtest)
[59]: pd.DataFrame({"Actual":ytest,"predicted":ypredtest_mlr}).head()
[59]:
          Actual predicted
     59
            18.4 18.127313
     5
             7.2 12.977800
            18.0 17.978211
     20
     198
            25.5 23.505852
     52
            22.6 20.635645
[60]: from sklearn.metrics import mean_squared_error
     print("The Rms value is test data")
     print("----")
     print(np.sqrt(mean_squared_error(ytest,ypredtest_mlr)))
     The Rms value is test data
     2.314415304470104
[61]: print(r2_score(ytest,ypredtest_mlr))
     0.8572463405728885
 []: # End of Linear regression
```

[]: