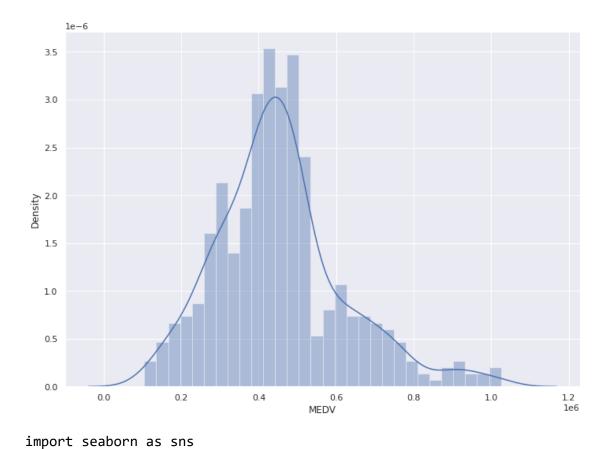
Assignment No 4

```
import pandas as pd
import numpy as np
import io
import seaborn as sns
import matplotlib.pyplot as plt
%matplotlib inline
from google.colab import files
uploaded=files.upload()
<IPython.core.display.HTML object>
Saving housing.csv to housing.csv
df=pd.read_csv(io.BytesIO(uploaded['housing.csv']))
print(df)
        RM LSTAT
                  PTRATIO
                             MEDV
0
    6.575
            4.98
                     15.3
                           504000
            9.14
                     17.8 453600
1
    6.421
2
    7.185
            4.03
                     17.8 728700
3
    6.998
            2.94
                     18.7 701400
4
    7.147
            5.33
                     18.7 760200
             . . .
                     21.0 470400
484 6.593
            9.67
485 6.120
            9.08
                     21.0 432600
486 6.976
            5.64
                     21.0 501900
487 6.794
            6.48
                      21.0 462000
488
    6.030
            7.88
                     21.0 249900
[489 rows x 4 columns]
df.info()
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 489 entries, 0 to 488
Data columns (total 4 columns):
    Column
             Non-Null Count Dtype
- - -
    -----
              -----
                             ____
0
    RM
             489 non-null
                             float64
             489 non-null
1
                             float64
    LSTAT
2
    PTRATIO 489 non-null
                             float64
    MEDV
             489 non-null
                             int64
dtypes: float64(3), int64(1)
memory usage: 15.4 KB
```

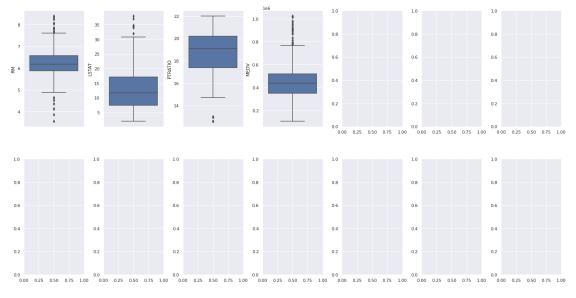
```
print(df.isnull().sum())
RM
           0
LSTAT
           0
PTRATIO
           0
MEDV
dtype: int64
print(df.isnull().sum().sum())
0
print(np.shape(df))
(489, 4)
print(df.describe())
               RM
                        LSTAT
                                  PTRATIO
                                                   MEDV
count 489.000000
                 489.000000 489.000000 4.890000e+02
                                18.516564 4.543429e+05
mean
        6.240288
                   12.939632
std
        0.643650
                    7.081990
                                 2.111268 1.653403e+05
                                12.600000 1.050000e+05
min
         3.561000
                     1.980000
25%
         5.880000
                    7.370000
                                17.400000 3.507000e+05
50%
        6.185000
                   11.690000
                                19.100000 4.389000e+05
75%
         6.575000
                   17.120000
                                20.200000 5.187000e+05
        8.398000
                    37.970000
                                22.000000 1.024800e+06
max
#set the size of the figure
sns.set(rc={'figure.figsize':(11.7,8.27)})
#Histogram for distribution of the target values
sns.distplot(df['MEDV'], bins =30)
plt.show()
/usr/local/lib/python3.7/dist-packages/seaborn/distributions.py:2619:
FutureWarning: `distplot` is a deprecated function and will be removed in a
future version. Please adapt your code to use either `displot` (a figure-
level function with similar flexibility) or `histplot` (an axes-level
function for histograms).
 warnings.warn(msg, FutureWarning)
```



```
import matplotlib.pyplot as plt
from scipy import stats

fig, axs = plt.subplots(ncols=7, nrows=2, figsize=(20, 10))
index = 0
axs = axs.flatten()
for k,v in df.items():
    sns.boxplot(y=k, data=df, ax=axs[index])
    index += 1
```

plt.tight_layout(pad=0.4, w_pad=0.5, h_pad=5.0)



```
fig, axs = plt.subplots(ncols=7, nrows=2, figsize=(20, 10))
index = 0
axs = axs.flatten()
for k,v in df.items():
    sns.distplot(v, ax=axs[index])
    index += 1
plt.tight layout(pad=0.4, w pad=0.5, h pad=5.0)
```

/usr/local/lib/python3.7/dist-packages/seaborn/distributions.py:2619: FutureWarning: `distplot` is a deprecated function and will be removed in a future version. Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

warnings.warn(msg, FutureWarning)

/usr/local/lib/python3.7/dist-packages/seaborn/distributions.py:2619: FutureWarning: `distplot` is a deprecated function and will be removed in a future version. Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

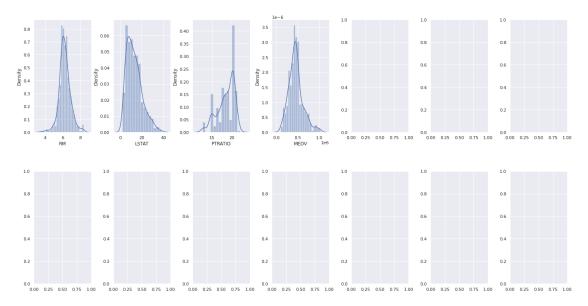
warnings.warn(msg, FutureWarning)

/usr/local/lib/python3.7/dist-packages/seaborn/distributions.py:2619: FutureWarning: `distplot` is a deprecated function and will be removed in a future version. Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

warnings.warn(msg, FutureWarning)

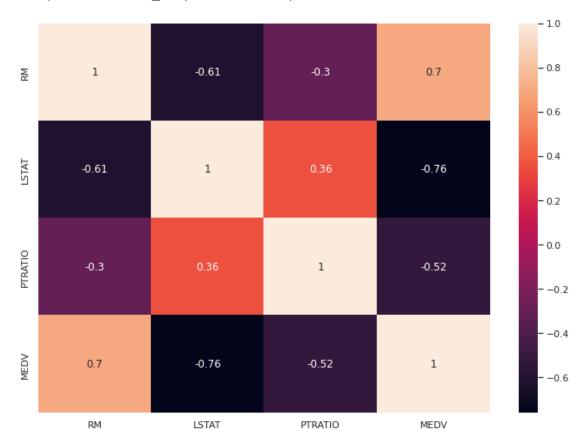
/usr/local/lib/python3.7/dist-packages/seaborn/distributions.py:2619: FutureWarning: `distplot` is a deprecated function and will be removed in a future version. Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

warnings.warn(msg, FutureWarning)



correlation_matrix = df.corr().round(2)
sns.heatmap(data=correlation_matrix, annot = True)

<matplotlib.axes._subplots.AxesSubplot at 0x7f9d65c7b450>



plt.figure(figsize=(20,5))

features = ['LSTAT','RM']

```
target = df['MEDV']
for i, col in enumerate(features):
  plt.subplot(1, len(features), i+1)
  x=df[col]
  y = target
  plt.scatter(x, y,marker='o')
  plt.title(col)
  plt.xlabel(col)
  plt.ylabel('MEDV')
#Prepare data for training
X = pd.DataFrame(np.c_[df['LSTAT'],df['RM']], columns = ['LSTAT','RM'])
Y = df['MEDV']
#split data into training and testing sets
from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test = train_test_split(X, y,test_size =
0.3, random_state = 42)
print(X_train.shape)
print(X_test.shape)
print(y_train.shape)
print(y_train.shape)
(342, 2)
(147, 2)
(342,)
(342,)
from sklearn import linear_model
# Train the model using sklearn linear regression
from sklearn.linear_model import LinearRegression
from sklearn.metrics import mean_squared_error, r2_score
lin_model = LinearRegression()
lin_model.fit(X_train,y_train)
LinearRegression()
# Model evaluation for training set
y_train_predict = lin_model.predict(X_train)
```

```
rmse = (np.sqrt(mean_squared_error(y_train, y_train_predict)))
r2 = r2_score(y_train,y_train_predict)
print("Model performance for training set")
print('RMSE is {}'.format(rmse))
print('R2 score is {}'.format (r2))
print("\n")
# Model evaluation for testing set
y_test_predict = lin_model.predict(X_test)
rmse = (np.sqrt(mean_squared_error(y_test, y_test_predict)))
r2 = r2_score(y_test,y_test_predict)
print("Model performance for testing set")
print('RMSE is {}'.format(rmse))
print('R2 score is {}'.format (r2))
print("\n")
Model performance for training set
RMSE is 97613.15525868809
R2 score is 0.6701982599508145
Model performance for testing set
RMSE is 92082.48814292479
R2 score is 0.628271585004582
# Plotting y test vs y pred
plt.scatter(y_test, y_test_predict)
plt.show()
```

