ASSIGNMENT -4

import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt

from google.colab import files
upload=files.upload()
df = pd.read_csv('abalone.csv')

abalone.csv

Choose Files

• **abalone.csv**(text/csv) - 191962 bytes, last modified: 10/29/2022 - 100% done Saving abalone.csv to abalone (1).csv

df.describe()

		v₩sēshā	รพยรัชอุต	wahaht	Height	Diameter	Length	
	4177	4177.000000	4177.000000	4177.000000	4177.000000	4177.000000	4177.000000	count
ı	(0.180594	0.359367	0.828742	0.139516	0.407881	0.523992	mean
ı	(0.109614	0.221963	0.490389	0.041827	0.099240	0.120093	std
ı	(0.000500	0.001000	0.002000	0.000000	0.055000	0.075000	min
ı	(0.093500	0.186000	0.441500	0.115000	0.350000	0.450000	25%
ı	(0.171000	0.336000	0.799500	0.140000	0.425000	0.545000	50%
ı	(0.253000	0.502000	1.153000	0.165000	0.480000	0.615000	75%
		0.760000	1.488000	2.825500	1.130000	0.650000	0.815000	max

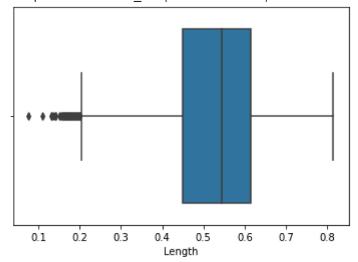
df.head()

	Sex	Length	Diameter	Height	Whole weight	Shucked weight	Viscera weight	Shell w
0	М	0.455	0.365	0.095	0.5140	0.2245	0.1010	
1	М	0.350	0.265	0.090	0.2255	0.0995	0.0485	
2	F	0.530	0.420	0.135	0.6770	0.2565	0.1415	
3	М	0.440	0.365	0.125	0.5160	0.2155	0.1140	
4	I	0.330	0.255	0.080	0.2050	0.0895	0.0395	

sns.boxplot(df.Length)

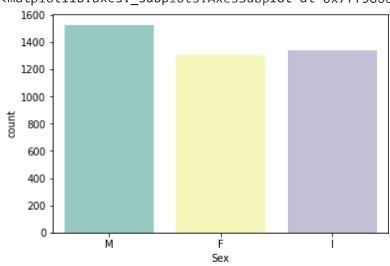
/usr/local/lib/python3.7/dist-packages/seaborn/_decorators.py:43: FutureWarning: Pass t FutureWarning

<matplotlib.axes._subplots.AxesSubplot at 0x7ff5877a5350>



sns.countplot(x = 'Sex', data = df, palette = 'Set3')

<matplotlib.axes._subplots.AxesSubplot at 0x7ff586e42c10>

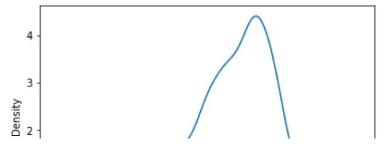


a = pd.read_csv('abalone.csv')

a['age'] = a['Rings']+1.5
a = a.drop('Rings',axis = 1)

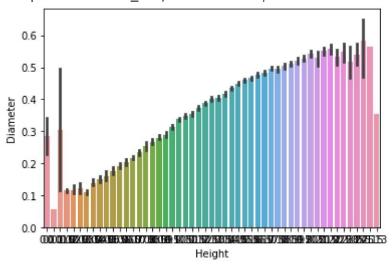
sns.kdeplot(a['Diameter'])

<matplotlib.axes._subplots.AxesSubplot at 0x7ff586d84350>



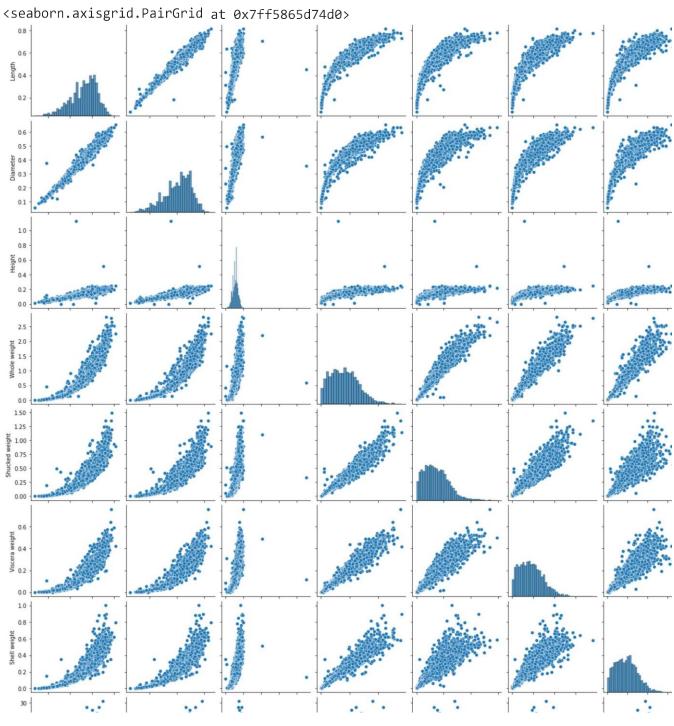
sns.barplot(x=df.Height,y=df.Diameter)

<matplotlib.axes._subplots.AxesSubplot at 0x7ff5868893d0>



sns.pairplot(a)

+



a.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 4177 entries, 0 to 4176
Data columns (total 9 columns):

	`	,	
#	Column	Non-Null Count	Dtype
0	Sex	4177 non-null	object
1	Length	4177 non-null	float64
2	Diameter	4177 non-null	float64
3	Height	4177 non-null	float64
4	Whole weight	4177 non-null	float64
5	Shucked weight	4177 non-null	float64
6	Viscera weight	4177 non-null	float64

7 Shell weight 4177 non-null float64 8 age 4177 non-null float64

dtypes: float64(8), object(1)
memory usage: 293.8+ KB

a['Diameter'].describe()

count mean 9.407.000000 mean 9.407881 std 9.099240 min 9.055000 25% 9.350000 50% 9.425000 75% 9.480000 max 9.650000

Name: Diameter, dtype: float64

a['Sex'].value_counts()

M 1528 I 1342

Name: Sex, dtype: int64

df['Height'].describe()

4177.000000 0.139516
0.041827
0.000000
0.115000
0.140000
0.165000
1.130000

Name: Height, dtype: float64

df[df.Height == 0]

	Sex	Length	Diameter	Height	Whole	Shucked	Viscera	
1257	I	0.430	0.34	0.0	w ei ght	we2965	wedg60	
3996	ı	0.315	0.23	0.0	0.134	0.0575	0.0285	

df['Diameter'].median()

0.425

df['Shucked weight'].skew()

0.7190979217612694

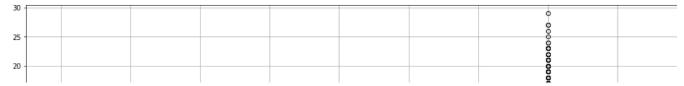
```
missing_values = df.isnull().sum().sort_values(ascending = False)
percentage_missing_values = (missing_values/len(df))*100
pd.concat([missing_values, percentage_missing_values], axis = 1, keys= ['Missing values', '%
```

	Missing values	% Missing
Sex	0	0.0
Length	0	0.0
Diameter	0	0.0
Height	0	0.0
Whole weight	0	0.0
Shucked weight	0	0.0
Viscera weight	0	0.0
Shell weight	0	0.0
Rings	0	0.0

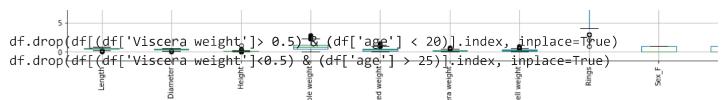
```
q1=df.Rings.quantile(0.25)
q2=df.Rings.quantile(0.75)
iqr=q2-q1
print(iqr)
```

3.0

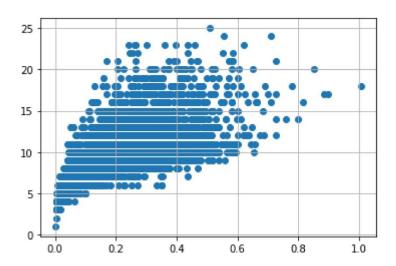
```
<matplotlib.axes._subplots.AxesSubplot at 0x7ff581e54fd0>
```



```
df['age'] = df['Rings']
df = df.drop('Rings', axis = 1)
```



```
var = 'Shell weight'
plt.scatter(x = df[var], y = df['age'])
plt.grid(True)
```



numerical_features = df.select_dtypes(include = [np.number]).columns
categorical_features = df.select_dtypes(include = [np.object]).columns

/usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:2: DeprecationWarning: `np Deprecated in NumPy 1.20; for more details and guidance: https://numpy.org/devdocs/rele

→

abalone_numeric = df[['Length', 'Diameter', 'Height', 'Whole weight', 'Shucked weight','Visce
abalone_numeric.head()

```
Length Diameter Height
                                                                                       age Sex_
                                       weight
                                                                  weight
                                                                              weight
                                                     weight
      0
          0.455
                     0.365
                                        0.5140
                                                     0.2245
                             0.095
                                                                   0.1010
                                                                                0.150
                                                                                        15
x = df.iloc[:, 0:1].values
y = df.iloc[:, 1]
У
     0
             0.365
             8:355
             0.365
     3
     4
             0.255
             0.450
     4172
     4173
             0.440
     4174
             0.475
     4175
             0.485
     4176
             0.555
     Name: Diameter, Length: 4150, dtype: float64
print ("\n ORIGINAL VALUES: \n\n", x,y)
      ORIGINAL VALUES:
      [[0.455]
      [0.35]
      [0.53]
      . . .
      [0.6]
      [0.625]
                        0.365
      [0.71]] 0
     1
             0.265
     2
             0.420
     3
             0.365
     4
             0.255
     4172
             0.450
     4173
             0.440
```

Whole

Shucked

Viscera

Shell

VALUES AFTER MIN MAX SCALING:

fieωmyskhearMax_scaler.fit_transform(x,y)

Name: Diameter, Length: 4150, dtype: float64

print ("\n VALUES AFTER MIN MAX SCALING: \n\n", new_y)

min max scaler = preprocessing.MinMaxScaler(feature range =(0, 1))

import preprocessing

4174

4175 4176 0.475
0.485

0.555

```
[[0.51351351]
      [0.37162162]
      [0.61486486]
      [0.70945946]
      [0.74324324]
      [0.85810811]]
X = df.drop(_{age'}, axis = 1)
v = df['age']
from sklearn.preprocessing import StandardScaler
from sklearn.model_selection import train_test_split, cross_val_score
from sklearn.feature_selection import SelectKBest
standardScale = StandardScaler()
standardScale.fit_transform(X)
selectkBest = SelectKBest()
X new = selectkBest.fit_transform(X, y)
X_train, X_test, y_train, y_test = train_test_split(X_new, y, test_size = 0.25)
X_train
     array([[0.475, 0.355, 0.12 , ..., 0.
                                             , 0.
                                                    , 1.
                                                            ],
            [0.5, 0.365,
            [0.58, 0.43, 0.13, ..., 0., 0.
                                                    , 1.
                                                            ],
                            0.125, ..., 0. , 0.
                                                            ],
            . . . ,
                                                    , 1.
            [0.49, 0.38, 0.135, ..., 0.
                                             , 0.
                                                    , 1.
                                                            ],
                           0.1 ,
            [0.5 , 0.31 , 0.115, ..., 0.
                                             , 1.
                                                    , 0.
                                                            ],
                                             , 1.
                                ..., 0.
                                                            ]])
                  , 0.37 ,
                                                    , 0.
y_train
     65
              8
     3826
             19
     1753
             10
     503
             13
             . .
     1829
             13
     12
             11
              7
     2127
              7
     2980
     Name: age, Length: 3112, dtype: int64
from
             import linear_model as lm
     sklearn.linear_model import
     sklearn
from
                                  LinearRegression
```

```
model=lm.LinearRegression()
results=model.fit(X train,v train)
accuracy = model.score(X_train, y_train)
print('Accuracy of the model:', accuracy)
     Accuracy of the model: 0.5385553745257212
lm = LinearRegression()
lm.fit(X_train, y_train)
y_train_pred = lm.predict(X_train)
y_train_pred
     array([ 8.7821132 , 8.69223506, 7.91346477, ..., 10.66419868,
             6.49661756, 7.32689668])
X_train
     array([[0.475, 0.355, 0.12 , ..., 0. , 0.
                                                            ],
            [0.5 , 0.365, 0.13 , ..., 0. , 0.
                                                    , 1.
                                                            ],
            [0.58, 0.43, 0.125, ..., 0., 0.
                                                    , 1.
                                                            ],
            [0.49, 0.38, 0.135, ..., 0., 0.
                                                    , 1.
                                                            ],
                 , 0.31 , 0.1 , ..., 0. , 1. , 0.37 , 0.115, ..., 0. , 1.
            [0.4
                                                    , 0.
                                                            1,
            [0.5
                                                            11)
                                                    , 0.
y_train
     65
              8
     3826
             10
     1753
             10
     503
             13
             13
     1820
     1902
             11
     12
             11
     2127
              7
              7
     2980
     Name: age, Length: 3112, dtype: int64
     sklearn.metrics import mean_absolute_error, mean_squared_error
frommean_squared_error(y_train, y_train_pred)
grint('Mean Squared error of training set :%2f'%s)
     Mean Squared error of training set :4.638811
y_train_pred = lm.predict(X_train)
y_test_pred = lm.predict(X_test)
y_test_pred
```

```
array([ 9.65587378, 5.56256054, 12.55940421, ..., 8.63884103,
             7.17360886, 9.97959562])
X_test
     array([[0.42 , 0.325, 0.115, ..., 0. , 0.
                                                             ],
             [0.31, 0.225, 0.05, ..., 0., 1.
                                                             ],
             [0.52, 0.415, 0.175, \ldots, 0., 0.
                                                     , 1.
                                                             ],
             [0.385, 0.305, 0.105, ..., 1. , 0.
                                                     , 0.
            [0.635, 0.495, 0.015, ..., 1. , 0. [0.55, 0.43, 0.145, ..., 0. , 1.
                                                             ],
                                                     , 0.
                                                             ],
                                                             ]])
                                                     , 0.
y_test
     542
             15
     2116
              6
     3223
              8
     1974
             10
     2554
             7
             9
     2525
     1956
             11
              7
     2287
              9
     1174
     2760
             10
     Name: age, Length: 1038, dtype: int64
    mean_squared_error(y_test, y_test_pred)
print('Mean Squared error of testing set :%2f'%p)
     Mean Squared error of testing set :5.150781
     sklearn.metrics import r2_score
$romr2_score(y_train, y_train_pred)
print('R2 Score of training set:%.2f'%s)
```

R2 Score of training set:0.54