# **Assignment -4**

Assignment Date	17 November 2022
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Student Roll Number	111619104139
Maximum Marks	4 Marks

```
import pandas as pd
import numpy as np
from matplotlib import pyplot as plt
import seaborn as sns
from sklearn.linear_model import LinearRegression

df=pd.read_csv("/content/drive/NyDrive/Colab Notebooks/abalone.csv")
```

#### Univariate Analysis

```
df.hist(figsize=(20,10), grid=False, layout=(2, 4), bins = 3B)
     array([[<matplotlib.axes. subplots.AxesSubplot object at 8x7f3d1b8fb698>,
              <matplotlib.axes._subplots.AxesSubplot object at 0x7f3dlade4d98>,
              <matplotlib.axes._subplots.AxesSubplot object at 0x7f3dladaa398>,
             <matplotlib.axes._subplots.AxesSubplot object at Bx7f3d1ad60998>],
            [<matplotlib.axes._subplots.AxesSubplot object at 0x7f3dladl6f98>,
             <matplotlib.axes._subplots.AxesSubplot object at Bx7f3d1acda5d8>,
             <matp1otlib.axes._subplots.AxesSubplot object at Bx7f3dlac8fc58>,
             <matplotlib.axes._subplots.AxesSubplot object at 8x7f3dlac53ld8>]],
           dtype=object)
                                                            Diameter
                                                                                                    Height
                                                                                                                                        Whole weight
                                                                                   1600
      400
                                             350
                                                                                   1400
      350
                                                                                                                           250
                                                                                   1200
      300
                                             250
                                                                                    1000
                                                                                                                           200
      250
                                             200
                                                                                    800
      200
                                                                                                                            150
                                             150
                                                                                    600
      150
                                                                                                                            100
      100
                                             100
                                                                                    400
                                                                                                                            50
       50
                                              50
                                                                                    200
              0.2
                      0.4
                             0.6
                                                   0.1
                                                       0.2
                                                            0.3
                                                                0.4
                                                                    0.5
                                                                                                 0.4
                                                                                                      0.6
                                                                                                           0.8
                                                                                                                                   0.5
                                                                                                                                        10
                                                                                                                                             1.5
                                                                                                                                                  2.0
                  Shucked weight
                                                          Viscera weight
                                                                                                  Shell weight
                                                                                                                                            age
      350
                                             350
                                                                                    350
                                                                                                                            600
      300
                                             300
                                                                                    300
                                                                                                                            500
      250
                                             250
                                                                                    250
                                                                                                                            400
      200
                                             200
                                                                                    200
      150
                                                                                                                           300
                                             150
                                                                                    150
                                                                                                                            200
      100
                                             100
                                                                                    100
                                                                                                                            100
       50
                                              50
                                                                                     50
         0.00 0.25 0.50 0.75 100 125 150
                                                                                             0.2
```

 Length
 Diameter
 Height
 whole weight
 Shucked weight
 Viscera weight
 Shell weight
 age

 Sex

 0.427746
 0.326494
 0.107996
 0.431363
 0.191035
 0.092010
 0.128182
 9.390462

 M
 0.561391
 0.439287
 0.151381
 0.991459
 0.432946
 0.215545
 0.281969
 12.205497

F 0.579093 0.454732 0.158011 1.046532

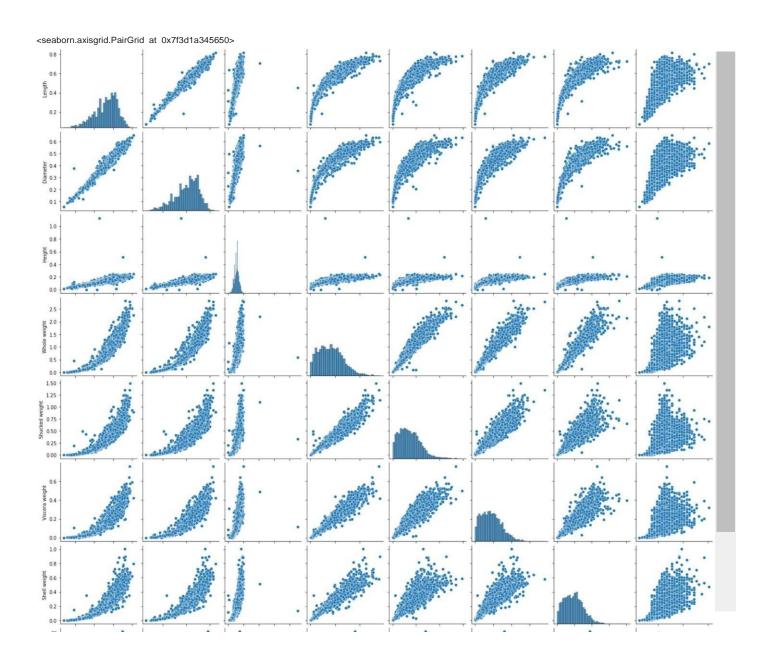
0.446188

0.230689

0.302010 12.629304

# Bivariate Analysis

numerical\_features = df.select\_dtypes(include = [np.number]).columns
sns.pairplot(df[numerical\_features])



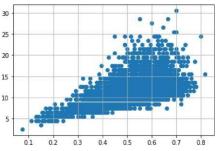
# Descriptive statistics

Length         Diameter         Height         whole weight         Shucked weight         viscera weight         Shell weight         age           count         4177.000000         0.139203         3.224169         0.139203         3.224169         0.00000         0.001500         0.001500         0.001500         0.001500         0.001500         0.001500         0.001500         0.001500         0.186000         0.005500         0.130000         9.500000         0.001500         0.170000         0.234000         10.50000 <td< th=""><th colspan="9">df.describe()</th></td<>	df.describe()								
mean         0.523992         0.407881         0.139516         0.828742         0.359367         0.180594         0.238831         11.433684           std         0.120093         0.099240         0.041827         0.490389         0.221963         0.109614         0.139203         3.224169           min         0.075000         0.055000         0.000000         0.002000         0.001000         0.000500         0.001500         2.500000           25%         0.450000         0.350000         0.115000         0.441500         0.186000         0.093500         0.130000         9.500000           50%         0.545000         0.425000         0.140000         0.799500         0.336000         0.171000         0.234000         10.500000           75%         0.615000         0.480000         0.165000         1.153000         0.502000         0.253000         0.329000         12.500000		Length	Diameter	Height	whole weight	Shucked weight	viscera weight	Shell weight	age
std         0.120093         0.099240         0.041827         0.490389         0.221963         0.109614         0.139203         3.224169           min         0.075000         0.055000         0.000000         0.002000         0.001000         0.000500         0.001500         2.500000           25%         0.450000         0.350000         0.115000         0.441500         0.186000         0.093500         0.130000         9.500000           50%         0.545000         0.425000         0.140000         0.799500         0.336000         0.171000         0.234000         10.500000           75%         0.615000         0.480000         0.165000         1.153000         0.502000         0.253000         0.329000         12.500000	count	4177.000000	4177.000000	4177.000000	4177.000000	4177.000000	4177.000000	4177.000000	4177.000000
min         0.075000         0.055000         0.000000         0.002000         0.001000         0.000500         0.001500         2.500000           25%         0.450000         0.350000         0.115000         0.441500         0.186000         0.093500         0.130000         9.500000           50%         0.545000         0.425000         0.140000         0.799500         0.336000         0.171000         0.234000         10.500000           75%         0.615000         0.480000         0.165000         1.153000         0.502000         0.253000         0.329000         12.500000	mean	0.523992	0.407881	0.139516	0.828742	0.359367	0.180594	0.238831	11.433684
25%       0.450000       0.350000       0.115000       0.441500       0.186000       0.093500       0.130000       9.500000         50%       0.545000       0.425000       0.140000       0.799500       0.336000       0.171000       0.234000       10.500000         75%       0.615000       0.480000       0.165000       1.153000       0.502000       0.253000       0.329000       12.500000	std	0.120093	0.099240	0.041827	0.490389	0.221963	0.109614	0.139203	3.224169
50%     0.545000     0.425000     0.140000     0.799500     0.336000     0.171000     0.234000     10.500000       75%     0.615000     0.480000     0.165000     1.153000     0.502000     0.253000     0.329000     12.500000	min	0.075000	0.055000	0.000000	0.002000	0.001000	0.000500	0.001500	2.500000
75% 0.615000 0.480000 0.165000 1.153000 0.502000 0.253000 0.329000 12.500000	25%	0.450000	0.350000	0.115000	0.441500	0.186000	0.093500	0.130000	9.500000
	50%	0.545000	0.425000	0.140000	0.799500	0.336000	0.171000	0.234000	10.500000
max 0.815000 0.650000 1.130000 2.825500 1.488000 0.760000 1.005000 30.500000	75%	0.615000	0.480000	0.165000	1.153000	0.502000	0.253000	0.329000	12.500000
	max	0.815000	0.650000	1.130000	2.825500	1.488000	0.760000	1.005000	30.500000

Check for missing values

df.isnull().sum()

```
df - pd.get dummies(df)
dummy_da ta = df . copy()
var = 'Viscera weight'
plt.scatter(x = df[var], y = df['age'],)
plt.grid(True)
# outliers removal
d-F. drop(df[ (d-F[ ' VI scera weight ' ] \gt 0. 5) & (df-[ ' age ' ] \lt 20) ] . Index, inp1ace=True)
\label{eq:dfdf} $$ df.drop(df[(df['Uiscera weight']<0.5) & (df['age'] \rightarrow 25)].index, inplace=True) $$
var - 'Shell weight'
plt.scatter(x = df[var], y = df['age'],)
plt.grid(True)
#Outliers removal
var = 'Shucked weight'
plt.scatter(x = df[var], y = df['age'],)
plt.grid(True)
#Outlier removal
var = ' Nhole weight '
pit . scatter (x = df-[var], y = df['age'])
p1t . grid(True)
df.drop(df[(df['Whole weight'] >= 2.5) &
          (df['age'] < 25)].index, inplace = True)</pre>
df. drop(df-[(df['Nhole weight']<2.5) & (
d-F['age'] \rightarrow 25)]. Index, 1nplace = True)
var = ' Diameter '
pit . scatter (x = df-[var] , y = df[ ' age ' ] )
p1t . grid(True)
df.drop (df-[(df['Diazeten'] <8.1) &
\label{eq:df-def} $$ (df['age'] < 5)].index, inplace = True) $$ df. drop(df-[(df['Diameter'] < 0.6) & (
d-F['age'] > 25)]. Index, 1nplace = True)
d-F-. drop(df-[ (d1°[ ' Diameter '] >=0. 6) & (
df-['age'] < 25)] . Index, 1nplace = True)
var = 'Height'
p1t . scatter (x - df[var] , y - df[ 'age '])
p1t.grid(True)
d-F. drop(d-I- [ (df-[ ' Height ' ] > 6 . 4) &
          (df[ 'age '] < 15) ] . Index, Inplace = True)
d-F. drop(df-[ (d-F[ 'Height '] <0. 4) & (
d-I^{\circ}['age'] > 25)] . index, 1nplace = True)
var = 'Length'
plt.scatter(x = df[var], y = df['age'])
plt.grid(True)
df.drop(df[(df['Leugth'] < 8.1) &
          (df['age'] < 5)].index, inplace = True)</pre>
dfdropd[df['Leugth]<0.8) & (
df['age'] > 25)].index, inplace = True)
df.dropd[df['Length]>=8.8) & (
df['age'] < 25)].iudex, inplace = True)</pre>
```



#### Categorical columns

 $numerical\_features = df.select\_dtypes(include = [np.number]).columns \\ categorica1\_features = df.select\_dtypes(include = [np.object]).columns$ 

/usr/local/lib/python3.7/dist-packages/ipykernel\_launcher.py:2: Deprecationwarning: 'up.object' is a deprecated alias for the builtin 'object' To siler Deprecated in NumPy 1.20; for more details and guidance: <a href="https://numpy.org/devdocs/release/1.20.8-notes.html#deprecations">https://numpy.org/devdocs/release/1.20.8-notes.html#deprecations</a>

#### numerical\_features

# categonica I\_featunes

```
Index(['Sex'], dtype='object')
```

#### **ENCODING**

from sklearn.preprocessing import LabelEncoder le=LabelEncoder() print(df.Sex.value\_counts())

M 1525 1 1341 F 1301

Name: Sex, dtype: int64

# x=df.iloc[:,:5]

1	Nhole we1ght	Helght	Diameter	Length	Sex		
	0.5140	0.095	0.365	0.455	М	0	
	0.2255	0.090	0.265	0.350	М	1	
	0.6770	0.135	0.420	0.530	F	2	
	0.5160	0.125	0.365	0.440	М	3	
	0.2050	0.080	0.255	0.330		4	
	0.8870	0.165	0.450	0.565	F	4172	
	0.9660	0.135	0.440	0.590	М	4173	
	1.1760	0.205	0.475	0.600	М	4174	
	1.0945	0.150	0.485	0.625	F	4175	
	1.9485	0.195	0.555	0.710	М	4176	

y=df.iloc[:,5:]

4167 rows • 5 columns

	Shucked weight	VIscera weight	Shell weight	age	10
0	0.2245	0.1010	0.1500	16.5	
1	0.0995	0.0485	0.0700	8.5	
2	0.2565	0.1415	0.2100	10.5	
3	0.2155	0.1140	0.1550	11.5	
4	0.0895	0.0395	0.0550	8.5	
4172	0.3700	0.2390	0.2490	12.5	
4173	0.4390	0.2145	0.2605	11.5	
4174	0.5255	0.2875	0.3080	10.5	
4175	0.5310	0.2610	0.2960	11.5	
4176	0.9455	0.3765	0.4950	13.5	
4167 ro	ws 4 columns				

 $from \ sk1earn.model\_selection \ import \ train\_test\_split \\ x\_train,x\_test,y\_train,y\_test=train\_test\_split(x,y,test\_size=0.2)$ 

Model Building

 $\label{thm:constraint} from sklearn.linear_model import LinearRegression \\ mlr=LinearRegression() \\ mlr.fit(x\_train,y\_train) \\$ 

Train and Test model

# x\_test [6:5]

	Sex	Length	Diameter	Height	Nhole we1ght
661		0.535	0.450	0.170	0.781
370	F	0.650	0.545	0.165	1.566
2272	М	0.635	0.510	0.210	1.598
1003	М	0.595	0.455	0.150	1.044
1145	М	0.580	0.455	0.195	1.859

# y\_test[0:5]

	Shucked we1ght	vlscera we1ght	Shell we1ght	age
661	0.3055	0.1555	0.295	12.5
370	0.6645	0.3455	0.415	17.5
2272	0.6535	0.2835	0.580	16.5
1003	0.5180	0.2205	0.270	10.5
1145	0.9450	0.4260	0.441	10.5

#### Feature Scaling

from sklearn.preprocessing import StandardScaler ss=StandardScaler() x\_train=ss.fit\_transform(x\_train) mlrpred=mlr.predict(x\_test[B:9]) mlrpred

Performance measure

I-rom sklearn .metric s Import r2\_score r2\_s core(m1r . predict (x\_test) , y\_test )

0.5597133867640833