

Assignment -4

Assignment Date	06 November 2022
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Student Roll Number	812719104023
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
```

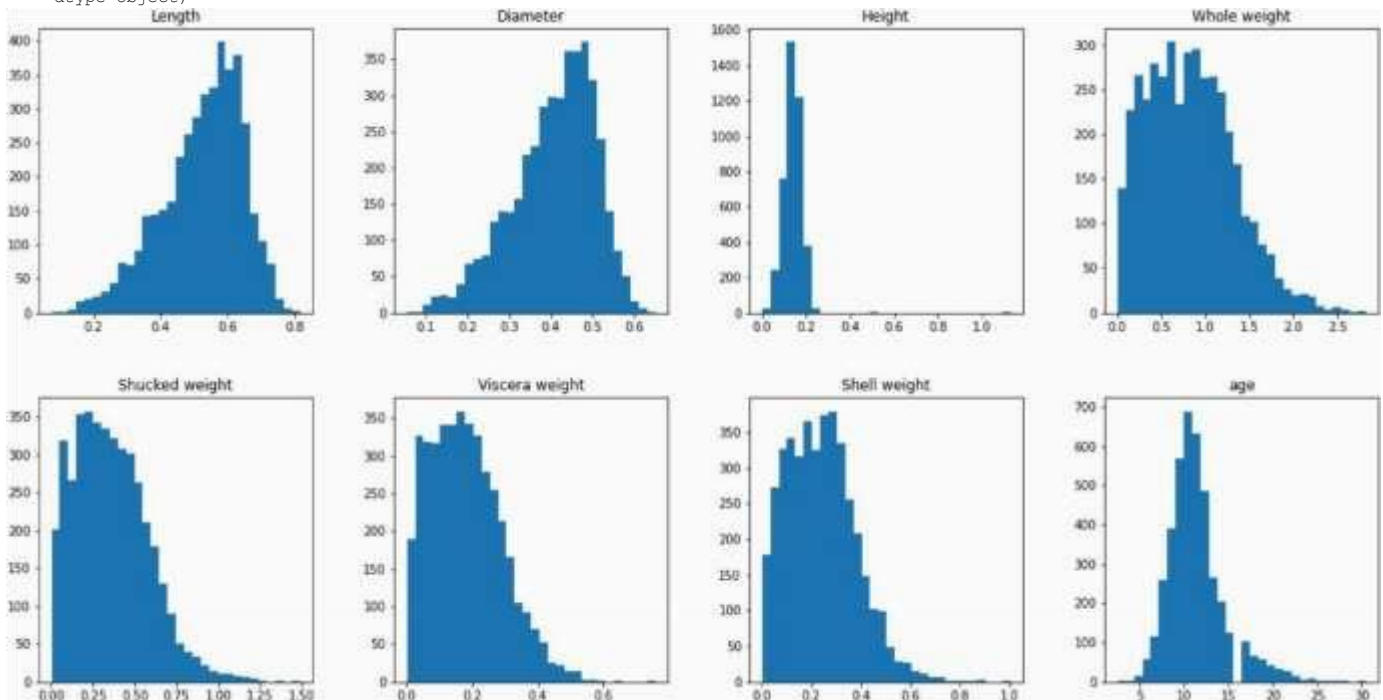
```
df=pd.read_csv("/content/drive/MyDrive/Colab Notebooks/shalone.csv")
```

```
d-F['age'] = d-F['Rings'] + 1.5
df = df.drop('Rings', axis = 1)
```

Univariate Analysis

```
df.hist(figsize=(20,10), grid=False, layout=(2, 4), bins = 38)
```

```
[ array([[<matplotlib.axes._subplots.AxesSubplot object at 8x7f3d1b8fb698>,
<matplotlib.axes._subplots.AxesSubplot object at 0x7f3d1ade4d98>,
<matplotlib.axes._subplots.AxesSubplot object at 0x7f3d1adaa398>,
<matplotlib.axes._subplots.AxesSubplot object at 8x7f3d1ad60998>],
[<matplotlib.axes._subplots.AxesSubplot object at 0x7f3d1ad16f98>,
<matplotlib.axes._subplots.AxesSubplot object at 8x7f3d1acda5d8>,
<matplotlib.axes._subplots.AxesSubplot object at 8x7f3dlac8fc58>,
<matplotlib.axes._subplots.AxesSubplot object at 8x7f3dlac531d8>]],
dtype=object)
```



```
df.groupby('Sex')[['Length', 'Diameter', 'Height', 'Whole weight', 'Shucked weight',
'Viscera weight', 'Shell weight', 'age']].mean().sort_values('age')
```

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

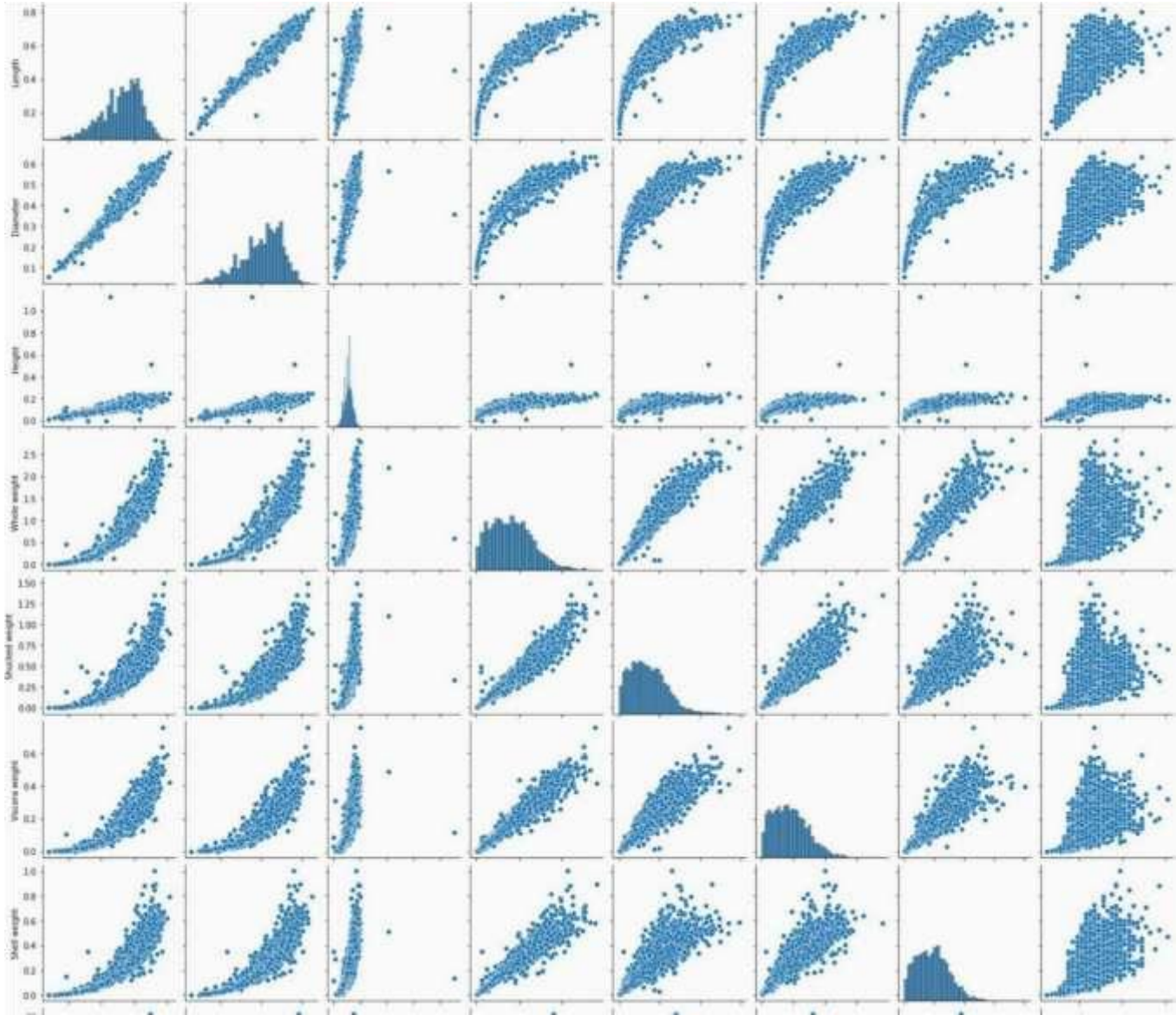
0.427746 0.326494 0.107996

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

df.describe()

	Length	Diameter	Height	whole weight	Shucked weight	viscera weight	Shell weight	age
count	417	417	417	417	417	417	417	417
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

max	0.815000	0.650000	1.130000	2.825500	1.488000	0.760000	1.005000	30.500000
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Check for missing values

```
df.isnull().sum()
```

Outlier handling

```
df = pd.get_dummies(df) dummy_data = df.copy()
var = 'Viscera weight' plt.scatter(x = df[var], y = df['age'],) plt.grid(True)
```

```
# outliers removal
d-F.drop(df[(d-F['Viscera weight'] > 0.5) & (df['age'] < 20)].Index, inplace=True)
df.drop(df[(df['Viscera weight'] < 0.5) & (df['age'] > 25)].index, inplace=True)
var = 'Shell weight'
plt.scatter(x = df[var], y = df['age'],) plt.grid(True) #Outliers removal
df.drop(df[(df['Shell weight'] > 0.6) & (df['age'] < 25)].index, inplace=True)
df.drop(df[(df['Shell weight'] < 0.8) & (df['age'] > 25)].index, inplace=True)
```

```
var = 'Shucked weight'
plt.scatter(x = df[var], y = df['age'],) plt.grid(True)
#Outlier removal df.drop(df[(df['Shucked weight'] >= 1) & (df['age'] < 28)].index, inplace=True) df.drop(df[(df['Shucked weight'] < 1) & (df['age'] > 28)].index, inplace=True)
```

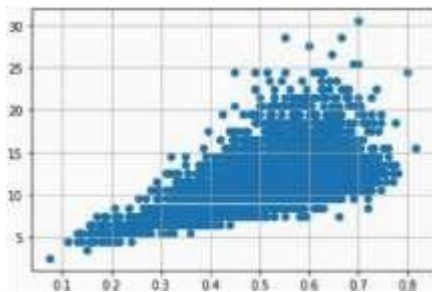
```
var = 'Diameter'
plt.scatter(x = df[var], y = df['age'])
plt.grid(True)

df.drop(df[(df['Diameter'] < 0.1) & (df['age'] < 5)].index, inplace = True)
df.drop(df[(df['Diameter'] < 0.6) & (df['age'] > 25)].index, inplace = True)
d-F.drop(df[(df['Diameter'] >= 0.6) & (df['age'] < 25)].index, inplace = True)
```

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

df.drop(df[(df['Whole weight'] >= 2.5) & (df['age'] < 25)].index, inplace = True)
df.drop(df[(df['Whole weight'] < 2.5) & (df['age'] > 25)].index, inplace = True) var = 'Height'
plt.scatter(x = df[var], y = df['age']) plt.grid(True)
d-F.drop(df[(df['Height'] > 6.4) & (df['age'] < 15)].index, inplace = True)
d-F.drop(df[(df['Height'] < 0.4) & (df['age'] > 25)].index, inplace = True)
```

```
var = 'Length'
plt.scatter(x = df[var], y = df['age']) plt.grid(True)
df.drop(df[(df['Length'] < 8.1) & (df['age'] < 5)].index, inplace = True)
df.drop(df[(df['Length'] < 0.8) & (df['age'] > 25)].index, inplace = True)
df.drop(df[(df['Length'] >= 8.8) & (df['age'] < 25)].index, inplace = True)
```



Categorical columns

```
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.object' is a deprecated alias for the builtin 'object'. To silence this warning, use 'object' instead of 'np.object' in the source. To silence this warning in NumPy, you should set the environment variable 'NPY_USE_DEPRECATED=0' before running the application. Deprecation is scheduled for removal in NumPy 1.20; for more details and guidance: <https://numpy.org/devdocs/release/1.20.0-notes.html#deprecations>

categorical_features

numerical_features

Index(['Length', 'Diameter', 'Height', 'Whole weight', 'Shucked weight', 'Viscera weight', 'Shell weight', 'age'], dtype='object')

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]
```

	Sex	Length	Diameter	Height	Whole weight
0	M	0.455	0.365	0.095	0.5140
1	M	0.350	0.265	0.090	0.2255
2	F	0.530	0.420	0.135	0.6770
3	M	0.440	0.365	0.125	0.5160
4		0.330	0.255	0.080	0.2050
		0.565	0.450	0.165	
4172	F				0.8870
4173	M	0.590	0.440	0.135	0.9660
4174	M	0.600	0.475	0.205	1.1760

4175	F	0.625	0.485	0.150	1.0945
4176	M	0.710	0.555	0.195	1.9485

4167 rows • 5 columns

Train, Test, Split

```
y=df.iloc[:,5:]
```

	Shucked weight	Viscera weight	Shell weight	age
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	11.5
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	11.5
4173	0.4390	0.2145	0.2605	11.5
4174	0.5255	0.2875	0.3080	11.5
4175	0.5310	0.2610	0.2960	11.5
4176	0.9455	0.3765	0.4950	11.5

4167 rows 4 columns

```
from sklearn.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

```
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 weight

		0.535	0.450	0.170	0.781
661					
370	F	0.650	0.545	0.165	1.566
2272	M	0.635	0.510	0.210	1.598
1003	M	0.595	0.455	0.150	1.044
1145	M	0.580	0.455	0.195	1.859

```
y_test[0:5]
```

	Shucked weight	viscera weight	Shell weight	age
661	0.3055	0.1555	0.295	11.5
370	0.6645	0.3455	0.415	17.5
2272	0.6535	0.2835	0.580	11.5
1003	0.5180	0.2205	0.270	10.5
1145	0.9450	0.4260	0.441	11.5

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
```

```
from sklearn.metrics import r2_score  
r2_score(m1r.predict(x_test), y_test)
```

0.5597133867640833