| Assignment Date | 29.10.2022 |
|---------------------|-------------------|
| Student Name | M.shakthi vignesh |
| Student Roll Number | 312319205146 |
| Maximum Marks | 2 Marks |

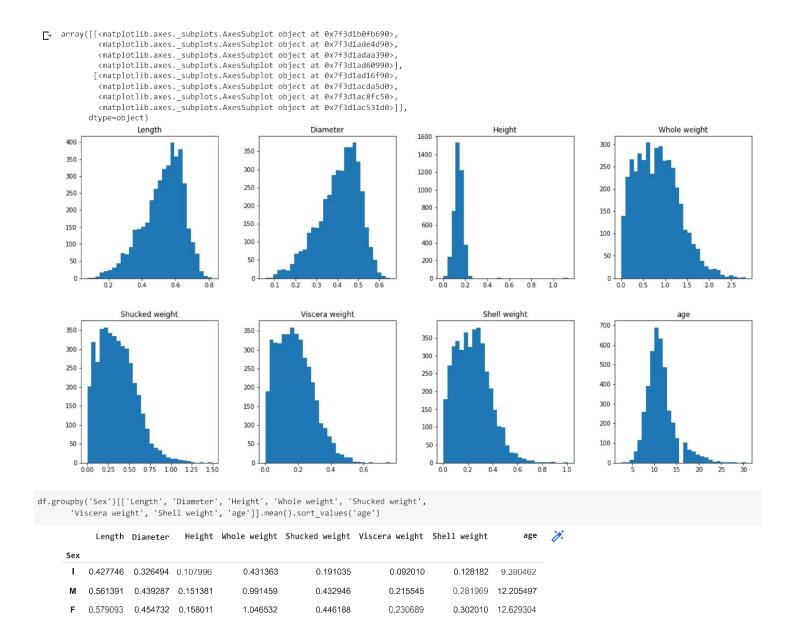
```
import pandas as pd
import numpy as np
from matplotlib import pyplot as plt
import seaborn as sns
from sklearn.linear_model import LinearRegression
```

 ${\tt df=pd.read_csv("/content/drive/MyDrive/Colab}\ \ {\tt Notebooks/abalone.csv")}$

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

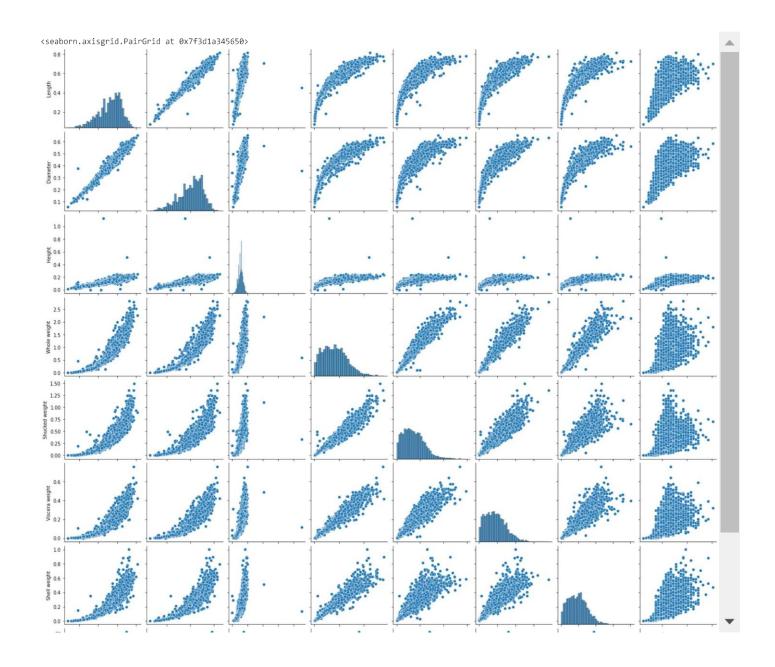
Univariate Analysis

```
df.hist(figsize=(20,10), grid=False, layout=(2, 4), bins = 30)
```



Bivariate Analysis

numerical_features = df.select_dtypes(include = [np.number]).columns
sns.pairplot(df[numerical_features])



Descriptive statistics

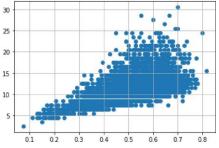
| f.describe(|
|-------------|
|-------------|

| | Length | Diameter | Height | Whole weight | Shucked weight | Viscera weight | Shell weight | age | 1 |
|-------|-------------|-------------|-------------|--------------|----------------|----------------|--------------|-------------|---|
| count | 4177.000000 | 4177.000000 | 4177.000000 | 4177.000000 | 4177.000000 | 4177.000000 | 4177.000000 | 4177.000000 | |
| 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 | |

Check for missing values

df.isnull().sum()

```
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
  \label{eq:df_def} $$ df.drop(df[(df['Viscera weight']> 0.5) \& (df['age'] < 20)].index, inplace=True) $$
  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
  \label{eq:dfdf} $$ df.drop(df[(df['Shell weight']> 0.6) \& (df['age'] < 25)].index, inplace=True) $$
  \label{eq:df_df_df_df_df_df} $$ df.drop(df[(df['Shell weight']<0.8) & (df['age'] > 25)].index, inplace=True) $$
  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'] < 20)].index, inplace=True)</pre>
  \label{lem:df_df_df_df_df_df_df} $$ df.drop(df[(df['Shucked weight']<1) & (df['age'] > 20)].index, inplace=True) $$
  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)</pre>
  df.drop(df[(df['Whole weight']<2.5) & (
  df['age'] > 25)].index, inplace = True)
var = 'Diameter'
plt.scatter(x = df[var], y = df['age'])
plt.grid(True)
df.drop(df[(df['Diameter'] <0.1) &</pre>
          (df['age'] < 5)].index, inplace = True)</pre>
df.drop(df[(df['Diameter']<0.6) & (</pre>
df['age'] > 25)].index, inplace = True)
df.drop(df[(df['Diameter']>=0.6) & (
df['age'] < 25)].index, inplace = True)</pre>
var = 'Height'
plt.scatter(x = df[var], y = df['age'])
plt.grid(True)
df.drop(df[(df['Height'] > 0.4) &
          (df['age'] < 15)].index, inplace = True)</pre>
df.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'] < 0.1) &
          (df['age'] < 5)].index, inplace = True)</pre>
df.drop(df[(df['Length']<0.8) & (
df['age'] > 25)].index, inplace = True)
df.drop(df[(df['Length']>=0.8) & (
df['age'] < 25)].index, inplace = True)</pre>
```



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 siler Deprecated in NumPy 1.20; for more details and guidance: https://numpy.org/devdocs/release/1.20.0-notes.html#deprecations

```
\blacktriangleleft
```

```
numerical_features
```

categorical_features

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

ENCODING

from sklearn.preprocessing import LabelEncoder
le=LabelEncoder()
print(df.Sex.value_counts())

M 1525 I 1341 F 1301

Name: Sex, dtype: int64

x=df.iloc[:,:5]

Х

| | Sex | Length | Diameter | Height | Whole weight |
|------|--------|--------|----------|--------|--------------|
| 0 | М | 0.455 | 0.365 | 0.095 | 0.5140 |
| 1 | М | 0.350 | 0.265 | 0.090 | 0.2255 |
| 2 | F | 0.530 | 0.420 | 0.135 | 0.6770 |
| 3 | М | 0.440 | 0.365 | 0.125 | 0.5160 |
| 4 | 1 | 0.330 | 0.255 | 0.080 | 0.2050 |
| | | | | | |
| 4172 | F | 0.565 | 0.450 | 0.165 | 0.8870 |
| 4173 | М | 0.590 | 0.440 | 0.135 | 0.9660 |
| 4174 | M 0.60 | 0.600 | 0.475 | 0.205 | 1.1760 |
| 4175 | F | 0.625 | 0.485 | 0.150 | 1.0945 |
| 4176 | М | 0.710 | 0.555 | 0.195 | 1.9485 |

4167 rows × 5 columns

y=df.iloc[:,5:]

| | Shucked weight | Viscera weight | Shell weight | age | 1. | |
|------|----------------|----------------|--------------|------|----|--|
| 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 rows × 4 columns

Train, Test, Split

 $\label{local_selection} 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[0:5]

| | Sex | Length | Diameter | Height | Whole weight |
|------|-----|--------|----------|--------|--------------|
| 661 | 1 | 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 weight | Viscera weight | Shell weight | age | 1 |
|------|----------------|----------------|--------------|------|---|
| 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[0:9])
mlrpred

Performance measure

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

0.5597133867640833