## PROJECT TITLE: Airlines Data Analytics for Avaition Industry

#### Team ID: PNT2022TMID29101

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

### 2.LOAD THE DATASET INTO COLLAB

```
df=pd.read_csv("/content/abalone.csv")

df['age'] = df['Rings']+1.5

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

#### 3.UNIVARIATE ANALYSIS

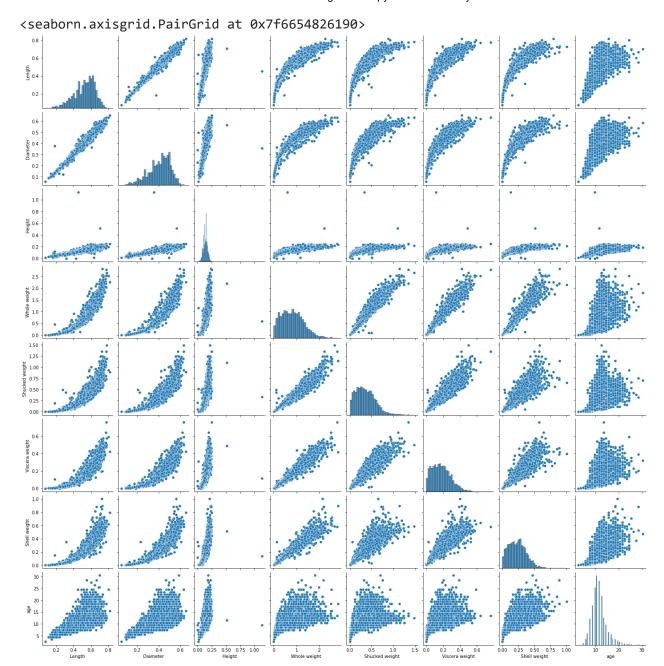
```
df.hist(figsize=(20,10), grid=False, layout=(2, 4), bins = 30)
```

```
array([[<matplotlib.axes. subplots.AxesSubplot object at 0x7f6654e8ffd0>,
              <matplotlib.axes._subplots.AxesSubplot object at 0x7f6654e80490>,
              <matplotlib.axes. subplots.AxesSubplot object at 0x7f6654e37a90>,
              <matplotlib.axes. subplots.AxesSubplot object at 0x7f6654dfa0d0>],
             [<matplotlib.axes._subplots.AxesSubplot object at 0x7f6654db06d0>,
              <matplotlib.axes. subplots.AxesSubplot object at 0x7f6654d67cd0>,
              <matplotlib.axes._subplots.AxesSubplot object at 0x7f6654d2b390>,
              <matplotlib.axes._subplots.AxesSubplot object at 0x7f6654ce28d0>]],
            dtype=object)
               Length
                                        Diameter
                                                                                       Whole weight
                                                      1600
                              350
                                                      1400
      350
                                                                               250
                              300
                                                      1200
      300
                              250
                                                                               200
                              200
      200
                                                                               150
                              150
                                                       600
      150
                              100
                                                       400
                                                       200
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								
	I	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
4									<b>•</b>

#### 3.BIVARIATE ANALYSIS & MULTIVARIATE 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
count	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
std	0.120093	0.099240	0.041827	0.490389	0.221963	0.109614
min	0.075000	0.055000	0.000000	0.002000	0.001000	0.000500
25%	0.450000	0.350000	0.115000	0.441500	0.186000	0.093500
50%	0.545000	0.425000	0.140000	0.799500	0.336000	0.171000
75%	0.615000	0.480000	0.165000	1.153000	0.502000	0.253000
4						<b>&gt;</b>

# 5. Check for Missing Values

df.isnull().sum()

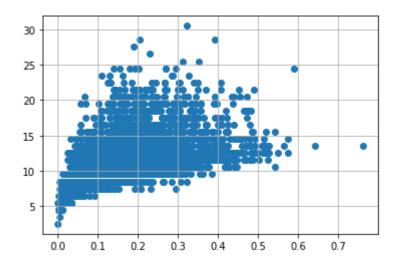
Sex	0
Length	0
Diameter	0
Height	0
Whole weight	0
Shucked weight	0
Viscera weight	0
Shell weight	0
age	0
dtype: int64	

# 6.OUTLIER HANDLING

df = pd.get\_dummies(df)

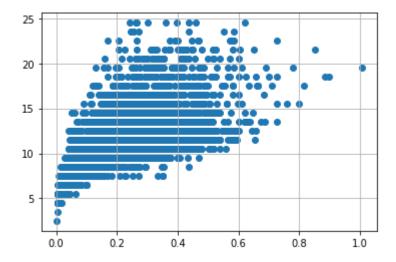
dummy\_data = df.copy()

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



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)

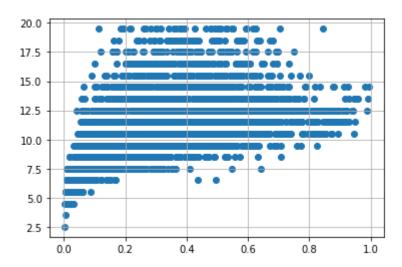
```
var = 'Shell weight'
plt.scatter(x = df[var], y = df['age'],)
plt.grid(True)
#Outliers removal
df.drop(df[(df['Shell weight']> 0.6) & (df['age'] < 25)].index, inplace=True)
df.drop(df[(df['Shell weight']<0.8) & (df['age'] > 25)].index, inplace=True)
```

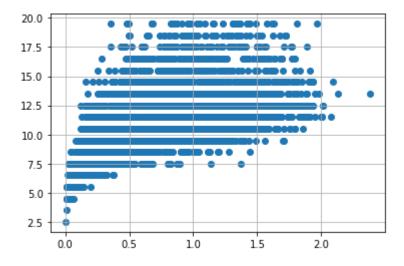


```
var = 'Shucked weight'
plt.scatter(x = df[var], y = df['age'],)
plt.grid(True)
```

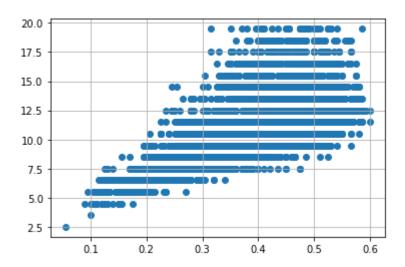
#Outlier removal

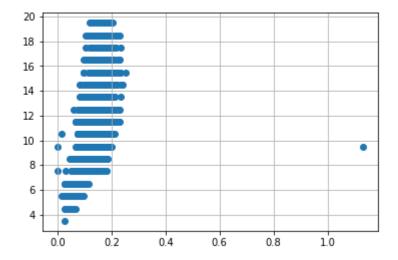
df.drop(df[(df['Shucked weight']>= 1) & (df['age'] < 20)].index, inplace=True)
df.drop(df[(df['Shucked weight']<1) & (df['age'] > 20)].index, inplace=True)

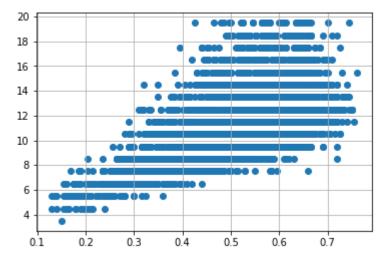




```
df.drop(df[(df['Diameter']>=0.6) & (
df['age'] < 25)].index, inplace = True)</pre>
```







### 7. 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 Deprecated in NumPy 1.20; for more details and guidance: <a href="https://numpy.org/devdocs/releaded-numpy.org/devdocs/rel

```
numerical_features
```

categorical\_features

Index([], dtype='object')

### **ENCODING**

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

0.575 93 0.625 91 0.580 89 0.550 89 0.620 83 ... 0.220 2 0.755 1 0.135 1 0.760 1

Name: Length, Length: 126, dtype: int64

# 8. Split the dependent and independent variables

	Length	Diameter	Height	Whole weight	Shucked weight
0	0.455	0.365	0.095	0.5140	0.2245
1	0.350	0.265	0.090	0.2255	0.0995
2	0.530	0.420	0.135	0.6770	0.2565
3	0.440	0.365	0.125	0.5160	0.2155
4	0.330	0.255	0.080	0.2050	0.0895
4172	0.565	0.450	0.165	0.8870	0.3700
4173	0.590	0.440	0.135	0.9660	0.4390
4174	0.600	0.475	0.205	1.1760	0.5255
4175	0.625	0.485	0.150	1.0945	0.5310
4176	0.710	0.555	0.195	1.9485	0.9455

3995 rows × 5 columns

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

	Viscera weight	Shell weight	age	Sex_F	Sex_I	Sex_M
0	0.1010	0.1500	16.5	0	0	1
1	0.0485	0.0700	8.5	0	0	1
2	0.1415	0.2100	10.5	1	0	0
3	0.1140	0.1550	11.5	0	0	1

# 9. Feature Scaling

#Scaling the Independent Variables
print ("\n ORIGINAL VALUES: \n\n", x,y)

#### ORIGINAL VALUES:

	Length	Diameter	Hoigh+	Whole weight	Shuck	rod woi	ah+			
0	•		_	•	Siluci		•			
0	0.455	0.365	0.095	0.5140		0.22				
1	0.350	0.265	0.090	0.2255		0.09				
2	0.530	0.420	0.135	0.6770		0.25	65			
3	0.440	0.365	0.125	0.5160		0.21	L55			
4	0.330	0.255	0.080	0.2050		0.08	395			
	• • •	• • •	• • •	• • •						
4172	0.565	0.450	0.165	0.8870		0.37	700			
4173	0.590	0.440	0.135	0.9660		0.43	390			
4174	0.600	0.475	0.205	1.1760		0.52	255			
4175	0.625	0.485	0.150	1.0945		0.53	310			
4176	0.710	0.555	0.195	1.9485		0.94	155			
[3995	rows x 5	columns]	Vi	scera weight	Shell	weight	age	Sex_F	Sex_I	Sex_M
[3995 0		columns] 0.1010		scera weight 00 16.5	Shell 0	weight 0	age	Sex_F	Sex_I	Sex_M
_		_		00 16.5		_	_	Sex_F	Sex_I	Sex_M
0		0.1010	0.15	00 16.5 00 8.5	0	0	1	Sex_F	Sex_I	Sex_M
0 1		0.1010 0.0485	0.15 0.07	00 16.5 00 8.5 00 10.5	0 0	0	1 1	Sex_F	Sex_I	Sex_M
0 1 2		0.1010 0.0485 0.1415	0.15 0.07 0.21	00 16.5 00 8.5 00 10.5 50 11.5	0 0 1	0 0 0	1 1 0	Sex_F	Sex_I	Sex_M
0 1 2 3		0.1010 0.0485 0.1415 0.1140	0.15 0.07 0.21 0.15 0.05	00 16.5 00 8.5 00 10.5 50 11.5	0 0 1 0	0 0 0 0	1 1 0 1	Sex_F	Sex_I	Sex_M
0 1 2 3 4		0.1010 0.0485 0.1415 0.1140 0.0395	0.15 0.07 0.21 0.15 0.05	00 16.5 00 8.5 00 10.5 50 11.5 50 8.5	0 0 1 0	0 0 0 0 1	1 1 0 1	Sex_F	Sex_I	Sex_M
0 1 2 3 4 		0.1010 0.0485 0.1415 0.1140 0.0395 	0.15 0.07 0.21 0.15 0.05	00 16.5 00 8.5 00 10.5 50 11.5 50 8.5 	0 0 1 0 0	0 0 0 0 1	1 0 1 0 	Sex_F	Sex_I	Sex_M
0 1 2 3 4		0.1010 0.0485 0.1415 0.1140 0.0395  0.2390 0.2145	0.15 0.07 0.21 0.15 0.05 0.24 0.26	00 16.5 00 8.5 00 10.5 50 11.5 50 8.5  90 12.5 05 11.5	0 0 1 0 0 	0 0 0 0 1	1 0 1 0 	Sex_F	Sex_I	Sex_M
0 1 2 3 4  4172 4173 4174		0.1010 0.0485 0.1415 0.1140 0.0395  0.2390 0.2145 0.2875	0.15 0.07 0.21 0.15 0.05 0.24 0.26 0.30	00 16.5 00 8.5 00 10.5 50 11.5 50 8.5  90 12.5 05 11.5 80 10.5	0 0 1 0 0 	0 0 0 0 1  0	1 0 1 0  0 1 1	Sex_F	Sex_I	Sex_M
0 1 2 3 4  4172 4173		0.1010 0.0485 0.1415 0.1140 0.0395  0.2390 0.2145	0.15 0.07 0.21 0.15 0.05 0.24 0.26	00 16.5 00 8.5 00 10.5 50 11.5 50 8.5  90 12.5 05 11.5 80 10.5 60 11.5	0 0 1 0 0 	0 0 0 0 1	1 0 1 0 	Sex_F	Sex_I	Sex_M

[3995 rows x 6 columns]

```
from sklearn import preprocessing
min_max_scaler = preprocessing.MinMaxScaler(feature_range =(0, 1))
new_y= min_max_scaler.fit_transform(x,y)
print ("\n VALUES AFTER MIN MAX SCALING: \n\n", new_y)
```

#### VALUES AFTER MIN MAX SCALING:

## 10. Split the data into training and testing

```
#Split the data into Training and Testing
X = df.drop('age', axis = 1)
y = df['age']
```

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