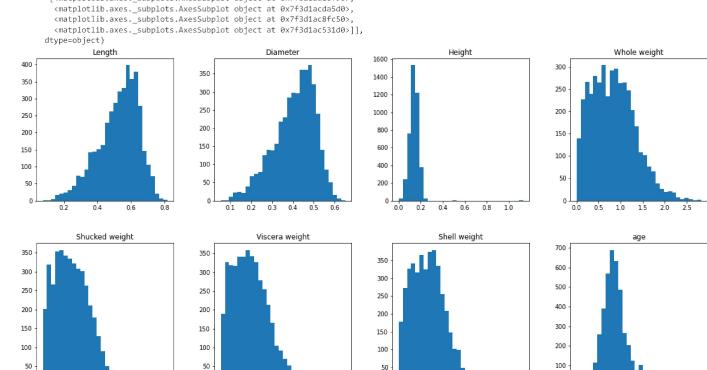
```
import pandas as pd
import numpy as np
from matplotlib import pyplot as plt
import seaborn as sns
from sklearn.linear_model import LinearRegression
df=pd.read_csv("/content/drive/MyDrive/Colab Notebooks/abalone.csv")
df['age'] = df['Rings']+1.5
df = df.drop('Rings', axis = 1)
Univariate Analysis
df.hist(figsize=(20,10), grid=False, layout=(2, 4), bins = 30)
array([[<matplotlib.axes._subplots.AxesSubplot object at 0x7f3d1b0fb690>,
             <matplotlib.axes._subplots.AxesSubplot object at 0x7f3d1ade4d90>,
             <matplotlib.axes._subplots.AxesSubplot object at 0x7f3d1adaa390>,
             <matplotlib.axes._subplots.AxesSubplot object at 0x7f3d1ad60990>],
            [<matplotlib.axes._subplots.AxesSubplot object at 0x7f3d1ad16f90>,
             <matplotlib.axes._subplots.AxesSubplot object at 0x7f3d1acda5d0>,
             <matplotlib.axes._subplots.AxesSubplot object at 0x7f3d1ac8fc50>,
             <matplotlib.axes._subplots.AxesSubplot object at 0x7f3d1ac531d0>]],
           dtype=object)
                                                                                                                                        Whole weight
                     Length
                                                           Diameter
                                                                                                   Height
                                                                                   1600
     400
                                                                                                                           300
                                             350
                                                                                   1400
```



00 02

	Length	Diameter	Height	Whole weight	Shucked weight	Viscera weight	Shell weight	age	1
Se	x								
ı	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	

0.2

0.4

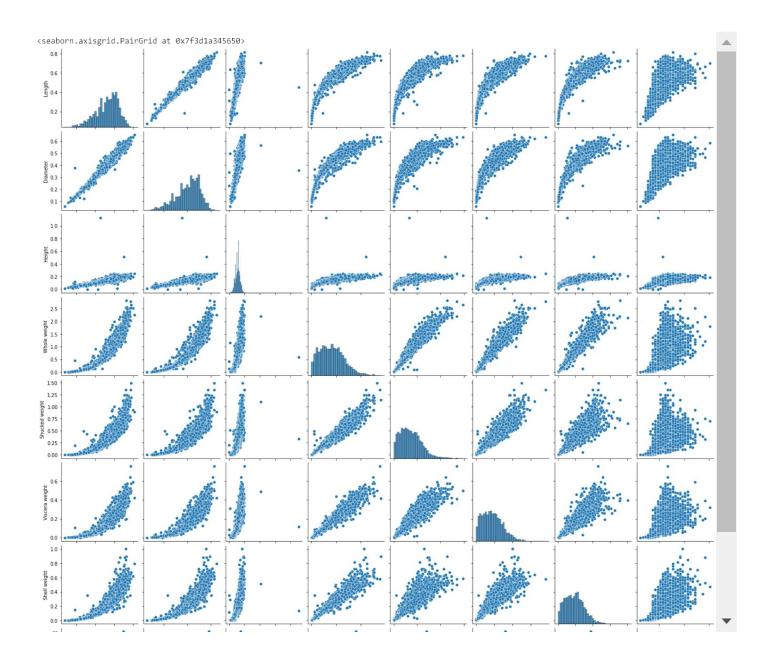
0.6

0.0

## Bivariate Analysis

0.00 0.25 0.50 0.75 1.00 1.25 1.50

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



## Descriptive statistics

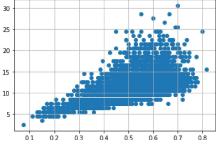
f	d	e	5	c	r	i	b	e	(	)

	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
df.drop(df[(df['Viscera weight']> 0.5) & (df['age'] < 20)].index, inplace=True)</pre>
\label{eq:df_def} $$ df.drop(df[(df['Viscera weight']<0.5) \& (df['age'] > 25)].index, inplace=True) $$
var = 'Shell weight'
plt.scatter(x = df[var], y = df['age'],)
plt.grid(True)
 #Outliers removal
 \label{eq:dfdf} $$ df.drop(df[(df['Shell weight']> 0.6) \& (df['age'] < 25)].index, inplace=True) $$
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} $$ df.drop(df[(df['Shucked weight']<1) & (df['age'] > 20)].index, inplace=True) $$
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) &
          (df['age'] < 5)].index, inplace = True)</pre>
df.drop(df[(df['Diameter']<0.6) & (
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>
```



```
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: <a href="https://numpy.org/devdocs/release/1.20.0-notes.html#deprecations">https://numpy.org/devdocs/release/1.20.0-notes.html#deprecations</a>

#### numerical\_features

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

## 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] x

	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	М	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 rc	ws × !	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

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	I	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