import pandas as pd In [2]: import numpy as np import seaborn as sns import matplotlib.pyplot as plt df = pd.read csv('abalone.csv') In [3]: df.describe() In [4]: Height Whole weight Shucked weight Viscera weight Shell weight Out[4]: Length Diameter Rings **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 9.933684 std 0.120093 0.099240 0.041827 0.490389 0.221963 0.109614 0.139203 3.224169 0.075000 0.055000 0.000000 0.002000 0.001000 0.000500 1.000000 min 0.001500 25% 0.450000 0.350000 0.115000 0.441500 0.186000 0.093500 0.130000 8.000000 50% 0.545000 0.425000 0.140000 0.799500 0.336000 0.171000 0.234000 9.000000 75% 0.615000 0.480000 0.165000 1.153000 0.502000 0.253000 0.329000 11.000000 0.650000 1.130000 2.825500 1.488000 0.760000 1.005000 29.000000 0.815000 max df.head() In [5]:

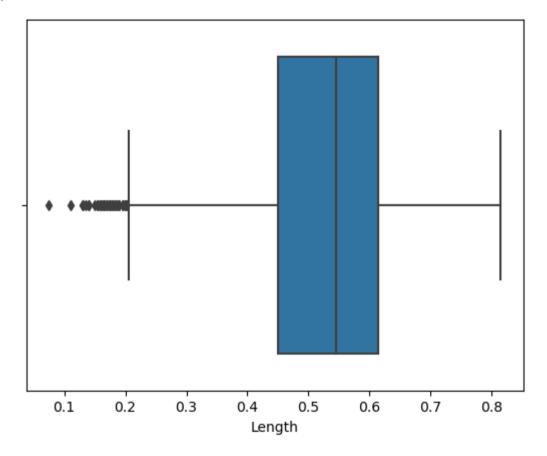
Out[5]: Sex Length Diameter Height Whole weight Shucked weight Viscera weight Shell weight Rings Μ 0 0.455 0.365 0.095 0.5140 0.2245 0.1010 0.150 15 0.350 0.265 0.090 0.2255 0.0995 0.0485 0.070 7 M 2 0.530 0.420 0.135 0.6770 0.2565 0.1415 0.210 9 Μ 0.440 0.365 0.125 0.5160 0.2155 0.1140 0.155 10 7 4 0.330 0.255 0.080 0.2050 0.0895 0.0395 0.055

In [6]: sns.boxplot(df.Length)

C:\Users\91904\anaconda3\lib\site-packages\seaborn_decorators.py:36: FutureWarning: Pass the following variable as a keyword arg: x. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation.

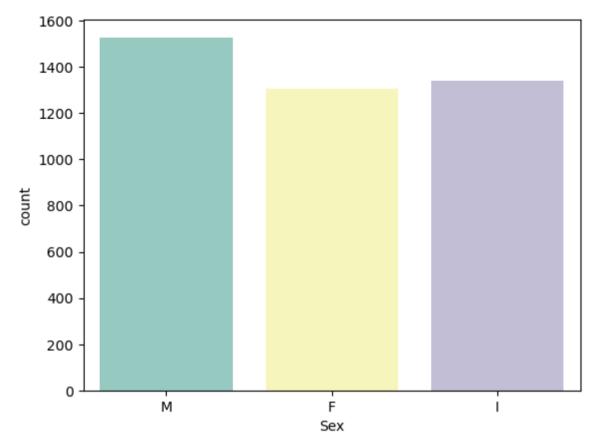
warnings.warn(

Out[6]: <AxesSubplot:xlabel='Length'>



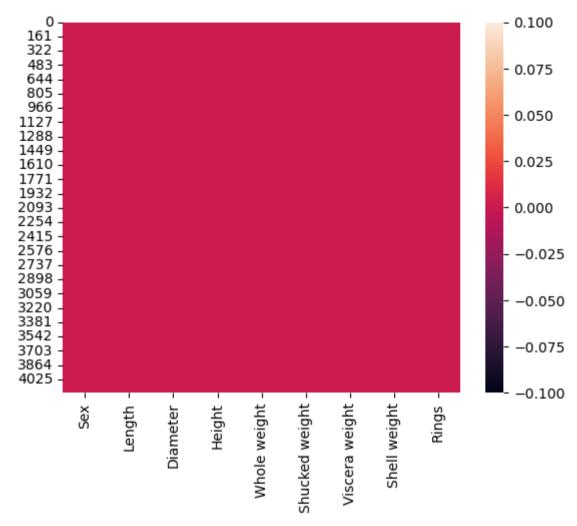
```
In [7]: sns.countplot(x = 'Sex', data = df, palette = 'Set3')
```

Out[7]. <AxesSubplot:xlabel='Sex', ylabel='count'>



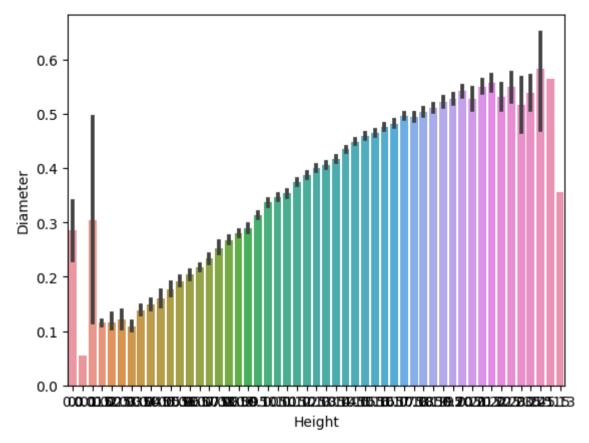
In [8]: sns.heatmap(df.isnull())

Out[8]: <AxesSubplot:>



In [9]: sns.barplot(x=df.Height,y=df.Diameter)

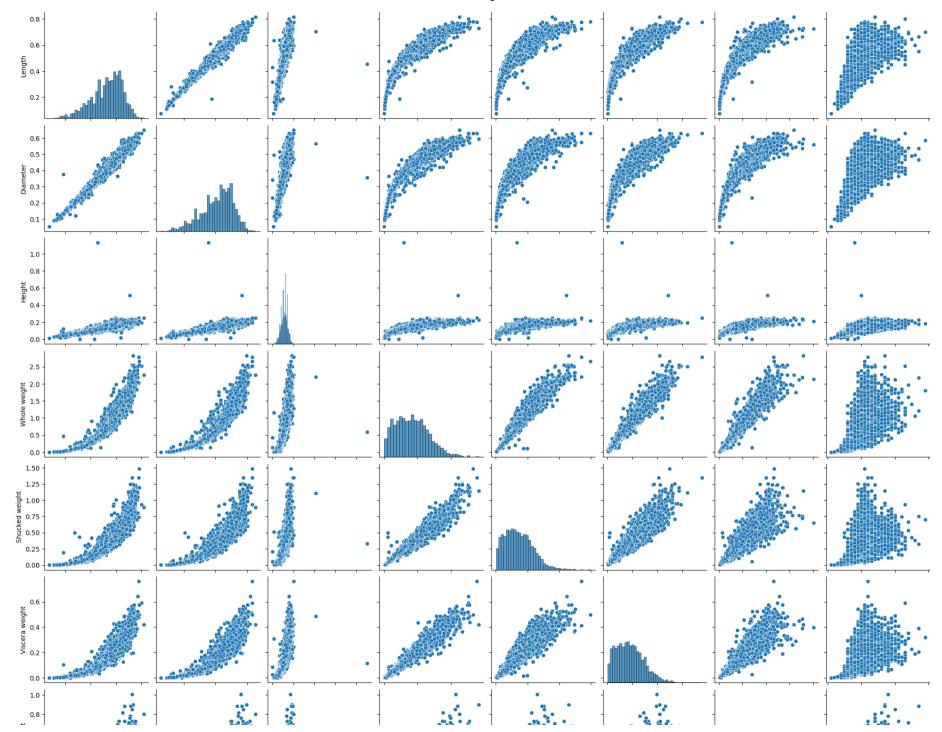
Out[9]: <AxesSubplot:xlabel='Height', ylabel='Diameter'>

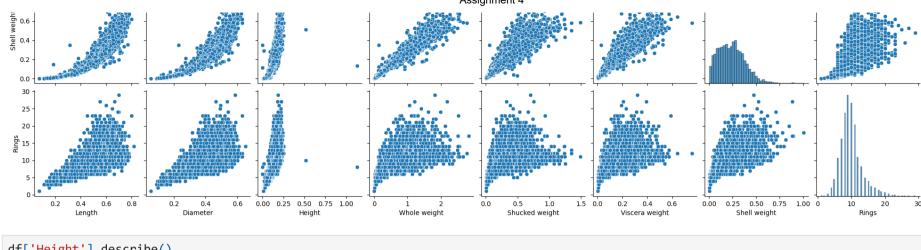




In [12]: sns.pairplot(df)

Out[12]. <seaborn.axisgrid.PairGrid at 0x1233a722280>





```
df['Height'].describe()
In [13]:
                   4177.000000
          count
Out[13]:
          mean
                      0.139516
         std
                      0.041827
                      0.000000
          min
         25%
                      0.115000
         50%
                      0.140000
         75%
                      0.165000
                      1.130000
          max
         Name: Height, dtype: float64
         df['Height'].mean()
In [14]:
         0.1395163993296614
Out[14]:
         df.max()
In [15]:
                                 Μ
         Sex
Out[15]:
         Length
                             0.815
         Diameter
                              0.65
         Height
                              1.13
         Whole weight
                            2.8255
         Shucked weight
                             1.488
         Viscera weight
                              0.76
         Shell weight
                             1.005
         Rings
                                29
         dtype: object
In [16]: df['Sex'].value_counts()
```

```
1528
Out[16]:
               1342
               1307
          Name: Sex, dtype: int64
          df[df.Height == 0]
In [17]:
                Sex Length Diameter Height Whole weight Shucked weight Viscera weight Shell weight Rings
Out[17]:
                                         0.0
                                                                                                      8
          1257
                      0.430
                                0.34
                                                    0.428
                                                                  0.2065
                                                                                0.0860
                                                                                            0.1150
          3996
                      0.315
                                0.23
                                         0.0
                                                    0.134
                                                                  0.0575
                                                                                0.0285
                                                                                            0.3505
                                                                                                      6
          df['Shucked weight'].kurtosis()
In [18]:
          0.5951236783694207
Out[18]:
          df['Diameter'].median()
In [19]:
Out[19]:
          df['Shucked weight'].skew()
In [20]:
          0.7190979217612694
Out[20]:
          df.isna().any()
In [21]:
                            False
Out[21]:
                            False
          Length
          Diameter
                            False
          Height
                            False
          Whole weight
                            False
          Shucked weight
                            False
          Viscera weight
                            False
          Shell weight
                            False
                            False
          Rings
          dtype: bool
         missing values = df.isnull().sum().sort values(ascending = False)
In [22]:
          percentage_missing_values = (missing_values/len(df))*100
          pd.concat([missing_values, percentage_missing_values], axis = 1, keys= ['Missing_values', '% Missing'])
```

.Z, 1Z.14 AIVI			
Out[22]:		Missing values	% Missing
	Sex	0	0.0
	Length	0	0.0
	Diameter	0	0.0
	Height	0	0.0
	Whole weight	0	0.0
	Shucked weight	0	0.0
	Viscera weight	0	0.0
	Shell weight	0	0.0
	Rings	0	0.0

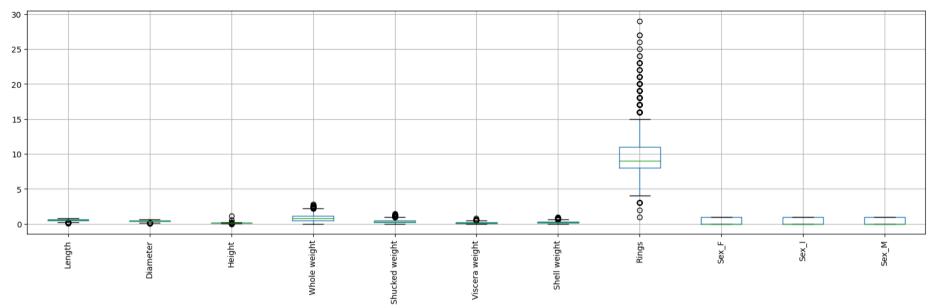
```
In [23]: q1=df.Rings.quantile(0.25)
q2=df.Rings.quantile(0.75)
iqr=q2-q1
```

```
In [24]: print(iqr)
```

3.0

```
In [25]: df = pd.get_dummies(df)
dummy_df = df
df.boxplot( rot = 90, figsize=(20,5))
```

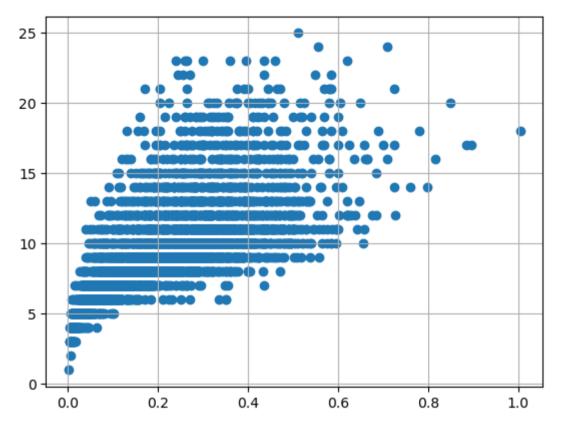
Out[25]: <AxesSubplot:>



```
In [26]: df['age'] = df['Rings']
    df = df.drop('Rings', axis = 1)

In [27]: 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)

In [28]: var = 'Shell weight'
    plt.scatter(x = df[var], y = df['age'])
    plt.grid(True)
```



numerical features = df.select dtypes(include = [np.number]).columns

```
categorical_features = df.select_dtypes(include = [np.object]).columns

C:\Users\91904\AppData\Local\Temp\ipykernel_15636\3796453440.py:2: DeprecationWarning: `np.object` is a deprecated alias for the builtin `object`. To silence this warning, use `object` by itself. Doing this will not modify any behavior and is safe. Deprecated in NumPy 1.20; for more details and guidance: https://numpy.org/devdocs/release/1.20.0-notes.html#deprecations categorical_features = df.select_dtypes(include = [np.object]).columns

In [30]: numerical_features 
    categorical_features

Out[30]: Index([], dtype='object')

In [31]: abalone_numeric = df[['Length', 'Diameter', 'Height', 'Whole weight', 'Shucked weight', 'Viscera weight', 'Shell weight', 'age',

In [32]: abalone_numeric.head()
```

In [29]:

```
Out[32]:
             Length Diameter Height Whole weight Shucked weight Viscera weight Shell weight age Sex_F Sex_I Sex_M
                               0.095
                                                                         0.1010
                                                                                             15
          0
              0.455
                        0.365
                                            0.5140
                                                           0.2245
                                                                                      0.150
                                                                                                     0
                                                                                                           0
              0.350
                        0.265
                               0.090
                                            0.2255
                                                           0.0995
                                                                         0.0485
                                                                                      0.070
                                                                                               7
                                                                                                     0
                                                                                                           0
          2
              0.530
                        0.420
                               0.135
                                            0.6770
                                                           0.2565
                                                                         0.1415
                                                                                      0.210
                                                                                               9
                                                                                                           0
                                                                                                                   0
                                                                                                     1
              0.440
                        0.365
                               0.125
                                            0.5160
                                                           0.2155
                                                                         0.1140
                                                                                      0.155
                                                                                             10
                                                                                                           0
                                                                                                                  1
          3
                        0.255
                                            0.2050
          4
              0.330
                               0.080
                                                           0.0895
                                                                         0.0395
                                                                                      0.055
                                                                                               7
                                                                                                     0
                                                                                                           1
                                                                                                                   0
In [33]: x = df.iloc[:, 0:1].values
In [34]: y = df.iloc[:, 1]
In [35]: y
                  0.365
Out[35]:
                  0.265
                  0.420
          2
                  0.365
          3
          4
                  0.255
                  . . .
          4172
                  0.450
                  0.440
          4173
          4174
                  0.475
                  0.485
          4175
          4176
                  0.555
          Name: Diameter, Length: 4150, dtype: float64
          print ("\n ORIGINAL VALUES: \n\n", x,y)
In [36]:
```

```
ORIGINAL VALUES:
          [[0.455]
          [0.35]
          [0.53]
          . . .
          [0.6]
          [0.625]
          [0.71]]0
                           0.365
                 0.265
         2
                 0.420
         3
                 0.365
                 0.255
                 . . .
         4172
                 0.450
         4173
                 0.440
         4174
                 0.475
         4175
                 0.485
         4176
                 0.555
         Name: Diameter, Length: 4150, dtype: float64
In [37]: 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:
          [[0.51351351]
          [0.37162162]
          [0.61486486]
          [0.70945946]
          [0.74324324]
          [0.85810811]]
In [38]: X = df.drop('age', axis = 1)
         y = df['age']
In [39]: from sklearn.preprocessing import StandardScaler
         from sklearn.model selection import train test split, cross val score
         from sklearn.feature_selection import SelectKBest
         standardScale = StandardScaler()
         standardScale.fit transform(X)
```

```
selectkBest = SelectKBest()
         X new = selectkBest.fit transform(X, y)
         X train, X test, y train, y test = train test split(X new, y, test size = 0.25)
         X train
         array([[0.47 , 0.37 , 0.18 , ..., 0. , 0. , 1.
Out[39]:
                [0.535, 0.435, 0.15, ..., 0., 0., 1.],
                [0.49, 0.4, 0.135, ..., 0., 1., 0.
                . . . ,
                [0.545, 0.4 , 0.13 , ..., 0. , 1. , 0.
                [0.59, 0.475, 0.155, \ldots, 1., 0., 0.]
                [0.55, 0.425, 0.15, ..., 0., 1., 0.]])
In [40]: y_train
         792
                  9
Out[40]:
         1304
                  9
         1577
                  8
         2772
                 10
         507
                 15
                 . .
         3494
                  9
                  9
         1295
         1608
                  9
         3283
                 11
         581
                 14
         Name: age, Length: 3112, dtype: int64
In [41]: from sklearn import linear model as lm
         from sklearn.linear model import LinearRegression
         model=lm.LinearRegression()
         results=model.fit(X train,y train)
In [42]: accuracy = model.score(X train, y train)
         print('Accuracy of the model:', accuracy)
         Accuracy of the model: 0.5284655210389322
In [43]: lm = LinearRegression()
         lm.fit(X_train, y_train)
         y train pred = lm.predict(X train)
         y train pred
```

```
array([10.552379 , 9.87098986, 8.41911783, ..., 8.34601774,
                11.92987898, 9.24299279])
In [44]: X_train
Out[44]: array([[0.47 , 0.37 , 0.18 , ..., 0. , 0. , 1. ],
               [0.535, 0.435, 0.15, ..., 0., 0., 1.],
               [0.49, 0.4, 0.135, ..., 0., 1., 0.
                . . . ,
                [0.545, 0.4 , 0.13 , ..., 0. , 1. , 0.
                [0.59, 0.475, 0.155, \ldots, 1., 0., 0.]
               [0.55, 0.425, 0.15, ..., 0. , 1. , 0. ]])
In [45]: y train
                  9
         792
Out[45]:
         1304
                  9
         1577
                  8
         2772
                 10
         507
                 15
                 . .
         3494
         1295
         1608
                  9
         3283
                11
                 14
         581
         Name: age, Length: 3112, dtype: int64
In [46]: from sklearn.metrics import mean absolute error, mean squared error
         s = mean squared error(y train, y train pred)
         print('Mean Squared error :%2f'%s)
         Mean Squared error :4.800536
In [47]: y train pred = lm.predict(X train)
         y test pred = lm.predict(X test)
In [48]: y test pred
        array([13.22005956, 12.63289321, 10.55556642, ..., 8.8203319,
Out[48]:
                 9.93986695, 4.99254467])
In [49]: X_test
```

```
Out[49]: array([[0.595, 0.5 , 0.18 , ..., 1. , 0. , 0.
                [0.53, 0.455, 0.165, ..., 1. , 0. , 0.
                [0.655, 0.515, 0.145, ..., 0. , 1.
                [0.54, 0.43, 0.14, ..., 0., 1., 0.
                [0.625, 0.485, 0.16, ..., 0., 1., 0.
                [0.185, 0.135, 0.045, ..., 0. , 1. , 0.
In [50]: y test
         3895
                 13
Out[50]:
         770
                 11
         2405
                 15
         2298
                 7
         2768
                 11
         1185
                  9
         2212
                 13
         1603
                  9
         3687
                 11
         3994
                  4
         Name: age, Length: 1038, dtype: int64
In [51]: p = mean squared error(y test, y test pred)
         print('Mean Squared error of testing set :%2f'%p)
         Mean Squared error of testing set :4.431685
In [52]: from sklearn.metrics import r2_score
         s = r2 score(y train, y train pred)
         print('R2 Score of training set:%.2f'%s)
         R2 Score of training set:0.53
In [53]: from sklearn.metrics import r2_score
         p = r2 score(y test, y test pred)
         print('R2 Score of testing set:%.2f'%p)
         R2 Score of testing set:0.56
 In [ ]:
```