

```
In [3]: # Import Libraries
import pandas as pd
import numpy as np
#data visualization
import seaborn as sns
import matplotlib.pyplot as plt
%matplotlib inline
```

```
In [6]: #Import the dataset
import os
os.chdir("C:/Users/ELCOT/Desktop/assignment 4")
```

```
In [7]: #1.Load the dataset into the tool
#add target(age) to dataset [rings+1.5=age]
data=pd.read_csv('abalone.csv')
data['age']=data.Rings+1.5
#remove rings variable
data.drop('Rings',axis=1,inplace=True)
print("Data loaded successfully!")
```

Data loaded successfully!

```
In [8]: df=pd.read_csv('abalone.csv')
```

```
In [9]: df
```

Out[9]:

	Sex	Length	Diameter	Height	Whole weight	Shucked weight	Viscera weight	Shell weight	Rings
0	M	0.455	0.365	0.095	0.5140	0.2245	0.1010	0.1500	15
1	M	0.350	0.265	0.090	0.2255	0.0995	0.0485	0.0700	7
2	F	0.530	0.420	0.135	0.6770	0.2565	0.1415	0.2100	9
3	M	0.440	0.365	0.125	0.5160	0.2155	0.1140	0.1550	10
4	I	0.330	0.255	0.080	0.2050	0.0895	0.0395	0.0550	7
...
4172	F	0.565	0.450	0.165	0.8870	0.3700	0.2390	0.2490	11
4173	M	0.590	0.440	0.135	0.9660	0.4390	0.2145	0.2605	10
4174	M	0.600	0.475	0.205	1.1760	0.5255	0.2875	0.3080	9
4175	F	0.625	0.485	0.150	1.0945	0.5310	0.2610	0.2960	10
4176	M	0.710	0.555	0.195	1.9485	0.9455	0.3765	0.4950	12

4177 rows × 9 columns

```
In [10]: 5.#check for missing values in the dataset and deal with them
df.isnull().sum()
```

```
Out[10]: Sex          0
Length          0
Diameter        0
Height          0
Whole weight    0
Shucked weight  0
Viscera weight  0
Shell weight    0
Rings          0
dtype: int64
```

```
In [11]: data.describe()
```

```
Out[11]:
```

	Length	Diameter	Height	Whole weight	Shucked weight	Viscera weight	Shell weight
count	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	0.238837
std	0.120093	0.099240	0.041827	0.490389	0.221963	0.109614	0.139203
min	0.075000	0.055000	0.000000	0.002000	0.001000	0.000500	0.001500
25%	0.450000	0.350000	0.115000	0.441500	0.186000	0.093500	0.130000
50%	0.545000	0.425000	0.140000	0.799500	0.336000	0.171000	0.234000
75%	0.615000	0.480000	0.165000	1.153000	0.502000	0.253000	0.329000
max	0.815000	0.650000	1.130000	2.825500	1.488000	0.760000	1.005000

```
In [12]: data['age'].isnull().sum()
```

```
Out[12]: 0
```

```
In [13]: data['age'].mean()
```

```
Out[13]: 11.433684462532918
```

```
In [14]: data['age'].replace(np.NaN , data['age'].mean()).head(15)
```

```
Out[14]: 0    16.5
1     8.5
2    10.5
3    11.5
4     8.5
5     9.5
6    21.5
7    17.5
8    10.5
9    20.5
10   15.5
11   11.5
12   12.5
13   11.5
14   11.5
Name: age, dtype: float64
```

```
In [15]: data['age'].median()
```

```
Out[15]: 10.5
```

```
In [16]: data['age'].mode()
```

```
Out[16]: 0    10.5
         Name: age, dtype: float64
```

```
In [21]: # 7.Check for categorical columns and perform encoding
```

```
In [22]: #preprocess our categorical data from words to number to make it easier for the con
         #understand
```

```
In [23]: from sklearn.preprocessing import OneHotEncoder
```

```
In [24]: encoder = OneHotEncoder(sparse=False)
         cat_cols = ['sex']
```

```
In [25]: from sklearn.preprocessing import StandardScaler
         # copying original dataframe
         df_ready = df.copy()
```

```
In [28]: scaler = StandardScaler()
         num_cols = ['Rings', 'Shell weight', 'Viscera weight', 'Shucked weight', 'Whole we:
```

```
In [29]: df_ready.head()
```

```
Out[29]:
```

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

```
In [30]: from sklearn.preprocessing import OneHotEncoder
```

```
In [31]: encoder = OneHotEncoder(sparse=False)
         cat_cols = ['Sex']
```

```
In [39]: # Encode Categorical Data
         df_encoded = pd.DataFrame(encoder.fit_transform(df_ready[cat_cols]))
         df_encoded.columns = encoder.get_feature_names(cat_cols)
```

```

-----
KeyError                                Traceback (most recent call last)
Input In [39], in <cell line: 2>()
      1 # Encode Categorical Data
----> 2 df_encoded = pd.DataFrame(encoder.fit_transform(df_ready[cat_cols]))
      3 df_encoded.columns = encoder.get_feature_names(cat_cols)

File C:\Program Files\AMD\Anaconda\lib\site-packages\pandas\core\frame.py:3511, in
DataFrame.__getitem__(self, key)
    3509     if is_iterator(key):
    3510         key = list(key)
-> 3511     indexer = self.columns._get_indexer_strict(key, "columns")[1]
    3513 # take() does not accept boolean indexers
    3514 if getattr(indexer, "dtype", None) == bool:

File C:\Program Files\AMD\Anaconda\lib\site-packages\pandas\core\indexes\base.py:5
782, in Index._get_indexer_strict(self, key, axis_name)
    5779 else:
    5780     keyarr, indexer, new_indexer = self._reindex_non_unique(keyarr)
-> 5782 self._raise_if_missing(keyarr, indexer, axis_name)
    5784 keyarr = self.take(indexer)
    5785 if isinstance(key, Index):
    5786     # GH 42790 - Preserve name from an Index

File C:\Program Files\AMD\Anaconda\lib\site-packages\pandas\core\indexes\base.py:5
842, in Index._raise_if_missing(self, key, indexer, axis_name)
    5840     if use_interval_msg:
    5841         key = list(key)
-> 5842     raise KeyError(f"None of [{key}] are in the [{axis_name}]")
    5844 not_found = list(ensure_index(key)[missing_mask.nonzero()[0]].unique())
    5845 raise KeyError(f"{not_found} not in index")

KeyError: "None of [Index(['Sex'], dtype='object')] are in the [columns]"

```

```

In [120...] # Replace Categorical Data with Encoded Data
df_ready = df_ready.drop(cat_cols, axis=1)
df_ready = pd.concat([df_encoded, df_ready], axis=1)

```

```

-----
KeyError                                Traceback (most recent call last)
Input In [120], in <cell line: 2>()
      1 # Replace Categorical Data with Encoded Data
----> 2 df_ready = df_ready.drop(cat_cols ,axis=1)
      3 df_ready = pd.concat([df_encoded, df_ready], axis=1)

File C:\Program Files\AMD\Anaconda\lib\site-packages\pandas\util\_decorators.py:31
1, in deprecate_nonkeyword_arguments.<locals>.decorate.<locals>.wrapper(*args, **k
wargs)
    305 if len(args) > num_allow_args:
    306     warnings.warn(
    307         msg.format(arguments=arguments),
    308         FutureWarning,
    309         stacklevel=stacklevel,
    310     )
--> 311 return func(*args, **kwargs)

File C:\Program Files\AMD\Anaconda\lib\site-packages\pandas\core\frame.py:4954, in
DataFrame.drop(self, labels, axis, index, columns, level, inplace, errors)
    4806 @deprecate_nonkeyword_arguments(version=None, allowed_args=["self", "label
s"])
    4807 def drop(
    4808     self,
    4809     (...)
    4815     errors: str = "raise",
    4816 ):
    4817     """
    4818     Drop specified labels from rows or columns.
    4819
    4820     (...)
    4952             weight  1.0      0.8
    4953     """
-> 4954     return super().drop(
    4955         labels=labels,
    4956         axis=axis,
    4957         index=index,
    4958         columns=columns,
    4959         level=level,
    4960         inplace=inplace,
    4961         errors=errors,
    4962     )

File C:\Program Files\AMD\Anaconda\lib\site-packages\pandas\core\generic.py:4267,
in NDFrame.drop(self, labels, axis, index, columns, level, inplace, errors)
    4265 for axis, labels in axes.items():
    4266     if labels is not None:
-> 4267         obj = obj._drop_axis(labels, axis, level=level, errors=errors)
    4269 if inplace:
    4270     self._update_inplace(obj)

File C:\Program Files\AMD\Anaconda\lib\site-packages\pandas\core\generic.py:4311,
in NDFrame._drop_axis(self, labels, axis, level, errors, consolidate, only_slice)
    4309     new_axis = axis.drop(labels, level=level, errors=errors)
    4310     else:
-> 4311         new_axis = axis.drop(labels, errors=errors)
    4312     indexer = axis.get_indexer(new_axis)
    4314 # Case for non-unique axis
    4315 else:

File C:\Program Files\AMD\Anaconda\lib\site-packages\pandas\core\indexes\base.py:6
644, in Index.drop(self, labels, errors)
    6642 if mask.any():
    6643     if errors != "ignore":

```

```
-> 6644         raise KeyError(f"{list(labels[mask])} not found in axis")
    6645         indexer = indexer[~mask]
    6646     return self.delete(indexer)
```

KeyError: "['Sex'] not found in axis"

```
In [42]: df_ready['Rings'] = df_ready['Rings'].apply(lambda x: 1 if x == 'yes' else 0)
```

```
In [43]: print('Shape of dataframe:', df_ready.shape)
```

Shape of dataframe: (4177, 8)

```
In [44]: df_ready.head()
```

```
Out[44]:
```

	Length	Diameter	Height	Whole weight	Shucked weight	Viscera weight	Shell weight	Rings
0	0.455	0.365	0.095	0.5140	0.2245	0.1010	0.150	0
1	0.350	0.265	0.090	0.2255	0.0995	0.0485	0.070	0
2	0.530	0.420	0.135	0.6770	0.2565	0.1415	0.210	0
3	0.440	0.365	0.125	0.5160	0.2155	0.1140	0.155	0
4	0.330	0.255	0.080	0.2050	0.0895	0.0395	0.055	0

```
In [45]: # 10.split the data into training and testing
         # 12.train the model
         # 13.test the model
```

```
In [46]: #Split Dataset for Training and Testing
         # Select Features
         feature = df_ready.drop('Rings', axis=1)
```

```
In [47]: # Select Target
         target = df_ready['Rings']
```

```
In [48]: # Set Training and Testing Data
         from sklearn.model_selection import train_test_split
         X_train, X_test, y_train, y_test = train_test_split(feature , target,
         shuffle = True,
         test_size=0.2,
         random_state=1)
```

```
In [49]: # Show the Training and Testing Data
         print('Shape of training feature:', X_train.shape)
         print('Shape of testing feature:', X_test.shape)
         print('Shape of training label:', y_train.shape)
         print('Shape of testing label:', y_test.shape)
```

Shape of training feature: (3341, 7)
 Shape of testing feature: (836, 7)
 Shape of training label: (3341,)
 Shape of testing label: (836,)

```
In [50]: X_train
```

Out[50]:

	Length	Diameter	Height	Whole weight	Shucked weight	Viscera weight	Shell weight
666	0.455	0.350	0.120	0.4835	0.1815	0.1440	0.1600
2813	0.255	0.195	0.055	0.0725	0.0285	0.0170	0.0210
1862	0.520	0.410	0.110	0.5185	0.2165	0.0915	0.1840
3684	0.620	0.470	0.155	0.9660	0.4470	0.1710	0.2840
551	0.615	0.490	0.155	0.9885	0.4145	0.1950	0.3450
...
2895	0.540	0.415	0.110	0.6190	0.2755	0.1500	0.1765
2763	0.550	0.425	0.135	0.6560	0.2570	0.1700	0.2030
905	0.320	0.240	0.090	0.1575	0.0700	0.0265	0.0425
3980	0.525	0.410	0.115	0.7745	0.4160	0.1630	0.1800
235	0.295	0.225	0.080	0.1240	0.0485	0.0320	0.0400

3341 rows × 7 columns

In [51]:

y_train

Out[51]:

666 0
2813 0
1862 0
3684 0
551 0
..
2895 0
2763 0
905 0
3980 0
235 0
Name: Rings, Length: 3341, dtype: int64

In [52]:

X_train.shape

Out[52]:

(3341, 7)

In [53]:

y_train.shape

Out[53]:

(3341,)

In [54]:

X_train = X_train.values.reshape((-1,1))

In [55]:

X_train

Out[55]:

array([[0.455],
[0.35],
[0.12],
...,
[0.0485],
[0.032],
[0.04]])

In [56]:

y_train

```
Out[56]: 666      0
          2813    0
          1862    0
          3684    0
          551     0
          ..
          2895    0
          2763    0
          905     0
          3980    0
          235     0
          Name: Rings, Length: 3341, dtype: int64
```

```
In [57]: X_test
```

```
Out[57]:
```

	Length	Diameter	Height	Whole weight	Shucked weight	Viscera weight	Shell weight
17	0.440	0.340	0.100	0.4510	0.1880	0.0870	0.1300
1131	0.565	0.435	0.150	0.9900	0.5795	0.1825	0.2060
299	0.370	0.280	0.105	0.2340	0.0905	0.0585	0.0750
1338	0.580	0.455	0.135	0.7955	0.4050	0.1670	0.2040
2383	0.525	0.390	0.135	0.6005	0.2265	0.1310	0.2100
...
1787	0.545	0.420	0.165	0.8935	0.4235	0.2195	0.2280
3075	0.680	0.520	0.185	1.4940	0.6150	0.3935	0.4060
2766	0.555	0.445	0.175	1.1465	0.5510	0.2440	0.2785
1410	0.665	0.530	0.180	1.4910	0.6345	0.3420	0.4350
2529	0.600	0.500	0.155	1.3320	0.6235	0.2835	0.3500

836 rows × 7 columns

```
In [58]: y_test
```

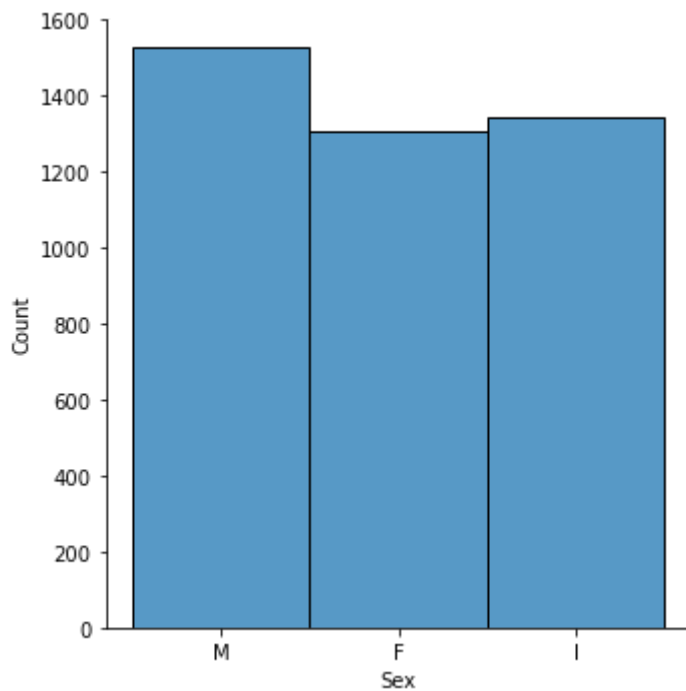
```
Out[58]: 17      0
          1131    0
          299     0
          1338    0
          2383    0
          ..
          1787    0
          3075    0
          2766    0
          1410    0
          2529    0
          Name: Rings, Length: 836, dtype: int64
```

```
In [60]: #perform Visualization
```

```
In [61]: #Univarient analysis
```

```
In [62]: sns.displot(df['Sex'])
```

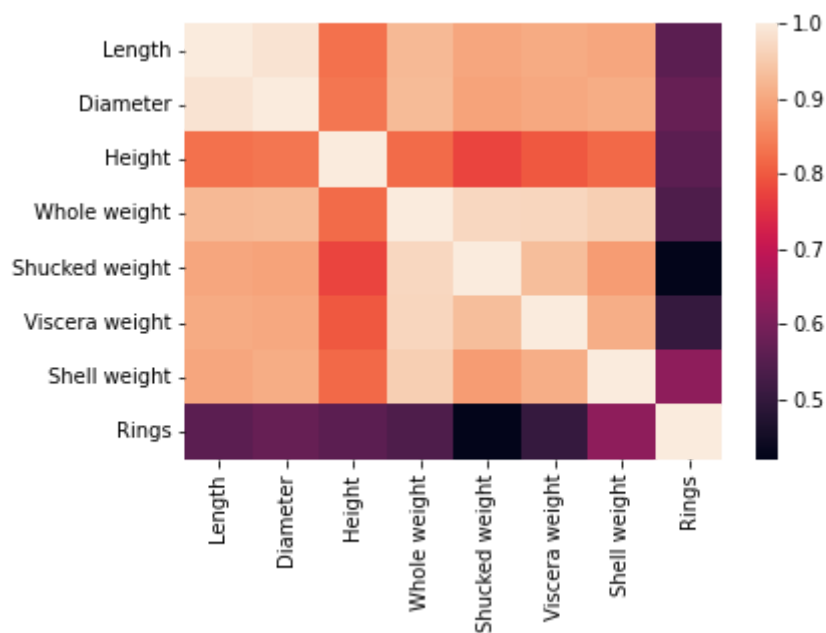
```
Out[62]: <seaborn.axisgrid.FacetGrid at 0x1479f8851c0>
```

In [63]: *#Multivariant analysis*

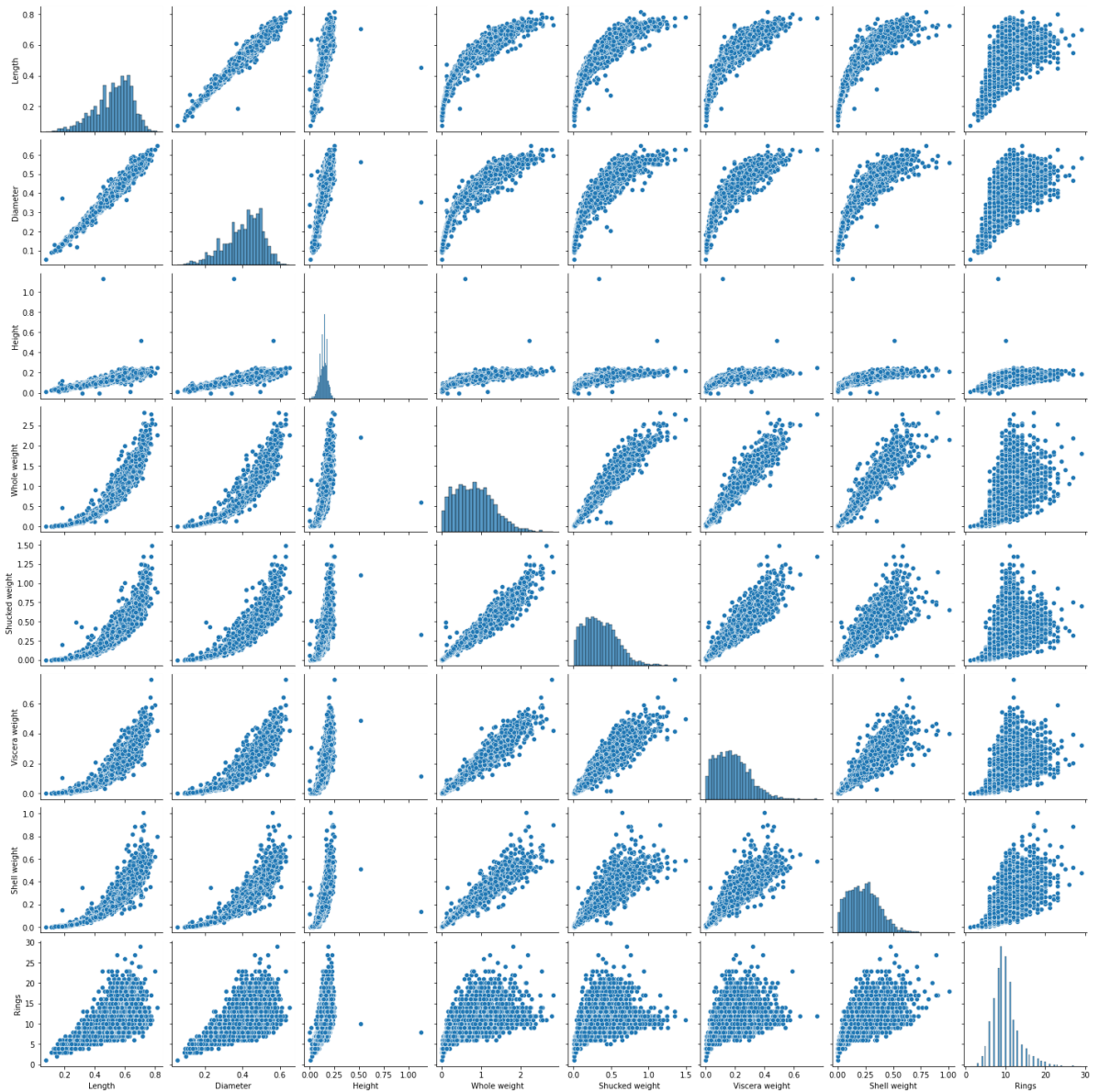
```
In [64]: corr = df.corr()
sns.heatmap(corr,xticklabels=corr.columns,yticklabels=corr.columns)
```

Out[64]: <AxesSubplot:>



```
In [71]: #Bi-variant analysis
sns.pairplot(df)
```

Out[71]: <seaborn.axisgrid.PairGrid at 0x147b3e33d30>



In [75]: `# 4.Descriptive statistics on the dataset`
`data.info()`

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 4177 entries, 0 to 4176
Data columns (total 9 columns):
#   Column      Non-Null Count  Dtype
---  ---
0    Sex         4177 non-null   object
1    Length      4177 non-null   float64
2    Diameter    4177 non-null   float64
3    Height      4177 non-null   float64
4    Whole weight 4177 non-null   float64
5    Shucked weight 4177 non-null   float64
6    Viscera weight 4177 non-null   float64
7    Shell weight 4177 non-null   float64
8    age         4177 non-null   float64
dtypes: float64(8), object(1)
memory usage: 293.8+ KB
```

In [76]: `data.describe()`

Out[76]:

	Length	Diameter	Height	Whole weight	Shucked weight	Viscera weight	Shell weight
count	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
std	0.120093	0.099240	0.041827	0.490389	0.221963	0.109614	0.139203
min	0.075000	0.055000	0.000000	0.002000	0.001000	0.000500	0.001500
25%	0.450000	0.350000	0.115000	0.441500	0.186000	0.093500	0.130000
50%	0.545000	0.425000	0.140000	0.799500	0.336000	0.171000	0.234000
75%	0.615000	0.480000	0.165000	1.153000	0.502000	0.253000	0.329000
max	0.815000	0.650000	1.130000	2.825500	1.488000	0.760000	1.005000

```
In [82]: df = pd.get_dummies(df)
dummy_df = df
```

```
In [115... import numpy as np
from collections import Counter

def detect_outliers(df, n, features):
    """
    Takes a dataframe df of features and returns a list of the indices
    corresponding to the observations containing more than n outliers according
    to the Tukey method.
    """
    outlier_indices = []

    # iterate over features(columns)
    for col in features:
        # 1st quartile (25%)
        Q1 = np.percentile(df[col], 25)
        # 3rd quartile (75%)
        Q3 = np.percentile(df[col], 75)
        # Interquartile range (IQR)
        IQR = Q3 - Q1

        # outlier step
        outlier_step = 1.5 * IQR

        # Determine a list of indices of outliers for feature col
        outlier_list_col = df[(df[col] < Q1 - outlier_step) | (df[col] > Q3 + outlier_step)]

        # append the found outlier indices for col to the list of outlier indices
        outlier_indices.extend(outlier_list_col.index)

        # select observations containing more than 2 outliers
    outlier_indices = Counter(outlier_indices)
    multiple_outliers = list(k for k, v in outlier_indices.items() if v > n)

    return multiple_outliers
```

```
In [85]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 4177 entries, 0 to 4176
Data columns (total 11 columns):
#   Column                Non-Null Count  Dtype
---  ---
0   Length                4177 non-null   float64
1   Diameter              4177 non-null   float64
2   Height               4177 non-null   float64
3   Whole weight         4177 non-null   float64
4   Shucked weight       4177 non-null   float64
5   Viscera weight       4177 non-null   float64
6   Shell weight         4177 non-null   float64
7   Rings                4177 non-null   int64
8   Sex_F                4177 non-null   uint8
9   Sex_I                4177 non-null   uint8
10  Sex_M                4177 non-null   uint8
dtypes: float64(7), int64(1), uint8(3)
memory usage: 273.4 KB
```

```
In [118... outliers=detection(df,["Length","Whole weight","Height","Diameter"])
df.loc[outliers]
```

```
-----
NameError                                Traceback (most recent call last)
Input In [118], in <cell line: 1>()
----> 1 outliers=detection(df,["Length","Whole weight","Height","Diameter"])
      2 df.loc[outliers]

NameError: name 'detection' is not defined
```

```
In [117... df=df.drop(outliers,axis=0).reset_index(drop = True)
```

```
-----
NameError                                Traceback (most recent call last)
Input In [117], in <cell line: 1>()
----> 1 df=df.drop(outliers,axis=0).reset_index(drop = True)

NameError: name 'outliers' is not defined
```

```
In [95]: df
```

Out[95]:

	Length	Diameter	Height	Whole weight	Shucked weight	Viscera weight	Shell weight	Rings	Sex_F	Sex_I	Sex_M
0	0.455	0.365	0.095	0.5140	0.2245	0.1010	0.1500	15	0	0	1
1	0.350	0.265	0.090	0.2255	0.0995	0.0485	0.0700	7	0	0	1
2	0.530	0.420	0.135	0.6770	0.2565	0.1415	0.2100	9	1	0	0
3	0.440	0.365	0.125	0.5160	0.2155	0.1140	0.1550	10	0	0	1
4	0.330	0.255	0.080	0.2050	0.0895	0.0395	0.0550	7	0	1	0
...
4172	0.565	0.450	0.165	0.8870	0.3700	0.2390	0.2490	11	1	0	0
4173	0.590	0.440	0.135	0.9660	0.4390	0.2145	0.2605	10	0	0	1
4174	0.600	0.475	0.205	1.1760	0.5255	0.2875	0.3080	9	0	0	1
4175	0.625	0.485	0.150	1.0945	0.5310	0.2610	0.2960	10	1	0	0
4176	0.710	0.555	0.195	1.9485	0.9455	0.3765	0.4950	12	0	0	1

4177 rows × 11 columns

In [96]: `# 8.split the data into dependent and independent variables`
`# 9.scale independent variable`

In [97]: `# x-independent variable & y-dependent variable`

In [98]: `x=df.iloc[:,1]`

In [99]: `x`

Out[99]:

	Length
0	0.455
1	0.350
2	0.530
3	0.440
4	0.330
...	...
4172	0.565
4173	0.590
4174	0.600
4175	0.625
4176	0.710

4177 rows × 1 columns

In [100]: `df=pd.read_csv('abalone.csv')`

```
In [101... df
y=df.iloc[:,1:]
```

```
In [103... y
```

```
Out[103]:
```

	Length	Diameter	Height	Whole weight	Shucked weight	Viscera weight	Shell weight	Rings
0	0.455	0.365	0.095	0.5140	0.2245	0.1010	0.1500	15
1	0.350	0.265	0.090	0.2255	0.0995	0.0485	0.0700	7
2	0.530	0.420	0.135	0.6770	0.2565	0.1415	0.2100	9
3	0.440	0.365	0.125	0.5160	0.2155	0.1140	0.1550	10
4	0.330	0.255	0.080	0.2050	0.0895	0.0395	0.0550	7
...
4172	0.565	0.450	0.165	0.8870	0.3700	0.2390	0.2490	11
4173	0.590	0.440	0.135	0.9660	0.4390	0.2145	0.2605	10
4174	0.600	0.475	0.205	1.1760	0.5255	0.2875	0.3080	9
4175	0.625	0.485	0.150	1.0945	0.5310	0.2610	0.2960	10
4176	0.710	0.555	0.195	1.9485	0.9455	0.3765	0.4950	12

4177 rows × 8 columns

```
In [104... # 11.Build the model
```

```
In [105... transformed_sex_feature = OneHotEncoder().fit_transform(df['Sex'].values.reshape(-1,1))
df_sex_encoded = pd.DataFrame(transformed_sex_feature, columns = ["Sex_"+str(int(i)) for i in range(2)])
df = pd.concat([df, df_sex_encoded], axis=1)
```

```
Input In [105]
df_sex_encoded = pd.DataFrame(transformed_sex_feature, columns = ["Sex_"+str(int(i)) for i in range(2)])
^
SyntaxError: invalid syntax
```

```
In [106... df.head()
```

```
Out[106]:
```

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

```
In [107... # 14. Measure the performance using Metrics
```

```
In [108... df['Age'] = df['Rings'] + 1.5
df['Age'].head(5)
```

```
Out[108]: 0    16.5
          1     8.5
          2    10.5
          3    11.5
          4     8.5
          Name: Age, dtype: float64
```

```
In [111]: '''Sex and Age Visualization'''
plt.figure(figsize = (20,7))
sns.swarmplot(x = 'Sex', y = 'Age', data = df, hue = 'Sex')
sns.violinplot(x = 'Sex', y = 'Age', data = df)
```

C:\Program Files\AMD\Anaconda\lib\site-packages\seaborn\categorical.py:1296: UserWarning: 56.2% of the points cannot be placed; you may want to decrease the size of the markers or use stripplot.

warnings.warn(msg, UserWarning)

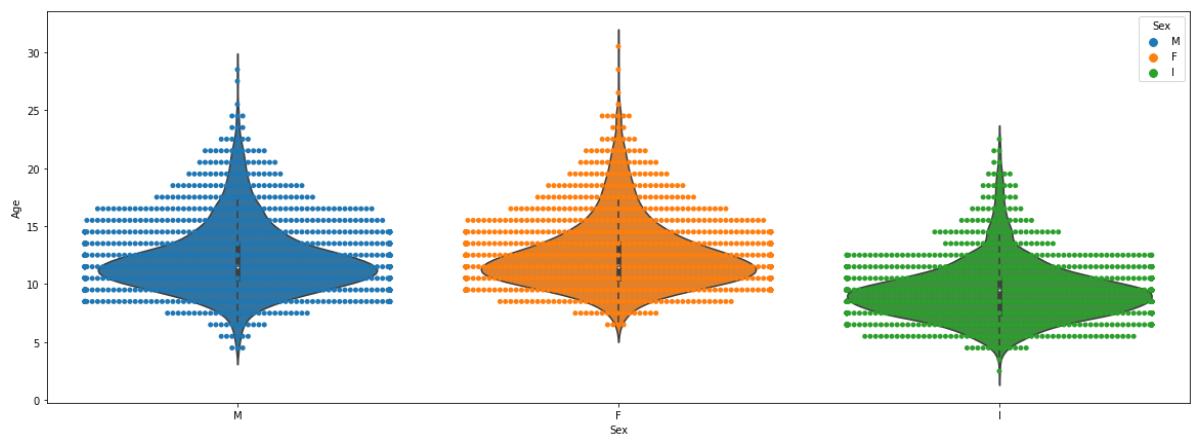
C:\Program Files\AMD\Anaconda\lib\site-packages\seaborn\categorical.py:1296: UserWarning: 52.2% of the points cannot be placed; you may want to decrease the size of the markers or use stripplot.

warnings.warn(msg, UserWarning)

C:\Program Files\AMD\Anaconda\lib\site-packages\seaborn\categorical.py:1296: UserWarning: 58.5% of the points cannot be placed; you may want to decrease the size of the markers or use stripplot.

warnings.warn(msg, UserWarning)

```
Out[111]: <AxesSubplot:xlabel='Sex', ylabel='Age'>
```



```
In [ ]:
```