```
In [1]:
            import numpy as np
            import pandas as pd
In [13]:
           df = pd.read csv("/Users/dev/Personal/DS & AI Class Notes/Data Sets/Adaboos
          col are 'Sex','Length','Whole weight','Viscera weight','Diameter','Height','Shucked
          weight','Shell weight','Rings'
In [34]:
           df.columns = ['Sex','Length','Whole weight','Viscera weight','Diameter','He
In [40]:
           df = df[['Length', 'Whole weight', 'Viscera weight', 'Diameter', 'Height',
                    'Shucked weight', 'Shell weight', 'Rings', 'Sex']]
In [41]:
           df
Out[41]:
                            Whole
                                      Viscera
                                                                  Shucked
                                                                               Shell
                 Length
                                               Diameter Height
                                                                                     Rings Sex
                            weight
                                       weight
                                                                    weight
                                                                             weight
              0
                  0.455
                             0.365
                                        0.095
                                                 0.5140 0.2245
                                                                     0.1010
                                                                             0.1500
                                                                                        15
                                                                                              M
              1
                  0.350
                             0.265
                                        0.090
                                                                             0.0700
                                                                                         7
                                                 0.2255 0.0995
                                                                    0.0485
                                                                                              M
              2
                  0.530
                             0.420
                                        0.135
                                                 0.6770 0.2565
                                                                     0.1415
                                                                             0.2100
                                                                                         9
                                                                                              F
                  0.440
                             0.365
                                        0.125
                                                 0.5160
                                                         0.2155
                                                                     0.1140
                                                                             0.1550
                                                                                        10
                                                                                              Μ
              4
                  0.330
                             0.255
                                        0.080
                                                 0.2050 0.0895
                                                                    0.0395
                                                                             0.0550
                                                                                         7
                                                                                              1
           4172
                  0.565
                             0.450
                                        0.165
                                                        0.3700
                                                                    0.2390
                                                                             0.2490
                                                                                              F
                                                 0.8870
                                                                                        11
           4173
                  0.590
                             0.440
                                        0.135
                                                 0.9660 0.4390
                                                                    0.2145
                                                                             0.2605
                                                                                        10
                                                                                              Μ
           4174
                  0.600
                             0.475
                                        0.205
                                                 1.1760 0.5255
                                                                    0.2875
                                                                             0.3080
                                                                                         9
                                                                                              Μ
           4175
                  0.625
                             0.485
                                        0.150
                                                 1.0945 0.5310
                                                                    0.2610
                                                                             0.2960
                                                                                        10
                                                                                              F
```

4177 rows × 9 columns

0.710

4176

Checking Balance of DF

0.555

```
In [43]:
df["Sex"].value_counts()
```

1.9485 0.9455

0.3765

0.4950

12

Μ

0.195

```
1528
Out[43]:
              1342
              1307
         F
         Name: Sex, dtype: int64
In [44]:
          def checkz(df):
              return df[df == 0].value_counts()
In [46]:
          for i in df.columns:
              print(checkz(df[i]))
         Series([], Name: Length, dtype: int64)
         Series([], Name: Whole weight, dtype: int64)
         Name: Viscera weight, dtype: int64
         Series([], Name: Diameter, dtype: int64)
         Series([], Name: Height, dtype: int64)
         Series([], Name: Shucked weight, dtype: int64)
         Series([], Name: Shell weight, dtype: int64)
         Series([], Name: Rings, dtype: int64)
         Series([], Name: Sex, dtype: int64)
In [47]:
          df.columns
         Index(['Length', 'Whole weight', 'Viscera weight', 'Diameter', 'Height',
Out[47]:
                 'Shucked weight', 'Shell weight', 'Rings', 'Sex'],
               dtype='object')
        Checking And Removing 0's
In [48]:
          clist = ['Length', 'Whole weight', 'Viscera weight', 'Diameter', 'Height',
                 'Shucked weight', 'Shell weight', 'Rings']
In [49]:
          def zeroremove(df):
              m = round(df.mean(), 2)
              df.replace(0,m,inplace = True)
In [50]:
          for i in clist:
              zeroremove(df[i])
In [51]:
          for i in df.columns:
              print(checkz(df[i]))
```

```
Series([], Name: Length, dtype: int64)
Series([], Name: Whole weight, dtype: int64)
Series([], Name: Viscera weight, dtype: int64)
Series([], Name: Diameter, dtype: int64)
Series([], Name: Height, dtype: int64)
Series([], Name: Shucked weight, dtype: int64)
Series([], Name: Shell weight, dtype: int64)
Series([], Name: Rings, dtype: int64)
Series([], Name: Sex, dtype: int64)
```

OD Tech With The help of Skew

```
In [52]:
          def odigr(df):
              q1 = df.quantile(0.25)
              q3 = df.quantile(0.75)
              iqr = q3 - q1
              low = q1 - (1.5 * iqr)
              high = q3 + (1.5 * iqr)
              m = df.mean()
              df = df.apply(lambda x : m if x < low else (m if x > high else x ) )
              return df
In [53]:
          def odmsd(df):
              m = round(df.mean(), 2)
              s = round(df.std(),2)
              low = round(m-(3*s), 2)
              high = round(m+(3*s),2)
              ft1 = df[df<low]</pre>
              ft2 = df[df>high]
              df = df.map(lambda x : low if x < low else (high if x > high else x ))
In [55]:
          for i in clist:
              print(f'{i} is {df[i].skew()}')
         Length is -0.639873268981801
         Whole weight is -0.6091981423290918
         Viscera weight is 3.1671068608877198
         Diameter is 0.5309585632523087
         Height is 0.7190979217612694
         Shucked weight is 0.5918521514155083
         Shell weight is 0.6209268251392077
         Rings is 1.114101898355677
In [60]:
          for i in clist:
              if df[i].skew() <= 0.5:</pre>
                   odmsd(df[i])
              else:
                   df[i] = odiqr(df[i])
```

```
In [61]:
    for i in clist:
        print(f'{i} is {df[i].skew()}')

Length is -0.47226034516491344
    Whole weight is -0.47042514918312894
    Viscera weight is -0.1854164788342902
    Diameter is 0.39235373317723926
    Height is 0.42806642403527495
    Shucked weight is 0.4557527876962757
    Shell weight is 0.36801505496676334
    Rings is 0.15139283233172263
```

Encoding of df["Sex"] with Label Encoder

```
In [63]: from sklearn.preprocessing import LabelEncoder
In [64]: le = LabelEncoder()
In [66]: le.classes_
Out[66]: array(['F', 'I', 'M'], dtype=object)
In [71]: df = pd.concat([ df, pd.DataFrame(le.fit_transform(df["Sex"]),columns=["SEXIN [72]: df
```

Out[72]:		Length	Whole weight	Viscera weight	Diameter	Height	Shucked weight	Shell weight	Rings	SEX
	0	0.455	0.365	0.095	0.5140	0.2245	0.1010	0.1500	15.0	2
	1	0.350	0.265	0.090	0.2255	0.0995	0.0485	0.0700	7.0	2
	2	0.530	0.420	0.135	0.6770	0.2565	0.1415	0.2100	9.0	0
	3	0.440	0.365	0.125	0.5160	0.2155	0.1140	0.1550	10.0	2
	4	0.330	0.255	0.080	0.2050	0.0895	0.0395	0.0550	7.0	1
	•••									
	4172	0.565	0.450	0.165	0.8870	0.3700	0.2390	0.2490	11.0	0
	4173	0.590	0.440	0.135	0.9660	0.4390	0.2145	0.2605	10.0	2
	4174	0.600	0.475	0.205	1.1760	0.5255	0.2875	0.3080	9.0	2
	4175	0.625	0.485	0.150	1.0945	0.5310	0.2610	0.2960	10.0	0
	4176	0.710	0.555	0.195	1.9485	0.9455	0.3765	0.4950	12.0	2

4177 rows × 9 columns

Splitting

```
In [73]:
          X = df.drop("SEX",axis=1)
In [74]:
          X.sample()
Out[74]:
                           Whole
                                                                Shucked
                                      Viscera
                                                                             Shell
                                              Diameter Height
                Length
                                                                                   Rings
                           weight
                                      weight
                                                                  weight
                                                                           weight
          2989
                  0.56
                            0.425
                                        0.135
                                                0.9415
                                                        0.509
                                                                   0.2015
                                                                            0.1975
                                                                                     9.0
In [75]:
           y = df["SEX"]
In [76]:
          y.sample()
          2791
                  2
Out[76]:
          Name: SEX, dtype: int64
In [77]:
          from sklearn.model_selection import train_test_split
          from sklearn.preprocessing import StandardScaler
          from sklearn.model_selection import KFold , cross_val_score
          from sklearn.svm import SVC
          from sklearn.tree import DecisionTreeClassifier
          from sklearn.ensemble import RandomForestClassifier , AdaBoostClassifier
```

```
In [78]:
          Xtrain, Xtest, ytrain, ytest = train_test_split(X, y, test_size=.30)
In [79]:
          X.shape , Xtrain.shape , Xtest.shape
          ((4177, 8), (2923, 8), (1254, 8))
Out[79]:
In [80]:
          y.shape , ytrain.shape , ytest.shape
          ((4177,), (2923,), (1254,))
Out[80]:
In [81]:
          kf = KFold(n splits=11)
In [82]:
          dct = DecisionTreeClassifier()
In [83]:
          rfc = RandomForestClassifier()
In [84]:
          algo = [ dct , rfc ]
         Without Feature Scaling
```

```
In [85]:
    for i in algo:
        i.fit(Xtrain,ytrain)
        s = i.score(Xtest,ytest)
        print(f'{i} = {s}')

DecisionTreeClassifier() = 0.48165869218500795
```

With Feature Scaling

RandomForestClassifier() = 0.562200956937799

```
In [86]: ss = StandardScaler()

In [87]: ss.fit(Xtrain)
Out[87]: StandardScaler()

In [88]: Xtrain_ss = ss.transform(Xtrain)

In [89]: Xtest_ss = ss.transform(Xtest)
```

```
In [90]:
    for i in algo:
        i.fit(Xtrain_ss,ytrain)
        s = i.score(Xtest_ss,ytest)
        print(f'{i} = {s}')

DecisionTreeClassifier() = 0.4904306220095694
```

With Cross Validation

RandomForestClassifier() = 0.5502392344497608

RandomForestClassifier() = 0.524532577546048

```
In [91]:
    for i in algo:
        s = cross_val_score(i, X, y, cv = kf)
        print(f'{i} = {s.mean()}')

DecisionTreeClassifier() = 0.4876816351264344
```

Boosting

Without Feature Scaling

AdaBoostClassifier()

```
In [99]:
    for i in algo1:
        i.fit(Xtrain,ytrain)
        s = i.score(Xtest,ytest)
        print(f'{i} = {s}')

RandomForestClassifier(max_depth=2, max_leaf_nodes=3, n_estimators=150) = 0
    .5223285486443381
```

With Feature Scaling

AdaBoostClassifier() = 0.5350877192982456

With Cross Validation (Boosting)

With GridSearch CV

```
In [102...
          from sklearn.model selection import GridSearchCV
In [108...
          dic = { 'n estimators' : [100,125,180], 'criterion': ['gini', 'entropy'],
                  ,'min samples leaf' : [ 1,5] }
In [109...
          kf1 = KFold(n splits=12)
In [110...
          gvc = GridSearchCV(RandomForestClassifier(),param_grid=dic,cv = kf1)
In [111...
          %%time
          gvc.fit(X,y)
         CPU times: user 2min 2s, sys: 754 ms, total: 2min 3s
         Wall time: 2min 3s
         GridSearchCV(cv=KFold(n_splits=12, random_state=None, shuffle=False),
Out [111...
                       estimator=RandomForestClassifier(),
                       param_grid={'criterion': ['gini', 'entropy'],
                                    'max_depth': [2, 3, 10], 'min_samples_leaf': [1, 5
          1,
                                    'n estimators': [100, 125, 180]})
In [112...
          gvc.best params
```