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0.2565	"2	F	0.530	0.420	0.135	0.6770
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0.2155	"3	M	0.440	0.365	0.125	0.5160
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0.0895	"4	I	0.330	0.255	0.080	0.2050
			0.0395	\n",		

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2.51-.94 2.06-2.06.94zm10 10l.94 2.06.94-2.06 2.06-.94-2.06-.94-.94-2.06-
.94 2.06-2.06.94z"/><path d="M17.41 7.96l-1.37-1.37c-.4-.4-.92-.59-1.43-
.59-.52 0-1.04.2-1.43.59L10.3 9.45l-7.72 7.72c-.78.78-.78 2.05 0 2.83L4
21.41c.39.39.9.59 1.41.59.51 0 1.02-.2 1.41-.59l7.78-7.78 2.81-2.81c.8-
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' +\n",
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href=https://colab.research.google.com/notebooks/data_table.ipynb>data
table notebook</a>'\n",
        "                + ' to learn more about interactive
tables.';\n",
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0.0995          1  M    0.350      0.265    0.090      0.2255
0.2565          2  F    0.530      0.420    0.135      0.6770
0.2155          3  M    0.440      0.365    0.125      0.5160
0.0895          4  I    0.330      0.255    0.080      0.2050
          0.0395  \n",
          " Shell_weight  Age  \n",
          0          0.150  16.5  \n",
          1          0.070   8.5  \n",
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```

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"    <td>I</td>\n",
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.94L8.5 2.51-.94 2.06-2.06.94zm10 10l.94 2.06.94-2.06 2.06-.94-2.06-
.94 2.06-2.06.94z\"/><path d=\"M17.41 7.96l-1.37-1.37c-.4-.4-.92-.59-1.43-
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21.41c.39.39.95 1.41.59.51 0 1.02-.2 1.41-.59l7.78-7.78 2.81-2.81c.8-
.78.8-2.07 0-2.86zM5.41 20L4 18.59l7.72-7.72 1.47 1.35L5.41 20z\"/>\n",
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```



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    "        buttonEl.style.display =\n",
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'none';\n",
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    "                                [key],
    {});\n",
    "            if (!dataTable) return;\n",
    "\n",
    "            const docLinkHtml = 'Like what you see? Visit the
' +\n",
    "                '<a target=\"_blank\"
href=https://colab.research.google.com/notebooks/data_table.ipynb>data
table notebook</a>'\n",
    "                + ' to learn more about interactive
tables.';\n",
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weight': 'Shucked_weight', 'Viscera weight': 'Viscera_weight',\n",
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        "### Histogram"
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            "metadata": {},
            "execution_count": 9
        },
        {
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            "data": {
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                    "<Figure size 432x288 with 1 Axes>"
                ],
                "image/png":          },
                "metadata": {
                    "needs_background": "light"
                }
            }
        },
    ],
    "source": [
        "sns.histplot(y=df.Age,color='green') "
    ]
},
{
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    "execution_count": 10,
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            "height": 296
        },
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        "outputId": "3975e559-6086-4fc3-de67-7e5d9e90a5da"
    },

```

```

},
"outputs": [
  {
    "output_type": "execute_result",
    "data": {
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      ]
    },
    "metadata": {},
    "execution_count": 10
  },
  {
    "output_type": "display_data",
    "data": {
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        "<Figure size 432x288 with 1 Axes>"
      ],
      "image/png":          },
    "metadata": {
      "needs_background": "light"
    }
  }
],
"source": [
  "sns.histplot(x=df.Age,color='yellow') "
]
},
{
  "cell_type": "markdown",
  "metadata": {
    "id": "ikzi_PFS2Sib"
  },
  "source": [
    "### Boxplot"
  ]
},
{
  "cell_type": "code",
  "execution_count": 11,
  "metadata": {
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      "height": 296
    },
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    "outputId": "cd3239ba-4a75-4493-99e4-b4bb1eb92764"
  },
  "outputs": [
    {
      "output_type": "execute_result",
      "data": {
        "text/plain": [
          "<matplotlib.axes._subplots.AxesSubplot at 0x7f105777ff10>"
        ]
      },
      "metadata": {},
      "execution_count": 11
    },
    {
      "output_type": "display_data",

```

```

      "data": {
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        ],
        "image/png":          },
        "metadata": {
          "needs_background": "light"
        }
      }
    ],
    "source": [
      "sns.boxplot(x=df.Age,color='orange') "
    ]
  },
  {
    "cell_type": "markdown",
    "metadata": {
      "id": "MtC8V3TW2YXt"
    },
    "source": [
      "### Countplot"
    ]
  },
  {
    "cell_type": "code",
    "execution_count": 12,
    "metadata": {
      "colab": {
        "base_uri": "https://localhost:8080/",
        "height": 296
      },
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      "outputId": "f9eb5356-0fab-4b7f-eb7c-52764efa6d84"
    },
    "outputs": [
      {
        "output_type": "execute_result",
        "data": {
          "text/plain": [
            "<matplotlib.axes._subplots.AxesSubplot at 0x7f1057717510>"
          ]
        },
        "metadata": {},
        "execution_count": 12
      },
      {
        "output_type": "display_data",
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            "<Figure size 432x288 with 1 Axes>"
          ],
          "image/png":          },
          "metadata": {
            "needs_background": "light"
          }
        }
      ]
    },
    "source": [
      "sns.countplot(x=df.Age) "
    ]
  },

```

```

{
  "cell_type": "markdown",
  "metadata": {
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  },
  "source": [
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  ]
},
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  "metadata": {
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  },
  "source": [
    "### Barplot"
  ]
},
{
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  "metadata": {
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      "height": 296
    },
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    "outputId": "06097670-ea52-4fab-e181-9abc9393ab6e"
  },
  "outputs": [
    {
      "output_type": "execute_result",
      "data": {
        "text/plain": [
          "<matplotlib.axes._subplots.AxesSubplot at 0x7f10576f5390>"
        ]
      },
      "metadata": {},
      "execution_count": 14
    },
    {
      "output_type": "display_data",
      "data": {
        "text/plain": [
          "<Figure size 432x288 with 1 Axes>"
        ],
        "image/png":

```

1. Importing the required libraries

In [1]:

```
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
import numpy as np
```

2. Loading the dataset

In [4]:

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

Out[4]:

	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 [6]:

```
df.shape
```

Out[6]:

```
(4177, 9)
```

In [7]:

```
Age=1.5*df.Rings
df["Age"]=Age
df=df.rename(columns = {'Whole weight':'Whole_weight', 'Shucked weight':
'Shucked_weight', 'Viscera weight': 'Viscera_weight',
'Shell weight': 'Shell_weight'})
df=df.drop(columns=["Rings"],axis=1)
df.head()
```

Out[7]:

	Sex	Length	Diameter	Height	Whole_weight	Shucked_weight	Viscera_weight	Shell_weight	Age
0	M	0.455	0.365	0.095	0.5140	0.2245	0.1010	0.150	16.5

	Sex	Length	Diameter	Height	Whole_weight	Shucked_weight	Viscera_weight	Shell_weight	Age
1	M	0.350	0.265	0.090	0.2255	0.0995	0.0485	0.070	8.5
2	F	0.530	0.420	0.135	0.6770	0.2565	0.1415	0.210	10.5
3	M	0.440	0.365	0.125	0.5160	0.2155	0.1140	0.155	11.5
4	I	0.330	0.255	0.080	0.2050	0.0895	0.0395	0.055	8.5

3. Performing Visualizations

(i) Univariate Analysis

Histogram

```
sns.displot(df["Age"], color='lightblue')
```

In [8]:

```
<seaborn.axisgrid.FacetGrid at 0x7f105a943e10>
```

Out[8]:

```
sns.histplot(y=df.Age,color='green')
```

In [9]:

```
<matplotlib.axes._subplots.AxesSubplot at 0x7f10579ca050>
```

Out[9]:

```
sns.histplot(x=df.Age,color='yellow')
```

In [10]:

```
<matplotlib.axes._subplots.AxesSubplot at 0x7f10578ae8d0>
```

Out[10]:

Boxplot

```
sns.boxplot(x=df.Age,color='orange')
```

In [11]:

```
<matplotlib.axes._subplots.AxesSubplot at 0x7f105777ff10>
```

Out[11]:

Countplot

```
sns.countplot(x=df.Age)
```

In [12]:


```
<matplotlib.axes._subplots.AxesSubplot at 0x7f1057717510>
```

Out[12]:

(ii) Bi-Variate Analysis

Barplot

```
sns.barplot(x=df.Height,y=df.Age)
```

In [14]:

```
<matplotlib.axes._subplots.AxesSubplot at 0x7f10576f5390>
```

Out[14]:

Linearplot

```
sns.lineplot(x=df.Age,y=df.Height, color='darkblue')
```

In [15]:

```
<matplotlib.axes._subplots.AxesSubplot at 0x7f10577137d0>
```

Out[15]:

Scatterplot

```
sns.scatterplot(x=df.Age,y=df.Height,color='green')
```

In [16]:

```
<matplotlib.axes._subplots.AxesSubplot at 0x7f10575d5150>
```

Out[16]:

Pointplot

```
sns.pointplot(x=df.Age, y=df.Height, color="blue")
```

In [17]:

```
<matplotlib.axes._subplots.AxesSubplot at 0x7f10573fee10>
```

Out[17]:

Regplot

```
sns.regplot(x=df.Age,y=df.Height,color='orange')
```

In [18]:

```
<matplotlib.axes._subplots.AxesSubplot at 0x7f105719f6d0>
```

Out[18]:

(

iii) Multi-Variate Analysis

Pairplot

```
In [20]:  
sns.pairplot(data=df[["Height", "Length", "Diameter", "Age", "Whole_weight", "Shucked_weight", "Viscera_weight", "Shell_weight"]])
```

```
Out[20]:  
<seaborn.axisgrid.PairGrid at 0x7f105710d850>
```

4. Perform descriptive statistics on the dataset

```
In [21]:  
df.describe(include='all')
```

```
Out[21]:
```

	Sex	Length	Diameter	Height	Whole_weight	Shucked_weight	Viscera_weight	Shell_weight	Age
count	4177	4177.000000	4177.000000	4177.000000	4177.000000	4177.000000	4177.000000	4177.000000	4177.000000
unique	3	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN
top	M	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN
freq	1528	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN
mean	NaN	0.523992	0.407881	0.139516	0.828742	0.359367	0.180594	0.238831	11.433684
std	NaN	0.120093	0.099240	0.041827	0.490389	0.221963	0.109614	0.139203	3.224169
min	NaN	0.075000	0.055000	0.000000	0.002000	0.001000	0.000500	0.001500	2.500000
25%	NaN	0.450000	0.350000	0.115000	0.441500	0.186000	0.093500	0.130000	9.500000
50%	NaN	0.545000	0.425000	0.140000	0.799500	0.336000	0.171000	0.234000	10.500000

	Sex	Length	Diameter	Height	Whole_weight	Shucked_weight	Viscera_weight	Shell_weight	Age
75 %	Na N	0.61500 0	0.48000 0	0.16500 0	1.153000	0.502000	0.253000	0.32900 0	12.5000 00
max	Na N	0.81500 0	0.65000 0	1.13000 0	2.825500	1.488000	0.760000	1.00500 0	30.5000 00

5. Check for Missing values and deal with them

```
df.isnull().sum()
```

In [22]:

Out[22]:

```
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. Find the outliers and replace the outliers

```
outliers=df.quantile(q=(0.25,0.75))
outliers
```

In [26]:

Out[26]:

	Length	Diameter	Height	Whole_weight	Shucked_weight	Viscera_weight	Shell_weight	Age
0.25	0.450	0.35	0.115	0.4415	0.186	0.0935	0.130	9.5
0.75	0.615	0.48	0.165	1.1530	0.502	0.2530	0.329	12.5

```
a = df.Age.quantile(0.25)
b = df.Age.quantile(0.75)
c = b - a
lower_limit = a - 1.5 * c
df.median(numeric_only=True)
```

In [32]:

Out[32]:

```
Length          0.5450
```

```
Diameter      0.4250
Height        0.1400
Whole_weight  0.7995
Shucked_weight 0.3360
Viscera_weight 0.1710
Shell_weight  0.2340
Age           10.5000
dtype: float64
```

In [33]:

```
df['Age'] = np.where(df['Age'] < lower_limit, 7, df['Age'])
sns.boxplot(x=df.Age, showfliers = False)
```

Out[33]:

```
<matplotlib.axes._subplots.AxesSubplot at 0x7f105535ead0>
```

7. Check for categorical columns and perform encoding

```
df.head()
```

Out[34]:

	Sex	Length	Diameter	Height	Whole_weight	Shucked_weight	Viscera_weight	Shell_weight	Age
0	M	0.455	0.365	0.095	0.5140	0.2245	0.1010	0.150	16.5
1	M	0.350	0.265	0.090	0.2255	0.0995	0.0485	0.070	8.5
2	F	0.530	0.420	0.135	0.6770	0.2565	0.1415	0.210	10.5
3	M	0.440	0.365	0.125	0.5160	0.2155	0.1140	0.155	11.5
4	I	0.330	0.255	0.080	0.2050	0.0895	0.0395	0.055	8.5

In [36]:

```
from sklearn.preprocessing import LabelEncoder

lab1 = LabelEncoder()
df.Sex = lab1.fit_transform(df.Sex)

df.head()
```

Out[36]:

	Sex	Length	Diameter	Height	Whole_weight	Shucked_weight	Viscera_weight	Shell_weight	Age
0	2	0.455	0.365	0.095	0.5140	0.2245	0.1010	0.150	16.5
1	2	0.350	0.265	0.090	0.2255	0.0995	0.0485	0.070	8.5
2	0	0.530	0.420	0.135	0.6770	0.2565	0.1415	0.210	10.5
3	2	0.440	0.365	0.125	0.5160	0.2155	0.1140	0.155	11.5
4	1	0.330	0.255	0.080	0.2050	0.0895	0.0395	0.055	8.5

8. Split the data into dependent and independent variables

In [39]:

```
y = df["Sex"]
y.head()
```

Out[39]:

```
0    2
1    2
2    0
3    2
4    1
Name: Sex, dtype: int64
```

In [40]:

```
x=df.drop(columns=["Sex"],axis=1)
x.head()
```

Out[40]:

	Length	Diameter	Height	Whole_weight	Shucked_weight	Viscera_weight	Shell_weight	Age
0	0.455	0.365	0.095	0.5140	0.2245	0.1010	0.150	16.5
1	0.350	0.265	0.090	0.2255	0.0995	0.0485	0.070	8.5
2	0.530	0.420	0.135	0.6770	0.2565	0.1415	0.210	10.5

	Length	Diameter	Height	Whole_weight	Shucked_weight	Viscera_weight	Shell_weight	Age
3	0.440	0.365	0.125	0.5160	0.2155	0.1140	0.155	11.5
4	0.330	0.255	0.080	0.2050	0.0895	0.0395	0.055	8.5

9. Scale the independent variables

```
from sklearn.preprocessing import scale
x_scaled = pd.DataFrame(scale(x), columns=x.columns)
x_scaled.head()
```

In [41]:

Out[41]:

	Length	Diameter	Height	Whole_weight	Shucked_weight	Viscera_weight	Shell_weight	Age
0	0.574558	0.432149	1.064424	-0.641898	-0.607685	-0.726212	-0.638217	1.577830
1	1.448986	1.439929	1.183978	-1.230277	-1.170910	-1.205221	-1.212987	0.919022
2	0.050033	0.122130	0.107991	-0.309469	-0.463500	-0.356690	-0.207139	0.294809
3	0.699476	0.432149	0.347099	-0.637819	-0.648238	-0.607600	-0.602294	0.017298
4	1.615544	1.540707	1.423087	-1.272086	-1.215968	-1.287337	-1.320757	0.919022

10. Split the data into training and testing

```
from sklearn.model_selection import train_test_split
X_train, X_test, Y_train, Y_test = train_test_split(x_scaled, y,
test_size=0.2, random_state=1)
```

In [42]:

```
X_train.shape, X_test.shape
```

In [44]:

```
((3341, 8), (836, 8))

Y_train.shape,Y_test.shape

((3341,), (836,))

X_train.head()
```

Out[44]:

In [45]:

Out[45]:

In [46]:

Out[46]:

	Length	Diameter	Height	Whole_weight	Shucked_weight	Viscera_weight	Shell_weight	Age
	-	-	-					
666	0.574558	0.583316	0.466653	-0.704101	-0.801435	-0.333880	-0.566371	0.329404
	-	-	-					
2813	2.240135	2.145375	2.020857	-1.542312	-1.490821	-1.492627	-1.565034	1.855341
	-	-	-					
1862	0.033246	0.021352	0.705762	-0.632721	-0.643732	-0.812890	-0.393939	0.606915
	-	-	-					
3684	0.799543	0.626020	0.370226	0.279929	0.394854	-0.087532	0.324524	0.329404
	-	-	-					
551	0.757903	0.827576	0.370226	0.325817	0.248416	0.131444	0.762786	0.953617

In [47]:

```
X_test.head()
```

Out[47]:

	Length	Diameter	Height	Whole_weight	Shucked_weight	Viscera_weight	Shell_weight	Age
	-	-	-					
17	0.699476	0.684094	0.944870	-0.770383	-0.772147	-0.853947	-0.781909	0.017298
	-	-	-					
1131	0.341509	0.273297	0.250672	0.328876	0.991872	0.017394	-0.235878	0.606915

	Length	Diameter	Height	Whole_weight	Shucked_weight	Viscera_weight	Shell_weight	Age
299	1.2824 28	1.2887 62	0.8253 16	-1.212942	-1.211462	-1.113981	-1.177064	0.2948 09
133 8	0.4664 27	0.4748 53	0.1079 91	-0.067795	0.205611	-0.124028	-0.250247	0.0172 98
238 3	0.0083 94	0.1802 04	0.1079 91	-0.465487	-0.598674	-0.452492	-0.207139	1.8899 36

```
In [48]:
Y_train.head()

Out[48]:
666      2
2813     1
1862     1
3684     1
551      1
Name: Sex, dtype: int64
```

```
In [49]:
Y_test.head()

Out[49]:
17      0
1131     2
299      2
1338     2
2383     0
Name: Sex, dtype: int64
```

11. Build the Model

```
In [50]:
from sklearn.ensemble import RandomForestClassifier
model = RandomForestClassifier(n_estimators=10,criterion='entropy')

In [51]:
model.fit(X_train,Y_train)

Out[51]:
RandomForestClassifier(criterion='entropy', n_estimators=10)

In [52]:
y_predict = model.predict(X_test)

In [56]:
y_predict_train = model.predict(X_train)
```

12. Train the Model


```
from sklearn.metrics import
accuracy_score, confusion_matrix, classification_report
```

In [54]:

```
print('Training accuracy: ', accuracy_score(Y_train, y_predict_train))
Training accuracy: 0.980544747081712
```

In [60]:

13. Test the Model

```
print('Testing accuracy: ', accuracy_score(Y_test, y_predict))
Testing accuracy: 0.5215311004784688
```

In [61]:

14. Measure the performance using Metrics

```
pd.crosstab(Y_test, y_predict)
```

In [62]:

Out[62]:

col_0	0	1	2
-------	---	---	---

Sex

0	122	31	107
---	-----	----	-----

1	26	198	40
---	----	-----	----

2	148	48	116
---	-----	----	-----

```
print(classification_report(Y_test, y_predict))
```

In [63]:

	precision	recall	f1-score	support
0	0.41	0.47	0.44	260
1	0.71	0.75	0.73	264
2	0.44	0.37	0.40	312
accuracy			0.52	836
macro avg	0.52	0.53	0.52	836
weighted avg	0.52	0.52	0.52	836

In []: