

In [1]: import pandas as pd import numpy as np import seaborn as sns import matplotlib.pyplot as plt import warnings

warnings.filterwarnings('ignore')

In [2]: df = pd.read_csv('penguins_binary_classification.csv')

In [3]: df.head()

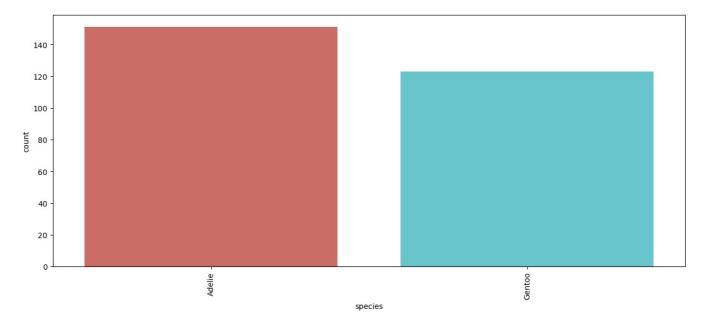
island bill_length_mm bill_depth_mm flipper_length_mm body_mass_g year Out[3]: species Adelie Torgersen 39.1 18.7 181.0 3750.0 2007 Adelie Torgersen 39.5 17.4 186.0 3800.0 2007 Adelie Torgersen 40.3 18.0 195.0 3250.0 2007 36.7 19.3 193.0 3450.0 2007 Adelie Torgersen Adelie Torgersen 39.3 20.6 190.0 3650.0 2007

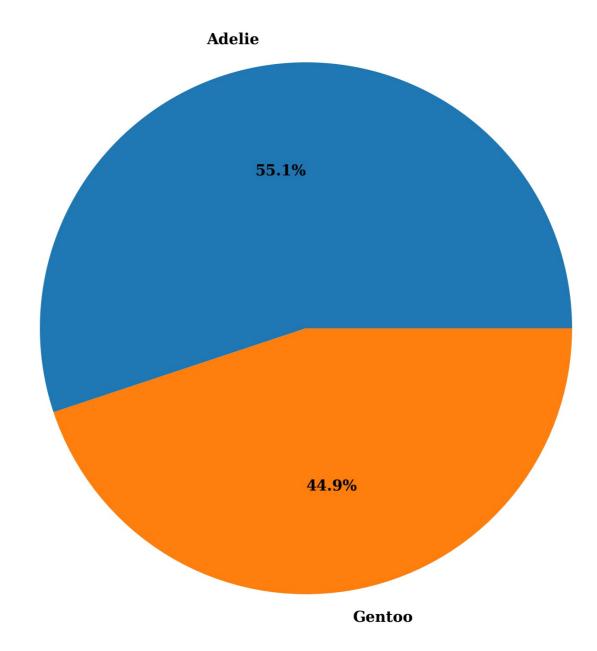
In [4]: df.tail()

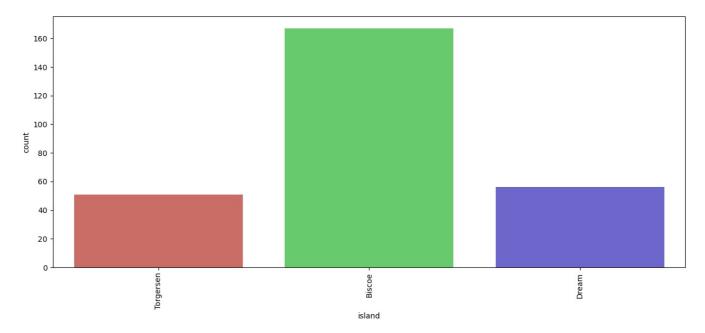
Out[4]: species island bill_length_mm bill_depth_mm flipper_length_mm body_mass_g year 269 47.2 13.7 214.0 4925.0 2009 Gentoo Biscoe 270 Gentoo Biscoe 46.8 14.3 215.0 4850.0 2009 5750.0 2009 271 Gentoo Biscoe 50.4 15.7 222.0 212.0 5200.0 2009 272 Gentoo Biscoe 45.2 14.8 Gentoo Biscoe 49.9 16.1 213.0 5400.0 2009

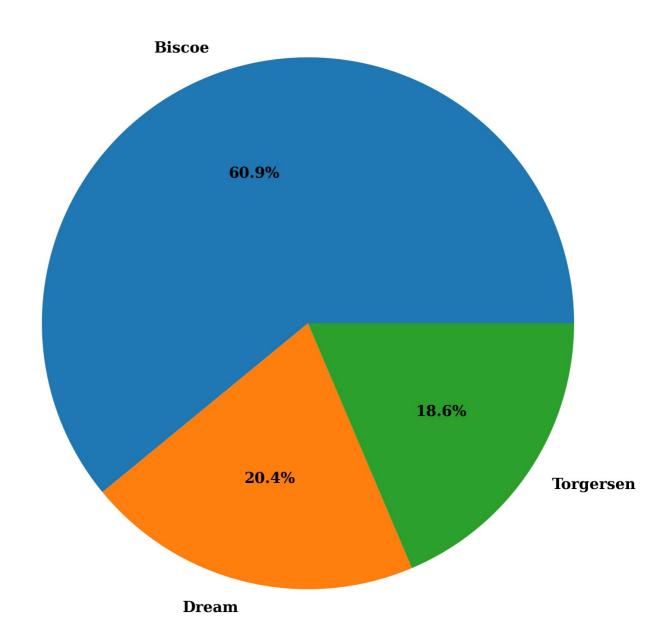
In [5]: df.describe()

```
bill_length_mm bill_depth_mm flipper_length_mm body_mass_g
 Out[5]:
                                                                               year
          count
                    274 000000
                                 274.000000
                                                  274.000000
                                                               274.000000
                                                                          274 000000
          mean
                     42.702920
                                  16.836131
                                                  202.178832
                                                              4318.065693 2008.043796
                      5.195566
                                   2.013410
                                                   15.047938
                                                               835.933105
                                                                            0.806281
            std
            min
                     32.100000
                                  13.100000
                                                  172.000000
                                                              2850.000000 2007.000000
                                                              3600.000000 2007.000000
           25%
                     38.350000
                                  15.000000
                                                  190.000000
           50%
                     42.000000
                                  17.000000
                                                  198.000000
                                                              4262.500000 2008.000000
           75%
                     46.675000
                                  18.500000
                                                  215.000000
                                                              4950.000000
                                                                         2009.000000
                     59.600000
                                  21.500000
                                                  231.000000
                                                              6300.000000 2009.000000
           max
 In [6]: df.isnull().sum()
          species
                                 0
 Out[6]:
          island
                                 0
          bill_length_mm
                                 0
          bill_depth_mm
                                 0
          flipper length mm
                                 0
          {\tt body\_mass\_g}
                                 0
                                 0
          year
          dtype: int64
 In [7]: df.dtypes
          species
                                  object
 Out[7]:
                                  object
          island
          bill\_length\_mm
                                 float64
          bill depth mm
                                 float64
          flipper_length_mm
                                 float64
          body_mass_g
                                 float64
          year
                                   int64
          dtype: object
 In [8]: df.columns
          dtype='object')
 In [9]: df.shape
          (274, 7)
 Out[9]:
          #from sklearn.preprocessing import LabelEncoder
In [11]:
          #object cols = df.select dtypes(include=['object']).columns
          #le = LabelEncoder()
          #for col in object_cols:
              #df[col] = le.fit transform(df[col])
In [10]: df.head()
Out[10]:
             species
                       island bill_length_mm bill_depth_mm flipper_length_mm body_mass_g year
          0
              Adelie Torgersen
                                       39.1
                                                    18.7
                                                                    181.0
                                                                                3750.0 2007
          1
              Adelie Torgersen
                                       39.5
                                                     17.4
                                                                    186.0
                                                                                3800.0 2007
          2
              Adelie Torgersen
                                       40.3
                                                     18.0
                                                                    195.0
                                                                                3250.0 2007
          3
              Adelie Torgersen
                                       36.7
                                                    19.3
                                                                    193.0
                                                                                3450.0 2007
              Adelie Torgersen
                                       39.3
                                                    20.6
                                                                    190.0
                                                                                3650.0 2007
In [11]: df['species'].unique()
          array(['Adelie', 'Gentoo'], dtype=object)
Out[11]:
In [13]: df['species'].value_counts()
          species
          Adelie
                     151
          Gentoo
                     123
          Name: count, dtype: int64
          plt.figure(figsize=(15,6))
In [14]:
          sns.countplot(x =df['species'], data = df, palette = 'hls')
          plt.xticks(rotation = 90)
          plt.show()
```

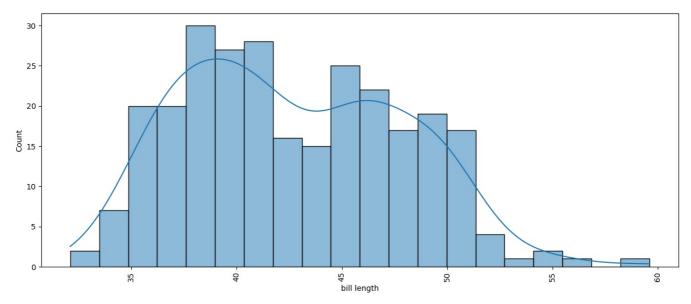




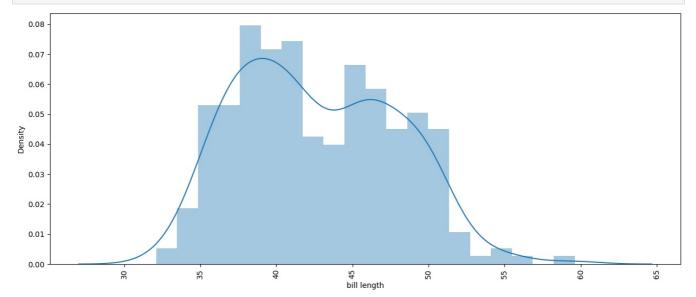




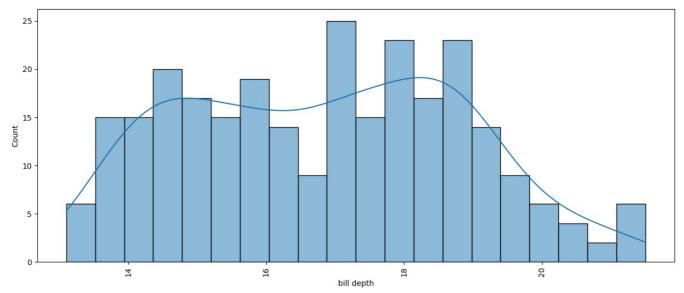
		.head()		h:11 la 4h	h:111 d4h	fliance langeth	h - d	
3]:	0	Species Adelie	Torgersen	39.1	18.7	flipper length	3750.0	
	1		Torgersen	39.5	17.4	186.0	3800.0	
	2	Adelie	Torgersen	40.3	18.0	195.0	3250.0	2007
	3	Adelie	Torgersen	36.7	19.3	193.0	3450.0	2007
	4	Adelie	Torgersen	39.3	20.6	190.0	3650.0	2007



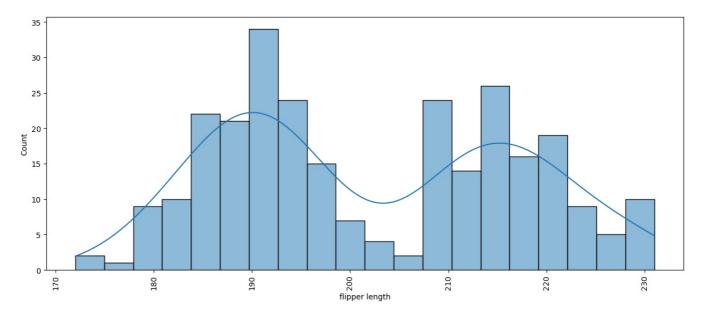
```
In [27]: plt.figure(figsize=(15,6))
    sns.distplot(df['bill length'], kde = True, bins = 20)
    plt.xticks(rotation = 90)
    plt.show()
```



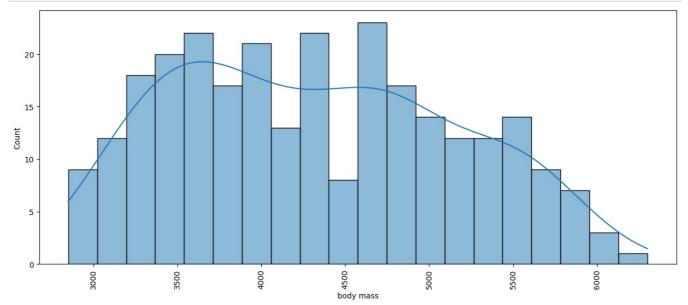
```
In [28]: plt.figure(figsize=(15,6))
    sns.histplot(df['bill depth'], kde = True, bins = 20, palette = 'hls')
    plt.xticks(rotation = 90)
    plt.show()
```



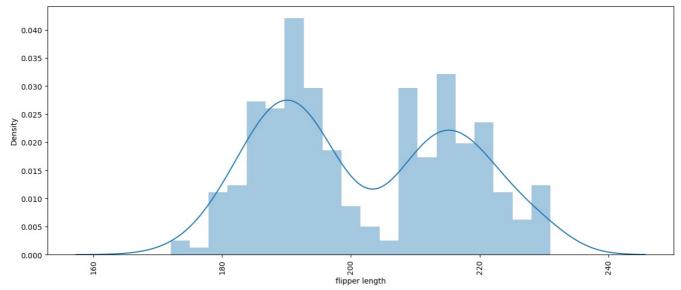
```
In [29]: plt.figure(figsize=(15,6))
    sns.histplot(df['flipper length'], kde = True, bins = 20, palette = 'hls')
    plt.xticks(rotation = 90)
    plt.show()
```



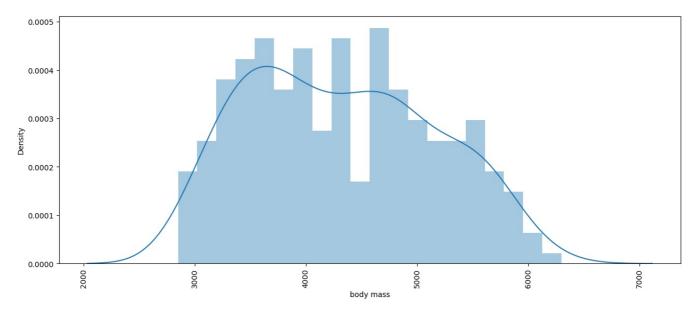
```
In [30]: plt.figure(figsize=(15,6))
    sns.histplot(df['body mass'], kde = True, bins = 20, palette = 'hls')
    plt.xticks(rotation = 90)
    plt.show()
```



```
In [31]: plt.figure(figsize=(15,6))
    sns.distplot(df['flipper length'], kde = True, bins = 20)
    plt.xticks(rotation = 90)
    plt.show()
```



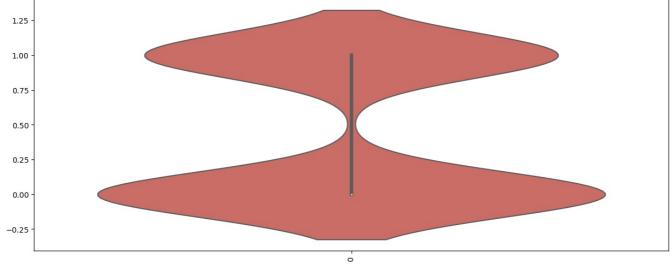
```
In [32]: plt.figure(figsize=(15,6))
    sns.distplot(df['body mass'], kde = True, bins = 20)
    plt.xticks(rotation = 90)
    plt.show()
```

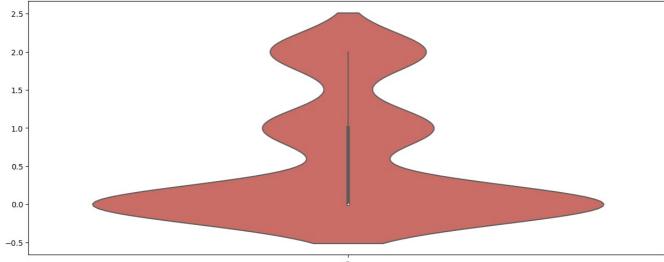


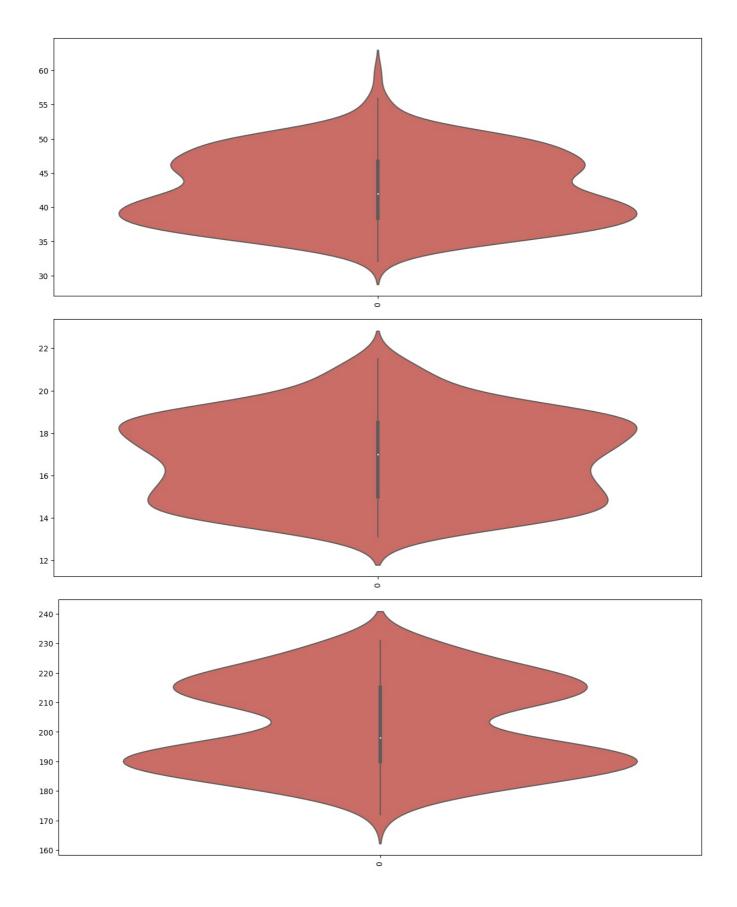
```
In [35]: from sklearn.preprocessing import LabelEncoder
   object_cols = df.select_dtypes(include=['object']).columns

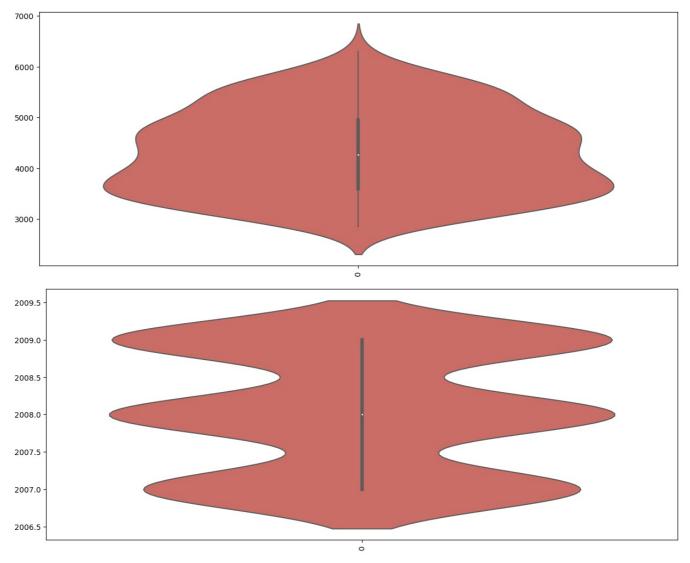
le = LabelEncoder()
   for col in object_cols:
        df[col] = le.fit_transform(df[col])
```

In [36]: for i in df.columns:
 plt.figure(figsize=(15,6))
 sns.violinplot(df[i], palette = 'hls')
 plt.xticks(rotation = 90)
 plt.show()

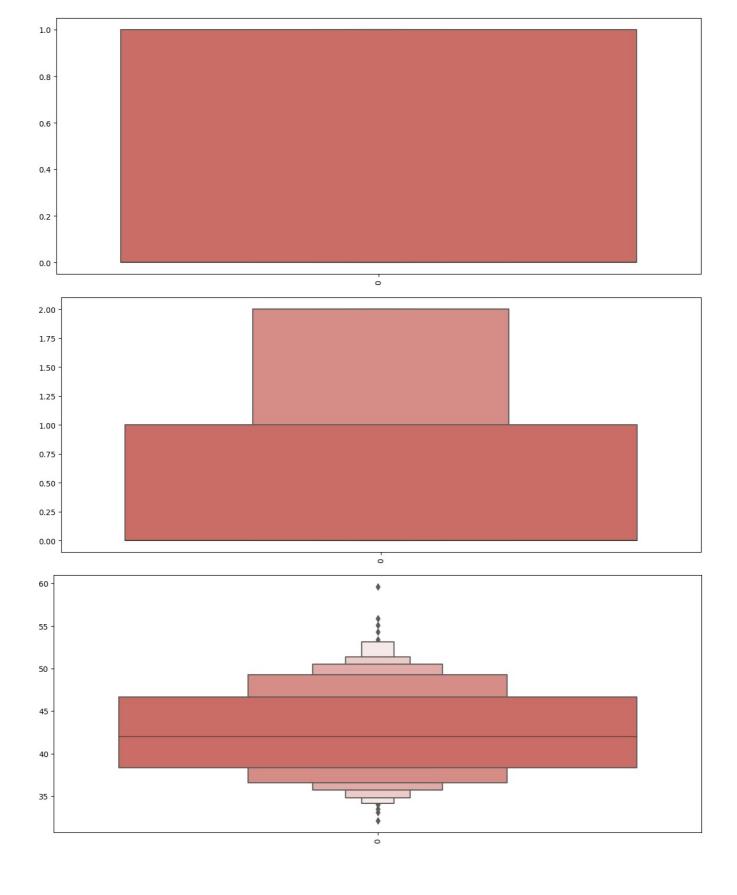


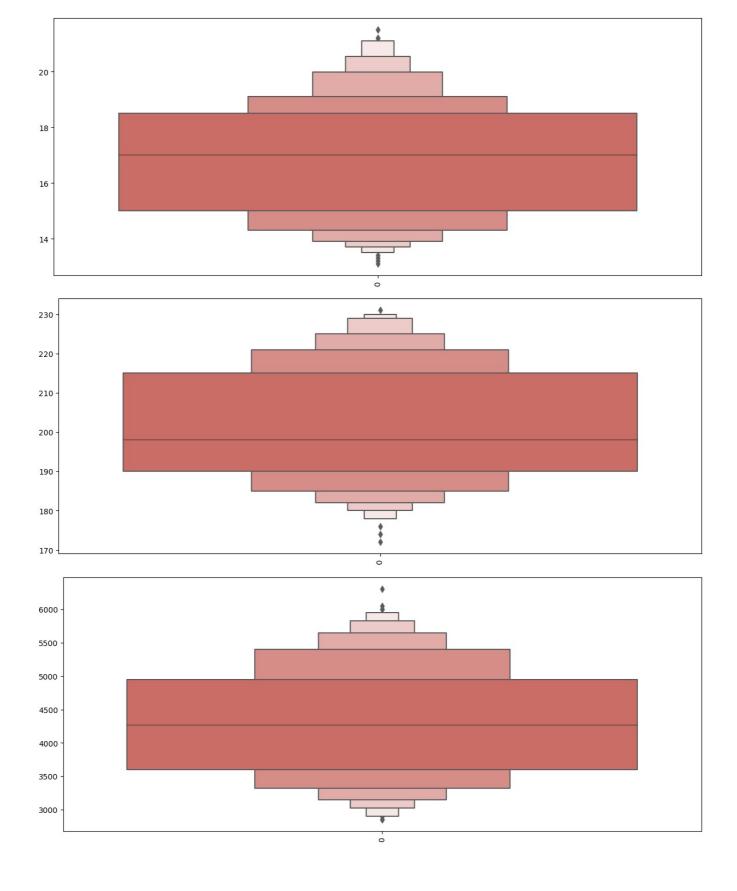




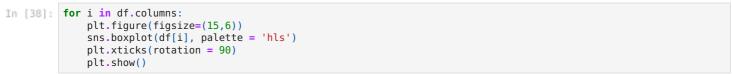


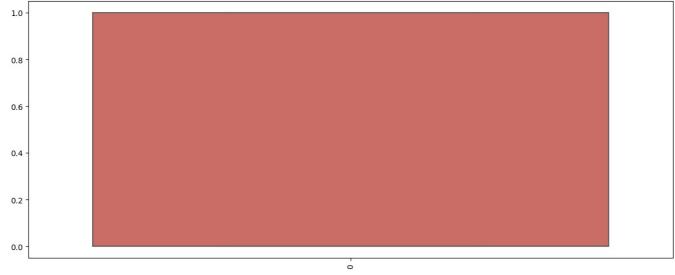
```
In [37]: for i in df.columns:
    plt.figure(figsize=(15,6))
    sns.boxenplot(df[i], palette = 'hls')
    plt.xticks(rotation = 90)
    plt.show()
```

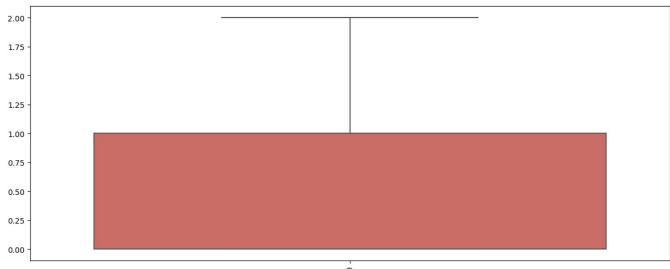


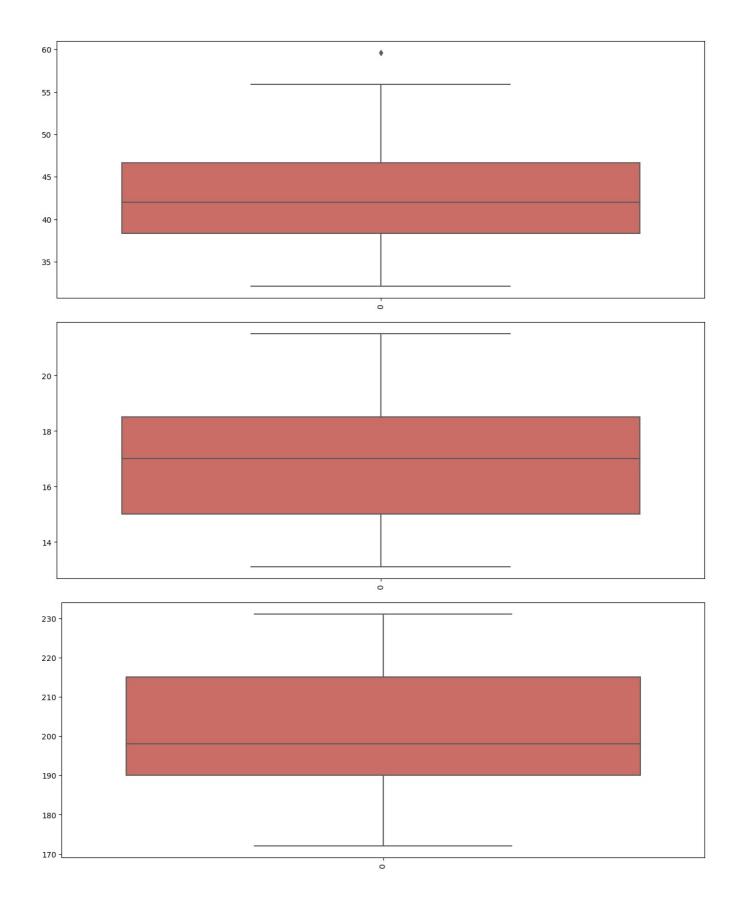














```
species
                                 1.000
                                 1.000
          island
          bill length
                                 8.325
          bill depth
                                 3.500
          flipper length
                                25.000
          body mass
                              1350.000
                                 2.000
          year
          dtype: float64
In [42]: df_{new} = df[\sim((df < (Q1 - 1.5 * IQR)) | (df > (Q3 + 1.5 * IQR))).any(axis = 1)]
In [43]: df_new.shape
          (273, 7)
Out[43]:
In [44]: df_corr = df.corr()
In [45]: df_corr
Out[45]:
                                   island bill length bill depth flipper length body mass
                        species
                                                                                        year
               species 1.000000 -0.662893
                                                                                    0.032942
                                           0.835687 -0.832606
                                                                 0.901796
                                                                           0.819837
                island -0.662893
                                1.000000
                                         -0.553917 0.559274
                                                                -0.562413
                                                                           -0.544047 -0.074625
             bill length
                       0.835687 -0.553917
                                           1.000000 -0.546050
                                                                 0.869026
                                                                           0.876905
                                                                                     0.086799
             bill depth -0.832606 0.559274 -0.546050 1.000000
                                                                           -0.483223 -0.051973
                                                                -0.640141
          flipper length 0.901796 -0.562413
                                          0.869026 -0.640141
                                                                 1.000000
                                                                           0.882262 0.150004
            body mass
                       0.819837 -0.544047
                                          0.876905 -0.483223
                                                                 0.882262
                                                                           1.000000
                                                                                    0.034148
                  year 0.032942 -0.074625 0.086799 -0.051973
                                                                           0.034148 1.000000
                                                                 0.150004
In [46]:
          plt.figure(figsize=(20, 17))
          matrix = np.triu(df_corr)
          sns.heatmap(df_corr, annot=True, linewidth=.8, mask=matrix, cmap="rocket");
          plt.show()
```

```
species
                                                                                                                                                    - 0.8
                                                                                                                                                    - 0.6
                                                                                                                                                    - 0.4
           length
                     0.84
                                                                                                                                                    - 0.2
           bill depth
                     -0.83
                                      0.56
            flipper length
                                                                                                                                                    -0.2
                                      -0.56
                                                                        -0.64
                      0.9
                                                       0.87
                     0.82
                                                       0.88
                                                                        -0.48
                                                                                         0.88
            year
                    species
                                      island
                                                     bill length
                                                                       bill depth
                                                                                      flipper length
                                                                                                        body mass
In [48]: from sklearn.model_selection import train_test_split
X_train,X_test,y_train,y_test = train_test_split(X,y)
In [63]: X_train.shape
            (205, 6)
Out[63]:
In [64]: X_test.shape
            (69, 6)
Out[64]:
In [47]: X = df.drop('species', axis=1)
            y = df['species']
In [50]: from sklearn.linear_model import LogisticRegression
            regressor = LogisticRegression()
In [60]:
            regressor.fit(X_train,y_train)
Out[60]: ▼ LogisticRegression
           LogisticRegression()
In [51]: model = LogisticRegression()
In [53]: model.fit(X_train,y_train)
Out[53]: ▼ LogisticRegression
            LogisticRegression()
```

In [54]: y_predict = model.predict(X_test)

```
y predict
Out[54]: array([0, 0, 0, 1, 0, 0, 0, 0, 1, 1, 1, 0, 0, 1, 0, 0, 1, 1, 0, 1, 0,
                 0, 0, 0, 1, 1, 1, 0, 0, 1, 1, 0, 1, 0, 1, 0, 0, 1, 1, 0, 1, 0, 0,
                 0, 1, 0, 1, 1, 1, 1, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 1, 0, 0,
                 0, 1, 0])
In [55]: from sklearn.metrics import accuracy_score, classification_report, confusion matrix
In [56]:
          accuracy score(y test,y predict)
Out[56]:
In [57]: print(classification_report(y_test,y_predict))
                         precision
                                      recall f1-score
                                                           support
                     0
                              1.00
                                         1.00
                                                   1.00
                                                                42
                     1
                              1.00
                                         1.00
                                                   1.00
                                                                27
                                                                69
                                                   1.00
              accuracy
                              1.00
                                         1.00
             macro avg
                                                   1.00
                                                                69
          weighted avg
                              1.00
                                         1.00
                                                   1.00
                                                                69
In [65]: df.head()
            species island bill length bill depth flipper length body mass year
                 0
                        2
                               39.1
                                        18.7
                                                   181.0
                                                            3750.0 2007
          1
                 0
                        2
                               39.5
                                        17.4
                                                   186.0
                                                            3800.0 2007
          2
                 0
                        2
                               40.3
                                        18.0
                                                   195.0
                                                            3250.0 2007
          3
                 0
                        2
                               36.7
                                        19.3
                                                   193.0
                                                            3450.0 2007
                        2
          4
                 0
                               39.3
                                        20.6
                                                   190.0
                                                            3650.0 2007
In [73]: df.tail()
              species island bill length bill depth flipper length body mass year
          269
                          0
                                                     214.0
                                 47.2
                                          13.7
                                                              4925.0 2009
          270
                         0
                                 46.8
                                          14.3
                                                    215.0
                                                              4850.0 2009
          271
                   1
                          0
                                 50.4
                                          15.7
                                                     222.0
                                                              5750.0 2009
          272
                          0
                                 45.2
                                          14.8
                                                     212.0
                                                              5200.0 2009
                          0
                                                              5400.0 2009
          273
                   1
                                 49.9
                                          16.1
                                                     213.0
In [66]: input data = (2,39.1,18.7,181.0,3750.0,2007)
          input data as numpy array = np.asarray(input data)
          input_data_reshaped = input_data_as_numpy_array.reshape(1,-1)
          prediction = regressor.predict(input_data_reshaped)
          print(prediction)
In [67]: print('The penguins species',prediction[0])
          The penguins species 0
In [68]:
          if (prediction[0])==0:
              print('The penguin species is Adelie')
              print('The penguin species is Gentoo')
          The penguin species is Adelie
In [69]: import pickle
In [70]:
          filename = 'trained model.sav'
          pickle.dump(regressor,open(filename,'wb'))
In [71]: loaded model = pickle.load(open('trained model.sav','rb'))
In [74]:
          input_data = (1,39.1,18.7,181.0,3750.0,2007)
          input_data_as_numpy_array = np.asarray(input_data)
          input data reshaped = input data as numpy array.reshape(1,-1)
          prediction = regressor.predict(input data reshaped)
```

```
print(prediction)

if (prediction[0])==0:
    print('The penguin species is Adelie')
else:
    print('The penguin species is Gentoo')

[0]
    The penguin species is Adelie

In [1]: import ultralytics

In [2]: ultralytics.__version__
Out[2]: '8.0.232'

In [3]: import supervision
    print("supervision.__version_:", supervision.__version_)
    supervision.__version_: 0.17.1

In []:
```

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