Import libraries

In [1]:

```
import numpy as np
import pandas as pd
import seaborn as sbn
from sklearn.utils import shuffle
import matplotlib.pyplot as plt
```

In [2]:

```
dataset = pd.read_excel("fish.xlsx")
```

We created one Excel table and then we filled the values into this file then saved as "fish.xlsx"

In [3]:

dataset

Out[3]:

	skin_color	weight	width_size	height	dorsal_fin_size	eye_diameter	class
0	blue	8.302707	1.167420	19.184339	6.193168	0.506914	0
1	blue	7.418877	2.198725	12.464850	6.498403	0.626358	0
2	blue	9.499399	2.568813	17.449198	6.360583	0.769239	0
3	blue	7.680885	1.191879	12.786238	6.447008	0.448620	0
4	blue	7.502223	2.806455	13.940972	6.355912	0.985559	0
95	red	128.601004	5.816665	20.686216	10.807237	0.582889	1
96	red	103.316794	3.490792	15.126622	11.214208	0.002602	1
97	red	110.608310	3.198861	29.323495	11.097603	1.199941	1
98	red	121.906324	4.445191	28.765103	6.704414	0.741137	1
99	red	131.909198	5.364021	23.689947	7.787630	1.070646	1

100 rows × 7 columns

In our dataset has 6 data values and 1 target value, in our dataset every data must be numerical so we should check it.

Preprocessing

In [4]:

dataset.dtypes

Out[4]:

skin_color object
weight float64
width_size float64
height float64
dorsal_fin_size float64
eye_diameter float64
class int64

dtype: object

As you see that, our skin color value is object so we should convert it to integer.

At this step, we can use LabelEncoder as well for convert to object data to integer.

In [5]:

```
from sklearn.preprocessing import LabelEncoder
le = LabelEncoder()
dataset["skin_color"] = le.fit_transform(dataset["skin_color"])
```

In [6]:

dataset.dtypes

Out[6]:

skin_color int32
weight float64
width_size float64
height float64
dorsal_fin_size float64
eye_diameter float64
class int64

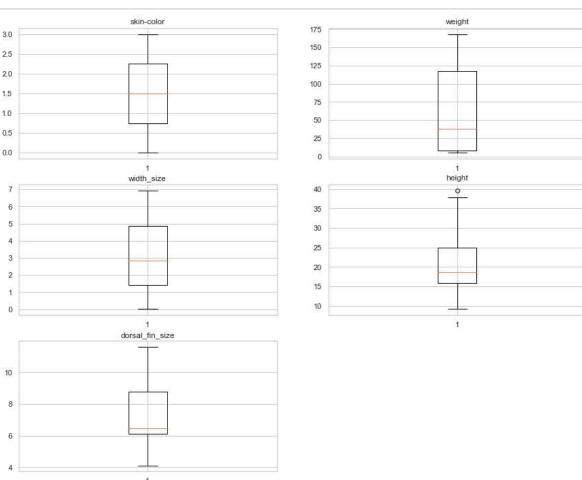
dtype: object

Visualization

Box Plot

In [7]:

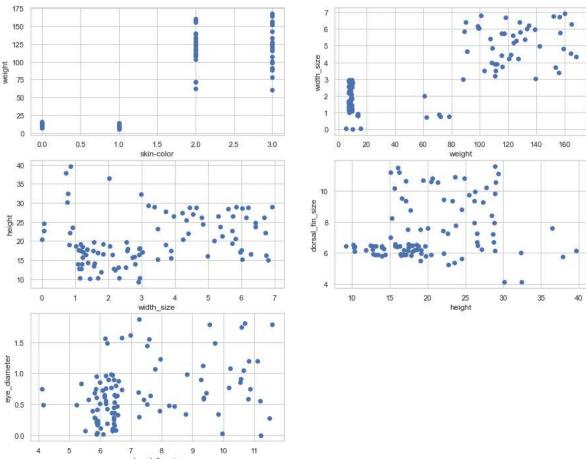
```
sbn.set_theme(style="whitegrid")
plt.figure(figsize=(15, 12))
plt.subplot(3, 2, 1)
plt.boxplot(dataset["skin_color"])
plt.title("skin-color")
plt.subplot(3, 2, 2)
plt.boxplot(dataset["weight"])
plt.title("weight")
plt.subplot(3, 2, 3)
plt.boxplot(dataset["width_size"])
plt.title("width_size")
plt.subplot(3, 2, 4)
plt.boxplot(dataset["height"])
plt.title("height")
plt.subplot(3, 2, 5)
plt.boxplot(dataset["dorsal_fin_size"])
plt.title("dorsal_fin_size")
plt.show()
```



Scatter Plot

In [8]:

```
plt.figure(figsize=(15, 12))
plt.subplot(3, 2, 1)
plt.scatter(dataset["skin_color"],dataset["weight"])
plt.xlabel("skin-color")
plt.ylabel("weight")
plt.subplot(3, 2, 2)
plt.scatter(dataset["weight"],dataset["width_size"])
plt.xlabel("weight")
plt.ylabel("width_size")
plt.subplot(3, 2, 3)
plt.scatter(dataset["width_size"],dataset["height"])
plt.xlabel("width_size")
plt.ylabel("height")
plt.subplot(3, 2, 4)
plt.scatter(dataset["height"],dataset["dorsal_fin_size"])
plt.xlabel("height")
plt.ylabel("dorsal_fin_size")
plt.subplot(3, 2, 5)
plt.scatter(dataset["dorsal_fin_size"],dataset["eye_diameter"])
plt.xlabel("dorsal_fin_size")
plt.ylabel("eye_diameter")
plt.show()
```



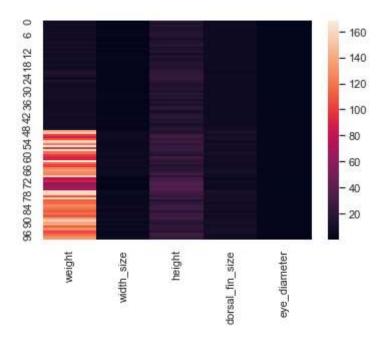
Correlation

In [9]:

```
sbn.heatmap(dataset[["weight","width_size","height","dorsal_fin_size","eye_diameter"]])
```

Out[9]:

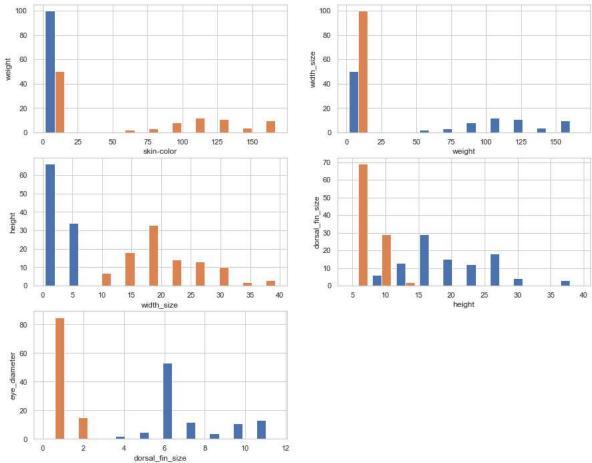
<AxesSubplot:>



Histogram

In [10]:

```
plt.figure(figsize=(15, 12))
plt.subplot(3, 2, 1)
plt.hist(dataset[["skin_color","weight"]])
plt.xlabel("skin-color")
plt.ylabel("weight")
plt.subplot(3, 2, 2)
plt.hist(dataset[["weight","width_size"]])
plt.xlabel("weight")
plt.ylabel("width_size")
plt.subplot(3, 2, 3)
plt.hist(dataset[["width_size", "height"]])
plt.xlabel("width_size")
plt.ylabel("height")
plt.subplot(3, 2, 4)
plt.hist(dataset[["height","dorsal_fin_size"]])
plt.xlabel("height")
plt.ylabel("dorsal fin size")
plt.subplot(3, 2, 5)
plt.hist(dataset[["dorsal_fin_size","eye_diameter"]])
plt.xlabel("dorsal_fin_size")
plt.ylabel("eye_diameter")
plt.show()
```



Now the dataset is looking better than before, after that step we can distinguish our datas and target value

In [11]:

```
X = dataset.drop(columns ="class",axis = 0).values
y = dataset[["class"]].values.ravel()
```

```
In [12]:
```

```
X,y = shuffle(X,y)
```

```
In [13]:
```

```
from sklearn.model_selection import train_test_split

X_train,X_test,y_train,y_test = train_test_split(X,y,test_size=0.3,random_state=41)
```

Above the code means that, we have a dataset with 100 rows and we should split our datas %30 testing and %70 training, then we can work on our 70 rows's values afterward we can test it with 30 percent of testing datas.

Decision Tree

In [14]:

```
from sklearn.tree import DecisionTreeClassifier

dt_classifier = DecisionTreeClassifier()

dt_classifier.fit(X_train,y_train)

y_predict = dt_classifier.predict(X_test)
```

In [15]:

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

We defined accuracy score and confusion matrix for see the results.

Accuracy Score which is how much percent accuracy we have in our values.

Confusion Matrix which is in our 30 testing datas that how many datas are true negative, false positive, false negative and ture positive

```
In [16]:
```

```
print("Accuracy score is: ",accuracy_score(y_test,y_predict))
print("Confusion matrix is(tn,fp,fn,tp) :\n",confusion_matrix(y_test,y_predict))

Accuracy score is: 1.0
Confusion matrix is(tn,fp,fn,tp) :
[[19 0]
[ 0 11]]
```

KNN

```
In [17]:
```

```
X,y = shuffle(X,y)
X_train,X_test,y_train,y_test = train_test_split(X,y,test_size=0.3,random_state=41)
from sklearn.neighbors import KNeighborsClassifier
knn_classifier = KNeighborsClassifier()
knn_classifier.fit(X_train,y_train)
y_predict = knn_classifier.predict(X_test)
```

In [18]:

```
print("Accuracy score is: ",accuracy_score(y_test,y_predict))
print("Confusion matrix is(tn,fp,fn,tp) :\n",confusion_matrix(y_test,y_predict))

Accuracy score is: 1.0
Confusion matrix is(tn,fp,fn,tp) :
[[15  0]
[ 0 15]]
```

Ensemble

In [19]:

```
X,y = shuffle(X,y)
X_train,X_test,y_train,y_test = train_test_split(X,y,test_size=0.3,random_state=41)
from sklearn.ensemble import RandomForestClassifier

rf_classifier = RandomForestClassifier(n_estimators=10)

rf_classifier.fit(X_train,y_train)

y_predict = rf_classifier.predict(X_test)
```

In [20]:

```
print("Accuracy score is: ",accuracy_score(y_test,y_predict))
print("Confusion matrix is(tn,fp,fn,tp) :\n",confusion_matrix(y_test,y_predict))

Accuracy score is: 1.0
Confusion matrix is(tn,fp,fn,tp) :
[[17 0]
[ 0 13]]

In [ ]:
```