**Species** 

0.2 Iris-setosa

Out[59]:

0 1

```
In [ ]:
           Pooja Dhumal
           LGM task 2
           Name-Iris flower classification ML project
In [54]:
           import pandas as pd
           import numpy as np
           import matplotlib.pyplot as plt
           import seaborn as sns
           sns.set(style='white', color codes=True)
           from sklearn.linear_model import LogisticRegression
           from sklearn.model selection import train test split
           from sklearn.neighbors import KNeighborsClassifier
           from sklearn import svm
           from sklearn import metrics
           from sklearn.tree import DecisionTreeClassifier
In [58]:
           df= pd.read_csv(r"C:\Users\Pooja\Downloads\Iris.csv")
In [22]:
           df
Out[22]:
                Id SepalLengthCm SepalWidthCm PetalLengthCm PetalWidthCm
                                                                                  Species
            0
                 1
                               5.1
                                             3.5
                                                            1.4
                                                                          0.2
                                                                                Iris-setosa
                 2
                               4.9
                                                                          0.2
                                                                                Iris-setosa
            1
                                             3.0
                                                            1.4
            2
                 3
                               4.7
                                                                          0.2
                                             3.2
                                                            1.3
                                                                                Iris-setosa
            3
                                             3.1
                                                            1.5
                                                                          0.2
                               4.6
                                                                                Iris-setosa
                               5.0
                                                                          0.2
                                             3.6
                                                            1.4
                                                                                Iris-setosa
                                                             •••
          145
              146
                               6.7
                                             3.0
                                                            5.2
                                                                          2.3 Iris-virginica
          146
              147
                               6.3
                                             2.5
                                                            5.0
                                                                          1.9 Iris-virginica
          147 148
                               6.5
                                             3.0
                                                            5.2
                                                                          2.0 Iris-virginica
          148
              149
                               6.2
                                             3.4
                                                            5.4
                                                                          2.3 Iris-virginica
                               5.9
          149 150
                                             3.0
                                                            5.1
                                                                          1.8 Iris-virginica
         150 rows × 6 columns
In [59]:
           df.head()
```

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1.4

Id SepalLengthCm SepalWidthCm PetalLengthCm PetalWidthCm

3.5

5.1

	ı	d Se	palLengthCm S	epalWidthCm	PetalLeng	thCm Per	talWidthCm	Species	
	1	2	4.9	3.0		1.4	0.2	Iris-setosa	
	2	3	4.7	3.2		1.3	0.2	Iris-setosa	
	2	Δ	46	2 1		1 5	0.2	lric-cetoca	
n [60]:	df	.tail	L ()						
ut[60]:		ld	SepalLengthCn	n SepalWidth(	Cm PetalL	.engthCm	PetalWidthC	m Species	
	145	146	6.	7	3.0	5.2	2	2.3 Iris-virginica	
	146	147	6.3	3	2.5	5.0	1	I.9 Iris-virginica	
	147	148	6.5	5	3.0	5.2	2	2.0 Iris-virginica	
	148	149	6.2	2	3.4	5.4	2	2.3 Iris-virginica	
	149	150	5.9	)	3.0	5.1	1	I.8 Iris-virginica	
n [61]:	df	.shap	pe						
ut[61]:	(150, 6)								
n [62]:	df.info()								
	<pre><class 'pandas.core.frame.dataframe'=""> RangeIndex: 150 entries, 0 to 149 Data columns (total 6 columns): # Column Non-Null Count Dtype</class></pre>								
		Id Sep Sep Per Per Sper	palLengthCm palWidthCm talLengthCm talWidthCm ecies float64(4), usage: 7.2+	150 non-nu 150 non-nu 150 non-nu 150 non-nu int64(1),	11 in 11 fl 11 fl 11 fl 11 fl 11 fl				
n [63]:	df	.isnu	ull().sum()						
ut[63]:	Sep Pet Pet Spe	alWid alLen alWid cies	ngthCm 0 dthCm 0 ngthCm 0 dthCm 0 dthCm 0 dthCm 0						
n [64]:	df	.desc	cribe()						
ut[64]:			ld Sepal	LengthCm Se <sub>l</sub>	palWidthCr	n PetalLe	engthCm Pet	talWidthCm	

		Id	SepalLengthCm	SepalWidthCm	PetalLengthCm	PetalWidthCm
	count	150.000000	150.000000	150.000000	150.000000	150.000000
	mean	75.500000	5.843333	3.054000	3.758667	1.198667
	std	43.445368	0.828066	0.433594	1.764420	0.763161
	min	1.000000	4.300000	2.000000	1.000000	0.100000
	25%	38.250000	5.100000	2.800000	1.600000	0.300000
	50%	75.500000	5.800000	3.000000	4.350000	1.300000
	75%	112.750000	6.400000	3.300000	5.100000	1.800000
In [65]:	df.nı	unique()				
Out[65]:	Sepal Petal Petal Speci	LengthCm WidthCm LengthCm WidthCm es : int64	150 35 23 43 22 3			
In [66]:	df['	SepalWidth	nCm'].value_co	ounts()		
Out[66]:	3.0 2.8 3.2 3.1 3.4 2.9 2.7 2.5 3.5 3.3 3.8 2.6 2.3 3.7 2.4 2.2 3.6 3.9 4.4 4.0 4.1 4.2 2.0 Name:	26 14 13 12 12 10 9 8 6 6 6 5 4 3 3 3 2 1 1 1 1 1 SepalWide	thCm, dtype:	int64		
In [67]:	df.c	orr()				

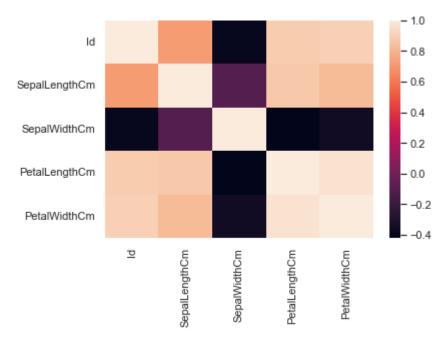
Out[67]: Id SepalLengthCm SepalWidthCm PetalLengthCm PetalWidthCm

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	ld	SepalLengthCm	SepalWidthCm	PetalLengthCm	PetalWidthCm
Id	1.000000	0.716676	-0.397729	0.882747	0.899759
SepalLengthCm	0.716676	1.000000	-0.109369	0.871754	0.817954
SepalWidthCm	-0.397729	-0.109369	1.000000	-0.420516	-0.356544
PetalLengthCm	0.882747	0.871754	-0.420516	1.000000	0.962757

```
In [68]: sns.heatmap(df.corr())
```

Out[68]: <AxesSubplot:>

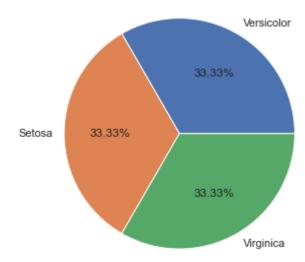


```
In [69]:
          df.columns
         Index(['Id', 'SepalLengthCm', 'SepalWidthCm', 'PetalLengthCm', 'PetalWidthCm',
Out[69]:
                'Species'],
               dtype='object')
In [70]:
         n = len(df[df['Species'] == 'Iris-versicolor'])
         print("No of Versicolor in Dataset:",n)
        No of Versicolor in Dataset: 50
In [71]:
         n1 = len(df[df['Species'] == 'Iris-versicolor'])
         print("No of Versicolor in Dataset:",n1)
        No of Versicolor in Dataset: 50
In [72]:
         n2 = len(df[df['Species'] == 'Iris-versicolor'])
         print("No of Versicolor in Dataset:",n2)
```

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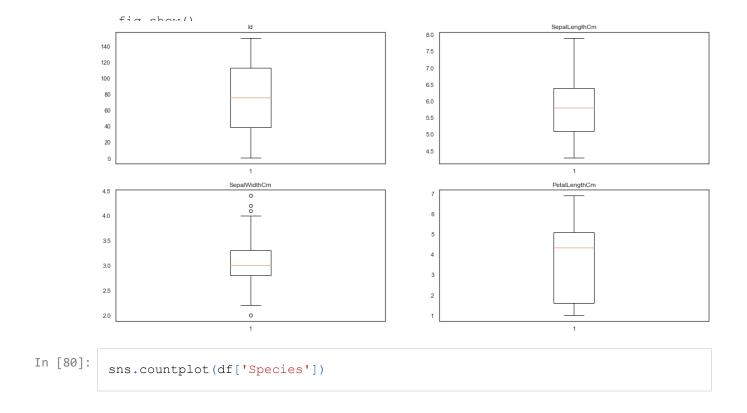
No of Versicolor in Dataset: 50

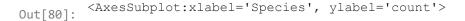
```
In [73]:
          df.isnull().sum()
Out[73]:
         SepalLengthCm
                          0
                          0
         SepalWidthCm
         PetalLengthCm
         PetalWidthCm
                          0
         Species
                          0
         dtype: int64
In [78]:
          fig = plt.figure()
         ax = fig.add_axes([0,0,1,1])
         ax.axis('equal')
         1 = ['Versicolor', 'Setosa', 'Virginica']
         s = [50, 50, 50]
         ax.pie(s, labels = 1,autopct= '%1.2f%%')
         plt.show()
```

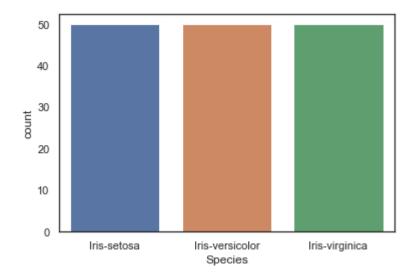


```
In [79]: # Checking outliers
   rows = 2
   cols = 2
   fig, axs = plt.subplots(rows,cols)
   index=-1
   for i in range(rows):
      for j in range(cols):
        index+=1
        axs[i,j].boxplot(df[df.columns[index]])
        axs[i,j].set_title(df.columns[index])
   fig.set_size_inches(10,5)
   fig.show()
```

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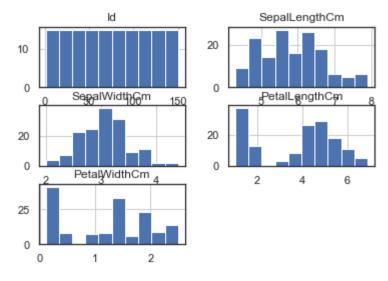






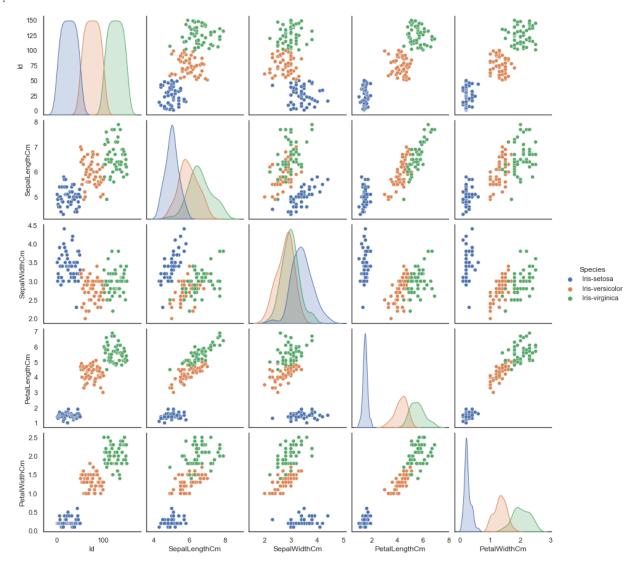
```
In [81]: df.hist()
    plt.show()
```

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In [82]: sns.pairplot(df, hue='Species')

Out[82]: <seaborn.axisgrid.PairGrid at 0x15a1fd24220>



```
In [83]:
           train, test = train test split(df, test size = 0.24)
           print(train.shape)
           print(test.shape)
          (112, 6)
          (38, 6)
In [106...
           train_X = df[['SepalLengthCm', 'SepalWidthCm', 'PetalLengthCm', 'PetalWidthCm
           train y = df.Species
           test X = df[['SepalLengthCm', 'SepalWidthCm', 'PetalLengthCm', 'PetalWidthCm'
           test y = df.Species
In [101...
           train X.head()
Out[101...
             SepalLengthCm SepalWidthCm PetalLengthCm PetalWidthCm
          0
                       5.1
                                    3.5
                                                  1.4
                                                               0.2
          1
                       4.9
                                    3.0
                                                  1.4
                                                               0.2
          2
                       4.7
                                    3.2
                                                  1.3
                                                               0.2
          3
                       4.6
                                    3.1
                                                  1.5
                                                               0.2
                       5.0
                                    3.6
                                                  1.4
                                                               0.2
In [107...
           test y.head()
               Iris-setosa
Out[107...
          1
               Iris-setosa
          2
               Iris-setosa
              Iris-setosa
               Iris-setosa
          Name: Species, dtype: object
In [108...
           #Using LogisticRegression
           model = LogisticRegression()
           model.fit(train_X, train_y)
           prediction = model.predict(test X)
           print('Accuracy:', metrics.accuracy score(prediction, test y))
          Accuracy: 0.9733333333333334
```

```
In [112...
#Using Support Vector
from sklearn.svm import SVC
model1 = SVC()
model1.fit(train_X, train_y)

pred_y = model1.predict(test_X)

from sklearn.metrics import accuracy_score
print("Acc=", accuracy_score(test_y, pred_y))
Acc= 0.97333333333333333334
```

In [113...

```
#Using KNN Neighbours
from sklearn.neighbors import KNeighborsClassifier
model2 = KNeighborsClassifier(n_neighbors=5)
model2.fit(train_X, train_y)
y_pred2 = model2.predict(test_X)

from sklearn.metrics import accuracy_score
print("Accuracy Score:,accuracy_score(test_y,y_pred2)")
```

Accuracy Score:, accuracy score(test y, y pred2)