

```
In [1]: #IMPORTING LIBRARIES
import warnings
warnings.filterwarnings("ignore")

import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt
%matplotlib inline
from sklearn.linear_model import LogisticRegression
from sklearn.model_selection import train_test_split
from sklearn.metrics import classification_report

In [23]: #LOADING DATA

In [5]: Iris = pd.read_csv(r"C:\Users\91918\Downloads\Iris.csv")

In [6]: Iris

Out[6]:
```

	Id	SepalLengthCm	SepalWidthCm	PetalLengthCm	PetalWidthCm	Species
0	1	5.1	3.5	1.4	0.2	Iris-setosa
1	2	4.9	3.0	1.4	0.2	Iris-setosa
2	3	4.7	3.2	1.3	0.2	Iris-setosa
3	4	4.6	3.1	1.5	0.2	Iris-setosa
4	5	5.0	3.6	1.4	0.2	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
149	150	5.9	3.0	5.1	1.8	Iris-virginica

150 rows × 6 columns

```
In [24]: # DATA DESCRIPTION

In [7]: Iris.describe()

Out[7]:
```

	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
max	150.000000	7.900000	4.400000	6.900000	2.500000

```
In [26]: # DATA EXTRACTION

In [25]: Data = Iris[['Id', 'SepalLengthCm', 'SepalWidthCm', 'PetalLengthCm', 'PetalWidthCm']]

In [9]: Data

Out[9]:
```

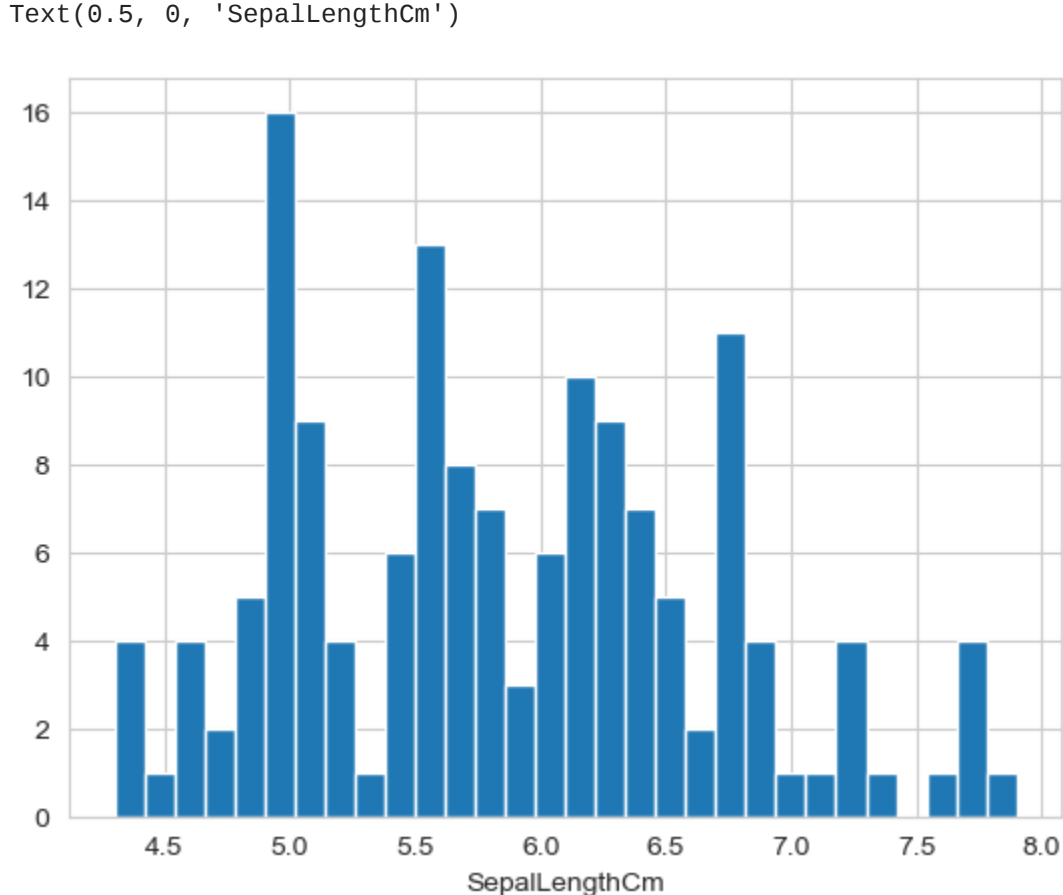
	Id	SepalLengthCm	SepalWidthCm	PetalLengthCm	PetalWidthCm
0	1	5.1	3.5	1.4	0.2
1	2	4.9	3.0	1.4	0.2
2	3	4.7	3.2	1.3	0.2
3	4	4.6	3.1	1.5	0.2
4	5	5.0	3.6	1.4	0.2
...
145	146	6.7	3.0	5.2	2.3
146	147	6.3	2.5	5.0	1.9
147	148	6.5	3.0	5.2	2.0
148	149	6.2	3.4	5.4	2.3
149	150	5.9	3.0	5.1	1.8

150 rows × 5 columns

```
In [27]: #VISUALIZATION OF DATA

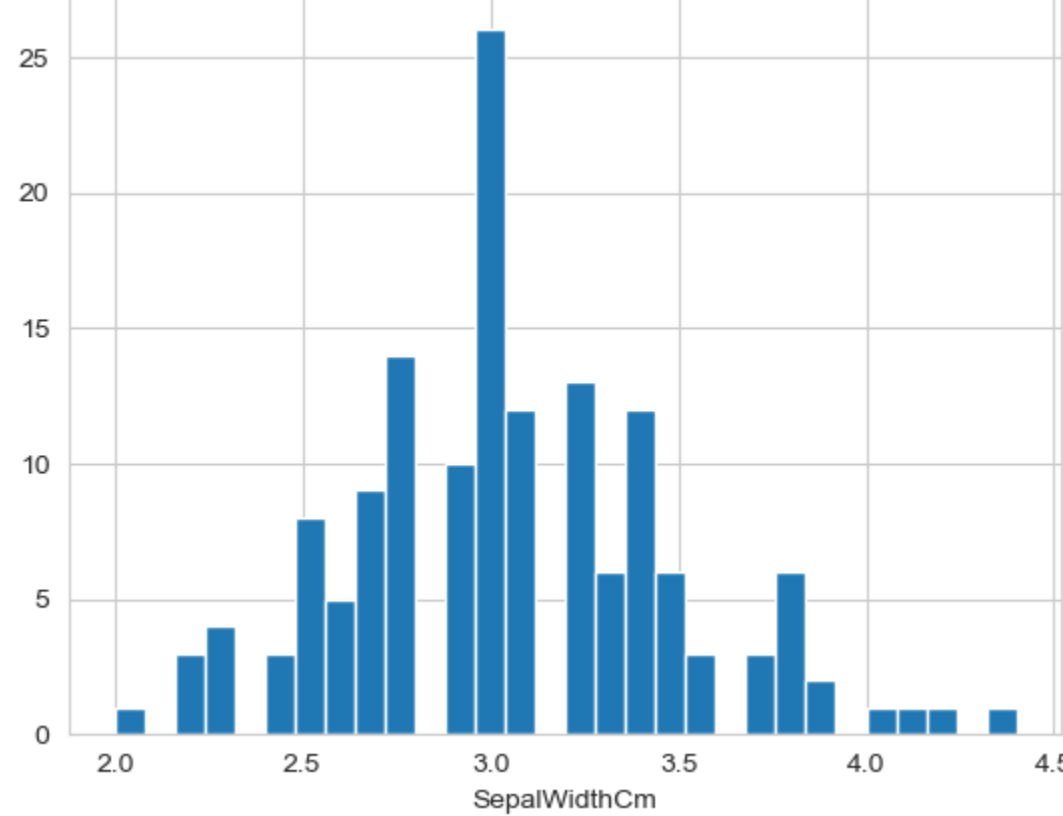
In [10]: sns.set_style('whitegrid')
Iris['SepalLengthCm'].hist(bins=30)
plt.xlabel('SepalLengthCm')

Out[10]:
```



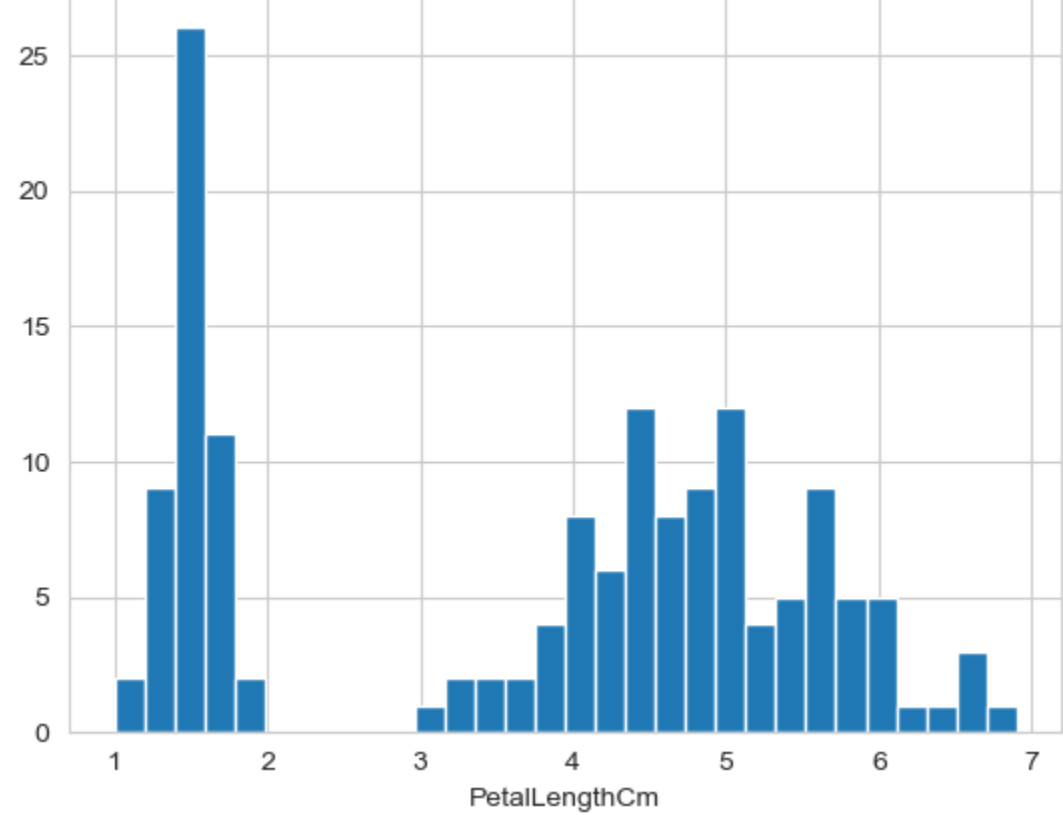
```
In [11]: sns.set_style('whitegrid')
Iris['SepalWidthCm'].hist(bins=30)
plt.xlabel('SepalWidthCm')

Out[11]:
```



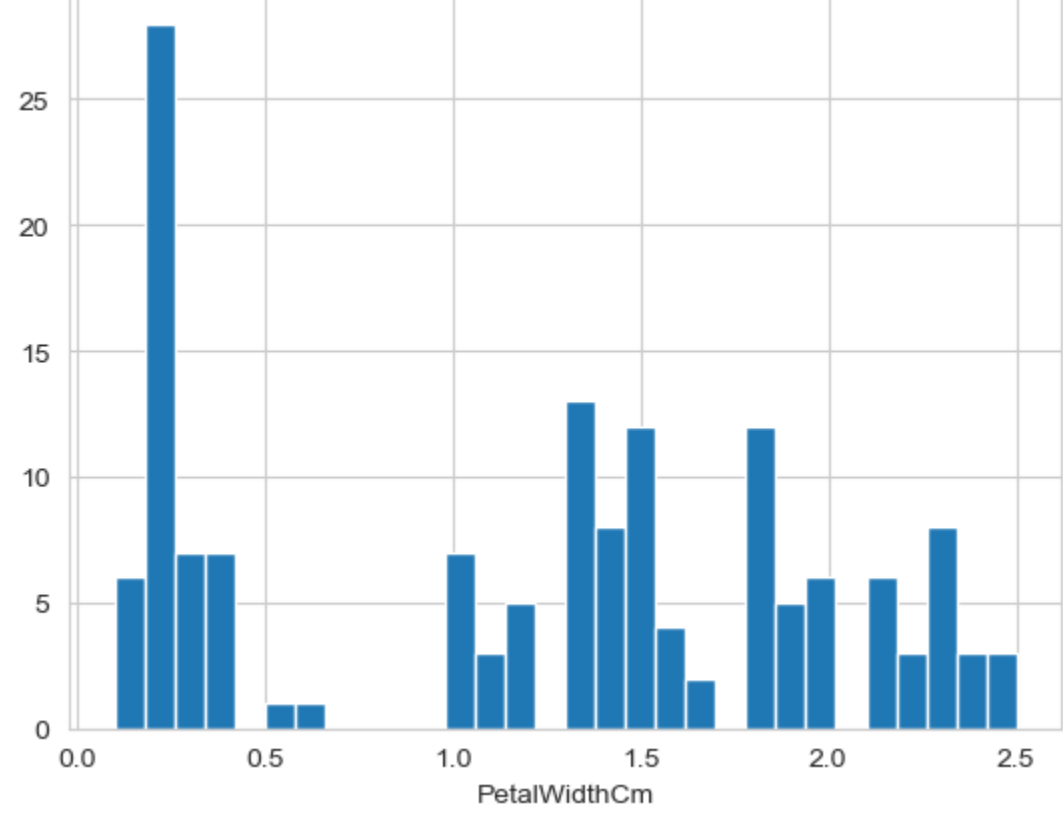
```
In [12]: sns.set_style('whitegrid')
Iris['PetalLengthCm'].hist(bins=30)
plt.xlabel('PetalLengthCm')

Out[12]:
```



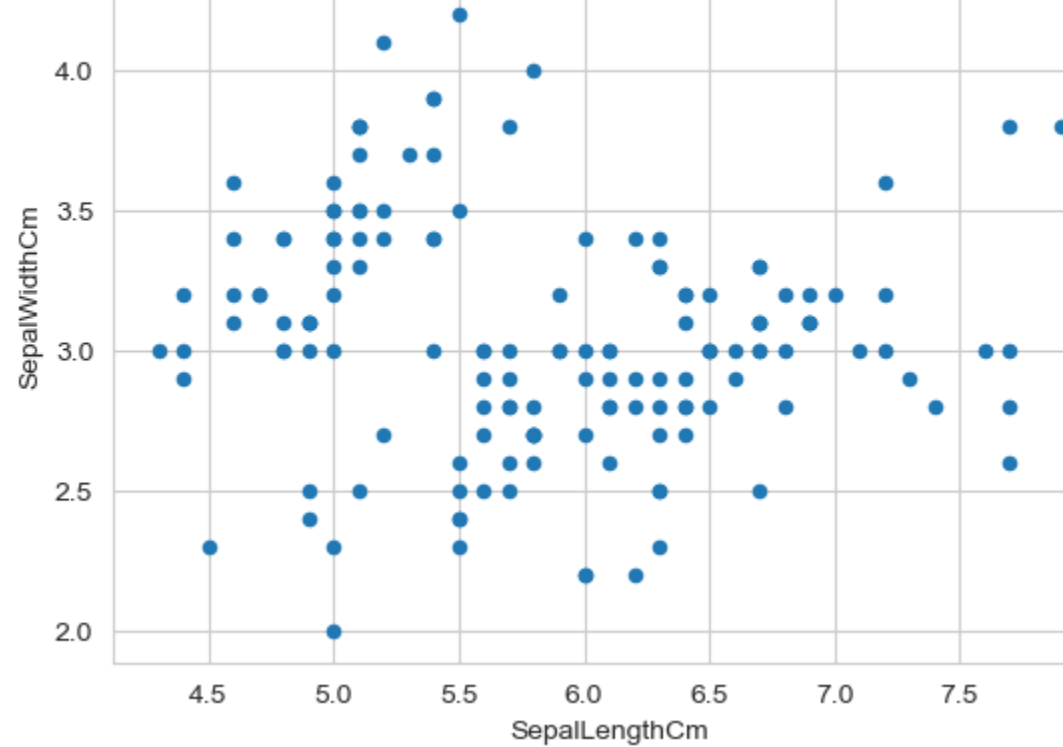
```
In [13]: sns.set_style('whitegrid')
Iris['PetalWidthCm'].hist(bins=30)
plt.xlabel('PetalWidthCm')

Out[13]:
```



```
In [14]: Iris.plot.scatter(x = "SepalLengthCm", y = "SepalWidthCm")

Out[14]: <Axes: xlabel='SepalLengthCm', ylabel='SepalWidthCm'>
```



```
In [ ]: #SPLITTING DATA

In [17]: x=Data
y=Iris['Species']

In [18]: x_train, x_test, y_train, y_test = train_test_split(x, y, test_size=0.33)

In [28]: #MODEL BUILDING BY LOGISTIC REGRESSION ALGO

In [29]: reg = LogisticRegression()

In [30]: reg.fit(x_train,y_train)

Out[30]: LogisticRegression

In [31]: # CLASSIFICATION OF REPORT

In [32]: prediction = reg.predict(x_test)

In [33]: print(classification_report(y_test,prediction))
```

	precision	recall	f1-score	support
Iris-setosa	1.00	1.00	1.00	16
Iris-versicolor	1.00	1.00	1.00	20
Iris-virginica	1.00	1.00	1.00	14
accuracy			1.00	50
macro avg	1.00	1.00	1.00	50
weighted avg	1.00	1.00	1.00	50

```
In [ ]:
```