

Importing Libraries

```
In [1]: 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
```

Loading Dataset

```
In [2]: iris = pd.read_csv('iris.csv')
iris
```

Out[2]:

	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

Data Exploration

In [3]: iris.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 150 entries, 0 to 149
Data columns (total 6 columns):
#   Column          Non-Null Count  Dtype
---  -
0   Id               150 non-null    int64
1   SepalLengthCm    150 non-null    float64
2   SepalWidthCm     150 non-null    float64
3   PetalLengthCm    150 non-null    float64
4   PetalWidthCm     150 non-null    float64
5   Species          150 non-null    object
dtypes: float64(4), int64(1), object(1)
memory usage: 7.2+ KB
```

Features Extraction

In [4]: features = iris[['Id', 'SepalLengthCm', 'SepalWidthCm', 'PetalLengthCm', 'PetalWidthCm']

In [5]: features

Out[5]:

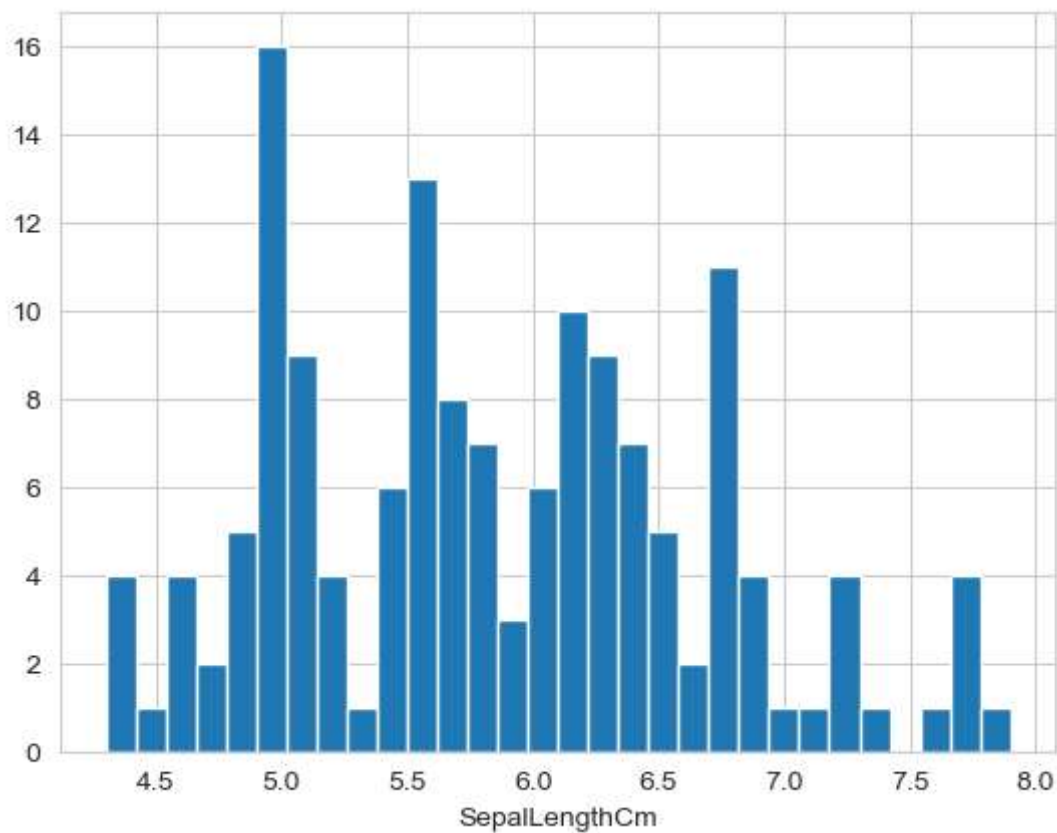
	Id	SepalLengthCm	SepalWidthCm	PetalLengthCm	PetalWidthCm
0	1	5.1	3.5	1.4	0.2
1	2	4.9	3.0	1.4	0.2
2	3	4.7	3.2	1.3	0.2
3	4	4.6	3.1	1.5	0.2
4	5	5.0	3.6	1.4	0.2
...
145	146	6.7	3.0	5.2	2.3
146	147	6.3	2.5	5.0	1.9
147	148	6.5	3.0	5.2	2.0
148	149	6.2	3.4	5.4	2.3
149	150	5.9	3.0	5.1	1.8

150 rows × 5 columns

Data Visualization

```
In [6]: sns.set_style('whitegrid')
iris['SepalLengthCm'].hist(bins=30)
plt.xlabel('SepalLengthCm')
```

```
Out[6]: Text(0.5, 0, 'SepalLengthCm')
```

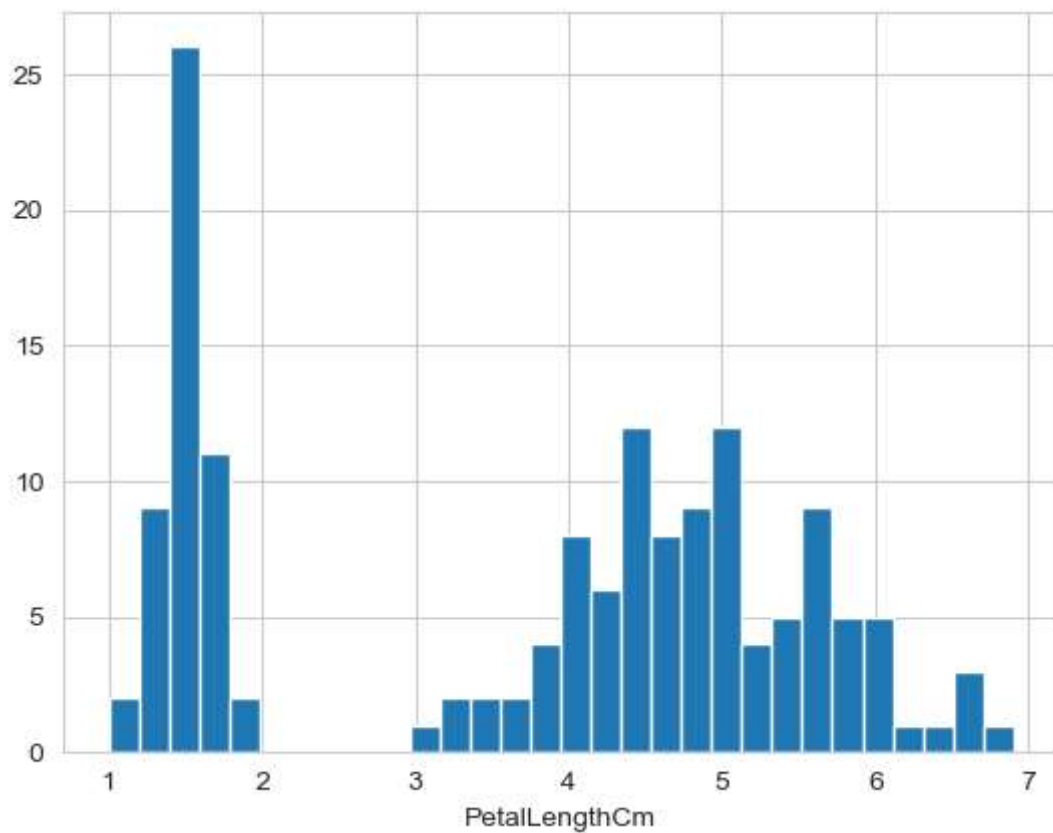


```
In [7]: sns.set_style('whitegrid')
iris['SepalWidthCm'].hist(bins=30)
plt.xlabel('SepalWidthCm')
```

...

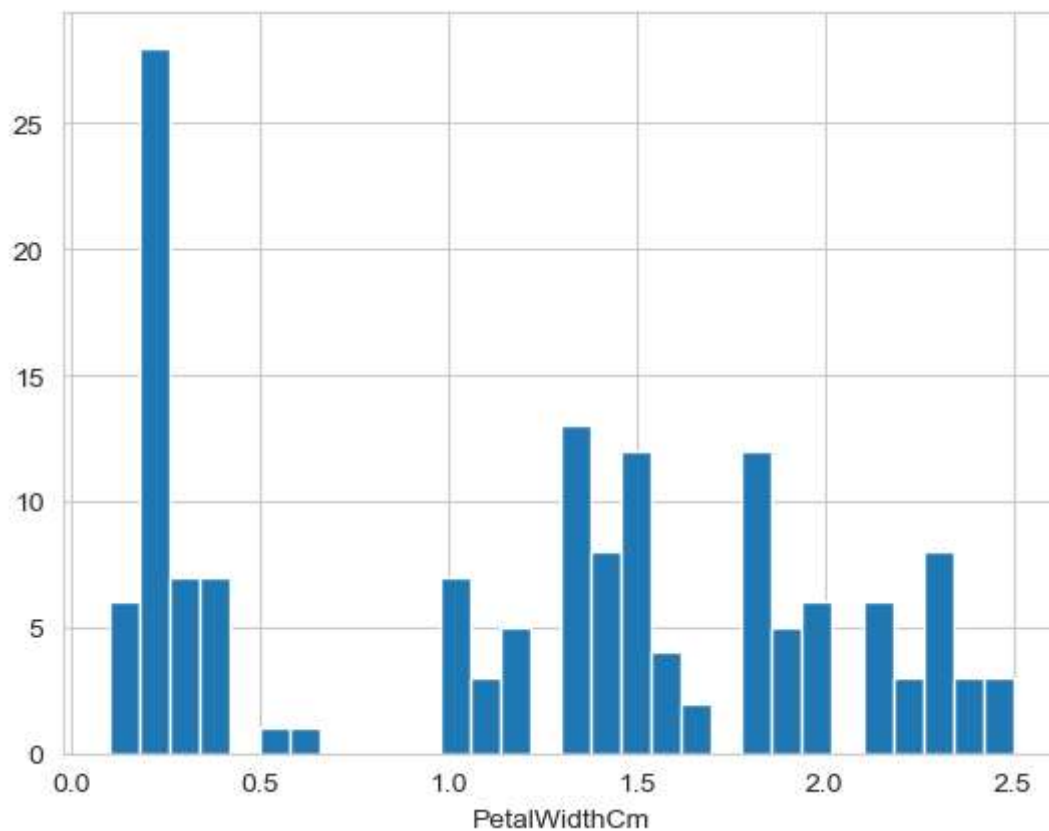
```
In [8]: sns.set_style('whitegrid')  
iris['PetalLengthCm'].hist(bins=30)  
plt.xlabel('PetalLengthCm')
```

```
Out[8]: Text(0.5, 0, 'PetalLengthCm')
```



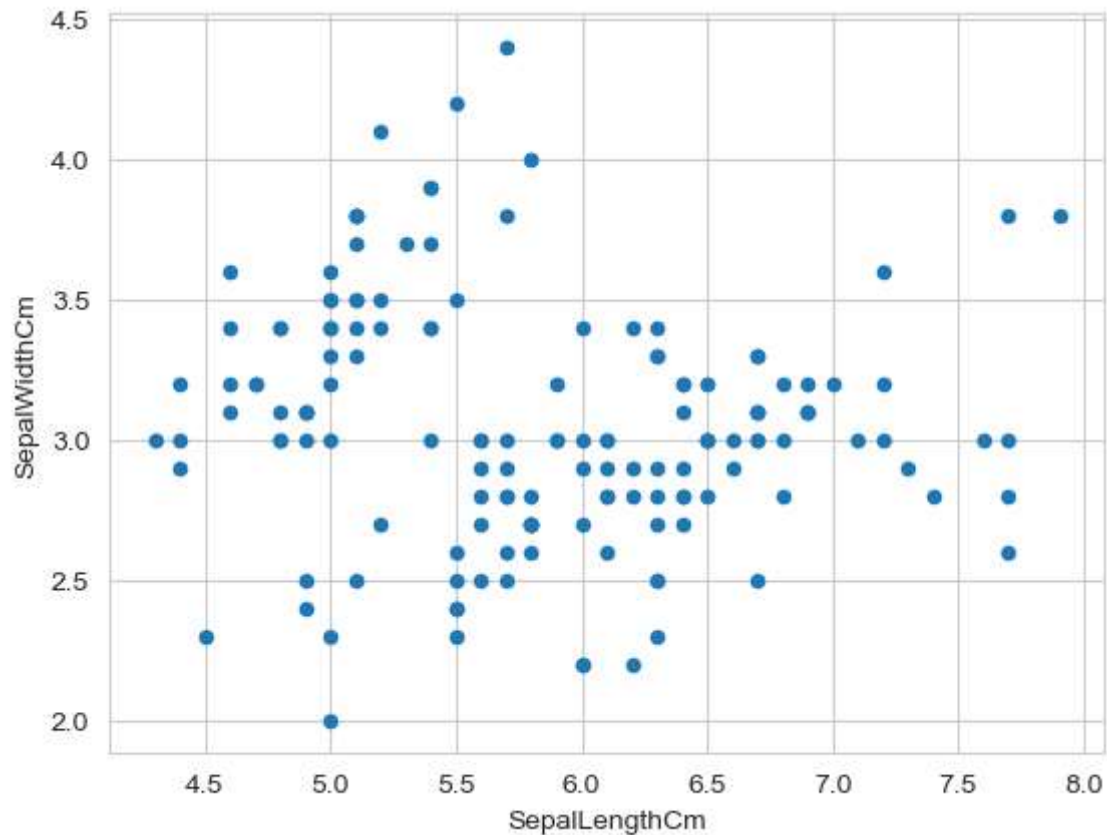
```
In [9]: sns.set_style('whitegrid')  
iris['PetalWidthCm'].hist(bins=30)  
plt.xlabel('PetalWidthCm')
```

```
Out[9]: Text(0.5, 0, 'PetalWidthCm')
```



```
In [10]: iris.plot.scatter(x = "SepalLengthCm", y = "SepalWidthCm")
```

```
Out[10]: <Axes: xlabel='SepalLengthCm', ylabel='SepalWidthCm'>
```



Splitting data for Training & Testing

```
In [11]: x=features  
y=iris['Species']
```

```
In [12]: x_train, x_test, y_train, y_test = train_test_split(x, y, test_size=0.33)
```

Model Building By Logistic Regression Algo

```
In [13]: reg = LogisticRegression()
```

```
In [14]: reg.fit(x_train,y_train)
```

```
Out[14]: LogisticRegression
```

Classification Report

```
In [15]: prediction = reg.predict(x_test)
```

```
In [16]: print(classification_report(y_test,prediction))
```

	precision	recall	f1-score	support
Iris-setosa	1.00	1.00	1.00	15
Iris-versicolor	1.00	1.00	1.00	21
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