import pandas as pd import numpy as np import seaborn as sns import matplotlib.pyplot as plt data =pd.read_csv("/content/Iris.csv") data.head() \overline{z} \blacksquare SepalWidthCm PetalLengthCm SepalLengthCm PetalWidthCm Species 0 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 4.6 3.1 1.5 0.2 Iris-setosa 3.6 0.2 Iris-setosa Generate code with data ? View recommended plots Next steps: data.tail() $\overline{2}$ H. SepalLengthCm SepalWidthCm PetalLengthCm PetalWidthCm **Species 145** 146 6.7 3.0 5.2 2.3 Iris-virginica ıl. 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 34 5.4 2.3 Iris-virginica 149 150 5.9 3.0 5 1 1.8 Iris-virginica data.shape **(150, 6)** data.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 SepalLengthCm 150 non-null 1 float64 SepalWidthCm 150 non-null float64 PetalLengthCm 150 non-null float64 PetalWidthCm 150 non-null float64 5 Species 150 non-null object dtypes: float64(4), int64(1), object(1) memory usage: 7.2+ KBdata.describe() $\overline{2}$ 扁 ${\bf SepalWidthCm} \quad {\bf PetalLengthCm}$ PetalWidthCm Id SepalLengthCm 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 1.000000 4.300000 2.000000 1.000000 0.100000 min 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 6 900000 150.000000 7 900000 4 400000 2 500000 max df.isnull().sum() Ιd

```
SepalLengthCm 0
SepalWidthCm 0
PetalLengthCm 0
PetalWidthCm 0
Species 0
dtype: int64
```

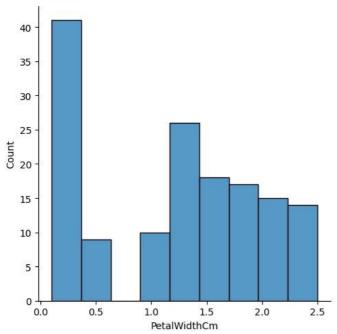
data['SepalLengthCm'].max()

₹ 7.9

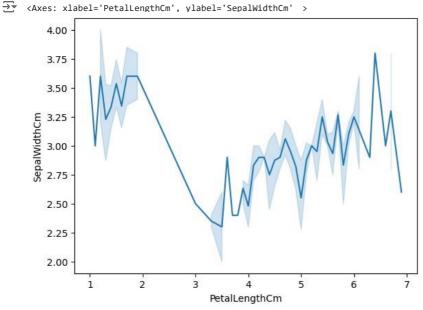
data.columns

sns.displot(data['PetalWidthCm'])

<seaborn.axisgrid.FacetGrid at 0x7f1b86cb 6200>

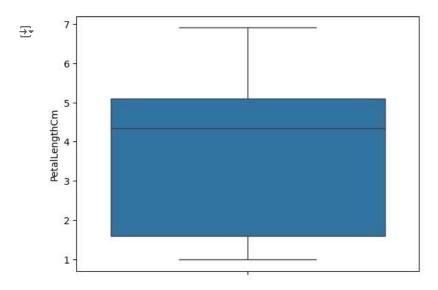


sns.lineplot(x=data['PetalLengthCm'],y=data['SepalWidthCm'])



sns.boxplot(data['PetalLengthCm'])

<Axes: ylabel='PetalLengthCm'>



numeric_columns=['SepalLengthCm','SepalWidthCm','PetalLengthCm','PetalWidthCm']

mean=data[numeric_columns].mean() mean

SepalLengthCm 5.843333
SepalWidthCm 3.054000
PetalLengthCm 3.758667
PetalWidthCm 1.198667
dtype: float64

median=data[numeric_columns].median() median

SepalLengthCm 5.80
SepalWidthCm 3.00
PetalLengthCm 4.35
PetalWidthCm 1.30
dtype: float64

mode=data[numeric_columns].mode() mode

₹		SepalLengthCm	SepalWidthCm	PetalLengthCm	PetalWidthCm	==
	0	5.0	3.0	1.5	0.2	+/

x=data.drop('Species',axis=1)

y=data['Species']

from sklearn.model_selection import train_test_split

 $\verb|x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.2,random_state=12)|$

from sklearn import svm

model=svm.SVC(kernel='linear')

 ${\tt model=model.fit(x_train,y_train)}$

 ${\tt predict=model.predict(x_test)}$

from sklearn import metrics
metrics.accuracy_score(predict,y_test)

1.0