```
In [1]: import pandas as pd
   import numpy as np
   import matplotlib.pyplot as plt
   import seaborn as sns
   import warnings
   warnings.filterwarnings('ignore')
```

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

In [3]: iris

Out[3]:

0 5.1 3.5 1.4 0.2 Iris-setos	
1 4.9 3.0 1.4 0.2 Iris-setos	sa
2 4.7 3.2 1.3 0.2 Iris-setos	sa
3 4.6 3.1 1.5 0.2 Iris-setos	sa
4 5.0 3.6 1.4 0.2 Iris-setos	sa
145 6.7 3.0 5.2 2.3 Iris-virginic	ca
146 6.3 2.5 5.0 1.9 Iris-virginic	ca
147 6.5 3.0 5.2 2.0 Iris-virginic	ca
148 6.2 3.4 5.4 2.3 Iris-virginio	ca
149 5.9 3.0 5.1 1.8 Iris-virginic	ca

150 rows × 5 columns

In [4]: iris.shape

Out[4]: (150, 5)

In [5]: iris.describe()

Out[5]:

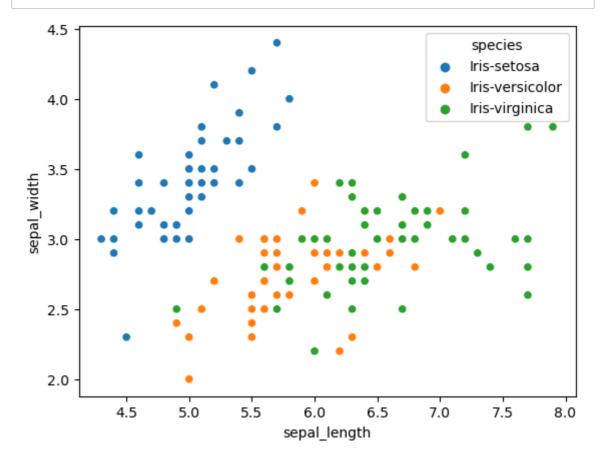
	sepal_length	sepal_width	petal_length	petal_width
count	150.000000	150.000000	150.000000	150.000000
mean	5.843333	3.054000	3.758667	1.198667
std	0.828066	0.433594	1.764420	0.763161
min	4.300000	2.000000	1.000000	0.100000
25%	5.100000	2.800000	1.600000	0.300000
50%	5.800000	3.000000	4.350000	1.300000
75%	6.400000	3.300000	5.100000	1.800000
max	7.900000	4.400000	6.900000	2.500000

In [6]: | iris.groupby('species').mean()

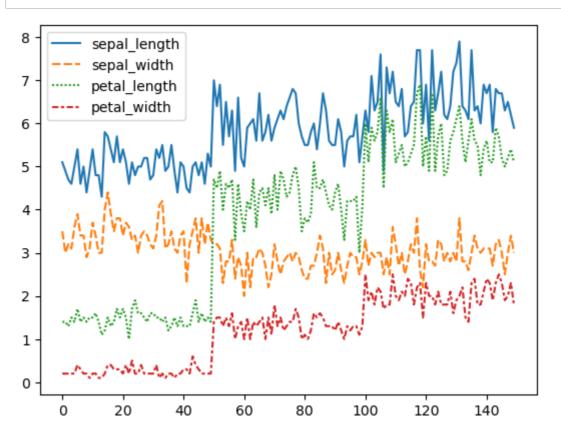
Out[6]: sepal_length sepal_width petal_length petal_width

species				
Iris-setosa	5.006	3.418	1.464	0.244
Iris-versicolor	5.936	2.770	4.260	1.326
Iris-virginica	6.588	2.974	5.552	2.026

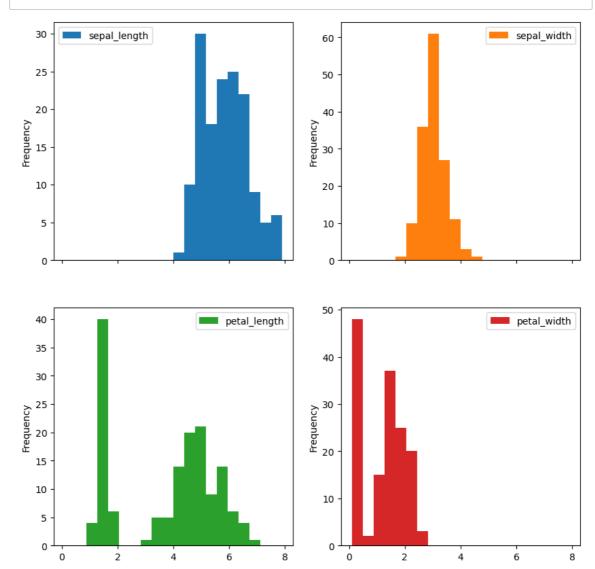
In [7]: sns.scatterplot(x='sepal_length', y='sepal_width', hue='species', data=iris
plt.show()



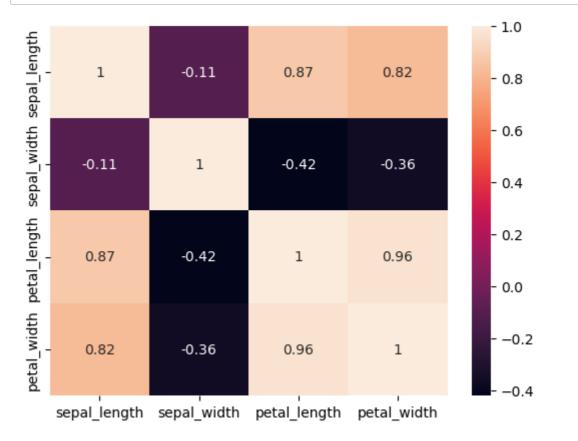
In [8]: sns.lineplot(data=iris.drop(['species'], axis=1))
 plt.show()

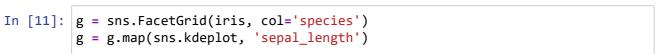


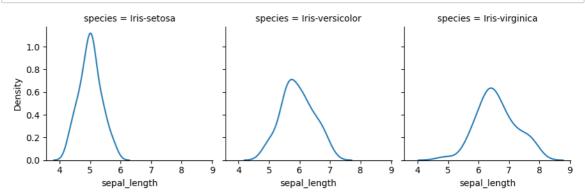
In [9]: iris.plot.hist(subplots=True, layout=(2,2), figsize=(10, 10), bins=20)
 plt.show()



```
In [10]: sns.heatmap(iris.corr(), annot=True)
plt.show()
```

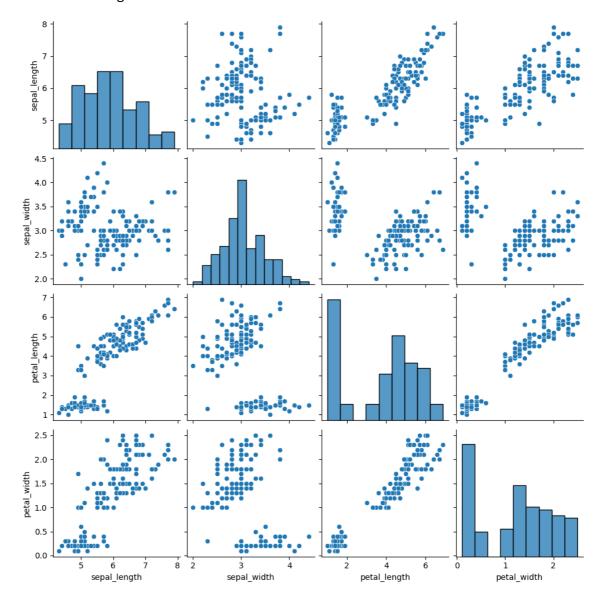




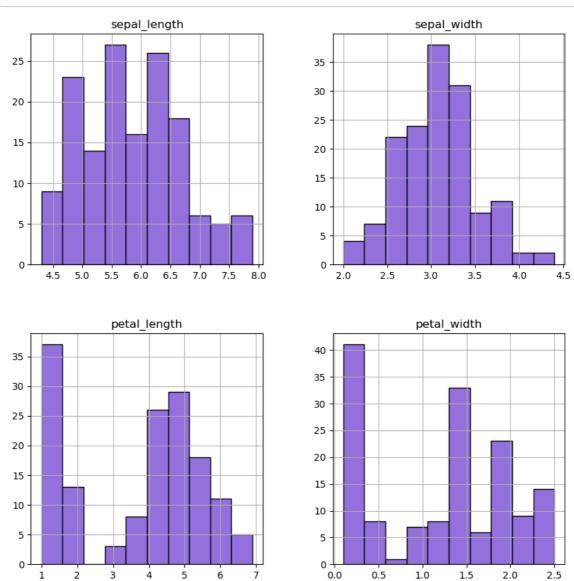


In [12]: sns.pairplot(iris)

Out[12]: <seaborn.axisgrid.PairGrid at 0x1cdc968cd90>



In [13]: iris.hist(color= 'mediumpurple' ,edgecolor='black',figsize=(10,10))
 plt.show()



In [14]: iris.corr().style.background_gradient(cmap='coolwarm').set_precision(2)

Out[14]:

	sepal_length	sepal_width	petal_length	petal_width
sepal_length	1.00	-0.11	0.87	0.82
sepal_width	-0.11	1.00	-0.42	-0.36
petal_length	0.87	-0.42	1.00	0.96
petal_width	0.82	-0.36	0.96	1.00

In [15]: from sklearn.model_selection import train_test_split
 from sklearn import metrics
 from sklearn.metrics import accuracy_score

```
In [16]: | x = iris.drop('species', axis=1)
         y= iris.species
         x_train, x_test, y_train, y_test = train_test_split(x, y, test_size=0.4, ra
In [17]: from sklearn.neighbors import KNeighborsClassifier
         knn = KNeighborsClassifier(n_neighbors=5, p=2, metric='minkowski')
         knn.fit(x_train, y_train)
         knn.score(x_test, y_test)
Out[17]: 0.966666666666667
In [18]: | from sklearn.linear_model import LogisticRegression
         logreg = LogisticRegression()
         logreg.fit(x, y)
         y_pred = logreg.predict(x)
         print(metrics.accuracy_score(y, y_pred))
         0.9733333333333334
In [19]: from sklearn.svm import SVC
         svm = SVC(kernel='rbf', random_state=0, gamma=.10, C=1.0)
         svm.fit(x_train, y_train)
         svm.score(x_test, y_test)
Out[19]: 0.9833333333333333
In [20]: |from sklearn.tree import DecisionTreeClassifier
         dtree = DecisionTreeClassifier()
         dtree.fit(x_train, y_train)
         dtree.score(x_test, y_test)
Out[20]: 0.966666666666667
 In [ ]:
```