


```
import pandas as pd
import seaborn as sns
```

```
dt = pd.read_csv('/content/iris.csv')
```


```
dt
```



	sepal.length	sepal.width	petal.length	petal.width	variety
0	5.1	3.5	1.4	0.2	Setosa
1	4.9	3.0	1.4	0.2	Setosa
2	4.7	3.2	1.3	0.2	Setosa
3	4.6	3.1	1.5	0.2	Setosa
4	5.0	3.6	1.4	0.2	Setosa
...	...	...	...	...	...
145	6.7	3.0	5.2	2.3	Virginica
146	6.3	2.5	5.0	1.9	Virginica
147	6.5	3.0	5.2	2.0	Virginica
148	6.2	3.4	5.4	2.3	Virginica
149	5.9	3.0	5.1	1.8	Virginica


150 rows × 5 columns

```
dt.describe()
```



	sepal.length	sepal.width	petal.length	petal.width
count	150.000000	150.000000	150.000000	150.000000
mean	5.843333	3.057333	3.758000	1.199333
std	0.828066	0.435866	1.765298	0.762238
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

```
dt.info()
```



```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 150 entries, 0 to 149
Data columns (total 5 columns):
#   Column          Non-Null Count  Dtype
---  -
0   sepal.length    150 non-null   float64
1   sepal.width     150 non-null   float64
2   petal.length    150 non-null   float64
3   petal.width     150 non-null   float64
4   variety         150 non-null   object
dtypes: float64(4), object(1)
memory usage: 6.0+ KB
```

```
dt['variety'].unique()
```



```
array(['Setosa', 'Versicolor', 'Virginica'], dtype=object)
```

```
from sklearn.preprocessing import LabelEncoder
```

```
l = LabelEncoder()
```

```
dt['variety'] = l.fit_transform(dt['variety'])
```

```
dt['variety'].unique()
```

```
array([0, 1, 2])
```

```
dt
```

```

sepal.length  sepal.width  petal.length  petal.width  variety
0            5.1          3.5           1.4           0.2          0
1            4.9          3.0           1.4           0.2          0
2            4.7          3.2           1.3           0.2          0
3            4.6          3.1           1.5           0.2          0
4            5.0          3.6           1.4           0.2          0
...          ...          ...           ...           ...          ...
145           6.7          3.0           5.2           2.3          2
146           6.3          2.5           5.0           1.9          2
147           6.5          3.0           5.2           2.0          2
148           6.2          3.4           5.4           2.3          2
149           5.9          3.0           5.1           1.8          2

```

150 rows × 5 columns

```
from sklearn.model_selection import train_test_split
```

```
x = dt.drop(['variety'],axis=1)
```

```
y = dt['variety']
```

```
x
```

```

sepal.length  sepal.width  petal.length  petal.width
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
4            5.0          3.6           1.4           0.2
...          ...          ...           ...           ...
145           6.7          3.0           5.2           2.3
146           6.3          2.5           5.0           1.9
147           6.5          3.0           5.2           2.0
148           6.2          3.4           5.4           2.3
149           5.9          3.0           5.1           1.8

```

150 rows × 4 columns

```
dt.corr()
```

```

sepal.length  sepal.width  petal.length  petal.width  variety
sepal.length    1.000000   -0.117570    0.871754    0.817941    0.782561
sepal.width    -0.117570    1.000000   -0.428440   -0.366126   -0.426658
petal.length    0.871754   -0.428440    1.000000    0.962865    0.949035
petal.width     0.817941   -0.366126    0.962865    1.000000    0.956547
variety         0.782561   -0.426658    0.949035    0.956547    1.000000

```

```
xtrain, xtest, ytrain, ytest = train_test_split(x,y, test_size=0.25)
```

xtest




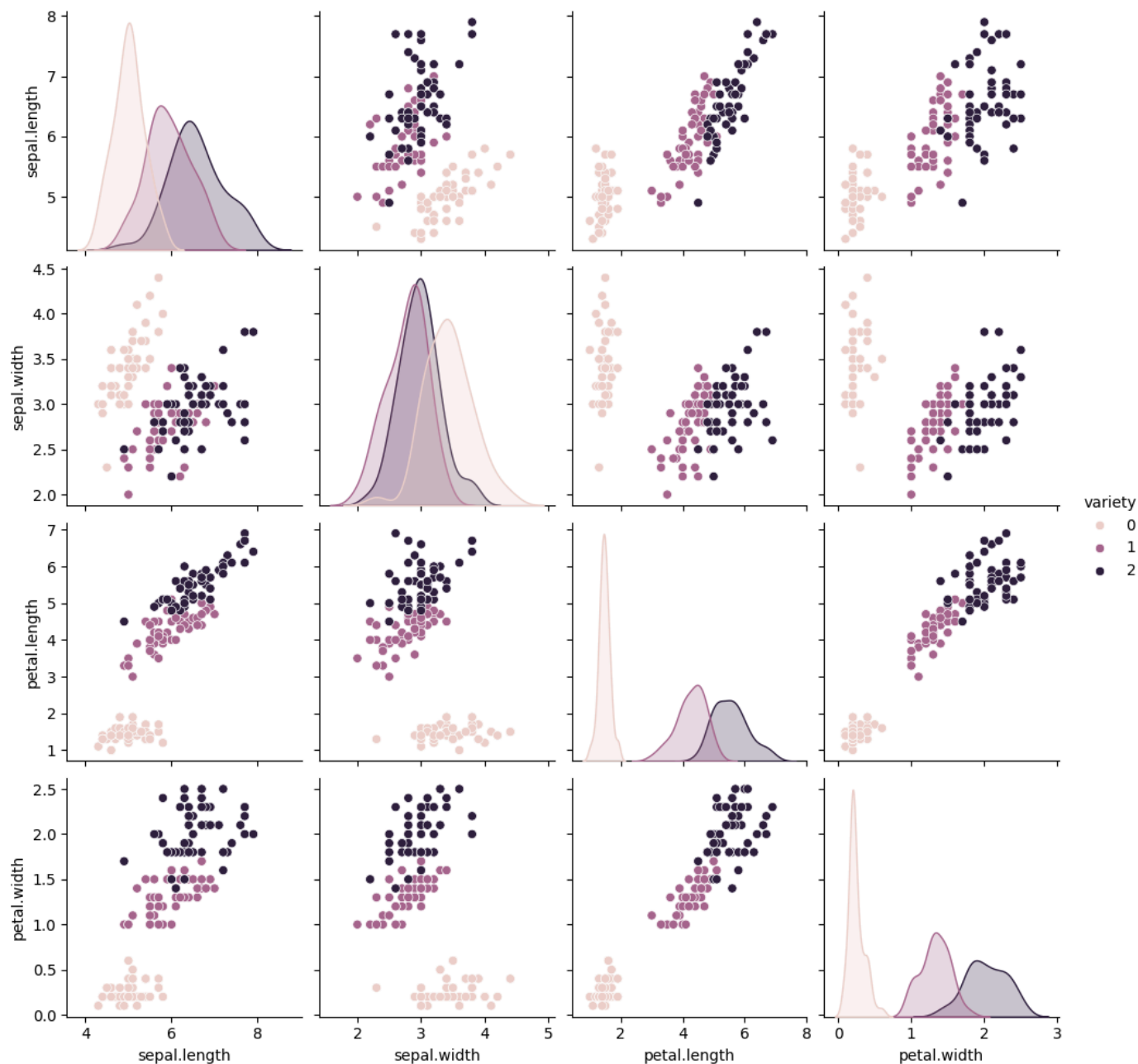
	sepal.length	sepal.width	petal.length	petal.width
--	--------------	-------------	--------------	-------------

80	5.5	2.4	3.8	1.1
67	5.8	2.7	4.1	1.0
6	4.6	3.4	1.4	0.3
103	6.3	2.9	5.6	1.8
118	7.7	2.6	6.9	2.3
127	6.1	3.0	4.9	1.8
111	6.4	2.7	5.3	1.9
141	6.9	3.1	5.1	2.3
0	5.1	3.5	1.4	0.2
104	6.5	3.0	5.8	2.2
89	5.5	2.5	4.0	1.3
149	5.9	3.0	5.1	1.8
30	4.8	3.1	1.6	0.2
144	6.7	3.3	5.7	2.5
52	6.9	3.1	4.9	1.5
117	7.7	3.8	6.7	2.2
142	5.8	2.7	5.1	1.9
62	6.0	2.2	4.0	1.0
131	7.9	3.8	6.4	2.0
143	6.8	3.2	5.9	2.3
107	7.3	2.9	6.3	1.8
147	6.5	3.0	5.2	2.0
51	6.4	3.2	4.5	1.5
83	6.0	2.7	5.1	1.6
124	6.7	3.3	5.7	2.1
135	7.7	3.0	6.1	2.3
69	5.6	2.5	3.9	1.1
113	5.7	2.5	5.0	2.0
81	5.5	2.4	3.7	1.0
121	5.6	2.8	4.9	2.0
137	6.4	3.1	5.5	1.8

sns.pairplot(dt, hue='variety')

60	5.0	2.0	3.5	1.0
74	6.4	2.9	4.3	1.3
2	4.7	3.2	1.3	0.2
39	5.1	3.4	1.5	0.2
106	4.9	2.5	4.5	1.7
138	6.0	3.0	4.8	1.8




 <seaborn.axisgrid.PairGrid at 0x7b8408561250>



```
from sklearn.linear_model import LogisticRegression
```

```
lr = LogisticRegression()
```


```
lr.fit(xtrain,ytrain)
```

 `LogisticRegression`    
`LogisticRegression()`

```
ypred = lr.predict(xtest)
```

```
from sklearn.metrics import accuracy_score
```

```
accuracy_score(ytest,ypred)
```

 0.8947368421052632

```
import pickle
```

```
pickle.dump(lr,open('iris.pkl','wb'))
```

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