

```
In [98]: import pandas as pd
import numpy as np
from matplotlib import pyplot as plt
import seaborn as sns
```

```
In [99]: df=pd.read_csv("IRIS.csv")
```

```
In [100]: df.head()
```

```
Out[100]:
```

| | sepal_length | sepal_width | petal_length | petal_width | species |
|---|--------------|-------------|--------------|-------------|-------------|
| 0 | 5.1 | 3.5 | 1.4 | 0.2 | Iris-setosa |
| 1 | 4.9 | 3.0 | 1.4 | 0.2 | Iris-setosa |
| 2 | 4.7 | 3.2 | 1.3 | 0.2 | Iris-setosa |
| 3 | 4.6 | 3.1 | 1.5 | 0.2 | Iris-setosa |
| 4 | 5.0 | 3.6 | 1.4 | 0.2 | Iris-setosa |

```
In [101]: df.tail()
```

```
Out[101]:
```

| | sepal_length | sepal_width | petal_length | petal_width | species |
|-----|--------------|-------------|--------------|-------------|----------------|
| 145 | 6.7 | 3.0 | 5.2 | 2.3 | Iris-virginica |
| 146 | 6.3 | 2.5 | 5.0 | 1.9 | Iris-virginica |
| 147 | 6.5 | 3.0 | 5.2 | 2.0 | Iris-virginica |
| 148 | 6.2 | 3.4 | 5.4 | 2.3 | Iris-virginica |
| 149 | 5.9 | 3.0 | 5.1 | 1.8 | Iris-virginica |

```
In [102]: df.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   species         150 non-null   object  
dtypes: float64(4), object(1)
memory usage: 6.0+ KB
```

```
In [103]: df.describe()
```

```
Out[103]:
```

| | 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 [104]: df.isnull().sum()
```

```
Out[104]: sepal_length    0
sepal_width    0
petal_length    0
petal_width    0
species        0
dtype: int64
```

```
In [105]: df.duplicated()
```

```
Out[105]: 0      False
1      False
2      False
3      False
4      False
...
145    False
146    False
147    False
148    False
149    False
Length: 150, dtype: bool
```

```
In [106]: df[df.duplicated()]
```

```
Out[106]:
```

| | sepal_length | sepal_width | petal_length | petal_width | species |
|-----|--------------|-------------|--------------|-------------|----------------|
| 34 | 4.9 | 3.1 | 1.5 | 0.1 | Iris-setosa |
| 37 | 4.9 | 3.1 | 1.5 | 0.1 | Iris-setosa |
| 142 | 5.8 | 2.7 | 5.1 | 1.9 | Iris-virginica |

```
In [107]: df.drop_duplicates()
```

```
Out[107]:
```

| | sepal_length | sepal_width | petal_length | petal_width | species |
|-----|--------------|-------------|--------------|-------------|----------------|
| 0 | 5.1 | 3.5 | 1.4 | 0.2 | Iris-setosa |
| 1 | 4.9 | 3.0 | 1.4 | 0.2 | Iris-setosa |
| 2 | 4.7 | 3.2 | 1.3 | 0.2 | Iris-setosa |
| 3 | 4.6 | 3.1 | 1.5 | 0.2 | Iris-setosa |
| 4 | 5.0 | 3.6 | 1.4 | 0.2 | Iris-setosa |
| ... | ... | ... | ... | ... | ... |
| 145 | 6.7 | 3.0 | 5.2 | 2.3 | Iris-virginica |
| 146 | 6.3 | 2.5 | 5.0 | 1.9 | Iris-virginica |
| 147 | 6.5 | 3.0 | 5.2 | 2.0 | Iris-virginica |
| 148 | 6.2 | 3.4 | 5.4 | 2.3 | Iris-virginica |
| 149 | 5.9 | 3.0 | 5.1 | 1.8 | Iris-virginica |

147 rows × 5 columns

```
In [108]: df.columns
```

```
Out[108]: Index(['sepal_length', 'sepal_width', 'petal_length', 'petal_width',  
                'species'],  
                dtype='object')
```

```
In [109]: df.describe(include=['O']).T
```

```
Out[109]:
```

| | count | unique | top | freq |
|---------|-------|--------|-------------|------|
| species | 150 | 3 | Iris-setosa | 50 |

```
In [110]: df['species'].unique()
```

```
Out[110]: array(['Iris-setosa', 'Iris-versicolor', 'Iris-virginica'], dtype=object)
```

```
In [111]: df['species'].replace(['Iris-setosa','Iris-virginica','Iris-versicolor'],[0,1,2])
print(df)
```

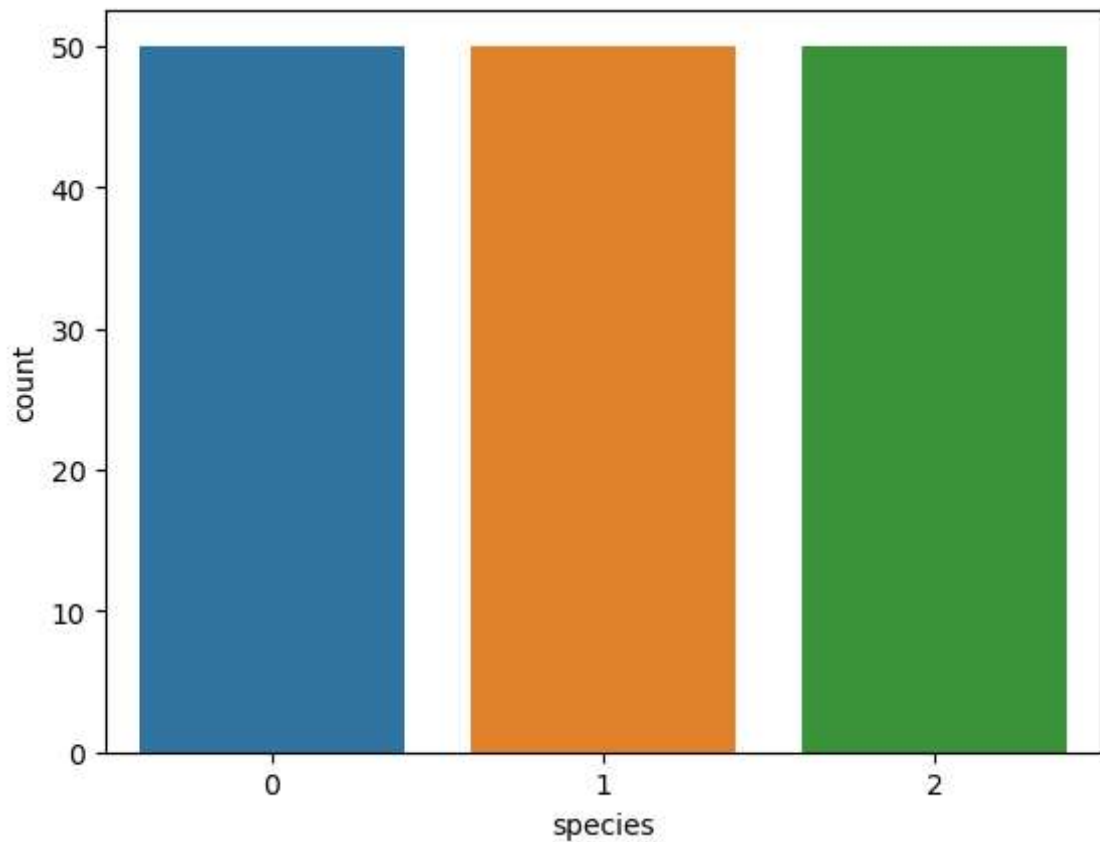
| | sepal_length | sepal_width | petal_length | petal_width | species |
|-----|--------------|-------------|--------------|-------------|---------|
| 0 | 5.1 | 3.5 | 1.4 | 0.2 | 0 |
| 1 | 4.9 | 3.0 | 1.4 | 0.2 | 0 |
| 2 | 4.7 | 3.2 | 1.3 | 0.2 | 0 |
| 3 | 4.6 | 3.1 | 1.5 | 0.2 | 0 |
| 4 | 5.0 | 3.6 | 1.4 | 0.2 | 0 |
| .. | ... | ... | ... | ... | ... |
| 145 | 6.7 | 3.0 | 5.2 | 2.3 | 1 |
| 146 | 6.3 | 2.5 | 5.0 | 1.9 | 1 |
| 147 | 6.5 | 3.0 | 5.2 | 2.0 | 1 |
| 148 | 6.2 | 3.4 | 5.4 | 2.3 | 1 |
| 149 | 5.9 | 3.0 | 5.1 | 1.8 | 1 |

[150 rows x 5 columns]

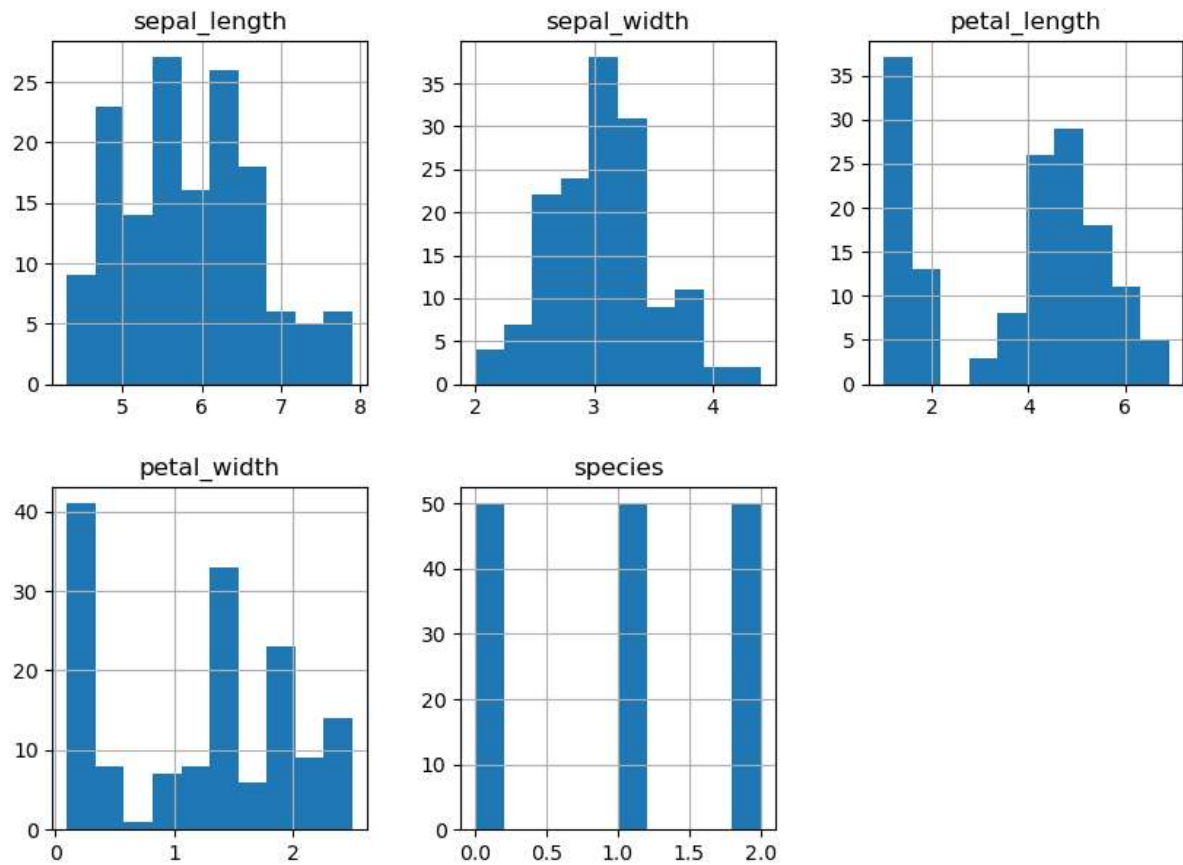
```
In [112]: print("Iris-setosa=0 Iris-virginica=1 Iris-versicolor=2")
sns.countplot(x='species',data=df)
```

Iris-setosa=0 Iris-virginica=1 Iris-versicolor=2

Out[112]: <Axes: xlabel='species', ylabel='count'>



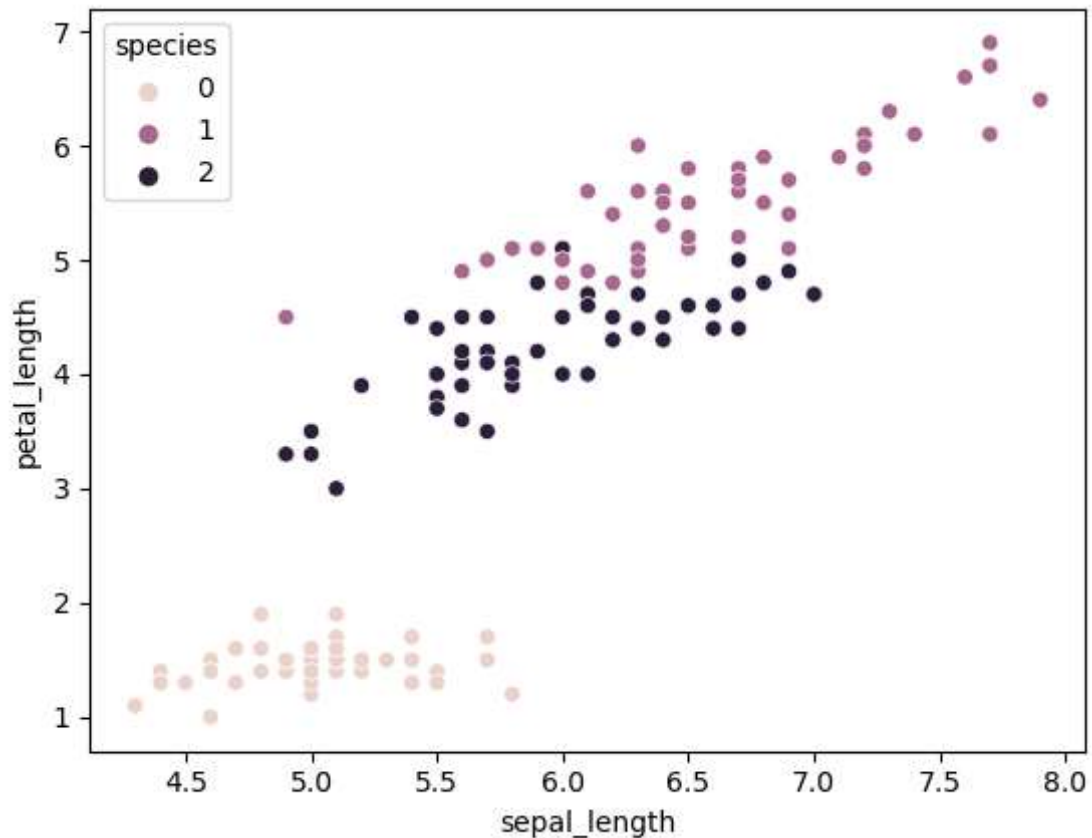
```
In [45]: df.hist(layout=(2,3),figsize=(10,7))  
plt.show()
```



```
In [113]: print("Iris-setosa=0 Iris-virginica=1 Iris-versicolor=2")
sns.scatterplot(x="sepal_length",y="petal_length",hue="species",data=df)
```

Iris-setosa=0 Iris-virginica=1 Iris-versicolor=2

```
Out[113]: <Axes: xlabel='sepal_length', ylabel='petal_length'>
```



```
In [46]: from sklearn.model_selection import train_test_split
from sklearn.metrics import accuracy_score
```

```
In [84]: x=df.drop('species',axis=1)
y=df['species']
```

```
In [85]: model_df={}
def model_val(model,x,y):
    x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.5,random_state=42)
    model.fit(x_train,y_train)
    pred=model.predict(x_test)
    print(f"{model} accuracy is {accuracy_score(y_test,pred)}")
```

```
In [89]: #Linear regression
from sklearn.linear_model import LogisticRegression
model=LogisticRegression()
model_val(model,x,y)
```

LogisticRegression() accuracy is 1.0

```
In [90]: #SVM  
from sklearn import svm  
model=svm.SVC()  
model_val(model,x,y)
```

SVC() accuracy is 0.9866666666666667

```
In [91]: #Decision Tree  
from sklearn.tree import DecisionTreeClassifier  
model=DecisionTreeClassifier()  
model_val(model,x,y)
```

DecisionTreeClassifier() accuracy is 0.9866666666666667

```
In [ ]:
```