

```
In [9]: import numpy as np
import pandas as pd
df = pd.read_csv("IRIS.csv")
df.head(10)
```

```
Out[9]:
```

	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
5	5.4	3.9	1.7	0.4	Iris-setosa
6	4.6	3.4	1.4	0.3	Iris-setosa
7	5.0	3.4	1.5	0.2	Iris-setosa
8	4.4	2.9	1.4	0.2	Iris-setosa
9	4.9	3.1	1.5	0.1	Iris-setosa

```
In [6]: df.shape
```

```
Out[6]: (150, 5)
```

```
In [13]: df_Setosa = df[:50]
df_versicolor = df[50:100]
df_versinica = df[100:]
```

```
In [18]: df_versicolor.shape
```

```
Out[18]: (50, 5)
```

```
In [19]: df_Setosa.shape
```

Out[19]: (50, 5)

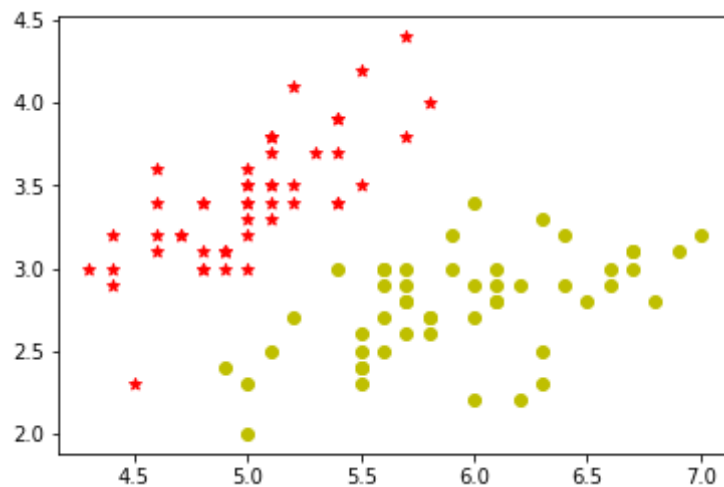
```
In [34]: from matplotlib import pyplot as plt
```

```
In [36]: x_vers = df_versicolor["sepal_length"]
```

```
In [37]: y_vers = df_versicolor["sepal_width"]
```

```
In [38]: plt.scatter(x,y,marker="o",color="y",label="Versicolor")  
plt.scatter(df_Setosa["sepal_length"],df_Setosa["sepal_width"],marker="*",color = "r",label="Setosa")
```

Out[38]: <matplotlib.collections.PathCollection at 0x15cae7c0460>



```
In [40]: from sklearn.model_selection import train_test_split  
x = df.drop(["species"],axis=1)  
y = df["species"]  
  
x_train, x_test, y_train, y_test = train_test_split(x,y,test_size=0.2,random_state=1)
```

```
In [42]: x_test.head()
```

Out[42]:

sepal_length	sepal_width	petal_length	petal_width
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	sepal_length	sepal_width	petal_length	petal_width
14	5.8	4.0	1.2	0.2
98	5.1	2.5	3.0	1.1
75	6.6	3.0	4.4	1.4
16	5.4	3.9	1.3	0.4
131	7.9	3.8	6.4	2.0

In [43]: `x_test.shape`

Out[43]: (30, 4)

In [44]: `from sklearn.svm import SVC`
`model = SVC()`

In [48]: `rissh = model.fit(x_train,y_train)`
`rissh.score(x_test,y_test)*100`

Out[48]: 96.66666666666667

In [49]: `x_test`

Out[49]:

	sepal_length	sepal_width	petal_length	petal_width
14	5.8	4.0	1.2	0.2
98	5.1	2.5	3.0	1.1
75	6.6	3.0	4.4	1.4
16	5.4	3.9	1.3	0.4
131	7.9	3.8	6.4	2.0
56	6.3	3.3	4.7	1.6
141	6.9	3.1	5.1	2.3
44	5.1	3.8	1.9	0.4

	sepal_length	sepal_width	petal_length	petal_width
29	4.7	3.2	1.6	0.2
120	6.9	3.2	5.7	2.3
94	5.6	2.7	4.2	1.3
5	5.4	3.9	1.7	0.4
102	7.1	3.0	5.9	2.1
51	6.4	3.2	4.5	1.5
78	6.0	2.9	4.5	1.5
42	4.4	3.2	1.3	0.2
92	5.8	2.6	4.0	1.2
66	5.6	3.0	4.5	1.5
31	5.4	3.4	1.5	0.4
35	5.0	3.2	1.2	0.2
90	5.5	2.6	4.4	1.2
84	5.4	3.0	4.5	1.5
77	6.7	3.0	5.0	1.7
40	5.0	3.5	1.3	0.3
125	7.2	3.2	6.0	1.8
99	5.7	2.8	4.1	1.3
33	5.5	4.2	1.4	0.2
19	5.1	3.8	1.5	0.3
73	6.1	2.8	4.7	1.2
146	6.3	2.5	5.0	1.9

```
In [51]: rissh.predict([[5.8,4.0,1.2,0.2]])    #Prediction of 14 row
```

```
Out[51]: array(['Iris-setosa'], dtype=object)
```

```
In [52]: rissh.predict([[5.7,2.8,4.1,1.3]])      #Prediction of 99 row
```

```
Out[52]: array(['Iris-versicolor'], dtype=object)
```

```
In [53]: rissh.predict([[6.3,2.5,5.0,1.9]])      #Prediction of 146 row
```

```
Out[53]: array(['Iris-virginica'], dtype=object)
```

```
In [ ]:
```