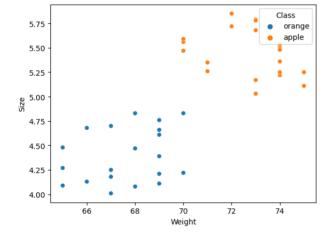
```
In [1]: import pandas as pd
    from sklearn.model_selection import train_test_split
    from sklearn.svm import SVC
    from sklearn.metrics import confusion_matrix
    from sklearn.preprocessing import LabelEncoder
```

## In [2]: data=pd.read\_csv('apples\_and\_oranges.csv') print(data)

```
Weight Size
69 4.39
69 4.21
65 4.09
72 5.85
67 4.70
73 5.68
75 5.11
74 5.36
65 4.27
73 5.79
70 5.47
74 5.53
68 4.47
74 5.22
65 4.48
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75 5.25
67 4.18
74 5.59
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68 4.08
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apple
                        73 5.03
                                                           apple
```

In [3]: import seaborn as sns
sns.scatterplot(x='Weight',y='Size',hue='Class',data=data)

## Out[3]: <AxesSubplot:xlabel='Weight', ylabel='Size'>



```
In [10]: train_set,test_set=train_test_split(data,test_size=0.2,random_state=1)
print("train:",train_set)
print("test:",test_set)
                                             Weight Size
74 5.50 apple
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In [11]: x train=train set.iloc[:,0:2].values
                         y_train=train_set.iloc[:,2].values
x_test=test_set.iloc[:,0:2].values
                         y_test=test_set.iloc[:,2].values
print(x_train,y_train)
print(x_test,y_test)
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   In [6]: c=SVC(kernel='rbf',random_state=1,C=1,gamma='auto')
                         c.fit(x_train,y_train)
  Out[6]: SVC(C=1, gamma='auto', random_state=1)
   In [7]: yp=c.predict(x_test)
print(yp)
                         ['orange' 'orange' 'apple' 'apple' 'orange' 'apple' 'orange' 'apple']
   In [8]: cm=confusion_matrix(y_test,yp)
                         print(cm)
                         accuracy=float(cm.diagonal().sum())/len(y_test)
print("model accuracy is:",accuracy*100,'%')
                         [[3 0]
[1 4]]
```

model accuracy is: 87.5 %