```
import csv
a = []
with open('enjoysport.csv', 'r') as csvfile:
    for row in csv.reader(csvfile):
        a.append(row)
    print(a)
print("\n The total number of training instances are : ",len(a))
num attribute = len(a[0])-1
print("\n The initial hypothesis is : ")
hypothesis = ['0']*num attribute
print(hypothesis)
for i in range(0, len(a)):
    if a[i][num attribute] == '1':
        for j in range(0, num_attribute):
            if hypothesis[j] == '0' or hypothesis[j] == a[i][j]:
                hypothesis[j] = a[i][j]
            else:
                hypothesis[j] = '?'
    print("\n The hypothesis for the training instance {} is :\n"
.format(i+1),hypothesis)
print("\n The Maximally specific hypothesis for the training instance is ")
print(hypothesis)
```

```
Dell@college-Grass MINGw64 /d/college Stuff/VI Sem/ML

$ python text.py
[['sky', 'airtemp', 'humidity', 'wind', 'water', 'forcast', 'enjoysport'], ['sunny', 'warm', 'normal', 'strong', 'high', 'strong', 'warm', 'change', '0'], ['sunny', 'warm', 'high', 'strong', 'cool', 'change', '1']]

The total number of training instances are: 5

The initial hypothesis is:
['0', '0', '0', '0', '0', '0']

The hypothesis for the training instance 1 is:
['0', '0', '0', '0', '0', '0']

The hypothesis for the training instance 2 is:
['sunny', 'warm', 'normal', 'strong', 'warm', 'same']

The hypothesis for the training instance 3 is:
['sunny', 'warm', '?', 'strong', 'warm', 'same']

The hypothesis for the training instance 4 is:
['sunny', 'warm', '?', 'strong', 'warm', 'same']

The hypothesis for the training instance 5 is:
['sunny', 'warm', '?', 'strong', '?', '?']

The Maximally specific hypothesis for the training instance is
['sunny', 'warm', '?', 'strong', '?', '?']
```

```
import numpy as np
from sklearn.datasets import load iris
from sklearn.model_selection import train_test_split
from sklearn.svm import SVC
from sklearn.metrics import accuracy_score, classification_report,
confusion matrix
# Load the Iris dataset
iris = load iris()
X = iris.data
y = iris.target
# Split the dataset into training and test sets
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3,
random state=42)
# Train the SVM classifier
svm = SVC(kernel='linear', C=1.0, random state=42)
svm.fit(X_train, y_train)
y_pred = svm.predict(X_test)
print(f"Accuracy: {accuracy_score(y_test, y_pred):.2f}")
print("Classification Report:")
print(classification_report(y_test, y_pred, target_names=iris.target_names))
print("\nConfusion Matrix:\n", confusion matrix(y test, y pred))
```

```
Dell@College-Grass MINGW64 /d/College Stuff/VI Sem/ML
$ python text2.py
Accuracy: 1.00
Classification Report:
             precision
                           recall f1-score
                                              support
                             1.00
                                       1.00
                                                    19
                   1.00
     setosa
                             1.00
                                       1.00
 versicolor
                   1.00
                                                    13
  virginica
                   1.00
                             1.00
                                       1.00
                                                    13
   accuracy
                                       1.00
                                                    45
  macro avq
                   1.00
                             1.00
                                       1.00
                                                    45
                                                    45
weighted avg
                   1.00
                             1.00
                                       1.00
Confusion Matrix:
[[19 0 0]
  0 13 0]
  0 0 1377
```

```
from sklearn.datasets import load_iris
from sklearn.model_selection import train_test_split
from sklearn.tree import DecisionTreeClassifier, plot_tree
from sklearn.metrics import accuracy_score, classification_report

# Load and split the Iris dataset
X, y = load_iris(return_X_y=True)
X_train, X_test, y_train, y_test = train_test_split(X, y,
test_size=0.35, random_state=42)

# Train and evaluate the Random Forest classifier
rf_classifier = DecisionTreeClassifier(
random_state=42).fit(X_train, y_train)
y_pred = rf_classifier.predict(X_test)

print(f'Accuracy: {accuracy_score(y_test, y_pred):.2f}')
print('Classification_Report:\n', classification_report(y_test,
y_pred))
```

Accuracy: 0.98

Classification Report:

	precision	recall	f1-score	support
0	1.00	1.00	1.00	19
1	0.94	1.00	0.97	17
2	1.00	0.94	0.97	17
accuracy			0.98	53
macro avg	0.98	0.98	0.98	53
weighted avg	0.98	0.98	0.98	53

6.

```
from sklearn.datasets import load_iris
from sklearn.model_selection import train_test_split
from sklearn.ensemble import RandomForestClassifier
from sklearn.metrics import accuracy_score, classification_report
# Load and split the Iris dataset
X, y = load_iris(return_X_y=True)
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.35, random_state=42)
# Train and evaluate the Random Forest classifier
rf_classifier = RandomForestClassifier(n_estimators=100, random_state=42).fit(X_train, y_train)
y_pred = rf_classifier.predict(X_test)
print(f'Accuracy: {accuracy_score(y_test, y_pred):.2f}')
print('Classification Report:\n', classification_report(y_test, y_pred))
```

<u>OUTPUT</u>

Dell@College-Grass MINGW64 /d/College Stuff/VI Sem/ML \$ python text.py Accuracy: 0.98							
Classification							
	precision	recall	f1-score	support			
	1 00	1 00	1 00	10			
0	1.00	1.00	1.00	19			
1	0.94	1.00	0.97	17			
2	1.00	0.94	0.97	17			
			0.00	5.3			
accuracy			0.98	53			
macro avg	0.98	0.98	0.98	53			
weighted avg	0.98	0.98	0.98	53			

```
from sklearn.datasets import fetch_20newsgroups
from sklearn.model selection import train test split
from sklearn.feature extraction.text import TfidfVectorizer
from sklearn.naive_bayes import MultinomialNB
from sklearn.metrics import accuracy_score, classification_report
# Load the 20 newsgroups dataset
newsgroups = fetch 20newsgroups(subset='all')
# Split the dataset into training and test sets
newsgroups.target, test_size=0.2, random_state=42)
# Transform the text data to feature vectors using TF-IDF
vectorizer = TfidfVectorizer(stop_words='english')
X_train_tfidf = vectorizer.fit_transform(X_train)
X test tfidf = vectorizer.transform(X test)
# Train the Naive Bayes classifier
clf = MultinomialNB()
clf.fit(X_train_tfidf, y_train)
y_pred = clf.predict(X_test_tfidf)
print(f"Accuracy: {accuracy score(y test, y pred):.2f}")
print("Classification Report:")
print(classification_report(y_test, y_pred,
target_names=newsgroups.target_names))
```

```
Dell@College-Grass MINGW64 /d/College Stuff/VI Sem/ML
$ python text.py
Accuracy: 0.88
Classification Report:
                            precision
                                          recall
                                                   f1-score
                                                               support
              alt.atheism
                                 0.85
                                            0.86
                                                       0.86
                                                                   151
                                            0.84
           comp.graphics
                                 0.88
                                                       0.86
                                                                   202
 comp.os.ms-windows.misc
                                 0.86
                                            0.85
                                                       0.85
                                                                   195
                                 0.65
                                            0.85
                                                       0.73
                                                                   183
comp.sys.ibm.pc.hardware
                                 0.94
                                                       0.90
   comp.sys.mac.hardware
                                            0.87
                                                                   205
                                 0.95
                                                                   215
                                            0.85
                                                       0.90
          comp.windows.x
             misc.forsale
                                 0.93
                                            0.72
                                                       0.81
                                                                   193
                                                                   196
                rec.autos
                                 0.91
                                            0.94
                                                       0.92
         rec.motorcycles
                                 0.89
                                            0.95
                                                       0.92
                                                                   168
                                 0.95
                                            0.95
                                                       0.95
                                                                   211
      rec.sport.baseball
        rec.sport.hockey
                                 0.90
                                            0.99
                                                       0.94
                                                                   198
                sci.crypt
                                 0.91
                                            0.97
                                                       0.94
                                                                   201
                                 0.92
                                            0.82
                                                       0.86
                                                                   202
         sci.electronics
                                            0.94
                                                       0.96
                                                                   194
                                 0.97
                  sci.med
                                            0.99
                                                       0.93
                                                                   189
                                 0.88
                sci.space
                                            0.99
                                                       0.83
                                                                   202
  soc.religion.christian
                                 0.72
                                            0.97
                                                                   188
                                                       0.88
      talk.politics.guns
                                 0.81
                                            0.99
   talk.politics.mideast
                                 0.94
                                                       0.97
                                                                   182
      talk.politics.misc
                                 0.96
                                            0.75
                                                       0.84
                                                                   159
      talk.religion.misc
                                 1.00
                                            0.31
                                                       0.47
                                                                   136
                                                       0.88
                                                                  3770
                 accuracy
                                                       0.87
                                                                  3770
                                 0.89
                                            0.87
                macro avg
            weighted avg
                                  0.89
                                            0.88
                                                       0.87
                                                                   3770
```

8.

```
import pandas as pd
from sklearn.model selection import train test split
from sklearn.naive bayes import GaussianNB
from sklearn.metrics import accuracy score, classification report
# Load the Dataset
url = "iris.csv"
column names = ['sepal length', 'sepal width', 'petal length', 'petal width',
'class'l
df = pd.read csv(url, header=None, names=column names)
#print(df.head())
X = df.drop('class', axis=1)
y = df['class']
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2,
random_state=42)
# Train the Model
model = GaussianNB()
model.fit(X train, y train)
# Evaluate the Model
y pred = model.predict(X test)
accuracy = accuracy_score(y_test, y_pred)
report = classification_report(y_test, y_pred)
print(f"Accuracy: {accuracy}\n")
print("Classification Report:")
print(report)
```

```
Dell@College-Grass MINGW64 /d/College Stuff/VI Sem/ML
$ python text.py
Accuracy: 1.0
Classification Report:
                 precision
                               recall f1-score
                                                   support
    Iris-setosa
                      1.00
                                 1.00
                                           1.00
                                                        10
Iris-versicolor
                                 1.00
                       1.00
                                           1.00
                                                         9
                      1.00
 Iris-virginica
                                 1.00
                                           1.00
                                                        11
                                           1.00
                                                        30
       accuracy
                       1.00
                                 1.00
                                           1.00
                                                        30
      macro avg
   weighted avg
                      1.00
                                 1.00
                                           1.00
                                                        30
```