

```
In [1]: import numpy as np

class NaiveBayes:

    def fit(self, X, y):
        n_samples, n_features = X.shape
        self._classes = np.unique(y)
        n_classes = len(self._classes)

        # calculate mean, var, and prior for each class
        self._mean = np.zeros((n_classes, n_features), dtype=np.float64)
        self._var = np.zeros((n_classes, n_features), dtype=np.float64)
        self._priors = np.zeros(n_classes, dtype=np.float64)

        for idx, c in enumerate(self._classes):
            X_c = X[y == c]
            self._mean[idx, :] = X_c.mean(axis=0)
            self._var[idx, :] = X_c.var(axis=0)
            self._priors[idx] = X_c.shape[0] / float(n_samples)

    def predict(self, X):
        y_pred = [self._predict(x) for x in X]
        return np.array(y_pred)

    def _predict(self, x):
        posteriors = []

        # calculate posterior probability for each class
        for idx, c in enumerate(self._classes):
            prior = np.log(self._priors[idx])
            posterior = np.sum(np.log(self._pdf(idx, x)))
            posterior = posterior + prior
            posteriors.append(posterior)

        # return class with the highest posterior
        return self._classes[np.argmax(posteriors)]

    def _pdf(self, class_idx, x):
        mean = self._mean[class_idx]
        var = self._var[class_idx]
        numerator = np.exp(-((x - mean) ** 2) / (2 * var))
        denominator = np.sqrt(2 * np.pi * var)
        return numerator / denominator
```

```
In [2]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
from sklearn.model_selection import train_test_split

iris = pd.read_csv("iris.csv")
```

```
In [3]: X = iris.iloc[:, :-1].values
y = iris.iloc[:, -1].values
```

```
In [4]: X_train, X_test, y_train, y_test = train_test_split(X, y, test_size = 0.2, random_state = 0)
```

```
In [5]: def accuracy(y_true, y_pred):
    accuracy = np.sum(y_true == y_pred) / len(y_true)
    return accuracy

nb = NaiveBayes()
nb.fit(X_train, y_train)
predictions = nb.predict(X_test)
```

```
In [6]: print("Naive Bayes classification accuracy", accuracy(y_test, predictions))
```

```
Naive Bayes classification accuracy 0.9666666666666667
```

```
In [7]: from sklearn.naive_bayes import GaussianNB
from sklearn.metrics import confusion_matrix
from sklearn.metrics import classification_report
classifier = GaussianNB()
classifier.fit(X_train, y_train)

y_pred = classifier.predict(X_test)

# Summary of the predictions made by the classifier
print(classification_report(y_test, y_pred))
print(confusion_matrix(y_test, y_pred))
# Accuracy score
from sklearn.metrics import accuracy_score
print('accuracy is', accuracy_score(y_pred, y_test))
```

```

precision    recall   f1-score   support
setosa       1.00     1.00     1.00      11
versicolor   0.93     1.00     0.96      13
virginica    1.00     0.83     0.91       6

accuracy          0.97     0.97     0.97      30
macro avg       0.98     0.94     0.96      30
weighted avg    0.97     0.97     0.97      30

[[11  0  0]
 [ 0 13  0]
 [ 0  1  5]]
accuracy is 0.9666666666666667

```

dibetes dataset

```

In [8]: di = pd.read_csv("diabetes_RF.csv")

In [9]: X = di.iloc[:, :-1].values
y = di.iloc[:, -1].values

In [10]: X_train, X_test, y_train, y_test = train_test_split(X, y, test_size = 0.2, random_state = 0)

In [11]: def accuracy(y_true, y_pred):
            accuracy = np.sum(y_true == y_pred) / len(y_true)
            return accuracy

nb = NaiveBayes()
nb.fit(X_train, y_train)
predictions = nb.predict(X_test)

In [12]: print("Naive Bayes classification accuracy", accuracy(y_test, predictions))
Naive Bayes classification accuracy 0.7922077922077922

In [13]: from sklearn.naive_bayes import GaussianNB
from sklearn.metrics import confusion_matrix
from sklearn.metrics import classification_report
classifier = GaussianNB()
classifier.fit(X_train, y_train)

y_pred = classifier.predict(X_test)

# Summary of the predictions made by the classifier
print(classification_report(y_test, y_pred))
print(confusion_matrix(y_test, y_pred))
# Accuracy score
from sklearn.metrics import accuracy_score
print('accuracy is', accuracy_score(y_pred, y_test))

precision    recall   f1-score   support
NO          0.84     0.87     0.85      107
YES         0.67     0.62     0.64       47

accuracy          0.79     0.79     0.79      154
macro avg       0.76     0.74     0.75      154
weighted avg    0.79     0.79     0.79      154

[[93 14]
 [18 29]]
accuracy is 0.7922077922077922

```

In []:

Loading [MathJax]/jax/output/CommonHTML/fonts/TeX/fontdata.js