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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")
```

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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	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

```

## diabetes 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	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

```

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In [ ]:
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