

Naïve-Bayes Classifier for prediction on PB1_test.csv:

To train the data using Naïve-Bayes classifier, import **GaussianNB()** from **sklearn.naive_bayes** package.
Let the given test data be $X=(\text{Height}=h1, \text{Age}=a1, \text{Weight}=w1)$.

Using Naïve-Bayes classifier:

Find $P(X | \text{Class}=0) = P(h1 | \text{Class}=0) + P(a1 | \text{Class}=0) + P(w1 | \text{Class}=0)$

$P(X | \text{Class}=1) = P(h1 | \text{Class}=1) + P(a1 | \text{Class}=1) + P(w1 | \text{Class}=1)$

$P(0 | X) = (P(X | 0) * P(0)) / P(X)$

$P(1 | X) = (P(X | 1) * P(1)) / P(X)$

If $P(0 | X) > P(1 | X)$ then, Class = 0, else Class=1 for the given test data.

The prediction values and accuracy are shown below.

Predicted Values on PB1_test.csv: [1. 1. 0. 1. 0. 1. 0. 0. 1. 0. 0. 1. 0. 1. 1. 1. 0. 1. 0. 0.]

Accuracy: 100.0 %

SVM Classifier for prediction on PB1_test.csv:

To train the data using SVM classifier, import **svm** from **sklearn** package.

Support Vector Machines (SVM) uses different kernels to train the data and to classify the test data.

In **linear** SVM, a largest margin hyperplane is searched. This margin is determined by the relation,

$$\mathbf{w} \cdot \mathbf{x} + \mathbf{b} = 0$$

where \mathbf{x} corresponds to attributes of training dataset and \mathbf{w} & \mathbf{b} are parameters of the model.

In **polynomial** kernel, a polynomial relation of certain **degree** is determined for the given training dataset.

In radial basis function (**rbf**) kernel, an exponential relation is determined for the given training dataset.
and accuracies are shown below.

Predicted Values on PB1_test.csv using SVM kernel=linear: [0. 1. 0. 0. 1. 0. 1. 1. 0. 0. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1.]

Accuracy: 40.0 %

Predicted Values on PB1_test.csv using SVM kernel=poly: [0. 0. 0. 0. 0. 0. 1. 1. 0. 0. 0. 0. 0. 0. 1. 0. 1. 1. 1. 1.]

Accuracy: 35.0 %

Predicted Values on PB1_test.csv using SVM kernel=rbf: [0. 0. 0. 0. 1. 0. 1. 1. 0. 0. 1. 0. 1. 0. 1. 1. 1. 1. 1. 1.]

Accuracy: 25.0 %

Naïve-Bayes Classifier for prediction on PB2_test.csv:

To train the data using Naïve-Bayes classifier, import **GaussianNB()** from **sklearn.naive_bayes** package.

Let the given test data be $X=(\text{Height}=h1, \text{Age}=a1, \text{Weight}=w1)$.

Using Naïve-Bayes classifier:

Find $P(X | \text{Class}=0) = P(h_1 | \text{Class}=0) + P(a_1 | \text{Class}=0) + P(w_1 | \text{Class}=0)$

$$P(X | \text{Class}=1) = P(h1 | \text{Class}=1) + P(a1 | \text{Class}=1) + P(w1 | \text{Class}=1)$$

$$P(0|X) = (P(X|0) * P(0))/P(X)$$

$$P(1|X) = (P(X|1) * P(1))/P(X)$$

If $P(0|X) > P(1|X)$ then, Class = 0, else Class=1 for the given test data.

The prediction and accuracy values are shown below:

Predicted Values on PB2_test.csv: [0. 0. 0. 0. 0. 0. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1.]

1. 1. 1. 1. 1. 1.]

Accuracy: 43.333333333333336 %

SVM Classifier for prediction on PB2_test.csv:

To train the data using SVM classifier, import **svm** from **sklearn** package.

Support Vector Machines (SVM) uses different kernels to train the data and to classify the test data.

In **linear** SVM, a largest margin hyperplane is searched. This margin is determined by the relation,

$$\mathbf{w} \cdot \mathbf{x} + \mathbf{b} = 0$$

where x corresponds to attributes of training dataset and w & b are parameters of the model.

In **polynomial** kernel, a polynomial relation of certain **degree** is determined for the given training dataset.

In radial basis function (**rbf**) kernel, an exponential relation is determined for the given training dataset.

The prediction values and accuracies are shown below.

Predicted Values on PB1_test.csv using SVM kernel=linear: [0. 0. 0. 1. 1. 0. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1.]

1. 1. 1. 1. 1. 1. 1.

1. 1. 1. 1. 1. 1.]

Accuracy: 43.333333333333336 %

Predicted Values on PB1_test.csv using SVM kernel=poly: [1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1.

1. 1. 1. 1. 1. 1.

1. 1. 1. 1. 1. 1.]

Accuracy: 36.666666666666664 %

Predicted Values on PB1_test.csv using SVM kernel=rbf: [1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1.]

1. 1. 1. 1. 1.

1. 1. 1. 1. 1. 1.]

Accuracy: 36.666666666666664 %