## Naive Baye's Theorem Assignment

# Difference between Bernoulli Naive Bayes and Multinomial Naive Bayes.

#### Bernoulli Naive Bayes:

- Bernoulli Naive Bayes is typically used when the input features are binary or Boolean variables.
- For example, in text classification, the input features can be binary indicators of whether a certain word appears in a document or not.
- In Bernoulli Naive Bayes, each input feature is treated as a binary variable, and the model estimates the probability of each feature being present in a document given the class label. It assumes that the presence or absence of a feature is independent of the presence or absence of any other feature.

#### **Multinomial Naive Bayes:**

- Multinomial Naive Bayes is typically used when the input features are discrete counts, such as the frequency of occurrence of each word in a document.
- In Multinomial Naive Bayes, the input features are treated as discrete variables, and the model estimates the probability distribution of the count of each feature given the class label.
- It assumes that the counts of each feature are independent of the counts of any other feature, but not necessarily independent of their presence or absence.

#### **Answer 3**

### Bernoulli Naive Bayes handle missing values .

• In Bernoulli Naive Bayes, missing values are typically handled by assuming that the missing values are equivalent to the feature being absent. This means that if a feature value is missing for a particular instance, the classifier will assume that the feature is not present for that instance.

#### For example.

let's say we have a binary feature "has\_experience" with values "yes" and "no", and some instances are missing this feature value. The Bernoulli Naive Bayes classifier will treat the

missing values as equivalent to "no", i.e., it will assume that the instances without a value for "has\_experience" do not have experience.

This approach to handling missing values can lead to some loss of information, as
the classifier assumes that missing values are equivalent to the feature being absent.
However, it is a simple and effective approach that is widely used in practice. If the
missing values are considered to be informative, it may be more appropriate to
impute the missing values based on other features or using more advanced
techniques, such as matrix factorization or k-nearest neighbor imputation

#### **Answer 4**

# Can Gaussian Naive Bayes be used for multi-class classification.

- Yes, Gaussian Naive Bayes can be used for multi-class classification. In fact, Gaussian Naive Bayes is one of the most commonly used algorithms for multi-class classification problems, where the goal is to predict the class label of an instance from among three or more possible classes.
- To use Gaussian Naive Bayes for multi-class classification, the algorithm needs to be extended to handle more than two classes.
- One way to do this is to use the "one-vs-all" approach, where a separate binary classifier is trained for each class, and the class with the highest probability is chosen as the final prediction.
- In this approach, the Gaussian Naive Bayes algorithm is trained to model the distribution of each feature for each class separately, and the class probabilities are estimated using Bayes' theorem.

#### **Another approach**

Another approach for multi-class classification using Gaussian Naive Bayes is to use the "one-vs-one" approach, where a separate binary classifier is trained for each pair of classes, and the class with the most votes is chosen as the final prediction.

• In this approach, the Gaussian Naive Bayes algorithm is trained to model the distribution of each feature for each pair of classes separately, and the class probabilities are estimated using Bayes' theorem.