

# 9th April Assignment

April 29, 2023

## 1 Assignment 64

**Q1. What is Bayes' theorem?**

**Ans.** Bayes' theorem is a method to determine conditional probabilities, that is probabilities of even A occurring given that another event has already occurred.

**Q2. What is the formula for Bayes' theorem?**

**Ans.**  $P(A | B) = P(B | A) * P(A) / P(B)$

where  $P(A | B)$  is the probability of event A occurring given that event B has occurred,  $P(B | A)$  is the probability of event B occurring given that event A has occurred,  $P(A)$  is the prior probability of event A, and  $P(B)$  is the prior probability of event B.

**Q3. How is Bayes' theorem used in practice?**

**Ans.** The bayes theorem used in-

- Medical Diagnosis
- Spam Filtering
- Weather Forecasting
- Stock Market Prediction

**Q4. What is the relationship between Bayes' theorem and conditional probability?**

**Ans.** Bayes' theorem is closely related to conditional probability. Conditional probability is the probability of an event A occurring, given that another event B has occurred. It is denoted as  $P(A | B)$ , which reads as "the probability of A given B." Bayes' theorem provides a way to calculate conditional probabilities by reversing the conditioning. Specifically, it states that:  $P(A | B) = P(B | A) * P(A) / P(B)$

**Q5. How do you choose which type of Naive Bayes classifier to use for any given problem?**

**Ans. The types of naive bayes classifier are:**

1. Gaussian Naive Bayes: This classifier assumes that the features follow a normal (Gaussian) distribution. It is often used when the features are continuous, and their distribution is not too skewed.
2. Multinomial Naive Bayes: This classifier is used when the features are discrete, such as word counts in text classification. It assumes that the features follow a multinomial distribution.
3. Bernoulli Naive Bayes: This classifier is similar to the Multinomial Naive Bayes, but it is used when the features are binary (i.e., present or absent). It assumes that the features follow a Bernoulli distribution.

**Q6. Assignment: You have a dataset with two features, X1 and X2, and two possible classes, A and B. You want to use Naive Bayes to classify a new instance with features X1 = 3 and X2 = 4. The following table shows the frequency of each feature value for each class:**

Class	X1=1	X1=2	X1=3	X2=1	X2=2	X2=3	X2=4
A	3	3	4	4	3	3	3
B	2	2	1	2	2	2	3

Assuming equal prior probabilities for each class, which class would Naive Bayes predict the new instance to belong to?

**Ans.** To predict the class of the new instance using Naive Bayes, we need to calculate the posterior probabilities of each class given the values of X1=3 and X2=4. Since we have equal prior probabilities for each class, the class with the highest posterior probability will be the predicted class.

**We can calculate the posterior probabilities for each class as follows:**  $P(A | X1=3, X2=4) = P(X1=3 | A) * P(X2=4 | A) * P(A) / P(X1=3, X2=4)$

$P(B | X1=3, X2=4) = P(X1=3 | B) * P(X2=4 | B) * P(B) / P(X1=3, X2=4)$

Using the Naive Bayes assumption of feature independence, we can calculate the likelihoods of each feature value given the class as follows:  $P(X1=3 | A) = 4/10 = 0.4$   
 $P(X1=3 | B) = 1/6 = 0.1667$

$P(X2=4 | A) = 3/10 = 0.3$   $P(X2=4 | B) = 1/3 = 0.3333$

To calculate the marginal probability of the feature values, we can sum over the product of the likelihoods and the prior probabilities for each class:  $P(X1=3, X2=4) = P(X1=3 | A) * P(X2=4 | A) * P(A) + P(X1=3 | B) * P(X2=4 | B) * P(B) = 0.4 * 0.3 * 0.5 + 0.1667 * 0.3333 * 0.5 = 0.0633 + 0.0278 = 0.0911$

**Therefore, we can calculate the posterior probabilities as:**  $P(A | X1=3, X2=4) = 0.4 * 0.3 * 0.5 / 0.0911 = 0.6596$   $P(B | X1=3, X2=4) = 0.1667 * 0.3333 * 0.5 / 0.0911 = 0.3404$

The class with the highest posterior probability is A, so the Naive Bayes classifier would predict that the new instance belongs to class A.