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6 Easy Steps to Learn Naive Bayes Algorithm (with codes in Python and R)




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 SUNIL RAY , SEPTEMBER 11, 2017 /  36SHARE    

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Note: This article was originally published on Sep 13th, 2015 and updated on Sept 11th, 2017

Introduction

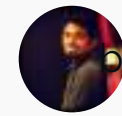
Here's a situation you've got into:

You are working on a classification problem and you have generated your set of hypothesis, created features and discussed the importance of variables. Within an hour, stakeholders want to see the first cut of the model.

What will you do? You have hundreds of thousands of data points and quite a few variables in your training data set. In such situation, if I were at your place, I would have used '**Naive Bayes**', which can be extremely fast relative to other classification algorithms. It works on Bayes theorem of probability to predict the class of unknown data set.

In this article, I'll explain the basics of this algorithm, so that next

4



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time when you come across large data sets, you can bring this algorithm to action. In addition, if you are a [newbie in Python or R](#), you should be overwhelmed by the presence of available codes in this article.

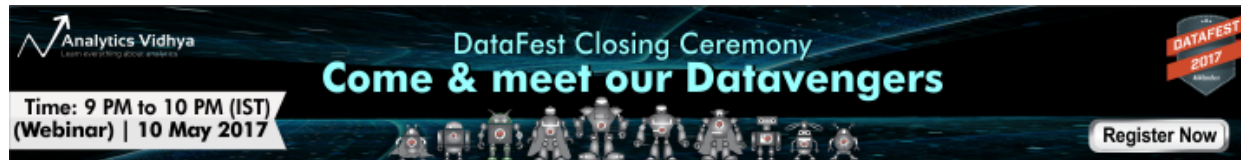


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What is Naive Bayes algorithm?

It is a classification technique based on [Bayes' Theorem](#) with an assumption of independence among predictors. In simple

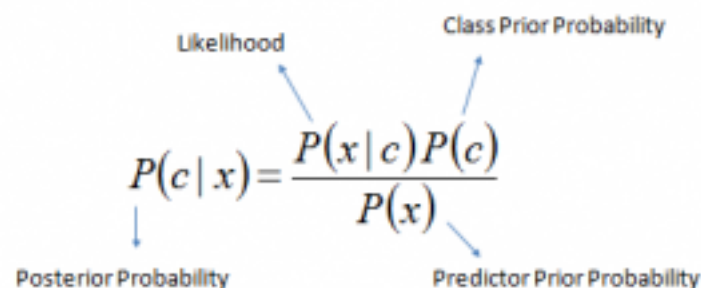
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terms, a Naive Bayes classifier assumes that the presence of a particular feature in a class is unrelated to the presence of any other feature. For example, a fruit may be considered to be an apple if it is red, round, and about 3 inches in diameter. Even if these features depend on each other or upon the existence of the other features, all of these properties independently contribute to the probability that this fruit is an apple and that is why it is known as 'Naive'.

Naive Bayes model is easy to build and particularly useful for very large data sets. Along with simplicity, Naive Bayes is known to outperform even highly sophisticated classification methods.

Bayes theorem provides a way of calculating posterior probability $P(c|x)$ from $P(c)$, $P(x)$ and $P(x|c)$. Look at the equation below:

$$P(c|x) = \frac{P(x|c)P(c)}{P(x)}$$


The diagram shows the equation $P(c|x) = \frac{P(x|c)P(c)}{P(x)}$ with four labels and arrows pointing to the corresponding parts of the equation:

- Likelihood** points to $P(x|c)$
- Class Prior Probability** points to $P(c)$
- Posterior Probability** points to $P(c|x)$
- Predictor Prior Probability** points to $P(x)$

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$$P(c | X) = P(x_1 | c) \times P(x_2 | c) \times \dots \times P(x_n | c) \times P(c)$$

Above,

- $P(c|x)$ is the posterior probability of class (c, target) given predictor (x, attributes).
- $P(c)$ is the prior probability of class.
- $P(x|c)$ is the likelihood which is the probability of predictor given class.
- $P(x)$ is the prior probability of predictor.

How Naive Bayes algorithm works?

Let's understand it using an example. Below I have a training data set of weather and corresponding target variable 'Play' (suggesting possibilities of playing). Now, we need to classify whether players will play or not based on weather condition. Let's follow the below steps to perform it.


Step 1: Convert the data set into a frequency table

Step 2: Create Likelihood table by finding the probabilities like

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Overcast probability = 0.29 and probability of playing is 0.64.

Weather	Play
Sunny	No
Overcast	Yes
Rainy	Yes
Sunny	Yes
Sunny	Yes
Overcast	Yes
Rainy	No
Rainy	No
Sunny	Yes
Rainy	Yes
Sunny	No
Overcast	Yes
Overcast	Yes
Rainy	No

Frequency Table		
Weather	No	Yes
Overcast		4
Rainy	3	2
Sunny	2	3
Grand Total	5	9

Likelihood table		
Weather	No	Yes
Overcast		4
Rainy	3	2
Sunny	2	3
All	5	9
	$\approx 5/14$	$\approx 9/14$
	0.36	0.64



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Step 3: Now, use Naive Bayesian equation to calculate the posterior probability for each class. The class with the highest posterior probability is the outcome of prediction.

Problem: Players will play if weather is sunny. Is this statement is correct?

We can solve it using above discussed method of posterior probability.

$$P(\text{Yes} \mid \text{Sunny}) = P(\text{Sunny} \mid \text{Yes}) * P(\text{Yes}) / P(\text{Sunny})$$

Here we have $P(\text{Sunny} \mid \text{Yes}) = 3/9 = 0.33$, $P(\text{Sunny}) = 5/14 =$



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0.36, $P(\text{Yes}) = 9/14 = 0.64$

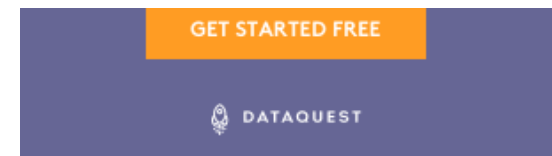
Now, $P(\text{Yes} \mid \text{Sunny}) = 0.33 * 0.64 / 0.36 = 0.60$, which has higher probability.

Naive Bayes uses a similar method to predict the probability of different class based on various attributes. This algorithm is mostly used in text classification and with problems having multiple classes.

What are the Pros and Cons of Naive Bayes?

Pros:

- It is easy and fast to predict class of test data set. It also perform well in multi class prediction
- When assumption of independence holds, a Naive Bayes classifier performs better compare to other models like logistic regression and you need less training data.
- It perform well in case of categorical input variables compared to numerical variable(s). For numerical variable,



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normal distribution is assumed (bell curve, which is a strong assumption).

Cons:

- If categorical variable has a category (in test data set), which was not observed in training data set, then model will assign a 0 (zero) probability and will be unable to make a prediction. This is often known as "Zero Frequency". To solve this, we can use the smoothing technique. One of the simplest smoothing techniques is called Laplace estimation.
- On the other side naive Bayes is also known as a bad estimator, so the probability outputs from `predict_proba` are not to be taken too seriously.
- Another limitation of Naive Bayes is the assumption of independent predictors. In real life, it is almost impossible that we get a set of predictors which are completely independent.

4 Applications of Naive Bayes Algorithms

- **Real time Prediction:** Naive Bayes is an eager learning classifier and it is sure fast. Thus, it could be used for making predictions in real time.
- **Multi class Prediction:** This algorithm is also well known for



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multi class prediction feature. Here we can predict the probability of multiple classes of target variable.

- **Text classification/ Spam Filtering/ Sentiment Analysis:** Naive Bayes classifiers mostly used in text classification (due to better result in multi class problems and independence rule) have higher success rate as compared to other algorithms. As a result, it is widely used in Spam filtering (identify spam e-mail) and Sentiment Analysis (in social media analysis, to identify positive and negative customer sentiments)
- **Recommendation System:** Naive Bayes Classifier and [Collaborative Filtering](#) together builds a Recommendation System that uses machine learning and data mining techniques to filter unseen information and predict whether a user would like a given resource or not

How to build a basic model using Naive Bayes in Python?

Again, scikit learn (python library) will help here to build a Naive Bayes model in Python. There are three types of Naive Bayes model under scikit learn library:

- **Gaussian:** It is used in classification and it assumes that features follow a normal distribution.
- **Multinomial:** It is used for discrete counts. For example, let's say, we have a text classification problem. Here we can consider bernoulli trials which is one step further and instead of "word occurring in the document", we have "count how often word occurs in the document", you can think of it as "number of times outcome number x_i is observed over the n trials".
- **Bernoulli:** The binomial model is useful if your feature vectors are binary (i.e. zeros and ones). One application would be text classification with 'bag of words' model where the 1s & 0s are "word occurs in the document" and "word does not occur in the document" respectively.

Based on your data set, you can choose any of above discussed model. Below is the example of Gaussian model.

Python Code

```
#Import Library of Gaussian Naive Bayes model
```

```
from sklearn.naive_bayes import GaussianNB

import numpy as np

#assigning predictor and target variables
x= np.array([[ -3,7],[1,5], [1,2], [ -2,0], [2,3], [ -4,0],
[ -1,1], [1,1], [ -2,2], [2,7], [ -4,1], [ -2,7]])
Y = np.array([3, 3, 3, 3, 4, 3, 3, 4, 3, 4, 4, 4])
```

```
#Create a Gaussian Classifier
model = GaussianNB()

# Train the model using the training sets
model.fit(x, y)

#Predict Output
predicted= model.predict([[1,2],[3,4]])
print predicted
```

Output: ([3,4])

R Code:

```
require(e1071) #Holds the Naive Bayes Classifier
Train <- read.csv(file.choose())
Test <- read.csv(file.choose())

#Make sure the target variable is of a two-class classification problem only

levels(Train$Item_Fat_Content)

model <- naiveBayes(Item_Fat_Content~., data = Train)
class(model)

pred <- predict(model,Test)
table(pred)
```

Above, we looked at the basic Naive Bayes model, you can improve the power of this basic model by tuning parameters and handle assumption intelligently. Let's look at the methods to improve the performance of Naive Bayes Model. I'd recommend you to go through [this document](#) for more details on Text classification using Naive Bayes.

Tips to improve the power of Naive Bayes Model

Here are some tips for improving power of Naive Bayes Model:

- If continuous features do not have normal distribution, we should use transformation or different methods to convert it in normal distribution.
- If test data set has zero frequency issue, apply smoothing techniques "Laplace Correction" to predict the class of test data set.
- Remove correlated features, as the highly correlated features are voted twice in the model and it can lead to over inflating importance.
- Naive Bayes classifiers has limited options for parameter tuning like `alpha=1` for smoothing, `fit_prior=[True|False]` to learn class prior probabilities or not and some other options (look at detail [here](#)). I would recommend to focus on your pre-processing of data and the feature selection.
- You might think to apply some classifier combination technique like ensembling, bagging and boosting but these methods would not help. Actually, "ensembling, boosting, bagging" won't help since their purpose is to reduce variance.

Naive Bayes has no variance to minimize.

End Notes


In this article, we looked at one of the supervised machine learning algorithm “Naive Bayes” mainly used for classification. Congrats, if you've thoroughly & understood this article, you've already taken your first step to master this algorithm. From here, all you need is practice.

Further, I would suggest you to focus more on data pre-processing and feature selection prior to applying Naive Bayes algorithm. In future post, I will discuss about text and document classification using naive bayes in more detail.

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which tool should I
learn?



Author

Sunil Ray

I am a Business Analytics and Intelligence professional with deep experience in the Indian Insurance industry. I have worked for various multi-national Insurance companies in last 7 years.

36 COMMENTS



Arun CR says:

REPLY



SEPTEMBER 14, 2015 AT 5:36 AM

Hi Sunil ,

From the weather and play table which is table [1] we know that

frequency of sunny is 5

and play when sunny is 3

no play when sunny is 2

so probability(play/sunny) is $3/5 = 0.6$

Why do we need conditional probability to solve this?

Is there problems that can be solved only using conditional probability. can you suggest such examples.

Thanks,

Arun



matt says:

REPLY

JANUARY 24, 2016 AT 4:42 AM

Arun,

An example of a problem that requires the use of conditional probability is the Monty Hall problem.

https://en.wikipedia.org/wiki/Monty_Hall_prob

Conditional probability is used to solve this particular problem because the solution depends on Bayes' Theorem. Which was described earlier in the article.



Erdem Karakoylu says:

REPLY

MARCH 17, 2016 AT 2:26 PM

It's a trivial example for illustration. The "Likelihood table" (a confusing misnomer, I think) is in fact a probability table that has the JOINT weather and play outcome probabilities in the center, and the MARGINAL probabilities of one variable (from integrating out the other variable from the joint) on the side and bottom.

Say, weather type = w and play outcome = p . $P(w,p)$ is the joint probabilities and $P(p)$ and $P(w)$ are the marginals. Bayes rule described above by Sunil stems from:

$$P(w,p) = P(w|p) * P(p) = P(p|w) * P(w).$$

From the center cells we have $P(w,p)$ and from the side/bottom we get $P(p)$ and $P(w)$.

Depending on what you need to calculate, it follows that:

(1): $P(w|p) = P(w,p) / P(p)$ and

(2:) $P(p|w) = P(w,p) / P(w)$, which is what you did with $P(\text{sunny}, \text{yes}) = 3/14$ and $P(w) = 5/14$, yielding $(3/14) (14/5)$, with the 14's cancelling out.

The main Bayes take away is that often, one of the two quantities above, $P(w|p)$ or $P(p|w)$ is much harder to get at than the other. So if you're a practitioner you'll come to see this as one of two mathematical miracles regarding this topic, the other being the applicability of Markov Chain Monte Carlo in circumventing some nasty integrals Bayes might throw at you. But I digress.

Best,
Erdem.



John Siano says:

JULY 21, 2016 AT 3:27 AM

REPLY

I had the same question. Have you found an

answer? Thanks!



Arun CR says:

REPLY

SEPTEMBER 14, 2015 AT 5:42 AM

Great article and provides nice information.



Nishi Singh says:

REPLY

DECEMBER 12, 2015 AT 10:35 AM

Amazing content and useful information



Leena says:

REPLY

FEBRUARY 8, 2016 AT 5:59 AM

I'm new to machine learning and Python. Could you please help to read data from CSV and to separate the same data set to training and test data



Ismael Ezequiel says:

REPLY

OCTOBER 15, 2016 AT 12:12 AM


```
import pandas as pd
```

```
person = pd.read_csv('example.csv')  
mask = np.random.rand(len(sales)) < 0.8  
train = sales[mask]  
test = sales[~mask]
```



Sushma honnidige says:

REPLY

MARCH 7, 2016 AT 2:44 PM

very useful article.



RAJKUMAR says:

REPLY

MARCH 16, 2016 AT 1:30 PM

Very nice....but...if u dont mind...can you please give me
that code in JAVA

...



jitesh says:

REPLY

APRIL 5, 2016 AT 3:20 AM

is it possible to classify new tuple in orange data mining

tool??



SPGupta says:

REPLY

APRIL 11, 2016 AT 4:43 PM

good.



devenir riche au Maroc says:

REPLY

APRIL 14, 2016 AT 5:13 PM

I am really impressed together with your writing skills as well as with the format to your weblog.

Is that this a paid theme or did you customiz it yourself?

Anyway stay up the excellent quality writing, it's uncommon tto see a great weblog like this one these days..



Patek Philippe says:

REPLY

APRIL 22, 2016 AT 3:52 AM

You should be a part of a contest for one of the most useful websites online.

I will highly recommend this blog!



TingTing says:

REPLY

APRIL 22, 2016 AT 3:18 PM

Thanks for tips to improve the performance of models,
that 's really precious experience.



Leanna Partridge says:

REPLY

MAY 11, 2016 AT 8:18 AM

Nice piece – Just to add my thoughts , people require a
CA OCF-1 , We used a sample document here
<http://goo.gl/ibPgs2>



Miguel Batista says:

REPLY

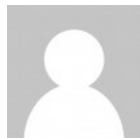
MAY 11, 2016 AT 9:54 AM

Hi,

I have a question regarding this statement:

'If continuous features do not have normal distribution,
we should use transformation or different methods to
convert it in normal distribution.'

Can you provide an example or a link to the techniques?
Thank you,
MB



bd tv says:

REPLY

JUNE 6, 2016 AT 2:45 AM

Can I simply just say what a comfort too find someone who actually knows what they're talking about over the internet. You certainly know how to bring an issue to light and make it important.

A lot more people ought to read this and understand this side of your story.

I can't believe you aren't more popular because you certainly have the gift.



nir says:

REPLY

JULY 13, 2016 AT 6:57 PM

Great article! Thanks. Are there any similar articles for other classification algorithms specially target towards textual features and mix of textual/numeric features?



Nick says:

REPLY

AUGUST 19, 2016 AT 6:29 AM

great article with basic clarity.....nice one



Catherine says:

REPLY

AUGUST 29, 2016 AT 5:18 PM

This article is extremely clear and well laid-out. Thank you!



pangavhane nitin says:

REPLY

AUGUST 31, 2016 AT 8:41 AM

ty



alfiya says:

REPLY

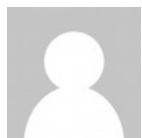
AUGUST 31, 2016 AT 9:26 AM

Explanation given in simple word. Well explained! Loved this article.

**Chris Rucker says:**[REPLY](#)

SEPTEMBER 1, 2016 AT 7:02 PM

The 'y' should be capitalized in your code – great article though.

**Akash Swamy says:**[REPLY](#)

SEPTEMBER 6, 2016 AT 7:24 AM

This is the best explanation of NB so far simple and short

**John says:**[REPLY](#)

SEPTEMBER 9, 2016 AT 5:20 PM

Great article! Really enjoyed it. Just wanted to point out a small error in the Python code.

Should be a capital "Y" in the predict like so : `model.fit(x, Y)`

Thanks!



Adnan says:

REPLY

SEPTEMBER 20, 2016 AT 5:12 AM

Is this dataset related to weather? I am confused as a newbie. Can you please guide?



bh says:

REPLY

OCTOBER 19, 2016 AT 5:40 PM

best artical that help me to understand this concept



Richard says:

REPLY

OCTOBER 25, 2016 AT 7:52 PM

Am new to machine learning and this article was handy to me in understanding naive bayes especially the data on weather and play in the table.

Thanks for sharing keep up



Lisa says:

REPLY



NOVEMBER 10, 2016 AT 8:29 AM

Thanks to you I can totally understand NB classifier.



T B says:

REPLY

JULY 3, 2017 AT 12:17 AM

Really nice article, very use-full for concept building.



AKshay says:

REPLY

JULY 4, 2017 AT 8:40 PM

I didn't understand the 3rd step. Highest probability out of which probability values?

>> Now, $P(\text{Yes} \mid \text{Sunny}) = 0.33 * 0.64 / 0.36 = 0.60$, which has higher probability.

Higher than what?



DN says:

REPLY

JULY 30, 2017 AT 8:29 AM

Concept explained well... nice Article



Rajeshwari says:

AUGUST 31, 2017 AT 2:36 PM

REPLY

thanks nice artical that help me to understand this concept



amit Kumar yadav says:

SEPTEMBER 18, 2017 AT 12:49 AM

REPLY

Good article and I am waiting for text and documents classification using naive base algorithm.



Stella says:

SEPTEMBER 20, 2017 AT 10:50 AM

REPLY

Superb information in just one blog.Enjoyed the simplicity.Thanks for the effort.

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