Practical Mork

Min: To demonstrate confliction rule process on dateset using rains bayer algorithm.



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nim: To domainstak classification state paners on distance using naives
hayes algorithm.
Theory:
maire sayes classifier algorithm:
· Maix Bayes algo in a supervised learning algo, which is based in
eages theorem and used for solving classification problems.
- It is mainly used in text classification that includes a high-
dimensional training datases.
· Maire Bayes classifier in one of the simple and most effecting
charification also.
. It is a probabilistic charifier, which means it predict on the
basis of the contribution of an income it predict on the
banis of the probability of an doject.
· Some popular example of naive payer algo, are spain filheration,
sentimental analysis and classifying axticles.
why is H could Maines Bayes?
The Maire Bayes algo, is comprised of two words you've and
layer, which can be desirabled as:
Maine: It in couled Naive because it assume that the occurrence of a
restruct feature is independent of the occurrence of other
features. Such as if the fault in identified on the bases of
color, shape and take then red, spherical, and suges fruit
is identifier or recognized as an apple, Hence each feature
individually contributes to identify that it is an apple
without depending on each other.
Bayes: It caused scupes because it depends on the principle
· marcoart 28 you to

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## output:

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		1.0
	Time taken to build model: O seconds	
	- Stratified cross-validation - Scrapy -	
	Correctly Classified Instances 8 57.1422 % Incorrectly Classified Instances 8 42.907 %	
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	Foot relative sourced error 50, 6452 % Total Number of Instances 16	
	Detailed Accuracy By Class and	
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	41   6 - 60	
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Date: Barles Theorem in also known as Barles, sine or Barles law in reprein wild to determine the probability of a hypothesis with philips unouledge. It depend on the conditional probability The formula for acuses theorem in given as: (a)a (a(a)a) = (a(a)a) scale) in bortecion bropapility; bropapility of photherin & ou the Observed event B. P(B/A) in historical probability: probability of the evidence given that the parobability of a hypothesis true. P(A) in pains probability: probability of hypothery before observing the evidence. P(10) in marginal probability: probability of evidence working of Maine Rayes' Navillier: wasting of Maine Bayes durafter can be understood with the help of the below example: Suppose we have dataset of weather conditions and corresponding target variable "play". So using this destanct we need to decide that whether we should play or not on a particular day according to the weather condition so to solve this problem we Meed to follow the below steps: convert the given dustaves into frequency tables Chenerals likelihood table by finding the probabilities of given Now, we payed theorem to colourate the posterior probability. tep 1: Initially, use have to pay the required during in the well to ! wing thouse file aptions. Here we selectling the weather-nomineed dutants to execute

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	Now we have to go the classify tech on the left side and with on the charge button and select the naive Bayerian
	Many to change the parameters white on the right side at the chance button and we are accepting the default values in
Step 5:	we chask the percentage Split as our measurement method from the "Test" choices in the main panel. since we don't have a separate test data collection, we'll we the percentage split of the percentage to the gest a good idea of the model's accuracy, our training and 5 being used for testing.  To generate the model, we now was 'hard' when the model is done, the percentage the model we now was 'hard' when the model is done, the production statistic will appear in xight panel.
	Reput:  Thus the classification on data set in performed by maines  eagles classification.  Viva questions:  wheat in a Maine pages charafter?  A naine cages classifier in an along that was pages to mile
<b>Q</b>	abjects. Naive Bayer classifiers assume strong or naive, independence between attributes of data points.  what are the basic assumption?  The basic assumption independent variable features.  Page No.  Page No.



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Ques.3)	what are the advantages of Maire Bayes champler?
-> 0	It is simple and easy to implement.
	TH doesn't require as much training day,
	It handle both discrete and continuous desta.
1	It in fact and can be used to make oreal time predictions.
	It in not sensitive to interevant fractures.
auer 4	Mame the different problem statement you can solve wing
•	Maire Baye's.
-2	A fruit may be considered to be an apple if it in rid, round and
	about sinch in diameter.
	If the weather is sunry, thin the player should play or not?
	set the former than set the
70.1	
*: 1	
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