



Naïve Bayes & K-Means Clustering for Detection of HD

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Abstract

In this paper, by utilizing data mining we can evaluate many patterns which will be used in future to make keenly intellectual systems and decisions. By data mining refers to sundry methods of identifying information or the adoption of solutions predicated on cognizance and data extraction of these data so that they can be utilized in sundry areas such as decision-making, the presage value for the presage and calculation. In our days the health industry has amassed astronomical amounts of patient data, which, infelicitously, is not "engendered" in order to give some obnubilated information, and thus to make efficacious decisions, which are connected with the base of the patient's data and are subject to data mining. This research work has developed a Decision Support in Heart Disease Presage System (HDPS) utilizing data mining modelling technique, namely, Naïve Bayes and Kmeans clustering algorithms that are one of the most popular clustering techniques; however, where the initial cull of the centroid vigorously influences the final result. Utilizing of medical data, such as age, sex, blood pressure and blood sugar levels, chest pain, electrocardiogram, analyzes of different study patient, etc. graphics can presage the likelihood of the patient. This paper shows the efficacy of unsupervised learning techniques, which is a k-betokens clustering to ameliorate edifying methods controlled, which is ingenuous Bayes. It explores the integration of K-designates clustering with verdant Bayes in the diagnosis of disease patients. It withal investigates different methods of initial centroid cull of the K-designates clustering such as range, inlier, outlier, arbitrary attribute values, and desultory row methods in the diagnosis of heart disease patients. The results designate that the integration of the K-betokens clustering with naïve Bayes with different initial centroid culling naïve Bayesian amend precision in diagnosis of the patient.

Keywords: Data Mining, Naïve Bayes, K-Means Clustering.



1. INTRODUCTION

Data mining this is revelation process in the raw data antecedently unknown, non-frivolous, virtually utilizable, the interpretation of the available erudition indispensable for decision-making in the sundry spheres of human activity. This search for relationship with subsisting astronomically immense associated data that are obnubilated among astronomically immense amounts of data and refers to the "mining" cognizance from sizably voluminous amounts of data. Subsisting systems are habituated to avail in decision-making, referred to as data mining. These systems represent an iterative sequence of pre-processing as cleaning, data integration, and data cull is veridical the pattern identification of data mining and erudition representation. Data mining is the search for relationships and ecumenical patterns that subsist in astronomically immense databases, but obnubilated among the plethoras of data. Computer diagnosis of diseases is the medico for the same instrument, the calculations for an engineer: design diagnostics does not supersede the medico, but it avails.[1] The practice of examining immensely colossal preexisting

data bases in order to engender incipient information. It coverts raw data into subsidiary information. It analyzes the data for relationships that have not antecedently been discovered. The steps of data mining are: Data cleaning, data integration, data cull, data transformation, data mining, pattern evaluation and cognizance representation. Medical data mining is a domain of lot of imprecision and skepticismity.[2,3] The clinical decisions are conventionally predicated on the medicos intuition. Consequently this may lead to disastrous consequences. Due to this there are many errors in the clinical decisions and it results in extortionate medical costs. Serialization is withal utilized in this system. It converts the data objects into streams of bytes and stores it into database.

2. RELATED WORK

Many hospital information systems are designed to fortify patient billing, inventory management and generation of simple statistics. Some hospitals use decision support systems, but they are largely constrained. They can answer simple queries like "What is the average age of patients who have heart disease?", "How many surgeries had resulted in hospital stays

longer than 10 days?”, “Identify the female patients who are single, above 30 years old, and who have been treated for cancer.” However, they cannot answer intricate queries like “Identify the paramount Preoperative prognosticators that increase the length of hospital stay”, “Given patient records on cancer, should treatment include chemotherapy alone, radiation alone, or both chemotherapy and radiation?”, and “Given patient records, soothsaying the probability of patients getting a heart disease.” Clinical decisions are often made predicated on doctors’ intuition and experience rather than on the erudition- opulent data obnubilated in the database.[5,8] This practice leads to unwanted biases, errors and extortionate medical costs which affects the quality of accommodation provided to patients. Wu, et al proposed that integration of clinical decision support with computerbased patient records could reduce medical errors, enhance patient safety, decrement unwanted practice variation, and ameliorate patient outcome. This suggestion is promising as data modeling and analysis implements, e.g., data mining, have the potential to engender a cognizance-opulent environment which can avail to significantly amend the quality of clinical decisions

3. IMPLEMENTATION

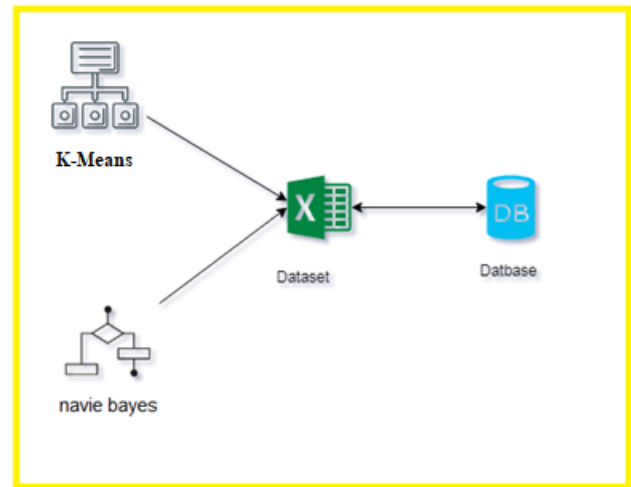


Fig: -1 System Architecture

Naïve Bayes Algorithm

Ingenuous Bayes classifier can be trained in supervised learning setting. It utilizes the method of maximum kindred attribute. It has been worked in involute authentic world situation. It requires iota of training data. It estimates parameters for relegation. Only the variance of variable need to be tenacious for each class not the entire matrix.[6] Naïve bayes is mainly used when the inputs are high. It gives output in more sophisticated form. The probability of each input attribute is shown from the prognosticable state. Machine learning and data mining methods are predicated on naïve bayes relegation. Naïve bayes will rudimentally soothsaying the output whether the patient will have chances of getting the heart disease or not. The

model dataset which we get after applying K-Betokens algorithm will compared the values of dataset with a trained dataset. It will apply the bayes theorem and the probability will be obtained whether the patient will have heart disease or not

Algorithm Steps

Which one is the best predictor ?				
Outlook	Temp	Humidity	Windy	Play Golf
Rainy	Hot	High	False	No
Rainy	Hot	High	True	No
Overcast	Hot	High	False	Yes
Sunny	Mild	High	False	Yes
Sunny	Cool	Normal	False	Yes
Sunny	Cool	Normal	True	No
Overcast	Cool	Normal	True	Yes
Rainy	Mild	High	False	No
Rainy	Cool	Normal	False	Yes
Sunny	Mild	Normal	False	Yes
Rainy	Mild	Normal	True	Yes
Overcast	Mild	High	True	Yes
Overcast	Hot	Normal	False	Yes
Sunny	Mild	High	True	No

Fig 2 Sample Medical Data Set values

Frequency Tables				
★ Outlook	Sunny	3 3/9	2 2/5	
	Overcast	4 4/9	0 0/5	
	Rainy	2 2/9	3 3/5	
Temp.	Hot	2 2/9	2 2/5	
	Mild	4 4/9	2 2/5	
	Cool	3 3/9	1 1/5	
Humidity	High	3 3/9	4 4/9	
	Normal	6 6/9	1 1/5	
Windy	False	6 6/9	2 2/5	
	True	3 3/9	3 3/5	

Fig 3 Frequency Tables from Data Set values

- The posterior probability can be calculated by first, constructing a frequency table for each attribute against the target
- Then, transforming the freq. tables to likelihood tables and finally using the Naive Bayesian equation to calculate the posterior probability for each class
- The class with the highest posterior probability is the outcome of prediction

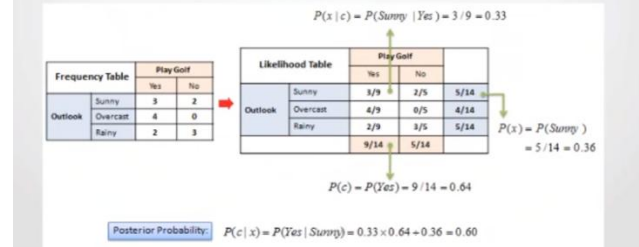


Fig 4 Posterior Probability Calculation

Let's assume we have a day with:

Outlook = Rainy
Temp = Mild
Humidity = Normal
Windy = True

Likelihood of Yes = $P(\text{Outlook}=\text{Rainy}|\text{Yes}) \cdot P(\text{Temp}=\text{Mild}|\text{Yes}) \cdot P(\text{Humidity}=\text{Normal}|\text{Yes}) \cdot P(\text{Windy}=\text{True}|\text{Yes}) \cdot P(\text{Yes}) =$
 $2/9 \cdot 4/9 \cdot 6/9 \cdot 3/9 \cdot 9/14 = 0.014109347$

Likelihood of No = $P(\text{Outlook}=\text{Rainy}|\text{No}) \cdot P(\text{Temp}=\text{Mild}|\text{No}) \cdot P(\text{Humidity}=\text{Normal}|\text{No}) \cdot P(\text{Windy}=\text{True}|\text{No}) \cdot P(\text{No}) =$
 $3/5 \cdot 2/5 \cdot 1/5 \cdot 3/5 \cdot 5/14 = 0.010285714$

Now we normalize:

$P(\text{Yes}) = 0.014109347 / (0.014109347 + 0.010285714) = 0.578368999$
 $P(\text{No}) = 0.010285714 / (0.014109347 + 0.010285714) = 0.421631001$

Fig 5 Result

4. K-MEANS CLUSTERING

Algorithm Steps

KNN is slow supervised learning algorithm, it take more time to get trained classification like other algorithm is divided into two step training from data and testing it on new instance . The K Nearest Neighbour working principle is based on assignment of weight to the each data point which is called as neighbour.[10] In K Nearest Neighbour distance is calculate for training dataset for each of the K Nearest data points now classification is done on basis of majority of

votes there are three types of distances need to be measured in KNN Euclidian, Manhattan, Minkowski distance in which Euclidian will be consider most one the following formula is used to calculate their distance.

$$\begin{aligned} \text{Euclidian Distance} = D(x, y) & \quad (1) \\ & = (x_i - y_i)_{2k_i} = 1 \end{aligned}$$

K=number of cluster

x , y=co-ordinate sample spaces

The algorithm for KNN is defined in the steps given below:

1. D represents the samples used in the training and k denotes the number of nearest neighbour.
2. Create super class for each sample class.
3. Compute Euclidian distance for every training sample
4. Based on majority of class in neighbour, classify the sample

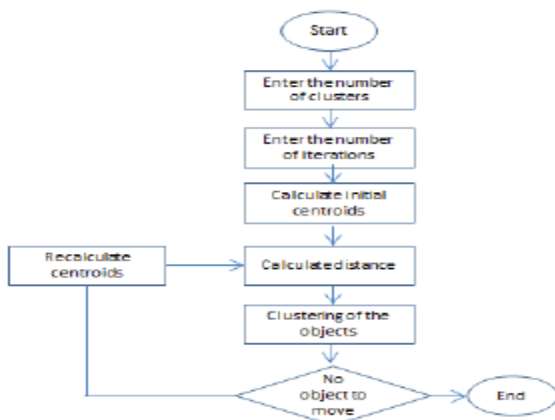
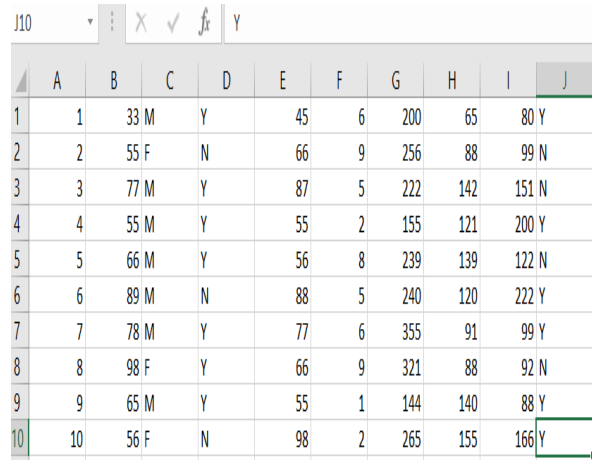


Fig 6K-means clustering algorithm

5. EXPERIMENTAL RESULTS



	A	B	C	D	E	F	G	H	I	J
1	1	33 M	Y		45	6	200	65	80	Y
2	2	55 F	N		66	9	256	88	99	N
3	3	77 M	Y		87	5	222	142	151	N
4	4	55 M	Y		55	2	155	121	200	Y
5	5	66 M	Y		56	8	239	139	122	N
6	6	89 M	N		88	5	240	120	222	Y
7	7	78 M	Y		77	6	355	91	99	Y
8	8	98 F	Y		66	9	321	88	92	N
9	9	65 M	Y		55	1	144	140	88	Y
10	10	56 F	N		98	2	265	155	166	Y

Fig 7 Medical Data Set

Frequency Table for Total data..

Yes (Total)	No (Total)	Total	Yes/Tot	No/Tot
359	240	599	0.5993322203672788	0.4006677796327212

Frequency Tables are Completed...

Fig 8Frequency Table for Age Data

Enter Patient Details

Age:

Gender:

Smoker:

Heart Rate:

Chest Pain:

Cholesterol:

Blood Pressure:

Blood Sugar:

Fig 9 Input values

Cholesterol	100-200	166.0	1.0
Bloodpressure	91-120	121.0	1.0
BloodSugar	80-150	181.0	181.0

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p(Yes)=(Yes_Score_of_age/Tot of Yes Score)*(Yes_Score_of_gen/Tot of Yes Score)*(Yes_Score_of_smoker/Tot of Yes Score)
*(Yes_Score_of_hr/Tot of Yes Score)*(Yes_Score_of_cp/Tot of Yes Score)*(Yes_Score_of_ch/Tot of Yes Score)*(Yes_Score_of_bp/Tot of Yes
Score)*(Yes_Score_of_bs/Tot of Yes Score)*(Yes Tot/Total)
p(Yes)=0.003082236996967234

p(No)=(No_Score_of_age/Tot of No Score)*(No_Score_of_gen/Tot of No Score)*(No_Score_of_smoker/Tot of No Score)
*(No_Score_of_hr/Tot of No Score)*(No_Score_of_cp/Tot of No Score)*(No_Score_of_ch/Tot of No Score)*(No_Score_of_bp/Tot of No
Score)*(No_Score_of_bs/Tot of No Score)*(Tot No Score/Total) p(No)=2.160313721510784E-9

p(Yes)+p(No)=0.0030822391572809553

p(Yes)/(p(Yes)+p(No))=0.9999992991089883

p(No)/(p(Yes)+p(No))=7.008910117852562E-7

RESULT: POSITIVE

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Fig 10 Result after Mining

6. CONCLUSION

In this paper we are proposing heart disease prognostication system utilizing naïve bayes and k-designates clustering. We are utilizing k-betokens clustering for incrementing the

efficiency of the output. This is the most efficacious model to prognosticate patients with heart disease. This model could answer intricate queries, each with its own vigor with deference to facilitate of model interpretation, access to detailed information and precision

7. FUTURE SCOPE

In future work we will improve this intelligent decision-making system by using other new models and apply them to other environments.

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