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Application of the *K-Nearest Neighbor* Method for Classification of Leiomyoma (Myoma)

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Abstract

Information technology is very important in the process of human life. Along with the growth opens up opportunities for relatively large data growth, one of which is hospitals. Mioma is a disease that continues to increase and has a major impact on health female reproduction. Myoma is a benign tumor that grows in or around the uterus. Mioma is a medical condition experienced by women of all ages, but is often experienced by women who have entered premenopause, myoma is also the second benign tumor in Indonesia by age range sufferers 20-50 years old. Sufferers rarely cause specific symptoms so women are rarely aware of them the presence of myoma growth in their uterus. This research classifies patient data with a purpose to classify types of myoma disease using the K-Nearest Neighbor method. There are several The attributes used in this research are diastolic blood pressure, systolic blood pressure, hemoglobin, ever been pregnant, symptoms 1 and symptom 2. The data used for this research amounted to 288 myoma patient data which will be divided into 2, namely 70% training data and 30% testing data. Then it is divided into 3 classes, namely intramural myoma, submucosal myoma, and subserosal myoma. Results of myoma classification using the K-Nearest algorithm Neighbor at Aceh Tamiang Regional Hospital used 87 test data or patient data, indicating people with the disease Intramural myoma are more numerous with 48 data, subserosal myomas 15 data and for subserosal myomas there are 25 data with a high accuracy rate of 93%.

Keywords: Classification, K-Nearest Neighbor, Myoma.

Introduction

Data mining is a data processing process using statistical techniques, mathematics, artificial intelligence and machine learning, to extract and identify useful information and relevant knowledge from large database[1]. There are several techniques that can be used in data mining to find information new information, one of the data mining techniques is classification[2].

Classification is a method of systematically grouping data according to rules and regulations that has been determined, Classification is also defined as grouping new data or objects that have not yet been determined class or category is known[3]. Classification is a statistical method for grouping or classify data that is arranged systematically. Grouping rules are used for classify new data into existing groups[4]. In this research the classification system is used is the K-Nearest Neighbor algorithm.

Myoma, leiomyoma, or uterine myoma is a medical condition in the form of abnormal tissue growth or tumors inside or around the uterus. Uterine myoma consists of muscle and fibrous tissue, Unlike malignant cancer, uterine myoma is a benign tumor. The size of a benign tumor that grows on an organ These uteruses generally vary, ranging from the size of a marble to a tennis ball[5]. Women's reproductive health has a huge impact and plays an important role in the continuity of the next generation, uterine myoma/mioma is one of the reproductive health problems that occurs in women continues to increase and is the second most common benign tumor in Indonesia, wrong the only one is in the Aceh Tamiang area. Myoma patients at the Aceh Tamiang Regional Hospital continue increases every year with an age range of around 20-50 years, the sufferers Myoma disease rarely causes specific symptoms so it is rare in women aware of the presence of myoma growth in their uterus.

For this reason, it is necessary to classify uterine myoma using the KNN method and can be done applied at the Aceh Tamiang Regional Hospital so that it can facilitate the resolution of existing problems and provide more accurate and effective information. Based on the background that has been described, then The proposed title is "Application of the K-Nearest Neighbor Method for Classification of Leiomyoma Disease(Myoma)".

Literature Review K-Nearest Neighbor Algorithm



The K-Nearest Neighbor algorithm is one of the methods used for classification within an object. K-Nearest Neighbor works with looking for the shortest distance between data and other data, K-Nearest Neighbor too including a supervised learning algorithm, where the results of a new query instance are classified based on the majority of categories in K-Nearest Neighbor[6].

Which performs classification based on the proximity of the location of one data to other data is the K-Nearest Neighbor method. Purpose of the K-Nearest algorithm Neighbor is to classify new objects based on features and training samples[7]. There are several steps to calculate the K Nearest Neighbor algorithm as follows:

- 1. Determine the K parameter (number of closest neighbors)
- 2. Calculate the Eulidean Distance of each object to the data training provided.
- 3. Then sort the objects into groups has the smallest distance.
- 4. Collecting category Y (K-Nearest Neighbor classification based on value k).

The advantages of the K-Nearest Neighbor algorithm for classification are:

1. Have strong consistency. When the amount of data approaches infinity The K-Nearest Neighbor algorithm guarantees a better error rate than Bayes error rate.

2. Resilient to noisy test data.

3. Highly nonlinear

The disadvantages of the K-Nearest Neighbor algorithm are:

- 1. Need to determine the parameter K (number of nearest neighbors).
- 2. Quite high computational load due to the distance between test data and all data training must be counted

Euclidean Distance

Euclidean distance has a function to test possible measurements used as an interpretation of the closeness of the distance between two objects. Euclidean distance is also often used to calculate distance. Method used in this research is the Euclidean Distance approach method. Following

the formula:

$$d(x, y) = \sqrt{\sum_{i=1}^{n} (x_i - y_i)^2}$$
(1)

Information:

d(x,y) = Distance

xi = Test Data

- yi = Training Data
- N = Data Dimension
- I = Data Variable

Confusion Matrix

Confusion matrix is a testing stage to estimate objects true or false. The confusion matrix testing sequence is the predicted class will be displayed at the top of the matrix and the observed class on the left. Each cell has a number indicating how many cases there are the actual value of the observed class to predict.

	Table 1. Confusion	1 Matrix
Mark	Actu	al Value
Prediksi	TP	FP
Frediksi	FN	TN

Materials & Methods

The research steps carried out by the author were to develop This final project was made using the waterfall method. First step starting with a literature study to system evaluation. The waterfall method consists of from several stages including: Literature study, needs analysis, design system, system implementation, system testing, and system evaluation. can be seen as follows:

1. Literacy Studies

Before conducting any research, it is necessary to conduct a literature study means to find the underlying theory. The theoretical basis can be found Obtain from various sources such as journals, articles, the internet, and various books.

2. Needs Analysis

Requirements analysis functions to obtain requirements specifications generality of the system being built. In this research, a needs analysis was carried out through data collection. After getting general needs data, the data will be used at the next stage of system design.

3. System Design

System design is carried out in several stages, namely design architecture, database design, and interface design.

Planning architecture involves creating context diagrams for database design, and interface design creating wireframes.

4. System Implementation

At the implementation stage of this system, a design is made at stage previously implemented in the programming language that will be used.

5. System Testing

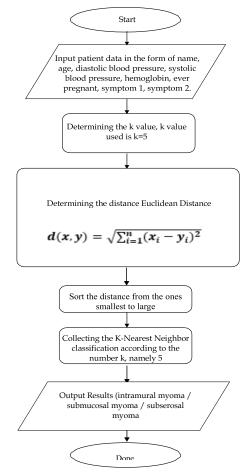
This stage carries out system testing that has been carried out at stage system implementation. This stage is useful to help determine whether system implementation that meets the requirements.

6. System Evaluation

After the system software is complete, system testing will be carried out that has been made. The purpose of testing is to find errors in system and then improve it in the hope that it is in accordance with the analysis done..

System Schema

The classification system scheme in the research entitled "Application of the K-Nearest Neighbor Method for Classification of Leiomyoma Disease (Myoma)" as follows:





Information:

- 1. The system process starts.
- 2. User inputs attributes, in the form of name, age, diastolic blood pressure, systolic blood pressure, hemoglobin, ever pregnant, symptom 1, symptom 2.
- 3. Then the user determines the k value, the k value used is k=5.
- 4. Then the user calculates the Euclidean distance.
- 5. Then the user sorts the Euclidean distance from the smallest to the smallest biggest
- 6. Next, the user collects the K-Nearest Neighbor classification according to the number k is 5.
- 7. After that, the results appear in the form of a diagnosis in the form of Intramural Mioma, Mioma Submucosal,
- Subserosal Mioma according to the symptoms suffered.
- 8. The system work process is complete.

Result and Discussion

The data used in this study is data on patients with myoma disease obtained from the Aceh Tamiang District Hospital for 5 years, from 2019-2023 with the classification of intramural myoma, submucosal myoma, and subserosal myoma. The following is the normalization table:



Table 2. Normalization Table				
	Normalization Table			
Classification	Weight			
Intramural Myoma	1			
Submucosal Myoma	2			
Subserosal Myoma	3			

The following is a table of criteria codes:

Table 3. Criteria Table			
Criteria Code Criteria			
X1	Diastolic Pressure		
X2	Systolic Pressure		
X3	Hemoglobin		
X4	Ever Pregnant		
X5	Symptom 1		
X6	Symptom 2		

The following is a table of diastolic blood pressure criteria:

Ta	ble 4. Diastolic Blood Pressure Crit	eria
	Diastolic Blood Pressure	
Attribute	Value	Weight
	100	1
	110	2
Diastolic Blood Pressure	120	3
Diastolic blood Pressure	130	4
	140	5
	150	6

The following is a table of systolic blood pressure criteria:

Table 5. Systolic Blood Pressure Criteria				
Systolic Blood Pressure				
Attribute Value Weight				
	70	1		
Systolic Blood Pressure	80	2		
-	90	3		

The following is a table of hemoglobin criteria:

Table 6. Hemoglobin Criteria			
	Hemoglobin		
Attribute	Value	Weight	
	6	1	
	7	2	
	8	3	
Hemoglobin	9	4	
-	10	5	
	11	6	
	12	7	

The following is a table of ever pregnant criteria:

	Table 7. Ever Pregnant Criteria				
	Ever Pregnant				
Attribute	Attribute Value Weight				
Ever Dreenent	Yes	1			
Ever Pregnant	No	2			

The following is a table of symptom 1 criteria:

Table 8. Symptom 1 Criteria

Symptom 1					
Attribute Value Weight					
Comparente and 1	Bleeding	1			
Symptom 1	Stomach Enlargement	2			

The following is a table of symptom 2 criteria:

Table 9. Symptom 2 Criteria					
	Symptom 2				
Attribute	Attribute Value Weight				
	Pelvic Pain	1			
Symptom 2	Stomach Pain	2			
	Changes in menstrual patterns	3			

From all the tables above, training data and testing data are obtained which have been weighted as follows:

			Т	able 10). Train	ing Da	ta			
No	Patient Name	X1	X2	Х3	X4	X5	X6		Classification	on
1	Nurjanah	3	3	6	1	1	1	Iı	ntramural My	yoma
2	Fatimah Hasan	4	2	7	1	1	1	I	ntramural My	yoma
3	Rasiyah	1	2	3	1	1	1	I	ntramural My	yoma
								•		
200	Jubaedah	4	3	2	1	1	1	I	ntramural My	yoma
			ſ	Table 1	1. Data	Testin	g			
No	Patient Name	Х	1	X2	2	X3		X4	X5	X6
1	Imrana	1	L	2		7		1	2	2
2	Sari Ayu	3	3	3		6		1	1	3
3	Lili Yuliani	5	5	3		3		1	1	1
88	Iin Supreh	4	1	3		7		1	2	2

From the training data and testing data above, distance testing was carried out from the 1st testing data to the 88th testing data with all training data using the Euclidean distance formula. After getting the distance value, the data will be grouped based on the predetermined k, namely k=5 and the data will be sorted based on the smallest distance value for each distance according to the number of k. The following is a table of test results for the value k=5:

Table 12. Table of Test Results K=5

No	Name	Actual	Predictions
1	Imrana	Subserosal Myoma	Subserosal Myoma
2	Sari Ayu	Submucosal Myoma	Submucosal Myoma
3	Lili Yuliani	Intramural Myoma	Intramural Myoma
4	Siti Ruqoyah	Intramural Myoma	Intramural Myoma
5	Nur Dewi	Intramural Myoma	Intramural Myoma
88	Iin Supreh	Subserosal Myoma	Subserosal Myoma

From the distance calculation results above, 5 data were obtained with different results. This indicates that calculating classification using the k-nearest neighbor method and using the Euclidean distance formula has a very low error value. The following confusion matrix performs tests to photograph the correct and incorrect objects:

Table 13. Confusion Matrix Testing						
Actual	Prediction					
Actual	Intramural Myoma	Submucosal Myoma	Subserosal Myoma			
Intramural Myoma	45	0	0			
Submucosal Myoma	0	15	3			
Subserosal Myoma	3	0	22			

Based on the confusion matrix above, the levels of precision, recall and accuracy are as follows:

	Table 14. Table of Recall,	Precision and Accuracy Values	
Class	Recall	Precision	Accuracy



1	100%	94%	
2	83%	100%	93%
3	88%	88%	

The results are visualized with a bar graph in the following image:

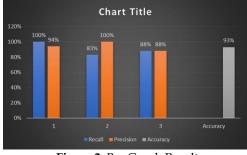


Figure 2. Bar Graph Result

System Implementasion

The following are the results of system implementation using the PHP programming language according to the system scheme that was previously designed:

admin 📃	Data	Training	3									
Dashboard Data Training	Tam	ibah Data						Pilih	File Tidak	ada file yanç	g dipilih	Unggah
Hasil Klasifikasi	No	Nama Pasien	Usia	Tensi Diastolik	Tensi Sistolik	Hemoglobin	Pernah Hamil	Gejala 1	Gejala 2	Jarak	Hasil Klasifikasi	Aksi
😐 Keluar	1	Nurjannah	39	120	90	11	Ya	Pendarahan	Nyeri Panggul	2.828427	Mioma Intramural	Edit Dele
	2	Fatimah Hasan	45	130	80	12	Ya	Pendarahan	Nyeri Panggul	3.316625	Mioma Intramural	Edit Dele
	3	Rasiyah	42	100	80	8	Ya	Pendarahan	Nyeri Panggul	4.242641	Mioma Intramural	Edit Date
	4	Yusniar	53	110	80	10	Ya	Pendarahan	Nyeri Panggul	2.645751	Mioma Intramural	Edit Dele
	5	Musdalifah Srg	41	120	90	11	Ya	Pembesaran Perut	Nyeri Perut	2.449490	Mioma Subserosal	Edit Dele

Klasifikasi Miom	=											admin
🧕 admin	No	Nama Pasien	Usia	Tensi Diastolik	Tensi Sistolik	Hemoglobin	Pernah Hamil	Gejala 1	Gejala 2	Jarak	Klasifikasi	Hapus
Dashboard	1	Imrana	47	100	80	12	Ya	Pembesaran Perut	Nyeri Perut	0.965685	Mioma Subserosal	Colete
III Data Training	2	Sari Ayu	46	120	90	11	Ya	Pendarahan	Perubahan Pola Menstruasi	1.331371	Mioma Submukosal	Delete
Hasil Klasifikasi	3	Lili Yuliani	41	140	90	8	Ya	Pendarahan	Nyeri Panggul	3.183633	Mioma Subserosal	Dointe
ビ Keluar	4	Siti Ruqoyah	41	130	90	8	Ya	Pendarahan	Nyeri Panggul	0.4	Mioma Intramural	Delete
	5	Nur Dewi	32	100	80	10	Ya	Pendarahan	Nyeri Panggul	0.482843	Mioma Intramural	Delete
	6	Ika Putri	31	110	80	9	Ya	Pendarahan	Nyeri Panggul	1	Mioma Intramural	Delete
	7	Nining	33	130	90	12	Ya	Pembesaran Perut	Nyeri Perut	0.2	Mioma Subserosal	Delete
	8	Farida Eka	51	100	80	6	Ya	Pendarahan	Nyeri Panggul	2.212899	Mioma	Celete

Figure 4. Classification Results Display

Figure 3 shows the training data menu which contains patient data that can be added, edited and deleted. Then in Figure 4 shows the classification results menu which can be deleted and managed by the admin.

Conclusions

From the results of the tests that have been carried out, it can be concluded that the data used was 288 patient data from the Aceh Tamiang District Hospital. There are 200 data used as training data and 88 data used for testing data and using 6 (six) criteria, namely diastolic blood pressure, systolic blood pressure, hemoglobin, ever pregnant, symptom 1, and symptom 2. Using the K-Nearest Neighbor algorithm, using the Euclidean Distance formula and Confusion Matrix calculations with a value of K=5, 88 data were tested to obtain classification output results in the form of 48 intramural myoma data, 15 submucosal myoma data and 25 subserosal myoma data in patient data with myoma at the Aceh Tamiang District Hospital. In its application, the myoma disease classification system which uses the K-Nearest

Neighbor algorithm and uses the Euclidean Distance formula produces a high accuracy value of 93%.

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