ARPN Journal of Engineering and Applied Sciences

©2006-2017 Asian Research Publishing Network (ARPN). All rights reserved.



www.arpnjournals.com

DEVELOPMENT OF AN EXPERT SYSTEM ALGORITHM FOR DIAGNOSING CARDIOVASCULAR DISEASE USING ROUGH SET THEORY IMPLEMENTED IN MATLAB

Aaron Don M. Africa

Department of Electronics and Communications Engineering, Gokongwei College of Engineering, De La Salle University, Manila, Philippines

E-Mail: aaron.africa@dlsu.edu.ph

ABSTRACT

Cardiovascular disease refers to conditions that involve narrow or blocked blood vessels. This disease when remained untreated may lead to a heart attack. When a person has a cardiovascular disease, the heart may not be able to pump enough blood to the body. When there is insufficient blood the brain or other organs may become damaged. Cardiovascular disease is challenging to diagnose because its symptoms may be mistaken for other diseases. Early detection, if a person has cardiovascular disease is a big advantage in combating the ailment. This is because diagnosing the disease early may reduce the complications it may bring. This research will develop an Expert System Algorithm for the diagnosis of cardiovascular disease. This research will guide the person diagnosing the disease to provide the appropriate recommendation. The Rough Set Theory will be used to reduce the rules so it can be easily diagnosed. This research will utilize the Statlog Heart Data Set of the UCI machine learning repository. Matrix Laboratory or MATLAB will be used to implement the system.

Keywords: cardiovascular disease, rough set theory, expert system, matrix laboratory, public health engineering.

1. INTRODUCTION

1.1 Background of the Study

Cardiovascular disease is a dangerous disease. This disease is one of the leading causes of deaths in the world. Cardiovascular disease is a general term that refers to conditions affecting the heart and blood vessels. This disease is associated with the buildup of atherosclerosis or fatty deposits inside the arteries. This disease can cause damage to other organs like heart, brain and kidneys especially if a heart attack occurs. A heart attack happens when blood flow to a part of a heart is blocked by a blood clot. If this clot clots completely the blood flow that part of the heart muscle supplied by the artery begins to die [1]. Many people survive their first heart attack but the problem is once you already had a heart attack you are not the same as before. Some of your organs may already have received damage and you are prone to another heart attack. The next heart attack a person may have can be fatal and may lead to further complications. The best way is to prevent a heart attack before it occurs in the first place [2]. One way to do it is to detect if a person has a cardiovascular disease. This disease is difficult to diagnose because it may be mistaken for other diseases [3]. This research will develop an Expert System algorithm that can be used to diagnose cardiovascular disease using Rough Set Theory. This algorithm will be implemented in MATLAB.

1.2 Statement of the Problem

Diagnosing if a person has cardiovascular disease is a difficult task. One way to make diagnosis is to use an Expert System. This system simulates the judgment of a human expert to give the correct diagnosis. The main problem in using an Expert System is the information that you should input to it should be complete. In order to make a diagnosis all of the condition attribute information should be known. The problem is there are instances when the value of the condition attribute is unknown. The data cannot be obtained therefore a diagnosis cannot be provided.

1.3 Significance of the Study

Developing an Expert System algorithm for the diagnosis of cardiovascular disease will help cardiologist in making diagnosis. Knowing if a person has this disease will greatly help because early treatment can be performed before it is too late. This research will also help in solving the problem with Expert Systems about dealing with incomplete information. Integrating the Expert System algorithm with Rough Set Theory will enable the system to handle incomplete information.

1.4 Objectives

1.4.1 Main objective

To develop an Expert System algorithm for the diagnosis cardiovascular disease.

1.4.2 Specific objective

- To implement the Expert System algorithm in **MATLAB**
- To use Empirical testing to validate the results
- To utilize the Statlog Heart Data Set of the UCI machine learning repository as the training data

©2006-2017 Asian Research Publishing Network (ARPN). All rights reserved.



www.arpnjournals.com

1.5 Assumptions, scope and limitations

- This study is about the diagnosis of cardiovascular disease only using the Statlog Heart UCI dataset
- The scope of this study is limited only to the attributes of the Statlog Heart UCI dataset

2. REVIEW OF THE RELATED LITERATURE

2.1 Expert system

An Expert System is defined as a computer program that represents knowledge and can reason with it. This program also simulates a human expert in giving a recommendation on the possible cause of a problem [4]. In order to solve expert level problems, an Expert System should have efficient access to a knowledge base. This system should also have a reasoning mechanism in order to apply the knowledge to the problems inputted into it. Generally, these systems are built on the ideas of knowledge representation. These systems also have production rules in order to give the correct possible cause. An Expert System usually has a shell which is an existing knowledge independent framework on which additional information can be inputted. [5]. Expert Systems have no limits on the problems it can solve. Expert systems are distinguished from typical computer systems because they simulate human reasoning, perform reasoning over the database oh human knowledge and solve problems by using approximate methods. The knowledge acquisition component allows the Expert System to be inputted with data from human experts. These data can be refined if required. Expert Systems currently are used in fields like artificial intelligence, digital processing and biomedical engineering [6].

2.2 Cardiovascular disease

Cardiovascular disease is generally known as damage to the heart and blood vessels [7]. The term cardiovascular disease includes stroke, heart disease and other diseases of the heart. Major blood vessels consist of arteries. These arteries carry blood away from the heart and veins which returns it. Capillaries are tiny vessels where the exchange of carbon dioxide and oxygen occurs. There are instances when fatty materials like cholesterol deposit on the walls of vessels. These deposits are known as plaque. These plaques clogs the artery, thus reduce the space for blood flow. This phenomenon is called arteriosclerosis [8]. If for example the plaque ruptures in the artery walls, platelets try to repair the damage. The problem here is more clots might form and overtime the walls of the blood vessels may lose their elasticity. The loss of elasticity will contribute to hypertension or high blood pressure. The blockage of an artery leads to a part of the body being starved by oxygen. This in turn will result to a heart attack or stroke [9]. It is better to detect if a cardiovascular disease is present early on so heart attack can be prevented.

2.3 Statlog heart data set of the UCI machine learning repository

The UCI machine learning repository are datasets compiled by University of California Irvine. They have different fields like annealing datasets and automobile datasets. Their datasets have various forms like categorical, numerical and mixed. Research subjects can be in the field of statistics, mathematics and artificial intelligence. Their repository is open source and free to download for scientific uses. The Statlog Heart dataset is a multivariate dataset which is real and categorical in form. This dataset slightly varies from the Hungarian UCI database which is also included in the UCI machine learning repository. It consist 13 attributes where the 13th attribute which is the thal, is the decision attribute [10].

3. THEORETICAL FRAMEWORK

In the early 1980's Rough Set Theory Zdzislaw I. Pawlak developed the Rough Set Theory [11]. This theory deals with the classification analysis of data tables. The primary goal of the Rough Set analysis is to synthesize concepts of approximation from the acquired data [12]. The general modelling process typically consist of sequence of various steps that all requires several degrees of tuning and fine adjustments. In order to make this function, an interactive management process is required. Rough Sets consist of information systems. An information system is where the data is represented in a table. Every column represents an attribute and the rows represent the number of cases. The intersection of the rows and columns are called objects. In an information system IS = (U,A) where U is a non-empty finite set of objects called the universe and A is a non-empty finite set of attributes. In an information system $a: U \to V_a$ for every $a \in A$. The set V_a is referred to as the value set of a[13].

Upper Approximation Set X Lower Approximation Equivalence class $[x]_p$

Figure-1. Rough set theory lower and upper approximations [14].

©2006-2017 Asian Research Publishing Network (ARPN). All rights reserved.



www.arpnjournals.com

Rough Set Theory also has a concept of lower approximation. Figure-1 shows representation. The lower approximation is also known as the positive region it is the union of all equivalence classes in [x]p and contains the target set. The upper approximation is the union of all the equivalence classes of [x]p and should have a non-empty intersection with the target set [14].

4. METHODOLOGY

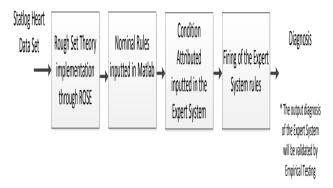


Figure-2. Block diagram of the methodology.

The Statlog Heart Data Set of the UCI machine learning repository will be used as the data. This database consists of 13 attributes which are the following:

- A) age
- B)
- C) chest pain type (4 values)
- D) resting blood pressure
- E) serum cholesterol in mg/dl
- F) fasting blood sugar > 120 mg/dl
- resting electrocardiographic results (values 0,1,2) G)
- maximum heart rate achieved H)
- exercise induced angina I)
- oldpeak = ST depression induced by exercise J) relative to rest
- the slope of the peak exercise ST segment K)
- L) number of major vessels (0-3) colored by fluoroscopy
- M) thal: 3 = normal: 6 = fixed defect: 7 = reversibledefect

Attributes 1 to 12 are the condition attributes while attribute 13 is the decision attribute. In order to determine the value of attribute 13, attributes 1 to 12 should be known. However there are instances where not all the values of the 12 attributes are known so the decision attribute cannot be determined. In order to solve this problem Rough Set theory is used so only the essential attributes are needed to determine the value of the decision attribute. The rules produced will where only the essential attributes are needed are called nominal rules.

Figure-2 shows the block diagram of the Methodology. The values of the Statlog Heart Data Set of the UCI machine learning repository will be inputted in

the Rough Set Data Explorer (ROSE). The ROSE will implement the Rough Set Theory algorithm then output the rules which nominal [15]. These rules will then be inputted in the MATLAB Expert System. Condition attributes will be inputted in the Expert System. After the attributes are inputted the correct Expert System rules will fire. The correct diagnosis will then be outputted by the System. Empirical testing will be used to validate the diagnosis of the Expert System [16].

Empirical testing is performed by:

- Count the total number of rules of the Expert System. The result will be variable a.
- The rules produced by the Expert System and the information system will be compared if they match.
- The total number of matched rules will be defined as variable b.
- The percent validity c will then be computed. It is obtained by using the formula $c = (b/a) \times 100$.

5. DATA AND RESULTS

5.1 Information system of the UCI Statlog heart data

The UCI Statlog Heart Data Set is then converted to an Information System a shown in Figures 1 and 2. For coding purposes a "0" is added at the end of the value of the condition attributes. The condition attributes are A to L while the decision attribute is the thal M. Knowing the correct value of the decision attribute will help the physician in diagnosing cardiovascular disease.

ARPN Journal of Engineering and Applied Sciences ©2006-2017 Asian Research Publishing Network (ARPN). All rights reserved.



www.arpnjournals.com

Table-1. Information system of the UCI Statlog heart data set from A to G.

| # | A | В | C | D | E | F | G |
|----|-----|----|----|------|------|----|----|
| 1 | 670 | 00 | 30 | 1150 | 5640 | 00 | 20 |
| 2 | 570 | 10 | 20 | 1240 | 2610 | 00 | 00 |
| 3 | 640 | 10 | 40 | 1280 | 2630 | 00 | 00 |
| 4 | 740 | 00 | 20 | 1200 | 2690 | 00 | 20 |
| 5 | 650 | 10 | 40 | 1200 | 1770 | 00 | 00 |
| 6 | 560 | 10 | 30 | 1300 | 2560 | 10 | 20 |
| 7 | 590 | 10 | 40 | 1100 | 2390 | 00 | 20 |
| 8 | 600 | 10 | 40 | 1400 | 2930 | 00 | 20 |
| 9 | 630 | 00 | 40 | 1500 | 4070 | 00 | 20 |
| 10 | 590 | 10 | 40 | 1350 | 2340 | 00 | 00 |
| 11 | 530 | 10 | 40 | 1420 | 2260 | 00 | 20 |
| 12 | 440 | 10 | 30 | 1400 | 2350 | 00 | 20 |
| 13 | 610 | 10 | 10 | 1340 | 2340 | 00 | 00 |
| 14 | 570 | 00 | 40 | 1280 | 3030 | 00 | 20 |
| 15 | 710 | 00 | 40 | 1120 | 1490 | 00 | 00 |
| 16 | 460 | 10 | 40 | 1400 | 3110 | 00 | 00 |
| 17 | 530 | 10 | 40 | 1400 | 2030 | 10 | 20 |
| 18 | 640 | 10 | 10 | 1100 | 2110 | 00 | 20 |
| 19 | 400 | 10 | 10 | 1400 | 1990 | 00 | 00 |
| 20 | 670 | 10 | 40 | 1200 | 2290 | 00 | 20 |
| 21 | 480 | 10 | 20 | 1300 | 2450 | 00 | 20 |
| 22 | 430 | 10 | 40 | 1150 | 3030 | 00 | 00 |
| 23 | 470 | 10 | 40 | 1120 | 2040 | 00 | 00 |
| 24 | 540 | 00 | 20 | 1320 | 2880 | 10 | 20 |
| 25 | 480 | 00 | 30 | 1300 | 2750 | 00 | 00 |
| 26 | 460 | 00 | 40 | 1380 | 2430 | 00 | 20 |
| 27 | 510 | 00 | 30 | 1200 | 2950 | 00 | 20 |
| 28 | 580 | 10 | 30 | 1120 | 2300 | 00 | 20 |
| 29 | 710 | 00 | 30 | 1100 | 2650 | 10 | 20 |
| 30 | 570 | 10 | 30 | 1280 | 2290 | 00 | 20 |

Table-2. Information system of the UCI Statlog heart data set from H to M.

| # | Н | I | J | K | L | M |
|----|------|----|----|----|----|---|
| 1 | 1600 | 00 | 16 | 20 | 00 | 7 |
| 2 | 1410 | 00 | 03 | 10 | 00 | 7 |
| 3 | 1050 | 10 | 02 | 20 | 10 | 7 |
| 4 | 1210 | 10 | 02 | 10 | 10 | 3 |
| 5 | 1400 | 00 | 04 | 10 | 00 | 7 |
| 6 | 1420 | 10 | 06 | 20 | 10 | 6 |
| 7 | 1420 | 10 | 12 | 20 | 10 | 7 |
| 8 | 1700 | 00 | 12 | 20 | 20 | 7 |
| 9 | 1540 | 00 | 40 | 20 | 30 | 7 |
| 10 | 1610 | 00 | 05 | 20 | 00 | 7 |
| 11 | 1110 | 10 | 00 | 10 | 00 | 7 |
| 12 | 1800 | 00 | 00 | 10 | 00 | 3 |
| 13 | 1450 | 00 | 26 | 20 | 20 | 3 |
| 14 | 1590 | 00 | 00 | 10 | 10 | 3 |
| 15 | 1250 | 00 | 16 | 20 | 00 | 3 |
| 16 | 1200 | 10 | 18 | 20 | 20 | 7 |
| 17 | 1550 | 10 | 31 | 30 | 00 | 7 |
| 18 | 1440 | 10 | 18 | 20 | 00 | 3 |
| 19 | 1780 | 10 | 14 | 10 | 00 | 7 |
| 20 | 1290 | 10 | 26 | 20 | 20 | 7 |
| 21 | 1800 | 00 | 02 | 20 | 00 | 3 |
| 22 | 1810 | 00 | 12 | 20 | 00 | 3 |
| 23 | 1430 | 00 | 01 | 10 | 00 | 3 |
| 24 | 1590 | 10 | 00 | 10 | 10 | 3 |
| 25 | 1390 | 00 | 02 | 10 | 00 | 3 |
| 26 | 1520 | 10 | 00 | 20 | 00 | 3 |
| 27 | 1570 | 00 | 06 | 10 | 00 | 3 |
| 28 | 1650 | 00 | 25 | 20 | 10 | 7 |
| 29 | 1300 | 00 | 00 | 10 | 10 | 3 |
| 30 | 1500 | 00 | 04 | 20 | 10 | 7 |

${\bf 5.2}$ Rule reduction of the UCI Statlog heart data set

The rules of the Information System were reduced using the Expert System algorithm with the aid of the Rough Set Data Explorer (ROSE).

©2006-2017 Asian Research Publishing Network (ARPN). All rights reserved.



www.arpnjournals.com

Table-3. Rough set rules of the UCI Statlog heart data set.

| Num. | Rule | Case | Value of symptoms in cases |
|------|---|------------------|-------------------------------|
| 1 | $(C = 40) & (D = 1120) \Rightarrow (M = 3)$ | M = 3 | C = 40, D = 1120 |
| 2 | $(C = 30) & (K = 10) \Rightarrow (M = 3)$ | M = 3 | C = 30, K = 10 |
| 3 | $(I = 10) & (K = 20) & (L = 0) \Rightarrow (M = 3)$ | M = 3 | I = 10, K = 20, L = 0 |
| 4 | $(A = 610) \Rightarrow (M = 3)$ | M = 3 | A = 610 |
| 5 | $(C = 20) & (G = 20) \Rightarrow (M = 3)$ | M = 3 | C = 20, G = 20 |
| 6 | $(E = 3030) \Rightarrow (M = 3)$ | M = 3 | E = 3030 |
| 7 | $(A = 560) \Rightarrow (M = 6)$ | $\mathbf{M} = 6$ | A = 560 |
| 8 | (B = 10) & (C = 40) & (I = 10) => (M = 7) | M = 7 | B = 10, C = 40, I = 10 |
| 9 | $(C = 30) & (I = 0) & (K = 20) \Rightarrow (M = 7)$ | M = 7 | C = 30, I = 0, K = 20 |
| 10 | $(C = 40) & (G = 20) & (I = 0) & (K = 20) \Rightarrow$ (M = 7) | $\mathbf{M} = 7$ | C = 40, G = 20, I = 0, K = 20 |
| 11 | $(A = 650) \Rightarrow (M = 7)$ | M = 7 | A = 650 |
| 12 | $(A = 400) \Rightarrow (M = 7)$ | M = 7 | A = 400 |
| 13 | $(D = 1240) \Rightarrow (M = 7)$ | M = 7 | D = 1240 |
| 14 | $(D = 1350) \Rightarrow (M = 7)$ | M = 7 | D = 1350 |

Table-3 shows the reduced rules of the Information System and the Empirical Testing Data.

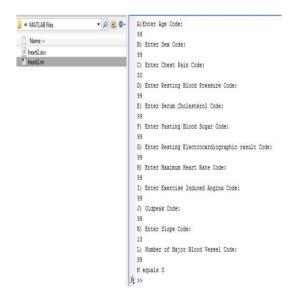


Figure-3. MATLAB implementation of the system.

Figure-3 shows the MATLAB implementation of the system. MATLAB [17] can be used as a tool to apply the rules created by the algorithm with the aid of ROSE.

5.3 Analysis of data

This research showed an Expert System algorithm for diagnosing cardiovascular disease using Rough Set Theory. A program was created in MATLAB to implement the algorithm. A total of 30 sample data from the UCI Statlog Heart Data Set was used to implement the algorithm. Empirical testing was used to verify the accuracy of the algorithm. From the data c = 14

and b = 14. Using the formula for percent validity c =(b/a) x 100. The percent validity of the algorithm is 100%.

6. CONCLUSIONS AND RECOMMENDATIONS

An expert System algorithm for use in the diagnosis of cardiovascular disease was created in this research. The dataset used was the UCI Statlog Heart Data Set. The implementation of the algorithm was done using MATLAB. A total of 30 dataset was used. When the algorithm was applied the rules were reduced to 14 giving a 53.33% reduction. This algorithm cab be used in cardiovascular diagnosis especially when there are missing data or information that cannot be verified. This algorithm can also speed up the process in diagnosis therefore saving precious time.

For future studies it is recommended to try the algorithm to a larger dataset sample. Theoretically the larger the dataset sample this algorithm is used, it will give a more accurate conclusion. This algorithm can also be used to other datasets and can be used as a tool for proper Expert System diagnosis.

REFERENCES

- [1] Falk E. 2006. A Pathogenesis of Atherosclerosis. ARPN Journal of Engineering and Applied Sciences. (47)8: 7-12.
- [2] Young L., Bamason S. and Kupzyk K. 2016. Mechanism of engaging self-management behavior in rural heart failure patients. Applied Nursing Research. (30)1: 222-227.

ARPN Journal of Engineering and Applied Sciences

©2006-2017 Asian Research Publishing Network (ARPN). All rights reserved.



www.arpnjournals.com

- [3] Nahar J., Imam T., Tickle K. and Chen Y. 2013. Association rule mining to detect factors which contribute to heart disease in males and females. Expert System with Applications. (40)4: 1086-1093.
- [4] Kim K., Roh M. and Ha S. 2015. Expert System based on the arrangement evaluation model for the arrangement design of a submarine. Expert System with Applications. (42)22: 8731-8744.
- [5] Jet M. and James M. 2001. An Autonomous Diagnostic and Prognostic Monitoring System for NASA's Deep Space Network. IEEE Aerospace.
- [6] Anto S. and Chandramathi S. 2015. An Expert System for Diabetes Diagnosis using Extreme Learning Machine and Simulated Annealing. International Journal of Applied Engineering Research. (10)2: 32087-32101.
- [7] Africa A. 2016. A Rough Set Based Data Model for Heart Disease Diagnostics. ARPN Journal of Engineering and Applied Sciences. (11)15: 9350-9357.
- [8] Zhou T., Ding J., Wang X. and Xheng X. 2016. Long noncoding RNA and Atherosclerosis. Atherosclerosis. (248)1: 51-61.
- [9] Lynch E., Cumming T., Jensenn H. and Bernhardt J. 2015. Early Mobilization after Stroke: Changes in Clinical Opinion despite an Unchanging Evidence Base. Journal of Stroke and Cerebrovascular Diseases. (26)1: 1-6.
- [10] Statlog Heart Dataset of the UCI Machine learning repository. https://archive.ics.uci.edu/ml/datasets/Statlog+%28He art%.
- [11] Palwak Z. 1982. Rough Sets. International Journal of Information in Computer Science. (11)5: 341-356.
- [12] Palwak Z. 1991. Rough Sets: Theoretical Aspects of Reasoning about data. Boston, MA: Kluwer.
- [13] Palwak Z., Slowinsky K. and Slowinsky R. 1998. Rough classification of patients after highly selective vagotomy for duodenal ulcer. International Journal of Information of Machine Studies. (24)1: 413-433.
- [14] Anitha K. and Venkatesan E. 2013. Feature Selection by Rough Quick Reduct Algorithm. International Journal of Innovative Research in Science, Engineering and Technology. (2)8: 3989-3993.

- [15] ROSE 2.0. 1999. http://www.idss.cs.put.poznan.pl/rose.
- [16] Preece A. 2001. Evaluating Verification Validation Methods in Knowledge Engineering. http://citeseerx.ist.psu.edu/viewdoc/summary?doi=10. 1.1.17.3766.
- [17] MATLAB 2017. http://www.mathworks.com.