Data Mining Classification on Hypo Thyroids Detection: Association Women Outnumber Men

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Abstract: Thyroid diseases are common worldwide and affecting health life. Health care is an inevitable task to be done in human life. In this paper, we have made an attempt to diagnosis hypothyroid. This paper analyzes few essential parameters affecting thyroid. Data mining acts as a solution to many thyroid healthcare problems. To overcome the problem we have come here with a novel solution approach to identify key factors affecting hypothyroid using WEKA tool 3.8. This paper presents thyroid data analysis, classification and prediction. The result of the proposed work used for the forthcoming identification to keep track on important factors affecting hypothyroid.

Keywords: Data Mining, Hypo Thyroidism, Prediction

I. INTRODUCTION

Thyroid one of the most popular disease across the world is not exception for the India. Various studies indicated 42 million people in India undergoes through thyroid disease. Ambika Gopalakrishnan et.al, identified various thyroid related disorder in India includes hypothyroidism, hyperthyroid, goiter, iodine shortage diseases, Hashimoto’s thyroiditis, thyroid cancer. Author worked on determining thyroid hormones normal indication range, particularly during pregnancy on Indian dataset[1].Report identified prevalence of Hypo Thyroidism is around 11% in India.[13]

In this paper, the work is furnished such as in the section – II the outcomes from the parallel researches are discussed. In the section – III Proposed Research Work Approach is discussed. In the next section, Section – IV implementation is discussed, Followed by the results are furnished and explained in the Section – V and the work presents the final conclusion in the Section – VI.

II. RELATED WORK

Sayyad Rasheeduddin et.al used unsupervised Graph Clustering ant Colony Optimization based Extreme Machine Learning technique to detect threat of thyroid. This approach identified factors affecting risk of thyroid. Author found approach superior than univariate logic. The ultrasound technique used for identification found popular, affordable and efficient [2]. Liyong Ma, Chengkuan Ma et.al proposed proficient way using convolutional neural network to detect disease illnesses on SPECT datasets. Suggested approach result worked better than existing methods [3].

K. Rajam et.al noticed data mining supervised functionalities Naïve bayes, decision tree, back propagation, Support vector machine identifies thyroid disease at former phase. Outcomes evaluated based on parameters speed, accuracy, performance and cost and found effective for treatment of the patient [4].Marissa Lourdes De Aitaide et.al applied a two multilayer perceptron classifier for classifying thyroid diseases into three classes as euthyroid, hyperthyroid and hypothyroid and to classify hypothyroid disease into primary, secondary and tertiary hypothyroid with focused on maximum accuracy in minimum time. This classifier gave good accuracy of classification [5].To reduce problem [6],[7],[8],[9], applied neural network for analyzing thyroid problem. Fatemeh Saiti et.al, applied Genetic Algorithms Using Support Vector Machine for thyroid verdict [10]. G. Rasitha Banu predicted problem using Linear Discriminant Analysis (LDA)- Data Mining approach[11].

III. PROPOSED WORK

Various author applied supervised and unsupervised approaches such as machine learning, neural network, Naive bayes, decision tree, back propagation, multilayer perceptron, Genetic Algorithm and Support vector machine. Further future work scope is available to expand on larger size datasets. Inadequate data blocks need detail classification and diagnosis of thyroid diseases. The data obtained from diagnoses by professional physicians can serve as fundamental correct source of data. We need sophisticated system to identify thyroid and also to recommend level that can further leads to serious problems like cancer. We are using in our proposed work data mining approach. We have taken datasets hypothyroid from website source github. Thus the proposed novel developed architecture for Hypo Thyroid Detection is available in Fig. 1.

Fig1. Proposed Architecture for Hypo Thyroid Detection and classification.

Our proposed system architecture Hypo Thyroid Detection and classification consist of following

A. Explore Dataset in WEKA Tool

We used free Hypothyroid dataset from github library and explore it in WEKA Data Mining Tool.

B. Preprocess dataset

Here dataset is preprocesses for all attributes. Preprocessing is the basic step in data mining whether we perform...
check for attributes and instances.

C. Analyze Attribute Type
We perform check for all attribute type as nominal numeric or class type.

D. Apply Info Gain attribute Eval Method
We applied Info Gain Attribute Eval Method and identified which attribute are more dangerous for Hypothyroid

E. Identify Important Attributes
Ranking of Attribute is considered here using Ranker Algorithm to identify important attribute.

F. Classification of Dataset
We have classified dataset under 4 labels such as negative, compensated_hypothyroid,primary_hypothyroid,secondary_hypothyroid to understand severity of disease and further to take suitable measure and care

G. Prediction of Dataset
The obtained parameter can further use to predict Hypothyroid for the future to keep track on important factors affecting hypothyroid.

H. Data Visualization
From the retrieved result BarGraph, PieChart used as a medium Visualization

IV. IMPLEMENTATION OF PROPOSED WORK
We considered dataset of 30 attributes which includes age, sex, on thyroxine, query on thyroxine, on antithyroid medication, sick, Pregnant, thyroid surgery,I131 treatment, query hypothyroid, query hyperthyroid, lithium, goiter, tumor, hypopituitary, psych, TSH measured,TSH,T3 measured,T3,TT4 measured,TT4,T4U measured,T4U,FTI measured, FTI,TBG measured, TBG, referral source, Class. We focused on Identification of Attribute Type, Identification of Important Attribute contributing to thyroid, Classify the dataset using ZeroR Classifier.

Identification of Attribute Type
The first outcome of this work is to identify attribute under 3 category nominal, numeric and class.
Nominal Attributes of the dataset is shown in Table-I:
Table-I: Nominal Attribute Identification

| Sr. No | Attribute Name | Attribute Category | Number | Label | Count |
|--------|----------------|--------------------|--------|-------|-------|
| 1      | Sex            | Nominal            | 1      | F     | 2480  |
| 2      | on thyroxline  | Nominal            | 1      | f     | 3308  |
|        |                |                    | 2      | t     | 464   |
| 3      | query on thyroxline | Nominal    | 1      | f     | 3722  |
|        |                |                    | 2      | t     | 50    |
| 4      | on antithyroid medication | Nominal | 1      | f     | 3729  |
|        |                |                    | 2      | t     | 43    |
| 5      | Sick           | Nominal            | 1      | f     | 3625  |
| 6      | Pregnant       | Nominal            | 1      | f     | 3719  |
| 7      | thyroid surgery | Nominal            | 1      | f     | 3719  |
| 8      | I131 treatment | Nominal            | 1      | f     | 3713  |

Classification attribute of the dataset is shown in Table-III:
Table-II: Numeric Attribute Identification

| Sr.No | Attribute Name | Attribute Category | Statistic | Value |
|-------|----------------|--------------------|-----------|-------|
| 1     | Age            | Numeric            | Minimum   | 1     |
|       |                |                    | Maximum   | 455   |
|       |                |                    | Mean      | 51.736|
|       |                |                    | StdDev    | 20.085|
| 2     | TSH            | Numeric            | Minimum   | 0.005 |
|       |                |                    | Maximum   | 530   |
|       |                |                    | Mean      | 5.087 |
|       |                |                    | StdDev    | 24.521|
| 3     | T3             | Numeric            | Minimum   | 0.05  |
|       |                |                    | Maximum   | 10.6  |
|       |                |                    | Mean      | 2.013 |
|       |                |                    | StdDev    | 0.827 |
| 4     | TT4            | Numeric            | Minimum   | 2     |
|       |                |                    | Maximum   | 430   |
|       |                |                    | Mean      | 108.319|
|       |                |                    | StdDev    | 35.604|
| 5     | T4U            | Numeric            | Minimum   | 0.25  |
|       |                |                    | Maximum   | 2.32  |
|       |                |                    | Mean      | 0.995 |
|       |                |                    | StdDev    | 0.195 |
| 6     | FTI            | Numeric            | Minimum   | 2     |
|       |                |                    | Maximum   | 395   |
|       |                |                    | Mean      | 110.47|
|       |                |                    | StdDev    | 33.09 |
| 7     | TBG            | Numeric            | Minimum   | NaN   |
|       |                |                    | Maximum   | NaN   |
|       |                |                    | Mean      | NaN   |
|       |                |                    | StdDev    | NaN   |
| 8     | referral source| Numeric            | SVHC      | 386   |
|       |                |                    | other     | 2201  |
|       |                |                    | SVI       | 1034  |
|       |                |                    | STMW      | 112   |
|       |                |                    | SVHD      | 39    |

Numerical Attributes of the dataset is shown in Table-II:
Table III: Classification Attribute Identification

| Sr. No | Classificaton | Attribute Type | Classification Type | Classification Result | 
|--------|----------------|----------------|----------------------|-----------------------|
| 1      | Class          | Numeric        | negative             | 3481                  |
|        |                |                | compensated_         | 194                   |
|        |                |                | hypothyroid          |                       |
|        |                |                | primary_             | 95                    |
|        |                |                | hypothyroid          |                       |
|        |                |                | secondary_           | 2                     |
|        |                |                | hypothyroid          |                       |

A. Identification of Important Attribute

The second outcome of this work is to identify Important attribute. We applied InfoGain Attribute Eval function on dataset which consist of 3772 instances and 30 attributes. Considering all train data with Information Gain Ranking Filter we got result of Selected attributes: 18, 26, 22, 20, 17, 29, 10, 24, 21, 12, 2, 16, 19, 6, 7, 8, 13, 25, 2, 11, 14, 9, 15, 28, 27, 1 : 29 Ranked attributes shown in Table IV:

Table IV: Ranked Attribute using InfoGain Attribute Eval Method

| Attribute Name | Info Gain Ranking Value |
|----------------|-------------------------|
| age            | 29                      |
| sex            | 11                      |
| On thyroxine   | 6                       |
| Query onthyroxine | 21                     |
| On antithyroidmedication | 20                 |
| sick           | 14                      |
| pregnant       | 15                      |
| Thyroid surgery | 16                    |
| I131 treatment | 25                      |
| Query hypothyroid | 8                    |
| Query hyperthyroid | 22                  |
| Lithium        | 23                      |
| Goiter         | 17                      |
| Tumor          | 24                      |
| hypothitary    | 26                      |
| Psych          | 12                      |
| TSH measured   | 5                       |
| TSH            | 1                       |
| T3 measured    | 13                      |
| T3             | 4                       |
| TT4 measured   | 10                      |
| TT4            | 3                       |
| T4U measured   | 18                      |
| T4U            | 9                       |

C. Classify the dataset

The third outcome of this work is to perform classification of dataset. We used ZeroR classifier and retrieved results on dataset name hypothyroid with 30 attributes and 3772 data values. Results are observed in 0.02 seconds. We found properly classified values as 3481 and improperly classified instances as 291. We obtained 92.2853 % of accuracy positive and 7.7147% negative accuracy of classification.

V. RESULT ANALYSIS

A. InfoGain Attribute Evaluation Results

The comparison is presented in Graphical Format.
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The results are been visualised graphically to understand the classification Fig. 3.

![Classified Data](image)

**Fig.3.Classification of Data**

C. Female and Mail Classification and graph

The graph shows female are more prone to hypothyroidism than male depicted in Fig.4.

![Fig. 4. Female and Male Classification](image)

VI. CONCLUSION

We used WEKA tool 3.8 and perform classification using Zero R on dataset Correctly Classified Instances 3481 i.e 92.2853 % and Incorrectly Classified Instances 291 7.7147 %.Under 4 category classified data with label negative3481(compensated_hypothyroid194,primary_hypothyroid95secondary_hypothyroid 2.This paper presents thyroid data analysis, classification and prediction. The outcome of the extracted data can be analyzed for the future to keep track on important factors affecting hypothyroid. Here we have also identified that Female contributes more comparatively than male towards hypothyroid. The obtain information serves useful and plays vital role in future for women to be careful towards thyroid.

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