

# **Machine Learning Prediction Algorithm for Chronic Diseases Detection over Bigdata**

Himanshu<sup>1</sup>, Kale Hanumanth<sup>2</sup>

\*1.2 Lecturer Department of Computer Science and Engineering, AL-KHATEEB Polytechnic College, Bangalore, Karnataka, India

#### ABSTRACT

Article Info Volume 7, Issue 4 Page Number : 527-531

**Publication Issue :** July-August-2021

Article History Accepted: 08 Aug 2021 Published : 16 Aug 2021 With big data growth in biomedical and healthcare communities, accurate analysis of medical data benefits early disease detection, patient care and community services. However, the analysis accuracy is reduced when the quality of medical data is incomplete. Moreover, different regions exhibit unique characteristics of certain regional diseases, which may weaken the prediction of disease outbreaks. In this paper, we streamline machinelearning algorithms for effective prediction of chronic disease outbreak in disease-frequent communities. We experiment the modified prediction models over real-life hospital data collected from central China in 2013- 2015. To overcome the difficulty of incomplete data, we use a latent factor model to reconstruct the missing data. We experiment on a regional chronic disease of cerebral infarction. To the best of our knowledge, none of the existing work focused on both data types in the area of medical big data analytics. Compared to several typical prediction algorithms, the prediction accuracy of our proposed algorithm reaches 94.8% with a convergence speed which is faster than that of the CNN-based unimodal disease risk prediction (CNN-UDRP) algorithm.

**Keywords** : Big data analytics, Machine Learning, Healthcare.

#### I. INTRODUCTION

According to a report by McKinsey [1], 50% of Americans have one or more chronic diseases and 80% many other countries. In China, chronic diseases are of American medical care fee is spent on chronic disease treatment. With the improvement of living standards, the incidence of chronic disease is increasing. The United States has spent an average of 2.7 trillion USD annually on chronic disease

treatment. This amount comprises 18% of the entire annual GDP of the United States. The healthcare problem of chronic diseases is also very important in the main cause of death, according to a Chinese report on nutrition and in 2015, 86.6% of deaths are caused due to chronic diseases. Therefore, it is essential to perform risk assessments for chronic diseases. With the growth in medical data [2], collecting electronic

**Copyright:** © the author(s), publisher and licensee Technoscience Academy. This is an open-access article distributed under the terms of the Creative Commons Attribution Non-Commercial License, which permits unrestricted non-commercial use, distribution, and reproduction in any medium, provided the original work is properly cited



health records (EHR) is increasingly convenient [3]. Besides,[4] first presented a bio inspired highperformance heterogeneous vehicular telemetric paradigm, such that the collection of mobile users' health related real-time big data can be achieved with the deployment of advanced heterogeneous vehicular networks. Chen et.al. [5]– [7] proposed a healthcare system using smart clothing for sustainable health monitoring.

# A. MOTIVATION

To solve these problems, it see the structured and unstructured data in healthcare field to assess the risk of disease. First, the system use Decision tree map algorithm to generate the pattern and causes of disease. It clearly shows the diseases and sub diseases. Second, by using Map Reduce algorithm for partitioning the data such that a query will be analyzed only in a specific partition, which will increase the operational efficiency but reduce query retrieval time.

### **B. PROBLEM STATEMENT**

- Chronic Disease Prediction by Machine Learning over Big Data from Healthcare Communities.
- The healthcare problem of chronic diseases is also very important in many other countries.
- In China, chronic diseases are the main cause of death, according to a Chinese report on nutrition and chronic diseases in 2015, 86.6% of deaths are caused by chronic diseases. Therefore, it is essential to perform risk assessments for chronic diseases.

# C. Objective of the Project

- The analysis accuracy is reduced when the quality of medical data in incomplete. Moreover, different regions exhibit unique characteristics of certain regional diseases, which may weaken the prediction of disease outbreaks.
- However, those existing work mostly

considered structured data. There is no proper methods to handle semi structured and unstructured.

• The proposed system will consider both structured and unstructured data. The analysis accuracy is increased by using Machine Learning algorithm.

# **II. LITERATURE SURVEY**

In [1], P. Groves, B. Kayyali, D. Knott, and S. V. Kuiken, based on his experience the stacks of healthcare data global are increasing exponentially. To supply high value healthcare at lower cost there is a need felt in the direction of implementing successful big data analytics pract ice to find insight from very large data sets and coordination, improving care quality and outcomes through improved analytics of healthcare big data.

In [2], M. Chen, S. Mao, and Y. Liu, he stated In this paper, we review the background and stateof- the-art of big data. We first introduce the general background of big data and review related technologies, such as could computing, Internet of Things, data centers, and Hadoop.

In [3], P. B. Jensen, L. J. Jensen, and S. Brunak, An electronic health record is a systematic collection of electronic health information about an individual patient or population.

In [4], D. Tian, J. Zhou, Y. Wang, Y. Lu, H. Xia, and Z. Yi, We propose a bio-inspired model for making handover decision in heterogeneous wireless networks. It is based on an extended attractor selection model, which is biologically inspired by the self-adaptability and robustness of cellular response to the changes in dynamic environments.



In [5] S. Bandyopadhyay, J. Wolfson, D. M. Vock, Survival prediction models most commonly use Cox Propor-tional Hazards (CPH) models, and are frequently used in medical statis-tics and clinical practice. However, such models underperform when thepredictor variables are missing. By building Bayesian networks we automatically construct a model with the most important risk factors and relationships between risk factors and Bayesian networks are able to inferthe likely values of missing data.

In [6]J. Wan, S. Tang, D. Li, S. Wang, C. Liu, Nowadays the use of Big Data is increasing in biomedical and healthcare communities, accurate analysis of medical data benefits early disease detection, patient care, and community services. Incomplete medical data reduces analysis accuracy. The framework proposes Decision Tree calculation and Naive Bayesian calculation for disease prediction. We are applying both Naïve Bayes and Decision Tree algorithm on the data set and by checking the accuracy of each algorithm, we are taking the output which is more accurate.

### **III. PROPOSED SYSTEM**

In a proposed system, we can first get the large volume of a healthcare big data, then that data is considered as training data. Naive Bayes algorithm is used for the clarification of the data. Then after the clarification, the hospital data similar type of data can be stored. Then CNN extract the text characteristics automatically. In that we use a CNN MDRP algorithm that uses both structured unstructured hospital data. Selecting the characteristics automatically form a large number of data. This improves the disease prediction than rather previously selected characteristics. CNN- MDRP algorithm helps to accuracy of the result of a disease prediction over a large volume of data from hospital.

An easy way to comply with the conference paper formatting requirements is to use this document as a template and simply type your text into it.

We combine the structured and unstructured data in healthcare field to assess the risk of disease. First, we used latent factor model to reconstruct the missing data from the medical records collected from a hospital in central China. Second, by using statistical knowledge, we could determine the major chronic diseases in the region. Third, to handle structured data, we consult with hospital experts to extract useful features.



Figure 1: Processing Steps in Disease Prediction

The disease risk model is obtained by the combination of structured and unstructured features. Through the experiment, we draw a conclusion that the performance of CNN- MDPR is better than other existing methods.



# IV. ARCHITECTURE



Figure 2: Classifying of Algorithms

The above figure shows Naive Bayes classification is a simple probabilistic classifier. It requires calculating the probability of feature attributes. In this experiment, we use conditional probability formula to estimate discrete feature attributes and Gaussian distribution to estimate continuous feature attributes. The KNN classification is given a training data set, and the closest k instance in the training data set is found. For KNN, it is required to determine the measurement of distance and the selection of k value. In the experiment, the data is normalized at first. Then we use the Euclidean distance to measure the distance. As for the selection of parameters k, we find that the model is the best when k = 10. Thus, we choose k = 10. We choose classification and regression tree (CART) algorithm among several decision tree (DT) algorithms. To determine the best classifier and improve the accuracy of the model, the 10-fold crossvalidation method is used for the training set, and data from the test set are not used in the training phase.

# V. BENEFITS

- 1) Higher Accuracy.
- We leverage not only the structured data but also the text data of patients based on the proposed CNN-MDPR algorithm.
- We find that by combining these two data, the accuracy rate can reach 94.80%, to better evaluate the risk of cerebral infarction disease.
- 4) To the best of our knowledge, none of the existing work focused on both data types in the area of medical big data analytics.

### VI.CONCLUSION

Machine learning Decision tree map algorithm by using structured and unstructured data from hospital. It also uses Map Reduce algorithm for partitioning the data. To the highest of gen, none of the current work attentive on together data types in the zone of remedial big data analytics. Compared to several typical calculating algorithms, the scheming accuracy of our proposed algorithm reaches 94.8% with an regular speed which is quicker than that of the CNNbased unimodal disease risk prediction (CNN-UDRP) algorithm and produces report. The report consists of possibility of occurrences of diseases.

### VII. REFERENCES

- A. B. Author, "Title of chapter in the book," in Title of His Published Book, xth ed. City of Publisher, Country if not
- [2] P. Groves, B. Kayyali, D. Knott, and S. V. Kuiken, "The 'big data' revolution in healthcare: Accelerating value and innovation," 2016.
- [3] M. Chen, S. Mao, and Y. Liu, "Big data: A survey," Mobile Networks and Applications, vol. 19, no. 2, pp. 171–209, 2014.



- [4] S. Bandyopadhyay, J. Wolfson, D. M. Vock, G. Vazquez-Benitez, G. Adomavicius, M. Elidrisi, P. E.Johnson, and P. J. O'Connor, "Data mining for censored qtime-to-event data: a bayesian network model for predicting cardiovascular risk from electronic health record data," Data Mining and Knowledge Discovery, vol. 29, no. 4, pp. 1033–1069, 2015..
- [5] D. Tian, J. Zhou, Y. Wang, Y. Lu, H. Xia, and Z. Yi, "A dynamic and self-adaptive network selection method for multimode communications in heterogeneous vehicular telematics," IEEE Transactions on Intelligent Transportation Systems, vol. 16, no. 6, pp. 3033–3049, 2015.
- [6] N. Nori, H. Kashima, K. Yamashita, H. Ikai, and Y. Imanaka, "Simultaneous modeling of multiple diseases for mortality prediction in acute hospital care," in Proceedings of the 21th ACM SIGKDD International Conference on Knowledge Discovery and Data Mining. ACM, 2015, pp. 855–864.
- [7] K. Hwang, M. Chen, "Big Data Analytics for Cloud/IoT and Cognitive Computing," Wiley, U.K., ISBN: 9781119247029, 2017.
- [8] S. M. Chu, W.-T. Shih, Y.-H. Yang, P.-C. Chen, and Y.-H. Chu, "Use of traditional chinese medicine in patients with hyperlipidemia: A population-based study in taiwan," Journal of ethnopharmacology, vol. 168, pp. 129–135, 2015.
- [9] B. Qian, X. Wang, N. Cao, H. Li, and Y.-G. Jiang, "A relative similarity based method for interactive patient risk prediction," Data Mining and Knowledge Discovery, vol. 29, no. 4, pp. 1070–1093, 2015.

#### Cite this article as :

Himanshu, Kale Hanumanth, "Machine Learning Prediction Algorithm for Chronic Diseases Detection over Bigdata ", International Journal of Scientific Research in Computer Science, Engineering and Information Technology (IJSRCSEIT), ISSN : 2456-3307, Volume 7 Issue 4, pp. 527-531, July-August 2021.

Journal URL : https://ijsrcseit.com/CSEIT2174125