



Machine Learning to provide automated, deeper understanding of health care data to improve patient outcomes

Rajeev Kapoor¹, Anitha.T², K. Karthikeyan³, Aleem Ali⁴

1. Assistant Professor in Punjabi University Neighbourhood Campus, Jaito, Faridkot, India.

2. Assistant Professor, Department of Science and Humanities, Sri Krishna College of Engineering and Technology, Kuniyathur, Coimbatore- 08. TamilNadu, India.

3. Associate Professor, Department of Electrical and Electronics Engineering, Ramco Institute of Technology, Rajapalayam - 626117, Tamil Nadu, India.

4. Department of Computer Science and Engineering, School of Technology, Glocal University Saharanpur. U.P (India)-247121.

Correspondence Email: kaps.rajeev@gmail.com

ABSTRACT

Machine learning has sparked a great deal of interest in the latest period that lower computer memory and processing, which makes it easier to store, interpret, & massive amounts of data. Improved techniques were already being developed and tested on massive sets of data to aid in the discovery of hidden insights and connections between data items that have not been visible to us. These analyses help businesses make better choices and enhance key performance measures. The reality that learning algorithms were agnostic to the field of applications, contributes to huge machine learning trend. For instance, classification methods that can be used to classify defects in windmill slides can be used to classify viewers in a study. The ability to customize and use these methods to special real concerns can be what gives machine learning its true value. The researchers use convolutional neural networks to assess tumour progression and aggressiveness using computerised categorization of cellular images in this study. Although this study is in its early stages, it demonstrates the importance of machine learning approaches to delivering fast, efficient and automated data processing. Machine learning provides hope for early detection of illness, helps clients make informed medical decisions, and improves the overall standard of living.

Keywords: Machine learning; Classification algorithms; automated data analysis

Received: 20.02.2022

Revised: 07.03.2022

Accepted: 18.04.2022

INTRODUCTION

Information science aimed to find new facts in the information that could be used to make smart decisions and to provide valuable intelligence. Technology like machine learning techniques has been at the heart of data science. These strategies aren't fresh; they've been around for years [1]. Machine learning has been impractical until now due to a lack of adequate information to build models & most critically, processing capacity [2]. Machine Learning has sparked a lot of interest in the last ten years, because of cheaper computer memory and processing, which makes it easier to store, process, & analyze information gathered from a variety of organizations [3]. The introduction of new database technology has dramatically reduced the education cost of collecting big amounts of information. Machine Learning has seen a resurgence of interest & increased funding in studies to develop abilities by leveraging newly available processing capacity [4-6]. By constructing a mathematical design that accurately describes the data, artificial intelligence provides techniques for evaluating information and creating important aspects of relationships among the variables [5]. It's now possible to build computer simulations rapidly & autonomously that could evaluate bigger, better big quantities of information & provide quicker, extra sufficient insight or analysis - even on a massive scale [6]. The underlying insights and correlations provided by modeling were frequently not visible to the naked eye or imaginable to an analyst [7].

This information aid firms in making important choices about operations, pricing, consumers, and other crucial variables. Symbol Intelligence was an area of artificially intelligent study that was built in high-level symbols of information [8-10]. Its origins seem to be in mathematics & psychology. From the mid-1950s through the late 1980s, symbol AI has been the dominant approach to AI technologies. The concept was dependent on the premise that manipulating symbols could attain several elements of knowledge

[11]. Issues can be resolved in this perspective by filling in the blanks in the present understanding [12]. Computer Algorithms, which consists of a series of multiple if-then-else principles linking respondents explained that could be browsed & recovered to make conclusions seem to be the most effective result of this concept. This paradigm focuses on our intuition interpretation, in which people reinforce our ideas as additional data become available [13].

REVIEW OF LITERATURE

Our opinions were founded on prior distributions, which represent past facts. The posterior distribution [14] describes human opinions development of new experiences. Utilizing Bayes' theorem [15], Bayesian analysis allows to create procedure yields in terms of the likelihood function. This method of Machine Education is based on neuroscience, to simulate brain activity that used a computing framework termed an adaptive intelligence network.

An Artificial Neural Network (ANN) was composed of multiple layers of networks linked by links with a weight that channel functions in adjacent cells. An ANN could have an operational or dormant node [16-18]. In an ANN, data were propagated by sending activity from one cluster to its neighbors. If a cluster becomes engaged, it could only stimulate its neighbors if their weights were high. An ANN, more like a biological equivalent, stores information by distributing weight & authorizations throughout the channel's vertices. This contrasts sharply with the Pictorial paradigm [19]. The capacity of a neural network to adapt, which would be accomplished through altering parameters, was one of its most important features. Machine Learning takes its cues from evolutionary, which would be nature's most basic learning process. It's a catchall term describing teaching styles related to biological evolutionary concepts, including genetic heritage & survival of the strongest [20]. Incremental development, like a developmental process in a group of potential answers, has been used in the evolutionary algorithm approach. To reach the required outcome, this group would then be chosen via natural selection depending on a fitness value. To investigate the optimum solution & offer variety in the answer, ideas like cross-pollination & mutations have been used.

MATERIAL AND METHODS

There are at least three fields of medical information that can benefit very quickly from data science technology. Each year, thousands of people were served in hospitals across the country, & billions more operations were conducted. Depending on forecasts, machine learning technology can help analyze trends in hospital stays, illnesses, & disasters, and also efficient strategies to minimize visits, population changes, & disease control. Knowing electronic health records could also assist healthcare professionals in making well-informed decisions about how to improve healthcare outcomes. Laboratory findings, consumer information, medical images, & physicians' comments were examples of clinical evidence that could be connected & categorized for enhanced clinical information systems [21]. Medical professionals could view the big image by combining information from diverse sources. With the growing population, clinical evidence would only continue to rise at a fast rate, considering issues associated with the manual examination of that kind of information, as it was currently doing, nearly unfeasible. An increasing number of tests are being done all around the world to better understand illnesses & strategies for dealing with them. Massive volumes of rich information would've been ideal for using data analysis methods to uncover ideas, correlations, & connections in the information.

RESULT AND DISCUSSION

The application of the Connection between financial to service was introduced of cells designed to check breast tumor growth. And throughout their lives, one in every eight American women would get invasive cancer. The tumor was best treated if it has been detected early, which would be usually done through routine screening checks. Ductal Carcinoma in Situ (DCIS) seems to be a type of early-stage breast cancer in which cells inside the channels had begun to develop quickly and have not yet breached the duct wall or disseminated to the outside layers. DCIS would be a type of pre-invasive, metastatic disease, and about 25% of DCIS patients go on to develop invasive cancer of the chest. Usual Ductal Hyperplasia (UDH) would be a harmless condition that requires no intervention. Lumpectomy, along with radiotherapy or hormonal treatment, or mastectomy has been the most common treatment for DCIS. The present categorization of different degrees of DCIS vs. UDH during biopsy was dependent on pathologists' observations of the cells under a microscope.

Earlier studies have shown Regression Analysis may be used to classify cells in two ways [13]. We intended to see if we could use cell information regarding photos for a 4-way classification utilizing nonlinear methods like Neural Networks in this study. The objective behind this study is that correctly categorizing the stage of DCIS will substantially assist patients in choosing the optimal treatment

option. The image extract features in [13] were our starting point. In [13], 374 characteristics were retrieved from around 200 images. As a result, an image was described in the input dataset by a vector of about values. Researchers did Dimension Reduction by selecting only one characteristic from every set of linked characteristics because multiple features seemed to be highly associated. Table 1 shows how researchers decreased the set of features to 55 elements & utilized them to train the system. [-1, +1] was used to equalize the input variables. The network was trained on the objective of successfully classifying images into one of four degrees of breast cancer aggressiveness.

Table 1. Inputs and Outputs

Mean_Area	Mean_AR	Mean_Solidity
Mean_skewness_R	SD_variance_R	Mean_SRE_R
SD_energy_R	SD_SRE_R	SD_LRE_R
Mean_LRE_G	Mean_LGLRE_G	SD_interia_G
Mean_SRE_B	Mean_LRE_B	SD_LRE_G
SD_SRE_B	Mean_SRE_HSV	Mean_LRE_HSV
Mean_kurtosis_R	Mean_SRE_G	SD_IDM_G
Mean_LGLRE_R	SD_skewness_B	SD_LRE_HSV
SD_Circumference	SD_mean_G	SD_Correlation_B
Mean_LRE_R	SD_SRE_G	SD_SRE_HSV
SD_LRE_HSV	Mean_LRE_Lab	Mean_GLN_Lab
SD_SRE_Lab	Mean_SRE_Lab	SD_LRE_Lab
Mean_SRE_Luv	Mean_LRE_Luv	SD_SRE_Luv
SD_LRE_Luv	SD_Mean	Mean_SRE_HE
Mean_LRE_HE	Mean_HGLRE_HE	SD_SRE_HE
SD_LRE_HE	SD_kurtosis_BR	Mean_SRE_BR
Mean_LRE_BR	SD_SRE_BR	SD_LRE_BR

The neural network was trained using 4 different value updating methods. The following are the details:

1. The Levenberg-Marquardt achieve ideal seems to be a classical machine learning algorithm for adjusting the network's connection weights. This has been the most efficient optimization strategy. It does, however, necessitate a greater amount of storage than other methods.

2. Resilient Backpropagation operates independently of every load & considers only its signature of the difference equation across all patterns (not the magnitude). That's a neural network learning algorithm that keeps track of weights & biases.

3. Bayesian inference A statistical procedure known as regularization transforms a prediction equation into a well-posed statistics issue. Throughout learning, it utilizes Levenburg-Marquardt optimization to adjust the weights & biases of a neural network. It identifies the proper mix of squares of the errors & values to construct a network that generalizes well by minimizing a mixture of squares of the errors & values. It takes a very long time to learn this technique.

4. Simplified Conjugate Gradient would be an approach that relies on the Conjugate Gradient Techniques, a family of optimization algorithms well enough in quantitative simulation. It makes use of 2nd data from the neural network, but only uses, storage, where N represents the size of connection weights. Weights values are updated using this method.

We reserved 10% for verification & 10% for evaluation in place to evade overtraining the system. The system was learned on the training dataset, which comprised 80% of the information; the test dataset has been used to decide when to cease learning, and also test set has been used to assess the network's performance. For every one of the previous paragraph training approaches, researchers performed this operation 100 times. Figures 1 to 4 illustrate the Neural Network's performance compared to four different training techniques. The findings show that Breast Cancer Malignancy Grades may be accurately identified using Neural Networks with greater than 90% reliability. Without any domain expertise of cellular disease, researchers were able to decrease 393 characteristics to 55 characteristics using the Dimensional Reduction technique. The Bayesian Regularization learning model provided the highest efficiency of the 4 kinds of Neural Networks tested for learning. The Bayesian Regularization process results in a well generalizing system. The LevenbergMarquardt Optimization technique took the least amount of time to learn.

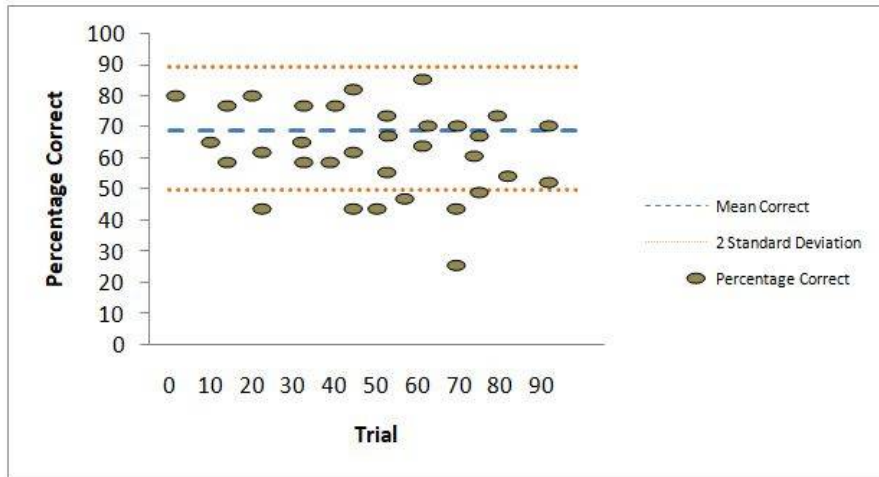


Figure 1: Comparison of propose with Gradient Descent Method

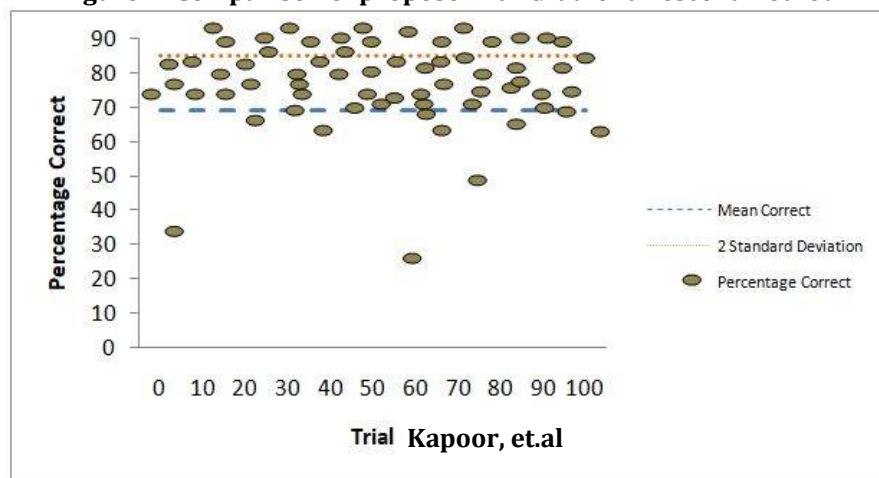


Figure 2: Comparison of propose with Levenburg-Marquardt

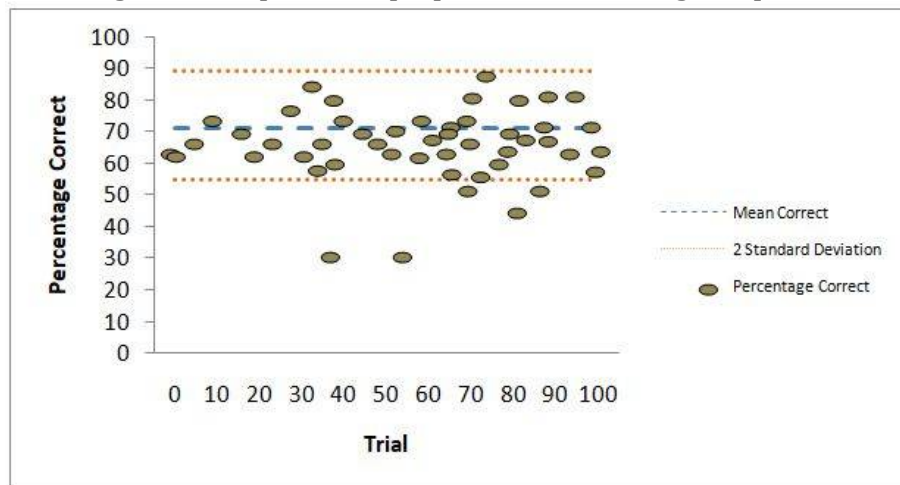


Figure 3: Comparison of propose with Resilient Backpropagation

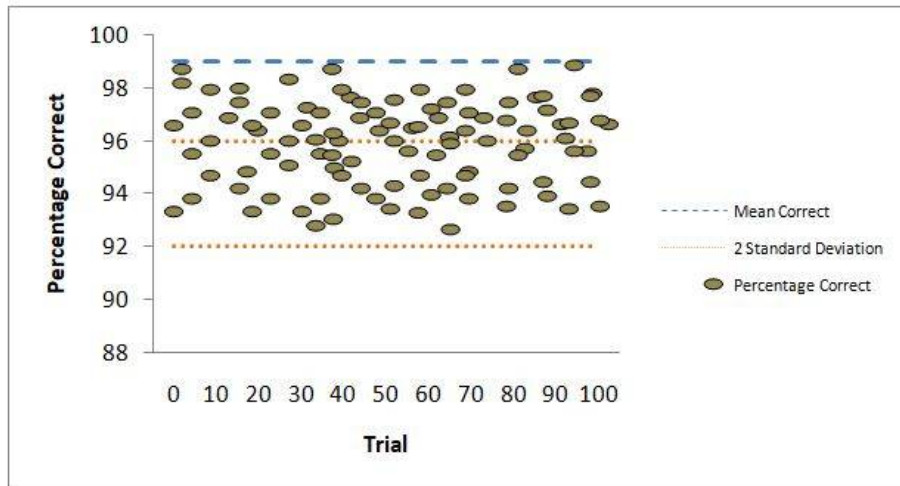


Figure 4: Comparison of propose with Bayesian Regularization

Apart from Bayesian Regularization, regressive approaches simply update the weights are adjusted levels, not the squared mistakes & values as in Bayesian Regularization. Although the Bayesian Regularization approach takes a long time to learn, it is the least prone to overtraining. The research presented here was conducted using a limited data set obtained from [14]. In addition, it was used to manually extract the characteristics of the image data. Researchers would rather use Deep Learning with CNN to learn the classifier model as a further step in this research, once we have a large-scale dataset. CNN's were capable of independently detecting objects of interest in images & classifying them depending on such characteristics. The use of CNN comes with the limitation that it implies the availability of enormous amounts of data. As a result, we were unable to apply this strategy to the tiny dataset we had. The ability to swiftly and accurately define the stage of DCIS can make a big difference for medical professionals, allowing them to make better medical decisions. Furthermore, doing such analyses on patients on a routine basis could reveal the rate at which cancer cells develop, assisting healthcare professionals in determining the extent of health interventions required to prevent cancer from spreading.

CONCLUSIONS

Researchers have detailed the automatic identification of cancer cells in this study. Disorders like Alzheimer's & cancer will be on the increase, & they understand the agony & suffering that they cause patients and family members. Any advancement in a framework that involves early detection of diseases, quantification of progression of the disease, & identification of disease severity would make a positive difference in the lives of patients and families. The research papers presented in this article were preliminary findings. To deliver more actual diagnoses, further effort needs to be done to broaden the analysis to larger and richer datasets. Medical analytics do not become a new concept, but today's modern methodologies and big data have boosted automated medical diagnostics to a degree that was unthinkable just a few decades earlier. With a population increase & vast quantities of medical data that is generated, Machine Learning devices are becoming increasingly important for providing fast, automatic, & better understanding of medical information to identify individuals' well-being.

ACKNOWLEDGEMENT

The authors acknowledge the subjects who were involved in the study.

CONFLICT OF INTEREST

The authors declare that there is no conflict of interest for this study.

REFERENCES

1. Manogaran, G., Shakeel, P. M., Hassanein, A. S., Kumar, P. M., & Babu, G. C. (2018). Machine learning approach-based gamma distribution for brain tumor detection and data sample imbalance analysis. *IEEE Access*, 7, 12-19.
2. Waring, J., Lindvall, C., & Umeton, R. (2020). Automated machine learning: Review of the state-of-the-art and opportunities for healthcare. *Artificial Intelligence in Medicine*, 104, 101822.
3. Buettner, R., Grimmeisen, A., & Gotschlich, A. (2020, January). High-performance diagnosis of sleep disorders: a novel, accurate and fast machine learning approach using electroencephalographic data. In *Proceedings of the 53rd Hawaii International Conference on System Sciences*.

4. Garikapati, P., Balamurugan, K., Latchoumi, T. P., & Malkapuram, R. (2021). A Cluster-Profile Comparative Study on Machining AlSi7/63% of SiC Hybrid Composite Using Agglomerative Hierarchical Clustering and K-Means. *Silicon*, 13(4), 961-972.
5. Mir, A., & Dhage, S. N. (2018, August). Diabetes disease prediction using machine learning on big data of healthcare. In *2018 fourth international conference on computing communication control and automation (ICCCUBEA)* (pp. 1-6). IEEE.
6. Latchoumi, T. P., Balamurugan, K., Dinesh, K., & Ezhilarasi, T. P. (2019). Particle swarm optimization approach for waterjet cavitation peening. *Measurement*, 141, 184-189
7. Ngiam, K. Y., & Khor, W. (2019). Big data and machine learning algorithms for healthcare delivery. *The Lancet Oncology*, 20(5), e262-e273.
8. Bolhasani, H., Mohseni, M., & Rahmani, A. M. (2021). Deep learning applications for IoT in health care: A systematic review. *Informatics in Medicine Unlocked*, 23, 100550.
9. Latchoumi, T. P., Ezhilarasi, T. P., & Balamurugan, K. (2019). Bio-inspired weighed quantum particle swarm optimization and smooth support vector machine ensembles for identification of abnormalities in medical data. *SN Applied Sciences*, 1(10), 1-10.
10. Guarin, D. L., Dusseldorp, J., Hadlock, T. A., & Jowett, N. (2018). A machine learning approach for automated facial measurements in facial palsy. *JAMA facial plastic surgery*, 20(4), 335-337.
11. Książek, W., Abdar, M., Acharya, U. R., & Pławiak, P. (2019). A novel machine learning approach for early detection of hepatocellular carcinoma patients. *Cognitive Systems Research*, 54, 116-127.
12. Islam, M. M., Nasrin, T., Walther, B. A., Wu, C. C., Yang, H. C., & Li, Y. C. (2019). Prediction of sepsis patients using machine learning approach: a meta-analysis. *Computer methods and programs in biomedicine*, 170, 1-9.
13. Stiglic, G., Kocbek, P., Fijacko, N., Zitnik, M., Verbert, K., & Cilar, L. (2020). Interpretability of machine learning-based prediction models in healthcare. *Wiley Interdisciplinary Reviews: Data Mining and Knowledge Discovery*, 10(5), e1379
14. Vellido, A. (2020). The importance of interpretability and visualization in machine learning for applications in medicine and health care. *Neural computing and applications*, 32(24), 18069-18083.
15. Ghassemi, M., Naumann, T., Schulam, P., Beam, A. L., Chen, I. Y., & Ranganath, R. (2020). A review of challenges and opportunities in machine learning for health. *AMIA Summits on Translational Science Proceedings, 2020*, 191.
16. Ahmed, Z., Mohamed, K., Zeeshan, S., & Dong, X. (2020). Artificial intelligence with multi-functional machine learning platform development for better healthcare and precision medicine. *Database*, 2020.
17. Latchoumi, T. P., Reddy, M. S., & Balamurugan, K. (2020). Applied machine learning predictive analytics to SQL injection attack detection and prevention. *European Journal of Molecular & Clinical Medicine*, 7(02), 2020.
18. Spasic, I., & Nenadic, G. (2020). Clinical text data in machine learning: a systematic review. *JMIR medical informatics*, 8(3), e17984.
19. Daghistani, T. A., Elshawi, R., Sakr, S., Ahmed, A. M., Al-Thwayee, A., & Al-Mallah, M. H. (2019). Predictors of in-hospital length of stay among cardiac patients: a machine learning approach. *International journal of cardiology*, 288, 140-147.
20. Roth, J. A., Battegay, M., Juchler, F., Vogt, J. E., & Widmer, A. F. (2018). Introduction to machine learning in digital healthcare epidemiology. *Infection Control & Hospital Epidemiology*, 39(12), 1457-1462.
21. Jones, L. D., Golan, D., Hanna, S. A., & Ramachandran, M. (2018). Artificial intelligence, machine learning and the evolution of healthcare: A bright future or cause for concern?. *Bone & joint research*, 7(3), 223-225.

CITATION OF THIS ARTICLE

Rajeev Kapoor, Anitha.T, K. Karthikeyan, Aleem Ali, Machine Learning to provide automated, deeper understanding of health care data to improve patient outcomes. *Bull. Env. Pharmacol. Life Sci.*, Vol 11[5] April 2022: 38-43