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Machine learning used in detection & diagnosis of human diseases

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Abstract--One of the most important topics in society is human health care. It looks to get the best diagnosis of a serious illness so they can get the care they need right away it is possible. Other fields, such as mathematics and computer science, are required to part of the search health as this awareness is often difficult. Function of following new trends is a challenge for these sectors, moving forward common. The actual number of new strategies makes it possible give a broad view that avoids certain features. In this case, we suggest a systematic analysis of human diseases related to machine learning. This paper focuses on existing strategies related to the growth of machine learning in which it is used diagnostics in the medical field to find interesting trends, make unimportant predictions, and assist in decision- making. People often feel it hesitation to go to the hospital or doctor or minor symptoms. However, in many places in some cases, these minor symptoms may cause serious health risks. Like online life advice is easily accessible, it can be a good start for users. Moreover, it provides online health care systems suffer from a lack of trust and accuracy. This system analyzes the symptoms given by the user as input and detects the disease and provides the diagnosis as per the symptoms. Predicting is done by using the Naive Baye's Algorithm. This paper attempts to analyze machine learning algorithms used in health care requests to build adequate decision support. To this end, we propose a methodical audit managing the Machine Learning

applied to the finding of human sicknesses. This survey centers around current strategies related to the advancement of Machine Learning applied to conclusion of human sicknesses in the clinical field, to find fascinating examples, making non-trifling expectations and helpful in navigation. Thusly, this work can assist analysts with finding and, if essential, decide the pertinence of the AI strategies in their specific strengths. Staff understands the extent to which AI improves diagnostics and how health care benefits from it. However, there are a few issues that need to be addressed before the successful use of AI in the diagnosis of diseases can be achieved. It can increase productivity as well as efficiency of care delivery and allowing health care systems to provide better and better caring for more people. AI can help improve health care knowledge doctors, which enables them to spend more time in direct patient care and reduction fatigue.

Keywords---human diseases, health care, prediction, machine learning, symptoms.

Introduction

In human society, health care is one of the most pressing issues, such as quality of life of people depends on it. The area of health care, however, is very large diverse, highly dispersed, and distinct. Adequate patient care delivery from a clinical point of view it requires access to appropriate patient information, it is not readily available if needed. Besides, the big difference in the sequence of tests for diagnostic purposes it indicates the need for adequate and appropriate collection tests. Extend this claim by suggesting that the greatest difference is found in the application of general pathology comes primarily to the individual differences in clinical practice are therefore likely to improve further making clear and informed decisions for doctors [1]. Therefore, medical data also contains many different variables found in the variety sources, such as demographics, medical history, medications, allergies, biomarkers, medical images, or genetic markers, each provide a different component a view of the patient's condition. And, among the sources, as mentioned earlier, mathematical structures are fundamentally different.

Researchers and staff are facing two challenges when analyzing such data: Curse of size (number of size and number of samples increasing steadily in the space of features) and the diversity of working resources and mathematical features. These factors contribute to delays and inaccuracies in the diagnosis therefore, patients were not able to receive appropriate care. So, there it is a strong requirement for a proper and orderly approach that makes it timely diagnosis and can be used as a medical decision- making tool. Therefore, the medical, computer, and mathematical fields face this challenge exploring new techniques to illustrate disease prediction and diagnosis, e.g. standard paradigms strive to respond to all of this information. Today, ML provides many useful resources for intelligent data analysis. Moreover, its technology is currently well adapted to study medical data. In particular, a variety of medical diagnostic services have been performed for specialist's diagnostic problems, where the first

ML applications were found [2]. ML Categories have been used successfully, for example, to distinguish between stability patients and those with Parkinson's disease, which is an important tool in clinics diagnosis. Indeed, in many important problems, many ML algorithms did very well.

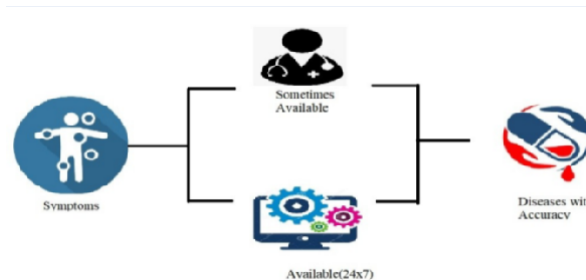


Fig 1. Proposed framework for detection of diseases

The specialist may not be accessible consistently when required. In any case, in the advanced time situation, as indicated by need one can generally utilize this expectation framework whenever. The manifestations of a person along with the age and sexual orientation can be given to the ML model to additional cycle. After starter handling of the information, the ML model uses the ebb and flow information, trains and tests the calculation bringing about the anticipated sickness [3]. As how much clinical information being carefully gathered and put away is immense and extending rapidly, the study of information the executives and examination is additionally progressing to change over this huge asset into data and information that assists them with accomplishing their targets.



The dataset comprising of orientation, side effects, and age of a person was preprocessed and taken care of as a contribution to various ML calculations for the expectation of the infection. The different ML models utilized were Fine, Medium and Coarse Decision trees, Gaussian Naive Bayes, Kernel Naive Bayes, Fine, Medium and Coarse KNN, Weighted KNN, Subspace KNN, and RUSBoosted trees. The result of the models is the sickness according to the side effects, age, and orientation is given to the handling mode [4]. Today, ML gives a few key instruments to astute information investigation. Moreover, its innovation is right now appropriate for dissecting clinical information and, specifically, there is a wide scope of works done in clinical determination in little concentrated analytic issues, where beginning utilizations of ML were brought up. For instance, ML classifiers have been effectively applied to recognize well being and Parkinson patients, which is a helpful instrument in clinical determination. Without a doubt, most ML calculations function admirably on a wide assortment of significant issues. Be that as it may, they have not prevailed with regards to tackling the principle issues in AI when they become incredibly troublesome also, the number in the information is high (the scourge of dimension [5]). In these cases, Big data (BD) innovation is fundamental. In this way, Deep Learning (DL) emerged as a particular sort of ML. Henceforth, the turn of events of DL was inspired and intended to defeat the disappointment of conventional calculations

to work with high-layered information, and learn complex capacities in high-layered spaces.

Literature review

Various examination works have been done for the forecast of the infections in view of the side effects shown by singular utilizing ML algorithm. Measurable model was intended to anticipate regardless of whether a patient had flu. They included 3744 unvaccinated grown-ups and young adult patients of flu who had fever and undoubtedly 2 different side effects of flu. Out of 3744, 2470 were affirmed to have flu by the research facility.

Table 1
Comparison of the methodologies

| Model Used | Accuracy |
|---|----------|
| KNN | 99.6 |
| Random Forest | 91.23 |
| SVM | 80.6 |
| HRFLM | 87.31 |
| Simple Cart, Naive Bayes, SVM and Random Forest | 78.46 |
| Coarse, Fine, Medium and Weighted KNN; SubSpace KNN; Gaussian Navie Bayes; Kernel Naïve Bayes; RUSBoost algorithm; Coarse, Medium and Fine Decision Tree | 93.7 |

In light of this information, their model gave a precision of 80 %. The arbitrary woodland AI calculation is utilized to anticipate the infection in view of the side effects. The framework brought about low time utilization and negligible expense for the expectation of infections. The calculation brought about an exactness of 83.7 %. Different apparatuses were created by to recognize Alzheimer's sickness. Information for 29 grown-ups was utilized for the preparation reason for the ML calculation. They had created grouping models to recognize solid outright changes in the scores with the assistance of SmoteBOOST and wRACOG calculations [6]. A assortment of ML methods like artificial neural network (ANNs), bayesian organizations (BNs), support vector machines (SVMs) and decision trees (DTs) have been broadly applied in disease research for the advancement of prescient models, coming about in viable and exact choice making. A heart problem prediction framework that utilizes the neural network backpropagation algorithm. 13 clinical elements were utilized as contribution for the ANN and afterward the neural network was prepared with the backpropagation calculation to foresee nonattendance or presence of coronary illness with an exactness of 94 %. Different AI calculations were smoothed out for the compelling expectation of a constant sickness flare-up. The information gathered for it was deficient to prepare reason. To beat this, an idle variable model was utilized. A new

convolutional neural network-based multimodal diseases risk prediction (CNN-MDRP) was organized. The calculation arrived at a precision of around 94.8 %.4 different profound learning models were utilized to be specific profound deep neural network (DNN), long short term memory (LSTM), ordinary least squares (OLS), an autoregressive integrated moving average (ARIMA) for observing 80 irresistible illnesses in 6 gatherings. Of the relative multitude of models utilized, DNN and LSTM models had a superior exhibition. The DNN model performed better as far as normal execution and the LSTM model gave close expectations when events were enormous. A data set was utilized that contained data about patients having any heart problems [7]. They removed highlights utilizing three choice calculations which are help, minimum redundancy, and maximum relevance (mRMR), furthermore, least outright shrinkage and determination administrator which was cross-checked by the K-fold technique. The separated highlights were shipped off 6 different MI algorithms and afterward it was characterized in view of the presence or absence of heart problems. A compelling heart illness forecast framework was created They accomplished an exactness level of 87.2 % through the forecast model for heart problems with the hybrid random forest with a linear model (HRFLM).Logistic regression (LR) was utilized to distinguish the gamble factors for diabetes infection. The in general precision of the ML-based framework was 91.53 %.

Methodology

From an open-source dataset, a sheet was made where we recorded down every one of the side effects for the separate sicknesses. After which relying upon the illness, age and orientation were determined as a piece of the dataset. We recorded down around 230 diseases with additional than 1000 one of a kind side effects altogether. The side effects, age and sex of an individual were utilized as contribution to different ML algorithms [8].

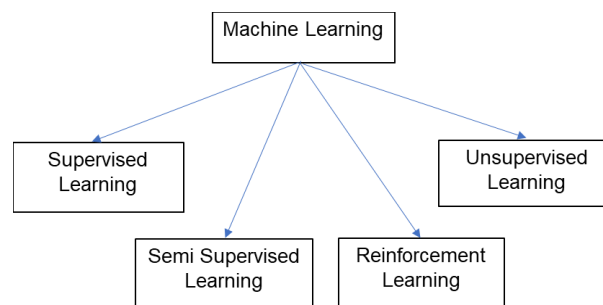


Figure 1. Different kinds of ML

Various techniques used by ML

Numerous researchers have fostered all of the different ML algorithms to analyze sicknesses. The scientist expresses that ML works productively to investigations different sicknesses. In the field of medication, there are a few degrees of ML. Protein- protein coordinated effort involved the observational space of modified learning in certain capacities such as clinical imaging, remedial information recovery, supportive decision help, and general patient organization. ML is utilized

to separate and assess pneumonia, threatening development of the lungs, and various diseases.

K-nearest neighbours (KNN)

The K-Nearest Neighbours (KNN) algorithm utilized as a sort of supervised ML algorithm. It basically determined the distance of one new location to all other locations. The distance can be of Euclidean or Manhattan type. After this, it chooses the K-nearest items, where K can be any whole number. Finally, it does out the location to the class to which most of K information focuses have a place.

Fine, Medium, and Coarse KNN

We have to relegate the whole number upsides of K to see as the distance. In this way, in our fine KNN model, we relegated a low worth of K which implies it roughly utilizes just one neighbor for the forecast. Essentially, the medium KNN model uses roughly 10 neighbors and the coarse KNN utilizes 100 neighbors. Since the neighbors for each, the model varies the precision rates likewise shifted with a wide reach. Among every one of the three models our fine KNN gave us an extremely high exactness while the coarse KNN brought about low forecast esteem.

Weighted KNN

It is a modified form of KNN. In KNN we picked an whole number boundary K and by utilizing that boundary we tracked down where the major anticipated qualities lied. In any case, if the worth of K is too little the calculation is considerably more delicate to the focuses that are exceptions. Likewise, if the worth of K is too enormous then every one of the focuses that are nearly exceptionally near the K worth is chosen. To defeat this issue the weighted KNN gave more weight to the focuses that were closest to the K worth and the less weight to the focuses that were farther away. We had the option to get the most elevated exactness utilizing this model. Additionally among all the KNN models, this model gave us the best outcomes

Naïve Bayes

It is AI estimation for portrayal issues and relies upon Bayes' probability speculation. The fundamental use of this is to do message portrayal which incorporates high layered getting ready enlightening assortments.

Gaussian Naïve Bayes

It follows a similar system with respect to Naive Bayes. Be that as it may, at Naive Bayes we want stage and information base data sets Gaussian Naive Bayes need a comprehensive data set constant highlight. Our arrangement of information included nonstop highlights of elements, age, and orientation so utilizing this model was compulsory. The precision of utilizing this model was not the most elevated esteem.

Kernel Naive Bayes

Our dataset had some mathematical properties like age so we inferred Kernel Naive Bayes to foresee the drugs. The means followed for this calculation are like the Naive Bayes. The significant advantage of utilizing this calculation is that it gives assessors that have a non parametric nature. Assuming there is no earlier information that the dataset utilized is parametric or not this model can give more exact outcomes. The outcomes given by this model were practically as old as by the Gaussian Naive Bayes.

Decision trees

Decision trees calculation has a place with the group of regulated learning calculations. It is utilized for relapse and characterization. In the choice tree, for forecast, it utilizes the strategy for tree graph at the top. It contains a root hub after which it gets parted in the prevailing info element and afterward it again gets parted. These cycles proceed till all information is set and at the hub, the super last hub contains .The loads on the foundations of these loads it characterizes the info. In a coarse tree, the greatest number of parts from every hub is 4. While in a Medium tree, the greatest number of parts from every node is 20. In a fine tree, the most extreme number of parts from every node is 100.

Results

Different AI models were utilized to analyze the expectation of illness for accessible information dataset. We utilized 11 different ML models for the expectation. Out of the 11 models we figured out how to get 50 % or on the other hand above exactness for 6 models. As displayed, among every one of the models, we acquired the most noteworthy precision for the Weighted KNN model of 94.2 %. The precision is high in light of the fact that the weighted KNN was high since in this model the worth of K changed [9]. This worth changed as indicated by our dataset for example it was little and enormous for the preparation set. Because of this variety, it ended up being the most dependable model when contrasted with the other ML calculations. We took crude data and recognized them based on orientation, age gathering, and side effects.

We have contrasted our approach and different philosophies revealed in the writing. A portion of the writing has utilized the SVM and the KNN model for the forecast of the infections. We have utilized 11 different ML models for the forecast of around 230 illnesses. We accomplished the most noteworthy exactness of 93.7 % which was high when contrasted with most of different approaches revealed. The most elevated precision was accomplished in light of the Weighted KNN model. The most noteworthy exactness of 99.6 % was accomplished utilizing the KNN model for the forecast and grouping of heart illnesses. The Random Forest model was utilized and accomplished a precision of 91.23 %. The SVM model demonstrated to have a higher exactness of 80.6 % when contrasted with different techniques [10]. This Section gives a synopsis of the couple of studies assessed in this paper necessities for getting the smart information examination acted in the clinical field.

Machine learning principle

Each learning cycle, profound or not, comprises of two stages: the assessment of obscure conditions in a framework from a given informational collection (input) and the utilization of assessed conditions to foresee new results of the framework. In this Subsection, we dissect the most well-known strategies utilized in the two stages.

Input definition and methods

The contribution of a ML cycle is a bunch of occurrences. These cases are the things that can be grouped, related, or grouped. Each occasion is an individual, i.e., autonomous illustration of the idea that should be learned. Furthermore, everyone is described by the upsides of a bunch of foreordained credits. Each dataset is addressed as a network of occurrences versus ascribes, which, in information base terms, is a level document (single connection) characterized the really normal learning techniques as unsupervised learning and supervised learning. In unsupervised learning, non-marked models are given and there is no thought of the result along the learning system. The point in this sort of learning is to investigate information before the finish of finding various classes or bunches, which permit us to sort out them. Delegate unsupervised learning grouping assignments are K-means Clustering, Density-Based Spatial Clustering of Applications with Noise (DBSCAN), Self-Organized Maps (SOMS), Similarity Network Fusion (SNF), Perturbation Clustering for Data Integration and Disease Subtyping (PINS), and Cancer Integration by means of Multikernel Learning (CIMLR) algorithms.

In supervised learning, a marked arrangement of preparing information is utilized with to gauge or guide the ideal yield. In any case, this reality can be cured through Active Learning (AL), which advances steadily by beginning with a couple of models and afterward asking in every emphasis the clinical master to just name the case that the calculation decides as the most useful. The AL methods have been effectively utilized in the clinical field. Tests had shown that this can diminish the quantity of named cases expected to each maximal exactness by 30-40% contrasted with standard techniques that begin with the completely named informational index. The most well-known calculations in managed learning are Support Vector Machine (SVM), Iterative Dichotomiser 3 (ID3), K-Nearest-Neighbour (KNN), Naïve Bayes, Bayesian Networks, straight relapse, and calculated relapse. SVM calculations utilize linear models to carry out nonlinear class limits. The ID3 calculation was designed by Quinlan is the forerunner of calculation, like C4.5 and J4.8 [11]. These calculations are utilized to produce choice trees from a dataset. Whether or not the learning strategy is unsupervised or supervised, the system is consistently the equivalent.

Output learning representation

There are a wide range of approaches to addressing the examples in information. Every one directs the sort of procedure that can be utilized to construe the result structure from information (in regulated learning) or then again basically addresses various bunches of information (unaided). The most

widely recognized portrayals are as per the following.

- Decision tables are one kind of data tables with a choice characteristic giving the decision classes for all articles. Table shows a basic illustration of a choice table, where the last segment addresses an ultimate conclusion. In decision tables, the approach to deducing the result is to make equivalent to the info. Along these lines, choice tables are one of the least complex approaches to addressing the result from a ML grouping model.
- Decision trees (DT) are prescient portrayals that can be utilized both for order also, relapse models. These are a levelled approach to parceling the space, where the objective is to make a model that predicts the worth of an objective variable in light of a few info factors. A DT learns by parting the source set into subsets that depend on a characteristic esteem test. This cycle is rehashed on each determined subset in a recursive manner, called recursive parceling. At the point when a DT is utilized for grouping errands, it is all the more suitably alluded to as a characterization tree.

Table 2
A Decisiion Table

| Patient | Fever | Head Ache | Flu | Muscles Ache |
|---------|-------|-----------|-----|--------------|
| A1 | Yes | No | Yes | Yes |
| A2 | No | Yes | Yes | No |
| A3 | Yes | No | Yes | Yes |
| A4 | No | Yes | No | Yes |
| A4 | No | Yes | Yes | No |

In design acknowledgment one attempt to observe an answer that prevails with regards to separating between focuses of various classes or planning the information into another variable. Furthermore, it should have the capacity to sum up the learning model to other new and already inconspicuous information sources [12]. To this end, the model should be prepared. Since unsupervised learning just tries to find putting together information, it doesn't need earlier preparation. Then, the training and the ensuing system of testing are point by point.

Discussion

In this paper, various instruments and approaches that are generally utilized in the clinical and medical services fields are portrayed. These instruments are inside AI and permit us to arrive at the fundamental point of ML, finding valuable examples in data sets, which help to make sense of and make non-paltry forecast about information. That's what we see, as a general rule; the most well-known task acted in the clinical field is characterization in all application regions exemplified. Notwithstanding, another normal assignment is relapse in the irresistible illness region. This undertaking is minimal utilized in regions, like Alzheimer's infection, Parkinson's infection, and hepatic sicknesses. Then again, the it is additionally slight considered to group task in hepatic and heart

infections, notwithstanding, it is many times utilized in Alzheimer's sickness and Parkinson's sickness [13]. At long last, the blend of brain organizations and other directed calculations is generally utilized in malignant growth, Alzheimer's infection, Parkinson's sickness, and renal illness application regions, however is only here and there applied in metabolic, hepatic, irresistible, and coronary illness application regions.

Machine learning uses in clinics

ML is an important apparatus for clinical experts in the counteraction, conclusion, furthermore, treatment of human illnesses. Notwithstanding, right now there are couple of instances of the effective use of these strategies in the specific area of clinical practice, in spite of the way that there are different ML methods produce great outcomes. It is said that a ML algorithms took in another undertaking, rather than saying that it basically extricated a bunch of factual examples from a bunch of preparing information, where these information are physically chosen and marked under the immediate management of somebody who picked which algorithms, boundaries, and work processes would be utilized for its turn of events. It is moreover said that a neural network accurately recognizes, for instance, pictures of canines and felines, by learning the qualities of those creatures, when it basically relates explicit gatherings of varieties and surfaces in the photos. Subsequently, assuming a picture veers off excessively far from the models that the brain network has seen, the expectation will fizzle, with unfortunate results assuming we address the location of malignant growth or a neurodegenerative infection. ML algorithms, through their portrayal, assessment, and improvement parts, benefit from the accessibility of a lot of information and strong equipment designs to address more intricate measurable peculiarities than conventional methodologies, while DL takes into consideration recognizing recently covered up designs, extrapolate drifts, and anticipate brings about a wide range of issues, attempting to "learn" a way to deal with some capacity [14].

ML strategies are as of now applied to clinical records in clinical practice to foresee, for instance, which patients are at most serious gamble of readmission to emergency clinic or who are probably not going to follow endorsed medicines. The applications are limitless in analysis, research, drug improvement furthermore, clinical preliminaries. Notwithstanding the enormous measure of digitized information, prescient models that are worked from clinical records are essentially founded on customary direct models and seldom think about more than 20 or 30 factors. Nonetheless, a critical benefit of ML is that scientists don't have to determine which potential prescient factors to consider and in which blends. A significant issue to think about while applying ML in the clinical practice is the consistency of information from heterogeneous sources. Every wellbeing framework might gather patient information diversely for comparative purposes. Hence, prior to applying ML, it is important to standardised the information. This would keep away from information over fitting and the trouble of applying similar method to different informational collections. The issue of inclination is additionally significant. This issue emerges when there is unfortunate inclusion of preparing information, prompting blunders while applying to minority gatherings. By and large, in clinical practice it is intriguing

to have unique furthermore, huge information sources to feature the particular highlights of each gathering of patients. At long last, the conceivability of the calculations is another key component [15]. A split the difference between execution and the interpretability must belaid out. Models with better execution (e.g., DL) are frequently challenging to make sense of, while models, for example, straight relapse or choice trees are more justifiable.

Conclusion

Keen information handling is a social need for distinguishing, straightaway, of helpful and strong diseases detection to give patients fitting consideration inside the briefest conceivable time. This recognition has been completed in ongoing a very long time by recognizing invigorating designs in information bases. Savvy information handling is arising as a necessity for viable and powerful illnesses to be found by society. Detection of patients giving the fundamental treatment straightaway inside the most limited conceivable period. This identification has been accomplished in ongoing a very long time through the technique for recognizing invigorating examples in information bases. A complete outline of shrewd information investigation apparatuses in the clinical area is surrendered in this paper. A few instances of certain algorithms utilized in these clinical field regions are moreover introduced, inspecting potential examples in view of the objective searched, the system utilized and the application field. Furthermore, we detail the benefits and weaknesses of every procedure depicted to help in a future foundation about which the strategy is generally reasonable for every genuine circumstance tended to by different creators. At last, Figure shows the connections between all procedures as well as all supervised and unsupervised learning point by point in this paper.

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