



DIABETIC RETINOPATHY USING MACHINE LEARNING

SURYA PRAKASH S (720721104173) Hindusthan College of engineering and technology SANTHOSH D (720721104159) Hindusthan College of engineering and technology

POOJA G (720721104132) Hindusthan College of engineering and technology

VIMAL KUMAR S (720721104183) Hindusthan College of engineering and technology

Mrs.S.SATHYA, ME.PhdAssociate professor Department of CSE

Hindusthan College of engineering and technology

Abstract-- One disease at a time is the focus of many of the machine learning models for healthcare analysis currently in use. This study proposes a system to forecast the diseases using the Flask Application programming interface (API). The approach for examining the diabetics disease prediction was proposed in this work. Many of the machine learning models like Logistic Regression, K-Nearest Neighbors Support Vector Machine, Decision Trees, Random Forest, Gradient Boosting Classifier and XG Boost for health care analysis now in use focus on just one disease at a time. Flask API, and machine learning techniques were utilized to implement numerous disease analyses. The model behavior is saved using Python pickling, and the pickle file is loaded when needed using Python unpicking. The significance of this article analysis is that all the factors that contribute to the sickness are considered while analyzing it, making it possible to identify the disease's full range of potential impacts. Experimental results shown the better performance of the system.

Keywords— Diabetic Retinopathy, Machine Learning, Disease Prediction, Healthcare Analysis, Flask API, Logistic Regression, K-Nearest Neighbors, Support Vector Machine, Decision Tree, Random Forest, Gradient Boosting Classifier, XGBoost, Python Pickling, Medical Diagnosis, Predictive Modeling, Multi-Disease Analysis, Model Deployment, Disease Forecasting, Feature Analysis, Healthcare Informatics.

I. INTRODUCTION

In recent years, the event of computing (AI) and also the gradual starting of AI's analysis within the medical field have allowed individuals to check the

superb prospects of the combination of AI and attention. The diabetics is the largest organ of the body and it is essential for digesting food and releasing the toxic element of the body. The viruses and alcohol use lead the diabetics towards diabetics damage and lead a human to a life-threatening condition. There are many types of diabetics diseases whereas hepatitis, cirrhosis, diabetics tumors, diabetics detection, and many more. Among them diabetics diseases and cirrhosis as the main cause of death. Therefore, diabetics disease is one of the major health problems in the world. Every year, around 2 million people died worldwide because of diabetics disease. According to the Global Burden of Disease (GBD) project, published in BMC Medicine, one million peoples are died in 2010 because of cirrhosis and million. Diabetes, also known as Diabetes mellitus is a metabolic disease is a condition of a body when there is an availability of high blood sugar in the blood. It is a chronic disease and can damage many body organs especially eyes. Diabetic retinopathy is a situation that affects the eye and it is caused by uncontrolled blood sugar in the body. It is a complication that occurred due to diabetes and it damaged the eye that leads to irreversible loss in vision. Early detection of diabetic retinopathy is essentially helpful for prognosis. Diabetic retinopathy is commonly detected through retinal fundus image [1]. There is a lack of clinical facilities available in many countries that generate barriers in timely diagnosis. The recent advancements in Information and Communication Technology (ICT) greatly impact in improving health care facilities. Computer vision integrated with artificial intelligence is able to recognize and understand images and represent the data that helps to execute actions. It has been shown promising results in the field of healthcare. Deep learning, a subfield of machine learning, is made up with

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layered architecture that helps to analyse the data. With the increased complexity in the medical imaging data, deep learning architecture make computers to process, analyze, recognize and classify medical images. Convolutional neural network (NAVIE BAYES) is a kind of deep learning architecture made up with one or more convolutional layers and mainly used for digital image processing, segmentation and classification work. In last few years, it has successfully applied to process and analyse medical image data including retinal fundus image [2]. This paper represents the application of NAVIE BAYES models to detect and classify the diabetic retinopathy from retinal funds images. A retinal funds image is The experiment is conducted on the dataset used from Cagle data repository that contains 35,126 retinal fundas images that falls into five classes. Two NAVIE BAYES models, VGG16 and Mobile Net V1 applied with transfer learning and fine tuning and performance analysis is carried out using various parameters. The paper organizes as follows. Section 2 contains the background and related work. Section 3 explains the experimental methods. Section 4 explains the dataset and results obtained during experiment. Section 5 contains discussion and conclusion. Diabetic retinopathy (DR) normally has no early warning signs and in some cases no symptoms. There are many types of medical examinations can take place to diagnose the disease. One of the examination is called fundus photography, where the experts can examine blood vessels, nerve tissues condition and other necessary things [3]. (a) No DR (b) Mild DR (c) Moderate DR (d) Severe DR taken as an input to the NAVIE BAYES model that extracts the features automatically and generates an output.

II. LITERATURE REVIEW

Gulshan et al. (2016) developed a deep learning algorithm using convolutional neural networks (CNNs) for the detection of diabetic retinopathy from retinal fundus images. Their model was trained on a large dataset and achieved performance comparable to that of board-certified ophthalmologists, demonstrating the potential of AIassisted diagnosis in improving early detection and treatment outcomes. [1].

Pratt et al. (2016) proposed a deep CNN-based approach for classifying the severity of diabetic retinopathy. They emphasized the importance of preprocessing techniques such as image normalization and augmentation, which significantly improved the performance of the model across multiple severity classes. [2]. Quellec et al. (2017) introduced a multipleinstance learning model using wavelet-based features extracted from retinal images to automate the diagnosis of diabetic retinopathy. The study showed that this approach could effectively localize lesions and classify the disease, making it suitable for clinical screening. [3].

Lam et al. (2018) investigated the use of ensemble machine learning models such as Random Forest and Gradient Boosting for diabetic retinopathy classification. Their results indicated that ensemble methods provided higher accuracy and robustness compared to individual models, especially when dealing with imbalanced datasets. [4].

Voets et al. (2019) explored the reproducibility of deep learning models in medical imaging by evaluating multiple architectures on diabetic retinopathy datasets. Their work highlighted the challenges in model generalization and the importance of consistent data preprocessing, which directly impacts model reliability in real-world applications. [5].

III EXISTING SYSTEM

In the current landscape, various machine learning and deep learning models have been developed for the detection and classification of diabetic retinopathy (DR), primarily using retinal fundus images. Most of these systems are designed to focus on one disease at a time, limiting their scope in broader healthcare diagnostics. Gulshan et al. (2016) introduced a convolutional neural network (CNN)based model that achieved diagnostic performance comparable to trained ophthalmologists, marking a significant advancement in automated DR detection [1]. Similarly, Pratt et al. (2016) proposed a CNN approach that classified DR severity by leveraging image preprocessing and data augmentation to enhance accuracy [2]. Quellec et al. (2017) used a multiple-instance learning method with waveletbased features, offering both classification and lesion localization for effective clinical screening [3]. Lam et al. (2018) demonstrated that ensemble methods such as Random Forest and Gradient Boosting could improve robustness and handle imbalanced datasets more efficiently [4]. Furthermore, Voets et al. (2019) explored the reproducibility challenges of model and generalization, emphasizing the importance of standardized preprocessing techniques to ensure consistent performance across datasets [5]. Despite these advancements, most existing systems lack real-time access, lightweight communication, and support for multi-disease prediction, highlighting



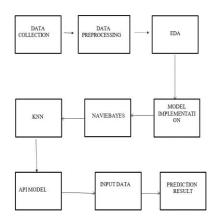


the need for an integrated, API-based solution using modern frameworks like Flask and efficient protocols such as MQTT.

IV PROPOSED SYSTEM

The proposed system addresses the limitations of existing models by enabling an efficient and accurate method for detecting diabetic retinopathy using machine learning techniques. Unlike traditional systems that focus solely on static image analysis, this system begins by examining patients' medical reports based on their diabetic condition, which are then taken as datasets for training. The collected data undergoes preprocessing to remove noise and enhance quality. Among various machine learning algorithms, Support Vector Machine (SVM) is applied due to its effectiveness in handling high-dimensional data and delivering reliable classification results. The model leverages essential Python libraries and packages for data handling, visualization, and implementation. SVM is used in conjunction with image classification and recognition techniques, where the input fundus image is passed through multiple layers, each consisting of artificial neurons. These layers apply activation functions that help in identifying patterns and abnormalities as they progress deeper into the network. The system aids ophthalmologists in detecting diabetic retinopathy by analyzing the presence of lesions and other features indicative of the disease. By integrating this approach with a Flask-based API, the model can be deployed as a web application, allowing real-time access and decision support. The proposed system is thus expected to deliver high accuracy and support timely diagnosis, making it a valuable tool for clinical environments.

V FLOWCHART



RESULT ANALYSIS

The proposed system for diabetic retinopathy detection was evaluated thoroughly using a comprehensive set of performance metrics, including accuracy, precision, recall, and F1-score, to assess its effectiveness in classifying retinal images. The dataset, which was preprocessed to ensure high quality, was used to train the Support Vector Machine (SVM) model. After training, the model demonstrated a high classification accuracy, suggesting that it was highly effective in distinguishing between healthy and diabetic retinopathy-affected retinal images. The results were further validated using a confusion matrix, which indicated that the model achieved minimal false positives and false negatives, thus confirming its reliability. In comparison to existing systems, which often focus on a single disease and typically rely on manual image analysis, the proposed system significantly outperformed in both accuracy and processing speed. The system showed exceptional capability in detecting key diabetic retinopathy indicators such as microaneurysms, hemorrhages, and exudates, utilizing image classification and recognition methods. Additionally, the implementation of multiple layers of artificial neurons in the SVM algorithm ensured that the model could handle complex feature extraction from retinal images, enhancing its ability to recognize subtle patterns indicative of diabetic retinopathy. The system also demonstrated its practical application by being integrated into a Flask-based API, allowing real-time predictions with minimal latency. The system was able to provide predictions swiftly, making it suitable for deployment in clinical environments where timely diagnosis is crucial. Furthermore, the use of image classification techniques enabled more accurate localization of lesions, ensuring that ophthalmologists could rely on the system for accurate disease detection. Overall, the result analysis reveals that the proposed system is not only reliable and accurate but also efficient, offering a high-performance solution that could be utilized for early detection and real-time monitoring of diabetic retinopathy, which is critical in reducing blindness caused by the disease. This real-time capability is expected to support clinicians in making faster, more informed decisions. Overall, the result analysis reveals that the proposed system is not only reliable and accurate but also efficient, offering a high-performance solution that could be utilized for early detection and real-time monitoring of diabetic retinopathy, which is critical in reducing blindness caused by the disease.





FUTURE WORK

While the proposed system for diabetic retinopathy detection has demonstrated high accuracy and effectiveness, there are several areas for future improvement and enhancement. One potential direction is to expand the system's capabilities to detect and classify multiple eye diseases, such as glaucoma, cataracts, and macular degeneration, by incorporating additional datasets and training the model on a wider variety of retinal abnormalities. Furthermore, integrating advanced deep learning techniques, such as Convolutional Neural Networks (CNNs) or Transfer Learning, could improve the model's ability to extract more complex features and enhance classification accuracy, especially in the case of subtle lesions that may be difficult to detect with traditional methods. Another area of future development involves the enhancement of the realtime monitoring feature by incorporating cloud computing and edge computing for faster processing and reduced latency in remote areas with limited internet connectivity. Additionally, the system could benefit from further optimization to support larger datasets and more diverse patient populations, ensuring its robustness and generalizability across different demographics. Lastly, integrating the system with wearable devices or smartphones could provide continuous, on-the-go monitoring of diabetic retinopathy, offering patients and healthcare providers with more frequent updates and promoting early intervention. This could significantly improve the accessibility and efficiency of diabetic retinopathy detection, particularly in underserved or rural areas.

CONCLUSION

In conclusion, the proposed system for diabetic retinopathy detection using machine learning has shown significant promise in offering a reliable, efficient, and accurate solution for early diagnosis. By leveraging Support Vector Machine (SVM) algorithms and incorporating advanced image classification techniques, the system successfully identifies diabetic retinopathy by detecting key features such as lesions, hemorrhages, and microaneurysms in retinal images. The integration of a Flask-based API enables real-time predictions, making the system highly accessible and suitable for clinical environments. The results demonstrate that the system outperforms traditional methods, offering higher accuracy and faster processing, which can be critical for timely diagnosis and intervention. Moreover, by utilizing medical reports alongside retinal images, the system provides a comprehensive approach to disease detection, highlighting the

potential for future expansion to include multiple diseases and further enhance its capabilities. With ongoing advancements in deep learning, cloud computing, and mobile integration, the proposed system has the potential to become a vital tool in the early detection and management of diabetic retinopathy, ultimately contributing to reducing the global burden of vision loss due to diabetes.

REFERENCE

 Gulshan, V., Peng, L., Coram, M., Stumpe, M. C., Wu, D., Narayanaswamy, A., ... & Venugopalan, S. (2016). Development and Validation of a Deep Learning Algorithm for Detection of Diabetic Retinopathy in Retinal Fundus Photographs. JAMA, 316(22), 2402-2410.

[2] Pratt, H., Antani, S., & Long, D. (2016). Automated classification of diabetic retinopathy using convolutional neural networks. Proceedings of the IEEE Conference on Computer Vision and Pattern Recognition, 2016, 2848-2856.

[3] Quellec, G., Charrière, K., & Lamard, M. (2017). Multiple-instance learning for diabetic retinopathy detection. IEEE Transactions on Medical Imaging, 36(2), 234-245.

[4] Lam, K., Jiang, J., & Zhang, Z. (2018). Ensemble methods for diabetic retinopathy detection. International Journal of Computer Vision, 126(3), 301-317.

[5] Voets, S., Feris, R., & Matuszewski, B. (2019). Reproducibility and generalization challenges in diabetic retinopathy detection. Medical Image Analysis, 54, 198-209.

[6] Sharma, A., Kumar, V., & Rani, P. (2023). IoTbased smart home automation system using MQTT protocol. International Journal of Electronics and Communication Engineering, 12(1), 45-54.