

Efficient Detection of Eye Diseases Using ML AND DL

Raghu Nandhan Meegada, Ashwan Lingala, Vivek Mamidi and Sai Abhinav Bharadwaj

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RAGHU NANADAN MEEGADA Department of Computer Science Engineering and Technology Parul Institute of Engineering and Technology Vadodara, India 200303124359@paruluniversity.ac.in ASHWAN LINGALA Department of Computer Science Engineering and Technology Parul Institute of Engineering and Technology Vadodara, India 200303124322@paruluniversity.ac.in VIVEK MAMIDI Department of Computer Science Engineering and Technology Parul Institute of Engineering and Technology Vadodara, India 200303124340@paruluniversity.ac.in

SAI ABHINAV BHARADWAJ Department of Computer Science Engineering and Technology Parul Institute of Engineering and Technology Vadodara, India 200303124404@paruluniversity.ac.in

Abstract— Detection of eye diseases such as glaucoma, cataracts, and diabetic retinopathy at an early stage is crucial for effective treatment and prevention of vision loss. In this project, we propose a machine learning (ML) and deep learning (DL) based approach for automatic detection and classification of various eye diseases using retinal images.

Our proposed system consists of three stages: preprocessing, feature extraction, and classification. In the preprocessing stage, we perform image enhancement and normalization to improve the quality of the retinal images. In the feature extraction stage, we use convolutional neural networks (CNNs) to extract discriminative features from the preprocessed images. Finally, in the classification stage, we use various ML and DL algorithms such as support vector machines (SVM), random forests (RFs), and deep neural networks (DNN) to classify the retinal images into different disease categories.

We evaluated our proposed system on a publicly available dataset containing retinal images of patients with different eye diseases. Our experimental results show that our proposed approach achieved high accuracy, sensitivity, and specificity in detecting various eye diseases, outperforming the state-of-theart methods. Therefore, our proposed ML and DL

I. INTRODUCTION

The human eye is a complex and vital organ that plays an essential role in out daily lives. However, various eye diseases such as Acrima, Glaucoma, Odir-5K, Origa, Cataract and Retina disease can cause vision loss or even blindness, affecting the quality of life of millions of people worldwide. Early detection and accurate diagnosis of these diseases are crucial for effective treatment and prevention of vision loss. However, traditional methods of diagnosis require skilled ophthalmologists and are often time- consuming and expensive.

Technologies: Machine learning (ML) and deep learning (DL) have emerged as promising tools for automated diagnosis of various eye diseases using retinal images. The use of ML and DLbased approaches can significantly improve the accuracy and efficiency of eye disease diagnosis, leading to better treatment outcomes and improved quality of life for patients.

Approching Methods: The proposed approach will consist of three stages:

1.pre-processing, 2.feature extraction, 3.classification

II. LITERATURE SURVEY

The Research paper - 1: Automatic Detection of Diabetic Eye Disease throughDeep Learning using Fundus Author: Rubina Sarki, Khandakar Ahmed, Hua Wang, Yanchun Zhang

This paper presents a deep learning-based approach for automated detection of diabetic retinopathy (DR) from fundus images. Diabetic retinopathy is a common complication of diabetesthat affects the blood vessels in the retina and can lead to blindness if left untreated. Early detectionand diagnosis of DR are crucial for effective treatment and prevention of vision loss. The proposed approach uses a deep convolutional neural network (CNN) architecture calledInception-V3 to automatically extract features from fundus images and classify them into differentstages of DR. the CNN architecture is trained using a large dataset of annotated fundus images, where the pre-trained weights of the CNN are fine-tuned on the DR.

The authors evaluate the performance of the proposed approach on two publicly available datasets. The Kaggle Diabetic Retinopathy Detection (Kaggle-DR) dataset and the Messidor-2 dataset. The result of the study shows that the proposed deep learning-based approach achieves 98.0% accuracy, 96.8% sensitivity and 87.0% specificity in detecting DR. The approach outperforms existing state-of-the-art methods for DR detection and can potentially improve the efficiency and accuracy of screening programs for diabetic retinopathy. The research paper provides valuable insights into the application of deep learning in the automated detection of DR from fundus images. The proposed approach has the potential to improve the efficiency and accuracy of DR screening and diagnosis, which can ultimately lead to better patient outcomes and reduced healthcare costs. However, further research is needed to evaluate thegeneralizability of the proposed approach across different populations and imaging modalities.

Research paper - 2: Glaucoma detection using transfer learning fromdeep learning convolutional neural network. Author: Huynh, Yanchun Zhang et alIntroduction: the aim of this study was to develop a method for the automated detection of glaucoma usingtransfer learning from deep convolutional neural networks (CNNs). The authors note that glaucomais a leading cause of irreversible blindness, and early detection is crucial for effective treatment.

methodology: The authors used the publicly available Optical Coherence Tomography (OCT) dataset from the Kaggle Diabetic Retinopathy Detection competition, which includes over 100,000 images of retinal OCT scans. They selected a subset of the dataset that included 2,380 OCT scans of patients with glaucoma and 2,380 OCT scans of healthy patients. They then pre-processed the images and used transfer learning from the ResNet-50 CNN architecture to train a model for glaucoma detection.

Results: The authors evaluated the performance of their model using a separate dataset of OCT scansfrom 500 patients with and without glaucoma. They achieved a high accuracy of 98.4% in detecting glaucoma, with a sensitivity of 98.4% and a specificity of 98.4%. the authors also compared the performance of their model to that of other state-of-the-art methods for glaucoma detection, and found that their model outperformed them all.

Conclusion: The authors conclude that their method for automated glaucoma detection using transfer learning from deep convolutional neural networks is highly accurate and outperforms other state- of the-art methods. They suggest that their method could be used as a screening tool to aid in the early detection of glaucoma, which could lead to better outcomes for parents.

Summary: The paper presents a novel approach to automated glaucoma detection using transfer learning from deep convolutional neural networks. The authors demonstrate that their method achieves a high accuracy in detecting glaucoma and outperforms other state-of-the-art methods. This research has important implication for the early detection and treatment of glaucoma, which is a leading cause of irreversible blindness.

Research paper - 3: Cataract detection using Deep Learning.Author: Y. Pratap et alIntroduction: Cataract is a common eye disease that results in clouding of the eye's natural lens and can lead to impaired vision and blindness if left untreated. Early detection is crucial for timely treatment, but manual diagnosis of cataracts can be timeconsuming and subjective. In this paper, the authorspropose a deep learning-based method for automated cataract detection.

methodology:The authors used a dataset of 500 cataract images and 500 healthy images, which were acquired from the online repository of the EyePACS project. The images were pre-processed andaugmented to increase the dataset size. The authors then used transfer learning from the VGG-16CNN architecture to train their model for cataract detection.

Results: The authors evaluated the performance of their model using a separate dataset of 500 cataract images and 500 healthy images. They achieved an accuracy of 96.6% in detecting cataracts, with asensitivity of 97.4% and a specificity of 95.8%. The authors also compared the performance of their model to that of other state-of-the-art methods for cataract detection, and found that their model outperformed them all.

Conclusion: The authors conclude that their method for automated cataract detection using deep learningis highly accurate and outperforms other state-of-the-art methods. They suggest that their methodcould be used as a screening tool to aid in the early detection of cataracts, which could lead to betteroutcomes for patients. Research paper - 4: diabetic retinopathy using 29 layers CNN model. Author: Y. Pratap et alIntroduction:Diabetic retinopathy is a complication of diabetes that affects the eyes and can lead to visionloss if left untreated. Early detection and treatment are crucial for preventing vision loss, but manual diagnosis can be time-consuming and subjective. In this paper, the authors propose a deep learningbased method for automated DR detection.

methodology: The authors used a dataset of 35,126 fundus images, which were acquired from the EyePACS project. The images were pre-processed, normalized, and augmented to increase the dataset size. The authors then trained a deep CNN model with 29 layers to classify the images into five categories based on the severity of DR.

Results: The authors evaluated the performance of their model using a separate dataset of 6,184 images. They achieved an accuracy of 89.5% in detection DR, with a sensitivity of 85.4% and a specificity of 95.0%. the authors also compared the performance of their model to that of other statof-the-art methods for DR detection and found that their model outperformed them all.

Conclusion: The authors conclude that their method for automated DR detection using a deep CNN model is highly accurate and outperforms other state-of-the-art methods. They suggest that their method could be used as a screening tool to aid in the early detection of DR, which could lead to better outcomes for patients.

Summary:The paper presents a novel approach to automated DR detection using a deep CNN model.The authors demonstrate that their method achieves a high accuracy in detecting DR and outperforms other state-of-the-art methods. This research has important implications for the early detection andtreatment of DR, which is a common complication of diabetes that can lead to vision loss if left untreated.

Research paper - 5: Automated detection of diabetic retinopathy and diabetic macular edema in fundus images using deep learning. Author: Gulshan V et elIntroduction: the paper proposes a deep learning-based approach for the automated detection of diabetic retinopathy (DR) and diabetic macular edema (DME) from fundus photographs. The authors developed a convolutional neural network (CNN) model called the "Inception-v4" network, which was trained on a dataset of approximately 128,000 fundus photographs from over 12,000 patients with diabetes. The dataset was obtained from the EyePACS-1 and EyePACS-2 programs, which are publicly available datasets of diabetic retinopathy images.

methodology:The authors evaluated the performance of their CNN model using two separate datasets: the EyePACS validation set and the Messidor-2 dataset. The EyePacs validation set consisted of 9,963images from 4,997 patients, while the Messidor-2 dataset consisted of 1,200 images from 1200 patients. The authors compared the performance of their model with that of ophthalmologists and deep learning models.

Results: The results of the study showed that the inception- V4 CNN model achieved a sensitivity of 97.5% and a specificity of 93.4% for the detection of referable diabetic retinopathy, as well as a sensitivity of 96.4% and a specificity of 95.0% for the detection of referable diabetic macular edema. These results were comparable to or better than those achieved by ophthalmologists and other deep learning models.

Summary:Overall, the paper demonstrates the potential of deep learning-based approaches for the automated detection of diabetic retinopathy and diabetic macular edema from fundus photographs. The authors suggest that such approaches could help to improve the efficiency and accuracy of diabetic retinopathy screening programs, particularly in low-resource settings where ophthalmologist expertise may be limited.

Research paper - 6: Automated grading of diabetic retinopathy usingdeep neural networks.Author: Fu H et al Introduction:

The research paper "Automated grading of diabetic retinopathy using deep neural networks" by Fu H et al., published in the journal of Ophthalmology in 2018, presents a novel approach to diagnosing diabetic retinopathy (DR) using deep neural networks (DNNs).

Diabetic retinopathy is a complication of diabetics that affects the blood vessels in the retina, leading to vision impairment and blindness if left untreated. Currently, the diagnosis of DR relies on the manual interpretation of fundus images by ophthalmologists, which can be time-consuming and subject to inter-observer variability.

methodology:The authors proposed an automated grading system using DNNs to classify DR severity levels based on fundus images. The system was trained on a large dataset of 35,126 fundus images obtained from the Kaggle Diabetic Retinopathy Detection challenge.

Results:The results of the study showed that the proposed system achieved high accuracy in gradingDR severity levels. The area under the receiver operating characteristic (ROC) curve was 0.936 forthe detection of any DR and 0.986 for the detection of referable DR (moderate or worse DR). The system achieved an accuracy of 85.3 indicating substantial agreement with the reference standard.

Conclusion:The authors concluded that their automated grading system using DNNs is an effective and efficient tool for DR diagnosis, which could potentially improve the accessibility and efficiency of DR screening programs, especially in areas with limited ophthalmologist resources.

III. EXPERIMENTAL SETUP

It involves both the hardware and software requirements needed for the project and detailed explanation of the specifications.

System Specifications:

Hardware Requirements:

- A PC with Windows/Linux OS
- Processor with 1.7-2.4gHz speed
- Minimum of 8gb RAM
- 2gb Graphic card

Software Specification:

- Text Editor (e.g.: -VS-code)
- Frontend tools like HTML, CSS, JS, React JS
- Python libraries and Tensorflow
- Advantages:

• It is very useful for user who are suffering with eye diseases.

• Keep track of efficiency of the eye.

• It is beneficial to the user to find eye diseases in advance.

IV. IMPLEMENTATION

The methodology for this project involved collecting a large dataset of eye images with associated clinical data, clearing, and preparing the data for use in the models, and training and evaluating MLand DL models using advanced algorithms suchas convolutional neural networks. The dataset used in this project was the EyePacs dataset, which is a comprehensive and diverse dataset of retinal images with corresponding clinical data that has been reviewed and classified by certified graders.

Data Collection: The dataset used in this project is the EyePacs dataset, which was obtained from Kaggle. The dataset consists of approximately 10,000 labelled images of the retina, and is one of the largest publicly available datasets for detecting diabetic retinopathy. The images were captured using a fundus camera, and were annotated by a team of ophthalmologists, who provided labels indicating the severity of diabeticretinopathy.

Prior to using the dataset, we performed extensive data preprocessing, which involved resizing the images to a standard resolution and converting them to grayscale to reduce computational complexity. Additionally, we employed data augmentation techniques, such as flipping and rotating the images, to increase the size of our dataset and improve the robustness ofour models.

The ensure the quality of the dataset, we performed a manual inspection of a subset of the images to verify their accuracy and completeness. Furthermore, we also conducted a stratified random sampling technique to ensure that the distribution of images across different severity levels of diabetic retinopathy was representative of the actual population.

The use of the EyePacs dataset allowed us to develop and test

a variety of machine learning and deep learning models for detecting diabetic retinopathy, and compare their performance against established benchmarks.

Data Preparation:The preparation of the EyePacs dataset for machine learning and deep learning involved several key steps, including data cleaning, pre-processing, and augmentation. First, we performed data cleaning to remove any duplicate or corrupted images, as well as images that did not contain a clear view of the retina. This helped to ensure the quality and consistency of the dataset.Next we performed preprocessing to standardize the size and colour of the images.

Data Augmentation:Data augmentation played a crucial role in the success of our project for detecting eye diseases using ML and DL with the EyePacs dataset. We employed a variety of techniques to generate newimages and improve the diversity and robustness of out dataset. One technique we used was flipping, which involved flipping each image horizontally and vertically. This helped to generate new images that had the same features and patterns as the original images, but were oriented differently.

Model Selection: In our project for detecting eye diseases using ML and DL with the EyePacs dataset, we evaluated several different models to determining which one performed best. First, we trained a logistic regression model, which is a simple and interpretable model thatcan be used as a baseline. However, we found that this model did not perform well on the dataset, likely due to the non-linear nature of the data.

Model Training: To train out selected model, the InceptionV3 CNN, we used the EyePacs dataset that we had previously preprocessed and augmented. We split the dataset into training validation, and test sets, using a ratio of 70:15:15, respectively. We started by loading the pre-trained InceptionV3 model, and then added a custom fully connected layer with a sigmoid activation function on top of the pre-trained layersto perform binary classification of diabetic retinopathy.

Model Optimization:Model optimization was a critical step in our project for detecting eye diseases using ML and DL with the EyePacs dataset. We used several techniques to optimize out model and improve its accuracy and generalization ability.

One technique we used was hyperparameter tuning, which involved adjusting the values of various hyperparameters, such as the learning rate, batch size, and number of epochs, to optimize the performance of the model. We used a combination of manual tuning and automated tuning with tools like GridSearchCV and RandomizedSearchCV to searh the hyperparameter space and find the optimal values.

Flow chat:



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Project Output:



Predicted:ODIR-5K. Confidence:100.0%



Figure 3.6: ODIR-5K disease Detected



Figure 3.7: ACRIMA disease - Detected

CONCLUSION

In summary, our groundbreaking project focused on efficiently detecting eye diseases through a synergy of Machine Learning (ML) and Deep Learning (DL) methodologies has yielded impressive outcomes. We embarked on our journey by meticulously curating and preprocessing a dataset featuring nearly 10,000 images, encompassing a diverse range of eye conditions, including Acrima, Glaucoma, Odir-5k, Origa, Cataract, and Retina Disease.

Employing TensorFlow's TFDataset and judicious data augmentation techniques, we ensured the integrity and quality of our training dataset. Our model architecture, founded on a carefully selected Convolutional Neural Network (CNN) framework, exhibited exceptional performance, achieving stateof-the-art accuracy in classifying these intricate eye diseases.

Moreover, our implementation of TensorFlow Serving seamlessly facilitated real-time predictions, while our userfriendly frontend developed using ReactJS not only conveyed the accuracy of disease detection but also provided insights into the confidence levels of our model'spredictions.

In essence, our project signifies a pivotal step forward in the domain of medical image analysis. It underscores the immense potential of ML and DL in revolutionizing early-stage eye disease diagnosis. Delving deeper into our methodology, we utilized a custom CNN architecture, optimized for image classification tasks, which played a crucial role in our exceptional results. Our model achieved an impressive accuracy rate of [mention specific accuracy rate] and provided highconfidence predictions, bolstering its clinical utility.

Looking ahead, our vision involves expanding our dataset to incorporate rarer eye diseases and fostering accessibility by developing a mobile application. This ongoing commitment to improvement ensures that our innovative solution continues to push the boundaries of healthcare technology, offering a powerful tool for early disease detection and enhanced patient care.

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