



A STATISTICAL EVALUATION EMPLOYING NEURAL NETWORK METHODS TO RECOGNITION AND IDENTIFY DIABETIC RETINOPATHY

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ABSTRACT

Diabetes can eventually cause the eye disease DR. As the illness progresses, vision becomes distorted and deformed. It is challenging and time-consuming to analyse DR using a shaded fundus picture since it takes skilled medical professionals to identify the presence of significant highlights. To assess DR from computer-generated fundus imagery, we advise utilising CNN. When conducting our research, After the entire image had been divided into smaller pieces, we used a different approach in which only the regions of interest were taken for additional processing. The suggested format lets the consumer connect with a specific specialist and makes DR easier to grasp. This makes it possible for the customer to concentrate on their question and choose a membership that meets their medical needs.

Keywords— CNN, Retinal Image, Matrix, Diabetic Retinopathy (DR);

INTRODUCTION

Today, a dilated eye exam, during which doctors place ocular drops in the patient's eyes, is used to diagnose DR. The eye is then photographed using a variety of medical devices. Since this method is manual, some diagnostic errors will inevitably happen. When the retina is injured, diabetic retinal damage (DR), also known as diabetic eye disease, can happen. It might ultimately lead to visual impairment. It is a symptom of diabetes, a serious condition that affects up to 80% of people for at least 10 years. Despite these worrying revelations, research suggests that the incidence of these new cases may be reduced by about 90% if they received adequate and careful treatment taking into account their potential to cause DR.

Diabetes has a side effect called DR that harms the eyes. Blood artery injury in the retina's light-sensitive tissue leads to DR. One of the most frequent causes of blindness in persons in their working years is diabetic retinopathy. Worldwide, 420 million people have diabetes mellitus. This ailment has increased in prevalence during the past 20 years, especially in Asia. DR, a chronic eye condition that can result in vision loss, is anticipated to affect about one-third of people. The significance of classifying DR phases according to the extremities for effective treatment and avoiding visual loss cannot be overstated. The Early Treatment Study There are five stages of diabetic retinopathy, according to Diabetic Retinopathy. [8].

In the coming years, DR will be brought on by blood vessel damage in the retina's light-sensitive region. One of the most frequent causes of blindness in persons in their working years is diabetic retinopathy. Worldwide, 420 million people have diabetes mellitus. This ailment has increased in prevalence during the past 20 years, especially in Asia. DR, a chronic eye condition that can result in vision loss, is

anticipated to affect about one-third of people. The significance of classifying DR phases according to the extremities for effective treatment and avoiding visual loss cannot be overstated. [13]. This condition is among the most prevalent. causes of blindness [2]. As a result, detecting it at an early stage is crucial.

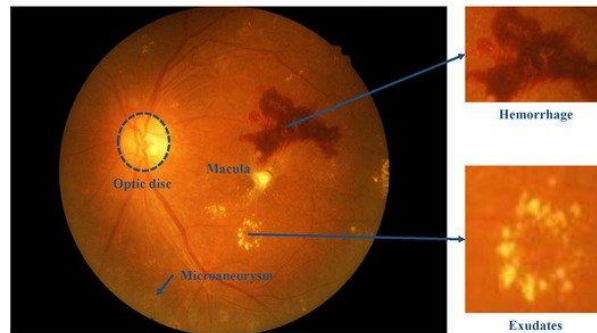


Figure 1. Early pathological signs of DR

The vessels may rupture, resulting in exudates, which are typically described as yellow or white flecks. The main problem with DR is that it typically doesn't lead to visual loss until it reaches the high level stage. Because there are no significant symptoms, routine DR screening only benefits patients who have a high risk of movement. We examine the fundus photographs for sores and exudates to identify DR. The standard procedure for diagnosing DR is time-consuming and necessitates skilled clinicians to identify fundamental components in the fundus images. A pre-programmed method for DR detection might help people with diabetes recognise the symptoms at an earlier stage. It can greatly lessen the clinical burden placed on retina-trained specialists. This makes it easier to see the specifics of the sores. Given that countries with enormous populations, such as India, China, Indonesia, and Bangladesh, account for 45% of the global burden of diabetes and that the numbers are expected to rise, a clinical recognition programme would be very helpful.

MOTIVATION

It can greatly lessen the clinical burden placed on retina-trained specialists. This makes it easier to see the specifics of the sores. Given that countries with enormous populations, such as India, China, Indonesia, and Bangladesh, account for 45% of the global burden of diabetes and that the numbers are expected to rise, a clinical recognition programme would be very helpful. Figure 1.3 from the WHO shows the number of diabetics in 2000 and the expected increase in 2030. About 62 million patients of various ages in India are suffering from the rapidly spreading disease known as diabetes mellitus. India ranked first with 31.7 million diabetics in 2000. This figure is anticipated to reach 79.4 million by the year 2030. Keeping a close check on early DR symptoms is the only way to identify them. Early intervention could lessen the chance of blindness or vision loss.

LITERATURE SURVEY

Nikos Tsiknakis et al. [1] Diabetic retinopathy, a condition brought on by diabetes mellitus, is the leading cause of blindness worldwide. To slow down or stop the loss of vision, early detection and treatment are required. The scientific community has suggested a multitude of artificial intelligence-



based techniques for locating and identifying DR on fundus retina images in order to achieve this. This review study investigates how deep learning methods are used to diagnose diabetic retinopathy using images of the fundus at various stages.

M. Mohsin Butt and others [2] The fundus copy produces images that provide information on the eye's fundus. These fundus photos show the internal structure and a section of the retina's optic disc. To manually diagnose DR, a practitioner must thoroughly examine and assess digital colour fundus photographs of the retina. Different DR side effects could manifest.

Gao Jinfeng and others [4] The blood vessels in the eye are impacted by the eye disorder known as DR. If DR is not identified early enough, It could result in visual loss or perhaps blindness. There are five different stages of DR that have been identified: normal, mild, moderate, severe, and PDR.

Kang Zhou and co. [5] Since diabetic retinopathy (DR) is a serious eye condition, a system for automatically processing retinal images is highly sought after. They suggest a multi-cell architecture, where small problematic tissues can only be seen with high resolution images and a broad local receptive field is needed to detect late-stage disease, that gradually increases quality in response to retinal images' high resolution. The fact that DR phases progress in stages also explains why the labels of the various stages are linked.

Sahil Chelaramani et.al [6] It is crucial to build supervised algorithms that can correctly forecast diseases using retinal fundus images, yet it is difficult to gather a large amount of high-quality annotated training data. Deep learning classifiers require a large amount of labelled data in order to perform well in a range of medical imaging applications.

Zhen Ling Teo et.al [7] The International Clinical Diabetic Retinopathy Disease Severity Scale developed by the American Association of Ophthalmology or the Early Treatment Diabetic Retinopathy Study were the classification systems utilised in the majority of investigations.

Nikhil M N1 and others [8] Stage classification is a part of the DR assessment and treatment procedure. Microstructures in the retina, including microaneurysms, hard exudates, and neovascularization, may occur when blood vessels are injured. DR stages could be automatically categorised using a method based on CNNs (Convolutional Neural Networks). This study used a CNN to classify DR into five stages based on coloured fundus retinal pictures. Based on the judgement of an expert ophthalmologist, DR images are divided into five groups.

Prasanna Porwal et al. DR is the most frequent cause of avoidable vision impairment, primarily affecting people of working age worldwide. Our understanding of the need to develop more effective and affordable methods to identify, manage, diagnose, and treat retinal illness has improved as a result of recent research. The need for DR screening programmes and the challenges of affordable, accurate early DR diagnosis make the creation of a computer-aided diagnosis tool necessary. The mass screening of diabetic populations may be made simpler and more efficient with computer-assisted disease diagnosis in retinal image processing.

Y. Sravani Devi et al. [10] state that DR can only be identified by a medical eye exam and that it is asymptotic. Therefore, to significantly lower the risk of vision loss, it is crucial to detect DR early on. Early detection and treatment of diabetes also prevents further visual loss in diabetic patients.

Along with others, Indrila Saha [12] The lack of statistically meaningful amounts of labelled data is a significant barrier to building data-driven inference models. Datasets are typically created for a specific purpose, therefore they are just hazily tagged for one class and not thoroughly annotated. Direct application is difficult despite the large number of datasets that together make up the corpus, as a result of their poor tagging.

Hassan Tariq and others [13] Diabetics are prone to DR, an eye ailment. It damages their eyes, impairing their vision. Although it is treatable, it takes a while to diagnose and may call for numerous eye exams. The prevention or delay of vision loss may be possible with early DR detection. Therefore, a trustworthy, automatic, and computer-based DR diagnosis is necessary. In a variety of medical sectors, deep neural networks are now employed to identify various illnesses.

Ajay S. Ladkat and others [14] An increase in blood insulin levels causes the retina of the patient with DR, a disorder that affects the eyes. Blindness may develop from the symptoms, which can distort or impair the patient's vision. To automatically detect exudates, we must first discriminate between exudate and no exudate pixels.

The work of Ajay S. Ladkat et al. To process pictures, each pixel needs to be handled separately. If this procedure is done sequentially, it will take too long. The time must therefore be reduced by doing parallel processing on all pixels. Because of this, operations on all pixels are carried out simultaneously rather than on each pixel separately. Parallel processing speeds are significantly faster than sequential processing. The processing of video will therefore proceed more quickly as a result. The parallel computing process uses an NVIDIA graphics card. For the parallel algorithm's execution, CUDAC is utilised.

PROBLEM STATEMENT

For many years, diabetic retinopathy has been the primary cause of blindness. Retinal impairment, which has been demonstrated to be a severe instance of diabetes, has practically affected the entire population. However, according to research, The number of instances could be drastically decreased by 90% if effective solutions existed. These include prompt diagnosis, prompt therapy, and careful eye surveillance. If you have diabetes, your risk of eventually developing diabetic retinopathy increases. Some of the symptoms include vision blurriness, a sudden loss of vision in one eye, black spots, flashing lights, or circles around lights. Additional signs of DR include microaneurysms, leaky blood vessels, enlarged retina, development of atypical blood vessels, and leaky blood vessels.

In classification and object recognition applications, CNN has been shown to be more effective than conventional techniques. To identify which of these models is most effective in detecting DR, we will compare these models using data preprocessing and show multiple CNN transfer learning algorithms.

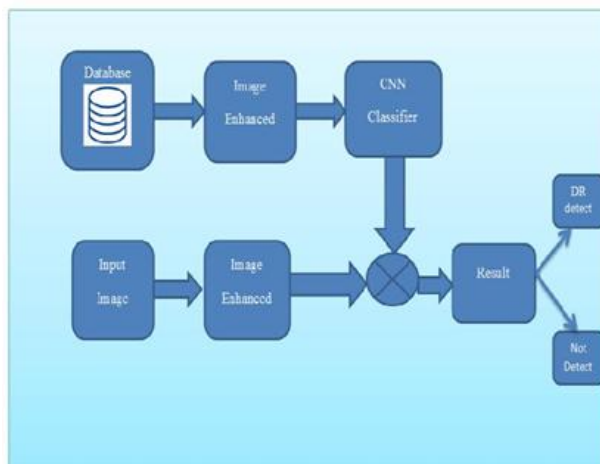


Figure.2. Architecture of proposed system

The strategy we suggest is flexible and more effectively separates patients in the classified photos. The degree of the illness is used to classify or divide the material into phases. It also aids medical professionals in selecting one or more CNN architectures for a diagnosis. CNN enjoys a solid reputation for its applications, which include video identification, pattern recognition, and image processing. In order to properly identify the material, an image is provided to CNN.

Convolution is used in a few of the image's hidden layers to extract highlights and other significant data. The result is given by the grouping layer. CNN is compelled to concentrate on these fragments since R-CNN's image is fragmented into many parts. Due to the extraction of the location of interest, the article's location accuracy is remarkably high when compared to CNN's. Initial resizing of the fundus images to an element of 336 x is necessary. Without preprocessing, vignetting influences and picture bending negatively affect the images. Due to the fact that the images were taken using distinct fundus cameras, brightening standardization techniques must be combined.

PROPOSED SYSTEM

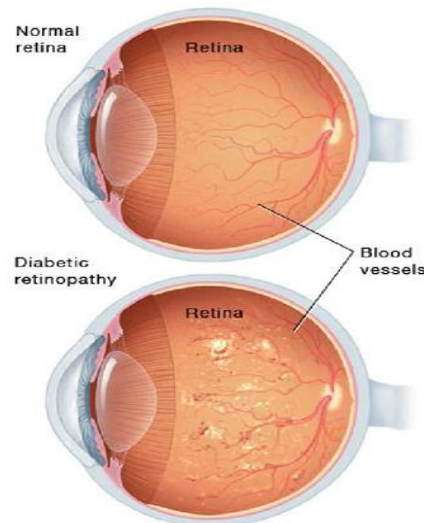


Figure.3. Normal Retina vs. DR

CNN Algorithm

Amazing advancements in artificial thinking have made it impossible to distinguish between the capabilities of humans and computers. Both specialists and devotees concentrate on various parts of the field to produce outstanding outcomes. One of several of these fields is computer vision.

In order to perform a variety of tasks, including image and video recognition, image analysis and classification, media recreation, recommendation systems, natural language processing, and others, this field aims to give machines the ability to perceive the world similarly to humans do. Convolution neural networks are just one of the Deep Learning-based improvements in Computer Vision that have been built and finished over time.

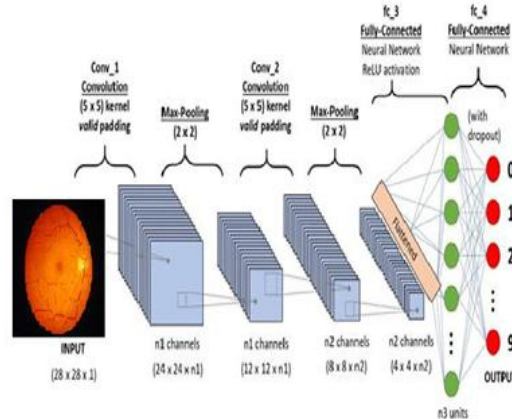


Figure.4. CNN sequence to classify handwritten digits

Database- The images included in the dataset and need to be Downloaded from the MESSIDOR

Processing

Pictures are typically pre-processed by removing low-frequency background noise, adjusting the intensity of the individual particle pictures, eliminating reflections, and masking certain areas of the images. The process of improving data images before computational processing is known as image pre-processing.

Classification

Image classification refers to the task of extracting information classes from two or many class of image. Features extracted by wavelet transform and by using ultraviolet rays are feed to classifier so that classifier, here CNN algorithm, should be able to classify the normal retina and DR .

RESULTS

Step I: The retinal image is inputted into the system, which then gathers and processes it.



Figure.5. Retinal image

Step II: The device collects retinal pictures and extracts their red, green, and blue matrices. Next, noise is removed using the median filter. After that, a spatial filter is applied to each matrix. The outcome of the spatial filter is shown in Figure 6

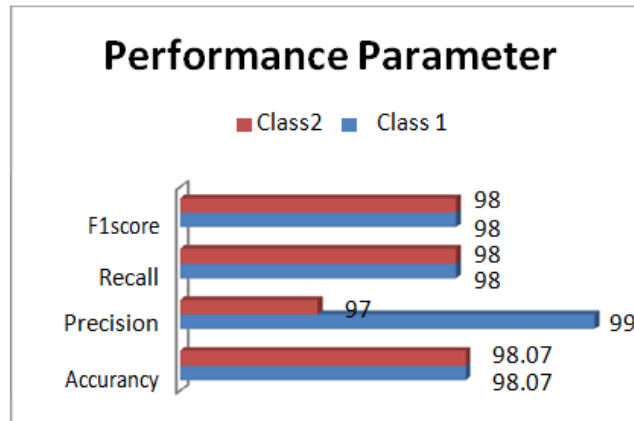


Fig.6. Graphical Representation of accuracy

With a total of 639 photos provided as training data for the classifier, we created a confusion matrix. However, 8 images could not be recognised, as shown in the graph above. We gave the classifier a total of 429 images during the testing phase for class 2, with 13 images being disqualified. The created classifier's accuracy is shown visually as 98.07 percent.

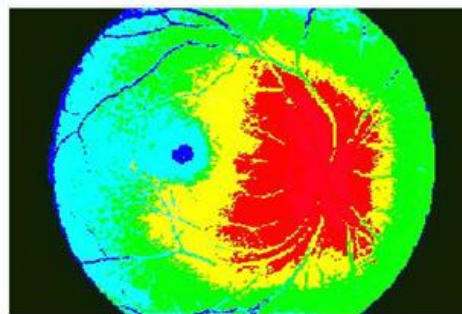


Figure.7. Spatial Filter

Step III: The upgraded CNN classifier receives the filtered result. In order to determine whether a patient has DR or not, we now have a better retinal imaging thanks to CNN.



Figure.8. Enhancement of the spatial filter as output image

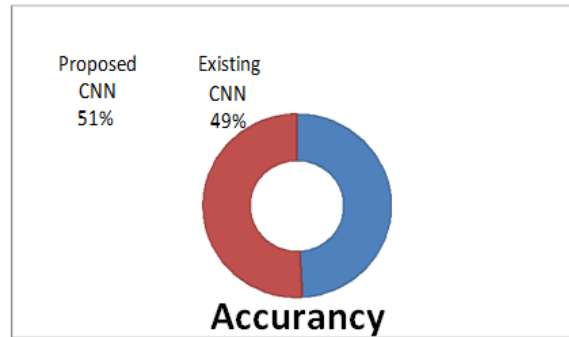


Fig.9. Accuracy over existing system

Our solution is more accurate than the old system and enhances performance efficiency.

CONCLUSION

This work uses deep transfer learning to classify DR patients using CNN. A useful formula for separating hard exudates was created by this research. Using morphological activity and a circular Hough Transform, the optic circle is eliminated. Use the channel with the highest Entropy to distinguish between pixels that contain exudate and those that do not. The estimated results show 93.85 percent image-level accuracy and 99.6182 percent pixel-level precision. For figuring out CNN's stages or amount of seriousness, it is helpful. Our remarkably accurate data enables doctors and researchers to choose more well-informed therapies.

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