# DIABETIC RETINOPATHY DETECTION

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## ABSTRACT

The blood vessels in the retina are harmed by a consequence of diabetes called diabetic retinopathy. High blood sugar levels over time might harm these blood vessels, leading to vision loss or even blindness may result from them leaking or becoming obstructed. Usually, the illness develops in phases, from mild no proliferative diabetic retinopathy to moderate and severe no proliferative diabetic retinopathy, and finally, proliferative diabetic retinopathy. Depending on the severity of the illness, diabetic retinopathy treatment options can range from less intrusive interventions like lifestyle changes to more invasive ones like laser surgery or injections. For people with diabetes to avoid vision loss and maintain their eye health, early detection and treatment are essential.

Keyword : - Diabetic Retinopathy, Proliferative, non-proliferative

## **1. INTRODUCTION**

A significant side effect of diabetes that affects the eyes, specifically the blood vessels in the retina, is known as diabetic retinopathy. It is the primary cause of adult blindness in affluent nations, and its incidence is rising globally as diabetes is becoming more and more common. Chronically high blood sugar levels are the root cause of diabetic retinopathy, which can harm the tiny blood vessels that feed the retina and cause vision loss or even blindness. Proliferative diabetic retinopathy is the final stage of the ailment, which normally develops in stages starting with mild no proliferative diabetic retinopathy, moderate, and severe proliferative diabetic retinopathy, and finally proliferative diabetic retinopathy. To prevent vision loss and maintain eye health, diabetic retinopathy must be identified and treated early. An overview of diabetes will be given in this introduction.

## 2. PROBLEM STATEMENT

Diabetic retinopathy is a significant public health problem that affects millions of individuals worldwide. It is a leading cause of blindness among adults, particularly those with diabetes, and its incidence is on the rise. Despite the availability of effective treatments, many individuals with diabetic retinopathy are not diagnosed or receive appropriate treatment until the condition has progressed to a more advanced stage, which can lead to irreversible vision loss. This delay in diagnosis and treatment is often due to lack of awareness among individuals with diabetes, as well as healthcare providers, about the importance of regular eye exams for early detection and treatment of diabetic retinopathy. Additionally, barriers to accessing healthcare, such as cost and transportation, can further

impede timely diagnosis and treatment. Thus, there is an urgent need to improve vision loss and improve the quality of life for individuals with diabetes.

## **3. LITERATURE REVIEW**

Over the times, multitudinous scholars and entrepreneurs have made multitudinous exchanges and disquisition on how to use deep knowledge technology to meliorate and manage the diabetes retinopathy.

Melaheh Imani. (1) addresses the visibility of the retinal vessels. Different DR related lesions can be found using the DR discovery system. This method is independent of the lesion's size, colour, or type. The MCA algorithm separates the healthy and pathological structures in the retinal image. Eventually, statistical factors of the retinal lesions are employed to differentiate between normal and atypical picture data. Analysing morphological elements is employed.

Ms. Ankita Gupta. (2) It is suggested that two methods—blood vessel segmentation and lesion identification—are used to discover DR. This study examines the experimental findings from various databases, including DRIVE, CHASE, and others, using metrics like as perceptive, clarity, sharpness, and total area under the wind (AUC). The problems with deep neural networks are likewise diverse.

Mohamed Shaban. (3) The use of a focus-deep CNN with three fully connected layers and 18 convolutional layers is advised for analysing fundus prints and automatically differentiating between the various colourful retinopathy scenarios.

According to WejdanL. Alyoubi (4), automated network styles significantly reduce the amount of time needed to make decisions, saving ophthalmologists' time and enabling quick case resolution. Deep knowledge approaches have been introduced, and closely accessible common fundus DR datasets have been presented.

According to Carson Lam. (5), this is because CNN is unable to recognise nuanced complaint elements. They found that there was a pre-processing step that involved an automatic DR rating system that could estimate prints in accordance with complaint pathology in four vividly harsh conditions.

## 4. METHODOLOGY

#### 4.1 Supervised learning model

We have used total 5 types of datasets in diabetic retinopathy.

- 1. No Dr: which gives the result as no diabetic retinopathy detected.
- 2. Mild: Which gives the result it's a phase 1 of diabetic retinopathy and can be cured if taken proper treatment.
- 3. Moderate: which gives the result it's a phase 2 of the diabetic retinopathy and should be treated under doctor's consultant.
- 4. Severe: Which gives the result that the eye has begun to damage and should be treated as soon as possible.
- 5. Proliferative: Which gives the result it's in last phase and already damaged some part of eye and can make the eye blind.

## 4.2 Proposed model



- Gather pictures as a dataset, then upload them using programming. The system was then developed using feature extraction, image processing, and classification techniques.
- The initial picture dataset that underwent pre-processing (Clean dataset, RGB image to Gray-to-Binary Conversion).
- After that, use the CNN approach to categorize the dataset by extracting its features (such as image dimensions, width, height, and label).
- The dataset was trained using this approach, which also produced the CNN model. Using This Model to Determine an Image's Diabetic Status.

#### 4.3 Model specification

Convolutional Neural Network (CNN):

Convolutional neural networks (CNNs, or ConvNets) are a type of artificial neural network (ANN) used most frequently in deep learning to analyse visual data. Due to their shared-weight design, which is built on convolution kernels or filters, CNNs are occasionally referred to as Shift Invariant or Space Invariant Artificial Neural Networks (SIANN). that slide along input features and produce feature maps that are translation equivariant in responses. Because most convolutional neural networks use a down sampling operation to the input, they are typically not translationally invariant. They have uses in the recognition of images and videos, recommender systems, classification and segmentation of images, analysis of images used in medicine, natural language processing, brain-computer interfaces, and time series of financial data.

Convolutional neural networks with an emphasis on applications for image and video recognition. CNN is primarily used for segmentation, object detection, and image recognition tasks in image analysis applications.

There are Four types of layers in Convolutional Neural Networks:

1.Convolutional Layer: In a traditional neural network, the next hidden layer is connected to each input neuron. In CNN, the hidden layer of neurons is coupled to a relatively small number of the input layer neurons.

2.Pooling Layer: The feature map's dimensionality is decreased using the pooling layer. There will be several activation and pooling layers inside the CNN's hidden layer.

3.Flattening: involves turning data into a one-dimensional array so that it may be entered into the following layer. The output of the convolutional layer is flattened to create a single, extensive feature vector.

4.Layer that is fully connected: The final several layers of the network are known as Fully Connected Layers. The last convolutional or pooling layer's output is sent to the fully connected layer, where it is flattened before being used.

## **5. PROJECT IMPLEMENTATION**

Working:

1. If a person comes out to be a new user, he/she can register the application. They have to fill all the personal detail in the form.





2. After registration user have to login the application using username and password.

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3. Uploading the image and then the fundus image is given as an input to the pre-trained model.



Fig: Upload image.

4. Getting the result after the image is uploaded and scrutinized, once gets an accurate result of intensity of his/her diabetic retinopathy.



Fig: NO DR Result

Then the parameters are as follows:

- 1. NO DR
- 2. Mild DR
- 3. Moderate DR
- 4. Severe DR
- 5. Proliferative DR

# 6. TOOLS AND TECHNOLOGY REQUIREMENT

- Hardware requirements:
  - 1. RAM: 8GB
  - 2. Hard disk: 40 GB
  - 3. Processor: Intel i5 processor
- Software requirements:
  - 1. IDE: Anaconda Navigator.
  - 2. Coding language: Python version 3.5
  - 3. Operating System: Windows 10
  - 4. Database: DB SQLite

## 7. RESULT AND DISCUSSION

We discussed a proposed system and the execution of our theory in the previous chapter. We have presented the data set we used, its description, visualization, and the techniques we employed. We are now upset by the outcomes of our investigations into the administration of this system. Our dataset has been split into training and testing datasets. This chapter will demonstrate how the training and testing dataset has grown. As previously indicated, four machine literacy algorithms were employed. First, we built a model and trained these four algorithms on our dataset. Additionally, we evaluated this model using our testing dataset. If the test set accuracy is close to the train set accuracy, we can say that we constructed a decent model. If the test set delicacy is close to the train set delicacy as well.

## 8. CONCLUSION

In conclusion, diabetic retinopathy is a significant complication of diabetes that affects the eyes, particularly the blood vessels in the retina. It's a leading cause of blindness among grown-ups in developed countries, and its prevalence is on the rise worldwide. The condition is caused by chronically high blood sugar situations, which can damage the small blood vessels that nourish the retina, leading to vision impairment or indeed blindness. Beforehand discovery and treatment of diabetic retinopathy are pivotal for precluding vision loss and conserving eye

health. Recent literature reviews have stressed the significance of regular eye examinations, early opinion, and timely treatment of diabetic retinopathy to help vision loss and ameliorate issues for individualities with diabetes. Advances in artificial intelligence and machine literacy have also shown pledge in the early discovery and opinion of diabetic retinopathy.

## 9. FUTURE WORK

Any investigation may always be improved upon. That is not an exception in our case. We have created a few spaces where this system can be used:

1. Beforehand discovery and opinion: One of the crucial challenges in managing DR is detecting it beforehand enough to help unrecoverable vision loss. There's a need for more accurate and effective webbing styles, similar as using artificial intelligence algorithms to dissect retinal images.

2. Case education and operation: There's a need for better education and support for cases with diabetes, particularly in terms of precluding and managing DR. This could involve developing individualized treatment plans, using digital health tools, and perfecting access to care for underserved populations.

3.Understanding the underpinning mechanisms: There's still important to be learned about the underpinning mechanisms of DR, including the part of inflammation, oxidative stress, and other factors in the development and progression of the complaint. More understanding these mechanisms could lead to new treatment targets and strategies.

4.cooperative sweats: Collaboration among experimenters, healthcare providers, and cases is essential for advancing the field of DR. This could involve sharing data and coffers, coordinating clinical trials, and engaging cases in exploration and advocacy sweats.

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