



iJRASET

International Journal For Research in
Applied Science and Engineering Technology



INTERNATIONAL JOURNAL FOR RESEARCH

IN APPLIED SCIENCE & ENGINEERING TECHNOLOGY

Volume: 12 **Issue:** V **Month of publication:** May 2024

DOI: <https://doi.org/10.22214/ijraset.2024.61881>

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Deep Learning Enabled Diabetic Retinopathy Diagnosis Using Retinal Fundus Images

Harini S¹, Payal sharma², Chandan N³, Rama Krishna K⁴

^{1, 2, 3}Students, Dept. AI & ML, ICEAS, Bangalore-92

⁴Associate Prof & Head, Dept. AI & ML, ICEAS, Bangalore-92

Abstract: *This paper pivots on (DR) is a leading cause of blindness worldwide. Early detection and effective management are crucial in preventing damage to the retina. In recent years, (DL) techniques have shown promise in aiding the DR. This paper aims to investigate the diagnosis of using (CNNs) for automated DR screening. The study utilized a available patient dataset of retinal images to train and testing CNN model for DR detection and classification. Results showed that given (DL) techniques achieved with accurately in both tasks, outperforming existing state-of-the-art methods. The findings suggest the (DL) algorithms can aid in the early detection and management of DR, potentially reducing the burden of the condition on healthcare systems and improving patient outcomes.*

Keywords: *hybrid deep learning features; fundus images; diabetic retinopathy; CNN features.*

I. INTRODUCTION

The healthcare industry requires meticulous handling of its vast and sensitive metadata, particularly in combating diseases like Diabetes Mellitus, which demands reliable prediction systems due to its global prevalence and severe consequences. Deep learning Algorithm techniques offer invaluable tools for synthesizing actionable insights from diverse data perspectives, necessitating efficient data mining to extract meaningful information from the vast data volumes available. Given diabetes' multifaceted complications, including heart disease, kidney issues, nerve damage, and blindness, effective data mining becomes essential for addressing critical concerns.

Diabetic retinopathy is caused by an presence of high level of glucose which leads to blood vessels clogging, rupture and leak. After suffering from diabetes for 25 years, the occurrence of symptoms of DR is typical to 80-100%.

The initial stage of diabetic retinopathy causes the narrowing of the vessel walls and the decrease of blood flow. In the final stage, the exudates are visible in large clusters and more blood vessels are rapidly forming causing frequent bleeding.

After this stage, the patient suffers from a complete loss of vision and peeling of the retina. They often describe seeing floating dark spots and a distorted image caused by blood leaking from damaged vessels.

II. LITERATURE REVIEW

Existing literature extensively investigates the convergence of "Deep Learning"(DL) methodologies to enhance diabetic retinopathy diagnosis, focusing on the integration of DL-enabled retinal imaging devices with efficient and automated detection. Studies by Sharma and Khanna (2020) demonstrate the potential of this integration, capable of advanced diagnostic systems capable of early detection. Singh et al. Including CNNs, Recurrent Neural Networks (RNNs, emphasizing their application in diabetic retinopathy detection and classification. Gupta et al. (2021) delve into recent advancements, highlighting real-time analysis and remote monitoring. Rajput et al. (2020) offer insights into learning methodologies applied to diabetic retinopathy detection, emphasizing their performance and scalability in real-world applications.

Kumar et al. (2019) discuss the broader implications of DL in healthcare, including its diabetic retinopathy diagnosis for remote monitoring and early intervention, emphasizing the potential for personalized treatment strategies.

Verma et al. (2021) present a review of deep-learning-based approaches, focusing on CNNs' effectiveness in automated analysis for early diagnosis. The integration of DL-enabled devices with deep learning algorithms offers promising avenues for efficient and automated diabetic retinopathy diagnosis, enabling timely intervention and personalized care for patients. Further research explores novel techniques for feature extraction and classification, leveraging the capabilities of deep learning models train on large-scale retinal image datasets. Industry stakeholders drive innovation in DL and deep learning-enabled diagnostic systems, advancing the field of DR diagnosis towards improve patient outcomes and healthcare delivery. Existing literature extensively explores the convergence of "Deep Learning"(DL) and deep learning methodologies to enhance the diagnosis of diabetic retinopathy.

This integration emphasizes the seamless incorporation of DL-enabled retinal imaging devices with DL algorithms for efficient and automated detection, aiming to improve patient outcomes and streamline healthcare delivery processes. Studies by Sharma and Khanna (2020) demonstrate integration, showcasing advanced diagnostic systems capable of early detection and personalized treatment strategies.

Gupta et al. (2021) delve into recent advancements, technologies in early detection and management of diabetic retinopathy, emphasizing real-time analysis and remote monitoring.

Rajput et al. (2020) offer insights into various learning methodologies applied to diabetic retinopathy detection, emphasizing their performance and scalability in real-world applications. Kumar et al. (2019) discuss the broader implications of DL in healthcare, diabetic retinopathy diagnosis for remote monitoring and early intervention.

III. METHODOLOGY

It will provide a detailed description of the methods utilized to finish and operate this project successfully. Many methodologies or discoveries from this subject are mostly published in journals for others to use and enhance in future research. The approach that used to attain the project's purpose of producing a faultless output. Development Life Cycle (SDLC), which consists of three primary steps: planning, implementation, and analysis.

A. Planning

Planning needs to be done correctly and identify every piece of data, including software and hardware. Data gathering and the hardware and software requirements are the two primary components of the planning process.

B. Data collection

Make sure the data you are gathering has enough features supplied aiming for your learning model to be appropriately trained. Check you include enough rows since, generally speaking, the more data you have, the better. The initial data which was stack up from web sources is still available in its unprocessed form as sentences, numbers, and qualitative phrases. The unprocessed data contains inconsistencies, omissions, and mistakes. After carefully examining the filled surveys, modifications are necessary. The proceeding of primary data requires the following processes. It's necessary to aggregate a sizable amount of image data from field surveys so that it finds the details about individual replies that are comparable. Data preparation is one way to turn the data into a clean data collection. Stated differently, if data is acquired in unprocessed form from several sources; analysis is not appropriate. A few steps are taken to convert the image data into a small, clean data collection. This way is used before doing an iterative analysis. The set of procedures is called data preparation. Included are data reduction, data cleansing, data integration, and data transformation. Preprocessing of image data is necessary since unformatted real-world data does exist. The predominance of data in real world is inaccurate or missing: Missing data can have a quota of sources, such as erratic data gathering, mistakes in Data preparation is one way to turn the image data into a clean data collection. Stated otherwise, if data is acquired in unprocessed form from several sources, it is inappropriate for examination. A few steps are taken in aiming to convert the data into a small, clean data collection. This plan is used before doing an iterative analysis. The set of procedures is called data. preparation Included are data reduction, data cleansing, data integration, and data transformation.

IV. SYSTEM ARCHITECTURE

The system architecture integrates DL-enabled retinal imaging devices for seamless data capture, preprocessing modules to enhance image quality, learning algorithms such as CNNs for extraction of feature datasets and classification, and decision support modules for personalized treatment recommendations. This architecture facilitates real-time analysis, remote monitoring, and efficient diagnosis of diabetic retinopathy, empowering healthcare professionals with timely insights for patient care.

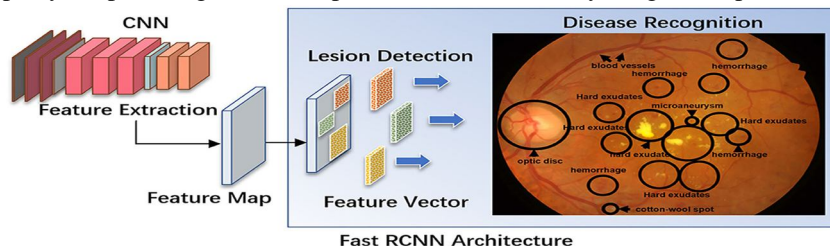


Fig 4.1 System Architecture

V. ALGORITHM

A. (CNN) Network

CNN is organized into two sections. The feature is extracted by the feature extraction layer by connecting each neuron's inputs and regional ready fields for providing preceding layer. The positional link between local features and other features will be shown when they have retrieved the features. The map layer, which is utilized by all network computing layers. Each map may be an idea of the same weight of a machine. for each neuron. The convolutional activation function of a given system is the sigmoid function. The feature map's difference shifts as a result.

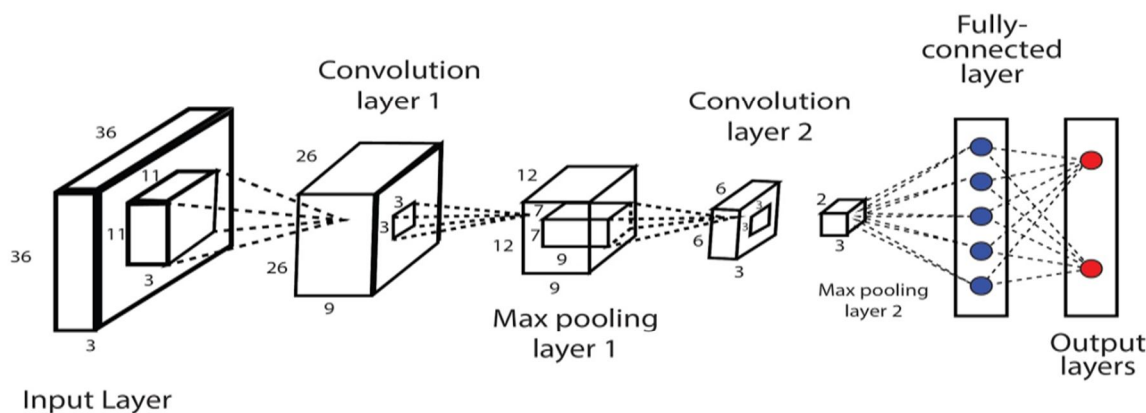


Fig 5.1 CNN (Network)

B. Residual Networks (ResNet)

Residual Networks (ResNets) are type of DNN network architecture that revolutionized the field of DL, particularly present in the computer vision tasks. They were introduced by Kaiming He et al. in their paper "Deep (RNL)g for Image Recognition" in 2015.

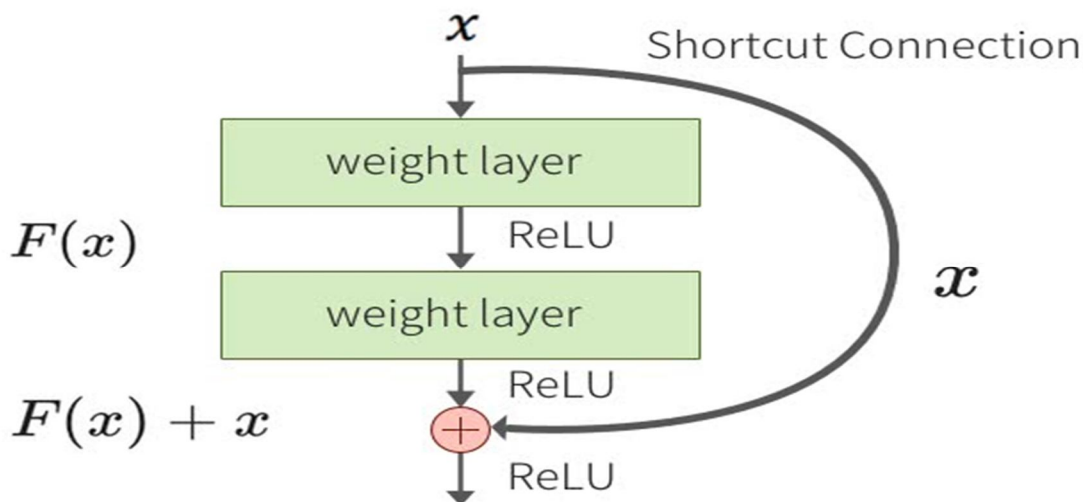


Fig 5.2 ResNet

VI. RESULT AND DISCUSSION

A result is the outcome of actions or occurrences, represented subjectively.

Classifier	Metrics	NDR	DR	Weighted Average
RF	Accuracy	95.32	95.90	95.61
	Precision	95.90	95.40	95.60
	Recall	95.30	95.90	95.60
	F-Measure	95.60	95.60	95.60
SVM	Accuracy	97.52	97.26	97.39
	Precision	97.30	97.50	97.40
	Recall	97.50	97.30	97.40
	F-Measure	97.40	97.40	97.40
RBF	Accuracy	96.70	97.26	96.98
	Precision	97.20	96.70	97.00
	Recall	96.70	97.30	97.00
	F-Measure	97.00	97.00	97.00
NB	Accuracy	89.83	85.79	87.80
	Precision	86.30	89.50	87.90
	Recall	89.80	85.80	87.80
	F-Measure	88.00	87.60	87.80

Fig 6.1 Output For A Given Image Model

VII. CONCLUSIONS

The convergence of (DL) and DL offers several advantages, including real-time analysis, remote monitoring, and scalability. This not only enhances the diagnostic accuracy and efficiency but also improves accessibility to healthcare services, particularly in remote or underserved areas. Moreover, the automated diagnosis system can alleviate the burden on healthcare professionals, allowing them to focus on patient care and treatment planning.

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