

ENHANCING IMAGE SEGMENTATION IN DIABETIC RETINOPATHY FUNDUS IMAGES USING DEEP LEARNING TECHNIQUE

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ABSTRACT

Diabetic retinopathy (DR), a severe complication arising from diabetes, make a significant threat to vision due to the deterioration of retinal blood vessels. This study presents an innovative methodology for automated detection, grading, and segmentation of diabetic retinopathy using deep learning, with a focus on residual encoder-decoder architecture. The study utilizes the Indian Diabetic Retinopathy Image dataset (IDRID), comprising 80 fundus images and labels, to rigorously evaluates the proposed methodology. By employing advanced image preprocessing techniques to enhance data quality, followed by a unified model capable of both segmentation and classification tasks. These findings contribute to the improvement of diagnostic accuracy and patient outcomes in retinal diseases, offering potential applications in clinical settings to support early detection and management of DR, thereby enhancing patient care and alleviating healthcare system burdens.

Keywords:

Diabetic retinopathy, Encoder-decoder model, Image processing, Segmentation, Classification.

INTRODUCTION

Diabetic retinopathy, a complication stemming from diabetes, poses a serious threat to vision due to the deterioration of retinal blood vessels. This condition progresses through stages such as non-proliferative diabetic retinopathy (NPDR) and proliferative diabetic retinopathy (PDR). NPDR involves the weakening of small retinal vessels, leading to fluid leakage, while PDR is marked by the growth of abnormal new blood vessels, increasing the risk of bleeding and detachment of the retina.

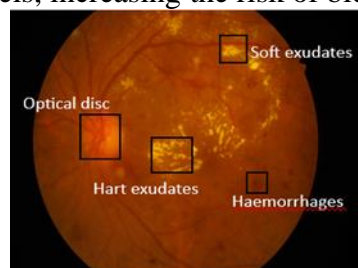


Figure.1: Diabetic retinopathy fundus image

Blurred vision, a notable symptom, highlights the importance of regular eye examinations for timely detection. The need for effective computer-assisted diagnosis systems for Diabetic Retinopathy (DR) is evident from the challenges posed by manual identification of retinal lesions. identification of retinal lesions urges the development of an efficient computer-assisted diagnosis system for Diabetic Retinopathy (DR). Recent research highlights a concerning trend of increasing retinal ailments among adults, significantly impacting vision and making DR a leading cause of blindness in diabetes-related diseases.

Recent studies indicate a troubling rise in retinal diseases among adults, significantly affecting vision and making diabetic retinopathy a leading cause of blindness in diabetes-related conditions. Advanced deep learning techniques, especially the Residual Encoder-Decoder model, are pivotal in detecting and analysing diabetic retinopathy. This model utilizes innovative residual connections inspired by the U-net architecture to extract complex features from retinal images and accurately identify different stages of DR. Moreover, machine learning models have demonstrated exceptional effectiveness in diagnosing a range of retinal diseases, thus streamlining screening processes and



supporting large-scale healthcare initiatives. The versatility of these models allows them to efficiently handle diverse datasets, thereby enhancing the precision and efficiency of detecting and analyzing diabetic retinopathy and related retinal diseases.

LITERATURE SURVEY

[1] Wang et al. (2022) introduced deep learning models for diabetic retinopathy (DR) detection and stage grading, emphasizing referable DR identification through lesion and stage integration. The study annotated fundus images by licensed ophthalmologists, partitioned into training, validation, and internal test sets, with an external test set for validation. Four deep learning models incorporating DR lesions and stages significantly improved sensitivity, specificity, and AUC metrics for referable DR identification, with high precision in detecting specific. The importance of integrating lesion detection and stage grading was highlighted for superior diagnostic accuracy, across diverse patient populations and imaging conditions requires further investigation and validation in clinical settings. [2] Lin and Wu (2023) presented a revised ResNet-50 model for DR detection, leveraging deep learning advancements to refine diagnostic capabilities specifically for diabetic retinopathy. Concurrently, Manan et al. (2023) explored semantic segmentation of retinal exudates using a residual encoder-decoder architecture, offering insights into lesion identification within DR. These studies demonstrate the evolving landscape of deep learning techniques in addressing challenges in DR diagnosis and emphasize the importance of interpretable and clinically applicable models. Nonetheless, broader datasets and robust validation in clinical settings are essential for advancing automated DR detection and classification. [3] In 2023, a study on semantic segmentation of retinal exudates using a residual encoder-decoder architecture underscored the importance of leveraging deep learning to automate the identification of exudates, a common sign of diabetic retinopathy. The study proposed a residual CNN with skip connections for robust segmentation accuracy, demonstrating promising results in precision, accuracy, sensitivity, specificity, and AUC metrics. This work highlights the potential of deep learning-based methods in enhancing the efficiency and accuracy of DR diagnosis and management.

EXISTING METHODS

The existing methods for segmenting and classifying diabetic retinopathy (DR) using deep learning techniques primarily involve sophisticated architectures tailored for medical image analysis. For segmentation, popular methods such as the U-Net architecture and residual encoder-decoder networks are utilized to accurately identify and delineate DR-related lesions like exudates and hemorrhages. These architectures leverage skip connections and residual connections to preserve spatial details and optimize gradient flow, resulting in precise lesion segmentation. In terms of classification, Convolutional Neural Networks (CNNs) including VGG, Res-Net, and Dense-Net are extensively employed to automatically learn discriminative features from retinal images and classify them into different DR severity levels or referable/non-referable DR cases. Transfer learning is often applied by fine-tuning pre-trained CNN models on DR datasets to enhance classification performance, particularly in scenarios with limited labeled data. Attention-based mechanisms further refine classification by focusing on salient image regions, collectively demonstrating the efficacy of deep learning in segmenting and classifying diabetic retinopathy for improved diagnosis and treatment planning.

PROPOSED METHODOLOGY

This introduces a holistic methodology to automate Diabetic Retinopathy (DR) detection, grading, and segmentation, aiming to improve early diagnosis and severity assessment for better patient outcomes. By leveraging advanced image preprocessing, deep learning feature extraction, and metric evaluation techniques, the goal is to develop an efficient DR grading system and precise image segmentation using the Indian Diabetic Retinopathy Image dataset (IDRiD), comprising 80 high-resolution images from clinical exams in India. Initial efforts focus on refining image quality through preprocessing,

followed by employing a Residual Encoder-Decoder model for deep learning analysis. This model excels in capturing subtle features indicative of different DR stages, offering a comprehensive solution to enhance the efficiency and accuracy of DR diagnosis and management, thereby advancing retinal disease diagnosis and improving patient care.

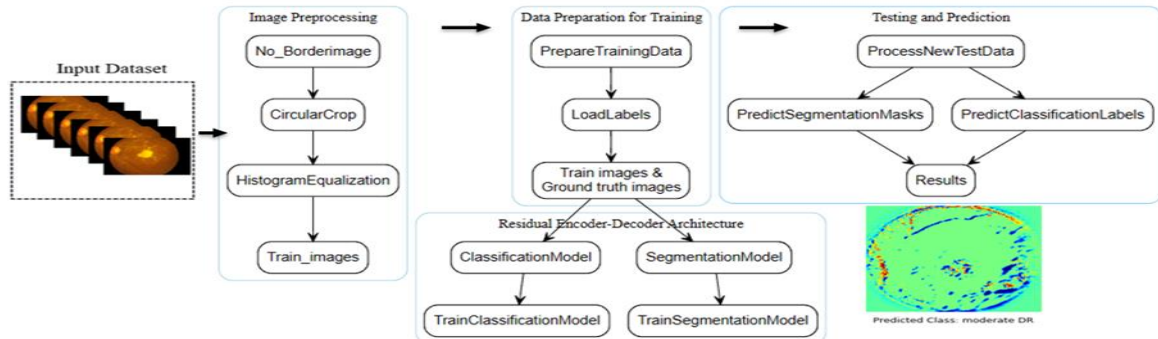


Figure.2. Visualization of the proposed model

IMPLEMENTATION

The implementation utilizes TensorFlow within the Spyder environment to enable early diagnosis and classification of Diabetic Retinopathy (DR) using fundus images. The integrated workflow begins with detailed image preprocessing techniques aimed at enhancing data quality. Image preprocessing techniques are used to enhance image quality and relevance for advanced analysis. This involves converting images to grayscale, applying binary thresholding to remove unwanted borders, and using



Figure.3: IDRID database results. (a) Original image. (b) No-border image. (c) Circular crop image. (d) Histogram equalization

contour analysis for precise border removal. Circular cropping isolates the retinal area, while histogram equalization improves contrast. These steps create high-quality images suitable for feature extraction and machine learning. The Residual Encoder-Decoder model, a neural network architecture, combines residual networks with encoder-decoder frameworks for tasks like image segmentation and classification. It's effective for complex image data due to residual connections preserving details. After preprocessing, images are processed using a semantic segmentation and classification model. Convolutional layers with pooling operations capture hierarchical features. The output includes pixel-wise segmentation masks. Additionally, a classification model with similar structure predicts class probabilities. Both models are trained and tested on the IDRID dataset using appropriate optimization and evaluation techniques to optimize performance on medical image analysis tasks.

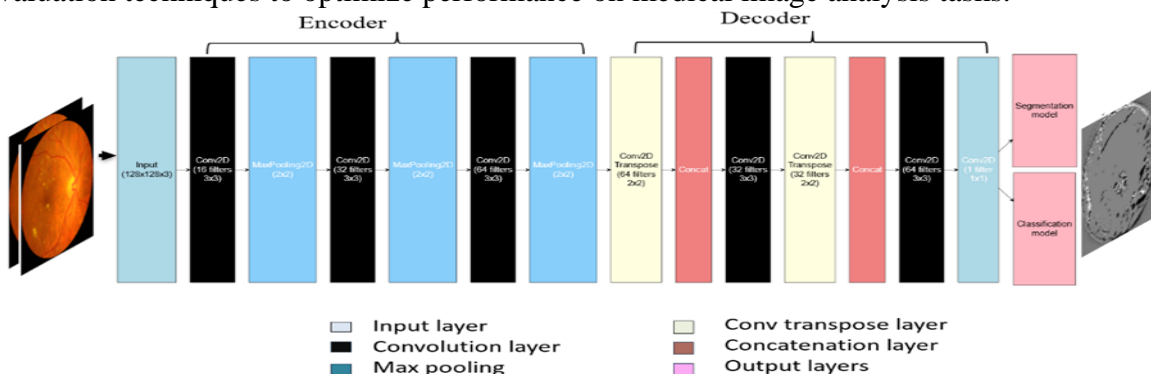


Figure.4: Residual encoder decoder architecture

RESULT ANALYSIS

The trained models were used to visualize segmented images and predicted classes for segmentation and classification tasks. Each new image from the IDRID dataset was processed by the segmentation model to generate masks highlighting regions of interest like exudates. These masks were overlaid on original images to assess segmentation accuracy visually. Meanwhile, the classification model assigned labels indicating severity levels to each image, enhancing context for the segmented regions. This demonstration illustrated the models' ability to produce interpretable visual results, with further quantitative assessments planned to gauge performance and potential for medical image analysis and diagnosis.

Table.1: Outputs showing the segmentation image and the predicted class using proposed architecture.

| Sno | Original image | Segmented image | predicted class |
|-----|----------------|-----------------|---------------------------------------|
| 1. | | | Predicted Class: proliferative DR |
| 2. | | | Predicted Class: severe DR |
| 3. | | | Predicted Class: moderate DR |
| 4. | | | Predicted Class: proliferative DR |

CONCLUSION

In conclusion, a comprehensive approach for automated diabetic retinopathy analysis using retinal images was implemented. It incorporates preprocessing techniques tailored to enhance image quality and relevance, including border removal, circular cropping, and histogram equalization, to optimize feature extraction for diabetic retinopathy detection. The proposed custom semantic segmentation and classification model, based on a residual encoder-decoder architecture, demonstrates



robust performance by accurately segmenting pathological regions like exudates and precisely classifying retinal conditions. Leveraging convolutional layers with max-pooling operations, transpose convolutions, and skip connections, this model effectively captures hierarchical features to improve segmentation accuracy. The model highlights the capability to produce meaningful visual outputs and achieve competitive performance metrics, showcasing its potential for advancing automated diagnostic tools in ophthalmology. This research contributes to the ongoing efforts in medical image analysis and underscores the importance of preprocessing techniques and deep learning architectures.

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