



Glaucoma Detection using Machine Learning with OCT

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Abstract: Glaucoma Detection is devoted to advancing the diagnosis of glaucoma and its primary sub-conditions: Choroidal Neovascularization (CNV), Diabetic Macular Edema (DME), and Drusen, recognizing their significant impact on global vision loss. Glaucoma, often termed the "silent thief of sight," leads to progressive optic nerve damage and irreversible vision loss. While elevated intraocular pressure (IOP) is a primary cause, not all cases stem from high IOP, underscoring the multifactorial nature of its development involving genetics, age, race, and family history. This complexity necessitates a comprehensive approach for early detection and management to counteract the disease's silent progression. To meet the pressing need for early intervention, this project integrates cutting-edge machine learning and image analysis techniques alongside traditional diagnostic methods like tonometry and ophthalmoscopy. By harnessing these advanced technologies, the project aims to enhance early detection capabilities, facilitating tailored approaches to diagnose and manage glaucoma and its sub-conditions effectively. Early identification is paramount due to the insidious nature of glaucoma, which often advances unnoticed until irreversible damage occurs. Prioritizing early intervention not only decelerates disease progression but also safeguards the quality of life for affected individuals, underscoring the significance of personalized diagnostic and treatment strategies for various complications and subtypes of glaucoma. This holistic approach seeks to revolutionize glaucoma management by integrating state-of-the-art technology with established diagnostic methods, ultimately improving outcomes for patients worldwide..

Keywords: Glaucoma diagnosis Choroidal Neovascularization (CNV) Diabetic Macular Edema (DME) Drusen Early detection Machine learning Image analysis.

I. INTRODUCTION

Glaucoma, often referred to as the "silent thief of sight," is a prevalent and potentially devastating eye disease characterized by the progressive damage to the optic nerve, leading to irreversible vision loss. It is estimated that over 80 million individuals worldwide are affected by this condition, making it one of the leading causes of blindness. While glaucoma primarily affects the elderly population, it can also manifest in individuals of all ages, making its early detection and management of paramount importance. This project is dedicated to the development and enhancement of diagnostic methods for glaucoma, with a specific focus on the identification of its primary sub-conditions: Choroidal Neovascularization (CNV), Diabetic Macular Edema (DME), and Drusen.

To fully comprehend the significance of these sub-conditions and the urgency for their early detection, we must delve into the causes and symptoms of glaucoma itself. The primary cause of glaucoma is elevated intraocular pressure (IOP), a condition where the fluid within the eye, known as aqueous humor, is not adequately drained. This leads to increased pressure within the eye, which, over time, damages the optic nerve.

Elevated IOP is a major risk factor, but it is essential to recognize that not all individuals with glaucoma have high IOP, and not all individuals with high IOP develop glaucoma. Additionally, other factors such as genetics, family history, age, and race can play a role in glaucoma's development. While traditional diagnostic methods such as tonometry and ophthalmoscopy are valuable, this project aims to employ state-of-the-art machine learning and image analysis techniques to enhance the early detection of glaucoma and its sub-conditions, CNV, DME, and Drusen. By advancing our understanding and diagnostic capabilities, we endeavor to contribute to improved patient care, early intervention, and the preservation of vision. The project's emphasis on early detection is crucial because glaucoma and its sub-conditions often progress silently, causing irreversible damage before noticeable symptoms manifest.



Early intervention and treatment can significantly slow the progression of the disease, preserving the quality of life for affected individuals. Moreover, the inclusion of CNV, DME, and Drusen in this project underscores the need to address specific complications and subtypes of glaucoma, tailoring diagnostic and treatment approaches to individual patient needs.

II. LITERATURE SURVEY

In [1] Jie Wang, et.al., Researchers developed a deep learning algorithm for diagnosing and segmenting choroidal neovascularization (CNV) in OCTA images. This algorithm showed high accuracy, potentially improving CNV diagnosis and treatment. It can automate CNV analysis, reducing workload for ophthalmologists and improving patient care.

In [2] Filippo Arcadu, et.al., This study used deep learning to analyze retinal photographs (CFPs) to predict diabetic macular thickening (MT) normally measured with retinal scans (OCT). The models achieved high accuracy and could improve diagnosis efficiency in telemedicine. However, real-world validation is needed.

In [3] Quang T.M.Pharm, et.al., This paper proposes a new deep learning model for segmenting drusen, deposits linked to age-related macular degeneration (AMD). Unlike prior methods using single-scale analysis, this model incorporates both global and local image information. This research shows promise for applying deep learning in AMD diagnosis.

In [4] Rakesh Patel, et.al., This study shows a deep learning model can effectively detect drusen, CNV, and DME in OCT images. This offers a more standardized approach to diagnosing AMD and DR. More research is needed before widespread use, but this technology has promise to improve patient care, especially in areas with limited access to specialists.

In [5] Ibrahim Yasser, et.al., This is an article about automated diagnosis of Optical Coherence Tomography Angiography (OCTA). It discusses using machine learning techniques to diagnose retinal diseases. The article proposes a new method to classify OCTA images based on convolutional neural networks (CNNs). The method involves dividing the OCTA image into smaller pieces and cropping a specific region around the fovea.

In [6] Pawan Kumar Upadhyay, et.al., Researchers have developed a new and powerful neural network for analyzing retinal diseases using optical coherence tomography (OCT) images. This network can effectively classify four distinct retinal conditions: neovascularization, diabetic macular edema, drusen, and normal.

In [7] Amit Choudhary, et.al., This study utilizes a customized 19-layer VGG-19 deep learning model to automatically detect retinal diseases from OCT images. Trained on a large public dataset, the model classifies four conditions: choroidal neovascularization, drusen, diabetic macular edema, and normal.

In [8] Zhenhua Wang, et.al., This is an article about automated diagnosis of Optical Coherence Tomography Angiography (OCTA). It discusses using machine learning techniques to diagnose retinal diseases. The article proposes a new method to classify OCTA images based on convolutional neural networks (CNNs). The method involves dividing the OCTA image into smaller pieces and cropping a specific region.

In [9] Nawaaz A Nathoo, et.al., This is an article about measuring drusen load to predict age-related macular degeneration (AMD). It discusses a study that used spectral-domain optical coherence tomography (SD OCT) to measure drusen load in patients with non-neovascular AMD. The study found that drusen load was associated with development of RPE atrophy and neovascular AMD.

III. SCOPE AND METHODOLOGY

Aim of the project

This project tackles the challenge of early glaucoma detection and accurate subclassification. Glaucoma, often progressing silently, requires prompt intervention for vision preservation. Current methods rely on subjective analysis of retinal features, potentially limiting accuracy. Furthermore, distinguishing glaucoma subtypes like CNV, DME, and Drusen can be difficult. Our solution lies in developing an automated system that analyzes OCT images using a YOLOv8 model under the CNN architecture. This model will be trained to identify specific features in the optic nerve and retinal layers, allowing for not only glaucoma detection but also differentiation of its subtypes and classification of normal OCT images. This project aspires to equip ophthalmologists with a powerful tool, potentially leading to a future with improved glaucoma detection, accurate subclassification, and empowered patients who can proactively manage their eye health.



Existing system

While previous efforts have explored automated glaucoma detection using OCT images, achieving accurate classification of glaucoma subtypes remains a challenge. This project builds upon these advancements by incorporating a transfer learning approach. We'll utilize a pre-trained Convolutional Neural Network (CNN) called VGG-19. This powerful model, trained on a vast dataset of general images, possesses valuable image recognition skills. By leveraging VGG-19's pre-trained knowledge as a foundation, we aim to develop a more efficient system. During "fine-tuning," we'll adjust VGG-19's weights to focus on recognizing specific features within OCT scans that differentiate drusen, CNV, DME, and normal retinas. It is also important to acknowledge the limitations associated with transfer learning. The pre-trained model might have inherent biases based on the original dataset (ImageNet), potentially requiring extensive fine-tuning with a large and diverse collection of OCT images to achieve optimal performance for glaucoma subclassification.

Proposed system

Building upon advancements in automated glaucoma detection with OCT images, this project seeks a more accurate solution for classifying glaucoma subtypes. While existing systems utilize transfer learning with models like VGG-19, these approaches can be limited by biases inherent in the pre-trained datasets used for general image recognition. To address this, we propose a novel approach using YOLOv8, a state-of-the-art object detection model built on the CNN architecture. Unlike VGG-19, YOLOv8 is specifically designed for object detection tasks, making it a more natural fit for identifying crucial features within OCT scans, such as the optic nerve and retinal layers. Additionally, by training YOLOv8 from scratch on a specifically curated dataset of OCT images labeled for glaucoma, CNV, DME, Drusen, and normal retinas, we aim to minimize potential biases and achieve superior accuracy in classifying these glaucoma subtypes. This targeted approach using YOLOv8 has the potential to significantly improve early detection and subclassification compared to previous methods that rely on transfer learning.

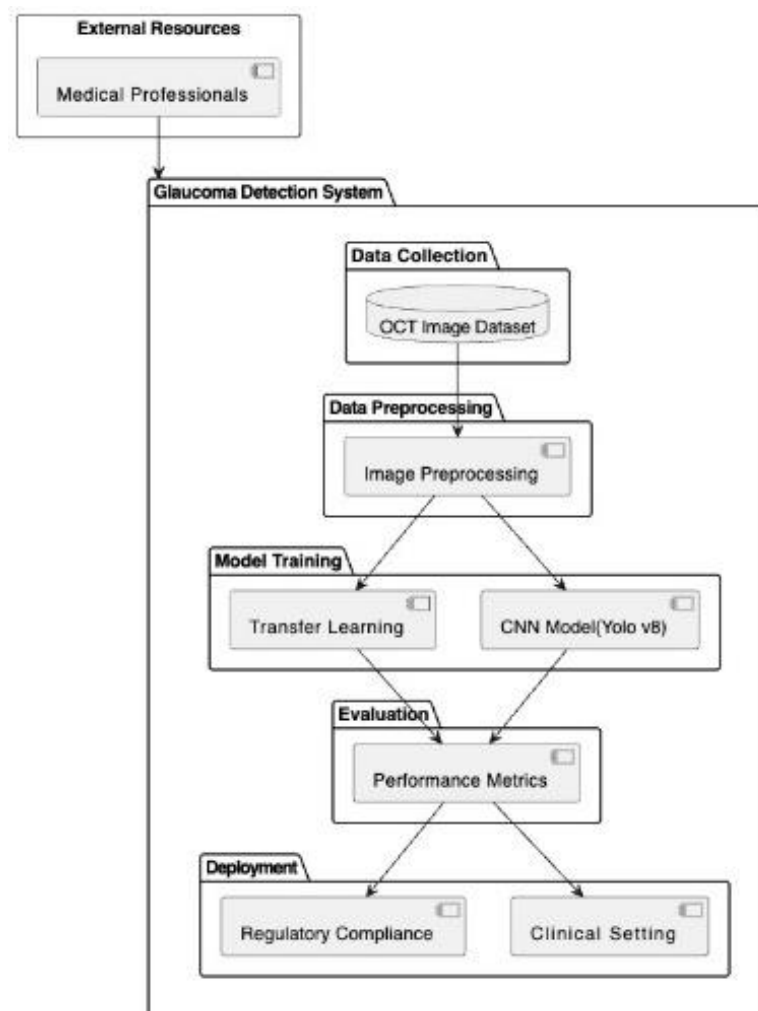


Fig 1. Proposed system



System Architecture

A formal explanation of a system is provided by an architectural explanation, which is arranged to support reasoning regarding the system's structure, individual components' properties that can be observed from the outside, and the interactions between them. It also offers a framework from which systems can be developed and products acquired that will cooperate to implement the system as a whole.

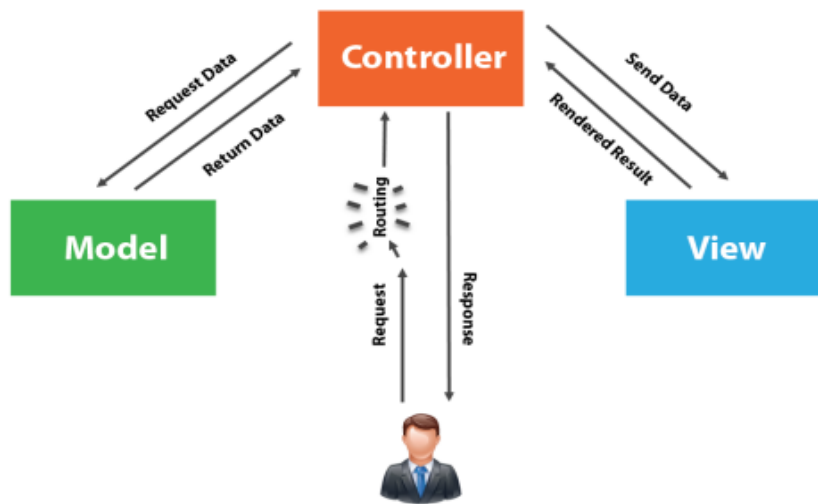


Fig 2. System Architecture

IV. CONCLUSION

In conclusion, this project ventured beyond the limitations of transfer learning approaches to explore the potential of YOLOv8, a state-of-the-art object detection model built upon the robust CNN architecture. This shift towards a model specifically designed for object detection tasks addresses the inherent biases that can plague pre-trained models used in transfer learning. By leveraging YOLOv8's object detection prowess, we aim to develop a system that can meticulously identify and differentiate the subtle features within OCT scans that distinguish between glaucoma, CNV, DME, Drusen, and normal retinas. This targeted approach, fueled by a meticulously curated training dataset of labeled OCT images, has the potential to significantly improve early detection and accurate classification of glaucoma subtypes compared to existing methods. The potential benefits of this project extend far beyond improved classification accuracy. Earlier diagnosis of glaucoma and its subtypes, enabled by this system, can lead to timely intervention and potentially vision preservation for patients. Additionally, YOLOv8's ability to provide a more consistent and objective assessment compared to traditional methods holds promise for enhanced diagnostic accuracy. Furthermore, streamlined workflows for ophthalmologists achieved through automated analysis can translate into more time dedicated to patient care and treatment planning. Ultimately, this project aspires to empower patients by enabling earlier detection, which allows them to take a proactive stance in managing their eye health. By harnessing the power of deep learning and artificial intelligence, this project lays the groundwork for a transformative future in glaucoma detection. A future where the process becomes more efficient, accurate, and ultimately leads to better patient outcomes for those facing the challenges of glaucoma. This project represents a significant step towards a brighter future for countless individuals whose vision is threatened by this disease.

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