



A COMPREHENSIVE FRAMEWORK FOR MULTI-DISEASE PREDICTION THROUGH RETINAL IMAGES

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ABSTRACT

This paper aims to develop a basic framework for identifying illnesses such as diabetes and heart attacks using retinal fundus images. The fundus picture shows two types of blood vessels: arteries and veins, which are closely associated with brain and heart function. This makes it crucial to distinguish between them for diagnosing different disorders. After preprocessing, the images undergo Convolutional Neural Network (CNN) analysis to recognize essential features and discriminate between arteries and veins. By analyzing retinal images, differences related to various disorders can be observed. Diseases like heart attacks and diabetes are leading causes of mortality, highlighting the importance of early detection. This paper leverages the retinal eye's significance due to its connection to retinal nerves that lead to the brain. The ResNet-18 architecture is employed for illness identification using retinal images.

Keywords: ResNet-18 architecture, Convolutional Neural Network, Deep Learning

INTRODUCTION

This paper presents a unique method for predicting multiple diseases using retinal images, providing a non-invasive and efficient approach for early detection and diagnosis of various ocular and systemic conditions. Retinal imaging has become invaluable in healthcare, offering detailed insights into structural and vascular changes related to diverse medical issues. By employing advanced image processing and deep learning techniques, this study aims to analyze retinal images and forecast diseases such as diabetic retinopathy and heart attacks. Detecting multiple diseases from a single imaging modality offers advantages in accessibility, cost-effectiveness, and patient convenience. The proposed framework includes data preprocessing, feature extraction, CNN-based classification, and user-friendly interfaces for healthcare providers. Through rigorous experimental validation, this approach demonstrates promising results in disease identification and classification from retinal images, contributing significantly to early disease detection and personalized healthcare, thereby improving patient outcomes and quality of life.

LITERATURE SURVEY

Reference: [1] Carlos Hernandez-Matasa, Antonis A. Argyros, Xenophon Zabulisa, "Retinal Image Preprocessing, Enhancement, and Registration," IEEE.

Carlos Hernandez-Matasa, Antonis A. Argyros, and Xenophon Zabulisa explore the historical evolution of retinal imaging and its role in medical diagnostics in their work "Retinal Image Preprocessing, Enhancement, and Registration" [1].

Fundus imaging originates from the invention of the ophthalmoscope, marking the start of storing and analyzing retinal images for diagnostics. Early efforts focused on analogue images and vessel detection using fluorescein angiography, which, although improving vessel visibility, was invasive, time-consuming, and costly due to the required fluorescent agents.

The shift to digital imaging and processing revolutionized retinal analysis, making it noninvasive and efficient for screening and diagnosis. Techniques such as fundus photography, optical coherence



tomography (OCT), and scanning laser ophthalmoscopy (SLO) expanded the scope of retinal image analysis in healthcare.

This review discusses the evolution of fundus imaging techniques and emphasizes digital advancements' role in enhancing accuracy and accessibility for medical professionals and computer-based systems.

Reference: [2] Umamageswari, J. Shiny Duel, K. Raja, "Identifying Diabetic Retinopathy using Deep Learning based Classification"

Umamageswari, J. Shiny Duel, and K. Raja investigate deep learning for Diabetic Retinopathy (DR) identification in their study "Identifying Diabetic Retinopathy using Deep Learning based Classification" [2]. Their research focuses on efficiently detecting exudates and vessels from retinal images to analyze retinal vasculature disorders and aid in early-stage diabetes detection, surpassing traditional methods like retinal blood vessel analysis.

Reference: [3] S.N Shivapriya, M. Ramya, D. Prasanth, "Disease Prediction based on Retinal Images"

S.N Shivapriya, M. Ramya, D. Prasanth's study, "Disease Prediction based on Retinal Images" [3], aims to develop a framework for identifying disorders like diabetes, hypertension, and heart attacks using retinal images. Retinal fundus images reveal arteries and veins crucial for diagnosing brain and heart-related

conditions. A median filter algorithm is used to classify vessels into arteries and veins, followed by a Convolutional Neural Network (CNN) to identify key features distinguishing between different diseases. This approach detects variations in retinal images for conditions such as heart attacks and diabetes, assisting in disease prediction by leveraging the significance of the retinal eye connected to the brain. The VGG16 architecture is employed for disease identification using retinal images.

Reference: [4] Lalitha Krishnasamy, Rajesh Kumar Dhanaraj, Monika Gupta. "Detection of Diabetic Retinopathy using Retinal Fundus Images." Presented at IEEE 2022.

Diabetic retinopathy (DR) is a complication of diabetes affecting the retina due to blood vessel damage, often leading to vision issues. Detecting DR involves identifying Hard Exudate in retinal images. While deep neural networks offer strength in image tasks, adding layers can reduce accuracy and increase complexity. Introducing Recurrent Neural Networks (RNNs) overcomes this, as they handle data sequences efficiently, ensuring faster training, stable model size, and improved accuracy, making them preferable for DR detection.

EXISTING METHOD

Existing methods for multi-disease prediction include traditional machine learning algorithms and simple image processing techniques. These methods often involve feature extraction and classification using Support Vector Machines (SVM). However, SVMs may struggle with complex image data due to linearity. Deep learning models like VGG16 offer a deep architecture for feature extraction but can be computationally expensive and prone to overfitting. In contrast, ResNet18's advantage lies in its residual learning approach, enabling easier training of deeper networks and mitigating the vanishing gradient problem.

PROPOSED METHOD

The proposed method for multi-disease prediction uses retinal imaging and the ResNet18 architecture to identify diseases like diabetes, hypertension, and heart attacks from retinal fundus images. It starts by preprocessing images to extract key features indicative of different disorders. These features are then input into a CNN model trained to recognize patterns and abnormalities related to various health conditions.



Leveraging deep learning and medical imaging, the paper aims to create a robust framework for early disease detection and healthcare diagnostics.

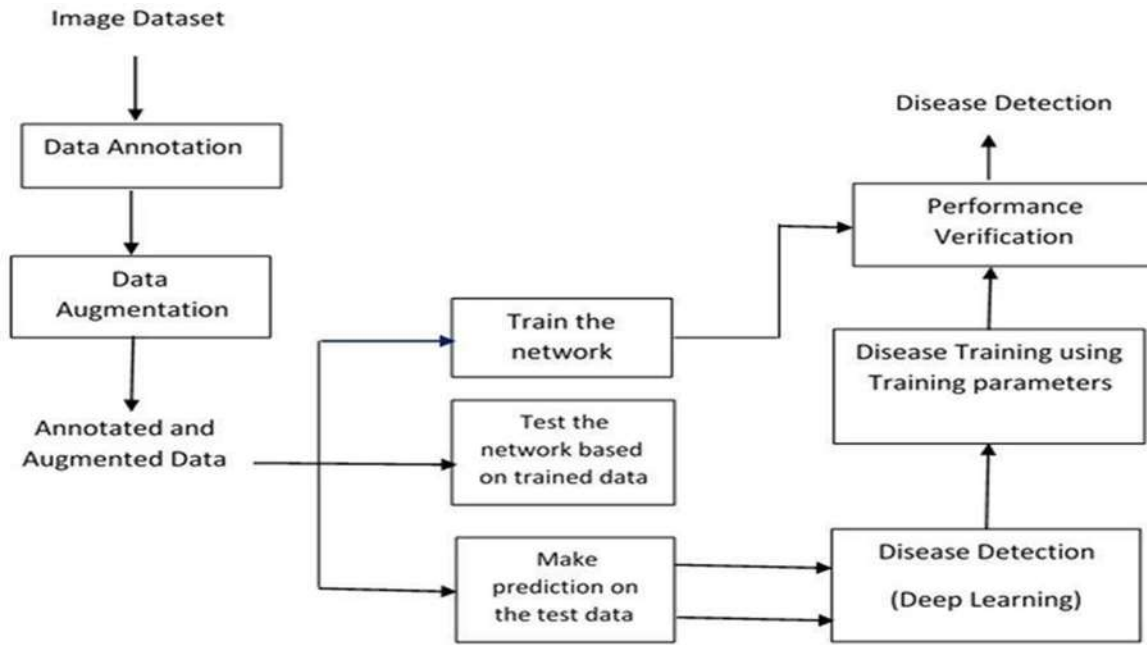
1. **Comprehensive Disease Prediction:** Our method accurately predicts diabetes, hypertension, and heart attacks using retinal imaging and CNNs.
2. **Non-Invasive Diagnostics:** By leveraging retinal imaging, our approach offers non-invasive disease detection.
3. **Advanced Deep Learning:** Integration of ResNet18 enhances feature extraction and prediction accuracy.
4. **Scalable Efficiency:** The system efficiently handles large volumes of images for potential clinical deployment.
5. **Early Intervention Potential:** Early prediction enables proactive healthcare measures, improving outcomes and reducing costs.

DESIGN STRUCTURE

The paper's design structure emphasizes a systematic approach to multi-disease prediction leveraging retinal images and heart health data. It includes robust data preprocessing pipelines to ensure data quality and consistency. The core architecture involves the implementation of ResNet-18 and related models, customized for accurate disease risk assessments. Model training, evaluation, and validation processes are meticulously designed to guarantee model reliability and predictive accuracy. Integration into a user-friendly deployment platform and comprehensive testing strategies further solidify the paper's design, aiming for a scalable and effective solution for healthcare diagnostics.

1. **Data Processing:** Develop robust pipelines for preprocessing retinal images and heart health data, incorporating loading, augmentation, and normalization methods.
2. **Model Architecture:** Employ ResNet-18 and other deep learning architectures for efficient feature extraction and precise prediction of diabetic retinopathy and heart attack risks, ensuring accuracy and interpretability.
3. **Training and Validation:** Train models using Adam optimization, validate using accuracy and loss metrics, and evaluate comprehensively on separate test datasets to ensure reliability and generalization.
4. **Integration and Deployment:** Create an integrated pipeline for combining predictions from both models, offering a unified interface for multi-disease risk assessment. Deploy using Streamlit for user-friendly access.
5. **Testing and Validation:** Conduct thorough unit testing, integration testing, and system validation to guarantee scalability, accuracy, and reliability in predicting multiple diseases simultaneously.

SYSTEM BLOCK DIAGRAM



RESULT ANALYSIS

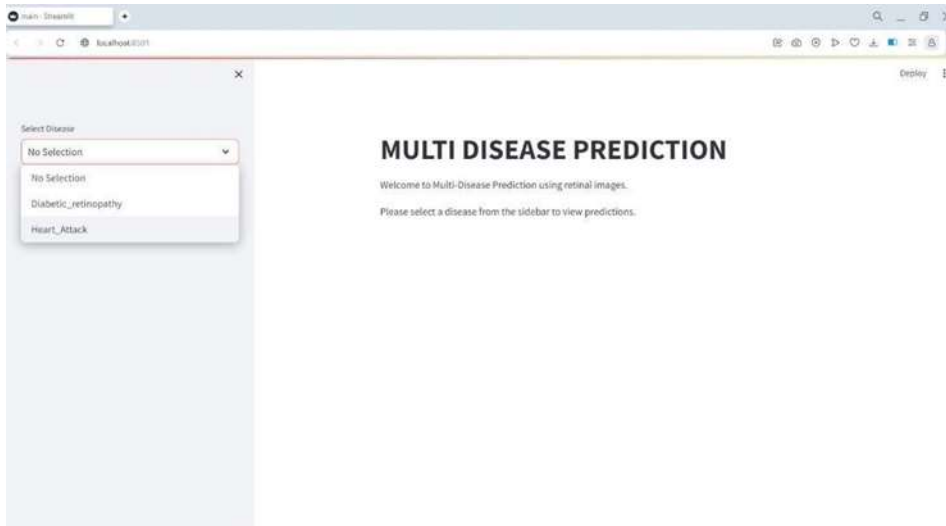


Fig 1: Select the Disease name for diagnosis

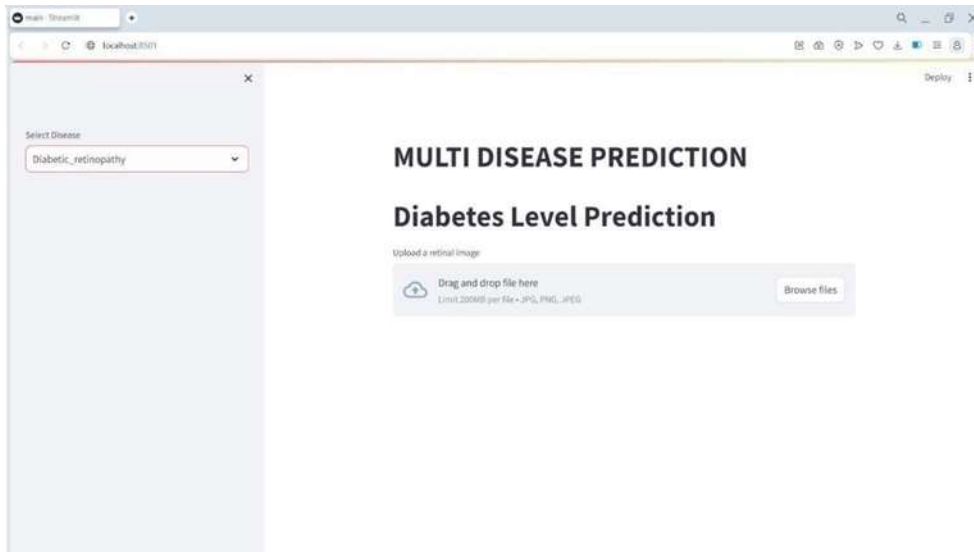


Fig 2: Input the image

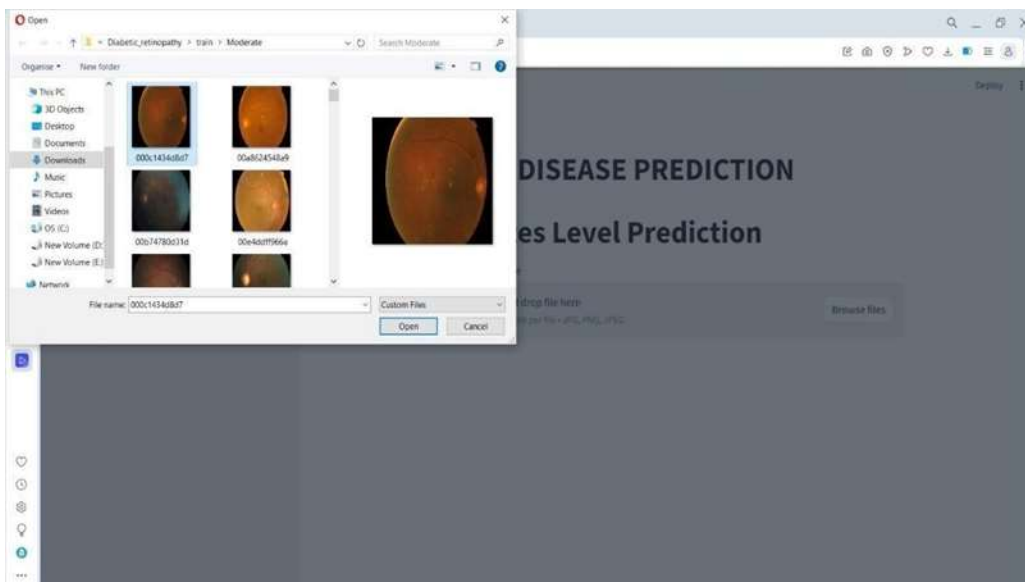


Fig 3: Upload the Fundus Image

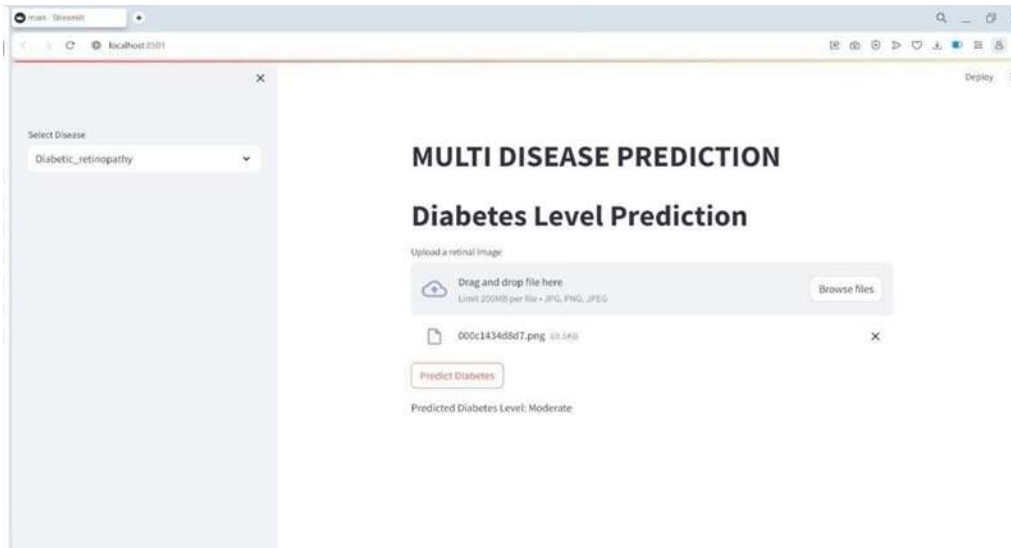


Fig 4: The level of Diabetics is Predicted

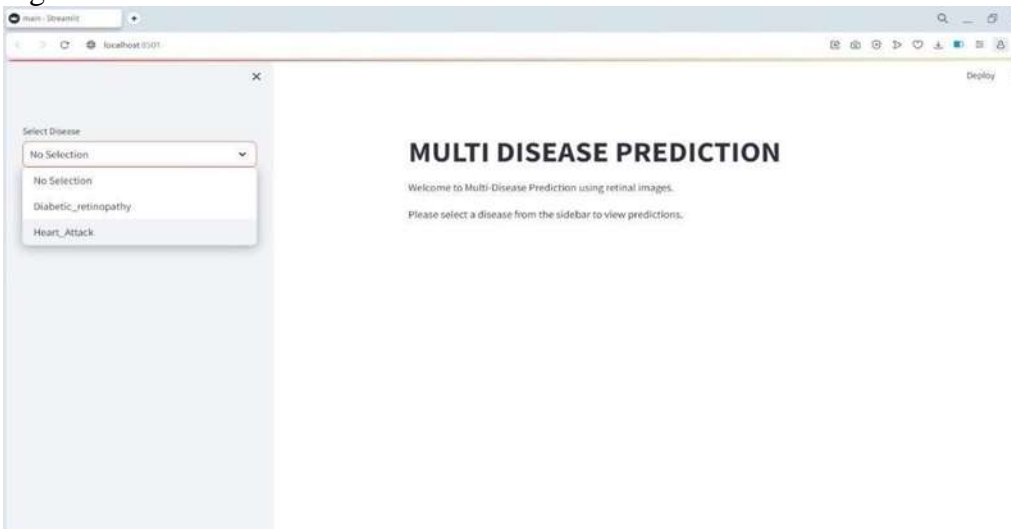


Fig 5: Select the Other Disease for Diagnosis

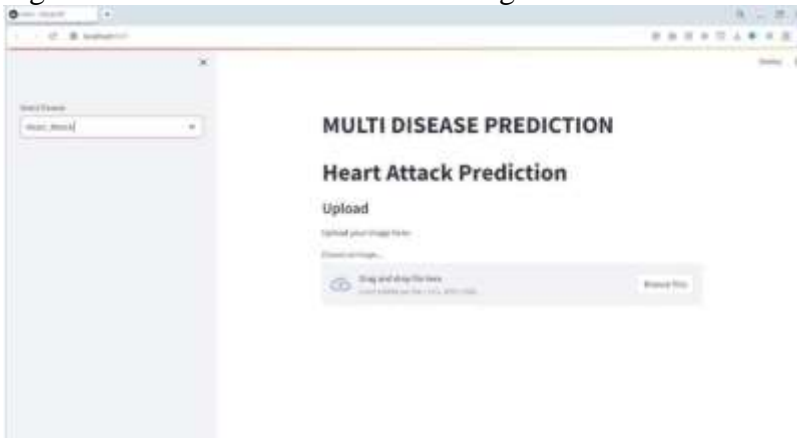


Fig 6: Input the Image

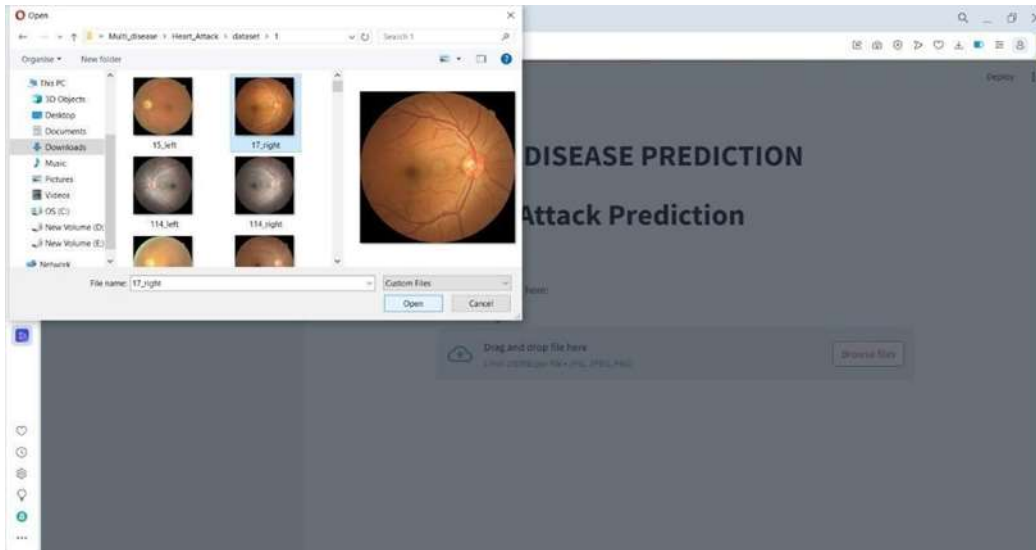


Fig 7: Upload the Fundus Image

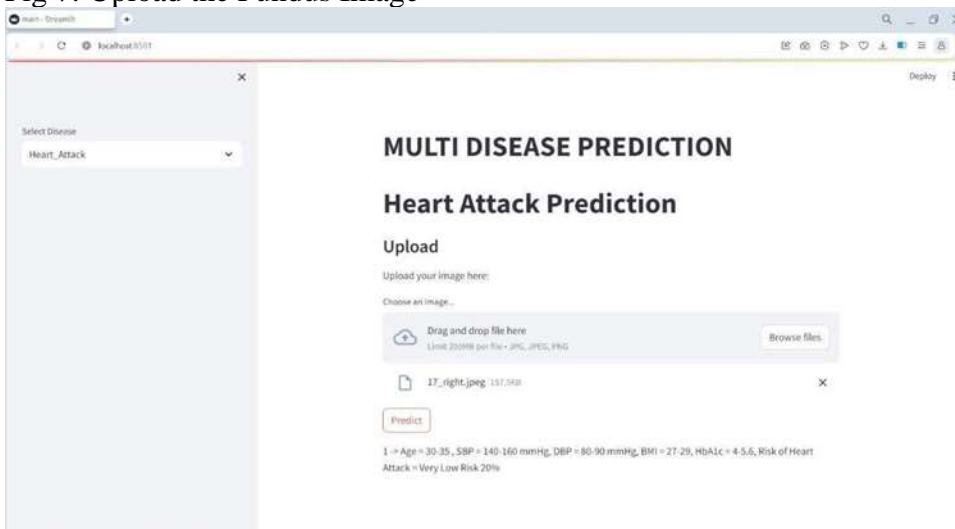


Fig 8: The chances of a Heart Attack is predicted

CONCLUSION

In conclusion, the multi-disease prediction paper has effectively met its primary goal of creating predictive models for heart attack and diabetic retinopathy. The paper encompassed several phases, including data preprocessing, designing model architecture, training, evaluation, and deployment using contemporary technologies and machine learning methodologies. The heart attack prediction model utilized a dataset containing cardiovascular risk factors to construct a classification model capable of forecasting the likelihood of a heart attack. Conversely, the diabetic retinopathy prediction model utilized retinal images and deep learning algorithms to categorize various levels of diabetic retinopathy severity. Thorough testing and validation procedures were conducted, demonstrating promising accuracy and reliability in the models' predictions. The deployment of the models as a web application via Streamlit highlighted their practical usability in real-world contexts.



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