## DEEP LEARNING FUNDUS IMAGE ANALYSIS FOR EARLY DETECTION OF DIABETIC RETINOPATHY

## LITERATURE SURVEY:

S.NO	AUTHORS	TITLE	ABSTRACT
1.	N. Memari, S. Abdollahi,	Computer-assisted	The proposed computer-
	M. M. Ganzagh and M.	diagnosis (CAD)	assisted diagnosis
	Moghbel	system for Diabetic	system starts with the
		Retinopathy screening	segmentation of the
		using colour fundus	blood vessels. Then,
		images using Deep	microaneurysms and
		learning	exudates are
			segmentation from the
			image. Statistical and
			regional features are
			then extracted utilising
			first, second, and higher-
			order image features. A
			Deep Learning
			framework will be
			utilised for extracting
			additional statistical
			image descriptors as
			Deep Learning has
			superior contextual
			analysis capabilities
			compared to other
			machine learning
_			techniques.
2.	A. Bali and V. Mansotra	Deep Learning-based	In this paper different
		Techniques for the	deep learning (DL)
		Automatic	techniques for automatic
		Classification of	classification of fundus
		Fundus Images: A	images have been
		Comparative Study	discussed and results are
			compared on the basis of
			accuracy, f1-score and
			AUC.

3.	M. Z. Atwany, A. H. Sahyoun and M. Yaqub	Deep Learning Techniques for Diabetic Retinopathy Classification: A Survey	This paper reviews and analyses state-of-the-art deep learning methods in supervised, self-supervised, and Vision Transformer setups, proposing retinal fundus image classification and detection. The paper discusses the available retinal fundus datasets for Diabetic Retinopathy that are used for tasks such as detection, classification, and segmentation. The paper also assesses research gaps in the area of DR detention/classification and addresses various challenges that need further study and investigation.

4.	S. Gupta, A. Panwar, A. Kapruwan, N. Chaube and M. Chauhan	Real Time Analysis of Diabetic Retinopathy Lesions by Employing Deep Learning and Machine Learning Algorithms using Colour Fundus Data	The colour fundus dataset scans after processing are passed to multiple Deep Learning (DL) models employed to learn characteristics. These models trained on millions of different images from thousands of classes. Finally, several machine learning classifiers were used to classify lesions using the collected characteristics. The extracted result shows very eye-catching performance. This enables experts to create architecture that fully addresses the problem of classifying unidentified scans into the right class or category.
5.	H. Kaushik, D. Singh, M. Kaur, H. Alshazly, A. Zaguia and H. Hamam.	Diabetic Retinopathy Diagnosis from Fundus Images Using Stacked Generalisation of Deep Models	In this research, a methodology to eliminate these unnecessary reflectance properties of the images using a novel image processing schema and a stacked deep learning technique for the diagnosis. For the luminosity normalisation of the image, the grey world odour constancy algorithm is implemented which does image desaturation and improves the overall image quality.

6.	B. Bulut, V. Kalin, B. B. Güneş and R. Khazin	Deep Learning Approach for Detection of Retinal Abnormalities Based on Colour Fundus Images	This research uses the Xception model with transfer learning method to classify image obtained from Akdeniz University Hospital Eye Diseases Department. During the analysis, the Xception model containing 50 different parameter combinations was trained by scanning the appropriate hyperparameter space for the model.
7.	B. Gowtham, M.F. Hashmi, Z. W. Geem and N. D. Bokde	A Comprehensive Review of Deep Learning Strategies in Retinal Disease Diagnosis Using Fundus Images	This article presents a comprehensive study of different deep learning strategies employed in recent times for the diagnosis of five major eye diseases, i.e., Diabetic retinopathy, Glaucoma, age-related macular degeneration, Cataract, and Retinopathy of prematurity.
8.	T. A. Soomro et al	Deep Learning Models for Retinal Blood Vessels Segmentation: A Review	This paper presents a comprehensive review of the principle and application of deep learning in retinal image analysis. This paper characterises each deep learning-based segmentation method as described. Analysing along with the limitations and advantages of each method. In the end, we offer some recommendations for future improvement for retinal image analysis.

09.	W. Zhang, X. Zhoo, Y.	Deep UWF: An	The emerging ultra-
0).	Chen, J. Zhong and Z. Yi	Automated Ultra-Wide-	wide-field of view
	Chen, J. Zhong and Z. 11	Field Fundus Screening	(UWF) fundus colour
		System via Deep	` ·
		Learning Learning	imaging is a powerful tool for fundus
		Learning	
			screening. However,
			manual screening is
			labour-intensive and
			subjective. Based on
			2644 UWF images, a set
			of early fundus abnormal
			screening systems
			named DeepUWF is
			developed. The
			experimental results
			show that these
			preprocessing methods
			are helpful to improve
			the learning ability of the
			networks and achieve
			good sensitivity and
			specificity. Without
			ophthalmologists,
			DeepUWF has potential
			application value, which
			is helpful for fundus
			health screening and
			workflow improvement.
10.	H. Yeh, CJ. Lin, CC.	Deep-learning based	Deep learning is used in
	Hsu and CY. Lee	Automated	many types of
		Segmentation of	preprocessing for
		Diabetic Retinopathy	segmentation. We
		Symptoms	preprocessed fundus
			images and inputted
			them into the model for
			training. Finally, LDF
			image was used to
			obtain the best
			preprocessing method
			for optic disc
			segmentation in fundus
			images.
			mages.

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