

"Glaucoma Stages Detection Using Fundus Images Through Deep Learning"*Synopsis*

*Submitted in partial fulfillment of the requirements for the award of the degree
of*

BACHELOR OF TECHNOLOGY**IN****DEPARTMENT OF CSE-ARTIFICIAL INTELLIGENCE**

Submitted by

Batch 20JR-07

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Assistant Professor – AI Department

**DEPARTMENT OF CSE- ARTIFICIAL INTELLIGENCE****KKR & KSR INSTITUTE OF TECHNOLOGY AND SCIENCES****(Autonomous)**

Approved by A.I.C.T.E New Delhi || Permanently Affiliated to JNTUK, Kakinada) || Accredited with 'A' Grade by NAAC

|| NBA Accreditation status for 5 B. Tech Programmes (Civil,CSE, ECE, EEE & Mech)

Vinjanampadu (Vil), Vatticherukuru (Md), Guntur (DT), A.P-522017.

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CERTIFICATE

This is to certify that this Major project titled "*Glaucoma Stages Detection Using Fundus Images Through Deep Learning*" is done by **Ms.Chennam HemaSirisha (20JR1A4302),Ms.Lavu SumaSri(20JR1A4317),Ms.Thokala MiniRose (20JR1A4328)** in the duration of January to April 2024, who carried out the work under my supervision and submitted in the partial fulfillment of the requirements for the award of the degree of **Bachelor of Technology** in Computer Science & Engineering-Artificial Intelligence from **JNTU-Kakinada**.

INTERNAL EXAMINER

HEAD OF THE DEPARTMENT

EXTERNAL EXAMINER

DECLARATION

We here by inform that this major project titled "*Glaucoma Stages Detection Using Fundus Images Through Deep Learning*" has been carried out by myself in the duration of January to April 2024 and submitted in partial fulfillment for the award to the degree of **Bachelor of Technology in Computer Science and Engineering-Artificial Intelligence** to **Jawaharlal Nehru Technological University Kakinada** under the guidance of **Ms. K. Radhika**, Assistant Professor, Dept. of Computer Science and Engineering-Artificial Intelligence.

20JR1A4302**20JR1A4317****20JR1A4328**

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We are very much thankful to the **college management** for their continuous support and facilities provided.

We render our deep sense of gratitude to **Dr. P. BABU, Principal & Dr K Hari Babu, Academic Director**, for permitting us to carry out major project. We would like to express our sincere thanks to Computer Science and Engineering-Artificial Intelligence faculty for lending us their time to help us and complete the work successfully.

We would also like to thank our staff, parents and friends for their enduring encouragement and assistance whenever required.

20JR1A4302

20JR1A4317

20JR1A4328

Institute Vision and Mission

INSTITUTION VISION

To produce eminent and ethical Engineers and Managers for society by imparting quality professional education with emphasis on human values and holistic excellence.

INSTITUTION MISSION

- To incorporate benchmarked teaching and learning pedagogies in curriculum.
- To ensure all round development of students through judicious blend of curricular, co-curricular and extra-curricular activities.
- To support cross-cultural exchange of knowledge between industry and academy.
- To provide higher/continued education and research opportunities to the employees of the institution.

Department of CSE-Artificial Intelligence

Vision of the Department

- To be a renowned department for education in Artificial Intelligence and empowering students into professional engineers with human values and holistic excellence.

Mission of the Department

- Impart rigorous training to acquire knowledge through the state-of-the-art concepts and technologies in Artificial Intelligence.
- Train students to be technically competent through innovation and leadership.
- Inculcate values of professional ethics, social concerns, life-long learning and environment protection.
- Establish centers of excellence in leading areas of computing and artificial intelligence.

Program Specific Outcomes (PSOs)

PSO1: Application Development

Apply the concepts in core area of Artificial Intelligence, Data Structure, Database System, Operating System, Networking and Intelligence System to solve futuristic problems.

PSO2: Computing Paradigms

Develop automated solutions for real world problems through laboratory experiments, projects and internship.

Program Educational Objectives (PEOs)

PEO:1	Graduates of Computer Science and Engineering – Artificial Intelligence shall apply appropriate theory, practices, and tools to provide solution for multidisciplinary challenges.
PEO:2	Graduates of Computer Science and Engineering - Artificial Intelligence shall have an ability to function effectively in the workplace for professional growth.
PEO:3	Graduates of Computer Science and Engineering shall have exposure to adapt, contribute and innovate new technologies in the key domains of Artificial Intelligence during higher studies or product development.

Program Outcomes

- 1. Engineering knowledge:** Apply the knowledge of mathematics, science, engineering fundamentals, and an engineering specialization to the solution of complex engineering problems.
- 2. Problem analysis:** Identify, formulate, research literature, and analyze complex engineering problems reaching substantiated conclusions using first principles of mathematics, natural sciences, and engineering sciences.
- 3. Design/development of solutions:** Design solutions for complex engineering problems and design system components or processes that meet the specified needs with appropriate consideration for the public health and safety, and the cultural, societal, and environmental considerations.
- 4. Conduct investigations of complex problems:** Use research-based knowledge and research methods including design of experiments, analysis and interpretation of data, and synthesis of the information to provide valid conclusions.
- 5. Modern tool usage:** Create, select, and apply appropriate techniques, resources, and modern engineering and IT tools including prediction and modeling to complex engineering activities with an understanding of the limitations.
- 6. The engineer and society:** Apply reasoning informed by the contextual knowledge to assess societal, health, safety, legal and cultural issues and the consequent responsibilities relevant to the professional engineering practice.
- 7. Environment and sustainability:** Understand the impact of the professional engineering solutions in societal and environmental contexts, and demonstrate the knowledge of, and need for sustainable development.

8. Ethics: Apply ethical principles and commit to professional ethics and responsibilities and norms of the engineering practice.

9. Individual and team work: Function effectively as an individual, and as a member or leader in diverse teams, and in multidisciplinary settings.

10. Communication: Communicate effectively on complex engineering activities with the engineering community and with society at large, such as, being able to comprehend and write effective reports and design documentation, make effective presentations, and give and receive clear instructions.

11. Project management and finance: Demonstrate knowledge and understanding of the engineering and management principles and apply these to one's own work, as a member and leader in a team, to manage projects and in multidisciplinary environments.

12. Life-long learning: Recognize the need for, and have the preparation and ability to engage in independent and life-long learning in the broadest context of technological change.

Project Course Outcomes

CO421.1:Analyze the System of Examinations and identify the problem.

CO421.2:Identify and classify the requirements.

CO421.3:Review the Related Literature

CO421.4:Design and Modularize the project

CO421.5:Construct, Integrate, Test and Implement the Project.

CO421.6:Prepare the project Documentation and present the Report using appropriate method.

Course Outcomes – Program Outcomes mapping

	1	2	3	4	5	6	7	8	9	10	11	12	PSO1	PSO2
CO421.1	3	2					2						2	
CO421.2		3		1		2					2			2
CO421.3						3	2	1	2					
CO421.4			3		2				2				3	
CO421.5				2			3				1		2	1
CO421.6									3	2		2		3
Average	3	2.5	3	1.5	2	2.5	2.33	1	2.33	2	1.5	2	2.33	2

3: High 2: Medium 1: Low

CO – PO Mapping with Reasons:

1.CO421.1 is mapped with PO1, PO2 and PO7 as basic knowledge of Engineering and problem Analysis activities are highly essential to conduct examinations on existing systems which have been using in industries as part of and to define the problem of proposed system.

2. CO421.2 is mapped with PO2, PO4, PO6 and PO11 as for identification, gathering, analysis and classification of requirements for the proposed system, basic knowledge of engineering, and Analysis steps along with complex problem analysis skills to meet the specific needs of the customer.

3.CO421.3 is mapped with PO6, PO7, PO8, and PO9 as to conduct the literature review and to examine the relevant systems to understand and identify the merits and demerits of each to enhance and develop the proposed as per the need.

4.CO421.4 is mapped with PO3, PO5 and PO9 because modularization and design of the project is needed after requirements elicitation. For Modularization and design of the project, Design skills, Modern tool usage and communication is needed between team members as different modules are designed individually before integration.

5.CO421.5 is mapped with PO5, PO8, PO10, PO11 and PO12 as to construct the project latest technologies are needed. The development of project is done individually and in groups with well defined communication by using the engineering and management principles.

6.CO421.6 is mapped with PO4, PO7 and PO11 because during and after completion of the project, documentation is needed along with proper methods of presentation through understanding and application of Engineering and management principles, which in turn needs well defined communication between the team members with all the ethical values. Even The project development team defines the future enhancements as a part of project development after identifying the scope of the project.

CO-PSOs Mapping with Reasons:

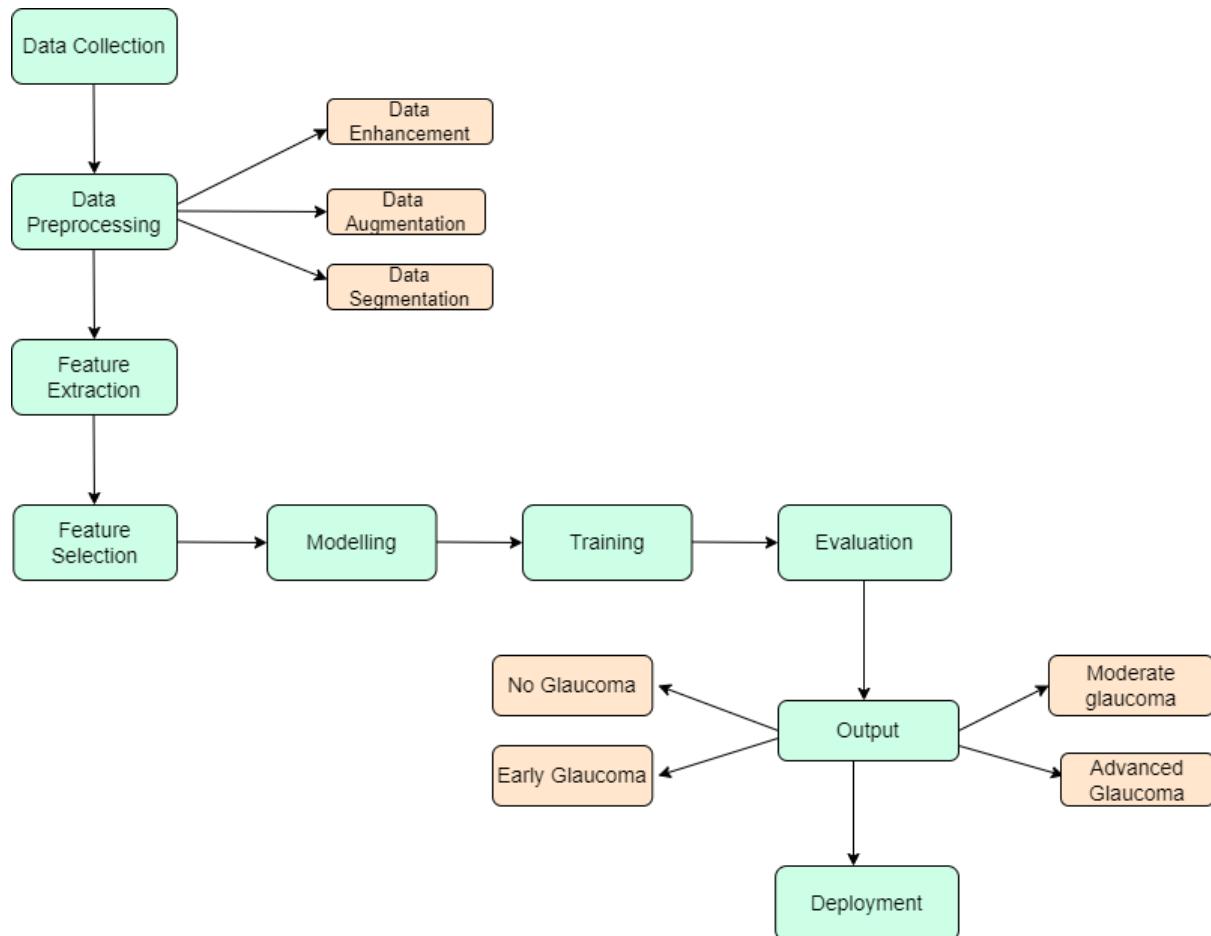
1. **CO421.1** is mapped with **PSO1** as examining of existing systems and identification of the problem is a part of Application Development Activity and identification of evolutionary changes in latest technologies.
2. **CO421.2** is mapped with **PSO2** as review of literature is a part of Application development activity with the practice of professional and leadership qualities.
3. **CO421.4** is mapped with **PSO1** because modularization and logical design is also a part of Application Development.
4. **CO421.5** is mapped with **PSO1 and PSO2** as Development, Testing and Integration of project activities are part of Application Development.
5. **CO421.6** is mapped with **PSO2** as for project documentation and presentation the project members apply the professional and Leadership qualities.

ABSTRACT

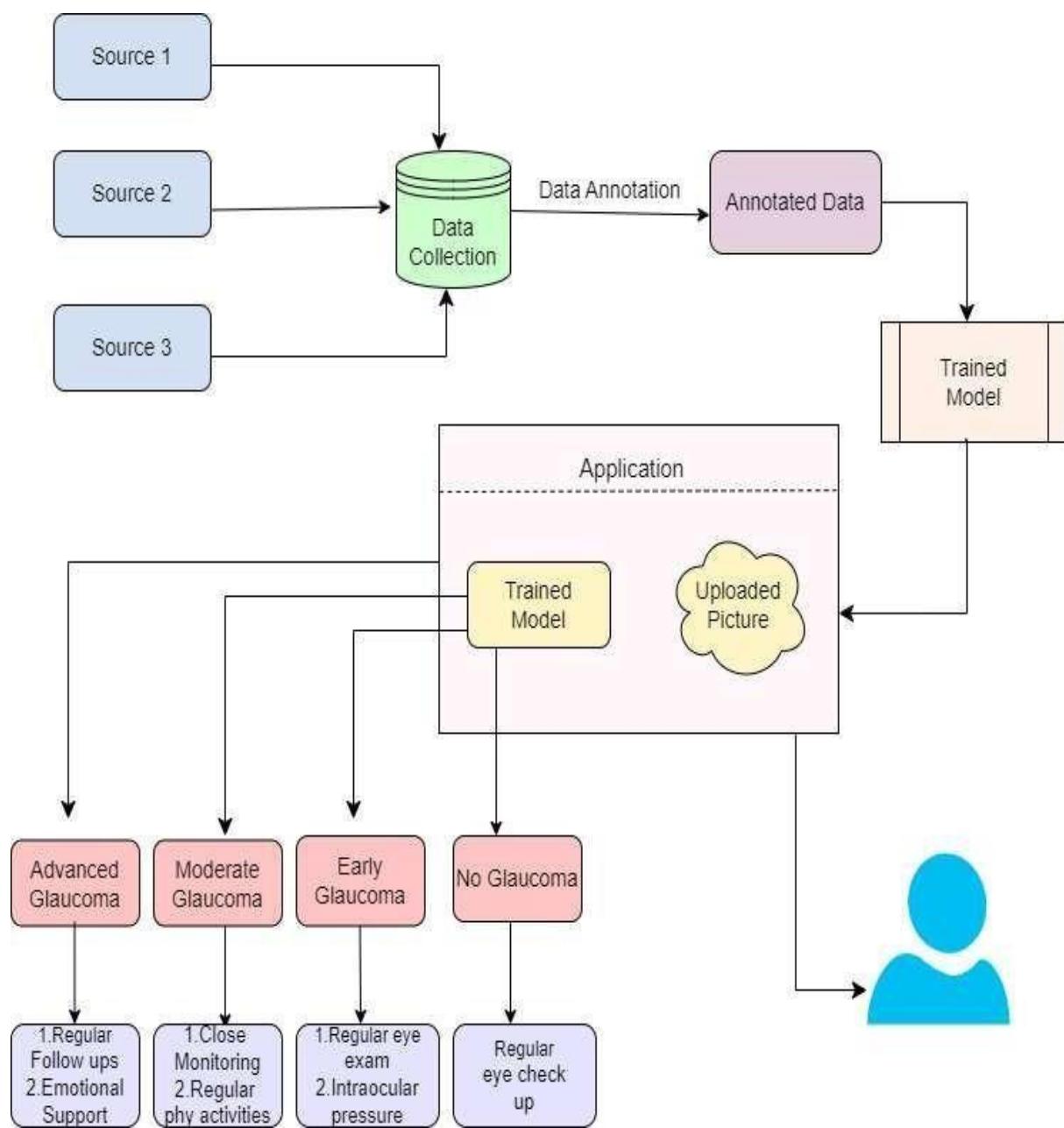
A chronic eye condition called glaucoma has a deleterious effect on the optical nerve, which links the brain and eye to transmit visual information. Early detection is essential for stopping the condition's progression. Glaucoma is one of the most prevalent eye conditions, and it's important to catch it early because it can cause blindness and neurological issues. In this study, a Deep Learning system is proposed for the early detection of glaucoma. The eye image undergoes preprocessing to eliminate any noise and prepare them for further analysis. The system utilizes enlarged images of the eyes as input data for the deep learning method. The suggested system classifies new eye images as No_glucoma, Glaucoma_early, Glaucoma_moderate, and Glaucoma_advanced and also provides respective preventive measures based on the features it learned during training.

Glaucoma, One of the leading causes of blindness worldwide is glaucoma, a longterm neurodegenerative eye disease. According to the WHO, an average of 65 million people around the world are affected by glaucoma. Given that the primary symptom of glaucoma, the loss of optic nerve fibers, may be asymptomatic, early diagnosis and treatment are crucial in preventing vision loss. This loss is caused by increased intracranial pressure or decreased blood flow into the optic nerve. High pressure in glaucoma is caused by increased reluctance to fluid expulsion into the drainage system of the eye. The fluids generated within the eye and those released are in equilibrium in healthy eyes.

ARCHITECTURE DIAGRAM



BLOCK DIAGRAM



CONCLUSION

Our project aimed to address the critical issue of early detection and management of glaucoma using advanced technology and user-friendly interfaces. By leveraging Roboflow for image preprocessing and MobileVNet architecture for glaucoma stage detection, we created an efficient and accurate solution for diagnosing glaucoma from input images. The integration of Streamlit for frontend development enabled us to create an intuitive and interactive user interface, allowing users to easily upload their images, receive prompt glaucoma stage assessments, and access preventive measures.

This project represents a significant advancement in the field of medical image analysis and healthcare technology. By providing a user-friendly platform for glaucoma detection and education on preventive measures, we aim to empower individuals to take proactive steps towards managing their eye health and reducing the risk of vision loss associated with glaucoma. Privacy, consent, and data security are paramount. We ensured robust anonymization and compliance. Transparency in model predictions fosters trust among users and clinicians.

Our project bridges the gap between technology and healthcare. By democratizing access to glaucoma assessment, we empower individuals to take proactive steps toward eye health. Furthermore, the use of streamlined image preprocessing techniques and lightweight convolutional neural network architectures demonstrates the feasibility of deploying such solutions in resource-constrained environments or mobile applications.

Moving forward, our project opens up avenues for further research and development in the field of computer-aided diagnosis and telemedicine. Future enhancements could include expanding the scope of the platform to cover other eye diseases, refining the accuracy of glaucoma stage detection algorithms, and incorporating additional features such as patient tracking and risk assessment. Overall, our project underscores the potential of technology-driven solutions in improving healthcare outcomes and promoting proactive health management.

FUTURE ENHANCEMENTS

LONGITUDINAL TRACKING AND PREDICTIVE ANALYTICS: Envision a future where longitudinal monitoring of glaucoma patients, facilitated by remote monitoring devices and wearable technologies, becomes commonplace. Predictive analytics algorithms leveraging longitudinal data could offer insights into future disease trajectories, enabling proactive interventions aimed at mitigating vision loss risks.

REAL-TIME MONITORING AND ALERTS: Implement a real-time monitoring system. Users can upload new eye images periodically (e.g., monthly), and the system alerts them if any significant changes occur. This proactive approach ensures timely medical attention.

PERSONALISED RISK ASSESSMENT: Develop an algorithm that assesses an individual's risk based on factors like age, family history, intraocular pressure, and other health conditions. Provide personalized recommendations accordingly.

Project Mapping with various courses of the Curriculum with attained POS

Name of Course from which Principles are applied in this project	Description of the Application	Attained POs
C324,C312	Thorough Examination of Existing systems and definition of the problem.	PO1,PO2,PO7, PO8
C311,C312,C313	Gathering, Analysis and classification of all the requirements for the proposed system.	PO1,PO2,PO4
C124, C222	Logical design is done by using logical tools.	PO1,PO2,PO3, PO5
C226,C227,C229	The physical design is done by using Streamlit, Jupyter Notebook.	PO5,PO9,PO11, PO12
C421, C314, C324	Each and every module is tested, integrated, and evaluated.	PO2,PO5,PO8, PO11,PO12
C421,C314,C324 C414	Implementation of the project .	PO1,PO3,PO5, PO7,PO8,PO11
C421, C3110, C317,C412	Documentation is done by all the team members with well defined communication with inclusion of Engineering and management Principles	PO6,PO7,PO8, PO9,PO10,PO11, PO12
C421, C3110, C3210	Presentation of the work in teams with proper method of presentation.	PO9,PO8,PO10, PO11

COURSE STRUCTURE (R13)
I YEAR- I SEMESTER

S No	Subject	T	P	Credits
C111	Problem Solving and Programming Using C	3	0	3
C112	Applied Chemistry	3	0	3
C113	Differential Equations	3	0	3
C114	Engineering Graphics	1	4	3
C115	Basics of Electrical and Electronics Engineering	3	0	3
C116	Problem Solving and Programming Using C Lab	0	3	1.5
C117	IT Workshop	0	3	1.5
C118	Applied Chemistry Lab	0	3	1.5
Total Credits				19.5

I YEAR- II SEMESTER

S No	Subject	T	P	Credits
C121	Communicative English	3	0	3
C122	Applied Physics	3	0	3
C123	Linear Algebra & Vector Calculus	3	0	3
C124	Digital Logic Design	3	0	3
C125	Python Programming	3	0	3
C126	Environmental Sciences	2	0	0
C127	Communicative English Skills Lab	0	3	1.5
C128	Applied Physics Lab	0	3	1.5
C129	Python Programming Lab	0	3	1.5
Total Credits				19.5

II YEAR- I SEMESTER

S No	Subject	T	P	Credits
C211	Probability & Statistics	3	0	3
C212	Mathematical Foundations of Computer Science	3	0	3
C213	Data Structures & Algorithms	3	0	3
C214	Object Oriented Programming through Java	3	0	3
C215	Introduction to Artificial Intelligence	3	0	3
C216	Constitution of India	2	0	0
C217	Data Structures &Algorithms Lab	0	3	1.5
C218	Object Oriented Programming through Java lab	0	3	1.5
C219	Introduction to Artificial Intelligence Lab	0	3	1.5
C2120	Skill Oriented Course -1	1	2	2.0
Total Credits				21.5

II YEAR- II SEMESTER

S No	Subject	T	P	Credits
C221	Numerical Methods & Transformations	3	0	0
C222	Computer Organization	3	0	0

C223	Database Management Systems	3	0	0
C224	Formal Languages and Automata Theory	3	0	0
C225	Managerial Economics and Financial Accountancy	3	0	0
C226	Database Management Systems Lab	0	0	3
C227	Web Application Development Lab	0	0	3
C228	R Programming Lab	0	0	3
C229	Skill Oriented Course -2	1	0	2
Total Credits				21.5

III YEAR- I SEMESTER

S No	Subject	T	P	Credits
C311	Design and Analysis of Algorithms	3	0	3
C312	Machine Learning	3	0	3
C313	Operating Systems	3	0	3
C314	Professional Elective -1	3	0	0
C315	Open Elective -1	3	0	3
C316	Skill Oriented Course - III	0	4	2
C317	Professional Ethics and HumanValues	2	0	0
C318	Summer Internship one Month (Mandatory) after second year(to be evaluated during V Semester	0	0	1.5
C319	Machine Learning Lab	0	3	1.5
C3110	Operating Systems Lab	0	3	1.5
Total Credits				21.5

III YEAR- II SEMESTER

S No	Subject	T	P	Credits
C321	Computer Networks and communications	3	0	3
C322	Deep Learning	3	0	3
C323	Expert Systems	3	0	3
C324	Professional Elective -2	3	0	3
C325	Open Elective -2	3	3	3
C326	Skill Oriented Course - IV (Soft Skills)	0	2	2
C327	Intellectual property rights and patents (IPR&P)	2	0	0
C328	Deep Learning Lab	0	3	1.5
C329	Computer Networks and communications Lab	0	3	1.5
C3210	Mini Project with seminar	1	2	1.5
Total Credits				21.5

IV YEAR- I SEMESTER

S No	Subject	T	P	Credits
C411	Professional Elective-III	3	0	3
C412	Professional Elective-IV	3	0	3
C413	Professional Elective-V	3	0	3
C414	Open Elective-III	3	0	3

C415	Open Elective - IV	3	0	3
C416	Management Science	3	0	3
C417	Skill Oriented Course -V	0	4	2
C418	Industrial/Research Internship one months (Mandatory) after third year (to be evaluated during VII semester)	0	0	1.5
Total Credits				23

IV YEAR- II SEMESTER

S No	Subject	T	P	Credits
C421	Project work - Phase II	0	0	12
Total Credits				12

OPEN ELECTIVES		
Open Elective - 1 (V Semester) Python Programming	Open Elective -2 (VI Semester) Fundamentals of Artificial Intelligence	Open Elective -3 (VII Semester) Human Computer Interaction
Open Elective -4 (VII Semester) Applications of Artificial Intelligence	Skill Oriented Course (Advanced)– IV 1. (MEAN Stack Technologies - Module I- MongoDB, Express.js, Angular JS, Node.js and AJAX 2. Big Data : Apache Spark 3. DevOPS	
PROFESSIONAL ELECTIVES		
Professional Elective – 1 (V Semester)	Professional Elective – 2 (VI Semester)	Professional Elective – 3 (VII Semester)
1. Software Engineering 2. Compile Design 3. Data Visualization 4. Design and Analysis of Algorithms	1. Software Project Management 2. Distributed Systems 3. Internet of Things 4. Data Ware Housing and Data Mining	1. Reinforcement Learning 2. Soft Computing 3. Cryptography and Network Security 4. NOSQL Databases 5. Natural Language Processing
Professional Elective – 4 (VII Semester)	Professional Elective – 5 (VII Semester)	
1. Robotic Process Automation 2. Cloud Computing 3. Big Data Analytics 4. Block Chain Technologies 5. Image & Video Analytics	1. Social Network Analysis 2. Recommender Systems 3. Computer vision 4. Object Oriented Analysis and Design 5. Semantic Web	

BOOKS REFERED

- "Deep Learning" by Ian Goodfellow, Yoshua Bengio, and Aaron Courville.
- "Hands-On Machine Learning with Scikit-Learn, Keras, and TensorFlow" by Aurélien Géron.
- "Deep Learning with Python" by François Chollet (creator of Keras).
- "TensorFlow 2.0 in Practice" by Sebastian Raschka and Josh Patterson.
- "Streamlit: A Quick Guide to Learning Streamlit for Data Science and Machine Learning" by Hancel Gonzalez.
- "Deep Learning for Vision Systems" by Rajalingappa Shanmugamani.

WEBSITES VISITED

Kaggle : Kaggle is a platform that hosts datasets, machine learning competitions, and resources for data science projects. You can find Glaucoma related fundus image datasets, kernels (code notebooks), and discussions related to Glaucoma stage detection.

GitHub: Explore GitHub repositories for open-source projects, code samples, and libraries related to Glaucoma stages detection, deep learning, and data analysis. You may find useful code snippets, frameworks, and tools for your project.

Roboflow : We utilized datasets sourced from Roboflow, comprising annotated images of various eye conditions, including glaucoma stages. These datasets offer a diverse range of images crucial for training our model to accurately detect different stages of glaucoma.

IEEE: We also leveraged datasets sourced from reputable academic journals such as IEEE, augmenting our collection with high-quality, peer-reviewed images pertinent to glaucoma stages detection. These datasets from esteemed publications enriched our training data, ensuring robustness and accuracy in our model's performance.

References

- [1] A. Navea, A. Diaz-Pinto, S. Morales, J. M. Mossi, V. Naranjo, and T. Köhler. A thorough evaluation of CNNs for glaucoma identification using images of the fundus. *biomedical engineering online*. 2019 Dec;18:1-9.
- [2] J. Wen, M. Hassan, N. Nasrullah, S. Sun, and S. Hayat, as well as M. Mateen. Analyses of datasets, methods, and evaluation metrics for the automatic diagnosis of diabetic retinopathy. *IEEE Access*.
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- [6] Automated glaucoma diagnosis employing segmentation of the retinal image and optic cupsin fundus photographs was developed by Yin, F., J. Liu, D. Wong, N. M. Tan, C. Cheung, M. Baskaran, T. Aung, and T. Y. Wong. The 25th annual International Conference on Computer-based Health Systems (CBMS) was held by the IEEE on June 20, 2012.(pp. 1-6). IEEE.
- [7] Wong TY, Wong DW, Tan NM, Tao D, Cheng CY, and Aung. Yin F, Cheng J, Liu J, Xu Y, and Yin F. Superpixel classification is used to separate the optic disc from the optic cup for glaucoma screening. 2013 Feb18;32(6):1019-32 in the *Journal of IEEE on Diagnostic Imaging*.
- [8] MD, S. Khilariwal, and R. Verma. An antenna with a notch design for 5 GHz high-speedLAN. The 23rd annual World Convention on Networking, Signal Processing, and Wireless Communications is called Wisp NET 2016.(pp. 999- 1002). IEEE.
- [9] Sampling techniques and clustering techniques for automatic segmentation of the optic discand cups are used to identify glaucoma.
Medical images and images on a computer. 2017 Jan 1;55:28-41. Zilly J, Buhmann JM,Mahapatra D



GLAUCOMA STAGES DETECTION USING FUNDUS IMAGES THROUGH DEEP LEARNING

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Abstract: A chronic eye condition called glaucoma has a deleterious effect on the optical nerve, which links the brain and eye to transmit visual information. Early detection is essential for stopping the condition's progression. Glaucoma is one of the most prevalent eye conditions, and it's important to catch it early because it can cause blindness and neurological issues. In this study, a Deep Learning system is proposed for the early detection of glaucoma. The eye image undergoes pre-processing to eliminate any noise and prepare them for further analysis. The system utilizes enlarged images of the eyes as input data for the deep learning method. The suggested system classifies new eye images as No_glucoma Glaucoma_early, Glaucoma_moderate, and Glaucoma_advanced and also provides respective preventive measures based on the features it learned during training.

Index Terms - Glaucoma, Deep Learning, pre-processed, Fundus Images.

I. INTRODUCTION

Glaucoma, One of the leading causes of blindness worldwide is glaucoma, a long-term neurodegenerative eye disease. According to the WHO, an average of 65 million people around the world are affected by glaucoma. Given that the primary symptom of glaucoma, the loss of optic nerve fibers, may be asymptomatic, early diagnosis and treatment are crucial in preventing vision loss. This loss is caused by increased intracranial pressure or decreased blood flow into the optic nerve. Visual data is transmitted via the optic nerve from the brain to the eye. Pathologically high intraocular pressure, which can suddenly rise to 60-70 mmHg is a symptom of glaucoma. Prolonged pressure of less than 25-30 mmHg can result from 2 in visual loss. High pressure in glaucoma is caused by increased reluctance to fluid expulsion into the drainage system of the eye. The fluids generated within the eye and those released are in equilibrium in healthy eyes. A common method used in ophthalmology to examine the human eye is taking a photo of the eye's fundus using a fundus camera. The medical professional takes the picture through the pupil to capture the eye's background. The photos are then analyzed, which can take several hours on a computer, but the results are not always accurate. Diagnosing glaucoma at home is a challenging task that requires determination and patience. We employed a supervised learning method classifier to distinguish between a healthy eye fundus and one affected by glaucoma. SVM aims to build a model, based on training and test data, which predicts the key features of the test data. SVM

is a popular supervised learning technique used for classification or regression problems. For classification issues, the SVM algorithm is a popular choice in machine learning. Its purpose is to create a boundary line or decision point that can divide high-dimensional spaces into classes, making it easier to categorize new data points in the future. This boundary line is referred to as a hyperplane[4]. The objective is to detect the abnormalities automatically and conditions with the least amount of error.

However, when used with SVM algorithms for images obtained with fast-rising spatial resolution, conventional image processing methods that were created and tested on low-resolution images have limits. A new set of methods must be devised for this purpose. Because Convolutional Neural Networks (CNNs) can handle high-resolution images with minimal processing expense, we use them. CNNs are one kind of neural network that is frequently employed for image recognition applications.

The network's convolutional layer lowers the high dimensionality of the images while retaining crucial data. Another similar model that extracts features through convolutional filters is the Convolutional Neural Network (CNN). In large datasets, CNNs have become the preferred method for efficient and accurate image classification.

II. LITERATURE SURVEY:

Glaucoma, a condition characterized by the loss of retinal cells and astrocytes, can be assessed through specific measurements related to the eye cup and the neuro-retinal rim. Researchers have extensively explored this topic using fundus images, with a primary focus on quantifying the size of the retinal ganglion cell head.

One study proposed a system for measuring the Cup-to-Disc Ratio (CDR) using position-set methods and optic cup masks. Their evaluation involved 104 images, aiming for a CDR difference of less than 0.2 points from ground truth. Another approach, based on anatomical features, identified the optic cup using blood vessel curvature at the cup boundary. Using a container shape and circular Hough transform, this method achieved a CDR error of 0.12 to 0.10 in locating the eye cup.

In a separate study, researchers Yin et al. employed the Circular Wavelet transform to segment the optic disc or cup in 325 fundus images, achieving average correlation measures of 0.92 and 0.81. Cheng and colleagues proposed an alternative method that utilized superpixels for retinal image and cup segmentation. Their system, tested on 650 images, yielded average Jaccard scores of 0.800 and 0.822 across two datasets.

Additionally, Liu et al. incorporated patient-specific and genetic information into their study. The loss of eye nerve fibers and astrocytes remains a key symptom of glaucoma, emphasizing the importance of accurate measurements of the eye cup length and neuroretinal rim viscosity. Overall, various techniques, including position-set methods, anatomical verification, and Circular Hough transform, have been explored for computing the CDR, yielding diverse results across different datasets.

III. PROPOSED SYSTEM:

3.1 INTRODUCTION:

The proposed system aims to detect stages of glaucoma using fundus images sourced from Roboflow. Leveraging deep learning models such as MobileNet and InceptionV3, the system seeks to achieve high accuracy in glaucoma stage classification. The system's architecture encompasses data preprocessing, model training, evaluation, and deployment, ensuring a comprehensive approach to glaucoma detection.

3.2 SYSTEM OVERVIEW:

1. Data Acquisition: Utilize Roboflow to acquire a diverse dataset of fundus images, annotated with glaucoma stage labels. Ensure sufficient representation of various stages of glaucoma for robust model training.

2. Preprocessing: Apply preprocessing techniques to standardize image quality and enhance relevant features. These may include resizing, normalization, and augmentation to improve model generalization.

3. Model Selection: Choose MobileNet and InceptionV3 as deep learning architectures for glaucoma stage detection. These models are well-suited for image classification tasks and offer a balance between accuracy and computational efficiency.

4. Training: Train MobileNet and InceptionV3 on the preprocessed dataset using transfer learning. Fine-tune the models on fundus images to adapt them to the task of glaucoma stage detection.

5. Evaluation: Evaluate the trained models using validation and test datasets to assess their performance in glaucoma stage classification. Measure metrics such as accuracy, sensitivity, specificity, and AUC-ROC to gauge model effectiveness.

6. Deployment: Deploy the trained models in a production environment for real-time glaucoma stage detection. Integrate the models into a user-friendly interface, allowing healthcare professionals to input fundus images and receive predicted glaucoma stages promptly.

3.3 ADVANTAGES OF PROPOSED SYSTEM:

1. Utilization of Roboflow Dataset: Leveraging a diverse dataset from Roboflow ensures comprehensive coverage of glaucoma stages, enhancing model generalization and robustness.
2. Efficient Model Architectures: MobileNet and InceptionV3 offer a balance between accuracy and computational efficiency, making them suitable for deployment in resource-constrained environments.
3. Transfer Learning: By employing transfer learning, the proposed system can leverage pre-trained models' knowledge, accelerating training and improving performance on the glaucoma detection task.
4. Real-time Deployment: The system enables real-time glaucoma stage detection, facilitating prompt intervention and treatment decisions by healthcare professionals.
5. User-friendly Interface: The integration of the models into a user-friendly interface simplifies the process of inputting fundus images and accessing predicted glaucoma stages, enhancing usability for healthcare practitioners.

3.4 PROPOSED SYSTEM WORKFLOW:

1. Data Collection: The system begins by sourcing a diverse dataset of fundus images annotated with glaucoma stage labels. These images are obtained from Roboflow, ensuring a varied representation of glaucoma stages for robust model training.

2. Data Preprocessing: Prior to model training, the fundus images undergo preprocessing steps to standardize image quality and enhance relevant features. Techniques such as resizing, normalization, and augmentation are applied to optimize image representation and improve model performance.

3. Model Selection: The system selects MobileVNet and InceptionV3 as deep learning architectures for glaucoma stage detection. These models are chosen for their ability to balance computational efficiency with high performance, making them suitable for deployment in resource-constrained environments.

4. Model Training: Utilizing transfer learning techniques, the selected models are trained on the preprocessed dataset. Transfer learning allows the models to adapt quickly to the nuances of glaucoma detection, thereby accelerating training and enhancing overall accuracy.

5. Model Evaluation: Trained models undergo thorough evaluation using validation and test datasets to assess their performance in glaucoma stage classification. Metrics such as accuracy, sensitivity, specificity, and area under the ROC curve are computed to evaluate model effectiveness and generalization.

6. System Deployment: Upon successful training and evaluation, the trained models are deployed in a production environment for real-time glaucoma stage detection. The system interface facilitates seamless

interaction, enabling healthcare professionals to input fundus images and promptly obtain predicted glaucoma stages.

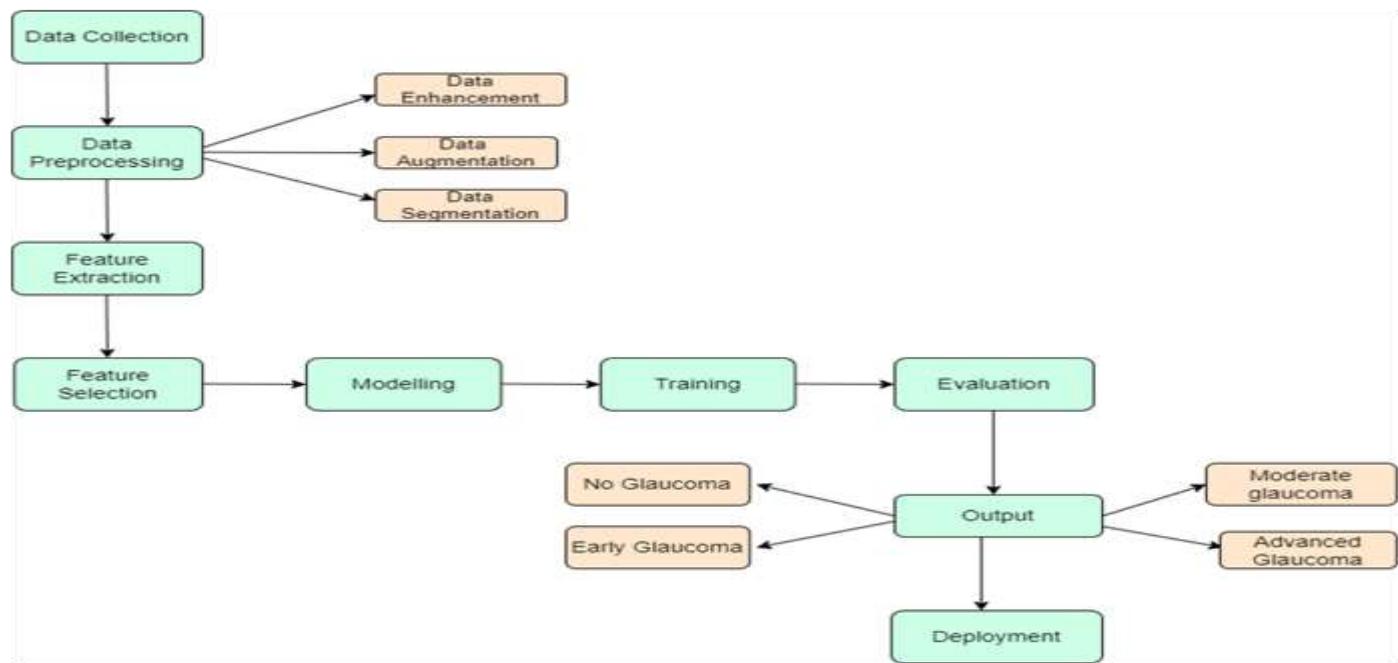


Fig 1: proposed system workflow diagram

3.5 ILLUSTRATIVE WORKFLOW:

- 1. User Interaction:** A healthcare practitioner accesses the system interface and uploads a fundus image of a patient's eye suspected of having glaucoma.
- 2. Image Processing:** The system preprocesses the uploaded image, standardizing its dimensions and enhancing its quality through normalization techniques.
- 3. Model Inference:** The preprocessed image is passed through both MobileVNet and InceptionV3 models for inference. The models analyze the image and provide predictions regarding the likelihood of the patient being in various stages of glaucoma.
- 4. Result Presentation:** The system presents the predicted glaucoma stage(s) along with associated confidence scores or probabilities to the healthcare practitioner. Additionally, it may offer recommendations for proactive measures based on the detected stage(s).
- 5. Clinical Decision-making:** Armed with the system's output, the healthcare practitioner can make informed decisions regarding further diagnostic procedures, treatment modalities, and patient management strategies, thereby enhancing the quality of care provided.

IV. LIBRARIES USED:

- 1. Deep Learning:** Deep learning techniques are employed for training the glaucoma detection models. These techniques involve neural networks with multiple layers that can automatically learn hierarchical representations of fundus images to identify glaucoma stages.

2. Roboflow: Roboflow is employed for managing and preprocessing the fundus image dataset used for training the glaucoma detection models. It offers tools for annotating, augmenting, and organizing image data, streamlining the data preparation process.

3. TensorFlow: TensorFlow serves as the primary deep learning framework for implementing and training the glaucoma detection models. It provides a comprehensive set of tools and APIs for building and optimizing deep neural networks.

4. Keras: Keras, as a high-level neural networks API, is likely utilized in conjunction with TensorFlow for rapid prototyping and experimentation with different model architectures. Keras simplifies the process of designing, training, and evaluating deep learning models.

5. Streamlit: Streamlit is utilized to create a user-friendly web application for interacting with the trained glaucoma detection models. It enables healthcare professionals to upload fundus images and receive predictions regarding the stages of glaucoma in real time.

V. METHODOLOGY

Introduction: The methodology proposed in this paper aims to leverage deep learning techniques, specifically InceptionV3 and MobileNet architectures, for the accurate detection of glaucoma stages from fundus images. Additionally, the methodology extends to providing personalized proactive measures based on the detected stage. This review outlines the key steps and approaches employed in the proposed methodology.

Data Collection: The methodology begins with the collection of a diverse dataset of fundus images, comprising images from individuals at various stages of glaucoma. Data collection efforts likely involved collaboration with healthcare institutions or utilizing publicly available datasets, ensuring adequate representation of different glaucoma stages for robust model training.

Preprocessing: Fundus images undergo preprocessing steps aimed at standardizing image quality and enhancing relevant features. Common preprocessing techniques include resizing, normalization, denoising, and contrast adjustment. These steps ensure consistency across the dataset and optimize image quality for subsequent analysis.

Feature Extraction: Extracting informative features from fundus images is crucial for accurate glaucoma stage detection. The methodology likely involves extracting optic disc parameters (e.g., cup-to-disc ratio, disc area), retinal nerve fiber layer thickness, and vessel morphology, among other relevant features associated with glaucoma progression.

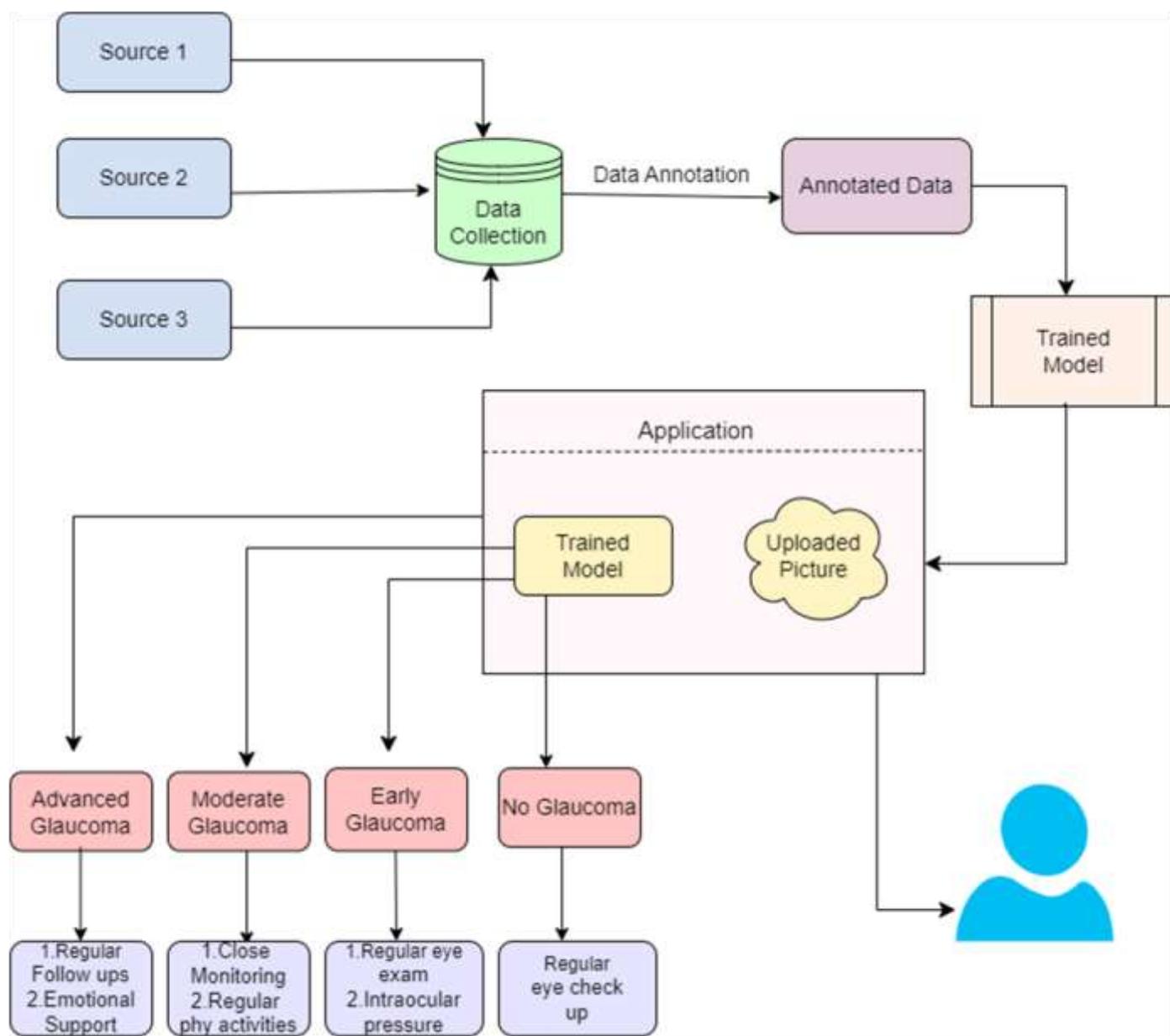
Model Selection: The methodology selects two deep learning architectures, InceptionV3 and MobileNet, for glaucoma stage detection. These architectures are known for their effectiveness in image classification tasks and are chosen for their ability to handle the complexity of fundus images efficiently. The selection process likely considers factors such as model complexity, computational resources, and performance metrics.

Training and Evaluation: The selected models undergo training using the preprocessed dataset, with appropriate validation and test splits. During training, model parameters are optimized to minimize classification errors and maximize performance metrics such as accuracy, sensitivity. The trained models are then evaluated using separate validation and test sets to assess generalization performance and obtain unbiased estimates of effectiveness in glaucoma stage detection.

Output Generation: Upon successful training and evaluation, the trained models generate predictions for glaucoma stage detection. The output includes predicted stages for each fundus image, accompanied by

corresponding confidence scores or probabilities. Visualizations of model predictions and performance metrics aid in interpretation and decision-making for clinicians and healthcare practitioners.

VI. ARCHITECTURE DESIGN:



VII. FUTURE PROSPECTS:

- Advancements in Imaging Modalities:** Anticipate ongoing enhancements in imaging technologies like optical coherence tomography (OCT) and adaptive optics, promising more detailed assessments of glaucoma-related structural changes. These advancements could significantly elevate diagnostic accuracy and refine staging capabilities.
- Integration of Multifaceted Data:** The fusion of various data streams, including fundus images, OCT scans, visual field assessments, and patient demographics, holds immense potential for a comprehensive understanding of glaucoma progression. Innovations in machine learning algorithms capable of synthesizing and interpreting diverse datasets may revolutionize staging precision and individualized treatment strategies.
- Deep Learning and Artificial Intelligence Innovations:** Continued evolution and fine-tuning of deep learning algorithms and artificial intelligence methodologies are projected to streamline and optimize glaucoma staging processes. Models trained on extensive and diverse datasets may exhibit improved adaptability and accuracy in identifying and categorizing distinct glaucoma stages.
- Longitudinal Tracking and Predictive Analytics:** Envision a future where longitudinal monitoring of glaucoma patients, facilitated by remote monitoring devices and wearable technologies, becomes

commonplace. Predictive analytics algorithms leveraging longitudinal data could offer insights into future disease trajectories, enabling proactive interventions aimed at mitigating vision loss risks.

5. Telemedicine Integration and Remote Consultation: Expect telemedicine platforms and remote consultation services to increasingly integrate automated staging algorithms. This integration has the potential to revolutionize glaucoma management, particularly in remote or underserved regions, by facilitating prompt diagnosis and personalized treatment planning.

6. Emphasis on Patient-Centered Care and Informed Decision-Making: Foresee a shift towards patient-centric care models emphasizing shared decision-making. Future tools and technologies should empower patients by providing personalized risk assessments, treatment options, and prognostic insights, thereby fostering active engagement in their care journey.

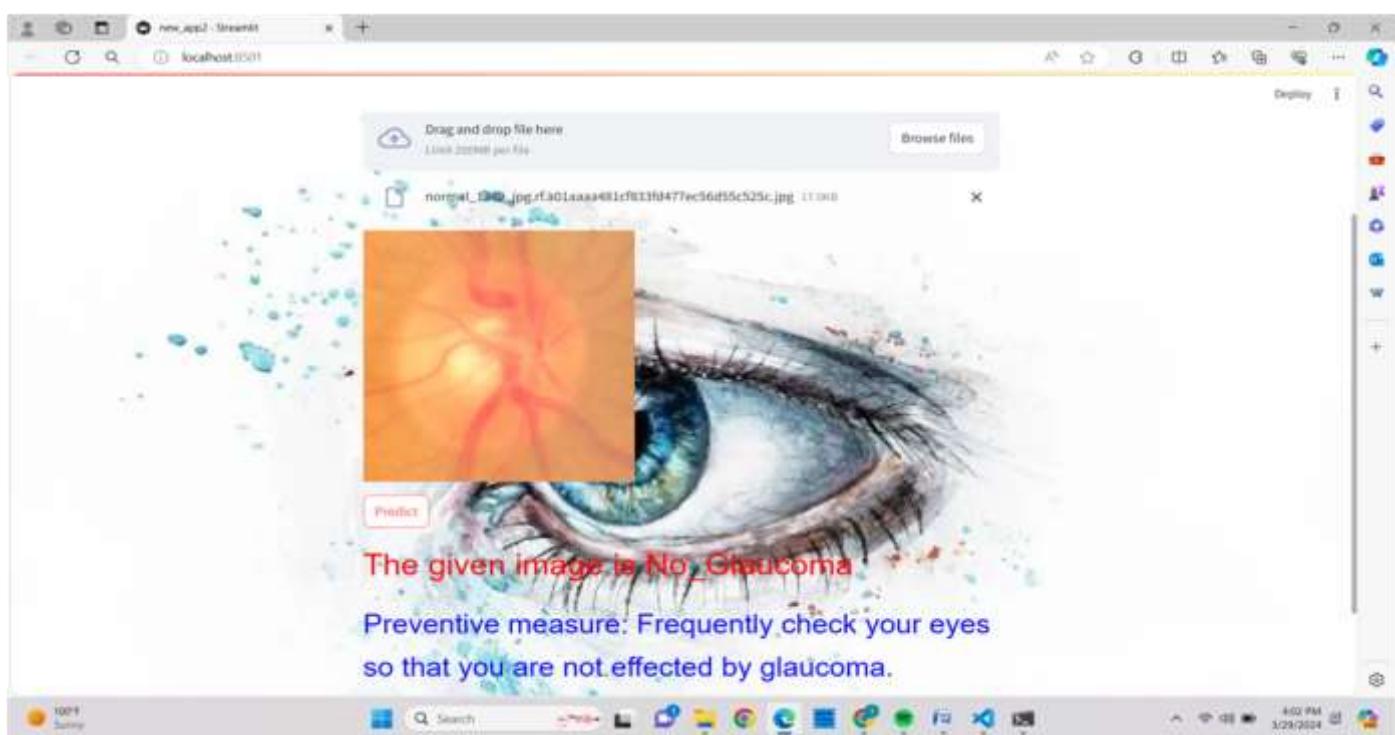
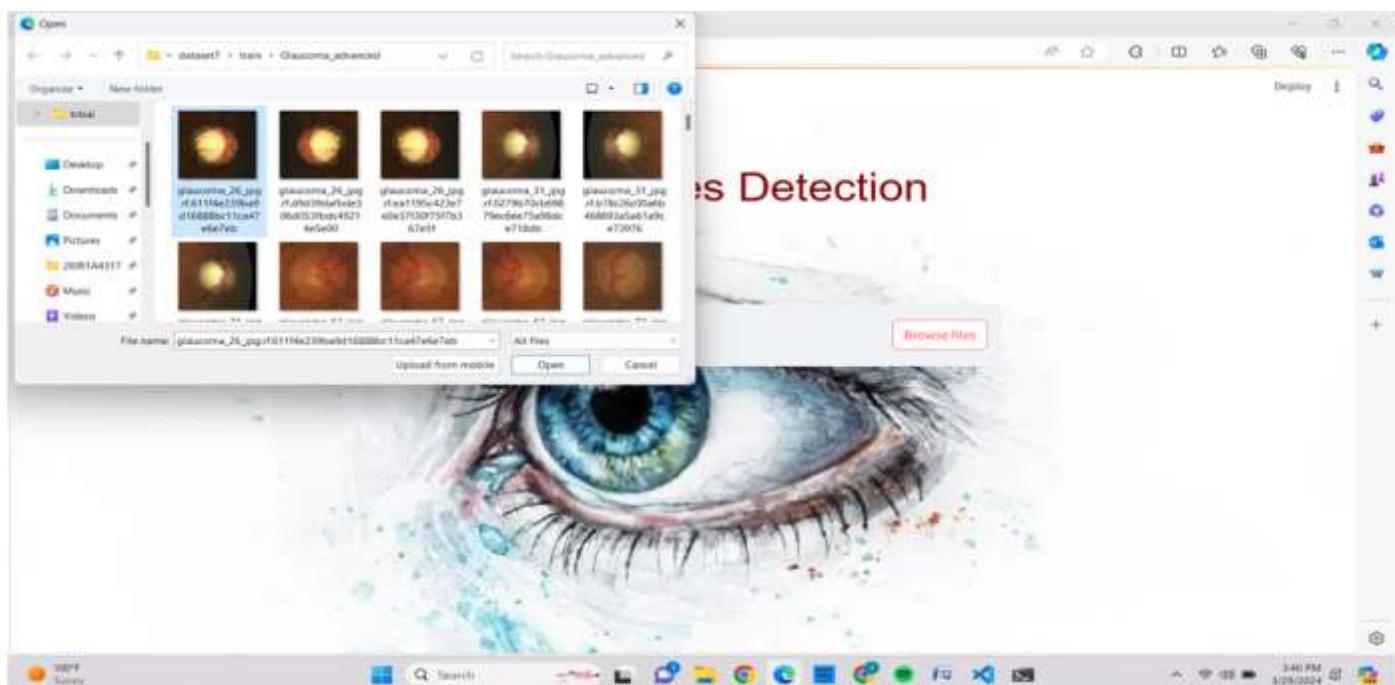
7. Collaborative Research and Clinical Collaboration: Collaboration across research institutions and concerted efforts in large-scale clinical trials will be pivotal for validating and fine-tuning emerging staging technologies. Multicenter studies involving diverse patient cohorts and longitudinal follow-ups will yield invaluable real-world insights into the efficacy and clinical applicability of novel staging methodologies.

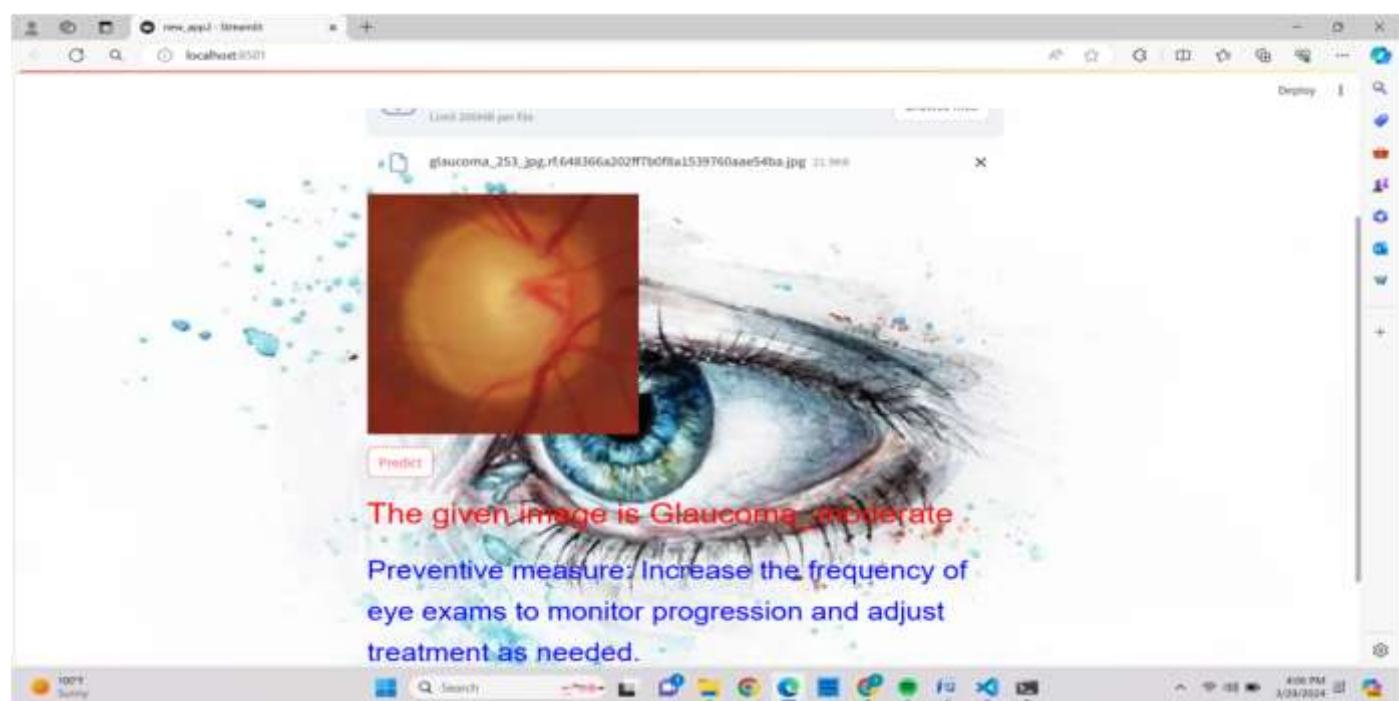
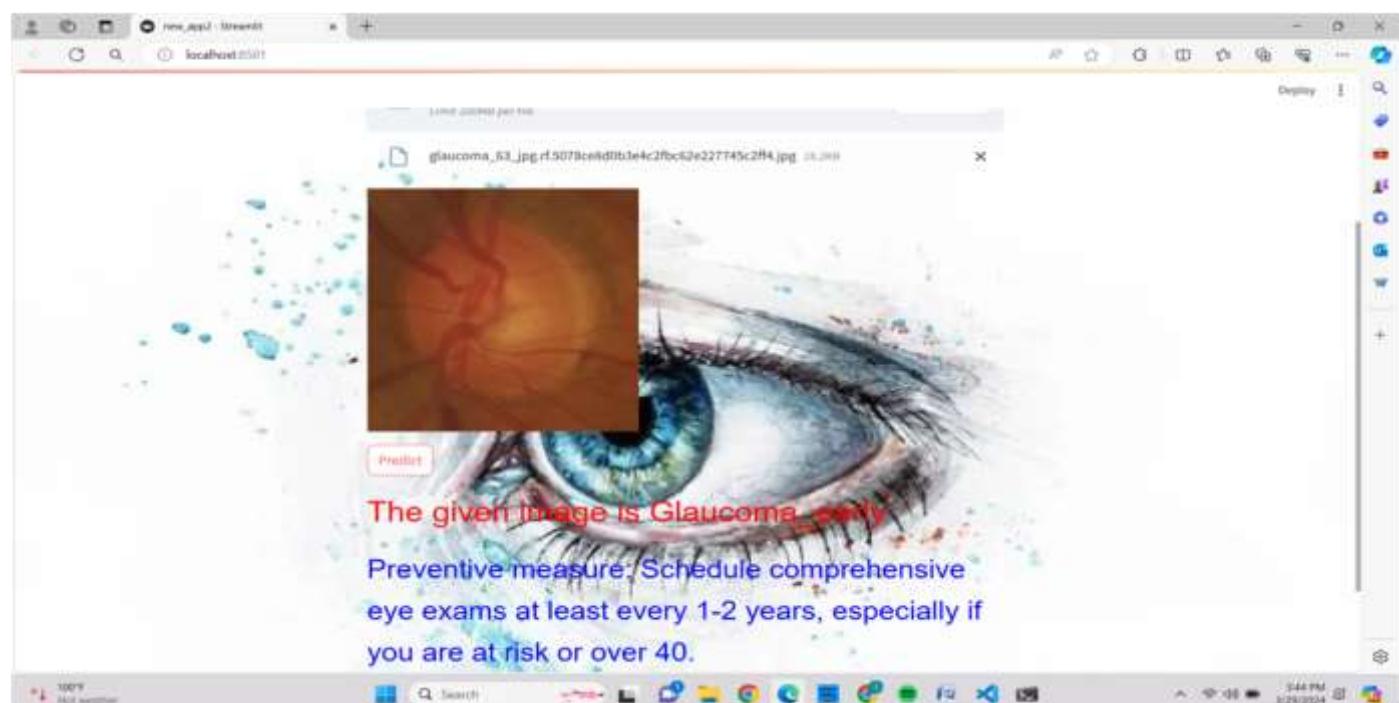
VIII. OUTPUT SCREENS:

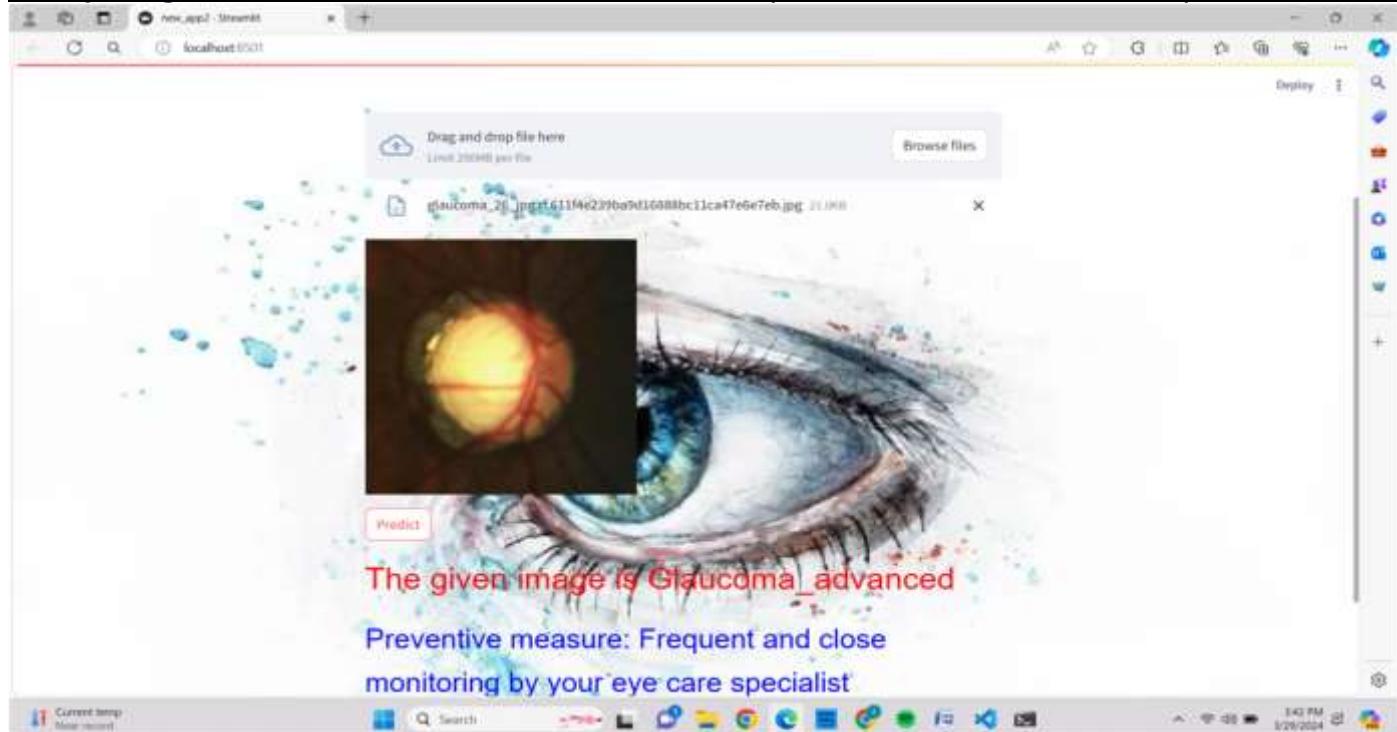
The output of the proposed system is given in 4 ways. There are as follows

1. No_Glaucoma
- 2.Glaucoma_early
- 3.Glaucoma_moderate
- 4.Glaucoma_advanced









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Time period	Description	Dates
Week -1	Python's Playground Kickoff Understanding Importance and Types of Data Analysis Python Interactive Visualization using Plotly for Dashboards	20-11-23 to 25-11-23
Week -2	Web Scraping using Python and Data Analysis	27-11-23 to 02-12-23
Week -3	Introduction to Machine Learning - Minds & Machines Convergence	04-12-23 to 09-12-23
Week -4	Model Deployment with Flask - Web Wizards Unleashed / Streamlit	11-12-23 to 16-12-23
Week -5	Advanced Machine Learning	18-12-23 to 23-12-23
Week -6	Introduction to Recommendation Systems - The Recommender Rendezvous	25-12-23 to 30-12-23
Week -7	End-to-End Machine Learning on AWS: From Dataset to Deployment using Amazon SageMaker	01-01-24 to 06/01/24
Week -8	Unsupervised Learning in the Machine Learning Ecosystem	08-01-24 to 13-01-24
Week -9	Computer Vision Basics - Visions of Pixels	15-01-24 to 20-01-24
Week -10	Deep Learning Dive - Into the Neural Abyss	22-01-24 to 27-01-24
Week -11	Getting into Convolutional Neural Networks in Depth Understanding Different Deep Learning Frameworks (Tensorflow, Keras)	29-01-24 to 03-02-24
Week -12	RNN - Recurrent Neural Networks	05-02-24 to 10-02-24
Week -13	Tensor Flow Lite Mastery: A Step-by-Step Guide to Mobile AI Integration	12-02-24 to 17-02-24
Week -14	Deep Unsupervised Learning using Tensor flow and Keras Recommender Systems using Restricted Boltzmann Machines	19-02-24 to 24-02-24
Week -15	Large Language Models and Generative AI Introduction to Generative AI	26-02-24 to 02-03-24
Week -16	Project Work and Optimization	04-03-24 to 09-03-24
Week -17	Project Work and Optimization	11-03-24 to 16-03-24
Week -18	Documentation	18-03-24 to 23-03-24
Week -19	Documentation	25-03-24 to 30-03-24
Week-20	Project Presentation and Conclusion	01-04-24 to 06-04-24