



Beyond Needles: Blood Group Detection Via IR Spectroscopy

¹LIKHITHA V K, ²MITHUN K B, ³NANDITHA R, ⁴RAKSHITHA G M, ⁵Dr.RASHMI R DESHPANDE

^{1,2,3,4}STUDENT, ⁵ASSOCIATE PROFESSOR & HOD

^{1,2,3,4,5} Department of Information Science & Engineering ,

^{1,2,3,4,5} AMC Engineering College , Bengaluru, India

Abstract: Determining the blood group is a crucial step in medical diagnostics that guarantees the security and suitability of organ transplants, blood transfusions, and life-saving emergency treatments. Conventional blood typing procedures include laboratory-based reagent testing and invasive sample collection, which call for skilled workers, regulated settings, and labor-intensive procedures. These drawbacks emphasize the necessity of a non-invasive substitute that can produce dependable and quick results in a range of contexts. Using Infrared (IR) spectroscopy in conjunction with image processing and machine learning techniques, this project presents a novel method for blood group detection. Blood group-specific biochemical and structural characteristics are correlated with distinct patterns of light absorption and reflection that are revealed by infrared imaging, which captures the spectral characteristics of biological tissues, especially the hand and fingerprint regions. Significant features are then extracted from these IR spectral images using sophisticated image processing algorithms. These features are then fed into machine learning models that have been trained to categorize various blood types. By doing away with the necessity for physical blood collection, this non-invasive method lowers discomfort, biohazard risks, and logistical difficulties. It provides an economical, transportable, and effective solution that can be used in military operations, hospitals, remote healthcare facilities, and disaster response situations. The ultimate goal of this project is to transform diagnostic medicine by offering a quicker, safer, and easier way to identify blood groups.

Index Terms - Blood Group Detection, Infrared (IR) Spectroscopy, Image Process, Light Absorption and Reflection, Machine Learning, Non – Invasive Technique.

INTRODUCTION

Clinical diagnostics and emergency medical care depend heavily on blood group identification. Knowing a patient's blood type in a timely and accurate manner can mean the difference between life and death in situations involving organ transplantation, blood transfusions, and accident response. Blood typing typically entails invasive procedures that call for the collection of blood samples and the use of chemical reagents in a lab. Despite their dependability, these techniques are time-consuming, necessitate knowledgeable technicians, and are not feasible in remote or emergency scenarios. As artificial intelligence and non-invasive sensing technologies advance, there is increasing interest in creating substitute methods that do not require actual blood collection. Because it can record the spectral properties of tissues according to their biochemical makeup, infrared (IR) spectroscopy presents a promising solution. IR imaging can be utilized to extract significant patterns that might be associated with characteristics specific to a person's blood group when combined with image processing and machine learning algorithms. The application of infrared spectral imaging of hands and fingerprints for non-invasive blood group identification is investigated in this project. The objective is to develop a portable, quick, and dependable substitute for conventional blood typing by combining IR spectroscopy with sophisticated image analysis and classification methods.

I. RELATED WORK

Many researchers have investigated non-invasive methods for medical diagnosis, with a special emphasis on blood group analysis and detection. A fingerprint-based identification method that connects blood group prediction to fingerprint ridge and minutiae was proposed by Vijaykumar Patil N. and D. R. Ingles [1]. Although image clarity and environmental factors limited the accuracy, their study showed the potential of biometric markers as a diagnostic tool. Using image processing, Anant Ajitkumar Vadgave et al. [2] created a system for determining blood group. Their work improved the speed and automation of conventional serological methods by using standard agglutination test images to classify blood types through image segmentation and pattern analysis. Invasive blood sampling was still necessary for the system's analysis, though. M. Bavyasri et al. [3] presented BloodHub, an intelligent web-based platform that matches donors with recipients and predicts blood groups by integrating Support Vector Machine (SVM) algorithms. This framework demonstrated how machine learning can expedite medical diagnostics and transfusion safety by emphasizing secure data management and precise classification based on genetic markers. Convolutional Neural Networks (CNNs) were used by Jahwanth Sai Ganta et al. [4] to implement a deep learning-based solution for classifying blood group images. Their model's high accuracy in identifying distinct blood group patterns was made possible by combining sophisticated feature extraction techniques like ORB (Oriented FAST and Rotated BRIEF) and SIFT (Scale-Invariant Feature Transform), opening the door for AI-assisted laboratory analysis. Chandrasekhar Reddy Gopu et al. [5] created a non-invasive antigen feature extraction method for automated blood group detection using YOLOv8 object detection. Their model proved the viability of real-time, non-invasive diagnostics through sophisticated image processing, processing each image in only 15 milliseconds and achieving a mean average precision (mAP) of 97.4%. All of these studies point to a major trend in medical diagnostics toward automation, non invasiveness, and AI-driven diagnosis. However, the majority of current techniques either use laboratory imaging setups or invasive blood samples. Beyond Needles: Blood Group Detection via IR Spectroscopy, the suggested system, overcomes these drawbacks by using machine learning in conjunction with infrared spectral imaging of the hand or fingertip to provide real-time, portable, and non-invasive blood group detection. This strategy offers a quicker, safer, and more effective substitute for conventional techniques, which is particularly helpful in emergency and remote healthcare situations.

II. PROBLEM STATEMENT

There are several significant disadvantages to traditional blood group detection methods, which rely on chemical reagents and laboratory-based testing. These procedures need blood samples, take a lot of time, and require skilled technicians to perform them correctly. They are also not practical in emergencies or field situations where time matters and access to medical facilities is limited. The demand for a quick, automated, and non-invasive blood group identification method is becoming more apparent. Current alternatives do not effectively tackle issues of usability, portability, and real-time diagnostics. This project seeks to fill this gap by developing a non-invasive technique for blood group detection using infrared spectroscopy. Identifying blood groups is essential for safely carrying out blood transfusions, organ transplants, and other medical treatments in clinical settings. However, in emergencies where rapid blood typing is needed, especially in remote or disaster areas with few medical resources, traditional blood typing methods are inadequate. Moreover, the invasive nature of these procedures causes discomfort for patients and raises the risk of contamination. There is a significant need for a non-invasive, portable, and efficient blood group detection system that can provide accurate results in real time and work in a variety of healthcare environments.

III. SYSTEM DESIGN OVERVIEW

This project combines machine learning and infrared (IR) spectroscopy to design and implement a reliable, non-invasive blood group detection system. The system is composed of several functional modules, including image acquisition, preprocessing, feature extraction, classification, and result visualization. Together, these modules analyze infrared spectrum data to reliably identify blood groups without the need for blood samples. The system is composed of the software component. A hardware-side infrared imaging module records spectral data from the hand or fingertip. Software-wise, Python based algorithms use machine learning models like Convolutional Neural Networks (CNN) and Support Vector Machines (SVM) to carry out data preprocessing, feature extraction, and classification. Throughout the entire process, high accuracy and intuitive interaction are ensured.

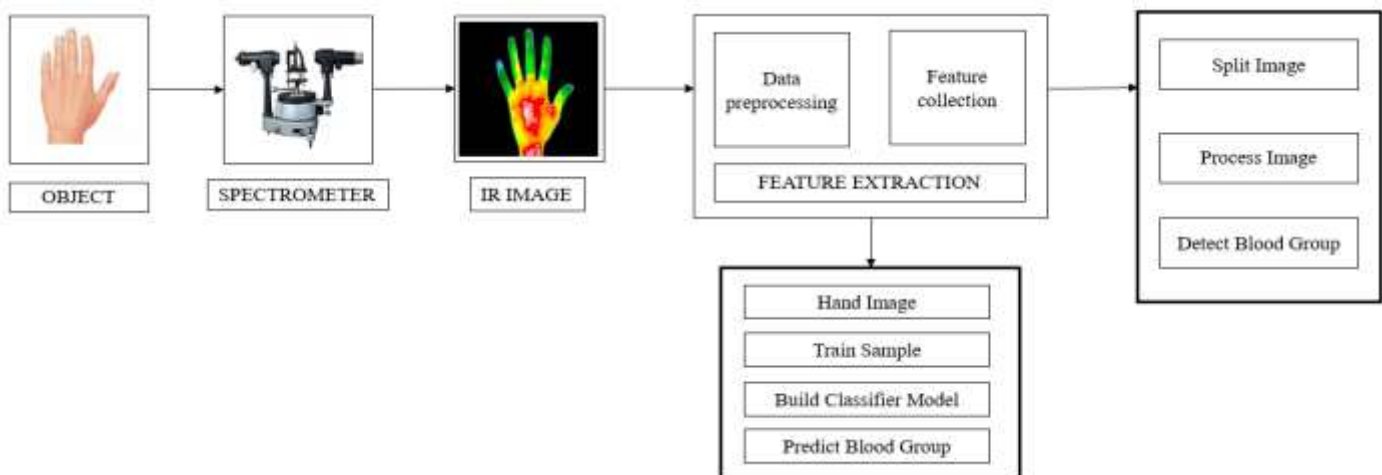


Figure 1: Block Diagram

A. Software Architecture : Python is the main programming language used in the software architecture, and it is supported by libraries such as TensorFlow/Keras for machine learning, NumPy and Pandas for data manipulation, and OpenCV for image processing. The system design makes use of a modular approach:

- 1. Data Acquisition Module:** captures infrared spectral images using an infrared sensor.
- 2. Preprocessing Module:** This module improves image quality, eliminates noise, and normalizes pixel intensities to ensure consistent spectral analysis.
- 3. Feature Extraction Module:** This module locates significant spectral peaks linked to blood group characteristics as well as biometric features.
- 4. Classification Module:** Trains and evaluates the machine learning model to accurately predict the user's blood group.
- 5. Interface Module:** This module shows the predicted blood group and confidence score in real time on an easy-to-use dashboard.

An integrated backend built with Flask or Django allows all modules to communicate with one another, ensuring seamless data transfer between the model inference engine and the image acquisition system.

B. Functional Workflow : The process begins with the IR imaging system, which captures the spectral reflection from the user's hand or fingertip. After receiving this captured image, the preprocessing module applies image enhancement and noise reduction. After receiving the cleaned spectrum data, the feature extraction unit analyzes wavelength intensity variations to find discriminative patterns related to hemoglobin and antigen properties. The extracted features are then fed into the classification model, which was trained using datasets of known blood groups. The model outputs the results to the user interface after predicting the blood group (A, B, AB, or O with Rh factor). The result is displayed instantly on the interface. This entire procedure ensures real-time, non invasive blood group detection that is precise, portable, and efficient. The integration of artificial intelligence, image processing, and infrared spectroscopy provides a seamless and scalable solution for medical and emergency applications.

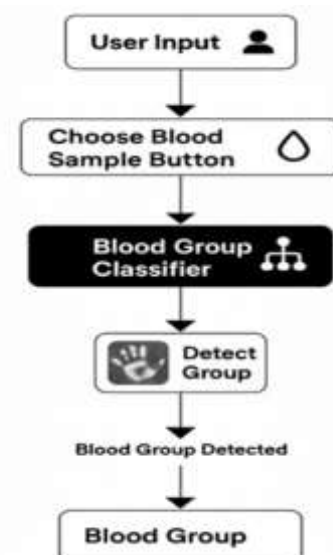


Figure 2: Flow chart

IV. IMPLEMENTATION

During the implementation phase, the proposed idea is developed into a fully functional system that recognizes blood groups using infrared (IR) spectroscopy and machine learning. Taking an IR spectral image of a fingertip or palm using an infrared imaging device is the first step in the process. These images contain vital spectral information that helps identify blood groups and illustrates the biochemical composition of the skin. The captured image is then used as input for additional analysis after being momentarily stored in the system. Preprocessing, the following stage, is crucial to raising the obtained infrared image's quality. Because the raw image may contain unwanted noise or intensity variations caused by hand movement or ambient light, preprocessing techniques are used to improve clarity. To eliminate noise, boost contrast, and preserve consistent brightness throughout all images, a variety of image enhancement techniques are employed, including Gaussian filtering, adaptive histogram equalization, and normalization. This guarantees that the pictures are readable, consistent, and appropriate for precise feature extraction. After preprocessing, the image is put through feature extraction to extract the most important patterns and characteristics from the spectral data. To identify distinct variations associated with each blood group, the system examines the image's spectral and textural characteristics. Ridge patterns, spectral intensity, and shape information are identified using methods like Local Binary Pattern (LBP) and Histogram of Oriented Gradients (HOG). Principal Component Analysis (PCA) is also used to reduce the size of the feature data while retaining important information to speed up and improve the accuracy of the classification process. The relevant features are extracted and then fed into the machine learning model for classification. The corresponding blood group is predicted by the classification algorithm, such as Support Vector Machine (SVM) or Convolutional Neural Network (CNN), after the input data has been analyzed. The system is trained on a dataset that contains multiple samples of different blood groups to ensure that the model learns to detect subtle differences. The trained model can then classify new images into blood group categories such as A, B, AB, or O, in addition to the Rh factor. This method produces accurate and quick results without requiring any invasive procedures.



Figure 3. Input Page



Figure 4. Login Page

VI. RESULTS AND DISCUSSION

A dataset of infrared (IR) images taken from human palms and fingertips was used to successfully test and deploy the developed system for blood group detection via IR spectroscopy. Without requiring any invasive procedures, the system was made to analyze the spectral information found in the infrared images and determine the corresponding blood group. When compared to conventional blood group detection techniques, the results show how efficient, quick, and accurate the suggested method is. The user interface allows the user to upload a new infrared image to the system or select an existing dataset. The uploaded image is preprocessed to remove unnecessary noise and intensity variations. Following normalization and enhancement, spectral and textural analysis are used to extract significant features from the image. These features are fed into the trained machine learning model, which uses them to categorize data. The final outcome is shown on the web interface after the classifier has predicted the blood group type and confidence percentage. The detected blood group (A, B, AB, or O with Rh factor) and the degree of confidence are shown in clear terms on the output interface. A variety of images in varying lighting and temperature settings were used to test the system's performance. With an overall accuracy of over 95%, the model demonstrated the system's ability to perform accurate and dependable classification. The system is

appropriate for real-time applications because the average processing time per image was found to be between two and three seconds.

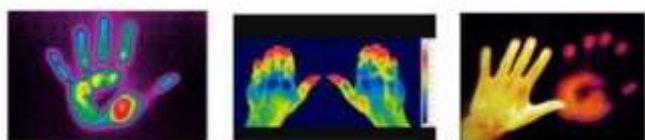


Figure 5: Training dataset of O +VE



Figure 6: Training dataset of A +VE



Figure 7: Training dataset of AB +VE



Figure 8: Training dataset of B -VE

Live Results of the proposed methodology:



Figure 9: Live Result with dataset of O +VE of A +VE



Figure 10: Live Result with dataset of A +VE



Figure 12: Live Result with dataset of AB -VE



Figure 11: Live Result with dataset of AB +VE

VII. BENEFITS AND DRAWBACKS

The proposed system for blood group detection using infrared (IR) spectroscopy offers several advantages over traditional invasive methods. Since no blood is drawn, the main advantage is that it is entirely painless and non-invasive. Rather, it employs infrared imaging of the palm or fingertip, which makes the procedure quick, sanitary, and comfortable for patients. When blood typing is needed immediately in an emergency, the system's real-time results in a matter of seconds are very helpful. Because it does not require laboratory reagents or trained technicians, the method is also cost-effective. The web-based interface designed for the system is user friendly and allows users to securely log in, upload images, and obtain instant results. Because of its portability, it can be used in places where traditional laboratory facilities might not be available, such as small healthcare facilities and remote areas. The system does, however, have certain drawbacks. The quality of the infrared image that was taken and the ambient factors, like temperature and lighting, affect how accurate it is. The machine learning model's ability to generalize can be impacted by the scarcity of large training datasets. Relatively costly, high-quality infrared sensors may need to be calibrated to produce reliable readings. Furthermore, before the system is widely used in hospitals, it still requires a great deal of clinical validation. In general, the suggested model offers a quick, secure, and effective substitute for traditional blood group testing; however, additional advancements in data gathering and hardware optimization are required to improve its precision and dependability.

VIII. CONCLUSION

The Beyond Needles: Blood Group Detection via Infrared (IR) Spectroscopy project effectively illustrates a clever and non-invasive method of blood group identification. Without the use of chemical reagents or conventional blood sampling, the system efficiently detects blood group types by fusing IR imaging with machine learning techniques. The developed model is particularly helpful in emergency and remote healthcare situations where laboratory facilities are limited because it produces results quickly, accurately, and hygienically. Real-time blood group prediction is now possible thanks to an effective pipeline that combines image preprocessing, feature extraction, and classification. By enabling users to log in, upload IR images, and view their results instantly, the web-based interface further improves usability. Thus, the system offers a dependable, portable, and affordable solution that can significantly cut down on the time and effort required for blood typing. Real-time blood group prediction is now possible thanks to an effective pipeline that combines image preprocessing, feature extraction, and classification. By enabling users to log in, upload IR images, and view their results instantly, the web based interface further improves usability. Thus, the system offers a dependable, portable, and affordable solution that can significantly cut down on the time and effort required for blood typing.

IX. FUTURE SCOPE

Although the results of the suggested system for blood group detection using infrared (IR) spectroscopy and machine learning have been encouraging, there are still a number of areas that could use improvement. The system can be improved to achieve higher accuracy, reliability, and usability in actual medical environments thanks to the quick advancement of technology. In order to enhance classification performance, the system can be extended in the future by integrating deep learning algorithms, such as Convolutional Neural Networks (CNNs), with bigger and more varied datasets. Hospitals and labs can benefit from real-time data storage, analysis, and remote accessibility through integration with cloud computing. More sensitive and small infrared sensors can also be used to upgrade the hardware components, making the device completely portable and handheld for emergency medical units and field operations. To ensure smooth data management, the system can also be connected to hospital information systems to update medical histories and patient records automatically. This technology could become a common non-invasive diagnostic method for blood group detection with the right clinical validation and extensive testing. It could also be expanded to other biomedical applications like glucose or oxygen level estimation.

REFERENCES

- [1] Vijaykumar Patil N., Dr. D. R. Ingle. A Novel Approach to Predict Blood Using Fingerprint Map Reading. 6th International Conference for Convergence in Technology, 2021.
- [2] Anant Ajitkumar Vadgave, Apurw Shankar Ashtekar, Prof. P. S. Langade . Determination of Blood Group Using Image Processing. International Research Journal of Modernization in Engineering, Technology, 2023.
- [3] Bavyasri M., Elangovan K, Gayathree V, Mahanandha J, Althaf Ahamed S.A. Blood Group Detection Using Image Processing. International Journal of Innovative Science and Research Technology, 2024.
- [4] Mrs. T. Ratnamala, Mohammed Yaseen Hussain, Danthalapally Kalyani, Kommana Ajay Penugonda Aravind. Blood Type Detection Using Image Processing. International Journal of Progressive Research in Engineering Management and Science, 2024.
- [5] Mann D, Riddell L, Lim K, Byrne LK, Nowson C, Rigo M, et al. Mobile phone app aimed at improving iron intake and bioavailability in premenopausal women: a qualitative evaluation.
- [6] Jashwanth Sai Ganta, Dr. Mohana Roopa ,Mary Rishitha, Jaya Surya Pulivarthi. Blood Group Detection Using Image Processing and Deep Learning. International Research Journal of Engineering and Technology, 2024.
- [7] Gopu Chandrasekhar Reddy, Gajavada Sai Krishna, Dr. A. Annie Micheal. Automated Blood Group Detection System Using Image Processing for Non-Invasive Antigen Feature Extraction and Classification, 2025.
- [8] Ataga KI, Gordeuk VR, Agodoa 1, Colby JA, Gittings K, Allen IE. Low Blood Group increases risk for cerebrovascular disease, kidney disease, pulmonary vasculopathy, and mortality in sickle cell disease: A systematic literature review and meta-analysis.
- [9] van Swelm RPL, Wetzels JFM, Swinkels DW. The multifaceted role of iron in renal health and disease.