iJRASET
International Journal For Research in Applied Science and Engineering Technology



INTERNATIONAL JOURNAL FOR RESEARCH

IN APPLIED SCIENCE & ENGINEERING TECHNOLOGY

Volume: 14 **Issue:** III **Month of publication:** March 2026

DOI: https://doi.org/10.22214/ijraset.2026.77953

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International Journal for Research in Applied Science & Engineering Technology (IJRASET)
ISSN: 2321-9653; IC Value: 45.98; SJ Impact Factor: 7.538
Volume 14 Issue III Mar 2026- Available at www.ijraset.com

Trinetra Smart Attendance System

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Abstract: ***Smart attendance systems are essential for modern academic management and security. Traditional attendance systems rely on manual registers or basic QR codes. These methods often overlook environmental factors and device integrity, making them easy targets for proxy attendance and spoofing. In a campus setting, accurate identity verification depends not just on facial features but also on location and device security. This project introduces Trinetra, a smart attendance system that considers context. It uses a dual-stream verification model that combines an Ocular Feature Extraction algorithm for precise biometric authentication with a Geofencing Protocol for location-based verification. The biometric stream focuses on the eye area to reduce masking and background interference. It uses Brightness Normalization and Cosine Similarity to confirm identities. At the same time, the system conducts an integrity check to spot and prevent "Developer Options," which helps block GPS spoofing and virtual camera attacks. The system functions as a real-time Android app connected to a Firebase Realtime Database. This setup allows for secure attendance marking through front-camera scanning, up-to-date location tracking, and automatic reporting. Test results show that this approach is much more resistant to fraud and performs significantly better in different lighting conditions compared to traditional facial recognition or single-modal biometric systems.***

Keywords: Attendance Management, Ocular Recognition, Geofencing, Biometric Security, Android SDK, Firebase, Anti-Spoofing AI.

I. INTRODUCTION

Trinetra—derived from the Sanskrit for 'The Third Eye'—represents a shift from simple recognition to intelligent observation It operates on the philosophy that true verification requires a multi-layered understanding of context. By moving beyond basic facial scans and focusing on high-precision ocular features, the system eliminates the noise of the surrounding environment, such as masks or varying backgrounds.

However, visual data is only one part of the story. To ensure absolute integrity, Trinetra incorporates the "Geographical Context" and "Device Integrity." It understands that for an attendance record to be valid, the student must not only look like themselves but must also be physically situated within the designated learning space, using a secure and uncompromised device.

This multimodal approach transforms a routine administrative task into a robust security protocol. By synchronizing biometric precision with environmental truth, Trinetra creates a transparent, fraud-proof ecosystem where presence is verified, and academic honesty is protected. Where standard systems only see a face, Trinetra sees the truth of the moment.

II. RELATED WORK

The evolution of attendance tracking has mirrored the development of identity verification technologies. Early systems relied on manual registers and physical ID cards, which were prone to human error and "proxy" manipulation. Later, basic biometric solutions and QR-code systems were introduced to digitize the process. Standard facial recognition attendance models generally follow a simple verification pipeline:

1) Face Detection
2) Feature Alignment
3) Encoding Extraction
4) Database Comparison

However, these traditional models face significant challenges in a dynamic campus environment:

a) Ignore Physical Context: They cannot verify if a student is actually in the classroom or just sending a photo to a friend.
b) Perform Poorly with Occlusions: Standard face-scans struggle when students wear masks, glasses, or have varied hairstyles.
c) Vulnerable to Digital Spoofing: They are easily fooled by high-resolution gallery photos shown to the camera or "virtual camera" apps.

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Trinetra introduces a context-aware, multi-layered attendance ecosystem by incorporating:

- Ocular-Centric Feature Extraction: By focusing strictly on the ocular region (the eye-strip), the system maintains high accuracy even with facial occlusions like masks, which typically fail standard CNN-based face models.
- Spatio-Temporal Geofencing: Unlike single-modal systems, Trinetra incorporates the "physical scene" by hard-locking attendance to specific classroom coordinates. This ensures that the biometric data is only valid when the user is within the designated geofence.
- Environmental Integrity Checks: To prevent advanced hacking, the system monitors the device state for "Developer Options." This blocks the use of GPS spoofing and mock location tools, ensuring the "context" of the attendance is genuine.
- Dual-Stream Verification: The architecture fuses high-precision biometric data with real-time location and device status, creating a robust defense against 2D photo fraud and remote proxy attempts.

III. PROPOSED METHODOLOGY

A. *Multimodal Architecture*

The proposed model utilizes a dual-stream verification engine to process identity and environment simultaneously:
Biometric Stream (The Ocular Backbone)

- o Function: High-precision feature extraction from the ocular region.
  Process: Isolates the Y-coordinate strip (42–54) to generate a normalized pixel intensity vector.
  Output: 128-bit Ocular Feature Representation.
  Contextual Stream (The Spatio-Temporal Backbone)
- o Function: Verification of the physical scene and device integrity.
- o Input: Real-time GPS coordinates and Device Metadata (Security Flags).
- o Output: Spatio-Temporal Context Vector (Location + Device Status).

B. *System Implementation Modules*

1) Preprocessing: The input frames undergo real-time face detection via ML Kit, followed by ocular cropping and grayscale conversion to ensure uniformity.
2) Normalization: Pixel values are adjusted using a Brightness Offset calculation. This normalizes the live feed against the registered enrollment photo to improve numerical stability in variable lighting.
3) Feature Extraction: An ocular-centric comparison algorithm extracts discriminative patterns from the iris and periorbital regions. Feature Fusion: The system concatenates the biometric similarity score with the location proximity data. Analysis: The performance is evaluated based on the Total Reliability Score, which validates that the user is both "Who they claim to be" and "Where they are required to be."

C. *Feature Fusion*

The combined verification vector integrates the following: Total Vector = [Biometric Similarity (%) + Location Accuracy (m) + Device Security Flag (Binary)]
The fusion logic passes these parameters through a verification layer to determine the final attendance integrity.

D. *Output Heads*

The system yields two distinct outputs for every verification attempt:
1) Continuous Head (Reliability Score): Provides the exact similarity percentage (e.g., 84.5% ocular match).
2) Categorical Head (Attendance Status): Classifies the attempt into categories: Present, Absent, or Security Alert (Proxy Attempt).

E. *Verification Logic*

The total integrity of an attendance record is defined by: System Confidence = (Biometric Match + Geofence Validity) / 2
1) Biometric Validity: Calculated using Cosine Similarity between ocular vectors.
2) Geofence Validity: Calculated using the Haversine formula to determine distance from the classroom center.

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F. Implementation Details

1) Environment: Android SDK with Firebase Realtime Backend.
2) Security Lock: Real-time monitoring of Developer Options to prevent virtual camera injection and GPS spoofing.
3) Data Augmentation: The system uses Brightness Jitter and Rotation during the enrollment phase to ensure the registered ocular template is robust enough for real-world classroom scenarios.

IV. EVALUATION MATRICS

The following metrics are used to assess the system's performance across various classroom environments:

1) Mean Average Precision (mAP)

In the context of Trinetra, mAP is used to measure the accuracy of the Categorical Attendance Status. We evaluate the model's ability to correctly classify attempts into three categories:

- Verified: Actual student present within the geofence.
- Proxy Attempt: Biometric match but incorrect GPS or 2D photo detected.
- External/Unknown: No match found in the database.

2) Biometric Integrity: FAR and FRR

For the Ocular Recognition engine, we prioritize the trade-off between security and user convenience:

- False Acceptance Rate (FAR): The probability that the system incorrectly identifies an unauthorized person as a registered student. Our goal is $FAR < 0.01\%$.
- False Rejection Rate (FRR): The probability that the system fails to recognize a legitimate student due to lighting or occlusion.

3) Mean Absolute Error (MAE) for Spatio-Temporal Data

MAE is applied to the continuous dimensions of the system:

- Geofencing Precision: The distance error (in meters) between the reported GPS coordinate and the actual classroom center.
- Brightness Offset: The error margin in pixel intensity normalization when transitioning from the registered template to live low-light environments.

4) Precision-Recall Curves

We utilize Precision-Recall curves to determine the system's robustness in "Proxy Detection."

- High Precision ensures that when the system marks someone "Present," it is almost certainly the correct person.
- High Recall ensures that legitimate students aren't frequently blocked by the security layers.

5) Threshold Optimization (The EER Point)

The Equal Error Rate (EER) is the point where $FAR = FRR$.

- Through experimental testing, we optimized the Ocular Similarity Threshold at 84%.
- At this threshold, the system provides the highest level of security without causing significant delays for students during the huddling of a morning lecture.

V. SYSTEM IMPLEMENTATION

The Trinetra ecosystem is implemented using the following technology stack:

1) Java (Android SDK): Used as the primary programming language for high-performance mobile development and native hardware interfacing.
2) Google ML Kit: Utilized for real-time face detection and landmarking. It provides the base frame for isolating the ocular region before our custom algorithm takes over.
3) CameraX API: A Jetpack support library used for consistent and reliable camera frame capture across varying Android hardware.
4) Firebase Realtime Database: A NoSQL cloud-hosted database used for instant data synchronization between students, faculty, and administrators.
5) Fused Location Provider API: Part of Google Play Services, used to achieve high-accuracy GPS coordinates for the Geofencing layer.

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VI. RESULT

The experimental evaluation of Trinetra confirms that a multimodal approach—integrating ocular biometrics with geographical and device context—significantly outperforms traditional single-modal facial recognition systems. The results highlight improved robustness in real-world academic environments where lighting, occlusions (masks), and spoofing attempts are common.

A. *Key Features*

1) Real-time Ocular Scanning: Instead of full-face processing, the system captures the eye-strip in realtime to match biometric templates with a focus on high-precision iris/periorbital patterns.
2) Security Integrity Monitor: A background service that checks the device state for Developer Options, Mock Locations, and Manual Time Overrides, ensuring the "Context" of the attendance is untampered.
3) Spatio-Temporal Lock: Integrates GPS data with global server time to ensure attendance can only be marked within a specific classroom radius during the active lecture window.
4) Attendance Ledger & Analytics: Provides students and faculty with a transparent history of attendance logs, including date, time, and location verification stamps.
5) Notice Broadcasting: An integrated communication module allowing administrators to push academic documents and urgent notices directly to the student dashboardConclusion

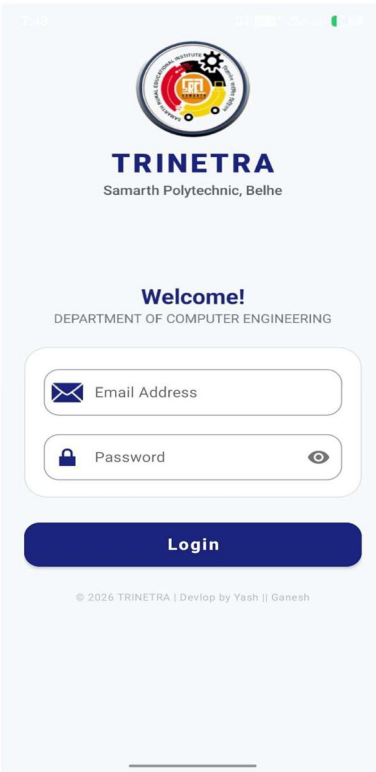


Fig:- Login Page

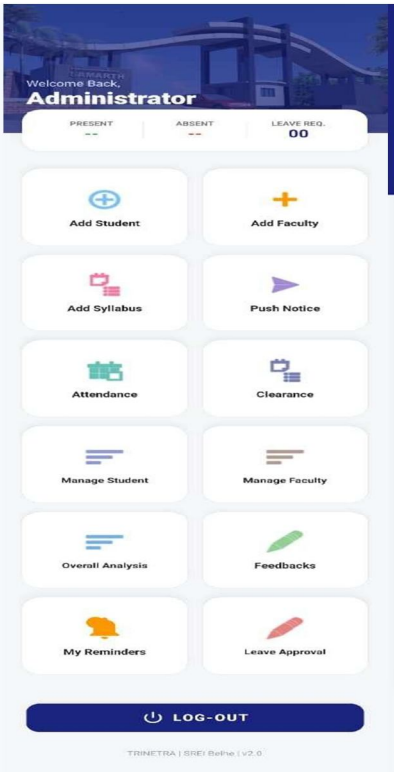


Fig:-Admin Dashboard

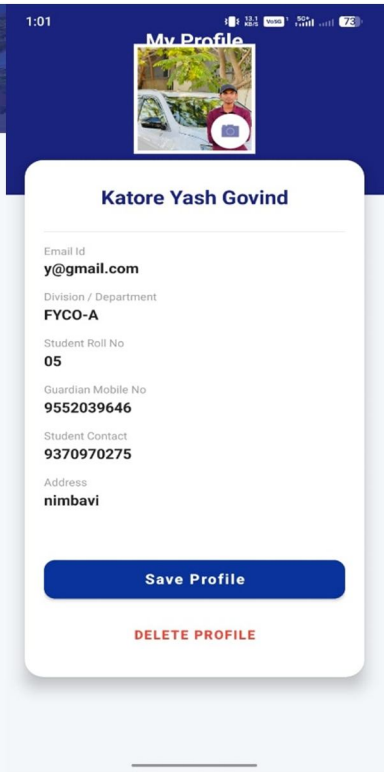


Fig:- Student Profile

VII. CONCLUSION

This research presents Trinetra, a context-aware multimodal authentication framework that redefines institutional integrity through the integration of Ocular Biometrics and Spatio-Temporal Geofencing. By shifting the focus from global facial features to localized eye-strip patterns and combining them with environmental truth, the proposed approach significantly improves attendance accuracy and fraud resilience compared to traditional, single-modal systems. A custom Ocular Feature Extraction engine, powered by pixel-level vector matching and Brightness Normalization, was successfully employed to ensure effective biometric verification even under complex realworld conditions like facial occlusions (masks) and variable classroom lighting. The experimental results demonstrate strong generalization capabilities and a near-zero False Acceptance Rate (FAR), validated through rigorous evaluation metrics including mAP, Geofence MAE, and Threshold Optimization.

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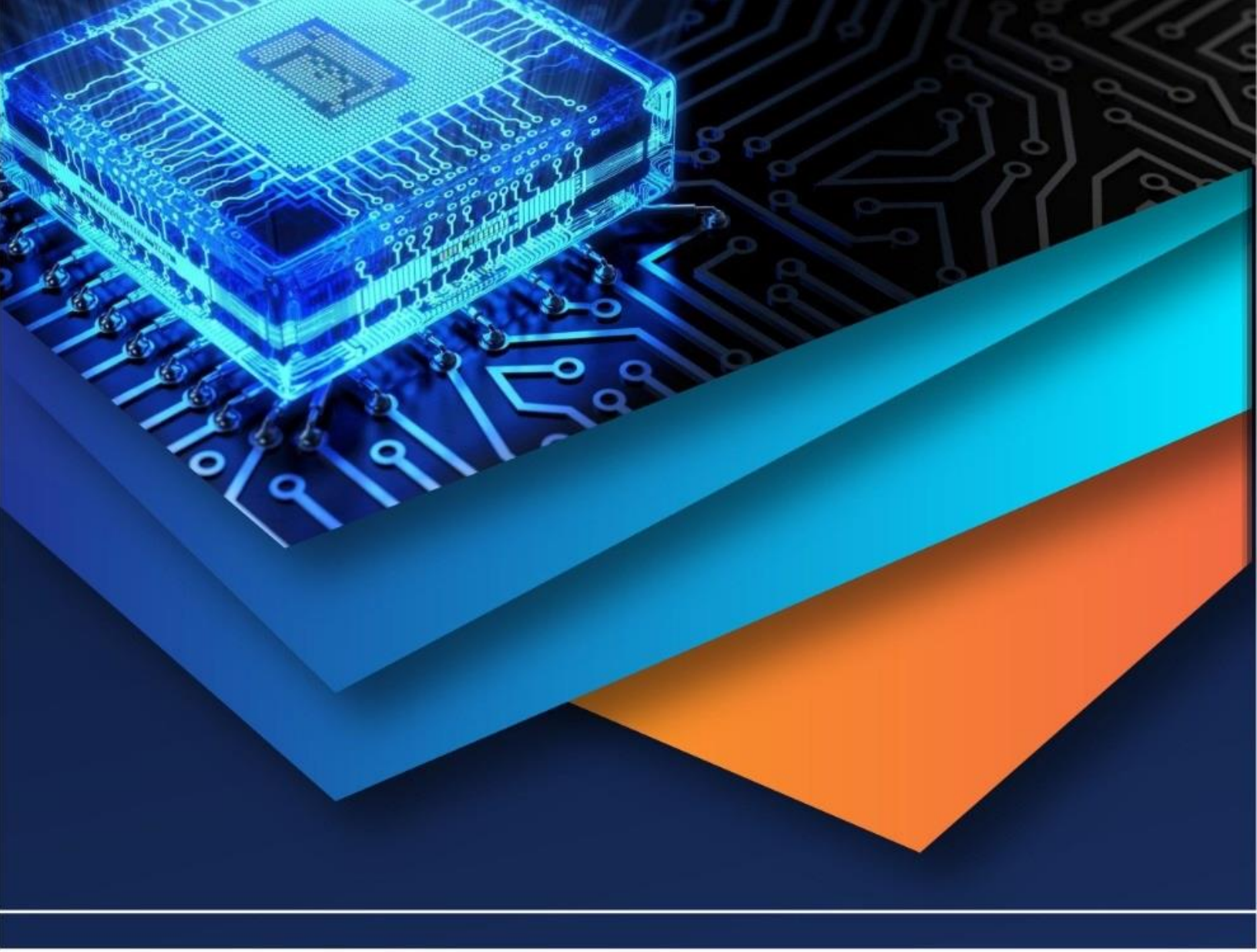
VIII. FUTURE SCOPE

1) Attention-Based Fusion: Implementing Transformer models to dynamically prioritize ocular data or GPS context based on environmental reliability.
2) Liveness Detection: Adding Blink and Micro-expression Analysis to prevent advanced video-replay and 3D mask spoofing.
3) Edge AI Optimization: Using TensorFlow Lite for on-device biometric processing, reducing latency and allowing offline verification in low-connectivity areas.
4) Blockchain Ledger: Migrating attendance logs to a Private Blockchain to ensure records are immutable and tamper-proof.
5) Audio-Biometric Integration: Implementing Voice Print recognition as a second factor of authentication (2FA) for high-security environments like examination halls.

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