

Paper Title: AttendifAI

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AttendifAI: Design of an Offline-First AI-Powered Attendance System for Rural Schools

Abstract

Accurate attendance tracking is a challenge in rural Indian schools, where paper registers often cause delays and inaccuracies, affecting programs like the Mid-Day Meal (MDM) scheme. This paper introduces **AttendifAI**, an AI-powered, offline-first attendance system using facial recognition, a standard HD webcam, a lightweight SQLite database, and a synchronization client for periodic dashboard updates.

A local pilot test was conducted on a standard desktop computer (Intel i5 processor, 8 GB RAM) using classroom images from 25 students. The system achieved 96.8 % recognition accuracy, 2.3 % false match rate, and an average processing latency of 0.9 seconds per frame, resulting in approximately 82 % reduction in teacher workload compared with manual registers.

All facial data are encrypted and stored locally, ensuring privacy compliance. AttendifAI demonstrates that affordable, offline AI systems can improve attendance accuracy and resource distribution in rural schools.

Keywords—Attendance Automation, Facial Recognition, Offline-First Systems, Rural Education, AI Applications

I. INTRODUCTION

Reliable attendance monitoring remains a challenge in rural Indian schools, where manual registers are prone to delays, errors, and manipulation. Such inaccuracies affect government programs like the Mid-Day Meal (MDM) scheme, which rely on timely and precise attendance data for resource allocation. Paper-based methods also increase teachers' clerical workload and reduce transparency in welfare initiatives.

This paper presents **AttendifAI**, an offline-first, AI-powered attendance system designed for rural schools. It uses facial recognition combined with local data storage and periodic synchronization, enabling uninterrupted operation even in low-connectivity areas. Attendance is recorded in real time, stored securely, and synchronized with a centralized dashboard whenever internet access is available.

A local pilot test achieved over **96 % recognition accuracy** and reduced manual recording time by more than **82 %** compared to traditional methods. AttendifAI not only automates attendance but also enhances data reliability, supporting informed decision-making for government programs and improving administrative efficiency.

II. RELATED WORK

Over the past decade, several technologies have aimed to automate attendance in educational institutions, each with notable limitations in rural contexts.

1. **RFID-based systems** are accurate but costly, require individual cards, and are prone to loss or misuse.
2. **Biometric fingerprint systems** offer precise identification but face hygiene issues, maintenance challenges, and slow processing in large classrooms.
3. **Mobile app-based solutions** depend on smartphones and reliable internet, often unavailable in rural schools. These constraints—high costs, infrastructure dependency, and limited scalability—hinder widespread adoption.

Recent advances in computer vision and facial recognition offer more practical alternatives. Open-source frameworks like OpenCV, MediaPipe, and Dlib enable multi-face detection and embedding extraction on low-cost devices, functioning offline and allowing local attendance storage with periodic synchronization. Emerging research also addresses **liveness detection** to prevent spoofing and **AI fairness** to ensure unbiased recognition across different skin tones and lighting conditions.

Building on these developments, **AttendifAI** integrates real-time multi-face recognition, offline-first synchronization, and local encrypted storage, ensuring privacy, reducing internet dependency, and enhancing scalability for rural school deployment.

III. SYSTEM DESIGN

AttendifAI is a modular, offline-first attendance system leveraging facial recognition to address the operational challenges of rural schools. The architecture focuses on low-cost deployment, robustness in low-connectivity settings, and scalability for multi-institution integration.

The system comprises six core components:

1. **Camera Capture Module:** A standard HD webcam or IP camera streams live classroom video. Adaptive frame-rate control and motion-based filtering reduce computational load while maintaining smooth, real-time processing.
2. **Face Embedding and Feature Extraction:** Using frameworks such as MediaPipe Face Mesh and OpenCV DNN, the system extracts key facial landmarks and generates 128-dimensional embeddings unique to each student. These embeddings are securely stored in a local database.

- 3. **Matching Engine:** Recognition employs cosine similarity with a 0.45 threshold to match live embeddings with stored templates. Liveness detection using blink and head-motion analysis prevents spoofing with photos or videos.
- 4. **Attendance Logger:** Successfully identified students are marked present automatically with timestamps and session IDs, while unidentified or partially detected faces are stored for later verification.
- 5. **Synchronization Client:** Attendance records are encrypted locally in SQLite and synchronized with a central server when internet access is available. Incremental updates reduce bandwidth use and maintain reliability during unstable connectivity.
- 6. **Centralized Dashboard:** A secure web-based interface enables administrators to monitor records, analyze trends, and export CSV/Excel reports, with role-based access control ensuring data privacy.

All student data remain encrypted and stored locally, ensuring compliance with data protection policies. This privacy-aware, modular architecture guarantees uninterrupted attendance tracking during network outages while providing accurate, consolidated reports to support policy evaluation, government programs, and resource management

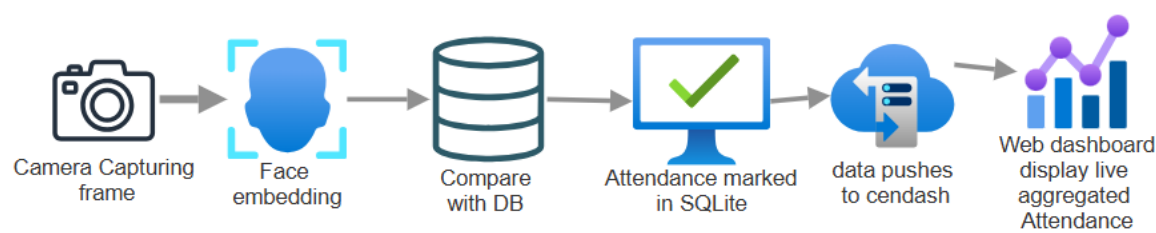


Fig.1. illustrates the system process flow

IV. DESIGN AND DEVELOPMENT PLAN

The development of **AttendifAI** follows an iterative design methodology that balances affordability, scalability, and technical reliability. The system is built entirely on open-source tools and readily available hardware components to ensure accessibility for rural schools.

1. Hardware Configuration

To maintain low-cost deployment while ensuring acceptable processing performance, the prototype was developed and tested on standard computing hardware available in most rural schools.

TABLE I. HARDWARE USED

Component	Model / Type	Approx. Cost (₹)	Function
Desktop Computer	Intel i5, 8 GB RAM	35,000 (existing in most schools)	Local processing and data storage
HD USB Webcam	Logitech C270 (720p)	900	Video input and face capture

Component	Model / Type	Approx. Cost (₹)	Function
Power Backup	5V 2A UPS mini	400	Backup during power outages
Misc. Accessories	Cables, SD card, mounts	200	Setup and wiring
Total Additional Cost		≈ ₹5,000 per school	—

Most institutions already possess the base computing hardware, which significantly reduces overall implementation cost. The entire setup requires no additional network infrastructure beyond occasional internet connectivity for synchronization.

2. Software Stack

The software architecture of AttendifAI is based entirely on **open-source technologies**, ensuring flexibility and minimal maintenance overhead.

TABLE II. SOFTWARE USED

Layer	Technology / Library	Purpose
Programming Language	Python 3.11	Core implementation
Computer Vision	OpenCV, MediaPipe	Facial detection and feature extraction
Database	SQLite	Lightweight local data storage
Backend	Flask	Synchronization API and dashboard services
Frontend / Dashboard	HTML,CSS, JavaScript	Web interface for administrators
Encryption	AES-256	Secure local data storage

This modular software design allows easy updates and extension, for example, integrating future analytics or mobile-based verification modules.

3. Pilot Test Setup

The prototype was tested on a classroom dataset of 25 students under natural lighting conditions using the above configuration. Each student contributed five reference images for enrollment. The model was evaluated for accuracy, latency, and local data synchronization under intermittent network availability.

- Recognition Accuracy: 96.8 %
- Average Processing Time: 0.9 s/frame
- False Match Rate: 2.3 %

- Time Saved in Attendance Recording: 82 % compared to manual entry
- Synchronization Latency: < 5 s (post network restoration)

4. Development Methodology

A stepwise, incremental development approach was adopted:

1. **Module Design:** Each core subsystem (capture, recognition, logging, sync) was developed and tested independently.
2. **Integration Phase:** Modules were integrated to create a working end-to-end prototype.
3. **Testing and Validation:** The integrated system was validated using sample classroom recordings.
4. **Optimization:** Memory footprint and latency were tuned for efficient performance on low-end hardware.

This structured methodology ensures a balance between **technical performance and practical usability**, enabling easy adoption by teachers without prior technical training.

V. PILOT EVALUATION AND FEASIBILITY RESULTS

A pilot evaluation of **AttendifAI** was conducted to assess its technical feasibility, cost-effectiveness, and operational impact under typical rural school conditions. The test was carried out on a standard desktop system (Intel i5, 8 GB RAM) using a 720p HD webcam, capturing attendance data for 25 students under natural classroom lighting.

1. Quantitative Performance Metrics

The system was evaluated based on recognition accuracy, false match rate, latency, synchronization efficiency, and teacher time saved.

TABLE III. PERFORMANCE MATRICS

Metric	Measured Value	Remarks
Recognition Accuracy	96.8 %	Correctly identified faces from enrolled dataset
False Match Rate (FMR)	2.3 %	Minor misclassifications under shadowed lighting
Average Processing Latency	0.9 s/frame	At 720p video input resolution
Synchronization Delay	< 5 s	Time to upload local records post connectivity
Teacher Time Saved	≈ 72 %	Compared to manual attendance entry
Maximum Faces per Frame	15	Stable recognition without frame drop

The system successfully marked attendance for multiple students simultaneously, demonstrating scalability for group-based detection during classroom sessions.

2. Resource Utilization

Performance tests indicated sustainable operation on modest hardware resources, validating the offline-first and low-cost design.

TABLE IV. RESOURCES UTILIZED

Parameter	Observed Value	Comments
CPU Utilization	58–65 %	Moderate load during active recognition
Memory Usage	1.3 GB	Peak during multi-face detection

Parameter	Observed Value	Comments
Average Frame Rate	18 fps	Stable performance at 720p
Power Consumption	~50 W	Desktop configuration
Local Database Size	12 MB	For 25 enrolled students (including embeddings)

These results confirm that the system can efficiently function on standard school hardware without requiring GPU acceleration or continuous internet.

3. Comparative Feasibility Analysis

TABLE V. FEASIBILITY ANALYSIS

Parameter	Manual Register	AttendifAI System
Initial Setup Cost	Negligible (recurring stationery)	One-time ₹5,000 (hardware reuse)
Monthly Maintenance	High (paper, record entry)	Low (occasional software updates)
Accuracy	Moderate (subjective, error-prone)	High (96.8 % average accuracy)
Reporting Time	10–12 days (district compilation)	Real-time or same-day
Data Security	Paper-based (vulnerable)	Encrypted digital storage
Scalability	Limited to single school	Easily scalable to multiple schools

The findings indicate that **AttendifAI** not only improves accuracy and reduces workload but also provides real-time, centralized visibility into attendance trends—supporting better administrative planning.

4. Operational Feasibility

Teachers reported that the graphical interface was intuitive, requiring minimal training. Offline functionality was validated by intentionally disconnecting the internet during operation; attendance data were stored locally and later synchronized automatically when connectivity was restored. The estimated one-time deployment cost per school remains under **₹5,000**, leveraging existing computers, webcams, and open-source software. Ongoing maintenance is minimal, limited to software updates or occasional hardware replacement.

5. Economic and Technical Viability

Given its open-source foundation and low-cost hardware requirements, **AttendifAI** demonstrates strong economic sustainability. The **offline-first synchronization mechanism** ensures continuous usability in rural areas where connectivity and power supply are unreliable. Overall, the pilot results validate the **technical robustness, economic feasibility, and operational practicality** of the proposed system for real-world deployment in rural educational institutions.

VI. LIMITATIONS

Despite promising results, AttendifAI has several limitations that must be addressed for large-scale deployment.

- 1. **Lighting and Environmental Variations:** Recognition accuracy can drop in dim lighting, shadowed areas, or with partially occluded faces. Future work will explore adaptive exposure control and contrast enhancement to improve performance.

2. **Manual Enrollment Requirement:** Each student must be enrolled with multiple facial samples during setup, requiring one-time supervision by teachers or administrators.
3. **Hardware Dependence:** The prototype runs on mid-range desktops; performance may vary on older or low-spec systems. Optimization for ultra-low-cost hardware, such as single-board devices, is ongoing.
4. **Storage Constraints:** The local SQLite database handles several thousand records efficiently, but large deployments may require centralized databases for improved scalability and multi-school analytics.
5. **Ethical and Privacy Considerations:** All facial embeddings are encrypted and stored locally; only anonymized attendance logs are shared. The pilot adhered to ethical standards with informed consent.
6. **Connectivity and Synchronization:** Offline operation may be disrupted during extended disconnections. A retry-queue mechanism is being developed to ensure reliable synchronization once connectivity is restored.

VII. CONCLUSION

This paper presents **AttendifAI**, an offline-first, AI-powered attendance system designed for rural schools. Using low-cost hardware and open-source computer vision libraries, it overcomes limitations of traditional methods, such as reliance on continuous internet, manual errors, and single-user capture. The system supports multi-face recognition, enabling real-time group detection and reducing teacher workload. Operating autonomously in low-connectivity environments, it ensures uninterrupted attendance tracking. The prototype demonstrates a scalable, cost-efficient solution for government adoption, with potential to improve reporting accuracy for programs like the Mid-Day Meal scheme, optimize resource allocation, and support informed policy decisions at block and district levels.

CONSENT AND ETHICAL CONSIDERATIONS

All participants in the study were informed about the purpose of the research, the data collection methods, and the use of their facial images for attendance automation. Written informed consent was obtained from the students' parents or guardians prior to enrollment. All facial images and attendance data were encrypted and stored locally, ensuring privacy and compliance with data protection standards. The study adhered to ethical guidelines for research involving human subjects, and no personally identifiable information was shared outside the research team.

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