

**CSAI 498 / CSAI 499**

**Initial System Design & Work Division**

**Smart Attendance System**

**• Team Members:**

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**• Semester / Year:**

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**• Team number:**

32

**• Submission Date:**

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## 1. Project summary:

This project introduces a **Smart Attendance System** that replaces manual attendance with an automated, AI-based solution.

Instead of analyzing full video streams, the system captures a **single image every minute**, detects and recognizes students, and calculates each student's **actual presence duration** based on how many frames they appear in.

Attendance records are stored in a **centralized database** and visualized through a **web/app dashboard**, allowing instructors to easily track attendance summaries and class insights in real time.

The system reduces manual effort, eliminates errors, and provides a more accurate measurement of student presence. Future enhancements include enabling **single-image student registration** to simplify onboarding and improve scalability.

## 2. Progress Since Proposal:

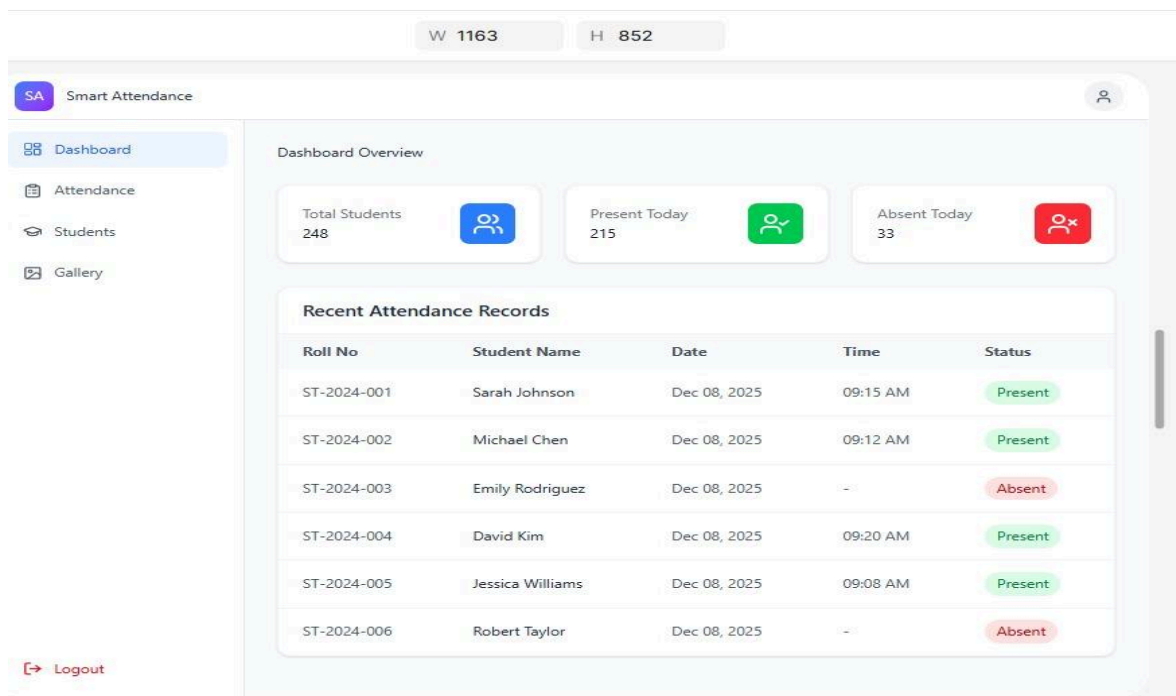
### 2.1 Completed Tasks (Since Week 4)

- **Data Collection:** All student images were collected and organized into structured class folders.
- **Image Preprocessing:** Cropped individual faces from the collected images using detected bounding boxes, added padding, and generated a clean dataset of face images.
- **Data Labeling:** Each cropped face was labeled with the corresponding student's **Name** and **ID** to prepare the dataset for training.
- **Model Training Started:** Initiated training of the **ResNet-based face recognition model** to learn student identities and return the correct Name/ID from an input image.
- **UI Development:**
  - Designed and implemented the **login page**.
  - Built the **dashboard** displaying attendance records.
  - Created the **student page**.
  - Developed preliminary **UI prototypes** outlining the overall system flow.
- **Repository Setup:** Initial project structure and commits were pushed to the GitHub repository.

- **Basic System Pipeline Testing:** Verified that the main pipeline works end-to-end:
  - Load image → detect face → crop → generate embedding → return identity prediction.

## 2.2 Research, Design, and Initial Coding

- Implemented a **two-step face detection pipeline** using **DNN SSD** as the primary detector with **Haar Cascades** as a fallback for missed detections.
- Designed the preprocessing workflow, including bounding box extraction, padding, cropping from the original image, and organizing the processed images into labeled folders.
- Started training the **ResNet-based face recognition model** to classify identities based on learned embeddings.
- Tested the full pipeline across multiple sample images to ensure detection, cropping, embedding, and prediction operate correctly.
- Developed initial UI components and **early mockups**, including the login interface, dashboard structure, and student attendance page.



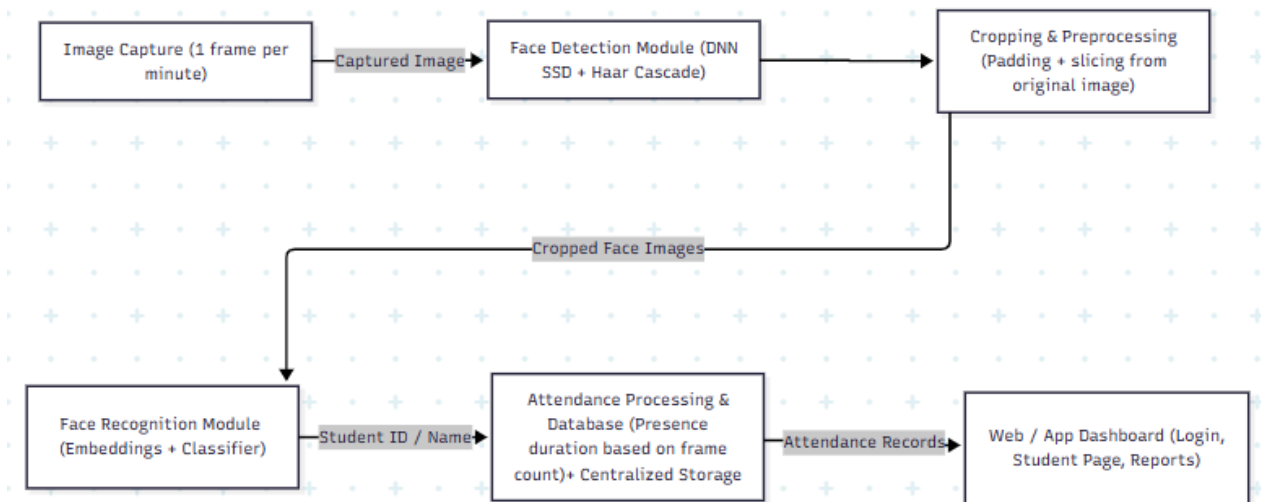
**Figure 1:** Smart Attendance System dashboard showing attendance statistics and recent records

### 3. System Architecture:

The Smart Attendance System is designed as a modular pipeline that processes periodically captured images, identifies students, and updates attendance records in a centralized database.

The system consists of five main components:

1. **Image Capture Module**
2. **Face Detection Module**
3. **Face Recognition Module**
4. **Attendance Processing & Database Module**
5. **Web/Application Dashboard**



**Figure 2:** *System Architecture Diagram*

## 4.Component Breakdown:

Component	Function	Input	Output	Main Functionality	Technologies Used
<b>1. Image Capture Module</b>	Periodically capture classroom images	Camera snapshot (every 1 min)	Raw image frame	Triggers timed image capture and forwards it to detection module	Python, OpenCV
<b>2. Face Detection Module</b>	Detect faces in each captured image	Raw image	Bounding boxes of detected faces	Uses DNN SSD to detect faces; Haar Cascades as fallback when SSD fails	DNN SSD (OpenCV DNN), Haar Cascades
<b>3. Preprocessing &amp; Cropping Module</b>	Prepare clean face crops	Bounding boxes + original image	Cropped face images	Apply padding, crop faces, resize, organize images for recognition	NumPy, OpenCV
<b>4. Face Recognition Module</b>	Identify student identity	Cropped face image	Student Name + ID	Generate embeddings using ResNet model and classify identity	ResNet, PyTorch / TensorFlow
<b>5. Attendance Processing Module</b>	Calculate presence duration & store records	Student ID per frame	Attendance entries	Count frames where each student appears, compute attendance duration	Python, Custom Logic
<b>6. Centralized Database</b>	Store student data & attendance logs	Attendance records	Structured data	Maintain secure, real-time attendance storage & retrieval	Firebase / MySQL
<b>7. Web/App Dashboard</b>	Visualize attendance and manage system	Database data	Front-end pages	Show login, attendance summaries, student pages, and class reports	HTML/CSS, Flask / Bootstrap
<b>8. User Authentication Module</b>	Secure login to dashboard	Credentials	Authenticated user session	Validate users and restrict access	Flask Auth / JWT

## 5. Data Flow:

### Image Capture → Face Detection

- Every minute, the Image Capture Module takes a classroom image.
- The raw image is sent to the Face Detection Module.

### Face Detection → Preprocessing

- DNN SSD detects faces and returns bounding boxes.
- If SSD fails, Haar Cascades provide fallback detections.
- Bounding boxes and the original image are passed to the Preprocessing Module.

### Preprocessing → Face Recognition

- The Preprocessing Module crops each face with padding and resizes it.
- Clean, labeled face crops are forwarded to the Face Recognition Module.

### Face Recognition → Attendance Processing

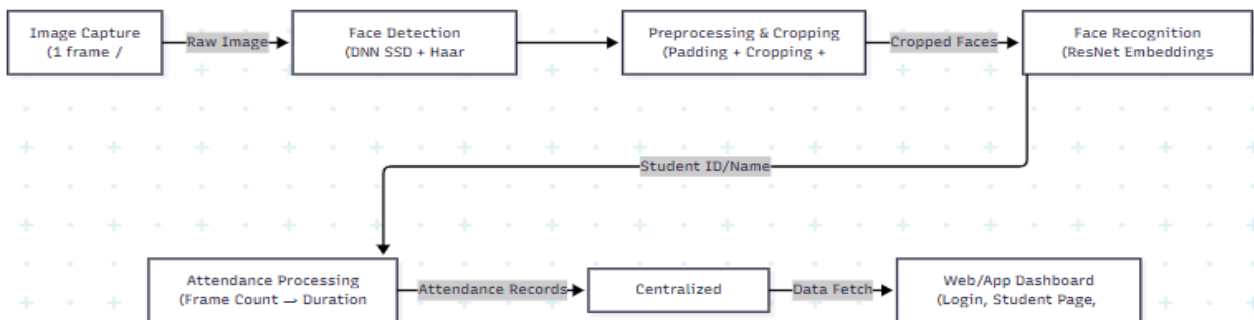
- The ResNet-based model generates an embedding for each face and identifies the student (Name + ID).
- Identified students are passed to the Attendance Processing Module.

### Attendance Processing → Database

- The system counts how many frames each student appears in.
- Presence duration is computed and stored in the centralized database as attendance records.

### Database → Dashboard

- The dashboard retrieves attendance data from the database.
- Displays summaries, individual student records, and class statistics to instructors.



**Figure 3:** System Data Flow Diagram

## 6. Work Breakdown Structure (WBS):

Task	Responsible Team Member	Start Date	End Date	Status
<b>Data Collection (Student Images)</b>	All Members	Week 4 fall	Until now	In Progress
<b>Face Detection Pipeline (SSD + Haar)</b>	Hania	Week 10	Week 3 spring	In Progress
<b>Image Preprocessing (Cropping + Padding)</b>	Farah	Week 10	Week 14	In Progress
<b>Dataset Labeling (Names + IDs)</b>	Hager	Week 10	Week 14	In Progress
<b>Face Recognition Model Setup</b>	Farah, Hager, Hania	Week 11	Until now	In Progress
<b>Model Training &amp; Evaluation</b>	Farah, Hager, Hania	Week 11	Until now	In Progress
<b>UI Development</b>	Yassmin	Week 11	Until now	In Progress
<b>Backend Integration (Recognition → Attendance Logic)</b>	All Members	Week 3 spring	End of spring	Planned
<b>Attendance Processing Module</b>	Farah, Hager, Hania	Week 1 spring	Week 6 spring	Planned
<b>Database Setup (Students + Attendance)</b>	Yassmin	Week 1 spring	Week 6 spring	Planned
<b>Dashboard Data Visualization</b>	Yassmin	Week 1 spring	Week 6 spring	Planned



<b>Final Testing (End-to-End Pipeline)</b>	ALL Members	Week 6 spring	Week 12 spring	Planned
<b>Deployment Setup (Local/Web)</b>	Yassmin	Week 13 fall	Week 9 spring	Planned
<b>Documentation &amp; Report Writing</b>	ALL Members	Week 4 Fall	Week 13 spring	In Progress

## 7.Risk Analysis:

Risk Type	Description	Impact	Mitigation Strategy
<b>Technical</b>	Low accuracy in face recognition due to poor image quality or occlusions	High	Use high-quality image dataset, implement padding, data augmentation, and ResNet embeddings.
<b>Technical</b>	Face detection failures in certain lighting conditions	Medium	Two-stage detection: DNN SSD as primary, Haar Cascade as fallback; preprocess images to normalize lighting
<b>Timeline</b>	Delays in UI or model integration	High	Parallelize module development, schedule weekly progress checks, prioritize critical components first
<b>Resource</b>	Team members unavailable due to personal reasons	Medium	Reassign tasks, maintain clear documentation, implement backup responsibilities
<b>Data</b>	Incomplete or inconsistent dataset	High	Validate data collection, standardize preprocessing, and double-check labeling accuracy

## 8. References:

- 1] Khaled, K. (n.d.). *Computer vision playlist* [Video playlist]. YouTube.  
[https://youtube.com/playlist?list=PL5JZLxl\\_tFCeB5bOLYGXNdjnmgRIU6Ov0](https://youtube.com/playlist?list=PL5JZLxl_tFCeB5bOLYGXNdjnmgRIU6Ov0)
- 2] OpenCV Team. (n.d.). *OpenCV library*. <https://opencv.org/>