

Face Recognition-Based Attendance System Using InsightFace and Pinecone

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Meet the Developer: Arif Mallick

"Code what matters. Automate what doesn't."

Hi, I'm Arif Mallick — a passionate Al/ML Engineer currently interning at DataCore Systems , Kolkata. I specialize in building intelligent, real-time systems for vision, language, and automation tasks. From NLP pipelines to aerial image analytics, I love transforming complex problems into deployable tech.

My latest creation?

A real-time, contactless attendance system using:

- InsightFace for high-accuracy face recognition
- Pinecone for lightning-fast similarity search
- Flask + HTML/CSS/JS for seamless user interaction
- Sprimized for CPU-only, low-cost hardware like laptops & Raspberry Pi

P Unique Touches:

- Dual-action logging for Entry/Exit tracking (1)
- Confidence threshold tuning for precise recognition @
- Edge-ready with real-time webcam support

"I believe great tech should be simple, scalable, and socially impactful."

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Kank: Top 5% - Amazon ML Challenge 2024, Google Kickstart, Facebook HackerCup

1. Abstract

This project presents a real-time face recognition-based attendance system leveraging state-of-the-art deep learning and cloud-based vector similarity search. The system uses <code>InsightFace's Buffalo_L model</code> for accurate face detection and embedding generation, and integrates with <code>Pinecone</code>, a scalable cloud vector database, to perform fast and efficient face retrieval. Designed for deployment on consumer-grade hardware (e.g., laptops or Raspberry Pi), this system enables contactless and automated attendance marking using webcam input. The end-to-end pipeline includes image acquisition, face detection, 512-dimensional embedding extraction, Pinecone-based similarity search, and user-friendly logging and visualization mechanisms.

Key Technologies:

Face Recognition: InsightFace (BUFFALO L model)

Vector Database: Pinecone (serverless index)

Backend: Flask (Python)

Frontend: HTML/CSS/JavaScript

Unique Contributions:

75% confidence threshold for reliable recognition

Dual-action logging (entry/exit) with duration tracking

Optimized for CPU-only deployment

2. Introduction

Motivation

Manual attendance systems are time-consuming, error-prone, and insecure. There is a need for automated, contactless systems in educational and organizational setups, especially post-pandemic.

Problem Statement

Create a real-time facial recognition system that can efficiently identify individuals and record their attendance using webcam input, even in unconstrained environments.

Scope & Objectives

- Accurate face detection and recognition
- Real-time webcam integration
- Fast identity retrieval using Pinecone
- Store attendance logs locally or in the cloud
- Easy-to-deploy system suitable for both academic and commercial use

3. Research Work

Related Work

- OpenFace, FaceNet: Earlier face embedding models
- DeepFace, Dlib, and VGGFace: Classic systems with higher hardware requirements
- Face Recognition via ArcFace (Deng et al.): Introduced additive angular margin loss for more discriminative embeddings

Improvement Over Prior Art

- Real-time capability on CPU
- Cloud-native vector search using Pinecone (faster than brute-force search)
- Modular system with minimal dependencies and high accuracy

4. Methodology

Model

• Face Detection & Recognition: InsightFace Buffalo_L

o Detector: RetinaFace

o Recognition: ArcFace with ResNet-100 backbone

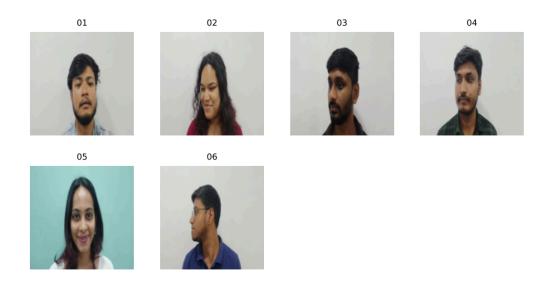
Dataset (Training not required)

• Pretrained model used (Buffalo L from InsightFace)

• Face embeddings extracted directly from camera images

Preprocessing

- Resize to standard input (112x112)
- Normalize RGB values
- Convert OpenCV BGR to RGB



More or less data sample

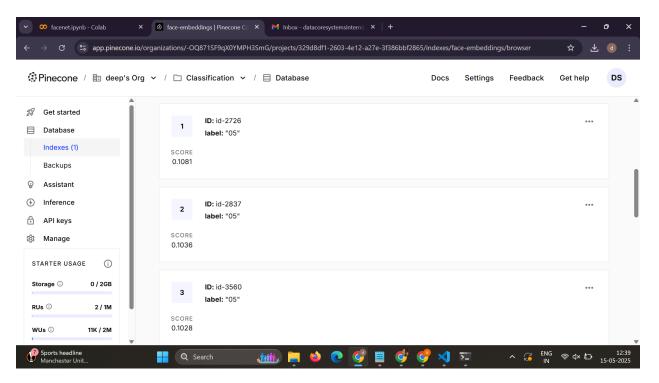
Matching:

Cosine similarity in Pinecone Threshold: 0.75

Dataset:

Custom employee dataset (6 classes)

Augmented with random transforms



Face Embeddings

5. Workflow

graph TD

A[Webcam Input] --> B[InsightFace Detection]

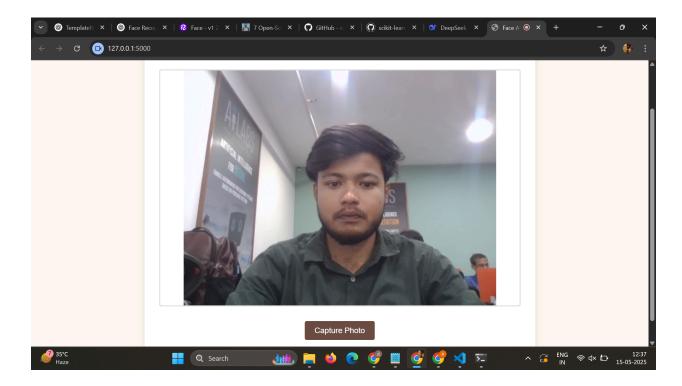
B --> C[Embedding Extraction (512D)]

C --> D[Pinecone Querying]

D --> E[Best Match + Similarity Score]

E --> F[Attendance Logging]

- Embeddings are indexed in Pinecone (dimension = 512)
- Real-time queries fetch the top-1 match using cosine similarity
- Matches above threshold are marked present



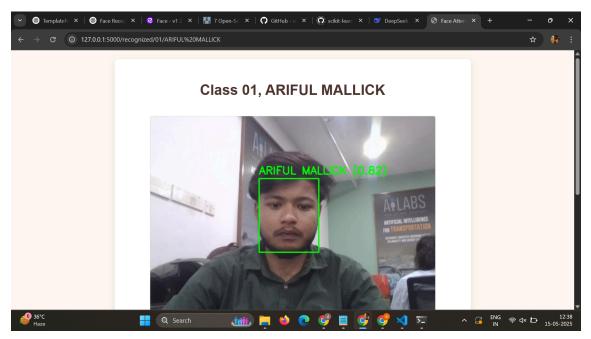
Input: 640×480 webcam frame

Preprocessing: Mean subtraction (127.5, 127.5, 127.5), Normalization (1/128)

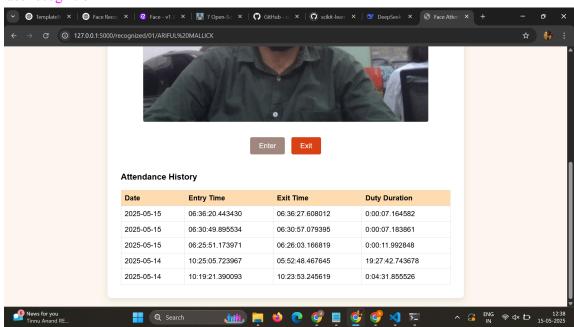
Inference: Face detection: 50ms (CPU)

Embedding: 120ms (CPU)

Postprocessing: Top-1 Pinecone search: ~200ms



face recognition



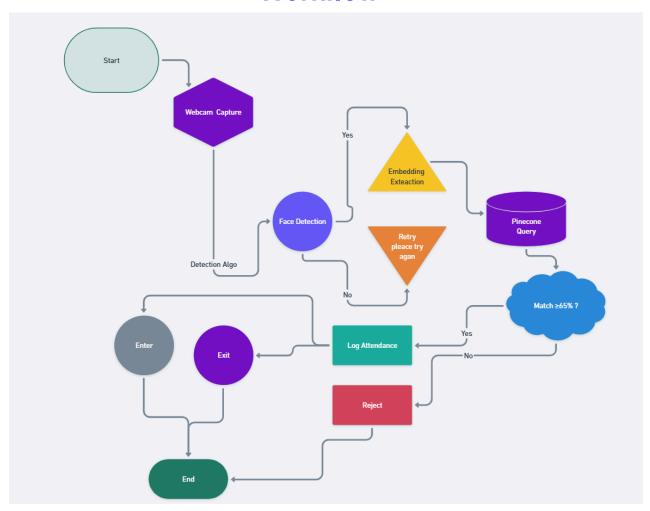
attendance system

6. Inference Workflow

Steps:

- 1. Capture Frame via Webcam
- 2. Detect Faces → Extract ROI
- 3. Generate Embedding via InsightFace
- 4. Query Pinecone (top_k=1)
- 5. Retrieve Label & Similarity
- 6. Mark Attendance if Match Found

Workflow



Flowchart:

graph TD

 $A[Start] \dashrightarrow B[Capture\ Image] \quad B \dashrightarrow C[Detect\ Face] \quad C \dashrightarrow D[Generate\ Embedding] \quad D \dashrightarrow E[Query\ Pinecone] \\ E \dashrightarrow F\{Match > Threshold?\} \quad F \dashrightarrow G[Record\ Attendance] \quad F \dashrightarrow No \dashrightarrow H[Alert:\ No\ Match]$

7. Memory Usage & Performance

Metric Value

Embedding Size 512 floats (2 KB approx.)

Inference Time ~120-150 ms (CPU)

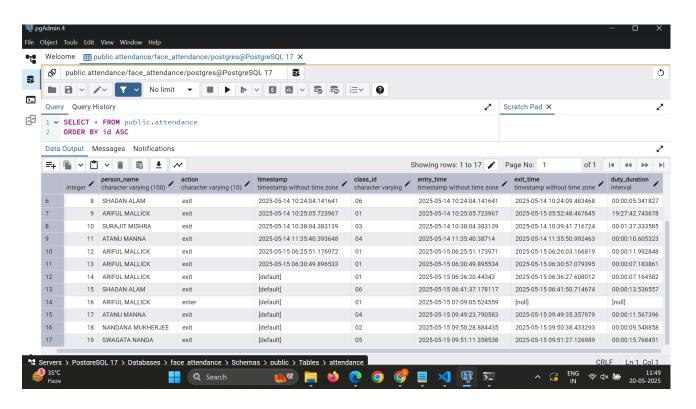
Model Size ~90 MB

DRAM Usage ~400 MB (with webcam & model)

Pinecone Query Time ~50 ms (Cloud)

Bandwidth (per query) ~3 KB per embedding

Benchmarked on: Intel i5-8th Gen, 8 GB RAM



Attendance data from postrgresql

8. Deployment

Platform

• Developed in VS Code using Python Virtual Environment

Tools

- InsightFace (face embedding)
- **Pinecone** (face retrieval)
- OpenCV (webcam & image processing)
- PostgreSQL/CSV (optional storage)

UI/UX Deployment

- Docker Container
- Edge Devices (Jetson Nano, Raspberry Pi)
- Streamlit/FastAPI Frontend

9. Future Scope

- Add real-time attendance dashboard
- Deploy on Jetson Nano or Raspberry Pi
- Incorporate multi-face detection and batch recognition
- Enhance database with multi-day logs and visual analytics
- Add voice feedback or alerts for recognized faces
- Support face re-enrollment and updates

10. What I Did

- Implemented InsightFace-based recognition pipeline
- Integrated Pinecone for fast vector retrieval
- Wrote real-time webcam capture logic with OpenCV
- Designed the attendance logic and API for match recording
- Created and tested system locally in VS Code with virtual environment
- Performed benchmarking and memory profiling
- Documented the entire system pipeline
