

Team : 92  
Mentor : Mr. Preshit Mangesh  
Desai  
Member : Sneha Pandit  
Institute : GLA University

# SENTIA

Emotional Drift Analysis in Temporal Text Streams Using AI



Single-modality emotion detection systems fail to accurately capture human emotions due to ambiguity in text, lack of contextual understanding in facial expressions, and absence of temporal emotional analysis.

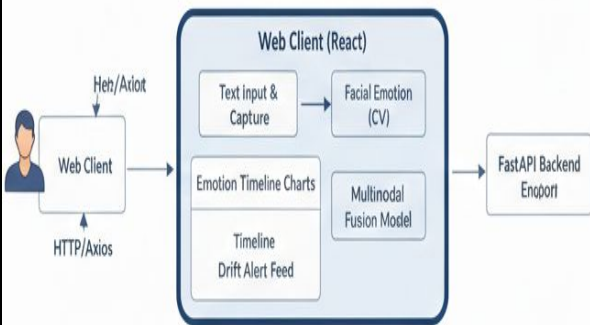
Goal: Build a multimodal emotion intelligence system combining text + facial emotion recognition + temporal drift analysis

Value Proposition: Improved accuracy, interpretability, and longitudinal emotion tracking

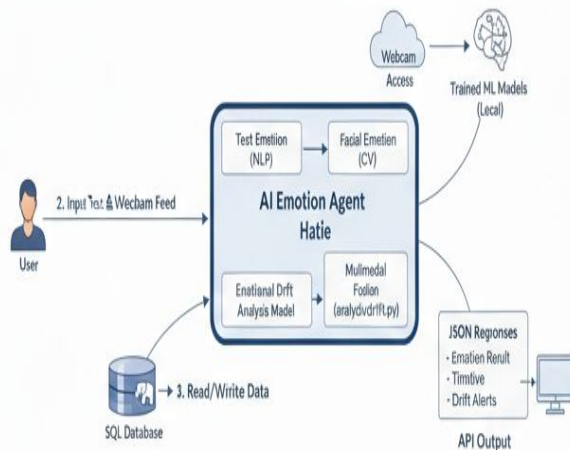
# PROBLEM STATEMENT

# SYSTEM ARCHITECTURE

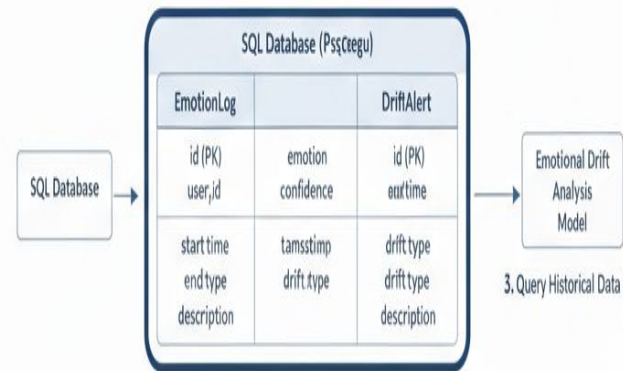
## Frontend Architecture Diagram



## Backend Architecture Diagram



## Database Architecture Diagram



## Frontend

React SPA (Axios for API calls)

## Backend

FastAPI (Modular services:  
Inference, Fusion, Drift, API Layer)

## Database

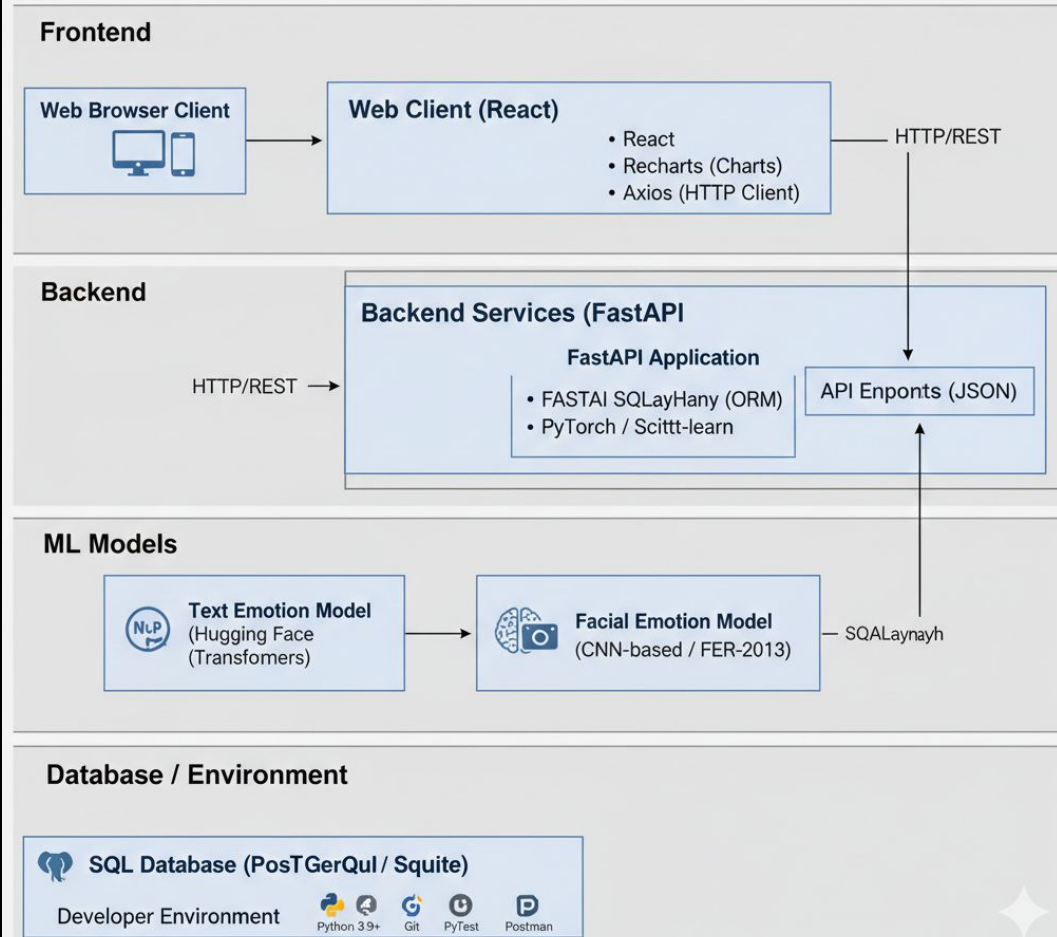
SQL-based (EmotionLog, DriftAlert)

# TECHNICAL STACK & COMPONENTS

- Frontend:
  - ◆ React
  - ◆ Recharts
  - ◆ Axios
- Backend:
  - ◆ FastAPI
  - ◆ PyTorch/Scikit-learn
  - ◆ SQLAlchemy
- ML Models:
  - ◆ Text Emotion: Hugging Face Transformers
  - ◆ Facial Emotion: CNN-based model (FER-2013)
- Database:
  - ◆ SQLite / PostgreSQL
- Environment:
  - ◆ Python 3.9+
  - ◆ Git
  - ◆ PyTest
  - ◆ Postman

## System Architecture Diagram - Technology Stack

WHO WE ARE



Approach: Rule-based fusion of text + facial emotion outputs

Process:

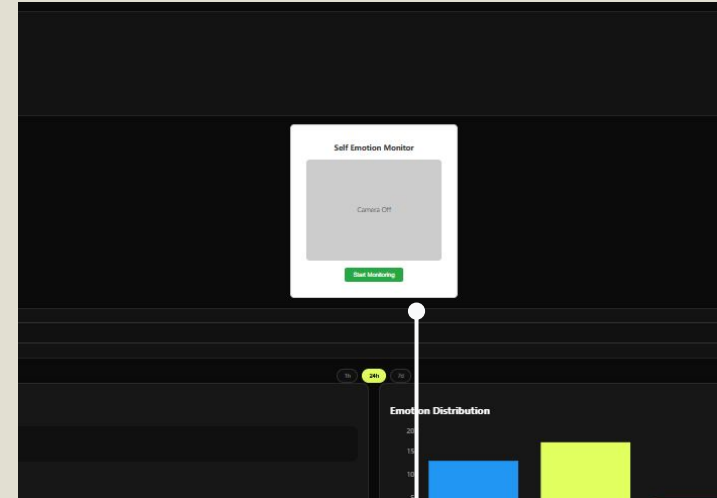
- Text input → NLP model → Emotion label + confidence
- Webcam input → Facial recognition model → Emotion label + confidence
- Fusion logic → Weighted confidence → Final emotion label
- Fallback Mechanism: Text-only mode if webcam fails

# Multimodal Fusion Logic

## Text Analysis



## History



## Face Emotion Detection

## Emotion Drift Detection - Technical Deep Dive

- Definition : Significant change in emotional pattern over time
- Method :
  - Sliding window analysis on historical emotion logs
  - Statistical thresholds for drift detection (e.g., confidence drop, label change)
- Output: Drift alerts stored in DB, Visualized in frontend

## Reports

**POST** /reports/generate Generate Report Api

**GET** /reports/download/{report\_id} Download Report

## auth

**POST** /auth/google Google Login

**POST** /auth/signup Signup

**POST** /auth/token Login For Access Token

**POST** /auth/forgot-password Forgot Password

**POST** /auth/reset-password Reset Password

## Analysis

**POST** /analyze/chat Analyze Chat Upload

**GET** /analyze/chat/status/{job\_id} Get Chat Analysis Status

## self-emotion

**POST** /self-emotion/capture Capture Emotion

**GET** /self-emotion/history Get History

## default

**GET** / Health

**POST** /predict Predict

**GET** /visualization/timeline Timeline

**GET** /visualization/distribution Distribution

**GET** /drift Drift

**GET** /alerts Get Alerts

**GET** /compare Compare



## API Design & Endpoints

RESTful APIs (FastAPI):

POST/ predict : Emotional inference

GET/ timeline : Historical emotion logs

GET/ drift : Drift detection results

Response Time: <2 seconds

Authentication: User ID- based data mapping



# Data Design & Privacy

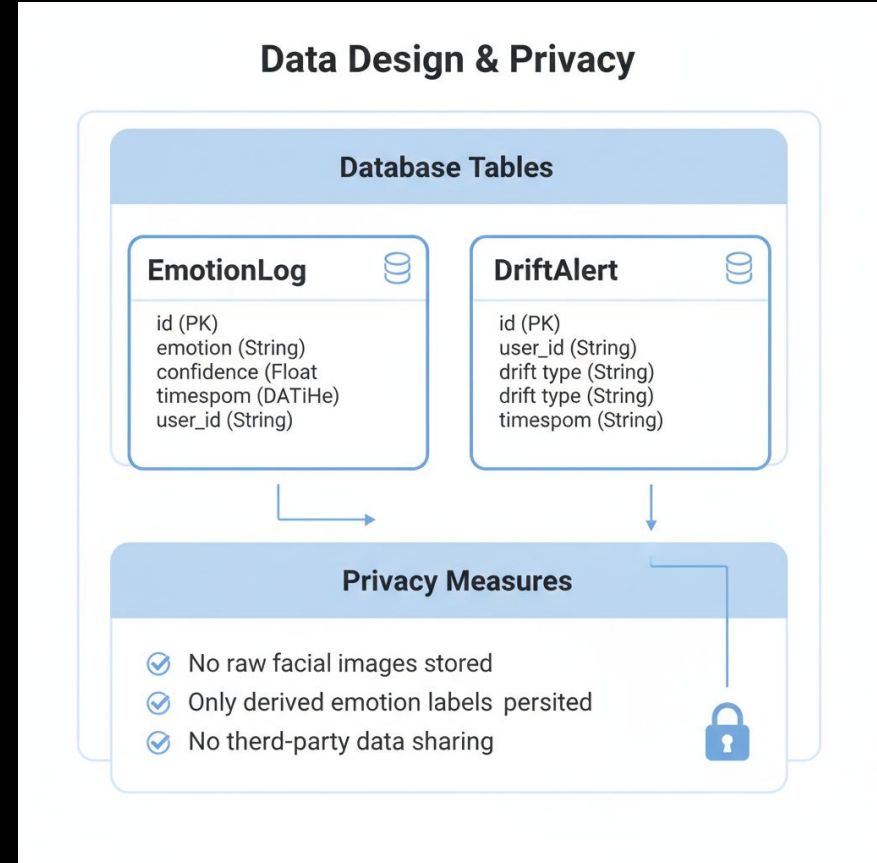
## Tables:

### EmotionLog

### DriftAlert

Privacy Measures

- No facial images
- Only derived emotions
- No third-parthy data sharing



# Research & Market Analysis

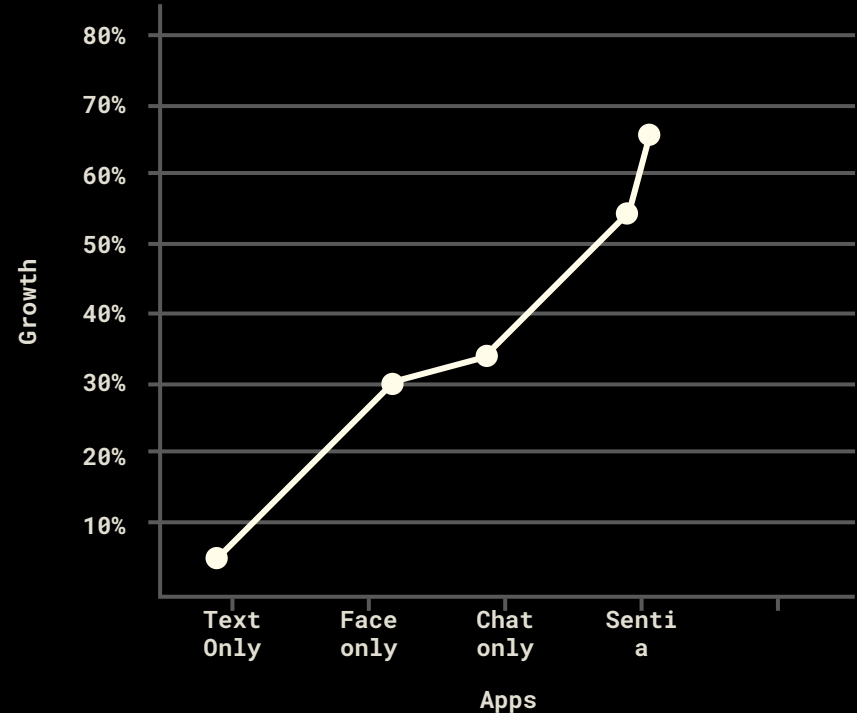
## Literature Review :

- Ekman's Basic Emotion
- Multimodal Sentiment Analysis
- CNN-based Facial emotion Recognition

## Competitor Analysis:

- Text-only tools (No facial cues)
- Facial-Only tools (No context)
- Chatbots (no drift analysis)

**Our Edge : Multimodal + temporal drift detection**



# Testing & Quality Assurance

## Non-Functional Requirements:

1. Latency  $\leq 2$  Sec
2. Uptime  $\geq 99\%$
3. Error rate  $\leq 1\%$

## Testing Strategy

1. Unit tests (PyTest)
2. Integration tests (Postman)
3. UI tests (Manual)

**Environments:** Local - test - Demo

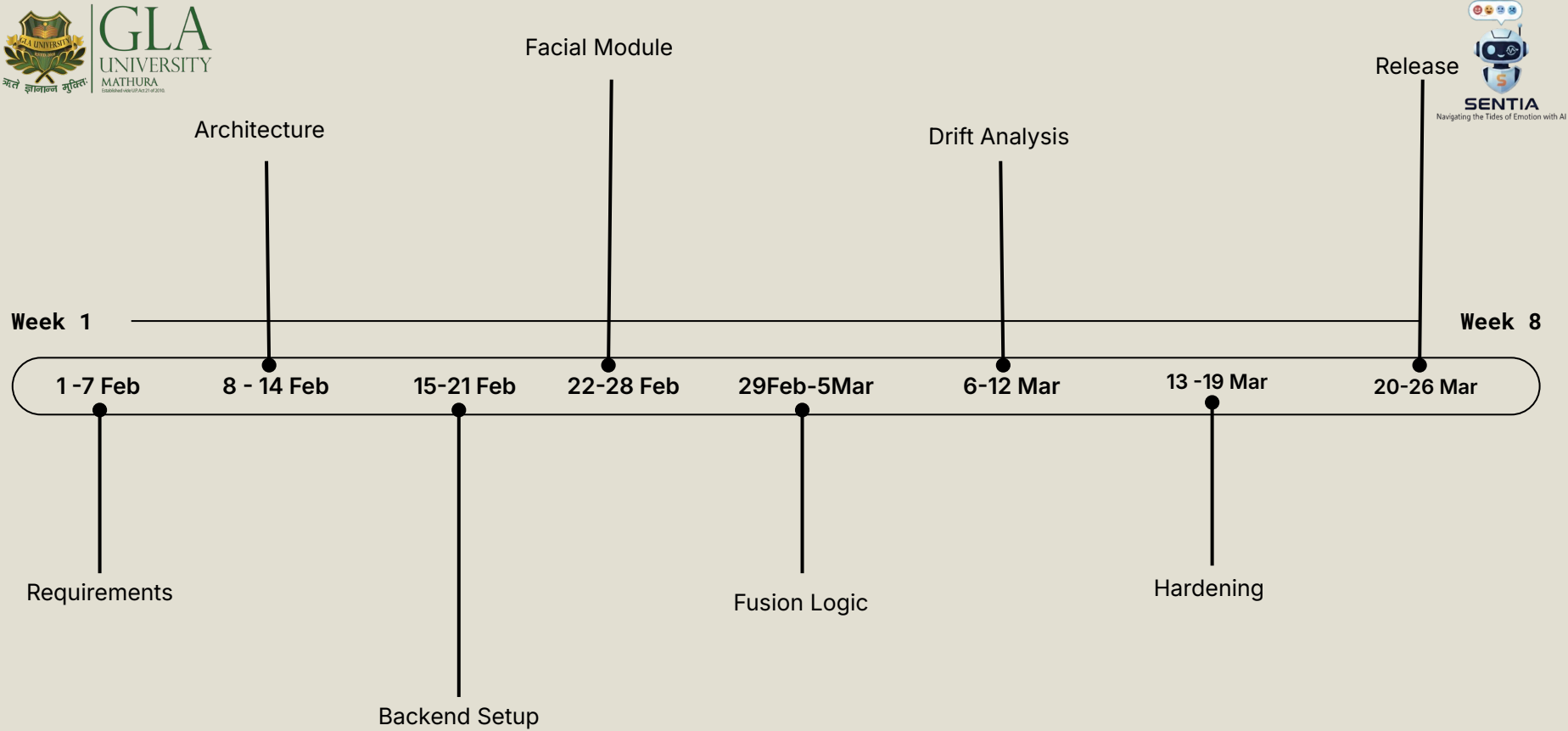
# Risk Mitigation & Challenges

## **Risks :**

- 1. Low detection accuracy:**
  - a. Model training**
  - b. validation**
- 2. Webcam failure**
  - a. Fallback to text-only mode**
- 3. Time constraints :**
  - a. Scope**
  - Prioritization**

## **Mitigation :**

- 1. Modular Design**
- 2. Incremental Development**
- 3. Continuous Validation**



## Week-wise Timeline



- **Achievements :**
  - Built a working multimodal emotion detection system
  - Implemented drift detection and visualization
  - Ensured Privacy-aware design
- **Future Work :**
  - Speech Emotion Integration
  - Cloud deployment
  - Mobile app deployment

**Conclusion and Future Scope**

