



GLA
UNIVERSITY
MATHURA
Established vide U.P. Act 21 of 2010.

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SENTIA

Emotional Drift Analysis in Temporal Text Streams Using AI





गृहे ज्ञानान्न मुकिता

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.

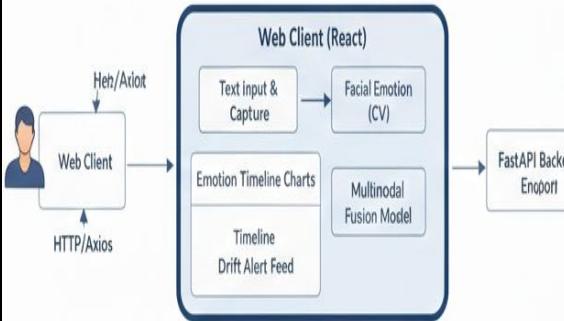
Goal: Build a multimodal emotion intelligence system combining text + facial emotion recognition + temporal drift analysis

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

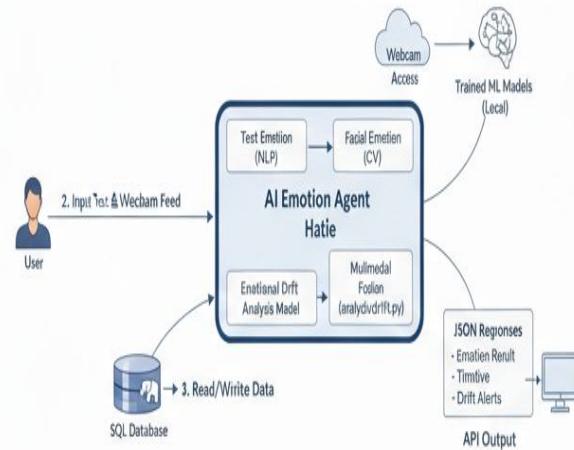
PROBLEM STATEMENT

SYSTEM ARCHITECTURE

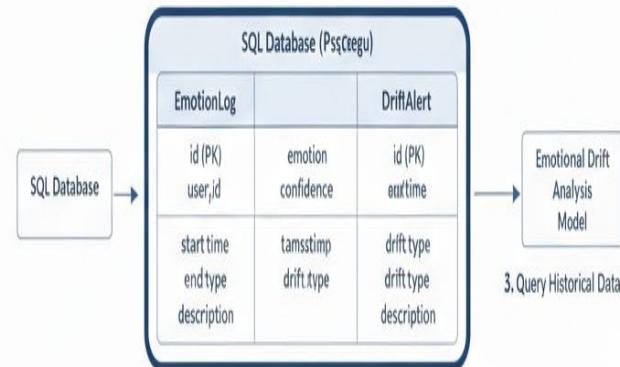
Frontend Architecture Diagram



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

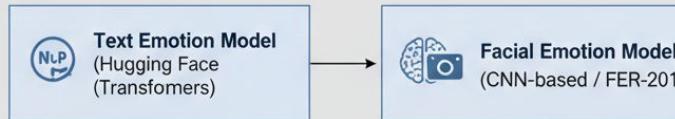
Frontend



Backend



ML Models



Database / Environment





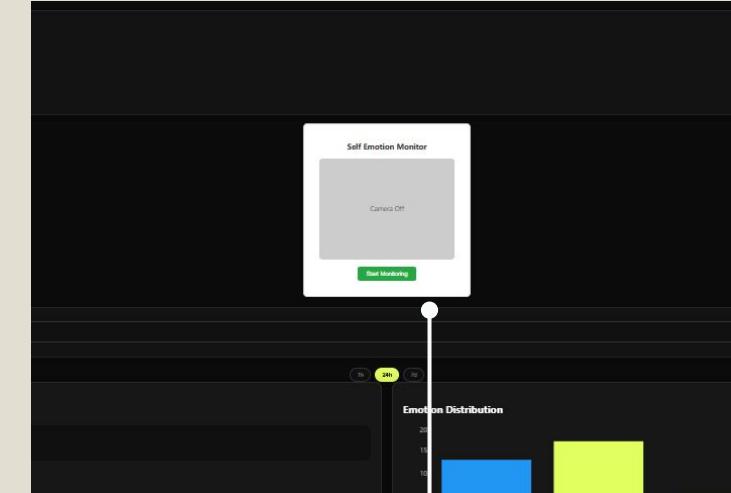
Approach: Rule-based fusion of text + facial emotion outputs

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



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

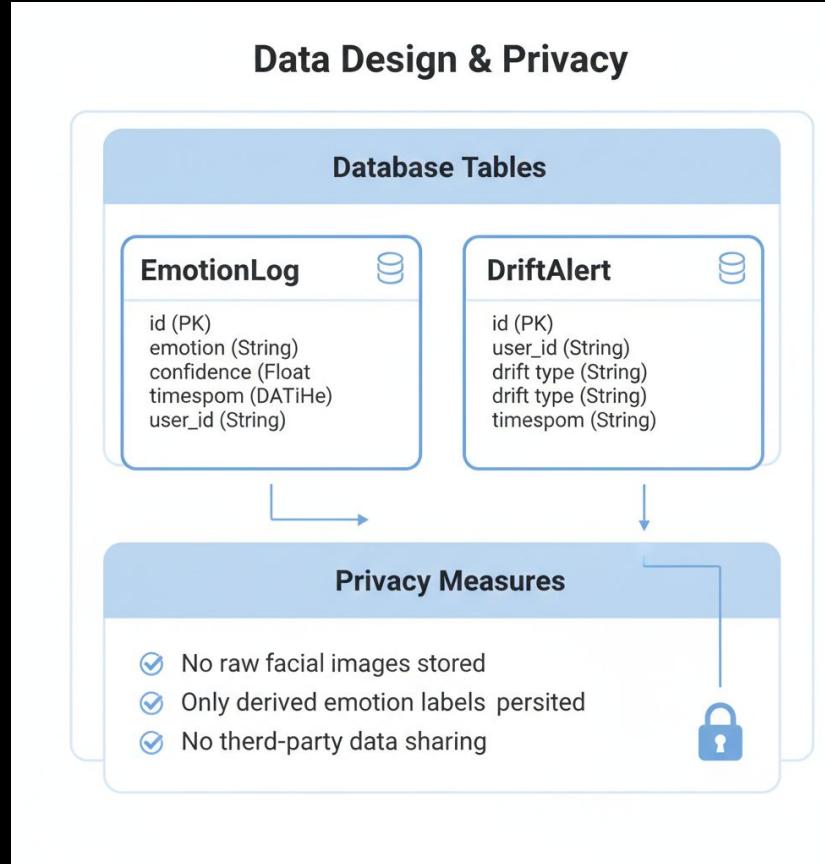
Data Design & Privacy

Tables:

- EmotionLog
- DriftAlert

Privacy Measures

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



Research & Market Analysis

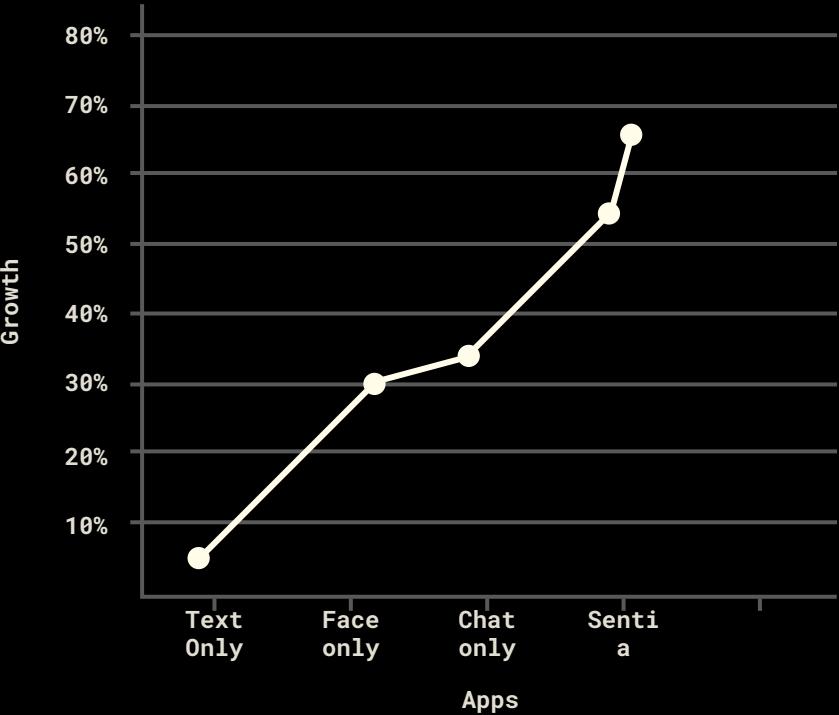
Literature Review :

- 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 <= 2 Sec
2. Uptime >= 99%
3. Error rate <= 1%

Testing Strategy

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

Environments: Local - test - Demo



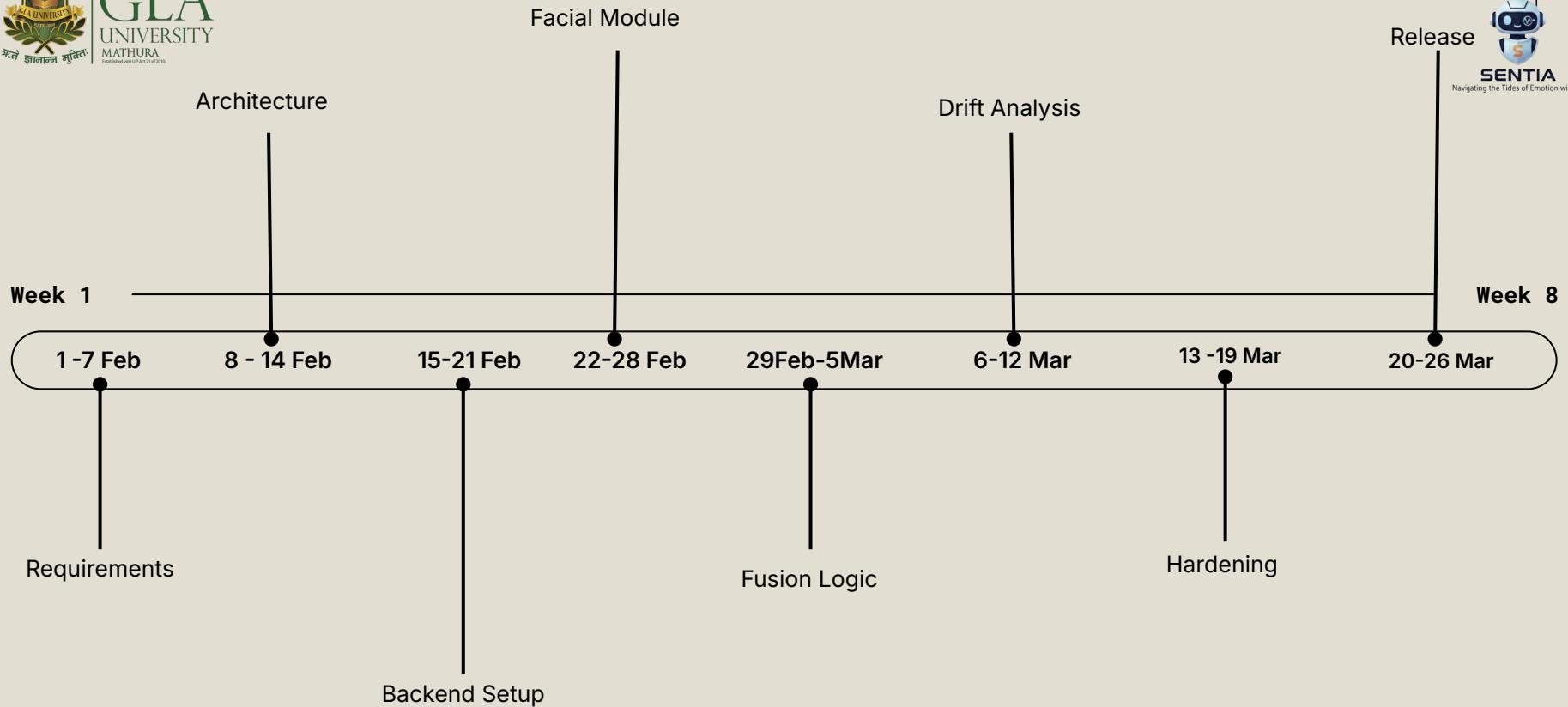
Risk Mitigation & Challenges

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

