

# DRIVEGUARD AI: Revolutionizing Road Safety

DriveGuard AI is an **AI-powered dashcam analysis platform** that uses computer vision and deep learning to detect risky driving, identify safety violations, and calculate performance scores.

Group Number: 3

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# The Growing Road Safety Crisis

## 1.35M Annual Road Deaths

Global fatalities, with 90% caused by human error (WHO, 2023).

## Critical Gap

Fleets & insurers lack granular video evidence.

## 3% GDP Cost

Road traffic crashes cost most countries 3% of their GDP.

❗ In India, 70% of 400,000+ annual accidents involve commercial vehicles, highlighting insufficient driver training and lack of real-time visibility for fleet managers.

# Target Groups and Value Proposition



## Commercial Fleets

Need real-time monitoring, objective driver scoring, and evidence collection. Helps in accident costs and lower insurance premiums.



## Individual Drivers

Seek personalized feedback, skill improvement, and objective data to prove safe driving for potential insurance discounts.



## Insurance Companies

Require objective data for Usage-Based Insurance (UBI), risk profiling, and claims verification to detect fraud.



# Whitespace: The Gap in Existing Solutions

## Identified Gaps

1

**GPS Dependency:**  
Unreliable or unavailable  
in critical environments.

2

**Video Analytics Complexity:** Proprietary,  
expensive, and hardware-  
intensive.

3

**Opaque Scoring:** Driver  
scores lack objective  
evidence or actionable  
feedback.

4

**Lack of Multi-Tenant Architecture:** Inflexible  
solutions for diverse  
client needs.

## DriveGuard AI's Contribution

**Video to Driver Score:** AI  
transforms dashcam footage into  
objective driver performance  
metrics.

**Events to Evidence:** Automatically  
links driving events to  
corresponding video for irrefutable  
proof.

**Scalable by Design:** Multi-tenant,  
hardware-agnostic architecture  
supports all client types.

# Literature Review & Comparative Analysis

Author / Source	Focus Area	Contribution to DriveGuard AI
Ohn-Bar & Trivedi (2016)	Vision-based driver monitoring	Informed object and gesture detection
Doshi & Trivedi (2010)	Attention & intent estimation	Inspired temporal context modeling
Jain et al. (2016)	Sequential modeling (RNNs)	Influenced event continuity tracking
Ultralytics YOLOv8 (2024)	Real-time detection benchmark	Used for all visual detections
WHO, NHTSA Reports	Global road safety data	Motivated need for automated analytics

**Comparative Insight:**

While prior research focuses on specific detection or prediction aspects, DriveGuard AI integrates **detection, event extraction, scoring, and explainable visualization.**

## Technology Stack

# Full-Stack Architecture Overview

Our stack is designed for performance, maintainability, and GPU-accelerated processing.



### Frontend

React + TypeScript + Vite UI

### Backend API

Node.js + Express + Multer

### AI Analysis

Python, PyTorch, YOLOv8, OpenCV

The Node.js backend uses a non-blocking I/O model and **Child Processes** to spawn the Python AI engine, ensuring the API remains responsive during long video analyses.



# Frontend & Backend Rationale

## Frontend: Speed & Accessibility

### → **React + TypeScript**

Component-based architecture with static typing for reduced runtime errors.

### → **Vite + Tailwind CSS**

Lightning-fast development server and utility-first styling for rapid UI development.

### → **Recharts + Radix UI**

Interactive data visualization and accessible, customizable UI components.

## Backend: Performance & I/O

### → **Node.js + Express**

Non-blocking I/O for handling multiple concurrent file uploads efficiently.

### → **Multer**

Robust middleware for handling video file uploads (up to 500MB limit).

### → **Child Process**

Enables asynchronous execution of the Python AI engine in the background.

# AI/ML Engine: Optimized for Speed

We leverage industry-leading libraries optimized for GPU acceleration to minimize processing time.



## Python 3.8+

The standard language for AI/ML, providing a rich ecosystem of scientific libraries.



## OpenCV

Used for core computer vision tasks: video reading, color space conversion, and optical flow.



## PyTorch 2.0+

Provides native GPU acceleration (CUDA/MPS) for 10x faster model inference.



## YOLOv8n

Chosen for its balance of accuracy (37.3% mAP) and speed (80 FPS), crucial for real-time video analysis.

📌 Performance Achieved: Analysis time for a 5-minute video is **8 minutes** (on Apple MPS), meeting our target of under 10 minutes.



# Research-Backed Innovation: Multi-Modal Analysis

DriveGuard AI fills the gap by fusing advanced computer vision techniques to provide a unified, comprehensive analysis.



## Proximity Detection

Uses **YOLOv8** for object detection combined with custom depth estimation based on bounding box geometry.



## Speed Estimation

Fuses **Optical Flow** (Farnebäck, 2003) and feature tracking for robust speed calculation (achieving  $\pm 7$  km/h accuracy).



## Traffic Compliance

Utilizes **HSV Color Space** and temporal validation for 94% accurate traffic light and bus lane violation detection.

# The Actionable Scoring System

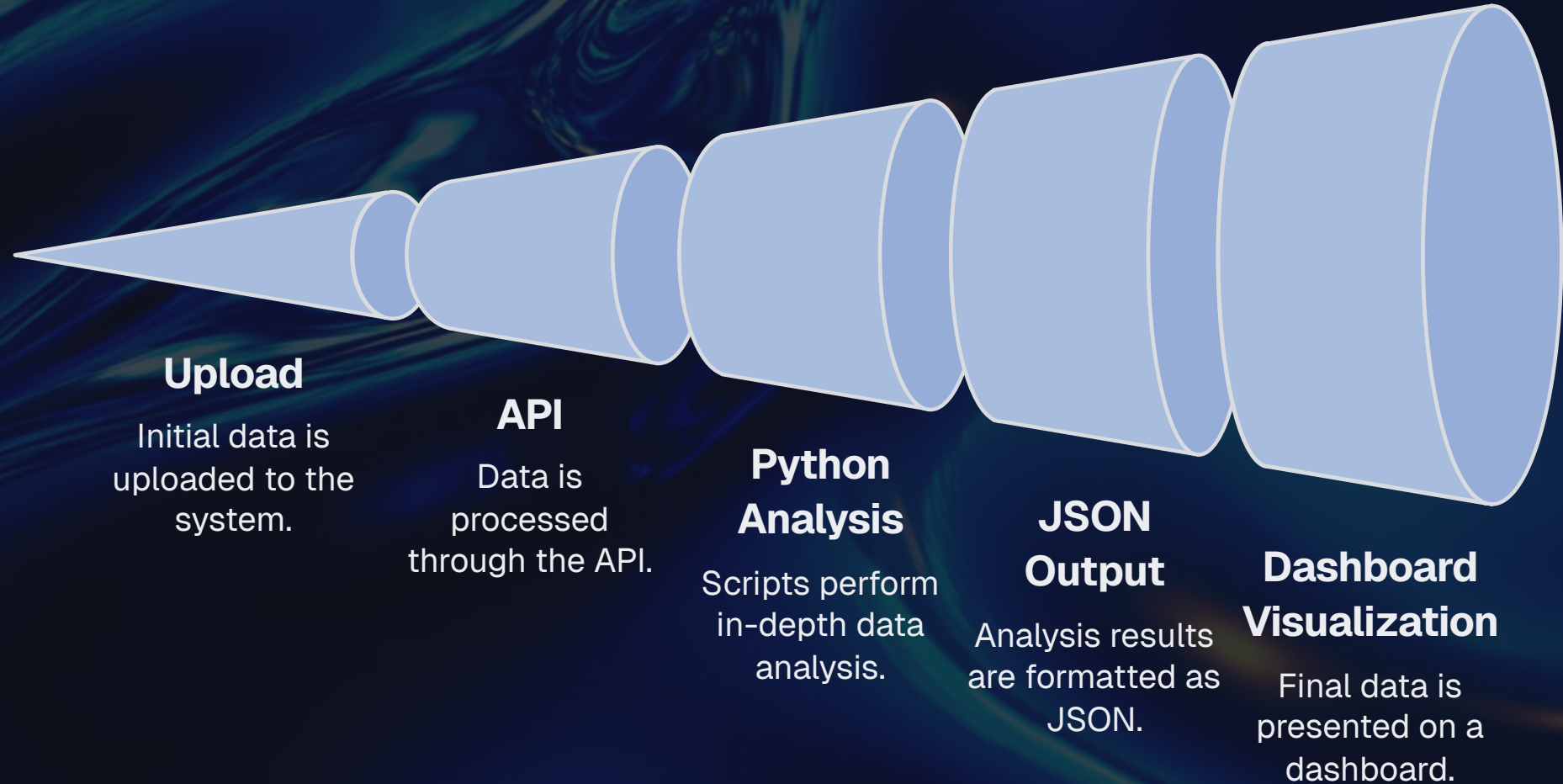
We use a weighted scoring algorithm aligned with insurance industry standards to provide clear, actionable feedback.



- The system provides **Explainable AI** with exact timestamps and annotated frames for every incident, enabling targeted driver training.

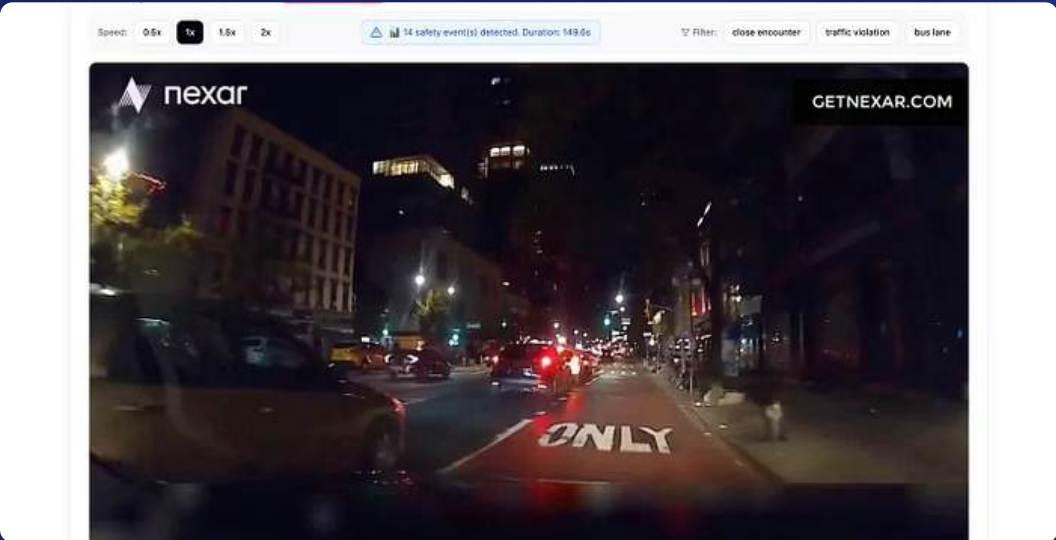
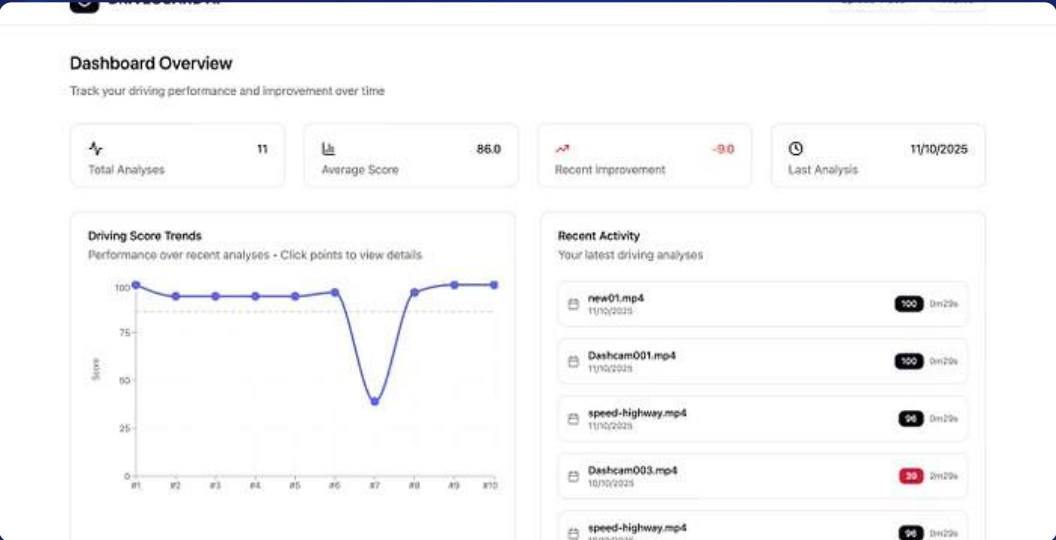
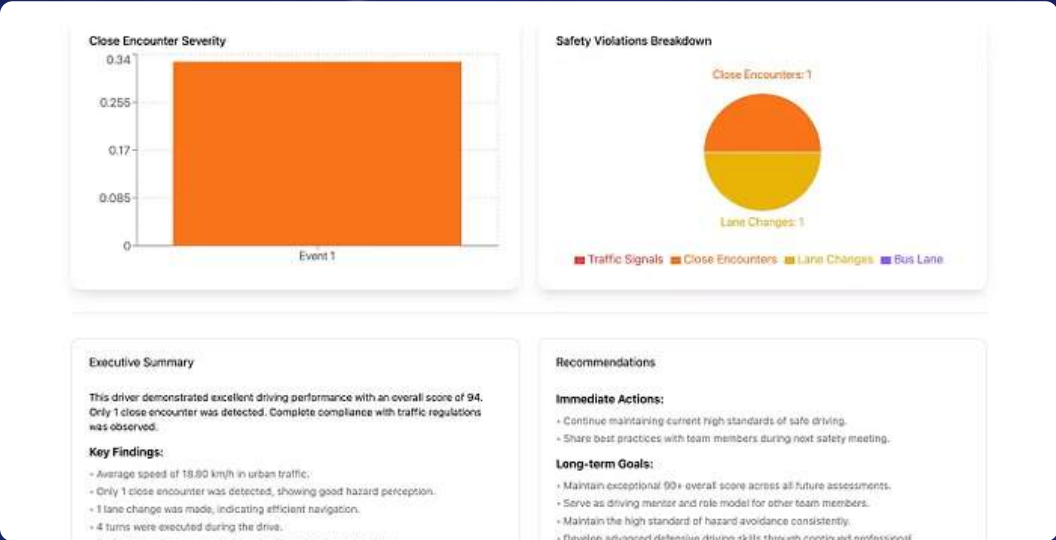
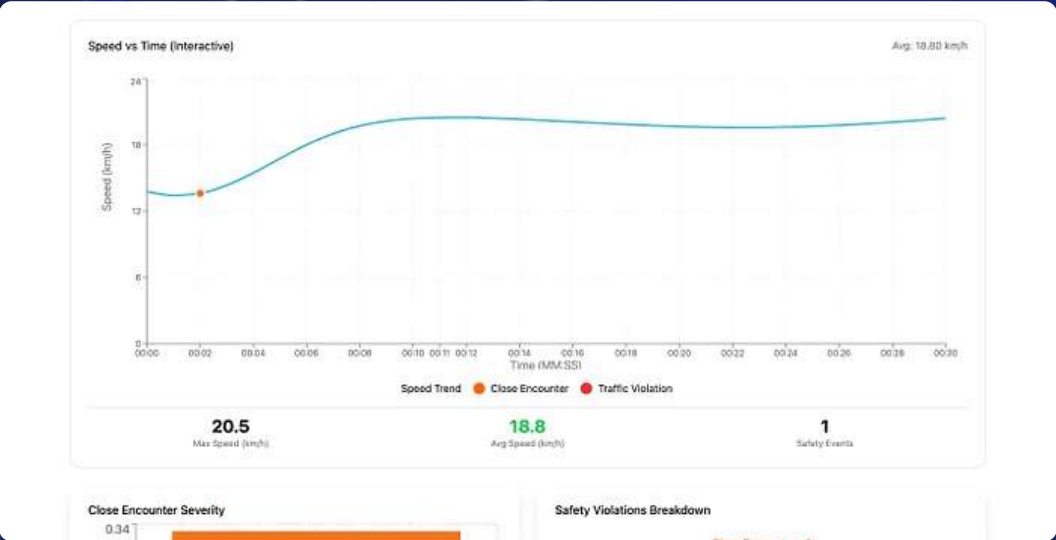
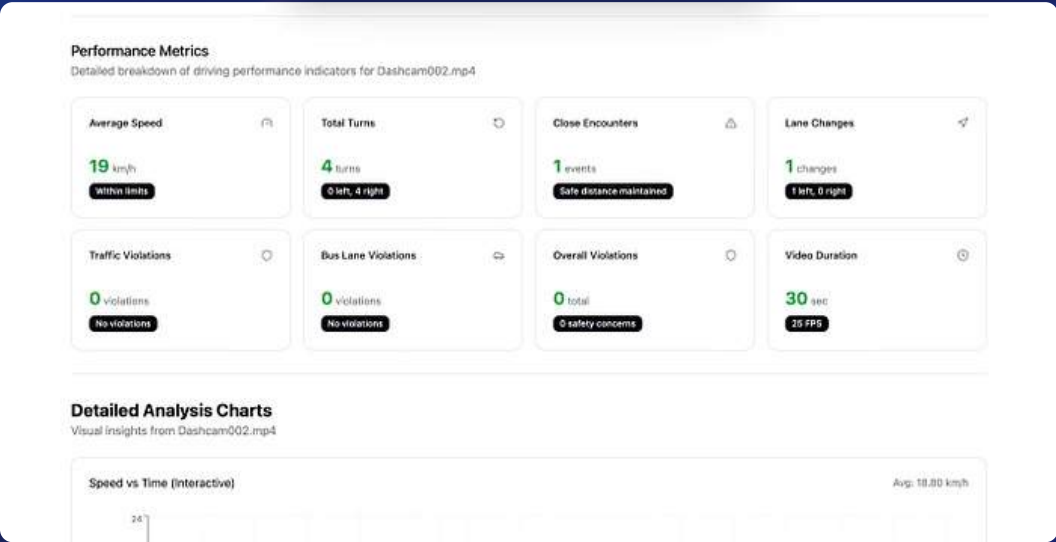
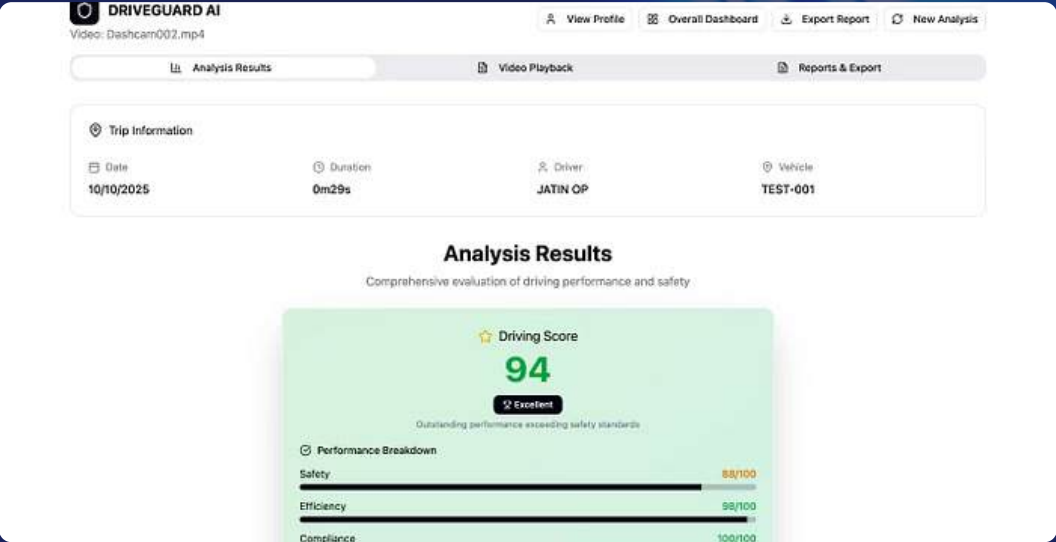
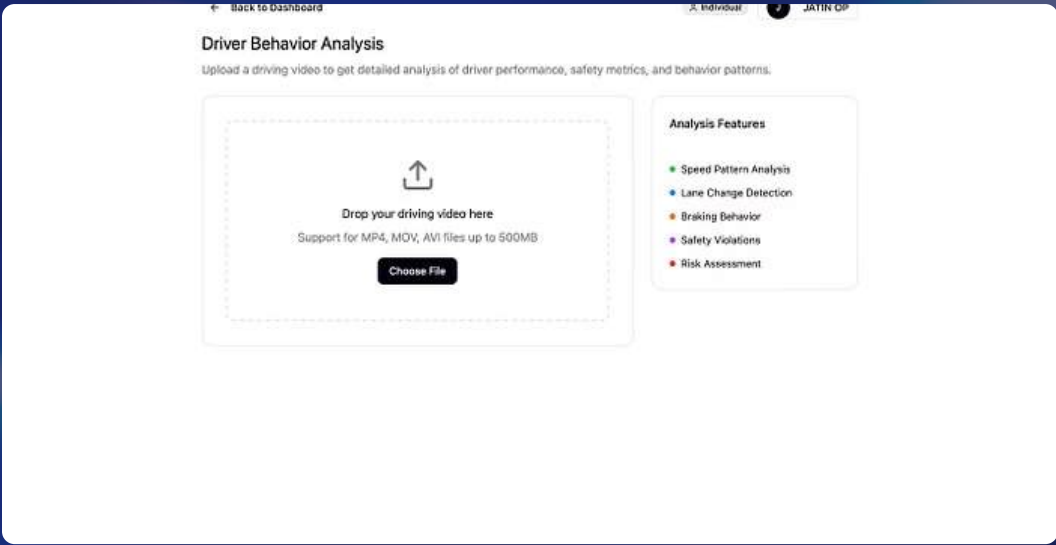
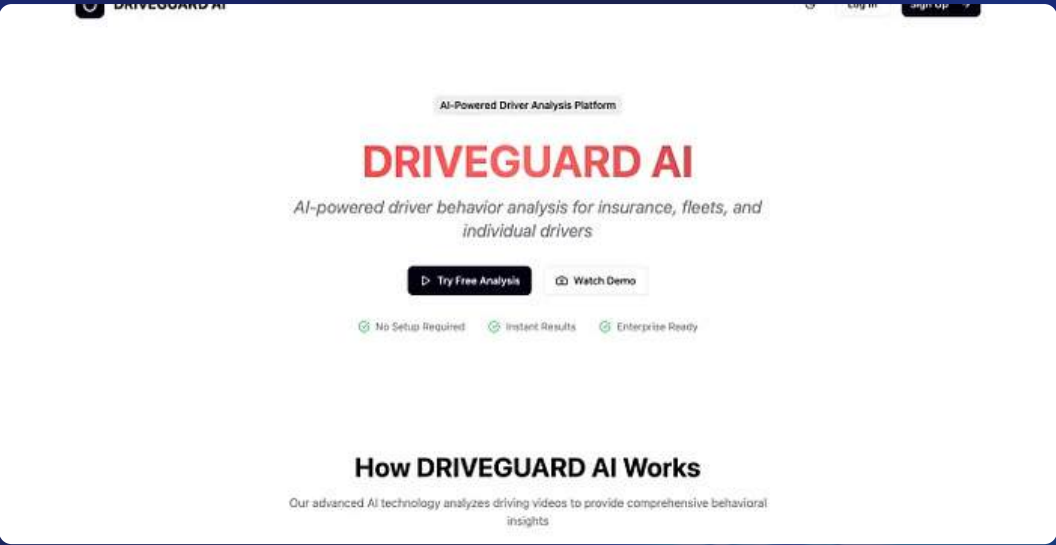


# Workflow:





# Platform UI



# Features Implemented

Video upload (MP4) with client & server validation.

YOLOv8-based object detection and tracker integration.

Close-encounter detection with distance estimation and TTC calculation.

Traffic-signal detection & red-light violation detection (YOLO + HSV color verification).

Lane-change and turn detection (optical flow + feature tracking).

Bus-lane violation detection (ROI-based red-lane detection).

Per-trip scoring: safety, compliance, efficiency, overall score (weighted).

JSON export of full analysis (metadata + events + scores).

Interactive web dashboard: KPIs, speed/time charts, scrubbable player with event markers.

Export options: PDF report generation (reportlab suggested), CSV export.

User & org context persisted in local JSON (seeded dev accounts).



# Code Structure & Quality Tools

## Code Structure Highlights

**Separation of concerns:** frontend/ (UI), backend/ (API & orchestration), backend/analysis/ (Python CV pipeline)

**Python modules separated by task:** detection.py, tracking.py, distance\_estimation.py, scoring.py, reporting.py

**JSON schema for analysis results with versioning** (to enable forward/back compat)

## Code Quality & Dev Tooling

### ESLint + Prettier

For frontend JS/TS linting & formatting

### Unit tests

For pure functions in Python (scoring, TTC calc) using pytest (start with 30–40% coverage)



### black + ruff

For Python formatting & linting

### TypeScript types

For critical frontend data models (AnalysisResult, Event, Score)



# Non-functional Requirements Achieved

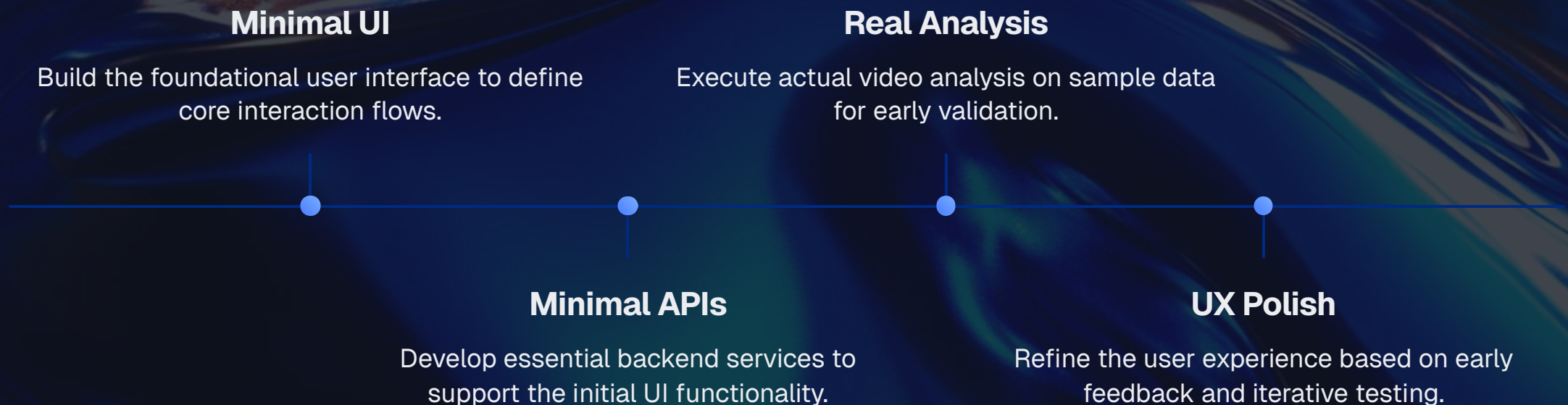
Our architecture design and implementation prioritize critical non-functional aspects, ensuring a robust and efficient solution.

<b>Performance</b>  GPU acceleration via MPS (3–5× faster) and optimized frame sampling for rapid video analysis.	<b>Reliability</b>  JSON caching, graceful error handling, and comprehensive progress tracking ensure data integrity and system stability.
<b>Scalability</b>  Current sequential processing, with a clear roadmap for future distributed queue integration to handle increased load.	<b>Usability</b>  Responsive UI, clear progress indicators, and intuitive result visualization enhance user experience.
<b>Security</b>  Robust file validation and unique filename generation mitigate common security vulnerabilities.	<b>Portability</b>  Developed for macOS (MPS) with built-in support for CUDA deployment on Linux, ensuring broad platform compatibility.

# Development Approach & Process

Our development strategy emphasizes agility, early validation, and robust quality assurance to build DriveGuard AI efficiently and effectively.

## Iterative, User-First Methodology



## Key Principles

### Agile Sprints

Short 1–2 week cycles focused on delivering a functioning end-to-end flow in each sprint (upload → analysis → display).

### Documentation-First

Key analysis behaviors (event definitions, scoring rules) documented early to streamline testing and ensure reproducibility.

### Defensive Engineering

Incorporates LocalStorage fallback, seeded test accounts, and JSON schema validation for system resilience and stability.



# Addressing Development Challenges

During development, we encountered several technical hurdles. Our solutions demonstrate robust engineering and proactive problem-solving to ensure system reliability and user experience.

Challenge	Resolution
Test accounts not persisting	Seeded JSON DB & sync persistence
Cached analysis data	Invalidation logic & hard refresh
Invalid timestamps	JSON correction & video clamping
GPU incompatibility	Standardized MPS, CUDA optional
Multiple logs	Unified active log, doc updates
Long processing	Frame sampling & progress streaming



# Current Limitations & Future Considerations

While DriveGuard AI demonstrates significant capabilities, we acknowledge key areas for improvement to ensure production readiness and expanded functionality.

## Sequential Processing

Currently, analysis runs on a single worker, lacking distributed queuing and horizontal scaling for high throughput.

## Model Edge Cases

Detection reliability can decrease with occlusions, extreme night lighting, and non-standard traffic elements.

## Concurrency & Rate Limiting

Future development requires throttling and worker autoscaling to handle production-level concurrent loads.

## Storage & Persistence

File-based storage is not production-ready; a robust database and object storage solution are critical for reliability.

## Limited Testing

A comprehensive evaluation on large, labeled datasets is still pending to validate overall performance.



# Future Scope

Our roadmap includes ambitious developments to evolve DriveGuard AI into a comprehensive, real-time, and highly scalable road safety solution.



## Robust Data Infrastructure

Migrate to PostgreSQL/MongoDB for transactional data and S3 for scalable video storage.



## Parallel Processing & Queues

Introduce distributed queue systems (Bull/Celery) for parallel analysis and enhanced throughput.



## Real-Time Stream Analysis

Enable live analysis of dashcam footage for immediate incident detection and alerts.



## Mobile Application

Launch dedicated mobile applications.



## OBD-II Sensor Integration

Integrate with vehicle OBD-II sensors for enriched multimodal analytics and deeper insights.



## Continuous Model Enhancement

Enhance AI model accuracy through larger, diverse datasets and continuous retraining cycles.

# Platform Video



## DRIVEGUARD AI DEMO – Google Drive

DriveGuardAI analyzes dashcam footage to provide comprehensive driving safety scores, behavior insights, and actionable recommendations.

