

# Vehicle Speed and Lane Usage Analysis Using Computer Vision

**Course:** TTENG564 | **Group:** B



# Problem Statement

## Critical Safety Challenge

Traffic safety remains a pressing issue worldwide, with vehicle speed and lane violations contributing significantly to accidents.

## Traditional Monitoring Limitations

Current methods are costly, require manual intervention, and provide limited scalability for comprehensive traffic analysis.

## Automation Need

Scalable, automated solutions are essential for simultaneous analysis of vehicle speeds and lane usage patterns.



# Objectives - Group B Focus



## Speed Estimation

Accurately estimate vehicle speeds using computer vision techniques with high precision and reliability.



## Lane Usage Analysis

Analyze vehicle lane usage patterns to identify traffic flow characteristics and potential safety concerns.



## Safety Insights

Provide actionable insights for traffic safety improvement and intelligent transportation management systems.



# Proposed Solution

Computer vision-based system integrating real-time detection, tracking, and analysis capabilities.

**Input Video**

**Vehicle Detection**

**Speed Estimation**

**Lane Analysis**

## Real-Time Detection

Detect and track vehicles in live video feeds using advanced computer vision algorithms.

## Speed Calculation

Estimate vehicle speed using calibrated distances and temporal analysis between frames.

## Pattern Monitoring

Monitor and analyze lane usage patterns for comprehensive traffic behavior assessment.



# Methodology



## Vehicle Detection

**YOLO (v5/v8)** implementation for robust real-time vehicle identification and classification in various traffic conditions.



## Camera Calibration

**Perspective correction** using reference objects to establish accurate spatial measurements and distance calculations.



## Speed Estimation

**Frame-to-frame analysis** calculating displacement over time intervals for precise velocity measurements.



## Lane Analysis

**Position mapping** comparing vehicle coordinates against defined lane boundaries for usage pattern identification.

# Technical Implementation



## Core Technologies

**OpenCV + Python** for comprehensive vehicle detection and image processing capabilities with robust performance.



## Advanced Tracking

**Kalman filtering** ensures smooth vehicle tracking across frames, handling occlusions and maintaining trajectory consistency.



## Spatial Correction

**Perspective transformation** converts camera view to real-world coordinates for accurate distance and speed measurements.



## Analysis Tools

**Pandas & Matplotlib** for comprehensive data analysis, statistical processing, and professional visualization output.





# Challenges & Limitations

## Environmental Factors

Variable lighting conditions between day and night operations significantly impact detection accuracy and system reliability.

## Camera Positioning

Variations in camera angles and mounting positions require adaptive calibration methods for consistent performance.

## Traffic Complexity

Vehicle occlusion in dense traffic scenarios poses challenges for continuous tracking and accurate speed estimation.

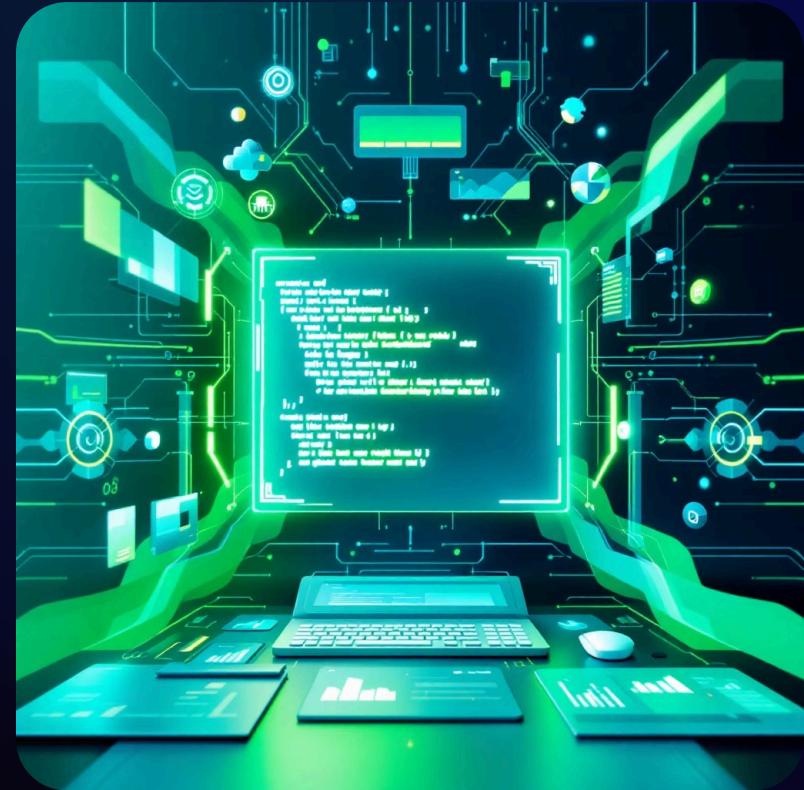
## Performance Balance

Optimizing the trade-off between real-time processing requirements and detection accuracy for practical deployment.

# Deliverables & Evaluation

## Project Deliverables

- Complete Python codebase with documentation
- CSV datasets with processed traffic data
- Statistical charts and visualization outputs
- Comprehensive technical report
- Professional presentation materials



$\leqslant 5\%$

### Speed Accuracy

Maximum acceptable error rate in speed estimation measurements

30

### FPS Target

Real-time processing performance requirement for live video analysis

95%

### Overall Accuracy

Target system accuracy for vehicle detection and tracking operations

# Resources & Timeline

## Technology Stack

- **Python 3.8+** - Core programming language
- **OpenCV** - Computer vision processing
- **YOLO v5/v8** - Object detection framework
- **TensorFlow/PyTorch** - Machine learning
- **Pandas & Matplotlib** - Data analysis
- **Git** - Version control system

## Data Requirements

Traffic camera feeds from multiple highway locations and calibration videos with known reference measurements.





# Expected Impact & Closing

## Real-World Application

Demonstrate practical computer vision implementation in transportation engineering for scalable traffic monitoring solutions.

## Foundation Technology

Establish robust groundwork for automated traffic monitoring systems that can be deployed across highway networks.

## Performance Targets

95% accuracy | 30 fps real-time | 4 lanes coverage

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# Thank You

## Questions & Discussion