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**Academy of  
Engineering**

(An Autonomous Institute Affiliated to Savitribai Phule University)

# TRAFFIC SURVEILLANCE: VEHICLE DETECTION & COUNTING USING YOLO-V8

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# INTRODUCTION

- Traffic surveillance is essential for managing road congestion and safety.
- Manual monitoring is slow and often inaccurate.
- Automated vehicle detection and counting provide real-time insights.
- Helps in traffic signal optimization, accident prevention, and smart city planning.
- Using YOLO( You Only Look Once), vehicles like cars, buses, trucks, bikes, and bicycles can be detected quickly and accurately.

# PROBLEM STATEMENT

The rapid increase in the number of vehicles has led to severe traffic congestion, frequent delays, and a rise in road accidents. Manual traffic monitoring methods are often inefficient, time-consuming, and susceptible to human errors. Additionally, existing automated solutions struggle to perform reliably under poor lighting or heavy traffic conditions. Therefore, there is a need for a robust, real-time AI-based vehicle detection and counting system to support effective traffic management and facilitate smart city planning.

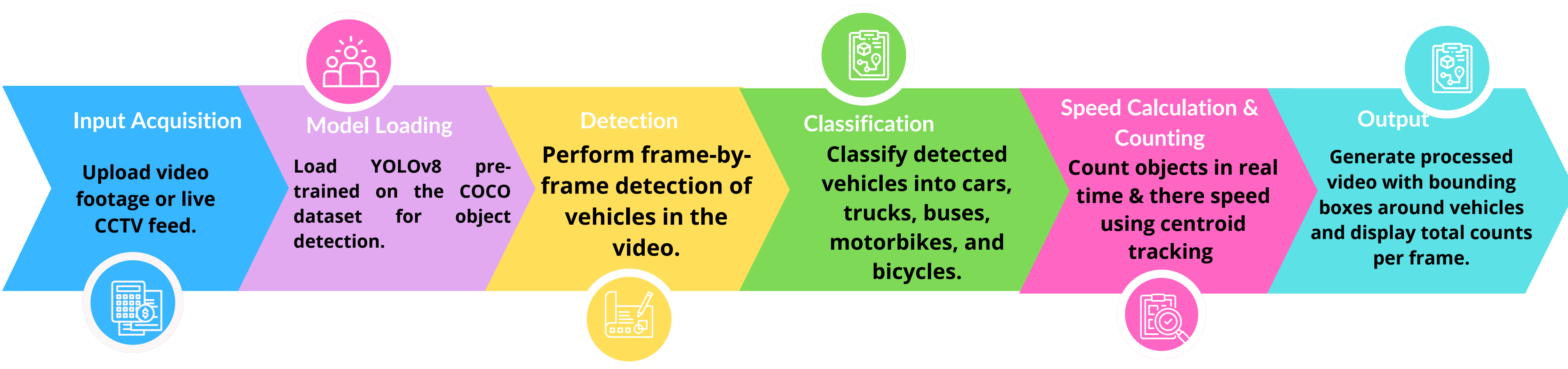
# OBJECTIVES

- Detect vehicles in real-time using CCTV or video feeds.
- Classify vehicles into categories: car, bus, truck, motorcycle, bicycle.
- Count total vehicles and provide per-category statistics.
- Measure vehicle speeds using frame-by-frame centroid tracking.
- Generate structured traffic data for analysis and decision-making.
- Assist in traffic signal optimization, congestion control, and accident prevention.

# DATASET ANALYSIS

- COCO Dataset (Common Objects in Context)
  - 118,000+ training images
  - 80 object categories
  - Includes vehicles: car, bus, truck, motorbike, bicycle
- YOLOv8 Model
  - Pre-trained on COCO dataset
- Evaluation Dataset
  - Real-world traffic video used for testing
  - Ensures robust training (COCO) + realistic testing (custom video)

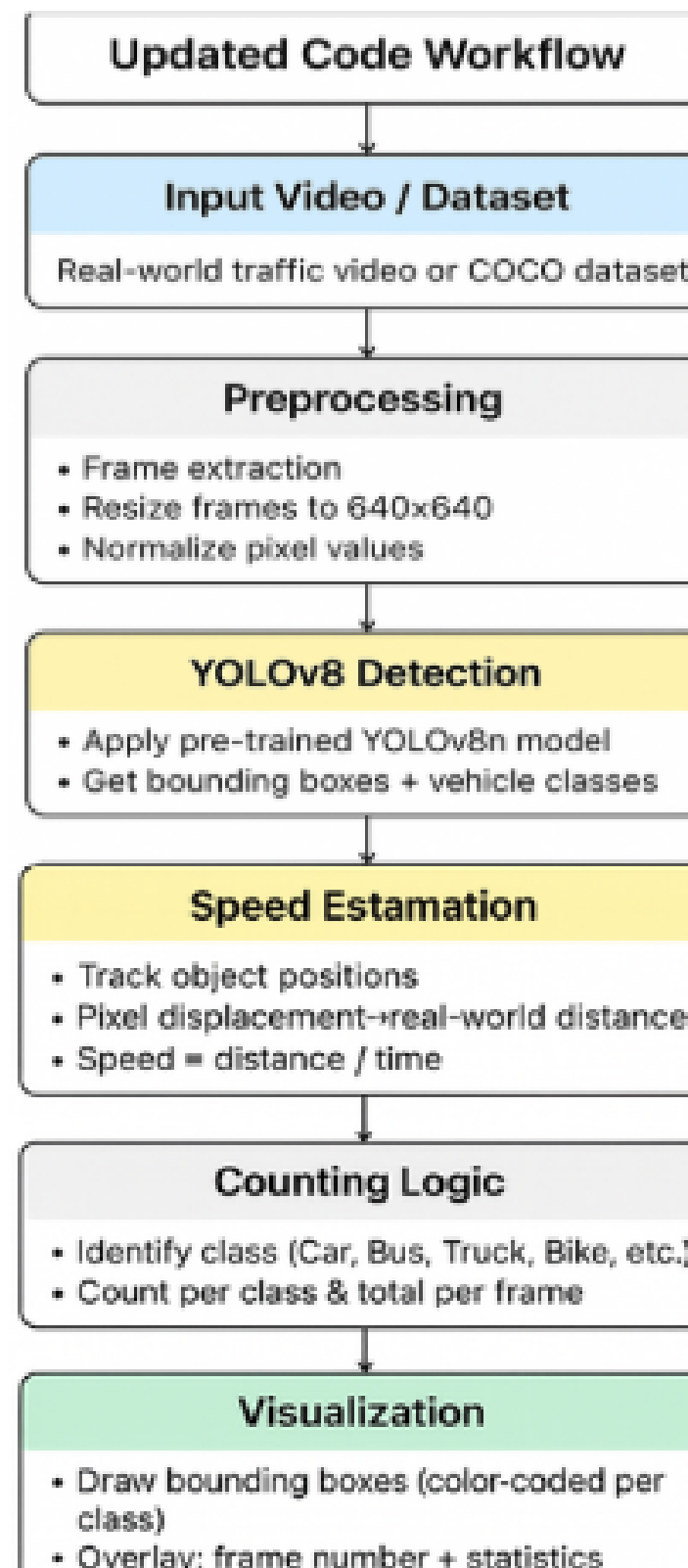
# METHODOLOGY



# TECH STACK

- **Python** – core programming language for scripting and analysis
- **Ultralytics YOLOv8** – state-of-the-art model for vehicle detection & classification
- **OpenCV** – video processing, frame extraction, image enhancement
- **Matplotlib & Seaborn** – data visualization and result plotting
- **Scikit-learn** – performance metrics and confusion matrix
- **Pandas & NumPy** – data manipulation and numerical operations
- **Google Colab** – cloud-based execution for GPU acceleration

# CODE WORKFLOW



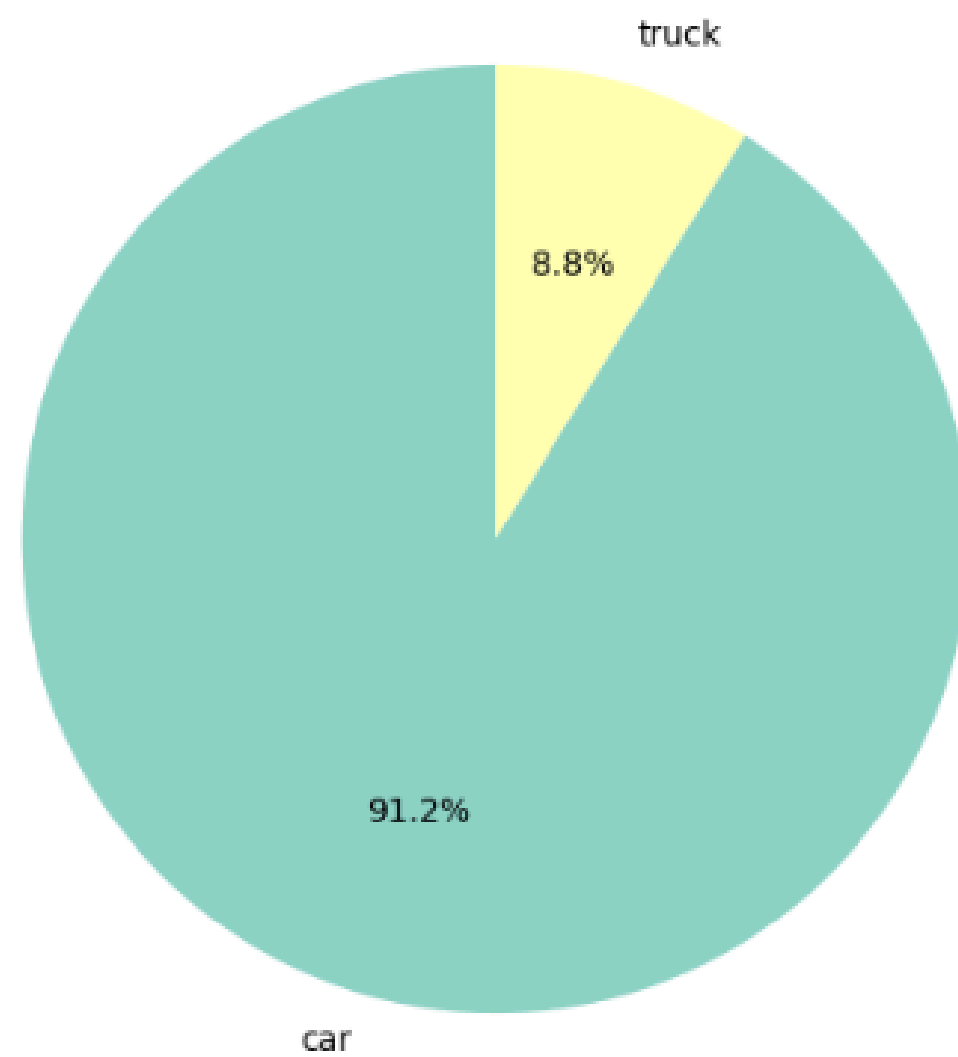


# CODE EXPLANATION

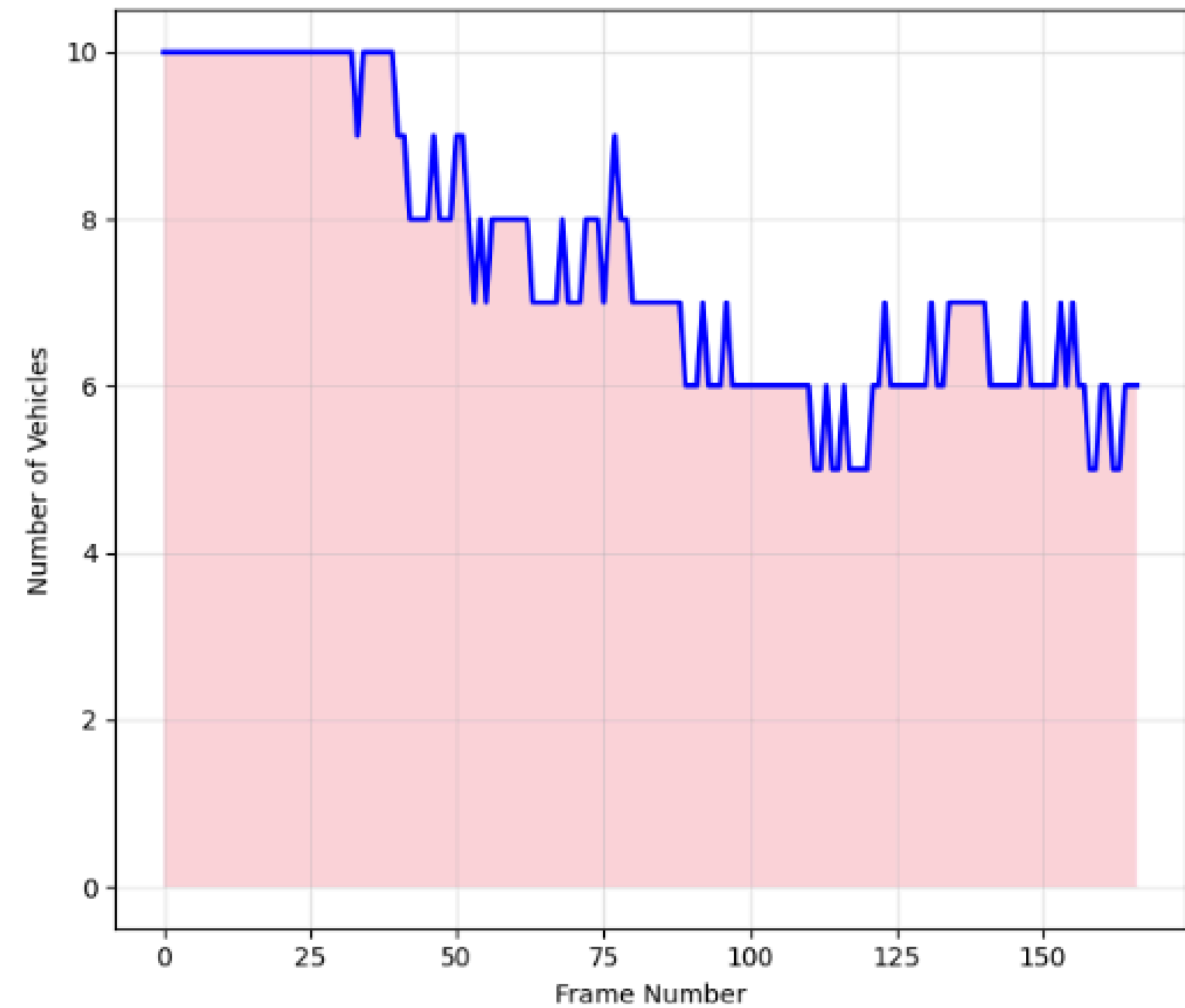
- Input video is preprocessed—frames are extracted and normalized for YOLOv8.
- YOLOv8x model detects vehicles on each frame, classifies their type.
- Centroid-based tracking is applied to maintain identities and track movement.
- Speed for each vehicle is estimated using centroid displacement and video calibration.
- Counts and speeds are overlaid per vehicle and aggregated across the video.
- Statistics, performance metrics, and visualizations are generated for analysis.

# EXPLORATORY DATA ANALYSIS

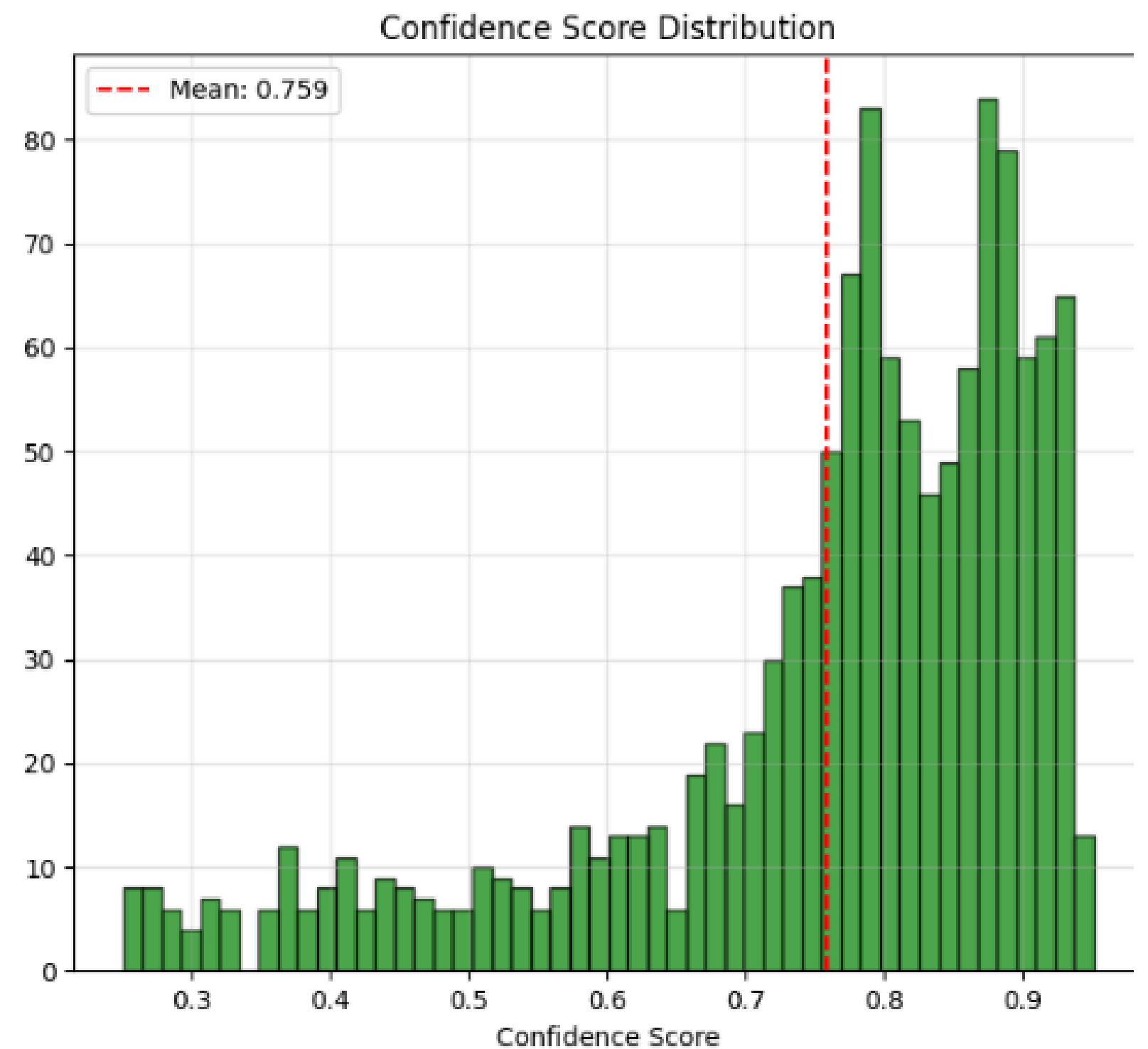
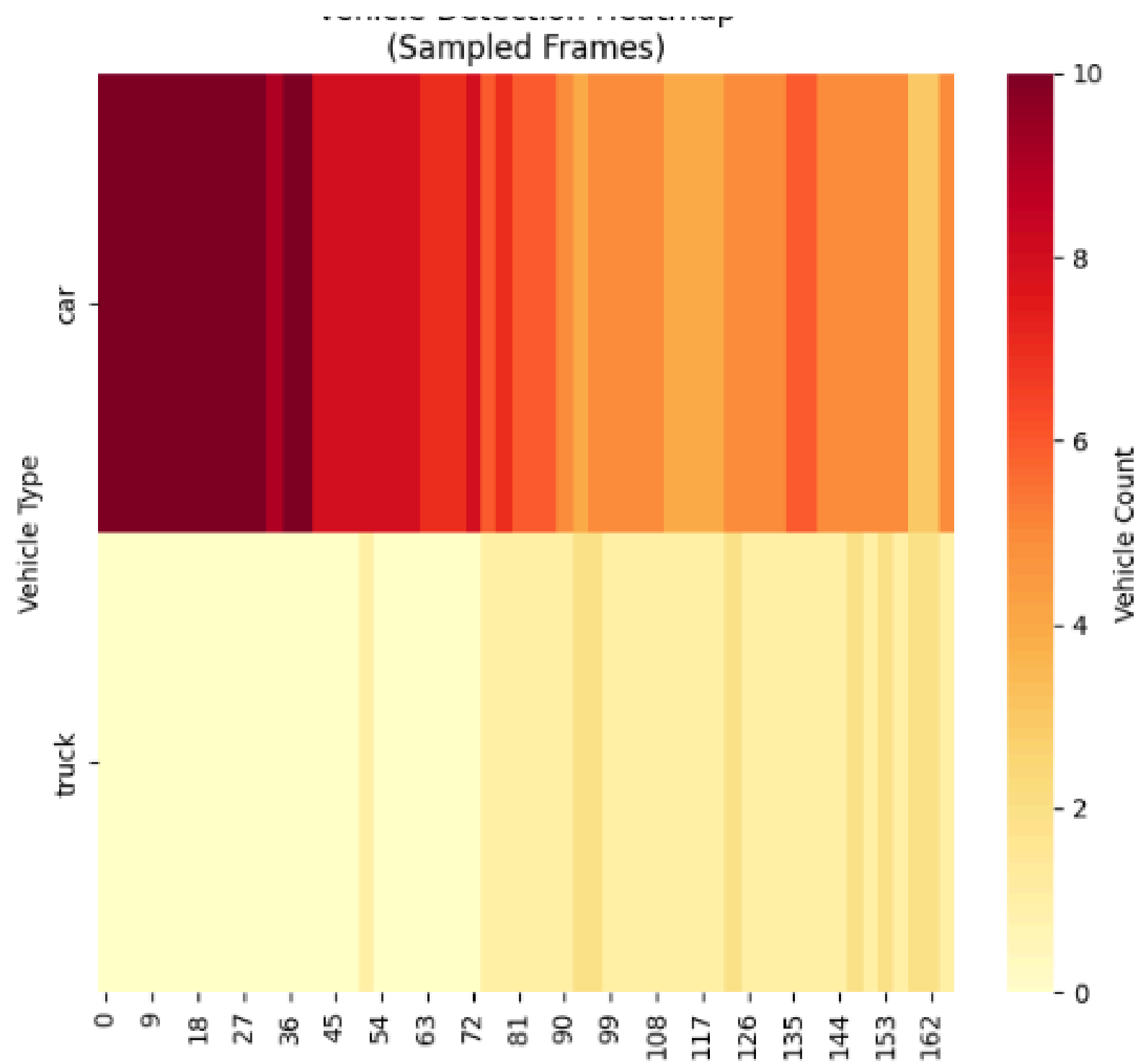
Overall Vehicle Distribution



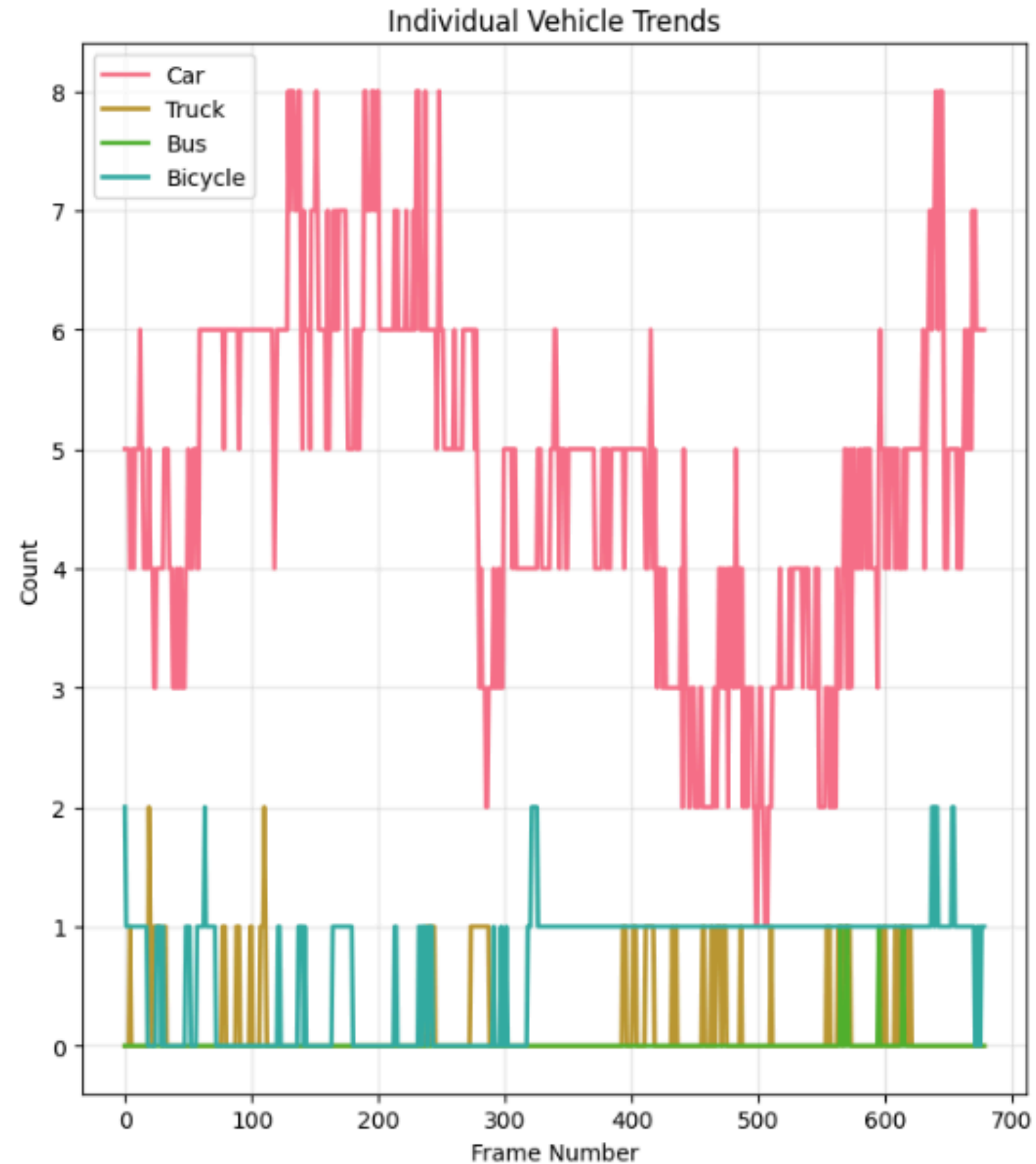
Vehicle Count Over Time



# EXPLORATORY DATA ANALYSIS



# EXPLORATORY DATA ANALYSIS



Vehicle	Total	Mean/Frame	Std Dev	Max/Frame
Car	3225	4.75	1.45	8
Truck	97	0.14	0.36	2
Bus	4	0.01	0.08	1
Bicycle	449	0.66	0.51	2

# MODEL EVALUATION

 Classification Metrics (Confidence Threshold: 0.25):

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Precision: 0.9102 (91.02%)

Recall: 1.0000 (100.00%)

F1-Score: 0.9530 (95.30%)

Accuracy: 0.9102 (91.02%)

 Detection Statistics:

True Positives: 1247

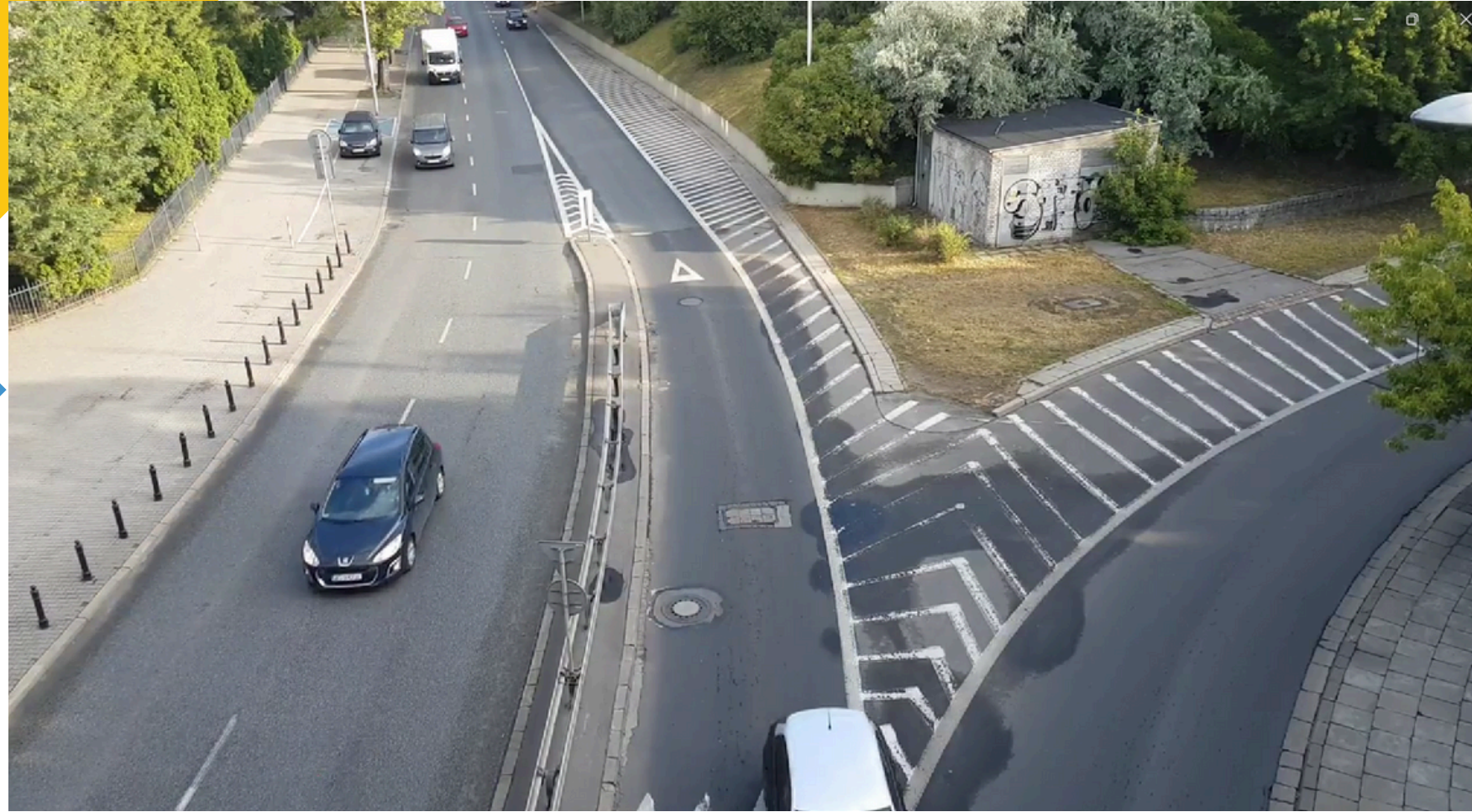
False Positives: 123

False Negatives: 0

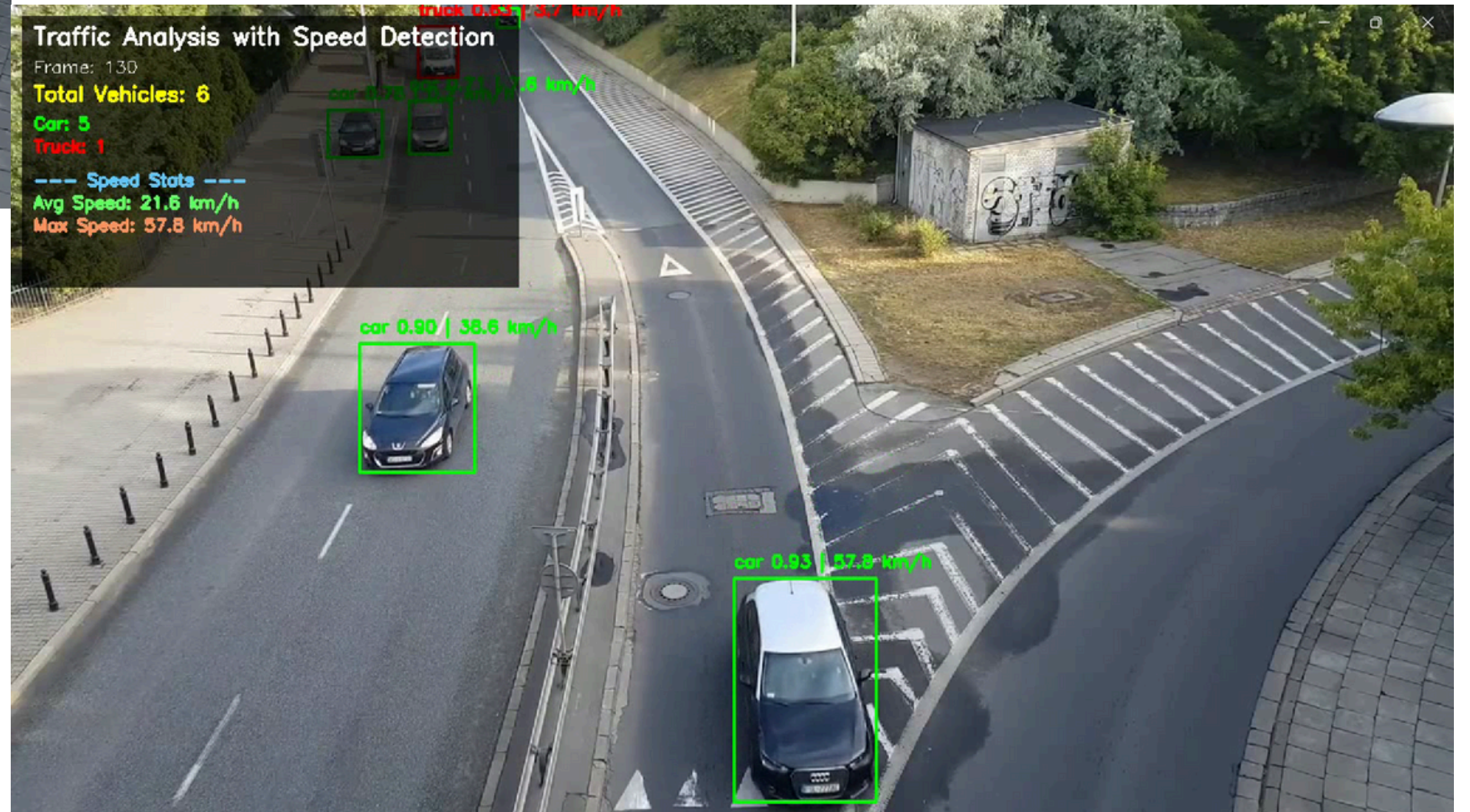
True Negatives: 0



# RESULTS



INPUT



OUTPUT



# CONCLUSION

- Automated real-time vehicle detection, classification, and counting achieved using YOLO with COCO pre-trained model.
- Accurately identifies vehicles: cars, buses, trucks, motorbikes, and bicycles.
- Reduces manual monitoring efforts and improves traffic management efficiency.
- Provides structured data for smart traffic control, accident prevention, and urban planning.
- Implementation in Google Colab demonstrates scalable and practical real-time surveillance.



**Thank You!**