



edunet  
foundation

# Smart Traffic Detection via YOLOv7

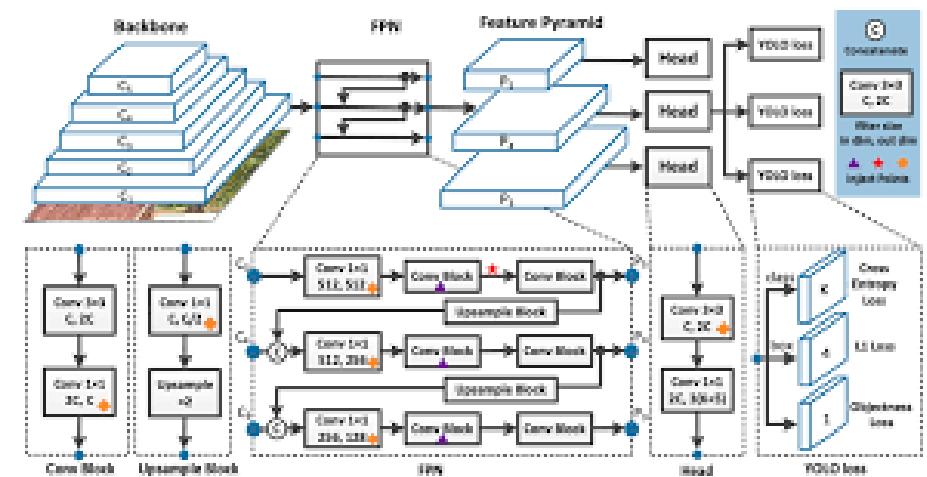
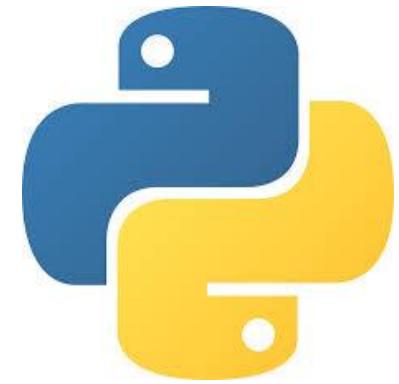
## Learning Objectives:

- To aid for **sustainable development** using **Machine-Learning**, especially, **Deep-Learning** models, to classify objects/actions.
- To gain hands-on experience in creating a **multi-process** pipeline that helps in sustainable **city-development**.
- Analyze how automation and data-overlays contribute to **SDG 11 (Sustainable Cities and Communities)**.
- Contribute to the analysis of **Fuel-Consumption** and **Emission** at traffic junctions, thereby helping in implementation of techniques that reduces them.



## Tools and Technology used:

- **OpenCV**(Initializes camera for real-time detection).
- **YOLOv7**(Trained with custom-made dataset).
- **Python**(Coding and Libraries utilized for architecture).
- **Tensorboard**(To track training rate and log similar criteria).
- **LabelIMG**(For creating custom dataset).
- **Tkinter**(GUI Development).



## Methodology:

- **Custom Dataset Preparation** – Annotated vehicle classes (Truck, Bus, Car, Motorcycle) using **LabelImg** via hand-picked traffic footage.
- **Model Training & Optimization** – Trained YOLOv7 on the custom dataset using **PyTorch**, monitored via **TensorBoard**, and exported weights.
- **Real-Time Detection**- Deployed **detect.py** to process live or recorded feeds and save class-wise label with confidence scores.
- **Threshold-Based Filtering** – Implemented a filter module (**class\_filter.py**) to compute rolling averages, and apply per-class thresholds and write structure JSON thresholds.
- **Signal Decision & Overlay** – Designed a decision module (**dig\_tr\_sgn.py**) to read JSON counts, and to trigger signals based on thresholds.
- **GUI-Orchestrated System Integration** - Built a Tkinter-based control panel to launch, monitor, and terminate subprocesses, enabling modular, user-friendly orchestration of the full pipeline.

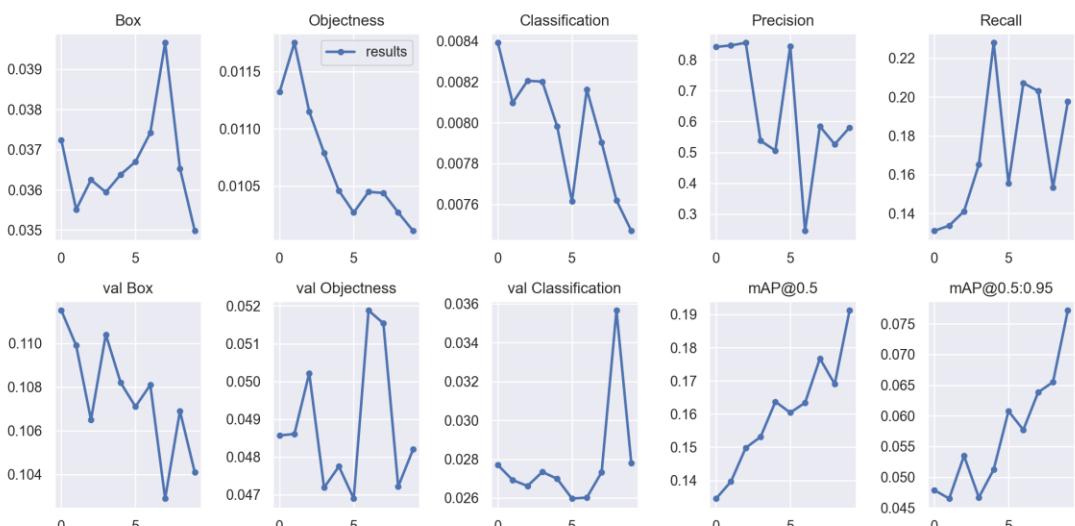
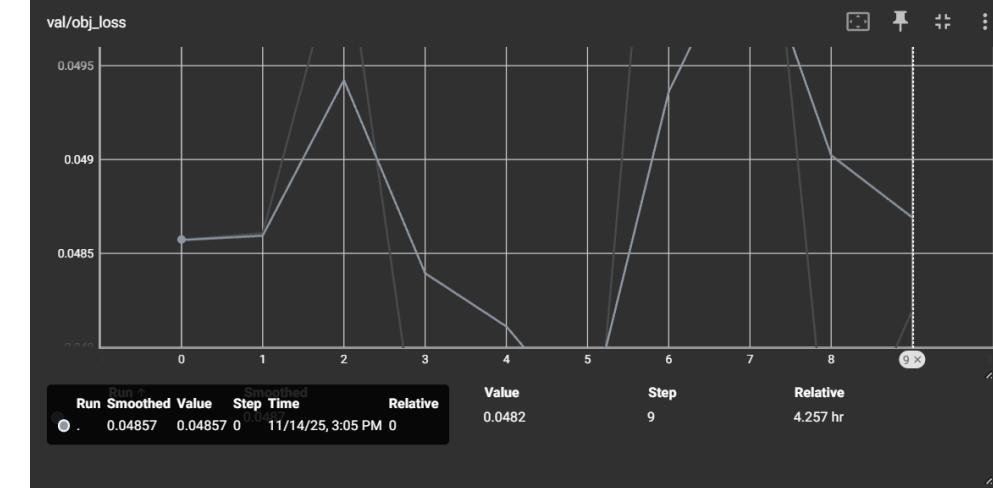
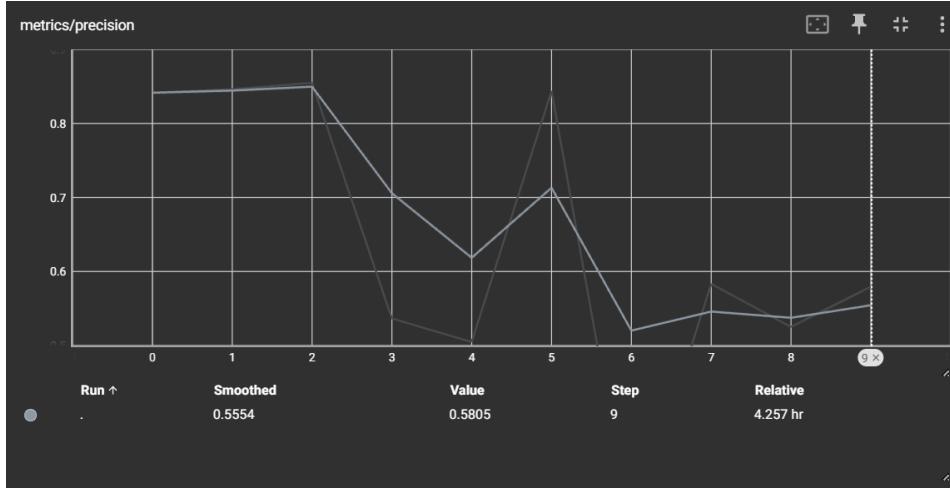
## Problem Statement:

- **Urban intersections** suffer from inefficient traffic signal timing, leading to prolonged idle times and increased fuel consumption.
- **Manual traffic control** systems lack real-time responsiveness, making it difficult to adapt to fluctuating vehicle densities and class types.
- **High congestion levels** contribute to elevated carbon emissions, noise pollution, and commuter frustration, undermining sustainable mobility goals.
- **Existing surveillance** infrastructure often lacks intelligent automation, resulting in underutilized data and missed opportunities for optimization.
- There is **limited integration** between detection, decision-making, and visualization, making it hard for operators to monitor and act dynamically.

## Solution:

- **Real-Time Vehicle Detection with YOLOv7** - Deployed a custom-trained YOLOv7 model to identify and classify vehicles (truck, bus, car, motorcycle) from live or recorded video feeds with high accuracy.
- **Rolling Average-Based Threshold Filtering** - Implemented a dynamic filtering mechanism that computes class-wise rolling averages over recent frames to smooth out noise and ensure reliable signal decisions.
- **Automated Traffic Signal Decision Logic** - Designed a rule-based system that triggers GREEN signals when specific vehicle class thresholds are met, reducing idle time and improving flow efficiency.
- **Live Overlay with Visual Feedback** - Used OpenCV to render real-time overlays on the video feed, displaying signal status, vehicle counts, and trigger reasons for operator transparency.
- **Modular, Multi-Process Architecture** - Separated detection, logging, and overlay into independent scripts communicating via JSON, enabling parallel execution and easy scalability.
- **User-Friendly GUI for System Control** - Built a Tkinter-based dashboard to launch, monitor, and stop the pipeline with configurable inputs, making the system accessible and operator-ready.

## Screenshot of Output:



**Traffic AI Control Panel**

Source (file path or camera index): N:\College\Anotha\_Internship\Stock\_Footage\Traffic Jam Stock Video

Config

- Confidence (float): 0.25
- Car threshold: 10
- Bus threshold: 1
- Truck threshold: 2
- Motorcycle threshold: 3

Status / Logs

```

DETECT_OUT: video_1/1 (6/383) N:\College\Anotha_Internship\Stock_Footage\Traffic Jam Stock Video.mp4: [DEBUG] Frame 0 -> det count: 15
DETECT_OUT: [DEBUG] Detected class=2, conf=0.2513427436351776
DETECT_OUT: [DEBUG] Detected class=2, conf=0.3017696142196655
DETECT_OUT: [DEBUG] Detected class=2, conf=0.3502651853546053
DETECT_OUT: [DEBUG] Detected class=2, conf=0.3302500733814524
DETECT_OUT: [DEBUG] Detected class=2, conf=0.3542840733847754
DETECT_OUT: [DEBUG] Detected class=2, conf=0.3976072072987788
DETECT_OUT: [DEBUG] Detected class=2, conf=0.414939247200012
DETECT_OUT: [DEBUG] Detected class=2, conf=0.44187188148498535
DETECT_OUT: [DEBUG] Detected class=2, conf=0.4518336057662564
DETECT_OUT: [DEBUG] Detected class=2, conf=0.4916859810458557
DETECT_OUT: [DEBUG] Detected class=2, conf=0.5271449493003546
DETECT_OUT: [DEBUG] Detected class=2, conf=0.631329178101196
DETECT_OUT: [DEBUG] Detected class=2, conf=0.7075069546699524
DETECT_OUT: [DEBUG] Detected class=2, conf=0.723440106834412
DETECT_OUT: [DEBUG] Labels written to: N:\College\Anotha_Project\yolov7\run\ct\gui_run_20251116_130328\labels\Traffic Jam Stock Video_9.txt
DETECT_OUT: [DEBUG] 15 cars, Done: (519.7ms) Inference, (1.0ms) NMS
DETECT_OUT: [DEBUG] video_1/1 (6/383) N:\College\Anotha_Internship\Stock_Footage\Traffic Jam Stock Video.mp4: [DEBUG] Frame 0 -> det count: 14
DETECT_OUT: [DEBUG] Detected class=2, conf=0.2529621906262384
DETECT_OUT: [DEBUG] Detected class=3, conf=0.2976831495761873

```

**Launch Pipeline** | **Stop Pipeline**



## Conclusion:

- **Successfully built a real-time traffic detection** and control system and integrated object detection, threshold logic, and visual overlays into a modular, fully functional pipeline.
- **Demonstrated the power of AI in urban mobility optimization** and displayed how smart automation can reduce congestion/traffic.
- **Engineered a scalable, multi-process architecture**, thereby enabling parallel execution of detection, logging, and visualization, adaptable to multi-feed deployments.
- **Aligned technical innovation with sustainability goals** resulting in reduced idle emissions and fuel waste, contributing to cleaner, smarter cities in line with **SDG 11**.