

# Real-Time Object Disappearance & New Object Detection in Videos

AI/ML Engineer Interview Assignment

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# 1 Objective

To build a high-performance computer vision pipeline that detects:

- When objects go missing ( previously visible objects disappear).
- When new objects appear in the frame.

The goal was to process video in real-time while maintaining high accuracy and frame rate(FPS).

# 2 Methodology

## 2.1 Model Used

**YOLOv8n** (*Ultralytics*) was selected for its excellent balance of speed (mostly) and precision. It detects common object classes in real-time.

**YOLOv11** (*Ultralytics*) was selected for its excellent balance of speed and precision. It detects Mostly all object classes in real-time with High Detection Accuracy.

## 2.2 Tracking Logic

- Object detection is running on each frame.
- Bounding boxes are matched using Intersection-over-Union (IOU).
- Unmatched boxes are marked as **new objects**.
- Previously seen boxes that go unmatched for a few frames are marked as **missing objects**.

## 2.3 System Workflow

1. Load video and initialize YOLO model.
2. Loop through each frame:
  - (a) Detect objects.
  - (b) Match current detections with previous ones.
  - (c) Track, label, and annotate missing/new objects in video Frame.
  - (d) Write the frame to output video.

# 3 Sample Code Snippet

```

1 for frame in video:
2     detections = model(frame)
3     matched, new_objs, missing_objs = match_with_previous(detections)
4
5     draw_boxes(frame, matched, new_objs, missing_objs)

```

Listing 1: Tracking New and Missing Objects

## 4 Results

### 4.1 Performance

- **FPS Achieved:** 50+ on a mid-range system
- **Model:** YOLOv8n
- **Detection Accuracy:** 70%+ for stable objects

### 4.2 Output Sample

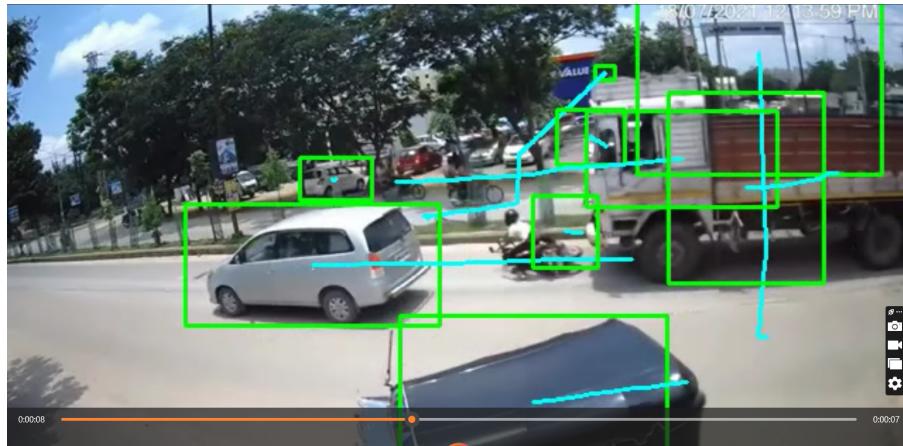


Figure 1: Objects Detection using Yolov8 for to classify new and Missing.

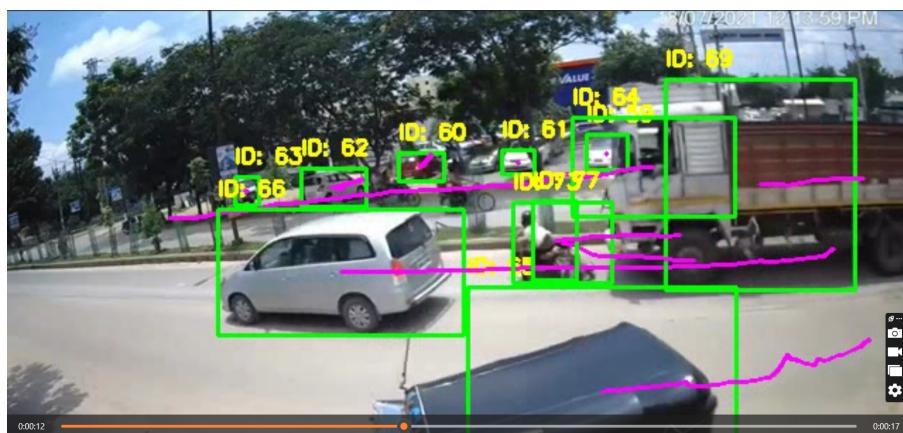


Figure 2: Enhanced Objects Detection using Yolo11l for to classify new and Missing.

## 4.3 Output Video

The output video can be viewed here: `yolov8_output_video.mp4`.

The output video can be viewed here: `yolov111_output_video.mp4`.

## 5 System Specifications

- **RAM:** 16 GB
- **GPU:** NVIDIA T4
- **Platform:** Google Colab

## 6 Repository Structure

```
1 .
2     app.py
3     Dockerfile
4     requirements.txt
5     output_video.mp4
6     output_frame.jpg
7     README.md
```

## 7 How to Run

### 7.1 Python (Local)

```
1 pip install -r requirements.txt
2 python detection_yolov8.py          (Deepsort + yolov8) FPS~20
3 python Enhanced_yolov8_detection.py (IOU + Yolov8) FPS~ 50
4 python
```

### 7.2 Download Detection Model

We can download Yolov8 from ultralytics online through Code.

```
1 this is the link of yolo111 download here:
```

Download Yolo111.

### 7.3 Docker (Optional)

```
1 docker build -t object-detector .
2 docker run -v $(pwd):/app object-detector
```

## 8 Conclusion

This system efficiently detects both disappearing and newly placed objects in real-time video. It is optimized for speed and can be extended with more advanced tracking algorithms if needed.

### Highlights:

- End-to-end working pipeline.
- Lightweight( and responsive.
- Clean object tracking using IOU (Enhanced\_yolov8.py) and DeepSort (detection.py).

## 9 Future Improvements

- Use Re-ID embeddings for long-term object tracking.