

# Automating Infrastructure Assessment

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IVORY Mid-Term Conference – “AI for Vision Zero in Road Safety”

Athens, 15 April 2026



This project has received funding from the European Union's Horizon Europe research and innovation programme under grant agreement No 101119590

# Introduction



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# Problem at Hand

- ❖ Most existing road assessment methods are manual
- ❖ High Cost and Resource Intensity
- ❖ Time consuming
- ❖ **The Shift Toward Automation:**  
by using **computer vision**, road infrastructure assessment can be automated and significantly accelerated.

Example of Time spend for detecting road attribute by manual/human analysis based approaches [1]

Attribute	Time [min/km]
Roadside severity Driver side object	15
Roadside severity Passenger side object	15
Intersection Type	9.4
Intersection channelization	9.4
Intersecting road volume	9.4
Intersection quality	9.4
Pedestrian crossing facilities – inspected road	9.4
Pedestrian crossing facilities – intersecting road	9.4
Pedestrian crossing quality	9.4
Land Use Driver Side	5
Land Use Passenger Side	5
Property access points	4.8
Sidewalk Driver side	3.7
Sidewalk Passenger side	3.7

[1] Saving Lives Assessing and Improving TEN-T Road Network Safety (SLAIN), Deliverable 7.3 & 7.4



# Contextualization

- ❖ Imagery data provides **meaningful information about the environment**
- ❖ From aerial imagery:
  - Land use
  - Density
  - Road design
- ❖ From street-view imagery (SVI):
  - Road quality
  - Roadside attributes
  - Visual disturbances
- ❖ Can be further improved with use of **drones** and **LiDAR**
- ❖ Challenging ground-truth **labeling**



# SVI Applications in Transportation Research

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- ❖ Researchers mostly rely on **pre-trained models**, although a few train **specific deep learning models** for crash prediction or perception generalization
- ❖ Results only sometimes converge
  - Higher **building density** leading to increase and decrease in crash risk
  - **Greenery** is consistently associated with higher perceived comfort and socioeconomic status
- ❖ On regression modelling, image-derived attributes tend to play a **secondary role**
- ❖ On the other hand, **clustering** those attributes provides meaningful contextual information with statistically significant effects
- ❖ What about **combining generic** features for context and **specialized** features for safety prediction?



# Road objects Detection & Segmentation



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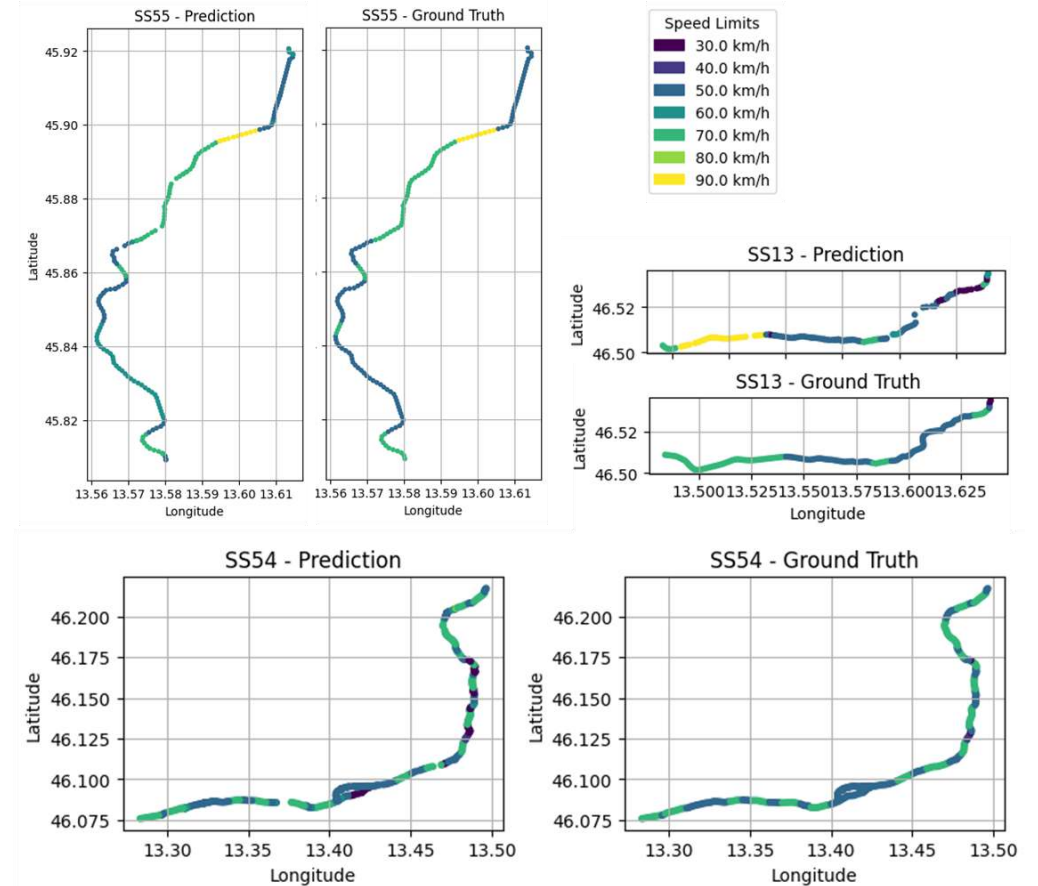
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# Trained Models for Fine-Grained Information

## ❖ Traffic speed limit:

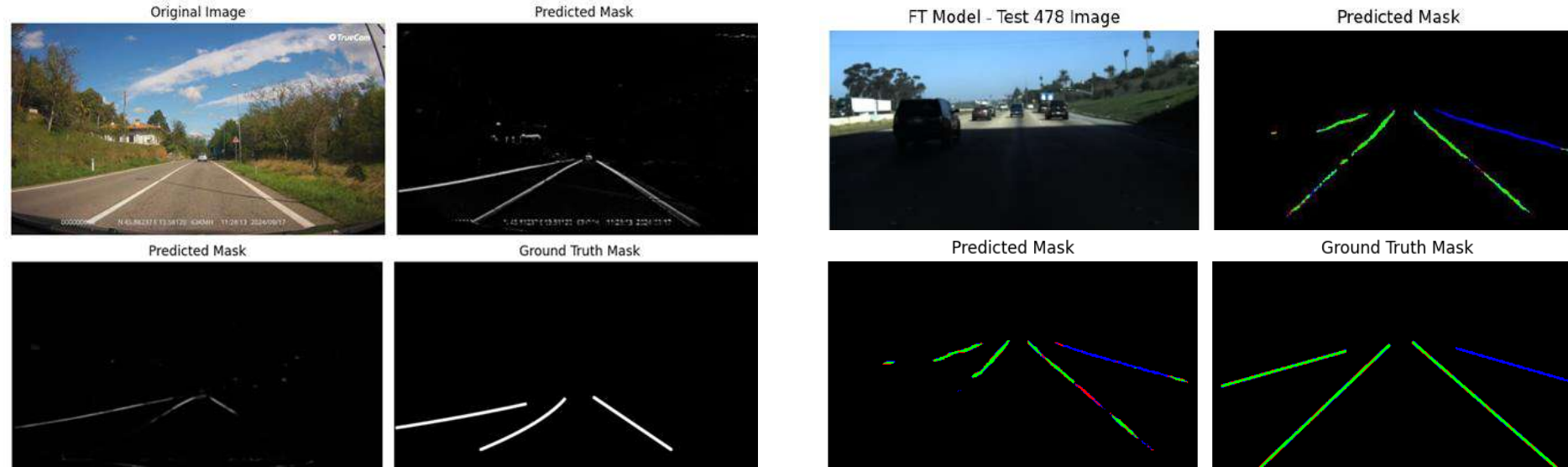
- YOLO and OCR model
- Inclusion of Cancellation Speed Limit signs
- Validation with iRAP-coded data



# Trained Models for Fine-Grained Information

## ❖ Road lane segmentation:

- Weak Supervision or Fine Tuning with Contrastive Loss
- Binary segmentation: label generator for proprietary dataset and TuSimple dataset
- Multiclass segmentation: continuous, dashed and unmarked from TuSimple dataset



# Roadside objects Detection & Distance Calculation using **Yolo model & Lidar data**

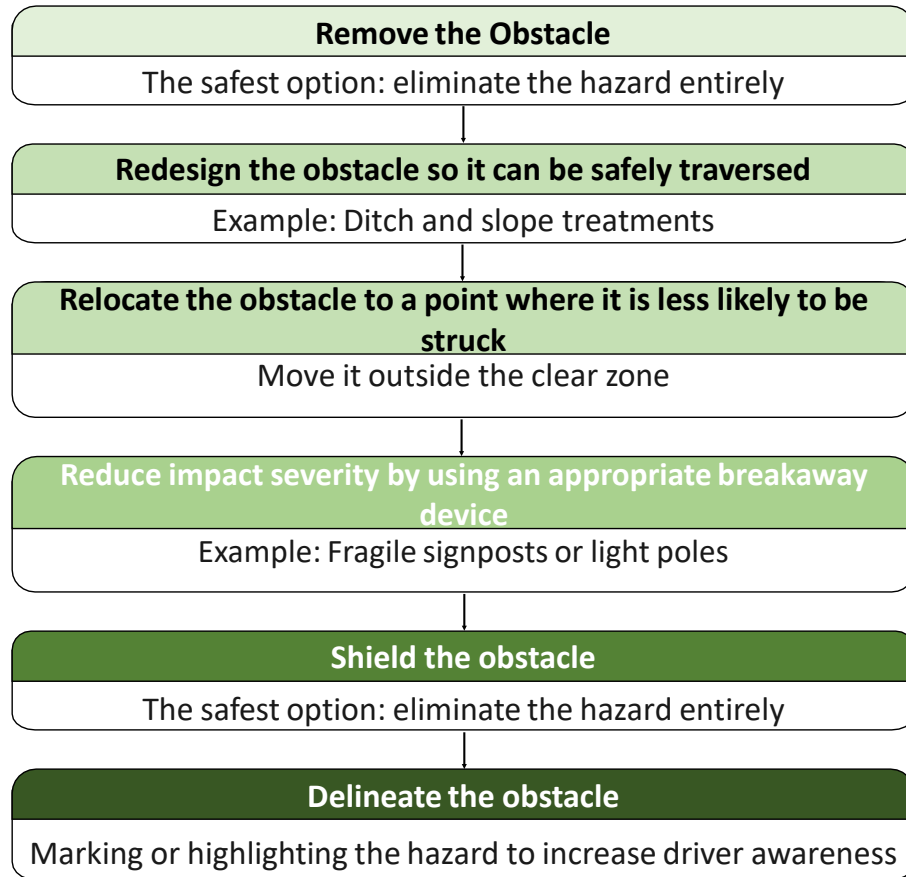


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# Forgiving Roadside Design



Remove and Relocate



Side Slope Improvements



Fragile Light Poles

Source: AASHTO. (2011). Roadside Design; Mitra et al., (2022). Guide to Integrating Safety into Road Design



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# Roadside Safety Attributes

- ❖ Yolo model is trained on Lidar road survey dataset
- ❖ 13 roadside severity objects are detected
  
- ❖ Roadside object, present varying levels of risk based on their distance from the road edge, with objects closer to the traveled way posing a significantly higher hazard to safety. Therefore, we have also calculated their distances as well.

## Road-Side Object Type (13 classes)

1. **Safety Barrier Metal**
2. **Safety Barrier Concrete**
3. **Rockface**
4. **Upward Slope Rollover**
5. **Upward Slope No Rollover**
6. **Drainage**
7. **Down Slope**
8. **Tree**
9. **Pole**
10. **Rigid**
11. **Semi Rigid**
12. **Safety Barrier End**
13. **Rock**



# Roadside severity objects detection using Lidar



RGB Image

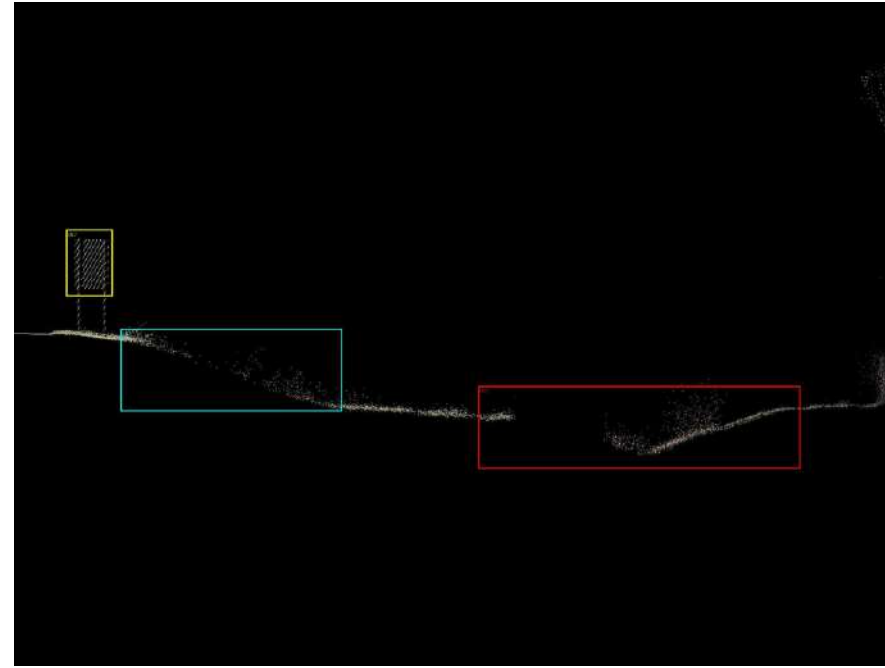
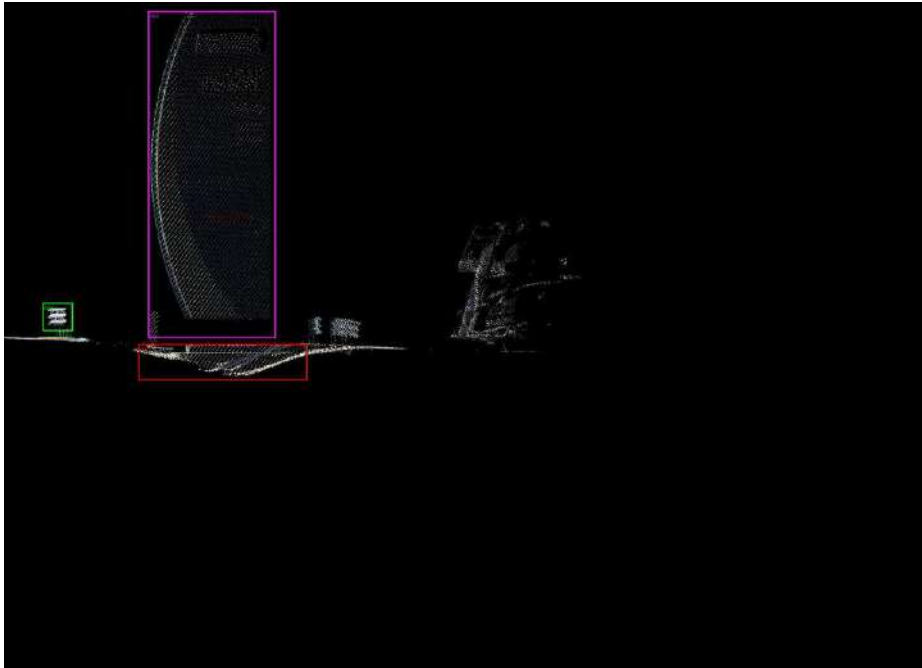


LiDar Image



# Dataset creation- bounding boxes

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LiDar Image – Road Side Objects Bounding Boxes



# Two-Stage Detection Pipeline



## Road Detection

Single-class YOLO model trained on road cross-section images · Outputs road bounding box + right edge X coordinate (road\_edge\_x)

## Edge Extraction & Crop

Road right edge (Xmax) defines transition from road to roadside · Roadside crop defined relative to road\_edge\_x for Stage 2 input

## Object Detection

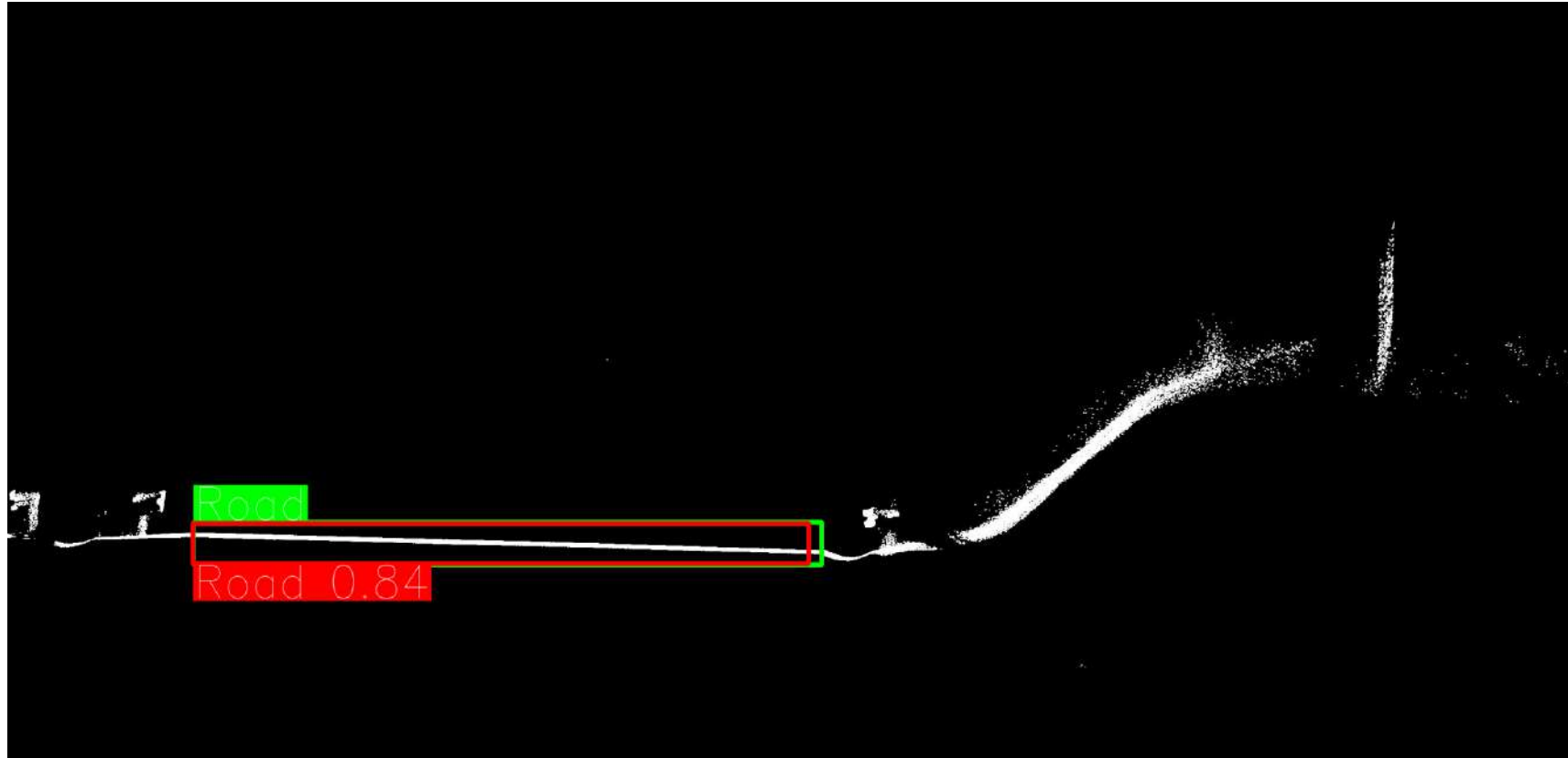
13-class YOLO model identifies roadside hazard type·

## Distance Calculation

Pixel distance from road\_edge\_x to object reference point



# Stage 1: Road detection

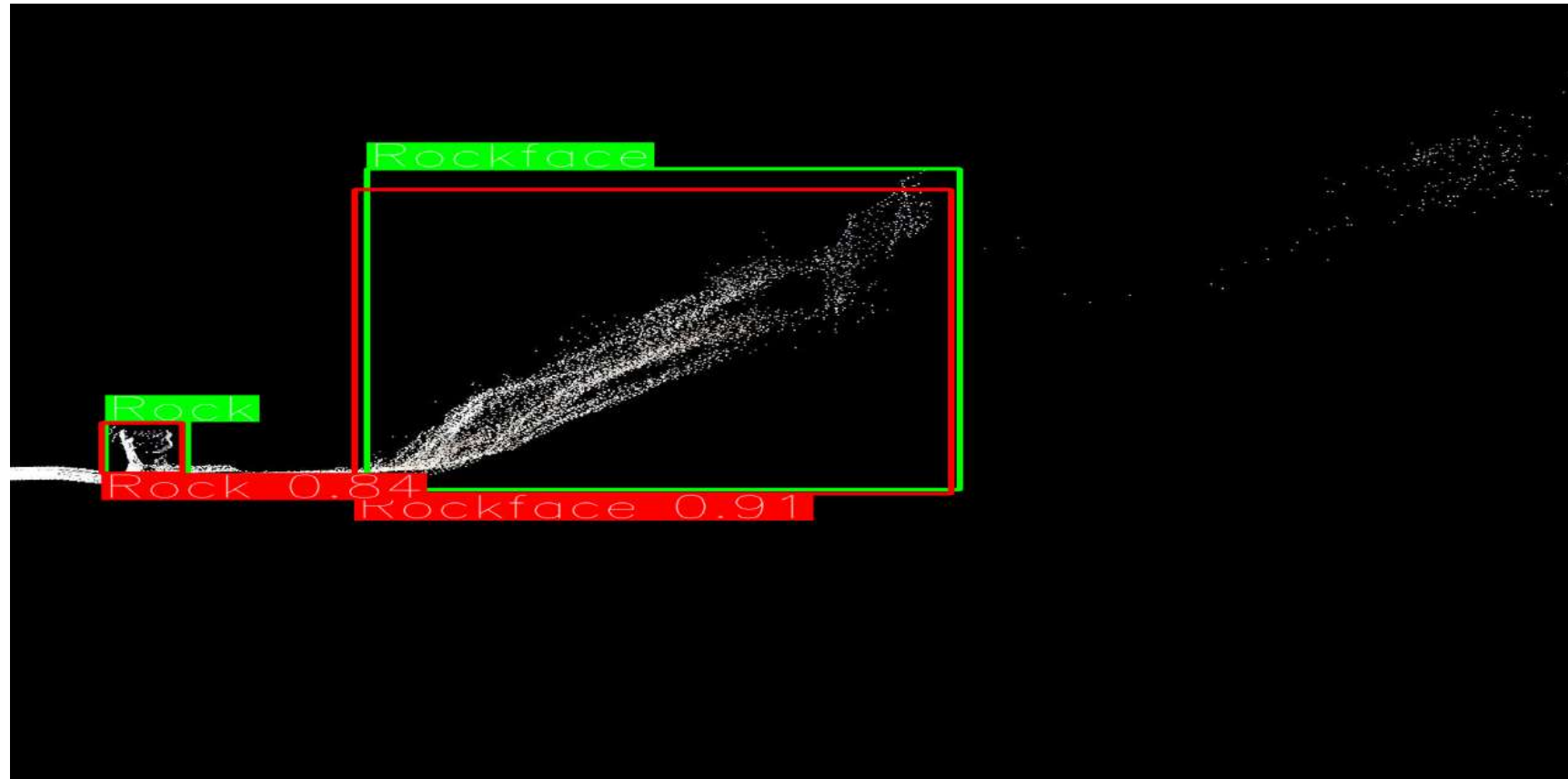


Ground truth

Prediction



# Stage 2: Roadside objects detection



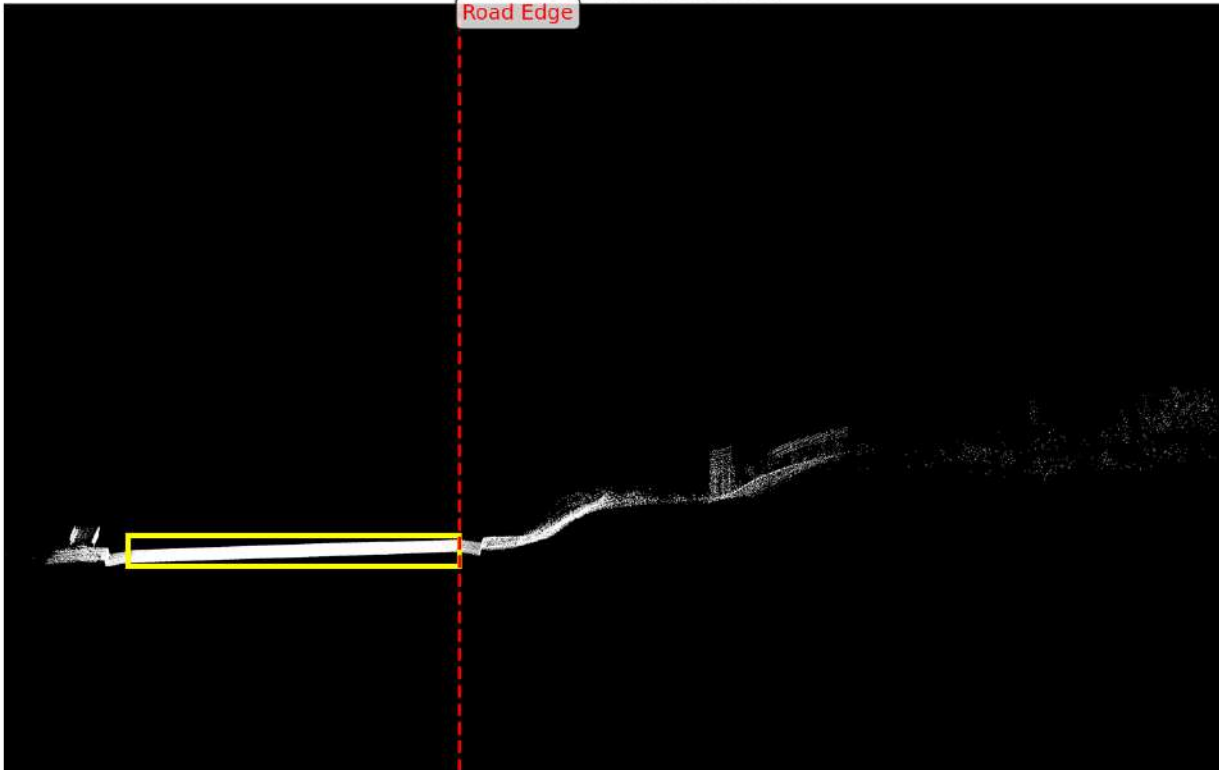
Ground truth

Prediction

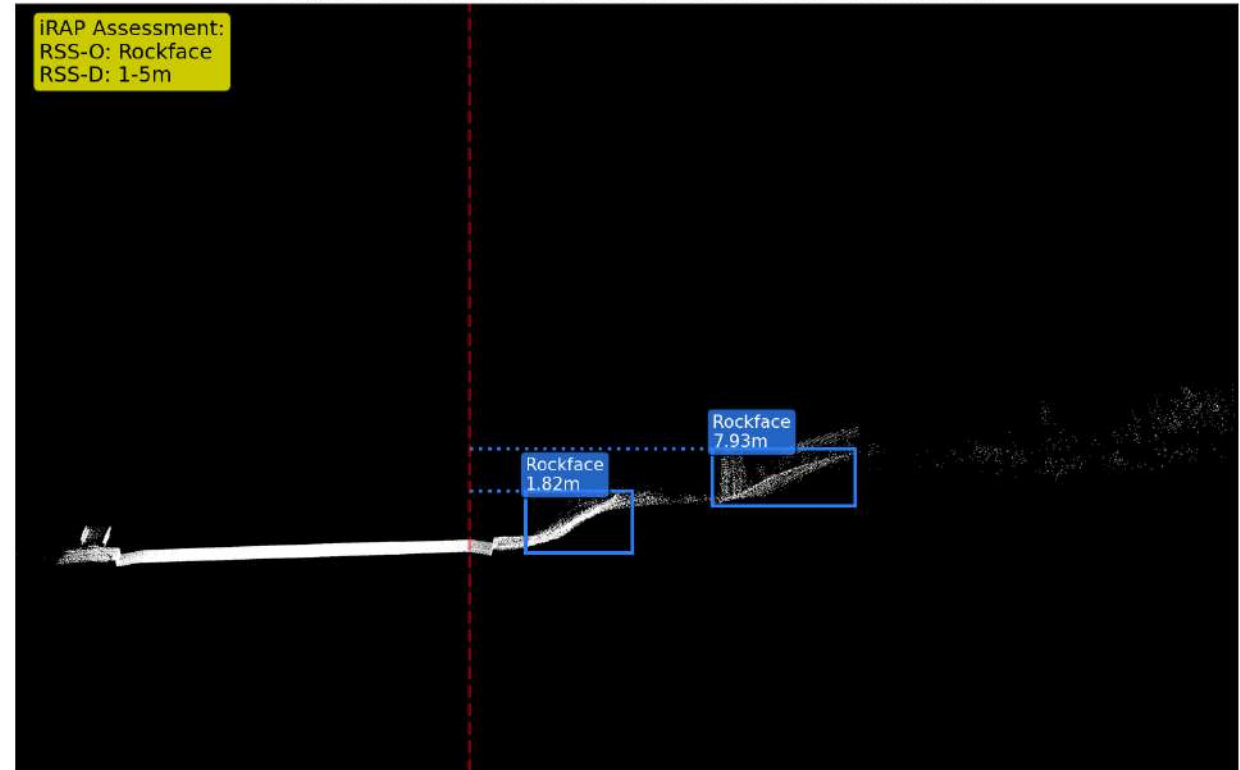


# Model Inference on Test Dataset

Stage 1: Road Detection



Stage 2-4: Severity Detection & Distance Calculation



Yolo model detected the road, then detected the roadside objects after that it calculated the distance of each roadside object to road edge line.



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# Thank you!

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