

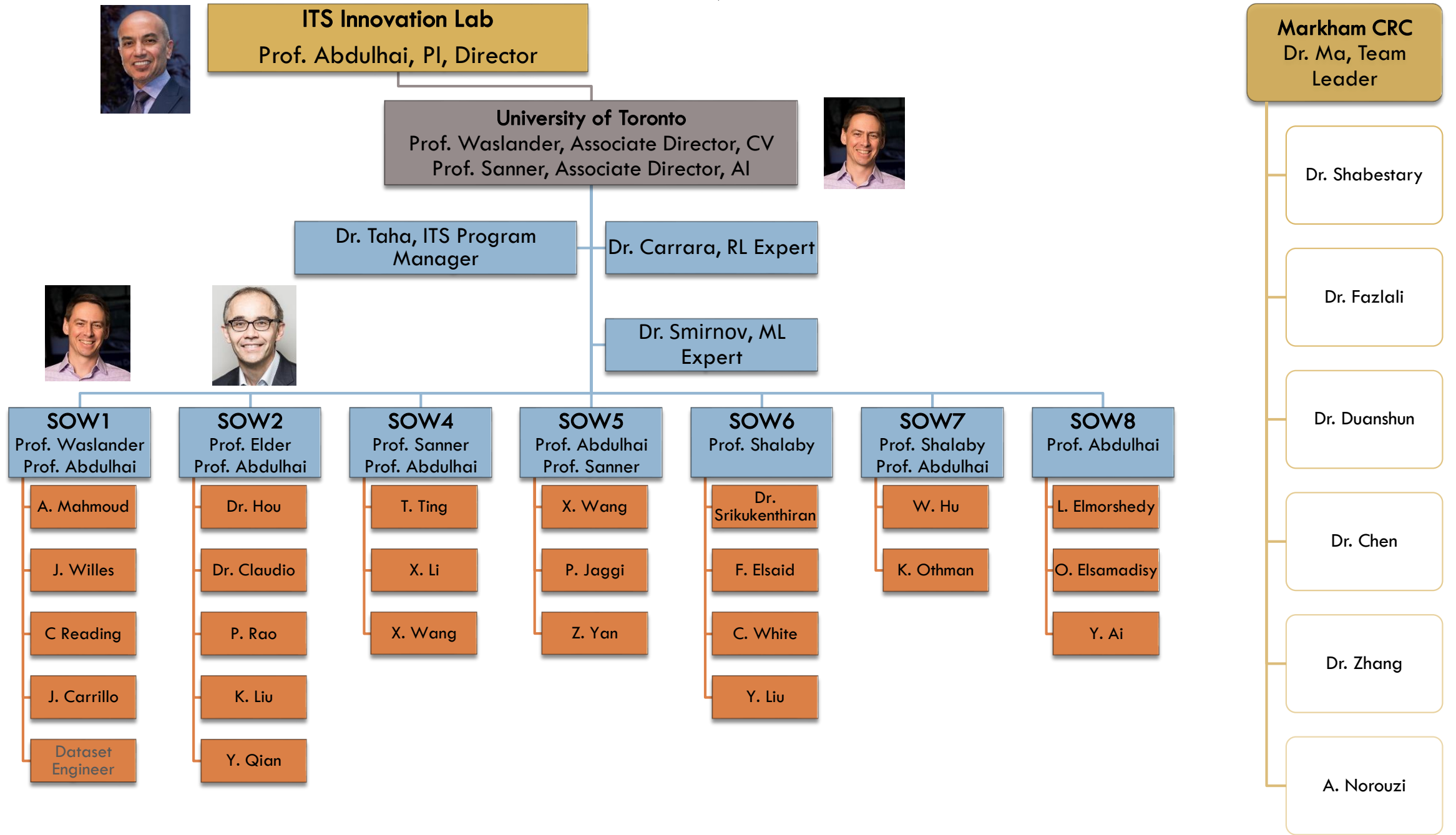


# VISUAL 3D UNDERSTANDING OF MIXED TRAFFIC IN BUSY INTERSECTIONS

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# TEAM



# Research Objectives

- Core Objective
  - ▣ Real-time 3D detection, classification, accurate geo-location and tracking of all road users
    - Cars, trucks, buses, pedestrians, bicycles,...
    - Diverse conditions: day/night, sun, cloud, rain, snow,...
- Derived Applications and Use Cases
  - ▣ Traffic counting and trajectory classification
  - ▣ Accurate speed measurement
  - ▣ Traffic anomaly and incident detection
  - ▣ Detection of near-misses
  - ▣ 3D visualization (digital intersection)
  - ▣ Automatic identification of and attention to vulnerable road users

# Field Site: Highway 7 & Leslie

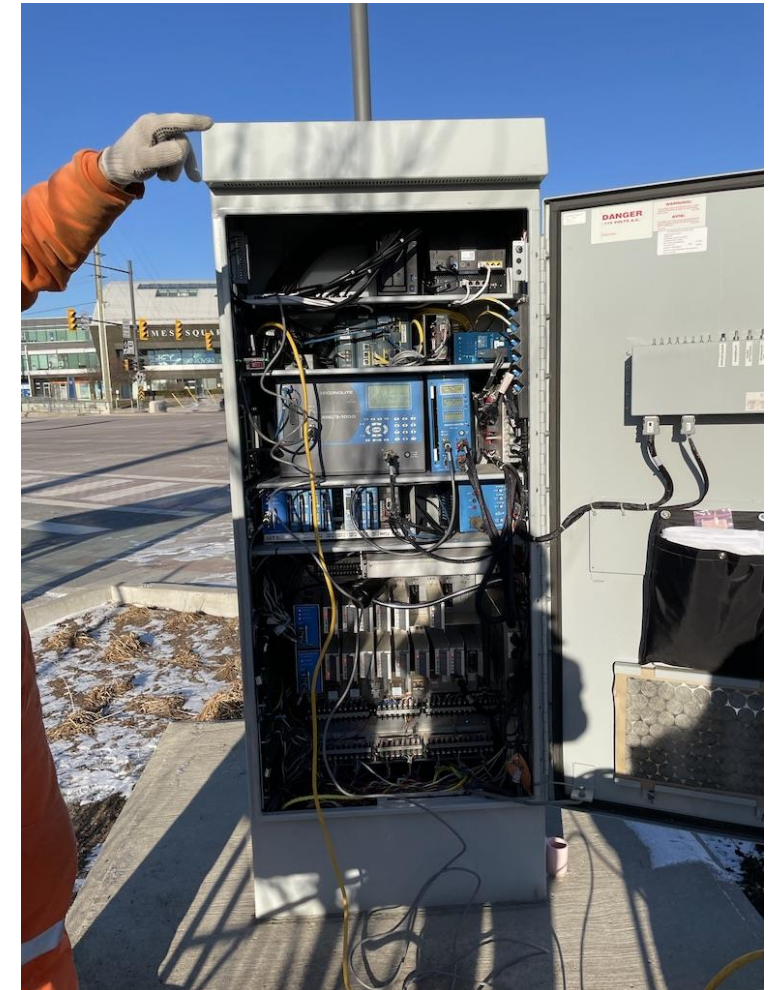
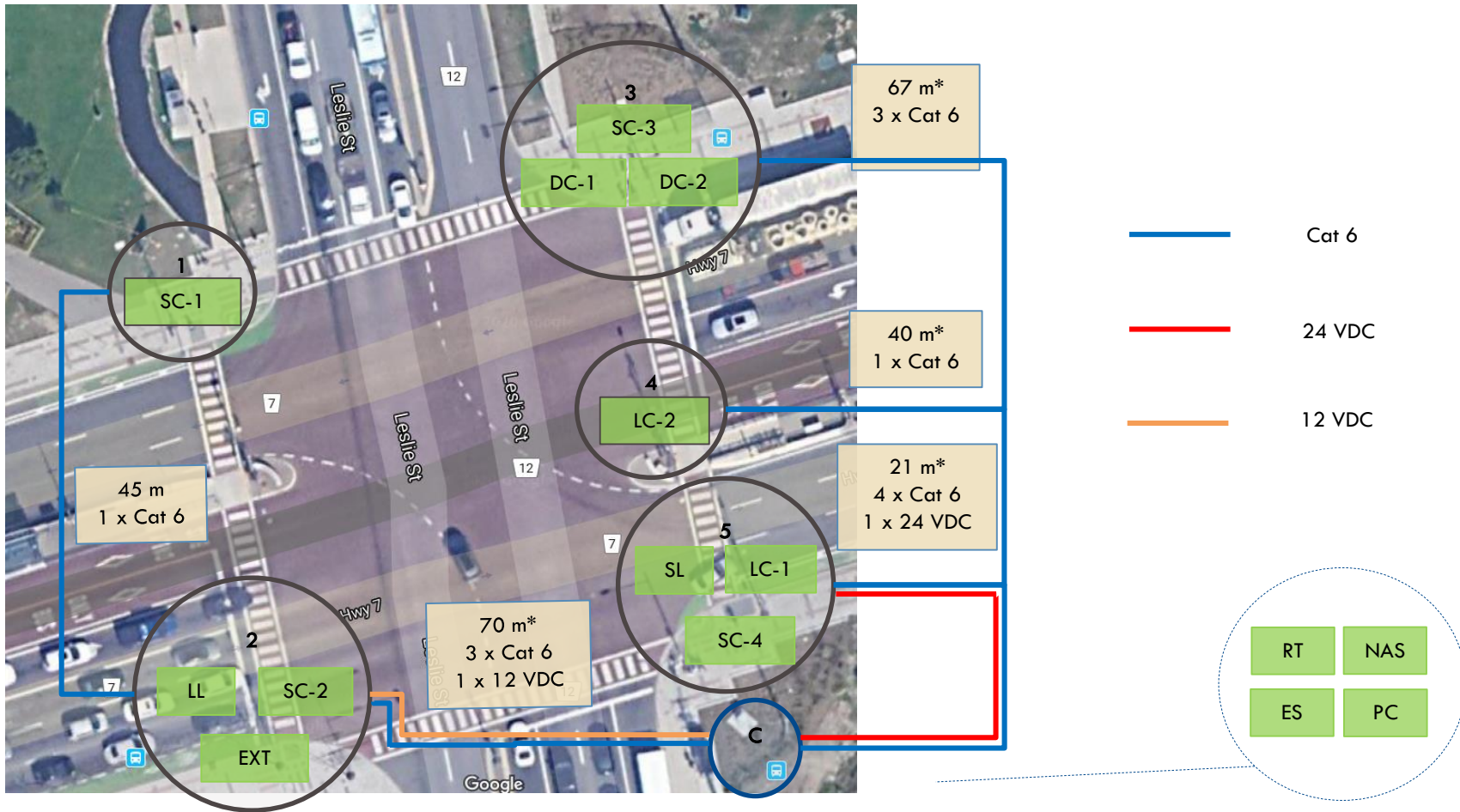
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Transit Stop



# Complete Hardware Layout



# Sensors

Axis Q1798 Camera (SOW 2)

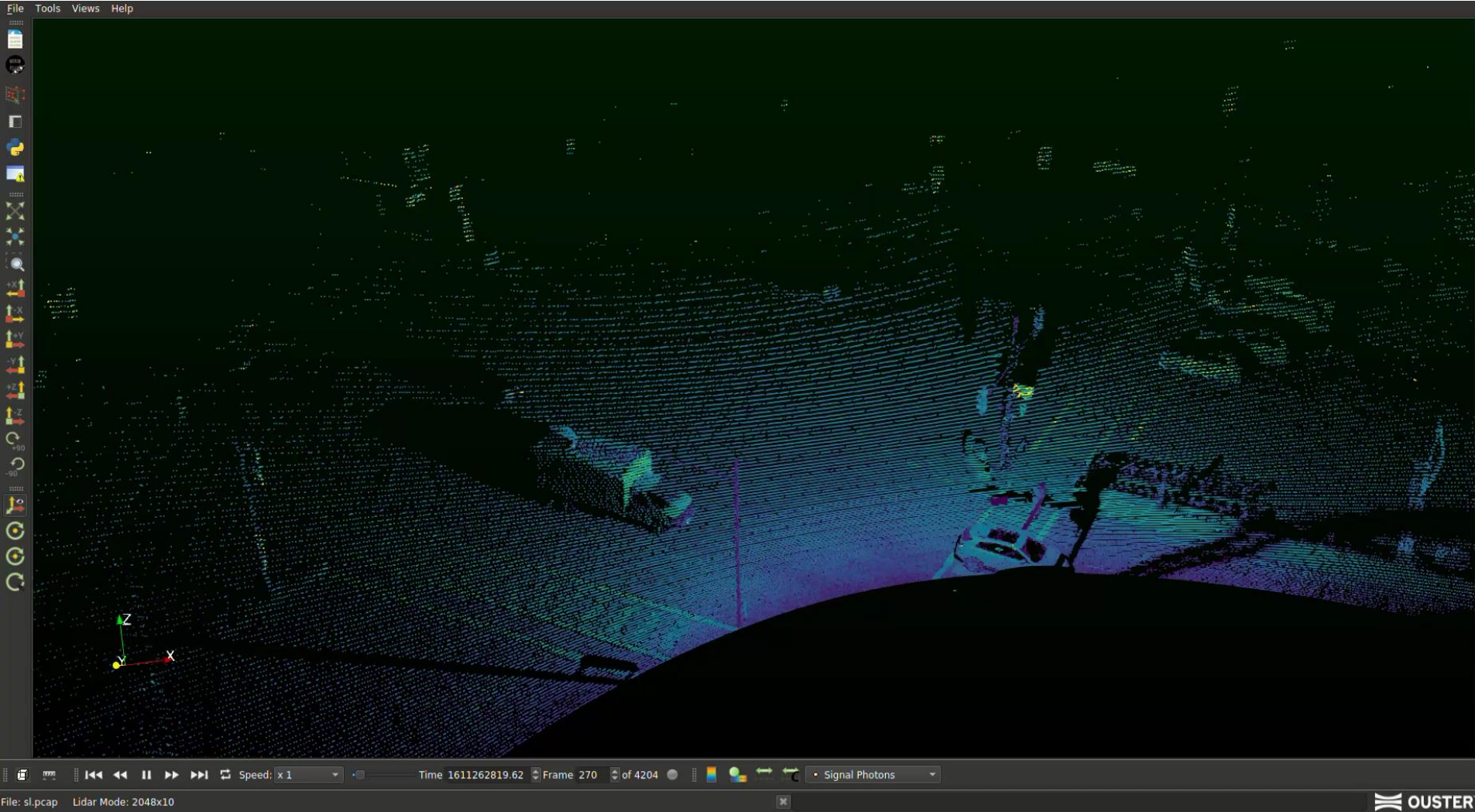


Cepton Vista X LiDAR (SOW 1)



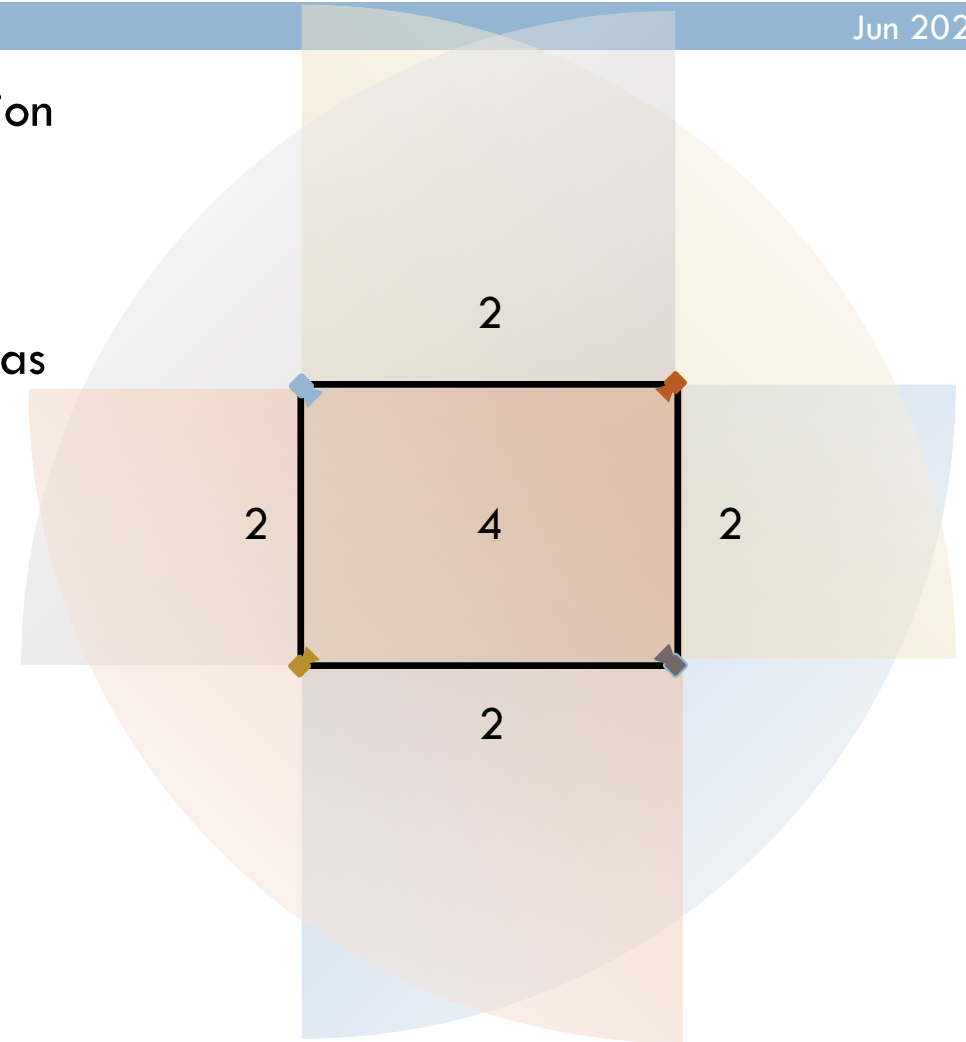


# Intersection LiDAR



# 4-Camera Geometry for Intersection

- ❑ Cameras mounted at 10m height at 4 corners of intersection
- ❑ Oblique angle expected to yield better recognition rates
- ❑ Objects interior to intersection seen by 4 cameras
- ❑ Objects entering and exiting intersection seen by 2 cameras

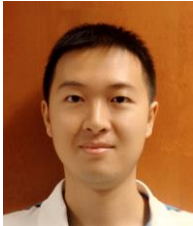
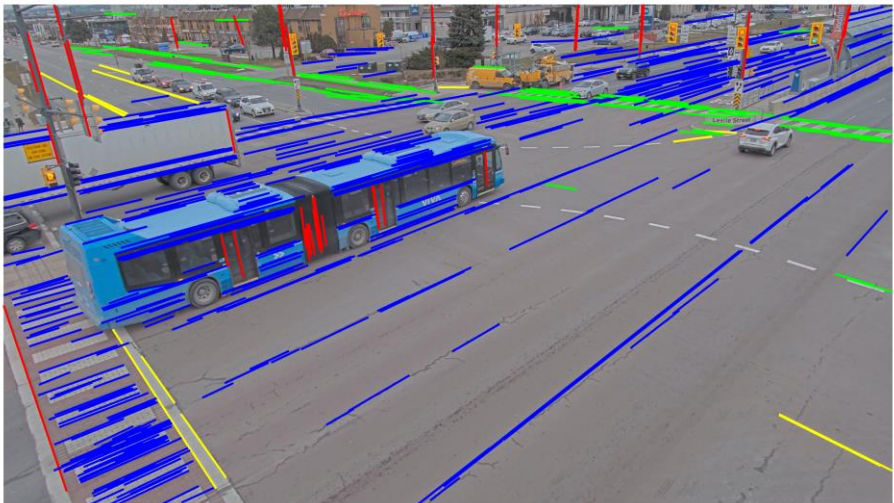
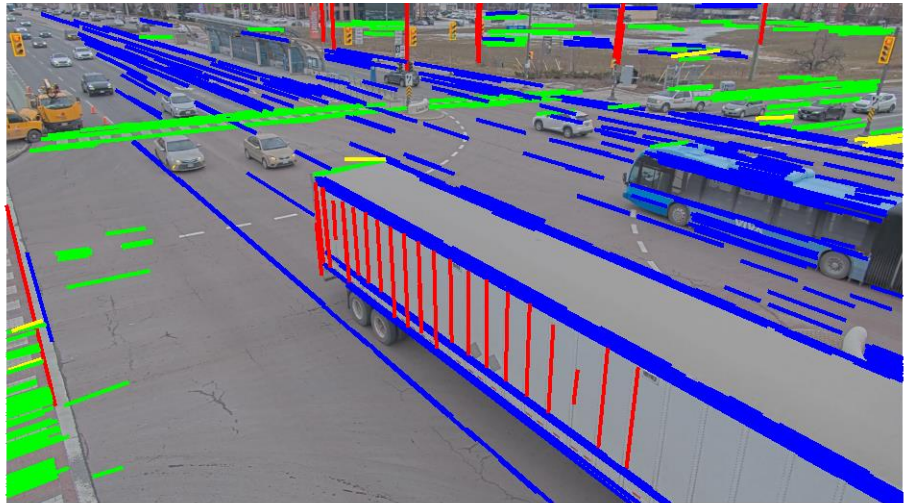
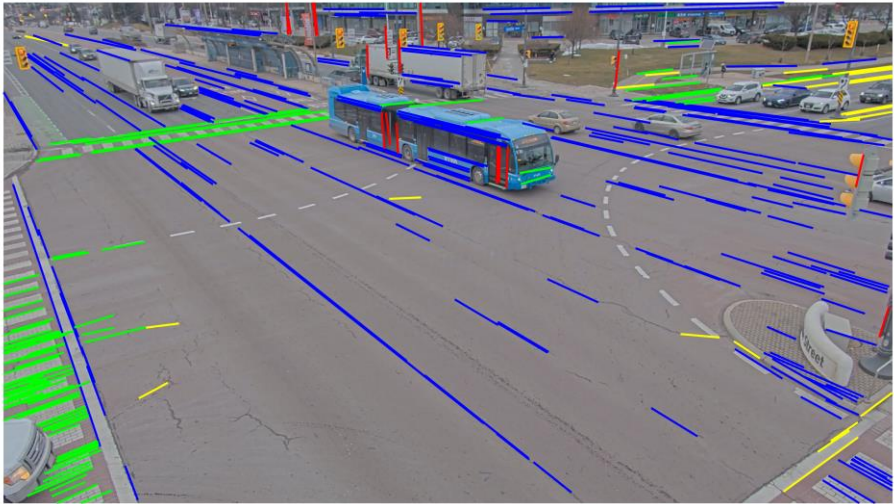
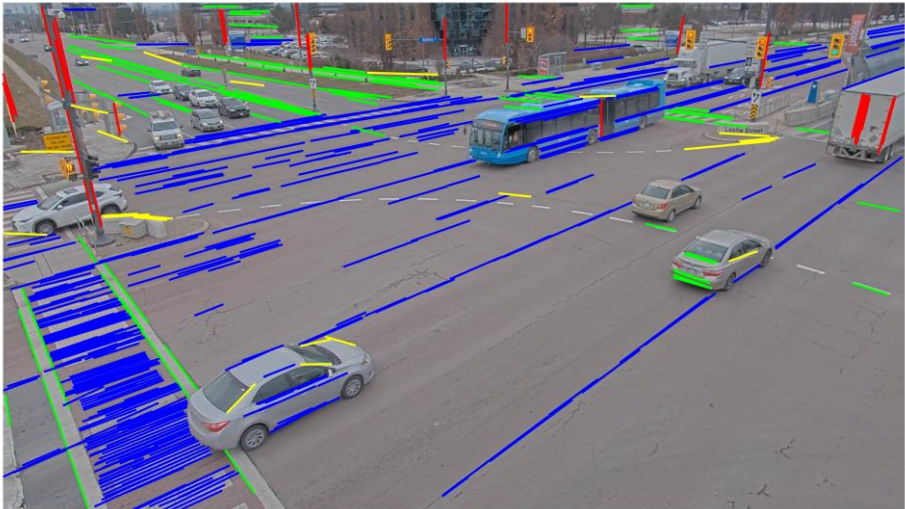




# Current Work

- Online camera calibration
- Object detection and segmentation
- Pipeline for trajectory classification
- Attending to vulnerable road users

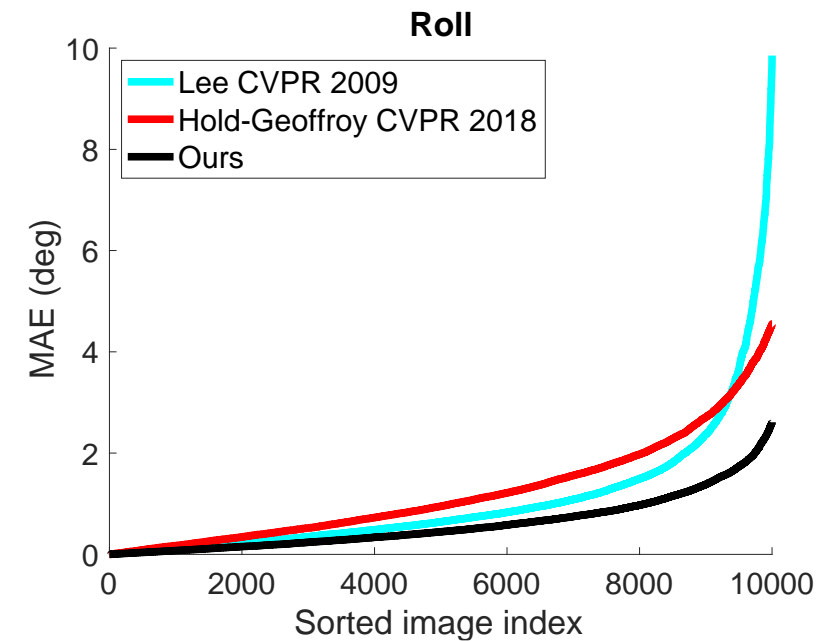
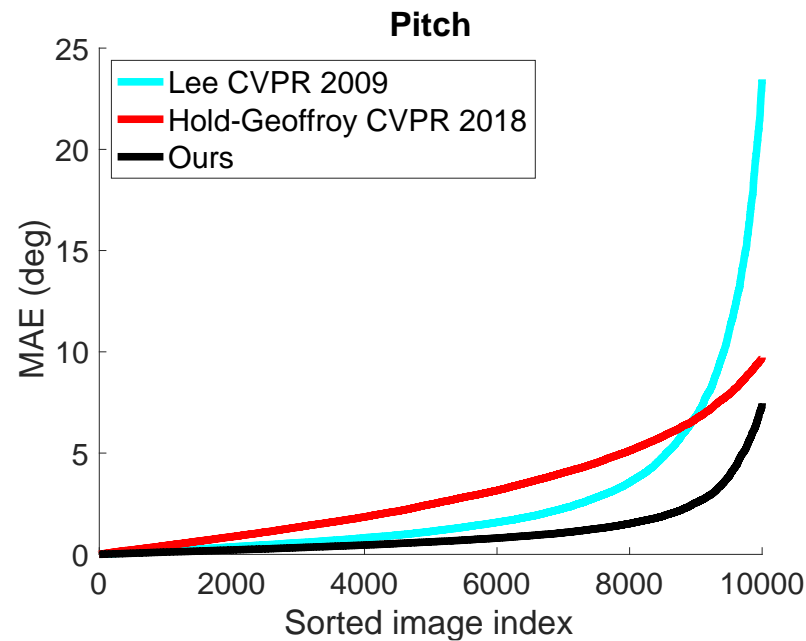
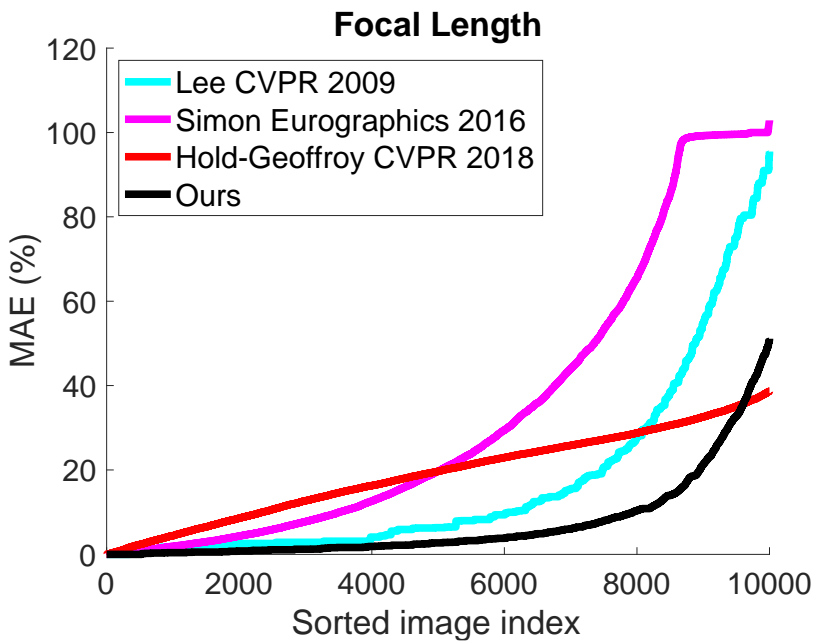
# Online Camera Calibration



Dr. Yiming Qian  
J. Elder

# Improving State-of-the-Art

Images sorted by MAE independently for each algorithm and parameter.





# 2D Vehicle Detection

## □ CenterNet with Resnet50 Backbone trained on MS COCO



Duan *et al.* Centernet: Keypoint triplets for object detection. ICCV 2019.



Keyi Liu

# 2D Vehicle Segmentation

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# 2D Figure/Ground Completion

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# Improving State-of-the-Art

- DTM-based method with sparse coding outperforms both Contour-based and Mask-based top performers
- All methods are trained for 12 epochs on MS COCO with ResNet-50 backbone

Representation	Method	Input Scale	Encoding Dim.	Active Dim.	Mask AP
Contour	ESE-Seg[1]	416	20	20	0.216
	PolarMask[2]	400	72	72	0.229
Mask	MEInst[3]	400	60	60	0.239
	<b>Ours</b>	<b>400</b>	<b>128</b>	<b>57</b>	<b>0.249(+1.0%)</b>
DTM	<b>Ours</b>	<b>400</b>	<b>128</b>	<b>56</b>	<b>0.257(+1.8%)</b>

[1] RuWenqiang Xu, Haiyang Wang, Fubo Qi, and Cewu Lu. Explicit shape encoding for real-time instance segmentation. In ICCV, 2019.

[2] Enze Xie, Peize Sun, Xiaogang Song, Wenhai Wang, Xuebo Liu, Ding Liang, Chun-hua Shen, and Ping Luo. PolarMask: Single shot instance segmentation with polar representation. In CVPR, 2020..

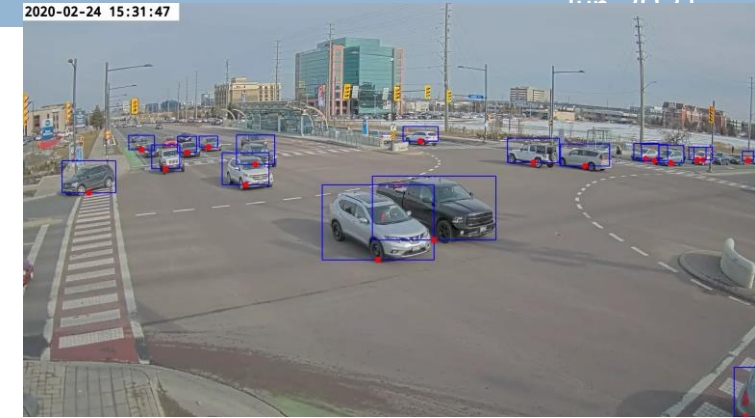
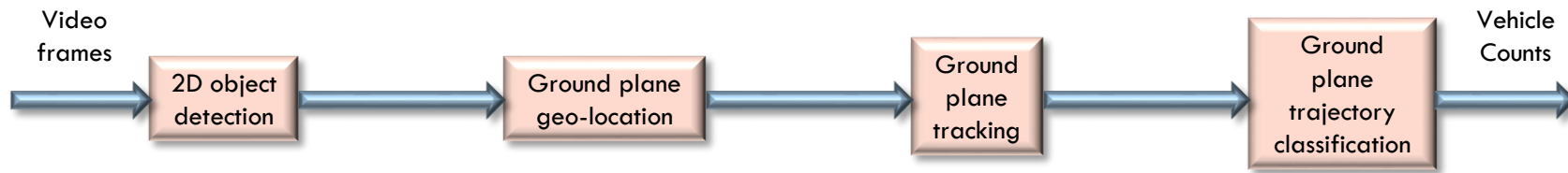
[3] Rufeng Zhang, Zhi Tian, Chunhua Shen, Mingyu You, and Youliang Yan. Mask Encoding for Single Shot Instance Segmentation. In CVPR, 2020.

# Base Pipeline for Trajectory Counting

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CATTS

- Implemented complete pipeline for trajectory classification of motor vehicles at intersection.
- Evaluated on small provisional dataset derived from field site.
- Results provided as extra deliverable for Y1



	Approaching from East	Approaching from West	Approaching from South	Approaching from North
Ground Truth	145	156	58	59
FRCNN	148	133	79	74
FRCNN error	+2.1%	-14.7%	+36.2%	+25.4%
SSD512	141	70	82	81
SSD512 error	-2.8%	-55.1%	+41.4%	+37.3%

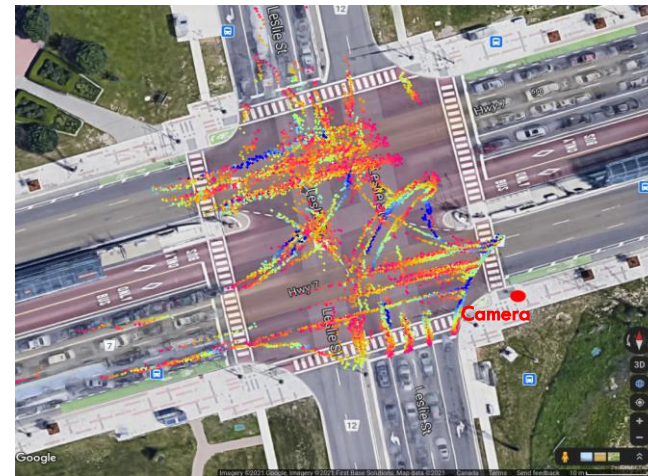
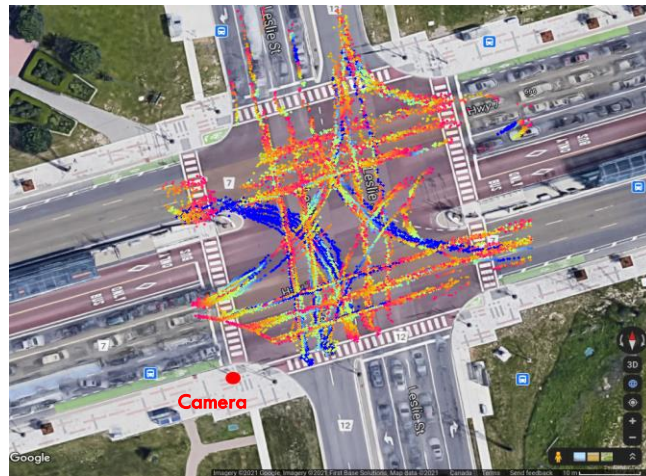
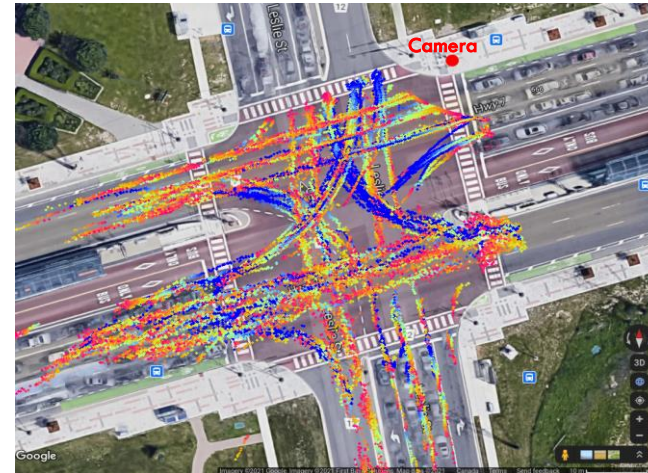
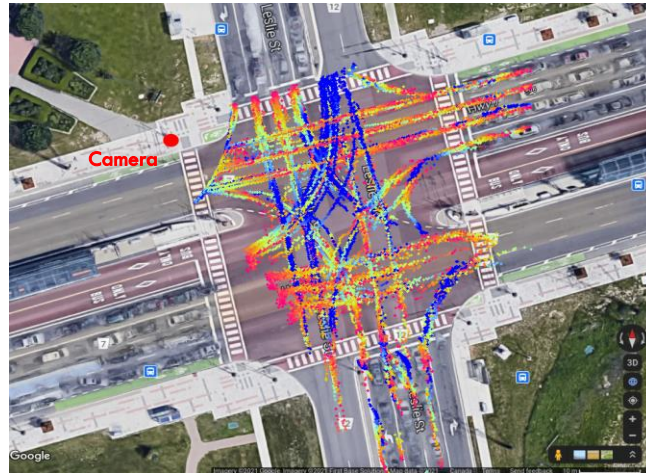


Poorna Rao

J. Elder



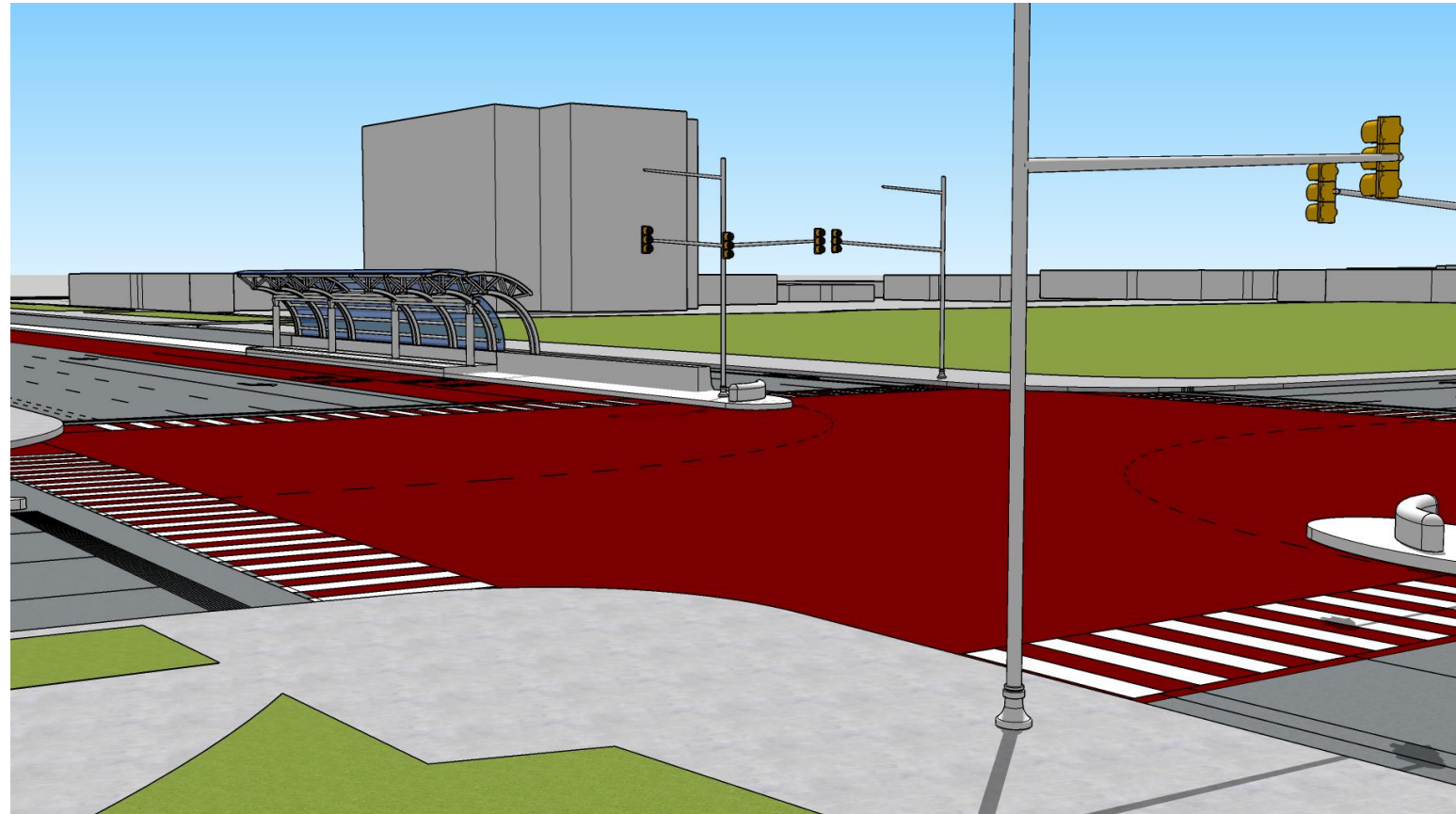
# Tracks from 4 Cameras





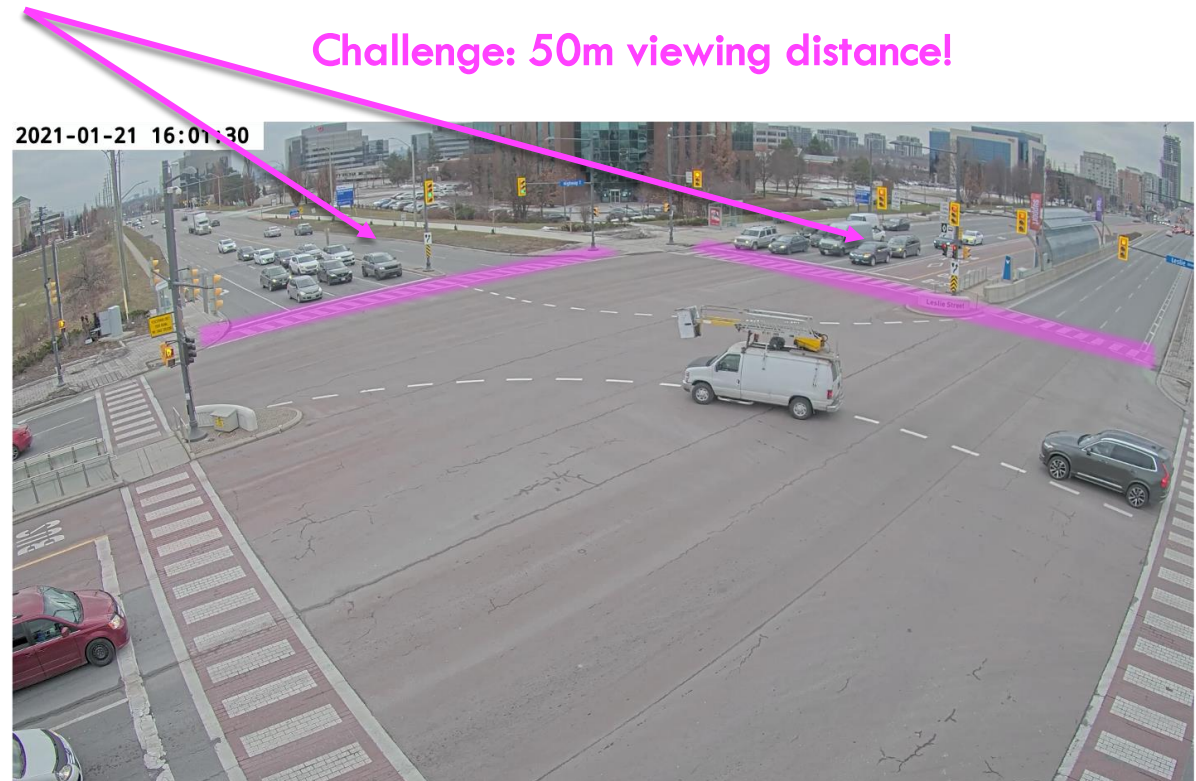
# Digital Intersection

- ✓ 1. Accurate road shape
  - ✓ 2. Traffic islands
  - ✓ 3. Textures: road markings
  - ✓ 4. Grass
  - ✓ 5. Building massing
  - ✓ 6. Traffic lights
  - 7. Trees
  - 8. Signs
  - 9. Bus stops
  - 10. Lamps
  - 11. Electricity poles
- LOD 1
- LOD 2
- LOD 3



# Attending to Vulnerable Road Users

- Goal: Detailed detection and fine-grained assessment of vulnerable road users over two crosswalks with a single camera system
- Example conditions of interest:
  - ▣ Wheelchair
  - ▣ Cane
  - ▣ Other mobility challenge
  - ▣ Elderly
  - ▣ Children
  - ▣ Stroller
  - ▣ Distracted (e.g., texting)





# Attentive Sensor: Indoor Prototype

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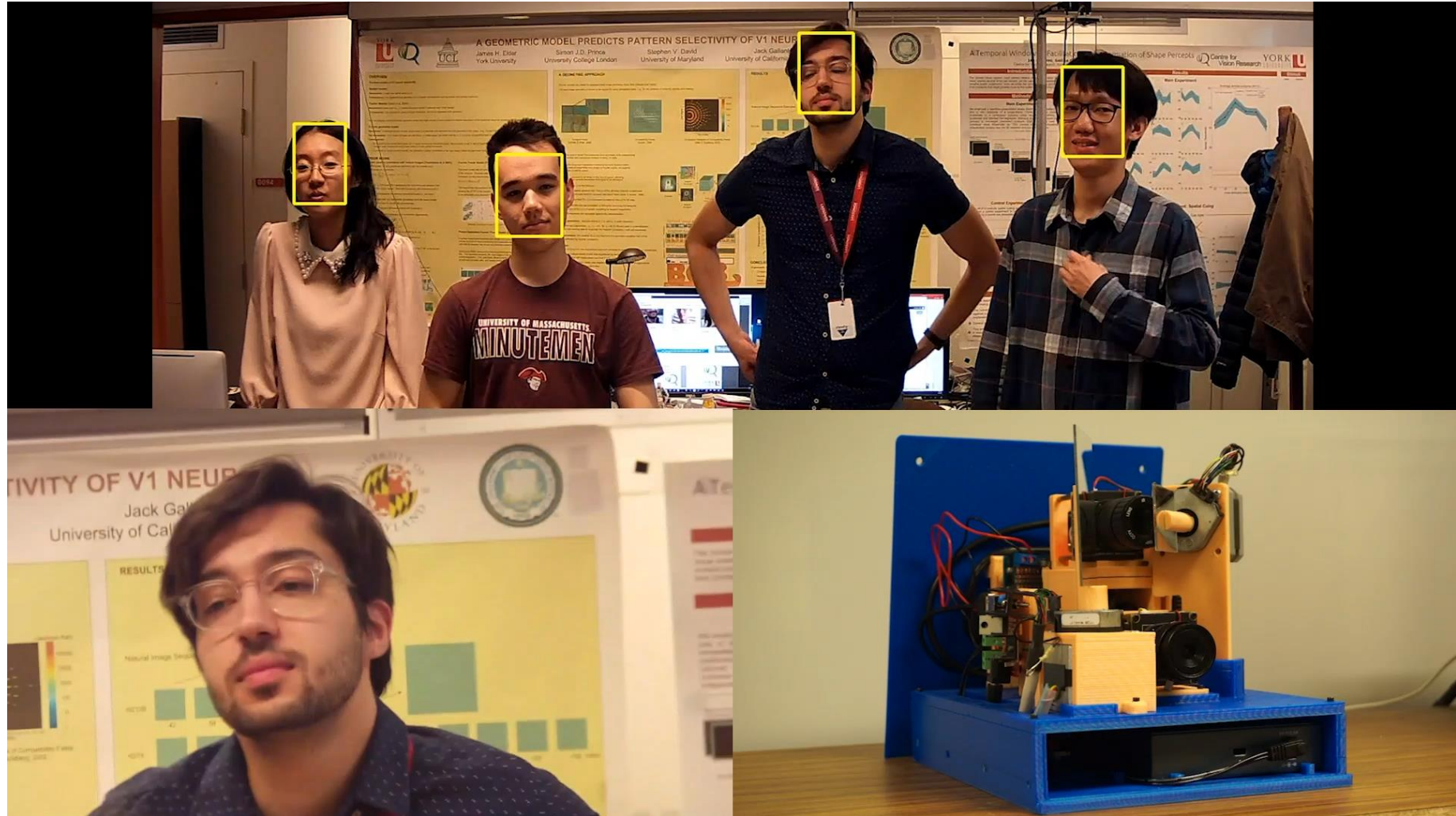
Jun 2021

## Pre-Attentive:

- 1920 x 1080p
- 75 deg FOV
- Using MTCNN face detector (Zhang et al 2016)

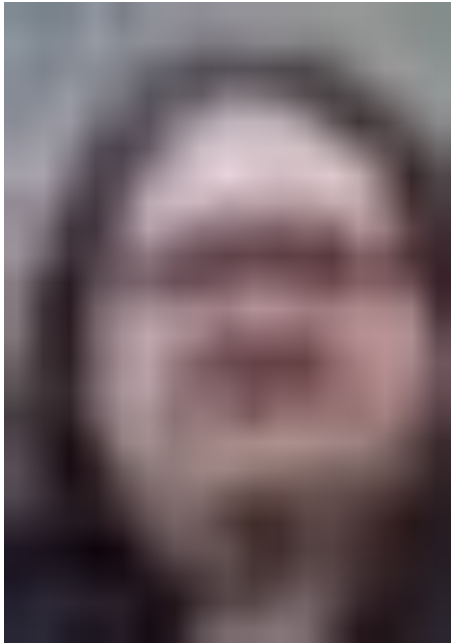
## Attentive:

- 1920 x 1200p
- 8 deg FOV
- Operates at 6 fixations/sec
- Human saccadic rate: 2-3 fixations/sec





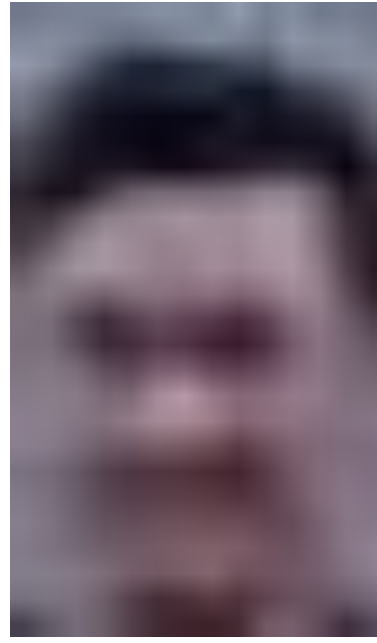
# Attentive Sensing Extends Visual Range



Pre-Attentive



Attentive



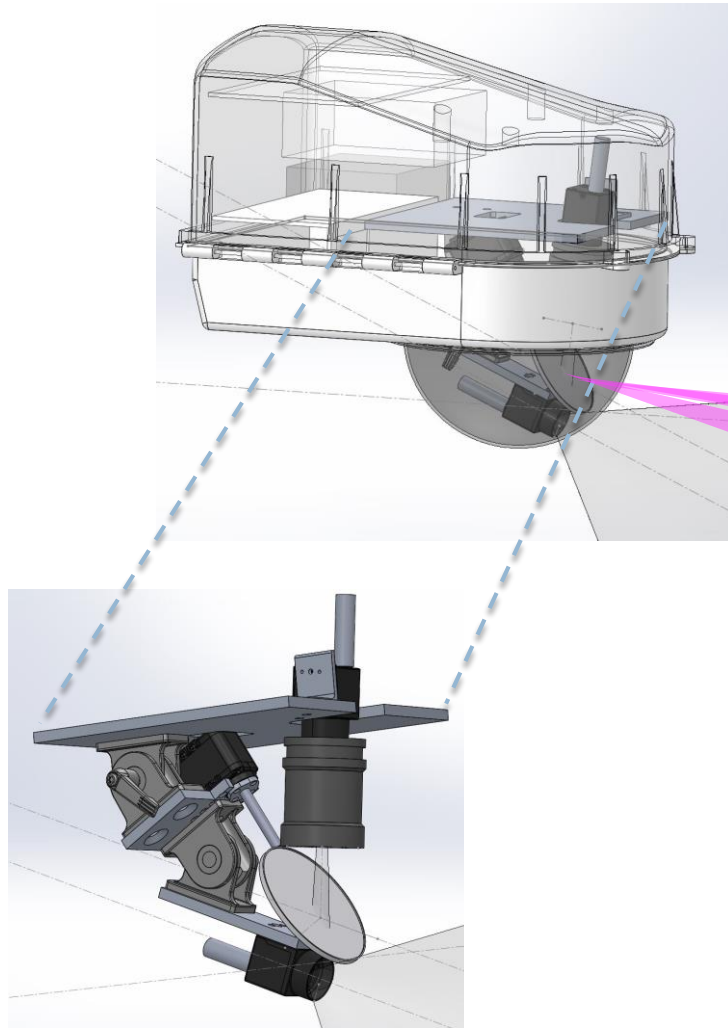
Pre-Attentive



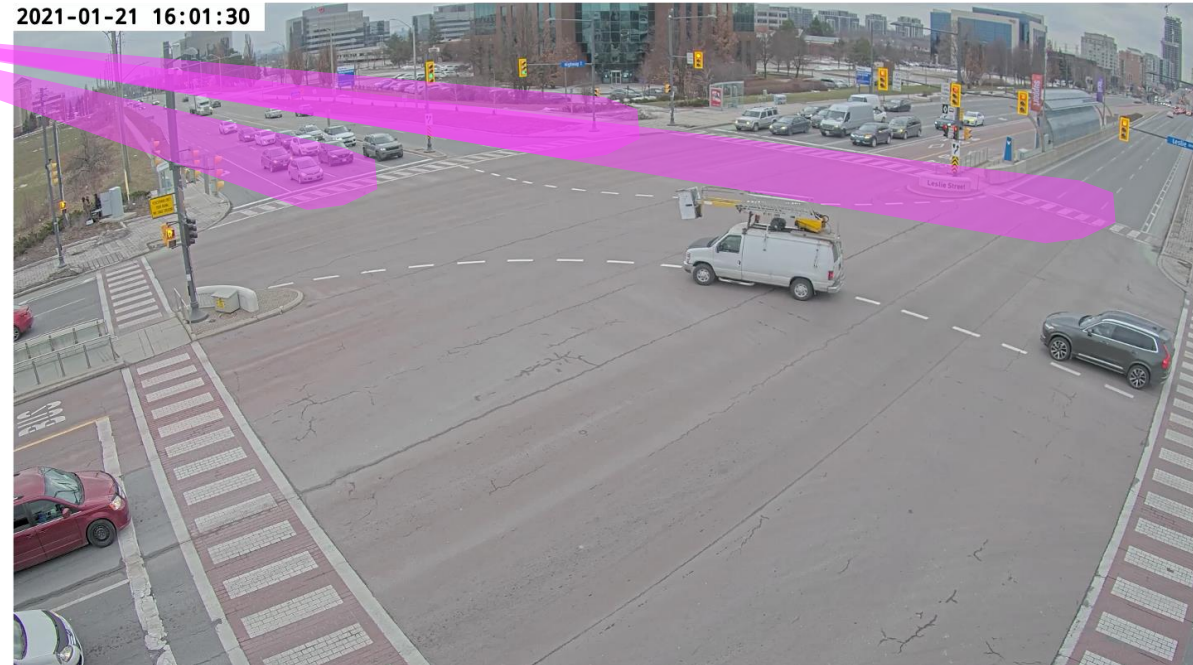
Attentive

Range: 27 metres

# Attentive Sensor for Pedestrian Analytics: Design



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Kartikeya Bhargava

# Next Steps

- Generalization to busy downtown intersection with dense pedestrian traffic.

