

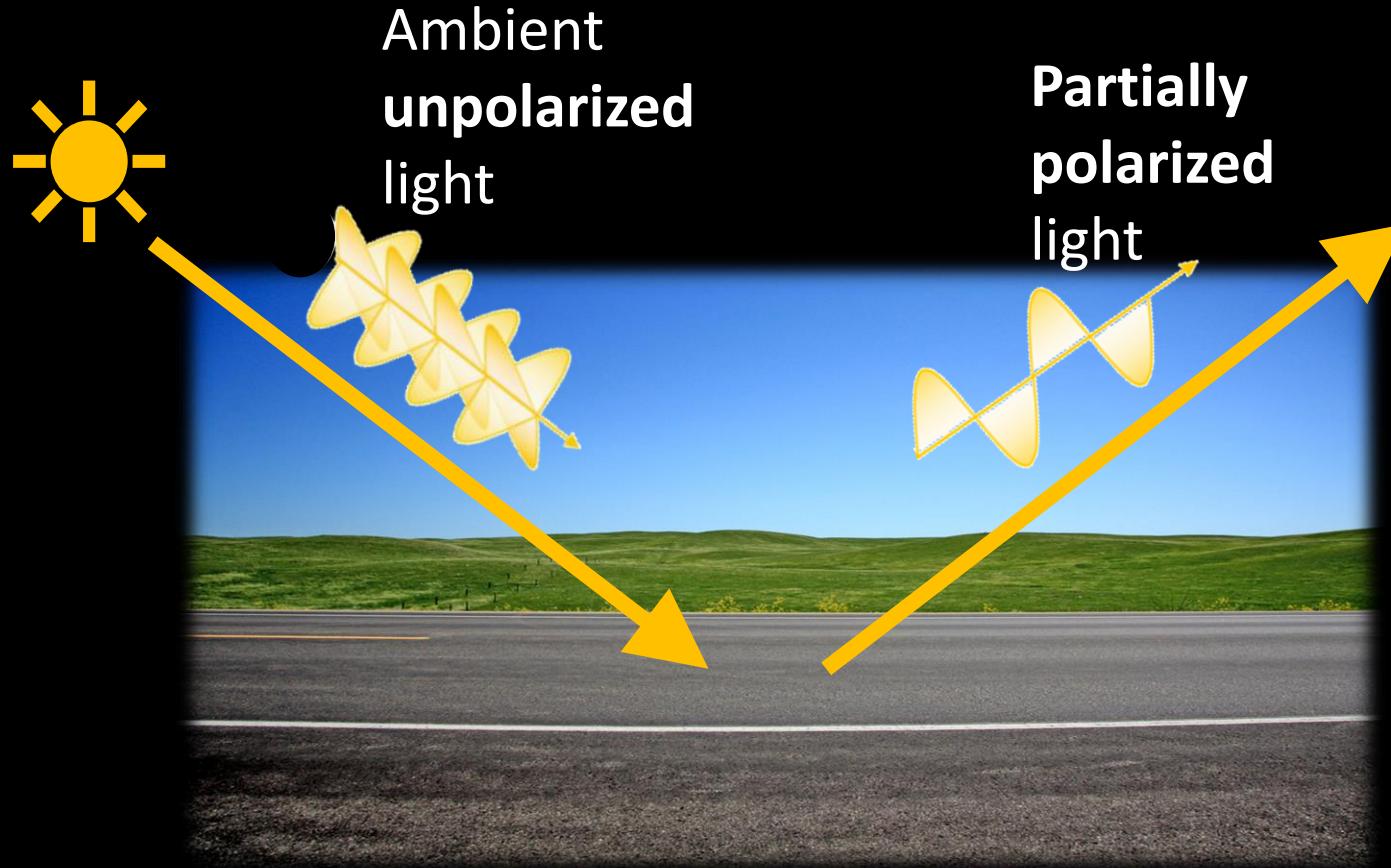


IMVC 2024

# Polarimetric Imaging for Perception

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General Motors

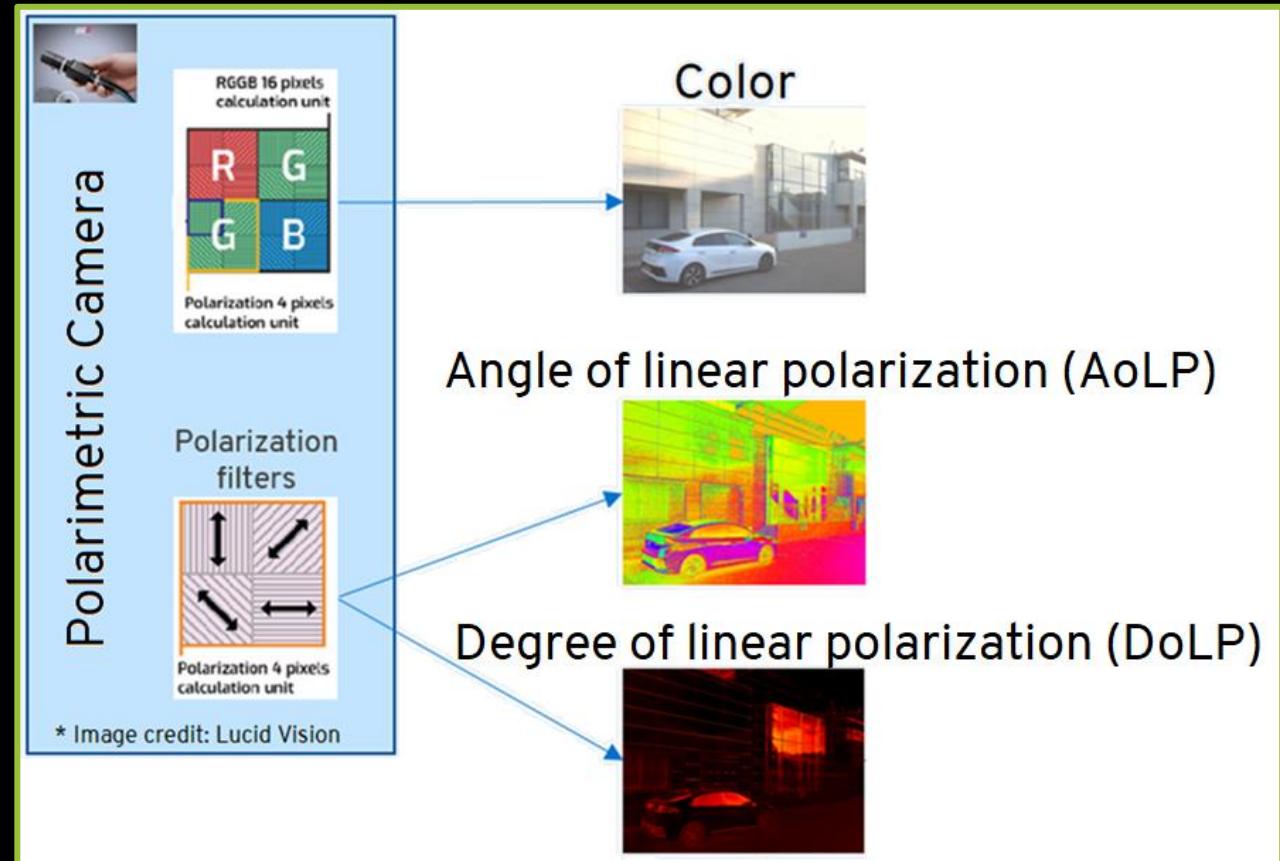
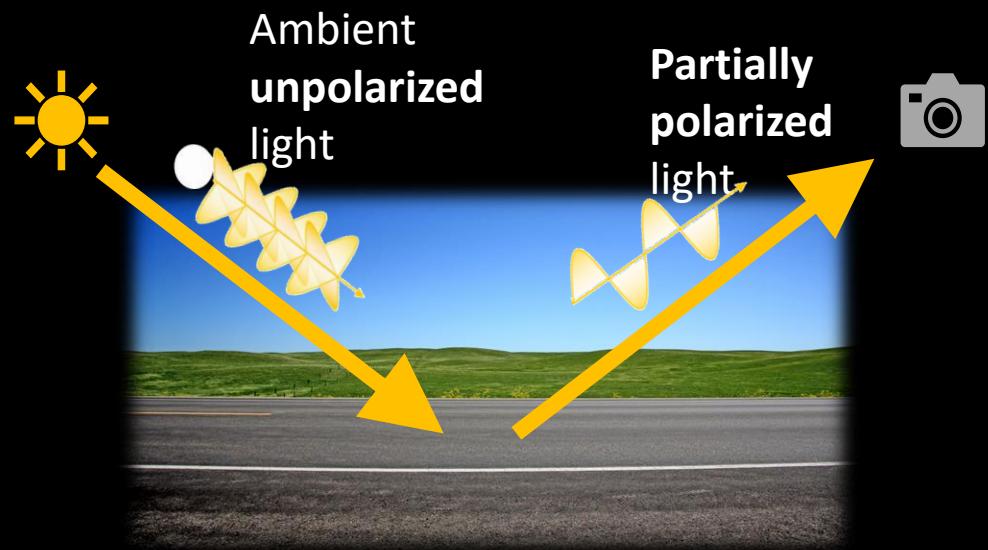
# Polarization of Light



## Polarization State

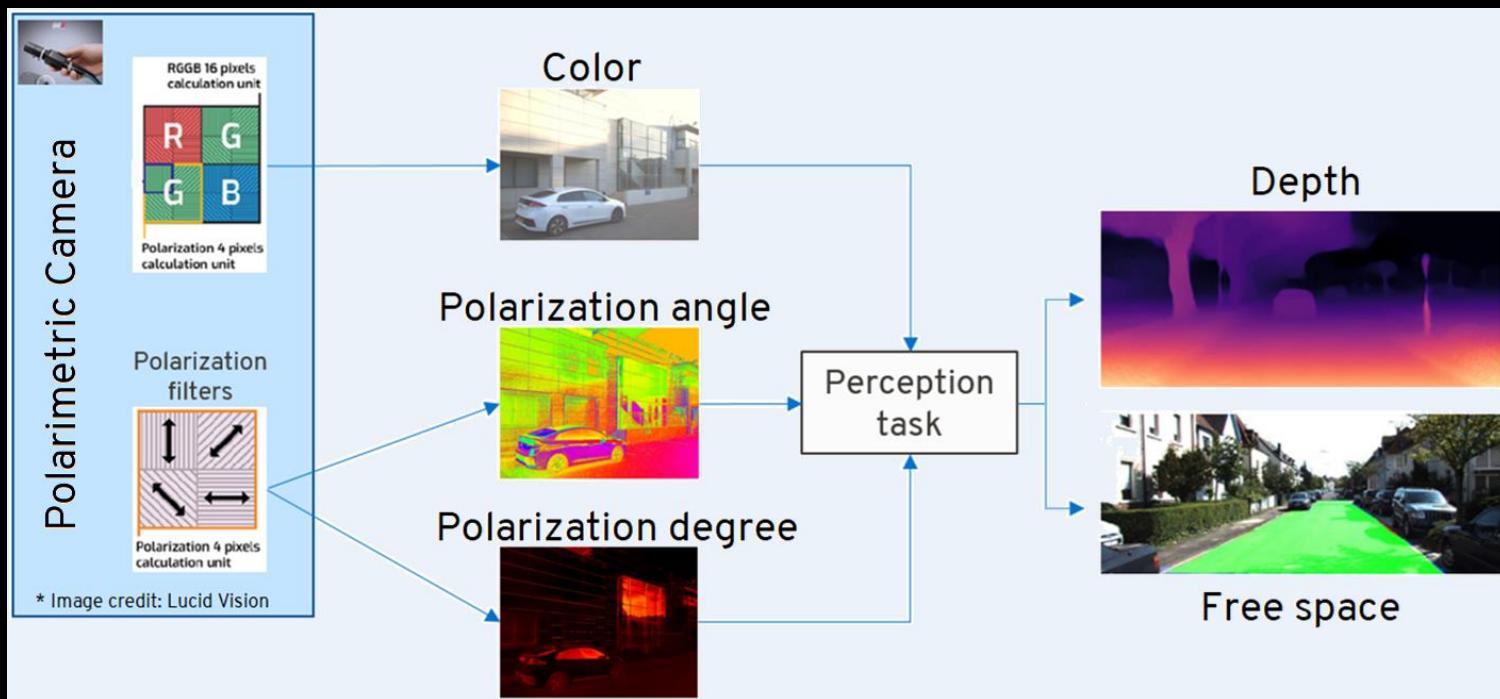
- Surface orientation
- Material properties
- Viewing direction

# Polarization of Light



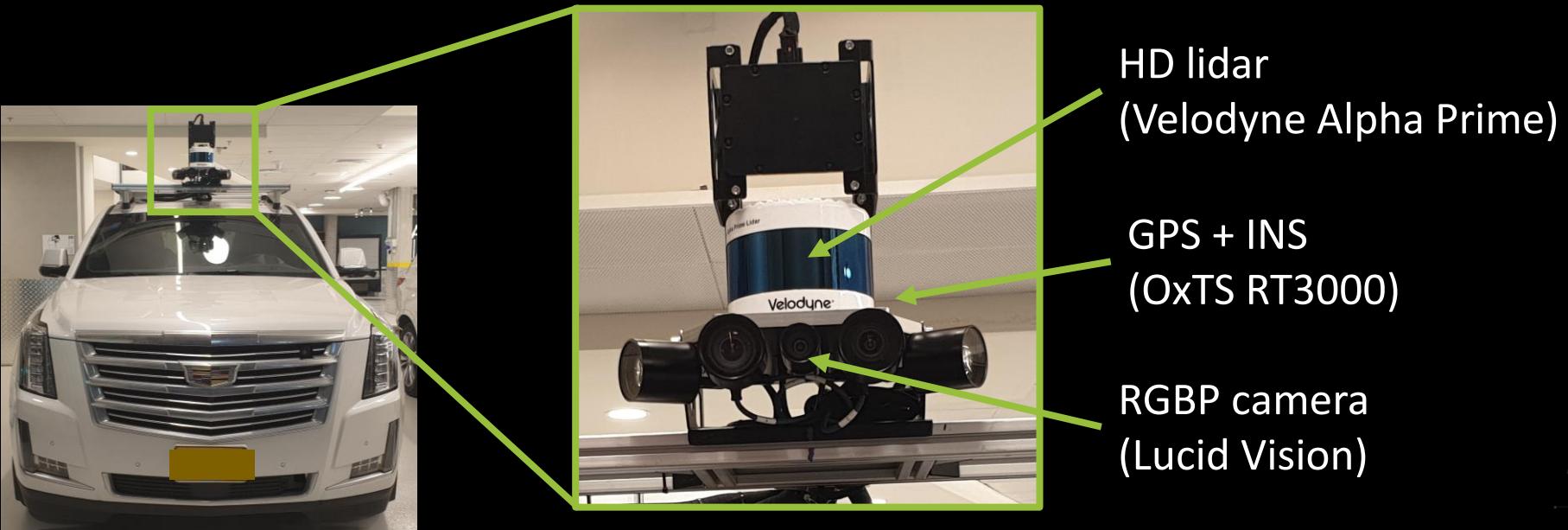
# Polarimetric Imaging for Perception

**Goal:** Explore the potential improvement of perception tasks, when using an RGBP camera.



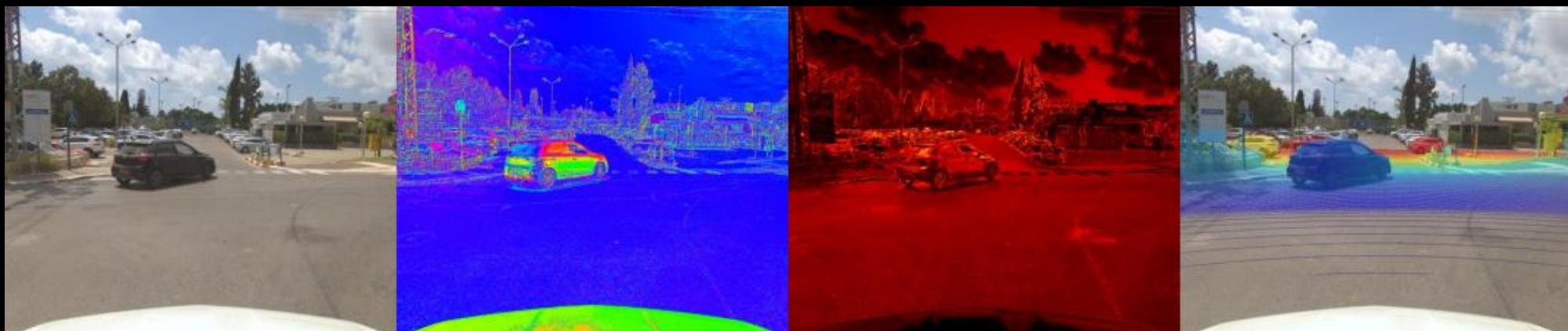
# RGBP Dataset – Data Collection

- Setup built in-house.
- Full calibration: camera – lidar – [GPS + inertial system].



# RGBP Dataset

- 12K samples: RGB, AoLP, DoLP, lidar and pose.
- 8K images with free space annotated.
- Noon-time to avoid heading dependency.



RGB

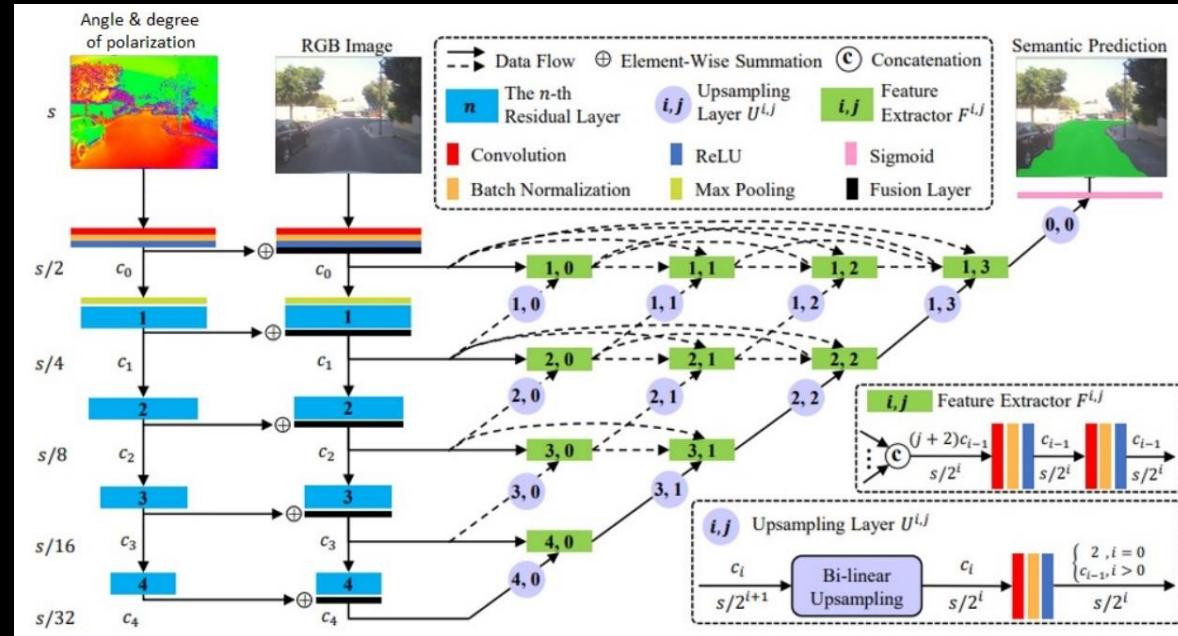
AoLP

DoLP

Lidar

# Free Space Estimation

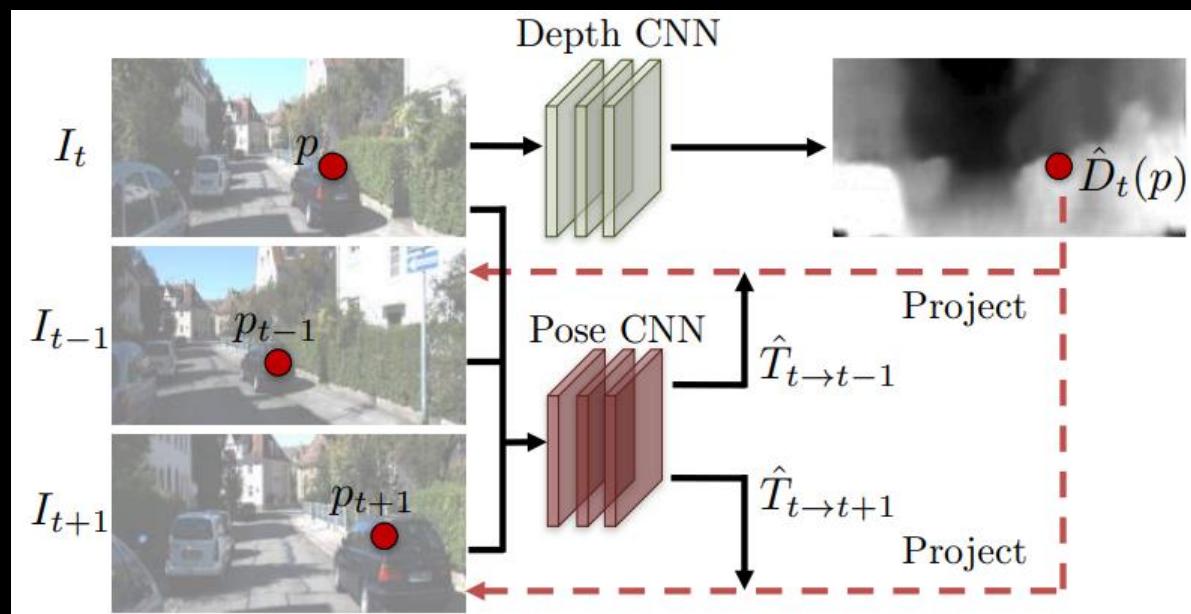
- Adapted SNE-RoadSeg network [1]. Removed normals estimation.
- Polarization features added:  $[\sin(2 \cdot AoLP), \cos(2 \cdot AoLP), 2 \cdot DoLP - 1]$ .



[1] Fan, Rui, et al. "Sne-roadseg: Incorporating surface normal information into semantic segmentation for accurate freespace detection." ECCV, 2020.

# Depth Estimation

- Adapted self-supervised monodepth v2 [2].
- Polarization features added:  $[\sin(2 \cdot AoLP), \cos(2 \cdot AoLP), 2 \cdot DoLP - 1]$ .



[2] Godard, Clément, et al. "Digging into self-supervised monocular depth estimation." ICCV, 2019.

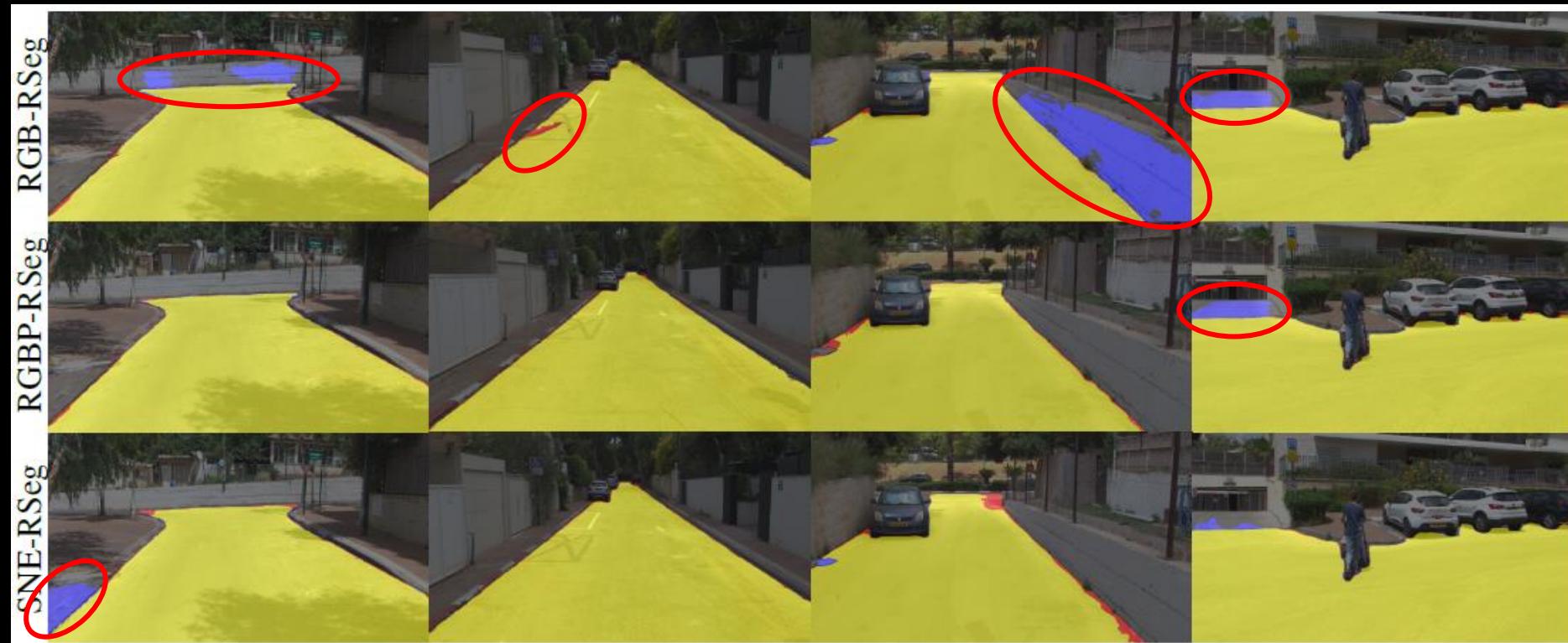
# Results – Free Space

RGBP outperformed RGB.

| Method              | Accuracy     | Precision    | Recall       | $F_{max}$    | IoU          | AP           |
|---------------------|--------------|--------------|--------------|--------------|--------------|--------------|
| RGB-RoadSeg         | 0.979        | 0.949        | 0.968        | 0.953        | 0.902        | 0.974        |
| P-RoadSeg           | 0.865        | 0.845        | 0.534        | 0.641        | 0.467        | 0.634        |
| <b>RGBP-RoadSeg</b> | <b>0.986</b> | <b>0.966</b> | <b>0.972</b> | <b>0.968</b> | <b>0.939</b> | <b>0.994</b> |
| SNE-RoadSeg         | 0.985        | 0.967        | 0.967        | 0.965        | 0.934        | 0.993        |

# Results – Free Space

RGBP outperformed RGB.



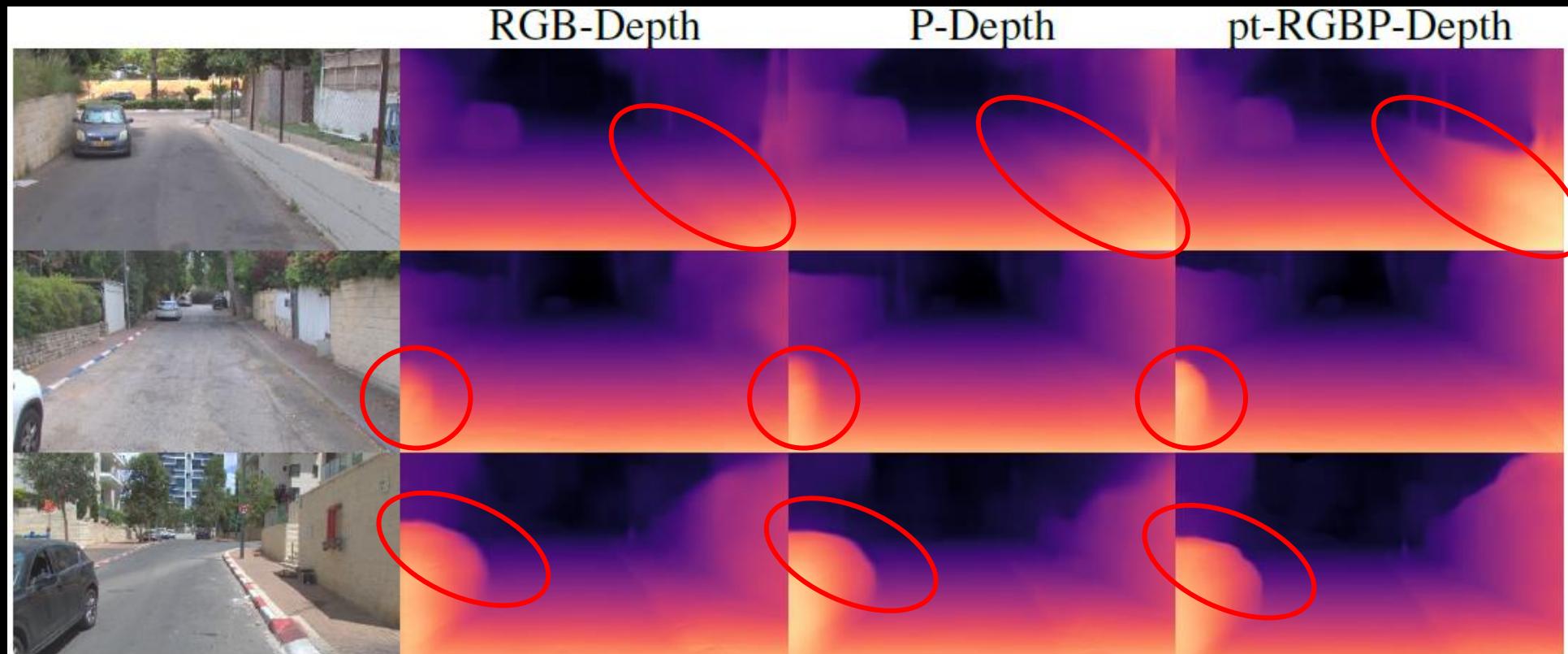
# Results – Depth Estimation

RGBP and RGBP pre-trained on RGB (pt-RGBP) outperformed RGB.

| Method        | Error metric ↓ |              |              |              | Accuracy metric ↑ |                   |                   |
|---------------|----------------|--------------|--------------|--------------|-------------------|-------------------|-------------------|
|               | Abs Rel        | Sq Rel       | RMSE         | RMSE Log     | $\delta < 1.25$   | $\delta < 1.25^2$ | $\delta < 1.25^3$ |
| RGB-Depth     | 0.094          | 0.838        | 6.389        | 0.166        | 0.904             | 0.964             | 0.984             |
| P-Depth       | 0.091          | 0.811        | 6.325        | 0.164        | 0.907             | 0.966             | 0.985             |
| RGBP-Depth    | 0.089          | 0.770        | 6.172        | 0.161        | 0.911             | <b>0.968</b>      | <b>0.986</b>      |
| pt-RGBP-Depth | <b>0.086</b>   | <b>0.767</b> | <b>6.109</b> | <b>0.158</b> | <b>0.915</b>      | <b>0.968</b>      | 0.985             |

# Results – Depth Estimation

RGBP and RGBP pre-trained on RGB (pt-RGBP) outperformed RGB.

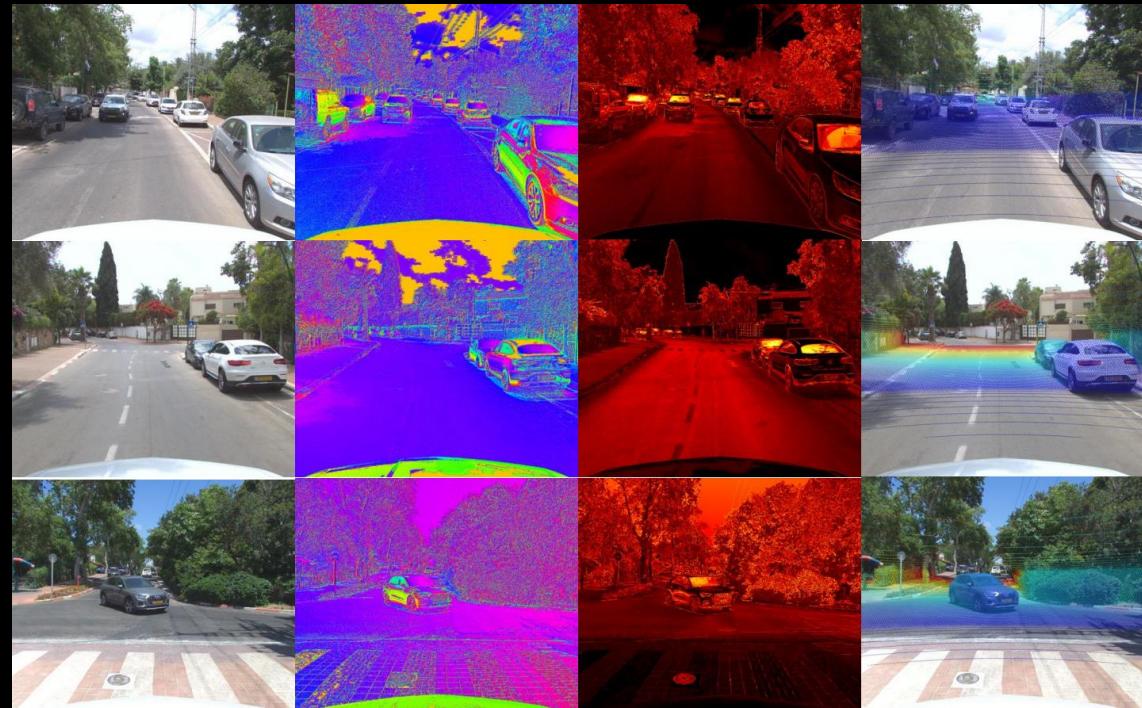


# Conclusions

- New polarimetric dataset released for perception tasks.
- Polarization data improved free space and depth estimation systems.
- Minimal architectural changes needed for quantitative improvement.

# Thanks!

Download the dataset!



RGB

AoLP

DoLP

Lidar

[https://michaelbaltaxe.github.io/polarimetric\\_perception](https://michaelbaltaxe.github.io/polarimetric_perception)