UNIVERSITÉ Clermont Auvergne

> **Ecole doctorale Sciences Pour l'Ingénieur**

2D-3D Fusion for Road Object Detection and Tracking on Autonomous Vehicles





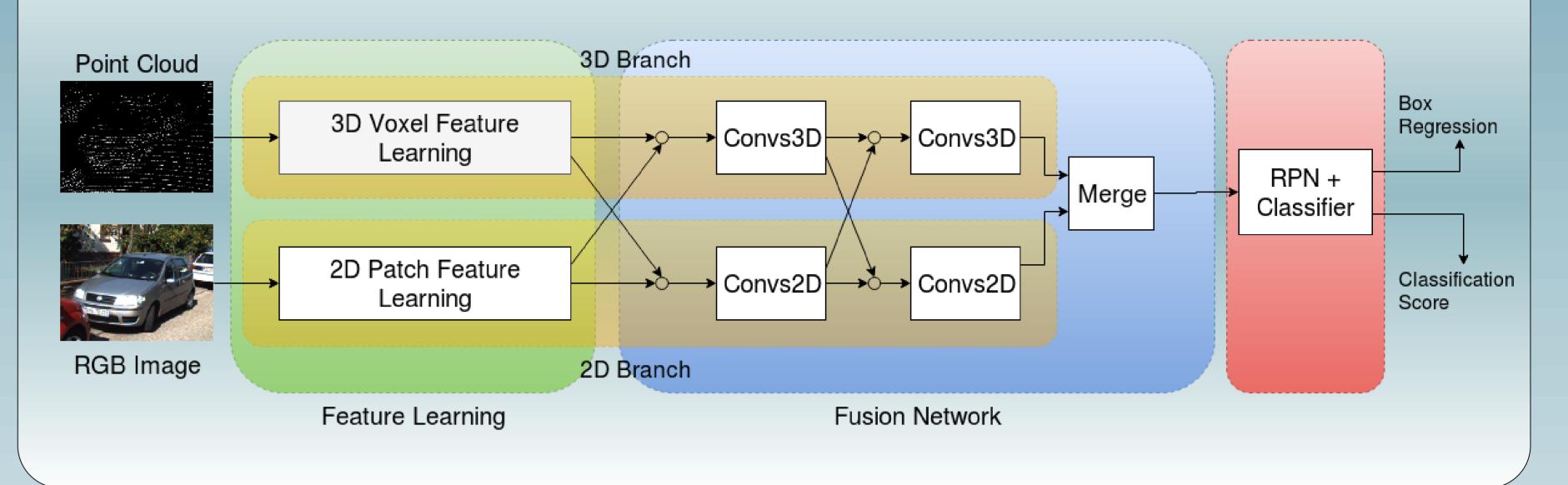
Ruddy THEODOSE¹, Dieumet DENIS¹, Christophe **BLANC², Paul CHECCHIN²** ¹Sherpa Engineering, ²Institut Pascal



Goals

Explore merging methods between point clouds from automotive LiDAR with video camera stream and artificial neural networks Develop a **obstacle detection** system using **both** modalities **Track** obstacles from detections and predict their trajectories

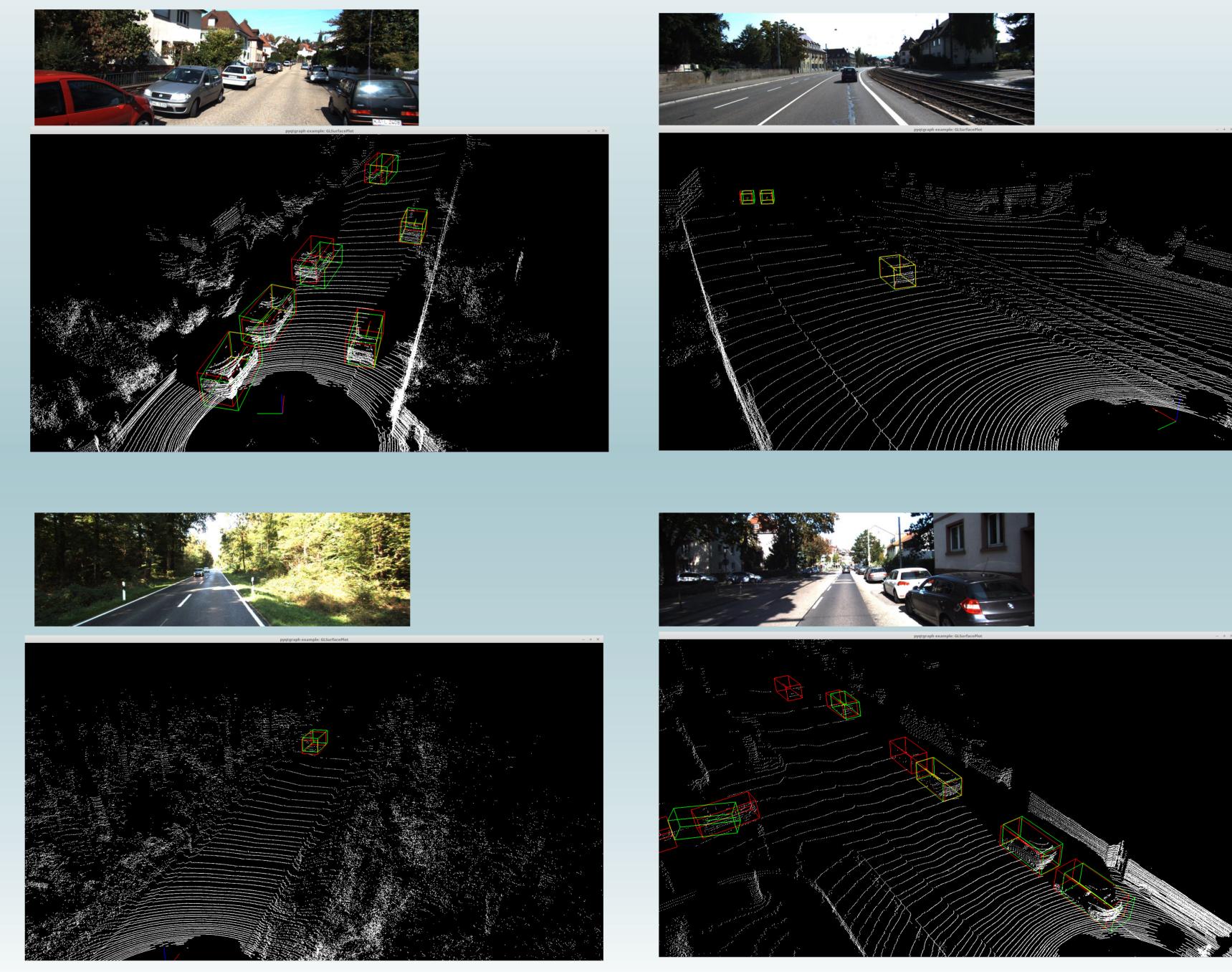




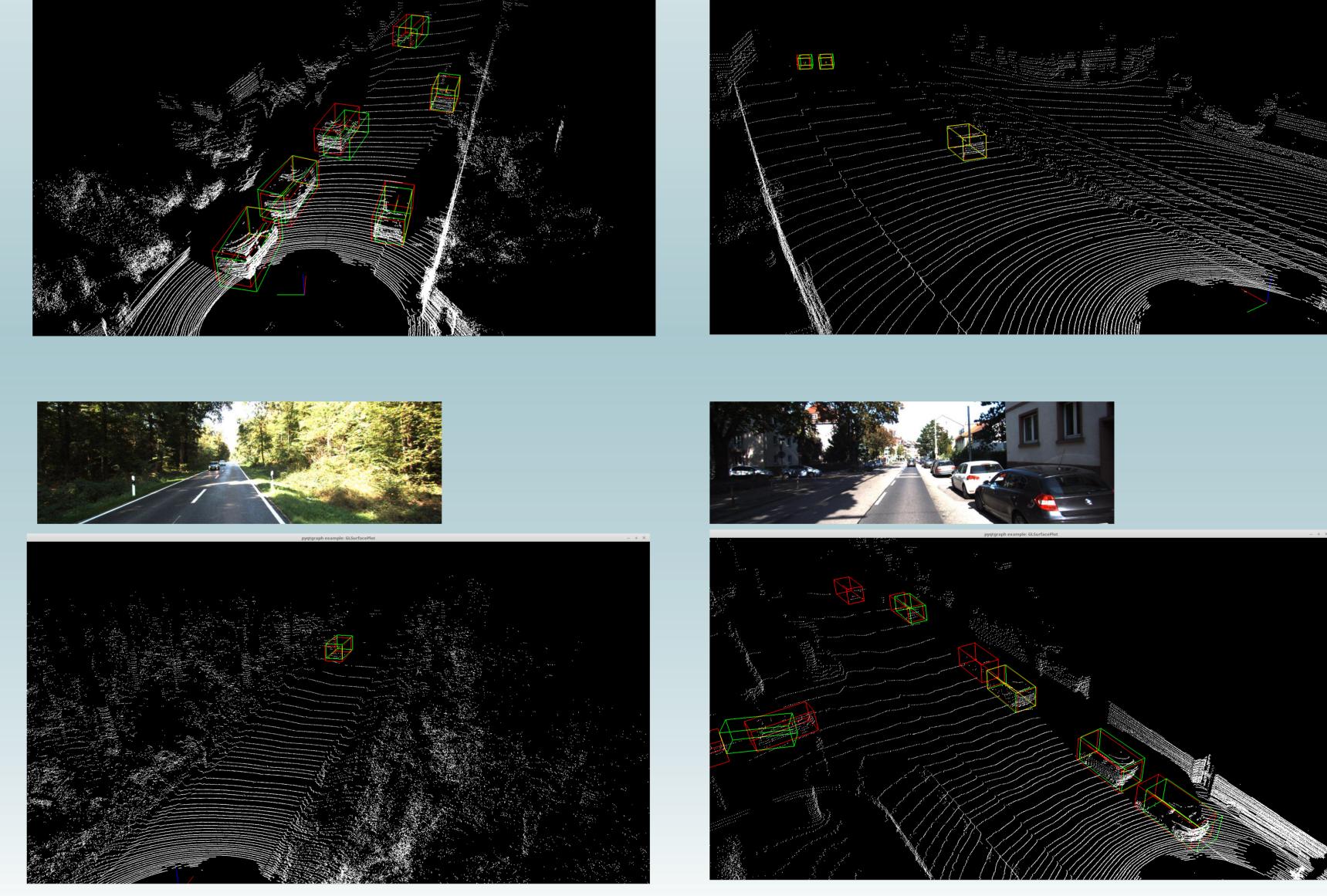
Detection System

- Image patches and corresponding 3D cells learn their own representation independently from the other patches/cells
- Spherical coordinates for Voxel Feature learning
- Voxels features augmented with probabilities inspired by occupancy grid

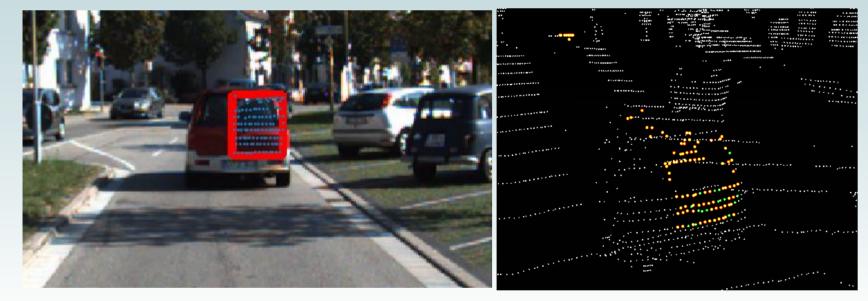
Experiments : Detection





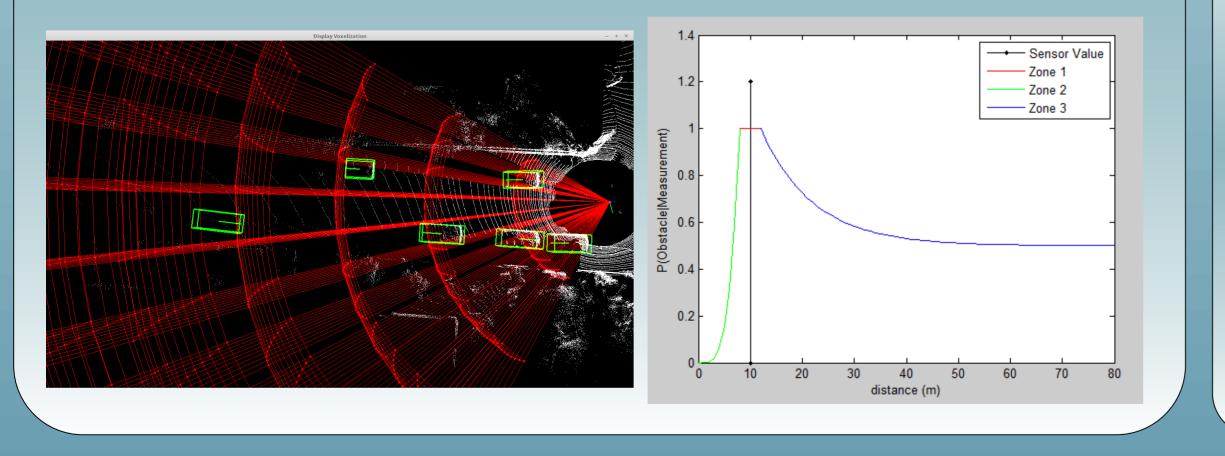


- approach
- Diffusion of each branch information to the other branch



Cell Probabilities

- For each patch, the related space region is divided into 3 zones depending on the sensor values and the distance
- Region 1 : Obstacle probably present
- Region 2 : Obstacle probably missing
- Region 3 : Missing information



Future Work

- Improve accuracy, execution time and stability on sensor loss
- Adaptation to different types of LiDAR (resolution, FOV...) Data augmentation : Information Densification (Depth) or Prediction (Surface Normals, Ego Motion Pose...) Obstacle tracking and trajectory estimation

Bibliography

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- Ku, J., Mozifian, M., Lee, J., Harakeh, A., & Waslander, S. (2017). Joint 3D Proposal Generation and Object Detection from View Aggregation.
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