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Benefits and Problems in Digital Map for Autonomous Driving: From Our Research Experiences

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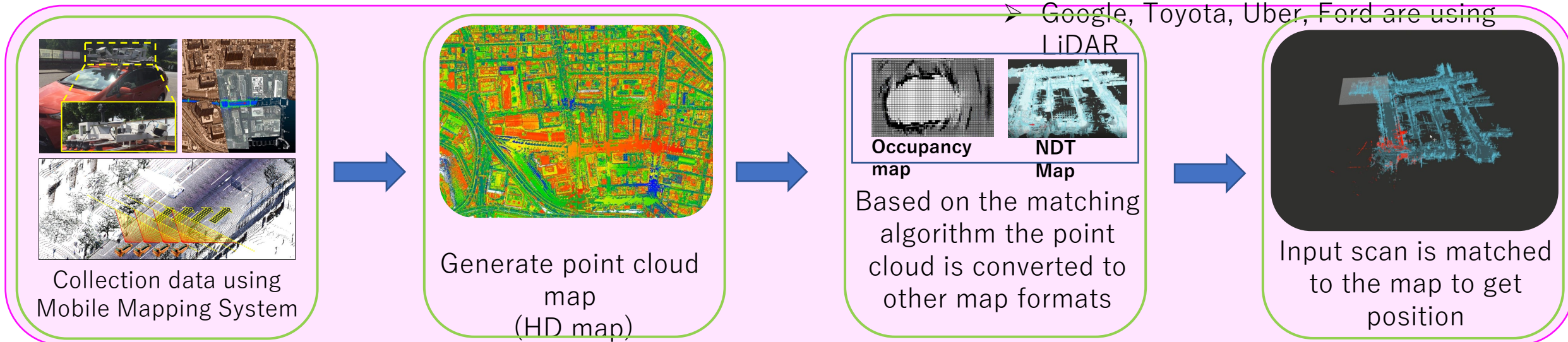
LiDAR-based vehicle localization

- LiDAR: One of the well-known sensor for localization
- LiDAR-based self-localization methods
 - SLAM (Simultaneous localization and mapping)
 - Map-based methods



3D Light Detection and Ranging (LiDAR) sensor

- 360 field of view
- Measuring error is less than 2cm in 100m
- Night condition without illumination
- Google, Toyota, Uber, Ford are using LiDAR



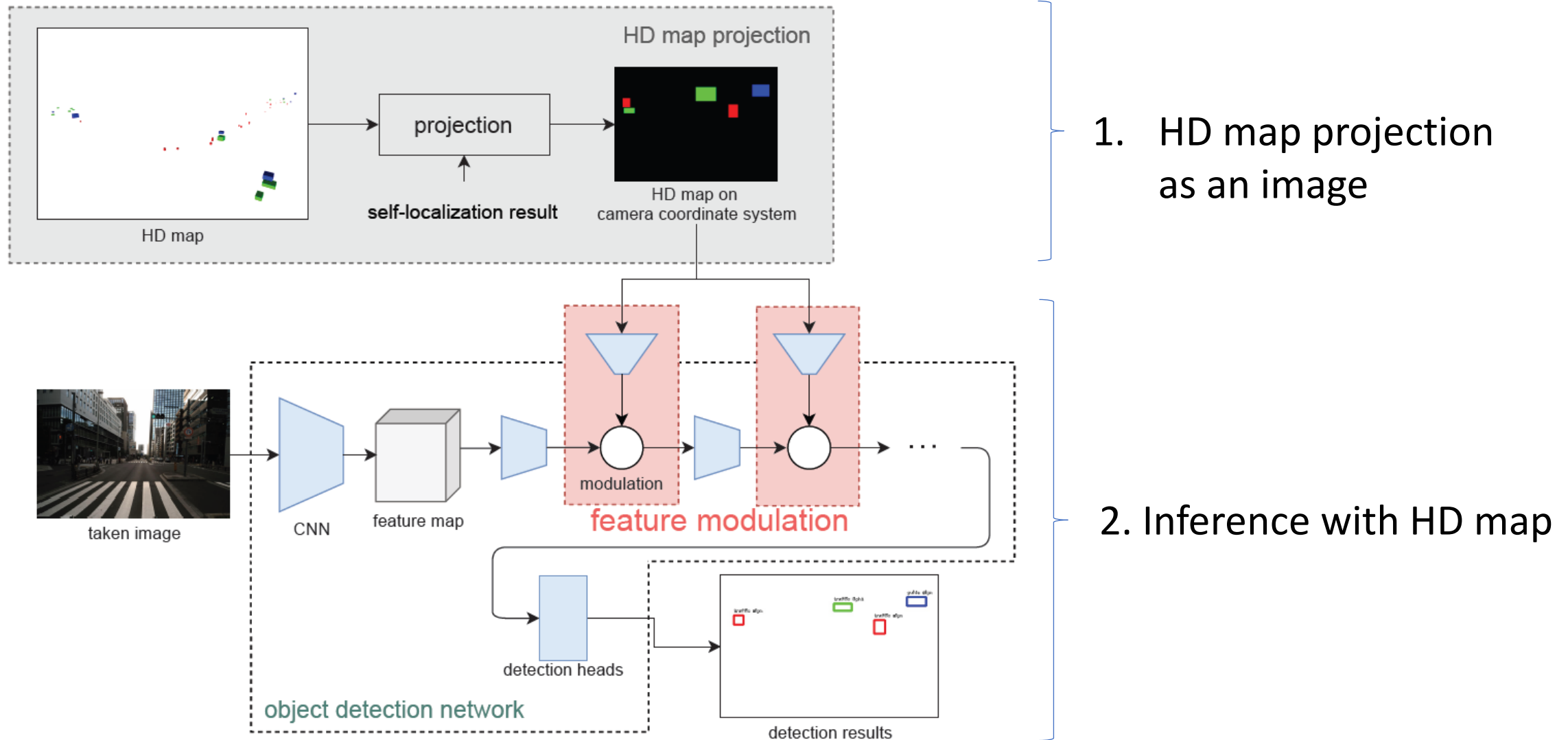
Benefits of Digital Map in Autonomous Vehicle Application

- Self-Localization referencing to point-cloud data.
- Static rules are annotated in the High-Definition map.
 - Buildings, road sided facilities, lanes, road markings, ...
 - lane semantics, speed limitation, traffic sign, stop line, pedestrian crossing, direction guidance, ...
 - Those information are useful for motion planning.
- Dynamic information are attached on the HD map, and delivered through digital network to the autonomous vehicles.
 - Macroscopic events of road construction, traffic regulation, damaged road, ...
 - Mesoscopic events as congestion, accident, ...
 - Those information are useful for travel planning.
- **Object detection aided by HD map**
 - HD map might comprise information of **traffic signals, traffic signs, variable message signs: 3D positions and bounding boxes.**
 - Improve the detection accuracy of those facilities in **adverse condition as rain, fog, night time, ...**

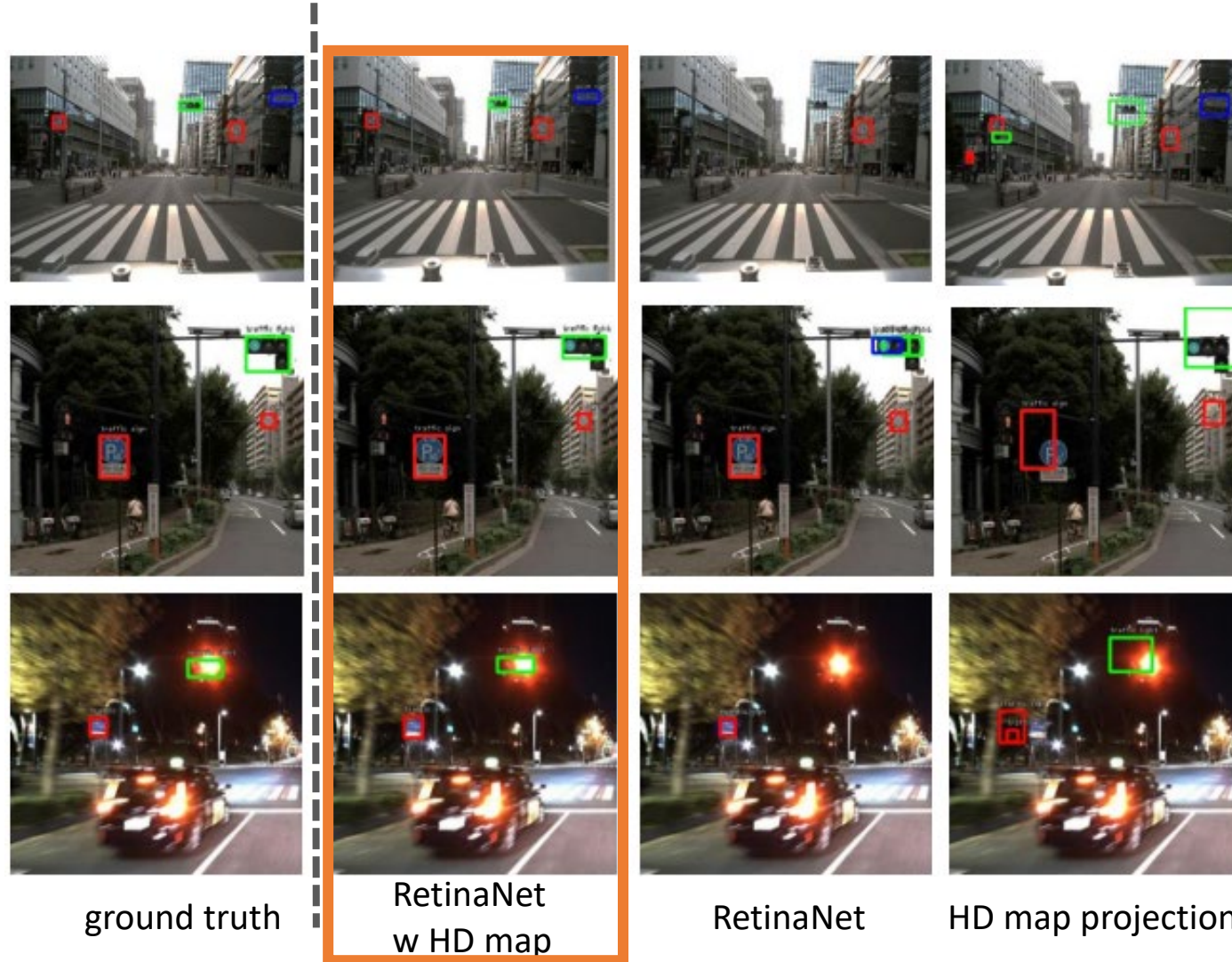
Problems of Digital Map in Autonomous Vehicle Application

- How the digital map can be updated?
 - Cloud sourcing or Tailor made?
- How the digital map should be standardized among countries, map providers, and OEMs?
 - What kind of **format** is available and suitable for the digital map?
point cloud, vector, polygon, ...
 - How the **quality of the digital map** should be defined, evaluated and secured?
 - and by whom?
- LiDAR is not reliable for localization in some scenarios.
 - **Adverse weather conditions**: heavy rain and fog
 - **Occlusion effects**: beams are interrupted by the surrounding tall vehicles
 - **Passive sensor fusion** mitigates the occlusion effects in conjunction with the digital map.

Algorithm Flow of the HD map aided Object Detection



Evaluation of Object Detection in Night Image



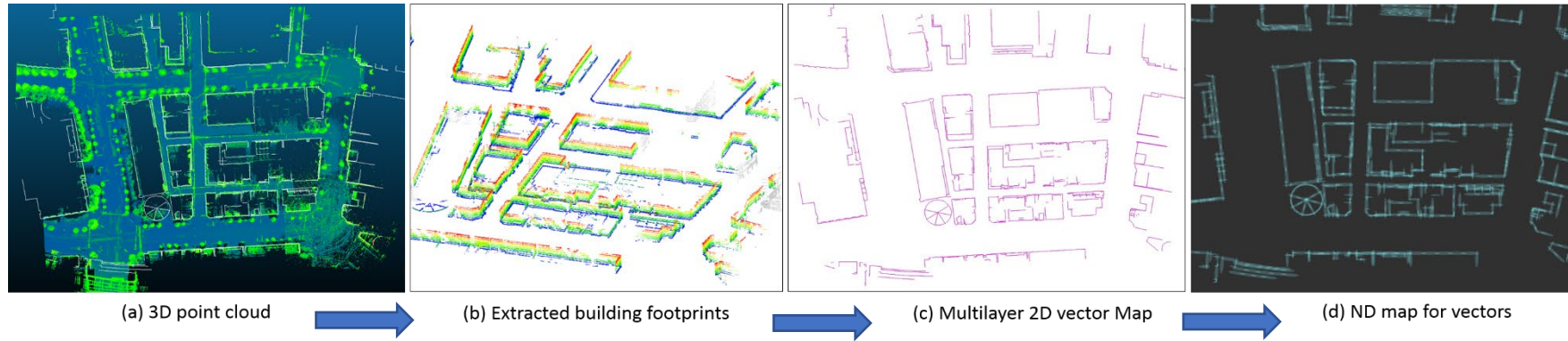
ground truth

RetinaNet
w HD map

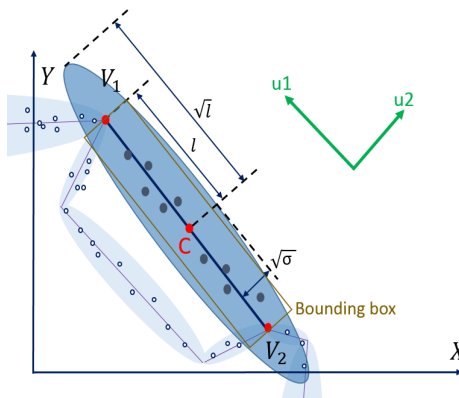
RetinaNet

HD map projection

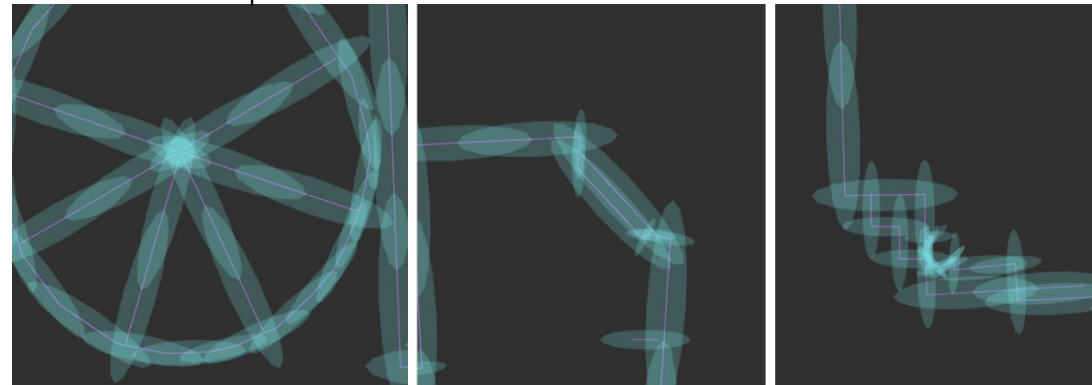
HD map comprising Vector Elements: Vector NDT



Points that made a vector form a normal distribution



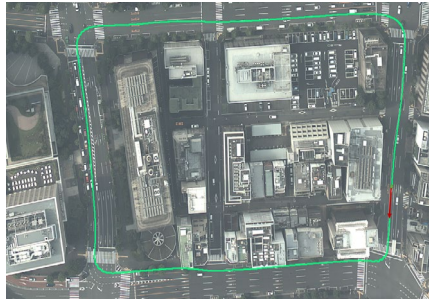
Generated Normal distribution form vector map



$$P(\vec{x}) = \frac{1}{2\pi\sqrt{|\Sigma|}} \exp\left(-\frac{(\vec{y}_k - \vec{\mu})^T \Sigma^{-1} (\vec{y}_k - \vec{\mu})}{2}\right)$$

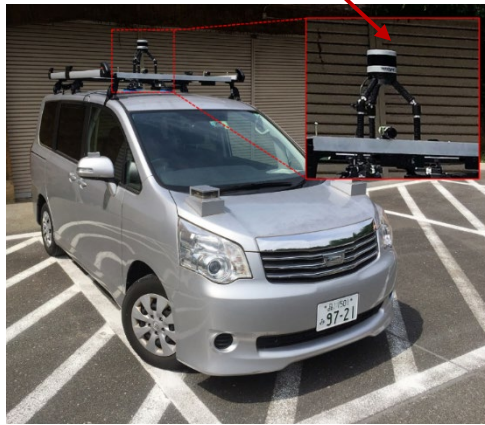
- $Y = \{\vec{y}_1, \dots, \vec{y}_n\}$ Points that made a vector segment
- $\vec{\mu}$ Mean of generated normal distribution
- Σ Covariance of generated normal distribution

Experimental results



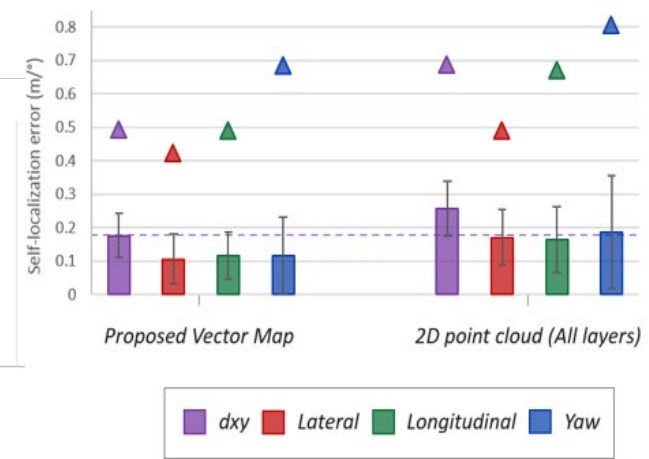
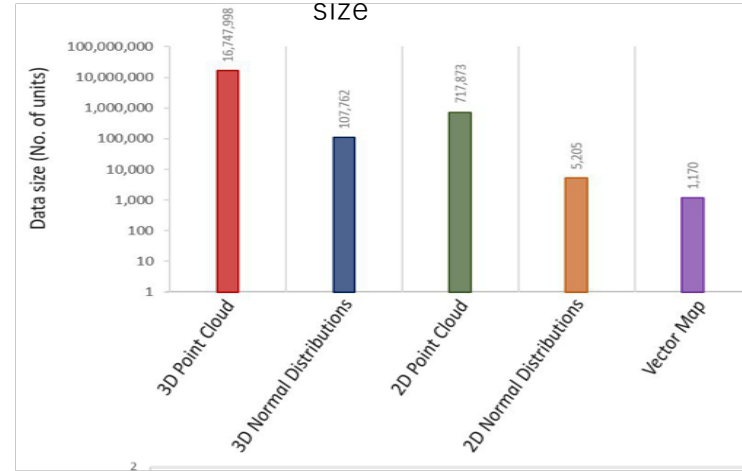
Route of experiments

Velodyne's VLP-16
(16 channel)

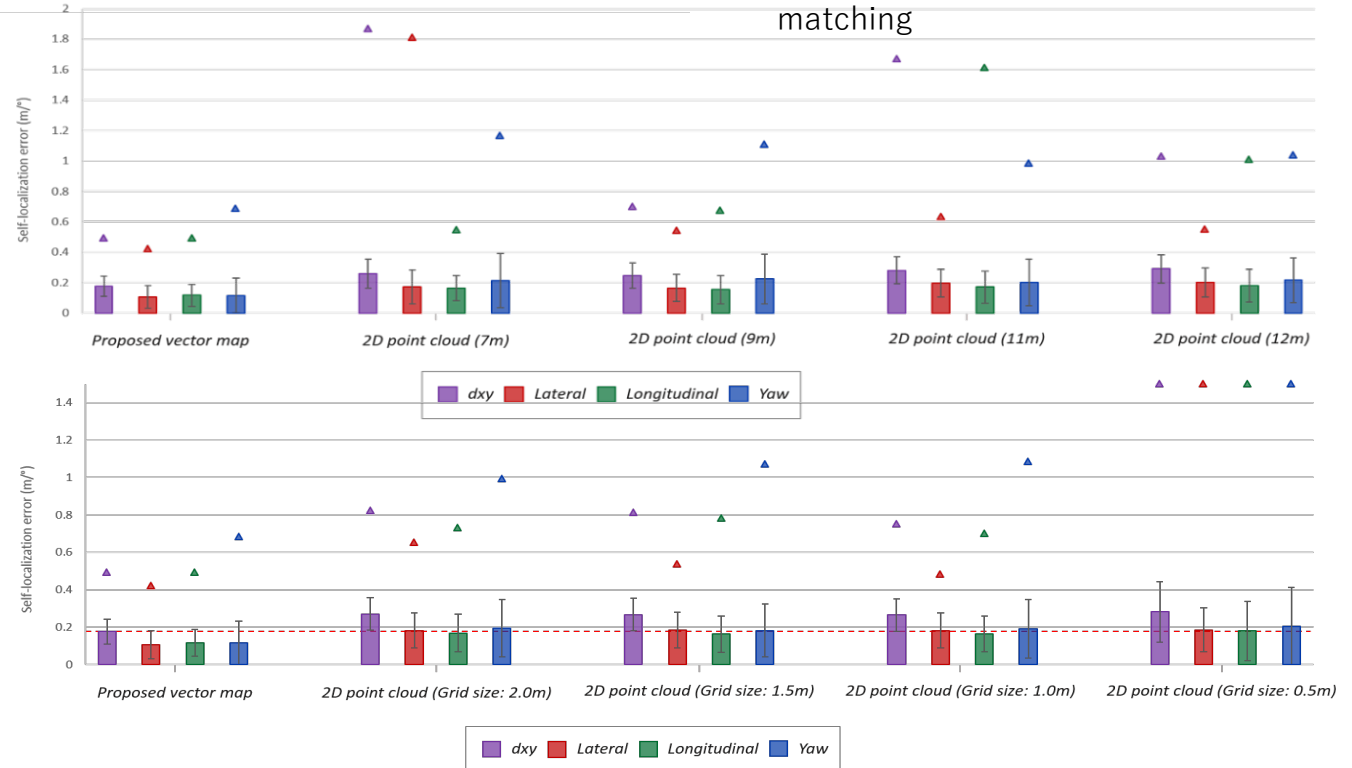


Our experimental vehicle

Evaluation of data size



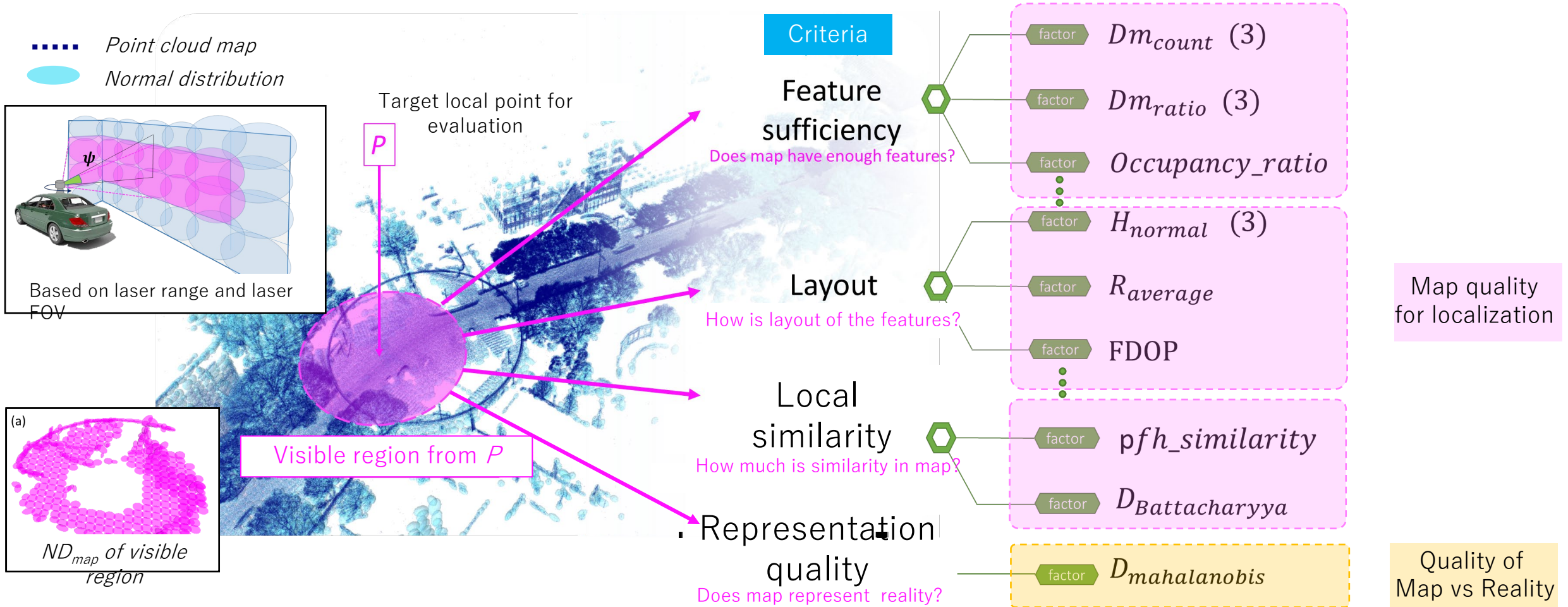
Evaluation of vector ND map-matching



Evaluation of multilayer 2D vector structure (comparison with conventional 2D methods)

Map evaluation criteria to formulate the Localization accuracy

- For each local point P on the map, a visible region is extracted
- To evaluate the map ability for localization for point P 4 criteria are introduced



Test-bed and Setups for the Experiments

Experimental Path



Experimental Path in
Shinjuku, Tokyo
1.2 km path



Point cloud map of
experimental path

Our experimental vehicle

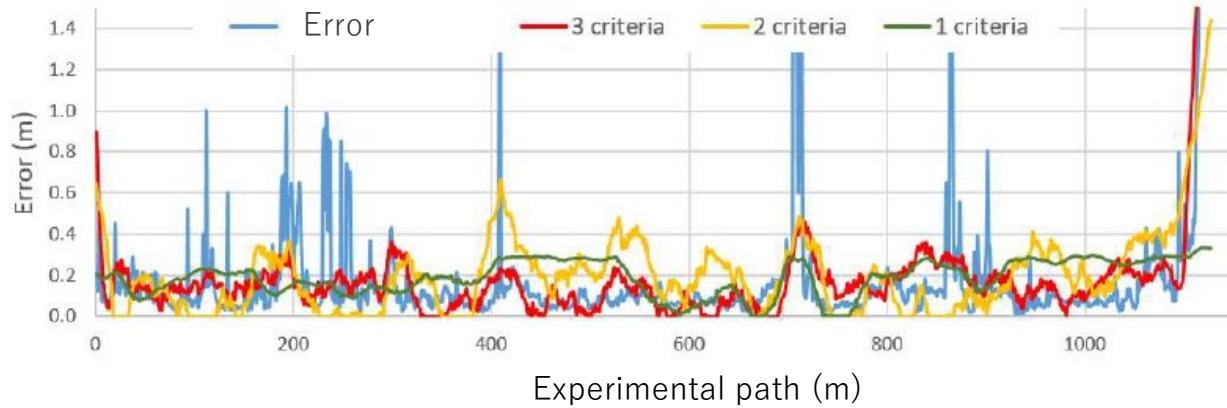


- ❖ Laser scanner range : 20m
- ❖ Frequency : 20Hz
- ❖ Driving speed : 10km/h
- ❖ (distortion is less than 7cm in each scan)

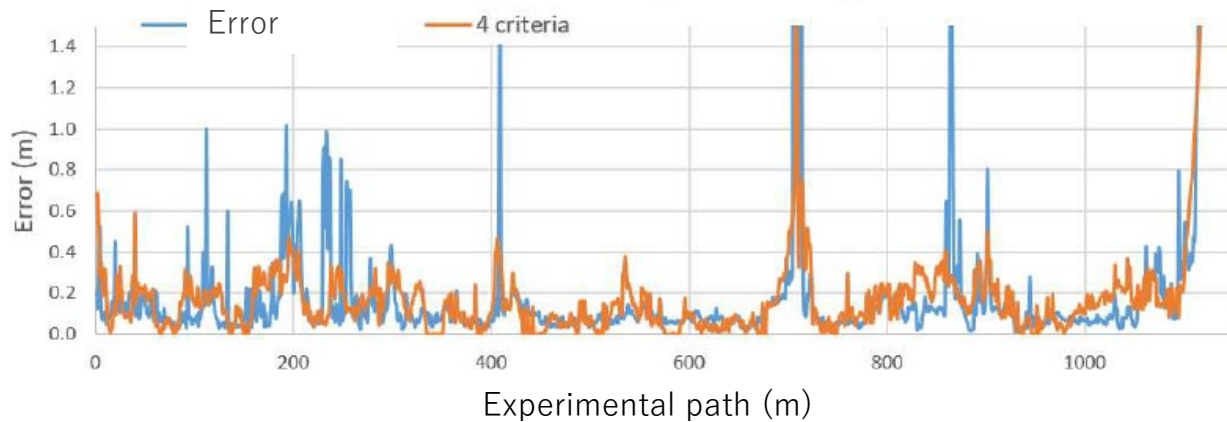
Results of error modeling

Localization error for 3.0m grid size NDT map

Modeled Error with PCR (without similarity) – Path #1



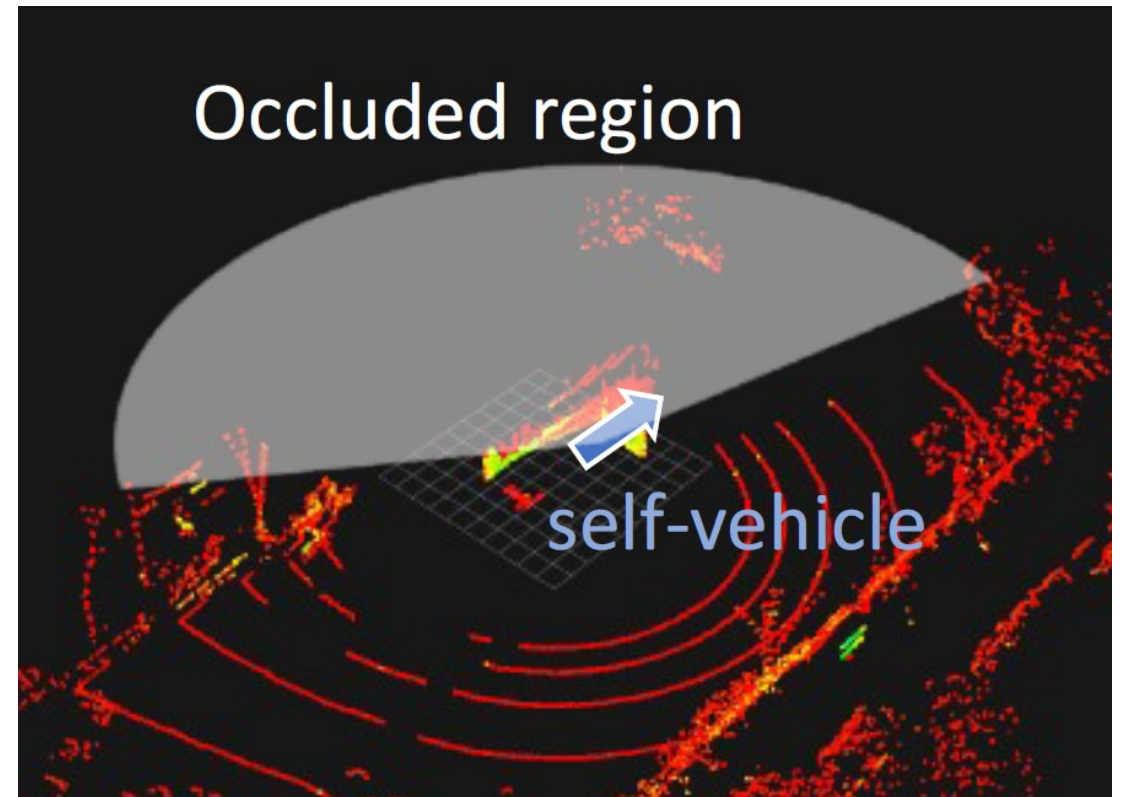
Modeled Error with PCR (with similarity) – Path #1



	R^2	RMSE	predErr < 10cm	predErr < 15cm	predErr < 20cm
Represent Quality	0.046	0.399	47.6%	65.2%	79.9%
+Feature Criteria	0.269	0.350	40.9%	60.2%	72.5%
+Layout Criteria	0.664	0.238	66.2%	83.3%	89.7%
+Local Similarity	0.747	0.207	70.7%	83.2%	88.9%

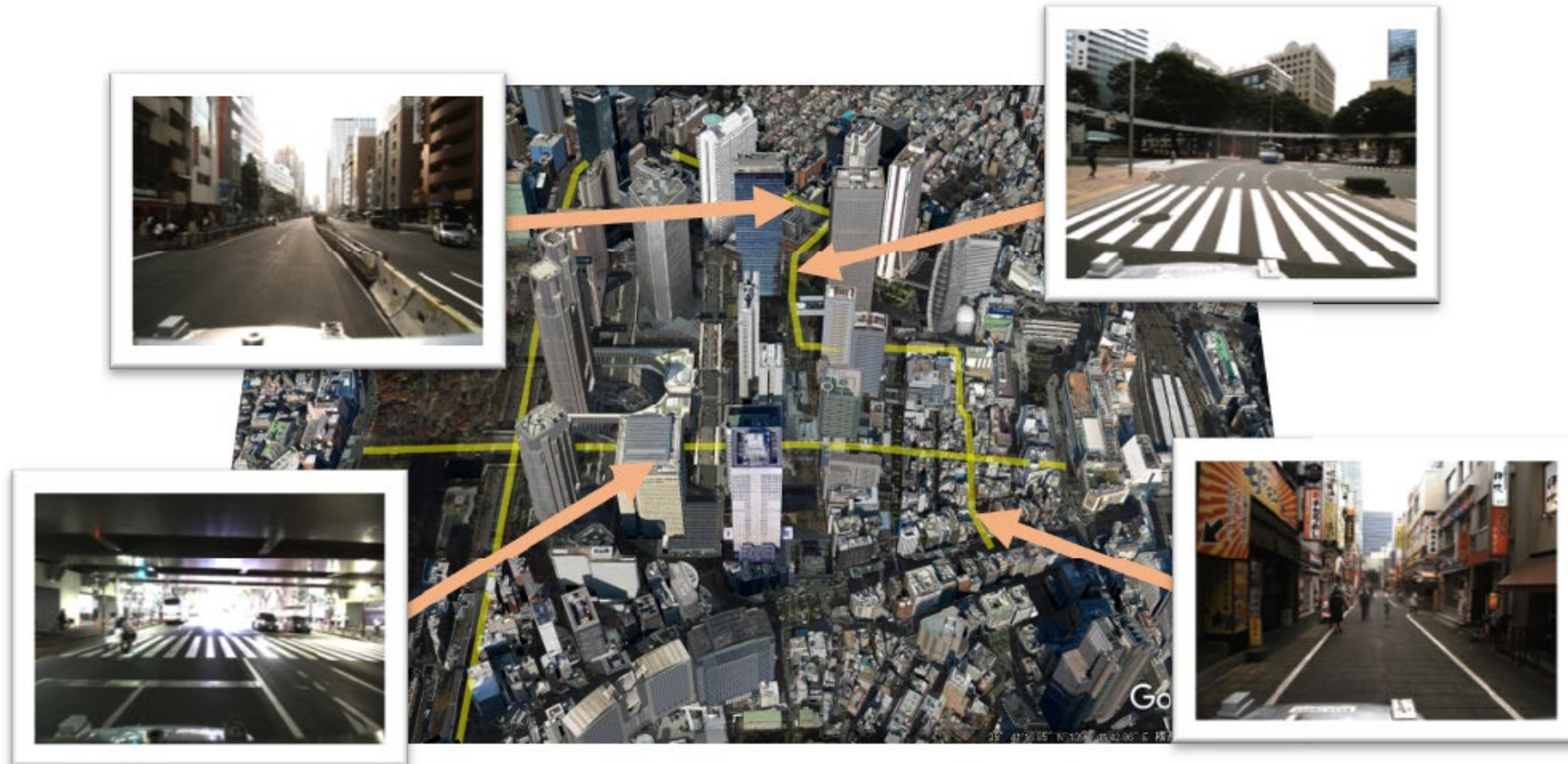
Occlusion Effects for Localization in Urban Scenario

- LiDAR beams are interrupted by tall vehicles in heavy traffics, and could not reach the reference infrastructures for the localization.
- The accuracy of the localization degenerates due to occlusion effects

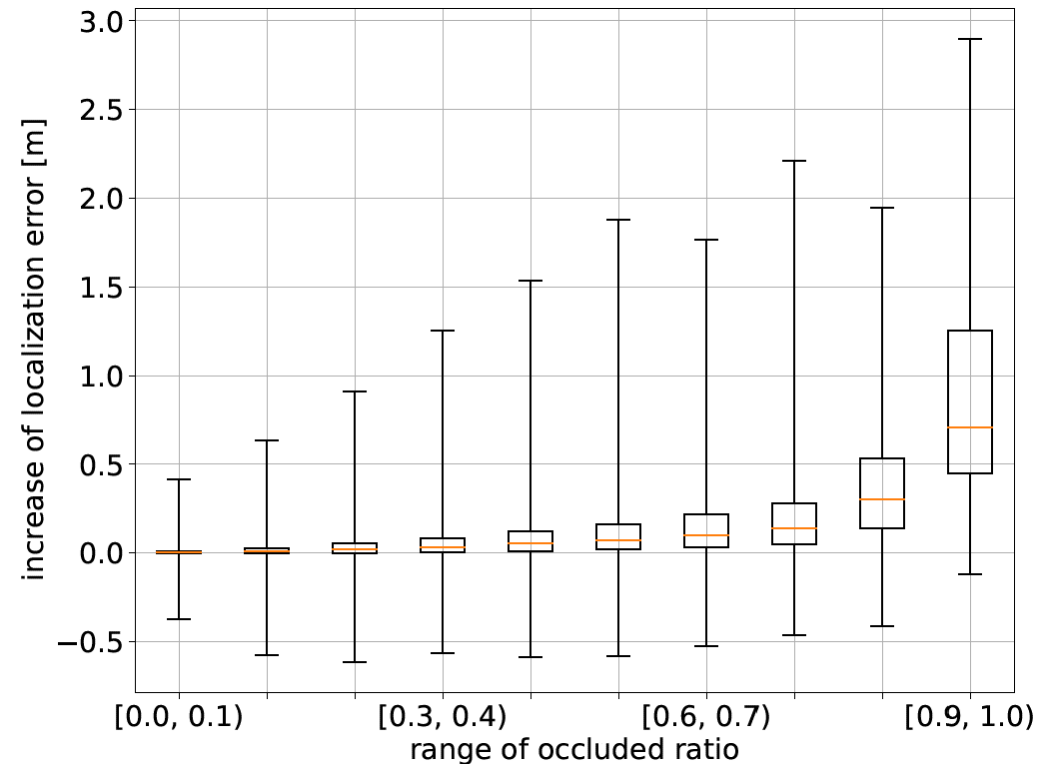
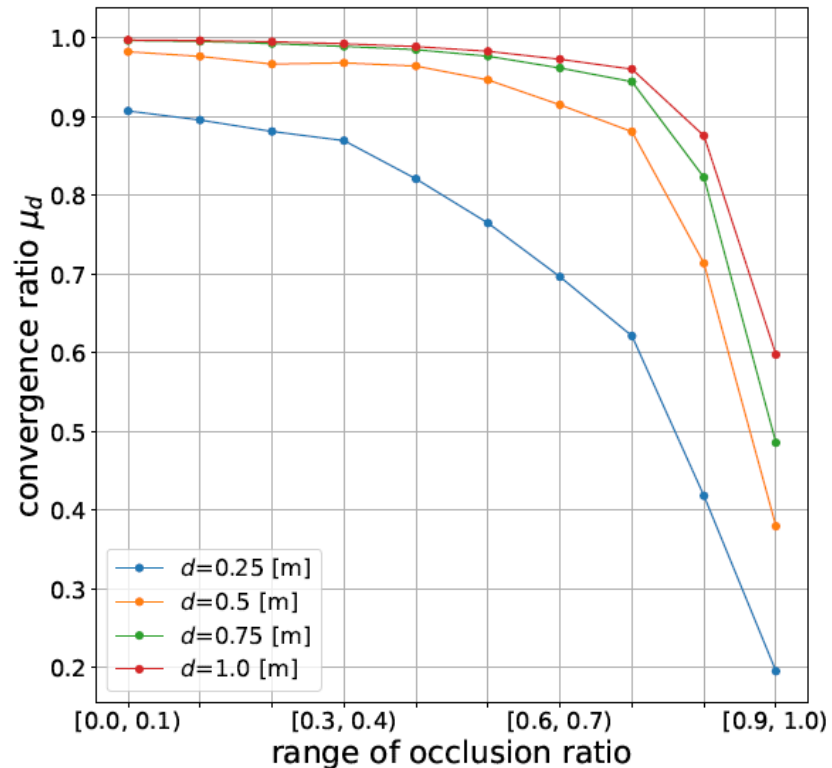


Test-bed to evaluate Occlusion effects in Urban Canyon

- The evaluations were performed in **Shinjuku, Tokyo**.
- Total length of the trajectory is **7.0km**.



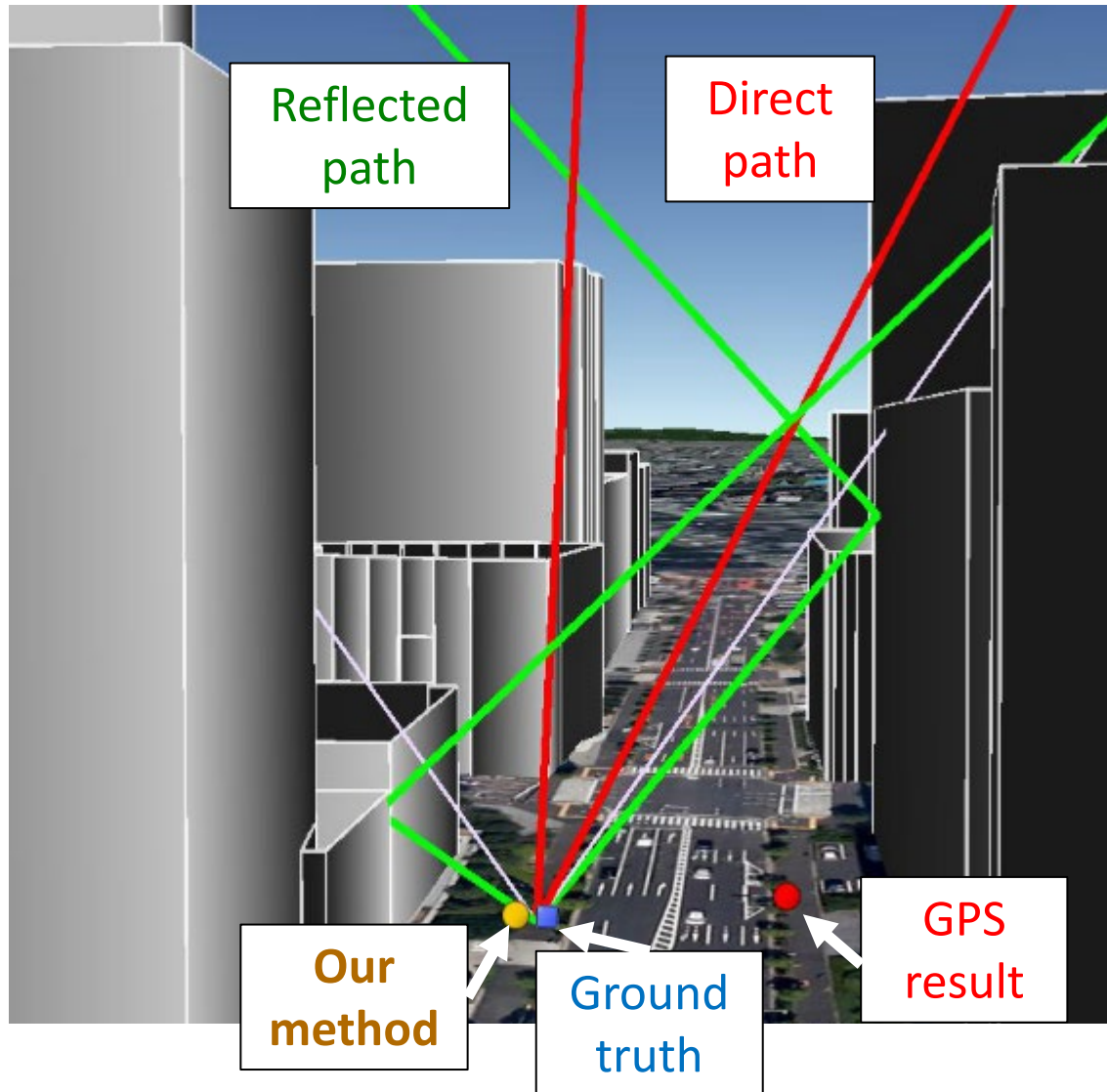
Evaluation of Occlusion effects: Convergence and Error



In the case which **occlusion ratio** is high:

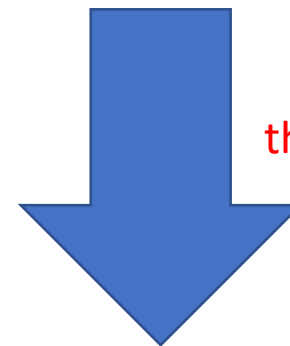
- **convergence ratio** becomes lower for the localization algorithm
- **localization error** becomes larger even in the converged case.

3D-GNSS Positioning in Urban Canyon



Signal Observation:

- Pseudorange
- RSSI (Received Signal Strength Indicator)
- Deceived positioning results



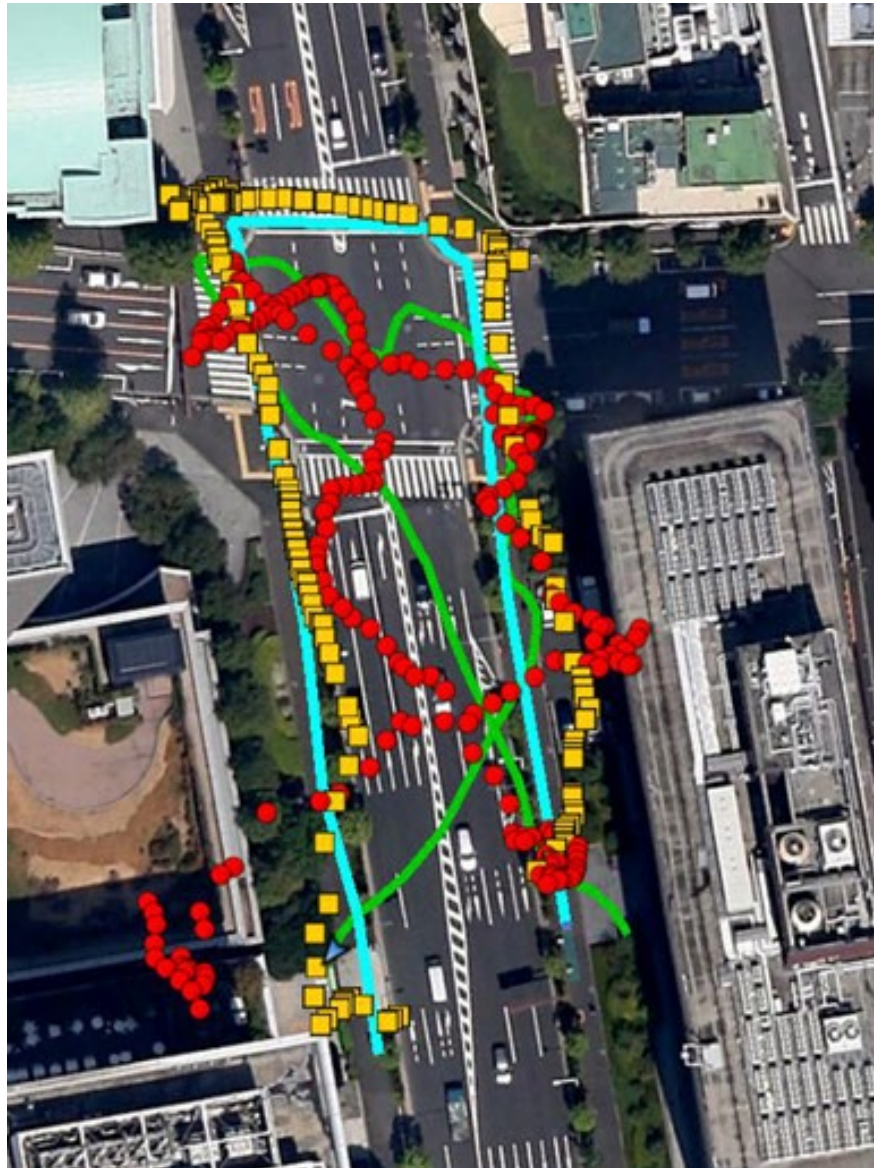
Find
the most consistent

Position Assumptions:

Estimated by ray tracing

- Pseudorange
- RSSI (Received Signal Strength Indicator)
- Deceived positioning results

Evaluations applying 3D method to GPS signals



- iPhone4S with WiFi
- u-blox NEO-6P
- Proposed (with NEO-6P)
- Ground Truth

Shunsuke Miura, Shoma Hisaka, and Shunsuke Kamijo,
"GPS Multipath Detection and Rectification using 3D
Maps", IEEE ITSC2013, pp.1528-1534, The Hague, The
Netherlands, Oct.6-9, 2013

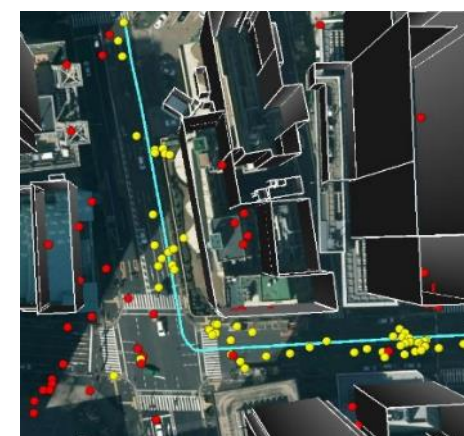
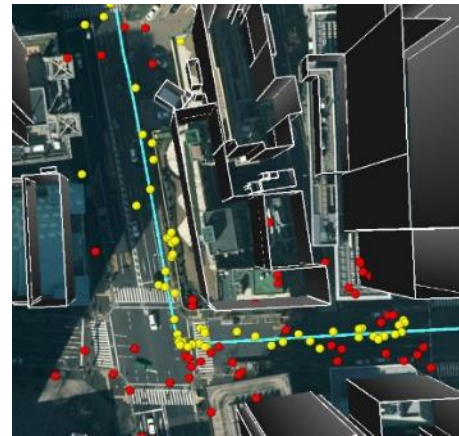
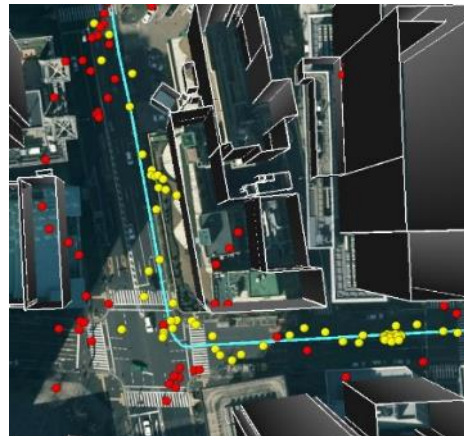
Experimental result: GPS

● Weighted least square (WLS) GPS
— Ground truth ● 3D map based GPS

Right turning



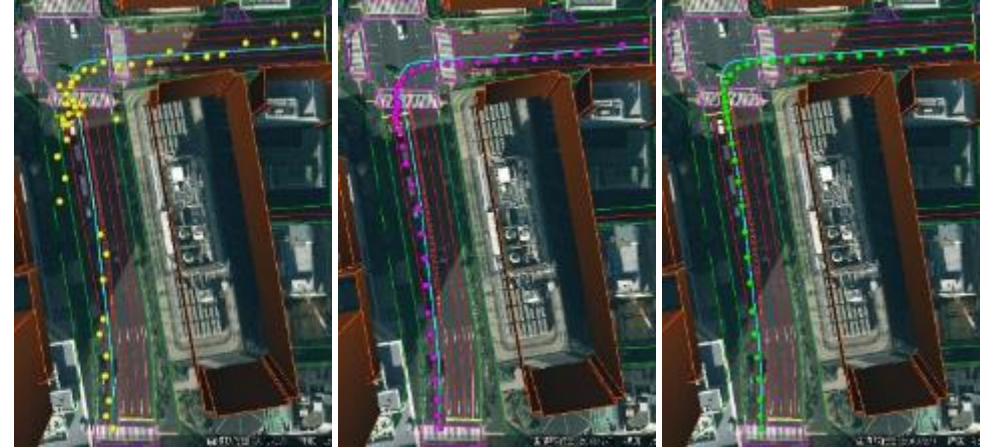
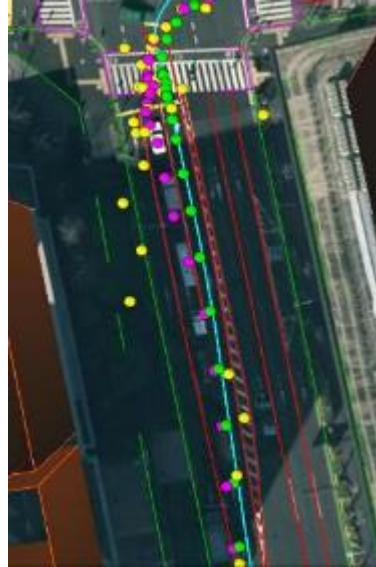
Left turning



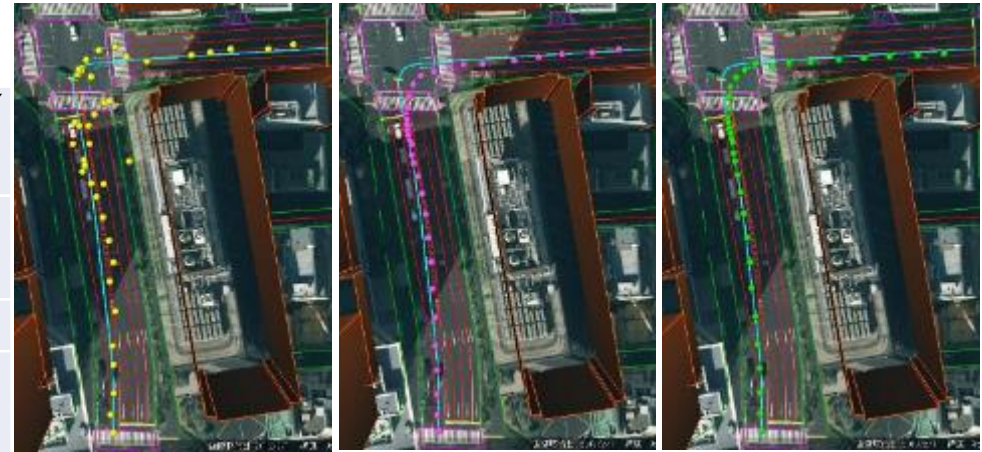
	Mean error (m)	Standard Deviation (m)	Maximum error (m)
WLS GPS	17.8	11.8	82.9
3D map based GPS	3.1	3.2	24.1

Experimental result for localization

- GNSS
- GNSS/INS
- GNSS/INS/lane detection
- Ground truth



(a) First right turning experiment



(b) Second right turning experiment

<i>Ex.</i>	<i>Evaluation method</i>	<i>GNSS</i>	<i>GNSS/INS</i>	<i>GNSS/INS/Lane Det.</i>
1	Positioning Error mean (meter)	3.44	1.79	0.73
	Correct lane rate	43.2%	59.1%	93.2%
2	Positioning Error mean (meter)	2.41	1.29	0.71
	Correct lane rate	55.0%	82.5%	95.0%

Case Study of Under the Bridge Environment

Experimental environment



Experimental equipment



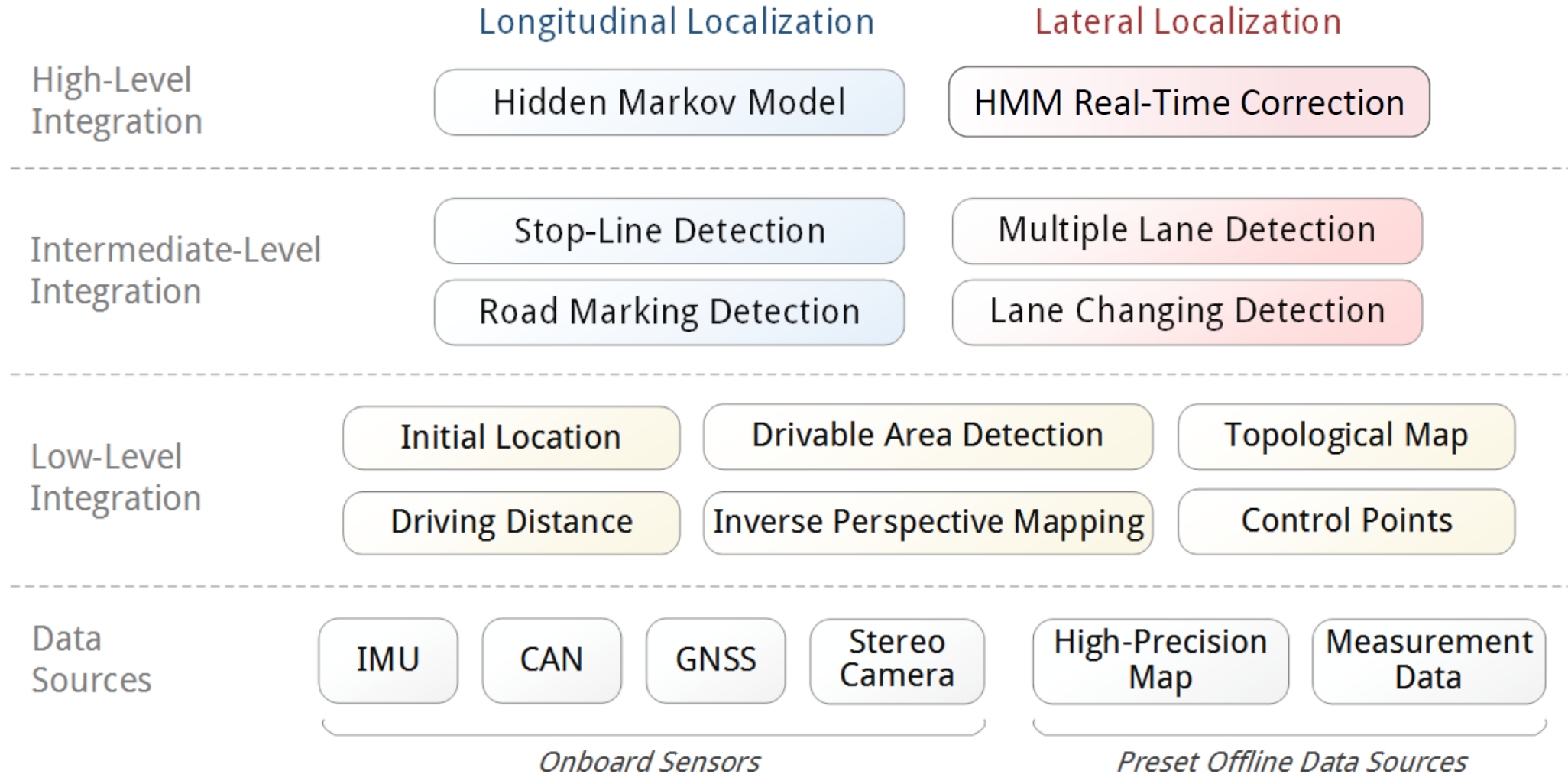
Stereo camera

Vehicle odometer



Proposed Vehicle Localization System

- ▶ Topological map & stereo camera & IMU/CAN → HMM → Position



Conclusions

- Benefits and problems of the digital map for autonomous vehicle application were discussed in this presentation.
- Problems and their solutions were exemplified from our research experiences with the evidence of experimental data.
- Need to stimulate the discussions on the digital map solutions for the successful cross bordered delivery and usage of autonomous vehicles.