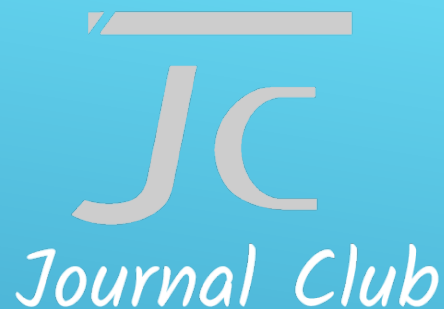




iMorpheus.ai
Weekly Journal Club

NDT localization outline

Friday 09 Mar 2018, 12:00PM



JOURNAL CLUB介绍与自动驾驶中定位方案相关的论文， 主要关注的方向有：
SLAM算法、点云数据的处理和压缩、特征地图、传感器数据处理和融合、GNSS信号处理等。我们一直关注领域前沿技术， 选取得到广泛认可的、或者是在我们的实际使用中结果比较好的论文， 与大家分享， 共同学习成长。

每周五 北京时间12点
<http://imorpheus.ai/journalclub>



扫码加入无人驾驶技术群

1. Problems in localization

2. Autoware VS Apollo

3. Options



1. Problems in localization

How to localize a moving vehicle with sensors && a computer.

- Without memory, GNSS || RTK+IMU is used.

- Noises || Weak signal || Multi-path Effect

- With memory ⇔ Maps, Sensors(GNSS/IMU, Lidar, Camera...) are used.

- How to build a HD map.

- How to localize with online data in a HD map.



2.1 Localization in autoware

Autoware & Manuals in github:

<https://github.com/CPFL/Autoware>

<https://github.com/CPFL/Autoware-Manuals>

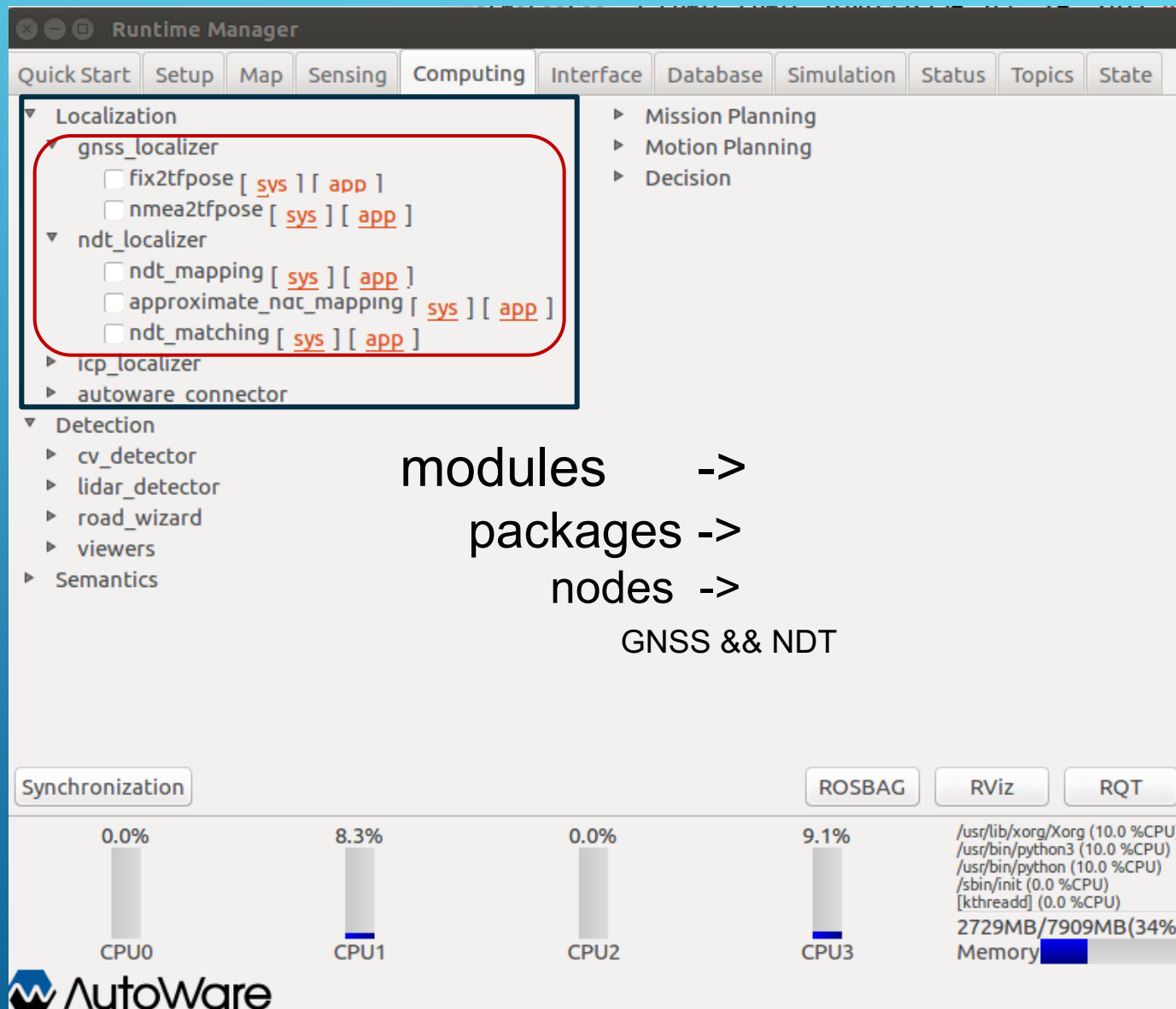
Website of autoware group:

<https://tier4.jp/en/>



iMorpheus.ai

2.1 Localization in autoware



The screenshot shows the Autoware Runtime Manager interface. The 'Localization' module is expanded, showing a tree structure of sub-modules. A red box highlights the 'gnss_localizer' and 'ndt_localizer' sub-modules. The 'gnss_localizer' sub-module contains 'fix2tfpose' and 'nmea2tfpose'. The 'ndt_localizer' sub-module contains 'ndt_mapping', 'approximate_ndt_mapping', and 'ndt_matching'. Each sub-module has a checkbox and a label indicating its status (sys or app). The 'Detection' module is also expanded, showing 'cv_detector', 'lidar_detector', 'road_wizard', 'viewers', and 'Semantics'. The bottom of the interface shows a 'Synchronization' section with four CPU bars (CPU0, CPU1, CPU2, CPU3) and a 'Memory' bar. The CPU usage is 0.0%, 8.3%, 0.0%, and 9.1% respectively. The memory usage is 2729MB/7909MB (34%).

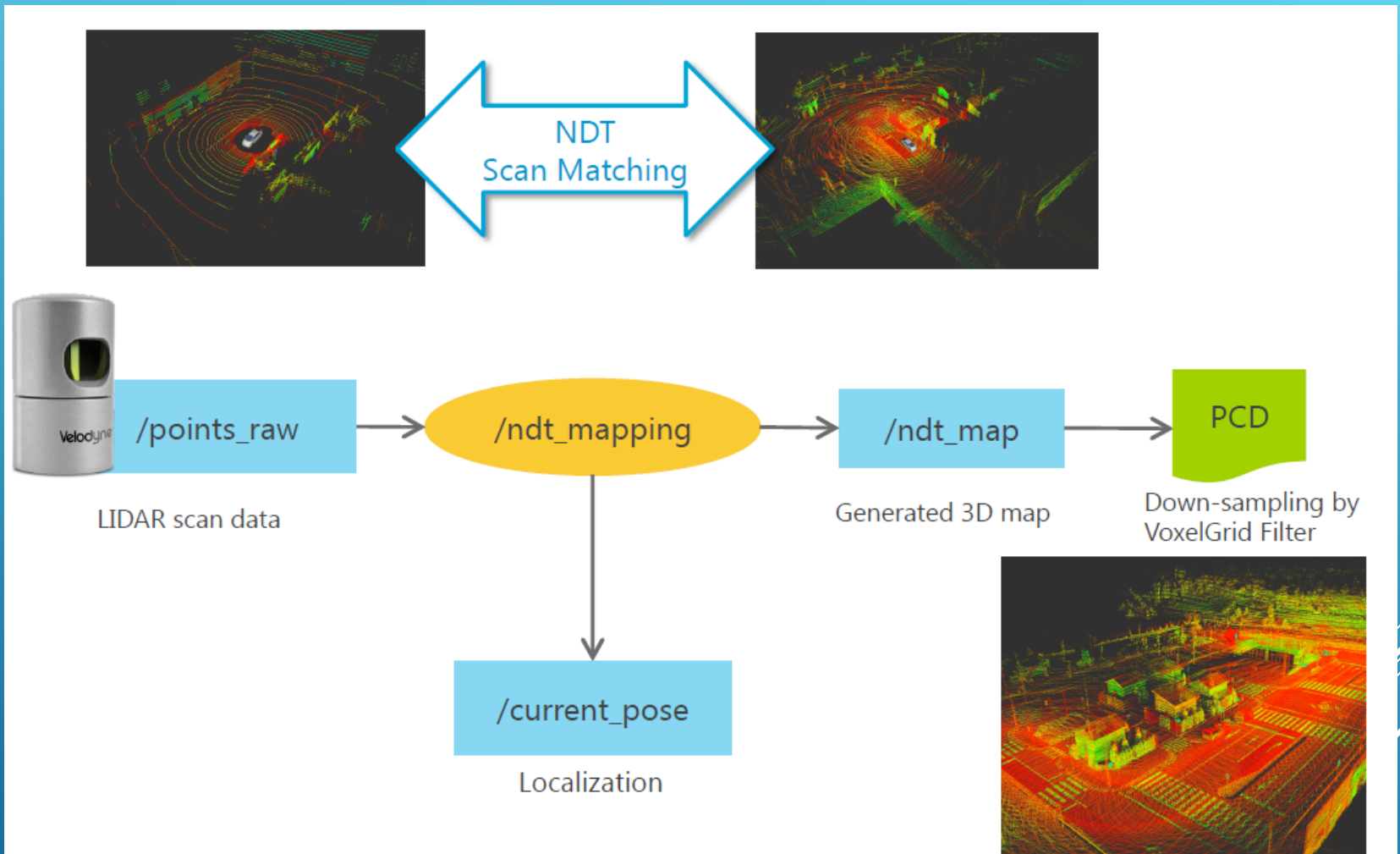
modules ->
packages ->
nodes ->
GNSS && NDT

CPU	Usage
CPU0	0.0%
CPU1	8.3%
CPU2	0.0%
CPU3	9.1%

Memory: 2729MB/7909MB (34%)

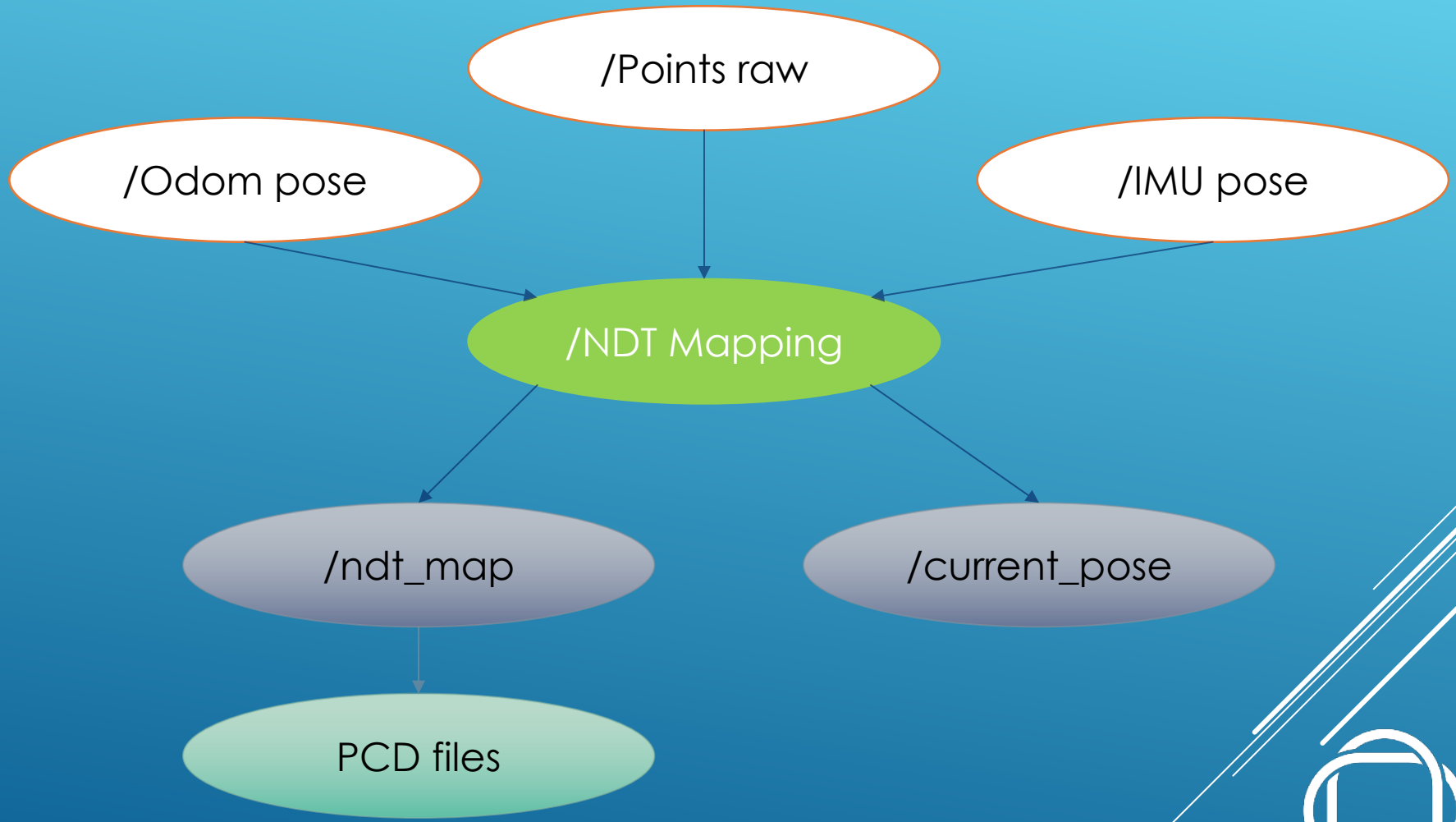


2.1 Localization in autoware-Mapping



Autoware manual of mapping, 2017

2.1 Localization in autoware-Mapping



2.1 Localization in autoware-Mapping

1. PCD file is in a local coordinate, you should transform the local map into a global map.
2. /map -> /world

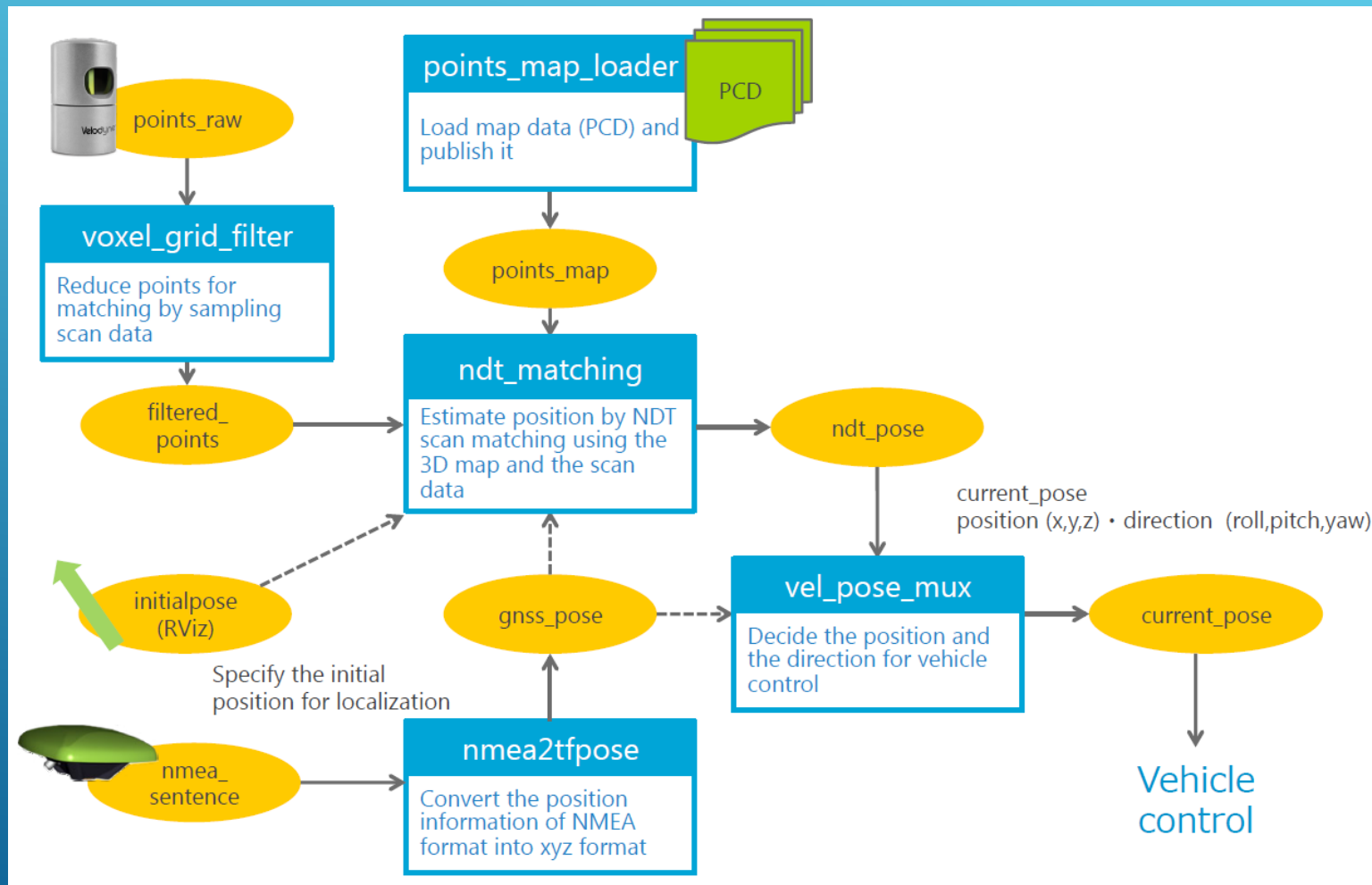
Autoware uses tf_mapping to do these job.

Given a initial 6-DOF value to construct rotation and translation matrix.

3. NDT Mapping -> same problems in SLAM.
4. Without RTK, even in fixed scenes.
5. No deal with dynamic objects-noises.



2.1 Localization in autoware-Matching



2.1 Localization in autoware-Matching

1. Input data

- ❑ online lidar pcd -> local coordinate
- ❑ offline pcd map -> global coordinate
- ❑ initial pose(6-DOF) || gnss_pose (gnss+ yaw, pitch, roll)

gnss_pose is needed in two cases.

2. Model in NDT matching

linear || quadratic || none

3. Initial value is critical important.

We have tested that: with 5-10meters errors, the software crashes.

4. Single localization scheme with simple algorithm.



2.2 MSF in apollo

Apollo in github:

<https://github.com/ApolloAuto/apollo>

For understanding MSF, three papers are needed:

Stanford

2007, Map-Based Precision Vehicle Localization in Urban Environments

2010, Robust Vehicle Localization in Urban Environments Using Probabilistic Maps

BaiDu

2017, Robust and Precise Vehicle Localization based on Multi-sensor Fusion in Diverse City Scenes



iMorpheus.ai

2.2 Localization in apollo

Apollo 1.0

RTK method -> without memory(map)

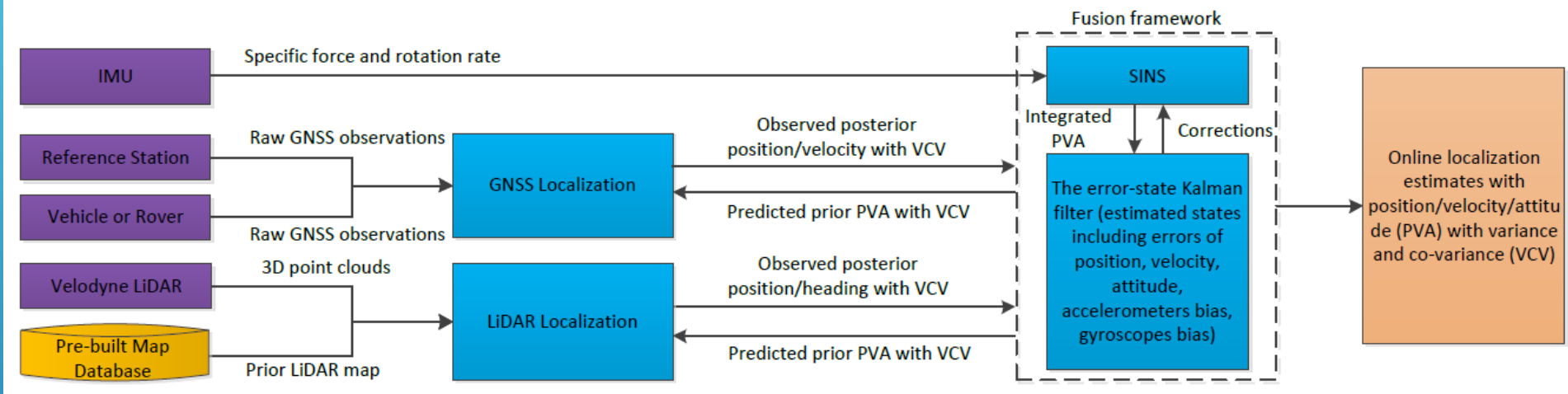
Apollo 2.0

MSF method -> with memory(map)

Map building && localization



2.2 MSF in apollo



1. Lidar based localization

- ❑ Intensity Map & Altitude Map
- ❑ Heading Estimation <- Lucas Kanade Algorithm
- ❑ Horizontal Localization <- Histogram filter

2. GNSS based localization

- ❑ Raw data received & RTK algorithm mentioned
- ❑ Tightly coupled with INS

3. Sensors fusion localization

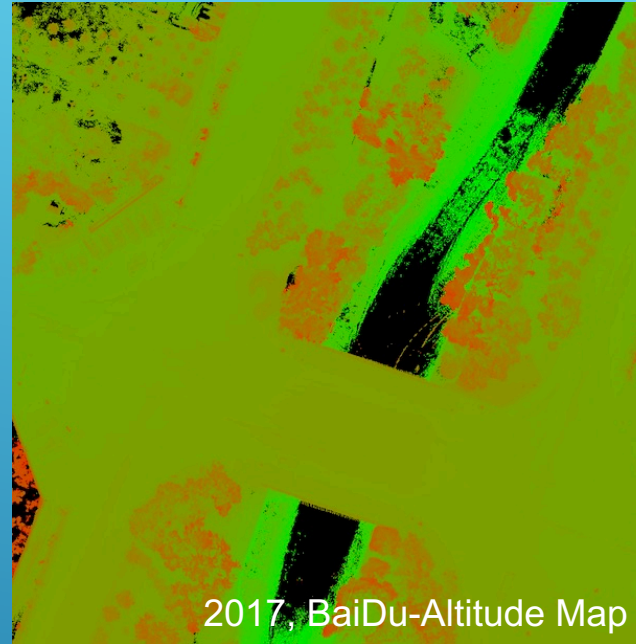
- ❑ An error-state KF is applied.



2.2 MSF in Apollo-Lidar based localization



2.2 MSF in Apollo-Lidar based localization



1. Altitude Map, 3D info is added.

2. Map Building Method

- With good GNSS signal reception

The GNSS/INS solution based on post-processing algorithms, such as the NovAtel Inertial Explorer, is able to produce enough accurate vehicle motion trajectories.

- In weak GNSS signal scenarios

BD treat it as a classic map reconstruction problem combining several techniques including NovAtel IE post-processing, LiDAR SLAM, loop closure, and the global pose-graph optimization.



2.2 MSF in Apollo-Lidar based localization

Algorithm 1 LiDAR-based localization

Input: Prior map m , online point cloud z , rough transformation $T_0 = (x_0, y_0, a_0, \phi_0, \theta_0, h_0)$ and search space X, Y .

Output: Best registration $(\hat{x}, \hat{y}, \hat{a}, \hat{h})$, and covariance matrix \mathcal{C}_{xy} .

```
1:  $\hat{h} \leftarrow$  heading angle estimation ▷ IV-A
2:  $\hat{a}_0 \leftarrow m(x_0, y_0)$  ▷ Get altitude from map
3: Transform  $z$  with the transformation  $(x_0, y_0, \hat{a}_0, \phi_0, \theta_0, \hat{h})$ 
4: for  $x_i, y_i \in \{x_0 + X, y_0 + Y\}$  do
5:    $P_r \leftarrow SSD_r(x_i, y_i, z, m)$  ▷ Equ. 6 8
6:    $P_a \leftarrow SSD_a(x_i, y_i, z, m)$  ▷ Equ. 7 8
7:    $P(z|x_i, y_i, m) \leftarrow (P_r)^\gamma \cdot (P_a)^{1-\gamma}$  ▷ Equ. 5 9 10
8:    $P(x_i, y_i) \leftarrow P(z|x_i, y_i, m) \cdot (\bar{P}(x_i, y_i))^{1/\kappa}$  ▷ Equ. 4
9: end for
10:  $(\hat{x}, \hat{y}) \leftarrow \{P(x_i, y_i)\}$  ▷ Equ. 11
11:  $\mathcal{C}_{xy} \leftarrow \{P(x_i, y_i)\}$  ▷ Equ. 12
12:  $\hat{a} \leftarrow m(\hat{x}, \hat{y})$  ▷ Get altitude from map
13: return  $(\hat{x}, \hat{y}, \hat{a}, \hat{h}, \mathcal{C}_{xy})$ 
```



2.2 MSF in Apollo-GNSS based localization

1. A RTK solution is proposed.

RTK-LIB is referenced in this part.

2. A tightly coupled solution is presented.

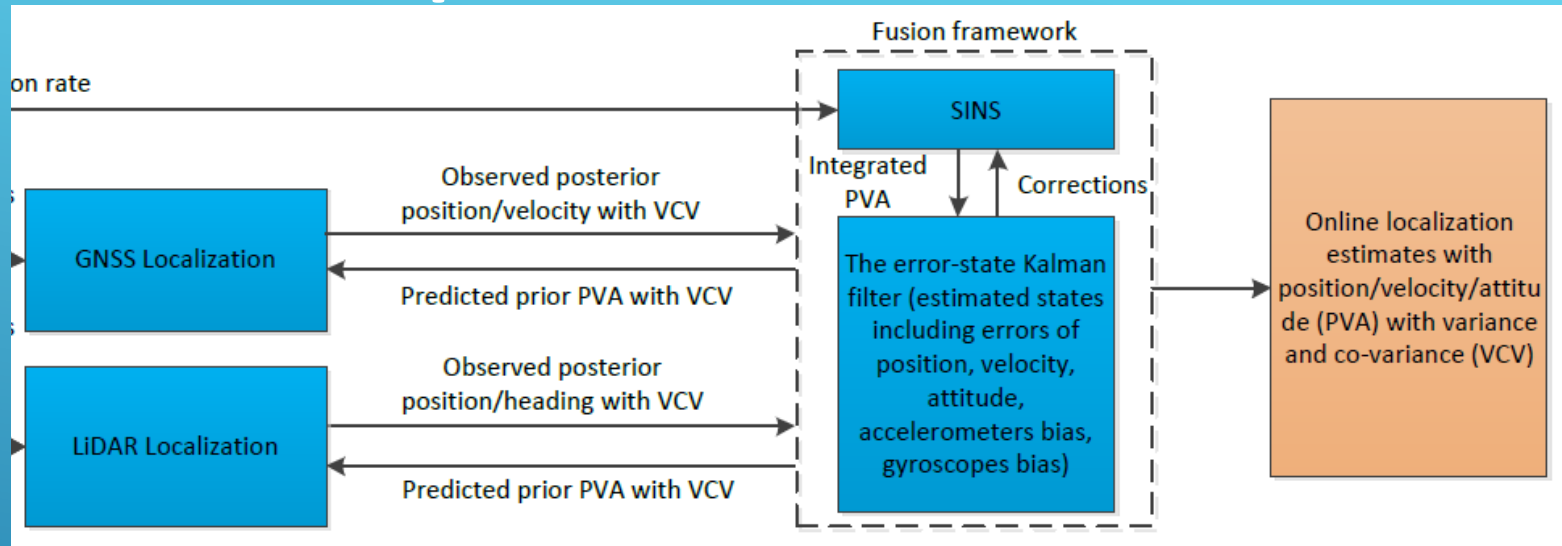
Currently, Baidu only loosely couples the sensor observations, which means GNSS results are come from hardware.

Option 1: NovAtel SPAN-IGM-A1

Option 2: NovAtel SPAN® ProPak6™ and NovAtel IMU-IGM-A1



2.2 MSF in Apollo-Fusion-Error-state KF



1. SINS Kinematics Equation And Error Equation
2. Filter State Equation
3. Measurement Update Equation
 - 3.1 LiDAR measurement update equation
 - 3.2 GNSS measurement update equation



3. Options

High Precision Localization absolutely needs memory \Leftrightarrow Map.

During building map, we should concern:

1. Which kind of map is needed.

Raw PCD || Feature Map || Intensity Map || Altitude Map...

2. How to control map precision.

RTK Solution && Valid Satellites Collection && MSF Algorithm...

3. How to deal with dynamic objects.

Ground Constrained && Statistics Features...



3. Options

High Precision Localization absolutely needs sensors.

During online localization, we should concern:

1. Robustness & Precision & Efficiency

GNSS/IMU & LiDAR & Cameras & MSF Algorithm





iMorpheus.ai Weekly Journal Club

Next Friday, 16/03/2018 12:00PM GMT+8

Hand-eye Calibration

关键词：手眼标定，Machine Vision（机器视觉），Sensor Fusion（传感器融合）

Website : <http://imorpheus.ai>
Email Address : live@imorpheus.ai



iMorpheus.ai