

Autoware Architecture Proposal

Architecture proposal for AWF by Tier IV Inc.

Agenda

1. Why we need a new architecture
2. Considered use cases
3. Architecture overview
 - Layered architecture
 - High level introduction of each module
4. Contribution steps to Autoware community
5. Conclusion



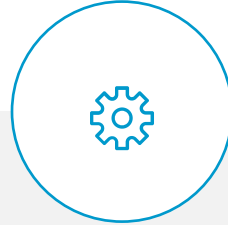
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Why we need a new architecture



Problem

- It's difficult to improve Autoware.AI capabilities



Why?

- No concrete architecture design
- A lot of technical debt
 - Tight coupling between modules
 - Unclear responsibility of modules



Tier IV's proposal

- Define a layered architecture
- Clarify the role of each module
- Simplify the interface between modules

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Considered use cases



Example use cases that were considered during architecture design:

Module	Use cases
Sensing	<ul style="list-style-type: none">• 360-degree sensing by the camera-LiDAR fusion
Perception	<ul style="list-style-type: none">• Recognition of dynamic objects and traffic lights
Localization	<ul style="list-style-type: none">• Robust Localization using multiple data sources
Planning	<ul style="list-style-type: none">• Route planning, dynamic planning based on vector map (not only waypoint following)• Automatic parking• Object avoidance
Control	<ul style="list-style-type: none">• High control performance on many kinds of vehicle-controllers

Features that are not considered yet (for the sake of development speed)

- Real-time processing
- HMI / Fail safe / Redundant system / State monitoring system / etc...

Will consider these items at AWF WGs

Module	Contents
Whole	<u>Scenario demo</u>
Sensing+Perception	<u>360° FOV</u> , <u>Prediction</u>
Localization	<u>Robustness of localization</u> , <u>Return from error</u>
Planning	<u>Lane change</u> , <u>Obstacle avoid</u> , <u>Parking</u>
Control	<u>Slow brake (normal stop)</u> , <u>Rapid brake (emergency stop)</u>

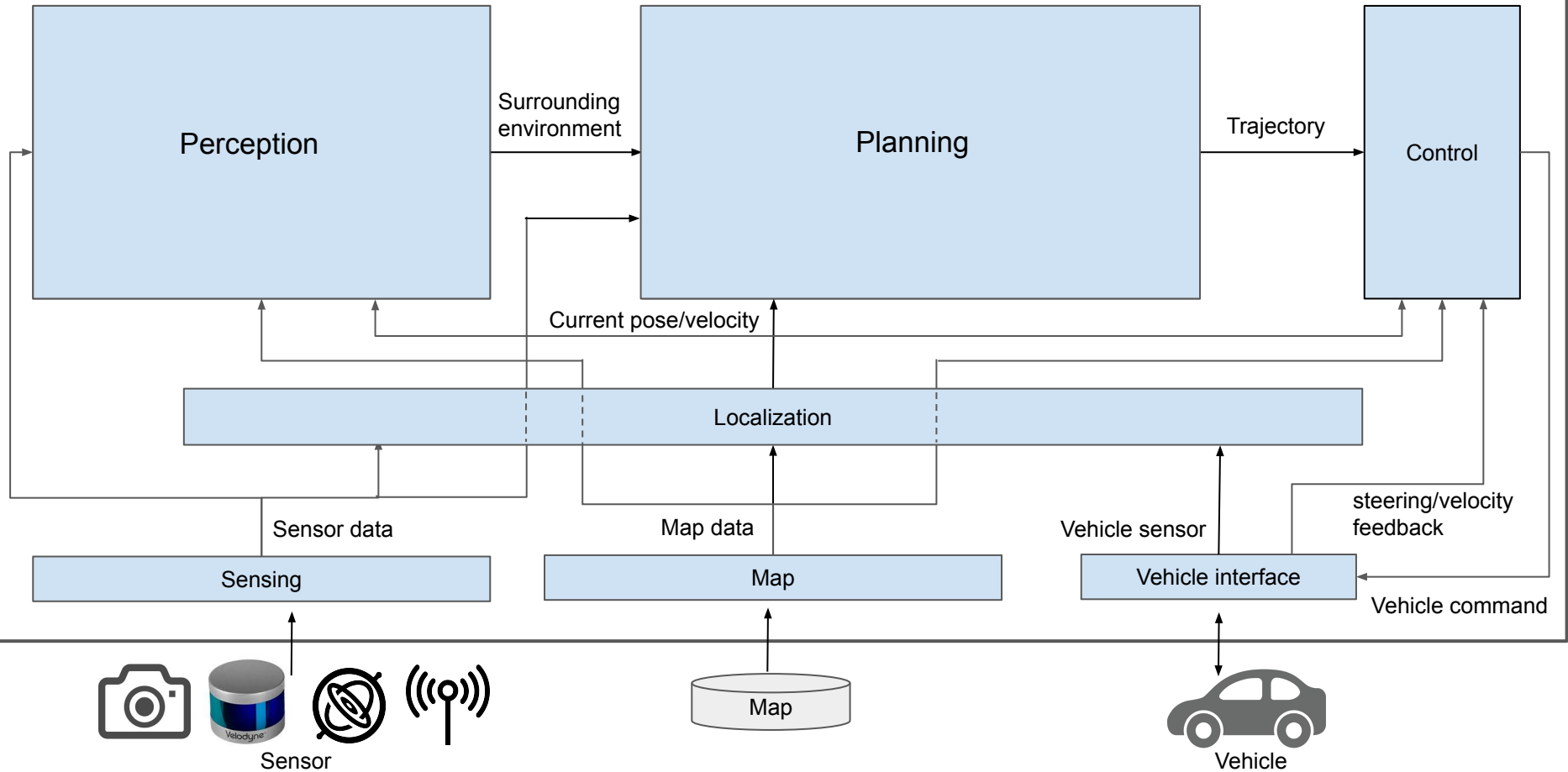
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Architecture overview

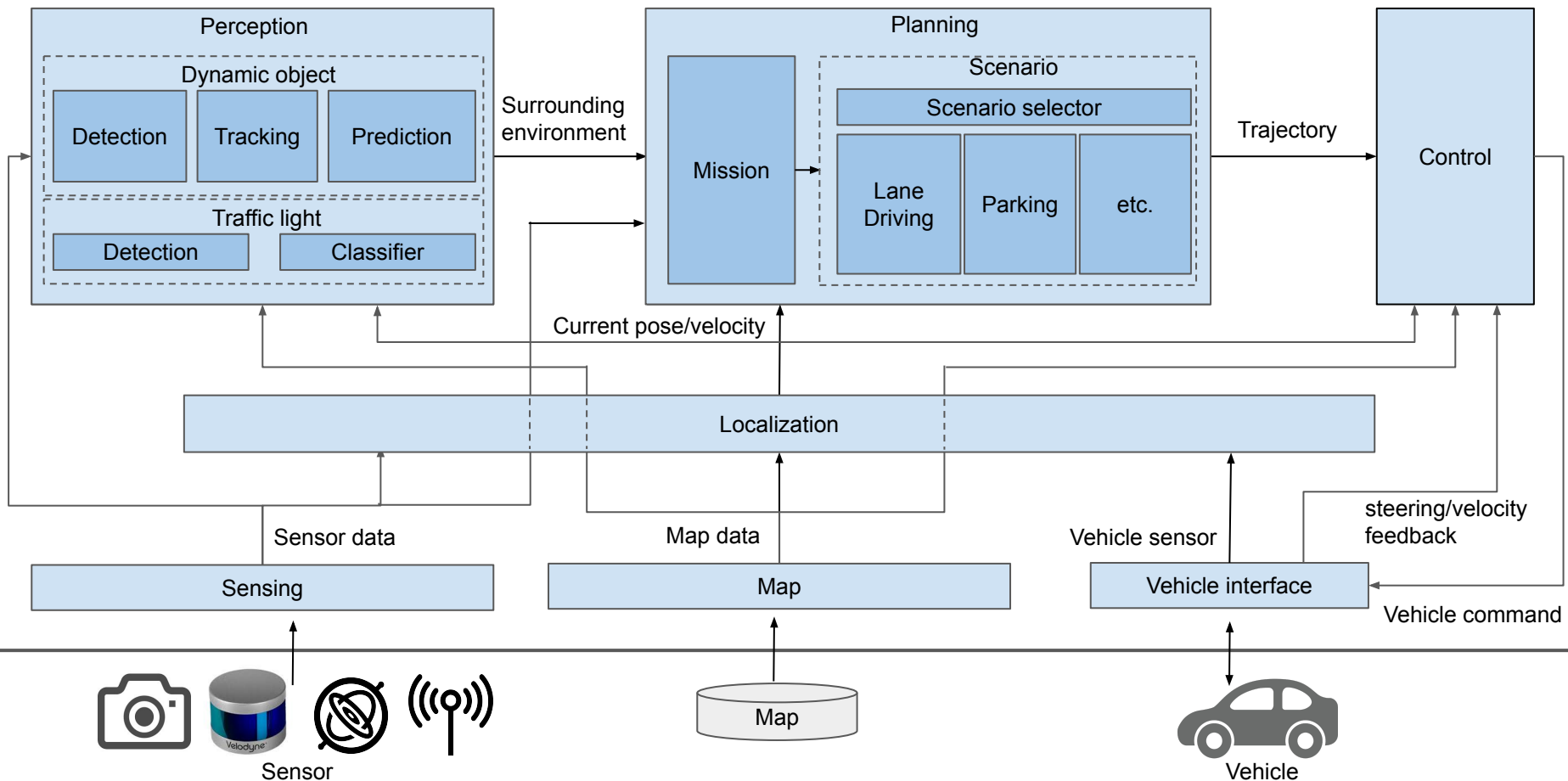
Layered architecture

High level introduction of each module

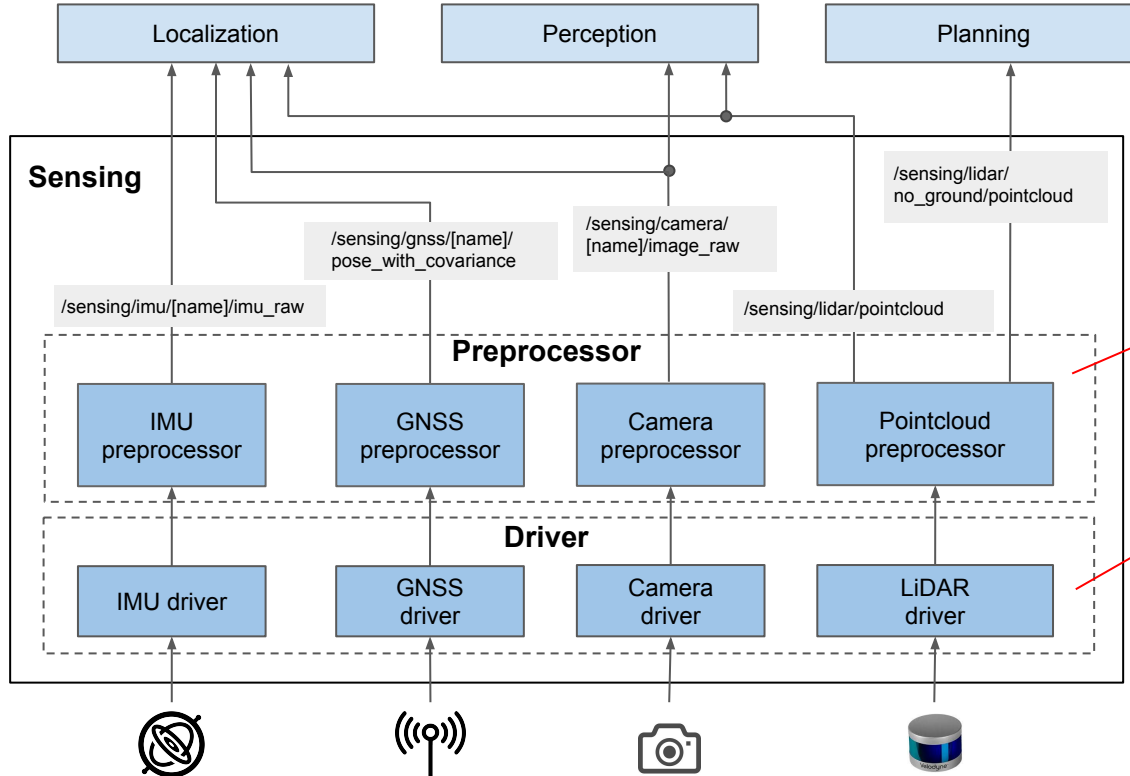
Autoware



Autoware



Role : Conversion of sensing data to ROS message

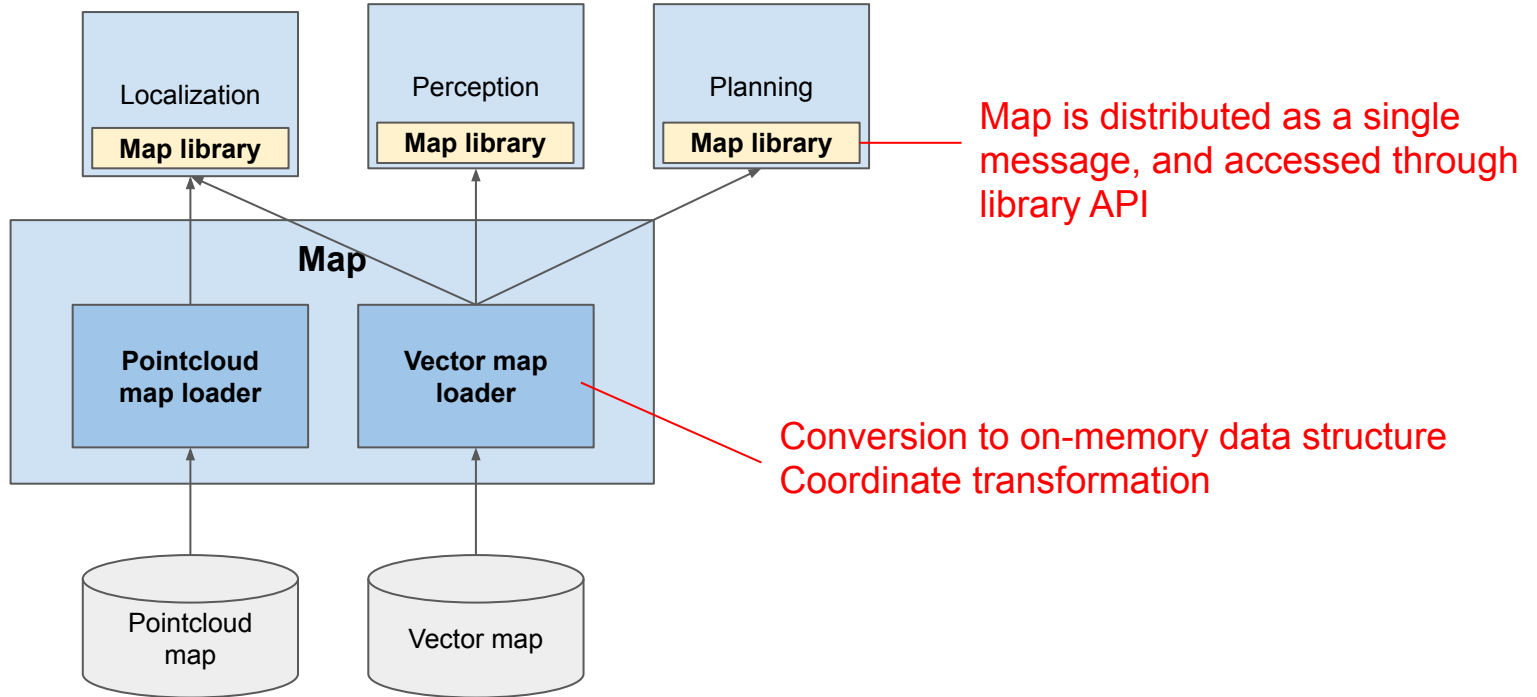


Examples of preprocessors

- Distortion correction
- Outlier filter
- Concat filter
- Ground filter

ROS drivers (mainly from upstream)

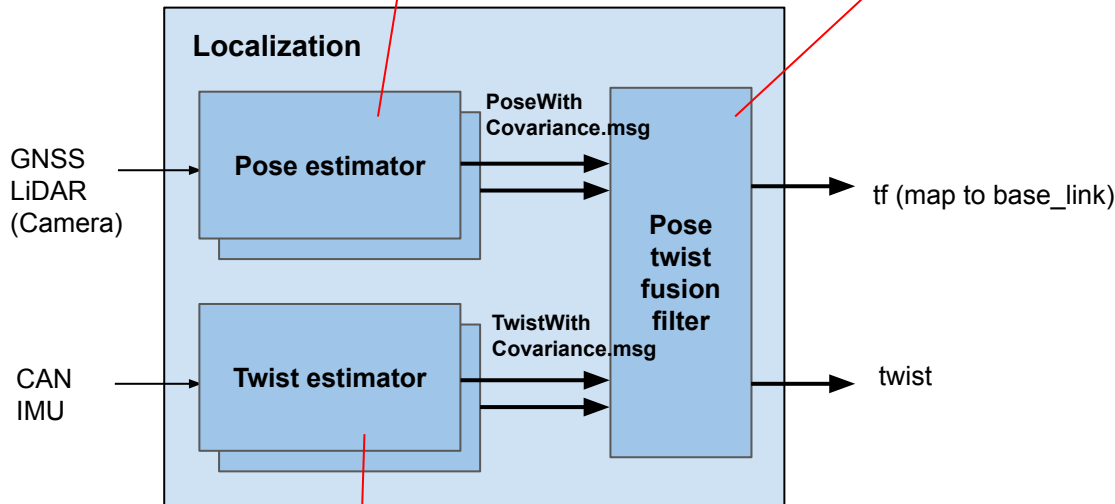
Role : Distribute static environment information to other modules



Role : Integration of each sensor data and estimation of self-pose and self-twist

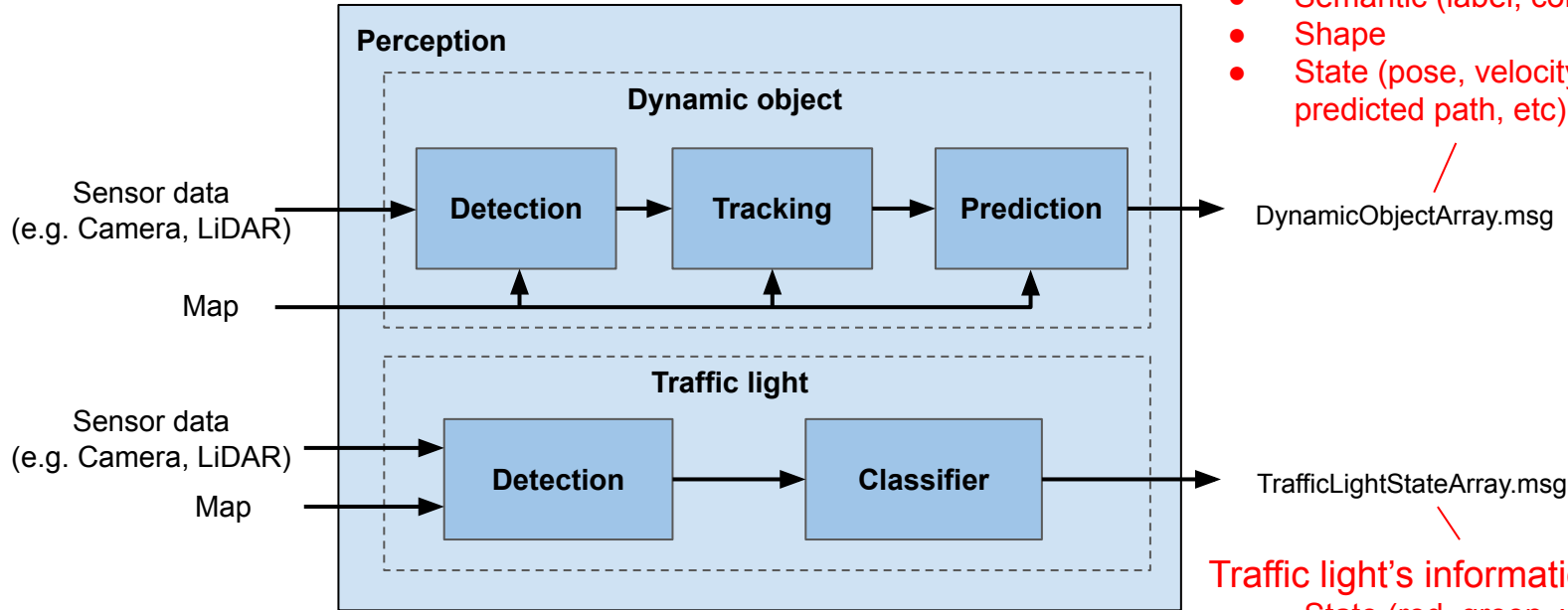
Support multiple localization methods
based on different sensors(LiDAR/Camera/GNSS)

Fuse pose and twist (e.g.EKF)



Merge sensor inputs

Role : Dynamic object recognition and traffic light recognition



Dynamic object's information including:

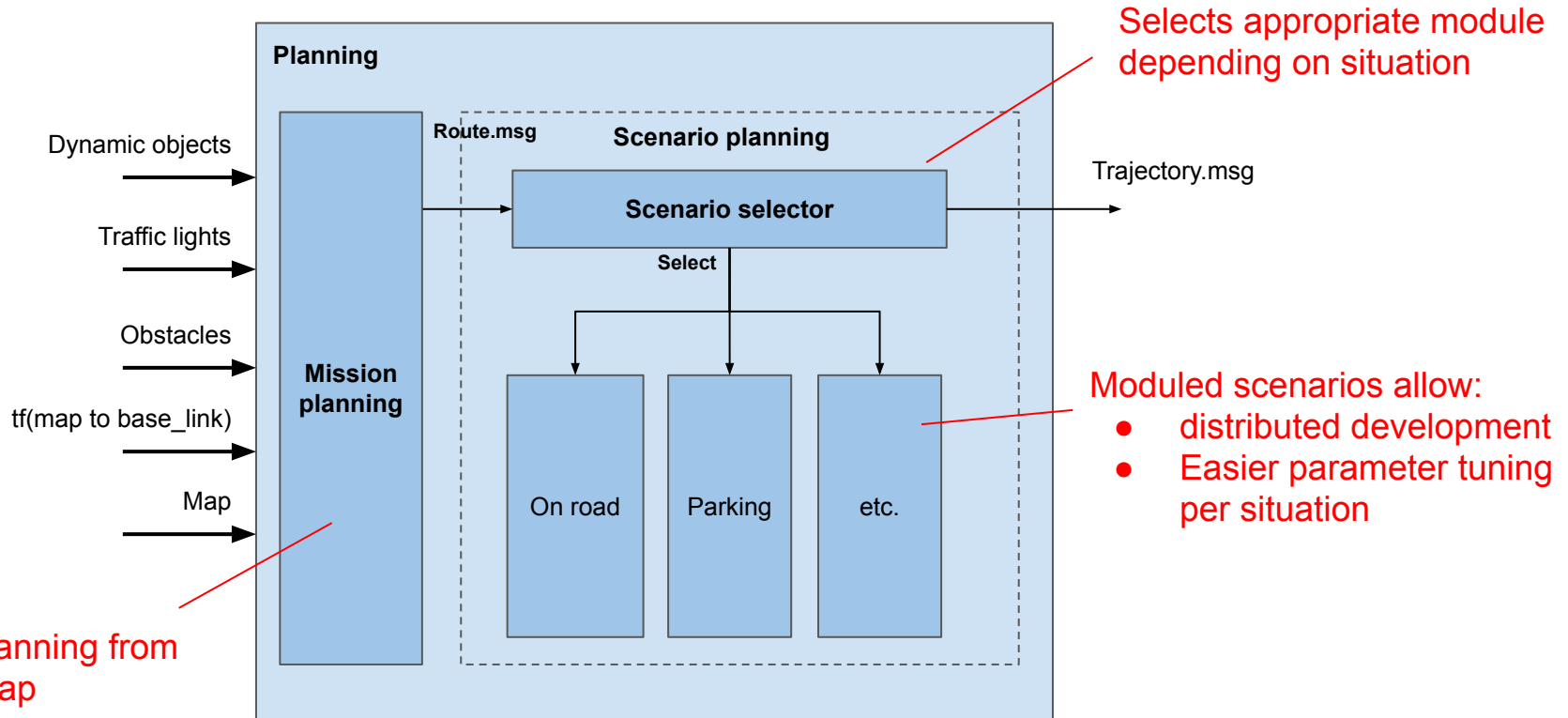
- ID
- Semantic (label, confidence)
- Shape
- State (pose, velocity, acceleration, predicted path, etc)

`DynamicObjectArray.msg`

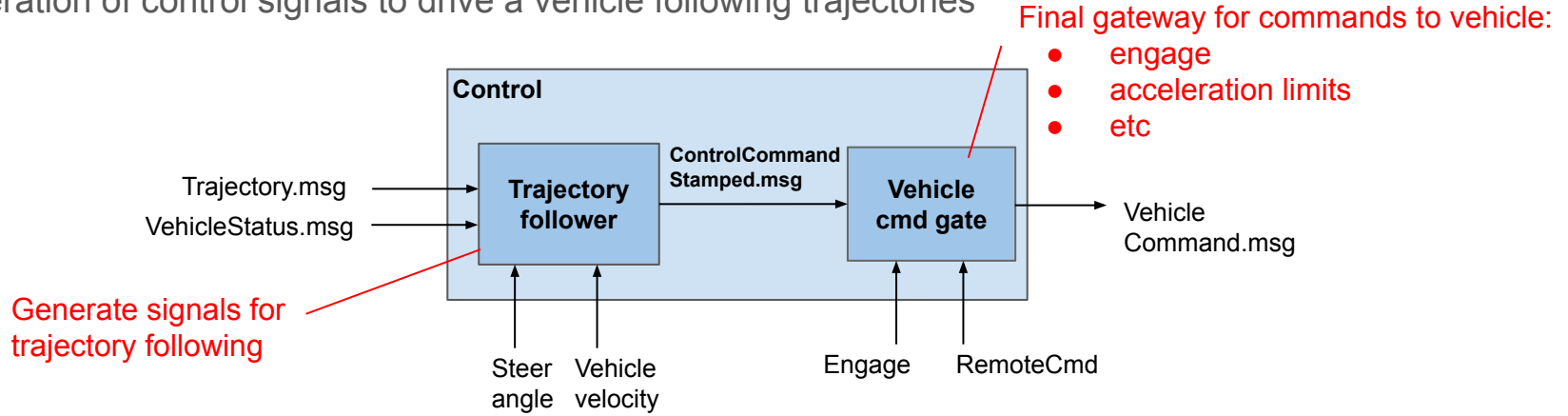
Traffic light's information including:

- State (red, green, yellow, etc)
- Could be customized for unique traffic light types.

Role : From a calculation of the route to the goal to drive trajectory generation



Role : Generation of control signals to drive a vehicle following trajectories



- engage
- acceleration limits
- etc

```
VehicleCmd.msg (autoware.ai)
-----
header
steer_cmd
-header
-steer
accel_cmd
-header
-accel
brake_cmd
-header
-brake
gear
twist_cmd
-header
-linear
-angular
ctrl_cmd
-linear velocity
-linear acceleration
-steering angle
```

Remove redundant output

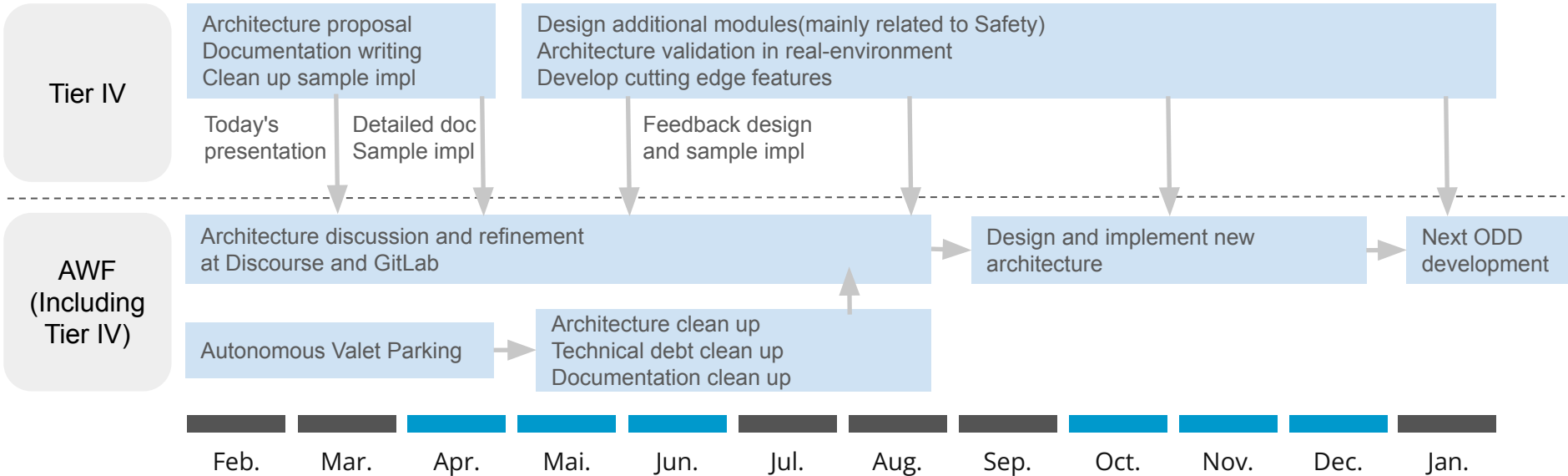
```
VehicleCommand.msg (proposed architecture)
-----
header
control
-steering angle
-velocity
-steering angle velocity
-acceleration
shift
-data
```

Support low level control

4

Contribution steps to Autoware community

Contribution steps to Autoware community



5

Conclusion



- We proposed a new architecture of Autoware
- Tier IV will contribute our achievements to Autoware community
 - Detailed design documents
 - Reference source code
- We would like to discuss the new architecture at WGs
 - Any feedback is welcome
- We will make an effort to implement the new architecture in Autoware.Auto

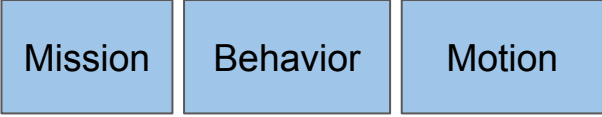
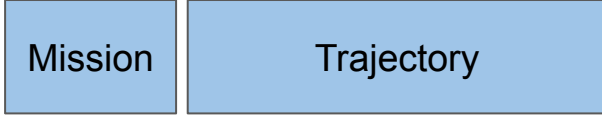
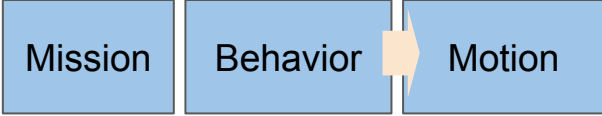
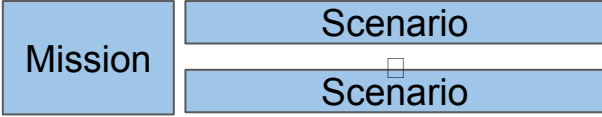
Appendix

- Design alternative
- Module implementation
 - Implementation details
 - Achievements

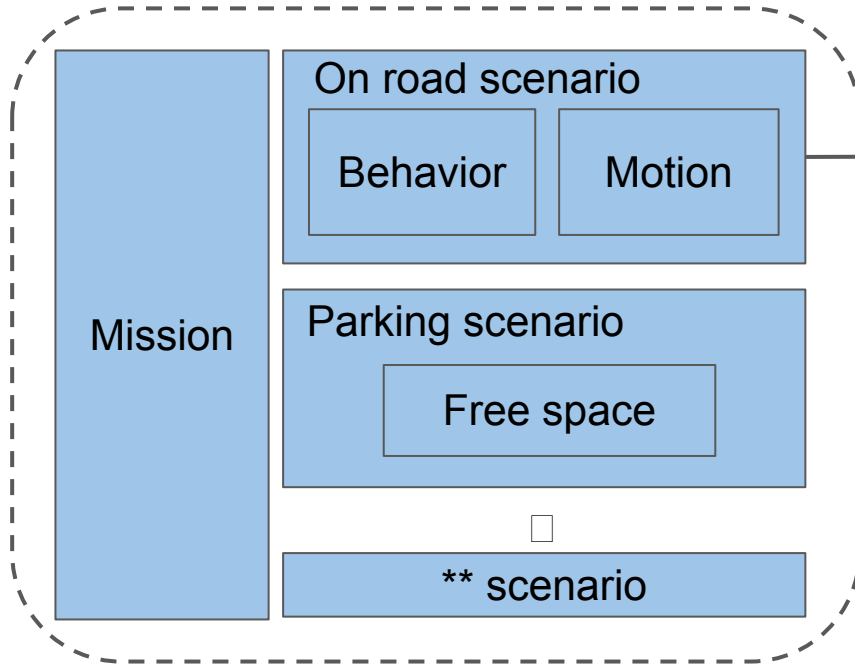
- Design alternative
- Module implementation
 - Implementation details
 - Achievements

[Design alternative] Planning (1/2)

Pre-study of the planning architecture

Type	Structure	Pros.	Cons.
BOSS type (Junior type)		<ul style="list-style-type: none"> • Structure is intuitive and easy to understand. • Easy to handle separately. 	<ul style="list-style-type: none"> • Behavior has to make a conservative decision in order to reduce discrepancies with a motion's decision.
CMU Dr. paper type		<ul style="list-style-type: none"> • It solves and optimizes both behavior and motion simultaneously to overcome a cons. of the BOSS type. 	<ul style="list-style-type: none"> • Difficult to improve if any issue was found. • Optimization of the whole autonomous driving functions is not realistic..
Victor Tango type		<ul style="list-style-type: none"> • Behavior and motion are tightly coupled to overcome a cons. of the BOSS type. • Easy to improve if any issue was found 	<ul style="list-style-type: none"> • Difficult to take a new research result into the modules.
Apollo type		<ul style="list-style-type: none"> • Scenario is one layer upper concept. • We can adopt preferred types of the planning structure according to each scenario. 	<ul style="list-style-type: none"> • In the scenario, cons. such as things described above are still remained. • If scenarios increased, switching mechanism can be complicated.

Finally, we adopted **Apollo + Boss type**



Lane following scenario : BOSS type

- It's clear and convenient to develop as OSS

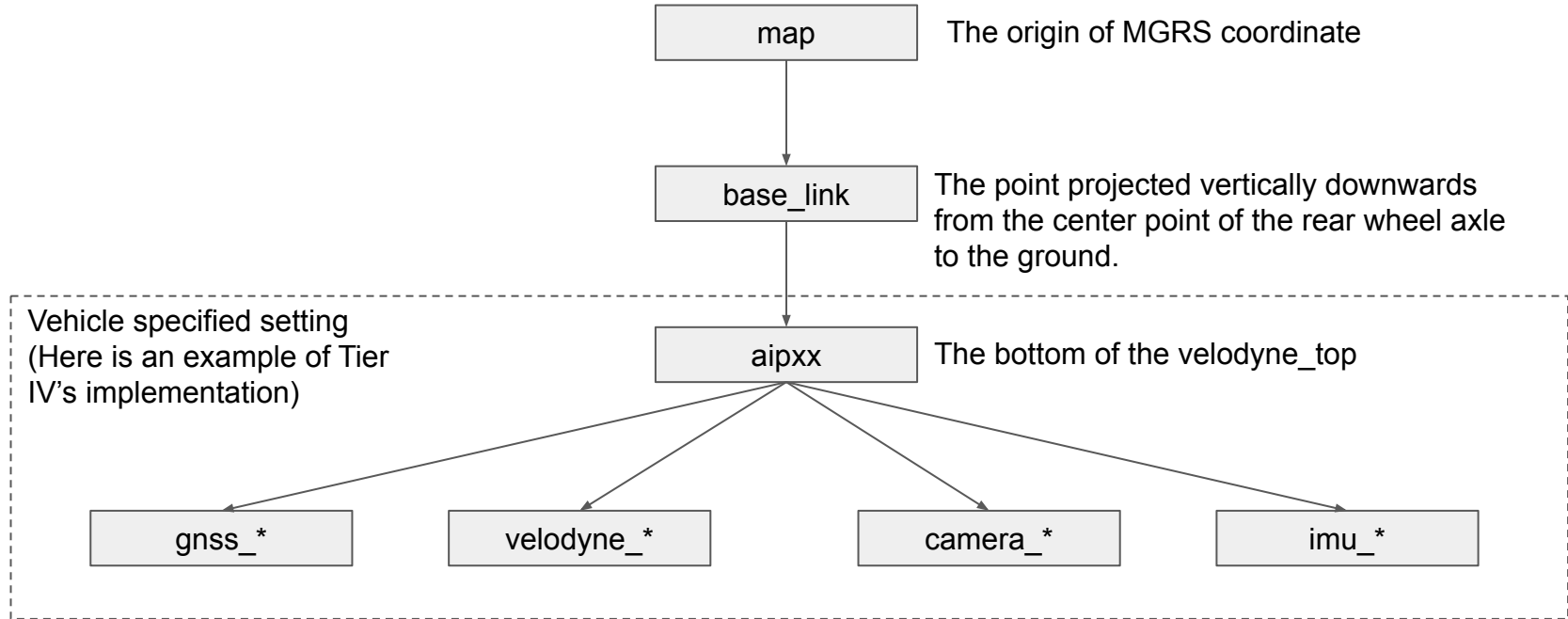
Whole module : Apollo type

- New scenario can be added as needed

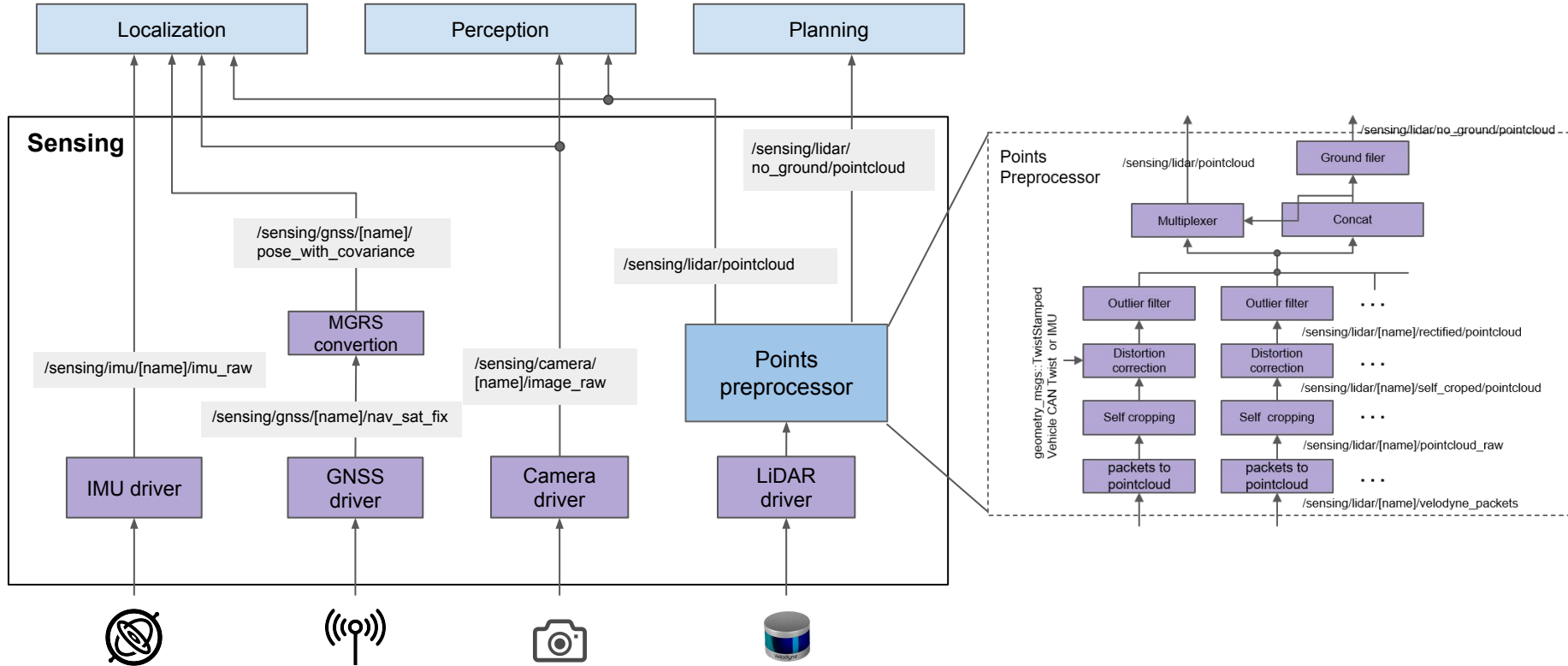
Future challenge

- In lane following scenario, behavior still makes a conservative decision. And an aggressive driving like changing lane by entering into crowded lane cannot be executed.

- Design alternative
- **Module implementation**
 - Implementation details
 - Achievements

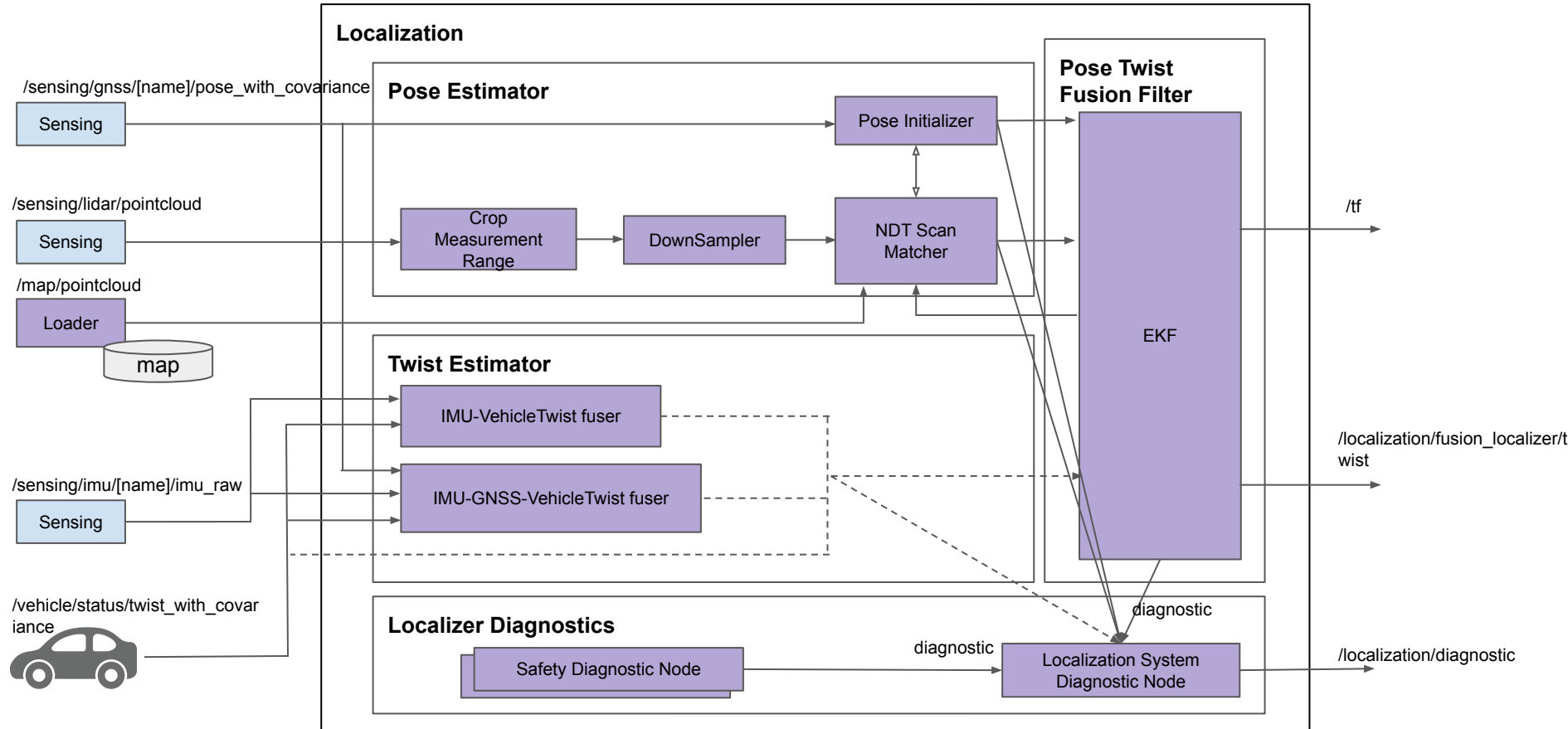


[Sensing] Components

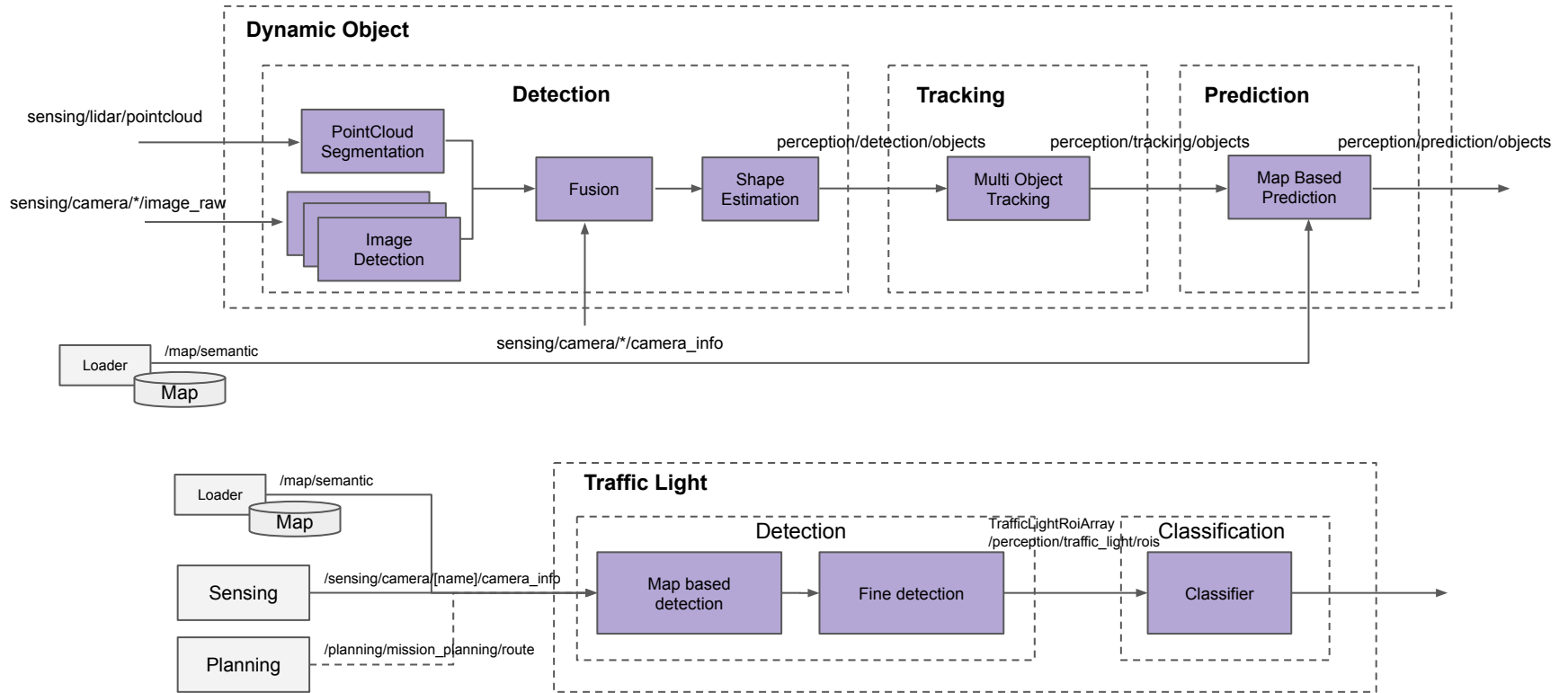


- **Distortion correction**
 - Corrects a distortion of the pointclouds due to an observation time gap
 - ⇒ Accuracy of the localization is improved
- **Ground filter**
 - Removes a ground from pointclouds (.ai has same feature)
- **Outlier filter**
 - Removes outliers by ring-base method ⇒ Reduce impacts of leaves and insects
- **Concat filter**
 - Integrates some pointclouds
 - Reduces an impact when a part of sensors stop
 - Corrects time gaps between each LiDAR with odometry ⇒ Accuracy of the localization is improved

[Localization] Components



- **Pose initializer**
 - Automatic initial self position estimation by GNSS + Monte-Carlo method
 - **NDT scan matcher**
 - Uses an estimated value of EKF as an initial position of the scan matching ⇒ If scan matching was failed, localization can be returned
 - Performance and accuracy are improved (Open-MP implementation, accuracy improvement of the initial position, improvement of the gradient method, distortion correction of pointclouds, and etc.)
 - **Scan matching failure judgement**
 - Monitors statuses of scan matching based on a score
 - If score is lower than a threshold, an estimated result isn't output
 - **EKF localization**
 - Integrates the estimated self position of the scan matching and the velocity of CAN + IMU
 - If scan matching broke down, the vehicle can drive a certain distance with odometry only
 - **IMU Vehicle-twist fusion**
 - Uses a translation velocity of CAN and yaw rate of IMU ⇒ Accuracy of odometry is improved
 - **Localizer Diagnostics**
 - Monitors a status of a whole localization module (Temporary implementation)
- ※ Only NDT is implemented in Pose Estimator now.
In future, other estimators like a white line recognition based one will be added.

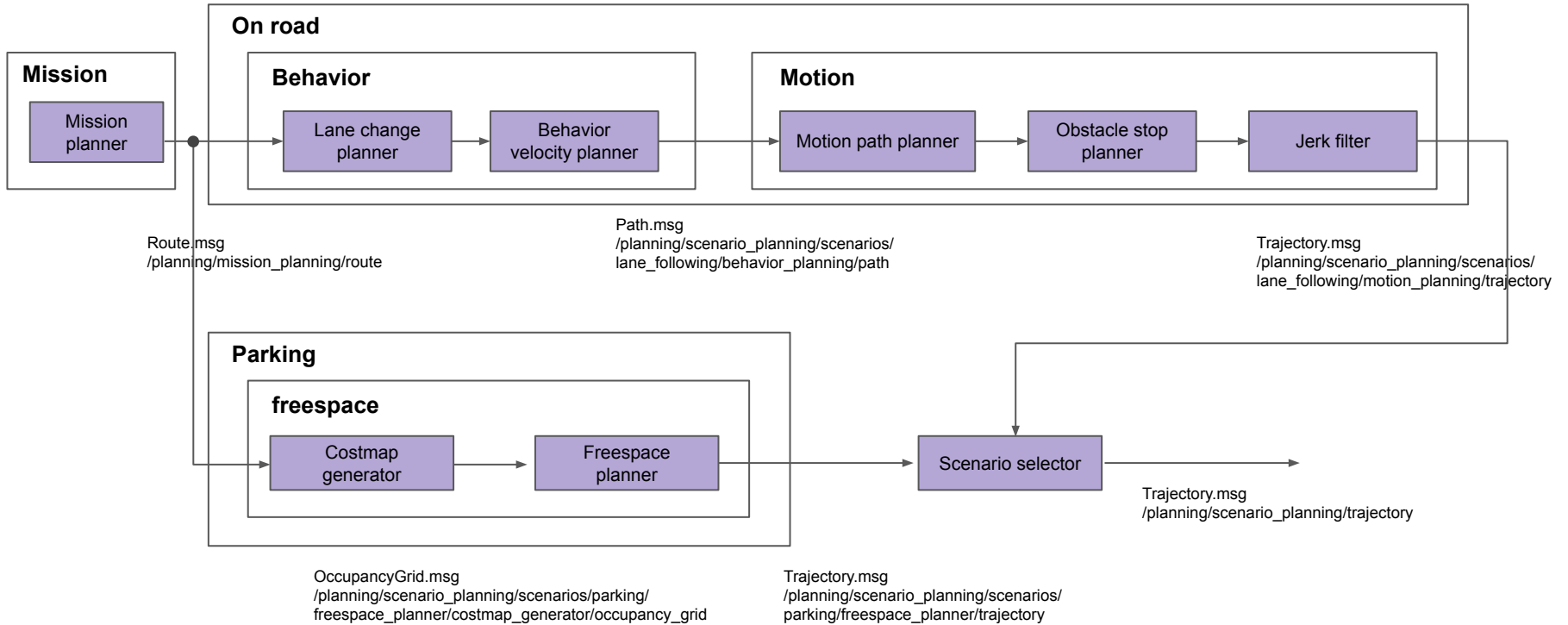


Dynamic Object

- **Point Cloud Segmentation** : Splits pointclouds into object clusters
 - Becomes faster by algorithm improvement (It can work in real-time on CPU)
- **Image Detection** : Detects objects on an image as ROI
 - Becomes to work on 40FPS on Jetson AGX by Tensor RT and int8 quantization
- **Fusion** : Matches the result of Pointclouds Segmentation and the ROI of the image detection result
 - Matching accuracy is improved by using IoU
- **Shape Estimation** : Geometrically approximates the whole size of the objects by Point Cloud Segmentation
 - Shape estimations by each object class
 - Accuracy improvement by changing fitting algorithm of Bounding Box
- **Multi-Object Tracking** : Assigns IDs based on time series data, estimates the velocity and the acceleration, and removes outliers
 - Performance improvement by changing a tracking model by class labeling
 - Data association considering the class label and the size
- **Map Based Prediction** : Predicts moving path of the objects with the lane information in a map
 - Infers objects' intent of the behavior and estimates each reliability of the predicted moving paths.

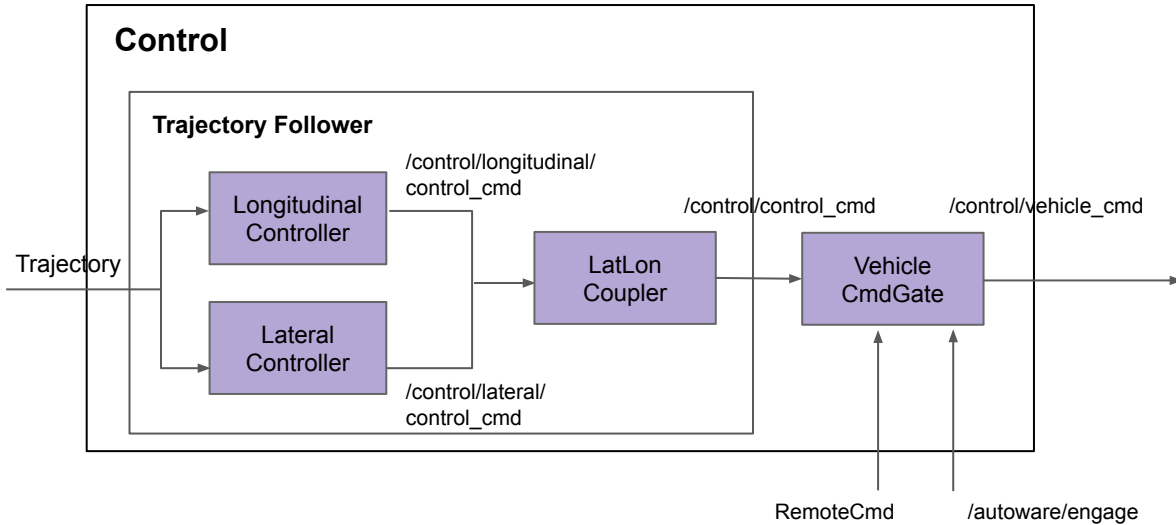
Traffic Light

- **Map Based Detection** : Extracts an ROI of the traffic light in camera image based on the self position and map data
 - Takes errors of a self position, a calibration and a hardware vibration into consideration
- **Fine Detection**
 - Extracts the ROI of the traffic light more precisely with a learner
- **Classifier** : Recognizes a state of the traffic light by a color information in the image
 - Reduces detection errors by fine noise removal process



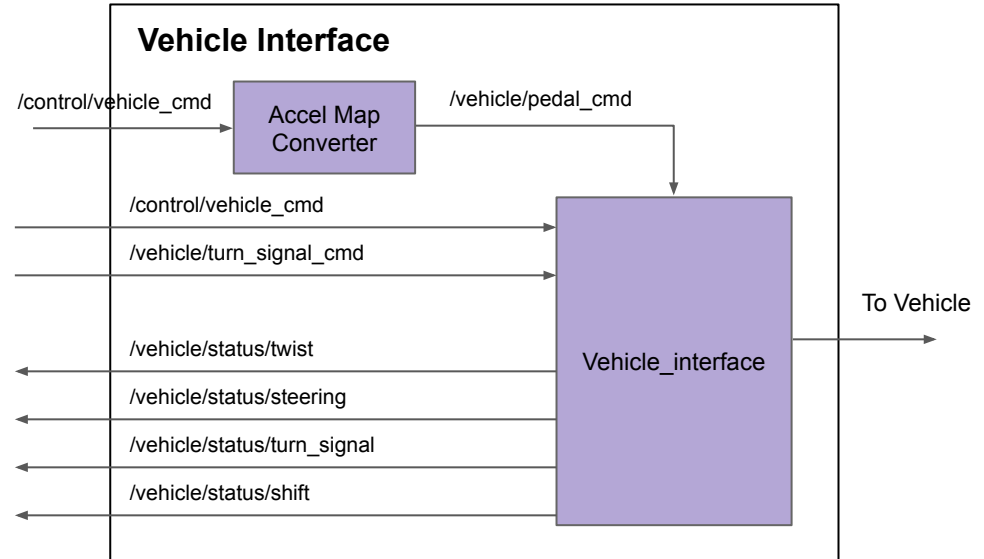
- **Scenario selector**
 - Chooses and executes “lane_following” or “parking” scenario according to the information of the mission
- **Mission planner**
 - Automatically searches the route from a self position to a goal via certain pass through points (by using lanelet function)
 - Automatically generates a route by using a map information
- **Lane change**
 - Judges a situation which needs lane change (Drive route limitation / street parking avoidance)
 - Judges whether a collision with a dynamic will happen obstacle and if a lane change is possible, it will be executed
- **Behavior Velocity Planner**
 - Plans a velocity based on traffic rules
 - Supports intersections, crosswalks, back-side check when turning right/left, stop with traffic lights and temporary stop line
 - Uses results of the perception (dynamic object information)
- **Obstacle avoidance (motion path planner)**
 - Plans a path to avoid the obstacle while driving ($A^* + QP$)
 - Generate a smooth path considering a drivable area and the obstacle
- **Obstacle stop (obstacle stop planner)**
 - Judges whether a collision with obstacle pointclouds considering the vehicle shape, and fill a stop velocity
- **Jerk filter**
 - Smoothens the velocity under the limitation of max speed, acceleration and lateral acceleration (QP)
 - Enables to switch rapid/slow acceleration and deceleration plans

- **Longitudinal Controller**
 - Calculates a target velocity and a target acceleration by PID
 - Supports a gradient correction and start/stop at a slope
 - Supports a smoothly stop
- **Lateral Controller**
 - Calculates a steering angle and an angular velocity (pure pursuit or MPC)
- **LatLon Coupler**
 - Integrates the longitudinal and lateral control command values
- **Vehicle Cmd Gate**
 - Switches the some command values like “remote” and “auto”
 - Attaches a limitation for the control command
 - longitudinal/lateral acceleration, longitudinal/lateral jerk



- In this implementation, longitudinal and lateral controls are separately calculated
- In future implementation, these might be calculated collectively

- **Accel Map Converter**
 - Converts a target acceleration to a vehicle specific accel/brake pedal value by Accel Map
 - If the vehicle_interface supports a velocity/acceleration command, this converter isn't necessary
- **Vehicle Interface (vehicle specific)**
 - Interface between Autoware and a vehicle
 - Communicates control signals
 - Obtains the vehicle information



- **Web Controller**
 - Through this controller, we can engage and specify the max velocity and also use “Go Home” button, “lane change OK” button and sensor Hz monitor
- **Autoware State Monitor**
 - State monitoring (Initializing / WaitForEngage / Driving / ...)

- Design alternative
- **Module implementation**
 - Implementation details
 - **Achevements**

- Automatic initial self pose estimation
- Robustness of self pose estimation improvement
 - CAN / imu fusion, EKF feedback, Gradient method algorithm improvement, pointclouds distortion correction, estimation speed improvement by Open-MP implementation
- 3D objects class recognition and shape estimation
- Performance improvement of dynamic objects tracking
- Moving path prediction of dynamic objects by using map information
- 360-degree sensing by the camera-LiDAR fusion
- Automatic lane change planning & decision
- Object avoidance (with lane change/ in the same lane)
- Intersection support (consider the oncoming vehicles and crossing vehicles)
- Person recognition at pedestrian crossing

- Performance improvement of the traffic light recognition
- Blind spot check when turning right/left
- Route generation from current position to a destination (via specified passing points)
 - With this feature, “Go Home button” can be implemented
- Automatic parking
- Slope support
- Slow / rapid brake planning support
- Performance improvement of the vehicle control
- Automatic deceleration at curve
- Collision judgement considering a vehicle shape