

An aerial night view of a city, showing a complex network of roads and highways illuminated with warm orange and yellow lights. The city buildings are scattered across the landscape, with some taller structures standing out. The overall scene is dark, with the city lights providing the primary illumination.

# TIER IV

2023 / 07 / 12

Development about  
Occupancy Grid Map Fusion

# CONTENTS

01 / Problem Statement

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# Related Links

PRs(~2023/April)

- **Solution A:** move scan origin
  - [PR#2939](#) : enable to select gridmap origin frame
  - [PR#3032](#) : add scan frame option and fix scan
- **Solution B:** create ogm in each sensor
  - [PR#3032](#) : Separate scan\_frame and gridmap\_origin
  - [PR#3054](#) : Filter obstacle pointcloud by raw pointcloud
  - [PR#3312](#) : Publish time synced raw pointcloud from sensing component (WIP)
- **Solution B:** OGM Fusion Node
  - [PR#3058](#) : Refactor OGM launcher
    - [PR#3340](#) : Bug fix
  - [PR#3107](#) : Add grid map fusion node (WIP)

Links

- [Issue #2906](#)



# Problem Statement

01



# About OGM(Occupancy Grid Map)

OGM example in sample

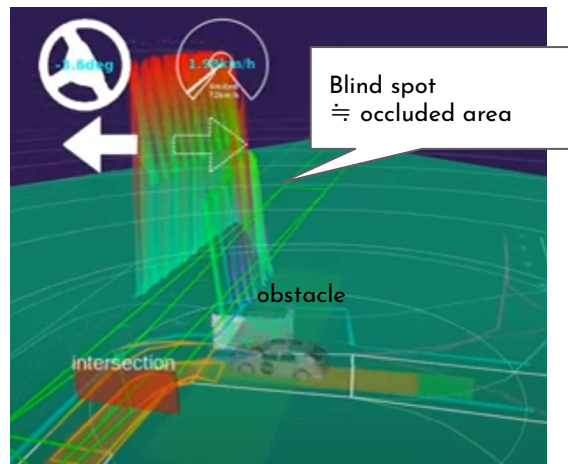
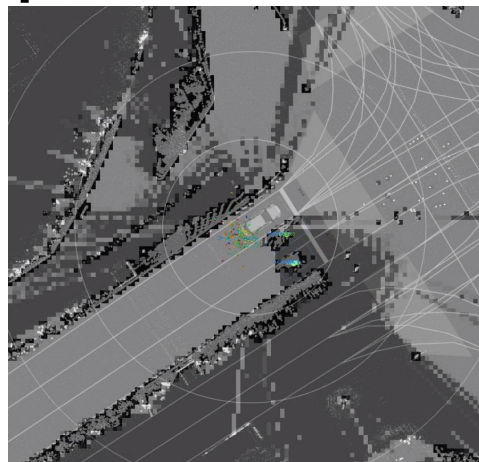
OGM is used to represent the presence of an obstacle in environment.

It is used in:

- Perception
  - object filtering
- Planning
  - intersection module
  - blind spot estimation...etc

OGM is created with

- raw lidar pointcloud
- obstacle lidar pointcloud



# (Appendix) OGM generation

Currently we only support two lidar based methods:

- laserscan based method
  - 🧑 only need obstacle pointcloud
  - 🧑 short range visibility (obstacle)
- Pointcloud based method
  - 🧑 more detailed
  - need raw and obstacle pointcloud

We recommend latter in many cases.

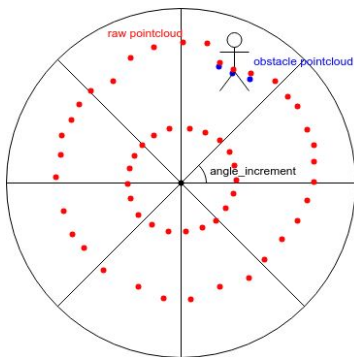
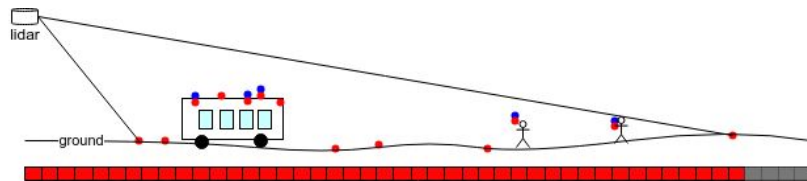
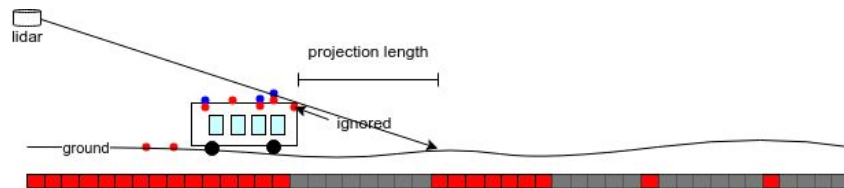


Fig. Pointcloud based OGM creation

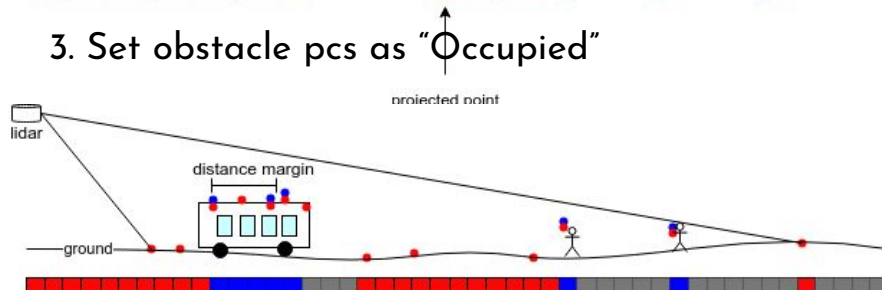
1. Fill visible range with "Free"



2. Set blind spot as "Unknown"



3. Set obstacle pcs as "Occupied"



# Problem statement

OGM is often scanned on `base_link` (rear side).

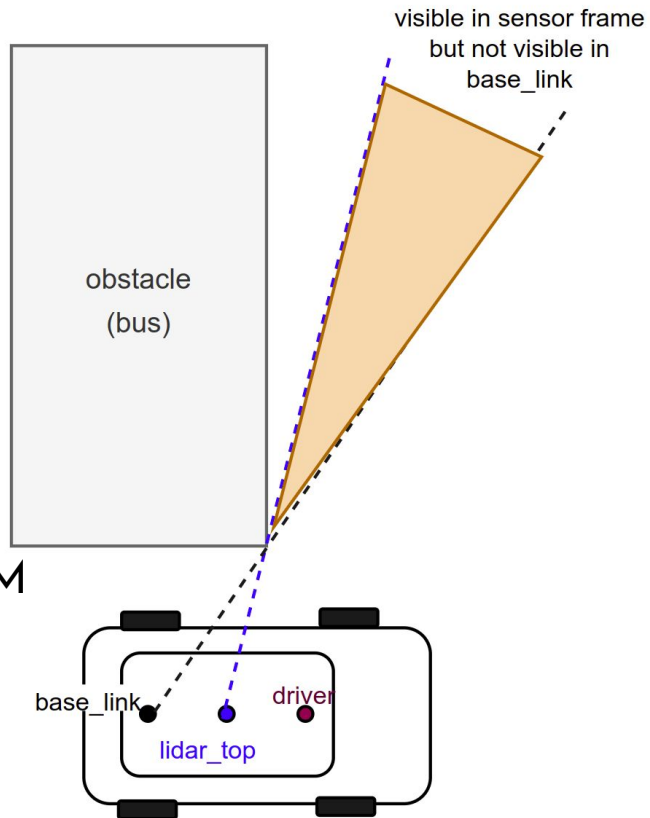
This sometimes result in **shorter range visibility in intersection**

It do not reflect true FOV of the each sensors because scanning origin is not on sensor

## Goals:

- Provide method to reflect true sensor FOV to OGM

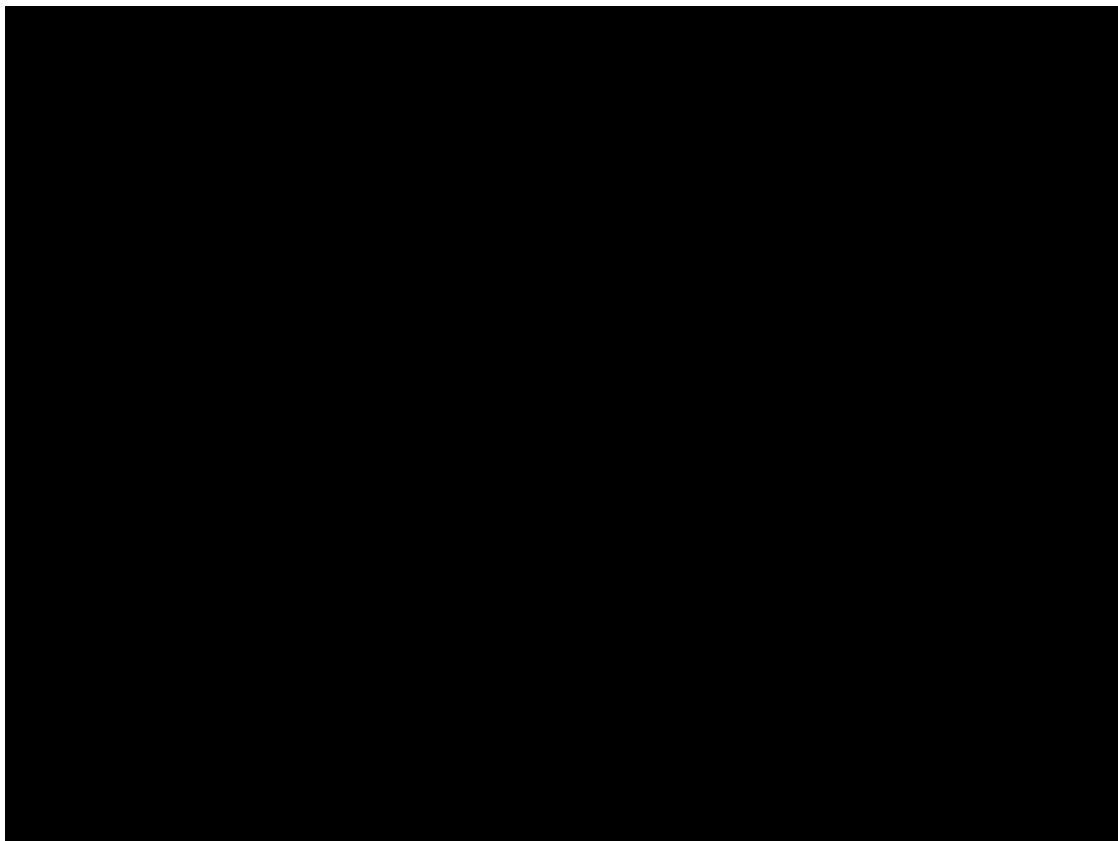
For details, please see [Issue](#)



# (Appendix): Problem statement in Video

Blind spot in intersection is displayed as area surrounded by walls.

Planning module tries to expose the vehicle head to see this blind spot.



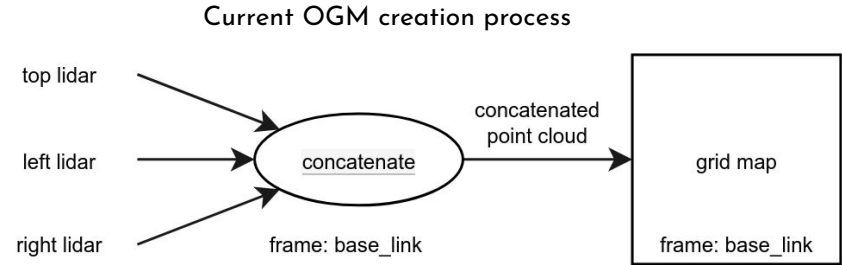


# Basic Idea and Solutions

Basic Idea:

The scan origin in OGM creation should be the same with the sensor origin.

- Solution A
  - Move scan origin
- Solution B
  - Create OGM in each sensor frame
  - Fuse multiple OGM output



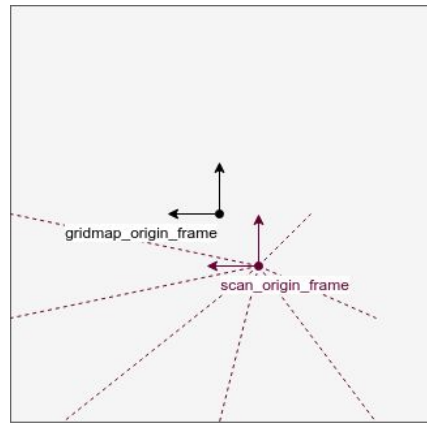
# Solution A: Move scan origin (Merged)

In the config/\*.yaml file you can set origin separately

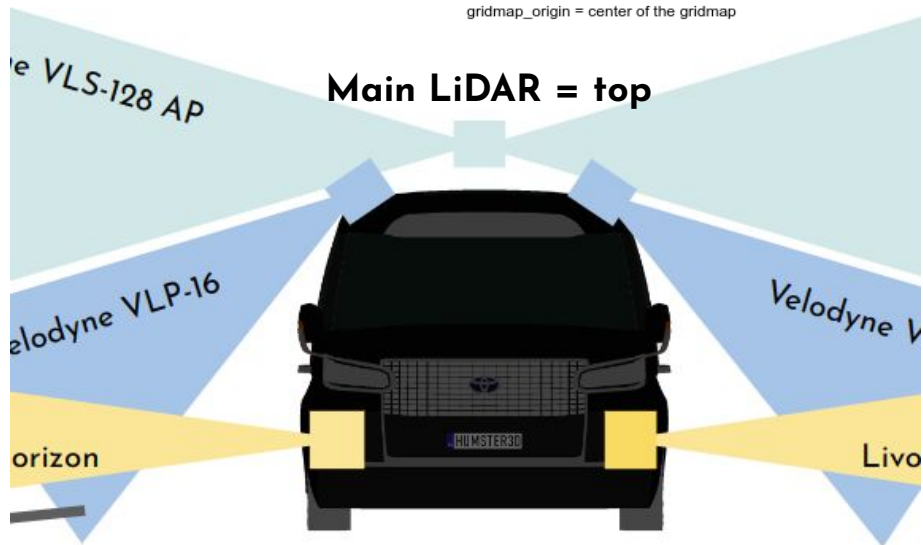
- `gridmap_origin_frame`
  - geometrical center of OGM
- `scan_origin_frame`
  - origin of virtual scan

See [README.md](#) for details.

This works well in small vehicle with one main-lidar and some sub-lidar builds.



gridmap\_origin = center of the gridmap

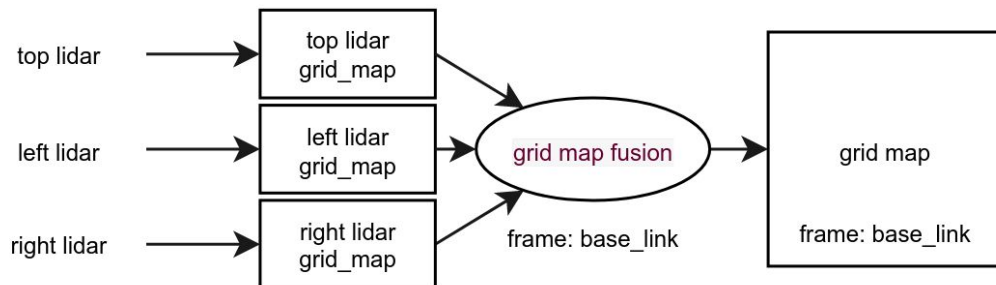


# Solution B: Create OGM in each sensor frame

In the larger vehicle, we can not choose appropriate origin.

We need to

- Create OGM in each sensor frame
- Fuse OGM



generate grid map in each sensor frame





# Detailed OGM Fusion Flow

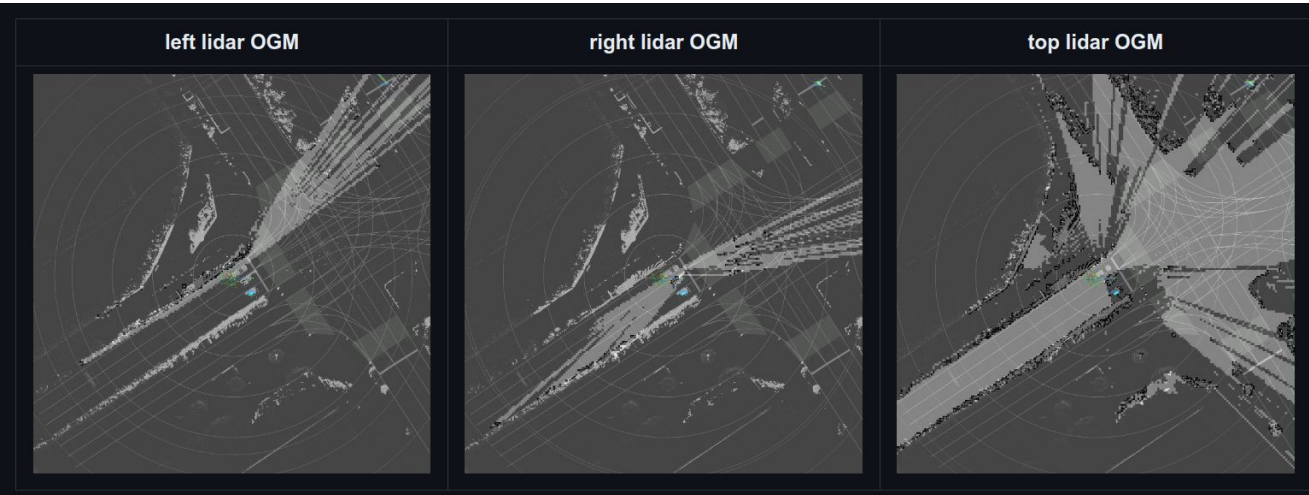
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02

# Solution B: Create OGM in each sensor frame

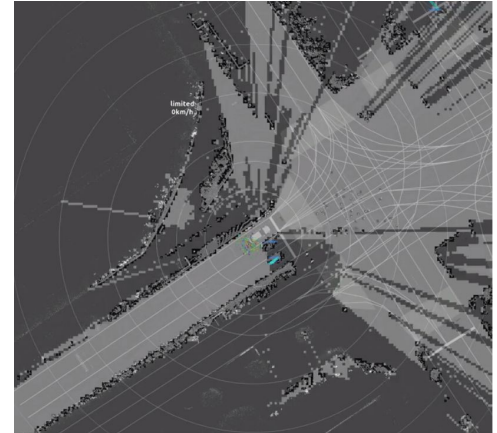
OGM in each sensor is created on same map\_origin and then fused.

See [document](#)(in PR). reference -> [[CARLOS GÁLVEZ 2015](#)]

Example in sample-rosbag in documentation



Fused OGM



# Create OGM in each sensor frame

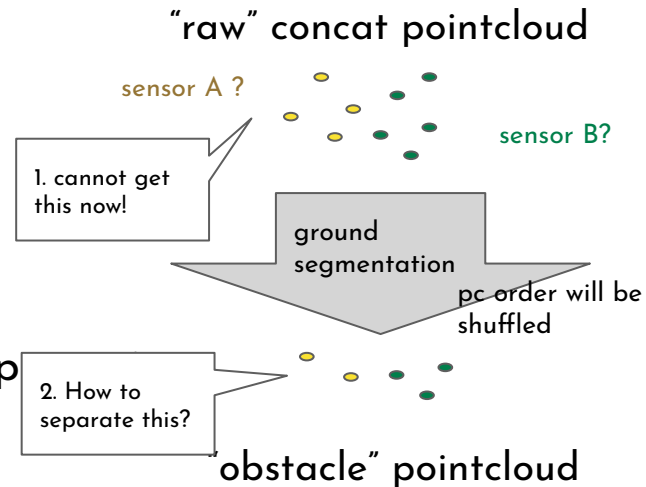
we need "raw" and "obstacle" pointcloud in each sensor

## 1. There are no "raw" pointcloud published

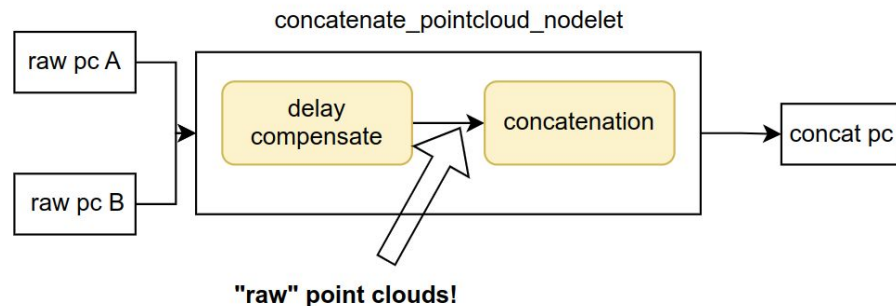
- point cloud concatenation includes delay comp
- need this PR

## 2. How to separate "obstacle" pointcloud

- Use "raw" pointcloud to filter concatenated obstacle pointcloud
- use `filter_obstacle_pointcloud_by_raw_pointcloud` option



Pointcloud concatenation process



# OGM fusion flow

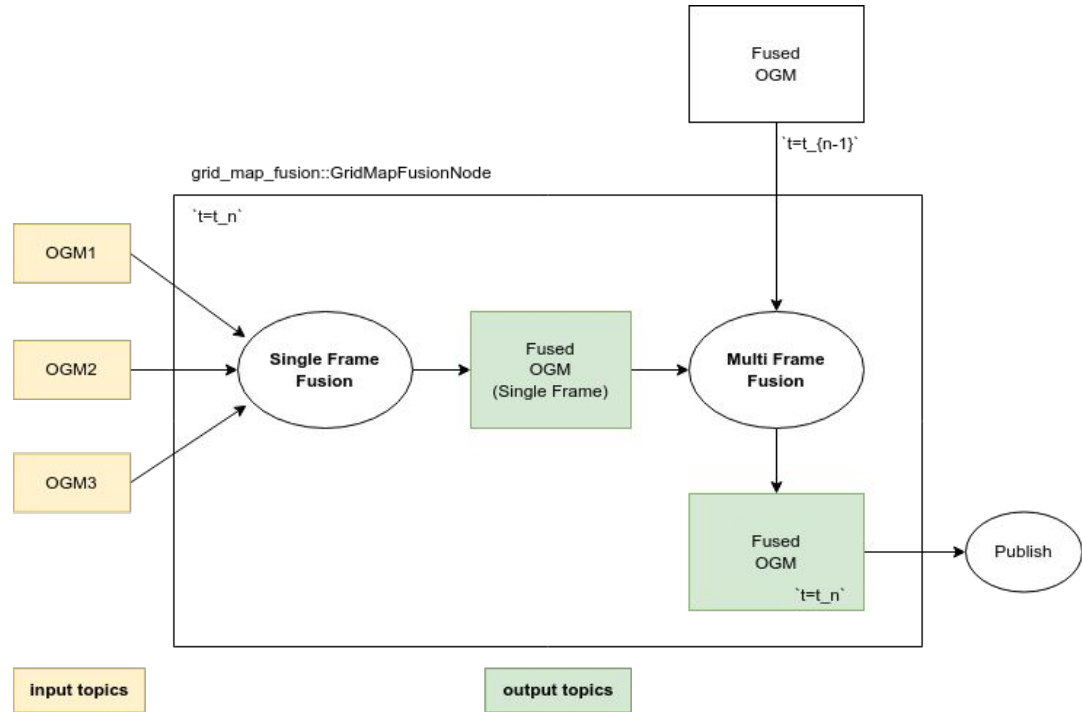
We have two choices:

## 1. synchronized fusion

- fuse in the certain frame
- 🧑🏻‍🔧 easier to debug
- 🧑🏻‍🔧 sensor output should be synchronized (for LiDAR)

## 2. async fusion

- Update fused OGM when message comes
- 🧑🏻‍🔧 no sync process
- better in multi modal OGM fusion
- not implemented yet... 🧑🏻‍🔧



# OGM fusion method [\[CARLOS GÁLVEZ 2015\]](#)

OGM fusion method requires:

- “Unknown” observations should not affect to Fusion.
- Able to handle each sensor weight
- Able to manage conflict
- Independent to input order

policy	description
<b>overwrite</b>	more critical state is overwritten: occupied -> free -> unknown
<b>log-odds</b>	most simple bayes rule
<b>dempster-shafer</b>	able to handle unknown information directly



# (Appendix) Log-Odds fusion

Log-Odds based method is one of the bayesian filter implementation

Basic equation is like:

$$l_t(m_{ij}) = \log \frac{p(m_{ij}|\mathbf{z}_{1:t}, \mathbf{x}_{1:t})}{1 - p(m_{ij}|\mathbf{z}_{1:t}, \mathbf{x}_{1:t})} = \log \frac{p(m_{ij}|\mathbf{z}_t, \mathbf{x}_t)}{1 - p(m_{ij}|\mathbf{z}_t, \mathbf{x}_t)} + l_{t-1}(m_{ij}) - \log \frac{p(m_{ij})}{1 - p(m_{ij})}$$

$$l(\mathbf{m}) = \log \frac{p(\mathbf{m})}{1 - p(\mathbf{m})} \quad ; \quad l_k(\mathbf{m}) = \log \frac{p_k(\mathbf{m})}{1 - p_k(\mathbf{m})}$$

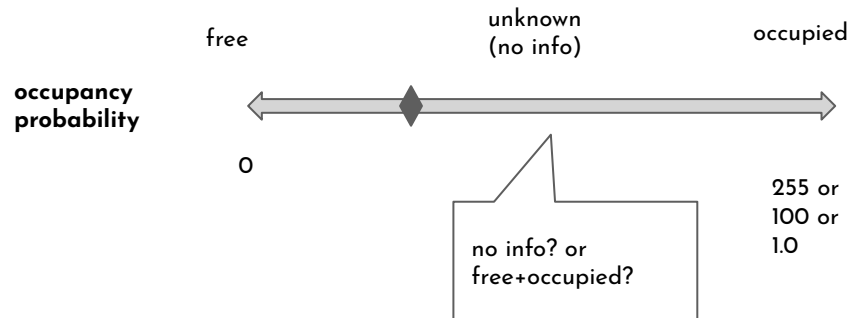
The fused log-likelihood then simply becomes:

$$l(\mathbf{m}) = \sum_k l_k(\mathbf{m})$$

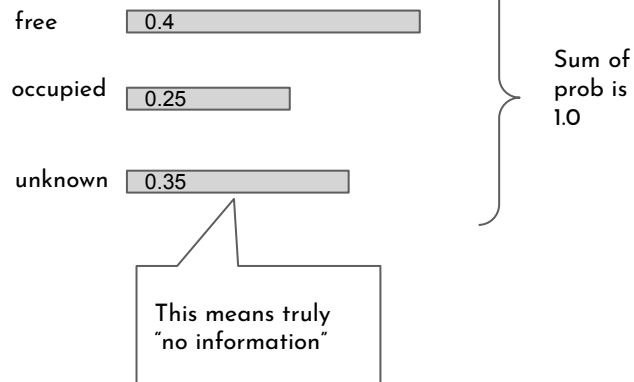
Actually, we use “weighted” log-odds

# (Appendix) Dempster Shafer Theory [\[TIER IV internal link\]](#)

- 🧑‍🎓 able to handle each state probability separately  
→ aware of difference in “no info” and “conflicted”
- 🧑‍🎓 can suppress unknown probability after fusion
- 🧑‍🎓 conflict sometimes lead to unintended result
- 🧑‍🎓 ros message limitation
  - output will be occupancy probability...



dempster shafer

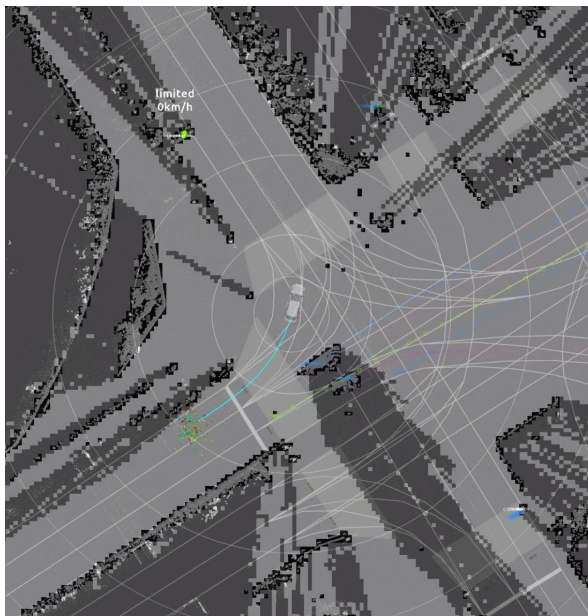


# Comparison: Single Frame Fusion

Each method does not differ so much. Because:

- **input ogm is binary-like:** free(0) or occupied(255)
- **overwrapped region is not large enough**

“overwrite” policy



“dempster-shafer” policy

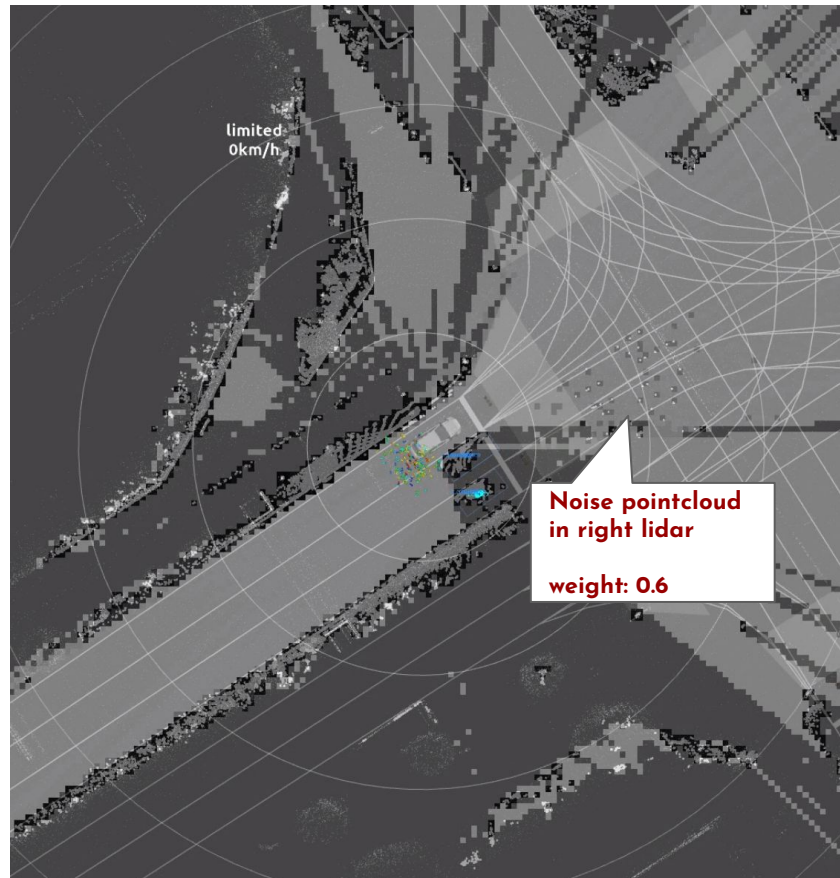


# (Appendix) use weights to suppress noise

Each sensor weight is another important parameter

We can suppress noise from less trustable sensor

```
each_ogm_output_topics:  
  - "/perception/occupancy_grid_map/top_lidar/map"  
  - "/perception/occupancy_grid_map/left_lidar/map"  
  - "/perception/occupancy_grid_map/right_lidar/map"  
each_ogm_sensor_frames:  
  - "velodyne_top"  
  - "velodyne_left"  
  - "velodyne_right"  
# reliability of each sensor (0.0 ~ 1.0) only work with "log  
each_ogm_reliabilities:  
  - 1.0  
  - 0.6  
  - 0.6
```



# Comparison: Processing time

Processing time:

`overwrite >> dempster-shafer ≐ log-odds`

Want faster processing

→ `overwrite`

Need sensor weighting

→ `log-odds, dempster-shafer`

Method	processing time [ms]
<b>overwrite</b>	mean: <b>4.162326</b> min: 3.000000 max: 7.546000
<b>log-odds</b>	mean: <b>8.159959</b> min: 5.859000 max: 13.731000
<b>Dempster-shafer</b>	mean: <b>7.489534</b> min: 5.049000 max: 16.239000

An aerial, high-angle photograph of a city at night, featuring a complex, multi-level highway interchange. The roads are illuminated by streetlights, and the movement of vehicles is captured as vibrant, multi-colored light trails in shades of orange, red, blue, and green. The surrounding urban landscape includes several tall, modern skyscrapers with lit windows, and smaller buildings with varied architectural styles. The overall scene is a dynamic and colorful representation of a bustling city at night.

Conclusion

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03

# Conclusion

- OGM virtual scan should be done in (main) sensor frame
- solution A
  - Move virtual scan origin (merged)
- solution B
  - publish raw pointcloud ([waiting for merge](#))
  - OGM fusion ([under construction](#))
    - move to new package -> ogm\_fusion or something?
    - fix conflicts with newer changes
    - (add asynchronous fusion node) (Evaluate in planning)
  - concerns:
    - increasing traffic, computational resources...  
(component container sometimes stops...)



# Appendix: Other TODO

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04



# Current costmap value settings

original settings in  
nav\_msgs/OccupancyGridMap messages:

data int8[]

-127~128

costmap in autoware S/P nodes:

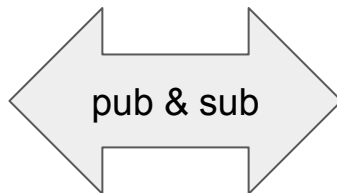
data char[]

0~255

0: Free

128: NO\_INFORMATION

255:Occupied



Existence Probability

messages in autoware

data char[]

1~99

1: Free

99:Occupied



Hard to explain sensor FOV by this existence probability

# Change costmap value settings

original settings in  
nav\_msgs/OccupancyGridMap messages:

data int8[]

-127~128

costmap in autoware S/P nodes:

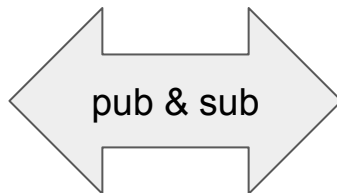
data char[]

0~255

0: Free

254: Occupied

255: NO\_INFORMATION



Existence Probability

messages in autoware

data char[]

-1, 0~100

-1: NO INFO

0: Free

100: Occupied

Add FOV information. cf. [navigation2](#)