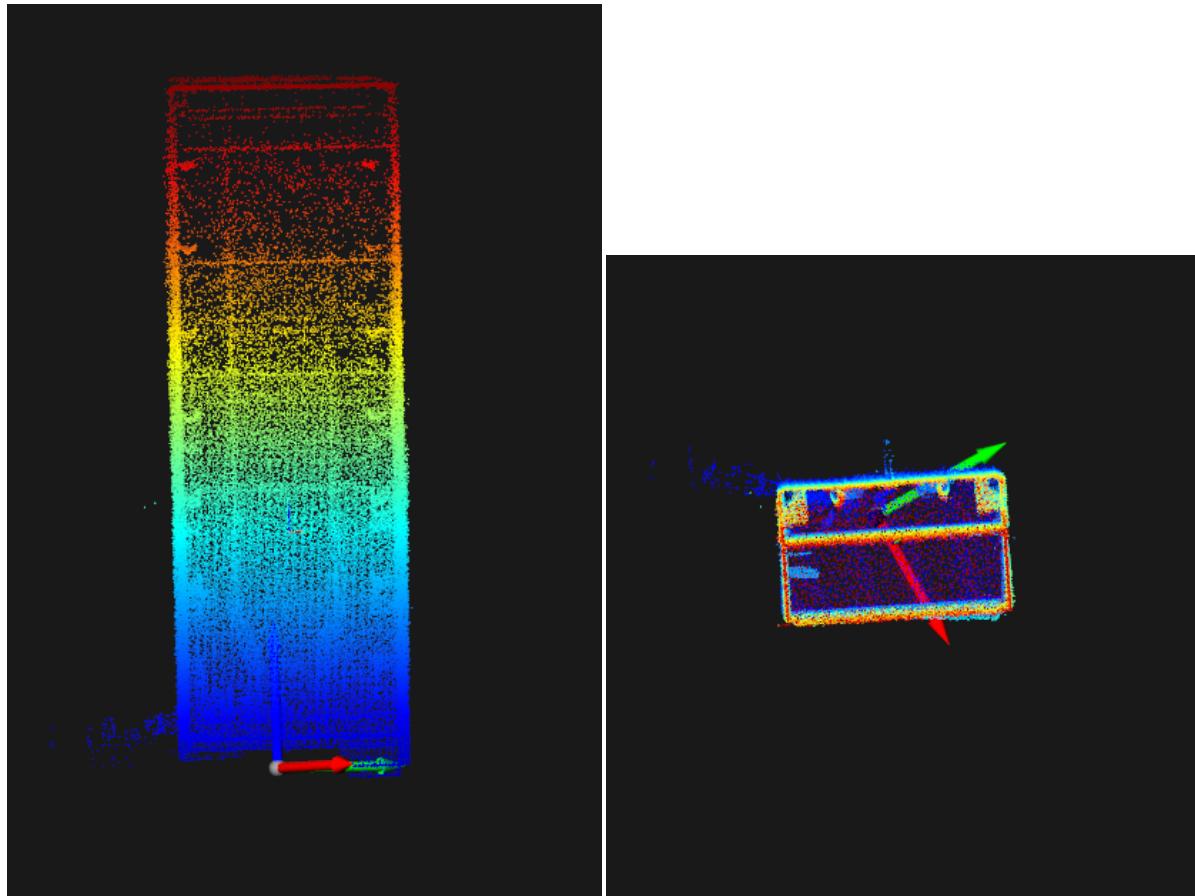


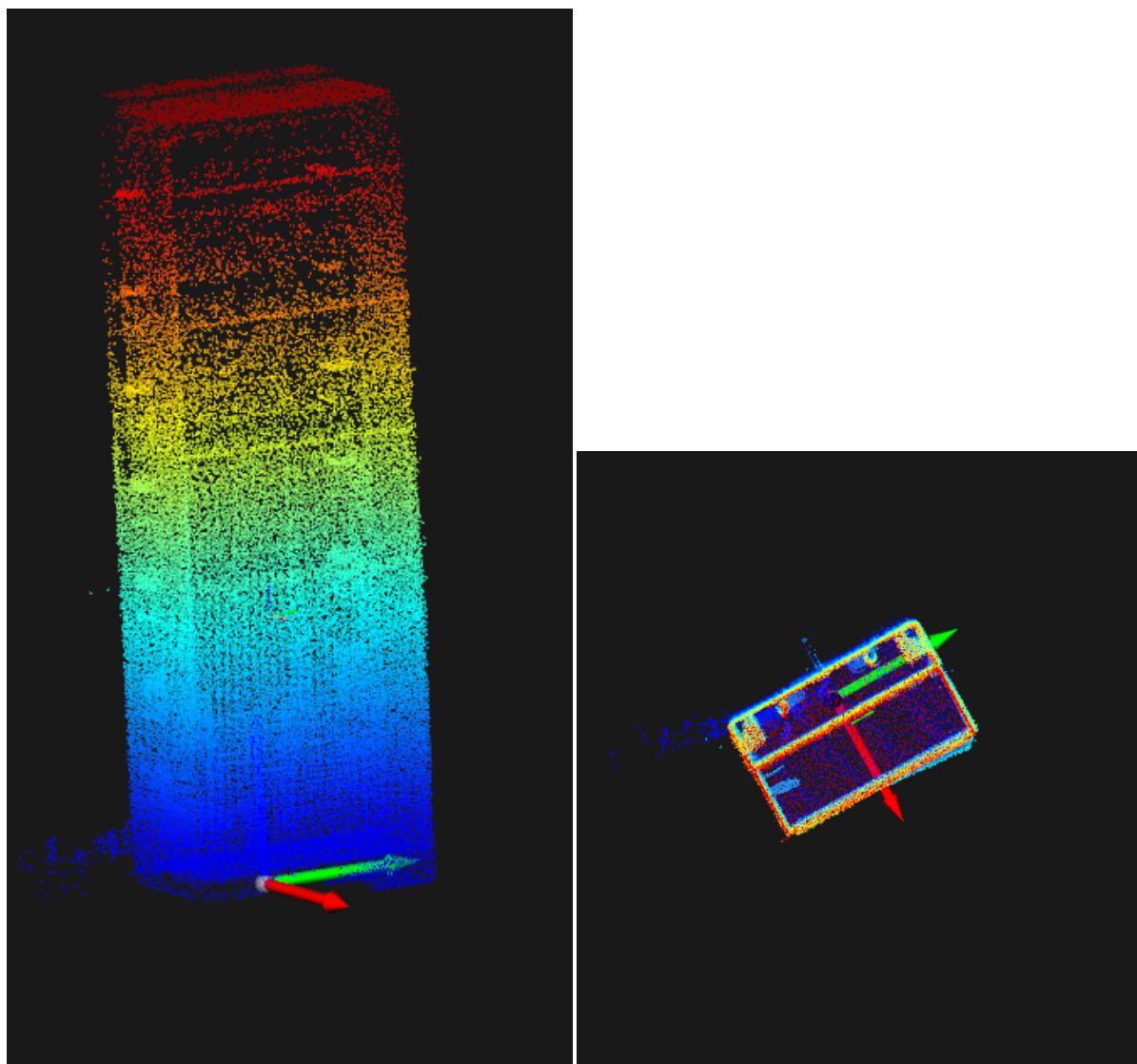
Localization at HRSG Manhole

Perception Pipeline

1. (offline) **COLLECT LiDAR and IMU data**, sweeping inside each manhole: field experiments
2. (offline) **BUILD MAP** and postprocess:
 1. identify the floor and level the map



2. manually rotate the map (yaw) so that axes are aligned (parallel) to the real world



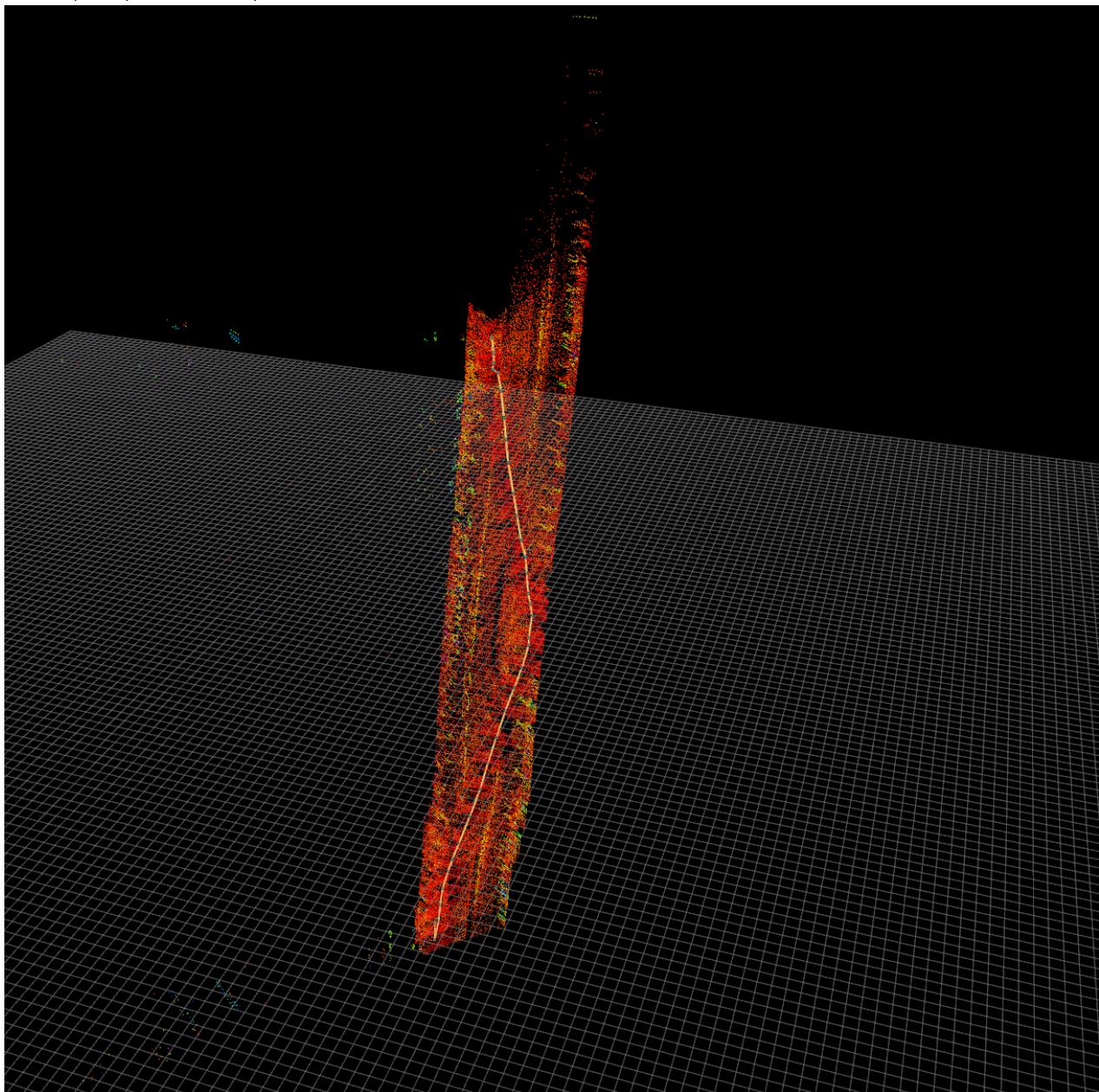
3. while building map, **construct NODE GRAPH**
 1. node: important (key) pose
 2. which involves LiDAR-inertial odometry (LIO) internally
4. (online) **LOCALIZE** drone
 1. initially by providing **INITIAL POSE**: where drone takes off (depends on the experiment), probably where **station** should be
 2. then consecutively:
 1. by identifying similar & revisited node (constructed offline, see 3)
 2. and applying LIO & fixed window smoothing (factor graph optimization)

Problems of HRSG Manholes

there were two problems:

1. while building map, internal LIO continuously failed
 - correspondence are hard to secured, due to the repeated feature
 - **correspondence**: where are the points (of current LiDAR scan) mapped to the built map?
 - basically **tunnel-like & featureless** structures are harsh environment for LIO

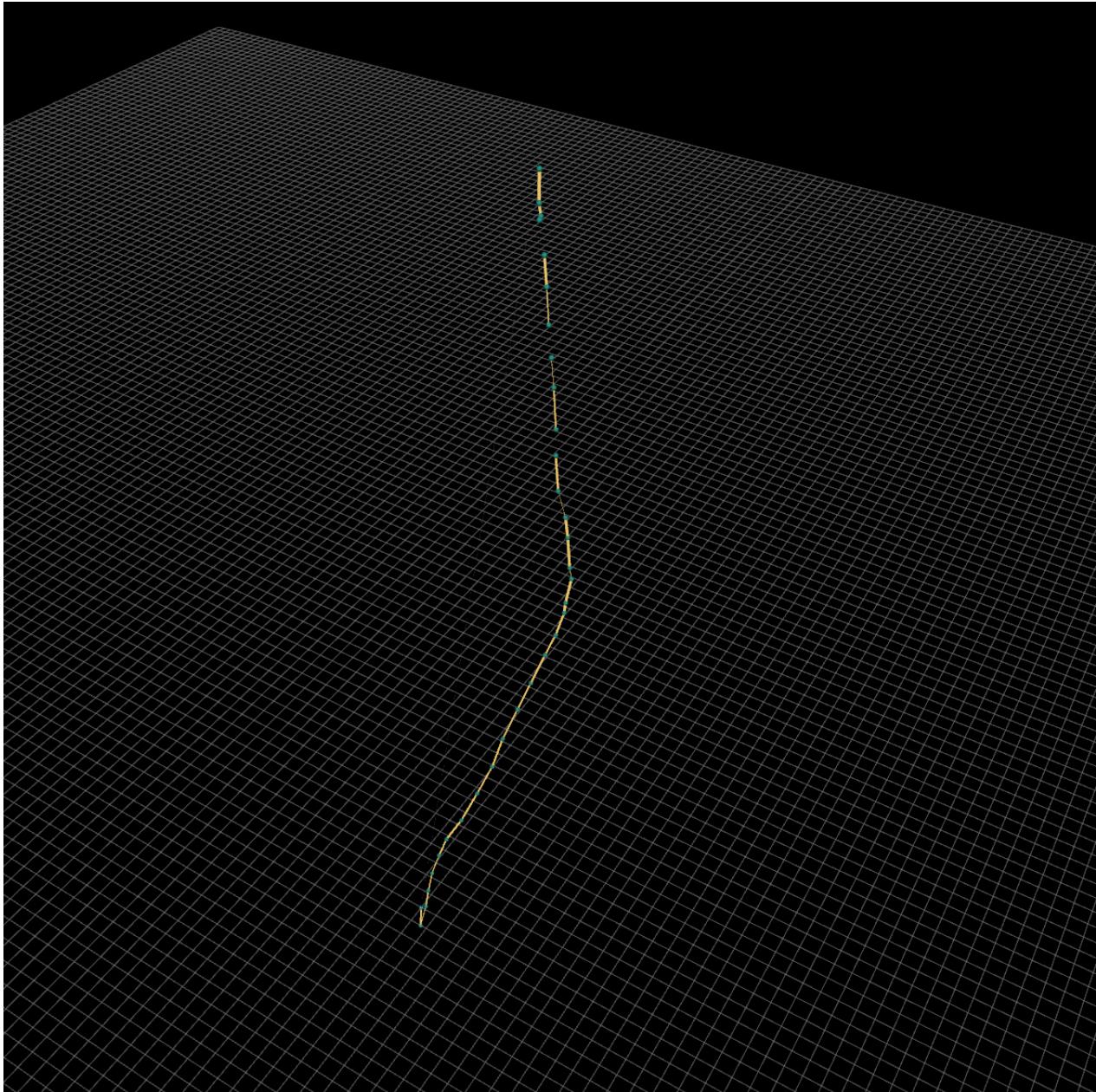
- an example of problematic map:



2. identifying node(s) and initial pose

- basically all nodes are similar
- the manhole structure was narrow and repeated, thus, there are many (multimodal) wrong localization solutions

- an example of node graph under ill-conditioned situation (where no vertical movements were involved in real world):



Solution Approaches

respective approach per problem is:

1. relax the correspondence checking condition

- there are basically tons of parameters:

```
/*
ros_parameters:
  bag_file: "" # absolute path of bagfile
  odometry_type: "fast_lio"
  keyframe_detection:
    max_range: 40.0 # maximum detection range
    max_time_diff: 60.0 # maximum time difference to maintain the latest keyframe
    motion_threshold: 5.0 # motion triggered keyframe generation
  loop_detection:
    nearby_frames: 3 # nearby keyframes to loop registration
    search_radius: 1.5 # search radius for loop candidate
    min_time_diff: 30.0 # minimum time difference to be a loop pair
  registration:
    resolution: 0.125 # loop registration sampling resolution
    max_iterations: 20 # maximum iteration
    k_corrs: 10 # nearest k
```

```

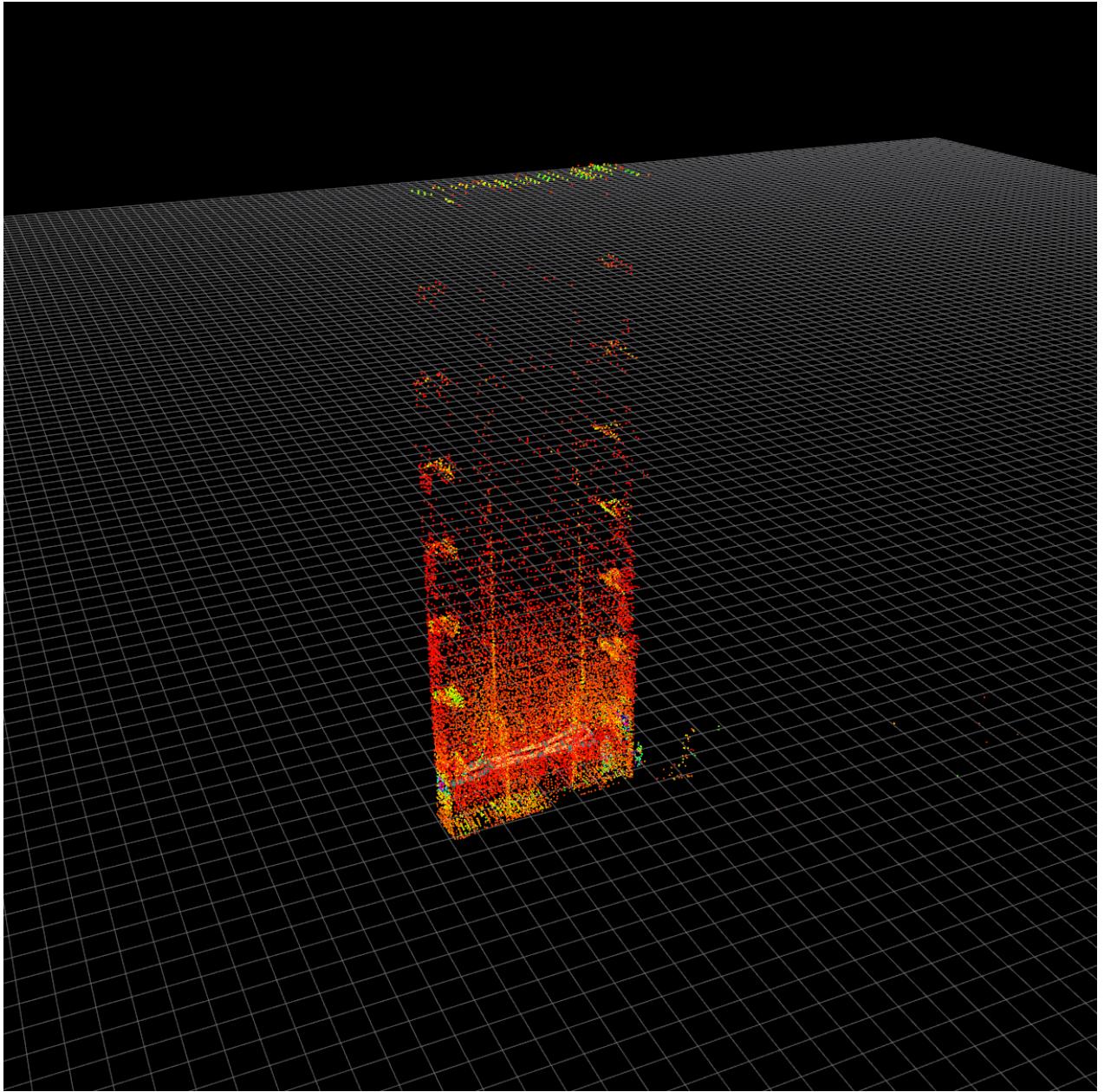
    kernel_sigma: 1.0 # m-estimator constant
    convergence_epsilon: 0.001 # convergence criterion
graph_optimization:
    max_iterations: 10 # maximum iteration
    abs_error_epsilon: 0.000001 # absolute residual epsilon
    rel_error_epsilon: 0.000001 # relative residual epsilon
    prior_orientation_noise_sigma: 0.000001 # rotational noise sigma for prior factor
    prior_position_noise_sigma: 0.000001 # translational noise sigma for prior factor
    odom_rotation_noise_sigma: 0.001 # rotational noise sigma for odometry factor
    odom_translation_noise_sigma: 0.01 # translational noise sigma for odometry factor
    loop_rotation_noise_sigma: 0.003 # rotational noise sigma for loop factor
    loop_translation_noise_sigma: 0.03 # translational noise sigma for loop factor

base_frame: "ids" # body frame name
map_size: 100.0 # edge length of local map
init_period: 0.5 # system initialization period
gravity_scale: 9.81 # scale of gravity vector
base_to_lidar: # base-to-lidar extrinsic params (for visualization)
    rotation: [1.0, 0.0, 0.0, 0.0, 1.0, 0.0, 0.0, 0.0, 1.0]
    translation: [0.0, 0.0, 0.0]
imu_to_lidar: # imu-to-lidar extrinsic params
    rotation: [1.0, 0.0, 0.0, 0.0, 1.0, 0.0, 0.0, 0.0, 1.0]
    translation: [0.0, 0.0, 0.0]
    fixed: true
preprocess:
    skip_stride: 2 # resampling stride
    resolution: 0.125 # downsampling resolution
    range:
        min: 0.5 # nearest range
        max: 40.0 # farthest range
registration:
    use_feature: true # use only planar neighbor points
    plane_threshold: 0.05 # planarity threshold
    k_corrs: 10 # nearest k to search neighbor points
    max_corr_dist: 1.0 # maximum correspondency distance
    min_corrs: 100 # minimum correspondency number
ikd_tree:
    resolution: 0.25 # downsampling resolution
    alpha_deletion: 0.5
    alpha_balance: 0.7
    min_size_rebuild: 10 # minimum subtree size to rebuild
    min_size_async: 1500 # minimum subtree size to async worker
ivox_map:
    resolution: 0.25 # downsampling resolution
    nearby_search: 4 # {1, EXACT}, {2, Nearby7}, {3, Nearby19}, {4, Nearby27}
    max_points_per_voxel: 5 # maximum points in single voxel
    max_decay: 60.0 # maximum decay time in seconds
ikf:
    max_iterations: 4 # maximum iteration
    convergence_epsilon: 0.005 # convergence criterion
    lidar_noise_sigma: 0.03 # sigma of point noise distribution
    imu_noise_sigma: [0.01, 0.03, 0.0001, 0.0001] # [gyro, accel, bias_gyro, bias_accel]

```

- by adjusting these parameters **PER-MANHOLE**:`loop_detection::registration::k_corrs, preprocess::resolution, registration::k_corrs, registration::resolution, ikd_tree::resolution`, successful maps were obtained

- an example for manhole 6 (with node graph) would be:



- yet, no consistent parameters were found that universally apply for all manholes: different manhole should use different set of parameters

2. fine-tune the parameters related to LIO and (re)localization

- however, still not solved: unstable solution
- a different set of parameters set is used here:

```

◦ map_path: "/home/junwoo/ids_mmu_workspace/src/perception/navigation/assets/mh6_rev"
keyframe_detection:
    max_range: 30.0          # maximum detection range
    max_time_diff: 30.0       # maximum time difference to maintain the latest keyframe
    motion_threshold: 5.0     # motion triggered keyframe generation
relocalization:
    period: 10.0
    search_radius: 5.0
registration:
    num_threads: 1
    resolution: 0.25          # loop registration sampling resolution
    max_iterations: 64         # maximum iteration
    k_corrs: 5                # nearest k
    kernel_sigma: 1.0          # m-estimator constant
    convergence_epsilon: 0.001 # convergence criterion
fixed_lag_smoker:

```

```

window_size: 30.0          # smoother lag in seconds
prior_orientation_noise_sigma: 0.01 # rotational noise sigma for prior factor
prior_position_noise_sigma: 0.1      # translational noise sigma for prior factor
odom_rotation_noise_sigma: 0.001     # rotational noise sigma for odom factor
odom_translation_noise_sigma: 0.01   # translational noise sigma for odom factor
relo_rotation_noise_sigma: 0.003     # rotational noise sigma for relo factor
relo_translation_noise_sigma: 0.03   # translational noise sigma for relo factor
lidar_topic: "/livox/lidar" # point cloud topic
imu_topic: "/livox/imu"      # imu topic
base_frame: "ids"           # base frame
num_threads: 5               # maximum number of threads for tbb
map_size: 100.0              # edge length of local map
init_period: 0.5              # system initialization period
gravity_scale: 9.81            # scale of gravity vector
base_to_lidar:                # base-to-lidar extrinsic params (for visualization)
    rotation: [ 1.0, 0.0, 0.0,
                 0.0, 1.0, 0.0,
                 0.0, 0.0, 1.0 ]
    translation: [ 0.0, 0.0, 0.0 ]
imu_to_lidar:                  # imu-to-lidar extrinsic params
    rotation: [ 1.0, 0.0, 0.0,
                 0.0, 1.0, 0.0,
                 0.0, 0.0, 1.0 ]
    translation: [ 0.0110, 0.02329, -0.04412 ]
    fixed: true
preprocess:
    skip_stride: 1             # resampling stride
    resolution: 0.125           # downsampling resolution
range:
    min: 0.5                   # nearest range
    max: 40.0                  # farthest range
registration:
    use_feature: false          # use only planar neighbor points
    plane_threshold: 0.05        # planarity threshold
    k_corrs: 5                  # nearest k to search neighbor points
    max_corr_dist: 10.0          # maximum correspondency distance
    min_corrs: 100               # minimum correspondency number
ikd_tree:
    resolution: 0.25            # downsampling resolution
    alpha_deletion: 0.5
    alpha_balance: 0.7
    min_size_rebuild: 10          # minimum subtree size to rebuild
    min_size_async: 1500          # minimum subtree size to async worker
ivox_map:
    resolution: 0.5              # downsampling resolution
    nearby_search: 4              # {1, EXACT}, {2, Nearby7}, {3, Nearby19}, {4, Nearby27}
    max_points_per_voxel: 5        # maximum points in single voxel
    max_decay: 30.0                # maximum decay time in seconds
ikf:
    max_iterations: 4             # maximum iteration
    convergence_epsilon: 0.005     # convergence criterion
    lidar_noise_sigma: 0.05         # sigma of point noise distribution
    imu_noise_sigma: [ 0.03, 0.3, 0.001, 0.01 ] # [gyro, accel, bias_gyro, bias_accel]

```

- by adjusting these parameters **PER-MANHOLE::relocalization::resolution, relocalization::k_corrs, ikd_tree::resolution, ikf::imu_noise_sigma, ikf::lidar_noise_sigma**, **ONLY FEW** cases were able to correctly estimate poses.
- below are some examples of successful and failed cases (videos):
 - [successful](#), [failed1](#), [failed2](#)
 - note: initial jump was intentional, by feeding true initial pose (by operator)
- common feature of failure cases:
 - it seems to track correctly but out of sudden **the solution jumps** to other nodes that was accidentally more fitted to current scan
 - repeated and hard-to-distinct feature of manhole yields **unstable** pose solution
 - of course we can TRY FLYING drones**, but safety (stability of accurate solution) should be secured first