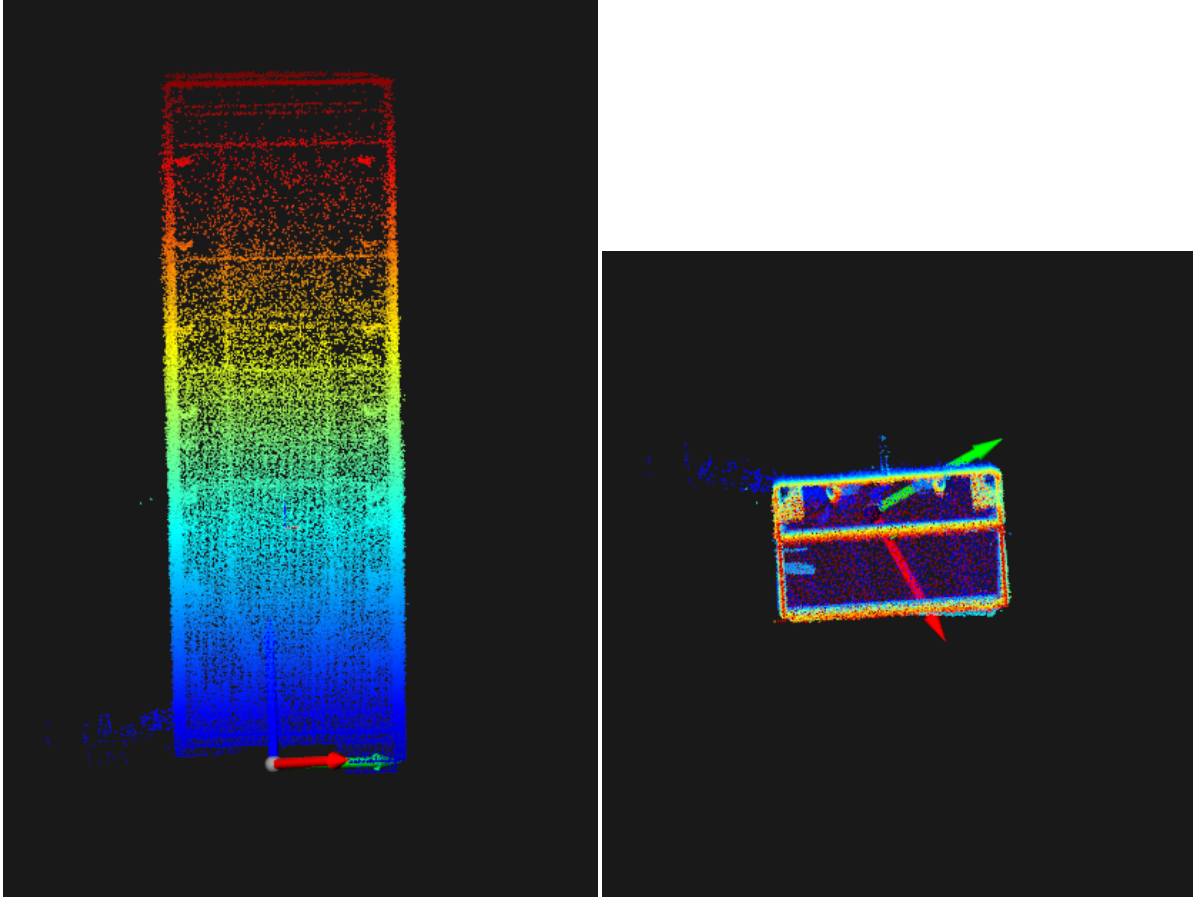


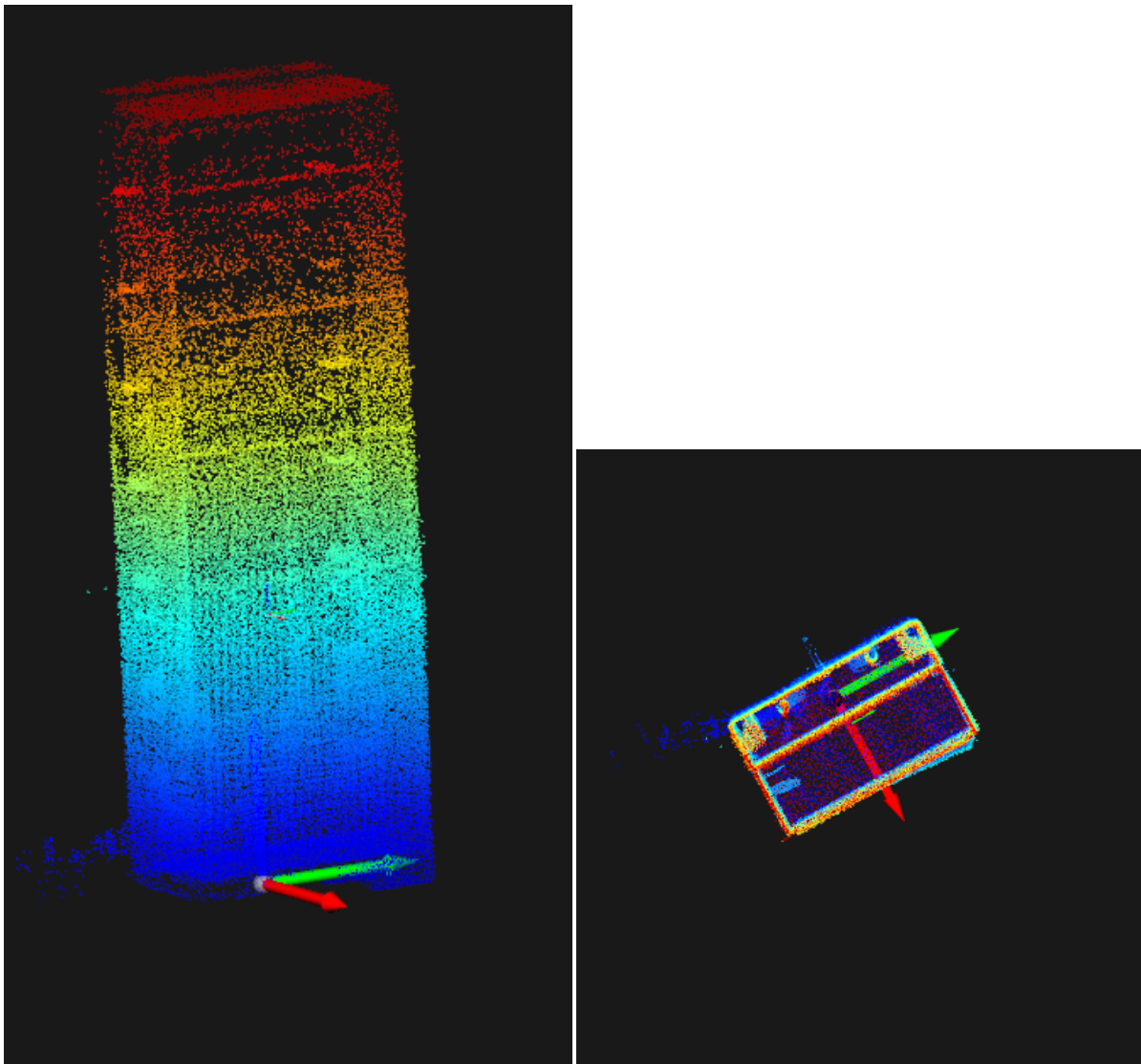
Localization at HRSG Manhole

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



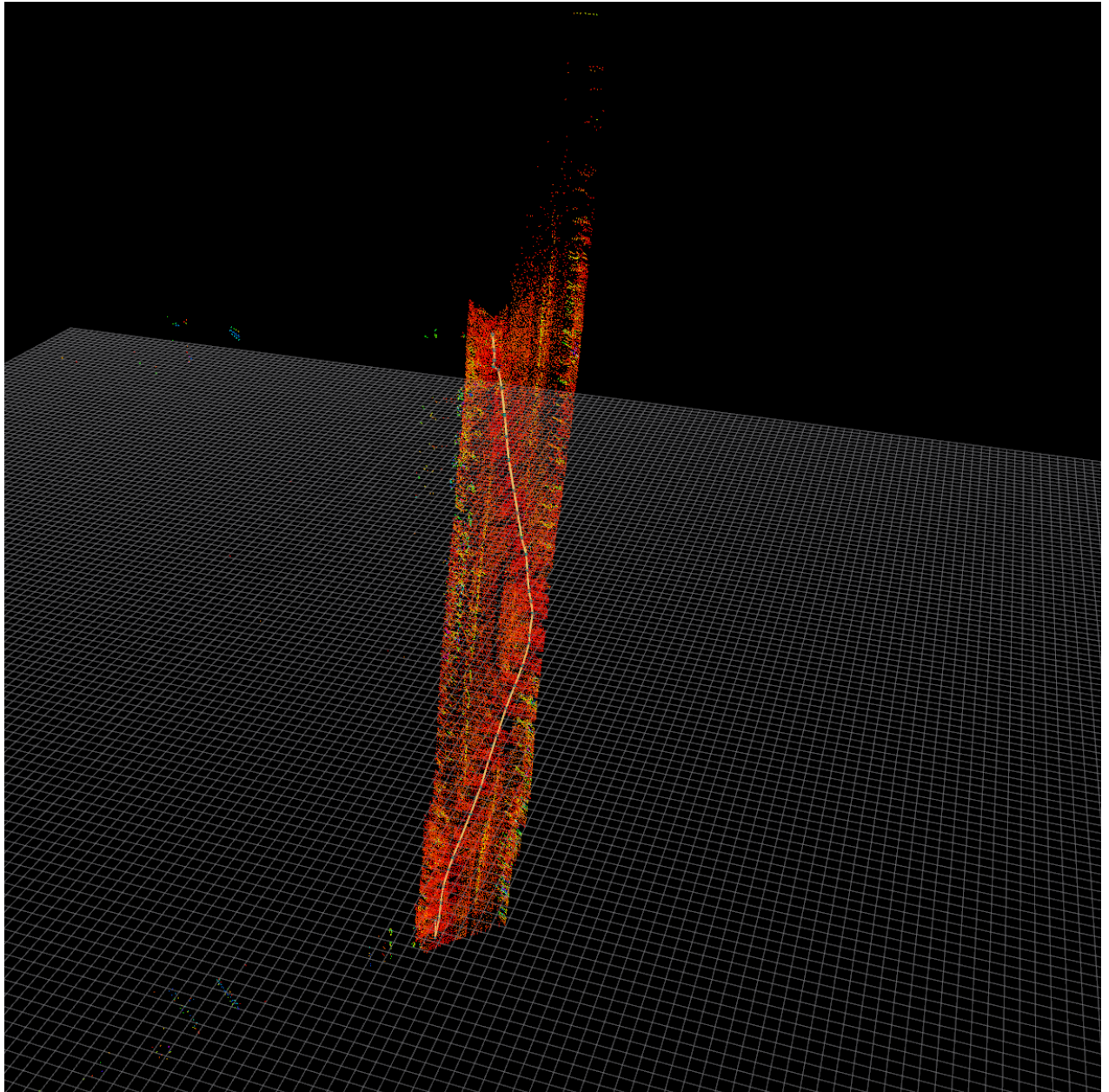
3. while building map, **construct NODE GRAPH**
 1. node: important (key) pose
 2. which involves LiDAR-inertial odometry (LIO) internally
4. (online) **LOCALIZE** drone
 1. initially by providing **INITIAL POSE**: where drone takes off (depends on the experiment), probably where **station** should be
 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

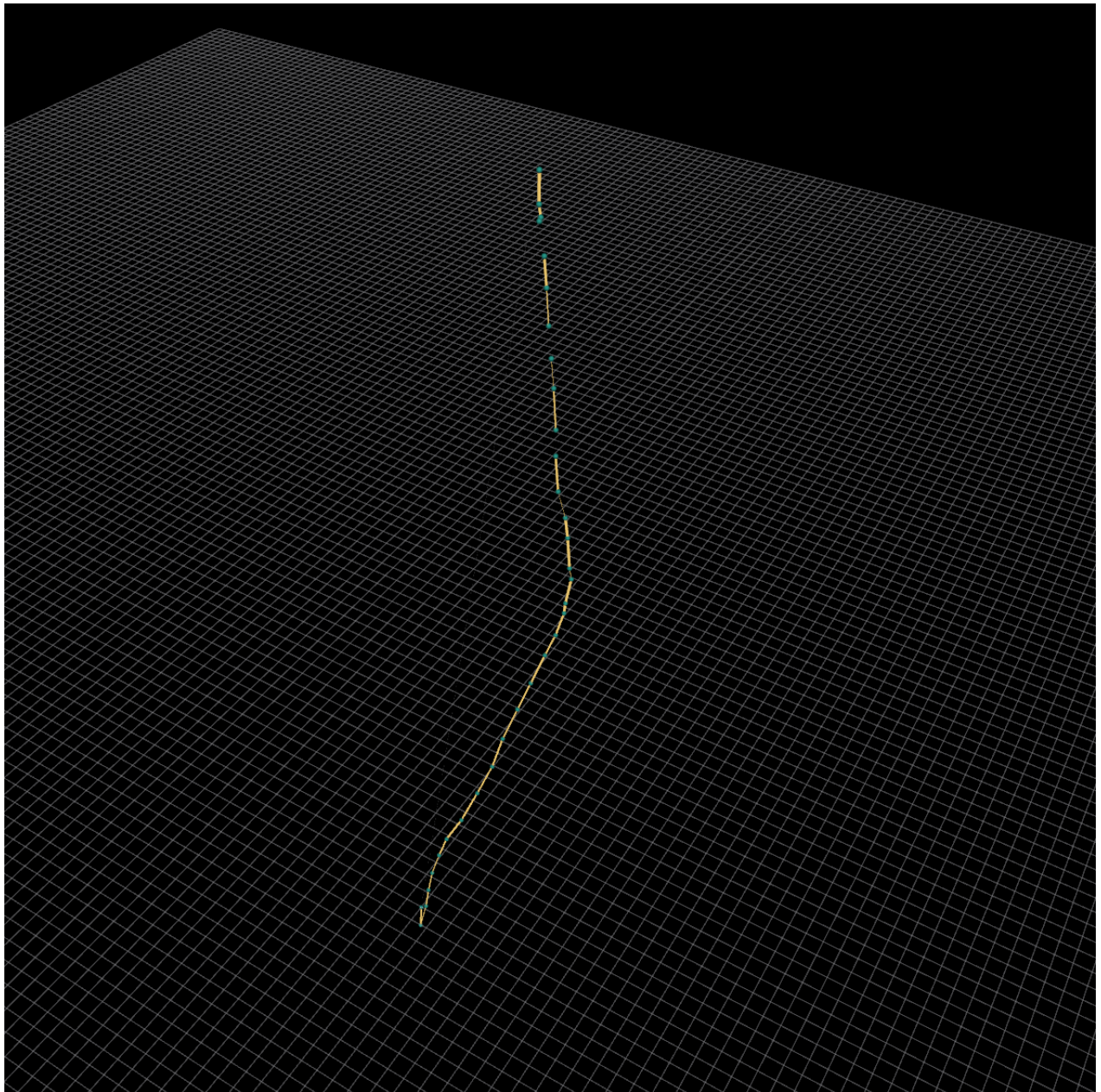
- an example of problematic map:



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
```

```

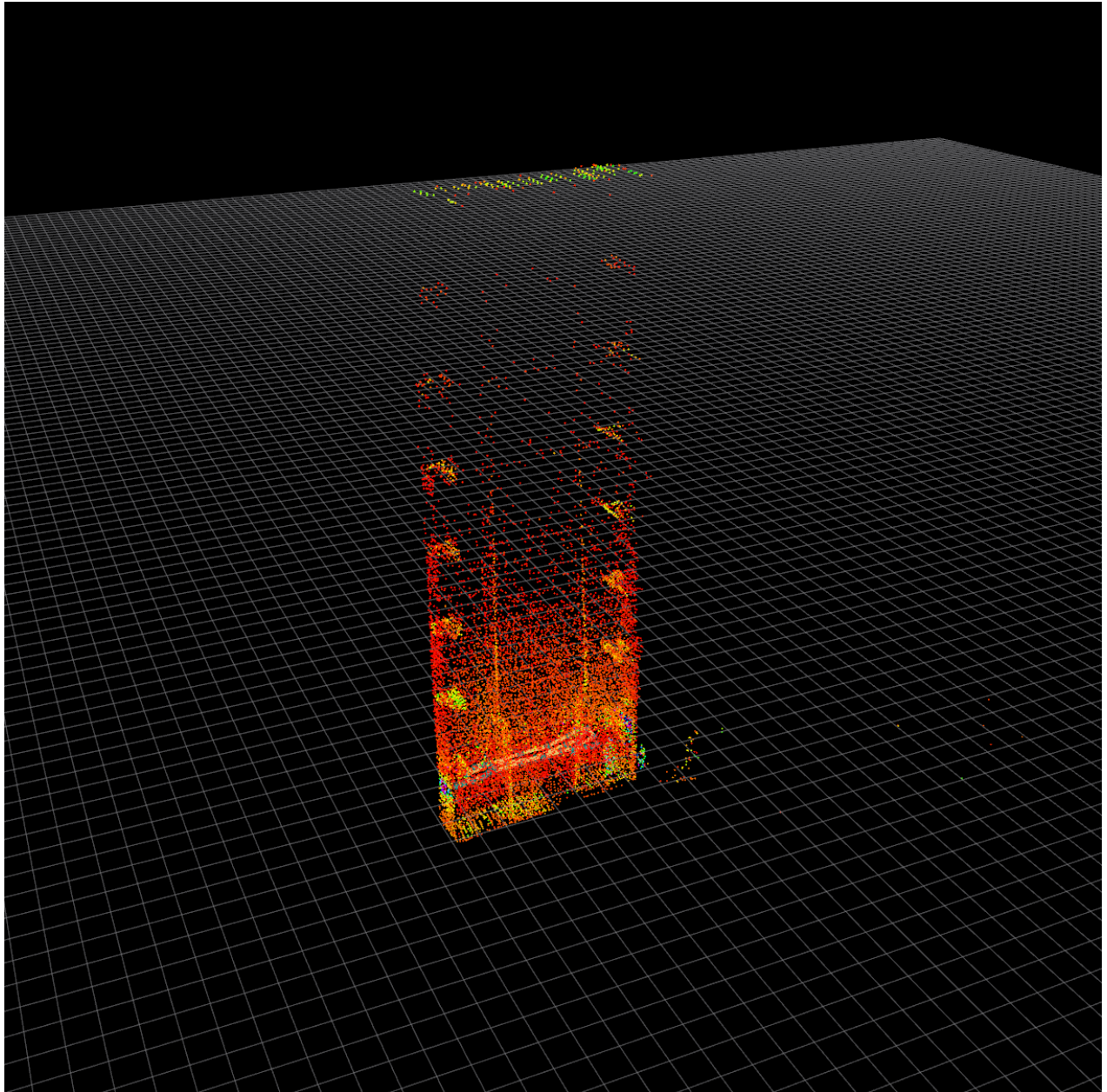
 kernel_sigma: 1.0 # m-estimator constant
 convergence_epsilon: 0.001 # convergence criterion
graph_optimization:
 max_iterations: 10 # maximum iteration
 abs_error_epsilon: 0.000001 # absolute residual epsilon
 rel_error_epsilon: 0.000001 # relative residual epsilon
 prior_orientation_noise_sigma: 0.000001 # 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 correspondence distance
 min_corrs: 100 # minimum correspondence 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_smoother:
```



```

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 correspondence distance
  min_corrs: 100            # minimum correspondence 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