

HW3

309512074 電控碩一 黃柏叡

```
ray@ray-ubuntu: ~/self-driving-car-2021/catkin_ws/src/309512074_hw3/bags 86x17
[RUNNING] Bag Time: 1542193271.762395 Duration: 30.257551 / 79.921261
[RUNNING] Bag Time: 1542193271.772451 Duration: 30.267607 / 79.921261
[RUNNING] Bag Time: 1542193271.782507 Duration: 30.277663 / 79.921261
[RUNNING] Bag Time: 1542193271.802624 Duration: 30.297780 / 79.921261
[RUNNING] Bag Time: 1542193271.812681 Duration: 30.307838 / 79.921261
[RUNNING] Bag Time: 1542193271.822739 Duration: 30.317895 / 79.921261
[RUNNING] Bag Time: 1542193271.832795 Duration: 30.327951 / 79.921261
[RUNNING] Bag Time: 1542193271.842852 Duration: 30.338008 / 79.921261
[RUNNING] Bag Time: 1542193271.852908 Duration: 30.348064 / 79.921261
[RUNNING] Bag Time: 1542193271.862964 Duration: 30.358120 / 79.921261
[RUNNING] Bag Time: 1542193271.873049 Duration: 30.368205 / 79.921261
[RUNNING] Bag Time: 1542193271.883104 Duration: 30.378260 / 79.921261
[RUNNING] Bag Time: 1542193271.893160 Duration: 30.388316 / 79.921261
[RUNNING] Bag Time: 1542193271.903218 Duration: 30.398373 / 79.921261
[RUNNING] Bag Time: 1542193271.913273 Duration: 30.408429 / 79.921261
[RUNNING] Bag Time: 1542193271.923331 Duration: 30.418487 / 79.921261

[ WARN ] [1634531738.161122082]: TF_REPEATED_DATA ignoring data with redundant timestamp
p for frame origin at time 1542193271.853527 according to authority unknown_publisher
[ WARN ] [1634531738.172708771]: TF_REPEATED_DATA ignoring data with redundant timestamp
p for frame origin at time 1542193271.865294 according to authority unknown_publisher
[ WARN ] [1634531738.182677078]: TF_REPEATED_DATA ignoring data with redundant timestamp
p for frame origin at time 1542193271.875361 according to authority unknown_publisher
[ WARN ] [1634531738.192764202]: TF_REPEATED_DATA ignoring data with redundant timestamp
p for frame origin at time 1542193271.885426 according to authority unknown_publisher
[ WARN ] [1634531738.202948432]: TF_REPEATED_DATA ignoring data with redundant timestamp
p for frame origin at time 1542193271.895490 according to authority unknown_publisher
[ WARN ] [1634531738.213440647]: TF_REPEATED_DATA ignoring data with redundant timestamp
p for frame origin at time 1542193271.906010 according to authority unknown_publisher
[ WARN ] [1634531738.223370802]: TF_REPEATED_DATA ignoring data with redundant timestamp
p for frame origin at time 1542193271.916078 according to authority unknown_publisher
[ WARN ] [1634531738.233643879]: TF_REPEATED_DATA ignoring data with redundant timestamp
p for frame origin at time 1542193271.926292 according to authority unknown_publisher

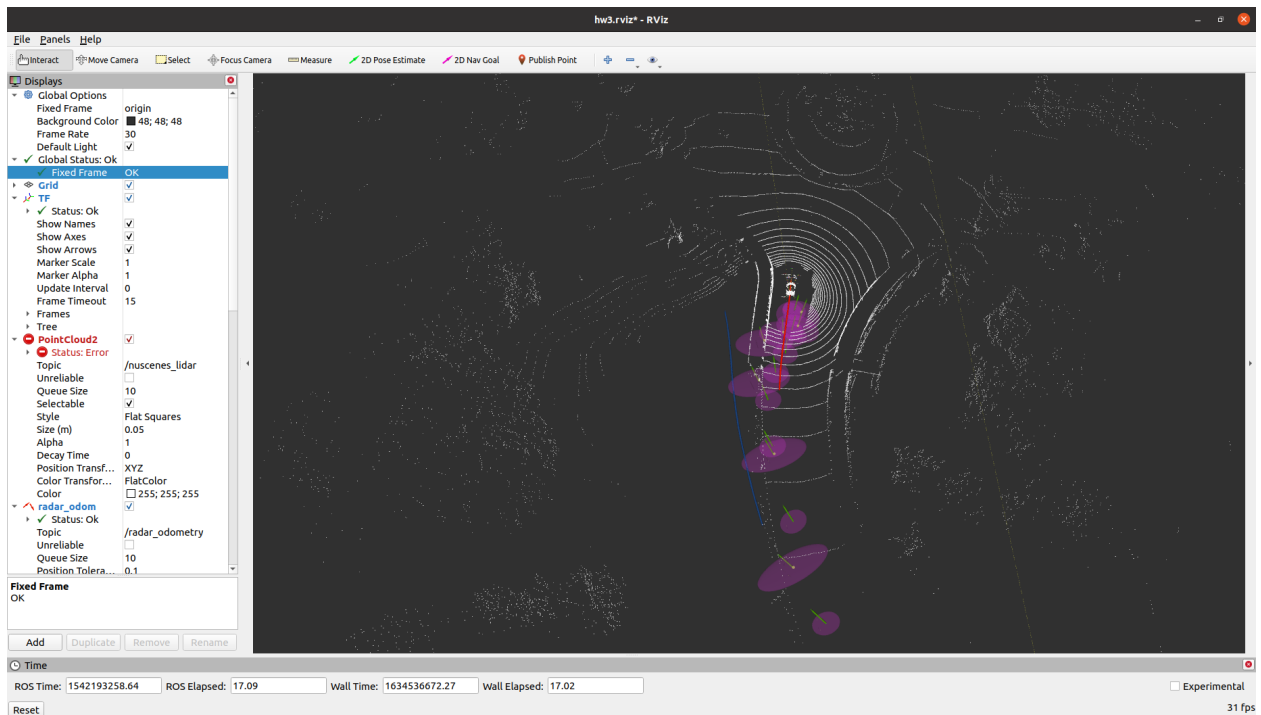
[2] /home/ry/self-driving-car-2021/catkin_ws/src/309512074_hw3/robot_localization/launch/ekf_template.launch http://
Warning: TF_REPEATED_DATA ignoring data with redundant timestamp for frame origin at tim
e 1542193271.885426 according to authority unknown_publisher
Warning: TF_REPEATED_DATA ignoring data with redundant timestamp for frame origin at tim
e 1542193271.895490 according to authority unknown_publisher
Warning: TF_REPEATED_DATA ignoring data with redundant timestamp for frame origin at tim
e 1542193271.906010 according to authority unknown_publisher
Warning: TF_REPEATED_DATA ignoring data with redundant timestamp for frame origin at tim
e 1542193271.916078 according to authority unknown_publisher
Warning: TF_REPEATED_DATA ignoring data with redundant timestamp for frame origin at tim
e 1542193271.926292 according to authority unknown_publisher

[2] roscore http://ray-ubuntu:11311/88x17
SUMMARY
*****
PARAMETERS
* /roslistro: noetic
* /rosversion: 1.15.11
NODES
auto-starting new master
process[master]: started with pid [801908]
ROS_MASTER_URI=http://ray-ubuntu:11311/
setting /run_id to a44ce840-2fie-11ec-b8e5-05a09df12c53
process[rosout-1]: started with pid [801933]
started core service [/rosout]
```

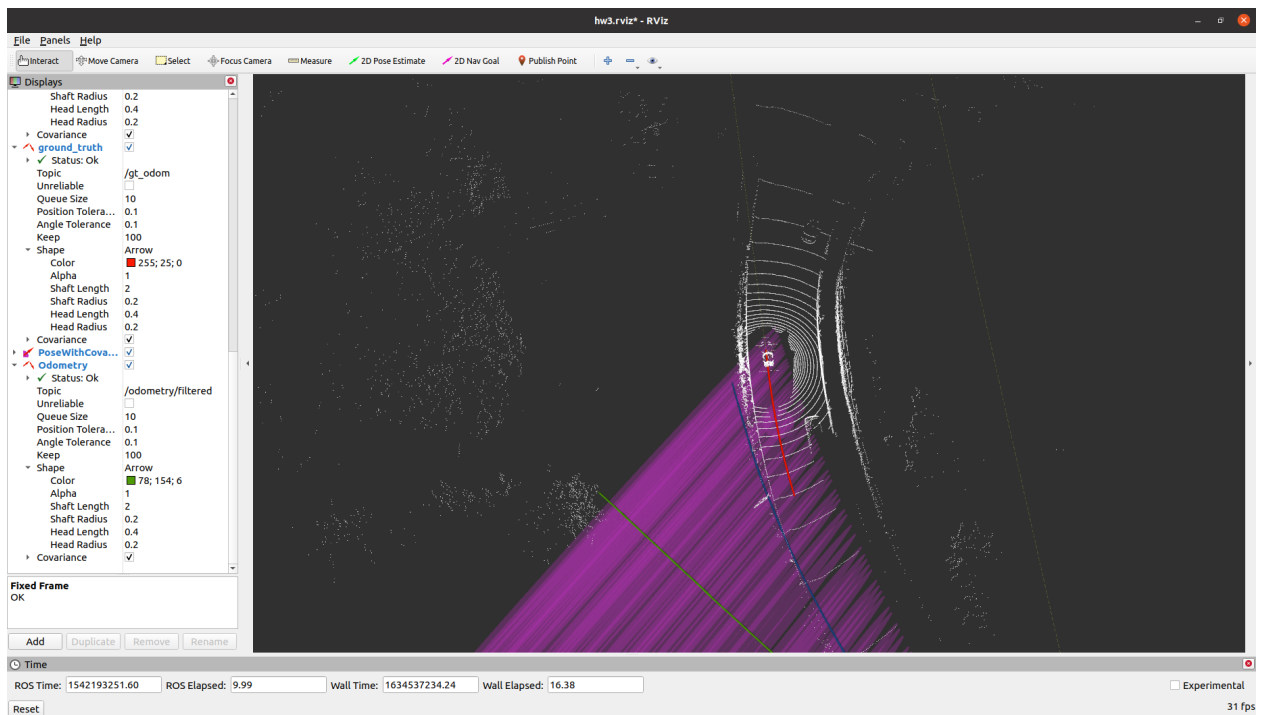
EKF Result

RED arrows are ground truth odometry, GREEN arrows are EKF result, BLUE arrows are radar odometry, PURPLE eclipse represents covariance.

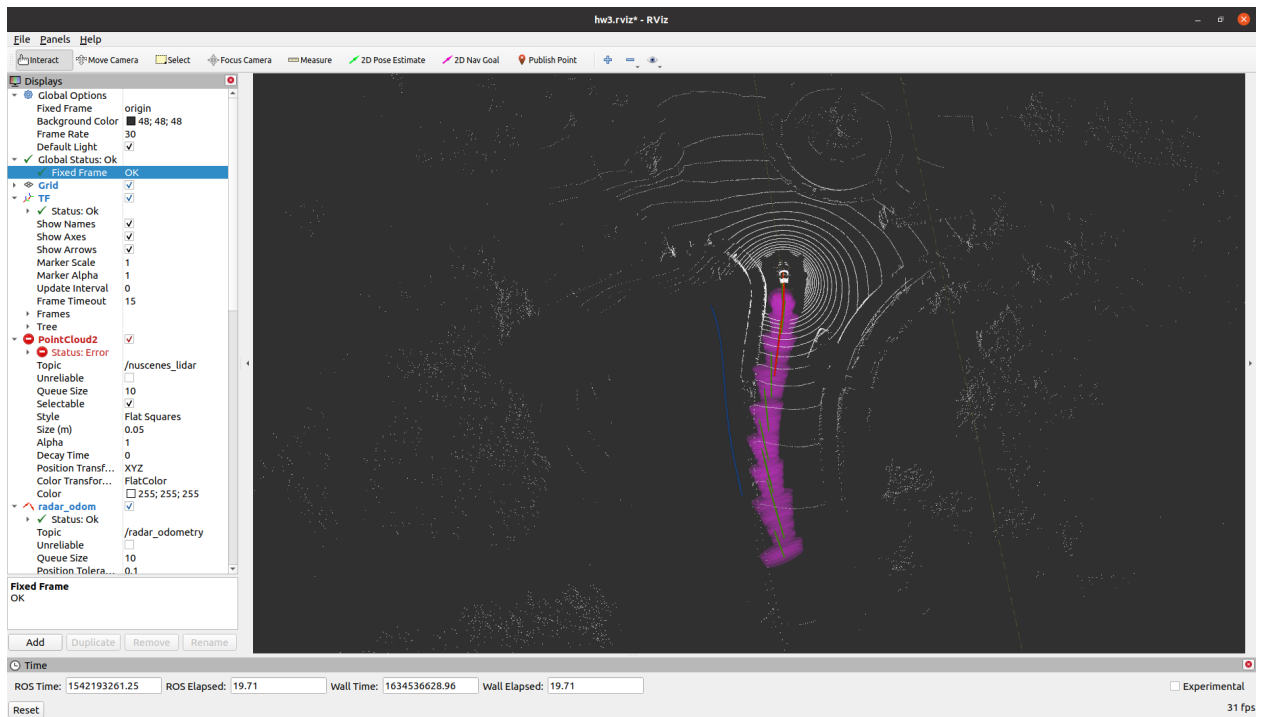
- Screenshot of the result using GPS in EKF



- Screenshot of the result using radar odometry in EKF

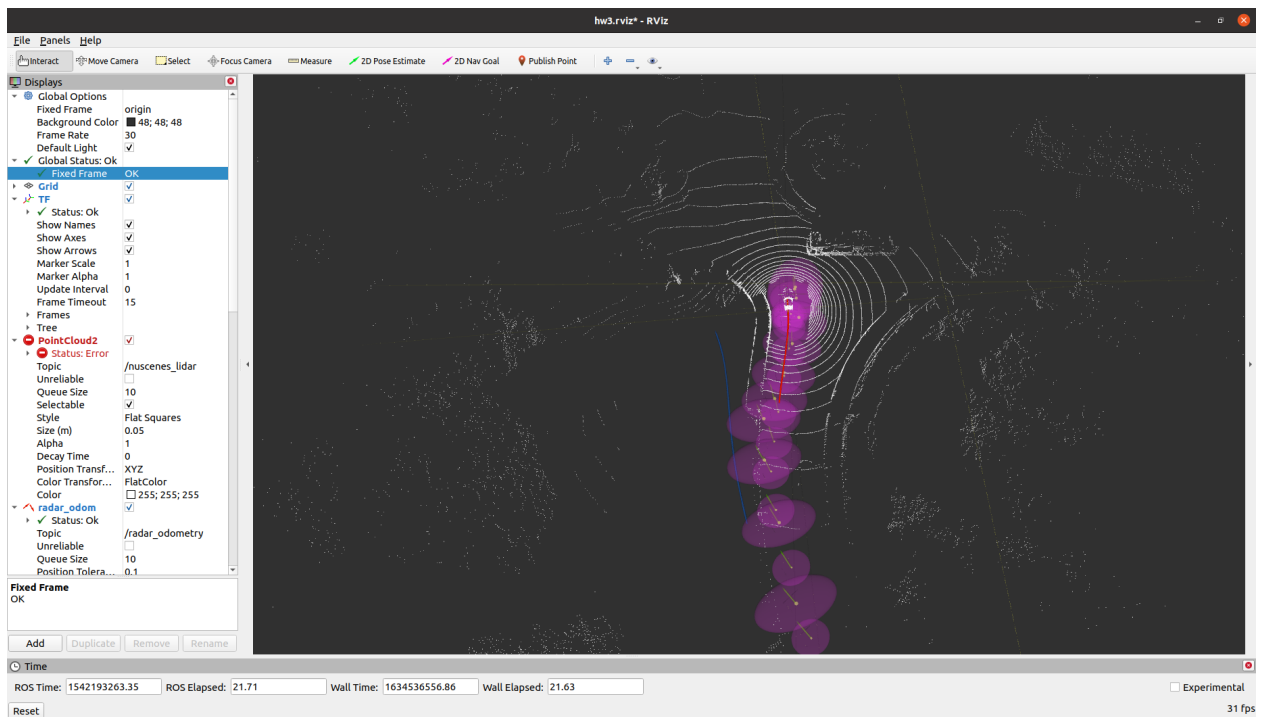


- Screenshot of the result fusing GPS and radar odometry in EKF

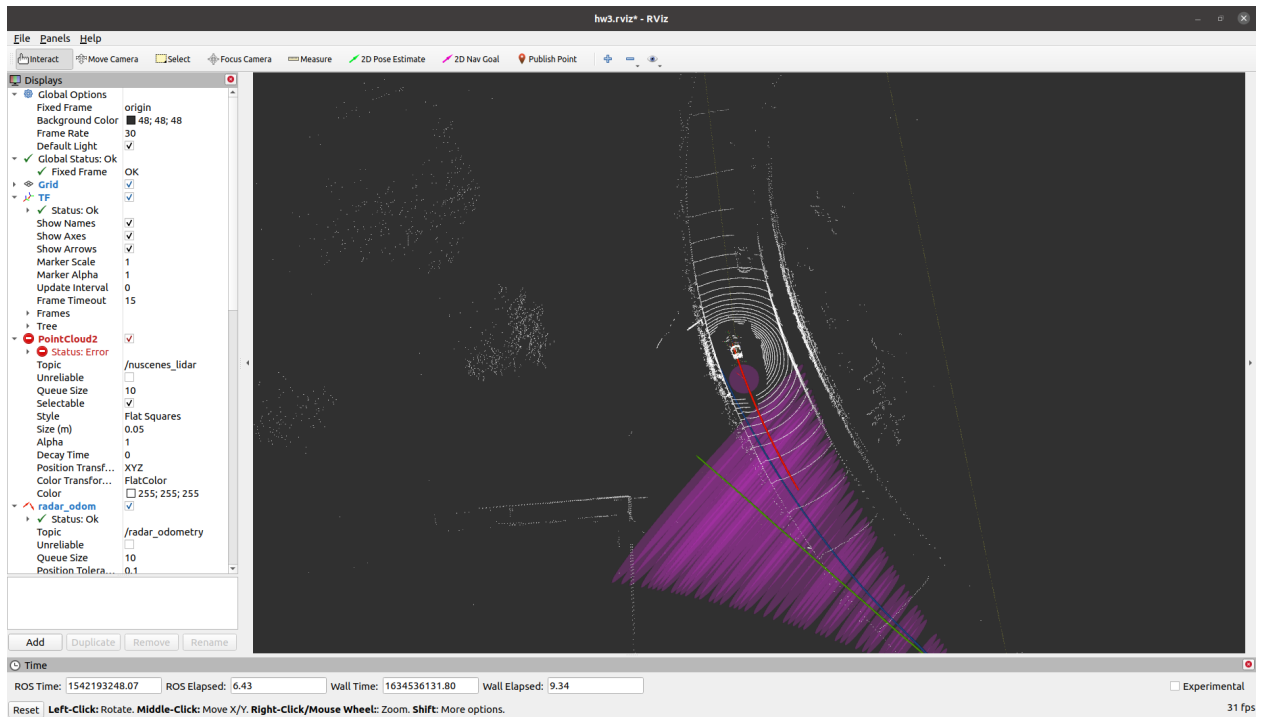


UKF Result

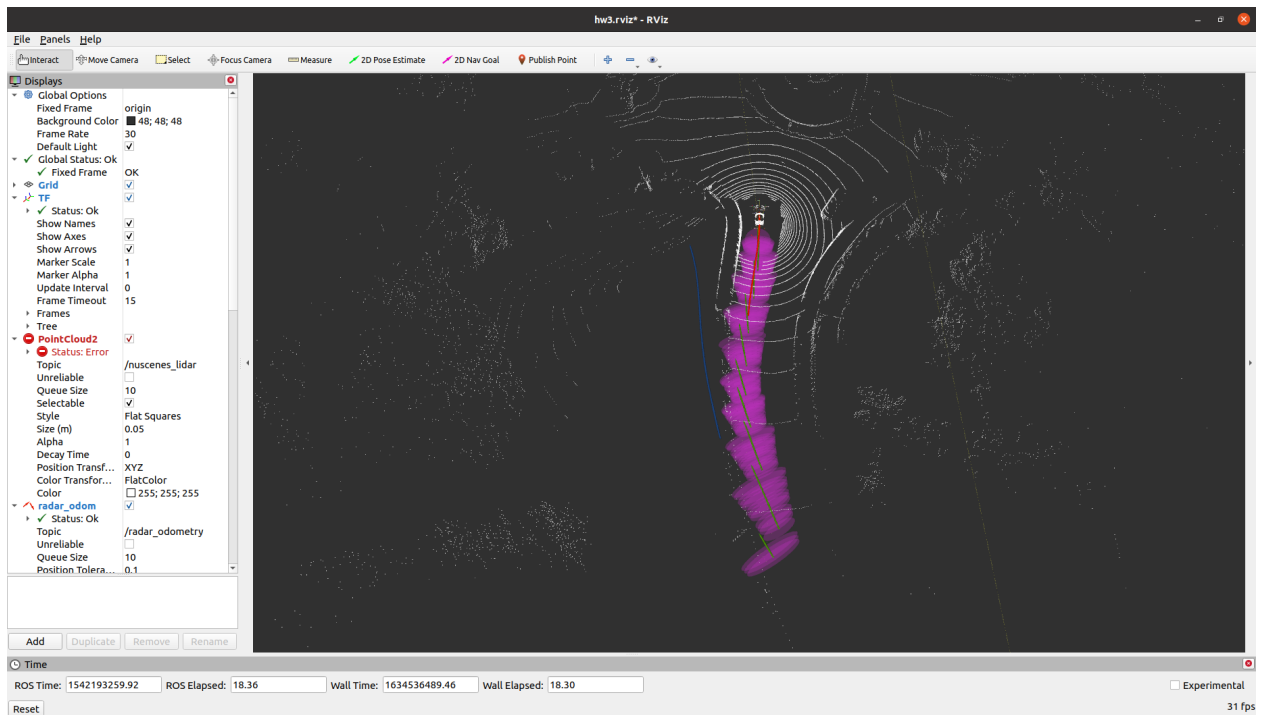
- Screenshot of the result using GPS in UKF



- Screenshot of the result using radar odometry in UKF



- Screenshot of the result fusing GPS and radar odometry in UKF



Discussion

- How do robot_localization package know the covariance matrix of GPS and radar odometry?

By subscriber, the rosbag will publish

geometry_msgs/PoseWithCovarianceStamped(/gps) and nav_msgs/Odometry(/radar_odometry) rostopic, therefore the robot_localization package can know the covariance matrix of GPS and radar odometry.

- What is the covariance matrix of GPS and what does it mean?

The covariance matrix is a powerful tool for expressing what remains uncertain about a group of variables that have already been measured many times. This tool can be applied to help make smart decisions about systems which are messy and complex because they have many different kinds of uncertainty piled on top of each other.

covariance matrix of GPS:

[illegible]

- In the yaml file, do you set differential parameter of odometry and GPS to true? or false? Why?

differential: When differential mode is enabled, all absolute pose data is converted to velocity data by differentiating the absolute pose measurements. These velocities are then integrated as usual.

In our case, we have two sources with absolute pose information: radar odometry and GPS. If the variances of the input sources are not configured correctly, the measurements may get out of sync and cause oscillations in the filter.

I set differential parameter of odom0(odometry) to true since it odometry message include velocity information(twist), and the differential parameter to pose0(gps) to false since it is not continuous, if not, the covariance of pose will be divergence.

Bonus

- Compare EKF result with UKF result. Describe your findings and explain why.

EKF: In real world, we have non linear equations, because we may be predicting in one direction while our sensor is taking reading in some other direction, so it involves angles and sine cosine functions which are non linear. So EKF takes helps of Taylor Series (and Jacobian Matrix further) to linearly approximate a non linear function around the mean of the Gaussian and then predict the values.

UKF: sigma points, we take some points on source Gaussian and map them on target Gaussian after passing points through some non linear function and then we calculate the new mean and variance of transformed Gaussian, and those sigma points are the representative of whole distribution.

The main difference between EKF and UKF is EKF only take one point(mean) and approximate, but in UKF we take a bunch of points(sigma points) and approximate, the more number of points it take, more precise the approximation will be.

In my case, the outcome of EKF and UKF is similar, but I found out the results of covariance in UKF is more continuous compare to the result of EKF.