## Record For Implementing UKF on Existing Round Move

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May 24, 2021

## 1 Implementing UKF

After connecting to the robot through ssh, then execute the following command:

1 husarion@husarion: \$ roscore

start 2nd. command line window and execute following command:

1 \$\ roslaunch rosbot\_ekf all.launch rosbot\_pro:=true

start 3rd. command line window and launch the *robot\_localization* through executing:

```
1    ~/pathTo/catkin_ws$ source ./devel/setup.bash
2    ~/pathTo/catkin_ws$ roslaunch playground start_filter.launch
```

Now our purpose is implementing UKF algorithm into existing  $round\_move.py$  script.

To launch the *UWB tag* through executing:

```
1     ~/pathTo/catkin_ws$ source ./devel/setup.bash
2     ~/pathTo/catkin_ws$ roslaunch localizer_dwm1001 dwm1001.launch
```

inside the *playground* package there's a **Python** script called *round\_move\_ukf.py*. Now execute:

The running process was: we let the robot to move back and forth twice with single distance of 2m, which makes a total distance of 8m, we repeat this experiment with the same starting position for the robot with different R and Q for UKF:

The 1st. combination of R and Q is:

The 2nd. combination of R and Q is:

```
1    std_x , std_y = .1 , .1
2    ukf.R = np.diag([std_x ** 2, std_y ** 2])
3    ukf.Q[0:2 , 0:2] = Q_discrete_white_noise(2, dt=dt , var=0.01 ** 2)
4    ukf.Q[2:4 , 2:4] = Q_discrete_white_noise(2, dt=dt , var=0.01 ** 2)
```

- (a) Four stable working anchors with worse coordinates precision under the 1st. combination of R and Q
- (b) we calibrated the coordinate of UWB more precisely (from this running on) with BOSCH Laser measure but only 3 stable working anchors under the 1st. combination of R and Q
- (c) With 4 stable working anchors under the 1st. combination of R and Q
- (d) With 4 stable working anchors under the 2nd. combination of R and Q

Afterwards, position data based on calculation and  $\mathit{UKF}$  are collected separately in files called:

- (a) pose1204.csv (1st. run)
- (b) pose2104.csv (2nd. run)
- (c) pose2204.csv
- (d) pose2304.csv

and after plotting, we get Fig. 1, Fig. 2, Fig. 3 and Fig. 4:

In Fig. 1 and 2, the *khaki trace* denotes the Cartesian coordinates based on *UKF* calculation, while the *blue trace* represents the raw *UWB tag* position, which is obviously not very stable. In Fig. 3 and 4 we can see only 1 trajectory because UKF track the position of tag so closely that we can not distinguish the trajectory of UKF and UWB tag and 4 colors are for better demonstration of 4 single rounds in general.

Our calculation process is:

- 1. Initial position was calculated based on UWB
- 2. Afterwards, every small step was calculated based on an *internal EKF* with only IMU and Odometry as input and an external UKF with all sensor inputs

At the same time, mean squared error of first two running was calculated through Python script MSE\_cal.py and the calculation result was:

```
1 $ MSE for 1st. run = 0.0014062992125984232
2 $ MSE for 2nd. run = 0.0003471074380165288
```

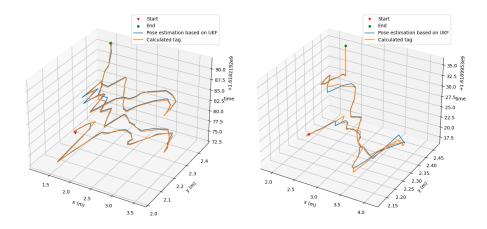


Figure 1: UKF vs. UWB alone

Figure 2: UKF with UWB calibration and 3 stable working anchors only vs. UWB alone

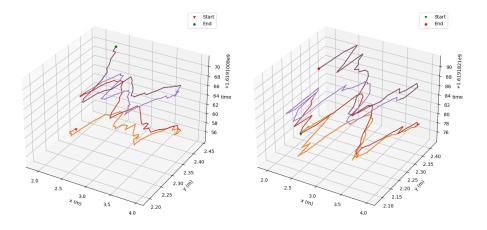


Figure 3: UKF with UWB calibration Figure 4: UKF with UWB calibration and 4 stable working anchors under the And Four Stable Working Anchors Unlst. comb of R and Q der the second Comb of R and Q

So from the comparison we can see that the calibration made to UWB anchors has brought us some measurement improvement, although far from enough.

## 2 Idea for next step

- 1. Try to get the orientation of the robot, e.g. from quaternion (subscribe from topic mpu9250) to euler, which can be extremely useful in the process which involves turning
- 2. Measure the process error through observation: Let the robot move forward with a velocity of  $0.5~\mathrm{m/s}$  and record its relative position after 1s, repeat e.g.  $1000~\mathrm{times}$ , which makes it reliable
- 3. Try to read IMU data from UWB tag sensor unit
- 4. Fuse uwb tag pos + IMU to eliminate pos errors
- 5. Conducting repeated experiment and at the same time, gather raw sensor data, which contains more noise definitely and try to reduce the noise
- 6. Consider using Graph Neural Networks,  $Particle\ Filter$ , not only UKF and EKF
- 7. Consider designing a random walk model for the final demonstration

## 3 Results and Conclusions

From all the experiments and figures above we can see, after conducting UKF, tag was closely followed, so the precision of the tag matters quite a lot. However, the actual movement was along the x-axis, the change in y-axis should be no more than 5cm theoretically, which is maximum around 30cm in the figures, should be eliminated.