

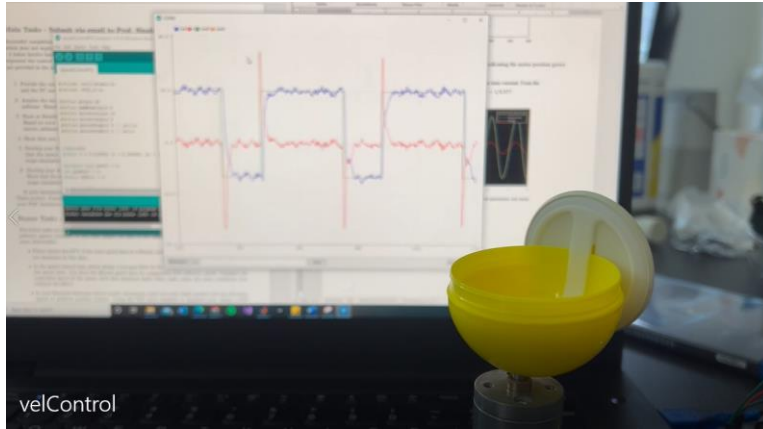


Hi! This Jing.

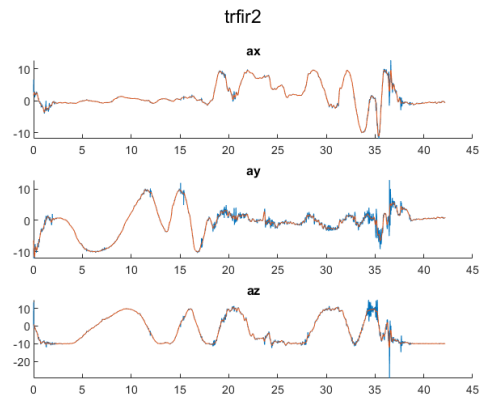
I put down some of my past projects, coursework and internship accomplishments here, involving controls, mechatronics, mechanics and data mining.

My portfolio GitHub Repo: <https://github.com/wwtse/folioS>

Projects



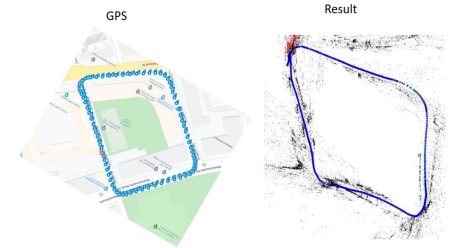
DCmotor control



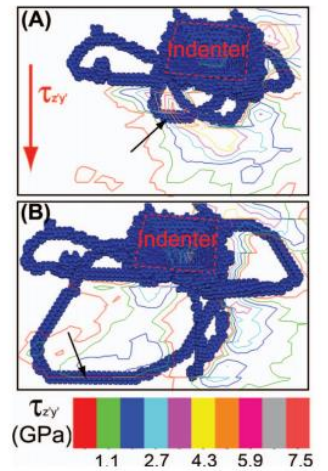
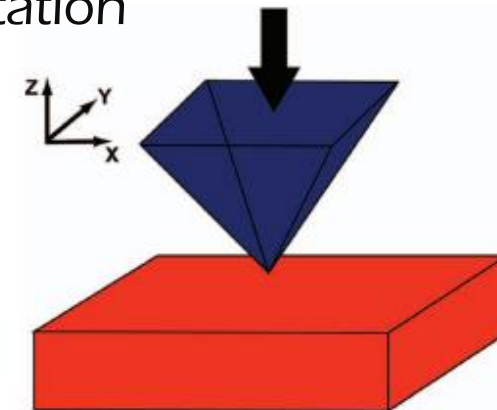
IMU filters



SLAM



Nanoindentation Simulation



Coursework

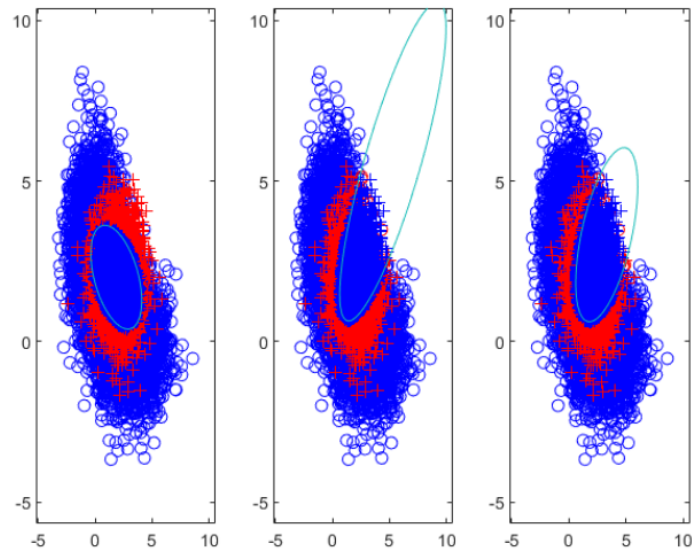
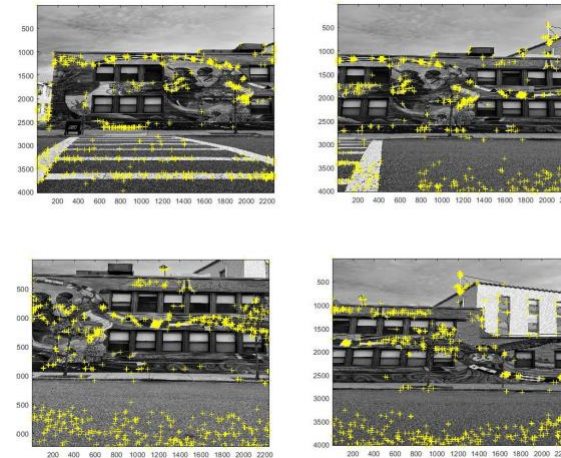


Figure 8: Validation Datasets with Decision Boundary by logistic-quadratic-function

Machine Learning



Sensor and Navigation

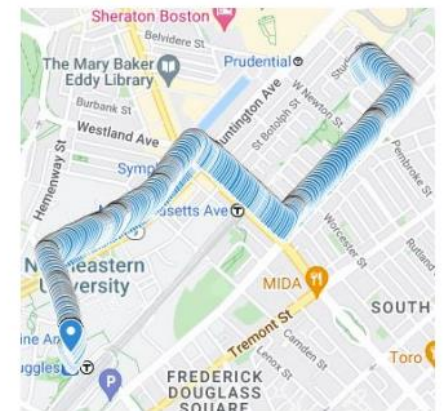
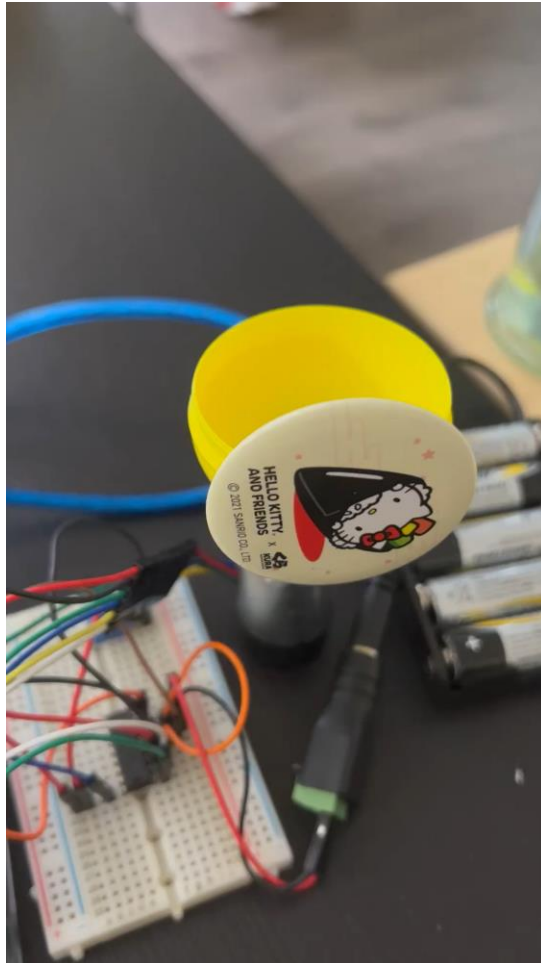


Figure 1: data present in Google Map

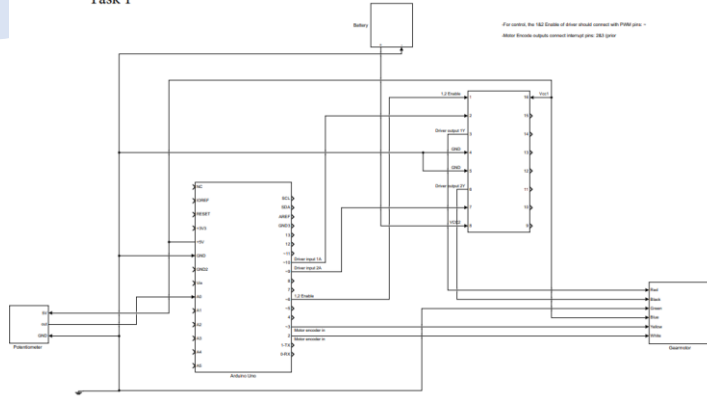
DCmotor Control



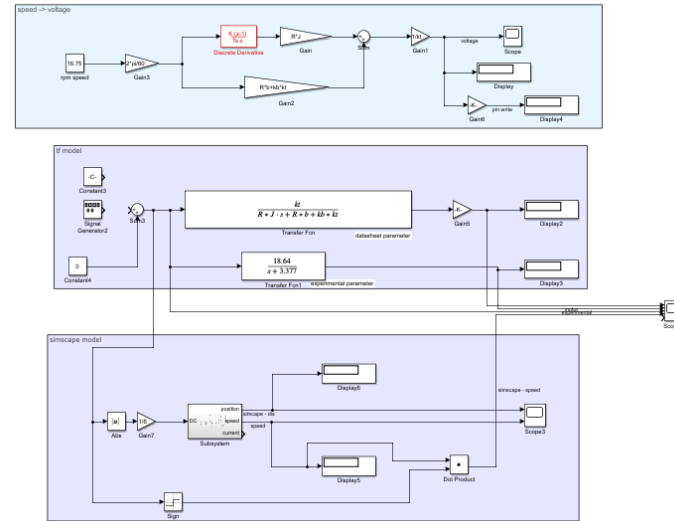
Video of DC motor position control

Demonstrate motor speed and position control with Arduino IDE and Simulink Real-Time Target.

Circuit Diagram Task 1



1. Design the circuit diagram



2. Simulation of DC motor with Simulink model

```

speedControlPID
double u = 0;
double e = 0;

//PID parameter
double rpmD = 10;
double kp = .2, ki = 3, kd = 0;
long eintegral = 0;

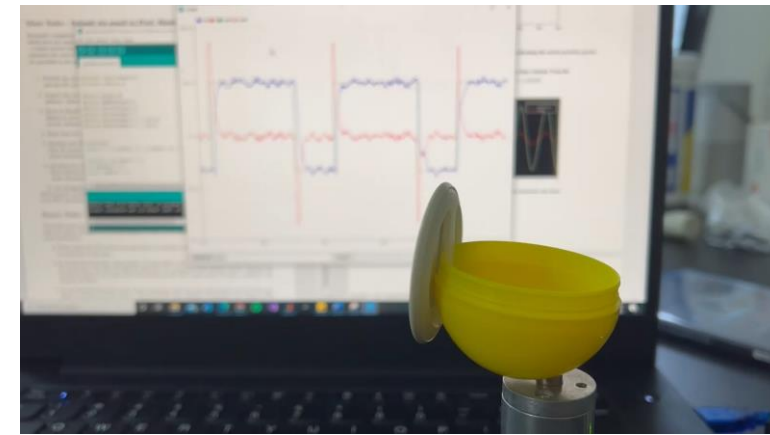
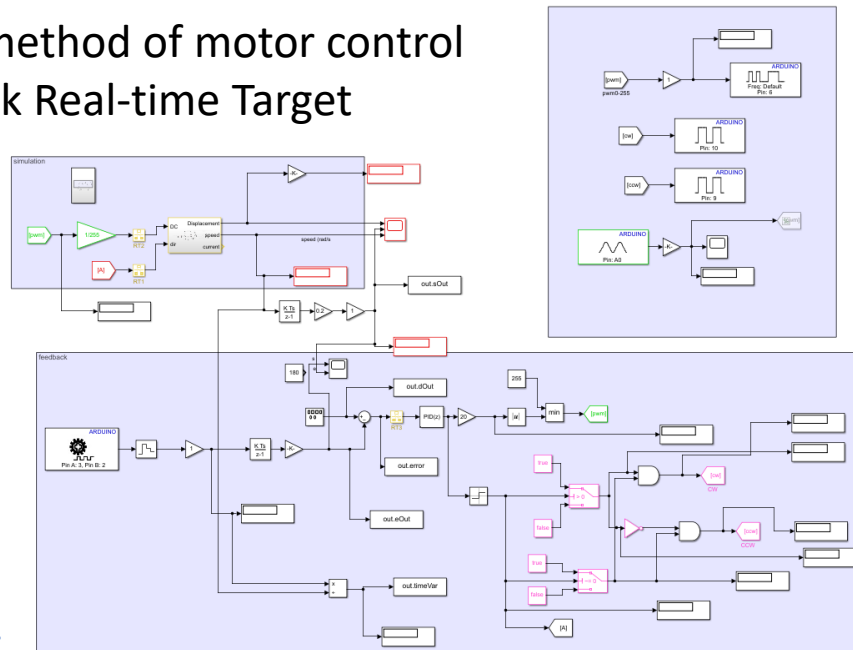
PID myPID(&rpm, &u, &rpmD, kp, ki, kd, DIRECT);

void setup() {
  Serial.begin(9600);
  pinMode(potpin, INPUT); // Potentiometer
  pinMode(motorOutApin, INPUT); // yellow
  pinMode(motorOutBpin, INPUT); // white
  pinMode(driverIn1pin, OUTPUT); //1Y
  pinMode(driverIn2pin, OUTPUT); //2Y
  pinMode(pwmEnablepin, OUTPUT); // pwmEnable

```

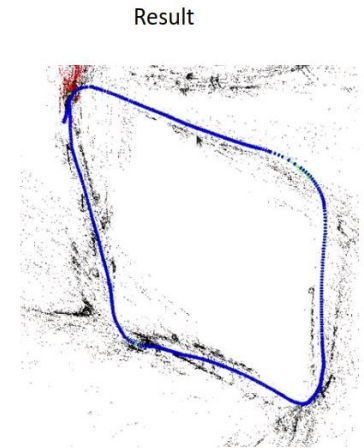
3. One method of motor control with Arduino IDE and C++

4. Another method of motor control with Simulink Real-time Target

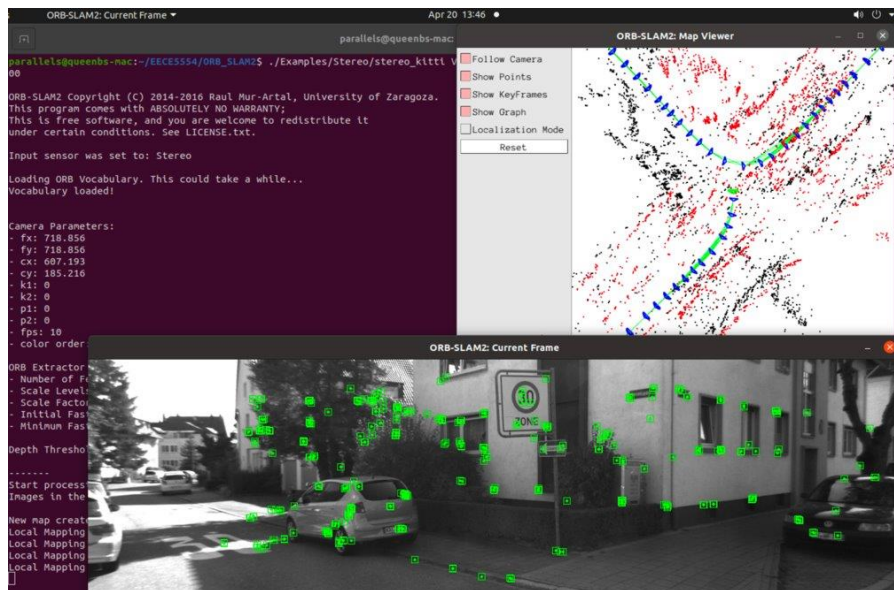


Video of DC motor speed control

SLAM



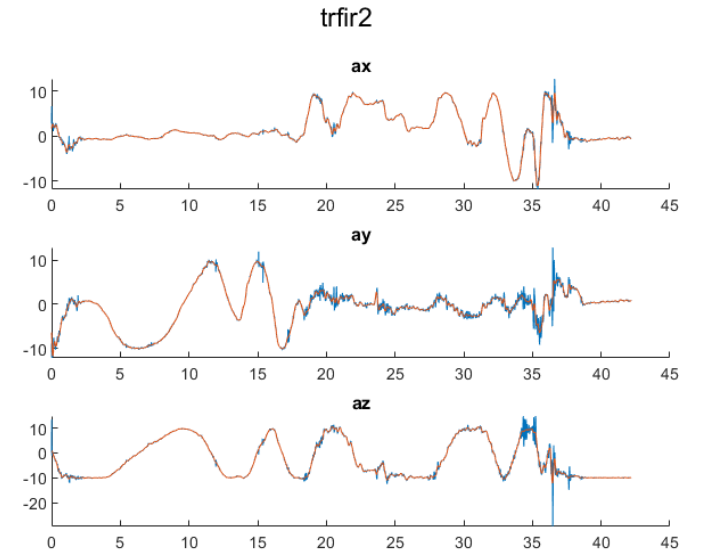
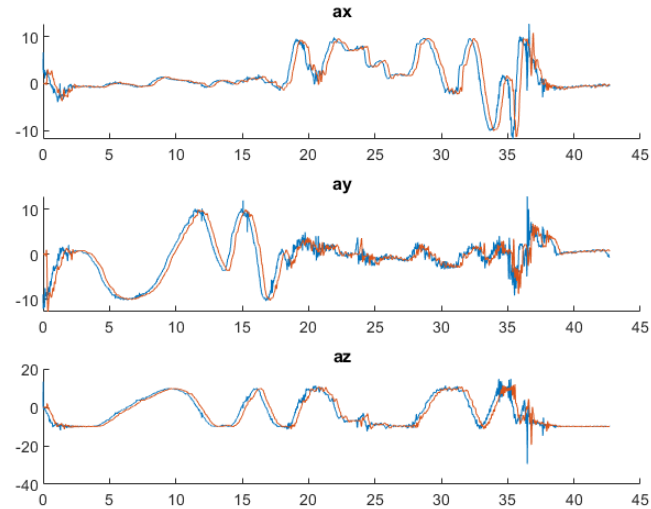
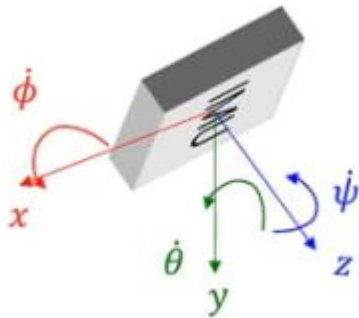
Map of sequence 00 of kitti



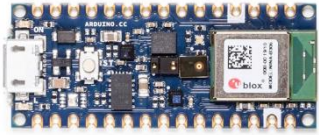
SLAM using stereo cameras dataset (kitti and NUance) with ORB-SLAM2 and ROS in Linux system.

[The Group Presentation Slides](#)

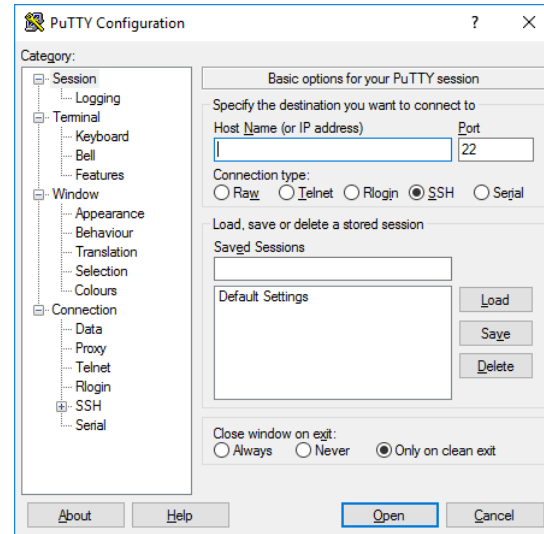
IMU Filters



Filter IMU raw data with the combination of FIR, IIR and complementary filters.

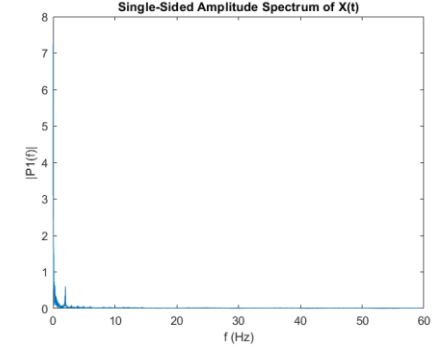
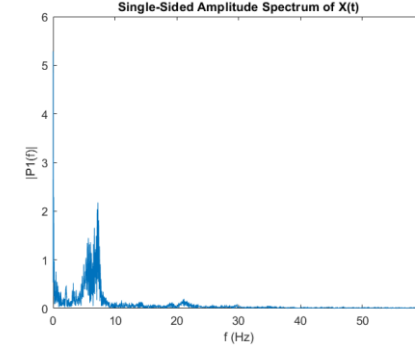


+

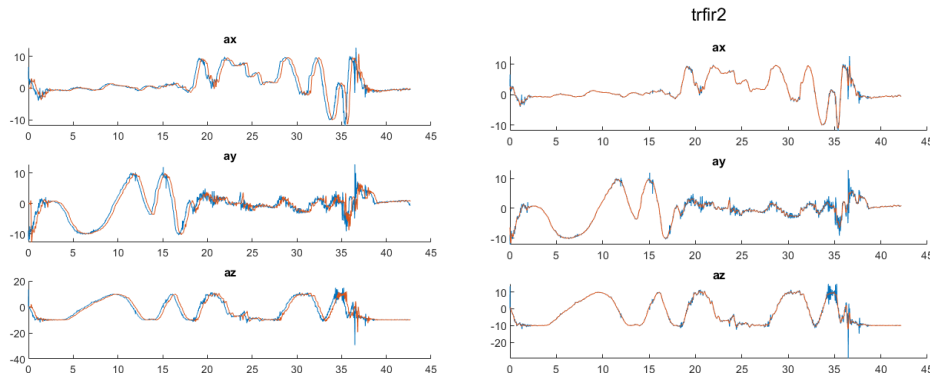


1. Datasets source: Parrot Mambo quadcopter and LSM9DS1 IMU. Terminal emulator PuTTY is used to log the signal.

```
1 clear;
2 close all;
3
4 % Load IMU_Hover/R5dataLine0904.mat;
5 fs = 119;
6 g = 9.81;
7
8 fo = 4; %S is useless
9 fname = 'putty_test'+fo;
10 data1 = table2array(readtable('IMU_Hover/'+fname));
11 ax = data1(1:end-1,1)*g;
12 ay = data1(1:end-1,2)*g;
13 az = data1(1:end-1,3)*g;
14 at = 1/fs*(0:(length(data1)-1));
15
16 S = ax;
17 T = at(2)-at(1);
18 Fs = fs;
19 L = length(S);
20 t = at;
21
22 X = S;
23 Y = fft(X);
24
25 figure;
26 P2 = abs(Y/L);
27 P1 = P2(1:L/2+1);
28 P1(2:end-1) = 2*P1(2:end-1);
29
30 f = Fs*(0:(L/2))/L;
31 plot(f,P1)
32 title('Single-Sided Amplitude Spectrum of X(t)')
33 xlabel('f (Hz)')
34 ylabel('|P1(f)|')
```



2. Analyze signal with frequency sampling method



$$y[n] = (1 - \alpha)x[n] + \alpha y[n - 1] \quad (15)$$

$$\alpha = \frac{\tau}{\tau + T_s} \quad (16)$$

when τ is the desired time constant, T_s is the sampling interval.

In IMU, $x[n]$ is the angle data calculated by gyro meter reading, $y[n - 1]$ is the angle data calculated by accelerometer reading.

3. Low pass filter design with window design method: 4 FIR filters and 1 IIR filter

4. Complementary filter design

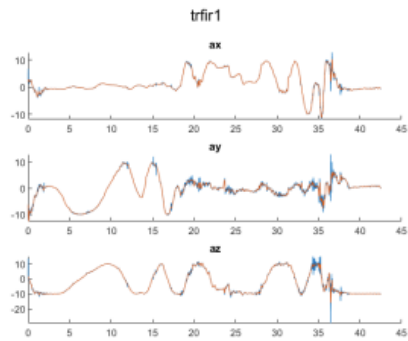


Figure 25: acceleration filter by tfir1

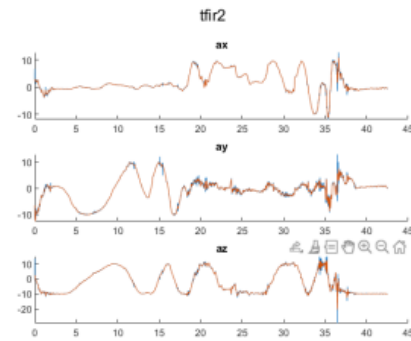


Figure 26: acceleration filter by tfir2

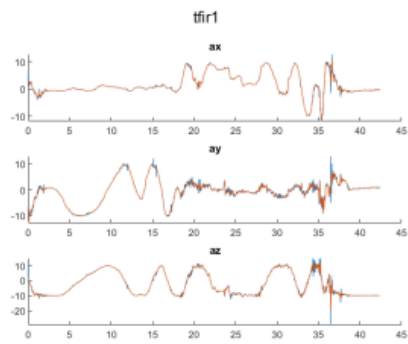


Figure 27: acceleration filter by tfir1

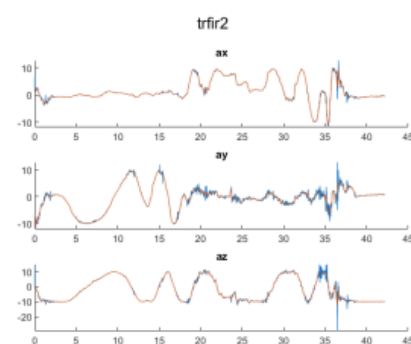


Figure 28: acceleration filter by tfir2

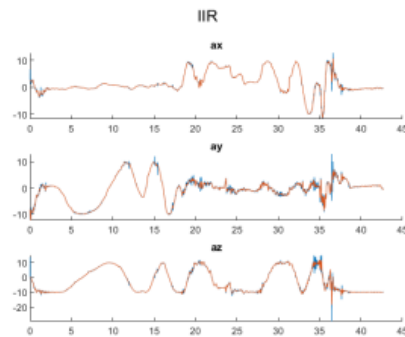


Figure 29: acceleration filter by IIR

5. Performance of the low-pass filters

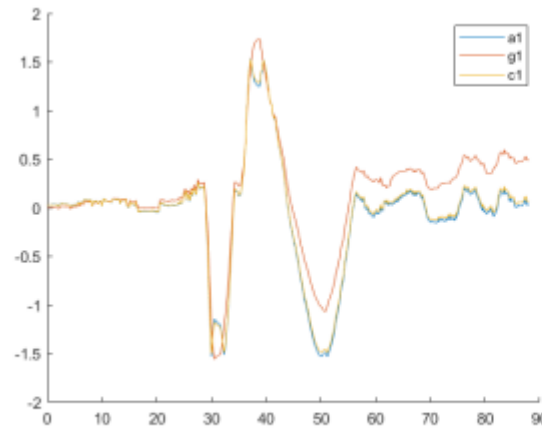
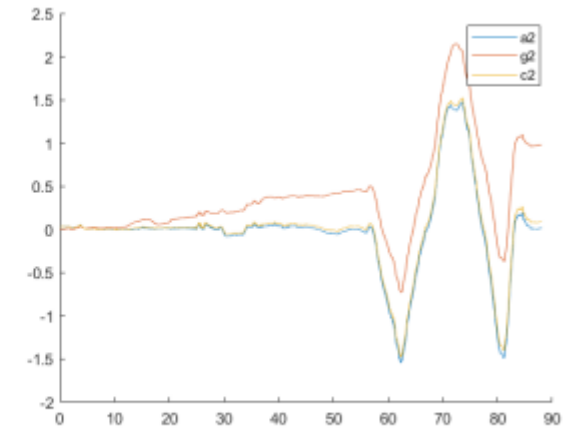
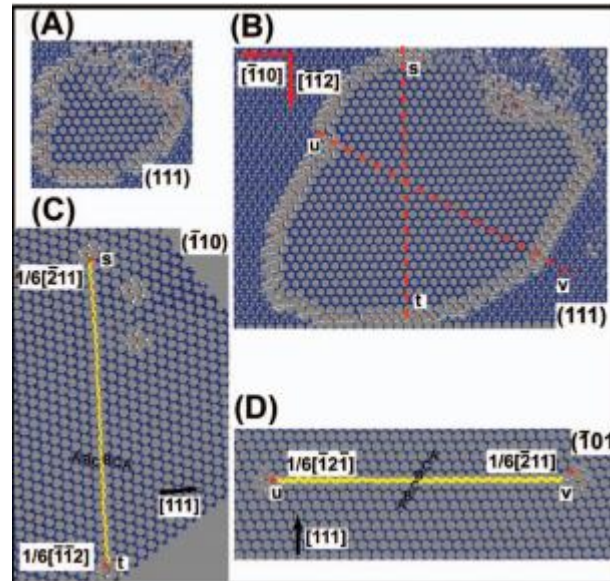
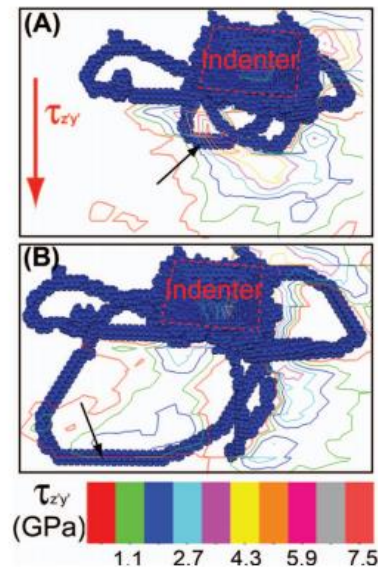
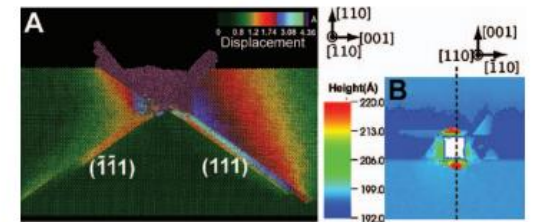
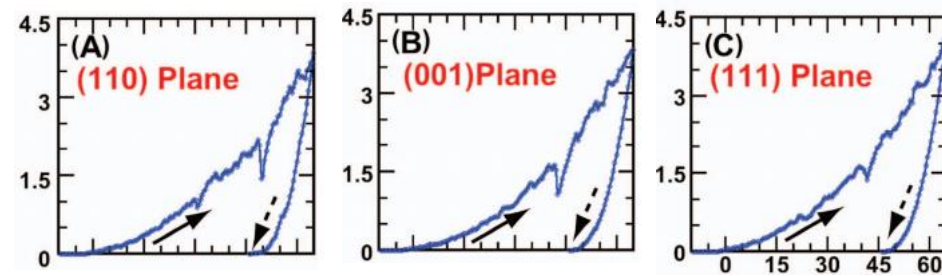
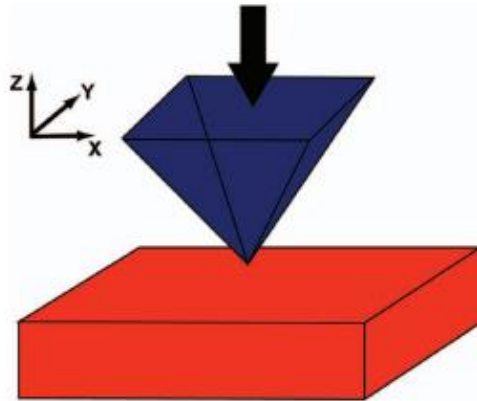


Figure 35: Angle calculated by accelerometer, gyro meter and complementary filter



6. Performance of the complementary filter

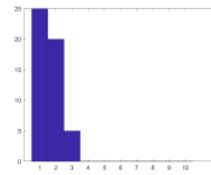
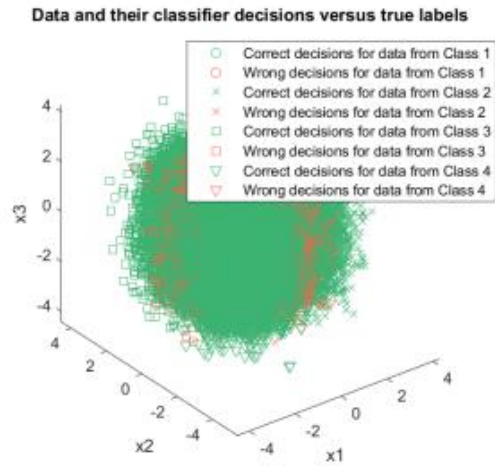
Nanoindentation Simulation



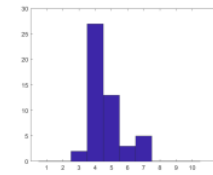
Bachelor Thesis

Multimillion-atom nanoindentation simulation of crystalline silicon carbide with spherical cavity.

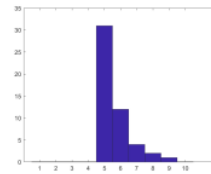
Machine Learning



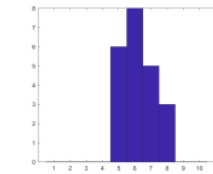
(a) N=100



(b) N=1000



(c) N=10000



(d) N=100000

Figure 8: kFold histogram

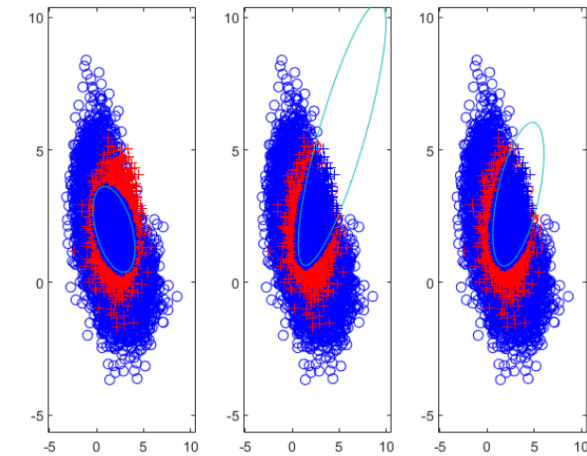


Figure 8: Validation Datasets with Decision Boundary by logistic-quadratic-function

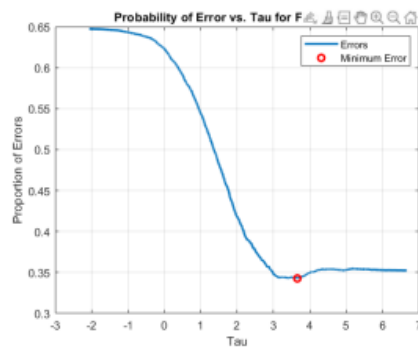


Figure 6: Fisher LDA Probability of Error vs. Tau

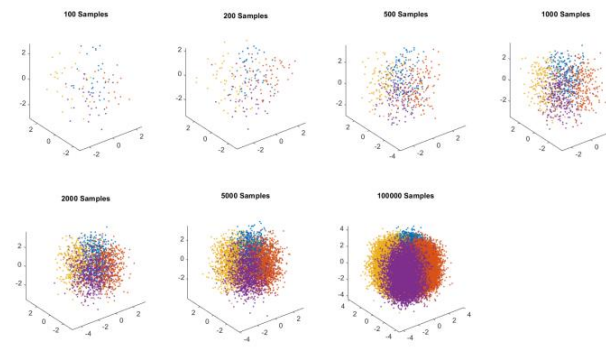


Figure 1: Data Distribution

Used Fisher LDA, ERM, MLE, MAP, Bayesian estimation, BIC and K-fold cross-validation to approximate model parameters. Trained 2-layer MLP.

Sensor and Navigation



Figure 5: 5 images using for mosaic



Figure 7: the panoramic mosaic of entire building

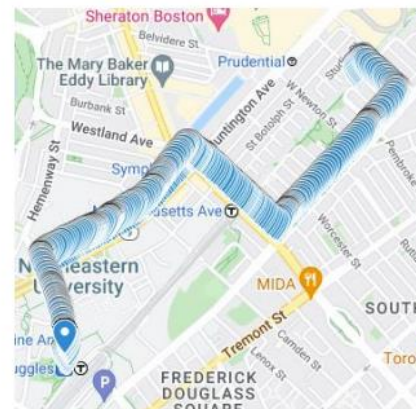


Figure 1: data present in Google Map

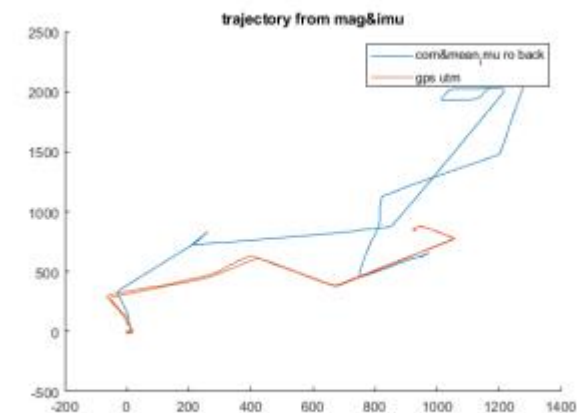


Figure 10: trajectory from calculation and gps

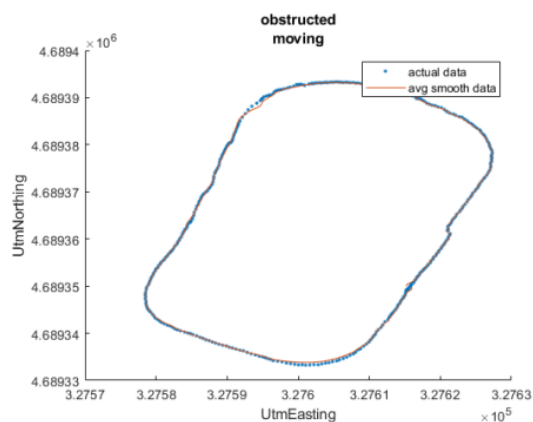


Figure 9: Obstructed moving data

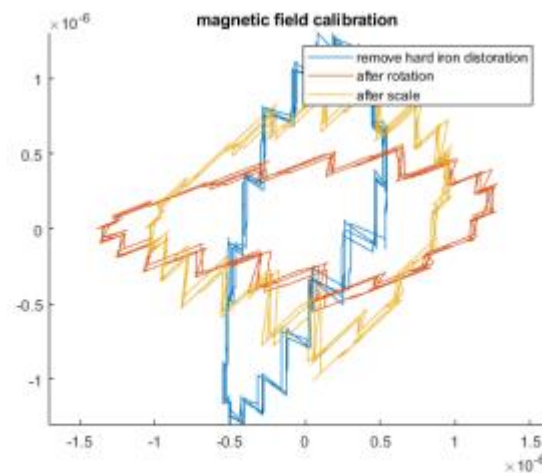


Figure 4: magnetometer calibration result

IMU, GNSS, magnetometer, camera and Lidar.

Internship

Unfortunately, I cannot directly show my work here for some reasons.

Here is the accomplished list of what I have done during my last internship.

- BLDC motor control MATLAB simulation without Simulink built-in Toolbox
- Drive programming and assist the SW team in familiarizing with the prototype machine
- CAN signal capture from i.MX8 and MathWorks Speedgoat
- Data Mining for Formative Study with the EM team to troubleshoot from unexpected performance
- Electrical schematic review with the EE team
- Presentation for internal training
- 8 pages document of knowledge transfer