## sensor fusion 3 22

## March 22, 2021

```
[25]: import math
     import numpy as np
     import pandas as pd
     import matplotlib.pyplot as plt
     TEST_NAME = "euler_angles"
     GRAVITY = 9.80665
     RAD_TO_DEG = 180 / math.pi
     dt = 1/960
[26]: # read test params from CSVP
     csvp = open(f"data/{TEST_NAME}.csvp")
     # create params array
     params = []
     for line in csvp: params.append(eval(line))
     params = np.array(params)
     print(params)
         1 3813
                                                     8 2000
                  0 300
                            0
                                      0 960
                                               96
                                                              92
                                                                   92
                                                                        63
        63
             14
                       86
                            86
                                      0 0 0
                                                     0 0
                                                               0
                                                                    0
                                                                         0
                  14
         0
              0
                  0]
[27]: # read data from CSV
     data = pd.read_csv(f"data/{TEST_NAME}.csv", names=["AccelX", "AccelY", "
      →"AccelZ", "GyroX", "GyroY", "GyroZ", "MagX", "MagY", "MagZ"],
      →index_col=False)
     sample_rate = params[7]
     # add time axis to data set
     time = np.arange(0, len(data)/sample_rate, 1/sample_rate)
     data.insert(0, "Time", time)
      # sign data
     data = data.applymap(lambda x: x-65535 if x > 32767 else x)
```

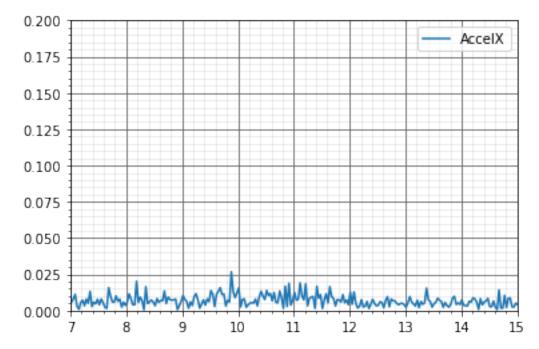
```
# apply accel sensitivity
acc_cols = ["AccelX", "AccelY", "AccelZ"]
acc_sens = params[9]
data[acc_cols] = data[acc_cols].applymap(lambda x: x * acc_sens * GRAVITY / ___
 →32768)
# apply gyro sensitivity
gyro_cols = ["GyroX", "GyroY", "GyroZ"]
gyro_sens = params[10]
data[gyro_cols] = data[gyro_cols].applymap(lambda x: x * gyro_sens / 32768)
# apply mag sensitivity
mag_cols = ["MagX", "MagY", "MagZ"]
mag_sens = 4800
data[mag_cols] = data[mag_cols].applymap(lambda x: x * mag_sens / 8192)
# create new mag dataframe by removing all NaNs
mag_data = data[mag_cols + ["Time"]].dropna()
# for some reason, the first mag data point is always erroneous, so remove it
mag_data = mag_data.iloc[1:]
# calculate offsets for each sensor (first 0.5s of data)
acc_offsets = data[acc_cols].head(480).mean()
gyro_offsets = data[gyro_cols].head(480).mean()
mag_offsets = mag_data[mag_cols].mean()
# apply offsets to each sensor (remove sensor bias)
# TODO: hold off on accel until actual IMU calibration is implemented
for i, axis in enumerate(gyro_cols):
    data[axis] = data[axis].map(lambda x: x - gyro_offsets[i])
for i, axis in enumerate(mag cols):
    data[axis] = data[axis].map(lambda x: x - mag_offsets[i])
print(acc_offsets, gyro_offsets, mag_offsets)
print(data.head(20))
AccelX
         0.579721
AccelY -0.262405
AccelZ -9.943548
dtype: float64 GyroX -1.288223
GyroY
       -3.983561
GyroZ
        14.923859
dtype: float64 MagX -316.491174
MagY
      132.152067
MagZ -164.916010
```

```
dtype: float64
                            AccelY
        Time
                 AccelX
                                       AccelZ
                                                    GyroX
                                                                GyroY
                                                                            GyroZ \
0
    0.000000 -0.299275 -0.936133 -10.685322
                                                -0.848007
                                                             2.518717 -12.238312
1
    0.001042
               1.431733 -0.639252 -10.120290
                                                                         9.673309
                                                 6.903458
                                                             0.809733
              0.555455
                         0.957681 -10.168174
2
    0.002083
                                                 4.095840
                                                            -8.467611
                                                                       -0.763702
3
    0.003125
              0.052672 -1.094150
                                    -9.512163
                                                 1.166153 -13.655599 -16.938019
4
    0.004167
               0.167594 -0.179565
                                    -9.890447
                                                -3.716660
                                                             3.983561
                                                                       -1.068878
5
    0.005208
              0.435745 -0.869095 -10.915165
                                                 0.922012
                                                            -1.936849
                                                                         5.217743
6
    0.006250
              0.699107 -0.940921
                                    -9.454702
                                                -0.542831
                                                            -1.814779
                                                                       17.180634
7
    0.007292
               1.115698
                        0.359130 -11.657368
                                                 3.180313
                                                            -7.674154
                                                                         1.006317
8
    0.008333
               1.288080 -1.127669
                                    -9.742007
                                                 0.738907
                                                             0.077311
                                                                       -8.637238
9
    0.009375
              0.541090
                         1.278504
                                    -9.512163
                                                -1.519394
                                                            -8.589681 -13.336945
    0.010417
               0.708684 -0.711078
                                    -9.028534
                                                 1.532364
                                                            -9.566243
                                                                        12.969208
10
              0.924162 -0.849942
                                    -9.818621
11
    0.011458
                                                -3.106308
                                                            -1.143392
                                                                        -8.820343
12
    0.012500
              0.967257
                         0.569820 -10.053253
                                                -0.481796
                                                            -4.988607
                                                                         3.264618
    0.013542
              0.885855
                         0.316035
                                    -9.220071
                                                -8.111191
                                                             4.838053 -11.261749
13
14
    0.014583
              0.612916
                         0.201113
                                    -9.449914
                                                -0.115585 -17.561849
                                                                         6.011200
15
    0.015625
               1.010353 -0.284910
                                    -9.224859
                                                -0.848007
                                                             1.420085
                                                                       13.640594
    0.016667
               0.397437 -1.319205
                                    -9.052476
                                                -2.007675
                                                            -1.387533
                                                                       -2.899933
16
17
    0.017708
              0.086191 -0.505177 -10.316615
                                                10.931778
                                                             5.692546
                                                                       -3.876495
                         0.459687
18
    0.018750
               0.119710
                                    -9.325415
                                                -1.824570
                                                            -3.157552
                                                                        -3.266144
                         0.612916
                                    -9.943119
19
    0.019792
              0.814029
                                                 2.020645
                                                             1.969401
                                                                       -3.998566
         MagX
                     MagY
                                 MagZ
0
    34.655237 -60.667692
                            45.970698
1
          NaN
                      NaN
                                  NaN
2
          NaN
                      NaN
                                  NaN
3
          NaN
                      NaN
                                  NaN
4
                      NaN
                                  NaN
          NaN
5
          NaN
                      NaN
                                  NaN
6
          NaN
                      NaN
                                  NaN
7
          NaN
                      NaN
                                  NaN
8
          NaN
                      NaN
                                  NaN
9
                                  NaN
          NaN
                      NaN
  -26.868201
                20.777620 -27.271490
10
11
          NaN
                      NaN
                                  NaN
12
                      NaN
                                  NaN
          NaN
13
          NaN
                      NaN
                                  NaN
14
          NaN
                      NaN
                                  NaN
                                  NaN
15
          NaN
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16
          NaN
                      NaN
                                  NaN
17
                      NaN
                                  NaN
          NaN
18
          NaN
                      NaN
                                  NaN
19
                                  NaN
          NaN
                      NaN
```

# denoising accelerometer

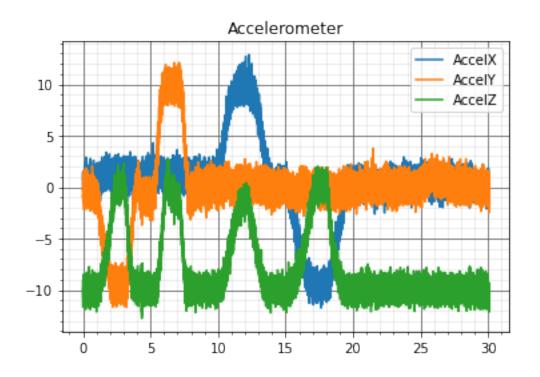
[39]:

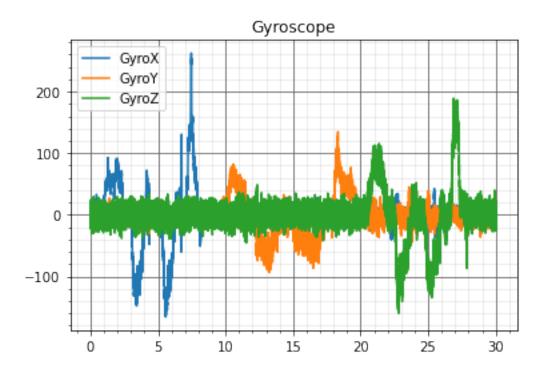
```
# calculate fft for AccelX
fourierTransform = np.fft.fft(data["AccelX"])/len(data["AccelX"])
tpCount = len(data["AccelX"])
values = np.arange(tpCount)
timePeriod = tpCount/sample_rate
frequencies = values/timePeriod
plt.plot(frequencies, abs(fourierTransform), label="AccelX")
# display the plot
plt.xlim(7,15)
plt.ylim(0,0.2)
plt.grid(b=True, which='major', color='#666666', linestyle='-')
plt.minorticks_on()
plt.grid(b=True, which='minor', color='#999999', linestyle='-', alpha=0.2)
plt.legend()
plt.show()
# apply simple moving average
# data[acc_cols] = data[acc_cols].rolling(window=100).mean().
 \rightarrow fillna(data[acc\_cols].iloc[49])
```

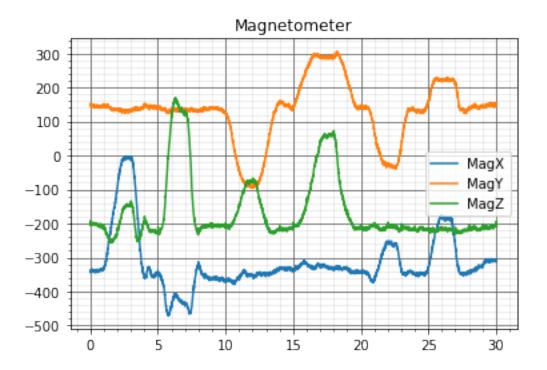


```
[29]: # plot acceleration
plt.plot(data["Time"], data["AccelX"], label="AccelX")
```

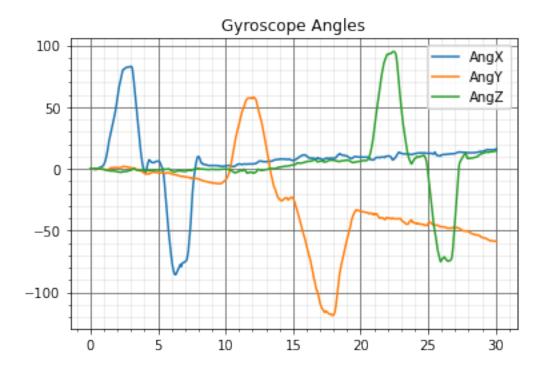
```
plt.plot(data["Time"], data["Accely"], label="Accely")
plt.plot(data["Time"], data["AccelZ"], label="AccelZ")
# display the plot
plt.title("Accelerometer")
plt.grid(b=True, which='major', color='#666666', linestyle='-')
plt.minorticks on()
plt.grid(b=True, which='minor', color='#999999', linestyle='-', alpha=0.2)
plt.legend()
plt.show()
# plot gyroscope
plt.plot(data["Time"], data["GyroX"], label="GyroX")
plt.plot(data["Time"], data["GyroY"], label="GyroY")
plt.plot(data["Time"], data["GyroZ"], label="GyroZ")
# display the plot
plt.title("Gyroscope")
plt.grid(b=True, which='major', color='#666666', linestyle='-')
plt.minorticks_on()
plt.grid(b=True, which='minor', color='#999999', linestyle='-', alpha=0.2)
plt.legend()
plt.show()
# plot magnetometer
plt.plot(mag_data["Time"], mag_data["MagX"], label="MagX")
plt.plot(mag_data["Time"], mag_data["MagY"], label="MagY")
plt.plot(mag_data["Time"], mag_data["MagZ"], label="MagZ")
# display the plot
plt.title("Magnetometer")
plt.grid(b=True, which='major', color='#666666', linestyle='-')
plt.minorticks_on()
plt.grid(b=True, which='minor', color='#999999', linestyle='-', alpha=0.2)
plt.legend()
plt.show()
```



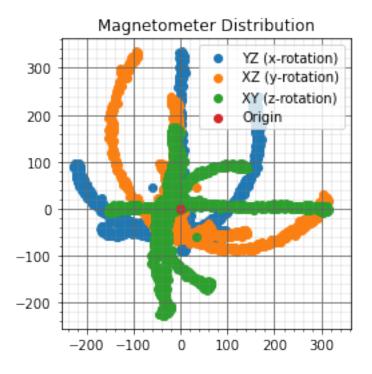




```
[30]: from scipy import integrate
      # calculate angles from gyroscope
      ang_x = integrate.cumtrapz(y=data["GyroX"], x=data["Time"], initial=0)
      ang_y = integrate.cumtrapz(y=data["GyroY"], x=data["Time"], initial=0)
      ang_z = integrate.cumtrapz(y=data["GyroZ"], x=data["Time"], initial=0)
      # plot gyroscope angles
      plt.plot(data["Time"], ang_x, label="AngX")
      plt.plot(data["Time"], ang_y, label="AngY")
      plt.plot(data["Time"], ang_z, label="AngZ")
      # display the plot
      plt.title("Gyroscope Angles")
      plt.grid(b=True, which='major', color='#666666', linestyle='-')
      plt.minorticks_on()
      plt.grid(b=True, which='minor', color='#999999', linestyle='-', alpha=0.2)
      plt.legend()
      plt.show()
```

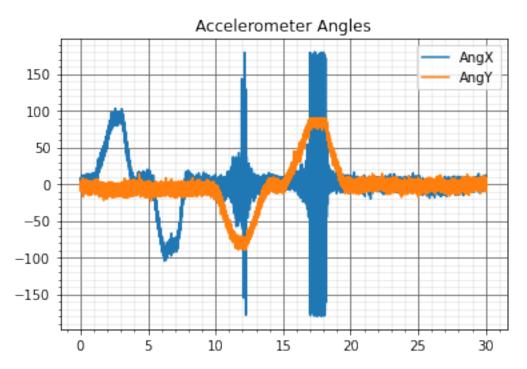


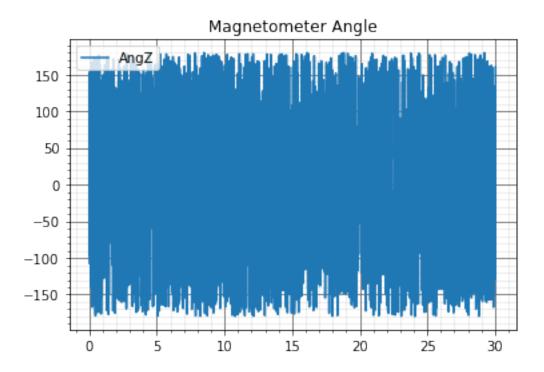
```
[31]: # hard iron calibration
      # plot x-axis rotation (MagY, MagZ)
      plt.scatter(data["MagY"], data["MagZ"], label="YZ (x-rotation)")
      # plot y-axis rotation (MaqX, MaqZ)
      plt.scatter(data["MagX"], data["MagZ"], label="XZ (y-rotation)")
      # plot z-axis rotation (MagX, MagY)
      plt.scatter(data["MagX"], data["MagY"], label="XY (z-rotation)")
      # plot origin (0,0)
      plt.scatter([0], [0], label="Origin")
      # display the plot
      plt.title("Magnetometer Distribution")
      plt.axis("square")
      plt.grid(b=True, which='major', color='#666666', linestyle='-')
      plt.minorticks_on()
      plt.grid(b=True, which='minor', color='#999999', linestyle='-', alpha=0.2)
      plt.legend()
      plt.show()
```



```
[32]: # OLD: calculate angles from accelerometer
     # acc ang x = np.arctan2(-data["AccelY"], -data["AccelZ"]) * RAD TO DEG
     # acc_ang_y = np.arctan2(-data["AccelX"], np.sqrt(data["AccelY"]**2 +_
      \rightarrow data["AccelZ"]**2)) * RAD_TO_DEG
     MU = 0.01
     # EXPERIMENTAL ANGLE CALCULATIONS
     # TODO: added 2 negatives to acc_ang_x, doesn't work without it: why?
     acc_ang_x = np.arctan2(-data["AccelY"], -np.sign(data["AccelZ"]) * np.
      acc_ang_y = np.arctan2(-data["AccelX"], np.sqrt(data["AccelY"]**2 +__
      →data["AccelZ"]**2)) * RAD_TO_DEG
     # calculate z-angle (yaw) from accelerometer & magnetometer
     M_x = mag_data["MagX"] * np.cos(acc_ang_y) + mag_data["MagZ"] * np.
      →sin(acc_ang_y)
     M_y = mag_data["MagX"] * np.sin(acc_ang_x) * np.sin(acc_ang_y) + ___
      →mag_data["MagY"] * np.cos(acc_ang_x) - mag_data["MagZ"] * np.sin(acc_ang_x)_⊔
      →* np.cos(acc_ang_y)
     # remove NaNs from mag calculations
     M_x = M_x[\sim np.isnan(M_x)]
     M_y = M_y[\sim np.isnan(M_y)]
```

```
mag_ang_z = np.arctan2(-M_y, M_x) * RAD_TO_DEG
# plot accelerometer angles
plt.plot(data["Time"], acc_ang_x, label="AngX")
plt.plot(data["Time"], acc_ang_y, label="AngY")
# display the plot
plt.title("Accelerometer Angles")
plt.grid(b=True, which='major', color='#666666', linestyle='-')
plt.minorticks_on()
plt.grid(b=True, which='minor', color='#999999', linestyle='-', alpha=0.2)
plt.legend()
plt.show()
# plot mag+accel z-axis angle
plt.plot(mag_data["Time"], mag_ang_z, label="AngZ")
# display the plot
plt.title("Magnetometer Angle")
plt.grid(b=True, which='major', color='#666666', linestyle='-')
plt.minorticks_on()
plt.grid(b=True, which='minor', color='#999999', linestyle='-', alpha=0.2)
plt.legend()
plt.show()
```



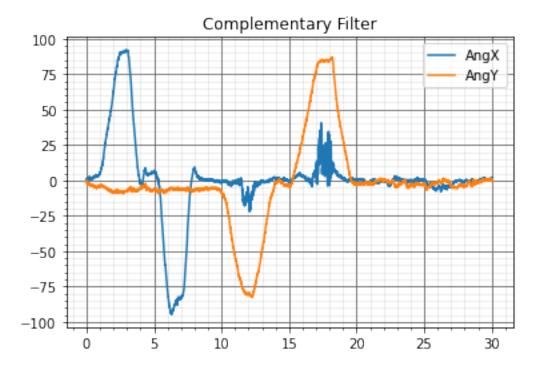


```
[33]: # complementary filter
      HP_weight = 0.98
      LP_weight = 0.02
      # create empty array w/ 3 axes
      # set first elements to 0 (will be removed)
      cf_ang = [[0],[0],[0]]
      # group all axes of calculated angles together
      # TODO: add z axis once it matches shape of the others
      gyro_ang = np.array([data["GyroX"].to_numpy(),data["GyroY"].to_numpy()])
      calc_ang = np.array([acc_ang_x, acc_ang_y])
      # pair the calculated arrays for each axis together and loop
      # TODO: to add z-axis, just add it to the tuple and as a zip() param
      for i, (gyro_arr, calc_arr) in enumerate(zip(gyro_ang, calc_ang)):
          # pair the individual samples together and loop
          for gyro, calc in zip(gyro_arr, calc_arr):
              cf_ang_prev = cf_ang[i][-1]
              cf_ang_samp = HP_weight * (cf_ang_prev + gyro * dt) + LP_weight * calc
              cf_ang[i].append(cf_ang_samp)
```

```
# remove initial 0 value
cf_ang[0].pop(0)
cf_ang[1].pop(0)

plt.plot(data["Time"], cf_ang[0], label="AngX")
plt.plot(data["Time"], cf_ang[1], label="AngY")

# display the plot
plt.title("Complementary Filter")
plt.grid(b=True, which='major', color='#666666', linestyle='-')
plt.minorticks_on()
plt.grid(b=True, which='minor', color='#999999', linestyle='-', alpha=0.2)
plt.legend()
plt.show()
```



[]: