

s6_t0

July 27, 2022

```
[ ]: import matplotlib.pyplot as plt
import pandas as pd
import numpy as np
import msgpack as mp
import msgpack_numpy as mpn
import os
import sys
import cv2
import glob
# directory reach
directory = os.path.abspath('')
append_path = os.path.dirname(directory) # go two folders up
from pykinect2 import PyKinectRuntime
from pykinect2 import PyKinectV2

kinectColor = PyKinectRuntime.PyKinectRuntime(PyKinectV2.FrameSourceTypes_Color)
import pickle
# setting path
sys.path.append(append_path)
from cv2 import aruco
from datetime import timedelta

from support.pd_support import *
from support.calculations_support import *
from support.generate_ar_data import get_ar_pose_data, detect_ar_markers, ↵
    ↪add_time_col
from support.camera_calibration import *
from support.ar_calculations import *
from support.mecanum_calculations import *
```

```
[ ]: _pth_calibration = r"C:
    ↪\Users\CMC\Dropbox\mira\mira_vellore\splitVideos\ARUCXXXXMARKU010120000000XXXXXXXXXX\Session
    ↪20-07-22_10-53-15_690"
```

```
[ ]: """ importing calibration file """

calib_pth = os.path.join(_pth_calibration, "AR_CALIBRATION.msgpack")
```

```

_calib_file = open(calib_pth, "rb")
unpacker = mp.Unpacker(_calib_file, object_hook=mpn.decode)
_calib = []
for unpacked in unpacker:
    _calib.append(unpacked)

cameraMatrix = _calib[0][0]
distCoeffs = _calib[0][1]

yPos = 112
xPos = 274

yRes = 736
xRes = 864
scalling = 2
yRes

```

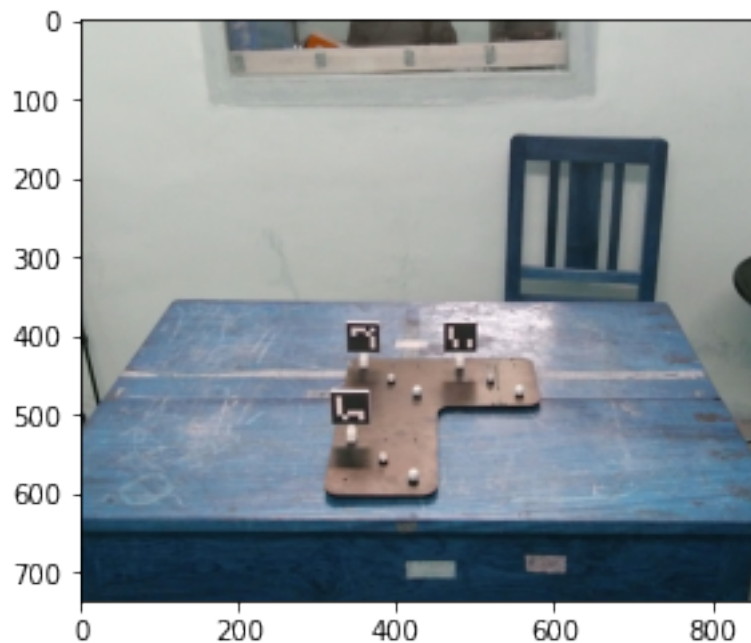
[]: 736

```

[ ]: """ importing l frame data """
calib_file = open("new_l_frame2.pickle", "rb")
calib_img = pickle.load(calib_file)
calib_file.close()
plt.imshow(calib_img)

```

[]: <matplotlib.image.AxesImage at 0x17bb4828b88>



```

[ ]: """detect the aruco markers"""
aruco_dict = aruco.getPredefinedDictionary(aruco.DICT_ARUCO_ORIGINAL)
arudo_parameters = aruco.DetectorParameters_create()

corners, ids, rejectedImgPoints = detect_ar_markers(calib_img,
↳arudo_parameters, aruco_dict)

rotation_vector, translation_vector, _ = cv2.aruco.
↳estimatePoseSingleMarkers(corners, 0.05, cameraMatrix, distCoeffs)
rotation_vector

"""

Idx: 6 - zvector
Idx: 9 - orgin
Idx: 10 - xvector
"""

print(ids)
zvec = translation_vector[0][0]
zvec = np.reshape(zvec, (3, 1))
org = translation_vector[2][0]
org = np.reshape(org, (3, 1))
xvec = translation_vector[1][0]
xvec = np.reshape(xvec, (3, 1))

translation_vector
zvec
rotMat = calculate_rotmat(xvec, zvec, org)
rotMat

t_xvec = xvec - org

translation_correction = np.array([0.045, -0.05, 0.045]).reshape(3, 1) # adding
↳the corrections in the new L frame

rotMat.T@t_xvec + translation_correction

[[ 6]
 [10]
 [ 9]]

```

```
[ ]: array([[ 0.19988459],
          [-0.05      ],
          [ 0.045      ]])
```

```
[ ]: # _data_pth = r"C:
      ↪ \Users\CMC\Dropbox\mira\mira_vellore\splitVideos\ARUCXXXXMARKU010120000000XXXXXXXXXX\Session
      ↪ 22-07-22_12-42-02_505" # session 5 test 1
      _data_pth = r"C:
      ↪ \Users\CMC\Dropbox\mira\mira_vellore\splitVideos\ARUCXXXXMARKU010120000000XXXXXXXXXX\Session
      ↪ 26-07-22_10-27-13_120" # session 6 test 0

      df = get_ar_pose_data(_data_pth, cameraMatrix, distCoeffs, process_raw=True)
      df
```

returning dataframe

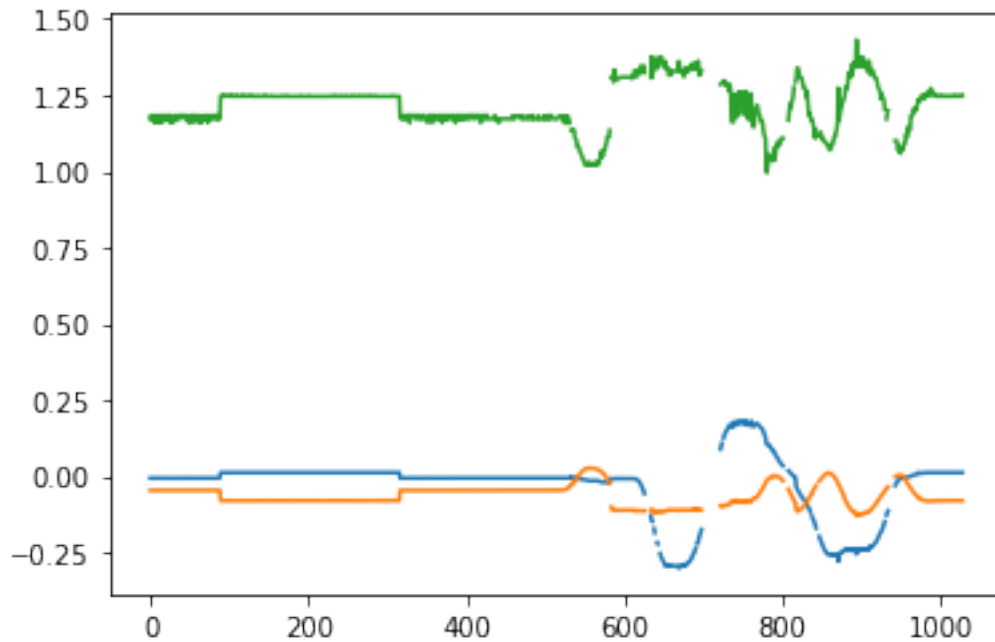
```
[ ]:      frame_id      x      y      z      yaw      pitch      roll
      0          5.0 -0.005015 -0.045195 1.172664 -0.053384 3.029512 0.795643
      1          5.0 -0.005337 -0.045384 1.184101 0.071220 -3.062403 -0.731841
      2          5.0 -0.005015 -0.045195 1.172664 -0.053384 3.029512 0.795643
      3          5.0 -0.005015 -0.045195 1.172664 -0.053384 3.029512 0.795643
      4          5.0 -0.005337 -0.045384 1.184101 0.071220 -3.062403 -0.731841
      ...      ...      ...      ...      ...      ...      ...
      1023      5.0 0.013014 -0.078591 1.246457 -0.055127 2.969929 0.735210
      1024      5.0 0.013026 -0.079034 1.249041 -0.066245 2.992277 0.776526
      1025      5.0 0.013026 -0.079034 1.249041 -0.066245 2.992277 0.776526
      1026      5.0 0.013014 -0.078591 1.246457 -0.055127 2.969929 0.735210
      1027      5.0 0.013026 -0.079034 1.249041 -0.066245 2.992277 0.776526
```

[1028 rows x 7 columns]

```
[ ]: df = add_time_col(df, _data_pth)
```

```
[ ]: plt.plot(df["x"])
      plt.plot(df["y"])
      plt.plot(df["z"])
```

```
[ ]: [<matplotlib.lines.Line2D at 0x17bb535f708>]
```



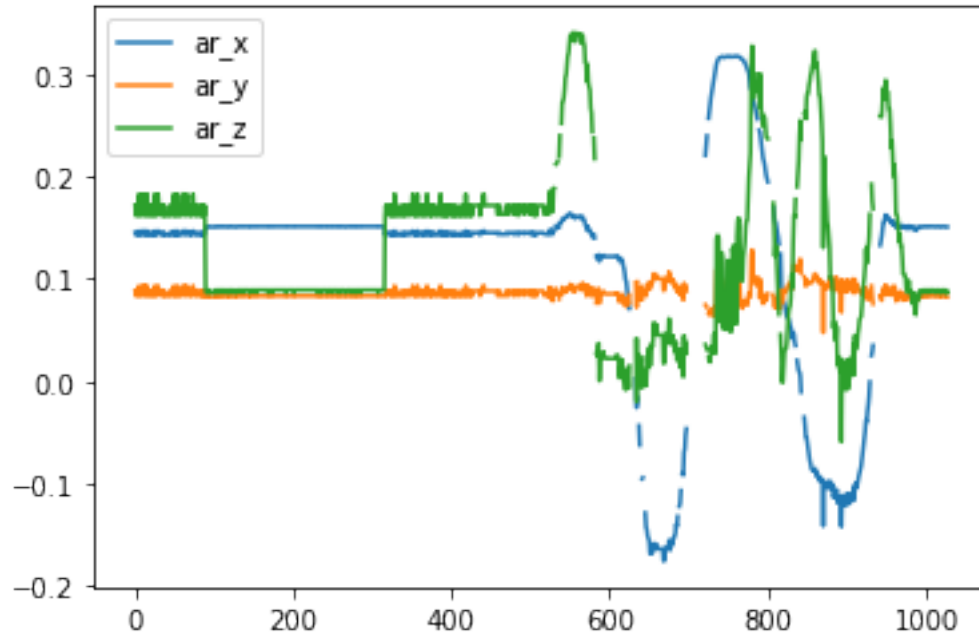
```
[ ]: ar_df = pd.DataFrame(columns=["x", "y", "z"])

for i in range(len(df)):
    _val = df[["x", "y", "z"]].iloc[i]
    _val = np.reshape(_val.values, (3, 1))
    _val_t = _val - org
    _xyz = rotMat.T*_val_t + translation_correction
    _xyz = _xyz.reshape(3)
    ar_df.loc[i] = _xyz
ar_df["time"] = df["time"]
```

```
[ ]: ar_df["time"] = pd.to_datetime(ar_df["time"])
```

```
[ ]: plt.plot(ar_df["x"], label="ar_x")
plt.plot(ar_df["y"], label="ar_y")
plt.plot(ar_df["z"], label="ar_z")
plt.legend()
```

```
[ ]: <matplotlib.legend.Legend at 0x17bb5906548>
```



```
[ ]: """ getting mc df"""
mc_df, mt_t = read_df_csv("ar_s6_t0.csv")

mc_df = add_datetime_col(mc_df, mt_t, "seconds")
plt.plot(mc_df["org_x"], label="mc_org_x")
plt.plot(mc_df["org_y"], label="mc_org_y")
plt.plot(mc_df["org_z"], label="mc_org_z")
plt.legend()

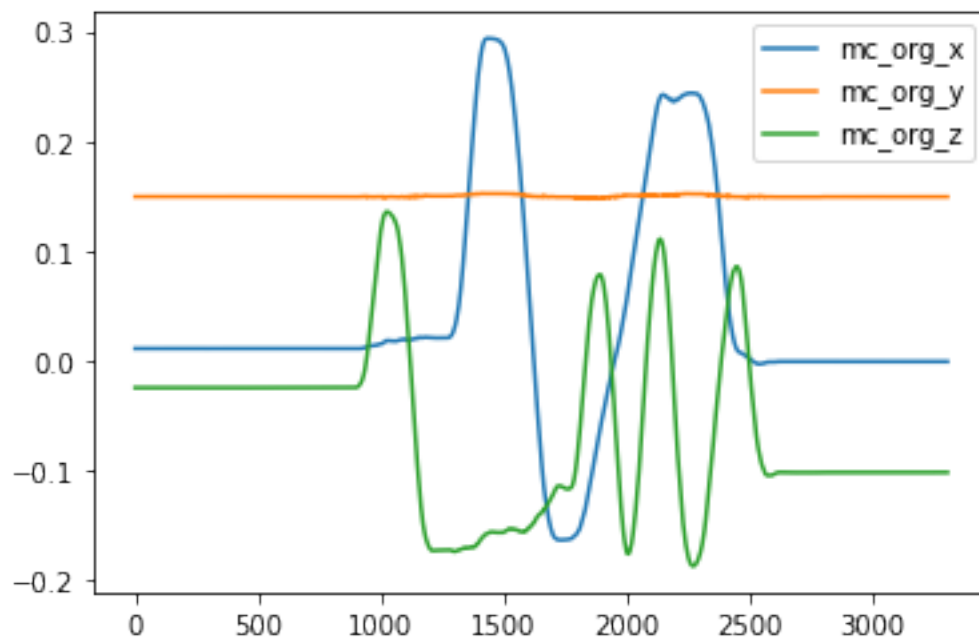
mc_df
```

```
[ ]:
```

	frame	seconds	org_x	org_y	org_z	xvec_x	xvec_y	\
0	0	0.00	0.011143	0.149485	-0.024517	0.143142	0.149919	
1	1	0.01	0.011144	0.149493	-0.024511	0.143157	0.149925	
2	2	0.02	0.011150	0.149491	-0.024513	0.143153	0.149866	
3	3	0.03	0.011144	0.149497	-0.024518	0.143150	0.149924	
4	4	0.04	0.011151	0.149489	-0.024520	0.143177	0.149888	
...	
3304	3304	33.04	-0.000707	0.149333	-0.101714	0.131616	0.150364	
3305	3305	33.05	-0.000742	0.149333	-0.101692	0.131609	0.150609	
3306	3306	33.06	-0.000714	0.149316	-0.101726	0.131606	0.150610	
3307	3307	33.07	-0.000716	0.149316	-0.101725	0.131616	0.150367	
3308	3308	33.08	-0.000731	0.149350	-0.101699	0.131627	0.150367	
			xvec_z	zvec_x	zvec_y	zvec_z	time	
0			-0.035744	0.033077	0.150649	0.054005	2022-07-26 10:27:35.849	

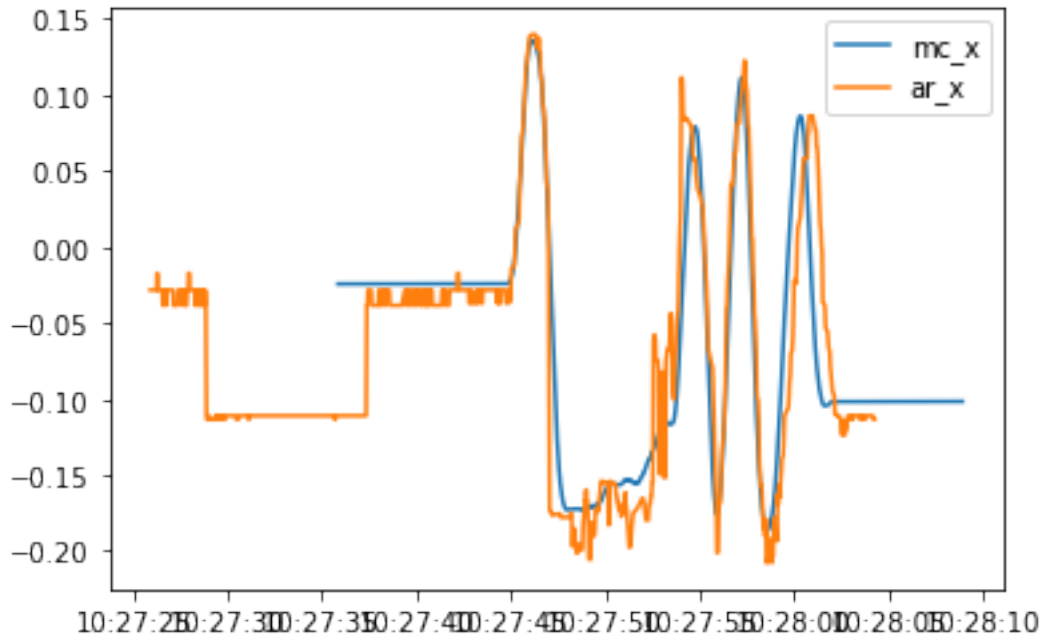
1	-0.035728	0.033075	0.150652	0.054001	2022-07-26	10:27:35.859
2	-0.035750	0.033079	0.150656	0.054007	2022-07-26	10:27:35.869
3	-0.035729	0.033093	0.150635	0.053989	2022-07-26	10:27:35.879
4	-0.035746	0.033082	0.150630	0.053976	2022-07-26	10:27:35.889
...
3304	-0.106933	0.017551	0.150260	-0.022183	2022-07-26	10:28:08.889
3305	-0.106972	0.017345	0.150243	-0.022308	2022-07-26	10:28:08.899
3306	-0.106979	0.017545	0.150280	-0.022169	2022-07-26	10:28:08.909
3307	-0.106934	0.017557	0.150264	-0.022183	2022-07-26	10:28:08.919
3308	-0.106927	0.017361	0.150256	-0.022305	2022-07-26	10:28:08.929

[3309 rows x 12 columns]



```
[ ]: plt.plot(mc_df["time"], mc_df["org_z"], label="mc_x")
plt.plot(ar_df["time"] - timedelta(seconds=8.4), ar_df["z"].interpolate().
    ↪rolling(3).median()-0.2, label="ar_x")
plt.legend()
```

```
[ ]: <matplotlib.legend.Legend at 0x17bb81b6688>
```



```
[ ]: """reading wheels data"""
_csv_pth = os.path.join(_data_pth, "imu01.csv")
cart_df = pd.read_csv(_csv_pth)

cart_df = add_datetime_diff(cart_df, mt_t, "sync", "mils", False)
cart_df

imu_df = cart_df[["sync", "mils", "ax", "ay", "az", "gx", "gy", "gz", "time"]]
```

```
[ ]: """ this calculates the robots position using mecanum wheels"""

radius = 47.5/1000
lx = 79 #half of the distance between the wheels
ly = 122.5/2

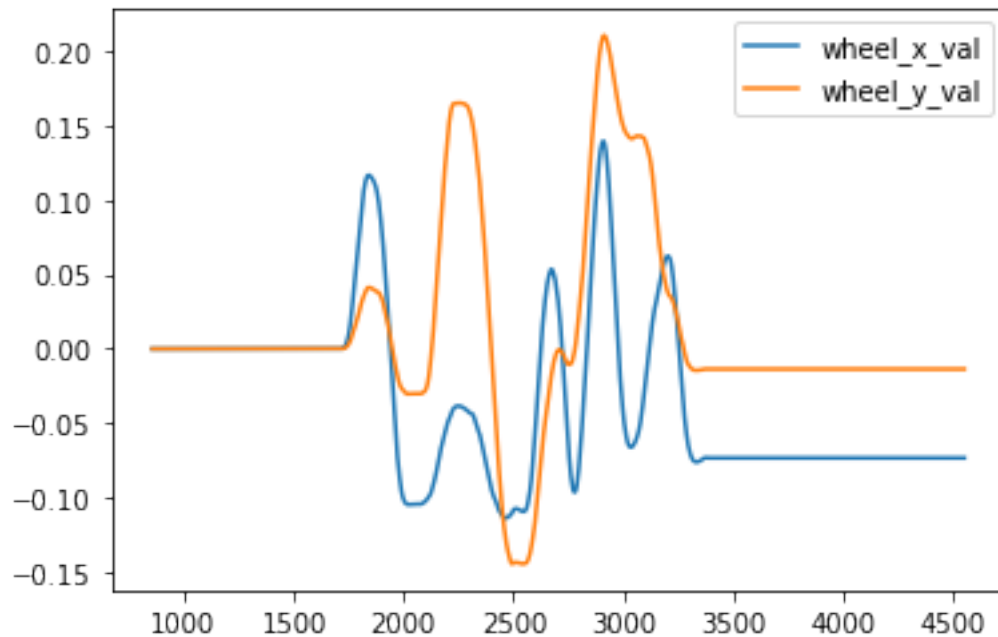
lx = lx/1000
ly = ly/1000

cart_df = set_zero(cart_df, column_name=["e_fr", "e_fl", "e_rr", "e_rl"])
cart_df, c_names = get_angular_velocity(cart_df, column_name=["e_fr", "e_fl",
↳ "e_rr", "e_rl"])
cart_df, c_names = get_directional_velocity(cart_df, c_names, radius, lx, ly)
cart_df, c_names = get_position(cart_df)
```



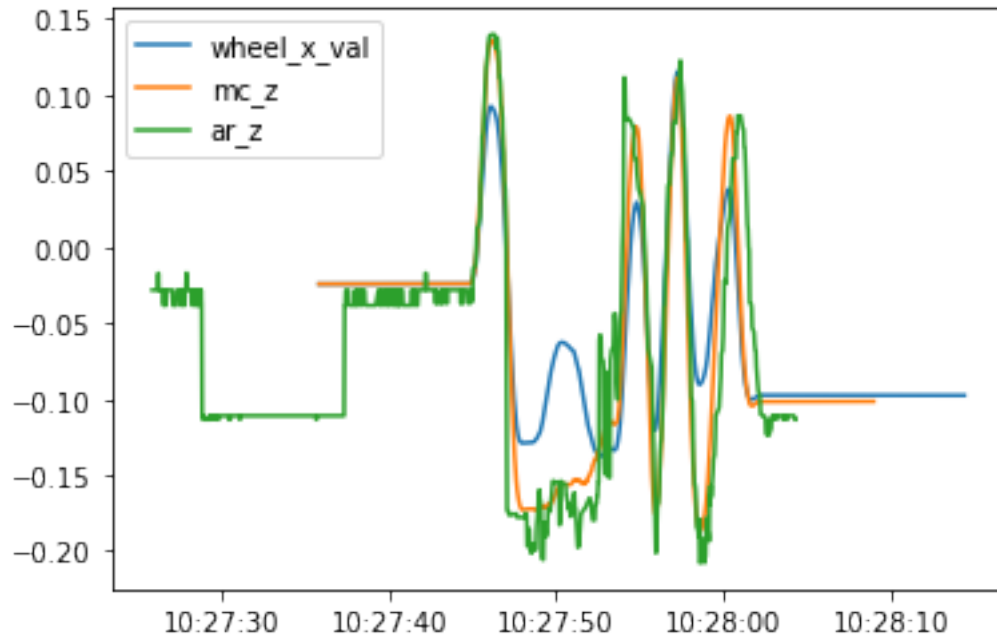
```
plt.plot(cart_df["x_val"], label="wheel_x_val")
plt.plot(cart_df["y_val"], label="wheel_y_val")
plt.legend()
```

[]: <matplotlib.legend.Legend at 0x17bb5991788>



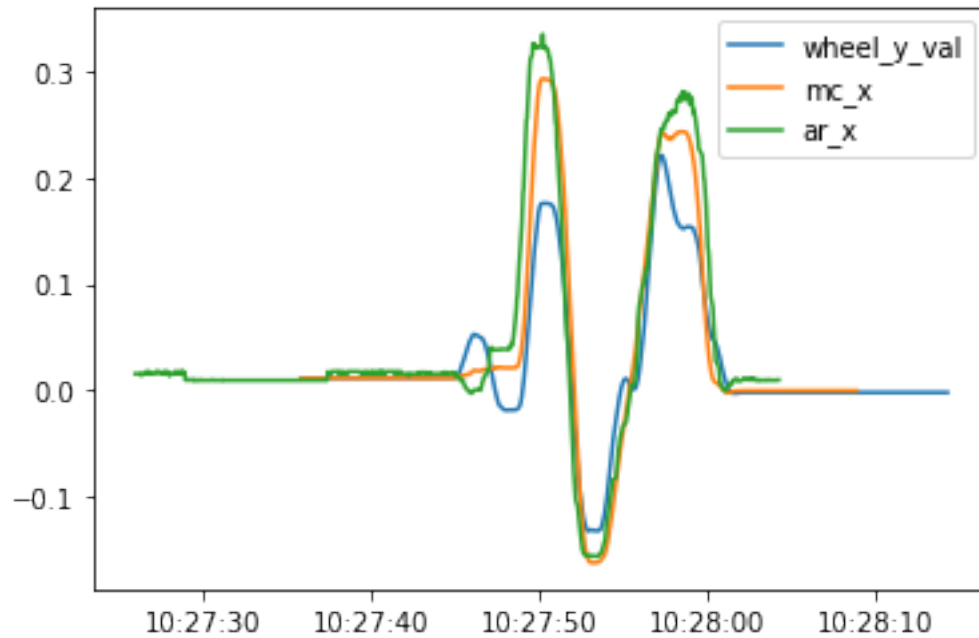
```
[ ]: plt.plot(cart_df["time"], cart_df["x_val"]-(cart_df["x_val"].iloc[0] -
↳mc_df["org_z"].iloc[0]), label="wheel_x_val")
plt.plot(mc_df["time"], mc_df["org_z"], label="mc_z")
plt.plot(ar_df["time"] - timedelta(seconds=8.4), ar_df["z"].interpolate().
↳rolling(3).median()-0.2, label="ar_z") #filtered data
plt.legend()
```

[]: <matplotlib.legend.Legend at 0x17bbabcf488>



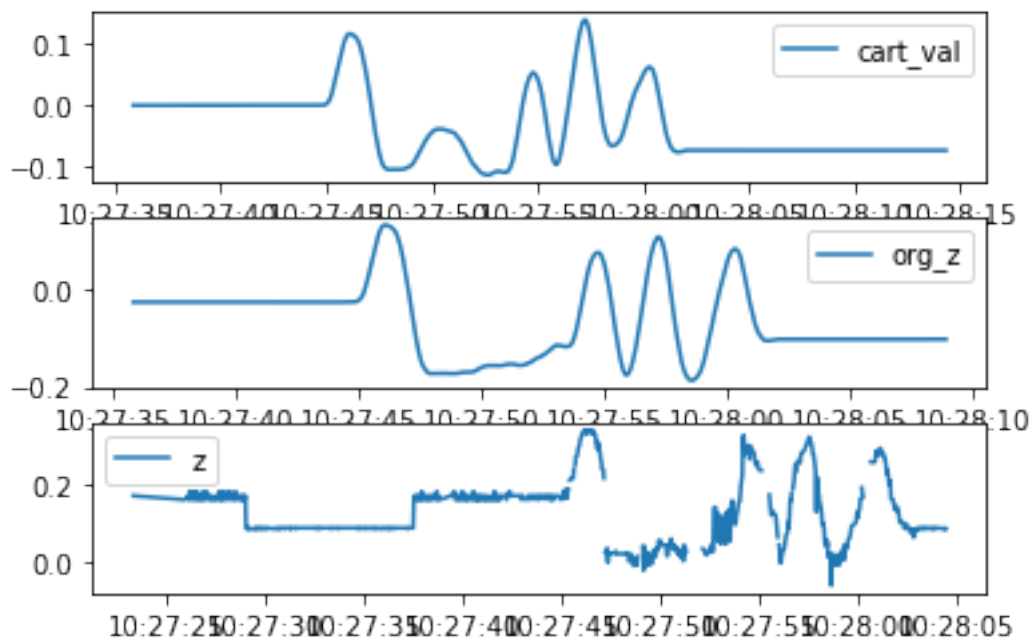
```
[ ]: plt.plot(cart_df["time"], cart_df["y_val"]-(cart_df["y_val"].iloc[0] -
↳mc_df["org_x"].iloc[0]), label="wheel_y_val")
plt.plot(mc_df["time"], mc_df["org_x"], label="mc_x")
plt.plot(ar_df["time"] - timedelta(seconds=8.4), (ar_df["x"].interpolate().
↳rolling(3).median()-0.16)*-1, label="ar_x") #filtered data
plt.legend()
```

```
[ ]: <matplotlib.legend.Legend at 0x17bbc307108>
```



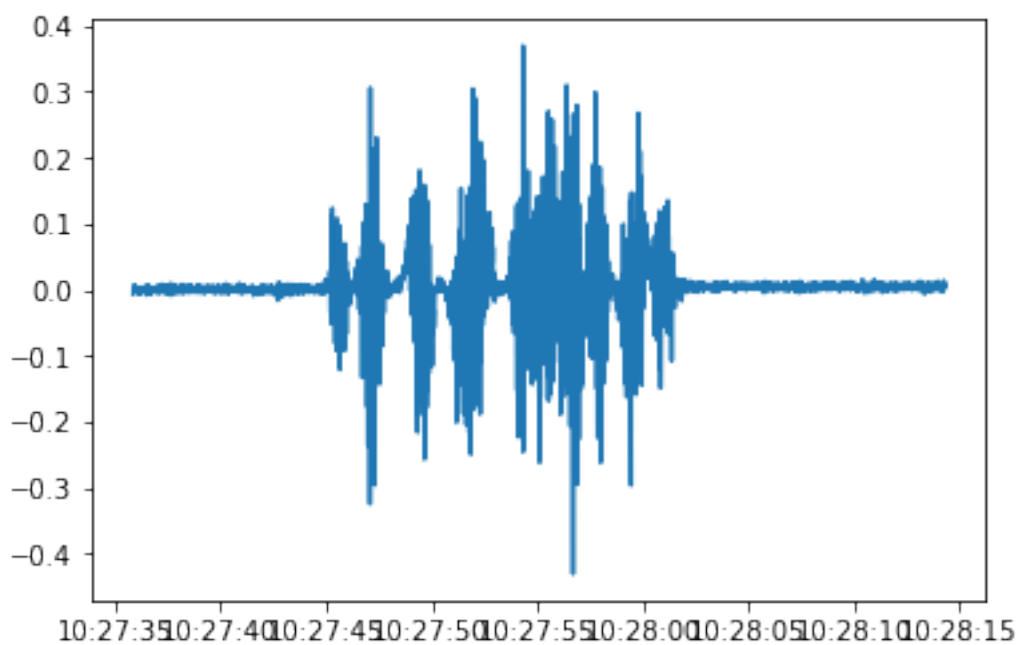
```
[ ]: plt.subplot(3, 1, 1)
plt.plot(cart_df["time"], cart_df["x_val"], label="cart_val")
plt.legend()
plt.subplot(3, 1, 2)
plt.plot(mc_df["time"], mc_df["org_z"], label="org_z")
plt.legend()
plt.subplot(3, 1, 3)
plt.plot(ar_df["time"]- timedelta(seconds=8.2), ar_df["z"], label="z")
plt.legend()
```

```
[ ]: <matplotlib.legend.Legend at 0x17bb7bdfec8>
```



```
[ ]: plt.plot(imu_df["time"], imu_df["ax"], label="ax")
```

```
[ ]: [ <matplotlib.lines.Line2D at 0x17bbc1f9bc8> ]
```



[]: