

Mobile Health Practical 1

Introduction to Signal Processing

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Goal of the Practical

- Intro to signal processing in Python
- Introduction to IMU data
- Python Notebook in Colab for data processing
 - Load and organise data
 - Data visualisation
 - Signal processing in Python
- Learn about the upcoming assignment



- You should be already familiar with the following concepts:
 - Analogue and digital signal
 - Nyquist theorem
 - Discrete Fourier transform and Fast Fourier transform
 - Spectrograms
 - Basics of filtering
- Most common tools for digital signal processing Python, MATLAB, C++





Python tools necessary for this practical

Data loader

Data organiser





Data visualiser

Actual signal processing



Python tools necessary for this practical

Data loader

Data organiser

- Need to load:
 - sensor data could be a CSV, WAV, etc.
 - metadata typically a CSV







Python tools necessary for this practical

Data loader

Data organiser

Display the data and transform it if needed lacksquare









- Library for data manipulation and analysis
- Supports and allows complex operations on large, multi-
- Library for working with regular expressions



Python tools necessary for this practical



Data organiser

- Always a good idea not to work with the data blindly
- The most straightforward library matplotlib.pyplot







Actual signal processing

import matplotlib.pyplot as plt

```
x = [1, 2, 3, 4, 5]
y = [2, 4, 6, 8, 10]
plt.plot(x, y)
plt.show()
```



Python tools necessary for this practical

Data loader

Data organiser

import numpy as np from scipy import signal

• Multiple libraries with predefined functions

fs = 10e3f = 1e3







Actual signal processing

generate a 1kHz sine wave

t = np.linspace(0, 1, fs, endpoint=False) x = np.sin(2 * np.pi * f * t)









Infinity AI IMU Fitness Dataset

Open Source Synthetic Dataset for Fitness Applications

- Paired video and IMU data (9 DoF: angular positions)
- We've extracted a subset of the data
- 20Hz sampling frequency

[1] https://marketplace.infinity.ai/products/imu-fitness-basic-dataset





Infinity AI IMU Fitness Dataset

Open Source Synthetic Dataset for Fitness Applications



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- 10 exercises
- 10 samples each
- 3 files per sample:
 - CSV
 - JSON
 - MP4

Importing Libraries

- **import** requests Allows to send HTTP requests
- **import zipfile** Enables work with ZIP archives
- Provides a way of using operating system (OS) dependent functionality
- **import numpy as np** → NUMerical PYthon
- **import** scipy SClentific PYthon, provides functions for stats and signal processing
- import matplotlib.pyplot as plt -> Plotting library



- **import pandas as pd** Used for working with datasets

Loading data

	rep_count_from_intermediate	rep_count_from_start	ref_xy_rotation	time	rotation_matrix_m11	rotation_matrix_m12	rotation_matrix_m13	rotation_matrix_m21	rotation_matrix_m22	rotation_matrix_m23	rotation_matrix_m31	rotation_matrix_m32	ro
0	0.4736842105263160	0.0	4.8182613644283	0.0	0.31345245242118800	-0.21886013448238400	-0.9240388870239260	0.09758635610342030	0.9753504395484920	-0.1979101151227950	0.9445763826370240	-0.028138162568211600	0
1	0.49027919940091000	0.016594988874594500	4.8182613644283	0.05	0.3190750181674960	-0.2161204069852830	-0.9227584600448610	0.09248825907707210	0.9761050343513490	-0.19663383066654200	0.9432057738304140	-0.022603364661335900	
2	0.5068740668841270	0.033189856357810800	4.8182613644283	0.1	0.33161893486976600	-0.21445204317569700	-0.9187160134315490	0.08539441972970960	0.9766470193862920	-0.19715073704719500	0.9395406246185300	-0.013074328191578400	0.3
3	0.5234688116608430	0.049784601134527000	4.8182613644283	0.150000	0.3504527509212490	-0.21328036487102500	-0.9119728207588200	0.07712976634502410	0.976990818977356	-0.19884651899337800	0.9333991408348080	-0.0006539252935908740	0.3
4	0.5400634317583760	0.06637922123206050	4.8182613644283	0.2	0.3798833191394810	-0.2133406549692150	-0.9000969529151920	0.06752893328666690	0.976841151714325	-0.20303015410900100	0.9225663542747500	0.016345201060175900	0.3
5	0.5566579245464840	0.08297371402016790	4.8182613644283	0.25	0.4221969544887540	-0.21402782201767000	-0.8808756470680240	0.05850962549448010	0.9761358499526980	-0.2091301530599590	0.9046139717102050	0.03675442561507230	0.4
6	0.5732522867373600	0.09956807621104460	4.8182613644283	0.300000	0.47755521535873400	-0.21485623717308000	-0.851925790309906	0.05239385738968850	0.974876880645752	-0.21649464964866600	0.8770380020141600	0.05875243619084360	C
	0.5898465143856410	0.1161623038593260	4.8182613644283		0.5444672703742980	-0.2160136103630070	-0.8104896545410160	0.050907816737890200	0.9729984998703000	-0.2251272201538090	0.8372358083724980	0.08131415396928790	



 otation_matrix_m33

 0.3270837366580960

 0.331439346075058

 0.34218764305114700

 0.35883939266204800

 0.38549232482910200

 0.42464426159858700

 0.4768146872520450

 0.5407626628875730

There are few problems with **Euler angles** (XYZ axes):

- Gimbal lock
- Cannot do smooth interpolation
- Computationally inefficient
- Unintuitive for rotations

So instead can use **quaternions** — may make mathematical representation for 3D rotations.





Image credit: Mark Hughes

x- Ax 15

AXIS

Quaternions

$$a + bi + cj + dk$$

 a, b, c, d — real numbers
 i, j, k — basic quaternions

$$i^2 = j^2 = k^2 = -1$$

Basic quaternions can be interpreted as unit-vectors pointing along X, Y and Z. Can convert rotation matrices easily to quaternions or Euler angles.



Multiplication table





Converting quaternions to Euler angles

Can convert rotation matrices easily to quaternions or Euler angles.





We've loaded the data and converted it to a form we understand. How do we now start to analyse and process it?





Time series visualisation — bicep curls





Time series visualisation — magnitude vector



Time series visualisation



... continued on next slide

Time series visualisation

... continuing from previous slide



Frequency spectrum







Frequency visualisation (via FFT)





Frequency visualisation (via FFT)







Frequency visualisation (via FFT)







Lets make it noisier

- We're going to add some artificial noise to our very clean synthetic dataset
- Add a 2Hz wave and a 4Hz wave



Artificial noisy signal

More indicative of real world data





Observations

- Noisy peaks overlaid with the signal of interest
- DC offset in the signal (centred around ~0.5)



FFT (again)

- Peaks at 2Hz and 4Hz interfering with the signal
- How can we remove the interference and the DC offset?





We've visualised the data. Now lets preprocess it!



Filtering

- Remove the **DC** offset first
- High pass filter
- From the FFT, we can see that we have peaks of interest from ~0.25Hz
 - We need to apply the filter at a **lower cutoff** than the signals of interest
 - Try some frequencies and look at the resulting FFTs





High Pass Filter



Apply High Pass Filter



Filtering

- Now lets remove the 2Hz and 4Hz noise
- Low pass filter
- From the FFT, we can see that our maximum peak of interest is at ~1Hz
 - We need to apply the filter at a higher cutoff than the signals of interest
 - Try some frequencies and look at the resulting FFTs



Pass Band -Stop Baı 0dB -3dB

Low Pass Filter



Apply Low Pass Filter



Signal after filtering



What is this peak?

Dataset analysis — how much variability across samples?

Assignment

- Uploaded on Moodle
- Due **20 February**
- Audio dataset of heart sounds
- Preprocessing techniques for IMU can also be used for audio!
- Feature extraction will differ between IMU and audio

