Luka Alhonen Exercise 2 1 Non-programming Tasks 1.1 Learning diary Reading data and synchronisation Because of the diversity of IoT applications, data from sensors is accessed in many ways, however the three main ways of accessing sensor data are; Streaming, Polling and Event-based. These different approaches are often used in combination in IoT applications, which leads to misaligned data since sensors may take samples at different times. To solve this issue, synchronisation is applied to align samples taken at different times, as well as with inconsistent or different sampling frequencies. For example, sensor 1 has sampling frequency 50 Hz and sensor 2 has sampling frequency 60 Hz. Synchronisation can the be used to match the corresponding samples from sensor 1 to sensor 2. This can be performed for example by interpolation with upsampling, where the sampling rate is increased to fill in the blanks or with downsampling where the sampling rate is decreased to discard some samples. Ground thruth In order to evaluate the performance of IoT applications, one needs to be able to compare the results of the application to a fundamental truth in order to determine if the result is correct. This fundamental truth is referred to as a ground truth. One example of a ground truth is the kilogram, which today, is measured using an atomic transition frequency, the speed of light and the Planck constant. Since ground truths can be difficult to access, gold standards, meaning measurements from highly accurate devices, referred to as reference instruments, are often used instead. Gold standards or ground truths can be divided into three types; discrete labels or classes, continuous value and regions of interest. 2 Programming tasks 2.1 Signal Synchronisation and Interpolation In []: import pandas as pd import numpy as np import matplotlib.pyplot as plt In []: # Read dataset into dataframe df = pd.read_csv('sensors_sample.csv', header=None) 1. Sample Synchronisation To synchronise the samples I first distribute the data points in the first two columns along the length of the last column and fill in the gaps with NaN. This then allows me to synchronise the samples by interpolating. I then assigned a time index to the dataframe and then used the resample() function from pandas to resample the data at a frequency of 50 Hz (20 ms) and interpolate the missing values. In []: # Count the number of non-NaN values in each column counts = df.count() # Find the maximum count max_count = counts.max() # Create a new DataFrame with the same number of rows as the longest column new df = pd.DataFrame(index=range(max count)) # For each column in the original DataFrame for col in df.columns: # Calculate the indices where the non-NaN values should be placed indices = np.linspace(0, max_count-1, counts[col]).astype(int) # Create a new series with NaNs at all indices new_series = pd.Series(np.nan, index=range(max_count)) # Place the non-NaN values at the calculated indices new_series[indices] = df[col].dropna().values # Add the new series to the new DataFrame new_df[col] = new_series In []: # Create time index time_index = pd.date_range(start='00:00:00', periods=len(new_df), freq='20ms')

Set time index new_df.index = time_index # Resample and interpolate resamp_df = new_df.resample('20ms').asfreq().interpolate(method='linear') # Reset index to make dataset look nice :) resamp_df = resamp_df.reset_index(drop=True) resamp_df.to_csv('out.csv') 2. Plotting measurements From the plots it's evident that after synchronisation, only column 2 (Plot 4.2) exhibits loss of accuracy (that is visible to the naked eye atleast), otherwise the signals have been interpolated well. In []: plt.figure(figsize=(12, 6)) # Plot before synchronisation plt.subplot(1, 2, 1) plt.plot(df.index, df.values) plt.title('Before Synchronisation (Plot 1)') xmin, xmax = plt.xlim() ymin, ymax = plt.ylim() # Plot after synchronisation plt.subplot(1, 2, 2) plt.plot(resamp_df.index, resamp_df.values) plt.title('After Synchronisation (Plot 2)') # Set the same x and y limits for the second plot plt.xlim([xmin, xmax]) plt.ylim([ymin, ymax]) # Plot individual column before synchronisation fig1, axs1 = plt.subplots(1, df.shape[1], figsize=(6 * df.shape[1], 6)) for i in range(df.shape[1]): axs1[i].plot(df.index, df.iloc[:, i]) axs1[i].set_title(f'column {i+1} (Plot 3.{i+1})') axs1[i].set_xlim([xmin, xmax]) axs1[i].set_ylim([ymin, ymax]) fig1.suptitle(f'Before Synchronisation (Plot 3)', fontsize=16) # Plot individual column after synchronisation fig2, axs2 = plt.subplots(1, resamp_df.shape[1], figsize=(6 * resamp_df.shape[1], 6)) for i in range(resamp_df.shape[1]): axs2[i].plot(resamp_df.index, resamp_df.iloc[:, i]) axs2[i].set_title(f'column {i+1} (Plot 4.{i+1})') axs2[i].set_xlim([xmin, xmax]) axs2[i].set_ylim([ymin, ymax]) fig2.suptitle(f'After Synchronisation (Plot 4)', fontsize=16) plt.show() Before Synchronisation (Plot 1) After Synchronisation (Plot 2) 0.004 0.004

0.002

0.000

-0.002

-0.004

-0.006

500

Before Synchronisation (Plot 3)

column 2 (Plot 3.2)

1000

1500

2000

2500

column 3 (Plot 3.3)

0.002

0.000

-0.002

-0.004

-0.006

500

1000

column 1 (Plot 3.1)

1500

2000

2500



