Assignment - 2

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CS724: SENSING COMMUNICATIONS AND NETWORKING FOR SMART WIRELESS DEVICES

IIT Kanpur
MTech CSE

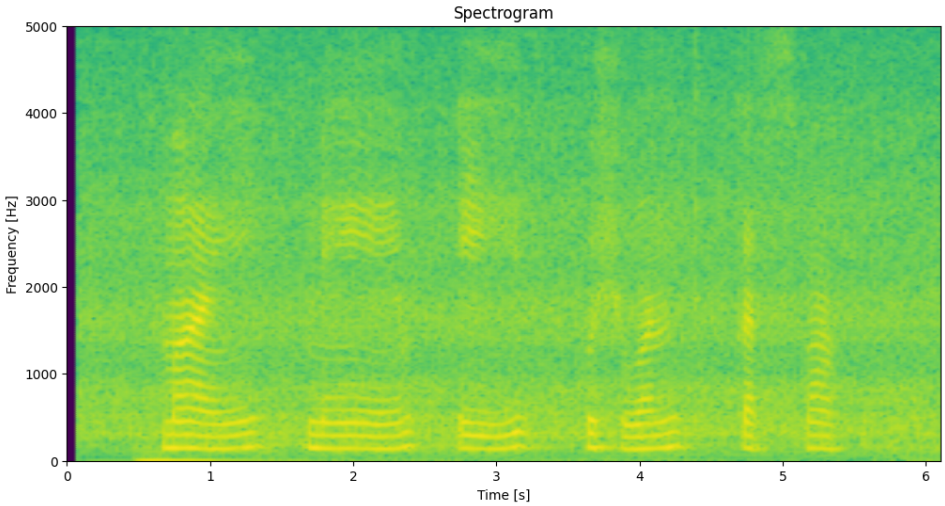
Question 1 : Spectrogram of Audio File

(a)

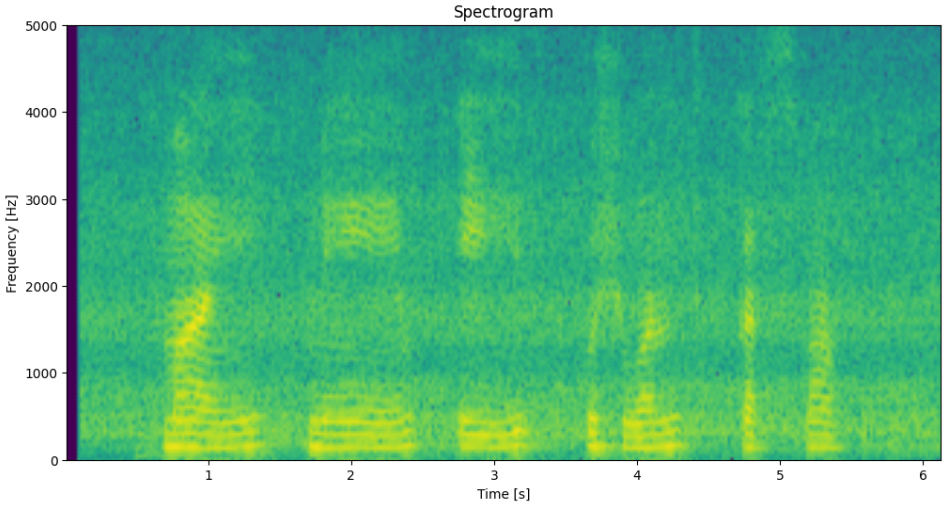
Recorded audo file of .wav format saying "My name is Khushwant Kaswan" and saved as "audio.wav"

(b)

Used numpy and Fourier Transform (numpy.fft) to create a spectogram and plotted using matplotlib.



To verify this, I plotted an direct spectogram of the audio file using `scipy`.



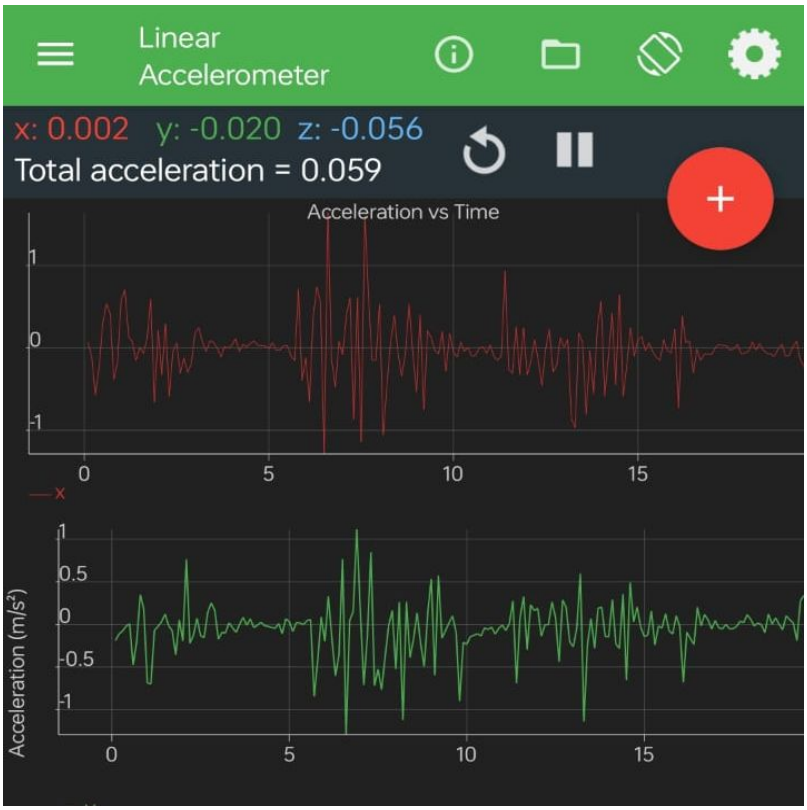
(c)

- Both spectograms(self and scipy generated) are completly same except the color difference.
- Horizontal lines or bands represent frequencies that remain constant during a particular time interval.
- Changes in frequency patterns over time (e.g., the appearance and disappearance of frequency bands) typically correspond to different spoken words or sounds in the audio.
- Each "block" of the spectrogram reflects the frequency content of the audio at that time ie individual words as they are spoken ie "My name is Khushwant Kaswan".
- So, the spectrogram indeed shows the time-frequency structure of speech, with each "spectrum" corresponding to a word or sound.

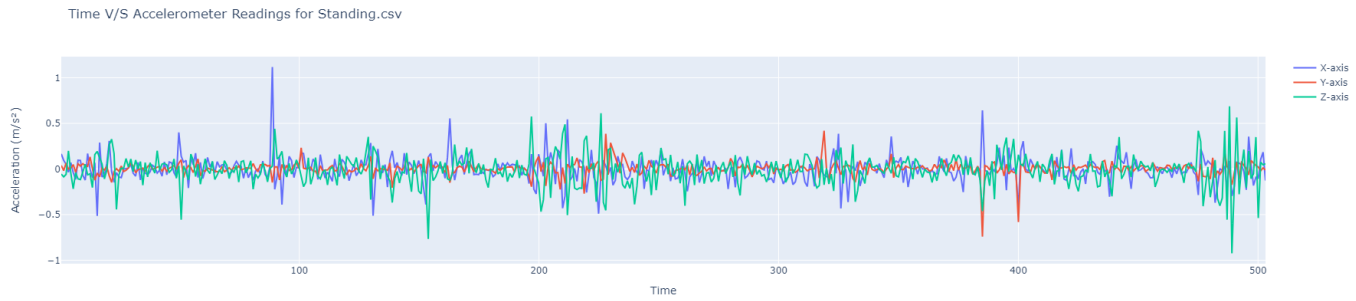
Question 2 : Human Posture Detection

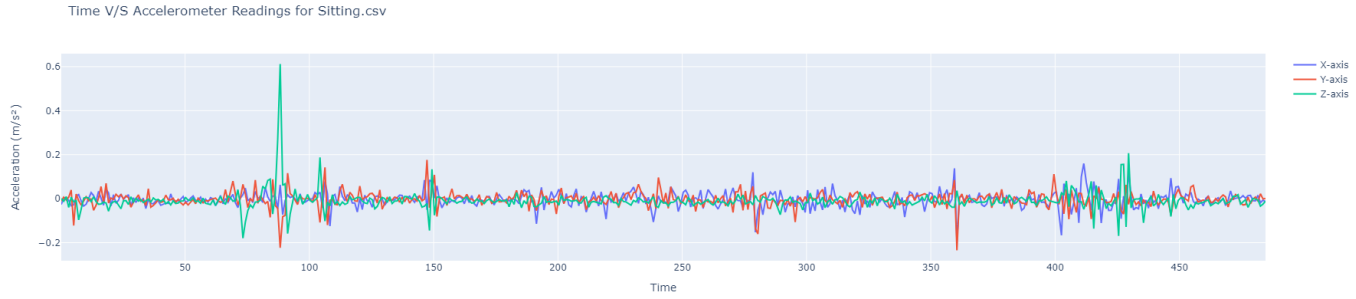
(a)

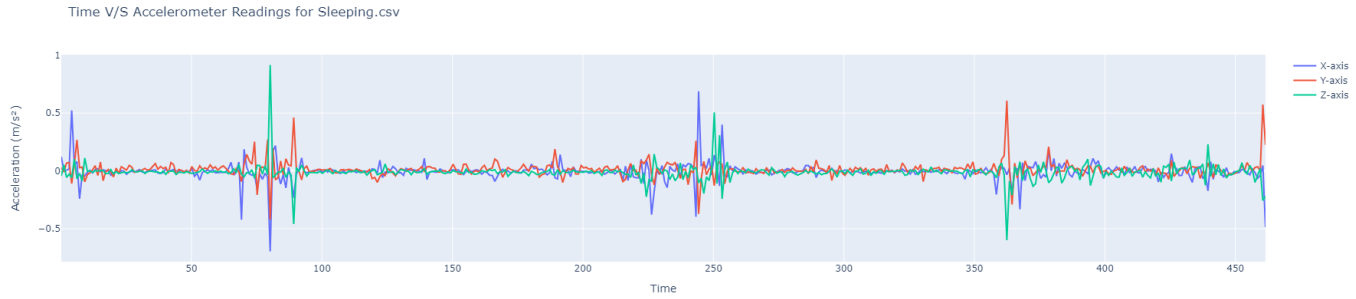
Used the Physics Toolbox app to record the accelerometer samples while sitting, standing and lying down straight.



Plotting them using python `plotly`.







(b)

Find out among these three human postures to which each data belongs to.

Created a test.csv containg 416 different test examples and we are going to classify them to "Sitting" "Sleeping" "Standing" and "Unknown".

We have used 3 different methods

- Using normal thresholding with threshold chosen as (min---max) values of the acceleration-axis.
- Using normal thresholding with threshold chosen as (mean-2*SD*---*mean*+*2*SD) values of the acceleration-axis.
- Using Random Forest Classifier based Machine Learning Approach.

Using the statistics of already available data, we have mean standard-deviation min max.

M1.

For each posture, we create a x,y,z thresholds using minimum and maximum value of that axis.

```
def calculate_thresholds_min_max(sitting_stats, standing_stats, lying_stats):
    thresholds = {
        'X_SLEEP_THRESHOLD': (lying_stats['min']['ax'], lying_stats['max']['ax']),
        'Y_SLEEP_THRESHOLD': (lying_stats['min']['ay'], lying_stats['max']['ay']),
        'Z_SLEEP_THRESHOLD': (lying_stats['min']['az'], lying_stats['max']['az']),

        'X_SITTING_THRESHOLD': (sitting_stats['min']['ax'], sitting_stats['max']['ax']),
        'Y_SITTING_THRESHOLD': (sitting_stats['min']['ay'], sitting_stats['max']['ay']),
        'Z_SITTING_THRESHOLD': (sitting_stats['min']['az'], sitting_stats['max']['az']),

        'X_STANDING_THRESHOLD': (standing_stats['min']['ax'], standing_stats['max']['ax']),
        'Y_STANDING_THRESHOLD': (standing_stats['min']['ay'], standing_stats['max']['ay']),
        'Z_STANDING_THRESHOLD': (standing_stats['min']['az'], standing_stats['max']['az']),
    }
    return thresholds
```

```
Thresholds using minmax:
X_SLEEP_THRESHOLD: (np.float64(-0.07601262981030535), np.float64(0.0688093308412332))
Y_SLEEP_THRESHOLD: (np.float64(-0.07622638670892268), np.float64(0.07004576815222166))
Z_SLEEP_THRESHOLD: (np.float64(-0.09945091017177374), np.float64(0.08130204419239229))
X_SITTING_THRESHOLD: (np.float64(-0.2855997487255026), np.float64(0.28449795946108675))
Y_SITTING_THRESHOLD: (np.float64(-0.14868076454295825), np.float64(0.15608235500021472))
Z_SITTING_THRESHOLD: (np.float64(-0.3549421168684677), np.float64(0.3283246218386467))
X_STANDING_THRESHOLD: (np.float64(-0.1745053045858415), np.float64(0.16223690631744322))
Y_STANDING_THRESHOLD: (np.float64(-0.13258737650349064), np.float64(0.15792720334331747))
Z_STANDING_THRESHOLD: (np.float64(-0.16315753219843665), np.float64(0.1403280949689994))
```

M2.

For each posture, we create a x,y,z thresholds between mean-2*SD to mean+2*SD for all axis.

```
def create_thresholds_mean_sd(lying_stats, sitting_stats, standing_stats):
    # mean ± 2 * SD.
    thresholds = {
        'X_SLEEP_THRESHOLD': (lying_stats['mean']['ax'] - 2 * lying_stats['std']['ax'],
                              lying_stats['mean']['ax'] + 2 * lying_stats['std']['ax']),
        'Y_SLEEP_THRESHOLD': (lying_stats['mean']['ay'] - 2 * lying_stats['std']['ay'],
                              lying_stats['mean']['ay'] + 2 * lying_stats['std']['ay']),
        'Z_SLEEP_THRESHOLD': (lying_stats['mean']['az'] - 2 * lying_stats['std']['az'],
                              lying_stats['mean']['az'] + 2 * lying_stats['std']['az']),

        'X_SITTING_THRESHOLD': (sitting_stats['mean']['ax'] - 2 * sitting_stats['std']['ax'],
                                sitting_stats['mean']['ax'] + 2 * sitting_stats['std']['ax']),
        'Y_SITTING_THRESHOLD': (sitting_stats['mean']['ay'] - 2 * sitting_stats['std']['ay'],
                                sitting_stats['mean']['ay'] + 2 * sitting_stats['std']['ay']),
        'Z_SITTING_THRESHOLD': (sitting_stats['mean']['az'] - 2 * sitting_stats['std']['az'],
                                sitting_stats['mean']['az'] + 2 * sitting_stats['std']['az']),

        'X_STANDING_THRESHOLD': (standing_stats['mean']['ax'] - 2 * standing_stats['std']['ax'],
                                 standing_stats['mean']['ax'] + 2 * standing_stats['std']['ax']),
        'Y_STANDING_THRESHOLD': (standing_stats['mean']['ay'] - 2 * standing_stats['std']['ay'],
                                 standing_stats['mean']['ay'] + 2 * standing_stats['std']['ay']),
        'Z_STANDING_THRESHOLD': (standing_stats['mean']['az'] - 2 * standing_stats['std']['az'],
                                 standing_stats['mean']['az'] + 2 * standing_stats['std']['az']),
    }
    return thresholds
```

```
Thresholds using meansd:
X_SLEEP_THRESHOLD: (np.float64(-0.6956), np.float64(0.6881))
Y_SLEEP_THRESHOLD: (np.float64(-0.421), np.float64(0.6074))
Z_SLEEP_THRESHOLD: (np.float64(-0.5984), np.float64(0.9153))
X_SITTING_THRESHOLD: (np.float64(-0.1936), np.float64(0.1597))
Y_SITTING_THRESHOLD: (np.float64(-0.2346), np.float64(0.1765))
Z_SITTING_THRESHOLD: (np.float64(-0.1791), np.float64(0.6119))
X_STANDING_THRESHOLD: (np.float64(-0.5145), np.float64(1.117))
Y_STANDING_THRESHOLD: (np.float64(-0.7421), np.float64(0.4199))
Z_STANDING_THRESHOLD: (np.float64(-0.9244), np.float64(0.6869))
```

Testing using M1 and M2

If all the x,y,z value lies betwee the threhold only then we classify the posture else Unknown.

```
def classify_posture(x_acc, y_acc, z_acc, thresholds):

    if (thresholds['X_STANDING_THRESHOLD'][0] <= x_acc <= thresholds['X_STANDING_THRESHOLD'][1] and
        thresholds['Y_STANDING_THRESHOLD'][0] <= y_acc <= thresholds['Y_STANDING_THRESHOLD'][1] and
        thresholds['Z_STANDING_THRESHOLD'][0] <= z_acc <= thresholds['Z_STANDING_THRESHOLD'][1]):
        return "Standing"

    elif (thresholds['X_SITTING_THRESHOLD'][0] <= x_acc <= thresholds['X_SITTING_THRESHOLD'][1] and
        thresholds['Y_SITTING_THRESHOLD'][0] <= y_acc <= thresholds['Y_SITTING_THRESHOLD'][1] and
        thresholds['Z_SITTING_THRESHOLD'][0] <= z_acc <= thresholds['Z_SITTING_THRESHOLD'][1]):
        return "Sitting"

    elif (thresholds['X_SLEEP_THRESHOLD'][0] <= x_acc <= thresholds['X_SLEEP_THRESHOLD'][1] and
        thresholds['Y_SLEEP_THRESHOLD'][0] <= y_acc <= thresholds['Y_SLEEP_THRESHOLD'][1] and
        thresholds['Z_SLEEP_THRESHOLD'][0] <= z_acc <= thresholds['Z_SLEEP_THRESHOLD'][1]):
        return "Sleeping"

    else:
        return "Unknown"
```

Using minmax thresholds

SNo	ax	ay	az	Posture
1	0.0031	-0.0253	0.1412	Sitting
2	0.4193	-4.9210	-0.8440	Unknown
3	-0.3709	-1.1664	0.9668	Unknown
4	2.7903	0.4226	1.3831	Unknown
5	1.1368	0.0837	0.4272	Unknown
6	2.6244	0.1059	-1.6480	Unknown
7	4.8386	-2.2122	4.5807	Unknown
8	3.4710	1.0699	-1.5718	Unknown
9	0.9477	2.1898	-1.9411	Unknown
10	-1.1141	0.7250	-0.2694	Unknown
11	-0.1970	-0.6247	1.2076	Unknown
12	-2.5819	0.1456	-0.6227	Unknown
13	-1.7483	-2.2014	-0.0063	Unknown
14	-3.1802	-2.3740	0.4016	Unknown
15	-2.9738	-3.9254	-0.0971	Unknown
16	-2.8121	-3.8173	0.9375	Unknown
17	-3.0900	-1.7953	1.3963	Unknown
18	-1.5236	0.6277	2.6021	Unknown
19	-1.4895	-0.4611	1.8119	Unknown
20	0.9882	-0.9263	-0.6596	Unknown
21	5.8187	0.9201	-0.9798	Unknown
22	6.3257	-0.0642	-0.2200	Unknown
23	2.2127	1.2019	-1.0511	Unknown
24	-0.9703	0.3318	0.0132	Unknown
25	-0.0382	-2.3205	0.7462	Unknown
26	-4.1653	-4.4957	0.2054	Unknown
27	-0.1614	-1.2996	0.1153	Unknown
28	-0.2621	-0.8977	0.7754	Unknown
29	0.0797	-1.3104	1.1307	Unknown
30	0.1608	-0.3590	1.1070	Unknown
31	-0.0952	-0.1178	0.1607	Sitting
32	-0.1310	-0.0464	0.0808	Standing
33	0.0031	0.0071	-0.0119	Standing
34	0.0039	-0.0066	0.0094	Standing

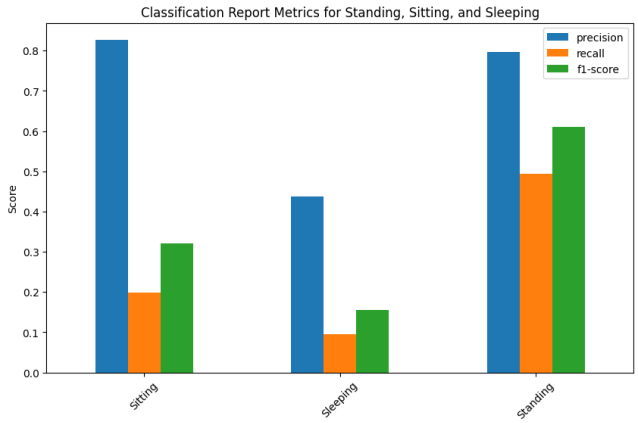
Using meansd thresholds

SNo	ax	ay	az	Posture
1	0.0031	-0.0253	0.1412	Standing
2	0.4193	-4.9210	-0.8440	Unknown
3	-0.3709	-1.1664	0.9668	Unknown
4	2.7903	0.4226	1.3831	Unknown
5	1.1368	0.0837	0.4272	Unknown
6	2.6244	0.1059	-1.6480	Unknown
7	4.8386	-2.2122	4.5807	Unknown
8	3.4710	1.0699	-1.5718	Unknown
9	0.9477	2.1898	-1.9411	Unknown
10	-1.1141	0.7250	-0.2694	Unknown
11	-0.1970	-0.6247	1.2076	Unknown
12	-2.5819	0.1456	-0.6227	Unknown
13	-1.7483	-2.2014	-0.0063	Unknown
14	-3.1802	-2.3740	0.4016	Unknown
15	-2.9738	-3.9254	-0.0971	Unknown
16	-2.8121	-3.8173	0.9375	Unknown
17	-3.0900	-1.7953	1.3963	Unknown
18	-1.5236	0.6277	2.6021	Unknown
19	-1.4895	-0.4611	1.8119	Unknown
20	0.9882	-0.9263	-0.6596	Unknown
21	5.8187	0.9201	-0.9798	Unknown
22	6.3257	-0.0642	-0.2200	Unknown
23	2.2127	1.2019	-1.0511	Unknown
24	-0.9703	0.3318	0.0132	Unknown
25	-0.0382	-2.3205	0.7462	Unknown
26	-4.1653	-4.4957	0.2054	Unknown
27	-0.1614	-1.2996	0.1153	Unknown
28	-0.2621	-0.8977	0.7754	Unknown
29	0.0797	-1.3104	1.1307	Unknown
30	0.1608	-0.3590	1.1070	Unknown
31	-0.0952	-0.1178	0.1607	Standing
32	-0.1310	-0.0464	0.0808	Standing
33	0.0031	0.0071	-0.0119	Standing
34	0.0039	-0.0066	0.0094	Standing
35	-0.0079	0.0070	-0.0397	Standing
36	0.0691	-0.1229	0.1577	Standing

M3.

Using Random Forest Based Machine Learning Classifier.

- Add posture labels to the each dataset
- Combine all 3 datasets into one dataset
- Split the data into training and validation sets
- Train a Random Forest classifier
- Predict on validation data to check performance
- For each prediction, if the prediction confidence is below the threshold(0.7), assign "Unknown"
- Predict on the test.csv data
- For each test prediction, assign "Unknown" if prediction confidence is below the threshold(0.7)
- Show the classification report and test data with predicted postures



Line	ax	ay	az	Posture
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2	0.4193	-4.9210	-0.8440	Standing
3	-0.3709	-1.1664	0.9668	Unknown
4	2.7903	0.4226	1.3831	Standing
5	1.1368	0.0837	0.4272	Standing
6	2.6244	0.1059	-1.6480	Standing
7	4.8386	-2.2122	4.5807	Unknown
8	3.4710	1.0699	-1.5718	Unknown
9	0.9477	2.1898	-1.9411	Unknown
10	-1.1141	0.7250	-0.2694	Unknown
11	-0.1970	-0.6247	1.2076	Unknown
12	-2.5819	0.1456	-0.6227	Standing
13	-1.7483	-2.2014	-0.0063	Unknown
14	-3.1802	-2.3740	0.4016	Unknown
15	-2.9738	-3.9254	-0.0971	Unknown
16	-2.8121	-3.8173	0.9375	Unknown
17	-3.0900	-1.7953	1.3963	Unknown
18	-1.5236	0.6277	2.6021	Unknown
19	-1.4895	-0.4611	1.8119	Unknown
20	0.9882	-0.9263	-0.6596	Standing
21	5.8187	0.9201	-0.9798	Unknown
22	6.3257	-0.0642	-0.2200	Standing
23	2.2127	1.2019	-1.0511	Unknown
24	-0.9703	0.3318	0.0132	Unknown
25	-0.0382	-2.3205	0.7462	Unknown
26	-4.1653	-4.4957	0.2054	Unknown
27	-0.1614	-1.2996	0.1153	Unknown
28	-0.2621	-0.8977	0.7754	Unknown
29	0.0797	-1.3104	1.1307	Unknown
30	0.1608	-0.3590	1.1070	Unknown
31	-0.0952	-0.1178	0.1607	Unknown
32	-0.1310	-0.0464	0.0808	Standing
33	0.0031	0.0071	-0.0119	Sitting
34	0.0039	-0.0066	0.0094	Sitting
35	-0.0079	0.0070	-0.0397	Unknown
36	0.0691	-0.1229	0.1577	Standing
37	-0.5145	-0.0446	0.1915	Standing
38	0.2884	-0.0869	0.1279	Standing
39	-0.0882	0.0098	0.0538	Standing
40	-0.0326	0.0393	0.0065	Unknown
41	-0.0275	-0.1219	-0.0138	Sitting