

## labeling-2

February 19, 2020

```
[1]: import pandas as pd
import numpy as np
import scipy
from scipy import stats
import matplotlib.pyplot as plt
```

```
[2]: df = pd.read_csv("windowing_results.csv")
```

```
[3]: df
```

```
[3]:      Unnamed: 0  ECG #      start_time      end_time  acc_count  \
0          svm      1  2019-11-21 09:35:27  2019-11-21 09:35:29    2271.0
1          svm      1  2019-11-21 09:35:28  2019-11-21 09:35:30    2287.0
2          svm      1  2019-11-21 09:35:29  2019-11-21 09:35:31    2288.0
3          svm      1  2019-11-21 09:35:30  2019-11-21 09:35:32    2294.0
4          svm      1  2019-11-21 09:35:31  2019-11-21 09:35:33    2283.0
...          ...      ...
14192       svm      6  2019-11-21 11:00:35  2019-11-21 11:00:37    2087.0
14193       svm      6  2019-11-21 11:00:36  2019-11-21 11:00:38    2277.0
14194       svm      6  2019-11-21 11:00:37  2019-11-21 11:00:39    2279.0
14195       svm      6  2019-11-21 11:00:38  2019-11-21 11:00:40    2276.0
14196       svm      6  2019-11-21 11:00:39  2019-11-21 11:00:41    1518.0

      acc_mean  acc_std  acc_min  acc_25%  acc_50%  ...  gyro_25%  \
0    1.488039  0.008815  1.464329  1.482331  1.487741  ...  0.595793
1    1.487708  0.008177  1.464329  1.482331  1.487741  ...  0.508012
2    1.484074  0.007623  1.462431  1.478181  1.482889  ...  0.451361
3    1.487552  0.007213  1.462431  1.482889  1.488377  ...  0.449286
4    1.486918  0.006790  1.462431  1.482331  1.488377  ...  0.533525
...          ...      ...      ...      ...      ...  ...
14192  1.601729  0.088580  1.501142  1.531028  1.566651  ...  1.838959
14193  1.584692  0.055840  1.501142  1.532493  1.579270  ...  2.090319
14194  1.634041  0.070983  1.502046  1.565519  1.651781  ...  2.356062
14195  1.701175  0.045277  1.584833  1.675359  1.706349  ...  2.275376
14196  1.727280  0.024522  1.678459  1.706349  1.721291  ...  2.325614

      gyro_50%  gyro_75%  gyro_max      gyro_eng  gyro_entropy  gyro_var  \
```

0	0.892834	1.257357	2.802878	2444.644866	7.727976	0.203803
1	0.779196	1.169292	2.802878	2203.410455	7.734996	0.208016
2	0.697942	1.106439	2.802878	2120.945536	7.735433	0.247652
3	0.673428	1.076882	2.895836	2354.323033	7.738052	0.314183
4	0.849516	1.426766	3.649751	3478.688197	7.733246	0.451992
...	...	...	...	...	...	...
14192	2.628791	3.736613	12.023601	25833.234557	7.643483	3.165173
14193	2.962076	3.977001	12.023601	30604.610546	7.730005	2.992684
14194	3.164181	4.072940	12.023601	33008.076912	7.730884	2.628023
14195	3.049675	3.954686	12.413456	29859.735732	7.729566	2.380862
14196	3.011691	3.944505	12.413456	21075.108897	7.325149	2.775818

	gyro_skew	gyro_kurt	gyro_mad
0	0.492299	0.014721	0.370100
1	0.772472	0.362000	0.374022
2	1.072809	0.951265	0.396192
3	1.283556	1.183364	0.433260
4	1.079871	0.786448	0.535704
...	...	...	...
14192	1.786256	4.765928	1.282278
14193	1.558248	4.272371	1.270297
14194	1.724908	5.127654	1.170044
14195	1.807932	6.491142	1.110759
14196	2.050351	6.801323	1.158021

[14197 rows x 32 columns]

```
[4]: df.columns
```

```
[4]: Index(['Unnamed: 0', 'ECG #', 'start_time', 'end_time', 'acc_count',
        'acc_mean', 'acc_std', 'acc_min', 'acc_25%', 'acc_50%', 'acc_75%',
        'acc_max', 'acc_eng', 'acc_entropy', 'acc_var', 'acc_skew', 'acc_kurt',
        'acc_mad', 'Activity', 'gyro_mean', 'gyro_std', 'gyro_min', 'gyro_25%',
        'gyro_50%', 'gyro_75%', 'gyro_max', 'gyro_eng', 'gyro_entropy',
        'gyro_var', 'gyro_skew', 'gyro_kurt', 'gyro_mad'],
        dtype='object')
```

```
[73]: df_filtered = df[(df['ECG #'] >= 5) | (df['ECG #'] <= 2) ]
```

```
[75]: df1 = df_filtered
df1 = df1.reset_index()
df1
```

```
[75]:
```

	index	Unnamed: 0	ECG #	start_time	end_time	\
0	0	svm	1	2019-11-21 09:35:27	2019-11-21 09:35:29	
1	1	svm	1	2019-11-21 09:35:28	2019-11-21 09:35:30	
2	2	svm	1	2019-11-21 09:35:29	2019-11-21 09:35:31	

3	3	svm	1	2019-11-21 09:35:30	2019-11-21 09:35:32
4	4	svm	1	2019-11-21 09:35:31	2019-11-21 09:35:33
...	...	...	...	...	...
10286	14192	svm	6	2019-11-21 11:00:35	2019-11-21 11:00:37
10287	14193	svm	6	2019-11-21 11:00:36	2019-11-21 11:00:38
10288	14194	svm	6	2019-11-21 11:00:37	2019-11-21 11:00:39
10289	14195	svm	6	2019-11-21 11:00:38	2019-11-21 11:00:40
10290	14196	svm	6	2019-11-21 11:00:39	2019-11-21 11:00:41

	acc_count	acc_mean	acc_std	acc_min	acc_25%	...	gyro_25%	\
0	2271.0	1.488039	0.008815	1.464329	1.482331	...	0.595793	
1	2287.0	1.487708	0.008177	1.464329	1.482331	...	0.508012	
2	2288.0	1.484074	0.007623	1.462431	1.478181	...	0.451361	
3	2294.0	1.487552	0.007213	1.462431	1.482889	...	0.449286	
4	2283.0	1.486918	0.006790	1.462431	1.482331	...	0.533525	
...	...	...	...	...	...	...	...	
10286	2087.0	1.601729	0.088580	1.501142	1.531028	...	1.838959	
10287	2277.0	1.584692	0.055840	1.501142	1.532493	...	2.090319	
10288	2279.0	1.634041	0.070983	1.502046	1.565519	...	2.356062	
10289	2276.0	1.701175	0.045277	1.584833	1.675359	...	2.275376	
10290	1518.0	1.727280	0.024522	1.678459	1.706349	...	2.325614	

	gyro_50%	gyro_75%	gyro_max	gyro_eng	gyro_entropy	gyro_var	\
0	0.892834	1.257357	2.802878	2444.644866	7.727976	0.203803	
1	0.779196	1.169292	2.802878	2203.410455	7.734996	0.208016	
2	0.697942	1.106439	2.802878	2120.945536	7.735433	0.247652	
3	0.673428	1.076882	2.895836	2354.323033	7.738052	0.314183	
4	0.849516	1.426766	3.649751	3478.688197	7.733246	0.451992	
...	...	...	...	...	...	...	
10286	2.628791	3.736613	12.023601	25833.234557	7.643483	3.165173	
10287	2.962076	3.977001	12.023601	30604.610546	7.730005	2.992684	
10288	3.164181	4.072940	12.023601	33008.076912	7.730884	2.628023	
10289	3.049675	3.954686	12.413456	29859.735732	7.729566	2.380862	
10290	3.011691	3.944505	12.413456	21075.108897	7.325149	2.775818	

	gyro_skew	gyro_kurt	gyro_mad
0	0.492299	0.014721	0.370100
1	0.772472	0.362000	0.374022
2	1.072809	0.951265	0.396192
3	1.283556	1.183364	0.433260
4	1.079871	0.786448	0.535704
...	...	...	...
10286	1.786256	4.765928	1.282278
10287	1.558248	4.272371	1.270297
10288	1.724908	5.127654	1.170044
10289	1.807932	6.491142	1.110759
10290	2.050351	6.801323	1.158021

[10291 rows x 33 columns]

```
[ ]: # sc_results.insert(0, 'GSR #', '2')
```

```
[76]: df2 = pd.read_csv('skin conductance.csv')
```

```
[77]: df2
```

```
[77]:
```

	Unnamed: 0	GSR #	start_time	\
0	GSR1_GSR_Skin_Conductance_CAL	1	2019-11-21 09:35:27	
1	GSR1_GSR_Skin_Conductance_CAL	1	2019-11-21 09:35:28	
2	GSR1_GSR_Skin_Conductance_CAL	1	2019-11-21 09:35:29	
3	GSR1_GSR_Skin_Conductance_CAL	1	2019-11-21 09:35:30	
4	GSR1_GSR_Skin_Conductance_CAL	1	2019-11-21 09:35:31	
...	...	...	...	
10286	GSR6_GSR_Skin_Conductance_CAL	6	2019-11-21 11:00:35	
10287	GSR6_GSR_Skin_Conductance_CAL	6	2019-11-21 11:00:36	
10288	GSR6_GSR_Skin_Conductance_CAL	6	2019-11-21 11:00:37	
10289	GSR6_GSR_Skin_Conductance_CAL	6	2019-11-21 11:00:38	
10290	GSR6_GSR_Skin_Conductance_CAL	6	2019-11-21 11:00:39	

  

	end_time	gsr_mean	gsr_std	gsr_mean.1	gsr_max	\
0	2019-11-21 09:35:29	1.355378	0.015984	1.332223	1.391275	
1	2019-11-21 09:35:30	1.339162	0.011172	1.318903	1.358863	
2	2019-11-21 09:35:31	1.326166	0.011614	1.306471	1.347763	
3	2019-11-21 09:35:32	1.313587	0.010649	1.294483	1.333999	
4	2019-11-21 09:35:33	1.302077	0.009201	1.286491	1.321123	
...	...	...	...	...	...	
10286	2019-11-21 11:00:37	1.873922	0.004225	1.857143	1.895238	
10287	2019-11-21 11:00:38	1.872439	0.004663	1.838095	1.895238	
10288	2019-11-21 11:00:39	1.869794	0.004626	1.838095	1.892308	
10289	2019-11-21 11:00:40	1.866879	0.004492	1.838095	1.889377	
10290	2019-11-21 11:00:41	1.865581	0.004135	1.842491	1.883516	

  

	gsr_variance	gsr_skewness	gsr_kurtosis	gsr_mad	Activity
0	0.000256	0.699293	-0.627312	0.012809	0
1	0.000125	-0.100332	-1.238669	0.009691	0
2	0.000135	0.010484	-1.243451	0.010120	0
3	0.000113	0.140383	-1.149240	0.009224	0
4	0.000085	0.139074	-1.145452	0.007970	0
...	...	...	...	...	...
10286	0.000018	1.241958	5.575090	0.002684	23
10287	0.000022	0.428993	5.263945	0.003213	23
10288	0.000021	0.490440	4.582487	0.003247	23
10289	0.000020	0.256191	3.974350	0.003274	23
10290	0.000017	0.676028	4.517975	0.002796	23

[10291 rows x 13 columns]

```
[82]: new_df = pd.concat([df1, df2], axis=1)
      new_df
```

```
[82]:
```

	index	Unnamed: 0	ECG #	start_time	end_time	\
0	0	svm	1	2019-11-21 09:35:27	2019-11-21 09:35:29	
1	1	svm	1	2019-11-21 09:35:28	2019-11-21 09:35:30	
2	2	svm	1	2019-11-21 09:35:29	2019-11-21 09:35:31	
3	3	svm	1	2019-11-21 09:35:30	2019-11-21 09:35:32	
4	4	svm	1	2019-11-21 09:35:31	2019-11-21 09:35:33	
...	...	...	...	...	...	
10286	14192	svm	6	2019-11-21 11:00:35	2019-11-21 11:00:37	
10287	14193	svm	6	2019-11-21 11:00:36	2019-11-21 11:00:38	
10288	14194	svm	6	2019-11-21 11:00:37	2019-11-21 11:00:39	
10289	14195	svm	6	2019-11-21 11:00:38	2019-11-21 11:00:40	
10290	14196	svm	6	2019-11-21 11:00:39	2019-11-21 11:00:41	

  

	acc_count	acc_mean	acc_std	acc_min	acc_25%	...	\
0	2271.0	1.488039	0.008815	1.464329	1.482331	...	
1	2287.0	1.487708	0.008177	1.464329	1.482331	...	
2	2288.0	1.484074	0.007623	1.462431	1.478181	...	
3	2294.0	1.487552	0.007213	1.462431	1.482889	...	
4	2283.0	1.486918	0.006790	1.462431	1.482331	...	
...	...	...	...	...	...	...	
10286	2087.0	1.601729	0.088580	1.501142	1.531028	...	
10287	2277.0	1.584692	0.055840	1.501142	1.532493	...	
10288	2279.0	1.634041	0.070983	1.502046	1.565519	...	
10289	2276.0	1.701175	0.045277	1.584833	1.675359	...	
10290	1518.0	1.727280	0.024522	1.678459	1.706349	...	

  

	end_time	gsr_mean	gsr_std	gsr_mean.1	gsr_max	\
0	2019-11-21 09:35:29	1.355378	0.015984	1.332223	1.391275	
1	2019-11-21 09:35:30	1.339162	0.011172	1.318903	1.358863	
2	2019-11-21 09:35:31	1.326166	0.011614	1.306471	1.347763	
3	2019-11-21 09:35:32	1.313587	0.010649	1.294483	1.333999	
4	2019-11-21 09:35:33	1.302077	0.009201	1.286491	1.321123	
...	...	...	...	...	...	
10286	2019-11-21 11:00:37	1.873922	0.004225	1.857143	1.895238	
10287	2019-11-21 11:00:38	1.872439	0.004663	1.838095	1.895238	
10288	2019-11-21 11:00:39	1.869794	0.004626	1.838095	1.892308	
10289	2019-11-21 11:00:40	1.866879	0.004492	1.838095	1.889377	
10290	2019-11-21 11:00:41	1.865581	0.004135	1.842491	1.883516	

  

	gsr_variance	gsr_skewness	gsr_kurtosis	gsr_mad	Activity
0	0.000256	0.699293	-0.627312	0.012809	0

1	0.000125	-0.100332	-1.238669	0.009691	0
2	0.000135	0.010484	-1.243451	0.010120	0
3	0.000113	0.140383	-1.149240	0.009224	0
4	0.000085	0.139074	-1.145452	0.007970	0
...	...	...	...	...	...
10286	0.000018	1.241958	5.575090	0.002684	23
10287	0.000022	0.428993	5.263945	0.003213	23
10288	0.000021	0.490440	4.582487	0.003247	23
10289	0.000020	0.256191	3.974350	0.003274	23
10290	0.000017	0.676028	4.517975	0.002796	23

[10291 rows x 46 columns]

```
[83]: # new_df.columns
result = new_df.copy()
result.drop(columns = ['index', 'Unnamed: 0', 'GSR #'], inplace = True)
result.columns
```

```
[83]: Index(['ECG #', 'start_time', 'end_time', 'acc_count', 'acc_mean', 'acc_std',
        'acc_min', 'acc_25%', 'acc_50%', 'acc_75%', 'acc_max', 'acc_eng',
        'acc_entropy', 'acc_var', 'acc_skew', 'acc_kurt', 'acc_mad', 'Activity',
        'gyro_mean', 'gyro_std', 'gyro_min', 'gyro_25%', 'gyro_50%', 'gyro_75%',
        'gyro_max', 'gyro_eng', 'gyro_entropy', 'gyro_var', 'gyro_skew',
        'gyro_kurt', 'gyro_mad', 'start_time', 'end_time', 'gsr_mean',
        'gsr_std', 'gsr_mean.1', 'gsr_max', 'gsr_variance', 'gsr_skewness',
        'gsr_kurtosis', 'gsr_mad', 'Activity'],
        dtype='object')
```

```
[84]: result1 = result.copy()
result1 = result1.T.drop_duplicates().T
result1 = result1.iloc[:, :-1]
result1.columns
```

```
[84]: Index(['ECG #', 'start_time', 'end_time', 'acc_count', 'acc_mean', 'acc_std',
        'acc_min', 'acc_25%', 'acc_50%', 'acc_75%', 'acc_max', 'acc_eng',
        'acc_entropy', 'acc_var', 'acc_skew', 'acc_kurt', 'acc_mad', 'Activity',
        'gyro_mean', 'gyro_std', 'gyro_min', 'gyro_25%', 'gyro_50%', 'gyro_75%',
        'gyro_max', 'gyro_eng', 'gyro_entropy', 'gyro_var', 'gyro_skew',
        'gyro_kurt', 'gyro_mad', 'gsr_mean', 'gsr_std', 'gsr_mean.1', 'gsr_max',
        'gsr_variance', 'gsr_skewness', 'gsr_kurtosis'],
        dtype='object')
```

```
[85]: df_target = result1.copy()
df_target
```

```
[85]:      ECG #      start_time      end_time acc_count acc_mean \
0      1  2019-11-21 09:35:27  2019-11-21 09:35:29      2271  1.48804
```

1	1	2019-11-21	09:35:28	2019-11-21	09:35:30	2287	1.48771
2	1	2019-11-21	09:35:29	2019-11-21	09:35:31	2288	1.48407
3	1	2019-11-21	09:35:30	2019-11-21	09:35:32	2294	1.48755
4	1	2019-11-21	09:35:31	2019-11-21	09:35:33	2283	1.48692
...	...	...	...	...	...	...	...
10286	6	2019-11-21	11:00:35	2019-11-21	11:00:37	2087	1.60173
10287	6	2019-11-21	11:00:36	2019-11-21	11:00:38	2277	1.58469
10288	6	2019-11-21	11:00:37	2019-11-21	11:00:39	2279	1.63404
10289	6	2019-11-21	11:00:38	2019-11-21	11:00:40	2276	1.70117
10290	6	2019-11-21	11:00:39	2019-11-21	11:00:41	1518	1.72728

	acc_std	acc_min	acc_25%	acc_50%	acc_75%	...	gyro_skew	\
0	0.00881535	1.46433	1.48233	1.48774	1.49491	...	0.492299	
1	0.00817652	1.46433	1.48233	1.48774	1.49345	...	0.772472	
2	0.00762258	1.46243	1.47818	1.48289	1.49008	...	1.07281	
3	0.00721254	1.46243	1.48289	1.48838	1.49297	...	1.28356	
4	0.00678993	1.46243	1.48233	1.48838	1.49183	...	1.07987	
...	...	...	...	...	...	...	...	...
10286	0.0885796	1.50114	1.53103	1.56665	1.65917	...	1.78626	
10287	0.0558403	1.50114	1.53249	1.57927	1.62692	...	1.55825	
10288	0.0709833	1.50205	1.56552	1.65178	1.70042	...	1.72491	
10289	0.0452765	1.58483	1.67536	1.70635	1.73144	...	1.80793	
10290	0.0245219	1.67846	1.70635	1.72129	1.74851	...	2.05035	

	gyro_kurt	gyro_mad	gsr_mean	gsr_std	gsr_mean.1	gsr_max	\
0	0.0147212	0.3701	1.35538	0.0159845	1.33222	1.39128	
1	0.362	0.374022	1.33916	0.0111716	1.3189	1.35886	
2	0.951265	0.396192	1.32617	0.0116139	1.30647	1.34776	
3	1.18336	0.43326	1.31359	0.0106492	1.29448	1.334	
4	0.786448	0.535704	1.30208	0.0092008	1.28649	1.32112	
...	...	...	...	...	...	...	...
10286	4.76593	1.28228	1.87392	0.0042255	1.85714	1.89524	
10287	4.27237	1.2703	1.87244	0.00466292	1.8381	1.89524	
10288	5.12765	1.17004	1.86979	0.00462558	1.8381	1.89231	
10289	6.49114	1.11076	1.86688	0.00449201	1.8381	1.88938	
10290	6.80132	1.15802	1.86558	0.00413547	1.84249	1.88352	

	gsr_variance	gsr_skewness	gsr_kurtosis
0	0.000255504	0.699293	-0.627312
1	0.000124804	-0.100332	-1.23867
2	0.000134884	0.0104838	-1.24345
3	0.000113405	0.140383	-1.14924
4	8.46548e-05	0.139074	-1.14545
...	...	...	...
10286	1.78548e-05	1.24196	5.57509
10287	2.17428e-05	0.428993	5.26394
10288	2.13959e-05	0.49044	4.58249

```

10289  2.01781e-05      0.256191      3.97435
10290  1.71021e-05      0.676028      4.51798

```

[10291 rows x 38 columns]

```

[89]: df_target['score_change'] = ""
      df_target

```

```

[89]:      ECG #      start_time      end_time  acc_count  acc_mean  \
0         1  2019-11-21 09:35:27  2019-11-21 09:35:29      2271  1.48804
1         1  2019-11-21 09:35:28  2019-11-21 09:35:30      2287  1.48771
2         1  2019-11-21 09:35:29  2019-11-21 09:35:31      2288  1.48407
3         1  2019-11-21 09:35:30  2019-11-21 09:35:32      2294  1.48755
4         1  2019-11-21 09:35:31  2019-11-21 09:35:33      2283  1.48692
...      ...
10286      6  2019-11-21 11:00:35  2019-11-21 11:00:37      2087  1.60173
10287      6  2019-11-21 11:00:36  2019-11-21 11:00:38      2277  1.58469
10288      6  2019-11-21 11:00:37  2019-11-21 11:00:39      2279  1.63404
10289      6  2019-11-21 11:00:38  2019-11-21 11:00:40      2276  1.70117
10290      6  2019-11-21 11:00:39  2019-11-21 11:00:41      1518  1.72728

```

```

      acc_std  acc_min  acc_25%  acc_50%  acc_75%  ...  gyro_kurt  \
0    0.00881535  1.46433  1.48233  1.48774  1.49491  ...  0.0147212
1    0.00817652  1.46433  1.48233  1.48774  1.49345  ...    0.362
2    0.00762258  1.46243  1.47818  1.48289  1.49008  ...  0.951265
3    0.00721254  1.46243  1.48289  1.48838  1.49297  ...   1.18336
4    0.00678993  1.46243  1.48233  1.48838  1.49183  ...  0.786448
...      ...
10286  0.0885796  1.50114  1.53103  1.56665  1.65917  ...   4.76593
10287  0.0558403  1.50114  1.53249  1.57927  1.62692  ...   4.27237
10288  0.0709833  1.50205  1.56552  1.65178  1.70042  ...   5.12765
10289  0.0452765  1.58483  1.67536  1.70635  1.73144  ...   6.49114
10290  0.0245219  1.67846  1.70635  1.72129  1.74851  ...   6.80132

```

```

      gyro_mad  gsr_mean      gsr_std  gsr_mean.1  gsr_max  gsr_variance  \
0         0.3701  1.35538  0.0159845    1.33222  1.39128  0.000255504
1         0.374022  1.33916  0.0111716    1.3189  1.35886  0.000124804
2         0.396192  1.32617  0.0116139    1.30647  1.34776  0.000134884
3         0.43326  1.31359  0.0106492    1.29448    1.334  0.000113405
4         0.535704  1.30208  0.0092008    1.28649  1.32112  8.46548e-05
...      ...
10286  1.28228  1.87392  0.0042255    1.85714  1.89524  1.78548e-05
10287  1.2703  1.87244  0.00466292    1.8381  1.89524  2.17428e-05
10288  1.17004  1.86979  0.00462558    1.8381  1.89231  2.13959e-05
10289  1.11076  1.86688  0.00449201    1.8381  1.88938  2.01781e-05
10290  1.15802  1.86558  0.00413547    1.84249  1.88352  1.71021e-05

```



	gsr_skewness	gsr_kurtosis	score_change
0	0.699293	-0.627312	
1	-0.100332	-1.23867	
2	0.0104838	-1.24345	
3	0.140383	-1.14924	
4	0.139074	-1.14545	
...	...	...	...
10286	1.24196	5.57509	
10287	0.428993	5.26394	
10288	0.49044	4.58249	
10289	0.256191	3.97435	
10290	0.676028	4.51798	

[10291 rows x 39 columns]

```
[118]: x = df_target.copy()
```

```
[119]: x = x.rename(columns={"ECG #": "Participant"})
x
```

```
[119]:
```

	Participant	start_time	end_time	acc_count	\
0	1	2019-11-21 09:35:27	2019-11-21 09:35:29	2271	
1	1	2019-11-21 09:35:28	2019-11-21 09:35:30	2287	
2	1	2019-11-21 09:35:29	2019-11-21 09:35:31	2288	
3	1	2019-11-21 09:35:30	2019-11-21 09:35:32	2294	
4	1	2019-11-21 09:35:31	2019-11-21 09:35:33	2283	
...	...	...	...	...	
10286	6	2019-11-21 11:00:35	2019-11-21 11:00:37	2087	
10287	6	2019-11-21 11:00:36	2019-11-21 11:00:38	2277	
10288	6	2019-11-21 11:00:37	2019-11-21 11:00:39	2279	
10289	6	2019-11-21 11:00:38	2019-11-21 11:00:40	2276	
10290	6	2019-11-21 11:00:39	2019-11-21 11:00:41	1518	

	acc_mean	acc_std	acc_min	acc_25%	acc_50%	acc_75%	...	\
0	1.48804	0.00881535	1.46433	1.48233	1.48774	1.49491	...	
1	1.48771	0.00817652	1.46433	1.48233	1.48774	1.49345	...	
2	1.48407	0.00762258	1.46243	1.47818	1.48289	1.49008	...	
3	1.48755	0.00721254	1.46243	1.48289	1.48838	1.49297	...	
4	1.48692	0.00678993	1.46243	1.48233	1.48838	1.49183	...	
...	...	...	...	...	...	...	...	
10286	1.60173	0.0885796	1.50114	1.53103	1.56665	1.65917	...	
10287	1.58469	0.0558403	1.50114	1.53249	1.57927	1.62692	...	
10288	1.63404	0.0709833	1.50205	1.56552	1.65178	1.70042	...	
10289	1.70117	0.0452765	1.58483	1.67536	1.70635	1.73144	...	
10290	1.72728	0.0245219	1.67846	1.70635	1.72129	1.74851	...	

gyro_kurt	gyro_mad	gsr_mean	gsr_std	gsr_mean.1	gsr_max	\
-----------	----------	----------	---------	------------	---------	---

0	0.0147212	0.3701	1.35538	0.0159845	1.33222	1.39128
1	0.362	0.374022	1.33916	0.0111716	1.3189	1.35886
2	0.951265	0.396192	1.32617	0.0116139	1.30647	1.34776
3	1.18336	0.43326	1.31359	0.0106492	1.29448	1.334
4	0.786448	0.535704	1.30208	0.0092008	1.28649	1.32112
...	...	...	...	...	...	...
10286	4.76593	1.28228	1.87392	0.0042255	1.85714	1.89524
10287	4.27237	1.2703	1.87244	0.00466292	1.8381	1.89524
10288	5.12765	1.17004	1.86979	0.00462558	1.8381	1.89231
10289	6.49114	1.11076	1.86688	0.00449201	1.8381	1.88938
10290	6.80132	1.15802	1.86558	0.00413547	1.84249	1.88352

	gsr_variance	gsr_skewness	gsr_kurtosis	score_change
0	0.000255504	0.699293	-0.627312	
1	0.000124804	-0.100332	-1.23867	
2	0.000134884	0.0104838	-1.24345	
3	0.000113405	0.140383	-1.14924	
4	8.46548e-05	0.139074	-1.14545	
...	...	...	...	...
10286	1.78548e-05	1.24196	5.57509	
10287	2.17428e-05	0.428993	5.26394	
10288	2.13959e-05	0.49044	4.58249	
10289	2.01781e-05	0.256191	3.97435	
10290	1.71021e-05	0.676028	4.51798	

[10291 rows x 39 columns]

```
[123]: x.loc[(x.Participant == 1) & (x.Activity == 3), 'score_change'] = 1
x.loc[(x.Participant == 1) & (x.Activity == 11), 'score_change'] = 0
x.loc[(x.Participant == 1) & (x.Activity == 12), 'score_change'] = 1
x.loc[(x.Participant == 1) & (x.Activity == 13), 'score_change'] = -1
x.loc[(x.Participant == 1) & (x.Activity == 21), 'score_change'] = 3
x.loc[(x.Participant == 1) & (x.Activity == 22), 'score_change'] = -1
x.loc[(x.Participant == 1) & (x.Activity == 23), 'score_change'] = -1
```

```
[126]: x.loc[(x.Participant == 2) & (x.Activity == 1), 'score_change'] = 0
x.loc[(x.Participant == 2) & (x.Activity == 11), 'score_change'] = 1
x.loc[(x.Participant == 2) & (x.Activity == 12), 'score_change'] = -1
x.loc[(x.Participant == 2) & (x.Activity == 13), 'score_change'] = 1
x.loc[(x.Participant == 2) & (x.Activity == 21), 'score_change'] = 1
x.loc[(x.Participant == 2) & (x.Activity == 22), 'score_change'] = -2
x.loc[(x.Participant == 2) & (x.Activity == 23), 'score_change'] = 2
```

```
[128]: x.loc[(x.Participant == 5) & (x.Activity == 1), 'score_change'] = 0
x.loc[(x.Participant == 5) & (x.Activity == 11), 'score_change'] = -2
x.loc[(x.Participant == 5) & (x.Activity == 21), 'score_change'] = 0
x.loc[(x.Participant == 5) & (x.Activity == 22), 'score_change'] = -1
```

```
x.loc[(x.Participant == 5) & (x.Activity == 23), 'score_change'] = -1
```

```
[130]: x.loc[(x.Participant == 6) & (x.Activity == 1), 'score_change'] = -2
x.loc[(x.Participant == 6) & (x.Activity == 2), 'score_change'] = 1
x.loc[(x.Participant == 6) & (x.Activity == 3), 'score_change'] = -1
x.loc[(x.Participant == 6) & (x.Activity == 11), 'score_change'] = -1
x.loc[(x.Participant == 6) & (x.Activity == 12), 'score_change'] = 0
x.loc[(x.Participant == 6) & (x.Activity == 13), 'score_change'] = 0
x.loc[(x.Participant == 6) & (x.Activity == 21), 'score_change'] = 0
x.loc[(x.Participant == 6) & (x.Activity == 22), 'score_change'] = 0
x.loc[(x.Participant == 6) & (x.Activity == 23), 'score_change'] = 0
```

```
[131]: y = x.loc[(x.Participant == 6)]
y
```

```
[131]:
```

	Participant		start_time		end_time	acc_count	\
7718	6	2019-11-21	09:41:46	2019-11-21	09:41:48	2270	
7719	6	2019-11-21	09:41:47	2019-11-21	09:41:49	2264	
7720	6	2019-11-21	09:41:48	2019-11-21	09:41:50	2274	
7721	6	2019-11-21	09:41:49	2019-11-21	09:41:51	2271	
7722	6	2019-11-21	09:41:50	2019-11-21	09:41:52	2275	
...	...	...	...	...	...	...	
10286	6	2019-11-21	11:00:35	2019-11-21	11:00:37	2087	
10287	6	2019-11-21	11:00:36	2019-11-21	11:00:38	2277	
10288	6	2019-11-21	11:00:37	2019-11-21	11:00:39	2279	
10289	6	2019-11-21	11:00:38	2019-11-21	11:00:40	2276	
10290	6	2019-11-21	11:00:39	2019-11-21	11:00:41	1518	

  

	acc_mean	acc_std	acc_min	acc_25%	acc_50%	acc_75%	...	\
7718	2.99992	0.0085862	2.97773	2.99389	2.99927	3.005	...	
7719	3.00216	0.0091704	2.97773	2.99663	2.99927	3.00767	...	
7720	3.00256	0.00917211	2.97773	2.99468	3.00114	3.00781	...	
7721	3.00541	0.00935355	2.97773	2.99744	3.00584	3.01144	...	
7722	3.00479	0.00875686	2.97773	2.99927	3.00405	3.00971	...	
...	...	...	...	...	...	...	...	
10286	1.60173	0.0885796	1.50114	1.53103	1.56665	1.65917	...	
10287	1.58469	0.0558403	1.50114	1.53249	1.57927	1.62692	...	
10288	1.63404	0.0709833	1.50205	1.56552	1.65178	1.70042	...	
10289	1.70117	0.0452765	1.58483	1.67536	1.70635	1.73144	...	
10290	1.72728	0.0245219	1.67846	1.70635	1.72129	1.74851	...	

  

	gyro_kurt	gyro_mad	gsr_mean	gsr_std	gsr_mean.1	gsr_max	\
7718	0.0765793	0.177507	1.27079	0.0019118	1.26163	1.28294	
7719	0.00492086	0.174414	1.27117	0.00189859	1.26163	1.28294	
7720	-0.061004	0.178038	1.27133	0.0019141	1.26163	1.28294	
7721	1.12681	0.324406	1.27119	0.00189484	1.26207	1.28205	
7722	-0.412198	0.366646	1.2709	0.00188809	1.26163	1.28116	

...	...	...	...	...	...	...
10286	4.76593	1.28228	1.87392	0.0042255	1.85714	1.89524
10287	4.27237	1.2703	1.87244	0.00466292	1.8381	1.89524
10288	5.12765	1.17004	1.86979	0.00462558	1.8381	1.89231
10289	6.49114	1.11076	1.86688	0.00449201	1.8381	1.88938
10290	6.80132	1.15802	1.86558	0.00413547	1.84249	1.88352

	gsr_variance	gsr_skewness	gsr_kurtosis	score_change
7718	3.65499e-06	1.07017	7.65327	
7719	3.60465e-06	1.13963	7.99129	
7720	3.66379e-06	1.11323	8.75814	
7721	3.59042e-06	1.08947	8.02794	
7722	3.5649e-06	1.20852	7.93972	

...	...	...	...	...
10286	1.78548e-05	1.24196	5.57509	0
10287	2.17428e-05	0.428993	5.26394	0
10288	2.13959e-05	0.49044	4.58249	0
10289	2.01781e-05	0.256191	3.97435	0
10290	1.71021e-05	0.676028	4.51798	0

[2573 rows x 39 columns]

```
[132]: x.to_csv('test.csv', mode='w', header=True)
```

```
[ ]:
```