```
In [1]:
         # collect data
         import os
         import numpy as np
         import pandas as pd
         root = "./Dog Collar IFT/"
         folder = os.listdir(root)
         datasets = {}
         for fold in folder:
             name = fold
             subfolder = os.listdir(root + fold)
             for subfold in subfolder:
                 if "Activity" in subfold:
                     activity_data = pd.read_csv(root + fold + "/" + subfold)
                     activity_data = pd.DataFrame(np.array(activity_data[['Packet Time', ' Activ
                 elif "Light" in subfold:
                     light_data = pd.read_csv(root + fold + "/" + subfold)
                     light_data = pd.DataFrame(np.array(light_data[['Packet Time', ' Light']]))
                 elif "Audio" in subfold:
                     audio data = pd.read csv(root + fold + "/" + subfold)
                     audio_data = pd.DataFrame(np.array(audio_data[['Packet Time',' Audio']]))
                 elif "IMU" in subfold:
                     imu_data = pd.read_csv(root + fold + "/" + subfold)
                     imu_data['action'] = imu_data[' X']**2 + imu_data[' Y']**2 + imu_data[' Z']
                     imu_data = pd.DataFrame(np.array(imu_data[['Packet Time', ' X', ' Y', ' Z',
                 elif "Pressure" in subfold:
                     pressure_data = pd.read_csv(root + fold + "/" + subfold)
                     pressure_data = pd.DataFrame(np.array(pressure_data[['Packet Time', ' Press
                 elif "Temp " in subfold:
                     temp_data = pd.read_csv(root + fold + "/" + subfold)
                     temp_data = pd.DataFrame(np.array(temp_data[['Packet Time', ' Temperature']
                 elif "Humid" in subfold:
                     humid data = pd.read csv(root + fold + "/" + subfold)
                     humid_data = pd.DataFrame(np.array(humid_data[['Packet Time', ' Humidity']])
             dataset = {'act':activity_data, 'light':light_data, 'audio':audio_data, 'imu':imu_d
             datasets.update({name:dataset})
         print(datasets.keys())
        dict_keys(['2022-10-05-08_29_20'])
```

## Low-Level Fusion (Data Level)

```
In [2]: import DFE_object

In [3]: # Create DFE object
dfeo = DFE_object.DFE_object()

# Upload data
a_data = datasets['2022-10-05-08_29_20']
dfeo.import_from_pandas(a_data['light'], t = 0)
dfeo.import_from_pandas(a_data['audio'], t = 0)
dfeo.import_from_pandas(a_data['imu'], t = 0)
dfeo.import_from_pandas(a_data['press'], t = 0)
dfeo.import_from_pandas(a_data['temp'], t = 0)
dfeo.import_from_pandas(a_data['temp'], t = 0)
dfeo.import_from_pandas(a_data['humid'], t = 0)
```

```
# Temporal Alignment
u_time = pd.DataFrame(a_data['imu'][0]) # we chose to have imu time as the universal ti
dfeo.temporal alignment(u time)
# Normalize (Z-score)
dfeo.normalize('Entry_0', ignore = [0]) # Z-score not the time column
dfeo.normalize('Entry_1', ignore = [0])
dfeo.normalize('Entry 2', ignore = [0])
dfeo.normalize('Entry_3', ignore = [0])
dfeo.normalize('Entry_4', ignore = [0])
dfeo.normalize('Entry_5', ignore = [0])
# Low-Level Data Fusion: Concatenation
dfeo.concatenate()
# Classification: Random Forest
dfeo.random forest()
dfeo.classification report()
```

C:\Users\dmarti22\Anaconda3\lib\site-packages\sklearn\metrics\\_classification.py:1245: U ndefinedMetricWarning: Precision and F-score are ill-defined and being set to 0.0 in lab els with no predicted samples. Use `zero\_division` parameter to control this behavior.

warn prf(average, modifier, msg start, len(result))

C:\Users\dmarti22\Anaconda3\lib\site-packages\sklearn\metrics\\_classification.py:1245: U ndefinedMetricWarning: Precision and F-score are ill-defined and being set to 0.0 in lab els with no predicted samples. Use `zero\_division` parameter to control this behavior.

\_warn\_prf(average, modifier, msg\_start, len(result))

C:\Users\dmarti22\Anaconda3\lib\site-packages\sklearn\metrics\ classification.py:1245: U ndefinedMetricWarning: Precision and F-score are ill-defined and being set to 0.0 in lab els with no predicted samples. Use `zero division` parameter to control this behavior.

warn prf(average, modifier, msg start, len(result)) recall f1-score

precision

•			
0.00	0.00	0.00	50
0.00	0.00	0.00	39
0.00	0.00	0.00	22
0.00	0.00	0.00	41
0.00	0.00	0.00	50
0.67	0.07	0.12	59
0.25	0.99	0.40	111
1.00	1.00	1.00	482
0.00	0.00	0.00	67
0.00	0.00	0.00	94
0.58	0.40	0.47	53
0.96	0.77	0.85	149
1.00	1.00	1.00	310
0.00	0.00	0.00	73
0.97	0.87	0.92	76
0.00	0.00	0.00	50
1.00	1.00	1.00	136
0.78	1.00	0.88	123
1.00	1.00	1.00	236
0.48	0.85	0.62	61
0.46	0.49	0.47	67
0.00	0.00	0.00	37
0.00	0.00	0.00	27
0.00	0.00	0.00	45
0.43	0.60	0.50	81
0.00	0.00	0.00	14
0.00	0.00	0.00	10
0.12	0.91	0.21	91
	0.00 0.00 0.00 0.67 0.25 1.00 0.00 0.58 0.96 1.00 0.97 0.00 1.00 0.78 1.00 0.48 0.46 0.00 0.00 0.43 0.00	0.00       0.00         0.00       0.00         0.00       0.00         0.00       0.00         0.67       0.07         0.25       0.99         1.00       1.00         0.00       0.00         0.00       0.00         0.58       0.40         0.96       0.77         1.00       1.00         0.00       0.00         0.97       0.87         0.00       0.00         1.00       1.00         1.00       1.00         1.00       1.00         0.48       0.85         0.46       0.49         0.00       0.00         0.00       0.00         0.43       0.60         0.00       0.00         0.00       0.00	0.00       0.00       0.00         0.00       0.00       0.00         0.00       0.00       0.00         0.00       0.00       0.00         0.00       0.00       0.00         0.67       0.07       0.12         0.25       0.99       0.40         1.00       1.00       1.00         0.00       0.00       0.00         0.00       0.00       0.00         0.58       0.40       0.47         0.96       0.77       0.85         1.00       1.00       1.00         0.97       0.87       0.92         0.00       0.00       0.00         0.97       0.87       0.92         0.00       0.00       0.00         1.00       1.00       1.00         1.00       1.00       1.00         1.00       1.00       1.00         0.78       1.00       0.88         1.00       1.00       1.00         0.48       0.85       0.62         0.46       0.49       0.47         0.00       0.00       0.00         0.00       0.00

```
28.0
                               0.00
                                                                   72
                                          0.00
                                                      0.00
                  29.0
                                                      0.99
                                                                   70
                               1.00
                                          0.97
                  30.0
                                                                   73
                               0.44
                                          0.32
                                                      0.37
                  31.0
                               0.00
                                          0.00
                                                      0.00
                                                                   57
                                                                   98
                  32.0
                               0.98
                                          0.49
                                                      0.65
                  33.0
                               0.00
                                          0.00
                                                                   11
                                                      0.00
                  34.0
                                          0.00
                               0.00
                                                      0.00
                                                                    8
                  35.0
                               0.00
                                          0.00
                                                      0.00
                                                                   13
                  36.0
                               0.00
                                          0.00
                                                      0.00
                                                                   13
                  37.0
                               0.43
                                          0.25
                                                      0.32
                                                                   91
                   38.0
                               0.00
                                          0.00
                                                      0.00
                                                                   26
                  39.0
                               0.00
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                                                                   16
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                                                                   17
                  41.0
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                                          0.00
                                                      0.00
                  42.0
                               0.16
                                          0.92
                                                      0.28
                                                                   89
                  43.0
                               1.00
                                          0.04
                                                      0.07
                                                                   54
                                                      0.00
                  44.0
                                          0.00
                                                                   51
                               0.00
                  45.0
                                          0.51
                                                      0.57
                                                                   68
                               0.65
                  46.0
                               0.61
                                          0.20
                                                      0.31
                                                                   83
                  47.0
                               0.94
                                                                   90
                                          0.18
                                                      0.30
                                                                   76
                  48.0
                               0.00
                                          0.00
                                                      0.00
                  49.0
                               0.00
                                          0.00
                                                      0.00
                                                                   44
                                                      0.56
                                                                 3789
              accuracy
                                          0.30
                                                      0.27
                                                                 3789
                               0.32
             macro avg
         weighted avg
                               0.57
                                          0.56
                                                      0.52
                                                                 3789
                                                       400
                                                       300
         True label
                                                       - 200
                                                       - 100
               OL23456789000294966789002224926789882993969499456789
                          Predicted label
Out[3]: (array([[ 0,
                         0,
                              0, ...,
                                            0,
                                                 0],
                    0,
                         0,
                                        0,
                                                 0],
                                            0,
                   [ 0,
                   [ 0,
                              0, ..., 16,
                         0,
                                        0,
                                            0,
                                                 0],
                    0,
                              0, ...,
                                                 0]], dtype=int64),
                                        0,
                                            0,
```

## Mid-Level Fusion (Feature Level)

```
In [4]: # Feature Extraction

dfeo.set_window_scheme(length = 30000, overlap = 5000) # set windowing scheme

dfeo.fe_average("Entry_0")
    dfeo.fe_average("Entry_1")
```

<sklearn.metrics. plot.confusion matrix.ConfusionMatrixDisplay at 0x2627e35cca0>)

```
dfeo.fe average("Entry 2")
 dfeo.fe average("Entry 3")
 dfeo.fe_average("Entry_4")
dfeo.fe_average("Entry_5")
dfeo.fe variance("Entry 0")
dfeo.fe variance("Entry 1")
dfeo.fe variance("Entry 2")
dfeo.fe_variance("Entry_3")
 dfeo.fe variance("Entry 4")
dfeo.fe variance("Entry 5")
 dfeo.fe skewness("Entry 0")
dfeo.fe_skewness("Entry_1")
dfeo.fe skewness("Entry 2")
dfeo.fe skewness("Entry 3")
dfeo.fe skewness("Entry 4")
dfeo.fe_skewness("Entry_5")
dfeo.fe kurtosis("Entry 0")
dfeo.fe kurtosis("Entry 1")
dfeo.fe_kurtosis("Entry_2")
dfeo.fe kurtosis("Entry 3")
dfeo.fe kurtosis("Entry 4")
dfeo.fe kurtosis("Entry 5")
dfeo.fe_peak_count("Entry_0")
dfeo.fe peak count("Entry 1")
dfeo.fe peak count("Entry 2")
dfeo.fe_peak_count("Entry_3")
dfeo.fe peak count("Entry 4")
dfeo.fe_peak_count("Entry_5")
dfeo.fe RMS("Entry 0")
dfeo.fe_RMS("Entry_1")
 dfeo.fe RMS("Entry 2")
dfeo.fe_RMS("Entry_3")
dfeo.fe RMS("Entry 4")
dfeo.fe RMS("Entry 5")
dfeo.classification windowing("Entry 6")
# Mid-Level Fusion: Concatenation
dfeo.concatenate() # should be able to specify that we're concatenating features for a
                    # defaults to current window scheme if unspecified
# Dimensional Reduction: PCA
dfeo.my_PCA("active")
# Classification: Random Forest
dfeo.random forest()
dfeo.classification report()
Calculated average on dataset Entry_0 at 2023-05-19 09:54:53.494914.
Calculated average on dataset Entry_1 at 2023-05-19 09:54:53.513347.
Calculated average on dataset Entry_2 at 2023-05-19 09:54:53.533341.
Calculated average on dataset Entry_3 at 2023-05-19 09:54:53.540405.
Calculated average on dataset Entry_4 at 2023-05-19 09:54:53.556066.
Calculated average on dataset Entry 5 at 2023-05-19 09:54:53.556066.
```

Calculated variance on dataset Entry 0 at 2023-05-19 09:54:53.571690.

```
Calculated variance on dataset Entry 2 at 2023-05-19 09:54:53.618533.
Calculated variance on dataset Entry 3 at 2023-05-19 09:54:53.645025.
Calculated variance on dataset Entry_4 at 2023-05-19 09:54:53.656547.
Calculated variance on dataset Entry_5 at 2023-05-19 09:54:53.672173.
Calculated skewness on dataset Entry_0 at 2023-05-19 09:54:53.772856.
Calculated skewness on dataset Entry 1 at 2023-05-19 09:54:53.871834.
Calculated skewness on dataset Entry 2 at 2023-05-19 09:54:54.213746.
Calculated skewness on dataset Entry_3 at 2023-05-19 09:54:54.292548.
Calculated skewness on dataset Entry_4 at 2023-05-19 09:54:54.393002.
Calculated skewness on dataset Entry_5 at 2023-05-19 09:54:54.476616.
Calculated kurtosis on dataset Entry_0 at 2023-05-19 09:54:54.593334.
Calculated kurtosis on dataset Entry_1 at 2023-05-19 09:54:54.720945.
Calculated kurtosis on dataset Entry_2 at 2023-05-19 09:54:55.093792.
Calculated kurtosis on dataset Entry 3 at 2023-05-19 09:54:55.193919.
Calculated kurtosis on dataset Entry_4 at 2023-05-19 09:54:55.293835.
Calculated kurtosis on dataset Entry_5 at 2023-05-19 09:54:55.396357.
Calculated peak_count on dataset Entry_0 at 2023-05-19 09:54:55.456080.
Calculated peak count on dataset Entry 1 at 2023-05-19 09:54:55.511138.
Calculated peak_count on dataset Entry_2 at 2023-05-19 09:54:55.729975.
Calculated peak_count on dataset Entry_3 at 2023-05-19 09:54:55.805342.
Calculated peak_count on dataset Entry_4 at 2023-05-19 09:54:55.873442.
Calculated peak_count on dataset Entry_5 at 2023-05-19 09:54:55.920318.
Calculated RMS on dataset Entry 0 at 2023-05-19 09:54:55.943787.
Calculated RMS on dataset Entry_1 at 2023-05-19 09:54:55.965563.
Calculated RMS on dataset Entry 2 at 2023-05-19 09:54:55.985561.
Calculated RMS on dataset Entry_3 at 2023-05-19 09:54:55.992622.
Calculated RMS on dataset Entry_4 at 2023-05-19 09:54:56.008249.
Calculated RMS on dataset Entry_5 at 2023-05-19 09:54:56.008249.
Calculated classification on dataset Entry 6 at 2023-05-19 09:54:56.086376.
C:\Users\dmarti22\Anaconda3\lib\site-packages\sklearn\metrics\ classification.py:1245: U
ndefinedMetricWarning: Precision and F-score are ill-defined and being set to 0.0 in lab
els with no predicted samples. Use `zero division` parameter to control this behavior.
   warn prf(average, modifier, msg start, len(result))
C:\Users\dmarti22\Anaconda3\lib\site-packages\sklearn\metrics\_classification.py:1245: U
ndefinedMetricWarning: Recall and F-score are ill-defined and being set to 0.0 in labels
with no true samples. Use `zero_division` parameter to control this behavior.
  _warn_prf(average, modifier, msg_start, len(result))
C:\Users\dmarti22\Anaconda3\lib\site-packages\sklearn\metrics\_classification.py:1245: U
ndefinedMetricWarning: Precision and F-score are ill-defined and being set to 0.0 in lab
els with no predicted samples. Use `zero division` parameter to control this behavior.
  _warn_prf(average, modifier, msg_start, len(result))
C:\Users\dmarti22\Anaconda3\lib\site-packages\sklearn\metrics\ classification.py:1245: U
ndefinedMetricWarning: Recall and F-score are ill-defined and being set to 0.0 in labels
with no true samples. Use `zero_division` parameter to control this behavior.
  _warn_prf(average, modifier, msg_start, len(result))
C:\Users\dmarti22\Anaconda3\lib\site-packages\sklearn\metrics\_classification.py:1245: U
ndefinedMetricWarning: Precision and F-score are ill-defined and being set to 0.0 in lab
els with no predicted samples. Use `zero division` parameter to control this behavior.
  warn prf(average, modifier, msg start, len(result))
C:\Users\dmarti22\Anaconda3\lib\site-packages\sklearn\metrics\ classification.py:1245: U
ndefinedMetricWarning: Recall and F-score are ill-defined and being set to 0.0 in labels
with no true samples. Use `zero_division` parameter to control this behavior.
  warn prf(average, modifier, msg start, len(result))
```

Calculated variance on dataset Entry 1 at 2023-05-19 09:54:53.587283.

	precision	recall	f1-score	support
5.0	0.00	0.00	0.00	1
6.0	0.00	0.00	0.00	1
7.0	0.00	0.00	0.00	1
12.0	0.50	0.67	0.57	3
13.0	0.00	0.00	0.00	1
14.0	0.00	0.00	0.00	1
16.0	1.00	0.50	0.67	2
17.0	0.00	0.00	0.00	0
18.0	0.40	1.00	0.57	2

```
20.0
                          0.00
                                    0.00
                                              0.00
                                                          1
                24.0
                          0.00
                                    0.00
                                                          2
                                              0.00
                27.0
                                                          2
                                    0.00
                          0.00
                                              0.00
                28.0
                          0.00
                                    0.00
                                              0.00
                                                          1
                                                          2
                31.0
                          0.00
                                    0.00
                                             0.00
                                                          0
                32.0
                          0.00
                                    0.00
                                             0.00
                37.0
                                    1.00
                                                          1
                          0.33
                                              0.50
                42.0
                          1.00
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                                              1.00
               45.0
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                                                          1
                46.0
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                                    0.00
                                              0.00
                                                          0
               47.0
                          0.00
                                    0.00
                                              0.00
                                                          0
                                              0.30
                                                         23
            accuracy
                                                         23
           macro avg
                          0.16
                                    0.21
                                              0.17
        weighted avg
                          0.24
                                    0.30
                                              0.25
                                                         23
                                               2.00
          1
2
3
4
5
6
7
8
9
10
11
12
13
14
15
16
17
18
                                               1.75
                                               1.50
                                              1.25
                                              - 1.00
                                               0.75
                                               0.50
                                               0.25
                                               0.00
             0 1 2 3 4 5 6 7 8 910111213141516171819
                      Predicted label
0, 0, 0,
                                          0,
                                             0,
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                                                      0,
                                       1,
                                          0,
                                             0,
                                                0,
                                                   0,
                                    0,
                                                      0,
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                                             0,
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                [0, 0, 1, 2, 0, 0, 0, 0, 0,
                                                      0, 0,
                [0, 0, 0, 0, 0, 0, 0, 0, 1,
                                          0, 0, 0, 0, 0, 0, 0, 0,
                [0, 0, 0, 0, 0, 0,
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                         0, 0, 1, 0, 0, 0,
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                                                0, 0, 0, 0, 1,
                            0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,
                [0, 0, 0, 1,
                            0, 0, 0, 0, 0, 1,
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                [0, 0, 0, 0, 0,
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                               0,
                                  0, 0, 0,
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                                                         0,
                                                               0,
                [0, 0, 0, 0,
                                                            1,
                                       0,
                                          0,
                                             0,
                                                0,
                                                   0,
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                [0, 0, 0, 0, 0, 0,
                                  0, 0,
                                                         0,
                [0, 0, 0, 0, 0, 0, 0, 0, 0,
                                          0,
                                             0,
                                                0,
                                                   0,
                                                      0, 0, 0,
                dtype=int64),
         <sklearn.metrics. plot.confusion matrix.ConfusionMatrixDisplay at 0x2627fa424c0>)
```

## Mid-Level Fusion (LDA instead of PCA and Random Forest)

```
In [5]:
         # Feature Extraction
         dfeo.set window scheme(length = 30000, overlap = 5000) # set windowing scheme
         dfeo.fe average("Entry 0")
         dfeo.fe_average("Entry_1")
         dfeo.fe average("Entry 2")
         dfeo.fe_average("Entry_3")
         dfeo.fe_average("Entry_4")
         dfeo.fe_average("Entry_5")
         dfeo.fe variance("Entry 0")
         dfeo.fe_variance("Entry_1")
         dfeo.fe_variance("Entry_2")
         dfeo.fe_variance("Entry_3")
         dfeo.fe variance("Entry 4")
         dfeo.fe_variance("Entry_5")
         dfeo.fe skewness("Entry 0")
         dfeo.fe_skewness("Entry_1")
         dfeo.fe skewness("Entry 2")
         dfeo.fe_skewness("Entry_3")
         dfeo.fe_skewness("Entry_4")
         dfeo.fe_skewness("Entry_5")
         dfeo.fe kurtosis("Entry 0")
         dfeo.fe kurtosis("Entry 1")
         dfeo.fe_kurtosis("Entry_2")
         dfeo.fe_kurtosis("Entry_3")
         dfeo.fe_kurtosis("Entry_4")
         dfeo.fe kurtosis("Entry 5")
         dfeo.fe_peak_count("Entry_0")
         dfeo.fe_peak_count("Entry_1")
         dfeo.fe peak count("Entry 2")
         dfeo.fe_peak_count("Entry_3")
         dfeo.fe_peak_count("Entry_4")
         dfeo.fe_peak_count("Entry_5")
         dfeo.fe RMS("Entry 0")
         dfeo.fe_RMS("Entry_1")
         dfeo.fe_RMS("Entry_2")
         dfeo.fe_RMS("Entry_3")
         dfeo.fe RMS("Entry 4")
         dfeo.fe RMS("Entry 5")
         dfeo.classification_windowing("Entry_6")
         # Mid-Level Fusion: Concatenation
         dfeo.concatenate() # should be able to specify that we're concatenating features for a
                             # defaults to current window scheme if unspecified
         # Dimensional Reduction and Classification: LDA
         dfeo.my_LDA(5)
         dfeo.classification report()
```

Calculated average on dataset Entry\_0 at 2023-05-19 09:55:00.013356. Calculated average on dataset Entry\_1 at 2023-05-19 09:55:00.029919. Calculated average on dataset Entry\_2 at 2023-05-19 09:55:00.046860.

```
Calculated average on dataset Entry 3 at 2023-05-19 09:55:00.057857.
Calculated average on dataset Entry_4 at 2023-05-19 09:55:00.068856.
Calculated average on dataset Entry 5 at 2023-05-19 09:55:00.079950.
Calculated variance on dataset Entry_0 at 2023-05-19 09:55:00.079950.
Calculated variance on dataset Entry_1 at 2023-05-19 09:55:00.095575.
Calculated variance on dataset Entry_2 at 2023-05-19 09:55:00.126827.
Calculated variance on dataset Entry 3 at 2023-05-19 09:55:00.142486.
Calculated variance on dataset Entry 4 at 2023-05-19 09:55:00.158110.
Calculated variance on dataset Entry_5 at 2023-05-19 09:55:00.173735.
Calculated skewness on dataset Entry_0 at 2023-05-19 09:55:00.257658.
Calculated skewness on dataset Entry_1 at 2023-05-19 09:55:00.351410.
Calculated skewness on dataset Entry_2 at 2023-05-19 09:55:00.638355.
Calculated skewness on dataset Entry_3 at 2023-05-19 09:55:00.724150.
Calculated skewness on dataset Entry_4 at 2023-05-19 09:55:00.823439.
Calculated skewness on dataset Entry 5 at 2023-05-19 09:55:00.911318.
Calculated kurtosis on dataset Entry_0 at 2023-05-19 09:55:00.989483.
Calculated kurtosis on dataset Entry_1 at 2023-05-19 09:55:01.062749.
Calculated kurtosis on dataset Entry_2 at 2023-05-19 09:55:01.359391.
Calculated kurtosis on dataset Entry_3 at 2023-05-19 09:55:01.439668.
Calculated kurtosis on dataset Entry_4 at 2023-05-19 09:55:01.517792.
Calculated kurtosis on dataset Entry_5 at 2023-05-19 09:55:01.605709.
Calculated peak count on dataset Entry 0 at 2023-05-19 09:55:01.676926.
Calculated peak count on dataset Entry 1 at 2023-05-19 09:55:01.739427.
Calculated peak_count on dataset Entry_2 at 2023-05-19 09:55:01.941345.
Calculated peak_count on dataset Entry_3 at 2023-05-19 09:55:02.003880.
Calculated peak count on dataset Entry 4 at 2023-05-19 09:55:02.066380.
Calculated peak_count on dataset Entry_5 at 2023-05-19 09:55:02.113218.
Calculated RMS on dataset Entry_0 at 2023-05-19 09:55:02.128879.
Calculated RMS on dataset Entry_1 at 2023-05-19 09:55:02.140536.
Calculated RMS on dataset Entry 2 at 2023-05-19 09:55:02.165893.
Calculated RMS on dataset Entry_3 at 2023-05-19 09:55:02.179853.
Calculated RMS on dataset Entry_4 at 2023-05-19 09:55:02.181671.
Calculated RMS on dataset Entry_5 at 2023-05-19 09:55:02.197326.
Calculated classification on dataset Entry 6 at 2023-05-19 09:55:02.212956.
C:\Users\dmarti22\Anaconda3\lib\site-packages\sklearn\metrics\ classification.py:1245: U
ndefinedMetricWarning: Precision and F-score are ill-defined and being set to 0.0 in lab
els with no predicted samples. Use `zero_division` parameter to control this behavior.
  _warn_prf(average, modifier, msg_start, len(result))
C:\Users\dmarti22\Anaconda3\lib\site-packages\sklearn\metrics\_classification.py:1245: U
ndefinedMetricWarning: Recall and F-score are ill-defined and being set to 0.0 in labels
with no true samples. Use `zero division` parameter to control this behavior.
  warn_prf(average, modifier, msg_start, len(result))
C:\Users\dmarti22\Anaconda3\lib\site-packages\sklearn\metrics\_classification.py:1245: U
ndefinedMetricWarning: Precision and F-score are ill-defined and being set to 0.0 in lab
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C:\Users\dmarti22\Anaconda3\lib\site-packages\sklearn\metrics\_classification.py:1245: U
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  warn prf(average, modifier, msg start, len(result))
C:\Users\dmarti22\Anaconda3\lib\site-packages\sklearn\metrics\ classification.py:1245: U
ndefinedMetricWarning: Precision and F-score are ill-defined and being set to 0.0 in lab
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  _warn_prf(average, modifier, msg_start, len(result))
C:\Users\dmarti22\Anaconda3\lib\site-packages\sklearn\metrics\ classification.py:1245: U
ndefinedMetricWarning: Recall and F-score are ill-defined and being set to 0.0 in labels
with no true samples. Use `zero_division` parameter to control this behavior.
  _warn_prf(average, modifier, msg_start, len(result))
              precision
                          recall f1-score
         5.0
                   0.00
                             0.00
                                       0.00
         6.0
                   0.50
                             1.00
                                       0.67
                                                    1
         7.0
                             0.00
                                                    1
                   0.00
                                       0.00
```

11.0

12.0

1.00

0.67

1.00

0.67

1.00

0.67

1

3

```
1.00
         13.0
                                  1.00
                                               1.00
                                                              1
         14.0
                                   1.00
                      0.50
                                              0.67
                                                              1
         16.0
                                   1.00
                                                              1
                      1.00
                                               1.00
         17.0
                      0.00
                                  0.00
                                              0.00
                                                              1
         18.0
                      1.00
                                  0.67
                                              0.80
                                                              3
                                                              0
         20.0
                      0.00
                                  0.00
                                              0.00
                                  1.00
                                                              1
         24.0
                      0.50
                                              0.67
         27.0
                      1.00
                                  1.00
                                              1.00
                                                              1
         28.0
                      0.00
                                  0.00
                                              0.00
                                                              1
                                                              2
         32.0
                      1.00
                                  0.50
                                              0.67
                                                              1
         42.0
                      0.50
                                   1.00
                                              0.67
         43.0
                      0.00
                                  0.00
                                                              0
                                              0.00
         44.0
                      0.00
                                  0.00
                                              0.00
                                                              1
                                                              2
         45.0
                      0.00
                                  0.00
                                              0.00
                                                              0
         46.0
                      0.00
                                  0.00
                                               0.00
         47.0
                      0.00
                                              0.00
                                                              0
                                  0.00
                                               0.57
    accuracy
                                                             23
   macro avg
                      0.41
                                   0.47
                                               0.42
                                                             23
                      0.57
                                               0.54
                                                             23
weighted avg
                                   0.57
                                                2.00
                                                1.75
  4
5
6
7
8
9
10
11
12
13
14
15
16
17
                                               1.50
                                                - 1.25
                                               1.00
                                               0.75
                                                0.50
  18
19
                                                0.25
```

0 1 2 3 4 5 6 7 8 91011121314151617181920 Predicted label

[0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 1, 0, 2, 0, 1, 0, [0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0], [0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0], [0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0],0, 0, 2, 0, 0, 0, 0, [0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, [0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, [0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, [0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 1, 0, [0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, [0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0], 0, [0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0], 0, 0, 1, 0], dtype=int64),

0.00

<sklearn.metrics.\_plot.confusion\_matrix.ConfusionMatrixDisplay at 0x2627fa2c5e0>)