

intensity_classification_model_backup

January 12, 2021

1 Intensity recognition

Activity intensity	MET range
Light	< 3.00
Moderate	3.00 - 5.99
Hard	6.00 - 8.99
Very hard	> 8.99

Activity count cut-points bu : Freedson et al.(1998)

```
[4]: from helpers          import pandas_helper as pdh, math_helper as mth
    from utils          import read_functions
    from sensors.activpal    import *
    from sensors.vyntus     import *
    from sklearn.model_selection import train_test_split
    from sklearn.linear_model import LogisticRegression
    from sklearn.ensemble    import RandomForestClassifier
    from sklearn.metrics     import confusion_matrix, classification_report, \
        accuracy_score, precision_score, recall_score, roc_auc_score, f1_score

    import matplotlib.pyplot as plt
    import pandas            as pd

    activpal = Activpal()
    vyntus = Vyntus()

    respondents_df = pdh.read_csv_respondents()
    test_users     = ['BMR032', 'BMR042', 'BMR098']
```

2 Preparing dataset

```
[5]: def get_vyntus_df(correspondent, start, stop):
    intensity_intervals = [0, 3, 5.99, 8.99, 1000]
    intensity_labels = ['Light', 'Moderate', 'Hard', 'Very Hard']

    vyntus_df = vyntus.read_data(correspondent, start, stop)

    if vyntus_df.empty:
        return pd.DataFrame()

    corr_number = int(correspondent.replace('BMRO', ''))

    weight = respondents_df['gewicht'][corr_number]
    length_m = respondents_df['lengte'][corr_number] / 100

    vyntus_df['vyn_V02'] = [float(vo2.replace(',', '.')) if type(vo2) == str
    → else vo2 for vo2 in vyntus_df['vyn_V02']]
    vyntus_df['met'] = mth.calculate_met(vyntus_df['vyn_V02'], weight)
    vyntus_df['weight'] = weight
    vyntus_df['bmi'] = mth.calculate_bmi(weight, length_m)

    vyntus_df = vyntus_df.resample('60s').mean()[:-1]

    vyntus_df['intensity'] = pd.cut(vyntus_df.met, intensity_intervals,
    → labels=intensity_labels)

    return vyntus_df;
```

```
[6]: def get_activpal_df(correspondent, start, stop):
    activpal_df = activpal.read_data(correspondent, start, stop)

    activpal_df = activpal_df[['pal_accX', 'pal_accY', 'pal_accZ']].apply(mth.
    → convert_value_to_g)

    x_abs = abs(activpal_df['pal_accX'])
    y_abs = abs(activpal_df['pal_accY'])
    z_abs = abs(activpal_df['pal_accZ'])

    activpal_df['mag_acc'] = mth.to_mag_acceleration(x_abs, y_abs, z_abs)

    return activpal_df.resample('60s').sum()[:-1]
```

```
[7]: def get_dataset_of_correspondent(correspondent):
    dataset_df = pd.DataFrame(columns=['sum_mag_of_acc', 'mean_met',
    → 'intensity', 'activity'], index=pd.to_datetime([]))
```

```

activities = ['lopen', 'rennen', 'springen', 'staan', 'traplopen', 'zitten']

activities_df = read_functions.read_activities(correspondent)

for activity_name in activities:
    activity = activities_df.loc[activity_name]

    if not activity.empty:

        start_time = activity.start
        stop_time = activity.stop

        activpal_df = get_activpal_df(correspondent, start_time, stop_time)
        vyntus_df = get_vyntus_df(correspondent, start_time, stop_time)

        if not vyntus_df.empty and not activpal_df.empty:
            activity_dataset_df = pd.DataFrame(index=activpal_df.index)

            activity_dataset_df['sum_mag_of_acc'] = activpal_df['mag_acc']
            activity_dataset_df['mean_met'] = vyntus_df['met']
            activity_dataset_df['weight'] = vyntus_df['weight']
            activity_dataset_df['bmi'] = vyntus_df['bmi']
            activity_dataset_df['intensity'] = vyntus_df['intensity']
            activity_dataset_df['activity'] = activity_name

            activity_dataset_df.dropna(how='any', inplace=True)

            dataset_df = pd.concat([activity_dataset_df, dataset_df])

return dataset_df

```

```

[8]: def create_dataset_from_correspondents(correspondents):
    dataset_df = pd.DataFrame(index=pd.to_datetime([]))

    for correspondent in correspondents:
        print("Extracting " + correspondent)

        correspondent_dataset_df = get_dataset_of_correspondent(correspondent)
        dataset_df = pd.concat([dataset_df, correspondent_dataset_df])

    print("Done creating dataset")

    return dataset_df

def create_dataset_from_all_correspondents(exclude_test_correspondent = True):

```

```

exclude = ['output', 'throughput', 'Test data', '.ipynb_checkpoints',
↳ 'BRM015', 'BMR035', 'BMR100', 'BMR051', 'BMR027']

if (exclude_test_correspondent):
    exclude = exclude + test_users

correspondents = []

for directory in os.walk('../..data'):
    if directory[0] == '../..data':
        correspondents = directory[1]

for exclude_item in exclude:
    if exclude_item in correspondents:
        correspondents.remove(exclude_item)

return create_dataset_from_correspondents(correspondents)

```

```
[9]: dataset = create_dataset_from_all_correspondents()
```

```

Extracting BMR099
Extracting BMR025
Could not read file: ../../data/BMR025/vyntus.csv
Could not read file: ../../data/BMR025/vyntus.csv
Could not read file: ../../data/BMR025/vyntus.csv
Could not read file: ../../data/BMR025/vyntus.csv
Could not read file: ../../data/BMR025/vyntus.csv
Could not read file: ../../data/BMR025/vyntus.csv
Extracting BMR060
No data for respondnet: BMR060
No data for respondnet: BMR060
No data for respondnet: BMR060
No data for respondnet: BMR060
No data for respondnet: BMR060
No data for respondnet: BMR060
Extracting BMR012
Extracting BMR030
Extracting BMR044
Extracting BMR043
Extracting BMR004
Extracting BMR011
Extracting BMR034
Extracting BMR014
Extracting BMR036
Extracting BMR052
Extracting BMR002

```

```
Extracting BMR031
Extracting BMR097
Extracting BMR008
Extracting BMR015
Extracting BMR033
Extracting BMR064
Extracting BMR055
Extracting BMR041
Extracting BMR053
Extracting BMR018
Extracting BMR058
Extracting BMR040
Done creating dataset
```

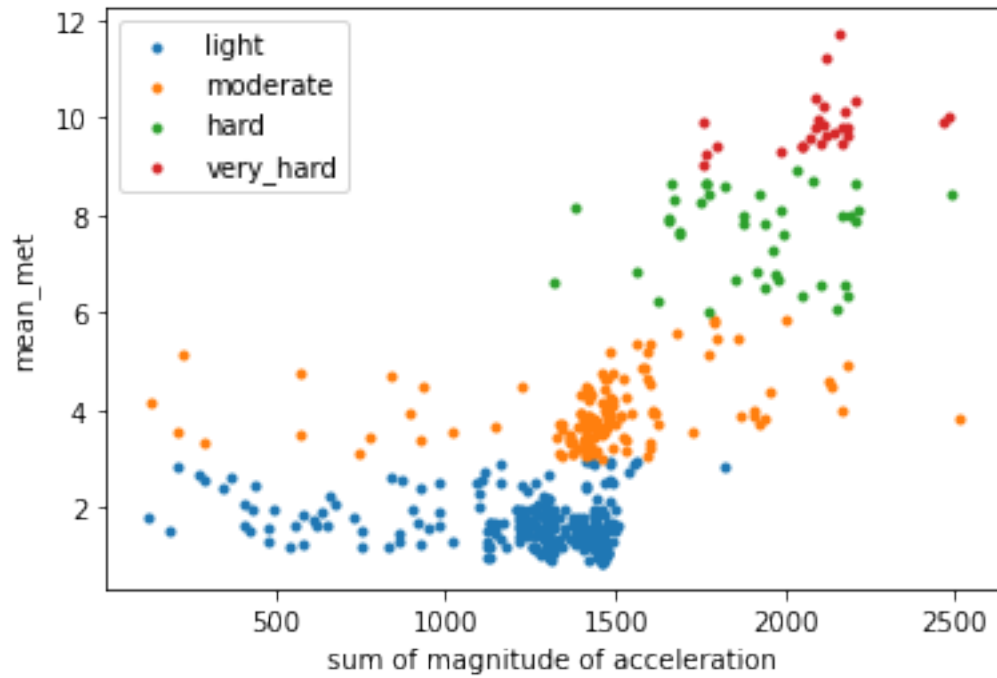
2.1 Dataset analysis

```
[10]: def plot_intensity_on_accel(dataset):
    light = dataset.loc[(dataset['intensity'] == 'Light') ]
    moderate = dataset.loc[(dataset['intensity'] == 'Moderate') ]
    hard = dataset.loc[(dataset['intensity'] == 'Hard') ]
    very_hard = dataset.loc[(dataset['intensity'] == 'Very Hard') ]

    plt.scatter(light['sum_mag_of_acc'], light['mean_met'], marker='.',
    ↪label='light')
    plt.scatter(moderate['sum_mag_of_acc'], moderate['mean_met'], marker='.',
    ↪label='moderate')
    plt.scatter(hard['sum_mag_of_acc'], hard['mean_met'], marker='.',
    ↪label='hard')
    plt.scatter(very_hard['sum_mag_of_acc'], very_hard['mean_met'], marker='.',
    ↪label='very_hard')

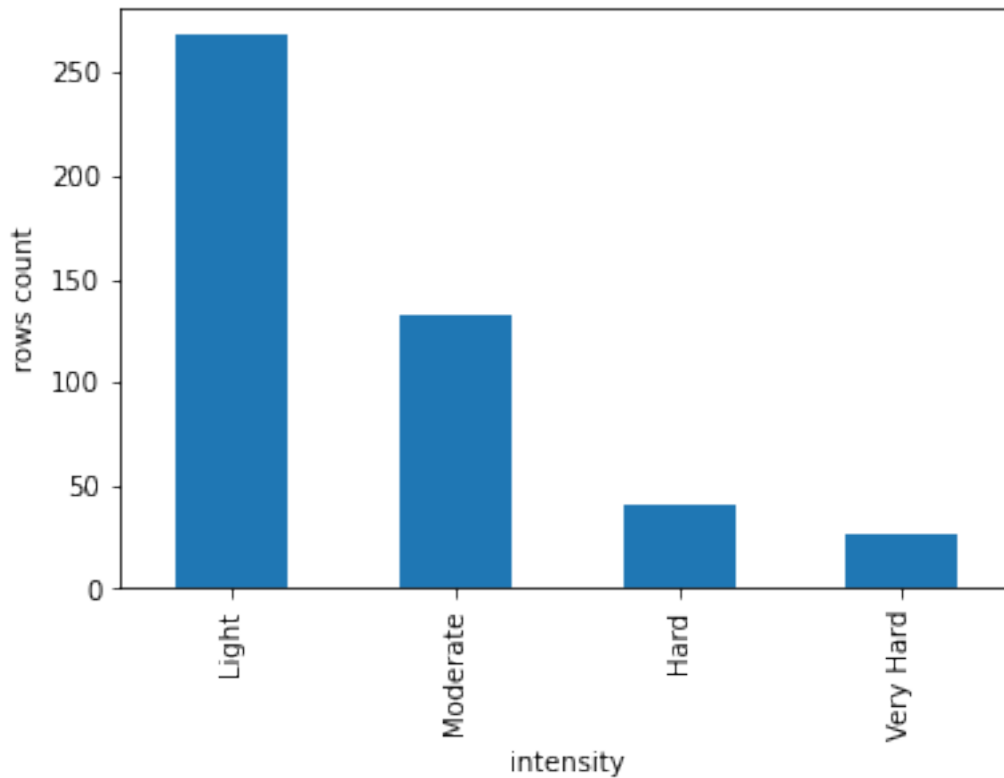
    plt.ylabel('mean_met')
    plt.xlabel('sum of magnitude of acceleration')
    plt.legend()
    plt.show()

plot_intensity_on_accel(dataset)
```



```
[11]: dataset['intensity'].value_counts().plot.bar(ylabel='rows_␣  
      ↳count',xlabel='intensity')
```

```
[11]: <matplotlib.axes._subplots.AxesSubplot at 0x7ff0a16f20b8>
```



2.2 Defining categories

```
[12]: intensity_columns = ['intensity_light', 'intensity_moderate', 'intensity_hard', 'intensity_very_hard']
dataset[intensity_columns] = 0

dataset.loc[(dataset['intensity'] == 'Light'), 'intensity_light'] = 1
dataset.loc[(dataset['intensity'] == 'Moderate'), 'intensity_moderate'] = 1
dataset.loc[(dataset['intensity'] == 'Hard'), 'intensity_hard'] = 1
dataset.loc[(dataset['intensity'] == 'Very Hard'), 'intensity_very_hard'] = 1

dataset.dropna(how='any', inplace=True)
```

2.3 Splitting dataset

```
[13]: x = dataset[['sum_mag_of_acc', 'bmi']]
y = dataset[intensity_columns]
```

```
train_x, valid_x, train_y, valid_y = train_test_split(x, y, test_size=0.2,
↳random_state=23, stratify=y)

len(x)
```

[13]: 467

2.4 Training dataset

```
[25]: rfc = RandomForestClassifier(n_estimators=23,random_state=0)
rfc.fit(train_x, train_y)
```

[25]: RandomForestClassifier(n_estimators=23, random_state=0)

```
[26]: prediction_y = rfc.predict(valid_x)
```

2.5 Result analysis

2.5.1 Validation

```
[27]: accuracy_score(valid_y, prediction_y, normalize=True)
```

[27]: 0.7446808510638298

```
[28]: print(classification_report(valid_y, prediction_y,
↳target_names=intensity_columns, zero_division=0))
```

	precision	recall	f1-score	support
intensity_light	0.87	0.96	0.91	54
intensity_moderate	0.80	0.44	0.57	27
intensity_hard	0.40	0.50	0.44	8
intensity_very_hard	0.40	0.40	0.40	5
micro avg	0.78	0.74	0.76	94
macro avg	0.62	0.58	0.58	94
weighted avg	0.78	0.74	0.75	94
samples avg	0.74	0.74	0.74	94

2.5.2 Test

```
[29]: test_dataset = create_dataset_from_correspondents(test_users)

test_dataset[intensity_columns] = 0

test_dataset.loc[(test_dataset['intensity'] == 'Light'), 'intensity_light'] = 1
test_dataset.loc[(test_dataset['intensity'] == 'Moderate'), 'intensity_moderate'] = 1
test_dataset.loc[(test_dataset['intensity'] == 'Hard'), 'intensity_hard'] = 1
test_dataset.loc[(test_dataset['intensity'] == 'VeryHard'), 'intensity_very_hard'] = 1

test_dataset.dropna(how='any', inplace=True)

x = test_dataset[['sum_mag_of_acc', 'bmi']]
y = test_dataset[intensity_columns]
```

Extracting BMR032
Extracting BMR042
Extracting BMR098
Done creating dataset

```
[30]: test_prediction_y = rfc.predict(x)
```

```
[31]: accuracy_score(y, test_prediction_y, normalize=True)
```

```
[31]: 0.6610169491525424
```

```
[32]: print(classification_report(y, test_prediction_y,
    target_names=intensity_columns, zero_division=0))
```

	precision	recall	f1-score	support
intensity_light	0.79	0.79	0.79	33
intensity_moderate	0.50	0.39	0.44	18
intensity_hard	0.60	0.60	0.60	5
intensity_very_hard	1.00	1.00	1.00	3
micro avg	0.71	0.66	0.68	59
macro avg	0.72	0.69	0.71	59
weighted avg	0.69	0.66	0.68	59
samples avg	0.66	0.66	0.66	59

2.6 Tune hyperparameteres

```
[22]: ##### Quick analysis
accuracy_scores = []
f1_scores = []
#precision_scores = []

n_estimator_numbers = range(1,300)

for i in n_estimator_numbers:
    rfc_t = RandomForestClassifier(n_estimators=i, random_state=0)
    rfc_t.fit(train_x, train_y)

    predictions = rfc_t.predict(valid_x)

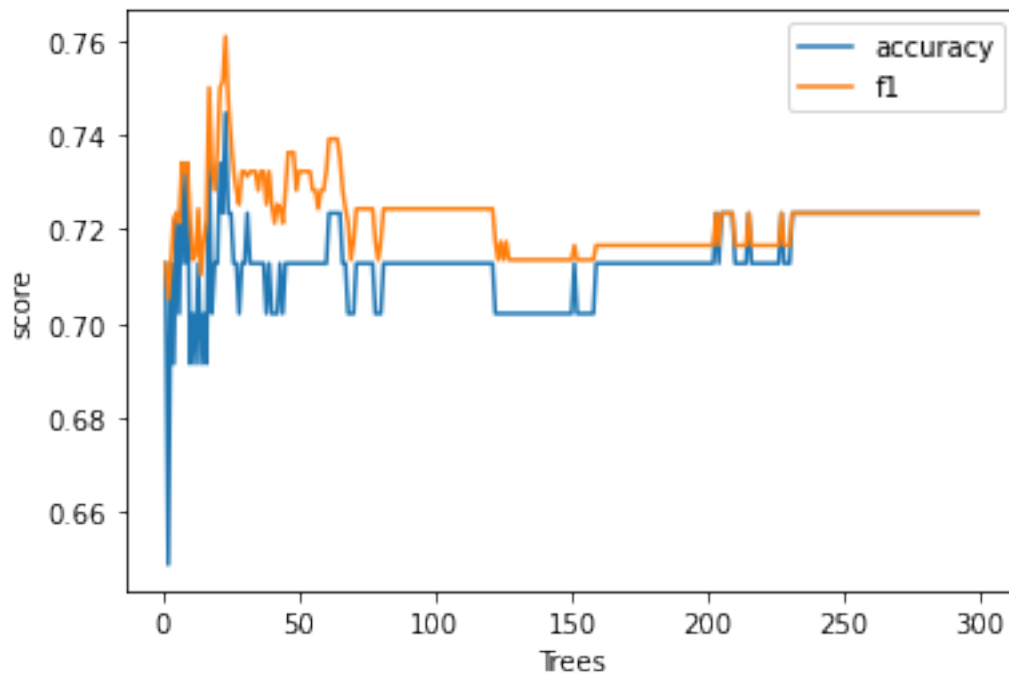
    accuracy_scores.append(accuracy_score(valid_y, predictions, normalize=True))
    f1_scores.append(f1_score(valid_y, predictions, average='micro' ))
    #precision_scores.append(precision_score(valid_y, predictions,
    →average='micro'))
```

```
[23]: plt.plot(n_estimator_numbers, accuracy_scores, label='accuracy')
plt.plot(n_estimator_numbers, f1_scores, label='f1')

plt.xlabel('Trees')
plt.ylabel('score')
#plt.plot(n_estimator_numbers, precision_scores, label='precision')

plt.legend()
```

```
[23]: <matplotlib.legend.Legend at 0x7ff0a182ac18>
```



```
[24]: np_accuracy_scores = np.array(accuracy_scores)
      np_f1_scores = np.array(f1_scores)

      best_accuracy_index = np.argmax(np_accuracy_scores)
      best_f1_index = np.argmax(np_f1_scores)

      print('accuracy: ', n_estimator_numbers[best_accuracy_index])
      print('f1: ', n_estimator_numbers[best_f1_index])
```

```
accuracy: 23
f1: 23
```

```
[ ]:
```