# $linear\_regression\_XYZ\_MET$ -sitting

January 12, 2021

```
[1]: from helpers import pandas helper as pdh
     from helpers import math_helper as mth
     from sensors.activpal import *
     from utils import read_functions
     from scipy.stats import linregress
     from sklearn.linear_model import LinearRegression
     from sklearn.model_selection import train_test_split
     from sklearn.metrics import mean_squared_error
     from sklearn.feature_selection import RFE
     import seaborn as sns
     import math
     import numpy as np
     import matplotlib.pyplot as plt
     import pandas as pd
     import datetime
     activpal = Activpal()
     activity_focus = 'zitten'
     respondents = ['BMR002', 'BMR011', 'BMR012', 'BMR014', 'BMR018', 'BMR030', |
      \hookrightarrow 'BMR031', 'BMR032', 'BMR033', 'BMR034', 'BMR036', 'BMR040', 'BMR041', \sqcup
      _{\hookrightarrow} 'BMR042', 'BMR043', 'BMR044', 'BMR052', 'BMR053', 'BMR055', 'BMR058', _{\sqcup}
      \hookrightarrow 'BMR064', 'BMR098']
     test_respondents = ['BMR004', 'BMR008', 'BMR097']
```

# 1 Defining functions

1.0.1 Method to retrieve DataFrame containing all information for regression

```
[2]: def get_regression_df(respondent, activity):
    start, stop = get_timestamps(respondent, activity)

# read in all dataframes necessary
```

```
respondents_df = pdh.read_csv_respondents()
   res_number = get_respondent_number(respondent)
    vyntus_df, min_index, max_index = get_vyntus_df(respondent,__
 →respondents_df['gewicht'][res_number], start, stop)
   raw_df = get_raw_df(respondent, min_index, max_index)
    # add met and mag acc to new dataframe
   new_df = pd.DataFrame(index=raw_df.index)
   new_df['mean_met'] = vyntus_df['met']
   new_df['sum_mag_acc'] = raw_df['sum_mag_acc']
   # add features to new dataframe
   new_df['length_cm'] = respondents_df['lengte'][res_number]
   new_df['weight_kg'] = respondents_df['gewicht'][res_number]
   new_df['mean_speed'] = raw_df['mean_speed']
   new_df['bmi'] = mth.calculate_bmi(new_df['weight_kg'], new_df['length_cm'])
   new_df['gender'] = int(respondents_df['geslacht'][res_number].
→replace('vrouw',str(0)).replace('man', str(1)))
   new_df['age_category'] = respondents_df['leeftijdscategorie'][res_number]
    convert_age_to_number(new_df, "age_category")
   new_df['is_sporter'] = respondents_df['sporter'][res_number]
   new_df['meets_activity_guidelines'] = int(respondents_df['voldoet aan_u
 →beweegrichtlijn 2nee17'][res_number].replace('ja', '1').replace('nee', '0'))
   new df['does muscle bone exercises'] = int(respondents df['voldoet aan,
→richtlijn bot en spierversterkende activiteiten'] [res number].replace('ja', |
→'1').replace('nee', '0'))
   new_df['meets_balance_guidelines'] = int(respondents_df['voldoet_aan_u
→richtlijn balansoefeningen'][res_number].replace('ja', '1').replace('nee', □
→'0'))
   return new df
def get_regression_dfs(respondents, activity):
   all_df = pd.DataFrame(index=pd.to_datetime([]))
   for cor in respondents:
        df = get_regression_df(cor, activity)
        all_df = pd.concat([all_df, df])
   all_df.sort_index(inplace=True)
   return all_df
```

#### 1.0.2 Helper method, returning start and stop timestamps for an activity

```
[3]: def get_timestamps(respondent, activity):
    activities_df = read_functions.read_activities(respondent)
    start = activities_df.loc[activity].start
    stop = activities_df.loc[activity].stop

return (start, stop)
```

## 1.0.3 Helper method, returning the number of respondent code

```
[4]: def get_respondent_number(respondent):
    if ('BMRO' in respondent):
        return int(respondent.replace('BMRO', ''))

if ('BMR' in respondent):
    return int(respondent.replace('BMR', ''))
```

#### 1.0.4 Helper method, returning Vyntus (lab data)

The DataFrame contains MET and other information necessary for regression

The DataFrame is resampled to minutes by mean

## 1.0.5 Helper method, returning Activpal20

The DataFrame contains sumation of magnitude of acceleration

The DataFrame is resampled to minutes by sumation

```
[6]: def get_raw_df(respondent, start, stop):
    df = activpal.read_data(respondent, start, stop)

mask = (df.index >= start) & (df.index < stop)
    df = df.loc[mask]

df = df[['pal_accX', 'pal_accY', 'pal_accZ']].apply(mth.convert_value_to_g)
    df['sum_mag_acc'] = mth.to_mag_acceleration(df['pal_accX'], df['pal_accY'],
    df['pal_accZ'])

df['mean_speed'] = get_speed(df['pal_accX'], df['pal_accY'], df['pal_accZ'])

df = df.resample('60s').agg({'sum_mag_acc':'sum', 'mean_speed':'mean'})[:-1]

return df</pre>
```

#### 1.0.6 Helper method, from Colin

```
[7]: def convert_age_to_number(dataframe, age_category):
    age_convertion = {
        age_category: {
            "15-19": 0, "20-24": 1, "25-29": 2, "30-34": 3,
            "35-39": 4, "40-44": 5, "45-49": 6, "50-54": 7,
            "55-59": 8, "60-64": 9, "65-69": 10, "70-74": 11, "75-79": 12
        }
    }
    return dataframe.replace(age_convertion, inplace=True)
```

#### 1.0.7 Helper method, returns speed, from Adnan

#### 1.0.8 Helper method, returns best features out of all features

```
[9]: def get_best_features(df, feature_columns, n_features_to_select):
    x_train, x_valid, y_train, y_valid = train_test_split(df[feature_columns],
    →df['mean_met'], test_size=0.2, random_state=0)

estimator = LinearRegression()
    selector = RFE(estimator, n_features_to_select=n_features_to_select)
    selector = selector.fit(x_train, y_train)

return x_train.columns[selector.support_]
```

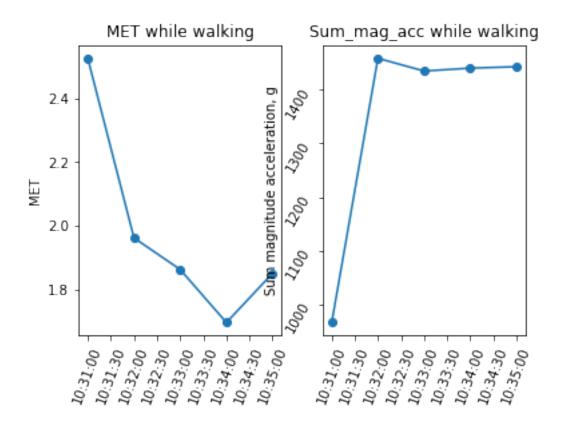
### 1.0.9 Helper methods to plot various graphs

```
[10]: def plot_met(met, title = 'MET while sitting'):
          plt.ylabel('MET')
          plt.plot(met, marker = 'o')
          plt.title(title)
          plt.xticks(rotation=70)
      def plot_mag_acc(mag_acc, title = 'Sum_mag_acc while sitting'):
          plt.ylabel('Sum magnitude acceleration, g')
          plt.plot(mag_acc, marker = 'o')
          plt.title(title)
          plt.xticks(rotation=70)
          plt.yticks(rotation=60)
      def plot_lin_reg(x, y, xlabel = 'acceleration, g', ylabel = 'MET'):
          linreg = linregress(x, y)
          fx = np.array([x.min(), x.max()])
          fy = linreg.intercept + linreg.slope * fx
          plt.xlabel(xlabel)
          plt.ylabel(ylabel)
          plt.plot(x, y, 'o')
          plt.plot(fx, fy, '-')
      def plot_heatmap(df):
          plt.figure(figsize=(15,10))
          sns.heatmap(df.corr(), annot=True, cmap=plt.cm.Reds)
          plt.show()
      def plot_pred_truth(pred_y, valid_y, title):
          plt.plot(range(len(pred_y)), pred_y, label='prediction', marker='o')
          plt.plot(range(len(valid_y)),valid_y,label='ground_truth', marker='o')
```

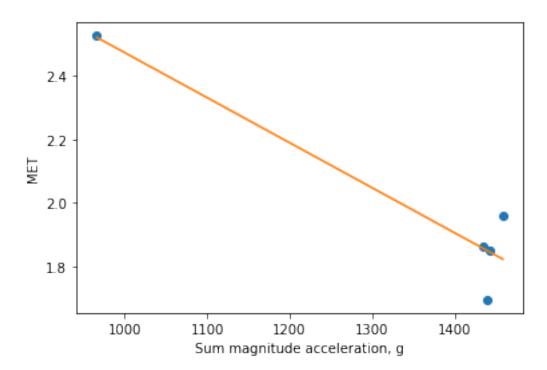
```
plt.title(title)
    plt.ylabel('met')
    plt.xlabel('Prediction Number')
    plt.legend()
def plot_ground_truth_vs_prediction(pred_y, valid_y, title='Predictions on_u
⇔validation dataset'):
    plt.figure(figsize=(20,10))
    bar_width = 0.35
    pred_index = np.arange(len(pred_y))
    y_index = np.arange(len(valid_y)) + bar_width
    plt.bar(pred_index, pred_y, bar_width, label='Prediction')
    plt.bar(y_index, valid_y, bar_width, label='Ground Truth')
    plt.xlabel('Prediction Number')
    plt.ylabel("MET")
    plt.title(title)
    plt.xticks(pred_index + 0.15, pred_index)
    plt.legend()
    plt.grid()
    plt.show()
```

# 2 Univariate linear regression

#### 2.0.1 One respondent



## r = -0.9463741861339477



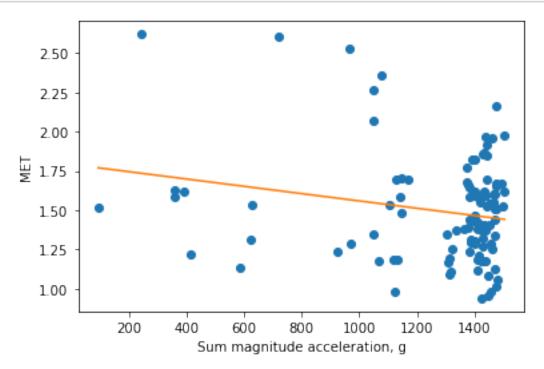
#### 2.0.2 All respondents

```
[12]: all_df = get_regression_dfs(respondents, activity_focus)

test_df = get_regression_dfs(test_respondents, activity_focus)

plot_lin_reg(all_df['sum_mag_acc'], all_df['mean_met'], xlabel='Sum magnitude

→acceleration, g')
```

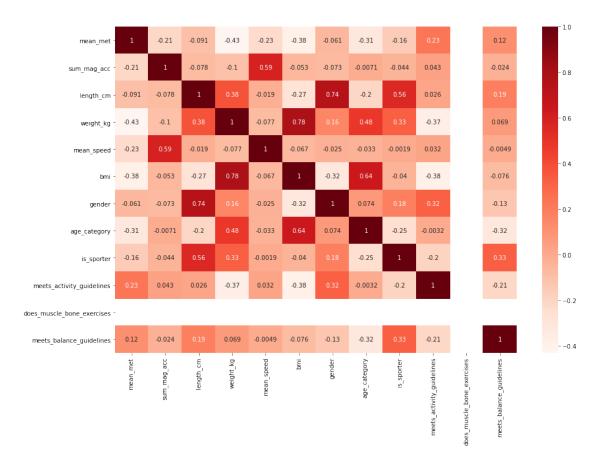


# 3 Multivariate Linear regression model training

# 3.1 Dimensionality reduction

## 3.1.1 Pearson Correlation confusion matrix

```
[13]: plot_heatmap(all_df)
```



#### 3.1.2 Drop missing values

```
[14]: all_df = all_df.dropna(how='any')
test_df = test_df.dropna(how='any')
```

#### 3.1.3 Drop variables with zero variation (unary)

```
[16]: print(all_df.var())

if 'does_muscle_bone_exercises' in all_df.columns:
    all_df = all_df.drop('does_muscle_bone_exercises', axis=1)

if 'does_muscle_bone_exercises' in test_df.columns:
    test_df = test_df.drop('does_muscle_bone_exercises', axis=1)

all_df.columns
```

```
0.113902
     mean_met
                                  93075.769009
     sum_mag_acc
     length_cm
                                     103.219609
     weight_kg
                                     167.392513
     mean speed
                                    8860.829164
                                      15.924856
     bmi
     gender
                                      0.252336
     age_category
                                      14.357080
     is sporter
                                      0.199436
     meets_activity_guidelines
                                      0.085523
     meets_balance_guidelines
                                      0.175630
     dtype: float64
[16]: Index(['mean_met', 'sum_mag_acc', 'length_cm', 'weight_kg', 'mean_speed',
             'bmi', 'gender', 'age_category', 'is_sporter',
             'meets_activity_guidelines', 'meets_balance_guidelines'],
            dtype='object')
```

### 3.1.4 Drop very low correlation variables

```
[17]: if 'gender' in all_df.columns:
    all_df = all_df.drop('gender', axis=1)

if 'gender' in test_df.columns:
    test_df = test_df.drop('gender', axis=1)

if 'length_cm' in all_df.columns:
    all_df = all_df.drop('length_cm', axis=1)

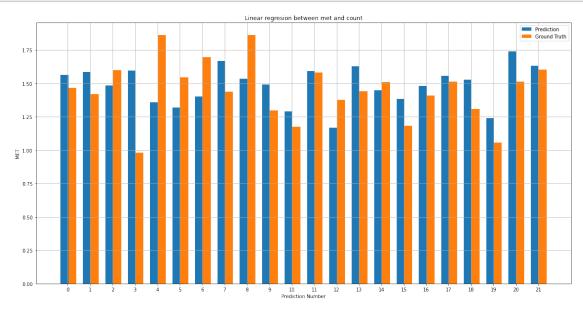
if 'length_cm' in test_df.columns:
    test_df = test_df.drop('length_cm', axis=1)
```

#### 3.1.5 Define remaining features

#### 3.1.6 Recursive Feature Elimination (RFE)

```
[25]: best_features = get_best_features(all_df, remaining_features, 9)
print(best_features)
```

## 3.2 Multivariate regression best features



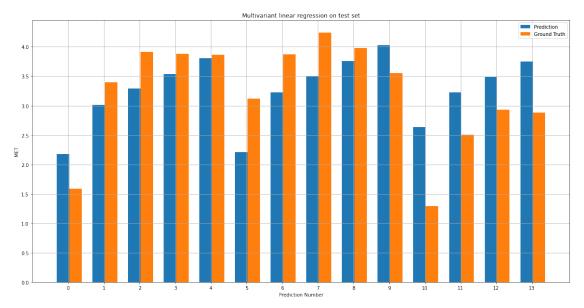
mean squared error = 0.05872498026548571

```
r squared = 0.3594060523477939
```

# 3.3 Hyperparameter tuning

[]:

# 3.4 Multivariate Linear Regression model on Test set



```
mean squared error = 0.4616545227293855
r squared = 0.39186986685653225
```

[]: