

# Capturing real-world walking speed in a lab-based protocol: validation in younger and older healthy adults

As part of the project:

Validation of a tool for digital assessment of mobility in the real world

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## Background

- Wearable sensors offer the potential to quantify real-world walking and subsequent adoption in healthcare to better diagnose, classify and predict pathologies related to a loss of mobility. However, validation of the accuracy and precision of the sensor's calculated outputs, referred to as digital mobility outcomes (DMOs), including walking speed, is crucial.
- A gold-standard for assessment of gait outputs, an optoelectronic stereophotogrammetric (SP) system, can be used to validate wearable sensors and calculated outputs, however this system is restricted to a laboratory space, which may limit the ability to capture the variation of the sensor's calculated outputs expected in real-world walking. In response, a task protocol that includes variation in walking conditions and complexity has been designed.

This study aims to:

- Assess if the variation in walking speed calculated by the SP system for two cohorts, younger healthy adults (YHA, 18 - 65 years old) and older healthy adults (OHA, > 65 years old), is representative of that reported for real-world.
- Assess if the walking speed data for the YHA and OHA cohorts is normally distributed to allow for parametric analysis of walking speed between the two cohorts.
- Assess the difference in walking speed between the YHA and OHA cohorts in easier and more complex walking tasks with normally distributed data.

This study:

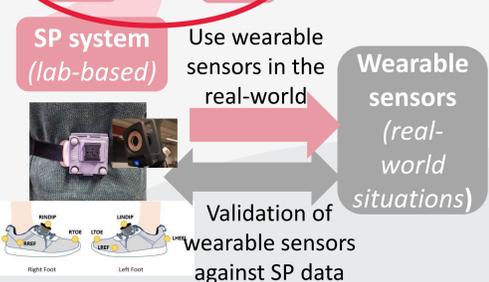
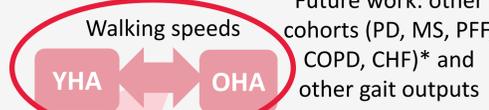
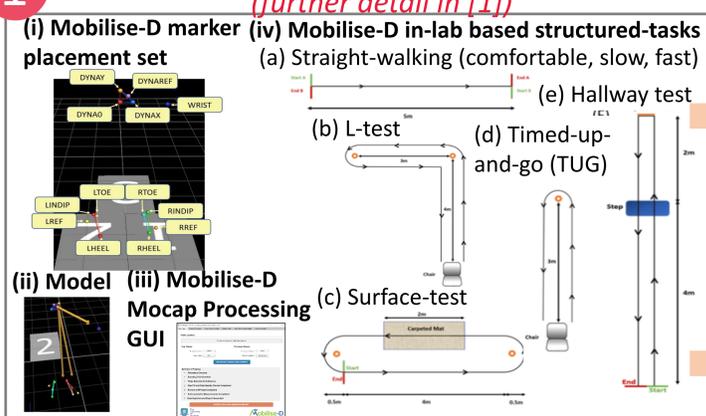


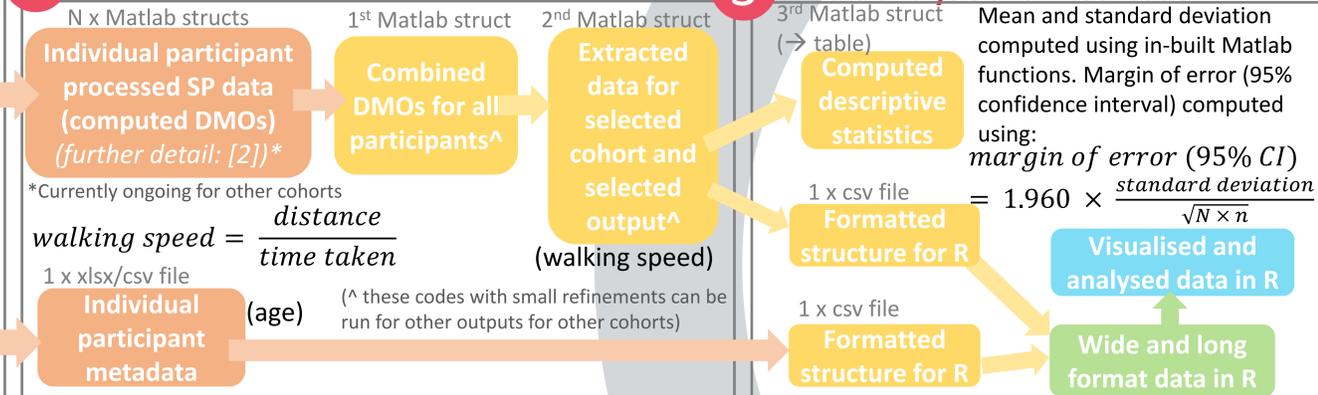
Figure 1. From lab to real-world assessment of mobility: the need for validation of wearable sensors

## Methods

1 Prior to this study: Mobilise-D in-lab protocol (further detail in [1])



2 During this study, a data analysis pipeline was developed and carried out:



For each of YHA and OHA cohorts: N = 20, n=1 for all activities except for each of the three conditions of straight-walking (SWalk) where N=20, n=2 (N = number of participants, n = number of repeats); with further exceptions resulting in: N x n = SWalk Comf OHA (28), LTest OHA (19), SWalk Slow OHA (38), SWalk Fast OHA (21) and SWalk Fast YHA (19)

Software used: Matlab 2020b (MathWorks Inc.) and R Studio (version 1.4.1717)

## Results

Aim 1 Table 1. Mean, standard deviation and margin of error (95% confidence interval) of the calculated walking speed for all tasks

|              | Mean (m/s) |      | SD (m/s) |      | Margin of error (95% CI) (m/s) |      |
|--------------|------------|------|----------|------|--------------------------------|------|
|              | YHA        | OHA  | YHA      | OHA  | YHA                            | OHA  |
| SWalk_Comf   | 1.24       | 1.06 | 0.15     | 0.15 | 0.07                           | 0.07 |
| L-Test       | 1.05       | 0.90 | 0.13     | 0.16 | 0.06                           | 0.07 |
| Surface-test | 1.08       | 0.99 | 0.11     | 0.15 | 0.05                           | 0.07 |
| Hallway test | 1.04       | 0.92 | 0.11     | 0.17 | 0.05                           | 0.08 |
| SWalk_Slow   | 0.80       | 0.69 | 0.18     | 0.13 | 0.08                           | 0.06 |
| SWalk_Fast   | 1.61       | 1.30 | 0.17     | 0.16 | 0.07                           | 0.07 |
| TUG          | 1.32       | 1.05 | 0.17     | 0.26 | 0.08                           | 0.11 |

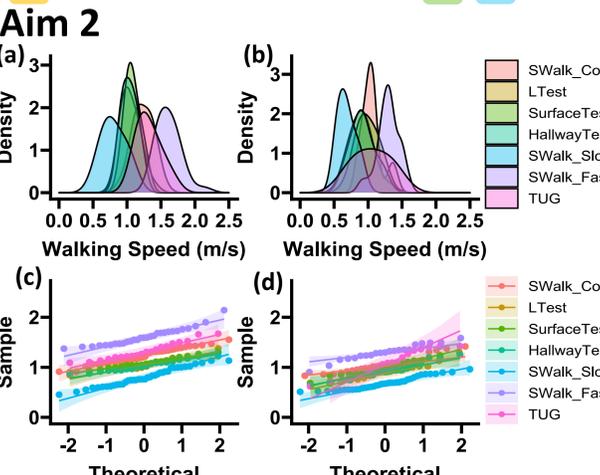


Figure 3. Density plots (a,b) and Q-Q plots (c,d) for YHA (a,c) and OHA (b,d) cohorts

Table 2. p-values for Shapiro-Wilk test and F-test

|              | Shapiro-Wilk (normality test) |       | F-test (variance test) |
|--------------|-------------------------------|-------|------------------------|
|              | YHA                           | OHA   |                        |
| SWalk_Comf   | 0.51                          | *0.01 | 0.96                   |
| L-Test       | 0.15                          | 0.52  | 0.38                   |
| Surface-test | 0.72                          | 0.50  | 0.14                   |
| Hallway test | 0.36                          | 0.62  | 0.06                   |
| SWalk_Slow   | 0.37                          | *0.02 | *0.04                  |
| SWalk_Fast   | *0.04                         | 0.69  | 0.77                   |
| TUG          | 0.19                          | 0.23  | 0.09                   |

\* Indicates statistical significance of p < 0.05 which Shapiro-Wilk on its own would indicate as non-parametric data

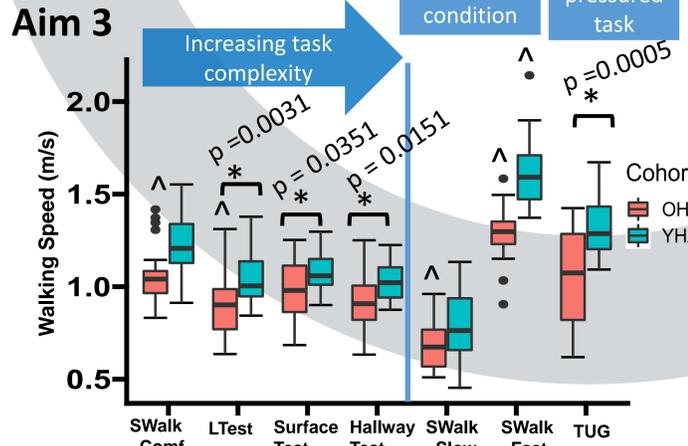


Figure 4. Boxplots for YHA and OHA walking speeds for all tasks \* Unpaired, two-sided student's t-test, 95% CI, unequal variances computed for normally distributed data. Further analyses required for non-parametric data (SWalk: Comf OHA, SWalk\_Slow OHA and SWalk\_Fast YHA)

^ Aforementioned exceptions to sample size (N x n) indicated = 28, 19, 38, 21, 29

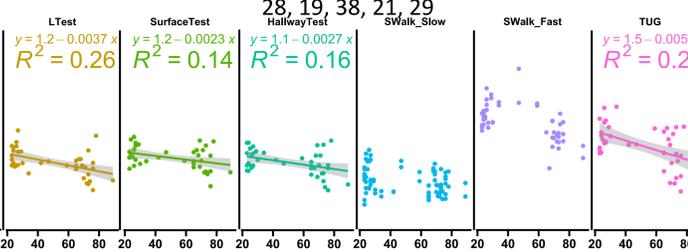


Figure 5. Scatter plots. Tasks with normally distributed data for both YHA and OHA, fitted with linear model showing equation and R-squared value R-squared value of 16 - 27% contribution of age to walking speed for tasks with normally distributed data, few data points exist for middle range of ages (Figure 5).

## Conclusions

Upon initial analysis for the YHA and OHA cohorts:

- Aim 1:** Variation in walking speed of 0.69 - 1.61 m/s (Table 1) consistent with real-world walking speed range reported in the literature ([2]).
- Aim 2:** Walking speeds for most tasks are normally distributed (Figure 3) except those marked in Table 2, Shapiro-Wilk test.
- Aim 3:** Significant differences (p < 0.05) in walking speeds between YHA and OHA cohorts were found for all normally distributed tasks (Figure 3).

## Future work

In order to determine the ages of participants required for validating the wearable sensors, further analysis is required for:

- In-lab tasks with non-parametric data (Figure 3, 4, Table 1)
- Middle-range age groups and confounding variables such as height and mass to produce a model with stronger fit to the data points (Figure 5)
- Time-varying walking speeds across the full gait cycle
- Different disease cohorts

## Acknowledgements

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## References

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