

PHYSICAL ACTIVITY, HEALTH AND EXERCISE



## Evaluation of wrist and hip sedentary behaviour and moderate-to-vigorous physical activity raw acceleration cutpoints in older adults

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### ABSTRACT

Wrist-based accelerometers are now common in assessing physical activity (PA) and sedentary behaviour (SB) in population-based studies, but there is a scarcity of raw acceleration cutpoints in older adults. The study aimed to determine and evaluate wrist-based GENEActiv (GA) and hip-based ActiGraph GT3X+ (AG) raw acceleration cutpoints for SB and moderate-to-vigorous PA (MVPA) in older adults  $\geq 60$  years of age. A laboratory-based calibration analyses of 34 healthy older adults involved receiver operator characteristic (ROC) curves to determine raw acceleration cutpoints for SB and MVPA. ROC analysis revealed an area under the curve (AUC) of 0.88 for GA SB and MVPA, and 0.90 for AG SB and 0.94 for AG MVPA. Sensitivity optimised SB and specificity optimised MVPA GA cutpoints of 57 mg and 104 mg, and AG cutpoints of 15 mg and 69 mg were also generated, respectively. Cross-validation analysis revealed moderate agreement for GA and AG SB cutpoints, and fair to substantial agreement for GA and AG MVPA cutpoints, respectively. The resultant cutpoints can classify older adults as engaging in SB or not engaging in MVPA but the sensitivity optimised SB cutpoints should be interpreted with a degree of caution due to their modest cross-validation results.

### ARTICLE HISTORY

Accepted 27 November 2018

### KEYWORDS

Accelerometry; activity monitors; calibration; cross-validation; ageing

### Introduction

There is strong evidence supporting the benefits of reducing sedentary behaviour (SB) and increasing physical activity (PA) in older adults (Devereux-Fitzgerald, Powell, Dewhurst, & French, 2016; Greaney, Lees, Blissmer, Riebe, & Clark, 2016; Kim, Im, & Choi, 2016; Lewis, Napolitano, Buman, Williams, & Nigg, 2017). Despite this, older adults (e.g., aged 60+ years; Heo, Chun, Kim, Ryu, & Lee, 2017) remain the most sedentary and physically inactive segment of society (Chastin et al., 2017). Older adults spend approximately 80% of their awake time engaged in SB which represents eight to 12 hours/day (Chastin et al., 2017). Furthermore, compliance with recommended PA guidelines is poor in the general population and even worse in older adults (Troiano et al., 2008). Health Survey for England data indicate that only 20% of men and 17% of women aged 65 to 74 years meet recommended PA guidelines (NHS, 2015). This contrasts with 49% of men and 35% of women aged 25 to 34 years (NHS, 2015). It is widely acknowledged that increasing age is associated with decreased time spent in PA (Berkemeyer et al., 2016; Harvey, Chastin, & Skelton, 2014; Martinez-Gomez et al., 2017; Wullems, Verschueren, Degens, Morse, & Onambélé, 2016) and this trend is apparent in older adults, particularly after retirement age (Martinez-Gomez et al., 2017; Strain et al., 2016). To exemplify this, Buman et al. (2010) reported that when comparing those between 66–69 years old with those aged 80 years and older,

engagement in MVPA decreased from 16.2 min-day<sup>-1</sup> to 10.7 min-day<sup>-1</sup>.

As described in the behavioural epidemiological framework (Sallis, Owen, & Fotheringham, 2000), accurate measurements of SB and PA are needed to detect potential correlates, identify relationships between such behaviours and associated health outcomes, and evaluate the efficacy of intervention strategies (Lewis et al., 2017). Accelerometry is now commonly adopted for monitoring older adults' SB and PA levels (Mañas et al., 2017; Oguma et al., 2017; Wullems, Verschueren, Degens, Morse, & Onambélé, 2017; Zhu et al., 2017), as it allows for valid, reliable, and accurate assessments of activity intensity, frequency, and duration (Mathie, Coster, Lovell, & Celler, 2004; Prince et al., 2008). Moreover, accelerometers are particularly appropriate for assessing PA in older adults as these devices require no input from the participant over the data collection period, and superior wearer compliance has been demonstrated when compared to younger age groups. For example, in the UK Biobank study of 103,578 adults and older adults, average wear time compliance of a wrist-worn triaxial accelerometers increased by an average of two hours 18 min-day<sup>-1</sup> (1.6%) for each decade from 40 to 79 years old (Doherty et al., 2017). Accelerometry also eliminates self-report questionnaire bias related to subjective recall of past events which is an ability that can decline with ageing (Barnett, van Den Hoek, Barnett, & Cerin, 2016).

The hip has been the conventional attachment site for accelerometers because of its proximity to the centre of mass (Troiano, McClain, Brychta, & Chen, 2014; van Hees et al., 2011). A limitation of hip-worn accelerometers in studies involving older adults (Copeland & Eslinger, 2009; Taylor et al., 2014) is that most commonly used cutpoints applied to data to classify PA intensity have been calibrated for younger adults (Falck, Davis, & Liu-Ambrose, 2016). As the energy expenditure associated with a given metabolic equivalent (MET) activity intensity threshold is lower in older adults compared to younger adults (Hall, Howe, Rana, Martin, & Morey, 2013), using accelerometer cutpoints developed in young adults would likely result in an underestimation of time spent in moderate-to-vigorous intensity PA (MVPA) (Barnett et al., 2016).

Recent accelerometer studies have suggested that the wrist may be a preferable attachment site as it can more accurately capture the arm motions of non-ambulatory based activities such as household chores (Evenson et al., 2015; Landry, Falck, Beets, & Liu-Ambrose, 2015), and is less influenced by atypical gait patterns and walking speed variability, which are both commonly observed in older adults (Ko, Jerome, Simonsick, Studenski, & Ferrucci, 2018). Wrist-worn accelerometers have demonstrated excellent validity against energy expenditure as the criterion measure, and in comparison to hip-worn monitors (Eslinger et al., 2011). Furthermore, wrist- and hip-worn accelerometers have demonstrated comparable free-living MVPA classification accuracy (Hargens, Deyarmin, Snyder, Mihalik, & Sharpe, 2017). Superior wearer compliance has also been reported for accelerometers worn on the wrist versus the hip in large population-based studies including the National Health and Nutrition Examination Survey (NHANES), Dallas Heart Study, and the UK Biobank study adopting wrist-worn accelerometer protocols (Doherty et al., 2017; Lakoski & Kozlitina, 2014; Troiano et al., 2014). Specifically, wrist-worn data from the 2011 to 2012 cycle of NHANES showed that 70–80% of participants provided at least six days of data with at least 18 hours of wear. This contrasts with 40–70% of participants who provided at least six days of hip-worn accelerometer data with at least 10 hours of wear in the 2003 to 2004 cycle of NHANES (Troiano et al., 2014).

A further development in accelerometer-based SB and PA research is the move toward raw acceleration signal processing. This advance in accelerometer-based PA monitoring, which has traditionally used accelerometer output reduced to dimensionless activity “counts” per user-specified period of time or epoch (Fairclough et al., 2016) is likely to provide greater methodological transparency in post-data collection analytical processes and improve comparability of data between different accelerometer models (Hildebrand, Van Hees, Hansen, & Ekelund, 2014). Devices such as the GENEActiv (Activinsights, Cambs, United Kingdom) and ActiGraph GT3X+ and GT9X (ActiGraph, Pensacola, FL) are capable of collecting and recording raw unfiltered accelerations, which can then be subject to researcher-driven data processing procedures (Welk, McClain, & Ainsworth, 2012). Interunit reliability is acceptable for both brands (Eslinger et al., 2011; Santos-Lozano et al., 2012). Although time spent in SB and light, moderate and/or vigorous intensity PA can be quantified from raw acceleration data (Matthew, 2005),

currently, no raw acceleration cutpoints for SB and MVPA exist for older adults. Consequently, the aim of this study was to determine laboratory-based wrist-worn GENEActiv and hip-worn ActiGraph GT3X+ raw acceleration cutpoints for SB and MVPA in older adults.

## Methods

### Study population

A homogenous purposive sample of 34 community dwelling white, British older adults (10 male), aged 60–86 years (mean age = 69.6, SD = 8.0 years) were recruited through leaflets distributed at local fitness centres/gyms and community centres, and through word of mouth referrals. A sample size of 30 participants was targeted so as to be comparable with recent calibration studies in older adults (Landry et al., 2015; Wullems et al., 2017). Individuals interested in participating in the study were pre-screened for inclusion criteria which set out that participants must be (1) ≥60 years of age, (2) be physically cleared for exercise using the modified Physical Activity Readiness Questionnaire (Modified PAR-Q; Cardinal & Cardinal, 2000; Cardinal, Esters, & Cardinal, 1996), (3) have the ability to walk briskly on a treadmill without assistance, and (4) not be taking any medications that would influence energy expenditure or their ability to perform ambulatory activity. Participants were excluded if they had a medical condition precluding them from exercise, were unable to wear a portable indirect calorimeter during testing, or had limited mobility such that they could not walk on a treadmill independently. Prior to the commencement of the study, institutional ethical approval was received (SPA-REC-2016–343) and all participants provided written informed consent prior to their inclusion.

### Anthropometrics

Participants' body mass (Seca mechanical scales, Birmingham, UK) and standing height (Holtain Limited stadiometer, Crymch, UK) were measured in light clothing without shoes. Resting blood pressure was measured upon arrival and immediately prior to commencement of the accelerometer calibration protocol using a Boso Medicus Prestige blood pressure monitor (Boso Bosch + Sohn, Germany).

### Study protocol

Participants completed two separate laboratory data collection visits (separated by one week) at the university site. Visit 1 was an initial familiarisation of the protocol structure, equipment, and laboratory. Participants also provided written informed consent and completed anthropometric measures, and a six-minute treadmill walk test (6MWT; ATS Committee on Proficiency Standards for Clinical Pulmonary Function Laboratories, 2002) to establish maximal walking speeds and to familiarise participants with treadmill walking at different speeds. Participants completed a laboratory-based protocol consisting of 16 activities (see Table 1 for protocol description). Older adults typically engage in SB

**Table 1.** Description of the sixteen structured activities.

Activity	Description of activity
Lying down	Lying in supine position awake, with arms at the side, avoiding bodily movement.
Reclined	Reclined in supine position reading newspaper turning the page every 20 seconds.
Reclined	Reclined in supine position using a mobile phone.
Sitting	Sitting in a chair with hands on knees, avoiding bodily movement.
Sitting	Sitting in a chair reading newspaper turning the page every 20 seconds.
Sitting	Sitting in a chair using a mobile phone.
Standing	Standing upright, with arms at the side, avoiding bodily movement.
Standing	Standing upright reading newspaper turning the page every 20 seconds.
Standing	Standing upright using a mobile phone.
Washing up	Standing upright at a sink washing (20 seconds) and drying (20 seconds) five items.
Shopping	Standing upright putting five items in an open cupboard with one hand, and then removing the same five items with the opposite hand.
Mopping	Standing upright dry mopping the floor (marked out area).
Stepping	Stepping up and down at 69 beats per minute on a 23 cm high step.
Walking	Walking on a treadmill at 65% maximal speed individually calibrated from the 6MWT.
Walking	Walking on a treadmill at 75% maximal speed individually calibrated from the 6MWT.
Walking	Walking on a treadmill at 85% maximal speed individually calibrated from the 6MWT.

6MWT, six-minute walk test.

and light intensity PA (LPA; Ku, Steptoe, Liao, Sun, & Chen, 2018) consisting of activities such as carrying light objects, walking slowly, and household chores (Public Health England, 2017). Thus, to represent these typical activity patterns and to make the distinction between sitting and standing, a range of sedentary, stationary, and LPA activities were included in the study protocol. MVPA activities reflected typical lifestyle behaviours of walking and stepping (e.g., stair climbing). All activities were performed in a standardised order as described in Table 1. Participants were provided with standardised instructions from a predetermined script delivered by the same researcher prior to beginning each activity. The activity protocol only was repeated at visit 2. Visit 1 lasted ~80 minutes whilst visit 2 lasted ~60 minutes as participants did not need to repeat the familiarization components and 6MWT.

Sedentary, stationary and LPA activities were performed for 3-minutes each, whilst the stepping and walking activities were performed for two minutes. Sedentary, stationary, LPA, and stepping activities were separated by a 1-minute transition period, whilst the treadmill walking activities were separated by 5-second transitions. To ensure consistency across measurement sessions, the first author was trained and led on all aspects of the measurement protocol. The start and end of each activity was timed by the first author with a stopwatch (Fastime, Leicestershire, UK). All instruments were synchronised to the same clock to ensure that criterion measurements for a given period of activity were matched with time-stamped accelerometer data for precisely the same duration of the activity. This allowed for appropriate data comparisons to be made across all recording devices.

### Accelerometers

Participants wore GENEActiv (GA) (ActivInsights Ltd., Kimbolton, Cambridgeshire, UK) and ActiGraph (AG) GT3X+ (ActiGraph, Pensacola, FL) triaxial accelerometers on the non-dominant wrist and left hip, respectively. Both monitors were set to collect raw triaxial accelerations at 60 Hz. Participants also wore an activPAL (PAL Technologies Ltd., Glasgow, UK) accelerometer on the left anterior thigh as the criterion measure for SB (Chastin, Culhane, & Dall, 2014; Varela Mato, Yates,

Stensel, Biddle, & Clemes, 2017). The activPAL monitor has been shown to accurately measure postural allocation when worn on the upper thigh (Kozey-Keadle, Libertine, Lyden, Staudenmayer, & Freedson, 2011). It is valid for step counts, time spent sitting, lying, standing, and walking, and is currently regarded as a gold standard for the objective measurement of sedentary/sitting time and patterning of SB (Chastin et al., 2018; Koster et al., 2016; Kozey-Keadle et al., 2011; Sellers, Dall, Grant, & Stansfield, 2016). The activPAL uses proprietary algorithms to classify activity into periods spent sitting, standing, and stepping. activPAL data were collected at a sampling frequency of 20 Hz. Participants wore the same monitors (matched by serial number) for both visits. A total of six different GA and AG, and four different activPAL monitors were used throughout the study. All devices were used and calibrated as per manufacturers' instructions and initialised approximately 10 minutes before the start of each data collection session.

Immediately after testing, the activity monitors were removed and the data downloaded to a single, secured computer. The GA data were downloaded using GENEActiv PC software version 2.9 (Activinsights, Cambs, United Kingdom) and saved in raw format as .bin files, whilst the AG data were downloaded using ActiLife version 3.13.3 (ActiGraph, Pensacola, FL), and saved in raw format as .gt3x files and converted to time-stamped .csv files to facilitate raw data processing. activPAL data were downloaded using activPAL3 software version 7.2.32 (PAL Technologies Ltd., Glasgow, UK), saved as .datx files and converted to .csv "Event" files for processing. Signal processing of raw GA .bin files and raw AG .csv files was completed using R-package GGIR version 1.5 (<https://cran.r-project.org/web/packages/GGIR/>) (van Hees et al., 2013b). GGIR facilitates data cleaning and the extraction of user-defined acceleration levels, which can then be set to reflect the intensity thresholds as derived in this study (Rowlands, Yates, Davies, Khunti, & Edwardson, 2016). Concurrent with previous studies (Fairclough et al., 2016; Hildebrand et al., 2014; Menai et al., 2017; Rowlands et al., 2016) and to help promote comparability between studies, the Euclidean Norm Minus One (ENMO) (van Hees et al., 2013b) was adopted to quantify acceleration relative to gravity ( $1 \text{ mg} = 0.00981 \text{ ms}^{-2}$ ), after which negative values were rounded to zero. The ENMO metric has proven competency and increasing use in the field (Vähä-Ypyä et al., 2015).



ENMO has been shown to be comparable to other raw acceleration signal metrics such as MAD and HFEN in predicting PA energy expenditure (Bakrania et al., 2016; van Hees et al., 2013b) and consequently, has been adopted in large scale studies for the analysis of raw acceleration data (Doherty et al., 2017; Menai et al., 2017). Furthermore, although there are differences in the design of the chosen GA and AG devices that may affect individual axis output, these differences affect the resultant ENMO values in a consistent manner with GA output approximately 10% higher than AG output (Rowlands et al., 2018).

Raw data were further reduced by averaging the ENMO values over 1-second epochs. All resulting values are expressed in milli ( $10^{-3}$ ) gravity-based acceleration units (mg), where  $1\text{ g} = 9.81\text{ m/s}^2$ . Although the ENMO metric can be sensitive to poor calibration (van Hees et al., 2013b), GGIR autocalibrates the raw triaxial accelerometer signal in order to reduce such calibration error (van Hees et al., 2014). Where insufficient non-movement periods were available for auto-calibration (due to the relatively short duration of the protocol), back-up calibration coefficients derived from older adult males' and females' free-living data collected with the same accelerometer units were used (Hildebrand, Hansen, van Hees, & Ekelund, 2017; Rowlands et al., 2018). Subsequently, no files were removed from the analytical sample as a result of calibration error. The activPAL "Event" files provided the exact time in seconds that posture changes occurred and each second was classified as sedentary, standing, or stepping. These files were then expanded using an Excel formula to obtain second-by-second data, with each second subsequently classified as either sedentary or not sedentary. These second-by-second activPAL files were synchronised with the 1-second ENMO values from GA and AG. To exclude any transitional movements, the middle 2-minutes of data from each 3-minute activity were extracted and subsequently utilised for analysis. The full 2-minutes for each of the stepping and walking activities were used. This included the 5-second transition period between each walking speed.

### Energy expenditure

As the criterion measure for MVPA, oxygen consumption ( $\text{VO}_2$ ;  $\text{ml}\cdot\text{kg}\cdot\text{min}^{-1}$ ) was measured with a portable indirect calorimetry system (MetaMax 3B-R2, CORTEX Biophysik GmbH, Leipzig, Germany). The Metamax interface and breathing mask (7600 Series V2, Hans Rudolph, Kansas) were set up and fitted as per the manufacturer instructions.  $\text{VO}_2$  was measured using breath-by-breath mode and in order to match time periods across devices, data were stored in second-by-second intervals. These measurements were used to determine energy expenditure (EE), which was then used to classify activity intensity in METs. Resting energy expenditure was measured at visit 2 during the first three lying activities of the protocol. The observed mean resting energy expenditure was  $2.89\text{ ml}\cdot\text{kg}\cdot\text{min}^{-1}$ , which was used to define 1 MET. This value is comparable with previous calibration studies in older adults (Barnett et al., 2016; Evenson et al., 2015; Sergi et al., 2010; Siervo et al., 2014), and is consistent with the expected decrease in RMR associated with ageing (Byrne, Hills, Hunter, Weinsier, & Schutz, 2005; Kwan, Woo, & Kwok, 2004). MVPA was defined as an intensity of 3 METs and above (e.g.,  $\geq 8.68\text{ ml}\cdot\text{kg}\cdot\text{min}^{-1}$ ) (Shephard, 2011).

### Data analysis

#### Cutpoint calibration

A randomly counter-balanced sample of 17 participants (12 female, five male) from visit 1, and 17 participants (12 female, five male) from visit 2 provided the calibration data ( $N = 34$ ). Descriptive statistics for all devices were calculated for each activity in the protocol. The activPAL sedentary events and 3 MET  $\text{VO}_2$  values were used as the criterion reference standards for SB and MVPA, respectively. SB and MVPA were each coded as either 0 or 1, where 1 represented the behaviour occurring and 0 represented the behaviour not occurring. Receiver Operating Characteristic (ROC) curve analyses (Jago, Zakeri, Baranowski, & Watson, 2007) were used to determine SB and MVPA cutpoints. Area under the curve (AUC) was calculated for each analysis as a measure of diagnostic accuracy. AUC values of  $\geq 0.90$  are considered excellent,  $0.80\text{--}0.89$  good,  $0.70\text{--}0.79$  fair, and  $<0.70$  poor (Metz, 1978). For each device two different pairs of cutpoints were generated by analysing combinations of sensitivity (Se) and specificity (Sp) on the ROC curves. Our aim was to determine a cutpoint that accurately captured SB and MVPA (Se) whilst limiting misclassification of SB and MVPA (Sp). Two approaches were adopted to achieve this. Firstly, ENMO values that maximised both Se and Sp (Youden index) (Perkins & Schisterman, 2006) were identified as one set of cutpoints ( $\text{SB}_{\text{Youden}}$  and  $\text{MVPA}_{\text{Youden}}$ ). The Youden index can however result in low positive predicted values (Evenson et al., 2015) and it is recommended that researchers consider the relative importance of Se and Sp (Welk, Laurson, Eisenmann, & Cureton, 2011), and implications of the selected cutpoints on the biobehavioural impact on the outcome variables (Mackintosh, Fairclough, Stratton, & Ridgers, 2012). Secondly, cutpoints were determined that emphasised Se over Sp for SB cutpoints ( $\text{SB}_{\text{Se}}$ ) to minimise the likelihood of classifying SB as PA, with Sp emphasised over Se for MVPA cutpoints ( $\text{MVPA}_{\text{Sp}}$ ) to reduce the likelihood of misclassifying light PA as MVPA. Both cutpoints reflected recommendations that the lower Se or Sp values should be  $\geq 60\%$  (Lugade, Fortune, Morrow, & Kaufman, 2014). This prioritization approach minimises the risk of individuals being misclassified in the target behaviour and is common in accelerometer calibration (Landry et al., 2015; Mackintosh et al., 2012; Nero, Wallén, Franzén, Ståhle, & Hagströmer, 2015) and fitness standards research (Welk, Laurson, et al., 2011). There was no *a priori* preference for selecting either Youden or Se/Sp-based cutpoints, but instead we wanted to examine the effect of both approaches and make an informed decision as to which may be most suitable once the data were analysed.

#### Cross-validation

To be consistent with good practice guidelines suggested by Welk (2005),  $\text{SB}_{\text{Youden}}$  and  $\text{MVPA}_{\text{Youden}}$ , and  $\text{SB}_{\text{Se}}$  and  $\text{MVPA}_{\text{Sp}}$  cutpoints were cross-validated. Cross-validation analyses were performed with data from the 17 participants (12 female, five male) whose visit 1 data was not used in the calibration analysis, and the 17 participants (12 female, five male) whose visit 2 data was not used in the calibration analysis ( $N = 34$ ). Two-by-two ( $2 \times 2$ ) contingency tables were used to check classification agreement. The criterion measure and ENMO data were first categorised into sedentary/not sedentary and MVPA/not MVPA binary codes. Computed Se and Sp, Cohen's

kappa coefficients, and percentage agreement between classifications were also assessed. Statistical analyses were performed using IBM SPSS Statistics, version 24 (IBM, Armonk, NY), with the level of statistical significance set at  $p \leq 0.05$ .

## Results

### Participant characteristics

Of the 24 female and 10 male participants, 20 were classed as healthy (five male), seven were classed as overweight (three male), and five were classed as obese (two male) (NHS, 2016). Further descriptive characteristics of the sample are presented in Table 2. The mean (SD) energy expenditure and accelerometer output from GA and AG accelerometers are presented in Table 3 for each of phase of the laboratory protocol.

### ROC curve analysis

ROC curve analysis revealed an AUC for the GA of 0.88 (95% CI: 0.87–0.88;  $P < 0.001$ ) for SB (Figure 1) and 0.88 (95% CI: 0.87–0.88;  $P < 0.001$ ) for MVPA (Figure 2). For the AG the AUC for SB was 0.90 (95% CI: 0.90–0.91;  $P < 0.001$ ) (Figure 3), and 0.94 (95% CI: 0.94–0.95;  $P < 0.001$ ) for MVPA (Figure 4).

### Cutpoint generation

Table 4 displays all the generated cutpoints. The GA cutpoints which maximised both Se and Sp using the Youden Index were  $SB_{Youden}$ : 20 mg (Se = 94%, Sp = 72%) and  $MVPA_{Youden}$ : 32 mg (Se = 88%, Sp = 77%). AG cutpoints were 6 mg (Se = 85%, Sp = 82%) for  $SB_{Youden}$  and 19 mg (Se = 86%, Sp = 92%) for  $MVPA_{Youden}$ . The cutpoints optimising Se and Sp for GA were 57 mg (Se = 99%, Sp = 60%) for  $SB_{Se}$  and 104 mg (Se = 60%, Sp = 89%) for  $MVPA_{Sp}$ . Respective AG cutpoints for  $SB_{Se}$  and  $MVPA_{Sp}$  were 15 mg (Se = 98%, Sp = 60%) and 69 mg (Se = 60%, Sp = 99%).

### Cross-validation

The classification agreement, sensitivity, specificity, and kappa coefficients between calibration and cross-validation data for SB and MVPA cutpoints are shown in Table 4. GA  $SB_{Youden}$

(Se = 47%, Sp = 92%) and  $MVPA_{Youden}$  (Se = 76%, Sp = 76%) cutpoints demonstrated moderate percentage agreement (73.1–76.2%) and moderate kappa scores (0.42–0.52). AG  $SB_{Youden}$  (Se = 62%, Sp = 93%) and  $MVPA_{Youden}$  (Se = 86%, Sp = 89%) cutpoints demonstrated high percentage agreement (83.3–87.3%) and moderate to substantial kappa scores (0.59–0.75). Comparatively, lower percentage agreement and kappa scores were observed for both GA  $SB_{Se}$  and  $MVPA_{Sp}$  (67.2–68.9%,  $k = 0.36$ –0.38), and AG  $SB_{Se}$  and  $MVPA_{Sp}$  (73.2–80.4%,  $k = 0.46$ –0.60) cutpoints, respectively.

## Discussion

This is the first study to determine GA wrist- and AG hip-worn raw acceleration cutpoints for SB and MVPA in older adults. ROC curve analyses revealed that wrist GA and hip AG accelerometer raw acceleration cutpoints provide good and excellent discriminations of SB and MVPA, respectively. The  $SB_{Se}$  and  $MVPA_{Sp}$  cutpoints were comparable to those reported previously in younger adults (Hildebrand et al., 2014; Hildebrand et al., 2017; Menai et al., 2017). The Se values of 99% for GA  $SB_{Se}$  and 98% for AG  $SB_{Se}$  cutpoints ensure that almost all older adults who are sedentary have ENMO values below the established cutpoints, and therefore have a very low risk of being misclassified as being physically active. The de-emphasis on Sp (e.g., % of older adults correctly identified as not being sedentary) acknowledges the risk that a proportion (up to 40%) of older adults who are physically active may be classified as sedentary. However, SB is an identifiable risk factor affecting physical (e.g., premature mortality, chronic diseases, and all-cause dementia risk) and psychosocial (e.g., self-perceived quality of life, wellbeing, and self-efficacy) determinants of health (Edwards & Loprinzi, 2016; Falck et al., 2016; Lewis et al., 2017) independent of PA (Tremblay et al., 2017). Such misclassification is likely to be beneficial if already physically active older adults were offered further opportunities to take part in interventions to reduce SB and increase PA levels (Chastin et al., 2017). Conversely, the higher Sp values for the GA  $MVPA_{Sp}$  cutpoint (89%) and AG  $MVPA_{Sp}$  cutpoint (99%) ensures that older adults not engaged in MVPA (e.g., ENMO values below the established cutpoints) are not falsely classified as being in MVPA and are correctly identified and targeted for PA-promoting interventions (Lyons, Swartz, Lewis, Martinez, & Jennings, 2017). The de-emphasis on Se suggests that up to 40% of older adults who are not in MVPA could have ENMO values above this cutpoint (due to the lower true positive rate, relative to the true negative rate (Sp)), and therefore their behaviour could be misclassified as MVPA (Nero et al., 2015). However, it is more likely to be harmful for an older adult to be wrongly classified as active rather than asking active older adults to take part in additional MVPA. Hence, the goal of the MVPA cutpoint to identify older adults who may have increased health risks due to

Table 2. Study sample characteristics.

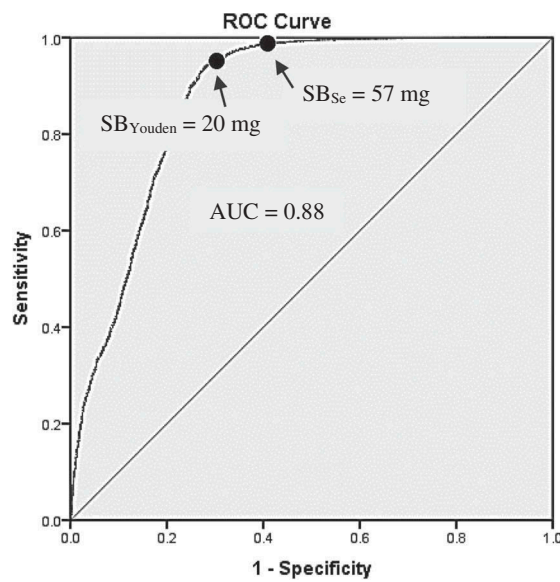
Sample Characteristics	(n = 34)
Age (years)	69.6 (8.0)
Body Mass (kg)	71.7 (17.5)
Body Height (cm)	164.3 (1)
BMI ( $\text{kg}^{-1} \cdot \text{m}^{-2}$ )	26.3 (4.7)
Blood Pressure (mmHg)	146/85 (22/11)
Maximal Walking Speed ( $\text{km} \cdot \text{h}^{-1}$ )	4.3 (1.5)

Values represent arithmetic mean (SD). SD, standard deviation; BMI, body mass index; mmHg, millimeters of mercury.

Table 3. Mean (SD) energy expenditure ( $\text{VO}_2$ ) and accelerometer output from GA and AG during each activity.

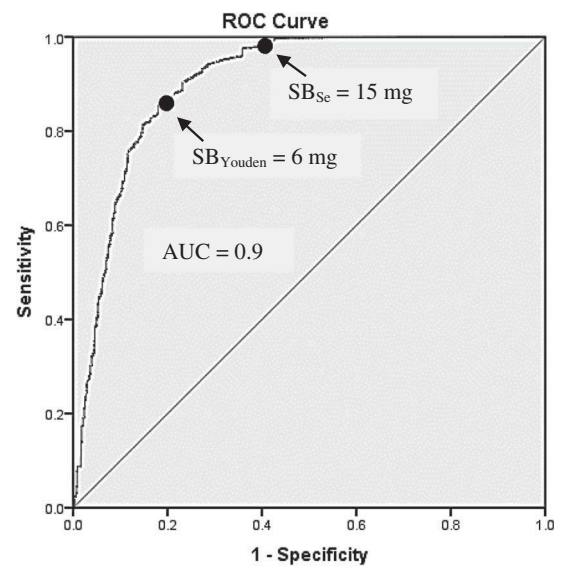
	Lying	Sitting	Standing	Household	Stepping	65% Walk	75% Walk	85% Walk
$\text{VO}_2$ ( $\text{ml} \cdot \text{kg}^{-1} \cdot \text{min}^{-1}$ )	4.08 (0.68)	4.09 (0.67)	4.43 (0.70)	7.57 (1.45)	16.23 (3.70)	11.23 (1.82)	12.65 (2.10)	13.81 (2.39)
GA wrist (mg)	6.12 (4.0)	8.14 (4.01)	12.24 (6.44)	128.05 (42.03)	99.66 (53.01)	155.89 (124.18)	185.65 (140.77)	208.90 (148.12)
AG hip (mg)	2.19 (2.92)	2.65 (3.01)	6.64 (4.69)	15.33 (7.00)	88.25 (38.10)	75.72 (32.64)	91.48 (37.58)	104.69 (47.76)

Note: 65%, 75% and 85% walk = 65%, 75% and 85% of the maximal speed obtained from the six-minute treadmill walk test (6MWT; ATS Committee on Proficiency Standards for Clinical Pulmonary Function Laboratories, 2002).



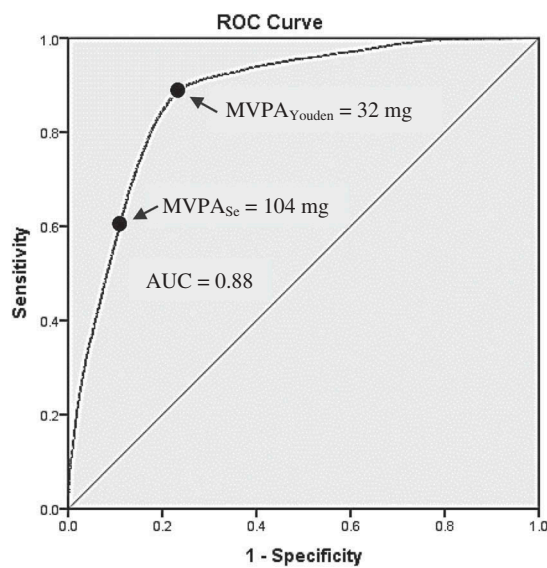
AUC, area under the curve.

Figure 1. ROC curve and associated GA  $SB_{Youden}$  and  $SB_{Se}$  cutpoints.



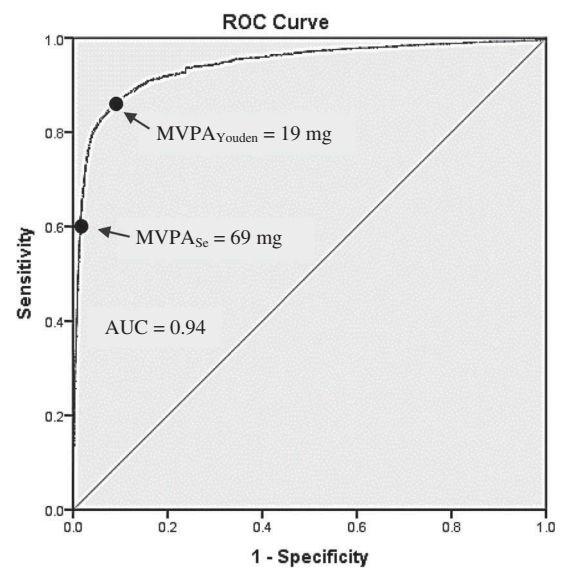
AUC, area under the curve.

Figure 3. ROC curve and associated AG  $SB_{Youden}$  and  $SB_{Se}$  cutpoints.



AUC, area under the curve.

Figure 2. ROC curve and associated GA  $MVPA_{Youden}$  and  $MVPA_{Se}$  cutpoints.



AUC, area under the curve.

Figure 4. ROC curve and associated AG  $MVPA_{Youden}$  and  $MVPA_{Se}$  cutpoints.

Table 4. Calibration cutpoints and cross-validation % agreement, kappa ( $\kappa$ ) and se and sp.

	Cutpoint (mg)	Cross-Validation % Agreement	Cross-Validation Kappa	Cross-Validation Se (%)	Cross-Validation Sp (%)
GA					
$SB_{Youden}$	$\leq 20$	73.1	0.42	47	92
$MVPA_{Youden}$	$\geq 32$	76.2	0.52	76	76
AG					
$SB_{Youden}$	$\leq 6$	83.3	0.59	62	93
$MVPA_{Youden}$	$\geq 19$	87.3	0.75	86	89
GA					
$SB_{Se}$	$\leq 57$	67.2	0.38	43	99
$MVPA_{Sp}$	$\geq 104$	68.9	0.36	81	65
AG					
$SB_{Se}$	$\leq 15$	73.2	0.46	48	97
$MVPA_{Sp}$	$\geq 69$	80.4	0.60	94	74

GA, GENEActiv; AG, Actigraph; MVPA, moderate-to-vigorous physical activity; Se, sensitivity; Sp, specificity.



being below this cutpoint (by favouring Sp over Se) appears to be justified (Nero et al., 2015; Welk, Going, Morrow, & Meredith, 2011).

The  $SB_{\text{Youden}}$  and  $MVPA_{\text{Youden}}$  cutpoints yielded greater cross-validation classification accuracy. However, they are significantly lower than existing ENMO adult SB cutpoints for GA wrist-worn (46 mg) and AG hip-worn (47 mg) accelerometers (Hildebrand et al., 2017), and MVPA cutpoints for GA wrist-worn accelerometers in adults (93 mg; Hildebrand et al., 2014) and older adults (100 mg; Menai et al., 2017), and AG hip-worn accelerometers in adults (69.1 mg; Hildebrand et al., 2014). The mean age of the participants of 69.6 years was older than participants in both the Hildebrand et al. (2014) and Hildebrand et al. (2017) studies (34.2 years). Thus, the lower SB and MVPA cutpoints were anticipated given the lower RMR associated in older adults (Byrne et al., 2005; Kwan et al., 2004). Notwithstanding this, the  $SB_{\text{Youden}}$  and  $MVPA_{\text{Youden}}$  cutpoints were substantially lower than those previously reported. If these cutpoints were applied in free-living environments it is likely that they would significantly underestimate SB and unrealistically overestimate MVPA, resulting in participants being falsely classified as being physically active when they are more likely to be sedentary.

Given that acceleration magnitudes are significantly lower for the AG GT3X+ relative to the GA (John, Sasaki, Staudenmayer, Mavilia, & Freedson, 2013; Rowlands, Frayse, & Catt, 2015; Rowlands et al., 2016), the higher wrist-worn cutpoints relative to the hip-worn cutpoints were consistent with those observed previously (Rowlands et al., 2015; Stiles, Griew, & Rowlands, 2013). Our protocol was comparable to previous calibration studies implemented in controlled settings (de Almeida Mendes, da Silva, Ramires Reichert, Martins, & Tomasi, 2017). However, laboratory calibration protocols rely on small deliberate increases in PA intensities and movement patterns within a limited period of time, compared to free-living activities over extended periods (van Hees, Golubic, Ekelund, & Brage, 2013a). Such protocols cannot fully reflect daily SB and PA patterns, and this may limit the accuracy of the SB and MVPA thresholds obtained for wrist- and hip-worn devices when they are applied in free-living environments (van Hees et al., 2013a). This may have partially explained some of the modest results observed from the cross-validation phase, which repeated the laboratory protocol with the same participants.

One of the main decisions to be made by researchers using either raw acceleration or count-based outcomes is monitor placement location (de Almeida Mendes et al., 2017). After comparing SB and PA estimates from wrist- and hip-worn monitors with energy expenditure, Rosenberger et al. (2013) concluded that SB and MVPA classification accuracy was superior for the hip-worn devices. Our cross-validation results support this due to the superior percent agreement and kappa scores for the hip-worn AG over the wrist-worn GA in classifying both SB and MVPA. However, our results also demonstrate that wrist-worn accelerometers can provide accurate estimates of SB and MVPA and the subsequent cutpoints performed reasonably well at discriminating both SB and MVPA (Troiano et al., 2014). Indeed, a recent systematic review of raw acceleration calibration studies reported no evidence of meaningful differences in the accuracy of wrist- and hip-worn accelerometers (de Almeida Mendes et al., 2017).

Considering the superior wear compliance associated with wrist-worn devices (Doherty et al., 2017), this attachment site may be the most suitable location during free-living protocols.

Several strengths and limitations should be noted when interpreting the results of this study. A main strength was the use of raw acceleration data from two commonly used devices positioned at wrist and hip wear sites. These cutpoints will be of utility to researchers using the raw data capabilities of the GA and AG accelerometers to study SB and PA in older adults. We used best practice analytical procedures to calibrate and cross-validate the cutpoints (Welk, 2005), which were specific to adults aged 60 years and over. We also directly measured resting energy expenditure to allow a sample-specific interpretation of three METs as the MVPA threshold, and used a validated separate criterion measure for SB (Kim, Barry, & Kang, 2015). There were also a number of limitations. The sample may not have been representative of the wider older adult population in respect of their fitness status and motivation to engage in PA, as we recruited a convenience sample of healthy older adults who answered advertisements and showed an interest in the study representing a broad age range. Furthermore, specific older adult populations (e.g., those with chronic diseases and impaired mobility) may require different SB and PA cutpoints (Landry et al., 2015) that reflect differences in RMR and energy cost during ambulatory PA across this age group (Miller, Strath, Swartz, & Cashin, 2010). Moreover, we incorporated activities that replicate everyday movements and tasks performed by older adults, but recognise that the laboratory setting limits the ecological validity of the resultant data (Hildebrand et al., 2017). The duration of the stepping and walking activities was relatively short and this may have limited the amount of MVPA data available for use in the analysis. Lastly, cross-validation was performed with the same participants using a repeated laboratory protocol rather than with an independent sample using a free-living or a simulated free-living protocol (Welk, 2005). It was felt that the challenges associated with having the participants wear the gas analysis system for an extended period in free-living situations were too great to warrant taking this approach. Consequently, the SB cutpoints which performed modestly in the cross-validation, should be interpreted with caution. Both sets of cutpoints should be further cross-validated in free-living environments (Welk, 2005) with independent samples over whole days using feasible criterion measures such as activPALs and wrist-worn heart rate monitors (Brage et al., 2015).

In conclusion, cutpoints varied dependent upon attachment site, with the wrist-worn GA cutpoints higher than those for the hip-worn AG. The identified GA and AG  $SB_{\text{Se}}$  and  $MVPA_{\text{Sp}}$  cutpoints can enable researchers to classify older adults as engaging in SB or not engaging in MVPA, but the  $SB_{\text{Se}}$  cutpoints should be interpreted with a degree of caution due to their modest cross-validation results. More cross-validation research in independent samples within free-living environments is needed to further test the utility of the  $SB_{\text{Se}}$  and  $MVPA_{\text{Sp}}$  cutpoints.

### Disclosure statement

No potential conflict of interest was reported by the authors.

## Funding

This work was funded by Sefton Metropolitan Borough Council and Edge Hill University and received institutional ethical approval from Edge Hill University under SPA-REC-2016-343.

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