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<https://doi.org/10.1136/bjsports-2018-099643>

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1 How well do activity monitors estimate energy expenditure? A systematic review and meta-
2 analysis.

3

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17 **Word count:**

18 4493

19 **Abstract**

20 **Objective** To determine the accuracy of wrist and arm-worn activity monitors' estimates of
21 energy expenditure (EE).

22

23 **Data sources** SportDISCUS (EBSCOHost), PubMed, Medline (Ovid), PsycINFO
24 (EBSCOHost), EMBASE (Ovid) and CINAHL (EBSCOHost).

25

26 **Design** A random effects meta-analysis was performed to evaluate the difference in EE
27 estimates between activity monitors and criterion measurements. Moderator analyses were
28 conducted to determine the benefit of additional sensors and to compare the accuracy of
29 devices used for research purposes with commercially available devices.

30

31 **Eligibility criteria** We included studies validating EE estimates from wrist or arm-worn
32 activity monitors against criterion measures (indirect calorimetry, room calorimeters and
33 doubly labelled water) in healthy adult populations.

34

35 **Results** 60 studies (104 effect sizes) were included in the meta-analysis. Devices showed
36 variable accuracy depending on activity type. Large and significant heterogeneity was
37 observed for many devices ($I^2 >75\%$). Combining heart rate or heat sensing technology with
38 accelerometry decreased the error in most activity types. Research-grade devices were
39 statistically more accurate for comparisons of total EE but less accurate than commercial
40 devices during ambulatory activity and sedentary tasks.

41

42 **Conclusions** EE estimates from wrist and arm-worn devices differ in accuracy depending on
43 activity type. Addition of physiological sensors improves estimates of EE and research-grade
44 devices are superior for total EE. These data highlight the need to improve estimates of EE
45 from wearable devices and one way this can be achieved is with the addition of heart rate to
46 accelerometry.

47

48 **Registration** PROSPERO CRD42018085016.

49

50 Keywords: *Energy expenditure, Accelerometer, Meta-analysis, Wrist, Validation.*

51

52 Device abbreviations: *Actical (ACT), Actigraph GT3X (AGT3X), Apple watch (AW), Apple*
53 *Watch series 2 (AWS2), Beurer (BA) Basis b1 (BB1), Bodymedia CORE armband (BMC),*
54 *Basis Peak (BP), Epson Pulsense (EP), ePulse Personal Fitness Assistant (EPUL), Fitbit*
55 *Blaze (FB), Fitbit Charge (FC), Fitbit Charge 2 (FC2), Fitbit Charge HR (FCHR), Fitbit*
56 *Flex (FF), Garmin Forerunner 225 (GF225), Garmin Forerunner 920XT (GF920XT),*
57 *Garmin Vivoactive (GVA), Garmin Vivofit (GVF), Garmin Vivosmart (GVS), Garmin*
58 *Vivosmart HR (GVHR), Jawbone UP (JU), Jawbone UP24 (JU24), LifeChek calorie sensor*
59 *(LC), Mio Alpha (MA), Microsoft band (MB), Misfit Shine (MS), Polar: AW360 (PA360),*
60 *Nike Fuel band (NF), Polar Loop (PL), Polar: AW200 (PO200), Samsung Gear S (SG),*
61 *SenseWear Armband (SWA), SenseWear Armband Pro 2 (SWA p2), SenseWear Armband Pro*
62 *3 (SWA p3), SenseWear Armband MINI (SWAM), TOMTOM Touch (TT), Vivago (V),*
63 *Withings Pulse (WP), Withings Pulse O2 (WPO).*

What is already known on this topic?

- Wrist or arm-worn devices incorporating multiple sensors are increasingly common and many devices provide estimates of energy expenditure. It is important to determine their validity overall and in different activity types.
- It is not clear which specific sensors or combinations of sensors provide the most accurate estimates of energy expenditure.
- It is unclear whether research-grade devices are more accurate than commercial devices.

What this study adds

- The accuracy in energy expenditure estimates from activity monitors varies between activities.
- Larger error is observed from devices employing accelerometry alone; the addition of heart rate sensing improves estimates of energy expenditure in most activities.
- In some activity types, research-grade devices are not superior to commercial devices.

65 **Introduction**

66 The prevalence of obesity has tripled in the last 40 years [1] and it has been estimated that by
67 2050, 60% of males and 50% of females may be obese [2]. Obesity is the result of a chronic
68 imbalance between energy intake (EI) and energy expenditure (EE) [3] driven by
69 physiological, psychological and environmental factors.

70 Doubly-labelled water (DLW) is considered the gold standard for the measurement of
71 free-living EE [4]; however, the considerable costs and analytical requirements limit its
72 feasibility in large cohort studies [5]. Indirect calorimetry methods represent the most
73 commonly employed criterion measure for assessment of the energy cost of an activity but
74 again are limited to structured activities usually within a laboratory [6]. Wearable activity
75 monitors are increasingly popular for the estimation of EE [7].

76 Wearable devices which use triaxial accelerometry to derive an estimate of EE have
77 been available for research purposes for some time [8]. These devices are worn on the hip,
78 thigh or lower back, as proximity to the centre of mass more accurately reflects the energy
79 cost of movement [9]; however, participant comfort and compliance is a recognised issue
80 [10] and therefore traditional wear devices have limited long-term, free-living measurement
81 capability. Use of wrist-worn activity monitors by both consumers and researchers has
82 dramatically increased [11] facilitated by improved battery longevity and miniaturization of
83 hardware required to produce interpretable data [12]. Recent consumer devices include
84 triaxial accelerometers, heat sensors and photoplethysmography heart rate sensors [13]. This
85 information can be incorporated to improve the estimation of EE relative to accelerometry
86 alone [14]. However, their accuracy compared with criterion measures is questionable [15]
87 and may vary with the type and intensity of activity [16].

88 This meta-analysis aimed to investigate the accuracy of EE estimates from current
89 wrist or arm-worn devices during different activities. Given the recent popularity wrist and
90 arm-worn activity monitors, it is critical to determine their validity for the estimation of EE
91 [17]. Secondary aims were to investigate the usefulness of specific sensors within devices,
92 and compare commercial and research-grade devices. We hypothesised that the addition of
93 physiological data to accelerometry within wearable devices will provide a more accurate
94 estimate of EE [18], compared with criterion measures, and that the performance of research-
95 grade devices would be superior to commercial devices.

96

97 **Methods**

98 This systematic review and meta-analysis adhered to PRISMA diagnostic test accuracy
99 guideline [19] (supplementary material 1) and was prospectively registered in the
100 PROSPERO database (CRD42018085016).

101

102 Search strategy

103 SportDISCUS (EBSCOHost), PubMed, Medline (Ovid), PsycINFO (EBSCOHost),
104 EMBASE (Ovid) and CINAHL (EBSCOHost) were searched for studies published up to 1st
105 December 2017 using terms relevant to the validation of EE estimates from activity monitors
106 against criterion measures with the following strategy ((tracker AND EE) AND validation).
107 The search was updated 15th January 2018. The specific keywords and the full search strategy
108 can be found in supplementary material 2. No language restrictions were applied and in the
109 case of studies available only as an abstract, attempts were made to contact the authors.

110

111 Inclusion criteria

112 We considered laboratory or field validation studies conducted in healthy adults (≥ 18 years)
113 comparing a criterion measure of EE to an estimate of EE in kilocalories (kcal), kilojoules
114 (kJ) or megajoules (MJ) from an activity monitor. We considered only wrist or arm-worn
115 devices. There is a clear tendency towards wrist worn devices amongst consumer devices and
116 devices worn on alternative anatomical locations produce different accelerometry patterns
117 and therefore estimates of EE [20]. For criterion validation, we considered DLW, indirect
118 calorimetry devices and metabolic chambers [6].

119

120 Exclusion criteria

121 Adults with conditions deemed to produce atypical movement patterns were excluded,
122 including Parkinson's disease, chronic obstructive pulmonary disease, cerebral palsy and
123 amputees. These conditions are often associated with abnormal gait pattern and thus reduce
124 accuracy in EE estimates [21]. Devices requiring external sensors or components were
125 excluded. Studies reporting only accelerometer counts or studies involving post-hoc
126 manipulation of the device output were excluded.

127

128 Study selection

129 Two authors (ROD and JT) independently assessed 100% of titles and abstracts for potential
130 inclusion, with 10% screened independently by a third author (GF). In the case of
131 disagreements between reviewers, the paper was retrieved in full-text and mutual consensus

132 was reached. Remaining articles were screened independently for inclusion at the full-text
133 level by two authors (ROD and JT), with a third author (SS) screening 10%. Similarly,
134 conflicts were resolved by discussion between reviewers.

135

136 Data extraction

137 From each of the included studies, characteristics of participants, validation protocol,
138 criterion measure and the devices tested including model, wear site and output were
139 extracted. Mean difference or EE estimates from the criterion measure and the device were
140 extracted, along with standard deviation (SD), standard error (SE) or 95% confidence
141 intervals (95% CI). If only SE was provided, SE was converted to SD. If data were not
142 provided, authors were contacted to request the raw data. Where values were only presented
143 in figures, a digitiser tool was used [22]. Data was extracted to a specialised spreadsheet and
144 entered into Comprehensive Meta-analysis (CMA) (CMA, version 2; Biostat, Englewood,
145 NJ) for analysis. Data was extracted by one author (ROD) and was cross-checked for data
146 extraction errors. A second author (JT) verified 100% of extracted data and data entered into
147 CMA.

148

149 Quality assessment

150 Risk of bias in included studies was determined using a modified version of the Downs and
151 Black checklist for non-randomised studies [23]. The Downs and Black instrument is an
152 established tool for determination of the quality of a study within a systematic review and
153 meta-analysis [24]. The modified version used in the present study carried a maximum score
154 of 18 and was quantified as: low (≤ 9 , $< 50\%$), moderate (> 9 –14 points, 50–79%), or high (≥ 15
155 points, $\geq 80\%$) [25]. It contained 17 questions, 10 related to reporting, three to external
156 validity and four to internal validity. The risk of bias assessment was performed
157 independently by two authors (ROD and JT), disagreements were resolved by discussion.

158

159 Statistical analysis

160 Descriptive statistics were calculated for studies included within the meta-analysis.
161 EE estimates from the device and criterion, SD or 95% CI, sample sizes and correlation
162 coefficients for within-activity comparisons for each device were used to calculate effect
163 sizes. Correlation coefficients were based on raw data from previously published studies or
164 were conservatively estimated based on the mean of similar devices (supplementary material
165 3). Where a study provided data for more than one comparison for one device, the selected

166 outcomes were pooled to provide a single mean and prevent overpowering of a single study.
167 Hedges' g (ES) [26] and 95% CIs were calculated using CMA, in accordance with the
168 majority of studies in the literature testing the mean bias between activity monitors and
169 criterion measures. A negative ES represents an underestimation relative to the criterion and
170 a positive value represents an overestimation. Interpretation of ES was as follows: <0.20 as
171 trivial, 0.20-0.39 as small, 0.40-0.80 as moderate and >0.80 as large [27]. A random effects
172 model was employed for all analyses based on the assumption that heterogeneity would exist
173 between included studies due to the variability in study design [28]. To determine
174 heterogeneity, the I^2 statistic [29] was utilised and >75% was considered to represent large
175 heterogeneity. To determine susceptibility to bias from one study, a leave one out analysis
176 was conducted where the removal of one study would leave at least three studies. The study
177 associated with the greatest change to significance of the effect is reported. To assist
178 interpretation of the error associated with each device, we calculated the percentage error
179 relative for each device using percentage difference and weight within each meta-analysis.

180

181 Exploration of small study effects

182 To examine small study effects, data were visually inspected with funnel plots and
183 subsequently quantified by using Egger's linear regression intercept [30]. A statistically
184 significant Egger's statistic indicates the presence of a small study effect.

185

186 Moderators and subgroups

187 As well as overall, which represents a combination of all subgroups, subgroup meta-analyses
188 were performed for specific activities/categories: 1) activity energy expenditure (AEE) which
189 included comparisons of EE estimates from the device to a criterion during non-specific
190 exercise protocols, circuits, arm ergometer, rowing and resistance exercises; 2) ambulation
191 and stair climbing; 3) cycling; 4) running; 5) sedentary behaviours and household tasks and
192 6) total energy expenditure (TEE), representing comparisons to DLW.

193

194 We conducted moderator analyses by sensors and all devices were grouped based on
195 the inclusion of the following sensor hardware: 1) accelerometry alone (ACC); 2) heart rate
196 alone (HR); 3) accelerometry and heart rate (ACC+HR); 4) accelerometry and heat sensing or
197 galvanic skin response (ACC+HS) and 5) accelerometry, heart rate sensors and heat sensing
198 or galvanic skin response sensors (ACC+HR+HS). Secondly, moderator analyses were
199 conducted by commercial and research-grade devices. Devices produced by Actical,

200 Actigraph and Bodymedia were considered as research-grade and all other devices included
201 in the analysis were considered commercial devices. Comparisons between each moderator
202 employed a random effects model.

203

204 **Results**

205 Overview

206 A total of 64 studies were included in the systematic review (Supplementary 4). Four studies
207 could not be synthesised by meta-analysis as mean difference between activity monitors and
208 criterion measurements were not provided [12,31–33]; thus, 60 studies were included in the
209 meta-analysis (figure 1) [10,13,41–50,20,51–60,34,61–70,35,71–80,36,81–88,37–40]. A total
210 of 1946 participants were included, with a mean age of 35 years (range 20 to 86 years). The
211 mean BMI was 24.9 kg/m² (range 21.8 to 31.6 kg/m²). Within the included studies, 104
212 comparisons between devices and a criterion were included. This represented 58 commercial
213 and 46 research-grade device comparisons. ACC was comprised of 35 comparisons, 1 in HR
214 devices, 20 in ACC+HR devices, 45 in ACC+HS and 3 in ACC+HR+HS. With regard to
215 activity performed, 35 comparisons were classed as AEE, ambulation and stairs included 55
216 comparisons, 23 were cycling tasks and 38 were running tasks. Sedentary and low-intensity
217 was comprised of 30 comparisons and TEE included 16 comparisons.

218

219 Devices

220 A total of 40 devices were tested in the included studies. One device was forearm-worn, 6
221 were worn on the upper arm (triceps) and 33 were wrist-worn. Characteristics of the devices,
222 number of studies and weighted percentage error for each device is shown in supplementary
223 materials 5.

224

225 Meta-analysis

226 Individual study effect sizes and allocation to moderator variables are provided in
227 supplementary materials 6. A minimum of three comparisons were required for meta-analysis
228 and as such, we report pooled ES for individual devices or moderators where three or more
229 comparisons were available. Statistical outputs for each device are presented in
230 supplementary materials 7.

231

232 Quality assessment

233 The modified Downs and Black scores revealed a median score of 13, with one study being
234 classed as low quality [69], 48 classed as moderate and 11 classed as high quality
235 (supplementary materials 8). The questions included in the modified tool and percentage of
236 studies fulfilling each question is shown in supplementary materials 9.

237

238 Overall

239 A forest plot of individual devices over all activities is shown in figure 2. Overall, devices
240 underestimated EE (ES: -0.23, 95% CI: -0.44 to -0.04; n=104; p=0.03) and showed
241 significant heterogeneity between devices ($I^2 = 92.18\%$; $p < 0.001$). Significant
242 underestimations relative to criterion measures were observed for the Garmin Vivofit (GVF;
243 ES: -1.09, 95% CI: -1.61 to -0.56; n=5; $p < 0.001$) and the Jawbone UP24 (JU24; ES: -1.16,
244 95% CI: -1.79 to -0.53; n=3; $p < 0.001$). The SenseWear Armband Pro3 (SWA p3) also
245 underestimated EE (ES: -0.32, 95% CI: -0.62 to -0.01; n=12; $p = 0.04$). Sensitivity analysis
246 revealed that the removal of six comparisons altered the significance of the SWA p3
247 ($p > 0.05$), the most influential of which decreased the ES to -0.19 (95% CI: -0.50 to 0.11;
248 $p = 0.21$) [81]. The Apple watch (AW) Bodymedia CORE armband (BMC), Fitbit charge HR
249 (FCHR), Fitbit Flex (FF), Jawbone UP (JU), Nike Fuelband (NF), SenseWear Armband
250 (SWA) SenseWear Armband Pro2 (SWA p2), and Mini (SWAM) did not differ significantly
251 from criterion measures. However, sensitivity analysis showed the FCHR differed
252 significantly with the removal of one study (ES: 0.34, 95% CI: 0.20 to 0.49; $p < 0.001$) [88].
253 The NF was the only device that did not display significant heterogeneity between studies (I^2
254 $= 25.44\%$; $p = 0.26$), with the remaining devices having I^2 values $\geq 66.91\%$ (all $p \leq 0.05$). No
255 device showed evidence of small study effects.

256

257 AEE

258 A forest plot of individual devices during activities classed as AEE is shown in
259 supplementary materials 10. For AEE, the pooled estimate of all devices was a non-
260 significant tendency to underestimate EE compared with criterion measures (ES: -0.34, 95%
261 CI: -0.71 to 0.04; n=35; $p = 0.08$) and significant heterogeneity was observed between devices
262 ($I^2 = 94.94\%$; $p < 0.001$). The SWA p2 underestimated EE (ES: -0.78, 95% CI: -1.48 to -0.08;
263 n=3; $p = 0.03$) and had moderate, non-significant heterogeneity ($I^2 = 64.19\%$; $p = 0.06$). The
264 BMC, NF, SWA and SWAM did not differ significantly from criterion measures but all
265 displayed significant heterogeneity. No device showed evidence of small study effects.

266

267

268 Ambulation and stairs

269 A forest plot of individual devices during ambulation and stair climbing is shown in figure 3.
270 The pooled estimate of all devices did not differ from criterion measures (ES: -0.09, 95% CI:
271 -0.45 to 0.27; n=55; p=0.62) and significant heterogeneity was observed between devices (I^2
272 =93.74%; p<0.01). The FCHR (ES: 0.78, 95% CI: 0.27 to 1.29; n=5; p=0.002) and FF (ES:
273 1.10, 95% CI: 0.43 to 1.77; n=3; p=0.001) overestimated EE. The GVF underestimated EE
274 (ES: -1.24, 95% CI: -1.86 to -0.62; n=4; p<0.01), however, sensitivity analysis revealed that
275 the removal of two comparisons significantly altered the mean effect (p>0.05) the most
276 influential significantly altered the mean effect to ES: -1.32 (95% CI: -2.73 to 0.08; p=0.07)
277 [34]. Further, there was evidence of small study effects (intercept= -13.76, 95% CI: -19.72 to
278 -7.80; p=0.01). The SWA overestimated EE (ES: 0.79, 95% CI: 0.25 to 1.33; n=5; p<0.01)
279 and sensitivity analysis revealed that the removal of four comparisons significantly altered
280 the mean effect (p>0.05) the most influential significantly altered the mean effect to ES: 0.33
281 (95% CI: -0.26 to 0.92; p=0.28) [56]. The AW, JU, SWA p3 and SWAM did not differ
282 significantly from criterion measures. The mean effect of the SWAM was significantly
283 altered by the removal of two studies; the removal of the most influential study yielded a
284 significant overestimation (ES: 0.57, 95% CI: 0.20 to 0.94; p=0.003) [87]. All devices
285 showed significant heterogeneity.

286

287 Cycling

288 A forest plot of individual devices during cycling is shown in supplementary materials 10.
289 The pooled estimate of all devices was significantly lower than criterion measures (ES: -0.73,
290 95% CI: -1.39 to -0.06; n=23; p=0.03) and significant heterogeneity was observed between
291 devices (I^2 =94.74%; p<0.01). The SWA did not differ significantly from criterion but
292 showed significant heterogeneity (I^2 =89.39%; p<0.001). The SWA p3 did not differ from
293 criterion measures and showed moderate heterogeneity (I^2 =54.95%; p=0.11).

294

295 Running

296 A forest plot of individual devices during running is shown in supplementary materials 10.
297 The pooled estimate was not statistically different from criterion measures (ES: -0.08, 95%
298 CI: -0.41 to 0.25; n=38; p=0.65) and significant heterogeneity was observed between devices
299 (I^2 =92.05%; p<0.001). The FCHR, GVF and SWA did not differ from criterion measures.

300 Sensitivity analysis revealed the removal of one study changed the overall effect for the
301 FCHR (ES: 0.59, 95% CI: 0.28 to 0.90; $p < 0.001$) [87]. Significant heterogeneity was
302 observed for the FCHR ($I^2 = 66.8\%$; $p = 0.03$) and SWA ($I^2 = 96.79\%$; $p < 0.001$), but not for the
303 GVF ($I^2 = 46.39\%$; $p = 0.15$).

304

305 Sedentary and household tasks

306 A forest plot of individual devices during sedentary and household tasks is shown in figure 4.
307 The pooled effect was not statistically different from criterion measures (ES: -0.09, 95% CI: -
308 0.51 to 0.32; $n = 30$; $p = 0.66$) and significant heterogeneity was observed between devices (I^2
309 $= 94.84\%$; $p < 0.001$). The AW, FCHR and SWAM were not statistically different from
310 criterion measures. The SWA p3 overestimated EE (ES: 0.67, 95% CI: 0.00 to 1.34;
311 $p = 0.049$). Sensitivity analysis revealed that the removal of three studies changed the mean
312 effect, the most influential of which decreased the ES to 0.41 (95% CI: -0.01 to 0.82; $p = 0.05$)
313 [42]. Observed heterogeneity was significant for the AW, SWA p3 and SWAM. The FCHR
314 had moderate, non-significant heterogeneity ($I^2 = 59.60\%$; $p = 0.60$).

315

316 TEE

317 A forest plot of individual devices for the measurement of TEE is shown in figure 5. The
318 pooled effect for TEE showed a significant underestimation of EE (ES: -0.68, 95% CI: -1.15
319 to -0.21; $n = 16$; $p = 0.005$) and significant heterogeneity was observed between devices (I^2
320 $= 92.17\%$; $p < 0.01$). The SWA p3 did not differ significantly from criterion measures and
321 showed significant heterogeneity ($I^2 = 94.20\%$; $p = 0.001$).

322

323 Moderator analyses

324 The results of moderator analyses are shown in table 1. Overall, there was a significant
325 difference between sensors ($p = 0.003$). Pooled estimate of EE from ACC+HR and ACC+HS
326 was not statistically different from criterion but ACC+HS showed a non-significant tendency
327 for underestimation, and ACC and ACC+HR+HS both significantly underestimated EE. In
328 the AEE comparison, there was no statistical difference between sensors, but ACC+HS
329 significantly underestimated EE, ACC showed a non-significant tendency for
330 underestimation and ACC+HR did not differ significantly from criterion measures. During
331 ambulation and stair climbing, a significant difference between sensors was observed, with
332 estimates of EE from ACC+HR and ACC+HS being significantly higher than criterion. In
333 cycling, significant differences were observed between sensors, with ACC devices

334 underestimating EE. During running activities, none of the pooled mean estimates were
335 significantly different from criterion. For sedentary and household tasks, a significant
336 difference was observed between sensors; ACC+HR was not different from criterion
337 measures whereas ACC and ACC+HS underestimated and overestimated EE respectively.
338 For TEE, sensors differed significantly; ACC underestimated EE, whereas ACC+HS did not
339 differ significantly from criterion.

340 When analysed by commercial and research-grade devices, no significant difference
341 was observed overall, for AEE, cycling or running. For both the ambulation and stairs
342 comparison and the sedentary and household tasks comparison, commercial devices were
343 closer to criterion measurements, with research grade devices significantly overestimating.
344 For TEE, research-grade devices were superior, with commercial devices significantly
345 underestimating EE.

346

Moderator variable	Subgroup level	p-value	Hedges' g (95% CI)
Overall activities			
Sensors	ACC (n=35)	<0.01	-0.36 (-0.55, -0.17)*
	ACC + HR (n=20)		0.06 (-0.18, 0.31)
	ACC + HR + HS (n=3)		-0.99 (-1.65, -0.33)*
	ACC + HS (n=45)		-0.151 (-0.32, 0.01)
Device grade	Commercial (n=58)	0.27	-0.269(-0.42, -0.12)*
	Research (n=46)		-0.141 (-0.31, 0.03)
AEE			
Sensors	ACC (n=8)	0.19	-0.40 (-0.84, 0.04)
	ACC + HR (n=9)		-0.04 (-0.47, 0.38)
	ACC + HS (n=16)		-0.32 (-0.63, -0.01)*
Device grade	Commercial (n=18)	0.62	-0.38 (-0.67, -0.08)*
	Research (n=17)		-0.27 (-0.57, 0.04)
Ambulation and stairs			
Sensors	ACC (n=24)	0.01	-0.23 (-0.51, 0.06)
	ACC + HR (n=10)		0.45 (0.02, 0.87)*
	ACC + HS (n=19)		0.40 (0.08, 0.72)*
Device grade	Commercial (n=35)	0.05	-0.04 (-0.28, 0.20)
	Research (n=20)		0.37 (0.05, 0.68)*
Cycling			
Sensors	ACC (n=3)	<0.01	-3.75 (-4.65, -2.85)*
	ACC + HR (n=9)		-0.04 (-0.47, 0.40)
	ACC + HS (n=9)		-0.41 (-0.84, 0.02)
Device grade	Commercial (n=14)	0.28	-0.82 (-1.30, -0.35)*
	Research (n=9)		-0.41 (-0.99, 0.17)
Running			
Sensors	ACC (n=19)	0.18	-0.06 (-0.364, 0.24)
	ACC + HR (n=7)		0.34 (-0.15, 0.82)
	ACC + HS (n=10)		-0.36 (-0.78, 0.05)
Device grade	Commercial (n=28)	0.08	0.06 (-0.18, 0.30)
	Research (n=10)		-0.36 (-0.76, 0.04)
Sedentary and household			
Sensors	ACC (n=6)	<0.01	-0.65 (-1.16, -0.13)*
	ACC + HR (n=9)		0.14 (-0.28, 0.57)
	ACC + HS (n=13)		0.41 (0.06, 0.75)*
Device grade	Commercial (n=17)	<0.01	-0.27 (-0.59, 0.05)
	Research (n=13)		0.41 (0.05, 0.77)*
TEE (DLW)			
Sensors	ACC (n=5)	<0.01	-1.24(-1.66, -0.81)*
	ACC + HS (n=10)		-0.13(-0.397, 0.32)
Device grade	Commercial (n=6)	<0.01	-1.13(-1.51, -0.76)*
	Research (n=10)		-0.13 (-0.39, 0.14)

348 **Discussion**

349

350 Given the clinical and consumer uptake of wrist and arm-worn activity monitors which can
351 be used for the estimation of EE, the aims of this meta-analysis were (i) to determine the
352 relative accuracy of current devices, (ii) to investigate the importance of specific sensors
353 within devices and (iii) to compare commercial and research-grade devices.

354 For devices with sufficient comparisons to be analysed separately from the main
355 pooled effect, significant error relative to criterion measures was observed for Garmin, Fitbit,
356 Jawbone and Bodymedia products. Garmin, Fitbit and Jawbone represent a major share of the
357 commercial wearable market [73] and Bodymedia products are widely used in research and
358 have been since 2004 [59]. Whilst it is initially encouraging that the ES for many devices was
359 not significantly different from criterion, the 95% CI observed in many cases indicates the
360 potential for these devices to produce erroneous estimates of mean EE and as such we would
361 be hesitant to consider any device sufficiently accurate. A 10% 'equivalence zone' has been
362 suggested previously [65] and with the exception of the Nike Fuel band, in which all three
363 studies reported a mean error <10% [65,79,82], no device pooled in this meta-analysis
364 consistently met this criteria. The SenseWear armband Mini was the most accurate device
365 overall but error reported in studies ranged from -21.27% [87] to 14.76% [39]. Studies in this
366 analysis followed the manufacturer's instructions for setup, with researchers ensuring the
367 position of the device and characteristics such as height, weight, sex and age were correct. In
368 free-living environments the lack of researcher presence could yield greater error than
369 observed in this analysis [17], as indicated by the moderate, significant underestimation for
370 the pooled effect in the TEE subgroup.

371

372 An accurate yet affordable measure of TEE, with a measure of change in energy storage,
373 could theoretically be used to retrospectively determine free-living EI in large cohorts [89].
374 In this context, TEE may be considered the most important activity subgroup in this meta-
375 analysis, however, the most variable and unpredictable component of TEE is EE during
376 activity [6]. In agreement with previous studies [13,45,52], we have shown that the accuracy
377 of devices differs by activity and this may be related to the inability of devices to differentiate
378 between activity types. For a device to accurately estimate TEE between individuals, it must
379 accurately estimate the energy cost of a wide range of activities however, some activities may
380 require greater focus. The majority of EE is attributable to rest or non-exercise activity [6] so
381 error here could have a great impact on the error in TEE. The Fitbit Charge HR was the most

382 tested commercial device in this analysis and it showed a trivial, non-significant ES overall
383 and during sedentary tasks but a moderate to large and significant overestimation during
384 ambulatory activity. Considering that ambulatory activity is central to public health
385 guidelines worldwide [90], the implications of this finding may be great for estimates of
386 TEE.

387 The observed error for different activity types may be because current algorithms do
388 not take physical activity type or bodily posture into account [91]. Indeed, activity
389 recognition is considered an important direction for wearable technology [11] and has been
390 used to improve estimates of EE [92]. Montoye et al have shown that accelerometers worn on
391 the wrists and thigh can be used to predict activity type [93]. The SenseWear software
392 employs complex pattern-recognition algorithms to determine activity type [45] which likely
393 contributed to the trivial or small ES observed for the SenseWear Armband Mini in all
394 comparisons. The challenges associated with activity recognition have been reviewed
395 recently [94] and as this technology develops, activity-specific EE prediction equations may
396 offer the opportunity to reduced errors associated with activity types.

397

398 Sensors

399 A 2012 review concluded that multisensory and triaxial accelerometry devices improve
400 estimates of EE, relative to uniaxial devices [21]. Due to recent technological advancements,
401 triaxial accelerometry, as well as heart rate or heat sensing technology are commonplace in
402 newer devices [48]. We hypothesised that the addition of this technology to accelerometry
403 would improve estimates of EE. Overall, this meta-analysis shows that the inclusion of heart
404 rate or heat sensors in devices can improve estimates of EE relative to accelerometry alone.
405 Indeed, it is established that accelerometry is limited for non-weight-bearing activities [84],
406 and accelerometry underestimated EE during cycling activities in our analysis. Significant
407 underestimations were also observed during sedentary and household tasks and TEE, which
408 is likely a product of the limited arm movements associated with these activities.

409 Accelerometry and heart rate devices moderately overestimated EE during ambulation
410 and stair climbing. Some of this error may be attributable to the individual variability in the
411 relationship between heart rate and EE. Individual calibration of this relationship in the
412 Actiheart device is associated with improved estimates of EE [95] and may offer a means for
413 further reducing the error observed in wrist and arm-worn devices. An alternative explanation
414 for this is the variability in estimates of heart rate from photoplethysmography heart rate
415 sensors. A recent study reported a small mean error of -5.9 bpm in the Fitbit Charge 2, but

416 wide limits of agreement of -28.5 to 16.8 bpm [96] and this variability is a common finding
417 [35,40].

418

419 Device Grade

420 The third aim of this meta-analysis was to compare commercial and research-grade devices.

421 Commercial devices may be developed with affordability and comfort as a primary focus,
422 and as a consequence it may be unreasonable to expect commercial devices to match the
423 validity of research-grade devices. Recent consumer monitors share similar technology with
424 established research-grade multi-sensor devices [48] and this is partially reflected in our
425 results. A benefit of research-grade devices for TEE was observed, but commercial devices
426 were statistically superior in ambulation and during sedentary tasks. Our results question the
427 use of wrist or arm-worn research-grade devices for the validation of newer devices.

428 Comparisons to criterion measures such as DLW or indirect calorimetry are more appropriate
429 when absolute accuracy is required [6]. Further, it is important to highlight that other
430 research-grade devices, for instance the Actiheart, which is worn on the chest [95], are likely
431 to be more accurate than research-grade devices included in this study [48]. Further research
432 is needed to establish whether research-grade devices that are worn in other locations such as
433 the chest, hip or thigh outperform consumer based devices.

434

435 Limitations

436 Separate pooled analyses to determine the accuracy of individual activity monitors were
437 performed for a limited number of devices due to the small number of comparisons available
438 for the remaining devices (i.e., less than three comparisons). This limitation is inevitable
439 considering the large number of activity monitors included in this review. Nevertheless, the
440 inclusion of all devices in the overall pooled analysis provides an extensive and robust
441 evaluation of the difference in EE outcomes between activity monitors and criterion
442 measures.

443 The majority of analyses conducted within this review demonstrated large
444 heterogeneity within and between devices which remained after moderating by specific
445 devices and activity. Such heterogeneity is not unexpected and in many cases may be
446 attributable to disparity in the protocols employed [97]. Indirect calorimetry systems were the
447 most commonly used criterion measure but EE estimates may differ by up to 5.2% depending
448 on the equations used [98]. EE is likely to be elevated in the period following higher intensity
449 exercise and the inclusion of only the steady state period may influence the extent to which

450 devices differ from criterion measures [56]. There is also the possibility that the discrepancy
451 between device estimates relates to populations studied [16] for example, a higher BMI
452 [35,40] or age related changes in movement patterns [69]. As few devices currently provide
453 open-access to EE algorithms, the potential for this to create heterogeneity remains uncertain.
454 Despite this, the statistically significant outcomes in many cases suggests a consistent
455 direction in effect sizes for many comparisons and the differences in statistical outcomes
456 between devices are supported by the magnitude of effect sizes.

457 External validity was low in 46 studies pooled in this meta-analysis, which must be
458 considered when interpreting the present results. It must also be noted that the present
459 analysis was limited to healthy individuals and therefore our results cannot be generalized to
460 populations with conditions that produce abnormal gait patterns.

461 Lastly, there is a lag between product release and testing in research environments
462 [40] and some of the devices included in this meta-analysis are no longer in production so the
463 continued validation of newer devices is imperative.

464

465 Conclusion

466 This meta-analysis collated studies evaluating the validity of EE estimates by wrist or
467 arm-worn devices. Devices vary in accuracy depending on activity type and the significant
468 heterogeneity means caution must be exercised when interpreting these results. Devices with
469 heart rate sensors often produced better estimates than devices using accelerometry only;
470 however, this was not consistent across all activities. Wrist and arm-worn research-grade
471 devices were more accurate than commercial devices for estimates of TEE but researchers
472 should be aware that such devices do not guarantee superior accuracy. Future research should
473 aim to understand and reduce the error in EE estimates from wrist or arm-worn devices in
474 different activity types. This may be achieved through activity recognition techniques,
475 incorporating physiological measures and exploring the potential for individual calibration of
476 these relationships.

477

478 **Funding**

479 The research was funded by a University of Leeds PhD studentship. This research received
480 no specific grant from any funding agency in the public, commercial or not-for-profit sectors.
481

482 **Conflicting interests**

483 None

484 Reference list

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783 **Legends:**

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785 **Table 1.** Moderation analysis for level of sensors and grade of device by subgroup. Data are
786 shown where at least 3 comparisons were included. *P*-value refers to a between subgroup
787 comparison. *Significant effect size at the subgroup level ($p < .05$). Abbreviations:
788 Accelerometry alone (ACC), accelerometry and heart rate (ACC+HR), accelerometry and
789 heart rate and heat sensing (ACC+HR+HS) and accelerometry and heat sensing (ACC+HS).
790 Activity energy expenditure (AEE), Total energy expenditure (TEE), Doubly labelled water
791 (DLW).

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793 **PLEASE INSERT FIGURE 1 AROUND LINE 216**

794 **Figure 1.** Flow diagram of study selection.

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796 **PLEASE INSERT FIGURE 2 AROUND LINE 254**

797 **Figure 2.** Pooled Hedges' *g* and 95% confidence intervals (CI) for estimates of energy
798 expenditure relative to criterion measures per device over all activities. Total refers to
799 number of effect sizes. A negative Hedges' *g* statistic represents an underestimation and a
800 positive Hedges' *g* represents an overestimation.

801 Abbreviations: *Actical (ACT)*, *Actigraph GT3X (AGT3X)*, *Apple watch (AW)*, *Apple Watch*
802 *series 2 (AWS2)*, *Beurer AS80 (BA)*, *Bodymedia CORE armband (BMC)*, *Basis Peak (BP)*,
803 *Epson Pulsense (EP)*, *ePulse Personal Fitness Assistant (EPUL)*, *Fitbit Blaze (FB)*, *Fitbit*
804 *Charge (FC)*, *Fitbit Charge 2 (FC2)*, *Fitbit Charge HR (FCHR)*, *Fitbit Flex (FF)*, *Garmin*
805 *Forerunner 225 (GF225)*, *Garmin Forerunner 920XT (GF920XT)*, *Garmin Vivoactive*
806 *(GVA)*, *Garmin vivofit (GVF)*, *Garmin vivosmart (GVS)*, *Garmin Vivosmart HR (GVHR)*,
807 *Jawbone UP (JU)*, *Jawbone UP24 (JU24)*, *LifeChek calorie sensor (LC)*, *Mio Alpha (MA)*,
808 *Microsoft band (MB)*, *Misfit Shine (MS)*, *Nike Fuel band (NF)*, *Polar Loop (PL)*, *Polar:*
809 *AW200 (PO200)*, *Polar: AW360 (PA360)*, *Samsung Gear S (SG)*, *SenseWear Armband*
810 *(SWA)*, *SenseWear Armband Pro 2 (SWA p2)*, *SenseWear Armband Pro 3 (SWA p3)*,
811 *SenseWear Armband MINI (SWAM)*, *TOMTOM Touch (TT)*, *Vivago (V)*, *Withings Pulse*
812 *(WP)*, *Withings Pulse O2 (WPO)*.

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814 **PLEASE INSERT FIGURE 3 AROUND LINE 284**

815 **Figure 3.** Pooled Hedges' *g* and 95% confidence intervals (CI) for estimates of energy
816 expenditure relative to criterion measures per device for ambulation and stair climbing.

817 Total refers to number of effect sizes. A negative Hedges' *g* statistic represents an
818 underestimation and a positive Hedges' *g* represents an overestimation.

819 Abbreviations: *Actigraph GT3X (AGT3X)*, *Apple watch (AW)*, *Beurer AS80 (BA)*, *Bodymedia*
820 *CORE armband (BMC)*, *Basis Peak (BP)*, *ePulse Personal Fitness Assistant (EPUL)*, *Fitbit*
821 *Charge (FC)*, *Fitbit Charge HR (FCHR)*, *Fitbit Flex (FF)*, *Garmin Forerunner 225 (GF225)*,
822 *Garmin Forerunner 920XT (GF920XT)*, *Garmin Vivoactive (GVA)*, *Garmin vivofit (GVF)*,
823 *Garmin vivosmart (GVS)*, *Jawbone UP (JU)*, *Jawbone UP24 (JU24)*, *Microsoft band (MB)*,
824 *Nike Fuel band (NF)*, *Polar Loop (PL)*, *Polar: AW200 (PO200)*, *SenseWear Armband*
825 *(SWA)*, *SenseWear Armband Pro 2 (SWA p2)*, *SenseWear Armband Pro 3 (SWA p3)*,
826 *SenseWear Armband MINI (SWAM)*, *Vivago (V)*, *Withings Pulse (WP)*, *Withings Pulse O2*
827 *(WPO)*.

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829 **PLEASE INSERT FIGURE 4 AROUND LINE 313**

830 **Figure 4.** Pooled Hedges' *g* and 95% confidence intervals (CI) for estimates of energy
831 expenditure relative to criterion measures per device for sedentary and household tasks.

832 Total refers to number of effect sizes. A negative Hedges' g statistic represents an
833 underestimation and a positive Hedges' g represents an overestimation.
834 Abbreviations: *Apple watch (AW)*, *Bodymedia CORE armband (BMC)*, *Basis Peak (BP)*,
835 *ePulse Personal Fitness Assistant (EPUL)*, *Fitbit Charge HR (FCHR)*, *Fitbit Flex (FF)*,
836 *Garmin Forerunner 225 (GF225)*, *Garmin vivofit (GVF)*, *Jawbone UP (JU)*, *Jawbone UP24*
837 *(JU24)*, *Microsoft band (MB)*, *SenseWear Armband Pro 2 (SWA p2)*, *SenseWear Armband*
838 *Pro 3 (SWA p3)*, *SenseWear Armband MINI (SWAM)*, *Vivago (V)*, *Withings Pulse (WP)*.

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840 **PLEASE INSERT FIGURE 5 AROUND LINE 320**

841 **Figure 5.** Pooled Hedges' g and 95% confidence intervals (CI) for estimates of energy
842 expenditure relative to criterion measures per device for total energy expenditure (TEE).

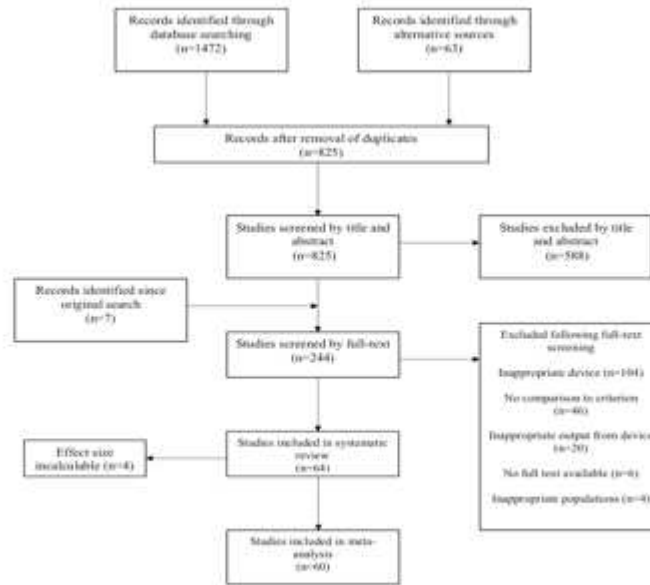
843 Total refers to number of effect sizes. A negative Hedges' g statistic represents an
844 underestimation and a positive Hedges' g represents an overestimation.

845 Abbreviations: *Epson Pulsense (EP)*, *Fitbit Flex (FF)*, *Garmin vivofit (GVF)*, *Jawbone UP24*
846 *(JU24)*, *Misfit Shine (MS)*, *SenseWear Armband (SWA)*, *SenseWear Armband Pro 2 (SWA*
847 *p2)*, *SenseWear Armband Pro 3 (SWA p3)*, *SenseWear Armband MINI (SWAM)*, *Withings*
848 *Pulse O2 (WPO)*.

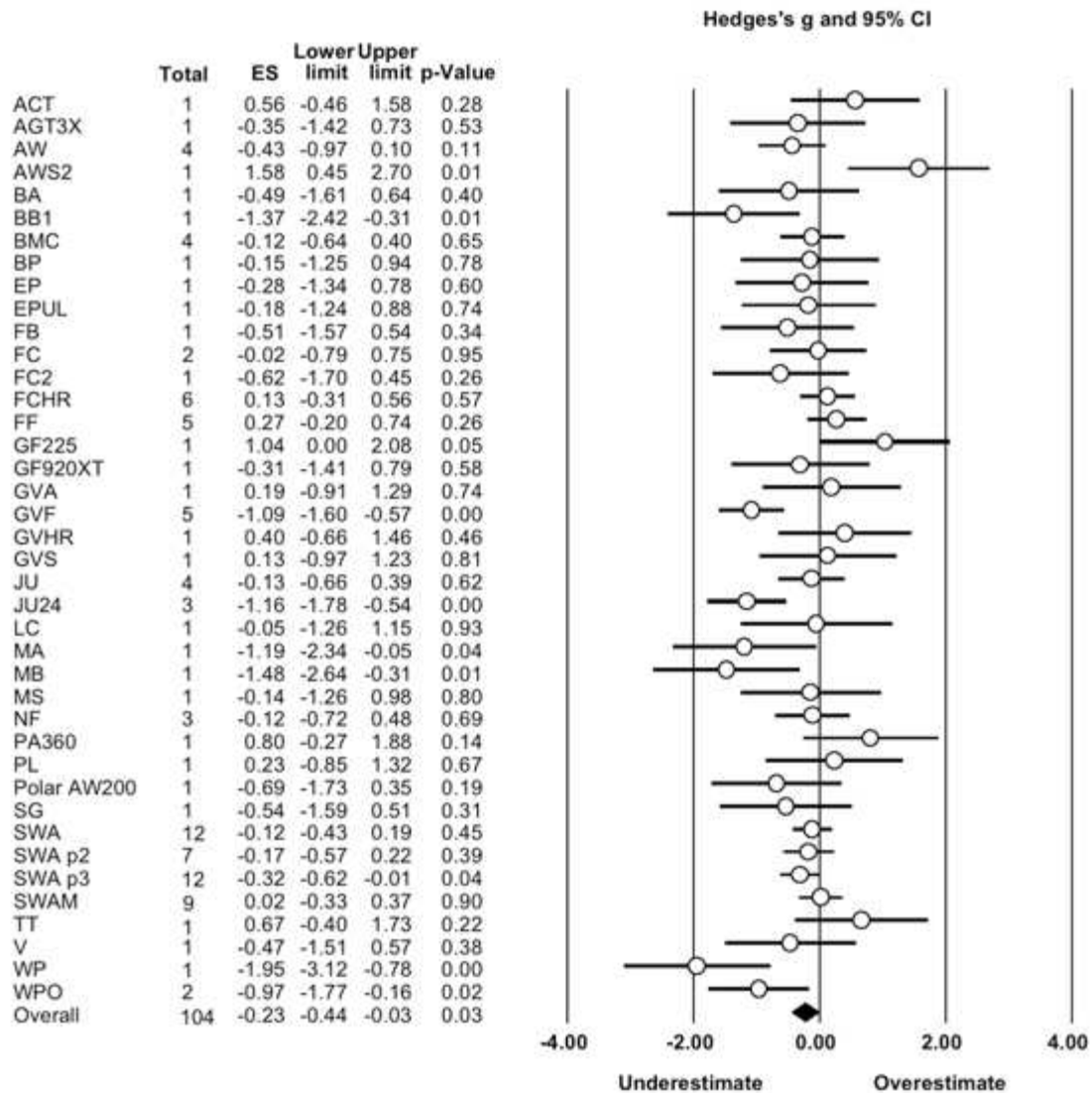
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851 *Figure 1:*



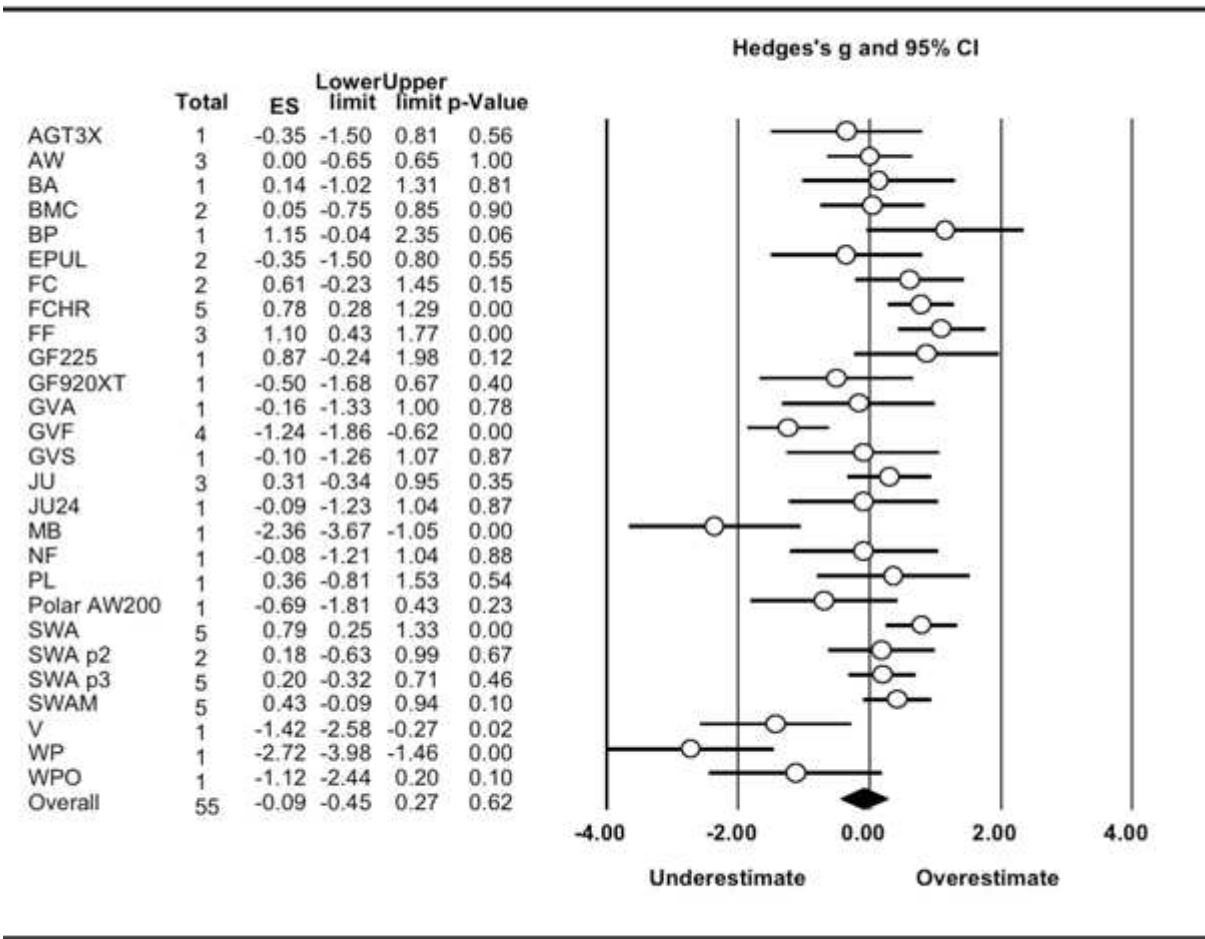
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 854 *Figure 2:*



Meta Analysis Overall

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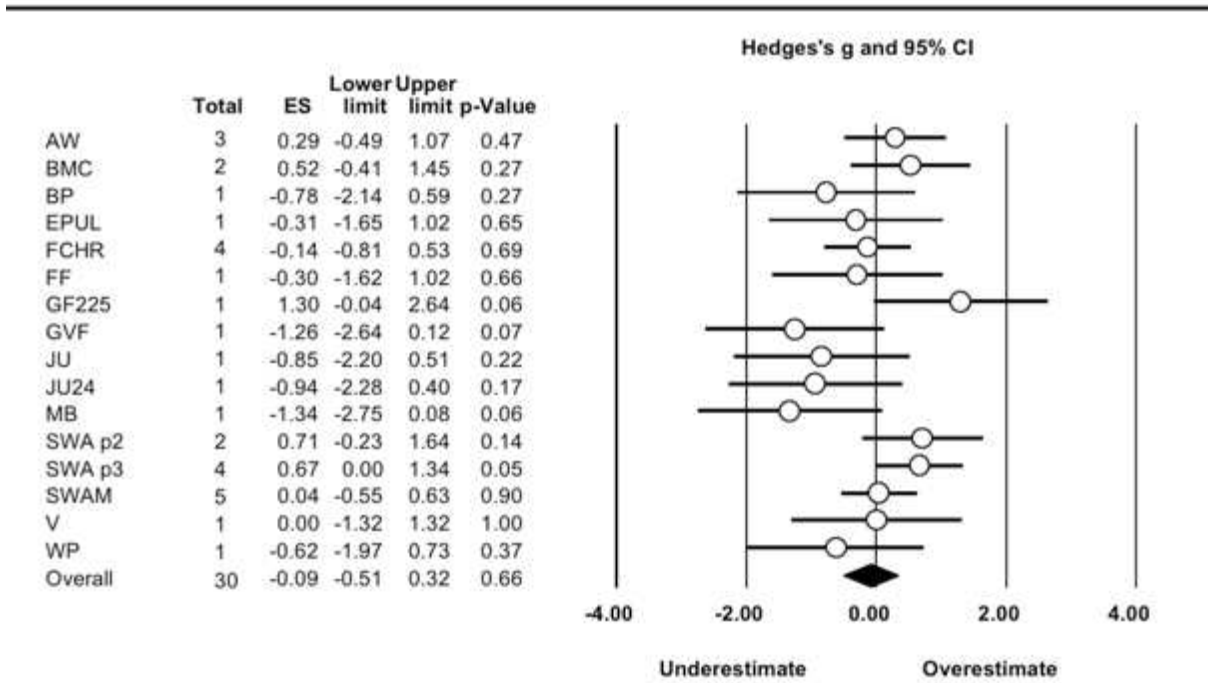
Figure 3:



Meta Analysis Ambulation and Stairs

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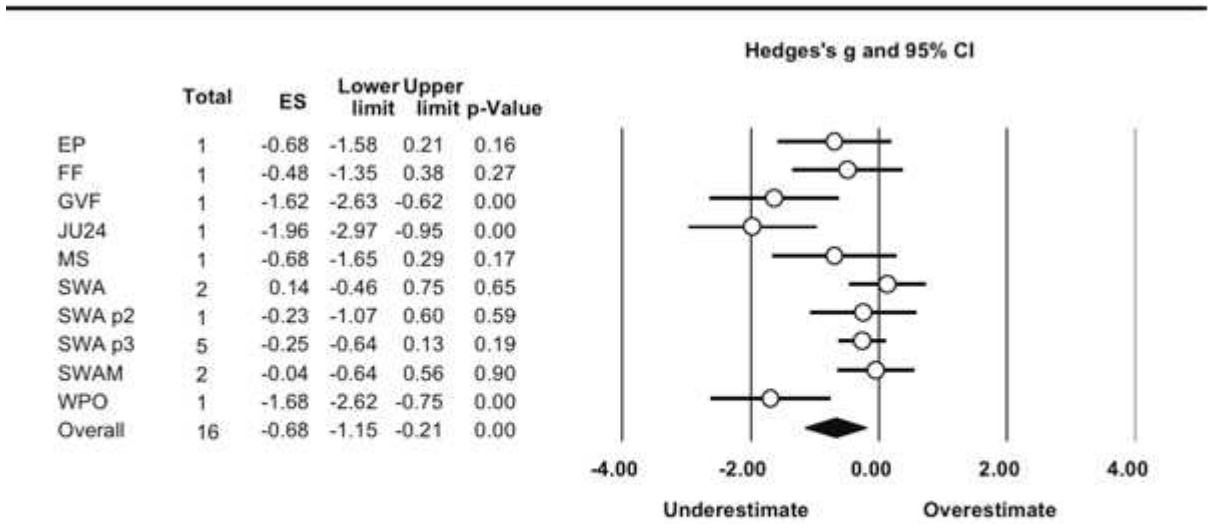
Figure 4:



Meta Analysis Sedentary and Household

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Figure 5:



Meta Analysis TEE (DLW)

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867 S1:



PRISMA-DTA Checklist

Section/topic	#	PRISMA-DTA Checklist Item	Reported on page #
TITLE / ABSTRACT			
Title	1	Identify the report as a systematic review (+/- meta-analysis) of diagnostic test accuracy (DTA) studies.	1
Abstract	2	Abstract: See PRISMA-DTA for abstracts.	2
INTRODUCTION			
Rationale	3	Describe the rationale for the review in the context of what is already known.	5 & 6
Clinical role of index test	D1	State the scientific and clinical background, including the intended use and clinical role of the index test, and if applicable, the rationale for minimally acceptable test accuracy (or minimum difference in accuracy for comparative design).	5 & 6
Objectives	4	Provide an explicit statement of question(s) being addressed in terms of participants, index test(s), and target condition(s).	5 & 6
METHODS			
Protocol and registration	5	Indicate if a review protocol exists, if and where it can be accessed (e.g., Web address), and, if available, provide registration information including registration number.	6
Eligibility criteria	6	Specify study characteristics (participants, setting, index test(s), reference standard(s), target condition(s), and study design) and report characteristics (e.g., years considered, language, publication status) used as criteria for eligibility, giving rationale.	6
Information sources	7	Describe all information sources (e.g., databases with dates of coverage, contact with study authors to identify additional studies) in the search and date last searched.	7
Search	8	Present full search strategies for all electronic databases and other sources searched, including any limits used, such that they could be repeated.	S2
Study selection	9	State the process for selecting studies (i.e., screening, eligibility, included in systematic review, and, if applicable, included in the meta-analysis).	7
Data collection process	10	Describe method of data extraction from reports (e.g., piloted forms, independently, in duplicate) and any processes for obtaining and confirming data from investigators.	7
Definitions for data extraction	11	Provide definitions used in data extraction and classifications of target condition(s), index test(s), reference standard(s) and other characteristics (e.g. study design, clinical setting).	7
Risk of bias and applicability	12	Describe methods used for assessing risk of bias in individual studies and concerns regarding the applicability to the review question.	7-8
Diagnostic accuracy measures	13	State the principal diagnostic accuracy measure(s) reported (e.g. sensitivity, specificity) and state the unit of assessment (e.g. per-patient, per-lesion).	8
Synthesis of results	14	Describe methods of handling data, combining results of studies and describing variability between studies. This could include, but is not limited to: a) handling of multiple definitions of target condition, b) handling of multiple thresholds of test positivity, c) handling multiple index test readers, d) handling of indeterminate test results, e) grouping and comparing tests, f) handling of different reference standards	

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869 S2:

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871 **Population:** Healthy adult populations (>18). Free from factors that impact physical movement.
872 **Intervention:** activity monitors + all research grade accelerometers (must be wearable on wrist or arm)
873 **Comparison:** Validated method: metabolic cart, DLW, DC, all IC systems,
874 **Outcome:** validity of energy expenditure (kcal/kj/met/correlation),
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	P	I	C	O
Key concepts	ADULTS	ACTIVITY MONITORS	VALIDATED METHOD	ENERGY EXPENDITURE
Related terms		FITNESS TRACKERS (CINHAL) ACCELEROMETRY (MESH) ACCELEROMETER AMBULATORY MONITOR* FITBIT ACTIVITY MONITOR	VALID* COMPAR* TEST	ENERGY METABOLISM (MESH) CALORIES ENERGY EXPENDITURE CALORIC EXPENDITURE TOTAL DAILY ENERGY EXPENDITURE TDEE AEE
Terms to include in search		<ol style="list-style-type: none"> 1. Activity tracker 2. Activity Monitor 3. Health tracker 4. Health monitor 5. Fitness tracker 6. Fitness monitor 7. Physical activity tracker 8. Physical activity monitor 9. Exercise tracker 10. Exercise monitor 	<ol style="list-style-type: none"> 1. Doubly labelled water 2. DIw 3. Indirect caliomet* 4. Caliomet* 5. Direct caliomet* 6. Metabolic chamber 7. Metabolic cart 8. Gold standard 9. Criterion 	<p>Energy expenditure</p> <ol style="list-style-type: none"> 1. Energy metabolism 2. Calori* 3. Calori* expenditure 4. Total energy expenditure 5. Activity energy expenditure 6. AEE 7. TDEE

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		11. Electronic tracker		
		12. Electronic monitor		
		13. acceleromet		
		14. Step tracker		
		15. Wearable		

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911 (Tracker AND EE) AND Validation

912 1. Activity tracker

913 2. Activity Monitor

914 3. Health tracker

915 4. Health monitor

916 5. Fitness tracker

917 6. Fitness monitor

918 7. Physical activity tracker

919 8. Physical activity monitor

920 9. Exercise tracker

921 10. Exercise monitor

922 11. Electronic tracker

923 12. Electronic monitor

924 13. acceleromet

925 14. Step tracker

926 15. Wearable

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928 AND

929 1. Energy expenditure

930 2. Energy metabolism

931 3. Calori*

932 4. Calori* expenditure

933 5. Total energy expenditure

934 6. Activity energy expenditure

935 7. AEE

936 8. TDEE

937 AND

938 1. Doubly labelled water

939 2. Dlw

940 3. Indirect caliomet*

941 4. Caliomet*

942 5. Direct caliomet*

943 6. Metabolic chamber

944 7. Metabolic cart

945 8. Gold standard

946 9. Criterion

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Database	Search	Results
Sport discus	(activity tracker or activity monitor or health tracker or health monitor or fitness tracker or fitness monitor or physical activity tracker or physical activity monitor or exercise tracker or exercise monitor or electronic tracker or electronic monitor or acceleromet* or step tracker or wearable tracker) AND (energy expenditure or energy metabolism or kalori* or kalori* expenditure or total energy expenditure or activ* energy expenditure or AEE or TDEE) AND (doubly labelled water or DLW or indirect caliomet* or caliomet* or direct caliomet* or metabolic chamber or metabolic cart or gold standard or criterion)	154
Pubmed	<i>((((((((((((((((activity tracker) OR activity monitor) OR health tracker) OR health monitor) OR fitness trackers) OR fitness monitor) OR physical activity tracker) OR physical activity monitor) OR exercise trained) OR exercise monitor) OR electronic trackers) OR electronic monitor) OR acceleromet*) OR step tracer) OR wearable trackers)) AND (((((((energy expenditure) OR energy metabolism) OR kalori*) OR kalori* expenditure) OR total energy expenditure) OR activ* energy expenditure) OR AEE) OR tdee))) AND (((((((doubly labelled water) OR DLW) OR indirect caliomet*) OR caliomet*) OR direct caliomet*) OR metabolic chamber) OR metabolic cart) OR gold standard) OR criterion).</i>	605
MEDLINE	((activity tracker or activity monitor or health tracker or health monitor or fitness tracker or fitness monitor or physical activity tracker or physical activity monitor or exercise tracker or exercise monitor or electronic tracker or electronic monitor or acceleromet* or step tracker or wearable tracker).mp.	228

	<p>AND (energy expenditure or energy metabolism or calori* or calori* expenditure or total energy expenditure or activ* energy expenditure or AEE or TDEE).mp.</p> <p>AND (doubly labelled water or DLW or indirect calimet* or calimet* or direct calimet* or metabolic chamber or metabolic cart or gold standard or criterion).mp. [mp=title, abstract, heading word, drug trade name, original title, device manufacturer, drug manufacturer, device trade name, keyword, floating subheading word]</p>	
Psycinfo	<p>((activity tracker or activity monitor or health tracker or health monitor or fitness tracker or fitness monitor or physical activity tracker or physical activity monitor or exercise tracker or exercise monitor or electronic tracker or electronic monitor or acceleromet* or step tracker or wearable tracker).mp.</p> <p>AND (energy expenditure or energy metabolism or calori* or calori* expenditure or total energy expenditure or activ* energy expenditure or AEE or TDEE).mp.</p> <p>AND (doubly labelled water or DLW or indirect calimet* or calimet* or direct calimet* or metabolic chamber or metabolic cart or gold standard or criterion).mp. [mp=title, abstract, heading word, drug trade name, original title, device manufacturer, drug manufacturer, device trade name, keyword, floating subheading word]</p>	26
Embase	<p>((activity tracker or activity monitor or health tracker or health monitor or fitness tracker or fitness monitor or physical activity tracker or physical activity monitor or exercise tracker or exercise monitor or electronic tracker or electronic monitor or acceleromet* or step tracker or wearable tracker).mp.</p> <p>AND (energy expenditure or energy metabolism or calori* or calori* expenditure or total energy expenditure or activ* energy expenditure or AEE or TDEE).mp.</p>	317

	AND (doubly labelled water or DLW or indirect caloriem* or caloriem* or direct caloriem* or metabolic chamber or metabolic cart or gold standard or criterion).mp. [mp=title, abstract, heading word, drug trade name, original title, device manufacturer, drug manufacturer, device trade name, keyword, floating subheading word]	
CINHAL	(activity tracker or activity monitor or health tracker or health monitor or fitness tracker or fitness monitor or physical activity tracker or physical activity monitor or exercise tracker or exercise monitor or electronic tracker or electronic monitor or acceleromet* or step tracker or wearable tracker) AND (energy expenditure or energy metabolism or calori* or calori* expenditure or total energy expenditure or activ* energy expenditure or AEE or TDEE) AND (doubly labelled water or DLW or indirect caloriem* or caloriem* or direct caloriem* or metabolic chamber or metabolic cart or gold standard or criterion)	142
Obtained from reference lists		63
		AFTER REMOVAL OF DUPLICATES: 825

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Exclusions:

- 1 = not comparison to criterion**
- 2 = not comparison to accelerometer**
- 3 = not healthy adult population**
- 4 = review**
- 5 = not kcal/kj**
- 6= duplicate**

974 S3:

	ae	walk	bike	run	sedentary and household	tee
Actical	0.77	0.8	0.3	0.8	0.7	0.7
Actigraph GT3X+	0.72	0.72	0.68	0.72	0.88	0.88
Apple watch	0.79	0.54	0.53	0.65	0.46	0.46
Basis b1	0.51	0.55	0.45	0.55	0.51	0.51
Basis Peak	0.51	0.55	0.45	0.66	0.49	0.49
Beurer AS80	0.44	0.44	0.44	0.44	0.44	0.44
BodyMedia FIT CORE	0.73	0.72	0.66	0.73	0.77	0.77
Epson Pulsense	0.71	0.71	0.71	0.71	0.71	0.71
ePulse Personal Fitness Assistant (ePulse)	0.24	0.24	0.24	0.24	0.24	0.24
Fitbit charge	0.32	0.68	0.44	0.68	0.41	0.41
Fitbit charge HR	0.77	0.75	0.53	0.68	0.41	0.41
Fitbit Flex	0.8	0.8	0.71	0.8	0.71	0.71
Fitbit Surge	0.77	0.75	0.53	0.68	0.41	0.41
Garmin Forerunner 225	0.35	0.35	0.35	0.35	0.35	0.35
Garmin Forerunner 920XT	0.35	0.35	0.35	0.35	0.35	0.35
Garmin vivoactive	0.75	0.75	0.75	0.35	0.75	0.75
Garmin vivofit	0.75	0.75	0.75	0.35	0.75	0.75
Garmin Vivosmart	0.75	0.75	0.75	0.35	0.75	0.75
Jawbone UP	0.82	0.8	0.73	0.74	0.53	0.53
Jawbone UP24	0.69	0.69	0.69	0.69	0.69	0.69
LifeChek calorie sensor	0.45	0.45	0.45	0.45	0.45	0.45
Microsoft band	0.54	0.55	0.46	0.54	0.44	0.44
Mio Alpha	0.46	0.46	0.46	0.46	0.46	0.46
Misfit Shine	0.41	0.41	0.41	0.41	0.41	0.41
Nike Fuel Band	0.77	0.77	0.77	0.77	0.77	0.77
Polar Loop	0.46	0.46	0.47	0.46	0.46	0.46
Polar: Activity Watch 200 (AW200)	0.7	0.7	0.7	0.7	0.7	0.7
PulseOn	0.45	0.45	0.4	0.45	0.45	0.45
Samsung Gear S	0.76	0.76	0.76	0.76	0.76	0.76
SenseWear Armband	0.73	0.72	0.66	0.73	0.77	0.77
SenseWear Mini Armband	0.73	0.72	0.66	0.73	0.77	0.77
SenseWear Pro 2 Armband	0.73	0.72	0.66	0.73	0.77	0.77
SenseWear Pro 3 Armband	0.73	0.72	0.66	0.73	0.77	0.77
TOM TOM TOUCH	0.2	0.3	0.3	0.3	0.3	0.3
Vivago	0.79	0.79	0.79	0.79	0.79	0.79
Withings Pulse	0.71	0.71	0.71	0.71	0.71	0.71
Withings Pulse 02	0.78	0.78	0.78	0.78	0.78	0.78

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976 Correlations imputed for specific devices and activities

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Study	Sample characteristics	Study protocol	Setting (Lab/Field)	Criterion comparison	Device	Device placement	Results (overall error relative to criterion)
Alsubheer, 2016	N=13 (5 F) Age: 40 ± 11.9 y	Subjects performed a graded	Lab	IC – Sable system (Sable Systems)	Garmin vivofit (Garmin Ltd, Olathe,	Wrist	Garmin vivofit: -41.63%

Bai, 2017	BMI: 27 ± 4.3 kg/m ² N=39 (16 F) Age: 32 ± 11 y BMI: 24.7 ± 4 kg/m ²	treadmill test. Subjects performed a semi-structured activity protocol consisting of sedentary activity, aerobic exercise, and light intensity physical activity on a treadmill.	Lab	International, Las Vegas NV) IC – Oxycon Mobile portable metabolic system (Erich Jaeger, Viasys Healthcare, Germany)	Kansas, USA) Apple watch (Apple Inc, Cupertino, California, USA) Fitbit charge HR (Fitbit Inc, San Francisco, California, USA)	Wrist	Apple Watch: -10.79% Fitbit Charge HR: 17.88%
Benito, 2012	N=29 (17 F) Age: 22.5 y BMI: 22 kg/m ²	Subjects performed circuits of resistance exercise at 30%, 50% and 70% of 15 repetition maximum.	Lab	IC – Oxycon Mobile portable metabolic system (Erich Jaeger, Viasys Healthcare, Germany)	SenseWear Pro2 Armband (HealthWe ar, Bodymedia , Pittsburgh, PA, USA)	Upper arm	SenseWear Pro2 Armband: -46.60%
Berntsen, 2010	N=20 (6 F) Age: 35 y BMI: 24 kg/m ²	Subjects performed lifestyle and sporting activities including strength exercises, ball games, occupational and home-based activities.	Lab	IC – MetaMax II (Cortex Biophysic, Leipzig, Germany)	SenseWear Pro2 Armband (HealthWe ar, Bodymedia , Pittsburgh, PA, USA)	Upper arm	SenseWear Pro2 Armband: -9.00%
Berntsen, 2012	N=29 (29 F) Age: 31 ± 4.1 y BMI: 27 ± 3.2 kg/m ²	Subjects participated in a period of sedentary behaviour. 9 subjects then performed callisthenics and cycling on a bicycle ergometer. The other 20 subjects performed outdoor walking followed by	Lab	IC – MetaMax II (Cortex Biophysic, Leipzig, Germany)	SenseWear Pro2 Armband (HealthWe ar, Bodymedia , Pittsburgh, PA, USA)	Upper arm	SenseWear Pro2 Armband: -10.34%

		relaxing, cycling and callisthenics.					
Bhammar , 2016	N=34 (26 F) Age: 30.1 ± 8.7 y BMI: 26.2 ± 5.1 kg/m ²	Subjects performed a semi structured and a structured routine. Semi- structured: 12 activities including 4 sedentary/light-intensity activities, 4 moderate-intensity activities, and 4 vigorous-intensity activities. The activities performed were randomly selected from a list of common activities. Structured: A period of rest, followed by 7 activities of 8 minutes each. The activities performed were randomly selected from a list of common activities.	Lab	IC – Oxycon Mobile portable metabolic system (Erich Jaeger, Viasys Healthcare, Germany)	SenseWear Mini Armband (HealthWear, Bodymedia , Pittsburgh, PA, USA)	Upper arm	SenseWear Mini Armband: 14.76%
Boudreaux, 2018	N=50 (28 F) Age: 22.4 y BMI: 26.5 kg/m ²	Subjects performed separate trials of graded cycling and 3 sets of 4 resistance exercises at a 10-	Lab	IC – Parvo TrueOne 2400 (Parvo Medics, East Sandy, UT, USA)	Apple Watch 2 (Apple Inc, Cupertino, California, USA) Fitbit Blaze (Fitbit Inc, San		Apple Watch 2: 48.20% Fitbit Blaze: 28.66% Fitbit Charge 2: -30.97%

		repetition maximum load.			Francisco, California, USA)		Garmin Vivosmart HR: 16.85%
					Fitbit Charge 2 (Fitbit Inc, San Francisco, California, USA)		Polar: the Activity Watch 360: 28.68%
					Garmin Vivosmart HR (Garmin Ltd, Olathe, Kansas, USA)		Tomtom Touch: 28.66%
					Polar: the Activity Watch 360 (Polar Electro Oy, Kempele, Finland)		
					Tomtom touch (TomTom, Amsterdam, the Netherlands)		
Brazeau, 2011	N=31 (16 F) Age: 26.7 y BMI: 27.5 kg/m ²	Subjects performed 45 minutes of stationary cycling at 50% VO _{2peak} .	Lab	IC – Ergocard exercise test station (MediSoft, Dinant, Belgium)	SenseWear Pro3 Armband (HealthWear, Bodymedia, Pittsburgh, PA, USA)	Upper arm	SenseWear Pro3 Armband: - 10.56%
Brazeau, 2014	N=38 (18 F) Age: 28.6 y BMI: 23.8 kg/m ²	Subjects performed 45 minutes of treadmill exercise at 40% VO _{2peak} then exercised on a stationary bike ergometer for 45 minutes at 50% VO _{2peak} .	Lab	IC – Ergocard exercise test station (MediSoft, Dinant, Belgium)	SenseWear Pro3 Armband (HealthWear, Bodymedia, Pittsburgh, PA, USA)	Upper arm	SenseWear Pro3 Armband: 14.94%

Brazeau, 2016	N=20 (0 F) Age: 26.2 ± 3.6 y BMI: 23.1 ± 2.3 kg/m ²	Subjects completed a field observation and a lab protocol. Field: 7-day comparison to DLW. Lab: Subjects performed 60 minutes rest followed by treadmill exercise for 45 minutes at 22-41% VO _{2peak} then stationary cycling for 45 minutes at 50% VO _{2peak} .	Lab/ Field	DLW – 7 days IC – Ergocard exercise test station (MediSoft, Dinant, Belgium)	SenseWear Pro3 Armband (HealthWe ar, Bodymedia , Pittsburgh, PA, USA)	Upper arm	SenseWear Pro3 Armband: 7.06%
Brugniaux, 2010	N=31 (16 F) Age: 42.9 y BMI: 22.7 kg/m ²	Subjects performed a 9.7km outdoor hike.	Field	IC – Metablograph with Hans Rudolph facemask (Hans Rudolph, Kansas City, MO, USA)	Polar: the Activity Watch 200 (Polar Electro Oy, Kempele, Finland)	Wrist	Polar: the Activity Watch 200: - 13.17%
Calabro, 2014	N=40 (19 F) Age: 27.4 y BMI: 22.8 kg/m ²	Subjects performed 60 minutes of structured activities including stationary biking, walking/ running on a treadmill, road biking, elliptical exercise and stair stepping and unstructured movements. The semi-structured measurement periods were	Lab	IC – Oxycon Mobile portable metabolic system (Erich Jaeger, Viasys Healthcare, Germany)	SenseWear Mini Armband (HealthWe ar, Bodymedia , Pittsburgh, PA, USA) SenseWear Pro3 Armband (HealthWe ar, Bodymedia , Pittsburgh, PA, USA)	Upper arm	SenseWear Mini Armband: 0.89% SenseWear Pro3 Armband: 2.33%

Calabro, 2015	N=29 (17 F) Age: 68.8 ± 6.3 y BMI: 26.3 ± 4.9 kg/m ²	performed in 5, 10, 10, 10, and 25-minute intervals and included sitting, walking, standing, stair climbing or light movements. 14-day comparison to DLW.	Field	DLW – 14 days	SenseWear Mini Armband (HealthWear, Bodymedia, Pittsburgh, PA, USA)	Upper arm	SenseWear Mini Armband: -0.86%
Casiraghi, 2013	N=18 (11 F) Age: 48.6 ± 21 y BMI: 24.6 ± 2.6 kg/m ²	Subjects performed a cycling protocol with three components: 1) Baseline where the subject sat on the cycle ergometer. 2) A 2-minute warm-up at 40 rpm at 40 watts. 3) Exercise increased to 60 rpm and intensity progressed by 7 watts/minute until exhaustion.	Lab	IC – SensorMedics Vmax 229 (SensorMedics Inc, Yorba Linda, CA, USA).	SenseWear Armband (HealthWear, Bodymedia, Pittsburgh, PA, USA)	Upper arm	SenseWear Armband: -8.00%
Chowdhry, 2017	N=30 (15 F) Age: 27 ± 1.6 y BMI: 23.4 ± 2.5 kg/m ²	Subjects performed two components: 1) A protocol of 4 activities of designed to replicate daily living tasks 2) 4 activities of	Lab	IC – COSMED K4b2 (COSMED, Rome, Italy)	Apple watch (Apple Inc, Cupertino, California, USA) Microsoft Band (Microsoft Corporation, Redmond,	Wrist Bodymedia core: Upper arm	Apple watch: -6.9% Microsoft Band: -49.15% Fitbit Charge HR: 15.49% Jawbone UP24:

		10 minutes in duration. These activities were walking on a treadmill, walking at the same speed with shopping bags, cycling on an ergometer and jogging on the treadmill.			Washington, USA)		-21.01%
					Fitbit Charge HR (Fitbit Inc, San Francisco, California, USA)		Bodymedia Core: 7.98%
					Jawbone UP24 (Jawbone, San Francisco, California, USA)		
					Bodymedia Core (HealthWe ar, Bodymedia , Pittsburg, PA, USA)		
Colbert, 2011	N=56 (45 F) Age: 74.7 ± 6.5 y BMI: 25.8 ± 4.2 kg/m ²	10-day comparison to DLW.	Field	DLW – 10 days	SenseWear Pro 3 Armband (HealthWe ar, Bodymedia , Pittsburgh, PA, USA)	Upper arm	SenseWear Pro 3 Armband: 58.53%
Correa, 2016	N=87 (72 F) Age: 42 ± 13 y BMI: 31.6 ± 4.5 kg/m ²	7-day comparison to DLW.	Field	DLW – 7 days	SenseWear Armband (HealthWe ar, Bodymedia , Pittsburgh, PA, USA)	Upper arm Wrist	SenseWear Armband –416.95 kcal Actual: 194.52 kcal
Diaz, 2015	N=23 (13 F) Age: N/A BMI: N/A	Subjects performed a treadmill protocol consisting of walking at slow, moderate and brisk	Lab	IC – Ultima CPX (Medgraphi cs, Saint Paul, MN, USA)	Actical (Phillips Respiroincs Inc, Murrysville , PN, USA) Fitbit Flex (Fitbit Inc, San Francisco, CA, USA)	Wrist	Fitbit Flex: 17.36%

Diaz, 2016	N=13 (13 F) Age: 32.0 ± 9.2 y BMI: 24.2 ± 3.4 kg/m ²	paces and jogging. Subjects performed a treadmill protocol consisting of walking at slow, moderate and brisk paces and jogging.	Lab	IC – Ultima CPX (Medgraphics, Saint Paul, MN, USA)	Fitbit Flex (Fitbit Inc, San Francisco, CA, USA)	Wrist	Fitbit Flex: 30.27%
Dondzila, 2016	N=19 (5 F) Age: 24.6 ± 3.1 y BMI: 28.0 ± 3.8 kg/m ²	Subjects performed 5-minute stages of jogging on a treadmill at increasing velocity.	Lab	IC – Parvo TrueOne 2400 (Parvo Medics, East Sandy, UT, USA)	Fitbit Charge (Fitbit Inc, San Francisco, California, USA)	Wrist	Fitbit Charge: -13.01%
Dooley, 2017	N=62 (36 F) Age: 22.46 y BMI: 24.86 kg/m ²	Subjects performed 4 stages of treadmill exercise followed by a seated recovery period. The activity routine consisted of an unmeasured warm-up walking period and measured stages of slow, then brisk walking and jogging.	Lab	IC – Parvo TrueOne 2400 (Parvo Medics, East Sandy, UT, USA)	Apple watch (Apple Inc, Cupertino, CA, USA) Fitbit charge HR (Fitbit Inc, San Francisco, CA, USA) Garmin Forerunner 225 (Garmin ltd, Olathe, Kansas, USA)	Wrist	Apple watch: 64.55% Fitbit charge HR: 18.70% Garmin Forerunner 225: 44.23%
Drenowatz, 2011	N=20 (10 F) Age: 24.3 y BMI: N/A	Subjects performed three treadmill runs at 65, 75, and 85% VO _{2max} .	Lab	IC – Oxycon Mobile portable metabolic system (Erich Jaeger, Viasys Healthcare, Germany)	SenseWear Armband (HealthWe ar, Bodymedia, Pittsburgh, PA, USA)	Upper arm	SenseWear Armband: -32.80%
Erdogan, 2010	N=43 (27 F) Age: 34.9 ± 5.5 y	Subjects performed rowing exercises at 50% and	Lab	IC – COSMED K4b2 (COSMED, Rome, Italy)	SenseWear Armband (HealthWe ar, Bodymedia	Upper arm	SenseWear Armband: 5.23%

	BMI: 31.2 ± 3.7 kg/m ²	70% VO _{2max} on an ergometer.						Pittsburgh, PA, USA)
Fruin, 2010	Experiment 1: N=13 (0 F) Experiment 2: N=20 (10 F) Age: 20.2 ± 1 y BMI: N/A	Experiment 1: Subjects performed two resting and a cycle ergometer session at 60% VO _{2peak} . Experiment 2: Subjects completed a treadmill protocol of jogging, running and uphill running.	Lab	IC – SensorMedics Vmax 229 (SensorMedics Inc, Yorba Linda, CA, USA).	SenseWear Armband (HealthWear, Bodymedia, Pittsburgh, PA, USA)	Upper arm	SenseWear Armband: -1.76%	
Furlanetto, 2010	N=30 (15 F) Age: 68 ± 7 y BMI: 25 ± 3 kg/m ²	Subjects performed a walking protocol on a treadmill at three intensities.	Lab	IC – VO ₂₀₀₀ aerograph (Medgraphics, Saint Paul, MN, USA)	SenseWear Armband (HealthWear, Bodymedia, Pittsburgh, PA, USA)	Upper arm	SenseWear Armband: -6.99%	
Gastin, 2017	N=26 (12 F) Age: 21.3 ± 2.4 y BMI: 23.2 ± 2 kg/m ²	Subjects performed a protocol involving resting periods, walking, jogging, running or a sport-simulated circuit.	Lab	IC – MetaMax 3b (Cortex Biophysic, Leipzig, Germany)	SenseWear Armband (HealthWear, Bodymedia, Pittsburgh, PA, USA)	Upper arm	SenseWear Armband: -19.90%	
Heierman n, 2011	N=32 (19 F) Age: 68.6 y BMI: 26.4 kg/m ²	Subjects were required to rest.	Lab	IC – Vmax Spectra (SensorMedics Viasys Healthcare, Bilthoven, The Netherlands)	SenseWear Pro2 Armband (HealthWear, Bodymedia, Pittsburgh, PA, USA)	Upper arm	SenseWear Pro2 Armband: 10.80%	
Imboden, 2017	N=30 (15 F) Age: 49.2 ± 19.2 y BMI: 26.2 kg/m ²	Subjects performed a semi-structured activity protocol, performing ≥12 activities for	Lab	IC – COSMED K4b2 (COSMED, Rome, Italy)	Fitbit flex (Fitbit Inc, San Francisco, California, USA) Jawbone UP24	Wrist	Fitbit flex: -15.29% Jawbone UP24: -40.00%	

		subject-selected duration and pace. Activities were selected from a list of sedentary, household activities ambulatory and cycling activities.			(Jawbone, San Francisco, California, USA)		
Jakicic, 2004	N=40 (20 F) Age: 23.2 ± 3.8 y BMI: 23.8 ± 3.1 kg/m ²	Subjects performed 4 separate exercise protocols including treadmill walking, stair stepping, cycle ergometry, and arm ergometry.	Lab	IC – SensorMedics Vmax 229 (SensorMedics Inc, Yorba Linda, CA, USA).	SenseWear Armband (HealthWear, Bodymedia, Pittsburgh, PA, USA)	Upper arm	SenseWear Armband: - 11.76%
Johannsen, 2010	N=30 (15 F) Age: 38.2 ± 10.6 y BMI: 24 ± 3.4 kg/m ²	14-day comparison to DLW.	Field	DLW – 14 days	SenseWear Pro3 Armband (HealthWear, Bodymedia, Pittsburgh, PA, USA)	Upper arm	SenseWear Pro3 Armband: - 2.48%
Kim, 2015	N=52 (19 F) Age: 23.8 ± 5.2 BMI: N/A	Subjects performed 15 activities including resting, stair climbing, cycling, walking and jogging. Each activity was performed for 5	Lab	IC – Oxycon Mobile portable metabolic system (Erich Jaeger, Viasys Healthcare, Germany)	SenseWear Mini Armband (HealthWear, Bodymedia, Pittsburgh, PA, USA) Bodymedia Core (HealthWear, Bodymedia, Pittsburgh, PA, USA)	Upper arm	Bodymedia Core: 5.80%

King, 2004	N=21 (10 F) Age: 37.55 y	minutes, with 1-minute resting intervals. Subjects performed 10 minutes of treadmill walking and running at various speeds.	Lab	IC – TrueMax 2400 (Consentius Technologies, Sandy, UT, USA)	SenseWear Armband (HealthWear, Bodymedia, Pittsburgh, PA, USA)	Upper arm	SenseWear Armband: 20.33%
Koehler, 2011	N=14 (0 F) Age: 30.4 ± 6.2 y BMI: 23.2 ± 1.4 kg/m ²	7-day comparison to DLW.	Field	DLW – 7 days	SenseWear Pro3 Armband (HealthWear, Bodymedia, Pittsburgh, PA, USA)	Upper arm	SenseWear Pro3 Armband: -1.83%
Lee, 2011	N=46 (21 F) Age: 24.8 ± 5.6 y BMI: 24.3 ± 3.6 kg/m ²	Subjects completed 4-minute periods of standing, walking, jogging, and running.	Lab	IC – Parvo TrueOne 2400 (Parvo Medics, East Sandy, UT, USA)	ePulse Personal Fitness Assistant (ePulse) (Impact Sports Technologies, San Diego, CA, USA)	Forearm	ePulse Personal Fitness Assistant - 3.46%
Lee, 2014	N=60 (30 F) Age: 26.4 y BMI: 23.05 kg/m ²	Subjects performed 13 activities for 5 minutes. Activities were categorized into sedentary, treadmill walking, treadmill jogging and moderate-to-vigorous activities (ascending and descending stairs, stationary bike, elliptical exercise, Wii tennis play,	Lab	IC – Oxycon Mobile 5.0 (Erich Jaeger, Viasys Healthcare, Germany)	BodyMedia CORE (BodyMedia Inc., Pittsburgh, PA, USA) Jawbone UP (Jawbone, San Francisco, California, USA) Basis B1 Band (Basis Science Inc, San Francisco, CA, USA) Nike Fuel Band (Nike Inc.,	Upper arm Wrist	BodyMedia CORE: -5.31% Jawbone UP: -6.92% Basis B1 Band: -31.65% Nike Fuel Band: -1.91%

		and basketball).			Beaverton, OR, USA)		
Lopez, 2017 ¹	N=36 (16 F) Age: 37.7 ± 9.8 y BMI: 23.4 ± 2.8 kg/m ²	Subjects performed a structured protocol including rest, computer use, standing, slow walking, running, basketball and overground cycling.	Lab	IC – MetaMax 3x (Cortex Biophysic, Leipzig, Germany)	SenseWear Mini Armband (HealthWear, Bodymedia, Pittsburgh, PA, USA)	Upper arm	SenseWear Mini Armband: - 16.00%
Mackey, 2011	N=19 (8 F) Age: 82 ± 3.3 y BMI: 28.1 ± 3.8 kg/m ²	12.5-day comparison to DLW.	Field	DLW – 12.5 days	SenseWear Armband (HealthWear, Bodymedia, Pittsburgh, PA, USA)	Upper arm	SenseWear Armband: - 0.05%
Martien, 2015	N=60 (47 F) Age: 85.5 ± 5.5 y BMI: N/A	Subjects performed activity for 4 minutes and separated by 4 minutes seated rest. Activities included: Walking, rising and sitting in chairs positioned 5 meters apart and moving light objects.	Lab	IC – Oxycon Mobile portable metabolic system (Erich Jaeger, Viasys Healthcare, Germany)	SenseWear Mini Armband (HealthWear, Bodymedia, Pittsburgh, PA, USA)	Upper arm	SenseWear Mini Armband: - 12.00%

Maschac, 2013 ¹	N=19 (13 F) Age: 55.65 y BMI: 31.5 ± 3.6 kg/m ²	Subjects performed three walking sessions on a treadmill with different combinations of speed and incline.	Lab	IC – VO ₂₀₀₀ aerograph (Medgraphics, Saint Paul, MN, USA)	SenseWear Pro 3 Armband (HealthWear, Bodymedia, Pittsburgh, PA, USA)	Upper arm	SenseWear Pro 3 Armband: 50.69%
McMinn, 2013	N=19 (6 F) Age: 30 y BMI: 23.6 kg/m ²	Subjects completed 3 treadmill walking trials at self-selected slow, medium, and fast speeds.	Lab	IC – Ultima CPX (Medgraphics, Saint Paul, MN, USA)	Actigraph GT3X+ (Actigraph Inc, Pensacola, FL, USA)	Wrist	Actigraph GT3X+ : - 8.84%
Melanson, 2009	N=7 (3 F) Age: 31.8 ± 7.2 y BMI: 27.8 ± 7.9 kg/m ²	Subjects performed individualised protocols, including bench stepping and stationary cycling.	Lab	MC – 22.8 hours	LifeChek Calorie Sensor (LifeChek, LLC, Pittsburgh, PA, USA)	Wrist	LifeChek calorie sensor -4.87%
Mikulic, 2011	N=19 (11 F) Age: 28 ± 6 y BMI: 23 ± 3 kg/m ²	Subjects performed in-line skating exercises on a circular track at a self-selected pace.	Field	IC – COSMED K4b2 (COSMED, Rome, Italy)	SenseWear Pro 3 Armband (HealthWear, Bodymedia, Pittsburgh, PA, USA)	Upper arm	SenseWear Pro 3 Armband : - 73.33%
Montoye, 2017	N=32 (14 F) Age: 23.7 y BMI: 25.5 kg/m ²	Subjects completed 14 exercises, 11 in the laboratory including walking, jogging and cycling ergometry and 3 track exercises included self-paced walking at both a leisure and brisk pace for 200 meters and self-paced jogging for	Lab	IC – Parvo TrueOne 2400 (Parvo Medics, East Sandy, UT, USA)	Fitbit Charge HR (Fitbit Inc, San Francisco, California, USA)	Upper arm	Fitbit Charge HR: 7.59%

		400 meters. Each was 5 minutes in duration.					
Murakami , 2016	N=19 (10 F) Age: N/A BMI: N/A	1) 12.5-day comparison to DLW. 2) 24 hours in metabolic chamber where subjects where subjects were required to perform deskwork, watch television, housework, treadmill walking, and sleeping.	Lab/ Field	DLW – 12.5 days MC – 24 hours	Withings Pulse O2 (Withings, Issy-les- Moulineau x, France) Garmin vivofit (Garmin Ltd, Olathe, Kansas, USA) Fitbit Flex (Fitbit Inc, San Francisco, California, USA) Misfit Shine (Misfit, San Francisco, California, USA) Epson Pulsense (Epson, Suwa, Nagano Prefecture, Japan)	Wrist	Withings Pulse O2: -22.03% Garmin vivofit: - 20.55% Fitbit Flex: - 1.04% Misfit Shine: - 2.36% Epson Pulsense: - 4.28%
Nelson, 2016	N=30 (15 F) Age: 48.9 ± 19.4 y BMI: 26.3 ± 5.2 kg/m ²	Subjects performed a structured protocol consisting of sedentary, household, and ambulatory activities.	Lab	IC – COSMED K4b2 (COSMED, Rome, Italy)	Jawbone UP (Jawbone, San Francisco, California, USA) Fitbit Flex (Fitbit Inc, San Francisco, California, USA)	Wrist	Jawbone UP: - 2.12% Fitbit Flex: 12.74%
Papazoglu, 2006	N=29 Age: N/A BMI: N/A	Subjects performed a resting	Lab	IC – SensorMedics Vmax	SenseWear Pro 2 Armband	Wrist	SenseWear Pro 2

		protocol in a larger sample and 29 of the obese subjects participated in low intensity modes of exercise including cycle ergometry, stair stepping and treadmill walking.		229 (SensorMedics Inc, Yorba Linda, CA, USA)	(HealthWe ar, Bodymedia , Pittsburgh, PA, USA)		Armband: 21.54%
Price, 2017	N=14 (3 F) Age: 23 y BMI: 22.8 kg/m ²	Subjects walked on a treadmill at increasing velocities.	Lab	IC – Parvo TrueOne 2400 (Parvo Medics, East Sandy, UT, USA)	Jawbone UP (Jawbone, San Francisco, California, USA)	Upper arm	Jawbone UP: 56.91% Garmin vivofit: 18.16%
Reece, 2015	N=22 (11 F) Age: N/A BMI: N/A	Subjects performed a protocol including rest, sedentary activities and walking.	Lab	IC – Oxycon Mobile portable metabolic system (Erich Jaeger, Viasys Healthcare, Germany)	Garmin vivofit (Garmin Ltd, Olathe, Kansas, USA) SenseWear Mini Armband (HealthWe ar, Bodymedia , Pittsburgh, PA, USA)	Wrist	SenseWear Mini Armband: -3.79%
Reeve, 2014 ¹	N: 18 (7 F) Age: 22.6 y BMI: 22.9 kg/m ²	Subjects performed 2 resistance training sessions that included 9 different exercises. The weight lifted was 70% of 1 repetition max with 90-second rest intervals.	Lab	IC – COSMED K4b2 (COSMED, Rome, Italy)	BodyMedia CORE (BodyMedia Inc., Pittsburgh, PA, USA) SenseWear Mini Armband (HealthWe ar, Bodymedia , Pittsburgh, PA, USA)	Upper arm	BodyMedia CORE: 13.8% SenseWear Mini Armband: 23.7%

Rousset, 2015	Free-living: N=41 (20 F) Lab: N=49 (26 F) Age: N/A BMI: N/A	1) 10-day comparison to DLW. 2) 24 hours in metabolic chamber, which included eating, deskwork, watching television, housework, treadmill walking, and sleeping.	Lab/ Field	DLW – 12.5 days MC – 17 hours	SenseWear Pro 3 Armband (HealthWe ar, Bodymedia , Pittsburgh, PA, USA)	Upper arm	SenseWear Pro 3 Armband: -2.80%
Shcherbina, 2017 ¹	N=60 (31 F) Age: 38.5 y BMI: 23.65 kg/m ²	Subjects performed treadmill flat and incline running and cycle ergometry at low and moderate intensity.	Lab	IC – COSMED Quark CPNET (COSMED, Rome, Italy)	Apple watch (Apple Inc, Cupertino, CA, USA) Basis Peak (Basis Science Inc, San Francisco, CA, USA) Fitbit surge (Fitbit Inc, San Francisco, CA, USA) Microsoft band (Microsoft Corporation, Redmond, WA, USA) PulseOn (PulseOn Oy, Espoo Finland)	Wrist	Apple watch: -38.23% Basis Peak: -12.94% Fitbit Surge: -3.86% Microsoft Band -19.64% PulseOn: -24.47%
Slinde, 2013	N=62 (62 F) Age: 33.2 ± 4.2 y	7-day comparison to DLW	Field	DLW – 7 days	SenseWear Pro 2 Armband (HealthWe ar,	Wrist	SenseWear Pro 2 Armband: -2.90%

	BMI: 30 ± 2.8 kg/m ²				Bodymedia, Pittsburgh, PA, USA)		
Smith, 2012	N=30 (30 F) Age: 29.0 ± 4.3 y BMI: 24.1 ± 3.0 kg/m ²	Subjects performed a series of activities of daily living activities and treadmill walking at increasing intensities.	Lab	IC – Parvo TrueOne 2400 (Parvo Medics East Sandy, UT, USA)	SenseWear Mini Armband (HealthWear, Bodymedia, Pittsburgh, PA, USA) Algorithm v2.2	Upper arm	SenseWear Mini Armband: 18.43%
Stackpool, 2014	N=20 (10 F) Age: N/A BMI: N/A	Subjects performed treadmill walking, treadmill running, elliptical exercise and an agility drills.	Lab	IC – Oxycon pro Mobile portable metabolic system (Erich Jaeger, Viasys Healthcare, Germany)	Nike Fuel Band (Nike Inc, Beaverton, OR, USA) Jawbone UP (Jawbone, San Francisco, California, USA)	Upper arm	Nike Fuel Band: -3.99% Jawbone UP: 3.09%
					Bodymedia Core (HealthWear, Bodymedia, Pittsburgh, PA, USA)		
St-Onge, 2007	N=45 (32 F) Age: 35.1 ± 14 y BMI: 23.9 ± 4.0 kg/m ²	10-day comparison to DLW.	Field	DLW – 10 days	SenseWear Armband (HealthWear, Bodymedia, Pittsburgh, PA, USA)	Upper arm	SenseWear Armband: 4.70%
Tucker, 2015	N=24 (13 F) Age: 28.4 ± 7.8 y BMI: 23.8 ± 3.9 kg/m ²	Subjects performed two, 60-minute semi-structured routines consisting of sedentary/light-intensity, moderate-intensity and	Lab	IC – Oxycon Mobile portable metabolic system (Erich Jaeger, Viasys Healthcare, Germany)	Nike Fuel Band (Nike Inc., Beaverton, OR, USA) SenseWear Armband (HealthWear, Bodymedia	Upper arm	Nike Fuel Band: 1.22% SenseWear Armband: -2.10%

Van Helst, 2012	N=21 (10 F) Age: 29.3 ± 5.1 y	vigorous-intensity physical activity. Subjects performed a treadmill protocol involving slow and moderate walking, running slowly, vigorously running and periods of rest.	Lab	IC – Gas analyzer (Respiromics Novamatrix Medical SystemW inc, NICO 7300, Wallingford, USA)	Vivago (Vivago Wellness, Paris, France)	Wrist	Vivago: -8.02%
Van Hoye, 2014	N=44 (20 F) Age: 21.1 ± 1.4 y BMI: 21.8 ± 1.4 kg/m ²	Subjects performed an incremental running test on a treadmill.	Lab	IC – Metalyzer 3B (Cortex Biophysic, Leipzig, Germany)	SenseWear Pro 3 Armband (HealthWe ar, Bodymedia, Pittsburgh, PA, USA)	Upper arm	SenseWear Pro 3 Armband: -32.96%
Van Hoye, 2015	N=39 (18 F) Age: 21.1 ± 1.4 y BMI: 21.8 ± 1.4 kg/m ²	Subjects performed exercise consisting of 5 minutes standing followed by alternating walking and running at 35% and 65% VO _{2max} .	Lab	IC – Metalyzer 3B (Cortex Biophysic, Leipzig, Germany)	SenseWear Armband (HealthWe ar, Bodymedia, Pittsburgh, PA, USA) Algorithm v2.2 SenseWear Armband (HealthWe ar, Bodymedia, Pittsburgh, PA, USA) Algorithm v5.2	Upper arm	SenseWear Pro 3 Armband: -15.23%
Vernillo, 2015	N=20 (8 F) Age: 30.1 ± 7.2 y BMI: 22.1 ± 2.4 kg/m ²	Subjects performed randomized pole walking activities at a constant speed and a variety of gradients.	Lab	IC – COSMED Quark b2 (COSMED, Rome, Italy)	SenseWear Pro 3 Armband (HealthWe ar, Bodymedia, Pittsburgh, PA, USA)	Upper arm	SenseWear Pro 3 Armband: -9.76% SenseWear Mini Armband: -12.50%

					SenseWear Mini Armband (HealthWear, Bodymedia, Pittsburgh, PA, USA)		
Wahl, 2017	N=20 (10 F) Age: 25.2 y BMI: 22.8 kg/m ²	Subjects performed a running protocol consisting of four 5-minute stages of treadmill running at different velocities followed by a period of intermittent running and then a 2.4 km outdoor run.	Lab/Field	IC – Metalyzer 3B (Cortex Biophysic, Leipzig, Germany)	SenseWear Mini Armband (HealthWear, Bodymedia, Pittsburgh, PA, USA) Beurer AS80 (Beurer GmbH, Ulm, Germany) Polar Loop (Polar Electro, Kempele, Finland) Garmin vivofit (Garmin Ltd, Olathe, Kansas, USA) Garmin vivosmart (Garmin Ltd, Olathe, Kansas, USA) Garmin vivoactive (Garmin Ltd, Olathe, Kansas, USA) Garmin Forerunner 920XT (Garmin Ltd, Olathe, Kansas, USA)	Upper arm/Wrist	SenseWear Mini Armband: -21.27% Beurer AS80: -58.07% Polar Loop: 18.05% Garmin vivofit: -13.67% Garmin vivosmart: 5.98% Garmin vivoactive: 3.42% Garmin Forerunner 920XT: -21.02% Fitbit Charge: 3.58% Fitbit charge HR: 7.58% Withings Pulse O2: -15.98%

					Kansas, USA)		
					Fitbit Charge (Fitbit Inc, San Francisco, California, USA)		
					Fitbit charge HR (Fitbit Inc, San Francisco, California, USA)		
					Withings Pulse (Withings, Issy-les- Moulineau x, France)		
Wallen 2016	N=22 (11 F) Age: 24.9 y BMI: 24.3 kg/m ²	Subjects performed a protocol including treadmill exercise and cycling ergometry.	Lab	IC – Metalyzer 3B (Cortex Biophysic, Leipzig, Germany)	Apple watch (Apple Inc, Cupertino, California, USA)	Wrist	Apple watch: -75.71 Fitbit charge HR: -26.31% Samsung Gear S: -9.98% Mio Alpha: -53.19%
					Fitbit charge HR (Fitbit Inc, San Francisco, California, USA)		
					Samsung Gear S (Samsung Electronics Co, Ltd, Suwon, South Korea)		
					Mio Alpha (Mio Global, Canada)		
Woodman , 2017	N=28 (8 F) Age: 24.85 y BMI: 24.25 kg/m ²	Subjects performed a range of activities including: supine rest, household	Lab/ Field	IC – Oxycon Mobile portable metabolic system (Erich	Withings Pulse (Withings, Issy-les- Moulineau x, France)	Wrist	Withings Pulse: - 133.33% Basis Peak: 0.59%

tasks, treadmill walking, stair stepping, outdoor walking, cycling, and running at a self-selected pace. Seated rest, and ergometer cycling.

Jaeger, Viasys Healthcare, Germany)

Basis Peak (Basis Science Inc, San Francisco, CA, USA)

Garmin vivofit: -80.59%

Garmin vivofit (Garmin Ltd, Olathe, Kansas, USA)

980 Characteristics of studies meeting inclusion criteria of systematic review. Results represents the mean
 981 percentage error between device measurements and criterion measurements.

982 ¹Not included in meta-analysis.

983 Abbreviations: Female (F), body mass index (BMI), indirect calorimetry (IC), metabolic chamber (MC), doubly
 984 labelled water (DLW), Kilocalories (Kcal)

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Device	Price	Wear site	Device grade	Input setup data	Sensors	Output	Battery life	Number of comparisons in meta-analysis	Weighted percent error
Actical (Phillips Respironics Inc, Murrysville, PN, USA)	€678 (incl. software)/ €321 (unit)	Hip, ankle, wrist	Research	Age, H, W	Accelerometer: Triaxial Heart rate: Heat sensors:	Activity intensity Kcals, steps	194 days	1	
Actigraph GT3X+ (Actigraph Inc, Pensacola, FL, USA)	\$250	Hip, ankle, wrist	Research	Age, Gender, Race, H, W	Accelerometer: Triaxial Heart rate: Heat sensors:	Activity intensity Kcals, sleep, steps	31 days	1	-8.84%
Apple watch (Apple Inc, Cupertino, California, USA)	£249	Wrist	Commercial	Age, Gender, H, W	Accelerometer: Triaxial Heart rate: Yes Heat sensors:	Steps, distance tracking, Kcals, HR, minutes of brisk	18 Hours	4	-6.59%

Apple watch 2 (Apple Inc, Cupertino, California, USA)	£315	Wrist	Commercial	Age, Gender, H, W	Accelerometer: Triaxial Heart rate: Yes Heat sensors:	Steps, distance tracking, Kcals, HR, minutes of brisk activity	18 Hours	1	48.20%	
Basis b1 (Basis Science Inc, San Francisco, CA, USA)	£149	Wrist	Commercial	Age, Gender, H, W	Accelerometer: Triaxial Heart rate: Yes Heat sensors: Yes	Steps, distance, Kcals, HR, active minutes, sleep	5 days	1	-31.65%	
Basis Peak (Basis Science Inc, San Francisco, CA, USA)	£170	Wrist	Commercial	Age, Gender, H, W	Accelerometer: Triaxial Heart rate: Yes Heat sensors: Yes	Steps, distance, Kcals, HR, active minutes, sleep	5 days	1	0.59%	
Beurer AS80 (Beurer GmbH, Ulm, Germany)	£29.99	Wrist	Commercial	Age, Gender, H, W	Accelerometer: Triaxial Heart rate: Heat sensors:	Steps, distance, Kcals, active minutes, sleep	14 days	1	-58.07%	
BodyMedia CORE (BodyMedia Inc., Pittsburgh)	\$150	Upper left arm	Research (commercially available)	Age, Gender, H, W	Accelerometer: Triaxial Heart rate:	Steps, activity intensity,	14 days	4	-1.06%	

, PA, USA)					Heat sensors: Yes	Kcals, sleep				
Epson Pulsense (Epson, Suwa, Nagano Prefecture, Japan)	£79.99	Wrist	Commercial	Age, Gender, H, W, RHR	Accelerometer: Triaxial Heart rate: Yes Heat sensors:	Steps, distance, kcals, active minutes, HR, sleep	36 hours	1		-4.28%
ePulse Personal Fitness Assistant (ePulse) (Impact Sports Technologies, San Diego, CA, USA)	\$129.95	Forearm	Commercial	Age, Gender, H, W, RHR	Accelerometer: Triaxial Heart rate: Yes Heat sensors:	Kcals, HR		1		-3.46%
Fitbit blaze (Fitbit Inc, San Francisco, California, USA)	£134.99	Wrist	Commercial	Age, Gender, H, W	Accelerometer: Triaxial Heart rate: Yes Heat sensors: Triaxial accelerometer, altimeter, optical HR	Steps, distance, Kcals, active minutes, sleep, HR, steps	5 days	1		28.66%
Fitbit charge (Fitbit Inc, San Francisco, California, USA)	£109.99	Wrist	Commercial	Age, Gender, H, W	Accelerometer: Triaxial Heart rate: Heat sensors:	Steps, distance, Kcals, active minutes, sleep	5 days	2		-5.06%

Fitbit charge 2 (Fitbit Inc, San Francisco, California, USA)	£109.99	Wrist	Commercial	Age, Gender, H, W	Triaxial accelerometer, altimeter Accelerometer: Triaxial Heart rate: Yes Heat sensors:	Steps, distance, Kcals, active minutes, sleep, HR, steps	5 days	1	-30.97%
Fitbit charge HR (Fitbit Inc, San Francisco, California, USA)	£139.99	Wrist	Commercial	Age, Gender, H, W	Accelerometer: Triaxial Heart rate: Yes Heat sensors:	Steps, distance, Kcals, active minutes, sleep, HR, steps	5 days	6	1.3%
Fitbit Flex (Fitbit Inc, San Francisco, California, USA)	£79.99	Wrist	Commercial	Age, Gender, H, W	Accelerometer: Triaxial Heart rate: Heat sensors:	Steps, distance, Kcals, active minutes, sleep	5 days	5	8.22%
Fitbit Surge (Fitbit Inc, San Francisco, California, USA)	£289.99	Wrist	Commercial	Age, Gender, H, W	Accelerometer: Triaxial Heart rate: Yes Heat sensors:	Steps, distance, Kcals, active minutes, altimeter, GPS	5 days		

Garmin Forerunner 225 (Garmin Ltd, Olathe, Kansas, USA)	£199.99	Wrist	Commercial	Age, Gender, H, W, RHR, HRmax	Accelerometer: Triaxial Heart rate: Yes Heat sensors:	Steps, HR, distance, Kcal, active minutes, altimeter, GPS	7-10 Hours	1	44.23%
Garmin Forerunner 920XT (Garmin Ltd, Olathe, Kansas, USA)	£450	Wrist	Commercial	Age, Gender, H, W, RHR, HRmax	Accelerometer: Triaxial Heart rate: Heat sensors:	Steps, distance, Kcal, active minutes, altimeter, sleep, HR, GPS	3 Days	1	-21.02%
Garmin vivoactive (Garmin Ltd, Olathe, Kansas, USA)	£250	Wrist	Commercial	Age, Gender, H, W	Accelerometer: Triaxial Heart rate: Heat sensors:	Steps, distance, Kcal, active minutes, altimeter, sleep, GPS	7 Days	1	3.42%
Garmin vivofit (Garmin Ltd, Olathe, Kansas, USA)	£79.99	Wrist	Commercial	Age, Gender, H, W	Accelerometer: Triaxial Heart rate: Heat sensors:	Steps, distance, Kcal, active minutes, sleep	1 Year	5	-26.09%
Garmin Vivosmart (Garmin Ltd, Olathe, Kansas, USA)	£139.99	Wrist	Commercial	Age, Gender, H, W	Accelerometer: Triaxial Heart rate: Heat sensors:	Steps, distance, Kcal, active minutes, sleep	7 Days	1	5.98%

Garmin Vivosmart HR (Garmin Ltd, Olathe, Kansas, USA)	£129.99	Wrist	Commercial	Age, Gender, H, W	Accelerometer: Triaxial Heart rate: Yes Heat sensors:	Steps, distance, Kcal, HR, intensity minutes, sleep	7 Days	1	16.85%
Jawbone UP (Jawbone, San Francisco, CA, USA)	£99.99	Wrist	Commercial	Age, Gender, H, W	Accelerometer: Triaxial Heart rate: Heat sensors:	Distance (app), Kcal, Steps, sleep	10 days	4	10.90%
Jawbone UP24 (Jawbone, San Francisco, CA, USA)	£89.99	Wrist	Commercial	Age, Gender, H, W	Accelerometer: Triaxial Heart rate: Heat sensors:	Distance (app), Kcal, Steps, sleep	14 Days	3	-29.58%
LifeChek calorie sensor (LifeChek, LLC, Pittsburgh, PA, USA)		Upper right arm	Commercial		Accelerometer: Triaxial Heart rate: Heat sensors: Yes	Kcal		1	-4.87%
Microsoft band (Microsoft Corporation, Redmond, WA, USA)	£169.99	Wrist	Commercial	Age, Gender, H, W	Accelerometer: Triaxial Heart rate: Yes Heat sensors: Yes	Steps, distance, kcal, active minutes, sleep, HR, GPS	48 Hours	1	-49.15%

Mio Alpha (Mio Global, Canada)	£119.99	Wrist	Commercial	Age, Gender, H, W, HRMAX, RHR	Accelerometer: Heart rate: Yes Heat sensors:	Kcals, HR	24 Hours	1	-53.19%
Misfit Shine (Misfit, San Francisco, California, USA)	£99.99	Wrist	Commercial	Age, Gender, H, W	Accelerometer: Triaxial Heart rate: Heat sensors:	Steps, distance, Kcals, active minutes, sleep		1	-2.36%
Nike Fuel Band (Nike Inc, Beaverton, OR, USA)	£129.99	Wrist	Commercial	Age, Gender, H, W	Accelerometer: Triaxial Heart rate: Heat sensors:	Steps, distance, Kcals, active minutes, sleep	4 days	3	-0.48%
Polar Loop (Polar Electro, Kempele, Finland)	£49.99	Wrist	Commercial	Age, Gender, H, W	Accelerometer: Triaxial Heart rate: Heat sensors:	Steps, distance, Kcals, active minutes, sleep	12 days	1	18.05%
Polar: AW200 (Polar Electro Oy, Kempele, Finland)	€152 (watch +software)	Wrist	Commercial	Age, Gender, H, W	Accelerometer: Triaxial Heart rate: Heat sensors: Triaxial accelerometer	Steps, distance, Kcals, active minutes		1	-13.17%

Polar: AW360 (Polar Electro Oy, Kempele, Finland)	£149.99	Wrist	Commercial	Age, Gender, H, W	Accelerometer: Triaxial Heart rate: Yes Heat sensors:	Steps, distance, Kcal, active minutes, sleep, HR	12 Days	1	28.68%
Samsung Gear S (Samsung Electronics Co, Ltd, Suwon, South Korea)		Wrist	Commercial	Age, Gender, H, W	Accelerometer: Triaxial Heart rate: Yes Heat sensors:	Steps, distance, Kcal, active minutes, sleep, HR, GPS	2 Days	1	-9.98%
SenseWear Armband (HealthWear, Bodymedia, Pittsburgh, PA, USA)	€800 (device)+ €1597 (software)	Upper right arm	Research	Age, Gender H, W,	Accelerometer: Biaxial Heart rate: Heat sensors: Yes	Steps, activity intensity, Kcal, sleep	14 days	12	-4.31%
SenseWear Mini Armband (HealthWear, Bodymedia, Pittsburgh, PA, USA)		Upper left arm	Research	Age, Gender H, W, smoking status	Accelerometer: Triaxial Heart rate: Heat sensors: Yes	Steps, activity intensity, Kcal, sleep	28 days	9	-1.44%
SenseWear Pro 2 Armband (HealthWear, Bodymedia, Pittsburgh, PA, USA)		Upper right arm	Research	Age, Gender H, W, smoking status	Accelerometer: Biaxial Heart rate: Heat sensors: Yes	Steps, activity intensity, kcal, sleep	14 days	7	-7.54%

SenseWear Pro 3 Armband (HealthWear, Bodymedia, Pittsburgh, PA, USA)		Upper right arm	Research	Age, Gender H, W, smoking status	Accelerometer: Biaxial Heart rate: Heat sensors: Yes	Steps, activity intensity, kcal, sleep	14 days	12	-4.56%
TomTom Touch (TomTom, Amsterdam, the Netherlands)	£129.99	Wrist	Commercial	Age, Gender H, W	Accelerometer: Triaxial Heart rate: Yes Heat sensors:	Steps, distance, activity intensity, Kcal, sleep, HR,	5 Days	1	28.66%
Vivago (Vivago Wellness W, Paris, France).		Wrist	Commercial		Accelerometer: Triaxial Heart rate: Heat sensors:	Steps, activity intensity, Kcal, sleep		1	-8.02%
Withings Pulse (Withings, Issy-les-Moulineaux, France)	£39.99	Wrist, pocket or clip on	Commercial	Age, Gender H,	Accelerometer: Triaxial Heart rate: (non continuous) Heat sensors:	Steps, distance, Kcal, sleep	14 days	1	-133.33%
Withings Pulse 02 (Withings, Issy-les-Moulineaux, France)	£79.99	Wrist	Commercial	Age, Gender H, W	Accelerometer: Triaxial Heart rate: (non continuous) Heat sensors:	Steps, distance, activity intensity, Kcal, sleep,	14 days	2	-19.42%

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	Activity	Hedges g	SE	Varian ce	Moderator	Moderator	Moderator
Alsubheen, 2016	AMBULAT ION	-1.26	0.32	0.10	GVF	ACC	Commercia l
Bai, 2017	AMBULAT ION MODERAT E	-0.12	0.15	0.02	AW	ACC + HR	Commercia l
Bai, 2017	RUN MODERAT E	-0.28	0.13	0.02	AW	ACC + HR	Commercia l
Bai, 2017	SEDENTA RY	-0.38	0.17	0.03	AW	ACC + HR	Commercia l
Bai, 2017	AMBULAT ION MODERAT E	0.72	0.13	0.02	FCHR	ACC + HR	Commercia l
Bai, 2017	RUN MODERAT E	0.37	0.13	0.02	FCHR	ACC + HR	Commercia l
Bai, 2017	SEDENTA RY	-0.56	0.18	0.03	FCHR	ACC + HR	Commercia l
Benito, 2012	AEE LIGHT	-0.75	0.15	0.02	SWA p2	ACC + HS	Research
Benito, 2012	AEE MODERAT E	-0.83	0.15	0.02	SWA p2	ACC + HS	Research
Benito, 2012	AEE VIGOROUS	-0.75	0.15	0.02	SWA p2	ACC + HS	Research
Berntsen, 2010	AEE MODERAT E	0.27	0.16	0.03	SWA p2	ACC + HS	Research
Berntsen, 2010	AEE VERY VIGOROUS	-2.22	0.30	0.09	SWA p2	ACC + HS	Research
Berntsen, 2010	AEE VIGOROUS	-1.52	0.24	0.06	SWA p2	ACC + HS	Research
Berntsen, 2011	AEE LIGHT	-0.35	0.16	0.03	SWA p2	ACC + HS	Research
Berntsen, 2011	AEE MODERAT E	-0.48	0.24	0.06	SWA p2	ACC + HS	Research
Bhammar, 2016	AEE	0.53	0.13	0.02	SWAM	ACC + HS	Research
Bhammar, 2016	AEE 2	0.66	0.14	0.02	SWAM	ACC + HS	Research
Bhammar, 2016	AMBULAT ION MODERAT E	0.57	0.14	0.02	SWAM	ACC + HS	Research
Bhammar, 2016	AMBULAT ION MODERAT E 2	0.78	0.14	0.02	SWAM	ACC + HS	Research

Bhammar, 2016	AMBULATION VIGOROUS	0.31	0.13	0.02	SWAM	ACC + HS	Research
Bhammar, 2016	AMBULATION VIGOROUS 2	0.58	0.14	0.02	SWAM	ACC + HS	Research
Bhammar, 2016	BIKE LIGHT	-0.68	0.15	0.02	SWAM	ACC + HS	Research
Bhammar, 2016	BIKE LIGHT 2	0.06	0.14	0.02	SWAM	ACC + HS	Research
Bhammar, 2016	HOUSEHOLD	0.78	0.13	0.02	SWAM	ACC + HS	Research
Bhammar, 2016	HOUSEHOLD 2	1.22	0.15	0.02	SWAM	ACC + HS	Research
Bhammar, 2016	RUN MODERATE	0.82	0.14	0.02	SWAM	ACC + HS	Research
Bhammar, 2016	RUN MODERATE 2	0.51	0.13	0.02	SWAM	ACC + HS	Research
Bhammar, 2016	SEDENTARY	0.65	0.13	0.02	SWAM	ACC + HS	Research
Bhammar, 2016	SEDENTARY 2	0.00	0.11	0.01	SWAM	ACC + HS	Research
Bhammar, 2016	SWEEP	0.65	0.13	0.02	SWAM	ACC + HS	Research
Bhammar, 2016	SWEEP 2	1.10	0.15	0.02	SWAM	ACC + HS	Research
Boudreaux, 2018	AEE	1.46	0.26	0.07	AWS2	ACC + HR	Commercial
Boudreaux, 2018	BIKE	1.70	0.27	0.07	AWS2	ACC + HR	Commercial
Boudreaux, 2018	AEE	-0.12	0.18	0.03	FB	ACC + HR	Commercial
Boudreaux, 2018	BIKE	-0.91	0.18	0.03	FB	ACC + HR	Commercial
Boudreaux, 2018	AEE	-0.20	0.19	0.04	FC2	ACC + HR	Commercial
Boudreaux, 2018	BIKE	-1.05	0.22	0.05	FC2	ACC + HR	Commercial
Boudreaux, 2018	AEE	0.65	0.20	0.04	GVHR	ACC + HR	Commercial
Boudreaux, 2018	BIKE	0.15	0.15	0.02	GVHR	ACC + HR	Commercial
Boudreaux, 2018	AEE	1.01	0.23	0.05	PA360	ACC + HR	Commercial
Boudreaux, 2018	BIKE	0.60	0.18	0.03	PA360	ACC + HR	Commercial
Boudreaux, 2018	AEE	0.57	0.19	0.04	TT	ACC + HR	Commercial
Boudreaux, 2018	BIKE	0.76	0.19	0.04	TT	ACC + HR	Commercial
Brazeau, 2011	BIKE LIGHT	-0.30	0.15	0.02	SWA p3	ACC + HS	Research
Brazeau, 2014	AMBULATION MODERATE B	1.30	0.22	0.05	SWA p3	ACC + HS	Research

Brazeau, 2014	AMBULAT ION MODERAT E C	0.50	0.18	0.03	SWA p3	ACC + HS	Research
Brazeau, 2014	BIKE LIGHT B	-0.09	0.18	0.03	SWA p3	ACC + HS	Research
Brazeau, 2014	BIKE LIGHT C	-0.88	0.22	0.05	SWA p3	ACC + HS	Research
Brazeau, 2014	SEDENTA RY B	2.16	0.27	0.07	SWA p3	ACC + HS	Research
Brazeau, 2014	SEDENTA RY C	0.85	0.18	0.03	SWA p3	ACC + HS	Research
Brazeau, 2016	AMBULAT ION LIGHT	1.09	0.21	0.04	SWA p3	ACC + HS	Research
Brazeau, 2016	AMBULAT ION MODERAT E	1.09	0.21	0.04	SWA p3	ACC + HS	Research
Brazeau, 2016	AMBULAT ION VIGOROUS	0.56	0.17	0.03	SWA p3	ACC + HS	Research
Brazeau, 2016	BIKE	-0.83	0.21	0.04	SWA p3	ACC + HS	Research
Brazeau, 2016	SEDENTA RY	0.81	0.17	0.03	SWA p3	ACC + HS	Research
Brazeau, 2016	TEE	0.18	0.15	0.02	SWA p3	ACC + HS	Research
Brugniaux, 2010	AMBULAT ION VIGOROUS	-0.69	0.15	0.02	Polar AW200	ACC	Commercial
Calabro, 2014	AEE	0.14	0.11	0.01	SWA p3	ACC + HS	Research
Calabro, 2014	AEE	0.03	0.11	0.01	SWAM	ACC + HS	Research
Calabro, 2015	TEE	-0.04	0.12	0.02	SWAM	ACC + HS	Research
Casiraghi, 2013	BIKE LIGHT	-0.19	0.19	0.04	SWA	ACC + HS	Research
Choudhry, 2017	AMBULAT ION	0.33	0.17	0.03	AW	ACC + HR	Commercial
Choudhry, 2017	AMBULAT ION VIGOROUS	-0.05	0.17	0.03	AW	ACC + HR	Commercial
Choudhry, 2017	BIKE	-0.09	0.15	0.02	AW	ACC + HR	Commercial
Choudhry, 2017	COMPUTE R	1.25	0.25	0.06	AW	ACC + HR	Commercial
Choudhry, 2017	HOUSEHO LD	-1.32	0.26	0.07	AW	ACC + HR	Commercial
Choudhry, 2017	RUN	0.59	0.16	0.03	AW	ACC + HR	Commercial
Choudhry, 2017	STAIRS	-1.00	0.21	0.04	AW	ACC + HR	Commercial
Choudhry, 2017	SWEEP	-1.14	0.24	0.06	AW	ACC + HR	Commercial
Choudhry, 2017	AMBULAT ION	0.34	0.14	0.02	BMC	ACC + HS	Research
Choudhry, 2017	AMBULAT ION VIGOROUS	0.52	0.14	0.02	BMC	ACC + HS	Research

Choudhry, 2017	BIKE	-0.53	0.16	0.02	BMC	ACC + HS	Research
Choudhry, 2017	COMPUTER	0.55	0.13	0.02	BMC	ACC + HS	Research
Choudhry, 2017	HOUSEHOLD	0.75	0.14	0.02	BMC	ACC + HS	Research
Choudhry, 2017	RUN	0.33	0.13	0.02	BMC	ACC + HS	Research
Choudhry, 2017	STAIRS	-1.01	0.17	0.03	BMC	ACC + HS	Research
Choudhry, 2017	SWEEP	0.93	0.15	0.02	BMC	ACC + HS	Research
Choudhry, 2017	AMBULATION	1.78	0.21	0.04	FCHR	ACC + HR	Commercial
Choudhry, 2017	AMBULATION	1.60	0.19	0.04	FCHR	ACC + HR	Commercial
Choudhry, 2017	VIGOROUS BIKE	-2.15	0.35	0.12	FCHR	ACC + HR	Commercial
Choudhry, 2017	COMPUTER	-0.24	0.20	0.04	FCHR	ACC + HR	Commercial
Choudhry, 2017	HOUSEHOLD	-0.29	0.20	0.04	FCHR	ACC + HR	Commercial
Choudhry, 2017	RUN	0.92	0.17	0.03	FCHR	ACC + HR	Commercial
Choudhry, 2017	STAIRS	-0.01	0.13	0.02	FCHR	ACC + HR	Commercial
Choudhry, 2017	SWEEP	0.89	0.23	0.05	FCHR	ACC + HR	Commercial
Choudhry, 2017	AMBULATION	0.91	0.17	0.03	JU24	ACC	Commercial
Choudhry, 2017	AMBULATION	0.21	0.14	0.02	JU24	ACC	Commercial
Choudhry, 2017	VIGOROUS BIKE	-3.67	0.40	0.16	JU24	ACC	Commercial
Choudhry, 2017	COMPUTER	-0.05	0.14	0.02	JU24	ACC	Commercial
Choudhry, 2017	HOUSEHOLD	-1.72	0.22	0.05	JU24	ACC	Commercial
Choudhry, 2017	RUN	0.92	0.17	0.03	JU24	ACC	Commercial
Choudhry, 2017	STAIRS	-1.39	0.20	0.04	JU24	ACC	Commercial
Choudhry, 2017	SWEEP	-1.05	0.18	0.03	JU24	ACC	Commercial
Choudhry, 2017	AMBULATION	-1.50	0.25	0.06	MB	ACC + HR + HS	Commercial
Choudhry, 2017	AMBULATION	-2.15	0.31	0.10	MB	ACC + HR + HS	Commercial
Choudhry, 2017	VIGOROUS BIKE	-0.90	0.22	0.05	MB	ACC + HR + HS	Commercial
Choudhry, 2017	COMPUTER	0.13	0.19	0.04	MB	ACC + HR + HS	Commercial
Choudhry, 2017	HOUSEHOLD	-2.13	0.35	0.12	MB	ACC + HR + HS	Commercial

Choudhry, 2017	RUN	0.17	0.17	0.03	MB	ACC + HR + HS	Commercial
Choudhry, 2017	STAIRS	-3.42	0.50	0.25	MB	ACC + HR + HS	Commercial
Choudhry, 2017	SWEEP	-2.01	0.33	0.11	MB	ACC + HR + HS	Commercial
Colbert, 2011	TEE	-1.10	0.11	0.01	SWA p3	ACC + HS	Research
Correa, 2016	AEE	0.56	0.11	0.01	ACT	ACC	Research
Correa, 2016	AEE	-0.43	0.11	0.01	SWA	ACC + HS	Research
Diaz, 2015	AMBULATION LIGHT L	-0.08	0.13	0.02	FF	ACC	Commercial
Diaz, 2015	AMBULATION LIGHT R	-0.15	0.13	0.02	FF	ACC	Commercial
Diaz, 2015	AMBULATION MODERATE L	0.95	0.16	0.02	FF	ACC	Commercial
Diaz, 2015	AMBULATION MODERATE R	0.96	0.16	0.02	FF	ACC	Commercial
Diaz, 2015	AMBULATION VIGOROUS L	1.44	0.19	0.03	FF	ACC	Commercial
Diaz, 2015	AMBULATION VIGOROUS R	0.94	0.15	0.02	FF	ACC	Commercial
Diaz, 2015	RUN L	0.57	0.14	0.02	FF	ACC	Commercial
Diaz, 2015	RUN R	0.27	0.13	0.02	FF	ACC	Commercial
Diaz, 2016	AMBULATION LIGHT	1.36	0.24	0.06	FF	ACC	Commercial
Diaz, 2016	AMBULATION MODERATE	3.04	0.41	0.17	FF	ACC	Commercial
Diaz, 2016	AMBULATION VIGOROUS	1.07	0.21	0.04	FF	ACC	Commercial
Diaz, 2016	RUN	0.78	0.19	0.04	FF	ACC	Commercial
Dondzilla, 2016	AMBULATION LIGHT	0.84	0.21	0.04	FC	ACC	Commercial
Dondzilla, 2016	AMBULATION MODERATE	-0.52	0.19	0.04	FC	ACC	Commercial
Dondzilla, 2016	RUN LIGHT	-0.51	0.19	0.04	FC	ACC	Commercial
Dondzilla, 2016	RUN MODERATE	-1.13	0.24	0.06	FC	ACC	Commercial

Dooley, 2017	AMBULAT ION LIGHT	0.49	0.13	0.02	AW	ACC + HR	Commercia
Dooley, 2017	AMBULAT ION MODERAT E	0.29	0.12	0.02	AW	ACC + HR	Commercia
Dooley, 2017	AMBULAT ION VIGOROUS	0.27	0.12	0.01	AW	ACC + HR	Commercia
Dooley, 2017	SEDENTA RY	1.48	0.19	0.04	AW	ACC + HR	Commercia
Dooley, 2017	STAND	1.80	0.21	0.05	AW	ACC + HR	Commercia
Dooley, 2017	AMBULAT ION LIGHT	1.60	0.13	0.02	FCHR	ACC + HR	Commercia
Dooley, 2017	AMBULAT ION MODERAT E	1.13	0.11	0.01	FCHR	ACC + HR	Commercia
Dooley, 2017	AMBULAT ION VIGOROUS	0.08	0.09	0.01	FCHR	ACC + HR	Commercia
Dooley, 2017	SEDENTA RY	0.28	0.14	0.02	FCHR	ACC + HR	Commercia
Dooley, 2017	STAND	-0.39	0.14	0.02	FCHR	ACC + HR	Commercia
Dooley, 2017	AMBULAT ION LIGHT	0.72	0.10	0.01	GF225	ACC + HR	Commercia
Dooley, 2017	AMBULAT ION MODERAT E	1.27	0.12	0.01	GF225	ACC + HR	Commercia
Dooley, 2017	AMBULAT ION VIGOROUS	0.63	0.10	0.01	GF225	ACC + HR	Commercia
Dooley, 2017	SEDENTA RY	1.45	0.20	0.04	GF225	ACC + HR	Commercia
Dooley, 2017	STAND	1.14	0.18	0.03	GF225	ACC + HR	Commercia
Drenowatz, 2011	RUN	-2.21	0.30	0.09	SWA	ACC + HS	Research
Drenowatz, 2011	RUN LIGHT	-1.10	0.20	0.04	SWA	ACC + HS	Research
Drenowatz, 2011	RUN MODERAT E	-1.95	0.28	0.08	SWA	ACC + HS	Research
Drenowatz, 2011	RUN VIGOROUS	-2.41	0.32	0.10	SWA	ACC + HS	Research
Erdogan, 2010	AEE	0.08	0.11	0.01	SWA	ACC + HS	Research
Erdogan, 2010	AEE VIGOROUS	-0.01	0.11	0.01	SWA	ACC + HS	Research
Fruin, 2004	AMBULAT ION MODERAT E	0.23	0.16	0.03	SWA	ACC + HS	Research
Fruin, 2004	AMBULAT ION	-0.21	0.16	0.03	SWA	ACC + HS	Research

	UPHILL VIGOROUS						
Fruin, 2004	AMBULAT ION	0.51	0.17	0.03	SWA	ACC + HS	Research
	VIGOROUS						
Fruin, 2004	BIKE EARLY	-0.50	0.23	0.05	SWA	ACC + HS	Research
Fruin, 2004	BIKE LATE	-0.05	0.21	0.05	SWA	ACC + HS	Research
Fruin, 2004	BIKE MIDDLE	-0.27	0.22	0.05	SWA	ACC + HS	Research
Furlanetto, 2010	AMBULAT ION LIGHT	-0.19	0.13	0.02	SWA	ACC + HS	Research
Furlanetto, 2010	AMBULAT ION MODERAT E	-0.04	0.13	0.02	SWA	ACC + HS	Research
Furlanetto, 2010	AMBULAT ION VIGOROUS	-0.11	0.13	0.02	SWA	ACC + HS	Research
Gastin, 2017	AEE 1	-1.71	0.22	0.05	SWA	ACC + HS	Research
Gastin, 2017	AEE 2	-1.83	0.23	0.05	SWA	ACC + HS	Research
Gastin, 2017	AEE 3	-1.61	0.22	0.05	SWA	ACC + HS	Research
Gastin, 2017	AMBULAT ION	2.87	0.33	0.11	SWA	ACC + HS	Research
Gastin, 2017	RUN LIGHT	1.06	0.18	0.03	SWA	ACC + HS	Research
Gastin, 2017	RUN MODERAT E	-0.68	0.16	0.02	SWA	ACC + HS	Research
Heiermann, 2011	REST	0.76	0.13	0.02	SWA p2	ACC + HS	Research
Imboden, 2017	AEE	-0.65	0.12	0.02	FF	ACC	Commercial
Imboden, 2017	AEE	-1.30	0.15	0.02	JU24	ACC	Commercial
Jakicic, 2004	AEE	2.43	0.25	0.06	SWA	ACC + HS	Research
Jakicic, 2004	AEE 1	0.90	0.14	0.02	SWA	ACC + HS	Research
Jakicic, 2004	AMBULAT ION MODERAT E	1.92	0.22	0.05	SWA	ACC + HS	Research
Jakicic, 2004	AMBULAT ION UPHILL MODERAT E	-0.31	0.13	0.02	SWA	ACC + HS	Research
Jakicic, 2004	AMBULAT ION UPHILL VIGOROUS	-1.58	0.20	0.04	SWA	ACC + HS	Research
Jakicic, 2004	BIKE LIGHT	-0.41	0.14	0.02	SWA	ACC + HS	Research
Jakicic, 2004	BIKE MODERAT E	-2.28	0.27	0.07	SWA	ACC + HS	Research
Jakicic, 2004	STAIRS LIGHT	-0.18	0.14	0.02	SWA	ACC + HS	Research

Jakicic, 2004	STAIRS MODERATE	-1.46	0.20	0.04	SWA	ACC + HS	Research
Johannsen, 2010	TEE	-0.20	0.12	0.01	SWA p3	ACC + HS	Research
Johannsen, 2010	TEE	-0.04	0.12	0.01	SWAM	ACC + HS	Research
Kim, 2015	AEE LIGHT	0.16	0.10	0.01	BMC	ACC + HS	Research
Kim, 2015	AEE MODERATE	-0.09	0.10	0.01	BMC	ACC + HS	Research
Kim, 2015	AEE VIGOROUS	0.29	0.10	0.01	BMC	ACC + HS	Research
Kim, 2015	SEDENTARY	0.30	0.09	0.01	BMC	ACC + HS	Research
King, 2004	AMBULATION LIGHT M	4.46	0.78	0.60	SWA	ACC + HS	Research
King, 2004	AMBULATION M	1.56	0.34	0.11	SWA	ACC + HS	Research
King, 2004	AMBULATION MODERATE F	1.94	0.37	0.14	SWA	ACC + HS	Research
King, 2004	AMBULATION MODERATE M	3.76	0.66	0.44	SWA	ACC + HS	Research
King, 2004	AMBULATION VIGOROUS F	2.05	0.39	0.15	SWA	ACC + HS	Research
King, 2004	AMBULETION F	0.72	0.24	0.06	SWA	ACC + HS	Research
King, 2004	RUN LIGHT F	2.20	0.45	0.20	SWA	ACC + HS	Research
King, 2004	RUN LIGHT M	2.73	0.56	0.31	SWA	ACC + HS	Research
King, 2004	RUN MODERATE F	1.77	0.39	0.15	SWA	ACC + HS	Research
King, 2004	RUN MODERATE M	1.69	0.39	0.15	SWA	ACC + HS	Research
King, 2004	RUN VERY VIGOROUS F	0.71	0.26	0.07	SWA	ACC + HS	Research
King, 2004	RUN VERY VIGOROUS M	0.69	0.27	0.07	SWA	ACC + HS	Research
King, 2004	RUN VIGOROUS F	1.32	0.33	0.11	SWA	ACC + HS	Research
King, 2004	RUN VIGOROUS M	0.48	0.25	0.06	SWA	ACC + HS	Research
Koehler, 2011	TEE	-0.06	0.17	0.03	SWA p3	ACC + HS	Research
Lee, 2011	AMBULATION LIGHT	-0.74	0.20	0.04	EPUL	ACC + HR	Commercial

Lee, 2011	AMBULAT ION MODERAT E	0.04	0.18	0.03	EPUL	ACC + HR	Commercia
Lee, 2011	RUN LIGHT	0.01	0.18	0.03	EPUL	ACC + HR	Commercia
Lee, 2011	RUN MODERAT E	0.12	0.18	0.03	EPUL	ACC + HR	Commercia
Lee, 2011	SEDENTA RY	-0.31	0.18	0.03	EPUL	ACC + HR	Commercia
Lee, 2014	AEE	-1.37	0.18	0.03	BB1	ACC + HR + HS	Commercia
Lee, 2014	AEE	-0.28	0.10	0.01	BMC	ACC + HS	Research
Lee, 2014	AEE	-0.34	0.08	0.01	JU	ACC	Commercia
Lee, 2014	AEE	-0.10	0.09	0.01	NF	ACC	Commercia
Mackey, 2011	TEE	0.05	0.15	0.02	SWA	ACC + HS	Research
Mackey, 2011	TEE 6	-0.06	0.15	0.02	SWA	ACC + HS	Research
Martien, 2015	AEE	-0.25	0.11	0.01	SWAM	ACC + HS	Research
Martien, 2015	SEDENTA RY	-0.99	0.11	0.01	SWAM	ACC + HS	Research
McMinn, 2013	AMBULAT ION LIGHT	-1.55	0.25	0.06	AGT3X	ACC	Research
McMinn, 2013	AMBULAT ION MODERAT E	0.04	0.16	0.03	AGT3X	ACC	Research
McMinn, 2013	AMBULAT ION VIGOROUS	0.48	0.17	0.03	AGT3X	ACC	Research
Melanson, 2009	TEE	-0.05	0.35	0.12	LC	ACC + HS	Commercia
Mikulic, 2011	AEE	-1.95	0.28	0.08	SWA p3	ACC + HS	Research
Montoye, 2017	AMBULAT ION LIGHT	0.23	0.12	0.02	FCHR	ACC + HR	Commercia
Montoye, 2017	AMBULAT ION MODERAT E	0.22	0.12	0.02	FCHR	ACC + HR	Commercia
Montoye, 2017	AMBULAT ION UPHILL	0.42	0.13	0.02	FCHR	ACC + HR	Commercia
Montoye, 2017	AMBULAT ION UPHILL LIGHT	0.45	0.13	0.02	FCHR	ACC + HR	Commercia
Montoye, 2017	AMBULAT ION UPHILL MODERAT E	0.47	0.13	0.02	FCHR	ACC + HR	Commercia
Montoye, 2017	AMBULAT ION VIGOROUS	0.36	0.13	0.02	FCHR	ACC + HR	Commercia
Montoye, 2017	BIKE	0.43	0.18	0.03	FCHR	ACC + HR	Commercia

Montoye, 2017	RUN	0.52	0.15	0.02	FCHR	ACC + HR	Commercial
Montoye, 2017	SITTING	0.06	0.19	0.04	FCHR	ACC + HR	Commercial
Montoye, 2017	STAND	0.00	0.19	0.04	FCHR	ACC + HR	Commercial
Montoye, 2017	SUPINE	-0.23	0.19	0.04	FCHR	ACC + HR	Commercial
Murakami, 2016	TEE	-0.68	0.19	0.04	EP	ACC + HR	Commercial
Murakami, 2016	TEE MC	0.12	0.17	0.03	EP	ACC + HR	Commercial
Murakami, 2016	TEE	-0.48	0.15	0.02	FF	ACC	Commercial
Murakami, 2016	TEE MC	0.38	0.11	0.01	FF	ACC	Commercial
Murakami, 2016	TEE	-1.62	0.30	0.09	GVF	ACC	Commercial
Murakami, 2016	TEE MC	-0.83	0.23	0.05	GVF	ACC	Commercial
Murakami, 2016	TEE	-1.96	0.30	0.09	JU24	ACC	Commercial
Murakami, 2016	TEE MC	-0.95	0.21	0.04	JU24	ACC	Commercial
Murakami, 2016	TEE	-0.68	0.27	0.07	MS	ACC	Commercial
Murakami, 2016	TEE MC	0.40	0.25	0.06	MS	ACC	Commercial
Murakami, 2016	TEE	-1.68	0.23	0.05	WPO	ACC	Commercial
Murakami, 2016	TEE MC	-0.93	0.18	0.03	WPO	ACC	Commercial
Nelson, 2016	AMBULATION	0.92	0.14	0.02	FF	ACC	Commercial
Nelson, 2016	HOUSEHOLD	-0.27	0.14	0.02	FF	ACC	Commercial
Nelson, 2016	SEDENTARY	-0.33	0.14	0.02	FF	ACC	Commercial
Nelson, 2016	AMBULATION	0.48	0.12	0.01	JU	ACC	Commercial
Nelson, 2016	HOUSEHOLD	-1.30	0.24	0.06	JU	ACC	Commercial
Nelson, 2016	SEDENTARY	-0.39	0.18	0.03	JU	ACC	Commercial
Papazoglou, 2006	AMBULATION	0.82	0.19	0.04	SWA p2	ACC + HS	Research
Papazoglou, 2006	BIKE	0.54	0.17	0.03	SWA p2	ACC + HS	Research
Papazoglou, 2006	STAIRS	0.88	0.17	0.03	SWA p2	ACC + HS	Research
Price, 2017	AMBULATION	-1.15	0.29	0.09	GVF	ACC	Commercial
Price, 2017	RUN	-0.50	0.31	0.09	GVF	ACC	Commercial
Price, 2017	AMBULATION	-0.12	0.16	0.03	JU	ACC	Commercial

Price, 2017	RUN	0.48	0.19	0.04	JU	ACC	Commercial
Reece, 2015	AMBULAT ION	0.40	0.16	0.03	SWAM	ACC + HS	Research
Reece, 2015	COMPUTE R	-1.22	0.19	0.04	SWAM	ACC + HS	Research
Reece, 2015	SEDENTA RY	0.20	0.14	0.02	SWAM	ACC + HS	Research
Reece, 2015	SITTING	0.19	0.14	0.02	SWAM	ACC + HS	Research
Reece, 2015	STAND	-0.58	0.15	0.02	SWAM	ACC + HS	Research
Reece, 2015	STAND COMPUTE R	-1.10	0.18	0.03	SWAM	ACC + HS	Research
Rousset, 2015	TEE CC	-0.49	0.10	0.01	SWA p3	ACC + HS	Research
Rousset, 2015	TEE DLW	-0.05	0.10	0.01	SWA p3	ACC + HS	Research
Ryan, 2013	AMBULAT ION	0.82	0.17	0.03	SWA p2	ACC + HS	Research
Ryan, 2013	AMBULAT ION MODERAT E	-0.08	0.14	0.02	SWA p2	ACC + HS	Research
Ryan, 2013	AMBULAT ION UPHILL MODERAT E	-2.27	0.27	0.08	SWA p2	ACC + HS	Research
Ryan, 2013	RUN	-0.38	0.15	0.02	SWA p2	ACC + HS	Research
Ryan, 2013	SEDENTA RY	0.65	0.14	0.02	SWA p2	ACC + HS	Research
Slinde, 2013	TEE	0.20	0.09	0.01	SWA p2	ACC + HS	Research
Slinde, 2013	TEE 6	-0.67	0.09	0.01	SWA p2	ACC + HS	Research
Smith, 2012	AMBULAT ION	1.29	0.18	0.03	SWAM	ACC + HS	Research
Smith, 2012	AMBULAT ION 2	1.96	0.23	0.05	SWAM	ACC + HS	Research
Smith, 2012	AMBULAT ION LIGHT	1.43	0.19	0.04	SWAM	ACC + HS	Research
Smith, 2012	AMBULAT ION LIGHT 2	1.77	0.22	0.05	SWAM	ACC + HS	Research
Smith, 2012	AMBULAT ION MODERAT E	1.57	0.20	0.04	SWAM	ACC + HS	Research
Smith, 2012	AMBULAT ION MODERAT E 2	1.43	0.19	0.04	SWAM	ACC + HS	Research
Smith, 2012	AMBULAT ION UPHILL MODERAT E	-0.16	0.13	0.02	SWAM	ACC + HS	Research
Smith, 2012	AMBULAT ION UPHILL MODERAT E 2	-0.22	0.13	0.02	SWAM	ACC + HS	Research

Smith, 2012	COMPUTE R	0.37	0.12	0.02	SWAM	ACC + HS	Research
Smith, 2012	HOUSEHOLD	1.25	0.16	0.03	SWAM	ACC + HS	Research
Smith, 2012	HOUSEHOLD 2	1.78	0.20	0.04	SWAM	ACC + HS	Research
Smith, 2012	SEDENTARY 2	0.36	0.12	0.02	SWAM	ACC + HS	Research
Smith, 2012	SWEEP	0.52	0.13	0.02	SWAM	ACC + HS	Research
Smith, 2012	SWEEP 2	1.50	0.18	0.03	SWAM	ACC + HS	Research
Stackpool, 2015	AEE	-0.77	0.18	0.03	BMC	ACC + HS	Research
Stackpool, 2015	AEE 1	-1.29	0.22	0.05	BMC	ACC + HS	Research
Stackpool, 2015	AMBULATION	0.16	0.17	0.03	BMC	ACC + HS	Research
Stackpool, 2015	RUN	-0.65	0.18	0.03	BMC	ACC + HS	Research
Stackpool, 2015	AEE	-1.15	0.17	0.03	JU	ACC	Commercial
Stackpool, 2015	AEE 1	0.00	0.13	0.02	JU	ACC	Commercial
Stackpool, 2015	AMBULATION	0.56	0.15	0.02	JU	ACC	Commercial
Stackpool, 2015	RUN	0.77	0.19	0.03	JU	ACC	Commercial
Stackpool, 2015	AEE	-0.63	0.16	0.03	NF	ACC	Commercial
Stackpool, 2015	AEE 1	-1.14	0.19	0.04	NF	ACC	Commercial
Stackpool, 2015	AMBULATION	-0.08	0.15	0.02	NF	ACC	Commercial
Stackpool, 2015	RUN	0.63	0.17	0.03	NF	ACC	Commercial
St-Onge, 2007	TEE	0.27	0.10	0.01	SWA	ACC + HS	Research
Tucker, 2015	AEE	0.04	0.13	0.02	NF	ACC	Commercial
Tucker, 2015	AEE	-0.08	0.15	0.02	SWA	ACC + HS	Research
Van Helst, 2012	AMBULATION	-1.42	0.20	0.04	V	ACC	Commercial
Van Helst, 2012	RUN MODERATE	-0.22	0.14	0.02	V	ACC	Commercial
Van Helst, 2012	RUN VIGOROUS	-0.23	0.14	0.02	V	ACC	Commercial
Van Helst, 2012	SEDENTARY	0.00	0.14	0.02	V	ACC	Commercial
Van Hoye, 2014	AMBULATION F	0.14	0.16	0.03	SWA p3	ACC + HS	Research
Van Hoye, 2014	AMBULATION LIGHT F	-0.52	0.17	0.03	SWA p3	ACC + HS	Research
Van Hoye, 2014	AMBULATION LIGHT M	-1.06	0.19	0.04	SWA p3	ACC + HS	Research
Van Hoye, 2014	AMBULATION M	-0.08	0.15	0.02	SWA p3	ACC + HS	Research

Van Hoyer, 2014	AMBULAT ION MODERAT E F	-0.47	0.17	0.03	SWA p3	ACC + HS	Research
Van Hoyer, 2014	AMBULAT ION MODERAT E M	-1.13	0.20	0.04	SWA p3	ACC + HS	Research
Van Hoyer, 2014	AMBULAT ION VIGOROUS F	-0.62	0.18	0.03	SWA p3	ACC + HS	Research
Van Hoyer, 2014	AMBULAT ION VIGOROUS M	-1.16	0.20	0.04	SWA p3	ACC + HS	Research
Van Hoyer, 2014	RUN LIGHT F	-1.30	0.22	0.05	SWA p3	ACC + HS	Research
Van Hoyer, 2014	RUN LIGHT M	-1.35	0.21	0.04	SWA p3	ACC + HS	Research
Van Hoyer, 2014	RUN MODERAT E F	-1.88	0.28	0.08	SWA p3	ACC + HS	Research
Van Hoyer, 2014	RUN MODERAT E M	-2.02	0.26	0.07	SWA p3	ACC + HS	Research
Van Hoyer, 2014	RUN VERY LIGHT F	-0.76	0.18	0.03	SWA p3	ACC + HS	Research
Van Hoyer, 2014	RUN VERY LIGHT M	-0.93	0.18	0.03	SWA p3	ACC + HS	Research
Van Hoyer, 2014	RUN VERY VIGOROUS M	-3.08	0.41	0.17	SWA p3	ACC + HS	Research
Van Hoyer, 2014	RUN VIGOROUS F	-2.24	0.33	0.11	SWA p3	ACC + HS	Research
Van Hoyer, 2014	RUN VIGOROUS M	-3.03	0.36	0.13	SWA p3	ACC + HS	Research
Van Hoyer, 2015	AMBULAT ION LIGHT	-0.36	0.12	0.01	SWA p3	ACC + HS	Research
Van Hoyer, 2015	AMBULAT ION MODERAT E	-0.28	0.12	0.01	SWA p3	ACC + HS	Research
Van Hoyer, 2015	RUN	-1.04	0.14	0.02	SWA p3	ACC + HS	Research
Van Hoyer, 2015	RUN MODERAT E	-0.77	0.13	0.02	SWA p3	ACC + HS	Research
Van Hoyer, 2015	STAND	0.23	0.11	0.01	SWA p3	ACC + HS	Research
Van Hoyer, 2015	STAND 1	0.62	0.12	0.01	SWA p3	ACC + HS	Research
Vernillo, 2015	AMBULAT ION	0.78	0.19	0.03	SWA p3	ACC + HS	Research
Vernillo, 2015	AMBULAT ION DOWNHIL L LIGHT	2.01	0.29	0.08	SWA p3	ACC + HS	Research

Vernillo, 2015	AMBULAT ION DOWNHIL L MODERAT E	2.28	0.31	0.10	SWA p3	ACC + HS	Research
Vernillo, 2015	AMBULAT ION DOWNHIL L VIGOROUS	2.09	0.29	0.09	SWA p3	ACC + HS	Research
Vernillo, 2015	AMBULAT ION MODERAT E	0.30	0.16	0.03	SWA p3	ACC + HS	Research
Vernillo, 2015	AMBULAT ION UPHILL VIGOROUS	-2.33	0.32	0.10	SWA p3	ACC + HS	Research
Vernillo, 2015	AMBULAT ION UPHILL MODERAT E	-1.07	0.20	0.04	SWA p3	ACC + HS	Research
Vernillo, 2015	AMBULAT ION UPHILL VERY VIGOROUS	-2.83	0.37	0.14	SWA p3	ACC + HS	Research
Vernillo, 2015	SEDENTA RY	0.00	0.15	0.02	SWA p3	ACC + HS	Research
Vernillo, 2015	AMBULAT ION	0.56	0.17	0.03	SWAM	ACC + HS	Research
Vernillo, 2015	AMBULAT ION DOWNHIL L LIGHT	2.29	0.32	0.10	SWAM	ACC + HS	Research
Vernillo, 2015	AMBULAT ION DOWNHIL L MODERAT E	2.28	0.31	0.10	SWAM	ACC + HS	Research
Vernillo, 2015	AMBULAT ION DOWNHIL L VIGOROUS	2.27	0.31	0.10	SWAM	ACC + HS	Research
Vernillo, 2015	AMBULAT ION MODERAT E	0.00	0.16	0.03	SWAM	ACC + HS	Research
Vernillo, 2015	AMBULAT ION UPHILL MODERAT E	-1.49	0.24	0.06	SWAM	ACC + HS	Research
Vernillo, 2015	AMBULAT ION UPHILL	-2.78	0.37	0.13	SWAM	ACC + HS	Research

Vernillo, 2015	VERY VIGOROUS AMBULATION UPHILL	-2.46	0.33	0.11	SWAM	ACC + HS	Research
Vernillo, 2015	SEDENTARY	0.00	0.15	0.02	SWAM	ACC + HS	Research
Wahl, 2017	AMBULATION	0.14	0.22	0.05	BA	ACC	Commercial
Wahl, 2017	RUN INTERMITTENT	-1.13	0.28	0.08	BA	ACC	Commercial
Wahl, 2017	RUN LIGHT	0.40	0.22	0.05	BA	ACC	Commercial
Wahl, 2017	RUN MODERATE	-0.68	0.24	0.06	BA	ACC	Commercial
Wahl, 2017	RUN OUTDOOR	-0.83	0.34	0.12	BA	ACC	Commercial
Wahl, 2017	RUN VIGOROUS	-0.82	0.25	0.06	BA	ACC	Commercial
Wahl, 2017	AMBULATION	1.12	0.28	0.08	FC	ACC	Commercial
Wahl, 2017	RUN INTERMITTENT	0.00	0.21	0.05	FC	ACC	Commercial
Wahl, 2017	RUN LIGHT	0.58	0.23	0.05	FC	ACC	Commercial
Wahl, 2017	RUN MODERATE	0.37	0.22	0.05	FC	ACC	Commercial
Wahl, 2017	RUN OUTDOOR	-0.19	0.29	0.09	FC	ACC	Commercial
Wahl, 2017	RUN VIGOROUS	-0.05	0.21	0.05	FC	ACC	Commercial
Wahl, 2017	AMBULATION	0.78	0.25	0.06	FCHR	ACC + HR	Commercial
Wahl, 2017	RUN INTERMITTENT	0.42	0.22	0.05	FCHR	ACC + HR	Commercial
Wahl, 2017	RUN LIGHT	0.11	0.22	0.05	FCHR	ACC + HR	Commercial
Wahl, 2017	RUN MODERATE	0.33	0.22	0.05	FCHR	ACC + HR	Commercial
Wahl, 2017	RUN OUTDOOR	-0.35	0.30	0.09	FCHR	ACC + HR	Commercial
Wahl, 2017	RUN VIGOROUS	0.31	0.22	0.05	FCHR	ACC + HR	Commercial
Wahl, 2017	AMBULATION	-0.50	0.23	0.05	GF920XT	ACC	Commercial
Wahl, 2017	RUN INTERMITTENT	-0.16	0.22	0.05	GF920XT	ACC	Commercial
Wahl, 2017	RUN LIGHT	-0.32	0.22	0.05	GF920XT	ACC	Commercial

Wahl, 2017	RUN MODERAT E	-0.19	0.22	0.05	GF920XT	ACC	Commercia
Wahl, 2017	RUN OUTDOOR	-0.47	0.31	0.09	GF920XT	ACC	Commercia
Wahl, 2017	RUN VIGOROUS	-0.22	0.22	0.05	GF920XT	ACC	Commercia
Wahl, 2017	AMBULAT ION	-0.16	0.22	0.05	GVA	ACC	Commercia
Wahl, 2017	RUN INTERMIT TENT	-0.01	0.21	0.05	GVA	ACC	Commercia
Wahl, 2017	RUN LIGHT	0.46	0.23	0.05	GVA	ACC	Commercia
Wahl, 2017	RUN MODERAT E	0.95	0.26	0.07	GVA	ACC	Commercia
Wahl, 2017	RUN OUTDOOR	-0.25	0.29	0.09	GVA	ACC	Commercia
Wahl, 2017	RUN VIGOROUS	0.15	0.22	0.05	GVA	ACC	Commercia
Wahl, 2017	AMBULAT ION	-0.13	0.22	0.05	GVF	ACC	Commercia
Wahl, 2017	RUN INTERMIT TENT	-0.50	0.23	0.05	GVF	ACC	Commercia
Wahl, 2017	RUN LIGHT	0.22	0.22	0.05	GVF	ACC	Commercia
Wahl, 2017	RUN MODERAT E	0.15	0.22	0.05	GVF	ACC	Commercia
Wahl, 2017	RUN OUTDOOR	-0.90	0.35	0.12	GVF	ACC	Commercia
Wahl, 2017	RUN VIGOROUS	-0.29	0.22	0.05	GVF	ACC	Commercia
Wahl, 2017	AMBULAT ION	-0.10	0.22	0.05	GVS	ACC	Commercia
Wahl, 2017	RUN INTERMIT TENT	0.03	0.21	0.05	GVS	ACC	Commercia
Wahl, 2017	RUN LIGHT	0.37	0.22	0.05	GVS	ACC	Commercia
Wahl, 2017	RUN MODERAT E	0.39	0.22	0.05	GVS	ACC	Commercia
Wahl, 2017	RUN OUTDOOR	-0.05	0.29	0.08	GVS	ACC	Commercia
Wahl, 2017	RUN VIGOROUS	0.14	0.22	0.05	GVS	ACC	Commercia
Wahl, 2017	AMBULAT ION	0.36	0.22	0.05	PL	ACC	Commercia
Wahl, 2017	RUN INTERMIT TENT	0.02	0.21	0.05	PL	ACC	Commercia
Wahl, 2017	RUN LIGHT	0.27	0.22	0.05	PL	ACC	Commercia

Wahl, 2017	RUN MODERATE	0.34	0.22	0.05	PL	ACC	Commercial
Wahl, 2017	RUN OUTDOOR	0.15	0.22	0.05	PL	ACC	Commercial
Wahl, 2017	RUN VIGOROUS	0.25	0.22	0.05	PL	ACC	Commercial
Wahl, 2017	AMBULATION	-0.11	0.22	0.05	SWAM	ACC + HS	Research
Wahl, 2017	RUN INTERMITTENT	-0.43	0.23	0.05	SWAM	ACC + HS	Research
Wahl, 2017	RUN LIGHT	-0.10	0.22	0.05	SWAM	ACC + HS	Research
Wahl, 2017	RUN MODERATE	-0.59	0.23	0.05	SWAM	ACC + HS	Research
Wahl, 2017	RUN OUTDOOR	-0.70	0.33	0.11	SWAM	ACC + HS	Research
Wahl, 2017	RUN VIGOROUS	-0.85	0.25	0.06	SWAM	ACC + HS	Research
Wahl, 2017	AMBULATION	-1.12	0.38	0.15	WPO	ACC	Commercial
Wahl, 2017	RUN INTERMITTENT	-1.86	0.51	0.26	WPO	ACC	Commercial
Wahl, 2017	RUN LIGHT	0.12	0.29	0.08	WPO	ACC	Commercial
Wahl, 2017	RUN MODERATE	0.04	0.29	0.08	WPO	ACC	Commercial
Wahl, 2017	RUN OUTDOOR	-0.15	0.29	0.08	WPO	ACC	Commercial
Wahl, 2017	RUN VIGOROUS	-0.27	0.30	0.09	WPO	ACC	Commercial
Wallen 2016	AEE	-2.44	0.27	0.07	AW	ACC + HR	Commercial
Wallen 2016	AEE	-0.80	0.16	0.03	FCHR	ACC + HR	Commercial
Wallen 2016	AEE	-1.19	0.28	0.08	MA	HR	Commercial
Wallen 2016	AEE	-0.54	0.16	0.03	SG	ACC + HR	Commercial
Woodman, 2017	AMBULATION	2.25	0.34	0.12	BP	ACC + HR + HS	Commercial
Woodman, 2017	AMBULATION UPHILL MODERATE	0.73	0.20	0.04	BP	ACC + HR + HS	Commercial
Woodman, 2017	BIKE LIGHT	-0.45	0.20	0.04	BP	ACC + HR + HS	Commercial
Woodman, 2017	BIKE MODERATE	-1.01	0.23	0.05	BP	ACC + HR + HS	Commercial
Woodman, 2017	COMPUTER	-0.37	0.20	0.04	BP	ACC + HR + HS	Commercial

Woodman, 2017	HOUSEHOLD	-0.78	0.22	0.05	BP	ACC + HR + HS	Commercial
Woodman, 2017	RUN	0.20	0.16	0.03	BP	ACC + HR + HS	Commercial
Woodman, 2017	SEATED	0.03	0.19	0.04	BP	ACC + HR + HS	Commercial
Woodman, 2017	SEDENTARY	-0.97	0.23	0.05	BP	ACC + HR + HS	Commercial
Woodman, 2017	STAIRS	0.49	0.19	0.04	BP	ACC + HR + HS	Commercial
Woodman, 2017	SWEEP	-1.80	0.31	0.10	BP	ACC + HR + HS	Commercial
Woodman, 2017	AMBULATION	-1.36	0.23	0.05	GVF	ACC	Commercial
Woodman, 2017	AMBULATION UPHILL MODERATE	-2.84	0.38	0.14	GVF	ACC	Commercial
Woodman, 2017	BIKE LIGHT	-6.59	0.84	0.71	GVF	ACC	Commercial
Woodman, 2017	BIKE MODERATE	-0.80	0.19	0.04	GVF	ACC	Commercial
Woodman, 2017	COMPUTER	-0.27	0.17	0.03	GVF	ACC	Commercial
Woodman, 2017	HOUSEHOLD	-2.31	0.32	0.10	GVF	ACC	Commercial
Woodman, 2017	RUN	-0.98	0.28	0.08	GVF	ACC	Commercial
Woodman, 2017	SEATED	-1.21	0.22	0.05	GVF	ACC	Commercial
Woodman, 2017	SEDENTARY	-0.32	0.17	0.03	GVF	ACC	Commercial
Woodman, 2017	STAIRS	-4.16	0.53	0.28	GVF	ACC	Commercial
Woodman, 2017	SWEEP	-2.18	0.31	0.10	GVF	ACC	Commercial
Woodman, 2017	AMBULATION	-1.88	0.24	0.06	WP	ACC	Commercial
Woodman, 2017	AMBULATION UPHILL MODERATE	-2.80	0.32	0.10	WP	ACC	Commercial
Woodman, 2017	BIKE LIGHT	-5.53	0.61	0.38	WP	ACC	Commercial
Woodman, 2017	BIKE MODERATE	-2.25	0.28	0.08	WP	ACC	Commercial
Woodman, 2017	COMPUTER	1.94	0.25	0.06	WP	ACC	Commercial
Woodman, 2017	HOUSEHOLD	-0.83	0.17	0.03	WP	ACC	Commercial
Woodman, 2017	RUN	-2.37	0.30	0.09	WP	ACC	Commercial
Woodman, 2017	SEATED	-1.41	0.20	0.04	WP	ACC	Commercial

Woodman, 2017	SEDENTARY	-1.05	0.18	0.03	WP	ACC	Commercial
Woodman, 2017	STAIRS	-3.47	0.39	0.15	WP	ACC	Commercial
Woodman, 2017	SWEEP	-1.76	0.23	0.05	WP	ACC	Commercial

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991 S7:

Overall activities	n	Heterogeneity		Effect size			Publication bias				
		I-squared (between studies)	P-value	Hedges' g (95% CI)	Lower Limit	Upper limit	P-value	Egger's intercept	Lower limit	Upper limit	P-Value
ACT	1.00	0.00	1.00	0.56	-0.46	1.58	0.28				
AGT3X	1.00	0.00	1.00	-0.35	-1.42	0.73	0.53				
AW	4.00	97.30	0.00	-0.43	-0.97	0.10	0.11	-19.41	-65.76	26.94	0.21
AWS2	1.00	0.00	1.00	1.58	0.45	2.70	0.01				
BA	1.00	0.00	1.00	-0.49	-1.61	0.64	0.40				
BB1	1.00	0.00	1.00	-1.37	-2.42	-0.31	0.01				
BMC	4.00	87.47	0.00	-0.12	-0.64	0.40	0.65	-2.60	-30.95	25.74	0.73
BP	1.00	0.00	1.00	-0.15	-1.25	0.94	0.78				
EP	1.00	0.00	1.00	-0.28	-1.34	0.78	0.60				
EPUL	1.00	0.00	1.00	-0.18	-1.24	0.88	0.74				
FB	1.00	0.00	1.00	-0.51	-1.57	0.54	0.34				
FC	2.00	74.54	0.05	-0.02	-0.79	0.75	0.95				
FC2	1.00	0.00	1.00	-0.62	-1.70	0.45	0.26				
FCHR	6.00	89.06	0.00	0.13	-0.31	0.56	0.57	-2.32	-20.26	15.62	0.74
FF	5.00	94.80	0.00	0.27	-0.20	0.74	0.26	13.81	-4.45	32.07	0.09
GF225	1.00	0.00	1.00	1.04	0.00	2.08	0.05				
GF920XT	1.00	0.00	1.00	-0.31	-1.41	0.79	0.58				

GVA	1.00	0.00	1.0	0.19	-	1.2	0.7					
			0		0.91	9	4					
GVF	5.00	79.33	0.0	-1.09	-	-	0.0	-11.66	-	1.42	0.06	
			0		1.60	0.5	0		24.7			
						7			5			
GVHR	1.00	0.00	1.0	0.40	-	1.4	0.4					
			0		0.66	6	6					
GVS	1.00	0.00	1.0	0.13	-	1.2	0.8					
			0		0.97	3	1					
JU	4.00	73.04	0.0	-0.13	-	0.3	0.6	2.81	-7.67	13.2	0.37	
			1		0.66	9	2			8		
JU24	3.00	66.91	0.0	-1.16	-	-	0.0	1.15	-	74.0	0.87	
			5		1.78	0.5	0		71.7	5		
						4			5			
LC	1.00	0.00	1.0	-0.05	-	1.1	0.9					
			0		1.26	5	3					
MA	1.00	0.00	1.0	-1.19	-	-	0.0					
			0		2.34	0.0	4					
						5						
MB	1.00	0.00	1.0	-1.48	-	-	0.0					
			0		2.64	0.3	1					
						1						
MS	1.00	0.00	1.0	-0.14	-	0.9	0.8					
			0		1.26	8	0					
NF	3.00	25.44	0.2	-0.12	-	0.4	0.6	-1.04	-	41.0	0.80	
			6		0.72	8	9		43.0	0		
									9			
PA360	1.00	0.00	1.0	0.80	-	1.8	0.1					
			0		0.27	8	4					
PL	1.00	0.00	1.0	0.23	-	1.3	0.6					
			0		0.85	2	7					
Polar AW200	1.00	0.00	1.0	-0.69	-	0.3	0.1					
			0		1.73	5	9					
SG	1.00	0.00	1.0	-0.54	-	0.5	0.3					
			0		1.59	1	1					
SWA	12.0	87.57	0.0	-0.12	-	0.1	0.4	-1.11	-6.72	4.49	0.67	
	0		0		0.43	9	5					
SWA p2	7.00	94.47	0.0	-0.17	-	0.2	0.3	-2.05	-	12.2	0.73	
			0		0.57	2	9		16.3	6		
									7			
SWA p3	12.0	93.03	0.0	-0.32	-	-	0.0	-0.49	-8.81	7.82	0.89	
	0		0		0.62	0.0	4					
						1						
SWAM	9.00	91.19	0.0	0.02	-	0.3	0.9	2.30	-8.15	12.7	0.61	
			0		0.33	7	0			6		
TT	1.00	0.00	1.0	0.67	-	1.7	0.2					
			0		0.40	3	2					
V	1.00	0.00	1.0	-0.47	-	0.5	0.3					
			0		1.51	7	8					
WP	1.00	0.00	1.0	-1.95	-	-	0.0					
			0		3.12	0.7	0					
						8						
WPO	2.00	71.58	0.0	-0.97	-	-	0.0					
			6		1.77	0.1	2					
						6						
Between			0.0									
			0									

AEE

	n	I-squared (between studies)	P- val ue	Effect size (Hedg es' g) (95% CI)	Lo wer Lim it	Up per limi t	P- val ue				
Overall	104.00	92.18		-0.23	-0.44	-0.03	0.03				
ACT	1.00	0.00	1.00	0.56	-0.61	1.73	0.35				
AW	1.00	0.00	1.00	-2.44	3.71	1.18	0.00				
AWS2	1.00	0.00	1.00	1.46	0.21	2.71	0.02				
BB1	1.00	0.00	1.00	-1.37	2.57	0.17	0.03				
BMC	3.00	92.83	0.00	-0.38	-1.06	0.30	0.28	-8.62	-94.18	76.94	0.42
FB	1.00	0.00	1.00	-0.12	-1.32	1.08	0.85				
FC2	1.00	0.00	1.00	-0.20	-1.40	1.01	0.75				
FCHR	1.00	0.00	1.00	-0.80	1.99	0.39	0.19				
FF	1.00	0.00	1.00	-0.65	1.82	0.52	0.28				
GVHR	1.00	0.00	1.00	0.65	-0.56	1.86	0.29				
JU	2.00	47.50	0.17	-0.46	1.28	0.37	0.28				
JU24	1.00	0.00	1.00	-1.30	2.48	0.12	0.03				
MA	1.00	0.00	1.00	-1.19	2.47	0.08	0.07				
NF	3.00	89.88	0.00	-0.31	-0.99	0.37	0.38	-5.94	-91.55	99.65	0.53
PA360	1.00	0.00	1.00	1.01	-0.22	2.24	0.11				
SG	1.00	0.00	1.00	-0.54	-1.73	0.65	0.38				
SWA	5.00	97.12	0.00	-0.10	-0.63	0.43	0.71	-12.22	-233.83	209.39	0.61

Ambulation and stairs

SWA p2	3.00	64.19	0.06	-0.78	-1.48	-0.08	0.03	-1.93	-90.60	86.75	0.82
SWA p3	2.00	97.87	0.00	-0.81	-1.67	0.05	0.06				
SWAM	3.00	91.43	0.00	0.12	-0.55	0.80	0.72	34.56	-115.76	184.88	0.21
TT	1.00	0.00	1.00	0.57	-0.64	1.77	0.36				
Between			0.00								
Overall	35.00	94.96		-0.34	-0.71	0.04	0.08				
	n	I-squared (between studies)	P-value	Effect size (Hedges' g) (95% CI)	Lower Limit	Upper Limit	P-value				
AGT3X	1.00	0.00	1.00	-0.35	-1.50	0.81	0.56				
AW	3.00	78.96	0.01	0.00	-0.65	0.65	1.00	-10.13	-68.37	48.11	0.27
BA	1.00	0.00	1.00	0.14	-1.02	1.31	0.81				
BMC	2.00	0.00	0.36	0.05	-0.75	0.85	0.90				
BP	1.00	0.00	1.00	1.15	-0.04	2.35	0.06				
EPUL	1.00	0.00	1.00	-0.35	-1.50	0.80	0.55				
FC	2.00	87.37	0.00	0.61	-0.23	1.45	0.15				
FCHR	5.00	76.12	0.00	0.78	0.28	1.29	0.00	1.63	-11.54	14.79	0.72
FF	3.00	82.83	0.00	1.10	0.43	1.77	0.00	5.87	-37.18	48.93	0.33
GF225	1.00	0.00	1.00	0.87	-0.24	1.98	0.12				
GF920XT	1.00	0.00	1.00	-0.50	-1.68	0.67	0.40				
GVA	1.00	0.00	1.00	-0.16	-1.33	1.00	0.78				
GVF	4.00	91.90	0.00	-1.24	-1.86	0.62	0.00	-13.76	-19.72	-7.80	0.01
GVS	1.00	0.00	1.00	-0.10	-1.26	1.07	0.87				

JU	3.00	83.19	0.0 0	0.31	- 0.34	0.9 5	0.3 5	-9.01	- 165. 97	147. 94	0.60
JU24	1.00	0.00	1.0 0	-0.09	- 1.23	1.0 4	0.8 7				
MB	1.00	0.00	1.0 0	-2.36	- 3.67	- 1.0	0.0 0				
NF	1.00	0.00	1.0 0	-0.08	- 1.21	1.0 4	0.8 8				
PL	1.00	0.00	1.0 0	0.36	- 0.81	1.5 3	0.5 4				
Polar AW200	1.00	0.00	1.0 0	-0.69	- 1.81	0.4 3	0.2 3				
SWA	5.00	95.95	0.0 0	0.79	0.25	1.3 3	0.0 0	9.82	1.24	20.8 8	0.07
SWA p2	2.00	96.06	0.0 0	0.18	- 0.63	0.9 9	0.6 7				
SWA p3	5.00	93.40	0.0 0	0.20	- 0.32	0.7 1	0.4 6	6.93	- 13.2 5	27.1 1	0.35
SWAM	5.00	81.80	0.0 0	0.43	- 0.09	0.9 4	0.1 0	-3.29	- 18.4 0	11.8 1	0.54
V	1.00	0.00	1.0 0	-1.42	- 2.58	- 0.2	0.0 2				
WP	1.00	0.00	1.0 0	-2.72	- 3.98	- 1.4	0.0 0				
WPO	1.00	0.00	1.0 0	-1.12	- 2.44	0.2 0	0.1 0				
Between			0.0 0								
Overall	55.0 0	93.74		-0.09	- 0.45	0.2 7	0.6 2				
	n	I-squared (between studies)	P- val ue	Effect size (Hedg es' g) (95% CI)	Lo wer Lim it	Up per limi t	P- val ue				
AW	1.00	0.00	1.0 0	-0.09	- 1.54	1.3 5	0.9 0				
AWS2	1.00	0.00	1.0 0	1.70	0.18	3.2 1	0.0 3				
BMC	1.00	0.00	1.0 0	-0.53	- 1.98	0.9 2	0.4 7				
BP	1.00	0.00	1.0 0	-0.73	- 2.21	0.7 4	0.3 3				
FB	1.00	0.00	1.0 0	-0.91	- 2.36	0.5 5	0.2 2				
FC2	1.00	0.00	1.0 0	-1.05	- 2.53	0.4 3	0.1 6				
FCHR	2.00	97.72	0.0 0	-0.76	- 1.83	0.3 1	0.1 6				

Cycling

GF920X	1.00	0.00	1.0	-0.27	-	0.9	0.6					
T			0		1.53	9	7					
GVA	1.00	0.00	1.0	0.26	-	1.5	0.6					
			0		1.01	3	9					
GVF	3.00	46.39	0.1	-0.58	-	0.1	0.1	-5.20	-	161.	0.76	
			5		1.33	7	3		171.	34		
									72			
GVS	1.00	0.00	1.0	0.18	-	1.4	0.7					
			0		1.08	4	8					
JU	2.00	15.55	0.2	0.63	-	1.5	0.1					
			8		0.24	0	6					
JU24	1.00	0.00	1.0	0.92	-	2.1	0.1					
			0		0.30	3	4					
MB	1.00	0.00	1.0	0.17	-	1.3	0.7					
			0		1.04	9	8					
NF	1.00	0.00	1.0	0.63	-	1.8	0.3					
			0		0.59	4	1					
PL	1.00	0.00	1.0	0.21	-	1.4	0.7					
			0		1.04	5	4					
SWA	3.00	96.79	0.0	-0.14	-	0.6	0.7	-0.73	-	176.	0.97	
			0		0.89	0	0		178.	57		
									03			
SWA p2	1.00	0.00	1.0	-0.38	-	0.8	0.5					
			0		1.59	3	4					
SWA p3	2.00	88.85	0.0	-1.34	-	-	0.0					
			0		2.22	0.4	0					
						6						
SWAM	2.00	94.20	0.0	0.10	-	0.9	0.8					
			0		0.77	8	2					
V	1.00	0.00	1.0	-0.23	-	0.9	0.7					
			0		1.43	8	1					
WP	1.00	0.00	1.0	-2.37	-	-	0.0					
			0		3.68	1.0	0					
						6						
WPO	1.00	0.00	1.0	-0.42	-	0.9	0.5					
			0		1.78	3	4					
Between			0.0									
			4									
Overall	38.0	92.05		-0.08	-	0.2	0.6					
	0				0.41	5	5					
Sedentary and household		n	I-squared (between studies)	P-value	Effect size (Hedges' g) (95% CI)	Lower Limit	Upper Limit	P-value				
	AW	3.00	97.07	0.0	0.29	-	1.0	0.4	3.11	-	396.	0.93
				0		0.49	7	7		389.	22	
										99		
	BMC	2.00	85.52	0.0	0.52	-	1.4	0.2				
				1		0.41	5	7				

BP	1.00	0.00	1.00	-0.78	-	0.5	0.2					
			0		2.14	9	7					
EPUL	1.00	0.00	1.00	-0.31	-	1.0	0.6					
			0		1.65	2	5					
FCHR	4.00	59.60	0.00	-0.14	-	0.5	0.6	-0.31	-	26.7	0.96	
			6		0.81	3	9		27.4	9		
									2			
FF	1.00	0.00	1.00	-0.30	-	1.0	0.6					
			0		1.62	2	6					
GF225	1.00	0.00	1.00	1.30	-	2.6	0.0					
			0		0.04	4	6					
GVF	1.00	0.00	1.00	-1.26	-	0.1	0.0					
			0		2.64	2	7					
JU	1.00	0.00	1.00	-0.85	-	0.5	0.2					
			0		2.20	1	2					
JU24	1.00	0.00	1.00	-0.94	-	0.4	0.1					
			0		2.28	0	7					
MB	1.00	0.00	1.00	-1.34	-	0.0	0.0					
			0		2.75	8	6					
SWA p2	2.00	0.00	0.60	0.71	-	1.6	0.1					
			0		0.23	4	4					
SWA p3	4.00	91.27	0.00	0.67	0.00	1.3	0.0	8.42	-	33.7	0.29	
			0			4	5		16.9	4		
									1			
SWAM	5.00	97.42	0.00	0.04	-	0.6	0.9	22.71	-	87.8	0.35	
			0		0.55	3	0		42.4	9		
									7			
V	1.00	0.00	1.00	0.00	-	1.3	1.0					
			0		1.32	2	0					
WP	1.00	0.00	1.00	-0.62	-	0.7	0.3					
			0		1.97	3	7					
Between			0.00									
			6									
Overall	30.00	94.84		-0.09	-	0.3	0.6					
	0				0.51	2	6					
TEE (DLW)												
	n	I-squared (between studies)	P-value	Effect size (Hedges' g) (95% CI)	Lower limit	Upper limit	P-value					
EP	1.00	0.00	1.00	-0.68	-	0.2	0.1					
			0		1.58	1	6					
FF	1.00	0.00	1.00	-0.48	-	0.3	0.2					
			0		1.35	8	7					
GVF	1.00	0.00	1.00	-1.62	-	-	0.0					
			0		2.63	0.6	0					
						2						
JU24	1.00	0.00	1.00	-1.96	-	-	0.0					
			0		2.97	0.9	0					
						5						
MS	1.00	0.00	1.00	-0.68	-	0.2	0.1					
			0		1.65	9	7					

SWA	2.00	57.21	0.1 3	0.14	- 0.46	0.7 5	0.6 5					
SWA p2	1.00	0.00	1.0 0	-0.23	- 1.07	0.6 0	0.5 9					
SWA p3	5.00	94.20	0.0 0	-0.25	- 0.64	0.1 3	0.1 9	7.03	- 31.0	45.0 7	0.60	
SWAM	2.00	0.00	0.9 9	-0.04	- 0.64	0.5 6	0.9 0					
WPO	1.00	0.00	1.0 0	-1.68	- 2.62	- 0.7	0.0 0					
							5					
Between			0.0 0									
Overall	16.0 0	92.71		-0.68	- 1.15	- 0.2	0.0 0					
							1					

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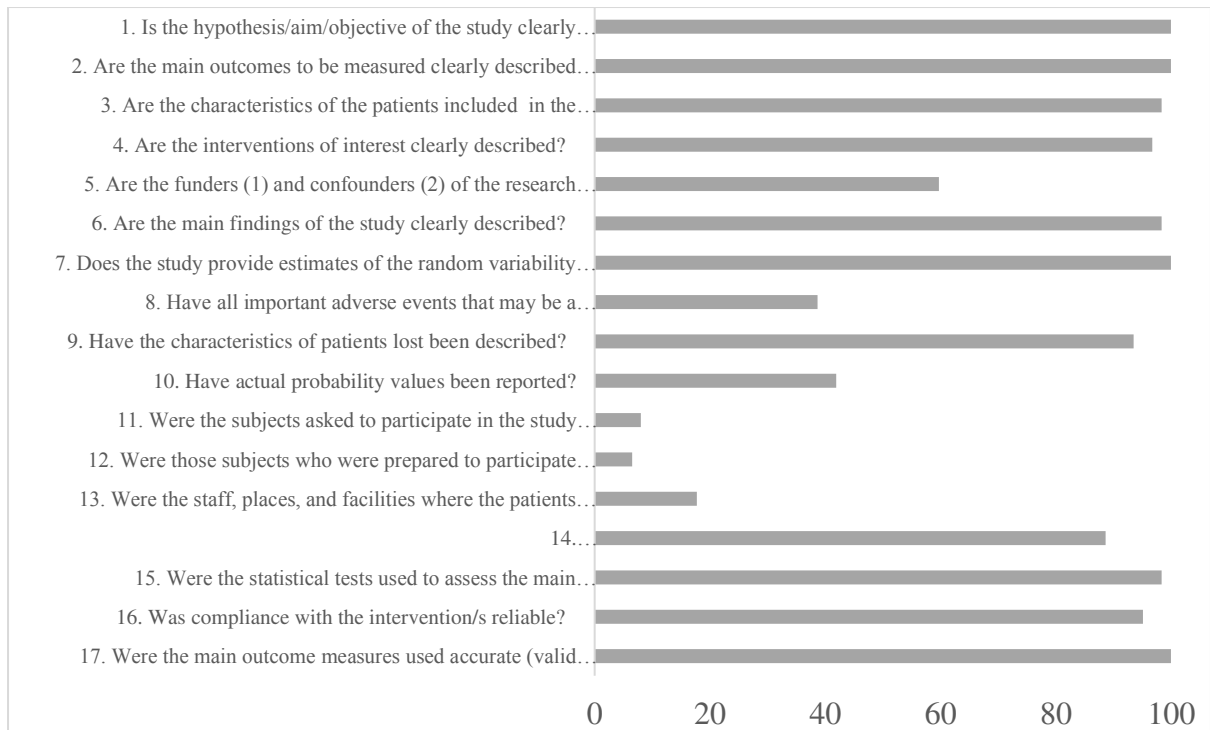
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	Reporting (/11)	External validity (/3)	Internal validity (/4)
Alsubheen, 2016	10	0	4
Bai, 2017	9	0	4
Benito, 2012	8	0	4
Berntsen, 2010	9	0	4
Berntsen, 2012	9	2	4
Bhammar, 2016	11	0	4
Boudreaux, 2018	10	0	4
Brazeau, 2011	10	0	4
Brazeau, 2014	11	0	3
Brazeau, 2016	11	1	4
Brugniaux, 2010	8	1	3
Calabro, 2014	9	0	4
Calabro, 2015	11	1	4
Casiraghi, 2013	11	0	4
Choudhry, 2017	9	0	4
Colbert, 2011	10	1	3
Correa, 2016	10	0	3
Diaz, 2015	7	0	4
Diaz, 2016	9	0	4
Donzilla, 2016	8	0	4
Dooley, 2017	10	0	4
Drenowatz, 2011	9	0	4
Erdogan, 2010	9	0	3
Fruin, 2010	9	0	3
Furlanetto, 2010	11	0	4

Gastin, 2017	8	0	4
Heiermann, 2011	8	2	4
Imboden, 2017	9	0	4
Jakicic, 2004	10	0	4
Johannsen, 2010	9	1	4
Kim, 2015	8	0	4
King, 2004	9	0	4
Koehler, 2011	10	1	4
Lee, 2011	9	0	4
Lee, 2014	9	0	4
Mackey, 2011	11	3	4
Martien, 2015	9	2	4
McMinn, 2013	9	0	4
Melanson, 2009	5	0	2
Mikulic, 2011	10	0	4
Montoye, 2017	10	0	4
Murakami, 2016	7	1	4
Nelson, 2016	10	0	4
Papazoglou, 2006	9	0	4
Price, 2017	9	0	4
Reece, 2015	9	0	4
Rousset, 2015	9	1	4
Ryan, 2013	10	0	2
Slinde, 2013	10	2	4
Smith, 2012	10	0	4
St-Onge, 2007	9	1	3
Stackpool, 2015	9	0	4
Tucker, 2015	11	0	4
Van helst, 2012	9	0	4
Van Hoye, 2014	9	0	4
Van Hoye, 2015	10	0	4
Vernillo, 2015	8	0	4
Wahl, 2017	9	0	4
Wallen 2016	9	0	4
Woodman, 2017	8	0	4

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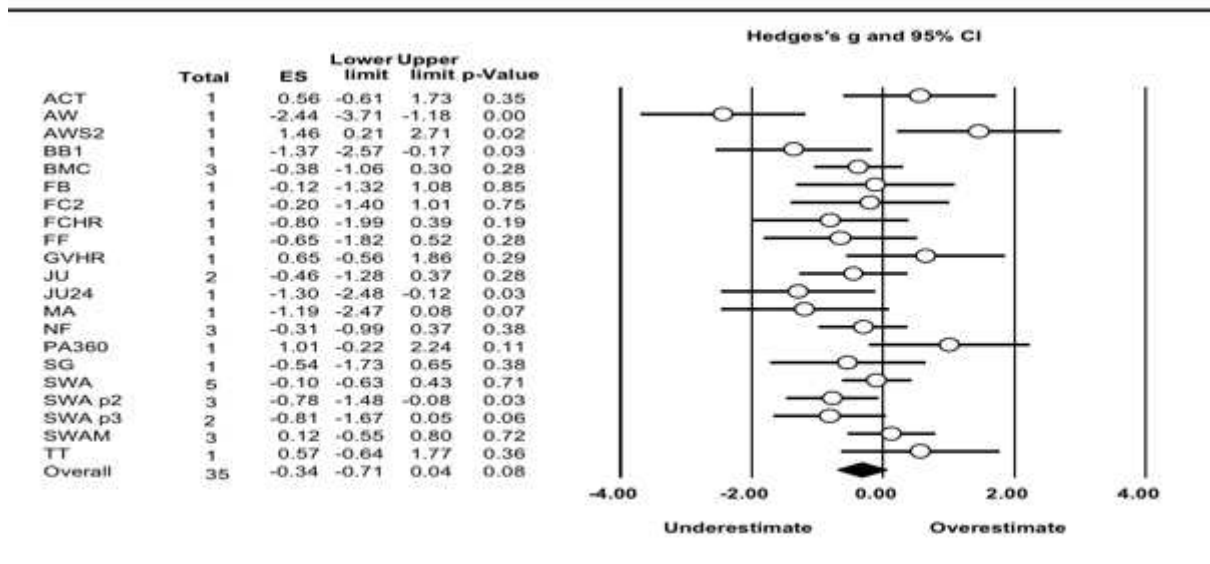
997 S9:



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Meta Analysis AEE

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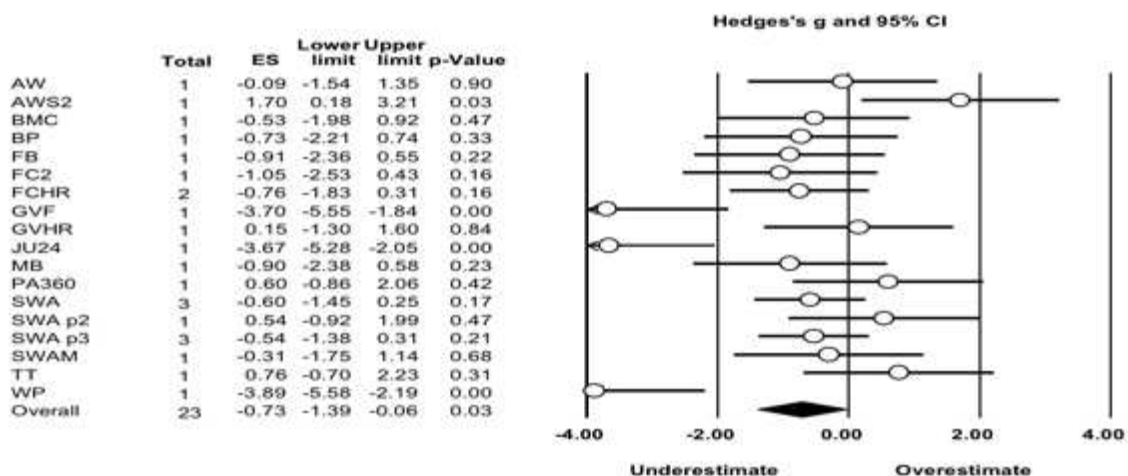
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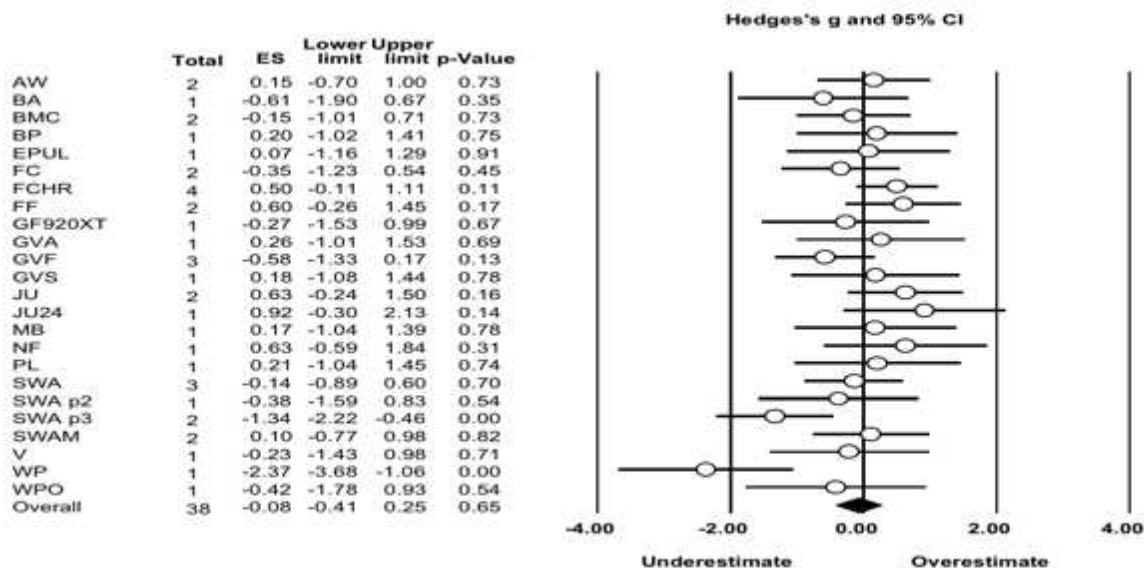
1008

Pooled Hedges' g and 95% confidence intervals (CI) for estimates of energy expenditure relative to criterion measures per device for AEE. Total refers to number of effect sizes. A negative Hedges' g statistic represents an underestimation and a positive Hedges' g represents an overestimation.



Meta Analysis Cycling

1009 Pooled Hedges' g and 95% confidence intervals (CI) for estimates of energy expenditure
 1010 relative to criterion measures per device during cycling. Total refers to number of effect
 1011 sizes. A negative Hedges' g statistic represents an underestimation and a positive Hedges' g
 1012 represents an overestimation.
 1013
 1014



Meta Analysis Running

1015 Pooled Hedges' g and 95% confidence intervals (CI) for estimates of energy expenditure
 1016 relative to criterion measures per device during running. Total refers to number of effect
 1017 sizes. A negative Hedges' g statistic represents an underestimation and a positive Hedges' g
 1018 represents an overestimation.
 1019
 1020