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1 **Pilot study assessing the influence of skin type on the heart rate measurements**
2 **obtained by photoplethysmography with the Apple Watch**

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6 **Running Head:** Skin Type and photoplethysmography

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28 **ABSTRACT**

29 Photoplethysmographic imaging (PPG) is currently used to measure heart rate (HR) and
30 the accuracy of PPG can be influenced by pigmentation of the skin; however, the effects
31 of skin color-related artifacts on PPG during exercise remain unclear. This study aimed
32 to assess the agreement between the Apple Watch photoplethysmography sensor and
33 a criterion, for measuring heart rate across a range of intensities during exercise and to
34 determine the influence of skin type on the accuracy of the measure. Forty-five males
35 (20-43 y) completed the Fitzpatrick Skin Scale and were classified into three different
36 skin type groups: a) types II (n=15), III (n=15) and IV (n=15). Participants performed a
37 graded incremental cycle-ergometer test while simultaneously wearing the Apple
38 Watch and a Polar monitor as a criterion measure. Data from both devices were
39 collected in 5-s epochs. Correlations between devices were very good (0.96-0.99 [95%CI:
40 0.94 to 0.99]). Significant differences were observed between skin types II and III when
41 the intensity of the exercise was increased, albeit with trivial to small effect sizes (ES:
42 0.05 to 0.28). All significant differences corresponded to <2% of relative difference
43 between both devices. Bland-Altman analyses showed a trivial but systematic
44 underestimation of HR in the Apple Watch compared to Polar for all skin types during
45 exercise. In conclusion, the Apple Watch accurately measures HR when cycling at
46 different intensities and certain types of skin seem not to influence these measures,
47 which may have important implications for controlling the intensity of exercise.

48

49 **Key Terms:** Heart rate; agreement; wearable sensors; exercise; skin type

50

51 **1. INTRODUCTION**

52 Heart rate (HR) is commonly used to monitor exercise intensity and therefore accurate
53 measures are important to provide individuals with precise estimates of cardiovascular-
54 based exercise intensity for safe and effective workouts. For many years, noninvasive
55 techniques for monitoring HR, such as portable electrocardiography (ECG) monitors,
56 have been analyzed. Most of these devices detect HR via a chest strap (e.g., Polar®),
57 which has been shown to be both a valid and a reliable method for determining HR
58 during rest and exercise [1,2]. However, some individuals are unable (e.g., sensitive skin)
59 or unwilling (e.g., the attachment of the electrode may be troublesome and the strap
60 needs to be worn on the skin and kept wet for accurate signal detection) to use these
61 methods [3]. Therefore, due to the problems experienced when fitting the strap and the
62 discomfort reported by many during exercise, especially when worn for extended
63 periods [4], the use of other alternatives has been recommended [5].

64

65 Optical methods, such as photoplethysmographic imaging (PPG), are also widely used
66 and have been investigated in recent years as an alternative to overcome the limitations
67 of traditional methods [5]. Numerous wearable activity trackers, such as the Apple
68 Watch®, incorporate optical LED sensors to non-invasively detect changes in the light
69 intensity with respect to the change in volume of blood flow and thus measure HR [5,6].
70 This technology consists of a light source to illuminate the skin tissue, and a photo-
71 detector to measure small variations in light intensity associated with changes in
72 perfusion in peripheral blood vessels [3]. The simplicity and easy accessibility of PPG has
73 meant that many use continuous HR monitoring to control exercise intensity without
74 being aware of the implications this may have on their performance or health.

75

76 PPG may be affected by several external factors such as anatomical placement (pressure
77 between a probe and the skin), environmental noise (ambient light), sweat and
78 especially motion artifacts, as signals are very sensitive to small changes in sensor
79 position which are considered an important obstacle when computing HR from PPG
80 [4,6,7]. Also, the accuracy of PPG can be dependent on the type and intensity of the
81 exercise [8]. Further, the sensitivity of the sensor may be influenced by pigmentation of
82 the skin [9]. However, while many signal processing techniques have been proposed to
83 remove motion artifacts during exercise, the effects of skin color-related artifacts on
84 PPG during this practice remain unclear [4].

85

86 Preliminary studies seem to indicate that variability caused by the amount of melanin
87 may affect characteristics of PPG signals [4,9,10] and while green wavelength, which is
88 the one used in the Apple Watch, bring greater signal resolution during exercise [11],
89 evidence of the accuracy of these sensors when measuring HR in people of varied skin
90 pigment is scarce [4]. Thus, in order to correctly monitor exercise intensity (i.e., HR)
91 there is a clear consensus on the importance of validation studies incorporating separate
92 analyses specific to subject skin color [8]. Therefore, the aims of this pilot study were to
93 (a) determine the validity of the Apple Watch PPG sensor when measuring HR across a
94 range of exercise intensities in reference to the Polar device criterion measure which
95 employed a chest strap based technology and has been shown to be highly correlated
96 to ECG [1,2] and (b) show some preliminary data on the influence of skin type on the
97 accuracy of the measure. We hypothesized that at certain intensities (i.e., high
98 intensities) the light reflectance would be different across skin types.

99

100 **2. MATERIALS AND METHODS**

101 **2.1 Participants**

102 Forty-five healthy males were recruited to this study. Participants were between the
103 ages of 20 and 43 years (24 ± 4 y; body mass: 72.2 ± 5.8 kg; stature: 1.77 ± 0.05 m) and
104 engaged in physical activities at least three times per week. The study was conducted at
105 the Sport Science Lab at the University of Seville. Participants were recruited by visiting
106 scheduled classes and asking for volunteers to complete the test. Participants were
107 eligible if they were between 18 and 45 years of age, and did not have any history of
108 injury or disease (e.g., peripheral circulatory failure) that would prevent them from
109 safely performing the study protocol. All participants refrained from smoking, caffeine
110 intake, alcohol consumption and extreme exercise for 12 h before the experiment to
111 minimize effects that could affect blood flow. The study was approved by the
112 Institutional Review Board of the University of Seville and after being informed of the
113 purpose, procedures, benefits and risks of the study, written informed consent was
114 obtained from each participant.

115

116 **2.2. Procedures**

117 As the skin perfusion also changes with environment, the experiments were performed
118 in the laboratory under temperature-controlled conditions. After arrival at the
119 laboratory, participants were required to complete the Fitzpatrick Skin Scale [12] where
120 the range consisted from type I = high photosensitivity to type VI = low photosensitivity.
121 After comparing with a photograph of each subject's forearm, participants were then
122 classified by an assistant not involved in the study into three different Fitzpatrick skin
123 type groups, ranging from type II to type IV (as no participants with type I, type V or type

124 VI skin photosensitivity participated in the study). All groups were equally sized with 15
125 participants in each.

126

127 The Apple Watch was placed on the forearm approximately 2 cm from the wrist bone
128 according to the manufacturer's specifications. The criterion measure of HR was
129 measured via a HR receiver (Polar RS800CX monitor, Polar Electro OY, Kempele, Finland)
130 that was placed on the left wrist and an accompanying chest strap that was applied as
131 per manufacturer's instructions. This device has been shown to be a valid gold-standard
132 measure of mobile HR monitoring technology when compared with ECG measurements
133 during exercise [1,2]. Body mass and stature were assessed and then participants were
134 kept in a quiet room in a seated position for 10 min while their resting HR was measured.
135 Then, each participant immediately started an incremental graded exercise test on a
136 cycle ergometer.

137

138 The incremental graded exercise test started after a standardized warm-up consisting
139 of 5 min of pedaling on a cycle ergometer (Ergoselect 200, Ergoline GmbH, Bitz,
140 Germany) at a load of 50 W. Then participants performed a maximal graded exercise
141 test at an initial load of 50 W (cadence of 60 rpm) that increased by 25 W every one min
142 until exhaustion. Data from the Polar and Apple Watch devices were collected in 5-s
143 epochs by reading each HR value from the watch face. HR values were independently
144 registered by two assistants. These values were used to calculate the mean HR over each
145 minute while performing the incremental protocol. For each group, HR was divided in
146 percentage zones from each individual peak HR to compare relative zones between both

147 devices (Zone 1 = 0-59%, Zone 2 = 60-69%, Zone 3 = 70-79%, Zone 4 80-89% and Zone 5
148 = 90-100%).

149

150 **2.3. Data analysis**

151 Data are represented as mean (SD) for each device and phototype. Prior to assessing the
152 relative HR within each zone from both devices for each group, normality was assessed
153 using the Kolmogorov-Smirnov test. All data violated the assumption of normality and
154 therefore a non-parametric Mann-Whitney U test was used to assess possible mean
155 differences between HR zones. A P value of < 0.05 was used to determine whether the
156 possible differences were statistically significant or not. In addition, the standard error
157 of the mean was calculated (SSE). To assess the magnitude of the differences, Cohen's
158 *d* effect size (ES) was calculated by dividing the pooled standard deviation by the mean
159 differences between both devices in each HR zone. The following magnitudes were used
160 to interpret the ES: trivial effect: < 0.20 , small effect: from 0.20 to 0.59, moderate effect:
161 from 0.60 to 1.19, large effect: from 1.20 to 1.99, very large effect: > 1.99 [13].

162

163 To determine the agreement between both instruments, two separate analyses were
164 conducted. First, Pearson product moment correlations and 95% confidence intervals
165 (CI) were used for each pair of HR data for each group. Prior to any plots analysis, data
166 were log transformed to reduce non-uniformity associated errors. The following
167 magnitudes were used to interpret the correlations: very poor ($r = 0.45$ to 0.69), poor (r
168 $= 0.70$ to 0.84), good ($r = 0.85$ to 0.94), very good ($r = 0.95$ to 0.994) and excellent ($r \geq$
169 0.995) [14]. Second, to calculate absolute systematic bias, Bland-Altman plots for
170 repeated measures were used for each group, together with the corresponding 95%

171 limits of agreement (LoA) following the guidance of Bland & Altman [15], using
172 calculations provided by Zou [16]. For all measures, the true value was assumed to vary.
173 Finally, the coefficient of correlation (r^2) of the plots were calculated to assess either if
174 bias was constantly along all the data ($r^2 < 0.1$) or tended to overestimate lower or higher
175 heart rates [17]. All calculations were conducted using SPSS (version 22.0, Chicaco, IL).

176

177 **3. RESULTS**

178 HR values together with the SEE and ES for each skin type at rest (Table 1) and during
179 the graded incremental exercise test (Table 2) are reported for both devices. At rest,
180 there were no significant differences between both devices for any skin type (trivial ES).

181

182 During exercise, type II participants showed significant differences between Apple
183 Watch and Polar in zones: 70-79% (135 ± 10 vs 138 ± 10 respectively, small ES), 80-89%
184 (154 ± 12 vs 157 ± 11 respectively, small ES) and 90-100% (175 ± 12 vs 178 ± 12
185 respectively, trivial ES). Type III showed significant differences between devices within
186 the same zones: 70-79% (138 ± 8 vs 140 ± 8 respectively, small ES), 80-89% (158 ± 9 vs
187 160 ± 8 respectively, small ES) and 90-100% (176 ± 9 vs 178 ± 9 respectively, small ES),
188 while non-significant differences were observed for Type IV in any zone. All significant
189 differences found corresponded to <2% of relative difference between both devices
190 (from 1.2 to 2.1% [CI: 1.1 to 1.8]).

191

192 Figures 1 and 2 displays the Pearson product-moment correlation coefficient (r) for the
193 Apple Watch showing excellent correlations with the criterion measure during exercise
194 (all $r=0.99$ [0.99-0.99 CI], $p<0.001$). Good to excellent correlations were also observed

195 during the rest condition in all groups (type II= 0.98 [0.97-0.99 CI], Type III= 0.96 [0.94-
196 0.98 CI], type IV= 0.98 [0.98-0.99 CI]) (Figure 2).

197

198 Bland-Altman analyses (mean difference and limits of agreement) are presented in
199 Figure 3.

200

201 There was a proportional systematic bias in the recorded HR between both devices for
202 all the skin types during the exercise condition (mean bias [95%LoA]): type II= -2(-8 to 8)
203 beats·min⁻¹, type III= -2(-8 to 4) beats·min⁻¹, type IV= -1(-6 to 4) beats·min⁻¹. In the resting
204 condition, participants in type II exhibited a mean bias of 0 (-5 to 4), type III= 0 (-5 to 5)
205 and type IV= 0 (-5 to 4).

206

207 **4. DISCUSSION**

208 While previous studies investigated the accuracy of wrist wearable technologies for
209 estimating HR at different intensities [5,18,19], to our knowledge this is the first study
210 to examine how well the Apple Watch wrist-worn device agrees with a criterion measure
211 of HR during rest and cycling at different intensities while examining the influence of a
212 range of skin types on this agreement. The results obtained in the current study suggest
213 that the Apple Watch agrees well with the criterion measure and therefore fulfills
214 published criteria for HR measurement provided in previous research [20]: a) A
215 correlation $r=0.90$ or greater between the test device and the criterion measure; b) A
216 mean bias less than 3 beats·min⁻¹; c) A standard error less than 5 beats·min⁻¹.

217

218 Our results showed good to excellent correlations during exercise (all $r > 0.9$) and mean
219 bias $< 2\%$. Therefore, and following previous recommendations coming from the
220 validation of consumer devices for accurate HR measurement [8] we can state that the
221 Apple Watch, which continually measures HR using PPG, can be used during a maximal
222 graded exercise test on a cycle ergometer. However, the question that arises is whether
223 the Apple Watch is still accurate depending on the skin type of the participants. In this
224 sense, we observed good to excellent correlations in all groups ($r > 0.93$) between the
225 devices with a mean bias $< 2 \text{ beats}\cdot\text{min}^{-1}$ and an absolute difference from criterion
226 measurements of $< 2\%$.

227

228 When analyzing the data as a whole, regardless of the skin type, our results are in
229 accordance with Wallen et al. [19] who also examined the accuracy of different wrist-
230 worn devices, including the Apple Watch, to measure HR. These authors reported that
231 the devices were within 1–9% of reference estimates. However, they also reported that
232 all devices underestimated HR. In the same line, Dooley et al. [18] recently reported that
233 the magnitude of errors across all intensities (treadmill exercise) for the Apple Watch
234 were between 1.1%-6.7%. In the current study the mean absolute percentage error
235 observed was $< 2\%$ at all intensities. These discrepancies can be attributable to
236 numerous factors such as the mode of exercise, intensity, and participant
237 characteristics. If we focus on intensity, Jo et al. [8] reported that the performance of
238 another wrist-worn device (Fitbit Charge HR) was poor during low intensity cycling (60
239 W) and usually, the accuracies were reduced with increasing exercise intensity. In fact,
240 it was recently reported that the correlation for the Apple Watch decreased at high
241 intensities [18,19], which contrasts slightly with the results of our study.

242

243 The mode of exercise may also reduce the correlation between both measures [4]. Thus,
244 Wallen et al. [20] revealed that HR measurement error tends to differ between treadmill
245 and cycle protocols and recently, Shcherbina et al. [21] also reported the lowest error in
246 measuring HR for the cycle ergometer task of 1.8% (0.9%–2.7%). These figures are in
247 agreement with the ones reported in our study - 1.59% (0.93%–2.09%). This could be an
248 explanation for our results since this tendency showing greater error when speeds were
249 increased was also observed by Lee & Gorelick [22] in the validation study of a different
250 smart watch and they suggested that this could be due to the greater disturbances by
251 the movement, the sensitivity of the device or even by the skin type of the participant
252 being studied. However, due to the large number of variables affecting PPG, it is not
253 practical to include all variables in a single study. For instance, the movement of the
254 Apple Watch was not varied in our study. This had the benefit of not introducing a
255 potentially confounding variable, but it needs to be considered that motion artifacts can
256 have powerful effects on the efficacy of the HR measurement [9].

257

258 Therefore, regarding the skin type, the Apple Watch was not significantly different from
259 the Polar HR monitor during baseline (all $P > 0.873$). On the basis of the Bland-Altman
260 analysis there was a systematic bias in the recorded HR between both devices for all the
261 phototypes during the exercise condition (bias ≤ 2 beats·min⁻¹ and the 95% limits of
262 agreement: -8 to 5 beats·min⁻¹) which is consistent with Wallen et al. [20] who previously
263 validated the Apple Watch showing a very good correlation ($r = 0.95$) with ECG and a
264 small mean bias of -1 beats·min⁻¹. Also during exercise, Spierer et al. [4] indicated that
265 some skin types could produce more error in some wrist-worn devices (Mio Alpha). In

266 the same line, Fallow et al. [11] demonstrated that a dark skin type (type V) attenuated
267 the signal in comparison with other skin types. It is known that melanin can absorb light
268 and thus attenuate the incident light wavelength [11]. However, while a dark
269 pigmentation was suggested to lead to a worse light reflection [11], in the current study
270 non-significant differences between both devices were observed for the darker skin type
271 analysed (type IV). This contrasts with Wallen et al. [20] who reported statistical
272 differences between correlations for HR based on skin colour (skin Type >IV was
273 statistically different to skin Type <IV). In any case, we observed diminished
274 performance when cycling at higher intensities (i.e. >70-79%) in participants with skin
275 type II and III. The measurement was accurate monitoring HR even with increasing
276 physical exertion, although some differences existed between types II and III, especially
277 in the higher intensity zones. An alternative explanation to these discrepancies can be
278 attributed to changes in the position of the sensor (the proximity of the device on the
279 wrist) derived from an isometric muscle contraction while holding on to handlebars
280 while cycling. It was suggested that in this activity, the fluctuations associated with
281 muscle actions might affect PPG signals [23].

282

283 There are a number of limitations that have to be considered when interpreting the
284 results. The first refers to the sample, where only healthy, relatively young (20–43 years)
285 individuals within the normal range of body composition were included, which could
286 limit the generalization of the results to other population groups (e.g. older adults).
287 Second, and despite having the same number of participants per group, we did not find
288 sufficient number of people classified as skin type I, V and VI, although this seems logical

289 taking into account the characteristics of the population in which the study was
290 developed.

291

292 Despite these limitations, the current study shows a strong agreement between the
293 Apple Watch and the criterion measure when exercising on a cycle ergometer at
294 different intensities, which together with a low systematic bias ensure that both devices
295 may be used inter-changeably for accurate HR measurements. Moreover, some
296 preliminary results on the effect of skin type suggest that the skin types analyzed have
297 no influence on the heart rate values obtained.

298

299 **PERSPECTIVES**

300 Heart rate monitors are widely used to control exercise intensity; however, many
301 athletes complain about having to use the chest bands. The results of this study provide
302 scientists, coach's and clinicians the error measurement of the Apple Watch when
303 cycling at different intensities. The study used a novel approach to measure accuracy of
304 this device for HR at specific bouts of exercise intensities according to the skin
305 pigmentation, which may be relevant in the sport medicine area when controlling or
306 prescribing physical activity by means of this PPG sensor. While the practicality of the
307 tested sensor has to be examined in a future studies, especially in a real-life setting or
308 with respect to different activities, we show important findings since this wrist-worn
309 device utilizing PPG offers both medical staff and performance coaches a valid method
310 to monitor HR while exercising on a cycle ergometer, which is essential to control the
311 exercise intensity.

312

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314 This study had no funding support.

315 **Compliance with Ethical Standards**

316 Authors declare that there are no conflicts of interest.

317 **Ethical Approval**

318 All procedures performed in studies involving human participants were in accordance
319 with the ethical standards of the institutional and with the 1964 Helsinki declaration and
320 its later amendments or comparable ethical standards. Informed Consent Informed
321 consent was obtained from all individual participants included in the study.

322

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388

389 **Figure legends**

390 **Figure 1.** Exercise data correlations plots. HR= heart rate. A= Type II; B= Type III; C= Type IV
391 **Figure 2.** Resting data correlations plots. HR= heart rate. A= Type II; B= Type III; C= Type IV
392 **Figure 3.** Bland-Altman plot showing the mean bias and 95% limits of agreement (with 95%
393 confidence intervals) for the absolute differences in heart rate (% HR_{max}) in participants with
394 skin type II (Figure 3a), III (Figure 3c) and IV (Figure 3e) and the relative differences in heart rate
395 (% HR_{max}) in participants with skin type II (Figure 3b), III (Figure 3d) and IV (Figure 3f).