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Estimating post-match fatigue in soccer: The effect of individualization of speed thresholds on perceived recovery

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Keywords:	external load, Global Positioning System, perceptual measures, soccer match, youth players

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1 ABSTRACT

2
3 **Purpose:** This study investigates the effectiveness of different individualization methods of
4 speed zones during match-play to estimate post-match perceptual recovery in soccer. **Methods:**
5 Twelve under 19 players undertook field-based assessments to determine their maximal aerobic
6 speed (MAS) and maximal sprint speed (MSS). External load (extracted from 10 Hz GPS over
7 10 official matches) was measured and classified into four categories: 1) low-speed running;
8 2) moderate-speed running; 3) high-speed running (HSR) and 4) sprinting. Match running
9 distribution into different speed zones were categorized using either: 1) MAS; 2) MSS; 3) MAS
10 and MSS as measures of locomotor capacities (LOCO); and 4) using absolute values (ABS).
11 Players perceived recovery status (PRS) was recorded post (Post), 24 (G+24H) and 48-hours
12 after each game (G+48H). **Results:** Different individualization methods resulted in distinct
13 match outputs within each locomotor category. The PRS was lower ($p<0001$) at Post (3.8 ± 1.32 ,
14 $95\%CI = 3.6-4.2$), G+24H (5.2 ± 1.48 , $95\%CI = 4.9-5.6$) and G+48h (6.0 ± 1.22 , $95\%CI = 5.7-$
15 6.3) compared to pre-match values (7.1 ± 1.05 , $95\%CI = 6.8-7.3$). The absolute PRS score was
16 better associated with HSR using LOCO method at Post (Beta = -1.7 , $95\%CI = -3.2, -0.22$,
17 $p=0.027$), G+24H (Beta = -2.08 , $95\%CI = -3.22, -0.95$, $p=0.001$) and G+48H (Beta = -1.32 ,
18 $95\%CI = -2.2, -0.4$, $p=0.004$) compared with other individualization methods. **Conclusions:**
19 Our results suggest that LOCO may better characterize the match intensity distribution
20 (particularly for the HSR and sprinting categories) and should be preferred over MAS and MSS
21 to estimate perceived recovery.

22
23 **Keywords:** external load; GPS; perceptual measures; soccer match; youth players.
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51 INTRODUCTION

52 The monitoring of the training load has been considered by practitioners as an important
53 strategy to assist athletes to enhance performance and to reduce injury risk in a variety of sports.
54 Within this context, research has previously established that the development of fatigue during
55 training and competitive phases impact player's responses to training and competition demands
56 ¹. Moreover, excessive fatigue may also compromise the capacity of the players to tolerate and
57 to recover from high training loads, and consequently increase the odds of injury ^{1,2}.
58 Accordingly, elite soccer clubs should seek to implement a fatigue monitoring system and
59 effective strategies to aid player's recovery as part of the training process and overall practice
60 organization ³.

61 Athlete self-reported measures (ASRM) are "paper-based or electronic records of an athlete's
62 perceived physical, psychological, and/or social well-being, completed on a regular, often daily
63 basis"⁴. Research has recognized that ASRM present the triple advantage of being easy to use,
64 cost effective and sensitive ^{3,5}. Recently, studies reported that specific ASRM such as wellness
65 questionnaires were effective to estimate the capacity of team-sport players to perform during
66 training and competition ⁶⁻⁸. As such, subjective measures have been generally supported as
67 the most efficient instrument to monitor player's fatigue/recovery status in team-sport setting
68 ^{6,7,8} compared to objective measures. This suggests that subjective measures may be more
69 appropriate or more sensitive to assess the stress imposed by training and competition ⁵.
70 Specifically, perceptual ratings (e.g. recovery, fatigue and soreness) collected subsequently to
71 the match (e.g. 24- and 48-hours post-match) may present a more holistic appraisal of the
72 internal stress (e.g. oxidative stress, muscle micro-trauma) induced by the previous match and
73 reflect on the extent of residual fatigue ⁹.

74 Alternatively, external output during match-play and training practices have been associated
75 with acute and residual changes in specific fatigue-related markers such as muscle damage,
76 neuromuscular readiness and perceptual state ¹⁰. Within this context, the use of arbitrary and
77 absolute thresholds for quantifying running speed during training and competition have been
78 predominantly applied ¹¹⁻¹³. This approach has also been adopted by previous investigations to
79 describe the impact of external load on post-match fatigue and recovery time-course ^{10,14,15}.
80 However, it is important to consider that running speed thresholds tailored (e.g. maximal
81 aerobic speed) to each player may better identify the relative physiological and neuromuscular
82 demands/load experienced during training and competition ^{16,17}. It is reasonable to assume
83 therefore, that a tailored approach to external load quantification may be appropriate to inform
84 decisions on training load management, as well as the time-course of training-induced
85 adaptations and recovery status ^{3,10,16,18}. Accordingly, the external load individualization method
86 has been commonly applied to "reassess" the competitive demands in soccer ^{16,17,19,20}. This
87 individualization approach is performed through the use of different physiological attributes
88 such as maximal aerobic speed (MAS) or maximal sprinting speed (MSS) ^{19,20}, or through a
89 combination of MAS, MSS and anaerobic speed reserve (ASR) ¹⁹⁻²¹.

90 In line with the basic principle of exposure-response relationship (i.e. workload-fatigue), a
91 tailored approach to quantify training load is recommended. This may allow an accurate
92 identification of the physiological strain experienced by each individual player during training
93 and competition ^{3,17}. This approach may consequently benefit practitioners, particularly for
94 programming daily training plan of each individual player or targeted groups (e.g. injured
95 players returning to sports participation and/or competition following rehabilitation process)²².
96 Recently, Scott & Lovell ⁹ observed that the individualization of speed thresholds was not
97 sufficient to enhance the dose-response determination in female soccer players. It is important
98 to highlight that the study ⁹ examined the dose-response relationship only during a training

99 camp without including matches. According to the authors, the speed zone individualization
100 may better elucidate the dose-response during matches⁹ since match load (external and/or
101 internal load during the competition) has been reported as a main determinant of a high weekly
102 training load²³. Consequently, it is still unclear which individualization method can be
103 considered as optimal to estimate players' fatigue and recovery responses. As stated previously,
104 the use of arbitrary and absolute thresholds have been applied to investigate the impact of
105 external load metrics on perceptual recovery measures (total quality recovery scale, and brief
106 assessment of mood)^{14,15}. However, only one study has examined the efficiency of an
107 individualized approach to quantified the dose-response to training (e.g. fatigue and soreness)⁹.
108 To our knowledge there is currently a paucity of studies examining the PRS as an indicator of
109 post-match fatigue and recovery. Moreover, this is the first study to examine the
110 individualization of external load and its utility to estimate fatigue and recovery in male soccer
111 players. Therefore, our objective was to investigate the effectiveness of different
112 individualization methods of speed zones to estimate post-match perceptual recovery in soccer.
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114

115

METHODS

Participants

117 Fourteen under 19 outfield male youth soccer players from a high-level youth academy from
118 Qatar (Al Sadd Sports Club) participated in this study. Players' displayed on average 9 years
119 of training experience. Data were collected during the competitive phase (10 official matches
120 played between February and March) of the season 2016/2017. The inclusion criteria were: 1)
121 a minimum of two matches per player where PRS and GPS match data were both recorded,
122 and 2) ≥ 75 -min of total match-play time. This resulted in a total of 78 individual match
123 observations from 12 players (age 18.9 ± 0.8 years, height 174.4 ± 0.51 cm, weight $66.4 \pm$
124 10.44 Kg, body mass index 21.8 ± 2.4 kg/m²) belonging to different playing positions (four
125 defenders, five midfielders and three forwards). All matches were played in 2 x 45 min with
126 15 min interval over an official natural soccer field (70m x 100m). Players played all the
127 matches under the same tactical formation (4-3-3) throughout the period of the study. The
128 temperature and humidity during the games were $24 \pm 3.1^{\circ}$ C and $55 \pm 11\%$ respectively. The
129 weekly micro-cycle during the competitive phase was comprised by one official match and 5
130 training sessions with a day off 24 hours after the match. Specifically, the post-match training
131 routine was comprised by a passive recovery day 24 hours after the match (G+24H, day-off)
132 and an active recovery training session 48 hours after the match (G+48H). This study has been
133 approved by the ethical committee of St Mary's University, England, and followed the ethical
134 recommendations suggested by the Declaration of Helsinki. Upon receiving the ethical
135 approval, the club provided no objection to the study and all participants and their respective
136 parents provided written informed consent.
137

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Study Design

139

140 A cohort observational study design was used to investigate the effectiveness of different
141 individualization methods of speed zones to estimate post-match perceptual recovery in soccer.
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Physical testing battery

144 A physical testing battery was used to measure the status of the player's performance (MAS =
145 16.1 ± 1.1 km/h, MSS = 31.6 ± 1 km/h, ASR = 15.5 ± 2.0 km/h) and to inform the individualized
146 speed thresholds. The physical assessments included the 30-15 intermittent fitness test (IFT)

147 and the 40 m straight line sprinting and were performed one week before the beginning of the
148 data collection ²⁴. Following a debrief of the assessments and a standardized warming up,
149 players started the 40 m sprinting test from a standing position with their front foot 0.5 m
150 behind the first timing gate and were instructed to accelerate maximally until they cross the
151 last timing gate (Race Time 2, Microgate S.r.l., Via Stradivari, 4, 39100 Bolzano – Italy). The
152 players performed two trials with at least three min of a passive rest between repetitions to
153 allow full recovery. Each split time (e.g. 10, 20 etc.) was measured to the nearest 0.01 s.
154 Subsequently, MSS was defined as the fastest 10m split obtained during test ²⁰. The Intraclass
155 Correlation Coefficient (ICC) values of the 40 m sprinting test have been observed to range
156 between 0.94 - 0.99 ²⁵. Following a recovering period of approximately 10 min, players were
157 required to perform the 30-15_{IFT}. The speed attained during the last completed stage of the 30-
158 15_{IFT} was taken as the final velocity (V_{IFT}) and MAS was estimated as 85% of the V_{IFT} (MAS
159 $= 0.85 * V_{IFT}$) ²⁶. This test has been shown to have good test-retest reliability with a typical
160 error of measurement of 0.3 km/h (ICC = 0.96), suggesting a potential difference of 1 stage or
161 0.5 km/h²⁴.
162

163 **Data individualization methods**

164 External load was measured with a 10 Hz GPS device (Optimeye S5, Catapult Innovations,
165 Australia). The devices were allocated at the players' upper back between the scapulae housed
166 in a tight-fitting garment to reduce movement artefact. The devices were turned on just before
167 the warming up (~ 30 min) prior to the matches to enable acquisition of the satellite signals.
168 During the period of data collection, players used the same GPS unit to reduce the measurement
169 error ²⁷. The mean number of satellites during data collection was 13.9 ± 1.1 and the mean
170 horizontal dilution of position was 0.7 ± 0.05 . Following each match, GPS data was
171 downloaded from Catapult Sprint v5.0.6 software. The raw data was then transferred to a
172 personalized Microsoft Excel spreadsheet (Microsoft, Redmond, USA). As described in Table
173 1, match running intensity was classified into four categories: 1) low-speed running (LSR); 2)
174 moderate-speed running (MSR); 3) high-speed running (HSR) and 4) Sprinting. Total high-
175 speed running (THSR) was calculated as the sum of HSR and Sprinting. Match running
176 distribution into different speeds zones were categorized either using relative or absolute
177 methods as per the following: 1) individualized using MAS *per se*; 2) individualized using
178 MSS *per se*; or 3) individualized using MAS and MSS as measures of locomotor capacities
179 (LOCO); and 4) using absolute values (ABS). The criteria used to individualize speed
180 thresholds is provided in Table 1 ²⁰. The individualization approach utilized by this study
181 followed the procedures previously applied by Hunter et al. ²⁰.

182 ***Insert Table 1 here***

183 **Perceived recovery scale**

184 The Perceived Recovery Status (PRS) is a sport-specific ASRM that was developed as a
185 convenient non-invasive marker with the primary goal to assess players recovery ^{4,28}. The
186 straight-forward response nature [0-10 scale (11-point), 0 (very poorly recovered/extremely
187 tired) to 10 (very well recovered/highly energetic)] favors the regular use of this empirical
188 scale in soccer ⁴. Additionally, the psychometric properties of the PRS (theoretical derived, a
189 documented instrument development and validity) make it highly regarded and accepted in the
190 literature ⁴. Previous studies revealed that PRS is sensitive to detect changes in sprint running
191 performance ²⁸ and showed to be acutely associated with the hormonal (testosterone) and
192 muscle damage (creatine kinase) responses to heavy resistance training ²⁹. The perceived
193 recovery of the players was measured through the following approach: the PRS scale has been

194 shown to each player followed by the question “how do you feel?”. This approach was adopted
195 approximately -30 minutes before starting the match (pre), 15 min after the match ends (Post),
196 24 hours (G+24H) and 48 hours (G+48H) after each single match. In case the 24 hours post-
197 match measurement coincided with a day off, the PRS was self-reported via mobile phone and
198 the respective result was added to the player profile. Players were familiarized with the PRS
199 scale (e.g. at least 7-months of exposure to the scale) and reported procedure (e.g. during the
200 off days) as part of the club fatigue monitoring protocol. PRS data was analyzed as absolute
201 values, and percentage of change described by the following equation: $((\text{Post-Pre})/\text{Pre}) * 100$.
202

203 **Statistical Analysis**

204 All data was continuous and presented as mean and standard deviation (mean \pm SD). The data
205 was screened and passed the test for normality using Shapiro-Wilk method. A linear mixed
206 model analysis was performed separately to compare the low speed running, moderate speed
207 running, high speed running, and sprinting using different individualization methods (MAS,
208 MSS, LOCO, ABS). Same statistical method was used to compare PRS over time (Pre, Post,
209 G24+H, G48+H). We adjusted for Bonferroni post-hoc pair-wise comparisons. Linear mixed
210 model analysis was again performed to generate parameter estimates and coefficients to predict
211 PRS at Post, G24+H, and G48+H separately using all individualization methods. The
212 parameter estimates and AIC statistics were reported for all models with significant
213 associations. Models with the lowest AIC was preferred and a more than two unit change in
214 AIC should be considered as meaningful change. Threshold values for Cohen effect size (ES)
215 were defined as trivial (< 0.2), small ($0.2 - 0.6$), moderate ($0.6 - 1.2$), large ($1.2 - 2.0$) and very
216 large (> 2.0)³⁰. A P-value at < 0.05 was considered as the threshold for statistical significance.
217 Data was analyzed using SPSS software (version 21.0, IBM SPSS Statistics, Chicago, IL,
218 USA).

219

220

221

222 **RESULTS**

223

224

225 ***Match Load according to individualization methods***

226 The different individualization methods resulted in distinct match outputs within each
227 locomotor category (Table 2). LSR and MSR was lower when using the MAS and LOCO
228 individualization methods vs MSS ($p < 0.001$, ES = 1.0 and 1.6 for LSR and MSR, respectively)
229 and ABS ($p < 0.001$, ES = 0.6 and 0.7 for LSR and MSR, respectively). In addition, a significant
230 greater distance in LSR and MSR was quantified by the MSS compared to ABS method
231 ($p < 0.001$, ES = 0.4 and 2.1, respectively). The different individualization methods resulted in
232 distinct match outputs for HSR ($p < 0.001$, ES = 2.6 for MAS vs. ABS, ES = 1.6 for MAS vs.
233 MSS, ES = 0.8 for MSS vs. ABS, ES = 0.9 for MSS vs. LOCO and ES = 0.6 for MAS vs.
234 LOCO). Sprinting distance was lower when using ABS and MSS than when adopting the MAS
235 ($p < 0.001$, ES = 1.3 and 1.4, respectively) and LOCO quantification ($p < 0.001$, ES = 2.6 and
236 2.8, respectively). Additionally, lower sprinting distance was covered when using MAS vs.
237 LOCO ($p < 0.001$, ES = 1.0).

238

239 ***Insert Table 2 here***

240

241 ***Time-course of PRS***

242 The PRS was lower ($p < 0.001$) at Post (3.8 ± 1.32 , 95% CI = 3.6 to 4.2, ES = 3.1), G+24H (5.2
243 ± 1.48 , 95% CI = 4.9 to 5.6, ES = 1.8) and G+48h (6.0 ± 1.22 , 95% CI = 5.7 to 6.3, ES = 1.0)
244 compared to before the match (7.1 ± 1.05 , 95% CI = 6.8 to 7.3).

245

246 *Estimating PRS using different quantification methods*

247

248 *PRS Immediately Post-match*

249 At post-match, absolute scores of PRS were associated with the HSR individualized to MSS
250 and LOCO methods. Using MSS method, the perceived recovery post-match was positively
251 associated (Table 3) with the total distance of sprinting ($\text{Sprint}_{\text{MSS}}$, Beta = - 1.53, 95% CI =
252 0.08 to 2.98, $p = 0.039$). Using LOCO method, the perceived recovery at post-match was
253 negatively associated (Table 3) with high-speed running distance (HSR_{LOCO} , Beta = - 1.73, 95%
254 CI = - 3.22 to - 0.23, $p = 0.027$). LOCO was the best method with the lowest AIC. At post-
255 match, the percentage change in PRS was not associated with any of the individualization
256 methods.

257 ***Insert Table 3 here***

258 *PRS 24 hours Post-match*

259 There was a negative association between match HSR individualized for MAS (HSR_{MAS} , Beta
260 = -1.74, 95% CI = - 3.22 to - 0.26, $p = 0.022$) and LOCO (Beta = - 2.37, 95% CI = - 3.86 to -
261 0.89, $p = 0.003$), with the absolute PRS score (Table 3) at 24-hours post-match recovery period.
262 This outcome was also negatively associated with the THSR individualized for LOCO
263 ($\text{THSR}_{\text{LOCO}}$, Beta = - 2.08, 95% CI = - 3.22 to - 0.95, $p = 0.001$). HSR_{LOCO} was found to be the
264 preferred method due to the lowest AIC compared to HSR_{MAS} and $\text{THSR}_{\text{LOCO}}$.

265 There was negative association between HSR_{MAS} (Beta = - 0.26, 95% CI = - 0.47 to - 0.04, p
266 = 0.019), HSR_{LOCO} (Beta = - 0.27, 95% CI = - 0.48 to - 0.05, $p = 0.016$), $\text{THSR}_{\text{LOCO}}$ (Beta = -
267 0.21, 95% CI = - 0.36 to - 0.06, $p = 0.009$) with the percentage of change of PRS scores at
268 G+24H (Table 3). $\text{THSR}_{\text{LOCO}}$ was the best fit compared to HSR_{MAS} and HSR_{LOCO} to estimate
269 variations in PRS at G+24H.

270 *PRS 48 hours Post-match*

271 Among the different factors, HSR_{MAS} (Beta = - 1.68, 95% CI = - 2.90 to - 0.47, $p = 0.008$),
272 HSR_{LOCO} (Beta = - 1.89, 95% CI = - 3.02 to - 0.75, $p = 0.002$), $\text{THSR}_{\text{LOCO}}$ (Beta = - 1.32, 95%
273 CI = - 2.20 to - 0.44, $p = 0.004$) were negatively associated with the PRS scores at G+48H
274 (Table 3). HSR_{LOCO} emerged as the best model compared to HSR_{MAS} and $\text{THSR}_{\text{LOCO}}$.

275 Other factors were also negatively associated with the percentage of change of PRS scores at
276 G+48H (Table 3) such as; HSR_{MAS} (Beta = - 0.19, 95% CI = - 0.35 to - 0.04, $p = 0.016$),
277 $\text{Sprint}_{\text{LOCO}}$ (Beta = - 0.34, 95% CI = - 0.6 to - 0.08, $p = 0.004$), $\text{THSR}_{\text{LOCO}}$ (Beta = - 0.013, 95%
278 CI = - 0.24 to - 0.02, $p = 0.017$) and HSR_{ABS} (Beta = - 0.03, 95% CI = - 0.52 to - 0.05, $p =$
279 0.017). At this time point, $\text{Sprint}_{\text{LOCO}}$ provided the lowest AIC compared to HSR_{MAS} ,
280 $\text{THSR}_{\text{LOCO}}$ and HSR_{ABS} to estimate variations in PRS.

281

282 **DISCUSSION**

283 The aim of the study was to compare the sensitivity of different individualization methods of
284 speed zones to estimate post-match perceptual recovery in soccer players. We found that
285 different individualization methods resulted in distinct match outputs within each locomotor

286 category. Moreover, independently of the outcomes analyzed, the LOCO quantification
287 method showed to have strongest association and should be therefore primarily used to estimate
288 players' perceived recovery. Among all the different individualization procedures, the HSR
289 category was shown to reflect the external load metric with better associations across different
290 combinations of PRS. Players performing higher HSR during the match showed lower PRS
291 during the recovery period (24h and 48h). Nevertheless, a more comprehensive analyzes using
292 HSR and Sprinting distance should be applied to estimate players perceived recovery at
293 G+48H. We did not find any correlation between total distance, LSR and MSR with any other
294 outcomes independent of the time point analyzed. In addition, the widely adopted ABS method
295 did not better explain PRS response compared with other individualization approaches.

296 Previous studies have analyzed the external load in soccer training and competition using
297 different approaches based on the individual's physical capacity to customize speed zones^{16,19}.
298 It has been suggested that the use of a combination of players' physical attributes should be
299 preferred instead of the use of a single one^{19,20}. The LOCO individualization method combines
300 physical measures of MSS, MAS and ASR and have been shown to better represent the relative
301 external load experienced by the player¹⁹. The MSS has been classified as the speed at which
302 an athlete can no longer accelerate when performing an 'all out' sprinting and reflects the
303 neuromuscular capacity²⁶. On the other hand, MAS reflects the maximum aerobic capacity and
304 combines VO₂ max and running economy into a single factor³¹. Research has suggested the use
305 of both MSS and MAS to determine individuals transition to HSR and sprinting, respectively,
306 as well as to quantify external training load pattern in soccer players^{19,20}.

307 It has been consistently shown that participation in a soccer match leads to acute (less than 3
308 hours post-match) and residual (still evident up to 72 h post-match) disturbances across
309 different parameters including physical, metabolic, biochemical and perceptual^{3,15}. As such,
310 perceptual measures have been suggested to quantify mental fatigue, effort, stress, and
311 motivation; all factors that seem to be important moderators of the relationship between
312 performance and fatigue³⁶. In our study, we found a significant decrease in players' perceptual
313 recovery post-match. Moreover, it is important to highlight that the perceptual values remained
314 lower up to 48h post-match. Previous research involving soccer players reported similar time
315 course for recovery; however, these studies have adopted different perceptual scales (e.g.
316 Hooper Index, DOMS, fatigue)^{3,32}. Interestingly, the time course of the PRS observed in our
317 study followed similar pattern of objective measures often reported in the literature, including
318 biochemical (e.g. muscle micro-trauma and inflammatory markers) and neuromuscular (e.g.
319 jump ability and eccentric muscle strength) variables³. Given previous research has also shown
320 association between PRS and biochemical (e.g. CK) and neuromuscular responses to training
321 (e.g. sprint running)^{28,29}, our results reinforce the importance of perceptual assessment
322 following training and match and its applicability to monitor fatigue and recovery in soccer
323 players.

324 We have identified a greater association between the LOCO method and players' perceived
325 recovery at G+24H and G+48H. Our findings are supported by recent research involving male
326 soccer players. According to Rago et al.²², there is a moderate to large association between the
327 session rating of perceived exertion and external training load when it is adjusted to individual
328 fitness capacities. However, it is important to highlight that the reported correlation between
329 internal load (e.g. RPE and heart-rate indices) and player's individualized external load has not
330 been confirmed by other studies⁹. Within this context, it is paramount to understand whether
331 specific match-play external load metrics reflect the acute and residual changes in post-match
332 perceptual recovery. This may allow practitioners to optimize training load in order to improve
333 performance and recovery capacity while minimize injury risk¹⁰. Despite the efficiency of self-

334 report perceptual measures to quantify match-related load ^{3,5}, only few studies have
335 investigated such relationship ^{8,14,15}. To our knowledge, just one study has been developed in
336 this line of research⁹. In this investigation similar within-player correlation coefficients were
337 recorded between the individualization approaches, arbitrary speed threshold and subsequent
338 day wellness ratings of fatigue and soreness⁹. The different results observed between the
339 present and the aforementioned study may be associated to several methodological aspects.
340 This includes differences in gender (male vs female), age (young vs adult), training scenarios
341 (official matches during in-season period vs training camp without matches), perceptual
342 measures (PRS vs fatigue and soreness), criteria used for entry the thresholds and different
343 approaches adopted to characterize the external load data. Hence, future research is warranted.

344 We have identified that HSR displayed better association with players perceived recovery
345 when the LOCO method was adopted. Although this association was evident throughout the
346 recovery phase (post, G+24 H and G+48H), we observed higher association values at G+24 H.
347 We did not find any correlation between the total distance covered, LSR and MSR amongst
348 any outcome analyzed independently of the time point. The trend of our results followed a
349 similar pattern of a recent systematic review with meta-analysis, which has reported strong
350 correlation between HSR ($> 5.5 \text{ m s}^{-1}$) and both biochemical and neuromuscular fatigue-related
351 makers ¹⁰. According to the authors, CK activity increased by 30% for every 100 m of HSR
352 running distance covered during a soccer match, while a decrement of 0.5% was observed for
353 CMJ_{PPO} at G+24 H ¹³.

354 We also observed that HSR was significantly different across the individualization methods of
355 speed zone, which might explain why LOCO exhibited better association compared to the other
356 variables. In fact, the LOCO resulted in 27 and 47% more distance covered at HSR compared
357 to MSS and ABS threshold respectively. On the other hand, it is important to highlight that the
358 LOCO resulted in 16% less distance covered at HSR compared with MAS individualization
359 method. The use of MSS *per se* may reduce the sensibility of external load individualization
360 process as it produces lower associations with external load measures and erroneous
361 interpretations of training load ^{9,19}. **This was further confirmed by our study where we found a
362 positive association between HSR_{MSS} and PRS (higher HSR_{MSS} resulted in a better perceived
363 recovery post-match).** On the other hand, as the LOCO method incorporates MAS (e.g. an
364 aerobic and anatomical trait dependent physical attribute) and MSS (e.g. a measure of
365 neuromuscular capacity), may allowed a better profiling of the player's phenotype; covering a
366 higher spectrum of fitness determinants thus more representative of the body functional
367 limits²⁰. Subsequently, this may result in an improved ability of the LOCO method to determine
368 the dose response relationship (external load reflecting the internal stress) in soccer matches
369 compared to MAS *per se*.

370 It is important to highlight that the physical fitness of the players may have changed over the
371 period of the present study and this is a limitation to consider. Nevertheless, improvement in
372 physical qualities was not the central aim of the training program during the period of data
373 collection (i.e. two months of competitive period). Another potential limitation of the study
374 was the small sample size (i.e. players belong to a single team) which unable generalization of
375 our results. Additionally, in order to estimate the degree of post-match fatigue, the inclusion of
376 individual acceleration thresholds is also recommended. Finally, to increase the buy-in of the
377 coaching staff and players we applied a sport-specific endurance test (e.g. intermittent nature).
378 This option resulted in the estimation of MAS from the 30-15 test and this obviously is not the
379 "gold standard" measure to assess MAS.

380

381 PRACTICAL APPLICATION

382 The current findings have direct application for practitioners involved in the area of training
383 load monitoring in soccer. Our results suggest the utilization of customized speed zones to
384 interpret players dose response. This approach may inform decision making on training load
385 management and recovery state. Amongst different individualization methods, PRS showed
386 better association with LOCO, particularly with external load metrics such as HSR and
387 Sprinting. In addition, different individualization methods result in distinct match outputs with
388 special evidence to the HSR. We also observed that a single physical capacity may overestimate
389 or underestimate players external load responses. Finally, we suggest the utilization of the
390 LOCO method for individualizing speed thresholds. It has also an advantage of being
391 comprised by field-based measures only, resulting in higher ecological validity, economical
392 and practical approach.

393 CONCLUSION

394 This is the first study that has assessed the dose-response relationship between a range of speed
395 zones individualization methods and the PRS in youth soccer players. According to our results,
396 the PRS is a cost-effective method to monitor perceptions of recovery and seems to be sensitive
397 to detect changes in the time course response following a soccer match-play. Furthermore, the
398 present study provides support for the utility of the LOCO quantification method and external
399 load measures such as HSR and Sprinting to estimate players' perceived recovery. Finally,
400 amongst the different individualization methods, LOCO showed to be more sensitive to
401 characterize the match intensity distribution.

402

403

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479

For Peer Review

Table 1. Classification of **speed** zones for different methods to determine the match-play intensity distribution.

	MAS	MSS	LOCO	ABS
LSR	< 79 % MAS	< 49 % MSS	< 79 % MAS	< 14.3 km/h
MSR	80 - 99 % MAS	50 - 59 % MSS	80 - 99 % MAS	14.4 – 19.7 km/h
HSR	100 - 139 % MAS	60 - 79 % MSS	100 % MAS – 29% ASR	19.8 – 25.1 km/h
Sprinting	≥ 140 % MAS	80 - 100 % MSS	30 % ASR - MSS	≥ 25.2 km/h
THSR	≥ 100 % MAS	≥ 60 % MSS	≥ 100 % MAS	≥ 19.8 km/h

LSR- low-**speed** running; **MSR**- moderate-**speed** running; **HSR**- high-**speed** running; **THSR** – total high **speed** running; **MAS**- maximal aerobic speed; **MSS** – maximal sprinting speed - **MAS** and **MSS** as measures of locomotor capacities; **ABS**- absolute

Table 2- Distances covered in meters (mean \pm SD) by different **speed** zones and quantification methods during match-play.

	MAS	ABS	MSS	LOCO
LSR	6635 \pm 888 ^{b,c}	7141 \pm 777 ^{a,c,d}	7470 \pm 837 ^{a,b,d}	6635 \pm 888 ^{b,c}
MSR	1011 \pm 208 ^{b,c}	1176 \pm 268 ^{a,c,d}	677 \pm 199 ^{a,b,d}	1011 \pm 208 ^{b,c}
HSR	961 \pm 242 ^{b,c,d}	430 \pm 161 ^{a,c,d}	587 \pm 234 ^{a,b,d}	807 \pm 257 ^{a,b,c}
Sprinting	259 \pm 148 ^{b,c,d}	94 \pm 59 ^{a,c}	99 \pm 64 ^{a,c}	410 \pm 142 ^{a,b,c}
THSR		8839 \pm 1008		

LSR- low-**speed** running; MSR- moderate-**speed** running; HSR- high-**speed** running; THSR – total high **speed** running; MAS: maximal aerobic speed; MSS: maximum sprint speed; ABS: absolute threshold; LOCO: locomotor speed zones incorporating MAS and MSS; a: significantly different from MAS ($p < 0.05$); b: significantly different from ABS ($p < 0.05$); c: significantly different from MSS ($p < 0.05$); d: significantly different from LOCO ($p < 0.05$)

Table 3 – Parameter estimates for predicting perceived recovery scale at post, 24h and 48h post-match using different quantification methods.

Distance covered (km) by quantification method	Perceived recovery scale					
	Absolute scores			Percentage of change		
	Post	G+24H	G+48H	Post	G+24H	G+48H
Maximal aerobic speed						
HSR _{MAS}		-1.74*	-1.68**		-25.5*	-19.2*
SPRINT _{MAS}						
<i>AIC</i>		285.0	263.5		1.2	-32.3
Absolute						
HSR _{ABS}						-2.9*
SPRINT _{ABS}						
<i>AIC</i>						-33.1
Maximal sprinting speed						
HSR _{MSS}		1.53*				
SPRINT _{MSS}						
<i>AIC</i>		288.1				
Locomotor Capacities						
HSR _{LOCO}		-1.73*	-2.37**	-1.89**	-26.7*	
SPRINT _{LOCO}						-34.4**
<i>AIC</i>		284.1	281.8	260.5	.5	-34.1
Total high speed running (THSR)						
THSR _{MAS}						
THSR _{LOCO}			-2.08***	-1.32**	-20.7**	-1.3*
THSR _{MSS}						
THSR _{ABS}						
<i>AIC</i>			290.9	262.9	.2	-31.5

*p<0.05; ** p<0.01; ***p<0.001 Dependent variable: PRS, Independent variables: HSR- high speed running; THSR -sum of HSR and sprinting; MAS- maximal aerobic speed; LOCO- locomotor capacities; MSS- maximal sprint speed; ABS- absolute. Bold parameter estimates represent best fit among the other quantification methods. Non-significant associations are not shown. AIC- Akaike's Information Criterion