

1 **Blinded Manuscript and Author Materials**

2 **Title/Subtitle**

3 **The complexity of defining and assessing the most demanding periods of play in team**

4 **sports: a current opinion**

5

6 **Running heading**

7 **Defining and assessing the most demanding periods**

8 **Keywords**

9 worst-case scenarios; peak periods; training load; match analysis; team sports; performance

10

11 **ABSTRACT**

12 In the context of training load monitoring, the most demanding periods of play (MDPs)
13 have increasingly caught the interest of researchers. However, the MDPs analysis is currently
14 embryonic, raising some conceptual and methodological questions. This current opinion
15 synthesizes the methods used for the MDPs analysis while highlighting conceptual and
16 methodological gaps and proposing a broader perspective on the topic. It is underlined that: 1)
17 the information available on the MDPs is mostly limited to external load (particularly running-
18 based metrics), with scarce research focused on internal load; 2) the metrics have been
19 analyzed in a univariate way, neglecting the multivariate scenarios from which the MDPs
20 emerge; 3) the MDPs are highly variable over time due to the complex interaction between
21 individual, tactical-technical and contextual factors; 4) scarce evidence is available regarding
22 the contextualization of the MDPs from a tactical-technical perspective. Thus, the MDPs would
23 benefit from cross-referencing external load with game moments and tactical actions while
24 avoiding the idea of fixed benchmarks given the inherent match-to-match variability.
25 Practitioners may consider replicating the MDPs (and their contexts) in (some?) training
26 sessions as a complementary prescription strategy (metaphorically, the cherry on top, not the
27 cake). However, the feasibility and effectiveness of such practices warrant investigation.
28

29 INTRODUCTION

30 In the last 20 years, research on match and training load monitoring has increased
31 considerably (19). In this context, several studies were published to analyze “the most intense
32 activity period (for an arbitrarily selected time frame) for a player within training or competition
33 settings” (43) has also increased (e.g., Australian rules football (11), basketball (15), futsal (14),
34 Gaelic football (24), handball (13), rink hockey (14), rugby union (10), and soccer (32, 37, 38)).
35 These periods have been defined as the most demanding periods of play (MDPs) but also as
36 worst-case scenarios (34), most demanding scenarios (15), maximal intensity period (50), peak
37 locomotor demands (3), most demanding passages (41) and peak match demands (51).
38 Regardless, all these concepts aim to identify the most intense physical activities experienced
39 during match-play (32, 51). The match is broken down into shorter timeframes of fixed length
40 (51), and different lengths can be applied (e.g., 10s – 10min) (12, 34, 51). However, the concept
41 of the MDPs seems to be the most adequate to express such moments by avoiding excessive
42 emphasis on the physical factors of performance (32) and allowing a more holistic perspective
43 of match demands. Moreover, the MDPs should best be viewed as complementary to common
44 training programming and prescription strategies, i.e., they should not replace programming
45 aiming to promote general training adaptations but complement such programming.

46 Why should practitioners identify the MDPs? In soccer, the relative 90-min locomotor
47 values (expressed in m/min) represented approximately 53% of total distance (TD), 16% of
48 high-speed running distance (HSRD), and 6% of sprint distance of the 1-min peak values (38).
49 These results unfold that the distances covered during the MDPs are considerably higher than
50 the mean match values. While practitioners may look at the MDPs as valuable insights into the
51 match's demands, it is unknown whether players' performance during the MDPs will improve in
52 response to specific training regimens, given the lack of studies in this direction (32). The MDPs
53 analysis may enable the planning of training sessions (or parts of it) according to match
54 demands, setting benchmarks to be reached during training drills (15, 34, 50, 51). However,
55 some researchers have criticized this approach, underlining the lack of clarity about what the
56 MDPs represent. Again, the perspective of the current work is to consider the MDPs as a
57 complement to usual programming strategies.

58 Some current methodological limitations in the MDPs analysis (3, 32, 45) may
59 compromise its practical use for exercise prescription: (1) the MDPs analysis has relied
60 excessively on external load, neglecting internal load (32, 50); (2) only running-based metrics
61 are usually analyzed (e.g., distance covered in predefined speed zones and
62 accelerations/decelerations [Acc/Dec]) (32, 51); (3) the MDPs analysis is usually univariate
63 (identifying the *peak periods* for each variable as separate constructs), despite the likely
64 multivariate nature of the MDPs (32, 51); (4) to date, the lack of studies carried out during
65 training sessions questions the effectiveness of using the MDPs to regulate training demands
66 (32). Likewise, tactical-technical and contextual factors influence players' performance (29, 30,
67 39, 47) and should be integrated with the MDPs analysis (32). Combined with within-player
68 performance variations, there is potential for considerable match-to-match variability, whereby
69 the MDPs should best be interpreted within a dynamic and evolving range of values (3, 32, 45).
70 In brief, these shortcomings mean that different researchers and practitioners may be referring
71 to different constructs despite using similar terminology. Limiting the analysis to a single load
72 metric is often done for convenience (e.g., GPS-related data that is readily available in most
73 cases). In par with overlooking contextual information underlying each action, this may restrict
74 the comprehension of when and how the MDPs emerge. Finally, players and teams are
75 dynamic entities, and therefore, relying on a single benchmark MDP value should be avoided.
76 Instead, these values should be seen as reference ranges, which are not set in stone and may
77 evolve (e.g., due to performance improvements resulting from adaptations to the training
78 process).

79 Thereby, practitioners could benefit from using a broader, more holistic conceptual and
80 methodological approach in the MDPs analysis. In this current opinion, the conceptual gaps in
81 the MDPs are discussed, highlighting the need to integrate physical load measures into broader
82 perspectives considering contextual and tactical-technical factors. Moreover, considering the
83 MDPs for training prescription is conceptualized as a complement to existing programming
84 strategies, not an alternative or replacement.

85

86 **THE MOST DEMANDING PERIODS OF PLAY: CURRENT CONCEPTS AND**
87 **TERMINOLOGIES**

88 Analyzing the MDPs aims to capture the most intense activities experienced during
89 match-play (32, 51). However, what do *the most intense activity periods* imply? The MDPs are
90 usually quantified based on external load measures (34, 37, 51). Training load is “a
91 multidimensional construct manifested by two causally related sub-dimensions: external and
92 internal load” (20). Thus, the MDPs should fit into this framework since, strictly speaking, it is the
93 manifestation of load within a specific time period and can be expressed by external and
94 internal parameters (19). Accordingly, the MDPs should integrate internal load measures with
95 external load metrics (32, 50) to provide a more holistic perspective. For example, for a given
96 MDP based on external physical metrics (e.g., TD), two players or the same player in a different
97 moment or scenario could present different internal loads and, consequently, different MDPs.
98 However, to the best of our knowledge, only a recent study with youth soccer players included
99 measures of internal load (heart rate [HR] expressed as average HR [HR_{ave} ; bpm] and
100 percentage of individual maximal HR [$\%HR_{max}$]) in the MDPs analysis (23).

101 Whitehead et al. (51) noted that studies on the MDPs mostly used running-based metrics
102 across different team sports, which indicates that the narrow scope of metrics included in the
103 MDPs analysis models is also a shortcoming. According to Impellizzeri et al. (19), the external
104 load measures should be specific to the nature of the exercise undertaken. This suggests that
105 the analysis models should be flexible to integrate the activities of each sport. For instance,
106 jumps are usually neglected in the analysis of many team sports, such as basketball (4) and
107 soccer (2), while limited studies, mostly with rugby teams, include collisions in the MDPs
108 assessment (49, 51). Collisions may potentially be identified by spikes in data from inertial
109 sensors in some GPS devices (e.g., Catapult Optimeye S5, Melbourne, Australia), or through
110 accelerometer- and gyroscope-derived variables (17, 31). However, without open-access details
111 regarding how these algorithms operate, it is difficult to assess how accurately they identify
112 collisions (31). Moreover, it is conceivable that any spike in data may be explained by different
113 factors, including error, and therefore, more research is warranted. Integrating external load
114 data with video analysis could potentially help assess which actions the MDPs correspond to.

115 Another shortcoming related to the analysis of the MDPs is that most models applied to
116 the MDPs analysis are univariate (10, 32), meaning a mixed model analysis is used for each
117 dependent variable (10). Analyzing the different activities in isolation fails to illustrate the
118 multiple scenarios that induce the MDPs during a specific period (cf. Novak et al. (32)). In a
119 recent exception, Kim et al. (23) used an alternative approach to identify whether external load
120 metrics (speed and acceleration data) matched an internal load metric (HR). While these
121 findings showed trivial differences between the external and internal load metrics (i.e., the effect
122 size was too small for the differences to be noticeable), this was a single study (thereby
123 requiring replication) and HR is only one of many possible internal load metrics. Moreover, this
124 still does not account for the context, i.e., when and how these MDPs occurred.

125 In summary, current MDPs concepts and analyses seem too narrow, denoting (i) a lack of
126 attention devoted to internal load measures, (ii) external load metrics limited to a restricted
127 number of variables (mostly running-based metrics); and (iii) univariate metrics analysis,
128 overlooking the interaction of the various physical factors that lead to the MDPs. However, the
129 concept of the MDPs can be expanded to provide a more holistic (and potentially useful) set of
130 information that practitioners can act upon.

131

132 **HOW ARE THE MOST DEMANDING PERIODS OF PLAY BEING ASSESSED?**

133 ***CURRENT QUANTIFICATION METHODS***

134 Methods such as fixed length and rolling average have been used to capture MDPs (34,
135 41). The fixed-length method implies splitting the match, from start to finish, into periods of a
136 fixed duration (e.g., 0-59", 1'-1.59") (34). Then, to determine the MDPs for each metric, the
137 period with the highest value is chosen (51). In turn, the rolling average requires the
138 instantaneous analysis of raw data, with the number of samples obtained per time unit
139 depending on the frequency of the device being used (10). For example, a GPS with a 10 Hz
140 sampling rate takes 600 samples for 1 min (10, 51). The MDPs can be estimated from the
141 beginning to the end of the match by an algorithm using the current sample and the preceding
142 599 samples (e.g., 0-600, 1-601, 2-602, 3-603) (10, 34, 51). Regardless of the method used
143 and the time window chosen, the MDPs can be identified by analyzing the whole match or the
144 longest period when the ball is in play (BiP), i.e., taking into account the time in which play is

145 ongoing before the ball leaves the pitch or the referee stops the play (41, 51). BiP has been
146 proposed as an alternative to analyzing the whole match, and it has been suggested as more
147 accurate in quantifying the match's demands (28, 48). The authors claim that BiP enables
148 practitioners to better perceive work-to-rest periods, valuable information for training prescription
149 (28, 48).

150 Both methods (fixed length vs. rolling average) have been compared in the literature,
151 considering different period lengths (1, 3, 5, and 10 min), and it is noted that the fixed length
152 method usually underestimates the MDPs, regardless of the physical metric or the playing
153 position analyzed (12, 34). Ferraday et al. (12) reported that the fixed length method
154 underestimates the MDPs for HSRD and TD in ~12-25% and ~7-10%, respectively, compared
155 to the rolling average. The longer the MDPs, the greater the observed differences between
156 methods (34). This can be explained because, given the unpredictable nature of team sports,
157 the MDPs are likely to not occur within the exact predefined and fixed periods (16). In this
158 sense, the rolling average is recommended in the assessment of the MDPs since it can better
159 detect intensity fluctuations (12, 34, 51). Within the rolling average method and considering the
160 same period length (1 min), differences were also observed in the MDPs when analyzing the
161 whole match or BiP (41). The whole match approach showed higher MDPs values for HSRD (48
162 vs. 36 m/min) and Acc/Dec (32 vs 20 m/min), while BIP showed higher values for TD (187 vs
163 293 m/min) (the values were extracted from figure 3 of the study conducted by Riboli et al. (41),
164 using the WebPlotDigitizer, version 4.6., developed by Ankit Rohatgi). The rapid tactical
165 adjustments by the players when the match stops could be the reason for the increase in
166 locomotor demands (41). It is important to note that the differences between approaches for
167 MDPs longer than 1 min are unknown and could possibly be time dependent. Moreover, BiP is
168 a time-consuming analysis technique, so further studies comparing both techniques are needed
169 to understand whether the cost-benefit is worthwhile.

170 While the MDPs are often analyzed based on measurable data from a single monitoring
171 tool (external and/or internal load), the so-called "Fine-Tuning" approach proposed by Boullosa
172 et al. (7) could allow improving the diagnosis and predictive capacity of monitoring practices.
173 This approach is defined as "the combined use of different monitoring tools (either objective or
174 subjective and external or internal) that practitioners experience" (7). Importantly, the fine-tuning

175 approach is a conceptual framework, not a specific, validated tool. In the case of the MDPs, the
176 fine-tuning approach could involve combining objective data from external and internal load
177 (e.g., the rolling average to identify the MDPs for both acceleration data and HR) with subjective
178 data (e.g., rating of perceived exertion), while integrating video footage (see section 5) to
179 provide context for the MDPs. This is currently a mere proposal and requires further research to
180 establish its feasibility and effectiveness.

181

182 **PERIOD LENGTHS FOR ASSESSMENT**

183 The MDPs are analyzed across incremental time intervals (e.g., 10 s – 10 min) (49). For
184 example, periods of 1, 3, 5, and 10 min in soccer are commonly analyzed (42). As expected, as
185 the analyzed period length increases, the relative intensity (distance covered per minute) tends
186 to decrease and vice-versa (25, 41). In Australian football, Delaney et al. (11) found that the
187 MDPs values reached a plateau in periods longer than 5-7 min (depending on the variable
188 chosen). This phenomenon can be explained by the players' being unable to physiologically
189 maintain the same intensity for a long period of time (51). The higher the probability of a break
190 in play (score, stoppage of time by the referee, or the ball going out of play) as the period
191 analyzed increases could also be behind the decrease in relative intensity (51).

192 To clarify the ideal period lengths to be analyzed, recently, the temporal self-containment
193 of shorter peak match running periods within longer windows was analyzed by Thoseby et al.
194 (46). Owing to the study's low to moderate levels of self-containment, the authors highlight the
195 need for athletes to train the MDPs across varying durations (46). In general, the MDPs of
196 shorter duration do not coincide with those of longer duration. Thus, the period length could be
197 set considering the adaptations practitioners would like to elicit in players (20).

198

199 **MATCH-TO-MATCH VARIABILITY**

200 Several studies have shown that match-to-match variability in physical performance is a
201 natural phenomenon in team sports (3, 8). In this regard, a recent study in women's soccer
202 showed that the match-to-match variability (represented by the coefficient of variation [CV]) was
203 higher in the 1-min peak periods compared to the absolute values of the entire match, for TD
204 (6.5% vs. 4.6%), HSRD (18.7% vs 15.9%) and Acc/Dec (12.9% vs 11.7%) (3). Other studies

205 have also reported high variability in the MDPs (32, 45). Figure 1 illustrates an example of a
206 soccer player's MDPs match-to-match variability for TD, HSRD, and sprint distance.

207 However, the match-to-match variability can provide useful insights to identify the
208 minimum and maximum MDPs values calculated using the CV, and consequently set a range of
209 values for the training prescription (45), i.e., instead of the idea of fixed benchmarks, the MDPs
210 can be understood under a dynamic framework. This approach can also be important for
211 identifying changes that appear normal and those that are lower or higher than expected (36) to
212 identify the reasons that led to such values. Identifying a range of values in which the MDPs
213 occur also allows quantifying the number of MDPs events players experience during the match.
214 Importantly, training adaptations during the season may potentially change the range values,
215 and it is possible that the upper limit of such range increased. In competitions with multiple
216 stages, the national stage may impose greater demands than the regional stage, which may
217 also impact the range of the MDPs observed in the matches.

218 ***Figure 1 near here***

219

220 **WHAT ARE THE TACTICAL-TECHNICAL BEHAVIORS BEHIND THE MOST DEMANDING** 221 **PERIODS OF PLAY?**

222 On this topic, the MDPs analysis has been developed in a blind context, without
223 considering the tactical behaviors that support the emergence of such periods (32).
224 Contextualizing the MDPs according to individual and collective tactical issues is necessary to
225 help practitioners create sport-specific drills and provide meaningful stimuli (5), i.e., knowledge
226 of context is paramount for a holistic understanding of the role of the MDPs. Although most
227 studies are limited to quantifying the MDPs (1, 10, 12, 32-35, 37), other studies have recently
228 attempted to contextualize them by synchronizing with video footage (21, 41). This involves
229 identifying the MDPs (e.g., rolling average) and simultaneously observing the tactical behavior
230 of the players during those timeframes (21). The tactical actions can be coded in line with a
231 systematic, integrated approach to quantifying match physical-tactical performance developed
232 by Ju et al. (22). Riboli et al. (41) showed that the MDPs occurred in different match phases
233 according to the players' positions. For example, without ball possession, the MDPs were higher
234 for central defenders (CD) (TD, HSRD, and Acc/Dec) and wide defenders (WD) (HSRD and

235 Acc/Dec) (41). Conversely, higher MDPs were observed for wide midfielders (WM) (Acc/Dec),
236 wide forwards (WF) (TD, HSRD, Acc/Dec), and forwards (FW) (TD and HSRD) when the team
237 possessed the ball, showing that different positions face different physical demands due to
238 tactical adjustments depending on the match phases (e.g., attack or defense) (41).

239 Previous research (21) revealed that the contextualized data during the MDPs are
240 position-dependent and that individual and collective tactical actions should be considered in the
241 analysis of these periods. For instance, the principal tactical actions for the CD during the MDPs
242 were *Covering* and *Recovery Run* (see detailed description in Ju et al. (21)). Both tactical
243 actions occurred without possession, and the data coincide with the findings of Riboli et al. (41).
244 *Close Down/Press*, *Run in Behind/Penetrates*, and *Support Play* were the key tactical actions for
245 FW (21). Note that in the study by Ju et al. (21), MDPs were only analyzed for the relative
246 distance above 5.5 m/s, not including metrics such as TD, sprint distance, and Acc/Dec.
247 Questions such as "Can the MDPs vary based on a team's model and game plan?" or "Can the
248 MDPs vary depending on how the models and game plans of the two opposing teams interact?"
249 are worth investigating in the future. In short, MDPs contextualization has the advantage of
250 enabling practitioners to be aware of the tactical actions where they emerge, facilitating the
251 design of training drills that replicate these actions based on the players' position, metrics, and
252 other potentially relevant factors. An expansion of the concept of MDPs to include the tactical-
253 technical context of their emergence may be required to avoid a reductionist approach to
254 understanding match demands and to engage in training prescription.

255

256 **INTEGRATING ADDITIONAL CONTEXTUAL FACTORS**

257 Additional contextual factors may influence the MDPs values (9), and should therefore be
258 considered when interpreting them (and when using them to prescribe training stimuli). Different
259 studies have analyzed the effects of factors such as the players' position, match location
260 (matches are divided into home or away), match outcome (defined as the final score of the
261 match), match status (defined as the momentary result during the match), starter vs non-starter,
262 among others in performance (1, 6, 16, 32, 37). For example, in soccer, the CD show lower
263 values than the other positions in TD, HSRD, and sprint distance (1, 37, 41). In turn, the central
264 midfielders (CM) have the highest TD values (32, 41). These differences seem to be due to the

265 specific tactical roles of the different positions (21), suggesting that the training drills should be
266 designed according to the specificity of each position (16).

267 Regarding the match location, the MDPs in away matches are higher than in home
268 matches (1, 37). In contrast, González-García et al. (16) found no significant difference between
269 home and away matches. It is important to note that the studies mentioned are limited to the
270 context and characteristics of the samples, constraining conclusions about practical application.
271 The match outcome and status also affect the MDPs (1, 37). Winning the match (i.e., match
272 outcome is defined as the final result of the match) has a positive impact (greater TD, HSRD,
273 and sprint distance) on the MDPs compared to drawing or losing when considering 1- and 3-min
274 periods (37). Concerning the match status (defined as the momentary result during the match),
275 MDPs seem to be higher when the score is a draw compared to winning or losing (1). Finally,
276 larger TD values were observed for non-starter players (32).

277 Therefore, the MDPs analysis could integrate the contextual factors, helping the
278 practitioners to identify the sources of variability of the MDPs, to improve the understanding of
279 these specific variables for practice and research. This raises additional questions, such as: (i)
280 can the MDPs vary depending on the presence of congested fixtures?; (ii) can different
281 competitive settings (domestic versus international competitions) influence the MDPs?; (iii) how
282 does travel affect the MDPs?; (iv) may weather conditions (especially temperature and relative
283 humidity) affect the MDPs?; and so on. Such gaps in knowledge limit our understanding of
284 MDPs emergence and, therefore, may result in MDPs being poorly replicated in training
285 sessions (i.e., divorced from their context).

286

287 **CAN THE MOST DEMANDING PERIODS HELP TO INFORM TRAINING PRESCRIPTIONS?**

288 Despite their current popularity, the MDPs should not be viewed as a panacea that will
289 resolve the problems of load quantification or revolutionize programming – they should best be
290 considered (cautiously) as potentially an extra piece of the puzzle, complementing the existing
291 strategies focused on progressive improvement of performance and/or recovery. Due to the
292 limited number of studies analyzing the MDPs occurrences within training, doubts have arisen
293 about how they could be applied in practice. The frequency, timing, and context of the MDPs
294 during training sessions, how they are distributed throughout the microcycle, and which drills

295 help to generate game-like MDPs are topics that represent valuable research opportunities. If,
296 on one hand, the MDPs (TD, HSRD, sprint distance, and Acc/Dec) are not always hit during
297 training sessions over the week (6, 15), on the other hand, exercises such as small-sided (SSG)
298 and large-sided games (LSG) might surpass the MDPs match values for Acc/Dec and sprint
299 distance, respectively (6, 26). For instance, Sansone et al. (43) developed a study on basketball
300 with youth players and used only one measure of external load (Impulse Load®). The authors
301 found exercises similar to the competitive context-induced higher MDPs values (43). This can
302 be explained because these drills replicate competition-like tactical scenarios (43).

303 Interestingly, the MDPs are usually higher in the training sessions corresponding to
304 match day minus 4 (MD-4) and minus 3 (MD-3) compared to the competition (match day [MD])
305 (6, 15). These training sessions occur on the central days of the microcycle, where the training
306 load is typically higher (27, 44). In this way, the practitioners can manipulate a set of variables in
307 the training drills (e.g., drill rules, pitch size, number of players, area per player) to expose the
308 players to appropriate MDPs within the microcycle (5, 40). However, special care must be taken
309 when interpreting the data described above, as the analysis failed to consider the range of
310 values for each player (i.e., they focus on point values instead of an interval of values). If the
311 MDPs are analyzed within a specific range of values, the probability of finding differences
312 between MDPs from training sessions and matches could decrease. Moreover, as previously
313 mentioned, those ranges may oscillate across the season, and, more importantly, the contextual
314 framing and understanding of those MDPs is paramount. Otherwise, prescribing training based
315 on MDPs may result in a vain search for a “physical indicator” divorced from its tactical-technical
316 context.

317 From a practical perspective, practitioners could identify the MDPs (and their context) to
318 optimize training. By understanding the highest demands (i.e., the MDPs) and in what contexts
319 they emerge, practitioners may potentially use that information to replicate match contexts and
320 improve training prescription to better prepare players for MDPs. Such optimization means
321 adjusting the activities to be performed during training sessions (external load) to generate a
322 specific internal load (mediating mechanism) that causes an outcome of interest (causal
323 exposure-outcome), for example improving aerobic capacity (20), but this should be done under
324 an appropriate tactical-technical context (50). However, practitioners should be careful

325 regarding the dose, i.e., over-emphasizing training prescription based on the MDPs may
326 potentially surpass the intended balance between training and recovering, something that
327 should also be addressed in future research. Importantly, MDPs are not an alternative to
328 general planning and programming, aiming to improve performance while ensuring proper
329 recovery. Instead, training programs focusing on MDPs should consider the frequency, volume,
330 and intensity of actions, offering a complementary approach for practitioners to integrate these
331 factors within a holistic perspective.

332

333 **FUTURE AVENUES FOR THE MOST DEMANDING PERIODS OF PLAY**

334 The composite construct developed by Novak et al. (32) provides a clear picture of the
335 MDPs and how they should be quantified. However, to be considered scientific, this construct
336 requires three essential characteristics – a label, a constitutive definition and an operational
337 definition (18). In accordance with what has been discussed throughout the article, this section
338 proposes to assign these qualities to the construct. As mentioned earlier, the MDPs would be
339 the construct label, given the holistic overview it presents. From a theoretical standpoint
340 (constitutive definition), the MDPs could be defined as the most intense periods that respond to
341 the external load and are influenced by complex interactions between individual, tactical-
342 technical, and contextual factors (instead of the prevalent, more reductionist approaches). For
343 practical applications (operational definition), the purpose of the MDPs is to capture the periods
344 of the match experienced by the players that fall within a range of values defined using a
345 multidimensional and dynamic approach through external and/or internal load measures, being
346 quantified by methods such as fixed length, or rolling average.

347

348 **PRACTICAL APPLICATIONS**

349 The main purpose of identifying the MDPs is to improve practitioners' knowledge of the
350 match's demands, and, consequently, to transfer the insights into practice. Importantly,
351 practitioners may still use their usual programming strategies – the MDPs are merely a
352 complement and should not blind practitioners to other goals (e.g., long-term learning and
353 training adaptations). Practitioners should be aware that the MDPs identification involves
354 measures of external and internal load, while also considering their tactical-technical context of

355 emergence. Accordingly, the rolling average appears to be a suitable method to identify the
356 MDPs. Nevertheless, studies analyzing the MDPs in a multivariate way are lacking, so
357 identifying the minute of the match in which the MDPs occur could be an alternative approach to
358 understand which activities lead to the highest internal response. Crossing these data with
359 video analysis, practitioners can better understand the contexts under which the MDPs emerge
360 (e.g., the when, how, and why). The match-to-match variability should also be assessed, and it
361 is suggested that practitioners use a database to record the MDPs so that the evolving range of
362 values can be monitored over time (updated every match). The linear mixed model can be a
363 method of statistical analysis to set the range of values based on CV, and to identify whether
364 the values of a given match are higher or lower than expected in comparison with the
365 retrospective match values, the method proposed by Oliva-Lozano et al. (36) can be used.

366 While hard conclusions are difficult to draw, it seems reasonable that practitioners may
367 wish to design parts of training sessions to ensure players hit the MDPs, considering both
368 *ingredients*: intensity and frequency. The use of the MDPs to prescribe parts of the training
369 sessions (or drills) should not, however, replace the core concept of designing training to
370 achieve specific adaptations, especially if non-contextualized MDPs values are used.
371 Considering the MDPs in this context is akin to putting the cherry on top of the cake – training
372 should first and foremost focus on the adaptations for the most common demands of a match,
373 and only secondly worry about the MDPs, which will occur only sparsely. Still, preparing the
374 players to face such demands may be important, not only from a performance perspective, but
375 also from an injury prevention perspective.

376 From a load management perspective, it could make sense they take place in the middle
377 of the week, as the training load tends to be higher on these days and decreases on the days
378 before the following match. However, this may vary depending on the momentary goals for the
379 microcycle and the stage of the competition. Some teams may, depending on how the
380 competition is organized, implement the MDPs more frequently in the microcycle. The
381 programming of the MDPs intensity and frequency for each training session weekly should be
382 tailored in accordance with the range of values set for the individual players. Finally, though it
383 may sound conflicting with what has been discussed in this article, univariate analysis can be
384 useful in specific settings when prescribing training (e.g., metrics such as sprint distance). For

385 example, in exercises that replicate match situations but are focused solely on speed. However,
386 this should be implemented with caution, to avoid an overly reductionist approach to training.
387 The introduction of video footage can assist practitioners to identify behavior patterns according
388 to players' positions during the MDPs, to be replicated during training drills. Figure 2
389 synthesizes how the MDPs can possibly be identified and applied in training sessions. It was
390 structured logically, starting from the assessment to the prescription of the MDPs.

391

392 ***Figure 2 near here***

393

394 **CONCLUSIONS**

395 Research on the MDPs would benefit from a broader conceptual and methodological
396 perspective, integrating internal load measures and exploring multivariate analysis. In addition
397 to the MDPs assessment for different external and internal load metrics, research should
398 integrate video analysis to contextualize the MDPs from a tactical-technical perspective and
399 avoid a reductionist approach to the MDPs. This way, practitioners can apply the analyzed data
400 to help monitor and develop certain training drills, allowing a better benchmark between the
401 training and match situations. However, given its limited nature and current limited research, the
402 MDPs should be considered a mere complement to usual training practices. Practitioners
403 should also interpret the MDPs dynamically to implement an evolving benchmarking process,
404 considering how it may change over time (e.g., match-to-match variability, within-season
405 evolution). Above all, the MDPs should not be looked at as a “single” approach to use to
406 monitoring load but rather as one element within a holistic framework that may provide valuable
407 insights for training prescription.

408

409 **Figure captions**

410 Figure 1. An example of a soccer player's 3-minute MDPs match-to-match variability for TD (A),
411 HSRD (B) and sprint distance (C) across thirty matches. The figure also shows the z-scores (D)
412 for each variable, so that zero represents the mean and one represents the standard deviation.
413 The z-scores are useful for comparing different variables, as well as for easier interpretation and
414 comparison of the MDPs performance on an individual basis. The data was generated in
415 RStudio (version 2023.12.0+369 "Ocean Storm", Boston, Massachusetts) based on the CV
416 reported by Novak et al. (32) for TD (6.2%), HSRD (25.2%) and sprint distance (46.1%).

417
418 Figure 2. Synthesis of how to analyze the most demanding periods of play (MDPs) and use
419 them in training prescription. The graph represents the match, and the red dots depict the
420 MDPs, demonstrating that they are higher than the mean values for the whole match. The
421 dialog boxes highlight the factors that influence the MDPs, the measures and the methods used
422 to identify the MDPs, the match-to-match variability, and the training prescription guidelines.

423

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