

Integrating physical and tactical factors in football using positional data: a systematic review

José Eduardo Teixeira^{1,2,3}, Pedro Forte^{1,2,4}, Ricardo Ferraz^{1,5}, Luís Branquinho⁵, António José Silva^{1,6}, António Miguel Monteiro², Tiago M Barbosa^{Corresp. 7}

¹ Research Centre in Sports Sciences, Health and Human Development, Vila Real, Portugal

² Instituto Politécnico de Bragança, Bragança, Portugal

³ Polytechnic Institute of Guarda, Guarda, Portugal

⁴ Institute of Educational Sciences, Penafiel, Portugal

⁵ University of Beira Interior, Covilhã, Portugal

⁶ University of Trás-os-Montes and Alto Douro, Vila Real, Portugal

⁷ Sport Sciences, Polytechnic Institute of Bragança, Bragança, Portugal

Corresponding Author: Tiago M Barbosa

Email address: barbosa@ipb.pt

Background: Positional data have been used to capture physical and tactical factors in football, however current research is now looking to apply spatiotemporal parameters from an integrative perspective. Thus, the aim of this article was to systematically review the published articles that integrate physical and tactical variables in football using positional data. **Methods and Materials:** Following the Preferred Reporting Item for Systematic Reviews and Meta-analyses (PRISMA), a systematic search of relevant English-language articles was performed from earliest record to August 2021. The methodological quality of the studies was evaluated using the modified Downs and Black Quality Index (observational and cross-sectional studies) and the Physiotherapy Evidence Database (PEDro) scale (intervention studies). **Results** The literature search returned 982 articles (WoS = 495; PubMed = 232 and SportDiscus = 255). After screening, 26 full-text articles met the inclusion criteria and data extraction was conducted. All studies considered the integration of physical and tactical variables in football using positional data ($n = 26$). Other dimensions were also reported, such as psychophysiological and technical factors, however the results of these approaches were not the focus of the analysis ($n = 5$). Quasi-experimental approaches considered training sets ($n = 20$) and match contexts ($n = 6$). One study analysed both training and play insights. Small sided-games (SSG) were the most common training task formats in the reviewed studies, with only three articles addressing medium-sided (MSG) ($n = 1$) and large-sided games (LSG) ($n = 2$), respectively. **Conclusions:** Among the current systematic review, the physical data can be integrated by player's movement speed. Positional datasets can be computed by spatial movement,

complex indexes, playing areas, intra-team and inter-team dyads. Futures researches should consider applying positional data in women's football environments and explore the representativeness of the MSG and LSG.

1 **Title page**

2 **Manuscript Title:**

3 Integrating physical and tactical factors in football using positional data: A Systematic Review.

4 **Running heading:**

5 Physical, physiological and tactical factors using positional data (Football).

6 **Authors:** José E. Teixeira^{1,2,3}, Pedro Forte^{1,2,5}, Ricardo Ferraz^{1,5}, Luís Branquinho⁵, António J.
7 Silva^{1,6}, António M. Monteiro^{1,2} and Tiago M. Barbosa^{1,2}

8 ¹ Research Centre in Sports Sciences, Health and Human Development, 5001-801 Vila Real, Portugal

9 ² Department of Sport Sciences and Physical Educations, Instituto Politécnico de Bragança, 5300-253 Bragança, Portugal;
10 jose.eduardo@ipb.pt (J.E.T.); barbosa@ipb.pt (T.M.B.); mmonteiro@ipb.pt (A.M.M.)

11 ³ Department of Sport Sciences, Polytechnic Institute of Guarda, 6300-559 Guarda, Portugal

12 ⁴ Department of Sport Sciences, Department of Sports, Douro Higher Institute of Educational Sciences, 4560-708 Penafiel,
13 Portugal; pedromiguel.forte@iscedouro.pt (P.F.); luis.branquinho@iscedouro.pt (L.B.)

14 ⁵ Department of Sports Sciences, University of Beira Interior, 6201-001 Covilhã, Portugal; rmpf@ubi.pt (R.F.)

15 ⁶ Department of Sports, Exercise and Health Sciences, University of Trás-os-Montes e Alto Douro, 5001-801 Vila Real, Portugal;
16 ajsilva@utad.pt (A.J.S.)

17 **Corresponding author:**

18 **E-mail:** barbosa@ipb.pt (T.M.B.)

ORCID: 0000-0001-7071-2116

19 **Word count (*Abstract*):** 308 words.

20 **Word count (*Text, including tables & figures*):** 9181 words.

21 **Word count (*Text, excluding tables & figures*):** 4213 words.

22 **Number of tables:** 6 (six).

23 **Number of figures:** 1 (one).

24 **Declarations**

25 **Funding:** This research was supported by Portuguese Foundation for Science and Technology,
26 I.P. (project UIDB04045/2021).

27 **Author contributions:** Conceptualization, J.E.T. and P.F.; Data curation, J.E.T., R.F. and P.F.;
28 Formal analysis, P.F., R.F., and L.B.; Funding acquisition, T.M.B.; Investigation, J.E.T., L.B.;
29 Methodology, P.F., L.B. and T.M.B.; Resources, P.F. and A.M.M.; Software, J.E.T., L.B. and
30 R.F.; Supervision, A.J.S. and A.M.M.; Validation, P.F., T.M.B. and R.F.; Writing – original draft,
31 J.E.T.; Writing – review & editing, P.F., R.F., L.B., A.J.S., T.M.B. and A.M.M. All authors have
32 read and agreed to the published version of the manuscript.

33 **Conflicts of Interest:** The authors declare no potential conflict of interest.

34 **Acknowledgements:** The authors express acknowledgement to the authors who provided
35 additional information for the studies included in this systematic review.

36

37 **Abstract**

38 **Background:** Positional data have been used to capture physical and tactical factors in football.
39 However, current research is now looking at applying spatiotemporal parameters from an
40 integrative perspective. The aim of this study was to systematically review the published articles
41 that integrate physical and tactical variables in football using positional data.

42 **Methods and Materials:** Following the Preferred Reporting Item for Systematic Reviews and
43 Meta-analyses (PRISMA), a systematic search of relevant English-language articles was
44 performed from earliest record to August 2021. The methodological quality of the studies was
45 evaluated using the modified Downs and Black Quality Index (observational and cross-sectional
46 studies) and the Physiotherapy Evidence Database (PEDro) scale (intervention studies).

47 **Results** The literature search returned 982 articles (WoS = 495; PubMed = 232 and SportDiscus
48 = 255). After screening, 26 full-text articles met the inclusion criteria and data extraction was
49 conducted. All studies noted the integration of physical and tactical variables in football using
50 positional data ($n = 26$). Other dimensions were also reported, such as psychophysiological and
51 technical factors. However, the results of these approaches were not the focus of the analyses ($n =$
52 5). Quasi-experimental studies focused on training sets ($n = 20$) and match settings ($n = 6$). One
53 study analysed both training and playing settings. Small sided-games (SSG) were the most
54 common training task formats in the retained studies, with only three articles addressing medium-
55 sided (MSG) ($n = 1$) and large-sided games (LSG) ($n = 2$), respectively.

56 **Conclusions:** Based on the current systematic review, physical data can be integrated by player's
57 movement speed. Positional datasets can be computed by spatial movement, complex indexes,

58 playing areas, intra-team and inter-team dyads. Future research should consider applying
59 positional data into women's football and, explore the representativeness of the MSG and LSG.

60 **Keywords:** tracking systems, movement, complexity, training, match

61 **1. Introduction**

62 Football can be characterized as a complex and dynamical system where the players collect
63 ecological information to make decisions, allowing them to gather numerical and spatial advantage
64 through the phases of play with a goal-orientation (Duarte et al., 2012; Folgado et al., 2014; Hewitt
65 et al., 2016). Hence, football players carry out intermittent movements to perform individual and
66 collective tactical actions (Clemente et al., 2020; Duarte et al., 2012; Low et al., 2019). Tracking
67 systems have been used to compute spatiotemporal measures and assess players' positions during
68 training and match settings (Lames et al., 2010; Marcelino et al., 2020; Pol et al., 2020). Positional
69 data can be captured at different frequencies by tracking systems, such as, global navigation
70 satellite systems (GNSS) or global positioning systems (GPS) (Beato et al., 2018; Rago et al.,
71 2019; Teixeira et al., 2021), local radio-based local positioning (LPM) (Hoppe et al., 2018; Leser
72 et al., 2011; Ogris et al., 2012) and computerized-video or optical-based tracking systems (Beato
73 & Jamil, 2018; Castellano et al., 2014; Di Salvo & Marco, 2006). The players and ball positioning
74 can be computed by Cartesian and Euclidian coordinates (xx, yy) contextualizing the physical
75 demands on the tactical behaviour (Carrilho et al., 2020; Clemente et al., 2013; Low et al., 2020;
76 Memmert et al., 2017). However, some of the above mentioned tracking methods do not allow to
77 gather information on the player-ball-goal position (Carrilho et al., 2020; Vidal-Codina et al.,
78 2022), opponent-adaptive play strategy (Memmert, 2021; Ranjitha et al., 2020) and individual
79 tactical behavior (Laakso et al., 2022; Reis & Almeida, 2020).

80 Furthermore, tracking systems generate a large amount and variety of data that can be used
81 for performance analysis in football (Rein & Memmert, 2016; Rojas-Valverde et al., 2019).
82 Notwithstanding, it is paramount implementing multidisciplinary frameworks underpinned by
83 sports science and computer science, making use of big data methodology, new computational

84 procedures to extract, process and analyse data that yield practical information with an impact on
85 training and match performance (Rico-González et al., 2021). However, integrating players'
86 physical performance with match-related contextual factors and tactical behaviours continues to
87 be a challenge in football science (Teixeira et al., 2022). Moreover, the performance analysis in
88 football needs a multidimensional approach to capture the adaptive individual and collective
89 behaviour (Carling et al., 2014; Gonçalves et al., 2019). This multifactorial phenomenon depends
90 on the interplay of physical, tactical and technical drivers (Bradley & Ade, 2018; Paul et al., 2015).
91 Thus, performance analysis in football is now focused on applying the spatiotemporal parameters
92 from an integrative perspective (Praça et al., 2022).

93 A growing number of reviews and meta-analyses have been published on this topic and
94 focusing on training and match settings (Low et al., 2020; Rago et al., 2020; Teixeira et al., 2021).
95 Nevertheless, previous reviews have considered each performance factor independently,
96 describing tactical behaviour independently from physical demands (Coito et al., 2022; Ometto et
97 al., 2018). Therefore, it is important to understand the main methodological procedures to conduct
98 an integrative analysis of physical and tactical performance in training and match in football. Also,
99 the published studies have calculated different physical and tactical measures by tracking
100 positional data, wherefore a procedural standardization is needed to progress towards integrative
101 approaches (Teixeira et al., 2022). Thus, the aim of this study was to systematically review the
102 published articles that integrate physical and tactical variables in football using positional data.

103 **2. Methods**

104 **2.1. Literature Search strategy**

105 The literature search strategy was registered on the International Platform of Registered Systematic
106 Review and Meta-Analysis Protocols with the number 202270030 (doi:
107 10.37766/inplasy2022.7.0030). The protocol was designed in accordance with ‘The Preferred
108 Reporting Items for Systematic Reviews and Meta-Analyses’ (PRISMA) guidelines and the
109 ‘Population-Intervention-Comparators-Outcomes’ (PICOS) (Page et al., 2021). The literature
110 search was conducted on three databases: PubMed/Medline, Web of Science (WoS, including all
111 Web of Science Core Collection: Citation Indexes), and SportDiscus. The studies were searched
112 using a Boolean string with specific keywords (Table 1).

113 **[PLACE TABLE 1 HERE]**

114 The literature search was performed between April and May 2022 by an independently author
115 (J.E.T) and checked by a second author (P.F.). Discrepancies between authors in the study
116 selection were solved by a third reviewer (T.M.B). Double-check review is recommended in
117 PRISMA guidelines (Page et al., 2021). The literature search was limited to peer-reviewed articles
118 and authors did not prioritize authors or journals.

119 **2.2. Selection criteria**

120 The selection criteria followed PICOS approach: (1) *Population*: amateur, semi-professional and
121 professional football players (aged ≥ 10 years); (2) *Intervention*: integration of physical and tactical
122 measures using spatiotemporal datasets; (3) *Comparison*: physical and tactical variables; (4)
123 *Outcomes*: tracking, positional and time-series data; (5) *Study design*: original experimental and

124 quasi-experimental trials (e.g., randomized controlled trial, cohort studies or cross-sectional
125 studies).

126 On this basis, the inclusion criteria used for article selection were: (1) original article
127 focused on adult and youth football players of both sexes; (2) studies with screening procedures
128 based on physical and tactical measures using tracking and positional data; (3) studies that used
129 spatiotemporal parameters to assess physical data; (4) studies that used positional and tracking
130 data to measures spatiotemporal and tactical variables through time-series; (3) other performance
131 factors as psychophysiological, technical and contextual factors were not excluded from the
132 present review if both variables of interest (i.e., physical and tactical measures) were part of the
133 experimental design; (5) studies of human physical performance in the field of sport science; (6)
134 original articles published in peer-review journals; (7) full text available in English; (8) reported
135 sample and screening procedures (e.g. data collection, study design, instruments, and the
136 outcomes).

137 Otherwise, exclusion criteria were: (1) original articles about positional data in individual
138 sports, team sports, and other football codes (e.g. Australian Football, Gaelic Football, Union
139 and/or Seven Rugby); (2) studies that analysed none or only one of the performance drivers (i.e.
140 only physical or tactical measures); (3) studies which integrate several performance factors, but
141 did not combine the two domains of interest, even if one single driver is integrated; (4) studies that
142 measured physical outcomes by field-based or laboratory tests rather than tracking and positional
143 data; (5) studies that reported tactical variables collected by notational analysis or other
144 methodological procedures that did not assess spatiotemporal time-series; (6) others research fields
145 and non-human participants; (7) articles with poor quality in the description of study sample and
146 screening procedures (e.g. data collection, study design, instruments, and the measures) according

147 to PEDro and Downs and Black scales; (8) reviews, conference abstract/papers, surveys, opinion
148 pieces, commentaries, books, periodicals, editorials, case studies, non-peer-reviewed articles,
149 masters dissertations and doctoral theses.

150 **2.3. Quality Assessment**

151 Methods quality was assessed by the modified Downs and Black Quality Index (cross-sectional
152 studies) and the Physiotherapy Evidence Database (PEDro) scale (intervention studies) as done in
153 previous systematic reviews (Downs & Black, 1998; Maher et al., 2003). For cross-sectional
154 studies, the modified Downs and Black Index was used and is a 14-item scale, with larger scores
155 deemed of studies with better quality. For intervention studies, the PEDro scale was assessed using
156 a 11-item scale that assesses randomized controlled trials from 0 to 1 in each item, where a score
157 of 6 is the cut-off values for high-quality studies (Bujalance-Moreno et al., 2019; García-Pinillos
158 et al., 2017). Previous research has reported a good test–retest ($r = 0.58–0.88$) and inter-rater
159 reliability ($r = 0.68$ to 0.75) for both qualitative indexes (Downs & Black, 1998; Maher et al.,
160 2003). For this systematic review, the quality assessment was independently performed by two
161 authors (J.E.T, P.F.) with subsequent inter-observer reliability analysis (Kappa index: 0.91; 95%
162 IC: 0.90–0.92).

163 **2.4. Study Coding and Data Extraction**

164 Data extraction of the reviewed articles was organized into the following topics: (1) sampling
165 characteristics by the study design, population, competitive level, sample (N), sex, age, expertise
166 level and quality score (Table 2); (2) summary of performance dimension, measures, measurement,
167 thresholds and/or metric formula in the reviewed articles; (3) references and ‘further reading’

168 reports the original studies where the methodology of the included articles were based; (4)
169 methodological approaches of the reviewed studies by reporting the study purpose, experimental
170 approach, methodological procedures, data collection, statistical and mathematical analysis. Data
171 were collected as previously described in ‘The Cochrane Data Extraction Template for Included
172 Studies’ using a Microsoft Excel sheet (Microsoft Corporation, Readmon, WA, USA (Synnot et
173 al., 2020).

174 **3. Results**

175 **3.1. Search Results and Study Selection**

176 A total of 982 titles were collected on three database (WoS = 495; Pub-Med = 232 and SportDiscus
177 = 255). After applying the selection criteria, 153 full-text articles were screened for eligibility,
178 having 26 articles been retained for final review. Figure 1 shows PRISMA flow diagram depicting
179 the screening procedures and search results.

180 **[PLACE FIGURE 1 HERE]**

181 **3.2. Participant Characteristics**

182 The reviewed articles were published between 2000 and 2022. Sample sizes ranged between 8–
183 148 participants with an observational, prospective and cross-sectional design ($n = 8$) and
184 randomized controlled trial ($n = 18$). Twenty-three articles focused on adult football players and
185 seven on youth counterparts. All articles studied male football players, particularly in elite ($n = 2$),
186 professional ($n = 8$), high-level ($n = 1$), national level ($n = 2$), amateur ($n = 2$) and volunteer ($n =$
187 1) performers. A total of 538 football players were analysed in this systematic review. Age and
188 expertise level were 16.81 ± 1.63 and 4.2 ± 3.83 years, respectively. Table 2 provides the
189 demographic characteristics of the participants in the retained studies.

190

[PLACE TABLE 2 HERE]

191 3.3. Quality Assessment

192 In the evaluation of methodological quality, the qualitative scores for cross-sectional studies
193 ranged from 8 (lowest quality) to 11 (highest quality) out of a maximum of 14 possible points in
194 the Downs and Black scale (Table 3). For intervention studies, the PEDro score ranged between 6
195 (lowest quality) and 9 (highest quality) out of 11 points (Table 4).

196

[PLACE TABLE 3 HERE]

[PLACE TABLE 4 HERE]

197 3.4. Main Findings

198 Table 5 presents the data extraction of the retained studies. Concerning the physical data, external
199 training load measures selected were based on movement speed, specifically: (i) total distance
200 covered ($n = 11$), (ii) distance covered at different speed zones ($n = 13$), (iii) game pace or average
201 speed ($n = 3$), (iv) accelerations and decelerations ($n = 3$), (v) locomotive-based ratios (e.g., ratio
202 between the distance covered at different intensities and distance) ($n = 1$). Otherwise, positional
203 and tactical variables reported in the included studies were based on the following independent
204 variables: (i) possession ball ($n = 1$), (ii) spatial exploration indexes ($n = 6$), (iii) LPW ratio ($n =$
205 1), (iv) stretch indexes ($n = 2$), (v) multiscale entropy ($n = 1$), (vi) synchronization indexes (i.e.,
206 longitudinal and lateral directions) ($n = 10$), (v) intra-team and opponent's dyads ($n = 3$), (vi)
207 dispersion and contraction indexes (i.e., length, width and speed) ($n = 4$), (vii) playing space and
208 effectiveness (i.e., effective playing space, longitudinal distance between GK and the closest
209 defender ($n = 3$), (viii) player's variability, regularity and coordination ($n = 10$) (i.e., entropy,
210 dynamic overlap, near-in-phase and near-anti-phase coordination, regularity zones occupied), (ix)
211 team centroid ($n = 2$).

212 Table 6 lists the purpose, game format, experimental approach, methodological procedures, data
213 collection, statistical and mathematical analysis of the studies included in this review. The data
214 organization respected the main purposes of this systematic review, specifically the integration of
215 physical and tactical variables in football using positional data ($n = 26$). Other dimensions were
216 also reported, such as psychological and technical factors. However, the results of these
217 approaches were not the focus of the analyses ($n = 5$). Non-positional variables were computed by
218 the reviewed studies for other performance dimension such as technical ($n = 4$), tactical ($n = 5$)
219 and psychophysiological variables ($n = 4$). Psychophysiological measures were reported by
220 exertion-based indexes (i.e., exertion index per minute, ratings of perceived exertion) ($n = 1$) and
221 heart rate-based methods (i.e., $\%HR_{max}$, $TRIMP_{MOD}$) ($n = 3$).

222 Quasi-experimental approaches studied training sets ($n = 20$) and match settings ($n = 6$).
223 One study analysed both training and play settings. Small sided-games (SSG) were the most
224 common training task formats ($n = 17$), with only three articles focusing on medium-sided (MSG)
225 ($n = 1$) and large-sided games (LSG) ($n = 4$). Regarding the methodological procedures, Matlab®
226 routines (Math-Works, Inc., Massachusetts, USA) were used by all authors for processing raw data
227 (xx, yy) ($n = 26$). All studies applied Butterworth low pass filter at sampling frequencies ranging
228 3–5 Hz, using 10–20 windows and, 1000–3000 points per data collect. Match analysis software
229 was used to extract technical variables in three studies, including the LongoMatch® software ($n =$
230 1), Match Analysis Camera Systems® ($n = 1$) and Lince software ($n = 1$). Data collection was
231 based on GPS ($n = 15$), LPM ($n = 2$) and optical-based tracking systems ($n = 5$) at 5-15 Hz. Also,
232 internal training load measures were collected at 1 Hz short-range radio telemetry ($n = 2$) and CR
233 10-scale ($n = 1$). A study used a portable optical timing system to measure neuromuscular
234 performance (i.e., countermovement jump, CMJ).

235 Null hypothesis statistical test (NHST) and magnitude-based inferences (MBI) were the
236 statistical procedures chosen in seven ($n = 7$) and five ($n = 5$) studies, respectively. The statistical
237 and mathematical analyses performed were the approximate entropy (ApEn) ($n = 7$), Boltzmann–
238 Gibbs–Shannon entropy measure ($n = 7$), Coefficient of variation (CV) ($n = 7$), dynamic overlap
239 ($\langle qd(t) \rangle$) ($n = 7$), effect sized Cohen's d ($n = 7$), effect sized eta-squared (η_p^2) ($n = 7$), hilbert
240 transform ($n = 7$), intraclass correlation (ICC) ($n = 7$), smallest worthwhile changes (SWC) ($n =$
241 7), standard error of measurement (SEM) ($n = 7$), standardized (Cohen) differences ($n = 7$),
242 structural coefficients (SC) ($n = 7$), and trapping strength ($n = 7$).

243 **[PLACE TABLE 5 HERE]**

244 **[PLACE TABLE 6 HERE]**

245 **4. Discussion**

246 The aim of this study was to systematically review the articles that integrated physical and tactical
247 variables using positional data in football. Physical data used to be analysed by the player's speed.
248 Otherwise, positional datasets were computed by spatiotemporal features such as spatial variability
249 or regularity of the player's movements, complex index, coordination/synchronization using intra-
250 team and inter-team dyads, playing space.

251 Positional datasets allows a more ecological insight on individual physical demands, if the
252 data interpretation considers the contextual factors and collective behaviour through a tactical
253 analysis (Marcelino et al., 2020; Teixeira et al., 2022; Teixeira et al., 2021). Several authors have
254 emphasized the need to expand the evidence produced in football on just one performance
255 dimension (i.e., physical/physiological, technical or tactical). It is important to apply
256 methodologies based on integrative approaches that analyse the interplay between technical
257 factors, key tactical/performance outcomes, collective behaviour and match-related contextual
258 drivers (Teixeira et al., 2022). Therefore, an integrative approach was expanded in 8 articles by
259 adding psychophysiological and technical outcomes (Bradley & Ade, 2018; Paul et al., 2015).
260 Considering the multifactorial phenomenon of performance in team sports, it is also important to
261 consider the influence of psychological variables on the control of physical capacities, pacing
262 behaviour, decision-making, self-regulation, and effort perception (Branquinho et al., 2020, 2021;
263 Ferraz et al., 2022). Also, bringing together observational methodologies should be considered
264 when positional data is to be made meaningful with skilled and technical aspects (Anguera &
265 Mendo, 2013; Preciado et al., 2019; Sarmiento et al., 2018).

266 The reviewed quasi-experimental studies researched training sets and match settings. Small
267 sided-games (SSG) was the most common training task in the studies, with only three articles
268 addressing medium-sided (MSG) and large-sided games (LSG) (Machado et al., 2020; Folgado et

269 al., 2018; Gonçalves et al., 2017; Jara et al., 2019; Nieto et al., 2022; Praça et al., 2016),
270 respectively. Thus, SSG formats have been further explored in the literature, mainly 5- and 6-sided
271 game formats. Indeed, these SSG-based formats were previously reported as useful tools to promote
272 significant variations in the training load, and likely in the improvement of the different domains
273 of football training (i.e., physiological, technical, and tactical dimensions) (Branquinho &
274 Marques, 2021; Clemente et al., 2021). Also, SSG and conditioned games (SSCG) are excellent
275 ways to enhance acquisition of motor efficiency and decision-making skills (Davids et al., 2013).
276 A research gap remains unexplored in MSG and LSG formats, as well as, its relationship with
277 formal game formats (i.e., 7-, 8-, 9- and 11-sided formats) (Baptista et al., 2020; Coutinho et al.,
278 2019, 2020; Ferraz et al., 2020; Figueira et al., 2018; Gonçalves et al., 2017; Sampaio et al., 2014).

279 Regarding the methodological procedures, Matlab® routines (Math-Works, Inc.,
280 Massachusetts, USA) were employed to process raw data (xx, yy) ($n = 26$), transforming data
281 points into the Universal Transverse Mercator (UTM) coordinate system (Folgado et al., 2014;
282 Sampaio & Maças, 2012). The most used correction guideline to reduce tracking signal noise was
283 a 3 Hz Butterworth low pass filter by applying non-linear logarithms using 20 windows of 3000
284 points per dataset (Coutinho et al., 2019, 2020; Figueira et al., 2018; Folgado et al., 2015, 2019;
285 Gonçalves et al., 2017, 2018). Other studies adopted smaller data windows such as 1500 data sets
286 and a sampling frequency for signal correction (i.e., 5 Hz) (Baptista et al., 2020). Nevertheless, the
287 sampling frequency and datasets is highly dependent on the type of non-linear method to be used,
288 and the use of higher time-series lengths can increase the consistency of the positional data
289 (Baptista et al., 2020; Richman & Moorman, 2000; Yentes et al., 2013). Approximate entropy
290 (ApEn) was the most noted non-linear variable for measuring the spatial movement
291 variability/regularity (Baptista et al., 2020; Coutinho et al., 2019, 2020; Ferraz et al., 2020;

292 Figueira et al., 2018; Gonçalves et al., 2017; Sampaio et al., 2014). Also, stretch index can be
293 based on the ApEn or distance, being the most reported complex index in the reviewed literature
294 (Coutinho et al., 2019; Olthof et al., 2018; Praça et al., 2021). Indeed, the entropy has been
295 extensively reported as outstanding informational parameter to describe the variability and
296 predictability of the players' movements (Teixeira et al., 2022). Hilbert transform was the most
297 frequent method, by computing the longitudinal and lateral directions through in- and anti-phase
298 (Coutinho et al., 2019, 2020; Fernandes et al., 2010; Figueira et al., 2018; Folgado et al., 2015,
299 2019; Gonçalves et al., 2017, 2018). Total surface area or playing space (m^2) can be provided
300 through trigonometry using the smallest convex hull and/or polygonal area delimited by the
301 peripheral players (Coutinho et al., 2017; Folgado et al., 2019; Olthof et al., 2018). Recently,
302 Teixeira et al. (2022) reported that it remains to be explored correlation matrixes, clustering
303 methods and fractality patterns. This review confirms this assertion and opens up the possibility
304 of exploring these metrics by integrating physical demands with individual and collective
305 behaviour.

306 Match analysis software was used in three studies for notational analysis to extract
307 technical variables, specifically the LongoMatch[®] software, Match Analysis Camera Systems[®]
308 (Gonçalves et al., 2018) and Lince software[®] (Canton et al., 2021; Ric et al., 2016). Further tactical
309 variables were selected in the retained studies using other methodological approaches (i.e.,
310 observational and notational analysis) such as metrics based on ball possession, team networks
311 and tactical actions classifications. Technical outcomes were mainly based on individual actions
312 and skills characterized in quantity and success (i.e., successful passes, dribbles and shots)
313 (Coutinho et al., 2020, 2022a; 2022b; 2022c; Folgado et al., 2019; Olthof et al., 2018; Ric et al.,
314 2017). Data collection was based on GPS, LPM and optical-based tracking systems ranging from

315 5 to 18 Hz. The first studies were mainly based on 5 Hz GPS (SPI-Pro X II, GPS ports, Canberra,
316 ACT, and Australia). However, the use of sampling frequency at 5 Hz must consider some
317 limitations in the measurement of non-linear and high-intensity paths (Portas et al., 2010; Teixeira
318 et al., 2021). Authors should prioritise tracking devices with sampling frequencies above 10 Hz
319 shape with an accelerometer (Gómez-Carmona et al., 2020; Rico-González et al., 2020). The latest
320 GPS devices already recommend a sampling frequency of 10-18 Hz, specifically 10 Hz GPS units
321 (WIMU PRO, RealTrack Systems, Almeria, Spain) (Canton et al., 2021; Machado et al., 2022),
322 10 Hz GPS (S5, Catapult Innovations, Australia) (Folgado et al., 2019; Nieto et al., 2022), 10 Hz
323 GPS (Polar, Team Pro, Kempele, Finland) (Praça et al., 2021), 15Hz GPS (SPIPRO, GPSports,
324 Canberra, ACT, Australia) (Coutinho et al., 2017; Praça et al., 2016; Ric et al., 2016) and 18.18
325 Hz GPS (GPEXE GK, Exelio SRL, Udine, Italy) (Jara et al., 2019). LPM devices and
326 semiautomatic video tracking system used in the surveyed studies were the Prozone[®] (ProZone
327 Holdings Ltd, UK) (Folgado et al., 2015) and the Inmotio Object Tracking[®] (BV., Amsterdam, the
328 Netherlands) (Olthof et al., 2018). The integration of the different tracking systems can be further
329 explored from an integrative perspective (Buchheit et al., 2014; Linke et al., 2018). Also, the
330 relationship between objective (i.e., tracking systems) and subjective (i.e., observational/notational
331 analyses) measures should be explored in future integrative approaches, in order to make the
332 integration of technical and tactical factors more effective (Teixeira et al., 2021b). Indeed, the non-
333 positional data can act as an added value to the positional raw data by making the information
334 gathered from the tracking systems more feasible and comprehensive (Praça et al., 2022, Teixeira
335 et al., 2021b)

336 Likewise, internal training load was collected by heart-rate-based measures and perceived
337 exertion. Although limitations have been reported in some studies, the perceived exertion and

338 heart-rate maintains its feasibility in elite and sub-elite football settings, it is cost-effective and
339 straightforward to employ (Achten & Jeukendrup, 2003; Teixeira et al., 2022; Teixeira, Forte, et
340 al., 2021b). A study used portable optical timing system to measure neuromuscular performance
341 (i.e., countermovement jump, CMJ). Field-based tests in football become more effective when
342 continuous control (i.e. monitoring) is integrated into the assessment (Clemente et al., 2022;
343 Gomez-Carmona et al., 2020).

344 Seven studies conducted null hypothesis statistical test (NHST) and five magnitude-based
345 inferences (MBI). Some authors used both types of analyses, besides statistical analyses such as
346 coefficient of variation (CV) ($n = 7$), effect sizes (ES) and smallest worthwhile changes (SWC)
347 (Bernards et al., 2017; Flanagan, 2013). The application of the two statistical procedures (NHST
348 and MBI) makes it more difficult to compare results of availed studies, and future pieces of
349 research should further analyse the potential overvaluation and bias of the study findings (Foster
350 et al., 2017; Welsh & Knight, 2015).

351 The current systematic review has limitations that should be considered. Firstly, the
352 interpretation of the studies was only qualitative, not have been done a meta-analysis. Secondly,
353 only studies that integrated physical and tactical measures were retained for further analysis,
354 instead of just one dimension. For this reason, other details on physical and behavioural data may
355 be absent (Duarte et al., 2013; Folgado et al., 2014; Vilar et al., 2014). Although this was made
356 clear in the inclusion and exclusion criteria, future systematic reviews should clarify which studies
357 used both one-dimensional and integrative approach. There are topics that can be explored in the
358 future: (i) the development of user-friendly interfaces to depict positional data, because Matlab®
359 routines requiring extensive training in coding and programming are used to process and display
360 data (Math-Works, Inc., Massachusetts, USA); (ii) developing tracking and wearables devices

361 enabling real-time feedback to increase the practical applicability and decision making of football
362 players and coaching staff; (iii) applying advanced data analytics and big data-based procedures
363 using artificial intelligence, machine learning and deep learning to compute automatically physical
364 and positional data; (iv) manipulating task constraints in MSG, LSG and different game-formats
365 can still be better exploited; (v) woman's football is still not analysed using physical and tactical
366 integration.

367 **5. Conclusions**

368 Based on this systematic review, physical and tactical factors can be integrated by positional data
369 using player's movement speed, spatial movement (and their variability, regularity or
370 predictability), complex indexes, playing areas, intra-team and inter-team synchronization dyads.
371 Futures research should consider applying positional data in women's football and explore the
372 representativeness of the MSG and LSG in youth training settings. Although positional data is
373 being extensively applied in semi-professional and professional football, user-friendly and real-
374 time interfaces streaming physical and tactical outcomes should be consider to enable the
375 widespread of this technology to all.

376 **6. References**

- 377 Abade, E. A., Gonçalves, B. V., Leite, N. M., & Sampaio, J. E. (2014). Time-motion and physiological profile of
378 football training sessions performed by under-15, under-17 and under-19 elite Portuguese players. *International*
379 *Journal of Sports Physiology and Performance*, 9(3), 463–470. <https://doi.org/10.1123/ijsp.2013-0120>
- 380 Abt, G., & Lovell, R. (2009). The use of individualized speed and intensity thresholds for determining the distance
381 run at high-intensity in professional soccer. *Journal of Sports Sciences*, 27(9), 893–898.
382 <https://doi.org/10.1080/02640410902998239>
- 383 Achten, J., & Jeukendrup, A. E. (2003). Heart rate monitoring: Applications and limitations. *Sports Medicine*
384 (*Auckland, N.Z.*), 33(7), 517–538.
- 385 Anguera, M. T., & Hernández Mendo, A. (2013). *Observational methodology in sport sciences*.
386 <https://dehesa.unex.es:8443/handle/10662/7361>
- 387 Arede, J., Cumming, S., Johnson, D., & Leite, N. (2021). The effects of maturity matched and un-matched opposition
388 on physical performance and spatial exploration behavior during youth basketball matches. *PLOS ONE*, 16(4),
389 e0249739. <https://doi.org/10.1371/journal.pone.0249739>
- 390 Balescu, R. (1975). Equilibrium and nonequilibrium statistical mechanics. *NASA STI/Recon Technical Report A*, 76,
391 32809.
- 392 Baptista, J., Travassos, B., Gonçalves, B., Mourão, P., Viana, J. L., & Sampaio, J. (2020). Exploring the Effects of
393 Playing Formations on Tactical Behavior and External Workload During Football Small-Sided Games. *The*
394 *Journal of Strength & Conditioning Research*, 34(7), 2024–2030.
395 <https://doi.org/10.1519/JSC.0000000000002445>
- 396 Batshchelet, E. (1981). *Circular Statistics in Biology*. Academic Press, 111 Fifth ave., New York, NY.
- 397 Beato, M., Coratella, G., Stiff, A., & Iacono, A. D. (2018). The validity and between-unit variability of GNSS units
398 (STATSports Apex 10 and 18 Hz) for measuring distance and peak speed in team sports. *Frontiers in Physiology*,
399 9. <https://doi.org/10.3389/fphys.2018.01288>
- 400 Beato, M., & Jamil, M. (2018). Intra-system reliability of SICS: Video-tracking system (Digital.Stadium®) for
401 performance analysis in soccer. *The Journal of Sports Medicine and Physical Fitness*, 58(6), 831–836.
402 <https://doi.org/10.23736/S0022-4707.17.07267-X>
- 403 Bernards, J. R., Sato, K., Haff, G. G., & Bazylar, C. D. (2017). Current Research and Statistical Practices in Sport
404 Science and a Need for Change. *Sports*, 5(4), 87. <https://doi.org/10.3390/sports5040087>
- 405 Bourbousson, J., Sève, C., & McGarry, T. (2010). Space–time coordination dynamics in basketball: Part 2. The
406 interaction between the two teams. *Journal of Sports Sciences*, 28(3), 349–358.
407 <https://doi.org/10.1080/02640410903503640>
- 408 Bradley, P., & Ade, J. (2018). Are current physical match performance metrics in elite soccer fit for purpose or is the
409 adoption of an integrated approach needed? *International Journal of Sports Physiology and Performance*, 13(5),
410 656–664. <https://doi.org/10.1123/ijsp.2017-0433>
- 411 Bradley, P. S., & Ade, J. D. (2018). Are current physical match performance metrics in elite soccer fit for purpose or
412 is the adoption of an integrated approach needed? *International Journal of Sports Physiology and Performance*,
413 13(5), 656–664. <https://doi.org/10.1123/ijsp.2017-0433>
- 414 Branquinho, L., Ferraz, R., & Marques, M. C. (2021). 5-a-Side Game as a Tool for the Coach in Soccer Training.
415 *Strength and Conditioning Journal*, 43(5), 96–108. <https://doi.org/10.1519/SSC.0000000000000629>
- 416 Branquinho, L., Ferraz, R., Travassos, B., Marinho, D. A., & Marques, M. C. (2021). Effects of different recovery
417 times on internal and external load during small-sided games in soccer. *Sports Health*, 1941738121995469.
418 <https://doi.org/10.1177/1941738121995469>
- 419 Branquinho, L., Ferraz, R., Travassos, B., & Marques, M. C. (2020). Comparison between continuous and fractionated
420 game format on internal and external load in small-sided games in soccer. *International Journal of*
421 *Environmental Research and Public Health*, 17(2), 405. <https://doi.org/10.3390/ijerph17020405>

- 422 Buchheit, M., Allen, A., Poon, T. K., Modonutti, M., Gregson, W., & Salvo, V. D. (2014). Integrating different
423 tracking systems in football: Multiple camera semi-automatic system, local position measurement and GPS
424 technologies. *Journal of Sports Sciences*, 32(20), 1844–1857. <https://doi.org/10.1080/02640414.2014.942687>
- 425 Busa, M. A., & van Emmerik, R. E. A. (2016). Multiscale entropy: A tool for understanding the complexity of postural
426 control. *Journal of Sport and Health Science*, 5(1), 44–51. <https://doi.org/10.1016/j.jshs.2016.01.018>
- 427 Campos-Vazquez, M. A., Mendez-Villanueva, A., Gonzalez-Jurado, J. A., León-Prados, J. A., Santalla, A., & Suarez-
428 Arrones, L. (2015). Relationships between rating-of-perceived-exertion- and heart-rate-derived internal training
429 load in professional soccer players: A comparison of on-field integrated training sessions. *International Journal*
430 *of Sports Physiology and Performance*, 10(5), 587–592. <https://doi.org/10.1123/ijspp.2014-0294>
- 431 Canton, A., Torrents, C., Gonçalves, B., Ric, A., Salvioni, F., Exel, J., & Sampaio, J. (2021). The diagonal positioning
432 of the goals modifies the external training load and the tactical behaviour of young football players. *Biology of*
433 *Sport*, 39(1), 135–144. <https://doi.org/10.5114/biolsport.2021.102929>
- 434 Carling, C., Wright, C., Nelson, L. J., & Bradley, P. S. (2014). Comment on ‘Performance analysis in football: A
435 critical review and implications for future research’. *Journal of Sports Sciences*, 32(1), 2–7.
436 <https://doi.org/10.1080/02640414.2013.807352>
- 437 Carrilho, D., Santos Couceiro, M., Brito, J., Figueiredo, P., Lopes, R. J., & Araújo, D. (2020). Using Optical Tracking
438 System Data to Measure Team Synergic Behavior: Synchronization of Player-Ball-Goal Angles in a Football
439 Match. *Sensors*, 20(17), 4990. <https://doi.org/10.3390/s20174990>
- 440 Castellano, J., Alvarez-Pastor, D., & Bradley, P. S. (2014). Evaluation of research using computerised tracking
441 systems (Amisco and Prozone) to analyse physical performance in elite soccer: A systematic review. *Sports*
442 *Medicine (Auckland, N.Z.)*, 44(5), 701–712. <https://doi.org/10.1007/s40279-014-0144-3>
- 443 Clemente, F. M., Afonso, J., Castillo, D., Arcos, A. L., Silva, A. F., & Sarmento, H. (2020). The effects of small-sided
444 soccer games on tactical behavior and collective dynamics: A systematic review. *Chaos, Solitons & Fractals*,
445 134, 109710. <https://doi.org/10.1016/j.chaos.2020.109710>
- 446 Clemente, F. M., Afonso, J., & Sarmento, H. (2021). Small-sided games: An umbrella review of systematic reviews
447 and meta-analyses. *PLOS ONE*, 16(2), e0247067. <https://doi.org/10.1371/journal.pone.0247067>
- 448 Clemente, F. M., Couceiro, M. S., Martins, F. M. L., Mendes, R., & Figueiredo, A. J. (2013). Measuring tactical
449 behaviour using technological metrics: Case study of a football game. *International Journal of Sports Science &*
450 *Coaching*, 8(4), 723–739. <https://doi.org/10.1260/1747-9541.8.4.723>
- 451 Clemente, F. M., Owen, A., Mustapha, A., van der Linden, C. M. I. (Niels), Ribeiro, J., Mendes, B., & Reichert, J.
452 (2018). Measurement of the Pitch Exploration Amongst Elite Professional Soccer Players: Official Match
453 Analysis. In R. Ghazali, M. M. Deris, N. M. Nawi, & J. H. Abawajy (Eds.), *Recent Advances on Soft Computing*
454 *and Data Mining* (pp. 191–199). Springer International Publishing. https://doi.org/10.1007/978-3-319-72550-5_19
- 456 Clemente, F. M., Owen, A., Serra-Olivares, J., Correia, A., Bernardo Sequeiros, J., Silva, F. G., & Martins, F. M. L.
457 (2018). The effects of large-sided soccer training games and pitch size manipulation on time–motion profile,
458 spatial exploration and surface area: Tactical opportunities. *Proceedings of the Institution of Mechanical*
459 *Engineers, Part P: Journal of Sports Engineering and Technology*, 232(2), 160–165.
460 <https://doi.org/10.1177/1754337117722658>
- 461 Clemente, F., Oliveira, R., Akyildiz, Z., Silva, R., Ceylan, H., Afonso, J., & Raya-González, J. (2022). *Field-based*
462 *Tests for Soccer Players: Methodological Concerns and Applications*.
- 463 Coito, N., Davids, K., Folgado, H., Bento, T., & Travassos, B. (2022). Capturing and Quantifying Tactical Behaviors
464 in Small-Sided and Conditioned Games in Soccer: A Systematic Review. *Research Quarterly for Exercise and*
465 *Sport*, 93(1), 189–203. <https://doi.org/10.1080/02701367.2020.1823307>
- 466 Collet, C. (2013). The possession game? A comparative analysis of ball retention and team success in European and
467 international football, 2007–2010. *Journal of Sports Sciences*, 31(2), 123–136.
468 <https://doi.org/10.1080/02640414.2012.727455>

- 469 Costa, I. T. da, Garganta, J., Greco, P. J., Mesquita, I., & Maia, J. (2011). Sistema de avaliação tática no Futebol
470 (FUT-SAT): Desenvolvimento e validação preliminar. *Motricidade*, 7(1), 69–84.
471 <https://doi.org/10.6063/motricidade.121>
- 472 Costa, M., Goldberger, A. L., & Peng, C.-K. (2005). Multiscale entropy analysis of biological signals. *Physical Review*
473 *E*, 71(2), 021906. <https://doi.org/10.1103/PhysRevE.71.021906>
- 474 Coutinho, D., Gonçalves, B., Folgado, H., Travassos, B., Santos, S., & Sampaio, J. (2022a). Amplifying perceptual
475 demands: How changes in the colour vests affect youth players performance during medium-sided games. *PLoS*
476 *ONE*, 17(1), e0262245. <https://doi.org/10.1371/journal.pone.0262245>
- 477 Coutinho, D., Gonçalves, B., Santos, S., Travassos, B., Folgado, H., & Sampaio, J. (2022b). Exploring how limiting
478 the number of ball touches during small-sided games affects youth football players' performance across different
479 age groups. *International Journal of Sports Science & Coaching*, 17(3), 545–557.
480 <https://doi.org/10.1177/17479541211037001>
- 481 Coutinho, D., Gonçalves, B., Santos, S., Travassos, B., Schöllhorn, W., & Sampaio, J. (2022c). The effects of
482 individual and collective variability on youth players' movement behaviours during football small-sided games.
483 *Research in Sports Medicine*, 0(0), 1–16. <https://doi.org/10.1080/15438627.2022.2042293>
- 484 Coutinho, D., Gonçalves, B., Santos, S., Travassos, B., Wong, D. P., & Sampaio, J. (2019). Effects of the pitch
485 configuration design on players' physical performance and movement behaviour during soccer small-sided
486 games. *Research in Sports Medicine*, 27(3), 298–313. <https://doi.org/10.1080/15438627.2018.1544133>
- 487 Coutinho, D., Gonçalves, B., Travassos, B., Abade, E., Wong, D. P., & Sampaio, J. (2019). Effects of pitch spatial
488 references on players' positioning and physical performances during football small-sided games. *Journal of*
489 *Sports Sciences*, 37(7), 741–747. <https://doi.org/10.1080/02640414.2018.1523671>
- 490 Coutinho, D., Gonçalves, B., Travassos, B., Folgado, H., Figueira, B., & Sampaio, J. (2020). Different marks in the
491 pitch constraint youth players' performances during football small-sided games. *Research Quarterly for Exercise*
492 *and Sport*, 91(1), 15–23. <https://doi.org/10.1080/02701367.2019.1645938>
- 493 Coutinho, D., Gonçalves, B., Travassos, B., Wong, D. P., Coutts, A. J., & Sampaio, J. E. (2017). Mental fatigue and
494 spatial references impair soccer players' physical and tactical performances. *Frontiers in Psychology*, 0.
495 <https://doi.org/10.3389/fpsyg.2017.01645>
- 496 Coutts, A. J., & Duffield, R. (2010). Validity and reliability of GPS devices for measuring movement demands of
497 team sports. *Journal of Science and Medicine in Sport*, 13(1), 133–135.
498 <https://doi.org/10.1016/j.jsams.2008.09.015>
- 499 Dalen, T., Jørgen, I., Gertjan, E., Geir Havard, H., & Ulrik, W. (2016). Player load, acceleration, and deceleration
500 during forty-five competitive matches of elite soccer. *The Journal of Strength & Conditioning Research*, 30(2),
501 351–359. <https://doi.org/10.1519/JSC.0000000000001063>
- 502 David Gomez-Carmona, C., Pino-Ortega, J., & Jose Ibanez, S. (2020). Design and validity of a field test battery for
503 assessing multi-location external load profile in invasion team sports. *E-Balónmano Com*, 16(1), 23–48.
- 504 Davids, K., Araújo, D., Correia, V., & Vilar, L. (2013). How Small-Sided and Conditioned Games Enhance
505 Acquisition of Movement and Decision-Making Skills. *Exercise and Sport Sciences Reviews*, 41(3), 154–161.
506 <https://doi.org/10.1097/JES.0b013e318292f3ec>
- 507 Di Salvo, C., & Marco, C. (2006). Validation of Prozone ®: A new video-based performance analysis system.
508 *International Journal of Performance Analysis in Sport*, 6(1), 108–119.
509 <https://doi.org/10.1080/24748668.2006.11868359>
- 510 Duarte, R., Araújo, D., Correia, V., & Davids, K. (2012). Sports teams as superorganisms: Implications of
511 sociobiological models of behaviour for research and practice in team sports performance analysis. *Sports*
512 *Medicine*, 42(8), 633–642. <https://doi.org/10.1007/BF03262285>
- 513 Duarte, R., Araújo, D., Correia, V., Davids, K., Marques, P., & Richardson, M. J. (2013). Competing together:
514 Assessing the dynamics of team–team and player–team synchrony in professional association football. *Human*
515 *Movement Science*, 32(4), 555–566. <https://doi.org/10.1016/j.humov.2013.01.011>

- 516 Duarte, R., Araújo, D., Davids, K., Travassos, B., Gazimba, V., & Sampaio, J. (2012). Interpersonal coordination
517 tendencies shape 1-vs-1 sub-phase performance outcomes in youth soccer. *Journal of Sports Sciences*, 30(9),
518 871–877. <https://doi.org/10.1080/02640414.2012.675081>
- 519 Duarte, R., Araújo, D., Folgado, H., Esteves, P., Marques, P., & Davids, K. (2013). Capturing complex, non-linear
520 team behaviours during competitive football performance. *Journal of Systems Science and Complexity*, 26(1),
521 62–72. <https://doi.org/10.1007/s11424-013-2290-3>
- 522 Duarte, R., Araújo, D., Freire, L., Folgado, H., Fernandes, O., & Davids, K. (2012). *Intra- and inter-group*
523 *coordination patterns reveal collective behaviors of football players near the scoring zone.*
524 <https://doi.org/10.1016/j.humov.2012.03.001>
- 525 Fernandes, O., Folgado, H., Duarte, R., & Malta, P. (2010). Validation of the tool for applied and contextual time-
526 series observation. *International Journal of Sport Psychology*, 41, 63–64.
- 527 Ferraz, R., Forte, P., Branquinho, L., Teixeira, J., Neiva, H., Marinho, D., & Marques, M. (2022). *The Performance*
528 *during the Exercise: Legitimizing the Psychophysiological Approach.*
529 <https://doi.org/10.5772/intechopen.102578>
- 530 Ferraz, R., Gonçalves, B., Coutinho, D., Marinho, D. A., Sampaio, J., & Marques, M. C. (2018). Pacing behaviour of
531 players in team sports: Influence of match status manipulation and task duration knowledge. *PLoS One*, 13(2),
532 e0192399. <https://doi.org/10.1371/journal.pone.0192399>
- 533 Ferraz, R., Gonçalves, B., Coutinho, D., Oliveira, R., Travassos, B., Sampaio, J., & Marques, M. C. (2020). Effects
534 of knowing the task's duration on soccer players' positioning and pacing behaviour during small-sided games.
535 *Int. J. Environ. Res. Public Health*, 11.
- 536 Ferraz, R., Gonçalves, B., Tillaar, R., Saiz, S., Sampaio, J., & Marques, M. (2017). Effects of knowing the task
537 duration on players' pacing patterns during soccer small-sided games. *Journal of Sports Sciences*, 36.
538 <https://doi.org/10.1080/24733938.2017.1283433>
- 539 Figueira, B., Gonçalves, B., Masiulis, N., & Sampaio, J. (2018). Exploring how playing football with different age
540 groups affects tactical behaviour and physical performance. *Biology of Sport*, 35(2), 145–153.
541 <https://doi.org/10.5114/biolsport.2018.71603>
- 542 Flanagan, E. P. (2013). The Effect Size Statistic—Applications for the Strength and Conditioning Coach. *Strength &*
543 *Conditioning Journal*, 35(5), 37–40. <https://doi.org/10.1519/SSC.0b013e3182a64d20>
- 544 Folgado, H., Bravo, J., Pereira, P., & Sampaio, J. (2019). Towards the use of multidimensional performance indicators
545 in football small-sided games: The effects of pitch orientation. *Journal of Sports Sciences*, 37(9), 1064–1071.
546 <https://doi.org/10.1080/02640414.2018.1543834>
- 547 Folgado, H., Duarte, R., Fernandes, O., & Sampaio, J. (2014). Competing with lower level opponents decreases intra-
548 team movement synchronization and time-motion demands during pre-season soccer matches. *PloS One*, 9(5),
549 e97145. <https://doi.org/10.1371/journal.pone.0097145>
- 550 Folgado, H., Duarte, R., Marques, P., & Sampaio, J. (2015). The effects of congested fixtures period on tactical and
551 physical performance in elite football. *Journal of Sports Sciences*, 33(12), 1238–1247.
552 <https://doi.org/10.1080/02640414.2015.1022576>
- 553 Folgado, H., Gonçalves, B., & Sampaio, J. (2018). Positional synchronization affects physical and physiological
554 responses to preseason in professional football (soccer). *Research in Sports Medicine*, 26(1), 51–63.
555 <https://doi.org/10.1080/15438627.2017.1393754>
- 556 Folgado, H., Lemmink, K. A. P. M., Frencken, W., & Sampaio, J. (2014). Length, width and centroid distance as
557 measures of teams tactical performance in youth football. *European Journal of Sport Science*, 14 Suppl 1, S487-
558 492. <https://doi.org/10.1080/17461391.2012.730060>
- 559 Fonseca, S., Milho, J., Travassos, B., & Araújo, D. (2012). Spatial dynamics of team sports exposed by Voronoi
560 diagrams. *Human Movement Science*, 31(6), 1652–1659. <https://doi.org/10.1016/j.humov.2012.04.006>
- 561 Foster, C., Rodriguez-Marroyo, J., & de Koning, J. (2017). Monitoring Training Loads: The Past, the Present, and the
562 Future. *International Journal of Sports Physiology and Performance*, 12, 2–8.

- 563 Frencken, W., De Poel, H., Visscher, C., & Lemmink, K. A. P. M. (2012). Variability of inter-team distances
564 associated with match events in elite-standard soccer. *Journal of Sports Sciences*, *30*, 1207–1213.
565 <https://doi.org/10.1080/02640414.2012.703783>
- 566 Frencken, W., Lemmink, K., Delleman, N., & Visscher, C. (2011). Oscillations of centroid position and surface area
567 of soccer teams in small-sided games. *European Journal of Sport Science*, *11*(4), 215–223.
568 <https://doi.org/10.1080/17461391.2010.499967>
- 569 Gabin, B., Camerino, O., Anguera, M. T., & Castañer, M. (2012). Lince: Multiplatform Sport Analysis Software.
570 *Procedia - Social and Behavioral Sciences*, *46*, 4692–4694. <https://doi.org/10.1016/j.sbspro.2012.06.320>
- 571 García-Pinillos, F., Soto-Hermoso, V. M., & Latorre-Román, P. A. (2017). How does high-intensity intermittent
572 training affect recreational endurance runners? Acute and chronic adaptations: A systematic review. *Journal of*
573 *Sport and Health Science*, *6*(1), 54–67. <https://doi.org/10.1016/j.jshs.2016.08.010>
- 574 Giménez, J. V., Del-Coso, J., Leicht, A. S., & Gomez, M.-Á. (2018). Comparison of the movement patterns between
575 small- and large-sided game training and competition in professional soccer players. *The Journal of Sports*
576 *Medicine and Physical Fitness*, *58*(10), 1383–1389. <https://doi.org/10.23736/S0022-4707.17.07343-1>
- 577 Gómez-Carmona, C. D., Bastida-Castillo, A., Ibáñez, S. J., & Pino-Ortega, J. (2020). Accelerometry as a method for
578 external workload monitoring in invasion team sports. A systematic review. *PLOS ONE*, *15*(8), e0236643.
579 <https://doi.org/10.1371/journal.pone.0236643>
- 580 Gonçalves, B., Coutinho, D., Exel, J., Travassos, B., Peñas, C., & Sampaio, J. (2019). Extracting spatial-temporal
581 features that describe a team match demands when considering the effects of the quality of opposition in elite
582 football. *PLOS ONE*, *14*, e0221368. <https://doi.org/10.1371/journal.pone.0221368>
- 583 Gonçalves, B., Coutinho, D., Travassos, B., Folgado, H., Caixinha, P., & Sampaio, J. (2018). Speed synchronization,
584 physical workload and match-to-match performance variation of elite football players. *PLoS One*, *13*(7),
585 e0200019. <https://doi.org/10.1371/journal.pone.0200019>
- 586 Gonçalves, B., Esteves, P., Folgado, H., Ric, A., Torrents, C., & Sampaio, J. (2017). Effects of pitch area-restrictions
587 on tactical behavior, physical, and physiological performances in soccer large-sided games. *The Journal of*
588 *Strength & Conditioning Research*, *31*(9), 2398–2408. <https://doi.org/10.1519/JSC.0000000000001700>
- 589 Gonçalves, B., Folgado, H., Coutinho, D., Marcelino, R., Wong, D., Leite, N., & Sampaio, J. (2018). Changes in
590 Effective Playing Space when Considering Sub-Groups of 3 to 10 Players in Professional Soccer Matches.
591 *Journal of Human Kinetics*, *62*, 145–155. <https://doi.org/10.1515/hukin-2017-0166>
- 592 Gonçalves, B., Marcelino, R., Torres-Ronda, L., Torrents, C., & Sampaio, J. (2016). Effects of emphasising opposition
593 and cooperation on collective movement behaviour during football small-sided games. *Journal of Sports*
594 *Sciences*, *34*(14), 1346–1354. <https://doi.org/10.1080/02640414.2016.1143111>
- 595 Gonçalves, B. V., Figueira, B. E., Maçãs, V., & Sampaio, J. (2014). Effect of player position on movement behaviour,
596 physical and physiological performances during an 11-a-side football game. *Journal of Sports Sciences*, *32*(2),
597 191–199. <https://doi.org/10.1080/02640414.2013.816761>
- 598 Gore, C. (2000). *Physiological Tests for Elite Athletes*. Human Kinetics.
- 599 Goto, H., Morris, J. G., & Nevill, M. E. (2015). Motion analysis of U11 to U16 elite English Premier League Academy
600 players. *Journal of Sports Sciences*, *33*(12), 1248–1258. <https://doi.org/10.1080/02640414.2014.999700>
- 601 Gréhaigne, J. F., Bouthier, D., & David, B. (1997). Dynamic-system analysis of opponent relationships in collective
602 actions in soccer. *Journal of Sports Sciences*, *15*(2), 137–149. <https://doi.org/10.1080/026404197367416>
- 603 Harbourne, R. T., & Stergiou, N. (2009). Movement variability and the use of nonlinear tools: Principles to guide
604 physical therapist practice. *Physical Therapy*, *89*(3), 267–282. <https://doi.org/10.2522/ptj.20080130>
- 605 Hewitt, A., Greenham, G., & Norton, K. (2016). Game style in soccer: What is it and can we quantify it? *International*
606 *Journal of Performance Analysis in Sport*, *16*, 355. <https://doi.org/10.1080/24748668.2016.11868892>
- 607 Hill-Haas, S., Rowsell, G., Coutts, A., & Dawson, B. (2008). The reproducibility of physiological responses and
608 performance profiles of youth soccer players in small-sided games. *International Journal of Sports Physiology*
609 *and Performance*, *3*(3), 393–396. <https://doi.org/10.1123/ijsp.3.3.393>

- 610 Hodgson, C., Akenhead, R., & Thomas, K. (2014). Time-motion analysis of acceleration demands of 4v4 small-sided
611 soccer games played on different pitch sizes. *Human Movement Science*, 33, 25–32.
612 <https://doi.org/10.1016/j.humov.2013.12.002>
- 613 Hoppe, M. W., Baumgart, C., Polglaze, T., & Freiwald, J. (2018). Validity and reliability of GPS and LPS for
614 measuring distances covered and sprint mechanical properties in team sports. *PLOS ONE*, 13(2), e0192708.
615 <https://doi.org/10.1371/journal.pone.0192708>
- 616 Hristovski, R., Davids, Araujo, D., Passos, P., Torrents, C., Aceski, A., & Tufekcjevski. (2013). *Creativity in sport*
617 *and dance: Ecological dynamics on a hierarchically soft-assembled perception-action landscape*. (pp. 259–271).
- 618 Jara, D., Ortega, E., Gómez-Ruano, M.-Á., Weigelt, M., Nikolic, B., & Sainz de Baranda, P. (2019). Physical and
619 Tactical Demands of the Goalkeeper in Football in Different Small-Sided Games. *Sensors*, 19(16), 3605.
620 <https://doi.org/10.3390/s19163605>
- 621 Laakso, T., Davids, K., Luhtanen, P., Liukkonen, J., & Travassos, B. (2022). How football team composition
622 constrains emergent individual and collective tactical behaviours: Effects of player roles in creating different
623 landscapes for shared affordances in small-sided and conditioned games. *International Journal of Sports Science*
624 *& Coaching*, 17(2), 346–354. <https://doi.org/10.1177/17479541211030076>
- 625 Lames, M., Erdmann, J., & Walter, F. (2010). Oscillations in football—Order and disorder in spatial interactions
626 between the two teams. *International Journal of Sport Psychology*, 41, 85–86.
- 627 Lee, K. A., Hicks, G., & Nino-Murcia, G. (1991). Validity and reliability of a scale to assess fatigue. *Psychiatry*
628 *Research*, 36(3), 291–298. [https://doi.org/10.1016/0165-1781\(91\)90027-m](https://doi.org/10.1016/0165-1781(91)90027-m)
- 629 Lefever, D. W. (1926). Measuring Geographic Concentration by Means of the Standard Deviation Ellipse. *American*
630 *Journal of Sociology*, 32(1), 88–94. <https://doi.org/10.1086/214027>
- 631 Leser, R., Baca, A., & Ogris, G. (2011). Local Positioning Systems in (Game) Sports. *Sensors*, 11(10), 9778–9797.
632 <https://doi.org/10.3390/s111009778>
- 633 Linke, D., Link, D., & Lames, M. (2018). Validation of electronic performance and tracking systems EPTS under
634 field conditions. *PLOS ONE*, 13(7), e0199519. <https://doi.org/10.1371/journal.pone.0199519>
- 635 Liu, H., Gómez, M.-A., Gonçalves, B., & Sampaio, J. (2016). Technical performance and match-to-match variation
636 in elite football teams. *Journal of Sports Sciences*, 34(6), 509–518.
637 <https://doi.org/10.1080/02640414.2015.1117121>
- 638 Los Arcos, A., Martínez-Santos, R., Yanci, J., Martín, J., & Castagna, C. (2014). Variability of objective and
639 subjective intensities during ball drills in youth soccer players. *Journal of Strength and Conditioning Research*,
640 28(3), 752–757. <https://doi.org/10.1519/JSC.0b013e3182a47f0b>
- 641 Low, B., Coutinho, D., Gonçalves, B., Rein, R., Memmert, D., & Sampaio, J. (2019). A systematic review of collective
642 tactical behaviours in football using positional data. *Sports Medicine*, 50, 1–43. <https://doi.org/10.1007/s40279-019-01194-7>
- 644 Machado, J. C., Góes, A., Aquino, R., Bedo, B. L. S., Viana, R., Rossato, M., Scaglia, A., & Ibáñez, S. J. (2022).
645 Applying Different Strategies of Task Constraint Manipulation in Small-Sided and Conditioned Games: How
646 Do They Impact Physical and Tactical Demands? *Sensors*, 22(12), 4435. <https://doi.org/10.3390/s22124435>
- 647 Machado, J., Barreira, D., Teoldo, I., Serra-Olivares, J., Góes, A., & José Scaglia, A. (2020). Tactical Behaviour of
648 Youth Soccer Players: Differences Depending on Task Constraint Modification, Age and Skill Level. *Journal of*
649 *Human Kinetics*, 75, 225–238. <https://doi.org/10.2478/hukin-2020-0051>
- 650 Mallo, J., Mena, E., Nevado, F., & Paredes, V. (2015). Physical Demands of Top-Class Soccer Friendly Matches in
651 Relation to a Playing Position Using Global Positioning System Technology. *Journal of Human Kinetics*, 47,
652 179–188. <https://doi.org/10.1515/hukin-2015-0073>
- 653 Malone, J. J., Jaspers, A., Helsen, W., Merks, B., Frencken, W. G. P., & Brink, M. S. (2018). Seasonal Training Load
654 and Wellness Monitoring in a Professional Soccer Goalkeeper. *International Journal of Sports Physiology and*
655 *Performance*, 13(5), 672–675. <https://doi.org/10.1123/ijsp.2017-0472>

- 656 Marcelino, R., Sampaio, J., Amichay, G., Gonçalves, B., Couzin, I. D., & Nagy, M. (2020). Collective movement
657 analysis reveals coordination tactics of team players in football matches. *Chaos, Solitons & Fractals*, 138,
658 109831. <https://doi.org/10.1016/j.chaos.2020.109831>
- 659 McGarry, T., Anderson, D. I., Wallace, S. A., Hughes, M. D., & Franks, I. M. (2002). Sport competition as a dynamical
660 self-organizing system. *Journal of Sports Sciences*, 20(10), 771–781.
661 <https://doi.org/10.1080/026404102320675620>
- 662 Memmert, D. (2021). *Match Analysis: How to Use Data in Professional Sport*. Routledge.
- 663 Memmert, D., Lemmink, K. A. P. M., & Sampaio, J. (2017). Current approaches to tactical performance analyses in
664 soccer using position data. *Sports Medicine*, 47(1), 1–10. <https://doi.org/10.1007/s40279-016-0562-5>
- 665 Mendes, R. S., Malacarne, L. C., & Anteneodo, C. (2007). Statistics of football dynamics. *The European Physical
666 Journal B*, 57(3), 357–363. <https://doi.org/10.1140/epjb/e2007-00177-4>
- 667 Nieto, S., Castellano, J., Echeazarra, I., & Fernández, E. (2022). Effects on collective behaviour and locomotor and
668 neuromuscular response in young players by varying the length of the pitch in 11-a-side football. *International
669 Journal of Sports Science & Coaching*, 17479541221101604. <https://doi.org/10.1177/17479541221101603>
- 670 O’Donoghue, P. (2009). *Research methods for sports performance analysis*. Routledge.
671 <https://doi.org/10.4324/9780203878309>
- 672 Ogris, G., Leser, R., Horsak, B., Kornfeind, P., Heller, M., & Baca, A. (2012). Accuracy of the LPM tracking system
673 considering dynamic position changes. *Journal of Sports Sciences*, 30(14), 1503–1511.
674 <https://doi.org/10.1080/02640414.2012.712712>
- 675 Olthof, S. B. H., Frencken, W. G. P., & Lemmink, K. A. P. M. (2018). Match-derived relative pitch area changes the
676 physical and team tactical performance of elite soccer players in small-sided soccer games. *Journal of Sports
677 Sciences*, 36(14), 1557–1563. <https://doi.org/10.1080/02640414.2017.1403412>
- 678 Ometto, L., Vasconcellos, F. V., Cunha, F. A., Teoldo, I., Souza, C. R. B., Dutra, M. B., O’Sullivan, M., & Davids,
679 K. (2018). How manipulating task constraints in small-sided and conditioned games shapes emergence of
680 individual and collective tactical behaviours in football: A systematic review. *International Journal of Sports
681 Science & Coaching*, 13(6), 1200–1214. <https://doi.org/10.1177/1747954118769183>
- 682 Page, M. J., McKenzie, J. E., Bossuyt, P. M., Boutron, I., Hoffmann, T. C.,
683 Mulrow, C. D., et al. (2021). The PRISMA 2020 statement: An updated guideline
684 for reporting systematic reviews. *BMJ* 372, n71. doi: 10.1136/bmj.n71
685 Palut, Y., & Zanone, P.-G. (2005). A
686 dynamical analysis of tennis: Concepts and data. *Journal of Sports Sciences*, 23(10), 1021–1032.
<https://doi.org/10.1080/02640410400021682>
- 687 Paul, D. J., Bradley, P. S., & Nassis, G. P. (2015). Factors affecting match running performance of elite soccer players:
688 Shedding some light on the complexity. *International Journal of Sports Physiology and Performance*, 10(4),
689 516–519. <https://doi.org/10.1123/IJSP.2015-0029>
- 690 Pincus, S. M. (1991). Approximate entropy as a measure of system complexity. *Proceedings of the National Academy
691 of Sciences*, 88(6), 2297–2301. <https://doi.org/10.1073/pnas.88.6.2297>
- 692 Pol, R., Balagué, N., Ric, A., Torrents, C., Kiely, J., & Hristovski, R. (2020). Training or synergizing? Complex
693 systems principles change the understanding of sport processes. *Sports Medicine - Open*, 6(1), 28.
694 <https://doi.org/10.1186/s40798-020-00256-9>
- 695 Portas, M. D., Harley, J. A., Barnes, C. A., & Rush, C. J. (2010). The validity and reliability of 1-Hz and 5-Hz global
696 positioning systems for linear, multidirectional, and soccer-specific activities. *International Journal of Sports
697 Physiology and Performance*, 5(4), 448–458. <https://doi.org/10.1123/ijsp.5.4.448>
- 698 Praça, G. M., Andrade, A. G. P., Bredt, S. da G. T., Moura, F. A., & Moreira, P. E. D. (2020). Progression to the target
699 vs. Regular rules in Soccer small-sided Games. *Science and Medicine in Football*, 0(0), 1–6.
700 <https://doi.org/10.1080/24733938.2020.1869811>
- 701 Praça, G. M., Folgado, H., Andrade, A. G. P. de, & Greco, P. J. (2016). Influence of additional players on collective
702 tactical behavior in small-sided soccer games. *Revista Brasileira de Cineantropometria & Desempenho Humano*,
703 18, 62–71. <https://doi.org/10.5007/1980-0037.2016v18n1p62>

- 704 Praça, G. M., Moreira, P. E. D., Dieguez, G. T. de O., Barbosa, T. de O., Brandão, L. H. A., & Custódio, I. J. de O.
705 (2021). The impact of match venue on performance indicators and tactical behaviour in youth soccer players.
706 *International Journal of Performance Analysis in Sport*, 0(0), 1–11.
707 <https://doi.org/10.1080/24748668.2021.1952831>
- 708 Praça, G., Moreira, P., Andrade, A., Clemente, F., Oliveira, W., & Demetrio Cunha, G. (2022). Integrating notational
709 and positional analysis to investigate tactical behavior in offensive and defensive phases of football matches.
710 *Proceedings of the Institution of Mechanical Engineers Part P Journal of Sports Engineering and Technology*,
711 17543371221122044. <https://doi.org/10.1177/17543371221122044>
- 712 Preatoni, E., Ferrario, M., Donà, G., Hamill,
713 J., & Rodano, R. (2010). Motor variability in sports: A non-linear analysis of race walking. *Journal of Sports
714 Sciences*, 28(12), 1327–1336. <https://doi.org/10.1080/02640414.2010.507250>
- 714 Preciado, M., Anguera, M. T., Olarte, M., & Lapresa, D. (2019). Observational Studies in Male Elite Football: A
715 Systematic Mixed Study Review. *Frontiers in Psychology*, 10. <https://doi.org/10.3389/fpsyg.2019.02077>
- 716 Rago, V., Brito, J., Figueiredo, P., Costa, J., Barreira, D., Krstrup, P., & Rebelo, A. (2020). Methods to collect and
717 interpret external training load using microtechnology incorporating GPS in professional football: A systematic
718 review. *Research in Sports Medicine*, 28(3), 437–458. <https://doi.org/10.1080/15438627.2019.1686703>
- 719 Rampinini, E., Impellizzeri, F. M., Castagna, C., Coutts, A. J., & Wisløff, U. (2009). Technical performance during
720 soccer matches of the Italian Serie A league: Effect of fatigue and competitive level. *Journal of Science and
721 Medicine in Sport*, 12(1), 227–233. <https://doi.org/10.1016/j.jsams.2007.10.002>
- 722 Ranjitha, M., Nathan, K., & Joseph, L. (2020). Artificial Intelligence Algorithms and Techniques in the computation
723 of Player-Adaptive Games. *Journal of Physics: Conference Series*, 1427(1), 012006.
724 <https://doi.org/10.1088/1742-6596/1427/1/012006>
- 725 Rein, R., & Memmert, D. (2016). Big data and tactical analysis in elite soccer: Future challenges and opportunities
726 for sports science. *SpringerPlus*, 5(1), 1410. <https://doi.org/10.1186/s40064-016-3108-2>
- 727 Reis, M. A. M. D., & Almeida, M. B. (2020). The role of somatic maturation in the tactical effectiveness, efficiency
728 and variability of young soccer players. *International Journal of Performance Analysis in Sport*, 20(2), 305–321.
729 <https://doi.org/10.1080/24748668.2020.1743165>
- 730 Ric, A., Hristovski, R., Gonçalves, B., Torres, L., Sampaio, J., & Torrents, C. (2016). Timescales for exploratory
731 tactical behaviour in football small-sided games. *Journal of Sports Sciences*, 34(18), 1723–1730.
732 <https://doi.org/10.1080/02640414.2015.1136068>
- 733 Ric, A., Torrents, C., Gonçalves, B., Sampaio, J., & Hristovski, R. (2016). Soft-assembled Multilevel Dynamics of
734 Tactical Behaviors in Soccer. *Frontiers in Psychology*, 0. <https://doi.org/10.3389/fpsyg.2016.01513>
- 735 Ric, A., Torrents, C., Gonçalves, B., Torres-Ronda, L., Sampaio, J., & Hristovski, R. (2017). Dynamics of tactical
736 behaviour in association football when manipulating players' space of interaction. *PLOS ONE*, 12(7), e0180773.
737 <https://doi.org/10.1371/journal.pone.0180773>
- 738 Richman, J. S., & Moorman, J. R. (2000). Physiological time-series analysis using approximate entropy and sample
739 entropy. *American Journal of Physiology-Heart and Circulatory Physiology*, 278(6), H2039–H2049.
740 <https://doi.org/10.1152/ajpheart.2000.278.6.H2039>
- 741 Rico-González, M., Los Arcos, A., Nakamura, F. Y., Gantois, P., & Pino-Ortega, J. (2020). A comparison between
742 UWB and GPS devices in the measurement of external load and collective tactical behaviour variables during a
743 professional official match. *International Journal of Performance Analysis in Sport*, 20(6), 994–1002.
744 <https://doi.org/10.1080/24748668.2020.1823153>
- 745 Rico-González, M., Pino-Ortega, J., Castellano, J., Oliva-Lozano, J. M., & Arcos, A. L. (2021). Reference values for
746 collective tactical behaviours based on positional data in professional football matches: A systematic review.
747 *Biology of Sport*, 39(1), 101–114. <https://doi.org/10.5114/biolsport.2021.102921>
- 748 Rojas-Valverde, D., Gómez-Carmona, C. D., Gutiérrez-Vargas, R., & Pino-Ortega, J. (2019). From big data mining
749 to technical sport reports: The case of inertial measurement units. *BMJ Open Sport & Exercise Medicine*, 5(1),
750 e000565. <https://doi.org/10.1136/bmjsem-2019-000565>

- 751 Russell, M., Sparkes, W., Northeast, J., Cook, C. J., Love, T. D., Bracken, R. M., & Kilduff, L. P. (2016). Changes in
752 Acceleration and Deceleration Capacity Throughout Professional Soccer Match-Play. *The Journal of Strength*
753 *& Conditioning Research*, 30(10), 2839–2844. <https://doi.org/10.1519/JSC.0000000000000805>
- 754 Sampaio, J. E., Lago, C., Gonçalves, B., Maçãs, V. M., & Leite, N. (2014). Effects of pacing, status and unbalance in
755 time motion variables, heart rate and tactical behaviour when playing 5-a-side football small-sided games.
756 *Journal of Science and Medicine in Sport*, 17(2), 229–233. <https://doi.org/10.1016/j.jsams.2013.04.005>
- 757 Sampaio, J., & Vitor, M. (2012). Measuring Tactical Behaviour in Football. *International Journal of Sports Medicine*,
758 33, 395–401. <https://doi.org/10.1055/s-0031-1301320>
- 759 Santos, S., Coutinho, D., Gonçalves, B., Schöllhorn, W., Sampaio, J., & Leite, N. (2018). Differential Learning as a
760 Key Training Approach to Improve Creative and Tactical Behavior in Soccer. *Research Quarterly for Exercise*
761 *and Sport*, 89, 1–14. <https://doi.org/10.1080/02701367.2017.1412063>
- 762 Sarmiento, H., Clemente, F. M., Araújo, D., Davids, K., McRobert, A., & Figueiredo, A. (2018). What Performance
763 Analysts Need to Know About Research Trends in Association Football (2012–2016): A Systematic Review.
764 *Sports Medicine*, 48(4), 799–836. <https://doi.org/10.1007/s40279-017-0836-6>
- 765 Saxton, M. J. (1996). Anomalous diffusion due to binding: A Monte Carlo study. *Biophysical Journal*, 70(3), 1250–
766 1262. [https://doi.org/10.1016/S0006-3495\(96\)79682-0](https://doi.org/10.1016/S0006-3495(96)79682-0)
- 767 Seifert, L., Button, C., & Davids, K. (2013). Key properties of expert movement systems in sport: An ecological
768 dynamics perspective. *Sports Medicine (Auckland, N.Z.)*, 43(3), 167–178. <https://doi.org/10.1007/s40279-012-0011-z>
- 770 Silva, P., Chung, D., Carvalho, T., Cardoso, T., Davids, K., Araújo, D., & Garganta, J. (2016). Practice effects on
771 intra-team synergies in football teams. *Human Movement Science*, 46, 39–51.
772 <https://doi.org/10.1016/j.humov.2015.11.017>
- 773 Silva, P., Duarte, R., Esteves, P., Travassos, B., & Vilar, L. (2016). Application of entropy measures to analysis of
774 performance in team sports. *International Journal of Performance Analysis in Sport*, 16(2), 753–768.
775 <https://doi.org/10.1080/24748668.2016.11868921>
- 776 Silva, P., Vilar, L., Davids, K., Araújo, D., & Garganta, J. (2016). Sports teams as complex adaptive systems:
777 Manipulating player numbers shapes behaviours during football small-sided games. *SpringerPlus*, 5(1), 191.
778 <https://doi.org/10.1186/s40064-016-1813-5>
- 779 Stagno, K. M., Thatcher, R., & van Someren, K. A. (2007). A modified TRIMP to quantify the in-season training load
780 of team sport players. *Journal of Sports Sciences*, 25(6), 629–634. <https://doi.org/10.1080/02640410600811817>
- 781 Stergiou, N., Buzzi, U., Kurz, M., & Heidel, J. (2004). *Nonlinear tools in human movement*. Champaign, IL, Human
782 Kinetics.
- 783 Teixeira, J. E., Alves, A. R., Ferraz, R., Forte, P., Leal, M., Ribeiro, J., Silva, A. J., Barbosa, T. M., & Monteiro, A.
784 M. (2022). Effects of chronological age, relative age, and maturation status on accumulated training load and
785 perceived exertion in young sub-elite football players. *Frontiers in Physiology*, 13, 832202.
- 786 Teixeira, J. E., Forte, P., Ferraz, R., Leal, M., Ribeiro, J., Silva, A. J., Barbosa, T. M., & Monteiro, A. M. (2021a).
787 Monitoring accumulated training and match load in football: A systematic review. *International Journal of*
788 *Environmental Research and Public Health*, 18(8), 3906. <https://doi.org/10.3390/ijerph18083906>
- 789 Teixeira, J. E., Forte, P., Ferraz, R., Leal, M., Ribeiro, J., Silva, A. J., Barbosa, T. M., & Monteiro, A. M. (2021b).
790 Quantifying sub-elite youth football weekly training load and recovery variation. *Applied Sciences*, 11(11), 4871.
791 <https://doi.org/10.3390/app11114871>
- 792 Teixeira, J. E., Leal, M., Ferraz, R., Ribeiro, J., Cachada, J. M., Barbosa, T. M., Monteiro, A. M., & Forte, P. (2021).
793 Effects of match location, quality of opposition and match outcome on match running performance in a
794 Portuguese professional football team. *Entropy*, 23(8), 973. <https://doi.org/10.3390/e23080973>
- 795 Teixeira, J., Forte, P., Ferraz, R., Branquinho, L., Silva, A., Barbosa, T., & Monteiro, A. (2022). *Methodological*
796 *Procedures for Non-Linear Analyses of Physiological and Behavioural Data in Football*.
797 <https://doi.org/10.5772/intechopen.102577>

- 798 Travassos, B., Araújo, D., Vilar, L., & McGarry, T. (2011). Interpersonal coordination and ball dynamics in futsal
799 (indoor football). *Human Movement Science*, 30(6), 1245–1259. <https://doi.org/10.1016/j.humov.2011.04.003>
- 800 Travassos, B., Davids, Araujo, D., & Esteves, P. (2013). Performance analysis in team sports: Advances from an
801 ecological dynamics approach. *International Journal of Performance Analysis in Sport*, 13, 89–95.
802 <https://doi.org/10.1080/24748668.2013.11868633>
- 803 Travassos, B., Gonçalves, B., Marcelino, R., Monteiro, R., & Sampaio, J. (2014). How perceiving additional targets
804 modifies teams' tactical behavior during football small-sided games. *Human Movement Science*, 38, 241–250.
805 <https://doi.org/10.1016/j.humov.2014.10.005>
- 806 Vidal-Codina, F., Evans, N., Fakir, B. E., & Billingham, J. (2022). *Automatic event detection in football using tracking*
807 *data* (arXiv:2202.00804). arXiv. <https://doi.org/10.48550/arXiv.2202.00804>
- 808 Vilar, L., Araújo, D., Davids, K., Travassos, B., Duarte, R., & Parreira, J. (2014). Interpersonal coordination
809 tendencies supporting the creation/prevention of goal scoring opportunities in futsal. *European Journal of Sport*
810 *Science*, 14(1), 28–35. <https://doi.org/10.1080/17461391.2012.725103>
- 811 Welsh, A. H., & Knight, E. J. (2015). “Magnitude-based Inference”: A Statistical Review. *Medicine and Science in*
812 *Sports and Exercise*, 47(4), 874–884. <https://doi.org/10.1249/MSS.0000000000000451>
- 813 Wisbey, B., Montgomery, P. G., Pyne, D. B., & Rattray, B. (2010). Quantifying movement demands of AFL football
814 using GPS tracking. *Journal of Science and Medicine in Sport*, 13(5), 531–536.
815 <https://doi.org/10.1016/j.jsams.2009.09.002>
- 816 Yentes, J. M., Hunt, N., Schmid, K. K., Kaipust, J. P., McGrath, D., & Stergiou, N. (2013). The appropriate use of
817 approximate entropy and sample entropy with short data sets. *Annals of Biomedical Engineering*, 41(2), 349–
818 365. <https://doi.org/10.1007/s10439-012-0668-3>
- 819 Yuill, R. S. (1971). The Standard Deviational Ellipse; An Updated Tool for Spatial Description. *Geografiska Annaler:*
820 *Series B, Human Geography*, 53(1), 28–39. <https://doi.org/10.1080/04353684.1971.11879353>

Table 1 (on next page)

Search terms and following keywords in the screening procedures of systematic review.

Search terms and following keywords in the screening procedures of systematic review.

1 **Table 1** – Search terms and following keywords in the screening procedures of systematic review.

Search Term	Keywords
Population	1 “soccer” OR “football” OR “Association football”
Intervention	2 “integrated” OR “integration” OR “comparison” OR “integration”
Comparison/ Outcomes	3 Physiological set: “training load” OR “external training load” OR “internal training load” OR “physical performance” OR “physiological performance” OR “physical response” OR “physical demands” OR “physiological response” OR “physiological demands” OR “activity profile” OR “time-motion” OR “workload” OR “work-rate” OR “loading” OR “match running performance” OR “match load” OR “match demands” OR “weekly load” OR “heart rate” OR “TRIMP” OR “perceived exertion” OR “distances” OR “sprint” OR “acceleration” OR “deceleration” OR “metabolic power” OR “energy cost” OR “high intensity” OR “running” OR “conditioning” OR “fitness” OR “biomechanics” OR “kinetic” OR “kinematic” OR “physiology” Positional data: “positional” OR “positioning” OR “behavioral data” OR “behaviour data” OR “tactical behavior” OR “tactical behaviour” OR “collective behavior” OR “collective behaviour” OR “team behavior” OR “team behavior” OR “movement behavior” OR “movement behaviour” OR “patterns” OR “constraints” OR “interpersonal coordination” OR “inter-personal coordination” OR “intra-team dyads” OR “inter-team dyads” OR “synchronization” OR “synergy” OR “tactical adjustments” OR “game dynamics” OR “dynamic” OR “variability” OR “stability” OR “regularity” OR “predictability” OR “spatial-temporal” OR “spatio-temporal” OR “complex systems” OR “dynamical systems” OR “complexity” OR “self-organization” OR “self-similarity” OR “self-organization” OR “chaos”
<i>Boolean phrase</i>	4 (((#4) AND #3) AND #2) AND #1

2

3

Table 2 (on next page)

Summary of the sampling characteristics in the studies included for systematic review and its quality score.

Summary of the sampling characteristics in the studies included for systematic review and its quality score.

1

Table 2 – Summary of the sampling characteristics in the studies included for systematic review and its quality score.

Reference (year)	Study design	Population, Competitive Level	Sample (N)	Sex	Age (y)	Expertise level (y)
Baptista et al. (2020)	RCT	Adult, Semiprofessional	23	Male	24.9 ± 6.5	12.6 ± 5.5
Canton et al. (2021)	RCT	Youth, High-Level	24	Male	U12: 11.3 ± 0.8	U12: 3.13 ± 1.5
Coutinho et al. (2017)	RCT	Youth, Amateur	12	Male	15.9 ± 0.8	8.9 ± 2.4
Coutinho et al. (2019)	RCT	Youth, Amateur	12	Male	15.9 ± 0.8	8.9 ± 2.4
Coutinho, et al. (2019)	RCT	Youth, ND	40	Male	U13 (n=20): 11.3 ± 0.8 U15 (n=20): 13.3 ± 0.6	U13 (n=20): 4.9 ± 2.7 U15 (n=20): 7.0 ± 1.6
Coutinho et al. (2020)	RCT	Youth, ND	10	Male	13.7 ± 0.5	6.1 ± 0.9
Coutinho et al. (2022)	RCT	Youth, ND	114	Male	U9: 7.9 ± 0.9 U11: 9.5 ± 0.9 U13: 11.6 ± 0.8 U15: 13.9 ± 0.6 U17: 16.2 ± 0.7 U19: 17.9 ± 0.4	U9: 2.7 ± 1.1 U11: 3.9 ± 1.2 U13: 4.9 ± 2.0 U15: 6.8 ± 2.5 U17: 7.9 ± 2.8 U19: 9.5 ± 2.1
Coutinho et al. (2022)	RCT	Youth, Regional Level	20	Male	16.1 ± 0.9	7.5 ± 3.4
Coutinho et al. (2022)	RCT	Youth, ND	21	Male	16.2 ± 0.6	8.3 ± 2.8
Ferraz et al. (2020)	RCT	Adult, Professional	20	Male	22.3 ± 2.1	10.3 ± 3.4
Figueira et al. (2018)	RCT	Youth, Elite	22	Male	U15 (n=22): 13.6 ± 0.4 U17 (n=22): 15.3 ± 0.4	U15 (n=22): 5.1 ± 1.3 U17 (n=22): 7.2 ± 1.4
Folgado et al. (2015)	Observational cohort	Adult, Professional	23	Male	25.5 ± 3.6	9.0 ± 3.7
Folgado et al. (2018)	Observational cohort	Adult, Professional	30	Male	23.7 ± 4.2	4.8 ± 4.2
Folgado et al. (2019)	RCT	Youth, National Level	20	Male	U15: 14.1 ± 0.5	U15: 6.4 ± 1.8
Gonçalves et al. (2014)	Observational cohort	Youth, Elite	22	Male	18.1 ± 0.7	9.4 ± 1.3
Gonçalves et al. (2017)	RCT	Adult, Professional	19	Male	25.1 ± 4.1	18.8 ± 5.3
Gonçalves et al. (2018)	Observational cohort	Adult, Professional	28	Male	24.7 ± 4.7	6.5 ± 4.7
Jara et al. (2019)	RCT	Adult, Elite	3	Male	24.7 ± 7.2	11.0 ± 4.7
Machado et al. (2022)	RCT	Youth, Recreational	10	Male	16.89 ± 0.11	ND
Nieto et al. (2022)	RCT	Youth, Elite	22	Male	14.6 ± 0.3	5.5 ± 0.5
Olthof et al. (2018)	Observational cohort	Youth, Professional	148	Male	U13 (n=36): 12.5 ± 0.5 U15 (n=43): 14.4 ± 0.5 U17 (n=28): 16.6 ± 3.2 U19 (n=43): 17.9 ± 1.0	ND
Praça et al. (2016)	Observational cohort	Youth, National Level	18	Male	16.4 ± 0.7	4.2 ± 0.0
Praça et al. (2021)	Observational cohort	Youth, National Level	50	Male	U17 (n=25): 16.79 ± 0.61 U20 (n=25): 19.08 ± 0.61	ND
Ric et al. (2016)	RCT	Adult, Professional	8	Male	26 ± 4.96	19.6 ± 4.9
Ric et al. (2017)	RCT	Adult, Professional	21	Male	25.1 ± 4.1	18.8 ± 5.3
Sampaio et al. (2014)	Observational cohort	Adult, volunteer	24	Male	20.8 ± 1.0	5.2 ± 1.3
All studies	-	-	764	-	16.81 ± 1.63	4.2 ± 3.83

Abbreviations: ND – Not described; U – Under; QS – Quality Score; RCT – randomized controlled trial; y – years.

2

Table 3 (on next page)

Modified Downs and Black scale for reviewed intervention studies.

Modified Downs and Black scale for reviewed intervention studies.

1
2**Table 3** – Modified Downs and Black scale for reviewed intervention studies.

Reference (year)	Item 1	Item 2	Item 3	Item 6	Item 7	Item 10	Item 12	Item 15	Item 16	Item 18	Item 20	Item 22	Item 23	Item 25	Total score (out of 14)
Folgado et al. (2015)	1	1	1	1	0	1	1	0	1	0	1	1	0	0	9
Folgado et al. (2018)	1	1	1	0	1	1	1	0	0	1	1	1	1	0	10
Gonçalves et al. (2014)	1	1	1	1	0	1	1	1	1	0	0	0	0	1	8
Gonçalves et al. (2018)	1	1	1	1	1	0	1	0	1	1	1	0	1	1	11
Olthof et al. (2018)	1	1	1	1	0	1	1	1	1	0	0	1	0	1	10
Praça et al. (2016)	1	1	1	1	0	1	1	0	1	1	0	0	1	0	9
Praça et al. (2021)	1	1	1	1	1	1	1	1	1	0	0	1	0	1	11
Sampaio et al. (2014)	1	1	0	1	1	0	1	0	1	1	0	0	1	0	8

Notes: 0 = no; 1 = yes; U = unable to determine. Item 1: clear aim/hypothesis; Item 2: outcome measures clearly described; Item 3: patient characteristics clearly described; Item 6: main findings clearly described; Item 7: measures of random variability provided; Item 10: actual probability values reported; Item 12: participants prepared to participate representative of entire population; Item 15: blinding of outcome measures; Item 16: analysis completed was planned; Item 18: appropriate statistics; Item 20: valid and reliable outcome measures; Item 22: participants recruited over same period; Item 23: randomised; Item 25: adjustment made for confounding variables.

3
4
5
6
7

Table 4(on next page)

Physiotherapy evidence database scale (PEDro) for reviewed intervention groups.

Physiotherapy evidence database scale (PEDro) for reviewed intervention groups.

1

Table 4 – Physiotherapy evidence database scale (PEDro) for reviewed intervention groups.

Reference (year)	Item 1	Item 2	Item 3	Item 4	Item 5	Item 6	Item 7	Item 8	Item 9	Item 10	Item 11	Total score (out of 11)
Baptista et al. (2020)	1	1	1	1	1	1	1	0	1	1	0	9
Canton et al. (2021)	1	1	1	1	1	0	1	1	1	0	0	8
Coutinho et al. (2017)	1	1	1	1	1	1	1	1	1	0	0	9
Coutinho et al. (2019)	1	1	1	1	1	1	1	0	1	1	0	9
Coutinho, et al. (2019)	0	1	1	1	1	0	0	1	1	0	0	6
Coutinho et al. (2020)	1	1	1	0	1	0	1	1	1	0	1	8
Coutinho et al. (2022)	1	1	1	1	1	1	1	0	1	1	0	9
Coutinho et al. (2022)	0	1	1	1	1	1	1	0	1	0	1	8
Coutinho et al. (2022)	1	1	1	1	1	1	1	0	1	0	1	9
Ferraz et al. (2020)	1	1	0	1	1	1	1	1	1	0	0	8
Figueira et al. (2018)	1	1	1	1	1	1	1	1	1	0	0	9
Folgado et al. (2019)	1	1	1	1	1	1	1	1	1	0	0	9
Gonçalves et al. (2017)	1	1	1	1	1	0	1	0	1	0	0	7
Jara et al. (2019)	1	1	1	1	0	1	1	1	1	0	0	8
Machado et al. (2022)	1	1	1	1	1	1	1	0	1	0	0	8
Nieto et al. (2022)	1	1	1	1	0	1	1	0	0	1	0	7
Ric et al. (2016)	1	1	1	1	1	1	1	0	1	1	0	9
Ric et al. (2017)	1	1	1	1	1	0	1	0	1	0	0	7

Notes: 0 = item was not satisfied; 1 = item was satisfied. Item 1: eligibility criteria were specified; Item 2: subjects were randomly allocated to groups; Item 3: allocation was concealed; Item 4: the groups were similar at baseline regarding the most important prognostic indicators; Item 5: there was blinding of all subjects; Item 6: there was blinding of all therapists who administered the therapy; Item 7: there was blinding of all assessors who measured at least one key outcome; Item 8: measurements of at least one key outcome were obtained from more than 85% of the subjects initially allocated to groups; Item 9: all subjects for whom outcome measurements were available received the treatment or control condition as allocated, or where this was not the case, data for at least one key outcome were analysed by “intention to treat”; Item 10: the results of between groups statistical comparisons are reported for at least one key outcome; Item 11: the study provides both point measurements and measurements of variability for at least one key outcome.

2
3
4
5
6
7
8
9
10

Table 5 (on next page)

Summary of performance dimensions, measures, measurements and their thresholds/metric formulas in the included articles.

Summary of performance dimensions, measures, measurements and their thresholds/metric formulas in the included articles.

Table 5 – Summary of performance dimensions, measures, measurements and their thresholds/metric formulas in the included articles.

Dimension	Measure	Measurement	Description, thresholds and/or metric formula	Reference	Further reading	
Physical data	External load	Movement speed	Higher Ratio (> 16 km·h ⁻¹), Moderate Ratio (10.0–15.9 km·h ⁻¹), Lower Ratio: 7.0–9.9 km·h ⁻¹) with distance covered at very low intensities (0.0–6.9 km·h ⁻¹)	(Canton et al., 2021; Coutinho et al., 2019; Coutinho et al., 2019)	(Abade et al., 2014; Bradley & Ade, 2018; Hodgson et al., 2014)	
			TD (m)	Walking (0.0–3.5 km·h ⁻¹), Jogging (3.6–14.3 km·h ⁻¹), Running (14.4–19.7 km·h ⁻¹), and Sprinting (>19.8 km·h ⁻¹)	(Coutts & Duffield, 2010; Ferraz et al., 2020; Figueira et al., 2018; Folgado et al., 2015, 2018; Gonçalves et al., 2017; Gonçalves et al., 2018; Ric et al., 2016; Ric et al., 2017)	(Duarte et al., 2013; Folgado et al., 2014; Giménez et al., 2018; Gonçalves et al., 2017)
				Zone 1 (0–6.9 km·h ⁻¹); zone 2 (7–9.9 km·h ⁻¹); zone 3 (10–12.9 km·h ⁻¹); zone 4 (13–15.9 km·h ⁻¹); zone 5 (16–17.9 km·h ⁻¹) and zone 6 (≥18 km·h ⁻¹).	(Sampaio et al., 2014)	(Hill-Haas et al., 2008)
		High Speed (km·h ⁻¹)	Distance covered in the high ratio/distance covered in walking multiplied by 100.	(Coutinho et al., 2019; Coutinho et al., 2019)	(Abade et al., 2014; Gonçalves et al., 2018)	
			Distance covered at high intensity (≥ 19.8 km·h ⁻¹) and number of sprints (frequency of displacements ≥ 25.2 km·h ⁻¹)	(Olthof et al., 2018)	(Abt & Lovell, 2009; Goto et al., 2015)	
			Distance covered at three speed zones (14.40–19.79 km·h ⁻¹ , 19.80–22.99 km·h ⁻¹ , higher than 23.00 km·h ⁻¹) and number of sprints (frequency of displacements ≥ 23.00 km·h ⁻¹)	(Praça et al., 2021)	(Mallo et al., 2015; Praça et al., 2020)	
		Game Pace or Average Speed (km·h ⁻¹ or CV)	Players' average speed displacement, expressed as meters or CV.	(Canton et al., 2021; Coutinho et al., 2019; Coutinho et al., 2019)	(Ferraz et al., 2017, 2018; Gonçalves et al., 2019)	
		ACC/DEC (m·s ⁻²)	ACC: 0.5-3.0 m·s ⁻² ; DEC: > -3.0 0 m·s ⁻²	(Coutinho et al., 2017)	(Dalen et al., 2016; Russell et al., 2016)	
		Body load	$\frac{\sqrt{(a_{y1} - a_{y-1})^2 + a_{x1} - a_{x-1})^2 + (a_{z1} - a_{z-1})^2}}{100}$	(Gonçalves et al., 2017)	(Buchheit et al., 2014)	
		Positional data	Spatial and temporal features	Spatial movement variability/regularity	CV	Magnitude of the variability in the distance between players', expressed by the coefficient of variation CV (%)
ApEn	Ranged 0 to 2, in which lower values correspond to more repeatable patterns). The imputed values used to compute were 2 to vector length (m) and 0.2*std to the tolerance (r).				(Baptista et al., 2020; Coutinho et al., 2020; Coutinho et al., 2019; Ferraz et al., 2020; Figueira et al., 2018; Gonçalves et al., 2017; Sampaio et al., 2014)	(Duarte et al., 2012; Gonçalves et al., 2016; Gréhaigne et al., 1997; Pincus, 1991; Preatoni et al., 2010; Richman & Moorman, 2000; Seifert et al., 2013; Silva, Duarte, et al., 2016; Stergiou et al., 2004; Yentes et al., 2013)
Boltzmann-Gibbs-Shannon entropy	Probabilities of configurations were calculated as limit (large N) relative frequencies for stationary distributions: $p_i = n_i/N$ where n_i and N is the frequency and number of the configuration respectively.				(Ric et al., 2016)	(Balescu, 1975)
MSE SamEn	SampEn and MSE curves to a range of different timescales, calculating the area under and complexity index.				(Canton et al., 2021)	(Busa & van Emmerik, 2016; M. Costa et al., 2005)
Complex Index	SEI			Width and length displacements from each positioning time series to the mean position.	(Canton et al., 2021; Figueira et al., 2018; Gonçalves et al., 2017; Praça et al., 2021)	(Arede et al., 2021; Clemente et al., 2018; Clemente et al., 2018; Travassos et al., 2014)
	Stretch Index (meters or ApEn)			Men of the distances between each player and the geometric centre of the team.	(Coutinho et al., 2019; Olthof et al., 2018; Praça et al., 2021)	(Bourbousson et al., 2010; Clemente et al., 2013; Clemente et al., 2018; Duarte et al., 2013; Lames et al., 2010; Travassos et al., 2014)
	Dynamic overlap <qd(t)>			Average cosine auto-similarity of the overlap between configurations with increasing time lag $\langle q_d(t) \rangle = (1 - q_{stat})t^\alpha + q_{stat}$	(Ric et al., 2016)	(Hristovski et al., 2013; Saxton, 1996)
	Trapping strength			Probability of remaining inside the same attractor that is a conditional probability of a configuration being subsequently repeated (i.e., trapping strength and behavioural flexibility).		

		Voronoi algorithms	Voronoi algorithms allow to compute a diagram represented by spatial cells for individual positional area (m ²).	(Baptista et al., 2020)	(Fonseca et al., 2012)	
Coordination/synchronization using intra-team dyads		Relative phase (Hilbert transform)	Longitudinal and lateral directions using near-in-phase synchronization of each dyad that was quantified by the percentage of time spent between -30° to 30° bin.	(Coutinho et al., 2020; Coutinho et al., 2019; Fernandes et al., 2010; Figueira et al., 2018; Folgado et al., 2015, 2019; Gonçalves et al., 2018; Gonçalves et al., 2017)	(Duarte et al., 2012; Duarte et al., 2012; Duarte et al., 2013; Folgado et al., 2014; Gonçalves et al., 2019; McGarry et al., 2002; Palut & Zanone, 2005; Sampaio & Vitor, 2012; Silva et al., 2016; Travassos et al., 2011a, 2011b, 2013)	
		Speed synchronisation	0.0–3.5 km·h ⁻¹ (low intensity); 3.6–14.3 km·h ⁻¹ (moderate intensity); 14.4–19.7 km·h ⁻¹ (high intensity); and >19.8 km·h ⁻¹ (very high intensity).	(Folgado et al., 2018; Gonçalves et al., 2018)	(Folgado et al., 2014)	
		Distance Player–Teammate	Interpersonal distance between each pair of players, both with teammates and opponents.			
Coordination/synchronization using inter-team dyads		Distance Player–Opponent	$D(a_{x(t)y(t)}, b_{x(t)y(t)}) = \sqrt{(a_{x(t)y(t)} - b_{x(t)y(t)})^2 + (a_{y(t)x(t)} - b_{y(t)x(t)})^2}$ Where D is the distance, a is the player, x and y are the coordinates, and t is the time, and b is the teammate or opponent.	(Ferraz et al., 2020; Olthof et al., 2018; Ric et al., 2017; Sampaio et al., 2014)	(Gonçalves et al., 2014; Low et al., 2020; Silva et al., 2016)	
		Distance from the target	Distance from the target according to ten categories: >37.45 m; 32.1 ± 37.45 m; 36.75 ± 32.1 m; 21.4 ± 26.75 m; 16.05 ± 21.4 m; 10.7 ± 16.05 m; 5.35 ± 10.7 m; 0 ± 5.35 m.	(Ric et al., 2017)	(Duarte, Araújo, Davids, et al., 2012)	
Playing Space		Total surface area or team effective playing space (m ²)	Smallest convex hull, that is the smallest polygonal area that it is delimited by the peripheral players	(Coutinho et al., 2017; Folgado et al., 2019; Olthof et al., 2018)	(Duarte et al., 2013; Folgado et al., 2014; Mendes et al., 2007; Ric, Torrents, et al., 2016; Russell et al., 2016; Sampaio & Vitor, 2012)	
		Ellipses: SEA and PEA areas	Spatial analysis for a set of points in a two-dimensional space, which boundaries will enclose about the 100(1 - α): $\bar{x} = \frac{1}{n} \sum_{i=0}^n x_i$, $\bar{y} = \frac{1}{n} \sum_{i=0}^n y_i$	(Jara et al., 2019)	(Batschelet, 1981; Leffever, 1926; Yuill, 1971)	
		Team's width and length				
		Team's speed contraction dispersion	Longitudinal position of team geometrical center (x axis) and lateral position of team geometrical center (y axis), expressed as meters, CV and length-per-width (LPW) ratio per team.	(Baptista et al., 2020; Canton et al., 2021; Coutinho et al., 2017; Gonçalves et al., 2017; Praça et al., 2021)	(Duarte et al., 2013; Folgado et al., 2014; Frencken et al., 2011; Mendes et al., 2007; Ric et al., 2016)	
		Team centroid				
Other dimensions (non-positional data)	Technical variables	Individual actions	Passes, dribbles and shots	Successful Passes (%), Successful Dribbles (%), Shots on Target (%), Goals (%)	(Coutinho et al., 2020)	(Liu et al., 2016; O'Donoghue, 2009; Santos et al., 2018)
				Distance covered at different intensities when dribbling; number of completed passes; completed passes distance; shots distance to the goal; distance between attacker and defender when shooting	(Folgado et al., 2019; Olthof et al., 2018)	(Rampinini et al., 2009)
				Successful pass reception and turnovers, goals scored and relative frequencies of players' passing interactions.	(Ric et al., 2017)	(Costa et al., 2005)
				Transition probabilities were calculated dividing the number of each player's passes to his teammates, turnovers and goals by the total number of player interactions.		
Tactical variables		Ball Possession	Offensive/Defensive Phases	Duration of possession, team width, team length and their ratio (LPWR), as well as their coefficient of variation	(Canton et al., 2021; Olthof et al., 2018; Ric, Hristovski, et al., 2016)	(Collet, 2013; Costa et al., 2011; Gabin et al., 2012)
		Network	Dyad Nodes	Relative phase analysis was also divided according to each dyad average speed in three levels: for the whole team; for dyads with similar synchronisation tendencies; and for each dyad.	(Folgado et al., 2015; Ric et al., 2017)	(McGarry et al., 2002; Travassos et al., 2011a, 2013)
		Tactical actions	Patterns/Categories	Tactical actions classified as: penetration, offensive coverage, depth mobility, width and length, offensive unity, delay, defensive coverage, balance, concentration, defensive unity.	(Ric et al., 2017)	(M. Costa et al., 2005)
Psychophysiological variables	Perceived Exertion	RPE (a.u.)	CR10-scale (0 to 10 arbitrary units).	(Coutinho et al., 2017)	(Lee et al., 1991)	

	Exertion index	Wisbey's formula: players' instantaneous speed (over 10 s and speed over 60 s).	(Folgado et al., 2015)	(Wisbey et al., 2010)
Heart Rate	HR _{max} (bpm)	Percentage of HR _{max} into intensity zones: Zone 1 (<75% HR _{max}), Zone 2 (75–84.9% HR _{max}), Zone 3 (85–89.9% HR _{max}), and Zone 4 (≥90% HR _{max}).	(Folgado et al., 2018; Sampaio et al., 2014)	(Abt & Lovell, 2009; Gore, 2000)
	Average HR (bpm)	Average beats per minute (BPM)	(Gonçalves et al., 2017)	ND
	TRIMP _{MOD}	Total TRIMP _{MOD} : zone 1 (65–71% HR _{max})*1.25; zone 2 (72–78% HR _{max})*1.71; zone 3 (79–85% HR _{max})*2.54; zone 4 (86–92% HR _{max})*3.61; and zone 5 (93–100% HR _{max})*5.16	(Folgado et al., 2018)	(Campos-Vazquez et al., 2015; Los Arcos et al., 2014; Stagno et al., 2007)

Abbreviations: $\langle \text{qd}(t) \rangle$ – Dynamic overlap; ACC/DEC – Accelerations and decelerations; ApEn – Approximate entropy; BPM – Average beats per minute; CR-10 – Borg CR10 scale; CV – Coefficient of variation; D – Distance; HR – Heart Rate; HR_{max} – maximum Heart Rate; LPWR – Team width, team length and their ratio or LPW – length-per-width ratio per team; MSE – Multiscale Entropy; PEA – Prediction Ellipse; RPE – Ratings of Perceived Exertion; SamEn – Sample Entropy; SEA – Standard Ellipse; SEI – Spatial exploration index; TD – Total distance; TRIMP – Training Impulse; TRIMP_{MOD} – modified Training Impulse.

Table 6 (on next page)

Methodological approaches of included articles.

Methodological approaches of included articles.

Table 6 – Methodological approaches of included articles.

Referente (year)	Study Purpose	Experimental Approach						Methodological procedures	Data Collection (Device specification)	Statistical and mathematical analysis
		Match-play	Training set	Game format	Physical/Physiological	Positional/Tactical	Other dimensions			
Baptista et al. (2020)	Identified the effects of playing formations on tactical behaviour and external workload during SSG.	×	SSG	GK + 7 vs 7 + GK	✓	✓	×	Matlab® routines (Math-Works, Inc., Massachusetts, USA)	Non-differential 5Hz GPS (ND)	Cohen' <i>d</i> SWC MBI ApEn Voronoi algorithme
Canton et al. (2021)	Identified how positioning the goals in diagonal configurations on the pitch modifies the external training load and the tactical behaviour during SSG.	×	SSG	GK + 5 vs 5 + GK	✓	✓	×	Matlab® routines (Math-Works, Inc., Massachusetts, USA) Lince software® (Gabin et al., 2012),	10 Hz GPS units (WIMU PRO, RealTrack Systems, Almeria, Spain)	Cohen' <i>d</i> SWC MBI NHST
Coutinho et al. (2017)	Examined the effects of mental fatigue and additional corridor and pitch sector lines on players' physical and tactical performances during SSG.	×	SSG	GK + 6 vs 6 + GK	✓	✓	✓	Matlab® routines (Math-Works, Inc., Massachusetts, USA)	CR10-scale (RPE) Portable optical timing system (Optojump, Microgate, Bolzano, Italy) 15Hz GPS (SPIPRO, GPSports, Canberra, ACT, Australia)	Cohen' <i>d</i> SWC NHST
Coutinho et al. (2019)	Identified the effects of adding spatial references during SSG on players' tactical and physical performance.	×	SSG	GK + 6 vs 6 + GK	✓	✓	✓	Matlab® routines (Math-Works, Inc., Massachusetts, USA)	5 Hz GPS (SPI-PRO, GPSports, Canberra, ACT, Australia)	CV ApEn MBI NHST
Coutinho et al. (2019)	Identified the effects of different pitch configurations on players' positional and physical performance.	×	SSG	GK + 5 vs 5 + GK	✓	✓	✓	Matlab® routines (Math-Works, Inc., Massachusetts, USA)	5 Hz GPS (SPI-PRO, GPSports, Canberra, ACT, Australia)	CV ApEn MBI NHST
Coutinho et al. (2020)	Compared players' performances when manipulating the external markings of the pitch during SSG.	×	SSG	GK + 5 vs 5 + GK	✓	✓	✓	Matlab® routines (Math-Works, Inc., Massachusetts, USA) LongoMatch software (Longomatch, version 1.3.7., Fluendo)	5 Hz GPS (SPI-PRO, GPSports, Canberra, ACT, Australia) Digital video camera (Sony NV-GS230)	CV ApEn MBI NHST
Coutinho et al. (2022)	Explored how the number of allowed ball touches per player possession affected the performance of different age groups during SSG.	×	SSG	GK + 6 vs 6 + GK	✓	✓	✓	Matlab® routines (Math-Works, Inc., Massachusetts, USA) LongoMatch software (Longomatch, version 1.3.7., Fluendo)	5 Hz GPS (SPI-PRO, GPSports, Canberra, ACT, Australia)	Cohen' <i>d</i> SWC NHST Hilbert transform
Coutinho et al. (2022)	Aimed to identify the effects of playing with additional individual, collective or individual-collective variability on players' performance during SSG.	×	SSG	GK + 6 vs 6 + GK	✓	✓	✓	Matlab® routines (Math-Works, Inc., Massachusetts, USA) LongoMatch software (Longomatch, version 1.3.7., Fluendo)	5 Hz GPS (SPI-PRO, GPSports, Canberra, ACT, Australia)	Cohen' <i>d</i> NHST Hilbert transform
Coutinho et al. (2022)	Explored how manipulating the colour of training vests affects footballers' individual and collective performance during SSG.	×	SSG	GK + 6 vs 6 + GK	✓	✓	✓	Matlab® routines (Math-Works, Inc., Massachusetts, USA) LongoMatch software (Longomatch, version 1.3.7., Fluendo)	5 Hz GPS (SPI-PRO, GPSports, Canberra, ACT, Australia)	Cohen' <i>d</i> NHST Hilbert transform
Ferraz et al. (2020)	Identified how the manipulation of knowledge regarding a training task duration constrains the pacing and tactical behaviour in SSG.	×	SSG	GK + 5 vs 5 + GK	✓	✓	✓	ND	5 Hz GPS (SPI-Pro X II, GPS ports, Canberra, ACT, and Australia)	ApEn MBI
Figueira et al. (2018)	Compared footballers' performances when playing with teammates and opponents from the same or different age groups	✓	-	GK + 11 vs 11 + GK	✓	✓	×	Matlab® routines (Math-Works, Inc., Massachusetts, USA)	5 Hz GPS (SPI-Pro X II, GPS ports, Canberra, ACT, and Australia)	CV ApEn MBI
Folgado et al. (2015)	Examined the physical and tactical performances under congested and non-congested fixture periods	✓	-	GK + 11 vs 11 + GK	✓	✓	×	Matlab® routines (Math-Works, Inc., Massachusetts, USA)	Semiautomatic tracking system (Prozone®, ProZone Holdings Ltd, UK).	CHAID Cohen' <i>d</i> NHST Hilbert transform
Folgado et al. (2018)	Identified changes in tactical, physical and physiological performances in LSG during the preseason.	×	LSG	GK + 8 vs. 8 + GK	✓	✓	×	Matlab® routines (Math-Works, Inc., Massachusetts, USA)	5 Hz GPS (SPI-Pro X II, GPS ports, Canberra, ACT, and Australia)	NHST Cohen' <i>d</i>
Folgado et al. (2019)	Compared players' performance during two SSG with different pitch orientation (i.e., 40x30m vs 30x40m).	×	SSG	GK + 4 vs 4 + GK	✓	✓	×	Matlab® routines (Math-Works, Inc., Massachusetts, USA)	10 Hz GPS (Minimax S5, Catapult Innovations, Australia). Digital Video Camera (Canon PowerShot SX720 HS, Canon Inc,	ICC SEM Cohen' <i>d</i> SWC

									Tokyo, Japan),	MBI
Gonçalves et al. (2014)	Identified differences in time-motion, modified training impulse, body load and movement behaviour between defenders, midfielders and forwards, during an 11-a-side simulated football game.	✓	-	GK + 11 vs 11 + GK	✓	✓	×	Matlab® routines (Math-Works, Inc., Massachusetts, USA)	5 Hz GPS (SPI-Pro X II, GPS ports, Canberra, ACT, and Australia) 1 Hz short-range radio telemetry (Polar Team Sports System, Polar Electro Oy, Finland)	Hilbert transform ApEn NHST
Gonçalves et al. (2017)	Identified how pitch area restrictions affect the players' tactical behavior, physical, and physiological performances during LSG.	×	LSG	GK + 9 vs. 9 + GK GK+ 10 vs. 10 + GK	✓	✓	×	Matlab® routines (Math-Works, Inc., Massachusetts, USA)	5 Hz GPS (SPI-Pro X II, GPS ports, Canberra, ACT, and Australia) 1 Hz short-range radio telemetry (Polar Team Sports System, Polar Electro Oy, Finland)	Hilbert transform CV ApEn Cohen' d SWC SWC MBI
Gonçalves et al. (2018)	Examined the changes in the players' speed synchronization and physical performance between the first and the second half (15- min time). Explored the match-to-match variation of players' speed synchronization performance.	✓	-	GK + 11 vs 11 + GK	✓	✓	×	Matlab® routines (Math-Works, Inc., Massachusetts, USA)	Match Analysis Camera System®.	Hilbert transform CV Cohen' d SWC MBI
Jara et al. (2019)	Analyzed how the modification of the pitch size in SSGs affects the GK's physical demands.	×	SSG MSG LSG	ND	✓	✓	×	Matlab® routines (Math-Works, Inc., Massachusetts, USA)	18.18 Hz GPS (GPEXE GK, Exelio SRL, Udine, Italy)	Cohen' d MBI NHST
Machado et al. (2022)	Investigated how different strategies of task constraint manipulation impact physical and tactical demands in small-sided and conditioned games (SSCG)	×	SSG	GK + 4 vs 4 + GK	✓	✓	×	Matlab® routines (Math-Works, Inc., Massachusetts, USA) and SPRO™ (RealTrack System, Almeria, Spain)	10 Hz GPS and inertial devices (WIMU Pro™ and GPS, RealTrack System, Almeria, Spain)	Effect size (ND) NHST
Nieto et al. (2022)	Described the effects on player's collective behaviour and physical response in three different pitch lengths (100, 75 and 50 m) keeping the width constant (60 m)	×	LSG	GK + 11 vs 11 + GK	✓	✓	×	Microsoft Excel Visual Basic for Applications (VBA) (Microsoft, Redmond, WA, USA)	10 Hz GPS (Minimax S5, Catapult Innovations)	Cohen' d MBI NHST SampEn
Olthof et al. (2018)	Investigated SSGs with a traditional small pitch and a match-derived relative pitch area in youth elite soccer players.	✓	SSG	GK + 4 vs 4 + GK GK + 11 vs 11 + GK	✓	✓	×	Matlab® routines (Math-Works, Inc., Massachusetts, USA)	Two HD video dome cameras (Bosch GmbH, Stuttgart, Germany) and one or two high resolution digital cameras (Canon HF100, Canon Inc., Tokyo, Japan; JVC Everio, JVC Kenwood Corporation, Kanagawa, Japan). LPM system (Inmotio Object Tracking BV., Amsterdam, the Netherlands)	NHST Effect sized eta-squared (η_p^2)
Praça et al. (2016)	Compared the collective tactical behavior between numerically balanced and unbalanced SSG.	×	SSG	3 vs 3 3 vs 3 + 2 4 vs 3	✓	✓	×	Matlab® routines (Math-Works, Inc., Massachusetts, USA)	15 Hz GPS (model SPI-Pro X2, GPSports, Canberra, Australia)	NHST Effect sized eta-squared (η_p^2)
Praça et al. (2021)	Analysed the effects of changing the match venue on match-related player's physiological, physical, and tactical responses with an age-dependent.	✓	-	GK + 11 vs 11 + GK	✓	✓	×	Matlab® routines (Math-Works, Inc., Massachusetts, USA)	10 Hz GPS device, with an embedded 200 Hz accelerometer and 1 Hz heart rate monitor (Polar®, Team Pro, Kempele, Finland).	NHST Effect sized eta-squared (η_p^2)
(Ric. et al., 2016)	Identified the dynamics of tactical behaviour emerging on different timescales in SSG. Quantified short- and long-term exploratory behaviour according to the number of opponents.	×	SSG	GK + 4 vs 3 + GK GK + 4 vs 5 + GK GK + 4 vs 7 + GK	✓	✓	×	Matlab® routines (Math-Works, Inc., Massachusetts, USA) Lince software® (Gabi net al., 2012)	Digital video camera for video recording and analysed an ad hoc instrument being used to notate tactical actions (Costa et al. 2011) 15 Hz GPS (SPI-ProX, GPS ports, Canberra, ACT, and Australia)	Dynamic overlap $\langle qd(t) \rangle$ Trapping strength Boltzmann–Gibbs–Shannon entropy NHST Cohen' d
(Ric et al., 2017)	Identified how players' spatial restrictions affected the exploratory tactical behaviour and constrained the perceptual-motor workspace in ball possession and the inter-player	×	ND	GK + 10 vs 9 + GK	✓	✓	×	Matlab® routines (Math-Works, Inc., Massachusetts, USA)	5 Hz GPS (SPI-Pro X II, GPS ports, Canberra, ACT, and Australia)	Dynamic overlap $\langle qd(t) \rangle$

 passing interactions.

(Sampaio et al., 2014)	Compared and discriminate the time-motion variables, heart rate and players' tactical behaviour according to game pace, status and team unbalance.	*	SSG	GK + 5 vs 5 + GK	✓	✓	*	Matlab® routines (Math-Works, Inc., Massachusetts, USA)	5 Hz GPS (SPI-Pro X II, GPS ports, Canberra, ACT, and Australia)	NHST SC
------------------------	--	---	-----	------------------	---	---	---	---	--	------------

Abbreviations: $D(t)>$ – Dynamic overlap; ApEn – Approximate entropy; CHAID – Chi-squared automatic interaction detection; Cohen's d – Standardized (Cohen) differences; CV – Coefficient of variation; GK – Goalkeeper; ICC – Intraclass correlation; LSG – Large-sided games; MBI – Magnitude-based inferences; MSG – Medium-sided games; ND – Not described; NHST – Null hypothesis statistical test; η_p^2 – Effect sized eta-squared; SampEn – Sample Entropy; SC – Structural coefficients; SEM – Standard error of measurement; SSG – Small-sided games; SWC – Smallest worthwhile changes; USA – United States; VSA – Visual Basic for Applications.

Figure 1

PRISMA flow diagram

