

What performance analysts need to know about research trends in association football (2012-2016): A systematic review

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What performance analysts need to know about research trends in Association Football (2012-2016) – A systematic review

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Abstract

Background

Evolving patterns of match analysis research need to be systematically reviewed regularly, since this area of work is burgeoning rapidly and studies can offer new insights to performance analysts if theoretically and coherently organized.

Objective

The purpose of this paper was to conduct a systematic review of published articles on match analysis in adult male football, identify and organize common research topics, and synthesize the emerging patterns of work between 2012 to 2016, according to PRISMA (Preferred Reporting Items for Systematic reviews and Meta-analyses) guidelines.

Methods

The Web of Science database was searched for relevant published studies and the following keywords were used: “football” and “soccer”, each one associated with the terms: “match analysis”, “performance analysis”, “notational analysis”, “game analysis”, “tactical analysis” and “patterns of play”.

Results

Of 483 studies initially identified, 77 were fully reviewed, and their outcome measures extracted and analysed. Results showed that research mainly focused on: (1) performance at set pieces – corner kicks; free kicks, penalty kicks; (2) collective system behaviours - captured by established variables such as team centroid (geometrical centre of a set of players), and team dispersion (quantification of how far players are apart), as

well tendencies for team communication (establishing networks based on passing sequences), sequential patterns (predicting future passing sequences), and group outcomes (relationships between match-related statistics and match final scores); (3) activity profile of players – playing roles, effects of fatigue, substitutions during matches, effects of environmental constraints on performance such as heat and altitude.

Conclusion

From the previous review, novel variables were identified that require new measurement techniques. It is evident that the complexity engendered during performance in competitive soccer requires an integrated approach that considers multiple aspects. A challenge for researchers is to align these new measures with the needs of the coaches, through a more integrated relationship between coaches and researchers to produce practical and usable information that improves player performance and coach activity.

Key-Points

There has been an increase in research examining performance in set-plays (free kicks, corner kicks, penalty kicks) in the last few years that have provided valuable information on variables that influence their effectiveness.

Investigations into match activity profiles have evolved to include the characterization of effort during congested fixture periods, in competition with extra time periods and effective use of substitutions.

Specific collective system measures (e.g., a team's geometric centre and dispersion) and analysis techniques (e.g., network, sequential and temporal-pattern analyses) have provided important information about how teammates and opponents interact during performance so that tactical behaviours can be better understood.

1. Introduction

A systematic review of the research articles that were published before 2011 provided a timely overview of the most common research topics, methodologies and evolutionary tendencies of research in Association Football [1]. More recently, studies have discussed current approaches to tactical performance analysis in football [2-4].

In the last five years there has also been some progress in terms of books dedicated entirely or partially to football [5-7], with one peer-reviewed journal, *Science and Medicine in Football* (Taylor and Francis Group) emerging as a stand-alone journal after three years of publication as regular supplement of the *Journal of Sport Sciences* (Taylor and Francis Group).

Additionally, match analysis as a methodological approach in sport science has progressively grown, based on proliferation of technological systems (e.g., GPS - global positioning system, Prozone - STATS, OPTA) to collect performance data. The interpretation of the data seeks to generate knowledge about team properties and the patterns that characterize their organization [8], with implications for coaches and sport analysts to design practice strategies and plan training sessions [9].

Progression of match analysis research in Association Football has increased exponentially since 2011, and recent literature since that date can offer new insights to the field if theoretically and systematically organized and interpreted. Systematically

reviewing research published in refereed journals contributes in several ways, such as: 1) informing researchers about the evolution of knowledge on match analysis; 2) the characterization of new techniques for gathering new information, and; 3) offering an evolving theoretical organization of the key topics and concepts researched in performance analysis in football.

The purpose of this article was to conduct a systematic review of published articles on match analysis in adult male football, identify and organize common research topics, and synthesize the emerging patterns of work between 2012 to 2016, predicated on findings from the previous review by Sarmiento et al. [1] of studies published up to 2011.

2. Methods

2.1 Search strategy: databases, inclusion criteria and process of selection

A systematic review of the available literature was conducted according to PRISMA (preferred reporting items for systematic reviews and meta-analyses) guidelines. The search strategy followed by Sarmiento et al. [1] was adopted in the current study.

The electronic database, Web of Science was searched on the 6th January 2017 for relevant articles published between 1st January 2012 and 31st December 2016, using the keywords “football” and “soccer”, each one associated with the terms: “match analysis”, “notational analysis”, “game analysis”, “tactical analysis”, and “patterns of play”.

The inclusion criteria for these articles were: (1) including relevant data concerning technical and tactical evaluation or statistical compilation, and time–motion analysis; (2) participants included amateur and/or professional adult male footballers; and (3), the

articles were published in English. Studies were excluded if they: (1) involved children or adolescents (under 18 years); (2) included females; (3) did not include relevant data for this study; and (4), were conference abstracts. If there was disagreement amongst authors regarding the inclusion of certain articles, a discussion was held until a consensus was found.

Two independent reviewers (HS, FC) independently screened citations and abstracts to identify articles potentially meeting the inclusion criteria. For those articles, full text versions were retrieved and independently screened by two reviewers to determine whether they met inclusion criteria. Disagreements about whether the inclusion criteria were met were resolved through discussion with the other authors, who analyzed the full text of the papers that induced doubts in the two main reviewers who performed the initial screening process. In two cases, direct communication with the authors of the original articles helped in the final decision making process. In this way, all final decisions resulted from a process of joint decision-making.

2.2 Extraction of data and quality of the studies

To evaluate quality of the studies, a risk-of-bias quality form was adapted for the specific context of research developed in match analysis, from the original version developed by Law et al. [10], following an evaluation process by five senior researchers with substantial experience (including relevant publications) in soccer performance analysis. Some minor suggestions were introduced in the final critical review form (16 items) according to their evaluation (see Electronic Supplementary Material Table S1). Articles were assessed based on: purpose (item 1), relevance of background literature (item 2), appropriateness of study design (item 3), sample studied (items 4 and 5), use

of informed consent procedure (item 6), outcome measures (item 7 and 8), method description (item 9), significance of results (item 10), analysis (item 11), practical importance (item 12), description of drop-outs (item 13), conclusions (item 14), practical implications (item 15), and limitations (item 16). All sixteen quality criteria were scored on a binary scale (0/1), wherein two of those criteria (items 6 and 13) presented the option: “If not applicable, assume 3”. The introduction of this option for item 6 “Was informed consent obtained?” and 13 “Were any drop-outs reported?” has been included because, in some studies, the investigators were not required to obtain informed consent (item 6), or report drop-outs (item 13). The introduction of the option “not applicable” allowed an appropriate score for the article, eliminating the negative effect of assuming the value “0” on a binary scale, when in fact that specific item was not applicable to that study. As in previous research [11, 12], to make a fair comparison between studies of different designs, the decision was taken to calculate a percentage score as a final measure of methodological quality. For this, the sum of the score of all items was divided by the number of relevant scored items for that specific research design. All articles were classified as: (1) low methodological quality - with a score $\leq 50\%$; (2) good methodological quality - between 51% and 75%, and; (3) excellent methodological quality - with a score $> 75\%$.

A data extraction sheet (adapted from Cochrane Consumers and Communication Review Group’s data extraction template – available at: <http://cccr.cochrane.org/author-resources>) was developed and tested with ten randomly-selected studies. First, one researcher extracted the data from included studies and then, a second researcher checked the extracted data. Disagreements were resolved by consensus.

3. Results

3.1 Search, selection and inclusion of publications

The initial search identified 483 titles in the described database. These data were then exported to reference manager software (EndNote X8), and any duplicates (189 references) were eliminated automatically. The remaining 294 articles were then screened according to the title and abstract for relevance, resulting in another 156 studies being eliminated from the database. The full text of the remaining 138 articles was read and another 61 were rejected due to a lack of relevance for the specific purpose of the current study. At the end of the screening procedure, 77 articles received further in depth reading and analysis for the systematic review (Figure 1).

The main reason for exclusion was that a published study did not concern match analysis ($n = 23$). Other reasons for exclusion included: (1) participants were youth players under 18 years of age ($n = 10$); (2) involvement of female players ($n = 8$); and (3), data were included from other team sports ($n = 20$), including rugby, futsal, handball, Australian Football, basketball, volleyball, field hockey, frisbee, floorball and waterpolo.

Sarmiento et al. [1] reviewed 53 articles published up to the end of 2011. Interestingly, in a much shorter period (2012-2016) the data revealed an increase ($n=77$) in the number of studies published on the selected topic.

****Insert Figure 1 here****

3.2 Quality of the studies

Sarmiento et al. [1] justified the quality of the papers included in their revision due to the use of Web of Science as the search database. In contrast, in the current review, we

evaluated the quality of the papers included for analysis. The results of the inter-observer reliability analysis, calculated by the Kappa index, was 0.97 (95% CI 0.97-0.98), indicating very good agreement between observers. The quality of indicators for the included papers was as follows: (1) the mean methodological quality score for the 77 selected articles was 89.8%; (2) twelve articles achieved the maximum score of 100%; (3) none of the articles scored below 50%; (4) two articles scored between 50% and 75% (good methodological quality); and (5), 75 articles achieved an overall rating of >75% (excellent methodological quality).

Possible deficiencies identified in the 77 studies were mainly related to two items on the criteria list: (1) for criterion 16 (see appendix S1), some studies failed to clearly acknowledge the limitations of the study; (2) some studies lacked information in relation to criterion 5, related to an explicit justification of the study sample size.

3.3 Data organization

The previous systematic review by Sarmento et al. [1] categorised research according to type of analyses performed (descriptive analysis, comparative analysis and predictive analysis), and type of variables analysed. In contrast, the present review grouped studies according to major match analysis research topics (categories) that emerged from the detailed analysis (Figure 2). This approach was adopted in order to contribute to a theoretical knowledge based on an ecological dynamics framework [7], without losing the bottom-up knowledge that emerges from the systematic analysis of studies review.

Two independent reviewers (HS, FC) independently classified the papers according to the different major research topics. Disagreements were resolved through discussion with the other co-authors until a consensus was found. The aim was not to produce categories that were mutually exclusive, since the same analysis can include topics that

relate to different categories. Thus, an article included in a specific “category”, could also be classified in another “category” whenever its content justified it.

****Insert Figure 2 here****

3.4 Major research topics

The following subsections describe the studies identified in each of the specific research topics. Their findings are outlined and discussed in more detail in section 4.

3.4.1 Set-plays

Set-plays like free kicks, penalty kicks, corners and throw-ins can provide match winning situations. There has been an increase in research examining these match events in the last few years [13] in different competitions (Table 1). Several studies have estimated that between 30% and 40% of goals are scored from set plays [14]. The importance of this type of situation is highlighted by professional coaches who identified an increased systematization of specific set-play training situations as an evolutionary trend of training/competition [15]. Given its importance to the match, dead ball specialists work on the training ground to perfect their techniques. However, it is important that defenders also be prepared to face set plays.

****Insert Table 1 here****

3.4.2 Activity profile

3.4.2.1 Playing roles

Modern professional soccer imposes more and more demanding requirements on players that relate to their pre-competition preparation [16], and specifically to their roles on field (i.e., goalkeepers, defenders, midfielders and forwards). There is a need to

understand how roles performed in soccer can affect performance, as estimated by specific variables (Table 2).

****Insert Table 2 here****

3.4.2.2 Fatigue influence

Match activity and fatigue during football matches has been a topic of increased research in recent years (Table 3). Recent research showed that physical performance during the match changes throughout the season and is related to players' training status [17]. Previous research has also revealed the evolution of the game over the last few decades, especially the increasing intensity of play [18] that is directly related to physical performance decrements from the first to the second half of elite soccer match play [19]. These trends could result in an inability of players in certain roles to repeatedly cover distances during critical situations and may also reduce technical capabilities that are related to match outcomes [20].

****Insert Table 3 here****

3.4.2.3 Substitutions

Substitutions have enormous impact and importance in modern football contexts, because coaches typically attempt to use well-timed substitutions to reduce the effects of fatigue across the team or in an effort to modify tactics. Consequently, some important research (Table 4) has analyzed this specific aspect of the match in the English Premier League [20], Spanish Professional Soccer League [21] and UEFA Champions League [22].

****Insert Table 4 here****

However, little scientific evidence is available and has only recently been introduced in the scientific community [20, 22, 21].

3.4.2.4 Altitude and environmental heat stress influence on performance

With the aim to examine the effect of altitude and environmental heat stress on football performance, some interesting studies (Table 5) have been developed with the national teams that participated in the FIFA 2010 [23] and 2014 World Cups [24].

****Insert Table 5 here****

3.4.3 Variables capturing group behaviors

3.4.3.1 Team centre

The team centre represents the geometric centre of a set of points that represent the current positioning of soccer players on field during competition [25]. This measure has been used to analyze the collective spatial-temporal synchronization between competing teams and to identify instabilities or transitions that might lead to the emergence of critical moments in a competitive match, such as an opportunity to score a goal [26, 27]. A summary of the studies that have examined dynamics of the collective variable 'team centroid' (geometrical centre of the players excluding goalkeeper and not considering the position of the ball) can be found in Table 6.

****Insert Table 6 here****

3.4.3.2 Team dispersion

The quantification of how far players are apart (dispersion) may help in understanding the nature of the space that emerges from the interactive dynamics of competitive matches, helping us to identify the natural expansion and contraction of soccer teams in attacking and defensive moments. Dispersion of the team can be estimated by

calculating the area covered by all points of the team (surface area), identifying the dispersion of the players from the geometrical centre (stretch index), calculating the Euclidian distance between each player and his teammates (a team's spread) or analyzing the effective defensive triangulations (effective area of play and defensive play area) [28, 25, 29]. These measures have been used to observe the oscillations of the areas of the team during attacking and defensive moments and to compare their variability throughout competitive matches (Table 7).

****Insert Table 7 here****

3.4.3.3 Team interaction/coordination networks

The communication process may occur in different ways, but in match analysis these have been used to classify the interactions between teammates when in possession of the ball, during passing sequences [30]. Social network analysis based on graph and digraph theories has been used to classify general properties of the network and specific centrality levels of players (nodes) [31, 32]. Identifying networks on field may help in understanding the specific relationships that emerge between teammates during attacking sequences and the general properties of collective team performance during passing sequences. A summary of the studies that analyzed the networks of soccer players can be found in Table 8.

****Insert Table 8 here****

3.4.3.4 Sequential patterns

Sequential patterns analyze sequential combinations of interactions that emerge between players during a match or a set of matches [33]. Some studies (Table 9) have analyzed the attacking patterns of teams in different competitions, considering criteria such as duration of an attacking sequence of play, number and role of the players involved in the attack, zone of pitch where the actions were performed, type of technical behaviors

and the number and co-location of players of both teams (interaction context) in the space adjacent to the ball on field [34, 35].

****Insert Table 9 here****

3.4.3.5 Group outcomes

Winning, drawing and losing may be constrained by or constrain some match-related statistics [36-38]. Based on these assumptions, some studies (Table 10) have been conducted to identify the variance that might exist between some playing actions and the final outcomes of a match.

****Insert Table 10 here****

4. Discussion

The aim of this paper was to systematically review the evolving patterns of match analysis in Association Football to organize research studies, published between 2012-2016, in a theoretically coherent way. After in-depth analysis, it was decided that the most appropriate way to discuss the results would be to categorize research topics according to similar themes (set-plays, group behavior and activity profile).

4.1 Set-Plays

4.1.1 Corner kicks

The reviewed studies which analyzed corner kicks were mainly focused on international competitions [39, 14] and the English Premier League [40, 41]. Corner kick effectiveness values of 2.6% [39], 2.2% [14], 4.1% [40] and 2.7% [41] were found, which means that, on average, between 24 to 45 corner kicks were needed to lead to a single goal scored. Casal et al. [14] reported that the likelihood of a shot on goal could be increased with the involvement of 3 or 4 attackers, a dynamic attacking move, and delivery of the ball to the far post. Pulling et al. [40] analyzed the importance of

defensive strategy and concluded that the one-to-one marking set-up did not concede any attempts at goal from 95.7% of corner kicks, whereas a zonal marking system did not concede goal attempts from 97.7% of corner kicks. In addition, the percentage of corner kicks resulting in a goal or attempt at goal was higher when the defending team used a one-to-one marking system (31.3%), compared to a zonal marking system (30.2%). The investigators highlighted that, although this finding may suggest that a zonal marking set-up is better for defending corners, the percentage difference between these systems is very small. Variables like the area (e.g., zones inside the penalty area) where the ball was delivered, the type of delivery (long and short corner kicks), and the influence of some situational variables (e.g., teams performed more short corners and took more short kicks and outswinging corner kicks when winning, but outstep and inswinging corner kicks when drawing and losing)[39] in the strategies used to perform the corner kick were also analyzed.

Data from the reviewed studies suggest that coaches should design training sessions that simulate the execution of more elaborate corner kicks which involve a short initial kick, followed by a dynamic interaction involving three or four players, before the ball is crossed to the far post. Concerning defensive strategies at corner kicks, there seem to be few differences in the effects of using zonal marking versus one-to-one marking systems. Additionally, coaches should be aware that the positioning of players on the goalpost(s), when defending corner kicks, does not significantly prevent goals from being scored. Rather this tactic actually increases the frequency of attempts on goal by the opposition. These findings suggest that the players positioned at goalposts could be “used” to carry out other defensive functions.

4.1.2 Penalty kicks

The penalty kick is a peculiar event involving a direct confrontation between two opponents directly functioning in a dyadic system: the penalty taker and goalkeeper [13]. It is one of the most pressured and intense moments in a competitive match. In male professional football approximately 70% of penalty kicks are scored [42].

The studies reviewed [13, 42-44] could help penalty takers and coaches improve their chances of successful outcomes as they provide information suggesting that: (1) the areas of the goal to which the ball is aimed is significantly important for penalty effectiveness [13, 42]; (2) saves depend mainly on the goalkeeper's reaction time but also on the ball speed in the penalty kick [42]; (3) situational factors (e.g., period in the match) may influence the success of penalty kicks [13]; (4) goalkeepers should wait longer in order to dive to the side of the goal to which the ball has been kicked [13]; and (5), penalty takers should use both a keeper-independent strategy and keeper-dependent strategy in order to increase their chances of success [43].

4.1.3 Free kicks

Despite the importance of set-piece goals in modern football, free kicks have not been extensively studied [45, 46]. The study by Link et al. [46] revealed an average of 34.9 ± 7.6 free kicks per match, while Casal et al. [45] concluded that on average each team takes three indirect free kicks aimed at scoring a goal per match. Of these, 21.8% ended in a shot, 9.3% ended in a shot between the posts and 2.9% ended in a goal. The type of attack and the number of players involved in the process has a direct influence on the outcome. Furthermore, Link et al. [46] analyzed variables like position (2D-location of the free kick on field) and zone (free kick location in the attacking third on the field of play (35m from goal), according to a specific categorization by the authors of the playing area. This included: density (number of free kicks in each 1m^2 sector on

field), interruption time (timespan between the foul that led to the free kick and the moment of ball contact when taking the free kick), distance to defensive wall (shortest distance between ball and defensive wall at the moment of ball contact), number of players participating in the wall, rule violation, type of play (shot on goal, cross, pass), and outcome shots (goal scored, header, save made by the goalkeeper, etc.). However, studies of this specific event remain rare and more research is needed to better understand the influence of different variables in the effectiveness of the free kick'.

Nevertheless, the reviewed scientific evidence suggests that coaches could design specific training sessions that aim to improve the effectiveness of free kicks. They could facilitate players working on elaborate kicks with the ball being played along the ground and involving interactions between three or four players (trying to reach the opposing team's penalty area using short passes and dribbles).

4.2 Activity profile

4.2.1 Playing roles

The relationship between a player's positional role and performance continues to be frequently studied [47-49, 16, 50-52]. However, Sarmento et al. [1] concluded in their analysis that previous investigations had grouped players according to different criteria, which made it difficult to accurately compare results regarding player roles.

In line with previous research findings (see Sarmento et al. [1], for a review), the results confirmed that midfielders covered the greatest average distance, followed by attackers, and then defenders. However, Clemente et al. [48] proposed an alternative way of conducting this analysis (distance that each player covered per minute) involving distance covered in possession of the ball and distance without the ball. This revision helped in understanding the running pace of the players (m/min) during the match.

Moreover, using such an approach, it was possible to compare the players who played less frequently (minutes on field) with those who played more. This relative measure based on time allowed us to make comparisons between all players in the competition, independent of their playing time. The results showed that the greatest distances, in possession of the ball, were achieved by midfielders, followed by forwards. Significant differences were also observed between defenders who played wider (more distance covered in possession of the ball) and central defenders. Without the ball, midfielders covered the greatest distances during play.

Additionally, Andrzejewski et al. [50] found that the mean total sprint distance covered by professional soccer players in the UEFA Cup ($\geq 24 \text{ km} \cdot \text{h}^{-1}$) amounted to $237 \pm 123 \text{ m}$. The sprint distance covered by the forwards was the highest ($345 \pm 129 \text{ m}$), 9% greater than midfielders ($313 \pm 119 \text{ m}$), and over 100% greater than the same value for central midfielders ($167 \pm 87 \text{ m}$). Elite footballers performed an average number of 11.2 ± 5.3 sprints per match, of which 90% were shorter than 5s duration and only 10% were longer than 5s. The results also revealed that forwards and wide playing midfielders performed a far greater number of short duration sprints, compared to central midfielders and central defenders. Also, wide playing defenders (e.g., wing backs) performed the highest number of long duration sprints, differing significantly from central midfielders, who performed fewest sprints.

A common mistake made by coaches preparing players for performance is application of the same workload to all players during training sessions. The reviewed studies identified specific physical load profiles for football players during a match, dependent on their specific playing positions, which can be used to design highly individualized training programmes for specific players.

4.2.2 Fatigue influence

Interest in fatigue in football has mainly focused on the impact of a congested fixture list [53, 54], extra time periods [55, 56] on player performance and variations in performance across a whole season [17]. Additionally, some researchers [57] investigated the fatigue rates and pacing strategies of players during matches by quantifying high-intensity running in rolling 5-min periods.

Studies of extra time clearly showed a greater decrement in physical performance markers during this period [55, 56]. Penas et al. [55] found that performance decrements affected players in all roles to a similar degree. Additionally, all of the physical markers under study showed a decline of 15-20% during the extra-time period in comparison to the first half performance, and an increase in low intensity activities in the second half. Russell et al. [56] verified that between 105 to 120 mins, acceleration and deceleration parameters reduced by >10% compared to the opening 15 mins.

Dellal et al. [54] examined three different congested fixture periods (6 matches in 21 days) and concluded that physical activities and technical performances were unaffected during these periods. Nevertheless, injury rate during match-play was significantly higher during congested periods in the fixture list. This difference between training and match-play can be explained by the low-intensity training promoted by coaches, the greater emphasis on recovery training programmes in the modern game, as well as players regulating their activity. The results reported by Soroka and Lago-Penas [53] are in line with data reported by Dellal et al. [54], though their study focused on a smaller congested period (3 consecutive matches separated by 4 days).

It is noteworthy that the study by Dellal et al. [54] surpassed the limitations of previous research that investigated sporadic congested fixture schedules and only analysed physical performance across two or three consecutive matches within a short time-

frame. They concluded that the overall distance covered was greater in the third period (October–November) of the season, whereas no differences were observed in the other speed thresholds (first period – August; second period – September). No studies have examined whether physical and technical activities decreased or varied according to stages of the season. However, a study by Silva et al. [17] examined match activity and the development of fatigue during competitive soccer matches in different periods across a whole season. They reported an association between muscle strength and power, and performance decrements in match-related physical parameters. Their results highlighted the importance of incorporating specific exercise programmes to improve the athletes' strength and power during performance.

The reviewed studies highlighted different fatigue-related mechanisms related to physical performance decrements throughout the duration of a normal match (i.e. 90+ mins), the duration of extra time (i.e. 120 min) in match play, or during a congested fixture period (e.g., when 3 games may be played in a week). The available knowledge seems to be useful for technical and medical staff in their implementation of specific strategies to minimize performance decrements during a match or a congested fixture period. Such strategies could include: 1) specific exercise programs to improve the athletes' aerobic capacity during the performance of soccer-specific activities; 2) nutritional supplementation protocols; 3) low intensity activities during training sessions and adequate rotation of players in congested fixture periods; 4) use of objective markers of fatigue combined with subjective measures of performance; and 5), adaptation of tactical strategies.

4.2.3 Substitutions

As a limited resource for tactical interventions, substitutions are assumed to be

important in football, although little scientific evidence is available on this issue [20, 22, 21]. Gomez et al. [21] concluded that most of the first and second substitutions occurred during the final third of the match (between 61-90 mins), while the third substitution occurred predominantly during the final quarter of the match (76-90 mins), in the Spanish first division. Conversely, in the English Premier League and UEFA Champions League a large number of substitutions occurred at halftime and between the 60-85 mins [20] and 57-78-mins [22], respectively.

The most substituted position is the central midfielder, followed by forwards, wide midfielders, full backs, and central defenders. Additionally, 'like-for-like' substitutions (the same playing position for player in and player out) were the most common and the defensive and offensive substitutions showed similar distributions [21]. Substitutions became more attack-minded as the 2nd half progressed [20].

Concerning match performance characteristics, Bradley et al. [20] found that the same players displayed more high-intensity running when they were introduced as substitutes compared with the equivalent period of the second-half, but not the first half-period when tracked from the start of the match. The distances covered by high-intensity running were greater for attacking substitutes. These results were interpreted according to perceived tactical options and specific physical demands of playing roles.

Effects of situational variables on timing and tactics of substitutions have also been analyzed [22, 21]. Rey et al. [22] presented a decision tree analysis that could be used to inform UEFA Champions League coaches by using the following heuristics: if losing a match, make the 1st substitution prior to the 53th minute, make the 2nd substitution prior to the 71st minute and make the 3rd substitution prior to the 80th minute; if the scores are tied or the team is ahead, make substitutions at will.

The results of these investigations can provide valuable information so that coaches

could optimize player and team performance, but more work needs to be undertaken to investigate the overall impact of substitutes on physical and technical indicators, and their contribution to key moments in matches [20, 22, 21].

4.2.4 Altitude and environmental heat stress influence on performance

Based on the assumption that exposure to altitude and environmental heat stress has a detrimental impact on exercise performance, some researchers have investigated their effects on football players [23, 24]. Nassis [23] conducted a study that examined effects of altitude on soccer performance during the 2010 World Cup in South Africa. The study verified a 3.1% decrease in the total distance covered by teams during matches played at 1200-1400m and 1401-1753m, compared with sea level. Through the analysis of environmental heat stress in the 2014 FIFA World Cup (Brazil), Nassis et al. [24] concluded that there was no difference in playing time (average of both teams), total distance traveled (m/min/player), number of goals scored and number of cards issued by referees, compared to matches played under different environmental stress categories (e.g., the risk of heat stress at 50% relative humidity is 'high' for wet-bulb globe temperature (WBGT) 28–33°C, 'moderate' for WBGT 24–28°C and 'low' for WBGT <24°C.). High intensity activity was lower under high compared to low environmental stress, and the rate of successfully completed passes was greater in the former compared to the latter.

4.3 Group behavior

4.3.1 Team centre

Different terminologies (centroid – geometrical centre of the players excluding goalkeeper and not considering the position of the ball; wcentroid – geometrical centre

of the team that attribute weight to the teammates based on their proximity to the ball; team centre - geometrical centre of the players excluding goalkeeper and not considering the position of the ball) have been used to describe the team centre which has been defined as the geometric centre of a team, considering the positioning of all players on the pitch [58, 27, 25]. The present review identified three main approaches for centroid analysis in soccer: 1) centroid of the team calculating the geometric centre without goalkeepers [27, 59]; 2) weighted centroid, considering the proximity of each player to the ball as the weight to move the centroid [25]; and 3), the centroid considering the middle point between the two teammates furthest apart [60]. Team centre has been used to assess intra- and inter-team coordination in soccer in a temporal sequence [27].

In most cases, an in-phase relationship between the competing teams' centre values during competitive performance has been investigated [26, 27, 61]. The study of elite European soccer suggested that team centroids moved synchronously both up and across the pitch [26]. Similar results in the variable relative phase were found in the final of the 2006 FIFA World Championship [62]. However, in the case of small-sided games (5 vs. 5), specific moments during performance may constrain the synchrony between teams, which can quickly turn into non-synchronization, and even a crossing of team centres that may relate to specific events in the match [63].

The idea of critical moments in competitive performance (e.g., goals, shots) was examined in competitive matches [26, 27]. The study of elite European soccer teams during official competitive 11-a-side matches [26] contradicted the evidence reported from observations of performance in small-sided games that some goals occurred at moments of non-synchronization or of crossing between centroids [63]. In a different approach, the investigators of the study of performance in small-sided games

investigated the hypothesis that inter-team variability would indicate critical moments in a competitive match [27]. However, the results from that study suggested that inter-team distances (differences between teams' centroids) were minimally related to emergence of critical moments in a match [27]. One possible explanation for this finding is that, during small-sided games, it is easier to remain close to other players and for the geometric midpoints of both teams to overlap.

The relationship between competing teams' centroids may provide information about the synchronization of the teams and identify when non-synchronization of team centroids may lead to critical events in a match. The distance between teams may be also used to design small- or medium-sided games that better simulate specific sub-phases of the game (e.g., direct attacking, indirect attacking) based on the dimensions of the pitch.

4.3.2 Team dispersion

The dispersion of a team can be defined by quantification of distances between teammates. The dispersion of the players on the pitch can be constrained by specific strategies and tactics that emerge during the match [25]. Regularly, players are more disposed to explore width and length of the pitch when attacking to exploit free space and to destabilise the opposition defense [64]. Conversely, distances between teammates tend to be smaller when defending to optimize cover and to contract space [28]. To examine some of these suggestions, some measures have been used, including: (1) stretch index; (2) weighted stretch index; (3) Frobenius norm (team's spread); (4) surface area; (5) effective area of play; (6) playing area; (7) team length and width; and (8) defensive play area and triangulations.

Stretch index can be described as the mean dispersion of the players from the team centre (non-weighted) [58, 59]. This measure can be quantified as the radius or only by the width or length axes. A similar concept was introduced by changing the weight of the centroid [8]. In a study of high-level European soccer teams smaller stretch index values were reported for defending teams, compared to attacking teams [26]. Nevertheless, the data did not consistently associate stretch index values with goals scored [26]. Using the same measure in an analysis of a competitive match, the evidence suggested that the variability of approximate entropy values decreased progressively during the match, with the exception of the transition from the last 15 minutes of the 1st half to the first 15 minutes of the 2nd half [59]. It was also reported that the stretch index was greater in home teams in most of the match time observed [59].

The weighted stretch index was also used to analyze performance variance between halves of the match [25]. It was found that values of dispersion were smaller in the 2nd half. In addition, the same group of researchers reported a greater weighted stretch index during drawn matches and no statistical differences between losing and winning situations [65].

The Frobenius norm was used as a measure of a team's spread in four studies [28, 26, 66, 67]. In a study conducted on Brazilian teams it was suggested that greater values of spread when defending were associated with emergence of critical moments such as shots on goal conceded. When attacking, this variable was greater when a team was closer to the opponent's goal [28]. However, the evidence reported on spread in attacking phases of play was not confirmed in a study of elite European teams [26]. More recently, a predominant in-phase relationship (linear association of both teams'

spread over time) was observed between the spread of teams and anti-phase periods at critical moments of play, such as when shots at goal emerged or goals were scored [67]. Surface area has been used as an alternative dispersion measure which uses the convex hull to determine the polygon generated by all the players [63]. In most cases it was found that surface area values were greater when attacking than when defending [68, 25]. It was also found that surface area was greater when a team competed against weaker teams [68] and when the scores were level in a match [65]. In a study conducted on a single match there was a progressive tendency of this measure to reduce in variability during the match [59].

The effective area of play was introduced as an alternative measure to the surface area providing a notion of effectiveness in the defensive triangulation generated when defenders covered each other [25]. Using this measure a significantly smaller space was covered in defence compared with attack, and there was a tendency for the effective area of play to be greater when the scores were level compared to when a team was losing or winning in matches [65].

In an alternative approach, playing area was introduced to verify the mean area covered per player and to identify the best value to use in designing small-sided games [69]. Individual playing area varied between 81.38 m² and 86.78 m² in observed competitive matches with the variability in these values influenced by the location of the ball in specific zones of the pitch. An individual playing area of 90 m² in small-sided games was suggested to develop build-up play or attacking play in the finishing phase, whereas 80 m² for small-sided games developed transition play [69].

Defensive play area was introduced as a tactical measure that determines the area covered by a group of players when defending [70]. It was found that the triangular positioning relationships generated when midfielders covered each other on field were

greater and for that reason the defensive playing area in the midfield area was significantly greater than upfront or at the back [70].

Dispersion measures can be directly used to evaluate the space required for defensive and attacking processes and to adjust the playing dimensions and format for small- or medium-sided games designed in training. The right measure of dispersion can be also used to identify width and length of attacking and defensive phases of play in teams. Such information can be used to characterize team performance and to help coaches plan better strategies to exploit opposition weaknesses or to reinforce playing patterns in their own team.

4.3.3 Team interaction/coordination networks

The network process can quantify the centrality level of a player (individual values per player), dependence between players (meso-level of analysis) and the general properties of a graph (that quantify a value of a specific network property of a team). General network properties have been studied in association with team performance variables such as shots, goals and successful outcomes in competition [30, 71, 72]. High passing rates were related to an increase in team performance and greater centralization was associated with a decrease, defined by the number of goals scored in an analysis of 760 matches from the English Premier League [30]. It has been reported that winning teams display statistically greater levels of general network measures, such as density, homogeneity or number of total links, with small-to-moderate associations with goals scored, overall shots taken and shots on goal in 64 matches from the 2014 FIFA World Cup [71].

Differences between centrality levels of players (individual level of analysis) have also been investigated using the social network approach [73, 31]. Similar evidence was

found in a specific analysis conducted on the Spanish national team during the same international competition [31]. In a study conducted on one team using the centroid measure, it was reported that the left back tended to be the dominant player during attacking sequences [74].

Variance of centrality measures between playing positions has also been analysed [32, 72]. Midfielders were classified as the most prominent players after observations of 64 official matches from the 2014 FIFA World Cup, independent of the specific team and tactical format used [32]. The specific analysis conducted on the German team revealed that midfielders had greater levels of intermediation (capacity of a player to link two or more teammates to each other during play on field) and dominance (capacity to be the player who most often participates in team networks) [72].

Full passing sequences have been analyzed in most cases, however, in an alternative study only the network interactions in passing sequences that resulted in scored or conceded goals were analysed [75]. The results suggested that attacking midfielders and wing forwards were the most prominent players for receiving the ball and the right back was the dominant player for passing. Analysis of pitch zones revealed that central and wing regions closer to the goal being attacked were mainly influential in network interactions during attacking phases of play that led to goals being scored [75].

Identification of prominent players engaged in specific types of playing interactions may be used by coaches to adjust performance strategies. In understanding defensive behaviours, a coach can identify the most prominent opponent and which players are well linked to him. Based on the information gained, the coach may design a strategy to mark a key player or to prevent teammates from passing the ball to him/her. Moreover, knowledge of interactions between teammates can be used to identify how they cooperate and can be used to classify the main networks in the team.

4.3.4 Sequential patterns

A temporal pattern can be described as a repeated temporal and sequential combination of the same order of events during a period, which are relatively invariant [33]. This kind of analysis, when focused on the sequence (and temporality) of events, supports the detection of patterns of play which have higher probabilities of occurrence than chance. The basis for any prediction model is that performance is repeatable, to some degree, suggesting that events that have previously occurred will occur again in some predictable manner.

Temporal analysis of attacking play has been adopted in many studies [34, 35, 76]. Performance criteria analysed have included the lateral position (the pitch is split into three longitudinal areas: right, centre and left), zone (ultra-defensive, defensive, central, offensive and ultra-offensive), possession (ball in play), interaction contexts (specific regions of the pitch in which interactions between players emerge), recovery and loss of ball possession and time that the ball is out of play [34, 35].

In a study of performances of FC Barcelona in the Spanish national league and in the UEFA Champions League, match to match and half to half patterns were reported [34]. Sixty-eight patterns were recognized during 10 matches observed. One of the identified patterns was an attacking structure that begins in the central defensive zone and progresses to the wings before entering the offensive zone. Another pattern was an attack that begins in the central defensive zone, progresses to the left side, moves back to the centre line and attacks again from the same side of the field [34]. A comparison of attacking patterns of play in a top Italian league team, when winning and losing matches, was analyzed [35]. One hundred and sixty-seven patterns emerged in 80% of the 19 matches studied. A greater volume of temporal patterns ($n = 101$) emerged when

the team was losing matches, compared to when winning ($n = 9$). It seemed that, in winning matches, the team was more likely to continue using the same playing pattern [35]. The variance of temporal patterns between halves was investigated, again in an elite Italian team [76]. A greater number of temporal patterns were found in the 2nd, compared to 1st halves. More patterns were found [59] and the length and level of passing sequences and their patterning were greater. Five playing patterns were observed in the 1st half, and nine in the 2nd. Temporal patterns also revealed at least one shot for each pattern (in the 2nd half), whereas in the 1st half, no such evidence was observed [76].

The study by Sarmento et al. [33], using sequential analysis, is the only study in this updated review that involved the expert opinions of professional coaches to interpret the data. This close relationship between researchers and professionals can be very fruitful for interpreting data in match analysis.

Identification of temporal patterns of play may provide information about structural behaviours which are independent of opposition play. Such patterns can be used by coaches and analysts to identify strategies to negate opposition strengths or to verify congruence between performance behaviours worked on in training and their execution in competitive games.

4.3.5 Group outcomes

Relationships between match-related statistics and match final scores have been analysed by several investigators [36-38]. A study of 177 matches from the FIFA World Cups in 2002, 2006 and 2010 found that total number of shots and shots on target were the main discriminatory variables to predict winning, losing and drawing matches [36]. Shots were also confirmed as the main discriminating variable for winning teams [77].

Moreover, a study conducted on 1900 Spanish league matches revealed that match winning teams displayed a greater number of ball recoveries and tended to perform longer passing sequences [78]. The patterns of ball recovery are also important to discriminate match winning teams from those who lost and drew matches [77].

An interesting observation from a study of 12 Spanish league matches revealed that match outcome influenced match-related statistics [79]. Teams who drew and won matches showed a decrease in the probability of reaching the penalty area in possession of the ball, in comparison to when they were losing a match [79]. In terms of defensive playing patterns, it was found that losing teams tend to defend in more advanced pitch zones. However, more successful teams tend to be more efficient in defensive pressure and ball recovery [80].

Identification of specific key indicators or use of modeling methods may provide information for coaches to re-prioritize playing styles and to also help re-design training exercises to include indicators of successful performance.

4.4 - Contextualizing performance

Contextualizing performance has been a concern of researchers in this field of study. Although it is possible to categorise studies according to “major topics” of research, in reviewed studies investigators analysed different variables (e.g., work rate, technical behaviours, ball possession) according contextual variables, including: (1) match half [25, 57, 81, 82]; (2) quality of opposition [68]; (3) match location [83, 21, 38, 22, 80]; (4) scoring first [38, 83]; (5) group stage vs. knockout phase [83, 22]; (6), intervals of 5 [57] or 15 minutes used to record data [82, 38]; (7) timing and tactical nature of substitutions [22, 21]; (8) competitive level [84, 85, 51], and (9), different competitions (different leagues and cups) [86].

Interpretation of player behaviours and match outcomes in specific contexts may help identify specific strategies or training designs that coaches could incorporate to prepare the team for different opposition strategies, circumstances and game scenarios.

4.5 Limitations

A possible limitation of this systematic review is that it only includes studies written in English from the Web of Science, thereby potentially overlooking other relevant publications in other languages.

5. Conclusion

Research on match analysis in adult male football players has been the subject of growing interest in the past five years. Nonetheless, some limitations remain in the published studies between 2012-2017, namely the lack of operational definitions and conflicting classifications of activity or playing positions that make it difficult to compare similar groups of studies. Additionally, some potential weaknesses may be apparent in more recent published research, such as the small sample sizes used in some studies. Nevertheless, researchers have developed new methods in order to better contextualize the performance of players and teams, which is likely to be essential for planning and application of training loads in modern professional soccer.

A progressive increase in group analysis based on positional data is one of the main new insights in comparison with the previous systematic review by Sarmiento et al. [1]. Positional data can be used to identify patterns of interaction between teammates and to explore the spatiotemporal patterns that emerge from a match [87, 9]. Team synchrony has been analyzed based on an in-phase relationship between teams using the measures of centre and dispersion, suggesting regularities in the dynamics of competing teams

and some disturbances that emerge at specific critical points of the matches (e.g., goals, shots, counter-attacks) [26, 66]. The new collective measures reviewed in this article could help identify the need for specific training conditions for the collective organization of a team, improving the efficacy of practice task design to augment the cognitions and perception of players regarding specific tactical behaviors [69]. Analysis of interactions between teammates may reveal collective properties that cannot be captured by players' individual movements. For example, synergetic properties of the team can offer theoretical guidance to capture system properties such as dimensional compression, patterns of interpersonal linkage, reciprocal compensation or degeneracy [88]. In summary, the collective measures identified in this review provide different information than that typically gained from traditional notational analyses. Knowledge about the spatiotemporal relationships formed by players during competitive performance may explain some behaviours that notational analysis cannot quantify. The exact positioning of players, occupied space and values of inter-personal distance can be more easily and objectively measured by collective measures, using tracking or GPS systems. Moreover, t-patterns (defined as a particular set of event-types recurring in the same order with significantly similar distance values between them) and network measures may be used to classify and rank players based on their importance in a competitive game, identifying specific interactions. Both of these methodologies can complement use of notational analysis.

These novel measures require new measurement techniques, and the complexity engendered during soccer matches requires an integrated approach that considers multiple aspects of performance [89]. A big challenge for researchers is to align these new measures with the needs of the coaching staff, through a more interactive relationship between all practitioners, to produce practically relevant information that

can improve performance through constant adaptations of training design. Reductionist methods and approaches should be avoided and multifactorial analyses must be conducted, integrating notational methods and computational collective measures to amplify knowledge and identify long-term patterns in performance dynamics during competition. The association between outcomes (notational analysis) and processes (spatiotemporal analyses) may also contribute to identify which patterns can be avoided or reinforced to increase possibilities for success. Additionally, future studies should promote real-world insights into optimal methodologies for player preparation through integrating sources of information about training requirements, periodization load, structure of competition, as well as player fitness and fatigue. Collecting and measuring a large volume of data (e.g., positional, physiological, psychological, environmental conditions, etc.) in real time, and compressing it into a smaller set of variables, providing objective information for coaches that facilitates, to some extent, the prediction of performance outcomes, seems to be a useful path in this specific area. An augmented perception analysis framework for football (ARCANE) [90] represents an interesting first step that may explain how to achieve this significant goal.

Compliance with Ethical Standards

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Table 1. Studies with predominantly corner kicks, penalty kicks and free kicks analysis.

Study	Competition	Number of events analyzed	Results/main findings	Practical applications	Quality score (%)
Corner kicks					
De Baranda and Lopez-Riquelme [39]	2006 FIFA World Cup.	653 corner kicks.	Teams perform: 1) more short corners and take more short kicks and outswing corner kicks when winning; 2) more outstep and inswing corner kicks when drawing and losing; 3) more shots that head toward the semicircle placed by the penalty area or do not use the centre when winning. However, teams head their shots toward the first and the second goalpost when drawing and losing. When teams that are winning perform the corner kick, the defending teams tend to have fewer players defending the goal line.	Allow coaches to design training exercises similar to the actual competition and adapt the game style with regard to match status.	100
Pulling et al. [40]	English Premier League.	436 corner kicks.	The most commonly used marking system was one-to-one marking, with zonal marking being used less often. There was no significant association between the marking set-up and the number of attempts at goal conceded when defending corner kicks. Teams who used zonal marking conceded fewer goals and fewer attempts at goal than teams who applied one-to-one marking. The most common set-up for defenders positioned at the goalposts was having a defender positioned only on the far post.	There are no differences in the effectiveness of goal scoring or goal attempts as a function of different defensive strategies (zonal marking vs one-to-one). There is no advantage in positioning players on the goalpost when defending corner kicks.	92.9
Casal et al. [14]	2010 FIFA World Cup, UEFA Euro 2012 and the UEFA champions league 2010-2011.	1139 corner kicks.	Just 2.2% of the corners ended in goal, but this goal was responsible for the team winning or drawing the match on 76% of occasions. In general, kicks are delivered through the air to the near post, with 1 or 2 intervening attackers. The attack is organized statically and the defense is a combination of zone and man-to-man. The following variables were significantly associated with corner kicks resulting in a goal: time, number of intervening attackers	More elaborate corner kicks (sent to the far post, following a short initial kick and the intervention of three or four players in a dynamic set-up) seems to be more effective.	92.9

Pulling [41]	English Premier League.	328 corner kicks.	and offensive organization. There was an association between the area where the ball was delivered to and the number of attempts at goal and the area the ball was delivered to and the number of defending outcomes. The area where a long corner kick is delivered to will influence how many attempts at goal can be achieved by the attacking team and how many defensive outcomes can be conducted by the defensive team.	The area of delivery is more important than the type of delivery for achieving attempts at goal from long corner kicks.	85.7
Penalty kicks					
Farina et al. [42]	World Cup (2010), Copa América (2011) and UEFA Euro (2012). Controlled situation in Club Atletico Peñarol (2010).	There is no reference of the number of penalties analyzed from 2010 World Cup, 2011 Copa América and UEFA Euro 2012. 50 penalty kicks 52425 simulations were analysed.	Saves were found to depend essentially on the beginning of the goalkeeper's motion according to the area. In the central areas of the goal, saves were statistically independent of ball speed and time of reaction. The goal region further away from the centre, where real shots are frequently saved, presented highest dispersion.	Penalty takers should be encouraged to direct the shot to areas near to the posts. Perceptual training related to the skill of anticipating the side (and precise area) of the goal the shot will be directed at may be essential for enhancing goalkeeper performance.	57.1
Noel et al. [43]	Controlled situation in a German club. FIFA World Cups (1986–2010) and UEFA European Football Championships (1984–2012).	84 penalty kicks. 322 penalty kicks.	The keeper-independent strategy was used much more frequently than the keeper-dependent strategy, but successes did not differ. Penalty takers should use both the strategies to be less predictable.	A goalkeeper could try to identify a penalty kick strategy by focusing on the fluency of the early parts of the run-up and the kicker's gaze. A penalty taker who tends to slow down, uses shorter strides and looks frequently at the goalkeeper is likely to use a keeper-dependent strategy. It is advisable for the goalkeeper to wait longer before starting to dive. If a penalty taker runs up steadily, while largely ignoring	85.7

Almeida et al. [13]	UEFA Champions and Europa leagues (2010-2011 to 2014-2015).	536 penalty kicks.	The probabilities of penalty kicks being saved significantly increased: i) in the middle of matches (30: 01-60: 00), and; ii) when the shots were directed to lower zones of the goal, in particular to the lower centre-left zone of the goal (penalty taker's perspective). The odds of missing the penalty substantially increased when the shot aimed the high zones of the goal.	the goalkeeper, a keeper-independent strategy is more likely. The goalkeeper is then advised to dive early to the kicker's natural side. Penalty takers should be encouraged to direct the shot to the upper corners of the goal and goalkeepers should wait longer in order to dive to the correct side of the ball.	100
Free kicks					
Casal et al. [45]	FIFA World Cup (2010), UEFA Champions League (2010) and UEFA European Championships (2010).	783 free kicks.	Almost 36 indirect free kicks are needed to score a goal, but that 64% of goals from indirect free kicks have a decisive influence on match outcome. Goals were more common when the attack was organized dynamically and three or four players touched the ball before a shot was taken.	Coaches should promote training session with elaborate kicks (more effective): the ball is played along the ground, touched by three or four players, and form part of a dynamically organized attack.	92.9
Link et al. [46]	German Bundesliga (2013/2014, 2014/2015).	1624 free kicks.	Centrality and proximity to the goal increased the variables players in wall, rule violations and interruption time, and the ratio of goals scored increased from 5.9% (central far) to 10.9% (central near). There was no statistical advantage for the defensive team when distance to wall was below 9.15 m or when there was a rule violation. Crosses had a success rate of 20.8%. Crosses from the right side outside the penalty box were 10% more successful than from the left side.	It might be more effective to increase the proportion of passes from side free kicks and to try to reach the opposing team's penalty area using short passes and dribbles. Since right-side free kicks tend to be more successful, it might be worthwhile practicing the defence of balls that come in from this side.	92.9
Corner kicks, free kicks and penalty kicks					

Siegle and Lames [44]	Sixteen matches from German first league.	Number of occurrences per match (mean): (1) free kicks - 39.69; (2) corner kicks – 10.0, and, (3) penalty kicks – 0.13.	There is an average of 108 interruptions per match. Throw-ins and free kicks were most frequent. Goal kicks, corner kicks, substitutions, and kick offs occurred less often. Drop balls, penalties, and injuries occurred least often.	Teams utilize match interruptions to run down the clock. While referees should take care to ensure that teams do not do this, coaches, on the other hand, should be aware of this phenomenon and adapt players' behaviour according to the state of the match.	85.7
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Table 2. Studies with predominantly activity profile analysis - playing roles.

Study	Sample	Categories of player positions	Analyzed variables	Main results	Practical applications	Quality score (%)
Andrzejewski et al. [16]	31 players participating in the Union of European Football Association cup competitions during the 2008-2009 season.	Defenders, Midfielders and forwards.	Distances covered at different intensities.	The midfielders covered the longest average distance during the match. This was 3% and 7% longer than the distance achieved by the attackers and defenders, respectively.	Training individualization: 1) the range of intensity during the training activities should oscillate within the limits of individual physical loads of soccer players; 2) speed load is especially important to forwards, that cover twice the distance in sprint than do the midfielders.	87.5
Andrzejewski et al. [50]	147 players who played in 10 matches of the 2008-2009 and 2010-2011 UEFA Europa League seasons.	Central defenders, external defenders, central midfield players, external midfield players, forwards.	Total number of sprints and total sprint distance covered.	Significant differences were found between all players' positional groups, in particular, between the forwards, external midfielders and external defenders as well as between central defenders and central midfielders. Differences in total numbers of performed sprints were found between between the forwards, external midfielders, central defenders and central midfielders.	There is a necessity to apply distances between 10 and 20 m and longer than 20 m in soccer speed training. The number of sprinting distances (in match situation) within these two ranges is twice as large in forwards and external midfielders than in central defenders and central midfielders. It seems unfounded to apply 40 m or even 50 m running distances in soccer speed training.	87.5
Di Salvo et al. [51]	26 449 observations (2006–07 to 2009–10) from Premier league	Attackers, central defenders, central midfielders, wide	Distances covered at different	Players from Championship league traveled more total match distance than players from the	Coaches need to develop training programs that address the physiological	87.5

	(n=13991) and Championship league (n=12458). 1241 players from Premier league and 1494 players from Championship league.	defenders, wide midfielders.	intensities.	Premiership. Championship players covered greater distances during jogging, running, high-speed running, and sprinting. Premiership players covered more distance walking.	requirements necessary to cover the identified distances (at different intensities) according to the specific competition. The most important characteristics of play in the highest division are based on quality (technical, tactical and mental aspects), and not quantity (physical performance) of measures.	
Clemente et al. [48]	443 players from FIFA World Cup in 2010.	Goalkeepers, central defenders, external defenders, midfielders and forwards.	Distances covered at different intensities.	Statistically significant differences among tactical positions were found, concluding that each position has its specific demands.	The specificity of playing positions should be considered by coaches to organize the workout and to adjust the demands to the characteristics of players.	92.9
Varley and Aughey [47]	29 elite Australian soccer players.	Central defenders, wide defenders, central midfielders, wide midfielder and forwards.	High-velocity running, sprinting and maximal acceleration.	The number of efforts performed in all categories were position dependent. Wide defenders performed more maximal accelerations and central defenders and midfielders performed less sprints compared to all other positions.	Positional differences in high-intensity movements should be accounted for when developing specific conditioning drills.	85.7
Padulo et al. [49]	10 male Italian goalkeepers from Italian third and fourth divisions.	Goalkeepers.	Number of frontal and lateral actions with distance covered and total distance covered during	Statistically significant differences among lateral actions, distance covered and total distance covered as function the competitive level.	Training programs should be determined by the identified running kinematic variable values in order to cope with the specific match demands. Measurement of kinematic	87.5

			match.		variables provides coaches with the necessary information to plan specific training sessions for goalkeepers, in order to induce competitive performance improvements.	
Liu et al. [52]	380 matches from Spanish first league (2012–2013).	Full back, central defenders, wide midfielder, central midfielder.	21 performance related match events and actions.	Technical performances differed between players of strong and weak teams from different perspectives across different field positions.	The players` technical profiles can be useful for talent development and player selection in the transfer market. Nevertheless, these types of player profiles need to be analyzed according to the specificities of player roles and the characteristics of current (or future) teams.	100

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Table 3. Studies with predominantly activity profile analysis - fatigue influence.

Study	Sample	Variables/tests	Period analysed	Main results	Practical applications	Quality score (%)
Silva et al. [17]	13 professional male soccer players competing in Portuguese soccer League.	Distances covered at different intensities; countermovement jump, 5- and 30-m sprints, change of direction, knee extensor and flexor isokinetic strength, and Yo-Yo intermittent endurance test-level 2.	Performance across 1 sporting season.	The footballers covered a greater high-intensity distance running in the last quarter of the season than in the second and the third quarters. The peak 5-min was higher in the last quarter than in the beginning of the season (first, second and third quarters). Soccer players covered more high-intensity distance running during the match and in the P5-min toward end of season.	There exists an association between muscle strength and power and performance decrements in game-related physical parameters. In this sense, football players' training should incorporate specific exercise programs to improve the athletes' strength and power during the performance of soccer-specific activities.	93.8
Dellal et al. [54]	16 outfield players from French soccer league.	Distances covered at different intensities, technical parameters, injury data.	Performance across a congested fixture period.	No differences were found across the successive matches in the congested period, and between no congested and the three congested periods for all the physical and technical activities. The injury rate during match-play was significantly higher during the congested period. The injury rate during training time was significantly lower during the congested period compared.	The coach, physiologist and medical staff should pay special attention to the recovery strategies when players participate in matches during congested fixture periods. Using low intensity activities during training sessions, rotation of players in games, attention to objective markers of fatigue (e.g., creatine kinase concentration) combined with subjective measures of performance, can be useful strategies at these times in	100

Penas et al. [55]	99 players from FIFA World Cup (2014).	Distances covered at different intensities. Time spent in low, medium and high activities, maximal running speed. Number of sprints per minute.	Influence of extra-time period on performance.	Total distance, high-intensity running, top speed and high intensity activities declines significantly from the first to the second half of the match. Low intensity activities increase in the second half and especially in the extra-time period in comparison with the first half. All of the physical markers under study showed a decline of 15-20% during the extra-time period in comparison to the first half of playing time.	the season. Technical coaching staff should be aware of the decline of physical performance in second half and extra-time period and consequently, influence the team from a tactical and physical focus within a training perspective.	100
Russell et al. [56]	5 professional soccer players from an English Premier League reserve team.	Distance covered and the number of sprints, accelerations and decelerations. Countermovement jump performance and creatine kinase concentrations were assessed.	Influence of extra-time period on performance.	From 105 to 120 min, acceleration and deceleration parameters reduced by >10% compared to the opening 15 min. Physical performance markers reduced throughout match play and countermovement jump performance was impaired, while creatine kinase remained elevated, for at least 48 h after the match.	Technical staff could use this information to inform team tactics and training sessions by implementing strategies that aim to minimise reductions in physical performance during the game and to enhance recovery in the days after a match, namely: 1) substitutions; 2) aerobic and anaerobic conditioning programmes, and; 3) nutritional supplementation protocols. To a certain extent, top players can cope with a busy match schedule without underperforming.	87.5
Soroka and Lago-Penas [53]	301 players from World Cup Brazil 2014.	Distances covered at different intensities.	Performance across a congested fixture period.	No differences were found across the analyzed matches in the distance covered in sprint, high-intensity running, moderate-intensity running and high-intensity running. There is not a clear tendency of how playing multiple matches modified the		86.7

				physical performance of players.		
Sparks et al. [57]	10 players from a South African University's first team.	Distances covered at different intensities. Yo-Yo intermittent endurance test-level 2.	Performance across periods of 5 minutes along the match.	The high activity groups showed: 1) moderate to large declines in distance covered between the first 15 min of the second half; 2) small to moderate declines in high-intensity running during the first 10 min of the second half. First-half activity profiles had a significant impact on recovery after the most intense 5-min periods.	First-half activity profiles had a significant impact on recovery after the most intense 5-min periods as well as on second-half performances, which may be attributed to the presence of transient fatigue.	75
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Table 4. Studies with predominantly activity profile analysis – substitutions.

Study	Sample	Variables	Main results	Practical applications	Quality score (%)
Bradley et al. [20]	1) An independent-measures analysis comparing the match-performance characteristics of players completing the entire match (n = 810) vs substitutes (n = 286) and the players they replaced (n = 286), English premier league; 2) A repeated-measures analysis comparing the same players completing full matches vs those in which they were introduced as a substitute (n = 94).	Distances covered at different intensities. Time and type of substitutions.	More substitutions occurred at halftime and between the 60- to 85-minute marks versus all first-half periods and the remaining second-half periods. More offensive substitutions involving attackers and wide and central midfielders were made between the 60- to 90-minute marks compared with defensive substitutes such as central defenders and fullbacks. Substitutes cover greater high-intensity-running distance. Both research designs indicated that attackers covered more high-intensity running than peers or their own performances when completing the entire match.	It seems that, from a work-rate perspective, substitution seems to be effective, but from a technical and outcome perspective this still needs to be established.	92.9
Rey et al. [22]	124 matches (677 substitutions) from UEFA champions league.	Timing and tactics of substitutions.	Coaches tended to hold onto substitutions later when the team was ahead, but made substitutions earlier when either tied or behind. The probability of the substitution being offensive in tactical terms increased when a team was behind in a match. When a team was ahead, coaches tended to make more defensive substitutions.	Coaches should be aware that changing a losing scenario appears to depend on changing tactics early in the match.	100
Gomez et al. [21]	50 matches from Spanish professional soccer league.	Substitutions minute and number. Yellow card, functional position, Ball, shots on target possession percentage and ball regained.	When the teams were losing they made the substitutions quicker within the match when compared to if they were in a tied or winning position. Home teams made the substitutions quicker than when compared to the away teams.	The data demonstrate how an effective timing of substitution strategy can improve team performance and modify the final match outcome. Additionally, this	85.7

When a team is losing the substitutions take place earlier, and when winning a match make their first substitution later than the away team. The better the quality of opposing team the later the substitutions will take place

study provides coaches with new information on substitution patterns that can be easily integrated into their overall coaching strategy.

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Table 5. Studies with predominantly activity profile analysis - altitude and environmental heat stress influence on performance.

Study	Sample	Aim	Main results	Practical applications	Quality score (%)
Nassis [23]	64 matches from the FIFA 2010 World Cup.	Examine the effect of altitude on football performance.	It is concluded that playing football above 1200 m had negative effects on endurance but not on technical skills during world cup 2010 matches. It seems that teams should follow several days of acclimatization before playing at altitude as low as 1200 m, to ameliorate the negative effects of altitude on physical performance.	Teams should not fly to altitudes of 1,500–1,700 m just 1–2 days before a football match, as is common practice, but should acclimatize for several days.	85.7
Nassis et al. [24]	64 matches from the 2014 FIFA World Cup Brazil.	Analyze performance data in relation to the environmental conditions to identify potential association.	Top-level players seem to modulate their activity pattern during matches in a hot and humid environment to preserve the global match characteristics.	Technical and medical staff should be aware that football players may alter their physical activity pattern when competing in the heat. Footballers may reduce their physical performance, especially the high-intensity distance covered, to preserve their ability in key physical and technical performance indicators.	100

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Table 6. Studies of group behavior analysis with the use of team centroid values.

Study	Sample	Aim	Variables	Results	Practical applications	Quality score (%)
Frencken et al. [27]	1 match from champions league	Analyze instantaneous variability of inter-team distance.	Difference between longitudinal and lateral inter team centroids.	Greater longitudinal and lateral inter team centroids were found in 1 st half (7.10 and 0.58 m, respectively) than in 2 nd half (6.54 and 1.18 m, respectively). However, greater variability was found in 2 nd half. Strong correlations values between centroids were found in both halves. Threshold values for the selection of critical match periods were established at 0.83m (variability of longitudinal inter-team distance), 0.84 m (variability of lateral inter-team distance), $1.62 \text{ m} \cdot \text{s}^{-1}$ (longitudinal rate of change) and $1.66 \text{ m} \cdot \text{s}^{-1}$ (lateral rate of change).	The technique allowed to identify that dynamical analysis of inter-team distances most strongly relates to match outcomes, thus it can be possible to quantify the ability of players of the defending team collectively to respond to explorations of the attacking team to create opportunities to move forward to score	86.7
Siegle and Lames [62]	1 match from the final of 2006 FIFA World Cup.	Analyze the relative phase between team's centroid	Team centroid	A tight coupling between both teams in the longitudinal team centroid in the 1 st half (relative phase: $0.00 \pm 5.25^\circ$). The coupling in lateral team centroid was even smaller (relative phase: $0.00 \pm 3.84^\circ$).	Relative-phase analysis can analyze coupling between teams, groups or players. Perturbations of the game may be also captured by using relative-phase analysis.	78.6
Clemente et al. [61]	1 match from Portuguese first league.	Analyze the relationship between the centroids of the teams.	Wcentroid.	Spearman test revealed strong relationships between lateral-to-lateral wcentroid of both teams ($r_s = 0.707$). Similar evidences were found in longitudinal wcentroid ($r_s = 0.781$).	This measure represents the possibility to be adjustable to the position of the ball given information about the weight of the players around the ball. Moreover, distance between centroids may characterize the synchronization of the	85.7

						teams during the match.	
Gonçalves et al. [91]	1 simulated match of 50 minutes.	Analyze the variance of players based on group centroid.	Centroid per position (defenders, midfielders and forwards).	Results showed that all players were nearer and more coordinated with their own position-specific centroid. The results showed stronger in-phase attractions in lateral than in longitudinal direction.	The data can be immediately interpreted in interaction with the physical and physiological responses, thus providing more possibilities to observe the game in a holistic and valid way.	81.3	
Aguiar et al. [92]	6 bouts of 2-, 3-, 4- and 5-a-side games.	Analyze the player's distance to the centroid	Centroid	The distance between centroids presented a small decrease from 2- to 4-a-side SSG and a moderate to nearly perfect increase to 5-a-side small-sided game. Larger formats of play increase the absolute distance to both team and opposition team centroid.	4- and 5-a-side games are the preferred formats to achieve team-related emergent and self-organized behavior. Smaller formats are more appropriate to increase the unpredictability.	80	
FIFA - Fédération Internationale de Football Association; Centroid: geometrical centre of the team that does not depend on the ball; wCentroid: weighted geometrical centre of the team that depends on the position of the ball; SSG: small-sided game.							

Table 7. Studies of group behavior analysis with the use of team's dispersion.

Study	Sample	Aim	Variables	Results	Practical applications	Quality score (%)
Bartlett et al. [26]	10 matches from elite European soccer.	Associate team's dispersion with notational variables.	Surface area, stretch index, Frobernius norm, centroid.	Positive correlations between attacking variables and dispersions variables were found. It was suggested that teams expand together or contract together in synchrony. Strong positive correlations were found between teams in both directions: goal-to-goal and lateral-to-lateral ($r = 0.99$ and 0.76 , respectively). Goal-to-goal correlations were significantly stronger for the 64 successful attacks than for 241 unsuccessful attacks. No crossing of the centroids of the two teams along the pitch leading up to any of the 14 goals being scored from open play in 50% of the matches was found.	In 11-a-side games the centroid and associated dispersion measures are not sensitive enough to signify critical events. It can be possible to analyze multi-dimensional coordination of groups of more than two players using self-organizing maps. The team length and width should not be used to analyze 11-a-side games.	92.9
Moura et al. [28]	8 matches from Brazilian first league.	Characterize the Brazilian team's coverage are and spread on the pitch.	Frobernius norm, team coverage area.	All the teams had greater values of Frobernius norm and coverage area in situations of ball possession compared to when they did not. In defending situations, the greater values of coverage area and Frobernius norm were observed when teams suffered shots on goal in comparison with tackles were performed. In attacking situations, greater values of Frobernius norm and team coverage area were observed when teams suffered tackles in comparison when shots to goal were performed.	Using automatic tracking systems would be possible to identify the organization on the pitch and systematize tactical strategies of the teams.	85.7
Duarte et al. [59]	1 match from English Premier League.	Analyze the variations in the patterns of collective behavior.	Surface area, stretch index, team length, team width, geometrical	Team length tended to be larger in the home team. Greater values of width were found in visiting team during the first two periods. The fourth and fifth periods had greater values of dispersion in the home team, after two goals scored. Approximate entropy	The five collective measures used in this study can be used to capture the idiosyncratic performance values of each team as	85.7

			centre.	applied in dispersion measures revealed a decrease of complexity with the course of the match. Approximate entropy of geometrical centre per time period: 0.36 (0-15min); 0.34 (15-30min); 0.30 (30-45min); 0.40 (45-60min); 0.34 (60-75min); 0.36 (75-90min). Larger average values of geometrical centre were found in 1 st half than in 2 nd half.	competitive performances unfold. Data also revealed that teams tended to be less complex over the time but increasing the magnitudes of variation in their organizational shape.	
Fradua et al. [69]	4 matches from Spanish first league.	Examine the playing area during official matches.	Playing area, length and width of the team, distance from defending and attacking goalkeeper.	Individual playing area: 1) ranged between 78.97 and 93.87 m ² ; 2) has changed based on the location of the ball; 3) decreased as the ball approached the middle of the pitch. Rectangle that included all outfield players became smaller as the ball was played in the middle of the pitch.	This study revealed that small-sided games can replicate the tactical aspects of full matches. The pitch sizes are valuable for ensuring training specificity of tactical factors. Build-up play and finishing phase should be played with individual player area of 90m ² and transition play should be designed with an individual area of 80m ² .	93.3
Castellano et al. [68]	6 matches from one team of the Spanish first league.	Identify collective tactical decisions from high-level teams.	Surface area, team length, team width.	Length and width were greater in offensive phase in the matches against weak opposition. Both variables were also greater during the defensive moments against stronger teams. The range values of surface area were 800-2800 m ² in offensive moments and 1000-2000 m ² in defensive moments.	The use of standardized pattern behavior in the attack can be influenced by the quality of opposition. Therefore, a more efficient coordination in defensive phase may be appropriate to reduce the effects of higher opponents' irregularity.	85.7
Clemente et al. [25]	3 matches from Portuguese first league.	Analyze the variance of weighted stretch index, surface area and effective area	Weighted stretch index, surface area, effective area of play.	Greater dispersion was found during offensive moments in comparison with defensive moments. Weighted stretch index, surface area and effective area of play were smaller in 2 nd half in situations with and without ball possession.	The use of collective measures allowed to identify that dispersion of the team decreased from the first to the second half of	92.9

		of play between halves.		With and without ball, longitudinal wcentroid was greater in 1 st half. Higher values of lateral-to-lateral wcentroid were found in 1 st half. Lateral-to-lateral wcentroid reveals a change of flank between the first and second halves with ball possession.	the match. An expansion-contraction relationship between the offensive and defensive moments was also detected by the collective measures.	
Frencken et al. [29]	4 small-sided games (4-a-side format).	Analyze the effect of three small-sided games in the surface area.	Surface area, difference between surface areas.	4-a-side played at 24×16 m had greater surface area ($38\pm31\text{m}^2$) in comparison with 30×20 m ($34\pm29\text{m}^2$), 30×16 m ($31\pm25\text{m}^2$) and 28×21 m ($28\pm21\text{m}^2$). Differences between surface areas (of both teams) were greater at 30×20 m and 24×16 m pitch.	Expert players can extract more pertinent information related to player movements more quickly than novices. Absolute values of inter-team distances are key performance indicators in which the strength of the relation is an indicator of playing level.	92.9
Moura et al. [66]	20 matches from Brazilian first league	Characterize the surface area and spread time series of soccer teams.	Surface area Frobernius norm.	Both surface area and Frobernius norm were distributed at low frequencies and the values decreased the median frequencies of the time series from the first to the second half.	Surface area and spread time series were distributed at low frequencies and the median frequencies decreased from the first to the second half. Results can provide valuable information about the player organization on the pitch and can be used by coaches to adjust tactical behaviors and design new tasks.	85.7
Clemente et al. [65]	3 matches from Portuguese first league.	Analyze the variance of weighted stretch index, surface area and effective area of play in the three-	Weighted stretch index, surface area, effective area of play, wcentroid.	Statistical greater values of weighted stretch index ($17.44\pm3.48\text{ m}$), surface area ($1735.63\pm511.99\text{ m}^2$) and effective area of play ($967.41\pm663.49\text{ m}^2$) were found in drawing score. No differences were found between losing and winning situations. Lateral-to-lateral wcentroid was statistically greater	Losing status reduced the space between teammates, increasing the compactness. In drawing situations, the dispersion of teammates was greater than in losing or	85.7

		possible match status.		in losing matches than in drawing or winning. Similar evidences were found in longitudinal wcentroid.	winning moments. The measures used in the study can provide valuable information about the collective organization of the teams.	
Clemente et al. [70]	3 matches from Portuguese first league.	Analyze the variance of defensive play area between halves and match status.	Defensive play area, number of triangulations per defensive play area.	Greater values of defensive play area (defensive backward region, defensive first half of the middle region, defensive second half of the middle region, defensive attacking region) were observed in loss matches and during 1 st half. No differences were found between drawing and winning status.	Defensive play area allowed to characterize the inter-sector process during defensive moments. The triangulations allowed to characterize the space between teammates and how they behave during the defensive process.	92.9
Moura et al. [67]	20 matches from Brazilian first league.	Test the anti-phase and in-phase of teams spread.	Frobernius norm.	An in-phase relationship between opposition teams spread was predominantly found. Sequences ending in shots on goal represent greater anti-phase.	Coaches should be aware that the way their own teams are organized may induce changes in the opponent's organization. Such results may be used as a tactical strategy by the coaches to influence the status and dynamics of the game.	92.9

Table 8. Studies of group behavior analysis with the use of network.

Study	Sample	Aim	Variables	Results	Practical applications	Quality score (%)
Grund [30]	76 matches for 23 teams of English premier league.	Analyze the association between network density and centralization and the number of goals.	Network density, network centralization.	Greater levels of network density lead to higher number of goals. Inversely, a more centralized interaction lead to decreased number of goals. A correlation value of 0.18 was obtained for the association between network intensity and goals. A negative correlation of -0.10 was obtained between centralization and goals.	Using this analysis could be possible to measure team performance and network structure at the same time. It can be argued that the realized passing structure represents an underlying pattern of orchestrated team play.	85.7
Cotta et al. [31]	3 matches from Spanish team during 2010 FIFA world cup.	Analyze which players tend to cluster together during passing sequences and the most important players.	Clustering coefficient, Centrality.	Xavi Hernández (central midfielder) and Xabi Alonso (defensive midfielder) were the dominant players during the quarter-finals. Clustering coefficient was greater between 10-20', 40-50' and 70-80' periods. In the semi-final, clustering coefficient was greater in the 1 st half. Once again, Xavi Hernández and Xabi Alonso were the dominant players of the team. Clustering levels were greater in the periods of 20-30', 60-70' and 80-90' of the final. Xavi Hernández and Cesc Fàbregas (central midfielder) (in the extra time) were the prominent players.	The measures used on the study allowed to capture the combinative nature of the “tiki-taka” style. The techniques can be promising in the future to find other properties associated with each team.	85.7
Gama et al. [74]	6 matches from Portuguese	Analyze the connection tendency between teammates.	Relative frequency odds method.	Left defender was the dominant player in to receive and in to pass the ball. Participation of players 5, 8 and 25	The approach allowed to capture intra-team interactions across	78.6

	first league			influenced the ball movement during attacking.	several games, suggesting that some structures of play are stable in the team independently from the opponent.	
Clemente et al. [32]	64 matches from 2014 FIFA world cup.	Analyze the variance of in-degree, out-degree, closeness and betweenness centralities between playing position and tactical line-up.	In-degree, out-degree, closeness centrality, betweenness centrality	Midfielders and external defenders had the higher values of in-degree, out-degree, closeness and betweenness centrality, especially in 1-4-3-3 and 1-4-2-3-1 tactical line-ups. The greater mean values per each measure were: in-degree (12.32), out-degree (13.46), closeness (10.17) and betweenness (14.40).	The use of centrality network measures can be used to classify the prominent players in an opposing team and develop defensive strategies to negate the actions of these players during a competitive game.	92.9
Clemente et al. [71]	64 matches from 2014 FIFA world cup.	Analyze the variance of general network measures between levels of the teams and associate the metrics with goals scored.	Total links, network density, network diameter, clustering coefficient.	Goals scored showed a small positive correlation with total links, network density and clustering coefficient. Shots had also a small positive correlation with total links, network density and clustering coefficient. Network diameter had negative correlations with goals scored and shots. Teams that achieved the highest stages in competition revealed greater capacity to disperse the ball for all teammates.	The study suggested that the farthest relationship between teammates can lead to decrease the possibility of scoring and shooting. Based on that, coaches may design games to increase the homogeneity of relationships and to improve the possibility to be successful in attacking situations.	85.7
Clemente et al. [93]	5 matches from Portuguese first league.	Characterize the general properties of network.	Network density, network heterogeneity,	Network density varied between 0.3199 and 0.3795. Network heterogeneity varied between 0.4182 and 0.4901. Network centralization varied between 0.2114 and	The general measures allowed to identify how players connect with each other and the kind	85.7

			network centralization.	0.3127.	of strength of the connections between them.	
David and Wilson [94]	1564 matches from national and international competitions .	Test the association between cooperation and winning; verify the contribution of each kind of cooperation for the winning; test the effect of competition on the cooperation; test the association between cooperation and shooting at goal.	Degree, strength, division of labor.	Winning teams had higher frequencies of network interactions. Teams with less division of labor were more successful. Greater: 1) levels of passes were found in international tournaments; 2) levels of network were associated with proficiency in shooting at goal; 3) volume of successful passes was associated with decrease in the opposition's number of network interactions.	This study allowed to show that greater activity in physical tasks may not translate into greater team success, showing that greater physical effort is not associated with winning. Rather, more passing interactions between particular players are likely to create better scoring opportunities to win matches.	85.7
Clemente et al. [72]	7 matches from 2014 FIFA world cup.	Test a new network software. Characterize the network process of the FIFA world cup.	Total links, graph density, degree centrality, degree prestige, betweenness centrality.	Midfielders had the greater values of degree prestige (12.83), degree centrality (12.58) and betweenness (13.50). The competition stage did not influence the prominence level of playing positions. Total links and network density did not significantly change between group matches and elimination matches.	The uPATO (ultimate performance analysis tool) software can be used by coaches and staff to quantify the teammate's interactions and produce adjacency matrices to be imported in social network software to generate information about the collective structure of the team.	100
Clemente et al. [75]	36 matches from Portuguese first league.	Characterize the network process during scored and conceded goals.	Clustering coefficient, indegree centrality,	Attacking midfielder, left and right forwards were the players with greater indegree centrality during attacks that resulted in scored goals. Higher values of	Using network analysis, it was possible to classify the most commonly used	

	outdegree centrality, closeness centrality, betweenness centrality.	outdegree were found in right defender. The greatest betweenness centrality was found in left and right forwards. The greater clustering coefficients were found in attacking midfielder and striker. Strikers and left forwards were the prominent playing positions in to receive the ball in the conceded goals. The greatest value of clustering coefficient was found in left defender during in the conceded goals.	interactions of a team to produce goals and how they conceded goals against opponents. Moreover, the passing trajectories allowed to identify that passing to teammates in forward sectors of the field increases the possibility of scoring.	92.9
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Table 9. Studies of group behavior analysis with the use of sequential patterns.

Study	Sample	Aim	Variables	Results	Practical applications	Quality score (%)
Camerino et al. [34]	10 matches from Spanish first league and UEFA champions league.	Describe the patterns during attacking processes.	Attacks categorized by lateral position, zone, possession and interaction contexts.	It was revealed regularities of Barcelona during the ball possession. It was suggested that the first patterns include to penetrate in the central zone before to moving forward and attack as close as possible of the opposition's goal.	T-patterns have enabled to identify and define playing offensive style of the team providing a notion of most representative interaction, the way of moving toward the opposing team's goal or the preference to pass the ball in the pitch.	85.7
Sarmento et al. [33]	36 matches from FC Barcelona (n=12), Manchester United (n=12) and Internazionale Milano (n=12).	Describe the patterns during attacking processes.	Attacks categorized by zone, possession, technical action and interaction contexts.	Different sequential patterns are identified as result of different philosophies of the match. Specific differences are identified regarding the starting zone, development and finalization of the offensive process.	The two analysis techniques helped to understand the tactical patterns of teams. It also identified individual characteristics of the players and the tactical technical and strategic aspects of the game. Based on that it can be possible to design more appropriate exercises. Results suggested that t-pattern analysis can be an effective tool to	92.9
Zurloni et al. [35]	19 matches from Italian first league.	Detect the dynamics of attack actions.	Attacks categorized by lateral position, zone, lateral passing, zone	Patterns were more consistent in winning situations. It was found a pattern to exploit the right		78.6

			passing, recovery and loss and ball out of play.	side to moving forward. In losing situations, patterns were less regular, maybe to test alternative strategies to gain the control of the match.	support research in sport performance analysis. It was possible to differentiate structures of play in won and lost games.	
Cavalera et al. [76]	3 matches from Italian first league.	Analyze the variance of temporal patterns between halves.	Attack situation with the following ending: goal, non-goal and permanent loss.	Number of patterns statistically increased from the first to the second half. Structure and distribution between won and lost matches were also different. Inversely with the 1 st half, patterns of attack during 2 nd half frequently ended with a shot.	It was possible to identify recurrent behaviors related to effective and non-effective playing actions and to show countermeasures to opposing teams	78.6

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Table 10. Studies of group outcomes.

Study	Sample	Categories	Variables	Results	Practical applications	Quality score (%)
Castellano et al. [36]	177 matches from FIFA World Cup 2002, 2006 and 2010.	Winning, drawing, losing.	Goals scored, total shots, shots on target, shots off target, ball possession, number of off-sides committed, fouls received and corners.	Total shots and shots on target (made and received) were the main discriminatory variables between winning, losing and drawing. Ball possession was not a discriminating variable.	The game models based on indirect style seems to have more chance of success. The effectiveness of attacking play and ball possession appear to be the key indicators for the success and for that reason coaches must consider such information to organize their models of play.	92.9
Gomez et al. [78]	1900 matches played over 5 seasons on the Spanish professional league.	Match location, final outcome	Goals, shots, committed fouls, turnovers, ball recovers, crosses.	Home matches lead to higher frequencies of goals, shots, committed fouls, turnovers, ball recovers and crosses. Winning teams had greater volume of ball recovery at zone 2 and tend to organize the offence through long passing sequences.	Home matches contribute to increase the frequencies of notational variables. Discriminating winning from losing teams may also provide information about which variables must be considered by coaches to increase the possibilities for success.	85.7
Lago-Ballesteros et al. [79]	12 matches from Spanish professional league.	Playing tactics, opposition interactions, situational	Success, duration, starting zone, team possession type, pass number,	When the team was drawing or winning, the probability of reaching the penalty area decreased dramatically compared with the	Direct attacks and counterattacks were more effective than elaborate attacks for	92.9

		variables.	players in possession, passing options, opposition, defensive pressure, match location, quality of opposition, match status.	losing situation. Teams often show a more defensive strategy when winning than when losing, and vice versa. Direct attacks and counterattacks were more effective than elaborate attacks. Playing against less than six defending players increased the offensive effectiveness Long possessions were more effective than short possessions.	increasing penalty area penetration (defined as an entry into the opposition penalty area with a high degree of control over the ball or when a set play is given to the attacking team as a result of penetration of the penalty area). A more defensive strategy is adopted by winning teams than by losing teams. Such information can be used by coaches to organize the training tasks and to anticipate situations of winning and losing.	
Collet [95]	6078 matches from clubs and 299 national teams.	Possession Outcome.	Completed passes, total passes, shots on goal, outcome.	Ball possession was a direct predictor of positive outcomes in domestic competitions in Europe as well as European club and national team competitions. Teams that possessed the ball more shot more and scored more. Frequent and accurate passing were strongly linked to shots, goals and points.	Tactical analysis, based on possession of the ball, should consider variations in quality and the competitive context. Coaches must adapt ball possession strategies to increase the efficiency of the team in different scenarios.	100
Russell et al. [82]	One team during the 2010/2011 English championship season (23 home	First vs second half.	Possessions, ball distributions, touches taken per possession, number of challenges.	Total possessions and number of ball distributions were lower in the second versus the first half of match-play. Analysis across 15-min intervals revealed reductions during the last 15-min of match-play in the total number of possessions and	Transient changes occurred in selected measures of technical performance during match-play. Individual possessions and passes performed were smaller	92.9

matches).

distributions. The number of touches taken per possession, number of challenges, percentage of challenges won, length of forward distributions and percentage success of distributions were all similar between halves and across 15-min intervals.

in 2nd half. Coaches must consider such decreases in performance to adjust the training plan and minimize such occurrences.

Almeida et al. [80]	28 matches from UEFA champions league.	Home vs away; winning, drawing, losing; better-ranked, similar-ranked, worse-ranked.	Ball recovery type and zone.	There is a tendency for home and losing teams to defend in more advanced pitch zones. Better-ranked teams were more effective than worse-ranked teams in applying defensive pressure in more advanced field positions.	Coaches must be aware that promoting the intention to gain directly the ball from the opponents and pressurizing the opposing team near its goal seems to be more effective for defensive moments.	92.9
Bradley et al. [81]	54 matches from English Premier League.	Home vs away, quality of opposition, team ranking, minutes winning/drawing		Playing against weak opposition was associated with an increase in time spent in possession while playing away decreased the time spent in possession by ~3%. Possession was increased when losing than winning or drawing. The better the ranking of a team, the higher the time spent in possession.	Technical, tactical or physical aspects of performance can be adversely influenced by specific situational variables. Coaches and analysts may be aware of these variables and adjust training exercises and strategies to avoid decrease in performance.	92.9
Harrop and Nevil [37]	46 matches played by a League One soccer team.	Winning, drawing, losing.	Offensive and defensive match-related statistics.	More passes and passes in the opposition half were made when the team lost compared to when they won and drew matches. Lower percentage of successful passes were completed when the team drew.	Direct play is a more effective approach for teams without sufficient skill levels to sustain ball possession in a succeeded way. Coaches	85.7

Moura et al. [77]	32 teams playing 3 matches at FIFA World Cup 2006.	Winning, drawing, losing.	Shots, shots on goal, goals performed, fouls committed/suffered, corner kicks, direct free kicks to goal, offside, own goals, yellow cards, expulsions due to second yellow cards, direct expulsions, actual playing time with possession of the ball and percentage of ball possession in relation to the total time played.	The variables “shots”, “shots on goal”, “playing time with ball possession” and “percentage of ball possession” are important to discriminate the winning teams from the drawing and losing ones.	must also be aware that the location of the match can affect the team’s performance and the possibility of success. Variables such as shots, shots on goal, playing time with ball possession and percentage of ball possession can be used to adjust training plans to improve the execution level.	92.9
Wallace and Norton [18]	Final match of each of the 12 World Cup soccer tournaments between 1966 and 2010.	Elements of match structure and speed.	Play and stop periods, ball speed, player density and passing rates.	The increased intensity of play is paralleled by longer stoppage breaks which allow greater player recovery and subsequently more intense play. Defensive strategies dominate over time as demonstrated by increased player density and congestion.	A rapid self-organization in a dynamical system, underpinned by prediction speed and accuracy, skill and fitness will increase the probability of success.	92.9
Castellano and Casamichana [84]	320 first-division matches and 335 second-division matches.	First vs second Spanish division.	Variables related to use of space, physical performance and technical-tactical actions.	The most successful teams in the first division of the Spanish football performed differently to the other teams analysed in terms of the majority of variables studied.	Successful and lesssuccessful teams can be differentiated by performance indicators. Attention to such indicators could make it possible for coaches to improve team playing patterns so that they become more successful.	92.9

Paixao et al. [96]	20 matches from UEFA champions league.	Winning, drawing, losing.	Length of passing sequences (number of passes and duration).	Teams used preferentially long passing sequences when they were losing or drawing, and short passing sequences when they were winning. Additionally, tended to differently adapt the length of their passing sequences according to the evolving score-line.	Enable to characterize opponent's teams and to provide useful information to increase tactical knowledge and to improve the design and organization of practice sessions by simulating competitive scenarios.	92.9
Garcia-Rubio et al. [83]	475 matches from UEFA champions league.	Home vs away, scoring first, group stage, knockout phase, quality of opposition.	Shots on goal, shots off target, off-sides, corners, fouls committed.	Home teams when scoring first win 62.3% and 55.8% in the group and knockout stages, respectively. Regression analysis has found match location as predictor in group stage but not in knockout stage. Scoring first or match location variables can dissolve quality of opposition variable due to interactive effects of all variables.	Identifying the interactive effects of situational variables (home vs away, scoring first, group stage, knockout phase and quality of opposition) on the performance indicators can be possible to identify and prepare the team to increase or reduce this effect based on the team's need.	92.9
Carling et al. [97]	38 league matches per season played over five consecutive seasons.	League rankings, points won, goals scored and conceded and match outcomes.	Ball possession and possession in opposition' half, passes, forward passes, completed passes and forward passes, crosses and completed crosses, goal attempts and goal attempts on target, successful final third entries, free-kicks and 50/50 duels won/lost. Total distance and distance covered at high-speeds.	Highest number of points obtained by the reference team was observed in the championship winning season. In all seasons except 2008/09, there was an increase in the number of points won per match for the second half of the season with the largest augmentation. In 2010/11 season, the reference team scored at least one goal, did not concede a goal and scored first on a greater number of occasions.	Higher player availability for selection linked to the ability to remain injury-free combined with good defensive rather than attacking performance may contribute to an increase in the possibility of success.	92.9

Liu et al. [85]	496 matches of UEFA champions league (2009-10 to 2012-13 sporting seasons).	Three	k-means cluster analysis based on UEFA season club coefficients. High-, Intermediate- and low-level teams.	Comparisons on performance profiles of teams of all three levels of strength, showed that high-level teams are characterized by stability of performance no matter play against which opposition, with result pressure or where they play. On the other hand, performances of intermediate- and low-level teams are associated with more variation to different situational variables.	The observational analysis of opponents must identify the specific circumstances of performance . Moreover, observational analysis of their own teams must be used to identify specific patterns and conditions that coaches should use to enhance performance in training.	100
Gonzalez-Rodenas et al. [98]	30 randomly selected matches from US major league soccer.	Playing tactics, match location, match status, match half.	Scoring opportunities.	Set plays created a higher volume of scoring opportunities than recoveries and restarts. Match location, match status and match half had correlations with scoring opportunities.	Soccer coaches should be aware that counterattacking effectiveness, performance of penetrative actions and penetration of the attacking third of the field with the ball depend on the quality of the opposition. Thus, coaches must use tasks to simulate different positions of opponents trying to improve the effectiveness and solutions of the team.	92.9
Fernandez-Navarro et al. [86]	97 matches from the Spanish La liga and the English premier	English vs Spanish first league teams.	19 performance indicators – 14 describing aspects of attacking play and five describing aspects to defensive play.	Six factors, representing 12 different styles of play (eight attacking and four defensive), had eigenvalues greater than 1 and explained 87.54% of the total variance. Direct and possession styles of play, were the	Playing styles profiling can be used on opponents to identify their dominant styles. By knowing such profiles, coaches may prepare the	85.7

	league.			most apparent styles.	most appropriate strategies and tactical behaviors that may disrupt the opponent's organization.	
Hoppe et al. [99]	306 matches from German Bundesliga	Success across the season.	Match running performance.	The match running performance alone was not significantly correlated with the final points accumulated. Positive-significant correlations were observed for the match running performance with ball possession.	Running performance in competitive matches alone is not sufficient to estimate successful performance outcomes in German teams. The relationship between running performance and technical/tactical skills with regard to ball possession is most prominent. Coaches must consider the need to organizetraining sessions based on this relationship.	100
Lago-Penas et al. [38]	1826 matches from: English Premier League, French league 1, Spanish La Liga, Italian serie A and German Bundesliga.	Home vs away; winning, drawing, losing; scoring first; quality of opposition; intervals of 15 min.	The team that scored first, the minute in which the goal was scored and quality of opposition.	Home teams scored first in 57.8 % of matches and went on the obtain 84.85% of points won in these matches. When the away team scored first, they obtained only 76.25% of subsequent points. Three independent variables were significant factors on the final outcome: (1) the quality of the opposition; (2) the minute in which the first goal is scored, and; (3) the team scoring first.	Scoring the first goal assumes great importance in elite football. However, the effects of scoring first depends on the opposition quality. Coaches must prepare the team for different scenarios to increase the capacity to deal with the unpredictability of match status.	100
Liu et al. [100]	320 close matches (goal	Match location.	16 football match events.	All the variables showed clear within-team relationships and 11	Generalized mixed linear modelling can be more	100

difference \leq
2).

events showed clear between-team relationships with the probability of winning. Ball possession had a small negative within-team effect but a small positive between-team effect on winning.

powerful than to identify key performance indicators. More objective information can be collected and provided to coaches.

Winter and Pfeiffer [101]	Twenty-seven of the 31 matches played by the 16 teams of the UEFA Euro 2012.	Winning, drawing, losing.	Goal scoring index, ball recovery index, successful transition index, ball recovery time, ball loss index, prevented transition index, ball loss time, offence efficiency and time, defence efficiency and time.	The tactical metrics presented can be summarized in four factors (match speed, transition play after ball loss, transition play after ball recovery and efficiency in open (offence) play). These factors are suited to discriminate winners, losers and drawers with 64.81% correct classifications.	The four factors can be used by coaches to prepare the teams, organize the strategy and plan exercises to increase the possibility of success.	92.9
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UEFA – Union of European Football Associations; FIFA - Fédération Internationale de Football Association

Figure Legends:

Figure 1. Preferred reporting items for systematic reviews flow diagram

Figure 2. Scopes of match analysis