This is an electronic reprint of the original article. This reprint may differ from the original in pagination and typographic detail.

Author(s): Araújo, Duarte; Passos, Pedro; Esteves, Pedro; Duarte, Ricardo; Lopes, José; Hristovski, Robert; Davids, Keith

Title: The micro-macro link in understanding sport tactical behaviours: Integrating information and action at different levels of system analysis in sport

Year: 2015

Version:

Please cite the original version:

All material supplied via JYX is protected by copyright and other intellectual property rights, and duplication or sale of all or part of any of the repository collections is not permitted, except that material may be duplicated by you for your research use or educational purposes in electronic or print form. You must obtain permission for any other use. Electronic or print copies may not be offered, whether for sale or otherwise to anyone who is not an authorised user.
The micro-macro link in understanding sport tactical behaviours: Integrating information and action at different levels of system analysis in sport

Duarte Araújo¹, Pedro Passos¹, Pedro Esteves¹, Ricardo Duarte¹, José Lopes¹, Robert Hristovski² & Keith Davids³,⁴

¹CIPER, Faculdade de Motricidade Humana, Universidade de Lisboa, Cruz Quebrada Dafundo, Portugal
²Faculty of Physical Culture, University of St. Cyril and Methodius, Republic of Macedonia
³Centre for Sports Engineering Research, Sheffield Hallam University, UK
⁴FiDiProProgramme, University of Jyväskylä, Finland
⁵CIDESD, Research Unit for Inland Development, Polytechnic Institute of Guarda, Portugal

Running Title: The micro-macro link

To be submitted to Movement and Sport Sciences (2015)
Abstract (120 words)

The micro-macro link is a central issue of human movement sciences because it directly refers to coordination and scalability of movement behaviour at different levels of analysis in a complex system. This feature of complexity is a key feature to consider decisions and actions as transitional behaviours at different levels during performance. Transitions expressed through bifurcations are an important part of a universal decision-making process. Here we explain how individuals perform decision-making behaviours in a minimal social performance unit (competing dyads), and how these two levels are linked in a sports team. Finally we discuss some applications of the micro-macro links for the design of training programmes in team sports.

Keywords (max 6)

Decision-making, micro-macro links, perception and action, player-team interdependence, phase transitions, group behaviours
Introduction

The challenge of understanding how individual actions and group or team behaviours emerge during interactions is a foundational issue for several sciences, including sport sciences. Developing understanding of the link between these micro and macro levels of analysis would result to substantial advances in training methods for enhancing dynamic and group-based social cooperation and competition. The result would be a deeper understanding of how individual, dyadic and group behaviours emerge during performance. The micro-macro linkage is a central issue of human movement sciences because it directly refers to such problems as coordination and scalability of behaviours. Conceptualisation of performance at a macro level often seeks to abstract from the individual and synthesize certain features of behaviours of a group of performers (performance, communication, division of labour, etc.). Alternatively it is aimed at enhancing understanding of organizational structures (e.g., in sport administration and management processes). Also, the macro level is itself a structure, which possesses to a certain degree its own autonomy (cf., the superorganism concept, see Duarte, Araújo, Correia, & Davids, 2013), maintaining such a system beyond individual behaviours (Schillo, Fischer, & Klein, 2001). Apart from the solely individual-oriented (e.g., schema theory; Schmidt & Lee, 2011) or social-oriented theories (e.g., social systems theory, Luhmann, 1995), there are some approaches that try to explain the connection between individual actions and social structure, such as the ecological dynamics approach.

The ecological dynamics of behaviour is rooted in an integration of key ideas from Gibsonian ecological psychology and dynamical systems theory (Gibson, 1979; Kelso, 1995; Kelso &Engström, 2006; Turvey, 1977; Warren, 2006), emphasizing the laws and symmetry conditions at nature’s ecological scale of analysis. This approach
signifies that behaviour emerges from an environment-performer system, and is not an output determined solely by the ideas, perceptions and actions of an individual. The implication is that cognition (and its related activities like strategic and tactical behaviours in sports) cannot be conceived as something that occurs in the mind between a stimulus and the initiation of an action. Cognition is the capacity to use information which is ecologically specific to the control of action (Reed, 1997), i.e., it is the capacity to actualize affordances (opportunities for action, Gibson, 1979). A recent formalization of this approach was developed by Warren (2006). He presented the performer-environment system as a dynamical system coupled by both mechanical forces (action) and information flow fields (perception). From intending-perceiving-acting cycles behavioural dynamics emerge. In this conceptualised system, attractors (stable system regions) are intended states, while repellers are states that the system seeks to avoid, and bifurcations (decision points) give rise to transitions in behaviours.

During intending-perceiving-acting cycles there emerge transitions in action modes. A phase transition is an impressively sudden shift in a system’s behaviours. Some time ago, we proposed that decision-making could be explained as a phase transition where initial increases in an identified control parameter have little or no effect on system behaviours (Araújo, Davids, &Hristovski, 2006). However, at a critical threshold point, tiny increases in the value of that control parameter can suddenly induce massive changes in system behaviours. Importantly, fine-grained time scales of measurement analysis expose details in the transition itself, uncovering the processes that enable the emergence of decision-making behaviours (Spivey, Anderson & Dale, 2009). If a dynamical system actually allowed extreme parameter changes from one instant in time to the next it would run the risk of oscillating around, rather than settling in a régime that it is moving toward during a phase transition (Spivey et al.,
Typically, therefore, most dynamical systems that appear to undergo sudden phase transitions from one stable regime to another may actually be functioning on a finer time scale than is being measured. That is, the processes that are carrying out the transition itself are doing so at an iteration-by-iteration scale that has the system spending a number of time steps in a region of parameter space that is in between two identified phases.

**The micro-macro link explained: The process of phase transitions**

Clearly, these ideas suggest the need for greater understanding of phase transition processes. Due to the interaction among their (microscopic) subsystems, open systems are able to (self) organize themselves by forming coherent, that is, spatially and temporally ordered patterns (Haken, 1996). The resulting patterns are macroscopic in nature and may be described by a small number of collective variables, conventionally referred to as order parameters. In many instances, spontaneous switches between macroscopic patterns occur as non-equilibrium phase transitions in equivalence with similar qualitative, structural changes in thermodynamical systems (where the term ‘phase’ refers to a system’s state). Switches in macroscopic patterns are a hallmark property of self-organization processes, as they involve no external agents imposing order from the outside. Phase transitions abound in both the inanimate and animate world, as exemplified by spontaneous shifts in gait patterns that are observed when velocity of locomotion gradually increases (e.g. from walking to jogging in humans).

The importance of non-equilibrium phase transitions can be appreciated when it is understood that qualitative changes in macroscopic patterns are accompanied by loss

---

1 This is true only if switches arise as a consequence of control parameter change. However, switches between patterns can arise also for a single value of a control parameter as a consequence of combination of noise (perturbations) and multistability, or through what Kelso (2012) terms metastability.
of stability of an original pattern. The stability of a system’s state (or phase) is predicated on the notion that every perturbation out of that state causes a system to return to it; when stability is lost, the system does not return but rather switches to another stable state (or phase). Close to points of (macroscopic) instability, the time to return to stability increases tremendously. As a result, the macroscopic state evolves rather slowly in response to perturbation, whereas the underlying microscopic components maintain their individual time scales of behaviour. The implication is that, from the perspective of the slowly evolving macroscopic state, the microscopic components change so quickly that they can adapt instantaneously to macroscopic changes. Thus, even though macroscopic patterns are generated by the subsystems, they order the subsystems or, put metaphorically, they ‘enslave’ their generating microscopic structures (Haken, 1977). Ordered states can always be described by very few variables (at least in the close neighbourhood of macroscopic changes), and consequently the state of the originally high-dimensional system can be summarized by a few, or even a single, collective variable(s): the order parameter(s). The relationship between the subsystems and the macroscopic structure, the latter generated by individual subsystems, and which in turn enslave the individual subsystems, implies a circular causality. This process effectively provides a low-dimensional description of the dynamical properties of system under study (Kelso & Engström, 2006).

**Transitions in tactical behaviour**

Transitions among stable performer-environment system states occur as a result of dynamic instability (Kelso & Engström, 2006). This process is the universal decision-making process for switching between, and selecting among, polarized states. Therefore, bifurcations function as selection mechanisms. If more functional action
solutions exist to fit performance circumstances, fluctuations will help performers
discover, and harness them.

In more detail, decision-making, or tactical behaviours, emerge from the
performer-environment system, when during the path of this system, they collapse into
a more attractive (stable) state in a landscape of possibilities for action (affordances)
available to a system. This process can occur without an explicit involvement of
additional inference mechanisms or mediating rules located within the performer, as
some organismic-centred theories of decision-making argue. We have previously
shown that searching for options (task solutions) during performance takes place within
a landscape of attractors, all with various depths or stabilities (Araújo et al., 2006).
From this viewpoint, the importance of each individual's exploratory behaviours is
emphasised as they actively seek to discover functional regions of the attractor
landscape of action possibilities to achieve a stable, performance solution. A viable
option to be selected is the strongest attractor. If the task is "familiar" to the performer,
this influence is reflected in the intrinsic dynamics of the individual. That is, the
individual, through learning and task experience, becomes perceptually attuned to
sources of information that specify goal achievement and a selected option is an
affordance that is picked up (Fajen, Riley & Turvey, 2009). Ignoring other options is a
dynamical consequence (i.e., a collapse of options), since if a system relaxes to one
attractor it concomitantly ignores other remaining options (attractors). That is, an
emergent decision is not a consequence of a 'costs-benefits' balance analysis calculated
in a performers’ head (McKinstry, Dale, & Spivey, 2008), but rather a selection that
emerges from the continuous interactions of an individual and a performance
environment. The fact that there is a stronger attractor does not completely eliminate the
influence of other attractors in the dynamic landscape of action possibilities. Under
dynamic performance conditions, other attractors (i.e., other options) can emerge and exert their attraction, emphasising the continuous nature of performer-environment reciprocity.

From a dynamical systems perspective, the same principles can describe the properties observed at different levels of the system, capturing its fractal nature (Renshaw, Davids, Shuttleworth, & Chow, 2009). This idea has important implications for understanding team sports performance since it captures the interdependence between individual and (sub)group levels of analysis. That is, changes at one level (individual) may influence performance at other levels (sub-groups and the team). Next, we present data from some previous studies that have exemplified how the decision making behaviours of performers (e.g., attackers or defenders in dyads) and sub-groups or teams might be constrained by environmental sources of information, and how all system levels are linked.

**Ecological constraints on dyadic behaviour in sports**

The work of Lagarde and colleagues (Lagarde, DeGuzman, Oullier, & Kelso, 2006) in boxing captured performers’ head motions as the basis to measure patterns of interpersonal behaviours that emerged between competing individuals. Data revealed that despite the random motions of the boxers, head radial distance oscillated between periods of contraction and expansion, which expressed continuous decision-making behaviours, sustained by information that was locally available from the continuous interactive movements of the performers. Lagarde et al. (2006) concluded that preferred modes of interaction (i.e., attractors) in competing boxing dyads could be characterized by the speed of change of head radial distance. Moreover, based on values of the boxers’ head velocity, results demonstrated patterns of interactive behaviours between
the performers. This observation was supported across a wide set of combined
individual actions, leading the authors to suggest the existence of degeneracy in the
boxer-boxer system, meaning that different system solutions were found for the same
performance outcome (Lagarde, et al., 2006).

Similar evidence of emergent decision making behaviours has been reported in
studies of sailing, a sport with different task constraints than boxing. Araújo and
colleagues (2003, 2006) analyzed the interactive behaviours of two competing boats
during the start in competitive regattas. Task and environmental constraints, such as
wind shift tendencies (intensity and direction), and the positioning of competing boats
during the start period, influenced racer' decision-making behaviours. During the
starting procedures, each sailor needed to continuously perform adaptive manoeuvres
aiming to achieve the most advantageous position at the precise moment of starting the
race. A boat's relative position (constrained by the direction and intensity of the wind),
at the starting moment, is crucial for gaining an advantage that is related to successful
performance in sailing regattas.

In team ball games, like basketball, futsal, soccer and rugby union, distances
between opposing players, between teammates, and distances to a target (basket or goal)
are task constraints that provide relevant information to guide players’ decision-making
behaviours. Original research in basketball highlighted how a task constraint, like the
value of interpersonal distance between an attacker and a defender, bounded their
decision-making behaviours in relation to the distance of the attacker-defender dyad to
the basket (Araújo et al.2002, 2004; Davids et al., 2006). To understand these emergent
decision-making behaviours, direct manipulation of task constraints has been very
informative (e.g., Cordovil et al., 2009). In basketball, Cordovil and colleagues (2009)
sought to establish the influence of task constraints, such as specific coaching
instructions, on a ball carrier's decision-making behaviours. The results pointed out that conservative coaching instructions constrained the ball carrier to keep the ball for longer, increasing the diversity of dribbling displacement trajectories, and taking more time to cross the mid-line of the court during play. Another important task constraint was the different height relationship that exist in specific attacker-defender dyads (e.g., tall players linked with shorter opponents). Cordovil and colleagues (2009) reported that the tallest attackers adopted a more conservative approach to dribble past the defenders (before mid-line court). The authors argued that the tallest attackers tend to gain an advantage as they move closer to the basket, and thus they do not need to take unnecessary risks (e.g., losing ball possession) at a distance from the basket. Individual constraints like the specific posture of the defender (i.e., right foot advanced, left foot advanced or neutral positioning) have also been observed to influence an attacker’s decision on the side to dribble past a defender, in a 1v1 basketball sub-phase (Esteves, de Oliveira, & Araújo, 2011). In addition, the time of exercise throughout a training session has been reported to positively influence the quality of the decision-making behaviours in a 3v3 basketball game. More time on the task (possibly inducing fatigue) was accompanied by a higher shooting efficacy, which with other performance indicators, highlighted the self-organizing, emergent tendencies of team decisional processes (Esteves, Araújo, & Barreto, 2007).

In team sports, several studies have investigated how players regulate their behaviours by using spatial-temporal information about their relative positioning (i.e., interpersonal distances) to the nearest opponent and the goal (Headrick, Davids, Renshaw, Araújo, Passos, et al., 2012). The work of Headrick and colleagues (2012) investigated how the proximity to a goal influenced the decision making behaviours of both a ball carrier and an immediate defender. The task design was a typical 1v1 soccer
subphase performed at three different locations on the pitch: i) attacking the goal (i.e., at the edge of the defender's penalty area; ii) in midfield; iii) advancing away from the goal (i.e., at the edge of the attacker's penalty area). Results demonstrated that the players’ decision-making behaviours were highly constrained by the location on the pitch where the ball carrier-defender interactions occurred. The defenders tended to display a more conservative behaviours closer to the goal (i.e., maintaining a higher value of interpersonal distance than when he/she was located further away from the goal). This decision making process was emergent since no specific instructions were provided to any of the participants and it is possible that a defenders perceived that a poor decision could lead to an advantageous situation for the ball dribbler.

An important task is to search for other variables that synthesize the dynamics of the dyad decision-making behaviours. This empirical strategy has led to the proposal of the angle composed by the attacker-defender vector with respect to the try line, as a key performance variable in rugby union (e.g., Passos, Araújo, Davids, Gouveia, Milho, et al., 2008). Data from this body of research have highlighted that the decision-making processes of attacker-defender dyads were consistently captured by this variable. Specifically, three different possible attractor states of an attacker-defender system were identified. In the same study, there was a tendency for greater irregularity in the movement of the dyad in successful tackles situations (i.e., providing the basis for a successful move by the defender) and smaller levels of irregularity when tries emerged (i.e., providing the basis for a successful move by the attacker) (see Araújo et al., 2013, for the complete mathematical model). Building on these observations, Vilar, Araújo, Davids, and Travassos (2012) analyzed the influence on interpersonal coordination tendencies of the relative positioning of competing attacker-defender dyads to the ball and the goal during performance in futsal. Data from 5v5 matches
showed that defenders’ and attackers’ coordination modes were preferentially in-phase, both for distances and angles to the ball and to the goal. However, defenders tended to position themselves closer to the goal with a lower angle, while attackers remained closer to the ball with a larger angle. Goals were scored when the distances to the goal by the attacker and the defender were very similar, but the angle to the goal of each was slightly different. These observations permitted us to conclude that the angular location of a defender and attacker to the goal (i.e., the values of their specific angles to the goal) provided greater influence on the decisional behaviours of the attacker-defender dyads than the ball location.

Examining the role of information manipulation, studies of performance rugby union by Correia, Araújo, Craig and Passos (2011) showed how the manipulation of initial starting locations (i.e., starting values of interpersonal distances) between two defenders near the try line influenced the ball carrier’s decision-making, and vice-versa. Results demonstrated how changes in the defenders’ starting values of interpersonal distance influenced their decision-making. The manipulation consisted of decreasing the space made available, between the defenders. For instance when there was a higher value of interpersonal distance between defenders, they tended to approach each other (i.e., decreasing interpersonal distance) and only after this procedure did they move forward towards the ball carrier. The data from the study of Correia and colleagues is in agreement with other studies in rugby union that showed that the value of interpersonal distance between attackers is a relevant task constraint to explain decision making and action (Passos et al., 2008). Decreasing values of interpersonal distance between players in a dyad is contextually dependent (Passos et al., 2008). The decreasing of interpersonal distance values moves seems to move a dyad towards critical performance regions, i.e., regions where contextual dependence is very high (Passos, Araújo,
Within these regions other key system parameters tend to exert strong influence over the behaviours of a dyadic system (Passos et al., 2008), such as the relative velocity of the competing players. Data have revealed that, if the value of relative velocity is increasing, a performance advantage tends to emerge for an attacker. If, on the other hand, it tends to decrease, a performance advantage often emerges for a defender. These data from previous research have all clearly demonstrated how information is locally created by players’ continuous interactions during performance.

The component parts of a team (the players) do not have any central commander that dictates players’ decision-making behaviours (Silva et al., 2013). The findings clarify how dyadic system behaviours in sport are emergent and predicated on self-organized tendencies. A most important empirical point to observe is that in all studies, no specific performance instructions are provided to participants apart from general information on scoring or preventing scoring within the laws/rules of each game. This empirical requirement in our work ensures that all perceptions, cognitions, decisions and actions are emergent in nature.

Ecological constraints on team decision-making in sport

Team ball sports are comprised by two opposing teams, which interact in order to fulfil their mutually exclusive goals (Lames &McGarry, 2007). These interactive behaviours, within and between teams, lead to the emergence of collective actions that characterise different game sub-phases (e.g., 1vs1, 2vs2, 3vs2). Due to continuous interactions between team-mates and opponent players, the totality of behaviours of such a complex system cannot be explained solely by a reductionist assumption of its individual contributions (McGarry, 2009). These macro behaviours are captured by
order parameters (i.e., collective variables) that accurately describe a system’s patterns of behaviour and their dynamics (Kelso & Engström, 2006).

Considering the micro-macro link between levels of analysis, it is apparent that, in contexts of sport performance, the interaction between performers (micro level of analysis) propels transitions between configurations of play (macro level of analysis) (Davids, Araújo & Shuttleworth, 2005). Remarkably, these transitions express emergent team decisions. Interpersonal dynamics directly influence the emergence of collective patterns of behaviour by destabilizing or (re)stabilizing the state of the system (Davids et al., 2005). Importantly the coupling between performers (the elements of the system) is based on information as well as by forces (Kugler & Turvey, 1987; Kelso & Engström, 2006). It is context dependency that guarantees stability of information over time. Therefore, spatial-temporal information can stabilize or destabilize the system (team-team) depending on the performance context.

Moreover, specific constraints like the players’ individual characteristics, a team’s traditions in a sport, strategy, coaches’ instructions, etc., may impact on the functional and goal-directed synergies formed by the players to shape the emergence of a particular performance behaviour. These informational constraints shape shared affordances available for perceptual systems, viewed as crucial for the assembly of synergies, that support the reduction of the number of independent degrees of freedom and enable fast, regulating actions (Riley et al., 2011). Another feature of a synergy is the ability of one of its components (e.g., a player) to lead changes in others (Riley et al., 2011). Thus, decision-making behaviours of the players forming a synergy should not be viewed as independent. In this context, social interpersonal synergies can be proposed to explain how multiple players can act in accordance with changing dynamic environments within fractions of a second. Therefore, the coupling of players’ degrees
of freedom into interpersonal synergies is based upon a social perception-action system that is supported by the perception of shared affordances (Silva et al., 2013). Next, we present some examples from previous research to illustrate these theoretical ideas.

Bourbousson and colleagues examined space-time patterns of basketball performers by decomposing their displacements into lateral (i.e., side-to-side) and longitudinal directions (i.e., basket-to-basket; see Bourbousson, Sève, & McGarry, 2010a). In this study, oscillatory movement patterns in both lateral and longitudinal directions were analyzed during performance in a 5v5 basketball game. Attackers and defenders in dyads exhibited stronger attraction to an in-phase coordination mode for longitudinal displacements, constrained by a personal marking system. In contrast to these results, in analysis of 5-vs-4+GK futsal game, Travassos and colleagues reported an in-phase attraction, in a lateral direction, but not longitudinal direction, of attacker-defender dyadic system displacements. Specific differences in the constraints of the game of futsal, in comparison to basketball, like the effects of a zonal marking defensive strategy, may explain these different results. This argument may also help us interpret the observation that ball dynamics, when attacking in futsal, demonstrated larger lateral displacement values than longitudinal displacements. Possibly, attackers’ decision-making behaviours were constrained to move the ball laterally in order to create possibilities to move into the opposition’s defensive area to create goal-scoring opportunities (Travassos, Araújo, Vilar, & McGarry, 2011).

Moreover, in 5vs4+GK Futsal subphases, Travassos, Araújo, Davids, Vilar, Esteves, et al. (2012) reported that defenders needed to continuously regulate their velocity in order to intercept opposition passes. Particularly, the values of the distance to the ball of the first defender and second defender, at the moment of pass initiation, constrained the success of ball interception. Similarly, in rugby union, passing
behaviours were investigated, using a spatiotemporal variable “tau”. This variable corresponds to changes in the value of the distance of the gap between an attacker and immediate defender (Correia et al., 2011). Results revealed that the initial value of tau was correlated with pass distance and pass duration, and consequently, with the type of pass performed (i.e., shorter and longer pass). This finding signifies that the decision on which pass to perform was constrained by the closing value of the distance to the defender.

Several attempts have been made to identify collective variables that capture the evolving dynamics of a whole team during performance in sport. For example, in soccer, values of the geometrical area of the team (i.e., mean positional coordinates of all players in each team), and the surface area of the team (i.e., polygonal covered area of each team), were analyzed in a four-a-side game of association football by selecting attacking situations that resulted in goal-scoring opportunities (Frencken, Lemmink, Delleman, & Visscher, 2011). Results showed a tendency for the geometrical centre of teams to overlap, accompanied by a sudden decrease in surface area of the attacking team or a decrease in surface area of the defensive team. It was also demonstrated that the crossing of centroids between teams, in a longitudinal direction, was related to emergence of goal scoring opportunities. A similar analysis was conducted on performance in 5v5 basketball games. Results indicated a stronger attraction to an in-phase coordination mode of the geometrical centre of both teams, for both lateral and longitudinal directions, with a higher stability in the longitudinal direction (Bourbousson, Sève, & McGarry, 2010b). In the same study, the stretch index, which measures the mean deviation of each player in a team from its spatial centre indicating the expansion and contraction in space of a team as the game unfolds, revealed an
attraction for in-phase coordination in a longitudinal, but not lateral, direction (Bourbousson et al., 2010b).

The interdependence between individual and team decision-making behaviours

Dynamical patterns observed in a game are created by the collective behaviours of the individual system elements. But at the same time, the behaviours of each individual element are constrained by the collective behaviours of the whole group. To understand how any self-organized pattern arises in a complex, dynamical system, one needs information about at least three levels (Kelso & Engström, 2006). At the “upper level”, the team, there are the constraints and parameters that act as boundary conditions on the potential patterns the system may produce. At the “lower level” are the individual elements (e.g., the players), each possessing his/her own intrinsic dynamics. At a middle level lie the patterns themselves (match sub-phases). Each level evolves on its own differing time scale, the upper level evolving most slowly and the bottom level evolving at the quickest rate due to quickly and unpredictably changing constraints between competing individuals. Under given performance constraints, a pattern is formed from the interactions among the elemental subsystems. Importantly, there is no ultimate preferred route or direction (top-down or bottom-up) to understanding behavioural patterns in team ball sports, because the idea is to view patterns as a complementary circular relation resulting from mutually reciprocating influences of different scales (both above and below). Interestingly, these principles of soft-assembled (or self-organized) multilevel dynamics in sports teams are in play as they are during individual performer-environment interactions (Hristovski, Davids, Araujo & Passos, 2011).

When individual parts and processes come together to form a team, novel patterns of behaviour arise. As a result of self-organization, the large number of
microscopic degrees of freedom is reduced to a much smaller set of relevant collective variables. Thus the information needed to describe the system’s behaviour is highly compressed. In order to capture the interdependence between individual and collective tactical behaviours in team sports, the cluster phase method was recently introduced in sports sciences (Duarte et al., 2013). The cluster phase method was developed to analyse synchrony within systems with a small number of oscillating components (Frank & Richardson, 2010). This method is based on the Kuramoto order parameter (Kuramoto & Nishikawa, 1987), which has been used to investigate phase synchronisation in systems with large number of oscillating components (e.g., the emergence of collective clapping in theatre audiences (Neda, Ravasz, Brechet, Vicsek, & Barabási, 2000). The Kuramoto model describes the collective synchronisation of oscillatory movement components (e.g., team players’ movement displacement trajectories) in a single collective variable. Frank and Richardson (2010) have adapted this model and showed its applicability using a rocking chair paradigm of only six oscillatory units (i.e., six individuals coordinating rocking chair movements). This method has also been used to assess the synchronisation dynamics emerging within association football teams during an entire match (Duarte et al., 2013). With this method, a specific measure of collective synchronisation – the cluster amplitude – is obtained, which captures how players collectively synchronise their on-field movement displacement trajectories. In a previous study, teams showed higher levels of collective synchrony in the longitudinal direction, than in the lateral direction of the field. Also, the relative phase of individual and whole group synchrony can be derived from this method, which can be a useful means to analyse the relationship between individual and collective (team) decision-making behaviours.
To illustrate how key performance constraints may influence decision-making behaviours at individual and team levels of analysis, it is useful to interpret data from an *in situ* experiment using small-sided games (Duarte, Travassos, Araújo, & Richardson, 2013). Twelve under-17 yrs elite footballers were divided into two teams, with a GK+1+3+1 diamond shape formation, with players assigned to their common positions (see Figure 1). Teams performed two small-sided games in a 40 x 42m of pitch size, in which the defensive playing method was manipulated. In the first condition/game, both teams used a zonal defence. In the second condition/game, the same players were paired and both teams used a personal marking defensive system. Positional data from the 10 outfield players were gathered using a Global Positioning System (GPSports SPI Elite system, GPSports, Canberra, Australia) with a sampling rate of 15 Hz (Gray, Jenkins, Andrews, Taaffe, & Glover, 2010).

![Figure 1](image_url)

Figure 1. Scheme of the GK+5vs5+GK small-sided games utilised in the field experiment reported in Duarte et al. (2013). In the first condition, teams used zone defence, while in the second condition they used man-to-man defence.

Time-series data of cluster amplitude (i.e., collective synchrony) are presented in the left panel of Figure 2. Mean data showed considerable higher values for zone
defence (0.63± 0.24) than for personal marking (0.47±0.21). These data indicated that zonal defending constrained players to collectively synchronise their on-field movement displacements trajectories in space and time. Acting according to a zonal defensive strategy seems to imply more cohesiveness within a team and to promote highly aggregated collective decision-making behaviours.

Figure 2. Data on the influence of varying the defensive playing method (zone defence vs. man-to-man defence): left panel shows the cluster amplitude of the collective synchrony of a team in the two conditions; right panel displays histograms of relative phase analyses between each individual player and whole group movement synchronisation. See text for further details.

Frequency histograms analyses presented in the right panel of Figure 2 also reveal important differences in the way individual players synchronised their movement displacement trajectories with the whole group. Adopting a personal marking defensive system implied less attraction for any specific mode of relation with greater dispersion within the entire spectrum of coordination possibilities. Also, we observed a slight shift for negative values compared with the zone defence condition, which suggested a lead-lag relationship attributed to the individual players’ movements. On the contrary, adopting a zonal defence strategy promoted higher attraction for relative phase values.
near in-phase mode of coordination (49.8% between -30 and 30 degrees). These data indicated that performance constraints imposed with a zonal defence influenced the players to be more synchronised with the collective movements of the team.

From a functional perspective, data showed that distinct performance constraints imposed on team players may change the emergent individual and team decision-making behaviours required to perform as a synchronised social unit. The relationship between individual and team decision-making behaviours revealed that, overall, by adopting zonal defence strategies, team players are more liable to achieve stable relationships with the collective behaviours of the whole team.

Applications

A founding statement of ecological dynamics approach to learning is that it can only occur through changes in properties to which players’ perceptual systems are attuned (Jacobs & Michaels, 2007). This means that, to improve decision-making behaviours, learning designs should manipulate those constraints, which induce changes in to-be-perceived properties or, in other words, the emergence of learning is coupled to the kind of informational (and other task) constraints present in a specific performance context (Kelso, 2008). The design of representative practice tasks requires an understanding of the interaction between key individual, task and environmental constraints of specific sports (Davids et al., 2012). Due to the characteristics of complex systems, there is no ideal decision that can be determined in advance (Passos et al., 2008). Through interactions with specific ecological constraints in learning environments, cognition, decision-making and actions emerge during performance. An important design strategy is the use of different types of variability (noise) to channel the learning process into meta-stable regions of the “learner–learning environment”
system to encourage adaptive behaviours. Here learners can be exposed to many functional and creative performance solutions during training. Functional variability in behaviours, invited by practice task design, is an important pedagogical principle, enabling players to become more adaptable, with concomitant repercussions at the individual, group and team levels of complexity.

Another important pedagogical principle is task simplification, where modified tasks in practice preserve the information-movement couplings used during performance (Renshaw, Chow, Davids, & Hammond, 2010). Key performance variables in team games such as the velocity of the displacement of the ball and opponents, number of players in the game and size of playing areas (Renshaw et al., 2009) are often manipulated to simplify a task for learners (e.g., in small-sided and conditioned games). Players’ decision-making behaviours are ruled by a nonlinear contextual dependence among neighbouring players, which affords new opportunities for behaviour when trying to adapt to upcoming circumstances (Davids et al., 2013). In practice sessions, making information available, or directing players’ attention to specific information sources relevant for governing interpersonal interactions, is a key feature of the learning process, particularly when system complexity increases, e.g., adding to the number of players involved, and manipulating time-to-act or space-to-act.

To promote this type of individual-to-team adaptations a nonlinear pedagogy (Chow, Davids, Hristovski, Araújo, & Passos, 2011) continues to be developed. Nonlinear pedagogy proposes that athletes exhibit purposive adaptive behaviours from the spontaneous patterns of interactions between system components. Nonlinear pedagogy emphasizes the manipulation of key task constraints (particularly informational constraints on action) during learning to allow functional movement behaviours to emerge in specific sports and physical activities. These manipulations
require skilful construction by pedagogists according to the theoretical principles of ecological dynamics, especially representative learning design.

One such principle is 'repeating without repetition' (Bernstein, 1967), where the learning system does not repeatedly practice the same movement pattern when seeking to resolve a specific performance problem (i.e., a particular situation of the game). Rather, this principle advocates that pedagogists induce a search in learners for new, functionally adaptive movement solutions for the same problem. The promotion of such functional variability is a crucial condition so that players can adapt to the unknown, emergent challenges of each, specific competition. During a match, teams have to make continuous adjustments to constraints such as the changing tactical landscape, stress, fatigue, weather conditions, injury or reduced performance of individual players, to maintain performance levels. The implication is that, rather than reducing variability, learning designs may increase the functional variability in practice conditions, in order to facilitate emergent decision-making behaviours and promote adaptability to changing constraints. At the same line, over-emphasizing prescription might lead learners to quickly stabilise their behaviours, but decreasing their ability to adapt to dynamic performance environments. By varying the constraints of the task, e.g., via small-sided and conditioned games, it is possible to influence key properties of the system, so that the individual-environment system produces different task solutions. The main concern and challenge when designing learning tasks is to ensure that task constraints are not so rigid that different possible behavioural solutions become severely reduced (Chow et al., 2011). Rather task constraints conducive to learning in sport should facilitate emergent behaviours at all times. In sum, to attain representative learning design, practitioners should design dynamic interventions that ensure that: (a) the degree of success of a performer’s actions are controlled for, and compared between contexts, and
(b), performers are able to achieve specific goals by actively exploring learning contexts designed to present affordances representative of those in the performance environment.
References


