Ad hoc procedure for optimising agreement between observational records

Javier Arana[1], Daniel Lapresa[2*], M. Teresa Anguera[3], y Belén Garzón[2]

[1] Universidad Internacional de La Rioja (UNIR) (España).
[2] Universidad de La Rioja (España).
[3] Universidad de Barcelona (España).

Título: Procedimiento *ad hoc* para optimizar el acuerdo entre registros observacionales.
Resumen: Habitualmente, los estudios observacionales desarrollados en el ámbito deportivo, se encuentran con la dificultad añadida que supone la elevada complejidad de las diferentes conductas que constituyen el registro, así como de la velocidad a la que éstas se suceden. El presente trabajo pretende satisfacer tres objetivos en relación con la fiabilidad de los datos observacionales. El primero consiste en implementar de forma conjunta en un mismo estudio, para facilitar así su comparación, formas de concordancia cuantitativa (coeficiente kappa de Cohen) y cualitativa (concordancia por consenso). El segundo persigue detallar el proceso de obtención del coeficiente kappa de Cohen, bajo el parámetro orden -mediante el software SDIS-GSEQ, versión 5.1.- con paquetes de datos constituidos por un número diferente de secuencias, incidiéndose en la forma en la que se garantiza su alineación. En el tercer objetivo, que persigue superar la confusión generada por dos registros diferentes, se presenta un procedimiento de mejora de la fiabilidad del dato. El procedimiento ha sido denominado "consultivo", al ser un nuevo observador el que determina qué conducta de las registradas es la que verdaderamente ha tenido lugar y, por lo tanto, procede incluir en el registro definitivo.
Palabras clave: Metodología observacional; fiabilidad; concordancia; kappa de Cohen; procedimiento consultivo.

Abstract: Observational studies in the field of sport are complicated by the added difficulty of having to analyse multiple, complex events or behaviours that may last just a fraction of a second. In this study, we analyse three aspects related to the reliability of data collected in such a study. The first aim was to analyse and compare the reliability of data sets assessed quantitatively (calculation of kappa statistic) and qualitatively (consensus agreement method). The second aim was to describe how, by ensuring the alignment of events, we calculated the kappa statistic for the order parameter using SDIS-GSEQ software (version 5.1) for data sets containing different numbers of sequences. The third objective was to describe a new consultative procedure designed to remove the confusion generated by discordant data sets and improve the reliability of the data. The procedure is called "consultative" because it involves the participation of a new observer who is responsible for consulting the existing observations and deciding on the definitive result.
Key words: Observational methodology, reliability, agreement, Cohen's kappa, consultative procedure.

Introduction

The extraordinary diversity of situations that can be systematically observed in sports and physical activity (Lapresa, Anguera, Alsasua, Arana & Garzón, 2013) requires the design of *ad hoc* observation tools suited to each behaviour of interest (Anguera, Blanco-Villaseñor, Losada & Hernández-Mendo, 2000). Garganta (2007) highlighted the difficulty of designing observation tools capable of capturing behaviours or events that are considered relevant yet may last only fractions of a second, and it is only logical that the reliability of data resulting from the observation of such fleeting events, which are common in the field of sport, has been questioned (O'Donoghue, 2007). In this article we address three issues related to the reliability of observational data.

First, as highlighted by Anguera and Hernández-Mendo (2013), the reliability of observational data can be estimated using agreement coefficients -proportion of concordant results- or correlation coefficients -measure of covariation- (Bland & Altman, 1986, 1990; Nevill & Atkinson, 1997). Reliability however, can also be measured qualitatively using the increasingly popular consensus agreement method (Anguera, 1990), designed to achieve agreement between observers based on the discussion of each code before it is recorded.

* Dirección para correspondencia [Correspondence address]:
Daniel Lapresa Ajamil. Universidad de la Rioja. Departamento de Ciencias de la Educación. Edificio Vives, C/ Luis de Ulloa s/n. 26004. Logroño, La Rioja (España). E-mail: daniel.lapresa@unirioja.es

The first aim of our study was to assess the reliability of sets of observational data both qualitatively and quantitatively, thereby facilitating the comparison of the two methods. For the quantitative analysis, we calculated Cohen's chance-corrected kappa statistic (1960) using SDIS-GSEG software, version 5.1 (Bakeman & Quera, 1992), while for the qualitative analysis, we used the consensus agreement method described by Anguera (1990).

Second, when a set of sequences of different units of behavior -for which there is no clear beginning or end- is independently coded by various observers or by the same observer at two different moments, the resulting number of sequences in each data block can vary, which is problematic as pairs of event sequences need to be aligned in order to calculate kappa (Quera, Bakeman, & Gnisci, 2007). While special alignment software programs exist for event sequences (Bakeman, 1978), such as Global Sequence Alignment (GSA) and Event Alignment (ELign) developed by Quera et al. (2007), they are not applicable to multievent data (Bakeman & Quera, 1995). The second aim of this study thus was to describe how we overcame the problem of misaligned data (soccer play sequences in our case) to calculate the kappa statistic using order as the primary parameter (Bakeman, McArthur & Quera, 1996).

Third, when researchers are faced with two different observations of the same event (from example, the area from which a shot was taken), they have no way of knowing with certainty which observation is correct. All they can be sure of on comparing the results is what each observer recorded.

Thus, the third objective of this study was to describe a new method of qualitative agreement that removes the confusion generated by two different interpretations. We call this new procedure "consultative", as it involves having a third expert observer analyse the event or behaviour in question, consult the conflicting observations, and produce what will be considered the definitive observation.

We addressed our three objectives in a study that analysed the movement of a football around a soccer field. We chose this setting as, according to Robinson and O'Donoghue (2007), events that occur within a playing area are commonly used as performance indicators in observational studies.

Method

We used an observational methodology (Anguera, 1979). The observational design (Anguera, Blanco-Villaseñor, Hernández-Mendo & Losada, 2011; Anguera, Blanco-Villaseñor & Losada, 2001) was point (three matches) with intrassesional follow-up (continuous monitoring of events throughout each observation session, or match) (Anguera & Hernández-Mendo, 2013); nomothetic (three teams of players aged 11-12); and multidimensional (different observation tool criteria). Observation was non-participatory and active, and the level of perceptibility was complete -direct observation- (Anguera et al., 2000).

Participants

The first three teams classified in the competition organized by a territorial soccer federation were intentionally selected. The players which comprise the teams are boys who turn 12 years old during this year. The study complied with the American Psychological Association's code of ethics and informed consent was obtained from all participants. A triangular tournament (Team A, Team B, Team C), in 7-a-side soccer discipline, was organised to perform the study. Each match lasted for 30 minutes and each minute of each match was analysed (90 minutes of play in total). Within-session consistency was guaranteed by establishing a list of minimum criteria to satisfy for all sessions observed.

Observation tool

We designed an observation tool to analyse the movement of the football around the playing field. It is a combination of field format and system of categories (Anguera, Magnusson & Jonsson, 2007; Herrero, 1992). Table 1 shows the criteria and categories used, and Figure 1 shows the zones into which the playing field was divided.

Recording and coding

The software SDIS-GSEQ was used to record and code the data with a purpose-designed template adapted to the

syntax rules and conventions of SDIS -Sequential Data Interchange Standard- (Bakeman & Quera, 2007); the template included all the observational criteria contemplated in the tool.

Table 1. Structure of observation tool

| No.	Criteria	Categories
1	Zone in which the play-sequence started (SZ)	SZ10, SZ20, SZ30, SZ40, SZ50, SZ60, SZ41, SZ51, SZ61, SZ70, SZ80, SZ90
2	Zone in which the play-sequence ended (EZ)	EZ10, EZ20, EZ30, EZ40, EZ50, EZ60, EZ41, EZ51, EZ61, EZ70, EZ80, EZ90
3	Zone in which the play-sequence continued (CZ)	Field format: referring to movement of ball around different zones of the field
4	Zone of momentary interception of the ball by an opponent (MIZ)	OI10, OI20, OI30, OI40, OI50, OI60, OI41, OI51, OI61, OI70, OI80, OI90
5	How the play-sequence started	Steal (STL); Goal kick (GK); Throw-in (TI); Corner kick (CK); Free kick (FK); Penalty (P); Kick-off (KO), Neutral bounce (NBS)
6	How the play-sequence ended	Interception (I); Off-side (OS); Ball out of play over goal line (OPGL); Foul (F); Ball out of play over touch line (OPTL); Neutral bounce (NBE)
7	Change of side (CS) (transfer of ball from one side of the field to the opposite)	Kick without the intervention of players in the centre corridor (CSK); without intervention of players in the centre corridor (CSC)
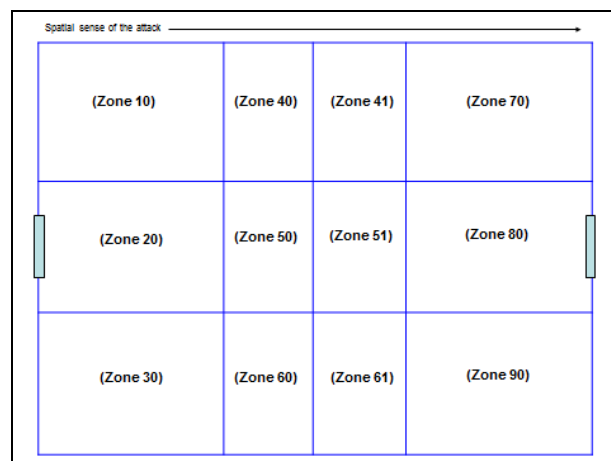
Figure 1. Playing field zones.

Based on Bakeman's (1978) definitions, the data were event-based and concurrent (type II). They were also multi-event (Bakeman & Quera, 1995), as the study design was multidimensional and the observation tool included field-format and category-based criteria.

Each team was observed separately, giving rise to two data blocks for each match analysed. A play sequence was considered to start or end when signaled by the referee or when

the team being analysed touched the ball at least twice in a row. A single touch by the team not being observed was interpreted as a momentary interception of the ball by an opponent (criterion 4) and therefore did not indicate the beginning or end of a play-sequence.

Procedure

Four observers were used. They were trained in three progressive stages using the approach described by Anguera (2003), which involves theoretical training followed by theoretical-practical training and practical training. The four observers were divided into three observation blocks, described below.

Data block 1

Three observers (none of whom are authors of this article) were used for data block 1 (B-1), which was used to analyse qualitative consensus agreement: different observers coming to an agreement on which observation to record (Anguera, 1990). The observers were informed that "agreement" referred to a decision taken on a given observation without strong objection from any of the observers. The observation group was provided with a set of guidelines designed to help them to resolve inevitable disagreements and overcome obstacles such as confrontation or submission that typically arise when different observers must agree on a result. The advice given was as follows: "Use logic", "Do not change your mind simply to avoid conflict", "Support ideas you partly agree with", "Consider differences in opinion to be positive (don't think in terms of win-lose)", "Argue your opinions", "Avoid strategies like: this time you, next time me", "Listen actively (i.e. listen and be heard)", "Do not criticize", "Make suggestions", "Remember that you are all equally responsible for the success of the process".

Data block 2

Data block 2 (B-2) was analysed by one of three observers from the first group. The observer was asked to re-analyse all the data from B-1 (the whole sample) and to record/code each action again. This step was designed to investigate quantitative agreement between the observations for B-1 and B-2.

Data block 3

A fourth observer (one of the authors of this article) was used for the third, and definitive, data block (B-3). This observer did not record all the data, just actions for which there were discrepancies between B-1 and B-2. Another of the authors of this article (who was responsible for training the observers as previously described) had provided this fourth observer with the video files and data sets containing the play sequences for which there were discrepancies. The agree-

ment procedure used is termed "consultative", as before generating the definitive data block, the observer consults the data produced by the other two observers and either choses one of these blocks or creates a new block of his own. The aim of this third step was to achieve qualitative consultative agreement on observations for which there were discrepancies. It also allowed us to obtain two quantitative coefficients of agreement: B1 vs B3 and B2 vs B3 (Figure 2).
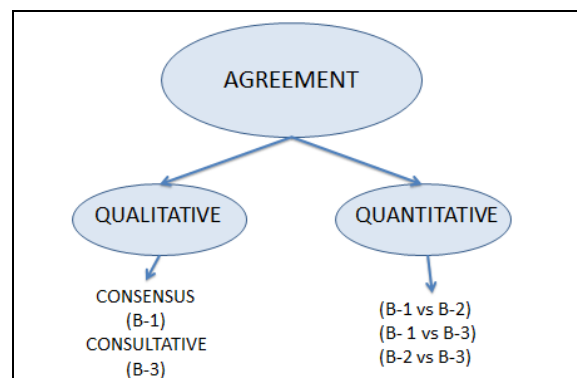

Figure 2. Structure of observation procedure showing data blocks

Analysis of agreement between data blocks

Reliability, in the form of agreement, was calculated quantitatively using Cohen's kappa statistic, which is widely used because of its satisfactory metric properties (Benavente, Ato & López, 2006). The kappa statistic is a measure of agreement for nominal classifications, with no ordering of data among the different categories. It is used to quantify inter-rater agreement and corrects for chance. It is expressed mathematically as $[\kappa = \sum_1^k \frac{(p_o - p_e)}{(1 - p_e)}]$, where p_o is the observed proportion of agreement, p_e the expected (chance) proportion of agreement, and k the different categories. Kappa was calculated using version 5.1 of the SDIS-GSEQ software program following the recommendations of Bakeman & Quera (1995, 2001, 2011).

The procedure designed to optimize the quality of data obtained using the observation tool built for the current study was divided into three phases, described in detail below.

a) Initial phase

In the initial phase, we compared observations from B-1 and B-2. All observations from each match/team observed were processed as a block of data. To calculate the kappa statistic for the whole block, we computed a weighted measure, taking into account the number of multi-events and the number of agreements.

b) Adjustment phase

The second stage was designed to reduce the effects that inter-sequence disagreements have on the calculation of kappa by match/team observed. These disagreements reduce the likelihood of a favorable agreement index, particularly when they involve sequences at the beginning of the data block. To reduce these effects, when a sequence was not rated by a particular observer, the data block was split at this point, allowing us to estimate agreement based on the same sequences (1 with 1, 2 with 2, etc.).

Next, we calculated a kappa statistic for each data block generated for each match/team observed, and then calculated a weighted measure to produce a single value for the whole block.

c) Modification phase

In the third phase, the modification phase, we searched for discrepancies affecting a particular action (intra-sequence disagreement) between B-1 and B-2. These were given to the fourth observer, whose job it was to choose the definitive observation for each case. This definitive data block (B-3) was used for the data analysis. To reduce observer-expectancy bias (Anguera, 1990), the fourth observer was unaware of which data belonged to B-1 and which belonged to B-2. He was additionally able to introduce changes in the definitive data block as he saw fit.

Results

Initial phase

Table 2 shows the kappa values for inter-block agreement (B-1 vs B-2) on the events for each match/team observed.

Table 2. Kappa statistics for the matches observed in the initial phase

Initial Phase	Match/Team	Kappa
	Team A_Team B	.5847
	Team A_Team C	.3523
	Team B_Team A	.9013
	Team B_Team C	.7037
	Team C_Team A	.4075
	Team C_Team B	.2983

To calculate the kappa statistic for the whole data block, we calculated the ratio between agreements and multi-events (no. of agreements / no. of multi-events) and the total ratio (Σ) (Table 3).

We then calculated the weight of each ratio (ratio / Σ ratios) and the weight of the kappa statistic for each data block (weight × kappa) (Table 4). We finally calculated the kappa statistic for the whole data block by adding the weighted values. The weighted kappa statistic for agreement between B-1 and B-2 in the initial phase was 0.5601.

Table 3. No. Agreements / No. Events or Multi-events and Total Ratio (Σ)

Match/Team	N° Multi-events	N° Agreements	N° Agreements / N° Multi-Events
Team A_Team B	324	277	.8549
Team A_Team C	346	275	.7948
Team B_Team A	310	305	.9839
Team B_Team C	388	361	.9304
Team C_Team A	361	291	.8061
Team C_Team B	342	264	.7719
			Σ = 5.1420

Table 4. Weight x Kappa

Match/Team	Weight (Ratio / Σ Ratios)	Kappa	(Weight x Kappa)
Team A_Team B	.1663	.5847	.0972
Team A_Team C	.1546	.3523	.0545
Team B_Team A	.1913	.9013	.1725
Team B_Team C	.1809	.7037	.1273
Team C_Team A	.1568	.4075	.0639
Team C_Team B	.1501	. 2983	.0448
			.5601

Adjustment phase

We calculated a kappa statistic for each of the blocks of data generated in the adjustment phase to ensure matching of comparable sequences in each of the matches/teams observed (Table 5).

Table 5. Kappa statistics for agreement between B-1 and B-2 by data block generated

Adjustment Phase	Match/Team	Kappa
	Team A_Team B	.8488
	Team A_Team B	.7857
	Team A_Team C	.8751
	Team A_Team C	.8516
	Team A_Team C	.9316
	Team A_Team C	.8305
	Team B_Team A	.9013
	Team B_Team C	.8699
	Team B_Team C	.9643
	Team B_Team C	.7946
	Team C_Team A	.8075
	Team C_Team A	.6263
	Team C_Team A	.7581
	Team C_Team B	.7997
	Team C_Team B	.8524
	Team C_Team B	.9133
	Team C_Team B	.8895

Using the same method as that described in the initial phase, we obtained a single weighted kappa value of 0.8420 for agreement between B-1 and B-2 in the adjustment phase.

Modification phase

Tables 6 and 7, respectively, show the results obtained on comparing B-3 (the definitive set of observations produced

by the fourth observer) with B-1 and with B-2 for each of the data blocks generated.

Table 6. Kappa statistics for agreement between B-1 and B-3 by data block generated

Modification Phase	Match/Team	Kappa
	Team A_Team B	.8997
	Team A_Team B	.8953
	Team A_Team C	.8517
	Team A_Team C	.8959
	Team A_Team C	.8991
	Team A_Team C	.9316
	Team B_Team A	1.0000
	Team B_Team C	.9867
Observer 1	Team B_Team C	.9043
	Team B_Team C	.9762
	Team C_Team A	1.0000
	Team C_Team A	.9484
	Team C_Team A	.7675
	Team C_Team B	.9529
	Team C_Team B	.9756
	Team C_Team B	.9539
	Team C_Team B	.9796

Table 7. Kappa statistics for agreement between B-2 and B-3 by data block generated.

Modification Phase	Match/Team	Kappa
	Team A_Team B	.8953
	Team A_Team B	.9356
	Team A_Team C	.9800
	Team A_Team C	.9533
	Team A_Team C	.9316
	Team A_Team C	1.0000
	Team B_Team A	.9115
	Team B_Team C	.9659
Observer 2	Team B_Team C	.9886
	Team B_Team C	.7956
	Team C_Team A	.8395
	Team C_Team A	.8605
	Team C_Team A	.8098
	Team C_Team B	.8246
	Team C_Team B	.9006
	Team C_Team B	.9350
	Team C_Team B	.8618

The resulting kappa statistics were 0.9309 for B-1 vs B-3 and 0.9056 for B-2 vs B-3, indicating very good agreement according to the criteria of Landis and Koch (1977) and Altman (1991). Table 8 shows the kappa statistics calculated for whole data block by phase and modality.

Table 8. Kappa statistics by phase and modality.

Initial Phase	Adjustment Phase	Modification Phase	
B-1 and B-2	B-1 and B-2	B-1 and B-3	B-2 and B-3
.5601	.8420	.9309	.9056

Discussion and Conclusions

As highlighted by Anguera and Hernández-Mendo (2013), once a set of data has been gathered, the observer needs to be confident of its quality (Anguera, 2003). Reliability is one of the most important indicators of the quality of a set of observational data. We will discuss the results of our study in the light of the three objectives established, each of which addressed a different aspect of data reliability.

We showed a high level of agreement between the data recorded by different observers in both our quantitative and qualitative assessment of reliability. For the quantitative analysis, we calculated the kappa statistic using SDIS-GSEQ, a program that has been widely used in the scientific literature for this purpose (Castellano, Hernández-Mendo, Morales & Anguera, 2007; Hernández-Mendo & Anguera, 2002; Lapresa, Arana, Anguera & Garzón, 2013; Lapresa, Ibañez, Arana, Garzón & Amatria, 2011). For the qualitative assessment, we analysed consensus agreement (Anguera, 1990), a measure that is gaining increasing acceptance in observational methodology. The consensus agreement approach has several advantages: not only does it produce a single observation but it also uses a strong observation tool to which clear definitions and additional information can be added as needed. It does, however, have some drawbacks, including the risk of confrontation during the consensus-building phase or compliant attitudes triggered by the presence of an "expert" in the group. It is noteworthy that in the modification phase, the kappa statistic obtained for B-1 vs B-3 was higher than that obtained for B-2 vs B-3, providing further evidence that the qualitative consensus agreement approach introduced by Anguera (1990) is an effective tool for improving the quality of data.

Our second objective was to describe how to calculate kappa, focusing on the order parameter (Anguera, 1989), for data sets containing a different number of sequences. As can be seen, the kappa values obtained in the initial phase (before alignment) indicated very poor agreement between the data sets. The application of our adjustment procedure, however, ensured alignment of sequences (Bakeman et al., 1996; Quera et al., 2007) and the kappa value for agreement between data packages B-1 and B-2 increased from 0.5601 in the initial phase to 0.8420 in the adjustment phase.

With regards to our third objective, our *ad hoc* consultative procedure achieved its goal: to guarantee high-quality data from an observation exercise. The consultative phase designed to improve the reliability of the data by producing a definitive set of observations (B-3) has three clear advantages:

1) It acts as an additional filter and therefore contributes to improving the quality of the data.
2) The definitive block can be used to assess inter-block agreement at additional levels (in our case, B-1 vs B-3 and B-2 vs B-3).
3) It saves time and effort as the observer responsible for producing the definitive set of observations only has to

analyse events on which there was disagreement. The cost of implementing a consultative procedure is not excessive considering the serious challenges that face researchers analysing multievent data sets (Bakeman & Quera, 1995) containing discordant or misaligned units. We consider that our procedure therefore has not only a positive cost-benefit ratio but also a positive cost-utility ratio, as it is likely to improve researchers' satisfaction considerably by providing a rigorous solution to the problem of discordant data sets.

The limitations of this study concern the first objective, i.e. they are related to the difficulty associated with generalizing the coefficients of agreement obtained in the modification phase. In brief, there was greater agreement between the

data block generated using the consensus agreement method (B-1) and the definitive data block (B-3) than between the data block generated by a single observer (B-2) and the definitive block. The procedure described in relation to the other two objectives of our study is easy to implement and provides a viable solution to an unresolved problem facing many researchers: the lack of agreement between data sets.

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