INTERPERSONAL EMOTIONS IN WOMEN'S FOOTBALL 1

Please cite as: Rumbold, J. L., Oliver, K-L., Madigan, D. J., Newman, J. A., Hobson, J. A., & Higham, A. 1 J. (in press, accepted 26 February 2025). Assessing the relationship between pre- and post-game interpersonal emotions in women's football teams. International Journal of Sport and Exercise Psychology. 2 https://doi.org/10.1080/1612197X.2025.2477167 3 Assessing the Relationship Between Pre- and Post-Game Interpersonal Emotions in 4 Women's Football Teams 5 James L. Rumbold, 6 Sheffield Hallam University, UK¹ Kelsey-Lee Oliver, 7 8 Sheffield Hallam University, UK 9 Daniel J. Madigan, York St John University, UK 10 11 James A. Newman, 12 Sheffield Hallam University, UK 13 Jennifer A. Hobson, Sheffield Hallam University, UK² 14 15 & 16 Andrew J. Higham, Sheffield Hallam University, UK³ 17 18 ¹ James L. Rumbold is now affiliated to the University of Derby, UK. ² Jennifer A. Hobson is now affiliated to Staffordshire University, UK. 19 20 ³ Andrew J. Higham is now affiliated to Heriot-Watt University, UK. 21 Corresponding Author: Dr James L. Rumbold, School of Sport and Exercise Science, College of Science and Engineering, University of Derby, Kedleston Road, Derby, Derbyshire, DE22 22 23 1GB, United Kingdom. Email address: J.Rumbold@derby.ac.uk; Twitter: @jameslrumbold; 24 LinkedIn: www.linkedin.com/in/james-rumbold-phd/

- 25 For the purpose of Open Access, the authors have applied a Creative Commons
- 26 Attribution (CC BY) licence to any Author Accepted Manuscript version arising from
- 27 this submission.

28

29

Abstract

30 Researchers have identified that sport emotions are interpersonal and can be transferred 31 between a team and its members. However, studies examining the transfer of emotions across 32 different phases of competition are limited. Consequently, the present study examined the cross-sectional, autoregressive (stability), and cross-lagged (bidirectional) relationships 33 34 between collective and group-based emotions over three consecutive football matches, whilst controlling for the performance outcome. Competitive female football players (N = 47, $M_{age} =$ 35 36 20.06 years; SD = 1.67) completed a sport emotion questionnaire before and immediately 37 after a match for three consecutive games. Players also completed a perfectionism towards 38 teammates questionnaire one week prior to data collection at football matches. Bayesian 39 dynamic structural equation modeling revealed that collective emotions were associated with 40 group-based emotions pre-game, but this was the case only for positive emotions. In addition, 41 perfectionism towards one's teammates was associated with group-based emotions at pre-42 game assessment. Emotions experienced at pre-game assessment were relatively stable at 43 post-game assessment. Finally, collective emotions at pre-game assessment predicted group-44 based emotions at post-game assessment. It would appear that while the performance 45 outcome strongly shapes players' group-based emotions following football matches, pregame collective emotions may offer earlier indications of the likely intensity of an 46 47 individual's group-based emotional response post-game; particularly when those emotions 48 are negative.

49

9 *Keywords*: Bayes; collective emotions; ecological momentary assessment;

50 group-based emotions; perfectionism

5	1

52

Assessing the Relationship Between Pre- and Post-Game Interpersonal Emotions in Women's Football Teams

53 Emotions in sport are inherently interpersonal (Tamminen et al., 2016). This is 54 because emotions are a consequence of interactions with various stakeholders (e.g., 55 teammates, coaches, parents, opposition) and, both an individual and collective response to 56 sporting events (Campo et al., 2019; Tamminen et al., 2024). Emotions that are formed 57 through any process of emotional exchange between individuals belonging to a group have 58 broadly been described as emotional dynamics (Smith & Mackie, 2016), or interpersonal 59 emotions. Interest in interpersonal emotions has increased in recent times, with sport researchers examining three distinct but interrelated interpersonal experiences: (a) how an 60 61 individual's emotions are experienced in response to sporting events that have relevance for a 62 team in which one identifies (i.e., group-based emotions; e.g., Rumbold et al., 2022), (b) how 63 teams converge on the same emotional responses to sporting events (i.e., collective emotions; e.g., Freemantle et al., 2022), and (c) the processes by which individuals recognise and mimic 64 65 the emotional expressions of others in teams (i.e., emotional contagion susceptibility; e.g., 66 Cotterill et al., 2020). Examining interpersonal emotions within sport dyads, teams, and 67 organisations is an important applied research endeavour, given the range of functions they serve. For example, the social sharing of positive emotions in groups can strengthen empathic 68 69 understanding, team integration, and team goals to enhance performance (Rimé, 2009). 70 Moreover, the social sharing of negative emotions can facilitate sense making and attempts at 71 emotional recovery for the sharer, through venting or social validation (Ma et al., 2024). 72 Shared positive and negative emotions can also provide an opportunity to reinforce groups' 73 identities (Goldenberg et al., 2020), regulate intergroup conflict (Halperin, 2014) and 74 strengthen group bonds (Wagstaff & Tamminen, 2021).

75

Despite a range of studies that have examined how different interpersonal emotion

76 experiences may operate in sporting dyads (e.g., Freemantle et al., 2022; Fritsch et al., 2024; 77 Stebbings et al., 2016) and teams (Cotterill et al., 2020; Van Kleef et al., 2019; Wergin et al., 78 2024), most studies have examined these interrelated interpersonal experiences (i.e., group-79 based emotions, collective emotions, emotional contagion) separately from one another, 80 rather than examining how they may influence and be affected by each other in varying 81 competition environments (Rumbold et al., 2022). In addition, given that emotions are 82 momentary responses to social interactions and events, it is surprising that there are limited 83 studies that have examined the relationships between interpersonal emotion experiences 84 temporally between competition phases (e.g., see Freemantle et al., 2022; Totterdell, 2000; Van Kleef et al., 2019). In the present study, we investigate how two interpersonal emotion 85 86 experiences (i.e., collective emotions and group-based emotions) interrelate during two 87 different phases of competition (pre- and post-competition).

88 Group-Based and Collective Emotions

89 Group-based emotions are individual-level emotions that occur in response to events, 90 that have perceived relevance for an individual's social group in which they identify as being 91 a member (Goldenberg et al., 2016). In this way, group-based emotions in sport are different 92 to individual emotions insofar that a person appraises events encountered based on their 93 social identity with a group (Campo et al., 2019). An example of individual emotions may 94 include a football player experiencing negative emotions (e.g., dejection) when he/she misses 95 a penalty. In contrast, group-based emotions may include a football player experiencing 96 dejection regarding the team, when the team have lost a penalty shootout.

97 Collective emotions represent macro-level group-based emotions that emerge from 98 interactions among group members who are feeling and responding to the same situation in 99 the same way, and at the same time (e.g., a football team experiencing dejection in relation to 100 their team after losing a penalty shootout; Goldenberg et al., 2020). As such, collective 101 emotions are different to group-based emotions. Group-based emotions reflect an individual's 102 emotional response in relation to their team following an event. In comparison, collective 103 emotions reflect a social group's emotional response in relation to their team following an 104 event (Goldenberg et al., 2014, 2020). In one of the first studies to illustrate collective 105 emotions in sport, Totterdell (2000) found that happy moods of individual cricket players 106 were linked to the team's average level of happiness during competition. More recently, 107 Freemantle et al. (2022) found that in table tennis dyads, collective within-dyad happiness, 108 dejection and anger were evident immediately following a match. In a recent scoping review 109 on convergence of emotions in sport, it was concluded that convergence of positive emotions is generally facilitative for performance, whilst convergence of negative emotions during 110 111 sporting events could be a maladaptive factor leading to team collapse (Fritsch et al., 2024). 112 Although collective emotions have been conceptualised by some researchers as a 113 convergence of individual emotions in response to an event (irrespective of individuals' 114 identity to their group; von Scheve & Ismer, 2013), we conceptualise collective emotions as 115 representing group-based emotions that are shared and felt concurrently by various members of a group that people identify with (Goldenberg et al., 2014, 2020). 116 117 From a theoretical perspective, social psychology researchers have provided explanations for how collective and group-based emotions may converge, based on conscious 118 119 (e.g., social identity, cognitive appraisal) and unconscious processes (mimicry, afferent 120 feedback) (Hatfield et al., 1994; Lazarus, 1991; Tajfel, 1982). The Emotions as Social 121 Information (EASI; Van Kleef, 2009) model blends these theoretical perspectives by

122 suggesting that collective emotional expressions regarding events provide relevant

123 information to group members, which may influence an individual's behaviour through

124 inferential processes (e.g., inferring and appraising emotional displays) and affective

125 reactions (mimicry, interpersonal liking). In addition, the strength with which inferential

126 processes and/or affective reactions may influence group-based emotions may depend on 127 information processing (e.g., the person's motivation and ability to process the information 128 from emotional expressions) or social-relational factors (e.g., the nature of group 129 relationships, emotional display rules of a group) (Van Kleef, 2009). 130 In demonstrating the relationship between collective and group-based emotions in 131 sport teams, Rumbold et al. (2022) found that for positive (e.g., excitement and happiness) 132 and negative emotions (e.g., anxiety, dejection and anger), collective emotions were strongly 133 associated with group-based emotions immediately following matches in male football teams, 134 irrespective of game outcome. In addition, the convergence between collective and group-135 based emotions post-game was more pronounced for negative emotions (e.g., dejection and 136 anger) than positive emotions (e.g., excitement, happiness). However, the relationship 137 between collective and group-based emotions was only assessed at post-game. Therefore, 138 research is needed to assess whether these significant relationships occur during other 139 temporal phases of competition when emotions may be particularly heightened (e.g., pre-140 game: Wolf et al., 2018), as this could have important implications for team performance 141 (Wergin et al., 2024), and future experiences of potentially unhelpful interpersonal emotions. 142 Moreover, controlling for performance outcome when assessing emotion convergence is 143 paramount to ensure that the relationship between individuals' emotional responses is not 144 simply due to individuals responding similarly to the same critical event (Fritsch et al., 2024; 145 Totterdell, 2000). In addition, recent research points to the performance outcome (especially when negative) being a trigger for worsening unhelpful emotional responses, which can be 146 147 transferred between team members (Wergin et al., 2024). 148 Hypothesis 1. Collective emotions will be associated with group-based emotions at

Hypothesis 1. Collective emotions will be associated with group-based emotions at
pre- and post-game competition phases.

150

In recent times, researchers have sought to determine whether achievement striving

151 dispositions may predispose teams to experience positive or negative emotions during phases 152 of competition. Indeed, emotional responses are believed to be the result of how individuals 153 appraise events in relation to their future achievement of goals (Lazarus, 1991). In addition, 154 according to Mackie et al. (2000), group-based emotions are influenced by a person's level of 155 identification with a social group and appraisals of events. Moreover, appraisals of events can 156 be determined by personality dispositions in addition to the type of event experienced (Ruiz et al., 2023). One achievement striving personality disposition that is relevant to influencing 157 158 sport emotions is perfectionism. Perfectionism has been described as a combination of 159 excessively high standards and a preoccupation with critical evaluations (Hewitt & Flett, 1991). In a series of studies in youth football teams, Donachie and colleagues (Donachie et 160 161 al., 2018, 2019) have shown that perfectionism predicts players' pre-competition negative 162 emotions (e.g., anxiety and anger). Although these studies have examined perfectionism and 163 emotions in the absence of perceived relevance for an individual's team, we believe that 164 assessing perfectionism in the context of one's teammates may provide some indication of an 165 individual's group-based emotions at different phases of a competition. Although our hypotheses are largely exploratory in this regard given the limited number of studies that 166 167 have assessed perfectionistic thoughts towards teammates (see Hill et al., 214), we contend that perfectionistic pressure on teammates could be associated with greater positive group-168 169 based emotions. Given the co-dependent nature of team sports, and the requirement for 170 shared expectations and aspirations, perfectionistic pressure on teammates may represent goal 171 strivings and subsequently positive feelings (e.g., happiness, excitement) about a team in 172 which an individual identifies. On the other hand, it is likely that negative reactions to 173 nonperfect performance of teammates could produce greater negative emotions (e.g., anger, 174 anxiety, dejection) about a team in which an individual identifies with. This is because 175 individuals with high perfectionistic concerns have a higher tendency towards being critical

and holding a sense of doubt over performances (Ruiz et al., 2023). This would coincide with
a large body work suggesting that highly critical negative reactions are largely maladaptive
for sport experiences (Hill et al., 2018).

Hypothesis 2. Individual's perfectionism towards teammates will be associated withgroup-based emotions.

181 It is generally accepted in psychology literature that emotions often involve short-term 182 responses to social exchanges and events that later fade away, or decline in intensity 183 (Goldenberg et al., 2016). However, in the context of collective and group-based emotions in 184 competitive sport environments, we believe that it might be possible for these interpersonal 185 emotions to remain relatively stable across phases of a competition (irrespective of game 186 outcome). This is because the EASI model (Van Kleef, 2009) posits that collective emotional 187 expressions provide social information which may influence future feelings and behaviours. 188 Future interpersonal group-based feelings may also be affected by individuals' understanding 189 of emotional display rules within a team. Furthermore, although social group identification is 190 an explanation for the generation of group-based emotions, it has also been argued that identification emerges as a by-product of collective emotions, which can then elicit new 191 192 collective emotions that help groups organise (Goldenberg et al., 2020). Limited research has 193 examined the transient nature of positive and negative collective (Freemantle et al., 2022) and 194 group-based emotions throughout different phases of competition. Research is therefore 195 warranted that assesses these variables in a time-lagged design, to evaluate whether 196 interpersonal emotions occur temporally.

Hypothesis 3. Collective and group-based emotions will show time-lagged (i.e.,
autoregressive) relationships between pre- and post-game competition phases.

Studies examining emotional dynamic experiences in sport have typically explored
the transfer of emotions as a unidirectional process from one interpersonal phenomenon (e.g.,

201 individual emotions) to another (e.g., collective emotions) (Fritsch et al., 2024; Moll et al., 202 2010; Totterdell, 2000; Van Kleef et al., 2019). For instance, in a multi field study conducted 203 by Van Kleef et al. (2019), coaches' happiness and anger predicted team sport performers' 204 happiness and anger before and during competition. However, there is likely to be degree of 205 mutual influence between interpersonal emotion experiences, such that previous positive and 206 negative collective emotions in response to events may elicit changes to an individual's 207 group-based emotions in the future, and vice-versa. This is because individuals in 208 relationships react to each other's actions and expressions and modify their interpersonal 209 behaviours in response (Pinus et al., 2025). As such, it is imperative that researchers consider 210 interpersonal emotion experiences in a bidirectional manner.

211 Hypothesis 4. Collective and group-based emotions will show bidirectional (i.e., 212 cross-lagged) relationships with one another between pre- and post-game competition phases. 213 In the current study, we make an original contribution to the interpersonal emotions in 214 sport literature in several ways. Firstly, using an experience sampling method (ESM; Hektner 215 et al., 2007) we investigate how positive and negative group-based and collective emotions interrelate at pre- and post-game competition phases. Secondly, we examine how individual 216 217 variability in positive and negative group-based emotions at different phases of competition 218 may be influenced by perfectionism dispositions regarding teammates. Thirdly, we examine 219 how the interrelationship between group-based and collective emotions may occur through 220 autoregressive and cross-lagged explanations. Finally, given that quantitative studies on 221 interpersonal emotions in sport to date have primarily researched male teams in isolation 222 (e.g., Freemantle et al., 2022; Moll et al., 2010; Rumbold et al., 2022; Totterdell, 2000), or 223 reported findings based on a small representation of females (e.g., Cotterill et al., 2020; Van 224 Kleef et al., 2019), we believe that quantitative studies examining interpersonal emotions in 225 female sport teams are lacking. Therefore, we examined the cross-sectional, autoregressive,

and cross-lagged relationships between positive and negative group-based and collectiveemotions in women's football teams over a series of competitive matches.

228

Materials and Methods

229 Participants and Procedure

The participants were 47 female football players ($M_{age} = 20.06$ years; SD = 1.67) who 230 231 played competitively for English university teams (n = 3). On average, players had trained 232 and competed for their university football teams for 1.55 years (SD = 1.66) and each team 233 was involved in a structured national university league competition. Aside from competing 234 for their university teams, 83% identified as playing for different teams across a range of levels within the English women's football pyramid. These included playing at club (55%), 235 236 county (2%), regional (4%) and national level (13%), or playing for Tier 2 (Women's 237 Championship, 4%) and Tier 3 (Women's National League, 4%) clubs. Seventeen percent of 238 the 47 football players (n = 8) identified as holding a leadership role (e.g., captain or vice-239 captain) in their university football team. Of the participants who provided information 240 pertaining to their indices of multiple deprivation (n = 43), 39.5% (n = 17) were raised by 241 families living in the top 30% of the least deprived areas of England, whilst 18.6% (n = 8) of 242 participants were raised by families living in the top 30% of the most deprived areas of 243 England. Following institutional ethics approval [University Ethics ID: ER39488057], 244 university football team coaches were approached by the second author who provided 245 coaches with a participation information sheet, inviting their players to take part in the study. 246 Female university players were then recruited via their respective coaches' request for 247 volunteers. For each of the three participating football teams, the full team roster was 248 represented by the participants. Each participant was provided with an information sheet 249 which outlined the aim of the study. Anonymity and confidentiality were assured, and players 250 were reminded of their right to withdraw from the study at any time.

251 Data were collected using paper questionnaires, which were distributed to participants 252 by the second author. Players first completed a background questionnaire one week prior to a 253 competitive game, providing demographic and sporting background information (e.g., age, 254 competitive standard, length of time playing for their clubs). During this time, participants 255 also completed a perfectionism questionnaire in relation to their teammates (cf., Stoeber et 256 al., 2006). Following completion of these questionnaire, players completed a sport emotion 257 questionnaire in their changing rooms 30 minutes prior to a competitive university league 258 game, and immediately after the game. This process was repeated for three games in total, 259 with one week separating each game. Regarding missing data, one player did not complete 260 the pre- or post-game emotion questionnaire at the second game for their team. This approach 261 was taken in line with ESM recommendations where event-contingent designs are employed 262 (e.g., see Rumbold et al., 2020; 2022). The changing room was used as the location for 263 collecting data on emotions, since measurement accuracy is enhanced and recall bias reduced 264 when measuring emotions as close as possible to the events when emotions are stimulated 265 (Hektner et al., 2007). In this context, the event or stimulus for triggering emotions was the team's preparation for a game (pre-game emotions) and the game outcome (post-game 266 267 emotions). Secondly, as the purpose of the study was to assess how collective team emotions may affect individual group-based emotions before and after football games (and vice-versa), 268 269 it was important for players to view and interpret the verbalised feelings and behaviours of 270 teammates in their natural environment (i.e., the team's changing room facility) (cf. 271 Csikszentmihalyi & Larson, 1987).

272 Measures

Group-based emotions. Participants' pre- and post-game emotions were assessed
using the 22-item Sport Emotion Questionnaire (SEQ, Jones et al., 2005). To measure
positive and negative group-based emotions before and after competitive games, players were

276	asked to indicate how they feel right now in relation to their team. The five subscales were
277	anxiety (5 items, pre $\alpha = .81$; post $\alpha = .89$), dejection (5 items, pre $\alpha = .79$; post $\alpha = .94$),
278	anger (4 items, pre $\alpha = .70$; post $\alpha = .94$), excitement (4 items, pre $\alpha = .78$; post $\alpha = .86$) and
279	happiness (4 items, pre $\alpha = .83$; post $\alpha = .94$). Previous research has demonstrated reliability
280	and validity for the SEQ (e.g., Arnold & Fletcher, 2015; Jones et al., 2005; Levillain et al.,
281	2025). Each participant's mean score for excitement and happiness was aggregated into a
282	mean score for positive group-based emotions. Similarly, mean scores for anxiety, dejection
283	and anger were aggregated into a mean score for negative group-based emotions. To
284	determine the amount of within and between team variance in positive and negative group-
285	based emotions, we calculated the intraclass correlations (ICCs) at each game time point. At
286	game time point 1, the ICCs were: pre-game negative group-based emotions = .01; post-game
287	negative group-based emotions = $.21$; pre-game positive group-based emotions = 0.07 ; and
288	post-game positive group-based emotions = .14. At game time point 2, the ICCs were: pre-
289	game negative group-based emotions = .00; post-game negative group-based emotions = .37;
290	pre-game positive group-based emotions = 0.01 ; and post-game positive group-based
291	emotions = .22. At game time point 3, the ICCs were: pre-game negative group-based
292	emotions = .02; post-game negative group-based emotions = .79; pre-game positive group-
293	based emotions = 0.07 ; and post-game positive group-based emotions = $.65$. Small ICC
294	values closer to zero suggest that within-team variance is much greater than between-team
295	variance, whereas larger ICCs provide empirical support for aggregation of within-person
296	data (group-based emotions) at the team level (collective group-based emotions) (Kenny &
297	La Voie, 1985).

298 Collective group-based emotions.¹ To compare each player's group-based emotion
299 scores to their team's collective emotion scores at each measurement timepoint, a team
300 aggregated mean score (excluding the individual player mean score this was being compared

301 to) was calculated for each of the three participating teams. This enabled the assessment of 302 emotion linkage between a player's group-based emotions and their team's collective 303 emotions about the team (Goldenberg et al., 2020) before and after three separate games. 304 **Covariates.** A selection of dispositional and situational variables were included as 305 time variant and invariant covariates. According to theories of emotion (Lazarus, 1991), 306 individual variability in emotional responses may be due to changes over time and can 307 typically be the result of interpreting and responding to an event. Therefore, the game outcome (e.g., 0 = 'no win', 1 = 'win') was dummy coded as a time variant covariate for each 308 309 of the three competitive games. Each of the three teams sampled had experienced a win and a loss across the three competitive matches.² Perfectionism was included as a time invariant 310 311 covariate, as the requirement for individuals to be perfect has been consistently linked to 312 explaining variability in performance-related emotions in sport (Donachie et al., 2018, 2019; 313 Ruiz et al., 2023). Given the team dynamic nature of the current study, perfectionism was 314 measured using two teammate-related subscales of the Multidimensional Inventory of 315 Perfectionism in Sport (MIPS; Stoeber et al., 2006). Perfectionistic pressure on teammates 316 (PPT; 8 items; $\alpha = .95$) measures an individual's pressure on their teammates to be perfect (e.g., "It is important to me that my teammates do everything perfectly"), whilst negative 317 318 reactions to nonperfect performance of teammates (NRNPT; 8 items; $\alpha = .95$) measures an individual's typical response when teammates do not meet their high expectations (e.g., "I get 319 320 annoyed with my teammates if their performance is not first class"). Participants rated to 321 what degree each statement characterised their perfectionistic attitudes towards their 322 teammates on a scale from 1 (strongly disagree) to 5 (strongly agree). Previous research has 323 demonstrated reliability for the PPT and NRNPT subscales (Stoeber et al., 2006). 324 **Data Analysis**

325

Bayesian dynamic structural equation modeling (DSEM) was utilised to examine the

326 cross-sectional, autoregressive, and cross-lagged relationships between group-based and 327 collective emotions across three competitive games. Bayesian DSEM is an appropriate 328 method to examine dynamic relationships between variables in small sample groups over 329 time (McNeish & Hamaker, 2020; Nelson et al., 2011). For readers interested in a deeper 330 understanding of Bayesian statistics, we refer readers to Zyphur and Oswald (2015) and Chen 331 et al. (2024) who introduce the foundations of Bayesian estimation and inference, and Myers 332 et al. (2018) and van de Schoot et al. (2014) who provide a helpful table outlining the key 333 differences between traditional frequentist and Bayesian principles. The main underlying 334 difference between Bayesian inference and frequentist approaches is how the probability of 335 something occurring is viewed and estimated. Bayesian inference interprets probability as a 336 subjective experience of uncertainty (akin to placing a bet on an event occuring), in 337 comparison to frequentist paradigms which employ infinitely repeating sampling of an event 338 (Slater, 2022). Secondly, in a frequentist paradigm, it is assumed that in the participant 339 population of interest there is only one true parameter (i.e., one true regression coefficient) 340 for a specific statistical relationship. With Bayesian analysis, all parameters are considered as uncertain and subsequently should be interpretated by way of a probability distribution for 341 342 each parameter (Chen et al., 2024).

All analyses were conducted using Mplus 7.0. First, we estimated a cross-sectional 343 344 model for group-based and collective emotions and an autoregressive model for the pre- and 345 post-game data. Separate models were estimated for positive and negative emotions. 346 Secondly, we added cross-lagged effects to assess whether pre-game group-based emotions 347 would predict post-game collective emotions, and whether pre-game collective emotions 348 would predict post-game group-based emotions. We then included time variant and invariant 349 covariates to the cross-lagged models. The game outcome for each of the three competitive 350 games was entered as a time variant covariate for post-game group-based emotions, and

perfectionism (PPT and NRNPT) was grand-mean centered as a time invariant covariate for
both pre- and post-game group-based emotions.

For both the autoregressive and cross-lagged models, we used Markov Chain Monte 353 354 Carlo simulation procedures with a Gibbs sampler and estimated the models with 50,000 355 iterations. Due to the difficulty of drawing on adequate informative priors from previous 356 research, we used the default uninformative prior distribution in Mplus. Uninformative priors 357 mimic frequentist maximum likelihood (ML) by estimating a likelihood for parameter 358 estimates based solely on the data collected (Ulitzsch et al., 2023). However, Bayesian 359 analysis differs to ML such that ML produces a single point estimate for each parameter, 360 whereas Bayesian produces a whole distribution of possible values for each parameter. This 361 distribution is known as the posterior probability distribution (McNeish & Hamaker, 2020). 362 In addition, Bayesian estimation produces posterior (probability) distributions for each 363 parameter estimate, and these posterior distributions form the subjective basis of a researcher's probability statement regarding the likelihood that a parameter estimate value is 364 365 likely given the dataset (Slater, 2022).

Posterior predictive p (PPp) and the 95% confidence interval were employed to 366 367 assess model fit. A low PPp value (e.g., < .05) closer to zero and a positive lower limit for the 95% credibility interval indicates a poor model fit (Winter & Depaoli, 2023). In contrast, 368 although there are no clear "cut-off" criteria for assessing adequate or good values, it is 369 370 generally accepted that PPp values around .50 indicate a well-fitting model (Chen et al., 371 2024; van de Schoot et al., 2014; Zyphur & Oswald, 2015). When comparing the 372 autoregressive and cross-lagged models we observed the deviance information criterion 373 (DIC) in which smaller DIC values indicate better fitting models. In addition, a potential scale reduction (PSR) factor of approximately 1 was considered as evidence of model 374 375 convergence (Zyphur & Oswald, 2015). For all parameter estimates, we observed the 95%

- 376 credibility interval ranges. In line with Zyphur and Oswald's (2015) recommendations, we
- rejected the null hypothesis if a moderate (e.g., > 70%) or large (e.g., > 90%) percentage of
- ach parameter's posterior distribution did not include zero.
- 379

Results

- Table 1 shows the means, standard deviations, reliabilities and correlations for group-based
 and collective emotions at pre- and post-game assessments, and covariates.
- **382 Positive Emotions**

383 Figure 1 illustrates the cross-sectional, autoregressive, and cross-lagged relationships 384 between positive group-based and collective emotions across three competitive games. Table 2 illustrates that the cross-lagged model with covariates represented an adequate fit to the 385 386 data (PPp = 0.31, 95% Confidence Interval [-22.29, 19.51], DIC = 1282.15).³ From this 387 model, the results provided strong evidence (probability = 0.996) that positive group-based 388 emotions were associated with positive collective emotions pre-game ($\beta = 0.31$). In addition, 389 there was strong evidence (probability = 0.898) that perfectionistic pressure on teammates 390 (PPT) was positively associated with positive group-based emotions pre-game ($\beta = 0.27$). 391 Conversely, negative reactions to nonperfect performances of teammates (NRNPT) was 392 inversely associated with positive group-based emotions pre-game ($\beta = -0.20$). The 393 probability that the 95% credible interval fell outside of zero was moderate at 78.4% (see 394 Table 3).

When assessing the relationship between pre- and post-game positive group-based emotions, there was strong evidence (probability = 0.969) of a positive association (β = 0.23). This finding indicates that positive group-based emotions were credible and stable between pre- and post-game assessments (see Figure 1). Moderate evidence (probability = 0.803) was found for a cross-lagged effect, such that positive collective emotions at pre-game inversely predicted positive group-based emotions post-game (β = -0.15) (see Table 3). Finally, there

401	was strong evidence (probability = 0.97	6-1.000) that wins wer	e associated with positive

402 group-based ($\beta = 0.40$) and collective emotions ($\beta = 0.88$) post-game.

403 We also explored the relationships between distinct positive interpersonal emotions 404 (i.e., happiness and excitement) in separate models. For happiness, a poorer model fit was

405 identified (PPp = 0.24, 95% Confidence Interval [-13.56, 28.60], DIC = 1368.66) in

406 comparison to the aggregated positive emotions model, and the findings were similar. For

407 excitement, an improved model fit was identified (PPp = 0.37, 95% Confidence Interval [-

408 17.74, 24.21], DIC = 1269.11). In addition, a stronger association was found between group-

409 based and collective emotions at pre-game assessment ($\beta = 0.40$), in comparison to the

410 aggregated positive emotions model ($\beta = 0.31$).

411 Negative Emotions

412 Figure 2 illustrates the cross-sectional, autoregressive, and cross-lagged relationships 413 between negative group-based and collective emotions across three competitive games. Table 414 2 shows that the cross-lagged model with covariates indicated a good fit to the data (PPp =415 0.56, 95% Confidence Interval [-15.88, 26.78], DIC = 790.23). From this model, there was 416 weak evidence of a relationship between negative group-based and collective emotions at 417 pre- ($\beta = 0.09$) or post-game ($\beta = 0.01$) assessments (probability = 0.492-0.663). On the other 418 hand, there was strong evidence (probability = 0.975) that NRNPT was associated with 419 negative group-based emotions pre-game ($\beta = 0.36$).

There was strong evidence (probability = 1.000) to suggest that both negative groupbased ($\beta = 0.48$) and collective emotions ($\beta = 0.60$) were stable between pre- and post-game assessments. In addition, strong evidence (probability = 0.918) was found for a positive cross-lagged effect, such that negative collective emotions at pre-game predicted negative group-based emotions post-game ($\beta = 0.29$). On the other hand, when assessing the reverse cross-lagged effect, there was moderate evidence (probability = 0.712) that negative groupbased emotions at pre-game predicted negative collective emotions post-game ($\beta = 0.11$) (see Table 3). Finally, there was strong evidence (probability = 0.991-1.000) that less wins were associated with higher negative group-based ($\beta = -0.52$) and negative collective emotions (β = -0.77) at post-game assessment.

430 We also explored the relationships between distinct negative interpersonal emotions 431 (i.e., anxiety, anger and dejection) in separate models. Model fit was not improved for any of 432 the specific emotion models. However, dejection demonstrated an adequate model fit (PPp =433 0.37, 95% Confidence Interval [-17.90, 24.64], DIC = 1228.15), and some findings were 434 markedly different to the aggregated negative emotions model. Firstly, there was strong evidence (probability = 1.000) to suggest that group-based dejection was associated with 435 436 collective dejection at post-game ($\beta = 0.59$), in comparison to the negative emotions model 437 which showed no relationship ($\beta = 0.01$). Secondly, the cross lagged effect whereby negative 438 collective emotions at pre-game predicted negative group-based emotions at post-game (β = 439 0.29) became non-probable when assessing the same cross-lagged effect for dejection only (β 440 = -0.04). Finally, when assessing dejection in isolation, we found that there was no evidence 441 of a statistical relationship between less wins and group-based dejection at post-game assessment ($\beta = -0.08$), in comparison to the aggregated negative emotions model ($\beta = -0.52$). 442

443

Discussion

This study examined the relationships between group-based and collective emotions in women's football teams. Using an Experience Sampling Method (ESM) to provide ecologically valid information at pre- and post-match assessments, we found partial support for Hypothesis 1, such that group-based emotions were associated with collective emotions pre-game, but only for positive emotions. These findings are supported by social functional theories of emotion (Van Kleef, 2009) which suggest that collective experiences regarding events provide social information to group members that can lead to convergence of emotions about one's group. It was surprising that the evidence linking group-based and collective
emotions at post-game assessment was weak, which contrasts previous findings with male
football teams (Rumbold et al., 2022).

454 One explanation might be the nature of events that are being appraised at pre- and 455 post-game assessments. In preparation for a game, players and teammates may be appraising 456 and responding to how they feel in anticipation of competing. As such, the social exchanges 457 and individual evaluations that occur in this context could lead to stronger convergence of 458 collective and group-based emotions (Goldenberg et al., 2020). In comparison, at post-game 459 assessment, there could be greater variability between team members in their group-based 460 emotions due to the game outcome (Fritsch et al., 2024). Players may appraise the importance 461 of the game outcome differently depending on whether the game outcome could harm the 462 achievement of individual and team goals, or harm how an individual socially identifies 463 within their team. Taken together, these findings highlight the importance of conducting 464 assessments of group-based and collective emotions across different phases of competition, 465 rather than assessing the convergence of emotions at single time points. Future studies could look to extend our ESM approach alongside a measurement of objective events encountered 466 467 and group-based cognitive appraisals.

We found partial support for Hypothesis 2, in so far that perfectionism dispositions 468 469 towards teammates was associated with pre-game emotions through PPT and NRNPT. 470 Although both perfectionism and emotions are group-referenced (i.e., the team) in this study, 471 the findings do provide support for previous research that has shown perfectionism to be 472 strongly associated with pre-competition emotions in youth footballers (Donachie et al., 473 2018, 2019). This can be explained by theories of appraisal and emotion (Lazarus, 1991) in 474 which emotional responses are largely influenced by the appraisals individuals make of events in relation to their goals, and personality dispositions can influence one's appraisals in 475

476 this regard (Mackie et al., 2000; Ruiz et al., 2023). In addition, given the requirement for 477 shared expectations and aspirations in the lead up to a competition, perfectionistic pressure on teammates could represent thoughts about team goal strivings and subsequently positive 478 479 feelings (e.g., happiness, excitement) about the team pre-competition. To our knowledge, 480 there are a limited number of studies that have utilised teammate-related measures of 481 perfectionism (e.g., see Hill et al., 2014). Therefore, this study offers some predictive validity 482 of the PPT and NRNPT subscales of the MIPS (Stoeber et al., 2006). Given that our findings 483 showed that NRNPT was showing signs of being related to post-game emotional responses, it 484 is possible that perfectionism dispositions towards teammates and the game outcome may have explained greater variance in group-based emotions post-game than collective emotions. 485 486 There are a couple of explanations for this. Firstly, perfectionism towards teammates 487 represents a stable disposition regarding how a person views others. Therefore, players 488 demonstrating a high degree of NRNPT are likely to hold these negative views for a 489 sustained period and draw on this tendency to be critical of others when opportunities arise 490 (following a win or loss). In comparison, collective emotions about the team are more 491 malleable to change in response to the events (win or loss) that influence how the collective 492 feels about the team in the moment. Negative reactions to the imperfection of teammates, 493 especially in the context of imperfect performance (particularly losses), will likely have 494 stronger implications for one's emotions, and certainly how one feels about others. There is a 495 growing body of evidence that other-oriented forms of perfectionism (i.e., a need for others to 496 be perfect rather than the self) are associated with angry reactions and antisocial behaviour 497 towards teammates (Grugan et al., 2020). Furthermore, there is a large body of evidence 498 showing that highly critical negative reactions to imperfection disrupt sport experiences (Hill 499 et al., 2018). This might be an interesting association for researchers to consider examining in 500 future assessments of sport teams' group-based emotions.

501 Limited research in sport has examined the episodic nature of group-based and 502 collective emotions throughout different phases of competition. Applying a time-lagged 503 approach this study makes an original contribution in supporting Hypothesis 3. Our study 504 makes a unique contribution to the interpersonal emotions in sport literature by showing that, 505 in three out of four autoregressive relationships examined, group-based and collective 506 emotions at pre-game assessment predicted group-based and collective emotions at post-507 game assessments, irrespective of perfectionism dispositions towards teammates and game 508 outcome. From a theoretical perspective (EASI: Van Kleef, 2009), this makes sense since 509 group-based and collective emotions that are identified prior to competition may positively 510 reinforce team identities (Mackie et al., 2000; Pinus et al., 2025) and future interpersonal 511 feelings about one's team (e.g., reciprocal liking), based on information processing (e.g., the 512 processing of emotional expressions) and social-relational factors (e.g., understood group emotional display rules). 513

514 To our knowledge, this is also the first study to examine collective and group-based 515 emotions in a cross-lagged manner across phases of sport competition. We found some initial 516 support for a cross-lagged relationship (hypothesis 4), insofar that negative collective 517 emotions pre-game predicted negative group-based emotions post-game. In addition, negative 518 group-based emotions pre-game showed some signs of credibly predicting negative collective 519 emotions following matches. These initial findings begin to answer questions around the bi-520 directional relationship of interpersonal emotions (Tamminen et al., 2024), alongside the 521 question of what causes collective emotions to occur in sport groups (Freemantle et al., 522 2022). Our cross-lagged findings illustrate that collective emotions can lead to later group-523 based emotions in competitive sport environments, but group-based emotions prior to a game 524 may go some way to predicting the collective emotions of female football teams following 525 matches. We recommend that researchers interested in examining emotional dynamics in

526 sport dyads and teams should explore these interpersonal phenomena together (i.e., rather 527 than in isolation from one another), to test these potential bi-directional relationships further. 528 Although our findings point to collective emotions being a stronger predictor of later group-529 based emotions (than vice-versa), the opposite could be just as apparent in contexts where 530 leaders' (e.g., team captains, coaches) expressions of emotions towards a team influence the 531 formation of stronger collective emotional responses (e.g., see Cotterill et al., 2020).

532 Limitations and Future Research

533 A strength of this study was the use of ESM to provide ecologically valid information 534 on transient group-based and collective emotions throughout phases of sport competition. 535 Whilst recognising the practical challenges, future research could attempt to follow the 536 methodological recommendations of Wagstaff and Tamminen (2021) regarding the need for 537 greater in-competition assessment when measuring sport emotions. We agree that this is an 538 important future research endeavour, but only as part of a wider range of ESM assessments of 539 interpersonal emotions across the training, competition, and sport team / organisational 540 environment. Indeed, it would be interesting to examine other time lagged approaches, such 541 as examining the influence of post-game interpersonal emotions on future interpersonal 542 emotions at subsequent competitions. Linked to the latter, case study time series designs 543 could be adopted to link episodic interpersonal emotions to future performance actions, to 544 identify which emotions are helpful or unhelpful for future performance for an individual 545 team. Regarding study limitations, we acknowledge that aggregating positive and negative 546 emotions doesn't tell researchers about the fluctuations in specific collective and group-based 547 emotions (e.g., happiness, anxiety), which is an approach researchers have employed when 548 examining these interpersonal emotions (e.g., Freemantle et al., 2022; Rumbold et al., 2022). 549 Our exploratory analysis of distinct emotions for group-based and collective emotions at pre-550 and post-game assessments does suggest that for some emotions (e.g., excitement, dejection),

551 the relationship between group-based and collective emotions could be more prominent. This 552 is worthy of further exploration in future research in sport and non-sport settings (Metzler et 553 al., 2023). In addition, from an applied perspective measuring fluctuations in the valence (i.e., 554 pleasantness, unpleasantness) of interpersonal emotions is perhaps more useful for intervention development in teams (see Pinus et al., 2025) than understanding which specific 555 556 emotions (e.g., happiness, anger, dejection) show convergence between individuals and their 557 team. Another consideration for future research could be to capture episodic collective 558 emotions differently, by examining the convergence of individuals' emotions following 559 events (von Scheve & Ismer, 2013), rather than the aggregation of a team's group-based emotions. 560

561 Although the sample size (N = 47) and number of observations across three 562 competitive matches (n = 6; df = 282) is not an issue for conducting complex Bayesian 563 DSEM (in comparison to traditional SEM), the impact of prior probability distributions (e.g., 564 informative, empirical, uninformative) on posterior distributions for parameter estimates can 565 be important for smaller sample sizes and can diminish as sample sizes increase (Chen et al., 2024). We used uninformative priors due to limited prior knowledge regarding pre- and post-566 567 game interpersonal emotions in women's football teams. Uninformative priors mimic frequentist maximum likelihood estimation, such that the estimation of posterior probabilities 568 569 for parameter estimates is dominated by the data only. Subsequently, the inclusion of prior 570 knowledge (e.g., informative priors) could help to strengthen the posterior probability 571 distributions (Winter & Depaoli, 2023). Extending our approach to a larger group of female 572 football teams, combined with greater observations across competitive match phases or 573 periods of the season may offer even greater insight into the cross-lagged effects of 574 interpersonal emotions, particularly if the current study data was used as informative priors to 575 strengthen the probabilistic interpretation of parameters.

576 Our findings do address Rumbold and colleagues' (2022) assertion regarding the 577 generalisability of group-based and collective emotion relationships in male team sport 578 populations. Specifically, we found that the relationship between group-based and collective 579 emotions post-game is different in female teams to what has currently been reported in male teams (see Rumbold et al., 2022). From an applied perspective, this may suggest that there 580 581 could be gender differences in how females and males appraise events in relation to their 582 team immediately following games. Moreover, the findings could hint at differences in how 583 females and males adhere to a different set of display rules for emotional expression in the 584 same sport team context.

585 Conclusion

586 In conclusion, we examined the cross-sectional, autoregressive, and cross-lagged 587 relationships between group-based and collective emotions in women's football teams over a 588 series of competitive matches. The findings provide support for social functional theories of 589 emotions in sport (Van Kleef, 2009) and further highlight that emotions in sport are a 590 consequence of social exchanges in relation to commonly experienced events. From an applied perspective, our findings suggest in the women's game, footballers should be wary of 591 592 the effect that collective displays of negative emotions can have for reinforcing their feelings 593 about the team, which may affect collective behaviour and team performance (Fritsch et al., 594 2024; Wergin et al., 2024). Collecting information on group-based and collective emotions in 595 a time series manner could serve as the basis for a series of team reflective and team re-596 appraisal exercises (e.g., see Pinus et al., 2025). Such reflective exercises could stimulate 597 team consensus on expectations regarding emotional display rules at different phases of 598 competition. They could also grow awareness for football players, their coaching and support 599 staff of how interpersonal emotions at different match phases may have sustained influence for future interpersonal emotions, and potentially performance. 600

- 601 Disclosure statement
- 602 The authors report no conflict of interest. 603
- 604 Data availability statement
- 605 The data and materials supporting the findings of this study are openly available on the open science framework 606 at https://doi.org/kkcd
- 607 608 Footnote
- 609 1. For brevity throughout the remainder of this paper, "collective group-based emotions" will heron be referred
- 610 to as "collective emotions".
- 611 2. In our analysis, we explored controlling for team membership in our cross-lagged model, but this negatively
- 612 affected the fit for the cross-lagged model, and there were no differences in the relationships assessed between 613 the original, better fitting model.
- 614
- 3. Despite this model appearing to demonstrate a poorer fit than the cross-lagged model without covariates 615 (PPp = 0.51, 95%) Confidence Interval [-14.31, 13.94, DIC = 840.66), we decided to report the cross-lagged
- 616 model with covariates due to the larger number of parameters in the model and due to the increase in r-squared
- 617 for emotion variables. We interpret the change in PPp and DIC from model 2 (the cross lagged model without
- 618 covariates) to model 3 (the cross-lagged model with covariates) for positive and negative emotions as a
- 619 consequence of the additional 11 parameter estimates being estimated, and the change in r-squared for positive
- 620 and negative emotions between models 2 and 3.

621	References
622	Arnold, R., & Fletcher, D. (2015). Confirmatory factor analysis of the Sport Emotion
623	Questionnaire in organisational environments. Journal of Sports Sciences, 33(2), 169-
624	179. https://doi.org/10.1080/02640414.2014.955520
625	Campo, M., Mackie, D. M., & Sanchez, X. (2019). Emotions in group sports: A narrative
626	review from a social identity perspective. Frontiers in Psychology, 10(666).
627	https://doi.org/gh8fq9
628	Chen, Q., Su, K., Feng, Y., Zhang, L., Ding, R., & Pan, J. (2024). A tutorial on Bayesian
629	structural equation modelling: Principles and applications. International Journal of
630	Psychology, 59(6), 1326-1346. https://doi.org/10.1002/ijop.13258
631	Cotterill, S. T., Clarkson, B. G., & Fransen, K. (2020). Gender differences in the perceived
632	impact that athlete leaders have on team member emotional states. Journal of Sports
633	Sciences, 38(10), 1181-1185. https://doi.org/ggqvz5
634	Csikszentmihalyi, M., & Larson, R. (1987). Validity and reliability of the experience-sampling
635	method. The Journal of Nervous and Mental Disease, 175, 526-536.
636	Donachie, T. C., Hill, A. P., & Hall, H. K. (2018). The relationship between multidimensional
637	perfectionism and pre-competition emotions of youth footballers. Psychology of Sport
638	and Exercise, 37, 33-42. https://doi.org/10.1016/j.psychsport.2018.04.002
639	Donachie, T. C., Hill, A. P., & Madigan, D. J. (2019). Perfectionism and precompetition
640	emotions in youth footballers: A three-wave longitudinal test of the mediating role of
641	perfectionistic cognitions. Journal of Sport and Exercise Psychology, 41(5), 309-319.
642	https://doi.org/10.1123/jsep.2018-0317
643	Freemantle, A. W. J., Stafford, L. D., Wagstaff, C. R. D., & Akehurst, L. (2022). Collective
644	emotions in doubles table tennis. Journal of Sport and Exercise Psychology, 44, 317-
645	326. https://doi.org/kjbk

- 646 Fritsch, J., Leis, O., Wohlfarth, D., Triller, N., & Jekauc, D. (2024). The role of emotional
- 647 convergence in sport: A scoping review. *Sport, Exercise, and Performance Psychology*,
 648 *13*(1), 39-58. <u>https://doi.org/10.1037/spy0000330</u>
- Goldenberg, A., Garcia, D., Halperin, E., & Gross, J. J. (2020). Collective emotions. *Current Directions in Psychological Science*, 29(2), 154-160. https://doi.org/gpzj7t
- 651 Goldenberg, A., Halperin, E., van Zomeren, M., & Gross, J. J. (2016). The process model of
- 652 group-based emotion: Integrating intergroup emotion and emotion regulation

653 perspectives. *Personality and Social Psychology Review*, 20, 118-141.

- 654 https://doi.org/f8h8rt
- 655 Goldenberg, A., Saguy, T., & Halperin, E. (2014). How group-based emotions are shaped by
- collective emotions: Evidence for emotional transfer and emotional burden. *Journal of Personality and Social Psychology*, *107*, 581-596. https://doi.org/f6xdb6
- Grugan, M. C., Jowett, G. E., Mallison-Howard, S. H., & Hall, H. K. (2020). The relationship
 between perfectionism, angry reactions, and antisocial behavior in team sport. *Sport*,
- 660 *Exercise, and Performance Psychology*, 9(4), 543-557. https://doi.org/f8ss
- Halperin E. (2014). Emotion, emotion regulation, and conflict resolution. *Emotion Review*, 6,
- 662 68-76. https://doi.org/ggwrmb
- Hatfield, E., Cacioppo, J., & Rapson, R. L. (1994). *Emotional contagion*. Cambridge
 University Press.
- Hektner, J. M., Schmidt, J. A., & Csikszentmihalyi, M. (2007). *Experience sampling method: Measuring the quality of everyday life*. Sage.
- 667 Hewitt, P. L., & Flett, G. L. (1991). Perfectionism in the self and social contexts:
- Conceptualization, assessment, and association with psychopathology. *Journal of Personality and Social Psychology*, 60, 456-470. https://doi.org/cw9
- Hill, A. P., Mallison-Howard, S. H., & Jowett, G. E. (2018). Multidimensional perfectionism in

INTERPERSONAL EMOTIONS IN WOMEN'S FOOTBALL 29

- 671 sport: A meta-analytical review. *Sport, Exercise, and Performance Psychology*, 7(3),
- 672 235-270. http://dx.doi.org/10.1037/spy0000125
- Jones, M. V., Lane, A. M., Bray, S. R., Uphill, M., & Catlin, J. (2005). Development and
- 674 validation of the Sport Emotion Questionnaire. *Journal of Sport & Exercise*
- 675 *Psychology*, 27(4), 407-431. https://doi.org/kjbn
- 676 Kenny, D. A., & La Voie, L. (1985). Separating individual and group effects. *Journal of*
- 677 *Personality and Social Psychology*, 48, 339-448. https://doi.org/d7b4j8
- 678 Lazarus, R. S. (1991). *Emotion and adaptation*. Oxford University Press.
- 679 Levillain, G., Martinent, G., Vacher, P., & Nicolas, M. (2025). Temporal invariance of Sport
- 680 Emotion Questionnaire scores using exploratory structural equation modelling.
- 681 *Movement & Sport Sciences*. Advance online publication.
- 682 https://doi.org/10.1051/sm/2025008
- Ma, G. W. S., Schöne, J. P., & Parkinson, B. (2024). Social sharing of emotion during the
- 684 collective crisis of COVID-19. *British Journal of Psychology*, *115*, 843-879.
- 685 https://doi.org/10.1111/bjop.12729
- Mackie, D. M., Devos, T., & Smith, E. R. (2000). Intergroup emotions: Explaining offensive
- 687 action tendencies in an intergroup context. *Journal of Personality and Social*
- 688 *Psychology*, 79, 602-616. https://doi.org/brkvz7
- 689 McNeish, D., & Hamaker, E. L. (2020). A primer on two-level dynamic structural equation
- 690 models for intensive longitudinal data in Mplus. Psychological Methods, 25(5), 610-
- 691 635. http://dx.doi.org/10.1037/met0000250
- 692 Metzler, H., Rimé, B., Pellert, M., Niederkrotenthaler, T., Di Natale, A., & Garcia, D. (2023).
- 693 Collective emotions during the COVID-19 outbreak. *Emotion*, 23(3), 844-858.
- 694 https://psycnet.apa.org/doi/10.1037/emo0001111
- Moll, T., Jordet, G., & Pepping, G-J. (2010). Emotion contagion in soccer penalty shootouts:

INTERPERSONAL EMOTIONS IN WOMEN'S FOOTBALL 30

- 696 Celebration of individual success is associated with ultimate team success. *Journal of* 697 *Sports Sciences*, 28(9), 983-992. https://doi.org/bzrwkv
- Myers, N. D., Ntoumanis, N., Gunnell, K. E., Gucciardi, D. F., & Lee, S. (2018). A review of
 some emergent quantitative analysis in sport and exercise psychology. *International Review of Sport and Exercise Psychology*, 11(1), 70-100. https://doi.org/gf4mgr
- 701 Nelson, T. D., Aylward, B. S., & Rausch, J. R. (2011). Dynamic P-technique for modeling
- patterns of data: Applications to pediatric psychology research. *Journal of Pediatric Psychology*, *36*, 959-968. https://doi.org/10.1093/jpepsy/jsr023

Pinus, M., Cao, Y., Halperin, E., Coman, A., Gross, J. J., & Goldenberg, A. (2025). Emotion

regulation contagion drives reduction in negative intergroup emotions. *Nature Communications*, 16: 1387. https://doi.org/10.1038/s41467-025-56538-x

Rimé, B. (2009). Emotion elicits the social sharing of emotion: Theory and empirical review.
 Emotion Review, 1(1), 60-85. https://doi.org/dbp5cq

709 Ruiz, M. C., Appleton, P. R., Duda, J. L., Bortoli, L., & Robazza, C. (2023). Perfectionism and

710 performance-related psychobiosocial states: The mediating role of competition

711 appraisals. *European Journal of Sport Science*, 23(5), 797-808. https://doi.org/n7xt

712 Rumbold, J. L., Fletcher, D., & Daniels, K. (2020). An experience sampling study of stress

713 processes and future playing time in professional sport. *Journal of Sports Sciences*,

714 38(5), 559-567. https://doi/.org/10.1080/02640414.2020.1717302

715 Rumbold, J. L., Newman, J. A., Foster, D., Rhind, D. J. A., Phoenix, J., & Hickey, L. (2022).

- 716 Assessing post-game emotions in soccer teams: The role of distinct emotional
- 717 dynamics. *European Journal of Sport Science*, 22(6), 888-896. https://doi.org/kdsx
- 718 Slater, M. (2022). Bayesian methods in statistics: From concepts to practice. Sage.

719 Smith, E. R., & Mackie, D. M. (2016). Group-level emotions. Current Opinion in Psychology,

720 *11*, 15-19. https://doi.org/gf3hb3

Stebbings, J., Taylor, I. M., & Spray, C. M. (2016). Interpersonal mechanisms explaining the
transfer of well- and ill-being in coach-athlete dyads. *Journal of Sport and Exercise*

723 *Psychology*, *38*(3), 292-304. https://doi.org/f88ks7

- Stoeber, J., Otto, K., & Stoll, O. (2006). *MIPS: Multidimensional Inventory of Perfectionism in Sport (English version)*. School of Psychology, University of Kent.
- Tajfel, H. (1982). Social psychology of intergroup relations. *Annual Review of Psychology*, *33*,
 1-9.
- 728 Tamminen, K. A., Palmateer, T. M., Denton, M., Sabiston, C., Crocker, P. R. E., Eys, M., &

Smith, B. (2016). Exploring emotions as social phenomena among Canadian varsity
athletes. *Psychology of Sport and Exercise*, 27, 28-38. https://doi.org/ggmwgh

731 Tamminen, K. A., Wolf, S. A., Dunn, R., & Bissett, J. E. (2024). A review of the interpersonal

experience, expression, and regulation of emotions in sport. *International Review of Sport and Exercise Psychology*, *17*(2), 1132-1169. https://doi.org/gq49nd

734 Totterdell, P. (2000). Catching moods and hitting runs: Mood linkage and subjective

- performance in professional sport teams. Journal of Applied Psychology, 85(6), 848-
- 736 859. <u>https://doi.org/crz45f</u>
- 737 Ulitzsch, E., Lüdtke, O., & Robitzsch, A. (2023). Alleviating estimating problems in small
- sample structural equation modeling A comparison of constrained maximum
- 739 likelihood, bayesian estimation, and fixed reliability approaches. *Psychological*

740 *Methods*, 28(3), 527-557. https://doi.org/10.1037/met0000435

- van de Schoot, R., Kaplan, D., Denissen, J., Asendorpf, J. B., Neyer, F. J., & van Aken, M. A.
- G. (2014). A gentle introduction to Bayesian analysis: Applications to developmental
 research. *Child Development*, *85*, 842-860. https://doi.org/f54sq7
- 744 Van Kleef, G. A. (2009). How emotions regulate social life. *Current Directions in*
- 745 *Psychological Science*, *18*(3), 184-188. https://doi.org/10.1111/j.1467-8721

- 746 Van Kleef, G. A., Cheshin, A., Koning, L. F., & Wolf, S. A. (2019). Emotional games: How
- 747 coaches' emotional expressions shape players' emotions, inferences, and team
- performance. *Psychology of Sport and Exercise*, 41(1), 1-11. https://doi.org/ggwq5j
- von Scheve, C., & Ismer, S. (2013). Towards a theory of collective emotions. *Emotion Review*,
 5(4), 406-413.
- Wagstaff, C. R. D., & Tamminen, K. A. (2021). Emotions. In R. Arnold & D. Fletcher (Eds.),
 Stress, well-being and performance in sport (pp. 97-113). Routledge.
- 753 Wergin, V. V., Wolf, S. A., Schwender, J. T., & Mallett. C. J. (2024). Experienced emotions,
- emotion regulation strategies, and regulation depletion in collective team collapse and
- performance recovery situations. *Journal of Applied Sport Psychology*, 1-22.
- 756 https://doi.org/10.1080/10413200.2024.2413996
- Winter, S. D., & Depaoli, S. (2023). Illustrating the value of prior predictive checking for
 bayesian structural equation modeling. *Structural Equation Modeling: A*

759 *Multidisciplinary Journal*, *30*(6), 1000-1021. https://doi.org/n7x9

- 760 Wolf, S. A., Harenberg, S., Tamminen, K., & Schmitz, H. (2018). "Cause you can't play this
- 761 by yourself": Athletes' perceptions of team influence on their precompetitive
- 762 psychological states. *Journal of Applied Sport Psychology*, *30*, 185-203.
- 763 https://doi.org/kjbm
- 764 Zyphur, M. J., & Oswald, F. L. (2015). Bayesian estimation and inference: A user's guide.
- 765 *Journal of Management*, 41(2), 390-420. https://doi.org/cbxb

	М	SD	α	1	2	3	4	5	6	7	8	9	10	11
1. Pre-game positive emotions	1.50	0.62	.88											
2. Pre-game collective positive emotions	1.50	0.28	-	.32										
3. Post-game positive emotions	1.32	0.96	.95	<u>.18</u>	04									
4. Post-game collective positive emotions	1.32	0.56	-	03	02	<u>.50</u>	—							
5. Pre-game negative emotions	0.65	0.39	.85	<u>.31</u>	<u>.18</u>	16	02							
6. Pre-game collective negative emotions	0.65	0.11	-	<u>.29</u>	<u>.75</u>	04	12	.05						
7. Post-game negative emotions	0.76	0.75	.95	.25	<u>.39</u>	<u>57</u>	<u>49</u>	.52	<u>.30</u>					
8. Post-game collective negative emotions	0.76	0.52	-	<u>.25</u>	<u>.57</u>	<u>41</u>	<u>75</u>	.12	<u>.58</u>	<u>.64</u>				
9. PPT	2.78	1.14	.95	.06	15	15	.04	.32	10	.11	09			
10. NRNPT	2.08	0.93	.95	05	<u>21</u>	<u>17</u>	.04	<u>.38</u>	12	.16	14	.81	_	
11. Win	0.56	0.50	-	01	.09	.44	<u>.71</u>	04	.12	<u>51</u>	<u>65</u>	.05	.08	_

Table 1. Means, standard deviations, internal consistencies, and correlations

Note. N = 47; N of observations = 140. Underlined values indicate significant correlations, p < .05. PPT = Perfectionistic pressure on teammates; NRNPT = Negative reactions to nonperfect performance of teammates.

Table 2. Model fit comparisons

Model	Parameters	PPp [95% CI]	DIC
Positive emotions			
Cross-sectional and autoregressive model	10	.51 [-14.60, 14.50]	838.66
Cross-lagged model	12	.51 [-14.31, 13.94]	840.66
Cross-lagged model with covariates	23	.31 [-22.29, 19.51]	1282.15
Negative emotions			
Cross-sectional and autoregressive model	10	.42 [-12.98, 14.73]	494.76
Cross-lagged model	12	.48 [-13.73, 15.24]	495.72
Cross-lagged model with covariates	23	.56 [-15.88, 26.78]	790.23

Note. PPp = Posterior predictive *p*; CI = 95% confidence intervals; DIC = Deviance information criteria

	Pos			3	Negative Emotions				
Parameter	2.5%	β	97.5%	P > 0	2.5%	β	97.5%	P > 0	
Pre-game group-based emotions									
Intercept	-0.16	0.70	1.61	1.000	0.15	1.08	2.02	1.000	
Pre-game collective emotions	0.16	0.31	0.45	0.996	-0.06	0.09	0.24	0.492	
PPT	-0.01	0.27	0.53	0.898	-0.23	0.03	0.29	0.486	
NRNPT	-0.46	-0.20	0.08	0.784	0.10	0.36	0.61	0.975	
R^2	0.05	0.14	0.25	0.796	0.07	0.17	0.29	0.912	
Post-game group-based emotions									
Intercept	-0.03	0.83	1.67	1.000	-1.77	-0.95	-0.08	0.983	
Pre-game group-based emotions	0.09	0.23	0.37	0.969	0.36	0.48	0.59	1.000	
Pre-game collective emotions	-0.28	-0.15	-0.01	0.803	0.01	0.29	0.56	0.918	
Post-game collective emotions	-0.15	0.14	0.43	0.671	-0.42	0.01	0.44	0.663	
Win	0.10	0.40	0.66	0.976	-0.86	-0.52	-0.17	0.991	
PPT	-0.28	-0.06	0.17	0.469	-0.27	-0.10	0.07	0.549	
NRNPT	-0.42	-0.19	0.04	0.796	-0.02	0.16	0.33	0.755	
R^2	0.28	0.40	0.51	1.000	0.56	0.65	0.73	1.000	
Post-game collective emotions									
Intercept	0.95	1.41	1.88	1.000	-1.63	-1.41	-1.18	1.000	
Pre-game collective emotions	-0.08	0.00	0.08	0.022	0.57	0.60	0.63	1.000	
Pre-game group-based emotions	-0.12	-0.03	0.05	0.068	0.07	0.11	0.15	0.712	
Win	0.85	0.88	0.91	1.000	-0.80	-0.77	-0.75	1.000	
R^2	0.73	0.79	0.83	1.000	0.93	0.95	0.96	1.000	
PPT with NRNPT	0.74	0.81	0.86	1.000	0.74	0.81	0.86	1.000	

Table 3. Parameter estimates and posterior distributions for positive and negative emotions.

Note. PPT = Perfectionistic pressure on teammates; NRNPT = Negative reactions to nonperfect performance of teammates; P > 0 = the posterior probability that the parameter estimate is greater than 0.

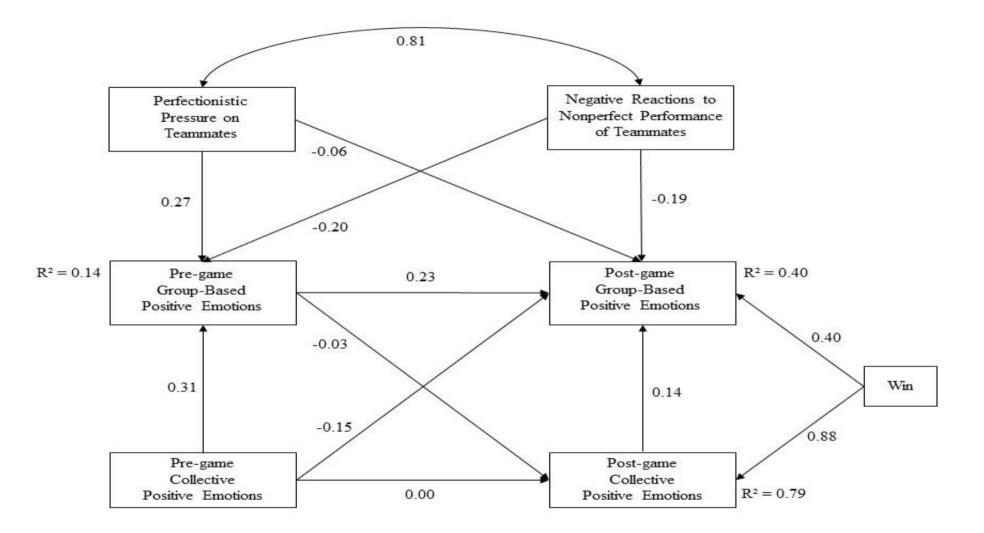


Figure 1. Cross-sectional, autoregressive and cross-lagged relationships between positively valanced pre- and post-game emotions. Standardized estimates are presented. PPp = 0.31, 95% Confidence Interval [-15.88, 26.78]

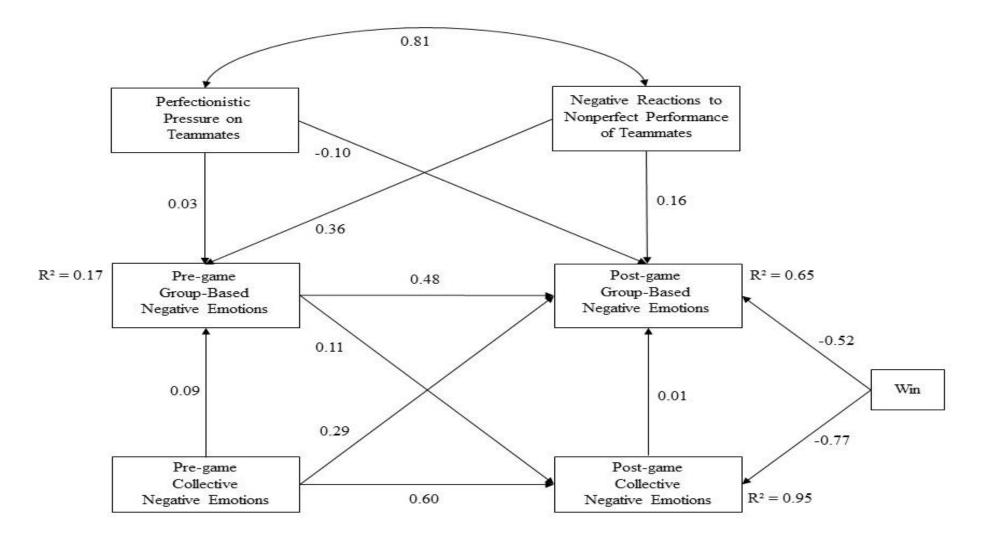


Figure 2. Cross-sectional, autoregressive and cross-lagged relationships between negatively valanced pre- and post-game emotions. Standardized estimates are presented. PPp = 0.56, 95% Confidence Interval [-22.29, 19.51]