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The Road to Victory in the UEFA Women's Champions League: A Multi-Level Analysis of Successful Coaches, Teams, and Countries

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8	The Road to Victory in the UEFA Women's Champions League:
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36	Abstract
37	Objectives: To explore coach-level, team-level, and country-level factors associated with
38	performance in the UEFA Women's Champions League.
39	Design: This study involved archival analysis of factual data on teams and coaches participating
40	in the UEFA Women's Champions League (2011-12 until 2015-16).
41	Method: Official data records were provided by UEFA. Hierarchical linear modeling analysis
42	was used to predict performance in the UEFA Women's Champions League. Specifically,
43	coaches' characteristics (level-1 variables), team factors (level-2 variables), and country
44	information (level-3 variables) were tested as predictors of performance (final rank, ranging from
45	1 to 32).
46	Results: Data analysis yielded a two- and three-level solution. The two-level solution was
47	deemed more realistic and applied, and was chosen as the omnibus final model. Within the two-
48	level solution, years coaching experience in Champions League at level-1 ($\gamma_{10} = -2.90$), and
49	number of times team has won Champions League (γ_{01} = -7.13) and number of international
50	<i>players</i> (γ_{02} = -1.08) at level-2, predict final performance at the UEFA Women's Champions
51	League (i.e., negative coefficient is indicative of performance improvement).
52	Conclusions: Our findings suggest that the quality of the team, positive cross-cultural effects
53	from an international roster, and the experience of the coach are positively associated with
54	performance in the UEFA Women's Champions League.
55	Keywords: coaching, expert performance, women's football, UEFA Champions League,
56	hierarchical linear modeling

57		Highlights
58	•	Over 85% of the coaches in the UEFA Women's Champions League are male.
59		
60	•	More experienced coaches are more likely to be successful or, alternatively, successful
61		coaches keep their jobs for a longer time.
62		
63	•	Successful teams at the UEFA Women's Champions League win because of the quality of
64		the "team as a whole".
65		
66	•	Internationalization is a good thing. For every international player on the team,
67		performance improved by about 1 position.
68		

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The Road to Victory in the UEFA Women's Champions League:

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A Multi-Level Analysis of Successful Coaches, Teams, and Countries

71 It is important to examine the profile of successful coaches as previous research has 72 suggested that coach behaviors influence team outcomes across domains of human performance 73 (Bloom, 1985; Côté & Gilbert, 2009). It is also imperative to consider the role of team factors, as 74 intra-team characteristics (e.g., number of international players) have been shown to influence 75 performance in sports (Filho, Gershgoren, Basevitch, & Tenenbaum, 2014). Broader 76 environmental factors, including country effects, have also been linked to expert performance in 77 sports (Salmela & Moraes, 2003). Given the paucity of research on the unique mechanisms of 78 expert performance in women's football, we aimed to examine the factors differentiating 79 successful coaches and teams from unsuccessful coaches and teams in the Union of European 80 Football Associations (UEFA) Women's Champions League, while accounting for the role of 81 country-level variables.

82 The uniqueness of the present study rests on examining excellence in women's 83 professional football, as most of the research in football has been focused on the men's game (for a review see Gledhill, Harwood, & Forsdyke, 2017). Previous research in sports has revealed 84 85 gender differences in team processes and performance (Carron, Colman, Wheeler, & Stevens, 2002; Duda, 1987; Eagly & Johnson, 1990; Filho, Tenenbaum, & Yang, 2015; Fransen, 86 87 Vanbeselaere, Cuyper, Vande Broek, & Boen, 2015; Leo, Sánchez-Miguel, Sánchez-Oliva, 88 Amado, & García-Calvo, 2013). Specifically regarding the UEFA Champions League, previous 89 research has revealed large gender differences in physical and technical match performance characteristics, such as distance covered by the players and pass completion rates (Bradley, 90 91 Dellal, Mohr, Castellano, & Wilkie, 2014).

92 We based our research on the notion that expert performance in sports is a multi-layered 93 phenomenon, and depends on inputs from various micro, meso, and macro levels of analysis 94 (Baker & Farrow, 2017). In fact, several frameworks have been proposed to explain consistent 95 superior performance in sports (i.e., *explanatory pluralism*; see Dale, Dietrich, & Chemero, 96 2009), with emphasis being placed on different variables and levels of analysis, including coach, 97 team, and contextual-level variables (for a review see Baker, Cobley, Schorer, & Wattie, 2017). 98 Notwithstanding, the various conceptual frameworks on expertise have been primarily informed 99 by two research frameworks, namely the *expert-novice paradigm* and the *expert performance* 100 approach (Filho & Tenenbaum, 2015). In the former, scholars seek to describe (descriptive 101 adequacy) variables related to expertise, whereas in the latter the focus is on identifying variables 102 that predict (explanatory adequacy) expert performance in sports. In the present study we sought 103 to both describe (through descriptive statistical analysis) and explain (through hierarchical linear 104 modeling) coach, team, and country-level factors related to performance at the UEFA Women's 105 Champions League.

106 Characteristics of Successful Coaches

107 Coach-level characteristics consider any variable related to the coach that may influence 108 sports performance, positively or negatively. To this extent, there is consensus that coaches with 109 different background characteristics (e.g., *age*, *nationality status*, *international playing* 110 *experience*) are more or less likely to be successful in sports (Côté & Gilbert, 2009; Gilbert, Côté 111 & Mallett, 2006; Starkes & Ericsson, 2003). Previous research suggests that expert coaches have 112 athletic experience, participate in formal and informal educational programs, and have extensive 113 coaching experience (Erickson, Côté, & Fraser-Thomas, 2007).

114 Previous research has also shown that competing in elite sport is an important

denominator of successful coaches (e.g., Cregan, Bloom, & Reid, 2007; Nash & Sproule 2009;
Schinke, Bloom, & Salmela, 1995). For instance, Gilbert and colleagues (2006) found that
successful coaches in interactive sports viewed themselves as high-performing athletes during
their playing careers. Overall, successful coaches tend to perceive themselves as "better than
average" athletes during their competitive careers (Gilbert, Lichtenwaldt, Gilbert, Zelezny, &
Côté, 2009).

121 In addition to playing experience, formal education in a sport-related domain (e.g., 122 Exercise Physiology, Physical Education, Sport Biomechanics, Sport Psychology), as well as 123 informal education opportunities, such as mentorship programs and networking with other 124 coaches, are important elements of successful coaches (Martens, 2012). Anderson and Gill 125 (1983) found that many expert coaches acquired their initial coaching knowledge while enrolled 126 in an undergraduate Physical Education degree. Interviews with high-performance coaches 127 across a range of team and individual sports revealed the importance of studying Physical 128 Education or Kinesiology and participating in formal coaching education courses (Carter & 129 Bloom, 2009; Erickson, Côté, & Fraser-Thomas, 2007).

130 Coaching experience also plays an important role in coaching development and expertise (Cregan et al., 2007; Schinke et al., 1995). In interviews with Olympic coaches, coaching 131 132 experience at the national and international level was the most frequently cited variable in 133 preparation to become an elite coach (Gould, Giannini, Krane, & Hodge, 1990; Gould, Hodge, 134 Peterson, & Giannini, 1989). Expert basketball coaches outlined several developmental stages 135 that led to their current position, including novice coaching, developmental coaching, national 136 elite coaching, and international elite coaching (Schinke et al., 1995). Moreover, time in a given 137 coaching position (i.e., coach tenure and turnover) has also been linked to the development of

expert performance in sports (De Paola & Scoppa, 2008). Long-term coaches have more time to
develop shared mental models with players, which in turn might increase the likelihood of
positive outcomes (for a review see Mohammed, Ferzandi, & Hamilton, 2010). A new coach, on
the other hand, may enhance motivation for players and provide beneficial changes in tactics and
playing styles (Höffler & Sliwka, 2003).

143 Characteristics of Successful Teams

144 Successful sport teams tend to share certain characteristics. In other words, the quality of 145 players on the team, rather than the quality of coaches, might determine sport success 146 (Szymanski & Kuypers, 1999). For instance, comparing a coach of an amateur team to a coach 147 of a professional team would not take into consideration the differences in the quality of the 148 players on each team. To this extent, previous studies examining football performance have 149 compared top to bottom teams, in line with the expert-novice paradigm (Filho & Tenenbaum, 150 2015; Hirotsu & Wright, 2003; Tenga, Holme, Ronglan, & Bahr, 2010; Yates, North, Ford, & 151 Williams, 2006). In the present study, we were interested in examining the effect of team quality 152 on performance in women's football, and thereby we explored the effects of the *number of times* 153 team has qualified for Champions League and number of times team has won Champions 154 League.

The bulk of empirical findings suggest that football teams can be successful by adopting different playing styles, akin to the equifinality principle in human movement sciences (Schmidt, McGown, Quinn, & Hawkins, 1986). For instance, one team may succeed by playing offensive style football (e.g., Brazil), whereas other teams may succeed by playing defensive football (e.g., Italy; see Filho, Basevitch, Yang, & Tenenbaum, 2013). Notwithstanding, having skilled players (i.e., players with task-related knowledge) on the roster seems paramount to team success 161 irrespective of the playing style adopted by a given team (Mohammed et al., 2010). As such, in 162 the present study we were interested in analyzing whether the *number of international players* 163 and number of players with national team experience increased the likelihood of successful 164 performance at the UEFA Women's Champions League. On one hand, athletes with national 165 team experience have access to high-level competition experiences and are more likely to be 166 experts in their respective sport domains (Côté, Salmela, & Russell, 1995). On the other hand, 167 international football players have been shown to perceive performance and team dynamics 168 differently than domestic athletes (Filho et al., 2014b), congruent with the overarching notion 169 that country-level effects influence athletic performance (Côté, Macdonald, Baker, & Abernethy, 170 2006).

171 Characteristics of Successful Countries

The country in which the team hails from may also play a role in success (Salmela & Moraes, 2003). Certain countries have reputations for excellence in football in general, and women's football in particular. In fact, since the inception of the men's FIFA World Cup in 1930, winners of the tournament have come from only eight countries (i.e., Argentina, Brazil, England, France, Germany, Italy, Spain, and Uruguay. The picture is similar for the women's game. Since the women's FIFA World Cup began in 1991 only four teams (i.e., Germany, Japan, Norway, and the United States) have won the tournament.

The rate of success of different countries may depend on various factors, including the popularity of the sport in the country (Salmela & Moraes, 2003). The size of the country may also impact the success of teams from that nation (Noll, 2002; Torgler, 2004). Teams from small countries may have a competitive disadvantage compared to teams that hail from large countries and thus benefit from having a greater number of football divisions, teams per division, and

184	professional players (Dejonghe & Vandeweghe, 2006).
185	Overall, myriad country-level variables may influence how teams play and coaches
186	develop and instruct, ultimately affecting the performance of club and national teams in
187	important international tournaments. Therefore, in the present study, we explored whether
188	several country-level variables (e.g., total number of divisions, number of teams in top division,
189	favorite team sport, budget for women's football) were predictive of performance at the UEFA
190	Women's Champions League.
191	Research Questions & Hypotheses
192	The overarching research question guiding the present study was: "What is the profile of
193	winning teams in the UEFA Women's Champions League?" This question was proposed as a
194	broad exploratory inquiry stemming from the notion that coach, team, and country characteristics
195	are associated with excellence in sports. The specific research questions were:
196	(1) What coaches' characteristics are associated with successful performance in the
197	UEFA Women's Champions League?
198	(2) What teams' characteristics are associated with performance in the UEFA Women's
199	Champions League?
200	(3) What country characteristics are associated with performance in the UEFA Women's
201	Champions League?
202	Congruent with the three research questions, the following three hypotheses were
203	proposed:
204	(H1) Coaches' characteristics (level-1) were expected to predict performance in the
205	UEFA Women's Champions League.

206 (H2) At least one team-level characteristic (level-2) was expected to add explanatory
 207 power to the final hierarchical linear model.

208 (H3) At least one country-level characteristic (level-3) was expected to add explanatory
 209 power to the final hierarchical linear model.

210 H1 is congruent with previous research suggesting that coach characteristics are linked to

211 expert performance in sports. H2 and H3 are aligned with the notion that expert performance in

212 sports depends on team and country factors, and consistent with current methodological

213 guidelines on parsimonious hierarchical linear model estimation in which level-2 and level-3

214 variables must be added "one by one" to allow for the development of a parsimonious robust

- 215 model (see Raudenbush & Bryk, 2002).
- 216

Methods

217 Design

218 This study involved archival analysis of factual data on teams and coaches participating 219 in the UEFA Women's Champions League (2011-12 until 2015-16). Country-level variables for 220 the same period were also taken into account. The final UEFA Women's Champions League 221 rank was the dependent variable, coaches' characteristics represented level-1 data, teams' 222 characteristics represented level-2 data, and country characteristics were included as level-3 data. 223 **Data Collection** 224 Official documents with information about the coaches, teams, and countries were 225 provided by UEFA, including team rosters and result sheets. Additional information was 226 gathered from the UEFA website, FIFA.com, and official country association websites. To this extent, previous exploratory research on the predictors of performance in professional football 227

has relied on factual, publicly available online sources (Filho et al., 2013; Hirotsu & Wright,

229 2003).

230 Inclusion criteria. After reviewing the qualifying procedures for the tournament and 231 noting the number of teams that attempted to qualify each season (i.e., over 50 teams competed 232 for a spot in the Round of 32 in 2015-16), it was decided that the data input and analysis would 233 measure only the knockout stage of the tournament (Round of 32). In this way, the dependent 234 variable for the regression model (i.e., UEFA Women's Champions League final rank) would 235 have the same range (i.e., 1 to 32) for all seasons. Furthermore, it is important to note that the 236 structure of the UEFA Women's Champions League allows teams to submit different rosters for 237 each part of the tournament (e.g., Oualifying Round, Round of 32, Round of 16, Ouarter-finals, 238 Semi-finals, and Final). Therefore, to be consistent across all teams, regardless of how far the 239 team advanced in the tournament, the coach- and team-level data was based on information for 240 the Round of 32.

241 With respect to the independent measures, all variables that were not consistently 242 recorded across levels of analyses for varying reasons (e.g., different countries reporting data 243 differently) were excluded from the data pool to ensure the analysis was performed on reliable 244 and valid information. Decisions on the inclusion/exclusion of any variable were made over a series of peer-debriefing meetings involving the authors and "external judges" from UEFA who 245 246 are not authors of this manuscript. Any issues were discussed until consensus was reached. In 247 total, 11 variables were excluded from analysis. A detailed explanation for the rationale 248 supporting the exclusion of each variable is provided as Supplementary Material (Part 1).

249 Data Input

The dependent variable and independent variables related to the coach (level-1), team
(level-2), and country (level-3) included in the analysis are described in detail next. Data before

the 2011-12 tournament was considered for coaches and teams. For instance, coach variables exceed the 5-year interval considered for the dependent variable. By doing so, we accounted for coaches and teams previous participation in the UEFA Women's Champions League since its inception in 2009-10.

256 Dependent variable. Final rank for the UEFA Women's Champions League was determined based on the official regulations of the game (see FIFA.com). Specifically, the 257 258 winner of the final match was ranked 1 and the finalist was ranked 2. All remaining teams were 259 ranked based on the official criteria put forth by FIFA: (1) Greatest combined goal difference in 260 all matches; (2) Greatest combined number of goals scored in all matches; and (3) If more than 261 one team remained level after applying the above criteria, their final ranking was determined based on how far the team that they were eliminated by advanced in the tournament. If the teams 262 263 that were tied were beaten by teams that advanced to the same round of the tournament, then the 264 greatest combined goal difference in all matches for the advanced team was used to separate the 265 tie.

266 **Independent coach-level variables.** Coach-level variables included *age*, *gender*,

267 *nationality status, former professional player, full national team playing experience,*

268 *international playing experience, position as a player, coaching experience of a national team,*269 *years coaching experience in Champions League, and time at current position* (Table 1).

270 *Age.* Age, in years, was calculated based on the date of birth for each coach listed on the
271 official UEFA roster.

272 *Gender.* Gender was included to examine whether differences exist between male and273 female coaches.

13

274 *Nationality status.* The coach's nationality status was coded according to whether they
275 coached a team from their native country or a team from outside their native country.

Former professional player. Whether the coach was a former professional football player
was included as a measure of playing experience. Of note, this variable represented the highest
level of playing experience the coach achieved during his/her career.

Full national team playing experience. The coach's involvement as a player in his/her
full national team was recorded based on information from national team rosters available online.

International playing experience. It was noted whether the coach competed at the
international level for his/her full national team (e.g., FIFA World Cup, Olympics, UEFA
Champions League).

284 Position as a player. It was also considered whether successful football coaches were 285 more likely to have played a certain position. Performance roles and expectations differ between 286 goalkeepers, defenders, midfielders, and forwards. Therefore, the position in which the coach 287 played during his/her career was coded for in the data.

288 Coaching experience of a national team. This variable took into consideration whether289 the coach had experience as the head coach of a national team, including a youth or full national290 team, from any country.

Years coaching experience in Champions League. The number of previous times each
 coach was involved in the UEFA Women's Champions League was recorded as a measure of
 previous coaching experience.

294 *Time at current position.* Time at current position, measured in years, was calculated for295 each coach to assess whether team performance was related to the length of time the coach has296 been in the position.

Number of times team has qualified for Champions League. The number of times the
 team has qualified for the UEFA Women's Champions League reflects the experience of the
 team in previous years.

Number of times team has won Champions League. The number of times the team has
 won the UEFA Women's Champions League title provides information about the past quality of
 the team.

Number of international players. The number of international players on the roster
 might be related to the financial capacity of the team. Wealthier teams have the financial means
 to recruit talent from overseas.

Number of players with national team experience. The total number of players with
 national team experience was included as an indicator of the football quality of the club team.
 Independent country-level variables. Country-level variables included *FIFA world*

312 ranking, total number of divisions, number of teams in top division, number of registered female 313 players, favorite team sport, and budget for women's football (Table 1). All country-level 314 variables, with the exception of FIFA world ranking that was gathered from FIFA.com, were 315 official records provided by UEFA.

316 *FIFA world ranking.* The FIFA world ranking for the country of which the team is from 317 was included in order to account for the strength of women's football in the given country. It was 318 deemed important to consider the ranking for each country at the point closest to the start of the 319 UEFA Women's Champions League, as it was expected that this most accurately reflects the quality of football in the country at the given time. The ranking used for the analysis was the one
issued most immediately preceding the start of the UEFA Women's Champions League
knockout round. For instance, for the 2015-16 competition, the rankings were from September
25, 2015 and the knockout stage started on October 7, 2015. The same procedure was applied to
all other seasons (i.e., 2011-12 to 2014-15).

325 *Total number of divisions.* To explore differences in league structures across countries,
 326 the total number of divisions in the domestic women's football league was included in the model.

Number of teams in top division. Given that the size of divisions also differs across
 countries, the total number of teams in the top national division was included in the model.

Number of registered female players. The total number of registered female players,
above 18 years of age, for the current year was used to measure the popularity of women's
football in each country.

Favorite team sport. Whether football was the favorite team sport, based on media,
exposure, marketing and spectators, was included in the model to explore the potential effect of
popularity of women's football on the dependent variable.

Budget for women's football. The budget (in Euros) for women's football for each
country was included in the data set to assess whether the general financial status of the sport in
the country was related to performance in the UEFA Women's Champions League.

338 Data Analysis

339 The first step in data analysis involved dealing with missing data. Subsequently,

340 descriptive and hierarchical linear modeling analyses were applied to the data set.

341 Missing data. Only two variables (i.e., former professional player; position as a player)
 342 were excluded from the data analysis due to a large percentage (over 30%) of non-available

information. Noteworthy, variables with up to 10% missing data points were treated, in line with recommendations for quantitative research analysis (see Creswell, 2008). Specifically, missing data was treated in three ways: (1) for dummy variables, missing data was coded as "0" ("no" or the absence of the attribute), thus reflecting a conservative approach in inference making; (2) for continuous variables, the median was computed to avoid inflation resulting from outliers; and (3) for *budget for women's football* interpolation was used on a case-to-case basis to determine the values for the missing data.

350 **Descriptive analysis.** Descriptive analysis is particularly informative in census-like 351 inquiries, such as in the case of the present study (Creswell, 2008). Accordingly, measures of 352 central tendency, namely mean, median, and standard deviation, as well as natural frequency 353 counts, were performed.

Hierarchical linear modeling. The data for all seasons was analyzed together as the goal was not to examine changes over time for particular teams but rather to conduct a census-like analysis of the factors linked to success in the UEFA Women's Champions League. Potential carry-over effects were not an issue as we explored the effects of level-1, level-2, and level-3 variables over the time span analyzed. It follows that a three-level hierarchical linear model was tested. Figure 1 is a schematic descriptive summary and graphic representation of all variables considered in the hierarchal linear modeling analysis.

For the null unconditional model, all dummy coded variables were treated as fixed effects, whereas continuous variables were initially conceptualized as random effects in the tested model. Furthermore, across the three levels of analysis, all variables were treated as raw, non-centered scores, given that there was (1) an interest in estimating the unique contribution of

365	each predictor, and (2) no occasion in which a value of zero represented either an undesirable or
366	an unreasonable score (see Raudenbush & Bryk, 2002).
367	Results
368	Congruent with the importance of describing (expert-novice paradigm) and explaining
369	(expert performance approach) potential mechanisms linked to expert performance in sports, we
370	first present the descriptive analysis applied to the final data set. Subsequently, we present the
371	multi-level analysis in a step-by-step mode, from the null unconditional model until the final
372	parsimonious model.
373	Descriptive Analysis for Coaches
374	For demographic factors (Table 2), the descriptive analysis revealed that the coaches
375	were in their early forties (M = 43.51; SD = 9.95), were mostly male (85.60%; $n = 137$), and
376	primarily coached a team in their native country rather than a foreign country. A post-hoc chi-
377	square analysis (see Garcia-Pérez & Núñez-Antón, 2003) confirmed that the proportion of male
378	coaches was statistically greater than the proportion of female coaches ($\chi 2$ (5) = 186.39, <i>p</i> <
379	.001), and the magnitude of this difference was found to be large (Cohen's $d = 2.03$).
380	With respect to coaches' previous experience as football players (see Table 2), the
381	majority of the coaches were not former professional players (54.10%, $n = 59$). Noteworthy, for
382	the most part (88.90%, $n = 136$) coaches with professional playing experience did not play at a
383	premier international level competition, such as the FIFA World Cup, Olympics, or UEFA
384	Champions League. Coaches with previous playing experience at any level were mostly
385	midfielders (43.10%, $n = 31$; see Figure 2). The proportion of midfielders was found to be
386	greater than the proportion of former goalkeepers and defenders $\chi^2(2) = 10.90$, $p < .01$, but did
387	not differ significantly from the proportion of forwards, $\chi^2(1) = 1.13$, $p = .29$.

With respect to the coaches' coaching experience (Table 2), the descriptive analysis revealed that most of them were at their current club in a head coach capacity for about three years (M = 3.36; SD = 4.51), and coaching for the first time in the UEFA Women's Champions League (M = 0.81; SD = 1.00). Over a third of the coaches (37.10%, n = 56) had previously led a youth or full national team.

393 Descriptive Analysis for Teams

394 Central tendency estimates and frequency counts for all level-2 team variables are 395 presented in Table 3. On average, teams had qualified for the UEFA Women's Champions 396 League two times (M = 1.79; SD = 1.56). Furthermore, the teams had a median of 13 players 397 with national team experience, and the average team size was approximately 23 players (M =398 22.71; SD = 2.19). The teams had around four international players on their rosters (M = 4.40; SD = 3.43). The majority of international players were from European countries (66% out of 703) 399 400 in total, n = 469), followed by North American (16%, n = 110), and African countries (9%, n = 100) 401 60; see Figure 3, Panel A). South American and Oceania countries accounted for 4% (n = 30) of 402 the international trade each, with Asian nations accounting for the remaining 1% (n = 9) of 403 foreign players. This trend was found to be consistent across all five years analyzed (Figure 3, 404 Panel B). The proportion of European players was found to be greater than all other continents, 405 $\chi^2(5) = 186.39$, p < .001. The number of players from North America was found to differ 406 significantly from the proportion of players coming from Africa, South America, Oceania, and 407 Asia, $\chi^2(4) = 20.48$, p < .001. No other statistically significant differences were observed when 408 comparing the proportion of international players across continents. 409

410 Descriptive Analysis for Countries

411 Central tendency estimates and frequency counts for all level-3 country variables are 412 presented in Table 4. Teams were from countries with a large range of FIFA world rankings. 413 Across countries, the average number of football divisions was approximately four (M = 4.21;414 SD = 2.06), with the average number of teams in the top division being about 10 (M = 10.55; SD = 2.60). The number of registered female football players, over age 18, varied greatly among 415 416 countries and was roughly 21,000 (M = 21,287; SD = 24,216). However, this value is not particularly informative as the variance was larger than the mean, likely because Europe is 417 418 comprised of countries with varying sizes and socio-economic characteristics. Also, noteworthy, 419 football was the favorite sport in approximately 60% of the countries (59.70%; n = 92), with the 420 budget allotted to women's football being, on average, close to four million Euros per year 421 (Median = 2,500,000; M = 3,953,011; SD = 4,152,050). Altogether, the country-level data was 422 marked by wide variability, thereby corroborating the importance of accounting for country 423 specificity in line with multi-level analysis guidelines.

424 Hierarchical Linear Modeling

First, correlation analyses were performed among the independent variables included in the analysis and the dependent variable (see Supplementary Material – Part 2). Overall, a linear relationship was observed, thus attesting the application of hierarchical linear modeling analysis to the data set (see Raudenbush & Bryk, 2002). For brevity, only the omnibus final model is defined in the text. The statistical definitions and coefficients for all models, including the intermediate models not detailed in the text, are given as Supplementary Material (Part 3) in the order in which they were ran.

432	Null unconditional model. Initially, the null unconditional model with two levels and no
433	independent variables was tested. The fixed and random effect estimates for the null
434	unconditional model are presented in Table 5. The reliability estimate for this model indicated
435	that 19% of the variance of final rank for the UEFA Women's Champions League was due to
436	between-team variables. The grand mean estimate was significant at 17.75 (CI = 19.72, 15.77),
437	and thus near the median value (final ranking = 16, as there are 32 teams) for the final ranking
438	across all teams. There was no significant effect for the variance components, thus suggesting the
439	adoption of a fixed effect model for the subsequent models.
440	Level-1 modeling. Model A included all level-1 coach variables. The coefficients,
441	standard errors, t-ratios and <i>p</i> -values for all tested variables are presented in Table 6. Based on
442	the results of Model A (Table 6), the next step involved advancing a more parsimonious model.
443	Specifically, congruent with guidelines on parsimonious statistical modeling (see Cohen, West,
444	& Aiken, 2002), Model B contained only the level-1 significant predictor of final rank: years
445	coaching experience in Champions League (see Table 7). Within Model B (Table 7), every
446	additional year of coaching experience in Champions League was found to improve final rank by
447	3.63 positions (γ_{70} = -3.63, <i>p</i> = .015). The intercept for Model B was estimated at 14.25 (CI =
448	11.66, 16.84) with the confidence interval encompassing the expected average value for final
449	ranking across all teams. Compared to Model A (Table 6), the reliability estimate for between-
450	teams decreased slightly to 17% after adding years coaching experience in Champions League to
451	the Model B. Nevertheless, computation of Pseudo R ² (see Raudenbush & Bryk, 2002) indicated
452	that Model B explained 6.84% more variance of final ranking than the null unconditional model
453	(Table 5) with no predictors.

454 **Level-2 modeling.** This step involved the consideration of team-level variables.

455 Congruent with guidelines on parsimonious hierarchical linear modeling (Raudenbush & Bryk,
456 2002), an a priori exploratory analysis was conducted to determine which significant level-2
457 predictors should be included in the model (see Supplementary Material – Part 3) in order to
458 advance the best, yet most parsimonious two-level model.

459 Level-2 variables were included on a "one to one basis" in the analysis, until a final 460 solution. wherein all predictors were statistically significant, was reached. Results for this model, 461 namely Model C (Table 8), suggested that *vears coaching experience in Champions League* at level-1, and number of times team has won Champions League and number of international 462 463 *players* at level-2, were significant predictors of final rank. Specifically, for every additional year 464 of experience coaching in the Champions League, final rank improved by approximately three positions ($\gamma_{10} = -2.90$, p = .038). Moreover, for every time a team raised the Champions League 465 trophy, final rank was estimated to improve by seven positions ($\gamma_{01} = -7.13$, p < .001). Finally, 466 467 every international player on the roster represented an improvement in final rank by about one 468 position ($\gamma_{02} = -1.08$, p < .001). The intercept for the model was significant at 24.56 (CI = 21.76, 469 27.36).

470 Level-3 modeling. To test whether a three-level model was required or whether a two471 level model would suffice, variance was fixed at ".19" (see Raudenbush & Bryk, 2002), which
472 was the reliability estimate for Model C (Table 8), and an exploratory analysis of all level-3
473 predictors was conducted (see Supplementary Material – Part 3).

474 The variables found to be statistically significant at level-1 (i.e., *years coaching*

475 *experience in Champions League*) and level-2 (i.e., *number of times team has won Champions*

476 *League*; *number of international players*) were then added to the hierarchical regression analysis,

along with *FIFA world ranking* at level-3, which was found to significantly predict final rank
(Table 9). The intercept for the model was estimated at 21.85 (CI = 18.86, 24.84), with the
reliability estimate for level-2 suggesting that 12% of the variation in the means of final rank was
due to true variation between countries. Importantly, in this three-level solution, *years coaching experience in Champions League* was no longer found to be a significant predictor of final rank.

482 Final model. Both the three-level solution given in Table 9 and the two-level solution 483 presented in Table 8 are suitable omnibus models to explain final rank for the UEFA Women's 484 Champions League. Importantly, reliance on statistical guidelines for model estimation does not 485 provide a straightforward answer for deciding between two alternative non-equivalent models 486 (Raudenbush & Bryk, 2002). On the one hand, arguments can be developed in favor of choosing 487 better-fit indices (see Stapleton, 2006), in which case the three-level solution given in Table 9 would be preferable as Pseudo R^2 computation indicates that this model accounted for an 488 489 additional 55.23% of the variance of final ranking scores. On the other hand, arguments can be 490 developed in favor of the more parsimonious two-level solution given in Table 8 (Gigerenzer, 491 2010; Tenenbaum & Filho, 2015). Every time you add factors to a model, the complexity of the 492 model increases (over parameterization) and its applicability tends to decrease.

To reach a decision between the two alternative solutions, the estimated impact of the level-3 and level-1 predictors on the criterion final rank were analyzed in detail. In regard to a three-level solution (Table 9), the median effect of *FIFA world ranking* on final ranking was close to a two-position downgrade ($\gamma_{001} = 0.09 * 17.5 = 1.58$), with numerous effects in between being possible (Figure 4, Panel A). Regarding a two-level solution (Table 8), the estimated average effect of years coaching experience in Champions League on final ranking is about a two-position upgrade ($\gamma_{10} = -2.90 * .81 = -2.35$). This effect was found to be linear over time

500	(Figure 4, Panel B), influencing final ranking by a maximum of approximately twelve positions
501	for coaches with four years of experience in the league ($\gamma_{10} = -2.90 * 4 = 11.60$), as per the
502	observed range for this variable (Table 2). Given that the impact of years coaching experience in
503	Champions League is more substantial than the impact of FIFA world ranking on final rank, a
504	final choice for a two-level solution is proposed herein (see Figure 5) and defined below:
505	Level-1 Model
506	Final rank _{ij} = $\beta_{0j} + \beta_{1j}$ *(Years coaching experience in Champions League) + r_{ij}
507	Level-2 Model
508	$\beta_{0j} = \gamma_{00} + \gamma_{01}^*$ (Number of times team has won Champions League) + γ_{02}^* (Number of
509	international players) + u_{0j}
510	$\beta_{1j} = \gamma_{10}$
511	β_{0j} : The predicted final rank mean controlling for the number of previous Champions League
512	wins and the number of international players on a given team j
513	β_{lj} : The predicted change in final rank for every year of coaching experience in the Champions
514	League for a given coach i in a given team j
515	γ_{00} : The grand mean for the dependent variable final rank across teams
516	γ_{01} : The average change in final rank for every time a given team j has won the Champions
517	League
518	γ_{02} : The average change in final rank for every international player on a given team j
519	r_{ij} : The deviation of final rank from its predicted value for a given coach i in a given team j
520	u _{0j} : A random effect for team j
521	The above-specified model, therefore, supports H1 and H2 but does not corroborate H3.
522	Had a three-level solution been selected, H3 and H2 would have been supported but not H1.

523	Considering the final coefficients estimated for this study (Table 8), the lowest "error free"
524	hypothetical final rank value consists of a coach with no previous experience in the league,
525	coaching a team with no previous league title, and without any international players on the roster
526	according to the equation:
527	Final rank = $24.56 + (-2.90) * (0) + -7.13 * (0) + -1.08 (0)$
528	Variations in the final rank value would depend on the number of previous years of
529	experience in the UEFA Women's Champions League by a given coach, a team with up to two
530	overall UEFA Women's Champions League titles within the past five years, and with a
531	maximum number of 15 international players on the roster. Again, the reported coefficients are
532	fixed rather than random and apply to the studied population given the range of the variables.
533	Discussion
534	The purpose of this study was to explore coach, team, and country factors linked to
535	performance in the UEFA Women's Champions League. To this end, descriptive statistics and
536	hierarchical linear modelling was applied to a data set spanning five seasons, for the three-
537	aforementioned levels of analysis. The main observed findings are discussed next.
538	
	Descriptive Analysis for Coaches
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546 management, in a classic study profiling the characteristics of over 1,000 executives, Bantel and 547 Jackson (1989) observed that CEOs from large corporations were in their forties on average. 548 Whereas previous experience seems to be essential to lead premier football clubs in the 549 UEFA Women's Champions League, the type of experience might differ across individuals. In 550 particular, the statistical analysis revealed that former professional players were not more likely 551 to coach in the league than those with no previous professional experience as a player. Thus, the 552 pathways to become a coach in the UEFA Women's Champions League seem to vary, akin to 553 the equifinality principle (see Von Bertalanffy, 1968), which purports that expert performance 554 can be reached through different routes. This finding bears implication for the on-going global 555 debate on coaching education (see Vargas-Tonsing, 2007), as it suggests that different types of 556 experience (e.g., former professional player, explicit academic training, formal coaching 557 education) can lead individuals to coaching at the highest competitive level. 558 It is noteworthy, however, that the majority of coaches with playing experience at any 559 level used to play as midfielders. Coaches who played as a midfielder might have a greater 560 chance of leading an elite women's football club in Europe. Midfielders have been found to 561 perceive performance requirements differently than players from other positions (i.e., 562 goalkeepers, defenders, and forwards) likely because midfielders are, in a sense, a hybrid 563 position that shares both defensive and offensive responsibilities (Filho et al., 2014b). As such, 564 former midfielders might have developed a better understanding of the game in both its 565 defensive and offensive requirements. Moreover, previous research has shown that athletes that 566 play in centralized positions have more access to information, and thus are more likely to facilitate team coordination and performance by communicating shared and complementary 567 568 information to their teammates (Filho, Gershgoren, Basevitch, Schinke, & Tenenbaum, 2014a).

569 Noteworthy, our analysis revealed that there were significantly more male than female 570 coaches in the league. This finding echoes previous research in the field in that women coaches' 571 report difficulties in progressing to a high-ranking coaching status in professional sports 572 (Norman, 2013). In particular, women coaches have noted that advancing to high-performance 573 coaching positions is difficult likely because of implicit gender biases, as coaching in sports is 574 dominated by men (Norman & Rankin-Wright, 2016; Rankin-Wright, Hylton, & Norman, 2017). 575 It is therefore important to support initiatives to increase the number of women in leadership 576 positions in sports and other domains of human performance (Blau, 2016). Particularly with 577 regards to women's football, policies should be continually developed to encourage former 578 female players to seek the necessary licenses and qualifications to pursue a career in coaching. 579 Examining the effectiveness of gender equality policies currently in place is also paramount to 580 ensure women take on leadership positions across domains of human performance (see Burton, 581 2015).

582 It is important to highlight that only about 10% of the coaches were from international 583 countries. There are limited financial resources in women's football and this might shed light on 584 the relatively low frequency of international coaches at the knockout round of the UEFA 585 Women's Champions League. Availability of financial resources may also explain job stability 586 in the analysed sample. Coaches were found to serve in their current position for over three years 587 on average, thus signalling a smaller coaching turnover than that observed in the men's game 588 (see De Paola & Scoppa, 2008). This finding might explain why coach tenure was positively 589 linked to performance. Over time, coaches get to know their player and teams, and thus are able 590 to foster the development of various team processes (e.g., cohesion, team mental models,

591 collective efficacy) while devising more effective performance strategies (Balduck, Prinzie, &

592 Buelens, 2010; Shamsie & Mannor, 2013).

593 Descriptive Analysis for Teams

594 Frequency counts revealed that only three teams had won the UEFA Women's 595 Champions League within the 5-year span analysed. Accordingly, there is evidence that "hubs of 596 expertise" occur and are dominant within the European league network. As per the Pareto law, 597 80% of outcomes tend to come from 20% of the inputs. It follows that qualitative analysis of 598 these highly successful cases is warranted as previous research suggests that studying the modus 599 operandi of a few expert teams can yield important insights to inform the development of less 500 successful teams (see Gershgoren, Filho, Tenenbaum, & Schinke, 2013).

601 Although few teams had earned the title, the teams had on average two years of 602 experience participating in the UEFA Women's Champions League. This suggests that the 603 quality of the team is paramount. Skill matters in the quest for success, which is why companies 604 from all domains seek to hire and retain highly qualified employees (Lockwood & Ansari, 1999). 605 In fact, the team-level data suggests that teams in the UEFA Women's Champions League have 606 top quality players, with an average of over 12 players with national team experience per team. 607 This finding opens another question pertaining to the direction of this putative 608 relationship: Do players that play for their national teams join the best club teams in Europe or 609 does playing on a strong team in the UEFA Women's Champions League increase a player's 610 chance of being invited to join her national team? It is likely that a reciprocal relationship occurs, 611 wherein playing on a top club team increases the players' visibility to join her respective national 612 squad and vice-versa: playing on a national team increases the chance of being hired by a leading 613 football club in Europe. Also noteworthy, countries with more or less tradition in football

614 produce players of more or less quality. In other words, hiring players from soccer powerhouse 615 countries (e.g., Brazil, England, Germany, the United States) might be more impactful than 616 hiring players from less traditional soccer nations. As discussed above, there is a grand influx of 617 players from the United States, which currently is the dominant country in women's football. 618 On average, teams had just over four international players on their squad. This figure is 619 likely constrained by the fact that European countries regulate the number of players outside 620 Europe that can play in their leagues (see Flores, Forrest, & Tena, 2010). While the number of 621 players is a constrained factor, the origin of the players is a "free parameter", mainly shaped by 622 the unique dynamics of women's football. Specifically, the majority of international players at 623 the UEFA Women's Champions League come from North America, particularly the United 624 States, who has been the major force in women's football for the past decade. As is the case with 625 many job markets, local protective measures along with the strength of the marketplace in other 626 countries establish the migration flow of workers around the globe (Greenwood, 2014).

627

Descriptive Analysis for Countries

628 Across the 35 countries represented in the UEFA Women's Champions League over the 629 5-year span analyzed, football was found to be the *favorite team sport* among women. In the 630 past, football has been stereotypically associated with male rather than female socially desirable 631 traits (Azzarito, Solmon, & Harrison, 2006). However, a positive shift has been noticed more 632 recently, with an increasing number of girls and women playing football around the globe (Lunz, 633 2007). It is important that researchers and practitioners continue to observe how societal and 634 cultural changes (e.g., gender rights movement) influence sport play and choice for women in 635 different countries.

636 All other country-level variables were characterized by wide variability. In fact, from the 637 FIFA world ranking to total number of divisions and number of teams in top division, great 638 dispersion in the data pool was the major trend observed. Scattered data patterns were also 639 noticed for number of registered female players and budget for women's football among the 35 640 countries that were analysed. Together, these findings suggest that heteroscedasticity in the 641 organization of national leagues as well as the economics of football is part of the women's game 642 in Europe. Hence, the recommendation derived from these findings is that scholars and 643 practitioners should continue to account for country-level factors when studying expertise among 644 individual sport actors, such as coaches in the present study, and teams at large.

645 Multi-Level Effects: Coaches within Teams within Countries

646 Agents at one level are systems at another level (Von Bertalanffy, 1968). For this reason, 647 mapping cross-level effects allows for a deeper understanding of optimal performance across 648 domains of human interest, including football (Filho et al., 2014b). In the multi-level analysis 649 applied herein, the results support the hypotheses that coach- and team-level variables are related 650 to performance in the UEFA Women's Champions League for a two-level solution, and that 651 team-level factors and country-level factors are paramount within a three-level solution. From a 652 three-level perspective, countries with higher FIFA world rankings have better teams that are 653 more likely to be successful regardless of their coaches, in comparison to weaker teams from less 654 traditional football countries. From a two-level view, coaches with more experience increase the 655 chances of victory in the UEFA Women's Champions League.

Experienced and successful coaches are also more likely to be recruited and retained by better teams. Altogether, "reciprocal determinism" (see Bandura, 1997) from a socio-cognitive standpoint or "affordances" (see Fajen, Riley, & Turvey, 2009) from a naturalistic account might

659	be at play here. Reciprocal determinism pertains to the notion that individual, group, and
660	contextual processes are intertwined and mutually influence one another. Within an affordance
661	view, changes to input throughout and output relations in a given system are more or less likely
662	depending on a set of constraints and initial values. For instance, it has been shown that success
663	in sports and other areas of human performance depends, in part, on place of birth (Côté et al.,
664	2006). In all, countries influence the development of teams and coaches. Likewise, hiring
665	experienced coaches may influence the development of strong teams, which in turn may
666	influence the development of football over time in a given country.
667	Regardless of which view is adopted (the two-level solution proposed herein or the
668	aforementioned three-level alternative solution), the quality of the teams was found to matter the
669	most in predicting performance at the UEFA Women's Champions League. In other words, the
670	strongest predictive effects originate from the team-level of analysis. A team that has won the
671	UEFA Women's Champions League before is more likely to succeed again. In fact, previous
672	performance accomplishments are a major predictor of efficacy beliefs, which in turn are major
673	predictors of performance in team sports in general (Feltz, Short, & Sullivan, 2008), and football
674	in particular (Filho, Tenenbaum, & Yang, 2014c; Leo et al., 2013). To put plainly, success boosts
675	confidence, which in turn increases the chance of further success. Additionally, more successful
676	teams are likely more attractive to high-quality athletes motivated by the best chances to win
677	titles (Sanderson & Siegfried, 1997).
678	The number of international players on the team was also found to predict final rank at
679	the UEFA Women's Champions League after analyzing several level-1 coach and level-3

681 performance differently, and apply different defensive and offensive tactics to football play

country relevant variables. International players aggregate value to the team, as they perceive

680

682 (Filho et al., 2014b). Moreover, international football players are usually top-level athletes that 683 have left their native countries to take on more prosperous job opportunities in foreign nations 684 (Kleven, Landais, & Saez, 2013). Similar to top-level engineers from around the world who are 685 hired by multinational corporations in Silicon Valley for instance, world-class foreign football 686 players are hired by European clubs to add value to their squads. To illustrate further, for part of 687 the 2015-16 season, Marta Da Silva (Brazil) and Carli Lloyd (United States), two the most 688 successful women football players of all times, played away from their homes for clubs in 689 Europe.

690 With respect to level-1 data, previous experience coaching in the UEFA Women's 691 Champions League was also found to predict final rank. Coaches that have competed in the 692 league before are likely more aware of the challenges that the competition imposes, such as 693 strategies to counter-act home field advantage and the away goals rule (i.e., goals scored at away 694 venues count more than goals scored at home). In effect, experience at the highest level of 695 competition is important in the development of expertise (Bloom, 1985; Côté et al., 1995; 696 Williams & Ericsson, 2005). Previous experience allows one to develop mental representations 697 that can be applied before, during, and after decisive moments in sport competitions (Filho & 698 Tenenbaum, 2015; Tenenbaum, Basevitch, Gershgoren, & Filho, 2013). Put differently, once 699 exposed to high-pressure situations, individuals develop mental skills that allow them to self-700 regulate and perform better the next time around.

With respect to level-3 data, expressive variability was observed across countries in all
measured variables. Hence, considering country-level factors is important in research on
women's football. However, the size and financial power of a country is not the major factor
predicting performance of teams at the UEFA Women's Champions League. In fact, previous

705 research has shown that the size and financial power of a country does not necessarily explain 706 performance in football (Hoffmann, Ging, & Ramasamy, 2002). Countries of smaller sizes and 707 budgets may also succeed in sports if the culture around that sport is strong enough. From the 708 present analysis, the only factor that might play a role in performance at the UEFA Women's 709 Champions League was the FIFA world ranking for a given country. More traditional countries 710 may perform better than less traditional ones. Thus, it is important to examine country-level 711 factors when studying performance in women's football. However, it is important to reiterate 712 that, for the present study, the quality of the team and the experience of the coach are paramount 713 for success in the UEFA Women's Champions League. That is, teams from less traditional 714 countries that have a winning story and an experienced coach may triumph in the end. The scope 715 of these findings, limitations, applied implications, and avenues for future research are discussed 716 next.

717 Limitations and Strengths

718 There are at least two limitations that need elaborating to orient future research in 719 women's football. As previously mentioned, the iterative model was fixed rather than random 720 and thus generalizability is limited to the variables tested within their respective ranges. 721 Moreover, this study was correlational in nature and, as such, inferences of causality are not 722 appropriate. Despite these limitations, this study advances the literature on women's football, as 723 the majority of research efforts in football have targeted the men's game. In the present study, an 724 inductive model of expert performance in women's football emerged from our data analysis 725 (Figure 5), addressing, at least in part, the call for empirical research geared towards developing 726 frameworks of expert performance in women's sports (see Gledhill et al., 2017). To this extent, 727 findings of this study contrasted many common notions in men's football, thus making it clear

728 that gender effects exist in the "beautiful game" and that guidelines derived from men's football 729 do not necessarily apply to high-performance women's football. Also, notwithstanding the cross-730 sectional nature of the study, the comprehensive census-like analysis presented herein provides 731 more than a "snapshot profile" of high-performance women's football in Europe. Natural 732 frequency counts revealed the current status of coaches, teams, and countries participating in the 733 league bringing to light, for instance, the small number of women coaches in the League. 734 Stakeholders should use the findings of this study to "think-act-reflect" (i.e., reflexive practice) 735 on best practice guidelines for coaches, teams, and countries. Awareness of factors related to 736 high-performance at the UEFA Women's Champions League is an important step to promote 737 positive (and evidence-based) changes in premier women's football.

738 Future Research

739 Future research could focus on studying expert coaches through qualitative lenses. As the 740 results have shown, the proportion of female coaches in the league is much smaller than the 741 proportion of male coaches. Accordingly, we echo the call for more studies on the challenges 742 that women face in pursuing a professional coaching career in sports (see Norman, 2013; 743 Norman & Rankin-Wright, 2016; Rankin-Wright et al., 2017). In particular, additional research 744 on the intersections of gender and other minority statuses (e.g., race/ethnicity) among football 745 coaches is warranted. More studies on the migration flow of international athletes are also 746 warranted. As the findings illustrate, the immigration flow of football players at the UEFA 747 Women's Champions League contrasts with what is known about the male player migration (see 748 Elliott & Harris, 2014). Also, the effect of the team's budget on performance variables should be 749 examined. In the present study, budget for women's football was modelled at the country-level 750 of analysis, not the team-level. It is likely that the quality and number of international players on

the team, factors that have been found significant in the present study, co-vary with the team's annual operating budget. However, it might be challenging to obtain this information, as teams might not be willing to disclose financial data.

Future research should look beyond the demographic characteristics of coaches and teams by addressing the multi-layered relationship among latent individual psychological factors ("I" factors, such as personality) and team processes (e.g., cohesion, collective efficacy). Furthermore, while it is unclear whether a general theory of expertise will ever be developed (Farrow & Baker, 2018), scholars should continue to work towards theoretical integration in research on expert performance in sports. An integrated yet parsimonious model of expertise in team sports might help to inform research and practice in sport psychology.

761 Applied Implications

762 Our analysis revealed that individual and team-level factors should be taken into account 763 by practitioners working in women's professional football. Foremost, team-level factors are most 764 important in predicting successful performance at the UEFA Women's Champions League. As 765 such, sport professionals should think about interventions that address "the team first". 766 Specifically, drafting players from traditionally successful teams as well as international players 767 may increase the chance of winning games at the UEFA Women's Champions League. Former 768 winners and international players bring the experience and confidence that propels performance 769 in high-level competitions. Practitioners wanting to promote peak performance in women's 770 football should also consider developing "cultural intelligence" interventions aimed at promoting 771 cross-cultural understanding in teams with numerous international players on their rosters. 772 The "team comes first", but our findings also revealed that coaching experience matters. 773 Accordingly, teams seeking to improve their performances in the UEFA Women's Champions

774 League should also consider hiring coaches who have previous experience in the competition. As 775 discussed, previous high-stake experience fosters the development of mental representations, 776 which are the basis for effective cognitive, affective, and behavioral patterns differentiating 777 expert individuals and teams from their less successful counterparts. Alternatively, teams could 778 work towards developing their coaches by reducing turnover and providing opportunities for 779 continued education and "learning the job while doing the job", rather than emphasizing an 780 immediate outcome. Repeated participation in the UEFA Women's Champions League may 781 equip coaches with the experience needed to help teams perform better over time. 782 Finally, the findings of this study reinforce the importance of governing bodies and 783 Football Associations in developing (a) coaching education programs tailored to the specific 784 needs of women's football; (b) initiatives to increase the number of women coaching high-785 performance football teams: and (c) campaigns publicizing the benefits of cultural diversity in 786 sports. Governing bodies should consider ways to promote "competitive balance" in order to 787 avoid a few teams consistently winning the championship, which negatively impacts the 788 economic sustainability of other teams (Sanderson & Siegfried, 2003). To conclude, we call for 789 comprehensive multi-levels of analysis studies on expert-performance across domains of human 790 interest. By examining multi-level effects it is possible to advance knowledge on how to foster

talent at the individual level of analysis, while promoting the development of expert teams, andadvancing country-level policies to promote quality sport play around the world.

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Coding and Description for Variables

Variable	Data Source	Coding Description
Coach-Level Variables		
Age	UEFA ^a	Continuous
Gender	Team website or sporting website	Dummy coded; $0 = male; 1 = female$
Nationality status	UEFA ^a	Dummy coded; $0 = \text{coaches team from}$ outside native country; $1 = \text{coaches team}$ from native country
Former professional player	Team website or sporting website	Dummy coded; 0 = did not play as a professional; 1 = played as a professional
Full national team playing experience	Team website or sporting website	Dummy coded; $0 = \text{did not play on full}$ national team; $1 = \text{played on full national}$ team
International playing experience	Team website or sporting website	Dummy coded; 0 = did not play internationally; 1 = played in World Cup, Olympics, or Champions League
<i>Position as a player</i> Goalkeeper; Defender; Midfielder; Forward	Team website or sporting website	Dummy coded; $0 = no; 1 = yes$
Coaching experience of a national team	Team website or sporting website	Dummy coded; $0 = \text{did not coach a youth or}$ full national team; $1 = \text{coached a youth or}$ full national team
Years coaching experience in Champions League	Team website or sporting website	Continuous
Time at current position	Team website or sporting website	Continuous

991 Table #1 – Continued

Variable	Data Source	Coding Description
Team-Level Variables		Continuous
Number of times team has qualified for Champions League	UEFA ^b	Continuous
Number of times team has won Champions League	UEFA ^b	Continuous
Number of international players	UEFA ^a	Continuous
Number of players with national team experience	Team website or sporting website	Continuous
Country-Level Variables		
FIFA world ranking	FIFA ^c	Continuous
Total number of divisions	UEFA ^d	Continuous
Number of teams in top division	UEFA ^d	Continuous
Number of registered female players	UEFA ^d	Continuous
Favorite team sport	UEFA ^d	Continuous
Budget for women's football	UEFA ^d	Continuous

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⁹⁹³ ^aData came from the UEFA Women's Champions League Player List, provided by UEFA.

⁹⁹⁴ ^bData came from the official UEFA Women's Champions League website (<u>http://www.uefa.com/womenschampionsleague/index.html</u>).

995 °Data came from the official FIFA website (<u>http://www.fifa.com/fifa-world-ranking/ranking-table/women/index.html</u>).

⁹⁹⁶ ^bData came from the Women's Football Across The National Associations yearly reports, provided by UEFA

Descriptive Statistics for Coach-Level Variables

Variables	Code or	Median	Mean (SD)	Valid % (n)	Missing % (n)	Included in HLM Model
Age	Range 27–71	43.00	43.51	99.40	.60(1)	Model Yes
Age	2/=/1	45.00	(9.95)	(159)	.00(1)	1 05
Gender	0/1		().)))	100	0 (0)	Yes
Sentuer	0/1			(160)	0 (0)	105
Male	0			85.60	-	
				(137)		
Female	1			14.40	-	
				(23)		
Nationality status	0/1			100	0 (0)	Yes
-				(160)		
Coaches team from outside native	0			8.10	-	
country				(13)		
Coaches team from native country	1			91.90	-	
				(147)		
Former professional player	0/1			68.10	31.90	No
				(109)	(51)	
Did not play as a professional	0			54.10	-	
				(59)		
Played as a professional	1			45.90	-	
				(50)		
Full national team playing	0/1			100	0 (0)	Yes
experience				(160)		
Did not play on full national team	0			86.90	-	
				(139)		
Played on full national team	1			13.10	-	
				(21)		
International playing experience	0/1			95.60	4.40(7)	Yes
				(153)		
Did not play internationally	0			88.90	-	
				(136)		
Played in World Cup, Olympics, or	1			11.10	-	
Champions League				(17)		
Position as a player				45.00	55.00	No
				(72)	(88)	
Goalkeeper	0/1			13.90	-	
				(10)		
Defender	1			13.90	-	
				(10)		
Midfielder	1			43.10	-	
				(31)		
Forward	1			29.10	-	
				(21)		

1002 1003

1005 1006 Table #2 – continued

Variables	Code	Median	Mean	Valid	Missing	Included in
	or		(SD)	% (n)	% (n)	HLM
	Range					Model
Coaching experience of a national	0/1			94.40	5.60 (9)	Yes
team				(151)		
Did not coach a youth/full national	0			62.90	-	
team				(95)		
Coached a youth/full national	1			37.10	-	
team				(56)		
Years coaching experience in	0–4	0.00	0.81	100	0 (0)	Yes
Champions League			(1.00)	(160)		
Time at current position	0-24	2.00	3.36	98.80	1.20(2)	Yes
*			(4.51)	(158)		

1011

Descriptive Statistics for Team-Level Variables

Variables	Range	Median	Mean (SD)	Valid % (n)	Missing % (n)	Included in HLM Model
Number of times team has qualified for Champions League	0-6	2.00	1.79 (1.56)	100 (160)	0 (0)	Yes
Number of times team has won Champions League	0-2	0.00	0.11 (.42)	100 (160)	0 (0)	Yes
Number of international players	0-15	4.00	4.40 (3.43)	99.40 (159)	.60 (1)	Yes
Number of players with national team experience	2 - 20	13.00	12.46 (3.85)	99.40 (159)	.60 (1)	Yes

1013

1015	Table	4

Descriptive Statistics for Country-Level Variables

Variables	Code or Range	Median	Mean (SD)	Valid % (n)	Missing % (n)	Included in HLM Model
FIFA world ranking	2-111	17.50	22.72	98.80 (158)	1.20 (2)	Yes
Total number of divisions	1 – 18	4.00	4.21 (2.06)	93.10 (149)	6.90 (11)	Yes
Number of teams in top division	5-20	10.00	10.55 (2.60)	96.90 (155)	3.10 (5)	Yes
Number of registered female players*	100 – 117,100	14,140	21,287 (24,216)	93.80 (150)	6.20 (10)	Yes
Favorite team sport	0/1	-	-	96.20 (154)	3.80 (6)	Yes
Any sport other than football	0			40.30 (62)		
Football	1			59.70 (92)		
Budget for women's football*	51,600 – 18,370,000	2,500,000	3,953,011 (4,152,050)	95.60 (153)	4.40 (7)	Yes

Multilevel Regression Estimates for the Null Unconditional Model

1024

Fixed Effect	Coefficient	SE	t-Ratio	<i>p</i> -value
Intercept, <i>y</i> ₀₀	17.75	1.01	17.61	< .001
Random Effect	Variance	df	x^2	<i>p</i> -value
Intercept, u_0	3.69	68	84.12	.090
Level-1 effect, r_{ii}	57.53			

Multilevel Regression Estimates for Two-Level Model A

Fixed Effect	Coefficient	SE	t-Ratio	<i>p</i> -value
Intercept, γ_{00}	14.25	6.63	2.15	.04
Age, γ_{10}	-0.01	0.10	-0.14	.90
Gender, <i>y</i> ₂₀	0.95	4.85	0.20	.85
Nationality status, γ_{30}	6.10	5.33	1.15	.26
Full national team playing experience, γ_{40}	1.50	5.11	0.29	.77
Coaching experience of a national team, γ_{50}	4.38	2.48	1.77	.08
International playing experience, <i>y60</i>	-4.30	3.91	-1.10	.28
Years coaching experience in Champions League, y ₇₀	-4.29	1.58	-2.71	.01
Time at current position, γ_{80}	-0.04	.37	-0.11	.91
Random Effect	Variance	df	x^2	<i>p</i> -value
Intercept, u_0	3.43	68	72.35	.34
Level-1 effect, r_{ii}	57.35			

1035 Multilevel Regression Estimates for Two-Level Model B

Fixed Effect	Coefficient	SE	t-Ratio	<i>p</i> -value				
Intercept, <i>γ</i> ₀₀	20.02	1.32	15.16	< .001				
Years coaching experience in Champions League, γ_{10}	-3.63	1.46	-2.49	.015				
Random Effect	Variance	df	x^2	<i>p</i> -value				
Intercept, u_0	3.53	68	82.62	.109				
Level-1 effect, r_{ij}	53.59							
Reliability estimate for level-1 = .19 Deviance = 476.77; Number of estimated parameters = 2								

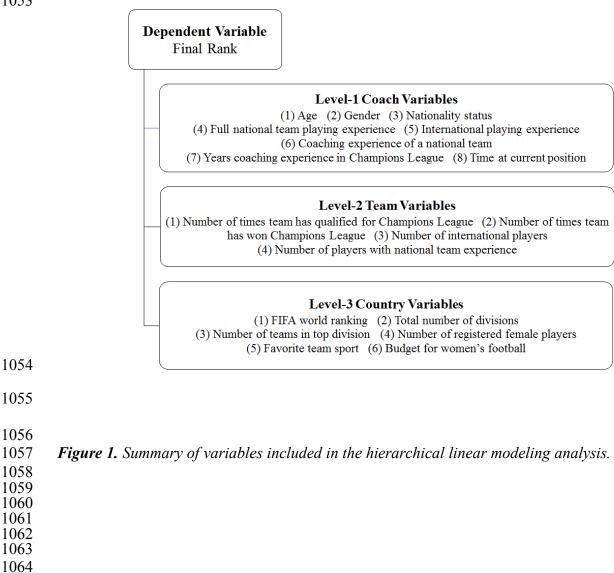
Multilevel Regression Estimates for Two-Level Model C

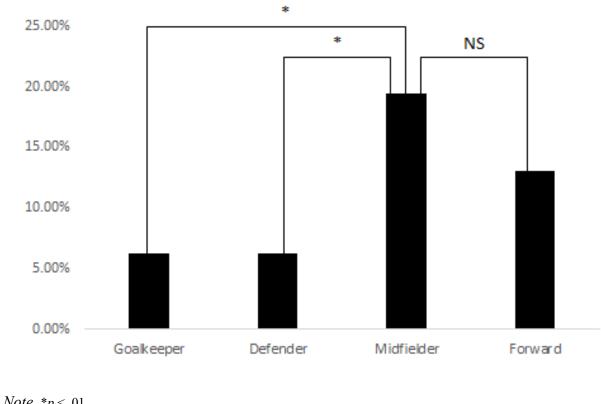
Fixed Effect	Coefficient	SE	t-Ratio	<i>p</i> -value
Intercept, γ_{00}	24.56	1.43	17.23	<.001
Number of times team has won Champions League, γ_{01}	-7.13	1.83	-3.89	< .001
Number of international players, γ_{02}	-1.08	0.25	-4.26	< .001
Years coaching experience in Champions League, γ_{10}	-2.90	1.37	-2.12	.038
Random Effect	Variance	df	x^2	<i>p</i> -value
Intercept, r_0	9.24	66	80.15	.113
Level-1 effect r_{ij}	39.64			
Reliability estimate for level-1= .19 Deviance = 451.28; Number of estimated parame	eters = 2			

047 Multilevel Regression Estimates for Three-Level Model D

Fixed Effect	Coefficient	SE	t-Ratio	<i>p</i> -value
Intercept, <i>y</i> ₀₀₀	21.85	1.53	14.25	< .001
FIFA world ranking, <i>y</i> 001	0.09	0.03	3.03	.005
Number of times team has won Champions League, γ_{010}	-5.79	1.87	-3.10	.004
Number of international players, <i>y020</i>	-1.25	0.25	-4.99	<.001
Years coaching experience in Champions League, γ_{100}	-0.81	1.49	-0.54	<i>p</i> >.05
Random Effect Level-3	Variance	df	x^2	<i>p</i> -value
Intercept 1/Intercept 2, u_{00}	1.80	32	37.52	.23
Reliability estimate for level- $1 = .99$				
Reliability estimate for level- $2 = .12$				
Deviance = 215.20; Number of estimated param	neters = 7			

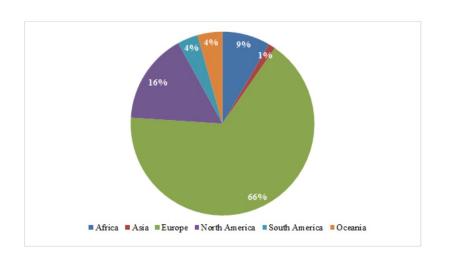






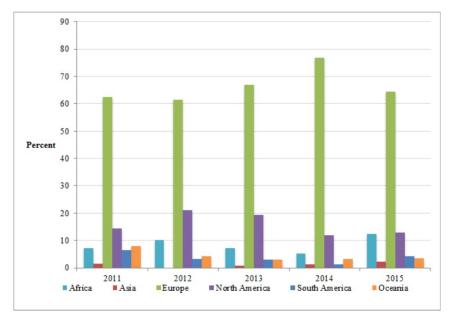
Note. **p* < .01

Figure 2. Playing position of coaches.



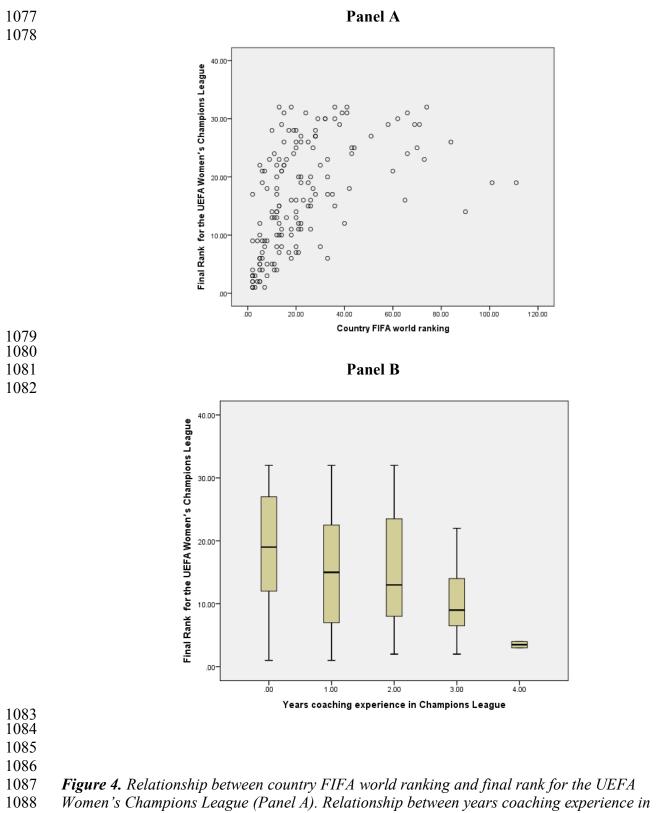






1073 *Figure 3.* Overall proportion of international players per continent competing in the UEFA

- 1074 Women's Champions League from 2011-12 to 2015-16 (Panel A). Proportion of international
- 1075 players per continent by year (Panel B).
- 1076



1089 Champions League and final rank for the UEFA Women's Champions League (Panel B).

- 1090
- 1091

