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1	Original Investigation
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4	Quantifying Technical Actions in Professional Soccer Using Foot-Mounted Inertial
5	Measurement Units
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8	Joshua Marris ^{a, b} , Steve Barrett PhD ^c , Grant Abt PhD ^a & Chris Towlson PhD ^a
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11	^{a.} Department of Sport, Health and Exercise Science, University of Hull, Kingston upon Hull, UK;
12	^{b.} Sports Science and Medicine Department, Hull City AFC, Kingston upon Hull, UK;
13	^{c.} Department of Sports Science and Research Innovation, PlayerMaker [™] , London, UK.
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15	
16	Corresponding Author
17	Joshua Marris ^{a, b}
18	Hull City AFC Training Ground, Millhouse Woods Lane, Cottingham, HU16 4HB, United Kingdom
19	joshua_marris@hotmail.co.uk
20	ORCiD: 0000-0001-8823-9283
21	Twitter: @JoshMarris11
22	
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29 Abstract

30 This study aimed to (i) establish the concurrent validity and intra-unit reliability of a foot-mounted 31 inertial measurement unit for monitoring soccer technical actions, (ii) quantify the within-microcycle 32 inter-positional differences in the technical actions of professional soccer training, and (iii) determine 33 the influence of drill category on the technical actions of professional soccer training. Twenty-one 34 professional soccer players' technical performance data (ball touches, releases, ball touches per minute, 35 releases per minute), collected during training sessions (i.e. match-day (MD) minus day number) 36 throughout twenty-four weekly microcycles, were analysed using general linear modelling. The inertial 37 measurement unit exhibited good concurrent validity ($P_A = 95.1\% - 100.0\%$) and intra-unit reliability $(P_A = 95.9\% - 96.9\%, CV = 1.4\% - 2.9\%)$ when compared with retrospective video analyses. The most 38 ball touches ($\overline{X} = 218.0$) and releases ($\overline{X} = 110.8$) were observed on MD - 1, with MD - 5 eliciting the 39 highest frequency of ball touches ($\overline{X} = 3.8$) and releases ($\overline{X} = 1.7$) per minute. Central midfielders 40 performed the most ball touches ($\overline{X} = 221.9$), releases ($\overline{X} = 108.3$), ball touches per minute ($\overline{X} = 3.4$) and 41 releases per minute ($\overline{X} = 1.6$). Small-sided games evoked more ball touches ($\overline{X}^{diff} = 1.5$) and releases per 42 minute ($\overline{X}^{diff} = 0.1$) than previously reported in match-play. The fewest ball touches ($\overline{X} = 1.2$) and releases 43 per minute ($\overline{X} = 0.5$) were observed during tactical drills. The results of this study provide a novel 44 45 understanding of the within-microcycle, inter-positional and drill category differences in the technical 46 actions performed by professional players during training.

47

48 Keywords: Professional Soccer Training, Technical Actions, Monitoring, Microcycle,
49 Microtechnology

50 Introduction

51 The multifactorial demands of professional soccer require the implementation of training programmes 52 that combine technical, tactical, physical and psychological components to enhance player performance 53 (Stølen et al., 2005). Technical (i.e. ball touches, passes, crosses, shots) and tactical components are 54 often prioritised by coaches during in-season training (Morgans et al., 2014), due to their association 55 with competition success (Castellano et al., 2012; Carling, 2013). Farrow and Robertson's (2017) skill 56 acquisition periodisation framework has enabled practitioners to systematically adjust players' 57 technical performance in training throughout numerous sports. For instance, soccer specificity may be 58 enhanced by comparing the extent that training mimics the technical demands of competition (Pinder 59 et al., 2011), with progression expedited by prescribing an increased frequency of technical actions 60 (Ericsson et al., 1993). However, technical actions are consistently neglected by practitioners during 61 player monitoring processes (Akenhead & Nassis, 2016; Malone et al., 2020), despite contributing to 62 players' overall external training load (Bradley & Ade, 2018).

63 The monitoring of technical actions is pertinent because the frequency of these actions executed by professional players during match-play has risen over time. Barnes et al. (2014) reported that the 64 65 frequency of ball touches and passes executed by English Premier League players increased, by 10.5% 66 and 39.9% respectively, over seven consecutive seasons. In the 2019/2020 season, Union of European 67 Football Associations Champions League players typically performed 60.2 ± 20.7 ball touches and 50.168 ± 25.7 releases (i.e. passes, crosses, shots, clearances) per match (Yi et al., 2020). Despite such insights 69 into match-play, examinations of technical actions during training scarcely appear within the literature 70 (Liu et al., 2016; Bradley & Ade, 2018). Quantifying technical actions often requires complex and 71 expensive infrastructure, such as semi-automated multiple camera tracking systems (e.g. ProZone®, 72 Castellano et al., 2014) or local positioning systems (e.g. Inmotio, Frencken et al., 2010; Kinexon, 73 Hoppe et al., 2018). Although these systems provide data that contextualises the multifaceted 74 determinants of player performance (Bradley & Ade, 2018), the significant financial investment 75 required hinders the transferability of such methods to the training environment (Akenhead & Nassis, 76 2016; Cardinale & Varley, 2017). In this setting, manual coding has been the prominent approach to 77 assessing players' technical performance (Wright et al., 2013), which not only quantifies the specific

78 technical actions performed (Wright et al., 2016), but also provides an understanding of players' pitch 79 location (Taylor et al., 2010) and associated action success (Bateman & Jones, 2019). Yet, this process 80 needs highly trained operators to limit measurement error (O'Donoghue, 2007) and to achieve sufficient 81 validity and reliability (Francis et al., 2019; Gong et al., 2019). Moreover, the substantial human 82 resources required has compelled practitioners to explore alternative approaches for quantifying 83 technical actions during training (Carling et al., 2014; Robertson, 2020). As a solution to these 84 problems, the implementation of wearable microtechnology, attached to players' boots (Edwards et al., 85 2019), may represent a time- and cost-efficient option for monitoring technical actions during weekly 86 training microcycles (Chambers et al., 2015; Nedergaard et al., 2017). However, while the efficacy of 87 such technology has been established from a time-motion perspective (Waldron et al., 2020), the 88 validity, reliability, and applicability of this method for quantifying technical actions remains to be 89 explored.

90 Quantifying technical actions during training would provide a broader understanding of the 91 periodisation strategies used to prepare professional players for competition. Throughout a typical 92 microcycle, external training load markers (e.g. total distance, mean speed) are consistently at their 93 lowest on the day immediately before competition (Anderson et al., 2015; Malone et al., 2015; Stevens 94 et al., 2017; Martín-García et al., 2018), with practitioners adopting this tapering approach to physically 95 unload players and increase readiness for competition (Malone et al., 2015; Owen et al., 2017). 96 Anecdotally, this unloading coincides with training becoming more technically and tactically oriented 97 (Martín-García et al., 2018; Walker & Hawkins, 2018). However, empirical evidence to support this is 98 lacking, necessitating an examination of the technical actions performed throughout professional soccer 99 training microcycles.

100 The periodisation of technical actions provides a macro view of training. However, the within-101 session distribution of technical actions also warrants attention. Despite numerous studies examining 102 technical actions during specific training drills (e.g. small-sided games) in isolation (Fradua et al., 2013; 103 Aguiar et al., 2015), little consideration has been given to the effect of drill category on the technical 104 actions executed by professional players (Barrett et al., 2020). Understanding these effects would allow 105 practitioners to manipulate players' technical actions to satisfy the aforementioned principles of skill 106 acquisition periodisation (Farrow & Robertson, 2017). Furthermore, to facilitate evidence-based 107 decisions regarding the inclusion of drills to achieve training objectives (Jaspers et al., 2017), and to 108 supplement coaches' feedback by providing objective insights into players' technical actions (Stodter 109 & Cushion; 2019; Nosek et al., 2021), the aims of the current study were to: (i) establish the concurrent 110 validity and intra-unit reliability of a foot-mounted inertial measurement unit for monitoring soccer 111 technical actions, (ii) quantify the within-microcycle inter-positional differences in the technical actions 112 of professional soccer training, and (iii) determine the influence of drill category on the technical actions 113 of professional soccer training.

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114 Methods

115 Validity and Reliability

116 Prior to the quantification of technical actions during training, the concurrent validity and intra-unit 117 reliability of a foot-mounted inertial measurement unit (IMU) was established. Twelve amateur soccer 118 players (mean \pm standard deviation (SD) age: 23.8 \pm 5.2 years; stature: 179.9 \pm 5.3 cm; body mass: 85.1 119 \pm 19.5 kg) collectively performed 8,640 ball touches and 5,760 releases, throughout a series of technical 120 soccer tasks, repeated over two pre-determined distances (Supplementary Table 1). The 1000 Hz IMU 121 microprocessor utilised zero-lag Butterworth and Kalman filters to transform raw accelerometer and 122 gyroscope traces, via proprietary soccer-specific event detection algorithms (Waldron et al., 2020), 123 prior to data being exported from the manufacturer's cloud-based software into Excel (v. 16.45; 124 Microsoft®, Redmond, USA) for analysis.

125 Concurrent validity was determined by calculating the proportion of agreement (P_A) between 126 the IMU data and retrospective video analyses (SportsCode Elite, v. 11.2.23, SportsTec, Warriewood, 127 Australia) (Cooper et al., 2007). Intra-operator reliability of the reference performance analyst, who 128 conducted all analyses, was established by coding three randomly selected repetitions of each soccer 129 task three times ($P_A = 100.0\%$). Intra-unit reliability was established using the same method (Cooper et 130 al., 2007), supplemented by a percentage coefficient of variation (CV), calculated as the standard 131 deviation of the between trial difference divided by the mean between trial difference. PA and CV values 132 were appraised as good (< 5.0%), moderate (5.0% - 10.0%) or poor (> 10.0%) (Scott et al., 2016). The 133 IMUs displayed good P_A (95.1% - 100%) and good CV (1.4% - 2.9%) values for measuring ball touches 134 and releases throughout all experimental conditions (Table 1).

135

136 *****TABLE 1 ABOUT HERE*****

137

138 Experimental Design

Technical actions were quantified during training sessions throughout a twenty-four-week (September
to February) mid-season period of the 2019/2020 English Football League Championship season
(Supplementary Figure 1), prior to competition disruption (Football Association, 2020a). This phase

142 ensured minimal changes to players' physiological fitness, such as those that typically occur during the 143 transition from pre-season to in-season, where coaches emphasise the continuation of physical 144 conditioning (Malone et al., 2015). Two microcycles were excluded as they fell within the Fédération 145 Internationale de Football Association International Match Calendar (Malone et al., 2015; Stevens et 146 al., 2017). Training sessions within one microcycle were categorised in relation to the number of days prior to a competitive fixture (i.e. match-day (MD) minus day number [MD - #]) (Malone et al., 2015). 147 148 Microcycles encompassing one fixture (n = 13, 54.2%) typically contained four training sessions, with 149 MD - 3 being a recovery day for all players. Fixtures were followed by a recovery day for all players. 150 According to their primary objective, training drills were assigned one of the following categories: 151 position specific; possession; small-sided games (SSG); tactical; technical; or warm-up (Supplementary 152 Table 2).

153

154 Exclusion Criteria

Players were required to have completed three full pitch-based sessions on each training day, and three repetitions of each drill, to facilitate comparative analyses. This resulted in twenty-seven players (e.g. academy scholars, trialists, players transferred in/out) being removed from the dataset through their intermittent involvement during the training programme. For eligible players, the twenty-four-week data collection period yielded 8,535 drill observations. 9.3% (n = 796) of these were removed having imposed various exclusion criteria (Figure 1) derived from comparable longitudinal monitoring studies (Malone et al., 2015; Stevens et al., 2017).

162

163 *****FIGURE 1 ABOUT HERE*****

164

A total of sixty-six training sessions, comprising 7,739 individual player observations, were included for analysis. Players completed a mean of 47.7 ± 13.2 training sessions, with 7.4 ± 2.1 drill observations per session. Sessions had a ball-in-play time of 61.8 ± 5.5 minutes, with recovery periods removed to provide an accurate representation of training intensity (Wass et al., 2020). Each player completed 351.8 ± 98.1 drills during the study, which did not influence the training content delivered.

170 Participants

171 Twenty-one professional soccer players (mean \pm SD age: 24.4 \pm 3.1 years; stature: 183.0 \pm 8.1 cm; 172 mass: 80.6 ± 9.6 kg), from one English Football League Championship club, participated in this study. 173 The sample size was constrained by the finite number of players with professional contracts, who were 174 available to participate in training, that satisfied the aforementioned exclusion criteria. As categorised 175 by the head coach, who typically employed a 4-2-3-1 formation, the sample of players comprised five 176 central defenders (CD), five wide defenders (WD), six central midfielders (CM), three wide midfielders 177 (WM) and two strikers (ST). The head coach and coaching staff remained consistent throughout, 178 alleviating the potential influence of a change in head coach on the technical requirements of the training 179 programme (Whitehead et al., 2018). This study obtained institutional ethical approval (FHS200), with 180 data collected as part of daily player monitoring procedures.

181

182 Inertial Measurement Units

Technical actions were quantified using commercially available foot-mounted IMUs (PlayerMakerTM, Tel Aviv, Israel). Each IMU incorporated two components from the MPU-9150 multi-chip motion tracking module (InvenSense, California, USA), being a 16 g triaxial accelerometer and a 2000°•sec⁻¹ triaxial gyroscope. Housed in manufacturer-supplied tightly-fitting silicone straps, each player was equipped with two IMUs (one for each foot), which were located at the lateral malleoli over the player's boots. To diminish issues related to inter-unit reliability, players used the same IMUs throughout the data collection period (Buchheit et al., 2014; Malone et al., 2020).

190

191 Statistical Analysis

Having verified the assumption of normality using a Q-Q plot (Schielzeth et al., 2020), general linear modelling was conducted within SPSS (v. 26; IBM, Chicago, USA) to establish estimated marginal mean values for the four fixed variables of interest: ball touches, releases, ball touches per minute, releases per minute. Random variables (e.g. player age, calendar month) were screened for covariance (Hopkins & Wolfinger, 1998), with Wald Z statistics (p > 0.05) indicating that no random intercept was required. In the event of a statistically significant *F* ratio, Sidak adjusted post-hoc pairwise comparisons

- 198 between the estimated marginal means were analysed. Cohen's *d* effect size (ES) statistics, using the
- 199 pooled standard deviation as the denominator, was computed to ascertain the magnitude of the within-
- 200 microcycle, inter-positional and drill category differences, with the following descriptors attached:
- 201 trivial (< 0.20); small (> 0.21 0.60); moderate (> 0.61 1.20); large (> 1.21 2.00); very large (> 2.01)
- 202 (Hopkins et al., 2009). Two-tailed statistical significance was established as p < 0.05.

203	Results
204	Fixture Proximity
205	There were main effects of fixture proximity on the frequency of technical actions ($F_{(4, 1019)} = 1705.05$
206	- 2026.17, $p < 0.001$; ES = 0.01 - 0.89), inter-positional differences in the absolute frequency of ball
207	touches and releases ($F_{(20, 1003)} = 347.19 - 416.34$, $p < 0.001$; ES = 0.00 - 0.83) (Figure 2) and the relative
208	frequency of ball touches and releases per minute ($F_{(20, 1003)} = 361.10 - 446.99, p < 0.001$; ES = 0.01 -
209	0.73) (Figure 3).
210	
211	***TABLE 2 ABOUT HERE***
212	***FIGURE 2 ABOUT HERE***
213	***FIGURE 3 ABOUT HERE***
214	
215	Playing Position
216	There were main effects of playing position ($F_{(5, 1018)} = 1301.82 - 1697.79, p < 0.001$; ES = 0.01 - 0.64)
217	on the frequency of ball touches, releases, ball touches per minute and releases per minute performed
218	during a typical training session (Table 3).
219	
220	***TABLE 3 ABOUT HERE***
221	
222	Drill Category
223	There were main effects of drill category ($F_{(6, 7728)} = 3801.45 - 4314.05, p < 0.001$; ES = 0.21 - 4.35)
224	on the frequency of ball touches and releases (Table 4), on the relative frequency of ball touches and
225	releases per minute ($F_{(6, 7728)} = 3709.50 - 4929.72$, $p < 0.001$, ES = 0.04 - 3.04) (Figure 4), and on the
226	inter-positional differences in the relative frequency of ball touches and releases per minute ($F_{(20, 1003)}$
227	= 361.10 - 446.99, <i>p</i> < 0.001; ES = 0.00 - 0.43) (Figure 5).
228	
229	***TABLE 4 ABOUT HERE***
230	***FIGURE 4 ABOUT HERE***

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*****FIGURE 5 ABOUT HERE*****

232 **Discussion**

233 The primary findings of this study were: (i) the foot-mounted IMU displayed consistently good 234 concurrent validity and intra-unit reliability for measuring ball touches and releases, (ii) players 235 typically performed the most ball touches and releases on MD - 1, (iii) training sessions on MD - 5 236 elicited the most ball touches and releases per minute, (vi) central midfielders generally performed the 237 highest frequency of ball touches, releases, ball touches per minute and releases per minute, (v) the 238 specificity of small-sided games for replicating the positional technical demands of match-play may be 239 limited, and (vi) regardless of playing position, the fewest ball touches per minute and releases per 240 minute were observed during tactical drills.

241 The foot-mounted IMU examined during this study displayed good concurrent validity and 242 intra-unit reliability for measuring ball touches and releases (Table 1). Such technological 243 advancements that quantify sport-specific non-locomotor activities may benefit practitioners (Lutz et 244 al., 2020), by supplementing player monitoring procedures with the integration of technical data 245 (Malone et al., 2020). The relative ease of implementation of foot-mounted IMUs (Starling & Lambert, 246 2018) provided a time-efficient automated alternative to laborious manual coding (Carling et al., 2014; 247 Robertson 2020), with the efficacy of the IMUs negating concerns regarding human measurement error 248 (O'Donoghue, 2007).

249 Previous research has demonstrated that players exhibit the lowest external training load on the 250 day immediately preceding competition (Anderson et al., 2015; Malone et al., 2015; Stevens et al., 251 2017; Martín-García et al., 2018). Conversely, the current study noted that the frequency of technical 252 actions performed during a typical microcycle peaked on MD - 1 (Table 2), which supports the notion 253 that, to physically unload players as competition approaches (Malone et al., 2015; Owen et al., 2017), 254 training objectives become more technical and tactical in nature (Martín-García et al., 2018; Walker & 255 Hawkins, 2018). It would appear that the coaches sought to facilitate this pre-competition physical 256 unloading by prescribing a greater proportion of position specific, tactical and warm-up drills on this 257 day (Supplementary Table 4), with such drills demonstrating significantly lower external training load 258 markers than small-sided games (Barrett et al., 2020). However, training sessions on MD - 1 resulted 259 in the average player performing almost four times the frequency of ball touches, and more than double 260 the frequency of releases, compared to previously reported match-play data from semi-automated 261 multiple camera tracking systems (Yi et al., 2020). Although previous research has emphasised caution 262 when comparing data from different monitoring systems (Buchheit et al., 2014; Taberner et al., 2020), 263 the foot-mounted IMU is not currently permitted during match-play under the Laws of the Game (Law 264 4.4; International Football Association Board, 2020). Therefore, albeit tentatively, the current study 265 begins to question whether players' readiness for the impending fixture may have been inadvertently 266 compromised (Anderson et al., 2016; Kelly et al., 2020), given the potential for neuromuscular fatigue 267 attributed to the heightened frequency of technical actions performed (Guex & Millet, 2013, Silva et 268 al., 2018). Nevertheless, the IMU could not differentiate between the types of release performed, nor 269 did the current study examine players' shank angular velocity during kicking (Lees et al., 2010), which 270 has demonstrated fatigue-related decrements (Ferraz et al., 2012; 2019). Future research considering 271 the magnitude of players' releases may, therefore, provide an insight into the metabolic cost 272 implications of performing specific technical actions (Osgnach et al., 2010; Russell et al., 2011; Walker 273 et al., 2016). By understanding the resulting biomechanical load imposed on the musculoskeletal system 274 pre-competition (Vanrenterghem et al., 2017), and associated mechanobiological response (Wisdom et 275 al., 2015), practitioners would be better placed to gauge players' holistic readiness to perform in 276 conjunction with current monitoring systems (Bradley & Ade, 2018; Verheul et al., 2020).

Relative to ball-in-play time, training sessions on MD - 5 elicited the most ball touches and 277 278 releases per minute. The greatest proportion of technical drills was also observed on MD - 5 279 (Supplementary Table 4), perhaps delivered in an attempt to compensate players for the lack of technical 280 stimuli through not participating in competition (Morgans et al., 2018). Although this study did not 281 account for levels of match participation, which has demonstrated large effects on external training load 282 markers (Anderson et al., 2015), the training group on MD - 5 often comprised non-starting and fringe 283 players, with those who started the previous fixture performing recovery activities (Morgans et al., 284 2014; Anderson et al., 2016). Practitioners frequently prescribe 'top-up' training immediately after a 285 fixture to atone for the insufficient external training load encountered by partial-match and unused 286 substitute players (Hills et al., 2018; Buchheit, 2019; Buckthorpe et al., 2019). However, such training 287 is solely physical in nature, with players rarely exposed to supplementary technical activities (Hills et 288 al., 2020a). This may be due to governing body pitch-usage restrictions permitting only fifteen minutes 289 of post-match activity (Rule 23.11i, Football Association, 2020b), team travel requirements (Hills et 290 al., 2020b) or a lack of available coaching staff (Hills et al., 2020b). As such, it would appear that 291 coaches attempt to limit the consequences of reversibility (Farrow & Robertson, 2017), by utilising 292 technical drills on MD - 5 to provide non-starting and fringe players with sufficient perceptual-cognitive 293 stimuli that is crucial for technical performance (Reilly et al., 2000; Williams & Hodges, 2005). 294 Nonetheless, the alternative tactical systems (Whitehead et al., 2018) and within-microcycle schedules 295 (Malone et al., 2015) employed by coaches may influence technical performance during specific 296 training programmes, limiting the generalisability of these results (Dalton-Barron et al., 2020).

297 Inter-positional differences in the technical actions of match-play are well documented within 298 the literature (Ade et al., 2016; Baptista et al., 2018). However, prior to the current investigation, 299 research examining these differences in the training environment were scarce. This study reported 300 trivial-to-moderate (ES = 0.01 - 0.64) inter-positional differences in the technical actions of professional 301 soccer training, with central midfielders performing the most absolute and relative ball touches and 302 releases during a typical training session (Table 3). This suggests that the technical actions performed 303 by CM during training are somewhat specific to those experienced during competition (Farrow & 304 Robertson, 2017), with this position typically performing the most ball touches and releases per match 305 (Yi et al., 2020). This is likely related to the tactical responsibilities of CM (Dellal et al., 2011), which 306 primarily involves coordinating attacking play and creating goal scoring opportunities (Gonçalves et 307 al., 2014; Bush et al., 2015). For instance, regardless of match status, 61.0% of passes originate from 308 the midfield third of the pitch (Taylor et al., 2010), likely contributing to central midfielders 309 demonstrating trivial differences in frequency of technical actions between playing at home versus 310 playing away, and trivial differences when playing against a higher quality of opposition (Yi et al., 311 2020). This highlights the importance of training specificity for CM (Farrow & Robertson, 2017), given 312 the apparent stability of the technical actions performed by this position during match-play.

Small-sided games are routinely employed in professional soccer training (Hill-Hass et al.,
2011), simultaneously providing players with technical, tactical, physical and psychological stimuli
similar to that encountered during competition (Halouani et al., 2014; Bujalance-Moreno et al., 2019).

316 Indeed, SSG accounted for 49.5% of individual drill observations during the current investigation, with 317 the highest proportion being observed on MD - 1 (Supplementary Table 4). The trivial-to-small (ES = 318 0.00 - 0.35) inter-positional differences in the frequency of ball touches and releases per minute 319 observed during SSG would imply that the specificity of these drills for replicating the inter-positional 320 technical actions of match-play may be limited (Farrow & Robertson, 2017). For example, the only 321 small differences during SSG were observed within CM, who performed more ball touches than CD 322 and ST. For all playing positions, SSG during training evoked more ball touches and releases per minute 323 than match-play (Yi et al., 2020), suggesting that SSG may facilitate progression through the elevated 324 frequency of technical actions performed (Farrow & Robertson, 2017), alongside the concurrent 325 decision-making and perceptual demands of these drills (Sampaio & Maçãs, 2012; Aguiar et al., 2015). 326 However, comparisons between training and match-play should be interpreted with caution, given the 327 problematic nature of quantifying performance with different systems in different environments 328 (Buchheit et al., 2014). Future research should explore the agreement between the foot-mounted IMUs 329 and semi-automated multiple camera tracking systems, to determine whether these approaches can be 330 used interchangeably throughout training and match-play (Taberner et al., 2020).

331 Drill category displayed trivial-to-very large (ES = 0.04 - 3.04) effects on the relative frequency 332 of technical actions performed during training (Table 4). Tactical drills (e.g. team shape, set pieces) are 333 arguably the most important training modality in professional soccer, with players' tactical roles being 334 a powerful determinant of match performance (Bradley & Ade, 2018). This study observed that, for all 335 positions, the fewest ball touches and releases per minute were observed during tactical drills. These 336 drills are intermittent in nature (Siegle & Lames, 2012), with coaches frequently interrupting to provide 337 instruction and management-related information (Ford et al., 2010). Although instruction and 338 management are crucial for delivering tactical messages (Cushion & Jones, 2001), previous research 339 has demonstrated the potential issues related to interrupting practice too frequently (Williams & 340 Hodges, 2005), which perhaps contributed to the lowest relative technical stimuli being provided by 341 tactical drills. Therefore, practitioners should seek alternative exercise modalities, such as incorporating 342 technical actions within warm-up drills, should a high technical output be required from a particular 343 session.

344 The application of foot-mounted IMUs is not unproblematic. Firstly, foot-mounted IMUs lack 345 the contextual information provided by manual coding (i.e. player location, action success/failure). 346 Practitioners will need to weigh up the time-saving ability of the IMUs, against the reduced contextual 347 information provided, when considering the implementation of this solution. Secondly, soccer players 348 often perform technical actions with alternative body parts (e.g. thigh, head) (Bloomfield et al., 2007), 349 which a foot-mounted device is unlikely to detect. Given that underreporting the frequency of technical 350 actions performed may have implications for skill acquisition periodisation (Farrow & Roberton, 2017), 351 practitioners should account for the disparity between the frequency of technical actions measured by 352 foot-mounted IMUs and those performed with alternative body parts when programming players' 353 technical actions during training. The inability of the IMU to quantify specific types of release 354 performed, which likely possess distinct metabolic requirements (Osgnach et al., 2010; Walker et al., 355 2016), may limit the subsequent usefulness of the monitoring system (Taberner et al., 2020). 356 Researchers might seek to develop a cumulative vector magnitude index of technical 'load', 357 incorporating players' shank angular velocity alongside the specific types of release performed (Lees 358 et al., 2010), to provide practitioners with an arbitrary value representative of the metabolic cost of 359 players' technical performance (Boyd et al., 2013; Barrett et al., 2015; Dalen et al., 2016). While the 360 current study provided an examination of the capabilities of the foot-mounted IMUs during controlled, 361 discrete, technical soccer tasks that are commonly observed during training, the validity and reliability 362 of the IMUs is yet to be comprehensively examined during match-play. The increasingly complex 363 scenarios observed during competition may involve instances of foot-to-foot contact between players, 364 potentially resulting in additional false positive ball touches and releases being recorded (Rossi et al., 365 2018). Further research establishing the efficacy of the foot-mounted IMU during competitive match-366 play scenarios, such as when physically contesting for possession of the ball (Drust et al., 2007), is 367 required.

This investigation has provided a novel understanding of within-microcycle, inter-positional and drill category differences in the technical actions of professional soccer training, which may be especially relevant to researchers and practitioners alike. Although the magnitude of players' releases was not quantified, which may yield a broader understanding of the metabolic and mechanobiological 372 implications of performing technical actions (Walker et al., 2016; Vanrenterghem et al., 2017), the 373 insights provided by this investigation have the potential to inform pre-competition recovery strategies 374 to negate the neuromuscular fatigue possibly induced through increased technical activity on MD - 1 375 (Rey et al., 2012a; 2012b; Nédélec et al., 2015). Lastly, coaches could manipulate the frequency (e.g. 376 CM performing additional releases during possession drills), and complexity (ST required to hit certain 377 zones during position-specific drills) of technical actions during training, to provide an optimal 378 challenge point that enhances the positional specificity and promotes progression according to the 379 principles of skill acquisition periodisation (Guadagnoli & Lee, 2004; Farrow & Robertson, 2017; 380 Mujika et al., 2018).

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- 383 and assistance during this study.

384 Declaration of Interest Statement

- 385 The second author of this study is now employed by the company who provided the foot-mounted IMUs
- 386 used to collect players' technical performance data. However, throughout the data collection and writing
- 387 up periods, the second author had the same affiliation as the first author and was not involved with the
- 388 company in any way. Despite this, and to remove the potential for bias, the second author was not
- involved in any statistical analyses or data interpretation conducted during the investigation.

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Tables with Captions

Table 1: A summary of the concurrent validity and intra-unit reliability of the foot-mounted IMUs for quantifying ball touches and releases throughout all experimental conditions.

		Concurre	ent Validi	ty		Intra-Unit Reliability								
Variable	SportsCode Mean ± SD	PlayerMaker™ Mean ± SD	P _A (%)	SE (%)	P _A (%) 95% CI	PlayerMaker™ Between Trial Mean ± SD Difference	P _A (%)	SE (%)	P _A (%) 95% CI	CV (%)				
Ball Touches	30.0 ± 0.0	29.9 ± 0.5	95.1	0.1	95.0 - 95.3	0.0 ± 0.4	96.9	0.0	96.8 - 96.9	1.8				
Releases	20.0 ± 0.0	20.0 ± 0.5	97.6	0.0	97.5 - 97.7	0.0 ± 0.4	95.9	0.2	95.5 - 96.2	2.3				

 $\textbf{N.B. SD} = standard \ deviation. \ P_A = proportion \ of \ agreement. \ SE = standard \ error. \ CI = confidence \ intervals. \ CV = coefficient \ of \ variation.$

Table 2: Estimated marginal mean (\pm SE) values representative of the absolute frequency of ball touches and releases, and the relative frequency of ball touches and	
releases per minute of ball-in-play time, performed by professional soccer players on each training day within a typical weekly microcycle.	

Variable	0	VERAL	L	MD - 5				MD - 4]		MD - 1			
Variable	EM Mean	SE	95% CI	EM Mean	SE	95% CI	EM Mean	SE	95% CI	EM Mean	SE	95% CI	EM Mean	SE	95% CI
Duration (mins)	62.5	0.8	61.0 - 64.1	41.5 ${}_{4}^{L}{}_{2}{}_{1}^{L}{}_{1}^{M}$	2.1	37.3 - 45.7	76.6 5 ^L 1 ^M	1.8	73.1 - 80.1	70.8 5 ^L 1 ^S	1.4	68.1 - 73.5	57.8 5 ^M 4 ^M 2 ^S	1.1	55.7 - 59.9
Ball Touches (f)	209.9	2.4	205.3 - 214.6	181.9 4 ⁸ 2 ⁸ 1 ⁸	7.0	168.2 - 195.7	209.0 5 ⁸	5.9	197.4 - 220.5	208.4 5 ⁸	4.4	199.7 - 217.1	218.0 5 ⁸	3.5	211.2 - 224.8
Releases (f)	103.0	1.3	100.4 - 105.5	${}^{80.9}_{4^{\rm S}2^{\rm S}1^{\rm M}}$	3.8	73.6 - 88.3	99.7 5 ⁸ 1 ⁸	3.2	93.5 - 105.9	100.6 5 ⁸ 1 ⁸	2.4	95.9 - 105.2	110.8 $5^{M}4^{S}2^{S}$	1.9	107.2 - 114.5
Ball Touches (f·min ⁻¹)	3.1	0.0	3.1 - 3.2	3.8 4 ^M 2 ^M 1 ^M	0.1	3.6 - 4.0	2.8 5 ^M 2 ^S 1 ^S	0.1	2.6 - 2.9	3.1 5 ^M 4 ^S	0.1	3.0 - 3.2	3.1 5 ^M 4 ^S	0.1	3.0 - 3.2
Releases (f·min ⁻¹)	1.5	0.0	1.4 - 1.5	1.7 4 ^M 1 ^S	0.1	1.5 - 1.8	1.3 5 ^M 2 ^S 1 ^S	0.0	1.2 - 1.4	1.5 4 ⁸	0.0	1.4 - 1.6	1.5 5 ⁸ 4 ⁸	0.0	1.4 - 1.5

N.B. EM = estimated marginal. SE = standard error. CI = confidence intervals. MD = match day. Statistically significant differences (p < 0.05) are depicted in bold: **5** = MD -

5, 4 = MD - 4, 2 = MD - 2, and 1 = MD - 1. Observed magnitude of effects are denoted as $[^{T}] = trivial$, $[^{S}] = small$, $[^{M}] = moderate$, $[^{L}] = large$, and $[^{V}] = very large$.

Table 3: Estimated marginal mean (\pm SE) values representative of the inter-positional differences in the absolute frequency of ball touches and releases, and the relative frequency of ball touches and releases per minute of ball-in-play time, performed by professional soccer players during a typical training session.

	CENTRA	ENTRAL DEFENDERS			WIDE DEFENDERS			CENTRAL MIDFIELDERS				CLDERS	STRIKERS			
Variable	EM Mean	SE	95% CI	EM Mean	SE 95% CI		EM Mean	SE	95% CI	EM Mean	SE	95% CI	EM Mean	SE	95% CI	
Ball Touches (f)	206.1	4.8	196.7 - 215.4	200.9 см ^s	4.8	191.5 - 210.2	221.9 wd ^s st ^s	4.4	213.3 - 230.4	218.3	6.4	205.6 - 230.9	195.4 см ⁸	7.5	180.7 - 210.2	
Releases (f)	102.2	2.6	97.1 - 107.3	97.5 см ⁸	2.6	92.4 - 102.6	108.3 wd ^s	2.4	103.6 - 112.9	106.9	3.5	100.0 - 113.8	97.2	4.1	89.2 - 105.3	
Ball Touches (f ⋅ min ⁻¹)	2.9 wd ^s см ^s	0.1	2.7 - 3.0	3.2 cd ^s st ^s	0.1	3.0 - 3.3	3.4 cd ^s st ^m	0.1	3.3 - 3.5	3.2 st ^s	0.1	3.0 - 3.4	2.7 wd ^s см ^m wm ^s	0.1	2.5 - 2.9	
Releases (f·min ⁻¹)	1.4 см ⁸	0.0	1.3 - 1.5	1.5 st ^s	0.0	1.4 - 1.6	1.6 cd ^s st ^s	0.0	1.5 - 1.7	1.4	0.0	1.3 - 1.5	1.3 wd ^s см ^s	0.1	1.2 - 1.4	

N.B. EM = estimated marginal. SE = standard error. CI = confidence intervals. Statistically significant differences (p < 0.05) are depicted in bold: **CD** = central defenders, **WD** = wide defenders, **CM** = central midfielders, **WM** = wide midfielders, and **ST** = strikers. Observed magnitude of effects are denoted as $[^{T}]$ = trivial, $[^{S}]$ = small, $[^{M}]$ = moderate, $[^{L}]$ = large, and $[^{V}]$ = very large. **Table 4:** Estimated marginal mean (\pm SE) values representative of the differences in the absolute frequency of ball touches and releases, and the relative frequency of ball touches and releases per minute of ball-in-play time, performed by professional soccer players throughout each category of training drill.

Variable	POSITION	IC	POSSESSION			SMALL-SIDED GAMES			TAC		TECH		WARM-UP					
variable	EM Mean	SE	95% CI	EM Mean	SE	95% CI	EM Mean	SE	95% CI	EM Mean	SE	95% CI	EM Mean	SE	95% CI	EM Mean	SE	95% CI
Duration (mins)	27.0 pos ^v ssg ^v tac ^v tec ^v wu ^v	0.2	26.6 - 27.5	12.7 PS ^V SSG ^L TAC ^S TEC ^M WU ^S	0.2	12.3 - 13.2	2.7 ps ^v pos ^l tac ^l tec ^l wu ^v	0.1	2.5 - 2.8	$10.3 \\ {}_{\text{PS}^{V} \text{ pos}^{S} \text{ ssg}^{L} \text{ tec}^{T} \text{ wu}^{M}}$	0.2	10.0 - 10.6	9.4 ps ^v pos ^s ssg ^l tac ^t wu ^m	0.2	9.1 - 9.7	15.2 ps ^v pos ^s ssg ^v tac ^m tec ^m	0.2	14.7 - 15.6
Ball Touches (f)	56.4 pos ^m ssg ^v tac ^l tec ^s wu ^v	0.9	54.6 - 58.2	30.9 PS ^M SSG ^M TAC ^M TEC ^L WU ^V	1.0	28.9 - 32.8	6.9 PS ^V POS ^M TAC ^S TEC ^V WU ^V	0.4	6.2 - 7.6	12.4 ps ^l pos ^m ssg ^s tec ^v wu ^v	0.7	11.1 - 13.6	61.2 ps ^s pos ^l ssg ^v tac ^v wu ^l	0.7	59.7 - 62.6	104.2 ps ^v pos ^v ssg ^v tac ^v tec ^L	0.9	102.4 - 106.0
Releases (f)	28.3 Pos ^l ssg ^v tac ^l tec ^s wu ^v	0.5	27.3 - 29.3	13.3 ps ^l ssg ^m tac ^m tec ^l wu ^v	0.5	12.3 - 14.4	2.8 ps ^v pos ^m tac ^s tec ^v wu ^v	0.2	2.4 - 3.2	$5.4 \\ {}_{PS^L POS^M SSG^S TEC^V WU^V}$	0.4	4.7 - 6.1	31.0 ps ^s pos ^l ssg ^v tac ^v wu ^l	0.4	30.2 - 31.8	55.2 ps ^v pos ^v ssg ^v tac ^v tec ^L	0.5	54.2 - 56.1
Ball Touches (f·min ⁻¹)	$2.4 \\ {}_{\rm TAC}{}^{\rm S}{}_{\rm TEC}{}^{\rm L}{}_{\rm WU}{}^{\rm V}$	0.1	2.2 - 2.5	2.5 ssg ^t tac ^m tec ^v wu ^v	0.1	2.3 - 2.6	$2.2 \\ \text{pos}^{T} \text{tac}^{s} \text{tec}^{v} \text{wu}^{v}$	0.0	2.1 - 2.2	1.2 PS ^M POS ^M SSG ^S TEC ^V WU ^V	0.1	1.1 - 1.3	6.2 ps ^v pos ^v ssg ^v tac ^v wu ^s	0.1	6.1 - 6.3	6.8 ps ^v pos ^v ssg ^v tac ^v tec ^s	0.1	6.6 - 6.9
Releases (f [.] min ⁻¹)	1.2 555 ⁸ tac ^m tec ^l wu ^v	0.0	1.2 - 1.3	$1.1\\\text{SSG}^{T}\text{TAC}^{S}\text{TEC}^{V}\text{WU}^{V}$	0.0	1.0 - 1.2	0.9 ps ^s pos ^s tac ^s tec ^v wu ^v	0.0	0.9 - 0.9	0.5 PS ^M POS ^S SSG ^S TEC ^V WU ^V	0.0	0.5 - 0.6	3.1 ps ^v pos ^l ssg ^v tac ^v wu ^s	0.0	3.1 - 3.2	3.6 ps ^v pos ^v ssg ^v tac ^v tec ^s	0.0	3.6 - 3.7

N.B. EM = estimated marginal. SE = standard error. CI = confidence intervals. Statistically significant differences (p < 0.05) are depicted in bold: **PS** = position specific, **POS** = possession, **SSG** = small-sided games, **TAC** = tactical, **TEC** = technical, and **WU** = warm-up. Observed magnitude of effects are denoted as $[^{T}]$ = trivial, $[^{S}]$ = small,

 $[^{\mathbf{M}}]$ = moderate, $[^{\mathbf{L}}]$ = large, and $[^{\mathbf{V}}]$ = very large.

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Figures

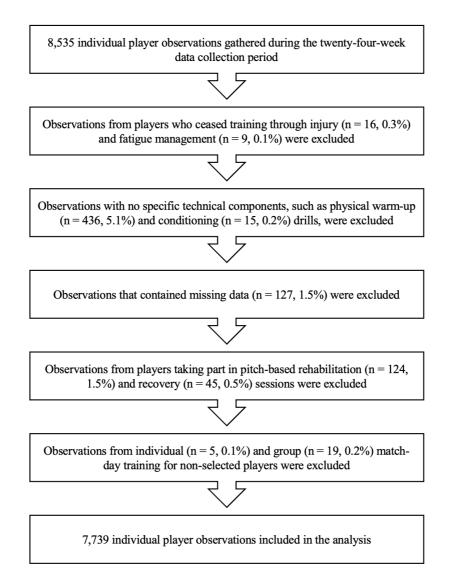


Figure 1: A schematic of the data exclusion process derived from comparable longitudinal monitoring studies (Malone et

al., 2015; Stevens et al., 2017).

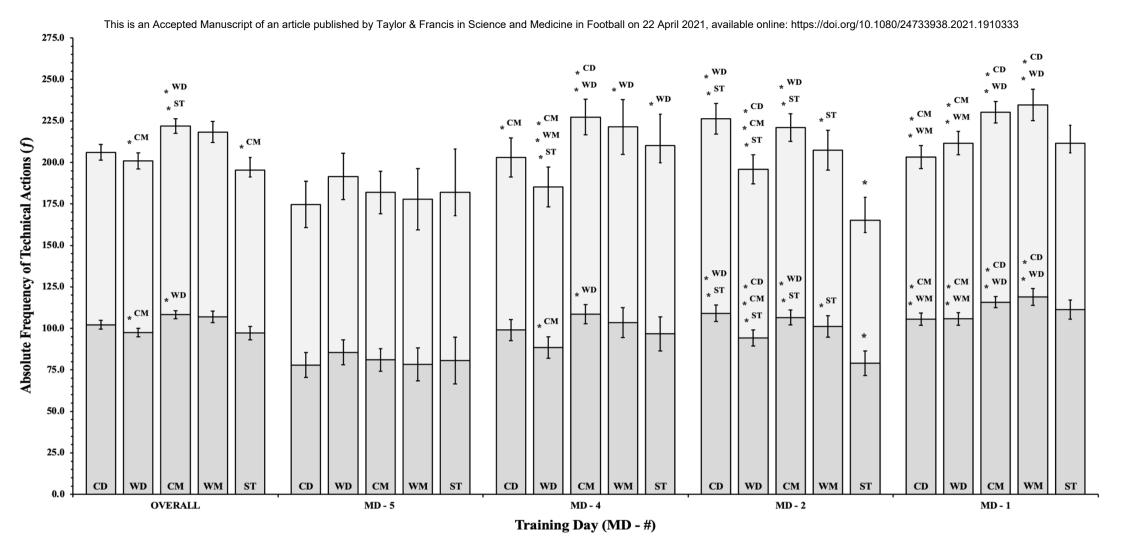


Figure 2: Estimated marginal mean (\pm SE) inter-positional differences in the absolute frequency of ball touches and releases performed by professional soccer players on each training day within a typical weekly microcycle. N.B. Each bar represents one playing position. Lighter shaded areas represent ball touches. Darker shaded areas represent releases. Statistically significant differences (p < 0.05) are displayed above SE bars. MD = match day. * = statistically significant difference to all other training days. *^{CD} = statistically significant difference to central defenders. *^{WD} = statistically significant difference to wide defenders. *^{CM} = statistically significant difference to wide midfielders. *^{CM} = statistically significant difference to strikers.

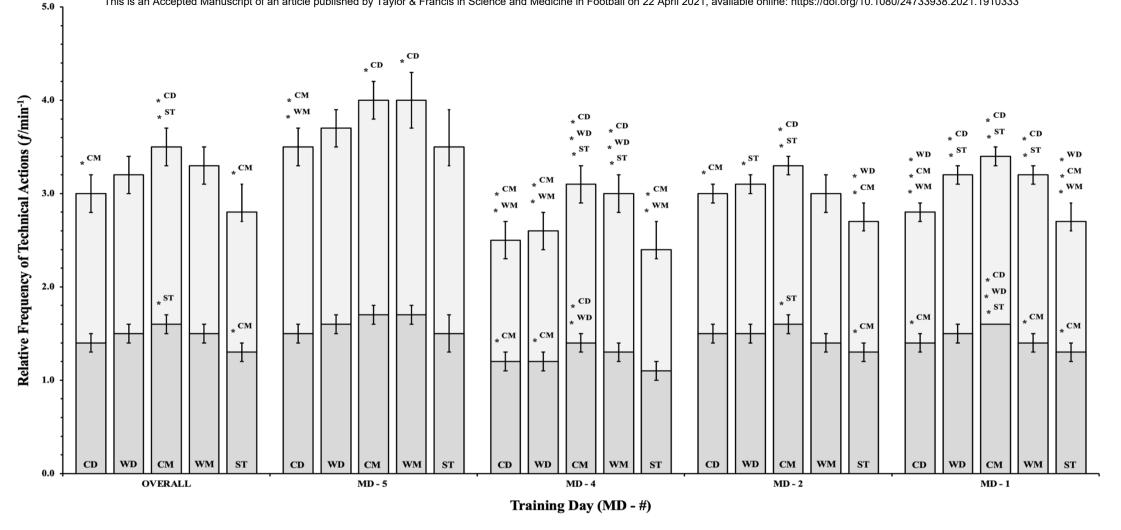


Figure 3: Estimated marginal mean (± SE) inter-positional differences in the relative frequency of ball touches and releases, per minute of ball-in-play time, performed by professional soccer players on each training day within a typical weekly microcycle. N.B. Each bar represents one playing position. Lighter shaded areas represent ball touches per minute. Darker shaded areas represent releases per minute. Statistically significant differences (p < 0.05) are displayed above SE bars. MD = match day. * = statistically significant difference to all other training days. * CD = statistically significant difference to central defenders. * WD = statistically significant difference to wide defenders. *^{CM} = statistically significant difference to central midfielders. *^{WM} = statistically significant difference to wide midfielders. *ST = statistically significant difference to strikers.

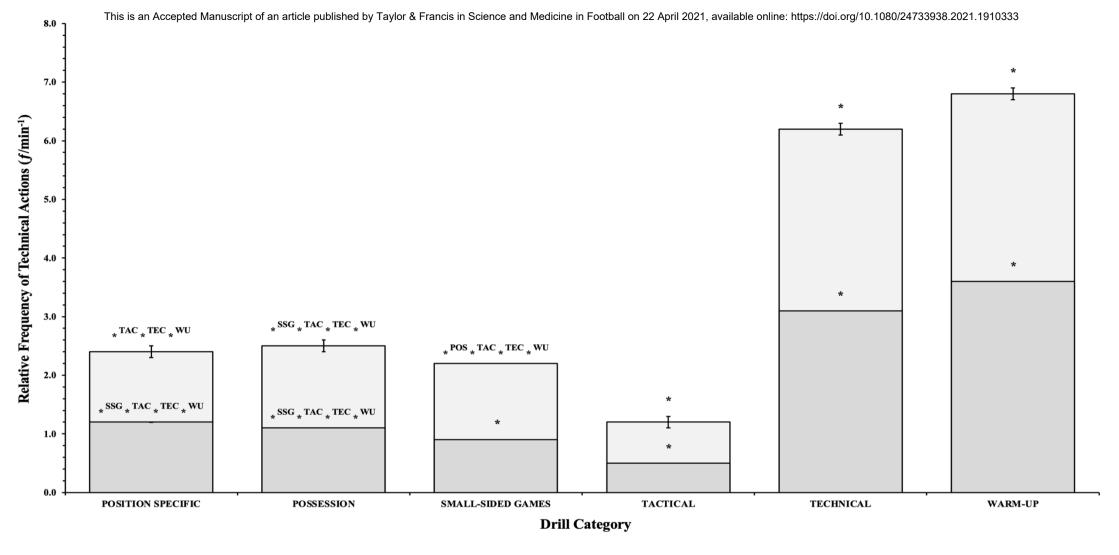
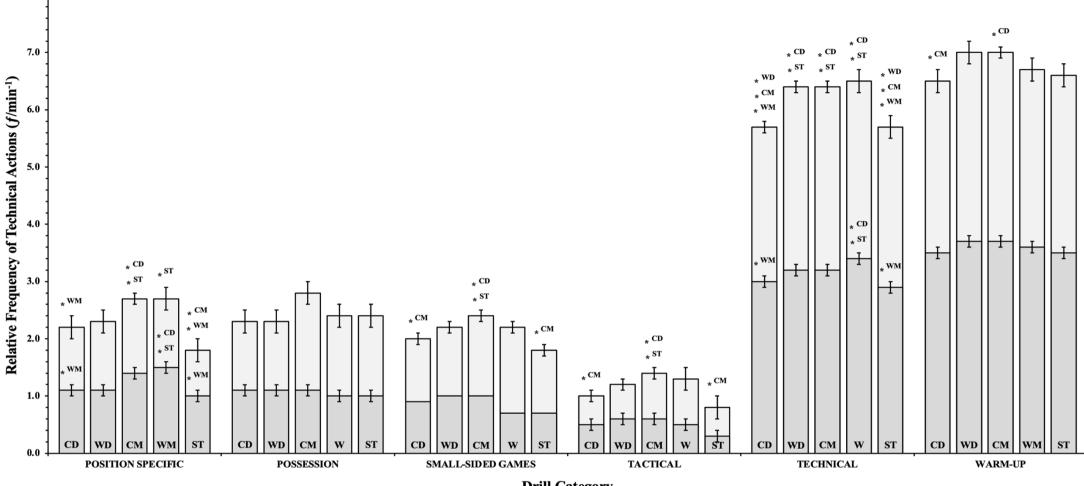


Figure 4: Estimated marginal mean (\pm SE) differences in the relative frequency of ball touches and releases, per minute of ball-in-play time, performed by professional soccer players throughout each category of training drill. N.B. Lighter shaded areas represent ball touches per minute. Darker shaded areas represent releases per minute. Statistically significant differences (p < 0.05) are displayed above SE bars. * = statistically significant difference to posterior. *^{POS} = statistically significant difference to possession. *^{SSG} = statistically significant difference to possession. *^{SSG} = statistically significant difference to warm-up.



Drill Category

Figure 5: Estimated marginal mean (\pm SE) inter-positional differences in the standardised frequency of ball touches and releases, per minute of ball-in-play time, performed by professional soccer players throughout each category of training drill. N.B. Each bar represents one playing position. Lighter shaded areas represent ball touches per minute. Darker shaded areas represent releases per minute. Statistically significant differences (p < 0.05) are displayed above SE bars. * = statistically significant difference to all other drill categories. *^{PS} = statistically significant difference to possession. *^{SSG} = statistically significant difference to small-sided games. *^{TAC} = statistically significant difference to tactical. *^{TEC} = statistically significant difference to technical. *^{WU} = statistically significant difference to warmup.

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Figure Captions

Figure 1: A schematic of the data exclusion process derived from comparable longitudinal monitoring studies (Malone et al., 2015; Stevens et al., 2017).

Figure 2: Estimated marginal mean (\pm SE) inter-positional differences in the absolute frequency of ball touches and releases performed by professional soccer players on each training day within a typical weekly microcycle. **N.B.** Each bar represents one playing position. Lighter shaded areas represent ball touches. Darker shaded areas represent releases. Statistically significant differences (p < 0.05) are displayed above SE bars. MD = match day. ***** = statistically significant difference to all other training days. *****^{CD} = statistically significant difference to central defenders. *****^{WD} = statistically significant difference to wide defenders. *****^{CM} = statistically significant difference to central midfielders. *****^{WM} = statistically significant difference to wide midfielders. *****ST = statistically significant difference to strikers.

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