

1 **Title: Planning the microcycle in elite football: to rest or not to rest?**

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32 **Abstract**

33

34 **Purpose:** To examine the association between the programming of days off (i.e., no pitch  
35 training, days off-feet) within turnarounds of varying length and injury rate in elite soccer.

36 **Methods:** Retrospective data from 56 team-seasons, belonging to 18 elite teams performing  
37 in top leagues including the EPL, the Italian Serie A, the Bundesliga, the Scottish  
38 Premiership, the MLS and the Dutch Eredivisie from January 2018 to December 2021 were  
39 analysed (total of 1578 players, 2865 injuries, 2859 non-international matches and 12939  
40 training session days). The turnarounds examined lasted from 3 to 8 days. Only injuries with  
41  $\geq 3$ -day time loss were retained for analysis. We then looked at the injury rate within each  
42 microcycle in relation to the presence of a day off or not, and its programming sequences in  
43 relation to the previous match (i.e., day off at D+1 vs D+2 for the day after the match or the  
44 following, respectively). **Results:** During 3- and 7-d turnarounds, the sequences including the  
45 day off-feet at D+2 were associated with 2 to 3 times lower overall non-contact injury rates  
46 than the other programming sequences (Cohens' d: 0.9 to 2.7). For the other turnarounds, the  
47 differences between the sequences were unclear. **Conclusion:** The programming of a day off  
48 (or at least 'off-feet') at D+2 may be associated with moderately-to-largely lower incidences  
49 of non-contact injuries, especially during 3- and 7-d turnarounds.

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51 **Key words:** planning; programming; elite football; injuries; rest day

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## 56 **Introduction**

57 Planning the microcycle is complex in elite (soccer) football.<sup>1</sup> While there are some  
58 informative data now available on programming practices in soccer,<sup>2-11</sup> these are generally  
59 representative of single club practices and only provide quantitative information (e.g.,  
60 external load dynamic based on GPS). Recently, in order to better understand the reasoning  
61 behind the choice and the drivers for planning and content selection, we surveyed 100 elite  
62 practitioners working in pro football.<sup>1</sup> The large majority of the responders confirmed  
63 collectively that balancing work and recovery from one day to the next across the microcycle  
64 was very likely required for optimised health and performance.<sup>1</sup> However, whether putting  
65 players at complete rest for one or two days affects injury rate during the same microcycle  
66 and the following match is still unknown. The question of what day to take off, and even  
67 whether to give a day off at all is something that has not been examined scientifically despite  
68 its immense importance in terms of recovery, compensation and psycho-social team  
69 dynamic.<sup>1,12</sup>

70 In order to shed light into this important topic, we examined in this descriptive study the  
71 association between the programming of days off and injury rate, using retrospective data  
72 from 18 elite teams performing in top leagues including the EPL, the Italian Serie A, the  
73 Bundesliga, the Scottish Premiership, the MLS and the Dutch Eredivisie from January 2018  
74 to December 2021. We more precisely also looked at the timing of these day(s) off within  
75 turnarounds of varying lengths. We then looked at the influence of prior match congestions  
76 on the above-mentioned associations. While the present observational study design precludes  
77 the examination of causal relationships, we believe that the information provided could help  
78 managers and performance staff to optimise the programming of their microcycles, within  
79 their own context.

## 81 **Methods**

### 82 *Data collection*

83 For this study, player characteristics, participation data and injury details were extracted from an  
84 online database (i.e., Kitman Labs platform, Dublin, Ireland) commonly used by all the football  
85 teams involved in the study. Each player and club is provided with an ID number on the platform.  
86 The researchers in charge of the analysis could only pull and analyze data associated with their  
87 IDs - no names included. Then, data was transformed and coded for injury occurrence (dates only  
88 used for assessing occurrences, such as during a match vs during training and when in relation  
89 from/to the previous match) and type (contact or non-contact injury, without any more details), to  
90 provide a final dataset.

91 The medical staff of each team registers injury details in the platform as a part of their daily player  
92 care management, including variables such as date of injury, type of injury and injury severity  
93 (days lost). Similarly, player game and training session participation are recorded as part of the  
94 team staff's daily monitoring. Additionally, the measures of training and competitive load are also  
95 added to the platform. The fact that all clubs used the same platform ensured the standardisation  
96 and the reliability of all types of entries, from medical information to exposure measures (e.g.,  
97 session duration and GPS data attached to the system calendar). We nevertheless ran a thorough  
98 data health check to ensure that all data retained for analysis met the same standard.

99 Permission was granted by the teams for their inclusion in this study, therefore ethics committee  
100 clearance was not required. The study conforms nevertheless to the recommendations of the  
101 Declaration of Helsinki.

102 Data were extracted from 18 teams belonging to EPL, the Italian Serie A, the Bundesliga, the  
103 Scottish Premiership, the MLS and the Dutch Eredivisie from January 2018 to December  
104 2021. This represented 82 team-seasons.

105 Since preliminary analysis didn't show any trends suggestive of differences between the  
106 different leagues or continents, all data were pooled together to increase sample size.

107 Team-seasons for which injury information was not accessible were not used for analysis.

108 Likewise, when there was not enough information about players on the platform (e.g. no  
109 exposure for less than 15 players over the entire season), the team season was not included.

110 The final data set included 56 team-seasons, including a total of 1578 players, 2865 injuries,  
111 2859 non-international matches and 12939 training session days.

112

### 113 *Data preparation*

114 A n-d turnaround was defined as a microcycle with n days between the first and second  
115 match, where n is the count of days from the first day after a match up to and including the  
116 following match day. The shortest observed turnaround was 3 days (3-d) e.g. playing a match  
117 on Sunday and again the following Wednesday, while the longest was 8 days (8-d) e.g.  
118 playing on Saturday and again the following Sunday. In total, 1871 turnarounds were  
119 extracted and were grouped by their respective length.

120 Turnarounds following at least one 3-d turnaround were considered as congested.<sup>13</sup>

121

122 In the absence of direct access to teams calendars and schedules, we assumed that a day off  
123 was a day without a game where the main 15 players of a given team did not have any  
124 training session exposure registered in the Kitman Labs platform. We considered that an  
125 exposure took place on a given day when there was information about either workload,  
126 duration or third-party metric with a game or training session event tag. We then considered  
127 that these exposures were accurate as they were extrapolated from the metrics consistently  
128 collected by the teams. Using this classification, it is very likely that non-exposure days were  
129 rest days, but we can't rule out that some light activities may have taken place at the club  
130 (i.e., recovery, mobility, football-tennis, etc), which, given their nature, were not registered as  
131 exposure. Therefore, considering those non-exposure days as "days off-feet" is likely the  
132 most accurate description of those specific days - this terminology was consequently used  
133 throughout the manuscript.

134

135 The 15 outfield players with the highest number of both pre- and in-season games played  
136 during a given season were considered as the main players. Note that these 15 main players  
137 can be different from one season to another for a given team. Days off-feet distribution  
138 patterns were examined within each microcycle. Days were first coded as 'x' for a day  
139 trained and as 'o' for a day off-feet; all possible combinations (e.g. x/x/x, o/x/x, x/o/x, x/x/o,  
140 o/o/x, o/x/o, x/o/o, o/o/o for 4-d turnarounds) were then created for each turnaround. Only the  
141 specific sequences with  $\geq 10$  occurrences within each turnaround were retained for analysis.

142

143 Injury is often defined as an occurrence sustained during either training or match-play which  
144 prevents a player from taking part in training or match-play for 1 or more days following the  
145 occurrence.<sup>14</sup> In this study we wanted to focus on non-contact injuries that substantially  
146 impact training and match participation and so only considered non-contact injuries that  
147 caused a minimum of 3 days of training/playing interruption i.e.  $\geq 3$ -day time loss. In fact, we  
148 excluded all mild injuries ( $< 2$  days lost) because injuries in this category could conceivably  
149 not have an impact on the next game availability or training dynamic within the same  
150 turnaround. Overall, this choice has allowed us not to include days lost due to potential  
151 training removal as a result of player management, as it sometimes happens in clubs.<sup>15</sup> If the  
152 medical staff registered injuries from the start to the end of the season, we considered that  
153 they strictly did it during the whole season, so we assumed that there was no missing data for  
154 this metric in this situation.

155 Considering all the above, there were 511 main players and 965 time-loss injuries, including  
156 559 non-contact ones (both match and training), as part of the 56 team-seasons.

157

### 158 *Data analysis*

159 Data was analysed in three consecutive steps, from a macro to a micro level.

160 1. Presence of a day off-feet *per se* and injury rate: we examined the potential difference in  
161 both training and match injury rates (per entire turnaround and per actual training day) with  
162 the presence or absence of at least a day off-feet in the turnaround - for all turnarounds  
163 pooled together, and then for each specific turnaround length separately.

164 2. Presence of match congestion (0, 1 or  $\geq 2$  of 3-d turnarounds) prior to turnarounds  
165 including a least one day off-feet, or not, and injury risk.

166 3. Distribution (i.e., when) of days off-feet during each turnaround length, and their  
167 association with training and match injuries.

168 For the later level of analysis, injuries were presented both per entire turnaround and per  
169 actual number of training days only; e.g. for x/x/x : overall non-contact injury rate per  
170 turnaround was calculated as follows: 0.15 non-contact game injuries per microcycle + 0.05  
171 non-contact training injuries per microcycle = 0.20 non-contact game and training injuries per  
172 microcycle; overall non-contact injury rate per training + match days only: 0.15 non-contact  
173 game injuries per day + 0.025 (=0.05/2 training days) non-contact training injuries per day =  
174 0.175 non-contact match and training injuries per day.

175

### 176 *Statistical analysis*

177 Results are presented as a mean and 95% confidence intervals. Substantial differences were  
178 assumed when the CIs did not overlap.<sup>16</sup> Cohen's d was then calculated to provide a  
179 magnitude of the differences, with thresholds of 0.2, 0.6, 1.2 and 2 considered as small,  
180 moderate, large and very large effects/differences.<sup>17</sup>

181

## 182 **Results**

183 Overall injury rate was 5 times greater during matches than during training (Table 1), with no  
184 difference between turnaround lengths, except for the 5-d turnaround which displayed fewer

185 injuries than all the others.

186 Training injuries were slightly lower for the 3- and 4-d turnarounds compared with the  
187 longest, but those differences were almost absent when expressed in relation to the actual  
188 numbers of days of exposure (Table 1). The differences in both training and match injury  
189 rates between turnarounds with and without a day off-feet were unclear. This was observed  
190 both when all turnarounds pooled together, and also when each specific turnaround was  
191 examined separately (all CIs overlapping, data not shown).

192

193 The number of congested turnarounds preceding the turnarounds of interest didn't have a  
194 clear relationship with either training or match injury rate, with or without day(s) off-feet -  
195 irrespective of the turnaround length (all CIs overlapping, data not shown).

196

197 The most represented training and days off-feet sequences within each turnaround are shown  
198 in Table 2. For all turnarounds up to 6-d the most frequent practice was to train all days of the  
199 microcycle (30 to 80%, with the shorter the turnaround, the less frequent the days off). For all  
200 these turnarounds, if a day off was programmed it occurred more commonly on D+1. For the  
201 two longest turnarounds, 7-d and 8-d, the most common practice was to give a day off-feet at  
202 D+1, followed by training every day.

203

204 We observed some substantial differences both in non-contact training and match injuries as  
205 a function of training and days off-feet sequences within some of the typical turnarounds  
206 (Figures 1 and 2). In our sample there were no non-contact training injuries during 3-d  
207 turnarounds when a day off-feet was included (irrespective of the day). The relative  
208 frequency of these turnarounds was quite low however (Table 2). The match injury rate  
209 across the entire 3-d turnarounds with a day off-feet at D+1 was about 50% of the rate for  
210 turnarounds with training every day. There were no match injuries at all for turnarounds with  
211 a day off-feet at D+2 (Figure 1).

212

213 During 5-d turnarounds, there were no non-contact training injuries when two days off-feet  
214 was programmed at both D+1 and D+2. During 6-d turnarounds, the lowest non-contact  
215 training injury rate was observed when there was either no day off-feet, and when the latter  
216 was programmed at D+2 (with unclear difference in match injury rate). During 7-d  
217 turnarounds, both non-contact training and match injuries were lower when the day off-feet  
218 was programmed at D+2 than when not programmed at all or at D+1. During 4- and 8-d  
219 turnarounds, all injury rate differences between the different sequences were unclear.

220

221 When looking at non-contact training and match injuries together (Figure 2), and focusing on  
222 the three most common programming practices only (i.e., no day off, a day off at D+1 or  
223 D+2), the sequences including the day off at D+2 (x/o/...) were associated with 2 to 3 times  
224 lower injury rates per day (moderate-to-very large Cohen's d) than the 2 other sequences for  
225 the 3- and 7-d turnarounds. For the other turnarounds, the differences between the main 3  
226 sequences were unclear.

227

228

## 229 Discussion

230 This is to our knowledge the first study to examine the association between the programming  
231 of day(s) off (at least days 'off-feet') within the training microcycle, and both training and  
232 match injury rates. While the present observational study design precludes the examination of  
233 causal relationships, the present findings suggest that while planning a day off *per se* may not  
234 share clear associations with injury rate (results not shown and Figure 2), the programming  
235 and distribution of the day off-feet within the microcycle (i.e. when the day off is scheduled),  
236 does, especially for 3- and 7-d turnarounds. Despite some variability between the different  
237 turnaround lengths, the sequences including the day off-feet at D+2 (x/o/...) were associated  
238 with 2 to 3 times lower injury rates per day (large to very large Cohen's d) than the 2 other  
239 sequences for the 3- and 7-d turnarounds (Figure 2). These associations with injury rate  
240 weren't affected by prior match congestion, suggestive of the robustness of the association  
241 between injury rate and this specific microcycle structure (i.e., x/o/...).

242

243 While there are always many ways to program the microcycle, training at D+1 and having a  
244 day off at D+2 may offer several advantages both on the performance and injury sides of  
245 things. At D+1, while the starters of the previous match can receive treatment and perform  
246 their recovery session, all benched players and substitutes also have the opportunity to train  
247 hard to compensate for the match they didn't play. This allows everyone to 'close the  
248 previous turnaround cycle' (recovery/compensation), and then rest for all the next day (D+2)  
249 before getting back fresh at D+3 for a new 'cycle' until the next match.

250 Conversely, when having the day off at D+1, the opportunities to care for starters and  
251 compensate for benched and substitute players are reduced, and potentially postponed. The  
252 consequence of this is that some starters may still need some treatment at D+2 and may  
253 therefore not be able to train, and subs may have been under a reduced training regime for 2-3  
254 consecutive days (light load at D-1, 0 to 30 min of play max on MD, and off at D+1),  
255 disturbing an optimal training dynamic and likely limiting their overall adaptation. Along  
256 these lines, when training is continually interrupted, substitutes end up lacking training  
257 stimulus, and especially with respect to some key elements of the game (e.g., sprinting  
258 distance<sup>18</sup>). They often tend to show reduced neuromuscular performance as the season  
259 progresses.<sup>19</sup> While not implying causality, our results may provide support to the common  
260 practice of having the (only) day off-feet at D+2, irrespective of the turnaround length, at  
261 least when injury is the consideration.

262

263 When looking at specific training and rest days distributions, it appeared that in all  
264 turnarounds up to 6-d the most frequent practice was to train on the pitch every day of the  
265 microcycle (Table 2). In addition, for all these turnarounds if a day off-feet was to be  
266 programmed, it was preferentially programmed on D+1. For the two longest turnarounds (7-  
267 and 8-d, Table 2), the preference was to give a day off-feet at D+1, followed by training  
268 every day. This contrasts with the results of our recent survey of 100 elite practitioners<sup>1</sup>  
269 where having the day off at D+2 (rather than D+1) was reported to be the optimal option.  
270 This may be related to the fact that when responding, the practitioners may have been biased  
271 toward their preferences rather than their actual practices (as per the data analysed in the  
272 present study). Therefore, the microcycle structure associated with the lowest injury rate was

273 not the most commonly programmed, irrespective of the turnaround length (e.g., the  
274 occurrence of the 'x/o/x/x/x/x/x' sequence was only 10% vs 25% for the o/x/x/x/x/x/x'  
275 sequence, Table 2). It's also worth noting that coaches may not always want to consider the  
276 'injury' argument as their first consideration when programming their microcycles; other  
277 factors including psycho-social team dynamics (players generally prefer D+1 to be off), the  
278 need to provide a greater overall training stimulus to players (very little rest or no days off at  
279 all during pre-season, returning from breaks) or to prepare tactically for important matches,  
280 and various external constraints (e.g., travels) may often need to be prioritised. Also, while  
281 having the day off at D+2 still allows for a complete training cycle post day off for the  
282 longest turnarounds e.g., 4 days left to prepare the next match during a 7-d turnaround, this  
283 may disrupt optimal match preparation during short turnarounds e.g., 1 day left to prepare the  
284 next match during a 4-d turnaround. In summary, coaches may not see the "rest at D+2  
285 option" as a relevant alternative in their own context even though it may be ideal on paper  
286 from a physiological and biological standpoint.

287

288 Finally, the reason for the lack of clear and consistent differences in injury rates between the  
289 different sequences within the 4-, 5-, 6- and 8-d turnarounds (Figure 2) is difficult to explain  
290 with the current data limited to exposure information. A simple first explanation is likely  
291 related to the lower samples for these turnarounds (Table 2), which directly increases the  
292 breadth of the CIs, making in turn some of the between-sequences differences unclear. It is  
293 also likely that other factors may share more associations with injury rate than the  
294 programming of days off *per se*, and, in turn, could have diluted/confounded the analysis.  
295 One important limitation is the univariate nature of the present analysis. While we thought to  
296 answer the simple question of the programming of rest days, it is clear that injuries are  
297 largely multifactorial in nature<sup>20</sup> and that different loading patterns, match exposures and  
298 minutes played within the same sequences may also directly affect injury rates. In fact, the  
299 data from the practitioners' survey<sup>1</sup> showed that while the current loading and training  
300 contents are pretty homogenous between teams for 7-d turnarounds, there is more variability  
301 in programming for 5- and 6-d turnarounds. This may partly explain why the association  
302 between days off-feet and injury rate was unclear for those turnarounds. We will certainly  
303 continue to investigate this and other topics related to planning the microcycle but given the  
304 lack of research in this area we encourage other researchers to think about experimental  
305 designs which would provide more insight on how best to adapt the training schedule to the  
306 fixture schedule.

307 Additionally, the simultaneous consideration of player profiles (e.g., age, injury history,  
308 strength, mobility or flexibility) and other measures of internal training load and responses to  
309 load should also improve the analysis - while making the current outputs less straightforward  
310 for practitioners. There is in fact a trade-off between the desire for simple questions to have  
311 simple answers (e.g, when is it best to rest?) and more sophisticated analytic approaches that  
312 may have more precision but require more effort to interpret in order to provide direct  
313 applications (i.e. results of multivariate analyses can be difficult to translate into simple  
314 yes/no answers).

315

316 *Limitations*



317 The present observational study design precludes the examination of causal relationships.  
318 Having a proper distinction between complete rest days, days off-feet and training days  
319 would have been ideal. In the absence of direct access to teams' calendars and schedules, days  
320 off were estimated based on exposure data (workload, duration or third-party metric with a  
321 game or training session event tag) and we interpreted those days off as at least days off-feet.  
322 Whether this perfectly reflects real practices remains impossible to verify e.g. a gym-based  
323 session with no measure of external load logged into the system could have been  
324 programmed on a day that was counted as 'off'. It is also worth mentioning that the number  
325 of observations for the x/o/x... sequences was consistently lower than that for the other  
326 sequences, irrespective of the turnaround lengths (see Table 2) and this should be considered  
327 when interpreting the results. Finally, the injury records used for analysis are as good as what  
328 practitioners may have registered. Relying on injuries based on practitioners' entries is  
329 however common practice,<sup>21</sup> and we believe that the value of the information provided,  
330 derived from a very large sample size (> 1800 turnarounds), partly outweighs those possible  
331 limitations. Also, the present data showed a 5 x greater injury rate during match than training  
332 (Table 1), which is highly consistent with the >24 vs 4 injuries / 1000 hrs of exposure  
333 generally reported.<sup>21</sup> Future research based on more detailed calendar entries and larger  
334 sample size for some of the day sequences would help improve the clarity of the current  
335 findings.

336

### 337 **Practical applications**

338 The present study showed for the first time, using a large pool of data from elite football, that  
339 while planning a day off (at least off-feet) *per se* may not share clear associations with injury  
340 rate, its programming (i.e., when) within the microcycle. In practice, at least for the 3- and 7-  
341 d turnarounds examined, programming the (only) day off-feet of the microcycle at D+2 was  
342 associated with 2 to 3 times less overall injury rates than either not having a day off-feet, or  
343 programming the latter at D+1. Future studies should also examine, within each turnaround  
344 length, the actual load of each training day in relation to the different day off programming  
345 strategies.

346

### 347 **Conclusion**

348 The programming a day off (or at least 'off-feet') at D+2 was associated with a moderate to  
349 large reduction of non-contact injuries, especially during 3- and 7-d turnarounds.

350

### 351 **Conflict of interest Statement**

352 The authors are all employed by Kitman Labs, a sports performance company used by sports  
353 organizations around the world to collect, centralize, and analyse data from multiple sources  
354 and aid decision-making ([www.kitmanlabs.com](http://www.kitmanlabs.com)). There is no direct gain for the authors or  
355 Kitman Labs, financial or otherwise, as a result of the findings of this study, or would there  
356 be if they had been different. However the Kitman Labs Performance Intelligence Research  
357 Initiative may influence organisations in their choice of technology partner. The authors are  
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365

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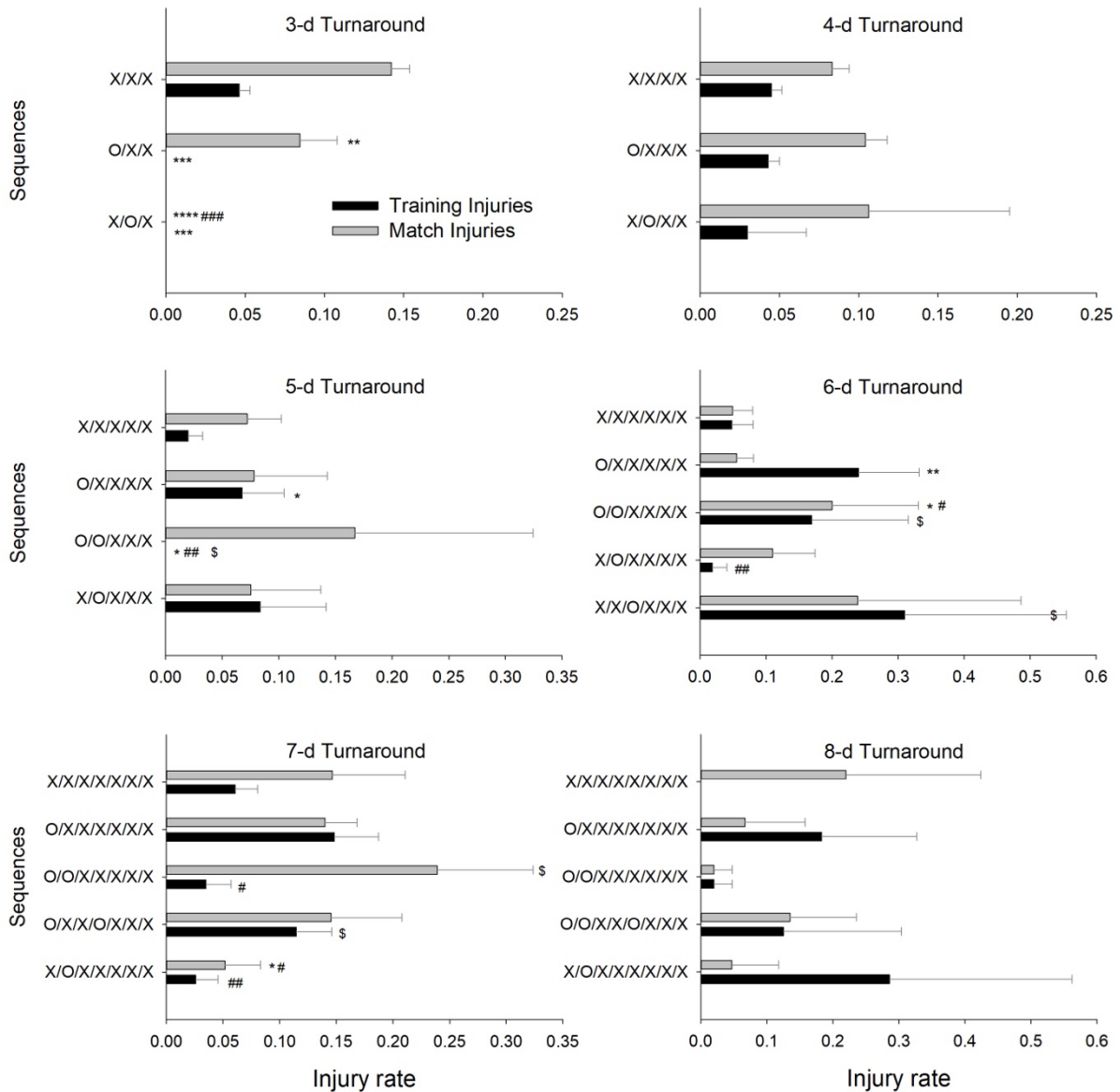
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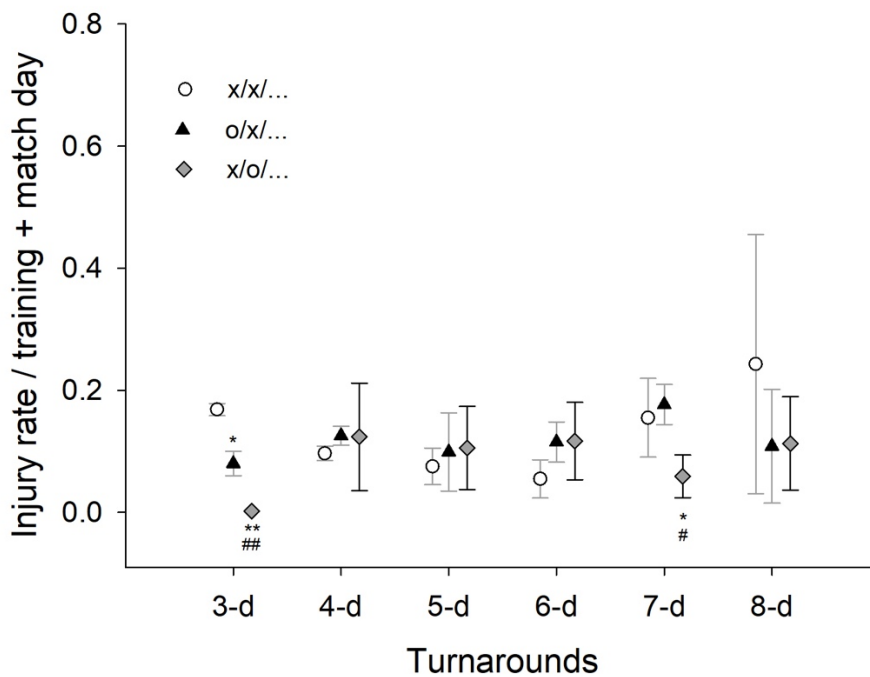
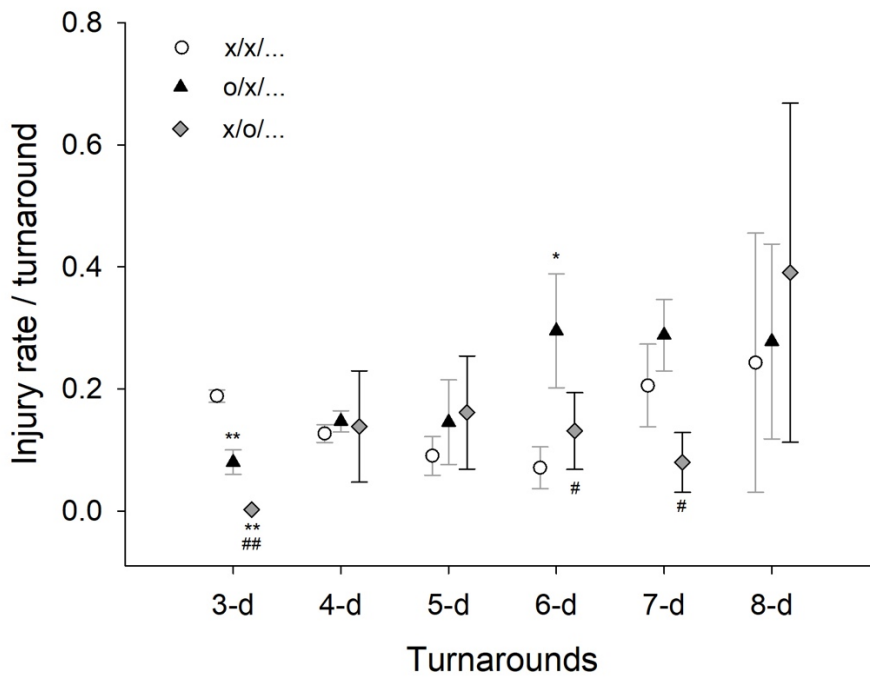
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**Figure 1.** Average (95% CI) non-contact training and match injury rate during the main training and off-foot day patterns observed within each of the 6 match turnarounds examined in the 18 teams. \*: stands for differences vs x/x/... sequence, #: vs o/x/... \$: vs x/o/... The number of symbols stands for small, moderate large and very large effects/differences.



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**Figure 2.** Average (95% CI) total (training + match) non-contact injury rate per turnaround (upper panel) and total (training + match) non-contact injury rate per training + match days only (lower panel) for the three main sequences including either no day off (x/x/...), or a unique day off either at D+1 (o/x/...) or D+2 (x/o/...) for all turnarounds. \* and \*\* stands for moderate and large differences vs x/x/... sequence, respectively. # and ## stands for moderate and large differences vs o/x/... sequences, respectively.

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<b>Turnarounds</b>	<b>Turnarounds (n)</b>	<b>Training injuries per turnaround</b>	<b>Training injuries / training day</b>	<b>Match injuries</b>
3 d	655	0.05 (0.01)*	0.03 (0.01)	0.25 (0.02)
4 d	577	0.06 (0.01)*	0.03 (0.01)#	0.21 (0.02)
5 d	195	0.13 (0.03)	0.04 (0.01)	0.14 (0.03)§
6 d	211	0.22 (0.04)	0.05 (0.01)	0.23 (0.05)
7 d	440	0.18 (0.02)	0.04 (0.01)	0.23 (0.02)
8 d	125	0.23 (0.06)	0.04 (0.01)	0.20 (0.06)

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**Table 1.** Number of observations for each turnaround length, and associated overall training and match injury rate (irrespective of the presence of days off or not, both contact and non-contact injuries together). \*: small substantial difference vs 5- to 8-d turnarounds. #: small substantial difference vs 6-d turnarounds. §: small substantial difference vs all other turnarounds.

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Turnaround	Planning Sequence	Frequency	Turnaround Proportion (%)
3-d	o/x/x	94	14
	x/o/x	18	3
	x/x/x	531	80
4-d	o/x/x/x	240	41
	x/o/x/x	28	5
	x/x/x/x	276	47
5-d	o/o/x/x/x	12	6
	o/x/x/x/x	63	29
	x/o/x/x/x	30	15
	x/x/x/x/x	70	34
6-d	o/o/x/x/x/x	25	11
	o/x/x/x/x/x	64	28
	x/o/x/x/x/x	25	11
	x/x/o/x/x/x	11	5
	x/x/x/x/x/x	68	29
7-d	o/o/x/x/x/x/x	53	11
	o/x/x/o/x/x/x	83	17
	o/x/x/x/x/x/x	116	25
	x/o/x/x/x/x/x	44	10
	x/x/x/x/x/x/x	69	15

8-d	o/o/x/x/o/x/x/x	15	12
	o/o/x/x/x/x/x/x	20	15
	o/x/x/x/x/x/x/x	21	16
	x/o/x/x/x/x/x/x	12	9
	x/x/x/x/x/x/x/x	16	12
Totals		2005	

472 **Table 2.** Frequency and proportion of the most represented training and days off-feet  
473 sequences within each turnaround. Note that since some less frequent sequences were not  
474 shown here, the proportions (right column) don't always sum up to 100%.

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