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7 Prediction of overuse injuries in professional U18-U21 footballers using metrics of training distance
8 and intensity.

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10 Bacon, Christopher S; Mauger, Alexis R

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21 ABSTRACT

22 The most common injury in professional football is an overuse injury to the lower limb. A significant
23 external risk factor of this injury is the mismanagement of training and match loads. The aim of the
24 current study was to examine the predictability of overuse injuries in professional youth soccer players
25 using volume and intensity variables derived from Global Positioning Systems (GPS). A total of 41
26 players (Age – 17.8 yrs \pm 1.1 yrs) training and match loads were assessed. These external loads were
27 measured over two competitive seasons for every training session and match for each individual. A
28 linear regression was used to test the predictability of the injury based on load, which were grouped
29 using loading groups calculated from squad weekly averages. The load groupings assigned were: Low
30 load = 1 SD below the squad mean score; Normal load = \pm 1 SD from the squad mean; High load = 1
31 SD above squad mean. The analysis demonstrated that total distance significantly predicted overuse
32 injury incidence rates (F(1, 39) = 6.482, p = 0.015), whereas high speed running meters could not (F(1,
33 39) = 1.003, p = 0.323). This study demonstrated that distance covered in training and matches can
34 impact on the incidence of overuse injury in youth soccer players. Coaches should seek to monitor
35 player training loads and incorporate this metric into their decision making for protecting players from
36 overuse injury.

37

38 **Key words:** Overuse Injury, GPS, External Loads, Total Distance, High Speed Running

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44 INTRODUCTION

45 Overuse injuries account for approximately 34% of all injuries sustained in soccer and these occur
46 throughout the competitive season (18,30,38). As this will directly impact on the quality of the team
47 fielded, it is important to reduce the amount of overuse injuries in order to maximize the chances of the
48 team's success (23). However, overuse injuries can be challenging to prevent due to the wide variety of
49 causative factors, many of which are difficult to manage. Generally, injury occurrences are categorized
50 by intrinsic (person related) and extrinsic (environment related) factors. Intrinsic factors that are
51 purported to have an increased effect on injury occurrence include age, career duration and history of
52 previous injury (12,17,32,35,38). The most prominent extrinsic risk factors causing overuse injuries are
53 suggested to be excess levels of external training loads, high training to match ratio and playing on a
54 hard surface with high friction (12,16,31). However, the consistent factor throughout the literature
55 related to overuse injury appears the mismanagement of external load on the working muscles (36).
56 This variable should be within complete control of the team's coaching staff, and so data is required to
57 go beyond establishing a link between training load with overuse injury and instead attempt to quantify
58 this relationship. It is important to note that although staff supervision in training can control part of the
59 external load experienced in a regular season week, within competitive matches it can only be monitored
60 through match data. Although certain aspects of training are more difficult to regulate (such as small-
61 sided games), these elements of a session can be monitored in real-time in order to control the external
62 load.

63 Most elite football clubs now use GPS (Global Positioning System) devices to monitor player
64 external loads and distance covered during training and matches across a season (33). GPS units collate
65 data into a system which has the potential to provide the user with detailed feedback regarding player
66 over- and under-loading, and this could be used to help reduce overuse injuries. However, there is
67 currently a lack of unity of procedures that elite club's follow in terms of data processing, inhibiting the
68 quality of data (16) and meaning the system is only as good as the user. The most recognized method
69 of data usage involves the comparison of each player's load for the session/match with the squad's
70 average for that session. This metric is then used to inform the degree of risk of an overuse injury in the

71 forthcoming sessions/matches for that player. The current method does not account for the internal load
72 that is likely to vary across a squad of players. Using a squad average for player's external load can
73 only be used for indicating a potential risk in terms of the external load that can be easily accounted for
74 in training/matches. Currently the method allows the users to make modifications to easily managed
75 external loads in order to alleviate potential high internal loads. Modifications in the player's training
76 program can then help to mitigate this risk (25).

77 The competitive football season generally lasts in excess of 9 months, and it is important to
78 take account the contributing effect of cumulative overloading to overuse injury (5). Some studies on
79 Australian Football have taken this cumulative training and match load meters completed factor into
80 account by monitoring the effect of 3-weekly cumulative loads on injury incidence rates (13). Colby et
81 al. (13) showed a higher injury risk with increased load and this was elevated with higher intensities.
82 The research also presented that the metrics of total distance (TD) and high speed running (HSR) were
83 the most plausible measures to be used in terms of injury prediction, which supports other prior research
84 by Castagna et al. (10). However, Australian Football has very different physiological demands to
85 football, and greater amounts of physical contact between players (13,18). This likely increases the
86 amount of injuries that are sustained in Australian Football compared to football (13,18), and limits the
87 translation of the data between the two sports.

88 There is a significant body of the research which uses GPS to provide movement analysis in
89 football (10). Indeed, Castagna et al. found players covered between 5098-7019 m in a match, with 15%
90 being accounted for as high intensity distance (10). It was also shown that players fatigued over the
91 course of the fixture by 3.8% in terms of their TD covered when compared to first half values (10).
92 Castellano and Casamichana (11) has examined the relationship between heart rate and GPS to define
93 fitness levels. Castellano and Casamichana (11) presented the average percentage of heart rate
94 maximum players worked at in relation to their distances covered from various GPS derived variables.
95 Whilst this information is interesting, there is a need to provide some application of this data. Brink et
96 al. (6) suggests that quantification of the relationship between cumulative training and match
97 load/intensity and overuse injury could be used to provide a framework for coaches to use in order to

98 reduce overuse injury risk. However, when applying GPS as a measure of load the percentage error
99 must be accounted for as this can range from 5-8% (14). Coutts and Duffield (14) demonstrated that the
100 accuracy of GPS devices is within an acceptable margin of error for validity of results, but it was noted
101 that measures of high intensity movements, such as HSR, have presented a potential error of 11.2-
102 32.4%. Deficiencies in the accuracy of the models can be attributed to devices which are less than 10
103 Hz in processing power (14,15). Accounting for this, data from appropriate GPS devices could provide
104 a basis for individual training norms for each player in a squad to be calculated and allow a more
105 intelligent means of guiding training prescription (6). Consequently, the aim of this investigation was
106 to monitor youth player training loads/intensities in a professional football club using GPS, and to
107 subsequently calculate the capacity of this data to predict overuse injury. It was hypothesized that TD
108 would predict overuse injury incidence rates and that HSR would not be able to significantly predict
109 overuse injury incidence rates.

110

111 **METHODS**

112 **Approach to the Problem**

113 Both 2012-13 and 2013-14 season's data were collated into a single data set for analysis. Player weekly
114 averages of TD and HSR within the 40-week time period were calculated. The calculation used did not
115 include weeks of training that were affected by injury (i.e. where a player was returning from injury
116 and training load/intensity was reduced). GPS (StatSports, Viper Pod, NI) data was acquired for every
117 training session and match that each individual player was involved in across the seasons. From this,
118 the metrics used in the current study were TD (volume of training) and HSR (intensity of training), used
119 in a similar study by Colby et al. (13). These metrics were used to represent external loads from training
120 and competitive matches. Injuries were collected from an injury audit with diagnosis and recording into
121 this dataset completed by a qualified physiotherapist. It is worth noting that there was an increase in
122 injuries between the seasons which can be potentially attributed to internal changes in the club - an
123 increase in coaching hours, an increase in the number of players and a change in coaching staff at the

124 club. This data was collected for two seasons in the Barclays U21/U18 Premier Leagues from 2012-13
125 and 2013-14. Data was collated for the 40 weeks of the competitive season in each year for both training
126 sessions and matches. The relationship between overuse injuries and external loads was explored using
127 a method similar to Colby et al. (14). Weekly training loads were assigned certain loading groups
128 dependent on the amount completed and then assessed to see the relationship to injury incidence rate.

129

130 **Subjects**

131 Over two seasons, data was collected from forty-one youth soccer players (n = 18 in 2012/13 season,
132 height: 175.2 cm \pm 4.5 cm, body mass: 72.4 kg \pm 3.1 kg, age: 18.7 yrs \pm 1.2 yrs; n = 23 in 2013/14
133 season, height: 181.3 cm \pm 6.1 cm, body mass: 74.9 kg \pm 8.7 kg, age: 17.0 yrs \pm 1.1 yrs). All players
134 were on a full-time training program (6 training sessions a week) and had either signed a youth
135 scholarship contract with the club or had signed a professional contract. All of the data obtained from
136 the professional football club was from pre-existing datasets which included both the GPS metric
137 measurements and the injury audit data. Access to data was granted with the consent from the
138 professional football club. All data was analyzed in an untraceable and anonymized format. Ethical
139 approval for the use of existing datasets was obtained from the University Research Ethics Committee.

140

141 **Procedures**

142 The squad average and standard deviation (SD) was calculated for both TD and HSR along with player
143 weekly averages of the same variables. The SD was used to assign player groupings dependent upon
144 their weekly average (for TD or HSR) compared to the rest of the squad. The groupings assigned were
145 as follows: Low load = 1 SD below the squad mean score ($x \leq 19404.30$ m for TD, $x \leq 538.17$ m HSR);
146 Normal load = ± 1 SD from the squad mean (19404.30 m $\leq x \leq 23700.62$ m for TD, 538.17 m $\leq x \leq$
147 890.63 m for HSR); High load = 1 SD above squad mean ($x \geq 23700.62$ m for TD, $x \geq 890.63$ m for
148 HSR). A second analysis was completed to test the effect of cumulative weekly loads on injury

149 incidence, in a similar manner to previous research (14). Cumulative player loads were calculated for
150 2, 3 and 4 week periods throughout the 40-week season, with players grouped according to SD and
151 squad averages, as described previously. In line with this, injury incidence rates were calculated for
152 each player in each season to allow comparison between load/intensity and overuse injury. Injury
153 incidence rates were reported by calculating the total number of overuse injuries (diagnosed by a
154 qualified physiotherapist) divided by the total 'on-leg' exposure time, and then reported as a figure per
155 1,000 training and match hours (20). Injury audits across the two seasons were collated and analyzed
156 with an X^2 analysis used to compare the frequency of injuries between each season.

157

158 **Statistical Analyses**

159 Statistical analyses were conducted using IBM SPSS 19 (SPSS Inc., Chicago, IL). Statistical
160 assumptions were checked using similar methods to Colby et al. (13) and Hulin et al. (27), and were
161 deemed plausible in all instances. HSR and TD data were collected from all training sessions and
162 matches that occurred during both seasons. The injury audit analysis contained detailed breakdown of
163 injury sites, injury types, contact or non-contact injuries, activity when injury occurred and the severity
164 of each injury. To ensure appropriateness of combining both seasons' data, a limits of agreement
165 analysis (2) was performed on data from youth players who had data across both seasons. A Pearson's
166 correlation test, in addition to the limits of agreement analysis, was run on the two seasons to test the
167 correlation between the two, in order to ensure that when combining the dataset neither season would
168 skew the data. Correlations were performed on; TD, HSR, Total Number of Injuries and Average Injury
169 Incidence Rate. A between groups one-way ANOVA was run to test the differences between playing
170 position and the variables; TD weekly average, HSR weekly average, whole season total distance
171 average, whole season high speed running meters average and average injury incidence rate. Players
172 were categorized based on the position they had played in for the majority of the competitive matches
173 and according to positions previously described (8). The categories were as follows: central defenders,
174 wide defenders, central midfielders, wide midfielders and forwards. Linear regression was used to test

175 the predictive capacity of HSR and TD groupings (Low, Normal and High) on injury incidence rates.
176 Odds ratios were used to assess the effect of cumulative weekly loads (2, 3 and 4 week) on injury risk,
177 using procedures as previously described (13,34). The reference for the odds ratio was set as the Normal
178 group (\pm 1SD of the squad mean). Significance was accepted at $P \leq 0.05$ with data expressed as means
179 \pm SD.

180

181 **RESULTS**

182 **Injury Audit Analysis**

183 In total there were 85 reported injuries in the cohort of players measured over the course of the 2012-
184 13 and 2013-14 seasons (Table 1). The majority of injuries sustained were located at the ankle (n = 26,
185 3.23 IIR, 30.59%) with no difference observed between the two seasons. Overall there was a significant
186 number of overuse injuries to the players involved (n = 16, 1.99 IIR, 18.82%) when compared to the
187 other injury types recorded over the time period. Within this injury type there was a significant
188 difference between the seasons for muscle strains ($X^2 = 7.514$, $p = 0.023$) with a substantial increase in
189 the 2013-14 season (n = 11, 2.64 IIR, 19.64%) when compared to the 2012-13 season (n = 1, 0.26 IIR,
190 3.45%). Contact and non-contact injuries presented a difference overall with the data showing more of
191 the injuries were non-contact (n = 44, 5.46 IIR, 51.76%). There was a significant difference in the total
192 number of injuries sustained from training between each season ($X^2 = 11.402$, $p = 0.010$), although
193 overall there were more injuries occurring in match scenarios (n = 47, 5.84, 55.29%) than in training
194 sessions (n = 30, 3.72 IIR, 35.29%). The majority of the injuries sustained were low in severity (n = 34,
195 4.22 IIR, 40.00%), with a significant difference between the two seasons in low severity injuries, with
196 the 2013-14 season (n = 23, 5.53 IIR, 41.07%) recording more than the 2012-13 season (n = 10, 2.57
197 IIR, 34.48%).

198

Please insert Table. 1. here

199

200

201 Season vs. Season Analysis

202 There was a positive correlation between the two seasons for the total number of injuries ($r = 0.382$, n
203 $= 41$, $p = 0.014$), and injury incidence rates ($r = 0.371$, $n = 41$, $p = 0.017$), although both r values were
204 relatively small. There was no correlation between the two seasons for TD or HSR (TD, $p = 0.093$;
205 HSR, $p = 0.914$).

206

Please insert Table. 2. here

207

208 Limits of agreement analysis revealed good agreement for TD across the two seasons (limits = 4636.32
209 to -4786.19, mean = -74.94) (Figure 1). The regression performed on the same data also showed that
210 data for both seasons worth for TDs were not significantly different ($t = 0.673$, $p = 0.515$), therefore
211 demonstrating agreement between the datasets ($p > 0.05$). HSR also showed good agreement (Figure
212 2). The regression completed on this variable confirmed there was no significant difference in the
213 seasons data ($t = -1.932$, $p = 0.079$). It was shown that there was agreement between the datasets, with
214 no proportional bias between the datasets ($p > 0.05$).

215

Please insert Figure. 1. and Figure. 2. here

216

217 Effect of Position

218 Table 3 demonstrates the positional mean \pm SD based on the two seasons analyzed, in addition to the
219 squad's descriptive data for the variables detailed. The ANOVA displayed a significant difference
220 between the positions for the variables; HSR weekly average ($f = 4.565$, $df = 4$, $p = 0.004$) and whole
221 season HSR total average ($f = 8.178$, $df = 4$, $p < 0.001$). For whole season HSR total average, the
222 multiple comparisons analysis displayed differences between: central defenders and wide midfielders

223 ($p < 0.001$); central midfielders and wide midfielders ($p < 0.001$); forwards and wide midfielders ($p =$
224 0.030); wide defenders and wide midfielders ($p = 0.048$).

225 Please insert Table. 3. here

226

227 **Cumulative Weekly Load Analysis**

228 When comparing high load groups to the reference normal group, there was close to significant levels
229 of increased risk of overuse injury for; TD - High Load Group (OR = 0.670, 95% CI = 0.395 – 1.137,
230 $p = 0.137$) and HSR - High Load Group (OR = 0.580, 95% CI = 0.330 – 1.021, $p = 0.059$) for 2 week
231 cumulative loads. The cumulative loads of 3 weekly and 4 weekly loadings showed no significant
232 differences.

233 Please insert Table. 4. here

234

235 **Overuse Injury Prediction Regression**

236 A simple linear regression was used to predict overuse injury incidence rates based on TD and HSR,
237 when players were assigned to the Low, Normal and High groups (Table 5). A significant regression
238 equation was found in only the TD variable, ($F(1, 39) = 6.482$, $p = 0.015$, $R^2 = 0.143$). Player's predicted
239 injury incidence rate was able to be significantly calculated by using their TD loading group. Injury
240 incidence rate per 1000 hours is decreased by -5.835 times when moving upward from one TD loading
241 group to the next. Thus, being in a higher TD loading group lowered the risk of an overuse injury
242 occurring. HSR loading groups were also analyzed using the same method with no significant
243 regression calculation found ($F(1, 39) = 1.003$, $p = 0.323$).

244 Please insert Table. 5. here

245

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247

248 **DISCUSSION**

249 This is the first study to use training and match loads, derived from GPS data, to predict overuse injury
250 risk in youth soccer players. The primary finding of the study was that overuse injury incidence rates
251 were decreased in youth soccer players who completed a higher weekly TD in training and matches. It
252 is important to note that these findings are based on average distances and intensities of the team from
253 which this data was derived and so these findings are likely specific only to these players. However,
254 Colby et al. (13) has also shown a similar result that an increase in weekly TD can lead to an increased
255 risk of injury. Intensity of training and matches (expressed as HSR) had no effect on overuse injury
256 incidence rates. This finding could be potentially due to the percentage error margin GPS devices
257 display when measuring high intensity bursts of speed (14). Despite this, the methods of analysis used
258 in this study could provide a framework on which other clubs can assess similar relationships within
259 their squads, and use this specific data to reduce risk of injury.

260 Although the results from the current study were unable to directly predict the occurrence of
261 overuse injuries, they do help to indicate the likelihood of these injuries occurring depending upon
262 training and match loads. However, predicting overuse injuries represents a significant challenge
263 primarily because of the numerous risk factors that contribute to this type of injury (25). The regression
264 model used in the current study predicts that the overuse injury incidence rate reduced by nearly 6 times
265 when the individual's TD is increased to that of players who achieve a high TD in relation to this squad.
266 Because high or low TD is relative to the squad of youth soccer players in this study, this finding should
267 not be generically interpreted as a higher loading reducing injury risk. Rather, it is hoped that the
268 methods outlined in this study can be used by other coaches and teams to assess their own squad's
269 competitive season loadings. By doing so, they can look to plan and modify training in individual
270 players who are at risk of overuse injury as a result of too high or too low intensity or load. This relates
271 to research by Gabbett (21) and Hulin et al. (27), where they examined the ratio of acute to chronic

272 workloads and the ability to predict injuries. Within this research both Gabbett (21) and Hulin et al. (27)
273 indicated that an increased risk of injury related to a sudden increase in workload ratio. Additionally,
274 they agree that training needs to be ‘smarter not harder’ in order to keep a the risk of injury from
275 mismanagement of loads reduced (21,27).

276 The findings in the present study can also potentially relate to literature focusing on the
277 relationship between overtraining and overreaching injuries. Existing research states that overreaching
278 injuries can be caused by an imbalance in the ratio of bouts of high intensity/load exercise to the amount
279 of recovery time between these bouts (29). Overreaching is an acute condition which can lead to chronic
280 overtraining due to a substantial period of mismanagement of external training loads (36). The present
281 research agrees with the existing research to the extent that significant cumulative loading through
282 overscheduling fixtures and training can cause a substantial increase to the risk of injury. Cumulative
283 weekly load analysis demonstrated that significant external loads over the course of a micro/meso cycle
284 start to increase injury risk. Although the results were not statistically significant, this may have been
285 due to the players involved being managed correctly. However, strong indications (TD - High Load
286 Group ($p = 0.137$); HSR - High Load Group ($p = 0.059$), for 2 week cumulative loads) were found
287 supporting existing research around overtraining and overreaching injuries, respectively.

288 This is especially the case in athletes at the younger ages of the experience spectrum to reduce
289 the chance of major overuse traumas later in their careers (5). Therefore, the important factor to take
290 from the analysis of the current research to the previous research is that training loads seem to display
291 reasonable association to injury risk. Monitoring athletes training load longitudinally will allow early
292 detection of overreaching to in turn reduce the risk of an overtraining injury (7). The present research
293 did not directly measure overreaching or overtraining symptoms, but additional metrics could be
294 integrated could assess this. Doing so would help further explore the relationship between
295 overreaching/training and injury. However, the findings of the current study show promise that GPS
296 derived external loads, especially the TD metric, can be analyzed against injury incidence rates within
297 professional sports over a longitudinal format.

298 GPS metrics have previously been used to demonstrate a link between injury risk and
299 cumulative weekly loads in Australian Football (13,20,34). These studies show that high 3 weekly
300 cumulative loads present a significantly elevated risk of injury. The current study suggests that in youth
301 football players, 2 weekly loads came closest to significantly increasing risk of injury (HSR meters – 2
302 Weekly Cumulative Load, $p = 0.059$). In addition to the effect of cumulative load, the research on
303 Australian Football indicates that the intensity of training and matches present a strong risk of injury
304 (13,20,34). The current study does not support this finding, with no effect of intensity found on overuse
305 injury incidence rates. An initial reason for this difference could be due to the sample age used in the
306 present study compared to previous others; present research – 17.8 ± 1.1 yrs, Australian Football
307 research average – 23.7 ± 3.4 yrs (13,20,34). This age difference in the samples will also indicate a
308 difference in years of training experience which has been shown to influence overuse injury incidence
309 rates (36). Additionally, the difference in the physical and physiological demands of the sport may
310 explain this result (26). These factors may also be responsible for the observed greater injury incidence
311 in Australian Football. Indeed, Colby et al. (13) presented figures of 297 injuries ($n = 46$ players)
312 recorded in the space of 1 season compared to the total of 85 ($n = 41$ players) recorded over the course
313 of two seasons found in this research. Additionally, movement analysis has shown that the proportion
314 of a match scenario in soccer spent at high intensities is higher (9) when compared to Australian Football
315 (22).

316 The data from the current study suggest that playing position should be taken into account when
317 assessing play training loads. It was shown that wide midfielders experience significantly increased
318 levels of HSR meters and TD meters throughout the season, and so a ‘one-size fits all’ approach to
319 player loading should not be used in a squad. Rather, training sessions need to reflect these differences
320 and ensure that each individual playing position is conditioned correctly to reduce overuse injury risk.
321 Within the squad analyzed for the current study, it appears as though training has potentially been
322 prescribed successfully as the injury incidence rate of the wide midfielders was the lowest out of the
323 positions analyzed (2.15 ± 2.49 IIR). Conversely, it was found that central midfielders had the highest
324 injury incidence rate (per 1000hrs) of all the positions with 14.22 ± 15.46 IIR. This may demonstrate

325 the physical demands of this position, which is supported by existing research based on position specific
326 movement analysis (3).

327 The present study has also examined the prevalence of different injuries within professional
328 football to evidence the impact of minor avoidable injuries (See Table 1.). Over the two season period
329 measured, the most common form of injuries were overuse injuries (n = 16, 1.99 IIR, 18.82%) and
330 muscle sprains (n = 16, 1.99 IIR, 18.82%). This finding is similar to previous research that has also
331 reported injury audits within professional football (1,12,19). Therefore, the findings in the current study
332 further demonstrates that a large majority of injuries sustained by a team are avoidable, which is an
333 important point to address in order to help maintain the team's performance (24). Additionally, the
334 current study demonstrated that 51.76% of all injuries sustained were non-contact indicating that even
335 with the sport being primarily contact based, injuries are just as likely to occur from non-contact actions.
336 The main location site for a significant number of the injuries were to the ankle region (n = 26, 3.23
337 IIR, 30.59%). This is similar to previous findings showing the relatively high frequency of ankle injuries
338 in football injury audits (24). It is also evident that a common trend appears in terms of when the injuries
339 took place, which this research found was in a match scenario (n = 47, 5.84 IIR, 55.29%). The injury
340 audit analysis has also established that the majority of injuries sustained are of minor level of severity
341 (≤ 7 Days Missed). However, it is important to note that even such minor severity injuries can still play
342 a significant part in a team's performance, as they can still result in a player missing an important fixture
343 (24).

344 It should be recognized that all team sport-based GPS units present a level of error in
345 measurement. The standard percentage error for GPS units normally lies between 5-8% (14,15). To
346 help overcome this in the current study, each GPS unit was assigned to a single player. This helped
347 ensure that all data error associated to the individual unit remained with individual players. In addition,
348 all devices were placed at the same location (top of the body on the trapezius) on each player and
349 recording during each session/match to help overcome inaccuracies as a result of differing placement
350 (37). Future research should look to examine a variety of GPS metrics, such as force loads (i.e. measures
351 of exerted power and momentum) which could also play a part in overuse injuries (13). Further

352 prediction models that include physiological or functional parameters such as heart rate variability,
353 creatine kinase levels and maximal force production should also be explored (4). Incorporating
354 physiological markers of overuse injuries to existing GPS derived metrics could potentially increase the
355 statistical power of an injury prediction model, allowing it to be more viable for use in the elite
356 environment.

357

358 **PRACTICAL APPLICATIONS**

359 The present findings provide a potential new approach for professional football clubs in analyzing youth
360 soccer player's GPS data. The present research shows that this cumulative approach has value,
361 particularly when assessing player injury risk. Initially, data should be analyzed over the course of a
362 season to provide individual player's training load baseline responses to training and matches (28).
363 Consequently, individual player data can be compared to the norm of the squad, and adjusted
364 accordingly. It is suggested that training sessions are not solely based on these loadings and instead
365 only be used as a guidance tool. Coaches should also look to individualize training sessions based on
366 playing positions, so that loads are specific to the position of the player.

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500 **Figure Legends**

501 Fig. 1. Limits of Agreement for 2012-13 and 2013-14 Seasons for Total Distance (m)

502 Fig. 2. Limits of Agreement for 2012-13 and 2013-14 Seasons for High Speed Running (m)

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520 **Table Legends**

521 Table. 1. Classification of Injuries in the 2012-13 and 2013-14 Season

522 Table. 2. 2012-13 Season vs. 2013-14 Season Correlation Analysis

523 Table. 3. Positional Breakdown between the seasons of 2012-13 and 2013-14 (n = 41)

524 Table. 4. Training and Match Load Metrics between the 2012-13 and 2013-14 seasons (n = 41)

525 Table. 5. Prediction of Overuse Injury Incidence Rates using Total Distance and High Speed Running
526 from GPS

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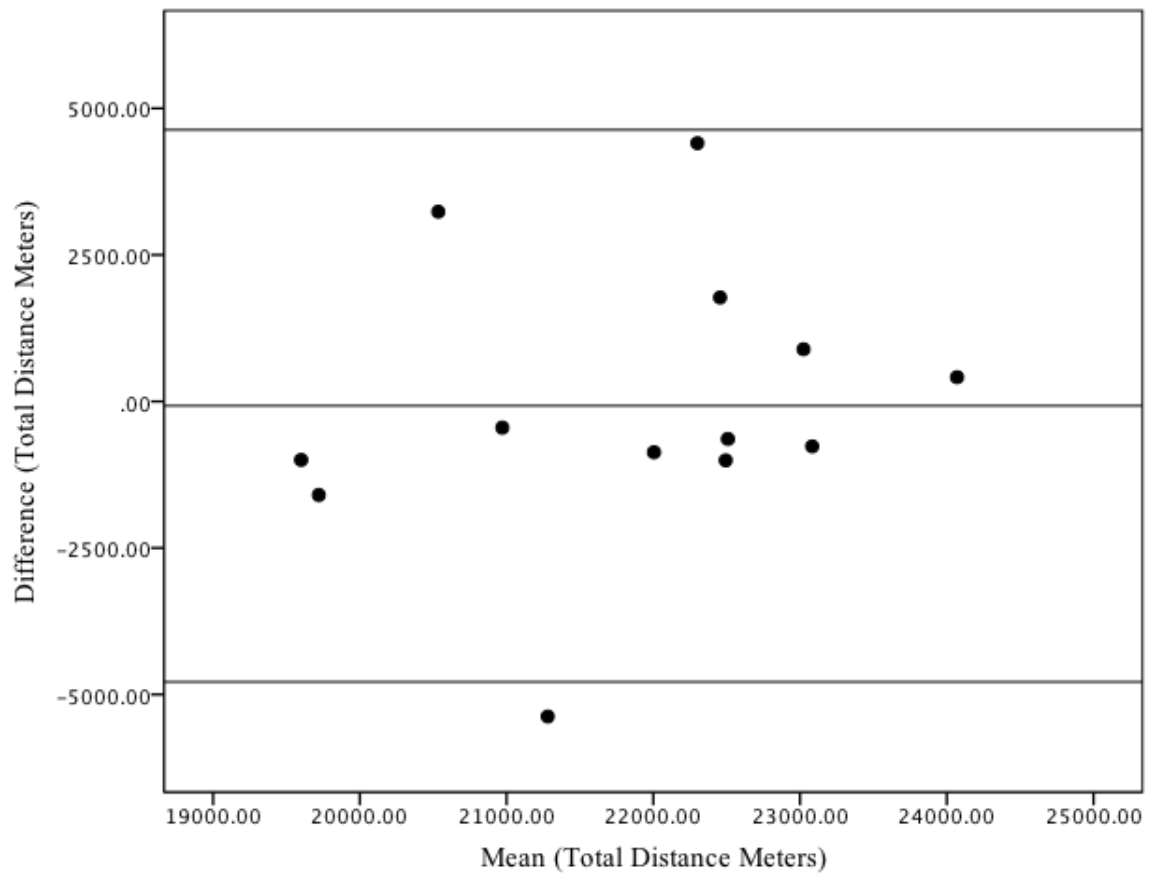


Figure 1. Limits of Agreement for 2012-13 and 2013-14 Seasons for Total Distance (m)

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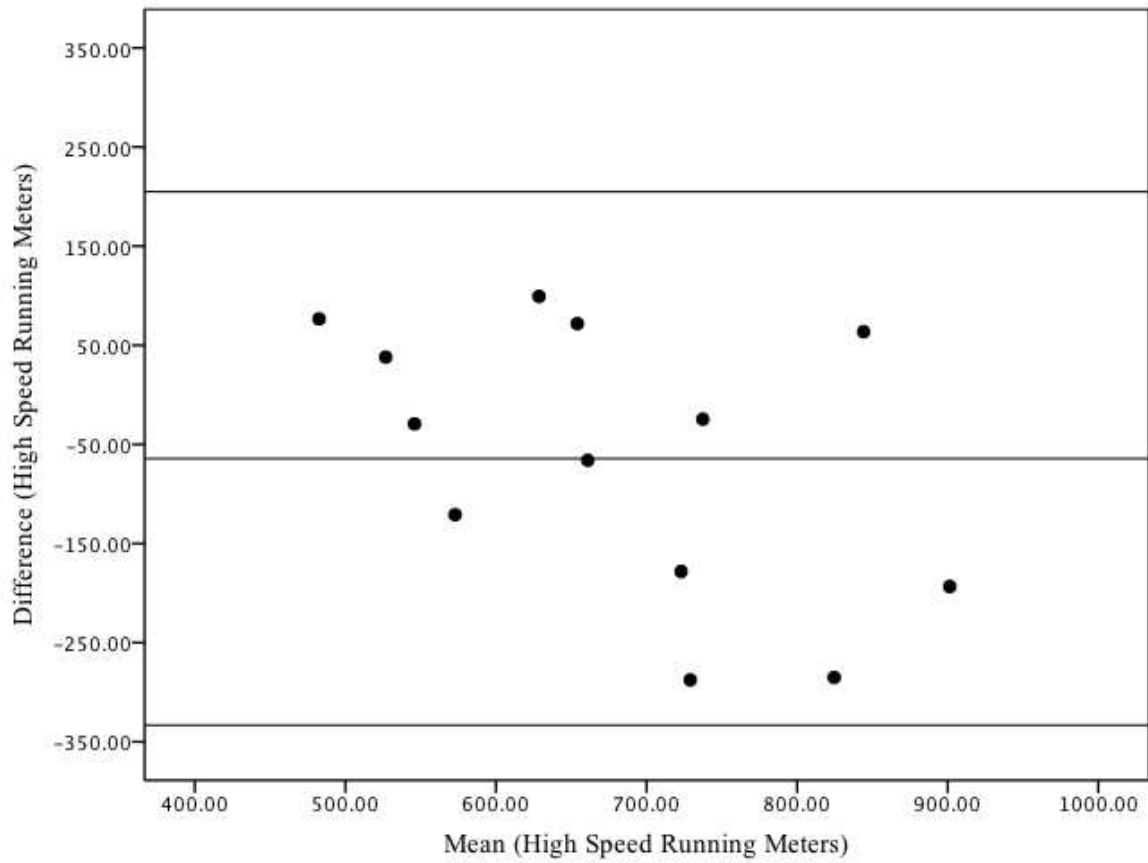


Figure 2. Limits of Agreement for 2012-13 and 2013-14 Seasons for High Speed Running (m)

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Prediction of overuse injuries in football

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|--|----|------|--------|----|------|--------|----|------|--------|---------|---------|
| Low (≤ 7 Days Missed) | 10 | 2.57 | 34.48% | 23 | 5.53 | 41.07% | 34 | 4.22 | 40.00% | 11.351* | 0.010† |
| Medium (≥ 8 Days Missed and ≤ 14 Days Missed) | 9 | 2.31 | 31.03% | 15 | 3.60 | 26.79% | 23 | 2.86 | 27.06% | 1.445* | 0.695** |
| High (≥ 15 Days Missed) | 10 | 2.57 | 34.48% | 18 | 4.33 | 32.14% | 28 | 3.48 | 32.94% | 4.432* | 0.218** |

Mean injury incidence rate reported per 1,000 hours 'on-leg' training and match exposure (95% confidence interval) difference between seasons ($P > 0.05$) †Significant difference between seasons ($P < 0.05$)

*Violated assumptions **No significant

Table 2. 2012-13 Season vs. 2013-14 Season Correlation Analysis

| Statistic | Season vs. Season (Distance) | Season vs. Season (HSR) | Season vs. Season (Injuries) | Season vs. Season (Injury Incidence Rate) |
|------------------------|---------------------------------|----------------------------|---------------------------------|--|
| Pearson Correlation | -0.266 | 0.017 | 0.382* | 0.371* |
| Sig. (2-tailed) | 0.093 | 0.914 | 0.014* | 0.017* |
| R ² | 0.071 | 0.000 | 0.146 | 0.138 |

*Significant Correlation (P < 0.05)

Table 3. Positional Breakdown between the seasons of 2012-13 and 2013-14 (n = 41)

| Position | Total Distance Weekly Average (m) | High Speed Running Weekly Average (m) | Whole Season Total Distance Average (m) | Whole Season High Speed Running Total Average (m) | Injury Incidence Rate (per 1000hrs) |
|--------------------------------|--|--|--|--|--|
| Central Defenders (n = 11) | 21804.30 ± 1556.93 | 648.09 ± 136.30 | 760933.96 ± 98010.22 | 22328.06 ± 4783.54 | 10.02 ± 7.89 |
| Wide Defenders (n = 9) | 21665.77 ± 2558.58 | 751.68 ± 180.20 | 792227.28 ± 112516.40 | 27475.57 ± 7027.14 | 7.78 ± 4.44 |
| Central Midfielders (n = 9) | 21758.86 ± 2048.98 | 604.93 ± 83.52 | 722652.21 ± 223869.70 | 19502.94 ± 5124.87 | 14.22 ± 15.46 |
| Wide Midfielders (n = 4) | 22048.57 ± 982.18 | 952.36 ± 187.93 | 876445.14 ± 41207.99 | 37924.02 ± 7844.734 | 2.15 ± 2.49 |
| Forwards (n = 8) | 20598.46 ± 2976.82 | 767.81 ± 176.02 | 731448.03 ± 143052.00 | 26657.11 ± 5145.59 | 8.11 ± 7.16 |
| Squad Average | 21575.19 | 744.97 | 776741.32 | 26777.54 | 8.46 |
| Standard Deviation | 564.03 | 134.70 | 62063.78 | 7026.08 | 4.36 |

Table 4. Training and Match Load Metrics between the 2012-13 and 2013-14 seasons (n = 41)

| | OR (Exp (B)) | 95% CI | | P |
|--|--------------|--------|-------|-------|
| | | Lower | Upper | |
| 2 Week Cumulative Loads | | | | |
| Total Distance – Normal Load (Reference) | 1.000 | - | - | - |
| Total Distance – Low Load | 1.264 | 0.164 | 9.769 | 0.822 |
| Total Distance – High Load | 0.670 | 0.395 | 1.137 | 0.137 |
| High Speed Running – Normal Load (Reference) | 1.000 | - | - | - |
| High Speed Running – Low Load | 0.993 | 0.381 | 2.588 | 0.989 |
| High Speed Running – High Load | 0.580 | 0.330 | 1.021 | 0.059 |
| 3 Week Cumulative Loads | | | | |
| Total Distance – Normal Load (Reference) | 1.000 | - | - | - |
| Total Distance – Low Load | 0.688 | 0.290 | 1.635 | 0.397 |
| Total Distance – High Load | 0.953 | 0.442 | 2.054 | 0.903 |
| High Speed Running – Normal Load (Reference) | 1.000 | - | - | - |
| High Speed Running – Low Load | 0.506 | 0.212 | 1.206 | 0.124 |
| High Speed Running – High Load | 1.049 | 0.543 | 2.029 | 0.886 |
| 4 Week Cumulative Loads | | | | |
| Total Distance – Normal Load (Reference) | 1.000 | - | - | - |
| Total Distance – Low Load | 0.688 | 0.290 | 1.635 | 0.397 |
| Total Distance – High Load | 0.953 | 0.442 | 2.054 | 0.903 |
| High Speed Running – Normal Load (Reference) | 1.000 | - | - | - |
| High Speed Running – Low Load | 0.506 | 0.212 | 1.206 | 0.124 |
| High Speed Running – High Load | 1.049 | 0.543 | 2.029 | 0.886 |

OR = Odds Ratio, OR = 1.50 is indicative of a 50% increased risk and vice versa. For an OR to be significant, 95% confidence intervals (CIs) did not contain the null OR of 1.00.

Table 5. Prediction of Overuse Injury Incidence Rates using Total Distance and High Speed Running from GPS

| | Total Number of Cases | Total Injuries Sustained | Injury Incidence Rate (per 1000hrs) | F (df) | P | R² | B |
|--|------------------------------|---------------------------------|--|------------------|----------|----------------------|----------|
| Total Distance (Volume of Training Variable) | | | | (1, 39) 6.482 | 0.015* | 0.143 | -5.835 |
| Low Group | 9 | 18 | 14.65 | | | | |
| Normal Group | 26 | 42 | 8.93 | | | | |
| High Group | 6 | 4 | 2.95 | | | | |
| High Speed Running (Intensity of Training Variable) | | | | (1, 39) 1.003 | 0.323 | 0.025 | -2.728 |
| Low Group | 5 | 10 | 11.29 | | | | |
| Normal Group | 29 | 46 | 9.75 | | | | |
| High Group | 7 | 8 | 6.08 | | | | |

* Denotes Statistical Significance (P < 0.05)