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An analysis of training loads in elite under 18 Australian Rule football players

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1 Abstract

2 Differences in training loads (TL) between under 18 (U18) Australian Rules football (AF)
3 State Academy selected and non-selected players were investigated. Players were categorised
4 relating to their highest representative level; State Academy selected (n = 9) and TAC Cup
5 level players (n = 38). Data were obtained from an online training-monitoring tool
6 implemented to collect player training and match information across a 20 - week period
7 during the regular season. Parameters modelled included AF skills, strength, and other sport
8 training sessions. Descriptive statistics (mean \pm SD) and between-group comparisons
9 (Cohen's d) were computed. A J48 decision tree modelled which TL variables could predict
10 selection level. Pooled data showed 60% of weekly training duration consisted of AF training
11 sessions. Similar AF TL were reported between State Academy and TAC Cup players (1578
12 \pm 1264 arbitrary units (AU) v 1368 \pm 872 AU; d = .05). While higher TL were reported for
13 State selected players comparative to TAC Cup in total training (d = .20), core stability (d =
14 .36), flexibility (d = .44), on-feet conditioning (d = .26), and off-feet conditioning (d = .26).
15 Decision tree analysis showed core stability duration and flexibility TL the most influential
16 parameters in classifying group selection (97.7% accuracy TAC Cup level; 35.8% accuracy
17 State Academy level). Insights of U18 AF players' weekly training structures, loads, and
18 characteristics of higher achieving players are provided. This study supports the application
19 of training diaries and session rating of perceived exertion (sRPE) for TL monitoring in
20 junior athletes.

21

22 **Key Words:** team sports; session RPE; talent identification; internal loads; junior athletes

23

24 INTRODUCTION

25 The Australian Football League (AFL) has established a talent development pathway for
26 junior players aimed at identifying, fostering, and progressing players towards an elite
27 Australian Rules football (AF) career. Levels including State Academies and National
28 Championships for age groups ranging from Under 14 to Under 18 years (U14 - U18 years),
29 are implemented nationwide and run along-side each State's participation pathways. In key
30 relevance to this study, the Transport Accident Commission (TAC) Cup is a Victoria state-
31 wide U18 representative competition for players to compete in high quality football and
32 developmental opportunities. The competition acts as one of the primary recruitment grounds
33 for selection into the Victorian State Metropolitan or Country teams, National Academy, and
34 scouting process for AFL clubs and semi-professional State league clubs.

35 Talent development and training practices for junior elite AF players are evolving to
36 incorporate a more scientific and measured approach as seen in the senior elite competitions.
37 The increased use of global positioning system (GPS) technology, individual athlete load
38 monitoring ⁽²⁵⁾, and online athlete self-reporting applications reflects a greater focus on grass
39 root development of AF players. An increased understanding of physical demands on players
40 from previous studies looking into junior elite AF match profiles ^(2, 21, 22) and athlete loads ^{(12,}
41 ¹³⁾ has also allowed for ongoing refinement of coaching practices and athlete management.
42 For example, match physical and technical differences between elite U16 and U18 AF
43 players have been reported ⁽²⁴⁾, including contested marks, clearances, total marks, and
44 relative distance ($\text{m}\cdot\text{min}^{-1}$). Greater statistical information of junior players could contribute
45 to improving progression and retention of talented players into the senior elite leagues. Apart
46 from the use of this data for match play performance enhancement, coaches could further
47 adapt training to suit age level, developmental stage, and playing position. Again, ensuring
48 appropriate loads are administered and effectively monitored.

49 Talented players may be exposed to higher training load (TL) in order to complete the
50 required tasks for selection at various levels of sport talent pathways ⁽¹⁰⁾. For example, U18
51 TAC Cup players may be involved in local club and school football competitions, or other
52 sports (e.g., basketball), whilst potentially being selected in State and National Academies.
53 The impacts of these additional training loads specifically on U18 AF player development is
54 not yet fully known. By using self-reported training measures, this study will examine the
55 training characteristics of U18 TAC Cup players throughout the 2016 playing season.
56 Previous studies have reported on the physical and match demands of TAC Cup players ^{(12,}
57 ¹³⁾. But it is not yet known the breakdown of total TL including extra training activities such
58 as participating in other organised sports simultaneously. Previous research on junior rugby
59 union players concluded that commitment to several levels of rugby teams, training and
60 matches, combined with outside sports participation created numerous high-load and impact
61 sessions throughout a week ⁽¹⁰⁾.

62 A previous systematic review ⁽⁷⁾ of the major football codes (American, AF, Gaelic,
63 rugby codes and soccer) examining the relationship between workloads, performance, injury,
64 and illness in adolescent male players acknowledged the need for further research in the area.
65 Particularly, training does-response relationships and effects of additional training. Results
66 indicated significant positive relationships between physical stress and traumatic injury,
67 furthermore that training duration was significantly associated with illness ⁽⁷⁾. Consistent
68 study results from multiple youth sports indicate a linear relationship between hours
69 participated and injury risk; greater than 16 hours per weeks specifically ⁽⁴⁾. Yet there are
70 changing views with evidence to suggest that appropriately prescribed and monitored high
71 TL will develop physical qualities in athletes that provide a protective effect against injury ⁽⁸⁾.

72

73 The aim of this study was to determine whether differences in TL existed between the
74 selection level of U18 AF players during the regular playing season. Furthermore, to
75 determine which combination of training type parameters would classify a player's training
76 week and level as either a TAC Cup player or higher selected State-team player. It was
77 hypothesized that higher selected State Academy players would record greater AF specific
78 training and associated developmental training such as strength sessions. This would be
79 accompanied by lower other outside sport involvement comparative to TAC Cup level
80 players.

81

82 **METHODS**

83 **Subjects**

84 A sample of 47 players registered with two TAC Cup clubs was available for participation in
85 the study (n = 17 club 1; n = 30 club 2). Participants were categorised into two groups based
86 on their highest representative level as supplied by the TAC Cup clubs; State Academy
87 selected (n = 9; male, age: 16.9 ± 0.3 years) or TAC Cup level (n = 38; male, age: $16.8 \pm .8$
88 years) therefore not selected in the higher State Academy level. The players trained and
89 competed in matches for their TAC Cup club, school team, local team, or State squad based
90 on coaches' selection, prior commitment requirements, and player availability during the data
91 collection period. Training sessions for both TAC Cup clubs were held on Monday, Tuesday
92 and Thursday evenings. The study and its methods were approved by the relevant Human
93 Research Ethics Committee. Parental or guardian signed consent was obtained for all players
94 under 18 years of age.

95

96

97

98 **Experimental Approach to the Problem**

99 Data were collected over a 20-week period during the regular playing season of the 2016
100 TAC Cup competition from rounds one to 16 inclusive (including four bye rounds).
101 Participants were provided with access to an online training monitoring tool (Smartabase:
102 Version 4.835, Fusion Sport, Queensland, Australia) for the purpose of self-reporting daily
103 training activity. Prior to the season, players were educated on how to correctly fill out the
104 diaries, including categorising training types and recording RPE scores. Players were
105 instructed to enter individual data each day related to all training undertaken throughout the
106 2016 TAC Cup competition (March to August) in the set questionnaire. The completion of
107 the diaries was self-directed from a player's perspective which may have created possibility
108 for players to misclassify certain sessions based on their own subjective interpretation of the
109 education mentioned above. The training load parameters included for modelling were: AF
110 training – scheduled sessions with their AF team; other sport training – any training or
111 competition undertaken with another sport outside AF; core stability – specific core work
112 conducted in an athlete's own time from a recommended program provided by the club's
113 strength coach; strength training – dedicated strength sessions either with their AF club or on
114 own; flexibility – dedicated flexibility sessions conducted on own from a recommended
115 program provided by the club's strength coach; on-feet conditioning – all dynamic
116 conditioning (e.g. run intervals, plyometrics); off-feet conditioning – all static or passive
117 conditioning work (e.g. stretching); total training – sum of all training conducted from each
118 training type.

119 **Procedures**

120 Internal TL was calculated through the session rating of perceived exertion (sRPE) method
121 by multiplying the total training duration (min) by the sRPE rating from the CR10 scale (AU)
122 ⁽⁵⁾. All raw data exported from the Smartabase software was imported into a custom designed

123 Microsoft Excel™ spreadsheet (Microsoft Corporation, Redmond, USA), and pre-processed
124 ⁽¹⁷⁾. Any identified abnormalities such as incorrectly entered time format data (reporting in
125 hours instead of minutes), or inconsistencies in recording a zero or leaving blank in entries
126 were rectified. Players were coded with an assigned identification number to de-identify the
127 data; and then level coded based on highest squad selection, State Academy (1) or TAC Cup
128 level (2). Cleaned data were organized to show all measures across a single row for each
129 player on each day of data entry provided, and weekly averages calculated. This resulted in
130 726 individual weekly load profiles for analysis.

131

132 **Statistical Analysis**

133 Descriptive data are presented as mean \pm standard deviation (SD). The effect size (ES) for
134 each measure for between group distances was calculated using Cohen's *d* statistic on a
135 customised Microsoft Excel™ spreadsheet, indicating a small or trivial ($d = 0 - .2$), moderate
136 ($d = .2 - .5$), large ($d = .5 - .8$), and very large ($d > .8$) effect ⁽³⁾. The confidence interval (CI)
137 was expressed as 90% representing the uncertainty in each effect and as probability that the
138 true effect was considerably positive or negative ⁽¹⁴⁾.

139 In addition to quantifying the differences between the two groups, a supervised
140 learning model was developed to provide a classification prediction for State Academy
141 selected and non-selected participants based on TL parameters. Given the uneven group
142 numbers, multiple blank events for some categories as well as 'zeroes' recorded in some
143 weeks, a number of data transformation techniques were attempted in order to normalise the
144 data. All of these were unsuccessful however, meaning that a non-parametric, machine
145 learning approach was implemented. Specifically, using the 'RWeka' package in R (R
146 Computing Environment) ^(15, 23). A J48 decision tree modelled each of the weekly load
147 profiles included in the dataset to classify player selection level in relation to TL measures.

148 All eight load parameters were included in the model, whilst a confidence value of 0.25 was
149 set and a minimum support of 10 instances required in order for a node to split. Model
150 performance was reported as classification accuracy of both groups and compared to the null
151 model.

152

153 RESULTS

154 The breakdown of weekly training duration types indicated that the majority of training for
155 this cohort was AF based sessions followed by strength training (Table 1); which is also
156 reflected in weekly sRPE TL (Table 1).

157

158 ****Table 1 near here****

159 ****Table 2 near here****

160

161 State Academy selected players in comparison to TAC Cup players had higher
162 weekly training durations in core stability (ES = 0.40; CI = -0.16 to -0.64), strength (0.23;
163 0.01 to -0.47), flexibility (0.37; -0.13 to -0.61), on-foot conditioning (0.28; -0.04 to -0.52), and
164 off-foot conditioning (0.26; -0.02 to -0.50) (Table 2). State Academy selected players also
165 showed higher weekly training loads in total training (ES = 0.20; CI = 0.04 to -0.44), core
166 stability (0.36; -0.12 to -0.60), flexibility (0.44; -0.20 to -0.68), on-foot conditioning (0.26; -
167 0.02 to -0.50), and off-foot conditioning (0.26; -0.02 to -0.50) (Table 2). In breaking down
168 training sRPE loads for each training type across four-week blocks between the two groups,
169 marked TL differences showed TAC Cup level players has larger loads in weeks 13, 14 and
170 15 compared to State selected players (Figures 1a and 1b). Other sports reported in the
171 training diaries included volleyball, rowing, swimming, soccer, hockey, tennis, athletics,
172 basketball, bike riding, own gym sessions, and netball.

173 ****Figures 1a, 1b, and 1c near here****

174

175 Decision tree evaluation analysed a total of 567 training weeks (78.1% of total
176 sample) including TAC Cup level players, and 159 weeks were reported including State
177 Academy selected participants. Results indicate that core stability duration and flexibility TL
178 are the most important interaction in parameters to classifying the two groups (Figure 2). This
179 is shown by the tree terminating down the right side at nodes 1 and 2 after just one branch
180 from the root node, weekly core stability duration greater than 33 minutes to weekly
181 flexibility TL. On the left side of the figure, the interaction between higher weekly off-feet
182 conditioning durations and weekly AF TL is also suggested as a strong predictor of player
183 selection level, classifying TAC Cup level 23 out of the 31 weeks (node 4) and State
184 Academy 10 out of the 12 weeks (node 5). The asymmetry in the decision tree output
185 indicates that TAC Cup level and State Academy training behaviour have different nuances.
186 There are greater interactions in parameters to classify TAC Cup level players based on their
187 training characteristics (nodes 2 – 4, 6, 7, 9) than State level players (nodes 1, 5, 8, 10).
188 Model performance was reported as 83.3%, which constituted only a moderate improvement
189 on the 78.1% null model. Of this, the model displayed an accuracy of 97.7% in classifying
190 TAC Cup level players (554 of 557 weeks) and 35.8% accuracy in classifying State Academy
191 players (51 of 157 weeks).

192

193 ****Figure 2 near here****

194

195

196

197

198 **DISCUSSION**

199 This study provides an insight into the internal TL of two elite U18 AF teams during the
200 regular playing season. These data provide a greater understanding of TL completed by elite
201 U18 AF players, which is currently underrepresented within the scientific literature. The
202 main findings were that State Academy selected players in comparison to TAC Cup level
203 players showed greater total weekly TL (AU) for total training, core stability, flexibility, on-
204 and off-feet conditioning ($d \geq .2$). Furthermore, greater total weekly training durations (min)
205 for core stability, strength, flexibility, on- and off-feet conditioning ($d \geq .2$).

206 This study's results are in agreement with previous studies showing that higher
207 selected players have greater AF weekly training durations and higher total training weekly
208 durations^(12, 13). Similarities also exist showing that higher selected players had lower other
209 football activity loads and training type variation⁽¹³⁾. It is common practice for players not
210 selected in their TAC Cup team for a weekend match to return to their local or school team
211 (football or other sports) and subsequently complete extra training sessions. This study
212 furthers the current knowledge by firstly examining selected State Academy level TAC Cup
213 players against non-State selected TAC Cup level players; and secondly breaking down their
214 training types for more descriptive measures.

215 Comparing sRPE loads between senior and junior elite players can be difficult
216 pertaining to a range of factors including differences in physiology, performance indicators
217⁽²⁾, and experience resulting in exertion perception variations⁽⁹⁾. Also, that senior elite AFL
218 clubs are professionally run entities with players employed as full-time athletes under strict
219 periodised training regimes. Previous study results⁽¹⁾ add that RPE is not linear in occurrence
220 and therefore each player's TL responses should take into account the context of previous,
221 current and future loading patterns.

222 Gaining information on training loads of junior players looking to progress into senior
223 elite tiers may be useful in assessing player development requirements in preparing for the
224 demands of senior AF.

225 Higher loads in the early in-season may be a continuation of pre-season loads as
226 reflected in periodisation strategies adopted by senior AFL teams ⁽¹⁸⁾. This periodisation
227 strategy sees higher conditioning and skills loading during the pre-season as preparation for
228 the playing season; which in contrast sees a majority of loading from weekly matches and
229 training focus shift to recovery, technical skills and conditioning maintenance ⁽¹⁸⁾. Higher
230 early in-season TL is also in part due to increased “other sports” TL (Figure 1b), which may
231 suggest players are still training and competing in their chosen summer sports, such as
232 rowing and soccer. Lower mid-season loads may have occurred for several reasons. It may
233 represent the league bye weekends in weeks eight, nine, 12, 17. Furthermore, State selected
234 players would likely have been competing in the National U18 Championship tournament
235 played during this time, which may imply minimal training was performed. Another reason
236 could be part due to compliance issues, and levels of education and guidance throughout the
237 season. Players may have been keen to complete the diaries early at its implementation, then
238 experienced a decline in motivation during the year. This lack of compliance and accuracy in
239 reporting may impact on the significance of the findings for the current study. Scope for
240 further investigation may be required to assess the accuracy and implementation complexity
241 of self-reported training diaries in U18 AF players. The use of external measures would
242 provide an objective measure for comparison to self-reported data. This would highlight any
243 problems with over- or under-estimating durations.

244 By comparing State Academy selected players to TAC Cup level players, the
245 Academy group engaged in a greater proportion of AF specific training, although the non-
246 State selected group showed slightly greater mean weekly AF TL, albeit trivial ($d = .05$).

247 An explanation for this may be the Academy players having greater on-feet
248 conditioning durations and lower RPE. Completing more conditioning work would imply that
249 the Academy players are more physically fit and therefore cope better with training demands,
250 hence rating sessions lower on the RPE scale ^(9, 20, 26). Notions of specialisation amongst State
251 Academy selected players is reflected in their greater emphasis and loading in AF training
252 considering the next stage of the talent pathway would be National Academy and Draft
253 selection in pursuit of a professional AFL career. Research results looking at junior elite
254 rugby union suggested evidence of deliberate practice in higher-level players could be seen in
255 the higher proportion of weekly training activities related to rugby ⁽¹¹⁾. In relation to training
256 load management and injury prevention, the importance of strength, conditioning and
257 functional movement training for both pre-and in-season aids to reduce the cited risk factors
258 for injury ⁽⁴⁾. These include lack of lean tissue mass, increased joint hypermobility and
259 imbalances from growth, have been emphasised for youth player development ⁽⁴⁾.

260 Applying a machine learning approach decision tree analysis showed multiple rules
261 capable of classifying selection level based on the TL measures (Figure 2). Weekly core
262 stability durations appeared to be an influencing factor in facilitating higher selection
263 classification, particularly showing a strong relationship with a weekly flexibility sRPE load
264 greater than 115 AU. It was not a stipulated requirement for State Academy players to be
265 completing extra core training outside of their TAC Cup or Academy team sessions. These
266 results suggest that higher selected players may take it upon themselves to complete these
267 extra conditioning sessions due to their motivation to achieve success within the sport. Other
268 rules included, if core stability duration is ≤ 33 min, weekly off-feet conditioning duration is
269 ≤ 40 min, flexibility load is > 115 min, but other sport duration is > 0 min will likely result in
270 TAC Cup level (12 out of 13 weeks identified). Decision trees provide a means to model non-
271 linear trends and provide visual representation for ease of interpretation ⁽¹⁹⁾.

272 This method for classification has previously been applied in senior AF to explain
273 match outcome (win/loss) based on team performance indicators ⁽¹⁹⁾. Previously it has been
274 acknowledged that addressing the research gaps in respects to effects of workloads by
275 incorporating non-linear models and/or machine learning techniques, internal and external
276 measurements, would lend to more efficient training practices for youth athletes ⁽⁷⁾. In this
277 study however, the poor performance of the model with respect to classification of State
278 Academy players suggests that further parameters are needed to improve the accuracy in
279 future research. This also suggests that it is likely that additional non-training load related
280 factors contribute to discriminating the two cohorts. With respect to the decision tree design,
281 although the minimum support instances could be increased, this would have resulted in a
282 reduced decision tree size, which may not have provided a full representation of the data.
283 Further work is also required to assess the generalisability of the model to subsequent years
284 and AF cohorts, as the results from this model are only applicable to the 2016 training data
285 collected from the participants included in this study.

286 Despite the findings, it is acknowledged that analysis only included two of the 12
287 teams competing within the TAC Cup competition; and therefore, the findings may be
288 specific to each team's training structure and coaching philosophy. A greater data input may
289 have been prevented due to a lack of compliance from athletes regularly filling out or failing
290 to correctly fill out training dairies on a regular basis during the season. Furthermore,
291 although both clubs received education on how to complete the training diaries including
292 using the RPE scale, the level of individual athlete understanding and consistency in self-
293 reporting throughout the year may have varied. Although the use of external load measures
294 such as GPS would have provided a more in-depth insight into these athletes' TL, resource
295 limitations and logistical practicality prevented the acquisition of significant data levels for
296 the analysis required.

297 Future work investigating the association between sRPE TL and external load
298 measures in juniors elite AF by similar methods as seen the professional AFL ⁽¹⁾ would be
299 beneficial in moving towards individualised athlete monitoring and training structures to
300 maximise performance.

301

302 **CONCLUSIONS**

303 This study has quantified the TL of elite U18 Australian Rules football players across extra
304 multiple session types. Also, assessed differences between State Academy selected and non-
305 State Academy selected TAC Cup level players. The results from this study showed State
306 Academy selected players are completing more AF specific training and accumulating greater
307 weekly loads. TAC Cup level U18 players are accumulating greater other sport weekly TL.
308 TAC Cup players rate (RPE method) their AF training harder as reflected in having lower
309 durations and higher sRPE TL compared to Academy players. Further analysis indicated that
310 core stability duration and flexibility TL were important factors in modelled classification for
311 group level selection. These findings add to the growing body of research in junior AF and
312 specifically provide greater insight into the player's weekly training structures.

313

314 **PRACTICAL APPLICATIONS**

315 The methods and outcomes of this study may assist coaching staff in making more informed
316 decisions on training structures in-line with a player's selection status. It may encourage
317 coaches to review player training management in terms of factoring in outside sport and TL
318 to ensure their players are training and competing at optimal levels for their TAC Cup club.
319 Furthermore, the results highlight the training characteristics of higher selected players.

320

321 This study reflects the practical application of self-reported training diaries and sRPE
 322 TL in junior sports as an effective low-cost method. Training diaries may provide
 323 complimentary information alongside objective measures, such as GPS. Or serve as a tool for
 324 player TL insight when objective measures may not be readily available in junior AF teams.
 325 Several studies have supported the use of the RPE method and training diaries for junior
 326 team-sport athletes⁽¹⁶⁾, junior AF^(12,13), junior soccer⁽⁶⁾ and junior rugby union^(10,11).

327

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332

333 **Disclosure of Interest**

334 The authors report no conflicts of interest.

335

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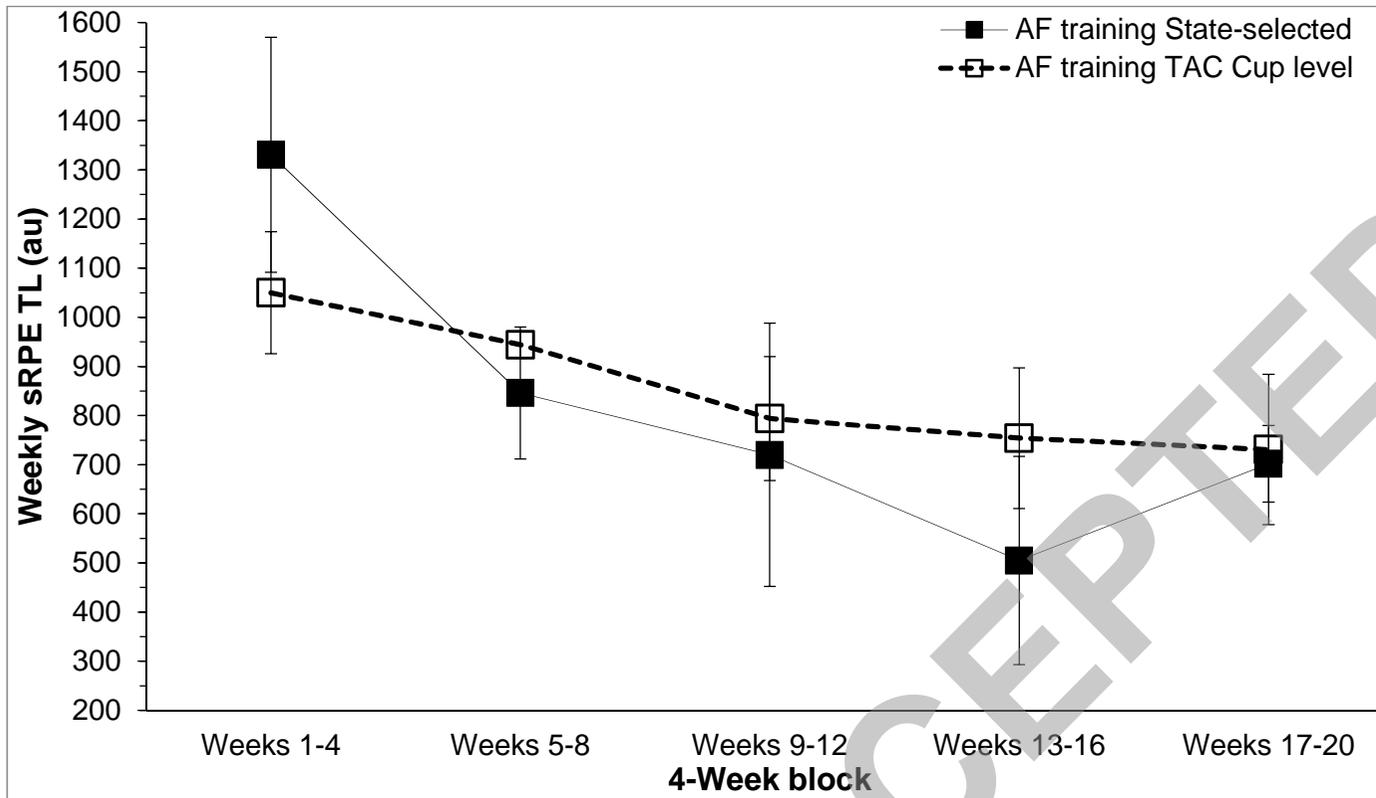
Table 1. Weekly training durations and sRPE TL of U18 TAC Cup players across both levels.

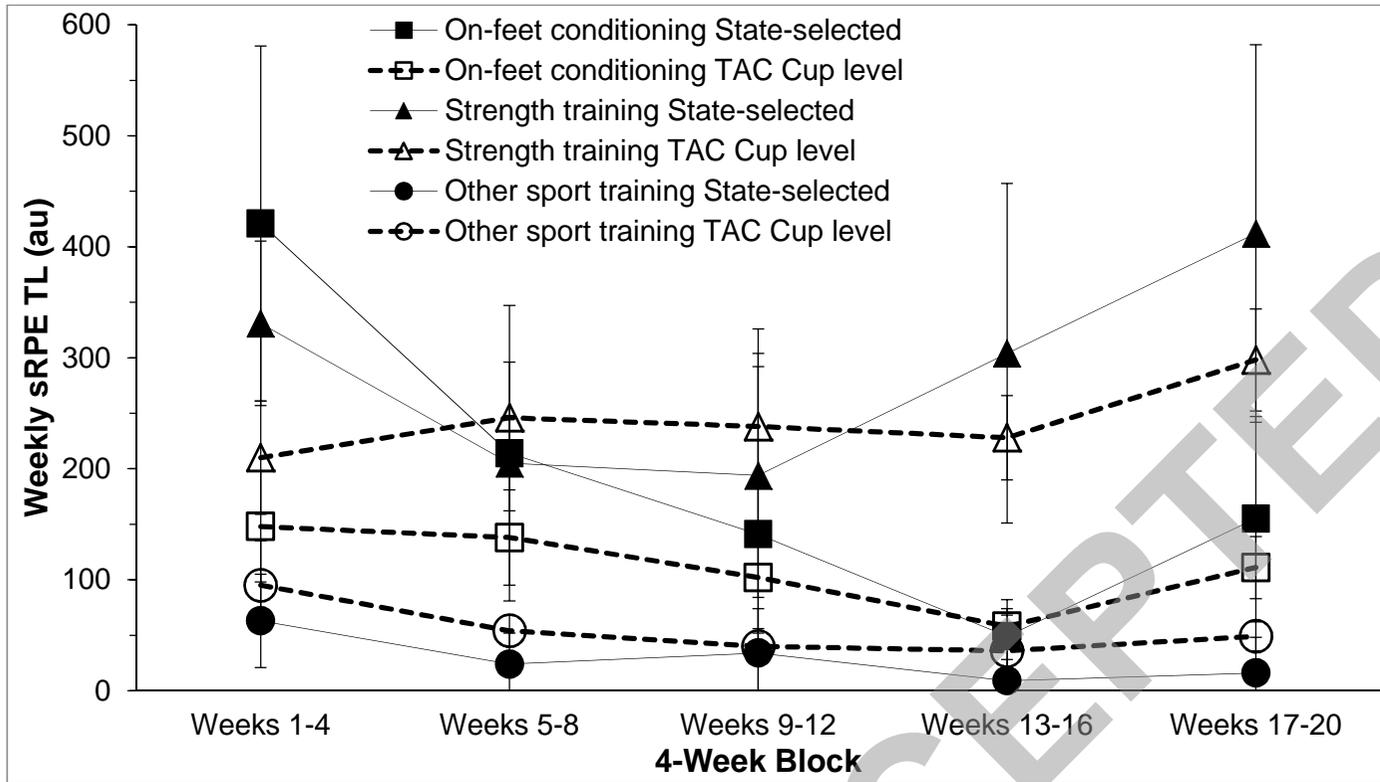
Training Type	Training Duration		sRPE Training Loads	
	Mean \pm SD (min)	% of total weekly training duration	Mean \pm SD (AU)	% of total weekly TL
Weekly total training	241 \pm 153		1414 \pm 940	
Weekly AF training	144 \pm 91	59.8	861 \pm 592	60.9
Weekly other sport training	8 \pm 27	3.4	49 \pm 173	3.5
Weekly core stability training	6 \pm 16	2.6	36 \pm 98	2.6
Weekly strength training	39 \pm 63	16.4	250 \pm 431	17.7
Weekly flexibility training	15 \pm 27	6.2	55 \pm 110	3.9
Weekly on-feet conditioning	23 \pm 50	9.4	132 \pm 312	9.4
Weekly off-feet conditioning	5 \pm 21	2.2	30 \pm 128	2.1

Table 2. Descriptive statistics of weekly TL and duration for each training type, TAC level and State Academy selected players. Data presented as mean \pm SD. The between group differences is presented as an effect size (Cohen's *d*), with 90% confidence intervals.

TL measure	State Academy selected	TAC Cup Level	<i>d</i> (90% CI)
Weekly total training sRPE load (AU)	1578 \pm 1264	1368 \pm 822	.20 (.04 to -.44)
Weekly AF sRPE TL (AU)	835 \pm 674	868 \pm 567	-.05 (.29 to -.19)
Weekly other sport sRPE TL (AU)	31 \pm 131	55 \pm 183	-.15 (.39 to -.09)
Weekly core stability sRPE TL (AU)	69 \pm 148	27 \pm 76	.36 (-.12 to -.60)
Weekly strength sRPE TL (AU)	284 \pm 427	241 \pm 432	.10 (.14 to -.34)
Weekly flexibility sRPE TL (AU)	95 \pm 128	44 \pm 101	.44 (-.20 to -.68)
Weekly on-feet conditioning sRPE TL (AU)	203 \pm 418	113 \pm 272	.26 (-.02 to -.50)
Weekly off-feet conditioning sRPE TL (AU)	62 \pm 202	22 \pm 97	.26 (-.02 to -.50)
Weekly total training duration (min)	285 \pm 214	228 \pm 128	.32 (-.08 to -.56)
Weekly AF training duration (min)	147 \pm 106	143 \pm 87	.05 (.19 to -.28)
Weekly other sport training duration (min)	6 \pm 24	9 \pm 28	-.12 (.36 to -.12)
Weekly core stability training duration (min)	13 \pm 25	5 \pm 12	.40 (-.16 to -.64)
Weekly strength training duration (min)	51 \pm 75	36 \pm 58	.23 (.001 to -.47)
Weekly flexibility training duration (min)	23 \pm 33	13 \pm 25	.37 (-.13 to -.61)
Weekly on-feet conditioning duration (min)	35 \pm 65	19 \pm 45	.28 (-.04 to -.52)
Weekly off-feet conditioning duration (min)	11 \pm 33	4 \pm 15	.26 (-.02 to -.50)

d is Cohen's effect size relative to the State selected players; Calculated using Cohen's *d* statistic, where an effect size of *d* = .20 was considered small, *d* = .50 moderate and *d* \geq .80 large (Cohen 1988).





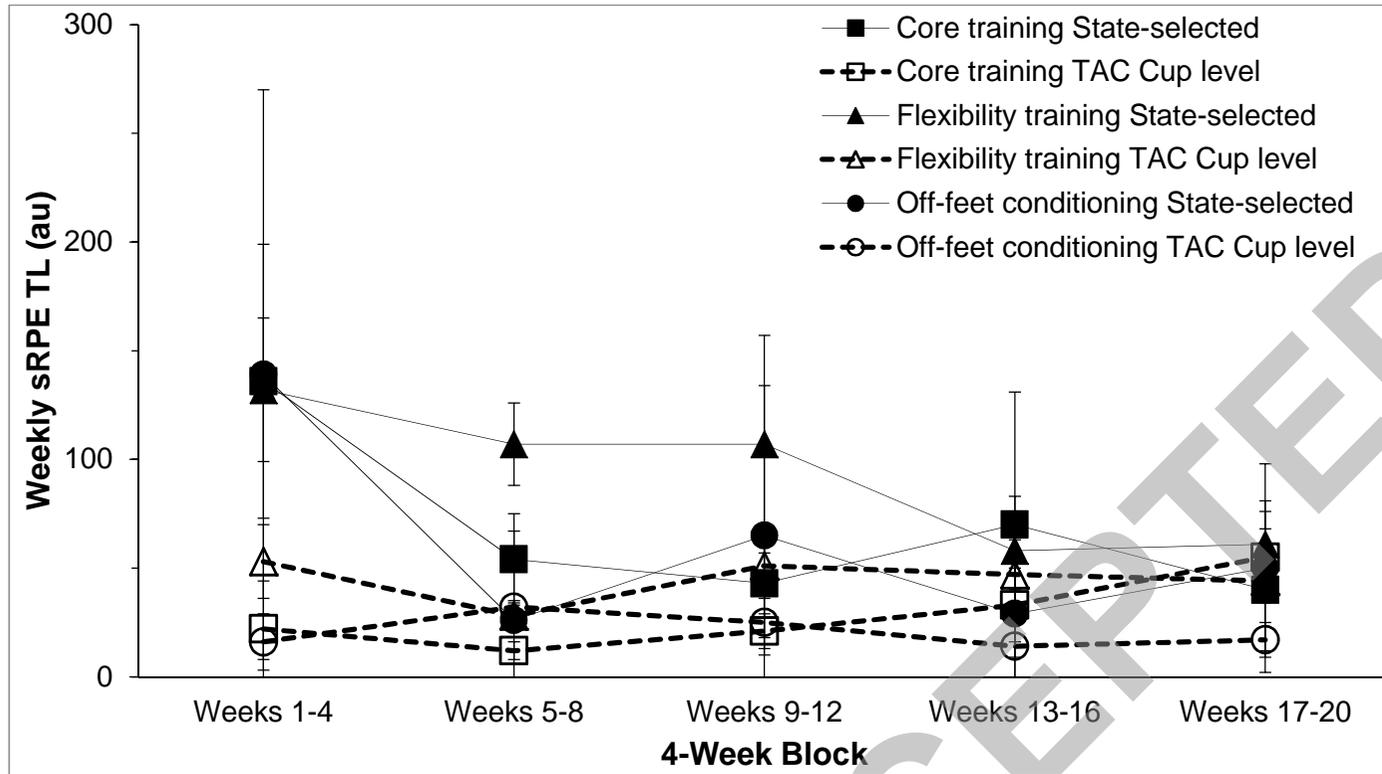


Figure 1. Weekly sRPE TL grouped in 4-week blocks for various training parameters between State-selected and TAC Cup level players. **Figure 1a:** AF training. **Figure 1b:** On-feet conditioning, Strength training, Other training. **Figure 1c:** Core stability, Flexibility training, Off-feet training. Data presented as mean \pm SD bars.

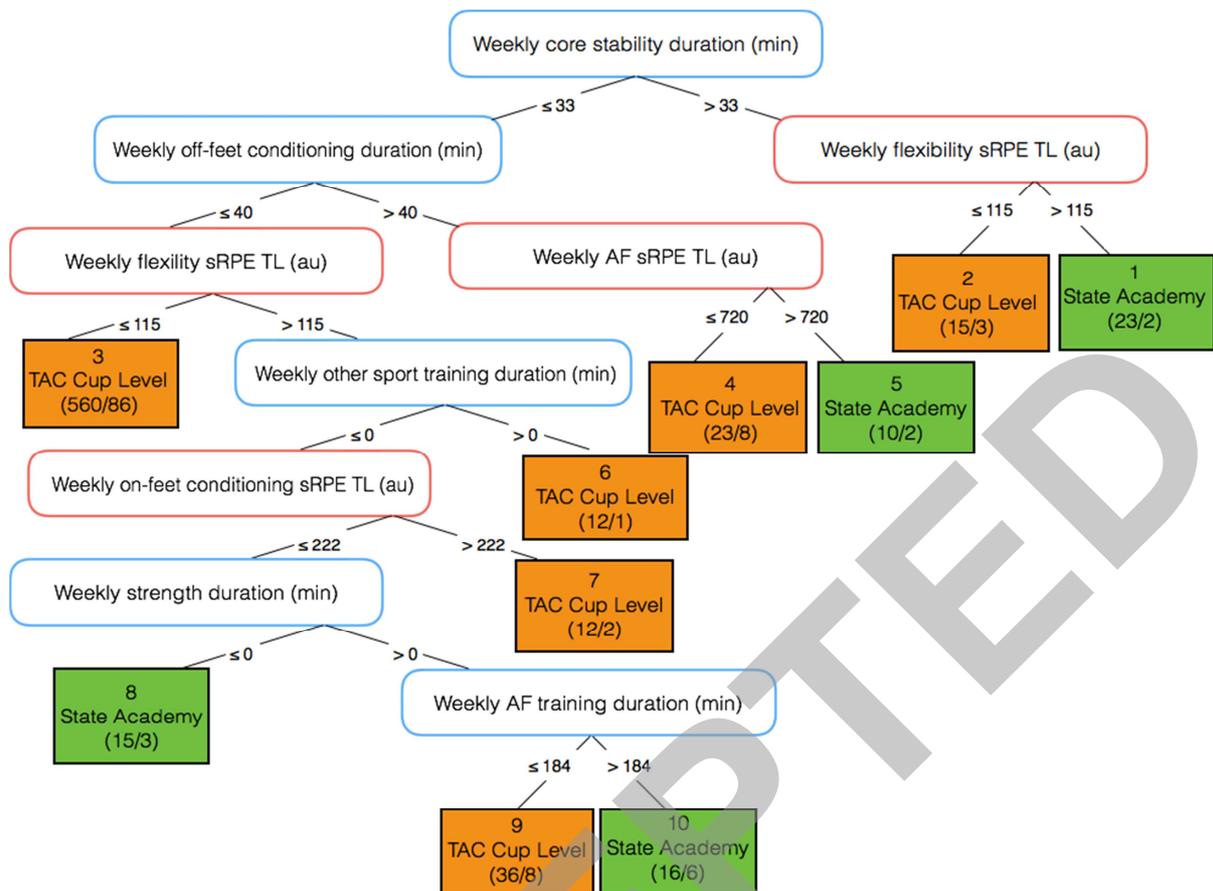


Figure 2. Decision tree analysis output explaining selection outcome based on reported training parameters. Leaf node class output reports correct/incorrect weeks reported according to identified player level, i.e. node 2: 15/3 classified TAC Cup level for 15 of the 18 weeks.