

1 **Technical testing and match analysis statistics as part of the talent**
2 **development process in an English football academy**

3 Adam L. Kelly^{1,2,3}, Mark R. Wilson², Daniel T. Jackson¹, and Craig A.
4 Williams²

5 *¹Research Centre for Life and Sport Sciences (CLaSS), School of Health Sciences, Department of*
6 *Sport and Exercise, Birmingham City University, Birmingham, West Midlands, United Kingdom;*

7 *²College of Life & Environmental Sciences, University of Exeter, Exeter, Devon, United*
8 *Kingdom; ³Exeter City Football Club, Exeter, Devon, United Kingdom*

9 Correspondence: Dr Adam L. Kelly, Department of Sport and Exercise, Birmingham City
10 University, City South Campus, Westbourne Road, Edgbaston, B15 3TN, UK. E-mail:
11 Adam.Kelly@bcu.ac.uk

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15 Technical ability is recognised as a fundamental prerequisite to achieve senior
16 professional status in football. However, research is yet to investigate what technical
17 attributes contribute to greater coach perceived potential within an academy
18 environment. Therefore, the aim of this study was to examine technical ability and skill
19 behaviour as contributing factors to coach potential ratings in an English football
20 academy. Ninety-eight outfield academy players (Foundation Development Phase
21 [FDP] under-9 to under-11 $n=40$; Youth Development Phase [YDP] under-12 to under-
22 16 $n=58$) participated in the study. Four football-specific technical tests were used to
23 measure technical ability, whilst eight match analysis statistics from competitive
24 match-play across an entire season were observed to measure skill behaviour. A
25 classification of 'higher-potentials' (top third) and 'lower-potentials' (bottom third)
26 were applied through coach rankings. Within the FDP, higher-potentials performed
27 significantly better ($P<0.05$) on the lob pass test, alongside greater reliability in
28 possession, pass completion, and total touches for match analysis statistics. Within the
29 YDP, higher-potentials performed significantly better ($P<0.05$) on all four technical
30 tests, alongside greater reliability in possession, dribble completion, and total touches
31 for match analysis statistics. Results suggest football-specific technical tests and 'in
32 possession' skill behaviours may provide discriminative tools that align with perceived
33 potential.

34 Keywords: Technical ability; Performance analysis; Skill behaviour; Talent
35 identification, Academy soccer, Football coaching

36 **Introduction**

37 Football is a sport that requires the repetition of many complex technical actions, such as
38 dribbling, passing, tackling, and shooting (Dardouri, Amin Selmi, Haj Sassi, Gharbi, Rebhi,
39 & Moalla, 2014; Figueiredo, Coelho-e-Silva, & Malina, 2011). Historically, objective
40 technical analysis was rarely monitored for talent development purposes (Abt, Zhou, &
41 Weatherby, 1998). For example, Ali (2011) states how there was a 'dearth' of studies on skill
42 execution within academic literature, particularly when it is readily acknowledged that

43 successful execution of skill is one of the most important aspects in football performance.
44 More recently however, the growing interest from practitioners, alongside an increase in
45 technology capabilities, has resulted in researchers focussing on technical tests and match
46 analysis statistics (e.g., Archer, Drysdale, & Bradley, 2016; Forsman, Grasten, Blomqvist,
47 Davids, liukkonen, & Konttinen, 2016; Pedretti, Pedretti, Fernandes, Rebelo, & Seabra,
48 2016).

49 Current research has illustrated the technical demands of contemporary football have
50 increased significantly in recent years (Barnes, Archer, Hogg, Bush, & Bradley, 2014).
51 Furthermore, there is a distinct association between greater ball possession and successful
52 results (Gomez, Mitrotasios, Armatas, & Lago-Penas, 2018; Liu, Hopkins, & Gomez, 2016;
53 Yang, Leicht, Lago, & Gomez, 2018). In addition, players from successful teams have been
54 regularly shown to complete more technical actions compared to their less successful
55 counterparts (Gomez et al., 2018; Rampinini, Impellizzeri, Castagna, Couus, & Wisloff,
56 2009). Therefore, from a talent development perspective, it may be important to monitor both
57 unopposed technique and skill behaviours in youth football, using technical tests and match
58 analysis data respectively, to measure these fundamental attributes to support greater
59 development strategies towards senior expertise.

60 ***Technical testing***

61 The acute motor skills of manipulating a ball effectively are vital factors in the professional
62 game of football and can be tested in isolation (Vaeyens, Malina, janssens, van Retergham,
63 Bourgois, Vrijens, & Philippaerts, 2006). Ali (2011) states the advantages of measuring these
64 technical attributes as: (a) facilitating initial talent identification, (b) providing a strategy for
65 skill acquisition, and (c) offering an alternative predictor for measuring technical ability
66 compared to a skilled execution during competitive match-play. The importance of technical

67 ability and successful football performance has been supported in previous studies, whereby
68 an association between technical capabilities and performance outcomes at varying
69 performance levels is demonstrated (e.g., Coelho-e-Silva et al., 2010; Figueiredo, Goncalves,
70 Coelho-e-Silva, & Malina, 2009; Huijgen, Elferink-Gamser, Lemmink, & Visscher, 2014;
71 Rebelo et al., 2013; Vaeyens et al., 2006).

72 Vaeyens and colleagues (2006) used a sequence of technical tests as part of their
73 research exploring the relationship between physical and technical performance
74 characteristics in youth football, revealing technical tests can distinguish ability groups in
75 youth football players at under-13 to under-16 age groups. Keller, Raynor, Bruce, and Iredale
76 (2016) used the Loughborough Short Passing Test, long passing test, shooting test, and speed
77 dribbling test to discriminate under-18 national 'elite', 'state elite', and 'sub-elite' youth
78 football players, reporting that the 'elite' group had higher scores compared to the others.
79 Huijgen, Elferink-Gemser, Post, and Visscher's (2010) longitudinal study also found that
80 dribbling performance during adolescence could discriminate between players who achieved
81 senior professional football status and those who reached amateur level. As a result, these
82 technical tests can be considered as valuable measures for assessing young football players'
83 potential.

84 Alongside ability groups, technical proficiency has been illustrated to improve with
85 age among youth football players, with the greatest developments shown to occur in pre-
86 pubertal years (Huijgen et al., 2010; Valente-dos-Santos et al., 2014; 2012; Wilson et al.,
87 2016). Additionally, some studies have reported growth and maturation status may also be
88 associated with technical skill development, with biological maturity impacting the technical
89 progression in young football players (Malina, Cumming, Kontos, Eisenmann, Ribeiro, &
90 Aroso, 2005; Malina, Ribeiro, Aroso, Cumming, Unnithan, & kirkendall, 2007; Valente-dos-
91 Santos et al., 2012; 2014). Moreover, time spent within practice activities, such as deliberate

92 practice, deliberate play, and multi-sports, has been allied with developing technical ability
93 within a football context (Huijgen, Elferink-Gemser, Ali, & Visscher, 2013; Huijgen et al.,
94 2010; Valente-dos-Santos et al., 2014). Consequently, this highlights the importance of
95 investigating technical ability from an age-specific perspective to support appropriate
96 developmental strategies in youth football.

97 *Match analysis statistics*

98 Football is characterised as a free-flowing team sport that requires the execution of many
99 aspects of skill in a dynamic context (Kempe, Vogelbein, Memmert, & Nopp, 2014). Thus,
100 although there are some ‘closed skills’ (i.e., penalty, corner, free-kick, throw-in), football is
101 an ‘open skill’ game; whereby players are required to perform the correct action at the right
102 moment to effectively operate (Carling, Williams, & Reilly, 2007). In addition, consistent
103 technique is required for a long period of time during a game, which has been shown to be
104 variable during the later stages of a game when fatigue sets in (Mohr, Krustup, & Bangsbo,
105 2003). Match analysis refers to the objective recording and examination of behavioural
106 events occurring during competition (Carling et al., 2007). The notational style of analysis,
107 generically applied within academies to recognise key skill behaviours, is an objective
108 method of providing data for player development (Appleby & Dawson, 2002; Hughes, 1988;
109 Hughes, Hughes, & Behan, 2007). The scientific analysis of sports performance aims to
110 advance understanding of game behaviour, with a view to improving future outcomes
111 (McGarry, 2009; Wright, Carling, & Collins, 2014). As such, match analysis, via recording
112 competitive games and objectively analysing them, provides both researchers and
113 practitioners useful data on individual skill execution in football.

114 Maintaining possession, through passing and preserving the ball within a team’s
115 control during competitive match-play, is associated with greater success at the highest levels

116 of professional football (Liu et al., 2016; Yang et al., 2018). Moreover, players from more
117 successful teams generally possess a greater pass completion percentage, alongside other
118 technical variables such as tackles, dribbles, and shots, during competitive match-play
119 (Rampinini et al., 2009; Yang et al., 2018). Gomez and colleagues (2018) also found greater
120 ball possession, more attacking actions, and lower individual challenges reflected a higher
121 league ranking at senior professional level. Although these characteristics are fundamental
122 skills in senior professional football, current research overlooks the potential significance
123 match analysis may provide for recognising and facilitating talent development in youth
124 football (Atan, Foskett, & Ali, 2014; James, 2006).

125 Whilst there are number of studies that have examined groups of youth athletes (i.e.,
126 ‘elite’ versus ‘non-elite’), which generally elicit superior technical abilities are possessed
127 within advanced cohorts (e.g., Vaeyens et al., 2006; Woods, Raynor, Bruce, & McDonald,
128 2015), there is no exploration regarding technical characteristics within an academy
129 environment that support developmental outcomes. Therefore, the purpose of this study was
130 to examine the discriminant function of technical ability (technical tests) and skill behaviours
131 (match analysis statistics) based on whether they could differentiate ‘higher-potentials’ and
132 ‘lower-potentials’ (coach potential rankings) from an age-specific perspective (Foundation
133 Development Phase [FDP] and Youth Development Phase [YDP]). It was hypothesised that
134 characteristics across the technical tests and match analysis statistics would differentiate
135 higher-potentials and lower-potentials within both age phases.

136 **Methods**

137 *Sample*

138 Ninety-eight participants were examined within their specific age phase; FDP (under-9 to
139 under-11; $n = 40$) and YDP (under-12 to under-16; $n = 58$). All participants were recruited

140 from the same Tier 4 English professional football club and their Category 3 academy. Only
141 outfield players were included due to the contrasting development pathway for goalkeepers
142 (Gil, Zabala-Lili, Bidaurrazaga-Letona, Aduna, Lekue, Santos-Concejero, & Granados,
143 2014). The Institutional Ethics Committee approved this study.

144 *Measures*

145 *Technical tests*

146 Four football-specific technical tests previously utilised in talent development research were
147 applied (Vaeyens et al., 2006). First, the slalom dribble test requires the player to control the
148 ball through nine cones (2 m apart) from the start to the end line and return. The timings are
149 recorded using timing gates (Brower TC Timing System, Draper, Utah, USA), with each
150 player completing two trials and the quicker of the two recorded for analysis. Second, the lob
151 pass test requires the player to kick the football from a distance of 20 m into a target area
152 divided into three concentric circles (3 m, 6 m, and 9.15 m in diameter). Each kick is scored
153 by the circle in which the ball initially landed (3, 2, and 1 point respectively). Ten attempts
154 (five with each foot) are attempted with a maximum of 30 points available. Third, the
155 shooting accuracy test requires the player to kick the ball at a 16 m wide goal target from a
156 shooting distance of 20 m and central to the goal. The goal was divided into five parallel
157 zones; centre, 2 m wide (3 points), two areas 3 m on each side of the centre (2 points), and
158 two areas 4 m wide at each extreme (1 point). Ten attempts (five with each foot) are
159 attempted with a maximum of 30 points available. Fourth, the ball juggling test requires the
160 player to keep a football off the ground with the total number of touches recorded. Two trials
161 are completed, with a maximum of 100 touches per attempt permitted, allowing a maximum
162 number of 200 touches. Each player completed these tests in an indoor sports hall with a
163 hard-wood floor, with generic training kit being worn. In addition, age group-specific balls

164 were used for the tests in-line with the Football Association regulations; size three for under-
165 9, size four for under-10 to under-13, and size five for under-14 to under-16.

166 *Match analysis statistics*

167 Video footage examined each player during competitive match-play as they performed each
168 skill behaviour. An average score of each skill behaviour is computed from across an entire
169 football season, including reliability in possession percentage, pass completion percentage,
170 number of tackles, number of blocks, number of loose balls retrieved, successful dribble
171 completion, total touches, and goals scored. As a standard pro-forma of match analysis
172 statistics within each academy varies based on its philosophy, this current study applied the
173 academy's existing protocol for its data collection. The specialist software Gamebreaker©
174 was used to perform participant analysis for each game and trained, club-appointed
175 Performance Analysts (who were not part of the research team and were blind to the grouping
176 of the study participants) adopted technical expert definitions (Table 1) to code behaviours (n
177 = 10). Twenty matches (25% of the data) of the matches that were included in the current
178 study were used to calculate the Performance Analysts' reliability (15-day test-retest
179 analysis). One match per team was randomly selected to carry out the intra- and inter-
180 reliability analysis. An intra-class correlation coefficient test was executed to analyse the
181 reliability levels (poor, <0.50; moderate, 0.50 to 0.75; good, 0.76 to 0.90; excellent, 0.91 to
182 1.00) (Koo & Li, 2016). Results showed the intra-observer reliability ranged from 0.76 to
183 1.00 and the inter-observer reliability ranged from 0.71 to 1.00 (Table 2).

184 *****Table 1 near here*****

185 *****Table 2 near here*****

186 Only home games were filmed and analysed unless an away team provided appropriate

187 footage (away footage accumulated 8.5% of overall footage). Each age group had a varied
188 number of games filmed and analysed ranging from seven to fourteen. Although all matches
189 analysed were performed on grass, weather and surface quality varied depending on the time
190 of the season. Additionally, as a result of age-specific development, match formats differed
191 throughout the season between age groups; for example, the under-9's generally played four
192 periods of 20 minutes with 5 vs. 5, compared to the under-16's who generally played two
193 periods of 40 minutes with 11 vs. 11. Age appropriate pitches and football size were also
194 applied. Eighty-one matches were filmed across the entire season, with each participant
195 playing a mean number of 7.3 games that were recorded for match analysis statistics. The
196 season accumulation subsequently supplied the match analysis statistics applied to this
197 research. The mean score for each skill behaviour was based on an 80 minute average in-line
198 with a full match duration (i.e., total number of skill behaviours divided by total number of
199 80 minute matches).

200 *Coach development rankings*

201 It is important to highlight that coach perception regarding talent development has been used
202 in previous empirical research (e.g., Kelly, Wilson, Jackson, Turnnidge, & Williams, 2020;
203 MacNamara & Collins, 2013). Indeed, coach observation and opinion is central to the
204 subjective nature of youth sport, with modern objective information readily available to
205 professional coaches to support their judgement (e.g., Sieghartsleitner, Zuber, Zibung, &
206 Conzelmann, 2019; Tangalos, Robertson, Spittle, & Gustin, 2015). Two coaches from each
207 age group ($n = 16$), who were deemed suitably qualified assessors (UEFA Pro, 'A', or 'B'
208 Licenced alongside either the FA Advanced Youth Award or the FA Youth Award), were
209 asked to rank their players from top to bottom in relation to their perception of the player's
210 potential to develop to senior professional status. This created a linear classification of

211 higher-potential players down to their lower-potential peers, with each age group then split
212 into thirds using tertiles. This created a group of 'higher-potentials', who represent the top
213 third, and a group of 'lower-potentials', who represent the bottom third. This enabled a
214 distinct comparison between the higher- and lower-potentials within each age group, with the
215 middle third discarded from the study ($n = 34$). For the purpose of this age-specific research,
216 the higher- and lower-potentials from the under-9 to under-11 were grouped together within
217 the FDP ($n = 26$), and the higher- and lower-potentials from the under-12 to under-16 were
218 grouped together within the YDP ($n = 38$). The results from the technical tests and match
219 analysis statistics were subsequently compared between the higher- and lower-potentials
220 throughout the FDP and YDP to observe any differences.

221 *Data analysis*

222 All data are expressed as mean \pm standard deviation. As a consequence of the potential
223 differing results between chronological age groups, such as older players generally
224 anticipated to record superior technical capabilities, data have been standardised using z-
225 scores within respective chronological age groups to allow comparisons between players
226 within both the FDP and YDP. Initial analysis investigated group differences between higher-
227 and lower-potentials using a MANOVA inclusive of all independent variables. Further post-
228 hoc analysis used an independent samples *t*-test to compare the higher- and lower-potentials'
229 mean scores of technical tests and match analysis statistics within the both FDP and YDP. A
230 binary logistic regression of the technical tests was also used to model higher- and lower-
231 potential status within the FDP and YDP, comprising of univariate and multivariate analyses
232 from the technical tests and match analysis statistics. Differences were considered significant
233 if $P < 0.05$. All analyses were conducted using IBM SPSS Version 23.

234 **Results**

235 The initial analysis using a MANOVA inclusive of all dependent variables revealed a
236 significant difference between groups of higher- and lower-potentials within the FDP
237 ($F(12,13) = 6.069, P = 0.001$; Wilk's $\Lambda = 0.151$, partial $\eta^2 = 0.849$) and YDP ($F(12,25) =$
238 $4.642, P = 0.001$; Wilk's $\Lambda = 0.310$, partial $\eta^2 = 0.690$).

239 *Technical tests*

240 Within the FDP, a significant difference was observed between the higher- and lower-
241 potentials for the lob pass test, with higher-potentials demonstrating a greater mean score (P
242 < 0.001). Within the YDP, significant differences were observed between higher- and lower-
243 potentials in the ball juggling test ($P = 0.012$), the slalom dribble test ($P = 0.003$), the
244 shooting accuracy test ($P = 0.005$), and the lob pass test ($P = 0.002$), with higher-potentials
245 demonstrating superior scores. The descriptive statistics of z -scores, t -tests, and non-
246 standardised mean results for all technical tests are displayed in Table 3.

247 *****Table 3 near here*****

248 The binary logistic regression of univariate factors from the technical tests within the FDP
249 showed a significant association between the lob pass test and higher-potentials, returning a
250 Cox and Snell R^2 of 0.542. Within the YDP, univariate regressions of the ball juggle test,
251 slalom dribble test, shooting accuracy test, and lob pass test showed significant associations
252 with higher-potentials, with Cox and Snell R^2 of 0.162, 0.214, 0.200, and 0.232 respectively.
253 The univariate logistic regressions of z -scores for technical tests are displayed in Table 4.

254 *****Table 4 near here*****

255 Further multivariate regression analysis was conducted to examine the relationship between
256 the higher-potentials and the series of technical tests. Correlation analysis showed low

257 collinearity between the technical tests, with the exception of the lob pass in the FDP, which
258 had a Pearson correlation coefficient of -0.604 for the ball juggle test ($P = 0.029$) and -0.605
259 for the slalom dribble test ($P = 0.029$). Thus, the lob pass test was excluded from the
260 multivariate regression for the FDP (Dormann et al., 2012). Results showed no significant
261 association for technical tests with **higher-potentials** ($\chi^2(3) = 6.010, P = 0.111$). The
262 explanatory power of the multivariate model did not improve upon the univariate models, and
263 only accounts for 20.6% of variance. The multivariate logistic regression within the YDP
264 showed a significant association between the technical tests and **higher-potentials** ($\chi^2(4) =$
265 $19.403, P = 0.001$), improving the explanatory power from univariate analysis to account for
266 40% of variance. The multivariate logistic regression models for the z -score of technical tests
267 are displayed in Table 5.

268 *****Table 5 near here*****

269 *Skill behaviours*

270 Within the FDP, there was a significant difference between **higher-** and **lower-potentials** for
271 reliability in possession ($P = 0.009$), pass completion ($P < 0.001$), and average touches ($P =$
272 0.030). Within the YDP, there was a significant difference between **higher-** and **lower-**
273 **potentials** for reliability in possession percentage ($P = 0.027$), dribble completion percentage
274 ($P = 0.001$), and average total touches ($P < 0.001$). **The descriptive statistics of z -scores, t -**
275 **tests, and non-standardised mean results for all skill behaviours are displayed in Table 3.**

276 The binary logistic regression of univariate factors from the skill behaviours within
277 the FDP showed significant associations between reliability in possession percentage, pass
278 completion percentage, and average total touches with **higher-potentials**, returning Cox and
279 Snell R^2 of 0.246, 0.405, and 0.206 respectively. Within the YDP, the univariate regressions
280 of dribble completion percentage and average total touches showed significant associations

281 with higher-potentials, returning Cox and Snell R^2 of 0.274 and 0.409, respectively. The
282 univariate logistic regression of z-scores for skill behaviours are displayed in Table 6.

283 ****Table 6 near here****

284 Further multivariate regression analysis was conducted to examine the relationship between
285 the higher-potentials and the series of skill behaviours within the FDP. Correlation analysis
286 showed some collinearity between the skill behaviours, thus those with a significant Pearson
287 correlation coefficient of greater than 0.5 with one or more variables were excluded from the
288 model. As a result, only reliability in possession percentage, average blocks, dribble
289 completion percentage, and average total touches were included in the model. Multivariate
290 logistic regression showed a significant association of technical tests with higher-potentials
291 ($\chi^2(4) = 12.475, P = 0.014$). The explanatory power of the skill behaviours multivariate
292 model improved upon the all univariate models, with the exception of pass completion
293 percentage, and accounts for 38.1% of variance. The multivariate logistic regression model
294 for the z-scores of skill behaviours are displayed in Table 7. Relationships between the
295 individual skill behaviours within the YDP showed high collinearity, thus multivariate
296 regression analysis was not conducted due to bias introduced upon variable selection and to
297 keep variables independent of one another (Myers, 1990).

298 ****Table 7 near here****

299 Discussion

300 This observational case study within a professional football academy presented the
301 opportunity to recognise technical factors that are associated with greater perceived
302 development from an age-specific perspective. Key findings in the FDP identified higher-
303 potentials had significantly greater lob pass ability, alongside reliability in possession

304 percentage, pass completion percentage, and average total touches, compared to lower-
305 potentials. Within the YDP, higher-potentials had significantly greater lob pass, slalom
306 dribble, shooting accuracy, and ball juggling abilities, alongside reliability in possession
307 percentage, dribble completion percentage, and average total touches, compared to lower-
308 potentials.

309 With regards to the technical testing within the FDP, the lob pass characterised the
310 single technical test that distinguished the groups, accounting for 54% of variance in the
311 univariate regression model. Perhaps due to the physical capabilities required for striking the
312 ball a relatively long distance for FDP players, a combination of technical proficiency and
313 physical abilities may partially explain why higher-potentials achieved greater scores on the
314 lob pass (Nicolai, Cattuzzo, Henrique, & Stodden, 2016). When compared to the FDP, the
315 technical tests were collectively a better discriminator of the groups in the YDP; although
316 they only accounted for a moderate variance in the model for all variables, multivariate
317 analysis did account for 21% of the between group variance. Consequently, this highlights
318 technical competency as an influential factor when discriminating talented football players
319 within this developmental context.

320 These results are comparable to those of Vaeyens et al. (2006) who, with the
321 exception of under-12's, studied the same age groups that are analysed in the YDP in this
322 current study. Since this current study incorporated the same battery of tests as Vaeyens and
323 colleagues (2006), it provides further evidence of the discriminative function of these
324 particular technical tests in youth football players. Similarly, the current findings also support
325 those of Keller et al. (2016), who found that their passing tests, shooting accuracy test, and
326 dribble speed test distinguished better performance in their YDP groups. Together, these
327 studies offer a range of literature to suggest that technical tests may prove useful in
328 identifying and developing youth football players within the YDP. Further, with technical

329 ability important for the future career progression of youth football players (Barnes et al.,
330 2014), these tests offer the option for academies to highlight specific technical abilities as key
331 developmental indicators as part of their talent development process (Hoare & Warr, 2000;
332 Rosch, Hodgson, Peterson, Graf-Baumann, Junge, Chomiak, & Dvorak, 2000; Vanderfold,
333 Meyers, Skelly, Stewart, & Hamilton, 2004).

334 The age-specific discrepancies in the technical testing results are likely explained by
335 the rate at which technical ability improves with age amongst youth football players. For
336 instance, it has been suggested that the greatest improvements are shown to occur in pre-
337 pubertal years, after which technical skills are gradually developed towards adulthood
338 (Huijgen et al., 2010; Valente-dos-Santos et al., 2014; 2012; Wilson et al., 2016).
339 Furthermore, with a greater discriminatory functions evident within the YDP, the results also
340 partially support previous studies that have revealed growth and maturation status to be
341 associated with technical skill development (Malina et al., 2005; 2007; Valente-dos-Santos et
342 al., 2014). In the context of this current study, as an example, greater slalom dribble speed
343 may be partially a result of enhanced growth and maturation status that subsequently allows
344 more mature players to run faster with the ball (see Kelly & Williams, 2020). Therefore, it
345 may be important to highlight the discriminating technical factors among youth football
346 players that may vary with the timing and tempo of growth, consequently adding to the
347 dynamic talent development process (Kelly, Wilson, & Williams 2018).

348 The outcome of a player's reliability in possession is based on the combined
349 execution of a technical action (i.e., pass or dribble) and a tactical decision (i.e., anticipation
350 and awareness). The ability to maintain possession, particularly under pressure, is an
351 important skill in senior professional football (Gomez et al., 2018; Liu et al., 2016; Yang et
352 al., 2018). Thus, the current findings show that being able to maintain the ball effectively
353 (reliability in possession) is also important from a talent development perspective. Likewise,

354 it is proposed that pass completion is a combination of technical execution and cognitive
355 function. For instance, a player requires the ability to execute a pass technically well (i.e.,
356 with the correct weight and angle), but also to select the correct option (i.e., decision-making
357 and positioning). Rampinini and colleagues (2009) also demonstrated players from more
358 successful senior professional football teams generally possess a higher pass completion rate
359 compared to their less successful counterparts during competitive match-play. As a result, the
360 feature of possessing superior pass completion appears to be a significant characteristic for
361 early talent development.

362 Within both the FDP and YDP, **higher-potentials** also possessed a greater number of
363 touches on the ball compared to their **lower-potential** counterparts. This may be due to a self-
364 fulfilling prophecy, whereby the better players play in positions where they receive the ball
365 more often; and as such, gain more technical development opportunities during competitive
366 match-play compared to **lower-potentials**. This finding supports the application of Fenoglio
367 (2004a; 2004b) and Thomas and Wilson's (2015) research, which reveals reducing player
368 numbers during competitive match-play in youth sport during childhood increases technical
369 outcomes. If players get more touches on the ball to try new skill behaviours, this provides
370 more opportunities to develop technical capabilities (Katis & Kellis, 2009). Therefore, it is
371 recommended that low player numbers (such as 4 vs. 4 to 6 vs. 6 formats) are utilised within
372 the FDP, to increase individual touches on the ball and subsequently technical development
373 opportunities for all.

374 Interestingly, average tackles completed, average blocks achieved, and average loose
375 balls retrieved revealed no significant difference when comparing **higher-** and **lower-**
376 **potentials** in either the FDP or YDP. These 'out of possession' factors do not require control
377 of the ball and may therefore be easier to execute or more cognitive in nature. These findings
378 concur with Gomez and colleagues (2018), who highlighted superior 'in possession' factors

379 (ball possession and ending actions) and a lower 'out of possession' factors (individual
380 challenges) were associated with a higher league ranking. Consequently, observing skill
381 behaviours in possession may provide greater reliability from a talent development
382 perspective in youth football; although position-specific requirements may also need to be
383 considered.

384 *Limitations and future directions*

385 It is important to recognise that observational case studies contain methodological limitations,
386 such as limited access to participants, who are often difficult to recruit (particularly for
387 technical observation), and low external validity (Morgan, Pullon, Macdonald, McKinlay, &
388 Gray, 2017). To address the former limitation, it is important to recognise the researchers
389 obtained the accessibility to a large enough group of professional football academy players.
390 In addition, statistical analyses procedures were applied to reduce potential bias introduced to
391 both the data and models. Thus, this research does not only provide a novel illustration of
392 technical attributes within the talent development process, it also offers a useful
393 benchmarking tool for other football academies. For the latter limitation of external validity,
394 the cultural and social dynamics in the English football talent pathways must be considered,
395 since the technical abilities of these Category 3 players may be different to youth football
396 players in other regions, countries, or categories. Thus, comparisons based on playing level,
397 location, and category status must be made with care.

398 Regarding the limitations of the measures applied, it may be argued technical tests
399 disregard the technical ability from an ecological perspective. For instance, these tests ignore
400 the physical and mental implications during the latter stages of a competitive game (Reilly,
401 1997; Russell, Benton, & Kingsley, 2010), whilst also applying an environment that differs to
402 the one that is applied to actual match-play. Nevertheless, the incorporation of a battery of

403 tests alongside match analysis statistics provides a dynamic context, thus supporting a greater
404 determination of technical ability. Furthermore, the variable number of matches that were
405 available for match analysis statistics should also be noted; although it is understood that this
406 **is representative** of the dynamic nature of academy development. Additionally, these
407 statistics may also provide as useful benchmarking figures for clubs, coaches, and players
408 alike.

409 Future research may offer further investigation into the technical ability and skill
410 behaviour of youth football players, while applying characteristics from other significant
411 talent development variables (i.e., physical performance and psychological characteristics).
412 Consequently, this will offer the novelty of a multidimensional approach required for
413 contemporary talent development literature, while gaining a complete impression of the talent
414 development process (Collins, MacNamara, & Cruickshank, 2018). Furthermore, collecting
415 these variables from a longitudinal perspective will also offer suggestions regarding what
416 technical abilities and skill behaviours are associated with individuals who achieve
417 professional status and those who do not. Additionally, the coaching process surrounding
418 how these technical qualities are developed, from an age-specific context, also requires
419 further investigation.

420 **Conclusion**

421 These results provide important insights on understanding the age-specific technical
422 abilities that are associated with coach development rankings. First, the results suggest
423 football-specific technical tests may provide discriminative tools to support the talent
424 development process from an age-specific perspective. Second, ‘in possession’ skill
425 behaviours, alongside gaining more touches on the ball during competitive match-play, may
426 support greater perceived development. Third, these descriptive variables offer a useful
427 benchmarking tool for practitioners to consider for developmental purposes. **In summary, a**

428 combination of technical tests and match analysis statistics provides a broader objective
429 context, thus offering a greater determination of technical ability. Thus, through coaches and
430 practitioners supporting these technical developmental outcomes during childhood and
431 adolescence, youth football players may possess greater developmental opportunities towards
432 senior expertise.

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438 **Disclosure statement**

439 The authors declare that they have no conflict of interest.

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