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Methods to predict the timing and status of biological maturation in male adolescent soccer players: A narrative systematic review

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Manuscript

- 1 Main title: Methods to predict the timing and status of
- 2 biological maturation in male adolescent soccer players: A
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30 Abstract

31 The aim of this review was to summarise the methods used to predict and assess maturity status 32 and timing in adolescent, male, academy soccer players. A systematic search was conducted on 33 PubMed, Scopus, Web of Science, CINAHL, Medline and SPORTDiscus. Only experimental studies 34 including male, academy players aged U9-U18 years registered with a professional soccer club were 35 included. The methodological quality of the included studies was assessed using guidelines from the 36 Framework of Potential Biases. Fifteen studies fulfilled our inclusion criteria. Studies were mainly 37 conducted in European countries (n = 12). In total, 4,707 players were recruited across all 15 studies, 38 with an age range of 8-18 years. Five studies were longitudinal, two studies were mixed-method 39 designs and eight studies were cross-sectional. Due to high heterogeneity within the studies, a meta-40 analysis was not performed. Our findings provided no equivalent estimations of adult height, skeletal 41 age, or age at PHV. Discrepancies were evident between actual and predicted adult height and age at 42 PHV. The Bayley-Pinneau (1952), Tanner-Whitehouse 2 (1983) and Khamis-Roche (1994) methods 43 produced estimates of adult height within 1cm of actual adult height. For age at PHV, both Moore 44 (2015) equations produced the closest estimates to actual age at PHV, and the Fransen (2018) equation 45 correlated highly with actual age at PHV (>90%), even when the period between chronological age 46 and age at PHV was large. Medical imaging techniques (e.g., Magnetic Resonance Imaging, X-Ray, 47 Dual energy X-ray Absorptiometry) demonstrated high intra/inter-rater reliability (ICC = 0.83-0.98) 48 for skeletal maturity assessments. The poor concordance between invasive and non-invasive methods, 49 is a warning to practitioners to not use these methods interchangeably for assessing maturational status 50 and timing in academy soccer players. Further research with improved study designs is required to 51 validate these results and improve our understanding of these methods when applied in this target 52 population.

54 Introduction

55 Professional football (soccer) clubs across the globe have academy infrastructure dedicated to 56 the identification and development of talented young players (1). A professional soccer academy is a 57 performance environment, where *potentially* talented youth players train, prepare and compete to attain 58 the soccer-specific skills (e.g., technical; physical; tactical; psychological) to progress to the first (i.e., 59 senior) team and succeed in competition (2-4). Most youth academies operate extensive data capture 60 systems where individual player's information is captured (e.g., training and match load, 61 anthropometric and injury data) on a daily basis (5,6). For example, in England and Wales the Player 62 Management Application (PMA) is an online system used by academy science and medicine 63 departments to record a range of information (e.g., training volume; training intensity; fitness testing 64 results), which is then provided to the Premier League. Departments also pay particular attention to 65 injury (i.e., incidence; type; location; burden) and anthropometric data that can be used to estimate 66 each player's growth and maturation status (7). This can subsequently facilitate the optimisation and 67 implementation of appropriate injury prevention plans that are specific to a player's stage of maturation 68 (8). One example of how clubs integrate maturation and training load data to develop young players is 69 presented by McBurnie et al., (8). This case study demonstrates how clubs can regularly use training 70 load data gathered via geographical positioning systems (GPS) as a measure of external load, in 71 combination with regular anthropometric and injury history datam to generate a 'risk profile' for each 72 player. This is used to create a 'decision-tree' process regarding the management of each player from 73 a training load and injury risk prevention perspective (8).

During the adolescent growth spurt, changes in lower limb length and limb mass continue until peak height velocity (PHV) growth rate is achieved (take-off), at which point a deceleration and eventual cessation in height occur (9). Male, youth soccer players typically undergo a phase of accelerated growth (i.e., 8-10cm) between 11-15 years of age, reaching PHV ~13-14 years of age (10). The growth spurt coupled with maturity-associated variations are among some of the injury risk factors
for the developing male athlete (10).

80 Injury incidence in youth academy soccer players competing in U16-U19 years are reported to 81 be as high as four injuries per 1000 hours of training and match exposure, however injury risk and 82 incidence is known to increase around reported mean ages (i.e., 13-14 years) at PHV (11,10). Evidence 83 of higher injury incidence, particularly microtraumatic damage to tissue (e.g., bone; muscle; tendon) 84 during the period of PHV and increased general injury burden (15 vs 7 days) compared to pre-PHV 85 (when the rate of growth in stature is at its slowest) (12) is reported in European youth soccer players 86 (13). Similarly, in a professional, male, Italian soccer academy, the highest injury incidence across 87 academy age groups was reported in the U13 years, followed by the U15 and U14 years, corresponding 88 to the period of PHV, yet caution is warranted when interpreting these findings as they are derived 89 from a single club in Europe (14). Professional soccer clubs worldwide are becoming increasingly 90 invested in monitoring injury rates and growth patterns of their players, particularly around PHV, due 91 to the associated increases in injury risk, incidence, and severity that predispose players during this 92 period (15,16). Previous research has also highlighted the importance of youth soccer players remining 93 'injury-free' during their academy years, due to the negative implications of deselection and loss of 94 athletic identity (16). It is suggested therefore that frequent monitoring of injury and maturation 95 patterns, particularly around PHV, will aid the design and implementation of targeted injury prevention 96 and training load strategies (15,8), thus supporting early maturing (skeletal age is older than 97 chronological age by at least one year), average (skeletal age is \pm one year of chronological age) and 98 late maturing players (skeletal age is younger than chronological age by at least one year) through the 99 maturation process (17).

100 Evidence suggests early maturing players have the highest overall injury risk (18), with growth-101 related injuries (e.g., apophysitis) generally occurring pre and circa-PHV, whilst muscular and 102 knee/ankle articular injuries occur post-PHV (11). A combination of high training volume and 103 relatively slow adaptation of muscles, tendons, and apophyses to changes in extremity length, mass, 104 and moments of inertia caused by PHV are possible explanations for these findings (11). Whereas 105 earlier maturing players have heightened injury severity pre-PHV, and later maturing players often 106 suffer more burdensome injuries during adulthood (19). This is reportedly due to the musculoskeletal 107 and neuromuscular alterations induced by the individual variation in the timing of PHV amongst 108 players within the same chronological age group (19). To optimise injury epidemiology associated with 109 growth and maturation within earlier, average, and later maturing players, performance staff employ 110 methods to measure maturity status (the stage of maturation at the time of observation, i.e., pre-, circa-111 , or post-PHV) and timing (the age at which PHV occurs i.e., early, average, or late) (11).

112 The 'gold standard' indicator for assessing biological maturation includes assessments of skeletal 113 age (20). However, this method is invasive and involves radiation exposure due to medical scanning 114 to assess the skeletal maturity of the hand/wrist (e.g., X-Ray, Dual energy X-ray Absorptiometry 115 (DXA); Magnetic Resonance Imaging (MRI) (21,22) and requires clinical expertise when applied in 116 youth environments. Furthermore, earlier work has demonstrated poor concordance for predictions of 117 skeletal age using the Tanner-Whitehouse 2 and 3 (23,24) methods for the same wrist/hand scan in 118 academy soccer players aged 11-17 years (25) whilst the Fels (26) method can reduce the estimation 119 of skeletal age (27). The systematic lowering of skeletal age associated with the Tanner-Whitehouse 3 120 (2001) vs. 2 method (1983) is reportedly as high as 1.06 years (25). These variations can be attributed 121 to the variance in reference samples from which the different methods were derived (28). For example, 122 the Greulich-Pyle (29) and Fels (1977) methods were developed in pediatric populations from high 123 socioeconomic areas in the United States (US), while the Tanner-Whitehouse 2 (1983) method was

124 developed using children in the United Kingdom (UK) (25). The Tanner-Whitehouse 3 (2001) method, 125 an extended version of the Tanner-Whitehouse 2 (1983) method, included children from the UK as 126 well as adolescents from other well established soccer nations such as Japan, Belgium, Argentina, and 127 Italy (25,30). A further consideration is that these methods differ in the types of hand/wrist bones used 128 for analysis. For instance, the Fels (1977) method uses the radius, ulna, short bones, and carpals to 129 predict skeletal age whereas the Tanner-Whitehouse 3 (2001) method uses the radius, ulna, metacarpals 130 and phalanges to provide a skeletal age assessment (28). Further differences are observed for the 131 statistical weighting and set of criteria for maturity indicators of bones within the hand/wrist between 132 the Tanner-Whitehouse 2 (1983) and 3 (2001) methods to calculate skeletal age (25). Given these 133 apparent discrepancies, non-invasive methods have been proposed as suitable alternatives for assessing 134 maturational status and timing of PHV (9).

135 Two non-invasive methods for estimating maturity status and timing that are typically utilised in 136 soccer academies are the percentage of estimated adult height and maturity offset methods (7). The 137 percentage of predicted adult height method provides an estimation of adult height and an estimate of 138 the current height of a player relative to their predicted adult height (31). The maturity offset method 139 provides an estimate of time (years) away from PHV and subsequently an estimate of age at PHV (32). 140 For predicting age at PHV, other alternative equations are available for practitioners working with 141 youth academy players. One equation proposed recently by Fransen et al. (33) has attempted to 142 improve the precision of estimates for age at PHV by using a maturity ratio (chronological age / age at 143 PHV) rather than a maturity offset (chronological age – age at PHV), which is considered a more 144 appropriate representation of the non-linear relationship between anthropometric variables and 145 maturity offset (34). Likewise, the Moore et al. (34) equations provide practitioners with other methods 146 for predicting age at PHV and is considered a modification of the original Mirwald (2002) equation, 147 however, the original regression equation used by Mirwald (2002) has been adjusted to create the Moore (2015) equations. Similarly, for predicting adult height, the Bayley-Pinneau (1952) method is widely used, as it aims to predict adult height from skeletal age and is based on the high correlation between skeletal ages attained from hand/wrist scans and the proportion of adult stature attained by adolescents at the time of the scan (35).

152 According to Towlson et al., (10), the Mirwald (2002) and Khamis-Roche (1994) are the most 153 popular methods for predicting age at PHV and adult height respectively, since they are facilitated by 154 organising bodies (e.g., the Premier League) and can be integrated into online PMAs. However, some 155 criticisms of these methods are that they require more than two years of longitudinal growth data (e.g., 156 total body height, annual growth velocity changes) and existing studies typically do not to track growth 157 rate data for this amount of time (36). Further limitations of these equation-based methods (e.g., 158 maturity offset) is the tendency to overestimate the timing of PHV in earlier maturing players and 159 underestimate the timing of PHV in later maturing players, although the accuracy of these methods 160 improves if applied promptly with data inputted at regular intervals (36). Parr et al., (9) has also 161 reported that the Khamis-Roche (1994) method has a greater prediction power compared to the 162 Mirwald (2002) equation for predicting the timing of PHV, despite being primarily used to predict 163 adult height. A limitation of the Khamis-Roche (1994) method is that it requires variables such as 164 decimal age, standing height (cm), body mass (kg), and an accurate stature (cm) of both biological 165 parents to provide an estimate of adult height. However, if parental height is unavailable, national 166 averages of stature for men and women are used in the equation from qualified anthropometric 167 assessments, which can potentially inflate the standard error (10).

168 Reliability concerns with these equation-based methods are associated with inconsistent research 169 designs, study quality, and recruited populations (36). Consequently, there is poor agreement between 170 invasive and non-invasive prediction methods of maturity status and timing (20). Nonetheless, 171 equation-based predictors remain the most practical option for practitioners working within professional soccer academies (7). Despite a wealth of individual empirical studies, there is currently limited review studies that synthesise the existing literature and establish the reliability of both invasive and non-invasive methods for assessing maturational status and timing in youth, academy soccer players.

176 To our knowledge, only one previous systematic review exists that examines the accuracy and 177 reliability of existing methods for predicting PHV in adolescents (36). This review reported that 178 radiograph-based methods appear to have the most value in predicting actual PHV and that the age of 179 PHV can be accurately predicted in males as young as 11 years. The review by Mills *et al.*, (36) was 180 conducted in healthy male and female adolescents from the general population and therefore, it is 181 unknown how well these methods perform in youth, academy soccer players. Further findings from 182 this review (36) suggest that equation-based methods offer some promise as surrogate measures of 183 maturity status, though the reliability of these methods is unknown, and the current state of the literature 184 makes such an investigation into the reliability of this particular method challenging, given the high 185 levels of heterogeneity within the datasets. Therefore, the aim of the present review is to narratively 186 summarise the reliability of method(s), both invasive and non-invasive, for assessing maturity status 187 and timing in adolescent, male, academy soccer players.

188 Materials and Methods

The systematic review was conducted in accordance with the Preferred Reporting of Items for Systematic Reviews and Meta-Analyses (PRISMA) guidelines (37). After several scoping searches, a comprehensive bibliographic search was conducted between June-September 2022 and re-run in May 2023 on the following academic databases: PubMed, Scopus, Web of Science, Cumulative Index to Nursing and Allied Health Literature (CINAHL), Medline, and SPORTDiscus. Search filters were limited to published literature in the English language, and articles that had full-text access. No filters 195 regarding publication year were included and 'grey literature' (e.g., student dissertations or theses, and 196 conference proceedings) were excluded from the search criteria.

197 Search syntax

The specific keywords and syntax terms for each database were agreed upon between members of the research team and a university librarian; a database specialist employed to support the review process. The following syntax were entered into each of the above databases: Method* OR procedure* AND Estimat* OR predict* OR calculat* OR measur* AND "Peak height velocity" OR "PHV" OR matur* OR "biological maturation" OR "growth spurt*" OR "maturity offset*" OR "skeletal age" OR "skeletal maturity" AND Youth* OR adolescent* OR teenage* AND Football OR soccer AND player*.

205 Inclusion/exclusion criteria (PICOSS)

206 The review was planned around the Population, Intervention, Comparator, Outcome variables, 207 Study design, Setting, (PICOSS) approach to capture appropriate quantitative studies. All members of 208 the research team participated in devising the inclusion criteria (**Table 1**) and exclusion criteria (**Table** 209 2) for candidate studies. The aim of the inclusion criteria was to capture as many relevant studies as 210 possible, that utilised either invasive (i.e., medical imaging or hand scans) or non-invasive (i.e., 211 predictive equations) methods to assess maturational status and timing of male, academy soccer players 212 from professional soccer clubs. The age range included U9-U18 players, in order to capture players 213 residing in different stages of maturation from across the academy system, as well as within individual 214 chronological age groups, with some players of the same chronological age group known to differ in 215 biological age by as much as 5-6 years (8).

216 217 218	Table 1. PICOSS study inclusion criteria.	
	Population	Male, academy soccer players aged U9-U18 years.
219		
220	Intervention/Comparator	Invasive/non-invasive methods used to predict or assess maturation status and timing.
221	Outcome variables	Maturity offset (years), age at PHV (years), skeletal age (years), maturity status, percentage of predicted adult height (%).
222	Study design	Longitudinal/cross-sectional, prospective/retrospective randomised control trials, cohort studies, case studies.
223	Study settings	Professional soccer club academies worldwide.
224		

Population	Female players.
	Amateur/non academy players.
	Adolescents from the general population.
Age	Academy players aged <u9 or="" years=""> U18 years.</u9>
Study characteristics	Non-English language published studies.
	Descriptive/anecdotal studies.
	Studies based on 'expert' opinion.
	Non-peer reviewed articles.
Outcome variables	Soccer-specific performance characteristics (i.e. passing, shooting, tackling).
	Physical performance characteristics (i.e. VO ₂ max, high-speed running distance, physical strength measures).

232 Screening process

Studies meeting the inclusion criteria were imported into a bibliographic management software system (i.e., EndNote) for stage one and stage two screening. Stage one screening consisted of title and abstract reviews which were completed by the lead author following the removal of duplicates. Two reviewers independently screened a random sample of 20 studies and inconsistencies were resolved by consensus. Stage two screening was conducted by the first author, whereby full-text papers were assessed against the eligibility criteria. Reasons for study exclusion included those that were irrelevant to study question, inappropriate study populations, outcome variables and study designs (**Fig 1**).

240 Risk of bias assessment

241 According to Mlinaric et al., (38) the threat of publication bias in academic research is 242 increasing, with a preference of current medical and scientific literature to publish seemingly more 243 positive study results. Furthermore, this bias could be because more 'successful' and 'productive' 244 studies are more interesting to read and are therefore perceived as being more valuable for publishers, 245 editors, and their audience. To offset this threat in the present review, a thorough and objective-based 246 inclusion criteria was provided, which used a variety of databases to capture as many relevant studies 247 as possible, all data was considered for analysis within each study and any missing data was requested 248 by the researchers. All included studies were quality assessed against a recognised objective framework 249 (A Measurement tool for Assessment of Multiple Systematic Reviews) (39) to assess for risk of bias. 250 Studies were independently assessed for risk of bias by two members of the research team with any 251 disagreements being resolved via a discussion and no arbitrary third assessor was required.

252

254 **Quality appraisal**

255 The quality of each study was assessed using the Framework of Potential Biases (40), which 256 has six criteria to assess for study bias, followed by a total quality score. Quality criteria is based on: 257 (1) study population; (2) study attrition; (3); use of valid and reliable instruments for predictors; (4) 258 having objectively measured outcome variables; (5) controlling for confounding factors (age, gender 259 etc.); and (6) using appropriate statistical analyses. If a criterion is fully satisfied, it receives a score of 260 two, if the criterion is partly satisfied a score of one is given, and if the criterion is not satisfied it 261 receives a score of zero. The score for each individual criterion is then added up to provide a total 262 quality score for each study. A low-quality study has a score ranging from 0-4 points, a medium-quality 263 study has a score ranging from 5-8 points and a *high*-quality study has a score ranging from 9-12 points.

264 **Data extraction**

Extracted data for individual study outcome variables were included but were not limited to: Pearson and Spearman-rank correlational values (R²), kappa and intra or inter-class coefficient values, mean differences between observed and predicted maturational status and timing variables (years/cm), level of concordance between invasive and non-invasive maturity estimates and between methods for maturity classification (%).

270 Data analysis

Due to assumptions of homogeneity not being satisfied and a high amount of heterogeneity within the data, a full meta-analysis was not performed (41). A high level of heterogeneity within the data was caused by variance within individual study characteristics (cross sectional vs. longitudinal designs), types of study data (dichotomous vs. continuous), and differences in outcome measures from individual studies (invasive vs. non-invasive outcome variables). Moreover, the vast differences in the number of included participants between studies, individual characteristics of these players not
being available, and lack of reported randomisation process for included study participants also made
a meta-analysis inappropriate (40). Due to these issues, a narrative review was preferred for the
study.

280 A measurement tool for assessment of multiple systematic reviews (AMSTAR) 2

Previous work by Shea *et al.*, (39) have commented that systematic reviews are subject to a range of biases due to the inclusion of non-randomised intervention studies, similar to the present review. A measurement tool for assessment of multiple systematic reviews (AMSTAR) model was developed to evaluate systematic reviews that utilised randomised studies (42) however it has since been updated (AMSTAR 2) to evaluate systematic reviews that have utilised non-randomised studies. The revised AMSTAR 2 tool has 16 items in total, consisting of binary '*yes*' or '*no*' questions relating to the quality of the systematic review, however it is not intended to generate an overall score.

288 Narrative synthesis of findings

In total, 15 publications fulfilled our inclusion criteria and were included in the final analysis. A group summary regarding participant recruitment, study design, outcome variables and country of origin of the included studies can be found in **Table 3**. **Table 3.** Narrative group summary of included studies.

293					
	Total number of participants	n = 4,707 players across 15 studies.			
294					
295	Age range	8 – 18 years.			
296	Country of origin	Non-European (i.e. Qatar, Brazil, Thailand, Japan, Mexico, Egypt) n = 3.			
297		European (i.e. Germany, UK, Belgium, Switzerland) $n = 12$.			
	Study design	Cross-sectional $(n = 8)$.			
298		Longitudinal ranging from five playing seasons up to 14 years $(n = 5)$.			
299		Mixed-method $(n = 2)$.			
		Entirely invasive $(n = 1)$.			
300		Entirely non-invasive $(n = 5)$.			
301		Invasive AND non-invasive combination $(n = 9)$.			
	Outcome variables	Skeletal age ($n = 10$).			
302		Predicted adult height/ percentage of predicted adult height ($n = 6$).			
303		Maturity ratio $(n = 3)$.			
		Maturity offset $(n = 8)$.			

304 ***Insert **Fig 1** about here***

305 **Quality scores**

306 An appraisal of study quality using the framework proposed by Rashid *et al.*, (40), revealed

- 307 that nine studies were medium quality (5-8 points), and six studies were high quality (9-12 points),
- 308 with no studies deemed low quality (0-4 points). See **Table 4**.

Table 4. Study quality assessment of included studies, in accordance with the Framework of Potential Biases (40).

Study	Representative sample of relevant population	Study attrition (loss to follow-up and response rate)	Valid and reliable instruments for predictors	Objectively measured outcomes	Controlled for age	Appropriate statistical analyses	Quality score
Abdelbary <i>et</i> <i>al.</i> , (43)	1	0	2	1	2	1	7 Moderate
Fransen <i>et al.</i> , (33)	1	0	0	2	2	2	7 Moderate
Leyhr <i>et al.</i> , (44)	1	0	1	2	2	2	8 Moderate
Lolli <i>et al.</i> , (45)	2	0	2	2	2	1	9 High
Malina <i>et al.</i> , (27)	2	0	1	1	2	2	8 Moderate

Malina <i>et al.</i> , (17)	1	0	1	1	2	2	7 Moderate
Malina <i>et al.</i> , (20)	1	0	1	1	2	1	6 Moderate
Malina <i>et al.</i> , (25)	2	0	1	1	2	2	8 Moderate
Malina <i>et al.</i> , (46)	2	2	1	2	2	2	11 High
Parr <i>et al.</i> , (9)	2	0	1	2	2	2	9 High
Romann and Fuchslocher (47)	2	2	1	1	2	1	9 High
Romann and Fuchslocher (22)	2	2	1	0	2	2	9 High
Ruf <i>et al.</i> , (48)	2	0	1	2	2	2	9 High
Salter <i>et al.</i> , (7)	2	0	0	2	1	1	6 Moderate

Teunissen et	2	0	0	2	2	2	8
al., (49)							
							Moderate

2 = criterion satisfied, 1 = criterion partly satisfied, 0 = criterion is not satisfied/ cannot be determined.

Maximum quality score = 12.

0-4 points = low quality, 5-8 points = medium quality, 9-12 points = high quality.

311 AMSTAR 2

In accordance with the framework and suggestions by Shea *et al.*, (39), the current review can be considered of moderate quality. The AMSTAR 2 assessment revealed that the current review contains more than one non-critical weakness (item 10 and 16, see **Table 5**) but no critical flaws. Therefore, the current review provides an accurate summary of the results from the included studies. **Table 5:** AMSTAR 2 systematic review assessment. Responses in **bold** are considered the key domains as suggested by

317 Shea *et al.*, (39).

Criterion	Response
1. Did the research question and inclusion criteria for the review include the components of PICO?	Yes
2. Did the report of the review contain an explicit statement that the review methods were established prior to the conduct of the review and did the report justify any significant deviations from the protocol?	Yes
3. Did the review authors explain their selection of the study designs for inclusion in the review?	Yes
4. Did the review authors use a comprehensive literature search strategy?	Yes
5. Did the review authors perform study selection in duplicate?	Yes
6. Did the review authors perform data extraction in duplicate?	Yes
7. Did the review authors provide a list of excluded studies and justify the exclusions?	Yes
8. Did the review authors describe the included studies in adequate detail?	Yes

9. Did the review authors use a satisfactory technique for assessing the risk of bias (RoB) in individual studies that were included in the review?	Yes
10. Did the review authors report on the sources of funding for the studies included in the review?	No
11. If meta-analysis was performed, did the review authors use appropriate methods for statistical combination of results?	N/A
12. If meta-analysis was performed, did the review authors assess the potential impact of RoB in individual studies on the results of the meta-analysis or other evidence synthesis?	N/A
13. Did the review authors account for RoB in primary studies when interpreting/discussing the results of the review?	Yes
14. Did the review authors provide a satisfactory explanation for, and discussion of, any heterogeneity observed in the results of the review?	Yes
15. If they performed quantitative synthesis did the review authors carry out an adequate investigation of publication bias (small study bias) and discuss its likely impact on the results of the review?	Yes
16. Did the review authors report any potential sources of conflict of interest, including any funding they received for conducting the review?	No

319 **Discussion**

320 Performance of invasive and non-invasive maturity indicators

321 Three of the included studies investigated adult height, with two of these studies (49,10) comparing 322 predicted adult height methods (e.g., Khamis-Roche, 1994; Bayley-Pinneau, 1952) to actual (observed) 323 adult height and one study (48) compared two predictive methods for adult height (Table 6). The 324 findings revealed that none of the methods produced the same estimation of adult height, with 325 discrepancies (-0.45cm to -2.1cm) evident between predicted and observed values of adult height. One 326 plausible explanation for these discrepancies could be attributed to the methods employed during the 327 anthropometric data collection within the studies. For example, it is unclear whether anthropometric 328 data was captured under International Society for the Advancement of Kinanthropometry (ISAK) 329 guidance or by single or multiple measurer(s) (45,9). Such methodological considerations may elevate 330 the poor level of agreement amongst these predictive methods (20). This finding is of relevance and 331 importance to practitioners in academy soccer (e.g., sport scientists) given the application of predicted 332 adult height to categorise players into maturity specific groupings (i.e., 'bio-banding') (50). Thus, 333 erroneous predictions of player maturity status may incur the mis-categorisation of players into such 334 groupings and afford players with unfair playing environments (e.g., competing against players who 335 matured earlier and subsequently possess enhanced anthropometric characteristics, or vice-versa). 336 Careful consideration should be taken by practitioners attempting to assess maturational status in 337 academy soccer players.

The evidence in this review suggests the Bayley-Pinneau (1952), Tanner-Whitehouse 2 (1983), and Khamis-Roche (1994) predictive methods performed well against observed adult height and produced estimates within 1cm of actual adult height. However, Tanner-Whitehouse 3 (2001) 341 estimates were 1-2 cm short of observed adult height. For predictive estimates, large agreements and 342 small systematic errors were observed between the Tanner-Whitehouse 2 (1983) and Khamis-Roche (1994) methods, which demonstrates the high level of concordance between these predictive methods 343 344 (48). The Khamis-Roche (1994) method produced a slightly higher estimate of adult height compared 345 to the Tanner-Whitehouse 2 (1983) method, with a difference of around 0.73cm, which is considered 346 acceptable (48). One potential reason for the observed differences between methods could be the 347 different player nationalities. Two studies were conducted in Europe (England and Germany) whilst 348 the other study was conducted in Qatar. These different nationalities and ethnicities could play a key 349 part in growth variables, such as proportions of sitting height and leg length to stature ratio, which 350 are known to vary among ethnic/racial groups and thus could influence the difference between 351 observed and predicted adult height values derived from different methods (20). For example, 352 previous work from Lopez et al., (51) concluded that adolescent soccer players in Chile were smaller 353 and lighter than the general South American population for any given age within adolescence and 354 demonstrate lower growth rates compared to Brazilian and Spanish soccer players. Given that many 355 of these existing predictive methods were derived using White and Caucasian populations from 356 middle-class backgrounds, the appropriateness of using current predictive equations for adult height 357 in culturally and ethnically diverse environments (e.g., professional soccer academies) currently 358 remains unknown (7). Therefore, validation of existing equations or proposal of new equations in this 359 target population may be more appropriate for soccer practitioners to use to assess and estimate adult 360 height in youth players.

361 One study investigated the use of Tanner-Whitehouse 2 (1983) and Tanner-Whitehouse 3 (2001) 362 methods to predict skeletal age (25). The finding of this study was that Tanner-Whitehouse 3 (2001) 363 skeletal ages were on average 1.06 years younger than skeletal ages derived from the Tanner-364 Whitehouse 2 (1983) method across the U12 to U17 age groups. The difference between Tanner-

365 Whitehouse 2 (1983) and Tanner-Whitehouse 3 (2001) skeletal ages was greatest between the U12 to 366 U14 age groups, a significant period during maturation that is associated with rapid increases in 367 skeletal and somatic growth (52). Given the systematic lowering of skeletal ages associated with the 368 Tanner-Whitehouse 3 (2001) method vs. Tanner-Whitehouse 2 (1983) method, this could elevate the 369 risk of incorrect maturity classification of players (25), as well as having implications for bio-370 banding in soccer tournaments, leading to players potentially being incorrectly included or excluded 371 in tournaments with peers of a similar skeletal or chronological age (17). These observed differences 372 between Tanner-Whitehouse 2 (1983) and Tanner-Whitehouse 3 (2001) methods could be explained 373 by the reference samples used to derive the estimates of skeletal age associated with each method. 374 For example, the Tanner-Whitehouse 2 (1983) method was developed in children in the UK, unlike 375 the Tanner-Whitehouse 3 (2001) method, which used a more heterogenous sample of children from 376 Spain, Italy, Belgium, Argentina, and Japan (45). The differences between the populations used to 377 derive these skeletal age estimates could partly explain the variance. According to Malina & 378 Bouchard (53) skeletal maturation in Hispanic adolescents occurs later than similarly aged Black and 379 White adolescents. Furthermore, Asian adolescents appear to be, on average, shorter, lighter and are 380 likely to be more skeletally immature compared to similarly aged adolescents of European ancestry. 381 Given these differences in maturational growth patterns between adolescents of various ethnicities 382 used within the reference samples, it is unsurprising that the Tanner-Whitehouse 2 and 3 (1983; 383 2001) methods produce inequivalent estimates of skeletal age. Other possible explanations for these 384 different skeletal ages could be due to the differences in the criteria for maturity indicators and the 385 associated statistical weighting provided to maturity indicators being different between Tanner-386 Whitehouse 2 and 3 (1983; 2001) methods, ultimately deriving different skeletal ages (17). Thus, the 387 most reliable method for estimating skeletal age remains unclear, yet the current review is supportive 388 of claims by Malina et al. (25) who advocated using Tanner-Whitehouse 2 (1983) rather than 3

(2001) due to the systematic lowering of skeletal ages and the potential negative consequences thismay have during maturational assessments when using the latter method.

391 Our findings suggest none of the estimated ages at PHV were equivalent to actual ages at PHV 392 with any of the proposed predictive methods (Teunissen et al., 49). On average, the Moore 2 (2015) 393 equation estimate was the closest to actual age at PHV (mean range = 0.3 years), followed by Moore 1 394 (2015, mean range = 0.6 years), Fransen (2018, mean range = 0.7 years), and Mirwald (2002, mean 395 range = 0.75 years). As these are group and not individual estimates, caution must be taken when 396 interpreting these findings, given that large inter-individual differences in maturational timing are 397 evident between players of the same chronological age group (9). Recent work from Teunissen et al. 398 (49) reported the Mirwald (2002) and Fransen (2018) equations provide the most stable estimates of 399 age at PHV over time, though none of these equations have longitudinal stability in more than 45% of 400 players, with evidently wide 95% confidence intervals (Fransen, 2018 = -0.38-0.25 years; Mirwald, 401 2002 = -0.29-0.12 years). The Fransen (2018) equation demonstrated higher correlative values with 402 actual age at PHV compared to the Mirwald (2002) equation (90% vs. 89% respectively), even when 403 the difference between age at PHV and observed chronological age was large (32). This could provide 404 some confidence to practitioners aiming to predict age at PHV in academy youth soccer players, as this 405 equation can be applied from an early chronological age without inflating the prediction error, though 406 more research is needed to support this claim. Recent criticism of the Fransen (2018) method has 407 emerged, which soccer practitioners using this method need to carefully consider. According to Nevill 408 & Burton (54), the Fransen (2018) method is flawed due to the inclusion of a player's chronological 409 age in both sides of the prediction equation, which the authors argue will inevitably result in high correlative R² values similar to the present review. Given that age at PHV is used by many professional 410 411 soccer club academies to assess their players (7), it is worth noting that all predictions have associated 412 errors when applied to individual players and therefore, the individual timing and rate of growth spurt 413 need to be considered for all players when selecting the appropriate predictive method to use for 414 deriving age at PHV on an individual basis (9). From a group perspective, both Moore 1 and 2 (2015) 415 equations produced the smallest amount of over/underestimation (0.3 and 0.6 years respectively) from 416 the observable age at PHV.

417 One study investigated biological age amongst four predictive equations (7). All four equations 418 were consistent in their estimates for biological age with a maximum difference of 0.3 years, suggesting 419 that there are tight limits of agreement. The tight limits of agreement between the maturity offset 420 methods (Fransen, 2018; Mirwald, 2002; Moore, 2015) is unsurprising given they all derived from the 421 same original regression equation. Still, the percentage of adult height equation was derived from a 422 different regression equation (31), therefore this could be an underlying reason for the higher biological 423 age with this method compared to the previous three (7). Furthermore, the Khamis-Roche (1994) 424 method contains a genetic component within the equation by including mid-parental height, a variable 425 that is not used with the other equations, which could also explain the slight difference in biological 426 age using the Khamis-Roche (1994) method in comparison to the maturity offset methods. One 427 criticism of this study is the lack of inclusion for any observed values of biological age to compare 428 these estimates against, therefore, the true reliability of these predictive methods remains unknown. 429 One final suggestion proposed by Salter et al. (7), which the present review supports, is to not use 430 maturity offset methods and predicted adult height methods interchangeably, given they provide 431 different estimates of biological age.

432 **Table 6.** Performance of invasive (i.e. medical imaging and hand scans) and non-invasive (i.e. predictive equations) maturity indicators.

433 *TW2 = Tanner-Whitehouse 2. TW3 = Tanner Whitehouse 3*

Study	Population	Population demographics	Country	Observed/ predicted	Method	Data
Lolli <i>et al.</i> , (45)	N = 103 youth academy players	Age (years): 11-17 at time of data collection. Sex: Male Body mass: Not reported Height (cm): 137.5 - 187	Qatar	Observed vs predicted	Actual - predicted adult height (BoneXPert, TW2, TW3)	Chronological age 12.5-17.5 years:BoneXPert (Bayley-Pinneau): -0.46cmTW2: -0.45cmTW3: -1.32cmSkeletal age 12.5- 17.5 years:BoneXPert (Bayley-Pinneau): -0.89cmTW2: -0.53cmTW3: -2.1cm

Adult height

Parr <i>et al.</i> , (9)	N = 23 youth soccer players	Age (years): Initial observation: 12.4 ± 0.6 Final observation: 15.4 ± 0.6 Sex: Male Height (cm): U13: 162.2 ± 7.6 U14: 167.8 ± 8.1 U15: 175.5 ± 7.0 U16: 178.8 ± 4.6 U17: 179.2 ± 4.2 Body mass: Not reported Ethnicity: N = 15 European N = 8 non-European	England	Observed vs predicted	Predicted (Khamis- Roche) - observed adult height	-0.9 cm
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	Ruf <i>et al.</i> , (48)	N = 114 youth soccer players	Age (years): U12: 11.4 \pm 0.3 U13: 12.6 \pm 0.3 U14: 13.5 \pm 0.2 U15: 14.6 \pm 0.3 U16: 15.5 \pm 0.4 U17: 16.5 \pm 0.4 U17: 16.5 \pm 0.4 Sex: Male Height (cm): U12: 146.4 \pm 6.2 U13: 153.5 \pm 6.6 U14: 167.0 \pm 8.3 U15: 171.1 \pm 5.8	Germany	Predicted vs predicted	BAUS (TW2) - Khamis- Roche: Predicted adult height difference Percentage of predicted adult height difference	-0.73cm 0.37%

	U16: 177.9 ± 6.6		
	U17: 174.7 ± 6.6		
	Body mass (kg):		
	U12: 37.3 ± 6.2		
	U13: 41.3 ± 5.1		
	U14: 56.6 ± 9.4		
	U15: 61.2 ± 7.4		
	U16: 69.0 ± 7.5		
	U17: 70.3 ± 7.3		
	Ethnicity: European, African, Middle- eastern		

	Skeletal age									
Study	Population	Population demographics	Country	Observed/ predicted	Method	Data				
Malina <i>et al.</i> , (25)	N = 1,831 youth soccer players	Age (years): 10-17 Body mass: Not reported Height: Not reported	Portugal Belgium Spain Japan Thailand Italy Mexico Brazil	Predicted vs predicted	TW3 – TW2 SA difference	11 years: -0.97 years 12 years: -1.13 years 13 years: -1.16 years 14 years: -1.09 years 15 years: -1.02 years 16 years: -1.00 years 17 years: -1.07 years Total: -1.06 years				

	Age at PHV									
Study	Population	Population demographics	Country	Observed/ predicted	Method	Data				
Parr <i>et al.</i> , (9)	N = 23 youth soccer players	Age (years):Initial observation: 12.4 ± 0.6 Final observation: 15.4 ± 0.6 Sex: MaleHeight (cm):U13: 162.2 ± 7.6 U14: 167.8 ± 8.1 U15: 175.5 ± 7.0 U16: 178.8 ± 4.6 U17: 179.2 ± 4.2 Body mass:Not reported	England	Observed vs predicted	Predicted (Mirwald) - observed age at PHV	0.89 years				

		Ethnicity: European and Non-European				
Teunissen <i>et al.</i> , (49)	N = 17 youth soccer players	Age (years): 11.9 ± 0.8 Sex: Male Height (cm): 149.7 ± 6.2 Body mass (kg): 38.9 ± 5.9 Ethnicity: European ancestry, African, Middle-Eastern	Netherlands	Observed vs predicted	Observed – predicted (Mirwald, Moore 1, Moore 2, Fransen) age at PHV	Observed age at PHV = 13.8 years Mirwald: 0.6 years Moore 1: 0.6 years Moore 2: 0.3 years Fransen: 0.7 years
Fransen <i>et al.,</i> (33)	N = 1,330 youth soccer players	Age (years): 8-17 Sex: Male Body mass: Not reported	Belgium	Predicted vs predicted	Maturity ratio vs maturity offset predictions of age at PHV	Maturity offset Standard error: 1.962 Correlation: 89.22%

He	eight:		Maturity ratio
No Eti Di Ca	ot reported hnicity: verse, mainly ucasian		Standard error: 0.051 Correlation:
			90.1970

Biological age

Study	Population	Population demographics	Country	Observed/ predicted	Method	Data
Salter <i>et al.</i> , (7)	N = 113 youth soccer players	Age (years): 14.3 ± 1.1 Sex: Male Height (cm): 170.1 ± 10.6 Body mass (kg): 58.7 ± 10.5 Ethnicity: 90% White- British, <10% from other ethnic minorities	England	Predicted vs predicted	Mirwald vs Moore vs Fransen vs Khamis-Roche	Mirwald: 14.4 years Moore: 14.3 years Fransen: 14.3 years Khamis-Roche: 14.7 years

435 **Concordance between invasive and non-invasive methods**

436 One of the major aims of the present review was to evaluate the level of agreement between 437 invasive and non-invasive methods for assessing maturational status and timing in academy soccer 438 players. Previously, relatively poor agreement between invasive and non-invasive methods for 439 assessing maturity status has been reported (20) and the present review supports this supposition. 440 Findings suggest a moderate agreement, at best, between invasive and non-invasive methods for 441 assessing maturational status and timing (Table 7). Due to the lack of concordance between invasive 442 and non-invasive methods, caution is required when interpreting correlative values based on non-443 significant and significant Spearman or Pearson factors. One criticism of the studies that investigated 444 the agreement between invasive and non-invasive methods, is the over reporting of correlative values 445 and inconsistent reporting of the size of agreement between these methods (44,20,48), in addition to a 446 lack of longitudinal follow up on the true relationship between these methods (44,20,47,48). The 447 studies in this review were largely inconsistent in the reporting of effect sizes, therefore the true nature 448 of the relationship(s) cannot be determined with confidence.

449 Invasive methods are considered as, the 'gold standard' for assessing biological maturation in 450 adolescent soccer players (20) and therefore it was unsurprising to observe moderate to high 451 correlations (skeletal age vs. pubic hair, r = 0.4; skeletal age assessed via Magnetic Resonance Imaging 452 and ultrasound techniques, r = 0.8) between invasive methods existed (20). Some differences were 453 evident between different age groups for the concordance between invasive and non-invasive methods 454 (e.g., U12, r = 0.62; U14, r = 0.67), with higher correlations for the concordance in the older vs. the 455 younger age group in some studies (44) but not others (U11-U12 = 62%; U13-U14 = 60%) (20). This 456 could be representative of a general maturity factor associated with maximal growth and biological 457 maturation typically seen with this older age group (20,44), moreover, it could also represent the 458 variation in the individual timing of maturation associated with players around this period, with 459 maturity status varying as much as 5-6 years for players of the same chronological age (8). Despite the 460 high correlative values, the size of agreement and associated effect sizes between invasive methods 461 was not reported by any of the included studies and warrants further investigation.

462 The analysis of non-invasive method results revealed only fair to moderate agreement with 463 invasive methods (20,47,48). Similar trends occurred whereby high correlative values did not translate 464 into similar levels of agreement for the relationship between invasive and non-invasive methods (48). 465 Methods such as percentage of predicted adult height ranged from 57-68% in agreement with invasive 466 methods (e.g., skeletal age) whereas age at PHV ranged from 55-65% in agreement with invasive 467 methods. Two notable findings from the data demonstrate a high level of agreement between two non-468 invasive methods (age at PHV and percentage of predicted adult height) ranging between 61-75%, 469 possibly due to the collection of similar anthropometric variables (44), and the use of 'coaches eye' 470 (i.e., a subjective estimation made by coaches on individual player maturity status) having moderate 471 levels of agreement with skeletal age (74%). However, the latter finding should be viewed with some 472 caution as this method is still prone to error and requires experienced staff members to make valid 473 estimations of player maturation (47). Furthermore, this study was also limited to a cross-sectional 474 study design, so the longitudinal stability of this method is yet to be determined. The disparity between 475 invasive and non-invasive methods may be explained by the population differences between the 476 reference samples used for developing the non-invasive methods currently used within current 477 professional soccer environments and modern academy youth players. (e.g., Mirwald, 2002; Khamis-478 Roche, 1994) The existing non-invasive equations for predicting age at PHV and adult height were 479 developed on adolescents of European ancestry from the general population (20). However, youth, 480 academy soccer players worldwide tend to mature earlier in comparison to adolescents from the general 481 population after 13 years of age (55). This advanced skeletal maturity is associated with transient 482 increases in body mass, muscular strength and power, and VO_2 max (55). Therefore, it is reasonable 483 to conclude that academy soccer players are not equivalent to the general adolescent population, and 484 the sample used for developing current non-invasive predictive equations (18). These population 485 differences question the validity and reliability of using these non-invasive methods within academy 486 soccer players and further investigative studies that take these population differences into consideration 487 are required. Collectively, these results indicate that invasive and non-invasive methods should not be 488 used interchangeably given their relatively poor agreement (43), therefore practitioners are advised not 489 to combine invasive and non-invasive methods when assessing maturational status and timing in 490 academy soccer players (44).

Table 7. Concordance between invasive and non-invasive methods. *MRI = Magnetic Resonance Imaging. US = Ultrasound. SA = Skeletal

492 age. $CA = Chronological age^*$

Study	Population	Population demographics	Country	Method	Correlation	Kappa coefficient	Percentage of agreement	Magnitude of agreement
Lehyr <i>et al.</i> , (44)	N = 63 German soccer players	Age (years): U12: 11.3 \pm 0.3 U14: 13.4 \pm 0.3 Sex: Male Body mass (kg): U12: 39.13 \pm 4.33 U14: 51.37 \pm 8.88 Height (cm): U12: 150.06 \pm 5.48	Germany	U12SA MRI vs SA USSA MRI vs MirwaldSA MRI vs Khamis-RocheU14SA MRI vs SA USSA MRI vs SA USSA MRI vs MirwaldSA MRI vs Mamis-RocheTotal SA MRI vs SA US	0.56 0.63 0.66 0.65 0.74 0.61 0.80	Not reported	Not reported	Not reported

		U14: 164.86 ± 10.23		SA MRI vs Mirwald SA MRI vs Khamis-Roche	0.84 0.81			
Malina <i>et al.</i> , (20)	N = 180 youth soccer players	Age (years): 10-15 Sex: Male Height: Not reported Body mass: Not reported	Portugal	11-12 yearsPercentage of predicted adult height vs SA-CA differenceAge at PHV vs SA-CA differenceAge at PHV vs Percentage of predicted adult height:SA-CA vs pubic hair stages 1-5Age at PHV vs pubic hair stages 1-5	0.27 0.43 0.26 0.40 0.50	0.23 0.11 0.12 Not reported Not reported	57% 55% 75% Not reported Not reported	Not reported Not reported Not reported Not reported Not reported

		Percentage of predicted adult height vs pubic hair stages 1-5	0.36	Not reported	Not reported	Not reported
		<u>13-14 years</u> Percentage of predicted adult height vs SA-CA difference	0.47	0.23	63%	Not reported
		Age at PHV vs SA-CA difference	0.29	0.13	57%	Not reported
		Age at PHV vs Percentage of predicted adult height	0.34	0.02	61%	Not reported
		SA-CA vs pubic hair stages 1-5	0.40	Not reported	Not reported	Not reported
		Age at PHV vs pubic hair stages 1-5	0.16	Not reported	Not reported	Not reported
		Percentage of predicted adult height vs pubic hair stages 1-5	0.34	Not reported	Not reported	Not reported

Romann & Fuchoslacher (47)	N = 119 youth soccer players N = 6 national coaches of U15-21 Swiss national team	Age (years): 14 ± 0.3 Sex: Male Height (cm): 164.9 ± 8.4 Body mass (kg): 53 ± 8.4	Switzerland	Skeletal age vs coaches eye Skeletal age vs age at PHV	0.62 0.42	0.48	73.9% 65.5%	Moderate Fair
Ruf <i>et al.</i> , (48)	N = 114 youth soccer players	Age (years): U12: 11.4 \pm 0.3 U13: 12.6 \pm 0.3 U14: 13.5 \pm 0.2 U15: 14.6 \pm 0.3 U16: 15.5 \pm 0.4 U17: 16.5 \pm 0.4	Germany	Z score 0.50: Percentage of predicted adult height vs SA-CA difference Z score 0.75: Percentage of predicted adult height vs SA-CA difference	0.52	0.37	65%	Not reported

Sex: Male	<u>Z score 1.00:</u>				
Height (cm): U12: 146.4 ± 6.2	Percentage of predicted adult height vs SA-CA difference	0.45	0.31	66%	Not reported
U13: 153.5 ± 6.6					
U14: 167.0 ± 8.3 U15:	BAUS software vs Khamis-Roche:				
171.1 ± 5.8 U16:	Predicted adult height	0.86	Not reported	Not reported	Not reported
177.9 ± 6.6 U17:	Percentage of predicted adult height	0.96	Not reported	Not reported	Not reported
Body mass (kg) :	Biological age	0.80	Not reported	Not reported	Not reported
U12: 37.3 ± 6.2					
U13: 41.3 ± 5.1					
U14: 56.6 ± 9.4					

U15: 61.2 ± 7.4			
U16: 69.0 ± 7.5			
U17: 70.3 ± 7.3			
Ethnicity: European, African, Middle-eastern			

494 Reliability of X-Ray, DXA and MRI scanning techniques

495 Investigative studies regarding the reliability of invasive scanning techniques in academy 496 soccer players remain limited, with only two studies included in the current review (43,22). However, 497 the findings from these studies report acceptable estimates for assessing skeletal age and maturity status 498 in academy soccer players (Table 8). Inter-observer agreement was considered excellent for using 499 DXA (intra class coefficient = 0.93) and X-Ray (intra class coefficient = 0.92) scanning to assess 500 skeletal maturity (22). Meanwhile, MRI (intra class coefficient = 0.828), inter-observer agreement was 501 considered very good. On the other hand, intra-rater reliability for DXA and X-Ray were also 502 considered excellent with intra class coefficients ranging from 0.95-0.97 for DXA and 0.98 for X-Ray, 503 respectively. Unfortunately, no values for intra-rater reliability were reported for MRI which can be 504 considered a limitation of the study (43). Collectively, the results demonstrate the efficiency of MRI, 505 X-Ray, and DXA scanning for assessing skeletal maturity in academy soccer players, yet further 506 validation of these methods is needed in players of different ethnicities as well as longer follow-up 507 periods to ensure long term reliability.

Despite the efficacy of these methods for assessing skeletal maturity and age in academy soccer players, subtle differences exist between the characteristics of these methods. For example, DXA scans are known to have significantly less radiation compared to MRI and X-Rays (22) and given the similar level of agreement between X-Ray and DXA scanning for assessing skeletal maturity, practitioners in soccer may be inclined to select DXA scanning instead of X-Ray scanning. However, DXA scanning is more time-consuming and expensive compared to X-Rays, which are additional considerations for academy soccer practitioners (22).

515 MRI has received more research attention than X-Ray and DXA scanning for assessing skeletal 516 maturity (21,43). MRI correlates highly with chronological age (21) and findings from Abdelbary *et* 517 *al.*, (43) support the use of MRI to assess skeletal maturity in academy soccer players. Further evidence 518 of the use of MRI includes shorter scanning times and higher image resolution, however high costs and 519 expertise required are potential disadvantages of this method. In sum, all the discussed invasive 520 methods report high cross-sectional reliability for assessing skeletal maturity in academy soccer 521 players, but further validation of these methods and exploration of other alternatives (e.g., ultrasound) 522 are needed.

Table 8. Reliability of X-Ray, DXA and MRI scanning techniques. *MRI = Magnetic Resonance Imaging. DXA = Dual energy X-ray

524 Absorptiometry*

Study	Population	Population demographics	Country	Outcome variables	ICC	Classification
Abdelbary <i>et al.</i> , (43)	N = 61 youth soccer players	Age (years): 13-18 Sex: Male Height: Not reported Body mass: Not reported	Egypt	Inter-rater reliability	<u>MRI grade of</u> <u>fusion vs</u> <u>actual age</u> 0.828	Very good
Romann & Fuchoslacher (22)	N = 63 youth soccer players	Age (years): 14 ± 0.3 Sex: Male Height (cm): 164.9 ± 8.4 Body mass (kg): 53 ± 8.7	Switzerland	Inter/intra-rater reliability	$\frac{Intra-rater}{DXA}$ R1: 0.97 R2: 0.95 $\frac{Inter-rater}{DXA}$ R1 + R2: 0.93	Excellent Excellent Excellent

		Intra-rater X- Ray	
		R1: 0.98	Excellent
		R2: 0.98	Excellent
		<u>Inter-rater X-</u> <u>Ray</u>	
		R1 + R2: 0.92	Excellent

526 **Concordance of maturity status classification**

527 Only a moderate agreement was found for the concordance of maturity status classifications 528 (Table 9). Utilising the Fels (1977) method compared to MRI to identify skeletally mature players, 529 revealed that more players were classed as skeletally mature using the Fels (1977) method compared 530 to MRI across the U15-U17 years, particularly for ages 16-17 years. A combined total of players aged 531 16-17 years reported that 62% of players were skeletally mature utilising the Fels (1977) method, 532 whereas only 22% were skeletally mature with MRI. Results are limited to three age groups and thus 533 may not represent the true discrepancies between these methods within the full academy system. One 534 explanation for these methodological discrepancies could be that MRI has six stages of fusion as 535 criteria to describe skeletal maturity, but the Fels (1977) method only has four (17), therefore, the 536 researcher interpretation of criteria to ascertain the degree of fusion for skeletal maturity at each stage 537 may differ between methods, with MRI fusion described via percentages and descriptive information 538 and the Fels (1977) method relying solely on descriptive information to assess skeletal fusion (17).

539 A moderate agreement (55%) between the Fels (1977) and Tanner-Whitehouse 3 (2001) 540 method was reported in the data, with a slightly lower level of agreement reported between the Tanner-541 Whitehouse 2 (1983) and 3 (2001) methods (52%). Differences were observed in the agreement for the 542 number of earlier (33%) and average (86%) maturing players between the Fels (1977) and Tanner-543 Whitehouse 3 (2001) method and the highest level of agreement between Tanner-Whitehouse 2 (1983) 544 and 3 (2001) methods were observed for the U11 (57%) and U14 (58%) years. Poor concordance 545 between the Fels (1977) and Tanner-Whitehouse 3 (2001) methods is expected given the differences 546 in reference samples used to develop each method, the different bones and criteria used to assess 547 skeletal maturity, and importantly the assignment of skeletal age given to a hand/wrist radiograph (27). 548 Given the relatively small sample size of the study (n = 40), we advise future studies use larger sample 549 sizes whilst including players of different ethnicities to validate these findings, which is vital given that 550 soccer academies across the world are becoming increasingly more diverse and consist of players with 551 different ethnicities, who undergo different patterns of maturation (20,53). A higher level of agreement 552 was observed in the U11 and U14 years, which is interesting as typically both U11 and U14 years are 553 considered pre-PHV and circa-PHV respectively (55). Therefore, the higher levels of agreement in 554 these age groups likely reflect the high proportion of average maturing players in these age groups. 555 However, the lower levels of agreement in the other age groups are a possible reflection of the variance 556 in the number of earlier and later maturing players in these age groups and therefore, individual timing 557 and growth rates are important factors to be considered for players within these age groups (9).

558 The analysis of non-invasive methods to classify player maturity status revealed a higher 559 amount of average maturing players using the Moore (2015; 50%) compared to the Mirwald (2002; 560 43%) method with a higher amount of later maturing players using the Mirwald (2002; 66%) compared 561 to the Moore (2015; 43%) method. Typically, substantial agreement (64-67%) was observed between 562 maturity offset methods, with only a moderate agreement (44-50%) seen between maturity offset and 563 predicted adult height methods, utilising a more conservative threshold of 85-95% of predicted adult 564 height. Interestingly, this level of agreement decreased when using a less conservative threshold of 88-565 93% of predicted adult height. The level of agreement was only moderate (58-60%) between maturity 566 offset methods and fair (31-43%) between maturity offset and predicted adult height methods when 567 using this less conservative threshold. The higher concordance between maturity offset methods is 568 expected given they are all derived from an identical regression equation (7) with the predicted adult 569 height equation deriving from an alternative regression equation. A lower agreement between maturity 570 offset and predicted adult height methods may also be reflective of the different variables that these 571 methods collect (adult height vs. time period from PHV). Together this re-iterates the premise that 572 maturity offset and predicted adult height methods should not be used interchangeably (7). The higher 573 level of agreement when utilising a more conservative threshold is unsurprising, as they account for 574 the error rate associated with assessments, as well as providing a broad range of players who are 575 classified as on time in their maturity classification (17). To sum, Salter *et al.*, (7) highlights the 576 differences in the classification of players maturity status using various invasive and non-invasive 577 methods. Practitioners are advised not to use these methods interchangeably and instead consider the 578 individual maturational timing of players. Longitudinal studies with larger sample sizes are required 579 to validate findings presented in this review. **Table 9.** Concordance of maturity status classification. * MRI = Magnetic Resonance Imaging. SA = Skeletal age. TW2 = Tanner-

Study	Population	Population demographics	Country	Invasive/non-invasive	Data
Malina <i>et al.,</i> (17)	N = 592 youth soccer players	Age (years): Series 1: 11-17 Series 2: 11-17 Series 3: 12-16 Sex: Male Height: Not reported Body mass: Not reported	Portugal, Spain	Invasive MRI grade of fusion vs Fels SA frequency of skeletally mature players	15 years Fels: 8% MRI: 3% 16 years Fels: 23% MRI: 7% 17 years Fels: 39% MRI: 15%
Malina <i>et al.</i> , (27)	N = 40 youth soccer players	Age (years): 12-16 U11-12: 12.78 \pm 0.18 U13-14: 14.1 \pm 0.39 U15-16: 15.7 \pm 0.32	Spain	Invasive Fels vs TW3 percentage of agreement	Late maturers 100% Average maturers 85.7% Early maturers 33.3% Mature 100%

		Sex: Male Height: Not reported Body mass: Not reported			Correlation: 0.66 Kappa coefficient: 0.59 Percentage of agreement: 55% Magnitude of agreement: Moderate
Malina et al., (25)	N = 1,831 youth soccer players	Age (years): 10-17 Body mass: Not reported Height: Not reported	Portugal, Belgium, Spain, Japan, Thailand, Italy, Mexico, Brazil	Invasive TW2 vs TW3	U11Late: 13.4%Average: 34.5%Early: 9%Percentage of agreement: 56.9% $U12$ Late: 13.6%Average: 17.5%Early: 19.7%Percentage of agreement: 51.1% $U13$ Late: 8.7%Average: 19.3%Early: 18.4%Percentage of agreement: 46.4% $U14$ Late: 4.8%Average: 26.7%Early: 26.7%Percentage of agreement: 58.1%

					U15Late: 4%Average: 33.3%Early: 7.3%Percentage of agreement: 44.6%Total:Late: 9.8%Average: 25.5%Early: 16.7%Percentage of agreement: 52%
Malina <i>et al.,</i> (46)	N = 58 youth soccer players	Age (years): 11-14 Sex: Male Height: Not reported Body mass: Not reported Ethnicity: European ancestry	Portugal	Non-invasive Mirwald vs Moore	Early Mirwald: 0% Moore: 3% <u>Average</u> Mirwald: 43% Moore: 50% <u>Late</u> Mirwald: 66% Moore: 43%
Salter <i>et</i> <i>al.</i> , (7)	N = 113 youth soccer players	Age (years): 14.3 ± 1.1 Sex: Male Height (cm): 170.1 ± 10.6	England	Non-invasive Moore vs Fransen vs Mirwald vs Khamis- Roche	<u>85-96% PAH</u> Moore – Mirwald kappa: 0.67 (substantial) Fransen – Mirwald kappa: 0.66 (substantial) Fransen – Moore kappa: 0.64 (substantial)

Body mass (kg):	Khamis-Roche – Mirwald kappa: 0.49
58.7 ± 10.5	(moderate)
	Khamis-Roche – Moore kappa: 0.50
Ethnicity:	(moderate)
90% White-	Khamis-Roche – Fransen kappa: 0.44
British, <10%	(moderate)
from other ethnic	
minorities	<u>88-93% PAH</u>
	Moore – Mirwald kappa: 0.60
	(moderate)
	Fransen – Mirwald kappa: 0.59
	(moderate)
	Fransen – Moore kappa: 0.58
	(moderate)
	Khamis-Roche – Mirwald kappa: 0.31
	(fair)
	Khamis-Roche – Moore kappa: 0.43
	(moderate)
	Khamis-Roche – Fransen kappa: 0.39
	(fair)
	<u>85-95% PAH</u>
	Maturity offset methods: 64-67%
	(substantial)
	Maturity offset vs PAH methods: 44-50%
	(moderate)
	<u>88-93% PAH</u>
	Maturity offset methods: 58-60%
	(moderate)
	Maturity offset vs PAH methods: 31-43%
	(fair)

583 Limitations of included studies

584 The critical appraisal of the included studies revealed a higher proportion of moderate 585 compared to high quality studies. A limitation of the current review is that we did not capture any 586 randomised controlled trials and alternatively reviewed studies adopting observational designs. The 587 studies included contained a higher proportion of cross-sectional (n = 8) compared to longitudinal 588 (n = 5) studies which could be considered a limitation of the current review. Longitudinal studies have 589 the potential to better describe the relationship between invasive and non-invasive methods over time 590 which is not possible with cross-sectional studies. Longitudinal studies also have the capacity to 591 determine the stability and reliability of some of the estimates and measures of maturation that are 592 produced by these invasive and non-invasive methods (49). A further assessment of the included 593 studies is the inconsistency in sample sizes, with a highly variable range of participants (n = 17-1831). 594 A limitation consistently reported from the authors of included studies was the underpowered sample 595 sizes which may have potentially influenced some of the findings presented in this review. However, 596 some of the included studies included predicted values for outcome variables such as age at PHV, 597 maturity offset, predicted adult height, and biological age, however, observed values were absent for 598 some of these variables and thus the true extent of the reliability of these estimates currently remains 599 unknown. Results for some of the included studies are restricted to a limited number of age groups 600 (e.g., number of skeletally mature players) hence some of the results may not be transferrable to other 601 age groups. Collectively, the number of limitations associated with the included studies suggests the 602 need for future research examining maturation status and timing in academy soccer players to consider 603 study design and data capture procedures.

604

606 Strengths and limitations of the review process

607 The present review included an objective framework (AMSTAR 2) to analyse the research 608 process and subsequently reduce any bias within the review. The use of this framework revealed many 609 strengths of the review process such as the inclusion criteria including all necessary components of 610 PICOSS (Population, Intervention/Comparator, Outcome, Study design, Setting), with all relevant 611 studies being captured using a comprehensive literature search, incorporating multiple databases. 612 Further, we used a detailed screening process completed by two members of the research team and all 613 included and excluded studies were justified by consensus. Data extraction was completed by two 614 members of the research team which aided in the synthesis of individual studies based on common 615 outcome variables, methods, and designs. An objective framework was used to assess study quality. 616 The review process accounted for the risk of bias (e.g., publication bias) when interpreting the 617 individual study findings.

618 The review process is not without its limitations, for example we only included studies in the 619 English language and therefore it is possible that some relevant articles may not have been captured 620 due to the filtering of English language search terms. Furthermore, we only included studies examining 621 male, youth soccer players aged U9-U18 years, consequently excluding female soccer players, male 622 amateur players and adolescents from the general population and players <U9 or >U18 years, which 623 makes the findings from the review applicable to only a small proportion of male soccer players. The 624 high heterogeneity in the data prevented a meta-analysis from being completed and the review is 625 limited to a narrative synthesis of the data. A final consideration is the confirmation that this review is 626 of moderate quality (see Table 5).

628 Comparisons to other reviews

To our knowledge, no existing systematic review for assessing maturational status and timing in academy soccer players is available, making comparisons to the present review challenging. Nonetheless, two reviews have been conducted on the general population from adolescents and focused on methods to predict PHV (timing) but not methods to assess maturational status (36) and aimed to provide a critical narrative summary of the methods to assess maturational status and timing in adolescents (28).

635 The results of the present narrative review demonstrate that the Mirwald (2002) equation 636 overestimated age at PHV by 0.6-0.9 years, a similar finding that was reported by Mills et al., (36) who 637 found in three studies that in the year before PHV, the Mirwald (2002) equation also overestimated age 638 at PHV. However, an extended finding from Mills et al., (36) was the increased accuracy of the 639 Mirwald (2002) equation for predicting age at PHV when data was acquired three years prior to the 640 actual age of PHV, which equated to age 11 years in boys. This finding was not reported in the present 641 review and shows some promise for these anthropometry-based methods (36). Another common 642 finding between the present review and Mills et al., (36) was the high reliability scores of radiograph-643 based methods (i.e., MRI, X-Ray/DXA scanning) and anatomical surrogate measures. The present 644 review differs from Mills et al., (36) as the radiograph-based methods in this review investigated 645 skeletal age and grade of fusion (maturity status) as opposed to age at PHV. However, the established 646 reliability of radiographic methods can give practitioners the confidence to consider these methods.

The present review concluded that Tanner-Whitehouse 3 (2001) skeletal ages are consistently lower than corresponding Tanner-Whitehouse 2 (1983) derived skeletal ages among youth athletes aged 11-17 years. This review concludes that the difference can be as much as 0.97-1.07 years for young soccer players aged 11-17 years and would support the argument that Tanner-Whitehouse 2 651 (1983) should be used instead of the Tanner-Whitehouse 3 (2001) (28). A noteworthy discrepancy 652 between the present review and Malina et al., (28) was maturity status classification. A reasonable 653 concordance for maturity status classification of soccer players was reported for skeletal age and age 654 at PHV, a comparison not made by any of the studies in the present review. This review reported only 655 a moderate concordance between invasive skeletal age methods and maturity offset vs. predicted adult 656 height methods. A substantial concordance was reported between maturity offset methods, which 657 reduced to only a moderate agreement when a less conservative banding threshold was used (88-93% 658 predicted adult height). This highlights a potential gap in the literature for future research to investigate 659 regarding the agreement of maturity status classification between invasive and non-invasive methods, 660 given the failure of the present review to address this.

661 Applied implications

662 The implications of the present review can benefit practitioners when assessing maturational status and 663 timing in academy soccer players. Although the findings reported in this review may not be 664 generalisable to amateur male or female players, they highlight some important considerations for 665 soccer clubs responsible for male academy players. Firstly, many of the non-invasive methods adopted 666 by soccer academies were developed using populations that significantly differ in ethnicity, 667 socioeconomic background and maturational status from modern academy players, which questions 668 the reliability of using these methods in the target population. Saying that, in the absence of any viable 669 alternatives, practitioners working in soccer academies are restricted to using these non-invasive 670 methods or opt for more invasive methods involving medical scanning and subsequent radiation 671 exposure for assessing maturational status and timing in academy players. Given the relatively poor 672 concordance between invasive and non-invasive methods for assessing maturational status and timing 673 in academy soccer players highlighted in this review and other reviews (8), it is recommended that 674 practitioners avoid using these methods interchangeably. It is worth noting that all non-invasive 675 methods have associated errors when applied to individual players, therefore, new predictive methods 676 or modifications to existing equations are warranted that carefully consider the individual timing and 677 rate of maturation amongst this culturally diverse and unique population (9).

678 Suggestions for future studies

679 Despite a wealth of studies using the general adolescent population, the investigation of 680 maturational assessments and associated performance effects within academy soccer is still in its 681 infancy. From a holistic perspective, practitioners and researchers in this field may need to look beyond 682 simply the methods they employ to assess maturational status and timing in academy soccer players 683 and consider the wider implications of their choices on issues such as injury risk. Recent work has 684 highlighted an increase in injury risk and incidence around reported mean ages (i.e., 13-14 years) at 685 PHV (14,11). Furthermore, the growth spurt coupled with its maturity-associated variations are among 686 some of the injury risk factors for the developing male athlete (10). Collectively, these findings 687 demonstrate the importance of using reliable methods to correctly assess a player's maturational status, 688 given the subsequent impact this can have for training load management and injury risk around the 689 time of PHV (8,11).

Given the high amount of heterogeneity in the available literature, future research should focus on the development of a homogeneous approach to data collection of maturity-related data and outcome variables during maturational research. Such data will enable a subsequent quantitative analysis to be completed, thus allowing researchers to better understand the reliability of these invasive and noninvasive methods. One possible solution to achieve a homogenous approach to future research within this area is to gain industry consensus on the rationale for professional clubs using specific types of maturational assessment methods when compared to alternatives. Gaining consensus on some of these 697 areas could facilitate the collection of some common outcome variables, which could eventually 698 facilitate the completion of a quantitative meta-analysis in this research area. Further training and 699 education of academy staff who are responsible for the collection of maturity-related data from players 700 across the academy may also be required to ensure reliable and accurate maturational data on an 701 individual and group basis is recorded.

702 The findings presented here suggest the need for more longitudinal studies, given the excess of 703 cross-sectional evidence, that are endorsed by governing bodies (e.g., the English Premier League and 704 the English Football Association) and continue to utilise both invasive and non-invasive methods to 705 monitor maturational status and timing amongst this large and ethnically diverse population. Given the 706 amount of heterogeneity within the results, combined with largely moderate study quality, the true 707 reliability of some of the most widely used methods to assess maturational status and timing in academy 708 soccer players cannot be determined. It is important that the true reliability of these methods is 709 established given the further implications of maturity on injury risk (10) and categorisation of academy 710 soccer players for bio-banding (50).

711 Conclusions

In this present review, we identified 15 studies that utilised invasive, non-invasive or a combination of both methods to assess maturational status and timing in academy soccer players. Despite the number of methods available to modern practitioners, no methods provided equivalent estimations of adult height, skeletal age, or age at PHV. Discrepancies were evident between actual and predicted adult height and actual vs predicted age at PHV. Practitioners utilising the Bayley-Pinneau (1952), Tanner-Whitehouse 2 (1983) or Khamis-Roche (1994) methods to predict adult height can be supported that these methods produce an estimated adult stature within 1cm of actual adult

719 height. Similarly, for age at PHV, practitioners may utilise either the Moore (2015) equations or the 720 Fransen (2018) equation in academy soccer players despite some recent criticism (54). The Moore 721 (2015) equations produced the closest estimates to actual age at PHV, however the Fransen (2018) 722 equation correlated highly with actual age at PHV (>90%), even when the period between chronological age and age at PHV was large. Practitioners should also be aware of the significantly 723 724 younger skeletal ages when using the Tanner-Whitehouse 3 (2001) assessment compared to 2 (1983) 725 method and are therefore advised to use the latter method for assessing skeletal age. The poor 726 concordance between invasive and non-invasive methods, despite high correlative values, is a 727 recommendation to practitioners that these methods should not be used interchangeably for assessing 728 maturational status and timing in academy soccer players. However, to understand the reliability of 729 these various types of methods, further research with improved study designs and reporting of 730 consistent outcome variables are needed to create a homogenous approach to research in this field. 731 Given the well documented association between injury risk and maturation, this review highlights the 732 importance of using reliable and accurate methods to assess maturational status and timing within youth 733 academy soccer players. This review demonstrates a bias towards single club studies (53); therefore, it 734 is our contention that better co-collaboration between clubs and performance staff such as sport 735 scientists would help clubs develop and implement alternative strategies to counteract this ongoing 736 problem.

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743 Supplementary Material

- 744 Please see attached files for supplementary material:
- 745 Supplementary file 1- PRISMA flow chart.
- 746 Supplementary file 2 PRISMA checklist.
- 747 Supplementary file 3- Cover letter.
- The authors confirm that there was no conflict of interest.
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