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Development and Initial Validation of the Perfectionistic Climate Questionnaire (PCQ-S)

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Abstract

Perfectionism has traditionally been conceptualised as a personality trait. However, researchers have recently proposed a new construct—*perfectionistic climate*—which focusses on the degree to which the social environment is experienced as perfectionistic. In the current study, we apply this new construct to sport and begin the development and validation of the first scale to measure perfectionistic climate – the Perfectionistic Climate Questionnaire-Sport (PCQ-S). This self-report scale includes five components of coach behaviour central to perfectionistic climate: expectations, criticism, control, conditional regard, and anxiousness. The development of the PCQ-S included five stages and data from four samples of youth athletes ($N = 730$ total; $M_{\text{age}} = 14.50$, $s = 1.83$). In stage one, outlined in study one, PCQ-S items were generated and then refined using feedback provided by academics, sport coaches, and youth athletes. In stage two, outlined in studies two and three, exploratory factor analysis, confirmatory factor analysis, and exploratory structural equation modelling techniques were used to examine the structure of PCQ-S items. In stage three, outlined in study four, discriminant validity was assessed by estimating the nomological network of relations between test scores on the PCQ-S and established coach climate measures. Then, in the final stage, outlined in study five, measurement invariance of the PCQ-S was examined across gender and age groups. The result of this process was a 20-item five-factor instrument that shows evidence of good factorial validity, can be differentiated from existing coach climate measures, and has measurement invariance. Based on these initial findings, the PCQ-S provides the first valid and reliable way of measuring perfectionistic climate in sport.

Keywords: perfectionism; climate; coach; scale development; youth athlete

Introduction

Perfectionism has traditionally been conceptualised as a personality trait – a personal quality of the individual. However, researchers have recently proposed a new construct: *perfectionistic climate*. Rather than focussing on the personal qualities of an individual that are perfectionistic, perfectionistic climate focusses on the degree to which the social environment can be considered perfectionistic. The theoretical basis for this new construct is provided by Flett et al.'s (2002) model of the development of perfectionism and defined using motivational climate work (Ames, 1992) guided by Achievement Goal Theory (AGT; Nicholls, 1984). In accord, perfectionistic climate is thought to be instrumental in the development of perfectionism and is defined as the informational cues and goal structures (i.e., components of the environment that emphasise what people are expected to accomplish and how they are to be evaluated) that align with the view that performances must be perfect and less than perfect performances are unacceptable (Hill & Grugan, 2019). The introduction of perfectionistic climate may offer an important new way to study the experience of the pressure to be perfect that complements existing approaches in sport and other domains. Herein, we sought to develop a scale to allow researchers and practitioners to begin to examine if this is the case.

Conceptualising Perfectionistic Climate

The notion that social environments can be perfectionistic is well grounded in models of perfectionism development. Specifically, in their seminal work on the topic, Flett, Hewitt, and colleagues (Flett et al, 2002; Hewitt et al., 2017) have outlined how the social environment is key in acting on biological vulnerabilities and shaping both early childhood experiences (e.g., asynchrony between child and caregiver) and formative development during adolescence (Chen et al., 2019; Flett et al., 1995; Ko et al., 2019). In their approach, the influence that key social agents have in shaping perfectionistic social environments is formalised via various pathways. These pathways focus on important behaviours, practices, and interpersonal styles (i.e., what key social agents do). Specifically, they reflect environments in which key social agents are considered as harsh, negligent,

1 and psychologically controlling, extremely difficult to please, and excessively worried about
2 mistakes and the implications of others not being perfect. This approach therefore accounts for the
3 role of social environments in the development of perfectionism and emphasises features that are
4 overtly perfectionistic. In doing so, it provides the theoretical basis for the construct of perfectionistic
5 climate as a particular type of social environment that encourages perfectionism.

6 With Flett et al.'s (2002) developmental model as its primary theoretical basis, perfectionistic
7 climate is situated alongside other climate constructs studied in achievement contexts. Specifically, it
8 is considered akin to the motivational climate constructs of AGT and is defined as such. Within
9 AGT, motivational climate represents the informational cues and goal structures that exist in the
10 environment and define what goals people are expected to achieve and the standards against which
11 they will be evaluated (Ames & Ames, 1984). The key factor that differentiates perfectionistic
12 climate from other existing climates is the focus on behaviours and practices that emphasise a strict
13 requirement for perfection. A highly perfectionistic climate is one in which people perceive that
14 performances must be perfect and less than perfect performances are unacceptable. However, in the
15 same way that other achievement climates exert a broad influence on the experiences people have,
16 we can expect a perfectionistic climate to both reinforce perfectionistic beliefs and give rise to
17 perfectionistic problems.

18 Perfectionistic climate is proposed to include five key components that reflect the various
19 practices and behaviours evident in developmental models (Hill & Grugan, 2019). The first two
20 components of perfectionistic climate are expectations and criticism. In context of perfectionistic
21 climate, the expectations component reflects the perception that key social agents hold and demand
22 unrealistic performance expectations of others. By contrast, the criticism component reflects the
23 perception that key social agents engage in harsh criticism whenever the performance of others is not
24 perfect. These components are drawn from existing models of trait perfectionism that identify
25 unrealistic expectations and harsh criticism from specific others as central dimensions of

1 perfectionism (e.g., parental expectations and parental criticism; Frost et al., 1990). By incorporating
2 these components within perfectionistic climate, Hill and Grugan (2019) propose that, rather than
3 being core features of the trait itself, such perfectionism dimensions are features of the social
4 environment. This proposal follows several claims that these components are developmental aspects
5 of perfectionism rather than core components of the perfectionism personality trait (see Frost et al.,
6 1990; Rhéaume et al., 2000; Shafran et al., 2002).

7 The next two components of perfectionistic climate are control and conditional regard. In
8 context of perfectionistic climate, the control component reflects the perception that key social
9 agents employ externally controlling strategies that place pressure on others to perform perfectly
10 (e.g., the routine use of punishment or sanctions for minor or inconsequential mistakes). By contrast,
11 the conditional regard component reflects the perception that key social agents employ internally
12 controlling strategies that place pressure on others to perform perfectly (e.g., providing approval,
13 acceptance, or support only when performances are perfect). These components represent distinct
14 forms of psychological control that are emphasised in traditional models of controlling behaviour
15 (e.g., Bartholomew et al., 2010; Soenens & Vansteenkiste, 2010). However, in contrast to existing
16 models, perfectionistic climate entails a narrow focus on controlling strategies that place pressure on
17 others to feel, think, and behave in accordance with a specific requirement for perfect performance.
18 This emphasis on perfection is a distinguishing feature of the controlling practices embedded in
19 models of perfectionism development (Flett et al., 2002).

20 The final component of perfectionistic climate is anxiousness. In the context of
21 perfectionistic climate, anxiousness reflects the perception that key social agents are extremely
22 worried and vigilant about mistakes and the consequences of others not performing perfectly. This
23 component is drawn from the anxious rearing model of perfectionism development (Flett et al.,
24 2002). This model identifies how an environment in which key social agents signal an excessive
25 focus on mistakes and the negative implications of imperfection may instil a need to be perfect in

1 others. The importance of including anxiousness as a central component of perfectionistic climate is
2 evident in research examining perfectionism and the parental motivational climate (Gustafsson et al.,
3 2016). This research shows that ego-involving aspects of the climate focused on worry and concerns
4 about failure are positively related to perfectionism in young athletes.

5 **Perfectionistic Climate in Sport**

6 In line with existing approaches to studying motivational climate, the study of perfectionistic
7 climate is likely relevant to various contexts including family, education, and sport settings
8 (Appleton & Curran, 2016; Flett et al., 2002). Here, we focus on the relevance of perfectionistic
9 climate in sport and place emphasis on the youth sport domain. We believe that youth sport provides
10 an ideal starting point for the study of perfectionistic climate. This is because sport is a popular
11 activity for young people and a context in which athletes often experience interpersonal pressure,
12 negative feedback, and contingent rewards (Fraser-Thomas et al., 2005; Gould, 2019; Ryan & Deci,
13 2017). Importantly, some youth athletes report that they experience their environment as excessively
14 demanding and that they frequently encounter performance standards are not just high, but ultimately
15 unrealistic and perfectionistic (see Krane et al., 1997; Lavallee & Robinson, 2007; Udry et al., 1997).
16 By focussing on perfectionistic climate in youth sport, we hope to better understand the experiences
17 of these athletes, identify the effects of training and performing in these environments, and ultimately
18 learn how to better safeguard athletes from any subsequent negative consequences.

19 This application of perfectionistic climate to sport was underpinned and guided by three key
20 proposals. This first key proposal is that the coach is a key figure responsible for shaping the extent
21 to which athletes experience an environment that is perfectionistic. This is an important
22 consideration as perfectionistic climate is introduced as a construct that can be shaped by various
23 significant others (e.g., parents, teachers, and coaches). However, the especial importance of the
24 coach is evident across research examining youth sport experiences (see Harwood et al., 2015).
25 Moreover, the role of the coach and specific coaching practices as key sources of pressure to be

1 perfect are also heavily emphasised in theory relating to the development of perfectionism in sport
2 (Appleton & Curran, 2016). The strongest empirical support in this regard is for the role of
3 unrealistic coach expectations and harsh coach criticism with numerous studies showing positive
4 relationships between these coach behaviours and perfectionism in youth athletes (e.g., Gotwals,
5 2011; Madigan et al., 2019; Sagar & Stoeber, 2009).

6 The second key proposal is that there is a need to differentiate components of coach pressure
7 to be perfect from core components of perfectionism. There are several sport-specific models of trait
8 perfectionism that incorporate a focus on the coach as a source of combined unrealistic expectations
9 and harsh coach criticism (e.g., Dunn et al., 2006). However, there are two key issues that apply to
10 such models. The first issue is that they may be confounding etiological factors with core
11 components of perfectionism (see Sirois & Molnar, 2016, for a related discussion). In this regard, the
12 application of perfectionistic climate to sport is important as it clearly positions components of coach
13 pressure as sources of interpersonal pressure that contribute to experiences of a perfectionistic
14 climate (as opposed to aspects of the trait). The second issue is that subscales measuring coach
15 expectations and criticism are typically collapsed into broader measures of coach pressure. This
16 differs from the model of perfectionistic climate which asserts that expectations and criticism should
17 be studied independently. This distinction may be important as previous research examining parental
18 pressure shows that expectations and criticism have unique effects and interact with each other to
19 predict important outcomes in youth athletes (see McArdle & Duda, 2008).

20 The final key proposal guiding the current application is that perfectionistic climate in sport
21 involves more than unrealistic expectations and harsh criticism. In line with theory relating to the
22 development of perfectionism, there are various other behaviours, practices, and relational styles that
23 influence the extent to which an environment is experienced as perfectionistic (see Appleton &
24 Curran, 2016). This is particularly relevant to the components of control, conditional regard, and
25 anxiousness, which have all been linked to the development of perfectionism in young athletes

1 (Barcza-Renner et al., 2016). Here we include each one of these additional components. In doing so,
2 we can test the construct validity of perfectionistic climate by ascertaining the degree to which each
3 of these components capture separate but related aspects of the same multifaceted construct.

4 **Perfectionistic Climate and Established Motivational Climate Constructs**

5 The perfectionistic climate construct is introduced alongside established climate constructs in
6 sport. The most relevant in this regard are coach climate constructs embedded in AGT and Self-
7 Determination Theory (SDT; Ryan & Deci, 2017). AGT includes task-involving and ego-involving
8 climate dimensions (Duda, 1992). In a task-involving climate, athletes perceive that the coach places
9 emphasis on effort and cooperation, views mistakes as central to personal development, and values
10 all performers regardless of ability. By contrast, in an ego-involving climate, athletes perceive that
11 the coach places emphasis on outperforming others, views mistakes as worthy of punishment, and
12 values only the best performers (Smith et al., 2008). As a complementary approach, SDT includes
13 autonomy supportive and controlling dimensions (Vallerand & Losier, 1999). In an autonomy
14 supportive climate, athletes perceive that the coach cares about the opinions of all performers and
15 provides athletes with meaningful information and choice (Mageau & Vallerand, 2003). By contrast,
16 in a controlling climate, athletes perceive that the coach uses coercive, manipulative, and
17 authoritarian practices that pressure performers to feel, think, and behave in a particular way
18 (Bartholomew et al., 2010).

19 The motivational climate dimensions emphasised in AGT and SDT can be studied
20 independently or integrated and studied under a hierarchical model (Appleton et al., 2016). The
21 model proposed by Duda (2013) differentiates between two higher-order coach climate dimensions:
22 an empowering climate and a disempowering climate. An empowering climate is characterised by
23 task involving features of AGT and autonomy supportive features of SDT, whereas an environment
24 that is disempowering is characterised by ego-involving features of AGT and controlling features of
25 SDT. This hierarchical model is useful as it allows the broad spectrum of climate dimensions

1 proposed in AGT and SDT to be considered simultaneously and has been argued to provide a fuller
2 account of the impact of coach climate in youth sport (Appleton et al., 2016). In support of this
3 approach, research shows that perceptions of an empowering climate are positively related to
4 indicators of athlete well-being (e.g., enjoyment and global self-worth), whereas perceptions of a
5 disempowering climate are positively related to indicators of athlete ill-being (e.g., symptoms of
6 athlete burnout and physical ill-health; Appleton & Duda, 2016).

7 In situating the construct of perfectionistic climate alongside these well-established climate
8 constructs, it is possible to identify key similarities and differences. In terms of areas of
9 commonality, both an ego-involving climate and perfectionistic climate emphasise punishment for
10 mistakes and recognition for better performers. Similarly, both a controlling climate and a
11 perfectionistic climate emphasise the controlling use of punishments and conditional reward of
12 approval. Importantly, however, perfectionistic climate focusses on a narrower set of cues and
13 behaviours that stress a requirement for perfection. For example, rather than emphasising
14 expectations that are high or even very high, perfectionistic climate emphasises expectations that are
15 rigid and unrealistic. Similarly, unlike existing climate models, perfectionistic climate emphasises a
16 particular style of criticism that is harsh, unreasonable, and unrelenting. In these regards, a
17 perfectionistic climate is altogether more extreme. If this is the case, and perfectionistic climate
18 offers an alternative to existing approaches, differences between climate measures should be evident
19 in the degree to which their components are related to each other and based on their relative positions
20 within a nomological network (e.g., Ziegler et al., 2013).

21 **The Present Study**

22 The aim of the present study was to develop a scale to measure perfectionistic climate in
23 sport – Perfectionistic Climate Questionnaire-Sport (PCQ-S). To this end, five studies are provided.
24 The first study involves initial item generation and refinement procedures. The second study
25 provides an exploratory examination of the factor structure of the PCQ-S items identified in study

1 one. The third study provides a further examination of the factor structure of PCQ-S items using
2 confirmatory and exploratory-confirmatory analyses. The fourth study provides an assessment of
3 discriminant validity by estimating the nomological network of relations between test scores on the
4 PCQ-S and established coach climate measures. The final study provides an initial exploratory
5 examination of measurement invariance across age and gender groups.

6 **Study 1**

7 The primary purpose of study one was to generate items that capture the five perfectionistic
8 climate components. The second purpose of the study was to then evaluate and refine the initial
9 PCQ-S item pool. The procedures outlined in the present study and subsequent studies were
10 approved by an institutional ethics review board. A detailed description of the approved recruitment
11 and consent procedures employed is included in the Supplemental Material.

12 **Initial Item Generation and Item Refinement**

13 Definitions of the five perfectionistic climate components were developed based on existing
14 theory (Appleton & Curran, 2016; Flett et al., 2002) and empirical research (e.g., Frost et al., 1990).
15 In line with the conceptual model of perfectionistic climate proposed by Hill and Grugan (2019),
16 each component incorporated a focus on perfect performance as opposed to high, very high, or
17 exceptionally high performance standards (e.g., "...unrealistic expectations that one should *perform*
18 *perfectly*"). A parallel list of key characteristics was also created to accompany the definition of each
19 component. In line with the recommendations of DeVellis (2017), the authors independently
20 generated items using the definitions and key characteristics of each component to guide the process.
21 This process yielded an initial pool of 127 items. These were then assessed by the authors for their
22 clarity, readability (assessed via Flesch-Kincaid grade level scores; Kincaid et al., 1975), relevance,
23 similarity to other items generated and items in existing scales, and the degree to which they adhered
24 to the specified criteria. This process led to a first revised pool of 50 items.

25 **Expert Panel Review**

1 The first revised item pool of 50 items was then subject to an external review process. The
2 first external panel consisted of five experts who were independent of the research group, had
3 worked extensively with constructs relevant to perfectionistic climate in sport (e.g., perfectionism,
4 motivational climate, and interpersonal relationships), and held both PhD degrees and senior
5 academic roles at universities in the UK or Sweden. The expert panel members were provided with a
6 document containing the definition of each perfectionistic climate component, the accompanying key
7 characteristics, and the revised pool of 50 items. In line with the recommendation of DeVellis
8 (2017), they were instructed to: (a) identify the component each item reflects; (b) rate the content
9 suitability of each item (high, moderate, or low); and (c) rate the comprehensibility of each item
10 (high, moderate, or low). The experts were also invited to provide comments regarding specific
11 items, recommend new items, and provide alternative wording for existing items.

12 The same revised pool of 50 items was also subject to an external review by a panel of seven
13 sport coaches (5 male, 2 female, $M_{age} = 43.14$ years, $s = 14.86$, range = 23-59). In this review, the
14 coaches were provided with a document containing the revised pool of items and instructed to: (a)
15 indicate whether they considered the content of each item to be applicable to the sport they coached
16 (applicable *versus* not applicable); and (b) rate the clarity of each item (high, moderate, or low).

17 Using feedback generated through the two external review panels, two members of the
18 research group carefully reviewed the feedback relating to all items before making an informed
19 decision about how to use the advice (DeVellis, 2017). This process helped to identify some
20 potentially problematic items that were addressed by deleting or re-wording items. This included
21 issues pertaining to item replication (e.g., “the coach expects flawless performances” versus “the
22 coach expects performances to be flawless”), content suitability (e.g., controlling items identified
23 that reflect internally rather than externally controlling contingencies), and clarity (e.g., items in
24 which it is not clear who the behaviour or practice is directed toward). In addition, in line with expert
25 feedback, we avoided the use of “flawless” which may be inappropriate for youth athletes and added

1 parallel items to ensure conditional regard captured negative (e.g., “the coach is less approachable
2 when ...”) internally controlling contingencies. This process led to a second revised pool of 56 items.

3 **Focus Group Review**

4 The third stage of the external review process involved conducting focus groups with three
5 independent groups of youth footballers (5 males, 15 females, $M_{age} = 14.10$ years, $s = 0.91$, range =
6 13-16). We followed common recommendations (e.g., Hennink, 2014) and assessed the readability,
7 comprehension, and content suitability of the second revised pool of 56 items (DeVellis, 2017). This
8 process was also conducted to assess the stability of item interpretation across athletes and identify
9 any items that athletes may be unwilling or unable to respond to (Collins, 2003). The feedback
10 generated across the three focus groups led to a third revised pool of 54-items.¹

11 **Study 2**

12 The primary purpose of study two was to explore the factor structure and psychometric
13 properties of the third revised pool of 54 items identified in study one. In doing so, we aimed to
14 identify a parsimonious model which: (a) adequately accounts for the correlations among the set of
15 items; and (b) incorporates factors which are interpretable. The secondary purpose of study two was
16 to assess the final set of items in relation to scale readability and reliability. To achieve these aims, a
17 convenience sample of youth athletes were invited to complete the revised pool of 54 items. This
18 development sample ($N = 487$) was split into two subsamples via block randomisation (Arifin,
19 2012). The first subsample ($n = 243$) was reserved exclusively for study two and the second
20 subsample ($n = 244$) was reserved exclusively for study three.

21 **Participants**

22 The subsample reserved for study two consisted of 243 youth athletes (110 males; 130
23 females; 3 missing; $M_{age} = 14.19$ years; $s = 1.72$; range = 10-18) who represented various individual

¹ Further details of procedures relating to initial item generation and item refinement, external panel review of items by academic experts and sport coaches, and focus group review of items by youth athletes are provided in the Supplemental Material.

1 and team sports (e.g., football, rugby, netball, athletics, and rowing) and levels of competition (e.g.,
2 recreational = 6, club = 136, county = 18, regional = 25, semi-professional = 4, professional = 11,
3 national = 17, international = 6, and missing = 20). On average, athletes had been competing in their
4 sport for 5.48 years ($s = 3.16$) and dedicated 7.30 hours ($s = 4.46$) to training and competition per
5 week. In comparison to other activities in their lives, athletes rated their sport as moderately-to-
6 extremely important ($M = 7.79$, $s = 1.51$; 1 = *extremely unimportant* to 9 = *extremely important*).

7 **Data analysis**

8 Exploratory factor analysis (EFA) was carried out with the robust variance-adjusted weighted
9 least squares estimation (WLSMV) method for categorical variables in *Mplus* 8.0 (Muthén &
10 Muthén, 1998-2017). The procedure was iterative and followed several common recommendations
11 (Fabrigar & Wegener, 2012). Factor retention was explored using eigenvalues, goodness of fit
12 statistics for competing models, and solution interpretability. This approach highlights the
13 importance of considering both relevant theory and statistical criterion when considering the
14 appropriate number of factors to retain. The factor solutions from EFA were assessed based on the
15 magnitude of parameter estimates ($\geq .30$ was considered meaningful), degree of cross loading (the
16 number of indicators loading meaningfully on more than one factor), and solution interpretability
17 (Morin et al., 2020).

18 In line with previous psychometric research (e.g., Appleton et al., 2016), we used multiple fit
19 indices to help evaluate overall model fit: chi-square statistic (χ^2), comparative fit index (CFI),
20 Tucker-Lewis index (TLI), root mean square error of approximation (RMSEA), and standardised
21 root-mean-square residual (SRMR). However, as the χ^2 is oversensitive to sample size and minor
22 model misspecifications, we focussed predominantly on the additional indices specified. In line with
23 established guidelines (Marsh et al., 2004), we considered models meeting the following criteria to
24 reflect adequate model fit: $> .90$ CFI, TLI, $< .08$ RMSEA, 90% CI $< .05$ to $< .08$; $< .08$ SRMR. To
25 aid assessment of the magnitude of factor correlations, descriptors for small ($.10 \leq |r| < .30$), medium

1 (.30 \leq $|r|$ < .50), and large ($|r| \geq .50$) effects were used (Cohen, 1988). Inter-factor correlations
2 exceeding .85 were considered to reflect poor discriminant validity (Brown, 2015).

3 In terms of readability, we assessed each item using the Flesch-Kincaid grade level score
4 method (Kincaid et al., 1975). We used this analysis to assess the approximate grade-level and
5 associated age range that each item is suitable for. For example, an item receiving a grade-level score
6 of 4.7 is typically suitable for individuals with at least a 4th grade reading level (in the UK, 9- to 10-
7 year-olds). In terms of reliability, we followed recent recommendations for estimating scale
8 reliability by estimating McDonald's omega (ω) over the more traditional Cronbach's alpha (α).
9 The ω estimates and corresponding 95% confidence intervals were calculated in *Mplus* 8.0 (see
10 Hayes & Coutts, 2020). We also examined inter-item correlations and corrected item-total
11 correlations in SPSS 25.0 (inter-item correlations $.20 < r < .70$ and item-total correlation $> .30$ were
12 used to guide assessment; Kidder & Judd, 1986).

13 **Exploratory Factor Analysis**

14 The iterative EFA procedure revealed that the most interpretable solution consisted of 20
15 items loading on five factors. This model yielded five eigenvalues with a magnitude of one or greater
16 (eigenvalues = 10.24 to 1.00) and provided significantly better fit than the alternative models
17 examined (see Table 1). In this model, well-defined factors for expectations, criticism, control,
18 conditional regard, and anxiousness were evident (see Table 2). This was demonstrated through
19 meaningful factor loadings for each item on their target factor (target $\lambda = .42$ to $.89$) and minimal
20 meaningful cross loadings (only four cases out of a possible 80). In two of these cases, items loaded
21 higher on their target factor (target $\lambda = .52$ to $.69$; non-target $\lambda = .33$ to $.35$). While the other two
22 cases provided a comparable or stronger loading on the non-target factor (target $\lambda = .44$ to $.50$; non-
23 target $\lambda = .44$ to $.54$), we decided to retain these items for further psychometric evaluation.

24 **Assessment of Scale Readability and Reliability**

25 In terms of scale readability, the Flesch-Kincaid grade level scores revealed that the 20 items

1 ranged from 4.7 (4th grade, typically suitable for 9- to 10-year-olds) to 11.2 (11th grade, typically
2 suitable for 16- to 17-year olds). Most item scores ($n = 15$ or 75%) were within 4th grade to 7th grade
3 reading ability range (typically suitable for 9- to 12-year-olds). Three items were within an 8th grade
4 reading ability range (typically suitable for 13-to 14-year-olds). The remaining two item scores were
5 9.0 (9th grade, typically suitable for 14- to 15-year-olds; item 12) and 11.2 (item six). Overall, based
6 on these scores, we believe the scale is likely to function best and be most appropriate for athletes
7 ages 12 years and older. While we arrived at this conclusion, it is important to note that we are aware
8 that two items (six and 12) may need further revision to improve readability for younger athletes
9 (e.g., athletes < 16 years). In relation to scale reliability, the ω estimates and corresponding 95%
10 confidence intervals for each PCQ-S factor identified in the EFA analysis were all acceptable:
11 expectations = .82 (95% CI range = .77 to .85); criticism = .85 (95% CI range = .81 to .88); control =
12 .84 (95% CI range = .80 to .88); conditional regard = .86 (95% CI range = .82 to .89); anxiousness =
13 .84 (95% CI range = .80 to .87). The inter-item correlations were also within recommended limits (r
14 = .37 to .70) and the corrected item-total correlations were all acceptable ($r = .51$ to .74).

15 **Study 3**

16 The primary purpose of study three was to further examine the five-factor 20-item structure
17 identified in study two. We examined three alternative measurement models using confirmatory
18 factor analysis (CFA) and exploratory structural equation modelling (ESEM) techniques
19 (Asparouhov & Muthén, 2009; Myers et al., 2011).

20 **Participants**

21 The subsample reserved for study three consisted of 244 youth athletes (114 males; 130
22 females; $M_{age} = 14.16$ years; $s = 1.72$; range = 10-18) who represented various individual and team
23 sports (e.g., cricket, basketball, rowing, golf, and tennis) and levels of competition (e.g., recreational
24 = 7, club = 138, county = 20, regional = 35, semi-professional = 6, professional = 7, national = 12,
25 international = 2, and missing = 17). On average, athletes had been competing in their sport for 5.71

1 years ($s = 3.09$) and dedicated 7.27 hours ($s = 5.91$) to training and competition per week. In
2 comparison to other activities in their lives, athletes rated their sport as moderately-to-extremely
3 important ($M = 7.71$, $s = 1.53$; 1 = *extremely unimportant* to 9 = *extremely important*).

4 **Data Analysis**

5 The first model we assessed involved using CFA to examine a structure in which: (a) items
6 were constrained to load on first-order target factors only; and (b) all latent factors were specified to
7 covary. However, as this first-order CFA specification is highly restrictive and may be inappropriate
8 for evaluating complex multidimensional structures, we also examined a second model. Specifically,
9 we used ESEM with oblique target rotation to examine a structure in which: (a) items were permitted
10 to load on all first-order factors; and (b) all latent factors were specified to covary. The third model
11 we tested was specified to explore the possibility that a hierarchical perfectionistic climate construct
12 may account for the covariation among the five first-order PCQ-S factors. This involved examining a
13 hierarchical ESEM model using ESEM-within-CFA (see Morin et al., 2020). Using this framework,
14 we examined a structure in which: (a) items were permitted to load on all first-order factors; and (b)
15 all latent first-order factors were specified to load on a second-order hierarchical perfectionistic
16 climate factor. In all analyses, we used the WLSMV estimation method for categorical variables in
17 *Mplus* 8.0 and assessed the magnitude and interpretability of parameter estimates using the
18 guidelines identified in study two. Similarly, the overall model fit of each model was evaluated using
19 the model fit indices and criteria previously identified.

20 **Alternative Measurement Models**

21 The first-order CFA model examined provided good fit to the data. In terms of parameter
22 estimates (see Table S1 in the Supplemental Material), all factor loadings were significant ($p < .001$)
23 and meaningful ($\lambda \geq .67$). In comparison to this model, the first-order ESEM model examined
24 provided improved model fit and further support for the hypothesised five-factor structure. In line
25 with the EFA results in study two, well-defined and discernible factors for expectations, criticism,

1 control, conditional regard, and anxiousness were evident (see Table 3). This was demonstrated
2 through meaningful factor loadings for each item on their target factor (target $\lambda \geq .42$) and minimal
3 meaningful cross loadings (only two cases out of a possible 80). In all cases of cross loading, items
4 loaded higher on their target factor (target $\lambda = .57$ to $.65$; non-target $\lambda = .31$ to $.40$). In the first-order
5 CFA model, standardised factor correlations were positive, significant, and large ($r = .68$ to $.84$). In
6 the first-order ESEM model, the standardised factor correlations were positive, significant, and
7 moderate-to-large in magnitude ($r = .45$ to $.71$). The hierarchical ESEM model examined also
8 provided excellent fit to the data. In terms of parameter estimates, well-defined factors for the five
9 PCQ-S components were evident. This was demonstrated through meaningful factor loadings for
10 each item on their target factor (target $\lambda \geq .42$) and minimal meaningful cross loadings (only three
11 cases out of a possible 80). In all cases of cross loading, items loaded higher on their target factor
12 (target $\lambda = .50$ to $.62$; non-target $\lambda = .30$ to $.44$). In this model, all second-order factor loadings were
13 also meaningful ($\lambda \geq .64$; see Table S2 in the Supplemental Material).

14 **Study 4**

15 The primary purpose of study four was to provide an assessment of discriminant validity by
16 estimating the nomological network of relations between test scores on the PCQ-S and established
17 coach climate measures. The secondary purpose of this study was to examine the stability of the
18 three PCQ-S measurement models highlighted and tested in study three using an independent dataset.

19 **Participants**

20 A convenience sample of 223 (183 males; 38 females; 2 missing; $M_{\text{age}} = 15.25$ years, $s =$
21 1.90 ; age range = 12–19) youth athletes recruited from sport clubs and schools in the United
22 Kingdom provided the data for study four. The participants represented various team sports (e.g.,
23 football, cricket, hockey, rugby, netball, and basketball) and levels of competition (recreational = 27,
24 club = 140, county = 18, regional = 15, semi-professional = 10, professional = 1, national = 4,
25 international = 1, and missing = 7). On average, athletes had been competing in their sport for 7.51

1 years ($s = 3.68$), dedicated 7.61 hours ($s = 4.49$) to training and competition per week, and had been
2 working with their current coach for 2.47 ($s = 2.37$) years. In comparison to other activities in their
3 lives, athletes rated their sport as moderately-to-extremely important ($M = 7.49$, $s = 1.87$; 1 =
4 *extremely unimportant* to 9 = *extremely important*).

5 **Measures**

6 **Perfectionistic Climate.** To measure youth athlete perceptions of perfectionistic coach
7 climate components, we used the five-factor 20-item model identified in the previous two studies. In
8 line with previous data collection, participants were instructed to think about the coach that they
9 spend the most time with and rate their level of agreement or disagreement with each statement using
10 a 5-point Likert scale (1 = *strongly disagree* to 5 = *strongly agree*).

11 **Empowering and Disempowering Coach Climate Dimensions.** To measure youth athlete
12 perceptions of empowering and disempowering coach climate dimensions, we used three measures.
13 The 12-item Motivational Climate Scale for Youth Sports (MCSYS; Smith et al., 2008) was used to
14 assess perceptions of task-involving (TASK) and ego-involving (EGO) cues emphasised in AGT.
15 The 15-item Controlling Coach Behaviour Scale (CCBS; Bartholomew et al., 2010) was used to
16 assess perceptions of four facets of controlling behaviour emphasised in SDT: controlling use of
17 rewards (CREW), negative conditional regard (NCR), intimidation (INT), and excessive personal
18 control (EPC). The 6-item Sport Climate Questionnaire-Short Form (SCQ-SF; Felton & Jowett,
19 2012) was used to assess perceptions of autonomy supportive behaviours (PAS) emphasised in SDT.
20 In line with the hierarchical model of coach-created motivational climate (Appleton et al., 2016), we
21 considered TASK and PAS to reflect empowering features of the coach climate and EGO, CREW,
22 NCR, INT, and EPC to reflect disempowering features of the coach climate.²

23 **Data Analysis**

² Example items for the MCSYS, CCBS, and SCQ-SF are provided in the Supplemental Material.

1 The study analyses were carried out using IBM Statistics SPSS 25.0, *Mplus* 8.0, and *R*
2 version 3.5.0. The first step of data analysis involved cleaning the raw data. Following this,
3 descriptive statistics and bivariate correlations were computed. The second step involved examining
4 an ESEM measurement model including all coach climate measures.³ In the third step, we then
5 estimated a regularised partial correlation network using *R* package *qgraph* (Epskamp et al., 2012).
6 This analysis provides a visualisation of the complex covariation between the variables under
7 investigation. In the current application of this technique, we estimated a network of coach climate
8 variables based on partial correlation coefficients (Epskamp & Fried, 2018). This means that the
9 resultant network includes nodes (circles) that represent different coach climate variables and edges
10 (lines connecting nodes together) that represent associations between variables after controlling for
11 all other variables in the network. To account for measurement error in the computation of the coach
12 climate variables, we used the extracted factor scores from the ESEM measurement model.

13 To limit the number of potentially spurious edges included in the final model, we used the
14 least absolute shrinkage and selection operator (LASSO) which limits the sum of absolute partial
15 correlation coefficients. This regularisation technique shrinks all partial correlations and constructs a
16 sparse network in which likely spurious associations are excluded (i.e., some partial correlations are
17 shrunk to exactly zero; see Epskamp & Fried, 2018). The edges not identified to be exactly zero can
18 therefore be considered sufficiently strong to be included in the estimated network (Epskamp et al.,
19 2018). This approach functions well in retrieving the true network structure and is particularly useful
20 for dealing with relatively small datasets when estimating psychological networks (Costantini et al.,
21 2015; Epskamp & Fried, 2018).

22 Once a psychological network is constructed there are several analyses that can be applied to
23 examine global structure and local patterns within the network (Costantini et al., 2015). In the current

³ The results pertaining to the data screening protocol and ESEM measurement model are reported in the Supplemental Material.

1 application, we examined: (a) the overall dimensionality of the network; and (b) the associations
2 between specific pairs of nodes. To examine network dimensionality, we employed exploratory
3 graph analysis (EGA; Golino et al., 2020) and followed the recommendations and *R* code provided
4 by Christensen et al. (2020). EGA has been found in simulation studies to yield comparable or
5 improved accuracy in identifying dimensionality in comparison to traditional factor analytic
6 techniques (Golino et al., 2020). To examine the associations between specific pairs of nodes, we
7 computed the edge weights (i.e., regularised partial correlation coefficients) of each association in
8 the estimated network. This allowed us to explore the relative strength of each retained edge weight.

9 **Descriptive Statistics, Bivariate Correlations, and Scale Reliability Estimates**

10 The descriptive statistics and bivariate correlation estimates are displayed in Table 4. The
11 bivariate correlations show that each of the five PCQ-S components shared positive relationships
12 with the disempowering coach climate subscales. The only exception to this was the non-significant
13 relationship between the criticism component of the PCQ-S and the EPC subscale of the CCBS. By
14 contrast, the five PCQ-S components shared either negative or non-significant relationships with the
15 empowering coach climate subscales. In line with previous research, the disempowering climate
16 subscales shared mainly inverse relationships with the empowering subscales examined (Appleton et
17 al., 2016). The scale reliability estimates and corresponding 95% confidence intervals for each coach
18 climate variable are also reported in Table 4 ($\omega = .68$ to $.90$).

19 **Regularised Partial Correlation Network**

20 The regularised partial correlation network is displayed in Fig. 1. The network features 12
21 nodes and 39 edges (59% of all possible associations). The LASSO regularisation technique resulted
22 in 27 of the possible 66 edges being removed (41% of all possible associations were shrunk to
23 exactly zero). The EGA identified that this network contained three dimensions. The dimensions
24 identified reflect perfectionistic (dimension one: five PCQ-S components), disempowering

1 (dimension two; five disempowering coach climate subscales), and empowering (dimension three;
2 two empowering coach climate subscales) features of the coach climate. The nodes within each
3 dimension were well connected and located closely to one another in the estimated network (see Fig.
4 1). The edge weights (i.e., regularised partial correlation coefficients) are reported in Table 4.

5 **Stability of Alternative Measurement Models**

6 The three models examined (i.e., first-order CFA, first-order ESEM, and hierarchical ESEM
7 model specifications) provided support for the stability of the 20-item five-factor PCQ-S. In line with
8 the findings from study two, all three models provided good fit to the data (see Table 1). The
9 findings were also consistent in that the first-order ESEM model provided superior model fit in
10 comparison to the first-order CFA model. In terms of parameter estimates, both models provided
11 support for the 20-item five-factor PCQ-S (see Table S1 for CFA results and Table 3 for ESEM
12 results). In terms of the ESEM parameter estimates, model support was based on the high percentage
13 of meaningful ($\lambda = > .30$) target factor loadings (95% of cases) and small percentage of meaningful
14 non-target factor loadings (only 7.5% of cases). In terms of the final model examined, we again
15 found support for a hierarchical perfectionistic climate construct accounting for the covariation
16 among the five first-order PCQ-S factors (see Table S2 in the Supplemental Material).

17 **Study Five**

18 The purpose of study five was to test the measurement invariance of the first-order five-factor
19 20-item ESEM model across meaningful subgroups defined based on gender and age. This testing
20 procedure was adopted as it provides a robust test of model generalisability and generates important
21 statistical information regarding the applicability of making gender- and age-based comparisons in
22 future research. To provide the data needed to examine measurement invariance, we created one
23 combined dataset ($N = 710$) comprised of data from the initial development sample used across
24 studies two and three ($N = 487$) and the independent sample used in study four ($N = 223$).

1 **Data Analysis**

2 In line with the approach to evaluating measurement invariance adopted by Jin (2020), we
3 examined three nested ESEM models to evaluate two aspects of measurement invariance per multi-
4 group analysis: metric invariance (i.e., invariance of factor loadings) and scalar invariance (i.e.,
5 invariance of factor loadings and thresholds). We used the WLSMV estimator and oblique target
6 rotation with delta parameterisation to analyse each model.⁴

7 Using the combined dataset, we initially examined measurement invariance across the two
8 independent samples. In line with our primary aim, we then examined measurement invariance of the
9 ESEM model across gender (male = 407; female = 298; total $N = 705$) and age (≤ 14 years = 360; \geq
10 15 years = 349; total $N = 709$) groups.⁵ In each assessment, the first stage involved examining the
11 overall fit of each specified model using the fit indices and corresponding criteria previously
12 identified. In the second stage, metric invariance and scalar invariance were evaluated by differences
13 in CFI (Δ CFI), TLI (Δ TLI), and RMSEA (Δ RMSEA) between nested models. To evaluate metric
14 invariance (i.e., metric versus configural model) and scalar invariance (i.e., scalar versus metric
15 model), we used cut-off values identified based on Jin's (2020) simulation study as an evaluative
16 guide: metric non-invariance (Δ CFI $> -.003$, Δ TLI $> -.004$, and Δ RMSEA $> +.014$) and scalar non-
17 invariance (Δ CFI $> -.001$, Δ TLI $> -.001$, and Δ RMSEA $> +.005$).

18 **Measurement Invariance**

19 The goodness of fit statistics for all invariance models are reported in Table 1. In terms of
20 measurement invariance across independent samples, the three increasingly restrictive models
21 provided a good level of fit. In terms of the specific cut-off values identified, support for metric

⁴ A detailed description of the three nested ESEM measurement invariance models is provided in the Supplemental Material (see also Jin, 2020).

⁵ In the combined dataset of 710 athletes, the youngest reported age was 10 years old and the oldest reported age was 19 years old. We therefore used ≤ 14 years and ≥ 15 years as the cut-off values to identify two groups with an equal age range representing younger versus older youth athletes.

1 invariance was provided by $\Delta RMSEA$ (+.011). By contrast, support for scalar invariance was
2 provided by ΔCFI (-.001), ΔTLI (+.001), and $\Delta RMSEA$ (-.001). The two scores that exceeded the
3 specific cut-off values for metric invariance either approximated ($\Delta CFI = -.013$) or satisfied ($\Delta TLI =$
4 $-.010$) more traditional criteria often applied in tests of measurement invariance (i.e., non-invariance
5 reflected by ΔCFI and $\Delta TLI > -.010$; Chen, 2007). While the applicability of traditional cut-off
6 values is unclear for assessing measurement invariance in this type of analysis, they have been found
7 to show some sensitivity to metric non-invariance with ESEM models that have a small or medium
8 degree of cross-loading (Jin, 2020). With this in mind, the two samples were combined for the
9 purpose of testing gender and age invariance.

10 In terms of measurement invariance across gender groups, the three increasingly restrictive
11 models provided good fit. In terms of the specific cut-off values identified, support for metric
12 invariance was provided by ΔCFI (-.003), ΔTLI (+.001), and $\Delta RMSEA$ (-.001). Similarly, support
13 for scalar invariance was provided by ΔCFI (0.000), ΔTLI (0.000), and $\Delta RMSEA$ (0.000). In sum,
14 this evidence shows that the PCQ-S model operates equivalently across the two identified gender
15 groups.

16 In terms of measurement invariance across age groups, the three increasingly restrictive
17 models provided a satisfactory level of fit. In terms of the specific cut-off values identified, support
18 for metric invariance was provided by ΔTLI (-.001) and $\Delta RMSEA$ (+.001). Similarly, support for
19 scalar invariance was provided by ΔTLI (-.001) and $\Delta RMSEA$ (+.002). The ΔCFI values used to
20 assess metric invariance (-.004) and scalar invariance (-.002) only marginally exceeded the specific
21 cut-off values identified. With this in mind, there is sufficient evidence the PCQ-S model operates
22 equivalently across the two age groups.

1 The aim of the present research was to develop a scale to measure perfectionistic climate in
2 sport. The five studies outlined above highlight key stages in the initial development and validation
3 of the PCQ-S. In stage one, outlined in study one, PCQ-S items were generated and refined. In stage
4 two, outlined in studies two and three, exploratory, confirmatory, and exploratory-confirmatory
5 analyses were used to further examine the structure of PCQ-S items. In stage three, outlined in study
6 four, discriminant validity was assessed by estimating the nomological network of relations between
7 test scores on the PCQ-S and established coach climate measures. Then, in the final stage, outlined in
8 study five, measurement invariance of the PCQ-S was examined across gender and age groups. The
9 result of this process was a five-factor 20-item scale.

10 **Key Proposals Underpinning the PCQ-S**

11 The development of the PCQ-S was underpinned and guided by several key proposals and
12 considerations. The first key proposal relevant in this regard is that we considered perfectionistic
13 climate in sport to be comprised of five distinct and interrelated components representing
14 expectations, criticism, control, conditional regard, and anxiousness. We found strong support for
15 this hypothesised five-factor structure across multiple measurement models and samples. In line with
16 previous research in sport (e.g., Appleton et al., 2016; Myers et al., 2011; Perry et al., 2015), the
17 ESEM model we examined provided the strongest support for the hypothesised first-order model
18 specification. This was reflected in better model fit and lower factor correlations relative to the
19 alternative CFA model. However, even the more restrictive CFA model provided good fit. In all, the
20 findings suggest that the PCQ-S matches the intended structure of perfectionistic climate with five
21 interrelated components focused on perceived perfectionistic coaching practices.

22 The second key proposal is that we considered unrealistic expectations and harsh criticism to
23 represent conceptually distinct constructs. While we found evidence for the construct validity of all
24 five PCQ-S factors, support for measuring expectations and criticism as independent factors is

1 particularly noteworthy. The psychometric analyses conducted across multiple measurement models
2 and analyses indicated that the two constructs are unique (Brown, 2015). This was particularly
3 evident in the ESEM model examined in study four which identified an inter-factor correlation that
4 was moderate in magnitude ($r = .43$). Therefore, rather than collapsing the two factors to create a
5 broader construct of coach pressure as has been the case in previous research (see Dunn et al., 2006),
6 the PCQ-S is the first domain specific scale in sport to capture unrealistic expectations and harsh
7 criticism separately. As such, the PCQ-S can be used to further examine the unique and interactive
8 effects of these two distinct practices in youth sport (see McArdle & Duda, 2008).

9 The third key proposal that we considered is that perfectionistic climate in sport involves
10 more than just unrealistic expectations and harsh criticism. The identified five-factor PCQ-S includes
11 the additional components of control, conditional regard, and anxiousness. This broader spectrum of
12 climate components better captures the related behaviours, practices, and relational styles central to
13 the development of perfectionism in athletes (Appleton & Curran, 2016). The most compelling
14 evidence in this regard is the strong evidence we found for a hierarchical perfectionistic climate
15 construct. Specifically, the hierarchical ESEM models examined in studies three and four suggest
16 that the five PCQ-S components can all be considered to be manifested as a result of a broader,
17 higher-order, latent perfectionistic climate factor. Researchers can therefore study the components of
18 perfectionistic climate as individual factors or statistically model them in a manner to study the
19 overall construct. This level of modelling flexibility is advantageous as it provides scope to examine
20 the relative influence of each PCQ-S factor in relation to a specified outcome or examine the broader
21 influence of an overall perfectionistic climate factor itself (see Myers et al., 2014).

22 The final key proposal is that we considered the five PCQ-S components to be conceptually
23 distinguishable from existing AGT and SDT coach climate measures. To examine this proposal, we
24 estimated a regularised partial correlation network. Network analysis offers a novel psychometric
25 approach to identifying the empirical relations between measures of constructs within an identified

1 nomological network (e.g., Ziegler et al., 2013). While this technique is common in areas of clinical,
2 personality, and health psychology, the present study offers one of the first applications of this
3 technique in sport and a means to assessing the discriminant validity of the PCQ-S. The network-
4 based analyses illustrated that while commonalities exist between the PCQ-S and existing climate
5 measures, the five PCQ-S components are empirically distinct from empowering and disempowering
6 coach climate dimensions. This evidence of discriminant validity lends support to our proposal that
7 the coaching practices captured by the PCQ-S reflect an unreasonable, harsh, and altogether more
8 extreme relational style than existing coach climate measures. Importantly, the results suggest that a
9 disempowering climate is not necessarily experienced as perfectionistic, and that a perfectionistic
10 climate is experienced differently than a climate that is disempowering.

11 **Implications of PCQ-S**

12 The application of perfectionistic climate to sport and development of the PCQ-S provides a
13 new and alternative way to study the experience of the pressure to be perfect. While the focus in
14 existing perfectionism measures is on the individual and the extent to which they display
15 perfectionistic traits, experience perfectionistic cognitions, or feel the need to appear perfect to
16 others, the PCQ-S focusses on the extent to which an environment is perfectionistic. This is an
17 important advance in this area. Notably, the construct of perfectionistic climate accounts for the
18 possibility that all athletes—regardless of their own levels of perfectionism—can experience the
19 consequences of the pressure to be perfect. In line with guiding theory, this may include the
20 internalisation of external pressure that encourages the development of personal perfectionism as
21 well as other outcomes that are particularly distressing for athletes.

22 The introduction of perfectionistic climate also builds on existing climate research to offer
23 another way of studying the social environment in sport. Against a backdrop of a reliance on a small
24 number of existing approaches (e.g., AGT), there have been calls for other approaches to be adopted

1 to better understand practices that shape different athlete experiences (Morgan, 2017). The
2 requirement to move beyond AGT, for example, is already evident in coach climate research drawing
3 upon key tenets of SDT (Mageau & Vallerand, 2003) and integrating multiple theoretical
4 frameworks (e.g., Appleton et al., 2016; Duda, 2013). This type of extension has been critical in
5 improving our understanding of the repercussions that different coach climate experiences may have
6 on athlete motivation, health, and psychological functioning (e.g., Appleton & Duda, 2016). The
7 development of the PCQ-S offers a further alternative that focuses exclusively on the practices that
8 shape environments so that athletes experience an extreme pressure to be perfect.

9 The development of the PCQ-S also has important implications as far as primary prevention
10 and intervention to help manage perfectionistic problems in sport are concerned. This includes
11 safeguarding mental health, maintaining motivation, and optimising performance. To date, research
12 on perfectionism intervention in sport has focussed on the individual and provided mixed success in
13 terms of intervention effectiveness (e.g., Donachie & Hill, 2020; Mosewich et al. 2013; Thompson et
14 al., 2011). To improve on current intervention strategies, researchers have emphasised that it is
15 important to incorporate a focus on the coaching environment (Appleton & Curran, 2016; Mosewich
16 et al., 2013). In this regard, a focus on coach education and changing current practice offers a new
17 avenue to explore how perfectionistic problems in athletes might be prevented as a routine part of
18 their experience. There is evidence in sport and other contexts that this type of intervention can be
19 effective (e.g., Braithwaite et al., 2011). Thus, the ability to measure the degree to which the social
20 environment is experienced as perfectionistic is an essential tool to help researchers and practitioners
21 assess and redress perfectionistic cues provided by the coach and test these types of interventions.

22 **Limitations and Future Research Directions**

23 The findings of the present research should be considered in relation to some limitations and
24 future research directions. Firstly, while we found strong evidence for the factorial validity of the

1 PCQ-S, researchers should continue to examine the hypothesised five-factor structure in diverse
2 samples of youth athletes. In doing so, the two items (item six and 12) that returned slightly higher
3 grade level reading scores in our assessment of readability should be closely monitored. Secondly,
4 while we found support that the PCQ-S operates equivalently across different age and gender groups,
5 it is important for researchers to examine whether this is also the case across other meaningful
6 groups. We also only examined metric and scalar invariance. Future research is therefore needed to
7 examine the invariance of structural parameters (e.g., equivalence of factor variances, covariances,
8 and means). Thirdly, in line with existing motivational climate measures, the PCQ-S provides
9 information pertinent to how athletes experience the coach climate in sport as opposed to how
10 coaches actually behave. To better understand perfectionistic climates in sport, researchers should
11 consider measuring both personality (e.g., other-oriented perfectionism) and objective behaviours of
12 the coach alongside perceptions of the perfectionistic climate. The final point relates to the structure
13 of data collected using the PCQ-S. When youth athletes are nested within groups that share the same
14 coach, it will be important for researchers to adopt a multilevel approach that examines both within-
15 and between-group variations in perceptions of the coach-created perfectionistic climate. This is
16 especially important given that climate-based constructs are inherently group-level constructs (see
17 Papaioannou et al., 2004).

18 **Conclusion**

19 The purpose of the present research was to apply the construct of perfectionistic climate to
20 sport and develop the first scale to measure it. Here, we have reported on key stages involved in the
21 development and initial validation of this new scale. Based on these analyses, the PCQ-S shows
22 evidence of good factorial validity, can be differentiated from existing coach climate measures, and
23 is reasonably invariant across different age and gender groups.

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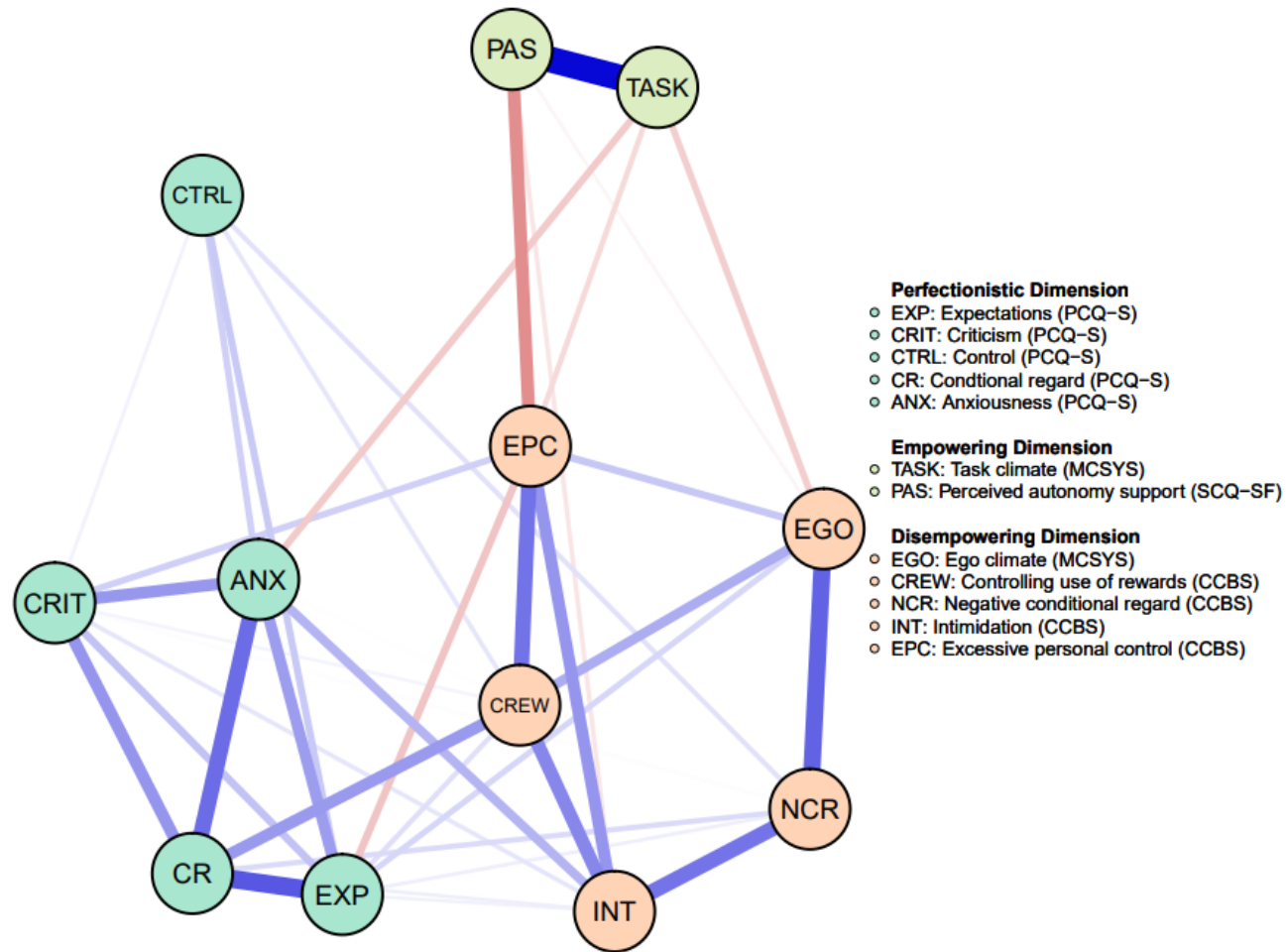
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1 Fig. 1: Regularised partial correlation network containing coach climate nodes (i.e., variables) from the Perfectionistic Climate Questionnaire-Sport (PCQ-S), Motivational
 2 Climate Scale for Youth Sports (MCSYS), Controlling Coach Behaviour Scale (CCBS), and Sport Climate Questionnaire-Short Form (SCQ-SF); Blue lines represent
 3 positive associations; Red lines represent negative associations; The width and saturation of an edge represent the strength of the association.

1 Table 1. Goodness of fit statistics for EFA (Study Two), CFA and ESEM (Studies Three and Four), and invariance (Study Five) measurement models.

	WLSMV χ^2 (<i>df</i>)	CFI	TLI	RMSEA	RMSEA 90% CI	SRMR	Δ CFI	Δ TLI	Δ RMSEA
Study Two (<i>N</i> = 243)									
EFA	152.80*** (100)	.991	.983	.047	[.031, .061]	.021	--	--	--
Study Three (<i>N</i> = 244)									
First-order CFA	415.66*** (160)	.964	.958	.081	[.071, .091]	.043	--	--	--
ESEM	179.65*** (100)	.989	.79	.057	[.043, .070]	.022	--	--	--
Hierarchical ESEM	175.13*** (105)	.990	.982	.052	[.038, .066]	.022	--	--	--
Study Four (<i>N</i> = 223)									
First-order CFA	384.12*** (160)	.947	.937	.079	[.069, .089]	.055	--	--	--
ESEM	180.14*** (100)	.981	.964	.060	[.046, .074]	.028	--	--	--
Hierarchical ESEM	171.71*** (105)	.984	.971	.053	[.038, .067]	.028	--	--	--
Study Five (<i>N</i> = 710)									
Sample Invariance (Initial Development Sample = 487; Independent Study Four Sample = 223; Total <i>N</i> = 710)									
Configural	540.61*** (260)	.984	.977	.055	[.049, .062]	.026	--	--	--
Metric	855.98*** (335)	.971	.967	.066	[.061, .072]	.040	-.013	-.010	+.011
Scalar	872.10*** (350)	.970	.968	.065	[.059, .070]	.040	-.001	+.001	-.001
Gender Invariance (Male = 407; Female = 298; Total <i>N</i> = 705)									
Configural	478.47*** (260)	.987	.981	.049	[.042, .056]	.024	--	--	--
Metric	608.49*** (335)	.984	.982	.048	[.042, .054]	.034	-.003	+.001	-.001
Scalar	628.95*** (350)	.984	.982	.048	[.042, .053]	.034	.000	.000	.000
Age Invariance (≤ 14 years = 360; ≥ 15 years = 349; Total <i>N</i> = 709)									
Configural	431.75*** (260)	.990	.985	.043	[.036, .050]	.021	--	--	--
Metric	566.86*** (335)	.986	.984	.044	[.038, .050]	.031	-.004	-.001	+.001
Scalar	609.34*** (350)	.984	.983	.046	[.040, .052]	.032	-.002	-.001	+.002

2 Note. EFA = Exploratory factor analysis; CFA= Confirmatory factor analysis; ESEM = Exploratory structural equation modelling; *df* = Degrees of freedom; CFI =

3 Comparative fit index; TLI = Tucker-Lewis index; RMSEA = Root mean square error of approximation; CI = Confidence interval; SRMR = Standardised root mean square

4 residual; ESEM models were estimated with target oblique rotation; *** $p < .001$; ** $p < .01$; * $p < .05$.

1 Table 2. Geomin rotated loadings from five-factor EFA in study two.

	<i>M</i>	<i>s</i>	F1	F2	F3	F4	F5
1. The coach expects performances to be perfect at all times.	3.20	1.16	.73* (0.10)	-.26* (.10)	.03 (0.06)	.00 (0.05)	.04 (0.06)
17. The coach expects performances to be perfect.	2.58	1.16	.73* (0.09)	.08 (0.09)	-.04 (0.06)	.01 (0.06)	.08 (0.05)
26. The coach expects performances to include no errors.	2.27	1.08	.59* (0.09)	.20* (0.08)	-.01 (0.04)	.21* (0.08)	-.08 (0.05)
29. The coach expects nothing less than perfect performance.	2.25	1.05	.70* (0.08)	.03 (0.04)	.07 (0.07)	.09 (0.07)	.07 (0.05)
12. The coach criticises even the best performances.	2.36	1.22	-.01 (0.05)	.79* (0.06)	.05 (0.06)	-.02 (0.05)	.06 (0.06)
15. The coach criticises performances that are not perfect.	2.87	1.15	.03 (0.07)	.57* (0.07)	.07 (0.07)	-.04 (0.06)	.25* (0.08)
20. The coach criticises all mistakes no matter how small.	2.55	1.12	.15* (0.08)	.69* (0.07)	-.01 (0.05)	.01 (0.05)	.11 (0.07)
34. The coach criticises performances all the time.	2.30	1.13	.07 (0.08)	.65* (0.07)	.16* (0.07)	.18* (0.07)	-.08 (0.05)
33. The coach uses his/her position unfairly to try to make performances perfect.	2.18	1.11	<u>.44* (0.07)</u>	.08 (0.05)	.44* (0.06)	.00 (0.05)	.02 (0.05)
41. The coach uses threats to try to stop mistakes in performances.	1.82	1.05	<u>.35* (0.11)</u>	-.01 (0.04)	.69* (0.07)	-.05 (0.05)	-.05 (0.05)
50. The coach uses punishment to try to make performances perfect.	2.17	1.20	.05 (0.07)	.01 (0.05)	.74* (0.07)	.05 (0.06)	.07 (0.06)
54. The coach withholds rewards if performances are not perfect.	2.14	1.11	.01 (0.05)	.01 (0.05)	.66* (0.06)	.16 (0.08)	.00 (0.05)
45. The coach is less approving when performances are not perfect.	2.28	1.10	.18 (.10)	.10 (0.07)	.25* (0.08)	.42* (0.06)	.07 (0.05)
48. The coach is friendlier when performances are perfect.	2.84	1.17	.03 (0.04)	.02 (0.03)	-.08 (0.06)	.83* (0.06)	.08 (0.05)
52. The coach is kinder when no mistakes are made when performing.	2.63	1.15	.05 (0.05)	-.09 (0.06)	.06 (0.04)	.89* (0.06)	-.05 (0.03)
53. The coach is less friendly when performances are not perfect.	2.34	1.14	-.08 (0.06)	.03 (0.04)	<u>.54* (0.07)</u>	.50* (0.08)	.06 (0.04)
6. The coach is anxious about the possibility of even small mistakes when performing.	2.24	1.05	.22* (0.09)	-.13 (0.07)	.15 (0.09)	-.00 (0.04)	.61* (0.06)
7. The coach is tense when mistakes are more likely to happen during performances.	2.67	1.09	.01 (0.06)	.01 (0.02)	.09 (0.08)	.05 (0.05)	.78* (0.05)
9. The coach is nervous that things will not go perfectly during performance.	2.51	1.14	.05 (0.06)	.11 (0.08)	-.05 (0.06)	-.01 (0.05)	.77* (0.06)
11. The coach is concerned about mistakes during performance.	2.94	1.08	-.05 (0.08)	<u>.33* (0.08)</u>	-.02 (0.05)	.15* (0.07)	.52* (0.08)
Geomin Factor Correlations			F1	.63*	.58*	.56*	.51*
			F2		.48*	.48*	.39*
			F3			.46*	.43*
			F4				.50*

2 *Note.* Study two data ($N = 243$); Bold typeface denotes meaningful loading ($\geq .30$) on target factor; Underlined typeface denotes meaningful cross-loading ($\geq .30$) on non-target factor; F1 =

3 Expectations; F2 = Criticism; F3 = Control; F4 = Conditional regard; F5 = Anxiousness; Standard errors reported in parentheses; * $p < .05$.

1 Table 3. Standardised model results and factor correlations for the ESEM models in studies three and four.

Item	<i>M</i>	<i>s</i>	Expectations	Criticism	Control	Conditional Regard	Anxiousness
1	3.34	1.14	.63*** (0.09) / .79*** (0.06)	-.03 (0.08) / -.05 (0.05)	.04 (0.07) / -.07 (0.06)	-.01 (0.08) / .11 (0.06)	.08 (0.06) / .01 (0.06)
17	2.67	1.15	.60*** (0.09) / .77*** (0.05)	-.00 (0.08) / .14** (0.04)	.12* (0.05) / .12** (0.04)	.12* (0.06) / -.07 (0.05)	.12* (0.05) / .09 (0.05)
26	2.42	1.08	.80*** (0.08) / .62*** (0.07)	.05 (0.07) / .14* (0.06)	-.01 (0.05) / .27*** (0.04)	.05 (0.05) / .01 (0.06)	-.01 (0.05) / -.02 (0.06)
29	2.77	1.28	.76*** (0.08) / .84*** (0.05)	.10 (0.07) / .08 (0.04)	.03 (0.05) / -.06 (0.04)	.06 (0.05) / .09* (0.04)	-.02 (0.04) / -.03 (0.05)
12	2.82	1.13	.09 (0.07) / -.11 (0.06)	.90*** (0.09) / .78*** (0.06)	-.25*** (0.06) / -.03 (0.05)	-.03 (0.05) / -.09 (0.06)	.04 (0.04) / .07 (0.06)
15	3.12	1.08	.12 (0.07) / .20** (0.06)	.62*** (0.08) / .59*** (0.07)	-.04 (0.07) / <u>-.30*** (0.06)</u>	.18*** (0.05) / .23*** (0.06)	.07 (0.05) / .17** (0.05)
20	2.55	1.11	.04 (0.06) / <u>.38*** (0.06)</u>	.63*** (0.08) / .39*** (0.07)	.17** (0.06) / .05 (0.07)	-.10 (0.05) / .02 (0.06)	.13** (0.04) / .06 (0.06)
34	2.44	1.08	-.13* (0.06) / .12* (0.06)	.65*** (0.07) / .59*** (0.07)	<u>.40*** (0.05) / .33*** (0.05)</u>	.01 (0.05) / -.10 (0.06)	.01 (0.04) / -.02 (0.06)
33	1.97	1.02	.29*** (0.06) / .23** (0.07)	.17** (0.06) / -.05 (0.07)	.49*** (0.06) / .46*** (0.07)	-.07 (0.06) / -.03 (0.08)	.07 (0.05) / .18** (0.06)
41	1.90	1.14	.00 (0.07) / .03 (0.06)	-.02 (0.06) / -.01 (0.06)	.75*** (0.07) / .58*** (0.07)	.04 (0.06) / .02 (0.07)	.18*** (0.05) / <u>.30*** (0.06)</u>
50	2.41	1.26	-.03 (0.08) / .04 (0.05)	.20** (0.07) / .06 (0.05)	.57*** (0.07) / .75*** (0.07)	.23*** (0.05) / <u>.30*** (0.05)</u>	-.02 (0.06) / -.13* (0.05)
54	2.44	1.02	.19* (0.08) / .17* (0.07)	.08 (0.06) / -.05 (0.07)	.42*** (0.06) / .14 (0.07)	.25*** (0.06) / .22** (0.08)	.06 (0.05) / .21** (0.07)
45	3.06	1.05	.13 (0.07) / -.01 (0.06)	.25*** (0.05) / .02 (0.04)	.17*** (0.05) / -.01 (0.06)	.46*** (0.06) / .79*** (0.07)	-.05 (0.05) / .10* (0.05)
48	3.58	1.15	-.08 (0.06) / .21** (0.07)	-.01 (0.05) / -.16** (0.06)	-.13** (0.04) / .00 (0.06)	1.00*** (0.06) / .68*** (0.07)	.03 (0.04) / .04 (0.07)
52	3.30	1.16	.10 (0.06) / -.01 (0.06)	-.14* (0.05) / .01 (0.05)	.04 (0.05) / .10 (0.06)	.70*** (0.06) / .81*** (0.06)	.22*** (0.04) / -.11* (0.05)
53	2.62	1.09	.19** (0.06) / -.04 (0.05)	.05 (0.06) / .09 (0.05)	<u>.31*** (0.04) / .24*** (0.05)</u>	.57*** (0.05) / .60*** (0.06)	-.06 (0.04) / .16** (0.06)
6	2.48	.98	.16* (0.06) / -.07 (0.06)	-.00 (0.06) / .17** (0.06)	.06 (0.05) / .22*** (0.06)	-.13** (0.05) / .15* (0.07)	.74*** (0.06) / .50*** (0.08)
7	2.92	.99	-.05 (0.06) / .11 (0.07)	-.06 (0.05) / .28*** (0.07)	.10* (0.05) / -.05 (0.07)	-.04 (0.05) / .24** (0.07)	.94*** (0.05) / .31*** (0.07)
9	2.38	.94	-.05 (0.07) / .02 (0.05)	.06 (0.06) / -.23*** (0.06)	-.06 (0.05) / .00 (0.06)	.13* (0.05) / -.11* (0.05)	.79*** (0.05) / .90*** (0.09)
11	2.94	.96	-.01 (0.09) / -.05 (0.06)	.26*** (0.07) / <u>.44*** (0.06)</u>	-.08 (0.07) / .10 (0.06)	.19** (0.06) / .12 (0.06)	.48*** (0.06) / .34*** (0.08)
Factor Correlations			F1				
			F2	.71 / .43			
			F3	.53 / .34	.55 / .36		
			F4	.62 / .59	.61 / .42	.47 / .40	
			F5	.59 / .46	.51 / .35	.45 / .39	.60 / .45

2 Note. Study three data (*N* = 244) to left; Study four data (*N* = 223) to right; Item mean (*M*) and standard deviation (*s*) values are for study four data; Bold typeface denotes meaningful loading (\geq

3 .30) on target factor; Underlined typeface denotes meaningful cross-loading (\geq .30) on non-target factor; F1 = Expectations; F2 = Criticism; F3 = Control; F4 = Conditional regard; F5 =

4 Anxiousness; Standard errors reported in parentheses; *** $p < .001$; ** $p < .01$; * $p < .05$.

1 Table 4: Descriptive statistics, bivariate correlations, edge weights, and reliability estimates

Subscale (Instrument)	<i>M</i>	<i>s</i>	ω	ω 95% CI	1.	2.	3.	4.	5.	6.	7.	8.	9.	10.	11.	12.
1. EXP (PCQ-S)	2.79	.98	.88	[.85, .90]		.10	.09	.28	.16	--	.07	.05	.03	.02	-.10	--
2. CRIT (PCQ-S)	2.73	.81	.74	[.68, .80]	.57***		.02	.17	.17	--	--	.02	.00	.04	.08	--
3. CTRL (PCQ-S)	2.15	.81	.74	[.68, .79]	.51***	.45***		--	.08	--	--	.04	.05	--	--	--
4. CR (PCQ-S)	3.12	.90	.84	[.80, .87]	.57***	.45***	.56***		.24	--	--	.17	.06	.03	--	--
5. ANX (PCQ-S)	2.67	.71	.74	[.67, .79]	.49***	.53***	.54***	.55***		-.09	--	.00	--	.12	--	--
6. TASK (MCSYS)	4.15	.60	.75	[.69, .81]	-.10	-.11	-.29***	-.11	-.11		-.08	--	--	--	-.06	.41
7. EGO (MCSYS)	2.47	.72	.68	[.54, .75]	.47***	.38***	.40***	.43***	.37***	-.23**		.14	.26	--	.09	-.02
8. CREW (CCBS)	3.50	1.24	.80	[.73, .85]	.40***	.25***	.45***	.35***	.34***	.03	.45***		--	.20	.23	--
9. NCR (CCBS)	3.23	1.29	.86	[.82, .89]	.34***	.39***	.41***	.46***	.41***	-.23**	.60***	.55***		.23	--	--
10. INT (CCBS)	2.63	1.29	.82	[.77, .86]	.25***	.31***	.54***	.32***	.38***	-.27***	.48***	.43***	.64***		.17	-.04
11. EPC (CCBS)	2.24	1.32	.86	[.81, .90]	.15*	.10	.38***	.16*	.16*	-.30***	.37***	.38***	.52***	.64***		-.19
12. PAS (SCQ-SF)	5.15	1.08	.90	[.87, .92]	-.07	.01	-.26***	-.21**	-.14*	.55***	-.21**	-.04	-.30***	-.30***	-.39***	

2 *Note.* Values below the diagonal are bivariate correlations; Values above the diagonal are edge weights (regularised partial correlation coefficients) from the regularised partial correlation
3 network analysis (see Figure 1); Missing values (--) indicates the LASSO estimation shrunk the partial correlation to exactly "0"; PCQ-S = Perfectionistic Climate Questionnaire-Sport; EXP =
4 Expectations; CRIT = Criticism; CTRL = Control; CR = Conditional regard; ANX = Anxiousness; MCSYS = Motivational Climate Scale for Youth Sports; Task = Task-involving climate;
5 EGO = Ego-involving climate; CCBS = Controlling Coach Behaviour Scale; CREW = Controlling use of rewards; NCR = Negative conditional Regard; INT = Intimidation; EPC = Excessive
6 personal control; SCQ-SF = Sport Climate Questionnaire; PAS = Perceived autonomy support; *N* = 211; **p* < .05; ***p* < .01; ****p* < .001.