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(Re)examining the effects of athlete brand image (ABI) on psychological commitment: an empirical investigation using structural equation modelling (SEM) and fuzzy set qualitative comparative analysis (fsQCA)

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ABSTRACT

Research question: How does Athlete Brand Image (ABI) affect psychological commitment (PC) when operationalised at the dimension- (attribute-) level, and measured using reflective indicators? Previous studies operationalise ABI at a higher-order construct level, and/or measure ABI using formative measures. Such operationalisations obscure potentially different relationships between ABI's image attributes and PC.

Research methods: A questionnaire was used to collect data from 197 UK respondents over a six-day period within two weeks of the Rio 2016 Olympics concluding. Data were analysed through structural equation modelling (SEM) and fuzzy set Qualitative Comparative Analysis (fsQCA) techniques.

Results and findings: Through SEM, the ABI attributes, *athletic expertise*, *life story*, *role model*, and *competition style* are positively related to PC, *sportsmanship* and *symbol* are negatively related, and *rivalry*, *physical attractiveness*, *body fit*, and *relationship effort* are nonsignificantly related. Many structural paths between ABI's attributes and PC are also significantly different. Through fsQCA, high PC exists under three complex ABI attribute configurations, while it is absent under four complex configurations.

Implications: Theoretically, finding different relationships between ABI's attributes and PC highlights why operationalising ABI at the dimension-level provides a more in-depth

understanding of athlete brand image's effects on PC. Managerially, the findings suggest athletes need only a subset of ABI attributes for high PC. Subsequently, managers can more-efficiently and effectively direct resources towards those attributes that best-capitalise on athletes' images.

Keywords: *Athlete Brand Image (ABI); operationalisation; structural equation modelling (SEM); fuzzy set Qualitative Comparative Analysis (fsQCA); reflective measurement*

Introduction

The professional sports arena is an intensely commercial industry, where competition from rivals (other athletes/clubs and other forms of entertainment) for audience viewing and spending is fierce (Ross, 2007). In this context, branding – the process by which organisations distinguish themselves from competition – plays an increasingly key role in sport management (Tsiotsou, Alexandris, & Cornwell, 2014), because it helps organisations stand-out from the crowd. Indeed, both academic and practitioner attention towards branding strategies ‘is warranted as customer participation and loyalty are vital at nearly every level of professional sport’ (Pifer, Mak, Bae, & Zhang, 2015, p.88). One way organisations brand themselves is through promoting professional athletes (Carlson & Donovan, 2013). Professional athletes are among the most valuable assets sport organisations have (Hodge & Walker, 2015), and are considered popular cultural products (Summers & Morgan, 2008) that can be branded in their own right (Thomson, 2006). Thus, in the highly competitive industry that is sport, managing athletes’ brands is becoming an essential task in enhancing both athletes’ (Arai, Ko, & Kaplanidou, 2013), and sport organisations’ (Gladden & Funk, 2002) marketing value.

Athlete brand management is a relatively new field of academic research, which has received little attention to date (Hodge & Walker, 2015). Brand management typically aims to create, develop or enhance brand image (Park, Jaworski, & MacInnis, 1986). Within this context, athlete brand image (ABI), defined as people’s perceptions of athlete brand attributes (Arai, Ko, & Ross, 2014), purportedly drives consumers’ attitudes towards athlete brand extensions (Walsh & Williams, 2017), identification with athletes’ teams (Carlson & Donovan, 2013), event and merchandise consumption (Braunstein & Zhang, 2005), loyalty towards endorsed firms (Bush, Martin, & Bush, 2004), donation intentions towards athletes’ causes (Kim & Walker, 2013) and psychological commitment (PC) towards athlete brands

(Arai et al., 2013). The ABI-PC relationship is particularly noteworthy because it suggests ABI affects athletes' marketing value through psychological commitment. That is, PC is generically linked to a brand's value (e.g. its unit price, marketing costs, and number of units sold) (Park, MacInnis, & Priester, 2014). Within athlete brand contexts, PC is connected to people's attitudes towards athletes (see Sassenberg, 2015), which is itself positively related to people's behavioural responses (e.g. following/purchasing) towards both athletes and athlete-endorsed products (cf. Lee & Park, 2014; McCormick, 2018).

Despite ABI's reported effects on consumers, scant literature explicitly focuses on the ABI construct (Arai et al., 2013). Arai and colleagues (2013; 2014) provide a first understanding of brand image of athletes, and propose a measurement instrument to capture ABI, including ten attributes. While some of these attributes (namely *athletic expertise*, *rivalry*, *physical attractiveness*, *life story*, and *role model*) are found in team contexts (Gladden & Funk, 2001; 2002; Ross, James, and Vargas, 2006), five athlete-specific attributes were also identified (namely, *sportsmanship*, *competition style*, *symbol*, *body fit*, and *relationship effort*). In turn, and following image-literature convention (e.g. Bachleda, Fakhar, & Elouazzani, 2016; Kunkel, Funk, & Lock, 2017), Arai and colleagues treated ABI as one, higher-order construct (that was subsequently linked to PC). However, recent classical measurement theory submits that multi-faceted constructs should be operationalised at the dimension-level, rather than at the higher-order construct-level: 'This approach is conceptually clean, and potentially adds richness to the research model' (Lee & Cadogan, 2013, p. 246). Hence, it appears our understanding of the ABI-PC relationship would be richer and more fine-grained if we operationalised ABI at its dimension (attribute) level, as specific and potentially different relationships between each ABI dimension and PC may emerge (e.g. it could be *sportsmanship* and *role model* are more strongly-related to PC than *physical attractiveness*, a finding that would not be identified with a holistic ABI measure).

Hence, this study's key research objective is to examine ABI's effects on psychological commitment towards an athlete when ABI is operationalised at the dimension-level.

Second, and given our limited knowledge of ABI's effects at the dimension-level, little is known about which ABI attribute combinations produce high PC. In team contexts, off-field attributes have an especially large impact on team commitment (cf. Bauer, Stokburger-Sauer, & Exler, 2008). Yet, it is unknown whether this holds for athlete contexts too. In practice, it is likely different ABI attribute combinations are equivalently relevant for PC because '(a) outcomes of interest rarely have any single cause, (b) causes rarely operate in isolation from each other, and (c) a specific causal attribute may have different and even opposite effects depending on context' (Greckhamer, Misangyi, Elms, & Lacey, 2008, p. 697). Hence, one subset of ABI attributes may be as important for PC as another subset, or certain ABI attributes may be beneficial on some occasions but detrimental on other occasions. Traditional regression-based analyses such as structural equation modelling (SEM) cannot examine such possibilities because their foundations are built on examining each antecedent's relationship with an outcome *after* other antecedents are controlled for (Woodside, 2013).

Conversely, qualitative comparative analysis (QCA) has different epistemological foundations. QCA is a partially inductive-oriented approach (Marx & van Hootegem, 2007) which attempts to explain whether configurations of causal conditions are necessary or sufficient for an outcome to occur (Vis, 2012): 'Necessary conditions are simple or complex causal recipes that are found in all instances of the outcome occurring; sufficiency conditions, however, are those conditions wherever they occur, the outcome is present and when the sufficiency condition does not occur, instances of the outcome condition both occur and do not occur' (Chang, Tseng, & Woodside, 2013, p. 96).

QCA's use within sport management (e.g. Clausen et al., 2018; Lucidarme, Cardon, & Willem, 2016; Winand, Rihoux, Robinson, & Zintz, 2013) and general brand settings (e.g. Mühlbacher, Raies, Grohs, & Koll, 2016) is growing but its use within ABI contexts is conspicuous by its absence. Yet, QCA seems pertinent to ABI contexts, given athletes are unlikely to concurrently possess all image attributes. That is, QCA can unearth whether or when ABI attributes are beneficial, detrimental or irrelevant to PC, when considered in conjunction with other ABI attributes. Hence, this study's second objective is to uncover different ABI attribute configurations that lead to high PC.

In achieving these objectives, two main theoretical contributions are made. First, through SEM, we demonstrate not all ABI attributes significantly relate to people's PC towards an athlete, and that, importantly, some are negatively related to PC. Significant differences also exist between many of these attributes' path-strengths. Such findings are not apparent if ABI is operationalised (and analysed) at its construct level. Second, through QCA, we demonstrate different bundles of ABI attributes produce equifinal routes to high PC. Moreover, minor variations in the composition of ABI attributes are the difference between high PC's presence and absence. Together, the complementarity of SEM's and QCA's findings is an additional contribution to sport management. That is, SEM enables researchers to examine relationship strengths between variables, while QCA allows researchers to observe different configurations of the same variables that lead to a specific outcome (Urueña & Hidalgo, 2016). Thus, despite differing epistemological foundations, the 'pragmatic scholar might argue that these differences...shed a distinct but hopefully complementary light on the research topic at hand' (Vis, 2012, p. 175). Subsequently, a more-rounded understanding of the linkages between ABI and PC materialises. Managerially, the results suggest athletes' representatives (e.g. agents and managers) should focus their resources on a

subset of ABI attributes if high PC is to materialise, rather than unnecessarily wasting resources trying to enhance (perceptions of) all ABI attributes.

The remainder of this paper is as follows. First, literature pertaining to *operationalising* ‘image’ in sport management is outlined, and subsequently applied to athlete brand image (ABI). Next, apposite psychological commitment literature is offered. Following this, our study’s methodology is summarised before our SEM results are produced. We then introduce our specific QCA approach before presenting these findings. A general discussion then precedes managerial implications and theoretical contributions. The paper concludes with limitations highlighting future research directions.

Literature Review

Operationalising Image in Sport Management

Operationalisation is defined as ‘the transition from theory to measurement’ (Hui & Triandis, 1985, p. 134). In this context, two operationalisation decisions need further review. First, sport management scholars *model* ‘image’ as either a single, first-order (e.g. Lohneiss & Hill, 2014), or as a higher-order (e.g. Gladden & Funk, 2002), construct. Second, sport management scholars *measure* ‘image’ as reflective (e.g. Gladden & Funk, 2001), formative (e.g. Kunkel et al., 2017), or both reflective and formative (e.g. Braunstein & Zhang, 2005) in nature. Often, scholars intertwine these two operationalisation decisions. For instance, when operationalising second-order constructs, scholars historically judged ‘(a) a first-order construct can have either formative or reflective indicators, and (b) those first-order constructs can, themselves, be either formative or reflective indicators of an underlying second-order construct’ (Jarvis, MacKenzie, & Podsakoff, 2003, p. 204). Subsequently,

researchers measured their respective constructs following this viewpoint. However, recent literature has cast doubt on such positions. First,

‘[a] higher-order measurement model that is truly unidimensional and conforms to the reflective measurement model is entirely redundant. As a result, unidimensional higher-order measurement models should be modelled as first-order measurement models. [Meanwhile, if] a higher-order measurement model is not unidimensional... the lower-order constructs should be treated as separate variables...’, because they may have different antecedents and outcomes (Lee & Cadogan, 2013, p. 246).

This suggests ABI should be modelled at the first-order (dimension) level. Conversely, Arai and colleagues (2013; 2014) grouped together the on-field attributes *athletic expertise* (defined as *an athlete’s achievement and athletic capability*), *competition style* (*specific characteristics of an athlete’s performance in a competition*), *rivalry* (*an athlete’s competitive relationship with other athletes*), and *sportsmanship* (*an athlete’s virtuous behaviour in terms of respect for the game, opponents, and teammates*) to produce the higher-order factor, ‘*athletic performance*’; while the off-field attributes *physical attractiveness* (*an athlete’s physical qualities and characteristics that spectators find aesthetically pleasing*), *symbol* (*an athlete’s attractive personal style or fashion*), and *body fit* (*how physically fit an athlete is for his or her sport*) made up the higher-order factor, ‘*attractive appearance*’, and *life story* (*the extent to which an athlete has an appealing, interesting off-field life story*), *role model* (*an athlete’s ethical behaviour that society has determined is worth emulating*) and *relationship effort* (*an athlete’s positive attitude toward interaction with fans, spectators, sponsors and media*) produced the higher-order factor, ‘*marketable lifestyle*’¹. These three higher-order factors then produced ‘athlete brand image’ (that linked to PC). Consequently,

¹Arai and colleagues name the ten attributes, ‘subdimensions’, and the three higher order factors, ‘dimensions’.

Arai et al. (2013) may have masked potentially different relationships between ABI's attributes and PC.

Second, formative measurement is labelled a fallacy (Edwards, 2011) because researchers cannot know how formative (latent) variables vary (Lee & Cadogan, 2013). Indeed, ontologically-speaking, 'the only plausible notion of measurement is a realist one, which assumes the existence of real attributes that cause variation in measurement devices' i.e. reflective measurement (Lee & Chamberlain 2016, p 106). Hence, irrespective of researchers' (reflective or formative) modelling approaches, measurement theory asserts measurement can only be reflective in nature (Markus & Borsboom, 2013) so researchers should use reflective indicators wherever possible (Lee & Chamberlain 2016). It appears ABI's respective dimensions should be measured accordingly. However, some of Arai and colleagues' original items suggest formative indicators. For example, two of Arai et al.'s (2013) original *competition style* scale items appear to capture 'distinctiveness' ('*The athlete's competition style is distinctive from other players*') and 'excitement' ('*The athlete's competition style is exciting to watch*'). Clearly 'distinctiveness' and 'excitement' are not reflective of a unidimensional measure. Again, this raises questions concerning the link between ABI's attributes and PC.

Psychological Commitment (PC)

According to Mahony, Madrigal, & Howard (2000), psychological commitment (PC) reflects consumers' attitudinal loyalty, which subsequently predicts behavioural loyalty. Hence, PC is linked to current and future consumption behaviours (Park et al., 2014). Within sport and recreational settings, PC is associated with people's ongoing attraction to a particular sport (e.g. Bee & Havitz, 2010), which is manifested in, for example, continued sport event

attendance, season ticket procurement, and higher levels of sports participation (see e.g. Inoue, Funk, & McDonald, 2017; Tokuyama & Greenwell, 2011). Similarly, PC helps explain people's resistance to changing recreational agencies (e.g. Iwasaki & Havitz, 2004). Consequently, capturing PC is particularly useful for marketers because it helps segment markets and facilitate predictions about individuals' behaviours (e.g. Mahony et al., 2000).

Study Overview

The Athlete Brand Image Construct

In light of the above discussion, this study investigates how ABI affects PC when ABI is operationalised at the dimension (attribute) level and measured using reflective indicators. We also consider whether the ten ABI attribute path-strengths are significantly different to each other. We use Arai and colleagues' (Arai et al., 2013; Arai et al., 2014) ABI construct because their conceptualisation provides 'the first comprehensive conceptual framework of athlete brand image' (Arai et al., 2014, p. 97), implying the construct is seminal to future ABI literature. Meanwhile, we capture PC because (a) of its connection to athletes' marketing values and (b) it allows us to contrast our ABI-PC findings against Arai et al. (2013) – if divergent results materialise we begin to demonstrate how operationalising ABI at the dimension-level can provide richer and more fine-grained knowledge. Figure I outlines the conceptual framework.

FIGURE I ABOUT HERE

Methodology

For comparison purposes, we closely follow Arai et al.'s (2013) research design as far as possible. First, we purposively selected a number of high-profile athletes to include in the questionnaire: Andy Murray (tennis), Mohamed Farah (track), Jessica Ennis-Hill (track), Laura Trott (cycling), Jason Kenny (cycling), and Nicola Adams (boxing) were chosen because they had each won at least one gold medal during the London 2012 Olympics, meaning they received substantial media attention in their home-country during the build-up, and throughout, the Rio 2016 Olympics. Further, each athlete won at least a silver medal during the Rio 2016 Olympics, ensuring media coverage continued after the Games had finished (when data collection commenced). This media coverage provided a plethora of opportunities for the public to create and reinforce perceptions of athlete brand attributes. Winning medals at two Olympic Games also meant these athletes demonstrated 'a sustained level of consistently superior athletic performance', meaning they are more likely to be product endorsers or spokespeople than less established athletes (Stone, Joseph, & Jones, 2003, p. 96). Consequently, any findings should have enhanced ecological validity. In line with Arai et al. (2013), respondents answered ABI questions on the athlete they were most familiar with.

Wherever possible, the ten image attribute measures came from Arai et al. (2013). However, whenever Arai and colleague did not use reflective indicators we followed their definitions to find related and established literature to ensure our dimensions were captured with reflective indicators. This approach aligns with the original scholars, who suggested their scale needed further modification and improvement (Arai et al., 2013). Seven-point Likert-type scales captured responses for all constructs. Table I outlines the retained scale items.

TABLE I ABOUT HERE

Protocol and debriefing analysis with five university students unaware of the study's motivation ensured the research instrument was comprehensible. Data were collected from 197 UK respondents over a six-day period within two weeks of the Rio 2016 Olympic Games concluding. The total response-rate was 67.2% (54.3% males, 43.1% female, 2.5% did not/preferred to not answer; mean age=29.4 years, s.d.=11.2 years). 56.9% of respondents identified tennis star, Andy Murray, as the most familiar athlete, followed by Mohamed Farah (18.3%), Jessica Ennis-Hill (15.2%), Laura Trott (4.6%), Jason Kenny (4.1%), and Nicola Adams (1.0%). Potential common method variance (CMV) concerns were alleviated via procedures outlined in Podsakoff, Mackenzie, Lee, & Podsakoff (2003). Procedurally, respondents were informed that there were no right or wrong answers, that responses were anonymous, and that they should answer questions as honestly as possible. Statistically, a structural model unmeasured latent method factor test was nonsignificant ($\Delta\chi^2(33)=41.472$), and no relationships changed. Hence, CMV is of little-to-no concern.

Structural equation modelling

Data were analysed using Lisrel 8.71, which took measurement error into account. Following standard theory-trimming procedures, confirmatory factor analysis (CFA) suggested the model fitted the data well overall: $\chi^2(409)=556.197$, RMSEA=.0372, SRMR=.0530, with GFI, NNFI and CFI all reaching or exceeding .9 (factor loadings and error variances are found in Table I). Meanwhile, construct reliabilities ranged from .80 to .95. Finally, average variances extracted (AVEs) ranged from .568 to .861, while the square root of the lowest AVE (.753, *life story*) was higher than the highest correlation (.639, between *athletic expertise* and *sportsmanship*). Thus, discriminant validity is considered upheld (Fornell & Larcker, 1981).

Next, SEM assessed structural paths from each of the ten ABI dimensions to PC. This effectively produced a saturated structural model (i.e. one where all paths are assumed to be nonzero), and thus is equivalent to a CFA with identical fit indices (Williams & O’Boyle, 2011). Structural path analysis indicates the standardised paths of *athletic expertise* ($\gamma=.295$, $p=.005$), *life story* ($\gamma=.489$, $p<.001$), and *role model* ($\gamma=.279$, $p=.008$) are each significantly and positively related to PC, while *competition style* ($\gamma=.135$, $p=.084$) is marginally significant and positively related to PC. Conversely, *sportsmanship* ($\gamma=-.262$, $p=.026$) is significantly and negatively related to PC, while *symbol* ($\gamma=-.146$, $p=.096$) is marginally significant and negatively related to PC. Finally, *rivalry* ($\gamma=-.001$, $p=.986$), *physical attractiveness* ($\gamma=.058$, $p=.512$), *body fit* ($\gamma=-.005$, $p=.951$), and *relationship effort* ($\gamma=.067$, $p=.425$) are nonsignificantly related to PC (all two-tailed).

To test whether significant differences exist between any of the ten attribute-paths to PC, we examined the significance of $\Delta\chi^2$ ($\Delta 1$ degree of freedom) when the two paths under investigation were constrained to be equal and remaining paths were fixed to their original regression-path estimates. Using a Bonferroni-corrected α -level of .001 (calculated by dividing the traditional .05 α -level by 45 observations), results suggest the path from (a) *life story* to PC is significantly different from the paths of *physical attractiveness* ($\Delta\chi^2=17.846$), *rivalry* ($\Delta\chi^2=21.723$), *sportsmanship* ($\Delta\chi^2=36.415$), *symbol* ($\Delta\chi^2=20.755$), and *body fitness* ($\Delta\chi^2=14.681$); (b) *athletic expertise* to PC is significantly different from the paths of *physical attractiveness* ($\Delta\chi^2=15.029$), *rivalry* ($\Delta\chi^2=12.801$), *sportsmanship* ($\Delta\chi^2=10.812$), and *symbol* ($\Delta\chi^2=24.083$); (c) *role model* to PC is significantly different from the paths of *symbol* ($\Delta\chi^2=15.417$) and *sportsmanship* ($\Delta\chi^2=15.894$), and (d) *sportsmanship* to PC is also significantly different from the paths of *physical attractiveness* ($\Delta\chi^2=12.056$) and *competition style* ($\Delta\chi^2=13.289$). The structural model explains 44.1% of PC.

The SEM results suggest when ABI is operationalised as ten separate dimensions not all of ABI's attributes are positively – or even significantly – related to PC. This is in sharp contrast to Arai et al. (2013) who treated ABI as one higher-order construct. Moreover, a number of dimensions have significantly different path-strengths to PC. Finally, and unlike team contexts (e.g. Bauer et al., 2008), the findings indicate some off-field attributes (in this case *physical attractiveness*, *body fit*, and *relationship effort*) may not be important for commitment by themselves. In conclusion, the results illustrate the importance of analysing ABI at the dimension-level.

That said, it is imprudent to disregard nonsignificant ABI dimension paths, especially as some of these attributes originate from Arai et al.'s (2013) inductive word-elicitation task. This exercise brought-to-mind people's most salient athlete attributes, suggesting these attributes are important for people's PC towards athletes. Moreover, the highest correlation between variables was below 0.70, implying asymmetric relations exist, and different configurations of independent variables may produce the same outcome (Urueña & Hidalgo, 2016). Hence, a more judicious approach is to assume each of the ten dimensions is important for PC under certain qualitatively-different circumstances. Pertinent to this paper, we suspect different ABI dimension subsets link to PC. Qualitative Comparative Analysis (QCA) enables us to investigate this possibility.

Qualitative Comparative Analysis (QCA)

Qualitative Comparative Analysis (QCA) is historically a qualitatively-rooted approach but survey and/or purely quantitative data are increasingly analysed through QCA techniques (see Wagemann, Buche, & Siewert, 2016). QCA is built on set-theoretic methods, which (a) work with membership scores of cases in sets, (b) perceive relations between social

phenomena as set relations, and (c) interpret relations in terms of necessary and sufficiency conditions (Schneider & Wagemann, 2013). Simple conditions (also known as configurations, causes, solutions or recipes) are analogous to independent, singular antecedents in SEM whereas complex conditions involve examining two or more antecedents together (Woodside, 2015). In this context, a single ABI dimension is a simple configuration while complex configurations incorporate two or more ABI dimensions.

Two main QCA techniques exist²: crisp set QCA (csQCA) and fuzzy set QCA (fsQCA). csQCA regards cases are either full members or full non-members of a set, while fsQCA assumes membership gradation. For example, cases can be completely in, more in than out, neither in nor out, more out than in, or completely out of a set (Schneider & Wagemann, 2013). Theoretically, these ‘fuzzy’ qualitative breakpoints avoid unnecessary information-loss which would otherwise occur if data were regarded as only ‘in’ or ‘out’ of a set (Tóth, Thiesbrummel, Henneberg, & Naudé, 2015). In practical terms, we captured people’s responses using 7-point Likert-type scales, which ‘let respondents make qualitative statements of agreement, disagreement and indifference’ (Emmenegger, Schraff, & Walter, 2014, p. 3), thereby complementing fsQCA’s membership gradation. For example, ‘neither in nor out’ cases correspond with ‘indifference’ responses. Due to this mapping, fsQCA is utilised in this study.

It is imperative researchers offer full disclosure of their (fs)QCA procedures so that studies are judged against minimum QCA standards (Wagemann et al., 2016). Moreover, as (fs)QCA is relatively new to sport management, a comprehensive overview of our steps taken is provided from a multitude of sources, so as to facilitate QCA’s advancement in sport management scholarship. First, as our measures are reflective in nature, it is imperative

²Multi-value QCA (mvQCA) also exists. However, there is debate surrounding its set-theoretic status. csQCA and fsQCA are also regarded as the main QCA approaches (see e.g. Schneider & Wagemann, 2013).

discriminant validity and acceptable construct reliabilities are upheld, which we demonstrate through CFA, as per Chang et al. (2013). Next, the 10 ABI dimensions' and PC's test scores should be calibrated, and fuzzy scores created. Calibration is a core activity of QCA (Wagemann et al., 2016), and involves demarcating an entity's qualitatively-different states, which are themselves, potentially important contexts for other conditions (Ragin 2008b). Our calibration procedures followed Leischnig and Kasper-Brauer (2015) and Rauch, Dekker, and Woodside (2015). First, we averaged each construct's multi-item measures to create single-item scores (SISs) for each respondent's ratings (Leischnig & Kasper-Brauer, 2015). As some SISs were not integers, we considered membership to be (increasingly more) 'out' when SIS was (decreasingly) less than 3.5, (increasingly more) 'in' when SIS was 4.5 or (increasingly) higher, and 'neither in nor out' (the crossover point) when $3.5 \leq \text{SIS} < 4.5$. Utilising these threshold boundaries allowed us to further distinguish between cases, which is itself an important consideration (see e.g. Chang et al., 2013). Subsequently, we created fuzzy scores by transforming each SIS to fall between 0 and 1, following Rauch et al. (2015), and used the same calibrations for PC and the 10 ABI dimensions.

Necessity

Once fuzzy scores are created, each simple configuration's necessity is assessed through its consistency, which represents the extent to which a target set is a subset of a configuration. Our target sets are (a) high PC present and (b) high PC absent. It is important to concomitantly investigate high PC's absence because 'the explanation of the absence of the outcome (i.e. the negation of the phenomenon under analysis)...[is not in general] directly derived from the explanation of the presence of the outcome' (Wagemann et al., 2016, p. 2533).

A simple configuration is considered necessary if its consistency is 0.9 or higher (e.g. Tóth et al., 2015). We find high *athlete expertise* (consistency score=0.9944), *competition style* (0.9105), *sportsmanship* (0.9908), *body fit* (0.9888), and *relationship effort* (0.9796), at first glance appear to be simple, necessary solutions for high PC's presence. Moreover, these simple configurations' coverages range from 0.4778 (*body fit*) to 0.6243 (*competition style*), indicating they are empirically relevant. However, high *athletic expertise* (0.9642), *sportsmanship* (0.9446), *body fit*, and *relationship effort* (0.9082) also appear to be simple necessary solutions for high PC's absence. When the same simple necessary configurations appear in both the presence and absence of the same outcome they are considered trivial and removed from further analysis. This is not to say these trivial configurations are unimportant for PC per se, but rather they cannot meaningfully explain differences between high PC's presence and absence in our study (see e.g. Lucidarme et al., 2016). Subsequently, we consider these attributes to be hygiene image attributes (i.e. necessarily needed but not necessarily PC-enhancing). This means only *competition style* is deemed a simple necessary solution for high PC's presence because it was not a necessary condition for high PC's absence (please see the online supplemental table for all attributes' consistency and coverage scores).

Sufficiency

Sufficiency should always be assessed after necessity (Wagemann et al., 2016). The first step towards identifying sufficiency is through a truth table (Tóth et al., 2015), which consists of 2^n configurations, where n represents the number of unique, simple configurations. Some configurations are found empirically in the data, while others exist logically but lack empirical instances. The latter are called logical remainders, and often occur in empirical studies.

Next, if simple necessary configurations are found, ‘truth table rows (no matter if existing ones or logical remainders) that do not show this condition can be automatically excluded’ from further consideration (Wagemann et al., 2016, p. 2535). Relating this to our study, high PC’s presence can only meaningfully exist when high *competition style* is present in truth table rows. Thus, any rows suggesting high *competition style*’s absence leads to high PC (either empirically or through logical remainders) should be removed from the truth table. Following this, researchers must choose an appropriate frequency threshold. This is the minimum level researchers accept configurations are empirically (ir)relevant and is usually based on a study’s number of empirical instances (Tóth et al., 2015). We chose a threshold of ‘2’ to increase empirical relevance (Ragin, 2008a). Researchers must also choose an appropriate sufficiency consistency threshold, which now represents the extent to which a configuration is a consistent subset of a target set. A sufficiency consistency threshold should be at least 0.75 or results are too inconsistent (e.g. Ragin, 2008b). We applied Ragin’s (2008b) threshold. We note that when we created our truth table for high PC’s (a) presence and (b) absence, contradictory truth table rows were found (i.e. the same configurations exist in both truth tables). This is quite normal in applied QCA and can be dealt with at the final analysis stage (e.g. Schneider & Wagemann, 2013). The truth tables for high PC’s presence/absence after necessity, sufficiency and frequency thresholds are accounted for are found in a supplemental table online.

The final stage undertaken before complex sufficient configurations are identified involves logical minimisation using a Boolean Algebra algorithm based on counter-factual analysis. The algorithm removes simple, conflicting configurations from otherwise alike, complex configurations. The minimisation process is informed by both the truth table and scholars’ handling of logical remainders. Logical remainders, which are also necessary conditions, are treated as present when a target set is present. Hence, high *competition style*

logical remainders are treated as present when high PC exists. Additionally, when ‘a cause is necessary for an outcome, then its negation is sufficient for the negation of the outcome’ (Ragin, 2000, footnote 4, p. 332). Hence, *competition style* logical remainders are treated as absent when high PC is absent³. Finally, logical remainders of all other ABI attributes are treated as being either present or absent when the target set is present. In this paper, fsQCA 2.5 software (see Ragin, 2008a) is used for counter-factual analysis, which provides complex, intermediate, and parsimonious solutions. The intermediate solution is usually the most interpretable (Ragin, 2008a), not least because necessary conditions are retained during the minimisation process (Ren, Tsai, & Eisingerich, 2016). We therefore report intermediate solutions.

The solution’s overall consistency for high PC’s presence is 0.7849, surpassing Ragin’s (2008a) 0.75 threshold. Meanwhile, the overall solution coverage, which ‘represents exploratory power and has a meaning similar to R-square values in regression analyses’ (Zacharias, Nijssen, & Stock, 2016, p. 4127), is 0.7146. This indicates the equifinal solutions account for 71.46% of cases where high PC is present.

High PC exists under three equifinal complex configurations: (a) high *competition style*’s, *rivalry*’s, *physical attractiveness*’s, *symbol*’s, and *life story*’s presence (‘High PC Present: Row 1’ in Table II); (b) high *competition style*’s, *physical attractiveness*’, *symbol*’s, and *role model*’s presence in conjunction with high *rivalry*’s and *life story*’s absence (‘High PC Present: Row 2’); and (c) high *competition style*’s, *rivalry*’s, *life story*’s, and *role model*’s presence in conjunction with high *physical attractiveness*’ and *symbol*’s absence (‘High PC

³Following the logic of Wagemann et al. (2016) and Ragin (2000) we would have removed truth table rows (i.e. before the final analysis stage) contradicting this sufficiency condition for high PC’s absence if they had existed empirically. However, whenever empirical cases of *competition style*’s absence occurred, empirical cases of high PC’s absence also occurred.

Present: Row 3'). The consistencies of the complex configurations are above 0.75, meaning they are deemed sufficient configurations (see Table II).

Meanwhile, the overall solution consistency for high PC's absence is 0.8150, and the overall coverage is 0.7208. Here, four equifinal complex configurations are found: (a) high *rivalry's*, *physical attractiveness*, *symbol's*, and *life story's* presence ('High PC Absent: Row 1' in Table II); (b) high *physical attractiveness*, *symbol's*, and *role model's* presence in conjunction with high *rivalry's* and *life story's* absence ('High PC Absent: Row 2'); (c) high *competition style's*, *rivalry's*, and *role model's* presence in conjunction with high *physical attractiveness* and *symbol's* absence ('High PC Absent: Row 3'); (d) high *rivalry's* and *life story's* presence in conjunction with high *competition style's*, *physical attractiveness*, *symbol's*, and *role model's* absence ('High PC Absent: Row 4'). The third solution leading to high PC's absence (i.e. 'High PC Absent: Row 3') is regarded as a necessary complex configuration due to its consistency level exceeding 0.9 and its raw coverage exceeding 0.5 (e.g. Mühlbacher et al., 2016). The other three configurations are deemed sufficient complex configurations (see Table II).

TABLE II ABOUT HERE

FsQCA Overview

Examining the complex configurations of high PC's presence and absence uncovers vital differences. For example, high *competition style* is always present when high PC exists (expectedly so given its necessity status). However, when high PC is absent, high *competition style* may be present, absent or even irrelevant depending upon other ABI attributes it is conjugated with. Similarly, high *role model* is never absent when high PC exists. Instead, it is present on two occasions and irrelevant on a third occasion. Conversely, high *role model's*

absence is found in one complex configuration for low PC. More interesting, a conjugation of high *competition style*, *rivalry*, *physical attractiveness*, *symbol*, and *life story* lead to high PC. However, if high *competition style* is absent from this configuration, high PC is absent. Likewise, high *life story* forms part of a sufficient configuration for high PC's presence but this dimension is absent from an otherwise exact configuration for high PC's absence. These results suggest subtle differences in complex configurations are the difference between high PC's presence and absence.

General Discussion

Overall, this study finds ABI's attributes differentially affect PC. Specifically, the SEM analysis shows PC is positively determined by *athletic expertise*, *life story*, *role model*, and *competition style*, negatively determined by *sportsmanship* and *symbol*, and nonsignificantly determined by *rivalry*, *physical attractiveness*, *body fit*, and *relationship effort*. Meanwhile, the fsQCA analysis demonstrates which ABI attributes are present, absent, irrelevant, or even trivial (hygiene attributes) for high PC. These findings are only noticeable because ABI is operationalised at the dimension-level.

The SEM and fsQCA findings are complementary (Vis, 2012). For example, net of other variables, the on-field attribute, *competition style* displays a positive relationship with PC, while high *competition style* is present when high PC is present. Similarly, the off-field attribute, *role model* is positively related to PC net of other variables, while high *role model* is present in two of three configurations (and irrelevant for the other) when high PC exists. Together, these findings generally suggest an athlete needs to be (perceived as) a role model and have a distinct competition style if high PC is to materialise. Furthermore, if these two attributes are considered alongside *athletic expertise* and *life story* (both of which were

significantly related to PC during SEM but trivially linked to high PC during fsQCA), the two analyses findings suggest a combination of these on-field and off-field attributes have at least some importance for (increasingly) high PC towards athlete brands.

Conversely, and given *sportsmanship* is negatively related to PC net of other variables (SEM) and high *sportsmanship* is trivially present when high PC exists (fsQCA), it appears individual athletes are expected to have a high baseline level of *sportsmanship*. Indeed, the manifest variables' mean for *sportsmanship*'s was 6.271. Even so, the negative association between *sportsmanship* and PC is still somewhat surprising. Perhaps this relationship arises from people's patriotism which is particularly salient during Olympic Games periods. Stated differently, once other ABI attributes are accounted for, perhaps people slightly prefer athletes with 'winning at all costs' mentalities relative to 'wholehearted sportsmanship' mentalities, despite what they may openly express in person. Patriotism may also explain why *symbol* was negatively linked to PC. For example, some people may believe Olympians should focus their efforts on winning medals rather than wasting time being *symbol* icons. Indeed, this belief may be especially true for UK Olympic medallists/medal hopefuls given they receive public funding. While we can only speculate on these relationships – particularly given investigations into people's perceptions of these newly-identified image attributes are limited – that we found them reinforces the importance of operationalising ABI at the dimension-level.

Managerial Implications

Together, the knowledge generated through SEM and fsQCA can help practitioners (e.g. athletes, agencies, and brand managers) become more efficient when directing resources towards improving (perceptions of) athletes' image attributes, thereby more effectively

capitalising on athletes' images. Specifically, the findings suggest practitioners should focus resources on respective ABI attributes that substantially enhance PC. In particular, the results indicate athlete brand managers should try to influence people's perceptions of athletes' distinct competition styles (an 'on-field' attribute). Concurrently, managers should complement this by exploring other ABI attributes athletes are currently most associated with, and then cross-reference these with the ABI configurations that lead to high PC's presence. Incongruities between athletes' perceived attributes and attributes found in the equifinal configurations (Table II) can then be targeted. For example, if people perceive an athlete has no *competition style*, *role model*, *physical attractiveness*, and *symbol* yet possesses high levels of all other attributes (i.e. 'High PC Absent: Row 4'), resources should only be directed towards enhancing and/or communicating the athlete's *competition style* and *role model*. High levels of these two attributes should convert people to high PC, despite *physical attractiveness* and *symbol* remaining absent (i.e. 'High PC Present: Row 3'). Indeed, assuming all other ABI attributes remain stable, further increases in an athlete's *competition style* and *role model* ought to generate even higher PC, given these attributes were also significant explanatory factors of PC (i.e. SEM). Subsequently, a higher marketing value (e.g. greater fanbase, and more lucrative endorsement opportunities) should follow. Similarly, if people perceive an athlete lacks *physical attractiveness* and *symbol* and consider their *life story* irrelevant, while simultaneously perceiving high levels of all other attributes exist (High PC Absent: Row 3), a natural and instinctive managerial decision may be to improve people's *physical attractiveness* and *symbol* perceptions. However, managers would be advised to concentrate solely on making the athlete's *life story* relevant – not only should high PC materialise (i.e. High PC Present: Row 3) but PC should continue to increase as *life story* does.

Theoretical Contributions

This study makes two primary theoretical contributions to sport management. Additionally, the use of both SEM and fsQCA techniques together provide a deeper and more-rounded understanding of ABI's effect on PC.

First, most ABI literature operationalises 'image' at the higher-order, construct-level (e.g. Arai et al. 2013). Our results highlight the benefits gained from operationalising the construct at the dimension-level. In particular, our theoretical contribution lies in demonstrating how some ABI dimensions are positively related to PC while others are negatively or nonsignificantly related. Further, and unlike team-based contexts (e.g. Bauer et al., 2008), we find some off-field attributes (in this case *relationship effort*, *body fit*, and *physical attractiveness*) are not important for PC net of other attributes' contributions, while *symbol* appears to be negatively impactful. Finally, we demonstrate that numerous ABI path-strengths to PC are significantly different from one another. These findings are not apparent if ABI is operationalised at the construct-level. Hence, the study's results support recent measurement theory contentions that higher-order, multidimensional constructs should be examined at the dimension-level (e.g. Lee & Cadogan, 2013). Hence, scholars should – as a minimum – consider operationalising constructs such as ABI at the dimension-level.

Our second theoretical contribution lies in fsQCA's utilisation within ABI contexts. Here, we find high PC occurs with three equifinal sufficient complex ABI configurations, but is absent when three equifinal sufficient complex configurations, and one necessary complex configuration, exist. Moreover, all ABI attributes play some role for high PC's existence in at least one configuration, whether that be trivially (*athletic expertise*, *sportsmanship*, *body fit*, *relationship effort*) or differentially (*competition style*, *rivalry*, *physical attractiveness*, *symbol*, *life story*, *role model*). Further, some ABI dimensions experience contrarian

conditions (i.e. the dimension is present in one complex configuration but absent in another). Hence, the findings adhere to Greckhamer et al.'s (2008) assertions that multiple causes of an outcome exist, individual causes rarely operate in isolation, and some causes may have opposing effects depending upon their respective conjugation.

Finally, by looking at both SEM and (fs)QCA results together, the findings suggest deeper insights are gained when both analysis techniques are used (Vis, 2012). That is, SEM helps scholars analyse how two or more independent variables affect an outcome while QCA informs scholars about intersections among the same independent variables that lead to the same outcome (Urueña & Hidalgo, 2016). Specifically, in addition to understanding which ABI dimensions are significant explanatory factors of PC (i.e. SEM), scholars can further understand which (non)significant antecedent conditions may still be important when analysed in conjunction with other antecedents (i.e. QCA). Conversely, had other regression models been analysed alongside SEM it is unlikely few if any further insights would emerge. To demonstrate this we reanalysed our data using (a) binary logistic regression (bLR) to mimic conditions' presence/absence, and (b) multinomial logistic regression (mLR) to – conceptually at least – mimic fsQCA's gradation (with our 'crossover point' acting as the baseline category). The bLR analysis suggests significantly more variance is explained than when no predictors are included [$\Delta\chi^2(10)=41.234, p<.001; R^2=.189$ (Cox & Snell), .264 (Nagelkerke)]. Subsequently, the individual parameter estimates suggest *role model* ($b=.478, p=.018, \text{Odds Ratio}=1.613$) and *life story* ($b=.385, p=.008, \text{Odds Ratio}=1.469$) significantly predict high PC. Meanwhile, including the ten ABI dimensions in the mLR model resulted in significantly more variance explained than in the baseline model [$\Delta\chi^2(20)=78.209, p<.001; R^2=.328$ (Cox & Snell), .369 (Nagelkerke)], and the Pearson ($p=.728$) and Deviance ($p=.754$) statistics were nonsignificant. Subsequently, the individual parameter estimates suggest *sportsmanship* ($b=.874, p=.014, \text{Odds Ratio}=2.396$) significantly predicts high PC, while *role*

model ($b=.441, p=.054, \text{Odds Ratio}=1.555$) is marginally significant. Conversely, *expertise* ($b=-.684, p=.048, \text{Odds Ratio}=.505$), *sportsmanship* ($b=.926, p=.003, \text{Odds Ratio}=2.524$), *relationship effort* ($b=-.658, p=.009, \text{Odds Ratio}=.518$), and *life story* ($b=-.641, p=.001, \text{Odds Ratio}=.527$) are significant predictors of high PC's absence.

We argue these results provide few – if any – additional insights compared to our fsQCA findings, not least because they tell us little about how combinations of nonsignificantly-related ABI attributes could produce high PC. In fact, the bLR results simply offer two significant predictors that were already found to be significant through SEM (i.e. *role model* and *life story*). Meanwhile, some mLR findings (e.g. *expertise*, *life story*) contradict our SEM results. Hence, both practitioners and scholars would be left wondering which direction to follow if SEM's and mLR's results were to be followed. Instead, (fs)QCA provides 'theoretically, empirically, and substantively highly informative' findings even if nonsignificant findings materialise from regression analyses (Grofman & Schneider, 2009, p. 666)⁴.

Limitations and Future Research

As with any study, a number of limitations exist. Building upon these will generate further knowledge in this area and begin the inductive process towards external validity. First, we used successful double-Olympians in our study. While this likely increases ecological validity, it may also reduce the dispersion of on-field attributes, such as *athletic expertise* and *sportsmanship*. Consequently, future research should consider using up-and-coming or close-

⁴We also reanalysed the data through bLR and mLR using only the ABI dimensions and the associated second-order interactions retained in the final fsQCA stage. For bLR, few significant predictors were found and some odds ratios appeared implausibly high. Similarly, few significant relationships were found in mLR and some odds ratios were questionable. This demonstrates the high demands placed on data when so many interactions are considered (Schneider & Wagemann, 2013). We do not report these results because we would only really know which ABI dimensions to include in the logistics regression models *after* fsQCA was undertaken (even with painstakingly looking for parsimony). In turn, this strengthens our view that fsQCA and SEM are complementary.

to-retiring athletes, and/or athletes with known unsportsmanlike behaviour. Second, following Arai and colleagues, we focused on ABI's attributes and PC. However, respondents' individual differences may also be important (i.e. boundary conditions). For example, a competitive person might embrace athletes' rivalries but an uncompetitive person might prefer athletes who display fellowship. Third, we measured ABI's dimensions using reflective indicators, following measurement theory. However, it could be argued that Arai and colleagues measured *some* of their dimensions in a formative manner (e.g. *competition style*). If some original ABI dimensions are formative in nature they would contain sub-dimensions, and these could be captured using reflective indicators too. We did not go to this sub-dimension level because we wanted to minimise respondent fatigue, and believe our reflective items appropriately capture the ABI definitions. Moreover, we suggest our findings are more robust and interpretable at the dimension-level compared to the sub-dimension-level, which substantially increases the number of conditions (see Wagemann et al., 2016). However, future research may consider balancing these issues against the potential knowledge generated from measuring ABI's sub-dimensions. Finally, we used 7-point Likert-type scales, which complements fsQCA's gradation membership (Emmenegger et al., 2014). However, Likert-type scales cannot capture aetiological antecedents of individuals' ratings. Further, we could not triangulate specific individual's responses with in-depth case analysis because we followed Podsakoff et al.'s (2003) recommendation of respondent anonymity for response-rate enhancement. Hence, while our QCA technique is perfectly acceptable (see e.g. Wagemann et al., 2016), future studies may consider combining qualitative and quantitative data, either to triangulate findings, or as an alternative approach to uncovering respective conditions. For example, qualitative interviews informed Winand et al.'s (2013) choice of simple configurations while quantitative data informed their performance outcome. Similarly, Clausen et al. (2018) used financial statements to inform their outcome-choice while simple

configurations were deduced from triangulating literature, documents, and interviews. That said, in studies such as ours, researchers must avoid activating socially desirable responses if they are also collecting qualitative data as this could mask otherwise unexpected findings, which then lead to future research questions (e.g. in our study, ‘why is *sportsmanship*’s relationship with PC negative?’). More broadly, the pragmatic scholar who utilises regression-based and QCA-based techniques must carefully balance each approach’s respective epistemological foundations in future studies.

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Table I: Items, Factor Loadings and Error Variances

Items and sources	Completely standardised loadings	Error variances
Athletic Expertise (Arai et al., 2013; Netemeyer & Bearden, 1992) $\bar{x}(\sigma)=6.602 (0.736)$		
When I think about the athlete's expertise in their sport, the athlete is...		
...very knowledgeable in their sport	.719	.483
...an expert in their sport	.940	.116
...accomplished in their sport	.694	.518
Competition Style (Arai et al., 2013; Carlson & Donovan, 2013) $\bar{x}(\sigma)=4.956 (1.132)$		
The athlete's competition style is distinctive from other players	.760	.422
The athlete is unique compared with other athletes	.710	.496
The athlete has unique competitive skills	.811	.342
Rivalry (Ross et al., 2006; Kilduff et al., 2010; Mael & Ashforth, 1992) $\bar{x}(\sigma)=4.841 (1.400)$		
When this person competes I think...		
...they see their fellow competitors as big rivals	.741	.451
...there is intensive rivalry between this athlete and their competitors	.853	.272
...they see their fellow competitors as big enemies	.653	.573
Sportsmanship (Arai et al., 2013) $\bar{x}(\sigma)=6.271 (0.929)$		
When I think about sportsmanship, the athlete shows...		
...sportsmanship in competition	.790	.376
...respect for their fellow sports competitors	.878	.229
...fair play	.878	.229
Physical Attractiveness (Arai et al., 2013) $\bar{x}(\sigma)=4.250 (1.535)$		
The athlete is physically attractive	.940	.116
The athlete is physically good looking	.887	.213
Physically the athlete is beautiful	.838	.298
Symbol (Arai et al., 2013) $\bar{x}(\sigma)=3.937 (1.252)$		
I imagine the athlete's private fashion (i.e. outside of when they compete) to be...		
...trendy	.880	.225
...stylish	.952	.095
... "in vogue"	.634	.598
Body Fit (Arai et al., 2013; Braunstein & Zhang, 2005) $\bar{x}(\sigma)=6.618 (0.753)$		
When I think of the physical fitness levels needed to compete at an elite level in this athlete's sport, I would say the...		
...athlete is in good shape	.948	.101
...athlete is physically fit	.928	.139
...athlete's body is well conditioned	.879	.228
Life Story (Arai et al., 2013; Braunstein & Zhang, 2005) $\bar{x}(\sigma)=3.919 (1.349)$		
Unique episodes in this athlete's personal life make them more	.722	.478

interesting to follow	.847	.282
This athlete's personal life is interesting to follow	.681	.537
This athlete's personal life increases interest in them		
Role Model* (Cadogan et al., 2009)		
$\bar{x}(\sigma)=5.368 (1.070)$		
This athlete shows good ethical standards in society	.943	.110
Being ethical in society is important for this athlete	.563	.683
Relationship Effort (MacKenzie & Lutz, 1989)		
$\bar{x}(\sigma)=6.014 (1.070)$		
When this athlete interacts with fans, their attitude appears...		
...positive	.928	.139
...good	.952	.093
...favourable	.903	.185
Psychological Commitment (Stokburger-Sauer et al., 2012)		
$\bar{x}(\sigma)=3.702 (1.565)$		
I am very fond of this athlete	.750	.438
I am very committed to this athlete	.903	.184
I consider myself to be very loyal to this athlete	.916	.161

All variables used fully-labelled 7-point Likert-type scales (anchored 'Strongly Disagree'/'Strongly Agree')

*two indicators are perfectly acceptable for this study (see e.g. Lee, Cornwell, & Babiak, 2012)

Table II:ABI Dimension Patterns in Presence/Absence of High PC

		ABI Patterns								
		Competition Style	Rivalry	Physical Attractiveness	Symbol	Life Story	Role Model	Raw Coverage	Unique Coverage	Consistency
High PC Present	Row 1	+	+	+	+	+		0.5738	0.1350	0.8062
	Row 2	+	-	+	+	-	+	0.4070	0.0613	0.7811
	Row 3	+	+	-	-	+	+	0.4550	0.0795	0.8118
	Number of +	3	2	2	2	2	1	<i>sol_cov = 0.7146</i>		
	Number of -	0	1	1	1	1	1	<i>sol_con = 0.7849</i>		
	Number of empty cells	0	0	0	0	0	1			
High PC Absent	Row 1		+	+	+	+		0.4978	0.0720	0.8415
	Row 2		-	+	+	-	+	0.3973	0.0473	0.9220
	Row 3	+	+	-	-		+	0.5225	0.1296	0.9017
	Row 4	-	+	-	-	+	-	0.2490	0.0090	0.9375
	Number of +	1	3	2	2	2	2	<i>sol_cov = 0.7208</i>		
	Number of -	1	1	2	2	1	1	<i>sol_con = 0.8150</i>		
	Number of empty cells	2	0	0	0	1	1			

Figures correct to 4 d.p.; *sol_cov*=solution coverage; *sol_con*=solution consistency; +/- indicates presence/absence of high (levels of) attribute; empty cells indicate simple irrelevant conditions.

Figure I: Conceptual Framework

