

**Coopetition in innovation ecosystems: A comparative analysis of knowledge transfer
configurations**

Emily Bacon

Michael D. Williams

Gareth Davies

School of Management

Swansea University

Bay Campus

Swansea

SA1 8EN

UK

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Abstract

The effective management of knowledge exchange is critical for open innovation in ecosystem contexts where organizations may partner with potential competitors. This study contributes to existing knowledge by detecting the conditions for knowledge transfer success between both cooperative and non-competitive ecosystem partners. The study uses a qualitative approach. Semi-structured interviews were conducted with 20 stakeholders across multi-industry ecosystems to compare the presence of knowledge transfer conditions between competitors and non-competitors. Through fuzzy-set qualitative comparative analysis (fsQCA), configurational recipes of conditions were identified, revealing the distinct configurations required of either cooperative or non-competitive partnerships in the context of innovation ecosystems. The findings show the need for organizations to tailor knowledge exchange practices to the competitive nature of each relationship. Notable theoretical and practical implications are provided for ecosystem stakeholders that engage in cooperative partnerships to develop innovations.

Keywords: Coopetition; Ecosystem; Open innovation; FsQCA; Knowledge transfer

1. Introduction

In an increasingly competitive, interconnected economy, in-house research and development (R&D) activities are being replaced by more open approaches to developing innovations (Chesbrough, 2003; Simpson, 2019). Firms increasingly pursue partnerships with external knowledge sources that possess more advanced expertise. Such sources may consist of potential competitors, whose similarities make them more suitable for sharing complementary knowledge assets (Gnyawali & Park, 2009). It is argued that organizational innovativeness relies on such partnerships, which involve simultaneous collaboration and competition, amalgamated to form “coopetition” (Bouncken & Kraus, 2013). Identified as a “secret of success” (Stockdale, 2015), coopetitive partnerships increase innovative capacity (Leitão & Pereira, 2015; Ribeiro-Soriano, Roig-Tierno, & Mas-Tur, 2016; Ritala & Hurmelinna-Laukkanen, 2013), with some innovations proving not to be feasible without coopetition (Baierl, Anokhin, & Grichnik, 2016). Innovation ecosystems offer multiple opportunities for knowledge exchange with both non-competitive and competitive partners. Comprising startups, small businesses, universities, and multinational corporations (Carayannis & Campbell, 2009), ecosystems are characterized by coopetition (Selander, Henfridsson, & Svahn, 2010) and occupy the core of open innovation activities (Bogers, Chesbrough, & Moedas, 2018).

Effective knowledge sharing is an important component of coopetitive ecosystem partnerships. It adds value to both partner organizations (Chin, Chan, & Lam, 2008) and is a critical factor in maintaining such relationships (Chua, 2003). However, knowledge exchange in coopetitive partnerships entails numerous complexities (Bouncken, Gast, Kraus, & Bogers, 2015) necessitating efficient management of the process (Baumard, 2009). Gast, Filser, Gundolf, and Kraus (2015) argue that mechanisms should be implemented to enable and sustain the sharing of necessary knowledge in coopetitive partnerships, acting as a preventive

measure for unwanted imitation and knowledge leakage. Thus, successful knowledge management is vital in contexts where competitors leverage the skills and resources of interorganizational partners (Oxley & Sampson, 2004). Despite the importance of this process, few studies have examined effective knowledge transfer for cooperative partners in innovation ecosystems (Bengtsson & Raza-Ullah, 2016; Charleton, Gnyawali, & Galavan, 2018; Czachon & Mucha-Kuś, 2014; Dorn, Schweiger, & Albers, 2016), justifying the need for a more comprehensive examination of these domains. Moreover, while the notion of cooperation has received widespread attention in recent years (Baierl et al., 2016; Bengtsson & Kock, 2014; Bouncken & Kraus, 2013; Yami, Castaldo, Dagnino, & Le Roy, 2010), scholars have noted that cooperation research is fragmented and limited, with some research areas being explored in depth yet others ignored (Ritala, Kraus, & Bouncken, 2016).

This study remedies the absence of such research by making the following contributions. First, because few analyses have adequately tackled the question of how open innovation fosters cooperative benefits (Bouncken et al., 2015), this research contributes to the growing body of literature on cooperation in open innovation ecosystems. Second, although 50% of all new alliances are between competitors (Bengtsson, Raza-Ullah, & Vanyushyn, 2016; Harbison & Pekar, 1998), the remaining proportion of interorganizational partnerships are non-competitive. Because these partnership types exhibit unique characteristics (Gnyawali & Charleton, 2018), detecting symmetrical distinctions in the knowledge transfer practices of these partners is necessary to further deconstruct how success can be achieved. Moreover, while research has begun to provide greater insights into interorganizational knowledge management in cooperation, further work is needed to ascertain the relationship-level and firm-level components of knowledge management for cooperation (Tidström, 2018). In view of this requirement, this study uses fuzzy-set qualitative comparative analysis (fsQCA) to detect how knowledge-, firm-, and relationship-

level characteristics contribute to knowledge transfer success in both cooperative and non-competitive partnerships. The application of fsQCA in this study constitutes a key contribution, following recommendations by Devece, Ribeiro-Soriano, and Palacios-Marqués (2019) and Ritala et al. (2016) to apply such novel analytical techniques to cooperation research. Doing so enables a comparative analysis of the conditions of knowledge transfer success in cooperative versus non-competitive partnerships, contributing to the cooperation literature by addressing the absence of this form of analysis. Accordingly, the following research question is posed: *How do the conditions for knowledge transfer success differ in cooperative versus non-competitive ecosystem partnerships?*

The remainder of this paper is structured as follows. Section 2 reviews the literature underlying this research. Section 3 presents the theoretical framework. Section 4 describes the sample, materials, and methods used in this study and justifies the application of fsQCA. Section 5 discusses the results of the fsQCA procedure and examines the main findings. Theoretical and managerial implications, along with study limitations and recommendations for future research, are provided in Section 6.

2. Literature review

Combining “simultaneous competition and cooperation,” cooperation involves partnerships between firms with “value creation intent” (Gnyawali & Charleton, 2018, p.2513). Many definitions emphasize the combination of the terms “cooperation” and “competition” (Bengtsson & Kock, 2000; Fernandez, Chiambaretto, Le Roy, & Czakon, 2018; Leitão & Pereira, 2015; Yami & Nemeh, 2014). The fact that the terms are harmonious and can occur concomitantly is stressed (Ritala et al., 2016). Although they are frequently merged, cooperation and competition entail separate logics. Cooperation entails the pursuit of collective interests and benefits, whereas competition involves private interests and

opportunistic behavior (Bengtsson et al., 2016). Competitors are identified as organizational actors who produce and market the same products (Bengtsson & Kock, 2000).

Choosing either a competitor or a non-competitor as a potential innovation partner commonly depends on the requirements of the partnership and the partner's ability to adhere to such requirements. Non-competitive partnerships are often sought for the provision of niche knowledge (Bougrain & Haudeville, 2002), complementary resources (Prashantham & Birkinshaw, 2008), or research-based knowledge (Nieto & Santamaría, 2007). Alternatively, organizations may seek a coopetitive partnership, the rationale for which varies. Commonly, coopetitive partnerships may derive from a need to carry out research and establish standards (Gueguen, 2009; Tether, 2002), address common problems (Nieto & Santamaría, 2007), achieve economies of scale (Gnyawali & Park, 2011), enter new markets (Gnyawali & Park, 2009), or share risks associated with introducing new products (Bouncken et al., 2015; Ritala & Hurmelinna-Laukkanen, 2009; Tether, 2002). Coopetitive partnerships are more likely than non-competitive partnerships to generate novel products (Ritala & Hurmelinna-Laukkanen, 2009).

Research has further shown that engaging with a wide *range* of competitors in an innovation ecosystem approach provides palpable benefits in the context of coopetition, including providing greater opportunities to observe and learn from the behaviors of competitors and facilitating open innovation through greater access to knowledge flows (Baierl et al., 2016). Numerous studies have illuminated the relationship between coopetition and innovation (Bouncken & Kraus, 2013; Gnyawali & Park, 2011; Leitão & Pereira, 2015; Ribeiro-Soriano et al., 2016; Ritala & Hurmelinna-Laukkanen, 2013; Ritala et al., 2016; Yami & Neme, 2014). Open innovation itself shares an important link with coopetition: simultaneous cooperative and competitive engagement between organizations relates strongly to the coupled mode of open innovation (Enkel, Gassmann, & Chesbrough, 2009), whereby

both outside-in and inside-out processes are employed in the innovation process. Thus, the relationship between open innovation and coopetition is an emergent area of notable interest (Ritala et al., 2016).

Coopetition has been identified as an integral component of open innovation ecosystems and is pursued in these contexts as a result of various lines of inquiry, including responding to threats, encouraging technology domination, and facilitating the diffusion of specialized activities (Gueguen, 2009). This importance has been echoed in previous empirical analyses of coopetition within ecosystems. Ritala, Agouridas, Assimakopoulos, and Gies (2013) investigated how leading firms facilitate value creation and capture with competitors across their ecosystems. Basole, Park, and Barnett (2015) demonstrated that coopetition in ICT ecosystems remains prevalent, with growing levels of value capture. Ritala, Golnam, and Wegmann (2014) analyzed the application of coopetition-based business models for value creation and capture in ecosystem contexts. Selander et al. (2010) explored the governance of coopetitive ecosystem relationships in the context of digital innovation. Gueguen and Isckia (2011) found that coopetitive strategies are particularly relevant in ecosystem contexts, with indirect coopetition also characterizing the ecosystem approach. Ansari, Garud, and Kumaraswamy (2016) analyzed how disruptive innovators negotiate coopetitive tensions within their ecosystems. Thus, research exploring coopetition in innovation ecosystems has progressed substantially in recent years. However, ecosystem-level analyses are still scarce (Ritala et al., 2016), belying the importance of these approaches to coopetition. Despite increasingly prevalent discourses about the positive interrelations between coopetition and innovation (Devece et al., 2019), few studies have evaluated the mechanisms employed by organizations for effective management of the knowledge sharing process in these contexts (Gast et al., 2015). With innovation ecosystems occupying an emergent area of interest in the coopetition literature, further explorations are required to

solidify the link between the two constructs, particularly because coopetitive and innovative capabilities are more important in complex, dynamic environments such as ecosystems (Roig-Tierno et al., 2018). This study addresses this gap by identifying the knowledge transfer conditions for coopetitive ecosystem partnerships.

3. Theoretical framework

According to a recent review (Devece et al., 2019), numerous theoretical frameworks have been applied in coopetition research, including game theory and the resource-based view. This broad use of theoretical approaches presents difficulties in relation to establishing a common or generally accepted theoretical lens for coopetition analyses. Moreover, knowledge sharing in coopetitive contexts requires a different theoretical framework depending on the level of analysis (Devece et al., 2019). Because of such inconsistencies, this study uses a previously validated theoretical lens that is appropriate for analyzing knowledge transfer success at the individual level in coopetitive contexts. This theoretical lens is based on the conceptual framework developed by Bacon, Williams, and Davies (2019). The framework reflects the importance of specific knowledge transfer conditions, identified through a review of the interorganizational knowledge transfer literature (Bacon et al., 2019). The most common conditions (i.e., with a citation frequency of 10 or higher) identified in the review are explored in this section. The conditions are classified into three groups: knowledge, relationship, and organizational characteristics.

3.1. Knowledge characteristics

The interorganizational literature reports that knowledge characteristics accelerate the transfer process (Easterby-Smith, Lyles, & Tsang, 2008). Knowledge is often typified as information-based and codifiable (explicit) or experience-based and personal (tacit; Nonaka

& Takeuchi, 1995). Because of the greater ease of translation, explicit knowledge is argued to be transferred more readily than its tacit counterpart (Al-Salti & Hackney, 2011). The degree of understanding possessed by the recipient organization thus depends on the nature of the knowledge. Accordingly, tacit versus explicit knowledge and an understanding of that knowledge are found to facilitate knowledge transfer. The framework applied in this research hence extends that of Bacon et al. (2019) by separating knowledge type into these separate components.

A further knowledge characteristic identified as affecting the transfer process is causal ambiguity, which involves a lack of clarity regarding the origins and components of knowledge (Szulanski, 1996). High causal ambiguity requires greater explanations, decelerating the transfer process (Al-Salti & Hackney, 2011). Tacit versus explicit knowledge, understanding, and low causal ambiguity formulate the knowledge category.

3.2. *Relationship characteristics*

The literature further posits that relationship characteristics between innovation partners encourage knowledge transfer. Enjoying the greatest popularity in the literature, trust involves an expectation that the innovation partner will fulfill their obligations as a transferee of knowledge (Inkpen, 1998). Trust is arguably a fundamental component of an ecosystem partnership because it encourages openness to communicate and share knowledge (Mazloomi Khamseh & Jolly, 2008). Moreover, a trustworthy partnership is crucial to achieving successful co-competition (Chin et al., 2008; Thomason, Simendinger, & Kiernan, 2013) because it enhances cooperative behavior and stimulates knowledge exchange (Planko, Chappin, Cramer, & Hekkert, 2019). Trust is further identified as an enabler of relationship strength (Narteh, 2008), alluding to another relationship characteristic: strong ties. These ties between organizations act as channels for knowledge flow (Easterby-Smith et al., 2008), encouraging

organizations to share more detailed knowledge and facilitating access to information (Narteh, 2008). Thus, an ecosystem that comprises strong partner ties is more likely to foster the successful exchange of high-quality knowledge (Van Wijk, Jansen, & Lyles, 2008). Trust and tie strength therefore encompass the relationship characteristics cited as expediting knowledge transfer.

3.3. *Organizational characteristics*

Other conditions related to the recipient organization are also of relevance for knowledge transfer success. The degree of cultural congruency between organizations in terms of shared beliefs, values, and practices is argued to advance knowledge transfer (Mowery, Oxley, & Silverman, 1996). Commonalities between a mutual vision and goals provide a crucial bonding mechanism (Van Wijk et al., 2008), encouraging organizations to become partners (Planko et al., 2019). Similarities between cultures thus underpin knowledge transfer success by delineating boundaries regarding the content and quality of knowledge (Narteh, 2008).

The willingness of the recipient organization to learn from an innovation partner is also cited as important for effective knowledge exchange (Inkpen, 1998). Learning does not occur automatically, and a determination and desire to achieve learning is therefore necessary (Mazloomi Khamseh & Jolly, 2008). Commitment to learning knowledge from an ecosystem partner ultimately assists in its successful transfer (Al-Salti & Hackney, 2011). However, this learning depends on the effective absorption of knowledge. The related concept of absorptive capacity is outlined as encompassing the recipient organization's ability to recognize the potential value of knowledge, diffuse it internally, and apply it beneficially (Cohen & Levinthal, 1990). Research has shown the importance of absorptive capacity for cooperation (Fredrich, Bouncken, & Kraus, 2019) by permitting organizations not only to absorb

knowledge effectively but also to transmit that knowledge across organizational boundaries (Easterby-Smith et al., 2008). Organizational characteristics thus comprise cultural similarity, learning intent, and absorptive capacity.

3.4. *Combinations of conditions*

The conditions explored in the conceptual framework shed light on how successful knowledge transfer is underpinned. However, their association with success does not necessarily mean that their absence will correlate with the absence of this outcome. Consequently, the investigation of which causes contribute to a given outcome conveys little about which causes contribute to the absence of an outcome, necessitating separate asymmetric models (Woodside, 2014). This causal asymmetry (Fiss, 2011) is an underlying tenet of fsQCA, providing a fundamental rationale for its application in this study: the fsQCA method enables an analysis that illuminates how instances of the presence and absence of success can be achieved.

Moreover, the multitude of conditions cited as underpinning knowledge transfer success are likely to result in multiple ways to achieve this outcome in cooperative and non-competitive partnerships. Termed equifinality, the presence of multiple, mutually exclusive pathways for accomplishing a given outcome underpins fsQCA (Fiss, 2011). Studies have provided contradictory results regarding the configurational nature of the knowledge transfer conditions. Despite the essential nature of relationship-, firm-, and knowledge-specific characteristics, they may be combined in different causal combinations to explain the outcome of success. Although some of the literature indicates that certain relationship-, knowledge-, or firm-specific characteristics form combinatory associations for cooperation (Fredrich et al., 2019; Ritala & Hurmelinna-Laukkanen, 2013; Planko et al., 2019), other research suggests that single conditions are of greater importance (Chin et al., 2008). These

inconsistencies in findings thus reflect a need for a more definite investigation into the configurational nature of knowledge transfer conditions for cooperative partnerships. The resulting conceptual model shows how the knowledge transfer conditions are expected to combine and overlap for the outcome of success (Figure 1).

Insert Figure 1 here.

4. Material and methods

4.1. Analytical technique

FsQCA was used to empirically analyze the conceptual framework. Its underlying tenets justify its suitability for this research. As outlined by Ragin (2009), fsQCA is a set-theoretic approach that detects the underlying causal conditions across cases that contribute to the expression of a given outcome. FsQCA reflects attempts to integrate the best features of qualitative and quantitative research (Ragin, 2009) by marrying the in-depth analyses associated with qualitative research with the accessible evaluations provided by quantitative analyses to effectively allow substantial comparisons across cases (Greckhamer, Misangyi, Elms, & Lacey, 2008). Cases are viewed as temporally and contextually embedded compositions of interrelated components rather than single entities (Pettigrew, 1990). FsQCA thus captures the diversity and heterogeneity of cases, in terms of their causally relevant conditions and contexts, by allowing cases to be compared as configurations (Ragin, 2009).

The configurational approach underpinning fsQCA is one of the fundamental justifications for selecting this method of analysis. This technique enables comparison of cooperative versus non-cooperative cases while generating highly representative characterizations of each case. It also enables detailed analyses that capture the heterogeneous nature of the partnerships explored in this study.

4.2. *Sample*

To ascertain the conditions that are present in cooperative and non-competitive partnerships, semi-structured telephone interviews were conducted with 20 members of an open innovation ecosystem. A snowball sampling strategy was employed (Goodman, 1961), and individuals from a large multinational services enterprise were contacted to request their perceptions of the conditions for knowledge transfer success. These participants then provided details of other suitable contacts across a variety of organisations, resulting in responses collected across a range of industries, including manufacturing, transport, technology, and finance.

Participants were placed into the cooperative or non-competitive category following classification of the partnership type. The cooperative category was composed of both multinational corporations and small businesses that were engaged in a partnership with an identified competitor. Participants in the non-competitive category were those whose company possessed no degree of competition with its innovation partner because they belonged to different industries or were of different organizational types.

4.3. *Interview schedule*

The interviews required participants to assess the presence of each condition on a 7-point scale that applied the principles of a semantic differential approach (Osgood, Suci, & Tannenbaum, 1957). Opposing adjectives were placed at either end of the scale to reflect the absence and presence of each condition. These scales were supplemented with additional questions asking participants to expand on their answers, allowing them to provide further information regarding each condition and generating a greater degree of freedom and flexibility in the responses. Table 1 displays the scales and their supplementary questions (adapted from Bacon et al., 2019).

Insert Table 1 here.

This study used the definition of knowledge transfer success given by Bacon et al. (2019) to assess whether each partnership could be characterized as having successfully achieved this outcome. The proposed definition was the “active exchange of knowledge between organizations, involving measurable and effective knowledge absorption, application and satisfaction by the recipient organization” (p. 380). This definition was embedded in the interviews as a further semantic differential measurement, and individuals were asked to rate their extent of agreement with five statements corresponding to the five separate constituents of knowledge transfer success.

4.4. *FsQCA calibration*

FsQCA operates on the basis of each case’s degree of membership in a given set. To establish degrees of membership, the data are calibrated into values ranging from 0 to 1. The calibration procedure requires specification of the following threshold values: full membership in the outcome, full non-membership, and maximum ambiguity (Fiss, 2011). The value for full membership was set at 6, the crossover point was set at 4.5, and full non-membership was set at 3. The use of semantic differential scales to assess this data set enabled natural breakpoints contained in the scale to be used as the three threshold values, which is an accepted approach to calibration (Ordanini, Parasuraman, & Rubera, 2014; Pappas, Kourouthanassis, Giannakos, & Chrissikopoulos, 2016; Woodside, 2013).

The conditions Tacit and Explicit Knowledge were originally calibrated as a single fuzzy-set condition. However, doing so meant that participants’ responses were inaccurately reflected because responses referring to both knowledge types could not be recorded on the aforementioned single continuum. Thus, it was deemed necessary to divide knowledge type into two sets: either tacit or explicit. In this manner, membership in “tacit knowledge” could

be coded as 1, a crossover point as 0.5, and non-membership as 0. Explicit knowledge was recoded in the same manner. The combination of explicit and tacit knowledge was indicated by representing membership in each condition in the analysis. The nature of these conditions meant that applying a 3-value fuzzy set was more suitable to capture participants' responses representatively.

Following calibration, the fsQCA software produces a truth table, which requires refinement according to the criteria of frequency and consistency. Frequency refers to the number of empirical observations of cases containing each specific combination of causal conditions. For a small data set, an acceptable frequency cutoff can be set at 1 (Ragin, 2009). Consistency reflects the degree to which the causal combinations represent a subset of the outcome (Ragin, 2009). The minimum acceptable consistency threshold for any solution is recommended as 0.75 (Ragin, 2006). Configurations with a frequency value of less than 1 and a consistency value of less than 0.8 were consequently removed from the truth table.

5. Results and discussion

The fsQCA results offer three types of solutions that contain the presence or absence of different conditions but ultimately lead to the same outcome: knowledge transfer success. The two solutions that are recommended for further discussion are the parsimonious and intermediate solutions (Fiss, 2011) because they incorporate logical remainders (configurations with no empirical evidence). Causal conditions that are present in both the parsimonious solution and the intermediate solution are referred to as core conditions; those that are solely present in the intermediate solution are peripheral conditions. Observing which conditions are present in either solution determines the causal essentiality of the conditions (Fiss, 2011). Core conditions are deemed essential, whereas peripheral conditions can be

considered of lesser importance because of their exchangeability with other conditions (Fiss, 2011).

5.1. Solutions for coepetitive partnerships

Table 2 displays the core and peripheral conditions that are present for the coepetitive partnership category.

Insert Table 2 here.

The results for the coepetitive category give four solutions for knowledge transfer success. Solution 1 combines the presence of Explicit Knowledge, Low Causal Ambiguity, Trust, Tie Strength, Learning Intent, and Absorptive Capacity with the absence of Tacit Knowledge and Cultural Similarity; Understanding is considered a redundant condition. The absence of Tacit Knowledge is a core condition here. Solutions 2 and 3 both combine the presence of Tacit Knowledge, Explicit Knowledge, Understanding, Low Causal Ambiguity, Tie Strength, Cultural Similarity, and Learning Intent. They differ in the redundancy of Absorptive Capacity in Solution 2 and Trust in Solution 3. Core conditions in both solutions are the presence of Cultural Similarity and Learning Intent. Solution 4 combines the presence of Tacit and Explicit Knowledge, Low Causal Ambiguity, Tie Strength, Cultural Similarity, and Learning Intent with the absence of Understanding, Trust, and Absorptive Capacity. Core conditions here are Cultural Similarity and Learning Intent.

The sole core condition in Solution 1 is the absence of Tacit Knowledge; all other conditions are peripheral. Knowledge that has an explicit quality is thus of greater importance in this solution for generating a successful transfer process. The literature states that explicit knowledge is more readily transferred than tacit knowledge (Nonaka & Takeuchi, 1995; Spender, 1996; Wang & Wang, 2012), implying that tacit knowledge may delay the transfer process. Loebecke, Van Fenema, and Powell (1999) argue that knowledge tacitness hinders the dynamics of knowledge exchange in coepetitive relationships. Participants stated that the

transferred knowledge commonly comprised “*logistical aspects of the process*” rather than experiential information (Director, Multinational Telecommunications Enterprise). The absence of Tacit Knowledge was further reiterated by the fact that the knowledge exchanged was more “*formalized knowledge surrounding key technologies*” (Coordinator, Multinational Transport Enterprise). This finding indicates that knowledge transfer success in Solution 1 is highly dependent on this absence of Tacit Knowledge. As reflected by participant responses, the fundamental goal of a cooperative partnership relates to the exchange of information regarding specific technologies for innovative purposes. Competitive partners may wish to share solely explicit information to avoid disclosure of the more idiosyncratic knowledge surrounding experiences specific to their own organization, which could be more valuable with a view to imitation. Accordingly, the absence of Tacit Knowledge is a core condition for knowledge transfer success between multinational organizations. This finding contradicts previous findings that indicate the importance of tacit knowledge exchange for cooperation (Bouncken & Kraus, 2013; Estrada, Faems, & de Faria, 2016; Ho & Ganesan, 2013); instead, the absence of this condition results in knowledge transfer success, indicating that its explicit counterpart is of greater importance for achieving this outcome.

Core conditions in Solutions 2, 3, and 4 are the presence of Cultural Similarity and Learning Intent, expressed as Cultfz*Learnfz, where “*” denotes the combination of two conditions. This finding indicates that the presence of both of these conditions has a strong causal relationship with the outcome. Participants stated that their culture was “*highly similar*” (Director, Small Technology Business) to their ecosystem partner’s, and their cultural similarity with this partner was argued to generate “*greater consistency and alignment of ideas*” (Department Head, Multinational Services Enterprise) when collaborating with a competitor. This cultural similarity helped facilitate an “*understanding of how they operate,*” making the competitor “*more appealing as an engagement partner*”

(Manager, Multinational Consultancy Enterprise). Such goal alignment and common vision have previously been cited as critical success factors for coopetition (Planko et al., 2019). Cultural similarity is therefore perceived to encourage competitors to co-create and to facilitate and maintain partnerships. The presence of Cultural Similarity across 75% of the solutions is to be expected: competitive partners often have greater cultural congruency because they are similar in size and scale, often residing in the same industry. Indeed, such similarities are what makes them competitors.

The combination of Cultfz*Learnfz in this solution arguably arises because participants identified Learning Intent as a “*requirement for co-creation*” (Director, Small Software Business). Cultural consistency between organizations could thus create a desire and intent to learn from a partner organization. This finding supports Lawson and Potter’s (2012) claim that an organization is encouraged to transfer knowledge if collaborating with a partner that perpetuates a desire to receive that knowledge. Such a desire is arguably embedded in the norms and values of an organization and is subsequently integrated into its overall culture. Thus, having similar cultures can act as an antecedent for Learning Intent, explaining the presence of these two conditions as a core configuration across Solutions 2, 3, and 4.

5.1.1. Necessary conditions

When assessing the presence of necessary conditions in a data set, consistency scores of 0.9 or higher are recommended (Ragin, 2009). The closer the consistency score is to 1, the greater the extent to which the condition leads to the outcome. In the coopetitive category, three conditions surpass the recommended consistency threshold: Explicit Knowledge (0.98), Tie Strength (0.98), and Learning Intent (0.90).

Explicit Knowledge. The presence of Explicit Knowledge as a necessary condition arguably derives from the nature of cooperative partnerships. Generally, organizations seek to collaborate with potential competitors within their ecosystem for the common purpose of addressing a shared problem (Nieto & Santamaría, 2007). Consequently, the knowledge exchanged with ecosystem partners is typically “*technical information*” (Director, Multinational Telecommunications Enterprise). The ease of transfer associated with explicit knowledge (Nonaka & Takeuchi, 1995; Spender, 1996; Wang & Wang, 2012) and its ability to facilitate interfirm learning in cooperative partnerships (Bouncken & Kraus, 2013) arguably explains its presence across all solutions; in the cooperative category, the exchange of highly specific knowledge with regard to an identified innovation requirement comprises the fundamental motive for engaging in the partnership.

Tie Strength. Tie Strength exists as a necessary condition across all solutions. Participants stated that they had “*well-established*” relationships (Director, Multinational Telecommunications Enterprise) with “*strong qualities*” (Director, Small Technology Business) that “*[elevated] the relationship*” (Department Head, Multinational Service Enterprise). Participants even specified that “*possessing a weak relationship can lead to the diminution of a relationship where the partner will leave*” (Department Head, Multinational Technology Enterprise). Additionally, the difficulties associated with engaging with a competitor and the associated erosion of trust mean that a strong relationship can often formulate the backbone of the partnership and can be a source of reliance where trust is potentially being diminished. The literature cites the importance of strong ties in facilitating cooperation (Brols, 2009; Choi, Garcia, & Friedrich, 2009). The presence of Tie Strength as a necessary condition in this data set substantiates this claim.

Learning Intent. The presence of Learning Intent as an additional core condition in this category arguably arises because “*a willingness and ability to innovate*” was cited as

fundamental for engaging in an ecosystem partnership (Department Head, Multinational Technology Enterprise). Participants expressed that their organizations fostered “*a culture of enjoying learning and seeing value from it*” (Director, Multinational Telecommunications Enterprise) and demonstrated this importance by ensuring that it was “*embedded in strategy*” (Director, Small Consultancy Business). The literature also states that a fundamental aim of a competitive partnership is to learn about the potential competencies of competitors (Hamel, Doz, & Prahalad, 1989). Research further accentuates the importance of learning intent for knowledge transfer success (Bandyopadhyay & Pathak, 2007; Inkpen, 1998; Tsang, 2002), with further studies postulating the ability of learning intent to offer increased opportunities to foster cooperative benefits (Ho & Ganesan, 2012). Thus, the empirical and theoretical evidence reinforces the importance of Learning Intent for cooperation in knowledge sharing contexts, supporting its presence as a necessary condition.

The four solutions for the cooperative category can be considered informative, with consistency values higher than 0.74 and coverage values between 0.25 and 0.90 (Oyemomi, Liu, Neaga, & Alkhuraji, 2016). These values support the authenticity of the solutions. Overall consistency is high, nearly reaching perfect consistency (Ragin, 2006). Overall solution coverage indicates that the causal configurations account for a large proportion of the instances of the outcome (Ragin, 2009).

The results for the cooperative category confirm the combinatory nature of the knowledge transfer characteristics. While Cultural Similarity and Learning Intent are core conditions, absorptive capacity fails to display the same level of importance. Thus, firm-specific characteristics cannot be considered more important for this category. Interestingly, the analysis reveals one condition from each characteristic as necessary, strengthening the implication of each condition grouping for knowledge transfer success. For cooperative

partnerships, knowledge-, firm-, and relationship-level characteristics form configurational associations for successful knowledge transfer.

5.2. *Solutions for non-competitive partnerships*

Table 3 shows the solutions for non-competitive partnerships.

Insert Table 3 here.

The results for the non-competitive category reveal different solutions for knowledge transfer success from those for the cooperative category. Solution 1 combines the presence of Tacit Knowledge, Explicit Knowledge, Understanding, Trust, Tie Strength, and Learning Intent with the absence of Low Causal Ambiguity and Absorptive Capacity; Cultural Similarity is considered a redundant condition. Core conditions here are Explicit Knowledge and Learning Intent. Solution 2 combines the presence of all conditions aside from Cultural Similarity, an absent condition, and Absorptive Capacity, a redundant condition. Core conditions in Solution 2 are Explicit Knowledge and Learning Intent. Solution 3 combines the presence of Tacit Knowledge, Low Causal Ambiguity, and Tie Strength with the absence of Explicit Knowledge, Understanding, Trust, Cultural Similarity, Learning Intent and Absorptive Capacity. The absence of Cultural Similarity and the absence of Absorptive Capacity are both core conditions. Solution 4 combines the presence of Tacit and Explicit Knowledge, Understanding, Low Causal Ambiguity, and Learning Intent, with the absence of Trust, Tie Strength, Cultural Similarity and Absorptive Capacity. Core conditions in Solution 4 are Explicit Knowledge and Learning Intent.

Solution 3 contains the causal configuration $\sim\text{Cultfz}*\sim\text{Acapfz}$, where “ \sim ” denotes the absence of the condition. This combination is core in this solution, perhaps because non-competitive partnerships are sought for different reasons than those for which cooperative partnerships are sought. Whereas cooperation is pursued to address knowledge gaps in a

specific domain (Yami & Neme, 2014) or to carry out research and establish standards (Gueguen, 2009; Tether, 2002), sole collaboration may be pursued for the purposes of differentiation. As indicated by one participant, innovation partners can “*offer different ways of working*” (Director, Small Consultancy Business). These inconsistencies between working practices may indicate that a lack of cultural similarity between organizations is beneficial for successful knowledge transfer. The combination $\sim\text{Cultfz}^*\sim\text{Acapfz}$ confirms the theoretical connections found in the literature: research states that the extent of an organization’s absorptive capacity derives from cultural compatibility between partners (Lane, Salk, & Lyles, 2001; Martinkenaite, 2011). Participants identified an organizational commitment to absorb knowledge but an overall “*reluctance to do so in practice*” (Business Development Officer, Small Technology Business). Additionally, participants cited absorptive capacity as a “*challenge for the organization*” because of the “*fast-paced growth of the industry and the organization itself,*” which presented difficulties in terms of “*struggling with the basics*” (Director, Small Consultancy Business). Nonetheless, the absence of Absorptive Capacity appears to facilitate successful knowledge transfer when combined with the absence of Cultural Similarity. A lack of cultural consistency between organizations may therefore discourage internal dissemination of the transferred knowledge within the recipient organization; however, this lack of cultural consistency does not affect knowledge transfer success.

In Solutions 1, 2, and 4, the causal configuration $\text{Learnfz}^*\text{Expfz}$ is core for knowledge transfer success. Research suggests that the process of organizational learning requires an overall intent to learn (Evangelista, 2007; Hamel, 1991; Mohr & Sengupta, 2002). Participants stated that their organizations were “*always open to learning new ideas*” (Chief Financial Officer, Small Financial Business). Like in the cooperative category, participants stressed the importance of learning intent, deeming it “*necessary for survival*” (Director,

Small Consultancy Business; Director, Small Software Business). The combination Learnfz*Expfz emphasizes that participants are more likely to learn from an innovation partner if the partner provides the required knowledge type.

5.2.1. Necessary conditions

The core condition Expfz is present in three out of the four solutions. Interestingly, Explicit Knowledge is a necessary condition for the cooperative category. Conversely, Tacit Knowledge has a consistency score of 0.97 in the analysis of necessary conditions for this category. Its presence across all four solutions further supports the necessity of this condition.

When exploring scale responses from participants in this category, all responses indicate the presence of either Tacit Knowledge alone or in conjunction with Explicit Knowledge. Participants expressed that the knowledge transferred included “*technical information*” as well as sharing “*pockets of expertise*” combined with “*personal experiences*” (Manager, Small Technology Business). Knowledge was argued to be explicit in terms of retaining “*market-specific*” qualities that were primarily “*technology and solution based*” and tacit in terms of “*experience and information... more informal, best practices, dos and don’ts*” (Director, Small Technology Business). Interestingly, participants believed that the stage of the innovation process influenced the type of knowledge exchanged, but there seemed to be discord between participants. Some participants perceived tacit knowledge exchange to occur initially to create a sense of “*personal alignment*” and “[*understanding*] of people,” which is “*necessary for business development*” (Manager, Small Technology Business). Evolution of the partnership arguably supports the exchange of more explicit knowledge because tacit knowledge needs to be present to “*facilitate understanding and create the relationship*” (Manager, Small Technology Business). In contrast, some participants stated that “*technology exchange occurred on the outset*” and then evolved into

experiential information pertaining to the achievement of “*performance and results*” (Chief Financial Officer, Small Financial Business). It thus seems that knowledge type has greater fluidity in the context of non-competitive ecosystem partnerships. This greater fluidity may derive from the fact that cooperative partnerships perpetuate a more superficial knowledge transfer process, occurring at surface level to eradicate the potential exploitation or imitation of the knowledge that is transferred. Non-competitive partnerships maintain a more complex relationship in terms of knowledge, retaining more flexible qualities that are subject to change over the course of the partnership.

Overall solution coverage and consistency values are similar to the values for the cooperative category (0.75 and 0.93, respectively), implying that all solutions are informative. Because these scores are almost the same for the two partnership types, their comparability is high. No condition grouping demonstrates sufficiency or necessity, again confirming that combinations of condition characteristics are responsible for success. This conclusion is further reinforced by the observation that cross-category conditions act as causal configurations (e.g., Learnfz*Expfz). Hence, conditions exert a combinatory effect on the occurrence of success.

6. Conclusions

This study establishes considerable distinctions between cooperative versus non-competitive partnerships operating in open innovation ecosystems. To capture the heterogeneity in these partnership types, fsQCA was employed to conduct a configurational analysis of knowledge transfer conditions and detect their combinations in the form of causal recipes. The findings reveal that although individual conditions are necessary across partnership types, the success of both cooperative and non-competitive partnerships ultimately depends on combinations of knowledge-, firm-, and relationship-related characteristics. No

solution exceeded the acceptable consistency threshold (0.75) for the absence of the outcome. Thus, although certain conditions are associated with the presence of success, further research is required to detect the antecedents of the absence of success.

6.1. *Theoretical implications*

This study has major theoretical implications for coopetition research. Existing literature has explored the link between competition and innovation (Bouncken & Kraus, 2013; Gnyawali & Park, 2011; Leitão & Pereira, 2015; Ribeiro-Soriano et al., 2016; Ritala & Hurmelinna-Laukkanen, 2013; Ritala et al., 2016; Yami & Nemeh, 2014). However, in terms of ecosystems, business ecosystems remain the predominant focus of such studies (Gueguen, 2009; Gueguen & Isckia, 2011; Ritala et al., 2014), with few studies addressing this link by shifting the focus to coopetition in open innovation ecosystems. This research thus contributes to the coopetition literature by empirically analyzing the construct in open innovation ecosystem contexts.

This research also extends coopetition research by presenting a comparative analysis of cooperative versus non-competitive partnerships. Notable studies evaluating coopetition in ecosystem contexts have typically relied on absolute analyses, solely examining coopetition itself (Ansari et al., 2016; Basole et al., 2015; Gueguen, 2009; Gueguen & Isckia, 2011; Ritala et al., 2013; Ritala et al., 2014; Selander et al., 2010). They have thus failed to ascertain how cooperative mechanisms differ from non-competitive mechanisms. In addressing this absence, this study compares various configurations of knowledge transfer success for cooperative and non-competitive partners, thus providing a more comprehensive examination of how success is affected by the competitive nature of ecosystem partnerships.

The elucidation of the differences between these partnerships, achieved using configurational approaches, is a further contribution of this research, whereby the application

of fsQCA has accentuated fundamental distinctions between cooperative and non-competitive ecosystem partners. The application of fsQCA in innovation-related research is becoming more common (see Kraus, Ribeiro-Soriano, & Schüssler, 2018), and fsQCA is an established methodological tool for cooperation studies (Adame-Sánchez, Caplliure, & Miquel-Romero, 2018; Bouncken, Fredrich, & Kraus, 2019; Fredrich et al., 2019; Ribeiro-Soriano et al., 2016). In response to Ritala et al.'s (2016) recommendation to analyze cooperation through novel methods such as fsQCA, this paper is, to the best of our knowledge, one of the first to use fsQCA to compare the causal configurations of cooperative and non-competitive relationship types in the context of innovation ecosystems. This research thus further contributes to a growing body of research that applies configurational approaches to innovation-related outcomes.

6.2. *Practical implications*

The findings of this research have noteworthy implications for practitioners engaging in open cooperative partnerships in innovation ecosystems. Equifinal causal configurations, combined with the lack of sufficient conditions in this data set, reflect the absence of a single model of best-fit for knowledge transfer success. Instead, success remains context-bound and depends on the nature of the partnership. Although a unifying approach is lacking, this study does identify the prominent role of necessary conditions for both cooperative and non-competitive partnerships. For cooperative partnerships, explicit knowledge, tie strength, and learning intent are necessary. Firms should thus introduce mechanisms to encourage the presence of these conditions in their partnerships. First, establishing measures for strengthening interfirm partnerships through, for example, increased interactions (Rejeb-Khachlouf, Mezghani, & Quélin, 2011) could be beneficial in contexts where weaknesses in interfirm relations are observed. Organizations could thus increase opportunities to interact

with partners (e.g., meetings, workshops, and visits) to foster stronger partnerships. To elicit the exchange of explicit knowledge, organizations could increase transparency in the information exchanged and attempt to disentangle any personalized or experiential components. Doing so would ensure that the information exchanged was solely explicit, thus increasing the likelihood of successful transfer when partnering with competitors. Finally, the emergence of learning intent as a necessary (and core) condition implies that corporations should encourage employees to be motivated to learn from partners, in turn increasing the learning intent of the organization as a whole.

This study further highlights the necessity of tacit knowledge in non-competitive ecosystem partnerships. Previous research has offered measures to accelerate tacit knowledge transfer, namely having increased interaction (Inkpen & Dinur, 1998), building strong relationships based on trust (Mu, Love, & Peng, 2008; Panteli & Sockalingam, 2005; Roberts, 2000), and establishing a common language (Carlile, 2004). Firms could use such mechanisms as a basis for encouraging tacit knowledge transfer by increasing opportunities to interact with partners, thereby creating and strengthening trust in the partnership.

Lastly, this research shows the prominence of learning intent as a core condition across both partnership types. Organizations should therefore incorporate learning intent in their business strategies and communicate its importance to ensure that information is exchanged successfully across the ecosystem. Identifying this commonality across both partnership types can mitigate the absence of a consistent approach to knowledge transfer success. In general, firms should be mindful of the idiosyncrasies of cooperative versus non-competitive partnerships and the different routes to successful knowledge transfer in each case. Innovation approaches should therefore be tailored specifically to the partnership type. The heterogeneity of knowledge transfer practices makes it difficult for organizations to ensure that knowledge transfer is accomplished. Ultimately, awareness of which conditions

are most important in the context of cooperation and which combinations of conditions contribute to success in these instances should help practitioners ensure that information is successfully exchanged with their ecosystem partners.

6.3. *Limitations*

As with all research, a number of limitations apply to this study and could be alleviated through further research. First, no solutions surpassed the consistency threshold for the absence of the outcome. Hence, further research is required to identify the conditions associated with the absence of success. Second, the sample of 20 respondents was relatively small. Although fsQCA has been identified as particularly suitable for small-N data sets (Woodside, 2013), the generalizability of the results is questionable. Future research could address this limitation through similar analyses on a wider scale using a larger sample of respondents. Additionally, the small sample size results in a limited empirical investigation on a range of industries, organization types, organizational positions, and innovative developments, all of which could shape the samples and contexts of future studies.

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Table 1:

Interview questions

Condition	Initial questions and semantic differentials	Follow-up questions (for each condition based on response to initial question)
Tacit/Explicit Knowledge	Would you say the knowledge you gained from your ecosystem partner was information highly personal and experiential, specific to your ecosystem partner (tacit)? Or was it more technical knowledge, in the form of perhaps manuals or policies (explicit)?	
Understanding	Please rate your level of understanding of the information you received, ranging from low level to high level.	
Low Causal Ambiguity	To what extent do you believe that the origins of the information gained from your ecosystem partner were unambiguous, ranging from unclear to clear?	Could you please tell me a bit more about why that is?
Trust	To what extent do you believe that you have a trustworthy relationship with your ecosystem partner, ranging from untrustworthy to trustworthy?	Why didn't you choose a lower/higher rating?
Tie Strength	To what extent do you believe that you have a strong relationship with your ecosystem partner, ranging from weak to strong?	Can you give me an example of this?
Cultural Similarity	To what extent do you believe that your organization possesses an organizational culture similar to that of your ecosystem partner, ranging from dissimilar to similar?	
Learning Intent	To what extent do you believe that your organization possesses a willingness to learn new knowledge, ranging from unwilling to willing?	I'd like to hear more about...
Absorptive Capacity	To what extent do you believe that your organization absorbs and acquires new knowledge, ranging from un-absorptive to absorptive?	

Table 2:

Solutions for coopetitive partnerships

Configuration	Solution			
	1	2	3	4
Tacit Knowledge	⊗	●	●	●
Explicit Knowledge	●	●	●	●
Understanding		●	●	⊗
Causal Ambiguity	●	●	●	●
Trust	●	●		⊗
Tie Strength	●	●	●	●
Cultural Similarity	⊗	●	●	●
Learning Intent	●	●	●	●
Absorptive Capacity of recipient firm	●		●	⊗
Consistency	0.98	0.92	0.91	0.96
Raw coverage	0.27	0.42	0.33	0.13
Unique coverage	0.21	0.15	0.06	0.06
Overall solution coverage	0.75			
Overall solution consistency	0.94			

Note: Black circles (●) indicate the presence of a condition; crossed out circles (⊗) indicate the absence of a condition. Large circles represent core conditions; small circles represent peripheral conditions; blank spaces indicate redundant conditions.

Table 3:

Solutions for non-competitive partnerships

Configuration	Solution			
	1	2	3	4
Tacit Knowledge	●	●	●	●
Explicit Knowledge	●	●	⊗	●
Understanding	●	●	⊗	●
Causal Ambiguity	⊗	●	●	●
Trust	●	●	⊗	⊗
Tie Strength	●	●	●	⊗
Cultural Similarity		⊗	⊗	⊗
Learning Intent	●	●	⊗	●
Absorptive Capacity of recipient firm	⊗		⊗	⊗
Consistency	0.94	0.88	0.95	0.95
Raw coverage	0.33	0.32	0.16	0.16
Unique coverage	0.24	0.22	0.11	0.07
Overall solution coverage	0.75			
Overall solution consistency	0.93			

Note: Black circles (●) indicate the presence of a condition; crossed out circles (⊗) indicate the absence of a condition. Large circles represent core conditions; small circles represent peripheral conditions; blank spaces indicate redundant conditions.

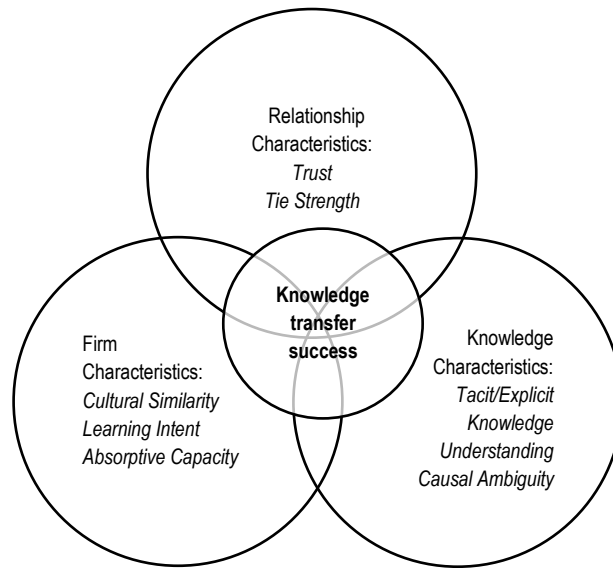


Figure 1:

Conceptual Framework