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Article

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How to innovate continuously? Conceptualizing generative capability

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ABSTRACT

This study conceptualizes the novel concept of generative capability, a unique capability by which firms organize experiential learning activities to achieve fast and iterative innovation. Employing a mixed-method research design, we first explored the theoretical framework of generative capabilities through an inductive case study to identify its dimensions. On this basis, we also developed measurement scales for generative capability. We then used survey data on a sample of 212 Chinese firms to test the construct and criterion validity of generative capability. Our findings show that generative capability can be captured by a second-order reflective construct with the three dimensions of knowledge acquisition, knowledge inheritance, and knowledge updating, and that capability is positively associated with firm innovation performance.

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Introduction

How can firms continuously and sustainably pursue innovation? The literature on product innovation has examined product innovation as a discrete event, focusing on the factors contributing to the development of a specific new product (Slater, Mohr & Sengupta, 2014; West & Bogers, 2014). Comparatively, less attention has been paid to continuous product innovations, which is defined as “the generation of multiple new products, as strategically necessary over time, with a reasonable rate of commercial success (Dougherty & Hardy, 1996, p. 1121)”. This temporal perspective of innovation is crucial for competing in an increasingly dynamic environment (Ancona, Okhuysen & Perlow, 2001).

While prior studies have found that discrete product innovation is associated with R&D resources (Song & Chen, 2014), R&D speed (Zhu, Xiao, Dong & Gu, 2019), market orientation (Joshi, 2016), and absorptive capacity (Santoro, Bresciani & Papa, 2020), it is unlikely that simply assembling these conditions is sufficient for firms to continuously introduce superior new products through successive innovation projects (Wang & Chen, 2018). Instead, other firm-level capabilities may drive the process and deliver the outcomes of continuous product innovation. This study proposes the concept of generative capability to incorporate key organizational routines that underlie continuous product innovation.

Specifically, this study conceptualizes, measures, and validates generative capability as an overarching capability that enables continuous product innovation. Scholars have considered innovation as the recombination of existing components (Schumpeter, 1939; Taylor, 2010) Fleming (2001). argued that the source of technological uncertainty is searching for processes with unfamiliar components and their combinations, which can result in both technological failure and breakthrough. Adopting a knowledge-based view, we propose that generative capability is manifested in a firm's high-level routines that reconfigure knowledge embodied within the product generation iteratively and re-bundle external and internal knowledge rapidly across product generations to develop the next-generation product. Generative capability enables firms to carry out cross-functional processes of redeploying internal and external competencies (Dougherty & Hardy, 1996; Garud, Gehman & Kumaraswamy, 2011; Soosay, Hyland & Ferrer, 2008), as well as the design processes of knowledge reconfiguration and design iteration (Eisenhardt & Bingham, 2017; Eisenhardt & Tabrizi, 1995; McKinley, Latham & Braun, 2014). Therefore, we believe that generative capability can enable firms to reduce technological uncertainty and the risk of R&D failure in the process of product development. This study aims to conceptually clarify the dimensions of generative capability as a driver of continuous product innovation and to empirically establish its construct and criterion validity.

To achieve this, we adopted a mixed-method design. We first conducted qualitative research to explore the theoretical framework of generative capability from in-depth interviews with eight high-tech

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companies in China. Findings from this qualitative phase allow us to establish the dimensions and structure of the construct of generative capability and propose measures for the dimensions. In the subsequent quantitative research phase, using survey data from a random sample of 212 Chinese high-tech companies, we empirically verified the proposed measurement and theoretical models to demonstrate the construct and criterion validities of generative capability. Our quantitative findings support generative capability as a second-order reflective construct, based on three key organizational routines: *knowledge acquisition, knowledge inheritance, and knowledge updating*.

The remainder of this paper is organized as follows. Following this introduction, we identify the literature gap in continuous product innovation and introduce the concept of generativity in different research fields. After linking generativity to product innovation literature, we develop an overarching framework for our theory development. Next, we proceed to the qualitative phase of this mixed-method research, where, through an inductive approach based on qualitative interviews, we elaborate on the conceptual structure and dimensions of generative capability. This is followed by the second phase, namely, a quantitative empirical study to validate the scale and test the hypotheses using survey data. We conclude the paper with a discussion of the main findings, theoretical contributions, implications for practice, limitations, and future research.

Literature review

Continuous product innovations

Despite their different research contexts, scholars of continuous product innovations¹ commonly emphasize its importance because of its function in coping with changing environments, seizing opportunities, and achieving sustained competitive advantage (Dougherty, 2001; Kim, Kim & Foss, 2016; Soosay et al., 2008). The central question in this research field is how firms can achieve continuous production innovation. The literature shows that achieving product innovation continuity is a complex process (Boer, Caffyn, Corso, Coughlan, Gieskes, & Magnusson, 2001; Dougherty & Hardy, 1996; Garud et al., 2011). Using qualitative data from 15 large organizations, Dougherty and Hardy (1996) found that large firms had difficulty achieving continuous innovation because they could not solve innovation-to-organization problems, such as the inability to connect new products with organizational resources, processes, and strategy (Dougherty, 2001). also articulated the image of work (in other words, an understanding of the work to be done) for sustained product innovation by comparing people's shared images of work in innovative and non-innovative organizations. His findings showed that, in contrast to the image of work in non-innovative organizations, which is based on passive maintenance of the system, people in innovative organizations understand that value creation is achieved by the situated practice of problem solving with customers, which requires the development and maintenance of long-term working relationships with customers.

To explore how firms can achieve continuous production innovation, other scholars utilize longitudinal case evidence. For instance, Lazonick and Prencipe (2005) analyzed the development history of

the Rolls-Royce Plc and found that strategic control and financial commitment play important roles in the process of continuous innovation. Top managers who make strategic decisions must have the capabilities and motivations to pour resources into innovation. Based on an in-depth study of innovation practices of 3 M Corporation, Garud et al. (2011): 760) identified how complexity arrangements, which they describe as “an intertwined set of arrangements capable of simultaneously addressing different complexities,” can facilitate sustained innovation. They found that through complex arrangements, 3 M can interweave actors, artifacts, and practices over time and enable productive nonlinear innovation.

In addition to the literature mentioned above, scholars also investigate continuous product innovations from different perspectives, such as organizational learning (Ballé, Morgan & Sobek, 2016), knowledge management (Kaminska & Borzillo, 2016; Kim et al., 2016), quality management (Cole, 2002), and supply chain management (Soosay et al., 2008). As such, existing insights into continuous product innovation are largely fragmented and primarily interpretative in nature. A clearly conceptualized and measurable construct pertaining to the core capability of enabling continuous product innovation is lacking. This hinders the bridging and synthesis of different research streams to better address the challenges of continuous product innovation. To address this knowledge gap, this study aims to explicate the conceptualization and operationalization of generative capability, which may serve as a foundation for further research to advance our understanding of continuous product innovation.

The notion of generativity

The concept of generative capability is centered on the notion of generativity, which has been introduced in multiple research fields. For instance, in linguistics, “generative grammar” has been developed as a theory that considers grammar as a system of basic rules that can generate infinite syntactical configurations (Chomsky, 2006). In the area of social policy research, Schön (1993) put forward the concept of “generative metaphor,” through which people's perspectives on the world can be changed, and new insights can be gained. Similarly, in the field of social psychology, Gergen (2012) proposed the notion of “generative capacity” to describe the ability of a theory or idea that challenges the status quo to enable out-of-the-box thinking (Avital and Te'Eni (2009: 349) also introduced the concept of “generative capacity” in the information systems field, by which they capture “one's ability to reframe reality and subsequently to produce something ingenious or at least new in a particular context”. Beyond the above-mentioned research fields, the notion of “generative” has also been applied in the areas of emergence (Lichtenstein, 2014), architecture (Alexander, 1999), computer design (Frazer, 2002), internet technology (Zittrain, 2008), and strategic partnerships (Remneland-Wikhamn, Ljungberg, Bergquist & Kuschel, 2011).

One consistency that emerges from this diverse literature suggests that generativity involves the transformation of patterns and the creation of new knowledge based on existing knowledge. In the product innovation context, the notion of “generativity” is particularly relevant because by reorganizing and upgrading the knowledge related to existing products, firms can continuously launch new-generation products, thus achieving continuous product innovation and gaining sustained competitive advantage. The new generation of products can be improved versions of the previous generation of products, or brand new products in a new market segment that spawned from the previous generation of products. In this regard, Ahuja, Lampert and Novelli (2013): 248) coined the concept of “generative appropriability,” which they define as “a firm's effectiveness in capturing the greatest share of future inventions spawned by its existing inventions”. An example of this concept is that Apple combined the innovative features of iPod and the functions of

¹ Also referred as sustained product innovation in some literature (e.g., Dougherty and Hardy, 1996; Dougherty and Deborah, 2001). Although both the terms continuous production innovation and sustainable production innovation share some kind of long-term innovation, they have different focal points. Sustainable product innovation is a research stem that emphasizes the sustainability of resource and environment during new product development process (Adams, Jeanrenaud, Bessant, Denyer & Overy, 2016; Varadarajan, 2017), while continuous production innovations focuses on the sustainability of firms' innovation behavior itself. Nevertheless, the two terms are not mutual exclusive. Continuous production innovations can be sustainable and vice versa.

telecommunication to create the iPhone. Inspired by previous research (Ahuja et al., 2013; Sun & Zou, 2019) and our observations of many leading firms, we propose the concept of generative capability to examine the mechanism of firms' continuous product innovation. From a knowledge-based perspective (Grant, 1996a, 1996b), we define generative capability as *a firm's high-level routine that reconfigures knowledge embodied within the product generation iteratively, and that re-bundles external and internal knowledge rapidly across product generations to develop the next-generation product.*

Generative capability is particularly relevant for multigenerational product development, which involves the evolution of firms' knowledge and behaviors. Through this process, firms can accumulate knowledge and improve their product development procedures (Eisenhardt & Tabrizi, 1995). Similar to situated learning (Lave, Wenger & Wenger, 1991), learning by doing through multiple iterations is better than less participative learning. By developing multiple generation products, firms' knowledge about the market, technology, and products can be enhanced because of the exploratory learning process (Krishnan & Ulrich, 2001; McGrath, 2001). For example, software companies add several new functions when they update their products to test customers' demand, and there is evidence suggesting that frequent product updates can lead to better performance (Lee & Raghu, 2014; McIlroy, Ali & Hassan, 2016). By integrating external knowledge (such as customer knowledge and competitor knowledge) and internal knowledge, firms can develop superior next-generation products; thus, new market opportunities can be seized (Wang & Chen, 2018). It is conceivable that multi-generational product development involves the development and configuration of various firm capabilities. According to Collis (1994) and Zollo and Winter (2002), capabilities can exist on different levels, where first-order capabilities refer to routines that reconfigure the organizational resource base and second-order capabilities are routines that reconfigure first-order capabilities. To facilitate theoretical precision, we further treat generative capability as a second-order "learning-to-learn capability" (Schilke, 2014).

Although we have proposed the concept of generative capability based on the literature and our observation of leading firms' practices, we still do not know what organizational routines substantiate a firm's generative capability. In other words, how is generative capability manifested in firms' experiential learning? To answer this question, we use a mixed-method research design (e.g., Molina-Azorin, 2012) to first explore the conceptualization and operationalization of generative capability, and then empirically validate them. The following sections report the qualitative and quantitative phases of the study.

Conceptual development through qualitative case study

Research setting and data collection

As the purpose of this phase is to investigate how firms can build their capability for continuous product innovation and to explore firms' behaviors that can reflect generative capability, theoretical sampling is employed to choose cases where the focal phenomenon is likely to occur. A valid case should meet the following criteria: (1) the firm has its own R&D department; (2) the firm has multigenerational product development experience and has launched at least two generations of products; and (3) the firm's products can be physical or software products. Following these criteria, we conducted 25 in-depth face-to-face semi-structured interviews with participants from eight leading Chinese high-technology companies, including Huawei, ZET, Tencent, and Baidu. (See Table 1) between June to August 2015. The interviews lasted for an average of 90 min each, and were audio recorded and later transcribed. Over 120,000 (Chinese) words of qualitative data consisting of interview transcripts and detailed notes were collected from the interviews.

All interviewees were middle or senior managers in the R&D departments of the case firms. For each interviewee, after securing his or her agreement, we held a face-to-face meeting of about two hours with recording. In creating the protocol, we use Kvale (1996) framework of conversational, qualitative interviewing as a guide to ensure that our interviews elicit information relevant to our research questions. Based on these guidelines, we developed an open-ended interview protocol² that focuses on (a) iterative product development, (b) knowledge iteration and integration mechanism, (c) internal factors related to generative capability, and (d) external factors affecting generative capability. During the interviews, we asked the interviewee questions such as "in the process of product development in your company, do you attach great importance to the multi-generational development of products, can you give an example?" and "how do you make sure the next generation of products is always better?"

Data analysis

The data is analyzed through an abductive process in which "empirical observations and surprises are connected to extant theoretical ideas to generate novel conceptual insight and distinctions (Langley, Smallman, Tsoukas & Van de Ven, 2013, p. 11)." Using NVivo 11, three coders of our research team coded the data independently to ensure the reliability of the data analysis. We began with open coding through which first-order codes are generated. We then conducted axial coding to structure first-order codes into second-order themes and higher-level aggregate dimensions (Gioia, Corley & Hamilton, 2013). Once the coding process was completed, the three coders extensively compared and discussed any discrepancies in the coding structure, shifting back to data coding whenever necessary until we reached a consensus on the final data structure Fig. 1. presents the final data structure resulting from the data analysis.³

Conceptual development – structure and dimensions of generative capability

Using an abductive approach that involves taking iterative steps between the data, literature, and a developing set of theoretical ideas (Miles, Huberman, Huberman & Huberman, 1994), we identify three dimensions of generative capability, namely "knowledge acquisition," "knowledge inheritance," and "knowledge updating".

Knowledge acquisition

Knowledge acquisition refers to the process through which firms obtain different types of knowledge from external sources. Exploitation of external knowledge is a critical component of a firms' innovation (Cohen & Levinthal, 1990; Jaziri, 2019). Knowledge acquisition has been studied extensively in the literature on absorptive capacity (Cohen & Levinthal, 1990; Zahra & George, 2002) and market orientation (Jaworski & Kohli, 1993; Morgan, Vorhies & Mason, 2009; Narver & Slater, 1990). The difference is that the absorptive capacity literature emphasizes the absorption of a wide range of external knowledge related to technology and firms' operations (Zahra & George, 2002) while the market orientation literature focuses on the transformation of customer and competitor knowledge into superior products and services (Atuahene-Gima, 2005; Johansson, Raddats & Witel, 2019). Firms' knowledge acquisition not only depends on the individual behaviors of its members, but also requires firm-level routines to identify and synthesize information from different sources and fields (Cohen & Levinthal, 1990). For start-up firms, knowledge acquisition through high-intensity R&D activities or merging is difficult because, at the early stage of development, their available

² Interview protocol is available upon request to the authors.

³ For detailed qualitative evidences, please see Appendix A.

Table 1
Interview sources.

Company	Industry	Main product	Interviewees	Hours
Huawei	Telecommunication equipment manufacturing	Transmission equipment, cellphone, chip	4	7.0
ZTE	Telecommunication equipment manufacturing	Router, base station, cellphone	4	5.0
Tencent	Internet	Software, video game	3	5.0
Baidu	Internet	Software, search service, cloud storage service	2	3.0
KONKA	Electronics	TV, LED, STB	4	4.0
NEPTUNUS	Pharmacy	Medicine	3	3.0
StateMicro	Chip	Micro controller, SOC	2	2.5
CMS	Financial	Financial services/products	3	3.0

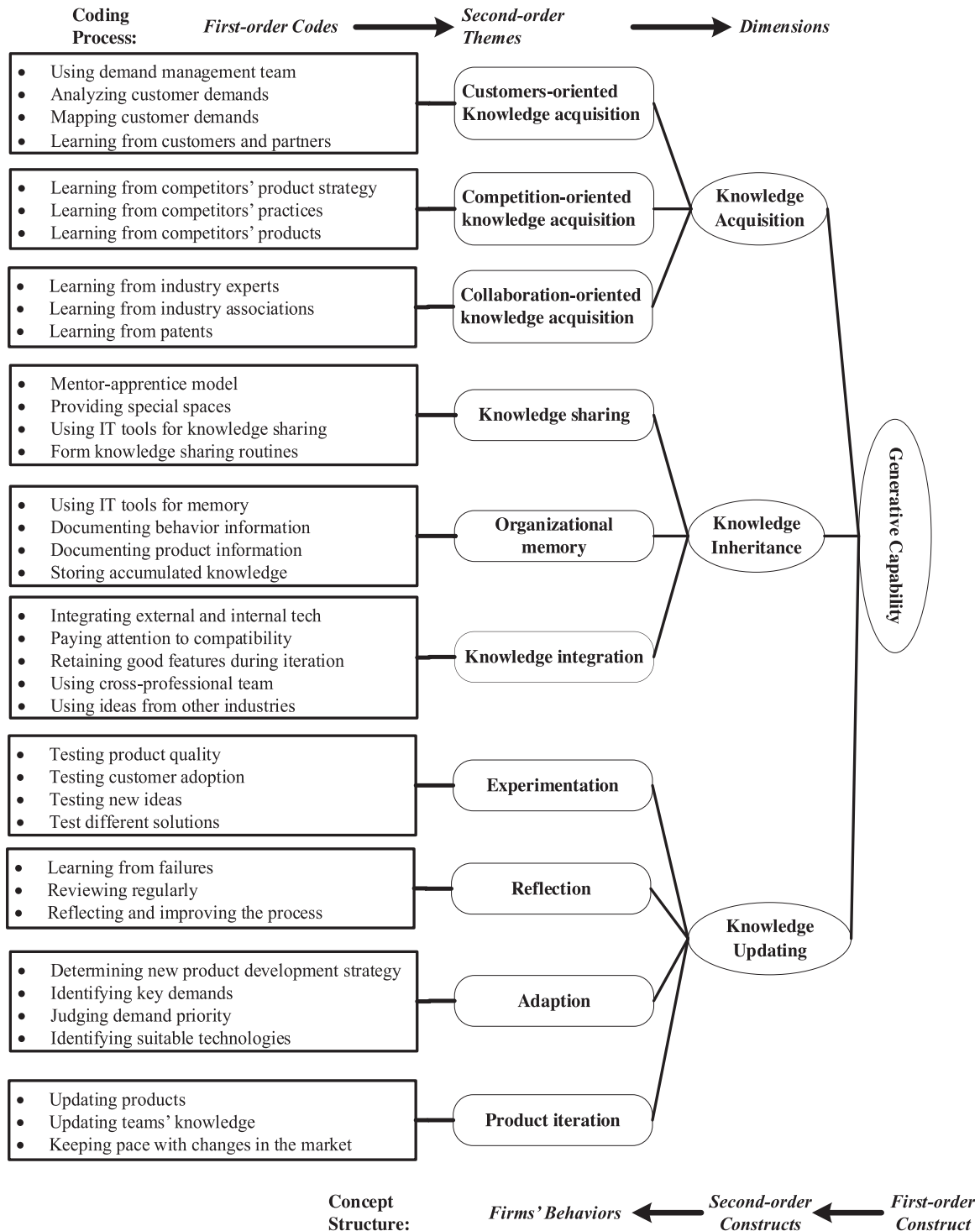


Fig. 1. Structure of generative capability.

resources and technological bases are limited. Under these circumstances, gaining knowledge from customers and competitors to develop multiple generations of products is a favorable strategy because through this process firms can gradually accumulate resources and improve their technological bases. Forming routines to acquire customer knowledge can help firms iteratively improve their products (Philipson, 2020). As a manager stated:

After we have made a product prototype, we go to various regions to test and communicate with users, operators, and partners. They give us feedback, and we continue to adjust and optimize our products.

Through the utilization of customer knowledge, the product can gain an accurate market positioning and better meet customer needs (Aishanty & Emeagwali, 2019).

Firms can also benefit from their competitors' knowledge. First, firms can conduct competition-oriented knowledge acquisition by learning from their competitors' practices and products. A competitor's knowledge may spill out after the competitor exploits the first-mover advantage in the marketplace (Cohen & Levinthal, 1990). Other firms can benefit from knowledge spillover (Operti & Carnabuci, 2014). As one interviewee said:

We also learn from our peer companies; for example, one of our competitors has a number of product lines. We analyzed their practices, and then some practices were used in our product lines after adjustment.

Forming routines to acquire competitors' knowledge can also help firms introduce superior products (Morgan et al., 2009). As a product manager put it,

We will thoroughly experience our products. . . Of course, we will also experience competitors' products to find out their advantages and disadvantages, which requires a proactive approach, that is, to pay close attention to the dynamics of the industry in addition to our work, and we must take the initiative to experience new products in the industry.

This kind of competitor knowledge will help firms set a clear target to surpass. With a deep understanding of rivals' strengths and weaknesses, firms can develop superior products more easily. Second, competitors can also facilitate peers' knowledge transfers. Firms can also conduct collaboration-oriented knowledge acquisition by learning from industry experts, associations, and authorized patents. As a manager stated:

We invite industry experts, sometimes even competitors to train us.

Another technical director also said:

To acquire external knowledge, we go to technical exchanges and training programs organized by the government or associations.

Such routines can help firms keep pace with industry technology trends and make better decisions regarding the direction of technology pre-research.

Knowledge inheritance

Knowledge inheritance refers to routinized organizational processes that facilitate the transfer of knowledge embodied within products across generations. Previous researches on knowledge inheritance are mainly focused on the phenomena of "spin-outs" (entrepreneurial ventures of ex-employees) (Argyres & Moffatt, 2016; Berry, 2015; Furlan & Grandinetti, 2016), that is the knowledge inheritance between incumbents and their spin-outs, while limited attention has been paid to the knowledge inheritance

across different product generations within firms. From our in-depth interviews, we find that knowledge inheritance is a key mechanism for product innovation. This enables firms to roll out new products continually. As a software architect articulated:

Some product features slowly become the advantages of the product; these advantages are often the label of this product, and [these features] will continue to be used in the subsequent product development.

Knowledge inheritance involves firms' learning behaviors, such as knowledge sharing, organizational memory, and knowledge integration. According to the knowledge-based view, Grant (1996a) argues that if employees are mobile, firms' capability depends more on its mechanisms of knowledge integration rather than the extent of employees' specialized knowledge, and the higher the level of common knowledge among the team, the more efficient integration is likely to be. Knowledge sharing is a key mechanism that can increase the level of common knowledge and facilitate knowledge inheritance. Besides its function on promoting innovation (Lin, 2007), knowledge sharing can also help minimize the influence of turnover on product development (Parise, Cross & Davenport, 2006). A senior manager said:

We often use the mentor-apprentice model to ensure that the old staff's experience and know-how knowledge can be inherited by new employees, so that some key technologies can be passed on.

Another programmer also put it as:

The code written by each team member must be reviewed by others. This ensures that if someone leaves, the project can continue as expected.

In addition to individual-level knowledge sharing, firm-level organizational memory is another key mechanism that ensures knowledge inheritance. Organizational memory refers to "actions organizations take to encode, store, and retrieve the lessons they have learned (Flores, Zheng, Rau & Thomas, 2012, p. 644)". It can also benefit knowledge inheritance across product generations (Argote & Ren, 2012; Ginja Antunes & Pinheiro, 2020). As a technical supervisor stated,

We have a technology library that stores all relatively mature technologies that we previously accumulated. They can be obtained at any time if required.

Knowledge in a library is no longer subject to personnel turnover; thus, it can promote the efficiency of knowledge integration Grant (1996a). uses the notion of "knowledge integration" to depict the source of organizational capability. He defines "knowledge" in this notion as "broadly to include both 'explicit' knowledge which can be written down, and 'tacit' knowledge which cannot" (Grant, 1996a, p. 377). The "knowledge integration" in this study mainly refers to the integration of internal knowledge that is already embodied within the existing product and the new externally acquired knowledge used to develop the next-generation product. Knowledge inheritance is also a process of knowledge integration.

An interviewee explained that:

When we plan products, we consider the technologies we have accumulated internally and externally. We gradually integrate external and internal technologies through product planning.

Knowledge integration can help firms improve their next-generation products and, simultaneously, new knowledge can be generated and accumulated during the knowledge inheritance process. In turn,

the accumulated knowledge can be used for the next round of product iterations.

Knowledge updating

Knowledge updating refers to a firm's efforts to continuously test, reflect, and adapt according to internal and external feedback, and to probe into the future (Brown & Eisenhardt, 1997; Miller & Sham-sie, 2001; Yang, Sun & Zhao, 2019). Experimentation involves firm members retaining actions that produce desired results and discarding those that do not (Argyris, 1976; Thomke, 1998). The iterative nature of the experimentation process allows firms to gradually improve the quality of their products (Sosna, Trevinyo-Rodríguez & Velamuri, 2010). A manager from a mobile phone company said:

Before the launch of the new mobile phone, we distributed it to every leader of our company. If there is any problem, they provide feedback immediately. . . This is a constant error correction process to ensure the final quality of the product.

Reflection allows firms to anticipate and exploit early information (Thomke, 2001) and implement continuous improvement through double-loop learning (Argyris & Schon, 1978; Knipfer, Kump, Wessel & Cress, 2013). As a senior manager stated:

Our team recorded the mistakes made during the previous generation of product development. After the project was completed, everyone discussed why we had made these mistakes. Why could we not avoid these mistakes at that time?

Through reflection, firms can obtain knowledge about the cause (why) and process (how) that leads to the success or failure of a product. This knowledge can lead to the development of a better next-generation product.

Adaption is another important activity that helps firms deal with the increasing complexities of fragmenting markets and decide which solution to adopt (Day, 2011). For example, rapid prototyping can be a useful tool for adaptation (Kelly, 2001). It tests copy versions with controlled experiments and continuously asks the market for ideas, concepts, and formulations that are working or failing, which is often an inexpensive and fast way to accelerate product development and meet customer needs (Day, 2011; Jaziri, 2019; Thomke, 1998). For example, a product manager articulated the following.

If there are too many options, such as customer needs, how to make a choice? Under these circumstances, the system engineer performs assessment and feasibility analysis, during which experiments are conducted to determine the feasible solution.

This idea is consistent with the classic evolutionary models of variation, selection, and retention. Using different variations of the product prototype, firms can conduct market testing on a small scale to weed out bad solutions and help them continuously adapt their solutions to find the optimal one. Essentially, trial and error is an important process for firms to "update" their knowledge stock. Adaption determines the direction of product innovation by reaching a decision about the priority of customer demands (Hoornaert, Ballings, Malthouse & Van den Poel, 2017), which is important because excessive unordered customer knowledge may cause cognitive overload and inhibit innovation performance (Tang & Marinova, 2019). The formation of a specific team or mechanism is helpful. As one interviewee stated:

Taking software development as an example, after a version is introduced, there will be some changes in demands, such as adding or removing some functions. We have a change control committee that judges and verifies the demands and decides which demand should

be integrated into the next version [of the software]. We follow this direction to develop the next version.

In summary, from the above qualitative case analysis, we find that firms' generative capability can be reflected in three dimensions: "knowledge acquisition," "knowledge inheritance," and "knowledge updating." Although we elaborate on the three dimensions separately, it should be noted that they are not independent of each other and there is no chronological order between the three dimensions. Knowledge acquisition provides an important knowledge base for knowledge inheritance and updating. During a firms' multigenerational product development process, knowledge inheritance can occur simultaneously with knowledge acquisition and updating. These three dimensions represent three sets of routines for firms to conduct continuous product innovation.

Quantitative examination of construct validities

Empirical setting and data collection

The goal of this phase is to develop measurement scales for generative capability and its components, and quantitatively verify the construct and criterion validity of this novel concept. For this purpose, we collect firm-level perceptual data from China, the world's largest emerging economy. The target population comprises firms with documented R&D activities from various industries. We used the industrial corporation database of the National Bureau of Statistics of China (NBS) to identify the target population. The NBS dataset contains in-depth information about the firm's characteristics and financial data (such as address, email, phone number, and R&D investment) with annual sales of over 5 million RMB, which has been previously utilized in strategic management research (Tan & Peng, 2003). We first eliminated firms with zero or missing R&D data from our sampling frame as well as those with incomplete or missing contact information. We then randomly selected 1000 firms as targets for our managerial survey following the procedures recommended by Dillman (2000).

Our survey questionnaire was developed in Chinese, which is the native language of the authors. In addition to the newly developed construct of generative capability, all other constructs included in our survey questionnaire were measured using existing scales validated in prior studies. To discriminate GC from other related measures, we collected data on incremental innovation capability (IIC), radical innovation capability (RIC), pure imitation (PI), creative imitation (CI), and organizational learning.⁴ Measurements of IIC and RIC were adopted from Subramaniam and Youndt (2005) and each construct had three items. Scales for PI and CI were derived from Lee and Zhou (2012) with three items for each scale. For organizational learning, we used the five-dimensional scale developed by Flores et al. (2012), which includes four items for information acquisition (IA), four items for information distribution (DT), four items for information interpretation (ITP), five items for information integration (ITG) and six items for organizational memory (MEMO). In cases where the existing scales are presented in English, we translate them into Chinese by following the procedures demonstrated by Brislin (1980) to preserve content validity. We pretested our questionnaire with a group of 24 executive MBA students whose feedback led us to adjust the wording of some questions. We conducted our survey through the internet during the period of January to August 2017. We targeted middle and senior managers of the R&D department as qualified respondents. Among the 1000 questionnaires delivered, 212 were completed and returned as usable responses, achieving an effective response rate of 21.2%. The firms included in our final research sample

⁴ Please refer to their references or Appendix B for detailed items.

Table 2
Sample firms.

Firm information		Frequency	Percentage
Size (number of employees)	<=20	15	7.08%
	20~100	46	21.70%
	100~300	65	30.66%
	300~1000	49	23.11%
	>=1000	37	17.45%
Industry	Electronics	38	17.92%
	Equipment manufacturing	32	15.09%
	Information technology	38	17.92%
	New material	11	5.19%
	New energy	10	4.72%
	Financial	15	7.08%
	Pharmacy	4	1.89%
	Others	64	30.19%
Age	<=1	2	0.94%
	1~5	35	16.51%
	5~10	49	23.11%
	>=10	126	59.43%

came from a wide range of industries, including equipment manufacturing, electronics, IT, new materials, and pharmacies. We perform non-response bias tests by comparing key attributes, such as size, age, and financial performance, between responding and non-responding firms, as obtained from the NBS database, as well as those between early respondents and late respondents. The t-tests showed that there are no significant differences between the groups regarding key attributes, such as age and industry. A summary of the sample firms' information is presented in Table 2.

Measuring generative capability

We used 17 newly developed items to measure the dimensions of generative capability. The development of these items is informed by both relevant literature concerning innovation, such as organizational learning (Flores et al., 2012) and market orientation (Jaworski & Kohli, 1993), and the qualitative data collected from the qualitative phase of this research. Examples of qualitative data in scale development include: "We will thoroughly experience our products... of course, we will also experience competitors' products to find out their advantages and disadvantages"; "[personnel] should be stable. Everyone likes to work with someone who has worked for years. There is a tacit understanding of each other... in comparison, if new members keep coming, you will have to continue tutoring so that they can adapt to the team. It can be annoying and affect work efficiency"; and "the initial idea was very rough... then you put it on the market and get recognized... the whole process is actually the course of trial and error."

To ensure content validity, the initial measurement items of generative capability were first sent to a panel of academics and managers to check for representativeness, ease of use, identify unclear wording, question double-barreled descriptions, and ensure that the measurement items would be interpreted accurately. Based on their feedback, some items were removed and others were modified. After several rounds, a final version of the 17 items was generated and included in the questionnaire. Among the 17 items, five items were intended to measure knowledge acquisition, six for knowledge inheritance, and six for knowledge updating. The completed survey data allowed us to perform a range of analyses to further purify the measurement scales and examine the construct and criterion validities of generative capability.

Formative vs reflective

According to MacKenzie, Podsakoff and Podsakoff (2011), constructs are not inherently formative or reflective in nature and mostly depend on the researcher's theoretical expectations. Regarding

multidimensional constructs, the main concern is the relationship between the sub-dimensions and the focal construct (Bollen & Lennox, 1991; Jarvis, MacKenzie & Podsakoff, 2003). Based on our conceptualization of generative capability, we propose that the empirical measurement of generative capability is best represented by a reflective first-order, reflective second-order model, whereby individual items (see Table 3) are reflective indicators of each dimension, and the dimensions are in turn reflective indicators of the latent second-order construct—generative capability (Jarvis et al., 2003; Zhang, Waldman, Han & Li, 2015). We arrive at this understanding for the following reasons. First, from an ontological perspective, "the realist interpretation of a latent variable implies a reflective model, whereas constructivist interpretations are more compatible with a formative model" (Borsboom, 2005, p. 63). We adopt the realist approach and view generative capability as existing at a deeper and more embedded level within organizations than its three subdimensions. These sub-dimensions are viewed as manifestations of generative capability. Second, when deciding the relationship between a construct and its sub-dimensions, scholars suggest that it is very helpful to ask, "Is it possible for a change in the focal construct to be associated with a change in only one of the sub-dimensions?" (Jarvis et al., 2003; MacKenzie et al., 2011; Wong, Law & Huang, 2008). If a change in only one of the sub-dimensions could be associated with a change in the focal construct, which means that each sub-dimension can independently influence the construct, then a formative structure is more appropriate (MacKenzie et al., 2011). If we look at the case of generative capability, we find that without the other two dimensions, changes in only one sub-dimension can hardly cause a change in generative capability. These three sub-dimensions must join forces to help firms roll out better products. In contrast, if we view the three sub-dimensions as manifestations of generative capability, which means that a change in generative capability would produce a change in all three sub-dimensions, then a reflective structure is more appropriate. In other words, the shared variance of the three sub-dimensions defines the variance in generative capability (Bollen, 1989). From the above, we believe that a reflective structure is more suitable for the relationship between generative capability and its three sub-dimensions.

Construct validity test

Using the SPSS random sample program, we separated the sample into two equal parts ($N = 106$). We used one subsample to conduct exploratory factor analysis (EFA) and another for confirmatory factor analysis (CFA). We first conducted an EFA procedure in SPSS, through which we removed three items with cross-loadings greater than 0.4. This procedure resulted in a 14-item solution: four for knowledge acquisition, four for knowledge inheritance, and six for knowledge updating. The association between the items and dimensions is consistent with our findings from the qualitative phase. The eigenvalues of the three factors range from 1.01 to 8.07, with 75.2% of the total variance explained. Then, we conducted CFA using Mplus software Table 3. shows the items, EFA factor loadings, and percentages of the variance explained.

We subject this proposed measurement model to confirmatory factor analysis using structural equation modeling (SEM), using the other subsample ($N = 106$), as well as the full sample ($N = 212$). This proposed measurement model achieves sufficient fitness to the data, as evidenced by the non-significant model chi-square and CFI/TLI above 0.95 (see Table 4). Construct consistency validity is illustrated by the significant factor loadings between items and their respective dimensions as well as between dimensions and generative capability.

To assess the discriminant validity of the dimensions, we compared our proposed measurement model with a number of alternatives. The first alternative is a first-order three-factor model, which is mathematically equivalent to a second-order three-factor model (Bollen, 1989). We used it as a basis to compare the fit of the other

Table 3
Exploratory and confirmatory factor analyses of generative capability.

Factors and items	EFA Loadings (N = 106)			CFA Loadings (N = 106)
	1	2	3	
1. Knowledge acquisition				0.83
(1). We regularly meet customers to learn about their current and potential needs for new products.	0.76	0.31	0.20	0.65
(2). We regularly use research techniques such as focus groups, surveys, and observations to gather customer information.	0.70	0.06	0.34	0.55
(3). We regularly collect and integrate information about the products and strategies of our competitors.	0.78	0.22	0.25	0.76
(4). Our knowledge of current and potential competitors' strengths and weaknesses are very thorough.	0.81	0.14	0.25	0.78
2. Knowledge inheritance				0.68
(1). We reconfigure the knowledge accumulated in the previous generations of products' development and apply them to the development of next-generation products.	0.27	0.80	0.30	0.78
(2). When we build an R&D team, we usually include personnel who have previous experience in the development of similar products.	0.08	0.84	0.28	0.62
(3). We attach great importance to customer feedback on existing products to meet customer needs in the next generation product.	0.27	0.79	0.28	0.77
(4). When we are developing the next generation products, we consciously integrate the previously accumulated knowledge and new knowledge we gained.	0.17	0.77	0.38	0.68
3. Knowledge updating				0.85
(1). During the product development process, we often set some indicators to examine the gap between product features and user needs.	0.27	0.36	0.75	0.64
(2). During the product development process, we often set some indicators to find out if the product development procedure can be improved.	0.33	0.36	0.75	0.81
(3). During the product development process, we usually set some assumptions and test them through experiments to find out whether the product meets the market demand.	0.26	0.33	0.77	0.73
(4). During the product development process, we usually invite some leading users to verify whether the product meets their demands.	0.36	0.30	0.77	0.71
(5). After the introduction of a new product, we often set some key market indicators (such as customer recommendation rate) to examine the fit between product and customer needs so that we can set new marketing strategies.	0.36	0.25	0.77	0.79
(6). For external new technologies, we usually conduct rigorous tests to verify whether they can be adopted by us.	0.18	0.26	0.76	0.64
% variance explained	57.60	10.40	7.20	

Note: The extraction method for EFA is principal component analysis. Varimax was used as the rotation method. The bold figures represent the highest factor loadings of indicators on factors from the EFA.

alternative models, thus illustrating the discriminant validity of the dimensions. The second alternative is a first-order one-factor model, in which all 14 items are associated with the generative capability factor, equivalent to Harman's one-factor test of common method variance. The other alternatives are first-order two-factor models, in which the three dimensions are combined two at a time. If the one- and two-factor alternatives show significantly worse model fit than the three-factor model, we can conclude that there is evidence of discriminant validity for the three dimensions of generative capability.

Table 4 reports the results of the tests on these alternative measurement models. The best-fitting models are the proposed measurement model and its first-order equivalent ($\chi^2 = 74.39/77.19$, $df = 65$, CFI = 0.99/0.98, TLI = 0.99/0.99 and RMSEA = 0.04/0.03). The one-

factor model shows the worst fit to the data ($\chi^2 = 205.46/302.21$, $df = 68$, CFI = 0.79/0.86, TLI = 0.71/0.82 and RMSEA = 0.14/0.13). The two-factor models achieve poor model fit as well, evidenced by significant model chi-squares and other indicators. Moreover, chi-square difference tests show that the alternative models had model fit significantly inferior to the proposed three-factor model. In the full sample, the Cronbach's alphas for knowledge acquisition, knowledge inheritance, and knowledge updating are 0.81, 0.87, and 0.91 respectively. The three dimensions' average variances extracted (AVE) are all above 0.5 (0.62, 0.59, and 0.62 respectively), indicating that the variances in these dimensions account for the majority of variance in the indicators. The square roots of AVE (from 0.77 to 0.79) are larger than the correlation coefficients among the three factors

Table 4
Comparison of alternative measurement models of generative capability.

Model	χ^2	df	p-value	$\Delta\chi^2$	CFI	TLI	RMSEA	SRMR
<i>Subsample for CFA (N = 106)</i>								
Second-order model (proposed)	74.39	65	0.199		0.99	0.98	0.04	0.06
First-order, three-factor model	74.39	65	0.199		0.99	0.98	0.04	0.06
One-factor model	205.46	68	0.000	131.07*** (3)	0.79	0.71	0.14	0.10
First-order, two-factor model (KA+KI, KU)	140.19	67	0.000	65.80*** (2)	0.89	0.85	0.10	0.08
First-order, two-factor model (KA+KU, KI)	125.87	67	0.000	51.48*** (2)	0.91	0.88	0.09	0.08
First-order, two-factor model (KA, KI+KU)	157.42	67	0.000	83.03*** (2)	0.86	0.81	0.11	0.09
<i>Full sample (N = 212)</i>								
Second-order model (proposed)	77.19	65	0.140		0.99	0.99	0.03	0.03
First-order, three-factor model	77.19	65	0.140		0.99	0.99	0.03	0.03
One-factor model	302.21	68	0.000	225.02*** (3)	0.86	0.82	0.13	0.07
First-order, two-factor model (KA+KI, KU)	202.15	67	0.000	124.96*** (2)	0.92	0.89	0.10	0.08
First-order, two-factor model (KA+KU, KI)	142.82	67	0.000	65.63*** (2)	0.96	0.94	0.07	0.05
First-order, two-factor model (KA, KI+KU)	236.79	67	0.000	159.60*** (2)	0.9	0.87	0.11	0.06

Note: KA = knowledge acquisition; KI = knowledge inheritance; KU = knowledge updating.

*** $p < 0.001$ (two-tailed).

Table 5

Confirmatory factor analysis results for scale validation of generative capability.

Model	χ^2	df	p-value	$\Delta\chi^2$	CFI	TLI	RMSEA	SRMR	AVEs
<i>Models used to discriminate GC from alternative measures</i>									
Three-factor model: GC and Subramaniam and Youndt (2005)'s incremental (IIC) and radical innovation capability (RIC)	35.41	24	0.060		0.99	0.98	0.05	0.03	AVE _{GC} =0.53
One-factor model: GC, IIC and RIC merged	97.79	27	0.000	62.38*** (3)	0.91	0.89	0.11	0.05	AVE _{IIC} =0.59
Two-factor model (IIC+RIC, GC)	60.50	26	0.000	25.09*** (2)	0.96	0.94	0.08	0.04	AVE _{RIC} =0.55
Two-factor model (IIC+GC, RIC)	63.89	26	0.000	28.48*** (2)	0.95	0.94	0.08	0.04	
Two-factor model (RIC+GC, IIC)	94.17	26	0.000	58.76*** (2)	0.92	0.89	0.11	0.05	
Three-factor model: GC and Lee and Zhou (2012)'s pure (PI) and creative imitation (CI)	31.06	24	0.150		0.99	0.99	0.04	0.04	AVE _{GC} =0.53
One-factor model: GC, PI and CI merged	433.03	27	0.000	401.97*** (3)	0.41	0.21	0.27	0.21	AVE _{PI} =0.59
Two-factor model (PI+CI, GC)	176.72	26	0.000	145.66*** (2)	0.78	0.70	0.17	0.15	AVE _{CI} =0.53
Two-factor model (PI+GC, CI)	409.78	26	0.000	378.76*** (2)	0.44	0.23	0.26	0.21	
Two-factor model (CI+GC, PI)	73.21	26	0.000	42.15*** (2)	0.93	0.91	0.09	0.08	
<i>Models used to validate GC dimensions</i>									
Two-factor model: KA dimension and information acquisition (IA)	23.73	10	0.008		0.98	0.95	0.08	0.03	AVE _{KA} =0.53
One-factor model: KA and IA merged	37.27	11	0.000	13.54*** (1)	0.95	0.91	0.11	0.04	AVE _{IA} =0.47
Four-factor model: KI, information distribution (DT), information integration (ITG) and organizational memory (MEMO)	250.37	144	0.000		0.95	0.94	0.06	0.05	AVE _{KI} =0.50
One-factor model: KI, DT, ITG and MEMO merged	539.81	150	0.000	289.44*** (6)	0.82	0.79	0.11	0.08	AVE _{DT} =0.57
Three-factor model: (KI+DT, ITG, MEMO)	428.54	147	0.000	178.17*** (3)	0.87	0.85	0.10	0.07	AVE _{ITG} =0.51
Three-factor model: (KI+ITG, DT, MEMO)	399.11	147	0.000	148.74*** (3)	0.88	0.86	0.09	0.07	AVE _{MEMO} =0.56
Three-factor model: (KI+MEMO, DT, ITG)	417.82	147	0.000	167.45*** (3)	0.87	0.85	0.09	0.07	
Two-factor model: KU dimension and information interpretation (ITP)	37.99	32	0.215		0.99	0.99	0.03	0.03	AVE _{KU} =0.57
One-factor model: KU and ITP merged	78.79	33	0.000	40.8*** (1)	0.96	0.94	0.08	0.04	AVE _{ITP} =0.54

Note: ***p < 0.001 (two-tailed).

(maximum 0.66), demonstrating the discriminant validity of the three-factor structure.

Following Zhang et al. (2015), to validate the overall construct, we used the dimensional scores of GC and item scores of IIC, RIC, PI, and CI to discriminate GC from the four measures in factor analyses. We ran two three-factor, two one-factor, and six two-factor models (Table 5). The comparison results showed that the GC differs from the four measures. The square root of GC's AVE (0.73) is greater than the correlations between GC and the other four measures (maximum 0.67), further demonstrating its high discriminant validity. Five dimensions of organizational learning were used to test the discriminant validity of the GC dimensions. As shown in Table 5, the KA dimension differs from IA; the KI dimension differs from DT, ITG, and MEMO; and the KU dimension differs from ITP. All results suggest that GC can be discriminated from other related measures and has high discriminant validity.

Criterion validity test

We verify the criterion validity of generative capability by testing its association with expected outcomes, namely product innovation performance. If our conceptualization of generative capability captures the higher-level capability of firms to succeed in continuous product innovation, we expect a strong positive association between generative capability and product innovation performance after controlling for other potential explanations of firm' innovation outcomes.

Theoretically, generative capability can foster innovation performance for two reasons: First, through the knowledge acquisition, inheritance, and updating processes, firms can gain a thorough understanding on the key market demand and available technology they can use (Van Oorschot, Eling & Langerak, 2018). Along this process, firms can build a close relationship with customers and foster customer engagement, which is beneficial for relevant marketing communication (Alvarez-Milán, Felix, Rauschnabel & Hinsch, 2018; Lee, Naylor & Chen, 2011). Using an iterative product development approach, the product team can learn thoroughly from concrete experimental results and customer feedback (Cui & Wu, 2017). Generative capability can enable firms to integrate key customer

demands and suitable technological knowledge to produce better next-generation products. Therefore, generative capability can help firms improve their products incrementally and create higher innovation performance.

Second, the relationship between technological innovation and time is a series of stair-stepping S-curves (Foster, 1986; Utterback, 1994). An existing technology (T1) initially promotes benefits rapidly when the technology is new and then slows as the technology matures. At a certain point in the history of T1, a new technology, T2, emerges. Initially, the benefits of T2 are inferior to those of T1. However, with the development of T2, its benefits increase rapidly and will surpass those of T1 at a point where the product based on T2 becomes a radical product innovation (Chandy & Tellis, 1998). Having experienced several product generations' development, the knowledge updating process would gradually help the R&D team gain a deep understanding on the utmost of the technology currently in use and the potential of new technology emerging in the industry (Dattée, Alexy & Autio, 2018). This prompts the R&D team to reserve promising technologies in advance. During the multi-generational product development process, listening to customers and learning from the past can help firms continually check and update their knowledge of customer needs and the technologies used to fulfill these needs. This knowledge provides the foundation for firms to break existing frames of understanding regarding customer needs and technology when a new technology emerges (Joshi, 2016). In other words, generative capability enables the search for a deeper understanding of customer needs and alternative technologies that can better address these needs. Thus, generative capability can improve innovation performance by stimulating radical innovation.

Given the above reasoning, to test this criterion validity, we constructed and tested a set of structural models linking generative capability to its expected outcomes. As Fig. 2 shows, we use two different sets of scales to measure firms' innovation performance. The first set of scales differentiates between incremental and radical innovation performance, each measured by four items adopted from prior studies (Atuahene-Gima, 2005; Chandy & Tellis, 1998). The second set is a reflective first-order and reflective second-order measurement for product innovation performance, based on seven items measuring

innovation efficacy and another four items measuring innovation efficiency adopted from prior studies (Alegre & Chiva, 2008; Alegre, Lapedra & Chiva, 2006). We also included six control variables based on previous research on the factors influencing firm innovation outcomes (Narver & Slater, 1990). Ease of entry (ENTRY) assesses the degree of difficulty of new entry into the market. Threat of substitutes (SUBS) measures the difficulty of product substitution. Bargaining power of buyers (BPOW) assesses the extent to which customers are able to negotiate lower prices in the industry. Bargaining power of suppliers (SPOW) measures the extent to which the manufacturer is able to negotiate lower prices from its suppliers. Relative size (RSIZE) examines the annual sales revenue of a firm relative to its largest competitor. Relative cost (RCOST) assesses a firm's operating costs relative to those of its largest competitor.

Table 6 presents the results of SEM analysis. We tested two models using different sets of innovation measurements: Model 1 uses incremental innovation performance and radical innovation performance as dependent variables, returns a χ^2 ratio of 1.46, a CFI of 0.96, a TLI of 0.95, an RMSEA of 0.05, and an SRMR of 0.07; Model 2 uses a second-order product innovation performance construct as the dependent variable, returns a χ^2 ratio of 1.34, CFI of 0.96, TLI of 0.96, RMSEA of 0.04, and SRMR of 0.07. Both models showed a good model fit according to the cutoff criteria proposed by Hu and Bentler (1999). In Model 1, there was a positive and significant relationship between generative capability and IIP ($b = 0.89, p < 0.001$), and generative capability had a positive and significant association with RIP ($b = 0.76, p < 0.001$). In Model 4, the relationship between generative capability and PIP is positive and significant ($b = 0.83, p < 0.001$). Overall, we found a strong positive association between generative capability and innovation performance, indicating the criterion validity of this novel construct.

Discussion

Theoretical contributions

This study makes three theoretical contributions to the literature: First, the literature on product innovation has focused on the factors contributing to the development a specific new product (Slater et al., 2014; West & Bogers, 2014). Comparatively less attention has been paid to continuous product innovation. From a knowledge-based perspective, we conceptualize a novel concept of generative capability, a unique capability by which firms organize experiential learning activities to achieve continuous product innovation. We elaborate on its three dimensions: "knowledge acquisition," "knowledge inheritance," and "knowledge updating." Knowledge acquisition can create a knowledge base for product iteration, knowledge inheritance depicts the knowledge transformation and integration process across different product generations, and knowledge updating can ensure that the direction of product innovation is in line with key customer demands and appropriate technology. Thus, we extend the product innovation literature by exploring the sources and implementation of firms' continuous product innovations. Simultaneously, this study enriches the literature on knowledge-based views.

Second, the conceptualization of generative capability contributes to resolving the paradox of organizational capability. Organizational capabilities are paradoxical because, on the one hand, they provide a reliable problem-solving architecture that can facilitate the formation of sustainable competitive advantages (Helfat & Peteraf, 2003). On the other hand, owing to path dependency and organizational inertia, capability-driven behaviors are likely to bind the organization to the past, which could be unfavorable in the face of a changing environment (Schreyögg & Kliesch-Eberl, 2007). We believe that generative capability can resolve this paradox to a certain extent. As a high-order capability, generative capability emphasizes the reconfiguration and iteration of knowledge and processes. Along with the multi-

generational product development journey, firms can continually upgrade and renew their knowledge and product development processes based on the information acquired and experimental results. Therefore, generative capability can relieve the capability paradox by reducing the influence of organizational inertia.

Third, previous studies on continuous production innovation are primarily interpretative and qualitative studies that have been fragmented in different research fields. This might be partly due to the lack of core concepts and scales in relation to continuous product innovation. In this study, we conceptualized generative capability and developed its measurement scales. According to Collis (1994) and Zollo and Winter (2002), generative capability is conceptualized as a second-order "learning-to-learn capability" (Schilke, 2014). Corresponding to this conceptualization, we develop a "reflective first-order, reflective second-order" measurement model for generative capability and test its reliability and validity. These efforts lay a solid foundation for further research on continuous production innovation.

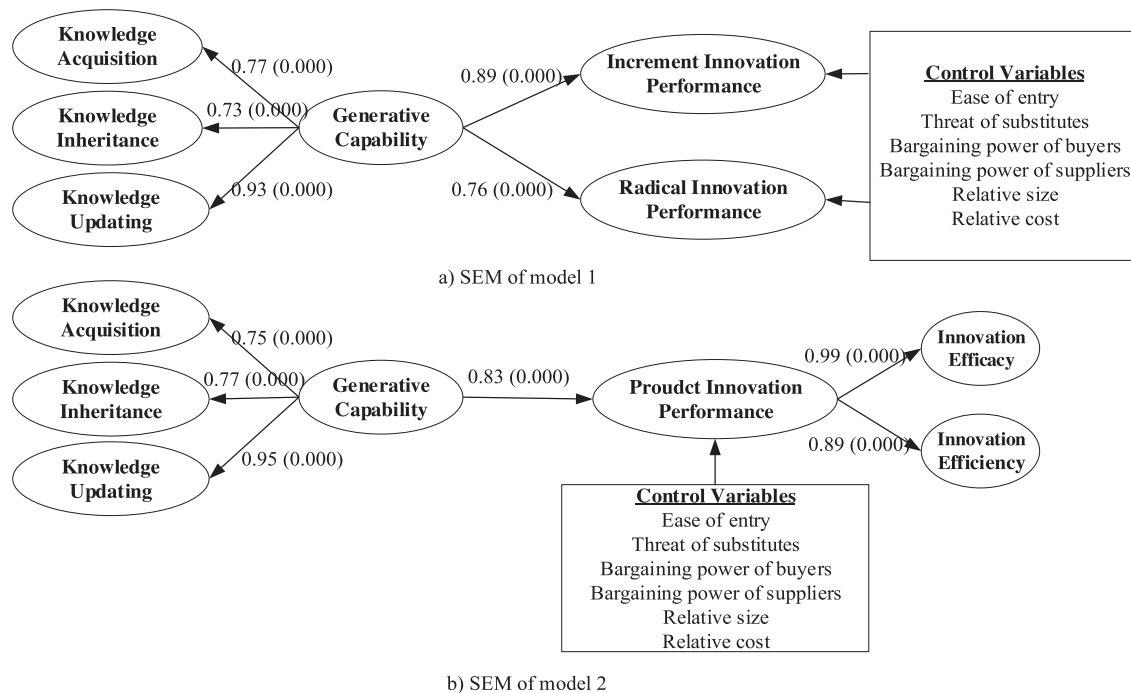
Managerial implications

Many firms consider R&D to be a bottomless pit that can never be filled because of its high investment and failure rate. Generative capability can help firms design reliable long-term product and technology roadmaps based on customer needs, competitor information, and experimental results. Firms can continuously improve their products and roadmaps during the multi-generational product development process, which can lower the R&D cost and make the investment more efficient because they are more likely to address the right customer needs and use appropriate technology. Firms' learning capabilities can also be simultaneously improved at the same time.

Our concept of generative capability can also guide managers in iterative product development. This means that in some cases, the quality of products can take a back seat. For example, a manager stated that:

Sometimes, to capture the market quickly, we push out a relatively low-quality but usable product and then gradually improve it based on customers' needs and rivals' actions. If we continue to improve quality instead of launching the product, our rivals will take over the market.

Our concept also grounds the idea of a *minimum viable product* (MVP) in lean startups on a solid theoretical foundation (Ries, 2011; Yang et al., 2019). MVP refers to "an early version of a new product that allows a team to collect the maximum amount of validated learning (learning based on real data gathering rather than guesses about the future) about customers" (Ries, 2011, p. 83). Firms should take the initiative to focus on customers' demands and rivals' movements and build a specific team (such as a demand management team) to find the right target to surpass and meet the core needs. Compared to the traditional R&D model, which does not launch the product in the market until it reaches a relatively high degree of completion, the practice of a minimum viable product not only has the advantage of quickly seizing the market, but also can lower the risk/cost of R&D failure. Before launching the first version of the products, managers should set the criterion for "minimum viability." Through market surveys or experiments, managers can determine which customer needs should be met and what functions should be achieved in an MVP. When launching early versions of products, it is important to let customers know the progress of product improvement and encourage them to keep using it and provide feedback. Managers can use incentive measures to promote the engagement of early users such as early bird discounts, rewards for reporting bugs, and special gifts.



Note: p-values are shown in brackets

Fig. 2. Structure equation models
Note: p-values are shown in brackets.

Table 6
SEM analysis for criterion validity test.

	MODEL 1 IIP	RIP	MODEL2 PIP
Generative capability	0.89*** (0.04)	0.76*** (0.04)	0.83*** (0.04)
Ease of entry	0.06 (0.06)	0.01 (0.06)	-0.01 (0.05)
Threat of substitutes	-0.11 (0.06)	-0.06 (0.06)	0.05 (0.05)
Bargaining power of buyers	0.02 (0.06)	0.05 (0.06)	0.00 (0.05)
Bargaining power of suppliers	0.06 (0.06)	0.05 (0.06)	0.16** (0.05)
Relative size	0.14* (0.07)	0.26*** (0.06)	0.27*** (0.05)
Relative cost	0.02 (0.06)	0.06 (0.06)	0.02 (0.05)
χ^2 (df)	313.96 (215)		494.46 (368)
CFI	0.96		0.96
TLI	0.95		0.96
RMSEA	0.05		0.04
SRMR	0.07		0.07

Note: Standard errors are shown in parentheses; IIP = incremental innovation performance; RIP = radical innovation performance; PIP = product innovation performance.

* $p < 0.05$.

** $p < 0.01$.

*** $p < 0.001$ (two-tailed).

Limitation and future research

Although this study has contributions and implications, it also has several limitations. First, in the generative process across multiple product generations, organizational semi-structures can support this continual change and a rhythm can be created (Brown & Eisenhardt, 1997). Future studies could explore which type of organizational structure provides better support for generative capability. Second, the sample survey was composed of samples from China, which constrains the generalization of the results to other countries.

It would be interesting to see if the findings of this study can be extended to other countries. Although most quantitative studies collect data from surveys, the use of multiple data sources may enhance the validity of the outcomes, which is also an avenue for future research. Third, the results of this study were based on product-related interviews or quantitative measures. Future studies could consider other innovation contexts to determine whether generative capability can be applied to generic discussions of innovation. Finally, the study applies a cross-sectional design, which implies that causal relationships may not be robust. Future research could adopt a longitudinal design to test these results.

Conclusion

How can firms continuously and sustainably conduct product innovation? We conceptualize generative capability, which involves knowledge acquisition, knowledge inheritance, and knowledge updating. We believe that generative capability can be a path toward continual innovation. Firms can ultimately build their own core competencies and gain sustained competitiveness through the development of generative capabilities.

Declaration of Competing Interest

none

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Appendix A. Qualitative evidence for generative capability

Dimensions	2nd order Codes	Selected Evidence on 1st order Codes
Knowledge acquisition	Customers-oriented Knowledge acquisition	<i>Using demand management team</i> “We have a dedicated demand management team. Before developing a product, we will fully explore customer needs, especially the needs of leading users. We must understand what they care about and what they need.”
		<i>Analyzing customer demands</i> “The first thing is to tap customer demand. Our system engineers and architects will perform in-depth analysis to decompose customer needs.”
	Competition-oriented knowledge acquisition	<i>Mapping customer demands</i> “We will make a roadmap for product development based on market demand. In fact, when the first generation of products was just launched, the second generation of products was already under development because our understanding of market needs is clear.”
		<i>Learning from customers and partners</i> “After we have made a product prototype, we go to various regions to test and communicate with users, operators, and partners. They give us feedback, and we continue to adjust and optimize our products.”
		<i>Learning from competitors' product strategy</i> “We will analyze our competitors' product strategy to see how they segment the market. In other words, we want to see what kind of products do they put out for the low, medium, and high-end market.”
		<i>Learning from competitors' practices</i> “We also learn from our peer companies; for example, one of our competitors has a number of product lines. We analyzed their practices, and then some practices were used in our product lines after adjustment.”
	Collaboration-oriented knowledge acquisition	<i>Learning from competitors' products</i> “We will thoroughly experience our products. . . Of course, we will also experience competitors' products to find out their advantages and disadvantages, which requires a proactive approach, that is, to pay close attention to the dynamics of the industry in addition to our work, and we must take the initiative to experience new products in the industry.”
		<i>Learning from industry experts</i> “We invite industry experts, sometimes even competitors to train us.”
Knowledge inheritance	Knowledge sharing	<i>Learning from industry associations</i> “To acquire external knowledge, we go to technical exchanges and training programs organized by the government or associations.”
		<i>Learning from patents</i> “We have a lot of cross-licensing patents with our competitors. In addition to reducing costs, it also provides opportunities to learn from each other”
		<i>Mentor-apprentice model</i> “We often use the mentor-apprentice model to ensure that the old staff's experience and know-how knowledge can be inherited by new employees, so that some key technologies can be passed on.”
		<i>Providing special spaces</i> “Our company has many public areas, such as cafes, gyms, smoking rooms, where everyone can relax. Many inspirations are collided and generated in these places.”
		<i>Using IT tools for knowledge sharing</i> “We have an internal sharing platform. After each person completes his own project, he must summarize his work and send it to the internal network for sharing. This document must be written in a straightforward manner.”
		<i>Form knowledge sharing routines</i> “The code written by each team member must be reviewed by others. This ensures that if someone leaves, the project can continue as expected.”
	Organizational memory	<i>Using IT tools for memory</i> “We use specialized software to store all kinds of bugs during product development.”
		<i>Documenting behavior information</i> “After finishing their work, everyone in our project team must deliver a document to make it clear what they have done.”
	Knowledge integration	<i>Documenting product information</i> “After we have completed the development of a generation of product, there will be a document a web page to describe the details of the product features. It is very convenient if we want to review later.”
		<i>Storing accumulated knowledge</i> “We have a technology library that stores all relatively mature technologies that we previously accumulated. They can be obtained at any time if required.”
		<i>Integrating external and internal technologies</i> “When we plan products, we consider the technologies we have accumulated internally and externally. We gradually integrate external and internal technologies through product planning.”
		<i>Paying attention to compatibility</i> “We attach great importance to the compatibility of product features in the product development process. In this way, if it is available, new technologies can be easily integrated into products through product upgrades.”
		<i>Retaining good features during iteration</i> “Some product features slowly become the advantages of the product; these advantages are often the label of this product, and [these features] will continue to be used in the subsequent product development.”

(continued)

Knowledge updating	Experimentation	<i>Using cross-professional team</i>
		"A cross-professional team will be formed in the product development process. There are members from market planning, R&D, logistics, quality, and testing. Different people can look at the problem from different professional perspectives."
		<i>Using ideas from other industries</i>
		"We produce televisions, but we will often look at other industries to see if there is any new idea that can be used on televisions."
		<i>Testing product quality</i>
	Reflection	"Before the launch of the new mobile phone, we distributed it to every leader of our company. If there is any problem, they provide feedback immediately. . . This is a constant error correction process to ensure the final quality of the product."
		<i>Testing customer adoption</i>
		"If we have a new product to be launched, we will not introduce it to large markets at first, such as Beijing, Guangzhou, and Shenzhen. We will first pick some small cities to try it."
		<i>Testing new ideas</i>
		"The initial idea was very rough. . . then you put it on the market and get recognized. . . the whole process is actually the course of trial and error."
	Adaption	<i>Test different solutions</i>
		"If there are too many options, such as customer needs, how to make a choice? Under these circumstances, the system engineer performs assessment and feasibility analysis, during which experiments are conducted to determine the feasible solution."
		<i>Learning from failures</i>
		"Our team recorded the mistakes made during the previous generation of product development. After the project was completed, everyone discussed why we had made these mistakes. Why could we not avoid these mistakes at that time?"
		<i>Reviewing regularly</i>
Product iteration	Adaption	"Before leaving the company every night, we will have a brief meeting where we discuss what each person has done and find out good and bad practices."
		<i>Reflecting and improving the process</i>
		"After the completion of a project, the project management team will conduct a reconciliation to summarize aspects of the project process that are problematic and find out why these problems arise. Then the process will be improved."
		<i>Determining new product development strategy</i>
		"When there are new products and technologies appear on the market, it requires judgment to determine whether to imitate or develop our own. One of our basic principles is to do only what we are good at."
	Product iteration	<i>Identifying key demands</i>
		"We cannot satisfy all the needs of our customers. We will only meet the needs of most customers and the needs of potential customers in the future. It requires judgment to distinguish different kinds of needs."
		<i>Judging demand priority</i>
		"Taking software development as an example, after a version is introduced, there will be some changes in demands, such as adding or removing some functions. We have a change control committee that judges and verifies the demands and decides which demand should be integrated into the next version [of the software]. We follow this direction to develop the next version."
		<i>Identifying suitable technologies</i>
	Product iteration	"We cannot blindly worship the most advanced technology. When a new technology emerges, we must ask ourselves: do we have enough reserve of talents to utilize the new technology? Is the cost controllable?"
		<i>Updating products</i>
		"We adopt an agile development model which emphasizes team openness, keeping up with customer needs and flat organization. This development model can ensure rapid response to the market and the continuous upgrading of products."
		<i>Updating teams' knowledge</i>
		"Along with the products' upgrading, the knowledge of our team will also be continuously updated. For example, the old product structure will be abandoned and a new product structure will be established. The entire team's knowledge has to keep up."
		<i>Keeping pace with changes in the market</i>
		"Nowadays the market is changing rapidly, our products must keep pace with changes in the market. . . Therefore, we continue to learn from the outside to promote the constant upgrading of our knowledge."

Appendix B. Key measures in survey

Dependent variables:

Radical innovation performance

- 1 Compared to your major competitor, this firm introduced more radical new products in the last three years (1 = “strongly disagree,” 5 = “strongly agree”).
- 2 This firm frequently introduced radical new products into markets totally new to the firm in the last three years (1 = “strongly disagree,” 5 = “strongly agree”).
- 3 Number of radical products introduced by the firm in the last three years (converted into a five-point scale).*
- 4 % of total sales from radical product introduced by your firm in the last three years (less than 5%, 5%–10%, 11%–15%, 16%–20%, >20%).*

Incremental innovation performance

- 1 Compared to your major competitor, this firm introduced more incremental new products in the last three years (1 = “strongly disagree,” 5 = “strongly agree”).
- 2 This firm frequently introduced incremental new products into new markets in the last three years (1 = “strongly disagree,” 5 = “strongly agree”).
- 3 Number of incremental products introduced by the firm in the last three years (converted into five-point scale).*
- 4 % of total sales from incremental product introduced by your firm in the last three years (less than 5%, 5%–10%, 11%–15%, 16%–20%, >20%).*

Product innovation performance

Could you please state the performance of your company compared to your competitors with regard to the following items? 1 represented much worse performance than competitors and 5 represented much better performance.

Product innovation efficacy

- 1 Replacement of products being phased out.
- 2 Extension of product range within main product field through new products.
- 3 Extension of product range outside main product field.
- 4 Development of environment-friendly products.
- 5 Market share evolution.
- 6 Opening of new markets abroad.
- 7 Opening of new domestic target groups.

Product innovation efficiency

- 1 Average innovation project development time.
- 2 Average number of working hours on innovation projects.
- 3 Average cost per innovation project.
- 4 Global degree of satisfaction with innovation project efficiency.*

Control variables: Adopted from Narver and Slater (1990)

- 1 *Ease of entry (ENTRY)*. How easy is it for new entrants to start competing in this industry? (1 very easy; 5=very difficult).
- 2 *Threat of substitutes (SUBS)*. How easy can a product or service be substituted in this industry? (1 very easy; 5 = very difficult).
- 3 *Bargaining power of buyers (BPOW)*. The extent to which our customers are able to negotiate lower prices in this industry is (1 very low; 5 = very high).
- 4 *Bargaining power of suppliers (SPOW)*. The extent to which we are able to negotiate lower prices from our suppliers in this industry (1 very low; 5 = very high).

- 5 *Relative size (RSIZE)*. The size of our annual sales revenues in the principal served market segment in relation to those of our largest competitor is: (1 = much smaller than the largest competitor; 5 = much larger than the largest competitor)
- 6 *Relative costs (RCOST)*. Our average total operating costs (administrative, production, marketing/sales, etc.) in relation to those of largest competitor in its principal served market segment are: (1=much lower than the largest competitor; 5 = much higher than the largest competitor).

Notes: * indicated that the item was deleted based on item-to-total correlation and factor analysis

References

- Ahuja, G., Lampert, C. M., & Novelli, E. (2013). The second face of appropriability: Generative appropriability and its determinants. *Academy of Management Review*, 38(2), 248–269.
- Alegre, J., & Chiva, R. (2008). Assessing the impact of organizational learning capability on product innovation performance: An empirical test. *Technovation*, 28(6), 315–326.
- Alegre, J., Lapidra, R., & Chiva, R. (2006). A measurement scale for product innovation performance. *European Journal of Innovation Management*, 9(4), 333–346.
- Alexander, C. (1999). The origins of pattern theory: The future of the theory, and the generation of a living world. *IEEE Software*, 16(5), 71–82.
- Alshanty, A. M., & Emeagwali, O. L. (2019). Market-sensing capability, knowledge creation and innovation: The moderating role of entrepreneurial-orientation. *Journal of Innovation & Knowledge*, 4(3), 171–178.
- Alvarez-Milán, A., Felix, R., Rauschnabel, P. A., & Hinsch, C. (2018). Strategic customer engagement marketing: A decision making framework. *Journal of Business Research*, 92(July), 61–70.
- Ancona, D. G., Okhuysen, G. A., & Perlow, L. A. (2001). Taking time to integrate temporal research. *Academy of Management Review*, 26(4), 512–529.
- Argote, L., & Ren, Y. Q. (2012). Transactive memory systems: A microfoundation of dynamic capabilities. *Journal of Management Studies*, 49(8), 1375–1382.
- Argyres, N., & Mostafa, R. (2016). Knowledge inheritance, vertical integration, and entrant survival in the early US auto industry. *Academy of Management Journal*, 59(4), 1474–1492.
- Argyris, C. (1976). Single-loop and double-loop models in research on decision making. *Administrative Science Quarterly*, 21(3), 363–375.
- Argyris, C., & Schon, D. (1978). *Organizational learning: A theory of action approach*. Reading, MA: Addison Wesley.
- Atuahene-Gima, K. (2005). Resolving the capability—Rigidity paradox in new product innovation. *Journal of Marketing*, 69(4), 61–83.
- Avital, M., & Te'eni, D. (2009). From generative fit to generative capacity: Exploring an emerging dimension of information systems design and task performance. *Information Systems Journal*, 19(4), 345–367.
- Ballé, M., Morgan, J., & Sobek, D. K. (2016). Why learning is central to sustained innovation. *MIT Sloan Management Review*, 57(3), 63.
- Berry, H. (2015). Knowledge inheritance in global industries: The impact of parent firm knowledge on the performance of foreign subsidiaries. *Academy of Management Journal*, 58(5), 1438–1458.
- Boer, H., Caffyn, S., Corso, M., Coughlan, P., Gieskes, J., Magnusson, M., et al. (2001). Knowledge and continuous innovation: The CIMA methodology. *International Journal of Operations & Production Management*, 21(4), 490–504.
- Bollen, K. (1989). *Structural equations with latent variables*. New York: Wiley.
- Bollen, K., & Lennox, R. (1991). Conventional wisdom on measurement: A structural equation perspective. *Psychological Bulletin*, 110(2), 305–314.
- Borsboom, D. (2005). *Measuring the mind: Conceptual issues in contemporary psychometrics*. Cambridge, UK: Cambridge University Press.
- Brislin, R. W., Triandis, H. C., & Berry, J. W. (1980). Translation and content analysis of oral and written material. *Handbook of cross-cultural psychology: Methodology*: 2 (pp. 389–444). Boston, MA: Allyn & Bacon.
- Brown, S. L., & Eisenhardt, K. M. (1997). The art of continuous change: Linking complexity theory and time-paced evolution in relentlessly shifting organizations. *Administrative Science Quarterly*, 42(1), 1–34.
- Chandy, R. K., & Tellis, G. J. (1998). Organizing for radical product innovation: The overlooked role of willingness to cannibalize. *Journal of Marketing Research*, 35(4), 474–487.
- Chomsky, N. (2006). *Language and mind*. Cambridge: Cambridge University Press.
- Cohen, W. M., & Levinthal, D. A. (1990). Absorptive capacity: A new perspective on learning and innovation. *Administrative Science Quarterly*, 35(1), 128–152.
- Cole, R. E. (2002). From continuous improvement to continuous innovation. *Total Quality Management*, 13(8), 1051–1056.
- Collis, D. J. (1994). How valuable are organizational capabilities? *Strategic Management Journal*, 15(S1), 143–152.
- Cui, A. S., & Wu, F. (2017). The impact of customer involvement on new product development: Contingent and substitutive effects. *Journal of Product Innovation Management*, 34(1), 60–80.

- Dattée, B., Alexy, O., & Autio, E. (2018). Maneuvering in poor visibility: How firms play the ecosystem game when uncertainty is high. *Academy of Management Journal*, 61(2), 466–498.
- Day, G. S. (2011). Closing the marketing capabilities gap. *Journal of Marketing*, 75(4), 183–195.
- Dillman, D. A. (2000). *Mail and internet surveys: The total design method*. New York: Wiley.
- Dougherty, D. (2001). Reimagining the differentiation and integration of work for sustained product innovation. *Organization Science*, 12(5), 612–631.
- Dougherty, D., & Hardy, C. (1996). Sustained product innovation in large, mature organizations: Overcoming innovation-to-organization problems. *Academy of Management Journal*, 39(5), 1120–1153.
- Eisenhardt, K. M., & Bingham, C. B. (2017). Superior strategy in entrepreneurial settings: Thinking, doing, and the logic of opportunity. *Strategy Science*, 2(4), 246–257.
- Eisenhardt, K. M., & Tabrizi, B. N. (1995). Accelerating adaptive processes: Product innovation in the global computer industry. *Administrative Science Quarterly*, 40(1), 84–110.
- Fleming, L. (2001). Recombinant uncertainty in technological search. *Management Science*, 47(1), 117–132.
- Flores, L. G., Zheng, W., Rau, D., & Thomas, C. H. (2012). Organizational learning: Sub-process identification, construct validation, and an empirical test of cultural antecedents. *Journal of Management*, 38(2), 640–667.
- Foster, R. N. (1986). *Innovation: The attacker's advantage*. New York: Summit books.
- Frazer, J., Bentley, P. J., & Corne, D. W. (2002). Creative design and the generative evolutionary paradigm. *Creative evolutionary systems* (pp. 253–274). San Francisco, CA: Morgan Kaufmann Publishers.
- Furlan, A., & Grandinetti, R. (2016). Spinoffs and their endowments: Beyond knowledge inheritance theory. *Journal of Intellectual Capital*, 17(3), 570–589.
- Garud, R., Gehman, J., & Kumaraswamy, A. (2011). Complexity arrangements for sustained innovation: Lessons from 3M corporation. *Organization Studies*, 32(6), 737–767.
- Gergen, K. J. (2012). *Toward transformation in social knowledge*. Berlin: Springer.
- Ginja Antunes, H. d. J., & Pinheiro, P. G. (2020). Linking knowledge management, organizational learning and memory. *Journal of Innovation & Knowledge*, 5(2), 140–149.
- Gioia, D. A., Corley, K. G., & Hamilton, A. L. (2013). Seeking qualitative rigor in inductive research: Notes on the Gioia methodology. *Organizational Research Methods*, 16(1), 15–31.
- Grant, R. M. (1996). Prospering in dynamically-competitive environments: Organizational capability as knowledge integration. *Organization Science*, 7(4), 375–387.
- Grant, R. M. (1996). Toward a knowledge-based theory of the firm. *Strategic Management Journal*, 17, 109–122 Winter Special Issue.
- Helfat, C. E., & Peteraf, M. A. (2003). The dynamic resource-based view: Capability life-cycles. *Strategic Management Journal*, 24(10), 997–1010.
- Hoornaert, S., Ballings, M., Malthouse, E. C., & Van den Poel, D. (2017). Identifying new product ideas: Waiting for the wisdom of the crowd or screening ideas in real time. *Journal of Product Innovation Management*, 34(5), 580–597.
- Hu, L. T., & Bentler, P. M. (1999). Cutoff criteria for fit indexes in covariance structure analysis: Conventional criteria versus new alternatives. *Structural Equation Modeling: A Multidisciplinary Journal*, 6(1), 1–55.
- Jarvis, C. B., MacKenzie, S. B., & Podsakoff, P. M. (2003). A critical review of construct indicators and measurement model misspecification in marketing and consumer research. *Journal of Consumer Research*, 30(2), 199–218.
- Jaworski, B. J., & Kohli, A. K. (1993). Market orientation: Antecedents and consequences. *The Journal of Marketing*, 57(3), 53–70.
- Jaziri, D. (2019). The advent of customer experiential knowledge management approach (CEKM): The integration of offline & online experiential knowledge. *Journal of Business Research*, 94, 241–256 January.
- Johansson, A. E., Raddats, C., & Witell, L. (2019). The role of customer knowledge development for incremental and radical service innovation in servitized manufacturers. *Journal of Business Research*, 98, 328–338 (Feb.).
- Joshi, A. W. (2016). When does customer orientation hinder (help) radical product innovation? The role of organizational rewards. *Journal of Product Innovation Management*, 33(4), 435–454.
- Kaminska, R., & Borzillo, S. (2016). Organizing for sustained innovation: The role of knowledge flows within and between organizational communities. *Knowledge Management Research & Practice*, 14(1), 46–54.
- Kelly, T. (2001). *The art of innovation*. New York: Doubleday.
- Kim, B., Kim, E., & Foss, N. J. (2016). Balancing absorptive capacity and inbound open innovation for sustained innovative performance: An attention-based view. *European Management Journal*, 34(1), 80–90.
- Knipfer, K., Kump, B., Wessel, D., & Cress, U. (2013). Reflection as a catalyst for organisational learning. *Studies in Continuing Education*, 35(1), 30–48.
- Krishnan, V., & Ulrich, K. T. (2001). Product development decisions: A review of the literature. *Management Science*, 47(1), 1–21.
- Kvale, S. (1996). *Interviews: An introduction to qualitative research interviewing*. Thousand Oaks, CA: Sage Publications.
- Langley, A., Smallman, C., Tsoukas, H., & Van de Ven, A. H. (2013). Process studies of change in organization and management: Unveiling temporality, activity, and flow. *Academy of Management Journal*, 56(1), 1–13.
- Lave, J., Wenger, E., & Wenger, E. (1991). *Situated learning: Legitimate peripheral participation*. Cambridge: Cambridge university press.
- Lazonick, W., & Prencipe, A. (2005). Dynamic capabilities and sustained innovation: Strategic control and financial commitment at Rolls-Royce plc. *Industrial and Corporate Change*, 14(3), 501–542.
- Lee, G., & Raghu, T. S. (2014). Determinants of mobile apps' success: Evidence from the app store market. *Journal of Management Information Systems*, 31(2), 133–170.
- Lee, R. P., Naylor, G., & Chen, Q. (2011). Linking customer resources to firm success: The role of marketing program implementation. *Journal of Business Research*, 64(4), 394–400.
- Lee, R. P., & Zhou, K. Z. (2012). Is product imitation good for firm performance? An examination of product imitation types and contingency factors. *Journal of International Marketing*, 20(3), 1–16.
- Lichtenstein, B. B. (2014). *Generative emergence: A new discipline of organizational, entrepreneurial, and social innovation*. Oxford, UK: Oxford University Press.
- Lin, H. F. (2007). Knowledge sharing and firm innovation capability: An empirical study. *International Journal of Manpower*, 28(3/4), 315–332.
- MacKenzie, S. B., Podsakoff, P. M., & Podsakoff, N. P. (2011). Construct measurement and validation procedures in MIS and behavioral research: Integrating new and existing techniques. *MIS Quarterly*, 35(2), 293–334.
- McGrath, R. G. (2001). Exploratory learning, innovative capacity, and managerial oversight. *Academy of Management Journal*, 44(1), 118–131.
- McIlroy, S., Ali, N., & Hassan, A. E. (2016). Fresh apps: An empirical study of frequently-updated mobile apps in the Google play store. *Empirical Software Engineering*, 21(3), 1346–1370.
- McKinley, W., Latham, S., & Braun, M. (2014). Organizational decline and innovation: Turnarounds and downward spirals. *Academy of Management Review*, 39(1), 88–110.
- Miles, M. B., Huberman, A. M., Huberman, M. A., & Huberman, M. (1994). *Qualitative data analysis: An expanded sourcebook*. Thousand Oaks, CA: SAGE.
- Miller, D., & Shamsie, J. (2001). Learning across the life cycle: Experimentation and performance among the hollywood studio heads. *Strategic Management Journal*, 22(8), 725–745.
- Morgan, N. A., Vorhies, D. W., & Mason, C. H. (2009). Market orientation, marketing capabilities, and firm performance. *Strategic Management Journal*, 30(8), 909–920.
- Narver, J. C., & Slater, S. F. (1990). The effect of a market orientation on business profitability. *Journal of Marketing*, 54(4), 20–35.
- Operti, E., & Carnabuci, G. (2014). Public knowledge, private gain: The effect of spillover networks on firms' innovative performance. *Journal of Management*, 40(4), 1042–1074.
- Parise, S., Cross, R., & Davenport, T. H. (2006). Strategies for preventing a knowledge-loss crisis. *MIT Sloan Management Review*, 47(4), 31.
- Philipson, S. (2020). Sources of innovation: Consequences for knowledge production and transfer. *Journal of Innovation & Knowledge*, 5(1), 50–58.
- Remmeland-Wikhamn, B., Ljungberg, J., Bergquist, M., & Kuschel, J. (2011). Open innovation, generativity and the supplier as peer: The case of iPhone and android. *International Journal of Innovation Management*, 15(1), 205–230.
- Ries, E. (2011). *The lean startup: How today's entrepreneurs use continuous innovation to create radically successful businesses*. New York: Random House.
- Santoro, G., Bresciani, S., & Papa, A. (2020). Collaborative modes with cultural and creative industries and innovation performance: The moderating role of heterogeneous sources of knowledge and absorptive capacity. *Technovation*, 92–93.
- Schilke, O. (2014). Second-order dynamic capabilities: How do they matter? *Academy of Management Perspectives*, 28(4), 368–380.
- Schön, D. A., & Ortony, A. (1993). Generative metaphor: A perspective on problem-setting in social policy. *Metaphor and thought* (pp. 137–163). Cambridge: Cambridge University Press.
- Schreyögg, G., & Kliesch-Eberl, M. (2007). How dynamic can organizational capabilities be? Towards a dual-process model of capability dynamization. *Strategic Management Journal*, 28(9), 913–933.
- Schumpeter, J. A. (1939). *Business cycles: A theoretical, historical, and statistical analysis of the capitalist process*. New York: McGraw-Hill.
- Slater, S. F., Mohr, J. J., & Sengupta, S. (2014). Radical product innovation capability: Literature review, synthesis, and illustrative research propositions. *Journal of Product Innovation Management*, 31(3), 552–566.
- Song, M., & Chen, Y. (2014). Organizational attributes, market growth, and product innovation. *Journal of Product Innovation Management*, 31(6), 1312–1329.
- Soosay, C. A., Hyland, P. W., & Ferrer, M. (2008). Supply chain collaboration: Capabilities for continuous innovation. *Supply Chain Management: An International Journal*, 13(2), 160–169.
- Sosna, M., Treviño-Rodríguez, R. N., & Velamuri, S. R. (2010). Business model innovation through trial-and-error learning: The naturhouse case. *Long Range Planning*, 43(2–3), 383–407.
- Subramaniam, M., & Youndt, M. A. (2005). The influence of intellectual capital on the types of innovative capabilities. *Academy of Management Journal*, 48(3), 450–463.
- Sun, S. L., & Zou, B. (2019). Generative capability. *IEEE Transactions on Engineering Management*, 66(4), 636–649.
- Tan, J., & Peng, M. W. (2003). Organizational slack and firm performance during economic transitions: Two studies from an emerging economy. *Strategic Management Journal*, 24(13), 1249–1263.
- Tang, Y., & Marinova, D. (2019). When less is more: The downside of customer knowledge sharing in new product development teams. *Journal of the Academy of Marketing Science*.
- Taylor, A. (2010). The next generation: Technology adoption and integration through internal competition in new product development. *Organization Science*, 21(1), 23–41.
- Thomke, S. H. (1998). Managing experimentation in the design of new products. *Management Science*, 44(6), 743–762.

- Thomke, S. (2001). Enlightened experimentation: The new imperative for innovation. *Harvard Business Review*, 79(2), 67–75.
- Utterback, J. (1994). *Mastering the dynamics of innovation*. Cambridge, MA: Harvard Business School Press.
- Van Oorschot, K., Eling, K., & Langerak, F. (2018). Measuring the knowns to manage the unknown: How to choose the gate timing strategy in NPD projects. *Journal of Product Innovation Management*, 35(2), 164–183.
- Wang, T., & Chen, Y. (2018). Capability stretching in product innovation. *Journal of Management*, 44(2), 784–810.
- West, J., & Bogers, M. (2014). Leveraging external sources of innovation: A review of research on open innovation. *Journal of Product Innovation Management*, 31(4), 814–831.
- Wong, C. S., Law, K. S., & Huang, G. H. (2008). On the importance of conducting construct-level analysis for multidimensional constructs in theory development and testing. *Journal of Management*, 34(4), 744–764.
- Yang, X., Sun, S. L., & Zhao, X. (2019). Search and execution: Examining the entrepreneurial cognitions behind the lean startup model. *Small Business Economics*, 52 (Jan), 667–679.
- Zahra, S. A., & George, G. (2002). Absorptive capacity: A review, reconceptualization, and extension. *Academy of Management Review*, 27(2), 185–203.
- Zhang, Y., Waldman, D. A., Han, Y. L., & Li, X. B. (2015). Paradoxical leader behaviors in people management: Antecedents and consequences. *Academy of Management Journal*, 58(2), 538–566.
- Zhu, X., Xiao, Z., Dong, M. C., & Gu, J. (2019). The fit between firms' open innovation and business model for new product development speed: A contingent perspective. *Technovation*, 86–87, 75–85.
- Zittrain, J. (2008). *The future of the internet—And how to stop it*. New Haven, CT: Yale University Press.
- Zollo, M., & Winter, S. G. (2002). Deliberate learning and the evolution of dynamic capabilities. *Organization Science*, 13(3), 339–351.