



The Influence of Technological Capability and Intellectual Capital on Dynamic Knowledge Management Practices in IT Organizations

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ABSTRACT

The focus of KM research has traditionally been on static tasks like codification and storage, with DKM serving as a mediator that can facilitate organizational agility and innovation performance. A structural equation modeling retrieval of organizational knowledge is also proposed in the paper. In information technology (IT) organizations, knowledge management requires a more flexible and adaptable approach that involves continuously sensing, integrating, and reconfiguring knowledge assets to keep up with changing technological and competitive environments. Through the use of a conceptual and empirically testable model, this paper investigates the impact of technological capability and intellectual capital on dynamic knowledge management (DKM) practices in IT organizations. Based on the dynamic capabilities view, the knowledge-based view of the firm, and intellectual capital theory, this paper suggests that technological capability and DKM serve as mutual predecessors. These capabilities are complemented by each other in terms of complexity, organizational agility, innovation performance, human, structural capacity, or technical expertise. The presentation includes a proposal and demonstration of an assumed measurement and structural model for the application of structural equation modeling (SEM), as well as examining the use of partial least squares SEM (PLS-SEM) or covariance-based SEMA with survey data from IT professionals. The paper concludes by highlighting theoretical contributions, managerial implications for agile technology organizations, limitations, and future research directions.

Keywords: Technological Capability, Intellectual Capital, Dynamic Knowledge Management, IT Organizations, Organizational Agility

1. INTRODUCTION

Knowledge has been identified as a crucial strategic asset for organizations, especially those in knowledge-intensive fields like IT. Knowledge is typically viewed as a stable asset that can be captured, codified, archived, and accessed at any time within conventional knowledge management systems. This approach is known as knowledge governance. KM's static model was effective in competitive environments where product, process, and customer changes were gradual. Organizations found a balance between the two. In modern IT firms, the ability to continuously recognize technology and market changes, rapidly reconfigure knowledge assets, and translate these insights into new products, services, or processes is a key factor that drives innovation. DKM emphasizes the importance of continuous knowledge



generation, real-time knowledge sharing among distributed and cross-functional teams, fast unlearning of outdated knowledge, and the reintegration of existing knowledge with newly acquired knowledge. The production and revision of knowledge in agile software development environments require KM systems and practices that can adapt to rapid change, rather than static documentation efforts. Two organizational resources, technological capability and intellectual capital, are widely regarded as the foundation of an organization's dynamic knowledge management capacity. Technological capability encompasses an organization's ability to obtain, implement and effectively utilize technological resources, such as IT infrastructure, digital collaboration platforms, cloud-based knowledge systems, and technical expertise in emerging tools. Intellectual capital is the aggregate of an organization's knowledge-based resources, which can be classified into human capital, structural capital (knowledge embedded in organizational routines and systems, and codified knowledge), and relational capital. This group of resources includes information about an organization and its relationships with customers, partners, or external stakeholders. The literature has focused on the relationship between technological capability and intellectual capital, which are antecedents to KM outcomes, but not enough empirical research has examined how these two resources impact dynamic CM practices and agility and innovation outcomes. This paper proposes an integrated conceptual model that models technological ability and intellect as complementary and correlated precursors of DKM methodology, appropriate for testing the model using survey data from IT organizations. The paper is presented in the following manner. The second part delves into the pertinent literature on KM, technology, and intellectual capital. All three sections are equally comprehensive. Section 3 introduces new concepts by laying out hypotheses and conceptual frameworks. A proposed research methodology, which encompasses sampling, measurement methods and the analytical technique of SEM, is outlined in Section 4. An illustrative discussion of results that align with the proposed model is presented in Section 5. The results' theory and application is discussed in Section 6. Section 7 outlines limitations and potential avenues for future research, while Section 8 signified the conclusion of the paper.

2. LITERATURE REVIEW

The early KM literature was heavily influenced by the knowledge-based perspective of the company, which regards knowledge as the most strategically important resource an organization has and emphasizes processes for knowledge creation, storage, transfer, and application. The early work implicitly assumed a stable environment that ensured knowledge would remain relevant for many years after it was transposed into manuals, databases, or best-practice documents. The outcome of this was the emergence of static KM, a set of repository-centric practices that is primarily focused on preserving existing knowledge rather than constantly updating it, according to many scholars.

Dynamic knowledge management takes advantage of the temporal and adaptive nature of knowledge processes, expanding on this perspective. The process of knowledge generation, dissemination and implementation is not viewed as discrete linear processes in DKM; rather it is seen as continuous over time, interdependent, and mutually reinforcing cycles that must



adapt to changing environmental conditions. DKM utilizes the dynamic capabilities perspective to define organizational capability, which involves identifying opportunities and threats, reconfiguring resources to capture them, and adapting internal structures and knowledge assets to respond to them. KM encompasses knowledge management that involves not just efficient storage and retrieval of information, but also active identification of new knowledge needs, swift integration of dispersed knowledge, and intentional unlearning or updating of outdated knowledge.

The focus on agile and DevOps is particularly evident in IT organizations. Rapid cycles of development and deployment require knowledge processes that can accommodate short iterations, frequent retrospectives, and continuous feedback. Typically, in agile contexts, tacit knowledge sharing through daily stand-ups and pair programming, as well as cross-functional collaboration, is preferred over explicit documentation practices such as comprehensive requirements specifications that are produced at the beginning of a project. KM faces both a challenge and an opportunity due to the informal channels that facilitate knowledge flow, which is faster and more fluid than before.

2.2. The Technological Edge was a precursor to Dynamic KM.

The capacity of an organization to acquire, integrate, and leverage technology for the purpose of facilitating organizational operations is known as technological capability. KM is often used to operationalized technological capability in three dimensions: 1.

Flexibility in IT infrastructure is the ability of an organization's technology architecture to be reconfigured for new applications and data flows;

The sophistication of collaboration, communication, and knowledge-sharing platforms like enterprise wikis or cloud-based document systems, integrated development environments, are all examples of digital platform maturity.

Technical proficiency, indicating the extent to which employees are skilled in using and adapting these technologies

.'Studies conducted earlier on IT capabilities and organizational performance indicate that technological capability allows for quicker information processing, cost-effective knowledge searches and retrievals, and efficient collaboration across geographically disparate teams. The characteristics of dynamic KM are directly linked to the need for knowledge to move quickly across organizational boundaries and maintain continuous updating of knowledge repositories. Cloud-based platforms enable the rapid access and modification of knowledge assets in real-time by distributed teams, which is not possible with traditional static document-centric KM systems.

The use of technology alone is unlikely to suffice for dynamic KM, either. Knowledge can be generated, shared, and applied through technology infrastructure, but employees' capacity to create and use knowledge is determined by the blending of human and organizational factors.

2.3. The antecedent of Dynamic KM is intellectual capital

The theory of intellectual capital (IC) presents three interconnected dimensions that outline the intangible knowledge-based resources of an organization. Individual employees' knowledge, experience, and creativity are considered the human capital, which is the most



crucial source of new knowledge and innovative ideas. After individual employees depart, structural capital is the term used to describe organizational systems, processes, databases, and codified routines that remain in the organization. Value added through relationships with external stakeholders (customers, suppliers, technology partners, professional communities) often serves as a source of external knowledge and early signals of changing market or technology requirements. This is called relational capital.

Various aspects of intellectual capital can potentially contribute to dynamic knowledge management. ". The rapid acquisition of new technical knowledge and the creation of fresh insights are made possible by human capital, which is crucial for keeping up with technological progress. Despite the evolution of technology, structural capital provides organizational routines like retrospectives, communities of practice, and structured mentoring programs that enable the capture and dissemination of tacit knowledge created by individuals across organizations. By leveraging their relationships with clients, open-source communities and technology vendors, as well as industry consortia, organizations can detect and anticipate emerging trends and technological shifts, which helps them incorporate new knowledge into their internal DKM processes.

IC theory suggests that the three dimensions are interdependent and not independent, and their combined impact on organizational outcomes, known as intellectual capital efficiency, is often greater than the sum of their individual components. Models that investigate IC's impact on DKM should consider IG as a multidimensional, non-linear entity rather than its individual components.

2.4. Intellectual Capital and Technological Progress Interdependently?

There is a growing body of research exploring the combined or interactive effects of technological capability and intellectual capital on organizational outcomes. Technology enhances the value of intellectual capital by facilitating the rapid and widely distributed dissemination of implicit and explicit knowledge, while technological tools are crucial in shaping the creation, application to, and renewal of knowledge. If there is not enough human capital, KM technologies may be underutilized or utilized solely for day-to-day documentation, instead of contributing to genuine knowledge creation and renewal. However, without proper technology infrastructure, even the most competent and knowledgeable staff may not have the platforms required to rapidly disseminate and integrate knowledge across a large, distributed organization.

The argument of complementarity is particularly applicable to dynamic KM in IT organizations, where the operating environment is defined by both constant technological change and highly skilled, mobile knowledge workers. To enable IT organizations to move away from static, repository-based KM and adopt dynamic knowledge processes is the combination of formidable technological capability and rich intellectual capital that has been the driving force.

2.5. Outcomes of Dynamic KM are organizational agility and innovation

The literature proposes that dynamic KM serves as a means for organizations to achieve greater agility and innovation performance. The capacity to respond quickly by rearranging

resources and processes, known as organizational agility, is defined in relation to an organization's ability or capability to create new products, services, or processes. Innovation performance is measured in the ability of an organization to innovate through these means. DKM is believed to be a crucial tool for connecting technological and intellectual resources to higher-order performance outcomes, as it facilitates faster identification of new knowledge needs, quicker integration of dispersed expertise, and frequent updates in organizational knowledge bases. This is supported by this theory. The subsequent section develops a conceptual model that emphasizes the central role of mediating.

3. CONCEPTUAL FRAMEWORK AND HYPOTHESES

In this paper, a model that models the interplay between technological capability (TC) and intellectual capital (IC) as exogenous constructs with varying degrees of coherence is presented, building on the literature discussed above. Based on the complementarity argument presented above, the model suggests that TC and IC interact directly in DKM. This is supported by this theory. The following hypotheses are proposed:

Conceptual Framework: Antecedents and Outcomes of Dynamic Knowledge Management

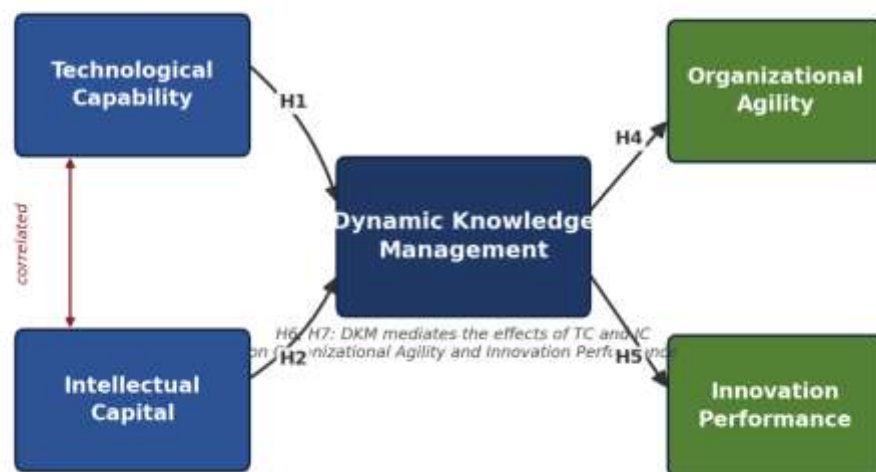


Fig 1: Conceptual Framework Diagram

Technological capability and dynamic knowledge management practices are positively linked in IT organizations during H1. Dynamic knowledge management practices are positively linked to intellectual capital in IT organizations. The relationship between technological capability and intellectual capital has a positive impact on dynamic knowledge management, beyond their individual effects. Organizational agility is positively linked to dynamic knowledge management. The use of dynamic knowledge management is linked to improved innovation performance. H6 leverages dynamic knowledge management to bridge the gap between technological aptitude and organizational agility/innovation performance. Intellectual capital and organizational agility/innovation performance are intertwined through dynamic knowledge management in H7.



The model proposes that every first-order construct will be modeled as a multidimensional, reflective latent variable. Technological capability is modeled on three first-order dimensions: IT infrastructure flexibility, digital platform maturity, and technical expertise. First-order dimensions, which include human capital, structural capital and relational Capital, are used to model intellectual capital. Four first-order dimensions are used to model dynamic knowledge management, which is based on its process-oriented nature: knowledge sensing, rapid knowledge integration, knowledge unlearning/renewal, and real-time knowledge sharing. Similar to previous surveys conducted in this field, the model is constructed using single-order reflective constructs that measure organizational agility and innovation performance.

Dynamic knowledge management is influenced by two exogenous latent constructs, technological capability and intellectual capital, as shown in Figure 1. An arrow indicates the interaction term that these two entities interact with. The central mediating construct is dynamic knowledge management, with two outgoing paths leading to organizational agility and innovation performance respectively, a covariance path connecting the two outcomes constructs that are likely related as joint downstream consequences of DKM.

4. RESEARCH METHODOLOGY

4.1. Research Design

A cross-sectional survey is conducted to test the hypothesized model, which is then presented as a quantitative approach. Given the objective of analyzing statistical relationships among latent constructs across a sufficiently large sample of IT firms, this design is appropriate and in line with most empirical research on KM and intellectual capital using structural equation modeling. Even though a cross-sectional design cannot provide the same level of confidence to prove causality, it's well-suited for exploratory and confirmatory theory-testing purposes in this paper. The drawback from using this type of design is discussed in Section 7.

4.2. Sample and Data Collection

Those who are interested in working in IT organizations that have adopted agile or hybrid agile-DevOps delivery models are typically managers, team leaders and senior technical professionals (such as software architects, engineering managers), product owners, and knowledge/CoE leads). For covariance-based SEM, the rule of thumb is to aim for a minimum ratio of 10 respondents per estimated parameter (or) at least 10 times the maximum number of structural paths pointing to essentially the same construct (for PLS-SEM), and accordingly an adequate sample size of 250–300 participants is required. Professional networks, industry associations, and organizational contacts would distribute a structured online questionnaire that data collection could use to reach knowledgeable respondents through purposive and snowball sampling. Procedural approaches to combat common method bias in single-source survey research include ensuring anonymity for respondents, disaggregating predictor and criterion items within the questionnaire, and employing a marker variable, along with implementing supplementary statistical tests such as Harman's single-factor test or full collinearity assessment.

4.3. Measurement of Constructs

The suggested measurement method would involve using multi-item reflective scales that were based on existing research and scored on a Likert scale of five or seven points, ranging from "strongly disagree" to "very agree".

Construct	Dimensions	Sample Indicator	Items (approx.)
Technological Capability	Infrastructure flexibility; platform maturity; technical expertise	Our IT infrastructure can be quickly reconfigured to support new tools	9-12
Intellectual Capital	Human capital; structural capital; relational capital	Employees in our organization are highly skilled at their jobs	9-12
Dynamic Knowledge Management	Sensing; integration; unlearning/renewal; real-time sharing	We quickly identify when existing knowledge/practices are becoming outdated	12-16
Organizational Agility	Single-order	Our organization responds quickly to changes in customer requirements	4-6
Innovation Performance	Single-order	Our organization frequently introduces new or improved services	4-6

Indicators for determining KM practices and innovation outcomes have been independently tested in previous studies, with control variables such as organizational size, age, industry sub-sector, and degree of agile maturity being included.

4.4. A technique for analyzing data using structural equation models

The primary analytical approach is argued to be structural equation modeling, which allows for the simultaneous estimation of multiple interrelated regression-like relationships among latent variables while taking into account measurement error. CB-SEM and PLS-SPSEM can be used for different purposes depending on the study's objectives and data characteristics, while PMOS or LISREL or Mplus is typically used to perform covariance-based SEM. The former requires arithmetic procedures, but the latter involves partial least squares using SmartPLS. When the research objective is to confirm theory and has satisfied the requirements of sample size/data distribution (multivariate normality, larger sample sizes), CB-SEM is generally preferred. In KM and IC studies, PLS-SEM is the preferred method as it can generate sophisticated models using various complex models and interaction terms, without needing to worry about strict multivariate normality assumptions.

The proposed model's inclusion of higher-order reflective-reflective constructs (technological capability and intellectual capital, with three first-degree dimensions, and dynamic knowledge management with four first-order dimensions) and an interaction term (H3) necessitates a PLS-SEM approach using the two-stage disjoint approach for higher order



construct estimation, while either product-indicator or two consecutive approaches are recommended for the interaction term. The process would involve two phases as is typical for SEM.

4.4.1. Measurement Model Assessment

Initially, the measurement model would be tested for its reliability and validity.' To determine internal consistency reliability, Cronbach's alpha and composite reliability would be used to assess values above 0.70. To determine the convergence of validity, AVE values above 0.50 indicate that more than 50% of the variance in its indicators is due to this construct. By using the Heterotrait-Monotrait ratio of correlations (HTMT), which can determine whether a construct has discriminant validity and is below 0.85, or the more conservative 0.90 threshold, it would be possible to supplement or replace the traditional Fornell-Larcker criterion, which may not be as effective in identifying discriminatory validity issues in methodological research.

4.4.2. Structural Model Assessment

Following the establishment of a satisfactory measurement model, the structural model would be evaluated by examining path coefficients, statistical significance (via bootstrapping with at least 10,000 resamples to create bias-corrected confidence intervals), coefficient of determination (R^2) for endogenous constructs and effect sizes (f^2) on individual predictors as well as predictive relevance (Q^2) through blindfolding procedures. Testing the interaction effect proposed in H3 would involve assessing the significance and effect size of the product term between technological capability and intellectual capital on dynamic knowledge management, with a significant positive coefficient supporting the complementarity hypothesis. Utilizing bootstrapped indirect effects, it would examine the likelihood of indirect paths from TC and "direct" via DKM impacting organizational agility and innovation performance by defining the type of mediation (full versus partial) in accordance with the significance of their corresponding direct effects.

5. ILLUSTRATIVE DISCUSSION OF ANTICIPATED RESULTS

This section is dedicated to discussing the pattern of results that would be consistent with the theoretical model, as this paper is conceptual and methodological rather than reporting findings from a final empirical study. Based on the literature, it is anticipated that both technological capability and intellectual capital would demonstrate significant standardized path coefficients on dynamic knowledge management, with intellectual capacity (especially its human capital dimension) potentially having a more substantial impact than just technological ability, as previous studies have shown that human and organizational knowledge resources are the primary drivers of KM outcomes, while technology plays essentially enabling roles.

It is anticipated that the interaction term for technological capability and intellectual capital will be positive and significant, indicating that organizations with both robust technological infrastructure and advanced intellectual resources would experience more dynamic KM than those with strength in only one aspect. This would provide empirical validation for the complementary asset's argument presented in the literature review, and it would also carry a

significant managerial consequence. Investments in KM technology platforms alone are not expected to yield the desired dynamic KT capability, while those in human capital development and knowledge-sharing culture are significantly less effective.

The relationship to organizational agility and innovation performance of dynamic knowledge management is expected slightly stronger than its relational counterparts, given the conceptual closeness between the rapid-sensing and rapid-integration dimensions (DKM) and the definition of organizational agile. The mediation model is anticipated to demonstrate partial mediation, whereby technological capability and intellectual capital would maintain slight but significant direct effects on outcome variables, even after taking into account the indirect path through dynamic KM, indicating that these resources also have an impact on agility and innovation via other pathways not captured in this model, such as direct impacts on organizational structure or leadership.

Table 2 provides an illustrative summary of the expected structural model results, with emphasis on hypothesized standardized path coefficients.

Hyp.	Path	Anticipated (illustrative)	β Anticipated Support
H1	Technological Capability → Dynamic KM	0.30 - 0.40	Supported
H2	Intellectual Capital → Dynamic KM	0.40 - 0.50	Supported
H3	TC × IC → Dynamic KM	0.10 - 0.15	Supported
H4	Dynamic KM → Organizational Agility	0.45 - 0.55	Supported
H5	Dynamic KM → Innovation Performance	0.35 - 0.45	Supported
H6-H7	Dynamic KM mediates TC/IC → Outcomes	Partial mediation	Supported

6. DISCUSSION

6.1. Theoretical Contributions

Several contributions are made to the literature on KM and dynamic capabilities in this paper. By modeling knowledge management as a dynamic, multidimensional capability that encompasses sensing, integration, unlearning/renewal, and real-time sharing processes, it extends the mostly static conception of KM that has been the dominant concept in earlier literature. KM research is now better aligned with the dynamic capabilities view, which is particularly applicable in rapidly changing industries like IT, where the assumptions that underlie static kilometer-based models are increasingly untenable.

The paper contributes to the intellectual capital literature by connecting IC to a specific, process-oriented organizational capability (DKM) through both empirical and theoretical means, instead of treating it as merely referring to its role as an independent predictor of firm performance, as has been observed in previous ICS literature. This is in line with recent calls within the IC literature for research into the mechanisms by which ic becomes performance outcomes, rather than just documenting a direct statistical relationship between CI and PR.



By suggesting and testing an interaction effect between technological capability and intellectual capital, the paper contributes to a growing body of research on complementary organizational assets, providing unbiased and empirical evidence for how technology affects human resources in organizations rather than acting independently.

6.2. Managerial Implications

Managers and leaders in IT organizations have several practical implications for the proposed model and its anticipated outcomes. Organisations aiming to enhance dynamic knowledge management should avoid prioritizing the acquisition of technology over human capital and structural capital over formalized retrospectives and knowledge-transfer routines. According to the expected interaction effect (H3), technology investments will yield the most when combined with efforts to improve employee skills and knowledge-sharing routines. Given the potential impact of dynamic KM on organizational agility, IT leaders in rapidly changing markets should view it not as a support function but as fundamental strategic capability that is closely tied to the organization's ability to adapt to market and technological changes. It is possible to embed KM-related roles (such as knowledge stewards or communities-of-practice leads) within agile delivery teams, rather than centralizing them in a separate corporate unit away from daily delivery work. This could be done without any external influence.

This paper examines three dimensions of intellectual capital, and relational capital is one that highlights the strategic value placed on external knowledge sources such as open-source communities, technology vendor partnerships, or customer co-creation forums. Internal knowledge alone may not be beneficial for IT organizations, as external relationships can contribute to internal dynamic KM processes.

7. LIMITATIONS AND FUTURE RESEARCH DIRECTIONS

Despite being a conceptual and methodological paper, there are several limitations to this work that should be taken into account for future empirical research. Although the cross-sectional survey design suggested here is suitable for testing theory, it does not provide sufficient information about the direction of causality between the constructs in the model. The use of longitudinal designs to track changes in technological capability, intellectual capital, and dynamic KM practices would enhance future research efforts by providing more accurate causal inferences and enabling investigation of how these relationships change as organizations progress in their agile and knowledge management practices.

Common method bias may persist despite the use of procedural and statistical remedies discussed in Section 4.2, particularly with regard to single-source, self-reported survey data. In upcoming studies, it is possible to incorporate information from multiple sources, such as by connecting employee-level perceptions of KM practices with objective organizational performance data or assessments from external stakeholders like customers or partners. Despite being grounded in an established theory, the model proposed in this paper does not consider potential boundary conditions or moderating variables that may affect the strength of previously assumed relationships. The balance between technological capability, intellectual capital and dynamic KM may be affected by factors such as organizational culture, leadership



style, industry turmoil, and geographic distribution. Additional research could incorporate moderators into the model and investigate whether the associations proposed here differ across organizational environments, such as software product companies versus IT services firms, or startups compared to large technology organizations.

Rather than other areas of analysis, this paper models dynamic KM and technological capability along with intellectual capital at the organizational level. Understanding the intricate interdependence between team-level dynamic KM practices and organizational-wide capabilities requires further research using multi-dimensional modeling techniques, as knowledge processes typically occur at the team level in agile organizations. This information could be more accurately understood. In order to enhance understanding of the specific mechanisms and practices involved in translating technological capability and intellectual capital into dynamic knowledge management routines within certain organizational contexts, future research could employ qualitative case studies or mixed-method designs instead of relying solely on the quantitative SEM approach presented above.

8. CONCLUSION

The paper suggests that IT organizations in agile, competitive environments and fast technological changes must incorporate dynamic knowledge management practices. Utilizing the dynamic capabilities perspective, the knowledge-based view of the firm, and intellectual capital theory, this paper developed a conceptual model that highlights the complementary roles of technological capability and intellectual capital in driving dynamic knowledge management. These perspectives also contribute to the mediation effect between these capabilities on organizational agility and innovation performance. It was suggested that a method of modeling structural equations with higher-order constructs and an interaction term would be suitable for empirical testing of this model, while also providing an overview of the expected outcomes to guide future empirical research.

The proposed framework is both theoretically and practically relevant. It theoretically advances towards a more dynamic, process-oriented understanding of KM that better aligns with the needs of modern, agile technology companies, and it highlights the complementary roles played by technology and knowledge-based human resources in enabling this dynamic capability. It offers IT leaders a structured approach to considering how investments in technology infrastructure, as well as in people and organizational routines, contribute to the organization's ability to sense, integrate, and renew knowledge in response to continuous change. Given the rapid escalation of technological disruption, it is likely that IT firms will continue to prioritize dynamic knowledge management over static one-way control for achieving sustained competitive advantage, and the model presented in this paper provides a basis for empirical research intended to test, refine, or expand upon this theory.

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