An Analysis of Global R&D Activities of Japanese MNCs in the US From the Knowledge-Based View

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Abstract—We examined 79 Japanese MNCs’ R&D subsidiaries in the US from the knowledge-based view. We found: 1) subsidiaries’ R&D strategies encouraged some types of knowledge flows, but discouraged other types; 2) autonomous R&D subsidiaries promoted knowledge flows from local environments to the subsidiary; 3) process-oriented incentives promoted technical knowledge flows between a subsidiary and its headquarters; 4) knowledge flows and R&D strategies contributed to a subsidiary’s knowledge accumulation; 5) organizational factors such as autonomy and process-oriented incentives contributed more to a subsidiary’s overall performance than the subsidiary’s R&D strategies did.

Index Terms—International subsidiaries, knowledge-based view, R&D management.

I. INTRODUCTION

EXTERNAL technological markets continue to become more global and efficient, primarily due to the advancements in IT-based communication and database methods. For example, web-based knowledge sharing sites, such as “Innovera,” started by Eli Lilly in 2001, have become popular outsourcing methods for technological problem-solving among multi-national corporations (MNCs) [93].

External markets, however, remain relatively ineffective mechanisms for knowledge sharing and transfer because: 1) the specialized knowledge of firms tends to be tacit and thereby difficult to transfer; 2) market-based transfers of knowledge are often associated with negative externalities such as involuntary expropriation and the risk of creating a new competitor [122].

Thus, in the last decade, MNCs seem to be globalizing their R&D activities by using internal governance mechanisms [36], [52], [64]. For example, a study by the National Science Foundation (NSF) shows that over the last ten years, U.S. companies’ investment in overseas R&D has increased three times faster than company funded domestic R&D—10.1% versus 3.4%. Overseas R&D now accounts for 12 percent of the U.S. industry’s domestic R&D expenditures [103].

However, there are only a few studies that shed light on such global R&D activities by MNCs, especially by Japanese MNCs. Furthermore, there are few theoretical and empirical studies that can answer the question: “How should global R&D activities be organized efficiently?.” In an effort to fill this gap in our understanding, the current study investigates global knowledge flows in Japanese R&D subsidiaries in the US, from the knowledge-based view of MNCs [36], [52].

We believe that this research sheds light on a general aspect of R&D globalization, because: 1) the US offers the most attractive “host” location for global R&D subsidiaries [110]; 2) many empirical studies have indicated that a majority of R&D facilities in the US owned by foreign firms are Japanese [29], [39], [117]; 3) a majority of Japanese MNCs’ overseas R&D facilities are reported to be located in the US [70], [72].

This paper attempts to answer the following four research questions: 1) what factors will determine knowledge flows among R&D subsidiaries, headquarters (HQ), and other R&D subsidiaries?; 2) what factors will determine knowledge accumulations of R&D subsidiaries?; 3) what factors will determine performance of R&D subsidiaries?; 4) how should R&D subsidiaries manage knowledge flows for higher performance?

In order to answer these questions, this paper includes the following four major sections. First, the paper discusses global R&D activities by surveying the relevant literature. Second, it constructs several hypotheses based on the knowledge-based view. Third, it tests these hypotheses by using survey data from 79 Japanese R&D subsidiaries in the US. Finally, it concludes with managerial implications and suggestions for future research.

II. THEORETICAL FRAMEWORK AND HYPOTHESES

A. Knowledge-Based View of MNCs

Hymer [64] originally argued that MNCs’ raison d’être lies in the ability to internally exploit knowledge more efficiently than would be possible through external market mechanisms. Such a perspective emphasizes that globally dispersed R&D operations provide MNCs with competitive advantages not available in single-country centralized R&D operations [20], [37], [111], and that such a competitive advantage is based on how efficiently MNCs share knowledge across HQs and R&D subsidiaries [36], [52]. Such a view can be called “the knowledge-based view (KBV) of MNCs.”

In spite of the criticality of “knowledge shares and transfers within MNCs”—labeled as “knowledge flows” henceforth—we have a limited understanding of how to efficiently manage such knowledge flows. This is because only a limited number of studies have empirically investigated the determinants of knowledge flows.

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The reason for such a paucity of studies could stem from difficulties in accessing actual knowledge flows in MNCs. However, we identified the following nine “exceptional” studies, based on the following two criteria: 1) studies must be based on KBV of MNCs; 2) studies must measure actual knowledge flows in MNCs. We will briefly review these nine studies below—Table I summarizes these studies.1

By examining 110 R&D subsidiaries in 15 Swedish MNCs, Nobel and Birkinshaw [98] argue that effective communication methods (e.g., face-to-face) and directions (e.g., with other global R&D subsidiaries) vary depending on specific types of R&D subsidiaries (e.g., global creator). In their extensive study of 374 subsidiaries in 75 MNCs in US, Europe and Japan, Gupta and Govindarajan [52] found that subsidiaries’ knowledge stock, absorptive capacity, motivational disposition and the richness of transmission channels promote knowledge flows.

By examining 91 new product development projects in 52 MNCs in US, Europe, Japan, and Korea, Subramaniam and Venkatraman [2001] found that a higher level of intra-MNCs integration promotes technological knowledge flows from subsidiaries to HQs.

1Egelhoff [38] examined knowledge flows between headquarters and subsidiaries of MNCs by focusing on control mechanisms. Rosenzweig and Nohria [115] also studied HR practices in MNCs and found that knowledge flows affect subsidiaries’ HR practices. However, we excluded these studies from our review, because these studies were not based on KBV of MNCs.

By examining 110 R&D subsidiaries in 15 Swedish MNCs, Birkinshaw, et al. [15] found that two dimensions of knowledge, observability—the extent to which it is easy to understand the activity by examining different aspects of the process or final product—and embeddedness—the extent to which knowledge is a function of the social and physical system in which it exists—stimulate knowledge flows. Almeida, et al. [6] also examined 21 MNCs in the semiconductor industry in ten countries, and argued that the superiority of MNCs stems from their ability to use multiple mechanisms of knowledge flows flexibly and simultaneously.

Furthermore, by examining 69 headquarters in American MNCs, Cummings and Teng [28] found that knowledge flow success was associated with 1) R&D units’ understanding where the desired knowledge resides, 2) the extent to which the parties share similar knowledge bases, 3) the extent of interactions between the source and the recipient, and 4) articulation processes through which the source’s knowledge is made accessible to the recipient.

By examining more than two thousand subsidiaries in seven countries, Foss and Pedersen [40] argue that MNC management can influence knowledge flows through choices regarding control, motivation and context. Finally, Minbaeva, et al. [91] examined 169 subsidiaries in US, Russia, and China, and found that knowledge flows tended to be promoted by investments in knowledge receivers’ ability and motivation through the extensive use of training, performance appraisal, performance-based compensation and internal communication.

There are several tendencies in these empirical studies on knowledge flows in MNCs. First, most of these studies investi-
gated general subsidiaries such as manufacturing or marketing subsidiaries within MNCs. Only two studies—i.e., Hakanson and Nobel [54] and Birkinshaw, et al. [15]—examined knowledge flows specifically in R&D subsidiaries within MNCs. Second, although Subramaniam and Venkatraman [119], Almeida, et al. [6], and Gupta and Govindarajan [52] examined Japanese MNCs as a part of their studies, there is no empirical study on knowledge flows, specifically focusing on Japanese MNCs and their R&D subsidiaries. Although there are some survey-based studies on recent global R&D activities in Japanese MNCs, these studies have primarily focused on “external factors,” such as when, how, and why Japanese MNCs globalized their R&D activities.2

Third, with the exception of Subramaniam and Venkatraman [119], and Gupta and Govindarajan [52], these studies measured technology-related knowledge flows, but did not measure market-related knowledge flows. We argue that both technology and market-related knowledge flows are important in order to examine the validity of KBV of MNCs, as will be elaborated in the following section.

Fourth, most of these studies measured either knowledge in-flows from HQs to subsidiaries—e.g., [28], [91]—or knowledge out-flows from subsidiaries to HQs—e.g., [15], [54]. Following Gupta and Govindarajan [52], we argue that both knowledge in-flows and out-flows should be measured in order to test the validity of KBV of MNCs.

Fifth, missions and functions of a variety of R&D subsidiaries have been theoretically discussed in many studies, and numerous typologies have been proposed—e.g., home-base-exploiting sites and home-base-augmenting sites [77], [109]; local adapters, international adapters, and international creators [98]. However, there are few empirical studies from the KBV that actually measured such typologies, except for Nobel and Birkinshaw [98].

Finally, most of these empirical studies either 1) used knowledge flows as performance indicators or 2) assumed that a high level of knowledge flows leads to high performance. We argue that knowledge flows are cost-involving activities—i.e., there must be an optimal level of knowledge flows. Our arguments will be elaborated further in the following section.

B. Knowledge Flow/Accumulation and Performance

The KBV of MNCs seems to assume: the higher the level of knowledge flows, the higher the MNCs’ (and/or subsidiaries’) performance, by extending information processing perspective

3For example, by studying 137 Japanese MNCs, Iwasa and Odagiri [68] found that HQ’s R&D expenditure, the US subs’ R&D expenditures, the presence of R&D in Europe, and the firm’s experience in the US, positively contributed to inventions (measured by patents) at home and in the US among firms, whose R&D subs mainly aim to research. On the other hand, such positive contributions were not observed in firms, whose R&D subs mainly aim to support local manufacturing and sales activities. Similar survey-based studies include Spiro and Dalton [108], Odagiri and Yasuda [105], Florida and Kenney [39], and Odagiri and Goto [104]. An exception is Asakawa [10], [11]. By surveying 44 Japanese MNCs and 66 R&D subs, Asakawa identified the evolving nature of headquarters-subsidiary relations during the whole process of R&D internationalization of Japanese MNCs. He found that the tension between HQs and R&D subs was more salient in information-sharing issues than in autonomy-control issues, and that R&D subs were more dissatisfied with the current level of information sharing and granted autonomy than HQs were.

[38], [41]. However, we cast serious doubt on its assumption for the following three reasons. First, the assumption disregards the intervening role of knowledge accumulation into the relationship between knowledge flows and performance.

Havelock [58] and Machlup and Mansfield [88] argue that “information” is a flow of messages, whereas “knowledge” is a stock. By extending their arguments, we argue that “knowledge accumulation” is a stocked assets, and that knowledge flows are activities, thus involving costs. Here, “knowledge accumulation” can be defined as “stored knowledge in forms of documents, papers, patents, copyrights or new products/services” [49], [90]. Furthermore, we argue that both market- and technology-related knowledge flows are critically important for effective knowledge accumulation.

The importance of market- and technology-related knowledge flows has been extensively examined in the new product development (NPD) literature—e.g., Tatikonda and Rosenthal [121], Sicotte and Langley [118], Ottum and Moore [107], and Kahn [75]. The NPD literature has been suggesting that technical knowledge flows either from in-house R&D efforts or from external sources—e.g., collaborative R&D projects or licensing—and market-related knowledge flows from customers should be efficiently converted into “accumulated knowledge.”

By extending the above arguments, Nonaka and Toyama [101], [102] argue that a firm can create new knowledge in “dialectic” manners by combining the externally acquired knowledge—mainly consisting of market-related knowledge—with the existing knowledge accumulated within the firm boundaries—mainly consisting of technology-related knowledge—only through the dynamic context.3

March [89] and Diericks and Cool [34] argue that the accumulation of strategic stocks (e.g., firm knowledge) is achieved over time through a constant flow of knowledge, resources and capabilities (e.g., effective problem-solving). Although all the knowledge flows may not necessarily lead to knowledge accumulation, we argue that there is a positive relationship between knowledge flows and knowledge accumulation.

Second, the assumption that a high level of knowledge flows leads to a high level of MNCs’ performance tends to disregard the qualitative aspects of knowledge flows and knowledge accumulation processes. For example, market-related knowledge flows concerning a specific customer’s hidden needs could have
various meanings and be interpreted differently according to MNCs’ or subsidiaries’ accumulated knowledge.4

Kagno [73] argues that “new meanings” will be created when knowledge flows enter into an organization’s accumulated knowledge, and that new meanings lead to new technologies, products or services. Such an argument is also made in interpretive sociology [116], symbolic interactionism [16], the sociology of knowledge [14], and cognitive social psychology [131].

Schutz [116] argues that interpretation may be one of the most important functions organizations perform. We argue that such interpretation can shape both knowledge flows and accumulations, and that both knowledge flows and accumulations shape the interpretation, as well. Such a unique organizational capability can be called “interpretative capacity.”

We assume that the essential nature of knowledge accumulations is in their interpretative functions rather than their activity-related functions such as their maintenance or retrieval. Thus, although knowledge accumulations may involve costs associated with their maintenance or retrieval activities, we assume that these costs are negligible.5

The literature on the resource-based view of the firms (RBV) has argued that accumulated resources (such as knowledge) could have negative effects on performance by proposing “core rigidity” [84] or “competence trap” [86]. The literature on knowledge management has also suggested such negative aspects of knowledge accumulation by proposing “knowledge depreciation” [30], or “organizational forgetting” [13]. For example, Darr et al. [30] argue that accumulated knowledge becomes obsolete. Benkard [13] and Argote and Eppele [8] also argue that firms tend to accumulate knowledge that is not related to their core capabilities.

However, most studies on knowledge depreciation or organizational forgetting were conducted in large-scale traditional operating situations—e.g., ship building during the World War [8], [124], wide-body airframe manufacturing [13], and pizza franchising [30]. We argue that such negative aspects of knowledge accumulation are negligible in this study, because overseas R&D subsidiaries are “genuine knowledge generating organizations,” and are generally small in size and new in their establishments.

Finally, the assumption that a high level of knowledge flows leads to a high level of MNCs’ performance disregards knowledge accumulation’s positive impact on performance. The accumulation of knowledge and its impact on innovative capabilities has received increasing attention from scholars of strategic management [62].

For example, the concept of “knowledge integration” [48], [65] emphasizes the ability to integrate different types of knowledge to develop new innovation-enhancing capabilities. Theories of “dynamic capabilities” [123] emphasize the role of knowledge as a source of innovation and competitive advantage.

Studies on social networks and inter-organizational relationships have also argued that knowledge flows among firms are essential to the innovation process (e.g., [87] and [114]). Furthermore, Helfat [61] found that during periods of environmental instability, firms with greater volumes of accumulated knowledge undertook greater amounts of R&D. Yli-Renko et al. also [134] found a significant positive relationship between knowledge accumulation and new product development.

We argue that accumulated knowledge leads to informed, efficient and timely decision-making or problem-solving, thus achieving high performance. Several empirical studies support our argument. For example, Lane et al. [82] found that knowledge acquired from alliance partners contributed significantly to alliance performance in their study on alliance. Furthermore, Moenaert, et al. [92] found that the knowledge accumulated by project teams tends to have a higher association with project performance than does the level of knowledge flows in their study on NPD.

In summary, we argue: 1) it is not “knowledge flow” but “knowledge accumulation” that contributes to the performance of MNCs and their subsidiaries; 2) “new meanings” will be created by uniquely associating knowledge flows with accumulated knowledge of MNCs or their subsidiaries; 3) both market and technology-related knowledge flows are needed for the new meanings to have values or competitive advantages—i.e., valuable, rare, inimitable and non-substitutable [130].

C. Factors Affecting Knowledge Flows

Strategic Factors: In spite of the various typologies proposed on foreign R&D subsidiaries, many studies are accepting that objectives or missions of foreign R&D subsidiaries are classified into two categories: 1) home-base-augmenting sites; 2) home-base-exploiting sites [25], [98], [109]. Home-base-augmenting sites can be defined as “sites established to enhance MNCs’ core capabilities by tapping technological knowledge from local environments—i.e., local universities, local governments, suppliers, and competitors [77]. Nobel and Birkinshaw [98] and Chiesa [25] argue that home-base-augmenting sites tend to become a “center of excellence,” by aiming at the basic end of the R&D spectrum—consisting of basic research, applied research and development research [4]—thus making the subsidiary a center of technical knowledge flows.

On the other hand, home-base-exploiting sites can be defined as “sites established to support foreign manufacturing facilities or to adapt products to the foreign market [77]. Papanastassiou and Pearce [108] argue that home-base-exploiting sites tend to develop products/services tailored to local markets’ needs, thus emphasizing market-related knowledge flows.
Accordingly, we expect that home-base-augmenting sites require high levels of technology-related knowledge flows, while home-base-exploiting sites require high levels of market-related knowledge flows. Thus, we propose the following two hypotheses.

**Hypothesis 1:** Home-base-augmenting R&D subsidiaries are likely to promote technology-related knowledge flows among the subsidiary, its parent company, other subsidiaries and its local environment.

**Hypothesis 2:** Home-base-exploiting R&D subsidiaries are likely to promote market-related knowledge flows among the subsidiary, its parent company, other subsidiaries and its local environment.

**Organizational Factors:** The literature on knowledge management has reported difficulties in encouraging knowledge sharing among employees—e.g., Garvin and March [42]. We argue that “Ba” or “communities-of-practices” emphasize democratic or egalitarian aspects of knowledge sharing processes, which are necessary for generating new meanings and innovations within MNCs, by mitigating “prisoner’s dilemma” situations [129]. Brown and Duguid [21], and Zaltman and Duncan [136] also support a similar argument.

Many studies on organizational communications have suggested that trustful environments encourage knowledge flows among various units. For example, Bresman et al. [19] and Argote [7] argue that quick and effective sharing of knowledge across units can be significantly facilitated, when units share visions and beliefs. Recent research on social capital has also suggested that shared vision and trust strongly influence inter-unit knowledge flows [94, 127].

Researchers on MNCs have argued that “autonomous subsidiaries”—i.e., with decentralized decision-making practices and localized employees—and “incentive practices” are two main factors that are related to such trustful environments—e.g., Gupta and Govindarajan [52] and Zander and Kogut [137].

We argue that autonomous subsidiaries tend to engage in more active knowledge flows from local environments—e.g., local universities or local suppliers—by building a local image for the subsidiary. Bartlett and Ghoshal [12], and Aghion and Tirole [1] agree with our argument in that autonomous subsidiaries are likely to contribute to new technologies or products by active knowledge flows.

However, prior studies have also suggested that there is a danger that autonomous subsidiaries in MNCs run by local managers tend to develop into their own little fiefdoms, thus inhibiting knowledge flows between headquarters and subsidiaries [43, 46]. According to the above arguments, we propose the following two hypotheses.

8Such a dilemma seems to also stem from the “public good” nature of information—i.e., the same information can be used simultaneously without additional costs to others [80].

9In the study of MNCs, “centralization-decentralization or localization” has been fundamental dimensions of organization designs [38]. “Centralization and decentralization” are concerned with decision-makings, while “localization” is concerned with people [2]. Thus, “autonomous” subsidiary has decentralized decision-making practices and more localized employees.

**Hypothesis 3:** The greater autonomy of an R&D subsidiary, the higher knowledge flows from its local environments to the subsidiary.

**Hypothesis 4:** The greater autonomy of an R&D subsidiary, the lower knowledge flows between the subsidiary and its parent company.

Another organizational factor affecting knowledge flows is related to “incentives.” Lane and Lubatkin [81] argue that incentive practices generally have positive associations with a firm’s inter-organizational learning. Brelade and Harman [18] and Davenport et al. [32] argue that a firm should emphasize both extrinsic rewards (e.g., bonuses) and intrinsic rewards (e.g., interesting or challenging projects) in order to encourage knowledge flows among employees. This argument seems valid, particularly for R&D personnel who tend to emphasize intrinsic rather than extrinsic rewards [4].

Studies on knowledge management have also reported that process-oriented incentives (i.e., based on leadership, teamwork, or commitment) may encourage knowledge sharing among employees, while outcome-based incentives (e.g., merit-based bonuses) may discourage knowledge sharing [53, 55]. The reason for this could be that knowledge is intimately connected with people’s egos.

Furthermore, Burgelman [22], Ghemawat et al. [45], and Gupta and Govindarajan [51] argue that incentives for R&D personnel should generally be flexible and loosely coupled to returns of R&D, particularly in basic research-related activities, where the returns of R&D activities are not certain, and often times, highly risky [85].

Accordingly, we argue that R&D subsidiaries’ incentives should be linked with long-term and process- (or behavior-) oriented measurements (e.g., leadership, teamwork, or commitment) rather than short-term and outcome-oriented measurements (i.e., bonuses based on number of patents or papers). Thus, we propose the following hypothesis.

**Hypothesis 5:** The more process-oriented incentive practices an R&D subsidiary employs, the higher knowledge flows between the subsidiary and its parent company.

**D. Knowledge Accumulation and Performance**

Evaluating the performance of an R&D subsidiary simply by financial measures such as ROI is not appropriate, because: 1) the result of R&D activities (i.e., knowledge) is not directly measurable, especially when tacit knowledge is involved [9]; 2) specific knowledge created in a subsidiary may have long-term pervasive impacts on HQs or other subsidiaries, occasionally without additional costs [45, 66]; 3) subsidiaries may specialize in different technological areas, and have different strategic priorities, thus requiring performance data to be adjusted for each subsidiary [50].

8By following Kuehmerle [78], we excluded knowledge flows from a subsidiary to local environments in our analyses, because: 1) such flows have been studied as “spillovers,” mainly form a public policy perspective; 2) it is usually difficult to specify local partners for such knowledge flows.

9For example, our interview with a director of an Japanese R&D subsidiary in the US emphasized the importance of “citation-index” as a performance measure of the subsidiary.
Thus, most of the empirical studies from the KBV of MNCs either: 1) used knowledge flows as performance indicators of HQs and/or subsidiaries; or 2) assumed that a high level of knowledge flows leads to high performance of HQs and/or subsidiaries, as discussed. However, we argue that knowledge flows are cost-involving activities—i.e., there must be an optimal level of knowledge flow. We also argue that knowledge flows lead to knowledge accumulations, and that knowledge accumulations lead to performance.

Recent studies on the relationship between information flows and performance in a NPD context support our argument. These studies found: 1) information flows had a positive effect on NPD performance under a specific condition—e.g., when levels of ambiguity are high [118]; 2) information which was exchanged and “used” had a positive effect on NPD performance [107]; 3) accumulated knowledge functions as an important “bridge” variable between information flows and NPD performance [79]; 4) accumulated knowledge lead to NPD performance, through informed, efficient and timely decisions [79].

Accordingly, we argue: it is not knowledge flows, but the knowledge that is actually accumulated and used which contributes to performance. Thus, we propose the following two hypotheses.

Hypothesis 6: The higher knowledge flows among an R&D subsidiary, its parent company, other R&D subsidiaries, and its local environments, the higher level of knowledge the subsidiary tends to accumulate.

Hypothesis 7: The higher level of knowledge an R&D subsidiary accumulates, the higher the subsidiary’s performance.

III. METHODOLOGY

A. Sample and Data Collection

In this study, we examine Japanese MNCs’ R&D subsidiaries in the US. There are some case studies that shed light on Japanese MNCs’ global R&D activities. These studies revealed the following characteristics of Japanese MNCs, compared with American and European counterparts: 1) established more recently smaller-scaled R&D subsidiaries [78]; 2) tend to co-locate with their manufacturing/marketing facilities [133], [135]; 3) tend to maintain stronger headquarter control over their R&D subsidiaries by sending expatriates [12]; 4) tend to give priority to development and design for manufacturability, and the associated speed of new product introduction [133]; 5) are developing international network of basic research [59].

However, there are no empirical studies on Japanese R&D subsidiaries in the US from the KBV, as discussed. Thus, the current study fills the gap among the previous studies on Japanese MNCs’ R&D activities abroad. Furthermore, we believe that this research sheds light on a general aspect of R&D globalization, because of the following three reasons regarding the population.

First, the US offers the most attractive “host” location for global R&D subsidiaries of MNCs. For example, Pearce and Papanastassiou [110] found that 530 global R&D laboratories owned by MNCs originated from a variety of countries, and that 20% of them are located in the US, followed by UK (11.5%), and Japan (9.4%).

Second, many empirical studies have indicated that a majority of R&D facilities in the US owned by foreign firms are Japanese. For example, Florida and Kenney [39] found that there were 174 R&D facilities in the US owned by foreign companies, and that 44.3 percent of them (77 facilities) were owned by Japanese MNCs.

Dalton et al. [29] also found that there were 715 R&D facilities in the US owned by foreign companies, and that 35% of them (250 facilities) were owned by Japanese MNCs, followed by German (15%) and Korean (14%) MNCs. Serapio and Dalton [117] also found that among the 25 largest R&D facilities in the United States, nine were Japanese.

Third, a majority of Japanese MNCs’ overseas R&D facilities are reported to be located in the US. For example, Iwata [70] reported that slightly more than 50% of all Japanese MNCs’ overseas R&D facilities are located in the US, followed by European and Asian countries. Furthermore, it is reported that some Japanese MNCs, especially those in the pharmaceutical industry, are increasing their R&D investments mainly to the US, in spite of the recent overall economic downturn [72]. Accordingly, we believe that Japanese R&D subsidiaries in the US represent (at least, quantitatively) the largest population—i.e., roughly 8% of the total number of MNCs’ R&D subsidiaries in the world.10

We collected data by surveying top managers (CEOs or presidents) of Japanese R&D subsidiaries. We supplemented this information with publicly available data. Because these subsidiaries were rather small in size—i.e., the average number of R&D personnel is 51—top management is assumed to have a thorough understanding of their subsidiaries. Therefore, the problem of “single respondent” bias was mitigated (although not perfectly resolved).

After developing the survey instrument in both English and Japanese, we asked seven managers of Japanese MNCs who were in charge of global R&D management to provide comments on our instrument, particularly on our definition of knowledge flows. Based on these comments, we revised some of the wording of the questions.

The Toyo Keizai’s directory on Japanese MNCs (Kaigai Shinshutsu Kyigou Souran) [125] identified 530 Japanese MNCs: 1) who were reported to conduct R&D activities in the US; 2) whose parent companies’ equity investments were higher than 50%. The directory identified not only independent R&D facilities, but also R&D activities conducted within manufacturing facilities.

By collaborating with the Japanese Ministry of Economy, Trade and Industry (METI), we mailed surveys to the top management of these subsidiaries. We received 85 responses—a response rate of 16 percent. However, we found that about one-half (42 companies) of the 85 respondents were either “planning” R&D activities, or not actually conducting any R&D activities. Thus, we mailed a second-wave of surveys to those MNCs that did not respond to our initial survey. We received 80 responses—a response rate of 15.4 percent. Again, we found that about one-half (38 companies) of the 80 respondents were

10$39\times(44.3\%+35\%)/2 = 79.7\%$ 20% is from Pearce and Papanastassiou [109]; 44.3% is Florida and Kenney [39]; 35% is from Dalton, et al. [29].
either planning R&D activities or not actually conducting R&D activities.

Accordingly, approximately half of the Japanese MNCs listed in the Toyo Keizai’s directory were not actually conducting significant R&D activities. This finding is consistent with Serapio and Dalton’s [117] finding—i.e., they found 249 Japanese R&D subsidiaries in the US. Six responses had significantly missing data and could not be used. Thus, the current study is based on the remaining 79 respondents (40 from the first-wave and 39 from the second-wave), representing roughly one-third of the population—i.e., 79 from 249 subsidiaries by Serapio and Dalton [117].

We checked non-respondent bias by observing the size of the subsidiaries and their parent companies that did not respond to our survey. We found that non-responding subsidiaries and their parent companies were not statistically different in size from those who did respond. We also checked those who responded to our first-wave survey with those who responded to the second-wave survey. There were no statistically significant differences between these two groups. Thus, we are confident that our sample represents the target population.

These R&D subsidiaries had, on average, 93.8% equity investments from Japanese parent companies. Twenty-one percent of these subsidiaries were established as merged or acquired firms. Seventy-two percent of these subsidiaries are co-located with either manufacturing or marketing facilities. The sample is composed of automobile-related companies (24%), electronics-related firms (24%), pharmaceutical companies (23%), and others such as chemical firms.

The average age of the subsidiaries was 12 years. There were, on average, 51 R&D personnel at each subsidiary, representing 5.5% of the total number of R&D personnel in the selected Japanese MNCs. Each subsidiary had, on average, 15 Japanese (expatriated) R&D personnel. Fifty-five percent of the respondents (i.e., top managers) were Japanese.

B. Measurements

Dependent Variables: Based on Kuemmerle [77], we measured four types of knowledge flows—i.e., Technical Knowledge Flow, Market Knowledge Flow, Sub-HQ Knowledge Flow, and Sub-Local Knowledge Flow—by asking respondents the extent to which their R&D subsidiaries send/receive (a) technology-related know-how/information, and (b) market-related know-how/information from/to a R&D subsidiary/HQ and from/to other R&D subsidiaries, and from local environments—i.e., local universities, local governments, customers, suppliers, competitors—by using ten 5-point Likert scales. (See Appendix for details about our measures.)

We conducted a factor analysis (varimax-rotation) on these ten scales. The factor analysis identified one more factor for which: 1) the eigenvalue exceeded one, and 2) the reliability index (Cronbach alpha) among the scores exceeded 0.70. The factor consisted of technology- and market-related knowledge flows between an R&D subsidiary and other subsidiaries. We included this factor—labeled as Sub-Sub Knowledge Flow—as an additional dependent variable in order to further explore our framework.

Accordingly, we have five types of knowledge flows. The first two types—i.e., Technical Knowledge Flow and Market Knowledge Flow—are regarding “what” or “contents” of flowing knowledge, while the other three types—i.e., Sub-HQ Knowledge Flow, Sub-Local Knowledge Flow and Sub-Sub Knowledge Flow—are regarding “where” or “places” of flowing knowledge.

Independent Variables: R&D Strategy consists of HBA Strategy and HBE Strategy. By following Kuemmerle [77], HBA Strategy was measured by two question items describing a home-base augmenting R&D subsidiary, while HBE Strategy was measured by three question items describing a home-base exploiting R&D subsidiary. Following Hedlund [60], Autonomy was measured using five questions items describing the degree to which each subsidiary is localized in its employment, and decentralized or autonomous in its decision-making. By extending Minbaeva et al. [91], Incentive was measured using four Likert-type questions regarding 1) attitude or commitment, 2) leadership, 3) teamwork, and 4) potential capabilities.

Control Variables: Yoshihara et al. [135] and Westney [133] suggested the tendency of Japanese R&D facilities to be located with their manufacturing/marketing facilities abroad. Thus, a control variable (a dummy), Co-location, was included in our analyses. Mergers/acquisitions and greenfield developments are the two most common methods of establishing R&D subsidiaries abroad [25]. Thus, we also included M&A as a control (dummy) variable. It is possible that the longer a subsidiary has been established in US, the more efficient it has become in its knowledge flow management. Thus, Subsidiary Experience was also measured by the number of years an R&D subsidiary had been established in US.

The larger subsidiaries may have more autonomy [132], or may require more parents’ attentions [60]. Thus, Sub Size was included, by measuring the number of R&D personnel in a subsidiary. We also measured the size of each parent company, HQ Size, by the total number of employees. Firms in pharmaceuticals, chemicals, plastics, ceramics, and metal products were assumed to have different attitudes toward intellectual properties from firms in electrical appliances, automobiles and parts, transport equipment, and precision machines [95]. Thus, firms in the former industries—considered “mono technology-based firms”— were coded as 0, while those in the latter industries—considered “system technology-based firms” were coded as 1.

Knowledge Accumulation and Performance: We argue that knowledge accumulations and subsidiaries’ performance should be measured by using a variety of objective measures—e.g., number of patents, copyrights, papers, or new products—as
well as subjective measures—e.g., Likert-scales—because subsidiaries in different industries must have different strategic priorities, thus requiring performance data to be adjusted [50]. Although patents are frequently used in previous studies to measure a level of knowledge accumulations [95], “meanings” of patents seem to differ significantly across firms. For example, firms with high manufacturing process technologies usually do not file patents because of fear of information leaks [26]. Our preliminary interviews with the seven managers in these R&D subsidiaries also revealed 1) home-base-augmenting subsidiaries tended to encourage paper publications; 2) home-base-exploiting subsidiaries tended to encourage paper publications; 3) subsidiaries in chemical and pharmaceutical industries generally tended to encourage both patent filing and paper publications; 4) subsidiaries in electronics and automobile industries generally tended to encourage patent filing, but discourage paper publications.

Studies on knowledge and technology transfer support our argument, as well. For example, Szulanski [120] used subjective measures of satisfaction with the process and/or outcome of knowledge transfers as proxy variables. Almeida et al. [6] also used combinations of quantitative and qualitative methods for knowledge transfer performance.

Furthermore, studies on multi-unit firms from the KBV also support our argument. For example, Tsai [126] employed an innovation achieved rate—i.e., the number of new products introduced in a unit in a particular year divided by the unit’s target number in that year—to measure the performance of innovation in each unit.

Accordingly, the level of knowledge accumulation—Knowledge Accumulation—was measured both quantitatively and qualitatively. Quantitative Knowledge Accumulation consisted of two parts: 1) the number of published papers—named as Quantitative Knowledge Accumulation-1—and 2) the total number of patents awarded, copyrights awarded, and new products/services created by utilizing knowledge developed in a subsidiary—named as Qualitative Knowledge Accumulation-2. Qualitative Knowledge Accumulation was measured by using three Likert-type question items asking respondents qualitatively about the level of knowledge accumulation at each subsidiary.

Finally, we measured subsidiaries’ overall performance, as hypothesized. Overall Subsidiary Performance was measured using seven Likert-type questions asking respondents the extent to which their subsidiary had been successful in 1) technologies, 2) speed, 3) technical goals achieved, 4) financial goals achieved, 5) technical contributions to their parent company, 6) financial contributions to their parent company, and 7) overall assessment of a subsidiary’s R&D performance.

We employed the following two methods to avoid common method bias. First, the dependent variables were placed after the independent variables in the survey to diminish the effects of consistency artifacts. Second, Harman’s single factor test was performed [57], [113]. The factor analysis revealed five factors with eigenvalues greater than 1.0 that accounted for 70.2% of the total variance. The first factor only accounted for 28.4% of the variance. These results suggested that common method variance was not a serious problem in our study.

Discriminant validity assesses the degree to which two theoretically similar constructs are distinct. Because of our multiple uses of the original ten scores in constructing the five knowledge flows, we conducted discriminant analyses by excluding the five knowledge flows. The results supported the discriminant validity, by using three criteria assessments proposed by Campbell and Fiske [23].

Fig. 1 represents our model. As discussed, the model consists of three independent variables—R&D Strategy, Autonomy, and

![Fig. 1. Determinants of global knowledge flows and performance.](image-url)

IV. RESULTS

Table II shows the descriptive statistics (mean and standard deviation) and correlation matrix among all of the defined variables—reliability indices (Cronbach’s alphas) are also reported on the diagonal. The mean scores in the defined seven knowledge flows indicate that Sub-Local Knowledge Flow had the highest score (3.622), while Sub-Sub Knowledge Flow had the lowest score (1.635)—the difference was statistically significant at the one percent level by our t-test. 14

Since Gupta and Govindarajan [52] did not report a similar difference among their four types of knowledge flows, our data may indicate that knowledge flows from local environments to R&D subsidiaries are generally high, while knowledge flows among overseas subsidiaries are generally low. Such low knowledge flows among subsidiaries may stem from Japanese MNCs’ lack of experience, as compared with American and European MNCs.

There are high correlations among the five knowledge flows—Technical Knowledge Flow, Market Knowledge Flow, Sub-HQ Knowledge Flow, Sub-Local Knowledge Flow, and Sub-Sub Knowledge Flow. These high correlations suggest that special care must be taken to avoid multicollinearity problems in the following multi-regression analyses.15

There are also moderately high correlations between Qualitative Knowledge Accumulation and Overall Performance, and between Qualitative Knowledge Accumulation and the five knowledge flows. These correlations seem to suggest a possibility that high levels of knowledge flows generally lead to moderate levels of knowledge accumulation and overall performance.

A correlation between M&A and Autonomy (0.265) suggests that a site started by M&A is likely to be more localized and decentralized. An association between Autonomy and Sub Size (0.193) implies that the longer a subsidiary operates in the US, the more localized and decentralized it becomes. An association between Incentive and HBA Strategy (0.335) also suggests that home-base-augmenting subsidiaries tend to have process-oriented incentive practices.

A positive correlation between HBA Strategy and Sub Size (0.271), and a negative correlation between HBA Strategy and Industry (−0.224) implies that large subsidiaries in mono-technology based industries tend to pursue home-base-augmenting strategies. There are also positive associations among HBE Strategy, Experience, and Industry. These associations imply that home-base-augmenting subsidiaries tended to be in system-technology-based industries such as automobile or electronics, and started their operations in the US earlier.

Co-Location indicated positive correlations with HQ Size (0.224) and Industry (0.181), suggesting that R&D subsidiaries with large parents in system-technology based industries tend to moderate levels of knowledge accumulation and overall performance.

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Numbers are Pearson correlation coefficients; *p<0.10, **p<0.05, ***p<0.01 (two-tailed). Reliability index (Cronbach’s alphas) are on the diagonal.

14The first discriminant validity criterion—i.e., to ascertain the level of general method variance contained within the matrix—was met because values within the validity diagonal were significantly higher than the values contained within the heterotrait-heteromethod triangles in 98% of the cases. The second discriminant validity criterion—i.e., to ascertain if the validity diagonal coefficients are greater than the heterotrait-monomethod coefficients—was satisfied in 95% of the cases. The third discriminant validity criterion—i.e., to ascertain if the same pattern of trait interrelationships is found in both the heterotrait-heteromethod coefficients and the heterotrait-monomethod coefficients—was met, because Kendall’s Coefficient of Concordance, was found to be .321 at p < .0001 which suggests that there is some similarity in the coefficient patterns. These results support the discriminant validity.

15Such high correlations were expected, because of our multiple uses of the original ten scores. Such high correlations among various types of knowledge flows were also reported by Gupta and Govindarajan [52].
to co-locate their facilities with their manufacturing facilities. Co-Location also indicated negative correlations with HBA Strategy (−0.266) and Sub Size (−0.273), suggesting that larger home-base-augmenting R&D subsidiaries are not likely to co-locate with their manufacturing facilities.

As we hypothesized, we have three models to test seven hypotheses. Hypotheses 1–5 were concerned with the determinants of knowledge flows. Hypothesis 6 is related to the determinants of knowledge accumulation. Finally, Hypothesis 7 is concerned with the determinants of the overall performance of subsidiaries. Thus, in order to test these hypotheses, we conducted three series of multi-regression analyses—i.e., knowledge flows, knowledge accumulation, and overall performance, as three dependent variables. Table III presents the first series of regression analyses.

Column 1 in Table III presents the relationships between the independent variables and Technical Knowledge Flow among a subsidiary, its parent company, other subsidiaries and its local environment. There is no significant beta coefficient between R&D Strategy (HBA Strategy and HBE Strategy) and Technical Knowledge Flow, thus lending no support for Hypothesis 1—home-base-augmenting R&D subsidiaries are likely to promote technology-related knowledge flows among the subsidiary, its parent company, other subsidiaries and its local environment.

Column 1 does indicate, however, that Technical Knowledge Flow has a highly positive association with Incentive, suggesting that process-based incentives encourage technical knowledge flows. Column 1 also indicates that larger subsidiaries tend to discourage such technical knowledge flows. We found that the size of the parent companies in the pharmaceutical industry were significantly smaller than those in other industries, and that our regression analyses when excluding these pharmaceutical companies did not show such a significant beta. Thus, the result seems to indicate that R&D subsidiaries in the pharmaceutical industry depend on more technical knowledge flows.

Column 2 in Table III reports the relationships between Market Knowledge Flow and the defined independent variables. The results provide support for Hypothesis 2—home-base-exploiting R&D subsidiaries are likely to promote market-related knowledge flows among the subsidiary, its parent company, other subsidiaries and its local environment.

Column 3 in Table III examines the relationships between HQ-Sub Knowledge Flow and the independent variables. The results do not support Hypothesis 4—the greater autonomy of an R&D subsidiary, the lower the knowledge flows between the subsidiary and its parent company. (The reason for this will be examined further in Section V.) However, the results in Column 3 do support Hypothesis 5—the more process-oriented incentive
TABLE IV
FACTORS AFFECTING KNOWLEDGE ACCUMULATION

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<tr>
<td>.790</td>
<td>.230***</td>
<td>.210**</td>
<td>.233***</td>
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</table>

Numbers are standardized beta; * p<0.10, **p<0.05, ***p<0.001

practices used in an R&D subsidiary, the higher the knowledge flows between the subsidiary and its parent company.

Column 4 in Table III reports the relationships between Sub-Local Knowledge Flow and the independent variables. Autonomy shows a significantly high association with Sub-Local Knowledge Flow, thus supporting Hypothesis 3—the greater autonomy of an R&D subsidiary, the higher the knowledge flows from its local environment.

Column 4 also shows that HBA Strategy tends to discourage Sub-Local Knowledge Flow. Sub-Local Knowledge Flow consists of two indicators—1) technology-related knowledge flows from local environments, and 2) market-related knowledge flows from local environments to the U.S. subsidiary. Further regression analyses using the two original dependent variables showed that market-related knowledge flows had a highly negative beta with HBA Strategy, while technology-related knowledge flow did not. Thus, results seem to partially support our assumption that HBA Strategy discourages market-related knowledge flows.

Column 4 indicates that Sub-Local Knowledge Flow tends to be greater when the subsidiary is large. This result is understandable because larger subsidiaries (measured by number of R&D personnel) tend to develop more networks with local environments such as local universities, local governments, customers, suppliers, and/or competitors.

Column 5 in Table III reports the relationships between Sub-Sub Knowledge Flow and the independent variables. Column 5 shows that Sub-Sub Knowledge Flow tends to be greater when the subsidiary employs HBE Strategy. This result supports Nobel and Birkinshaw [98] in the sense that some types of R&D subsidiaries encourage “lateral communications” among global R&D subsidiaries, while others do not.

The above results can be summarized as follows: 1) since two of the five hypotheses were not supported, our results cast doubt on our types of knowledge flows; 2) a company’s R&D strategy seemed to determine levels of knowledge flows, especially market-related and lateral (subsidiary-subsidiary) knowledge flows; 3) Autonomy and Incentive had positive effects on knowledge flows; 4) control variables—i.e., Co-Location, M&A, Experience, and Industry—generally had no significant effect on knowledge flows.

Knowledge Accumulation: Table IV examined the factors affecting these types of knowledge accumulations—i.e., Quantitative Knowledge Accumulation-1, Quantitative Knowledge Accumulation-2, and Qualitative Knowledge Accumulation. Since correlations among the five types of knowledge flows were very high, we conducted seven separate regression analyses for the three types of knowledge accumulations.

Column 1 to Column 5 examined factors affecting the number of papers in subsidiaries. Our regression analyses found that home-base-augmenting subsidiaries tend to have higher numbers of papers, because three out of the five regressions showed highly positive betas on HBA Strategy. These regression analyses also indicated that Industry had highly negative associations with the number of papers, suggesting that firms in pharmaceuticals, chemicals, plastics, ceramics, and metal products tend to publish more papers than those in electrical appliances, automobiles and parts, transport equipment, and precision machines.

Furthermore, Sub-Local Knowledge Flow in Column 4 of the table showed a highly positive association, thus, partially
TABLE V
FACTORS AFFECTING R&D SUBSIDIARIES’ OVERALL PERFORMANCE

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<td>2.877***</td>
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<tr>
<td>Adjusted (R^2)</td>
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<td>.294</td>
<td>2.988***</td>
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</tbody>
</table>

|                          |        |                  |      |
| F                        |        |                  | 3.034***|

Numbers are standardized beta; * p<0.10, **p<0.05, ***p<0.01

supporting Hypothesis 6—the higher knowledge flows among a subsidiary, its parent company, other R&D subsidiaries and its local environment, the higher the level of knowledge a subsidiary tends to accumulate.

Column 6 to Column 10 of Table IV indicates that there was no statistically significant beta regarding Quantitative Knowledge Accumulation-2—i.e., total number of patents, copyrights, and new products/services. Specifically, we did not find any association with the four types of knowledge flows, thus lending no support to Hypothesis 6. (The reason for this will be examined further in the following section.)

Column 11 to Column 15 of the table indicates highly positive associations between Qualitative Knowledge Accumulation and the five types of knowledge flows—Technical Knowledge Flow, Market Knowledge Flow, HQ-Sub Knowledge Flow, Sub-Local Knowledge Flow, and Sub-Sub Knowledge Flow—thus, partially supporting Hypothesis 6.

Highly positive betas were also found regarding HBE Strategy, suggesting that home-base-exploiting subsidiaries tend to accumulate a high level of qualitative knowledge. It should also be noted that strategic factors, such as HBA Strategy and HBE Strategy, contributed positively to knowledge accumulations, while organizational factors, such as Autonomy and Incentive, did not.16

Overall Performance: Table V shows our regression analyses on subsidiaries’ overall performance. Since correlations among the five types of knowledge flows were very high, we conducted

16Tests for multicolinearity did not reveal any serious problems in the 21 regression analyses in Table IV. The values for the variance inflation factor (VIF) used to check for multicolinearity were all below 1.9.
five separate regression analyses. The table indicated that three out of the five types of knowledge flows had significantly positive betas—i.e., Market Knowledge Flow, Sub-HQ Knowledge Flow, and Sub-Local Knowledge Flow.

The table also indicates that knowledge accumulation, especially Qualitative Knowledge Accumulation, significantly contributes to the subsidiaries’ overall performance, moderately supporting Hypothesis 7—the higher the level of knowledge an R&D subsidiary accumulates, the higher the subsidiary’s performance.

Since the three types of knowledge accumulations seem to contribute more to the overall performance than do the five types of knowledge flows, these results may suggest the validity of our assumptions in this study—i.e., “knowledge flows are cost-involving activities” and “knowledge accumulations lead to high performance.”

Furthermore, Table V indicates that both Autonomy and Incentive—organizational factors—had highly positive betas, suggesting that autonomous R&D subsidiaries with process-oriented incentives tend to have high overall performance, while strategic factors such as HBA Strategy and HBE Strategy do not. This finding suggests the importance of organizational factors (rather than strategic factors) for subsidiaries’ overall performance.17

V. DISCUSSION

This study attempts to fill a gap in the literature on the knowledge-based view of MNCs by measuring the impact of knowledge flows on knowledge accumulations and overall performance in R&D subsidiaries of Japanese MNCs in the US. The study’s survey results provide partial support for our model in Fig. 1, because three out of the seven hypotheses (Hypothesis 2, 3 and 5) were fully supported, and two hypotheses (Hypothesis 6 and 7) were partially supported.

However, Hypothesis 1—the more basic research-oriented an R&D subsidiary, the higher the technology-related knowledge flows among the subsidiary, its parent company, other R&D subsidiaries and its local environment—and Hypothesis 4—the greater autonomy of an R&D subsidiary is, the lower the knowledge flows between the subsidiary and its parent company—were not substantiated. Although not reported here, we conducted further analyses to help us understand these findings. We found a similar reason why these two hypotheses were not supported.

Hypothesis 1 was constructed based on the assumption that a more basic research-oriented R&D subsidiary would encourage two-way (or six-way) knowledge flows among the subsidiary, its local environment and its parent company. However, further analyses suggested that home-base-augmenting subsidiaries tended to encourage technology-related knowledge flows from the subsidiary to its parent company, while discouraging technology-related knowledge flows from the parent company to the subsidiary. Thus, HBA Strategy tended to cancel the summed effects on the technology-related knowledge flows.18

Hypothesis 4—the greater autonomy of an R&D subsidiary, the lower the knowledge flows between the subsidiary and its parent company—was constructed based on our assumption that greater autonomy of R&D subsidiaries would discourage two-way knowledge flows between the R&D subsidiaries and their parent companies.

However, further analyses suggested that the greater autonomy of R&D subsidiaries encouraged knowledge flows from the subsidiary to its parent company, while discouraging knowledge flows from the parent company to the subsidiary. We constructed our hypotheses based on the assumption that knowledge flows are reciprocal—i.e., not unidirectional. However, our results revealed that this assumption is not valid in the relationship between HQs and subsidiaries.19

By including three types of knowledge accumulation indicators, our analyses partially supported Hypothesis 6—the higher the knowledge flows among an R&D subsidiary, its parent company, other R&D subsidiaries and its local environments, the higher the level of knowledge the subsidiary tends to accumulate. Although Qualitative Knowledge Accumulation supported the hypothesis, Quantitative Knowledge Accumulation-2 did not support the hypothesis.

Although not reported here, we conducted further regression analyses, by 1) using the original three indices—patents, copyrights, and new products/services—and 2) including curvilinear items for the four types of knowledge flows. We found some statistically significant curvilinear associations, for example, between New Products/Services and Sub-Local Knowledge Flow. Fig. 2 shows this relationship graphically. The results seem to partially substantiate our assumption that knowledge flows are cost-involving activities.

In the new product development (NPD) literature, several empirical studies have shown that too much information or knowledge flows occasionally hurt NPD performance—e.g., Kurokawa et al. [79], Sicotte and Langley [118], and Ottum and Moore [107]. Thus, our current study may suggest “an optimum level of knowledge flows.”

Hypothesis 7—the higher the level of knowledge an R&D subsidiary accumulates, the higher the subsidiary’s performance—seems to be moderately supported, because Qualitative Knowledge Accumulation and Quantitative Knowledge Accumulation generally showed positive associations with subsidiaries’ overall performance, as seen in Table V. Table V also shows that three out of the five types of knowledge flows had

17In order to test mediation effects between knowledge flows and accumulations on performance, we conducted regression analyses by including interaction terms. All the variables were mean centered in order to avoid multicollinearity problems [112]. We found that there are some mediation effects between knowledge flows and accumulations. However, these interaction effects were statistically significant at the 10 percent level, but not significant at the 5 percent level.

18We conducted regression analyses by using original knowledge flow scores, as dependent variables. Our analyses indicated that HBA Strategy had a highly negative beta (−0.201) on technology-related knowledge flow from HQ to sub, while HBA Strategy had a highly positive beta (0.286) on technology-related knowledge flow from sub to HQ.

19We conducted regression analyses by using original knowledge flow scores, as dependent variables. Our analyses indicated that Autonomy had a highly negative beta (−0.234) on technology-related knowledge flow from HQ to sub and a highly negative beta (−0.256) on market-related information flow from HQ to sub. On the other hand, our analyses also indicated that Autonomy had a highly positive beta (0.278) on technology-related knowledge flow from sub to HQ and a highly positive beta (0.289) on market-related information flow from HQ to sub.
highly positive associations with overall performance—i.e., the higher the knowledge flows, the higher the performance. Thus, our results seem to substantiate the knowledge-based view of MNCs [52], [64].

Table V indicates that Market Knowledge Flow had a highly positive association with subsidiaries’ overall performance, while Technical Knowledge Flow did not. Our finding supports Subramaniam and Venkatraman [119] in that market-related knowledge flows tended to contribute to performance more than technology-related knowledge flows did. In the NPD literature, studies also showed that market-related information or knowledge flows tend to contribute to NPD performance more than technical information or knowledge flows do—e.g., Kurokawa et al. [79].

Furthermore, Table V shows that vertical knowledge flows (i.e., Sub-HQ Knowledge Flow) had highly positive associations with overall performance, while horizontal knowledge flow (i.e., Sub-Sub Knowledge Flow) did not. These results support Gupta and Govindarajan [52, p. 490] in that “the parent corporation continues to serve as the most active creator and diffuser of knowledge within the corporation.”

As discussed, Japanese MNCs tend to: 1) establish more recently smaller-scaled R&D subsidiaries; 2) co-locate with their manufacturing/marketing facilities; 3) maintain stronger headquarter control over their R&D subsidiaries by sending expatriates; 4) give priority to development and design for manufacturability, and the associated speed of new product introduction; 5) be developing international network of basic research.

We confirmed the first and second characteristics of Japanese R&D subsidiaries by our data. However, the rest of the characteristics were neither confirmed nor suggested by our data. Thus, in order to supplement our findings, we interviewed ten parent companies in Japan by randomly selecting from the survey respondents.

By reflecting our survey population, we selected three companies in the automobile industry, four in the electronics industry, and three in the pharmaceutical industry. We interviewed with either chief technology officers (CTOs) or top managers in charge of global R&D subsidiaries for approximately two hours.

We took unstructured (open-end) approach in order to find factors unidentified in our survey results. Most of interviews were tape-recorded with the approval of interviewees.

We found the following three tendencies from these interviews at HQs: 1) managers at HQs generally tended to underestimate the performance of their R&D subsidiaries in the US, in line with [135]; 2) managers at HQs tended to point out serious communication and language problems between HQs and their subsidiaries, as well as between Japanese expatriates and American researchers, in line with [10] and [11]; 3) MNCs with strong corporate culture—e.g., Toyota, Canon and Panasonic—tended to develop special methods to solve knowledge flow problems, in line with [133].

These findings seem to be consistent with the characteristics of Japanese R&D subsidiaries reported in previous case studies, in the sense that Japanese MNCs attempt to maintain stronger headquarter control over their global R&D subsidiaries, by setting high performance goals for prompt introductions of new products in a global scale.20

VI. CONCLUSION

This study attempts to fill a gap in the literature on the knowledge-based view of MNCs by focusing on R&D subsidiaries of Japanese MNCs. We had the following five research questions: 1) what factors will determine knowledge flows among R&D subsidiaries, headquarters (HQ), and other R&D subsidiaries?; 2) what factors will determine knowledge accumulations of R&D subsidiaries?; 3) what factors will determine performance of R&D subsidiaries?; 4) how should R&D subsidiaries manage knowledge flows for higher performance?

Regarding the first research question—i.e., factors determining knowledge flows—we found that R&D strategies generally determine levels of knowledge flows, especially market-related and horizontal knowledge flows. We also found that R&D strategies tended to cancel the summed effects of the technology-related knowledge in-flows and out-flows between HQ and subsidiaries.

Furthermore, we found that process-oriented incentives stimulate knowledge flows. The previous studies on the KBV of MNCs found that autonomous R&D subsidiaries promote knowledge flows [52] and that network-linked (i.e., HQ or other subsidiaries-linked) incentives stimulate knowledge flows [40], [91]. Thus, our findings may be unique, although Iwata [70] also found similar results among American and European R&D subsidiaries in Japan.21

Regarding the second research question—i.e., factors determining knowledge accumulations—we found that strategic factors, such as home-base-exploiting or home-base-augmenting

20Furthermore, we found the following three industry-specific tendencies from the interviews: 1) knowledge flows from HQs to subsidiaries tended to be well managed in the automobile industry; 2) knowledge flows from subsidiaries to HQs tended to be poorly managed in the electronics industry; 3) knowledge flows between subsidiaries and HQs tended to be poorly managed in the pharmaceutical industry.

21The previous studies have identified that Japanese companies generally tend to emphasize process-oriented incentives than American and European counterparts—e.g., Kagono et al. [74]. Thus, our finding may simply reflect such Japanese companies’ characteristics.
strategies, were more likely to determine knowledge accumulations than were organizational factors, such as autonomous subsidiaries, or incentive practices.

Regarding the third research question—i.e., factors determining performance—we found that market-related and vertical knowledge flows tend to contribute more to overall performance than do technology-related and horizontal knowledge flows. Furthermore, we found that organizational factors were more likely to determine subsidiaries' overall performance than were strategic factors.

The above findings suggest that strategic factors tend to determine knowledge accumulations, while organizational factors tend to determine performance. Although we hypothesized that accumulated knowledge would directly lead to high performance, organizational factors may be intervening between accumulated knowledge and subsidiaries’ overall performance.

Regarding the fourth research question, this study has several implications for managers in charge of global R&D activities. First, our data analyses found that knowledge flows are more likely to be determined by managerial factors (i.e., R&D strategies, autonomy and incentives) than contextual (control) variables such as co-location, M&A, experience, and industry.

These findings suggest: 1) purposes and missions of R&D strategies should be clearly defined before launching global R&D sites, 2) subsidiaries should be allowed a higher level of autonomy for higher knowledge flows and performance, and 3) process-oriented incentive practices rather than outcome-oriented incentives should be emphasized.

Second, knowledge flows among HQs, R&D subsidiaries, and other subsidiaries should be encouraged, because the level of knowledge flows had positive associations with knowledge accumulation and subsidiary performance. Specifically, market-related and vertical knowledge flows should be emphasized more than technology-related and horizontal knowledge flows. However, managers should also be cautious about costs of knowledge flows, because we found too much knowledge flow may endanger knowledge accumulations.

Finally, the amount of specific knowledge accumulations, such as number of patents, copyrights, and papers, should not be overemphasized as performance indicators, because these numbers were not necessarily related with subsidiaries performance. Such numbers seemed to vary depending upon specific companies or industries, as discussed.

Our theoretical contribution to the study of MNCs is to substantiate the knowledge-based view of MNCs by our three-stage model. First, we identified factors affecting knowledge flows. Next, we examined the extent to which knowledge flows lead to knowledge accumulations. Finally, we examined the extent to which knowledge accumulations lead to subsidiaries’ overall performance. In general, we found that high levels of knowledge flows lead to high levels of performance.

Accordingly, this study provides moderate support for the knowledge-based view of MNCs. We also found that the defined control factors—e.g., co-location, M&A, experience, and industry—did not significantly affect knowledge flows, knowledge accumulations, or subsidiaries’ overall performance. These results seem to further substantiate the generality of the knowledge-based view of MNCs.

However, we also found that high levels of knowledge flows occasionally do not lead to high levels of knowledge accumulations and performance, supporting our assumption that knowledge flows are cost-involving activities. For example, an R&D subsidiary might be able to achieve high performance without frequent knowledge flows from HQ, because of its highly experienced top management who has already accumulated enough knowledge from running the subsidiary.

Our methodological contributions to the study of MNCs from the KBV were to measure 1) subsidiaries’ R&D strategies, 2) both technology- and market-related knowledge flows, 3) both vertical and horizontal knowledge flows, 4) both knowledge in-flows and out-flows, and 5) the effect of knowledge flows on knowledge accumulations and overall performance of subsidiaries.

However, the findings in this paper must be interpreted with care, because of the following limitations. First, although we argued that Japanese R&D subsidiaries in the US represent the largest population, the largest population does not necessarily represent the average population. Thus, the external validity of this study could be severely limited.

Second, the long-term effects of knowledge flows on subsidiary performance, as well as the reverse effects—i.e., subsidiary’s performance on knowledge flows—cannot yet be ascertained. Our interviews also revealed that managers at HQs generally tended to underestimate the performance of their R&D subsidiaries. Finally, there are also some concerns about the discriminant validity of the five knowledge flows.

Accordingly, the opportunity for further research into global R&D management is substantial. For example, further research on German or Korean R&D facilities in the US or Europe, as well as Japanese R&D subsidiaries in Asia or Europe, would broaden our understanding of this subject. Several in-depth case studies would also clarify further the relationships among the variables defined in this study. Without such research, it is uncertain whether these findings are specific to the Japanese MNCs in the current study or if they represent MNCs in general.

APPENDIX

MEASUREMENT

Dependent Variables

Based on Kuemmerle [77], [78], we measured ten knowledge flows by asking respondents the extent to which their R&D subsidiaries send/receive (a) technology-related know-how/information, and (b) market-related know-how/information, from/to the subsidiaries/HQ. We used the word, “know-how/information flow” instead of “knowledge flow” in our original questions. Although “knowledge flow” is popular in academic worlds, we found that it was not popular among practicing managers during our preliminary study with seven managers. We used 5-point Likert scales (1=not significant at all, 5=highly significant). In order to illustrate an extreme case where there is no information flow at all, we added “0” for no information flows at all.

Technical Knowledge Flow was measured by: 1) technology-related know-how/information flows from local environments to the U.S. subsidiary, 2) technology-related know-how/information flows from the R&D subsidiary to HQ, 3) technology-related know-how/information flows from HQ to the U.S. subsidiary, 4) technology-related know-how/information flows from the U.S. subsidiary to other R&D subsidiaries,
and 5) technology-related know-how/information flows from other R&D subsidiaries to the U.S. subsidiary. The Cronbach alpha among these five variables was 0.735.

**Market Knowledge Flow** was measured by: 1) market-related know-how/information flows from local environments to the U.S. subsidiary, 2) market-related know-how/information flows from the R&D subsidiary to HQ, 3) market-related know-how/information flows from HQ to the U.S. subsidiary, 4) market-related know-how/information flows from the U.S. subsidiary to other R&D subsidiaries, and 5) market-related know-how/information flows from other R&D subsidiaries to the U.S. subsidiary. The Cronbach alpha among these five variables was 0.760.

**Sub-HQ Knowledge Flow** was measured by: 1) technology-related know-how/information flows from HQ to the U.S. subsidiary, 2) market-related know-how/information flows from HQ to the U.S. subsidiary, 3) technology-related know-how/information flows from the U.S. subsidiary to HQ, and 4) market-related know-how/information flows from the U.S. subsidiary to HQ. The Cronbach alpha among these four variables was 0.694.

**Sub-Local Knowledge Flow** was measured by 1) technology-related know-how/information flows from local environments—e.g., local universities, local governments, customers, suppliers, competitors—to the U.S. subsidiary, and 2) market-related know-how/information flows from local environments to the U.S. subsidiary. The Cronbach alpha between these two variables was 0.689.

**Sub-Sub Knowledge Flow** was measured by 1) technology-related knowledge flows from the focused R&D subsidiary to other R&D subsidiaries, and 2) technology-related knowledge flows from other R&D subsidiaries to the R&D subsidiary, 3) market-related knowledge flows from the focused R&D subsidiary to other R&D subsidiaries, 4) market-related knowledge flows from other R&D subsidiaries to the R&D subsidiary. This variable was constructed by our factor analysis (varimax-rotation) among the ten scales of knowledge flows. We included this variable as an additional dependent variable in order to further explore our framework. The Cronbach alpha was 0.889.

**Quantitative Knowledge Accumulation** was measured by: 1) the number of patents awarded, 2) the number of copyrights awarded, and 3) the number of published papers, and 4) the number of new products by using technologies and designs developed in a R&D subsidiary. We conducted a factor analysis (Varimax rotation) and found two factors. One factor is the number of published papers. Another factor consists of the number of patents, the number of copyrights, and the number of new products. This factor is called Quantitative Knowledge Accumulation-1. The Cronbach alpha among the three indicators was 0.689—called Quantitative Knowledge Accumulation-2.

**Qualitative Knowledge Accumulation** was measured by the following three questions based on a 5-point Likert scale (1—not at all, 5=significantly): 1) to what extent, have you produced unique R&D results, by combining knowledge in your subsidiary and knowledge from your HQ?; 2) to what extent, have you produced unique R&D results by combining knowledge in your subsidiary and knowledge from your marketing and manufacturing sites in the US?; 3) to what extent have your R&D personnel advised or helped your parent company and other R&D subsidiaries? The Cronbach alpha among the three indicators was 0.722.

**Overall Subsidiary Performance** was measured by 7 Likert-type questions, by asking respondents the extent to which their R&D subsidiary had been successful in the following: 1) unique technologies developed, 2) speed to produce visible technical results, 3) technical goals achieved, 4) financial goals achieved, 5) technical contributions to their parent company, and 6) financial contributions to their parent company (1=very low, 5=very high). We also asked respondents for their overall assessment of the subsidiary’s R&D performance (1=not successful at all, 5=very successful). The Cronbach alpha for this measure was 0.801.

**Independent Variables**

**R&D Strategy** was measured by asking respondents the extent to which they agreed with the following five statements on five-point scales (1=strongly agree, 5=strongly disagree): 1) we started this site to strengthen our R&D capabilities by conducting R&D in the US where some areas of technologies are advanced, 2) we started this site to take advantage of a better R&D environment in the United States, 3) we started this site to employ and utilize researchers/engineers in the United States, 4) we started this site to be able to promptly respond to the needs of the U.S. market, and 5) we started this site to establish an integrated system ranging from R&D to production to sales in the U.S. market. The first three questions describe a home-base enhancing R&D subsidiary, while the fourth and fifth questions depict a home-base exploiting R&D subsidiary [77], [78]. The Cronbach alpha for the first three questions was 0.801 and 0.697 for the fourth and fifth questions. We call the first R&D strategy HBA Strategy, and the second, HBE Strategy.

**Autonomy** was measured by using five variables: 1) the number of U.S. researchers/engineers as a percentage of the total number of researchers/engineers in the R&D subsidiary, 2) a dummy variable coded “one” if the top manager was American and “zero” if he/she was Japanese (there were no third country nationals), 3) we asked respondents how autonomous their R&D activities were from the parent company in terms of fund allocation, objectives and methods, on a 5-point scale (1= completely autonomous, 5=fully controlled), 4) we asked respondents what type of human resource management system was used in the subsidiary, and 5) we also asked respondents what type of R&D management system was used in the subsidiary. The latter two measures used a 5-point scale (1=100% Japanese style, 5=100% U.S. style). After normalizing each variable, we took an average. The Cronbach alpha for this construct was 0.734.

**Incentive** was measured by asking respondents about whether R&D personnel are evaluated on: 1) specific R&D performance such as patents or papers, 2) achieved goals 3) leadership, 4) attitude or commitment, 5) teamwork, and 6) potential capabilities (1=not at all; 5=significantly). We also asked respondents a five-point scale (1=not at all; 5=highly direct) the extent to which an R&D employee’s evaluation results are linked with promotion and pay raises. We conducted a factor
analysis (varimax-rotation) on these seven variables. This analysis resulted in one factor for which the eigen value exceeded one. This factor consisted of attitude or commitment, leadership, teamwork and potential capabilities. The Cronbach alpha for this construct was 0.690.

Control Variables

Co-location was measured by a dummy variable: coded “1” if a R&D subsidiary is located within a manufacturing facility, coded “0” if a R&D subsidiary is located as an independent facility. M&A, was coded “1” if subsidiaries were established by M&As, and coded were coded as “0” if subsidiaries are by greenfield development. Experience was measured by the number of years an R&D subsidiary had been established in the HQ Size was measured by the total number of employees. Industry was coded “0” when firms are in pharmaceuticals, chemicals, plastics, ceramics, and metal products, and coded “1” when firms are in electrical appliances, automobiles and parts, transport equipment, and precision machines.

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