Enhancing Supplier Development: an Efficiency Perspective

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20 1. Introduction

Supplier development, a buyer-initiated effort, has been shown to improve supplier performance effectively (Humphreys et al., 2004; Krause, 1997; Krause et al., 2007; Li et al., 2012; Modi and Mabert, 2007). However, the efficiency of supplier development tends to vary. For example, a survey finds that the reported reduction in order fulfillment cycle time due to supplier development could vary from 30% to 80%.¹ In addition, the reported reduction of product defects varies even more significantly from 5% to 90%. This evidence, though anecdotal, suggests that some buyers are more efficient at utilizing their supplier development practices to

¹ https://scm.ncsu.edu/scm-articles/article/supplier-development-strategies-and-outcomes

develop their suppliers than others. That is, some buyers are better at enhancing their supplier 1 2 development efficiency, which we define as the extent to which a buying firm (i.e., buyer)² 3 utilizes its supplier development practices (the input) to increase its supplier's performance (the 4 output). This definition is consistent with the broader concept of efficiency, which is generally defined as the extent to which inputs are transformed into output (Green and Mayes, 1991). 5 Given that buyers may differ in their efficiency of transforming their supplier development into 6 (improved) supplier performance, to identify interfirm differences in their supplier development 7 efficiency, we follow the same stochastic frontier approach adopted by Lieberman and Dhawan 8 (2005) in which they assess the impact of resources on firm performance using the Battese and 9 Coelli (1995) model (cf. Lieberman and Dhawan, 2005). While details about this stochastic 10 frontier approach are discussed later, measuring efficiency with the stochastic frontier approach 11 12 is superior to the traditional output-input ratio approach. Stochastic frontier explicitly considers firms' heterogeneity across production factors (Chen et al., 2015) and is capable of estimating 13 14 both efficiency frontier and a firm's efficiency using cross-sectional data (Lieberman and Dhawan, 2005). 15

16 We deliberately differentiate between efficiency (doing things right) and effectiveness 17 (doing the right things) in this study. Given that existing studies have documented abundant 18 evidence pertaining to the effectiveness of supplier development practices on improving supplier performance, the premise of this study is that supplier development is effective; that is, supplier 19 20 development is a right thing to do. Yet, existing studies have remained silent on how to do the things (i.e., supplier development) right, and factors that could potentially enhance a buyer's 21 22 efficiency of utilizing supplier development practices are unexplored—a gap this study aims to 23 fill. More specifically, the findings of this study shed lights on "how well the resources 24 expended are utilized" (Fugate et al., 2010, p. 45) such that buying firms not only do the right 25 thing (by investing in supplier development programs) but also do things right. Consistent with the literature, supplier development represents a specific asset a buyer 26

- invested in their suppliers (Krause 1997) that the ending of a given relationship will render little
 value of such an asset. Supplier development, as such, exposes a buyer to transaction costs (i.e.
 transaction risk and coordination costs) from a Transaction Cost Economics (TCE) perspective.
 - ² In this study, we use the term "buying firm" and "buyer" interchangeably. Similarly, the term "supplier" in this study refers to a supplying firm.

To enhance supplier development efficiency, buyers can reduce the transaction risk and 1 coordination costs associated with their supplier development. This study investigates three such 2 3 factors: relational norms, supplier information sharing, and buyer information sharing. There are 4 two considerations that we focus on relational norms, defined as a relationship between exchange partners based on mutual trust, concerns and shared values (Heide and John, 1992). First and 5 theoretically, relational norms is built with the intent of reducing transaction costs stemming 6 from asset specificity, i.e., supplier development (buyer's relationship-specific investment) in 7 this study. Second and extending from the first consideration, whether the reduced transaction 8 9 costs associated with relational norms offsets the increased transaction costs associated with asset specificity is, nevertheless, an empirical question (Artz and Brush, 2000). To the best of 10 our knowledge, the dialectical opposition between relational norms and asset specificity (i.e., 11 12 supplier development in this study) has not been explored. Thus, by examining the effect of relational norms in a supplier development context, findings of this study can provide managers 13 14 with guidance about prioritizing resources and make theoretical contributions to better understand transaction costs empirically, one of TCE's key concepts. 15

16 With respect to information sharing, TCE has been documented as one of the most prevalent theoretical lenses (Kembro and Näslund, 2014). Whereas buyer information sharing 17 can reduce transaction costs by reducing coordination costs associated with conflict resolution 18 (Li et al., 2012) and/or misunderstanding resulting from divergent viewpoints (Forker et al., 19 20 1999), supplier information sharing can reduce transaction costs by reducing supply uncertainty (Li and Lin, 2006) and by reducing a buyer's coordination costs (Dyer, 1997). In short, TCE 21 22 informs our choice of factors. In addition, our focus of these three factors, particularly relational 23 norms, further our understanding of TCE and its applicability in a supplier development context. 24 Further, existing supplier development literature often examines information sharing as a 25 single factor, specifically, information shared by the buying firm (Humphreys, Li, and Chan, 2004; Krause, Handfield, and Tyler, 2007). This study extends the existing literature by 26 considering information sharing from both the buyer's and supplier's perspectives. In the 27 28 existing literature, the role of a buyer is to initiate supplier development and a supplier responds 29 to such an initiative (Krause et al., 2007). Specifying information sharing into buyer versus supplier better captures the subtle yet crucial differences between a strategic (proactive) act and a 30 31 collaborative (reactive) response. Second, differentiating between buyer and supplier improves

the understanding of information sharing both conceptually and empirically. Conceptually, a 1 2 closer look at the existing supplier development literature indicates the concept of information 3 sharing is, in fact, buyer information sharing. For example, "effective communication" is described as "open and frequent communication between buying firm personnel and their 4 suppliers" (Humphreys et al. 2004, p.134), or "buying firm respondents were asked to specify the 5 extent of their willingness to share information with the supplier" (Krause et al. 2007, p.536). 6 Empirically, differentiating information sharing into two separate concepts can provide both the 7 buyer and the supplier with more specific guidance in a situation where the same act from two 8 sides may have different effects. In sum, differentiating information sharing between buyer and 9 supplier highlights another major difference between this study and the existing supplier 10 development literature. 11

12 This study utilizes the Stochastic Frontier Analysis (SFA) (Battese and Coelli, 1995). SFA allows researchers to relax the assumption that firms are technically efficient (always 13 14 produce the maximum amount of output with a given set of inputs) in estimating a production function, an assumption that is necessary using OLS regression (Greene, 2008). In other words, 15 16 SFA allows us to model the actual (in)efficiency, as opposed to assuming firms are fully efficient in their use of inputs—an assumption that is rarely warranted in reality. Since we assume that a 17 18 firm's efficiency regarding producing output (e.g., supplier performance) using supplier development practices as inputs vary and influenced by other factors, SFA is more appropriate 19 20 than traditional regression approach. Further, SFA allows us to infer an objective measure of supplier development efficiency (deviation from the efficiency frontier) rather than directly 21 22 assessing efficiency in a subjective manner (Chen et al., 2015).

Figure 1 shows the theoretical model examined in this study. Using a sample of 261 manufacturing plants from eleven countries, this study finds that relational norms and supplier information sharing each improve supplier development efficiency. Contrasting to expectation, we find that buyer information sharing could degrade supplier development efficiency. Robustness analyses using an alternative two-stage approach and Data Envelopment Analysis (DEA), an alternative frontier methodology, are consistent with the main results. Discussions and future directions conclude this study.

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--Figure 1 here---

1 2. Literature Review

2 2.1 Practices related to supplier development

3 The term supplier development was first used as early as 1966 (i.e., Leenders 1966) to refer to 4 manufacturers' efforts to increase viable suppliers and improve the subsequent supplier 5 performance. Such an effort was triggered by the "Buy Canadian" policy in the early 1960s, 6 during a time when many Canadian suppliers had poor quality. Poor supplier quality 7 compounded with the weak Canadian dollar (against US dollar) at that time render "...the 8 buyer's responsibility...not to select, but to create a satisfactory source" and as such, a 9 "purchaser does not select supplier development as an appropriate technique or tool; it is the only 10 course left, apart from in-plant manufacture" (Leenders, 1966, p. 54). Since then, supply chain management researchers have discussed supplier development process (Hahn et al., 1990; 11 Hartley and Choi, 1996; Krause et al., 1998), practices (Krause et al., 1998; Reed and Walsh, 12 2002; Wagner, 2006), factors preceding buyers investment in supplier development programs 13 (Krause, 1999), and results of supplier development programs and from which identifying 14 successful factors (Krause and Ellram, 1997a, 1997b) and barriers (Lascelles and Dale, 1989). 15 16 Strategic oriented supplier development involves those practices in which a buying firm takes an active role and dedicates physical and human resources directly to a specific supplier with the 17 goal of increasing supplier performance to better meet the buying firm's needs. Exemplary 18 19 practices include but not limited to providing formal evaluation, supplier site visits, feedback, and supplier training (Krause, 1997; Krause et al., 2000). 20

Table 1 summarizes exemplary work on supplier development. We highlight some 21 22 observations from Table 1. First, previous studies largely focus on the effectiveness of supplier 23 development by examining relationships between supplier development practices and business performance (Wagner, 2011), supplier performance (Modi and Mabert, 2007), and buyer 24 25 performance (Krause et al., 2000). Very few studies discuss how to enhance the extent of efficiency with which a firm utilizes those practices to achieve ideal supplier performance. 26 27 Second, in terms of methodology evolutions, early works are conceptual in nature and much of them built on case studies; empirical studies became more dominant as time progresses, and 28 29 survey appears to be the most commonly adopted approach for data collection. Besides the 30 content novelty (efficiency in this study vs. effectiveness in the extant literature), our stochastic frontier approach adds another novelty to the stream of supplier development literature. The 31

1 third observation pertains to the studied contexts and subsequent generalizability. As

2 aforementioned, the term supplier development was first coined by Leenders (1966) for studying

3 Canadian manufacturers. With this genesis, the first wave of studies are North American

4 dominant (particularly auto industry related) with sporadic attention from Asia (e.g., Humphreys

5 et al., 2004) and finally, there was a systematic interest from Europe, notably by Wagner (2011,

6 2010, 2006) who collected data from Germany, Switzerland, and Australia and published a series

7 of studies on this subject of matters. The data used in this study encompass responses from 11

8 countries across three continents, so the findings of this study can improve generalizability.

9

INSERT TABLE 1 about Here

10 2.2 A Transaction Cost Economics view of supplier development

This study uses Transaction Cost Economics (TCE), one of the most commonly evoked theories 11 12 in the existing literature (e.g., De Toni and Nassimbeni, 2000; Humphreys et al., 2004; Krause, 1999; Krause et al., 2000), as the theoretical lens. TCE is appropriate for this study concerns a 13 14 strategic oriented supplier development program (Krause et al. 1998). Two key assumptions characterize TCE: bounded rationality and opportunism. Both are relevant to the context of 15 16 supplier development (Rindfleisch and Heide, 1997). Bounded rationality refers to both the neurophysiological and language limits of individuals (Simon, 1955). Decision makers, while 17 18 attempting to act rationally under conditions of uncertainty, are constrained by bounded rationality. As such, bounded rationality gives rise to transaction costs, defined as the sum of 19 20 coordination costs and transaction risk, that need to be minimized (Williamson, 1991). Opportunism refers to behaviors in which individuals act in their own self-interest (Williamson, 21 1985). Similar to bounded rationality, the existence of opportunism gives rise to transaction 22 costs (Williamson, 1985, 1981). Since a supplier development program is often tailored for a 23 24 given supplier, from a buyer's perspective, supplier development practices are often relationship-25 specific (i.e., high asset specificity that has low transferability to other relationships). Therefore, supplier development practices can be seen as relationship-specific assets with high asset 26 specificity (Krause, 1997; Modi and Mabert, 2007). These relationship-specific assets inevitably 27 expose buyers to transaction risk and coordination costs. For instance, the supplier might take 28 29 advantage of the buyer's sunken investment that increases transaction risk (e.g. supplier opportunism, supply uncertainty) (Grover and Malhotra, 2003). Developing a supplier also 30 31 necessitates high coordination costs to the buyer, ranging from the ex-ante supplier selection

(search cost) and contracting efforts (contracting cost) to the ex-post efforts in verifying supplier
compliance and evaluating performance (monitoring cost and enforcement cost if sanctions are
levied) (Grover and Malhotra, 2003). In summary, the TCE perspective prescribes that the
buyers would incur coordination costs and transaction risk with the implementation of supplier
development practices due to opportunism and bounded rationality. Therefore, a buyer can
increase its supplier development efficiency by reducing the transaction costs accompanying
with supplier development practices.

8 3. Hypothesis Development

Per TCE, the high asset specificity nature of supplier development practices entails 9 10 coordination costs and transaction risk to a buyer, which affect the utilization of resources. Building relational norms could help curb supplier opportunism and lower transaction costs 11 (Frazier et al., 1988; Heide and John, 1988), and hence, increase supplier development efficiency. 12 When high relational norms exists, a buyer could spend less time and resources on monitoring 13 efforts and thus reduces buyer's coordination costs (Dyer and Chu, 2003; Nyaga et al., 2010). In 14 addition, high relational norms serve as a better safeguard if a given relationship continues, 15 16 reducing the transaction risk. In a sense, a high degree of relational norms allows buyers to focus more on realizing the full benefits of the supplier development practices without worrying 17 about suppliers taking advantage of the buyer's investment. This is in line with the extant 18 19 supplier development literature that the buying firms' perspective toward suppliers has been identified as one of the antecedents to supplier development programs (Krause, 1999), and that 20 21 transaction-specific supplier development is influenced by trust and long-term commitment (Li 22 et al., 2012). In contrast, buyers in a relationship characterized by a low degree of relational 23 norms may not realize the full potential of supplier development practices because their concern of supplier opportunism would entail more ex-post efforts and incur higher monitoring and 24 25 enforcement costs, negatively impacting supplier development efficiency. Further, suppliers in a relationship characterized by a high degree of relational norms may be more willing to provide 26 27 their expertise and work with buyers on realizing the benefits of supplier development practices. Conversely, when lack of trust with a low level of relational norms, suppliers may refrain 28 29 themselves from participating in buyers-initiated supplier development activities, suspecting 30 buyers may consistently act in their own interest in disguise (Nagati and Rebolledo, 2013). Therefore, a relationship with high relational norms helps buyers better realize benefits from 31

1 supplier development practices and helps suppliers to fully participate in those practices and

2 ultimately, improve their supplier performance. In sum, we posit that high relational norms

3 reduces transaction costs associated with supplier development practices and improves a buyer's

4 supplier development efficiency.

5 Hypothesis 1 (H1): Relational norms is positively associated with supplier development
6 efficiency.

7 Information sharing between a buyer and a supplier is another key approach to encouraging conflict resolution (Li et al., 2012; Spekman, 1988) and has been identified as a 8 critical element in successful supplier development (Krause and Ellram, 1997a). Buyer 9 information sharing is the extent to which a buyer openly shares information with a supplier 10 (Paulraj et al., 2008). Buyers who share information with suppliers could help their suppliers 11 12 better realize the benefits of supplier development practices, improving the supplier development efficiency. To start with, open communication pertaining to the scope and goal of supplier 13 14 development helps reduce contracting efforts, thereby reducing coordination costs. For example, buyers who share product quality information may help their suppliers better utilize the quality 15 16 control training offered by the buyer. Similarly, buyers who share demand information may help the supplier better utilize the forecasting training provided by the buyers. Also, divergent 17 18 viewpoints may lead to different understandings about supplier development (Forker et al., 1999), which increases transaction risk. Buyers who share information with suppliers could reduce the 19 20 likelihood of misunderstanding, assuring the suppliers that buyers are interested in the success of the supplier development. Thus, we propose the following: 21

Hypothesis 2 (H2): Buyer information sharing with suppliers is positively associated with
 supplier development efficiency.

24 Inter-firm communication is a two-way street. Supplier information sharing, a mirrored concept of buyer information sharing, is the extent to which a supplier openly shares information 25 with a buying firm from which the supplier receives assistance. From the TCE perspective, 26 supplier information sharing could increase supplier development efficiency by reducing 27 28 transaction risk and the associated coordination costs. A supplier sharing information with its 29 buyers could reduce supply uncertainty (Li and Lin, 2006; Yu et al., 2013). Supply uncertainty can preclude a buyer's ability to verify and ensure supplier compliance and gives rise to 30 31 transaction risk and increases coordination costs. Suppliers sharing information such as updates

on delivery schedules, production cost, and product quality reduce supply uncertainties for 1 2 buyers (Grover and Malhotra, 2003; Lin et al., 2002). Supplier sharing those pieces of 3 information may also indicate supplier commitment, which has been found as one of the critical antecedents to the success of supplier development programs (Krause, 1999). Supplier 4 information sharing could also reduce a buyer's coordination costs (Dyer, 1997). For example, 5 6 suppliers sharing production and delivery schedule information could reduce a buyer's costs in coordinating internal activities. Past study also shows that sharing product quality information 7 could reduce a buyer's costs related to ensuring supplier compliance (Dyer, 1997). Thus, we 8 propose the following: 9

Hypothesis 3 (H3): Supplier information sharing with buyers is positively associated with
supplier development efficiency.

Lastly, we theorize that relational norms and information sharing by buyer and by supplier could jointly influence supplier development efficiency by reducing transaction costs. For buyers, supplier development practices subject a buyer to supplier opportunism, increasing transaction costs for buyers. A high level of relational norms reduces transaction costs and encourages frequent and open information sharing with their suppliers (Heide and John, 1992), which further reduces transaction risk and coordination costs.

Similarly, from a supplier's viewpoint, suppliers sharing confidential information
exposes them to buyers' opportunism on one hand; on the other hand, suppliers' willingness to
share information signals their trust in buyers and characterizes a high level of relational norms,
which further curbs opportunism and reduces transaction costs (Nagati and Rebolledo, 2013).

Specific to the context, we posit that information sharing and relational norms together 22 promote a better environment for joint problem-solving (McEvily and Marcus, 2005; Modi and 23 24 Mabert, 2007; Watts and Hahn, 1993), which increases supplier development efficiency. 25 Relational norms promotes involving not just purchasing personnel but also engineers in the supplier development process (Reed and Walsh, 2002). Further, relational norms encourages 26 behaviors toward collective goals and mutual benefits. Information sharing increases both buyer 27 and supplier's awareness of potential supplier development problems and opportunities. Joint 28 29 problem-solving becomes more efficient with the exchange of sensitive and/or proprietary information and with the exchanged partners both willing to listen and accept suggestions. Taken 30

1 together, relational norms and information sharing jointly create an environment that facilitates

2 joint problem-solving, which helps increase supplier development efficiency.

3 Hypothesis 4a (H4a): There exists a synergistic effect between relational norms and buyer

4 information sharing on supplier development efficiency.

5 Hypothesis 4b (H4b): There exists a synergistic effect between relational norms and supplier

- 6 information sharing on supplier development efficiency.
- 7

8 **4. Data and Sample**

9 This study uses part of the data collected by the fourth round of the High-Performance

10 Manufacturing (HPM) research project (Schroeder and Flynn, 2001). HPM is a large-scale

11 global research project that involves a team of international researchers. HPM examines

12 manufacturing practices at the plant level in three specific industries: electronics, machinery, and

13 transportation. Many studies have been published based on previous rounds of HPM project.

14 The fourth round HPM data involves manufacturing plants in eleven countries (China, Korea,

15 Japan, Taiwan, Brazil, Germany, Sweden, Finland, Israel, Italy, and Spain). Participation in

16 HPM requires extensive efforts on the part of the plant. To increase a firm's willingness to

17 participate, research team members spent a significant amount of time with prospective firms to

18 explain the benefits of participation in the study. Researchers rely on their own personal or on

19 university networks to contact prospective firms. Once a firm agrees to participate, the firm

20 manager identifies a high-performing plant with at least 100 employees for the survey. To

21 reduce the potential impact of unobserved firm-level variations, each participating firm is limited

to one plant. The plant coordinator receives a package that consists of 12 survey questionnaires,

which cover different manufacturing and environmental aspects (e.g. supply chain, operations,

human resources, competitive environment, etc.) for distribution to the appropriate respondents.

To increase the accuracy of information, the survey requires the respondents' expertise or job titles to be directly related to the survey questions. In this study, two upstream supply chain

27 managers (e.g. purchasing manager) respond to the survey questions. Measurement items are

mixed in the questionnaires to reduce the context effects (Tourangeau et al., 2003). Using

29 multiple respondents and mixed survey items help reduce the common method bias (Podsakoff et

al., 2003). The final data set consists of 261 plants. Consistent with the previous round of HPM

project (e.g. Zhang et al., 2012), the response rate varies across countries and is approximately

60 percent as indicated in a recent study (Turkulainen et al., 2017). Table 2 provides the sample
 distribution across countries and industries.

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---Insert Table 2 about here------Insert Figure 2 about here---

6 **5. Methodology**

7 SFA estimates an individual unit's inefficiency as the distance to the efficient frontier (Battese 8 and Coelli, 1995; Chen et al., 2015; Lieberman and Dhawan, 2005). Specifically, SFA constructs an efficient frontier, which is the ideal output level achieved given a specific set of 9 inputs among a group of units (see Figure 2). The inefficiency term infers a unit's efficiency of 10 transforming "the inputs available to it ... and converts them into whatever output it desires..." 11 (Dutta et al., 2005, p. 278). Therefore, the closer a unit to the frontier, the better the unit's 12 relative efficient use of resources. In this study, we operationalize supplier development 13 (in)efficiency using the inefficiency term described in the following paragraph. 14 **5.1 Stochastic frontier analysis** 15

SFA decomposes the observed output into three elements: the ideal output (i.e. the desired outcome) determined by a set of inputs (i.e. resources), the random error term, and the
inefficiency term (Aigner et al., 1977; Meeusen and van Den Broeck, 1977). The stochastic
frontier model has the following formation:

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Y_i = f(\mathbf{X}_i; \boldsymbol{\beta}) + v_i - u_i \quad [1]
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where Y_i is the actual output; $f(X_i; \beta)$ is the production function, which denotes ideal output as a function of a set of inputs X_i with unknown coefficients β . The production function in [1] represents the idealized efficient frontier—the maximum expected output given inputs X, common to all sample organizations. Variable V_i captures the random errors affecting outputs due to unobserved inputs or measurement errors in data. Finally, u_i depicts the relative distance to the efficient frontier, that is, the inefficiency of an individual unit i regarding transforming inputs to an output.

Figure 2 illustrates the notion of the efficient frontier with one input and one output, where the value of the input X is shown on the horizontal axis and output Y is shown on the vertical axis. The deterministic efficient frontier reflects the existence of diminishing returns to scale. As illustrated, firm A uses input X_A to produce the observed output Y_A . If firm A is fully efficient in utilizing resources (i.e., there is zero inefficiency: u = 0), then firm A should achieve the ideal output $Y_A^* = f(x) + v$. When the random noise (v) is positive, the ideal output lies above the deterministic efficient frontier.

6 The main contrast of SFA and traditional least square regression approach is that the 7 traditional approach attributes the deviation from the efficient frontier to random error V_i only, while SFA recognizes firm-specific inefficiency \mathbf{u}_i as a potential cause of the deviation. SFA 8 9 also allows exogenous variables to affect the extent of inefficiency u_i . The three most common assumptions of the inefficiency term in SFA are exponential, half-normal, and truncated-normal 10 (Aigner et al., 1977; Stevenson, 1980). The choice of distribution assumption is typically 11 computational rather than theoretical (Coelli et al., 2005) for SFA. Since the truncated-normal 12 13 assumption is not able to converge on our data, we opt for a more parsimonious half-normal model. We consider the half-normal assumption by Caudill and Ford (1993), Caudill et al. 14 (1995), and Hadri (1999), which allows heteroscedasticity in the distribution of \mathbf{u}_{i} to analyze the 15 effects of exogenous factors on inefficiency u_i . Equations [2] and [3] specify the half-normal 16 inefficiency model. 17

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$\mathbf{u} \sim \mathbf{N}(0, \sigma_{\mathbf{u}}^2) $	[2]
$\sigma_{ui}^2 = \exp(\delta Z_1)$	[3]

In Equation [3], a positive value of the δ coefficient indicates that as the level of exogenous variable Z increases, so does the variance of technical inefficiency (σ_u). A negative δ indicates that the exogenous variable reduces the variance of technical inefficiency. Lastly, we adopt a one-step approach and estimate the production function and the inefficiency model simultaneously following recommendations by Chen et al. (2015).

25 **5.2 Measurements**

26 **5.2.1 Inputs and output of the production function**

27 Prior studies have used managerial practices as inputs because they represent resources utilized

in generating outcomes (e.g. Narasimhan et al., 2001). This study views supplier development

29 practices as the inputs X and supplier performance as the output Y of the production function in

[1] because the focus is on the buyer's supplier development efficiency. Several supplier 1 2 development related practices have been discussed in the literature. The construct of supplier 3 development initiative captures the practices that require a buying firm's active involvement, 4 such as providing technical assistance, training, and joint meetings (Krause, 1997; Narasimhan et al., 2009; Swink et al., 2005). The construct of supplier evaluation measures the extent of a 5 6 buying firm having a formal supplier evaluation system, which is an important practice in supplier development (Krause et al., 1998). The construct of supplier performance captures 7 multiple supplier performance dimensions such as conformance, cost, on-time delivery, and 8 willingness to meet a buyer's requirements (Johnston et al., 2004). Lastly, we include both 9 country and industry dummies in the production function to control for the potential effects on 10 supplier performance. Please see Appendix 1 for detailed survey items. 11

12 **5.2.2** Hypothesized factors in the inefficiency model

The measurements of hypothesized factors **Z** in the inefficiency model [3] are also adapted from 13 14 the existing supply chain literature. The relational norms construct, adapted from Min, Mentzer, and Ladd (2007), assesses the extent of supplier relationship in terms of supplier benevolence 15 (Kumar et al., 1995; Min et al., 2007) and shared understanding between buyer and supplier. We 16 conceptualize the extent of information sharing as a firm's willingness to share all kinds of 17 18 information. To reflect this conceptualization, we model both buyer' and supplier' information sharing as reflective scales-high information sharing indicates high extent of willingness to 19 20 share all relevant information, as opposed to formative scales—high information sharing can be achieved by high extent of willingness to share selective information. Specifically, the buyer 21 22 information sharing construct measures several types of information (e.g. production, delivery, and scheduling etc.) shared by buyers with their suppliers (Zhou and Benton, 2007). We asked 23 24 the purchasing managers the information they shared with their major suppliers. Similarly, the supplier information sharing construct measures the information shared by suppliers with their 25 buyers. We asked the purchasing managers whether they have access to the information from 26 their major suppliers. Rather than focusing on a firm's willingness to share information 27 (Monczka et al., 1998), these two constructs focus on the content of the information shared by 28 29 suppliers and buyers (Zhou and Benton, 2007).

Finally, plant size may influence the efficiency of utilizing supplier development
 practices. For example, a large plant may have more personnel and capital to invest in supplier

development activities (Blonska et al., 2013). Therefore, plant size is included as a control
 variable in the inefficiency function and measured as the natural logarithm of the number of
 employees in a plant (Dean and Snell, 1991).

4 5.3 Country differences across measurement items

Since the survey data come from multiple countries, the respondents across countries could 5 6 interpret the measurement items differently due to culture or language differences. To ensure measurement invariance, we use the recently develop alignment method for multiple-group 7 factor analysis (Asparouhov and Muthén, 2014), which is a procedure implemented in Mplus 7.1. 8 The alignment method is an optimization approach that identifies an optimal pattern of 9 measurement invariance solution and detects groups that deviate from the optimal pattern. The 10 alignment method is a two-step approach. In the first step, the alignment method fits a 11 12 configural invariance model across groups using maximum likelihood method with loadings and intercepts freed, factor means fixed at zero, and factor variances fixed at one. In the second step, 13 14 the method estimates the factor means and variances to minimize the total amount of measurement variances across all parameters by applying a simplicity function that works as the 15 16 rotation criteria for the exploratory factor analysis (Asparouhov and Muthén, 2014, pp. 496–498). 17 The estimation procedure is an iterative approach with different starting values until the 18 procedure reaches an optimal and stable solution. Finally, the alignment method reports the optimal solution (the best measurement invariance pattern) and provide information to assess the 19 20 degree of non-invariance across groups. The results of the alignment method indicate that the number of groups with approximate measurement invariance in factor loading is high. 21 22 Asparouhov and Muthén (2014) suggest the number of measurement non-invariance items should be lower than 20% of the total items for the factor means to be comparable across groups. 23 24 Our result is well below the 20% cutoff (only one item shows sign of measurement non-25 invariance: please see Appendix for details). As a result, we proceed with confirmatory factor 26 analysis.

27 5.4 Psychometric properties of constructs

28 This research uses confirmatory factor analysis (CFA) to assess the psychometric properties of

29 the survey constructs. We drop the items with standardized factor loadings below the normally

accepted level of 0.6 (Fornell and Larcker, 1981; Shah and Goldstein, 2006). Most of the

Average Variance Extracted (AVE) values are greater than 0.5, which indicates convergent

validity (Hair et al., 2006), except supplier development initiative (0.475). The AVE is slightly 1 2 below 0.5 due to the lower factor loading of one survey item (we offer the necessary training to 3 our suppliers). We decided to keep this item for a theoretical reason since this item represents an 4 activity of supplier development that requires buyers' active participation. Composite reliability coefficients ranged from 0.7 to 0.9, which exceed the recommended 0.7 benchmark for construct 5 6 reliability (Henseler et al., 2009). The square root of the AVE for each hypothesized factor is also greater than the correlations between the hypothesized factors (Gefen and Straub, 2005; Hair 7 et al., 2006). The overall model fit statistics are all above the recommended standards 8 (RMSEA=0.055 with p-close value=0.167, CFI=0.946, TLI=0.938, SRMR=0.052) (Hu and 9 Bentler, 1999, 1995). We also perform a bootstrapping procedure with 1000 samples to address 10 potential multivariate non-normality. The model fit statistics are qualitatively similar. Appendix 11 1 documents the measurement item loadings, composite reliability, and AVE of the constructs in 12 this study. 13

14 We utilize the factor scores of each construct for subsequent analysis rather than the averages following suggestions from past research (Edwards and Wirth, 2009). Factor scores are 15 16 better than averaging indicators because averaging requires stronger assumptions regarding 17 psychometric properties (averaging requires the indicators be parallel, whereas factor scores only 18 require indicators be congeneric) and factor scores often have more symmetrical distributions 19 (Calantone et al., 2017; Edwards and Wirth, 2009). Factor scores are even more beneficial when 20 researchers have data from multiple countries and concerns about measurement invariance (Flora et al., 2008). 21

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---Insert Table 3 about here------Insert Table 4 about here---

24 **6. Results**

We present the results with the half-normal distribution assumption for the inefficiency term. We also analyze the data with the exponential distribution assumption and the results are consistent. Table 4 reports the estimation results regarding the production function and inefficiency model using SFA. We use STATA 14 to perform the SFA analysis. The production function, which estimates the efficient frontier, includes the two supplier development practices (supplier development initiative and supplier evaluation) as inputs and supplier performance as output. We use the original factor scores for the production function without any transformation

instead of the commonly used Cobb-Douglas production function since transforming the scores 1 2 using natural logs changes the underlying equal spacing assumption of the Likert scales. 3 Country and industry dummies are included in the production function as control variables. The estimated coefficients of the production function change only slightly across all four models 4 (Table 4), with different specifications of the inefficiency model. This indicates a stable efficient 5 frontier. The coefficients of supplier development initiative are significant across all models, 6 suggesting a valuable resource for supplier performance. Conversely, supplier evaluation is not 7 all significant across all models. 8

Model 0 represents the baseline case. For the inefficiency model, Model 1-3 each include 9 one of the hypothesized factors; Model 4 includes all main effects and Model 5 includes the 10 interactions. The model diagnostics statistics (AIC/BIC) show improvements when comparing 11 Model 1-3 each to Model 4 but not Model 5. The primary focus of this study is the estimated 12 coefficients in the inefficiency model. As discussed before, a negative estimated coefficient in 13 14 the inefficiency model [3] indicates a decrease in the variance of firm-specific inefficiency u_i , which denotes a positive effect on supplier development efficiency. Plant size, a control variable, 15 16 is insignificant across all models. The estimated coefficient of relational norms is negative in Model 1 ($\delta = -2.284$, p<.01), suggesting that relational norms has a positive effect on supplier 17 development efficiency, but this effect is not robust to the inclusion of the information sharing 18 predictor variables (Table 4, Model 4, $\delta = 0.022$, n.s.), which provides only partial support for 19 H1. The coefficient of buyer information sharing is not significant in Model 2 ($\delta = -0.169$, n.s.). 20 Further, Model 4 shows the coefficient is positive, indicating a negative effect on supplier 21 development efficiency ($\delta = 2.767$, p<.001). The results do not support H2. The estimated 22 coefficients of supplier information sharing are negative and significant in both Model 3 (δ = -23 24 1.487, p<.01) and Model 4 (δ = -3.688, p<.05), suggesting that supplier information sharing has a positive effect on supplier development efficiency, which provides support for H3. Finally, 25 Model 5 shows the interaction effects between relational norms, buyer information sharing, and 26 27 supplier information sharing are not significant. The results do not support H4a, nor H4b. We discuss implications of the results further in the Discussion and Conclusion section. 28 29 ---Insert Table 5 about here---

30 6.2 Robustness analysis

We present results using an alternative two-stage approach advocated by certain researchers (Banker and Natarajan, 2008) as a robustness test. Researchers find that the coefficients remain statistically consistent even when the first-stage and second-stage variables are correlated (Johnson and Kuosmanen, 2012). In the first stage, we estimate the efficiency score using SFA but excluding the hypothesized factors in the inefficiency function. In the second stage, we regress the logarithm of the efficiency score on the hypothesized factors (see the results in Table 5). The overall results are consistent with the main results using a one-step approach.

Additionally, we use DEA as an alternative frontier approach to examine the robustness 8 of the study results (Banker et al., 1984). Prior studies often consider SFA and DEA as two 9 common alternative frontier methodologies (Chen et al., 2015). We consider input-oriented 10 11 DEA because the focus is on better resource utilization by changing the level or the mix of inputs. 12 Further, this study assumes variable return-to-scale (VRS) since increases in the extent of 13 managerial practices do not necessarily result in a proportional change in perceived supplier performance. We apply a two-stage DEA estimation approach. In the first stage, we obtain each 14 plant's supplier development efficiency score using DEA with supplier development initiative 15 and supplier evaluation, both adjusted by industry mean, as two inputs, and supplier performance 16 17 as the output. In the second stage, we follow the bootstrap procedure described in Simar and 18 Wilson (2007), a widely used procedure in the field of economics, to overcome the finite sample bias of the naïve two-stage approach (regression analysis using the DEA efficiency score as 19 dependent variable). This procedure estimates standard errors and confidence intervals by 20 independently drawing pseudo errors from the truncated normal distribution using a parametric 21 22 bootstrap procedure (Simar and Wilson, 2007). STATA module simarwilson is used for estimation (Badunenko and Tauchmann, 2018). Table 6 shows the results of the Simar and 23 Wilson (2007) two-stage estimation approach. The overall results are consistent with prior 24 25 findings using SFA. Relational norms ($\beta = 0.256$, p<.001) and supplier information sharing ($\beta =$ 0.298, p<.001) are positively associated with supplier development efficiency; buyer information 26 sharing has a negative effect ($\beta = -0.215$, p<.001) (Table 6: Model 2), which are consistent with 27 28 the findings from the SFA method. In addition, the interactions are not significant, which are 29 also consistent with the SFA results.

30

---Insert Table 6 about here---

31 **7. Discussion and Conclusion**

1 7.1 Theoretical contributions

Existing empirical studies often focus on either identifying best practices or on explaining
variations in performance in terms of best practices. Few studies focus on the efficiency of
utilizing best practices. We demonstrate that relational norms and supplier information sharing
can increase a buyer's supplier development efficiency, which prior studies have not yet
considered. We also provide a theoretical foundation indicating the reduction in transaction
costs is the underlying mechanism for changes in supplier development efficiency.

The finding that information sharing by suppliers is more influential as to supplier 8 development efficiency than information sharing by buyers merits further discussion. Though in 9 theory, information sharing could potentially reduce coordination costs for efficiency gains, 10 existing literature has mixed findings with different attributions and thus, calls for future 11 investigations (Blonska et al., 2013; Carr and Kaynak, 2007; Hult et al., 2004; Krause et al., 12 2007). Our study answers this call by separating information sharing into sharing on both sides 13 14 and finds that in practice, supplier information sharing sends a stronger signal to the buyer regarding the supplier's commitment to the supplier development program. Because supplier 15 16 development represents a buyer's specific investment in a supplier, the reciprocity of a buyer's goodwill from the supplier is more meaningful than efforts originated by the buyer. Supplier 17 18 information sharing, in this regard, signals a supplier's willingness to reciprocate a buyer's investment, which further reduces the buyer's transaction risk. In contrast, buyer information 19 20 sharing may degrade supplier development efficiency when the cost of sharing outweighs its benefit in situations where suppliers experience information overload (Meier, 1963). A supplier 21 22 may also choose to neglect the information provided by buyers to avoid information overload, particularly when the information shared by buyers is self-contradictory. Finally, we did not find 23 24 consistent interaction effect between relational norms and information sharing on efficiency gains in both SFA and DEA models, which required further research to investigate this issue. 25 Overall, by differentiating information shared by a supplier versus a buyer, this research 26 casts doubts on the conventional wisdom that information sharing is always helpful (Dyer, 1996; 27 Takeishi, 2001). With much of the existing SD literature either taking a buyer or supplier 28 29 viewpoint, there is a call for dyadic studies to better our understanding of SD (Krause, 1999). This study is one step closer to that call: by differentiating supplier information sharing from 30 buyer information sharing, we consider both views in the same study and find that supplier 31

information sharing is more beneficial to supplier development efficiency than information
sharing on a buyer side. Besides, compared to the existing empirical studies that based on
samples from one country or a region (e.g., studies by Wagner), our sample came from 11

4 countries across three continents, improving the generalizability of the study findings.

5 **7.2 Managerial implications**

6 Practitioners can benefit from this study as well. The supplier development initiative is subsumed under the broader supply management program with an ultimate goal of improving a 7 buyer's competitive advantages via improved supplier performance. Due to limited human and 8 financial capitals, buyers often struggle with allocating internal resources to supplier 9 development in a hope of improving supplier performance. The study results show that supplier 10 development initiative is a more valuable resource than supplier evaluation for supplier 11 12 performance. Therefore, we suggest that training and continuous improvement activities with suppliers should be prioritized for managerial attention and resource allocation than building a 13 14 supplier evaluation system. Another thing that purchasing professionals should be aware of is that contrast to the common belief, buyer information is not a panacea for supplier development 15 16 efficiency. While buyer information sharing may reduce the bullwhip effect in a supplier, buyers 17 should not expect that their sharing of information with the supplier will increase in supplier 18 development efficiency. Rather, buyer information sharing may reduce supplier development efficiency due to information overload experienced by the supplier. In short, buyers should be 19 20 aware of the potential downside of sharing information. In contrast, supplier information sharing can increase the efficiency of buyer-initiated supplier development programs. Therefore, buyers 21 22 should focus on encouraging and designing incentives to motive suppliers to share information with buyers. 23

24 The other implication practitioners can take away regards building relational norms with 25 suppliers and supplier information sharing, the two key factors that each enhance supplier development efficiency. Though there lacks a complementary effect between these two, we can 26 still draw inference about the dynamism between the two. First, managers should understand 27 28 that relational norms acts as the lubricant to reduce the wastes (i.e., opportunism, needs of 29 monitoring or coordination) generated from the transformation process that transforms supplier development initiative and supplier evaluation system to supplier performance. Specifically, 30 31 increased trust and commitment leads to reduced business partners' desire for secrecy and

seeking self-interest in disguise, motivating information sharing (Modi and Mabert, 2007).
Moreover, the effect of relational norms alone has a greater magnitude in improving supplier
development efficiency than supplier information sharing. Taken together, the results of this
study suggest that to improve efficiency when facing limited resources, purchasing and supply
chain managers should first invest in building relational norms with their business partners; once
a high level of relational norms is established, suppliers would naturally and more likely to
discard their guards and share information freely, enhancing efficiency even further.

8 Overall, we suggest that relational norms and supplier information sharing help explain 9 why some buyers are more efficient at utilizing their supplier development practices to develop 10 their suppliers than others. In sum, if supplier development practices are the hardware for 11 developing suppliers, relational norms and supplier information sharing act as the software that 12 increases the hardware efficiency.

13 7.3 Limitations and future research

14 The HPM project is comprehensive and requires a significant amount of commitment from both researchers and industry participants; however, it is not without limitation. The participating 15 16 firms mainly were mainly recruited through team researchers' personal networks. The measurement scales used in this study are part of the larger HPM project. They were not 17 18 specifically designed for this study, and therefore, are confined to the existing survey items. In addition, the research team relied on firm managers' judgments to identify high-performing 19 20 plants to be included in the HPM project. As the data focus on high-performing plants, the sample is not random. Nonetheless, including only high-performing plants is particularly 21 22 suitable for this research since the purpose of SFA is to estimate an efficient frontier, and the focus of this study is on proactive and strategic-oriented supplier development, as opposed to 23 24 reactive-oriented supplier development. An efficient frontier constructed by a group of highperforming plants is more representative and informative than a frontier constructed by a group 25 of randomly selected plants. Another limitation is that the survey data only come from the buyer 26 side. The results that only supplier information sharing has a significant effect may also reflect 27 dyadic differences in buyer-supplier collaborations. Future research is encouraged to collect data 28 29 from the supplier side to verify the results of this study. Another interesting avenue of future research is to compare our findings with findings from a supplier-initiated context. For 30

1	programs that are initiated by the suppliers, we suspect that buyer information sharing could be
2	more influential because a supplier would expect reciprocity from the buyer.
3	In summary, this study establishes a framework and identify factors for supplier
4	development efficiency. Supplier development consumes significant buyer resources, but the
5	outcomes often vary. We view supplier development from an efficiency perspective and
6	examine factors that help firms use resources more efficiently to achieve better outcomes.
7	Supplier development literature is not short of studies regarding best practices and their effects
8	on performance. Nonetheless, very few studies examine the follow-up question: how to increase
9	the extent of efficiency of using such practices. We hope this research stimulates interest and
10	encourages future researchers to examine managerial practices from an efficiency perspective.
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Figure 1. Theoretical model



1 Figure 2. Illustration of the efficient frontier



4 Table 1: Supplier Development: An Exemplary Review of Literature

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2 3

Reference	Nature of the Study	Main Content
Leenders, 1966	Conceptual	-first use of the term "supplier development" (SD) as part of the supplier selection process,
—		particularly (but not limited to) selecting new source
O		-use SD to refer to (Canadian) manufacturers' efforts to increase viable supply source and
		improve subsequent supplier performance
		-defines a broader view of SD as "situation in which the purchaser can see unusual benefits
0		for the potential supplier (of which the supplier is unaware) if he can be persuaded to
S		undertake the contract" (p.54)
Hahn et al., 1990	Conceptual	-discuss both a "narrow" perspective of SD that aims to develop a supplier that has not
		delivered products to a focal firm before, and a "broad" perspective that pertains to
		developing an extant supplier
M		-present a two-dimensional matrix and categorize SD activities. One dimension is related
		areas (product-, process-, and operating systems related), and the other dimension is
		capabilities (technical, quality, delivery, and cost)
		-present a conceptual model that outlines the organizational decision process of a supplier
		development program
Watts and Hahn,	Empirical (Survey)	-survey to determine the level of involvement in SD and the results indicated that large
1993		firms are more likely to be involved.
		-assess the success of the SD programs and the results confirmed the importance of formal
1		supplier evaluation in the SD process.
		-assess the effectiveness of SD by rating a series of questions using 7-point Likert scales,
		with quality and delivery related activities receiving highest ratings.
Hartley and Choi,	Conceptual model	-describe supplier development process
1996	(interview with	-identify four factors that instrumental to sustain and spread improvement activities

	automakers)	throughout a supplier organization: 1) hands-on training of supplier team members, 2)
		regular follow-up and measurement by the customer, 3) fit between the improvement
Ţ		approach and supplier firm's culture, 4) support in the supplier's organization
Krause and	Empirical (survey)	-survey firms' experience about their supplier development programs and different firms
Ellram, 1997(a)		into "fallen short" and "exceeded" groups.
		-show that the two groups are significantly different in formal evaluation, providing
0		feedback of evaluation results, use of a supplier certification program, site visits to the
S		suppliers, visits to the buying firm by supplier's representatives, supplier recognition,
		training and education of the supplier's personnel, and investment in the supplier's
		operation by the buying firm.
Krause and	Empirical: survey	-majority of buyers involved in SD will perceive their suppliers as partners
Ellram, 1997 (b)		-buyers involved in SD emphasize on sharing information with suppliers, top management
		involving in buyer-supplier relationship, cross-functional teams, and purchase a larger
		percentage of the supplier's annual sales
Krause et al., 1998	1. Empirical: part of a	-differentiate between reactive vs strategic SD: the former aims to improve poor supplier
<u> </u>	larger empirical research	performance and eliminate deficiencies, whereas the latter aims for continuous
	effort, i.e., the Global	improvement of supply base with a focus on long-term competitive advantages
	Procurement and Supply	-describe a generic 10-step process model of SD (i.e., Section 4.2 and Figure 2)
	Chain Electronic	-differentiate between supplier assessment on performance and assessment system of
1	Benchmarking Network	capabilities
	initiative.	-present three propositions regarding circumstances/conditions that under which firms are
	2. Descriptive statistics	more likely to participate in strategic SD programs
Krause 1999	1. Empirical: survey	-apply transaction cost theory and classify antecedents into three groups: the environmental
	2. Structural equation	and influence factors (market competition, rate of tech change, top management support,

	modeling (SEM)	importance of inputs to the buying firm); the attitudinal factor (buyer's attitude toward
		suppliers), and the barrier factors (supplier commitment, inter-firm communication, buyer's
		expectation of relationship continuity)
		-the proposed model is mostly supported except that rate of tech change doesn't influence
		buyer's perspective toward suppliers and that buyer's expected relationship continuity
		doesn't influence SD activities
Krause et al., 2000	1. Empirical: survey from	-apply transaction cost theory, resource-based theory and internalization theory to
S	279 manufacturing firms	categorize SD strategies as internalized (conceptualized using the transaction cost theory)
	2. SEM	vs. externalized (conceptualized using the resource-based theory) activities
		-examine the impact of SD activities on supplier performance and show that while direct
		involvement activities (i.e., internalized SD strategy) play a critical role in performance
m		improvement, external activities (supplier incentives and supplier assessment) exert their
		impact via direct involvement
De Toni and	1. Empirical: survey	-document and show that which dimension of SD are important for successful JIT
Nassimbeni, 2000	2. Multivariate analysis	implementation; among the three, the "design link" is most closely connected to SD
	(canonical correlation,	-specifically, all three links ("quality link," "logistics link," and "design link") are
	discriminant analysis)	connected to the "supplier assistance and training" SD practice, and both "logistics and
		design links" are connected to the "organizational integration" SD practice, and only the
		"design link" is connected to the "contractual incentives" SD practice.
Reed and Walsh,	Case study (12 interviews)	-describe and differentiate SD as a practice that is either reactive (to deal with poor supplier
2002		performance) or strategic (to enhance the supply base's long-term capability). This
		differentiation is in line with Krause et al.(1998)
		-propose that buyers prioritize supplier capabilities on quality, cost, and delivery over
		technological capability in their SD programs

Humphreys et al.,	1. Empirical: survey	-transaction-specific SD predicts buyer-supplier performance improvement
2004	2. regression analysis	-infrastructure factors such as trust, supplier strategic objectives and effective
		communication play an important role of buyer-supplier performance
O		-direct SD is correlated to infrastructure factors of SD, indicating SD is not an isolated
		behavior of the buying firms
Wagner 2006	1. Empirical (case studies	-combine qualitative (case studies) with quantitative (survey across firms in three German-
0	and survey)	speaking countries: Germany, Switzerland, and Austria) approach
S	2. Exploratory factor	-differentiate SD into direct (internalized) and indirect (externalized): the buying firm plays
	analysis	an active role and dedicates human and/or capital resources to a specific supplier in the
		former, whereas the buying firm commits little or no resources to a specific supplier in the
		latter.
m		-direct SD activities involve human and capital investment, whereas indirect SD activities
		involve both ad hoc (occasional) and regular supplier evaluations, evaluation system and
		process, communication.
Wagner 2010	1. Empirical paper (survey	-differentiate SD activities into direct vs. indirect and investigate their respective
	sent to industrial and	relationships to supplier performance.
	service firms in three	-discuss various theoretical lens applied in the existing studies, including theories relevant
	European countries)	to the links between indirect SD activities and performance (goal-setting theory, the concept
	2. Regression analysis	of "influence strategy") and those relevant to the links between direct SD activities and
<u>+</u>		performance (TCE and knowledge-based view of the firm)
		-found that 1) indirect SD improves suppliers' performance and capabilities, 2) direct SD
		improves suppliers' capabilities only, and 3) firms should engage in either indirect or direct
7		SD but not in both simultaneously.
Wagner 2011	1. Empirical: survey across	-draw on social capital theory to show that the length of the buyer-supplier relationship can

	three German-speaking	better explain the links between direct SD activities and performance: relationship length is
	countries	used as a proxy for the life-cycle of a given relationship
—	2. regression analysis	-the effectiveness of a given SD activity is moderated by relationship life-cycle
Nagati and	1. Empirical: survey	-survey manufacturing sectors in Canada
Rebolledo, 2013	2. partial least square	-adopt a supplier viewpoint and show that supplier's trust, preferred customer status, and
	(variance-based SEM)	environmental dynamism all impact SD activities and the subsequent supplier performance
0		·,
S		
0		
		

Country	Electronics	Machinery	Transportation	Total
Brazil	5	6	9	20
China	10	10	10	30
Finland	6	6	5	17
Germany	6	13	9	28
Israel	21	5	0	26
Italy	7	17	5	29
Japan	6	7	8	21
Spain	6	8	11	25
Sweden	4	4	1	9
South Korea	8	5	13	26
Taiwan	19	9	2	30
Total	98	90	73	261

1 Table 2: Sample distribution across industry and country

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Table 3: Summary statistics and correlations

	Mean	S.D.	1	2	3	4	5	6
1. Supplier performance	3.663	0.513	1					
2. Plant Size	844.117	1994.74	0.084	1				
3. Supplier development initiative	3.786	0.595	0.575**	0.156*	1			
4. Supplier evaluation	3.859	0.905	0.438**	0.161*	0.591*	1		
5. Relational norms	3.867	0.540	0.563**	0.037	0.645**	0.453**	1	
6. Buyer information sharing	3.194	0.953	0.330**	0.038	0.284**	0.264**	0.143*	1
7. Supplier information sharing	3.143	0.899	0.483**	0.009	0.355**	0.382**	0.241**	0.663**

† p<0.1, * p<0.05, **p< 0.01, *** *p*<.001

	Model 0	Model 1	Model 2	Model 3	Model 4	Model 5
Production function						
Supplier development	1.162***	0.912***	1.151***	1.144***	1.159***	1.151***
initiative	(0.100)	(0.112)	(0.109)	(0.092)	(0.111)	(0.119)
Supplier evaluation	-0.185	-0.111	-0.176*	-0.223**	-0.235***	-0.232
	(0.071)	(0.056)	(0.085)	(0.068)	(0.063)	(0.064)
Industry dummies	included	included	included	included	included	included
Country dummies	included	included	included	included	included	included
Inefficiency model						
Relational norms (RN)		-2.284**			0.022	-0.107
		(0.757)			(0.787)	(1.088)
Buyer information			-0.169		2.767***	2.760***
sharing			(0.255)		(0.699)	(0.733)
Supplier information				-1.487***	-3.688***	-3.641**
sharing				(0.298)	(0.617)	(0.664)
RN*Buyer information						-0.293
sharing						(1.265)
RN*Supplier						-0.076
information sharing						(1.063)
Plant size		0.193	0.073	0.081	0.191	0.178
U		(0.165)	(0.219)	(0.165)	(0.167)	(0.150)
Variance parameters						
σ_{v}	0.256	0.268	0.259	0.251	0.238	0.238
χ^2	550.44***	207.16***	486.41***	445.21***	260.82***	200.34***
Log-pseudo likelihood	-64.69	-58.92	-62.27	-46.96	-27.90	-27.74
N	261	261	261	261	261	261
AIC	163.37	155.85	166.54	131.91	97.81	101.49
BIC	223.97	223.58	234.27	199.64	172.66	183.48

Table 4: Estimates of the stochastic frontier analysis with half-normal distribution

* p<.05, ** p<.01, *** p<.001 (two-tail)

Note: Dummy variables of country and industry are included in the production function. Numbers in parentheses are standard errors.

We obtain comparable results using the exponential distribution of the inefficiency term.

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	Model 1	Model 2	Model 3	Model 4
Hypothesized factors				
Relational norms (RN)	0.044**			-0.012
	(0.018)			(0.017)
Buyer information sharing		0.012		-0.116***
		(0.010)		(0.015)
Supplier information sharing			0.064***	0.162***
()			(0.010)	(0.016)
RN*Buyer info sharing				0.068
()				(0.040)
RN*Supplier info sharing				-0.063
				(0.039
Control variables				
Plant size	-0.003	-0.001	-0.005	-0.007
	(0.005)	(0.006)	(0.005)	(0.005)
Industry dummies	included	included	included	included
F-statistics	3.09***	0.63	20.34***	18.64***
Within R ²	0.024	0.005	0.141	0.314
Between R ²	0.360	0.066	0.087	0.029
Overall R ²	0.014	0.001	0.087	0.274
N	261	261	261	261

Table 5: Results using two-stage estimation approach (Robustness test)

* p<.05, ** p<.01, *** p<.001 (two-tail)

Note: Dummy variables of industry are included as control variables. Fixed-effect regression with country as the cluster variable.

Numbers in parentheses are standard errors.

Dependent variable: ln(Efficiency score from SFA)

Table 6: Simar and Wison (2007) two-stage estimation approach with bootstrap standard errors

	Model 1	Model 2
Hypothesized factors		
Relational norms (RN)	0.249***	0.256***
	(0.053)	(0.052)
Buyer information sharing	-0.187***	-0.215***
	(0.047)	(0.047)
Supplier information sharing	0.288***	0.298***
	(0.052)	(0.051)
RN*Buyer info sharing		-0.095
U		(0.134)
RN*Supplier info sharing		0.256
		(0.136)
Control variables		
Plant size	-0.023	-0.020
T	(0.014)	(0.013)
Industry dummies	included	included
Wald χ^2	86.06***	102.33***
Ν	253	253

* p<.05, ** p<.01, *** p<.001 (two-tail)

Note: Dummy variables of industry are included.

Numbers in parentheses are bootstrap standard errors with 2000 replications.

Appendix 1. Measurement Scales

Alignment method for factor loading with a total of 11 countries (groups)

Production function

Supplier development initiative	Fit function	Groups with	Approximate
	Contribution	approx.	measurement invariance
	of factor	measurement	(non-invariance) for
	loading	invariance	groups
We encourage our suppliers to continuously improve their production processes.	-29.47	11	1 2 3 4 5 6 7 8 9 10 11
We offer the necessary training to our suppliers.	-21.83	11	1 2 3 4 5 6 7 8 9 10 11
We share our vision and supply chain policy with our key suppliers.	-28.68	11	1 2 3 4 5 6 7 8 9 10 11
As our suppliers strive to improve their processes, we provide assistance.	-36.43	11	1 2 3 4 5 6 7 8 9 10 11
\mathbf{O}			
Supplier evaluation			
We have a formal supplier certification program.	-32.54	11	1 2 3 4 5 6 7 8 9 10 11
Our company has a formal system for tracking the performance of the suppliers	-25.81	11	1 2 3 4 5 6 7 8 9 10 11
that we deal with.			
Our plant has a formal system for evaluating and rewarding suppliers.	-29.19	11	1 2 3 4 5 6 7 8 9 10 11
We assess the performance of our suppliers through formal evaluation, using	-25.74	11	1 2 3 4 5 6 7 8 9 10 11
established guidelines and procedures.			
We provide our suppliers with feedback about the results of their evaluation.	-31.63	11	1 2 3 4 5 6 7 8 9 10 11
Supplier performance			
Fast delivery	-39.04	11	1 2 3 4 5 6 7 8 9 10 11

On-time delivery	-27.47	11	1 2 3 4 5 6 7 8 9 10 11
Product liability	-29.98	11	1 2 3 4 5 6 7 8 9 10 11
Reputation for corporate social responsibility	-33.66	11	1 2 3 4 5 6 7 8 9 10 11
Service level	-31.29	11	1 2 3 4 5 6 7 8 9 10 11
Technical skill	-35.09	11	1 2 3 4 5 6 7 8 9 10 11
Use of sustainability practices	-29.15	11	1 2 3 4 5 6 7 8 9 10 11
Willingness to adapt processes to meet your changing needs	-32.87	11	1 2 3 4 5 6 7 8 9 10 11
Willingness to adapt products to meet your changing needs	-35.95	11	1 2 3 4 5 6 7 8 9 10 11
Willingness to participate in your plant's new product development	-48.78	11	1 2 3 4 5 6 7 8 9 10 11

Hypothesized factors in the inefficiency model

Relational norms	Fit function	Groups with	Approximate
	Contribution	approx.	measurement invariance
	of factor	measurement	(non-invariance) for
	loading	invariance	groups
When we share our problems with our suppliers, we know they will respond with	-37.51	11	1 2 3 4 5 6 7 8 9 10 11
understanding.			
We can count on our suppliers to consider how their decisions and actions will	-22.23	11	1 2 3 4 5 6 7 8 9 10 11
affect us.			
We are able to anticipate our suppliers' actions in specific situations.	-38.82	11	1 2 3 4 5 6 7 8 9 10 11
Buyer information sharing			
(Our key suppliers have access to the following information about our plant)			

Cost information	-55.38	11	1 2 3 4 5 6 7 8 9 10 11
Demand change information	-30.36	11	1 2 3 4 5 6 7 8 9 10 11
Demand forecast information	-33.15	10	1 2 3 4 5 6 7 8 9 (10) 11
Plant capability information	-25.04	11	1 2 3 4 5 6 7 8 9 10 11
Inventory information	-26.70	11	1 2 3 4 5 6 7 8 9 10 11
Production capacity information	-30.61	11	1 2 3 4 5 6 7 8 9 10 11
Schedule information	-38.97	11	1 2 3 4 5 6 7 8 9 10 11
Supplier information sharing			
(Our plant has access to the following information about our key suppliers)			
σ			
Cost information	-32.92	11	1 2 3 4 5 6 7 8 9 10 11
Demand change information	-27.69	11	1 2 3 4 5 6 7 8 9 10 11
Demand forecast information	-48.91	11	1 2 3 4 5 6 7 8 9 10 11
Inventory information	-25.83	11	1 2 3 4 5 6 7 8 9 10 11
Production capacity information	-32.58	11	1 2 3 4 5 6 7 8 9 10 11
Productivity information	-36.14	11	1 2 3 4 5 6 7 8 9 10 11
Schedule information	-24.31	11	1 2 3 4 5 6 7 8 9 10 11

Psychometric properties

Production function

Supplier development initiative	Factor
AVE=0.475, CR=0.728, alpha=0.713	loading
We encourage our suppliers to continuously improve their production processes.	.658
We offer the necessary training to our suppliers.	.616
We share our vision and supply chain policy with our key suppliers.	.782
As our suppliers strive to improve their processes, we provide assistance. **	
Supplier evaluation	
AVE=0.594, CR=0.814, alpha=0.844	
We have a formal supplier certification program.	.787
Our company has a formal system for tracking the performance of the suppliers that we deal with.	.709
Our plant has a formal system for evaluating and rewarding suppliers. **	
We assess the performance of our suppliers through formal evaluation, using established guidelines and	.812
procedures.	
We provide our suppliers with feedback about the results of their evaluation. **	
Supplier performance	
AVE=0.508, CR=0.861, alpha=0.857	
Fast delivery **	
On-time delivery	.729
Product liability	.698
Reputation for corporate social responsibility **	
Service level **	
Technical skill	.703

Use of sustainability practices	.716
Willingness to adapt processes to meet your changing needs	.734
Willingness to adapt products to meet your changing needs	.697
Willingness to participate in your plant's new product development **	

Hypothesized factors in the inefficiency model

Relational norms	Factor
AVE=0.504, CR=0.752, alpha=0.755	loading
When we share our problems with our suppliers, we know they will respond with understanding.	.643
We can count on our suppliers to consider how their decisions and actions will affect us.	.705
We are able to anticipate our suppliers' actions in specific situations.	.775
Buyer information sharing	
(Our key suppliers have access to the following information about our plant)	
AVE=0.512, CR=0.840, alpha=0.865	
Cost information **	
Demand change information	.715
Demand forecast information	.675
Plant capability information **	
Inventory information	.753
Production capacity information	.695
Schedule information	.737

Supplier information sharing	
(Our plant has access to the following information about our key suppliers)	
AVE=0.626, CR=0.909, alpha=0.914	
Cost information	.676
Demand change information	.797
Demand forecast information	.799
Inventory information	.807
Production capacity information	.823
Productivity information **	
Schedule information	.833

(**) excluded due to low factor loadings (<0.6)

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