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Information technology capability and firm performance in the era of Big Data analytics: Evidence from the United States

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Abstract

Research Question: This study investigates the relationship between information technology (IT) capability and firm performance in the 2010s, the era of big data analytics (BDA), in the context of US companies.

Motivation: With the evolution of business intelligence and the proliferation of analytic tools that further improve IT capability, it is more important than ever to understand whether firms with stronger IT capabilities perform better.

Idea: After categorizing firms into pairs of IT leaders and control groups, the performance of each pair of firms in each group was compared.

Data: All data are publicly available from Compustat and InformationWeek (IW) 500.

Tools: The Wilcoxon signed-rank test and regression analysis were used to examine how the performance of IT leaders and control groups changed during the 2010–2017 period.

Findings: This results show no significant relationship between a firm's IT capability and its performance in the sample of US companies during this period.

Contribution: This study will help academicians and practitioners to better understand how the adoption and application of BDA derived from IT capability affects firm performance.

Keywords: Big data analytics (BDA), information technology (IT) capability, firm performance, IW 500, IT business value

JEL codes: M41, L25

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1. Introduction

Understanding how a firm's information technology (IT) capability affects its business performance has been controversial since the 1990s, and a wealth of literature on the value of IT in business has been developed since then (Chan, 2000; Dehning & Richardson, 2002; Kohli & Devaraj, 2003; Mahmood & Mann, 2000; Melville *et al.*, 2004; Wade & Hulland, 2004). Despite the doubts over IT's direct influence on firm performance (Carr, 2003; Clemons, 1986, 1991; Clemons & Row, 1991; Powell & Dent-Micallef, 1997), information systems (IS) researchers agree that a firm with superior IT capabilities has a huge competitive edge (Chae *et al.*, 2014). However, the IT capability–firm performance relationship is mixed (Mata *et al.*, 1995; Ray *et al.*, 2005). The IS community acknowledges that a firm's superior IT capabilities and improve firm performance per se, but can serve as a catalyst for higher-order organizational capabilities and improved firm performance (Ales *et al.*, 2016).

Whereas the 1990s were characterized by proprietary information systems, the 2000s were mainly governed by standardized and identical information systems, thanks to the rapid adoption of enterprise resource planning (ERP) and web technologies (Wang, 2010). Therefore, some studies have investigated the relationship between a firm's IT capability and its performance during different time periods. Bharadwaj (2000), for instance, has been cited 5, 141 times according to the data from Google Scholar, and is one of the most convincing studies that provides strong evidence for the link between IT capability and firm performance. This study claims that firms with superior IT capabilities performed better in business than their counterparts during 1991–1994. Using the resource-based view (RBV), it argues that firms categorized as IT leaders can make use of IT-related resources to produce unique IT capabilities that can lead to superior business performance and consequently, become the basis for their future competitive edge. Santhanam and Hartono (2003), an extension of Bharadwaj (2000), confirm that the performance of IT leaders is superior to that of corresponding paired firms in the control group (firms that match IT leader firms in terms of industry and size). Their study also finds that a firm's competitive advantage generated through its IT capability can be sustained. Following the widespread changes in IT and the way in which organizations leverage IT, another study, that of Chae et al. (2014) updates the above studies with fresh data from the early 2000s. Contrary to the previous findings, this study failed to find a link between IT leaders and superior business performance, as measured by profit and cost ratios, irrespective of whether this relationship is assessed by the industry-benchmark test, the pair-wise comparison test, or by regression analysis. Moreover, there is no evidence to support the sustainability of IT capability.

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The IT business value literature has continued to stress the potential and the ability of information systems to assist in decision making and enhance company performance (Davern & Kauffman, 2000; Mithas *et al.*, 2011; Melville *et al.*, 2004; Bhattacharya *et al.*, 2010). The capacity of IS to help a firm make timely decisions, generate insights that elevate a firm's comparative advantage, and serve as a tool to deal with environmental uncertainties has been acknowledged in firm performance research (Popovič *et al.*, 2014). As a result, firms depend on IS to obtain high-quality information that is relevant, reliable, accurate and timely (Popovič *et al.*, 2012; Wixom & Todd, 2005), which improves the quality of their decision making and boosts their business performance (Mithas *et al.*, 2011). Using this insightful information, firms are increasingly investing in a variety of technologies and incorporating them into their business activities and processes (Chen *et al.*, 2012).

However, over the years, business intelligence and analytics and big data analytics (BDA) have become increasingly important in both business and academic circles (Chen et al., 2012). The changes and transformation in business and society have been partly brought about by big data (Marr, 2016). From the academic community's perspective, BDA research has grabbed the attention of widely read scientific outlets, such as Science and the Proceedings of the National Academy of Sciences, because of its nature and significance (Agarwal & Dhar, 2014). Moreover, firms are constantly attempting to generate insights from the expanding value, variety, veracity, velocity, and volume of data to better utilize data and improve decision making (Lavalle et al., 2011; White, 2012). In addition to using data to solve known issues, firms are using data to uncover trends that they failed to notice earlier (Fosso Wamba et al., 2015). The tremendous opportunities associated with data, information, and analysis in various organizations have aroused keen interest in BDA, which is commonly recognized as the technologies, systems, techniques, methodologies, practices, and applications that analyze various critical business data to enable firms to better assess their business and markets, and to make timely and efficient business decisions (Gandomi & Haider, 2015; Mcafee & Brynjolfsson, 2012). BDA mainly derives from a firm's IT-based resources such as physical IT infrastructure, human IT resources, and intangible resources (Agrawal, 2013; Grant, 1991). IT capability is predominantly defined as a firm's capability to mobilize, deploy, and use its IT-based resources to enhance its business performance (Santhanam & Hartono, 2003), which plays a significantly positive role in BDA adoption and application.

Therefore, in this study we investigate the link between a firm's IT capability and its business performance in the 2010s (the BDA era) in US companies during 2010–2017. The timeframe begins with 2010 for the following two reasons. First, the number of academic publications that focused on BDA increased exponentially from 2010 and their focus was on big data-enabled business value, that is, to store, compute, analyze, visualize, and integrate the growing quantity of data from various sources and transform these data into business insights and value. (McAfee &

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Brynjolfsson, 2012; Wamba *et al.*, 2015). Second, this study serves as an extension of the previous studies that deployed data from 1990 through 1997 (Bharadwaj, 2000; Santhanam & Hartono, 2003) and 2000 to 2007 (Chae *et al.*, 2014) to reexamine the relationship between a firm's IT capability and its business performance. We use US-listed companies as our sample for several reasons. While BDA has both direct and indirect impacts on the financial performance of companies in China (Wamba *et al.*, 2017), its impact on the financial performance of companies in the US has not been investigated. As an ongoing study that aims to make a direct and valid comparison with the previous studies mentioned above, we choose US-listed firms for consistency. Moreover, as unlisted companies' relevant financial data are difficult to obtain and listed companies' financial data are publicly available, we select only listed companies in the United States as our sample.

Our hypotheses state that IT leaders are associated with superior business performance and sustained superior business performance compared with the control groups, both before and after adjusting for the impacts from previous years' financial performance through the following years' financial performance. Based on a sample of 55 sets of IT leaders and control groups selected from IW 500 and their financial data from 2010 through 2017, this study finds that although IT leaders showed and sustained higher profit ratios than the control groups before the financial halo effects adjustment, such a correlation no longer existed after the adjustment.

This study makes two contributions to the literature on the business value of IT. First, it determines the relationship between a firm's IT capability and its business performance by adding a new construct, which is the influence of BDA since the 2010s. BDA-supported decision making enables decision makers to use robust and timely data to enhance the effectiveness of business decisions, which leads to better business performance. Second, it updates past studies with fresh data from the 2010s. Past studies use data from the early 1990s and 2000s to investigate the relationship between a firm's IT capability and its business performance, and therefore, their findings may not be relevant after a decade of rapid change in the use of technology in business, especially BDA. Replications and updates are vital in the scientific community because they examine the robustness of theories and fortify experimental beliefs into agreed knowledge (Santhanam & Hartono, 2003). However, researchers often do not make full use of replicative studies to reexamine what has been scientifically explored (Berthon *et al.*, 2002).

The remainder of this paper is organized as follows. Section 2 reviews the relevant literature and proposes the corresponding hypotheses. Section 3 outlines the research approach, data sources, and data analysis procedures. Section 4 presents the main findings of the research. Section 5 describes some possible explanations and unavoidable limitations of this study, and offers some avenues for future research.

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2. Past research and development of hypotheses

2.1 IT capability and firm performance

IT capability is largely defined as a firm's capability to mobilize, deploy, and use IT-based resources to enhance its business performance (Santhanam & Hartono, 2003). According to Porter (2011), firms can increase their business performance by utilizing their IT capabilities to reduce costs, increase revenues, or both. First, firms with superior IT capabilities can increase product differentiation by leveraging web technologies to generate higher sales (Hitt & Brynjolfsson, 1996). For example, some banks have launched financial services that rely on websites to differentiate their products and services based on customers' profiles. In addition, firms with superior IT capabilities can (1) improve their sales performance by generating valuable resources such as patents and other intellectual property (Fahy & Hooley, 2002); (2) significantly reduce their marketing costs by increasing switching costs and improving customer loyalty (Davenport, 2006); (3) gain exclusive access to customers' information and personal preferences, which can potentially save future business research costs (Straub & Watson, 2001). This proprietary information could save firms wanting to expand their business to new markets from incurring heavy expenditure.

Grant (1991) classified IT-based resources into three categories: tangible resources consisting of the physical IT infrastructure elements, human IT resources with managerial and technical IT skills, and intangible IT-enabled resources such as knowledge assets, customer orientation, and synergy.

According to the RBV, firms can innovate swiftly and improve their products by using tangible resources (Bharadwaj, 2000). Compared with a firm with a disaggregated and less developed IT infrastructure, a firm with an integrated IT infrastructure can launch innovative IT applications quicker than its competitors (Bharadwaj, 2000). Therefore, tangible resources play a significant role in the BDA adoption process.

Human resources have technical IT skills and managerial skills. As BDA adoption entails significant changes in business processes and IT infrastructure, managerial capability plays a significant role in coordinating activities pertaining to process redesign (Zhu *et al.*, 2006).

Intangible resources consist of customer orientation, knowledge assets, and synergy (Bharadwaj, 2000). Past studies have indicated that customer orientation plays a significant role in innovation adoption (Agrawal, 2013). "Knowledge assets" refer to the integration of intellectual resources, namely, employees' knowledge, skills, and experiences into an organization's processes, policies, and information repositories

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(Bharadwaj, 2000). Knowledge assets also play a critical role in BDA adoption because if employees' knowledge and skills are well developed, they are more likely to assimilate new innovations. Synergy is the organization-wide pooling of resources and capabilities to achieve a combined result that is superior to its individual components (Bharadwaj, 2000). Firms that are able to share information and knowledge across different functional and business units are more agile and respond faster to situations. As BDA is empowered to share information across all key functions such as R&D, purchasing, production, warehousing, and marketing, it paves the way for sharing information and resources (Agrawal, 2013). Therefore, the synergy from intangible resources is positively related to BDA adoption and application. Based on the discussion mentioned above, for a firm to demonstrate IT capability, BDA adoption should be significantly and positively related to physical IT infrastructure, human IT resources, and intangible resources.

The above discussion is an example of how IT capability influences a firm's business performance in the era of the Internet and BDA. Specifically, the business value of BDA, which is reflected in business performance, can be defined as the transactional, informational, and strategic benefits derived from BDA and brought to firms (Akter & Wamba, 2016). For example, the large volume of online consumer reviews generated on shopping platforms enables organizations to analyze and understand different consumers' preferences and behavior, thus enabling the optimization of the organizations' service and sales performance (Salehan & Kim, 2016). As with any other technology, it is significant for organizations to understand the relevant mechanisms, processes, and results of big data, through which business value is developed and added (Mikalef *et al.*, 2018). Although big data can give rise to various implications and applications in terms of valuation, legal protection, policy, and commercialization for organizations (Iqbal et al., 2020), only the implications that are reflected in organizations' financial performance are discussed in this study. Therefore, the following two hypotheses are formulated wherein financial ratios are leveraged to understand the impacts of BDA.

H1: Firms that had superior IT capabilities in the 2010s are associated with higher average profit ratios than those of control firms.

H2: Firms that had superior IT capabilities in the 2010s are associated with lower average cost ratios than those of control firms.

2.2 Sustainability of IT capability and business performance

The important question of whether firms' successful business performance resulting from IT capability can be sustained in the long run has become more difficult to answer. On the one hand, unlike the early stages of IT development that were characterized by proprietary information systems, the modern IT environment

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mainly consists of highly standardized IS applications due to the quick adoption of web technology and ERP, which makes it easier for a firm to counter and even surpass its competitors' IT capabilities. The time and costs involved in developing IS are also significantly reduced, thanks to advancements such as off shoring, readily available web search engines, and outsourcing (Porter, 2001). However, all of the IT-based competitive advantages obtained from these developments are short lived (Carr, 2003).

On the other hand, owing to a large and increasing amount of sensor-generated, mobile, and web-based data being obtained from IS, a vast amount of relevant, contextualized, and detailed content is made available to firms, which guides their short-term and long-term development and which is beneficial for BDA value creation and realization (Agarwal & Dhar, 2014; Chen *et al.*, 2012; Grover *et al.*, 2018). IT-enabled intangible assets, such as knowledge assets, are likely to enhance technological competence and capabilities, thus influencing firm performance and IT projects (Barney, 1991; Bharadwaj, 2000; Mata *et al.*, 1995; Melville *et al.*, 2004; Santhanam & Hartono, 2003; Wade & Hulland, 2004). Strategies that are driven by BDA contribute to business growth with sustained performance, which is also reflected in the internal processes (Singh & El-Kassar, 2019).

Despite the changing IT environment in which a firm can easily obtain applications and analytical insights (Chae *et al.*, 2014), a firm's IT capability is likely to be linked to sustainable business performance. Therefore, we propose the following two hypotheses:

H3: Firms that had superior IT capabilities in the 2010s are associated with higher average profit ratios than those of control firms in all subsequent years.

H4: Firms that had superior IT capabilities in the 2010s are associated with lower average cost ratios than those of control firms in all subsequent years.

2.3 The financial halo effect

Past studies have found that the results of popular industry rankings, for instance *Fortune*'s Most Admired Companies, can be greatly affected by past financial performance, and the influence of past financial performance on a reputation survey is defined as a financial halo effect (Brown & Perry, 1994). A previous study (Bharadwaj, 2000) used a method whereby a firm is selected as an IT leader on the basis of its IT capability rather than its previous financial performance, and that there is no evidence of a financial halo effect. It is very important to ensure that a firm is recognized as an IT leader because of its superior IT capability and not because of its previous financial performance (Chae *et al.*, 2014). Otherwise, the reliability and

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validity of business performance comparisons between IT leaders and control groups becomes questionable.

A follow-up study (Santhanam & Hartono, 2003) further investigated the financial halo effect. They argued that use of the financial halo effect in Bharadwaj (2000) to select IT leaders may not be sufficient, as the researcher made use of a firm's previous five years' financial performance despite the fact that the firm's ranking was affected most by its immediate past performance. Therefore, they claimed that the influence of IT capability should be reexamined via a more conservative and rigorous method by adjusting for a firm's past financial performance. Their concerns were legitimate, because the much better performance of the IT leaders than the control groups in the pair-wise comparison test was significantly weak when firms' previous financial performance was adjusted for.

In line with Santhanam and Hartono (2003), Chae *et al.*, (2014) also used a method to ensure that IT leaders' capabilities, rather than their past financial performance, differentiates them from control companies. Consistent with their research, a firm's previous year performance has a huge impact on its current year performance.

Similar to Santhanam and Hartono's (2003) approach, this study also seeks to ascertain that IT capability, and not previous financial performance, differentiates IT leaders from control companies. Consequently, additional tests controlling the effects of previous financial performance on the results should be performed. The following hypotheses are therefore proposed:

H5: Firms that had superior IT capabilities in the 2010s are associated with higher average profit ratios than those of control firms after adjustment for prior financial performance.

H6: Firms that had superior IT capabilities in the 2010s are associated with lower average cost ratios than those of control firms after adjustment for prior financial performance.

H7: After adjustment for previous financial performance, firms that had superior IT capabilities in the 2010s are associated with higher average profit ratios than those of control firms in the subsequent years.

H8: After adjustment for previous financial performance, firms that had superior IT capabilities in the 2010s are associated with lower average cost ratios than those of control firms in the subsequent years.

As this study was conducted after those of Bharadwaj (2000), Santhanam and Hartono (2003), and Chae (2014), and their research methodologies are heavily referenced, a comparison between those studies is presented in Table 1.

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Studies	Sample	Benchmark (Control Group)	Measures
Bharadwaj	IT leaders from	IT leader versus	Business performance measured
(2000)	500 listings	similar size and industry	to 1994.
Santhanam	IT leaders from	IT leaders versus	Business performance measured
and Hartono	1991–1994 IW	industry average	by profit and cost ratios from 1991
(2003)	500 listings		to 1994 2. Sustainability of superior business performance from 1005 to 1007
Chae et al	IT leaders	IT leaders versus	Business performance measured
(2014)	selected from	control companies	by profit and cost ratios from 2001
(_*-)	2001–2004 IW	of similar size and	to 2004 2. Sustainability of
	500	industry	superior business performance
			from 2005 to 2007.
This study	IT leaders	IT leaders versus	Business performance measured
	selected from	control companies	by profit and cost ratios from 2010
Rahman and	2010–2013 IW	of similar size and	to 2013 2. Sustainability of
Zhao (2020)	500	industry	superior business performance
			from 2014 to 2016.

3. Methods and data collection

Bharadwaj's (2000) "matched sample comparison group" is used in this study to test the formulated hypotheses empirically. This technique selects a treatment sample and a control sample, and then compares the interest variables' levels between these two groups. The treatment sample consists of IT leaders with superior IT capabilities and the control sample consists of firms that match the control sample in terms of industry and size. Next, the IT leaders' performance measurement results are compared with those of the control group.

3.1 Data collection

IS researchers have been using IW 500 for years to identify IT leaders with superior IT capabilities. Since 1989, IW 500 has selected and recognized 500 companies as annual leaders in business technology innovation. Its criteria for choosing IT leaders have evolved to better reflect ever-changing business and technology advancements and the refinement of its benchmarking power. Despite the frequent changes in its selection criteria, IW 500 continues to be recognized as a reliable indicator of firms' IT capabilities and has been cited in previous studies (Bharadwaj, 2000; Hitt & Brynjolfsson, 1996; Chae *et al.*, 2014; Santhanam & Hartono, 2003; Stoel &

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Muhanna, 2009). Firms with superior IT capabilities are those recognized by IW 500 during 2010–2013. To ensure the selection of firms with sustainable IT capabilities, the IT leader sample was limited to firms that were listed in IW 500 for at least two years during those four years.

The matching control firms were selected on the basis of sales levels—those that had the closest five-year average sales to the IT leaders were chosen. In line with Bharadwaj (2000), the control firms' average sales should be between 70% and 130% of those of the leader firms. Essentially, the control firm should be from the same industry as the IT leader, and its average sales from 2005 through 2009 should be between 70% and 130% of the leader's average sales during the same period. Table 2 summarizes the sample selection in a systematic manner.

Stop	Ducasduna	Number
step	Frocedure	of Companies
1	Identify companies listed in the IW 500 from 2010 to 2013.	2,000
2	Restrict the IT leader sample to those that were listed at leas twice in the IW 500 from 2010 to 2013.	t 569
3	Select IT leaders with comparable companies.	276

Table 2. Sample selection methods

It is critical to choose a control firm from the same industry as the IT leader and ensure that it is of the same size as the IT leader for the following reasons. First, operating performance differs considerably across different industries and firm sizes. Differences in performance caused by the variance in industry and firm size are significantly reduced by using matching samples (Chae *et al.*, 2014). Moreover, according to the accounting literature, industry type and firm size are widely used by accounting methods such as amortization and depreciation, in the calculation of costs (Bharadwaj, 2000). Therefore, effective control of the effects of variance from accounting methods leads to are liable comparison between profitability and cost ratios.

We selected a single control firm (benchmark) for each IT leader firm and did not consider all of the other firms in that industry for comparison (Chae *et al.*, 2014) for the following reasons. Unlike the studies conducted by Bharadwaj (2000) and Santhanam and Hartono (2003) that used only 56 leader and control pairs, this study uses 276 groups of samples. Therefore, it may not be appropriate to incorporate all firms in an industry to calculate the benchmark that is arrived at with a single set of benchmark firm data (Chae *et al.*, 2014). As a result, each IT leader firm is paired

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with a control group using the above methods. As Table 3 indicates, firms categorized as IT leaders are large, with total assets of US\$24.4 billion and average sales of US\$7.5 billion. This research shows that an IT leader is almost twice as large as the average firm in its respective industry. As discussed earlier, a firm's size directly influences its IT budget and resources, and IT leaders could have greater IT resources than other firms in the industry.

Table 3. Comparison	between th	<u>ie IT leader</u>	r group and	the control g	roup
Deservite time Vervie bles	IT Leader	Sample	Contr	ol Sample	
Descriptive variables	Mean	Median	n Mean	Media	an
Sales (billion \$)	7.	512	1.489	4.315	1.943
Assets (billion \$)	24.	398	2.145	12.826	2.596
Number of Employees (thousand)	20.	173	5.364	16.252	6.386

The variables chosen in this study to reflect business performance are the same as those in Bharadwaj (2000). Superior business performance is defined by profit and cost performance parameters such as higher profit ratios and lower cost ratios. Specifically, the profit ratios category, which incorporates the measurement of firm size based on sales performance, assets value, expenses, and the number of employees, includes the following.

- 1) Return on assets (ROA). ROA has been widely adopted in the IT business value literature to measure a firm's profitability and is correlated with other profitability measures (Kanodia *et al.*, 2016).
- Return on sales (ROS). ROS is also widely used in the IT business value literature as an indicator of the net profit margin of a firm (Kanodia *et al.*, 2016).
- 3) Operating income to assets (OI/A)
- 4) Operating income to sales (OI/S). As the operating income excludes revenue streams from other sources such as interest income, the measurement of OI/A and OI/S are appropriate to indicate the business value of IT.
- 5) Operating income to employees (OI/E). As Bharadwaj (2000) did not clearly define operating income and earnings before interest payment, the earnings before interest and taxes (EBIT) is used as the operating income in this study. This ratio is used to measure the relative profitability of effective IT users.

The cost ratios category is detailed below:

- 1) Total operating expenses to sales (OPEXP/S). Total operating expenses is selected because it is recognized as the most universal and comprehensive measure of the cost of a firm's operations (Enache *et al.*, 2018).
- 2) Cost of goods sold to sales (COG/S) and

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3) Selling and general administrative expenses to sales (SGA/S). Both the cost of goods sold and the general administrative expenses are widely accepted in accounting as measures of the costs related to production and overheads (Bharadwaj, 2000). Operating expenses are computed as the sum of SGA and COG, in line with Bharadwaj (2000).

It is widely accepted that the consistent application of accounting treatment is important in an accounting-related context (Wüstemann *et al.*, 2000). Although the measurement of the above ratios is subject to changes due to modifications in the accounting treatment from time to time, these changes tend to improve the reliability and accuracy of accounts (Chulle *et al.*, 2015). Therefore, the comparison of the ratios during different years focuses more on what the relative accounts stand for than on how the accounts are calculated.

The Compustat database was the source of financial data. The data were used to determine whether IT leaders have lower costs and higher profits than their corresponding control firms during 2010–2013.

The sustainability of superior business performance was assessed by whether IT leaders hold a higher profit and lower cost over time. Both the IT leaders' and control firms' profit and cost ratios were analyzed for the 2014–2017 period. The constructs and corresponding measurements that were used are summarized in Table 4.

However, after extracting all IT leaders and control groups' financial data from Compustat's database, only 56 sets of IT leaders and control groups remained valid for subsequent research because of the following reasons. Not all of the companies selected by IW 500 were listed companies, and the financial data of unlisted companies were not available on Compustat. As a result, 13 sets of groups were eliminated. In addition, because this study requires consecutive data from 2010 through 2017, any variable missing in any year makes this group invalid. Most of the missing values were those of employee numbers and selling and general administrative expenses. As a result, only 56 sets of groups that contained IT leaders and control groups were retained.

Category	Constructs	Definition	Measurement	Sources
Independent	IT Capability	A firm's ability	The firms ranked	Bharadwaj
variable		to assemble,	Information Week	(2000); Rai et
		integrate and	500 more than twice	al. (1997)
		deploy IT-	from 2010 to 2013	
		based resources.		
Dependent	Superior	Higher profit	ROA, ROS, OI/A,	Bharadwaj
variables	Business	and lower cost	OI/S, OI/E, COG/S,	(2000); Porter
	Performance	than the control	SGA/S, OPEXP/S	(1985)
			,	()

Table 4. Summary of constructs and measurement

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Category	Constructs	Definition	Measurement	Sources
		group.	from 2010 to 2013	
	Sustained	Sustained	ROA, ROS, OI/A,	Bharadwaj
	Superior	higher profit	OI/S, OI/E, COG/S,	(2000);
	Business	and lower cost	SGA/S, OPEXP/S	Santhanam
	Performance	than the control	from 2014 to 2017	and Hartono
		group.		(2003)

3.2 Statistical tests

This study investigates whether the firms with higher IT capabilities are likely to show superior financial performance as compared with the matching control firms. Comparing the mean value of variables for the IT leader group with that of the control sample via a standard *t*-test was one of the methods used to test the hypotheses. However, because the sample is not normally distributed, the Wilcoxon signed-rank test, which is a non-parametric test, was used. This test classifies one IT leader firm and a corresponding control firm into one pair and measures the differences between the pairs. By adopting the Wilcoxon signed-rank test, the variance in performance caused by the differences in industry can be effectively removed (Chae *et al.*, 2014).

Consistent with Santhanam and Hartono (2003), the same regression analysis methods for testing hypotheses H5, H6, H7, and H8 were adopted. Each financial performance measurement was conducted by two separate regression analyses as follows. The first analysis investigated how previous performance affects the current year's performance, and therefore, a regression analysis that regresses previous years' performance on current performance was performed. An additional binary variable was used in the second analysis to distinguish IT leaders from control firms (1 for IT leaders, and 0 for the control group). The two models are expressed as follows:

 $FP_t = \beta_0 + \beta_1 FP_{(t-1)}$

$$FP_t = \alpha_0 + \alpha_1 FP_{(t-1)} + \alpha_2 D$$

Where,

FP denotes financial performance;

t denotes the time period; D denotes the (0,1) binary variable;

 α_1 , α_2 , and β_1 are regression coefficients; and

 β_0 and α_0 are the intercepts.

 α_2 , the significance of the coefficient of the dummy variable in the second model, determines whether IT capability has a significant influence on business performance after adjusting for the effects of previous financial performance on current performance.

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4. Empirical results

4.1 Empirical results for superior business performance and sustained superior business performance

The statistical results of the Wilcoxon signed-rank test from 2010 through 2017 are presented in Tables 5 and 6. Both the mean and median of the performance measurement are displayed. The statistical test results are reported as P values and Z values. H1, which states that the IT leaders have higher profit ratios than those of the control groups, is supported.

The IT leaders group was associated with better performance in terms of profit ratios including ROA, ROS, OI/A, and OI/E for all four years from 2010 through 2013. Notably, the ROA and ROS showed high significance levels from 2011 to 2013. H2, which pertains to cost ratios, holds partially true. IT leaders' OPEXP/S was higher in three of the four years, COG/S was higher in all four years, and only SGA/S was lower in all four years.

The results for H3 and H4, which examine whether superior business performance driven by IT capability can be sustained over time, are displayed in Table 6. IT leaders did show higher profit ratios than those of control firms in all four years from 2014 through 2017, and thus hypothesis H3 is fully supported.

In line with H1, the ROA and ROS showed high significance levels from 2014 through 2017. In accordance with the results for H4, IT leaders had a higher COG/S than that of the control group in all four years but experienced lower OPEXP/S and SGA/S in all of the four years. Consequently, hypothesis H6 is partially supported.

4.2 Empirical results for the adjustment of financial halo effects

To understand how IT capability affects business performance after adjustment for previous years' firm performance, we conducted regression analysis. The results are reported in Table 7. Specifically, the regression coefficients of the models in equations 1 and 2 and changes in R-squares are presented in Table 7.

Consistent with the study's expectation, the previous year's organizational performance had a significant impact on the current year's organizational performance. However, according to the significance of the coefficient of the dummy variable for profit ratios and cost ratios in the second model, only the ROA in 2011 with a positive coefficient was significant at the 5% level. Therefore, H5, H6, H7, and H8 are not supported. The results of the hypotheses tests are summarized in Table 8.

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			Ä	010			5	011			50	012			20	13	
		Mean	Median	Z Value	P Value	Mean	Median	Z Value	P Value	Mean	Median	Z Value	P Value	Mean	Median	Z Value	P Value
ROA	Control Group	-0.135	0.189	3151	010	0.005	0.206	Lat 0	1000 KK	0.052	0.204	000 0	1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1	0.033	0.214	500	***
	IT Leaders	0.184	0.247	כוניו-	0.188	0.381	0.356	1977-	770-0	0.414	0.371	668-2-	0.004	0.401	0.373	174-7-	**C10.0
ROS	Control Group	-0.225	0.216	1 204	0106	-0.027	0.211	202	***0000	0.065	0.211	7647	0.011 **	0.084	0.258	2005	**700
	IT Leaders	0.164	0.249	+7C-1-	001-0	0.481	0.337	000-7-	600.0	0.504	0.368	1407-	110.0	0.523	0.415	C20.7-	000.0
OI/A	Control Group	0.056	0.069	1 026	0.050	0.075	0.079	2000	******	0.082	0.080	612.1	101.0	0.063	0.075	2001	0.050 K
	IT Leaders	0.102	0.086	CC4.1-	CCU.U	0.111	0.093	050.2-	0.042	0.112	0.092	740'1-	101.0	0.100	0.099	CC4.1-	* CCU.U
S/IO	Control Group	0.055	0.083	L23 I		0.129	0.107	1106	0900	0.136	0.127	202.0	0.500	0.133	0.120		696.0
	IT Leaders	0.126	0.111	/001-	/11.0	0.158	0.126	001.1-	607.0	0.149	0.126	050.0-	760.0	0.143	0.129	705-0-	c0/-0
OI/E	Control Group	-12.189	24.600	2361	175	55.344	30.686	2131	0110	40.545	35.311	101.0	102.0	44.309	30.274	0970	069.0
	IT Leaders	47.034	26.816	/001-	C/T-0	76.507	42.437	/101-	671.0	62.538	39.956	-0.474	170.0	64.801	35.622	-01-102	600.0
OPEXP/S	Control Group	0.845	0.856	0070	0 5 0	0.802	0.843	2100	100.0	0.785	0.820		121.0	0.797	0.831		201.0
	IT Leaders	0.826	0.827	070'0	טכביט	0.804	0.812	/10.0	102.0	0.812	0.840	161-0-	0.401	0.818	0.840	670-0-	U-407
COG/S	Control Group	0.541	0.530		236 V	0.510	0.508	1031	0100	0.497	0.486	2001	*0.00	0.493	0.458	500	0.0548
	IT Leaders	0.714	0.546	776'0-	/000	0.741	0.582	+co.1-	701.0	0.755	0.571	cc <i>4</i> -1-		0.715	0.567	1761-	- +cu.u
SGA/S	Control Group	0.304	0.280	1 266	0170	0.292	0.283	1 373	100.0	0.288	0.296	1 1 7 3	110.0	0.303	0.318	1 001	
	IT Leaders	0.243	0.215	000-1	7/1-0	0.237	0.198	C77-1	177.0	0.243	0.199	C/1-1	147.0	0.258	0.210	100.1	007-0

Accounting and Management Information Systems Table 5. Results of test on superior business performance

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Mean Median X value Median X value Mean Median X value <	2016		6	017
ROA Control Group 0.07 0.323 2.086 0.037^{*} 0.236 0.237 0.236 0.236 0.239 0.339 0.339 0.339 0.339 0.336 0.326 0.326 0.339 0.336 0.336 0.336 0.336 0.336 0.336 0.336 0.336 0.367 0.366 0.367 0.366 0.336 0.336 0.336 0.336 0.336 0.336 0.336 0.336 0.336 0.336 0.336 0.336 0.336 0.336 0.336 0.336 0.336 0.336 0.336 0.336 0.336 0.336 0.336 0.336 0.336 0.336 0.336 0.336 0.336 0.336 0.336 0.336 0.336 0.336 0.336 0.336 0.336 0.336 0.336 0.336 0.336 0.336 0.336 0.336 0.336 0.336 0.336 0.336 0.336 0.336	lean Median Z Value	P Value Me	an Median	Z Value P
$ \begin{array}{ c c c c c c c c c c c c c c c c c c c$	256 0.203 7 320	0.070**	06 0.199	0 102 0
$ \begin{array}{ c c c c c c c c c c c c c c c c c c c$	369 0.384	0.3	74 0.390	0 176-7-
$ \begin{array}{ c c c c c c c c c c c c c c c c c c c$	308 0.228 2.720	0.076** 0.2	12 0.312	1 650 0
$ \begin{array}{ c c c c c c c c c c c c c c c c c c c$.613 0.476	0.0	80 0.472	
$\begin{tabular}{ c c c c c c c c c c c c c c c c c c c$	059 0.069 1 300	0.0	74 0.070	1173
$ \begin{array}{ c c c c c c c c c c c c c c c c c c c$	0.095 0.091	0.0	95 0.089	
$ \begin{array}{ c c c c c c c c c c c c c c c c c c c$.067 0.101 1 0.77	0.1* 0.1	44 0.112	0.012
$ \begin{array}{c c c c c c c c c c c c c c c c c c c $	181 0.144	0.1	78 0.147	
$\begin{tabular}{ c c c c c c c c c c c c c c c c c c c$	1.930 30.048 2.028	0.043** 51.(092 31.487	VII 1-
$ \begin{array}{c c c c c c c c c c c c c c c c c c c $	5.663 41.653	80.0	222 39.973	
IT Leaders 0.792 0.815 0.767 0.811 0.777 0.811 0.778 0.813 0.772 0.813 0.775 0.811 0.778 0.813 0.772 0.679 0.778 0.813 0.772 0.675 1.12 0.490 0.556 0.499 0.749 0.493 0.472 0.679 0.748 0.731 0.748 0.813 0.747 0.679 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.	865 0.852 1.039	0.790 0.7	94 0.833	0 553 0
COG/S Control Group 0.526 0.492 0.611 0.490 0.243 0.493 0.472 0.679 IT Leaders 0.703 0.561 0.656 0.439 0.630 0.495 0.679 0.679 SGANS Control Group 0.293 0.294 1.785 0.074* 0.313 0.326 1.617 0.106 0.373 0.315 2.044 Tritorion 0.444 0.414 0.414 0.326 1.617 0.106 0.375 0.345 2.044	778 0.813	0.7	79 0.804	
IT Leaders 0.703 0.561	493 0.472	0.497 0.4	98 0.474	-0.628
SGA/S Control Group 0.293 0.294 1.785 0.074* 0.313 0.326 1.617 0.106 0.373 0.315 2.044 retrieved 0.344 0.311 0.306 0.352 0.332 2.044	.630 0.445	0.0	07 0.496	070-0-
	373 0.315 7 0.44	0.041** 0.2	96 0.276	1 340
	263 0.239	0.2	48 0.227	

Table 6. Results of test on superior business performance sustainability

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2011				
	<u>Model</u>	<u>R Square Change</u>	2010 Financial Performance	<u>Dummy</u> Variable
ROA	1	0.027	0.130	
	2	0.099	0.088	0.348**
ROS	1	0.054	0.202	
	2	0.120	0.157	0.448
OI/A	1	0.017	0.116	
	2	0.044	0.086	0.033
OI/S	1	0.058	0.137	
	2	0.064	0.130	0.020
OI/E	1	0.008	0.051	
	2	0.013	0.046	18.431
OPEXP/S	1	0.008	0.087	
	2	0.008	0.087	0.003
COG/S	1	0.000	-0.010	
	2	0.048	-0.054	0.240
SGA/S	1	0.004	0.058	
	2	0.027	0.038	-0.053
<u>2012</u>				
	Model	<u>R Square Change</u>	2011 Financial Performance	<u>Dummy</u> Variable
ROA	1	0.093	0.284	<u>, anabic</u>
	2	0.144	0.220	0.280
ROS	1	0.246	0.476	0.200
	2	0.262	0.438	0.217
OI/A	1	0.002	0.039	
0.011	2	0.027	0.011	0.030
OI/S	-	0.035	0.210	0.020
010	2	0.036	0.206	0.007
OI/E	-	0.062	0.253	0.007
	2	0.066	0.248	16.748
OPEXP/S	1	0.011	0.101	10., 10
	2	0.019	0.101	0.027
COG/S	-	0.022	-0.146	0.027
	2	0.100	-0.207	0.306
SGA/S	-	0.007	0.081	0.000
	2	0.022	0.062	-0.042
2013	-	==		0.012
	Model	R Square Change	2012 Financial Performance	<u>Dummy</u> Variable
ROA	1	0.135	0.406	
	2	0.166	0.343	0 243
	2	0.100	018 18	0.215
ROS	2	0.155	0.416	0.215

Table 7. Regression analysis

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$\begin{array}{c c c c c c c c c c c c c c c c c c c $					
2 0.051 0.219 0.030 0I/S 1 0.001 0.025 2 0.002 0.024 0.010 OI/E 1 0.043 0.163 1 2 0.010 0.086 0.019 2 2 0.010 0.086 0.019 2 COG/S 1 0.042 0.153 0.182 SGA/S 1 0.079 0.290 2 2 0.088 0.277 -0.032 2014 2 0.096 0.444 0.278 ROA 1 0.083 0.504 278 ROA 1 0.002 0.055 0.204 0.387 OI/A 1 0.0015 -0.182 0.053 0.053 OI/S 1 0.015 -0.182 0.027 0.027 2 0.023 -0.068 101.259 0.027 0.027 0.027 COG/S 1 0.001 -0.02	OI/A	1	0.036	0.246	
$\begin{array}{c c c c c c c c c c c c c c c c c c c $		2	0.051	0.219	0.030
2 0.002 0.024 0.010 OI/E 1 0.043 0.163 2 0.049 0.158 17.018 OPEXP/S 1 0.006 0.092 2 0.010 0.086 0.019 COG/S 1 0.042 0.153 0.182 SGA/S 1 0.079 0.290 0.277 0.032 2014 0.088 0.277 0.032 Z014 0.083 0.504 0.278 ROA 1 0.083 0.504 0.278 ROS 1 0.052 0.261 0.278 ROS 1 0.052 0.261 0.278 QI/A 1 0.002 0.055 0.021 2 0.031 -0.182 0.027 0.027 QI/E 1 0.001 -0.029 0.027 QI/E 1 0.001 0.0275	OI/S	1	0.001	0.025	
$\begin{array}{c c c c c c c c c c c c c c c c c c c $		2	0.002	0.024	0.010
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	OI/E	1	0.043	0.163	
$\begin{array}{c c c c c c c c c c c c c c c c c c c $		2	0.049	0.158	17.018
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	OPEXP/S	1	0.006	0.092	
$\begin{array}{ccccc} {\rm COG/S} & 1 & 0.042 & 0.195 \\ & 2 & 0.071 & 0.153 & 0.182 \\ {\rm SGA/S} & 1 & 0.079 & 0.290 & \\ & 2 & 0.088 & 0.277 & -0.032 \\ \hline \end{array} \\ \hline \begin{array}{c ccccccccccccccccccccccccccccccccccc$		2	0.010	0.086	0.019
$\begin{array}{c c c c c c c c c c c c c c c c c c c $	COG/S	1	0.042	0.195	
$\begin{array}{c c c c c c c c c c c c c c c c c c c $		2	0.071	0.153	0.182
2 0.088 0.277 -0.032 2014 Model R Square Change 2013 Financial Performance $Dummy Variable ROA 1 0.083 0.504 Variable ROA 1 0.083 0.504 Variable ROS 1 0.052 0.261 Variable QUA 1 0.002 0.055 0.387 OI/A 1 0.002 0.055 0.053 QU/S 1 0.015 -0.182 0.057 QU/E 1 0.001 -0.024 -0.027 OPEXP/S 1 0.001 -0.024 -0.027 COG/S 1 0.030 0.177 2 QPEXP/S 1 0.079 0.275 2 QUA 0.049 0.146 0.144 SGA/S 1 0.077 0.750 2 QUA 0.062 0.248 0.511 ROA 1 0.077$	SGA/S	1	0.079	0.290	
2014 Model R Square Change 2013 Financial Performance $Dummy Variable ROA 1 0.083 0.504 Variable ROS 1 0.052 0.261 Variable ROS 1 0.002 0.055 Variable QU/A 1 0.0015 -0.182 Variable QU/E 1 0.000 -0.021 Variable QU/E 1 0.001 -0.029 Variable QPEXP/S 1 0.001 -0.029 Variable QU/P 0.049 0.146 0.144 SGA/S 1 0.079 0.275 QU/P 0.062 0.248 0.511 ROA 1 $		2	0.088	0.277	-0.032
$\begin{tabular}{ c c c c c c } \hline Model & R Square Change & 2013 Financial Performance & Variable \\ \hline ROA & 1 & 0.083 & 0.504 & & & \\ 2 & 0.096 & 0.444 & 0.278 \\ ROS & 1 & 0.052 & 0.261 & & & \\ 2 & 0.087 & 0.204 & 0.387 \\ 0!/A & 1 & 0.002 & 0.055 & & & \\ 2 & 0.034 & 0.021 & 0.053 \\ 0!/S & 1 & 0.015 & -0.182 & & & \\ 2 & 0.031 & -0.188 & 0.057 \\ 0!/E & 1 & 0.000 & -0.021 & & \\ 2 & 0.023 & -0.068 & 101.259 \\ OPEXP/S & 1 & 0.001 & -0.029 & & \\ 2 & 0.006 & -0.024 & -0.027 \\ 2 & 0.006 & -0.024 & -0.027 \\ COG/S & 1 & 0.030 & 0.177 & & \\ 2 & 0.049 & 0.146 & 0.144 \\ SGA/S & 1 & 0.079 & 0.275 & & \\ 2 & 0.091 & 0.261 & -0.037 \\ \hline \end{tabular}$	2014				
ROA 1 0.083 0.504 2 0.096 0.444 0.278 ROS 1 0.052 0.261 2 0.087 0.204 0.387 OI/A 1 0.002 0.055 0.053 2 0.034 0.021 0.053 OI/S 1 0.015 -0.182 0.057 2 0.031 -0.182 0.057 0.057 OI/E 1 0.000 -0.021 0.057 2 0.023 -0.068 101.259 OPEXP/S 1 0.001 -0.029 -0.027 COG/S 1 0.030 0.177 -0.027 2 0.049 0.146 0.144 SGA/S 1 0.079 0.275 2 0.091 0.261 -0.037 Dummy Variable ROA 1 0.040 0.291 2 0.062 0.248 0.5111 <t< td=""><td></td><td>Model</td><td>R Square Change</td><td>2013 Financial Performance</td><td><u>Dummy</u> Variable</td></t<>		Model	R Square Change	2013 Financial Performance	<u>Dummy</u> Variable
Model 2 0.096 0.444 0.278 ROS 1 0.052 0.261 0.387 OI/A 1 0.002 0.055 0.053 QI/S 1 0.015 -0.182 0.057 QI/S 1 0.015 -0.182 0.057 QI/E 1 0.000 -0.021 0.057 QI/E 1 0.000 -0.021 0.027 QUE 1 0.001 -0.029 0.027 QOG/S 1 0.030 0.177 0.027 COG/S 1 0.030 0.177 0.037 2 0.049 0.146 0.144 SGA/S 1 0.079 0.275 0.037 2 0.091 0.261 -0.037 0.037 2015 2 0.062 0.248 0.511 ROA 1 0.040 0.291 0.261 -0.037 2 0.062 0.248 0.511	ROA	1	0.083	0.504	
ROS 1 0.052 0.261 2 0.087 0.204 0.387 OI/A 1 0.002 0.055 2 2 0.034 0.021 0.053 OI/S 1 0.015 -0.182 2 2 0.031 -0.188 0.057 OI/E 1 0.000 -0.021 2 2 0.023 -0.068 101.259 OPEXP/S 1 0.001 -0.029 2 2 0.049 0.146 0.144 SGA/S 1 0.079 0.275 2 2 0.049 0.261 -0.037 OILS Model R Square Change 2014 Financial Performance Dummy Variable ROA 1 0.040 0.291 2 2 0.062 0.248 0.511 ROS 1 0.040 0.291 2 2 0.062 0.248 0.511		2	0.096	0.444	0.278
2 0.087 0.204 0.387 OI/A 1 0.002 0.055 0.053 2 0.034 0.021 0.053 OI/S 1 0.015 -0.182 0.067 2 0.031 -0.182 0.057 OI/E 1 0.000 -0.021 0.057 QPEXP/S 1 0.001 -0.029 0.027 2 0.030 0.177 0.0261 -0.027 COG/S 1 0.030 0.177 0.0261 -0.037 2 0.049 0.146 0.144 0.047 SGA/S 1 0.079 0.261 -0.037 2 0.091 0.261 -0.037 2 0.062 0.248 0.511 ROA 1 0.040 0.291 2 2 0.062 0.248 0.511 ROS 1 0.077 0.750 2 2 0.018 0.127	ROS	1	0.052	0.261	
$\begin{array}{c cccccc} \text{OI/A} & 1 & 0.002 & 0.055 & 0.055 \\ & 2 & 0.034 & 0.021 & 0.053 \\ \text{OI/S} & 1 & 0.015 & -0.182 & 0.057 \\ & 2 & 0.031 & -0.188 & 0.057 \\ \text{OI/E} & 1 & 0.000 & -0.021 & 0.029 & 0.0068 & 101.259 \\ & 2 & 0.023 & -0.068 & 101.259 \\ \text{OPEXP/S} & 1 & 0.001 & -0.029 & 0.0024 & -0.027 \\ & 2 & 0.006 & -0.024 & -0.027 \\ & 2 & 0.049 & 0.146 & 0.144 \\ \text{SGA/S} & 1 & 0.079 & 0.275 & 0.037 \\ \hline \hline \\ \hline \hline \\ \hline \hline \\ \hline \\ \hline \\ \hline \\ \hline \\ \hline $		2	0.087	0.204	0.387
Model R Square Change 2014 0.021 0.053 OI/S 1 0.015 -0.182 0.057 OI/E 1 0.000 -0.021 0.021 2 0.023 -0.068 101.259 OPEXP/S 1 0.001 -0.029 2 0.030 0.177 2 0.049 0.146 0.144 SGA/S 1 0.079 0.275 0.037 2 0.091 0.261 -0.037 2015 - -0.037 2015 - - ROA 1 0.040 0.291 2 0.062 0.248 0.511 ROS 1 0.077 0.750 2 0.062 0.248 0.511 ROS 1 0.003 0.209 2 0.018 0.127 0.132 OI/A 1 0.004 0.202 2 0.023	OI/A	1	0.002	0.055	
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$		2	0.034	0.021	0.053
Model 0.031 0.182 0.057 OI/E 1 0.000 -0.021 2 2 0.023 -0.068 101.259 OPEXP/S 1 0.001 -0.029 2 2 0.006 -0.024 -0.027 COG/S 1 0.030 0.177 2 2 0.049 0.146 0.144 SGA/S 1 0.079 0.275 2 2 0.091 0.261 -0.037 2015 Model R Square Change 2014 Financial Performance Dummy Variable ROA 1 0.040 0.291 2 0.062 0.248 0.511 ROS 1 0.077 0.750 2 0.062 0.248 0.511 ROS 1 0.003 0.209 2 0.018 0.127 0.132 OI/S 1 0.004 0.202 0.132 OI/S 1	OI/S	1	0.015	-0.182	01000
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Model R Square Change 2014 Financial Performance Dummy Variable ROA 1 0.040 0.291 2 0.062 0.248 0.511 ROS 1 0.077 0.750 2 0.094 0.662 0.726 OI/A 1 0.003 0.209 2 0.018 0.127 0.132 OI/S 1 0.004 0.202 2 0.023 0.148 0.195 OI/E 1 0.000 0.050 2 0.016 -0.008 260.108 OPEXP/S 1 0.008 0.274 2 0.025 0.246 -0.150	2015	2	0.071	0.201	0.037
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$\begin{array}{cccccccccccccccccccccccccccccccccccc$	ROA	1	0.040	0.291	
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$\begin{array}{cccccccccccccccccccccccccccccccccccc$	ROS	1	0.077	0.750	
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2 0.018 0.127 0.132 OI/S 1 0.004 0.202 2 0.023 0.148 0.195 OI/E 1 0.000 0.050 2 0.016 -0.008 260.108 OPEXP/S 1 0.008 0.274 2 0.025 0.246 -0.150	OI/A	1	0.003	0.209	
OI/S 1 0.004 0.202 2 0.023 0.148 0.195 OI/E 1 0.000 0.050 2 0.016 -0.008 260.108 OPEXP/S 1 0.008 0.274 2 0.025 0.246 -0.150		2	0.018	0.127	0.132
2 0.023 0.148 0.195 OI/E 1 0.000 0.050 2 2 0.016 -0.008 260.108 OPEXP/S 1 0.008 0.274 2 0.025 0.246 -0.150	OI/S	1	0.004	0.202	
OI/E 1 0.000 0.050 2 0.016 -0.008 260.108 OPEXP/S 1 0.008 0.274 2 0.025 0.246 -0.150		2	0.023	0.148	0.195
2 0.016 -0.008 260.108 OPEXP/S 1 0.008 0.274 2 2 0.025 0.246 -0.150	OI/E	1	0.000	0.050	
OPEXP/S 1 0.008 0.274 2 0.025 0.246 -0.150		2	0.016	-0.008	260.108
2 0.025 0.246 -0.150	OPEXP/S	1	0.008	0.274	
		2	0.025	0.246	-0.150

Information technology capability and firm performance in the era of Big Data analytics: Evidence from the United States

	Α	ccounting and Mana	gement Information Systems	
COG/S	1	0.007	0.124	
000/5	2	0.007	0.124	0.023
SGA/S	1	0.000	0.088	0.025
50705	2	0.037	0.060	-0.062
2016	2	0.037	0.001	0.002
2010	Model	R Square Change	2015 Financial Performance	<u>Dummy</u> Variable
ROA	1	0.191	0.496	
	2	0.199	0.478	0.328
ROS	1	0.453	1.997	
	2	0.453	2.003	-0.166
OI/A	1	0.064	0.055	
	2	0.079	0.052	0.029
OI/S	1	0.093	0.155	
	2	0.107	0.146	0.084
OI/E	1	0.094	0.046	
	2	0.119	0.043	50.482
OPEXP/S	1	0.062	0.154	
	2	0.070	0.146	-0.064
COG/S	1	0.076	0.187	
	2	0.093	0.184	0.129
SGA/S	1	0.212	1.014	
	2	0.215	0.990	-0.045
2017				
	<u>Model</u>	<u>R Square Change</u>	2016 Financial Performance	<u>Dummy</u> Variable
ROA	1	0.039	0.056	
	2	0.087	0.045	0.239
ROS	1	0.012	0.013	
	2	0.067	0.010	0.449
OI/A	1	0.094	0.216	
	2	0.101	0.207	0.013
OI/S	1	0.019	0.056	
	2	0.028	0.050	0.028
OI/E	1	0.010	0.053	
	2	0.033	0.037	26.670
OPEXP/S	1	0.014	0.050	
	2	0.016	0.048	-0.011
COG/S	1	0.049	0.202	
	2	0.057	0.191	0.082
SGA/S	1	0.026	0.071	
		0.020		

	Hypothesis	Results
H1	Firms that had superior IT capabilities in the 2010s are associated	Supported
	with higher average profit ratios than those of control firms.	
H2	Firms that had superior IT capabilities in the 2010s are associated	Partially
	with lower average cost ratios than those of control firms.	Supported
H3	Firms that had superior IT capabilities in the 2010s are associated	Supported
	with higher average profit ratios than those of control firms in all subsequent years.	
H4	Firms that had superior IT capabilities in the 2010s are associated	Partially
	with lower average cost ratios than those of control firms in all	Supported
	subsequent years.	
H5	Firms that had superior IT capabilities in the 2010s are associated	Not
	with higher average profit ratios than those of control firms after	Supported
	adjustment for prior financial performance.	
H6	Firms that had superior IT capabilities in the 2010s are associated	Not
	with lower average cost ratios than those of control firms after	Supported
	adjustment for prior financial performance.	
H7	After adjustment for previous financial performance, firms that	Not
	had superior IT capabilities in the 2010s are associated with	Supported
	higher average profit ratios than those of control firms in the	
	subsequent years.	
H8	After adjustment for previous financial performance, firms that	Not
	had superior IT capabilities in the 2010s are associated with lower	Supported
	average cost ratios than those of control firms in the subsequent	
	years.	

Table 8. Summary of hypotheses tests

5. Discussion and conclusions

5.1 Conclusions

Despite insufficient evidence on the subject, we are motivated by the growing worldwide attention to whether a firm's IT capability affects its financial performance in the era of BDA. We investigated this issue in the context of the United States by (1) selecting IT leaders and corresponding control groups from IW 500 and comparing their profit ratios and cost ratios from 2010 through 2017 and (2) adjusting for the financial halo effects to remove the effects from the previous years' financial performance on the following years' financial performance. The findings indicate that although IT leaders did show and sustain higher profit ratios than control groups before adjusting for the financial halo effects, such a relationship no longer exists after the adjustment. Since the 2010s, both academic and business communities have become increasingly interested in big data-enabled business value

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(McAfee & Brynjolfsson, 2012; Wamba *et al.* 2015), and the storage, computation, analysis, visualization, and integration of big data relies heavily on a firm's IT capability.

Therefore, this research serves as a future reference for the business and academic communities to consider a BDA strategy as part of IT capability to positively influence business performance. It augments Bharadwaj (2000), Santhanam and Hartono (2003), and Chae (2014), by taking the BDA factor into consideration with the latest data from 2010 through 2017. Nevertheless, the results of the study should be cautiously interpreted because of the existence of certain limitations and issues, such as the categorization of IT leaders and control groups that needs the attention of future researchers.

This research has opened a Pandora's box filled with questions worthy of investigation by future researchers. For example, alternate methodologies that are more robust and reliable should be developed to define firms with superior IT capabilities. Moreover, other valid variables and constructs that could possibly influence the link between IT capability and financial performance should be identified and incorporated in future studies. In addition, it could be insightful to examine the relationship at sublevels to consider the different organizational characteristics such as industry type and size. Future research is needed, but is not limited to the above perspectives, and we hope that this study can be an inspiration for future studies that elaborate its findings.

5.2 Reliability and validity

This study follows the methodology and instructions from a model paper published in *MIS Quarterly*, a leading journal of management information systems (Chae *et al.*, 2014). Although our data were extracted from different years, the results of the hypotheses tests are similar to those of the model paper. As discussed earlier, the model paper also found no link between IT capability and business performance after adjusting for previous years' firm performance. Therefore, this research is reliable. In terms of validity, the results of this study can only be generalized to US-listed firms. First, IW 500 only selects firms from the United States. In addition, 13 firms acknowledged as IT leaders by IW 500 were excluded because they were not listed companies, which resulted in their financial data being inaccessible and not included in this study. A further 207 firms' financial data were not fully available in Compustat's database (i.e., missing values in certain years for certain variables), and thus were not considered. The firms eliminated due to the reasons given above could have made a significant difference to this study's results, but it is not feasible to discuss their impacts under the current conditions.

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5.3 Limitations

Although our findings indicate that strong IT capability does not necessarily result in its superior and sustained business performance in the BDA era, it is premature to make such a claim. More research is needed to understand the mechanisms of IT capability and business performance in the BDA era, and we highlight the following limitations to help understand the results critically. To systematically assess the limitations of this research, first the sample selection should be investigated. As shown in Table 3, the average sales and assets of the IT leaders' group were almost twice as much as those of the control group. Therefore, firm size, dissimilar IT budgets, or even slight accounting maneuvers in depreciation and amortization might have affected their net income and therefore, their financial measurements (Chae et al., 2014). Second, errors could have crept in during the sample selection process. Although choosing samples from IW 500 has advantages such as quality respondents and a high response rate, IW 500's frequent and inevitable changes in the selection process have posed challenges for researchers who pay attention to the benchmarking criteria. In fact, since IW 500 published its first list in 1989, its selection criteria for inclusion have continuously changed and the process is not consistent (Chae et al., 2014). Moreover, the firm selection procedure adopts a binary measure that identifies firms as leaders or not leaders. Such a measurement method makes the evaluation of the influence of incremental improvements in IT capability on firm performance impossible (Santanam & Hartnono, 2003).

5.4 Theoretical contributions

This research adds value to two aspects of the IT literature. First, it attempts to further the understanding of how a firm's IT capability relates to its business performance by introducing the impact of BDA since the 2010s. Second, it extends previous studies with fresh data from the 2010s. The previous research that discussed the relationship between a firm's IT capability and its business performance, was conducted with data from the early 1990s and 2000s; it is doubtful that those findings still hold true after decades of dramatic change in the business use of technologies, especially the growing popularity and application of big data analytics as part of business strategy.

5.5 Discussion

This research attempts to answer a vital question: does the relationship between a firm's IT capability and its business performance still hold true after experiencing a large technology advancement such as BDA? Both positive links and no positive links between IT capability and organizational performance have been found in previous studies (Bharadwaj, 2000; Chae *et al.*, 2014; Santhanam & Hartono, 2003). However, as this past research uses data from the early 1990s and 2000s, updates on our understanding are necessary. Therefore, this study replicates, extends, and

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updates the three previous studies with the data from the 2010s to investigate how the advantage of being an IT leader has changed since the early 1990s. The significance of this method is that it demonstrates that IS researchers generally underutilize replications and updates, which are important elements of scientific studies (Berthon *et al.* 2002).

The expected results are that IT leaders from 2010 through 2013 were associated with higher profit ratios and that they also sustained this from 2014 through 2017. However, IT leaders did not necessarily achieve lower cost ratios from 2010 through 2017, which was unexpected. Moreover, after adjusting for the financial halo effect, which aimed to remove the effects of the previous years' financial performance in the following years, IT leaders were neither associated with superior business performance nor sustained superior business performance.

To analyze both the expected and unexpected results, it is necessary to understand the role of IT capability and BDA in business performance since the 2010s. As discussed, IT capability plays a significant and positive role in BDA adoption and application. Due to the hypercompetitive nature of modern business environments, firms 'attention has been drawn to utilize IT capability-enabled information to develop organizational capabilities (Chakravarty *et al.*, 2013). The application of BDA particularly plays an important role in assisting operational and strategic decision making, and thus enhances organizational performance (Kiron *et al.*, 2014). Insights derived from BDA can improve real-time business process monitoring and measurement, enhance quality control (Waller and Fawcett, 2013; Davenport *et al.*, 2012), reinforce customer relationships, manage operational risks, elevate operational efficiency and effectiveness, and facilitate product or service delivery (Kiron, 2013; Zelbst *et al.*, 2011). Therefore, the results of H1 and H3, which show that IT leaders have higher profit ratios and that such advantages are sustained, can be validated.

However, transforming the advantages of IT capability and BDA into business value is challenging. The challenges start from the senior level, where top-level decision makers need to embrace evidence-based decision making and encourage all organizational members to redefine its understanding of "judgment" (McAfee & Brynjolfsson, 2012). In addition, readiness factors and organizational design are critical factors for IT utilization and therefore, developing IT capability may not necessarily enhance business performance (Hong & Kim, 2002; Dezdar & Sulaiman, 2010). To sum up, utilizing BDA (such as data sourcing, access, integration, delivery, advanced analytics capacity and capabilities, and human resources) and organizational factors (in terms of top management support, BDA strategy, financial resources, and people engagement) can both inhibit and facilitate the effective utilization of BDA in operations. Thus, the benefits of BDA utilization do not translate into business value quickly and easily (Popovič *et al.*, 2014). This study speculates that the high costs associated with BDA adoption and

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application are also reflected in a firm's cost ratios, which explains the results of H2 and H4.

None of the hypotheses from H5 through H8 are supported, which is interesting because the regression conducted in this study can be regarded as more conservative and rigorous than the pair-wise comparison, as the Wilcoxon signed-rank test (Santhanam & Hartono, 2003) was adopted. Although the results are surprising, similar patterns were also found in previous studies (Chae *et al.*, 2014; Santhanam & Hartono, 2003). These researchers also found that the significant effect of IT capability on organizational performance presented via a pair-wise comparison became less apparent when it was measured using regression analysis.

References

- Agarwal, R. and Dhar, V. (2014) Big data, data science, and analytics: The opportunity and challenge for IS research. *Information Systems Research*, vol. 25: 443-448
- Agrawal, K.P. (2015) "Investigating the determinants of Big Data Analytics (BDA) adoption in emerging economies", *Academy of Management Proceedings*, vol. 2015, no. 1: 11290
- Akter, S. & Wamba, S.F. (2016) "Big data analytics in E-commerce: a systematic review and agenda for future research", *Electronic Markets*, vol. 26, no. 2: 173-194.
- Berthon, P., Pitt, L., Ewing, M. & Carr, C.L. (2002) "Potential research space in MIS: A framework for envisioning and evaluating research replication, extension, and generation", *Information Systems Research*, Vol. 13, no.4: 416-427.
- Bharadwaj, A.S. (2000) "A resource-based perspective on information technology capability and firm performance: an empirical investigation", *MIS Quarterly*, vol. 24, no. 1:169-196.
- Brown, B. & Perry, S. (1994) "Removing the financial performance halo from Fortune's "most admired" companies", *Academy of Management Journal*, vol. 37, no. 5: 1347-1359.
- Carr, N. (2003) "IT doesn't matter", Educause Review, vol. 38: 24-38
- Chae, H.C., Koh, C.E. & Prybutok, V.R. (2014) "Information technology capability and firm performance: contradictory findings and their possible causes", *MIS Quarterly*, vol. 38, no.1: 305-326
- Chakravarty, A., Grewal, R. & Sambamurthy, V. (2013) "Information technology competencies, organizational agility, and firm performance: Enabling and facilitating roles", *Information Systems Research*, vol. 24, no. 4: 976-997
- Chan, Y.E. (2000) "IT value: The great divide between qualitative and quantitative and individual and organizational measures", *Journal of Management Information Systems*, vol. 26, no.4: 225-261

- Chullen, A., Kaltenbrunner, H. & Schwetzler, B. (2015) "Does consistency improve accuracy in multiple – based valuation", *Journal of Business Economics*, vol. 85, no. 6: 635-662
- Clemons, E.K. & Row, M.C. (1991) "Sustaining IT advantage: The role of structural differences", *MIS Quarterly*, vol. 15, no. 3: 275-292
- Clemons, E.K. (1986) "Information systems for sustainable competitive advantage", *Information & Management*, vol. 11, no.3: 131-136
- Davenport, T.H. (2006) "Competing on analytics", *Harvard Business Review*, vol. 84, no. 1: 98
- Davenport, T.H., Barth, P. & Bean, R. (2012) "How 'big data' is different", *MIT* Sloan Management Review, vol. 54, no. 1: 22-24
- Davern, M.J. & Kauffman, R.J. (2000) "Discovering potential and realizing value from information technology investments", *Journal of Management Information Systems*, vol. 16, no.4: 121-143
- Dehning, B. & Richardson, V.J. (2002) "Returns on investments in information technology: A research synthesis", *Journal of Information Systems*, vol. 16, no. 1: 7-30
- Dezdar, S. & Ainin, S. (2011) "The influence of organizational factors on successful ERP implementation", *Management Decision*, vol. 49, no. 6: 911-926
- Enache, L. & Anup, S. (2018) "Should intangible investments be reported separately or commingled with operating expenses? New evidence", *Management Science*, vol. 64, no. 7: 3446-3468
- Fahy, J. & Hooley, G. (2002) "Sustainable competitive advantage in electronic business: towards a contingency perspective on the resource-based view", *Journal of Strategic Marketing*, vol. 10, no. 4: 241-253
- Gandomi, A. & Haider, M. (2015) "Beyond the hype: Big data concepts, methods, and analytics", *International Journal of Information Management*, vol. 35, no. 2: 137-144
- Grant, R.M. (1991) "The resource-based theory of competitive advantage: implications for strategy formulation", *California Management Review*, vol. 33, no. 3: 114-135
- Grover, V., Chiang, R.H., Liang, T.P. & Zhang, D. (2018) "Creating strategic business value from big data analytics: A research framework", *Journal of Management Information Systems*, vol. 35, no.2: 388-423
- Hitt, L.M. & Brynjolfsson, E. (1996) "Productivity, business profitability, and consumer surplus: three different measures of information technology value", *MIS Quarterly*, vol. 20, no. 2: 121-142
- Hong, K.K. & Kim, Y.G. (2002) "The critical success factors for ERP implementation: an organizational fit perspective", *Information & Management*, vol. 40, no. 1: 25-40
- Iqbal, R., Doctor, F., More, B., Mahmud, S. & Yousuf, U. (2020) "Big data analytics: computational intelligence techniques and application areas", *Technological Forecasting and Social Change*, vol. 153: 119-253

- Kanodia, C., and Haresh, S. (2016) "A real effects perspective to accounting measurement and disclosure: Implications and insights for future research", *Journal of Accounting* Research, vol. 54, no. 2: 623-676
- Kiron, D., Prentice, P.K. & Ferguson, R.B. (2014) "Raising the bar with analytics", *MIT Sloan Management Review*, vol. 55, no. 2: 29
- Kohli, R. & Devaraj, S. (2003) "Measuring information technology payoff: A meta-analysis of structural variables in firm-level empirical research", *Information Systems Research*, vol. 14, no. 2: 127-145
- LaValle, S., Lesser, E., Shockley, R., Hopkins, M.S. & Kruschwitz, N. (2011) "Big data, analytics and the path from insights to value", *MIT Sloan Management Review*, vol. 52, no. 2: 21-32
- Mahmood, M.A., Mann, G.J. & Zwass, V. (2000) "Impacts of information technology investment on organizational performance", *Journal of Management Information Systems*, vol. 16, no. 4: 3-10
- Marr, B., (2016) Big data in practice: How 45 successful companies used big data analytics to deliver extraordinary results, John Wiley & Sons.
- Mata, F.J., Fuerst, W.L. & Barney, J.B. (1995) "Information technology and sustained competitive advantage: A resource-based analysis", *MIS Quarterly*, vol. 19, no. 4: 487-505
- McAfee, A., Brynjolfsson, E., Davenport, T.H., Patil, D.J. & Barton, D. (2012) "Big data: the management revolution", *Harvard Business Review*, vol. 90, no. 10: 60-68
- Melville, N., Kraemer, K. & Gurbaxani, V. (2004) "Information technology and organizational performance: An integrative model of IT business value", *MIS Quarterly*, vol. 28, no. 2: 283-322
- Mikalef, P., Pappas, I.O., Krogstie, J. & Giannakos, M. (2018) "Big data analytics capabilities: a systematic literature review and research agenda", *Information Systems and e-Business Management*, vol. 16, no. 3: 547-578
- Mithas, S., Ramasubbu, N. & Sambamurthy, V. (2011) "How information management capability influences firm performance", *MIS Quarterly*, vol. 35, no. 1: 237
- Popovič, A., Hackney, R., Coelho, P.S. & Jaklič, J. (2012) "Towards business intelligence systems success: Effects of maturity and culture on analytical decision making", *Decision Support Systems*, vol. 54, no. 1: 729-739
- Popovič, A., Hackney, R., Coelho, P.S. and Jaklič, J. (2014) "How information-sharing values influence the use of information systems: An investigation in the business intelligence systems context", *The Journal of Strategic Information Systems*, vol. 23, no. 4: 270-283
- Popovič, A., Hackney, R., Tassabehji, R. & Castelli, M. (2018) "The impact of big data analytics on firms' high value business performance", *Information Systems Frontiers*, vol. 20, no. 2: 209-222
- Powell, T.C. & Dent-Micallef, A. (1997) "Information technology as competitive advantage: The role of human, business, and technology resources", *Strategic Management Journal*, vol. 18, no. 5: 375-405

- Ray, G., Muhanna, W.A. & Barney, J.B. (2005) "Information technology and the performance of the customer service process: A resource-based analysis", *MIS Quarterly*, vol. 29, no.4: 625-652
- Salehan, M. & Kim, D.J. (2016) "Predicting the performance of online consumer reviews: A sentiment mining approach to big data analytics", *Decision Support Systems*, vol. 81, no. C: 30-40
- Santhanam, R. & Hartono, E. (2003) "Issues in linking information technology capability to firm performance", *MIS Quarterly*, vo. 27, no. 1: 125-153
 Singh, S.K. & El-Kassar, A.N. (2019) "Role of big data analytics in developing
- Singh, S.K. & El-Kassar, A.N. (2019) "Role of big data analytics in developing sustainable capabilities", *Journal of Cleaner Production*, vol. 213: 1264-1273
- Stoel, M. D. &Muhanna W. A. (2009) "IT capabilities and firm performance: A contingency analysis of the role of industry and IT capability type", *Information & Management*, vol. 46, no. 3:181
- Straub, D.W. & Watson, R.T. (2001) "Research commentary: Transformational issues in researching IS and net-enabled organizations", *Information Systems Research*, vol. 12, no. 4: 337-345
- Wade, M. & Hulland, J. (2004) "The resource-based view and information systems research: Review, extension, and suggestions for future research", MIS Quarterly, vol. 28, no.1: 107-142
- Waller, M.A. & Fawcett, S.E. (2013) "Data science, predictive analytics, and big data: a revolution that will transform supply chain design and management", *Journal of Business Logistics*, vol. 34, no. 2: 77-84
- Wamba, S.F., Akter, S., Edwards, A., Chopin, G. & Gnanzou, D. (2015) "How 'big data' can make big impact: Findings from a systematic review and a longitudinal case study", *International Journal of Production Economics*, vol. 165, no. C: 234-246
- Wamba, S.F., Gunasekaran, A., Akter, S., Ren, S.J.F., Dubey, R. & Childe, S.J. (2017) "Big data analytics and firm performance: Effects of dynamic capabilities", *Journal of Business Research*, vol. 70: 356-365.
- Wang, P. (2007) "Chasing the hottest IT: Effects of information technology fashion on organizations", MIS Quarterly, vol. 34, no. 1: 1-7
- White, M. (2012) "Digital workplaces: Vision and reality", *Business Information Review*, vol. 29, no. 4: 205-214.
- Wixom, B.H. & Todd, P.A. (2005) "A theoretical integration of user satisfaction and technology acceptance", *Information Systems Research*, vol. 16, no. 1: 85-102
- Wüstemann, Jens, & Wüstemann, S. (2010) "Why consistency of accounting standards matters: A contribution to the rules-versus-principles debate in financial reporting", *Abacus*, vol. 46, no. 1: 1-27
- Zhu, K., Kraemer, K.L. & Xu, S. (2006) "The process of innovation assimilation by firms in different countries: a technology diffusion perspective on e-business", *Management Science*, vol. 52, no. 10: 1557-1576