

Untangling operational performance implication of ambidextrous blockchain initiatives: An empirical investigation of Chinese manufacturers

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<u>manufacturers</u>

Abstract

Purpose - Blockchain has been considered as a disruptive technology and every industry stands to benefit from it. According to Deloitte, increasing number of managers claim that their firms will suffer loss in competitive advantage if they don't adopt blockchain. Despite this motivation, insufficient knowledge on how can blockchain add value inhibits the commitment to its adoption. To better understand the possible influence of blockchain, two dimensions (exploitation and exploration) of blockchain initiatives are distinguished. This research aims to examine the impact of ambidextrous blockchain initiatives on firms' operational efficiency, as well as the moderating role of unabsorbed slack and environmental munificence.

Design/methodology/approach – In this paper, secondary panel data were collected from multiple sources to test the proposed hypotheses. Content analysis was adopted to operationalise ambidextrous blockchain initiatives from firms' official announcements.

Findings - We find that firms' ambidexterity in blockchain initiatives have significantly positive impact on operational efficiency. Furthermore, when firms have lower unabsorbed slack or in more munificent environments, they may benefit from managing a balance between exploration and exploitation.

Originality/value – This research offers some insightful theoretical and managerial implications of ambidextrous blockchain initiatives. The findings of this study contribute to blockchain literature by providing theory-driven empirical research.

Keywords: blockchain; organisational ambidexterity; operational efficiency; unabsorbed slack; environmental munificence

1. Introduction

Blockchain – defined as 'a decentralized, distributed, shared, and immutable database ledger that stores registry of assets and transactions across a peer-to-peer network' (Khan and Salah, 2018, p.405) – is listed as one of the top 10 strategic technology trends, with the potential to provide significant business opportunity (Cearley and Burke, 2018). Manufacturing and operations management, in particular, are perceived as the most promising domains for adopting blockchain (Lohmer and Lasch, 2020). According to a report published by PwC (2019), blockchain is potential to revolutionise manufacturing in designing, engineering, making and scaling products. Realising such potentials, manufacturing firms are also being out front in developing blockchain (PwC, 2019). Growing interest has been shown in using blockchain to address operational challenges (Jabbar and Dani, 2020). Blockchain is anticipated to address the long-standing pain points in manufacturing industry such as materials provenance, counterfeit detection, engineering design for complex products and so on. For example, for aircraft manufacturers, blockchain can continuously update the condition of each part, which lead to shortened time on routine inspection and aircraft maintenance (PwC, 2019). More generally, blockchain is expected to benefit operations in many ways, including reducing costs, enhancing productivity and alleviating papertrail inefficiencies.

Despite blockchain research continuously grow, the extant literature remains largely technical or conceptual, and has mostly been composed of theoretical expositions, framework proposals and case studies and shows a relative lack of

quantitative empirical studies (Karakas *et al.*, 2021; Queiroz *et al.*, 2019). Since the application of blockchain is still at a rapid growth period, blockchain's capacity to help firms with value creation needs to be examined more closely. And according to Renee Ure, Lenovo's Chief Operating Officer, operational efficiencies can be seen through the use of blockchain.

In the light of this, this study aims to investigate the impact of blockchain initiatives, which refers to firms' use of blockchain, on firms' operational efficiency. Operational efficiency refers to a firm's managerial competence to transform diverse inputs into value-added outputs in the process of production, which is crucial to firms' success and must be taken seriously (Li et al., 2010). Hasan et al. (2020) pioneered in addressing this issue by directly investigating the association between the adoption of blockchain and operational efficiency. However, due to the fact that blockchain is one of the primary information technology innovations (Clohessy and Acton, 2019), blockchain initiatives can manifest themselves as varied innovation activities, such as incremental innovation in daily operations and business processes, or more radical in product and service. As suggested by ambidexterity theory, different patterns of innovation, i.e. exploitation and exploration, should be considered. On the one hand, firms can adopt existing blockchain services or products to 'enhance, optimize, secure and streamline many existing business and industrial processes' (Al-Jaroodi and Mohamed, 2019, p.36500), which represents an exploitative orientation. On the other hand, they can also develop their own blockchain-based R&D projects to create new business opportunities, processes and product development (Benzidia et al., 2021),

which represents an exploratory orientation. However, there is still a lack of understanding on the business value associated with different patterns of blockchain initiatives. As noted by Queiroz and Wamba (2019), the inadequate understanding of blockchain's value will be the important barrier for firms to adopt blockchain. In addition, technology exploration and exploitation could cross-fertilise each other (Geerts *et al.*, 2018). The separation investigation of the two activities could jeopardise the understanding of blockchain's effect. To address this concern, this research takes a further step by distinguishing blockchain initiatives into two categories: exploitative and exploratory, and employs organisational ambidexterity theory as the research lens.

Following organisational ambidexterity theory, we focus on the influence of ambidextrous blockchain initiatives, which is a metaphor defined as a firm's ability to successfully pursue both exploitative blockchain initiatives (EtBI) and exploratory blockchain initiatives (ErBI). We distinguish EtBI and ErBI by adapting the definitions of exploitation and exploration. In the context of strategic alliances, exploitation refers to a situation wherein a firm can achieve incremental improvements in its current products and processes by leveraging its partners' existing specialised knowledge (Hoang and Rothaermel, 2010). Exploration, in contrast, involves the discovery of something new, such as the development of a new product. Thus, we propose that EtBI refers to the utilisation of existing blockchain knowledge, competencies and opportunities through applying existing blockchain services or products. In contrast, ErBI mainly focuses on the creation of new blockchain knowledge, competencies and opportunities through the development of blockchain knowledge, projects.

And in line with the literature on organisational ambidexterity (He and Wong, 2004), a firm can achieve an ambidextrous blockchain initiative in two ways. Accordingly, a complementary blockchain initiative occurs when a firm scores high on both EtBI and ErBI. It corresponds to the type of strategy fit – 'fit as moderating', which implies that EtBI and ErBI can mutually interact. In addition, a balanced blockchain initiative occurs when a firm places relatively equal emphasis on the two dimensions, and it harmonises another type of strategy fit – 'fit as matching', indicating the case of matched exploration and exploitation.

Moreover, to better explain the value creation mechanism of blockchain, contextual factors which condition the actual business value of blockchain initiatives should be investigated. As the least-absorbed organisational resource, unabsorbed slack, which can help firms to achieve their development goals, plays a particularly important role in firms' growth (Wang *et al.*, 2021). In addition, firms need to take their industry environment into account, as underlying performance among firms is likely to vary under different industry conditions. With regard to this concern, environmental munificence is quite crucial as it defines firms' growth possibilities (Dess and Beard, 1984). Particularly, unabsorbed slack and environmental munificence also exemplify the resource accessible to a firm from inside and outside the organisation, respectively. And prior research has also highlighted that a firm's ability to benefit from ambidexterity critically depends on its available resources (Cao *et al.*, 2009). However, it also remains unclear how resource-relevant factors can hamper or promote the value associated with blockchain initiatives. Thus, this research attempts to fill this gap by quantitatively investigating the impact of blockchain initiatives on firms' operational efficiency – as well as the moderation effect of two contingencies – from the perspective of organisational ambidexterity. We propose the following two research questions:

- RQ1: Can the implementation of ambidextrous blockchain initiatives promote firms' operational efficiency?
- RQ2: How do organisational and environmental contingencies-namely, unabsorbed slack (internal resources) and environmental munificence (external resources)-influence the relationship between ambidextrous blockchain initiatives and operational efficiency?

To answer these questions, we collected secondary data from firms' official announcements and the China Stock Market & Accounting Research Database (CSMAR). We select China market as the main data source since China is a key player in the development of blockchain in the world (Kuo and Shyu, 2021). Thus, studying the phenomenon of Chinese firms provides certain representative significance (Chen *et al.*, 2021).

Our research makes several significant contributions. First, since the power of blockchain has drawn increasing attention from academia and more theory-driven and rigorously empirical studies are called for (Rossi *et al.*, 2019), this study enriches the blockchain literature by providing empirical evidence on the positive effect of blockchain initiatives. Second, differing from prior studies which explore the influence of adopting blockchain (e.g. Chen *et al.*, 2021; Hasan *et al.*, 2020; Pan *et al.*, 2020), to

our best knowledge, this research is the first attempt to apply the idea of organisational ambidexterity to distinguish among different types of blockchain implementations. Third, the inclusion of two moderators offers a more comprehensive understanding on the effect of contingencies in blockchain value creation. Finally, from a managerial perspective, this research sheds light on the importance of adopting blockchain–specifically, implementing blockchain in an ambidextrous manner – and provides insights for firms regarding how to derive better outcomes given specific internal and external resource conditions.

The remainder of this research is structured as follows. Section 2 outlines the relevant literature and develops research hypotheses. Section 3 describes the adopted methodology. The empirical results are presented in Section 4, and Section 5 offers a discussion.

2. Literature review and hypothesis development

2.1 Organisational ambidexterity

In brief, exploitation includes 'refinement, choice, production, efficiency, selection, implementation, and execution', while exploration is characterised by 'search, variation, risk taking, experimentation, play, flexibility, discovery, and innovation' (March, 1991, p.71). The concepts of exploitation and exploration have been applied and expanded in various contexts, including technology and innovation management (e.g. Gu *et al.*, 2021; Tai *et al.*, 2019)

Although exploration and exploitation have different underlying logics and tensions arise as they compete for scarce resources, they are also synergistic (He and Wong, 2004). Exclusive engagement in exploration will cause a firm to suffer from a lack of gains obtained through existing knowledge, whereas at the other extreme, exclusive involvement in exploitation will trap a firm in obsolescence (Levinthal and March, 1993). Thus, the optimal solution is an ambidextrous organisation that can simultaneously pursue exploitation and exploration (Tushman and O'Reilly, 1996). A fruitful stream of research offers insights and empirical evidence regarding the significant role of organisational ambidexterity in firms' performance. For example, analysing a sample of 206 manufacturing firms, He and Wong (2004) found that both complementary and balanced innovation strategies were positively related to firms' sales growth rates.

2.2 Ambidextrous blockchain initiatives and firms' operational efficiency

With respect to EtBI, manufacturing firms can directly leverage current blockchain services or products in the market, such as adopting or connecting with established blockchain platforms to integrate inventory, capital and information flows across companies. In this way, efficiency improvements can be gained through speeding up organisational processes, reducing operational costs and improving operating quality (Pan *et al.*, 2020). ErBI mainly manifest as innovative processes or products. Innovated processes can enhance efficiency by helping firms reduce process costs and time, and blockchain-driven ones have similar effects. For innovation in products, blockchain can

be added as one key element of a new product, achieving product upgrading and differentiation, resulting in higher operational efficiency. However, implementing solely EtBI or ErBI may not achieve a favourable outcome. And the concept of organisational ambidexterity proposes that ambidextrous organisations which synchronously develop their exploration and exploitation capacities are optimal (March, 1991).

On the one hand, organisational ambidexterity theory supports that exploitation and exploration can complementarily interact with each other (Tushman and O'Reilly, 1996; Wong et al., 2013). Accordingly, EtBI and ErBI can add value to each other and concurrent pursuit of them can lead to better results than implementing them separately. EtBI can stimulate ErBI, reducing the amount of time needed to develop blockchain R&D projects and increasing their success rate, and ErBI can expand the available blockchain-relevant resources for EtBI. Firms that successfully achieve a dual orientation of exploration and exploitation are more attuned to important opportunities to promote operational efficiency (Gedajlovic *et al.*, 2012). On the other hand, overemphasising one dimension at the expense of the other can be risky. In terms of McDermott and Prajogo (2012), firms with an extreme focus on EtBI could drive them into a 'success trap', ignoring the potential benefits derived from ErBI. And an overreliance on EtBI indicates overdependence on external blockchain expertise and partners, which may lead to a loss of certain flexibility in decision-making, resulting in unnecessary additional management costs and redundant management procedures. Alternatively, firms that overemphasise ErBI are in intensive resource-consuming yet

without immediate reward in sight, resulting a 'failure trap'. According to Melander (2018), to succeed in ErBI R&D projects, it is critical to have a supporting level of EtBI. Otherwise, it can be resource inefficient as the development cycle will be longer and the failure rate higher. Prior studies have also supported the efficiency-related synergies of ambidexterity (see Kortmann *et al.*, 2014). Based on the above discussion, we present the following hypothesis:

H1: Complementary blockchain initiatives are positively related to firms' operational efficiency.

H2: Balanced blockchain initiatives are positively related to firms' operational efficiency.

2.3 Organisational and environmental contingencies

Unabsorbed slack. Slack is defined as the pool of resources within an organisation and denotes an excess of the minimum level of resources needed for ordinary operations (Nohria and Gulati, 1996). Accordingly, unabsorbed slack, which is easy to reconfigure for various uses and assign to any purpose, refers to excess, uncommitted, available and ready-to-deploy liquid resources in organisations, such as cash and raw material inventory

Ample resources relax the tension between exploitation and exploration (Jin *et al.*, 2015) and resource availability is also a key constraint when using blockchain (Al-Jaroodi and Mohamed, 2019). For instance, Nandi *et al.* (2021) reasoned that firms with greater financial and slack capacity were more capable of building blockchain-

enabled circular economy system. Since EtBI focuses on utilising existing blockchain knowledge, whereas ErBI emphasises the creation of new blockchain knowledge, simultaneous development of EtBI and ErBI is complex, as it requires fast absorption and coordination of diverse knowledge. Therefore, substantive resources need to be committed. Unabsorbed slack offers such uncommitted resources to facilitate complementary blockchain initiatives by flexibly and timely allocating resources whenever and wherever needed. As confirmed by Cao *et al.* (2009), the positive effect of combined ambidexterity is highly related to the availability and allocation of resources. Thus, we posit:

H3a: Unabsorbed slack positively moderates the relationship between complementary blockchain initiatives and firms' operational efficiency.

The loss in efficiency from overreliance on EtBI can be attributed to inflexibility in adjusting strategies and adapting to changing environments. However, organisational slack can promote advanced strategic initiatives, enabling firms to better adapt to complex and competitive situations (Jin *et al.*, 2015). On the other hand, exclusive engagement in ErBI is risky due to long R&D cycles and higher failure rates and is also inefficient in resource distribution. Nevertheless, slack can alleviate such risks, as it can facilitate search, experimentation and innovation and even avoid some serious consequences of failure (Lavie *et al.*, 2010). Consistent with Cao *et al.* (2009), we argue that, the risk of unbalanced EtBI and ErBI is less threatening when a firm has a larger resource base, as those resources can cushion such risks. Therefore, we posit:

 H3b: Unabsorbed slack negatively moderates the relationship between balanced blockchain initiatives and firms' operational efficiency.

Environmental munificence. Munificent environments are characterised by low-level external threats, higher growth rates and abundant environmental resources (Bloom and Michel, 2002), whereas declining demand, fierce competition and high financial pressure depict the situation of less munificent environments (Choi *et al.*, 2020).

According to Heeley *et al.* (2006), firms in more munificent settings could realise greater returns from a given stock of resources devoted to blockchain initiatives. Consistent with this, as noted by Vitari and Raguseo (2020), munificent environment could expand the potential of complementary blockchain initiatives in promoting operational efficiency when such initiatives match the requirements of external environment. For example, adopting blockchain to store customer data can facilitate manufacturers' capacities to better manage their production. Such reductions in operational costs and cycle time are likely to be more pronounced in munificent industries with growing demand. On the contrary, firms in high-pressure environments may avoid implementing either EtBI or ErBI because, in such conditions, firms prefer to conserve resources. We hypothesise the following relationship:

H4a: Environmental munificence positively moderates the relationship between complementary blockchain initiatives and firms' operational efficiency.

Munificent environments provide firms with greater chances of survival, more opportunities to grow and more strategic choices (Li *et al.*, 2013). In such

circumstances, firms can fully leverage the opportunities to balance EtBI with ErBI, resulting in the greatest efficiency. Conversely, as suggested by Moss *et al.* (2014), firms in more hostile environments will be more likely to flexibly change their strategic emphasis on EtBI or ErBI. Deploying ambidexterity can be dangerous for firms with low environmental munificence; in such conditions, firms are recommended to emphasise exploitation to better pursue efficiency (Dolz *et al.*, 2019). A struggle to maintain balanced blockchain initiatives may cause unexpected waste in resources and in turn lead to loss of efficiency. Thus, we present:

H4b: Environmental munificence positively moderates the relationship between balanced blockchain initiatives and firms' operational efficiency.

The conceptual framework of this study is shown in Figure 1.

[Insert Figure 1 here]

3. Methods

3.1 Data collection

Secondary panel data were collected from multiple sources to test the proposed hypotheses. We derived information regarding firms' blockchain initiatives from firms' official announcements, which are more authoritative than newspaper reports (Liu *et al.*, 2020). Specifically, we searched for firms' official announcements from Wind Economic Database ('Wind'), as Wind contains all official announcements of listed firms. The data collection process is shown in Figure 2. The sample was limited to publicly listed manufacturing firms in China. First, we obtained a list of Chinese

manufacturing firms from CSMAR, a leading data source for Chinese listed firms. We followed the industrial classification standard proposed by China Securities Regulatory Commission (CSRC), which resulted in 2,860 listed firms (two-digit CSRC codes C13-C43). Second, we manually searched Wind to determine whether each firm had released any official announcements with the keyword 'blockchain'. In this step, 1,490 blockchain-relevant official announcements from 302 firms were found. Lastly, by reading through all the announcements, we identified firms' blockchain initiatives. Similar to the filtering criteria in Xiong et al. (2021), who identified blockchain adoption in supply chains, we first eliminated announcements not related to firms' blockchain initiatives, such as the appointments of executives with blockchain background. Then, since same blockchain initiatives could be identified in different announcements, repeated records were dropped to avoid duplicates. Finally, we coded 141 firms with at least one blockchain initiative between 2015 and 2020. The reason for exploring the influence of blockchain initiatives between 2015 and 2020 is that in China, blockchain did advance in this period. The starting year of 2015 is chosen as this year has seen the turning point for firm digitisation under the encouragement of the government (Chen et al., 2021).

[Insert Figure 2 here]

Descriptive statistics of the 141 firms are presented in Table 1. In particular, Panel A shows that the number of blockchain initiatives has continuously increased in recent years. The majority (85.686%) were launched in 2018, 2019 and 2020. These 141 manufacturing firms represent 21 industries based on two-digit CSRC codes,

indicating that the firms are from a wide range of industries. Panel B of Table 1 shows the top 10 industries. The top five industries take 62.411% of the total industries, and interestingly, four of which are associated with equipment manufacturing. In addition, organisational characteristics in terms of sales, total assets, number of employees and age are provided in Panel C of Table 1.

[Insert Table 1 here]

Next, for other research variables, operational data were collected from CSMAR. Due to missing data, not all firms had the full six-year consecutive data. Therefore, the final sample consisted of 395 firm–year observations, which included data from 141 firms between 2015 and 2020. The dataset was an unbalanced panel data.

3.2 Variable measurements

The details of the measurement procedures are summarised in Table 2.

[Insert Table 2 here]

Ambidextrous blockchain initiatives. Scholars have developed a variety of approaches to operationalise exploration and exploitation, such as survey-based analyse (e.g. Gu *et al.*, 2021; He and Wong, 2004), and using proxies such as patent data (see Kang and Kim, 2020). These approaches have been challenged for lacking generalisability and applicability, as well as lack of clarity regarding whether they are consistent with the concept of exploration and exploitation (Uotila *et al.*, 2009). To address such concerns, this study employed content analysis for three reasons: 1) we can directly derive exploration and exploitation from official announcements; 2) using

content analysis allowed us to operationalise data over an extended time period; 3) content analysis can ensure that research findings are applicable across different industries (Gatti *et al.*, 2015; Uotila *et al.*, 2009).

We classified EtBI and ErBI based on aforementioned definition. Some examples are shown in Table 3. Referencing Lam *et al.* (2016), we counted the numbers of the two types of initiatives to quantify EtBI and ErBI. To ensure the coding reliability of content analysis, the inter-coder reliability test was conducted. Considering that the collected announcements contain large texts needed to be coded, following Tangpong (2011), a subsample of 10% of the total text units was randomly selected and coded by another independent and well-trained coder. The results indicated adequate levels of inter-coder reliability for both EtBI (Cohen's kappa=0.845) and ErBI (Cohen's kappa=0.810).

[Insert Table 3 here]

Complementary blockchain initiative was measured as the interactive term: EtBI \times ErBI (Wong *et al.*, 2013). Before generating this product, EtBI and ErBI were mean centred to avoid potential multicollinearity. Balanced blockchain initiatives was measured as the absolute difference between exploration and exploitation, which was reversed to facilitate interpretation: 3-[EtBI – ErBI].

Operational efficiency. Stochastic frontier estimation (SFE) was adopted to measure operational efficiency by modelling a firm's relative efficiency in converting operational input resources into operational output (Li *et al.*, 2010). SFE provides a comprehensive measurement, as it considers various operational inputs rather than a

single indicator; it also better fits the definition of operational efficiency from the perspective of the OM domain (Lam *et al.*, 2016). Taking an SFE approach, the efficiency score was calculated as follows:

$$\ln(Operating \ Income)_{it} = \beta_0 + \beta_1 \ln(Number \ of \ Employees)_{it} + \beta_2 \ln(Cost \ of \ Goods \ Sold)_{it} + \beta_3 \ln(Capital \ Expenditure)_{it} + \varepsilon_{it} - \eta_{it}$$
(1)

where ε_{it} represents the purely stochastic random error term; η_{it} is the technical inefficiency. Thus, the operational efficiency of firm *i* in year *t* can be expressed as follows (Zhu *et al.*, 2021):

$$Operational \ efficiency_{it} = e^{-\eta_{it}}$$
(2)

The efficiency measurement ranges from 0 to 1; 0 denotes the lowest level of efficiency and 1 the optimal level in the conversion process.

Unabsorbed slack and environmental munificence. Consistent with Iyer and Miller (2008), we used the current ratio (i.e. current assets divided by current liabilities) as a proxy for unabsorbed slack. Following Goll and Rasheed (2005), we adopted 10-year average sales growth as the measurement for environmental munificence. For each industry and sample year, the industry-level total sales for the previous 10 years were regressed on time. Munificence was measured as the regression slope coefficient divided by the mean sales in the same time period (see Jacobs *et al.*, 2015).

Control variables. Four control variables – firm size, firm age, firm profitability and firm cash-to-cash cycle – were included because they might influence operational

efficiency. Firm size was measured as the natural logarithm of a firm's sales (Lee and Huang, 2012). Firm profitability was operationalised as a firm's return on assets (Fong *et al.*, 2010). Firm age was defined as years since the firm's incorporation (Kotha *et al.*, 2011). In line with Lam (2018), firm cash-to-cash cycle was calculated as in Equation (3) and standardised based on its industry. Finally, year dummies were also controlled in consideration of unobserved time-specific effects.

Cash to cash $cycle_{it} = Days \ of \ inventory_{it} + Days \ of \ accounts \ receivables_{it}$ - Days of $accounts \ payables_{it}$ (3)

3.3 Research modelling

The sample data represent a short unbalanced panel. In such cases, pooled regression is applicable and more likely to provide a reliable result (Ben-Jebara and Modi, 2021). Following He and Wong (2004), we analysed the effects of alternate ambidextrous blockchain initiatives in separate models, where only one of the two dimensions was included, as shown in Equation (4):

 $Operational \ efficiency_{i(t+1)} = \beta_0 + \beta_1 Ambidextrous \ blockchain \ initiatives_{it}$

 $+ \beta_{2} Exploitative blockchain initiatives_{it}$ $+ \beta_{3} Exploratory blockchain initiatives_{it}$ $+ \beta_{4} Firm age_{it} + \beta_{5} Firm size_{it} + \beta_{6} Firm profitability_{it}$ $+ \beta_{7} Firm cash - to - cash cycle_{it} + Year dummies + \varepsilon_{it}$ (4)

To test the moderating effect, we then added the two moderators and their interactions with ambidextrous blockchain initiatives in Equation (5) as follows:

 $Operational \; efficiency_{i(t+1)} = eta_0 + eta_1 Ambidextrous \; blockchain \; initiatives_{it}$

 $+ \beta_{2} Moderator$ $+ \beta_{3} Ambidextrous blockchain initiatives \times Moderator$ $+ \beta_{4} Exploitative blockchain initiatives_{it} (5)$ $+ \beta_{5} Exploratory blockchain initiatives_{it}$ $+ \beta_{6} Firm age_{it} + \beta_{7} Firm size_{it} + \beta_{8} Firm profitability_{it}$ $+ \beta_{9} Firm cash - to - cash cycle_{it} + Year dummies + \varepsilon_{it}$

To avoid multicollinearity, the independent variable and moderator were mean centred to compute the interaction. To reflect the causal relationship, all independent variables and control variables were lagged by one year from the dependent variable. In addition, to account for heteroscedasticity, cluster-robust standard error at the firm level was adopted in the study.

4. Results

The correlations, means and standard deviations of all variables are shown in Table 4. To exclude the potential influence of multicollinearity, the variance inflation factor (VIF) for each regression model is calculated and the results show that the VIF values for all models are less than 10, indicating no problems of multicollinearity. Furthermore, the normality of the residuals in all estimated models is tested through skewness and kurtosis tests. The results show that p-values are greater than 0.05, failing to reject the null hypotheses of the normally distributed residuals.

[Insert Table 4 here]

4.1 Impact of ambidextrous blockchain initiatives on operational efficiency

Table 5 presents the pooled regression outcomes of the main effect. Among the control

variables, firm size and firm profitability remained positively significant across the three models, indicating that larger and more profitable firms tend to be more operationally efficient. In addition, firms with long cash turnover cycles are less efficient. The coefficients of complementary ($\beta = 0.087$, p < 0.01) and balanced ($\beta = 0.057$, p < 0.05) blockchain initiatives are both positively significant. These results suggest that firms' ambidextrous blockchain initiatives improve operational efficiency. Thus, H1 and H2 are supported.

[Insert Table 5 here]

4.2 Moderating effect of unabsorbed slack

Model (5) shows that the interaction between complementary blockchain initiatives and unabsorbed slack is not statistically significant ($\beta = 0.006$, p > 0.1), meaning that the impact of complementary dimension on operational efficiency does not rely on firms' unabsorbed slack. Thus, H3a is not supported. However, there is a negatively significant interaction between balanced dimension and unabsorbed slack ($\beta = -0.027$, p < 0.05), as shown in Model (7). Thus, H3b is supported.

[Insert Table 6 here]

To clearly demonstrate the moderating effect, we conducted simple slope analysis and plotted simple slopes at the high (+1 standard deviation above the mean) and low (-1 standard deviation above the mean) levels of unabsorbed slack. As shown in Figure 3, when a firm has a high level of unabsorbed slack, the simple slope is not statistically significant ($\beta = -0.063$, p > 0.1), whereas the simple slope is positively

significant ($\beta = 0.124$, p < 0.001) with low unabsorbed slack. These results indicate an interference effect of unabsorbed slack.

[Insert Figure 3 here]

4.3 Moderating effect of environmental munificence

As shown in Model (9), the moderating effect of munificence on complementary blockchain initiatives is not significant ($\beta = -0.386$, p > 0.1). Thus, H4a is rejected. In contrast, the coefficient of the interaction between the balanced ambidexterity and munificence is positive and significant ($\beta = 0.853$, p < 0.05), suggesting a reinforcement effect of munificence. Thus, H4b is supported.

[Insert Table 7 here]

Consistent with our predictions, simple slope analysis (see Figure 4) shows that in a more munificent environment, balanced blockchain initiatives will lead to a higher level of operational efficiency ($\beta = 0.090$, p < 0.05). However, no significant influence on a firm's operational efficiency ($\beta = 0.011$, p > 0.1) is shown under a low level of environmental munificence.

[Insert Figure 4 here]

4.4 Robustness tests

We also employed alternative measurements of the dependent and control variables to check the robustness of our findings. Specifically, we generated alternative operational efficiency by replacing one input element (namely, cost of goods sold) with inventory. For the controls, we computed firm age as the number of years since IPO listing and firm size as the natural logarithm of total assets. Overall, the results of the robustness tests were consistent with our findings, further supporting our arguments.

5. Discussion

This study has two key findings. First, it confirms that a firm's ambidexterity in blockchain initiatives, whether complementary or balanced, is associated with increased operational efficiency. The result is consistent with prior empirical studies on the business value of using blockchain – to name a few, improved operational capabilities (Pan *et al.*, 2020), lower systemic risks and increased investment efficiency (Chen *et al.*, 2021) and increased stock returns (Xiong *et al.*, 2021). Our result also supports the benefits of ambidexterity in the realm of technology, which have been confirmed by extant research on enhanced supply chain resilience due to IT ambidexterity (Gu *et al.*, 2021), and increased operational support due to information systems ambidexterity (Tai *et al.*, 2019). This supports the credibility of both our findings and those of previous researchers.

Second, we empirically test the moderating effects of organisational and environmental contingencies (i.e. unobserved slack and environmental munificence). With regard to complementary blockchain initiatives, neither unobserved slack nor environmental munificence has a significant moderation effect. A possible explanation is that resources, whether obtained internally or externally, might affect EtBI and ErBI in opposite manners. Li *et al.* (2013) demonstrated that Chinese manufacturing firms have long preferred to absorb advanced knowledge through exploitation, whereas exploration is far from being fully realised due to a lack of resources and capabilities. However, product innovation advantage is one foremost concern for manufacturing firms, which indicates a significant role of exploration. Therefore, according to Li et al. (2013), when firms are situated in munificent environments or have more unabsorbed slack, they prefer to devote more resources and capabilities to ErBI, as the risk of failure in innovative programmes is reduced. As a result, the effectiveness of ErBI is improved, and the positive influence of ErBI on operational efficiency is enhanced. On the contrary, the marginal effect of EtBI on operational efficiency is comparatively lower, as exploitation has been leveraged in a previous long time. Consequently, manufacturing firms might ignore the influence of EtBI, thus weakening the efficiency promotion mechanism of EtBI. Taken together, the contradictory effects are neutralised. This aligns with Pan et al. (2018), who argue that higher environmental munificence erodes the complementary effect of exploitative technological diversification on firm performance. This finding is also similar to that of Lee et al. (2020), who confirmed that firm size (which represents the quantity of a firm's resources) did not significantly moderate the positive relationship between ambidextrous knowledge sharing and financial performance.

Furthermore, balanced blockchain initiatives generate greater operational efficiency under conditions of less unabsorbed slack or greater environmental munificence. Firms with more unabsorbed slack or in less munificent environments may not accrue efficiency promotion from the relative balance of EtBI and ErBI but would rather benefit from a more focused one-sided strategy.

5.1 Theoretical implications

The study contributes to the literature in several ways. First, given that blockchain initiatives can be classified into two categorises (i.e. exploitative and exploratory), and separated technology exploration and exploitation could hamper their synergies (Geerts *et al.*, 2018), an insufficient investigation on this issue could cause ambiguity and bias on understanding the actual magnitude of the effect of blockchain on operational efficiency. Therefore, instead of simply examining the effect of applying blockchain, this study further distinguished between two types of blockchain initiatives. This aligns with Zhu *et al.*'s (2021) suggestion to categorise big data and analytics implementations as exploration-oriented or exploitation-oriented when studying their value creation.

Second, the connotation of organisational ambidexterity is employed to explore the effect of blockchain initiatives on operational efficiency. This not only provides a better understanding of blockchain, providing a theoretical lens to explain the mechanism of how coordinating exploitation and exploration can be beneficial, but affirms the applicability of organisational ambidexterity in blockchain literature. While organisational ambidexterity has been widely used to investigate organisational-level phenomena (e.g. Luger *et al.*, 2018; Luo *et al.*, 2016), its application in the blockchain literature remains underexplored and its inclusion helps extend the knowledge in this area. In particular, our findings show that complementary and balanced blockchain initiatives are both positively associated with operational efficiency, reinforcing the theory of organisational ambidexterity. Moreover, this work can inform future research

on how to integrate organisational ambidexterity theory when studying other emerging technologies.

Second, this study also examines under what conditions firms can achieve operational efficiency. Indeed, management practices which are optimal for all instances are non-existent, and the results of specific firm initiatives are actually dependent on the internal and external context of a firm. In this circumstance, to reach insightful conclusions, contextual factors need to be examined. Thus, we further investigated the role of two contingencies: an internal, organisational factor (unabsorbed slack) and an external, environmental factor (environmental munificence). Interestingly, whilst the benefit of complementing EtBI and ErBI does not vary across different levels of unabsorbed slack and environmental munificence, the efficiency promotion effect of the balanced dimension is moderated in opposite ways by these two factors. These results are different from those in traditional ambidexterity studies (e.g. Luo et al., 2016; Wu and Liu, 2018), which have found that both unabsorbed slack and environmental munificence can improve the performance of complementarily ambidextrous organisations while weakening the benefits of balanced ambidexterity. The perspective of both internal and external contingencies provides a more comprehensive viewpoint when studying the effects of blockchain. On the other hand, this work also echoes Rossi et al. (2019), who called for more empirical effort in 'test[ing] existing theories that should be revisited in light of blockchain' (p.1396). This study thus suggests the necessity of grounding additional novel theories in future blockchain literature.

5.2 Managerial implications

The present research has several managerial implications. Given that efficiency is quite important for manufacturing firms, we suggest that decision-makers remain highly focused on implementing blockchain initiatives. Although firms have shown increasing interest in blockchain, they are still sceptical of its business value (Xiong *et al.*, 2021). However, performance expectancy is an important driver of adopting blockchain (Queiroz and Wamba, 2019), and doubt about blockchain's value can serve as an important barrier to its adoption. With respect to this concern, the empirical results of this study confirm that blockchain can have a positive impact on operational efficiency. Thus, this study provides decision-makers with justification for investing in blockchain.

Furthermore, our results show that, rather than being contradictory, EtBI and ErBI mutually enhance one another in promoting operational efficiency. The synergistic effect of EtBI and ErBI further indicates the need for firms to pursue exploitation and exploration simultaneously. Thus, apart from simply embracing blockchain, manufacturing firms are also recommended to appropriately allocate resources to achieve the dual development of EtBI and ErBI. In terms of He and Wong (2004), if conditions permit, practitioners are recommended to manage the tension between EtBI and ErBI in a more continuous process, rather than focusing on just one dimension in discrete periods.

In the meantime, this study can advise blockchain executives to determine whether to maintain a balance between EtBI and ErBI under different organisational and environmental contexts. Our findings suggest that managers should evaluate their

internal and external contingencies before making blockchain investments to determine their strategic emphasis. Firms with less unabsorbed slack and those in more munificent industry contexts, may benefit from balanced blockchain initiatives. Managers should also be cautious about maintaining a balance to avoid the risks associated with overemphasising one aspect. In this regard, despite the complexity and delicacy involved in pursuing such a balance, managers should endeavour to execute ambidexterity in using blockchain to seek better operational efficiency. To achieve it, purposeful training should be launched to foster employees' skills, competencies and desire to participate in challenging blockchain programmes. In addition, top management teams need to develop their capacity to wisely and timely adjust strategic focus, and take actions in resource orchestration.

5.3 Limitations and future research

Although this study makes some theoretical and practical contributions, some limitations should be considered in future research. First, the data used for empirical analysis were collected from listed Chinese manufacturing firms, which could reduce the generalisability of the results. The level of blockchain technological development and applications varies across countries and industries. The benefits of ambidextrous blockchain use can thus be different in different contexts. We therefore encourage future researchers to conduct studies in diversified settings. Second, although we used secondary data derived from different sources, which could deal with common method bias, our measurements still have limitations. We measured EtBI, ErBI and ambidexterity through firms' official announcements. However, as Luger *et al.* (2018) pointed, for the sake of confidentiality, firms have a tendency to withhold information of some exploratory initiatives rather than releasing them publicly. In the future, researchers can use additional data sources such as primary data through in-depth surveys and interviews to facilitate the information of secondary data. Finally, given the difficulties associated with collecting data about firms' blockchain adoption and application, content analysis was manually carried out by trained coders to measure the independent variables. Despite the effort to minimise biased coding, some degree of subjectivity is still inevitable. Inspired by extant studies adopting computer-aided text analysis to measure exploration and exploitation (e.g. Jancenelle, 2020), we recommend that future researchers attempt to advance our method of quantifying firms' different implementations of blockchain by improving objectivity in this manner.

5.4 Conclusion

In conclusion, blockchain initiatives can help firms reduce costs, shorten process time, eliminate the risks associated with human error and so on. Therefore, blockchain can provide firms with the opportunity to improve operational efficiency. Grounded in organisational ambidexterity, this study theorised and empirically tested the role of blockchain in improving operational efficiency and how the promotion effect was moderated by organisational and environmental contingencies. The empirical analysis shows that both complementary and balanced blockchain initiatives enable firms to be operationally efficient. In addition, the effect of balanced blockchain initiatives is

contingent: it contributes to higher operational efficiency when firms operate with less unabsorbed slack or in more munificent environments. These findings provide a theoretical understanding of the benefits of blockchain from the perspective of organisational ambidexterity. This study also provides insights for manufacturing firms regarding how to promote operational efficiency via complementary and balanced blockchain initiatives as well as how to flexibly adjust their emphasis on ErBI and EtBI in light of their organisational characteristics and industry environments.

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Table 1. Descriptive statistics of 141 manufacturing firms

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C-4 Ornelati squipment manufacturing 9 0.53378 C13 Texitic gamment and apper product industry 6 4.255% C20 Instrument and meter manufacturing 5 3.546% C21 Instrument and meter manufacturing 5 3.546% C31 Industry of rubber and plastic products 6 4.255% C40 Instrument and meter manufacturing 5 3.546% C41 Alcohol, beverage and refined tea manufacturing 2 17.021% Panel C: Characteristics of firms 24 17.021% Variable Millions 12052.387 44061.962 155.322 408697.219 Total assets Millions 12052.387 44061.962 155.322 408697.219 Total assets Millions 12082.387 4.939 4.000 39.000 State Years 19.035 4.939 4.000 39.000
C10 1 exiling and appare industry 8 3.0.4% C22 Industry of rubber and plastic products 6 4.255% C39 Industry of rubber and plastic products 6 4.255% C40 Instrument and meter manufacturing 5 3.546% C15 Alcohol, beverage and refined tea manufacturing 4 2.837% Other codes Other industries 24 17.021% Panel C: Characteristics of firms 1 12052.387 Standard deviation Minimum Maximum Variable Unit Millions 16186.671 43748.110 382.652 301955.406 Number of employees Thousands 7.219 21.606 0.118 209.671 Age Years 19.035 4.039 4.000 39.000
C22 reperimaning and paper product industry 0 4.255% C30 Instrument and meter manufacturing 5 3.546% C40 Instrument and meter manufacturing 5 3.546% C15 Alcohol, beverage and refined tea manufacturing 4 2.837% Panel C: Characteristics of firms 24 17.021% Panel C: Characteristics of firms 24 17.021% Variable Millions 1018,6671 43748,110 382,652 301955,406 Number of employces Years 19.035 4.939 4.000 39.000
C-29 industry of trubber and plastic products 0 4.255% C40 Instrument and meter manufacturing 4 2.837% Other codes Other industries 24 17.021% Panel C: Characteristics of firms 1 17.021% 17.021% Variable Unit Mean Standard deviation Minimum Maximum Variable Unit Mean Standard deviation Minimum Maximum Sales Millions 16186.671 43748.110 382.652 301955.406 Number of employees Thousands 7.219 21.006 0.118 209.671 Age Years 19.035 4.939 4.000 39.000
C40 Instrument and meter manufacturing 5 3.546% Other codes Other industries 24 17.021% Panel C: Characteristics of firms Variable Unit Mean Standard deviation Minimum Maximum Sales Millions 12052.387 44061.962 155.322 408697.219 Total assets Millions 12052.387 44061.962 0.115.322 408697.219 Age Years 19.035 4.939 4.000 39.000
C15 Alcohol, beverage and refined lea manufacturing <u>4</u> 2,837% <u>Other codes</u> <u>0ther industries</u> <u>24</u> 17.021% <u>Panel C: Characteristics of firms</u> <u>Variable Unit Mean Standard deviation Minimum Maximum</u> <u>Sales Millions 16186,671 43748,110 382,652 301955,406</u> <u>Number of employees Thousands 7,219 21,606 0.118 209,671</u> <u>Age Years 19,035 4,939 4,000 39,000</u>
Other codes Other industries 24 17.021% Panel C: Characteristics of firms Variable Unit Mean Standard deviation Minimum Maximum Sales Millions 10523 387 44061.962 155.322 408697.219 Total assets Millions 17.19 21.606 0.118 209.671 Age Years 19.035 4.939 4.000 39.000
Panel C: Characteristics of firms Variable Unit Mean Standard deviation Minimum Maximum Sales Millions 16186.671 43748.110 382.652 301955.406 Age Years 19.035 4.066 0.118 209.671 Age Years 19.035 4.030 39.000
Variable Unit Mean Standard deviation Minimum Maximum Sales Millions 12052.387 44061.962 155.322 408697.219 Total assets Millions fol38.651 43748.110 382.652 301955.406 Number of employees Thousands 7.219 21.606 0.118 209.671 Age Years 19.035 4.939 4.000 39.000
Sales Millions 12052.387 44061.962 155.322 408697.219 Total assets Millions 16186.671 43748.110 382.652 301955.406 Age Years 19.035 4.939 4.000 39.000
Total assets Millions 16186.671 43748.110 382.652 301955.406 Age Years 19.035 4.939 4.000 39.000
Number of employees Thousands 7.219 21.606 0.118 209.671 Age 4.939 4.000 39.000 30.000 30.000 39
Age Years 19.035 4.939 4.000 39.000
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Table 2. Variable descriptions

Variable name	Measurement	Data source	Reference
EtBI/ErBI	Identify and code a firm's blockchain initiative based on its announcement	Announcements	Developed
Ambidextrous blockchain initiative Complementary Balanced	Interaction: EtBI×ErBI Absolute difference (reversed): 3- ErBI — EtBI	Announcements	Wong, Wong, and Boonitt (2013)
Operational efficiency	Technical efficiency calculated using stochastic frontier estimation	CSMAR	Li <i>et al.</i> (2010)
Munificence	Slope coefficient generated by regressing sales over 10-year time period/mean sales in same time period	CSMAR	Goll and Rasheed (2005), Jacobs, Swink, and Linderman (2015)
Unabsorbed slack	Current ratio	CSMAR	Iyer and Miller (2008)
Firm size	Natural logarithm of a firm's sales	CSMAR	Lee and Huang (2012)
Firm profitability	ROA	CSMAR	Fong, Misangyi and Tosi (2010)
Firm age	Number of years since incorporation	CSMAR	Kotha, Zheng, and George (2011)
Firm cash-to-cash cycle	Days of inventory + days of accounts receivables - days of accounts payables	CSMAR	Lam (2018)

Table 3. Examples of EtBI and ErBI

Company	Text extracted from official announcements	Blockchain
name	e e e e e e e e e e e e e e e e e e e	initiative type
China High- Speed Railway Technology	The firm's professional maintenance and inspection will be upgraded using blockchain to efficiently collect multi-dimensional comprehensive data on maintenance and inspection.	
STEP	The firm further improved and upgraded its elevator control system manufacturing plant. Binding the elevator computer accessories to the original elevator manufacturer through a blockchain encryption algorithm.	Exploitative
Julong	The firm upgraded for its serial number management system. Superimposing blockchain to launch an RMB circulation management solution, which enables parallel triggers and immutability.	
Linyang	Linyang continues to innovate product solutions based on blockchain. It completed the verification test of blockchain smart meters, which can be subsequently applied to power energy spot transactions.	4
Tellhow Sci-Tec	The firm developed the first Chinese blockchain-based power spot trading technical support system and became the core system supplier for power grid firms' power transactions.	Exploratory
AISINO	AISINO completed the research and development of the electronic invoice system based on the full life cycle of the blockchain, and output 8 blockchain tax-related application solutions.	

*Note: The text in our examples was translated from Chinese to English. To provide useful examples, we paraphrased text from the announcements rather than quoting them directly, as the quotes are generally too long and difficult to understand out of context.

Table 4. Correlations and statistics of examined variables

Variables	1	2	3	4	5	6	7	8	9	10	11
1. Operational efficiency											
2. Complementary blockchain initiative	0.042	1									
3. Balanced blockchain initiative	0.073	-0.122**	1								
4. Exploitative blockchain initiative	-0.120**	0.487***	-0.346***	1							
5. Exploratory blockchain initiative	0.007	0.460***	-0.862***	0.318***	1						
6. Munificence	0.050	-0.058	0.099**	0.016	-0.108**	1					
7. Unabsorbed slack	0.019	0.040	0.024	-0.010	0.006	0.052	1				
8. Firm age	0.039	-0.038	0.007	-0.073	-0.023	0.039	-0.123**	1			
9. Firm size	0.197***	0.027	-0.049	-0.01	0.079	-0.129**	-0.283***	0.172***	1		
10. Firm profitability	0.396***	-0.135***	0.198***	-0.145***	-0.147***	0.065	0.077	-0.048	0.132***	1	
11. Firm cash-to-cash cycle	-0.088*	0.029	-0.049	0.0240	0.067	0.013	-0.112**	0.050	0.197***	-0.000	1
Mean	0.425	0.037	2.813	0.061	0.187	0.153	2.469	16.5	21.53	0.033	-0.032
Standard deviation	0.158	0.256	0.445	0.239	0.484	0.047	3.472	4.803	1.530	0.072	1.097
Jotes: * <i>p</i> < 0.1, ** <i>p</i> < 0.05, *** <i>p</i> < 0.	01 (two-taile	ed tests)						Ŷ	S	Ste	ns.

onplementary blockchain initiative 0.087** alanced blockchain initiative 0.037** initiative blockchain initiative 0.010 initiative blockchain initiative 0.011 initiative blockchain initiative 0.011 initiative blockchain initiative 0.011 initiative 0.011 initiative 0.011 initiative 0.031 initiative 0.032 initiative 0.035 initiative 0.035 initiative 0.035 initinitiative 0.022	Variable	Model (1)	Model (2)	Model (3)
ompenenerary blockchain initiative (3.19) halmeed blockchain initiative (0.57) ixploitative blockchain initiative (0.260) ixploitative blockchain initiative (0.922) ixploitative blockchain initiative (0.921) ixploitative blockchain initiative (0.922) ixploitative blockchain initiative (0.921) ixploitatitit			0.087***	
alanced blockchain initiative 0.057° (2.13) sploratory blockchain initiative 0.001 0.001 0.001 im age 0.081 0.084) 0.632) (2.30) im age 0.081 0.084) 0.632) (2.46) im age 0.081 0.084) 0.632) (2.46) im age 0.081 0.0844 0.797° (0.187) im age 0.081 0.8844 0.797° (0.187) im age/mathinitiative 0.019** 0.018** -0.018** -0.018** -0.019** im age/mathinitiative 0.018** -0.018*** -0.018** -0.019** -0.019** im cash-to-cash cycle 0.0122 0.007 0.176 -0.022 0.007 0.176 ons -0.022 0.007 0.176 -0.022 0.007 0.176 ons 0.185 0.202 0.007 0.176 -0.05 +** -0.016 -0.02 0.007 0.176 ont ** 0.185 0.202	omplementary blockchain initiative		(3.199)	
aaanced boecchain initiative (2.113) xploritative blockchain initiative (0.075°) (0.114) im age (0.081) (0.081) (0.077°) im age (0.081) (0.581) (0.584) (0.632) im size (2.146) (0.018°) (2.325) (1.144) im profitability (0.581) (0.584) (0.632) (2.325) im profitability (0.579) (0.8460) (0.617) (0.118°) im cash-to-cash cycle (3.052) (3.033) (3.152) (-0.018°** -0.018°** im cash-to-cash cycle (3.052) (-0.040) (-0.987) (-0.018°** -0.018°** idjusted R ² 0.05, ***p < 0.01 (two-tailed tests); r-statistics in parentheses; all independent varia				0.057**
spionative blockchain initiative 0,000 pionatory blockchain initiative 0,000 mage 0,000	alanced blockchain initiative			(2.113)
cp. (o) (a) (b) (b) (c) (c) (c) (c) (c) (c) (c) (c) (c) (c	valoitativo blookahain initiativo		-0.076**	-0.034
sporatory blockchain initiative0001000100180077"im age0001000100107"0018"im size019**018**018**019**im profilability0108**0108**0108**0108**im cash-to-cash cycle01018**01018**01018**01018**im cash-to-cash cycle01018**01018**01018**01018**im cash-to-cash cycle01018**01018**01018**01018**im cash-to-cash cycle01018**01018**01018**01018**im cash-to-cash cycle0118**0118**0118**0118**im cash-to-cash cycle0118**0118**0118***<	Exploitative blockchain initiative		(-2.607)	(-1.144)
$\frac{(0,022)}{(1,022)} (2,000) \frac{(0,001)}{(0,001)} \frac{(0,001)}{(0,00$	Exploratory blockchain initiative		0.018	0.077***
im age 0.001 0.001 0.001 im size 0.19** 0.017* 0.018* im profitability 0.87** 0.834** 0.019** im cash-to-cash cycle 0.018** -0.018** -0.018** im cash-to-cash cycle 0.022 -0.007 -0.176 cast 0.022 -0.007 -0.176 cast 0.128 0.022 -0.007 -0.176 cast 0.012 -0.0400 (-0.987) daysted R ² 0.128 0.202 0.007 -0.176 cast 0.128 0.202 0.007 -0.176 daysted R ² 0.185 0.202 0.017 0.176 daysted R ² 0.101 (two-tailed tests); r-statistics in parentheses; all independent varia ables one-year lagged. 0.185 0.202 0.196	Exploratory blockenam initiative		(0.922)	(2.960)
0.0581) 0.0584) 0.0542) imm size 0.019** 0.017** 0.018** imm profitability 0.807*** 0.834** 0.797*** irm cash-to-cash cycle (3.052) (3.083) (3.152) rear dummies included included included cons -0.018** -0.007** -0.010*** cons -0.012 -0.007** -0.010*** cons -0.012 -0.007** -0.010*** cons -0.012 -0.007 -0.176 cons -0.012 -0.0040) (0.987) digusted R ³ 0.185 0.202 0.196 cist 4.786 5.352 5.424 0.1, **p < 0.05, ***p < 0.01 (two-tailed tests); t-statistics in parentheses; all independent varia	Firm age	0.001	0.001	0.001
im size 0.019** 0.017** 0.018** im profitability (4.579) (4.860) (4.641) im cash-to-cash cycle -0.018*** -0.018*** -0.018** im cash-to-cash cycle (3.053) (3.083) (3.152) ier dummics Included Included Included cons -0.022 -0.007 -0.176 i/a 0.185 0.202 0.196 i/a 0.185 0.202 0.196 i/a 0.185 0.202 0.196 i/a 0.185 0.202 0.196 i/a 1.85 0.202 0.196 i/a 1.85 0.202 0.196 i/a 1.7*p < 0.05, ***p < 0.01 (two-tailed tests), r-statistics in parentheses; all independent varia	i mi ugo	(0.581)	(0.584)	(0.632)
(2.346) (2.322) (2.232) irm profitability (4.579) (4.360) (4.641) irm cash-to-cash cycle (-0.018***) -0.018***) -0.019*** 'ear dummies Included Included Included 'oans (-0.128) (-0.040) (-0.987) // coss (-0.128) (-0.040) (-0.987) // dysted R ² 0.155 5.352 5.424 0.1, **p < 0.05, ***p < 0.01 (two-tailed tests); <i>t</i> -statistics in parentheses; all independent varia ables one-year lagged.	Firm size	0.019**	0.017**	0.018**
imp profitability 0.807*** 0.834*** 0.797*** irm cash-to-cash cycle (3.052) (4.860) (4.641) irm cash-to-cash cycle (3.052) (3.083) (-3.152) (car dummics Included Included Included (cons (-0.128) (-0.040) (-0.987) (d)usted R ² 0.185 0.202 0.0007 -0.176 (d)usted R ² 0.185 0.320 0.196* -0.176 (d)usted R ² 0.185 0.322 0.424 -0.218* -0.128 (d)usted R ² 0.185 0.322 0.196* -0.176 (d)usted R ² 0.15, ***p < 0.01 (two-tailed tests); <i>t</i> -statistics in parentheses; all independent varia ables one-year lagged.		(2.346)	(2.232)	(2.235)
(4,5/9) (4,560) (4,641) imm cash-to-cash cycle -0.018*** -0.018** -0.019** 'car dummies included Included Included \lambda \frac{1}{2003} (-0.128) (-0.007) 0.176 \lambda \frac{1}{2003} (-0.128) (-0.040) (-0.987) \lambda \frac{1}{2003} 0.185 0.202 0.196 \lambda \frac{1}{2003} 0.185 0.202 0.176 \lambda \frac{1}{2003} 0.176 0.176 0.176	Firm profitability	0.807***	0.834***	0.797***
imm cash-to-cash cycle -0.018^{***} -0.018^{**} -0.018^{**} -0.015^{**} -0.015^{**} -0.015^{***} -0.015^{***} -0.015^{***} -0.012^{***} -0.012^{***} -0.012^{***} -0.012^{***} -0.012^{***} -0.012^{***} -0.012^{***} -0.012^{***} -0.012^{***} -0.012^{***} -0.012^{***} -0.012^{***} -0.012^{***} -0.012^{****} -0.012^{****} -0.012^{****} -0.012^{****} -0.012^{****} -0.012^{*****} $-0.012^{************************************$		(4.579)	(4.860)	(4.641)
The domain of the second seco	Firm cash-to-cash cycle	-0.018***	-0.018	-0.019***
Incluced	Vaar dummias	(-3.052) Included	(-3.083) Included	(-3.152) Included
$\frac{(0.128)}{(0.040)} + \frac{(0.040)}{(0.0987)} + \frac{(0.128)}{395} + \frac{(0.040)}{395} + \frac$	i cai dummies	Included	included	included
(4) (4) (4) (4) (4) (4) (4) (4) (4) (4)	Cons	-0.022	-0.007	-0.1/6
4 dipusted R ² 0.185 0.202 0.196 4.786 5.352 5.424 0.1, **p < 0.05, ***p < 0.01 (two-tailed tests); <i>t</i> -statistics in parentheses; all independent varia ables one-year lagged.	N	(-0.128)	(-0.040)	(-0.987)
Arges ar - 0.185 0.02 0.196 - 4.786 5.352 5.424 0.1, ** <i>p</i> < 0.05, *** <i>p</i> < 0.01 (two-tailed tests); <i>t</i> -statistics in parentheses; all independent varia ables one-year lagged.	IV A divisted <i>D</i> ²	395	395	395
4, roo 3, 352 5, 42 0.1, **p < 0.05, ***p < 0.01 (two-tailed tests); t-statistics in parentheses; all independent varia ables one-year lagged.		0.185	0.202	0.196
ables one-year lagged.	0.1, ** $p < 0.05$, *** $p < 0.01$ (two-tailed)	ed tests); <i>t</i> -statistics ir	n parentheses; all	independent var
4				
		4		

Table 5. Impact of ambidextrous blockchain initiatives on operational efficiency

Table 6. Moderating effect of unabsorbed slack

Variable	Model (4)	Model (5)	Model (6)	Model (7)
Complementary blockchain initiative	0.086***	0.085***		
	(3.116)	(3.161)		
Balanced blockchain initiative			0.057**	0.031
			(2.079)	(1.102)
Unabsorbed slack	0.001	0.001	0.001	0.013**
	(0.212)	(0.246)	(0.248)	(2.380)
Complementary blockchain		0.006		
initiative × unabsorbed slack		(0.657)		
Balanced blockchain				-0.027**
initiative × unabsorbed slack				(-2.550)
Exploitative blockchain initiative	-0.076**	-0.079**	-0.034	-0.035
	(-2.587)	(-2.546)	(-1.141)	(-1.362)
Exploratory blockchain initiative	0.018	0.017	0.076***	0.058**
	(0.919)	(0.875)	(2.928)	(2.209)
Firm age	0.001	0.001	0.001	0.001
	(0.602)	(0.564)	(0.653)	(0.476)
Firm size	0.018**	0.018**	0.018**	0.020**
	(2.173)	(2.186)	(2.187)	(2.405)
Firm profitability	0.829***	0.823***	0.791***	0.768***
	(4.747)	(4.673)	(4.551)	(4.555)
Firm cash-to-cash cycle	-0.018***	-0.018***	-0.019***	-0.017***
	(-3.106)	(-3.076)	(-3.173)	(-2.944)
Year dummies	Included	Included	Included	Included
Comp	-0.020	-0.017	-0.189	-0.055
Cons	(-0.109)	(-0.093)	(-1.007)	(-0.315)
N	395	395	395	395
Adjusted R ²	0.200	0.199	0.194	0.204
F	4.914	4.760	4.986	5.796

Notes: *p < 0.1, **p < 0.05, ***p < 0.01 (two-tailed tests); *t*-statistics in parentheses; all independent variables and control variables one-year lagged.

.0.1k .1.007) 395 0.194 4.986 .s in parentheses; all indeper.

Variable	Model (8)	Model (9)	Model (10)	Model (11)
Complementary blockchain initiative	0.090***	0.076**		
	(3.289)	(2.382)		
Balanced blockchain initiative			0.058**	0.051*
			(2.094)	(1.816)
Iunificence	0.212	0.210	0.183	-0.108
	(1.045)	(1.041)	(0.883)	(-0.407)
omplementary blockchain		-0.386		
nitiative × munificence		(-1.171)		
alanced blockchain				0.853**
itiative × munificence				(1.985)
xploitative blockchain initiative	-0.083***	-0.078**	-0.038	-0.030
	(-2.774)	(-2.556)	(-1.218)	(-0.945)
xploratory blockchain initiative	0.018	0.017	0.078***	0.059**
	(0.919)	(0.857)	(2.923)	(2.021)
rm age	0.001	0.001	0.001	0.001
	(0.465)	(0.434)	(0.530)	(0.571)
rm size	0.018**	0.018**	0.018**	0.019**
	(2.286)	(2.291)	(2.278)	(2.395)
rm profitability	0.827***	0.842***	0.790***	0.821***
· · · · · · · · · · · · · · · · · · ·	(4.820)	(4.651)	(4.587)	(4.615)
irm cash-to-cash cycle	-0.019***	-0.018***	-0.019***	-0.019***
	(-3.106)	(-3.078)	(-3.169)	(-3.170)
ear dummies	Included	Included	Included	Included
	-0.055	-0.021	-0.218	-0.069
ons	(-0.309)	(-0.119)	(-1,177)	(-0.398)
	395	395	395	395
	570	0.202	0.196	0.203
further the test state R^2	0 203	U 202		
djusted R^2 * $p < 0.1$, ** $p < 0.05$, *** $p < 0.01$ (to variables one-year lagged.	0.203 5.277 two-tailed tests); ,	5.065	5.270 entheses; all indep	4.458 endent variables and
Adjusted R^2 7^{7} p < 0.1, **p < 0.05, ***p < 0.01 (not contract on the second seco	0.203 5.277 two-tailed tests); ,	5.065 t-statistics in pare	5.270 entheses; all indep	4.458 endent variables and
Adjusted R^2 F s: * $p < 0.1$, ** $p < 0.05$, *** $p < 0.01$ (rol variables one-year lagged.	0.203 5.277 two-tailed tests);	5.065 t-statistics in pare	5.270 entheses; all indep	4.458 endent variables and
Adjusted R^2 F s: * $p < 0.1$, ** $p < 0.05$, *** $p < 0.01$ (rol variables one-year lagged.	0.203 5.277 two-tailed tests);	5.065 t-statistics in pare	5.270 entheses; all indep	4.458 endent variables and
Adjusted R^2 <u>F</u> s: * $p < 0.1$, ** $p < 0.05$, *** $p < 0.01$ (rol variables one-year lagged.	0.203 5.277 two-tailed tests);	5.065 t-statistics in pare	5.270 entheses; all indep	4.458 endent variables and
Adjusted R^2 <i>F</i> s: * <i>p</i> < 0.1, ** <i>p</i> < 0.05, *** <i>p</i> < 0.01 (not consider the set of the set o	0.203 5.277 two-tailed tests); ;	5.065 t-statistics in pare	5.270 entheses; all indep	4.458 endent variables and
Adjusted R^2 : * $p < 0.1$, ** $p < 0.05$, *** $p < 0.01$ (and the second seco	0.203 5.277 two-tailed tests);	5.065 t-statistics in pare	5.270 entheses; all indep	4.458 endent variables and
Adjusted R^2 r s: $*p < 0.1$, $**p < 0.05$, $***p < 0.01$ (not control of variables one-year lagged.	0.203 5.277 two-tailed tests); ,	5.065 t-statistics in pare	5.270 entheses; all indep	4.458 endent variables and
Adjusted R^2 F s: $*p < 0.1$, $**p < 0.05$, $***p < 0.01$ (not explored) of variables one-year lagged.	0.203 5.277 two-tailed tests);	5.065 t-statistics in pare	5.270 entheses; all indep	4.458 endent variables and
Adjusted R^2 <u>F</u> es: * $p < 0.1$, ** $p < 0.05$, *** $p < 0.01$ (rol variables one-year lagged.	0.203 5.277 two-tailed tests);	5.065 t-statistics in pare	5.270 entheses; all indep	4.458 endent variables and
Adjusted R^2 <i>F</i> s: * <i>p</i> < 0.1, ** <i>p</i> < 0.05, *** <i>p</i> < 0.01 (not consider the second sec	0.203 5.277 two-tailed tests); .	5.065 t-statistics in pare	5.270 entheses; all indep	4.458 endent variables and
djusted R^2 : * $p < 0.1$, ** $p < 0.05$, *** $p < 0.01$ (not set in the set of the set	0.203 5.277 two-tailed tests);	5.065 t-statistics in pare	5.270 entheses; all indep	4.458 endent variables and

Table 7. Moderating effect of environmental munificence





Figure 3. Simple slope analysis of unabsorbed slack



Figure 4. Simple slope analysis of environmental munificence