

Does bribery sand or grease the wheels of firm level innovation: Evidence from Latin American countries

Nirosha Hewa Wellalage

School of Accounting, Finance & Economics

The University of Waikato,

Hamilton, New Zealand

nirosha@waikato.ac.nz

Sujani Thrikawela

Waikato Institute of Technology

Hamilton, New Zealand

sujani.thrikawala@wintec.ac.nz

Cite this article

Wellalage, N., Thrikawala, S. Does bribery sand or grease the wheels of firm level innovation: evidence from Latin American countries. *J Evol Econ* **31**, 891–929 (2021). <https://doi.org/10.1007/s00191-020-00717-0>

Abstract

This paper examines the effects of bribery on product and process innovation at the firm level in Latin American countries. We provide insights into the heterogeneous effects of bribery on the types of innovations across firms. Using the locality-sector average of bribe payments as a percentage of firm annual sales, we control for endogeneity bias, which is a recurring econometric problem in corruption-innovation studies. Overall, we find evidence to support the ‘sand the wheels’ hypothesis and argue that this negative effect is slightly higher in process innovation compared to product innovation. Using interaction terms to see the effects of bribery on different firm sizes and institutional structures, we find that bribery hurts firm innovation for micro and small firms, but not for large firms. The results also suggest that the negative

effect of corruption on firm level innovation is stronger for firms in weak institutional environments than for firms in strong institutional settings. Practical interventions to promote strong institutional governance are not easy to arrange when corruption is endemic. Greater awareness of the extortionist environment of companies in these countries in the international community may bring pressures to bear for reform.

Keywords: Bribes, innovations, endogeneity, emerging economies; Latin America

JEL Classification: D73, O3, G30, L25

1 Introduction

Firm innovations are essential for sustainable economic development and growth as these help in value creation and a country's infrastructure development and employment (Anokhin and Schulze 2009; Ayyagari et al. 2014). However, innovators from emerging economies face considerably high bribing pressure from corrupt public officials (Murphy et al. 1993) and little or no institutional support for innovative firms (Kotabe et al. 2017). In this study, we estimate the effect of bribe payments on firm level innovations in Latin American economies. Additionally, we provide insights into the heterogeneous effects of bribery on the types of innovations across firms and institutions. Given that innovation has been recognized as the main driver of long-term economic growth, a clear understanding of how corruption affects innovation can assist development in emerging economies.

We seek to integrate transaction cost theory (Williamson 1979; Williamson 1987), institutional theory (Scott 2001; Scott 2013) and legitimacy theory (Deegan 2002; Deegan 2014) to understand the effects of bribery on firm level innovation. The literature suggests two competing views: corruption as 'sand in the wheels', where corruption is an involuntary tax (See Fungáčová et al. 2015; Mauro 1995; Reinikka and Svensson 2005), and corruption as 'greasing the wheels' (See Hanousek and Kochanova 2016; Luo 2005; Voskanyan 2000). When firm operators are in a location and sector of prolific bribery, the indirect competitive effects of bribes can affect firm level innovations in two ways. That is, for bribes to facilitate innovation in a locality sector, they should benefit all or most of the bribing firms. Conversely, if bribery helps only a minority of bribing firms and creates negative externalities (Wei and Kaufmann 1999), then firm innovation will deteriorate in a sector or local environment where bribery is commonplace.

Our examination of the relationship between innovation and corruption uses firm-level data for 6,040 firms in Latin American countries. Intrigued by the seemingly high levels of corruption and low levels of business innovation in Latin American countries (Lederman et al. 2014), we undertake an examination of the key variables involved. As a starting point, we investigate whether corruption engenders low innovation or vice versa, or whether both are shaped by other factors. The insights gained from this study potentially offer practical guidance for policymakers, businesses, financial institutions and international bodies. It is important to safeguard the private sector firm in the Latin American region by encouraging innovation, strengthening human capital, increasing productivity, promoting competition, reducing corruption, and the costs of bureaucratic processes.

After controlling for firm characteristics and accounting for the unobservable characteristics of firms which are typically linked to innovation decisions, we find that bribery has a negative effect on firm level innovations, supporting the theory that corruption ‘sands the wheels’ of firm innovation. The empirical findings in our study show that the negative effect of bribe payments on innovations is slightly higher in process innovation outcomes than in product innovations. We also provide insights into the heterogeneous effects of bribery on innovation across firms. In particular, our results indicate that large and medium size firms are not as severely affected as micro and small firms and may even potentially gain from bribery activities. This indicates that sometimes, bribery is a rational strategic response of large enterprise innovators, compensating for a lack of kinship or political affiliations, and hedging against political risk, although bribery acts as an involuntary “tax” for smaller firms. Further, we find that the negative effect on bribery on firm level innovation is much stronger for firms in weak institutional environments than for firms in strong institutional settings.

This paper expands three different aspects of prior research. Growing evidence in research on economic growth and performance stresses the significant impact of bribery

in firm-level innovation activities (See:Ayyagari et al. 2011; Ayyagari et al. 2014; D’Este et al. 2012). However, the measures of innovation used in those studies rarely distinguish between types of innovations and between types of firms. In other words, insight into the heterogeneous effects of bribery on firm level innovation remains limited in the literature. The only two exceptions to this are Mahagaonkar (2010) who distinguished corruption effects on four types of innovation in African firms and Habiyaemye and Raymond (2018), who distinguished various components of innovation activities and corruption practices in Eastern European and Central Asian firms. While recent Latin American studies like Birhanu et al. (2016) and Gaviria and Pagés (2002) have attempted to model the effects of corruption on economic outcomes, there has been minimal empirical evidence to demonstrate the heterogeneous effects of bribery on firm level innovation, a gap that our paper seeks to address. In addition to extending the analysis of corruption and innovation by exploring the heterogeneous effects of bribes and the relationship between firm level innovation, innovation types, and firm size, we also analyze the corruption behavior of firms in different institutional settings and at different levels of sector- locality corruption. We provide a fresh perspective on earlier studies of corruption in emerging countries and highlight the importance of heterogeneity in policy formulations.

Our second contribution is related to measurements of innovation and bribery. Studies to date largely use indirect proxies such as R&D expenses and patents to measure innovations and country level corruption perception indices to measure corruption (see (Dreher and Gassebner 2013; Lööf and Nabavi 2016; Mancusi and Vezzulli 2010), which do not adequately capture the corruption-innovation nexus of firm behavior in emerging economies. This study employs two direct measures of innovation for product and process, following the recommendation of the *Oslo Manual*, as this provides a more robust measure of firm-level innovation for firms, especially those in non-mature economies. We use firm level estimates of bribery, which captures the non-uniform distribution of bribe payments within a

country, while others reflect aggregate, country-level, views on the extent of bribery¹. The specific costs of bribery payments for firms directly link the effect of bribery in fiscal terms to the performance of firms.

Our third contribution is methodological. The perceived and actual level of corruption depends on how society and institute accept and follow the rules (Melgar et al. 2010), which depends on unobserved individual characteristics, such as values and moral views. Therefore, the validity of any conclusions drawn about the relationship between corruption and innovation rests on a proper treatment of the potential endogeneity problem. Using the locality-sector average of bribe payments as a percentage of firm annual sales, we control for endogeneity bias, which is a recurring econometric problem in innovation studies.

The remainder of this paper is organized in sections. The next section reviews prior research while Section 3 reports the data, sample and measures used. Section 4 then describes the methods of analysis and Section 5 reports the results from the analysis. Section 6 reports on the robustness testing of this study before the final section offers a discussion and conclusion.

2 Literature Review, Theoretical Background and Hypotheses Development

2.1 Literature Review

Corruption remains the greatest obstacle for sustainable economic development (United Nations 2006) and has been a topic of debate for over 50 years. In recent years, corruption has attracted much attention by economists and policymakers as it underpins social, cultural, and political discrimination by weakening access to public resources, information and decision-

¹ Cooray and Schneider (2018) and Mauro (1995) used country level corruption index as a proxy for corruption.

making. Corruption is defined as acceptance or extortion of money for private benefit by government officials (Tanzi and Davoodi 1997). However, it remains a point of debate as to whether such illegal payments are harmful to economic activities such as productivity, trade, foreign investment, entrepreneurship, or innovation (Krammer, 2017). In the growing stream of research on corruption, the relationship between corruption and innovation has received scant attention (Anokhin & Schulze, 2009) and remains unclear (Wang & You, 2012).

The literature on the effects of corruption on innovation has two opposing views. One branch of literature depicts corruption as being sand in the wheels, wherein corruption is likely to have detrimental effects on firms (Avnimelech et al. 2014; Dincer 2019; Fungáčová et al. 2015; Mauro 1995; Reinikka and Svensson 2005). In particular, a bribe amounts to a tax on firms and constitutes an obstacle to firm growth (Fungáčová et al. 2015), increases uncertainty (Shleifer and Vishny 1993), limits access to credit (Hewa Wellalage et al. 2019), and reduces foreign investments (Cuervo-Cazurra 2006). Corruption undermines the foundations of institutional trust, which is necessary to develop innovative activities (Habiyaemye and Raymond 2013). It has been argued that corruption increases transaction costs, which makes innovative opportunities much less attractive (Luo 2005).

The opposing view argues that corruption greases the wheels, claiming that in emerging economies, corruption is likely to have beneficial effects for firms suffering from obstructive private monopolies and government practices (Goedhuys et al. 2016; Kabadurmuş 2017; Voskanyan 2000). Hanousek and Kochanovac (2016) argue that the positive effects of bribery on a firm can outweigh the negative effects. Corruption can speed up the slow-moving permit queue (Luo, 2005), and enable licenses to be easily and quickly issued (Bertrand, Djankov, Hanna, & Mullainathan, 2007; Huntington, 2006), thereby reducing the transaction costs associated with the new product or process development. A recent study by Krammer (2017) confirms the greasing the wheel effect by employing data from over 6,000 firms in 30 emerging

markets. In support of this view, Krastanova (2014) finds that for Bulgarian firms, paying bribes minimises bureaucratic hurdles and speeds up product innovation.

Table 1 review the most recent literature related to corruption and innovation studies.

<<INSERT Table 1 here>>

Given that the effects of bribery on firm level innovation remains a subject of debate, further work is required to understand how bribes impact on firm level innovation in emerging economies.

2.2 Theoretical Considerations and Derived Hypotheses

This paper draws on three main theoretical frameworks: transaction cost theory, institutional theory, and legitimacy theory, all of which lead to the development of the hypotheses in this study.

Transaction cost theory suggests that every transaction in an organization produces costs in terms of coordinating, monitoring, controlling and managing transactions but that firms can minimize these costs with an optimum organization structure (Williamson 1979; Williamson 1987). Corruption increases costs in relation to searching for information, negotiation, monitoring and enforcement costs, all of which increase transaction costs and the final cost of the given goods or services (Lambsdorff 2002). On the other hand, corruption can decrease costs if it enables the organization to avoid significant delays in obtaining official approvals. Transaction costs exist in an innovative environment (Ayyagari et al. 2011; Hsieh et al. 2016), which suggests that the relationship between corruption and transaction costs is involved in explaining innovation outcomes (Barasa 2018).

Institutional theory suggests that an organization's activities and behavior are affected and encircled by the political, social and economic systems surrounding the organization (Scott

2001). This theory provides a rich and complex view of an organization and has spread rapidly due to the influence of institutional actors on organizational and decision-making processes (Hoskisson et al. 2000). The institutional environment can have two dimensions: formal and informal. The development of innovations is centered on a formal institutional environment (Williams and Vorley 2014). The formal environment refers to codified rules and regulations that govern the interaction among the different economic actors. The literature indicates that a well-developed formal institutional environment deters the rent-seeking behavior of economic participants (Tambunan 2007) by creating a strong platform for innovation (Manolova et al. 2008).

According to legitimacy theory, the actions of a firm should be appropriate within the socially constructed system of values, norms and beliefs of their respective society (Suchman 1995). However, these bounds and norms are not fixed and change over time. Firms continuously seek to ensure whether their activities are acceptable to the community in which they operate (Deegan 2002; Deegan 2014). In other words, legitimacy is a relative concept to the social system in which the company operates. To determine what establishes legitimacy, firms rely on indications from the institutional environment (Chizema and Pogrebna 2019) and act in strict compliance with the terms of their societal values (Deegan 2002). The level of social and political pressure placed on firms as they carry out their operations signal the extent to which their activities are deemed acceptable by society (Deegan 2014). If a firm is in an institutional environment where high corruption and unethical practices are rife, corruption may have a significant effect on a firm's actions in its search for institutional legitimacy' (Keig et al. 2015).

We propose that legitimacy theory has a role in explaining perceptions of bribery and firm level innovation. That is, the behaviors of a firm, such as the use of bribes, are subject to institutional pressures in the market, especially corrupt officials in government bureaus.

Furthermore, to overcome various institutional pressures in the emerging markets on innovations and to thrive in the business environment, entrepreneurs must use bribes (Krammer 2017). Therefore, bribery will become the norm of competence when deploying new products into the market (Krammer 2017; Luo 2005).

2.1.1 Hypotheses development

Corruption has a significant impact on firm level innovation activities, which in turn affects the country's economic environment (Ayyagari et al. 2014; Barasa 2018). It is generally agreed that corruption has a negative effect on a country's growth, and it undermines the foundations of institutional trust, which is necessary to develop innovative activities (Habiyaemye and Raymond 2013). According to transaction costs theory, corruption will increase transaction costs in the market, which leads to hindering entrepreneurial innovation activities (Chadee and Roxas 2013; Zhu et al. 2012). These high transaction costs arising from corruption may lead to a decrease or a complete withdrawal from innovation by firms (Fan et al. 2017; O'Connell 2016).

Compared with non-innovators, innovating firms have a greater likelihood of paying more bribes as they interact more with government employees. In particular, Ayyagari et al. (2014) find that, in a sample of 57 countries, corruption acts as a tax on innovation because bribe payment is higher among innovating firms than in non-innovating firms. Furthermore, they confirm that when obtaining government services, the innovating firms that pay bribes do not receive greater benefits than firms that do not pay bribes. Innovating firms must pay a higher percentage of the contract value as a bribe and this discourages innovations. Nevertheless, even if bribing expedites individual transactions with bureaucrats, its purpose will fail in the long term, since the number of such future transactions will increase and offset any likely gains (Wei and Kaufmann 1999). Thus, we formulate our first hypothesis as follows:

H1: Corruption has a negative impact on firm-level innovation

The impact of corruption may vary depending on firm size (See Hanousek and Kochanova 2016; Paunov 2016). In particular, the literature indicates that the detrimental effects of corruption are reduced when firm size is increased (See Paunov 2016; Shleifer and Vishny 1993). Small innovative firms have a high level of vulnerability to public corruption compared with firms that are already established (Zhou and Peng 2012; Zhu et al. 2012). Unlike larger well-established firms, small innovative firms are continually involved in innovative activities, such as changing factory layouts, acquiring equipment, importing new products, or registering a new logo for a new product. As a result, government officials who have power over these processes receive attention from these types of firms (Murphy et al. 1993).

Using firm level data for 48 developing and emerging countries, Paunov (2016) emphasizes that paying informal gifts and payments to obtain business license/quality certificates have a stronger negative impact on smaller firms than for larger firms. These smaller firms normally have less resources to draw on when bribing public bureaucrats and they always receive fewer benefits from patents and quality certificates. The literature reports that corruption may temper the negative effect of small firm innovation rather than for large firms (See Clarke and Xu 2004; Schiffer and Weder 2001). This may be the result of small firms engaging in forceful bribery whereas large firms engage in strategic bribery (Zhou and Peng 2012). Overall, we can conclude that an increase in firm size decreases the negative impact on firm level innovation. Thus, we formulate our hypothesis as follows:

H2: An increase in firm size positively moderates the relationship between paying bribes and firm level innovation.

Institutional theory suggests that firms seek external legitimacy by complying with their institutional context (Glynn and Abzug 2002). Corruption affects firm legitimacy through its

effects on formal and informal activities (Rodriguez et al. 2005). In a business environment where rules and regulations are weak, corruption and bribery often exist in day-to-day life, making firms susceptible to this value system (Krammer 2017).

According to transaction costs theory, corruption can either improve or weaken firm performance through transaction costs (Fan et al. 2017; O'Connell 2016). Corruption is less damaging and is sometimes beneficial within a weak institution's structure for several reasons (Dreher and Gassebner 2013; Méon and Sekkat 2005). In fact, corruption may be positively associated with overcoming bureaucratic processes, and unclear or complex regulations (Agrawal and Knoeber 2001; Khwaja and Mian 2005) in a weak institutional environment where the costs of prevention usually outweigh the benefits (Acemoglu and Verdier 2000). Emerging economies report that bribes facilitate innovation by reducing queues (Krammer 2017; Nguyen et al. 2016). Bribes present as an attractive alternative to circumventing red-tape and reducing the negative effects of nepotism and political favouritism by shielding innovative firms against the political risk in many emerging markets (Kasuga 2013). Krastanova (2014), for instance, finds that for Bulgarian firms, paying bribes minimises bureaucratic hurdles and speeds up product innovation. Our next hypothesis is as follows:

H3: Better developed institutional environments weaken the relationship between paying bribes and firm level innovation

3 Data and sample

The current study extracted data from the World Bank 2016 and 2017 Enterprise Surveys². The World Bank's Enterprise Survey database includes both qualitative and quantitative data at the

² The World Bank's Enterprise Surveys offer an expansive array of economic data on 135,000 firms in 139 countries. Its website (<http://www.enterprisesurveys.org/data>) provides details as to how the surveys are

firm level. These surveys comprise a representative random sample of firms with data collected across the world by using the same core questionnaire and the same sampling method. The most recent Enterprises survey provides data for eleven Latin American countries (Argentina, Bolivia, Colombia, Ecuador, El Salvador, Paraguay, Peru, Guatemala, Honduras, Nicaragua and Uruguay), which are all included in our study sample. The surveys employ stratified random sampling techniques with identical questionnaires across all countries. This stratified random sampling ensures a representative sample, which minimizes sample selection biases in data analysis. The strata for Enterprise Surveys are firm size, business sector, and geographic region within a country.

Table 2 displays the total study sample of the paper. The net result is a sample of 6,040 firm-level observations. The sample covers micro, small, medium and large firms.

Table 2 also reports the bribe payments (as a percentage of total sales) level of each country in the sample. Peru pays more bribes than other countries in the sample. The second largest bribe-paying country is Bolivia. We also report the 2017 Corruption Perception Index (CPI) values for each country. The index reports that, with the exception of Uruguay, all countries score below 39, which, in 2017, is below the average world score of 43.

<< INSERT Table 2 here >>

3.1 Variables

Following Wellalage and Fernandez (2019), Gorodnichenko and Schnitzer (2013), and Mahagaonkar (2010), this study uses product innovation and process innovation (dummy

conducted. An Enterprise Survey is a firm-level survey of a representative sample of an economy's private sector. The surveys cover a wide range of business environment topics including innovation, corruption, crime, access to finance, firm ownerships, infrastructure, competition and performance measure.

variables) as proxies for firm level innovation. Product and process innovation variables were derived from the following survey question(s):

During the last three years, has this establishment introduced new or significantly improved products or service methods of manufacturing products or offering service?

The direct measures of innovation are more appropriate for our study for two reasons. First, in a given context, most sample firms are not engaged in R&D and patents (approximately 10% report R&D expenditure and no patents information available for this sample of firms). This aligns with the view that the traditional measures of innovation (such as R&D expenditure and patents) are less likely to be observable in domestically owned small firms (Acemoglu et al. 2006) and in emerging markets (Gorodnichenko and Schnitzer 2013). Second, indirect proxies such as R&D expenses and patents fail to adequately capture the innovative behavior of firms in developing economies (Löf and Nabavi 2016; Mancusi and Vezzulli 2010). An important way for firms in emerging markets to catch up with developed markets is by adopting innovation from elsewhere (Acemoglu et al. 2006; Ayyagari et al. 2011; Ayyagari et al. 2014). Proxies provided by the *Oslo Manual* adequately capture both innovations new to the firm and innovations new to the world (Wellalage and Fernandez 2019).

However, following prior research, we acknowledge the subjectivity of these self-reported measures of innovation, compared to more objective measures, such as patent and R&D investments (See Gorodnichenko and Schnitzer 2013; Wellalage and Fernandez 2019). In order to verify the robustness of our results, we use R&D expenditures as a proxy for innovation as well.

Our main independent variable is *Bribes(p)*, This takes a value ranging from 0 to 100 as a percent of annual sales. In terms of corruption perception indicators, this monetary value

of corruption improves its accuracy (Méndez and Sepúlveda 2009; O'Toole and Tarp 2014). It has been derived from the following interview question in the Enterprise Survey:

“It is said that establishments are sometimes required to make gifts or informal payments to public officials to 'get things done' about customs, taxes, licenses, regulations, services, etc. On average, what percentage of total annual sales, or estimated total annual value, do establishments like this one pay in informal payments or gifts to public officials for this purpose?”. (World Bank Enterprise Survey, 2016/2017)

By their very nature, corrupt activities do not offer much in the way of paper trails. Therefore, this question represents the best available information to create a proxy for bribes. The World Bank data collection is carried out independently, is unconnected with government officials, and ensures the anonymity of participants. This minimizes measurement errors in the *Bribes(p)* variable. Also, in referring to ‘establishments like this’ the question serves to elicit candid responses (Billon and Gillanders 2016). This indirect formulation of the question implies that respondents are not asked to admit that their firm has paid bribes but to gauge the behavior of similar types of firms. Although seemingly less precise, this approach reduces self-censorship biases, which proliferate in socially sensitive research.

Table 1 in the Appendix presents definitions and metrics for all variables included in the models.

Table 3 panel A shows descriptive statistics of the full study sample. Panel B reports descriptive statistics of the sample, based on the size (large, medium, small and micro) of respective firms. Approximately 24% of firms are large firms, 34% are medium size, and 42% belong to the small and micro category. Table 3 shows that the mean value of the product and process innovation is 64% and 47% respectively for the full sample. However, Panel B illustrates that the percentage of firms engaged with firm level innovations are reduced

according to firm size. In fact, the mean value indicates that about 72% of large firms are engaged in product innovations and 52% of large firms are engaged in process innovations. Meanwhile, only 56% of micro firms are engaged in product innovations and 36% in process innovations. Of the entire sample, about 18% of annual sales are paid out in the form of informal payments or gifts to public officials. This indicates the existence of a high level of corruption, where firms make some form of payment and regular payment. This percentage is higher in large firms, which is approximately 19%. Panel B shows that medium size firms pay 18%, small firms pay 17%, and micro firms pay 13% of their annual sales as bribes. A reason for this pattern may be that large firms, because they are likely to have more opportunities to interact with government officials, encounter more instances of requests for bribes from government officials. In the same vein, micro and small firms have far less contact with government officials and therefore lack the same opportunities to engage in corruption.

<<INSERT Table 3 here>>

4 Method

Given the dichotomous nature of innovation dependent variables (*Product_Inno* and *Process_Inno*), we use a discrete probability model

$$Pr(Inno_{i,j} = 1) = \Phi(\alpha + \beta Bribes\%_{i,j} + Control\ Variables + \varepsilon_{i,j}) \quad (1)$$

where *Innovation* is a dummy variable which takes the value of one if the firm introduced a new or significant improvement of the product (*Product_Inno*) or process (*Process_Inno*), and zero otherwise. The subscript *i* and *j* refer to firm and country respectively. Φ is the standard normal cumulative distribution. *Bribes%*_{*i,j*} refers to the informal payment percentage for the firm. The range of firm and country controls are included as the vector of control variables.

4.1 Endogeneity

The main concern in micro econometric analyses on the impact of corruption is endogeneity.

Apart from the omitted variable bias and measurement errors that occur when corruption is jointly determined with innovation, there is a simultaneity problem, which leads to a spurious relationship between innovation and corruption (Hewa Wellalage et al. 2019). Bribe payments may contribute to a reduction in the innovation of firms as they pose additional costs to firms. A bribe effectively amounts to a tax on firms and so creates an obstacle to innovation. Conversely, highly innovative firms may use a high level of corruption to overcome bureaucratic processes and complex regulations in accessing limited resources. A firm's decisions to pay bribes and engage in innovations can be unobservable, and these unobservable factors would be correlated with both innovation and bribes simultaneously, giving rise to an endogeneity problem.

Angrist and Krueger (2001) explain that instrumental variables (IVs) methods are widely used to control simultaneity and measurement errors in an endogenous relationship. In addition to this, recent studies use IVs to overcome an omitted variables problem (Fisman and Svensson 2007; Qi and Ongena 2018; Wellalage et al. 2018). We follow the approach of Fisman and Svensson (2007) and Qi and Ongena (2018), using the Bribes(p) by a locality-sector average of bribery (Avg_Bribes(p)) as an instrument.

For each country, we have firms in 4 localities and 27 industries. Altogether, our sample contains 108 variations in the instrument for each country, indicating the strength of the instrument. Locality-sector average of bribes(p) captures the institutional environment in the locality, business methods, and rent extraction preferences of bureaucrats, which are factors exogenous to the firm. However, instrumentation of firm-level bribery existence by a locality-sector average of bribes(p) eliminates the omitted unobservable biases correlated with bribery at the firm level, but not at the locality-sector level (Qi and Ongena 2018). Econometrically,

the most common guideline for determining the relevance of selected instruments are in the first stage equation in the F-test for joint significance. If the F-test for joint significance exceeds the value 10, this indicates that the selected instrument is accepted as relevant.

Unlike control variables, an instrumental variable ($Avg_Bribes(p)$) is assumed to have no direct effect on the outcome (Innovation). Instead, the effect of the $Avg_Bribes(p)$ on the Innovation (Product_Inno and Process_Inno) is entirely mediated via its effect on treatment assignment. In other words, this indicates that our instrument variable is satisfied with an exclusion restriction condition (Heckman et al. 2006). To determine whether bribes(p) have a causal effect on innovation, we use the instrumental variables $Avg_Bribes(p)$.

5 Results

5.1 The impact of bribes on firm level innovation

Table 4 column 1 and 5 report probit estimation (marginal effect) results in assuming that the model contains no endogenous regressors. The marginal effects of our main variable of interest, $Bribes(p)$, turns out to be positive and significant for both product and process innovations, indicating that increases in bribe payments lead to higher firm level innovations. This incoherent positive effect may be due to the endogeneity of the $Bribes(p)$ variable. Also, prior studies allude to the way in which innovation studies are influenced by common factors of unobserved heterogeneity (See Savignac 2008). The unobservable risks for the firm on the outcome of innovation may also influence the bribe payment of the firm. On the one hand, innovative firms are more likely to be paying high bribes to facilitate innovative activities, while on the other hand, bribes may also discourage firms from engaging in innovation. To correct the problem of endogeneity mentioned above, we re-estimate Equation (1) using the instrumental variable approach.

In the first stage, the endogenous variable (Bribes(p)) is regressed on the instrument (Avg_Bribes(p)), along with any other exogenous variables (controls) and these are reported in column 2 and 6 of Table 4. We also report diagnostic tests for the weak instrument. We have only one endogenous regressor; therefore, we can compare Stock and Yogo's thresholds (in here 10%) and Cragg-Donald F statistic. Our Cragg-Donald F statistics (F statistics for Product Innovation is 144.71 and process innovation is 140.99) are higher than the Stock and Yogo's thresholds (16.38), enabling us to conclude that the instruments are not weak. The under-identification test (Anderson canon. corr. LM statistic) indicates that the instrument is relevant.

After correcting for the endogeneity bias, the coefficient of our main variable of interest turns out to be negative and significant (Column 3 and 7). This indicates that with every one unit (percent in the case of our study) increase in the Bribes(p) on average, there is a decrease of 0.036 (3.6% in here) in the probability of the product innovation taking the value of 1. Likewise, every one unit (again, percent in our study) increase in the Bribes(p) on average leads to a decrease of 0.039 (3.9% in here) in the probability of the process innovation taking the value of 1. This indicates that firms paying bribes are less likely to be innovative. Our finding supports the argument that bribing has a sand in the wheels effect on firm innovation (Fungáčová et al. 2015; Mauro 1995; Reinikka and Svensson 2005). In particular, the study by Fisman and Svensson (2007) found similar results in that a one percentage point increase in the bribery rate reduces firm growth in three percentage points. Based on the foregoing discussion of transaction costs, our results indicate that corruption increases the transaction costs, which hinders the potential scale and scope of the innovation. Based on the institutional theory argument, corruption may hinder innovation in two ways. First, when the corruption level increases, the public loses its trust in the system as a whole (Zelekha 2013). When trust in institutions is weak, firms become reluctant to proceed with risky, innovative projects which project ambiguous future profits. On the other hand, corruption is ex post opportunistic. In a

weak institutional environment, a public official can demand additional payments even after the initial undertaking because bribing behavior is not supported by any legal institutions. Thus, we accept our H1: Corruption has a negative impact on firm-level innovation.

A comparison between the two types of innovations shows that bribes have a particularly strong effect on process innovation outcome. In fact, the magnitude differences show that bribing has 0.3% more negative impact on process innovation than on product innovation. Process innovation may be difficult to identify compared to product innovation. The number of new product ideas or patents can be directly or partially evaluated by external parties. Therefore, the opaqueness and information asymmetry in product innovation is less, and consequently, decreases the ability of public officials to extract rent from product innovators than from process innovators. Thus, we can argue that the negative effect of bribing is higher in process innovation than in product innovation.

The Bribes(p) proxy is inherited from the survey question which purposefully contains the phrase 'establishments like this.' This implies that respondents are not asked to confess that their own establishment has paid bribes but to gauge the behavior of other establishments. Therefore, one can argue that any disclosures of bribery are not necessarily confessions of personal involvement. That said, such disclosures may still come from firms which may find themselves paying bribes in the near future (Birhanu et al. 2016), because when operating in the same institutional environment, non-bribe paying firms may be at a disadvantage. Because of the sensitive nature of bribery activities, prior research relies on this as a proxy for firm-level corruption. Therefore, following prior researchers (Ayyagari et al. 2014; Birhanu et al. 2016; Hewa Wellalage et al. 2019; Wellalage et al. 2018), we interpret this measure as firms that have paid bribes. However, skeptical readers can interpret this as firms that think their peers have paid bribes.

5.2 Heterogenous effects of bribery on firms

Columns 4 and 8 of Table 4 show the heterogeneous effects of bribery on innovation across firms.

The coefficient on Avg_Bribes(p), the main effect for micro and small firms, is negative and significant, indicating that bribery has a detrimental effect on the innovation performance of micro and small firms. In particular, this negative effect is higher in process innovation ($\beta = .076$) than in product innovation ($\beta = .075$). In product innovation results, for large and medium size firms, the marginal effects of the bribes variable and of the interaction between the bribes variable and the two size variables cancel each other out (i.e. $\beta = -.075 + \beta = .075$). In other words, the net effect is inexistence. In relation to process innovation, for large firms, the marginal effects of the bribes and size interaction are slightly positive ($\beta = -.076 + \beta = .077$) and for medium firms, the marginal effects of the bribes and size interaction cancel each other out (i.e. $\beta = -.076 + \beta = .076$).

Table 5 reports the sub-sample analysis. Large firms show an insignificant but positive effect of bribes on firm product innovation ($\beta = .021$). When firm size decreases, the significant negative effect of Bribes(p) gets stronger (β values for firms are as follows: medium firm = $-.043$, small firm = $-.048$ micro firm = $-.058$). Large and medium size firms show a positive relationship between bribes and firm process innovation, while small and micro firms show a negative relationship. This indicates that large and medium size firms are not as severely affected as micro and small firms and potentially, they may even gain from bribery activities.

The institutional theory argument holds that when operators in micro and small firms are in a weak institutional environment and have weak property rights, the smallness of their organisations increases the risk of government interference and expropriation (Zhou and Peng 2012). Put simply, large firms may be less vulnerable to corruption than smaller firms. In

addition, the negative impact of bribery on small firms is especially pronounced because the same amount of money paid for a bribe drains a disproportionately larger chunk of resources for smaller firms than for larger firms (Zhou and Peng 2012). Thus, we accept our H2: An increase in firm size positively moderates the relationship between paying bribes and firm-level innovation.

<<INSERT Table 4 here>>

<<INSERT Table 5 here>>

5.3 Analysis of institutional environment effects on the relationship between bribes and innovation

Rather than considering country level institutional environment factors and categorizing institutional environments as weak or strong, we used micro level factors to determine institutional quality. In this study, we used micro and small firms as well, so even in strong institutional environments, they may encounter obstacles to their business growth. For such cases, the quality of the institutional environment at country level is not appropriate for micro and small firms. Therefore, we create an *Institutional quality* variable using levels of obstacles for the following five items: tax rate, tax administration, business licensing and permits, macroeconomic stability, and political instability.

The coefficient on Avg_Bribes(p), is negative and significant, indicating that regardless of firm institutional environment quality, bribery has a sand in the wheels effect on the product innovation ($\beta = -.009$) and process innovation ($\beta = -.009$) of firms. The significant and negative relationship is shown in interaction variable (Bribes(p) X level of institutional weakness) and firm level product innovations ($\beta = -.119 + -.009 = -.128$) and process innovations ($\beta = -.113 + -.009 = -.122$). This indicates that bribes have more detrimental effects in firms in weak institutional environments compared to firms in strong institutional environments. The results

show that the negative effect is slightly more compelling in product innovation than in process innovation. Firms engaging in product innovation often need patents and licensing, requirements which necessitate interactions with government officials, more so than firms engaging in process innovation. In a weak institutional environment, government officials hold a high level of discretion over law enforcement and resource allocation (Zhu et al. 2012) which provides more opportunities for public officials to abuse their power for private benefits and to solicit illegal payments from firms (Murphy et al. 1993).

<<INSERT Table 6 here>>

5.4 The relationship between innovation and bribes at different levels of sector-locality corruption

The relationship between bribery and firm level innovation varies according to the level of corruption in the operating sector and locality. Figure 1 shows the predicted probabilities of being a product or process innovator and paying bribes³ at different levels of sector-locality corruption. Figure 1 (A) shows that regardless whether a firm pays bribes or not, (bribes=1 or 0), the probability of product innovation decreases for firms when they operate in severely corrupt sectors and localities. The impact is high for non-bribe paying firms. For bribe paying firms, this ranges from 77.7% probability of innovation when Sec_Avg is reported at its minimum level to 39.5% when Sec_Avg is at its maximum level with a confidence level of 95%. For non-bribe paying firms, this ranges from 72% probability of innovation when sec_Avg is reported at its minimum level to 19.6% when Sec_Avg is at its maximum level with a confidence level of 95%.

³ Determines the existence of a bribe or not: 1= yes; 0= otherwise.

Figure 1 (B) shows that the probability of process innovation increases for firms engaged in bribes (bribes=1) when they operate in highly corrupted sector and locality, ranging from 53.1% probability of innovation when Sec_Avg is reported at its minimum level to 64.7% when Sec_Avg is at its maximum level with a confidence level of 95%. In contrast, for non-bribe paying firms, the probability of process innovation decreases, ranging from 49.6% probability of innovation when Sec_Avg is reported at its minimum level to 29% when Sec_Avg is at its maximum level with a confidence level of 95%.

Overall, this result indicates that the firms paying bribes acquire a competitive advantage over other firms when they operate in highly corrupt sectors and localities. This result suggests an indirect “greasing” effect of bribes in overcoming bureaucratic barriers.

<<INSERT Figure 1 in here>>

Table 7 reports the summary of empirical results and related hypotheses.

<<INSERT Table 7 in here>>

6 Robustness

(i) Alternative proxy of bribes

We use an alternative measure of bribes as a proxy for Bribe(p) and re-run the regression. The alternative measure *Obs_Corruption* takes value one if firms report that corruption is a major or very severe obstacle to the current operations of the firm, and zero otherwise. The descriptive statistics indicate that 66.15% of sample firms have indicated that corruption is a major or very severe obstacle to their current operations. We first use baseline probit to estimate the effect of corruption on firm-level innovation, assuming an exogenous relationship between explanatory variables and Product_Inno and Process_Inno. However, to address endogeneity bias, we use the locality-industry sector average of the total time spent by senior management

on dealing with requirements imposed by government regulations in a typical week (*Avg_Mgr_time*) as an instrumental variable. This is based on the reasoning that, in an environment with high levels of corruption, managers need to spend more time dealing with regulations.

The estimates of the coefficients in the regressions with the new bribery proxy are negative and significant -- similar to those derived from our baseline regression. However, the marginal effects found in our results indicate that the IVProbit magnitudes are slightly smaller. This suggests that our results are not influenced by the construction of the bribery proxy.

<<INSERT Table 8 here>>

(ii) Propensity score matching using bribes as a dummy variable

Nonresponsive and measurement errors are clearly visible in the bribe amount variables (Ayyagari et al., 2014). Therefore, we reanalyze our data using a dichotomous variable of bribes, instead of using *Bribes(p)*. Innovative firms may exaggerate or under report bribe amounts, therefore, a dummy variable of ‘bribes’ may be more accurate.

Using the propensity score method (PSM), we compare the innovation of firms exposed to no treatment $T=0$ (non-bribe payments) and the innovation of firms exposed to treatment $T=1$ (paying bribes). Since only one of these outcomes is observed for each firm, we estimate the average treatment on the treated (ATT), that is, the difference in innovation between those treated and those with the same probability of being treated (Cox-Edwards and Rodríguez-Oreggia 2009; Ichino et al. 2008). We use four matching methods: stratification matching; nearest neighbor matching; kernel matching; and radius matching, following (Becker and Ichino 2002). The standard error for the ATT is calculated, using a bootstrapping procedure, from the standard deviation of the ATT involving 100 bootstrap replications. In all matching models, the ATT is significant and negative for both product and process innovations. For

product innovations, the magnitude of the ATT ranges from 0.105 with the radius matching method to 0.115 with the nearest neighbor matching method. This indicates that the estimated average negative effect of paying bribes on product innovation outcomes for firms which engaged in bribery is 11%. The estimated average negative effect of paying bribes on process innovation for firms which engaged in bribery is 10% (excluding nearest neighbor matching). By comparing the propensity score method with IV probit, we see that the same order of magnitude in IV probit and ATT, confirming the robustness of our main results.

<<INSERT Table 9 here>>

We report the covariate balance summary for product and process innovations to confirm the validity of our PSM model. The matched sample results (Table 2 in the Appendix) indicates that matching on the estimated propensity score balanced the covariates. In “Matched” group, the standardized differences are all close to zero for each covariate for product and process innovation. This confirms that the covariate balance is satisfied.

Also, we run a series of t-tests comparing the means of the treatment and control group variable by variable after the matching. The mean differences between treated and control variables are not significant at the 5% level, which indicates all covariates are balanced after matching. This is about what we expect by chance, so it seems randomization has succeeded in generating balance.

(iii) Instrumental variables (2SLS) regression with R&D as a dependent variable

We verify our results using R&D as our innovation proxy. Studies to date largely use R&D expenses as an indirect proxy of innovations (see (Dreher and Gassebner 2013; Lööf and Nabavi 2016; Mancusi and Vezzulli 2010). Although our results were insignificant for both OLS and 2SLS, the coefficient sign indicates a significant negative relationship between Bribes(p) and R&D expenses.

<<INSERT Table 10 here>>

7 Discussion and conclusion

This study empirically examines the relationship between corruption and firm level innovation in Latin American firms. We further analyze the heterogeneity effects of bribery on the types of innovations across firms. In addition, this paper examines whether a weak institutional environment increases the adverse effects of corruption on firm level innovation. Although corruption is a major concern in emerging economies, how corruption affects innovation has received limited attention. The results suggest that bribing has a significant negative effect on firm level innovation. The negative effect is slightly higher (0.3%) for process innovation than for product innovation. This result is robust to various specifications.

Further, we find that the increases in firm size positively moderate the relationship between paying bribes and firm level innovation. In other words, corruption has a more detrimental effect on small firm innovation than on large firm innovation. In the results for product innovation for large and medium size firms, the marginal effects of the bribes variable and of the interaction between the bribes variable and the two size variables cancel out. In the results for process innovation for large firms, the marginal effects of the bribes and size interaction are slightly positive whereas for medium firms, the marginal effects of the bribes and size interaction cancel out. The interaction terms on bribes and weak institutional environment have a negative and significant effect on the product and process innovation proxies, implying that a weaker institutional environment magnifies the negative effect of bribes on innovation. Finally, we find that the relationship between bribery and firm level innovation may vary according to the operating sector of the firm and the corruption level of its locality.

Overall, we find that corruption negatively affects firm level innovations, which aligns with the view of development experts. Even though corruption is less damaging for large firms, it remains a pervasive obstacle to a nation's overall economic and social development. This obstacle can be overcome by having strong regulatory and governance mechanisms, as negative effects of corruption are magnified in weak institutional structures. A weak institutional environment can lead to corruption and may encourage the practice of high levels of bribery to circumvent red tape, nepotism and political favoritism. A highly professional civil service with attractive fringe benefits and improved government regulatory systems with more protocols can control corruption in this region. Practical interventions to promote strong institutional governance are not easily established when corruption is endemic. Greater awareness in the international community of the extortionist environment of companies in these countries may bring pressures to bear for reform.

There are consequences for adopting institutional level strategies in product and process development when engaging in bribery in a weak institutional environment. To encourage entrepreneurs to engage in more innovative activities, emerging markets should have well-structured national level policies and efficient administrative mechanisms, such as preferential tax structures, efficient customs procedures, a consistent policy framework, and stringent regulations to eliminate malpractices by reducing bribery that adversely impacts the country's innovations and firm performances. Once there is more robust legislative control and monitoring of corruption, this will allow firms to mitigate any discriminatory treatment of bureaucrats without resorting to bribery. Ultimately, this will erode the relationship between corruption and innovation in emerging markets.

Our study has limitations, some of which may be possible avenues for future research. In this paper, we contribute to the firm-level empirical research on bureaucratic corruption and firm innovation. Future studies can consider macro level cross country analysis on the

incidence and depth of bribery and its relationship to levels of innovation. Also, as future data on corruption becomes more available, studies may seek to further unbundle the types of corruptions that exist and to establish more proxies for innovation. This would allow researchers to further their understanding of the differences between the various types of bribes that exist and their respective impacts on innovations.

Compliance with Ethical Standards

Conflict of interest: The authors declare that they have no conflict of interest.

Appendix

Table 1: Definition and measurement of variables included in the models.

Explanatory Variables	Definition	Measurement
<i>Product_Inno</i>	Product innovation designates that a firm has introduced new or significantly improved products or services during the last three years	1 = yes; 0 = otherwise
<i>Process_Inno</i>	The process innovation designates that a firm has introduced new or significantly improved methods of manufacturing products or offering services during the last three years	1 = yes; 0 = otherwise
<i>Bribes(p)</i>	Total annual bribe payments as a percentage of firm annual sales	Percentage
<i>Formal_Fund</i>	The proportion of the firm's working capital financed by banks and non-bank financial institutions	Percentage
<i>Informal_Fund</i>	The proportion of the firm's working capital financed by trade credit, money lenders, friends and family	Percentage
<i>Own_Fund</i>	The proportion of the firm's working capital financed by internal funds or retained earnings	Percentage
<i>Sole_Prop</i>	Legal status is a sole proprietorship	1 = yes; 0 = otherwise
<i>Partnership</i>	Legal status is a partnership or limited partnership	1 = yes; 0 = otherwise
<i>Company</i>	Legal status is a shareholding company (private or public).	1 = yes; 0 = otherwise
<i>Other_firms</i>	Legal status is other than that of the above three groups	1 = yes; 0 = otherwise
<i>Domestic_Owned</i>	Percentage owned by private domestic individuals, companies or organizations	Percentage
<i>Foreign_Owned</i>	Percentage owned by private foreign individuals, companies or organizations	Percentage
<i>Government_Owned</i>	Percentage owned by the government or state	Percentage
<i>Other_Owned</i>	Percentage owned by others outside of the above three groups	Percentage
<i>Female_own</i>	At least one female owner, amongst the owners of the firm	1 = yes; 0 = otherwise
<i>Firm_age</i>	Number of years from the date of establishment	Years
<i>Email</i>	Firm has work email	1 = yes; 0 = otherwise

<i>Website</i>	Firm has company website	1 = yes; 0 = otherwise
<i>Political_Instability</i>	Severity of political instability as an obstacle to the current operations of the firm	0 to 4
<i>Court_Instability</i>	Severity of impartial court system as an obstacle to the current operations of the firm	0 to 4
<i>Manufacturing</i>	Firm representing the manufacturing industry	1 = yes; 0 = otherwise
<i>Retail</i>	Firm representing the retail industry	1 = yes; 0 = otherwise
<i>Other_Industry</i>	Firm representing an industry other than manufacturing and retail	1 = yes; 0 = otherwise
<i>Large</i>	100 or more employees	1 = yes; 0 = otherwise
<i>Medium</i>	Between 20 and 99 employees	1 = yes; 0 = otherwise
<i>Small</i>	Between 6 and 19 employees	1 = yes; 0 = otherwise
<i>Micro</i>	Between 0 to 5 employees	1 = yes; 0 = otherwise

Table 2: Covariate balance summary

	Product innovation		Process innovation	
	Raw	Matched	Raw	Matched
Number of Obs	3980	7960	3978	7956
Treated Obs	2700	3980	2697	3978
Control Obs	1280	3980	1281	3978

	Product Innovation Standardized differences		t-test (Treated and control)*	Process Innovation Standardized differences		t-test (Treated and control)*
	Raw	Matched		Raw	Matched	
Formal_Fund	.078	.019	-.480 (.631)	.078	-.019	.977 (.482)
Infomal_Fund	.143	-.026	-1.56 (.116)	.144	-.008	-.783 (.433)
Own_Fund	-.178	-.001	1.71 (0.09)	-.180	.018	.174 (.298)
Sole_Prop	-.061	.002	-1.18 (.237)	-.061	-.017	-1.26 (.207)
Partnership	-.028	-.007	.575 (.565)	-.026	-.011	.263 (.206)
Company	.066	.007	.015 (.988)	.065	.022	-.570 (.568)
Domestic_Own	.061	.001	-1.54 (.121)	.059	.022	-.664 (.506)
Foreign_Own	-.032	-.007	.875 (.600)	-.030	-.009	.832 (.405)
Government_Own	.034	.011	.381 (.174)	.034	.003	.264 (.237)
Female_Own	.001	-.003	.965 (.334)	-.001	-.002	-.146 (.882)
Firm_Age	-.016	-.007	.787 (.430)	-.017	-.016	-.451 (.651)
Email	.117	-.002	-.474 (.635)	.116	-.001	-.804 (.421)
Website	.156	.004	1.57 (.116)	.156	.015	.547 (.110)
Weak_Intuitional Quality	-.397	.005	-.245 (.806)	-.397	.020	.230 (.817)
Manufacturing	.113	-.024	-.374 (.708)	.117	-.011	-.240 (.809)
Retail	-.134	.013	.549 (.109)	-.133	-.025	1.62 (.104)
Other_Industry	-		.781 (.075)			1.62 (.272)
Large	.029	.025	.542 (.587)	.025	.008	1.62 (.140)

Medium	.018	-.027	1.60 (.109)	.021	-.015	1.62 (.497)
Small& Micro	-	-	-1.60 (.109)	-	-	1.62 (.497)

*t-tests comparing the means of the treatment and control group variable by variable after the matching. P values in parentheses

References

- Acemoglu D, Aghion P, Zilibotti F (2006) Distance to frontier, selection, and economic growth *Journal of the European Economic Association* 4:37-74
doi:10.1162/jeea.2006.4.1.37
- Acemoglu D, Verdier T (2000) The choice between market failures and corruption *American economic review* 90:194-211 doi:DOI: 10.1257/aer.90.1.194
- Agrawal A, Knoeber CR (2001) Do some outside directors play a political role? *The Journal of Law and Economics* 44:179-198 doi:10.1086/320271
- Angrist J, Krueger AB (2001) Instrumental variables and the search for identification: From supply and demand to natural experiments *Journal of Economic Perspectives* 15:69-85 doi:10.1257/jep.15.4.69
- Anokhin S, Schulze W (2009) Entrepreneurship, innovation, and corruption *Journal of Business Venturing* 24:465–476 doi:<https://doi.org/10.1016/j.jbusvent.2008.06.001>
- Avnimelech G, Zelekha Y, Sharabi E (2014) The effect of corruption on entrepreneurship in developed vs non-developed countries *International Journal of Entrepreneurial Behavior, Research & Politics* 20:237-262 doi:<https://doi.org/10.1108/IJEBr-10-2012-0121>
- Ayyagari M, Demirgüç-Kunt A, Maksimovic V (2011) Firm innovation in emerging markets: the role of finance, governance, and competition *Journal of Financial and Quantitative Analysis* 46:1545-1580
- Ayyagari M, Demirgüç-Kunt A, Maksimovic V (2014) Bribe payments and innovation in developing countries: Are innovating firms disproportionately affected? *Journal of Financial and Quantitative Analysis* 49:51-75
- Barasa L (2018) Corruption, transaction costs, and innovation in Africa *African Journal of Science, Technology, Innovation* 10:811-821
doi:<https://doi.org/10.1080/20421338.2018.1519061>
- Becker SO, Ichino A (2002) Estimation of average treatment effects based on propensity scores *The stata journal* 2:358-377
- Billon S, Gillanders R (2016) State ownership and corruption *International Tax and Public Finance* 23:1074–1092
- Birhanu AG, Gambardella A, Valentini G (2016) Bribery and investment: Firm-level evidence from Africa and Latin America *Strategic Management Journal* 37:1865-1877 doi: <https://doi.org/10.1002/smj.2431>
- Chadee D, Roxas B (2013) Institutional environment, innovation capacity and firm performance in Russia *Critical perspectives on international business* 9:19-39
doi:doi.org/10.1108/17422041311299923
- Chizema A, Pogrebna G (2019) The impact of government integrity and culture on corporate leadership practices: Evidence from the field and the laboratory *The Leadership Quarterly* 30:101303 doi:<https://doi.org/10.1016/j.leaqua.2019.07.001>

- Clarke GR, Xu LC (2004) Privatization, competition, and corruption: how characteristics of bribe takers and payers affect bribes to utilities *Journal of Public economics* 88:2067-2097 doi:[10.1016/j.jpubeco.2003.07.002](https://doi.org/10.1016/j.jpubeco.2003.07.002)
- Cox-Edwards A, Rodríguez-Oreggia E (2009) Remittances and labor force participation in Mexico: an analysis using propensity score matching *World Development* 37:1004-1014 doi:<https://doi.org/10.1016/j.worlddev.2008.09.010>
- Cuervo-Cazurra A (2006) Who cares about corruption? *Journal of International Business Studies* 37:807-822
- D'Este P, Iammarino S, Savona M, von Tunzelmann N (2012) What hampers innovation? Revealed barriers versus deterring barriers *Research policy* 41:482-488 doi:<https://doi.org/10.1016/j.respol.2011.09.008>
- Deegan C (2002) Introduction: The legitimising effect of social and environmental disclosures—a theoretical foundation *Accounting, Auditing and Accountability Journal* 15:282-311 doi:<https://doi.org/10.1108/09513570210435852>
- Deegan C (2014) *Financial accounting theory*. McGraw-Hill Education (Australia) Pty. Ltd, North Ryde, N.S.W. :
- Dincer O (2019) Does corruption slow down innovation? Evidence from a cointegrated panel of US states *European Journal of Political Economy* 56:1-10 doi:<https://doi.org/10.1016/j.ejpoleco.2018.06.001>
- Dreher A, Gassebner M (2013) Greasing the wheels? The impact of regulations and corruption on firm entry *Public Choice* 155:413-432 doi:<https://doi.org/10.1007/s11127-011-9871-2>
- Fan JP, Huang J, Morck R, Yeung B (2017) Institutional determinants of vertical integration in China *Journal of Corporate Finance* 44:524-539 doi:<https://doi.org/10.1016/j.jcorpfin.2014.05.013>
- Fisman R, Svensson J (2007) Are corruption and taxation really harmful to growth? Firm level evidence *Journal of Development Economics* 83:63-75 doi:[10.1016/j.jdeveco.2005.09.009](https://doi.org/10.1016/j.jdeveco.2005.09.009)
- Fungáčová Z, Kochanová A, Weill L (2015) Does money buy credit? Firm-level evidence on bribery and bank debt *World Development* 68:308-322 doi:[10.1016/j.worlddev.2014.12.009](https://doi.org/10.1016/j.worlddev.2014.12.009)
- Gaviria A, Pagés C (2002) Patterns of crime victimization in Latin American cities *Journal of Development Economics* 67:181-203 doi:[https://doi.org/10.1016/S0304-3878\(01\)00183-3](https://doi.org/10.1016/S0304-3878(01)00183-3)
- Glynn MA, Abzug R (2002) Institutionalizing identity: Symbolic isomorphism and organizational names *Academy of Management Journal* 45:267-280 doi:<https://doi.org/10.5465/3069296>
- Goedhuys M, Mohnen P, Taha T (2016) Corruption, innovation and firm growth: firm-level evidence from Egypt and Tunisia *Eurasian Business Review* 6:299-322 doi:<https://doi.org/10.1007/s40821-016-0062-4>
- Gorodnichenko Y, Schnitzer M (2013) Financial constraints and innovation: Why poor countries don't catch up *Journal of the European Economic Association* 11:1115-1152 doi:[10.1111/jeea.12033](https://doi.org/10.1111/jeea.12033)
- Habiyaremye A, Raymond W (2013) Transnational corruption and innovation in transition economies
- Habiyaremye A, Raymond W (2018) How do foreign firms' corruption practices affect innovation performance in host countries? Industry-level evidence from transition economies *Innovation* 20:18-41 doi:<https://doi.org/10.1080/14479338.2017.1367626>

- Hanousek J, Kochanova A (2016) Bribery environments and firm performance: Evidence from CEE countries *European Journal of Political Economy* 43:14-28
doi:<https://doi.org/10.1016/j.ejpoleco.2016.02.002>
- Heckman JJ, Urzua S, Vytlačil E (2006) Understanding instrumental variables in models with essential heterogeneity *The Review of Economics and Statistics* 88:389-432
doi:<https://doi.org/10.1162/rest.88.3.389>
- Hewa Wellalage N, Locke S, Samujh H (2019) Firm bribery and credit access: Evidence from Indian SMEs *Small Business Economics* Forthcoming
doi:<https://doi.org/10.1007/s11187-019-00161-w>
- Hoskisson RE, Eden L, Lau CM, Wright M (2000) Strategy in emerging economies. *The Academy of Management Journal* 43:249-267
- Hsieh C-T, Huang H-C, Lee W-L (2016) Using transaction cost economics to explain open innovation in start-ups *Management Decision* 54:2133-2156
doi:<https://doi.org/10.1108/MD-01-2016-0012>
- Ichino A, Mealli F, Nannicini T (2008) From temporary help jobs to permanent employment: What can we learn from matching estimators and their sensitivity? *Journal of applied econometrics* 23:305-327 doi:<https://doi.org/10.1002/jae.998>
- Kabadurmuş FNK (2017) Corruption and Innovation: The Case of EECA Countries *Journal of Entrepreneurship and Innovation Management* 6:51-71
- Kasuga H (2013) Why Do Firms Pay Bribes? Firm-Level Evidence from the Cambodian Garment Industry *Journal of International Development* 25:276-292
- Keig DL, Brouthers LE, Marshall VB (2015) Formal and informal corruption environments and multinational enterprise social irresponsibility *Journal of Management Studies* 52:89-116 doi: <https://doi.org/10.1111/joms.12102>
- Khwaja AI, Mian A (2005) Do lenders favor politically connected firms? Rent provision in an emerging financial market *The Quarterly Journal of Economics* 120:1371-1411
doi:10.1162/003355305775097524
- Kotabe M, Jiang CX, Murray JY (2017) Examining the complementary effect of political networking capability with absorptive capacity on the innovative performance of emerging-market firms *Journal of Management* 43:1131-1156
doi:<https://doi.org/10.1177/0149206314548226>
- Krammer SMS (2017) Greasing the wheels of change: Bribery, institutions, and new product introductions in emerging markets *Journal of Management* 45:1889-1926
doi:<https://doi.org/10.1177/0149206317736588>
- Krastanova P (2014) Greasing the wheels of innovation: How corruption and informal practices of firms impact the level of innovation in Bulgaria *Central European University*
- Lambsdorff JG (2002) How confidence facilitates illegal transactions: An empirical approach *American Journal of Economics and Sociology* 61:829-853
- Lederman D, Messina J, Pienknagura S, Rigolini J (2014) Latin American Entrepreneurs: Many Firms but Little Innovation. International Bank for Reconstruction and Development, The World Bank, Washington,DC
- Lööf H, Nabavi P (2016) Innovation and credit constraints: evidence from Swedish exporting firms *Economics of Innovation and New Technology* 25:269-282
doi:10.1080/10438599.2015.1076196
- Luo Y (2005) An organizational perspective of corruption *Management and Organization Review* 1:119-154 doi:
<https://doi.org/10.1111/j.1740-8784.2004.00006.x>

- Mahagaonkar P (2010) Corruption and innovation Money and ideas.
doi:https://doi.org/10.1007/978-1-4419-1228-2_5
- Mancusi ML, Vezzulli A R&D, innovation and liquidity constraints. In: CONCORD 2010 Conference, Sevilla, 2010. pp 3-4
- Manolova TS, Eunni RV, Gyoshev BS (2008) Institutional environments for entrepreneurship: Evidence from emerging economies in Eastern Europe
Entrepreneurship Theory and Practice 32:203-218 doi:<https://doi.org/10.1111/j.1540-6520.2007.00222.x>
- Mauro P (1995) Corruption and growth The quarterly journal of economics 110:681-712
doi:10.2307/2946696
- Melgar N, Rossi M, Smith TW (2010) The perception of corruption International Journal of Public Opinion Research 22:120-131 doi:<https://doi.org/10.1093/ijpor/edp058>
- Méndez F, Sepúlveda F (2009) What do we talk about when we talk about corruption? The Journal of Law, Economics 26:493-514
- Méon P-G, Sekkat K (2005) Does corruption grease or sand the wheels of growth? Public choice 122:69-97 doi:<https://doi.org/10.1007/s11127-005-3988-0>
- Murphy KM, Shleifer A, Vishny RW (1993) Why is rent-seeking so costly to growth? The American Economic Review 83:409-414
- Nguyen NA, Doan QH, Nguyen NM, Tran-Nam B (2016) The impact of petty corruption on firm innovation in Vietnam Crime, Law and Social Change 65:377-394
doi:<https://doi.org/10.1007/s10611-016-9610-1>
- O'Connell D (2016) Risk management: Transaction cost economics, outsourcing and corruption prevention Governance Directions 68:658
- O'Toole CM, Tarp F (2014) Corruption and the efficiency of capital investment in developing countries Journal of International Development 26:567-597
doi:<https://doi.org/10.1002/jid.2997>
- Paunov C (2016) Corruption's asymmetric impacts on firm innovation Journal of Development Economics 118:216-231
doi:<https://doi.org/10.1016/j.jdeveco.2015.07.006>
- Qi S, Ongena S (2018) Will money talk? Firm bribery and credit access Financial management Forthcoming:40 doi:<https://doi.org/10.1111/fima.12218>
- Reinikka R, Svensson J (2005) Fighting corruption to improve schooling: Evidence from a newspaper campaign in Uganda Journal of the European economic association 3:259-267 doi:10.1162/jeea.2005.3.2-3.259
- Rodriguez P, Uhlenbruck K, Eden L (2005) Government corruption and the entry strategies of multinationals Academy of management review 30:383-396
- Savignac F (2008) Impact of financial constraints on innovation: What can be learned from a direct measure? Econ Innov New Techn 17:553-569
doi:10.1080/10438590701538432
- Schiffer M, Weder B (2001) Firm size and the business environment: Worldwide survey results vol 43. World Bank and the International Finance Corporation, Washington, DC
- Scott WR (2001) Institutions and organizations. Sage Publications, Thousand Oaks, CA:
- Scott WR (2013) Institutions and Organizations. Ideas, Interests, and Identities. Stanford University, USA,
- Shleifer A, Vishny RW (1993) Corruption The quarterly journal of economics 108:599-617
- Suchman MC (1995) Managing legitimacy: Strategic and institutional approaches Academy of Management Review 20:571-610
- Tambunan T (2007) The role of government in technology transfer to SME clusters in Indonesia: Micro-level evidence from the metalworking industry cluster in Tegal,

- Central Java South East Asia Research 15:385-406
doi:<https://doi.org/10.5367/000000007782717768>
- Tanzi V, Davoodi HR (1997) Corruption, public investment, and growth. Washington, DC
United Nations (2006) Millennium Development Goals UNIS.
<http://www.unis.unvienna.org/unis/en/topics/2013/mdg.html>.
- Voskanyan F (2000) A study of The effects of Corruption on Economic and Political
Development of Armenia Un published Master thesis, American University Of
Armenia, USA
- Wei S-J, Kaufmann D (1999) Does Grease Money Speed Up the Wheels of Commerce? The
World Bank. doi:<https://doi.org/10.1596/1813-9450-2254>
- Wellalage NH, Fernandez V (2019) Innovation and SME finance: Evidence from developing
countries International Review of Financial Analysis
doi:<https://doi.org/10.1016/j.irfa.2019.06.009>
- Wellalage NH, Locke S, Samujh H (2018) Corruption, Gender and Credit Constraints:
Evidence from South Asian SMEs Journal of Business Ethics:1-14
doi:<https://doi.org/10.1007/s10551-018-3793-6>
- Williams N, Vorley T (2014) Economic resilience and entrepreneurship: lessons from the
Sheffield City Region Entrepreneurship Regional Development 26:257-281
doi:<https://doi.org/10.1080/08985626.2014.894129>
- Williamson OE (1979) Transaction-cost economics: the governance of contractual relations
The journal of Law and Economics 22:233-261
- Williamson OE (1987) Transaction cost economics: The comparative contracting perspective
Journal of economic behavior and organization 8:617-625
doi:[https://doi.org/10.1016/0167-2681\(87\)90038-2](https://doi.org/10.1016/0167-2681(87)90038-2)
- Zelekha Y (2013) The effect of immigration on entrepreneurship Kyklos 66:438-465
doi:<https://doi.org/10.1111/kykl.12031>
- Zhou JQ, Peng MW (2012) Does bribery help or hurt firm growth around the world? Asia
Pacific Journal of Management 29:907-921 doi:<https://doi.org/10.1007/s10490-011-9274-4>
- Zhu Y, Wittmann X, Peng MW (2012) Institution-based barriers to innovation in SMEs in
China Asia Pacific Journal of Management 29:1131-1142
doi:<https://doi.org/10.1007/s10490-011-9263-7>

Table 1: Recent literature related to corruption and innovation

<u>Author/s, Title</u> <u>Journal Name and Year</u>	<u>Journal name</u>	<u>Research Question</u>	<u>Data and</u> <u>Methodology</u>	<u>Findings/ Results</u>	<u>Conclusion</u>
<u>Dincer (2019)</u> <u>Does corruption slow</u> <u>down innovation?</u> <u>Evidence from a</u> <u>cointegrated panel of US</u> <u>states</u>	<u>European</u> <u>Journal of</u> <u>Political</u> <u>Economy</u>	<u>How corruption</u> <u>impacts innovative</u> <u>activities in the U.S.?</u>	<u>Annual data from 48</u> <u>contiguous U.S.</u> <u>states between 1977-</u> <u>2006</u> <u>Fully Modified</u> <u>Ordinary Least</u> <u>Squares (FMOLS)</u>	<u>Corruption hurts</u> <u>innovation.</u>	<u>Corruption does slow down</u> <u>innovation in long-run.</u>
<u>Kabadurmuş (2017)</u> <u>Corruption and</u> <u>Innovation: The Case of</u> <u>EECA Countries</u>	<u>Journal of</u> <u>Entrepreneurship</u> <u>and Innovation</u> <u>Management</u>	<u>Is sand</u> <u>or grease effect of</u> <u>corruption dominant</u> <u>for Eastern Europe and</u> <u>Central</u> <u>Asia region?</u>	<u>Firm-level data for</u> <u>27 Eastern European</u> <u>and Central Asian</u> <u>countries.</u> <u>Logit and Tobit</u> <u>regression.</u>	<u>Corruption has a positive</u> <u>effect on the rate of</u> <u>innovation</u>	<u>Grease effect of corruption</u> <u>is dominated in this region.</u>
<u>Krammer (2017)</u> <u>Greasing the wheels of</u> <u>change: Bribery,</u> <u>institutions, and new</u> <u>product introductions in</u> <u>emerging markets</u>	<u>Journal of</u> <u>Management</u>	<u>How bribing affects</u> <u>the innovative</u> <u>performance of firms</u> <u>in emerging markets</u> <u>and How this</u> <u>relationship moderates</u> <u>by the quality of the</u> <u>institutions?</u>	<u>Over 6,000 firms in</u> <u>30 emerging markets</u> <u>in Central Asia and</u> <u>Eastern Europe</u> <u>Probit model</u>	<u>Firm bribery positively</u> <u>impacts on their ability to</u> <u>introduce new products.</u> <u>However, it depends on</u> <u>the formal and informal</u> <u>pressures in firms'</u> <u>environments.</u>	<u>Bureaucratic obstacles and</u> <u>uncertainty help firms to be</u> <u>innovative through illegal</u> <u>cost minimization strategies.</u> <u>Norms and informal</u> <u>practices moderate the</u> <u>efficiency of firm strategies</u> <u>in emerging markets.</u>
<u>Nguyen et al. (2016)</u> <u>The impact of petty</u> <u>corruption on firm</u> <u>innovation in Vietnam</u>	<u>Crime, Law and</u> <u>Social</u> <u>Change</u>	<u>How petty corruption</u> <u>impact on firm</u> <u>innovation activities?</u>	<u>SMEs in Vietnam</u> <u>from 2005 to 2011</u> <u>Logit model.</u>	<u>Corruption in the form of</u> <u>bribery has a positive</u> <u>effect on firm innovation.</u>	<u>Bribery payments by</u> <u>innovative SMEs in</u> <u>Vietnam grease the wheel of</u> <u>the bureaucratic machinery.</u>

<u>Paunov (2016)</u> <u>Corruption's asymmetric impacts on firm innovation</u>	<u>Journal of Development Economics</u>	<u>How corruption impacts on the ownership of quality certificates and patents for smaller- and larger-sized firms?</u>	<u>Data for 48 developing and emerging countries for the period 2007–2011</u> <u>Linear probability regression</u>	<u>There is a negative effect between corruption and firms' ownership of quality certificates and it is stronger for smaller firms than larger firms.</u>	<u>Corruption is impacting on the firm's investment in innovation. Therefore, it is important to make policies to fight for corruption.</u>
<u>Goedhuys et al. (2016)</u> <u>Corruption, innovation and firm growth: firm-level evidence from Egypt and Tunisia</u>	<u>Eurasian Business Review</u>	<u>How corruption effect on innovative behavior of firms when doing severe bureaucratic and institutional obstacles exist?</u>	<u>Analyse 3489 firm-level data from Egypt and Tunisia.</u> <u>Conditional recursive mixed-process model (CMP)</u>	<u>Corruption has a direct negative effect on innovative firms, but a positive effect when there is an institutional obstacle.</u>	<u>Corruption can identify as a mechanism that bypasses the bureaucratic obstacles and resonate the firm growth and product innovations.</u>
<u>Ayyagari et al. (2014)</u> <u>Bribe payments and innovation in developing countries: Are innovating firms disproportionately affected?</u>	<u>Journal of Financial and Quantitative Analysis</u>	<u>How bribery is associated with firm innovation and whether innovating firms pay more bribes than non innovators?</u>	<u>25,000 firms in 57 countries</u> <u>OLS regression.</u>	<u>Innovating firms pay more bribes than non innovators. This difference is larger in countries with more bureaucratic regulation and weaker governance.</u>	<u>Innovators are more likely to be victims of corruption than criminals as they pay more in bribes than their gain by underreporting revenues to tax authorities. Corruption acts as a tax on innovation</u>
<u>Avnimelech et al. (2014)</u> <u>The effect of corruption on entrepreneurship in developed vs non-developed countries</u>	<u>International Journal of Entrepreneurial Behavior, Research & Politics</u>	<u>How corruption is linked to productive entrepreneurship in general and is it depends on countries' specific characteristics in particular?</u>	<u>Data set of entrepreneurial activity within 176 countries, collected from the professional networking site LinkedIn</u> <u>OLS regression</u>	<u>Highly corrupted countries usually face low levels of productive Entrepreneurship. Further, this negative effect is much more significant for developed countries than in developing countries and depends on country specific characteristic.</u>	<u>It is important to fight for corruption developed and western developing country to reduce the negative impact of it.</u>

<u>Anokhin and Schulze (2009)</u> <u>Entrepreneurship, innovation, and corruption</u>	<u>Journal of Business Venturing</u>	<u>How the level of corruption affects rates of entrepreneurial activities and innovation across nations?</u>	<u>Data from 64 countries for a period of 1996 to 2002</u> <u>Quantile regression</u>	<u>There is a positive concave relationship between the control of corruption and domestic entrepreneurial and innovative activities. Further, this relationship moderates by foreign direct investment.</u>	<u>Better control of corruption can help to improve the level of innovation and entrepreneurship.</u>
<u>(Mahagaonkar 2010)</u> <u>Corruption and innovation</u>	<u>Book Chapter Money and ideas</u>	<u>How corruption impacts firm level innovation?</u>	<u>Sample of 3477 firms in the African continent</u> <u>Probit and instrumental variable probit models.</u>	<u>Corruption has a negative impact on the product and organizational innovation, a positive impact on marketing innovations but no effect on process innovation.</u>	<u>Corruption is more disruptive for innovative firms due to increased cost of corruption, imperfect financial markets and corrupt officials.</u>

Table 2: Total Study Sample

<u>Country</u>	<u>Micro & Small</u>	<u>Medium</u>	<u>Large</u>	<u>Total</u>	<u>Bribes</u>	<u>CPI score in 2017*</u>
<u>Argentina</u>	<u>353</u>	<u>324</u>	<u>254</u>	<u>931</u>	<u>17%</u>	<u>39</u>
<u>Bolivia</u>	<u>181</u>	<u>98</u>	<u>82</u>	<u>361</u>	<u>27%</u>	<u>33</u>
<u>Colombia</u>	<u>396</u>	<u>357</u>	<u>214</u>	<u>967</u>	<u>20%</u>	<u>37</u>
<u>Ecuador</u>	<u>145</u>	<u>129</u>	<u>78</u>	<u>352</u>	<u>16%</u>	<u>32</u>
<u>El Salvador</u>	<u>346</u>	<u>224</u>	<u>147</u>	<u>719</u>	<u>11%</u>	<u>33</u>
<u>Guatemala</u>	<u>113</u>	<u>133</u>	<u>97</u>	<u>345</u>	<u>13%</u>	<u>28</u>
<u>Honduras</u>	<u>177</u>	<u>92</u>	<u>61</u>	<u>332</u>	<u>13%</u>	<u>29</u>
<u>Nicaragua</u>	<u>151</u>	<u>122</u>	<u>58</u>	<u>333</u>	<u>17%</u>	<u>26</u>
<u>Paraguay</u>	<u>118</u>	<u>129</u>	<u>115</u>	<u>362</u>	<u>19%</u>	<u>29</u>

<u>Peru</u>	<u>470</u>	<u>300</u>	<u>232</u>	<u>1002</u>	<u>33%</u>	<u>37</u>
<u>Uruguay</u>	<u>123</u>	<u>133</u>	<u>80</u>	<u>336</u>	<u>10%</u>	<u>70</u>
<u>Total observations</u> <u>/Average</u>	<u>2573</u>	<u>2041</u>	<u>1418</u>	<u>6040</u>	<u>17.9%</u>	

Corruption Perception Index- 2017. The index, which ranks 180 countries and territories by their perceived levels of public sector corruption according to experts and businesspeople, uses a scale of 0 to 100, where 0 is highly corrupt and 100 is very clean.

Table 3: Descriptive statistics

<u>Variables</u>	<u>Panel A</u> <u>Full firm sample</u>			<u>Panel B</u> <u>Large</u>			<u>Panel B</u> <u>Medium</u>			<u>Panel B</u> <u>*Small</u>			<u>Panel B</u> <u>*Micro</u>		
	<u>Obs</u>	<u>Mean</u>	<u>Std</u>	<u>Obs</u>	<u>Mean</u>	<u>Std</u>	<u>Obs</u>	<u>Mean</u>	<u>Std</u>	<u>Obs</u>	<u>Mean</u>	<u>Std</u>	<u>Obs</u>	<u>Mean</u>	<u>Std</u>
<u>Product_Inno</u>	<u>6040</u>	<u>.6361</u>	<u>.4812</u>	<u>1418</u>	<u>.7165</u>	<u>.4509</u>	<u>2041</u>	<u>.6619</u>	<u>.4732</u>	<u>2503</u>	<u>.5706</u>	<u>.4951</u>	<u>70</u>	<u>.5573</u>	<u>.5008</u>
<u>Process_Inno</u>	<u>6040</u>	<u>.4698</u>	<u>.4991</u>	<u>1418</u>	<u>.5719</u>	<u>.4950</u>	<u>2041</u>	<u>.4785</u>	<u>.4997</u>	<u>2503</u>	<u>.4067</u>	<u>.4913</u>	<u>70</u>	<u>.3770</u>	<u>.4887</u>
<u>Bribes(p)</u>	<u>5231</u>	<u>17.97</u>	<u>20.12</u>	<u>1215</u>	<u>19.03</u>	<u>20.70</u>	<u>1770</u>	<u>18.24</u>	<u>19.84</u>	<u>2180</u>	<u>17.29</u>	<u>20.03</u>	<u>66</u>	<u>13.25</u>	<u>19.19</u>
<u>Formal_Fund</u>	<u>5398</u>	<u>20.76</u>	<u>27.40</u>	<u>1400</u>	<u>25.08</u>	<u>28.76</u>	<u>1939</u>	<u>21.44</u>	<u>26.95</u>	<u>2000</u>	<u>17.85</u>	<u>26.63</u>	<u>59</u>	<u>15.76</u>	<u>26.39</u>
<u>Infomal_Fund</u>	<u>5398</u>	<u>28.65</u>	<u>30.85</u>	<u>1400</u>	<u>28.72</u>	<u>29.73</u>	<u>1939</u>	<u>29.16</u>	<u>30.06</u>	<u>2000</u>	<u>28.11</u>	<u>32.02</u>	<u>59</u>	<u>30.90</u>	<u>34.15</u>
<u>Own_Fund</u>	<u>5396</u>	<u>50.57</u>	<u>36.70</u>	<u>1400</u>	<u>46.22</u>	<u>35.44</u>	<u>1939</u>	<u>49.37</u>	<u>36.47</u>	<u>1998</u>	<u>53.99</u>	<u>37.23</u>	<u>59</u>	<u>53.34</u>	<u>38.83</u>
<u>Sole_Prop</u>	<u>5411</u>	<u>.0991</u>	<u>.2988</u>	<u>1400</u>	<u>.0218</u>	<u>.1462</u>	<u>1852</u>	<u>.0593</u>	<u>.2364</u>	<u>2098</u>	<u>.1685</u>	<u>.3744</u>	<u>61</u>	<u>.3606</u>	<u>.4842</u>
<u>Partnership</u>	<u>5411</u>	<u>.3110</u>	<u>.4629</u>	<u>1400</u>	<u>.1491</u>	<u>.3563</u>	<u>1852</u>	<u>.3182</u>	<u>.4659</u>	<u>2098</u>	<u>.3971</u>	<u>.4894</u>	<u>61</u>	<u>.3934</u>	<u>.4925</u>
<u>Company</u>	<u>5411</u>	<u>.5847</u>	<u>.4928</u>	<u>1400</u>	<u>.8271</u>	<u>.3782</u>	<u>1852</u>	<u>.6197</u>	<u>.4856</u>	<u>2098</u>	<u>.4249</u>	<u>.4944</u>	<u>61</u>	<u>.2459</u>	<u>.4341</u>
<u>Other_firms</u>	<u>5411</u>	<u>.0052</u>	<u>.0720</u>	<u>1400</u>	<u>.0019</u>	<u>.0435</u>	<u>1852</u>	<u>.0027</u>	<u>.0516</u>	<u>2098</u>	<u>.0094</u>	<u>.0968</u>	<u>61</u>	<u>0</u>	<u>0</u>
<u>Domestic_Own</u>	<u>5420</u>	<u>91.50</u>	<u>26.34</u>	<u>1400</u>	<u>82.19</u>	<u>36.18</u>	<u>1852</u>	<u>93.54</u>	<u>22.78</u>	<u>2004</u>	<u>95.15</u>	<u>20.84</u>	<u>61</u>	<u>93.61</u>	<u>22.95</u>
<u>Foreign_Own</u>	<u>5419</u>	<u>7.077</u>	<u>24.09</u>	<u>1400</u>	<u>16.64</u>	<u>35.21</u>	<u>1301</u>	<u>5.475</u>	<u>21.01</u>	<u>2004</u>	<u>2.962</u>	<u>15.84</u>	<u>61</u>	<u>3.115</u>	<u>15.00</u>
<u>Government_Own</u>	<u>5420</u>	<u>.1446</u>	<u>3.173</u>	<u>1400</u>	<u>.3605</u>	<u>5.19</u>	<u>1801</u>	<u>.1726</u>	<u>3.272</u>	<u>2004</u>	<u>0</u>	<u>0</u>	<u>61</u>	<u>0</u>	<u>0</u>

<u>Other Own</u>	<u>5419</u>	<u>1.295</u>	<u>10.97</u>	<u>1400</u>	<u>.9002</u>	<u>8.880</u>	<u>1801</u>	<u>.7801</u>	<u>8.327</u>	<u>2004</u>	<u>1.887</u>	<u>13.41</u>	<u>61</u>	<u>3.278</u>	<u>17.96</u>
<u>Female Own</u>	<u>5657</u>	<u>.6301</u>	<u>.4828</u>	<u>1219</u>	<u>.6359</u>	<u>.4814</u>	<u>1987</u>	<u>.6382</u>	<u>.4807</u>	<u>2400</u>	<u>20.80</u>	<u>14.78</u>	<u>51</u>	<u>.6067</u>	<u>.4925</u>
<u>Firm Age</u>	<u>6040</u>	<u>26.88</u>	<u>19.99</u>	<u>1418</u>	<u>37.10</u>	<u>24.73</u>	<u>2041</u>	<u>27.12</u>	<u>19.01</u>	<u>2503</u>	<u>.6209</u>	<u>.4852</u>	<u>70</u>	<u>25</u>	<u>12.80</u>
<u>Email</u>	<u>5656</u>	<u>.9525</u>	<u>.2128</u>	<u>1218</u>	<u>.9867</u>	<u>.1144</u>	<u>1988</u>	<u>.9787</u>	<u>.1444</u>	<u>2400</u>	<u>.9168</u>	<u>.2762</u>	<u>50</u>	<u>.7705</u>	<u>.4240</u>
<u>Website</u>	<u>5657</u>	<u>.7266</u>	<u>.4458</u>	<u>1218</u>	<u>.9146</u>	<u>.2796</u>	<u>1988</u>	<u>.7776</u>	<u>.4159</u>	<u>2400</u>	<u>.5831</u>	<u>.4932</u>	<u>50</u>	<u>.4590</u>	<u>.5024</u>
<u>Political Instability</u>	<u>5275</u>	<u>.4779</u>	<u>.4996</u>	<u>1400</u>	<u>.5682</u>	<u>.4955</u>	<u>1877</u>	<u>.4782</u>	<u>.4997</u>	<u>1940</u>	<u>.4276</u>	<u>.4948</u>	<u>58</u>	<u>.1929</u>	<u>.3981</u>
<u>Court Instability</u>	<u>5334</u>	<u>.1477</u>	<u>.3548</u>	<u>1400</u>	<u>.1851</u>	<u>.3885</u>	<u>1909</u>	<u>.1503</u>	<u>.3574</u>	<u>1968</u>	<u>.1221</u>	<u>.3276</u>	<u>57</u>	<u>.3793</u>	<u>.4894</u>
<u>Manufacturing</u>	<u>6040</u>	<u>.5023</u>	<u>.5000</u>	<u>1418</u>	<u>.5896</u>	<u>.4921</u>	<u>2041</u>	<u>.5276</u>	<u>.4994</u>	<u>2503</u>	<u>.4318</u>	<u>.4954</u>	<u>70</u>	<u>.4590</u>	<u>.5024</u>
<u>Retail</u>	<u>6040</u>	<u>.1772</u>	<u>.3819</u>	<u>1418</u>	<u>.1213</u>	<u>.3266</u>	<u>2041</u>	<u>.1371</u>	<u>.3440</u>	<u>2503</u>	<u>.2389</u>	<u>.4265</u>	<u>70</u>	<u>.3115</u>	<u>.4669</u>
<u>Other Industry</u>	<u>6040</u>	<u>.3203</u>	<u>.4667</u>	<u>1418</u>	<u>.2891</u>	<u>.4536</u>	<u>2041</u>	<u>.3353</u>	<u>.4722</u>	<u>2503</u>	<u>.3292</u>	<u>.4701</u>	<u>70</u>	<u>.2295</u>	<u>.4240</u>
<u>Large</u>	<u>6040</u>	<u>.2385</u>	<u>.4262</u>	<u>-</u>											
<u>Medium</u>	<u>6040</u>	<u>.3398</u>	<u>.4736</u>												
<u>Small</u>	<u>6040</u>	<u>.4078</u>	<u>.4914</u>												
<u>Micro</u>	<u>6040</u>	<u>.0137</u>	<u>.1166</u>												
<u>Instrumental variable</u> <u>Avg_Bribes(p)</u>	<u>4423</u>	<u>17.96</u>	<u>3.53</u>	<u>1055</u>	<u>17.96</u>	<u>3.56</u>	<u>1503</u>	<u>18.15</u>	<u>3.42</u>	<u>1804</u>	<u>17.83</u>	<u>3.57</u>	<u>61</u>	<u>17.16</u>	<u>4.00</u>

Table 4: Probit model with continuous endogenous regressors estimation results of innovation and bribes% relationship -Full sample

<u>Variables</u>	<u>Panel A: Product Innovation</u>				<u>Panel B: Process Innovation</u>			
	<u>Probit</u> <u>(Marginal</u> <u>effects)</u>	<u>First stage</u> <u>Probit</u>	<u>IV Probit</u> <u>dy/dx</u> <u>(Marginal</u> <u>effects)</u>	<u>IV Probit</u> <u>Interaction</u> <u>dy/dx</u> <u>(Marginal</u> <u>effects)</u>	<u>Probit</u> <u>(Marginal</u> <u>effects)</u>	<u>First stage</u> <u>Probit</u>	<u>IV Probit</u> <u>dy/dx</u> <u>(Marginal</u> <u>effects)</u>	<u>IV Probit</u> <u>Interaction</u> <u>dy/dx</u> <u>(Marginal</u> <u>effects)</u>
	<u>1</u>	<u>2</u>	<u>3</u>	<u>4</u>	<u>5</u>	<u>6</u>	<u>7</u>	<u>8</u>
	<u>Product Inno</u>	<u>Bribes(p)</u>	<u>Product Inno</u>	<u>Product Inno</u>	<u>Process Inno</u>	<u>Bribes(p)</u>	<u>Process Inno</u>	<u>Process Inno</u>
<u>Bribes(p)</u>	<u>.001**</u> <u>(.001)</u>				<u>.001</u> <u>(.001)</u>			
<u>+Avg Bribes(p)</u>		<u>1.13***</u> <u>(.094)</u>	<u>-.036**</u> <u>(.019)</u>	<u>-.075***</u> <u>(.012)</u>		<u>1.12***</u> <u>(.094)</u>	<u>-.039**</u> <u>(.016)</u>	<u>-.076***</u> <u>(.008)</u>
<u>Formal Fund</u>	<u>.001***</u> <u>(.001)</u>	<u>.047**</u> <u>(.012)</u>	<u>.002***</u> <u>(.001)</u>	<u>.002</u> <u>(.001)</u>	<u>.001**</u> <u>(.001)</u>	<u>.046***</u> <u>(.012)</u>	<u>.002**</u> <u>(.001)</u>	<u>.001</u> <u>(.001)</u>
<u>Infomral Fund</u>	<u>.001**</u> <u>(.001)</u>	<u>.015</u> <u>(.010)</u>	<u>.002**</u> <u>(.001)</u>	<u>.001</u> <u>(.001)</u>	<u>.001**</u> <u>(.001)</u>	<u>.013</u> <u>(.010)</u>	<u>.002*</u> <u>(.001)</u>	<u>.001</u> <u>(.001)</u>
<u>Own Fund</u>	=	=	=		=	=	=	
<u>Sole Prop</u>	<u>.085</u> <u>(.116)</u>	<u>-3.03</u> <u>(4.61)</u>	<u>-.166</u> <u>(.385)</u>	<u>-.474</u> <u>(.345)</u>	<u>.045</u> <u>(.135)</u>	<u>-2.87</u> <u>(4.64)</u>	<u>-.264</u> <u>(.143)</u>	<u>-.511**</u> <u>(.291)</u>
<u>Partnership</u>	<u>.067</u> <u>(.122)</u>	<u>-.474</u> <u>(.454)</u>	<u>-.084</u> <u>(.338)</u>	<u>-.282</u> <u>(.307)</u>	<u>.012</u> <u>(.132)</u>	<u>-.529</u> <u>(4.57)</u>	<u>-.183</u> <u>(.308)</u>	<u>-.319</u> <u>(.270)</u>
<u>Company</u>	=	=	=	=		=		
<u>Domestic Own</u>	<u>-.001</u> <u>(.006)</u>	<u>-.022</u> <u>(.030)</u>	<u>-.002</u> <u>(.002)</u>	<u>-.004**</u> <u>(.002)</u>	<u>.001*</u> <u>(.001)</u>	<u>-.021</u> <u>(.030)</u>	<u>-.001</u> <u>(.003)</u>	<u>-.003</u> <u>(.003)</u>
<u>Foreign Own</u>	<u>.002**</u> <u>(.001)</u>	<u>-.038</u> <u>(.033)</u>	<u>.001</u> <u>(.004)</u>	<u>-.002</u> <u>(.004)</u>	<u>.002**</u> <u>(.001)</u>	<u>-.039</u> <u>(.033)</u>	<u>.001</u> <u>(.004)</u>	<u>-.002</u> <u>(.004)</u>
<u>Government Own</u>	<u>-.002</u> <u>(.003)</u>	<u>.019</u> <u>(.125)</u>	<u>-.004</u> <u>(.008)</u>	<u>-.006</u> <u>(.007)</u>	<u>.004</u> <u>(.003)</u>	<u>.017</u> <u>(.126)</u>	<u>.006</u> <u>(.008)</u>	<u>-.002</u> <u>(.009)</u>
<u>Other Own</u>	=	=	=	=	=	=	=	=

<u>Female Own</u>	<u>.022</u> (.017)	<u>-.432</u> (.656)	<u>.027</u> (.048)	<u>-.008</u> (.052)	<u>.001</u> (.017)	<u>-.442</u> (.660)	<u>-.012</u> (.038)	<u>-.021</u> (.035)
<u>Firm Age</u>	<u>-.001</u> (.001)	<u>.026</u> (.080)	<u>-.001</u> (.001)	<u>-.001</u> (.001)	<u>-.001</u> (.001)	<u>-.005</u> (.002)	<u>-.005</u> (.001)	<u>-.002</u> (.001)
<u>Email</u>	<u>.101***</u> (.041)	<u>2.85**</u> (1.58)	<u>.349***</u> (.095)	<u>.458***</u> (.128)	<u>.193***</u> (.038)	<u>2.80**</u> (1.58)	<u>.495**</u> (.165)	<u>.507**</u> (.261)
<u>Website</u>	<u>.162***</u> (.019)	<u>-.072</u> (.746)	<u>.272</u> (.177)	<u>.110</u> (.251)	<u>.081***</u> (.019)	<u>-.035</u> (.752)	<u>.123</u> (.094)	<u>.056</u> (.119)
<u>Political Instability</u>	<u>-.033**</u> (.017)	<u>-.200**</u> (.890)	<u>.102**</u> (.047)	<u>.056</u> (.044)	<u>.001</u> (.012)	<u>.619</u> (.456)	<u>.075</u> (.051)	<u>.046</u> (.039)
<u>Fair Instability</u>	<u>.017</u> (.019)	<u>.055</u> (.176)	<u>.034</u> (.087)	<u>.040</u> (.070)	<u>-.036**</u> (.017)	<u>-1.89**</u> (.894)	<u>.042</u> (.083)	<u>.042</u> (.067)
<u>Manufacturing</u>	<u>.057***</u> (.018)	<u>.331</u> (.702)	<u>.087</u> (.078)	<u>.016</u> (.098)	<u>.077***</u> (.018)	<u>.171</u> (.706)	<u>.102</u> (.094)	<u>.024</u> (.116)
<u>Retail</u>	<u>.057**</u> (.022)	<u>.106</u> (.953)	<u>.049</u> (.102)	<u>-.025</u> (.112)	<u>-.001</u> (.024)	<u>.999</u> (.959)	<u>-.057</u> (.057)	<u>-.062</u> (.049)
<u>Other Industry</u>	=	=	=			=	=	
<u>Large</u>	<u>.047**</u> (.023)	<u>1.28</u> (.889)	<u>.118**</u> (.065)	<u>-1.30***</u> (.296)	<u>.093***</u> (.024)	<u>1.41</u> (.896)	<u>.184**</u> (.093)	<u>-1.32***</u> (.240)
<u>Medium</u>	<u>.039**</u> (.018)	<u>.130</u> (.739)	<u>.064</u> (.062)	<u>-1.31***</u> (.247)	<u>.032*</u> (.019)	<u>.224</u> (.744)	<u>.048</u> (.054)	<u>1.34***</u> (.170)
<u>Small & Micro</u>	=	=	=		=		=	
<u>Bribes(p)* Large</u>				<u>.075***</u> (.012)				<u>.077***</u> (.006)
<u>Bribes(p)* Medium</u>				<u>.075***</u> (.010)				<u>.076***</u> (.007)
<u>Constant</u>		<u>-3.33**</u> (5.90)				<u>-2.99**</u> (5.92)		
<u>Summary statistics</u>								
<u>Log likelihood</u>			<u>-19967</u>	<u>-18276</u>			<u>-20131</u>	<u>-18428</u>
<u>LR Chi2 /Wald Chi2</u>			<u>511.0</u>	<u>301.1</u>			<u>492.4</u>	<u>3940.5</u>

<u>P value</u>			<u>0.00</u>	<u>0.00</u>			<u>0.00</u>	<u>0.000</u>
<u>Sample size</u>		<u>4043</u>	<u>3966</u>	<u>3966</u>		<u>4041</u>	<u>3964</u>	<u>3964</u>
<u>Cragg-Donald Wald F statistic</u>		<u>144.7</u>				<u>141.0</u>		
<u>Stock-Yogo weak ID test critical values: 10% maximal IV size</u>		<u>16.38</u>				<u>16.38</u>		
<u>Anderson canon. corr. LM statistic</u>		<u>140.3***</u>				<u>136.8***</u>		
<u>Wald test of exogeneity</u>			<u>7.30**</u>	<u>4.19***</u>			<u>7.61**</u>	<u>4.18**</u>

Note for Table 4 and 5 :⁺ Avg *Bribes(p)* is an instrumental variable for *Bribe(p)*. ⁺⁺ (*) dy/dx is for discrete change of dummy variable from 0 to 1. The main dependent variable is a dummy variable taking one if the firm is innovative and zero otherwise. Panel A reports product innovation results and Panel B reports process innovation results. In each panel, column 1 reports the marginal effect of the probit model with exogenous regressors regression results; column 2 reports first stage probit results, column 3 reports the marginal effect of IV probit regression results. Column 4 reports the marginal effect of IV probit regression with the interaction of bribe payment and firm size. Bribe percentage is a main explanatory variable and IV probit use, locality-sector-average of bribe percentage as an instrument. These models provide standard errors, which are in parentheses. The Wald test of exogeneity is reported in the last row as a chi-squared statistic with 1 degree of freedom. * Significant at 10% level, **Significant at 5% level, ***Significant at 1% level.

Table 5: Probit and IV probit estimation results of innovation and bribes% relationship -Large, medium, and small and micro firms

Panel A : Large firms							
	Variables	Probit (Marginal effects)	First stage Probit	IV Probit (Marginal effects)	Probit (Marginal effects)	First stage Probit	IV Probit (Marginal effects)
		<u>1</u>	<u>2</u>	<u>3</u>	<u>4</u>	<u>5</u>	<u>6</u>
		<u>Product Inno</u>	<u>Bribes(p)</u>	<u>Product Inno</u>	<u>Process Inno</u>	<u>Bribes(p)</u>	<u>Process Inno</u>
	<u>Bribes(p)</u>	<u>.001**</u> (.001)			<u>.002**</u> (.001)		
	<u>+Avg Bribes(p)</u>		<u>1.06***</u> (.150)	<u>.021</u> (.510)		<u>1.06***</u> (.151)	<u>.006***</u> (.003)
	<u>++Control variables</u>	<u>YES</u>	<u>YES</u>		<u>YES</u>	<u>YES</u>	<u>YES</u>
	<u>Constant</u>		<u>-9.24</u> (11.4)			<u>-6.95</u> (11.3)	
	<u>Summary statistics</u>						
	<u>Log likelihood</u>			<u>4565</u>			<u>4640</u>
	<u>LR Chi2/Wald Chi2</u>			<u>188.0</u>			<u>119</u>
	<u>P value</u>			<u>0.00</u>			<u>0.00</u>
	<u>Sample size</u>			<u>925</u>			<u>924</u>
	<u>Cragg-Donald Wald F statistic</u>		<u>49.43</u>			<u>49.3</u>	
	<u>Stock-Yogo weak ID test</u> <u>critical values: 10% maximal</u> <u>IV size</u>		<u>16.38</u>			<u>16.4</u>	
	<u>Anderson canon. corr. LM</u> <u>statistic</u>		<u>47.71***</u>			<u>47.6***</u>	
	<u>Wald test of exogeneity</u>			<u>9.69***</u>			<u>6.01**</u>
Panel B : Medium firms							
	Variables	Probit (Marginal effects)	First stage Probit	IV Probit (Marginal effects)	Probit (Marginal effects)	First stage Probit	IV Probit (Marginal effects)
		<u>(I)</u>	<u>(II)</u>	<u>(III)</u>	<u>(IV)</u>	<u>(V)</u>	<u>(VI)</u>
		<u>Product Inno</u>	<u>Bribes(p)</u>	<u>Product Inno</u>	<u>Process Inno</u>	<u>Bribes(p)</u>	<u>Process Inno</u>
	<u>Bribes(p)</u>	<u>.001**</u> (.007)			<u>.043</u> (.001)		
	<u>+Avg Bribes(p)</u>	=	<u>1.05***</u> (.157)	<u>-.043***</u> (.118)		<u>1.03***</u> (.158)	<u>.040**</u> (.018)
	<u>++Control variables</u>	<u>YES</u>	<u>YES</u>	<u>YES</u>	<u>YES</u>	<u>YES</u>	<u>YES</u>
	<u>Constant</u>		<u>.078</u> (13.6)			<u>.760</u> (13.7)	<u>.095**</u> (.126)
	<u>Summary statistics</u>						

<u>Log likelihood</u>			<u>-6799.</u>			<u>-6876.9</u>
<u>LR Chi2/Wald Chi2</u>			<u>148.3</u>			<u>99.89</u>
<u>P value</u>			<u>0.00</u>			<u>0.000</u>
<u>Sample size</u>	<u>1389</u>		<u>1361</u>			
<u>Cragg-Donald Wald F statistic</u>		<u>45.02</u>			<u>42.53</u>	
<u>Stock-Yogo weak ID test</u> <u>critical values: 10% maximal</u> <u>IV size</u>		<u>16.38</u>			<u>16.38</u>	
<u>Anderson canon. corr. LM</u> <u>statistic</u>		<u>44.1***</u>			<u>41.74***</u>	
<u>Wald test of exogeneity</u>			<u>4.52***</u>			<u>5.71**</u>

Panel C : Small firms

	<u>Variables</u>	<u>Probit</u> <u>(Marginal</u> <u>effects)</u>	<u>First stage</u> <u>Probit</u>	<u>IV Probit</u> <u>(Marginal</u> <u>effects)</u>	<u>Probit</u> <u>(Marginal</u> <u>effects)</u>	<u>First stage</u> <u>Probit</u>	<u>IV Probit</u> <u>(Marginal</u> <u>effects)</u>
		<u>(I)</u>	<u>(II)</u>	<u>(III)</u>	<u>(IV)</u>	<u>(V)</u>	<u>(VI)</u>
		<u>Product</u> <u>Inno</u>	<u>Bribes(p)</u>	<u>Product</u> <u>Inno</u>	<u>Process</u> <u>Inno</u>	<u>Bribes(p)</u>	<u>Process</u> <u>Inno</u>
<u>Bribes(p)</u>		<u>.013**</u> <u>(.001)</u>			<u>.004</u> <u>(.001)</u>		
<u>+Avg Bribes(p)</u>			<u>1.10***</u> <u>(.135)</u>	<u>-.048***</u> <u>(.001)</u>		<u>1.09***</u> <u>(.135)</u>	<u>-.018*</u> <u>(.068)</u>
<u>++Control variables</u>		<u>YES</u>	<u>YES</u>	<u>YES</u>	<u>YES</u>	<u>YES</u>	<u>YES</u>
<u>Constant</u>			<u>-.838</u> <u>(7.13)</u>			<u>-.847</u> <u>(7.13)</u>	<u>-1.43</u> <u>(1.64)</u>
<u>Summary statistics</u>							
<u>Log likelihood</u>				<u>8291.6</u>			<u>-8308.2</u>
<u>LR Chi2/Wald Chi2</u>				<u>972.16</u>			<u>64.93</u>
<u>P value</u>				<u>0.0000</u>			<u>0.000</u>
<u>Sample size</u>	<u>1671</u>	<u>1671</u>	<u>1671</u>	<u>1671</u>	<u>1671</u>	<u>1671</u>	<u>1671</u>
<u>Cragg-Donald Wald F statistic</u>			<u>64.44</u>			<u>65.55</u>	
<u>Stock-Yogo weak ID test</u> <u>critical values: 10% maximal</u> <u>IV size</u>			<u>16.38</u>			<u>16.38</u>	
<u>Anderson canon. corr. LM</u> <u>statistic</u>			<u>64.44</u>			<u>63.63</u>	
<u>Wald test of exogeneity</u>				<u>3.37**</u>			<u>2.11***</u>

Panel D : Micro firms

	<u>Variables</u>	<u>Probit</u> <u>(Marginal</u> <u>effects)</u>	<u>First stage</u> <u>Probit</u>	<u>IV Probit</u> <u>(Marginal</u> <u>effects)</u>	<u>Probit</u> <u>(Marginal</u> <u>effects)</u>	<u>First stage</u> <u>Probit</u>	<u>IV Probit</u> <u>(Marginal</u> <u>effects)</u>
		<u>(I)</u>	<u>(II)</u>	<u>(III)</u>	<u>(IV)</u>	<u>(V)</u>	<u>(VI)</u>
		<u>Product</u> <u>Inno</u>	<u>Bribes(p)</u>	<u>Product</u> <u>Inno</u>	<u>Process</u> <u>Inno</u>	<u>Bribes(p)</u>	<u>Process</u> <u>Inno</u>
<u>Bribes(p)</u>		<u>-.005</u> <u>(.005)</u>			<u>-.016**</u> <u>(.008)</u>		
<u>+Avg Bribes(p)</u>			<u>-.947*</u>	<u>-.058***</u>		<u>-.947**</u>	<u>-.013**</u>

		(.816)	(.006)		(.816)	(.057)
<u>++Control variables</u>	<u>YES</u>	<u>YES</u>	<u>YES</u>	<u>YES</u>	<u>YES</u>	<u>YES</u>
<u>Constant</u>			<u>1.01</u> <u>(.681)</u>		<u>-9.50</u> <u>(2.80)</u>	<u>-.689**</u> <u>(.7851)</u>
<u>Summary statistics</u>						
<u>Log likelihood</u>			<u>-252.11</u>			<u>-237.07</u>
<u>LR Chi2/Wald Chi2</u>			<u>95.12</u>			<u>31.36</u>
<u>P value</u>			<u>0.00</u>			<u>0.00</u>
<u>Sample size</u>	<u>55</u>	<u>55</u>	<u>55</u>		<u>56</u>	<u>56</u>
<u>Cragg-Donald Wald F statistic</u>		<u>18.96</u>			<u>18.73</u>	
<u>Stock-Yogo weak ID test</u> <u>critical values: 10% maximal</u> <u>IV size</u>		<u>16.38</u>			<u>16.38</u>	
<u>Anderson canon. corr. LM</u> <u>statistic</u>		<u>17.39</u>			<u>19.66</u>	
<u>Wald test of exogeneity</u>			<u>2.67**</u>			<u>.364***</u>

Table 6: Innovation and Institutional environment

	<u>Product Innovation</u>		<u>Process Innovation</u>	
<u>Variables</u>	<u>Probit</u>	<u>IV Probit</u> <u>Interaction</u>	<u>Probit</u>	<u>IV Probit</u> <u>Interaction</u>
	<u>(I)</u>	<u>(III)</u>	<u>(V)</u>	<u>(VIII)</u>
<u>Bribes(p)</u>	<u>.005*</u> <u>(.003)</u>		<u>.005*</u> <u>(.003)</u>	=
<u>+Avg Bribes(p)</u>	=	<u>-.009***</u> <u>(.003)</u>		<u>-.009***</u> <u>(.004)</u>
<u>Formal Fund</u>	<u>.001</u> <u>(.001)</u>	<u>.001</u> <u>(.001)</u>	<u>.001</u> <u>(.001)</u>	<u>.001</u> <u>(.001)</u>
<u>Infomal Fund</u>	<u>-.002*</u> <u>(.001)</u>	<u>-.001</u> <u>(.001)</u>	<u>-.002**</u> <u>(.001)</u>	<u>-.001</u> <u>(.001)</u>
<u>Own Fund</u>			=	
<u>Sole Prop</u>	<u>.231</u> <u>(.339)</u>	<u>.055</u> <u>(.320)</u>	<u>.111</u> <u>(.336)</u>	<u>.006</u> <u>(.301)</u>
<u>Partnership</u>	<u>.177</u> <u>(.335)</u>	<u>-.047</u> <u>(.317)</u>	<u>.041</u> <u>(.332)</u>	<u>-.102</u> <u>(.297)</u>
<u>Company</u>	=	=		=
<u>Domestic Own</u>	<u>.003</u> <u>(.002)</u>	<u>.003</u> <u>(.002)</u>	<u>.004**</u> <u>(.002)</u>	<u>.005***</u> <u>(.002)</u>
<u>Foreign Own</u>	<u>.006***</u> <u>(.002)</u>	<u>.006**</u> <u>(.003)</u>	<u>.006**</u> <u>(.002)</u>	<u>.006***</u> <u>(.002)</u>
<u>Government Own</u>	<u>-.005</u> <u>(.008)</u>	<u>.004</u> <u>(.009)</u>	<u>.011</u> <u>(.009)</u>	<u>.012</u> <u>(.008)</u>
<u>Other Own</u>	=	=	=	=
<u>Female Own</u>	<u>.059</u> <u>(.044)</u>	<u>.027</u> <u>(.049)</u>	<u>.008</u> <u>(.043)</u>	<u>.003</u> <u>(.038)</u>
<u>Firm Age</u>	<u>-.115</u> <u>(.005)</u>	<u>.004</u> <u>(.126)</u>	<u>.194</u> <u>(.321)</u>	<u>.045</u> <u>(.244)</u>
<u>Email</u>	<u>.301***</u>	<u>.026</u>	<u>.544***</u>	<u>.195</u>

	<u>(.104)</u>	<u>(.224)</u>	<u>(.112)</u>	<u>(.336)</u>
<u>Website</u>	<u>.412***</u> <u>(.049)</u>	<u>.244</u> <u>(.209)</u>	<u>.181***</u> <u>(.049)</u>	<u>.134</u> <u>(.088)</u>
<u>Weak Intuitional Quality</u>	<u>-.007</u> <u>(.007)</u>	<u>.160**</u> <u>(.064)</u>	<u>.005</u> <u>(.006)</u>	<u>.158**</u> <u>(.004)</u>
<u>Manufacturing</u>	<u>.141***</u> <u>(.050)</u>	<u>.079</u> <u>(.082)</u>	<u>.180***</u> <u>(.046)</u>	<u>.120**</u> <u>(.092)</u>
<u>Retail</u>	<u>.148**</u> <u>(.063)</u>	<u>.008</u> <u>(.082)</u>	<u>-.008</u> <u>(.062)</u>	<u>.039</u> <u>(.058)</u>
<u>Other Industry</u>	<u>=</u>	<u>=</u>		
<u>Large</u>	<u>127**</u> <u>(.061)</u>	<u>.080</u> <u>(.079)</u>	<u>.210***</u> <u>(.058)</u>	<u>.139</u> <u>(.109)</u>
<u>Medium</u>	<u>.099**</u> <u>(.050)</u>	<u>.066</u> <u>(.062)</u>	<u>.063</u> <u>(.049)</u>	<u>.048</u> <u>(.050)</u>
<u>Small& Micro</u>	<u>=</u>	<u>=</u>	<u>=</u>	<u>=</u>
<u>Bribes(p)*Weak Intuitional Quality</u>	<u>-.001</u> <u>(.001)</u>	<u>-.119***</u> <u>(.040)</u>	<u>-.001</u> <u>(.001)</u>	<u>-.113***</u> <u>(.046)</u>
<u>Constant</u>	<u>-.5631</u> <u>(.405)</u>	<u>-2.50***</u> <u>(.620)</u>	<u>-1.35***</u> <u>(.408)</u>	<u>-2.88***</u> <u>(.428)</u>
<u>Summary statistics</u>				
<u>Log likelihood</u>	<u>-2458.8</u>	<u>15798</u>	<u>-2626.8</u>	<u>-15983</u>
<u>LR Chi2 /Wald Chi2</u>	<u>210.01</u>	<u>153.9</u>	<u>150.91</u>	<u>591.9</u>
<u>P value</u>	<u>0.00</u>	<u>0.00</u>	<u>0.00</u>	<u>0.00</u>
<u>Sample size</u>	<u>6908</u>	<u>6908</u>	<u>6909</u>	<u>6909</u>
<u>Wald test of exogeneity</u>		<u>2.89**</u>		<u>2.77**</u>

Table 7 : Summary of empirical results

	<u>Results</u> $Pr(Inno_{i,j} = 1)$ $= \phi(\alpha + \beta Bribes\%_{i,j} + Control Variables + \varepsilon_{i,j})$	<u>Hypothesis</u>	<u>Hypothesis Accepted or Rejected</u>	<u>Table Reference</u>
<u>Full Sample</u>				
<u>Product_Inno</u>	<u>(-) Significant</u> <u>($\beta = -.036$)</u>	<u>H1: Corruption has a negative impact on firm-level innovation</u>	<u>Accepted</u>	<u>Table 4</u>
<u>Process_Inno</u>	<u>(-) Significant</u> <u>($\beta = -.039$)</u>	<u>H1: Corruption has a negative impact on firm-level innovation</u>	<u>Accepted</u>	<u>Table 4</u>
<u>Large firms only</u>				
<u>Product_Inno</u>	<u>Net effect is inexistence</u>	<u>H2: Increase firm size positively moderates the relationship between paying bribes and</u>		<u>Table 4</u>

		<u>firm-level innovation.</u>		
<u>Process_Inno</u>	<u>(+) Significant</u> <u>($\beta = .001$)</u>	<u>H2: Increase firm size positively moderates the relationship between paying bribes and firm-level innovation</u>	<u>Accepted</u>	<u>Table 4</u>
<u>Medium firms only</u> <u>Product_Inno</u>	<u>Net effect is inexistence</u>	<u>H2: Increase firm size positively moderates the relationship between paying bribes and firm-level innovation.</u>	<u>=</u>	<u>Table 4</u>
<u>Process_Inno</u>	<u>Net effect is inexistence</u>	<u>H2: Increase firm size positively moderates the relationship between paying bribes and firm-level innovation.</u>	<u>=</u>	<u>Table 4</u>
<u>Large firms only</u> <u>Product_Inno</u>	<u>(+) insignificant</u> <u>($\beta = -.021$)</u>	<u>H2: Increase firm size positively moderates the relationship between paying bribes and firm-level innovation.</u>	<u>=</u>	<u>Table 5</u>
<u>Process_Inno</u>	<u>(+) Significant</u> <u>($\beta = .006$)</u>	<u>H2: Increase firm size positively moderates the relationship between paying bribes and firm-level innovation.</u>	<u>Accepted</u>	<u>Table 5</u>
<u>Medium firms only</u> <u>Product_Inno</u>	<u>(-) Significant</u> <u>($\beta = -.043$)</u>	<u>H2: Increase firm size positively moderates the relationship between paying bribes and firm-level innovation.</u>	<u>Rejected</u>	<u>Table 5</u>
<u>Process_Inno</u>	<u>(+) Significant</u> <u>($\beta = .040$)</u>	<u>H2: Increase firm size positively moderates the relationship between paying bribes and</u>	<u>Accepted</u>	<u>Table 5</u>

		<i>firm-level innovation.</i>		
<u>Small firms only</u> <u>Product Inno</u>	(-) Significant ($\beta = -.048$)	<i>H2: Increase firm size positively moderates the relationship between paying bribes and firm-level innovation.</i>	<u>Accepted</u>	<u>Table 5</u>
<u>Process Inno</u>	(-) Significant ($\beta = .018$)	<i>H2: Increase firm size positively moderates the relationship between paying bribes and firm-level innovation.</i>	<u>Accepted</u>	<u>Table 5</u>
<u>Micro firms only</u> <u>Product Inno</u>	(-) Significant ($\beta = -.947$)	<i>H2: Increase firm size positively moderates the relationship between paying bribes and firm-level innovation.</i>	<u>Accepted</u>	<u>Table 5</u>
<u>Process Inno</u>	(-) Significant ($\beta = .013$)	<i>H2: Increase firm size positively moderates the relationship between paying bribes and firm-level innovation.</i>	<u>Accepted</u>	<u>Table 5</u>
<u>Full Sample</u> <u>Product Inno</u>	(-) Significant ($\beta = -.028$)	<i>H3: The better developed institutional environment weakens the relationship between paying bribes and firm-level innovation</i>	<u>Accepted</u>	<u>Table 6</u>

Table 8: Probit and IV probit estimation results of innovation and alternative proxy of bribes (Obs corruption) relationship

<u>Variables</u>	<u>Probit</u>	<u>First stage Probit</u>	<u>IV Probit</u>	<u>Probit</u>	<u>First stage Probit</u>	<u>IV Probit</u>
------------------	---------------	---------------------------	------------------	---------------	---------------------------	------------------

	(I)	(II)	(III)	(IV)	(V)	(VI)
	<u>Product Inno</u>	<u>Obs Corruption</u>	<u>Product Inno</u>	<u>Process Inno</u>	<u>Obs Corruption</u>	<u>Process Inno</u>
<u>Obs Corruption</u>	<u>-.004</u> <u>(.081)</u>			<u>-.042**</u> <u>(.017)</u>		
<u>+Avg Mgr time</u>		<u>631**</u> <u>(.312)</u>	<u>-.030*</u> <u>(.285)</u>		<u>.278</u> <u>(.286)</u>	<u>-.041**</u> <u>(.143)</u>
<u>++Control variables</u>	<u>YES</u>	<u>YES</u>		<u>YES</u>	<u>YES</u>	<u>YES</u>
<u>Constant</u>		<u>.379***</u> <u>(.136)</u>			<u>.381**</u> <u>(.135)</u>	
<u>Summary statistics</u>						
<u>Log likelihood</u>			<u>-5228</u>			<u>-5393</u>
<u>LR Chi2/Wald Chi2</u>			<u>3186</u>			<u>221.3</u>
<u>P value</u>			<u>0.00</u>			<u>0.00</u>
<u>Sample size</u>			<u>4141</u>			<u>4141</u>
<u>Cragg-Donald Wald F statistic</u>		<u>17.02</u>			<u>17.34</u>	
<u>Stock-Yogo weak ID test</u> <u>critical values: 10% maximal</u> <u>IV size</u>		<u>16.38</u>			<u>16.38</u>	
<u>Anderson canon. corr. LM</u> <u>statistic</u>		<u>15.03***</u>			<u>15.35***</u>	
<u>Wald test of exogeneity</u>			<u>8.12***</u>			<u>7.89**</u>

+ Avg Mgr time is an instrument variable for Obs Corruption . ++ (*) dy/dx is for discrete change of dummy variable from 0 to 1. The main dependent variable is a dummy variable taking one if the firm is innovative and zero otherwise. Panel A reports product innovation results and Panel B reports process innovation results. In each panel, column I report the marginal effect of probit model with exogenous regressors regression results; column II reports first stage probit results, column III reports the marginal effect of IV probit regression results. Obs Corruption is a main explanatory variable and IV probit use, locality sector average percentage of total senior management's time spent on dealing with requirements imposed by government regulations in a typical week over the last year as an instrument. These models provide standard errors, which are in parentheses. The Wald test of exogeneity is reported in the last row as a chi-squared statistic with 1 degree of freedom. * Significant at 10% level, **Significant at 5% level, ***Significant at 1% level.

Table 9: The impact on bribes on innovation: propensity score matching (PSM)

	<u>No. treated</u>	<u>No. contr.</u>	<u>ATT^a</u>	<u>Std.Err</u>	<u>t</u>
<u>Product innovation</u>					
<u>Nearest neighbour matching</u>	<u>264</u>	<u>247</u>	<u>-0.115</u>	<u>0.051</u>	<u>2.272</u>
<u>Kernel matching</u>	<u>223</u>	<u>2582</u>	<u>-0.111</u>	<u>0.028</u>	<u>3.933</u>
<u>Radius</u> <u>matching*(radius=0.01)</u>	<u>61</u>	<u>2575</u>	<u>-0.105</u>	<u>0.052</u>	<u>1.992</u>
<u>Stratification matching</u>	<u>222</u>	<u>2943</u>	<u>-0.110</u>	<u>0.030</u>	<u>3.666</u>
	<u>No. treated</u>	<u>No. contr.</u>	<u>ATT^a</u>	<u>Std.Err</u>	<u>t</u>
<u>Process innovation</u>					
<u>Nearest neighbour matching</u>	<u>223</u>	<u>224</u>	<u>-0.055</u>	<u>0.045</u>	<u>1.211</u>
<u>Kernel matching</u>	<u>223</u>	<u>2582</u>	<u>-0.096</u>	<u>0.035</u>	<u>2.738</u>

<u>Radius matching</u> (radius=0.01)	<u>224</u>	<u>3368</u>	<u>-0.095</u>	<u>0.029</u>	<u>3.283</u>
<u>Stratification matching</u>	<u>222</u>	<u>2943</u>	<u>-0.095</u>	<u>0.028</u>	<u>3.340</u>

Note: an ATT means the average treatment effect on the treated. Firm level covariates are included in all models

The standard error used to compute the t statistics is the standard deviation of the ATT after 100 bootstrap replications.

**In this matching, each treated unit is matched only with the control units whose propensity score falls into a predefined neighborhood of the propensity score (radius 0.01 in here) of the treated unit. Therefore, it is possible that some treated units are not matched because the neighborhood does not contain control units" (See <https://www.stata-journal.com/sjpdf.html?articlenum=st0026>)*

Table 10: Instrumental regression results

<u>Variables</u>	<u>OLS</u>	<u>2SLS</u>
<u>(I)</u>	<u>(II)</u>	<u>(III)</u>
<u>Bribe(p)</u>	<u>-.001</u> <u>(.001)</u>	
<u>⁺Avg_Bribe(p)</u>		<u>-.007</u> <u>(.013)</u>
<u>⁺⁺Control variables</u>	<u>YES</u>	<u>YES</u>
<u>Constant</u>	<u>10.01***</u> <u>(.074)</u>	<u>18.93***</u> <u>(.418)</u>
<u>Summary statistics</u>		
<u>P value</u>	<u>0.000</u>	<u>0.000</u>
<u>Sample size</u>	<u>930</u>	<u>930</u>
<u>Adjusted Rsq</u>	<u>0.998</u>	<u>0.998</u>