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Performance of Newly Listed Firms: Evidence from Japanese firm and venture capital data

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Abstract

This paper investigates the effect of Initial Public Offerings (IPOs) on firm performance. To single out the economic effects on firm performance brought about by issuing IPOs, we employ propensity-score matching difference-in-differences estimation. Using a unique firm-level panel dataset that allows us to identify newly listed firms and those keeping unlisted status, we find that the former showed better performance than their never-listed counterparts prior to their IPO while the difference in performance partly diminished after the IPO. This implies that firms' distorted behavior originating from, for example, empire building motives prevents newly listed firms from performing. This result is mainly driven by the sample firms going public during a "hot market". Using the information on venture capital (VC) investment, we also find that the participation of VC in investments exacerbates such negative impacts of IPOs on firm performance. The adverse impact on firm performance is more sizable among IPO firms which are invested in by VC syndicates consisting of a smaller number of less heterogeneous VC, or not including foreign VC. These results suggest that the timing of going public and the composition of VC syndicates are related to the post-IPO performance of newly listed firms.

Keywords: IPO; Firm performance; Venture capital; Distorted firm behavior

JEL classification: G24, G32, D24, L25

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1. Introduction

Firms go public with various prospective motivations. To illustrate, expanding financial channels to stock market through Initial Public Offering (IPO), firms alleviate financial friction. Being recognized by broader business entities and consumers, newly listed firms also enjoy benefit in the context of marketing and public relations. Listed firms, however, could also suffer from obstacles, which distort their optimal behaviors. Such distortion includes short-termism originating from the pressures in stock market (Stein 1989), manager's empire building motives (Stulz 1990), or inactive behavior of firm managers (Bertrand and Mullainathan 2003). In response to such theoretical discussion, a few extant studies have tried to identify the impact of being listed on firms' behavior and performance (Asker et al. 2012). More precisely, it has been an important empirical issue whether the economic causal impact running from being listed is positive or negative.

Although the horserace between the two competing empirical implications obtained from theoretical discussions is conceptually straightforward, researchers find it difficult to examine the relationship due to identification problems: while going public generates various impacts on firms' performance, the performance of firms also affects the likelihood of IPO. Such a concern about the simultaneous determination of firm performance and the decision of going public makes it difficult to establish causality running from being listed to firms' performance. This paper tackles this problem by taking advantage of the rich panel dataset consisting of newly listed firms and a number of unlisted firms which have not yet established IPO. This dataset allows us to match a newly listed firm (i.e., *treatment*) with a firm which (i) has similar characteristics to the treated firm in terms of the likelihood of IPO but (ii) not yet established IPO (i.e., *control*).

As another key theme of this paper, we also study the role of Venture Capital (VC) in the context of the economic impact of being listed. VC is a class of financial intermediaries that finances venture firms mainly through equity investment (Gompers and Lerner 2001) prior to IPO. It provides

funds, screens investment targets, and gives various advices aiming at adding value to the firms. The object of VC is successful exits from investments with higher return through, for example, IPO. Although VCs are supposed to employ their strategic, management, marketing, and administrative expertise to achieve the successful exits (Cumming et al. 2005), it is still not self-evident whether their participation in investments and the provision of their expertise lead to long-term improvement in firm performance or not. Specifically, during the hot market right after the introduction of Japanese emerging stock markets (e.g., TSE-MOTHERS), it had been said some of VCs induced unpromising venture firms to go public for making profits from selling their shares at the timing of IPO. If this sort of VCs' motivation is dominating, VC participation might not lead to better performance of client firms. Thus, our interest is in how and to what extent VC contributes to firms' post-IPO performance through their presumed roles.

This paper focuses on the IPO of Japanese firms after 2001. We examine whether IPOs of firms with or without being invested by VCs had a positive or negative impact on the performance of newly listed firms in the comparison with unlisted firms which show similar characteristics to listed firms but not establish IPO. To do so, we construct and use a unique firm-level dataset compiled from various sources. The dataset includes the information that allows us to identify firms' IPO timing, and the information of VC investment to each sample firm. Thus, our sample consists of four groups of firms: IPO firms not invested by VCs, IPO firms invested by VCs, non-IPO firms not invested by VCs, and non-IPO firms invested by VCs.¹ By comparing the change in the performance of IPO firms with/without being invested by VCs and non-IPO firms without invested by VCs, we can single out the effect of going public on the performance of firms.

Our main findings can be summarized as follows. First, newly listed firms showed better performance than never-listed control firms prior to IPO while the difference in the performance

¹ Precisely speaking, we will exclude the last group from our analysis, and focus on two separate comparisons between A) IPO firms not invested by VCs and non-IPO firms not invested by VCs, and B) IPO firms invested by VCs and non-IPO firms not invested by VCs.

partly diminished after establishing IPO. This implies that firms' distorted behavior originating from being listed possibly prevents newly listed firms from performing. Namely, in the case of IPO firms without being invested by VCs prior to IPO, the return on asset of these treated firms is 5-6% higher than control firms (i.e., counterpart non-IPO firms without being invested by VCs) before going to public. Such a margin of performance, however, partly diminishes after IPO. Since we chose the control firms so that the likelihood of IPO for not IPO firm in a counter-factual environment is same as that for the treated firm, such a change in the performance difference between treated and control is not a consequence of selection bias. Note that the result does not either reflect unobservable time-invariant firm-specific effect since we focus on the change in performance. We confirm, thus, the result implies that the change of firms' listed status had a significant causal adverse effect on firm performance. Second, we find that this result is mainly driven by the firms going public during the hot market from 2001 to 2005. On the other hand, there is no adverse impact found in the case of firms going public after 2005. Third, using the information on Venture Capitals (VCs) investment, we also find that the participation of VCs in investments exacerbates such negative impact of IPO on firms' performance. More precisely, in the case of IPO firms being invested by VCs prior to IPO, the return on asset (ROA) of these treated firms is 3-4% higher than control firms (i.e., counterpart non-IPO firms without being invested by VCs) before going to public. Surprisingly, the adverse effect (around -4 to -5%) on firm performance originating from going public not only offsets the performance advantage prior to IPO but also dominates it. Given another finding that the asset size of IPO firms is confirmed to become larger than their counterpart non-IPO firms, we conjecture that the overinvestment of VC-backed newly listed firms after going public is the main source of their deteriorated post-IPO performance. Lastly, we find that this adverse impact originating from IPO with VCs are more apparent in the case of VC syndicates with a small number of less heterogeneous VCs. It is also found that VC syndicates including foreign VCs do not exhibit this mechanism.

The contributions of this paper to the literature are at least twofold. First, by using a rich panel dataset containing not only newly listed firms but a number of unlisted firms, we can implement propensity-score matching difference-in-differences estimation, which allows us to circumvent the identification problem faced by many existing studies. More specifically, the present paper establishes a causal mechanism that going public prevents firms performing. Asker et al. (2012) examines the capital investment behavior of listed and unlisted firms using a firm level dataset including both the listed and unlisted U.S. enterprises, but our uniqueness rests on the fact that we investigate the negative impact of going public to firms not only on firms' behavior but also on their long-run performance as well. Second, we are able to provide evidence for the long-run impact of the participation of venture capitals in a clearer manner than many of the previous studies did in the past. Namely, many of the studies that investigate the effects of VC investments use stock price or short-run performance change after IPO and thus are unable to clarify the long-run effects of IPO on individual firms.

The rest of the paper is structured as follows. Section 2 reviews the related literature and our contribution in greater detail. Section 3 describes our data. Sections 4 and 5 report our methodology and results for the impact of going public on firm performance, respectively. Section 6 extends the analysis to several sub-sample analyses. Section 7 summarizes the results and concludes.

2. Literature Review

2.1 Economic Implication of Going Public

A vast literature has been documenting the empirical facts about the impact of going public. For example, reflecting the conjecture that going public alleviates financial constraints faced by firms, a number of studies confirm that private firms suffer from higher external finance costs (e.g.,

Brav 2009; Saunders and Steffen 2011). One important research strand in this context concentrates on friction faced by listed firms. The idea is that firms going public encounter managerial myopia discussed in Stein (1989), which is a pressure for company executives to show higher performance measure by, for example, the level of profit. Such distorted motivation leads to insufficient investment, which could result in poor long-run performance. Empire-building motives originating from manager's entrenchment behavior (Stulz 1990) and inactive attitudes aiming at "quiet-life" (Bertrand & Mullainathan 2003) would also generate suboptimal choice from the standpoint of firms' value-maximization.

Due to the increasing availability of unlisted firm data, for example, by Sagemworks, a growing number of studies have been examining the implication of firms' listed status in the context of firm behavior.² Asker et al. (2013) examine capital investment of listed and unlisted firms and find that the former invests much less than the latter. This result is consistent with the empirical finding already established by using smaller size of datasets, for example, in Sheen (2009).

The paper is most closely related to a group of literature that examines the impact of listed status onto firm performance. In this context (e.g., accounting profit), a number of extant studies have pointed out that managers of listed firms tend to avoid negative earnings surprise (Baber et al. 1991; Skinner and Sloan 2002; Bhojraj et al. 2009). Compared to these studies, our uniqueness rests on identifying a clear causal relation running from IPO to firm performance by using a rich panel dataset containing a number of listed and unlisted firms.

2.2 Contribution of Venture Capitals

From border perspective apart from direct impact of going public, there is a vast literature that examines empirically the roles of VC investment in the context of firm performance. Most of the extant studies in this line focus on the stock prices before and after IPO (e.g., Megginson and

² <https://www.sagemworksinc.com/default.aspx>

Weiss 1991; Jain and Kini 1995; Brav and Gompers 1997; Kutsuna et al. 2002; Da Silva Rosa et al. 2003; Wang et al. 2003; Tykvová and Walz 2003; Tykvová 2004; Lee and Wahal 2004; Florin 2005; Arthurs and Busenitz 2006).

One important theme in this field is the implication of the characteristics of each VC and syndication of VCs. Many papers find that the experience and reputation of VCs affect the performance of their investment (e.g., Megginson and Weiss 1991; Lerner 1994; Jain and Kini 1995; Gompers 1996; Brander et al. 2002; Wang et al. 2003; Rindermann 2003; Chang 2004; Lee and Wahal 2004; Arthurs and Busenitz 2005; Giot and Schwienbacher 2006; Hochberg et al. 2007; Miyakawa and Takizawa 2012). As another strand of research, there is growing group of studies examining the relationship between the performance of investment and the types of VCs. For example, Tykvová and Walz (2007) find that the involvement of independent or foreign-owned VCs contributes to better performance of investments. Hamao et al. (2000) and Tykvová (2004) also emphasize the role of including independent VC in syndication.

Among limited examples studying Japanese firms, Kutsuna et al. (2000) find a positive correlation between the share of VC investments and the performance of firms after IPO. Kutsuna et al. (2002) also compares the change in difference in firm performance before and after IPO but the result is not supporting significant effect originating from VC participation. Okamuro and Hisa (2005) implement the similar analysis and confirm positive contribution of VC syndicate.

3. Data

3.1 Data sources

We rely primarily on two groups of firm-level data sources. First, information on unlisted and listed firms' financial characteristics is obtained from the *Basic Survey of Business Structure and*

Activities (BSBSA; *Kigyō Katsudō Kihon Chōsa* in Japanese) compiled by the Ministry of Economy, Trade and Industry, and *Development Bank of Japan (DBJ) Financial Databank System*. The main purpose of the former survey is to gauge quantitatively the activities of Japanese enterprises, including capital investment, exports, foreign direct investment, and investment in research and development. To this end, the survey covers the universe of enterprises in Japan with more than 50 employees and with paid-up capital of over 30 million yen. From this data source, we obtain the major financial characteristics of unlisted firms. The latter dataset contains all the information about listed firms.

Second, we use the firm and investment round-level dataset provided by Japan Venture Research Co., LTD (JVR), a growing business data bureau focusing on start-up firms in Japan. The data covers all the IPOs dated from 2001 to 2011. The data consist of, for example, firm identification, IPO date, and the market where the firms are initially listed. An important feature of this data is that it stores the list of all VCs investing to each firm and the investment amount from each VC to the firm in each investment round. The data also store a part of the characteristics of each VC and entrepreneurial firms such as industry classification and location.

These datasets allows us to construct a unique firm-level dataset, which consists of four groups of firms: A) IPO firms invested by VCs, B) IPO firms not invested by VCs, C) non-IPO firms not invested by VCs, and D) non-IPO firms invested by VCs. The first group is identified by the list of firms in JVR data. Since DBJ data contains the list of all the newly listed firms, we can also identify the second group. A separate dataset provided by JVR contains the list of unlisted firms which have been invested by VCs, which leads to the last group. The residual firms are treated as the third group. Figure 1 illustrates the sample sizes of each group.

Using the financial characteristics stored in the datasets, we can match a newly listed firm (i.e., *treatment*) with a firm which has similar characteristics to the treated firm in terms of the

likelihood of IPO but not establishing IPO (i.e., *control*). As we detail later, by using the propensity of IPO conditional on firm characteristics, we match IPO firms with/without being invested by VCs with non-IPO firms not invested by VCs. In this sense, we ensure the exogeneity of IPO without being invested by VCs as well as the exogeneity of IPO with being invested by VCs.³ By comparing the change in the performance of IPO firms with/without being invested by VCs and non-IPO firms without invested by VCs, we attempt to single out the effect of going public with/without VCs on the performance of firms.

4. Empirical Strategy

4.1 Two Competing Stories on the Impact of Going Public

A vast literature has been discussing the existence of financial friction faced by firms. For example, Brav (2009) finds that compared to listed firms, unlisted firms face more expensive equity cost due to, for example, severer information asymmetry. In this sense, it is conjecture that firms alleviate financial friction by being listed. Listed firms, however, could also suffer from different type of obstacles, which distort their optimal behaviors. As discussed and confirmed in Asker et al. (2012) in the context of capital investment, short-termism originating from the pressure of stock market (Stein 1989), manager's empire building motives (Stulz 1990), or inactive behavior of firm managers (Bertrand and Mullainathan 2003) could a source of such distortion. This motivates to test whether newly listed firms show better or worse performance compared to their counterpart firms.

4.2 Propensity-Score Matching Difference-in-Differences Estimation (PSM-DID)

In order to run the horserace of the above-mentioned two stories, first, we compute the

³ If we are mainly interested in the role of VCs without considering IPO, we should match non-IPO firms invested by VCs with non-IPO firms not invested by VCs, and implement the difference-in-differences estimation, which is out of scope of the presenting paper.

propensity score defined in Rosenbaum and Rubin (1983), which is the conditional probability of assignment to a particular treatment (i.e., IPO without VCs and IPO with VCs) given the pre-treatment characteristics:

$$(1) P(x) \equiv \Pr\{z = 1|x\} = E\{z|x\}$$

In this formulation, $z = \{0,1\}$ is the indicator of receiving the treatment and x is a vector of observed pretreatment characteristics. Rosenbaum and Rubin (1983) show that if the recipient of the treatment is randomly chosen within cells defined by x , it is also random within cells defined by the values of the single-index variable $P(x)$. Therefore, for each treatment case i , if the propensity score $P(x_i)$ is known, the Average effect of Treatment on the Treated (ATT) can be estimated as follows:

$$\begin{aligned} (2) \hat{\alpha}_{ATT} &= E\{y_{1i} - y_{0i}|z_i = 1\} \\ &= E\{E\{y_{1i} - y_{0i}|z_i = 1, p(x_i)\}\} \\ &= E\{E\{y_{1i}|z_i = 1, p(x_i)\} - E\{y_{0i}|z_i = 0, p(x_i)\}|z_i = 1\} \end{aligned}$$

In this formulation, y_1 and y_0 denote the potential outcomes in the two counterfactual situations of treatment and no treatment, respectively. Therefore, according to the last line of equation (2), the ATT can be estimated as the average difference between the outcome of recipients and non-recipients of the treatment whose propensity scores $P(x_i)$ are identical.

In the case of the presenting study, we consider two types of treatment: IPO without being invested by VCs and IPO with being invested by VCs. Therefore, we focus on the difference in *ex post* performance between firms going public without being invested by VCs and firms that remain

private (non-IPO firms) as well as between firms going public with being invested by VCs and private firms. x is a vector of various characteristics of a firm such as firm size, liquidity, leverage, *ex ante* performance, etc.

By separately estimating two logit models (i.e., IPO without VCs and IPO with VCs) at the first stage, we investigate important determinants of going public and compute the propensity score (i.e., the probabilities of a firm going public) for each firm. Making use of this result, we conduct propensity score matching and compare the change in the performance of firms within the pairs of observations matched on the propensity score. In our matching process, firms are matched separately for each year using one-to-one nearest neighbor matching.⁴

In the second stage, we estimate a difference-in-differences (DID) estimator to evaluate the causal effect of IPO on a set of performance variables of interest. Once we match treated and control firms, the only difference between public and private firms is their listed status. Therefore, we focus on the Average effect of Treatment on the Treated (ATT). The ATT can be estimated as equation (2) above, which, in the case of this study, is equivalent to the following equation:

$$(3) \quad \hat{\alpha}_{ATT} = \frac{1}{n} \sum_1^n (y_{IPO\ year+s}^{treated} - y_{IPO\ year+s}^{control}) - \frac{1}{n} \sum_1^n (y_{Pre\ IPO\ year}^{treated} - y_{Pre\ IPO\ year}^{control}) \quad s = \{1,2,3\}$$

In this formulation, n denotes the number of observations and y denotes outcome variables

4.3 Performance Measure

The first performance variable we employ is firms' Total Factor Productivity (TFP), which is calculated using the multilateral TFP index method developed by Good et al. (1997). Details on the TFP measure are provided in the Appendix. The second performance measure used in this presenting

⁴ Our matching procedure is implemented in Stata 11 using a modified version of the procedure provided by Leuven and Sianesi (2001). As we match firms separately for each year and industry (52 manufacturing industries and 56 non-manufacturing industries), we had to modify the program.

paper is return on asset, defined as a ratio of firms' current profit to total asset. The last variable we use is the level of sale per employees. We also interested in the size of firms' assets since it reflects the difference in investment behaviors of newly-listed and never-listed firms, which we could use to discuss the source of performance differences between such two groups.

4.4 Explanatory variables for Propensity Score

Let us now describe the explanatory variables for our estimation in detail. Basic statistics of all variables are provided in Table 1. Following the extant studies examining the decision of going public, to estimate the propensity of going public $P(x)$ in (1), we employ firm size measured by the natural logarithm of firms' total asset (LN_ASSET), liquidity measured by the ratio of cash to total asset ($CASH_RATIO$), debt dependence measured by the ratio of debt to total asset ($DEBT_RATIO$), and pre-IPO TFP (TFP) as the determinants. For all these explanatory variables, we use a one-year lag to eliminate possible endogeneity problems originating from the reverse causality running from the dependent variable to the independent variables. In order to control for year-specific effect capturing, for example, the state of stock markets (Ritter 1984, 1991; Baker and Wurgler 2000), we also include the year dummy variable in the list of our explanatory variables. To control for industry-level shocks that affect the firm's IPO decision (Giot and Schwiendbacher 2007), we classify the firms into 21 industries and add four industry dummies accordingly.

Since we will compare IPO firms without being invested by VCs and non-IPO firms as well as IPO firms with being invested by VCs and non-IPO firms separately, we estimate the propensity scores associated with going public without being invested by VCs and with invested by VCs, separately. In both estimations, we use non-IPO firms without being invested by VCs as control group.

5. Empirical Results

In the following subsections, we (1) show the result of the logit estimation on the determinants of IPO (Section 5.1); and (2) examine the *ex post* performance differences between A) IPO firms without being invested by VCs and private firms (Section 5.2) as well as between B) IPO firms with being invested by VCs and private firms, using matched samples (Section 5.3).

5.1 Propensity Score

The estimated results for the probability of going public are shown in Table 2 (a) and (b), the former and latter of which account for the case of going public without and that with being invested by VCs, respectively.

In the case of going public without being invested by VCs, first, we find that *LN_SIZE* and *CASH_RATIO* have positive and significant coefficients. On the other hand, *DEBT_RATIO* takes positive and significant coefficient. These results imply that the probability of going public for larger and liquid firms that were not depending too much on debt is higher. One puzzling result is that TFP of IPO firms tend be lower than the counterpart unlisted firms. All the first two results are obtained for the case of going public without being invested by VCs with almost identical coefficients associated with the variables. One difference is in the coefficient associated with firm size. Namely, the coefficient in the case of IPO with VCs is much smaller than the case of IPO without VCs. This implies that the size of firms going public with VCs tend to be much more diverse than the firms going public without VCs. We use this result in our matching process where firms are matched separately for each year using one-to-one nearest neighbor matching.

5.2 Difference-in-Differences Estimation

The results for the difference-in-differences estimation associated with IPO firms not invested by VCs are shown in Table 3. First, from the upper panel of the table showing the difference-in-differences coefficients in our estimation, we find a negative and significant negative impact running from IPO to firms' ROA in the case of the comparison between $t - 1$ to $t + 3$ while there is no impact on TFP and sales-to-employee ratio. This is somewhat consistent with the results in Kutsuna et al. (2000). The magnitude of this negative impact in our estimation is around -4%, which is economically sizable since the average level of ROA in our sample is 9.7 % in the case of IPO firms with VCs. Second, the lower panel of the table, which tabulates the treatment effect of IPO, shows that firms going public tend to show higher ROA prior to IPO. The advantage in ROA is estimated around 5 to 6%. These two results imply that a large part of the performance difference prior to IPO diminished after IPO in our sample.

According to the discussion in, for example, Asker et al. (2012), corporate managers exposed to the pressure in stock market could take suboptimal decisions. It is one interpretation of this result that the drop in firm performance is due to this distorted firm behavior. Note that although not shown in the table, the difference-in-differences coefficient in the case of the comparison between $t - 1$ to $t - 4$ and $t - 1$ to $t - 5$ are -0.043 and not significantly different from zero, respectively. To search for the sources of the above-mentioned result in greater detail, we split the sample into two groups by using the information of the timing of IPO. Table 4 and 5 show the difference-in-differences coefficients in the case of firms going public until 2005 and after 2005, respectively. We split the sample in this way since the famous scandal issue in the emerging market (i.e., "Livedoor shock") occurred during the hot market periods in the first half of 2000s. It is claimed that it became much difficult for unlisted firms to go public after 2005. We conjecture that such a structural change certainly affected the post-IPO performance of newly listed firms. Notably, the coefficients in Table 4 exhibit the similar patten as in Table 2 while there is virtually no negative

DID effect found in Table 5. Considering the fact that the treatment effect in Table 4 and 5 are almost identical, the results imply that not only the firms showing better performance prior to IPO but also having a certain prospect after going public were allowed to be listed after 2005.

We should note that similar to Table 3, the difference-in-differences coefficients in the case of the comparison between $t - 1$ to $t - 4$ and $t - 1$ to $t - 5$ in the case of Table 4 are -0.043 and not significantly different from zero, respectively. This necessitates us to be cautious about the interpretation of the negative impact found in Table 4.

5.3 Impact of Venture Capital Investments

How the participation of VCs in pre-IPO investment is interacted with the result obtained above? The results for the difference-in-differences estimation associated with IPO firms invested by VCs are shown in Table 6. First, from the upper panel of the table, we find a negative and significant impact from IPO on firms' ROA in the case of the comparison between $t - 1$ to $t + 2$ and $t - 1$ to $t + 3$ while there is no impact on TFP. We also confirm that IPO has a significant and positive impact on sales-to-employee ratio in the case of the comparison between $t - 1$ to $t + 3$. Note that the result associated with the sales-to-employee ratio is found to be positive only in this period and not significant in the case of the comparison between $t - 1$ to $t + 4$ and $t - 1$ to $t + 5$ while the negative impact on ROA keeps being significant in such long-run comparison (i.e., -0.044 and -0.032). This means that the participation of VCs in investments exacerbates the negative impact of IPO on firms' performance (i.e., ROA). More specifically, the negative impact is found not only to be short-term but also long-term. As discussed in the extant literature, VCs might have different motivation from firms' long-run growth. This result implies that such negative impact overwhelms the positive impact potentially coming from VCs' strategic, management, marketing, and administrative expertise. Such an analysis on firms' long-term performance dynamics also gives us a

conjecture that the drop of performance reflects the large size of capital investment which becomes feasible thanks to IPO. For testing this conjecture, we implement the same DID estimation for firms' asset size (*LN_ASSET*). Evidently, we confirm a positive and significant impact from IPO on firms' size in the case of all the comparisons in Table 6. This implies that VC-backed IPO firms tended to invest more than their counterpart non-IPO firms. As the most important result, however, this overinvestment could not lead to the improvement in ROA as mentioned above. This is consistent with the Empire-building motives discussed in Stulz (1990).

Second, from the lower panel of Table 6, we find that IPO firms with being invested by VCs tend to perform well compared to their counterpart unlisted firms. It is important to note that the performance advantage prior to IPO (i.e., 3 to 4%) is completely offset by the adverse impact of IPO under the investment by VCs (i.e., 4 to 5%). This fact becomes more apparent in the case of firms going public until 2005 (see Table 7 and 8). This confirms the validity of the claim that some of VCs induce unpromising venture firms to go public based on other motivations.

6. Subsample Analyses

One important feature of VC investment is syndication. Brander et al. (2002) reports that 60% of VC investments in Canada were syndicated in 1993. According to Wright and Lockett (2003), the shares of syndicated VCs are 30% in Europe and 60% in the U.S. (in 2000s). In our data, 89% of Japanese venture firms accomplishing IPO were financed by syndicated VCs in the last decade.

First, the average number of VCs included at the first stage of investment is 2.4 while that at the timing IPO is 4.1, which means VC syndication tends to be exposed to certain dynamics in terms of its size. From a difference perspective, second, we find that VC syndication contains various types

of VC. The type of VC consists of, for example, bank-dependent, security firm-dependent, insurance company-dependent, trade company-dependent ("Shosha" in Japanese), corporate (i.e., non-financial firm-dependent), mixed origination, foreign-owned, foreign-located, independent, university, government, and others. The average number of VC types included at the first stage of investment is 1.8 while that at the timing IPO is 2.5, which means the member VCs in a syndicate tends to be dynamic and heterogeneous.

In this section, first, we test how the composition of VC syndication (i.e., the number of VCs and the number of VC types) are interacted with the negative impact of IPO established in the previous section. Second, we also examine whether inclusion of a specific type of VC alleviates the negative impact. Namely, following the discussion of Tykvová and Walz (2007), we conjecture that the involvement of independent and/or foreign-owned VCs contributes to better performance of IPO firms after going public.

6.1 Size of Syndication and Heterogeneity of VCs

Table 9 shows the similar difference-in-differences estimation for the case of IPO firms with being invested by many VCs or a small number of VCs at the first investment round. We split the sample at the median level of VC number as of first round (i.e., one). The result clearly shows that the adverse impact from going public is found in the case of being invested by a small number of VCs. Interestingly, IPO firms with being invested by a small number of VCs tend to show higher performance prior to IPO, which is to large extent offset by the adverse impact of IPO. Table 10 repeats the same exercise by splitting the sample at the median level of the number VC types included at the first round investment. The obtained implication is almost identical to the ones mentioned above.

As discussed in a number of extant literature (e.g., Giot and Schvienbacher 2006; Cumming

2006; Miyakawa and Talizawa 2012), the characteristics and the composition of VC syndicate are crucial for the investment performance done by VCs. The presenting result is consistent with these empirical discussions.

6.2 Inclusion of Specific Types of VCs

Table 11 and 12 shows the similar difference-in-differences estimation for the case of IPO firms with being invested by independent VCs or not, and foreign VCs or not. The result shows that the inclusion of foreign VCs alleviates the adverse impact from going public while the impact originating from the inclusion of independent VCs is sustained for relatively short-term.

As Tykvová and Walz (2007) discusses, among various types of VCs, foreign-based VCs are keen to the prospect of firms' future performance. The current results confirm their views.

7. Conclusion

This paper examined the effect of Initial Public Offering (IPO) on firms' performance by utilizing a unique firm-level panel dataset that allows us to identify newly listed firms and firms keeping unlisted status. We find newly listed firms showed better performance than their counterpart of never-listed firms prior to IPO while the difference in the performance partly diminished after IPO. This implies that firms' distorted behavior originating from prevents newly listed firms from performing. As one important finding, this result is mainly driven by the sample firms going public right after the introduction of the emerging market in the early 2000s. We also find that the participation of VCs in investments exacerbates such negative impact of IPO on firms' performance. The adverse impact on the firm performance is more sizable among IPO firms which are invested by VC syndicate consisting of smaller number of less heterogeneous VCs, or not including foreign VCs.

These results suggest that the timing of going public, and the composition of VC syndicates are related to the post-IPO performance of newly listed firms.

The research presented in this paper could be expanded in a number of directions. One such direction would be to examine other financial characteristics of newly listed firms. It is one promising direction to apply the same methodology employed in this paper to, for example, firms' leverage, bank-dependence, and/or cash holding behavior. Second, another interesting analysis would be to focus on bank-dependent VCs. As studied in a few extant studies (Hamao et al. 2000; Hellmann et al. 2008), bank-dependent VCs could have different motives for the participation to investment (e.g., post-IPO bank lending by affiliated banks). It is an important research issue to examine the motivation of these VCs more extensively.

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Appendix: The multilateral TFP index

As detailed in Fukao et al. (2011), the TFP level of firm i in industry j in year t , $TFP_{i,j,t}$ is defined in comparison with the TFP level of a hypothetical representative firm in the benchmark year t_0 in industry j . In the presenting paper, the benchmark year t_0 is set to the year 1995 and the firm-level TFP level is calculated as follows, using the multilateral TFP index method developed by Good et al. (1997):

$$\begin{aligned} \text{LN}(TFP_{i,j,t}) &= \left\{ \text{LN}(Q_{i,j,t}) - \overline{\text{LN}(Q_{j,t})} \right\} - \sum_{k=1}^n (S_{i,k,j,t} + \overline{S_{k,j,t}}) \left\{ \text{LN}(X_{i,k,j,t}) - \overline{\text{LN}(X_{k,j,t})} \right\} \\ &\quad \text{for } t = t_0 \\ \text{LN}(TFP_{i,j,t}) &= \left\{ \text{LN}(Q_{i,j,t}) - \overline{\text{LN}(Q_{j,t})} \right\} - \frac{1}{2} \sum_{k=1}^n (S_{i,k,j,t} + \overline{S_{k,j,t}}) \left\{ \text{LN}(X_{i,k,j,t}) - \overline{\text{LN}(X_{k,j,t})} \right\} \\ &+ \sum_{s=t_0+1}^t \left\{ \overline{\text{LN}(Q_{j,s})} - \overline{\text{LN}(Q_{j,s-1})} \right\} - \sum_{s=t_0+1}^t \sum_{k=1}^n \frac{1}{2} (\overline{S_{k,j,s}} + \overline{S_{k,j,s-1}}) \left\{ \overline{\text{LN}(X_{k,j,s})} - \overline{\text{LN}(X_{k,j,s-1})} \right\} \\ &\quad \text{for } t > t_0 \\ \text{LN}(TFP_{i,j,t}) &= \left\{ \text{LN}(Q_{i,j,t}) - \overline{\text{LN}(Q_{j,t})} \right\} - \frac{1}{2} \sum_{k=1}^n (S_{i,k,j,t} + \overline{S_{k,j,t}}) \left\{ \text{LN}(X_{i,k,j,t}) - \overline{\text{LN}(X_{k,j,t})} \right\} \\ &- \sum_{s=t+1}^{t_0} \left\{ \overline{\text{LN}(Q_{j,s})} - \overline{\text{LN}(Q_{j,s-1})} \right\} + \sum_{s=t+1}^{t_0} \sum_{k=1}^n \frac{1}{2} (\overline{S_{k,j,s}} + \overline{S_{k,j,s-1}}) \left\{ \overline{\text{LN}(X_{k,j,s})} - \overline{\text{LN}(X_{k,j,s-1})} \right\} \\ &\quad \text{for } t < t_0 \end{aligned}$$

where $Q_{i,j,t}$ stands for the real output (real sales) of firm i (in industry j) in year t , $X_{i,k,j,t}$ represents the real input of production factor k of firm i (in industry j) in year t , and $S_{i,j,k,t}$ is the cost share of production factor k at firm i (in industry j) in year t .⁵ $\overline{\text{LN}(Q_{j,t})}$ denotes the arithmetic average of the log value of the output, in year t , of all firms in industry j to which firm i belongs, while $\overline{\text{LN}(X_{k,j,t})}$ stands for the arithmetic average of the log value of the input of production factor k , in year t , of all firms in industry j to which firm i belongs. Finally, $\overline{S_{k,j,t}}$ is the arithmetic average of the cost share of the input of production factor k , in year t , of all firms in industry j to which firm i belongs.

⁵ Since our dataset does not contain the labor input prior to IPO, we use the labor input as of the timing of establishing IPO instead of this.

Figures and Tables

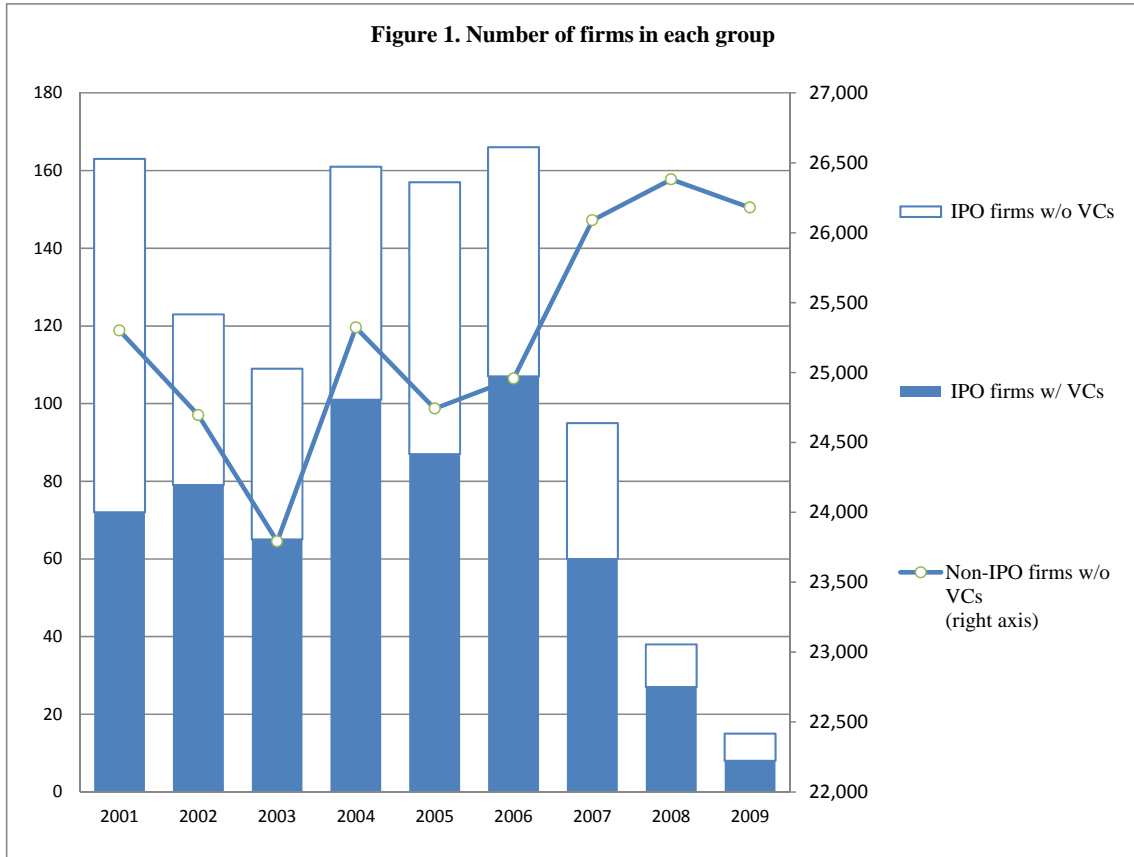


Table 1: Summary statistics

IPO firms w/ VCs	Mean	Median	Maximum	Minimum	Standard Deviation	Number of Sample Firms
<i>CASH_RATIO</i>	0.110	0.108	0.533	-1.115	0.112	606
<i>DEBT_RATIO</i>	0.122	0.072	0.704	0.000	0.133	606
<i>LN_ASSET</i>	8.376	8.290	12.103	5.601	0.987	606
<i>TFP</i>	0.023	0.026	1.777	-2.450	0.321	605
<i>ROA</i>	0.097	0.092	0.533	-1.142	0.113	606
<i>SALES-to-EMPLOYEE RATIO</i>	73.946	38.147	1168.540	0.000	119.196	606

IPO firms w/o VCs	Mean	Median	Maximum	Minimum	Standard Deviation	Number of Sample Firms
<i>CASH_RATIO</i>	0.117	0.106	0.472	-0.333	0.077	421
<i>DEBT_RATIO</i>	0.131	0.094	0.667	0.000	0.129	421
<i>LN_ASSET</i>	9.258	9.069	14.562	5.885	1.392	421
<i>TFP</i>	0.043	0.027	1.299	-2.144	0.282	419
<i>ROA</i>	0.104	0.092	0.472	-0.343	0.076	421
<i>SALES-to-EMPLOYEE RATIO</i>	93.973	52.557	1248.300	1.582	124.171	421

Non-IPO firms	Mean	Median	Maximum	Minimum	Standard Deviation	Number of Sample Firms
<i>CASH_RATIO</i>	0.071	0.059	174.767	-7.980	0.384	227,332
<i>DEBT_RATIO</i>	0.709	0.727	360.000	-1.175	1.052	227,332
<i>LN_ASSET</i>	8.083	8.008	14.903	1.099	1.214	227,332
<i>TFP</i>	-0.015	-0.009	3.086	-7.546	0.537	213,900
<i>ROA</i>	0.036	0.027	174.767	-36.167	0.387	227,332
<i>SALES-to-EMPLOYEE RATIO</i>	48.556	27.093	13440.510	0.000	116.709	227,456

Table 2: Logit estimation (IPO w/o and w/ VCs)

Panel (a) IPO without VCs

Dependent variable	IPO w/o VC dummy	
	Coef.	Std. Err
<i>CASH_RATIO (t-1)</i>	1.307	0.277 ***
<i>DEBT_RATIO (t-1)</i>	-12.324	0.501 ***
<i>LN_ASSET (t-1)</i>	0.641	0.051 ***
<i>TFP(t-1)</i>	-0.278	0.132 **
<i>Const.</i>	-16.132	666.489
Year dummy	Yes	
Industry dummy	Yes	
Number of obs	133529	
LR chi2(55)	2289.49	
Prob > chi2	0	
Pseudo R2	0.5184	

Panel (b) IPO with VCs

Dependent variable	IPO w/ VC dummy	
	Coef.	Std. Err
<i>CASH_RATIO (t-1)</i>	1.251	0.250 ***
<i>DEBT_RATIO (t-1)</i>	-12.782	0.399 ***
<i>LN_ASSET (t-1)</i>	0.074	0.045 *
<i>TFP(t-1)</i>	-0.453	0.103 ***
<i>Const.</i>	-10.260	466.945
Year dummy	Yes	
Industry dummy	Yes	
Number of obs	138830	
LR chi2(55)	3730.63	
Prob > chi2	0	
Pseudo R2	0.5386	

Note: ***, ** and * show statistical significance at the 1%, 5% and 10% level.

Table 3: DID effect (i.e., IPO w/o VC vs. non-IPO) & Treatment effect

DID effect (w/o VCs)

		Coef.	Std. Err	t-value	Obs.
2 windows((t-1)-(t+1))	<i>TFP</i>	-0.013	0.072	-0.180	606
	<i>ROA</i>	-0.019	0.013	-1.410	608
	<i>SALES_{t0}-EMPLOYEE RATIO</i>	21.126	24.824	0.850	608
	<i>LN_ASSET</i>	0.290	0.2181	1.33	608
3 windows((t-1)-(t+2))	<i>TFP</i>	-0.016	0.081	-0.190	522
	<i>ROA</i>	-0.024	0.015	-1.610	524
	<i>SALES_{t0}-EMPLOYEE RATIO</i>	14.120	27.170	0.520	524
	<i>LN_ASSET</i>	0.369	0.2334	1.58	524
4 windows((t-1)-(t+3))	<i>TFP</i>	0.025	0.105	0.240	406
	<i>ROA</i>	-0.037 **	0.017	-2.130	408
	<i>SALES_{t0}-EMPLOYEE RATIO</i>	65.411	55.304	1.180	408
	<i>LN_ASSET</i>	0.336	0.2857	1.17	408

Treatment effect

		Coef.	Std. Err	t-value	Obs.
2 windows((t-1)-(t+1))	<i>TFP</i>	-0.014	0.051	-0.270	606
	<i>ROA</i>	0.052 ***	0.009	5.530	608
	<i>SALES_{t0}-EMPLOYEE RATIO</i>	14.633	17.496	0.840	608
	<i>LN_ASSET</i>	0.307 **	0.1537	2	608
3 windows((t-1)-(t+2))	<i>TFP</i>	0.003	0.057	0.050	522
	<i>ROA</i>	0.057 ***	0.011	5.390	524
	<i>SALES_{t0}-EMPLOYEE RATIO</i>	18.146	19.175	0.950	524
	<i>LN_ASSET</i>	0.307 *	0.1647	1.75	524
4 windows((t-1)-(t+3))	<i>TFP</i>	0.057	0.074	0.770	406
	<i>ROA</i>	0.063 ***	0.012	5.200	408
	<i>SALES_{t0}-EMPLOYEE RATIO</i>	23.417	39.010	0.600	408
	<i>LN_ASSET</i>	0.351 *	0.2016	1.74	408

Table 4: 2001-2005 DID effect (i.e., IPO w/o VC vs. non-IPO) & Treatment effect

DID effect (2001-2005)

		Coef.	Std. Err	t-value	Obs.
2 windows((t-1)-(t+1))	<i>TFP</i>	-0.014	0.081	-0.180	386
	<i>ROA</i>	-0.023	0.017	-1.320	388
	<i>SALES-to-EMPLOYEE RATIO</i>	7.370	15.843	0.470	388
	<i>LN_ASSET</i>	0.302	0.2568	1.17	388
3 windows((t-1)-(t+2))	<i>TFP</i>	-0.024	0.083	-0.290	338
	<i>ROA</i>	-0.025	0.018	-1.400	340
	<i>SALES-to-EMPLOYEE RATIO</i>	1.852	18.830	0.100	340
	<i>LN_ASSET</i>	0.365	0.2778	1.31	340
4 windows((t-1)-(t+3))	<i>TFP</i>	-0.017	0.105	-0.160	302
	<i>ROA</i>	-0.041 **	0.019	-2.100	304
	<i>SALES-to-EMPLOYEE RATIO</i>	1.250	18.185	0.070	304
	<i>LN_ASSET</i>	0.314	0.3018	1.04	304

Treatment effect

		Coef.	Std. Err	t-value	Obs.
2 windows((t-1)-(t+1))	<i>TFP</i>	-0.006	0.057	-0.110	386
	<i>ROA</i>	0.058 ***	0.012	4.850	388
	<i>SALES-to-EMPLOYEE RATIO</i>	32.197 ***	11.145	2.890	388
	<i>LN_ASSET</i>	0.445 **	0.1807	2.46	388
3 windows((t-1)-(t+2))	<i>TFP</i>	-0.020	0.058	-0.350	338
	<i>ROA</i>	0.057 ***	0.013	4.500	340
	<i>SALES-to-EMPLOYEE RATIO</i>	28.400 **	13.275	2.140	340
	<i>LN_ASSET</i>	0.452 **	0.1958	2.31	340
4 windows((t-1)-(t+3))	<i>TFP</i>	0.024	0.074	0.320	302
	<i>ROA</i>	0.064 ***	0.014	4.650	304
	<i>SALES-to-EMPLOYEE RATIO</i>	37.278 ***	12.858	2.900	304
	<i>LN_ASSET</i>	0.496 **	0.2134	2.32	304

Table 5: 2006- DID effect (i.e., IPO w/o VC vs. non-IPO) & Treatment effect

DID effect (2006-)

		Coef.	Std. Err	t-value	Obs.
2 windows((t-1)-(t+1))	<i>TFP</i>	-0.011	0.141	-0.080	218
	<i>ROA</i>	-0.012	0.021	-0.560	220
	<i>SALES-to-EMPLOYEE RATIO</i>	46.191	62.587	0.740	220
	<i>LN_ASSET</i>	0.276	0.3987	0.69	220
3 windows((t-1)-(t+2))	<i>TFP</i>	-0.002	0.175	-0.010	182
	<i>ROA</i>	-0.021	0.026	-0.810	184
	<i>SALES-to-EMPLOYEE RATIO</i>	36.520	69.407	0.530	184
	<i>LN_ASSET</i>	0.385	0.4192	0.92	184
4 windows((t-1)-(t+3))	<i>TFP</i>	0.153	0.278	0.550	102
	<i>ROA</i>	-0.023	0.037	-0.630	104
	<i>SALES-to-EMPLOYEE RATIO</i>	246.315	208.638	1.180	104
	<i>LN_ASSET</i>	0.401	0.6949	0.58	104

Treatment effect

		Coef.	Std. Err	t-value	Obs.
2 windows((t-1)-(t+1))	<i>TFP</i>	-0.026	0.099	-0.260	218
	<i>ROA</i>	0.041 ***	0.015	2.700	220
	<i>SALES-to-EMPLOYEE RATIO</i>	-17.079	44.256	-0.390	220
	<i>LN_ASSET</i>	0.061	0.2819	0.22	220
3 windows((t-1)-(t+2))	<i>TFP</i>	0.046	0.123	0.380	182
	<i>ROA</i>	0.055 ***	0.019	2.990	184
	<i>SALES-to-EMPLOYEE RATIO</i>	-1.066	49.078	-0.020	184
	<i>LN_ASSET</i>	-0.016	0.2964	-0.05	184
4 windows((t-1)-(t+3))	<i>TFP</i>	0.151	0.192	0.780	102
	<i>ROA</i>	0.063 **	0.026	2.410	104
	<i>SALES-to-EMPLOYEE RATIO</i>	-14.947	146.103	-0.100	104
	<i>LN_ASSET</i>	-0.070	0.4866	-0.14	104

Table 6: DID effect (i.e., IPO w/ VC vs. non-IPO) & Treatment effect

DID effect (w/ VCs)

		Coef.	Std. Err	t-value	Obs.
2 windows((t-1)-(t+1))	<i>TFP</i>	-0.048	0.063	-0.760	888
	<i>ROA</i>	-0.003	0.023	-0.140	896
	<i>SALES-to-EMPLOYEE RATIO</i>	17.145	11.248	1.520	896
	<i>LN_ASSET</i>	0.368 **	0.1523	2.42	896
3 windows((t-1)-(t+2))	<i>TFP</i>	-0.059	0.068	-0.870	786
	<i>ROA</i>	-0.048 ***	0.014	-3.360	792
	<i>SALES-to-EMPLOYEE RATIO</i>	13.583	9.762	1.390	792
	<i>LN_ASSET</i>	0.479 ***	0.1672	2.86	792
4 windows((t-1)-(t+3))	<i>TFP</i>	-0.083	0.074	-1.110	650
	<i>ROA</i>	-0.043 ***	0.014	-3.050	656
	<i>SALES-to-EMPLOYEE RATIO</i>	20.120 **	9.652	2.080	656
	<i>LN_ASSET</i>	0.544 ***	0.1846	2.95	656

Treatment effect

		Coef.	Std. Err	t-value	Obs.
2 windows((t-1)-(t+1))	<i>TFP</i>	-0.056	0.044	-1.260	888
	<i>ROA</i>	0.016	0.016	1.010	896
	<i>SALES-to-EMPLOYEE RATIO</i>	36.675 ***	7.927	4.630	896
	<i>LN_ASSET</i>	0.315 ***	0.1074	2.94	896
3 windows((t-1)-(t+2))	<i>TFP</i>	-0.059	0.048	-1.240	786
	<i>ROA</i>	0.035 ***	0.010	3.500	792
	<i>SALES-to-EMPLOYEE RATIO</i>	30.952 ***	6.885	4.500	792
	<i>LN_ASSET</i>	0.291 **	0.1179	2.47	792
4 windows((t-1)-(t+3))	<i>TFP</i>	-0.012	0.052	-0.220	650
	<i>ROA</i>	0.033 ***	0.010	3.340	656
	<i>SALES-to-EMPLOYEE RATIO</i>	26.174 ***	6.773	3.860	656
	<i>LN_ASSET</i>	0.315 **	0.1296	2.43	656

Table 7: 2001-2005 DID effect (i.e., IPO w/ VC vs. non-IPO) & Treatment effect

DID effect (2001-2005)

		Coef.	Std. Err	t-value	Obs.
2 windows((t-1)-(t+1))	<i>TFP</i>	-0.088	0.071	-1.230	570
	<i>ROA</i>	-0.017	0.014	-1.220	572
	<i>SALES_{-t0}-EMPLOYEE RATIO</i>	20.140	12.260	1.640	572
	<i>LN_ASSET</i>	0.432 **	0.1900	2.27	572
3 windows((t-1)-(t+2))	<i>TFP</i>	-0.091	0.070	-1.300	554
	<i>ROA</i>	-0.036 **	0.017	-2.180	556
	<i>SALES_{-t0}-EMPLOYEE RATIO</i>	18.118	11.160	1.620	556
	<i>LN_ASSET</i>	0.570 ***	0.1968	2.9	556
4 windows((t-1)-(t+3))	<i>TFP</i>	-0.084	0.076	-1.100	552
	<i>ROA</i>	-0.042 ***	0.016	-2.640	556
	<i>SALES_{-t0}-EMPLOYEE RATIO</i>	23.199 **	11.120	2.090	556
	<i>LN_ASSET</i>	0.599 ***	0.2004	2.99	556

Treatment effect

		Coef.	Std. Err	t-value	Obs.
2 windows((t-1)-(t+1))	<i>TFP</i>	0.035	0.050	0.700	570
	<i>ROA</i>	0.040 ***	0.010	4.150	572
	<i>SALES_{-t0}-EMPLOYEE RATIO</i>	34.403 ***	8.639	3.980	572
	<i>LN_ASSET</i>	0.387 ***	0.1339	2.89	572
3 windows((t-1)-(t+2))	<i>TFP</i>	0.031	0.049	0.620	554
	<i>ROA</i>	0.039 ***	0.012	3.350	556
	<i>SALES_{-t0}-EMPLOYEE RATIO</i>	32.984 ***	7.877	4.190	556
	<i>LN_ASSET</i>	0.353 **	0.1389	2.54	556
4 windows((t-1)-(t+3))	<i>TFP</i>	0.044	0.053	0.830	552
	<i>ROA</i>	0.035 ***	0.011	3.100	556
	<i>SALES_{-t0}-EMPLOYEE RATIO</i>	30.417 ***	7.806	3.900	556
	<i>LN_ASSET</i>	0.345 **	0.1407	2.45	556

Table 8: 2006- DID effect (i.e., IPO w/ VC vs. non-IPO) & Treatment effect

DID effect (2006-)

		Coef.	Std. Err	t-value	Obs.
2 windows((t-1)-(t+1))	<i>TFP</i>	0.019	0.122	0.160	318
	<i>ROA</i>	0.021	0.058	0.360	324
	<i>SALES_{t0}-EMPLOYEE RATIO</i>	11.861	22.447	0.530	324
	<i>LN_ASSET</i>	0.256	0.2549	1	324
3 windows((t-1)-(t+2))	<i>TFP</i>	0.014	0.155	0.090	234
	<i>ROA</i>	-0.077 ***	0.028	-2.770	236
	<i>SALES_{t0}-EMPLOYEE RATIO</i>	2.847	19.570	0.150	236
	<i>LN_ASSET</i>	0.261	0.3154	0.83	236
4 windows((t-1)-(t+3))	<i>TFP</i>	-0.080	0.238	-0.340	98
	<i>ROA</i>	-0.049 *	0.026	-1.850	100
	<i>SALES_{t0}-EMPLOYEE RATIO</i>	1.811	8.564	0.210	100
	<i>LN_ASSET</i>	0.224	0.4708	0.48	100

Treatment effect

		Coef.	Std. Err	t-value	Obs.
2 windows((t-1)-(t+1))	<i>TFP</i>	-0.217 **	0.085	-2.550	318
	<i>ROA</i>	-0.026	0.041	-0.630	324
	<i>SALES_{t0}-EMPLOYEE RATIO</i>	40.686 **	15.823	2.570	324
	<i>LN_ASSET</i>	0.190	0.1797	1.06	324
3 windows((t-1)-(t+2))	<i>TFP</i>	-0.269 **	0.109	-2.480	234
	<i>ROA</i>	0.026	0.019	1.360	236
	<i>SALES_{t0}-EMPLOYEE RATIO</i>	26.170 *	13.779	1.900	236
	<i>LN_ASSET</i>	0.148	0.2220	0.67	236
4 windows((t-1)-(t+3))	<i>TFP</i>	-0.319 *	0.165	-1.930	98
	<i>ROA</i>	0.024	0.019	1.280	100
	<i>SALES_{t0}-EMPLOYEE RATIO</i>	2.571	5.995	0.430	100
	<i>LN_ASSET</i>	0.145	0.3295	0.44	100

Table 9: #(VCs) DID effect (i.e., IPO w/ VC vs. non-IPO) & Treatment effect
 #(VCs) at first investment round High

DID effect		Coef.	Std. Err	t-value	Obs.
2 windows((t-1)-(t+1))	<i>TFP</i>	-0.002	0.098	-0.020	330
	<i>ROA</i>	0.023	0.058	0.410	332
	<i>SALES-10-EMPLOYEE RATIO</i>	20.874	21.385	0.980	332
	<i>LN_ASSET</i>	0.365	0.2512	1.45	332
3 windows((t-1)-(t+2))	<i>TFP</i>	0.007	0.104	0.070	302
	<i>ROA</i>	-0.037	0.028	-1.320	304
	<i>SALES-10-EMPLOYEE RATIO</i>	16.461	20.713	0.790	304
	<i>LN_ASSET</i>	0.373	0.2712	1.37	304
4 windows((t-1)-(t+3))	<i>TFP</i>	-0.058	0.112	-0.520	234
	<i>ROA</i>	-0.030	0.028	-1.090	236
	<i>SALES-10-EMPLOYEE RATIO</i>	27.116	17.715	1.530	236
	<i>LN_ASSET</i>	0.478	0.3224	1.48	236

Treatment effect		Coef.	Std. Err	t-value	Obs.
2 windows((t-1)-(t+1))	<i>TFP</i>	-0.067	0.069	-0.980	330
	<i>ROA</i>	-0.017	0.041	-0.430	332
	<i>SALES-10-EMPLOYEE RATIO</i>	31.681 **	15.076	2.100	332
	<i>LN_ASSET</i>	0.230	0.1771	1.3	332
3 windows((t-1)-(t+2))	<i>TFP</i>	-0.065	0.073	-0.890	302
	<i>ROA</i>	0.026	0.020	1.290	304
	<i>SALES-10-EMPLOYEE RATIO</i>	37.615 **	14.645	2.570	304
	<i>LN_ASSET</i>	0.361 *	0.1918	1.88	304
4 windows((t-1)-(t+3))	<i>TFP</i>	-0.036	0.079	-0.460	234
	<i>ROA</i>	0.027	0.020	1.390	236
	<i>SALES-10-EMPLOYEE RATIO</i>	18.387	12.473	1.470	236
	<i>LN_ASSET</i>	0.235	0.2270	1.03	236

#(VCs) at first investment round Low

DID effect		Coef.	Std. Err	t-value	Obs.
2 windows((t-1)-(t+1))	<i>TFP</i>	-0.096	0.094	-1.020	464
	<i>ROA</i>	-0.027 **	0.013	-2.060	468
	<i>SALES-10-EMPLOYEE RATIO</i>	17.704	15.044	1.180	468
	<i>LN_ASSET</i>	0.433 **	0.2061	2.1	468
3 windows((t-1)-(t+2))	<i>TFP</i>	-0.119	0.101	-1.190	400
	<i>ROA</i>	-0.068 ***	0.018	-3.870	404
	<i>SALES-10-EMPLOYEE RATIO</i>	14.684	10.587	1.390	404
	<i>LN_ASSET</i>	0.647 ***	0.2301	2.81	404
4 windows((t-1)-(t+3))	<i>TFP</i>	-0.108	0.111	-0.970	342
	<i>ROA</i>	-0.059 ***	0.018	-3.330	344
	<i>SALES-10-EMPLOYEE RATIO</i>	19.815	13.437	1.470	344
	<i>LN_ASSET</i>	0.697 ***	0.2459	2.83	344

Treatment effect		Coef.	Std. Err	t-value	Obs.
2 windows((t-1)-(t+1))	<i>TFP</i>	-0.062	0.066	-0.940	464
	<i>ROA</i>	0.037 ***	0.009	3.990	468
	<i>SALES-10-EMPLOYEE RATIO</i>	44.715 ***	10.591	4.220	468
	<i>LN_ASSET</i>	0.299 ***	0.1451	2.06	468
3 windows((t-1)-(t+2))	<i>TFP</i>	-0.068	0.071	-0.960	400
	<i>ROA</i>	0.044 ***	0.012	3.600	404
	<i>SALES-10-EMPLOYEE RATIO</i>	29.999 ***	7.449	4.030	404
	<i>LN_ASSET</i>	0.204	0.1619	1.26	404
4 windows((t-1)-(t+3))	<i>TFP</i>	0.007	0.077	0.090	342
	<i>ROA</i>	0.039 ***	0.012	3.170	344
	<i>SALES-10-EMPLOYEE RATIO</i>	33.305 ***	9.392	3.550	344
	<i>LN_ASSET</i>	0.243	0.1719	1.42	344

Table 10: #(VC TYPES) DID effect (i.e., IPO w/ VC vs. non-IPO) & Treatment effect
 #(VC TYPES) at first investment round High

DID effect		Coef.	Std. Err	t-value	Obs.
2 windows((t-1)-(t+1))	<i>TFP</i>	-0.001	0.108	-0.010	286
	<i>ROA</i>	0.032	0.066	0.480	288
	<i>SALES₋₁₀-EMPLOYEE RATIO</i>	13.305	21.428	0.620	288
	<i>LN_ASSET</i>	0.356	0.2682	1.33	288
3 windows((t-1)-(t+2))	<i>TFP</i>	0.002	0.114	0.020	262
	<i>ROA</i>	-0.046	0.032	-1.430	264
	<i>SALES₋₁₀-EMPLOYEE RATIO</i>	14.659	23.058	0.640	264
	<i>LN_ASSET</i>	0.417	0.2908	1.44	264
4 windows((t-1)-(t+3))	<i>TFP</i>	-0.097	0.132	-0.740	190
	<i>ROA</i>	-0.045	0.033	-1.350	192
	<i>SALES₋₁₀-EMPLOYEE RATIO</i>	26.164 *	13.880	1.890	192
	<i>LN_ASSET</i>	0.499	0.3575	1.4	192

Treatment effect		Coef.	Std. Err	t-value	Obs.
2 windows((t-1)-(t+1))	<i>TFP</i>	-0.079	0.076	-1.040	286
	<i>ROA</i>	-0.021	0.047	-0.440	288
	<i>SALES₋₁₀-EMPLOYEE RATIO</i>	31.260 **	15.099	2.070	288
	<i>LN_ASSET</i>	0.144	0.1890	0.76	288
3 windows((t-1)-(t+2))	<i>TFP</i>	-0.065	0.080	-0.810	262
	<i>ROA</i>	0.031	0.023	1.380	264
	<i>SALES₋₁₀-EMPLOYEE RATIO</i>	37.631 **	16.304	2.310	264
	<i>LN_ASSET</i>	0.283	0.2056	1.38	264
4 windows((t-1)-(t+3))	<i>TFP</i>	-0.032	0.093	-0.340	190
	<i>ROA</i>	0.034	0.023	1.440	192
	<i>SALES₋₁₀-EMPLOYEE RATIO</i>	9.399	9.763	0.960	192
	<i>LN_ASSET</i>	0.050	0.2515	0.2	192

#(VC TYPES) at first investment round Low

DID effect		Coef.	Std. Err	t-value	Obs.
2 windows((t-1)-(t+1))	<i>TFP</i>	-0.088	0.088	-1.010	508
	<i>ROA</i>	-0.028 **	0.013	-2.180	512
	<i>SALES₋₁₀-EMPLOYEE RATIO</i>	22.300	15.350	1.450	512
	<i>LN_ASSET</i>	0.432 **	0.1978	2.19	512
3 windows((t-1)-(t+2))	<i>TFP</i>	-0.105	0.094	-1.110	440
	<i>ROA</i>	-0.060 ***	0.016	-3.670	444
	<i>SALES₋₁₀-EMPLOYEE RATIO</i>	15.966	10.310	1.550	444
	<i>LN_ASSET</i>	0.596 ***	0.2201	2.71	444
4 windows((t-1)-(t+3))	<i>TFP</i>	-0.083	0.100	-0.830	386
	<i>ROA</i>	-0.049 ***	0.016	-2.990	388
	<i>SALES₋₁₀-EMPLOYEE RATIO</i>	21.100	14.429	1.460	388
	<i>LN_ASSET</i>	0.663 ***	0.2331	2.84	388

Treatment effect		Coef.	Std. Err	t-value	Obs.
2 windows((t-1)-(t+1))	<i>TFP</i>	-0.055	0.062	-0.900	508
	<i>ROA</i>	0.034 ***	0.009	3.850	512
	<i>SALES₋₁₀-EMPLOYEE RATIO</i>	43.840 ***	10.811	4.060	512
	<i>LN_ASSET</i>	0.342 **	0.1393	2.45	512
3 windows((t-1)-(t+2))	<i>TFP</i>	-0.068	0.066	-1.030	440
	<i>ROA</i>	0.040 ***	0.012	3.440	444
	<i>SALES₋₁₀-EMPLOYEE RATIO</i>	30.673 ***	7.258	4.230	444
	<i>LN_ASSET</i>	0.264 *	0.1549	1.7	444
4 windows((t-1)-(t+3))	<i>TFP</i>	0.000	0.070	0.000	386
	<i>ROA</i>	0.035 ***	0.011	3.050	388
	<i>SALES₋₁₀-EMPLOYEE RATIO</i>	36.066 ***	10.098	3.570	388
	<i>LN_ASSET</i>	0.333 **	0.1631	2.04	388

Table 11: Independent VC or not DID effect (i.e., IPO w/ VC vs. non-IPO) & Treatment effect
Independent VC in syndicate

DID effect		Coef.	Std. Err	t-value	Obs.
2 windows((t-1)-(t+1))	<i>TFP</i>	-0.195	0.182	-1.070	178
	<i>ROA</i>	0.048	0.102	0.470	180
	<i>SALES₋₁₀-EMPLOYEE RATIO</i>	21.566	18.259	1.180	180
	<i>LN_ASSET</i>	0.408	0.3221	1.27	180
3 windows((t-1)-(t+2))	<i>TFP</i>	-0.111	0.200	-0.550	158
	<i>ROA</i>	-0.050 **	0.024	-2.110	160
	<i>SALES₋₁₀-EMPLOYEE RATIO</i>	9.315	12.689	0.730	160
	<i>LN_ASSET</i>	0.664	0.3401	1.95	160
4 windows((t-1)-(t+3))	<i>TFP</i>	-0.149	0.260	-0.570	116
	<i>ROA</i>	-0.034	0.022	-1.500	116
	<i>SALES₋₁₀-EMPLOYEE RATIO</i>	8.981	17.224	0.520	116
	<i>LN_ASSET</i>	0.408	0.4341	0.94	116

Treatment effect		Coef.	Std. Err	t-value	Obs.
2 windows((t-1)-(t+1))	<i>TFP</i>	-0.096	0.127	-0.760	178
	<i>ROA</i>	-0.060	0.072	-0.840	180
	<i>SALES₋₁₀-EMPLOYEE RATIO</i>	19.818	12.839	1.540	180
	<i>LN_ASSET</i>	0.043	0.2265	0.19	180
3 windows((t-1)-(t+2))	<i>TFP</i>	-0.158	0.140	-1.130	158
	<i>ROA</i>	0.036 **	0.017	2.180	160
	<i>SALES₋₁₀-EMPLOYEE RATIO</i>	23.441 ***	8.916	2.630	160
	<i>LN_ASSET</i>	0.109	0.2390	0.45	160
4 windows((t-1)-(t+3))	<i>TFP</i>	-0.074	0.182	-0.410	116
	<i>ROA</i>	0.034 **	0.016	2.190	116
	<i>SALES₋₁₀-EMPLOYEE RATIO</i>	28.083 **	12.074	2.330	116
	<i>LN_ASSET</i>	0.483	0.3043	1.59	116

Not Independent VC in syndicate

DID effect		Coef.	Std. Err	t-value	Obs.
2 windows((t-1)-(t+1))	<i>TFP</i>	-0.019	0.071	-0.270	616
	<i>ROA</i>	-0.022 *	0.014	-1.660	620
	<i>SALES₋₁₀-EMPLOYEE RATIO</i>	18.272	15.179	1.200	620
	<i>LN_ASSET</i>	0.405 **	0.1829	2.21	620
3 windows((t-1)-(t+2))	<i>TFP</i>	-0.052	0.073	-0.710	544
	<i>ROA</i>	-0.056 ***	0.019	-2.950	548
	<i>SALES₋₁₀-EMPLOYEE RATIO</i>	17.152	13.374	1.280	548
	<i>LN_ASSET</i>	0.488 **	0.2041	2.39	548
4 windows((t-1)-(t+3))	<i>TFP</i>	-0.072	0.076	-0.950	460
	<i>ROA</i>	-0.051 ***	0.018	-2.750	464
	<i>SALES₋₁₀-EMPLOYEE RATIO</i>	26.156 **	12.709	2.060	464
	<i>LN_ASSET</i>	0.655 ***	0.2197	2.98	464

Treatment effect		Coef.	Std. Err	t-value	Obs.
2 windows((t-1)-(t+1))	<i>TFP</i>	-0.054	0.050	-1.100	616
	<i>ROA</i>	0.037 ***	0.010	3.840	620
	<i>SALES₋₁₀-EMPLOYEE RATIO</i>	44.983 ***	10.699	4.200	620
	<i>LN_ASSET</i>	0.337 ***	0.1289	2.61	620
3 windows((t-1)-(t+2))	<i>TFP</i>	-0.039	0.052	-0.760	544
	<i>ROA</i>	0.037 ***	0.013	2.720	548
	<i>SALES₋₁₀-EMPLOYEE RATIO</i>	36.171 ***	9.439	3.830	548
	<i>LN_ASSET</i>	0.319 **	0.1440	2.22	548
4 windows((t-1)-(t+3))	<i>TFP</i>	0.006	0.053	0.100	460
	<i>ROA</i>	0.034 ***	0.013	2.670	464
	<i>SALES₋₁₀-EMPLOYEE RATIO</i>	27.052 ***	8.909	3.040	464
	<i>LN_ASSET</i>	0.180	0.1541	1.17	464

Table 12: Foreign VC or not DID effect (i.e., IPO w/ VC vs. non-IPO) & Treatment effect
Foreign VC in syndicate

DID effect		Coef.	Std. Err	t-value	Obs.
2 windows((t-1)-(t+1))	<i>TFP</i>	0.292	0.397	0.740	36
	<i>ROA</i>	0.049	0.036	1.350	36
	<i>SALES-10-EMPLOYEE RATIO</i>	16.804	41.751	0.400	36
	<i>LN_ASSET</i>	0.558	1.1256	0.5	36
3 windows((t-1)-(t+2))	<i>TFP</i>	0.308	0.386	0.800	36
	<i>ROA</i>	-0.015	0.032	-0.460	36
	<i>SALES-10-EMPLOYEE RATIO</i>	17.908	38.416	0.470	36
	<i>LN_ASSET</i>	0.826	0.9880	0.84	36
4 windows((t-1)-(t+3))	<i>TFP</i>	0.339	0.429	0.790	10
	<i>ROA</i>	-0.086	0.264	-0.330	12
	<i>SALES-10-EMPLOYEE RATIO</i>	54.848	43.784	1.250	12
	<i>LN_ASSET</i>	0.653	1.3761	0.47	12

Treatment effect		Coef.	Std. Err	t-value	Obs.
2 windows((t-1)-(t+1))	<i>TFP</i>	-0.365	0.281	-1.300	36
	<i>ROA</i>	0.003	0.026	0.130	36
	<i>SALES-10-EMPLOYEE RATIO</i>	31.551	29.522	1.070	36
	<i>LN_ASSET</i>	0.513	0.7959	0.64	36
3 windows((t-1)-(t+2))	<i>TFP</i>	-0.455	0.273	-1.670	36
	<i>ROA</i>	0.005	0.023	0.240	36
	<i>SALES-10-EMPLOYEE RATIO</i>	31.482	27.164	1.160	36
	<i>LN_ASSET</i>	0.656	0.6986	0.94	26
3 windows((t-1)-(t+3))	<i>TFP</i>	-0.331	0.286	-1.160	10
	<i>ROA</i>	-0.061	0.186	-0.330	12
	<i>SALES-10-EMPLOYEE RATIO</i>	34.758	30.960	1.120	12
	<i>LN_ASSET</i>	1.327	0.9730	1.36	12

Not Foreign VC in syndicate

DID effect		Coef.	Std. Err	t-value	Obs.
2 windows((t-1)-(t+1))	<i>TFP</i>	-0.073	0.069	-1.060	758
	<i>ROA</i>	-0.009	0.026	-0.340	764
	<i>SALES-10-EMPLOYEE RATIO</i>	19.168	12.927	1.480	764
	<i>LN_ASSET</i>	0.399 **	0.1588	2.51	764
3 windows((t-1)-(t+2))	<i>TFP</i>	-0.085	0.074	-1.150	666
	<i>ROA</i>	-0.057 ***	0.016	-3.470	672
	<i>SALES-10-EMPLOYEE RATIO</i>	15.295	11.168	1.370	672
	<i>LN_ASSET</i>	0.511 ***	0.1768	2.89	672
4 windows((t-1)-(t+3))	<i>TFP</i>	-0.095	0.081	-1.180	566
	<i>ROA</i>	-0.047 ***	0.015	-3.140	568
	<i>SALES-10-EMPLOYEE RATIO</i>	22.038 **	10.914	2.020	568
	<i>LN_ASSET</i>	0.605 ***	0.1973	3.07	568

Treatment effect		Coef.	Std. Err	t-value	Obs.
2 windows((t-1)-(t+1))	<i>TFP</i>	-0.050	0.048	-1.030	758
	<i>ROA</i>	0.015	0.019	0.820	764
	<i>SALES-10-EMPLOYEE RATIO</i>	39.682 ***	9.105	4.360	764
	<i>LN_ASSET</i>	0.259 **	0.1118	2.32	764
3 windows((t-1)-(t+2))	<i>TFP</i>	-0.046	0.052	-0.880	666
	<i>ROA</i>	0.038 ***	0.012	3.300	672
	<i>SALES-10-EMPLOYEE RATIO</i>	33.384 ***	7.873	4.240	672
	<i>LN_ASSET</i>	0.251 **	0.1247	2.01	672
4 windows((t-1)-(t+3))	<i>TFP</i>	-0.004	0.057	-0.070	566
	<i>ROA</i>	0.036 ***	0.010	3.510	568
	<i>SALES-10-EMPLOYEE RATIO</i>	27.105 ***	7.649	3.540	568
	<i>LN_ASSET</i>	0.217	0.1383	1.57	568