Continual equipment investment is the only option for all global semiconductor manufacturers?

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Abstract
In this paper, we focus on the semiconductor manufacturing industry and investigate whether the firms’ sales growth and their Return on Equity (ROE) have an asymmetric nonlinear relationship in different debt ratio regimes by using financial statement-based data and employing a more powerful panel threshold regression model developed by Hansen (1999) over the 1999 first quarter (1Q) to 2005 third quarter (3Q) period. The result shows that there exists a double threshold effect and that the threshold values of debt ratio are \( \hat{\gamma}_1 = 0.4935 \) and \( \hat{\gamma}_2 = 0.6847 \). When the debt ratio is below 68.47%, ROE will be increased by continual equipment investment. However, when the debt ratio is above 68.47%, ROE will be decreased by continual equipment investment. These findings could be valuable for both the market investors to search their target of investment and corporation managers who can utilize them to adjust their production strategy and investment decision for increasing their business performance.

Keywords: Panel-unit root test; Panel threshold regression model;

1. INTRODUCTION

In the past few decades, the Information Technology industry has experienced the most rapid growth among all industries worldwide, with semiconductor manufacturers having played an arguably crucial role in the supply chain.

In the field of semiconductor wafer manufacturing alone, the innovative technologies of the manufacturing process and the enlargement of wafer size are two main motive powers that contributed to the industry’s continued growth. The semiconductor manufacturing industry is a typical oligopolistic industry that has capital and technical thresholds. In order to maintain the entry barrier and decrease production costs, semiconductor manufacturers often propose new investment plans, among which is enlarging the wafer size. However, firms need considerable funds to support these plans. In practice, limited internal capital compels financial managers to adopt external financing for fund-raising. These investment plans, which are intended to fill enormous funds demands, must gain the unanimous approval of both the shareholders and debt-holders. Furthermore, financial managers must propose powerful financial prospects to acquire their accommodators’ financial assistance.

In this paper, we collect 23 listed multi-national semiconductor manufacturers and focus on the firms’ output aspect. We used the sales revenue data contained in their financial report to measure the firms’ output level. We attempt to utilize the firms’ sales growth rate as the proxy for the variation of a semiconductor manufacturer’s products. However, the firm’s sales revenue is the product of price by quantity thus, factors that influence increase in a firm’s sales revenue is either a raise in price or an

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1 The building cost of 8-inch wafer fab and 12-inch wafer fab needs $1 billion and $3 billions, respectively.
increase in production. According to Moore’s Law, the complexity and efficiency of IC chip will increase to twice its current capacity while its price will decrease by half every 18 months. Furthermore, Gordon (2000) points out that the decreasing rate of price change in computer hardware (including peripherals) were at an average rate of -14.7 percent between 1987 and 1995 and -31.2 percent between 1996 and 1999. In addition, Oliner et al. (2003) show that the relative inflation rate in semiconductor sector were at an average rate of -28.9 percent between 1974 and 1990, -21.8 percent between 1991 and 1995 and -44.7 percent between 1996 and 2001. Hence, the price of computer hardware appeared to have a tendency of accelerating depreciation after 1995. To be brief, the price of semiconductor may rise temporarily in the short run, but it will drop in the long run.

Consequently, increasing production is the only way firms to raise sales revenue drastically. For this reason, the firm’s production strategy is either to enlarge wafer size or accelerate innovative technologies of manufacturing process. However, both strategies require considerable funds to be carried out. We consider both strategies as continual equipment investment. As stated earlier, due to the limited internal fund supply, it becomes necessary for firms to seek external financing. Figure 1 illustrates the growth rate of the semiconductor industry in total sales; from here we can see that there is no question as to whether there have been striking variations in total demand, many of which have been unanticipated. Since the semiconductor manufacturers face an extremely fluctuating market demand and because the characteristics of higher capital threshold exist in the semiconductor manufacturing industry, they will require external financing to meet the demands for huge funds. The question then becomes whether there exists other options for semiconductor manufacturers or is it limited to continual equipment investment.

2 The relative inflation rate is output price inflation in Semiconductor sector minus that in the “other final-output” sector (for further details, see Oliner et al., 2003).

Modigliani and Miller (1958, 1963) published seminal papers on the capital structure, weighted average cost of capital, and corporate valuation. The major difference between the assumptions of these two seminal papers is that the former assumed no taxes and the latter considered corporate income tax deductibility of interest (tax shields effect). Miller (1977) modified this assumption by introducing personal taxes as well as corporate taxes into the gain-to-leverage model. Jensen and Meckling (1976) utilized agency costs to discuss the conflict between principals and agents. They suggested that there is an optimum ratio of debt to equity that will be chosen because it minimizes total agency costs. Consequently, the optimal capital structure can result in a trade-off between the tax shields benefit of debt and agency costs. Castanias (1983) argued that if managers increase financial leverage then the possibility of bankruptcy also increases and the probability of bankruptcy has a negative effect on the firm’s value. Therefore, the optimal ratio of debt to equity is determined by a trade-off between interest tax shields and bankruptcy costs. Leland (1994) and Leland and Toft (1996) modeled corporate valuation by assuming that the present value of business disruption costs and the present value of lost interest tax shields are affected by different capital structures of firm. The result is similar, an optimal capital structure as a trade-off between the tax deductibility of interest expenses and business disruption costs.

By combining the above arguments with the trade-off model, we predict that a few semiconductor manufacturers who will face an extremely fluctuating market demand will naturally need large external financing. In the future, if these firms encounter economic recession, they may incur huge losses have higher debt ratios. However, in general, most debt-holders will be reluctant to lend money to debt-heavy firms; hence, there will be an increase in financing difficulty and capital cost of firms who have higher debt ratios.

Financial ratios are widely employed as explanatory variables in the field of empirical finance to explain the
investment decisions of firms (Barnes (1987); Beatty (1993); and Cleary (1999) among others). Additionally, numerous empirical researchers have utilized ROE as an indicator or proxy variable to evaluate corporate business performance or firm value (e.g. see Mramor and Mramor Kosta, 1997; Pahor and Mramor, 2001; Easton, 2004; and Nieh et al., 2008 among others). In this paper, we utilize some financial ratios to explain the business performance (ROE) of firms. These ratios include sales growth rate, debt ratio and total asset turnover rate.

In recent years, Hansen’s panel threshold regression model has been widely employed in the field of empirical finance to explore corporate investment and financing constraints (e.g. see Chen, 2003; Nieh and You, 2005; Shen and Wang, 2005; Yeh et al., 2007; and Nieh et al., 2008 among others). The purpose of this paper is to focus on the semiconductor manufacturing industry and investigate whether the sales growth and ROE of firms in the industry have asymmetric nonlinear relationship in different debt ratio regimes. We presume that the semiconductor manufacturing industry has different debt ratio thresholds that will allow for the division of all of the firms into groups, making the firms’ sales growth rate and their ROE exist asymmetric nonlinear relationship in the different debt ratio regimes. When the sales growth rate and ROE have significant positive relationship in one regime, this implies that continual equipment investment could improve a firm’s business performance. On the contrary, when they have significant negative relationship in another regime, this implies that continual equipment investment will damage a firm’s business performance. Furthermore, when the sales growth rate and ROE have an insignificant relationship in one regime, this implies that continual equipment investment might not guarantee the improvement of a firm’s business performance.

The remainder of this paper is structured as follows. Section II presents a description of the data we use. Section III describes the methodology we employ and discusses the empirical results. It also points out some policy implications. Finally, Section IV concludes the paper.

II. DATA

In this paper, we collect data from 1999 Q1 to 2005 Q3 for 23 listed multi-national firms operating primarily in the semiconductor manufacturing industry. We collect these data from COMPSTAT and the Taiwan Economic Journal (henceforth TEJ) database. We select our sample data from certain semiconductor market research institutions (including iSuppli Corp., IC insights Corp. and Gartner Dataquest Corp.), which provide the rankings of the top 25 semiconductor market shares worldwide in the DRAM industry (the top 12), and the Foundry industry (the top 7). This information is summarized in Table 1. We exclude Japanese companies that started trading publicly after 1999 and IC design houses (Fabless). Our empirical analysis utilize three financial ratios to explain a firm’s business performance (ROE), which are sales growth rate, debt ratio, and total asset turnover rate. Table 2 reports the summary statistics of the four financial ratios. As shown in Table 2, the Jarque-Bera test results show that the

3 The COMPSTAT database provides quarterly data for NYSE- and NASDAQ-listed companies but only provides annual data for Korea and Taiwan listed companies. The TEJ database presents quarterly data for Korea and Taiwan listed companies.

4 We exclude Japanese semiconductor firms largely because most (NEC, Hitachi, Toshiba, Mitsubishi, among others) have split off their semiconductor manufacturing operations and jointly set up new companies with other firms in recent years. These new companies have been listed since 1999. Furthermore, these Japanese firms tend to be made up of financial groups making it difficult to distinguish semiconductor divisions from others.

5 Freescale Semiconductor and Renesas Technology, which were both listed after 1999, are good examples. Motorola split off its semiconductor manufacturing division and in 2004 set up a new company, Freescale. Mitsubishi and Hitachi split off their semiconductor manufacturing divisions and in 2002 merged them to form a new separate legal entity, Renesas.

6 Fab and Fabless companies are extremely different in terms of the structure of their assets; while Fab companies belong to the manufacturing industry, Fabless companies engage in the design, development, and marketing of their chips and adopt outsourcing strategies to have these chips manufactured.
distribution of all financial ratios approximate non-normal.

III. METHODOLOGY AND EMPIRICAL RESULTS

A. Hansen Panel Threshold Autoregressive Model

3.1 Panel Unit Root models

According to the trade-off theory of capital structure, the optimal ratio of debt to equity is determined by a trade-off between the interest tax shields and leverage related costs. Therefore, this paper presumes that there is a feasible debt ratio regime in the semiconductor manufacturing industry. We employ the panel threshold regression model developed by Hansen (1999) to estimate this regime and explore the asymmetric nonlinear relationship between the firms’ sales growth and their ROE in different debt ratio regimes. The results should help corporation managers adjust the firms’ capital structure or change their production strategies. From a methodological point of view, if we apply Hansen’s panel threshold regression model, we should first use panel unit root tests to verify that all financial ratio series are stationary series in order to avoid the so-called spurious regression. Thus, we first apply several panel-based unit root tests and examine the null of a unit root in four financial ratios of our sample firms. To avoid small-sample biases, we calculate the critical values based on Monte Carlo simulations, performing 10,000 times for each test. These results are given in Table 3. We find that both the Levin-Lin-Chu (Levin et al., 2002) and Im-Pesaran-Shin (Im et al., 2003) panel-based unit root tests reject the null of non-stationarity for these ratios and indicate that they are stationary.

3.2 Threshold Autoregressive model

The results of the panel unit root test of each variable confirm that all series are stationary. We thus perform Hansen’s panel threshold regression model and hypothesize that the firms’ sales growth rate and their ROE exist asymmetric nonlinear relationship in different debt ratio levels. Since Tong (1978) proposed the threshold autoregressive model, the utilization of this nonlinear time series model has been widely used in economic and financial research. In estimating the threshold autoregressive model, a test to determine whether or not there are threshold effects must first be conducted. If the null cannot be rejected, then the threshold effect does not exist. To avoid the “Davies problem”, Hansen (1999) recommended a bootstrap method to simulate the asymptotic distribution of the likelihood ratio test and compute the critical values in order to test the significance of threshold effect. Furthermore, when the null cannot be accepted, which means the threshold effect does exist, Chan (1993) demonstrated strong evidence that the OLS estimation of the threshold is super-consistent and that it can derive the asymptotic distribution. Hansen (1999) proposed a simulation likelihood ratio test to derive the asymptotic distribution when testing for a threshold and he applied the two-stage OLS method to estimate the panel threshold model. The steps are as follows. First, for any given threshold value , the sum of square errors (SSE) is computed separately.

Second, the threshold estimator is estimated by minimizing SSE. In this paper, we employ this two-stage OLS method to estimate and then utilize to estimate the coefficients of every regime then analyze the results.

3.3 The Threshold Regression Model Construction

A Single Threshold Model can be set up as follows.

\[
y_n = \mu + \Theta X_n + \beta_1 s_n I(d_n \leq \gamma) + \beta_2 s_n I(d_n > \gamma) + \epsilon_n \quad (1)
\]

7 Granger and Newbold (1974) argued that spurious regression is the estimation of the relationship among non-stationary series is without difficulty of getting higher R2 and t statistics.

8 “Davies’ Problem” exists when the testing statistics follow a non-standard distribution because of the presence of nuisance parameters (Davies, 1977, 1987).
Taking the difference between (2) and (3)

\[ y_i^* = \Theta X_i^* + \beta s_d(\gamma) + \varepsilon_i^* \]  

where  

\[ y_i = \mu_i + \Theta X_{i,\gamma} + \beta s_d(\gamma) \]  

\[ s_d(\gamma) = \begin{cases} 1 & (d_i \leq \gamma) \\ 0 & (d_i > \gamma) \end{cases} \]

\[ y_i = \mu_i + \omega k_i (\gamma) + \varepsilon_i \]

where  

\[ \beta = (\beta_1, \beta_2) \]  

\[ \omega = (\Theta, \beta) \]  

\[ k_i = (d_i, s_d(\gamma)) \]

The observations are separated into two groups dependent on whether the threshold variable \( d_i \) is larger or smaller than threshold value \( \gamma \). We can acquire the different regression slope estimators, \( \beta_1 \) and \( \beta_2 \), from two different regimes and apply given \( y_i \) and \( s_d(\gamma) \) to estimate these parameters that include \( \gamma, \beta, \Theta \) and \( \sigma^2 \).

### 3.4 Estimation

\[ \hat{y}_i = \hat{u}_i + \Theta \hat{X}_i + \beta \hat{s}_d(\gamma) + \hat{\varepsilon}_i \]  

where  

\[ \hat{y}_i = \frac{1}{T} \sum_{t=1}^{T} y_{it}, \quad \hat{X}_i = \frac{1}{T} \sum_{t=1}^{T} X_{it} \]

and  

\[ \hat{\varepsilon}_i = \frac{1}{T} \sum_{t=1}^{T} \varepsilon_{it} \]

The equation (5) represents the major estimation model of threshold effect. The panel threshold regression model use two-stage OLS method. On the first stage, for any given \( \gamma \), the slope coefficient \( \beta \) can be estimated by OLS. That is,  

\[ \hat{\beta}(\gamma) = \left( S_D^*(\gamma) S_D^*(\gamma) \right)^{-1} S_D^*(\gamma) Y^* \]  

The vector of regression residuals is  

\[ \hat{\Phi}^*(\gamma) = Y^* - \Theta X^* - \hat{\beta}(\gamma) S_D^*(\gamma) \]  

and the sum of squared errors, SSE is  

\[ \text{SSE}(\gamma) = \hat{\Phi}(\gamma) \hat{\Phi}(\gamma) = (Y^* - \hat{\beta}(\gamma) S_D^*(\gamma))^2 \]  

Chan (1993) and Hansen (1999) recommend estimation of \( \gamma \) by least squares method and achieve by minimization of the concentrated sum of squared errors (9). Hence the least squares estimators of \( \gamma \) is  

\[ \hat{\gamma} = \arg \min_{\gamma} \text{SSE}(\gamma) \]

Once \( \hat{\gamma} \) is obtained, the slope coefficient estimation is  

\[ \hat{\beta} = \hat{\beta}(\hat{\gamma}) \]

The residual vector is  

\[ \hat{\Phi}^* = \hat{\Phi}^*(\hat{\gamma}) \]

and the estimator of residual variance is
\[ \hat{\sigma}^2 = \hat{\sigma}^2(\hat{\gamma}) = \frac{1}{n(T - 1)} \hat{\Phi}(\hat{\gamma}) \hat{\Phi}(\hat{\gamma}) = \frac{1}{n(T - 1)} \text{SSE}(\hat{\gamma}) \] (10)

where \( n \) represents the number of sample, \( T \) represents the periods of sample.

3.5 Testing for a Threshold

This paper hypothesizes that there exists the threshold effect of the debt ratio between the firm’s sales growth and ROE. It is important to detect whether the threshold effect is statistically significant. The null and alternative hypothesis can be represented as follows:

\[
\begin{align*}
H_0 &: \beta_1 = \beta_2 \\
H_1 &: \beta_1 \neq \beta_2
\end{align*}
\]

When the null is rejected, the coefficient \( \beta_1 \neq \beta_2 \) the threshold effect of the debt ratio exists between the firms’ sales growth and their ROE. Alternatively, when the null is accepted, the coefficient \( \beta_1 = \beta_2 \) the threshold effect of the debt ratio doesn’t exist.

Under the null hypothesis of no threshold, the model is

\[ y_{it} = \mu_t + \Theta X_{it} + \beta_1 s_d_{it}(\gamma) + \varepsilon_{it} \] (11)

After the fixed-effect transformation is made, we can obtain

\[ G^* = \Theta X^* + \beta_1 S^*_{D}(\gamma) + \Phi^*_{it} \] (12)

By OLS method estimate (12) that can yield the estimator \( \hat{\beta}_1 \), residuals \( \hat{\Phi}^*_{it} \) and the sum of square errors \( \text{SSE}_0 = \hat{\Phi}^* \hat{\Phi}^* \).

Hansen (1996) suggests that we utilize the F test approach to determine the existence of threshold effect, and use the sup-Wald statistic to test the null.

\[
F = \sup F(\gamma) = \frac{(\text{SSE}_0 - \text{SSE}_1(\gamma))/1}{\text{SSE}_1(\gamma)/n(T - 1)} \] (13)

\[
F(\gamma) = \frac{(\text{SSE}_0 - \text{SSE}_1(\gamma))/1}{\text{SSE}_1(\gamma)/n(T - 1)} \] (14)

However, under the null, the pre-specified threshold \( \gamma \) dose not exist, thus, the nuisance parameter exists. Therefore, in order to avoid the “Davies problem”. Hansen (1996) recommended a bootstrap method to simulate the asymptotic distribution of the likelihood ratio test and compute the p-values in order to test the significance of threshold effect. Treat the regressors \( k \), and the threshold variable \( d_{it} \), as given, holding their values fixed in repeated bootstrap samples. Take the regression residuals \( \hat{\Phi}^*_{it} \), and group them by individual: \( \hat{\Phi}^*_{it} = (\hat{\varepsilon}_{i1}, \hat{\varepsilon}_{i2}, \ldots, \hat{\varepsilon}_{ir}) \). Treat the samples \( \{\hat{\Phi}^*_{i1}, \hat{\Phi}^*_{i2}, \ldots, \hat{\Phi}^*_{ir}\} \) as the empirical distribution to be used for bootstrap procedure. Draw a sample of size \( n \) from the empirical distribution and utilize these errors to create a bootstrap sample under the null. This sample is used to estimate the model under the null (12) and alternative (4), and the bootstrap values of the likelihood ratio statistic \( F(\gamma) \) (14) are calculated. This procedure is repeated a large number of times, and calculate the percentage of draws for which the simulated statistic exceeds the actual. This is the bootstrap estimate of the asymptotic p-value for \( F(\gamma) \) under the null.

\[ P = P(\tilde{F}(\gamma) > F(\gamma)|\phi) \] (15)

where \( \phi \) is the conditional mean of \( \tilde{F}(\gamma) > F(\gamma) \)

The null of no threshold effect is rejected if the p-value is smaller than the desired critical value.

3.6 Asymptotic Distribution of Threshold Estimate

Chan (1993) and Hansen (1999) demonstrated that when the null of no threshold effect is rejected, \( \hat{\gamma} \) is consistent for \( \gamma_0 \), and that the asymptotic distribution is highly nonstandard. Hansen (1999) suggested that apply simulation likelihood ratio test to develop the confidence interval and asymptotic distribution of testing for \( \gamma \). The null and alternative hypothesis can be represented as follows:

\[
\begin{align*}
H_0 &: \gamma = \gamma_0 \\
H_1 &: \gamma \neq \gamma_0
\end{align*}
\]

The likelihood ratio test is represented as follows:

\[
\text{LR}_1(\gamma) = \frac{\text{SSE}_1(\gamma) - \text{SSE}_1(\hat{\gamma})}{\hat{\sigma}^2} \] (16)

Hansen (1999) pointed out that when is too large and
the p-value exceeds the confidence interval, the null is rejected. In addition, Hansen (1999) indicated that under some specific assumptions⁹ and \( H_0 : \gamma = \gamma_0 \),

\[
LR_1(\gamma) \xrightarrow{n \to \infty} d \psi
\]

(17)

where \( \psi \) is a random variable with distribution function \( P(\psi \leq x) = \left(1 - \exp\left(-\frac{x}{\sqrt{2}}\right)\right)^2 \)

(18)

The asymptotic p-value can be estimated under the likelihood ratio. According to the proof of Hansen (1999), the distribution function (18) has the inverse function \( c(\alpha) = -2\log\left(1 - \sqrt{1 - \alpha}\right) \)

(19)

We can utilize (19) to calculate the critical values. For a given asymptotic level \( \alpha \), the null hypothesis \( \gamma = \gamma_0 \) is rejected if \( LR_1(\gamma) \) exceeds \( c(\alpha) \).

3.7 Multiple Thresholds Model

If there is a double threshold, the panel threshold model can be modified as:

\[
y_{it} = \mu + \Theta X_{it} + \beta_1 s_{it} I(d_{it} \leq \gamma_1) + \beta_2 s_{it} I(\gamma_1 < d_{it} \leq \gamma_2) + \beta_3 s_{it} I(d_{it} > \gamma_2) + \epsilon_{it}
\]

(20)

where threshold value \( \gamma_1 < \gamma_2 \). This method can be extended to multiple thresholds model \((\gamma_1, \gamma_2, \gamma_3, \cdots, \gamma_n)\).

B. Empirical Results

In this paper, we use the observed balanced panel data and choose four financial ratios then conduct these variables as dependent variable (ROE), threshold variable (Debt ratio), or control variables (the others). We employ the panel threshold regression model to detect whether they have a rational debt ratio level, which might result in threshold effect and asymmetrical relationships between the sales growth rate and firm’s ROE. If the threshold estimate of \( \gamma \) is demonstrated in addition to a significant relationship between the sales growth rate and firm’s ROE, then corporation managers should increase their firm’s business performance by adjusting the firm’s production or investment strategy. We utilize the multiple threshold regression model.

\[
ROE_{it} = \mu_i + \theta_0 d_{it} + \theta_1 d_{it} + \beta_0 s_{it} I(d_{it} \leq \gamma_1) + \beta_1 s_{it} I(\gamma_1 < d_{it} \leq \gamma_2) + \beta_2 s_{it} I(d_{it} > \gamma_2) + \epsilon_{it}
\]

\[
= \mu_i + \sum_{i=0}^{2} \hat{\theta}_i d_{it} + \sum_{i=0}^{2} \hat{\beta}_i s_{it} I(d_{it} \leq \gamma_i) + \epsilon_{it}
\]

(21)

We perform the bootstrap method to attain the approximation of F statistics and determine the critical value and p-value. Table 4 presents the empirical results for single, double and triple threshold effects. After performing bootstrap procedure 10,000 times for each of the panel threshold test, we find that the test for a double threshold is significant with a bootstrap p-value of 0.0497 and both tests for single and triple threshold are insignificant with bootstrap p-value of 0.1173 and 0.2664, respectively. Therefore, we conclude that there is a double threshold effect in this empirical model.

Table 5 shows the double threshold estimate values \((\hat{\gamma}_1, \hat{\gamma}_2)\) are \(\hat{\gamma}_1 = 0.4935\) and \(\hat{\gamma}_2 = 0.6847\). When there is a double threshold effect of the debt ratios on firms’ ROE, all of the observations can be split into three regimes depending on whether the threshold variable \( d_{it} \) is smaller or larger than the double threshold estimate values \((\hat{\gamma}_1, \hat{\gamma}_2)\). The regimes are distinguished by differing estimate values \(\hat{\gamma}_1\) and \(\hat{\gamma}_2\).

Figures 2 and 3 show the confidence interval of the estimators and the first and second threshold parameters’ estimators in double threshold model, respectively.

Table 6 reports the coefficient estimate \(\Theta\) of the regression parameters \(X_{it}\), conventional OLS standard errors, and White-corrected standard errors for three different regimes.¹⁰ The estimated model from our empirical result can be expressed as follows:

\[
ROE_{it} = \mu_i + 0.0361 d_{it-1} + 0.6890 s_{it} + 0.0356 s_{it} I(d_{it} \leq 0.4935) + 0.2104 s_{it} I(0.4935 < d_{it} \leq 0.6847) - 0.3105 s_{it} I(0.6847 < d_{it}) + \epsilon_{it}
\]

\[
\hat{\gamma}_1\ and\ \hat{\gamma}_2\ divides\ the\ observations\ into\ three\ regimes.\ In\ \textit{endnotes}⁹\ Refer\ to\ Hansen\ (1999)\ Appendix:\ Assumptions\ 1-8.

¹⁰ After we remove the control variable \( d_{it} \) from the regression and then employ this model, we find the empirical results are similar. However, all of the results are available from the author upon request.
the first regime, where the debt ratio is below 49.35%, the estimate of coefficient $\hat{\beta}_1 = 0.0560$ is significant, which implies ROE will increase by 5.6% with the 1% increase of the sales growth rate. In the second regime, where the debt ratio is between 49.35% and 68.47%, the estimate of coefficient $\hat{\beta}_2 = 0.2104$ is significant, which implies that when the sales growth rate is increased by 1%, ROE will be increased by 21.04%. In the third regime, where the debt ratio is above 68.47%, the estimate of coefficient $\hat{\beta}_3 = -0.3106$ is significant, both statistically and economically, which means that ROE will be decreased by 31.06% with the 1% increase of the sales growth rate.

According to Table 6, the estimates $\hat{\beta}_1 = 0.0560$, $\hat{\beta}_2 = 0.2104$ and $\hat{\beta}_3 = -0.3106$ are all highly significant at the 1% level under the consideration of both homogenous and heterogeneous standard errors. In other words, when the debt ratio is higher than 68.47%, ROE will decrease with the increase of sales growth rate. On the other hand, when the debt ratio is lower than 68.47%, ROE will increase with the increase of sales growth rate. Concerning other control variables, the empirical results show that the total asset turnover rate is statistically significant and the debt ratio is insignificant. The point estimate reveals that firms’ ROE is significantly positive related to total asset turnover rate. In practice, the total asset turnover rate is frequently utilized to measure the changes in the business strategies of firms. Therefore, the higher the total asset turnover rate is, the better the firm’s business strategy, and the greater the firm’s ROE is. Furthermore, the point estimate shows that the firms’ ROE is insignificantly related to debt ratio. The result implies that the firms’ debt ratio and their ROE have no direct relation. Since the semiconductor manufacturing industry is such a competitive and capital-intensive industry that firms must possess excellent financing ability to meet enormous funds demands, a firm's financial capacity will therefore be constrained when a firm's debt ratio is too high. In the future, when these firms encounter economic recession, they may incur huge losses and gain higher debt ratios. This may cause the interruption of a number of investment projects.

The empirical results demonstrate the vision that if firms use financial leverage excessively, this may damage the firms’ ROE by increasing equipment investment (increase sales growth). In contrast, the moderate use of financial leverage can improve the firms’ ROE noticeably by increasing equipment investment (increase sales growth). Moreover, in the semiconductor manufacturing industry, a conservative financing strategy can improve the firms’ ROE slightly by increasing equipment investment (increase sales growth). The empirical results are consistent with the trade-off theory of financial leverage; firms that adopt conservative financing strategy can improve their ROE slightly while those with a moderate degree of debt ratio can improve their ROE noticeably. However, owing to the excessive use of financial leverage, firms will increase the possibility of financial distress and damage their ROE. These findings are valuable for both the market investors to search their target of investment and for corporation managers who can utilize them to adjust their production strategy and investment decision to increase their business performance.

IV. CONCLUSIONS

In this study, we focus on the semiconductor manufacturing industry and find a more rational debt ratio in this industry by performing Hansen’s panel threshold regression model. We employ the nonlinear regression model with endogenous threshold instead of traditional linear regression model. It is worth noting that there exists a double threshold effect and that the threshold values of the debt ratio are $\gamma_1 = 0.4935$ and $\gamma_2 = 0.6847$. When the debt ratio is below 68.47%, ROE will increase with the increase of sales growth. On the other hand, when the debt ratio is below 49.35% and between 49.35% and 68.47%, ROE in the former will increase noticeably with the
increase of sales growth, while the ROE in the latter will increase slightly with the increase of sales growth. In contrast, when the debt ratio is above 68.47%, ROE will decrease with the increase of sales growth.

The empirical results verify that the conservative and moderate degree of the debt ratio can guarantee the improvement of the semiconductor manufacturers’ ROE by continual equipment investment. On the contrary, excessive financial leverage will damage the firm’s ROE by continual equipment investment. Therefore, we show that the equipment-race is not the only channel in the global semiconductor manufacturing industry. When the firm’s debt ratio is above 68.47%, they should employ the ‘fab-lite’ style or partial outsourcing strategies to diminish their financial leverage. On the contrary, when the firm’s debt ratio is lower than 68.47%, they should adopt more aggressive financing strategies to increase their financial leverage. These findings are valuable for both the market investors to search their target of investment and corporation managers who can utilize them to adjust their production strategy and investment decision to increase their business performance.

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Table 1. Market Share Rankings of the Worldwide Top Semiconductor Manufacturers

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<td>3</td>
<td>Texas Instruments</td>
<td>Hynix</td>
<td>IBM</td>
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<td>SANYO a</td>
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Data Source: ISuppli, Dataquest, IC Insights, Dataquest, Dataquest

Notes: a, b, and c denote Japanese companies that started trading publicly after 1999 and that belong to the IC design house (Fabless).
Table 2. Summary Statistics of Candidate Financial Ratios

<table>
<thead>
<tr>
<th>Financial Ratio</th>
<th>Mean</th>
<th>Std</th>
<th>Max.</th>
<th>Min.</th>
<th>Skewness</th>
<th>Kurtosis</th>
<th>J-B</th>
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<tbody>
<tr>
<td>ROE</td>
<td>0.00451</td>
<td>0.05662</td>
<td>0.12224</td>
<td>-0.32396</td>
<td>-1.5390</td>
<td>8.6796</td>
<td>879.84***</td>
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<tr>
<td>Sales Growth Rate</td>
<td>0.02467</td>
<td>0.11176</td>
<td>0.45002</td>
<td>-0.31577</td>
<td>-0.1044</td>
<td>4.3234</td>
<td>37.841***</td>
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<tr>
<td>Debt Ratio</td>
<td>0.59298</td>
<td>0.15736</td>
<td>0.88111</td>
<td>0.14369</td>
<td>-0.3208</td>
<td>2.6793</td>
<td>17.470***</td>
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<tr>
<td>Total Asset Turnover  Rate</td>
<td>0.16688</td>
<td>0.14722</td>
<td>0.85535</td>
<td>0.02005</td>
<td>0.0447</td>
<td>13.130</td>
<td>2945.13***</td>
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</tbody>
</table>

Notes: Std denotes standard deviation, and J-B denotes the Jarque-Bera test for Normality. ***, ** and * indicate significance at the 0.01, 0.05 and 0.1 level, respectively.

Table 3. Panel-based Unit Root Test Results

<table>
<thead>
<tr>
<th>Financial Ratios</th>
<th>method</th>
<th>Statistics</th>
<th>P-value</th>
<th>Critical value 1%</th>
<th>5%</th>
<th>10%</th>
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</thead>
<tbody>
<tr>
<td>Return on Equity</td>
<td>Levin, Lin &amp; Chu</td>
<td>-11.63***</td>
<td>0.0027</td>
<td>-10.43</td>
<td>-9.666</td>
<td>-9.233</td>
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<td>IPS $\phi_t$</td>
<td>-5.691**</td>
<td>0.0614</td>
<td>-6.426</td>
<td>-5.777</td>
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<td>$\phi_{LM}$</td>
<td>6.495***</td>
<td>0.0854</td>
<td>7.839</td>
<td>6.866</td>
<td>6.342</td>
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<tr>
<td>Sales Growth Rate</td>
<td>Levin, Lin &amp; Chu</td>
<td>-13.96***</td>
<td>0.0001</td>
<td>-10.69</td>
<td>-9.944</td>
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<td>IPS $\phi_t$</td>
<td>-7.172***</td>
<td>0.0895</td>
<td>-10.234</td>
<td>-8.0762</td>
<td>-6.933</td>
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<td>$\phi_{LM}$</td>
<td>7.706***</td>
<td>0.0083</td>
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<td>5.815</td>
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<td>Debt Ratio</td>
<td>Levin, Lin &amp; Chu</td>
<td>-11.32***</td>
<td>0.0041</td>
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<td>IPS $\phi_t$</td>
<td>-9.259***</td>
<td>0.0848</td>
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<td>$\phi_{LM}$</td>
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<td>Total Asset Turnover  Rate</td>
<td>Levin, Lin &amp; Chu</td>
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<td>IPS $\phi_t$</td>
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<td>$\phi_{LM}$</td>
<td>8.589***</td>
<td>0.0000</td>
<td>3.258</td>
<td>0.641</td>
<td>-0.644</td>
</tr>
</tbody>
</table>

Notes: 1. ***, ** and * indicate significance at the 0.01, 0.05 and 0.1 level, respectively.
2. The critical values are calculated using Monte Carlo simulations with 10,000 times.