



Estimating the Effects of Large Shareholders Using a Geographic Instrument

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September 15, 2009

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Abstract

Large shareholders may play an important role for firm performance and policies, but identifying this empirically presents a challenge due to the endogeneity of ownership structures. We develop and test an empirical framework which allows us to separate selection from treatment effects of large shareholders. Individual blockholders tend to hold blocks in public firms located close to where they reside. Using this empirical observation, we develop an instrument – the density of wealthy individuals near a firm’s headquarters – for the presence of a large, non-managerial individual shareholder in a firm. These shareholders have a large impact on firms, controlling for selection effects.

1. Introduction

A public firm's shareholders have extensive legal control rights in the corporation, but in practice much of this control is delegated to managers. In companies with small, dispersed shareholders, owners may find it costly to coordinate and exercise monitoring and control, leaving management with considerable discretion. The resulting agency problems between owners and managers may be severe (e.g., Berle and Means (1932) and Jensen and Meckling (1976)).¹ By concentrating a block of shares in the hands of a single decision maker, large shareholders may play a beneficial role in remedying these agency problems and facilitating effective owner control (e.g., Grossman and Hart (1980) and Shleifer and Vishny (1986)). A large shareholder may restrain managers from extracting rents and thereby improve firm performance, benefitting all shareholders. Other theories predict that a block in a firm may be costly. Large shareholders may, e.g., reduce the liquidity of a firm's shares.

Testing these theories about the effects of large owners is challenging. Researchers in corporate finance have recognized for more than 20 years (at least since Demsetz and Lehn (1985)) the difficulty of causal inference in studies of ownership and performance. Suppose for example that blockholders systematically target firms with poor performance because of the potential performance increases due to improved monitoring, or that they select more liquid stocks because of the relatively lower cost of creating a block. Suppose also, as predicted by theory, that conditional on such selection, blockholder monitoring improves operating profitability and reduces the liquidity of the firm's shares. If this is the case, then OLS regressions of performance or stock liquidity on block presence in a firm confound selection effects and treatment effects of block ownership. In this paper, we develop and test a new instrumental variable framework which allows us to separate these effects for a large group of blockholders. This allows us to quantify the impact of large owners on several key aspects of firm behavior.

¹ There is a large literature on manager-shareholder conflicts. Demsetz (1983) argues that a manager may extract non-pecuniary benefits to the detriment of shareholders. Jensen (1986) and Jensen (1989) argue that managers expropriate small, dispersed shareholders by diverting corporate resources for empire building or perk consumption.

Our analysis focuses on *individual non-managerial blockholders* and excludes, e.g., blocks held by mutual funds and other institutions. This is motivated by the limitations of our identification strategy (i.e. what our instrument predicts), but is also appealing when testing predictions from theories of large shareholders. Non-management individuals who hold a large block of shares in a firm seem to naturally fit with the notion of having a principal present in the firm because they are generally less constrained than institutional investors from being active monitors of management. They do not face the regulatory constraints of many mutual funds and pension funds (e.g., Black (1990) and Roe (1990)), they are not afraid of losing money management business (e.g., Brickley, Lease and Smith (1988) and Davis and Kim (2007)), and they come with fewer agency problems of their own (e.g., Diamond (1984)).² Managers who are also blockholders have interests that may differ dramatically from that of other owners (see e.g. Barclay and Holderness (1989)). Thus, non-management individual blocks appear to fit well to theories of outside monitors.

Our instrument for the presence of a large shareholder in a firm is the number of high net worth individuals in the state where a firm is headquartered divided by the number of public firms headquartered in the state, which we refer to as *wealth density*.³ This variable is a valid instrument for the presence of blocks in local firms if it is correlated with the frequency of such blocks (relevancy condition) but impacts the dependent variables only through its effect on block ownership (exclusion restriction). Our instrument reflects economic forces important for the presence of a block in a firm. The relevancy condition is easiest to evaluate. First, blockholders exhibit a preference for firms headquartered near where they live, presumably because of lower monitoring costs or asymmetric information.⁴ The median distance between a non-managerial individual blockholder and the firm in which they have a block is 42.5

² While institutions are not generally active owners, some institutions are more active than others. There is evidence from the 1980s and early 1990s that some public pension funds, e.g., CalPERS, were active in corporate governance (e.g., Smith (1996) and Del Guercio and Hawkins (1999)). Also, Brav, Jiang, Partnoy and Thomas (2008) compare firms before and after an activist hedge fund accumulates a stake and find significant changes in performance and firm policies.

³ To the best of our knowledge, there is no database available that reliably captures U.S. wealth distribution on a finer scale than state level (such as Metropolitan Statistical Areas).

⁴ See Lerner (1995) for evidence on local bias among venture capitalists, Coval and Moskowitz (1999), (2001) for mutual funds, and Bailey, Kumar and Ng (2008) for retail investors.

miles. Second, personal wealth is likely a determinant of an individual's ability to invest a large stake in a public firm. Finally, there is significant cross-state variation in the wealth density instrument, possibly reflecting exogenously attractive areas or the fact that wealthy people prefer to live near other wealthy individuals because of social reasons or similar tastes (e.g., Glaeser (1998)). We find that our instrument predicts the presence of a block in a firm with considerable power. The economic magnitude of the estimated effect is large: moving from the 25th to the 75th percentile wealth density increases the predicted probability of a block in a local firm from 8.7 percent to 12.8 percent, i.e., it increases the probability by about 50 percent.

While there is no statistical test of the validity of the exclusion restriction in the presence of a single instrument, we provide several empirical observations and arguments to support its validity. First, our instrument comes from inheritance data, i.e., deceased wealthy individuals, so it reflects "old money", limiting the impact of local wealth-creating events such as initial public offerings of local firms, which may directly affect current firm performance and policies. Second, if wealth density was correlated with firm-specific variables measuring the demand for a block in a firm, we would expect it to also predict the presence of blocks other than non-managerial individual blocks. We find that wealth density does not predict the presence of blocks by managers or institutional investors. We return to a more detailed discussion of the validity of our instrument in Section 4.4.

Our empirical evidence is consistent with theories of large shareholders as monitors of management. We find that blocks are not randomly allocated to firms, but large shareholders systematically target firms based on where the benefits from additional monitoring are expected to be more significant, e.g., smaller and relatively poorly performing firms. Controlling for such block selection, we find that the presence of a large shareholder significantly reduces a firm's investments, reduces corporate cash holdings, increases payouts to shareholders, reduces total top-executive pay, and increases firm performance. This evidence is consistent with theories of large shareholders mitigating agency problems between managers and owners. Firms with blockholders also have significantly more outside directors on their boards. In addition to monitoring benefits we also find that block presence

comes with some costs. One such cost is less liquid publicly traded shares. This may reflect a smaller float as well as the presence of privately informed traders (i.e., the blockholders).

The paper contributes to existing research on large shareholders' impact on firms by explicitly addressing the endogeneity of the ownership structure. Our instrumental variables approach supports a causal interpretation of block impact on firm behavior. In his survey, Holderness (2003) emphasizes endogeneity problems in analyzing block ownership and concludes that “[s]urprisingly few major corporate decisions have been shown to be different in the presence of a blockholder” (p. 60). Controlling for selection effects dramatically changes our inference about block impact on firm behavior. We show that large shareholders indeed play a large role for corporate governance in large public U.S. firms, and can have large effects on several firm policies. For example, large shareholders target firms with poor performance, but conditional on doing so, they improve performance, consistent with monitoring theories of blockholders. In addition to effects on performance, we test for a large shareholder's impact on corporate policies, many of which have not been previously analyzed, and we find evidence consistent with blockholder monitoring causing changes in these policies. On the other hand, capital structure does not appear much affected by large shareholders. In general, our findings are consistent with the interpretation that firms in areas with a relatively low supply of potential block holders display the characteristics of firms with more severe agency problems between owners and managers.

One contribution of our paper is to show that the local supply of potential blockholders (i.e., the local supply of one potentially important governance mechanism) in the area where a firm is headquartered may have significant effects on firm behavior, i.e. that geography matters for the ownership structure of firms (see Coval and Moskowitz (1999) for related patterns for institutional stock ownership). We use these patterns to empirically model the supply side of large shareholders. This allows us to examine whether ownership structure impacts firms' behavior, controlling for selection effects. The economics of the supply of blocks is largely unexplored in the governance literature, perhaps because the focus has mainly been on partial equilibrium analysis and results (see, however, Gorton and Kahl (2008) for a recent exception).

Our paper also relates to work in corporate governance which attempts to address endogeneity problems. Other papers in this area include Bertrand and Mullainathan (2003) (changes in antitakeover laws) and Pérez-González (2005) (the abolishment of foreign majority ownership regulations in Mexico). Several papers use the passage of the Sarbanes-Oxley Act as an exogenous change (e.g., Chhaochharia and Grinstein (2007) and Hochberg, Sapienza and Vissing-Jorgensen (2007)). In contrast to these papers, which use changes in laws (i.e., quasi-natural experiments) to address endogeneity concerns, we develop an instrumental variable for a specific governance mechanism (block presence) in a firm.

The paper is organized as follows. Section 2 reviews the role of large shareholders in the governance of public corporations. Section 3 describes our data. Section 4 defines the instrument, describes the econometric model that we use to identify and quantify the impact of blocks on firms, and examines the validity of the instrument. Section 5 analyzes the impact of large shareholders on firm performance and corporate policies using our two-stage selection model. Section 6 concludes.

2. Large Shareholders and Corporate Governance

2.1. Monitoring and the effects of large shareholders

Theory suggests that large shareholder monitoring may have a significant impact on several aspects of firm behavior, including a firm's operating profitability, corporate investment policy, and choice of financial and executive compensation policies. In this section, we review theories of large shareholders and discuss the empirical predictions which we test in the remainder of the paper.

Theories with a beneficial view of large shareholders typically start by positing agency problems between owners and managers, and by assuming that large owners are able to reduce these. Due to the size of their stake, they either have better incentives to monitor or more power to influence decisions than small, dispersed owners.⁵

⁵ Some recent theories argue that large shareholders can exert governance even if they do not intervene in a firm's operations (e.g., Admati and Pfleiderer (2009) and Edmans (2009)).

Large shareholders use their influence over firm behavior in several ways. They may affect operating decisions by monitoring management. This can result in improved project selection and investment levels, and possibly reduced waste. Demsetz and Lehn (1985) argue that blockholder monitoring restrains managers from consuming firm resources, thereby improving performance. Also, the presence of a block may mitigate overinvestment by management and improve the investment efficiency of the targeted firm (Jensen (1986)). To test these predictions empirically, we examine the impact of large shareholders on measures of operating performance (e.g., ROA), costs (such as SG&A expenses), and investment behavior.

Blockholders may also affect firms' financial policies. Again, the starting point of the argument is that managers may want to deviate from optimal capital structures (i.e. optimal in the view of owners). Managers may prefer lower leverage to reduce firm risk and protect their underdiversified human capital (e.g., Fama (1980)). Alternatively, managers may prefer to increase leverage in order to inflate the voting power of their equity stakes (see Harris and Raviv (1988) and Stulz (1988)). In either case, there is a role for large shareholder monitoring related to capital structure decisions. In terms of payout policy and cash holdings, managers may prefer lower payouts as they reduce the amount of resources under management's control and subject managers to capital markets discipline when the firm requires external financing (e.g., Easterbrook (1984) and Jensen (1986)). Monitoring by a large shareholder may restrict managers from reducing payouts or holding too much cash.

CEO pay is perhaps the most obvious potential source of conflicts between management and owners. Whether CEOs are paid excessively and set their own pay is a question which has been subject to considerable debate and academic research (see, e.g., Bebchuk and Fried (2004) for a review of this work). Because large shareholders may restrain rent extraction by a firm's CEO, we examine whether the presence of a block reduces CEO pay.

2.2. The costs of large shareholders

There are several potential costs of having a large owner. For example, monitoring may reduce managerial initiative (e.g., Burkart, Gromb and Panunzi (1997)). Even when the presence of a block is ex post efficient, in the sense that the owner will improve investment decisions, the possibility of future intervention may reduce a manager's ex ante incentive to exert effort and to search for value-increasing investment opportunities (in the Burkart et al. model, this happens because the manager will sometimes be overruled by active owners, reducing his utility). The net effect of concentrated ownership, i.e. the benefits of monitoring versus the cost of reduced managerial incentives, is indeterminate. In the model, the net effect depends on the size of the monitoring shareholder's stake. This theory thus predicts that the impact of blockholders on firm financial performance is ambiguous.

Another potential cost of large shareholders is reduced liquidity in a firm's publicly traded shares. Glosten and Milgrom (1985) suggest that one cause of stock market illiquidity is the presence of privately informed traders. One such group of potentially informed traders is blockholders (especially in our sample, which precedes Regulation FD). Holmstrom and Tirole (1993) model how concentrated ownership reduces incentives for information production in the stock market by reducing liquidity. Bhidé (1993) argues that large shareholders who reduce agency problems by monitoring management may also reduce stock liquidity by increasing informational asymmetries. Based on these theoretical predictions, we test whether blocks impact the liquidity of a firm's shares using measures of stock liquidity (e.g., Amihud's illiquidity measure).

Other theories argue that blockholder-liquidity causality might go in the opposite direction of the above arguments. For example, Kahn and Winton (1998) model the choice to exit or monitor/intervene to improve firm performance, and predict that intervention is less likely when the firm's stock is more liquid because of the lower cost of exit. Maug (1998) argues that liquidity may reduce the cost of creating a block because it makes it less costly to purchase more shares, so we would expect more blocks in firms

with more liquid stocks. Our instrumental variables approach will allow us to test and isolate the causal effect of blocks on liquidity without selection effects confounding the estimates.⁶

3. Data

3.1. Data on large shareholders

We analyze an unbalanced panel of large publicly traded U.S. corporations from 1996-2001. The sample consists of non-dual class S&P 1500 firms and was originally compiled by Dlugosz, Fahlenbrach, Gompers and Metrick (2006) (DFGM) who identify a firm's 5% blockholders, if any.⁷ The raw database contains 1,919 different firms, 18,818 blockholder-firm-year observations, and approximately 7,600 firm-year observations.⁸ We exclude financial firms and utilities, which reduces the sample to approximately 6,000 firm-year observations. Firms are included in our sample whether or not they have any large shareholder.

We define a large shareholder as a *non-managerial individual blockholder* if it meets both of the following criteria. First, an individual has the sole voting power of the block. Second, this individual is neither a current nor a former officer of the firm (e.g., we exclude Bill Gates' block in Microsoft). We identify current and former officers through proxy statements and a Dow Jones Interactive news search. We exclude holdings by current and former officers' spouses or trusts (e.g., the Bill and Melinda Gates Trust). We exclude blocks held by second generation family members only if a family member is an officer (e.g., the block by the Hewlett and Packard families is included as the family has non-executive

⁶ Several theories explicitly consider the trade-offs that result from the simultaneous benefits and costs of large shareholders. For example, Bolton and von Thadden (1998) argue that large shareholders reduce the threat of takeovers.

⁷ Available from Wharton Research Data Services, <http://wrds.wharton.upenn.edu>. The use of annual data may underestimate the true number of large shareholders because some blockholders can enter and exit our panel within a year, thus not showing up in firms' proxy statements. If some large shareholders choose to own a stake below the 5% reporting threshold, they also do not show up in the proxy statements and in the DFGM database.

⁸ The database has 20,975 blockholder entries. However, not all of these entries are 5% blockholders. For example, "Karen Smith" and "Peter Smith" are recorded in the raw data, but they jointly hold shares through the "Smith Family Trust." Conditioning on the final blockholder flag in the DFGM database will eliminate Karen Smith and Peter Smith, and retain only the Smith Family Trust.

directorships but none of the family members are officers of HP). Our dataset contains 715 firm-year observations with at least one non-managerial individual blockholder.

Who are these non-managerial individual large shareholders? In Appendix A, we provide details for six examples of large, non-managerial, individual shareholders in our sample. Such an analysis cannot be exhaustive and will be very descriptive, but gives an idea of the characteristics of the large shareholders we study and sheds some light on the reasons behind their block investments. Also, some of the examples show the channels through which large shareholders impact the firms in which they hold a large stake.

3.2. The location of large shareholders

We obtain data on the geographic location (zip code) of each block in our sample from firms' proxy statements. We exclude blocks located abroad. For some individuals who are officers or directors, the proxy statements do not give an address, but state "Same address as company". We thus underestimate the actual distance between the location of some individuals and firm headquarters. Also, when individuals have multiple residences, we use the one reported in the proxy. We obtain zip codes of firm headquarters from the *Execucomp* database, the *Compact Disclosure* discs, or hand-collect them from proxy statements. Finally, we identify the latitude and longitude of each blockholder and firm by matching the zip codes to data from the U.S. Census Bureau's Gazetteer Place and Zip Code Database.

We compute the distance (in miles) from a blockholder to each of its firms' headquarters using the methodology of Coval and Moskowitz (1999). Table 1 reports a mean distance of 385.4 miles and a small median distance of 28.5 miles for non-managerial individual blockholders. More than half (61%) are located in the same state as the firm in which they hold a block. Not surprisingly, the median distance between blockholder residence and firm headquarters for managerial blocks is zero and more than 90% of all managerial large shareholders are located in the same state as the corporate headquarters. The results are significantly different for institutional investors. The median distance between mutual fund (money manager) headquarters and the headquarters of the firms in which they own large stakes is 1,158 miles

(871 miles). Less than 10% of all mutual fund and money manager headquarters are located in the same state as the corporate headquarters of their block investments.⁹

3.3. Geographic data

Our data for the geographic distribution of high net worth individuals are from the Statistics of Income (SOI) program at the Internal Revenue Service (IRS).¹⁰ It reports the number of high net worth individuals in each state, estimated from estate tax return filings. The SOI reports estimates of the number of high net-worth individuals by state in 1995, 1998 and 2001 as well as their estimated average wealth. The definition of high net worth varies somewhat from year to year. In 1995, it is net wealth of \$0.6 to \$10 million (in terms of counting the total number of high net worth individuals, the upper cutoff makes a minor difference). In 1998 and 2001, the lower limit is \$1 million and there is no upper limit. However, the three cross-sections are very similar. We use 1995 data because they predate the block observations in our sample, but we have also used 1995 data for 1996-1998 and 1998 data for 1999-2001, as well as 1998 data for all years, with very small differences in results.

We also collect several control variables that vary at the state level. We obtain education, population and income data from the Census Bureau. We collect corporate income tax rates for 2000 (we have used other years as well) from various sources. When there are multiple brackets we use the rates applying for highest income levels.

3.4. Firm variables

The theories of large shareholders discussed in Section 2 predict that large shareholders can impact a broad set of corporate decisions:

⁹ Gaspar and Massa (2007) analyze the local ownership of mutual funds and find a local bias. Their analysis is not inconsistent with ours because they aggregate local ownership of many different mutual fund managers working for different firms to calculate a measure of total ownership in the geographic proximity of a firm's headquarters. In contrast, we aggregate shares across different fund managers within the same mutual fund family and attribute the total ownership to the location of the headquarters of the mutual fund family.

¹⁰ For details, see IRS, Statistics of Income Bulletin, Winter 2002-2003, Publication 1136 (Rev. 4-2003).

- (i) Operating performance. We analyze return on assets (ROA) and return on sales (ROS). We also analyze cost cutting policy (scaled SG&A expenses) and investments (scaled capital expenditures).
- (ii) Capital structure. We analyze measures of book and market leverage.
- (iii) Resources and payout policy. We analyze total payout as a fraction of EBITDA and cash holdings.
- (iv) Managerial compensation and board structure. We analyze the log of total CEO pay and the fraction of pay in equity. We also analyze the number of outside directors on the board.
- (v) Liquidity. We analyze trading volume and the illiquidity measure of Amihud (2002).

Our data source for annual accounting variables is Compustat. We exclude observations with explanatory variables in the top and bottom 1%, but we have also used winsorizing at the 1% level with very small differences in results. Our data sources for CEO compensation and board structure are the Execucomp database and the IRRC director database, respectively. The data used to calculate the liquidity measures come from the Center for Research in Security Prices (CRSP). All variables are defined in Appendix B.

Table 2 reports summary statistics. Panel A shows that there exists at least one non-managerial individual block in 11.8% of all firm-year observations. A manager (either current or former) holds a block in 21.9% of all firm-year observations. More than two-thirds of all sample observations have at least one large mutual fund shareholder. Panel B reports means, medians, and standard deviations for the variables we use in Section 5.¹¹

¹¹ The number of blockholders in Table 3 does not correspond to those in Table 1, because Table 1 is the raw database and allows for more than one large shareholder per firm-year.

4. Instrument and empirical methodology

4.1. *The instrumental variable*

We define our instrument, the density of high net worth individuals, as the number of high net worth individuals in 1995 divided by the total number of firms headquartered in a state, using the 1995 SOI data and information on firm headquarters from the Compustat database. Throughout the analysis, we normalize the measure so that it is measured in units of thousands of individuals per firm headquarter.

The significant geographic variation in the density of high net worth individuals is illustrated in Figure 1. It is worth pointing out that states with a large number of high net worth individuals, e.g., California, will not necessarily have the highest density measure when a large number of firms are also headquartered in that state. Across states, the median wealth density is 565 (Oklahoma). The mean is 664. The 25th and 75th percentiles are 419 (Arizona) and 733 (Idaho), respectively. The lowest densities are 228 (Colorado) and 247 (Utah) and the highest are 1,857 (North Dakota) and 2,500 (Alaska). Of the top five states in terms of number of headquarters, California has a density of 442, Texas 299, New York 473, Florida 603 and Massachusetts 270. Across sample firms, the mean density is 460 individuals per firm and the standard deviation is 151.

4.2. *Estimated wealth distributions*

Substantial personal wealth is required to buy a 5% or larger stake in one of the firms in our sample. Our data on high net worth individuals counts many wealthy people who do not in fact have the resources to buy a 5% stake in most (or any) listed firms. This raises two questions: first, is our variable informative about the very top of the wealth distribution, i.e. is it correlated with the density of billionaires as well as millionaires? Second, are there enough wealthy individuals to explain the observed frequency and size of the stakes we classify as individually held? To assess whether the number and wealth of high net worth individuals are large enough to be consistent with the value of blocks in our sample, we have to extrapolate from wealthy to *very* wealthy individuals. We use the fact that empirical

wealth distributions closely follow a power law, at least for the top 5-10% of individuals by wealth (e.g., Levy and Solomon (1997) and Klass, Biham, Levy, Malcai and Solomon (2006)). Such a power law implies a particular relationship between the wealth rank of an individual and her wealth. Using this empirical fact, we may estimate the number of people of any particular wealth level (as long as it is in the top 5-10% of individuals) using the number of people above the cut-off for the SOI wealth data.

We use the power law parameter of 1.36, as estimated by Levy and Solomon (1997), to estimate the top of the wealth distribution.¹² We infer the number and wealth of the very wealthiest individuals in each state in 1998. Table 3 shows our estimates of the wealth distributions for Oregon (with the median number of millionaires: 27,000), Georgia (75th percentile: 64,000), New York (2nd highest: 243,000), and California (highest: 412,000). In these four states, the estimated wealth of the hundred richest individuals was \$17.4 billion, \$32.8 billion, \$87.4 billion, and \$128.8 billion, respectively. The estimated number of ultra-wealthy individuals – with more than \$100 million in personal wealth – is 51 in Oregon, 122 in Georgia, 463 in New York and 785 in California. The table also reports the number of sample firms in each state, the total market value of equity of these firms, and the median, average, and largest firm market values in 1999. Five percent of the median firm's market value across the four states is \$43 million, \$70 million, \$171 million, and \$58 million, respectively, and there are an estimated 162, 198, 223, and 1,659 individuals with enough estimated wealth to buy such a stake if they invested all their wealth. These estimates suggest that there is enough money in the hands of the wealthiest individuals for them to potentially play a significant role as owners of large blocks of shares.

The estimates in Table 3 depend on the accuracy of the power law distribution. The following exercise shows that an extrapolation using the power law distribution works remarkably well and that the variable we use tracks the whole top of the wealth distribution, including its far tail. Forbes publishes an annual list of the wealthiest Americans and their principal residences. The minimum wealth required to make the 1998 list was \$430 million. We count the number of high net worth individuals in each state.

¹² Levy and Solomon (1997) use data from the 1996 Forbes 400 list of the richest people in the U.S. to back out the parameter of 1.36. Klass et al. (2006) calculate an average parameter of 1.49 using the Forbes 400 lists during 1988-2003. We use the 1996 parameter as it predates our sample.

We then use the power law and the SOI data to predict, for each state, the number of individuals with wealth greater than \$430 million. Finally, we compare our estimates to the actual number of ultra-wealthy individuals. Figure 2 plots the actual against the predicted number of individuals with more than \$430 million in each state. The correlation coefficient is 0.89 and highly statistically significant. Thus, the power law distribution can be used to accurately predict the geographic distribution of ultra-wealthy individuals using the distribution of moderately wealthy individuals.

4.3. Two-stage selection model

Most of the existing empirical literature in corporate finance on the economic effects of large shareholders has estimated an equation of the following form:

$$y_{it} = \gamma d_{it} + \delta_{j(i)} + \lambda_t + \beta X_{it} + \varepsilon_{it} \quad (1)$$

where i , and t index firms, and years, respectively and $j(i)$ indicates the industry to which firm i belongs. y_{it} is a firm performance or corporate policy variable, $\delta_{j(i)}$ are industry fixed effects, λ_t are year fixed effects, and X_{it} is a vector of time-varying firm-level controls, e.g., the size or age of the firm. d_{it} is an indicator variable that is equal to 1 if a large shareholder is present in firm i in year t , and 0 otherwise.

It is not possible to draw causal inferences about impact of blockholders on firm behavior from estimates of γ in equation (1), because there are economic reasons to expect that blocks are not randomly allocated to firms. For example, firm size or growth options are potential determinants of corporate policy choices and firm performance, but the same variables may also be correlated with the presence of a large shareholder. If X_{it} does not include all relevant variables, the large shareholder indicator variable d_{it} will be correlated with the error term in equation (1), and OLS estimates of γ will be biased. For example, if the selection and treatment effects go in opposite directions, γ will be biased towards zero.

To explicitly address this endogeneity problem, we use a two-stage selection model, following Heckman and Robb (1985). The model we estimate differs from standard two-stage least squares (TSLS)

models because of the binary nature of the endogenous variable of interest (the presence of a block). As in standard TSLS, the estimation involves two equations. The first stage equation is:

$$d_{it}^* = gw_{it} + a_{j(i)} + c_t + bX_{it} + e_{it}$$

$$d_{it} = \begin{cases} 0 & d_{it}^* > 0 \\ 1 & d_{it}^* \leq 0 \end{cases} \quad (2)$$

where d_{it}^* is a latent variable and w_{it} is our instrument, i.e., the number of high net worth individuals per firm headquarter in a state. $a_{j(i)}$ are industry fixed effects, c_t are year fixed effects, and X_{it} is a vector of firm-level control variables. We report estimates of the first stage using both linear probability models and probit models.¹³

In the second stage, firm performance or corporate policies are modeled and it is assessed how they are affected by the predicted probability of block presence in a firm. The second stage equation is:

$$y_{it} = \gamma_{Block} \hat{d}_{it} + \delta_j + \lambda_t + \beta X_{it} + \varepsilon_{it} \quad (3)$$

Since the only endogenous variable in equation (3) is an indicator variable, this model is sometimes referred to as a “dummy endogenous variable model” (Heckman (1978)). The vector of control variables, X_{it} , is identical to the one included in the first stage. We assume that e_{it} and ε_{it} follow a joint normal distribution with a correlation coefficient of ρ . The model is estimated by maximizing the joint likelihood function (see also Maddala (1983)).¹⁴ The estimated coefficient γ_{Block} captures the treatment effect of a large shareholder on y_{it} .

If the error terms e_{it} and ε_{it} are uncorrelated ($\rho = 0$), then the effect of large shareholders on firm performance or corporate policies can be consistently estimated with ordinary least squares (OLS). If ρ is

¹³ Similar model specifications have recently been estimated in the corporate finance literature (see, e.g., Li and Prabhala (2007)). Our specification with a first stage probit (and a variable in the first stage which is not included in the second stage), is similar to the treatment effects model estimated in Villalonga and Amit (2006).

¹⁴ The MLE model requires the assumption of joint normality of the error terms. An alternative approach would be to model the first stage as a probit model and then use traditional instrumental variable techniques where the IV is created from the probit fitted first stage values. We have also estimated these regressions, and the overall results from the probit-IV regressions are similar in significance and magnitude to our reported MLE results.

positive, then the coefficient γ estimated through OLS is overstated; if ρ is negative, then it will be understated. We test whether ρ is zero with a Chi-square test which is reported in all tables.

4.4. Validity of the instrument

4.4.1. Relevancy condition

The first condition for wealth density to be a valid instrument is that it significantly predicts the presence of large non-managerial individual shareholders in local firms. There are economic reasons to expect such a relation. First, we found above that blockholders exhibit a preference for firms headquartered near where they live. Second, we also found that there is significant geographic variation in the density of wealthy individuals. This is important because personal wealth is likely a key determinant of an individual's ability to accumulate a large equity stake in a public firm.

It is important to run regressions of block presence on wealth density with firm- and state-level controls to rule out the possibility that our instrument is correlated with some firm-level determinant of blocks such as firm size or state level variables such as corporate tax rates. But the simplest way to assess the validity of the relevancy condition is to collapse data by state. Figure 3 plots the frequency with which firms in a state have a large individual non-management shareholder against our instrument. Each state is represented by a circle whose area is determined by the number of firms used to calculate the state block frequency.¹⁵ States that are large in terms of firms, such as California and Texas, appear large (918 and 649 observations, respectively). On the other hand, South Dakota (the low outlier just below 1.5 on the x-axis) or Iowa (the high outlier with block frequency around 55%) are small (six and 40 observations, respectively). The graph shows an estimated regression line where observations have been weighted by the number of firms used in calculating the frequency (i.e. the circle sizes). The estimated slope is 0.190 (heteroskedasticity-robust standard error of 0.05; t -statistic of 3.77).

¹⁵ The block frequency is an average across our sample years to maximize the number of observations used to calculate the average. Using only the 1996 cross-section of blocks and firms increases the significance and estimated slope.

4.4.2. *Exclusion restriction*

The second condition for our instrument's validity is that it impacts the dependent variables only through its effect on the endogenous block variable, i.e., wealth density must not affect the dependent variables directly, and not through any other mechanism. That is, the correlation between the instrument and the second stage error term must be zero. The validity of this condition cannot be tested with only a single instrument, but we provide several empirical observations and arguments to support its validity.

One concern may be that areas with wealthy individuals experienced recent wealth creation (e.g., Microsoft executives in the State of Washington) so that there is a direct effect of wealth density on firm performance, or potentially reverse causality. Three arguments alleviate this concern. Our instrument reflects "old money" because it is based on inheritance data, and is not related to contemporaneous firm performance and recent IPOs, which reduces concerns of reverse causality with firm performance. Also, our block variable does not include executives (current or former officers), i.e., those most likely to have wealth created in IPOs and through superior firm performance.¹⁶ In addition, we measure our wealth density instrument with a lag, i.e., prior to our sample period, and while our instrument is measured at the state-level, we examine performance and policies at the firm-level.

Second, while the density of high net worth individuals predicts individual non-managerial blocks, it does not predict the presence of large shareholdings by either managers or institutional investors. It therefore seems unlikely that our instrument measures the "demand for blocks". If it did, we would expect wealth density to predict the presence of several different types of blocks, and not only non-managerial individual blocks.

¹⁶ For example, we do not analyze Bill Gates's block in Microsoft. We do include Gates's block in Washington Water & Power because he is a non-management individual blockholder in that company according to our definition. Microsoft clearly made many executives in the State of Washington wealthy. Some of these executives may choose, like Bill Gates, to own blocks in local firms. However, to explain our results, there must also be a relation with firm performance and policies. Washington Water & Power did not relocate to the State of Washington because of Microsoft – the company was founded in Spokane, WA, in 1889. While it is possible that the success of Microsoft made Washington Water & Power more profitable, it is not clear that we can explain our other firm policy evidence, e.g., why there would be a reduction in corporate cash holdings or top-executive pay in Washington Water & Power because of Microsoft's success in the region.

Third, it appears unlikely that location choices of firms and individuals are motivated by the market for blocks in listed firms. Firm headquarter relocations are extremely rare events. Pirinsky and Wang (2006) identify only 115 relocations among more than 5,000 firms between 1992 and 1997. Therefore, firm locations must typically be determined by early decisions that long pre-date the ownership structure in the very large firms in our sample.

Wealthy individuals are presumably more mobile. Various agglomeration economies affect location choices (see e.g., Puga and Duranton (2004) or Glaeser (2008) and the references therein). Wealthy people likely want to live near other wealthy individuals for social reasons and because they have correlated tastes. Such agglomeration seems consistent with the apparent concentration of wealthy people (New York's Upper East Side, Aspen, Beverly Hills, East Hampton), and is not primarily about the production side of the economy, but about the consumption side, because that is where the major externalities are (this is unlike location choices of the middle class, where productivity externalities may dominate). In particular, expectations about future policies (e.g., leverage or payout policy) in local firms are probably less likely to figure much in location decisions of wealthy individuals. Thus, we do not believe endogenous firm or individual/block location choices affect our results to a significant extent.

Finally, because we study a very broad set of corporate policies, any theory of a direct relation between wealth density and a particular policy also has to explain a large number of relationships, i.e., both the relation between wealth density and block presence and the one with all the dependent variables. We are aware of no economic theory which predicts a direct relation between wealth density and the multitude of dependent variables we study.

4.5. Weak instrument problem

If instrumental variables are insufficiently correlated with endogenous variables, identification with IV suffers from “weak instrument problem” (e.g., Bound, Jaeger and Baker (1995), Nelson and Startz (1990), and Staiger and Stock (1997)). Weak instruments may lead to two problems. First, IV coefficient estimates are biased toward OLS coefficients. Second, reported confidence intervals may be

understated. The problem of weak instruments is serious in settings with a large number of instruments, which are individually weakly correlated with the endogenous variable(s), but which as a group are able to explain some of its variation (see, e.g., Angrist, Imbens and Krueger (1999)). When the number of instruments is low, however, IV results are neither particularly biased nor generate excess precision. Angrist and Krueger (2001) state that if the number of instruments is equal to the number of endogenous variables, there can be no weak instrument bias. Given this result and the fact that we use a single instrument, weak instrument bias should not be a concern.¹⁷

4.6. Heterogeneity of large shareholder effects

The impact of large shareholders is likely to vary across firms and owners. Our results should be interpreted with this in mind. With homogenous effects (conditional on the control variables), performance and corporate policies will be similarly affected by a large shareholder across all firms. In that case, the second stage large shareholder coefficients we estimate can be interpreted as the effect for any firm with a large shareholder, including the majority of firms who do not actually have such a shareholder. If the homogeneity assumption does not hold, however, our regressions estimate the average impact of blocks across a subset of firms that actually have such blocks. This is called a local average treatment effect. Specifically, our estimates reflect the mean effect for those firms that have a large shareholder because they are located in states with a high wealth density measure.¹⁸ It seems highly likely that the homogeneity assumption is unwarranted, and that our estimated effects are local, i.e., they would not generally represent the average effect of blocks across the universe of firms. Two forms of heterogeneity may be important. First, the potential impact of a blockholder likely varies across firms, for example because other governance mechanisms are in place in some firms but not others, or simply because some firms are well run and some less so. Our estimated effects therefore have to be thought of

¹⁷ Following Staiger and Stock (1997), it is common practice to examine the strength of the instrument in the first stage using F-statistics. Based on cut-off values of Stock, Wright and Yogo (2002) and Stock and Yogo (2005) and using the Kleibergen and Paap (2006) test statistic that allows for clustered standard errors, we can reject the null hypothesis of weak instruments in all our reported first-stage regressions.

¹⁸ For further discussion, see Imbens and Angrist (1994) and Imbens, Angrist and Rubin (1996).

as average effects for those firms which are marginal targets, i.e., not the most obvious targets (because those firms have blocks in all states), nor the firms with the lowest potential impact of a block (because those firms do not have blockholders even in states with very high wealth density). Second, different blocks are probably formed for different reasons, e.g., to force management to raise leverage, to solidify a commercial relationship, to contain some agency problem, or to force a change in strategy. We estimate average effects across many policies, so there may be larger effects in a subset of firms with blocks.¹⁹

5. Results

5.1. Wealth density and the presence of a large shareholder in a firm

We start by reporting results from the first stage regression, in which we predict the presence of a non-managerial individual blockholder in a firm. Table 4 reports several model specifications. We find that the density of wealthy individuals in a state significantly predicts the presence of a large shareholder in firms headquartered in that state. In all regressions predicting the presence of a non-managerial block, our instrumental variable's t -statistics are significant at the 1%-level. The heteroscedasticity-consistent standard errors are clustered by state. In unreported regressions, we have clustered standard errors by firm and estimate consistently smaller standard errors for the instrument.

A linear probability model of having a non-managerial individual block estimated by ordinary least squares (OLS) is reported in column (1). The estimated coefficient on the density of high net worth individuals, 0.190, implies that a one standard deviation increase in the instrument (0.420) is predicted to increase the probability of having a block by 7.6 percentage points.²⁰ This effect is large compared to the

¹⁹ We may miss potential effects of blockholders if heterogeneity leads to effects of opposite signs (see, e.g., Cronqvist and Fahlenbrach (2008)). If blockholders increase leverage in some firms but reduce it in others, the estimated net effect might be indistinguishable from zero. While perhaps possible for leverage, this possibility seems less likely for, e.g., profits or liquidity. In any case, this suggests that there may be more effects of large shareholders than we identify, but does not suggest ever finding any effects spuriously.

²⁰ This standard deviation of the instrument is calculated across 50 states. Calculating the standard deviation across firms yields a lower number (0.151) because few firms are located in the states with extreme wealth densities. An increase in the density of 0.151 increases the predicted probability of a block by 1.5 percentage points according to column one.

unconditional mean probability of 11.8%. Because block probabilities are close to zero for most firms, the linear probability model is possibly a poor modeling choice and we therefore turn to other model specifications.

In column (2), we estimate probit regressions. Our previous finding that block presence in a firm is significantly predicted by the density of local wealth is robust and does not change across specifications. Around the sample mean probability of having a block, the effect of increasing the density by one standard deviation is to increase the block probability by about 4.5 percentage points, a 38% increase. That is, the magnitude of the effect remains large in this model. In column (3), time-varying firm controls, as well as year and industry-fixed effects are included.²¹ The estimated effect of the instrument is slightly lower, but remains significant at the 1%-level. This regression corresponds closely to the first stage estimation used to identify the effect of blockholders on a range of dependent variables below. Size has a negative correlation with block probabilities, and age and sales growth have positive correlations.

In the remaining columns of Table 4, we report robustness checks and placebo tests. First, we consider alternative ways of defining the instrument and dependent variable. In column (4), we replace the numerator of the wealth density measure (the number of wealthy individuals) with the total wealth held by these individuals, thus incorporating information on the high net worth individuals' average wealth (based on data from the 1995 SOI). At the sample mean, a one standard deviation increase in the wealth per listed firm variable increases the predicted probability of having a block by ten percent. In column (5), the dependent variable is the percent of common stock owned by individual shareholders with large blocks, adding up across blocks if there is more than one such block (18% of the firms with individually held blocks have more than one such block, most often two). Since this is a continuous variable, the regression is estimated with OLS. The coefficient on the density variable is highly significant and predicts that a one standard deviation increase in the instrument results in a 0.96%

²¹ Industries are based on Fama and French (1997). We have checked that our results are similar when using 2-digit North American Industry Classification System (NAICS) industries.

increase in the average fraction of shares held by large individuals shareholders (the unconditional average is 1.79% of shares). We conclude from the tests reported so far that there is a robust effect of local wealth on individually held blocks in local firms, across various ways of measuring both wealth and ownership, and controlling for important firm level variables.²²

It is important for our identification not just that the instrument predicts individual blocks, but also that it does not predict the presence of other types of blocks (some small negative effect is to be expected, since the ownerships shares of all owner categories add up to one). If it did, we might be concerned that any effects we identify might reflect those other owners. In column (6), we report a probit specification with a dummy for the presence of a managerial blockholder as the dependent variable. The coefficient on the instrument is positive, but insignificant. Age does not seem to be related to having a mutual fund-owned block, but profitability has a large positive coefficient. Column (7) shows a negative, but statistically insignificant, effect of the instrument on the propensity to have a block owned by a mutual fund or money manager.

There are two important conclusions from Table 4. First, non-managerial individual blocks are not randomly allocated to firms, but economic mechanisms systematically affect block allocations. For example, large shareholders select smaller and older firms with poor past performance, but which have significant growth in their sales. Conditional on firm size, firm age is positively correlated with the presence of a blockholder in a firm, suggesting that non-managerial individuals do not invest in the youngest firms. These are characteristics of firms where the agency problems between owners and managers may be particularly severe. One interpretation is that the evidence so far is consistent with economic theories of large shareholders as monitors, and that blocks are systematically allocated to firms based on where the demand for monitoring is more significant. In particular, our instrument strongly predicts the presence of a large individual shareholder. Meanwhile, the placebo tests suggest that our

²² As a further robustness check, in unreported regressions, we have used the log of the number of high net worth individuals and the log of the number of firms as separate independent variables, or the number of high net worth individuals per capita and the number of firms per capita. Both variables are always significant of opposite sign and similar magnitudes.

instrument is not identifying a set of firms with more blockholders in general, but very specifically predicts the presence of individually held blocks. Other block types are, if anything, less common when there are many local high net worth individuals.

Second, causal inference regarding the impact of blocks of shares on firms is not possible from standard econometric models such as equation (1) and OLS. Because several of the variables which explain the presence of a blockholder in a firm (e.g., past performance, firm size) can also be expected to play a role in explaining firm performance and policy choices in the second stage regressions, standard models will confound selection and treatment effects of blocks. For example, we find that large shareholders select firms with relatively poor recent past performance, but conditional on such selection, we expect blockholders to improve performance through monitoring of management. This is precisely why it is important to use an instrument when quantifying the impact of block presence on firms.

5.2. State-level control variables

Because the variation in our instrument comes from state-level differences, a concern is that we so far have captured differences across states that are correlated with blockholder presence but have nothing to do with the economic explanation we provide. If a variable is both a determinant of block presence and correlated with our wealth density measure, our identification is not valid. Therefore, we re-estimate the first stage with controls which may be correlated with blocks: population density, the state-level corporate tax rate, per capita income, and educational attainment.

Table 5 shows that none of these state-level control variables significantly predict the presence of a non-managerial individual blockholder in a firm. Also, the inclusion of state-level control variables does not affect the estimated effect of our instrument much. Columns (1) – (4) include the state-level controls one by one. In column (5), which includes all the state-level controls, we find that the estimated coefficient on wealth density is 0.626 and statistically significant at the 5%-level. That is, measures of economic performance or the quality of the labor force in a state do not seem to explain our evidence.

5.3. Empirical analysis of the effects of large shareholders

5.3.1. Firm performance

Table 6 reports evidence on the effect of large shareholders on performance. Column (1) estimates effects for ROA. Age, sales growth and dividend yield have positive coefficients, while age squared has a negative coefficient.²³ The coefficient on the instrumented block indicator variable is 0.037, significant at the 10%-level. This implies a predicted 3.7 percentage points higher profitability, all else equal, with the presence of a large shareholder in a firm. The magnitude of this effect is large given that the mean ROA is 4.7%. However, the second-stage regression estimates a coefficient for the *predicted probability* of having a large non-managerial individual shareholder as opposed to the existence of a large shareholder. Therefore, the magnitude of the second-stage coefficient measures the difference between a firm that is predicted to have a large shareholder with certainty and one that is predicted with certainty to not have a large shareholder.²⁴ It is important to note that the relatively large profit effect estimated here is identified from the marginal firm having a block when our instrument is high, but not if it were low, and that these firms may see larger improvements from having a block than the general population of firms. We cannot use the 0.037 coefficient to predict that if we added an individual non-managerial block to the 88% of sample firms which do not have one, they would experience an average ROA improvement of 3.7%. This effect refers to firms that are possible targets, i.e., that actual block holders decide to own a stake in.

The estimated correlation coefficient of the error terms in the first and second stage equations is negative (-0.193) and statistically significant at the 5%-level. That is, there is a systematic inverse relation between the unexplained (based on our model in the first stage) probability of having a block in a

²³ We include the dividend yield and dividend payer dummy in the reported specification for profitability, following Fama and French (2000). Payout policy is potentially endogenous and affected by large shareholders. We find that the statistical significance of the block indicator variable is somewhat stronger if we exclude the dividend variables from the ROA and ROS specifications.

²⁴ For a less extreme effect, one may multiply the estimated coefficient with, e.g., the interquartile range of the distribution of predicted probabilities (0.16). For expositional clarity, we discuss our results in Section 5.3 as moving from a probability of a large block of 0 to a probability of a large block of 1.

firm and the unexplained part from our operating performance regression model. An economic interpretation of this result is that large shareholders tend to target firms with relatively poor operating performance, holding other firm-level variables constant.

In column (2) of Table 6, we replace ROA with ROS (profits scaled by sales). The impact of large shareholders is robust and remains positive and significant (this time at the 1%-level). Again, the effect appears economically large.

There are several potential explanations for a positive block impact on profitability and firm cash flows. We study two of the most straightforward ones: cost reduction and investment policy. In column (3), we find that SG&A expenditures, a measure of cost cutting, do not seem to be significantly impacted by blocks. In column (4), we find that investments (scaled capex) of firms are lower by around 7% of assets in the presence of a block, with a coefficient significant at the 5%-level. This can be compared to mean investment of 28% of assets. The estimated error term correlation between the first and second stage equations is positive, suggesting that large shareholders select firms that invest more.

The evidence in Table 6 is consistent with theories of large shareholders as monitors. First, theory predicts that blockholders may monitor and restrain management from consuming firm resources (e.g., Demsetz and Lehn (1985) and Shleifer and Vishny (1986)), thereby improving return on assets and other operating performance measures. Blockholders target poorly performing firms, but conditional on such selection, they have a positive impact on profitability.²⁵ We also infer that the positive effect of blockholder presence in a firm empirically must dominate the reduction in managerial incentives effect identified by Burkart, Gromb and Panunzi (1997). One economic source through which large shareholders seem to systematically affect firm performance is improved investment efficiency.

²⁵ Our evidence complements existing evidence on blocks from event studies. For example, Holderness and Sheehan (1985) report that the stock market reacts positively to the announcement of block presence in a firm by six controversial so-called “raiders.” Holthausen, Leftwich and Mayers (1987) study block trades more generally and find that price effects are permanent for buyer-initiated transactions, but temporary for seller-initiated transactions. Barclay and Holderness (1991) study negotiated block trades (mainly corporate blocks) and find that stock price increases are larger when the new blockholder achieves control. Slovin and Sushka (1993) find that the sale of a large inside block to an outside blockholder triggered by the death of an insider leads to positive announcement returns..

Blockholders target firms which invest more than otherwise similar firms, but then the large shareholders work to reduce the overall capex of these firms by restricting management from overinvesting, i.e., selecting and investing in projects with lower returns than their respective cost of capital.

5.3.2. Capital structure

There is a role for large shareholders to monitor leverage choices because theory predicts that managers may deviate from optimal capital structures because of underdiversified human capital (e.g., Fama (1980)) or to inflate their control (e.g., Harris and Raviv (1988) and Stulz (1988)). Some existing evidence is consistent with Stulz's (1988) theory that managers take on excessive leverage to increase the voting power of their equity (e.g., Berger, Ofek and Yermack (1997)). The findings by Jung, Kim and Stulz (1996) support the view that agency problems make some firms issue equity when a debt issuance would have been better for firm value.

Table 7 reports our evidence on blockholder impact for three different leverage measures: book leverage, long term book leverage, and market leverage. Across all measures, we find that the coefficient on the instrumented block indicator variable is negative, although only the one for long-term book leverage is significant (at the 10%-level). This coefficient implies 3.7 percentage points lower long-term book leverage for firms with a large blockholder, which may be compared to a mean leverage ratio of 21% in our sample. Overall, there is little evidence that blockholder presence impacts firms' capital structures. This is an average effect, and if some blockholders reduce leverage while others increase it, the average result might be zero. From a theoretical point of view, this might be plausible because owners' risk aversion and lack of diversification might point to lower leverage, while control motives might favor debt financing over equity issues. The results are also consistent with the absence of any effect of large shareholders on capital structure.

5.3.3. *Resources and payout policy*

Table 8 reports evidence on the effect of large shareholders on payout policy and cash holdings. In column (1), we study the total payout ratio (dividends and repurchases scaled by EBITDA).²⁶ The estimated coefficient on the instrumented block indicator is positive and significant. Hence, large shareholders appear to cause firms to substantially increase their total payout. The effect is large. A block in a firm is estimated to approximately double the payout ratio compared to the sample mean. Moreover, the estimated error term correlation between the first and second stage equations is negative and significant, which suggests that the blockholders we analyze tend to select firms with low payout ratios compared to other similar firms. But conditional on selecting these low payout firms, blockholders seem to cause a significant increase in the amount paid out to shareholders.²⁷

We next consider block impact on corporate cash holdings. In column (2), we find that the coefficient on the block indicator is -0.049 (significant at the 10%-level). Thus, cash holdings of a firm with a block are smaller by 4.9 percent of total assets (sample mean cash holdings are 18%). There is thus some evidence that the presence of a large shareholder in a firm reduces cash holdings by about a third of average holdings.

Theory suggests that managers may prefer larger cash holdings and lower payout ratios because payouts reduce the amount of resources and assets under management's control and potentially subject managers to more capital markets discipline (e.g., Easterbrook (1984) and Jensen (1986)). In analyzing the determinants of corporate cash holdings, Opler, Pinkowitz, Stulz and Williamson (1999), conclude that "management accumulates excess cash if it has the opportunity to do so" (p. 44). A possible problem with excess cash, as found by Harford (1999), is that it may allow management to pursue acquisitions which do not increase shareholder value. Our evidence suggests that blockholder monitoring restrains managers from such resource grabbing.

²⁶ Scaling the total dollar payout by book value of assets or market value of assets does not change our conclusions.

²⁷ In unreported regressions, we separate payouts through dividends from payouts through repurchases and find that the increase in payouts is mostly driven by increases in dividends. One possible explanation is that dividends are a firmer commitment for management (e.g., Jagannathan, Stephens, and Weisbach (2000)).

5.3.4. *Managerial compensation and governance*

Table 9 reports evidence on the impact of large shareholders on managerial compensation contracts and governance structures related to a firm's board. In column (1), where the dependent variable is the log of total CEO pay, the coefficient on the instrumented block indicator variable is -0.96, significant at the 1%-level, implying that CEO pay is reduced by 0.8 of one standard deviation in the presence of a blockholder, which for the typical CEO corresponds to about \$1.5 million annually in lower pay. The estimated error term correlation between the first and second stage equations is positive (0.540) and significant at the 1%-level, suggesting that large shareholders seem to target firms with more generous executive pay packages compared to those of otherwise similar firms.²⁸

This large effect of blockholders seems consistent with the joint hypotheses that CEOs extract rents through compensation (e.g., Holderness and Sheehan (1988), Bertrand and Mullainathan (2000, 2001) and Bebchuk and Fried (2003, 2004)), and that blockholder monitoring can mitigate or eliminate such agency problems. Note that firm heterogeneity may be important, for example if shareholders target poorly managed firms. We cannot infer that firms without blockholders would respond with drastic pay reductions if a blockholder appeared.

We next analyze the composition of pay. In column (2), we find that incentive pay is a significantly smaller fraction of total compensation by 0.446 with a blockholder present in the firm. The reduction in total CEO pay because of large shareholder presence is almost all in terms of the incentive pay component. Many CEOs of large U.S. public firms receive \$1M in cash pay for tax reasons, which may induce a mechanical relation between incentive pay and pay levels (incentive compensation is a lower fraction of the total pay package for CEOs with lower total compensation). This could explain our finding. Alternatively, the evidence of less incentive pay in a firm with a blockholder is consistent with substitution of different governance mechanisms. Firms with a large shareholder as a monitor do not

²⁸ Whatever selection bias affects OLS results would seem unlikely to cloud these large estimated effects of blocks on CEO pay. Indeed, CEO pay is one of the few variables where OLS results are significant in the same direction as our instrumental variables (this is consistent with findings in Bertrand and Mullainathan (2003) that blockholders reduce CEO pay). See Table 11 below for a systematic comparison of OLS and instrumental estimates.

require as much pay-for-performance sensitivity to align the interests of owners and management as do firms with no block.

We also analyze the impact of large shareholders on a firm's board composition. Blockholders may monitor management in several different ways, but one of the most important means of corporate governance and monitoring is a seat on a firm's board. In column (3), where the dependent variable is the number of outside directors of the firm, the coefficient on the block indicator variable is 2.1, significant at the 1%-level. A block increases the number of outside directors by two, which may be compared to the mean number of outsiders of 7.44 among the firms in our sample. That is, blockholders seem to work to increase the number of outside director on the firm's board so that a firm's board composition tends to be substantially different in the presence of a large shareholder in the firm.²⁹ This evidence suggests that large shareholders monitor firms by either serving on the board of directors themselves, or by supporting other outside directors.³⁰

5.3.5. *Liquidity and risk*

Theory suggests that a potential cost of a large shareholder in a firm is reduced liquidity in the firm's publicly traded shares (e.g., Holmstrom and Tirole (1993), Bolton and von Thadden (1998)). Table 10 reports our evidence for three variables: trading volume, Amihud's (2002) illiquidity measure, and share price volatility. We find that a large shareholder in a firm causes the firm's shares to see lower trading volume, have lower volatility, and be less liquid. These effects are significant at the 1%-level and the magnitudes are large. For example, for the trading volume regression in column (1), the estimated regression coefficient on the instrumented block indicator is -0.004, implying that the presence of a blockholder causes a fall in trading volume by 0.6 standard deviations. The evidence of lower trading

²⁹ Outside directors are often thought of as positive from a governance perspective (see e.g. Rosenstein and Wyatt (1990) who report significantly positive share-price reactions surrounding outside director appointments). But researchers have not found much significant evidence of an effect of outside directors on longer-term firm performance measures (e.g., Hermalin and Weisbach (1991) and Bhagat and Black (2002)), possibly because of the endogeneity of board composition (e.g., Hermalin and Weisbach (1998)).

³⁰ Overall board size increases as well (not reported).

volume may suggest that large shareholders mechanically reduce liquidity by keeping shares off the stock market, by shrinking the float and lowering noise trading. However, the likely drop in float is much smaller than the estimated drop in trading volume. A possible additional source of lower volume (and higher illiquidity) is the presence of privately informed traders. This result supports theoretical predictions by, e.g., Glosten and Milgrom (1985) and also Bhidé (1993). They argue that large shareholders who reduce agency problems by monitoring management may also reduce stock liquidity by increasing informational asymmetries. In the context of liquidity, our empirical evidence suggests that the cost of the presence of a large shareholder significantly outweighs any potential benefits for small shareholders.

The evidence also indirectly supports theories arguing that stock market liquidity may reduce the cost of creating a block (e.g., Maug (1998)). The estimated error term correlation between the first and second stage equations in the first and second columns are significant at the 1%-level, suggesting that large shareholders target firms whose stocks are more liquid compared to those of other firms. Because these selection effects are strong, it is important to take them into account when empirically analyzing block impact on liquidity.³¹ The block effects on stock liquidity highlight that there are costs of large shareholders which accompany the benefits of improved monitoring of management and reduction of agency problems.

Column (3) suggests that stock price volatility falls when a firm has an individual non-managerial block, although the effect is less dramatic than the volume and liquidity effects. A block is predicted to reduce annual volatility by 7.5%, which is a little less than a fifth of average and median values. This is consistent with large shareholders influencing corporate policies in a conservative direction.

³¹ Below, we discuss comparisons of OLS and instrumental coefficient estimates (see Table 11 below). Only for one of these dependent variables (trading volume) is the OLS coefficient significant and of the same sign as the instrumental variables estimate, and in that case it is much smaller. This is an indication of the selection effects.

5.4. Robustness checks

We have performed several robustness checks. One concern is that the significance of our instrument is driven by a few states with the most extreme wealth densities. This is not the case because we find that dropping the 10 states with the highest or lowest wealth density has little effect on the economic and statistical significance of our results. The same is true if we drop the 10 states with the lowest number of firm headquarters. A related concern is that New York City is driving all our results. Many wealthy individuals may consider New York City an attractive city to live in, and at the same time, hold blocks in one of the many firms in New York. When we exclude New York City, we find that the economic and statistical significance of the estimated first stage coefficient of the instrument increases.

We have also included additional state-level control variables: a measure of state-level income inequality, a state of Delaware incorporation indicator variable, and the G-index of Gompers, Ishii and Metrick (2003) as a measure of anti-takeover provisions. We have also included fixed effects for the nine Census divisions (regions). None of the variables significantly predict block presence in a firm, and none of them affect the estimated effect of wealth density.

The block indicator may be proxying for non-linear firm size effects. But when we add $[\text{Log}(\text{assets})]^2$ and $[\text{Log}(\text{assets})]^3$ in the regressions, as well as alternative measure of size such as the log of market value and the log of sales, our results do not change (for predicting blockholders, higher powers of size tend not to be significant). Also, individual and institutional blocks may be substitutes. We have added an indicator variable for the presence of a large mutual fund, money manager, or other institutional owner as a control, but it does not affect any of our results.

The denominator of our instrument uses the number of firm headquarters in a state, which may be correlated with factors unrelated to the supply of wealthy individuals who are potential blockholders. We therefore include the two components of our instrument separately in the first stage, i.e., the number of individuals with high net worth and one divided by the number of firm headquarters in a state. Both

measures are statistically significant. The effect of wealth is large and highly significant, showing that our results are not driven by the number of local firms.

Finally, we check the robustness of our results for different classifications of large non-managerial, individuals shareholders. Second-generation family members who are not current or former officers are included in our measure, but we find that excluding these blocks does not materially affect our results. Also, some blockholders who are directors do not provide their private address in firms' proxy statements, but report "Same address as company." Excluding these does not change any of our results although significance is sometimes slightly reduced.

6. Discussion and Conclusion

We develop and test a new instrumental variable framework which allows us to separate selection effects from treatment effects for a large group of blockholders and to quantify their impact on several aspects of firm behavior. We start by documenting that non-managerial individual shareholders hold blocks in firms that are headquartered close to where they live. We then use this empirical fact to create an instrumental variable (the geographic variation in the density of wealthy individuals) for the presence of a large shareholder in a publicly traded U.S. firm. This instrument predicts the presence of a block in a firm with surprising power, and it is robust to the inclusion of variables that vary geographically, reducing concerns about its validity.

Our evidence contributes in three ways to existing research on blockholders and more generally to corporate governance research. First, we find that blocks are not randomly allocated to firms, but are systematically allocated based on where the benefits to additional monitoring are more significant. This result confirms a suspicion about block ownership endogeneity that researchers have had at least since Demsetz and Lehn (1985). Importantly, the non-randomness of blocks matters. The inferences about the impact of large shareholders change significantly once we control for selection effects. Table 11 summarizes this result by showing that OLS estimates of blockholder effects combine selection and treatment effects. For example, while ROS is positively affected by a block according to our instrumental

variables estimates, the LS effect is negative and insignificant. We infer that blocks are more likely in firms with otherwise low profitability, but that the presence of a blockholder tends to raise profits (back toward the mean).

After controlling for block selection effects, we find that large shareholders impact a broad set of firm policies in addition to performance. The presence of a large shareholder reduces the firm's investments, reduces corporate cash holdings, increases payouts to shareholders, and reduces total top-executive pay. These results are economically and statistically significant. Firms with blockholders are also found to have more outside directors on their boards. Many theories predict that large shareholders mitigate agency problems between managers and owners by monitoring of management, but our paper is first to show that blocks cause changes in firm behavior which are consistent with theories of blockholders as monitors.

Finally, we find that large shareholders come with benefits as well as costs. While large shareholders are improving firm performance through monitoring, we also find costs of their presence to other shareholders. One cost is a less liquid publicly traded stock because of a smaller float, and because the presence of privately informed blockholders who might sell, increases informational asymmetries and as a result increases illiquidity, consistent with theoretical predictions (e.g., Bhidé (1993)).

We want to recognize some limitations of our analysis. We have sidestepped the issue of how large shareholders other than individuals, e.g., institutions, impact corporate performance and policies, because our instrument is not expected to (and does not) predict the presence of such owners. It is perhaps even more challenging to identify a valid instrument for the presence of institutional investors than for individual non-management blocks. Also, we have not analyzed interactions of individual blocks with other governance mechanisms because testable predictions have not been worked out theoretically (see Zwiebel (1995) for an exception) and we are concerned about reintroducing endogeneity issues in the analysis. In addition, our empirical framework does not separate the effect of the presence of a large shareholder from the threat of future block presence in a firm. As suggested by Manne (1965), the threat

of actions by blockholders can influence managerial behavior, and if firms in high wealth density areas face a higher threat of block creation, this may partly explain our results.

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Appendix A: Examples of large shareholder monitoring

The data for this Appendix are from Factiva / Dow Jones News searches and the forms SC 13D that are mandatory SEC filings for any person obtaining a stake in excess of 5% in a public firm.

For some blocks in our dataset, it is evident how excess returns may be generated and through which channels influence can be exerted. George Lucas owns, through Lucasfilm Ltd., a 7.4% stake in toymaker Hasbro Inc. This stake was obtained in exchange for licenses to certain intellectual property rights (e.g., StarWars figure toys) and the right of first refusal to others. Clearly, the threat of cancellation of licensing agreements or the threat not to license future film figures and to cooperate with different toy makers appears credible.

Ty Warner, the inventor of the Beanie Babies, bought in February 1998 a 7.3% stake in Cyrk, Inc., a corporate promotion company, after he agreed that Cyrk launches and maintains the “Beanie Baby Official Club.” At the time of the investment, Cyrk Inc. was in difficulties, having lost about a quarter of its work force. Ty Warner bought the stock at \$13, and the stock declined significantly in the following year, down to \$6.50 in February 1999. For the twelve months ending December 31st, 1999, Cyrk reported an increase in net earnings of almost 30%, part of which was attributed to the licensing agreement with Ty Warner, which boosted the stock price to almost \$12.

Denver billionaire and private investor Donald Sturm owns a 5.1% stake in Level 3 Communications, Inc., headquartered in Broomfield, Colorado. Sturm spent a considerable part of his career at Peter Kiewit Sons’, Inc., a construction, coal mining, and telecommunications company that spun off Level 3 Communications in 1998. Although he had retired from Peter Kiewit in 1991, he decided to hold a 5.1% stake in the newly created company. In May 1998 his stake was worth \$31 per share. In August of 2000, he sold 0.4% of his holdings for approximately \$87 per share.

Robert Coates, founder, chairman, and CEO of Management Insights, Inc., a management consulting firm, doubled his stake in Inprise Corporation to 6.3% in April 1999. Inprise is the owner of the Borland family of embedded database tools. Robert Coates’ block purchase is a good example of the power that a non-managerial shareholder, although the owner of only a 6.3% stake, can have. In his

initial form SC 13D filed on April 9th, 1999, he notes under purpose of transaction the following: *“The market value of the common stock of Inprise Corporation has been steadily decreasing. Recently, the Chief Executive Officer and Chief Financial Officer have resigned from Inprise Corporation. As an interested and significant shareholder, Robert Coates and Management Insights, Inc. will take all actions necessary as may be deemed appropriate to protect their investment in Inprise.”*

In fact, Robert Coates put up a shareholder proposal critical of management for the 1999 annual meeting. This triggered a response from the corporation, and in an amended form SC 13D filed May 17th, 1999, it is reported that Robert Coates entered into an agreement with Inprise Corp. The agreement specifies that Coates is put up as a nominee for the board of directors. In exchange, he withdraws the earlier stockholder proposal and contractually agrees to the following: *“The members of the Coates Group and their respective Affiliates and Associates and the Company shall publicly support and recommend that the Company's stockholders vote for the election of each of the 1999 Nominees at the 1999 Annual Meeting and for each of the other matters being presented by the Board for a vote of stockholders at the 1999 Annual Meeting.”*

Finally, in early 2000, Robert Coates stepped down as a board member to protest the proposed acquisition of Inprise by Corel Corporation. Coates felt that the terms of the merger were not advantageous to Inprise shareholders, and actively fought the merger. Three months later, the merger fell through.

For some blocks, it is not obvious why it was bought or is maintained. For example, Microsoft-founder Bill Gates owns a 5.3% stake in Avista Corporation (formerly Washington Water and Power). Minnesota-based entrepreneur Richard Born, founder and CEO of Born Information Services, Inc., bought in 2000 a 5.8% stake in cross-town firm Analysts International group, which could be deemed a competitor. Born's initial statement of ownership indicates that he bought the block for investment purposes. He has neither sought nor obtained a board seat, nor did the two firms ever enter into merger talks.

Appendix B: Variable definitions

Variable	Definition
Assets (log)	Log of total assets (Compustat item 6)
Average share price	Mean daily closing share price previous year
Cash holdings	Cash and short-term investments (item 1) divided by lagged net property, plant, and equipment (item 8)
CEO incentive pay	Fraction of total CEO pay (Execucomp item TDC1) that is paid in either stock or options
CEO pay (log)	Log of the sum of cash salary, cash bonus, and the Black-Scholes value of options granted during a fiscal year to the CEO (Execucomp item TDC1)
Dividend yield	Ratio of the sum of common dividends (item 21) and preferred dividends (item 19) over book value of common equity (item 60)
Dividend payer	Indicator variable that is equal to 1 if the sum of common dividends (item 21) and preferred dividends (item 19) > 0, and 0 otherwise
Firm age	Log of months since first listing on stock exchange
Illiquidity	Yearly average of absolute daily return divided by daily dollar volume, as calculated by Amihud (2002). Data come from CRSP
Industry fixed effects	The 10 Fama-French industries, FF10
Institutional block ownership	Aggregate ownership by institutions, counting only blocks of 5% or more of common equity
Investment	Capital expenditures (item 128) divided by lagged net property, plant, and equipment (item 8)
Lagged return	Stock market return over the previous fiscal year
Leverage	Long-term debt (item 9) plus current liabilities (item 34) divided by long-term debt plus current liabilities plus book value of common equity (item 60) (book leverage) or divided by long-term debt plus current liabilities plus market value of common equity (market leverage)
Market capitalization	Number of shares outstanding times calendar year closing price (item 25 x item 199)
Nasdaq listing	Dummy variable equal to one for firms listed on NASDAQ, and zero otherwise
Number of outsiders on board	Number of non-affiliated directors on firm's board, from the IRRC director database
Payout ratio	Common dividends (item 21) plus dollar expenditures on the purchase of common and preferred stock (item 115) minus the reduction in the value of the net number of preferred shares outstanding (changes in item 56) divided by EBITDA
Payout yield	Common dividends (item 21) plus dollar expenditures on the purchase of common and preferred stock (item 115) minus the reduction in the value of the net number of preferred shares outstanding (changes in item 56) divided by book value of assets
Return on assets	Ratio of EBITDA (item 18) divided by lagged total assets (item 6)
Return on sales	Ratio of EBITDA (item 18) divided by lagged total sales (item 12)
Sales growth	Net sales (item 12) divided by lagged net sales, minus 1
S, G & A	Selling, general, and administrative expenses (item 189) divided by sales (item 12)
Trading volume	Yearly average of number of shares traded per month divided by total shares outstanding
Volatility	Annualized standard deviation of monthly share price, calculated over previous 24 months

Figure 1: The geographic distribution of high net worth individuals

The figure shows estimates of the number of high net worth individuals per Compustat firm headquarter for each U.S. state. The wealth data are from the Internal Revenue Service’s Statistics of Income (SOI) data which report the estimated number of individuals with a high net worth in 1995. The data for firms’ headquarters are from Compustat in 1995.

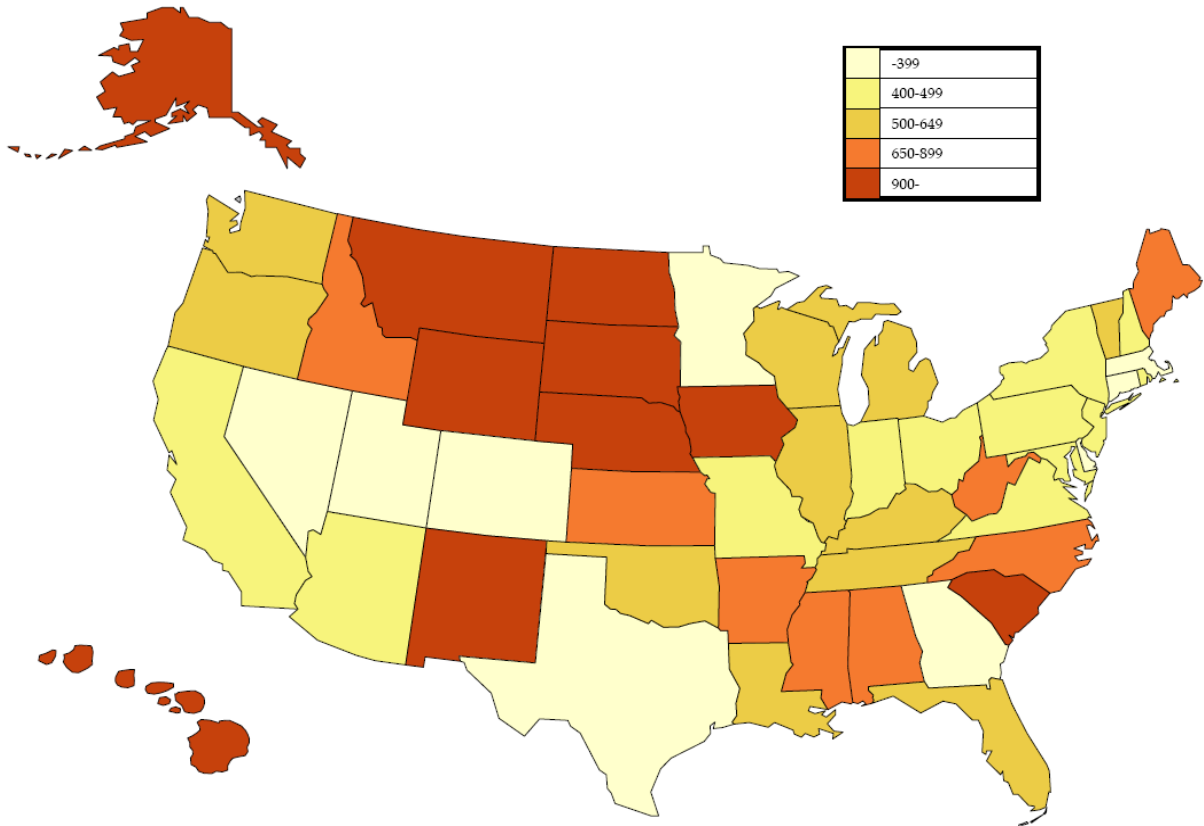


Figure 2: Actual versus predicted number of individuals on the Forbes 400 list

The figure plots state by state, the actual number of individuals on the *Forbes* list of the wealthiest Americans (in 1998) against the predicted number of individuals. We use the wealth data from the Internal Revenue Service's Statistics of Income (SOI) database (1998) and the power law distribution with a parameter of 1.36 estimated by Levy and Solomon (1997) to predict the number of individuals on the list (minimum wealth required to be on the Forbes 1998 list was \$430 million). The correlation coefficient between actual and predicted number of ultra-wealthy individuals is 0.89.

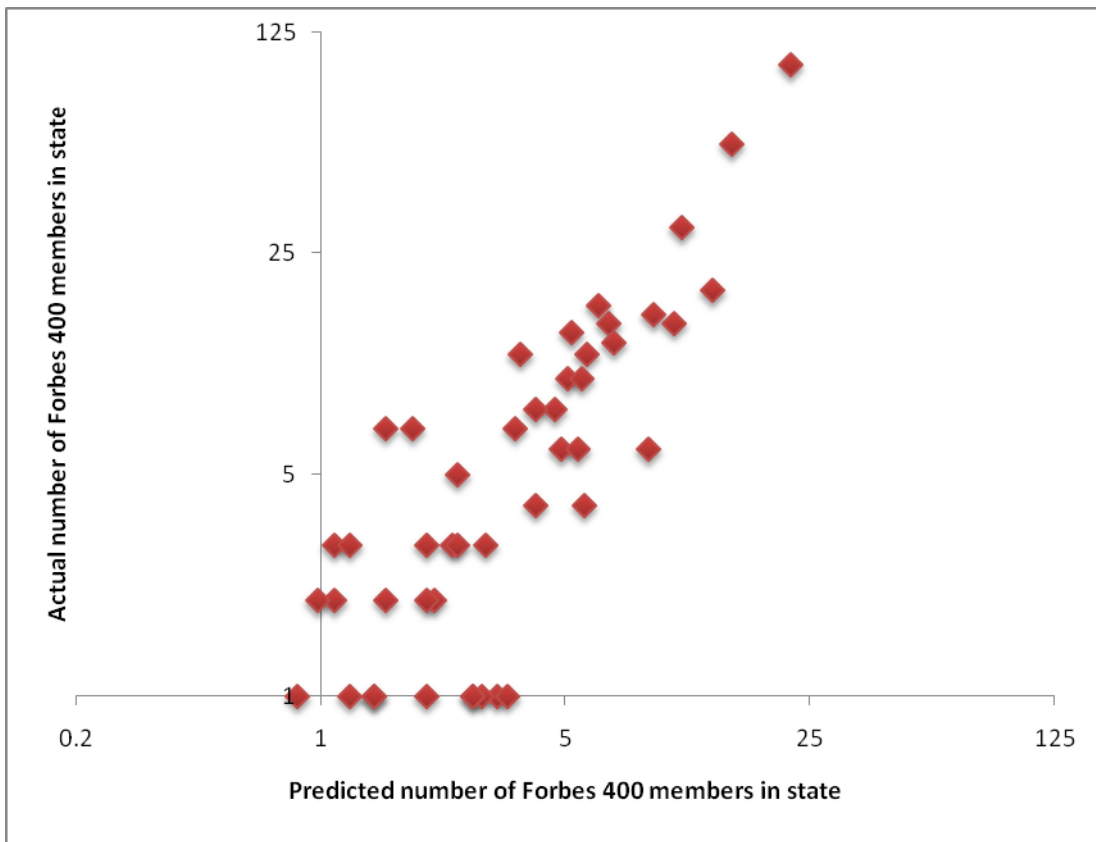


Figure 3: Frequency of individual blocks by U.S. state

The figure plots the frequency of large, individually held, non-managerial blocks in each state (across all sample years) against the density of high net worth individuals in 1995. Each circle represents one state. The area of a circle is proportional to the number of firms in the state. The dashed line represents the regression line from a weighted least squares regression (weights are the number of firms). This regression has an estimated intercept of 0.031 (standard error 0.026), an estimated slope of 0.190 (standard error 0.051). The slope is significantly different from zero at the 0.1%-level. The adjusted R-squared is 0.236.

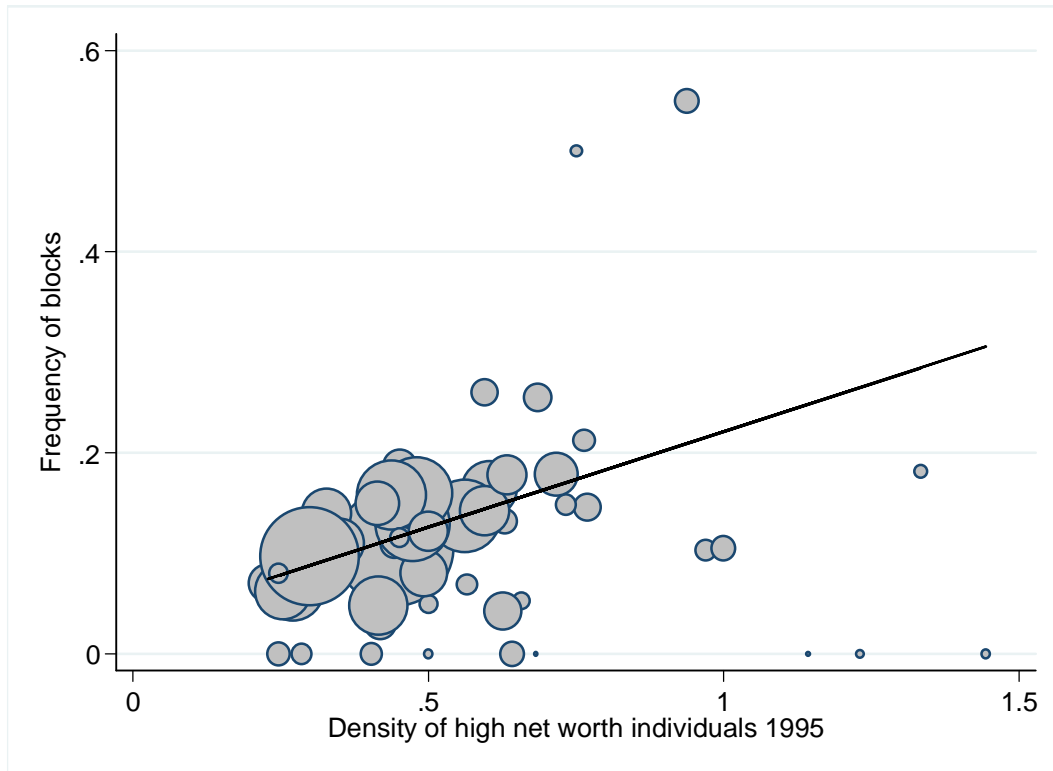


Table 1: Distance between large shareholders and firms' headquarters

The table shows summary statistics for the distance in miles between the location of a blockholder and the headquarters of the firm in which the block is held for four different types of large shareholders. The four types of blockholders are non-managerial individuals, managers (both current and former), mutual funds, and money managers. The summary statistics are based on firm-year observations. If a firm has multiple blockholders of the same category in a given year, the distance in miles is calculated as the average distance between firm headquarters and the locations of the blockholders. Firm-years are only included in the summary statistics below if there is at least one blockholder of the respective category. The sample is approximately 6,000 firm-year observations of non-dual class S&P 1500 firms during the period 1996-2001, excluding financial firms and utilities. Blockholders are entities that own at least 5% of the outstanding shares. The blockholder zip codes are hand-collected from proxy statements. The firms' zip codes come from the *Execucomp* database, the *Compact Disclosure* discs, or are hand-collected from proxy statements. The distances are calculated for U.S. based blockholders (excluding Alaska and Hawaii) only, using latitude and longitude data from the *U.S. Census Bureau's Gazetteer Place and Zip Code Database*. The indicator variable "Same State" is equal to one if the blockholder and firm headquarters are located in the same state, and zero otherwise.

Type of blockholder	N	Distance in miles		Same State
		Mean	Median	
Non-managerial individual	715	385.4	28.5	61.3%
Managers	1,329	52.1	0	93.2%
Mutual funds	4,102	1,158.5	1,121.38	9.5%
Money managers	1,843	1,061.9	871.4	8.6%

Table 2: Summary statistics

The table shows sample summary statistics. Panel A shows the number of observations and frequency for four types of blockholders: Non-managerial individuals, managers (both current and former), mutual funds, and money managers. Panel B shows mean, median and standard deviation for all other variables. The sample size (N) varies across variables because of data availability and because outliers are excluded from regressions. All variables are defined in the Data Appendix. The sample is non-dual class S&P 1500 firms during the period 1996-2001.

Panel A: Blockholders				
Type of blockholder	N	Frequency		
Non-managerial individuals	5984	0.118		
Management	5984	0.219		
Mutual funds	5984	0.677		
Money managers	5984	0.305		

Panel B: Regression variables				
Variables	N	Mean	Median	Standard deviation
Firm age	5979	24.0	20.1	19.13
ROA	5787	0.047	0.055	0.104
ROS	5784	0.035	0.047	0.166
S,G&A	5318	0.251	0.213	0.179
Investment	5632	0.282	0.221	0.221
Leverage, BV	5751	0.252	0.246	0.182
Leverage, long-term only, BV	5755	0.210	0.195	0.169
Leverage, MV	5744	0.278	0.220	0.277
Total payout ratio	5377	0.257	0.202	0.355
Cash holdings	5771	0.146	0.043	0.270
Number of outsiders on board	5584	7.440	7.000	2.812
CEO pay (log)	5092	7.820	7.775	1.188
CEO incentive pay	5092	0.512	0.553	0.285
Trading volume	5966	0.006	0.004	0.007
Annualized stock return	5951	0.054	0.089	0.475
Volatility	5893	0.450	0.392	0.248
Illiquidity	5890	0.031	0.004	0.088

Table 3: Estimated wealth distributions

The table shows estimates of the number and wealth of high net worth individuals for four U.S. states (Oregon, Georgia, New York, California). The wealth estimates are based on data from the Internal Revenue Service's Statistics of Income (SOI) data which reports the estimated number of individuals with a high net worth (\$1M and up) in 1998. The wealth of subsamples of these individuals (100 richest, number with \$100 million or more in net wealth) is estimated assuming individual net wealth follows a power law distribution with $W_n = A \times n^{-1/1.36}$ for wealth W , and individual rank n . A is a constant. The parameter of 1.36 is estimated by Levy and Solomon (1997) (LS (1997) below). The 1999 end-of-year market value and number of listed firms are constructed from Compustat data. The sample is non-dual class S&P 1500 firms.

	Oregon	Georgia	New York	California	Data source
Number of individuals with a high net worth	27,000	64,000	243,000	412,000	SOI (1995)
Wealth of 100 richest individuals (\$M)	17,367	32,759	87,373	128,817	Estimate using
their wealth as fraction of state	140%	8%	5%	9%	LS (1997)
market value 1999					
Number of individuals with \$100M in wealth	51	122	463	785	Estimate using
their wealth as fraction of state	110%	9%	8%	18%	LS (1997)
market value 1999					
Number of sample firms in state (1999)	7	26	70	153	Compustat
Total market value of equity in 1999 (\$M)	12,420	392,538	1,827,770	1,420,483	Compustat
Median market value in 1999 \times 5% (\$M)	43	70	171	58	Compustat
Average market value in 1999 \times 5% (\$M)	89	755	1,015	597	Compustat
Largest market value in 1999 \times 5% (\$M)	259	7,200	17,500	9,700	Compustat
No. of individuals with enough wealth to buy 5% of:					Estimates using LS (1997)
median firm in state	162	198	223	1,659	
average firm in state	61	8	20	69	
2 nd largest firm in state	27	0.6	0.6	1.9	
largest firm in state	14.1	0.4	0.4	1.6	

Table 4: First stage results

The table shows results from linear probability models and probit regressions of three large shareholder variables on measures of the density of high net worth individuals in a state in 1995 and controls. The dependent variables are defined as follows. The *large shareholder dummy* variable is equal to one if an individual non-managerial large shareholder is present in the firm, and zero otherwise (columns 1 through 4). *Block ownership* is the percentage ownership by individual non-managerial blockholders (column 5). The *managerial block dummy* variable is equal to one if a large individual managerial shareholder is present in the firm, and zero otherwise (column 6). The *Fund / MM block dummy* variable is equal to one if a mutual fund or money manager is present in a firm, and zero otherwise (column 7). The sample is non-dual class S&P 1500 firms during the period 1996-2001. Blockholders are entities that own at least 5% of outstanding shares. All other variables are defined in the Data Appendix. Robust standard errors (shown in parentheses) are clustered by state. Statistical significance at the 1%, 5%, and 10% level is indicated by ***, **, and *, respectively.

Table 4 – cont'd

Dependent variable	Large shareholder dummy				Block ownership	Managerial block dummy	Fund / MM block dummy
	(1)	(2)	(3)	(4)	(5)	(6)	(7)
Regression type	OLS	Probit	Probit	Probit	OLS	Probit	Probit
Density of high net worth individuals, 1995	0.190*** (0.060)	0.892*** (0.250)	0.728*** (0.281)		2.29*** (0.560)	0.281 (0.216)	-0.291 (0.212)
Wealth per listed firm, 1995				0.513*** (0.174)			
Lagged return			-0.071 (0.052)	-0.071 (0.052)	-0.129 (0.212)	0.008 (0.036)	-0.116*** (0.052)
Log (assets, book value)			-0.159*** (0.039)	-0.158*** (0.039)	-0.304*** (0.069)	-0.221*** (0.042)	-0.142*** (0.027)
Age			0.016* (0.009)	0.016* (0.009)	0.057** (0.015)	-0.007 (0.009)	-0.009 (0.006)
Age squared			-0.00021 (0.00013)	-0.00020 (0.00013)	-0.0008*** (0.0002)	-0.0002 (0.0002)	0.00008 (0.0007)
Sales growth			0.132* (0.070)	0.124* (0.070)	0.182 (0.315)	0.037 (0.065)	0.131** (0.063)
Return on assets (ROA)			-0.250 (0.309)	-0.224 (0.311)	-0.903 (0.870)	0.690*** (0.215)	-0.526 (0.447)
Year fixed effects	No	No	Yes	Yes	Yes	Yes	Yes
Industry fixed effects	No	No	Yes	Yes	Yes	Yes	Yes
R-squared	0.008	N/A	N/A	N/A	0.026	0.035	N/A
N	5,972	5,972	5,754	5,754	5,754	5,754	5,754

Table 5: First stage results with state-level controls

The table shows probit regressions of a blockholder indicator variable on the density of high net worth individuals in 1995 and other variables that vary at the state level. The dependent variable is equal to one if a large individual non-managerial shareholder is present in the firm, and zero otherwise. The sample is non-dual class S&P 1500 firms during the period 1996-2001. The state-level variables are defined in Section 3.3. Each specification includes the firm-level controls of Table 4, column 3. Robust standard errors (shown in parentheses) are clustered by state. Statistical significance at the 1%, 5%, and 10% level is indicated by ***, **, and *, respectively.

Dependent variable	(1)	(2)	(3)	(4)	(5)
Regression type	Probit	Probit	Probit	Probit	Probit
Density of high net worth individuals	0.709** (0.279)	0.725*** (0.281)	0.657** (0.283)	0.737*** (0.279)	0.626** (0.2683)
Population density	-0.005 (0.005)				-0.004 (0.004)
Corporate tax rate (state)		0.011 (0.014)			0.014 (0.014)
Per capita income			-0.001 (0.002)		-0.001 (0.002)
Educational attainment - % high school graduates				0.003 (0.005)	0.003 (0.005)
Firm control variables	Yes	Yes	Yes	Yes	Yes
Year fixed effects	Yes	Yes	Yes	Yes	Yes
Industry fixed effects	Yes	Yes	Yes	Yes	Yes
N	5,754	5,754	5,754	5,754	5,754

Table 6: Operating performance and large shareholders

The table shows results from the second stage regression of operating performance measures on the instrumented large shareholder dummy and control variables. The large shareholder dummy is equal to one if a large individual non-managerial shareholder is present in the firm, and zero otherwise. It is instrumented in the first stage regression with the density of high net worth individuals in 1995. The last two rows of the table show the coefficient and standard error for the density of high net worth individuals from the first stage probit regression of the large shareholder dummy on the density measure and controls. $\hat{\rho}$ is the estimated correlation between first and second stage error terms. The sample is non-dual class S&P 1500 firms during the period 1996-2001. All other variables are defined in the Data Appendix. Robust standard errors (shown in parentheses) are clustered by state. Statistical significance at the 1%, 5%, and 10% level is indicated by ***, **, and *, respectively.

Dependent variable	ROA	ROS	S, G & A	Investment
	(1)	(2)	(3)	(4)
Second stage results				
Large shareholder dummy	0.0371*	0.0782***	-0.0625	-0.0704**
<i>[instrumented]</i>	(0.0221)	(0.0251)	(0.0390)	(0.0297)
Age	0.0006*	0.0004	-0.002	-0.0036***
	(0.0003)	(0.0003)	(0.001)	(0.001)
Age squared (/1000)	-0.008*	-0.008**	0.00002	0.00004***
	(0.004)	(0.004)	(0.00001)	(0.00001)
Assets, log	-0.0024	0.0048*	-0.015***	-0.024***
	(0.0023)	(0.0026)	(0.004)	(0.0036)
Sales growth	0.0163***	0.0025	0.010	0.106***
	(0.0040)	(0.0053)	(0.011)	(0.027)
Dividend payer dummy	0.0008	-0.002		
	(0.0040)	(0.005)		
Dividend yield	0.327***	0.219***		
	(0.044)	(0.057)		
Lagged ROA			0.038	0.372***
			(0.039)	(0.046)
Lagged return	0.033***	0.027***	0.009	0.059***
	(0.003)	(0.004)	(0.007)	(0.007)
Year fixed effects	Yes	Yes	Yes	Yes
Industry fixed effects	Yes	Yes	Yes	Yes
Estimated $\hat{\rho}$	-0.250	-0.449	0.282	0.184
Chi-square test ($\hat{\rho} = 0$)	3.01*	10.73***	3.61*	6.83***
N	5,503	5,509	5,172	5,485
First stage results				
Density of high net worth individuals	0.722***	0.725***	1.013***	0.887***
	(0.281)	(0.257)	(0.360)	(0.313)

Table 7: Capital structure and large shareholders

The table shows results from the second stage regression of three measures of leverage on a large shareholder dummy and control variables. The large shareholder dummy is equal to one if a large individual non-managerial shareholder is present in the firm, and zero otherwise. It is instrumented in the first stage regression with the density of high net worth individuals in 1995. The last two rows of the table show the coefficient and standard error for the density of high net worth individuals from the first stage probit regression of the large shareholder dummy on the density measure and controls. $\hat{\rho}$ is the estimated correlation between first and second stage error terms. The sample is non-dual class S&P 1500 firms during the period 1996-2001. All other variables are defined in the Data Appendix. Robust standard errors (shown in parentheses) are clustered by state. Statistical significance at the 1%, 5%, and 10% level is indicated by ***, **, and *, respectively.

Dependent variable	Leverage, BV	Long term leverage, BV	Leverage, MV
	(1)	(2)	(3)
Second stage results			
Large shareholder dummy	-0.0323	-0.0372*	-0.0330
<i>[instrumented]</i>	(0.0246)	(0.0222)	(0.0436)
Controls (first, second stage)	Age, age squared, log (assets), sales growth, ROA, dummy for dividend payers, lagged return		
Year fixed effects	Yes	Yes	Yes
Industry fixed effects	Yes	Yes	Yes
Estimated $\hat{\rho}$	0.142	0.138	0.029
Chi-square test ($\hat{\rho} = 0$)	5.42*	5.06**	1.13
N	5,654	5,657	5,555
First stage results			
Density of high net worth individuals	0.692**	0.684**	0.747***
	(0.286)	(0.284)	(0.292)

Table 8: Payout policy, cash holdings, and large shareholders

The table shows results from the second stage regression of five measures of payout policy on a large shareholder dummy and control variables. The large shareholder dummy is equal to one if a large individual non-managerial shareholder is present in the firm, and zero otherwise. It is instrumented in the first stage regression with the density of high net worth individuals in 1995. The last two rows of the table show the coefficient and standard error for the density of high net worth individuals from the first stage probit regression of the large shareholder dummy on the density measure and controls. $\hat{\rho}$ is the estimated correlation between first and second stage error terms. The sample is non-dual class S&P 1500 firms during the period 1996-2001. All other variables are defined in the Data Appendix. Robust standard errors (shown in parentheses) are clustered by state. Statistical significance at the 1%, 5%, and 10% level is indicated by ***, **, and *, respectively.

Dependent variable	Total payout ratio	Cash holdings
	(1)	(2)
Second stage results		
Large shareholder dummy	0.265***	-0.0488*
<i>[instrumented]</i>	(0.091)	(0.0258)
Controls (first, second stage)	Age, age squared, log (assets), sales growth, ROA, lagged return	
Year fixed effects	Yes	Yes
Industry fixed effects	Yes	Yes
Estimated $\hat{\rho}$	-0.336	0.120
Chi-square test ($\hat{\rho} = 0$)	5.58**	2.30
N	5,178	5,461
First stage results		
Density of high net worth individuals	0.624**	0.782***
	(0.269)	(0.284)

Table 9: Managerial compensation, governance mechanisms, and large shareholders

The table shows results from the second stage regression of CEO pay and board structure on a large shareholder dummy and control variables. The large shareholder dummy is equal to one if a large individual non-managerial shareholder is present in the firm, and zero otherwise. It is instrumented in the first stage regression with the density of high net worth individuals in 1995. The last two rows of the table show the coefficient and standard error for the density of high net worth individuals from the first stage probit regression of the large shareholder dummy on the density measure and controls. $\hat{\rho}$ is the estimated correlation between first and second stage error terms. The sample is non-dual class S&P 1500 firms during the period 1996-2001. All other variables are defined in the Data Appendix. Robust standard errors (shown in parentheses) are clustered by state. Statistical significance at the 1%, 5%, and 10% level is indicated by ***, **, and *, respectively.

Dependent variable	CEO pay (log)	CEO incentive pay	No. of outsiders on board
	(1)	(2)	(3)
Second stage results			
Large shareholder dummy	-0.960***	-0.446***	2.112***
<i>[instrumented]</i>	(0.192)	(0.048)	(0.156)
Controls (first, second stage)	Age, age squared, log (assets), sales growth, ROA, dummy for dividend payers, lagged return)		
Year fixed effects	Yes	Yes	Yes
Industry fixed effects	Yes	Yes	Yes
Estimated $\hat{\rho}$	0.540	0.741	-0.368
Chi-square test ($\hat{\rho} = 0$)	15.76***	56.14***	1.83
N	4,936	4,947	5,368
First stage results			
Density of high net worth individuals	0.922***	0.620**	0.480***
	(0.308)	(0.278)	(0.174)

Table 10: Liquidity and large shareholders

The table shows results from the second stage regression of measures of liquidity on a large shareholder dummy and control variables. The large shareholder dummy is equal to one if a large individual non-managerial shareholder is present in the firm, and zero otherwise. It is instrumented in the first stage regression with the density of high net worth individuals in 1995. The last two rows of the table show the coefficient and standard error for the density of high net worth individuals from the first stage probit regression of the large shareholder dummy on the density measure and controls. $\hat{\rho}$ is the estimated correlation between first and second stage error terms. The sample is non-dual class S&P 1500 firms during the period 1996-2001. All other variables are defined in the Data Appendix. Robust standard errors (shown in parentheses) are clustered by state. Statistical significance at the 1%, 5%, and 10% level is indicated by ***, **, and *, respectively.

Dependent variable	Trading volume	Illiquidity	Stock price volatility
	(1)	(2)	(3)
Second stage results			
Large shareholder dummy	-0.0039***	0.124***	-0.075***
<i>[instrumented]</i>	(0.0009)	(0.017)	(0.028)
Controls (first, second stage)	Age, age squared, log (assets), sales growth, lagged return, ROA, dummy for dividend payers, institutional block ownership, dummy of Nasdaq membership, average share price		
Other controls	Volatility	Trading volume	Volatility, Trading volume
Year fixed effects	Yes	Yes	Yes
Industry fixed effects	Yes	Yes	Yes
Estimated $\hat{\rho}$	0.364	-0.667	0.290
Chi-square test ($\hat{\rho} = 0$)	145.11***	22.09***	11.15***
N	5,636	5,633	5,614
First stage results			
Density of high net worth individuals	0.824***	0.388*	0.578**
	(0.277)	(0.218)	(0.273)

Table 11: Comparison of instrumental variable and OLS estimates

The table compares instrumental variable (IV) and OLS estimates of non-managerial individual blockholder effects on firms. Each reported number corresponds to a separately estimated coefficient on the large shareholder dummy variable for the different dependent variables. We use the same control variables in each regression as those in Tables 6-10. The IV column shows the estimated second stage coefficients on the instrumented large non-managerial individual shareholder dummy variable. It is instrumented with the density of high net worth individuals in 1995. The OLS column shows the corresponding OLS estimate of the large non-managerial individual shareholder dummy variable. All variables are defined in the Data Appendix. Robust standard errors (shown in parentheses) are clustered by state. Statistical significance at the 1%, 5%, and 10% level is indicated by ***, **, and *, respectively.

Dependent variables	IV	OLS
ROA	0.0362*	0.0138
ROS	0.0782***	-0.0021
S, G & A	-0.0624	0.0082
Investment	-0.0704*	-0.0071
Leverage, BV	-0.0323	0.0094
Long term leverage, BV	-0.0372*	-0.0014
Leverage, MV	-0.0330	0.0008
Total payout ratio	0.265***	0.058**
Cash holdings	-0.0488*	-0.0160*
CEO pay (log)	-0.960***	-0.193***
CEO incentive pay (log)	-0.446***	-0.049***
Number of outside directors	2.112***	0.017
Trading volume	-0.0039***	-0.0014***
Volatility	-0.075***	0.0072
Illiquidity	0.124**	0.0014