

The surge in late stage VC funding with non-traditional investor participation: analyzing the impact on post-IPO gains

Hadi Zaklouta

The Leonard N. Stern School of Business
Glucksman Institute for Research in Securities Markets
Faculty Advisor: Professor Alexander Ljungqvist
April 1, 2014

EXECUTIVE SUMMARY

With late stage venture capital rounds recently on the rise, and with the increased involvement of non-traditional investors, typically hedge funds or mutual funds, the discussion in the venture capital community centers around whether these two related trends could impact valuations and ultimately post-IPO stock performance of the respective portfolio companies.

The hypothesis presented in this paper is that non-traditional investor participation in pre-IPO funding rounds may ultimately lead to higher valuations, which in turn could suppress post-IPO gains to pre-IPO investors. One may also expect the increase in non-traditional investor to impact public investor gains – this hypothesis is also explored in detail.

This study confirms the cited trends and tests the hypothesis on the correlation between post-IPO performance and non-traditional investor involvement. The 1-day and 3-day returns of companies that went public in the time period of 2004-2013, and across multiple industries, were analyzed against both funding-round characteristics and participating investor characteristics. The same analysis was also done on a subset of tech related companies, being a segment that has become particularly attractive for hedge funds and mutual funds.

The regression results indicate no statistically significant correlations. That is to say that non-traditional investor participation in the investment history of companies going public, and more particularly participation in the late funding stages, is not likely to have an effect on the performance of the company's stock in the period immediately after the IPO. This study was not without data limitations, and while it offers initial insights on this topic, further analysis on a larger dataset is recommended.

I. INTRODUCTION

The post-IPO stock performance of venture-backed companies has been studied extensively in an attempt to determine the key factors explaining cross-company performance differences. The most prominent studies analyze stock performance as a function of investing firm characteristics. Megginson and Weiss (1991) finds that VC-backed IPOs have higher returns than those without VC involvement. More recently, both Sorenson (2007) and Krishnan et al. (2009) analyze post-IPO performance as a function of VC reputation and find a strong correlation indicating that VCs with strong reputation tend to select better performing firms. The strong interest in relating stock performance to investor characteristics is clear: understanding which investors achieve higher returns helps companies seeking to go public partner with the right investor and helps Limited Partners identify the highest yield funds for investment. Yet research in this field is still in its nascent phase and the impact of investor characteristics on returns is not fully understood. In this paper, I will examine the potential impact of a recent trend in the VC industry, the participation of non-traditional investors, on post-IPO returns.

The VC industry is currently experiencing fast growth, with US VC investments at their highest since 2001 and 2014 Q1 investments totaling \$10.7B (Garland, 2014). In particular, there is evidence of a sharp increase in later stage funding rounds, accounting for \$5.8B of funds raised in Q1 of 2014 and having grown ~30% since last year (ibid.). This surge appears to be the result of higher participation from non-traditional investors, notably hedge funds and mutual funds (CB Insights, March 2014), reminiscent of the expansion of the mezzanine investment market in the 1990s. Prominent examples of such investors include BlackRock and T. Rowe Price, which have increased their participation in late-stage tech funding from 1 and 6 investments respectively in 2011 to 8 and 16 investments in 2014 (CB Insights, March 2015).

CB Insights reports that non-traditional investors have contributed over \$2.4B in funding in 2013 alone (CB Insights, January 2014). Their most common form of participation in funding-rounds has historically been to invest in the funds of VC firm, which in turn are used to invest in portfolio companies. Yet recently their participation model seems to be shifting towards more direct involvement, where they bypass the VC intermediaries altogether and invest directly in companies. Perhaps these non-traditional investors are gaining confidence and the skills needed in direct investing, or perhaps they are merely seeking new opportunities for high returns – there are many possible explanations. Logically, late stage rounds also appear to be the most appealing for these investors to target since companies at those stages are more mature and present lower risk.

The potential implications of this surge in late stage funding, and more specifically non-traditional investor involvement, are also being discussed in the VC community. This trend may have an impact on the realized returns of all players in the value chain including post-IPO investors in public markets.

One hypothesis is that an increased participation of hedge funds and mutual funds may drive up valuations in the rounds close to the IPO as these highly diversified investors have lower return requirements and, perhaps, are more likely to overpay. This is an explanation offered by many Venture Capitalists, including Glenn Solomon from GGV Capital who claims that "as public investors, their return expectations are lower than those of VC's,.. so ...they are also often willing to pay higher prices and are less focused on deal terms than VC's" (TechCrunch, February 2014). A recent Bloomberg article attributes a current surge in tech startup valuations to a "flood of money from unconventional sources", where "hedge funds and mutual funds are paying 15 to 18 times projected sales for the year ahead...[compared] with 10 to

12 times five years ago (Carey and Saitto, 2015). If this observation is true and widespread, there could be a negative impact on their post-IPO prices. The argument here is that one may observe a significant drop from the IPO offer price to the IPO opening price (first trade price) with the former perhaps influenced by high pre-IPO valuations, a prevalent environment of overvaluations in the industry or the perceived surge in demand of non-traditional investor at the IPO. Meanwhile, the opening price reflects the market's valuation of the company. This argument would also be counter to the common trend that IPOs tend to be underpriced on average, with stock market prices rising above the IPO offer price (Lowry et al., 2010). A counterargument can be made that IPO investors may disagree with pre-IPO valuations which would exert pressure to suppress the IPO offer price close to market equilibrium and from which the opening price may move in any direction.

Another hypothesis with regard to the effect of non-traditional investor involvement on post-IPO returns is that their involvement may affect public investor returns. One discussed possibility is that these investors are making more calculated and risk-averse investment decisions than VC's, overall betting on healthier companies where stock prices are more likely to increase beyond the IPO opening price. A contrasting argument is that non-traditional investors may be more inclined to push for earlier IPOs where companies may not yet be ready to go public. Hence, companies with non-traditional investor participation may exhibit poorer stock performance in the days immediately following the IPO.

The rest of the paper is organized as follows. In the first section I set up the data and methodology for my analysis. I then present the basic statistics and baseline findings from the analyzed dataset. Subsequently, I conduct the regression analyses of return measures against a set of company and investor characteristics. Finally, research limitations are discussed in detail.

II. DATA & METHODOLOGY

The hypotheses will be tested by analyzing data from venture-backed companies that have exited through an IPO in the time period 2004-2013. Note that 2014 was excluded as data for 2014 are not yet available in CRSP. Although the cited surge in non-traditional investor participation seems to have been observed very recently and reported in early 2014, a statistical examination of past VC deals may shed some light on its significance and implications. The analysis will span multiple industries and will include regressions of post-IPO returns against factors relating to the characteristics of participating investors and the funding rounds the companies received.

IPO Sample Construction

Core to this research is adequate data on post-IPO public investor returns and company funding rounds. For the former, I use post-IPO stock price data from the Center for Research in Security Prices (CRSP). For the latter, the Thomson Reuters Securities Data Company database (SDC Platinum) provides company funding history data, including investor names and other characteristics. Capital IQ is used to supplement SDC Platinum data on investor characteristics to more accurately identify non-traditional investors.

SDC Platinum contains data on venture backed deals, an industry that is generally difficult to get reliable information on. From SDC, I obtain data on 1648 venture backed companies that have exited via IPO in the years 2004-2014 (the majority based in the US). This dataset includes companies that are still listed and companies ultimately delisted by being taken private or being acquired. Available information on investor characteristics includes investor participant name, type, preferred sector and stage involvement, size and number of funds raised,

number of companies invested in, and the minimum, maximum and average investment size. Some variables in the data were excluded from the regression due to incompleteness (see Table 2 for the list of variables used). Meanwhile, data on company and investment rounds include company sector classification, date and investment amount of all funding rounds, the name of participants in all rounds and the IPO date.

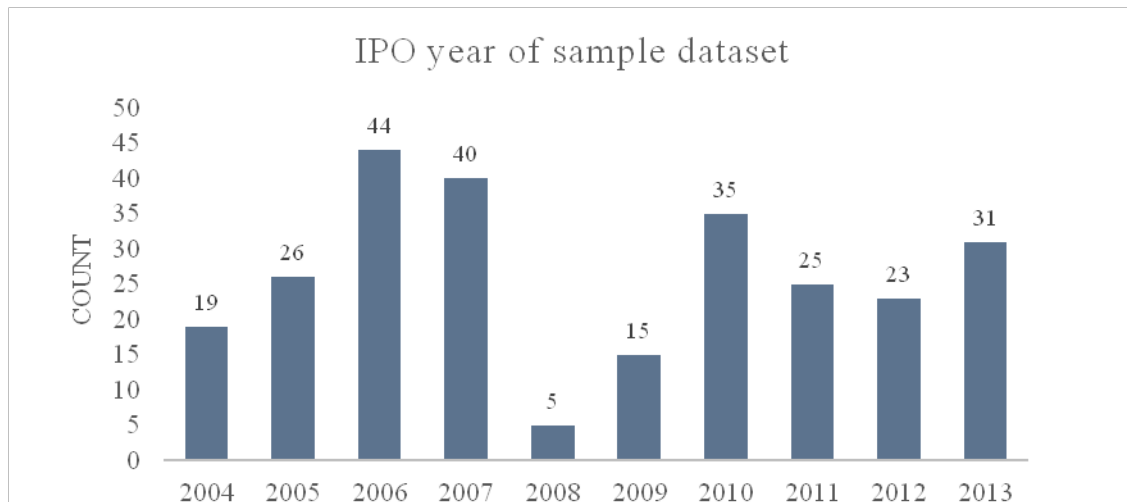
While the SDC data are likely to be the best source of data on VC deals, they are not complete. The identity of investors as well as the round and individual investor amounts are often not disclosed. Added to this limitation is the lack of share price data for 2014 in CRSP. The combined limitations of SDC and CRSP reduce the sample size to 591 companies, of which 263 company entries have complete information on all participating investors. To ensure the highest possible accuracy given data constraints, my analysis focuses on these 263 companies (see Appendix for a list of my sample companies). Note that even here, data on investment amounts are not always complete.

Table 1: Analyzed sample characteristics

Industry¹	Samples
Medical/Health/Life Science	34
Biotechnology	23
Communications and Media	17
Computer Related	48
Non-High-Technology	122
Semiconductors/Other Elect	19
Total	263

¹ The breakdown follows the industry classification provided by SDC Platinum, but amended to correct some mis-categorizations

Exhibit 1: IPO year distribution of analyzed sample set



Measuring Post-IPO Performance

For the purpose of my analyses, I use two measures of post-IPO performance. To analyze the impact of non-traditional investor participation on pre-IPO investor gains, I define post-IPO performance as the change in price from the offer price to the closing price on Day 1 of trading, allowing for the price to adjust to the equilibrium, efficient market level. Meanwhile, to test whether non-traditional investor participation in pre-IPO rounds may affect public investor gains, I define post-IPO performance as the change in price from the IPO opening price (first trade price) to the closing price on Day 3. To prevent skewness and outliers from influencing the results, the returns are defined mathematically as the difference between the natural logarithms of the two prices, $\ln(P_T) - \ln(P_0)$, where P_0 refers to the offer or opening price, and $T=1$ or $T=3$, depending on what measure of return is being implemented. Furthermore, the return data need to be normalized for time related factors by normalizing against returns on the S&P 500, defined as $\ln(P_{S\&P,T}) - \ln(P_{S\&P,0})$. Overall the market-adjusted return is measured as $[\ln(P_T) - \ln(P_0)] - [\ln(P_{S\&P,T}) - \ln(P_{S\&P,0})]$. Note that the 3 day measurement period is arbitrary – there is no

evidence that results vary with another chosen time period as long as the IPO opening price and the trend thereafter is captured.

Regression variables

The variables against which the measures of return are regressed include both funding round and investor specific variables. These are summarized below.

Table 2: Regression variables summary

Variable	Type	Comments
Time to IPO (years)	Funding rounds	Time between first round and IPO date measured in years
Number of rounds	Funding rounds	Number of rounds before IPO
Average number of investors per round	Funding rounds	Average number of investor participants
Known total investment amount in company ²	Funding rounds	Total registered funds raised prior to IPO
Presence of bank affiliated investors	Investor	Binary variable to investor type participation
Presence of investment management firm	Investor	Binary variable to investor type participation
Presence of insurance firm affiliate	Investor	Binary variable to investor type participation
Presence of endowment, foundation or pension	Investor	Binary variable to investor type participation
Presence of NT investors	Investor	Binary variable to investor type participation
Presence of NT investors in late stage	Investor	Binary variable to investor type participation in rounds 3+ ³

² Not all round amount information was disclosed in SDC Platinum

³ The definition of late stage was set to rounds Series C +

The category ‘non-traditional investor’ consists of four subgroups of investors. These investors are classified as per SDC Platinum’s definition into the categories of ‘Investment Management Firms’, ‘Insurance Firm Affiliates’, ‘Endowments, Foundations and Pension Funds’, as well as ‘Bank Affiliates’. Note that the latter category was filtered using Capital IQ classifications to exclude the venture capital arms of banks.

III. RESULTS

The results are organized in two sections. In the first section, trends relating to venture capital investment rounds are analyzed. Specifically, two questions are answered: has there been a noticeable surge in late stage rounds in the period 2004-2013, and is there evidence to suggest an increased participation of non-traditional investors? In the second section, the relationship between the measures of post-IPO performance and the characteristics of funding round and participating investors are explored.

Trends in venture capital investments rounds

Data from SDC Platinum on 1648 venture backed IPOs demonstrates evolving exit characteristics. Average time to IPO has close to doubled to over 9 years in the period 2004-2014, while the average time between rounds has nearly halved to 8 months, suggesting that companies are receiving significantly more funding rounds before going public and marking an increase in late stage financing.

Exhibit 2: Average time to IPO in years 2004-2014

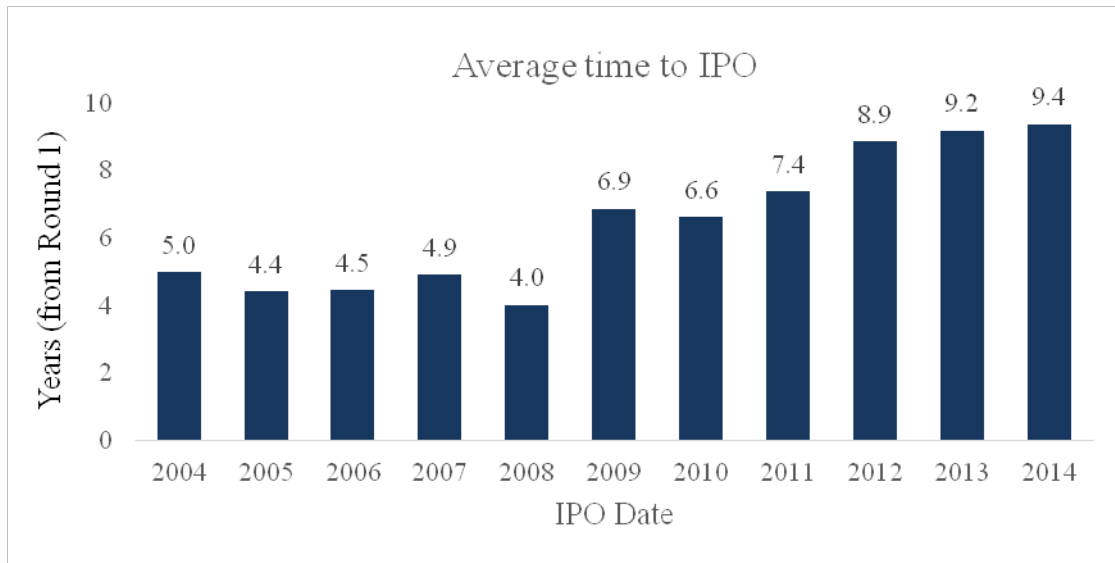
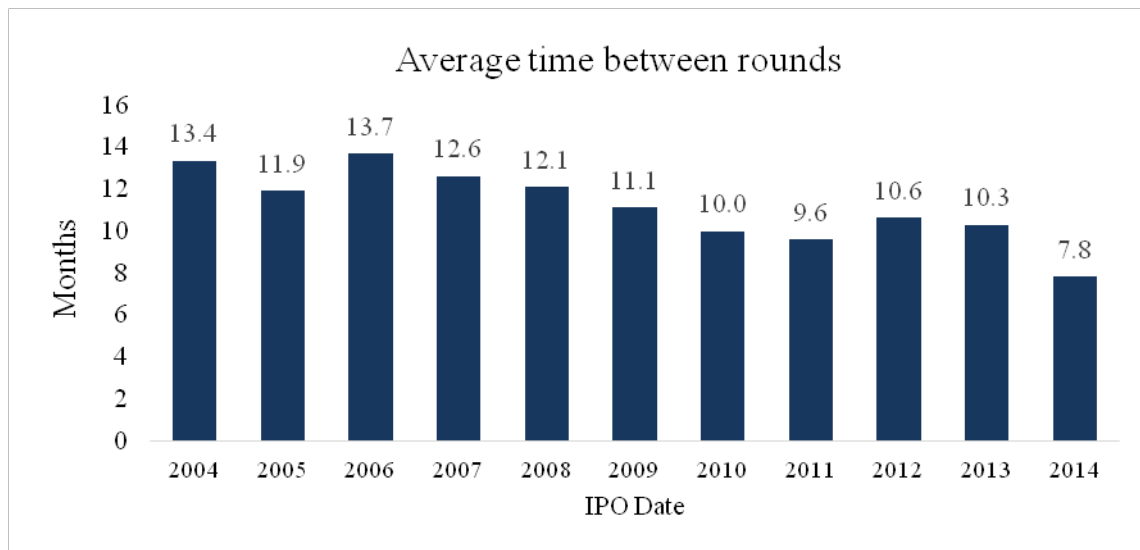
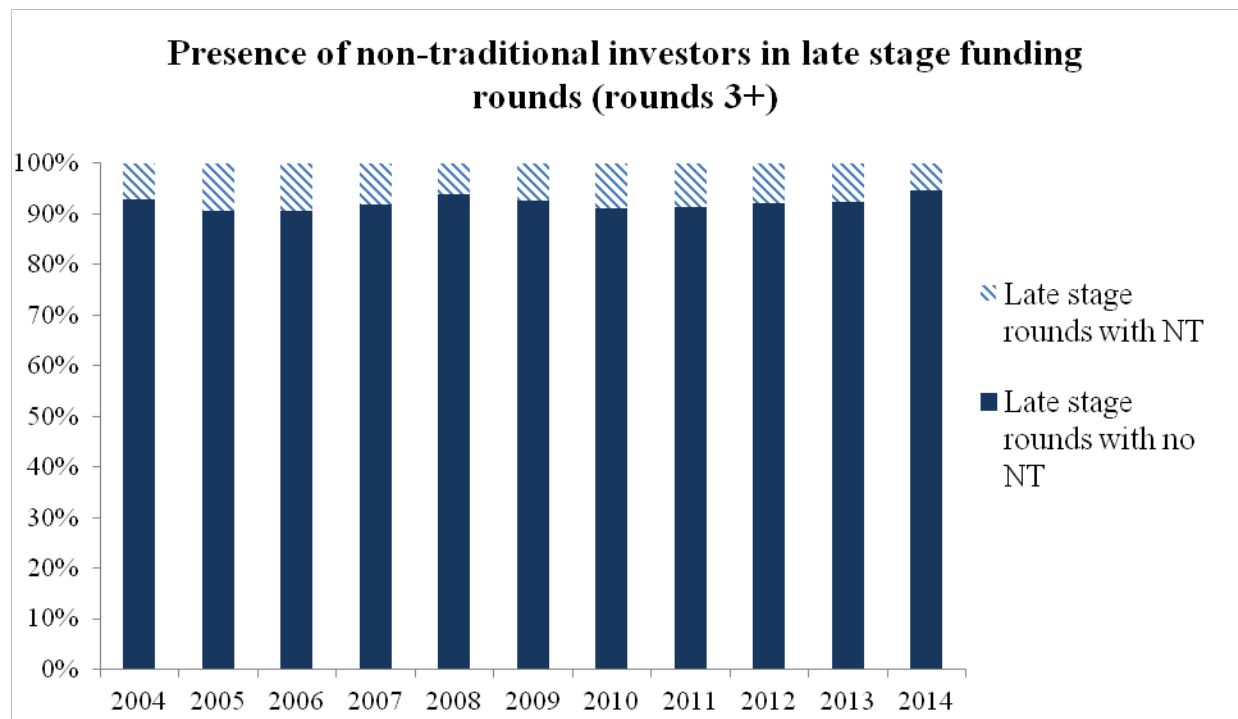


Exhibit 3: Average time between funding rounds in years 2004-2014



Participation of non-traditional investors in late stage rounds (Series C+) remains at a low ~10% of deals by count and has been consistent throughout the examined time period. Nevertheless, the observed increase in the average number of rounds to IPO implies a higher degree of non-traditional investor participation in an average company's funding history.

Exhibit 4: Non-traditional investor involvement in late stage rounds



Regression Analysis

The combined datasets of CRSP and SDC Platinum provide a sample size of 263 IPOs in the time period 2004-2013. This sample spans multiple industries including both High Tech and non-High Tech companies.

The first regression was run on the 1-day return from the IPO offer price versus the set of funding round characteristics as well as a binary indicator of whether at least one non-traditional investor was present in the company's pre-IPO funding history. The goal of the regression is to see if non-traditional investor participation has an effect on pre-IPO investor returns.

The results are summarized in Table 3 below. The estimated coefficients are economically small and statistically insignificant, implying that the examined variables alone are

not sufficient to explain the trend in 1-day post-IPO returns. Most importantly, the presence of a non-traditional investor does not seem to impact post-IPO performance in this sample.

Table 3: Regression of 1-day returns against presence of non-traditional investors

	Adj SS	Adj MS	F-Value	P-Value	Coefficient
Regression	0.144	0.029	0.680	0.641	
Time to IPO (years)	0.011	0.011	0.260	0.611	0.002
Number of rounds	0.053	0.053	1.240	0.267	0.004
Average number of investors per round	0.079	0.079	1.850	0.175	-0.007
Known total investment amount in company	0.001	0.000	0.010	0.915	0.000
Presence of non-traditional investors	0.004	0.003	0.080	0.775	0.009
R ²	1.30%				
Adjusted R ²	0.00%				

To determine whether these results reflect an overly broad definition of non-traditional investor participation, I run a regression where the indicator of participation takes on a value of unity only if the investors were present in later funding rounds (Series C+). The results are shown in Table 4 below. They further support the observation that these variables are insignificant contributors to post-IPO performance.

Table 4: Regression of 1-day returns against presence of non-traditional investors in late stage rounds (Series C+)

	Adj SS	Adj MS	F-Value	P-Value	Coefficient
Regression	0.1736	0.03472	0.82	0.536	
Time to IPO (years)	0.0061	0.0061	0.1400	0.7040	0.0014
Number of rounds	0.0355	0.0355	0.8400	0.3610	0.0037
Average number of investors per round	0.0967	0.0967	2.2800	0.1320	-0.0073
Known total investment amount in company	0.0002	0.0002	0.0100	0.9420	0.0000
Presence of NT investors in late stage	0.0334	0.0334	0.7900	0.3750	0.0410
R ²	1.57%				
Adjusted R ²	0.00%				

Finally, another version of the regression was run in an attempt to understand whether specific types of non-traditional investors can be shown to be significantly impacting returns (Table 5). Again, results indicate no significant correlations.

Table 5: Regression of 1-day returns against presence of specific non-traditional investor types

	Adj SS	Adj MS	F-Value	P-Value	Coefficient
Regression	0.2324	0.029046	0.68	0.707	
Time to IPO (years)	0.0194	0.019382	0.46	0.501	0.0024
Number of rounds	0.0665	0.066506	1.56	0.213	0.0051
Average number of investors per round	0.0759	0.0759	1.7800	0.1830	-0.0064
Known total investment amount in company	0.0000	0.0000	0.0000	0.9880	0.0000
Presence of bank affiliated investors	0.0037	0.0037	0.0900	0.7700	-0.0123
Presence of investment management firm	0.0082	0.0082	0.1900	0.6620	0.0159
Presence of insurance firm affiliate	0.0397	0.0397	0.9300	0.3350	-0.0929
Presence of endowment, foundation or pension	0.0417	0.041741	0.98	0.323	-0.211
R ²	2.10%				
Adjusted R ²	0.00%				

Technology companies

A set of analyses was also run on a subset of the sample, including all types of technology companies but excluding the non-high tech category. Tech companies have notably captured hedge and mutual fund interest and are therefore worth exploring separately. The sample size here is 141 companies including Google, Groupon and Kayak.

Table 6: Regression of 1-day returns against presence of non-traditional investors in tech companies

	Adj SS	Adj MS	F-Value	P-Value	Coefficient
Regression	0.29127	0.058254	1.4	0.229	
Time to IPO (years)	0.1242	0.1242	2.9800	0.0870	0.0100
Number of rounds	0.0002	0.0002	0.0000	0.9490	0.0004
Average number of investors per round	0.0711	0.0711	1.7100	0.1940	-0.0071
Known total investment amount in company	0.0173	0.0173	0.4100	0.5210	0.0000
Presence of non-traditional investors	0.0339	0.0339	0.8100	0.3690	0.0396
R ²	4.92%				
Adjusted R ²	1.40%				

Table 7: Regression of 1-day returns against presence of non-traditional investors in late stage rounds (Series C+) in tech companies

	Adj SS	Adj MS	F-Value	P-Value	Coefficient
Regression	0.26087	0.052173	1.25	0.292	
Time to IPO (years)	0.1243	0.1243	2.9700	0.0870	0.0101
Number of rounds	0.0000	0.0000	0.0000	0.9820	-0.0001
Average number of investors per round	0.0522	0.0522	1.2400	0.2670	-0.0060
Known total investment amount in company	0.0226	0.0226	0.5400	0.4640	0.0000
Presence of non-traditional investors in late stage	0.0035	0.0035	0.0800	0.7730	0.0161
R ²	4.41%				
Adjusted R ²	0.87%				

Table 8: Regression of 1-day returns against presence of specific non-traditional investor types in tech companies⁴

	Adj SS	Adj MS	F-Value	P-Value	Coefficient
Regression	0.348	0.058	1.400	0.221	
Time to IPO (years)	0.123	0.123	2.960	0.088	0.0100
Number of rounds	0.000	0.000	0.000	0.973	0.0002
Average number of investors per round	0.056	0.056	1.340	0.250	-0.0061
Known total investment amount in company	0.009	0.009	0.220	0.639	0.0000
Presence of bank affiliated investors	0.090	0.090	2.170	0.143	0.0879
Presence of investment management firm	0.000	0.000	0.000	0.964	-0.0024
R ²	5.88%				
Adjusted R ²	1.66%				

Although the R² values appear to be higher on a sample subset level, the results on the tech company data are similarly statistically insignificant. Only the variable ‘Time to IPO (years)’ appears to have a low P-value; its coefficient is positive, implying that tech companies that go public later are more likely to experience a price increase from the IPO offer price level.

The final set of analysis uses 3-day returns from the IPO opening price (first trade price) as the measure of return in the regressions, in order to study the effect of non-traditional investor participation on public investor post-IPO returns. The results are provided below in Tables 9-11.

⁴ Two variables, presence of insurance firm affiliates, and presence of endowment, foundation or pension, were dropped in this analysis due to low sample size

Table 9: Regression of 3-day returns against presence of non-traditional investors

	Adj SS	Adj MS	F-Value	P-Value	Coefficient
Regression	0.2125	0.042499	0.33	0.896	
Time to IPO (years)	0.0096	0.009592	0.07	0.786	0.002
Number of rounds	0.002	0.002002	0.02	0.901	0.001
Average number of investors per round	0.1144	0.114377	0.88	0.349	-0.008
Known total investment amount in company	0.0595	0.059489	0.46	0.499	0.000
Presence of non-traditional investors	0.0053	0.005305	0.04	0.840	0.011
R ²	0.63%				
Adjusted R ²	0.00%				

Table 10: Regression of 3-day returns against presence of non-traditional investors in late stage rounds (Series C+)

	Adj SS	Adj MS	F-Value	P-Value	Coefficient
Regression	0.3441	0.068818	0.53	0.752	
Time to IPO (years)	0.0018	0.001819	0.01	0.906	0.0007
Number of rounds	0.0009	0.000892	0.01	0.934	-0.0006
Average number of investors per round	0.1701	0.17006	1.31	0.253	-0.0096
Known total investment amount in company	0.0508	0.050781	0.39	0.531	0.0000
Presence of non-traditional investors in late stage	0.1369	0.136899	1.06	0.305	0.0831
R ²	1.02%				
Adjusted R ²	0.00%				

Table 11: Regression of 3-day returns against presence of specific non-traditional investor types

	Adj SS	Adj MS	F-Value	P-Value	Coefficient
Regression	0.5908	0.073851	0.57	0.803	
Time to IPO (years)	0.0103	0.010256	0.08	0.779	0.0018
Number of rounds	0.0023	0.002346	0.02	0.893	0.0010
Average number of investors per round	0.1467	0.146726	1.13	0.289	-0.0089
Known total investment amount in company	0.0222	0.022234	0.17	0.679	0.0000
Presence of bank affiliated investors	0.1734	0.173417	1.34	0.249	0.0850
Presence of investment management firm	0.0552	0.055183	0.42	0.515	0.0412
Presence of insurance firm affiliate	0.1633	0.163344	1.26	0.263	-0.188
Presence of endowment, foundation or pension	0.0474	0.047373	0.36	0.546	-0.225
R ²	1.76%				
Adjusted R ²	0.00%				

Again, the regressions indicate no significant correlations between post-IPO public investor gains (measured from the IPO opening price to the closing price on Day 3 of trading) and funding round or investor characteristics. Running the same analysis on 10-day returns provides similar statistically insignificant results and was therefore excluded from this paper.

IV. CONCLUSION

The analysis presented in this paper confirms the trends taking shape in the venture capital markets. In the time period 2004-2014, the average time to IPO has nearly doubled to 9 years and the average inter-round intervals have halved to 8 months. This implies a surge in late stage funding rounds. The overall participation of non-traditional investors in late stage rounds

remains at a mere ~10% of deals by count, but the increase in late stage rounds has led to more companies having at least one non-traditional investor in their pre-IPO funding history.

With the trends confirmed, this paper then sought to show whether this increase in non-traditional investor activity has any impact on post-IPO stock returns from both a pre-IPO investor and a public investor point of view. On an aggregate level, using a sample size of 263 publicly traded firms, there seem to be no conclusive correlations. I find low R^2 values in all regressions, indicating that the analyzed variables alone are not sufficient to explain post-IPO stock performance. More importantly, I cannot reject the null hypothesis that the presence of a non-traditional investor makes no difference to a VC-backed company's post-IPO stock performance in the overall sample.

Analyzing post-IPO gains for the subset of 141 companies operating in the tech sector, I find similarly weak correlations. The P-values remain too low to conclusively establish a significant correlation with pre-IPO investor characteristics.

My initial hypothesis, that the increased participation of non-traditional investors could lead to overvaluation in the rounds immediately preceding IPO and could therefore suppress post-IPO returns from a pre-IPO investor point of view, thus does not find support in the data, whether I analyze the overall sample or tech companies specifically. The hypothesis that the increased participation of non-traditional investors could impact the post-IPO gains from a public investor point of view also does not find support in the data.

It is important to note that my results could be influenced by limited data availability or accuracy. I retrieved data from multiple sources, with many companies having incomplete data or undisclosed investment amounts and investor names. It is also possible that the expected

effect is not yet observable; the venture capital community has only recently begun discussing the impact of investment by non-traditional investors, starting in early 2014. Non-traditional investors may only have started to invest small amounts in the deals, not significantly large enough to influence term sheets and valuations. In any case, this study merits further investigation in the future when more data points become available.

Limitations

As with many venture capital research efforts, my analysis faced many data related challenges. In this section, I highlight the limitations my research has encountered in the hope that further research can be conducted on this topic when these limitations can be overcome.

- *Data availability:* This was the largest obstacle in this study. Both CRSP and SDC Platinum, two of the best known finance sources, had a limited datasets. Not all IPOs in the investigated time period were captured in SDC Platinum, while CRSP lacked IPO data on many companies in the original sample. On a more granular level, the databases had a limited set of information available on a company. Where the variables existed in the database, data was missing for many companies. This is likely to be due to the lack of data disclosure by venture capital firms. After filtering companies for completeness of data, the final sample included only 263 companies.
- *Data accuracy:* There were many inconsistencies between databases on many dimensions including IPO dates and company classifications. This is a typical challenge when data is fragmented between multiple sources and there often is no workaround except manual consolidation.

- *Limited research in this field:* It would be inaccurate to say that there is an abundance of precedent research papers in this topic from which to formulate baseline research efforts and seek comprehensive sources of data. This is likely due in part to well-known data limitations, an issue not faced by securities research.

REFERENCES

- Benner, K. April 2014. "Surge in Late Stage Investing Crimps Post-IPO Gains", The Information, <https://www.theinformation.com/Surge-in-Late-Stage-Investing-Crimps-Post-IPO-Gains>
- Carey, D. and S. Saitto. March 2015. "Hedge Funds are Boosting Tech Valuations to Dangerous Heights", Bloomberg, <http://www.bloomberg.com/news/articles/2015-03-23/dizzying-pre-ipo-tech-values-spurred-by-rush-of-hedge-fund-money>
- CB Insights. March 2014. "Are Hedge Funds and Mutual Funds Pushing Later-Stage VCs out of Hot Late-Stage Deals?". <http://www.cbinsights.com/blog/late-stage-vc-investors>
- CB Insights. January 2014. "Hedge Funds and Mutual Funds Increase Investment Pace to Private Tech Companies". <https://www.cbinsights.com/blog/hedge-mutualfund-tech-investments/>
- CB Insights. March 2015. "The Rise of Hedge Funds and Mutual Funds in Tech Startup Investing in Two Charts". <https://www.cbinsights.com/blog/hedge-fund-mutual-fund-tech-startups/>
- Garland, R. April 2014. "US Venture Capital Investment Highest since 2001". <http://blogs.wsj.com/venturecapital/2014/04/17/us-venture-capital-investment-highest-since-2001/>
- Krishnan , C.N.V., V.I. Ivanov, R.W. Masulis and A.K. Singh. December 2009. "Venture Capital Reputation, post- IPO performance and Corporate Governance". Working Paper 265, ECGI.
- Lowry, M., M.S. Officer and G.W. Schwert. 2010. "The Variability of IPO Initial Returns". *The Journal of Finance*, 65, 425-465.
- Meggison, W. L. and K. A. Weiss. 1991. "Venture Capitalist Certification in Initial Public Offerings." *Journal of Finance*, 46, 879–903.
- Solomon, G. February 2014. "Hedge Fund Rising". TechCrunch. <http://techcrunch.com/2014/02/22/venture-investing-hedge-funds-rising/>
- Sørensen, M. 2007. "How Smart Is Smart Money? A Two-Sided Matching Model of Venture Capital." *Journal of Finance*, 62, 2725–2762.

APPENDIX: REGRESSION SAMPLE INFORMATION

Company Name	Ticker	Industry group (SDC)	IPO year
Accentia BioPharmaceuticals, Inc.	ABPI	Medical/Health/Life Science	2005
AcelRx Pharmaceuticals, Inc.	ACRX	Medical/Health/Life Science	2011
Achillion Pharmaceuticals, Inc.	ACHN	Medical/Health/Life Science	2006
Acme Packet, Inc.	APKT	Communications and Media	2006
Acorn International, Inc.	ATV	Communications and Media	2007
Actions Semiconductor Co., Ltd.	ACTS	Semiconductors/Other Elect	2005
Addus Homecare Corporation	ADUS	Medical/Health/Life Science	2009
Advanced Life Sciences Holdings, Inc.	ADLS	Medical/Health/Life Science	2005
Aegerion Pharmaceuticals, Inc.	AEGR	Medical/Health/Life Science	2010
AerCap Holdings N.V.	AER	Non-High-Technology	2006
Affymax Inc	AFFY	Biotechnology	2006
AGA Medical Corporation	AGAM	Medical/Health/Life Science	2009
Agria Corporation	GRO	Non-High-Technology	2007
Air Lease Corporation	AL	Non-High-Technology	2011
Aircastle, Ltd.	AYR	Non-High-Technology	2006
Alexza Pharmaceuticals, Inc.	ALXA	Medical/Health/Life Science	2006
Alimera Sciences, Inc.	ALIM	Biotechnology	2010
Allion Healthcare, Inc.	ALLI	Medical/Health/Life Science	2005
Alphatec Holdings Inc	ATEC	Medical/Health/Life Science	2006
Altra Holdings, Inc.	AIMC	Non-High-Technology	2006
American Public Education, Inc.	APEI	Computer Related	2007
Amicus Therapeutics, Inc.	FOLD	Biotechnology	2007
Amyris, Inc.	AMRS	Non-High-Technology	2010
Anacor Pharmaceuticals, Inc.	ANAC	Medical/Health/Life Science	2010
Animal Health Holdings Inc	AHII	Biotechnology	2007
Annie's, Inc.	BNNY	Non-High-Technology	2012
Anthera Pharmaceuticals, Inc.	ANTH	Medical/Health/Life Science	2010
Approach Oil and Gas Inc.	AREX	Non-High-Technology	2007
ARAMARK Corp	ARMK	Non-High-Technology	2013
Archipelago Learning LLC	ARCL	Computer Related	2009
Arcos Dorados, S.A.	ARCO	Non-High-Technology	2011
Artisan Partners, L.P.	APAM	Non-High-Technology	2013
ARYx Therapeutics, Inc.	ARYX	Biotechnology	2007

Aspreva Pharmaceuticals Corporation	ASPV	Medical/Health/Life Science	2005
ATA, Inc.	ATAI	Computer Related	2008
Athlon Energy LP	ATHL	Non-High-Technology	2013
AtriCure Inc	ATRC	Biotechnology	2005
AuthenTec, Inc.	AUTH	Semiconductors/Other Elect	2007
Avago Technologies, Ltd.	AVGO	Semiconductors/Other Elect	2009
Aventine Renewable Energy, Inc.	AVR	Non-High-Technology	2006
AVEO Pharmaceuticals, Inc.	AVEO	Medical/Health/Life Science	2010
Aviv REIT, Inc.	AVIV	Non-High-Technology	2013
AxoGen Inc	AXGN	Medical/Health/Life Science	2013
BankUnited Inc	BKU	Non-High-Technology	2011
Benefitfocus.Com Inc	BNFT	Computer Related	2013
Berry Plastics Corporation	BERY	Non-High-Technology	2012
BG Medicine, Inc.	BGMD	Biotechnology	2011
Bill Barrett Corp	BBG	Non-High-Technology	2004
Biodel, Inc.	BIOD	Medical/Health/Life Science	2007
BioMimetic Therapeutics, Inc.	BMTI	Biotechnology	2006
Blackbaud Inc	BLKB	Computer Related	2004
BlueLinx Holdings Inc	BXC	Non-High-Technology	2004
Booz Allen Hamilton Holding Corporation	BAH	Non-High-Technology	2010
Brookdale Senior Living Inc	BKD	Medical/Health/Life Science	2005
C&J Energy Services Inc	CJES	Non-High-Technology	2011
Cadence Pharmaceuticals, Inc.	CADX	Medical/Health/Life Science	2006
Caesars Entertainment Corporation	CZR	Non-High-Technology	2012
Camelot Information Systems, Inc.	CIS	Computer Related	2010
Capital Bank Financial Corporation	CBF	Non-High-Technology	2012
Carbonite, Inc.	CARB	Computer Related	2011
Cardica Inc	CRDC	Biotechnology	2006
Cardtronics, Inc.	CATM	Computer Related	2007
Cavium, Inc.	CAVM	Semiconductors/Other Elect	2007
Celanese Corp	CE	Non-High-Technology	2005
Cellu Tissue Holdings, Inc.	CLU	Non-High-Technology	2010
Chart Industries, Inc.	GTLS	Non-High-Technology	2006
Chesapeake Midstream Partners, L.P.	CHKM	Non-High-Technology	2010
China Finance Online Co., Ltd.	JRJC	Computer Related	2004
China GrenTech Corporation	GRRF	Semiconductors/Other Elect	2006
China Kanghui Holdings	KH	Biotechnology	2010
China Nepstar Chain Drugstore, Ltd.	NPD	Non-High-Technology	2007
China Nuokang Bio-Pharmaceutical, Inc.	NKBP	Biotechnology	2009
China Sunergy Co., Ltd.	CSUN	Semiconductors/Other Elect	2007
Chuy's Holdings, Inc.	CHUY	Non-High-Technology	2012
Cinemark Holdings, Inc.	CNK	Communications and Media	2007
Clayton Holdings LLC	CLAY	Non-High-Technology	2006

Clean Energy Fuels Corp	CLNE	Non-High-Technology	2007
Clearwire Corporation	CLWR	Communications and Media	2007
Cleveland Biolabs, Inc.	CBLI	Biotechnology	2006
Clovis Oncology, Inc.	CLVS	Biotechnology	2011
Clubcorp Inc	MYCC	Non-High-Technology	2013
CNInsure Inc.	CISG	Non-High-Technology	2007
Cobalt International Energy, Inc.	CIE	Non-High-Technology	2009
Compellent Technologies, Inc.	CML	Computer Related	2007
Complete Production Services, Inc.	CPX	Non-High-Technology	2006
Comverge, Inc.	COMV	Computer Related	2007
Concord Medical Services Holdings, Ltd.	CCM	Medical/Health/Life Science	2009
Constant Contact, Inc.	CTCT	Computer Related	2007
Copano Energy LLC	CPNO	Non-High-Technology	2004
CoTherix, Inc.	CTRX	Biotechnology	2004
Coty US Inc	COTY	Non-High-Technology	2013
Country Style Cooking Restaurant Chain Co., Ltd.	CCSC	Non-High-Technology	2010
Criteo SA	CRTO	Communications and Media	2013
CTC Media, Inc.	CTCM	Communications and Media	2006
Cumberland Pharmaceuticals, Inc.	CPIX	Medical/Health/Life Science	2009
CVR Energy, Inc.	CVI	Non-High-Technology	2007
Data Domain, Inc.	DDUP	Computer Related	2007
Del Frisco's Restaurant Group, Inc.	DFRG	Non-High-Technology	2012
Demand Media, Inc.	DMD	Communications and Media	2011
Design Within Reach Inc	DWRI	Non-High-Technology	2004
Diamondback Energy, Inc.	FANG	Non-High-Technology	2012
Dice Holdings, Inc.	DHX	Non-High-Technology	2007
Digital Realty Trust, Inc.	DLR	Computer Related	2004
DivX, Inc.	DIVX	Computer Related	2006
Dollar Financial Group Inc	DLLR	Non-High-Technology	2005
Dollar General Corporation	DG	Non-High-Technology	2009
Doubleline LLC	DSL	Non-High-Technology	2013
Double-Take Software, Inc.	DBTK	Computer Related	2006
Dover Saddlery Inc	DOVR	Non-High-Technology	2005
DreamWorks Animation SKG, Inc.	DWA	Communications and Media	2004
Dresser-Rand, Inc.	DRC	Non-High-Technology	2005
Duff & Phelps Corp	DUF	Non-High-Technology	2007
Dunkin Brands Group, Inc.	DNKN	Non-High-Technology	2011
Dyncorp International Inc	DCP	Non-High-Technology	2006
Eagle Bulk Shipping, Inc.	EGLE	Non-High-Technology	2005
Eagle Rock Energy Partners LP	EROE	Non-High-Technology	2006
Educate Inc	EEEE	Non-High-Technology	2004
Ehealth Inc	EHTH	Medical/Health/Life Science	2006
eLong, Inc.	LONG	Computer Related	2004
Eloqua, Inc.	ELOQ	Computer Related	2012

Emdeon Inc	EM	Medical/Health/Life Science	2009
Emergency Medical Services	EMS	Medical/Health/Life Science	2005
Endocyte, Inc.	ECYT	Biotechnology	2011
Endurance International Group Inc	EIGI	Computer Related	2013
Energy Transfer Equity LP	ETE	Non-High-Technology	2006
EnergySolutions Inc	ES	Non-High-Technology	2007
EnerNOC, Inc.	ENOC	Computer Related	2007
EnteroMedics, Inc.	ETRM	Medical/Health/Life Science	2007
EPAM Systems, Inc.	EPAM	Computer Related	2012
Epizyme Inc	EPZM	Biotechnology	2013
Erickson Air-Crane, Inc.	EAC	Non-High-Technology	2012
Eschelon Telecom, Inc.	ESCH	Communications and Media	2005
Eurand NV	EURX	Medical/Health/Life Science	2007
EV Energy Partners, L.P.	EVEP	Non-High-Technology	2006
ev3, Inc.	EVVV	Biotechnology	2005
EverBank Financial Corporation	EVER	Non-High-Technology	2012
Evertec Group LLC	EVTC	Computer Related	2013
ExactTarget Inc	ET	Computer Related	2012
ExlService Holdings, Inc.	EXLS	Computer Related	2006
Express, Inc.	EXPR	Non-High-Technology	2010
Fabrinet, Inc.	FN	Semiconductors/Other Elect	2010
Fairway Group Holdings Corporation	FWM	Non-High-Technology	2013
Fastclick, Inc.	FSTC	Communications and Media	2005
Fate Therapeutics Inc	FATE	Biotechnology	2013
Fidelity and Guaranty Life Insurance Company	FGL	Non-High-Technology	2013
Fidelity National Information Services Inc	FIS	Computer Related	2006
First Mercury Financial Corporation	FMR	Non-High-Technology	2006
First NBC Bank Holding Company	NBCB	Non-High-Technology	2013
Flagstone Reinsurance Holdings, S.A.	FSR	Non-High-Technology	2007
FleetCor Technologies, Inc.	FLT	Computer Related	2010
Fleetmatics Group PLC	FLTX	Computer Related	2012
Fluidigm Corporation	FLDM	Biotechnology	2011
Focus Media Holding Ltd	FMCN	Communications and Media	2005
Fortegra Financial Corp	FRF	Non-High-Technology	2010
Fortress Investment Group LLC	FIG	Non-High-Technology	2007
Fox Factory Inc	FOXF	Non-High-Technology	2013
Francesca's Collections, Inc.	FRAN	Non-High-Technology	2011
Freescale Semiconductor, Ltd.	FSL	Semiconductors/Other Elect	2011
FXCM, Inc.	FXCM	Non-High-Technology	2010
Generac Holdings, Inc.	GNRC	Non-High-Technology	2010
Genpact Ltd	G	Non-High-Technology	2007
GeoMet, Inc.	GMET	Non-High-Technology	2006
Gigamon LLC	GIMO	Communications and Media	2013
Global Brass and Copper, Inc.	BRSS	Non-High-Technology	2013
Global Cash Access Holdings, Inc.	GCA	Non-High-Technology	2005

Global Education and Technology Group	GEDU	Non-High-Technology	2010
Glu Mobile, Inc.	GLUU	Computer Related	2007
GNC Corporation	GNC	Non-High-Technology	2011
Goodman Global, Inc.	GGL	Non-High-Technology	2006
Google, Inc.	GOOG	Computer Related	2004
Gordmans Stores, Inc.	GMAN	Non-High-Technology	2010
Groupon, Inc.	GRPN	Computer Related	2011
GSI Technology, Inc.	GSIT	Semiconductors/Other Elect	2007
Guidewire Software, Inc.	GWRE	Computer Related	2012
Gundle/SLT Environmental, Inc.	GSE	Non-High-Technology	2012
Hannon Armstrong Capital LLC	HASI	Non-High-Technology	2013
Hansen Medical, Inc.	HNSN	Medical/Health/Life Science	2006
HD Supply Holdings, Inc.	HDS	Non-High-Technology	2013
HealthSpring, Inc.	HS	Medical/Health/Life Science	2006
Herbalife Ltd	HLF	Non-High-Technology	2004
Heritage-Crystal Clean, Inc.	HCCI	Non-High-Technology	2008
hhgregg, Inc.	HGG	Non-High-Technology	2007
Hilton Worldwide Inc	HLT	Non-High-Technology	2013
Horizon Technology Finance Corporation	HRZN	Non-High-Technology	2010
Hyatt Hotels Corporation	H	Non-High-Technology	2009
ICx Technologies, Inc.	ICXT	Semiconductors/Other Elect	2007
Ideal Power Inc	IPWR	Non-High-Technology	2013
Ignite Restaurant Group, Inc.	IRG	Non-High-Technology	2012
Ikanos Communications, Inc.	IKAN	Semiconductors/Other Elect	2005
ImaRx Therapeutics, Inc.	IMRX	Biotechnology	2007
Imperva, Inc.	IMPV	Computer Related	2011
InnerWorkings, Inc.	INWK	Communications and Media	2006
Innophos Holdings Inc	IPHS	Non-High-Technology	2006
Intralinks Holdings, Inc.	IL	Computer Related	2010
Intrepid Potash, Inc.	IPI	Non-High-Technology	2008
iSoftStone Holdings, Ltd.	ISS	Computer Related	2010
J Crew Group Inc	JCG	Non-High-Technology	2006
J G Wentworth & Company Inc	JGW	Non-High-Technology	2013
Jackson-Hewitt, Inc.	JTX	Non-High-Technology	2004
Jazz Pharmaceuticals Inc	JAZZ	Medical/Health/Life Science	2007
JinkoSolar Holding Co., Ltd.	JKS	Semiconductors/Other Elect	2010
Jive Software, Inc.	JIVE	Computer Related	2011
K & F Industries, Inc.	KFI	Non-High-Technology	2005
Kanbay International Inc	KBAY	Computer Related	2004
KAR Auction Services	KAR	Non-High-Technology	2009
KayakCom	KYAK	Computer Related	2012
Kinder Morgan, Inc.	KMI	Non-High-Technology	2011
Kosmos Energy, Ltd.	KOS	Non-High-Technology	2011
Laredo Petroleum, Inc.	LPI	Non-High-Technology	2011
LeMaitre Vascular, Inc.	LMAT	Medical/Health/Life Science	2006

LHC Group, Inc.	LHCG	Medical/Health/Life Science	2005
Limelight Networks, Inc.	LLNW	Communications and Media	2007
Lincoln Educational Services Corp	LINC	Non-High-Technology	2005
Liquidity Services, Inc.	LQDT	Non-High-Technology	2006
LogMeIn, Inc.	LOGM	Computer Related	2009
LoopNet, Inc.	LOOP	Computer Related	2006
LPL Financial LLC	LPLA	Non-High-Technology	2010
Lumber Liquidators Holdings, Inc.	LL	Non-High-Technology	2007
Luna Innovations, Inc.	LUNA	Semiconductors/Other Elect	2006
M/A-COM Technology Solutions, Inc.	MTSI	Semiconductors/Other Elect	2012
Maidenform Brands, Inc.	MFB	Non-High-Technology	2005
MAKO Surgical Corporation	MAKO	Medical/Health/Life Science	2008
Marketo Inc	MKTO	Computer Related	2013
Mattress Holding Corporation	MFRM	Non-High-Technology	2011
Mecox Lane, Ltd.	MCOX	Non-High-Technology	2010
MedGenics Inc	MDGN	Biotechnology	2011
MediaMind Technologies, Inc.	MDM D	Communications and Media	2010
Medidata Solutions, Inc.	MDSO	Computer Related	2009
Meru Networks, Inc.	MERU	Computer Related	2010
Metals USA Holdings Corporation	MUSA	Non-High-Technology	2010
Micrus Endovascular Corp	MEND	Biotechnology	2005
Midstates Petroleum Company, Inc.	MPO	Non-High-Technology	2012
Millennial Media, Inc.	MM	Communications and Media	2012
Mindray Medical International, Ltd.	MR	Medical/Health/Life Science	2006
MISTRAS Group, Inc.	MG	Semiconductors/Other Elect	2009
Molycorp, Inc.	MCP	Non-High-Technology	2010
Momenta Pharmaceuticals, Inc.	MNTA	Medical/Health/Life Science	2004
Montage Technology Co Ltd	MONT	Semiconductors/Other Elect	2013
Morningstar, Inc.	MORN	Non-High-Technology	2005
Motricity, Inc.	MOTR	Communications and Media	2010
MRC Global, Inc.	MRC	Non-High-Technology	2012
MYR Group, Inc.	MYRG	Non-High-Technology	2008
Nationstar Mortgage Holdings, Inc.	NSM	Non-High-Technology	2012
NeoPhotonics Corporation	NPTN	Semiconductors/Other Elect	2011
Ness Technologies Inc	NSTC	Computer Related	2004
NeurogesX, Inc.	NGSX	Biotechnology	2007
New Oriental Education & Technology Group Inc	EDU	Non-High-Technology	2006
New River Pharmaceuticals, Inc.	NRPH	Medical/Health/Life Science	2004
Newkirk Realty Trust, Inc.	NKT	Non-High-Technology	2005
NewStar Financial, Inc.	NEWS	Non-High-Technology	2006
Nextest Systems Corporation	NEXT	Semiconductors/Other Elect	2006
Nielsen Company, The	NLSN	Non-High-Technology	2011

Ninetowns Internet Technology Group Co., Ltd.	NINE	Computer Related	2004
Niska Gas Storage	NKA	Non-High-Technology	2010
Noodles & Co	NDLS	Non-High-Technology	2013
Noranda Aluminum Holdings Corporation	NOR	Non-High-Technology	2010
NYMEX Holdings, Inc.	NMX	Non-High-Technology	2006
SemiLEDs Corporation	LEDS	Semiconductors/Other Elect	2010
ServiceNow, Inc.	NOW	Computer Related	2012
SKY-MOBI, Ltd.	MOBI	Computer Related	2010
Springleaf Financial Services Inc	LEAF	Non-High-Technology	2013
Sungy Mobile Ltd	GOMO	Computer Related	2013
Tableau Software LLC	DATA	Computer Related	2013