# Trade Size and the Cross-Sectional Relation to Future Returns 

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#### Abstract

This paper uses trade clusters, centered around 100-share (and increments of 100 shares), 500 -share, and 1000 -share categories, to analyze the relationship between trade size clusters and the cross-section of future stock returns across momentum portfolios. We find that the winner-loser momentum portfolios that have a high concentration of 500 or 1000 -share trade size clusters earn an alpha of almost $1 \%$ per month which is almost double the performance of the momentum strategy not predicated on these trade size clusters. The performance of the 500 and 1000 -share trade size clusters is not matched by any other trade size cluster. This ability of the 500 and 1000 -share trade clusters in better predicting momentum returns persists regardless of the decimalization in stock quotes and is more resilient to "momentum crashes" that plague the conventional momentum strategy. We also link the momentum and value strategies by showing that firms with high past levels of 500 -share trade clusters exhibit glamour characteristics while the 100 -share trade clusters exhibit more value characteristics.


Keywords: Trade size, Intra-Day trades, Returns
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## 1 Introduction

Jegadeesh and Titman's (1993) seminal article documents a momentum effect where stocks with good (bad) recent performance continue to outperform (underperform) for up to a one-year horizon. This paper investigates a relatively unexplored area of whether trade size interacts with past momentum returns in the prediction of future returns. A vast literature has emerged relating momentum profits to a behavioral perspective ${ }^{1}$ that lends itself to focusing on the impact of small, retail trades on momentum profits (Hvidkjaer, 2006). But little attention has been focused on whether large trades, and, in particular, large trade size clusters, are associated with momentum. Trade size and stock returns are jointly determined by theory (Easley and O'Hara, 1987), yet trade size is invariably treated separately from returns.

Using the momentum anomaly as a basis, we find that portfolios that exhibit a high concentration of 100-share trades underperform relative to the base momentum strategy while portfolios that are dominated by 500 and 1000-share trade clusters significantly outperform the base momentum strategy. We further find that portfolios dominated by 500 and 1000 -share trade clusters do not evidence a seasonality in momentum returns (Jegadeesh and Titman, 1993) and they continue to earn significant momentum profits even when the base momentum strategy crashes (Daniel and Moskowitz, 2013). These results persist in the post-decimalization period and are strongest for the stocks that experience the highest level of trade volume. Our results indicate that given a weak-form information signal, investors trading in 500 or 1000-share trade clusters trade in a manner consistent with the anomalous signal with greatly enhanced momentum performance, while investors trading round-lots or non-clustered trades exhibit a contrarian trading behavior with reduced momentum performance.

Trade size has been examined by Barclay and Warner (1993) who argue that trade sizes from

[^0]500 shares to 9,900 shares per trade contain more value relevant information than do smaller trade sizes. Hasbrouck (1995) and Chakravarty (2001) document the presence of stealth trading by institutional investors and show that medium-sized trades tend to have a disproportionately greater aggregate price impact that is attributable to informed traders disguising at least some of their trades. Keim and Madhavan (1995) provide empirical evidence that institutions often break up their orders into discrete trade sizes and fill them over time and Chordia and Subrahmanyam (2004) model conditions in which traders find it optimal to break up their orders to minimize price impact. Battalio and Mendenhall (2005) use the 500 -share trade size as a minimum delineation for more informed trades and contend that trade sizes of 100 to 400 correspond to the trading interests of less informed traders. Alexander and Peterson (2007) find that 500 and 1000 share trade size clustering is consistent with the actions of stealth traders ${ }^{2}$ who tend to use medium sized rounded transactions in an attempt to disguise their trades. Given the potential that 500 and 1000 -share trade size clusters evidence more informed trade, it is an empirical question whether they are useful in predicting future returns. We explicitly study whether medium size trade clusters, as opposed to one round-lot and non-clustered trades, better predict momentum returns. We refer to this as the stealth trading hypothesis.

Existing studies on trade size are often predicated on inferring trade direction ${ }^{3}$ to allow inferences to be drawn from the order imbalance (Lee, 1992; Battalio and Mendenhall, 2005; Hvidkjaer, 2006) or by using trade volume to identify different investor clienteles in the market (Lee and Swaminathan, 2000), in conjunction with earnings or momentum tests. However, inferring trade direction is exceptionally difficult in the period after the decimalization of stock quotes because of the proliferation of quotes that makes relating the trade price to the relevant quote or the National Best-Bid and Offer (hereafter, NBBO) reference quote ${ }^{4}$ challenging. Additionally, the evidence is

[^1]mixed on the usefulness of this classification in predicting future returns. Our approach is decidedly different in that we do not require the bid and ask quotes to play a role in our measure of trade size making our trade size measure easily estimable in both the pre and post-decimalization ${ }^{5}$ periods, nor do we attach any significance to dollar volume. The post-decimalization period also affected volume measures as investors significantly increased trade volume in reflection of reduced trading costs. We find that turnover fails to predict future returns after the decimalization of stock quotes in 2001: a period that coincides with momentum crashes noted by Daniel and Moskowitz (2013) leading to robustness issues with volume as a viable predictor of momentum profits.

Easley and O'Hara (1987) theorize that trade size matters because it is correlated with private information about the security's true value. What is relevant for asset pricing is not the number of informed trades, but rather the fraction of trades that come from informed traders. In this model, based on an exogenous signal, large trade sizes (principally block trades) matter because they change the perception of the value of a security. Yet much of the literature is focused on small trades that presumably reflects retail or uninformed trades. Lee (1992) and Battalio and Mendenhall (2005) note the importance of small trades in predicting future returns in a post-earnings drift environment. Hvidkjaer (2006) argues large traders show no evidence of underreaction and large trade imbalances have little impact on subsequent returns, concluding that the results suggest that momentum could partly be driven by the behavior of small traders. We take a far different view, namely that stealth traders require the breaking up of larger orders into smaller trade sizes. The empirical regularity that we capitalize on is that these trades may be characterized by trade size clusters of 500 and 1000 -shares. We propose that trades clustered at 500 and 1000 -shares contain more value relevant information for momentum returns than other non-clustered trades or round-lot trades and we view this as a test of the stealth trading hypothesis.

The conventional wisdom is that focusing on segments ${ }^{6}$ of investor trading is useful in predicting

[^2]future returns. We further rely on the findings of Alexander and Peterson (2007) who note that trades increasingly cluster at 100,500 , and 1000 -shares, where, in particular, the 500 and 1000 -share trades are representative of more informed trading. We sum all 100 -share, 500 -share, and $1000-$ share trades ${ }^{7}$ for each firm and day over the month and then divide by the total number of trades each day. This daily ratio is then averaged over the month for each firm to determine an average daily trade size cluster. We next sort portfolios into past winner and loser momentum portfolios and then within each momentum portfolio we further sort on each trade size cluster. Conditional on momentum, we perform sort tests spanning the period 1983 to 2012 using raw returns and characteristic-adjusted returns (Daniel, Grinblatt, Titman, and Wermers, 1997; Wermers, 2003) to show the differential predictability of future returns for the large trade size and small trade size cluster portfolios.

We use the momentum anomaly as a weak-form signal to differentiate investors that trade with the anomaly with those that trade against the anomaly. We reason that the momentum "signal" is easily observed from past prices and that all traders can at least earn the base momentum profit by following the momentum strategy. However, if a group of traders can more profitably and consistently exploit the momentum anomaly, then this is evidence of stealth trading by more informed investors. We document that the base six month momentum strategy earns a monthly characteristic-adjusted return of $0.45 \%$ over the period 1983 to 2010 . We show that relatively uninformed trades congregate in one-round lot and non-clustered trades earning the base momentum return. However, in support of the stealth trading hypothesis, the hedged portfolios dominated by 500 or 1000 -share trades yields a significant characteristic-adjusted return of $0.78 \%$ per month, significantly outperforming the base momentum strategy, while portfolios avoided by the 500 and 1000 -share trades earn a characteristic-adjusted return of only $0.20 \%$ per month, significantly underperforming the base momentum strategy. Portfolios dominated by 500 and 1000 -share trades contain firms that outperform the base momentum strategy indicating the stock selectivity is enhanced by following these 500 and 1000 -share traders.

[^3]These results are robust to portfolio decile or tercile portfolios or whether we focus on NYSE/Amex of NASDAQ firms separately. Our results are strengthened when we filter by dollar trading volume in addition to price filters. We further find that large trade size portfolios formed before the momentum portfolio formation period also better predict future returns than does the base momentum strategy mitigating concerns about endogeneity bias (or feedback effects) on our results.

We find that 500 and 1000 -share traders act strategically before and after the decimalization ${ }^{8}$ in stock quotes. Before decimalization, 500 and 1000-share traders concentrate on smaller market capitalization firms, but after decimalization these large traders tend to concentrate on larger market capitalization firms. Surprisingly, the after bid-ask spread costs returns are higher after decimalization than before decimalization for the 500 and 1000-share portfolios. Interestingly, Frazzini, Israel, and Moskowitz (2013) show that actual institutional trading costs are less than a tenth as large as the quote bid-ask spread, and therefore the potential profitability of these trade cluster-based strategies is more than an double that suggested using the bid-ask spread as the relevant cost of trade.

The results also show that these two trade cluster portfolios maintain pricing ability that is remarkably consistent over specific time periods when the base momentum strategy crashes. We are able to show that the 500 -share trade clusters experience a characteristic-adjusted six-month return of $0.75 \%$ per month from 2001 to 2010, while the base momentum strategy earns an insignificant $0.30 \%$ per month. This level of return is remarkably consistent with that earned in the 1983 to 2000 period. We also find that portfolios dominated by 500 -share, and to a lesser extent $1000-$ share, portfolios are more resilient to the seasonality of momentum returns outlined by Jegadeesh and Titman (1993). Portfolios dominated by 500-share trades experience the same risk-adjusted momentum return in January as they do from February to December.

We advance the notion that 500 and 1000-share trade portfolios evidence distinct behavioral

[^4]biases, a domain which previously been dedicated to small, retail traders. Overall, our results could be construed as supportive of the conservatism bias as argued by Barberis, Shleifer, and Vishny (1998) or the self-attribution bias proposed by Daniel, Hirshleifer, and Subrahmanyam (1998). Barberis et al. (1998) discuss how a "conservatism bias" might lead investors to underreact to information, giving rise to momentum profits. The conservatism bias suggests that investors tend to underweight new information when they update their priors. If investors act in this way, prices will slowly adjust to information, but once the information is fully incorporated in prices there is no further predictability about stock returns. Conversely, Daniel et al. (1998) propose a "selfattribution" bias that is consistent with price momentum and return reversals. Daniel et al. (1998) argue that investors observe positive signals about a set of stocks, some of which perform well after the signal is received. Because of their cognitive biases, the investors attribute the performance of ex-post winners to their stock selection skills and that of the ex-post losers to bad luck. As a result, these investors become overconfident about their ability to pick winners and thereby overestimate the precision of their signals for these stocks. Based on their increased confidence in their signals, they push up the prices of the winners above their fundamental values. The delayed overreaction in this model leads to momentum profits that are eventually reversed as prices revert to their fundamentals.

Our evidence confirms that portfolios concentrating on trades clustered at 500 shares are more consistent with a self-attribution bias in that they "break" after one year, but trades clustered at 1000 shares show evidence of a conservatism bias whereby these traders drive the price to its intrinsic value over a six-month period with no evidence of a "break" in the returns subsequent to this period. Given that these two trade size clusters account for more than $45 \%$ of all large trades, the results appear to point to a distinct set of behavioral biases for traders engaged in each trade size cluster. A cause for concern for the behavioral perspective is that the 500 and 1000 -share trade cluster returns exceed the underlying bid-ask spread violating the limits to arbitrage constraint. This is most in evidence during the period after decimalization where liquidity cost are known to have fallen dramatically. Taken as a whole, the evidence is not fully consistent with any particular behavioral perspective, and given that the strategy earns an after-transaction cost return, the limits
to arbitrage may not be as binding a constraint as the behavioral perspective would require.
This study is important for the following reasons. We extend the line of research into trade size, but in a very different direction. By obviating the necessity of identifying trade direction and focusing on unidirectional trade size, we show that an easily estimable trade size portfolio can enhance the profitability of momentum based trading strategies. Rather than focusing on trade imbalances that are institutionally based (Kaniel, Liu, Saar, and Titman, 2012) or utilizing intraday dollar-volume based small trades (Lee, 1992), we would contend that large trade cluster portfolios are related to a vast array of anomalies (Hou, Xue, and Zhang, 2015; Novy-Marx and Velikov, 2014). The issue of momentum profitability across international markets first popularized by Rouwenhorst (1998) and then more recently examined by Asness, Moskowitz, and Pedersen (2013) may benefit from examining trade size portfolios. If the evidence presented in this paper is robust to foreign markets that are known to experience momentum pricing, then these markets provide a fertile ground for future study.

The paper is organized as follows. Section 2 outlines the estimation of the trade size portfolios and the various control variables. Section 3 presents the double sorts of momentum and trade size deciles and terciles as well as the Fama-French factor tests to identify the sources of the momentum profits for each trade size cluster. Section 4 presents robustness checks on our results using trade size deciles in the sort tests and dollar trading volume filters for firms. Section 5 splits the results into NYSE/Amex and NASDAQ listed firms. Section 6 controls for endogeneity and feedback in our trade size results by using trade size portfolios formed before the formation of the momentum portfolios. Section 7 presents results for the pre and post-decimalization in stock quotes that are known to have a large influence on trade size. Section 8 presents pre and post decimalization results using volume turnover as proposed by Lee and Swaminathan (2000) as affecting the profits from momentum trading strategies. Section 9 concludes.

## 2 Trade Size and Firm Attribute Controls

The sample includes all ordinary common stocks listed on the NYSE and the American Stock Exchange (AMEX) in the period January 1983 through December 2012. Transactions data on NASDAQ stocks became available in January 1987, hence those stocks are included in the sample from that time on. Real estate investment trusts, stocks of companies incorporated outside the U.S., and closed-end funds are eliminated from the sample. Return data and unsigned share volume data are from the Center for Research in Security Prices (CRSP) files. We employ characteristic-adjusted returns as developed by Daniel et al. (1997) and Wermers (2003). ${ }^{9}$

Transactions data are obtained from the Institute for the Study of Security Markets (ISSM) and the Trade And Quote (TAQ) data sets. The ISSM data set includes all trades for stocks listed on NYSE/AMEX from 1983 to 1992 and on NASDAQ from 1987 to 1992, while TAQ covers 1993 to present for all exchanges. Trades with irregular terms are excluded and trades are run through a simple price-based error filter to exclude likely erroneous prices. We only focus on the trades database for both ISSM and TAQ negating the need to match the trade with the prevailing quote due to our focus on trade size. We do utilize the quote database to calculate the bid-ask spread applicable to the closing price to estimate the costs of implementing the trade.

The trade size ratios are the sum of intraday 100 -share, 500 -share, and 1000 -share trades over a month divided by the total number of trades that month to derive monthly firm-level ratios within each trade size category. We also analyze trade size increments between 100 and 500 shares, and between 500 and 1000 shares, between 1000 and 5000, 5000-share trades, and greater than 5000 share trades.

To be included in our sample we require a stock to have available information on past returns, trading volume, market capitalization, and stock price. Turnover is calculated as the monthly trading volume divided by the number of shares outstanding at the beginning of the month.

At the beginning of each month, from January 1983 to December 2010 we sort stocks by past

[^5]returns and past trade size. The stocks are assigned to one of three portfolios based on past returns over the previous J months, where J ranges from one to 12 -months, and one of ten portfolios based on each of our three trade size portfolios. We focus our attention on the monthly returns of extreme winner and loser tercile momentum portfolios over the next K months, where K equals 6 and 12 . We also examine $\mathrm{K}=13$ to 24 months after the portfolio formation period and this is replicated for various momentum quintiles and trade size terciles. In all of these tests, we skip the month immediately after the portfolio formation period to avoid any microstructure issues in our K performance periods. Consequently, we use the ( $\mathrm{J}, 1, \mathrm{~K}$ ) nomenclature when describing the separate formation and performance periods.

Similar to Jegadeesh and Titman (1993), the monthly return for the K-month holding period is based on an equally-weighted average of portfolio returns from strategies implemented in the current month and the previous K-1 months. Thus, the monthly return for a six month holding period, averages the portfolio returns from this month's strategy, and then from the prior five months, all on an equally-weighted basis. This allows for a distribution to determine significance for monthly returns.

Finally, we delete all stocks with a price less than $\$ 5$ and greater than $\$ 1,000$ during the last month of the portfolio formation period. The $\$ 5$ price threshold that we impose on our firms mitigates any microstructure issues or regulatory concerns due to price that may impede investability by institutions. In addition, in our robustness tests, we also filter on monthly dollar trade value whereby we delete stocks that do not experience at least $\$ 2,000,000$ in dollar trade value in the last month of the portfolio formation period. This also focuses on investability for institutional traders.

## 3 Initial Trade Size Predictive Sort Results

We discuss the empirical results for trade size-based momentum strategies. In Section 3.1, we present our three trade size ratios across each momentum portfolio and illustrate the association with respect to price and market capitalization. In Section 3.2, we introduce trade size-based price
momentum portfolios, where trade size and momentum are broken down into deciles and terciles, respectively. We then examine the predictive power of trade size over six-months, one-year, and from one-to-two years from the momentum formation period. Section 3.3 examines the sources of the momentum profits for the price momentum trade size portfolios from a Fama-French factor perspective. Here, we attempt to identify the factor loadings that are associated with each trade size portfolio as well as to confirm our characteristic-adjusted return results.

### 3.1 Price Momentum Summary

Table 1 summarizes results from several price momentum portfolio strategies. We present decile and tercile portfolio assignments for the momentum portfolios with these shown in Panels A and B, respectively. Each month, stocks are ranked and grouped into decile (Panel A) or tercile (Panel B) portfolios on the basis of their returns over the previous three, six, nine, and 12 months. We report results for the extreme decile portfolio of losers (R1) and winners (R10), and one intermediate portfolio (R5). For brevity we do not present the remaining portfolios, but the results are consistent with both Jegadeesh and Titman (1993) and Lee and Swaminathan (2000).

For each portfolio, Panel A of Table 1 reports decile momentum portfolios with associated mean returns, the monthly average of the 100 -share, 500 -share, and 1000 -share trade size portfolios. We also present the median size decile of the portfolio based on NYSE/AMEX cutoffs (SzRnk), and the stock price at the end of the portfolio formation period (Price). At the portfolio formation date, stocks in the winner portfolio are larger and have higher price than stocks in the loser portfolio, although firms in either of these portfolios are smaller and of lower price than those of the intermediate portfolio. For instance, $\mathrm{J}=6$ formation period's R 1 price is $\$ 13.99$ and the R10 price is $\$ 33.42$. The lower price registered for the R1 portfolio coincides with negative return, shown as $-7.20 \%$ per month, earned by the loser portfolio, while the higher price given for the R10 portfolio is consistent with the positive return, shown as $10.57 \%$ per month, earned by the winner portfolio.

However, trade size exhibits some dispersion between the extreme and intermediate momentum
portfolios. The extreme momentum portfolios exhibit higher percentages of 500 -share and 1000share trades than are noted for the intermediate portfolios. As shown for $\mathrm{J}=6$, approximately $12 \%$ of all trades are for exactly 1000-shares in the extreme momentum R1 and R10 portfolios, compared to $9 \%$ for the intermediate portfolios. Similar evidence, although more muted, is shown for the 500 -share trades. This differentiation between the extreme momentum portfolios does not extend to the 100 -share trades. For small trade sizes, it appears that they are more concentrated in the intermediate portfolios than the extreme R1 or R10 portfolios. For example, the $\mathrm{J}=6$ formation period shows that approximately $34 \%$ of all trades are for exactly 100 shares for the intermediate portfolio and $32 \%$ for the extreme momentum portfolios. Overall, it appears that larger trades are more prevalent in the extreme momentum portfolios than are smaller trades.

Turning to the average monthly returns where we report the return followed by its t-statistic in parentheses. These results are segregated by four separate holdings period, i.e., $\mathrm{K}=3,6,9$, and 12 months. In unreported results, we show that consistent with prior research that the return breaks after one-year. The extreme momentum portfolios earn highly significant abnormal returns across the spectrum of holdings periods, although the levels are reduced from that reported by Lee and Swaminathan (2000). Regardless, the returns earned by momentum portfolios are robust to the 1983 to 2010 time period and they are economically significant. The $\mathrm{J}=6$ formation period appears to produce the most consistent momentum returns as evidenced by the increased abnormal performance across the four holdings periods relative to the remaining formation periods. For this reason our subsequent tests will focus only on the $\mathrm{J}=6$ formation period.

Panel B of Table 1, reporting momentum terciles, shows similar although less differentiated results across the momentum portfolios than is evident in the decile splits of Panel A. However, the quantitative nature of price, firm size, and trade size appear robust to this split of the momentum portfolios.

### 3.2 Trade Size Based Price Momentum

We first sort the portfolios at the beginning of each month based on their returns over the past $\mathrm{J}=6$ months, divide them into three portfolios spanning losers (L) and winners (W), and then sort the firms within each of the three momentum portfolios into deciles by the trade size categories. We measure trade size as of the last month prior to the performance period. We report the trade size decile as well as the number of firms compromising that portfolio. We also report the average price, firms size, bid-ask spread, and price impact measure. We complete the table by reporting the monthly holding returns for month 1-6, 1-12, and then from months 13-24. For each of these holding periods, the characteristic adjusted returns are reported based on firm size, book-to-market, and momentum. For comparison, the base momentum strategy in addition to our trade size portfolios is also displayed.

We separate our initial sort results into a test using all firms and then again with a filter that eliminates any overlapping firms in the extreme 100 -share, 500 -share, and 1000 -share trade size cluster portfolios for the winner and loser momentum portfolios. This occurs when a firm is assigned to the lowest (or highest) decile for the 100-share trade size portfolio and simultaneously assigned to the highest (or lowest) decile for the 500 and/or 1000-share trade size portfolio(s). Having the same firm in either portfolio may cloud the inferences as to the return predictability specific to each trade size cluster. The resulting sort tests necessarily have a smaller number of firms in the 100 -share portfolios than are evident in the 500 or 1000 -share portfolios. We begin with the filter tests without the overlapping firms as shown in Table 2 and then conclude as a robustness check the tests without any filters on firms in the extreme deciles as shown in Table 3.

For each of these tables, Panel 1 shows the clustered trade size portfolios that contain the 100share trade size ratio, the 500 -share trade size ratio, and the 1000 -share trade size ratio. Panel 2 shows the non-clustered trade size portfolios greater than 500 shares and Panel 3 shows the nonclustered trade size portfolio for trades less than 500 shares but greater than 100 shares. Within each trade size category, decile 1 represents the lowest trade size ratio, while decile 10 represents the highest trade size ratio. We will term the return earned by a hedged momentum portfolio
as W-L, and the return earned by that hedged momentum portfolio for each trade size decile as W-L (Decile 10) or W-L (Decile 1) to reflect either the highest or lowest trade size ratio portfolios, respectively.

We begin with the sample that eliminates any firm in the extreme decile of the 100,500 , and 1000-share portfolios with two-way sort results between momentum and trade size shown in Table 2 . As shown in Panel A of Table 2, across the 1983 to 2010 time period, the momentum portfolio earns a $0.53 \%$ return that is matched by a $0.45 \%$ characteristic-adjusted return for a six month holding period. The return declines to $0.40 \%$ and the characteristic-adjusted return declines to $0.26 \%$ over a one-year period. A break in the return is observed in months 13-24. The bid-ask spreads and the price impact measures are all lower for the winner than for the loser portfolios, although they do eclipse the returns across each of the winner-loser portfolios.

In Panel B of Table 2, the 100-share trade size portfolios show some stylized facts. First, decile 10 trade size portfolios are composed of smaller firms than are decile 1 trade size portfolios, but regardless of the smaller size, decile 10 trade size portfolios (i.e. more dominated by 100 -share trade sizes) are more liquid, with both lower bid-ask spreads and lower price impact costs being registered, than are the shown by decile 1 trade size portfolios. This supports the conventional wisdom that small traders are liquidity providers for the market and this is shown despite the inverse relation with firm size.

Turning to returns, we see that conditional on past returns, portfolios dominated by a large percentage of 100 -share trades earn higher returns than do portfolios dominated by low percentages of 100 -share trades. This is seen for the $\mathrm{K}=1-6$ period where the winner ( W ) trade size decile 10 portfolio earns $1.47 \%$ and the loser ( L ) trade size decile 10 portfolio earns $0.97 \%$ with the winnerloser portfolio (W-L) earning $0.51 \%$. The relatively large return for the loser portfolio appears to indicate that 100 -share traders are bidding up the value of loser firms, exactly opposite to a momentum strategy.

The portfolios that experience much smaller concentrations of 100 -share trades, shown by decile 1 results, indicate the winner trade size decile 1 portfolio earns $1.25 \%$ and the loser decile 1 trade
size portfolio earns $0.56 \%$, with the W-L portfolio earning $0.69 \%$ per month. Momentum returns are more in evidence for stocks that are avoided by 100 -share traders, rather than stocks that have concentrated 100 -share trades.

Across both the Decile 1 and Decile 10 results, Panel B Table 2 shows that the characteristicadjusted returns are substantially less than the unadjusted returns, but the winner-loser (W-L) portfolios are generally of the same level and of the same significance. This is seen in the W-L (Decile 10) characteristic-adjusted return portfolio earning $0.44 \%$ per month and the W-L (Decile 1) portfolio earning $0.61 \%$ per month.

Overall segregating by small trades produces no return improvement when compared to the base momentum strategy. This is shown by the W-L (Decile 10) - W-L (Base) and the W-L (Decile 1) - W-L (Base) where we report statistics to directly compare profits earned in excess of the base momentum strategy. As is shown, none of the 100 trade size portfolios earn a return in excess of the base momentum strategy with the W-L (Decile 10 - Base) portfolio returning $-0.03 \%$ per month and the W-L (Decile 1 - Base) returning a slightly positive premium of $0.16 \%$ per month. These results closely match those of based on the characteristic-adjusted returns. None of these return differences are statistically significant.

Turning to the 500 -share trade size portfolios for the $\mathrm{K}=1-6$ month period, shown in Panel C of Table 2, we see that the momentum winner decile 10 portfolio earns $1.47 \%$ per month, while the momentum loser decile 10 portfolio earns $0.66 \%$. Larger traders appear to better predict those loser stocks that will be bid down in the future as well as predict those winner stocks that will increase in value. This effect is noted nicely in the characteristic-adjusted returns whereby the winner decile 10 portfolio earns a significant $0.27 \%$ return, while the loser decile 10 portfolio earns a significant $-0.50 \%$ return. Portfolios dominated by 500 -share traders do indeed earn a return in excess of risk and it should be noted that both the long side and the short side of the trade are significant in the relation to momentum returns. The decile 10 portfolios experience a lower cost of trade as measured by the price impact measure when compared to the decile 1 portfolios. However, it has been shown by Collin-Dufresne and Fos (2015) that price impact measures are not be associated with informed trading.

The decile 10 winner portfolios do experience a higher level of bid-ask spread costs than do the decile 1 portfolios. The increased level of bid-ask spread costs are consistent with more informed trading. We conclude that 500 -share trades do convey informed trading even though the price impact measure is lower than any other trade size decile. These overall results are consistent with Alexander and Peterson (2007) who argue that 500 -share size trade clusters evidence informed trading as measured by the bid-ask spread, and with Collin-Dufresne and Fos (2015) who find that low frequency price impact measures are not able to detect the presence of informed trading.

Portfolios that are most concentrated (decile 10) in 500-share trades also show better subsequent return performance than do portfolios that are less concentrated (decile 1) in 500-share trades. This is noted in the W-L Decile 10 return that is shown as $0.81 \%$ per month with a highly significant characteristic-adjusted return of $0.78 \%$ per month, while the W-L Decile 1 return is only $0.17 \%$ per month with an insignificant characteristic-adjusted return of $0.11 \%$. Comparing these returns to the base momentum strategy, we clearly see that portfolios dominated by 500 -share trade clusters produce a significant return improvement. This is shown by the W-L (Decile 10 - Base) and the W-L (Decile 1 - Base) where we report statistics to directly compare profits earned in excess of the base momentum strategy. As is shown, the decile 10 portfolios earn a $0.28 \%$ return (characteristicadjusted return of $0.33 \%$ ) more than the base momentum strategy, while the portfolio avoided by 500 -share traders, decile 1 , earns a $-0.36 \%$ return (characteristic-adjusted return of $-0.33 \%$ ) less than the base momentum strategy. The evidence suggests that conditional on past returns, portfolios that are dominated by 500 -share traders earn a far higher level of risk-adjusted (and raw) return than is earned by those portfolios avoided by these traders.

In addition, the portfolios dominated by 500 -share trades see persistence in the abnormal returns for up to one-year after the formation period. As shown in Panel C of Table 2, over one-year the WL decile 10 portfolio return is $0.57 \%$ (characteristic return of $0.45 \%$ ) per month, and this portfolio earns a significant $0.16 \%$ (characteristic return of $0.19 \%$ ) per month return in excess of the base momentum strategy as shown by the W-L (decile 10 - Base) strategy. The persistence in the return "breaks" in month $13-24$ yielding an insignificant negative return. This return pattern is not consistent with the self-attribution bias of Daniel et al. (1998) in that the 500-share traders are
driving the price to its equilibrium level with an insignificant return reaction after one year.
The one-year W-L decile 1 portfolios yield monthly returns of only $0.21 \%$ (characteristicadjusted returns of $0.11 \%$ ), which significantly underperforms relative to the base momentum strategy, given by W-L (Decile 1 - Base) as $-0.19 \%$ per month (characteristic-adjusted return of $-0.15 \%)$.

Examining the 1000-share trades with results, shown in Panel D of Table 2, it is noted that the 1000 -share trade dominated portfolios (i.e. decile 10) are slanted toward much larger firms than the 500 -share trade size portfolios. This is also met with a vastly reduced trading cost, compared to the 500 -share trade size portfolios, as evidenced by either the bid-ask spread or the price impact measure. But, as is shown, the returns and characteristic-adjusted results are substantively similar to those obtained by focusing only on 500-share trades. Over the first six months, the W-L Decile 10 portfolios earn returns of $0.79 \%$ per month with characteristic-adjusted returns of $0.74 \%$ per month. Comparing these returns to the base momentum strategy, the W-L (Decile 10- Base) earns a $0.26 \%$ return in excess of the base momentum strategy, while the W-L (Decile 1 - Base) portfolio earns $0.15 \%$ per month less than the base momentum strategy. The characteristic-adjusted returns report virtually identical findings. Extending the analysis to the first twelve months, shows a significant return of continuation with returns of $0.63 \%$ (characteristic-adjusted returns of $0.37 \%$ ) per month. Interestingly, the return over months $13-24$ is slightly negative, but it demonstrates no break in the return sequence. Portfolios dominated by 1000 -share trades appear to drive returns to their intrinsic value with no further return predictability. This is most consistent with the conservatism bias expounded by Barberis et al. (1998).

This is reinforced by a comparison to the base momentum strategy where we see no significant improvement in holding the 1000-share portfolios over twelve months. The W-L (Decile 10 - Base) returns are $0.13 \%$ per month (characteristic-adjusted returns of $0.11 \%$ ) in excess of the base momentum strategy up to twelve months are $0.11 \%$ per month. The decile 1 performance over the base momentum strategy is also insignificant.

In sum, the results indicate momentum returns over the 12 months subsequent to the portfolio
formation are more pronounced for portfolios that are favored by 500 and 1000-share traders than for 100 -share traders. We find that portfolios dominated by 500 or 1000 -share traders earn abnormal returns that are nearly double that earned by the base momentum strategy. This out-performance does not extend to the 100 -share trade size portfolios.

We now examine the performance of the non-clustered large and small trade portfolios which are shown in Panels 2 and 3 of Table 2, respectively. For all the non-clustered large trade portfolios, shown in Panels E through G, we see that enhanced momentum performance is observed for the decile 1 portfolios rather than the decile 10 portfolios. For instance, for trade sizes between 500 shares and 1000 shares, shown in Panel E, we see six-month W-L decile 10 portfolio characteristicadjusted returns of $0.21 \%$, but the decile 1 characteristic-adjusted returns of $0.48 \%$. However, none of these portfolios out-performs the base momentum return as shown by either the W-L (Decile 10 - Base) or the W-L (Decile 1 - Base) returns. These results indicate that focusing on non-clustered trades greater than 500 shares yields no significant enhancement in return predictability.

Exemplifying the importance of the 500 -share trade categories on the large trade portfolio, we see that the portfolio that combines all large trades into one portfolio, shown in Panel H of Table 2, shows returns quite similar to the 500 -share portfolio, but with greatly reduced liquidity costs. While this portfolio reduces return predictability, it does support the importance of the 500 and 1000 -share trade sizes in return predictability.

Finally, Panel 3 of Table 2 reports non-clustered small trade portfolios, comprised of trades greater than 100 but less then 500 shares. As is shown, there is little return predictability over that contained in the base winner-loser momentum portfolios. None of the W-L (Decile 10 or 1) portfolios are significantly different from the base momentum strategy.

As a robustness check, we now use all firms regardless of whether the same firm "overlaps" in the extreme portfolios for the 100,500 , and 1000 share trade size cluster portfolios. These results are shown in Table 3. We see that our prior results are conclusively verified using all firms with balanced sorted portfolios. As shown in Table 3, we see again that portfolio dominated by 100share trades (W-L Decile 10 - Base) significantly underperforms relative to the base momentum
strategy earning $-0.17 \%$ less per month (over the first six-months), but that the stocks avoided by 100 -share traders (W-L Decile 1 - Base) earn $0.25 \%$ monthly risk-adjusted return more than the base momentum strategy. Evidently the overall 100-share traders are contrarian traders relative to a momentum strategy.

This is in marked departure to the behavior of the 500 and 1000-share clustered trades. Over the first six months, Panels C and D of Table 3 show that the W-L (Decile 10 - Base) portfolio of the 500 -share clustered trades earn $0.31 \%$ characteristic-adjusted monthly returns in excess of the base momentum strategy, and the W-L (Decile 10 - Base) portfolio of the 1000-share clustered trades earn $0.27 \%$ characteristic-adjusted monthly return in excess of the base momentum strategy. The W-L (Decile 1 - Base) portfolios for both the 500 and 1000-share portfolios significantly underperform the base momentum strategy. For instance, the 500 -share W-L (Decile 1 - Base) portfolio earns $-0.30 \%$ less that he base momentum return on a characteristic-adjusted return basis. These returns are significant persist over the one-year interval and then "break" past the one-year time period.

Examining the non-clustered trades, shown in Panels 2 and 3 of Table 3, we see clearly that only the amalgamated trades of greater than and equal to 500, yielding a W-L (Decile 10 - Base) characteristic-adjusted return of $0.24 \%$ per month over the first six months, produces significantly improved returns over the base momentum strategy, while none of the non-clustered trade groups produce returns in excess of the base momentum strategy.

On the one hand, these results support that notion that 100 (and non-clustered trades) do not follow a momentum strategy, evidencing a contrarian trading behavior. Indeed, the momentum stocks that 100 -share traders avoid (decile 1) most readily demonstrate momentum returns, while the most concentrated (decile 10) holdings are insignificantly associated with subsequent momentum returns. On the other hand, the results also support the claim that the 500 and 1000 -share trade clusters do follow the momentum strategy, whereby the portfolios that are most concentrated in 500 or 1000-share trades evidence the most persistent momentum returns and the portfolios that evidence the least concentration of 500 or 1000 -share trades produce the least momentum profits.

For the remainder of our tests we will focus only on the no-overlapping sample to ensure that
all of the inferences attributed to the 100,500 , or 1000 -share portfolios are specific to only these trade size clusters.

### 3.3 Fama-French Regressions with Price Momentum and Trade Size Based Portfolios

Table 4 provides additional evidence on the source of abnormal returns for the various price momentum-trade size strategies. In this table, we report the results from time-series regressions based on the Fama and French (1993) three-factor model where we run the following time-series regression using monthly portfolio returns:

$$
r_{i}-r_{f}=a_{i}+b_{i}\left(r_{m}-r_{f}\right)+s_{i} S M B+h_{i} H M L+\epsilon_{i}
$$

where $r_{m}-r_{f}$ is the excess return on the one-month value-weighted return on the market, HML is the high-minus-low book-to-market (value) factor, and SMB is the small-minus-big size factor. The term $a_{i}$ represents the abnormal performance for each portfolio. All returns and market return are stated on a percentage basis. The coefficients, $b_{i}, s_{i}$, and $h_{i}$ are the corresponding factor loadings and they are stated on a percentage basis. We report the portfolio formation and holding periods for the $\mathrm{J}=6, \mathrm{~K}=6$ frequency. We skip one-month after the $\mathrm{J}=6$ portfolio formation period and separate our findings with respect to the portfolios that concentrate on trade size $=100$ shares $($ Panel A), trade size $=500$ shares $($ Panel B), and trade size $=1000$ shares (Panel C). The subsequent results only focus on the 500 -share and the 1000 -share large trade portfolios because the prior results showed significance for only these two trade sizes. We retain the 100 -share portfolio because it is the largest (by number of trades) trade size in the market as well as to provide a comparison to the large trade size portfolio results. Within each of these categories, we present the lowest decile (D1), the middle decile (D5), and highest decile (D10) trade size ratios. For each trade size portfolio, we first present the estimated intercept coefficient followed by the estimated coefficients for $b_{i}, s_{i}$, and $h_{i}$, respectively. We also report the goodness-of-fit with the adjusted$R^{2}$ for each regression. The estimated intercept coefficients from these regressions $\left(a_{i}\right)$ are the
risk-adjusted return of the portfolio relative to the three-factor model.

As is shown in Table 4, the abnormal performance measures embodied in the intercept estimates confirm that our prior results are not dependent on the use of characteristic-adjusted returns and they cannot be explained by the standard Fama-French factors. For the 100 -share trade size category, shown in Panel A, we note that while the winner-loser portfolios are all positive and significant for each of the trade size deciles, there is no differential in abnormal returns between deciles 10 and 1 . This is seen more clearly by the W-L D10-D1 portfolio that shows an aggregate momentum return ${ }^{10}$ earned by investing across the high and low deciles of 100 -share trades to be $-0.01 \%$. This alludes to the fact that sorting on the 100 -share trade size does not yield any abnormal returns over and above the base momentum strategy. This reinforces our average return and characteristic-adjusted return results that shows that small trades do little to differentiate performance in momentum portfolios.

The loadings on the SMB factor show that across all the momentum portfolios, the D10 portfolios are more slanted to small stocks than are the D1 portfolios. The loadings on the HML factor show no differential between value/glamor except for the winner momentum portfolio where we see a clear value stock preference for the D1 portfolio.

These results are in marked departure from those of the larger trade size clusters. As shown in Panels B and C, we now find significant dispersion in abnormal performance across the decile 1 and decile 10 trade size portfolios for the winner-loser (W-L) momentum strategy. In Panel B, the 500-trade share portfolio based W-L decile 10 portfolio yields abnormal returns of $0.91 \%$ while the decile 1 portfolio yields abnormal returns of $0.24 \%$ per month, respectively, with a significant return differential of $0.67 \%$ per month. Substantively similar results are found for the 1000 -share trade size portfolio. As shown in Panel C, the 1000-share based W-L decile 10 portfolio yields abnormal returns of $0.92 \%$ while the decile 1 portfolio yields abnormal returns of $0.45 \%$ per month, respectively, with a significant return differential of $0.47 \%$ per month.

The SMB loadings for the 500 and 1000 -share trade portfolios, shown in Panels B and C ,

[^6]demonstrate the the D10 deciles are more concentrated in smaller firms across the winner-Loser momentum portfolios than are the D1 decile portfolios. Given that most of the abnormal returns from the momentum strategy accrue to the D10 decile, it appears that the 500 and 1000 -share traders are concentrating on somewhat smaller firms to derive larger abnormal profits. It is surprising that even the 1000 -share trades, shown in Panel C, are also concentrating on smaller firms. The distinction in firm size across the D10 and D1 trade size deciles is made evident by the monotonic trend across the trade size deciles that exhibits robust significance for the decile 10 - decile 1 portfolios, but with no significance across the winner - loser portfolios. The results indicate that decile 10 portfolios, or those portfolios dominated by large trades, earn more of the improvement in performance as noted by the abnormal returns do so by focusing on smaller firms.

Turning now to the HML loadings, we see for both the 500 -share and 1000 -share trade size portfolios that the D1 portfolio is more weighted to value firms than is the D10 portfolio. In essence, the decile 1 portfolios are comprised of more value stocks, or ones with higher book-to-market valuations. Conversely, the decile 10 portfolios demonstrate much lower loadings on book-to-market indicating they are slanted more towards glamour stocks, or ones with lower book-to-market valuations. This is reinforced by the negative coefficients for the decile 10 - decile 1 portfolios.

### 3.4 Seasonality and Price Momentum

Jegadeesh and Titman (1993) find a striking seasonality in momentum profits, a result that is substantiated in Jegadeesh and Titman (2001). They document that the losers significantly outperform the winners in January, while the winners outperform losers in all months except January. By extension, they also document that momentum profits are not in evidence in January, but are robust from February to December. We examine the out-of-sample momentum performance whereby we interact trade size with that of the momentum anomaly in January versus the rest of the year. We first sort the portfolios at the beginning of each month based on their returns over the past $\mathrm{J}=6$ months, dividing them into three portfolios spanning losers (L) and winners (W), and then sort the firms within each of the three momentum portfolios into deciles by the trade size
categories. We measure trade size as of the last month prior to the performance period. We then separate the performance exclusively in January and then from February to December.

The results are presented in Table 5. As shown in Panel A of Table 5, the January momentum average return underperforms relative to the rest of the year, a result consistent with the prior literature. The January returns are $-0.96 \%$ per month, while the February to December returns are $0.66 \%$ per month. This underperformance extends to the characteristic-adjusted returns that shows returns in January of only $0.01 \%$ per month compared to $0.49 \%$ per month across the rest of the year. This general results is consistent with the original findings of Jegadeesh and Titman (1993).

Focusing on 100-share trades, shown in Panel B of Table 5, we see that in January, the W-L decile 10 portfolios demonstrate a $-0.95 \%$ return per month ( $0.19 \%$ characteristic-adjusted return), while the W-L Decile 1 portfolios show a $-0.15 \%$ return per month ( $0.62 \%$ characteristic-adjusted return). This level of return is not statistically different from the base momentum strategy on a characteristic-adjusted return basis as evidenced by either the W-L (Decile 10 or Decile 1) - W-L (Base). However as found for the base momentum strategy, the February to December period shows momentum profits are more in evidenced in the portfolios that are avoided by the 100 -share trades. The Decile 10 risk adjusted returns are $0.32 \%$ while the Decile 1 risk-adjusted returns are $0.71 \%$. These results mirror the monthly average returns. The Decile 1 portfolios earn $0.21 \%$ more than the base momentum portfolio, while the decile 10 portfolios earn $-0.17 \%$ less than the base momentum portfolio. The returns demonstrate the contrarian nature of the one round lot trades.

The 500-share trades, shown in Panel C of Table 5, demonstrates a distinctly different return behavior. In January, the W-L decile 10 portfolios demonstrate a - $0.37 \%$ return per month ( $0.70 \%$ characteristic-adjusted return), and the W-L Decile 1 portfolios show a $-0.97 \%$ return per month (-0.03\% characteristic-adjusted return). While the risk-adjusted return for the Decile 10 portfolio is insignificant, this most probably results from a low power test. Relative to the Base momentum strategy, the W-L (Decile 10) portfolio represents a substantial improvement in the momentum performance. The W-L (Decile 10) - W-L (Base) is $0.69 \%$ (significant at the $5 \%$ level), while the W-L (Decile 1) - W-L (Base) is $-0.04 \%$.

The performance is improved in the January to December period, but on a risk-adjusted basis the overall return is surprisingly similar. The risk-adjusted return is $0.77 \%$ and significant at the $1 \%$ level. Comparing this to the $0.70 \%$ risk-adjusted return found in January, it can be easily argued that portfolios dominated by 500 -share trades do not experience a significant fall-off in performance in January as found in the base momentum strategy. As expected, the Decile 10 portfolios perform much better than the Decile 1 portfolios relative to the base momentum strategies as evidenced in the W-L Decile 10, 1) - W-L (Base) portfolios where the Decile 10 portfolios earn $0.28 \%$ more than the Base momentum strategy and the Decile 1 portfolios earn $-0.32 \%$ less than the base momentum strategy.

The 1000-share trades, shown in Panel D of Table 5, shows a return behavior that is similar to that exhibited by the 500 -share trades. In January, the W-L decile 10 portfolios demonstrate a $-0.50 \%$ return per month ( $0.50 \%$ characteristic-adjusted return), while the W-L Decile 1 portfolios show a $-0.98 \%$ return per month ( $0.07 \%$ characteristic-adjusted return). Relative to the Base momentum strategy, the W-L (Decile 10) - W-L (Base) is $0.49 \%$ (significant at the $10 \%$ level), and the W-L (Decile 1) - W-L (Base) is $0.06 \%$.

As found previously, the January to December period, the Decile 10 portfolio shows a riskadjusted return of $0.74 \%$ and significant at the $1 \%$ level, while the Decile 1 portfolio shows a risk-adjusted return of $0.30 \%$ per month. These returns as closely aligned to those noted for the 500 -share trades. As found for the 500 -share trades, the Decile 10 portfolios perform much better than the Decile 1 portfolios relative to the base momentum strategies. The W-L Decile 10) - W-L (Base) portfolio earns $0.25 \%$ more than the Base momentum strategy and the W-L Decile 10) -W-L (Base) portfolio earns $-0.20 \%$ less than the Base momentum strategy.

The results indicate that the 500 and 1000-share trade dominated portfolios trade with the momentum anomaly, and do so with vastly improved performance relative to the contrarian 100share trade size portfolios. The results also indicate that the seasonality in risk-adjusted returns is muted for the 500 share portfolios with January risk-adjusted returns mirroring those obtained from February to December. The performance of the portfolio dominated by 500 -share, and to a lesser extent the 1000 -share, trades does not appear to exhibit the severity in the seasonality.

## 4 Robustness Tests: Trade Size Quintiles and Price Momentum and Dollar Volume Trading Filters

We employ robustness checks on our results by splitting trade size into quintiles as opposed to deciles, but keeping momentum separated into terciles. This is done in order to increase the cross-sectional coverage of firm as well as reduce the associated liquidity costs of each trade size portfolio. We also increase the filtering of firms by requiring all firms have at least $\$ 2,000,000$ in dollar volume in the last month of the portfolio formation period. Section 4.1 presents the trade size quintile sort results and Section 4.2 presents the additional dollar volume filter results. This dollar volume limit is in addition to the $\$ 5.00$ price limit that is already imposed on the portfolios. This additional filter will ensure that the firms included in the sample are investable by institutions. We again focus on the original momentum tercile and trade size decile sorts that formed the basis of our prior tests. Also, as with all of our tests we eliminate any firm that is in both the decile (quintile) one and/or decile 10 (quintile 5) of both the 100 and 500-share (and 1000-share) trade size portfolios. We concentrate only on the characteristic-adjusted returns.

### 4.1 Trade Size Quintiles and Price Momentum

Table 6 reports results on the basis of a two-way sort between momentum and trade size. In this test, we sort momentum into terciles and trade size into quintiles in order to increase the span of firms in the sample as well as reducing the associated liquidity costs of the traded sample. As previously described, we first sort the portfolios at the beginning of each month based on their returns over the past $\mathrm{J}=6$ months and then divide them into three portfolios, spanning losers (L) and winners (W). We then sort the firms within each of the three momentum portfolios into quintiles by either the 100 -share trade size ratio, the 500 -share trade size ratio, or the 1000 -share trade size ratio. Within each trade size category, quintile 1 represents the lowest trade size ratio, while quintile 5 represents the highest trade size ratio. We will term the return earned by a hedged momentum portfolio as W-L, and the return earned by that hedged momentum portfolio for each trade size quintile as W-L (Quintile 5) or W-L (Quintile 1) to reflect either the highest or lowest
trade size ratio portfolios, respectively. To avoid repeated sampling, we eliminate any firm that is in the extreme trade size portfolios for both the 100 -share, 500 , and 1000 -share trade size portfolios. This is accomplished after our initial sort on trade size, hence our final sample will have unbalanced numbers of firms within each momentum/trade size portfolio. This filter results in a steep falloff in the number of firms but the restricted sample is of much higher liquidity. We report the trade size quintile as well as the number of firms compromising that portfolio. We also report the average price, firms size, bid-ask spread, and price impact measure.

As shown in Panel A of Table 6, across the 1983 to 2010 time period, the momentum portfolio remains priced earning a monthly $0.46 \%$ characteristic-adjusted return for a six month holding period. The characteristic-adjusted monthly return declines to $0.27 \%$ over a one-year period. A break in the return is observed in months 13-24. The bid-ask spreads and the price impact measures are all lower for the winner and loser portfolios, although they now do not eclipse the returns across each of the winner-loser portfolios.

In Panel B of Table 6 we review the findings for the 100 -share size portfolios. As shown, we see a concentration of momentum profits for the Quintile 5 portfolios where the winner momentum portfolio earns $0.33 \%$ per month and the loser momentum portfolio earns $-0.19 \%$ per month. This translates to a W-L momentum return across the Quintile 5 portfolios of $0.53 \%$. The Quintile 1 portfolios earn less than the Quintile 5 portfolios yielding a W-L Quintile 1 characteristic-adjusted return of $0.34 \%$. However, compared to the base momentum strategy, these returns are not statistically different from those earned by the base strategy as shown by the insignificant W-L (Quintile 5 or 1 ) - W-L (Base) portfolio returns. The evidence shows that concentrated trades by 100 -share traders have no incremental impact on the subsequent price performance of momentum strategies.

However, 500 -share trade clusters exhibit a far different behavior. Panel C of Table 6 shows that the W-L (Quintile 5) portfolio earns a significant characteristic-adjusted return of $0.87 \%$ per month over six months and a significant $0.53 \%$ per month over twelve months. This is $0.42 \%$ more than that earned by the base momentum strategy over six months and $0.26 \%$ more than that earned by the base strategy over twelve months. The portfolios avoided by large traders do not perform as well. The quintile 1 portfolios earn an insignificant characteristic-adjusted return of $0.18 \%$ over six
months and an slightly larger characteristic-adjusted return of $0.19 \%$ over 12 months.

The outperformance of the Quintile 5 portfolios lies mainly in the short position. The winner quintile 5 portfolio earns $0.30 \%$ per month, while the loser quintile 5 portfolio earns $-0.57 \%$ per month. The 500 -share traders appear to better predict those stocks that will decline in value for the loser momentum portfolio.

Finally, Panel D of Table 6 shows the 1000 -share portfolios exhibit similar behavior to that reported by the 500 -share trade portfolios, except that the performance over the base momentum strategy is not as robust past the first six months. As shown in Panel D of Table 6, the momentum characteristic-adjusted return earned by the portfolios dominated by 1000 -share trades, W-L (Quintile 5), is significant at $0.81 \%$ per month, as is the twelve month return of $0.40 \%$ per month. Over the first six months, the quintile 5 performance is $0.35 \%$ more than that earned by the base momentum strategy and over the twelve month period, the quintile 5 performance is $0.13 \%$ more than that earned by the base momentum strategy. Conversely, the quintile 1 portfolio earns significantly less than the base strategy over the six month period, shown as W-L (Quintile 1) - W-L $($ Base $)=-0.25 \%$.

### 4.2 Dollar Trade Size Filtered Trade Size and Price Momentum

For these set of results in addition to our prior price filter of prices greater than $\$ 5.00$, we also require a stock to have at least $\$ 2,000,000$ in dollar trading volume in the month prior to the momentum trade. This filter is added to ensure investability. We again use momentum terciles and trade size cluster deciles for the sort tests.

Panel A of Table 7 shows that the base momentum strategy is remarkably consistent earning a characteristic-adjusted return of $0.43 \%$ and $0.24 \%$ per month over six and twelve months, respectively. Filtering on dollar volume does little to affect the performance of the base momentum strategy.

The small trade size category, Panel B of Table 7 shows that again, the 100 -share trade size portfolios do not outperform the base momentum strategy. The characteristic-adjusted returns are
closely aligned to the base momentum strategy. For instance, over the first six-months the W-L (decile 10) characteristic-adjusted returns are $0.48 \%$ and the W-L (decile 1 ) characteristic-adjusted returns are $0.51 \%$. But the performance over the base momentum strategy is not significantly different from zero. This is given as $0.05 \%$ and $0.08 \%$, respectively for the the W-L (Decile 10 Base) and the W-L (Decile 1 - Base) strategy, i.e. the returns over the base momentum strategy.

The 500-share trade portfolios are dramatically different in performance. Panel C of Table 7 shows significance for decile 10 portfolios in the characteristic-adjusted returns over six-months and up to one-year after the formation period. Indeed, the returns are more robust than previously reported using only the $\$ 5.00$ price filter. Now the Decile 10 trade size cluster portfolios earn characteristic-adjusted returns of $1.02 \%$ per month over six months. This is a significant improvement over the base momentum strategy. Examining the W-L (Decile 10-Base) results we see that we earn more than $0.59 \%$ more on a characteristic-adjusted return basis.

These returns are far in excess of the underlying costs of immediacy. The round-trip costs of trade are $3.37 \%$ (Decile 10 bid-ask costs of $1.60 \%$ plus $1.77 \%$ ). The after-liquidity cost profits are more than $2.75 \%$ over six months. Given that these cost estimates are undoubtedly higher than would be experienced by institutions, the profits from this strategy appear substantial. These results are in stark contrast to those reported by the Decile 1500 trade-size portfolios. For these portfolios, the characteristic-adjusted returns over 6 -months and twelve-months are $0.02 \%$ and $0.05 \%$, respectively, neither of which are significant. The lack of significance extends to the twelve month performance period as well.

Finally, the 1000 -share trade size portfolios, shown in Panel D of Table 7 illustrate that the decile 10 portfolios see a significant characteristic-adjusted return of $0.84 \%$ over six months as well as a significant monthly characteristic-adjusted return of $0.39 \%$ over 12 months. However, there is no significance in the return for months 13-24. The lack of a significant break in the returns is an indication that 1000 -share traders, as do the 500 -share traders, push prices to their fundamental value and then the returns are stabilized at that level.

In terms of return improvement over the base momentum strategy, we do observe a dissipation
of momentum profits after only six months for the 1000 -share portfolios. Over the first six months, the W-L (decile 10 - Base) monthly characteristic-adjusted monthly returns are $0.42 \%$, while the 12 month characteristic-adjusted return is only $0.15 \%$ which is not statistically significant. These results would indicate that the 1000 -share portfolio more correctly prices any underpricing inherent in momentum portfolios, and they do so much faster than the 500 -share portfolios, removing any improvements in six months versus the 12 months experienced by the 500 -share trade size portfolios.

## 5 NYSE/Amex Versus NASDAQ Splits

In this section we examine NYSE/Amex and NASDAQ separately in order to assess the effects of exchange listing on our results. This test will ensure that a few small stocks are not driving the results. As implemented previously, we delete any firm with a price less than $\$ 5.00$ as determined at the end of the portfolio formation period. We again focus on the original momentum tercile and trade size decile sorts that formed the basis of our prior tests. This will ensure sufficient power for the tests. Also, as with all of our tests we eliminate any firm that is in both the decile one and/or decile 10 of both the T100 and T500-share trade size portfolio. Panel 1 of Table 8 outlines the results for NYSE/Amex listed firms and Panel 2 of Table 8 outlines the results for only NASDAQ listed firms.

### 5.1 NYSE/Amex listed Firms: Trade Size and Price Momentum

Panel 1 of Table 8 shows that the base momentum return strategy declines considerably when examining only NYSE/Amex firms. The characteristic-adjusted return falls to $0.27 \%$ per month over months 1-6 and the characteristic-adjusted return is insignificant over the 12 -month trading horizon falling to just $0.18 \%$ per month. The characteristic-adjusted return in months $13-24$ is negative, but insignificant, and recorded at $-0.07 \%$.

The small trade size category, Panel B of Table 8 again shows that portfolios that are dominated by 100 -share trade sizes underperform the base momentum strategy. Indeed, the Decile

10 portfolio earns a characteristic-adjusted return of $0.09 \%$ less than the base momentum strategy, while the decile 1 small trade size portfolio earns a characteristic-adjusted return of $0.12 \%$ per month more than the base momentum strategy. The profitably of momentum strategies, even at the NYSE/Amex level is not dependent on the trade behavior of small traders. Rather, the performance appears to be concentrated in the portfolio shunned by small traders.

Examining the performance of the larger trade size portfolios reveals results consistent with the prior findings. Namely, portfolios dominated by 500 and 1000-share trades exhibit very consistent significant characteristic-adjusted returns for up to one-year after the portfolio formation period. In panel C of Table 8, we see that the 500-share trade size portfolios earn significant monthly riskadjusted returns of $0.72 \%$ over six months and $0.34 \%$ over 12 months. These results extend to the 1000-share trade size portfolios that also show robust performance up to one-year after the portfolio formation period. The break in the return after one-year is insignificant, and it demonstrates the same behavior for either the 500 or 1000 -share trade size portfolios.

The returns earned by either large trade size portfolio indicates an economic and statistical improvement over the base strategy as evidenced by the W-L (decile 10 - base) return. The 500share portfolios, shown in Panel C of Table 8, demonstrates a per month characteristic-adjusted return improvement of $0.46 \%$ and $0.16 \%$ over six and 12 -months, respectively. Also, the decile 1 500 -share portfolio clearly under-performs relative to the decile 10500 -share portfolio with regard to the base momentum strategy. The decile 1500 -share portfolio earns $0.19 \%$ per month less than the base momentum portfolio with insignificance found at 12 months.

The 1000-share portfolio, presented in Panel D of Table 8, shows that only the W-L (Decile 10 - Base) has any improvement in performance over the base momentum strategy. The six month return, recorded at $0.31 \%$ per month, is significantly above that earned by the base momentum strategy. The out-performance is robust to a 12 month trading horizon with insignificant monthly characteristic-returns of $0.17 \%$. The performance of the 1000 -share trade size portfolios mimics that of the 500 -share trade size portfolio indicating that for NYSE/Amex firms, focusing on either trade size will out-perform that base momentum strategy. Focusing on 500 and 1000 -share trade size clusters is a viable avenue to better exploit the momentum profits.

### 5.2 NASDAQ listed Firms: Trade Size and Price Momentum

Panel 2 of Table 8 shows that the base momentum strategy strengthens by focusing only on NASDAQ firms. Persistence of the momentum profits is now evident up to a 12 -month trading horizon regardless of raw or characteristic-adjusted returns. The characteristic-adjusted return is highly significant at $0.68 \%$ over six months, but falls to $0.37 \%$ over the whole 12 -month trading horizon period. However, the break in the one-year performance is marginally significant indicating that NASDAQ firms exhibit characteristics consistent with the self-attribution behavioral bias (Daniel et al., 1998).

As was found with the NYSE/Amex firms, the small trade size portfolios, shown in Panel F of Table 8, do not exhibit any meaningful difference from that earned by the base momentum strategy. As shown by the W-L (Decile 10 - Base) or the W-L (Decile 1 - Base), the abnormal returns are essentially zero with the maximum recorded at at $0.18 \%$ per month (over the twelve-month holding period). Small one-round lot based trade portfolios do not substantially improve the performance of NASDAQ-based momentum portfolios.

The performance of the larger trade portfolios is distinctly improved relative to the small trade portfolios. As shown in Panel G of Table 8, the 500-share trade size portfolio earns $0.96 \%$ characteristic-adjusted return per month in the first six months and $0.56 \%$ per month over a oneyear holding period. The W-L (Decile 10 - Base) characteristic-adjusted return is significant at $0.28 \%$ per month over six months and $0.19 \%$ per month over one-year. However, the negative break experienced one year after the portfolio formation is insignificant indicating that the initial underreaction is not matched by a subsequent overreaction. Rather, the results appear to indicate that for NASDAQ firms, 500-share traders price momentum to its intrinsic level and do so over one-year.

These results are matched by the Decile 10, 1000-share trade size portfolio that experiences a significant characteristic-adjusted return of $0.87 \%$ and $0.50 \%$ per month over the six and twelve month trading horizons after the portfolio formation period, respectively. The W-L (Decile 10 Base) strategy yields a significant return improvement with the six month characteristic-adjusted
return seen as significant at $0.19 \%$ per month, but falls to insignificance over the twelve month period. It again appears that portfolios dominated by much larger trade sizes price momentum more quickly with returns over the base momentum strategy only significant up to six months and these portfolios experience some return continuation as evidenced by the positive, although insignificant, return in months 13-24 after the portfolio formation period.

## 6 Trade Size Portfolios Formed Before Momentum Portfolios

We examine the effect of forming trade size portfolios prior to the price momentum portfolio formation. The measurement of trade size prior to the momentum portfolio formation will address any concern that momentum itself may cause trade size portfolios to develop. We measure trade size and assign them into portfolios in the month prior to the $\mathrm{J}=6$ price momentum portfolios. As implemented previously, we delete any firm with a price less than $\$ 5.00$ as determined at the end of the portfolio formation period and we form momentum tercile and trade size decile portfolios in our sort tests.

Due to our data filter for non-overlapping firms, we reexamine the base momentum strategy. Panel A of Table 9 shows that the base momentum return strategy continues to produce significant characteristic-adjusted, $0.41 \%$, returns over the six month holding period. The significance of the base momentum strategy extends to the 12-month holding period with characteristic-adjusted returns of $0.26 \%$ per month.

Panel B of Table 9 shows 100-share trade size portfolio performance. Interestingly, the Decile 10, 100 -share portfolio evidences a significant characteristic-adjusted return of $0.34 \%$, but this is lower than the return earned ay the Decile 1 portfolio recored at $0.46 \%$ per month over six months. But neither of these returns are statistically distinguishable from the base price momentum strategy. This is represented by an insignificant W-L (Decile 10 - Base) six month portfolio return that is shown as $-0.07 \%$ for the decile 10 and $0.05 \%$ for the decile 1 portfolio. Overall then results indicate that forming trade size portfolios prior to the momentum portfolio formation affects the subsequent performance of the 100-share trade size portfolios, but again there is no significant enhancement
over the base momentum strategy, i.e. the 100 -share trade size portfolios simply replicate the performance of the base momentum strategy.

Panels B and C of Table 9 show the 500 and 1000-share trade size portfolios. As demonstrated in Panels B and C, for both the 500 or 1000-share trade portfolios, the characteristic-adjusted returns remain significant for up to one-year after the momentum portfolio formation period. Neither portfolio experiences a significant break in subsequent one-year holding period. For instance the W-L Decile 10, 500 -share portfolio earns a characteristic-adjusted return of $0.76 \%$ per month and a $0.42 \%$ per month of the six and twelve month periods, respectively. The W-L Decile 1 portfolios earn substantially less than the Decile 10 portfolios. It is notable that the Decile 10 portfolios, for both the 500 -share and the 1000 -share trade size categories, see a consistent return improvement over the base momentum strategy. The 500 -share W-L (Decile 10 - Base) shows a characteristicadjusted return improvement of $0.35 \%$ for six months and a return improvement of $0.17 \%$ for twelve months over the base momentum strategy. This is substantively similar to that earned by the 1000 -share trade size portfolio, except that the W-L Decile 10 - base 1000-share portfolios are insignificantly different after the first six months.

In conclusion, the results indicate that focusing on larger trade portfolios shows continued improvement in the momentum profits even if they are formed before the start of the momentum portfolio formation. This result does not imply that large trades cause momentum, but rather that large trades are not exhibiting any feedback effect from momentum whereupon large trade could be concentrating their trades on the previously disclosed momentum. Portfolios dominated by 500 (and 1000)-share trades appear to demonstrate persistent improvement in the momentum profits that are maintained for up to one-year. Importantly, the results do not evidence any support for an endogeneity or feedback bias that may cloud the inferences concerning the return predictability of large trade portfolios.

## 7 Decimalization Effects

In this section we examine the effect of momentum crashes on the relation between trade size and price momentum. Daniel and Moskowitz (2013) argue that momentum crashes have become more frequent after the year 2000. This coincides with the crash in NASDAQ stocks in 2000 and the Great Recession in 2007 and 2008. The crashes in momentum coincide with the decimalization in stock quotes, although the two are not explicitly tied together. We note that decimalization has significantly affected trade size. The NYSE Fact book reports statistics that show an average trade sizes falling dramatically after stock decimalization. The average trade size in 1999 for NYSE-listed firms was 1,205 shares per trade. In 2004, the average trade size was significantly reduced to just over 390 shares per trade, while in 2010 the average trade size had dwindled to 220 shares per trade and in 2014 the average trade size was approximately 140 shares per trade. To control for the drop in average trade size and to incorporate the cases of momentum crashes, we separate our sample into two periods with the first based on the period 1983 to 2000 and the second based on the period 2001 to 2010.

### 7.1 Pre-Decimalization Period

Panel 1 of Table 10 shows that the base momentum return strategy during the 1983 to 2000 is very profitable yielding characteristic-adjusted returns of $0.62 \%$ per month and $0.39 \%$ per month for the six and twelve-month periods, respectively. The return then breaks after one year and experiences a $-0.16 \%$ per month decline.

The small trade size category, Panel B of Table 10 shows that W-L Decile 10, 100-share portfolios are principally composed of much larger stocks. The winner stocks have average market valuations of $\$ 2.757$ billion while the loser portfolio has a market valuation of $\$ 1.721$ billion. The W-L decile 10 portfolios yield significant characteristic-adjusted returns of $0.66 \%$ and $0.45 \%$ over six months and one year, respectively. The Decile 1 T100 portfolio, composed of much smaller stocks, experiences higher characteristic-adjusted returns recorded at $0.83 \%$ and $0.47 \%$ per month over the first six and twelve months, respectively. However, the W-L (Decile 10 - Base) or the W-L (Decile 1 - Base)
do not earn a robust return improvement over the base momentum strategy.
The larger trade categories, i.e. the 500 and the 1000 -share trade size portfolios given in Panels C and D, respectively, show that the Decile 10 portfolios are composed of much smaller stocks with the 500-share portfolios dominated by very small stocks. Regardless, the W-L (decile 10) portfolios earn a significant return of $0.87 \%$ per month over the first six months and the significance in the characteristic-adjusted return measured at $0.52 \%$ per month, persists for up to one-year. On the one hand, the Decile 10, 500 -share portfolio returns, shown in Panel C of Table 10, break to a negative and significant $0.26 \%$ monthly return over the 13-24 month period. This behavior is consistent with the self-attribution behavioral bias. On the other hand, the 1000-share trade size portfolio earns a significant characteristic-adjusted return of $0.85 \%$ for six months and $0.43 \%$ per month for 12 months after the portfolio formation period. But the subsequent break in the return is negative but insignificant. This behavior is consistent with the conservatism behavioral bias.

Panels C and D show that the enhanced performance is restricted to those portfolios dominated by 500 and 1000 -share trades, or decile 10 portfolios, but only for the first six months. The W-L (decile 10 - Base) 500-share and 1000-share strategies earn monthly returns of $0.25 \%$ and $0.23 \%$, respectively, more than the base momentum strategy over six months. The W-L (decile 10 - Base) characteristic-adjusted returns fall to insignificance over the full twelve month period earning $0.13 \%$ and $0.04 \%$ for the 500 and 1000 -share portfolios, respectively. This behavior is distinctly different than that seen for the Decile 1 portfolios. The W-L (decile 1 - Base) earn characteristic-adjusted returns that are $-0.35 \%$ and $-0.23 \%$ for 500 and 1000 -share portfolios, respectively. This underperformance persists up to one year.

### 7.2 Post-Decimalization Period

Turning to the post decimalization period from 2001 to 2010 that is shown in Panel 2 of Table 10, we can clearly see that the base momentum strategy is now insignificant evidencing the crashes found in Daniel and Moskowitz (2013). The insignificance in the characteristic-adjusted returns are found regardless of holding period. Indeed, this is replicated with the 100 -share size portfolios
that also experience no significant performance over any holding period. This is found regardless of examining the Decile 10 or the Decile 1 portfolios. Also, we note that the average firm size that makes up these 100 -share portfolios is now slanted to extremely small firm with market valuations of $\$ 547$ million for the Loser portfolio and $\$ 670$ million for the Winner portfolio.

The 500 -share portfolio, shown in Panel G of Table 10, continues to earn a significant characteristic-adjusted return over the first year with a $0.75 \%$ per month return earned over the first six months and a $0.47 \%$ per month return earned over the year. An insignificant break in the return is noted after one-year. Interestingly, the average firm size of these extreme 500 -share portfolios is now increased to over a billion dollars. It appears that traders concentrating in this trade size portfolio are focusing on larger market capitalization firms and only these larger firms outperform the base momentum strategy as evidenced by the W-L (Decile 10- Base) results that shows a $0.45 \%$ per month return improvement over six months and a smaller $0.25 \%$ return improvement over one-year. The Decile 1, 500-share portfolio significantly underperforms the base momentum strategy indicating that only those portfolios that experienced higher concentrations of 500 -share trade sizes earn a significant momentum return.

Finally, the Decile 10, 1000-share size cluster portfolio, illustrated in Panel H of Table 10, also earn significant monthly returns of $0.68 \%$ over six months, but they fall to insignificance over one year. The larger trade size portfolios appear to price momentum more quickly with significance only observed in the first six months after the momentum portfolio formation. The decile 10 portfolios for both the winner and loser side of the trade, shows clustering around much larger firms with the Winner portfolio composed of firms with a market capitalization of $\$ 1.529$ billion and the Loser portfolio composed of firms with a market capitalization of $\$ 3.211$ billion. Additionally, the 1000share trade portfolio out-perform the base momentum strategy by $0.38 \%$ which is significant at the $5 \%$ level. This does not persist over the one-year period.

## 8 Volume Turnover

As a robustness check, we now analyze volume turnover as proposed by Lee and Swaminathan (2000). To this end, we analyze all NYSE/Amex firms in Table 11 both in the pre and post decimalization periods expressly as performed in Lee and Swaminathan (2000). As is shown in Panel A of Table 11, the pre-decimalization period produces significant characteristic-adjusted returns that are very robust. For instance, the six-month return is $0.78 \%$ per month which is significant at the $1 \%$ level. The return significance does not persist over the one-year period. The W-L (Decile 10 - Base) return indicates that the concentrating on high volume stocks can yield a $0.40 \%$ per month return improvement over the base momentum strategy. This is consistent with the tone of the results shown in Lee and Swaminathan (2000).

However, focusing on the post-decimalization period now shows no abnormal performance for turnover regardless of holding period. These results clearly illustrate that focusing on turnover cannot yield significant returns over the post decimalization period. This corresponds to the base momentum strategy that also experiences no abnormal return in the post-decimalization period. Turnover is not sufficient at separating out firms that can produce significant momentum profits in the post-decimalization period.

## 9 Conclusions

We show that past trade size clusters are an important determinant in price-based momentum. We consider the effect of trade size clusters within a momentum strategy and show that trade size clusters of 500 and 1000 -shares are important in the expectation of future returns. However, we further show that this pricing ability is not shared by any other trade size cluster. It appears that informed traders use these distinct 500 and 1000-share size when splitting up larger orders supporting the stealth trading hypothesis. We show conclusively momentum profits can be significantly increased by focusing on 500 and 1000-share trade size clusters rather than any other trade size cluster.

The large trade cluster results are robust to the portfolio formation techniques that would focus on deciles, quintiles, or terciles for portfolio formations alleviating concerns about sample size or portfolio composition. Additionally, these large trade size cluster portfolios produces significantly improved returns if we also condition on much larger dollar trade volumes in addition to price filters. Finally, the large trade size cluster portfolios do not experience "momentum crashes" that is typical of the base momentum strategy in the post-decimalization period. Large trade portfolios continue to earn significant returns in both the pre and post decimalization period. Traders concentrating in these portfolios appear to act strategically by focusing on larger market capitalization firms and avoiding the issues with falling trade size as a result of the decimalization of quotes that drove average trade size down significantly.

This paper attempts to expand upon the research that explores trade size, but we are distinctive in that we combine two separate strains of literature that have been previously explored separately. This includes the retail trade literature that requires separate buy and sell volume and is behaviorally based and an older literature that explores the pricing of more informed trades, but using a stealth trading hypothesis to form the expectation. We combine these two separate literature streams into one picture showing the importance of both categories when using trade size clusters, but we obviate the necessity of determining separate buys and sells for small trades by focusing exclusively on the trades clusters into pre-identified categories.

The main features of this path of research is the potential identification of a set of trades embodied in trade size clusters that may evidence more informed trading. More study on trade size and its relation to informed trade is necessary, but the ease in identification of informed segments of the market and the sheer significance of the results should foster more empirical research as well as theoretical investigation. We view these results as fundamental to better understanding the source of momentum profits through the actions of stealth traders and our results strongly support the stealth trading hypothesis. This line of research of who is informed in the market and how they trade is important in understanding the source of anomalies, that now number nearly 80 (Hou et al., 2015), and why they persist.

## References

Alexander, G., and M. Peterson. 2007. An analysis of trade-size clustering and its relation to stealth trading. Journal of Financial Economics.

Asness, C. S., T. J. Moskowitz, and L. H. Pedersen. 2013. Value and Momentum Everywhere. Journal of Finance. 68:929-985.

Barberis, N., A. Shleifer, and R. Vishny. 1998. A model of investor sentiment. Journal of Financial Economics. 49(3):307-343.

Barclay, M., and J. Warner. 1993. Stealth Trading and Volatility. Journal of Financial Economics. $34(2): 281-305$.

Battalio, R. H., and R. R. Mendenhall. 2005. Earnings expectations, investor trade size, and anomalous returns around earnings announcements. Journal of Financial Economics. 77:289-319.

Chakravarty, S.. 2001. Stealth-trading: Which traders' trades move stock prices?. Journal of Financial Economics. 61(2):289-307.

Chordia, T., and A. Subrahmanyam. 2004. Order imbalance and individual stock returns: Theory and evidence. Journal of Financial Economics. 72(3):485-518.

Collin-Dufresne, P., and V. Fos. 2015. Do prices reveal the presence of informed trading?. The Journal of Finance.

Daniel, K., M. Grinblatt, S. Titman, and R. Wermers. 1997. Measuring mutual fund performance with characteristic-based returns. Journal of Finance. 52:1035-1058.

Daniel, K., D. Hirshleifer, and A. Subrahmanyam. 1998. Investor Psychology and Security Market Under- and Overreactions. The Journal of Finance. 53(6):1839-1885.

Daniel, K., and T. Moskowitz. 2013. Momentum Crashes. CBS Working Paper. pages 1-38.

Daniel, K., and S. Titman. 1999. Market Efficiency in an Irrational World. Financial Analysts Journal. 55:28-40.

Easley, D., and M. O'Hara. 1987. Price, Trade Size, and Information in Security Markets. Journal of Financial Economics. 19:69-90.

Fama, E. F., and K. R. French. 1993. Common Risk Factors in the Returns on Stocks and Bonds. Journal of Financial Economics. 33:3-56.

Frazzini, A., R. Israel, and T. Moskowitz. Trading Costs of Asset Pricing Anomalies. URL http: //ssrn.com/abstract=2294498. Working paper, University of Chicago. 2013.

Hasbrouck, J.. 1995. One Security, Many Markets: Determining the Contributions to Price Discovery. The Journal of Finance. 50(4):1175-1199.

Holden, C. W., and S. Jacobsen. 2014. Liquidity measurement problems in fast, competitive markets: expensive and cheap solutions. The Journal of Finance. 69(4):1747-1785.

Hong, H., T. Lim, and J. Stein. 2000. Bad News Travels Slowly: Size, Analyst Coverage, and the Profitability of Momentum Strategies. Journal of Finance. 55:265-295.

Hou, K., W. Xiong, and L. Peng. 2006. R2 and Price Inefficiency. Fisher College of Business Working Paper Series. nov.

Hou, K., C. Xue, and L. Zhang. Sept. 2015. Digesting Anomalies: An Investment Approach. Review of Financial Studies. pages 650-705.

Huh, S.-W., and A. Subrahmanyam. 2005. Order Flow Patterns around Seasoned Equity Offerings and their Implications for Stock Price Movements*. International Review of Finance. 5(1-2): 75-111.

Hvidkjaer, S.. 2006. A trade-based analysis of momentum. Review of Financial Studies. 19(2): 457-491.
——. 2008. Small Trades and the Cross-Section of Stock Returns. Review of Financial Studies. 21(3):1123-1151.

Jegadeesh, N., and S. Titman. 1993. Returns to Buying Winners and Selling Losers; Implications for Stock Market Efficiency. Journal of Finance. 48:65-91.
-_ . 2001. Profitability of momentum strategies: An evaluation of alternative explanations. The Journal of Finance. 56(2):699-720.

Kaniel, R., S. Liu, G. Saar, and S. Titman. 2012. Individual Investor Trading and Return Patterns around Earnings Announcements. Journal of Finance. 67:639-680.

Kaniel, R., G. Saar, and S. Titman. 2008. Individual Investor Trading and Stock Returns. Journal of Finance. LXIII, No. 1:273-310.

Keim, D., and A. Madhavan. 1995. Anatomy of the trading process empirical evidence on the behavior of institutional traders. Journal of Financial Economics. 37(3):371-398.

Lee, C. M.. 1992. Earnings News and Small Traders. Journal of Accounting and Economics. 15: 265-302.

Lee, C. M. C., and B. Swaminathan. 2000. Price Momentum and Trading Volume. Journal of Finance. 55:2017-2069.

Malmendier, U., and D. Shanthikumar. Do Security Analysts Speak in Two Tongues? Forthcoming, Review of Financial Studies. 2014.

Novy-Marx, R., and M. Velikov. A Taxonomy of Anomalies and their Trading Costs. URL http: //www.nber.org/papers/w20721.pdf. NBER Working Paper No. 20721. December 2014.

Rouwenhorst, K. G.. 1998. International Momentum Strategies. Journal of Finance. 53:267-284.

Verardo, M.. 2009. Heterogeneous Beliefs and Momentum Profits. Journal of Financial and Quantitative Analysis. 44:795822.

Wermers, R. Is Money Really 'Smart? New Evidence on the Relation Between Mutual Fund Flows, Manager Behavior, and Performance Persistence. URL http://papers.ssrn.com/sol3/ papers.cfm?abstract_id=414420\#PaperDownload. Working Paper, University of Maryland. 2003.

Zhang, X. F.. 2006. Information Uncertainty and Stock Returns. The Journal of Finance. 61:105137.
Table 1: Momentum Portfolio Returns
We present average monthly returns in percentage for price momentum portfolio strategies involving NYSE/Amex/NASDAQ stocks for the time period 1983 to 2010. The sample consists of all NYSE/AMEX and beginning in 1987 NASDAQ stocks with a price $>\$ 5$. At the beginning of each month starting in 1983, all stocks in the NYSE/Amex/NASDAQ exchanges are sorted based on their previous J months' cumulative returns and divided into 10 equally-weighted portfolios (shown in Panel A) or into three equally-weighted portfolios (shown in Panel B). R1 represents the portfolio with the lowest returns (loser), and R10 (or R3 for the tercile portfolios) represents the portfolio with the highest returns (winner) during the previous J months. The monthly holding period is represented by K , where $\mathrm{K}=$ three, six, nine, or 12 months. Monthly holding period returns are computed as an equal-weighted average of returns from strategies initiated at the beginning of the month and the past J months. Returns are the average return across the monthly returns and are expressed in percentages. T100 represents the proportion of all 100 share daily trades divided by the total number of trades averaged over the month. T 500 represents the proportion of all 500 share daily trades divided by the total number of trades averaged over the month. T1000 represents all 1000 share daily trades divided by the total number of trades averaged on the month. $S z R n k$ is the time-series average of the median size decile of the portfolio of NYSE/Amex stocks in the sample. Price represents the time-series average dollar stock price of the portfolio on the portfolio formation date. Newey-West robust estimators with four lags specifying the t -statistics are presented in the parentheses. Significance is reported with an * ( $10 \%$ significance), an ** ( $5 \%$ significance), or an $* * *$ ( $1 \%$ significance).
Panel A: Momentum Portfolio Deciles

| J | Portfolio | Return | T100 | T500 | T1000 | SzRnk | Price | Monthly Returns |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  |  |  |  |  |  |  | $\mathrm{K}=3$ |  | $\mathrm{K}=6$ |  | $\mathrm{K}=9$ |  | $\mathrm{K}=12$ |  |
| 3 | R1 (Loser) | -10.31 | 0.33 | 0.09 | 0.11 | 2.99 | 15.53 | 0.73* | (1.77) | 0.61 | (1.50) | 0.57 | (1.41) | 0.58 | (1.42) |
|  | R5 | 0.34 | 0.34 | 0.09 | 0.09 | 4.09 | 51.17 | $1.13{ }^{* * *}$ | (4.17) | $1.14{ }^{* * *}$ | (4.28) | 1.09*** | (4.05) | $1.11{ }^{* * *}$ | (4.12) |
|  | R10 (Winner) | 15.04 | 0.32 | 0.10 | 0.12 | 3.42 | 27.28 | $1.26{ }^{* * *}$ | (3.17) | $1.28^{* * *}$ | (3.26) | 1.21 *** | (3.11) | $1.15{ }^{* * *}$ | (2.93) |
|  | R10-R1 |  |  |  |  |  |  | $0.53 * *$ | (1.98) | $0.67^{* * *}$ | (2.87) | 0.63*** | (2.99) | $0.57^{* * *}$ | (2.92) |
| 6 | R1 (Loser) | -7.20 | 0.32 | 0.09 | 0.11 | 2.91 | 13.99 | 0.52 | (1.22) | 0.49 | (1.15) | 0.45 | (1.05) | 0.54 | (1.27) |
|  | R5 | 0.49 | 0.34 | 0.09 | 0.09 | 4.12 | 45.44 | $1.13{ }^{* * *}$ | (4.22) | $1.13{ }^{* * *}$ | (4.26) | 1.08*** | (4.04) | 1.09*** | (4.05) |
|  | R10 (Winner) | 10.57 | 0.33 | 0.10 | 0.12 | 3.44 | 33.42 | $1.58{ }^{* * *}$ | (3.90) | $1.53^{* * *}$ | (3.81) | 1.35*** | (3.34) | 1.23 *** | (3.07) |
|  | R10-R1 |  |  |  |  |  |  | $1.06{ }^{* * *}$ | (3.40) | $1.04{ }^{* * *}$ | (3.57) | 0.90*** | (3.32) | 0.70*** | (2.76) |
| 9 | R1 (Loser) | -5.76 | 0.32 | 0.09 | 0.11 | 2.81 | 13.18 | 0.47 | (1.05) | 0.46 | (1.04) | 0.49 | (1.12) | 0.6 | (1.39) |
|  | R5 | 0.56 | 0.34 | 0.09 | 0.08 | 4.23 | 50.88 | 1.09*** | (4.06) | $1.12{ }^{* * *}$ | (4.23) | 1.09*** | (4.08) | $1.10^{* * *}$ | (4.13) |
|  | R10 (Winner) | 8.75 | 0.33 | 0.10 | 0.12 | 3.53 | 33.15 | $1.64{ }^{* * *}$ | (3.89) | 1.50*** | (3.58) | 1.29*** | (3.07) | $1.14{ }^{* * *}$ | (2.79) |
|  | R10-R1 |  |  |  |  |  |  | $1.17{ }^{* * *}$ | (3.41) | $1.05 * * *$ | (3.15) | 0.79** | (2.53) | 0.54* | (1.87) |
| 12 | R1 (Loser) | -4.89 | 0.32 | 0.09 | 0.11 | 2.75 | 12.58 | 0.43 | (0.96) | 0.5 | (1.12) | 0.56 | (1.27) | 0.67 | (1.51) |
|  | R5 | 0.59 | 0.34 | 0.09 | 0.08 | 4.24 | 58.08 | 1.09 *** | (4.14) | $1.12{ }^{* * *}$ | (4.30) | 1.10*** | (4.21) | $1.10{ }^{* * *}$ | (4.20) |
|  | R10 (Winner) | 7.66 | 0.33 | 0.10 | 0.12 | 3.61 | 28.35 | $1.48^{* * *}$ | (3.38) | $1.35{ }^{* * *}$ | (3.13) | 1.13 *** | (2.66) | $1.04 * *$ | (2.50) |
|  | R10-R1 |  |  |  |  |  |  | $1.06{ }^{* * *}$ | (2.87) | 0.85** | (2.42) | 0.57* | (1.74) | 0.37 | (1.23) |

Panel B：Momentum Portfolio Terciles

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Table 2: No-Overlap Monthly and Characteristic-Adjusted Returns: Trade Size and Price Momentum
This table presents average monthly and characteristic-adjusted returns from portfolios strategies formed on the basis of two-way sorts based on past trade size and price momentum for the period 1983 to 2010. We delete any firm that is in both the decile 1 (Decile 10 ) 100 -share portfolio and the decile 1 (decile 10 ) 500 and 1000 -share portfolios to eliminate overlapping firms. We also delete any firm with a share price $<\$ 5.00$. At the beginning of each month all available stocks listed on the NYSE/Amex/NASDAQ markets are first sorted based on the past 6 months returns, and divided into three portfolios. The loser portfolio represents the worst prior price
performance and the winner portfolio represents the best prior price performance. Within each momentum portfolio, the stocks are then sorted into deciles based on the average trade size over the month just prior to the monthly evaluation period. We skip one month between the portfolio formation period and the performance period. We analyze performance of three separate holding periods using a six-month period (1-6), a twelve-month period (1-12), and a holding period that spans months 13 to 24 from the formation period. Monthly returns are computed based on the portfolio rebalancing strategy described in Table 1 and characteristic-adjusted returns are based the base momentum strategy (W-L Base), the extreme Winner-Loser trade size deciles (W-L Deciles 1 or 10), and then we finally compare winner-loser trade portfolio returns to the base momentum strategy (W-L Decile 1 or 10 - Base). We report the average bid-ask spread and the price impact measure of Amihud, firm size expressed in millions of dollars, and the average price of the portfolio of stocks at the end of the month prior to the performance period. Panel 1 presents the primary trade size with four lags specifying the t-statistics are presented in the parentheses. Significance is reported with an * ( $10 \%$ significance), an ${ }^{* *}(5 \%$ significance), or an $* * *(1 \%$ significance).

|  |  |  |  |  |  | Panel 1: | Cluster | d Trad |  |  |  |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Momentum Tercile | Trade Size Decile | Number of Firms | Average Price | $\begin{gathered} \text { Firm } \\ \text { Size } \\ (\$ \text { millions }) \end{gathered}$ | Bid-Ask Spreads (\%) | Amihud's measure | \% Monthly Holding Period Returns |  |  |  |  |  | \% Monthly Characteristic-Adjusted Returns |  |  |  |  |  |
|  |  |  |  |  |  |  | Return ${ }^{\text {\% }}{ }^{\text {1-6 Monthly }}$ |  | Return | T | Return | T | Return | T | Return | T | Return | T |
|  | Panel A: Base Six Month Momentum Strategy, Single-Sort by previous six-month returns |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| Losers (L) |  | 933 | 19 | 1740 | 2.06 | 0.13 | 0.81** | (2.31) | 0.82** | (2.34) | $1.14{ }^{* * *}$ | (3.23) | -0.26*** | (-2.74) | -0.17* | (-1.96) | 0.08 | (1.07) |
| 2 |  | 932 | 26 | 2733 | 1.77 | 0.13 | 1.14*** | (4.27) | 1.11*** | (4.13) | 1.11*** | (4.04) | -0.02 | (-0.29) | -0.01 | (-0.14) | 0.02 | (0.27) |
| Winners (W) |  | 932 | 27 | 2194 | 1.75 | 0.09 | $1.34 * * *$ | (4.17) | $1.22^{* * *}$ | (3.83) | 1.00*** | (3.15) | $0.18{ }^{* * *}$ | (3.43) | 0.09** | (2.35) | -0.04 | (-0.57) |
| W-L (Base) |  |  |  |  |  |  | 0.53*** | (2.94) | 0.40** | (2.58) | -0.13 | (-1.23) | $0.45 * * *$ | (3.22) | 0.26** | (2.45) | -0.12 | (-1.34) |
| Panel B: Double Sort by previous six-month returns, then previous one-month trade size $=100$ shares |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| Losers (L) | 1 | 48 | 18 | 3751 | 3.23 | 0.48 | 0.56* | (1.76) | 0.61* | (1.90) | $1.06{ }^{* * *}$ | (3.25) | -0.42*** | (-3.40) | -0.29** | (-2.54) | 0.02 | (0.20) |
|  | 10 | 43 | 23 | 1304 | 2.07 | 0.12 | 0.97*** | (3.09) | 0.95*** | (2.9) | 1.16*** | (3.58) | -0.13 | (-1.25) | -0.07 | (-0.70) | 0.09 | (1.00) |
| 2 | 1 | 46 | 23 | 4256 | 3.11 | 0.54 | 0.89*** | (3.26) | 0.91*** | (3.34) | 1.13*** | (4.20) | -0.2 | (-1.56) | -0.15 | (-1.25) | 0.02 | (0.15) |
|  | 10 | 45 | 31 | 2352 | 1.65 | 0.09 | $1.14{ }^{* * *}$ | (4.5) | $1.122^{* *}$ | (4.47) | 1.06*** | (4.1) | -0.03 | (-0.29) | -0.02 | (-0.25) | -0.02 | (-0.17) |
| Winners (W) | 1 | 48 | 21 | 3296 | 2.79 | 0.32 | $1.25{ }^{* * *}$ | (4.04) | $1.133^{* * *}$ | (3.7) | 0.98*** | (2.96) | 0.19* | (1.71) | 0.07 | (0.78) | -0.05 | (-0.47) |
|  | 10 | 42 | 34 | 2015 | 1.69 | 0.08 | $1.47{ }^{* * *}$ | (4.32) | $1.37^{* * *}$ | (4.21) | 1.01*** | (3.35) | $0.31^{* *}$ | (2.46) | $0.24 * *$ | (2.29) | 0.02 | (0.16) |
| W-L (Decile 10) |  |  |  |  |  |  | 0.51** | (2.26) | 0.42* | (1.95) | -0.15 | (-1.04) | 0.44** | (2.54) | 0.30** | (2.00) | -0.08 | (-0.66) |
| W-L (Decile 1) |  |  |  |  |  |  | 0.69*** | (4.38) | 0.52*** | (3.57) | -0.08 | (-0.55) | 0.61*** | (4.16) | 0.36*** | (3.24) | -0.07 | (-0.56) |
| W-L (Decile 10) - W-L (Base) |  |  |  |  |  |  | -0.03 | (-0.17) | 0.01 | (0.15) | -0.02 | (-0.26) | -0.01 | (-0.03) | 0.04 | (0.44) | 0.04 | (0.57) |
| W-L (Decile 1) - W-L (Base) |  |  |  |  |  |  | 0.16 | (1.45) | 0.11 | (1.16) | 0.06 | (0.52) | 0.16 | (1.39) | 0.10 | (1.05) | 0.04 | (0.35) |
|  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| Losers (L) | 10 | 63 | 21 | 817 | 3.09 | 0.38 | 1.05*** | (3.7) | 1.01*** | (3.52) | $1.07^{* * *}$ | (3.83) | -0.08 | (-0.71) | -0.05 | (-0.48) | -0.03 | (-0.32) |
|  | 10 | 72 | 13 | 1018 | 2.96 | 0.29 | 0.66* | (1.8) | 0.63* | (1.77) | 0.91*** | (2.63) | -0.50*** | (-3.66) | -0.43*** | (-3.85) | -0.16* | (-1.77) |
| 2 | 1 | 63 | 27 | 850 | 2.88 | 0.36 | 1.13*** | (4.62) | $1.122^{* *}$ | (4.66) | 1.08*** | (4.57) | -0.09 | (-0.69) | -0.04 | (-0.35) | -0.04 | (-0.34) |
|  | 10 | 71 | 17 | 1421 | 2.74 | 0.26 | $1.25{ }^{* * *}$ | (4.34) | $1.15{ }^{* * *}$ | (3.98) | 0.94*** | (3.15) | 0.01 | (0.07) | -0.06 | (-0.53) | -0.19* | (-1.77) |
| Winners (W) | 1 | 63 | 31 | 1616 | 2.32 | 0.25 | $1.22{ }^{* * *}$ | (4.7) | $1.22^{* * *}$ | (4.77) | $1.15{ }^{* * *}$ | (4.35) | 0.03 | (0.28) | 0.06 | (0.59) | 0.12 | (1.21) |
|  | 10 | 71 | 18 | 637 | 2.57 | 0.18 | $1.47{ }^{* * *}$ | (4.03) | 1.20 *** | (3.3) | 0.74** | (2.08) | 0.27** | (2.45) | 0.02 | (0.19) | -0.35** | (-2.56) |
| W-L (Decile 10) |  |  |  |  |  |  | 0.81*** | (3.95) | $0.57^{* * *}$ | (3.61) | -0.18 | (-1.49) | 0.78*** | (4.5) | 0.45*** | (3.97) | -0.18* | (-1.68) |
| W-L (Decile 1) |  |  |  |  |  |  | 0.17 | (1.45) | 0.21* | (1.92) | 0.08 | (0.87) | 0.11 | (1.1) | 0.11 | (1.39) | 0.14* | (1.86) |
| W-L (Decile 10) - W-L (Base) |  |  |  |  |  |  | 0.28*** | (2.65) | $0.16^{* *}$ | (2.2) | -0.04 | (-0.51) | 0.33*** | (3.13) | 0.19** | (2.49) | -0.06 | (-0.79) |
| W-L (Decile 1) - W-L (Base) |  |  |  |  |  |  | -0.36*** | (-3.51) | -0.19** | (-2.38) | $0.21^{* * *}$ | (2.6) | -0.33*** | (-3.70) | -0.15** | (-2.17) | $0.26^{* * *}$ | (3.2) |
| Losers (L) | Panel D: Double Sort by previous six-month |  |  |  |  |  | returns, then previous one-month trade size $=1000$ shares |  |  |  |  |  |  |  |  |  |  |  |
|  | 1 | 59 | 23 | 648 | 2.79 | 0.41 | 0.89*** | (3.26) | 0.91*** | (3.29) | 1.06*** | (3.91) | -0.27** | (-2.05) | -0.18 | (-1.53) | -0.06 | (-0.58) |
|  | 10 | 63 | 12 | 1335 | 2.56 | 0.21 | 0.6 | (1.52) | 0.62 | (1.59) | $1.01^{* *}$ | (2.47) | -0.49*** | (-3.37) | -0.35*** | (-2.84) | -0.01 | (-0.08) |
| 2 | 1 | 60 | 28 | 549 | 2.76 | 0.39 | $1.122^{* * *}$ | (4.96) | 1.11*** | (4.92) | $1.09^{* * *}$ | (4.71) | -0.13 | (-0.83) | -0.08 | (-0.54) | -0.05 | (-0.33) |
| Winners (W) | 10 | 62 | 16 | 1526 | 2.41 | 0.22 | $1.22^{* * *}$ | (3.88) | $1.16^{* * *}$ | (3.63) | $1.08^{* * *}$ | (3.23) | 0.02 | (0.17) | , | (-0.04) | 0 | (0.05) |
|  | 1 | 57 | 32 | 996 | 2.25 | 0.24 | $1.27^{* * *}$ | (5.13) | $1.21^{* * *}$ | (4.93) | $1.06{ }^{* * *}$ | (4.22) | 0.04 | (0.33) | 0.04 | (0.31) | 0 | (-0.04) |
|  | 10 | 63 | 18 | 1002 | 2.07 | 0.14 | 1.39*** | (3.47) | $1.15 * * *$ | (2.94) | 0.99** | (2.53) | 0.25* | (1.85) | 0.02 | (0.14) | -0.06 | (-0.45) |
| W-L (Decile 10) |  |  |  |  |  |  | 0.79*** | (3.74) | $0.53 * * *$ | (3.21) | -0.02 | (-0.13) | $0.74 * * *$ | (4.15) | $0.37^{* * *}$ | (2.87) | -0.05 | (-0.37) |
| W-L (Decile 1) |  |  |  |  |  |  | 0.38*** | (3.06) | 0.30*** | (2.78) | 0.01 | (0.08) | 0.31*** | (2.81) | 0.21*** | (2.65) | 0.06 | (0.82) |
| W-L (Decile 10) - W-L (Base) |  |  |  |  |  |  | 0.26** | (2.44) | $0.13$ | (1.45) | $0.12$ | (1.05) | $0.29^{* * *}$ | (2.77) | $0.11$ | (1.21) | 0.07 | (0.69) |
| W-L (Decile 1) - W-L (Base) |  |  |  |  |  |  | -0.15 | (-1.29) | -0.1 | (-1.10) | 0.14* | (1.88) | $-0.14$ | (-1.35) | -0.05 | (-0.60) | 0.18** | (2.58) |

Panel 2: Non-Clustered Large Trades

| Momentum Tercile | Trade Size Decile | Number of Firms | Average Price | $\begin{gathered} \text { Firm } \\ \text { Size } \\ \text { (\$ millions) } \end{gathered}$ | Bid-Ask Spreads (\%) | Amihud's measure | \% Monthly Holding Period Returns <br> 1-6 1-12 13-24 |  |  |  |  |  | \% Monthly$1-6$Characteristic-Adjusted Returns$1-12$$13-24$ |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  |  |  |  |  |  | Return | T | Return | T | Return | T | Return | T | Return | T | Return | T |
| Panel E: Double Sort by previous six-month returns, then previous one-month trade size > 500 and < 1000 shares |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| Losers (L) | 1 | 70 | 20 | 439 | 3.46 | 0.38 | 0.85*** | (2.81) | 0.86*** | (2.83) | $1.11^{* * *}$ | (3.71) | -0.34*** | (-2.61) | -0.25** | (-2.16) | -0.01 | (-0.11) |
|  | 10 | 82 | 17 | 2947 | 2.73 | 0.29 | 0.84*** | (2.63) | $0.84^{* * *}$ | (2.66) | $1.04^{* * *}$ | (3.41) | -0.21* | (-1.72) | -0.14 | (-1.20) | -0.04 | (-0.42) |
| 2 | 1 | 70 | 25 | 498 | 3.23 | 0.35 | 1.26*** | (4.96) | $1.20{ }^{* * *}$ | (4.74) | $1.13{ }^{* * *}$ | (4.41) | -0.02 | (-0.17) | -0.04 | (-0.28) | -0.04 | (-0.32) |
|  | 10 | 83 | 23 | 4007 | 2.48 | 0.30 | 1.08*** | (4.26) | $1.05 * * *$ | (4.13) | $1.06{ }^{* * *}$ | (4.07) | -0.06 | (-0.49) | -0.04 | (-0.37) | -0.04 | (-0.37) |
| Winners (W) | 1 | 69 | 28 | 759 | 2.78 | 0.25 | 1.39*** | (4.82) | $1.27^{* * *}$ | (4.34) | $0.97^{* * *}$ | (3.3) | 0.14 | (1.36) | 0.05 | (0.56) | -0.11 | (-0.91) |
|  | 10 | 82 | 23 | 2939 | 2.33 | 0.19 | $1.13{ }^{* * *}$ | (3.89) | 1.11*** | (3.86) | 1.02*** | (3.33) | -0.01 | (-0.06) | 0 | 0 | -0.03 | (-0.31) |
| W-L (Decile 10) |  |  |  |  |  |  | 0.28** | (1.97) | $0.27^{* *}$ | (2.41) | -0.02 | (-0.20) | 0.21* | (1.77) | 0.14* | (1.75) | 0.01 | (0.09) |
| W-L (Decile 1) |  |  |  |  |  |  | 0.55*** | (3.56) | $0.41^{* * *}$ | (2.87) | -0.13 | (-1.17) | $0.48^{* * *}$ | (3.57) | 0.30*** | (2.89) | -0.1 | (-0.98) |
| W-L (Decile 10) - W-L (Base) <br> W-L (Decile 1) - W-L (Base) |  |  |  |  |  |  | -0.25*** | (-2.98) | -0.14* | (-1.82) | 0.11 | (1.41) | -0.24*** | (-3.31) | -0.13* | (-1.78) | 0.13* | (1.69) |
|  |  |  |  |  |  |  | 0.02 | (0.16) | 0 | (0.04) | 0 | (-0.01) | 0.03 | (0.32) | 0.04 | (0.55) | 0.02 | (0.34) |
| Losers (L) | Panel | F: Dou | ble Sort | by previous | six-mon | th return | s , then p | previous | one-mo | nth tra | de size $>$ | 1000 | and < 5000 | 0 share |  |  |  |  |
|  | 1 | 72 | 22 | 722 | 3.05 | 0.37 | 0.77** | (2.53) | 0.82*** | (2.69) | 1.12*** | (3.73) | -0.36*** | (-2.76) | -0.23* | (-1.87) | 0.03 | (0.29) |
|  | 10 | 78 | 16 | 4061 | 2.72 | 0.24 | 0.66* | (1.94) | 0.67 ** | (1.98) | $1.03^{* * *}$ | (3.01) | -0.36*** | (-3.32) | -0.25** | (-2.40) | 0.01 | (0.07) |
| 2 | 1 | 72 | 27 | 732 | 2.83 | 0.31 | 1.17*** | (4.97) | 1.16*** | (4.93) | $1.07^{* * *}$ | (4.45) | -0.08 | (-0.57) | -0.04 | (-0.31) | -0.05 | (-0.38) |
|  | 10 | 79 | 22 | 5550 | 2.47 | 0.23 | $0.98{ }^{* * *}$ | (3.57) | $0.96{ }^{* * *}$ | (3.48) | $1.08{ }^{* * *}$ | (3.76) | -0.12 | (-1.36) | -0.09 | (-0.98) | 0 | (0.02) |
| Winners (W) | 1 | 72 | 32 | 1311 | 2.55 | 0.24 | $1.47{ }^{* * *}$ | (4.75) | $1.34 * * *$ | (4.34) | 0.97*** | (3.29) | 0.25** | (2.33) | 0.16* | (1.76) | -0.04 | (-0.31) |
|  | 10 | 78 | 21 | 3145 | 2.35 | 0.15 | $1.12{ }^{* * *}$ | (3.41) | 1.02*** | (3.13) | 0.89** | (2.57) | 0.08 | (0.8) | -0.03 | (-0.32) | -0.15 | (-1.43) |
| W-L (Decile 10) |  |  |  |  |  |  | 0.46 *** | (2.85) | $0.34^{* * *}$ | (2.61) | -0.14 | (-1.36) | $0.44^{* * *}$ | (3.37) | 0.22** | (2.37) | -0.15 | (-1.53) |
| W-L (Decile 1) |  |  |  |  |  |  | 0.70*** | (3.75) | 0.52*** | (2.95) | -0.15 | (-1.08) | 0.62*** | (3.95) | 0.39*** | (2.92) | -0.07 | (-0.59) |
| W-L (Decile 10) - W-L (Base) |  |  |  |  |  |  | -0.07 | (-0.79) | -0.06 | (-0.75) | -0.01 | (-0.10) | -0.01 | (-0.13) | -0.04 | (-0.49) | -0.03 | (-0.45) |
| W-L (Decile 1) - W-L (Base) |  |  |  |  |  |  | 0.17* | (1.71) | 0.12 | (1.38) | -0.02 | (-0.21) | 0.17* | (1.79) | 0.13 | (1.61) | 0.05 | (0.73) |
| Losers (L) |  | Panel G | : Double | Sort by pr | evious si | x-month | returns, | then pr | evious o | ne-mon | th trade | size $=$ | 5000 shar |  |  |  |  |  |
|  | 1 | 139 | 14 | 118 | 4.35 | 0.38 | $1.02{ }^{* * *}$ | (3.06) | 1.19*** | (3.53) | $1.111^{* * *}$ | (3.72) | -0.61*** | (-2.99) | -0.34* | (-1.93) | -0.2 | (-1.33) |
|  | 10 | 86 | 16 | 2139 | 2.19 | 0.09 | 0.78** | (2.07) | 0.70** | (1.89) | $1.07^{* * *}$ | (2.92) | -0.32*** | (-2.73) | $-0.31^{* * *}$ | (-2.91) | -0.02 | (-0.20) |
| 2 | 1 | 176 | 20 | 135 | 4.11 | 0.28 | 1.18*** | (4.09) | $1.03^{* * *}$ | (3.16) | 0.99*** | (2.94) | 0.03 | (0.1) | 0.05 | (0.21) | -0.2 | (-0.91) |
|  | 10 | 85 | 20 | 2577 | 1.95 | 0.10 | $1.17{ }^{* * *}$ | (3.83) | $1.10{ }^{* * *}$ | (3.55) | 1.12*** | (3.58) | -0.01 | (-0.09) | -0.04 | (-0.47) | -0.03 | (-0.41) |
| Winners (W) | 1 | 131 | 20 | 171 | 3.56 | 0.30 | 1.50*** | (5.36) | $1.36{ }^{* * *}$ | (4.95) | 0.99*** | (2.94) | 0.2 | (1.34) | 0.05 | (0.37) | -0.02 | (-0.12) |
|  | 10 | 85 | 21 | 1348 | 1.87 | 0.07 | 1.29*** | (3.75) | $1.16{ }^{* * *}$ | (3.31) | $1.04^{* * *}$ | (2.87) | 0.16* | (1.7) | 0.04 | (0.5) | -0.07 | (-0.74) |
| W-L (Decile 10) |  |  |  |  |  |  | 0.52 *** | (2.84) | 0.46 *** | (3.14) | -0.03 | (-0.26) | $0.48{ }^{* * *}$ | (3.24) | $0.34^{* * *}$ | (3.48) | -0.05 | (-0.53) |
| W-L (Decile 1) |  |  |  |  |  |  | $0.74 * * *$ | (5.34) | 0.45*** | (3.42) | 0.09 | (0.77) | 0.72*** | (5.68) | 0.38*** | (3.58) | 0.15 | (1.58) |
| W-L (Decile 10) - W-L (Base) |  |  |  |  |  |  | -0.01 | (-0.15) | 0.06 | (0.96) | 0.1 | (1.28) | 0.03 | (0.46) | 0.08 | (1.52) | 0.07 | (0.97) |
| W-L (Decile 1) - W-L (Base) |  |  |  |  |  |  | 0.06 | (0.34) | 0.04 | (0.38) | $0.27^{* * *}$ | (2.96) | 0.08 | (0.56) | 0.07 | (0.76) | $0.28{ }^{* * *}$ | (3.16) |
| Losers (L) |  | Panel | H: Doubl | e Sort by p | revious s | ix-month | returns, | then pr | revious | ne-mon | th trad | e size $\geq$ | 500 share |  |  |  |  |  |
|  | 1 | 57 | 26 | 1027 | 2.35 | 0.23 | 0.85*** | (2.92) | 0.85*** | (2.92) | 1.09 *** | (3.78) | -0.28** | (-2.08) | -0.19 | (-1.60) | 0 | (0.03) |
|  | 10 | 62 | 15 | 3755 | 2.59 | 0.25 | 0.51 | (1.43) | 0.54 | (1.55) | $1.06{ }^{* * *}$ | (2.88) | -0.52*** | (-4.46) | -0.40*** | (-3.71) | 0.02 | (0.14) |
| 2 | 1 | 59 | 33 | 1185 | 2.06 | 0.20 | 1.09*** | (4.86) | 1.07*** | (4.75) | $1.06{ }^{* * *}$ | (4.58) | -0.09 | (-0.67) | -0.07 | (-0.55) | -0.03 | (-0.20) |
|  | 10 | 61 | 19 | 4531 | 2.44 | 0.27 | 1.00*** | (3.47) | $0.98{ }^{* * *}$ | (3.38) | $1.122^{* * *}$ | (3.6) | -0.12 | (-1.23) | -0.1 | (-1.11) | 0 | (0.05) |
| Winners (W) | $1$ | 55 | 36 | 1279 | 2.00 | 0.16 | $1.25{ }^{* * *}$ | (4.34) | 1.20*** | (4.23) | $0.96{ }^{* * *}$ | (3.41) | 0.07 | (0.69) | 0.07 | (0.7) | -0.03 | (-0.29) |
|  | 10 | 61 | 19 | 2332 | 2.20 | 0.15 | $1.24{ }^{* * *}$ | (3.4) | 1.09*** | (3.04) | 0.90** | (2.39) | 0.16 | (1.24) | -0.01 | (-0.05) | -0.19* | (-1.70) |
| W-L (Decile 10) |  |  |  |  |  |  | 0.73 *** | (3.73) | 0.55*** | (3.64) | -0.15 | (-1.28) | 0.67 *** | (4.23) | 0.40*** | (3.65) | -0.21* | (-1.84) |
|  |  |  |  |  |  |  | 0.40** | (2.25) | 0.36** | (2.21) | -0.13 | (-1.05) | 0.35** | (2.43) | 0.26** | (2.21) | -0.04 | (-0.40) |
| W-L (Decile 10) - W-L (Base) <br> W-L (Decile 1) - W-L (Base) |  |  |  |  |  |  | 0.20* | (1.92) | 0.15* | (1.72) | -0.02 | (-0.21) | 0.22** | (2.18) | 0.13 | (1.48) | -0.09 | (-1.00) |
|  |  |  |  |  |  |  | -0.13 | (-1.41) | -0.05 | (-0.65) | 0.01 | (0.1) | -0.1 | (-1.19) | 0 | (-0.05) | 0.08 | (1.28) |

Panel 3: Non-Clustered Small Trades

| Momentum Tercile | Trade Size Decile | Number of Firms | Average Price | FirmSize(\$ millions) | Bid-Ask Spreads (\%) | Amihud's measure | \% Monthly Holding Period Returns$1-6 \quad 1-12 \quad 13-24$ |  |  |  |  |  | \% Monthly  <br> $1-6$ $1-12$ |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  |  |  |  |  |  | Return | T | Return | T | Return | T | Return | T |  |  | Return T |  |
| Panel I: Double Sort by previous six-month returns, then previous one-month trade size > 100 and < 500 shares |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| Losers (L) | 1 | 71 | 17 |  | 2.71 | 0.23 | 0.78** | (2.36) | 0.82** | (2.39) | 1.20 *** | (3.43) | $-0.38^{* * *}$ | (-3.69) | $-0.24 * * *$ | (-2.70) | 0.09 | (1.18) |
|  | 10 | 82 | 19 | 1910 | 2.77 | 0.34 | 0.82*** | (2.7) | 0.80*** | (2.64) | $1.05{ }^{* * *}$ | (3.47) | -0.28** | (-2.16) | -0.22* | (-1.83) | -0.03 | (-0.34) |
| 2 | 1 | 70 | 24 | 877 | 2.45 | 0.20 | 1.12*** | (3.98) | $1.13{ }^{* * *}$ | (4.03) | $1.18{ }^{* * *}$ | (4.09) | -0.13 | (-1.44) | -0.08 | (-0.95) | 0 | (-0.01) |
|  | 10 | 84 | 24 | 2398 | 2.55 | 0.32 | 1.09*** | (4.58) | $1.08{ }^{* * *}$ | (4.5) | $1.04{ }^{* * *}$ | (4.28) | -0.06 | (-0.41) | -0.05 | (-0.39) | -0.07 | (-0.50) |
| Winners (W) | 1 | 70 | 25 | 822 | 2.16 | 0.15 | 1.40*** | (4.19) | $1.31^{* * *}$ | (3.93) | $1.08^{* * *}$ | (3.22) | $0.17{ }^{*}$ | (1.9) | 0.1 | (1.25) | -0.04 | (-0.43) |
|  | 10 | 83 | 25 | 2266 | 2.37 | 0.21 | 1.33*** | (4.97) | $1.26{ }^{* * *}$ | (4.69) | 1.01*** | (3.52) | 0.16 | (1.54) | 0.12 | (1.24) | -0.05 | (-0.36) |
| W-L (Decile 10) |  |  |  |  |  |  | 0.52*** | (3.77) | $0.45 * * *$ | (3.7) | -0.03 | (-0.34) | $0.44^{* * *}$ | (3.98) | 0.34*** | (3.92) | -0.01 | (-0.12) |
| W-L (Decile 1) |  |  |  |  |  |  | 0.62*** | (3.43) | $0.49^{* * *}$ | (2.94) | -0.12 | (-0.98) | $0.55^{* * *}$ | (3.73) | $0.34^{* * *}$ | (2.9) | -0.13 | (-1.26) |
| W-L (Decile 10) - W-L (Base) |  |  |  |  |  |  | -0.01 | (-0.12) | 0.05 | (0.73) | 0.10* | (1.79) | -0.01 | (-0.06) | 0.08 | (1.18) | 0.11** | (2.05) |
| W-L (Decile 1) - W-L (Base) |  |  |  |  |  |  | 0.09 | (0.99) | 0.09 | (1.14) | 0.02 | (0.27) | 0.1 | (1.09) | 0.08 | (1.15) | -0.01 | (-0.14) |

Table 3: Monthly and Characteristic-Adjusted Returns: Trade Size and Price Momentum
This table presents the average monthly returns and characteristic-adjusted returns from portfolios strategies formed based on two-way sorts based on past trade size and price momentum for the period 1983 to 2010. We delete stocks priced $<\$ 5.00$. At the beginning of each month all available stocks listed on the NYSE/Amex/NASDAQ markets are sorted based on the past 6 months returns, and divided into 3 portfolios. The loser portfolio represents the worst prior return performance and the winner portfolio represents the best prior return performance. The stocks are then sorted based on the average trade size for the last month of the portfolio formation period. We ithe six-month period (1-6), a twelve-month period (1-12), and a holding period that spans months 13 to 24 from the formation period. Monthly returns are computed based on the portfolio rebalancing strategy described in Table 1 and characteristic-adjusted returns are based on firm size and book-to-market and uses the portfolios derived from Daniel et al. (1997). We report the winner-loser portfolios for the base momentum strategy (W-L Base), then based on the extreme trade size deciles (W-L Deciles 1 or 10 ), and finally relative to the base momentum strategy ( W -L Decile 1 or 10 minus the base momentum strategy). For each extreme portfolio, we report the average end of the month prior to the performance period. Simple t-statistics are presented in the parentheses. Significance is reported with an * $(10 \%$ significance $)$, an $* *(5 \%$ significance), or an ${ }^{* * *}$ ( $1 \%$ significance).

Panel 2: Non-Clustered Large Trades

Panel 3: Non-Clustered Small Trades

| Momentum Tercile | $\begin{gathered} \text { Trade } \\ \text { Size } \\ \text { Decile } \end{gathered}$ | Number of Firms | Average Price | $\begin{gathered} \text { Firm } \\ \text { Size } \\ (\$ \text { millions }) \end{gathered}$ | Bid-Ask Spreads (\%) | Amihud's measure | \% Monthly$1-6$Holding$1-12$ |  |  |  |  |  | \% Monthly Characteristic-Adjusted Returns$1-6$$1-12$ |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  |  |  |  |  |  | Return | T | Return | T | Return | T | Return | T | Return | T | Return | T |
| Panel I: Double Sort by previous six-month returns, then previous one-month trade size $>100$ and $<500$ shares |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| Losers (L) | 1 | 105 | 19 | 504 | 2.79 | 0.25 | 0.73** | (2.29) | 0.79** | (2.39) | $1.18{ }^{* * *}$ | (3.44) | $-0.42^{* * *}$ | (-4.01) | -0.27*** | (-2.89) | 0.07 | (0.96) |
|  | 10 | 105 | 19 | 2209 | 2.82 | 0.36 | 0.77** | (2.56) | 0.77** | (2.55) | 1.01*** | (3.37) | -0.32** | (-2.44) | -0.25** | (-2.04) | -0.07 | (-0.60) |
| 2 | 1 | 105 | 26 | 845 | 2.53 | 0.21 | $1.07{ }^{* * *}$ | (3.87) | $1.10{ }^{* * *}$ | (4.01) | 1.18*** | (4.16) | -0.19* | (-1.96) | -0.11 | (-1.28) | 0 | (-0.02) |
|  | 10 | 105 | 24 | 3059 | 2.61 | 0.34 | $1.08{ }^{* * *}$ | (4.53) | $1.07^{* * *}$ | (4.46) | $1.03{ }^{* * *}$ | (4.2) | -0.08 | (-0.54) | -0.06 | (-0.47) | -0.08 | (-0.59) |
| Winners (W) | 1 | 105 | 27 | 800 | 2.21 | 0.16 | $1.34 * * *$ | (4.06) | 1.25 *** | (3.82) | $1.06^{* * *}$ | (3.19) | 0.12 | (1.3) | 0.04 | (0.51) | -0.05 | (-0.62) |
|  | 10 | 105 | 26 | 2423 | 2.42 | 0.26 | 1.30*** | (4.86) | 1.20 *** | (4.41) | $0.97 * * *$ | (3.35) | 0.14 | (1.28) | 0.07 | (0.7) | -0.08 | (-0.61) |
| W-L (Decile 10) |  |  |  |  |  |  | 0.53*** | (4.13) | 0.43*** | (3.67) | -0.04 | (-0.37) | $0.47^{* * *}$ | (4.62) | 0.32*** | (4.05) | -0.01 | (-0.16) |
| W-L (Decile 1) |  |  |  |  |  |  | 0.61*** | (3.48) | 0.46 *** | (2.86) | -0.12 | (-1.07) | $0.54^{* * *}$ | (3.78) | 0.31*** | (2.71) | -0.12 | (-1.29) |
| W-L (Decile 10) - W-L (Base) |  |  |  |  |  |  | 0 | (0.03) | 0.03 | (0.45) | 0.09 | (1.65) | 0.01 | (0.15) | 0.06 | (0.91) | 0.10* | (1.84) |
| W-L (Decile 1) - W-L (Base) |  |  |  |  |  |  | 0.08 | (0.9) | 0.06 | (0.89) | 0.02 | (0.27) | 0.09 | (0.97) | 0.04 | (0.68) | -0.01 | (-0.15) |

# Table 4: Fama-French Three-Factor Regressions of Monthly Excess Returns on Price-Momentum-Trade Size Portfolios 

This table presents regressions of monthly returns on price momentum and trade size for $\mathrm{J}=6$ (formation period) and $\mathrm{K}=6$ (months after the formation period) portfolios. We also skip one-month between each of these periods. We repeat the regressions for each of the three trade size portfolios, i.e. 100 -share, 500 -share, and 1000-shares. Each of the regression results are presented in separate panels. We delete any firms with a share price $<\$ 5.00$. At the beginning of each month all available stocks are sorted based on the past 6 months returns, and divided into 3 portfolios. The loser portfolio represents the worst prior return performance and the winner portfolio represents the best prior return performance. The stocks are then sorted based on the average trade size over the last month of the momentum portfolio formation period. We divide the trade size portfolios into 10 portfolios. The three-factor regression model is given as:

$$
r_{i}-r_{f}=a_{i}+b_{i}\left(r_{m}-r_{f}\right)+s_{i} S M B+h_{i} H M L+\epsilon_{i}
$$

where $r_{m}-r_{f}$ is the excess return on the one-month value-weighted return on the NYSE/Amex/NASDAQ market, HML is the High-minus-low book-tomarket (value) factor, and SMB is the small-minus-big size factor. The term $a_{i}$ represents the abnormal performance for each portfolio. All returns and market return are stated on a percentage basis, hence the estimated slope coefficients are stated on a percentage basis. We report the winner-loser portfolios for each of the three trade size portfolios. The sample period spans January 1983 to December 2010. Newey-West robust estimators with four lags specifying the t-statistics are presented in the parentheses. Significance is reported with an * ( $10 \%$ significance), an $* *$ ( $5 \%$ significance), or an $* * *$ ( $1 \%$ significance). Panel A: Trade Size $=100$ Shares

|  | D1 | D5 | D10 | D10-D1 | D1 | D5 | D10 | D10-D1 | D1 | D5 | D10 | D10-D1 |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Portfolio | a |  |  |  | b |  |  |  | s |  |  |  |
| Losers (L) | $\begin{gathered} -0.51^{* * *} \\ (-2.85) \end{gathered}$ | $\begin{gathered} -0.48^{* * *} \\ (-3.32) \end{gathered}$ | $\begin{aligned} & -0.18^{*} \\ & (-1.67) \end{aligned}$ | $\begin{aligned} & 0.34^{*} \\ & (1.82) \end{aligned}$ | $\begin{gathered} 0.95^{* * *} \\ (13.83) \end{gathered}$ | $\begin{aligned} & 1.20^{* * *} \\ & (26.91) \end{aligned}$ | $\begin{gathered} 1.03^{* * *} \\ (16.68) \end{gathered}$ | $\begin{gathered} 0.07 \\ (0.79) \end{gathered}$ | $\begin{gathered} 0.42^{* * *} \\ (3.03) \end{gathered}$ | $\begin{gathered} 0.67^{* * *} \\ (4.35) \end{gathered}$ | $\begin{gathered} 0.66^{* * *} \\ (6.69) \end{gathered}$ | $\begin{gathered} 0.25^{* * *} \\ (2.73) \end{gathered}$ |
| 2 | $\begin{gathered} -0.07 \\ (-0.41) \end{gathered}$ | $\begin{gathered} 0.05 \\ (0.57) \end{gathered}$ | $\begin{gathered} 0.1 \\ (1.24) \end{gathered}$ | $\begin{gathered} 0.17 \\ (0.99) \end{gathered}$ | $\begin{gathered} 0.76^{* * *} \\ (11.56) \end{gathered}$ | $\begin{gathered} 0.95 * * * \\ (31.21) \end{gathered}$ | $\begin{gathered} 0.84^{* * *} \\ (27.21) \end{gathered}$ | $\begin{gathered} 0.08 \\ (0.98) \end{gathered}$ | $\begin{gathered} 0.34^{* * *} \\ (3.36) \end{gathered}$ | $\begin{gathered} 0.40^{* * *} \\ (5.54) \end{gathered}$ | $\begin{gathered} 0.49 * * * \\ (8.75) \end{gathered}$ | $\begin{gathered} 0.15 \\ (1.63) \end{gathered}$ |
| Winners (W) | $\begin{gathered} 0.21 \\ (1.57) \end{gathered}$ | $\begin{gathered} 0.28^{* * *} \\ (3.56) \end{gathered}$ | $\begin{gathered} 0.54^{* * *} \\ (3.24) \end{gathered}$ | $\begin{gathered} 0.33 \\ (1.43) \end{gathered}$ | $\begin{gathered} 0.87^{* * *} \\ (14.38) \end{gathered}$ | $\begin{aligned} & 1.00^{* * *} \\ & (26.22) \end{aligned}$ | $\begin{gathered} 0.90^{* * *} \\ (21.86) \end{gathered}$ | $\begin{gathered} 0.02 \\ (0.27) \end{gathered}$ | $\begin{gathered} 0.64^{* * *} \\ (7.53) \end{gathered}$ | $\begin{gathered} 0.74^{* * *} \\ (21.62) \end{gathered}$ | $\begin{gathered} 0.84^{* * *} \\ (7.37) \end{gathered}$ | $\begin{gathered} 0.22 \\ (1.19) \end{gathered}$ |
| W-L | $\begin{gathered} 0.72^{* * *} \\ (4.63) \end{gathered}$ | $\begin{gathered} 0.76^{* * *} \\ (4.27) \end{gathered}$ | $\begin{gathered} 0.71^{* * *} \\ (3.24) \end{gathered}$ | $\begin{gathered} -0.01 \\ (-0.03) \end{gathered}$ | $\begin{gathered} -0.08 \\ (-1.38) \end{gathered}$ | $\begin{gathered} -0.20^{* * *} \\ (-2.77) \end{gathered}$ | $\begin{aligned} & -0.13^{*} \\ & (-1.69) \end{aligned}$ | $\begin{gathered} -0.05 \\ (-0.93) \end{gathered}$ | $\begin{gathered} 0.23^{* * *} \\ (2.86) \end{gathered}$ | $\begin{gathered} 0.08 \\ (0.54) \end{gathered}$ | $\begin{gathered} 0.19 \\ (0.94) \end{gathered}$ | $\begin{gathered} -0.03 \\ (-0.19) \end{gathered}$ |
|  | D1 | D5 | D10 | D10-D1 | D1 | D5 | D10 | D10-D1 |  |  |  |  |
| Portfolio |  | h |  |  | Adj-Rsqr |  |  |  |  |  |  |  |
| Losers (L) | $\begin{gathered} 0.26^{* *} \\ (2.38) \end{gathered}$ | $\begin{gathered} 0.26^{* *} \\ (2.14) \end{gathered}$ | $\begin{gathered} 0.26^{* *} \\ (2.18) \end{gathered}$ | $\begin{gathered} 0 \\ (0.02) \end{gathered}$ | 0.70 | 0.82 | 0.81 | 0.07 |  |  |  |  |
| 2 | $\begin{gathered} 0.26^{* *} \\ (2.55) \end{gathered}$ | $\begin{gathered} 0.38^{* * *} \\ (6.01) \end{gathered}$ | $\begin{gathered} 0.30^{* * *} \\ (4.44) \end{gathered}$ | $\begin{gathered} 0.04 \\ (0.33) \end{gathered}$ | 0.69 | 0.91 | 0.87 | 0.04 |  |  |  |  |
| Winners (W) | $\begin{aligned} & 0.19^{*} \\ & (1.87) \end{aligned}$ | $\begin{gathered} 0.1 \\ (1.22) \end{gathered}$ | $\begin{aligned} & -0.17^{*} \\ & (-1.94) \end{aligned}$ | $\begin{gathered} -0.36^{* *} \\ (-2.35) \end{gathered}$ | 0.77 | 0.91 | 0.85 | 0.16 |  |  |  |  |
| W-L | $\begin{gathered} -0.07 \\ (-0.68) \end{gathered}$ | $\begin{gathered} -0.16 \\ (-0.89) \end{gathered}$ | $\begin{gathered} -0.43^{* * *} \\ (-2.65) \end{gathered}$ | $\begin{gathered} -0.36^{* * *} \\ (-3.75) \end{gathered}$ | 0.06 | 0.05 | 0.13 | 0.10 |  |  |  |  |




# Table 5: Seasonality in Momentum Return and Trade Size 

This table presents average monthly and characteristic-adjusted returns from portfolios strategies separated into January and then February to December. All portfolios are formed on the basis of two-way sorts based on past trade size and price momentum for the period 1983 to 2010 . We delete any firm that is in both the extreme 100 -share and the 500 (or the 100 share and the 1000 ) share trade cluster portfolios to eliminate overlapping firms. We also delete any firm with a share price $<\$ 5.00$. At the beginning of each month all available stocks listed on the NYSE/Amex/NASDAQ markets are sorted based on the past 6 months returns, and divided into three portfolios. The loser portfolio represents the worst prior return performance and the winner portfolio represents the best prior return performance. The stocks are then sorted based on the average trade size over the month just prior to the monthly evaluation period. We divide the trade size portfolios into decile portfolios. Monthly returns are computed based on the portfolio rebalancing strategy described in Table 1 and characteristic-adjusted returns are based on firm size, book-to-market, and momentum using the portfolios derived from Daniel et al. (1997) and Wermers (2003). We report the winner-loser (W-L) portfolios for the base momentum strategy (W-L Base), then based on the extreme trade size quintiles (W-L quintiles 1 or 10 ), and finally relative to the base momentum strategy (W-L quintile 1 or 10 minus the base momentum strategy). T-statistics are presented in the parentheses and these are Newey-West corrected with four lags. Significance is reported with an ${ }^{*}\left(10 \%\right.$ significance), an ${ }^{* *}$ ( $5 \%$ significance), or an ${ }^{* * *}$ ( $1 \%$ significance).

| Period | Portfolio | Average | hly Return | Characteristic-Adjusted Return |  |
| :---: | :---: | :---: | :---: | :---: | :---: |
| Panel A: Base Strategy Momentum Portfolios Returns |  |  |  |  |  |
| Jan. | W-L | -0.96 | (-1.52) | 0.01 | (0.02) |
| Feb.-Dec. | W-L | $0.66^{* * *}$ | (3.03) | 0.49*** | (3.03) |
| Panel B: Double Sort on Momentum and 100-Share Trade Size |  |  |  |  |  |
| Jan. | W-L (Decile 10) | -0.95* | (-1.86) | 0.19 | (0.40) |
| Jan. | W-L (Decile 1) | -0.15 | (-0.20) | 0.62 | (0.87) |
| Jan. | W-L (Decile 10) - W-L (Base) | 0.01 | (0.03) | 0.18 | (0.41) |
| Jan. | W-L (Decile 1) - W-L (Base) | 0.81* | (1.81) | 0.61 | (1.47) |
| Feb.-Dec. | W-L (Decile 10) | $0.47^{* *}$ | (2.02) | 0.32* | (1.66) |
| Feb.-Dec. | W-L (Decile 1) | 0.86 *** | (4.32) | 0.71 *** | (4.43) |
| Feb.-Dec. | W-L (Decile 10) - W-L (Base) | -0.19** | (-2.21) | -0.17* | (-1.88) |
| Feb.-Dec. | W-L (Decile 1) - W-L (Base) | 0.19* | (1.85) | $0.21 * *$ | (2.00) |
| Panel C: Double Sort on Momentum and 500 Share Trade Size |  |  |  |  |  |
| Jan. | W-L (Decile 10) | -0.37 | (-0.59) | 0.7 | (1.29) |
| Jan. | W-L (Decile 1) | -0.97** | (-2.13) | -0.03 | (-0.05) |
| Jan. | W-L (Decile 10) - W-L (Base) | 0.58* | (2.04) | 0.69** | (2.45) |
| Jan. | W-L (Decile 1) - W-L (Base) | -0.02 | (-0.04) | -0.04 | (-0.12) |
| Feb.-Dec. | W-L (Decile 10) | 0.92 *** | (4.99) | $0.77^{* * *}$ | (5.16) |
| Feb.-Dec. | W-L (Decile 1) | 0.30** | (2.00) | 0.17 | (1.54) |
| Feb.-Dec. | W-L (Decile 10) - W-L (Base) | $0.26{ }^{* *}$ | (2.38) | $0.28 * * *$ | (2.64) |
| Feb.-Dec. | W-L (Decile 1) - W-L (Base) | -0.37*** | (-3.50) | -0.32*** | (-3.51) |
| Panel D: Double Sort on Momentum and 1000 Share Trade Size |  |  |  |  |  |
| Jan. | W-L (Decile 10) | -0.5 | (-0.77) | 0.5 | (0.89) |
| Jan. | W-L (Decile 1) | -0.98* | (-1.90) | 0.07 | (0.13) |
| Jan. | W-L (Decile 10) - W-L (Base) | 0.46 | (1.54) | 0.49* | (1.75) |
| Jan. | W-L (Decile 1) - W-L (Base) | -0.02 | (-0.06) | 0.06 | (0.18) |
| Feb.-Dec. | W-L (Decile 10) | 0.92 *** | (4.05) | $0.74 * * *$ | (4.17) |
| Feb.-Dec. | W-L (Decile 1) | $0.42{ }^{* *}$ | (2.53) | 0.30 ** | (2.21) |
| Feb.-Dec. | W-L (Decile 10) - W-L (Base) | $0.25 * * *$ | (2.67) | $0.25 * * *$ | (2.82) |
| Feb.-Dec. | W-L (Decile 1) - W-L (Base) | $-0.24^{* *}$ | (-2.36) | -0.20** | (-2.14) |

## Table 6: Portfolios Formed on Trade Size Quintiles and Price Momentum Terciles

This table presents average monthly and characteristic-adjusted returns from portfolios strategies formed on the basis of twoway sorts based on past trade size and price momentum for the period 1983 to 2010. We delete any firm that is in both the extreme 100 -share and the 500 (or the 100 share and the 1000 ) share trade cluster portfolios to eliminate overlapping firms. We also delete any firm with a share price $<\$ 5.00$. At the beginning of each month all available stocks listed on the NYSE/Amex/NASDAQ markets are sorted based on the past 6 months returns, and divided into 3 portfolios. The loser portfolio represents the worst prior return performance and the winner portfolio represents the best prior return performance. The stocks are then sorted based on the average trade size over the month just prior to the monthly evaluation period. We divide the trade size portfolios into quintile portfolios. We analyze performance of three separate holding periods using a six-month period (1-6), a twelve-month period (1-12), and a holding period that spans months 13 to 24 from the formation period. Monthly returns are computed based on the portfolio rebalancing strategy described in Table 1 and characteristic-adjusted returns are based on firm size, book-to-market, and momentum using the portfolios derived from Daniel et al. (1997) and Wermers (2003). We report the winner-loser (W-L) portfolios for the base momentum strategy (W-L Base), then based on the extreme trade size quintiles (W-L quintiles 1 or 5 ), and finally relative to the base momentum strategy ( $\mathrm{W}-\mathrm{L}$ quintile 1 or 5 minus the base momentum strategy). For each extreme portfolio, we report the average bid-ask spread and the price impact measure of Amihud. We also present firm size expressed in millions of dollars and the average price of the portfolio of stocks at the end of the month prior to the performance period. T-statistics are presented in the parentheses and these are Newey-West corrected with four lags. Significance is reported with an ${ }^{*}(10 \%$ significance $)$, an ${ }^{* *}$ ( $5 \%$ significance), or an ${ }^{* * *}$ ( $1 \%$ significance).

| Momentum Tercile | Trade Size | Number of Firms | Average Price | $\begin{gathered} \text { Firm } \\ \text { Size } \\ (\$ \text { millions }) \end{gathered}$ | Bid-Ask <br> Spreads <br> (\%) | Amihud's measure | $\begin{aligned} & \text { \% Monthly C } \\ & 1-6 \end{aligned}$ |  | $\begin{aligned} & \text { Characteristic-Adjusted Returns } \\ & 1-12 \end{aligned}$ |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | Quintile |  |  |  |  |  | Return | T | Return | T | Return | T |
| Panel A: Base Six Month Momentum Strategy, Single-Sort by previous six-month returns |  |  |  |  |  |  |  |  |  |  |  |  |
| Losers (L) |  | 761 | 18 | 1519 | 2.10 | 0.14 | $-0.26^{* * *}$ | (-2.68) | ) $-0.17^{*}$ | (-1.85) | 0.09 | (1.1) |
| 2 |  | 759 | 26 | 2475 | 1.80 | 0.13 | -0.01 | (-0.26) | 0 | (-0.06) | 0.02 | (0.4) |
| Winners (W) |  | 759 | 26 | 2070 | 1.78 | 0.10 | $0.19{ }^{* * *}$ | (3.59) | $0.10^{* *}$ | (2.55) | -0.04 | (-0.53) |
| W-L (Base) |  |  |  |  |  |  | $0.46{ }^{* * *}$ | (3.24) | $0.27^{* *}$ | (2.43) | -0.12 | (-1.36) |
| Panel B: Double Sort by Previous six-month returns, then previous one-month trade size $=100$ shares |  |  |  |  |  |  |  |  |  |  |  |  |
| Losers (L) | 1 | 67 | 19 | 2650 | 2.80 | 0.37 | $-0.23 * *$ | (-2.06) | - $0.19^{*}$ | (-1.86) | 0.09 | (1) |
|  | 5 | 63 | 19 | 844 | 2.08 | 0.11 | -0.19* | (-1.80) | -0.15 | (-1.56) | 0.05 | (0.51) |
| 2 | 1 | 63 | 25 | 3091 | 2.63 | 0.36 | -0.14 | (-1.31) | ) -0.09 | (-0.86) | 0.01 | (0.09) |
|  | 5 | 65 | 27 | 1888 | 1.65 | 0.08 | 0.02 | (0.25) | 0.03 | (0.38) | -0.03 | (-0.32) |
| Winners (W) | 1 | 66 | 24 | 2654 | 2.40 | 0.23 | 0.11 | (1.2) | 0.08 | (0.89) | -0.1 | (-0.89) |
|  | 5 | 62 | 29 | 1561 | 1.71 | 0.08 | $0.33^{* * *}$ | (2.89) | 0.20** | (2.19) | 0 | (0.01) |
| W-L (Quintile 5) |  |  |  |  |  |  | 0.53 *** | (2.93) | 0.35** | (2.38) | -0.05 | (-0.43) |
| W-L (Quintile 1) |  |  |  |  |  |  | 0.34** | (2.58) | $0.27 * * *$ | (2.62) | -0.19* | (-1.68) |
| W-L (Quintile 5) - W-L (Base) |  |  |  |  |  |  | 0.07 | (0.71) | 0.08 | (1.05) | 0.07 | (1.11) |
| W-L (Quintile 1) - W-L (Base) |  |  |  |  |  |  | -0.12 | (-1.22) | ) 0 | (0.02) | -0.07 | (-0.86) |


| Losers (L) | 1 | 103 | 21 | 1507 | 2.48 | 0.25 | -0.07 | (-0.69) | -0.07 | (-0.79) | 0.01 | (0.19) |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | 5 | 120 | 14 | 901 | 2.51 | 0.17 | $-0.57^{* * *}$ | (-4.44) | $-0.46^{* * *}$ | (-3.94) | -0.07 | (-0.63) |
| 2 | 1 | 101 | 29 | 2136 | 2.24 | 0.25 | -0.06 | (-0.62) | -0.01 | (-0.05) | 0.02 | (0.19) |
|  | 5 | 116 | 19 | 1810 | 2.26 | 0.15 | 0 | (-0.04) | -0.03 | (-0.41) | -0.13 | (-1.36) |
| Winners (W) | 1 | 101 | 31 | 2291 | 1.90 | 0.16 | 0.11 | (1.17) | 0.12 | (1.43) | 0.08 | $(1.08)$ |
|  | 5 | 117 | 19 | 1230 | 2.22 | 0.12 | 0.30*** | (3.69) | 0.07 | (0.92) | $-0.27 * *$ | (-2.34) |
| W-L (Quintile 5) |  |  |  |  |  |  | $0.87 * * *$ | (5.6) | $0.53^{* * *}$ | (4.64) | -0.21* | (-1.87) |
| W-L (Quintile 1) |  |  |  |  |  |  | 0.18 | (1.62) | 0.19** | (2.23) | 0.07 | (0.9) |
| W-L (Quintile 5) - W-L (Base) |  |  |  |  |  |  | $0.42^{* * *}$ | (5.73) | 0.26*** | (4.43) | -0.09 | (-1.52) |
| W-L (Quintile 1) - W-L (Base) |  |  |  |  |  |  | $-0.27^{* * *}$ | (-3.40) | -0.08 | (-1.27) | $0.19 * * *$ | (2.78) |


| Panel : Double Sort by Previous six-month returns, then previous one-month trade size $=1000$ shares |  |  |  |  |  |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Losers (L) | 1 | 92 | 24 | 1217 | 2.15 | 0.25 | -0.13 | (-1.24) | -0.1 | (-1.08) | -0.03 | (-0.37) |
|  | 5 | 100 | 13 | 875 | 2.35 | 0.14 | $-0.51 * * *$ | (-3.49) | -0.31** | (-2.44) | 0.06 | (0.44) |
| 2 | 1 | 95 | 30 | 1279 | 2.06 | 0.24 | -0.11 | (-0.93) | -0.05 | (-0.46) | -0.01 | (-0.10) |
|  | 5 | 96 | 18 | 1659 | 2.18 | 0.13 | 0 | (0.04) | 0 | (-0.06) | 0.04 | (0.58) |
| Winners (W) | 1 | 91 | 33 | 1895 | 1.74 | 0.15 | 0.07 | (0.68) | 0.06 | (0.59) | 0.03 | (0.39) |
|  | 5 | 98 | 19 | 1323 | 1.91 | 0.09 | 0.30** | (2.55) | 0.09 | (0.83) | -0.08 | (-0.72) |
| W-L (Quintile 5) |  |  |  |  |  |  | 0.81*** | (4.6) | 0.40*** | (3.22) | -0.14 | (-1.23) |
| W-L (Quintile 1) |  |  |  |  |  |  | 0.20* | (1.81) | 0.16* | (1.82) | 0.07 | (0.97) |
| W-L (Quintile 5) - W-L (Base) |  |  |  |  |  |  | $0.35 * * *$ | (4.86) | $0.13 * *$ | (2) | -0.02 | $(-0.29)$ |
| W-L (Quintile 1) - W-L (Base) |  |  |  |  |  |  | $-0.25 * * *$ | (-2.93) | -0.11* | (-1.69) | 0.19*** | (2.99) |

## Table 7: Price and Dollar Trading Volume Filters

This table presents average monthly and characteristic-adjusted returns from portfolios strategies formed on the basis of two-way sorts based on past trade size and price momentum for the period 1983 to 2010 . For this table, we delete any firm with a share price $<\$ 5.00$ and with a total monthly dollar volume of trade $\leq \$ 2,000,000$. Additionally, we delete any firm that is in both the extreme 100-share and the 500 share trade portfolios to eliminate overlapping firms. At the beginning of each month all available stocks listed on the NYSE/Amex/NASDAQ markets are sorted independently based on the past 6 months returns, and divided into 3 portfolios. The loser portfolio represents the worst prior return performance and the winner portfolio represents the best prior return performance. The stocks are then sorted based on the average trade size for the last month of the portfolio formation period. We skip one month between the portfolio formation period and the performance period. We divide the trade size portfolios into 10 portfolios. The stocks at the intersection of the two sorts are grouped together to form portfolios based on past returns and past trade size. We analyze performance of three separate holding periods using a six-month period (1-6), a twelve-month period (1-12), and a holding period that spans months 13 to 24 from the formation period. Monthly returns are computed based on the portfolio rebalancing strategy described in Table 1 and characteristic-adjusted returns are based on firm size, book-to-market, and momentum using the portfolios derived from Daniel et al. (1997) and Wermers (2003). We report the winner-loser portfolios for the base momentum strategy ( W -L Base), then based on the extreme trade size deciles (W-L Deciles 1 or 10), and finally relative to the base momentum strategy (W-L Decile 1 or 10 minus the base momentum strategy). For each extreme portfolio, we report the average bid-ask spread and the price impact measure of Amihud. We also present firm size expressed in millions of dollars and the average price of the portfolio of stocks at the end of the month prior to the performance period. T-statistics are presented in the parentheses and these are Newey-West corrected with four lags. Significance is reported with an ${ }^{*}(10 \%$ significance $)$, an ${ }^{* *}$ ( $5 \%$ significance), or an ${ }^{* * *}$ ( $1 \%$ significance).

| Momentum Tercile | Trade Size Decile | Number of Firms | Average Price | $\begin{gathered} \text { Firm } \\ \text { Size } \\ (\$ \text { millions }) \end{gathered}$ | Bid-Ask <br> Spreads (\%) | Amihud's measure | $\begin{aligned} & \text { \% Monthly } \\ & 1-6 \end{aligned}$ |  | Characteristic-Adjusted Returns$1-12$ |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  |  |  |  |  |  | Return | T | Return | T | Return | T |
| Panel A: Base Six Month Momentum Strategy, Single-Sort by previous six-month returns |  |  |  |  |  |  |  |  |  |  |  |  |
| Losers (L) |  | 724 | 20 | 2022 | 1.47 | 0.01 | $-0.27^{* * *}$ | (-2.67) | ) $-0.17^{*}$ | (-1.82) | 0.11 | (1.18) |
| 2 |  | 707 | 29 | 3348 | 1.17 | 0.01 | -0.01 | (-0.32) | ) 0 | (-0.02) | 0.04 | (0.91) |
| Winners (W) |  | 760 | 29 | 2564 | 1.27 | 0.01 | $0.16{ }^{* * *}$ | (2.71) | ) $0.07^{*}$ | (1.8) | -0.02 | (-0.37) |
| W-L (Base) |  |  |  |  |  |  | $0.43 * * *$ | (2.98) | ) $0.24^{* *}$ | (2.2) | -0.13 | (-1.36) |
| Panel B: Double Sort by Previous six-month returns, then previous one-month trade size $=100$ shares |  |  |  |  |  |  |  |  |  |  |  |  |
| Losers (L) | 1 | 36 | 21 | 5192 | 1.88 | 0.02 | -0.38** | (-2.53) | ) -0.19 | (-1.40) | 0.29** | (2.1) |
|  | 10 | 32 | 25 | 1682 | 1.31 | 0.01 | -0.16 | (-1.55) | ) -0.07 | (-0.87) | 0.08 | (0.75) |
| 2 | 1 | 33 | 28 | 7017 | 1.64 | 0.02 | -0.08 | (-0.98) | ) -0.07 | (-0.73) | 0.17** | (2) |
|  | 10 | 32 | 34 | 3271 | 1.03 | 0.01 | -0.01 | (-0.16) | ) -0.01 | (-0.19) | -0.03 | (-0.34) |
| Winners (W) | 1 | 39 | 25 | 4773 | 1.74 | 0.02 | 0.13 | (1.29) | 0.04 | (0.39) | -0.08 | (-0.67) |
|  | 10 | 33 | 37 | 2422 | 1.12 | 0.01 | 0.32** | (2.34) | ) $0.25 * *$ | (2.29) | 0.02 | (0.18) |
| W-L (Decile 10) |  |  |  |  |  |  | $0.48^{* * *}$ | (2.61) | ) $0.32^{* *}$ | (2.21) | -0.06 | (-0.44) |
| W-L (Decile 1) |  |  |  |  |  |  | $0.51^{* * *}$ | (2.82) | ) 0.23 | (1.56) | $-0.37 * * *$ | (-2.68) |
| W-L (Decile 10) - W-L (Base) |  |  |  |  |  |  | 0.05 | (0.41) | ) 0.08 | (0.81) | 0.07 | (0.89) |
| W-L (Decile 1) - W-L (Base) |  |  |  |  |  |  | 0.08 | (0.69) | ) -0.02 | (-0.17) | $-0.24^{* *}$ | (-2.32) |


| Panel C: Double Sort by Previous six-month returns, then previous one-month trade size $=500$ shares |  |  |  |  |  |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Losers (L) | 1 | 50 | 26 | 1574 | 1.55 | 0.02 | 0.01 | (0.1) | 0.02 | (0.25) | 0.05 | (0.6) |
|  | 10 | 57 | 16 | 1320 | 1.77 | 0.02 | $-0.72^{* * *}$ | (-4.71) | $-0.62^{* * *}$ | (-4.57) | -0.17 | (-1.27) |
| 2 | 1 | 48 | 35 | 2369 | 1.27 | 0.02 | 0.06 | (0.68) | 0.09 | (0.96) | 0.06 | (0.59) |
|  | 10 | 54 | 22 | 2270 | 1.50 | 0.02 | -0.03 | (-0.47) | -0.06 | (-0.84) | -0.1 | (-1.17) |
| Winners (W) | 1 | 51 | 36 | 2355 | 1.20 | 0.01 | 0.03 | (0.33) | 0.08 | (0.91) | 0.12 | (1.5) |
|  | 10 | 59 | 21 | 1163 | 1.60 | 0.02 | 0.30** | (2.51) | 0.02 | (0.23) | -0.33** | (-2.42) |
| W-L (Decile 10) |  |  |  |  |  |  | $1.02{ }^{* * *}$ | (5.65) | $0.65 * * *$ | (4.93) | -0.16 | (-1.27) |
| W-L (Decile 1) |  |  |  |  |  |  | 0.02 | (0.18) | 0.05 | (0.6) | 0.07 | (0.97) |
| W-L (Decile 10) - W-L (Base) |  |  |  |  |  |  | 0.59*** | (5.76) | 0.40 *** | (4.76) | -0.03 | (-0.37) |
| W-L (Decile 1) - W-L (Base) |  |  |  |  |  |  | $-0.41^{* * *}$ | (-4.26) | $-0.19^{* *}$ | (-2.52) | 0.20 *** | (2.75) |


| Losers (L) | 1 | 44 | 30 | 1438 | 1.12 | 0.01 | -0.11 | (-1.05) | -0.08 | (-0.86) | -0.01 | (-0.11) |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | 10 | 46 | 13 | 1372 | 1.70 | 0.02 | $-0.58^{* * *}$ | (-3.15) | -0.37** | (-2.30) | 0.06 | (0.31) |
| 2 | 1 | 44 | 37 | 1401 | 1.01 | 0.01 | -0.01 | (-0.09) | 0.02 | (0.18) | -0.01 | (-0.12) |
|  | 10 | 43 | 19 | 2280 | 1.48 | 0.02 | -0.01 | (-0.15) | -0.01 | (-0.15) | 0.14 | (1.42) |
| Winners (W) | 1 | 46 | 39 | 1792 | 0.98 | 0.01 | -0.03 | (-0.28) | 0 | (-0.04) | 0.02 | (0.27) |
|  | 10 | 48 | 20 | 1401 | 1.40 | 0.02 | 0.26* | (1.72) | 0.02 | (0.16) | -0.02 | (-0.16) |
| W-L (Decile 10) |  |  |  |  |  |  | $0.84 * * *$ | (4.43) | 0.39*** | (2.89) | -0.08 | (-0.54) |
| W-L (Decile 1) |  |  |  |  |  |  | 0.08 | (0.63) | 0.07 | (0.82) | 0.03 | (0.4) |
| W-L (Decile 10) - W-L (Base) |  |  |  |  |  |  | 0.42 *** | (3.67) | 0.15 | (1.52) | 0.05 | (0.47) |
| W-L (Decile 1) - W-L (Base) |  |  |  |  |  |  | $-0.35^{* * *}$ | (-3.11) | $-0.17 * *$ | (-2.18) | 0.16** | (2.4) |

## Table 8: Trade Size and Price Momentum: NYSE/Amex and NASDAQ Listed Firms Separately

This table presents average monthly and characteristic-adjusted returns from portfolios strategies formed on the basis of twoway sorts based on past trade size and price momentum for the period 1983 to 2010 . We delete any firm that is in both the extreme 100 -share and the 500 (or greater then 500) share trade portfolios to eliminate overlapping firms. We also delete any firm with a share price $<\$ 5.00$. Panel 1 contains only NYSE/Amex listed firms and Panel 2 contains only NASDAQ listed firms. At the beginning of each month all available stocks listed on the NYSE/Amex (Panel 1) or NASDAQ (Panel 2) markets are sorted based on the past 6 months returns, and divided into 3 portfolios. The loser portfolio represents the worst prior return performance and the winner portfolio represents the best prior return performance. The stocks are then sorted based on the average trade size over the month just prior to the monthly evaluation period. We divide the trade size portfolios into decile portfolios. We analyze performance of three separate holding periods using a six-month period (1-6), a twelve-month period (1-12), and a holding period that spans months 13 to 24 from the formation period. Monthly returns are computed based on the portfolio rebalancing strategy described in Table 1 and characteristic-adjusted returns using the characteristics of firm size, book-to-market, and momentum. We report the winner-loser (W-L) portfolios for the base momentum strategy (W-L Base), then based on the extreme trade size deciles ( $\mathrm{W}-\mathrm{L}$ deciles 1 or 10 ), and finally relative to the base momentum strategy ( $\mathrm{W}-\mathrm{L}$ deciles 1 or 10 minus the base momentum strategy). For each extreme portfolio, we report the average bid-ask spread and the price impact measure of Amihud. We also present firm size expressed in millions of dollars and the average price of the portfolio of stocks at the end of the month prior to the performance period. Newey-West robust estimators with four lags specifying the t-statistics are presented in the parentheses. Significance is reported with an * $(10 \%$ significance $)$, an $* *$ ( $5 \%$ significance), or an ${ }^{* * *}$ ( $1 \%$ significance).

| Momentum <br> Tercile | Trade Size Decile | Number of Firms | Average Price | Firm <br> Size <br> (\$ millions) | Bid-ask Spreads (\%) | Amihud's measure | $\begin{aligned} & \text { \% Monthly } \\ & 1-6 \end{aligned}$ |  | Characteristic-Adjusted Returns$1-12 \quad 13-24$ |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  |  |  |  |  |  | Return | T | Return | T | Return | T |
| Panel A: Base Six Month Momentum Strategy, Single-Sort by previous six-month returns |  |  |  |  |  |  |  |  |  |  |  |  |
| Losers (L) |  | 420 | 23 | 3384 | 1.38 | 0.03 | -0.17 | (-1.51) | ) -0.11 | (-0.97) | 0.06 | (0.53) |
| 2 |  | 474 | 32 | 4916 | 1.07 | 0.02 | 0 | (0.01) | -0.01 | (-0.13) | -0.01 | (-0.11) |
| Winners (W) |  | 404 | 32 | 4078 | 1.15 | 0.02 | 0.1 | (1.24) | 0.08 | (1.16) | -0.01 | (-0.17) |
| W-L (Base) |  |  |  |  |  |  | 0.27* | (1.88) | 0.18 | (1.6) | -0.07 | (-0.76) |



| Losers (L) | 1 | 25 | 28 | 878 | 1.61 | 0.11 | -0.21 | (-1.58) | -0.13 | (-1.08) | -0.07 | (-0.60) |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | 10 | 26 | 11 | 3337 | 2.25 | 0.07 | -0.43** | (-2.31) | -0.33** | (-2.17) | -0.13 | (-1.07) |
| 2 | 1 | 28 | 35 | 995 | 1.35 | 0.08 | -0.07 | (-0.51) | -0.07 | (-0.58) | -0.05 | (-0.39) |
|  | 10 | 27 | 16 | 4666 | 1.98 | 0.06 | 0.05 | (0.46) | -0.07 | (-0.79) | $-0.32^{* * *}$ | (-2.85) |
| Winners (W) | 1 | 23 | 37 | 1199 | 1.29 | 0.07 | 0.07 | (0.57) | 0.05 | (0.47) | 0.02 | (0.21) |
|  | 10 | 22 | 15 | 1552 | 2.15 | 0.06 | 0.16 | (1.2) | 0.03 | (0.24) | $-0.31 * *$ | (-2.31) |
| W-L (Decile 10) |  |  |  |  |  |  | $0.58^{* * *}$ | (2.73) | 0.36** | (2.35) | -0.18 | (-1.30) |
| W-L (Decile 1) |  |  |  |  |  |  | 0.28 ** | (2.01) | 0.18* | (1.68) | 0.09 | (1) |
| W-L (Decile 10) - W-L (Base) |  |  |  |  |  |  | 0.31** | (2.07) | 0.17 | (1.64) | -0.11 | (-0.82) |
| W-L (Decile 1) - W-L (Base) |  |  |  |  |  |  | 0.01 | (0.11) | 0 | (-0.04) | 0.16** | (2.16) |


| Panel 2: NASDAQ Listed Firms |  |  |  |  |  |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Momentum Tercile | Trade Size Decile | Number of Firms | Average Price | FirmSize$(\$$ millions $)$ | Bid-ask Spreads (\%) | Amihud's measure | $\begin{aligned} & \text { \% Monthly } \\ & 1-6 \end{aligned}$ |  | Characteristic-Adjusted Returns 1-12$13-24$ |  |  |  |
|  |  |  |  |  |  |  | Return | T | Return | T | Return | T |
| Panel E: Base Six Month Momentum Strategy, Single-Sort by previous six-month returns |  |  |  |  |  |  |  |  |  |  |  |  |
| Losers (L) |  | 614 | 15 | 649 | 2.69 | 0.22 | $-0.39^{* * *}$ | (-3.21) | $-0.25 * *$ | (-2.19) | 0.17 | (1.57) |
| 2 |  | 551 | 20 | 777 | 2.58 | 0.24 | -0.06 | (-0.75) | -0.03 | (-0.33) | 0.05 | (0.69) |
| Winners (W) |  | 631 | 21 | 938 | 2.24 | 0.15 | $0.29^{* * *}$ | (3.48) | 0.13* | (1.87) | -0.05 | (-0.54) |
| W-L (Base) |  |  |  |  |  |  | $0.68^{* * *}$ | (4.08) | $0.37^{* * *}$ | (2.79) | -0.21* | (-1.82) |
| Panel F: Double Sort by Previous six-month returns, then previous one-month trade size $=100$ shares |  |  |  |  |  |  |  |  |  |  |  |  |
| Losers (L) | 1 | 31 | 11 | 848 | 4.36 | 0.79 | $-0.67^{* * *}$ | (-3.98) | $-0.48^{* * *}$ | (-2.95) | -0.08 | (-0.56) |
|  | 10 | 30 | 21 | 438 | 2.91 | 0.20 | $-0.34^{* *}$ | (-2.26) | -0.22* | (-1.87) | 0.19 | (1.6) |
| 2 | 1 | 29 | 14 | 606 | 4.39 | 0.86 | $-0.43 * *$ | (-2.23) | -0.33* | (-1.77) | -0.05 | (-0.33) |
|  | 10 | 29 | 29 | 513 | 2.65 | 0.19 | 0 | 0 | 0.03 | (0.26) | -0.01 | (-0.04) |
| Winners (W) | 1 | 33 | 13 | 737 | 3.67 | 0.54 | 0.14 | (0.98) | -0.02 | (-0.18) | -0.07 | (-0.49) |
|  | 10 | 29 | 33 | 726 | 2.22 | 0.13 | 0.49*** | (2.91) | 0.33** | (2.33) | -0.05 | (-0.36) |
| W-L (Decile 10) |  |  |  |  |  |  | $0.82^{* * *}$ | (3.53) | 0.56 *** | (2.86) | -0.24 | (-1.43) |
| W-L (Decile 1) |  |  |  |  |  |  | $0.81 * * *$ | (4.64) | 0.46 *** | (2.85) | 0.01 | (0.06) |
| W-L (Decile 10) - W-L (Base) |  |  |  |  |  |  | 0.14 | (1.12) | 0.18* | (1.76) | -0.03 | (-0.25) |
| W-L (Decile 1) - W-L (Base) |  |  |  |  |  |  | 0.13 | (0.78) | 0.08 | (0.58) | 0.22 | (1.62) |
| Panel G: Double Sort by Previous six-month returns, then previous one-month trade size $=500$ shares |  |  |  |  |  |  |  |  |  |  |  |  |
| Losers (L) | 1 | 42 | 18 | 364 | 4.16 | 0.62 | -0.22 | (-1.41) | -0.14 | (-1.00) | -0.07 | (-0.60) |
|  | 10 | 44 | 12 | 371 | 4.08 | 0.54 | -0.63*** | (-3.79) | $-0.54^{* * *}$ | (-3.50) | -0.23* | (-1.70) |
| 2 | 1 | 38 | 22 | 292 | 4.25 | 0.64 | -0.17 | (-0.94) | -0.13 | (-0.72) | -0.13 | (-0.80) |
|  | 10 | 40 | 15 | 641 | 3.92 | 0.49 | -0.06 | (-0.38) | -0.09 | (-0.62) | -0.12 | (-0.88) |
| Winners (W) | $1$ | 44 | 26 | $594$ | $3.10$ | $0.40$ | $-0.05$ | $(-0.37)$ | 0.01 | (0.04) | $0.08$ | (0.62) |
|  |  |  | 16 |  |  |  | $0.33^{* *}$ | $(2.14)$ | 0.02 | (0.17) | -0.35* | (-1.94) |
| W-L (Decile 10) |  |  |  |  |  |  | 0.96*** | (4.89) | 0.56*** | (3.57) | -0.12 | (-0.74) |
| W-L (Decile 1) |  |  |  |  |  |  | 0.17 | (1.31) | 0.15 | (1.44) | 0.15 | (1.3) |
| W-L (Decile 10) - W-L (Base) |  |  |  |  |  |  | 0.28** | (1.99) | 0.19* | (1.67) | 0.1 | (0.81) |
| W-L (Decile 1) - W-L (Base) |  |  |  |  |  |  | -0.51*** | (-4.33) | $-0.22^{* *}$ | (-2.29) | 0.36 *** | (3.58) |
| Panel H: Double Sort by Previous six-month returns, then previous one-month trade size $=1000$ shares |  |  |  |  |  |  |  |  |  |  |  |  |
| Losers (L) | 1 | 39 | 18 | 284 | 4.79 | 0.77 | $-0.43^{* *}$ | (-2.27) | $-0.36{ }^{* *}$ | (-2.13) | -0.13 | (-0.83) |
|  | 10 | 43 | 12 | 461 | 2.79 | 0.35 | $-0.64^{* * *}$ | (-3.67) | -0.45*** | (-2.97) | 0.03 | (0.14) |
| 2 | 1 | 38 | 22 | 163 | 4.85 | 0.75 | -0.23 | (-1.12) | -0.13 | (-0.63) | -0.09 | (-0.48) |
|  | 10 | 41 | 16 | 566 | 2.70 | 0.37 | -0.19 | (-1.39) | -0.06 | (-0.49) | 0.17 | (1.5) |
| Winners (W) | 1 | 38 | 25 | 448 | 3.95 | 0.47 | -0.13 | (-0.72) | -0.07 | (-0.42) | -0.09 | (-0.54) |
|  | 10 | 45 | 18 | 554 | 2.14 | 0.20 | 0.23 | (1.28) | 0.05 | (0.31) | 0.1 | (0.57) |
| W-L (Decile 10) |  |  |  |  |  |  | $0.87^{* * *}$ | (4.17) | 0.50*** | (3.17) | 0.07 | (0.43) |
| W-L (Decile 1) |  |  |  |  |  |  | $0.31 * *$ | (2.07) | 0.29 *** | (2.77) | 0.04 | (0.39) |
| W-L (Decile 10) - W-L (Base) |  |  |  |  |  |  | 0.19** | (1.96) | 0.13 | (1.1) | 0.29** | (2.23) |
| W-L (Decile 1) - W-L (Base) |  |  |  |  |  |  | $-0.37 * * *$ | (-2.65) | -0.08 | (-0.74) | 0.26** | (2.51) |

## Table 9: Trade Size Portfolios Formed Before Momentum Portfolios

This table presents average monthly and characteristic-adjusted returns from trade size portfolios formed prior to the price momentum portfolio formation for the period 1983 to 2010 . We delete any firm that is in both the extreme 100-share and the 500 share trade portfolios to eliminate overlapping firms. We also delete any firm with a share price $<\$ 5.00$. At the beginning of each month all available stocks listed on the NYSE/Amex/NASDAQ markets are sorted independently based on the past 6 months returns, and divided into three portfolios. The stocks are first sorted based on the average trade size over the month just prior to the price momentum portfolio formation where the loser portfolio represents the worst prior price performance and the winner portfolio represents the best prior price performance. We divide the trade size portfolios into 10 portfolios. We skip one month between the momentum portfolio formation period and the performance period. We analyze the performance across three separate holding periods using a six-month period (1-6), a twelve-month period (1-12), and a holding period that spans months 13 to 24 from the formation period. Monthly returns are computed based on the portfolio rebalancing strategy described in Table 1 and characteristic-adjusted returns are based on firm size, book-to-market, and momentum characteristics. We report the winner-loser portfolios for the base momentum strategy ( W -L Base), then based on the extreme trade size deciles (W-L Deciles 1 or 10 ), and finally relative to the base momentum strategy (W-L Decile 1 or 10 minus the base momentum strategy). For each extreme portfolio, we report the average bid-ask spread and the price impact measure of Amihud. We also present firm size expressed in millions of dollars and the average price of the portfolio of stocks at the end of the month prior to the performance period. Newey-West robust estimators with four lags specifying the t-statistics are presented in the parentheses. Significance is reported with an * ( $10 \%$ significance), an ${ }^{* *}$ ( $5 \%$ significance), or an ${ }^{* * *}$ ( $1 \%$ significance).

| Momentum Tercile | Trade Size Decile | Number of Firms | Average Price | Firm Size (\$ millions) | Bid-ask Spreads (\%) | Amihud's measure | \% Monthly Characteristic-Adjusted Returns$1-6$$1-12$ |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  |  |  |  |  |  | Return | T | Return | T | Return | T |
| Panel A: Base Six Month Momentum Strategy, Single-Sort by previous six-month returns |  |  |  |  |  |  |  |  |  |  |  |  |
| Losers (L) |  | 856 | 19 | 1796 | 1.98 | 0.12 | $-0.26^{* * *}$ | (-2.69) | -0.17* | (-1.90) | 0.08 | (1.16) |
| 2 |  | 859 | 27 | 2894 | 1.67 | 0.11 | -0.02 | (-0.39) | -0.01 | (-0.19) | 0.02 | (0.39) |
| Winners (W) |  | 773 | 29 | 2541 | 1.53 | 0.07 | 0.15*** | (2.65) | 0.09** | (2.15) | 0.01 | (0.19) |
| W-L (Base) |  |  |  |  |  |  | $0.41^{* * *}$ | (2.93) | $0.26{ }^{* *}$ | (2.37) | -0.07 | (-0.83) |


| Panel B: Double Sort by Previous six-month returns, then previous one-month trade size $=100$ shares |  |  |  |  |  |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Losers (L) | 1 | 44 | 18 | 3898 | 2.99 | 0.42 | $-0.37^{* * *}$ | (-2.91) | -0.28** | (-2.27) | 0.06 | (0.64) |
|  | 10 | 39 | 22 | 1375 | 1.97 | 0.10 | -0.15 | (-1.34) | -0.11 | (-1.04) | 0.01 | (0.08) |
| 2 | 1 | 42 | 24 | 4483 | 2.77 | 0.40 | -0.18 | (-1.43) | -0.11 | (-0.92) | -0.08 | (-0.81) |
|  | 10 | 41 | 31 | 2618 | 1.57 | 0.08 | 0.01 | (0.07) | -0.01 | (-0.07) | -0.06 | (-0.61) |
| Winners (W) | 1 | 39 | 25 | 3826 | 2.35 | 0.24 | 0.09 | (0.95) | 0.01 | (0.14) | -0.08 | (-0.84) |
|  | 10 | 36 | 36 | 2942 | 1.52 | 0.06 | 0.18 | (1.61) | 0.15 | (1.59) | 0 | (-0.02) |
| W-L (Decile 10) |  |  |  |  |  |  | $0.34 * *$ | (2.1) | 0.26** | (1.98) | -0.01 | (-0.10) |
| W-L (Decile 1) |  |  |  |  |  |  | 0.46 *** | (3.56) | 0.30*** | (2.79) | -0.15 | (-1.17) |
| W-L (Decile 10) - W-L (Base) |  |  |  |  |  |  | -0.07 | (-0.77) | 0 | (0.04) | 0.06 | (0.78) |
| W-L (Decile 1) - W-L (Base) |  |  |  |  |  |  | 0.05 | (0.42) | 0.04 | (0.39) | -0.08 | (-0.78) |


| Losers (L) | 1 | 58 | 21 | 1249 | 2.84 | 0.38 | -0.11 | (-0.94) | -0.1 | (-0.96) | -0.07 | (-0.67) |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | 10 | 67 | 13 | 907 | 2.84 | 0.27 | $-0.63^{* * *}$ | (-4.98) | $-0.47^{* * *}$ | (-3.82) | -0.19* | (-1.89) |
| 2 | 1 | 58 | 28 | 1131 | 2.62 | 0.33 | -0.04 | (-0.31) | 0 | (-0.00) | -0.05 | (-0.39) |
|  | 10 | 66 | 18 | 1574 | 2.60 | 0.25 | -0.09 | (-0.81) | -0.14 | (-1.23) | -0.20* | (-1.74) |
| Winners (W) | 1 | 52 | 32 | 1530 | 2.19 | 0.17 | 0.14 | (1.26) | 0.13 | (1.24) | 0.03 | (0.34) |
|  | 10 | 59 | 21 | 1025 | 2.22 | 0.15 | 0.13 | (1.32) | -0.04 | (-0.46) | -0.26 ** | (-1.99) |
| W-L (Decile 10) |  |  |  |  |  |  | 0.76 *** | (5.63) | $0.42^{* * *}$ | (3.66) | -0.08 | (-0.78) |
| W-L (Decile 1) |  |  |  |  |  |  | $0.25 * *$ | (2.4) | 0.23 *** | (2.62) | 0.1 | (1.12) |
| W-L (Decile 10) - W-L (Base) |  |  |  |  |  |  | 0.35*** | (4.38) | $0.17{ }^{* *}$ | (2.36) | 0 | (-0.03) |
| W-L (Decile 1) - W-L (Base) |  |  |  |  |  |  | -0.16 | (-1.64) | -0.03 | (-0.37) | $0.18 * *$ | (2.27) |


| Losers (L) | 1 | 53 | 22 | 601 | 2.66 | 0.39 | -0.21 | (-1.58) | -0.18 | (-1.47) | -0.08 | (-0.64) |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | 10 | 58 | 13 | 1201 | 2.41 | 0.20 | $-0.45 * * *$ | (-3.42) | -0.31** | (-2.58) | -0.03 | (-0.21) |
| 2 | 1 | 55 | 29 | 641 | 2.50 | 0.34 | -0.07 | (-0.50) | -0.04 | (-0.25) | -0.09 | (-0.63) |
|  | 10 | 57 | 17 | 2040 | 2.31 | 0.23 | -0.09 | (-0.95) | -0.1 | (-1.09) | 0.03 | (0.38) |
| Winners (W) | 1 | 49 | 33 | 1112 | 2.08 | 0.19 | 0.03 | (0.22) | 0.06 | (0.55) | 0.02 | (0.19) |
|  | 10 | 52 | 21 | 1419 | 1.82 | 0.11 | $0.24 * *$ | (2.05) | 0.06 | (0.58) | -0.02 | (-0.17) |
| W-L (Decile 10) |  |  |  |  |  |  | $0.69{ }^{* * *}$ | (4.39) | $0.37^{* * *}$ | (3.1) | 0.01 | (0.04) |
| W-L (Decile 1) |  |  |  |  |  |  | $0.24 * *$ | (2.16) | $0.25 * * *$ | (2.87) | 0.1 | (1.22) |
| W-L (Decile 10) - W-L (Base) |  |  |  |  |  |  | $0.28{ }^{* * *}$ | (3.16) | 0.12 | (1.48) | 0.08 | (0.9) |
| W-L (Decile 1) - W-L (Base) |  |  |  |  |  |  | $-0.17^{*}$ | (-1.70) | -0.01 | (-0.14) | $0.17^{* *}$ | (2.23) |

## Table 10: Decimalization Effect on Trade Size and Price Momentum

This table presents average monthly and characteristic-adjusted returns from portfolios strategies formed on the basis of two-way sorts based on past trade size and price momentum for the period 1983 to 2010 . We split the sample into the pre-decimalization time period from 1983 to 2000, shown in Panel 1, and into the post-decimalization period from 2001 to 2010, shown in Panel 2. We delete any firm that is in both the extreme 100 -share and the 500 share trade portfolios to eliminate overlapping firms. We also delete firms with a price $<\$ 5.00$ in the month prior to the performance period. At the beginning of each month all available stocks listed on the NYSE/Amex/NASDAQ markets are sorted based on the past 6 months returns, and divided into 3 portfolios. The loser portfolio represents the worst prior return performance and the winner portfolio represents the best prior return performance. The stocks are then sorted based on the average trade size over the month just prior to the monthly evaluation period. We divide the trade size portfolios into 10 portfolios. We analyze performance of three separate holding periods using a six-month period (1-6), a twelve-month period (1-12), and a holding period that spans months 13 to 24 from the formation period. Monthly returns are computed based on the portfolio rebalancing strategy described in Table 1 and characteristic-adjusted returns are based on firm size, book-to-market, and momentum characteristics. We report the winner-loser portfolios for the base momentum strategy (W-L Base), then based on the extreme trade size deciles (W-L Deciles 1 or 10 ), and finally relative to the base momentum strategy (W-L Decile 1 or 10 minus the base momentum strategy). For each extreme portfolio, we report the average bid-ask spread and the price impact measure of Amihud. We also present firm size expressed in millions of dollars and the average price of the portfolio of stocks at the end of the month prior to the performance period. Newey-West robust estimators with four lags specifying the t-statistics are presented in the parentheses Significance is reported with an * ( $10 \%$ significance $)$, an ${ }^{* *}(5 \%$ significance $)$, or an ${ }^{* * *}(1 \%$ significance $)$.

| Momentum Tercile | Trade Size Decile | Number of Firms | Average Price | $\begin{gathered} \text { Firm } \\ \text { Size } \\ (\$ \text { millions }) \end{gathered}$ | Bid-Ask Spread (\%) | Amihud's measure | $\begin{gathered} \text { \% Monthly } \\ 1-6 \\ \text { Return } \end{gathered}$ | y Characteristic-Adjusted 1-12 |  |  | $\begin{gathered} \text { Returns, Pre-2001 } \\ 13-24 \end{gathered}$ |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  |  |  |  |  |  |  | T | Return | T | Return | T |
| Panel A: Base Six Month Momentum Strategy, Single-Sort by previous six-month returns |  |  |  |  |  |  |  |  |  |  |  |  |
| Losers (L) |  | 940 | 18 | 951 | 2.80 | 0.12 | $-0.37 * * *$ | (-3.39) | $-0.28^{* * *}$ | (-3.09) | 0.01 | (0.21) |
| 2 |  | 937 | 25 | 1784 | 2.34 | 0.10 | -0.03 | (-0.34) | -0.04 | (-0.50) | -0.04 | (-0.54) |
| Winners (W) |  | 942 | 26 | 1617 | 2.34 | 0.09 | $0.25{ }^{* * *}$ | (4.94) | 0.11** | (2.53) | -0.15* | (-1.87) |
| W-L (Base) |  |  |  |  |  |  | $0.62^{* * *}$ | (4.17) | $0.39^{* * *}$ | (3.63) | -0.16 | (-1.58) |
| Panel B: Double Sort by previous six-month returns, then previous one-month trade size $=100$ shares |  |  |  |  |  |  |  |  |  |  |  |  |
| Losers (L) | 1 | 55 | 13 | 431 | 3.87 | 0.23 | $-0.52^{* * *}$ | (-3.65) | $-0.40^{* * *}$ | (-2.97) | 0.06 | (0.51) |
|  | 10 | 46 | 22 | 1721 | 2.96 | 0.18 | -0.11 | (-0.88) | -0.08 | (-0.65) | 0.05 | (0.49) |
| 2 | 1 | 51 | 18 | 541 | 3.66 | 0.23 | -0.16 | (-1.02) | -0.17 | (-1.12) | 0.01 | (0.07) |
|  | 10 | 46 | 32 | 3254 | 2.34 | 0.13 | 0.04 | (0.3) | -0.03 | (-0.20) | -0.03 | (-0.21) |
| Winners (W) | 1 | 57 | 17 | 512 | 3.40 | 0.17 | 0.31** | (2.42) | 0.07 | (0.6) | -0.19 | (-1.62) |
|  | 10 | 46 | 36 | 2757 | 2.40 | 0.12 | $0.55^{* * *}$ | (3.31) | $0.37^{* * *}$ | (2.96) | -0.07 | (-0.45) |
| W-L (Decile 10) |  |  |  |  |  |  | 0.66 *** | (2.76) | 0.45** | (2.23) | -0.12 | (-0.94) |
| W-L (Decile 1) |  |  |  |  |  |  | $0.83 * * *$ | (4.47) | $0.47^{* * *}$ | (3.13) | -0.26 | (-1.57) |
| W-L (Decile 10) - W-L (Base) |  |  |  |  |  |  | 0.04 | (0.24) | $0.07$ | $(0.5)$ | $0.04$ | $(0.42)$ |
| W-L (Decile 1) - W-L (Base) |  |  |  |  |  |  | 0.20* | (1.69) | 0.08 | (0.84) | -0.09 | (-0.71) |
| Panel C: Double Sort by previous six-month returns, then previous one-month trade size $=500$ shares |  |  |  |  |  |  |  |  |  |  |  |  |
| Losers (L) | 1 | 71 | 18 | 771 | 3.87 | 0.31 | -0.08 | (-0.59) | -0.08 | (-0.55) | 0.04 | (0.34) |
|  | 10 | 80 | 13 | 122 | 3.80 | 0.26 | -0.56 *** | (-3.76) | $-0.46^{* * *}$ | (-3.36) | -0.22* | (-1.97) |
| 2 | 1 | 69 | 24 | 878 | 3.48 | 0.27 | 0.01 | (0.06) | 0 | (0.03) | -0.01 | (-0.08) |
|  | 10 | 78 | 18 | 194 | 3.38 | 0.22 | 0 | (-0.03) | -0.07 | (-0.48) | -0.23 | (-1.56) |
| Winners (W) | 1 | 72 | 28 | 1886 | 2.82 | 0.18 | 0.19 | (1.41) | 0.13 | (1.17) | 0.11 | (0.82) |
|  | 10 | 81 | 19 | 201 | 3.22 | 0.20 | $0.31^{* * *}$ | (2.91) | 0.05 | (0.52) | $-0.48^{* * *}$ | (-2.83) |
| W-L (Decile 10) |  |  |  |  |  |  | $0.87^{* * *}$ | (5.33) | 0.52*** | (3.97) | $-0.26 * *$ | (-2.03) |
| W-L (Decile 1) |  |  |  |  |  |  | $0.27 * *$ | (2.48) | $0.21 * *$ | (2.56) | 0.07 | (0.72) |
| W-L (Decile 10) - W-L (Base) |  |  |  |  |  |  | 0.25** | (2.11) | 0.13 | (1.4) | -0.1 | (-1.01) |
| W-L (Decile 1) - W-L (Base) |  |  |  |  |  |  | $-0.35^{* * *}$ | (-3.20) | $-0.18^{* *}$ | (-2.54) | 0.23 *** | (2.68) |


| Losers (L) | 1 | 58 | 21 | 639 | 3.33 | 0.29 | -0.26 | (-1.56) | -0.16 | (-1.00) | -0.01 | (-0.09) |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | 10 | 70 | 13 | 301 | 3.26 | 0.17 | $-0.54 * * *$ | (-3.34) | -0.34** | (-2.34) | 0.14 | (0.72) |
| 2 | 1 | 58 | 26 | 489 | 3.17 | 0.26 | -0.05 | (-0.26) | -0.04 | (-0.19) | -0.02 | (-0.13) |
|  | 10 | 68 | 18 | 773 | 2.89 | 0.16 | -0.02 | (-0.18) | 0.01 | (0.17) | 0.05 | (0.59) |
| Winners (W) | 1 | 57 | 29 | 960 | 2.65 | 0.18 | 0.14 | (0.86) | 0.08 | (0.61) | -0.06 | (-0.39) |
|  | 10 | 72 | 22 | 712 | 2.49 | 0.12 | 0.31* | (1.95) | 0.09 | (0.65) | 0 | (0.03) |
| W-L (Decile 10) |  |  |  |  |  |  | $0.85 * * *$ | (4.36) | $0.43 * * *$ | (3.1) | -0.14 | (-0.83) |
| W-L (Decile 1) |  |  |  |  |  |  | 0.40 *** | (3.35) | $0.24^{* * *}$ | (2.76) | -0.05 | (-0.60) |
| W-L (Decile 10) - W-L (Base) |  |  |  |  |  |  | 0.23* | (1.93) | 0.04 | (0.41) | 0.03 | (0.23) |
| W-L (Decile 1) - W-L (Base) |  |  |  |  |  |  | -0.23* | (-1.80) | -0.14* | (-1.68) | 0.11 | (1.56) |

Panel 2: Post-Decimalization Period - 2001 to 2010


## Table 11: Decimalization Effect on Turnover and Price Momentum Restricted to NYSE/Amex Firms

This table presents characteristic-adjusted returns from portfolios strategies formed on the basis of two-way sorts based on past one-month turnover and price momentum for the period 1983 to 2010, but only using NYSE and Amex listed firms. We delete any firms with a share price $<\$ 5.00$. Turnover is defined as the monthly volume of shares traded divided by the number of shares outstanding at the beginning of the month. We split the sample into the pre-decimalization time period from 1983 to 2000, shown in Panel A, and into the post-decimalization period from 2001 to 2010. At the beginning of each month all available stocks are sorted based on the past 6 months returns, and divided into 3 portfolios. The loser portfolio represents the worst prior return performance and the winner portfolio represents the best prior return performance. The stocks are then sorted based on the average trade size over the month just prior to the monthly evaluation period. We divide the trade size portfolios into 10 portfolios. We analyze performance of three separate holding periods using a six-month period (1-6), a twelve-month period (1-12), and a holding period that spans months 13 to 24 from the formation period. Monthly returns are computed based on the portfolio rebalancing strategy described in Table 1 and characteristic-adjusted returns are based on firm size, book-to-market, and momentum characteristics. We report the winner-loser portfolios for the base momentum strategy (W-L Base), then based on the extreme trade size deciles (W-L Deciles 1 or 10), and finally relative to the base momentum strategy (W-L Decile 1 or 10 minus the base momentum strategy). For each extreme portfolio, we report the average bid-ask spread and the price impact measure of Amihud. We also present firm size expressed in millions of dollars and the average price of the portfolio of stocks at the end of the month prior to the performance period. Newey-West robust estimators with four lags specifying the t-statistics are presented in the parentheses. Significance is reported with an * ( $10 \%$ significance), an ** ( $5 \%$ significance), or an ${ }^{* * *}$ ( $1 \%$ significance).

| Momentum Tercile | Turnover <br> Decile | Number of Firms | Average Price | FirmSize$(\$$ millions $)$ | Bid-Ask <br> Spread (\%) | Amihud's measure | $\begin{aligned} & \text { \% Monthly } \\ & 1-6 \end{aligned}$ |  | Characteristic-Adjusted Returns 1-12 |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  |  |  |  |  |  | Return | T | Return | T | Return | T |
| Panel A: Base Six Month Momentum Strategy |  |  |  |  |  |  |  |  |  |  |  |  |
| Losers (L) |  | 511 | 27 | 1566 | 1.94 | 0.05 | -0.32*** | (-2.75) | ) $-0.26^{* *}$ | (-2.44) | -0.07 | (-0.76) |
| 2 |  | 571 | 44 | 2815 | 1.52 | 0.03 | -0.06 | (-0.71) | ) -0.07 | (-0.84) | -0.08 | (-0.81) |
| Winners (W) |  | 484 | 45 | 2811 | 1.60 | 0.03 | 0.07 | (0.77) | 0.02 | (0.23) | -0.13 | (-1.46) |
| W-L (Base) |  |  |  |  |  |  | $0.39^{* * *}$ | (3.19) | $0.28^{* * *}$ | (3.10) | -0.06 | (-0.64) |
| Panel B: Sort by Previous Six-Month Returns and then One-Month Turnover |  |  |  |  |  |  |  |  |  |  |  |  |
| Losers (L) | 1 | 51 | 75 | 324 | 2.73 | 0.20 | -0.36* | (-1.85) | ) -0.26 | (-1.34) | -0.13 | (-0.81) |
|  | 10 | 51 | 22 | 1320 | 1.61 | 0.01 | -0.88*** | (-4.17) | ) $-0.62 * * *$ | (-3.52) | -0.11 | (-0.62) |
| 2 | 1 | 57 | 157 | 526 | 2.22 | 0.14 | -0.02 | (-0.10) | ) -0.04 | (-0.21) | -0.1 | (-0.55) |
|  | 10 | 57 | 32 | 2254 | 1.30 | 0.00 | -0.30*** | (-2.62) | ) $-0.31 * * *$ | (-3.07) | -0.25** | (-2.45) |
| Winners (W) | 1 | 48 | 166 | 570 | 2.29 | 0.14 | 0.15 | (0.82) | ) 0.04 | (0.23) | -0.09 | (-0.54) |
|  | 10 | 48 | 33 | 1769 | 1.44 | 0.01 | -0.1 | (-0.56) | ) $-0.33^{* *}$ | (-2.10) | $-0.39^{* * *}$ | (-2.66) |
| W-L (Decile 10) |  |  |  |  |  |  | $0.78^{* * *}$ | (3.20) | 0.29 | (1.64) | -0.29* | (-1.77) |
| W-L (Decile 1) |  |  |  |  |  |  | 0.51 *** | (3.67) | $0.29^{* * *}$ | (2.60) | 0.04 | (0.37) |
| W-L (Decile $10-$ Base) |  |  |  |  |  |  | 0.40** | (2.49) | 0.02 | (0.14) | $-0.24 * *$ | (-2.22) |
| W-L (Decile 1 - Base) |  |  |  |  |  |  | 0.13 | (0.99) | 0.02 | (0.23) | 0.08 | (0.90) |

Panel 2: Post-Decimalization Period - 2001 to 2010


Panel B: Sort by Previous Six-Month Returns and then One-Month Turnover

| Losers (L) | 1 | 42 | 567 | 1059 | 1.41 | 0.14 | -0.09 | (-0.32) | 0.02 | (0.07) | 0.23 | (1.01) |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | 10 | 42 | 27 | 3344 | 0.26 | 0.00 | -0.31 | (-0.66) | -0.17 | (-0.40) | 0.23 | (0.50) |
| 2 | 1 | 49 | 1187 | 1844 | 1.26 | 0.12 | -0.11 | (-0.44) | -0.06 | (-0.26) | -0.01 | (-0.05) |
|  | 10 | 49 | 37 | 4338 | 0.19 | 0.00 | -0.03 | (-0.14) | 0.01 | (0.05) | 0.2 | (0.80) |
| Winners (W) | , | 42 | $542$ | 1212 | 1.34 | 0.14 | 0.03 | (0.10) | 0.08 | (0.35) | 0.01 | (0.05) |
|  | 10 | 42 | $40$ |  | 0.25 | 0.00 | 0.22 | (0.77) | 0.15 | (0.58) | 0.27 | (0.95) |
| W-L (Decile 10) |  |  |  |  |  |  | 0.53 | (0.88) | 0.31 | (0.69) | 0.04 | (0.13) |
| W-L (Decile 1) |  |  |  |  |  |  | 0.11 | (0.41) | 0.06 | (0.28) | -0.22 | (-1.27) |
| W-L (Decile $10-$ Base) |  |  |  |  |  |  | 0.34 | (0.90) | 0.22 | (0.68) | 0.1 | (0.53) |
| W-L (Decile 1 - Base) |  |  |  |  |  |  | -0.08 | (-0.26) | -0.03 | (-0.13) | -0.16 | (-0.98) |


[^0]:    ${ }^{1}$ Momentum returns are higher for stocks that are small and have low analyst coverage (Hong, Lim, and Stein, 2000), high analyst forecast dispersion (Zhang, 2006; Verardo, 2009), low return (Hou, Xiong, and Peng, 2006), and high market-to-book ratios (Daniel and Titman, 1999). Lee and Swaminathan (2000) find an interaction between momentum and share turnover and suggest that turnover provides information on the extent to which investor sentiment favors a stock at a particular point in time. Since these characteristics are commonly used to proxy for information uncertainty and limits to arbitrage, these findings are often interpreted as evidence in support of behavioral explanations of momentum.

[^1]:    ${ }^{2}$ Since large orders are likely to involve informed institutions, given the analysis of Hasbrouck (1995) and Chakravarty (2001), the subsequent medium-sized rounded trades are more likely to be information-based.
    ${ }^{3}$ Kaniel, Saar, and Titman (2008) employ a similar procedure in obtaining a net individual investor trading by subtracting the value of the shares sold by individuals from the value of shares bought and standardized by the average daily dollar volume yielding a Net Individual Trading measure.
    ${ }^{4}$ This is extensively studied by Holden and Jacobsen (2014) who find that severe distortions in the quote can occur when using the more common monthly quote file on the Trade and Quote (hereafter, TAQ) database. They argue that using the daily TAQ can lead to very different empirical outcomes from the research using the monthly TAQ. However, they also note the difficulty in using the daily TAQ file whereby millions of quotes are noted on a daily basis making the identification of the NBBO difficult and its association to the price file impractical in large

[^2]:    time-series datasets.
    ${ }^{5}$ It should be noted that Battalio and Mendenhall (2005) only study NASDAQ firms from 1993 through 1996 and Hvidkjaer $(2006,2008)$ restricts his sample to the period before the decimalization of stock quotes in 2001.
    ${ }^{6}$ There is long-standing empirical evidence of systematic trading behavior among various investor groups. For instance, small and large investors respond differently to exogenous information events such as earnings releases (Lee, 1992), seasoned equity offerings (Huh and Subrahmanyam, 2005), and analyst recommendations (Malmendier and Shanthikumar, 2014).

[^3]:    ${ }^{7}$ We also analyze trade size increments between 100 and 500 shares, and between 500 and 1000 shares, between 1000 and 5000, and 5000-share trades. Our results point to only two trade size categories, namely 500 -share and 1000-share trade sizes, that are associated with significant improvements in momentum profits.

[^4]:    ${ }^{8}$ The NYSE Fact book reports statistics showing average trade sizes falling dramatically after stock decimalization. The average trade size in 1999 for NYSE-listed firms was 1,205 shares per trade. After decimalization in 2004, the average trade size was significantly reduced to just over 390 shares per trade. In 2010, the average trade size had dwindled to 220 shares per trade and in 2014 the average trade size was approximately 140 shares per trade.

[^5]:    ${ }^{9}$ Russ Wermer's website: http://www.smith.umd.edu/faculty/rwermers/ftpsite/Dgtw/coverpage.htm contains the characteristic-adjusted returns.

[^6]:    ${ }^{10}$ By observation, it is apparent that the base monthly momentum returns are approximately $0.75 \%$.

