The Passive Ownership Share Is Double What You Think It Is*

Alex Chinco† and Marco Sammon‡

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Abstract

Each time a stock gets added to or dropped from an index, we ask: “How much money would have to be tracking that index to explain the huge spike in rebalancing volume we observe on reconstitution day?” While index funds held 16% of the US stock market in 2021, we put the overall passive ownership share at 33.5%. Our headline number is twice as large because it reflects index funds as well as other kinds of passive investors, such as institutional investors with internally managed index portfolios and active managers who are closet indexing.

Keywords: Passive Ownership, Index-Linked Investing, Prearranged Trades, Index Funds, Internal Indexing, Reconstitution Day

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†Baruch College, Zicklin School of Business; alexchinco@gmail.com.
‡Harvard Business School; mcsammon@gmail.com.
Introduction

Index funds manage a lot more money today than they did twenty years ago (Investment Company Institute, 2022, ICI). Index funds had combined assets under management (AUM) of just $0.4t in 2000, roughly 3% of the value of the US stock market. By 2021, these funds had $7.2t in AUM, which represented 16% of the market’s value. We know this because US equity index mutual funds and exchange-traded funds (ETFs) have to regularly disclose their holdings. So it is relatively straightforward to calculate the index-fund ownership share.

While this is a phenomenal amount of growth, index funds are not the only kind of passive investor. Many institutional investors manage index-tracking portfolios on their own behalf, a practice known as “internal indexing”.\(^1\) Investors can also invest through separately managed index-like accounts. This is called “direct indexing”, and it is often done for tax reasons.\(^2\) Finally, active managers are evaluated relative to an index and sometimes engage in “closet indexing”.\(^3\)

These other kinds of passive investors do not face the same disclosure requirements. So in this paper we propose an alternative way to estimate the total amount of money held by all kinds of passive investors, not just the amount of money held by index funds. We find that passive investors tracking five popular indexes collectively owned 33.5% of the US stock market in 2021.

Our 33.5% estimate for the US passive ownership share is more than double the ICI’s index-fund ownership share of 16%. It implies that, for every $1 held by an index fund, there is another $1 held by another kind of passive investor. The result holds even though our estimate reflects only a subset of indexes; whereas, the ICI’s numbers include all index funds no matter what index they track.

The logic behind our approach is simple. Most passive rebalancing occurs in a huge spike right at market close on reconstitution days. So each time that a stock gets added to or dropped from an index we ask: “How much money would

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\(^1\)For example, Wigglesworth (2021, p17) writes that “many big pension plans and sovereign wealth funds have huge internal index-tracking strategies.”

\(^2\)Separately managed accounts are used by institutional investors and are also available to individuals with as little as $100k (see https://www.wealthfront.com/tax-loss-harvesting).

\(^3\)For example, see Cremers and Petajisto (2009) and Petajisto (2013).
Figure 1. Volume for Yeti Holdings (YETI) around Russell reconstitution day 2021 in millions of shares. Solid bars represent total volume each day. White bars represent volume from 4:00pm to 11:59pm. On June 4th (green), FTSE Russell announced that Yeti would join the Russell 1000 following market close on June 25th (red). June 18th (blue) was a triple witching day on the 3rd Friday in June. Grey region is average daily volume from June 4th to 24th.

have to be tracking that index to explain the enormous burst in closing volume on reconstitution day that we observe in the data?”

For example, Yeti Holdings (YETI) migrated from the Russell 2000 up to the Russell 1000 at the end of trading on Friday June 25th 2021. The company entered the Russell 1000 with a weight of 0.019%. To give investors time to prepare for the change, FTSE Russell made an official announcement three weeks in advance on Friday June 4th. But Yeti was so far above the Russell 1000’s size cutoff that it was possible to predict its migration months before.4

In an effort to minimize tracking error, a Russell 1000 ETF would likely build its 0.019% position in Yeti right at market close on reconstitution day. However, other kinds of passive investors were free to rebalance a little early. An active manager with a portion of her AUM closet indexed to the Russell 1000 could have bought Yeti shares at any point during the week prior to reconstitution day if a good opportunity arose. A university endowment that was internally indexing would have the same sort of freedom. So you might have expected Yeti’s volume to gradually increase in the days prior to June 25th 2021.

4A back-of-the-envelope calculation suggests that, as of the end of April, Yeti would have had to lose $200m in market cap by rank day to not migrate up to the Russell 1000.
Figure 2. Black line depicts the percent of the US stock market owned by passive investors tracking the S&P 500, S&P MidCap 400, Russell 1000, Russell 2000, and Nasdaq 100. Calculation is based on total volume experienced by index additions and deletions on reconstitution day. Black ribbon shows that percent of the US stock market owned by domestic index equity mutual funds and ETFs according to the Investment Company Institute (ICI). Sample: 2000 to 2021.

That is not what happens in Figure 1. After remaining flat in the lead up, Yeti’s volume suddenly spikes to 11.0m shares on reconstitution day itself, with 9.2m trading in the closing auction or immediately after. Given that we see no additional volume in the days immediately before and after reconstitution, active managers who were closet indexing likely did their rebalancing at the close on June 25th just like Russell 1000 ETFs did. The same goes for university endowments that were internal indexing.

Suppose Yeti’s entire spike in volume on June 25th came from Russell 1000 rebalancing. Yeti’s closing price was $92.07 per share on June 25th 2021. So, under this assumption, passive investors tracking the Russell 1000 would have spent 0.019% of their wealth purchasing $11.0m \times $92.07 \approx $1.0b in Yeti shares

\[
\text{value of shares purchased} = \frac{11.0m \times \$92.07}{\text{ReconDayVolume \times Price}} \quad \text{value of required position} = \frac{\text{AUMindexed} \times \text{IndexWeight}}{0.019\%}
\]  

Hence, this group of investors must have had 11.0m \times (\frac{$92.07}{0.019\%}) = $5.3t in AUM.

We estimate the US passive ownership share by performing this same exercise for every stock added to or dropped from five indexes: the Russell 1000,
the Russell 2000, the S&P 500, the S&P MidCap 400, and the Nasdaq 100. Each addition and deletion produces a separate estimate for the total AUM tracking an index. Each year we average the estimates for the AUM indexed to a given index. The black line in Figure 2 reports the sum of these annual averages for our five indexes as a percent of total US stock-market capitalization.

As a point of comparison, Figure 2 also reports the ICI’s index-fund ownership share. We use these numbers as a benchmark because these are the numbers that academic researchers typically point to as evidence for the rise of passive ownership. The “you” in the title refers researchers and anyone else who “forgets that open-ended investment funds only hold a slice of markets, and conflate passive’s mutual fund industry market share with its overall market ownership.”

The ICI’s 16% is not really the right benchmark for market participants. Industry reports suggest the number is higher. A 2017 BlackRock report put the passive-ownership share at 25.6% (Novick et al., 2017). Recent research from Bloomberg Intelligence says passive investors own at least 19% of the market (Seyffart, 2023). And even if they did not read these reports, institutional investors who were internally indexing in 2021 must have known that the overall passive ownership share was higher than 16%. The same goes for direct indexers and active managers engaged in closet indexing.

However, prior to our paper, it was difficult to know how much additional money was tracking an index outside of the index-fund universe. Market participants could only make an educated guess based on public holdings data. As illustrated by the examples above, these guesses tended to be lower than our headline 33.5%. Previous estimates based on holdings data were also relatively imprecise, yielding a broad range of values. By contrast, under our new approach, each individual stock that gets added to or deleted from an index yields a separate point estimate for the amount of money tracking that index. This allows us to calculate standard errors and cross-validate our results.

The ICI’s 16% is also not a lower bound for our calculation. The ICI includes all index funds tracking any index. We only have index weights for five indexes. Using fund holdings, we estimate that if all passive investing was index-fund

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investing, we would have estimated a value of around 6% not 16% (see appendix A). In 2021 Vanguard alone had roughly $2.25t in equity holdings tracking CRSP indexes, which we lack data on. This $2.25t represents 5% of ICI’s baseline 16%. If we were to include this $2.25t, our 33.5% would rise to 38.5%.

It is important to be clear about who we are calling passive. There is just the single spike in volume at market close on reconstitution day. This suggests that all passive investors rebalance in the same way. Our procedure counts any holdings that get rebalanced at this one point in time as passive holdings; it does not matter who actually owns these assets. As such, our numbers likely includes passive investments made by people who would not consider themselves to be passive investors. For example, suppose an active manager had $250b in total AUM, and 20% of this money was closet indexed to the Russell 1000. If the manager bought $50b \times \left( \frac{0.019\%}{82.07} \right) = 103k$ Yeti shares at market close on June 25th 2021, then we would count his $50b as AUM indexed to the Russell 1000.

Other kinds of active investors could be strategically delaying trades to take advantage of all the additional volume on reconstitution days. But there is no dip in volume for index additions and deletions in the days prior to reconstitution. Moreover, by definition, each reconstitution event must involve at least two stocks. And the variation in reconstitution-day volume across changes to the same index on the same day is almost entirely explained by passive demand.

For example, Sunrun Inc (RUN) also migrated up to the Russell 1000 on June 25th 2021. But, compared to Yeti, Sunrun entered the index with a higher weight, 0.027%, and a lower price, $54.38. So if Sunrun had realized the same reconstitution-day volume as Yeti, 11.0m, its addition would have implied that there was just $11.0m \times \left( \frac{54.38}{0.027\%} \right) = 2.2t$ in AUM tracking the Russell 1000. To produce the same estimate as Yeti, $5.3t$, Sunrun needed to have much more reconstitution-day volume, $5.3t \times \left( \frac{0.027\%}{54.38} \right) = 26.4m$. This is exactly what the data show! Sunrun’s realized a volume 26.8m on June 25th 2021.

This pattern holds more generally. On average, when looking across changes to the same index on the same day, our estimates are within ±1% of each other. This is true even though each calculation involves a different price and index weight. Additions and deletions do not just have higher reconstitution-day
volumes. Each stock's volume increases by the precise amount needed by passive rebalancing. There is very little left for active managers to explain.

In principle, brokers could be inflating reconstitution-day volumes in precisely the right proportions. For example, suppose that Russell 1000 investors bought 5.5m shares of Yeti and 13.2m shares of Sunrun from a broker on the afternoon of June 25th 2021. If their broker purchased these shares from Russell 2000 investors in the morning on the same day, each stock's volume on June 25th would be double the amount needed by Russell 1000 investors.

In practice, there is no time to buy from one passive investor and then sell to another. Figure 1 shows 9.2m of 11.0m Yeti shares were traded in the closing auction or immediately after. 7.7m of these were executed in trades specifically tied to the closing auction. Trading happens all at once. Brokers cannot be intermediating these trades by getting between buyers and sellers on the same day. Our estimates based on narrower definitions of rebalancing volume around the close are still well above the index-fund ownership share.

Trading is so concentrated on reconstitution days because passive investors preschedule their rebalancing trades. They contract with an intermediary to rebalance at the closing price on reconstitution day, whatever that price happens to be. Until recently, liquidity providers were earning large profits for committing to rebalancing trades weeks or months in advance. At peak, Goldman’s index rebalancing desk reportedly “[generated] more revenue per employee than almost any other” at the company.

This is one reason why the prices of index additions and deletions no longer move much on reconstitution days. “For the New York Stock Exchange, Russell reconstitution…is the greatest show on earth.” Moreover, passive investors execute much of this volume right at market close. But, because of the way that passive investors rebalance, their price impact occurs before their demand shock,

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6Alex Morrell “It can’t get much worse than this’: A trail of portfolio managers have resigned as the once lush index-rebalance strategy dries up.” Business Insider. May 11, 2023.
at the point in time where they prearrange their rebalancing trades.

Tesla’s addition to the S&P 500 on December 18th 2020 provides a helpful case study showing what happens when this trading apparatus breaks down. S&P Dow Jones’ initial announcement on November 17th caught many people by surprise. There was also a lot of uncertainty about “which current constituent Tesla [would] replace [or] how Tesla [would] be added. (S&P Dow Jones Indices, 2020b)” As a result, passive investors could not prearrange much of their rebalancing, and Tesla saw its price jump on reconstitution day.

Theorists often model passive investors as uninformed traders in Grossman and Stiglitz (1980). This paradigm assumes that the passive ownership share is common knowledge and that these investors choose their demand after observing the price. Neither of these assumptions is correct. The passive ownership share is not common knowledge among market participants, and many passive investors go out of their way to preschedule rebalancing trades before seeing the closing price on reconstitution day.

Our results also speak to the demand-system asset-pricing literature (Koijen and Yogo, 2019). Naïve estimates seem to imply that markets are extremely elastic on reconstitution days because there is not much same-day price pressure. But nothing could be further from the truth. Prices do not move much on reconstitution days because passive investors’ rebalancing trades are typically scheduled weeks in advance. The correct elasticity calculation needs to reflect the price change that occurs when these trades are arranged. Given that passive investors’ demand is inelastic, one might expect larger effects of flows into and out of passive investments on prices (Coval and Stafford, 2007; Lou, 2012). As an example, our results suggest that the growth of ESG investing may lead to asset price changes that are larger than in the past.

Related Work

This paper builds on several strands of literature connected to index-linked investing (Wurgler, 2011). Index inclusions used to generate predictable price pressure on reconstitution day (Harris and Gurel, 1986; Shleifer, 1986; Beneish and Whaley, 1996; Wurgler and Zhuravskaya, 2002; Madhavan, 2003; Petajisto,
These events also affect correlations and liquidity (Barberis, Shleifer, and Wurgler, 2005; Greenwood, 2008; Baker, Bradley, and Wurgler, 2011; Chang, Hong, and Liskovich, 2015; Burnham, Gakidis, and Wurgler, 2018; Brogaard, Ringgenberg, and Sovich, 2019).

ETFs have experienced explosive growth in recent years (Madhavan, 2016; Lettau and Madhavan, 2018), leading to higher closing volumes, more volatility, and lower information production for the stocks they hold (Israeli, Lee, and Sridharan, 2017; Ben-David, Franzoni, and Moussawi, 2018; Da and Shive, 2018; Chinco and Fos, 2021; Bogousslavsky and Muravyev, 2023). ETFs are some of the most actively traded assets (Robertson, 2019; Huang, O’Hara, and Zhong, 2021). We document that other passive investors trade like ETFs. This is separate from how other investors trade ETF shares.

Active investors sometimes park a fraction of their holdings in passive-investment vehicles (Cremers and Petajisto, 2009; Cremers, Ferreira, Matos, and Starks, 2016; Pavlova and Sikorskaya, 2023). Unlike an ETF, active managers have no obligation to rebalance right at market close on reconstitution day. Nevertheless, we find that they rebalance just like an ETF would. This is evidence supporting Gabaix and Koijen (2022)’s Inelastic Markets Hypothesis.

Our findings run contrary to order-execution models which predict that investors will smooth their demand to limit price impact (Kyle, 1985; Bertsimas and Lo, 1998; Almgren and Chriss, 2001). Unlike in a sunshine-trading model (Admati and Pfleiderer, 1991), markets are less liquid for other kinds of investors during normal trading hours on reconstitution days. A hot-potato model (Lyons, 1997) cannot explain our results since most rebalancing trades get executed all at once during the close. Bessembinder, Carrion, Tuttle, and Venkataraman (2016) finds different results when studying commodity indexes.

Many theory papers use Grossman and Stiglitz (1980) to model the rise of passive investing (e.g., see Baruch and Zhang, 2021; Bond and García, 2022; Buss and Sundaresan, 2023; Buffa, Vayanos, and Woolley, 2022; Lee, 2021; Schmalz and Zame, 2023). We argue that this is the wrong framework because it assumes that the passive share is common knowledge and that investors choose their demand after observing prices. Coles, Heath, and Ringgenberg (2022) fixes half
the problem by assuming inelastic demand in a Grossman and Stiglitz setup.

Finally, our analysis connects to the literature looking at how passive investors affect firm decisions (Appel, Gormley, and Keim, 2016; Bebchuk, Cohen, and Hirst, 2017; Edmans and Holderness, 2017; Azar, Schmalz, and Tecu, 2018; Heath, Macciocchi, Michaely, and Ringgenberg, 2022; Lewellen and Lewellen, 2022). While this literature focuses on index-fund ownership, we point out that there are other ways to passively invest.

1 Data Description

This section describes the data we use in our analysis. Subsection 1.1 details the five indexes in our study. Subsection 1.2 discusses variable construction. And subsection 1.3 provides summary statistics.

1.1 Five Indexes

We estimate the combined AUM of passive investors tracking five popular indexes: the S&P 500, the S&P MidCap 400, the Russell 1000, the Russell 2000, and the Nasdaq 100.

**S&P 500 and MidCap 400.** The S&P 500 is a float-adjusted value-weighted index that, loosely speaking, tracks the 500 largest public US companies. The index is maintained by S&P Dow Jones (S&P Dow Jones Indices, 2022). The S&P MidCap 400 is an analogous index tracking the next largest 400 US companies. A committee decides who gets added to and dropped from each index, and this committee makes its decision based on more than just firm size. For example, a firm must have positive earnings the quarter before being added.

S&P Dow Jones regularly reconstitutes the S&P 500 and MidCap 400 on a quarterly basis. These scheduled events take place on the third Friday of March, June, September, and December which represent triple-witching days when stock options, index options, and index futures all expire at the same time. However, the index provider also makes ad hoc changes at other times during the quarter due to corporate events like bankruptcies or mergers. We find similar point estimates for the total AUM indexed to the S&P 500 and MidCap 400 when using
For the S&P 500, we have quarterly index membership and changes directly from S&P Dow Jones. For the S&P MidCap 400, we have quarterly index membership and changes from Siblis Research. We use these data to interpolate daily membership and weights in each index. For the S&P 500, we include a float-adjustment factor directly from the index provider. Our weights for the S&P MidCap 400 are based on market capitalization in CRSP and do not include a float-adjustment factor.

We treat migrations between the S&P 500 and MidCap 400 as signals about AUM indexed to the S&P 500. By contrast, when estimating the AUM indexed to the S&P MidCap 400, we only include direct additions to and deletions from the index. Figure 3 shows that we have data on 38 changes to the S&P 500 in 2021 (both adds and drops); whereas, there were 75 stocks directly added to or dropped from the S&P MidCap 400 in our data set.

**Russell 1000 and 2000.** The Russell 1000 and 2000 are float-adjusted value-weighted indexes, which are provided by FTSE Russell. The Russell 1000 tracks the 1000 largest stocks in the Russell 3000E universe, and the Russell 2000

The entire Russell family of US indexes reconstitutes on the last Friday in June each year. FTSE Russell ranks stocks by market capitalization in late May. The index provider then formally announces changes to the Russell 1000 and 2000 roughly two weeks prior to reconstitution day. That being said, it is usually possible to predict which stocks will move long before this announcement.

Russell reconstitution day occurs on the fourth Friday in June each year. For years 2000 to 2008, we get end-of-month index membership from FTSE Russell. We use this end-of-month data to interpolate daily index membership and weights. These weights are based on market capitalization in CRSP and do not include a float-adjustment factor. Starting in 2009, we have daily data on index membership and weights directly from FTSE Russell.

Figure 3 shows that in 2021 FTSE Russell added 55 and 274 stocks to the Russell 1000 and 2000 respectively. While the index provider rarely makes ad hoc changes prior to reconstitution day, some passive investors must divest in response to certain corporate events, such as a bankruptcy. They cannot wait until reconstitution day to do their rebalancing. For this reason, we do not use stocks that exit the Russell 3000E universe in our estimation procedure.

**Nasdaq 100.** The Nasdaq 100 is a modified value-weighted index provided by the Nasdaq. The index tracks securities issued by the 100 largest non-financial stocks that are exclusively listed on the Nasdaq exchange. Although it has been around since the mid-1980s, the Nasdaq 100's popularity has grown along with the rise of Invesco’s QQQ ETF. Like with the S&P 500 and MidCap 400, there is a selection committee that decides membership in the Nasdaq 100. Since 2014 the committee has included companies with multiple share classes in the index.

Nasdaq regularly reconstitutes the Nasdaq 100 on the third Friday in December. Figure 3 shows that in 2021 there were 17 changes to the Nasdaq 100. This annual rebalancing event lines up with the final witching day of the calendar year. However, like with the S&P 500 and MidCap 400, there are also ad hoc changes to the Nasdaq 100 at other times during the year. For example, Honeywell International (HON) replaced Alexion Pharmaceuticals (ALXN) on
July 20th 2021. We find similar point estimates for the AUM indexed to the Nasdaq 100 when using both regularly scheduled and ad hoc changes.

We get quarterly data on Nasdaq 100 index membership and changes from Siblis Research. Use these data to interpolate daily index membership and weights. Our weights for the Nasdaq 100 are based on market capitalization in CRSP and do not include a float-adjustment factor.

**Other Indexes.** In an ideal world, we would be able to include data on other popular indexes in our study, too. Our 33.5% estimate does not reflect the passive AUM tracking the MSCI World or the CRSP Total Market. For example, using data from Thomson S12, we estimate that in December 2021 Vanguard had $2.25 trillion in equity holdings tracking CRSP indexes, which we lack data on. If we were to include this $2.25 trillion, our 33.5% would climb to 38.5%.

To estimate the AUM tracking a particular index, we need accurate data on each constituent’s weight in the index. For the Russell 1000 and Russell 2000, we purchased daily weights directly from FTSE Russell for $7.5 million. We are able to interpolate the daily weights for the S&P 500, S&P MidCap 400, and Nasdaq 100 from known quarterly values. For the S&P 500, these quarterly values come directly from S&P Dow Jones.

We have approached other index providers about purchasing similar data. When we talked to MSCI, they quoted us a price of $240 million for the daily data from 2000 to 2021. When we asked CRSP for the same thing, they refused to provide data on daily weights at any price. This pattern of events is consistent with the observation that index providers charge sizable licensing fees (An, Benetton, and Song, 2023). We are not arguing these fees are excessive. We are pointing out that it is expensive to acquire information about indexes in a timely fashion. The passive investors who pay these fees are not uninformed traders.

### 1.2 Variable Construction

Each time a stock gets added to or dropped from a index, we compute the dollar value of the spike in volume it experiences on reconstitution day. Then, under the assumption that this spike represents passive rebalancing, we back out the total AUM of passive investors given the stock’s weight.
### Table 1a. Characteristics of changes to S&P 500, S&P MidCap 400, and Russell 1000.

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<th>S&amp;P 500</th>
<th></th>
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<th>S&amp;P MidCap 400</th>
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<th>Russell 1000</th>
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Let $IndexWeight_{i,n}(t_{Recon})$ denote the $n$th stock's weight in index $i$ on reconstitution day $t_{Recon}$. For additions, $IndexWeight_{i,n}(t_{Recon})$ represents the stock's initial weight in index $i$ when markets open on the following trading day ($t_{Recon} + 1$). For deletions, $IndexWeight_{i,n}(t_{Recon})$ represents the $n$th stock's final weight in index $i$ at market close on $t_{Recon}$.

We use several variables to capture the spike in volume experienced by index additions and deletions on reconstitution day. $DailyVolume_{n}(t)$ denotes the $n$th stock's volume on day $t$ as reported in CRSP. This daily data covers our entire sample period from 2000 through 2021. When comparing reconstitution-day
### Stock characteristics, continued

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<th>Drops</th>
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<tr>
<td>$MCap \ [$1b$]</td>
<td>12.0</td>
<td>18.6</td>
<td>15.8</td>
</tr>
<tr>
<td>$IndexWeight \ [bps]$</td>
<td>29.8</td>
<td>39.3</td>
<td>38.3</td>
</tr>
<tr>
<td>$ADV \ [1m]$</td>
<td>6.4</td>
<td>23.6</td>
<td>5.6</td>
</tr>
<tr>
<td>$PastRet \ [%]$</td>
<td>11.3</td>
<td>48.9</td>
<td>30.3</td>
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</tbody>
</table>

**Table 1b.** Characteristics of changes to Russell 2000 and Nasdaq 100. $MCap$: Market cap on reconstitution day in billions of dollars. $IndexWeight$: Weight in index in basis points. $ADV$: Average volume during the 22 trading days prior to reconstitution in millions of shares per day. $PastRet$: Return during the 6 months prior to reconstitution in percent. 2000 to 2021.

volume across stocks, we normalize by average daily volume during the previous 22 trading days, $ADV_n = \frac{1}{22} \cdot \sum_{\ell=1}^{22} DailyVolume_n(t_{Recon} - \ell)$.

Since much of the spike in reconstitution-day volume is tied to the closing price, we also use TAQ’s millisecond-level daily-update consolidated trade database. This data starts on September 10th 2003, which is after Russell reconstitution day that calendar year. So we only use it from 2004 through 2021. We remove observations flagged with “M” and “Q” sale conditions, which represent duplicate observations produced by Nasdaq’s trade-reporting protocol (Tuttle, 2013). We also remove corrected trades. The remaining TAQ volume each day matches daily volume reported in CRSP.

Let $VolumeAtClose_n(t)$ denote the $n$th stock’s volume at the closing auction on day $t$. As we discuss in subsection 3.2, passive investors often use prescheduled trades to rebalance. These trades get executed at the price determined by the closing auction at 4:00pm on reconstitution day. So they typically hit the tape
some time after hours. For this reason, our preferred intraday measure of passive rebalancing volume is \( \text{Volume}_{1600\text{to}2359}_n(t) \), which represents the \( n \)th stock's volume from 4:00pm through 11:59pm on reconstitution day.

\( \text{VolumeAtClosingPrice}_n(t) \) denotes the \( n \)th stock's volume executed at the closing price on day \( t \) as indicated by trade condition “6” in TAQ. Earlier in our sample, prescheduled trades sometimes included price improvement, meaning that \( \text{VolumeAtClosingPrice} \) is likely too conservative. For example, FTSE Russell added Maxim Integrated Products (MXIM) to the Russell 1000 on June 26th 2009. A Russell 1000 investor might have prearranged on May 8th to buy 10k Maxim shares at $0.01 below the closing price on June 26th 2009. These 10k shares would not be captured by \( \text{VolumeAtClosingPrice}_{\text{MXIM}}(\text{June 26th 2009}) \).

We explore a variety of proxies for passive rebalancing volume in subsection 2.3. None of these other measures is perfect. However, by looking at a wide range of proxies, we are able to get a better sense of the true scale of passive ownership as well as how much uncertainty there is about this level.

### 1.3 Summary Statistics

Tables 1a and 1b describe the characteristics of stocks that got added to or dropped from each of our five indexes. As expected, index additions are different from index deletions. For example, index additions tend to be larger and have higher returns over the past 6 months.

We are exploiting the difference between an index switcher’s reconstitution-day volume and its own prior volume. We are not comparing index switchers to stocks that just missed getting added or dropped. We are not using the Russell 1000 cutoff for identification (Chang, Hong, and Liskovich, 2015; Appel, Gormley, and Keim, 2020). We know S&P Dow Jones strategically chooses which companies to add (Beneish and Whaley, 1996; Bennett, Stulz, and Wang, 2022).

Tables 2a and 2b then describe the reconstitution-day volume experienced by these index additions and deletions. We normalize each stock's volume measures by the stock's average daily volume during the previous 22 trading days, \( \text{ADV}_n \). For example, the top panel of Table 2a indicates that, on average, changes to the S&P 500 see 12.3 days worth of volume on reconstitution day. We also report
### Reconstitution-day volume

<table>
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<th>S&amp;P MidCap 400</th>
<th>Russell 1000</th>
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<td>All Drops</td>
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<td>Avg</td>
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<td>8.3</td>
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<td>VolumeAtClosingPrice</td>
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<td>VolumeAtClosingPrice</td>
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<table>
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<td>3.3</td>
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<td>VolumeAtClose</td>
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</table>


Summary statistics for closing volume on reconstitution day, volume from 4:00pm to 11:59pm, and volume at the closing price.

## 2 Passive Ownership

This section reports our headline numbers for the US passive ownership share. Subsection 2.1 describes our estimation strategy. Subsection 2.2 reports estimates based on daily volume. Subsection 2.3 gives a range of alternative estimates based on other proxies for rebalancing volume. Subsection 2.4 assesses

potential sources of measurement error. Subsection 2.5 compares our findings with public data on fund holdings. Finally, subsection 2.6 applies a Kalman filter to the passive ownership share.

2.1 Estimation Strategy

Here is the core intuition behind our approach. Suppose stock ADD replaced stock DROP in index \( i \) at market close on day \( t_{Recon} \). Let \( AUM_{indexed,i}(t_{Recon}) \) denote the total AUM held by passive investors tracking this index on reconstitution day. Further suppose that ADD initially represented \( IndexWeight_{i,ADD} \) of the index. If passive investors perfectly matched this portfolio weight, then they had to build new positions worth

\[
IndexWeight_{i,ADD} \times AUM_{indexed,i}(t_{Recon}) \tag{2}
\]

Now imagine that passive investors are the only people trading ADD on
reconstitution day and that these passive investors do all their trading at market close. In this scenario, \( DailyVolume_{\text{ADD}}(t_{\text{Recon}}) \) as reported in CRSP would capture all passive rebalancing volume. And these trades would be worth

\[
DailyVolume_{\text{ADD}}(t_{\text{Recon}}) \times Price_{\text{ADD}}(t_{\text{Recon}})
\]

where \( Price_{\text{ADD}}(t) \) denotes ADD’s closing price per share on day \( t \).

We impute the total AUM tracking index \( i \) by equating (2) and (3) and solving for \( AUM_{\text{Indexed}}(t_{\text{Recon}}) \)

\[
\tilde{AUM}_{\text{Indexed},\text{ADD}}(t_{\text{Recon}}) \leftarrow \frac{DailyVolume_{\text{ADD}}(t_{\text{Recon}}) \times Price_{\text{ADD}}(t_{\text{Recon}})}{IndexWeight_{i,\text{ADD}}}
\]

The tilde indicates that \( \tilde{AUM}_{\text{Indexed},\text{ADD}}(t_{\text{Recon}}) \) is an implied value, and the ADD subscript indicates that this implied value is based on a single addition.

Figure 4 shows the average implied \( \tilde{AUM}_{\text{Indexed},i,n} \) across all stocks added to
Passive ownership share for each index

Figure 5. Solid line is percent of the US stock market owned by passive investors when using \textit{DailyVolume} as proxy for passive rebalancing; 2000 to 2021. Black ribbon denotes percent owned by index funds according to the Investment Company Institute. Dotted line is percent owned by passive investors when using \textit{Volume1600to2359} as proxy for passive rebalancing; 2004 to 2021.

or dropped from a given index $i$ in year $y$:

$$
\overline{\text{AUMindex}ed_i(y)} = \text{Avg} \left( \overline{\text{AUMindex}ed_{i,n}} \mid \text{stock } n \text{ was added to or dropped from index } i \text{ in year } y \right)
$$

(5)

The solid lines represent estimates for passive AUM where we proxy for passive rebalancing volume with \textit{DailyVolume} as described in Equation (4). The dotted lines perform the same calculation with \textit{Volume1600to2359}. We estimate that there was $15.1t$ in passive AUM during 2021 when treating all reconstitution-day volume as rebalancing volume. When using only the volume from 4:00pm to 11:59pm, we get a value of $11.0t$. 
2.2 Headline Numbers

To compute the passive ownership share associated with index $i$ in year $y$, we divide $\text{AUM}_{\text{Indexed}}(i, y)$ by US stock-market capitalization

$$\%\overline{\text{Indexed}}_i(y) = 100 \times \frac{\text{AUM}_{\text{Indexed}}(i, y)}{\text{TotalMarketCap}(y)}$$

(6)

Figure 5 reports these estimates for each index. The solid lines represent calculations using $\text{DailyVolume}$. The dotted lines use $\text{Volume1600to2359}$.

The bottom-right panel reports the sum across all five index we study

$$\%\overline{\text{Indexed}}(y) = \%\overline{\text{Indexed}}_{\text{S&P 500}}(y) + \%\overline{\text{Indexed}}_{\text{S&P MidCap}}(y)$$
$$+ \%\overline{\text{Indexed}}_{\text{Russell 1000}}(y) + \%\overline{\text{Indexed}}_{\text{Russell 2000}}(y)$$
$$+ \%\overline{\text{Indexed}}_{\text{Nasdaq 100}}(y)$$

(7)

The solid line corresponds to the headline numbers reported in Figure 2 from the introduction. When using $\text{DailyVolume}$ as the proxy for passive rebalancing on reconstitution days, we find that 33.5% of the US stock market was held by passive investors in 2021. When using $\text{Volume1600to2359}$ instead of $\text{DailyVolume}$, we still put the US passive ownership share at 24.6% in 2021.

The bottom-right panel in Figure 5 also reports the share of the US stock market owned by index funds. This percentage comes from annual reports made by the Investment Company Institute (ICI; e.g., see Figure 2.9 in Investment Company Institute, 2022). The ICI’s numbers reflect the combined holdings of all domestic equity index mutual funds and ETFs. These numbers include the AUM of index funds that track indexes other than the five in our study, such as Vanguard funds tracking the CRSP Total Market index. In December 2021 Vanguard had roughly $2.25t tracking CRSP indexes. If we were to remove this $2.25t from ICI’s asset base, their 16% would fall to 11%. If we were to include this $2.25t, our 33.5% would climb to 38.5%.

There is a 2009 spike in the overall passive ownership share in the bottom-right panel of Figure 5. However, notice that there is no corresponding spike in
the index-fund ownership share. This is consistent with investors focusing on internal indexing or direct indexing, rather than index-fund investing. One possible explanation is that investors were engaging in tax-loss harvesting following the financial crisis. It could also be that passive investors held a larger share of the market following the financial crisis because active managers reduced their leverage after suffering losses.

Table 3 report the specific numerical values underpinning Figure 5. Since DailyVolume comes from CRSP, we can use this proxy to impute \( \%\text{Indexed}_i(y) \) for each of our five indexes all the way back to 2000 in Table 3. We report standard errors clustered by announcement to account for the fact that stocks can be added to and dropped from the S&P 500, S&P MidCap 400, and Nasdaq 100 at different times throughout the year. All additions to the Russell 1000 and Russell 2000 occur simultaneously on the last Friday in June each year. For these two indexes, clustering has no effect on our standard errors.

2.3 Range of Estimates

Our headline numbers indicate that, given on all the volume experienced by index additions and deletions on reconstitution days, passive investors likely held 33.5% of the US stock market in 2021. Even if we look only at the volume from 4:00pm onward on reconstitution days, we still get a passive ownership share of 24.6%. This is well above the percentage owned by index funds as reported by the ICI, 16%. Moreover, the ICI's numbers include index funds that track indexes not included in our sample, like the CRSP Total Market index.

DailyVolume and Volume1600to2359 are not perfect proxies for passive rebalancing volume on reconstitution days. On one hand, both measures ignore passive rebalancing done prior to reconstitution day. Some passive investors may choose to rebalance months ahead of time. And, to the extent that this happens, it will cause us to underestimate the true passive ownership share.

On the other hand, DailyVolume and Volume1600to2359 could be capturing reconstitution-day volume coming from active investors. This would lead us to overestimate the true passive ownership share. The data are not consistent with active investors delaying trades to take advantage of the high liquidity.
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<th>S&amp;P 500</th>
<th>S&amp;P MidCap</th>
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<th>Russell 2000</th>
<th>Nasdaq 100</th>
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Table 3. Percent of the US stock market owned by passive investors when using DailyVolume to proxy for passive rebalancing volume. Numbers in parentheses are standard errors clustered by announcement. 2000 to 2021.
experienced by index additions and deletions at market close on reconstitution day (see subsection 3.1). Admati and Pfleiderer (1991)’s sunshine-trading story would not apply in a world where passive investors prescheduled all their rebalancing trades months in advance via intermediaries.

In Table 4 we report a range of estimates for the US passive ownership share based on different proxies for passive rebalancing on reconstitution days. That way, readers can judge for themselves how much extra volume our headline numbers might be capturing. Column (1) corresponds to the total reported in Table 3 based on DailyVolume. Column (2) reports the results of the same calculation but using DailyVolume – ADV. This column looks at an index addition/deletion’s volume on reconstitution day in excess of its volume on a typical trading day. It is unlikely that non-Russell investors trade the same way on Russell reconstitution day as they would on any other day during the previous month. Russell reconstitution day is now “generally considered [to be] the single-biggest trading day in US markets.” However, if other investors kept on trading index additions and deletions in the exact same way on reconstitution days, then column (2) suggests that our headline numbers might be overestimating the true passive ownership share by somewhere between 2%pt and 5%pt.

Column (3) in Table 4 shows that, on a typical day, most trading activity occurs during normal market hours. This was true for Yeti Holdings (YETI) in Figure 1 from the introduction, and in subsection 3.1 we show this pattern holds more generally. However, reconstitution days are different. On reconstitution days, the bulk of trading occurs after hours. Thus, our passive ownership share based on Volume1600to2359 is only slightly below our estimate using DailyVolume.

One important reason for this pattern is that passive investors often preschedule rebalancing trades to get executed at the closing price on reconstitution day. So we report the %Indexed(y) implied by VolumeAtClosingPrice in column (4) in Table 4. These estimates look very similar to the ones in column (3) based on Volume1600to2359, which is consistent with the idea that much of the volume from 4:00pm-11:59pm comes from prescheduled trades. There is a larger gap

---

9Rolf Agather, managing director of North America research at FTSE Russell, quoted by Victor Reklaitis in “Why Friday could be the year’s biggest trading day.” MarketWatch. Jun 26, 2015.
<table>
<thead>
<tr>
<th>Year</th>
<th>All day</th>
<th>Minus ADV</th>
<th>4:00pm-11:59pm</th>
<th>Closing price</th>
<th>Closing auction</th>
</tr>
</thead>
<tbody>
<tr>
<td>2004</td>
<td>14.26 (1.02)</td>
<td>11.84 (1.02)</td>
<td>9.35 (0.67)</td>
<td>5.47 (0.48)</td>
<td>1.09 (0.21)</td>
</tr>
<tr>
<td>2005</td>
<td>13.52 (1.19)</td>
<td>11.36 (1.05)</td>
<td>7.18 (0.69)</td>
<td>4.40 (0.46)</td>
<td>0.97 (0.22)</td>
</tr>
<tr>
<td>2006</td>
<td>16.92 (1.50)</td>
<td>14.12 (1.39)</td>
<td>10.30 (0.77)</td>
<td>6.63 (0.45)</td>
<td>1.80 (0.34)</td>
</tr>
<tr>
<td>2007</td>
<td>16.33 (1.28)</td>
<td>13.37 (1.27)</td>
<td>10.41 (1.20)</td>
<td>7.42 (1.02)</td>
<td>2.51 (0.61)</td>
</tr>
<tr>
<td>2008</td>
<td>25.31 (2.07)</td>
<td>20.42 (1.72)</td>
<td>14.20 (0.94)</td>
<td>10.83 (2.74)</td>
<td>2.85 (0.53)</td>
</tr>
<tr>
<td>2009</td>
<td>31.85 (2.14)</td>
<td>27.64 (1.96)</td>
<td>18.15 (1.50)</td>
<td>9.91 (0.86)</td>
<td>4.92 (0.86)</td>
</tr>
<tr>
<td>2010</td>
<td>24.32 (2.02)</td>
<td>20.61 (1.62)</td>
<td>16.12 (1.91)</td>
<td>9.62 (0.99)</td>
<td>4.58 (0.88)</td>
</tr>
<tr>
<td>2011</td>
<td>25.90 (1.62)</td>
<td>21.86 (1.44)</td>
<td>17.21 (1.47)</td>
<td>10.78 (0.69)</td>
<td>3.81 (0.76)</td>
</tr>
<tr>
<td>2012</td>
<td>25.54 (1.65)</td>
<td>22.33 (1.67)</td>
<td>17.76 (1.64)</td>
<td>9.87 (0.84)</td>
<td>4.17 (0.67)</td>
</tr>
<tr>
<td>2013</td>
<td>23.21 (1.90)</td>
<td>19.51 (1.77)</td>
<td>15.82 (1.65)</td>
<td>10.31 (0.89)</td>
<td>3.54 (0.77)</td>
</tr>
<tr>
<td>2014</td>
<td>26.76 (2.18)</td>
<td>23.08 (2.00)</td>
<td>19.09 (1.74)</td>
<td>12.02 (0.88)</td>
<td>4.24 (1.13)</td>
</tr>
<tr>
<td>2015</td>
<td>22.06 (1.52)</td>
<td>19.92 (1.31)</td>
<td>15.47 (1.30)</td>
<td>11.86 (1.04)</td>
<td>5.52 (0.96)</td>
</tr>
<tr>
<td>2016</td>
<td>25.92 (1.39)</td>
<td>22.83 (1.35)</td>
<td>19.04 (1.26)</td>
<td>15.42 (0.99)</td>
<td>6.45 (1.12)</td>
</tr>
<tr>
<td>2017</td>
<td>28.11 (1.94)</td>
<td>24.67 (1.68)</td>
<td>21.09 (1.64)</td>
<td>17.46 (1.35)</td>
<td>9.99 (1.35)</td>
</tr>
<tr>
<td>2018</td>
<td>29.43 (1.59)</td>
<td>25.98 (1.40)</td>
<td>22.56 (1.26)</td>
<td>19.06 (1.03)</td>
<td>11.38 (1.26)</td>
</tr>
<tr>
<td>2019</td>
<td>31.53 (3.49)</td>
<td>25.12 (3.08)</td>
<td>23.29 (1.98)</td>
<td>21.71 (3.16)</td>
<td>10.79 (2.06)</td>
</tr>
<tr>
<td>2020</td>
<td>32.19 (3.26)</td>
<td>26.24 (2.71)</td>
<td>21.37 (1.77)</td>
<td>19.35 (1.37)</td>
<td>11.54 (1.37)</td>
</tr>
<tr>
<td>2021</td>
<td>33.50 (2.91)</td>
<td>28.65 (2.91)</td>
<td>24.62 (2.70)</td>
<td>22.53 (2.10)</td>
<td>13.31 (1.96)</td>
</tr>
</tbody>
</table>

between columns (3) and (4) earlier in our sample. As previously noted, it used to be common for prearranged trades to include some price improvement.

Finally, column (5) in Table 4 gives our most conservative estimate for the US passive ownership share based only on VolumeAtClose. These numbers help address concerns about possible double counting of active trades that get covered after hours. However, if a large institutional investor places an upstairs block order for an index addition to be executed at the closing price, this order will not hit the tape until after market close. So there is good reason to think that the estimates for \(\%\text{Indexed}(y)\) in column (5) is too low. Even still, there is no statistically measurable difference between our point estimate for \(\%\text{Indexed}(2021)\) based on VolumeAtClose, 13.31\% \pm (1.96\%), and ICI’s estimate based on index-fund holdings, 16\%.

This observation underscores the importance of passive investors who operate “outside the public universe of index funds and ETFs.” We know that closing volume omits some rebalancing trades and that our data only includes a subset of indexes. Nevertheless, the numbers in column (5) are on par with the ICI’s estimates for index-fund ownership.

2.4 Measurement Error

Our headline numbers suggest that the true passive ownership share is roughly double the index-fund ownership share. Given the economic magnitude of this difference, it is important to thoroughly examine potential sources of error in our calculations. We do this in several ways.

First, in an ideal world, when ADD replaces DROP in index \(i\) on day \(t_{Recon}\), both of these changes would yield the same value for \(AUM_{Indexed_i}(t_{Recon})\)

\[
\overline{AUM_{Indexed_i,ADD}(t_{Recon})} = \overline{AUM_{Indexed_i,DROP}(t_{Recon})}
\] (8)

When ADD replaces DROP in index \(i\), it will generally do so with a different weight and closing price. So these two stocks cannot simply have higher volumes on reconstitution days. They much each realize higher volumes in exactly the right proportion needed by passive investors. Every stock added to or
### Average absolute measurement error

<table>
<thead>
<tr>
<th></th>
<th>S&amp;P 500</th>
<th>S&amp;P MidCap</th>
<th>Russell 1000</th>
<th>Russell 2000</th>
<th>Nasdaq 100</th>
<th>Full Sample</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>%Error</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>%Error</td>
<td>5.49</td>
<td>0.35</td>
<td>1.87</td>
<td>0.23</td>
<td>0.74</td>
<td>1.00</td>
</tr>
</tbody>
</table>

**Table 5.** Average absolute difference between the passive ownership share implied by each index change and the average passive ownership share implied by all changes to that index in the same year. Sample: 2000 to 2021.

dropped from index $i$ on day $t_{Recon}$ should yield the same point estimate for $\text{AUM Indexed}_i(t_{Recon})$. We use this requirement to gauge the magnitude of our measurement errors.

Let $\%Error_{i,n}(t_{Recon})$ denote the difference between the passive share implied by a single addition or deletion and the average passive share implied by all additions and deletions in the same year

$$\%Error_{i,n}(t_{Recon}) = \overline{\text{Indexed}_{i,n}(t_{Recon})} - \overline{\text{Indexed}_i(y)}$$  \hspace{1cm} (9)

For example, Yeti’s volume on June 25th 2021 implied that 11.72% of the US stock-market was owned by Russell 1000 investors. The reconstitution-day volume for the average Russell 1000 addition in 2021 implied a passive ownership share of 10.77%. Thus, the measurement error associated with Yeti’s addition to the Russell 1000 on June 25th 2021 was $11.72\% - 10.77\% = 0.95\% pt$.

Table 5 reports the average magnitude of the measurement error for each index. The typical Russell 1000 addition yields an estimate for $\overline{\text{Indexed}_{i,n}(t_{Recon})}$ that is $\pm 1.87\% pt$ of the average for the year. The precision of estimate based on Yeti’s addition to the Russell 1000 was representative of all Russell 1000 additions.

Table 5 says that the typical S&P 500 addition/deletion yields an estimate that is $\pm 5.49\% pt$ away from the annual average, which would imply a standard error of $\sqrt{\# \text{ changes each year}}$. Figure 3 tells us that our data contains 48 changes to the S&P 500 each year. By comparing our original standard errors to the ones implied by this within-event analysis, we find that roughly $4/5$ of our uncertainty is coming from within-event differences. For example, to convert the numbers in Table 5 into standard errors, you need to divide by a factor of $\sqrt{\# \text{ changes each year}}$. For example, Yeti’s volume on June 25th 2021 implied that 11.72% of the US stock-market was owned by Russell 1000 investors. The reconstitution-day volume for the average Russell 1000 addition in 2021 implied a passive ownership share of 10.77%. Thus, the measurement error associated with Yeti’s addition to the Russell 1000 on June 25th 2021 was $11.72\% - 10.77\% = 0.95\% pt$.

Table 5 reports the average magnitude of the measurement error for each index. The typical Russell 1000 addition yields an estimate for $\overline{\text{Indexed}_{i,n}(t_{Recon})}$ that is $\pm 1.87\% pt$ of the average for the year. The precision of estimate based on Yeti’s addition to the Russell 1000 was representative of all Russell 1000 additions.

Table 5 says that the typical S&P 500 addition/deletion yields an estimate that is $\pm 5.49\% pt$ away from the annual average, which would imply a standard error of
\[
\frac{5.49\%}{\sqrt{48}} = 0.80\% \text{pt. This value is roughly } \frac{4}{5} \text{ of the average standard error on the S&P 500’s ownership share in Table 3, 1.06\% pt.}
\]

Next, we explore how our measurement errors are related to characteristics of the stocks being added or dropped. We do this by running regressions

\[
\%\text{Error}_{i,n}(t_{Recon}) = \hat{\alpha} + \hat{\beta} \cdot X_{i,n}(t_{Recon}) + \hat{\epsilon}_{i,n}(t_{Recon})
\]

(10)

where \(X_{i,n}(t_{Recon})\) represents one of the following variables: a stock’s market cap the day prior to reconstitution, its reconstitution-day return, its dollar volume on reconstitution day, its weight in the index, an indicator for whether the stock is an addition, an indicator for whether the stock was migrated.

Table 6 reports the results of these regressions. The negative coefficient on \(MCap\) in column (1) says that we tend to slightly underestimate the passive ownership share when analyzing the reconstitution-day volume of larger stocks. A $1b increase in an index addition/deletion’s market cap is associated with a 2bps underestimate. The zero coefficient on \(Ret\) in column (2) implies that, when an index addition or deletion has a large reconstitution-day return, this does not cause us to over- or underestimate the passive ownership share.

Column (3) in Table 6 suggests that, when a stock has more dollar volume on reconstitution day, we tend to overestimate the passive ownership share. Whereas, column (4) implies that, when a stock represents a larger share of the index, we tend to underestimate the true passive ownership share. It is noteworthy that, while we have spent most of our time so far worrying about how reconstitution-day volume might be overstating passive rebalancing volume, there is only a significant coefficient in column (4) on the index weights. We will return to this point shortly.

Finally, columns (5) and (6) in Table 6 show that our estimates for the passive ownership share are 67bps and 65bps higher for additions and migrations relative to deletions. These two point estimates are statistically significant but economically small. 67bps is 20× smaller than the US passive ownership share in 2000, 13.4%. These effects likely stem from how our sample is constructed. Direct deletions often follow corporate events like bankruptcies or mergers, meaning
## Over- and underestimates in the passive ownership share

<table>
<thead>
<tr>
<th>Dep variable: %Error</th>
<th>(1)</th>
<th>(2)</th>
<th>(3)</th>
<th>(4)</th>
<th>(5)</th>
<th>(6)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Intercept</td>
<td>0.07**</td>
<td>0.02</td>
<td>-0.06</td>
<td>0.08***</td>
<td>-0.58***</td>
<td>-0.16***</td>
</tr>
<tr>
<td></td>
<td>(2.09)</td>
<td>(0.89)</td>
<td>(1.26)</td>
<td>(2.85)</td>
<td>(4.20)</td>
<td>(3.95)</td>
</tr>
<tr>
<td>MCap</td>
<td>-0.02**</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>(2.14)</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Ret</td>
<td>0.00</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>(0.08)</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>$Volume</td>
<td></td>
<td>0.19</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td></td>
<td>(1.43)</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>IndexWeight</td>
<td></td>
<td></td>
<td>-0.01***</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td>(4.46)</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>IsAddition</td>
<td></td>
<td></td>
<td></td>
<td>0.67***</td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>(4.73)</td>
<td></td>
<td></td>
</tr>
<tr>
<td>IsMigration</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>0.65***</td>
<td></td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>(5.11)</td>
<td></td>
</tr>
<tr>
<td># Obs</td>
<td>11168</td>
<td>11087</td>
<td>11168</td>
<td>11168</td>
<td>11168</td>
<td>11168</td>
</tr>
<tr>
<td>Adj. $R^2$</td>
<td>0.66%</td>
<td>-0.01%</td>
<td>1.62%</td>
<td>0.28%</td>
<td>0.74%</td>
<td>1.16%</td>
</tr>
</tbody>
</table>

**Table 6.** Each column reports the results of a separate univariate regression. The dependent variable is always the difference between the passive ownership share implied by each index change and the average passive ownership share implied by all changes to that index in the same year, %Error. The right-hand-side variable is different in each column. MCap is the market cap of the stock being added or dropped on the day before reconstitution in billions of dollars. Ret is the realized return of the stock being added or dropped on reconstitution day in percent. $Volume$ is the dollar volume of the stock being added or dropped on reconstitution day in billions of dollars. IndexWeight is the weight of the stock being added or dropped in basis points. IsAddition is an indicator variable that is one if a stock is being added to the index and zero otherwise. IsMigration is an indicator variable that is one if a stock is being moved between indexes and zero otherwise. Numbers in parentheses are $t$-stats clustered by announcement. Sample: 2000 to 2021.
How outliers affect headline numbers

Figure 6. Grey line represents average passive ownership share tracking each index when using DailyVolume as proxy for passive rebalancing volume. White dots are estimates for passive ownership share implied by specific add/drop events, which we classify as outliers. Black dots represent the average passive ownership share implied by all remaining changes to the same index on the same date. Black line is the total passive ownership share across all five indexes excluding outliers. Sample: 2000 to 2021.

that some passive investors cannot wait until reconstitution day to rebalance.

At the end of the day, most of our measurement error seems to be coming from uncertainty about the precise weights used by the indexes. We get IndexWeight_{i,n}(t_{Recon}) directly from FTSE Russell starting in 2009. We have to interpolate these values from quarterly or monthly observations for the rest of our sample—i.e., for the S&P 500, the S&P MidCap 400, the Nasdaq 100, and the Russell 1000/2000 prior to 2009. Our measurement error all but disappears when we have precise index weights directly from the index provider.

The gray lines in Figure 6 show the passive ownership share tracking each index. These lines correspond to the solid lines reported in Figure 5. The white dots in each panel correspond to individual estimates for %Indexed_{i,n}(t_{Recon}) that are either 4× larger or smaller than the average for the year. There are almost no outliers for the Russell 1000 and Russell 2000 starting in 2009 when we
have index weights directly from FTSE Russell.

Whenever we have to interpolate index weights, we use an extremely conservative approach. Most of the outlier white dots are below the annual estimates for each index’s ownership share. If we were to omit these outliers from our sample, our headline numbers for the US passive ownership share would go up as shown by the black line in the bottom-right panel.

2.5 Quarterly Fund Holdings

In the past, researchers have used quarterly fund holdings data to study passive ownership. The usual source for this data is Thomson S12. This data exists because the Investment Company Act of 1940 requires mutual funds, closed-end funds, exchange-traded funds (ETFs), and unit investment trusts (UITs) to disclose snapshots of their holdings every quarter. In this section, we discuss the advantages of using reconstitution-day volume rather than quarterly fund holdings to estimate the US passive ownership share.

First, our approach reflects passive investments made outside of 40 Act funds. In the past, market participants have made educated guesses about the overall passive ownership share based on holdings data. However, as discussed above, these guesses tended to be well below our headline 33.5% number. Moreover, when market participants made different guesses from one another, it was difficult to reconcile their estimates. By contrast, under our new approach, each individual stock that gets added to or deleted from an index yields a separate point estimate for the amount of money tracking that index. This allows us to calculate standard errors and cross-validate our results.

It is possible to use quarterly fund holdings to infer which mutual funds are likely benchmarked to a particular index. However, “being benchmarked to” is not the same thing as “mimicking the precise portfolio weights at market close on reconstitution day.” We are the first to document the fact that most rebalancing volume gets executed in a short window of time around market close on reconstitution day. Our data suggest that all passive investors rebalance like an ETF would.

The amount of money benchmarked to our five indexes is also far below the
total amount of money being indexed. For example, Pavlova and Sikorskaya (2023) estimates that the collective AUM of mutual-fund managers who were benchmarked to the S&P 500 was $2.5t in 2018. We find a similar estimate in Figure 7 when we compute the combined AUM of mutual funds that bought shares of S&P 500 additions and sold shares of deletions in more than 50% of reconstitution events during the past 5 years. These estimates imply that the AUM of S&P 500-benchmarked mutual funds is $2.0t less than our $4.5t estimate for the AUM passively tracking the S&P 500 in Figure 4. If we use an 80% threshold for benchmarking, the gap grows to $3.2t. And, once again, not all mutual funds that are benchmarked to the S&P 500 are passive.

While we cannot see the holdings of non-index-fund passive investors, our approach sheds light on the economic forces behind their decisions. While we see a large spike in 2009 in the overall passive ownership share in Figure 2, there is no corresponding spike in the index-fund ownership share in Figure 2 or benchmarking intensity in Figure 7.

This suggests that, although the index-fund ownership share is slow moving, the ownership share of other kinds of passive investors is not. One possible explanation is that large institutional investors turned to direct indexing in 2009.
as a way to do tax-loss harvesting following the financial crisis. Another possible explanation is that active managers substantially reduced their leverage after suffering losses during the financial crisis. Both interpretations are consistent with Cremers and Petajisto (2009)’s results on the active share among US mutual funds. Their updated data reveals a 2009 drop in the active share as documented in Mauboussin, Callahan, and Majd (2017).

2.6 State-Space Framework

The share of the market owned by index funds has grown steadily over the past twenty years. By contrast, our estimates indicate that the market share of other kinds of passive investors increased sharply in 2009. Suppose you think that this spike is implausible. Under the assumption that index-fund ownership is a constant fraction of overall passive ownership, it is possible to use a Kalman filter to strip out noise from our estimates.

We study the data-generating process below

\[
m_i(y) = F x_i(y) + \epsilon_i(y) \quad \text{(11a)}
\]

\[
x_i(y) = G x_i(y - 1) + \epsilon_i(y) \quad \text{(11b)}
\]

\[
m_i(y) = [\%\text{IndexFundOwned}_i(y), \%\text{Indexed}_i(y), 1]^\top \quad \text{is the measurement vector, which contains the observable index-fund ownership share and our estimate for the passive ownership share.}
\]

\[
x_i(y) = [\%\text{Indexed}_i(y), \%\text{Indexed}_i(y - 1), 1]^\top \quad \text{is the hidden state vector containing the “true” passive ownership share.}
\]

We parameterize the coefficient matrices and noise terms based on the results of time-series and cross-sectional regressions, which we report in appendix A. The coefficient matrix \(F\) and the noise term \(\epsilon_i(y)\) are given by

\[
F = \begin{pmatrix} 
\lambda & 0 & \kappa \\
1 & 0 & 0 \\
0 & 0 & 1 
\end{pmatrix}
\quad \text{and} \quad
\epsilon_i(y) \sim \text{Normal}
\begin{pmatrix} 
0 \\
\zeta^2 & 0 & 0 \\
0 \\
0 & 0 & \sigma^2 \\
0 \\
0 & 0 & 0 
\end{pmatrix}
\]

(12)

The 1 in the second row of \(F\) captures the idea that our estimation procedure
yields a noisy measurement of the true passive ownership share, \( \%\text{Indexed}_i(y) = \%\text{Indexed}_i(y) + \epsilon_i(y) \), with \( 1/\sigma^2 \) denoting the precision of our estimates.

The coefficient matrix \( G \) and the noise term \( \epsilon_i(y) \) are given by

\[
G = \begin{pmatrix}
1 + \theta & -\theta & \mu \\
1 & 0 & 0 \\
0 & 0 & 1
\end{pmatrix}
\quad \text{and} \quad
\epsilon_i(y) \sim \text{Normal}\left( \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} \sigma^2 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix} \right)
\]

This structure yields a state-transition equation with a single lag in first differences, 
\( \Delta\%\text{Indexed}_i(y) = \mu + \theta \cdot \Delta\%\text{Indexed}_i(y - 1) + \epsilon_i(y) \).

We use maximum likelihood to fit the 7 parameters, \( \{\mu, \theta, \sigma, \kappa, \lambda, \zeta, \sigma_t\} \), to the annual data for each of our five indexes. Then we make one-step-ahead forecasts of the true passive ownership share, \( \%\text{Indexed}_i(y) \). These forecasts reflect our baseline estimates for the passive ownership share, \( \%\text{Indexed}_i(y) \), as well as the assumption that a constant fraction of all passive holdings are owned by index funds. Because the index-fund ownership share grows at a relatively stable rate, these filtered forecasts will be less volatile than our baseline estimates.

The gray dotted lines in Figure 8 show the one-step-ahead forecasts produced by a Kalman filter. The black lines correspond to our baseline estimates as shown in Figure 5. The filtered values are much smoother than our baseline estimates, displaying no spike in ownership following the 2008 financial crises. But, again, this is expected. The index-fund ownership share increases at a steady rate, and the Kalman filter adjusts our baseline estimates to reflect an assumption that index funds own a constant fraction of all passive investments.

Both time series start with the same initial value in 2001, but they do not have to end in the same place. Indeed, the filtered time series for the Nasdaq 100 and the Russell 2000 ends slightly below our baseline estimates of the passive ownership share for those indexes. But this is not the overarching pattern. Our estimates for the passive ownership share of Russell 1000 and S&P 500 investors are slightly above their filtered counterparts in 2021. The same is true for our 2021 estimate of total passive ownership. This suggests that the main takeaway
Figure 8. Dashed gray lines represent one-step-ahead forecasts produced by a Kalman filter. These filtered values represent our estimates for the passive ownership share after adjusting for the information in the current index-fund ownership share. Solid black lines depict our baseline estimates for the passive ownership share using DailyVolume as proxy for passive rebalancing. They are identical to the black lines reported in Figure 5. Sample: 2001 to 2021.

from our analysis is not driven by the spike in passive ownership in 2009.

3 Trading Volume

The previous section used the spike in reconstitution-day volume experienced by index additions and deletions to impute the US passive ownership share. In this section, we provide more information about the spike itself. Subsection 3.1 describes how trading volume jumps up on reconstitution day after being nearly flat in the days immediately prior. Then, in subsection 3.2, we describe the ecosystem that has emerged to allow passive investors to rebalance all at once following the closing bell on reconstitution day. Finally, in subsection 3.3, we look at an example of what happens when this ecosystem collapses. We include an analysis of the cost to rebalancing all at once on reconstitution day in appendix B.
Daily volume around reconstitution events

<table>
<thead>
<tr>
<th>Nasdaq 100</th>
<th>Russell 1000</th>
<th>Russell 2000</th>
</tr>
</thead>
<tbody>
<tr>
<td>×ADV</td>
<td></td>
<td></td>
</tr>
<tr>
<td>4.7</td>
<td>5.4</td>
<td>14.2</td>
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<td>1.0</td>
<td>1.3</td>
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<td>S&amp;P MidCap</td>
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<tr>
<td>1.0</td>
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</tr>
</tbody>
</table>

Figure 9. Average daily volume for index additions and deletions on days \( t \in \{ t_{\text{Recon}} - 22, \ldots, t_{\text{Recon}} + 7 \} \). We normalize volume on day \( t \) by \( \text{ADV} \) during the 22 trading days before reconstitution. All panels have same scale. Red bars and numbers denote reconstitution day. Blue bars denote the triple-witching day on the Friday before Russell reconstitution. Sample: 2000 to 2021.

3.1 Reconstitution Day

Figure 9 shows the average \( \text{DailyVolume}_n(t) \) for index additions and deletions on each day covering a 30-day window around reconstitution. To make volume numbers comparable, we normalize the values for each stock by \( \text{ADV}_n \). The typical index addition/deletion sees \( 9.5 \times \) its normal volume on reconstitution day. There is no index for which reconstitution days look ordinary. And some indexes have truly outstanding levels of reconstitution-day volume. For example, additions to the Russell 2000 see 14.2 days worth of volume on reconstitution day.

Figure 10 depicts how the magnitude of the reconstitution-day spike in volume has evolved over time for each index in our study. There is not an obvious common pattern across all five indexes. Additions to and deletions from both the S&P 500 and the S&P MidCap 400 have seen more and more reconstitution-day volume over time. Whereas, we find a qualitatively different pattern for changes to the Russell 1000 and Russell 2000.
Figure 10. Black bars denote the average volume on reconstitution day experienced by index additions and deletions in a given year. We normalize each stock’s reconstitution-day volume by its ADV during the 22 trading days prior to reconstitution. The highest and lowest y-axis labels in black represent the maximum and minimum annual values. The middle y-axis label in red represents the time-series average over entire sample. Sample: 2000 to 2021.

Figure 11 shows the fraction of $\text{DailyVolume}_{n}(t)$ for index additions and deletions that gets executed from 4:00pm-11:59pm on each day in the 30-day window around reconstitution. If $\frac{\text{Volume}_{1600\text{to}2359}\_{n}(t)}{\text{DailyVolume}_{n}(t)} = 100\%$, then there was not a single share of the $n$th stock traded during normal trading hours on day $t$.

In the 22 days prior to reconstitution, 7.1% of daily volume got traded from 4:00pm to 11:59pm for a typical index addition/deletion across all five indexes. On reconstitution day, 65.0% of all volume for adds and drops got executed either during the closing auction or after hours. For the typical Russell 2000 addition, this number is as high as 77.5%.

Since Russell reconstitution day falls on the fourth Friday in June each year, there is always a triple witching day on the Friday before. Stock options, index options, and index futures all expire at market close on the third Friday in June. We would expect to observe higher fraction of volume to occur at the close on these days, and the blue bars in Figure 11 confirm that this is indeed the case.
Figure 11. Percent of daily volume executed from 4:00pm-11:59pm for index additions and deletions in days $t \in \{t_{Recon} - 22, \ldots, t_{Recon} + 7\}$. All panels have same scale. Black y-axis labels represent the percent executed from 4:00pm-11:59pm on a typical day prior to reconstitution. Red bars and numbers are connected to reconstitution day. Blue corresponds to the Friday before Russell reconstitution, which falls on a triple witching day. Sample: 2004 to 2021.

The triple witching day on the Friday before Russell reconstitution day only yields $\frac{Volume_{1600to2359,n(t-5)}}{DailyVolume_{n(t-5)}} = 14.9\%$ for Russell 2000 additions.

It is not optimal for passive investors to wait until market close on reconstitution day to do all their rebalancing in textbook order-execution models such as Kyle (1985), Bertsimas and Lo (1998), and Almgren and Chriss (2001). And if markets really were this liquid on reconstitution days, then sunshine-trading models like Admati and Pfleiderer (1991) would predict that other traders should also want to get in on the action. Yet we see no evidence of investors delaying trades to take advantage of reconstitution-day liquidity. Neither class of model seems to describe what happens in our data. Moreover, because most reconstitution-day volume happens all at once, it cannot be driven by hot-potato trading between intermediaries like in Lyons (1997).
3.2 Prearranged Trades

ETFs aim to minimize their tracking error, so they wait until market close on reconstitution day to rebalance. However, other kinds of passive investors do not face the same end-of-day timing pressure. An active manager who is closet indexing could gradually rebalance if they wanted to. The same goes for a large institutional investor who was engaged in internal indexing or had an external direct-indexing account.

Nevertheless, in spite of the fact that they do not face the same timing constraints as ETFs, we provide evidence that all passive investors rebalance like ETFs. There is only a tiny increase in volume in the days immediately prior to reconstitution. Then, there is a huge spike in volume on reconstitution day, and most of these trades get executed from 4:00pm onward. To be able to trade this way, passive investors get help from an entire ecosystem of other investors.

Market participants begin preparing for reconstitution events months ahead of time. For example, Russell reconstitution day occurs on the last Friday in June each year. And Madhavan, Ribando, and Udevbulu (2022) suggests that from March to May: “Rebalance facilitators... use publicly available market information to predict anticipated changes to the index to estimate the size of the upcoming index rebalance. And liquidity providers, such as hedge funds, use the index predictions to establish trade positions in anticipation of supplying liquidity on the rebalance effective date.”

Anecdotally, we have heard from market participants that rebalance facilitators and liquidity providers begin preparing for Russell reconstitution day in January. And our data are consistent with these stories. Figure 12 shows the average daily volume for index additions and deletions in the months (rather than days like in Figure 9) around reconstitution. The height of each bar represents the percent difference between the typical addition/deletion's daily volume in month $m$ and its daily volume 11 months prior to reconstitution, $100 \times \left( \frac{\text{DailyVolume}_{m}(m)}{\text{DailyVolume}_{m}(m_{\text{Recon}-11})} - 1 \right)$. We see volumes begin to rise 6 months prior to Russell reconstitution.

While some passive investors do gradually rebalance during the months prior
Volume in months around reconstitution

Nasdaq 100  
Rushell 1000  
Russell 2000  
S&P 500  
S&P MidCap  
Full Sample

Figure 12. Average daily volume for index additions and deletions in months $m \in \{m_{\text{Recon}} - 10, \ldots, m_{\text{Recon}} + 3\}$. y-axis reports the percent difference between a stock’s average daily volume in month $m$ and its average daily volume 11 months prior to reconstitution, $100 \times \left( \frac{\text{DailyVolume}_{n}(m)}{\text{DailyVolume}_{n}(m_{\text{Recon}}-11)} - 1 \right)$. All panels have same scale. Red y-axis label denotes the month of reconstitution. Black y-axis label denotes the month prior to reconstitution. Sample: 2001 to 2021.

to reconstitution, most of the extra volume in Figure 12 comes from rebalancing facilitators (e.g., JP Morgan, Goldman Sachs, etc) and liquidity providers (e.g., hedge funds). These traders are making preparations so that passive investors can rebalance all at once on reconstitution day.

Some passive rebalancing is organized the day of reconstitution via market-on-close orders. However, many passive investors prefer to prearrange their rebalancing trades. An internal indexer tracking the Russell 1000 might contact, say, JP Morgan in February to set up rebalancing trades which will be executed at the closing price on the last Friday in June. JP Morgan would then line up liquidity providers—i.e., a group of hedge funds who are willing to sell each Russell 1000 addition and a group who is willing to buy each deletion. The deal would get finalized months ahead of reconstitution. On reconstitution day, these trades would get executed as large upstairs transactions.

Early in our sample period, it was common for prearranged rebalancing
trades to include price improvement. For example, in 2007 it would not have
been unusual for JP Morgan to sell each Russell 1000 addition to the internal
indexer at the closing price on Russell reconstitution day minus $0.01. Our
understanding is that this practice is much less common today.

As we will see shortly, these prearranged trades explain why there is so little
price impact on reconstitution day. “The industry does a good job of forecasting
and facilitating index demand. […] Despite the huge volumes, the annual Russell
reconstitution is usually a relatively orderly close. A few stocks typically see
some market impact late in the day, but in general the index trades are matched
up pretty well by liquidity providers. (Mackintosh, 2020)”

For theorists, there are two particularly noteworthy things about this trading
arrangement. First, it is explicitly designed so that passive investors can trade a
specific quantity regardless of the prevailing price. This is the exact opposite of
Grossman and Stiglitz (1980) where all traders observe the equilibrium price
before choosing their demand.

Second, passive investors devote substantial resources to managing reconsti-
tution events. They are not uninformed traders in a Grossman and Stiglitz (1980)
model. There are sell-side analysts specializing in index reconstitution events in
the same way that there is sell-side research on firm fundamentals (Nomura,
2022). These information providers have been around for decades.¹¹

3.3 Case Study: Tesla

One way to highlight the importance of prearranged rebalancing is to look at
what happens in a situation where passive investors could not preschedule their
rebalancing trades. Market events conspired to construct exactly this sort of
situation for Tesla Inc (TSLA)’s addition to the S&P 500 in December 2020. Here
we will lay out the facts and document the ways that Tesla’s rebalancing differed
that of a typical S&P 500 addition. In the following section, we will explore how
these differences in trading patterns manifested in prices.

Investors usually have a good idea about who will be added to the S&P 500

¹⁰e.g., see also www.bloomberg.com/what-goes-into-maintaining-an-equity-index/.
Figure 13. Volume for Tesla Inc (TSLA) around its addition to the S&P 500 in millions of shares. Solid bars represent total volume each day. White bars represent volume from 4:00pm to 11:59pm. On November 17th (green), S&P Dow Jones announced that Tesla would join the S&P 500 following market close on December 18th (red). Black y-axis label denotes Tesla’s average daily volume in 6 months prior to this announcement. Grey region is Tesla’s average daily volume from November 17th to December 17th.

before S&P Dow Jones makes its formal announcement. However, it came as something of a surprise when the index provider announced on November 17th that Tesla would get added to the S&P 500 on December 18th 2020. Even though the company was the 6th largest US firm, S&P Dow Jones had “passed [Tesla] over in several previous index reshuffles.” Many did not expect the company to get added in December 2020 either.

In addition to being surprised, S&P 500 investors also had relatively little time to prepare. While S&P Dow Jones made a formal announcement 22 trading days prior to Tesla’s inclusion, this was all the time that investors got to prepare. By contrast, for a normal event, investors are able to predict the change months ahead of time. Even if investors cannot predict exactly which stocks will be added to or dropped from the S&P 500, they usually have a shortlist of candidates.

To further complicate matters, rebalancing facilitators found it hard to line up liquidity providers. S&P Dow Jones’ initial press release did not say “which current constituent Tesla [would] replace [or] how Tesla [would] be added. (S&P Dow Jones Indices, 2020b)” And liquidity providers were still feeling the


In short, the usual trading apparatus behind passive rebalancing broke down when Tesla was added to the S&P 500. Only 222.1m/52.2m = 4.2x Tesla's average daily volume got traded on December 18th 2020, and roughly half of this volume took place during normal trading hours, (222.1m – 116.4m)/222.1m = 47.6%. Rather than using prearranged trades, S&P 500 investors were forced to do much of their rebalancing during the month prior to reconstitution. Tesla’s average daily volume from November 17th through December 18th, 52.2m shares, was nearly double its average daily volume during the 6 months prior, 29.0m shares.

4 Asset Prices

Passive investors often prearrange their rebalancing trades to be executed at market close on reconstitution day. We now look at how this practice affects prices. In subsection 4.1 we document how there is no longer much price pressure on reconstitution days. Prices move when trades are arranged not when they are executed. In subsection 4.2 we show how this disconnect helps explain the wide range of elasticity estimates found in the literature.

4.1 Abnormal Returns

Given that passive investors are doing some much trading at market close on reconstitution day, you might expect to see large price effects on reconstitution day. There used to be one (Harris and Gurel, 1986; Shleifer, 1986; Beneish and Whaley, 1996; Wurgler and Zhuravskaya, 2002; Madhavan, 2003; Petajisto, 2011). However, as Greenwood and Sammon (2022) document, there is no longer much predictable price pressure on reconstitution days.

To illustrate this fact, we form two kinds of value-weighted portfolios on $t_{Recon}$. We create one for index additions

$$\text{RetAdds}_i(t_{Recon}) = \frac{\sum_{n \in \text{Adds}_i} MCap_{n'}(t_{Recon} - 1) \cdot \text{Ret}_n(t_{Recon})}{\sum_{n' \in \text{Adds}_i} MCap_{n'}(t_{Recon} - 1)}$$

and another for index deletions, $\text{RetDrops}_i(t_{\text{Recon}})$.

Figure 14 reports the average value of $\text{RetAdds}_i(t_{\text{Recon}})$ and $\text{RetDrops}_i(t_{\text{Recon}})$ in excess of the market for all reconstitution events for a given index each year. The bottom-right panel show that the returns to buying index additions and the returns to selling index deletions have steadily converged to nearly zero. This has happened in spite of the fact that the overall passive ownership share has grown from 13.4% to 33.5% during this same period.

To be clear: index inclusion does affect prices. The effect just does not come by way of passive rebalancing demand on reconstitution days. Greenwood and Sammon (2022) document that, even in the modern era there are announcement-day returns associated with direct additions and deletions. Under normal conditions, passive investors have already scheduled most of their rebalancing trades well in advance of reconstitution day.

Tesla’s addition to the S&P 500 on December 18th 2020 was an exception to
Figure 15. Tesla Inc (TSLA)’s closing price in the days around its addition to the S&P 500. On November 17th (green), S&P Dow Jones announced that Tesla would join the S&P 500 following market close on December 18th (red). Percentages reported in the figure are realized returns on announcement day and reconstitution day. Black y-axis label is Tesla’s closing price on January 8th.

this rule. S&P 500 investors had a hard time prearranging enough rebalancing trades, so Tesla realized a 5.96% reconstitution-day return in Figure 15. When we include Tesla, the returns to S&P 500 additions in 2020 are 5.14%pt higher than the market return (Figure 14; white dot, lower left panel). All other S&P 500 additions in 2020 have abnormal reconstitution-day returns of just 1.20% (Figure 14; corresponding black bar).

4.2 Elasticity Estimates

Because they are often prescheduled, passive investors’ trades can affect prices long before they get executed. As a result, if a researcher uses too narrow a time window around reconstitution day, she may miss the price impact associated with passive demand, making the market seem too elastic. This observation can explain the wide range of elasticity estimates associated with index reconstitution events. For example, Wurgler and Zhuravskaya (2002, Table 4) reports values from $-11.72$ (more elastic) to $-1.00$ (less elastic).

To highlight this issue, we compute demand elasticities for direct additions using price changes at different horizons. We focus on direct additions to get the cleanest signal. When a stock gets migrated from, say, the Russell 2000 up to the Russell 1000, some of the buying pressure coming from Russell 1000 investors
will be met by Russell 2000 investors who need to sell. The same goes for moves between the S&P 500 and S&P MidCap 400. We do not include direct drops because these changes are often due to an acquisition or bankruptcy, making it unclear how much of the associated price change is due to passive rebalancing.

Let \( \text{Multiplier}_{\text{ADD}}(h) \) denote the increase in a direct addition’s price during the \( h \) trading days leading up to reconstitution day divided by the stock’s volume on reconstitution day as a fraction of shares outstanding

\[
\text{Multiplier}_n(h) = \left( \prod_{\ell=1}^{h} \left[ 1 + \text{AbnRet}_n(t_{\text{Recon}} - \ell) \right] \right)^{60/h} - 1
\]  

(15)

The demand shock is the same at every horizon: a direct addition’s reconstitution-day volume as a percent of shares outstanding. The only thing that changes as we extend the horizon is the length of time over which we measure the price impact. So that longer horizons do not look artificially inelastic, we calculate each stock’s return in excess of the market, \( \text{AbnRet}_n(t) = \text{Ret}_n(t) - \text{Mkt}(t) \).

We look at horizons \( h \in \{60, \ldots, 1\} \) where \( h = 60 \) denotes the quarter before reconstitution day and \( h = 1 \) denotes just reconstitution day itself. For each index we first estimate the median price multiplier at each horizon. We report the associated demand elasticity values in Figure 16 where

\[
\text{Elasticity}_i(h) = -\frac{1}{\text{Multiplier}_i(h)}
\]  

(16)

If \( \text{Elasticity}_i(h) = -5 \), demand would fall by 5% were prices to increase by 1%.

The red dot and \( y \)-axis label in each panel represents \( \text{Elasticity}_i(1) \). This calculation incorrectly assumes that the price impact of passive investors’ demand will occur on reconstitution day itself. As a result, it makes markets look far too elastic. The bottom-right panel in Figure 16 indicates that, on average, a 1% increase in a direct addition’s price on reconstitution day is associated with a 39.2% decrease in demand for the stock on reconstitution day.

By contrast, demand looks less elastic once we extend the horizon to include the price impact from prearranged trades. The bottom-right panel in Figure 16 indicates that, on average, a 1% increase in a direct addition’s price over
Figure 16. Median demand elasticity for direct additions at a particular horizon. The demand shock is reconstitution-day volume as a percent of its shares outstanding. The price change is cumulative excess return over the past $h \in \{60, \ldots, 1\}$ trading days (x-axis). The red dot corresponds to the estimate at $h = 1$, which only use a stock’s price increase on reconstitution day. The left-most point in each panel uses a stock’s price increase over the past quarter. The vertical white line in each panel denotes the average time between announcement day and reconstitution: 5 trading days for the S&P 500, S&P MidCap 400, and Nasdaq 100; 15 trading days for the Russell 1000 and Russell 2000.

the quarter leading up to reconstitution is associated with a 1.1% decrease in demand on reconstitution day.

What’s more, if our story is correct, then we should see a kink in the elasticity estimates at horizons where lots of rebalancing trades get prearranged. To test this hypothesis, we exploit differences in the timing and information content of announcement days across indexes. The vertical white line in each panel represents the point at which an index provider typically makes a formal announcement about upcoming changes.

FTSE Russell typically makes an announcement 15 trading days before Russell reconstitution day. However, it is usually possible to identify direct additions well before then. So there is no sharp change in the elasticity estimates
for the Russell 1000 and Russell 2000. The values in these two panels gradually fall down and to the right.

S&P Dow Jones and Nasdaq typically announce changes to their indexes 5 trading days prior to reconstitution. Moreover, their announcements typically contain new information. Prior to the announcement, investors have an idea about a handful of stocks that might be added. The announcement then reveals which stock from this shortlist was selected for inclusion. Hence, many prearranged rebalancing trades get arranged at the announcement. This is when passive investors’ demand shock shows up in prices, and using a shorter event window leads to spuriously low elasticity estimates in Figure 16.

We are able to reproduce the range of elasticity estimates (from $-39.2$ to $-1.1$) found in the literature. However, once you recognize how passive investors trade, it is clear that values near $-1$ reflect the demand elasticity of rebalancing facilitators, not passive investors. Passive investors are perfectly inelastic. They prearrange to trade a fixed quantity of additions and deletions at whatever the price happens to be on reconstitution day at the close. S&P 500 investors did not buy less Tesla because its price went up post announcement.

**Conclusion**

Each time a stock gets added to or dropped from a popular index, we ask: “How much money would have to be tracking that index to explain the huge spike in rebalancing volume we observe on reconstitution day?” We find that passive investors held 33.5% of the US stock market in 2021. This headline number is roughly double the index-fund ownership share in 2021, 16%, because index funds are not the only kind of passive investor.

The particular way that we estimate the US passive ownership share also gives theorists guidance on how to model the rise of passive investing going forward. To start with, there is not widespread agreement among market participants about the true passive ownership share. This number is not common knowledge. We should not be modeling investors as choosing between active and passive strategies based on a broad understanding of how many other investors have made the same decision.
Some market participants were aware that there was a lot of money being invested outside of the public index-fund universe. For example, in a 2017 white paper, researchers at BlackRock estimated that index funds held $5.0t in combined AUM while internal indexers held $6.8t (Novick et al., 2017). The Investment Company Institute is also clearly aware that index funds are not the only kind of passive investor. They are in no way misleading market participants. The title of Figure 2.9 in Investment Company Institute (2022) is “Index Fund Share of US Stock Market Is Small.”

However, in spite of this disclaimer, “people often [forgot] that open-ended investment funds only [held] a slice of markets, and [conflated] passive’s mutual fund industry market share with its overall market ownership.” Prior to this paper, it was difficult to gauge how much additional money was being passively invested outside of index funds. And there was not broad appreciation of how strictly passive investors track their benchmarks.

The way that passive investors rebalance does not match up with the usual noisy rational-expectations paradigm (Grossman and Stiglitz, 1980), which assumes that investors observe the equilibrium price before choosing their demand. Instead, theorists should focus on passive investors’ inelastic demand à la Haddad, Huebner, and Loualiche (2022). These traders often prearrange rebalancing trades to be executed at the closing price on reconstitution day, regardless of what that price is. They are sophisticated traders who dedicate substantial resources to managing reconstitution.

References


A Regression Analysis

To motivate our Kalman-filter specification, we run time-series regressions to understand the auto-regressive structure of our baseline estimates. Panel (a) in Table A1 reports regressions of the form below

\[
\%\text{Indexed}_i(y) = \hat{\mu} + \sum_{\ell=1}^{3} \hat{\theta}_\ell \cdot \%\text{Indexed}_i(y - \ell) + \hat{\eta} \cdot y + \hat{\epsilon}_i(y)
\] (A.1)

Columns (1)-(5) in Table A1 look at each of our five indexes on its own. Column (6) looks at the total passive ownership share for all five indexes. There are 22 years from 2000 through 2021, but we cannot compute lags the first 3 years. So these first six columns involve \(22 - 3 = 19\) annual observations. Column (7) is a pooled specification with all five in a given year, \(19 \times 5 = 95\) observations in total. We include a separate intercept term for each index but force the lag coefficients and time trend to be the same for all indexes. None of the lags has a coefficient larger than unity, \(|\hat{\theta}_\ell| < 1\), and only the first lag is ever significant.

However, the passive ownership share has been increasing over the past twenty years. The time trend is statistically significant in panel (a) of Table A1. So in panel (b) we estimate an analogous specification in differences

\[
\Delta\%\text{Indexed}_i(y) = \hat{\mu} + \sum_{\ell=1}^{3} \hat{\theta}_\ell \cdot \Delta\%\text{Indexed}_i(y - \ell) + \hat{\epsilon}_i(y)
\] (A.2)

Again, there is no long-term auto-correlation structure. This suggests that we only need to include a single lag when estimating our Kalman filter, which is extremely helpful given how short our time series is.

If you thought that index funds always owned a constant share of all passive investments, then you could use this additional assumption to constrain the evolution of our baseline estimates via a Kalman filter. But, before fitting a Kalman filter to the data, we first report what this relationship looks like on average. What fraction of all passive investments are held by index funds on average during our sample period?
### Time-series regressions

#### (a) %Indexed

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</table>

Index FE: \( N \ N \ N \ N \ N \ N \ Y \)

# Obs: 19 19 19 19 19 19 95

\[ R^2 = 68.6\% \ 73.2\% \ 71.1\% \ 69.5\% \ 70.3\% \ 76.1\% \ 97.5\% \]

#### (b) Δ%Indexed

<table>
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<td>Lag 2</td>
<td>-0.24</td>
<td>-0.19</td>
<td>0.13</td>
<td>0.12</td>
<td>-0.05</td>
<td>-0.12</td>
</tr>
<tr>
<td>Lag 3</td>
<td>-0.10</td>
<td>0.31</td>
<td>0.20</td>
<td>0.25</td>
<td>-0.15</td>
<td>0.08</td>
</tr>
<tr>
<td>Year</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

Index FE: \( N \ N \ N \ N \ N \ N \ Y \)

# Obs: 18 18 18 18 18 18 90

\[ R^2 = 7.3\% \ 31.9\% \ 16.1\% \ 10.2\% \ 4.7\% \ 5.9\% \ 8.6\% \]

**Table A1.** Each column reports the results of a separate time-series regression using annual observations. Columns (1)-(6) include one observation per year. Column (7) is a pooled specification that includes five observations per year—i.e., one for each index. The state variable in panel (a) is our estimate for the passive ownership share. Sample: 2003 to 2021. The state variable in panel (b) is the change in these estimates for the passive ownership share. Lag # is the coefficient on a lagged value of the state variable. Year is the coefficient on the time trend. Numbers in parentheses are t-stats. Sample: 2004 to 2021.
Table A2 reports the results of regressing the index-fund ownership share on the overall passive ownership share in a given year

\[
\%\text{IndexFundOwned}_i(y) = \hat{k} + \hat{\lambda} \cdot \%\text{Indexed}_i(y) + \hat{\epsilon}_i(y)
\]  

(A.3)

The first row shows results using the ICI's numbers for the index-fund ownership share. The ICI’s numbers reflect all index funds that track any index, including ones we do not have data on. The next five rows use Thomson S12 data to compute the ownership share of index funds tracking the Russell 1000, the Russell 2000, the S&P 500, the S&P MidCap 400, or the Nasdaq 100. We follow Appel, Gormley, and Keim (2016) and use a name-based classification system. For example, when considering the S&P 500, we look for combinations of “S&P”, “S & P”, “SandP”, “S and P”, and “SP” together along with “500”. The final row uses the combined AUM of index funds that track one of our five indexes.

The 0.54 slope coefficient in panel (a) of Table A2 implies that growth in index investing is responsible for roughly half of the overall growth in passive investing. Index-fund holdings contribute 54 cents of each $1 increase in overall passive ownership. Panel (b) shows analogous results for each individual index. As expected, the slope coefficients are now much smaller. Index funds are responsible for just 23 cents of each $1 increase in passive AUM tracking the S&P 500. Russell 1000 index funds account for just 4 cents out of every $1 increase in AUM tracking the Russell 1000.

In the introduction, we suggested that if all passive investing were done via index funds, then we would have estimated a passive ownership share of just 6% in 2021. This remark is based on the slope coefficient of 0.16 in the last row of panel (b), which implies that \(0.16 \times 33.5\% \approx 6\%\) of the US stock market was owned by the subset of index funds tracking our five indexes in 2021. The ICI’s 16% figure is not a lower bound because it is based on more indexes.

B Investor Costs

Many passive investors use prearranged trades to execute their rebalancing orders right at market close on reconstitution day. We do not know of any official
### Index-fund share of passive ownership

**Panel (a) ICI Factbook Intercept Slope Adj. $R^2$ # Obs**

<table>
<thead>
<tr>
<th>All index funds that track any index</th>
<th>Intercept</th>
<th>Slope</th>
<th>Adj. $R^2$</th>
<th># Obs</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>$-4.01^{**}$</td>
<td>$0.54^{***}$</td>
<td>$70.7%$</td>
<td>22</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>(b) Thomson S12 Funds that track the...</th>
<th>Intercept</th>
<th>Slope</th>
<th>Adj. $R^2$</th>
<th># Obs</th>
</tr>
</thead>
<tbody>
<tr>
<td>S&amp;P 500</td>
<td>$-0.97^{**}$</td>
<td>$0.23^{***}$</td>
<td>$67.8%$</td>
<td>22</td>
</tr>
<tr>
<td>S&amp;P MidCap</td>
<td>$-0.19^{***}$</td>
<td>$0.43^{***}$</td>
<td>$67.7%$</td>
<td>22</td>
</tr>
<tr>
<td>Russell 1000</td>
<td>$-0.09^{*}$</td>
<td>$0.04^{***}$</td>
<td>$73.6%$</td>
<td>22</td>
</tr>
<tr>
<td>Russell 2000</td>
<td>$-0.05^{*}$</td>
<td>$0.25^{***}$</td>
<td>$76.3%$</td>
<td>22</td>
</tr>
<tr>
<td>Nasdaq 100</td>
<td>$0.11^{***}$</td>
<td>$0.08^{***}$</td>
<td>$45.0%$</td>
<td>22</td>
</tr>
<tr>
<td>Funds that track one of these five indexes</td>
<td>$-1.17^{**}$</td>
<td>$0.16^{***}$</td>
<td>$76.4%$</td>
<td>22</td>
</tr>
</tbody>
</table>

**Table A2.** Coefficients estimates from regressing the index-fund ownership share on the overall passive ownership share. Each row represents a separate regression using annual observations. Panel (a): Left-hand side is the ICI’s index-fund ownership share, which reflects all index funds tracking any index. Right-hand side is the ownership share of passive investors tracking the Russell 1000, the Russell 2000, the S&P 500, the S&P MidCap 400, or the Nasdaq 100. Panel (b): Left-hand side is ownership share of index funds tracking just these five indexes. Right-hand side is ownership share of passive investors tracking the same index. Numbers in parentheses are $t$-stats. Sample: 2000 to 2021.

Data, but market participants have quoted us fees in the range of 5bps. Hence, for there to be a huge spike in reconstitution-day volume, the cost of rebalancing in the days prior to reconstitution must be at least this high.

We start in subsection B.1 by looking at whether an S&P 500 investor could improve his risk-return profile by rebalancing his portfolio the day after S&P Dow Jones announces each reconstitution event. He cannot. Hence, any gains from rebalancing early must come from cost savings. In subsection B.2, we then show that the volume-weighted spread for S&P 500 index additions and deletions is roughly 5bps on the day following the announcement.
**Daily returns to S&P 500 early-rebalancing portfolio**

<table>
<thead>
<tr>
<th></th>
<th>Full Sample</th>
<th>Relative to Tesla’s Addition</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Start date</td>
<td>End date</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>$\text{S&amp;P 500}$</td>
<td>4.37**</td>
<td>13.44</td>
</tr>
<tr>
<td></td>
<td>(1.77)</td>
<td>(0.10)</td>
</tr>
<tr>
<td>L/S Strategy [bps]</td>
<td>0.02***</td>
<td>2.67***</td>
</tr>
<tr>
<td></td>
<td>(0.01)</td>
<td>(0.03)</td>
</tr>
<tr>
<td># days</td>
<td>5036</td>
<td>4754</td>
</tr>
</tbody>
</table>

Table B1. Daily returns to the S&P 500 and a zero-cost strategy that rebalances the S&P 500’s weights the day after S&P Dow Jones announces an upcoming reconstitution. Numbers in parentheses are standard errors. The first column reports results for the full sample: January 2nd 2002 through December 31st 2021. The next three columns report results for different subperiods: before Tesla’s addition was announced; the day after the announcement through reconstitution day; and, the rest of the sample period.

### B.1 Average Returns

Consider a private wealth manager who is internal indexing to the S&P 500. The S&P 500 is a value-weighted index, and its constituent stocks were worth $10.4t on January 2nd 2002. Suppose the manager has $1b in AUM at the start of 2002. In that case, the manager’s assets would equate to roughly 0.01% of the total S&P 500 universe. Hence, to replicate the S&P 500, he would need to hold 0.01% of the float for each S&P 500 stock.

Changes to the S&P 500 are typically announced 5 trading days prior to reconstitution. Following each announcement, the manager could continue to hold 0.01% of existing S&P 500 constituents until market close on reconstitution day. The first row of Table B1 shows that, from January 2nd 2002 through December 31st 2021, the S&P 500 averaged returns of 4.37bps per day. Each $1 invested at the start of the sample period turned into $6.20 by the end, which amounts to a 9.55% annualized return and a 0.57 annualized Sharpe ratio.

However, the private wealth manager could also rebalance early. On the day after each S&P Dow Jones announcement, he could rebalance his portfolio so that
Figure B1. Cumulative returns to a zero-cost strategy that starts off with long/short $1 on January 2nd 2002. The sample period ends December 31st 2021. The long leg is the returns to an early-rebalancing version of the S&P 500, which adjusts its weights to reflect each upcoming reconstitution the morning after the event is announced. The short leg is the returns to the S&P 500. On November 17th 2020, S&P Dow Jones announced that Tesla would be joining the S&P 500 after markets closed on December 18th 2020.

The second row of Table B1 reports the difference between the returns to this early-rebalancing portfolio and the actual S&P 500. Over the entire sample period, this new portfolio outperforms the official S&P 500 by just 0.02bps per day. A strategy that went $1 long the early-rebalancing portfolio and $1 short the S&P 500 on January 2nd 2002 produced just $0.05 in profit by December 31st 2021 as shown in Figure B1.

Table B1 tells us the private wealth manager cannot boost his average return by rebalancing early. For most of the sample period, rebalancing the S&P 500 early would have done nothing except increase the volatility of the manager’s returns. The cumulative returns to the long/short strategy were negative on the Monday before Thanksgiving in 2020. The entire $0.05 gain was earned during the remaining 13 months of the sample period. The third column in Table B1 shows that half of this $0.05 gain came during the 22 trading days following Tesla’s announcement. And this is all before considering trading costs.
B.2 Marginal Spread

We have just seen that an S&P 500 investor cannot boost his risk-return profile by rebalancing early. Hence, any gains from rebalancing early must come from cost savings. The effective spread is the difference between a stock's bid and ask prices divided by the midpoint. Figure B2 shows the volume-weighted average effective spread each year for index additions and deletions on the day after the change was announced by the index provider.

The bottom-left panel reports results for the S&P 500. The numbers indicate that, taking prices as given, an investor that was internal indexing to the S&P 500 would have to pay an effective spread of between 4bps and 6bps in recent years if he were to rebalance early. The numbers are slightly higher for the S&P MidCap 400 and Russell 1000. They are much higher for the Russell 2000 since this index is composed of smaller stocks.

Hence, the observed spreads during normal trading hours on the day after announcements are on par with what market participants have told us about the cost of prearranging a trade. This does not mean that the practice of prearranging
trades is unimportant. The exact opposite is true. If all $7.5t in AUM that was tracking the S&P 500 in 2021 had instead tried to rebalance the day after announcements, we would not have estimated an effective spread of 6bps in Figure B2. The practice allows passive investors to push huge volumes through equity markets at a cost of 5bps, which is tantamount to the marginal spread on the day following an announcement.