



INTERNATIONAL  
BUSINESS SCHOOL  
The Rosenberg  
Institute of  
Global Finance

Vanessa Countryman, Secretary

Securities and Exchange Commission

100 F Street, NE

Washington, DC 20549-1090

June 7, 2022

Re: File No. S7-13-22 Special Purpose Acquisition Companies, Shell Companies, and Projections

Dear Secretary Countryman,

We are finance researchers at Brandeis University and Dickinson College and affiliated researchers with the Rosenberg Institute of Global Finance at Brandeis International Business School. We hereby submit this comment letter to the SEC in response to its published notice of proposed rules for SPACs and related matters (the “Proposed Rules”).

Our comments are drawn from findings in our two research papers on SPACs: *“Is Going Public via SPAC Regulatory Arbitrage? A Textual Analysis Approach”* (Wen and Zhu, 2022); *“The Going Public Decision and Retail Investors in SPACs vs. IPOs”* (Nandy and Wen, 2021).

We examined SPAC IPOs and de-SPAC transactions completed through the end of 2020 and continue to update the analyses. We highlight main findings relevant to the Proposed Rules:

**1. Enhancing investor protections and transparency in SPAC IPOs and the subsequent business combinations is warranted.**

In Nandy and Wen (2021), we find that retail investors' interest in SPACs is significantly lower prior to the announcement of the business combination target, but increases quickly after the target announcement. The increase in retail interest is mainly driven by the stock return volatility rather than target firm fundamentals. Applications of the Plain English disclosure rules will assist retail investors to understand the risks involved in the investment. Overall, financial returns to retail investors depend critically on when they decide to invest in SPACs (prior or after the target firm announcement).

**2. We support standardized and specialized disclosure rules for SPACs and the subsequent business combination to align disclosures more closely with those in traditional IPOs.**

In Wen and Zhu (2022), we compare SEC filings between operating companies that are fundamentally similar but went public through two different processes (de-SPAC vs. traditional IPO). We apply textual analysis techniques to systematically compare the disclosure sentiments and information content at multiple stages of the going-public process<sup>1</sup>. We find that regulatory arbitrage exists in the de-SPAC process. In particular, de-SPAC merger filings contain 44% less pessimistic statements, 59% more definitive statements, and 56% fewer unique financial words, based on the comparison between the de-SPAC DEF 14A and traditional IPO S-1. This indicates that, given similar firm

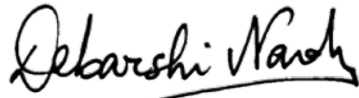
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<sup>1</sup> Between SPAC IPO and traditional IPO, we compare (1) changes made between Form DRS and S-1 during the confidential review stage; (2) Form S-1 during the public roadshow stage; (3) and, de-SPAC's DEF 14 to traditional IPO's S-1 during the post-deSPAC announcement

fundamental characteristics, de-SPAC disclosures tend to be overconfident and less informative than IPOs disclosures. Additionally, we show that the differences in the above-mentioned aspects become negligible in post-public annual report filings, where identical disclosure requirements are applied, and regulatory arbitrage no longer exists.

We attach a copy of the two papers and would be happy to present or further discuss our research. Please feel free to reach out to us if there are any questions or if we can provide any further information.

Sincerely,



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# Is Going Public via SPAC Regulatory Arbitrage? A Textual Analysis Approach

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March 25, 2022

## Abstract

We study firm's incentives and disclosure practice of going public via Special Purpose Acquisition Company (SPAC). We argue that going public via deSPAC merger is regulatory arbitrage, through which firms circumvent the SEC review process in conventional IPO registration and provide less comprehensive but overconfident disclosures to the public. Using textual analysis, we compare the information content of prospectuses and merger proxy statements in deSPAC mergers to their propensity-score-matched IPO peers. We find that deSPAC merger filings contain 44% less pessimistic statements, 59% more definitive statements, and 56% fewer unique financial words. Additionally, we show that the differences in the above-mentioned aspects become negligible in post-public annual report filings, where identical disclosure requirements applied and regulatory arbitrage no longer exists. Lastly, we document the impact of regulatory arbitrage on firms' litigation risk and stock performance after going public. We call for unified regulations overseeing alternative going-public methods to maintain information transparency and a fair public capital market.

**JEL:** G14, G18, G24

**Keywords:** SPAC, IPO, Regulatory Arbitrage, Textual Analysis

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# 1 Introduction

In this paper, we study the firm disclosure practice and consequences with alternative going-public methods in the U.S., comparing conventional IPO and deSPAC merger. We find that there exists regulatory arbitrage and call for unified regulations.

Since 2018, the U.S. public market has experienced an unprecedented surge of SPAC (Special Purpose Acquisition Company) activities. SPACs are publicly listed shell companies, whose purpose is to merge with private operating companies such that the resulting entity becomes publicly traded. Such a transaction is often referred to as a deSPAC merger. In 2020, 27% of private operating companies that went public through the deSPAC merger instead of the conventional IPO process. At the end of 2021, over 200 SPACs are actively searching for a merger target. Hence, understanding firm's incentives and the impact of this surge of alternative going-public method can shed light on the efficiency of public market, as we expect that there will continue to be a pipeline of SPAC activities in the near future.

The conventional IPO and deSPAC merger are both means for private operating companies to debut in the public market. However, the two methods are not subject to unified regulations and procedures, particularly with respect to regulator oversight. In a conventional IPO, the issuing firm files a registration statement with the Securities and Exchange Commission (SEC) about the initial public offering and undergoes a mandatory full cover-to-cover review by the SEC. In the deSPAC merger process, the transaction is typically structured as a merger between the public shell company (i.e., SPAC) and the private operating company that consequently obtains access to the public capital market as a post-merger entity. A variety of filings, such as proxy statements, are filed with the SEC during the deSPAC merger process, in which the SEC review these transactional filings on a selective basis.<sup>1</sup> In terms of regulatory oversight, a conventional IPO is subject to mandatory scrutiny by the SEC, while a deSPAC merger is subject to selective SEC review that may be less extensive in scope.

A natural question, yet overlooked in current literature, is whether there exists reg-

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<sup>1</sup>The Division of Corporation Finance of the SEC “selectively reviews filings made under the Securities Act of 1933 and the Securities Exchange Act of 1934 to monitor and enhance compliance with the applicable disclosure and accounting requirements.” See [sec.gov/divisions/corpfin/cffilingreview.htm](https://www.sec.gov/divisions/corpfin/cffilingreview.htm)

ulatory arbitrage between the two going-public methods and whether there is harm to investors if the arbitrage exists. Regulatory arbitrage commonly refers to the attempt to exploit loopholes and gaps in regulatory regime in order to gain business opportunities or to decrease regulatory burdens. Whether going public via SPAC is regulatory arbitrage is an ongoing debate among legal and capital markets practitioners, is being contemplated by the regulator, and a critical question to entrepreneurs considering taking their firms public. This paper addresses this question with a systematic comparison of firm disclosure, litigation risk, and post-public stock performance between the firms using the two going-public methods. We construct a comparable sample that contains firms went public using the conventional IPO method (IPO firms) and firms went public via the deSPAC merger (deSPAC firms). Next, five IPO peer firms are identified for each deSPAC firm applying propensity-score-matching (PSM) method to mitigate selection bias. Then, we measure and compare the information content in SEC filings during the going-public process using textual analysis.

In a nutshell, we find evidence in support of the existence of regulator arbitrage when going public via SPAC. First, we compare the information content of registration statements in SPAC vs. non-SPAC IPOs (i.e., shell companies vs. operating companies). As expected, SPAC prospectus primarily focuses on legal matter and the information production by SEC is sparse. Secondly, we compare the proxy statement in deSPAC merger to the prospectus of an operating company in the conventional IPO and find that deSPAC firms' disclosures are substantially less pessimistic and overconfident, and contain less informative content compared to the IPO peers. After going public, deSPAC firms also experience higher incidence of securities class action lawsuits and larger decline in share prices. In a placebo test, we focus on the first annual report that is subject to identical regulations after going public and no longer observe significant differences in the same textual measures between firms using the two different going-public methods.

Our results suggest that, given the current regulatory environment, deSPAC firms tend to disclose less information and avoid using unfavorable languages compared to its propensity-score-matched IPO peers. However, the differences vanish when identical regu-

lations apply in the post-public periodic filings. Therefore, we recommend that the regulator improves the existing review procedure in deSPAC transactions, such that the level of information transparency and regulatory oversight is comparable to the conventional IPO process.

This paper contributes to the going-public and the financial regulation literature. First, our work is part of the growing body of research on the current generation of SPACs, which have examined the SPAC instrument structure, investment return, sponsor characteristics, and more. Inspired by the ongoing policy debate on SPAC regulations, this study investigates firm disclosure and investor protection in deSPAC mergers, where we provide the first comprehensive comparison of SEC filings of deSPAC firms and matched IPO peer firms in the going-public process. Our paper complements research on the usage of financial projections in deSPAC mergers, but with important distinctions, by which we introduce alternative methodology and insights to the collective understanding in this area. [Blankspeer et al. \(2021\)](#) provides evidence that deSPAC mergers often include highly optimistic projections and have long-horizon projections up to five years or more. [Dambra et al. \(2021\)](#) finds positive correlations between projected revenue growth and both market returns and abnormal retail trading around investor presentation of the deSPAC merger. Finally, [Chapman et al. \(2021\)](#) constructs five measures of the forward-looking content in investor presentations and finds no evidence that more forecasts are associated with negative return or damage to investors.<sup>2</sup>

While these studies and ours share the focus on firm disclosures in the deSPAC process, our study has two main differences with respect to methodology and empirical setting. First, we compare deSPAC and IPO firms, so as to answer the question of regulatory arbitrage, while other studies mainly examine variations within the SPAC sample. Our analysis adds to the understanding of if and how the current practice of firm disclosure is different in deSPAC merger, given the existing regulatory environment. We show that deSPAC merger related filings tend to be less informative than IPO filings using propensity-score-matched samples. The second important distinction is that we consider a broader information set

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<sup>2</sup>Despite the recent rise in popularity, SPACs have also been studied in earlier periods (see, e.g. [Floros and Sapp, 2011](#); [Rodrigues and Stegemoller, 2014](#); [Cumming et al., 2014](#); [Kolb and Tykova, 2016](#)).

available to investors during firm’s entire going-public process. We analyze SEC filings at different stages of a SPAC’s life cycle, rather than narrowly focusing on forward-looking statements in investor presentations.<sup>3</sup> In addition, we provide an overarching view of what these documents entail using textual analysis approach which is advantageous in that the quantitative measures are objective and applicable to large-scale firm filings with high efficiency and accuracy.

Secondly, this paper is related to the broader literature in regulatory arbitrage. Typical definition of regulatory arbitrage refers to firms attempt to relocate transactions to jurisdictions of more favorable regulations. In the going-public context, although there is only one dominating regulator (i.e., SEC), deSPAC merger and conventional IPO are not subject to unified regulatory procedures. Hence, the question of regulatory arbitrage arises. We compare the intensities of regulatory oversight in the two alternative going-public methods by quantifying the information content of related SEC filings. The literature suggests that when regulatory burdens are high, alternative transactions that are not subject to the same regulations tend to grow.<sup>4</sup> Analogous to the banking industry, we hypothesize that the recent surge of SPACs is partly attributed to the lessened disclosure burden in the going public process. Our paper provides new evidence that contributes to the group of literature focusing on the post-public firm performance and investor protection. For example, [Klausner et al. \(2020\)](#) analyzes the factors that affect post-merger returns of all 47 SPACs that merged between January 2019 and June 2020. They caution the high dilution to public shareholders’ equity due to sponsor promote, public share redemption, and warrant issuance. [Gahng et al. \(2021\)](#) focuses on the return on common shares and warrants of exchange-traded SPACs since 2010. They also document that sponsors and underwriters forfeit shares and warrants, or structure earn-out compensation in order to induce business combination. In terms of the aggregate SPAC activity cycle, studies show that SPAC activity correlates with positive sentiment in the equity markets ([Bai et al., 2020](#)) and negatively correlated with market uncertainty index (VIX) and time-varying

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<sup>3</sup>Investor presentations are released at the early stage during the deSPAC merger process and the format or content structure varies significantly across companies.

<sup>4</sup>For example, shadow banking is more likely to expand where traditional banks face more regulatory constraints ([Buchak et al., 2018](#)).



risk aversion (Blomkvist and Vulanovic, 2020). Lin et al. (2021) finds that SPAC sponsor network centrality based on their overlap in work experience at PE/VC firms positively correlate with investment return prior to the deSPAC event. Through the comparison to the matched IPO peer firms, our results suggests that there is significant harm to the investors due to the regulatory arbitrage in deSPAC merger method. We show deSPAC firms face higher litigation risk and experience larger decline in stock returns.

Lastly, we contribute to the application of textual analysis in empirical finance literature. We use word lists from Loughran and McDonald (2011) and Bodnaruk et al. (2015) to construct sentiment measures for firm disclosures. The Loughran-McDonald word lists (LM word lists) are developed from firms' annual reports to indicate the tones of the disclosure, thus more suitable for analyzing financial documents than alternatives such as the Harvard IV dictionary. Although this methodology has been applied to analyze a variety of popular financial documents, this paper is the first to examine SEC filings related to the entire process of going public via SPAC.

The rest of the paper is organized as follows: Section 2 describes institutional background about the two going-public methods and their related disclosures. Section 3 explains our hypotheses and their rationale. Section 4 provides details of the sample and variable construction and the peer firm matching strategy. Section 5 discusses the empirical findings and hypotheses testing results. Section 6 concludes with a policy recommendation.

## 2 Institutional Background

We focus on two going-public methods in the U.S.: conventional IPO and deSPAC merger. In the conventional IPO, the issuing firm raises capital from public investors through a well-established process existed for over 40 years. The issuing firm employs underwriters, who conduct due diligence and prepare materials for the regulator and investors, and facilitate institutional sales. In contrast, firms going public through deSPAC merger do not directly raise capital from public investors. Instead, the deSPAC merger target, a private operating firm who intends to go public, merges with a publicly listed shell company (i.e., SPAC) and thus receives cash infusion and consequently become public. It is worth nothing that the public shell company goes public and raises capital through the conventional IPO process. In this section, we compare the two going-public methods and describe the differences in the registration process and related SEC filings that we use to construct quantitative measures using textual analysis.

### 2.1 Conventional IPO Process and Related Disclosure

To conduct an initial public offering (IPO), firms must publicly file a registration statement (Form S-1 or S-1) with the SEC. Under the JOBS Act of 2012, IPO firms can initiate a confidential review process with the SEC by submitting a draft registration statement (Form DRS or DRS) before the S-1, through which the firm receives SEC feedback while not disclosing to the public. During 2013-2020, 86% of IPO firms confidentially filed for IPO, making it a new standard for going public in the post-JOBS Act era. Therefore, DRS is the first document that contains ample information regarding the private firm that intends to go public.

After SEC’s confidential review, IPO firm publicly files S-1, which includes a detailed description of its properties, business, security being offered and risks involved, among other matters.<sup>5</sup> The content change from DRS to S-1 reflects information production facilitated by the regulator and contains valuable information about the intrinsic value of the business (Zhu, 2021). Given that our entire sample falls into the post-JOBS Act era, we consider

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<sup>5</sup>DRS becomes publicly available simultaneously at this point.

both DRS and S-1 as the main documents when examining information content of the disclosures for conventional IPO process.

## **2.2 SPAC IPO, deSPAC Merger, and Related Disclosure**

The deSPAC method is structured as a merger between a SPAC and an operating company. SPACs are publicly listed shell companies, whose purpose is to merge with private operating companies such that the resulting entity becomes publicly traded. There are three critical events in a SPAC's lifecycle: the SPAC IPO, deSPAC target announcement, and the deSPAC event. The SPAC IPO, the IPO stage of the shell company, follows the conventional IPO process, where the SPAC files a draft registration statement (DRS) to initiate a confidential review, then publicly files S-1 to formally launch the IPO process.

After the IPO, SPACs typically have up to two years to identify a merger target or otherwise will liquidate. Among deSPAC transactions completed in 2015-2020, the average time from SPAC IPO to target announcement is 11 months. When the SPAC and deSPAC target(s) enter into a definitive agreement for their potential merger, a Form 8-K must be filed with the SEC to notify material changes of the shell company, which can be used to identify target announcement events.

The lag between target announcement and the deSPAC vary in lengths and last several months. During this time, three groups of stakeholders are involved in frequent negotiations: the SPAC sponsors, the target firm, and PIPE (Private Investment in Public Equity) investors who provide additional capital. As such, details of the merger agreement are subject to substantial revisions after the initial merger announcement. Before the deSPAC merger officially completes, the shell company holds a shareholder meeting, where several proposals are voted on, including the transaction itself and governance structure and executive compensations of the post-merger entity. The final proxy statement (Form DEF 14A or DEF 14A) prior to this shareholder meeting provides the most comprehensive disclosure about the merger transaction and the post-merger entity. Since DEF 14A is the last disclosure before the deSPAC transaction completes, we consider this final proxy statement as the counterpart to the formal prospectus (S-1) in conventional IPO process.

In summary, we investigate and compare the information content of disclosures related to the two methods of going public, respectively. For the conventional IPO method, we construct textual measures with DRS and S-1; for the deSPAC merger, we use DRS, S-1, and DEF 14A. Figure 1 illustrates the major steps in the two going-public methods.

### 2.3 Regulatory Oversight in the Going-public Process

As detailed in Section 1, transactional filings related to the deSPAC merger are only subject to SEC review on a selective basis, whereas IPO registration statements are subject to full cover-to-cover review. In March 2020, a public statement issued by the SEC states that “de-SPAC merger filings are potentially subject to review by the SEC and companies may receive comment letters in a similar fashion to a traditional IPO.”<sup>6</sup> To measure the extent of regulator involvement in deSPAC transactions, we inspect SEC comment letter quantity, as each comment letter indicates one iteration of SEC review.

We collect SEC comment letters for both the conventional IPO and deSPAC firms in 2015-2020. The comment letter sample is constructed as the following. First, we download all “UPLOAD” filings (in PDF format) associated with Central Index Keys (CIKs) of IPO, SPAC, and deSPAC entities from SEC’s Electronic Data Gathering, Analysis, and Retrieval (EDGAR) system. The PDF documents are parsed programmatically, from which we extract the firm filings that the comment letter is pertinent to from the letter titles: for example, Form S-1, Form 10-K, etc. Then, the comment letters are screened for their relevance to the going-public process. Specifically, for conventional IPOs, we keep comment letters issued for Form S-1 and DRS. For deSPAC firms, we keep comment letters issued for Proxy Statement, Proxy Materials, Schedule 14, PREM14, and PRER14. We compare the comment letter filing date to IPO date, deSPAC announcement date, and deSPAC merger completion date, to ensure that the comment letters are not issued for irrelevant transactions with similar filings.<sup>7</sup> Lastly, we exclude clarification letters that indicate SEC review has not been conducted, which contain the verbatim “we have not reviewed and will

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<sup>6</sup>See [sec.gov/news/public-statement/munter-spac-20200331](https://www.sec.gov/news/public-statement/munter-spac-20200331)

<sup>7</sup>For example, a SPAC may file a proxy statement to request an extension of their deSPAC target search; an IPO firm may file Form S-1 to issue shares upon exercise of warrants. Comment letters issued for such filings are excluded.

not review.”

On average, comment letter quantities show that there is significantly less regulator involvement in deSPAC transactions compared to conventional IPOs, despite the recently heightened attention to SPAC activities by the SEC. Almost all conventional IPOs received at least one comment letter, with an average of 3.5 letters per firm, similar to the average of 3.8 in [Lowry et al. \(2020\)](#) based on an earlier sample period.<sup>8</sup> On the other hand, only 48% of the deSPAC transactions have received any comment letter. Among the deSPAC firms who received comment letters, the average number of letters is 3.2, which is 0.3 less than the IPO sample average. Figure 2 shows the distribution of number of SEC comment letters for IPO and deSPAC firms, respectively. These statistics corroborate the common belief and our conjecture that there is less regulator oversight in deSPAC merger compared to conventional IPOs.

### 3 Regulatory Arbitrage and Hypotheses Development

Regulatory arbitrage refers to attempts to exploit loopholes and gaps in regulatory regime in order to reduce regulatory burdens. In the going-public context, private firms have strong incentives to avoid disclosing private information during the going-public process, especially for unfavorable information. The substantial less number of SEC comment letters received in deSPAC merger is one strong indication of potential regulatory arbitrage. Therefore, we conjecture that the content of firm disclosures is less informative in deSPAC merger, due to the separation of the IPO entity and the operating entity. We compare the information production process in conventional IPO and deSPAC merger by focusing on the required SEC filings. We develop our specific hypotheses regarding different stages of the going-public process in this section.

Regardless of which method is used to go public, firms must submit associated filings to fulfill disclosure requirements. Therefore, SEC’s scrutiny, or the interactions between the firm and the SEC during the registration process can be viewed as a regulation-driven

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<sup>8</sup>Five of the IPO sample firms do not have a relevant comment letter based on our algorithm. However, upon manual inspection, they either have response letters or a confidential treatment order. In other words, they likely received a comment letter but the letters are not yet publicly available.

information production process. Recent studies have shown that the SEC review process produces valuable information relevant to the pricing strategy of the offering, the underpricing phenomenon, firm's fundamental risk, the rationale of withdrawal decision, as well as the post-IPO stock and operating performances (see, [Lowry et al., 2020](#); [Zhu, 2021](#)). As discussed in the previous section, the distinct feature of the deSPAC merger method is the separation of the IPO entity and the operating business. Hence, deSPAC method and conventional method are not subject to unified level of regulator scrutiny, although both of them are regulated by the SEC.

Specifically, during the confidential review stage, the information content of the disclosure is dramatically different in the deSPAC method, although same types of filings are submitted to the SEC (i.e., DRS and S-1). When a shell company registers for IPO, there is no business operation to be included in the DRS and S-1.<sup>9</sup> By the nature of SPACs, SEC's confidential review is unable to inspect the business of future operating firm, to be funded by public investors, under the deSPAC method. Therefore, one should expect the confidential review to be less relevant in terms of information production; and the additional information produced, if any, to be less informative. Naturally, the confidential review period should be shorter, since there is no business operation to be reviewed. SPAC IPO disclosures mainly discuss the security structure, procedures related to share redemption and warrant conversion, and SPAC sponsors. During the confidential review, SEC is likely to require additional explanations to make the disclosure more readable and understandable to regulator investors (see, [Securities and of Investor Education, 1998](#)). As such, we hypothesize:

- H1. *In deSPAC merger method, SEC's review produces less information and lasts shorter during the IPO stage.*
- H2. *In deSPAC merger method, SEC's review focuses more on legal matters during the IPO stage.*

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<sup>9</sup>Besides, identifying potential targets at this stage is not permitted in SPAC IPOs.

Besides the regulation-driven information production process, reflected by the changes in information content from DRS to S-1, the literature has shown that the word content in the formal S-1 also contains valuable information. [Hanley \(1993\)](#) focuses on the proposed offer price range in the prospectus (S-1) and documents the “partial adjustment” in firm’s valuation. [Hanley and Hoberg \(2010\)](#) decomposes information in S-1 into standard and informative components. [Loughran and McDonald \(2013\)](#) links the tone of S-1 to offer price adjustment, first-day returns and post-IPO return volatility. Unlike in conventional IPO of an operating company, the information content in the shell company’s prospectus concentrates on the security structure and participants’ rights and responsibilities. Intuitively, one can expect ample discussions on legal matters in DRS and S-1. Although there is no business operation in the shell company, the issuer commonly includes its targeted industries or areas of expertise to provide a scope of the future merger. To attract public investors, SPACs must distinguish itself from hundreds of other shell companies. Hence, we expect that, the narrative in a SPAC IPO prospectus is confident and optimistic, such that its future value appears more promising to public investors. This implies:

- H3. *In deSPAC merger method, the prospectus contains less information and focuses more on legal matters during the IPO stage.*
- H4. *In deSPAC merger method, the tone of prospectus is less pessimistic and more confident during the IPO stage.*

In the deSPAC merger method, the operating company attempting to go public only comes into play at target announcement. Upon target announcement, information about business operation will be included in the required public disclosures. As discussed in previous section, we focus our analysis on the last proxy statement before an approved merger (i.e., DEF 14A). Our main conjecture is that deSPAC method is regulatory arbitrage whereby firms lessen disclosure burden. Without SEC’s confidential review and comments, firms are less likely to be pressured to disclose unfavorable information. Hence, we expect that the tone of DEF 14A is less pessimistic and more confident, comparing to conventional IPOs, which go through SEC’s review process. Moreover, we expect DEF 14A to be less

informative in both qualitative and quantitative aspects, because of the lack of oversight. This implies:

- H5. *In deSPAC merger method, the disclosure related to business operation contains less information during the deSPAC stage.*
- H6. *In deSPAC merger method, the disclosure related to business operation is less pessimistic and more confident during the deSPAC stage.*

So far, we have developed several hypotheses to test our conjecture of regulatory arbitrage under the deSPAC merger method. The aforementioned arguments focus on the disclosure filings during the IPO stage and prior to deSPAC completion. Furthermore, we extend the analyses to the post-public period.

Once the deSPAC merger completes, the operating business becomes public and is then subject to the same compliance requirements as other public firms, regardless the going-public method. This creates a unique setting where we can validate our empirical strategy when regulations are standardized and regulatory arbitrage no longer exists. If the previous comparison between the matched deSPAC merger and conventional IPO samples indicates regulatory arbitrage, we expect that the differences will greatly reduce or disappear altogether after completion of the going-public process.

The most important disclosure that all public firms must submit is the annual report (Form 10-K). In 10-K, firms provide a comprehensive overview of the business operation and financial condition as well as audited financial statements. Numerous research has used 10-K as the source of information content to measure firm's intrinsic risk (see, e.g. [Loughran and McDonald, 2011](#); [Hoberg and Maksimovic, 2015](#); [Bodnaruk et al., 2015](#)). Since there is no differential regulations after going public, we expect minimal or no significant differences in the tone of 10-K between firms between different going-public methods. This implies:

- H7. *There is little or no differences in the standardized post-public disclosures between firms that have gone public using different methods.*

Finally, we consider the consequences of potential regulatory arbitrage. Lessened disclosure burden has its costs and benefits. On one hand, less disclosure obstructs information



transparency and may harm investors. On the other hand, it makes going-public more accessible and may facilitate capital formation as a result. Both investor protection and capital formation facilitation fall into the objectives of the SEC. Because our empirical strategy matches deSPAC and conventional IPO firms based on their fundamentals, we control for the ability to raise capital. Hence, we specifically focus on the investor protection aspect to examine the implications of possible regulatory arbitrage.

After going public, firms continue to provide information to the public on an ongoing basis. As more information becomes available to the market, differences between expectations or “promises” made before going public and actual performance may emerge. As hypothesized previously, without stringent regulator inspection of the going-public filings, issuers have incentive to avoid unfavorable content and languages. The federal securities laws provide investors means to recover their losses if there are any wrong-doings in the public firm, including fraud, material omissions, and manipulations in stock trading activities. We expect that deSPAC firms has higher likelihood to face securities class action lawsuit because of the understated risk and over-confident projections made prior to going public. Furthermore, stock value should drop more if actual firm performance falls short of expectations by a larger degree. We hypothesize:

- H8. *In deSPAC merger method, firms face higher post-public litigation risk.*
- H9. *In deSPAC merger method, firms experience larger share value decline (or smaller share value growth).*

## 4 Data and Methodology

Our main sample consists of firms that went public via the conventional IPO or deSPAC method between 2015 and 2020. Our main data sources include Thomson Reuters SDC Platinum, FactSet, Compustat, SEC filings, CRSP, Crunchbase, VentureXpert, and Stanford Securities Class Action Clearinghouse. Hereafter, we refer to firms that went public via conventional IPOs as the “IPO sample”, firms that went public via deSPAC method as the “deSPAC sample”, and both combined as the “aggregate sample”. In the comparative

analyses, we use propensity-score-matching to construct paired deSPAC and IPO samples matched on pre-public fundamentals to mitigate selection bias.

## 4.1 Sample Construction and Summary Statistics

The IPO sample is constructed from SDC New Issues Database and contains 573 firms that completed their IPOs after going through the SEC’s confidential review process during 2015-2020. Following the IPO literature, a standard screening process is applied. Specifically, IPOs with an offer price of less than \$5 per share, financial firms (i.e., banks, savings, and loans), American Depository Receipts (ADRs), unit or rights offers, real estate investment trusts (REITs), natural resource limited partnerships and closed-end funds are excluded.

For the sample of companies going public via deSPAC merger, we first identify the universe of SPACs listed on U.S. exchanges from using SDC New Issues Database and FactSet. We then use the SDC M&A Database, FactSet and SEC filings to determine its deSPAC transaction status and the merger target identity. In 2003-2020, there are 633 SPAC IPOs. Among them, 254 completed the deSPAC transaction by the end of 2020. We focus on the subset of deSPAC mergers completed in 2015-2020, with the similar screening as in IPO sample. This sample construction process results in 573 firms that went public through the conventional IPO method (including a SEC confidential review process) and 116 firms went public via deSPAC merger in 2015-2020. Figure 3 summarizes the yearly count of operating firms going public through the conventional IPO and deSPAC merger method between 2015 and 2020.

## 4.2 Textual Analysis Measures

We collect disclosures related to the two methods of going public: DRS, S-1, and 10-K filings for the IPO sample and DRS, S-1, DEF 14A, and 10-K for the SPAC sample. The full-text version filings are downloaded by specifying “Form Type” and “Central Index Key (CIK)” criteria on EDGAR. Additional screening is applied for the SPAC sample to ensure relevant disclosures are retrieved. Specifically, the DRS and S-1 filings are filed for the shell

company that successfully went public. The targeted DEF 14A filings are the last proxy statement submitted before a successful merger occurred, while the targeted 10-K filings are the first annual report filed by the combined entity after merger completed.

This study primarily uses seven categories of word lists as its measures of tone in the firm disclosures, while it uses the changes in the proportion of word lists as the measure of the changes in content during the confidential review process. In recent years, the implementation of textual analysis in finance has become one of the most important techniques in both academic research and industry practices.<sup>10</sup> Among word lists applied in the literature, we use the sets of word lists developed in [Loughran and McDonald \(2011, 2013, 2016\)](#) to measure tone in the DRS, S-1, and 10-K, as well as the change in tone between DRS and S-1. Unlike alternative word lists, the Loughran-McDonald word lists (LM word lists) are constructed specifically using firms' annual reports, making the word lists perfectly suitable for gauging sentiment in mandatory financial disclosures.<sup>11</sup> With the addition of the Constraining word list introduced in [Bodnaruk et al. \(2015\)](#), we select six categories of LM word lists: Litigious, Strong Modal, Constraining, Positive, Negative, and Uncertain. It is worth noting that there is a high correlation between Uncertain and Negative word lists. To address this concern, consistent with [Loughran and McDonald \(2011, 2013, 2016\)](#), an additional Risky word list is constructed as an aggregate measure of uncertainty and negative words.<sup>12</sup>

Each category of word lists serves as a lexicon-based dictionary to measure the frequency of specific group of words that appear in the content of interest. Since the nature of the measure is objective, the frequency of word lists is not affected by potential misinterpretation. The proportion of each of the seven word lists appeared in the disclosure is measured in percentage scaled by the number of words in the target document: %Litigious, %StrongModal, %Constraining, %Positive, %Negative, %Uncertain, and %Risky. Similarly, we construct the percentage changes for each word list to capture the changes in

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<sup>10</sup>See [Loughran and McDonald \(2020\)](#) for an updated review of the literature.

<sup>11</sup>[Hu and Liu \(2004\)](#) creates a word list for sentiment analysis in social media. [Chen et al. \(2019\)](#) creates a crypto-specific lexicon including emojis, slang, and profanity.

<sup>12</sup>We decide not to use Moderate Modal and Weak Modal word list due to the ambiguous definition relative to Strong Modal.

content during the confidential review process as in [Zhu \(2021\)](#). Specifically, we calculate the percentage changes ( $\Delta\%$ ) in word list proportions as the percentage changes in the proportion of each LM word list from DRS to the corresponding S-1 filings. See Appendix A for the definitions and examples of words included in each LM word lists used.

We report summary statistics for the textual analysis measures for the IPO sample and SPAC sample in Table 1. Panel A reports the levels of textual analysis measures in S-1, while Panel B shows the changes in textual analysis measures during SEC’s confidential review process.<sup>13</sup> For the IPO sample, on average, S-1 has about 3500 more LM words than in the SPAC sample. For both samples, the proportions of each LM word list in the DRS and S-1 are at the level of 0.5%-1.8%, very similar to the level (i.e., 0.5%-1.4%) reported in [Loughran and McDonald \(2013\)](#). However, there are notable variations across the two samples. For example, SPAC sample has relatively higher level of litigious words and strong modal words, while less negative and uncertain words. It seems, on average, SPAC firms’ disclosures are less pessimistic and more involved with legal matters. Panel B shows similar results that almost all LM word list proportions decrease after SEC’s scrutiny. However, the magnitudes for SPAC firms are substantially lower than IPO firms, indicating a less effective review process by the SEC. Although those unconditional comparisons provide evidence in support of our conjecture, we conduct a comprehensive comparisons using propensity-score-matched sample to avoid contamination due to potential selection bias.

### 4.3 Characteristics of Class Action Lawsuits

We focus on class action lawsuits as measure of litigation risk because they incur material costs to the defendant firm. A securities class action is a case brought on behalf of a group of persons who have purchased the securities of a particular firm that may have violated one or more federal or state securities laws. Class action lawsuits consolidate allegations by a large number of individual plaintiffs. Hence, the financial penalty and settlements can potentially be so enormous that firm managements must relocate tremendous amount of time, energy, and financial resources away from business operation to defend the case. This

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<sup>13</sup>There are 35 SPACs (including 9 deSPAC firms) did not go through the optional confidential review process, which are excluded in this comparison.

disruptive event can also cause indirect costs for the firm, such as damages on reputation, loss of customers and suppliers, and failures on growth opportunities.

After the aggregate sample is constructed, we search the Stanford Securities Class Action Clearinghouse (SCAC) database for cases that were filed against firms in our samples. We use CIK as unique identifier and verify it by checking company names and tickers. Our lawsuit sample consists of all firms that completed an IPO between 2015 and 2020, and were sued for violations of securities laws. We report the summary statistics of the lawsuits in Table 2. We find that 112 (19.5%) of the 573 firms that went public using conventional method were sued. For the SPAC sample, none of the 303 shell companies that have not complete merger were sued. This is consistent with our expectation and the fact that shell companies share values are guaranteed by the redemption option. For the 116 SPACs who completed merger, or deSPAC firms, we find 23 (20%) of them were sued. These proportions suggest similar litigation risk under different going-public methods. However, it is worth noting that this is an unconditional comparison where firm's fundamental risks and characteristics are ignored. In the propensity-score-matched comparison in later section, we show deSPAC firms face significantly higher litigation risk. In addition, we want to emphasize that the long preparation time in securities class actions may cause an underestimation of the litigation risk for deSPAC firms. Since most of our deSPAC firms have completed a merger in the past two years, we expect that more cases will be filed. The underestimation of litigation risk for deSPAC firms will make our results more conservative, and thus, more robust when longer horizon is available for our analysis. We investigate the impact of choosing different going-public methods on the probability of having class action suit in more details in Section 5.

Panel B of Table 2 shows the distribution of allegations of the lawsuits, specifically with respect to Section 11 of Securities Acts of 1933. Lawsuits under Section 11 generally argue that shareholders have suffered significant losses due to material untruths or omissions in the IPO prospectus. For all 112 cases against conventional IPOs, 46 (41.1%) of them were sued under Section 11, while only 2 (8.7%) of the 23 cases were sued against deSPAC firms under Section 11. This considerable difference in proportion is consistent with the fact

that, the lack of information about business operation in a SPAC IPO, provides very little basis for identifying untruths or omissions in the prospectus.

Panel C of Table 2 reports the distribution of case status as of data collection. Consistent with the reasons discussed earlier, about 61% of cases against deSPAC firms are ongoing. The only case that was settled with deSPAC firm is a case against Akazoo S.A.. The case was filed on April 24, 2020, and settled with \$35 millions on September 10, 2021. Although this is only one observation regarding deSPAC IPOs, it provides “a noteworthy data point for the discussion about SPAC-related litigation exposures.”<sup>14</sup> It is worth noting that the magnitude of litigation cost can be huge, even though data availability constraints prevent us from quantifying the cost of lawsuits. Hence, we focus on comparing the likelihood of being sued between different going-public methods in Section 5.

#### 4.4 Identifying Peer Firms

This paper asks the question of whether the alternative going-public methods have similar regulatory burden and investor protection effectiveness, given the existing legislatures. Our empirical approach is to test the nine hypotheses developed in Section 3 by comparing the information content of related disclosures and post-public outcomes in each going-public method.

A confounding factor is that different types of firms naturally differ in fundamental risks and characteristics. For instance, a high-growth software company in its early stages has different risk profile compared to, say, a family-run food manufacturer that has a long operating history. It is possible that each going-public method attracts clusters firms with similar characteristics that are correlated with the outcomes of interest. Anecdotally, dozens of electric vehicles and LiDAR technology related firms went public via deSPAC merger instead of through conventional IPO in the past two years. This is possibly driven by overlapping networks of financiers who advised these transactions. To control for potential selection bias in the going public method, we use a propensity-score-matching approach, which is implemented in two stages. First, we estimate a probit model where the dependent

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<sup>14</sup>See [dandodiary.com/2021/04/articles/securities-litigation](https://dandodiary.com/2021/04/articles/securities-litigation)

variable indicates whether a firm went public via deSPAC merger, conditional on pre-public firm characteristics. Second, for each deSPAC firm, we identify its matched peer firms from the IPO sample based on the predicted values from the probit model (i.e., propensity scores).

We estimate the following probit model to predict whether a firm chooses to public via deSPAC merger,

$$\text{Going Public via SPAC}_i = \beta_0 + \text{FirmCharacteristics}_i \beta_1 + \text{GoingPublicYearDummy}_i \alpha_1 \quad (1)$$

The firm characteristic variables include financial statistics in the last full fiscal year prior to going public, firm age, and VC backing status. Financial statistics are selected to represent firm size, liquidity, profitability, R&D and investment intensity. These features are selected as they are considered relevant in several models on the going-public choice (see, [Chemmanur and Fulghieri, 1999](#); [Maksimovic and Pichler, 2001](#); [Spiegel and Tookes, 2007](#)). The data for variables construction are drawn from multiple sources, including Compustat, VentureXpert, Crunchbase, the first post-public 10-K filing. Table 3 shows the definition of the firm characteristic variables.

The probit regression estimates suggest that, compared to the IPO firms, the deSPAC firms are younger, have received VC financing, have lower current ratios, and not in a high-tech industry, particularly biotechnology. Table 4 summarizes the regression estimates. Intuitively, these dimensions could affect the disclosure, litigation risk, and stock performance measures that we compare between the deSPAC and IPO samples. Using a propensity-score-matched sample helps alleviate the selection bias.

Then, to create the matched peer firm sample, we use the estimated propensity scores to identify the five closest matches. The average propensity score for the closest matched IPO sample (0.316) is close to that of the deSPAC sample (0.359). After identifying the five closest propensity-score-matched conventional IPO peers, we create two synthetic peers as “Average Peer” and “Median Peer”. The values of all textual analysis measures and other variables are calculated based on the average and median values of the corresponding variables for “Average Peer” and “Median Peer”, respectively. Along with the “Closest

Peer”, we have three comparable peers for each deSPAC firms in our sample. This matching and constructing process results a sample of 107 deSPAC-IPO-Peer pairs for each of the three peer types.

## 5 Discussion of Empirical Results

### 5.1 Comparison During IPO Stage: SPAC vs. non-SPAC

#### 5.1.1 Changes between DRS and S-1 during SEC Confidential Review

First, we discuss the empirical results related to hypotheses H1 and H2. We compare the information production during the confidential review process during an IPO (i.e. SPAC IPOs and non-SPAC IPOs, corresponding to the deSPAC and conventional method respectively). We find evidence supporting our hypothesis H1 and H2: in deSPAC method, SEC’s confidential review process produces less information, lasts shorter and focuses more on legal matters.

In the conventional method of going public, most firms initiate a confidential review process with SEC, where SEC provides feedback, raises concerns, and requests revisions regarding the draft prospectus.<sup>15</sup> Hence, this regulator-driven information production process serves the important function of ensuring information transparency and enforcing disclosure requirements. In the deSPAC method, the IPO of the shell company, or the SPAC IPO, goes through the same review process as the conventional method.

Panel A of Table 5 compares the length of the confidential review and shows that on average, SPACs spend about 43 days less than conventional IPOs and 22 days faster to complete the IPO after the review. In total, the shell company takes an average of 107 days to complete the IPO, while conventional IPOs of an operating business needs 65 days more to be listed. The results support our hypothesis H1 that, by design, SPACs circumvent the SEC’s scrutiny at the initial stage of raising public capital by filing as a shell company. This results in a much faster time to complete an IPO. When comparing

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<sup>15</sup>In [Zhu \(2021\)](#), SEC confidential review process is defined as the period from submitting draft registration statement (DRS) to submitting formal prospectus (S-1).



the IPO registration process for all the SPAC firms in Panel B, we find that there is no significant difference between SPAC firms that have completed deSPAC merger and those have not. This suggests that the SEC review process generally does not cover any activity regarding searching for potential target. This is not surprising as SEC's review has little impact on the final outcome of the shell companies' merger activities at this stage. Panel C reports a comparison of entire going public process by different methods used. Clearly, at the IPO stage, using deSPAC merger shortens the period by about 53 days. However, when adding the lengthy deSPAC period, the entire going-public process becomes considerable long. This may not be a concern for the private operating firm, but for the investors who have invested in the shell company, it raises potential uncertainty for their returns.

We further test hypothesis H2 by comparing the outcome of the SEC review process. The literature suggests that the confidential review enables improved transparency to the public market. We want to investigate if this still applies in the case of shell companies. Recall that the SEC may provide comment and request clarifications and additional information throughout the review. If shell companies produce similar DRS and S-1, then the confidential review process does not lead to better information environment for the public and hence is a less productive activity of the SEC.

In Panel D of Table 5, we report the comparison of changes in content during the confidential review process (i.e. between DRS and S-1) for the propensity-score-matched sample. During the confidential review, there is an increase of word content, measured by total number of LM financial words. On average, the total number of LM words increased by 7.6% for deSPAC firms. This increase is significantly lower than their matched peers. Specifically, Closest Peers add 13.8% more word content, while 10.4% and 11.8% for Median Peers and Average Peers, respectively. The differences of about 6% of word content is statistically significant at 5% level. This is consistent with hypothesis H1 that less additional information is produced during SEC confidential review process for shell companies.

Another interesting finding is that the changes in %Litigious is very different for shell companies and their matched peers. After SEC's scrutiny, the proportion of litigious words

increased by 2.2%, instead of decreasing for their matched peers. For instance, the decreases of litigious word proportions range from 1.9% to 2.7% among Closest Peers, Median Peers and Average Peers. Clearly, this opposite change for deSPAC firms confirms our expectation in hypothesis H2. Given the complex legal structure of SPACs, the disclosure mainly describes the legal structure of the entity and its securities, such as share conversion and redemption procedure at merger or liquidation. As such, SEC’s review focuses on legal jargons and requesting more explanations on those related procedures. As stated in the SEC’s guide on how to create clear SEC disclosure documents, it outlines SEC’s effort to reduce “the most common problems in disclosure documents,” such as “passive voice,” “legal and financial jargon,” and “abstract words”. This is consistent with our observation in Panel D. Moreover, this shows that SEC’s review is still able to induce valuable information in the case of shell companies, although limited information has been provided.

### 5.1.2 Sentiment Levels in S-1

Following prior research on IPO prospectus, we also compare the information content of the prospectus, as measured by sentiment levels. The objective is to investigate whether the sentiment levels in the prospectus of shell and operating businesses vary, and thus, testing our hypotheses H3 and H4. Since all firms in our comparison filed a draft version (i.e., DRS) and a formal version (i.e., S-1) of prospectus during the IPO period, regardless of going public method. We report the comparison between samples using DRS and S-1 in Panel A and B of Table 6, respectively. To be consistent with literature, we mainly discuss our findings with S-1 in this section.<sup>16</sup>

First, we find evidence supporting hypothesis H3 that the shell company prospectus contains less information and focuses more on legal matters. The total number of LM financial words in S-1 is significantly smaller than any of the three matched peer groups. In addition, the number of unique LM financial words is at least 1,000 fewer than any of the three matched peer groups. Specifically, there are 9,332 fewer LM financial words (or 1,084 fewer unique LM financial words) than the Closest Peer. Regardless of content quality, shell

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<sup>16</sup>The results for DRS are almost the same as S-1.

company prospectus contains much less information than operating companies. Although the amount of numbers used and the raw file size in SPAC IPOs S-1 is not statistically different from non-SPAC IPOs, they are smaller in the comparison of DRS. It appears that shell companies tend to initially provide an even simpler disclosure in their draft (DRS), which may reflect the true intention of the issuer. SEC’s review, however, leads to adding much more numbers in shell company’s S-1. Overall, both DRS and S-1 of shell companies show a much less informative prospectus. As expected, the proportion of litigious words are at least 12 basis points more for SPACs. This is consistent with our discussions in SEC review process, which shows the majority of content in prospectus is legal matters.

Secondly, the level of pessimistic sentiment and confidence are significantly different between SPAC and non-SPAC IPOs. In both DRS and S-1, the proportion of strong modal words is around 1.1%, which is about 50% higher than any of the matched peers (0.5%). Examples of strong-modal words include “always”, “definitely”, and “never”. In the literature, %Strong-modal in prospectus is often used as an ex-ante measure of managerial hubris. The interpretation of the large amount of strong modal words is that using overconfident language may be perceived as overselling the firm’s potential. Intuitively, shell companies have the incentive to be optimistic and confident to attract investors. Similarly, the tone of shell companies’ S-1 is relatively less pessimistic. For example, %Risky for deSPAC firm (2.8%) is 32 basis point smaller than Closest Peer’s (3.15%).

## **5.2 Comparison During deSPAC Stage: deSPAC Merger vs. Conventional IPO**

So far, we have shown the information content of prospectus during IPO stage, comparing SPAC IPO and non-SPAC IPOs. The results do not reject hypotheses H1-H4. Next, we consider the deSPAC stage, where disclosures related the operating company becomes available in the deSPAC method. Our conjecture is that if deSPAC method is regulatory arbitrage, the disclosure related to business operation should be less informative, less pessimistic, and more confident, in the deSPAC stage.

We compare the disclosure related to business operation to test hypotheses H5 and

H6. We use the last proxy statement (DEF 14A) before deSPAC merger completion as the disclosure document for deSPAC firms. As described in Section 2, merger agreement is subject to frequent and significant changes between target announcement and the shareholder meeting. By using the last proxy statement, we avoid using information that was not relevant to the deSPAC transaction.<sup>17</sup> Then, the comparison is conducted between deSPAC firm’s DEF 14A and matched IPO peer’s S-1.

In Table 7, there are considerable differences across all variables of interest, except %Litigious. Overall, we see strong evidence supporting hypotheses H5 and H6. deSPAC firm’s business description is much simpler than conventional IPO’s S-1. Comparing to Closest Peer, the size of the disclosure is 75% smaller. Specifically, the total number of LM financial words are 72% less, while there are 2,436 unique LM financial words less in deSPAC firm’s business description. In contrast, the proportion of strong modal words is 58% higher than Closest Peer (0.51%). The overconfidence in business operation is also reflected in less pessimistic tone. %Risky of deSPAC firm’s business description is 1.8%, comparing to the 3.1% in Closest Peer or 3.2% in Median Peer and Average Peer. One potential reason for this striking difference is the lack of SEC’s systemic review of proxy statements. Unlike IPO process where SEC’s review and comments are mandatory, the proxy statement submitted after the shell company went public is subject to a “random review” as all other filings. As explained by SEC, the disclosures that are required for public firms are randomly selected to be reviewed by the SEC. There is no guarantee that the proxy statement will be selected for review given large pool of public disclosures. Therefore, deSPAC firms do not have as much obstructions in including overconfident business description in their proxy statements, which could encourage shareholders approval for the proposed merger.

Based on abovementioned results, we find strong evidence supporting our hypotheses H1-H6. We conclude that going public via SPAC is regulatory arbitrage. During the IPO stage, the shell company submits relatively simple disclosure to the SEC, where the SEC review is irrelevant to the future operating business. The prospectus contains less pessimistic and overconfident information to attract investors, resulting a relatively less costly process of raising external funding for the eventual merger. In addition, deSPAC

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<sup>17</sup>This screening requirements excluded 20 pairs in our sample.

firm’s proxy statement contains business descriptions that are overoptimistic compared to its IPO peers, in turn, increasing public investor interest.

### 5.3 Placebo Test When Unified Regulations Exist

To validate our methodology and reinforce the conclusion of regulatory arbitrage by deSPAC firms, we conduct a placebo test in this section. Once the merger completes, the combined entity is subject to the same regulations as any other public firms, where opportunity for regulatory arbitrage no longer exists. The hypothesis is that there should be little or no differences in standardized post-public disclosures between firms went public under different methods (i.e., hypothesis H7), given the similar firm characteristics of the matched samples.

The design of the placebo test is to compare the first annual report (10-K) filing after the operating firm goes public. Specifically, we compare the sentiments in the first 10-K filed as a public firm for conventional IPOs and the first 10-K filed as a combined entity for deSPAC firms. This is a unique setting because 10-K filings, as one of the most important corporate filings, have very detailed and standardized formatting. SEC has issued several guidance and specified required items to be included in a 10-K document.

In Table 8, we repeat the same comparison as in Table 7, but using 10-K. Firstly, it shows the word content in 10-K for deSPAC firms is still less than matched IPO peers and the difference is statistically significant. However, the differences are much smaller than using DEF 14A where regulatory arbitrage opportunity exists. For example, the total number of LM financial words is 14% less than Closest Peer when comparing 10-K, while the difference is 72% less in the deSPAC stage. Similarly, the difference in unique financial words is 238 in 10-K, much lower than the 2,436 in DEF 14A. This substantial improvements in word content indicates that in an regulatory environment where no arbitrage exists, deSPAC firms comply and provide disclosure with much more comprehensive information similar to their conventional IPOs peers matched on fundamentals.

Further, we find consistent results in sentiment levels of 10-K filings. %Strong-modal is heavily reduced from 0.8% in DEF 14A (or 1.1% in S-1) to 0.35% in 10-K. When comparing

to matched IPO peers, there is no significant differences in %Strong-modal across Closest Peer, Median Peer and Average Peer. This means the overconfident tone observed in S-1 and DEF 14A is no longer present in the standardized annual report. Comparison of measures of the negative and uncertain tone of also suggests convergence of the deSPAC and IPO samples. The proportion of risky words (%Risky) in deSPAC firm's 10-K is at a similar level (3.8%) as their closest matched IPOs (3.9%). It is worth noting that the %Risky in 10-K almost doubles from the level in DEF 14A. The large increase in negative and uncertain words can be viewed as a sign of truthfully disclosing underlying risks involved in business operation. Intuitively, firms provide less optimistic disclosure to hedge against the litigation risk, because of the detailed requirements in 10-K filings and the lack of arbitrage opportunity. Overall, the placebo test indicates that there is little or no differences in post-public disclosures between firms going public under different methods, due to standardized regulations and hence the lack of arbitrage opportunity.

#### **5.4 Consequences of Regulatory Arbitrage: Litigation Risk and Stock Return**

Finally, we investigate the investor protection implications of regulatory arbitrage in the going-public process. Thus far, the empirical results provide ample evidence to suggest that deSPAC firms take advantage of regulatory inconsistency in their public filings. Is such practice harmful to investors? We compare the litigation risk and stock return in the post-public period of the deSPAC and their IPO peers in this section.

One important role of regulator in the public market is to maintain fairness and information transparency. Circumventing SEC oversight leads to a higher degree of information asymmetry. However, this gap in regulation is alleviated after the going public process completes since identical regulations apply regardless of the going-public method in this period. As more information becomes available to establish the intrinsic value of a public firm, the expectation based on pre-public disclosures is confronted with the reality. When the pre-public information asymmetry is more severe (i.e. in the case of deSPAC firm), we hypothesize that there is a larger mismatch between the expectations and the reality. As a

result, there is a higher litigation risk for deSPAC firms than their matched IPO peers and their share prices experience larger decline (or smaller growth) compared to its IPO peers (i.e., hypotheses H8 and H9).

In Table 9 Panel A, we compare deSPAC firms and matched IPO peers with respect to the probability of being named defendant in securities class action lawsuits. Because it takes time for investors to seek legal actions, we report separate statistics based on the time since going public. Overall, we see a 9.5% percentage points higher in litigation risk for deSPAC firms comparing to their matched IPO peers. There is a 19.8% probability that a deSPAC firm is involved in a class action lawsuit, while the probability for the Closest Peer is only 10.3%. Figure 4 plots the litigation risk with 95% confidence intervals. Although the probability of being sued within 1 year is only significant when comparing to Median Peer, we find a monotonic increase in the probability in the number of years being public. As mentioned in Section 4, most of deSPAC firms went public in 2019 and 2020. We believe the increased litigation risk of deSPAC firms would be more prominent in future years.

Figure 5 reports the average buy-and-hold returns of deSPAC firms and their matched IPO peers in 30-day, 90-day, and 180-day after going public, with 95% confidence intervals. To adjust for market returns, we compute excess return relative to CRSP, which is defined as the buy-and-hold return minus that of CRSP Equal Weighted Index in matched investment period. The deSPAC sample returns are calculated based on the purchase price on the first deSPAC trading day. The IPO sample returns are calculated based on the average of the highest and the lowest trading prices on the first trading day.<sup>18</sup>

Newly publicly traded companies generally underperform the market benchmark (Ritter, 1991, see, e.g.), and the recent sample in this study demonstrates a similar pattern. Based on the point estimates, deSPAC firms on average generate better return in the initial month of public trading, but underperforms compared to the IPO peers thereafter.

The empirical results related to litigation risk and stock return do not reject our hypotheses, and suggest that the regulatory arbitrage causes real consequences from investors'

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<sup>18</sup>We use CRSP data through the end of 2020; if the event window extends into 2021, an observation is dropped from the comparison. If either data for the deSPAC firm or matched IPO are unavailable, the pair is dropped from the average calculation.

perspective. As investor protection is one of SEC’s missions, such evidence should prompt the regulator to take actions to eliminate the opportunities for regulatory arbitrage.

## 6 Policy Recommendation and Conclusion

In Section 5, we find that there is no difference in textual sentiments between alternative going-public methods when in the context of standardized filings (Table 8). To address the regulatory arbitrage opportunity deSPAC mergers, we call for a mandatory SEC review process that is similar that in a convention IPO to enhance the information transparency. To examine whether filings requirements affects the effectiveness of investor protection, we use a logit regression to estimate the likelihood of security class action lawsuits conditional on disclosure content (Table 9 Panel B). The rationale is that, the method of going public should not matter if equivalent disclosure is provided to the public.

In Table 9 Panel B, the dependent variable is a dummy variable indicating having a class action suit after being public. The key explanatory variable of interest is deSPAC dummy, which equals one if the going public method is through deSPAC merger. Each column shows the regression results from four types of key filings discussed in the paper, as well as the content changes during the SEC confidential review. From Columns 1-4, we find the deSPAC dummy is not statistically significant when controlling information content of DRS, S-1 and DEF 14A. This result shows that, when the disclosure sentiment is controlled for, the deSPAC method itself does not have a higher likelihood of causing investor dispute. In contrast, deSPAC is strongly associated with litigation risk when 10-K is used (Column 5). The interpretation is that given a standardized disclosure, going public via deSPAC contributes to litigation risk, likely due to the insufficient disclosure ex-ante.

We further investigate whether the regulator-induced change in disclosure during the IPO stage has any long-term effect on the post-public litigation risk. In column 5, the estimation shows that  $\Delta\%Litigious$  and  $\Delta\%Strong-modal$  is positively correlated to litigation risk. In other words, during the confidential review process, if more explanation related to legal matters are required by SEC, or if more definitive and overconfident tone is revealed after SEC’s review, there is a higher probability that the deSPAC entity is



sued after going public. This results suggest that regulator scrutiny facilitates information production that can predicts the long-term performance of a deSPAC transaction. This resonates with findings in the literature. [Mahoney \(1995\)](#) finds that external monitoring is necessary to prevent an upward-biased disclosure by issuers. In more-recent work, [Dechow et al. \(2016\)](#), [Ryans \(2020\)](#), and [Lowry et al. \(2020\)](#) document the importance of comment letters generated during the communications between the SEC and the firms. Overall, all results in Table 9 Panel B consistently point to a direction of removing differences in SEC review requirements in alternative going-public methods.

The SEC has closely monitored SPAC activities since 2020 and expressed the stance that different going-public methods should be subjected to identical regulatory obligations. Updated regulation is being considered and underway. However, as of early 2022, there still lacks a unified regulatory framework that oversees alternative ways of going public. Our findings suggest that, as the current regulations stand, the information content in deSPAC proxy statement is less rich and more overly optimistic than that in IPO registration statements of similar firms. Subsequently, deSPAC firms experience higher security class actions filing rate and worse stock return performance, compared to matched IPO firms. This paper provides empirical evidence in support of a regulatory update that will better align the requirements of different going-public methods so that investors are equally protected regardless of the mechanics in the public listing transaction.

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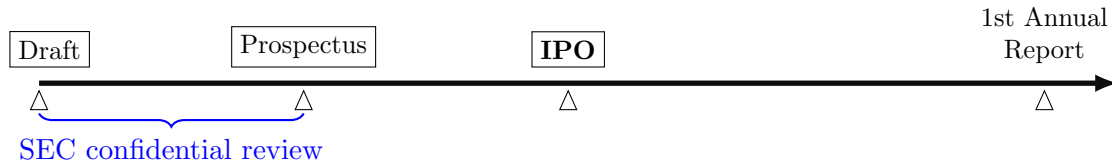
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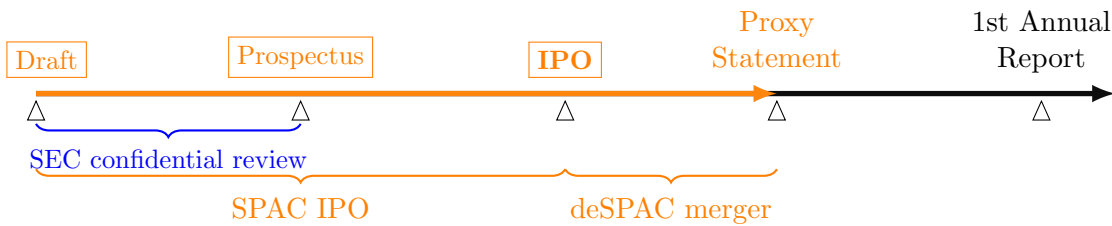
Figure 1: Going-public Methods and Related Disclosure

This figure describes the major steps in the two going-public methods, respectively. The top timeline describes a typical IPO registration process in the post-JOBS Act era. The bottom timeline describes going-public via deSPAC merger method. Black color indicates process involved with the operating firm that intends to go public; Orange color indicates process involved with the shell company (SPAC); Blue color indicates the confidential review process during the IPO period.

**Conventional IPO method:**



**deSPAC Merger method:**



- Black color: indicates process involved with the operating firm that intends to go public
- Orange color: indicates process involved with the shell company (SPAC)
- Blue color: indicates the confidential review process during the IPO period

Figure 2: Distribution of SEC Comment Letter Counts for Conventional IPOs and deSPAC Mergers, 2015-2020

Figure 2 reports histograms of the number of SEC comment letters received by Conventional IPOs and deSPAC firms in the going-public process. Description of the comment letter sample construction procedure is detailed in Section 2. Between 2015-2020, 48% of deSPAC transaction received at least one comment letter, while 99% conventional IPOs received at least one comment letter. Despite the recently heightened attention to SPAC activities by the SEC, the extent of regulator scrutiny is much less in a deSPAC transaction.

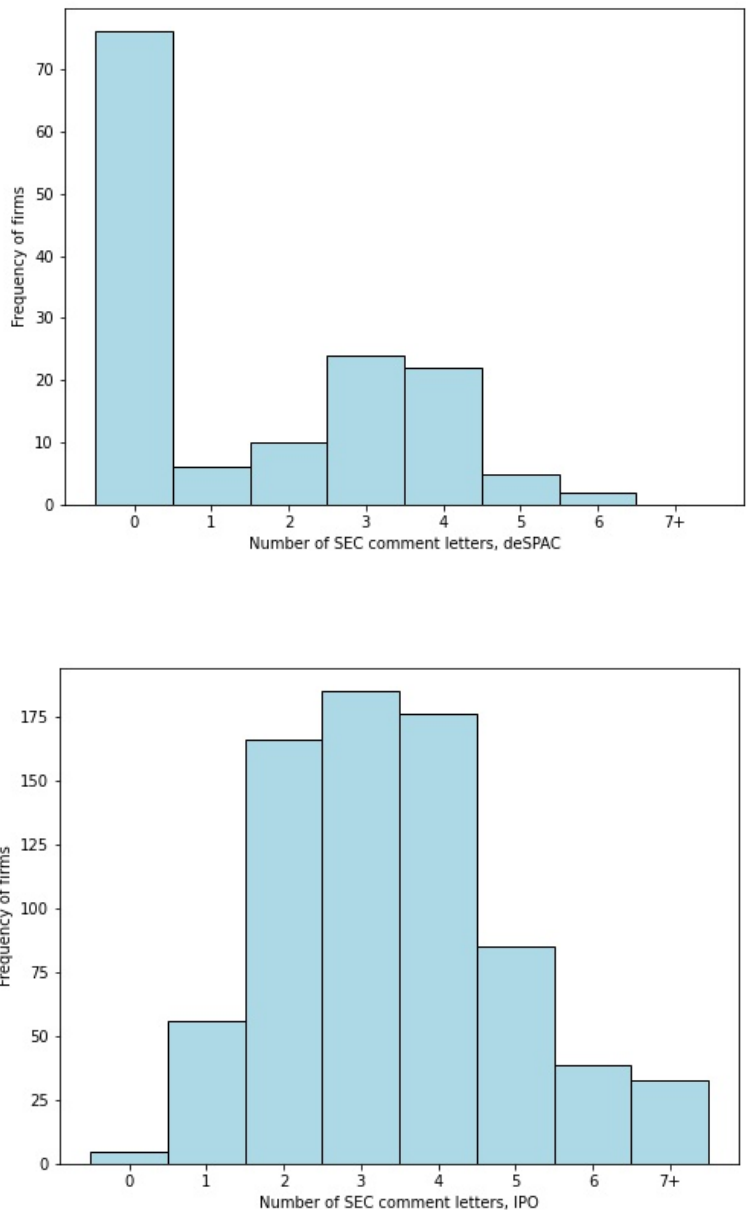


Figure 3: Number of Firms Going Public via Conventional IPOs and deSPAC Mergers, 2015-2020

Figure 3 reports the yearly count of operating firms going public in the U.S. through conventional IPO and deSPAC merger between 2015 and 2020. The sample includes companies traded on three major exchanges - NYSE, Nasdaq, or AMEX (NYSE American). The IPO sample excludes offerings with offer price below \$5, ADRs, units, REITs, closed-end funds, banks, ETFs, SPACs, direct listings, and stocks not included in CRSP such as OTC issues.

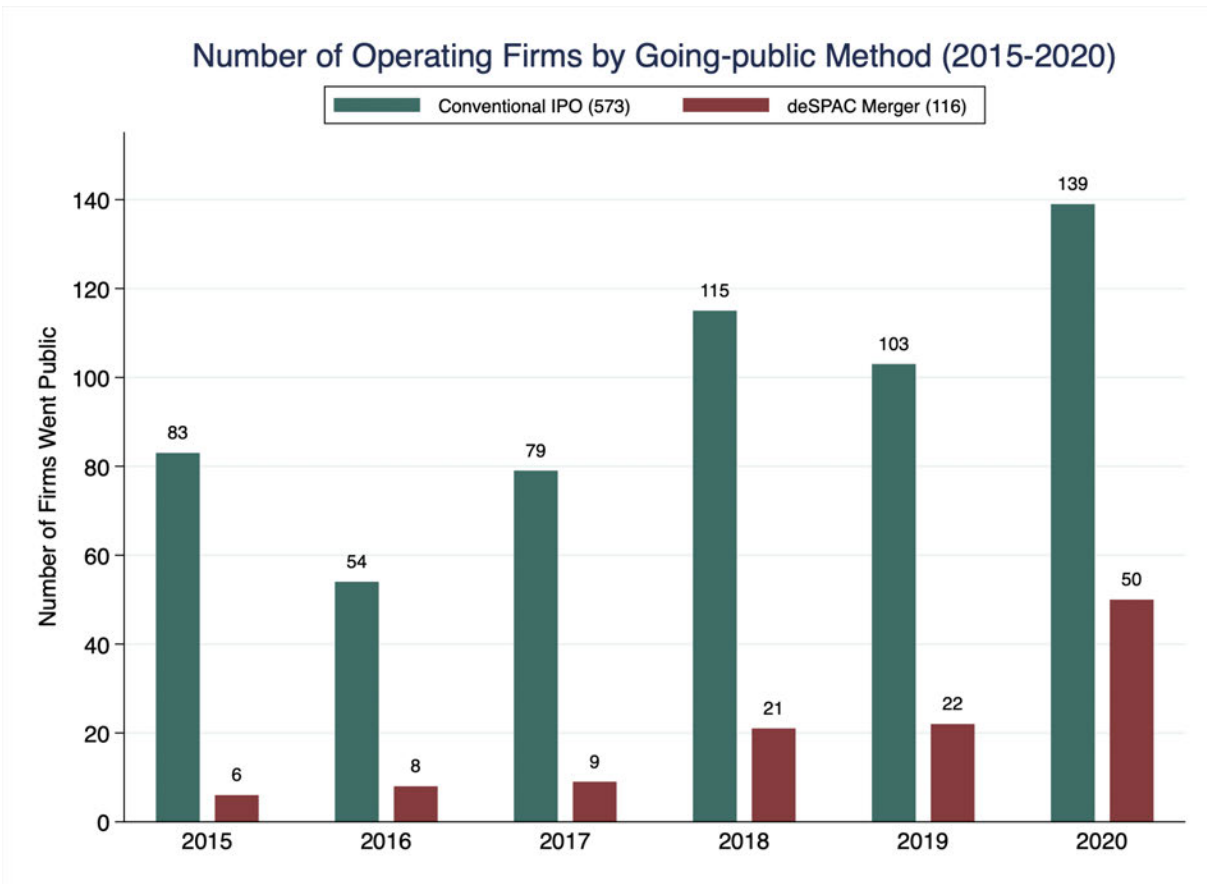


Figure 4: Comparison of Litigation Risk

This figure compares the average probability of being prosecuted in securities class action suits of deSPAC firms and their propensity-score-matched IPO peers. The mean is reported with 95% confidence intervals. The Average Peer, Median Peer, and Closest Peer of each deSPAC firm is defined as the average, median and the closest five propensity-score-matched regular IPOs in our sample, respectively. See Section 4 for the details of the peer matching methodology.

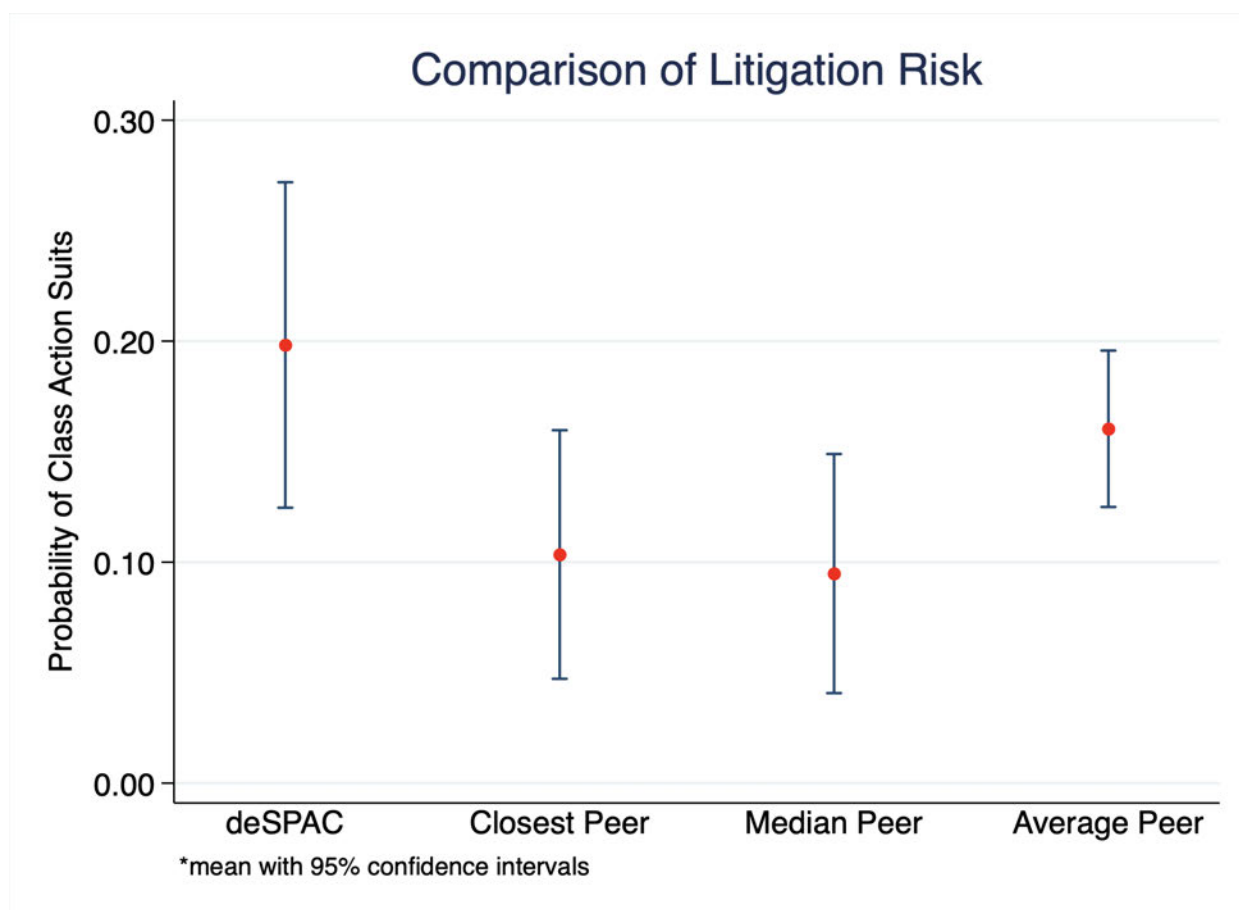


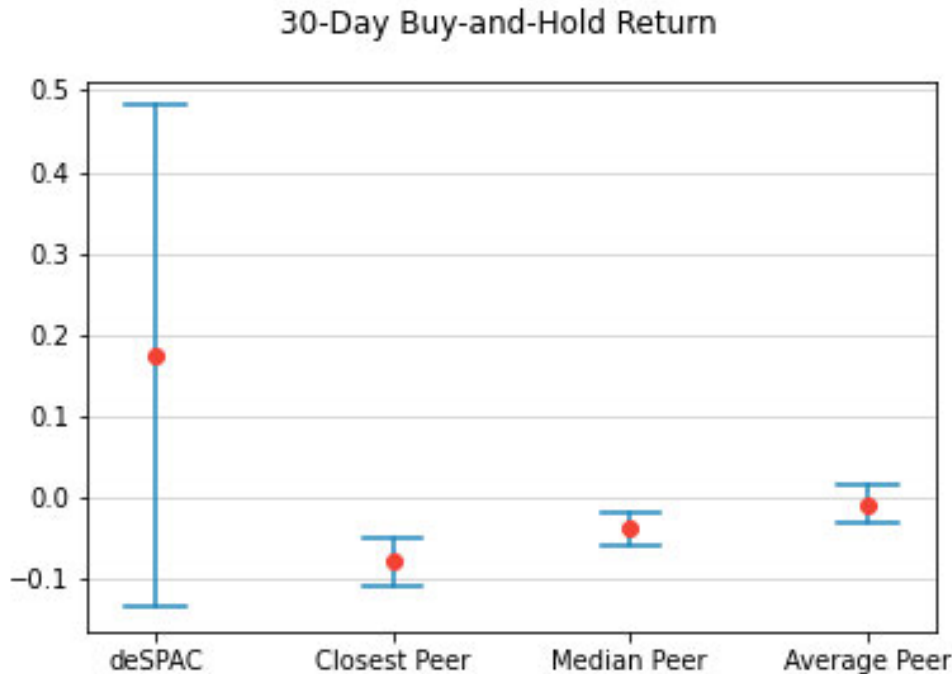


Figure 5: Comparison of Buy-and-Hold Excess Return

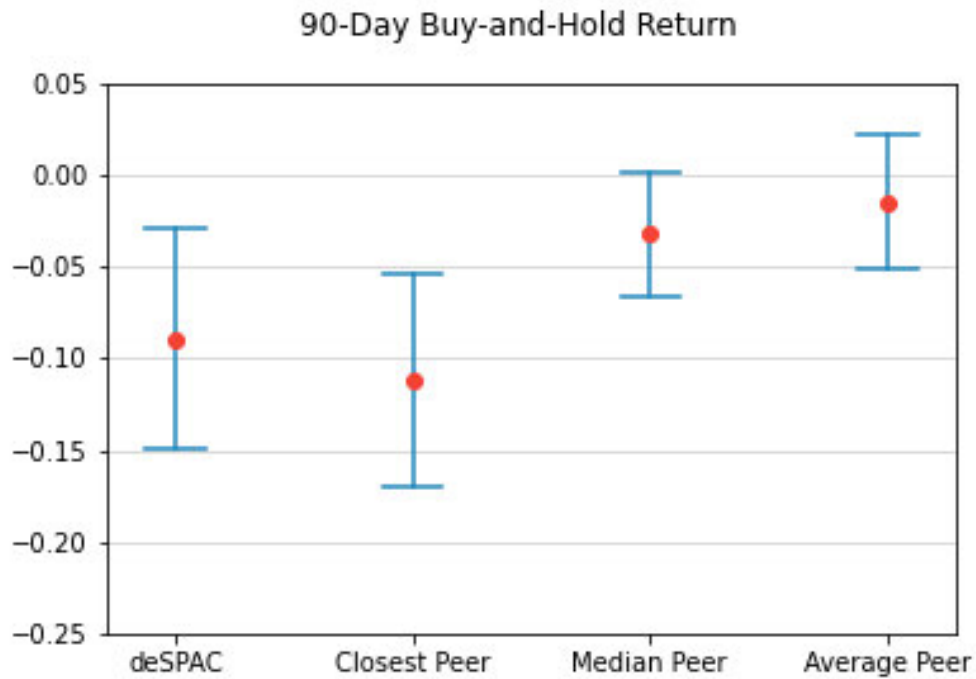
Figure 5 compares the average buy-and-hold excess return of deSPAC firms and their propensity-score-matched IPO peers, in the post-going-public period. We consider three holding periods: 30-day, 90-day, and 180-day after a firm's going public date. Compared to matched IPO firms, deSPAC firms generate better short-term average return, but underperform beyond the first 30-day of going public.

The excess return is defined as the buy-and-hold return minus that of CRSP Equal Weighted Index in matched investment period. The deSPAC sample returns are calculated based on the purchase price on the first deSPAC trading day. The IPO sample returns are calculated based on the average of the highest and the lowest trading prices on the first trading day. We use CRSP data through the end of 2020; if the event window extends into 2021, an observation is dropped from the comparison. If either data for the deSPAC firm or matched IPO are unavailable, the pair is dropped from the average calculation. See Section 4 for the details of the peer matching methodology.

(a) Holding period: through the first 30 days after going public, excess return relative to CRSP Equal Weighted Index



(b) Holding period: through the first 90 days after going public, excess return relative to CRSP Equal Weighted Index



(c) Holding period: through the first 180 days after going public, excess return relative to CRSP Equal Weighted Index

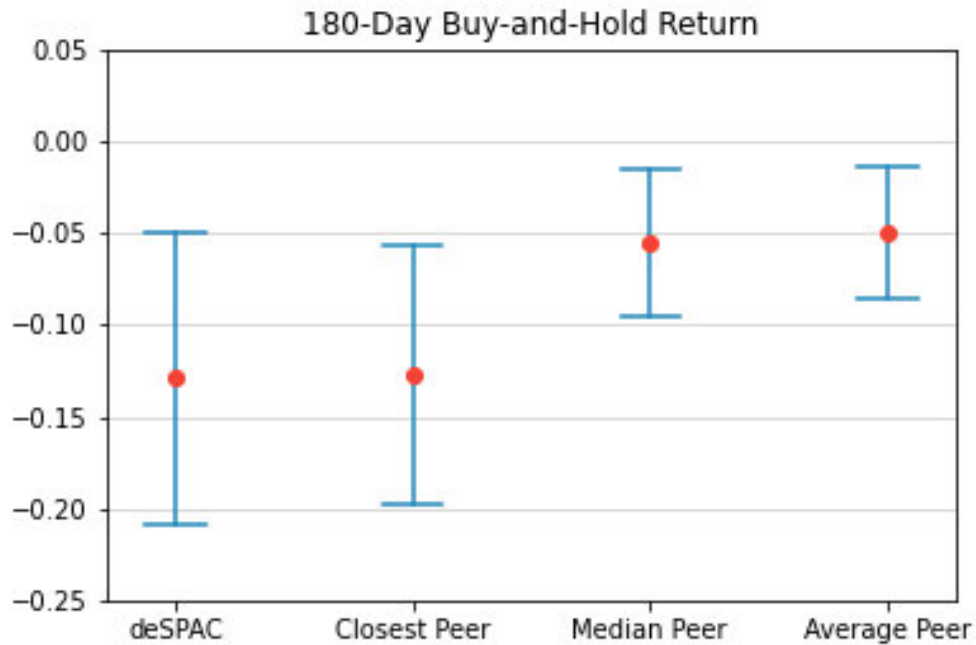


Table 1: Summary Statistics of Textual Measures

Panel A: Levels of Textual Analysis Measures in S-1		Obs.	Mean	SD	5th	25th	Median	75th	95th
<b>IPO</b>									
S-1:	Size of Raw File in MB	573	0.89	0.21	0.58	0.77	0.87	0.98	1.25
S-1:	Count of Total LM Words	573	107619.65	23730.07	72389	92639	106179	121470	146303
S-1:	Count of Unique LM Words	573	4453.49	457.58	3715	4159	4457	4770	5134
S-1:	Count of Numbers Used	573	5287.79	2055.96	2878	3990	4848	5977	9751
S-1:	%Litigious	573	1.15	0.19	0.85	1.00	1.17	1.29	1.45
S-1:	%StrongModal	573	0.54	0.09	0.40	0.48	0.54	0.59	0.69
S-1:	%Constraining	573	0.79	0.09	0.63	0.72	0.79	0.85	0.93
S-1:	%Positive	573	0.85	0.12	0.66	0.77	0.85	0.93	1.07
S-1:	%Negative	573	1.85	0.29	1.36	1.64	1.87	2.07	2.28
S-1:	%Uncertain	573	1.61	0.18	1.33	1.48	1.60	1.75	1.91
S-1:	%Risky	573	3.43	0.43	2.77	3.11	3.43	3.76	4.12
<b>SPAC</b>									
S-1:	Size of Raw File in MB	419	0.74	0.61	0.50	0.66	0.73	0.77	0.84
S-1:	Count of Total LM Words	419	103330.50	18121.89	71746	95169	106217	112334	121049
S-1:	Count of Unique LM Words	419	3338.95	221.30	3014	3217	3340	3454	3661
S-1:	Count of Numbers Used	419	2815.06	13192.78	1601	1898	2097	2299	2555
S-1:	%Litigious	419	1.23	0.12	1.10	1.18	1.22	1.26	1.34
S-1:	%StrongModal	419	1.09	0.06	1.01	1.07	1.10	1.12	1.16
S-1:	%Constraining	419	0.84	0.05	0.77	0.82	0.84	0.87	0.92
S-1:	%Positive	419	0.66	0.05	0.58	0.62	0.66	0.69	0.75
S-1:	%Negative	419	1.40	0.08	1.28	1.36	1.41	1.45	1.52
S-1:	%Uncertain	419	1.41	0.08	1.32	1.36	1.41	1.46	1.55
S-1:	%Risky	419	2.80	0.13	2.62	2.72	2.80	2.89	3.00

Table 1: (Cont.) Summary Statistics of Textual Measures

Panel B: Changes in Textual Analysis Measures during the Confidential Review Process		Obs.	Mean	SD	5th	25th	Median	75th	95th
<b>IPO</b>									
pctChng:	Size of Raw File in MB	573	11.81	13.64	-0.70	3.98	8.84	16.8	31.1
pctChng:	Count of Total LM Words	573	10.81	12.02	0.51	4.28	8.48	14.5	26.8
pctChng:	Count of Unique LM Words	573	2.88	2.93	0.23	1.03	2.05	3.98	7.69
pctChng:	Count of Numbers Used	573	28.79	32.29	-4.70	7.02	23.9	41.7	81.1
pctChng:	%Litigious	573	-2.43	4.25	-9.44	-4.82	-2.37	-0.15	4.33
pctChng:	%StrongModal	573	-0.59	6.40	-9.26	-3.68	-0.96	2.02	9.53
pctChng:	%Constraining	573	-2.21	3.97	-7.90	-4.43	-2.29	-0.35	4.16
pctChng:	%Positive	573	-1.51	4.78	-8.31	-3.95	-1.68	0.40	5.43
pctChng:	%Negative	573	-2.10	3.71	-7.38	-3.99	-2.22	-0.51	3.33
pctChng:	%Uncertain	573	-2.84	4.48	-8.53	-4.96	-2.89	-0.86	2.36
pctChng:	%Risky	573	-2.47	3.48	-7.28	-4.29	-2.35	-0.87	2.22
<b>SPAC</b>									
pctChng:	Size of Raw File in MB	384	8.73	104.70	-2.43	0.47	1.92	4.03	9.55
pctChng:	Count of Total LM Words	384	3.54	14.19	-0.74	0.53	1.83	3.75	8.91
pctChng:	Count of Unique LM Words	384	1.75	4.38	-0.48	0.15	1.07	2.27	4.76
pctChng:	Count of Numbers Used	384	52.27	784.66	-0.75	1.92	4.91	9.99	22.1
pctChng:	%Litigious	384	0.02	8.89	-4.03	-1.79	-0.55	0.23	4.55
pctChng:	%StrongModal	384	-1.14	4.37	-3.83	-1.62	-0.67	0.0057	1.93
pctChng:	%Constraining	384	-1.31	3.66	-5.60	-1.95	-0.84	-0.18	2.12
pctChng:	%Positive	384	0.03	4.17	-4.27	-1.02	-0.077	0.78	4.85
pctChng:	%Negative	384	-0.78	2.91	-3.61	-1.54	-0.53	0.058	2.00
pctChng:	%Uncertain	384	-0.82	3.74	-4.13	-1.46	-0.52	0.094	2.48
pctChng:	%Risky	384	-0.80	3.14	-3.31	-1.42	-0.58	0.016	1.58

Table 2: Summary Statistics for Class Action Suits

Panel A: Distribution of Class Action Suits by Firm Types

Class Action Suits?	IPO		Non-deSPAC		deSPAC	
	Obs.	Pct.	Obs.	Pct.	Obs.	Pct.
No	461	80.5	303	100	93	80.2
Yes	112	19.5	0	0	23	19.8
Total	573	100	303	100	116	100

Panel B: Distribution of Class Action Suits Under Section 11

Sued Under Section 11?	IPO		deSPAC	
	Obs.	Pct.	Obs.	Pct.
No	66	58.9	21	91.3
Yes	46	41.1	2	8.7
Total	112	100	23	100

Panel C: Distribution of Class Action Suits Outcomes

Case Status	IPO		deSPAC	
	Obs.	Pct.	Obs.	Pct.
Dismissed	37	33.0	8	34.8
Ongoing	43	38.4	14	60.9
Remanded	3	2.7	0	0
Settled	29	25.9	1	4.3
Total	112	100	23	100

Table 3: Variable Definitions in Propensity Score Matching Model

Variable Name	Definition
Firm Characteristics	
LN_TOTAL_ASSET	$\ln(\text{total asset in \$M})$
LN_SALES	$\ln(\text{total revenue in \$M})$
CURRENT_RATIO	current asset/current liability
LT_DEBT/ASSET	long term debt/total asset
NET_INCOME/ASSET	net income/total asset
CAPX/ASSET	capital expenditure/total asset
AGE	company age in the going public year
R&D_I	indicator variable that R&D expenditure statistics are available
R&D/ASSET	R&D expenditure/total asset. =0 if R&D_I==0.
HIGH_TECH	=1 if high-tech industry. Dummy variable based on SIC codes, following definitions in <a href="#">Puri and Zarutskie (2012)</a> .
VC_BACKED	=1 if company has received VC financing prior to going public
Notes:	
Variables are winsorized at 1% and 99% before log transformation.	
Ratios are winsorized at 1% and 99%.	

Table 4: Probit Model of Going Public Method Choice: IPO vs. deSPAC

Table 4 reports regression results of probability of going public through either IPO or deSPAC merger on firm characteristics. Sample includes operating firms that went public in 2015-2020 using either method.

Dependent Variable	deSPAC Method Indicator
LN_TOTAL_ASSET	-0.045 (-1.12)
CURRENT_RATIO	-0.049 <sup>+</sup> (-1.85)
LT_DEBT/ASSET	0.257 (1.58)
NET_INCOME/ASSET	0.026 (0.41)
CAPX/ASSET	-0.939 (-1.01)
AGE	-0.006 (-1.23)
R&D_I	-0.846*** (-4.84)
R&D/ASSET	-0.086 (-0.50)
HIGH_TECH	-0.923*** (-5.04)
VC_BACKED	0.633*** (3.87)
Constant	-0.448 (-1.31)
Going Public Year FE	Yes
Observations	809
Pseudo $R^2$	0.261

*t* statistics in parentheses. <sup>+</sup>  $p < 0.10$ , \*  $p < 0.05$ , \*\*  $p < 0.01$ , \*\*\*  $p < 0.001$

Table 5: Summary Statistics for IPO Registration Process

Panel A: Length of IPO Stage (IPOs vs. SPACs)

	Obs.	IPO Firm	Obs.	SPAC Firm	Difference
Number of days from DRS to S-1	573	119.7	384	76.5	43.2***
Number of days from S-1 to IPO	573	52.5	384	30.8	21.7***
Number of days from DRS to IPO	573	172.3	384	107.3	64.9***

+0.10 \*0.05 \*\*0.01 \*\*\*0.001; diff=IPO-SPAC

Panel B: Length of IPO Stage (non-deSPACs vs. deSPACs)

	Obs.	Non-deSPAC Firm	Obs.	deSPAC Firm	Difference
Number of days from DRS to S-1	277	72.8	107	86.1	-13.3
Number of days from S-1 to IPO	277	29.8	107	33.7	-3.90
Number of days from DRS to IPO	277	102.5	107	119.7	-17.2

+0.10 \*0.05 \*\*0.01 \*\*\*0.001; diff=NondeSPAC-deSPAC

Panel C: Length of Entire Going Public Process by Method Used

	Obs.	Conventional IPO	Obs.	deSPAC Merger	Difference
Number of days from DRS to S-1	573	119.7	107	86.1	33.7**
Number of days from S-1 to IPO	573	52.5	107	33.7	18.9*
Number of days from IPO to deSPAC	573	0	107	577.8	-577.8***
Total number of days until went public	573	172.3	107	697.5	-525.3***

+0.10 \*0.05 \*\*0.01 \*\*\*0.001; diff=NondeSPAC-deSPAC



Table 5: (Cont.) Summary Statistics for IPO Registration Process

Panel D: Comparison of Changes Between DRS and S-1

	(1)			(2)			(3)			(4)		
	Obs.	deSPAC	Diff.	Closest Peer	Diff.	Median Peer	Diff.	Average Peer	Diff.			
pctChng: Size of Raw File in MB	214	26.3	10.4	15.9	10.4	10.9	15.5	13.3	13.0			
pctChng: Count of Total LM Words	214	7.60	-6.17*	13.8	-6.17*	10.4	-2.82	11.8	-4.19			
pctChng: Count of Unique LM Words	214	2.75	-0.79	3.54	-0.79	2.75	0.00094	3.16	-0.40			
pctChng: Count of Numbers Used	214	172.0	134.7	37.2	134.7	26.7	145.3	33.8	138.2			
pctChng: %Litigious	214	2.22	4.90**	-2.68	4.90**	-2.13	4.35*	-1.92	4.14*			
pctChng: %StrongModal	214	-1.99	-1.07	-0.92	-1.07	-1.02	-0.97	-0.32	-1.67*			
pctChng: %Constraining	214	-1.83	1.01+	-2.84	1.01+	-2.21	0.38	-2.09	0.26			
pctChng: %Positive	214	-1.13	1.01	-2.14	1.01	-2.00	0.87	-1.89	0.76			
pctChng: %Negative	214	-1.23	1.67**	-2.90	1.67**	-2.35	1.13*	-2.25	1.02*			
pctChng: %Uncertain	214	-1.76	2.56***	-4.32	2.56***	-3.83	2.06**	-3.74	1.98**			
pctChng: %Risky	214	-1.50	2.12***	-3.62	2.12***	-2.77	1.27*	-2.98	1.48**			

+0.10 \*0.05 \*\*0.01 \*\*\*0.001; diff=deSPAC-Peer

Table 6: Comparison of Sentiments in Disclosures during IPO Registration Process

	Panel A: LM Word List Proportions in DRS						
	(1)	(2)	(3)	(4)			
	Obs.	deSPAC	Closest Peer	Median Peer	Average Peer	Diff.	Diff.
DRS: Size of Raw File in MB	214	0.64	0.78	0.73	0.75	-0.084***	-0.11***
DRS: Count of Total LM Words	214	92346.7	94138.5	88846.1	91313.0	3500.6+	1033.7
DRS: Count of Unique LM Words	214	3181.1	4179.3	4156.6	4185.4	-975.5***	-1004.3***
DRS: Count of Numbers Used	214	1832.6	4321.2	3919.0	4438.9	-2086.4***	-2606.3***
DRS: %Litigious	214	1.21	1.22	1.21	1.23	0.0067	-0.021
DRS: %StrongModal	214	1.10	0.52	0.51	0.52	0.60***	0.59***
DRS: %Constraining	214	0.86	0.81	0.78	0.79	0.081***	0.071***
DRS: %Positive	214	0.65	0.80	0.81	0.81	-0.16***	-0.16***
DRS: %Negative	214	1.43	1.68	1.67	1.69	-0.24***	-0.26***
DRS: %Uncertain	214	1.45	1.58	1.58	1.59	-0.13***	-0.14***
DRS: %Risky	214	2.87	3.23	3.23	3.25	-0.36***	-0.38***

+0.10 \*0.05 \*\*0.01 \*\*\*0.001; diff=deSPAC-Peer

	Panel B: LM Word List Proportions in S-1						
	(1)	(2)	(3)	(4)			
	Obs.	deSPAC	Closest Peer	Median Peer	Average Peer	Diff.	Diff.
S-1: Size of Raw File in MB	214	0.80	0.92	0.89	0.89	-0.082	-0.083
S-1: Count of Total LM Words	214	99108.5	108440.1	105286.5	105015.8	-6178.0*	-5907.2*
S-1: Count of Unique LM Words	214	3268.3	4352.0	4364.0	4341.6	-1095.7***	-1073.3***
S-1: Count of Numbers Used	214	4807.8	6432.3	6172.8	6375.1	-1364.9	-1567.2
S-1: %Litigious	214	1.24	1.09	1.11	1.12	0.13***	0.12***
S-1: %StrongModal	214	1.08	0.51	0.49	0.50	0.59***	0.58***
S-1: %Constraining	214	0.85	0.77	0.75	0.76	0.097***	0.085***
S-1: %Positive	214	0.64	0.78	0.78	0.79	-0.14***	-0.14***
S-1: %Negative	214	1.41	1.66	1.68	1.68	-0.27***	-0.27***
S-1: %Uncertain	214	1.43	1.51	1.52	1.51	-0.092***	-0.087***
S-1: %Risky	214	2.82	3.15	3.16	3.16	-0.33***	-0.33***

+0.10 \*0.05 \*\*0.01 \*\*\*0.001; diff=deSPAC-Peer

Table 7: Comparison of Sentiments in Disclosures Related to Business Operation

	(1)		(2)		(3)		(4)	
	Obs.	deSPAC	Closest Peer	Diff.	Median Peer	Diff.	Average Peer	Diff.
DEF: Size of Raw File in MB	174	0.23	0.93	-0.70***	0.90	-0.67***	0.90	-0.66***
DEF: Count of Total LM Words	174	29853.3	110055.9	-80202.7***	106814.9	-76961.6***	106550.0	-76696.7***
DEF: Count of Unique LM Words	174	1938.6	4375.1	-2436.6***	4397.3	-2458.7***	4378.8	-2440.3***
DEF: Count of Numbers Used	174	1265.2	6451.2	-5186***	6191.7	-4926.5***	6378.6	-5113.4***
DEF: %Litigious	174	1.25	1.10	0.15 <sup>+</sup>	1.12	0.12 <sup>+</sup>	1.13	0.12
DEF: %StrongModal	174	0.81	0.51	0.30***	0.49	0.32***	0.50	0.31***
DEF: %Constraining	174	0.49	0.77	-0.28***	0.75	-0.26***	0.76	-0.27***
DEF: %Positive	174	0.57	0.78	-0.21***	0.79	-0.22***	0.79	-0.22***
DEF: %Negative	174	0.98	1.67	-0.69***	1.70	-0.72***	1.69	-0.71***
DEF: %Uncertain	174	0.79	1.51	-0.71***	1.52	-0.73***	1.52	-0.73***
DEF: %Risky	174	1.77	3.14	-1.37***	3.18	-1.41***	3.18	-1.41***

+0.10 \*0.05 \*\*0.01 \*\*\*0.001; diff=deSPAC-Peer

Table 8: Comparison of Disclosure Sentiments in Annual Reports

LM Word List Proportions in the First Post-Public 10-K of the Operating Company

	(1)		(2)		(3)		(4)	
	Obs.	deSPAC	Closest Peer	Diff.	Median Peer	Diff.	Average Peer	Diff.
10K: Size of Raw File in MB	174	0.44	0.50	-0.066**	0.48	-0.048**	0.50	-0.060**
10K: Count of Total LM Words	174	54267.5	63300.7	-9033.2***	60882.1	-6614.6**	61593.3	-7325.8***
10K: Count of Unique LM Words	174	3481.3	3719	-237.7**	3735.6	-254.4***	3731.4	-250.1***
10K: Count of Numbers Used	174	3420.8	3854.5	-433.7*	3335.2	85.6	3652.6	-231.8
10K: %Litigious	174	1.13	1.19	-0.055	1.21	-0.075*	1.20	-0.074*
10K: %StrongModal	174	0.35	0.35	-0.0055	0.34	0.0040	0.35	-0.0055
10K: %Constraining	174	0.88	0.90	-0.020	0.89	-0.012	0.89	-0.0068
10K: %Positive	174	0.74	0.81	-0.067*	0.85	-0.11***	0.84	-0.10***
10K: %Negative	174	2.10	2.15	-0.043	2.25	-0.14**	2.21	-0.11*
10K: %Uncertain	174	1.75	1.76	-0.017	1.85	-0.10***	1.82	-0.076*
10K: %Risky	174	3.81	3.87	-0.064	4.05	-0.25***	3.99	-0.19**

+0.10 \*0.05 \*\*0.01 \*\*\*0.001; diff=deSPAC-Peer

Table 9: Comparison of Litigaton Risk

Panel A: Proportion t-test of Security Class Action Lawsuits

	(1)		(2)		(3)		(4)	
	Obs.	deSPAC	Closest Peer	Diff.	Median Peer	Diff.	Average Peer	Diff.
Prob. of Been Sued	232	0.198	0.103	0.095*	0.095	0.103*	0.160	0.038
Prob. of Suits within 1yr	232	0.095	0.086	0.009	0.017	0.078*	0.095	-0.000
Prob. of Suits within 2yr	232	0.172	0.103	0.069	0.034	0.138***	0.117	0.055
Prob. of Suits within 3yr	232	0.181	0.103	0.078 <sup>+</sup>	0.078	0.103*	0.147	0.034

+0.10 \*0.05 \*\*0.01 \*\*\*0.001; diff=deSPAC-Peer

Panel B: Likelihood of Security Class Action Lawsuits Conditional on Disclosure Content

	Class Action Suits Dummy				
	(1) DRS	(2) S-1	(3) R&R	(4) Def	(5) 10-K
deSPAC Dummy	0.050 (0.03)	-0.558 (-0.40)	0.677 (1.54)	1.143 (1.05)	1.137** (2.66)
%Litigious	-1.332 (-1.07)	0.958 (1.02)	0.025 <sup>+</sup> (1.76)	0.342 (0.60)	0.001 (0.00)
%StrongModal	1.621 (0.69)	2.010 (0.93)	0.060 <sup>+</sup> (1.93)	1.155 (1.08)	-0.759 (-0.37)
%Constraining	1.777 (0.58)	1.251 (0.41)	-0.162* (-2.06)	1.877 (1.19)	-2.214 (-1.21)
%Positive	1.591 (0.76)	1.489 (0.65)	-0.027 (-0.50)	1.335 (0.87)	1.130 (0.78)
%Risky	0.111 (0.13)	-0.495 (-0.58)	0.155 <sup>+</sup> (1.80)	-0.189 (-0.27)	0.718 (1.46)
Constant	-4.537 (-1.30)	-4.853 (-1.48)	-2.170*** (-5.50)	-5.050* (-2.13)	-3.711 <sup>+</sup> (-1.71)
Observations	223	223	223	203	203
Pseudo $R^2$	0.034	0.031	0.064	0.058	0.082

<sup>+</sup>  $p < 0.10$ , \*  $p < 0.05$ , \*\*  $p < 0.01$ , \*\*\*  $p < 0.001$

## Appendix A: Variable Definitions

Variable Name	Definition
File Size	The size of the <i>.txt</i> file, in megabyte (MB).
Number of LM Words	The number of words appear in the Loughran and McDonald (2011) word list dictionary.
Word Complexity	The average syllables per word for the words appear in the Loughran and McDonald (2011) word list dictionary.
%Positive	Percentage of words in DRS, or S-1, or specific section that are classified as positive based on Loughran and McDonald (2011) word list. Examples of positive words include accomplish, enable, good, stable, success.
%Negative	Percentage of words in DRS, or S-1, or specific section that are classified as negative based on Loughran and McDonald (2011) word list. Examples of negative words include loss, failure, closing, terminate, deficit.
%Uncertain	Percentage of words in DRS, or S-1, or specific section that are classified as uncertain based on Loughran and McDonald (2011) word list. Examples of uncertain words include approximate, uncertain, possible, unsure, assumed.
%Litigious	Percentage of words in DRS, or S-1, or specific section that are classified as legal based on Loughran and McDonald (2011) word list. Examples of legal words include settlement, plaintiff, bail, investigate, testify.
%Strong-modal	Percentage of words in DRS, or S-1, or specific section that are classified as strong-modal based on Loughran and McDonald (2011) word list. Examples of strong modal words include always, clearly, definitely, must, never.
%Constraining	Percentage of words in DRS, or S-1, or specific section that are classified as constraining based on Loughran and McDonald (2011) word list. Examples of constraining words include bond, covenant, forbid, impose, restrict.
%Risky	Percentage of words in DRS, or S-1, or specific section that are classified as negative or uncertain based on Loughran and McDonald (2011). Defined as %Negative plus %Uncertain.
Change in %LM Word Lists	Defined as the %LM word list in S-1 minus the %LM word list in DRS for each category of LM word lists in DRS, or S-1, or specific section, in basis points.

# The Going Public Decision and Retail Investors in SPACs vs. IPOs

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## Abstract

This is the first paper to document investor trading activity immediately following a firm's public market debut, via an IPO and SPAC backed acquisition. Our analysis compares firms going public through SPACs vs. IPOs and finds that SPAC backed firms tend to be younger, VC-backed, have lower current ratios, relative to IPO firms. While the average returns on SPAC common shares outperforms the market after target announcement, they generate larger losses on average, after the target firm goes public, relative to firms going public via IPOs. The popularity of SPACs among retail investors tend to be driven by their return volatility following the announcement of the deSPAC target. Overall, our results provide a methodical comparison between these alternate ways of going public, suggesting that in the long-run, SPACs might be the third leg in the going-public process after IPOs and direct listings.

**JEL:** G24, G32, G34

**Keywords:** SPAC, IPO, Public Equity Offering, retail investors

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## 1 Introduction

Over the last several years, Special Purpose Acquisition Companies (SPACs) have gradually gained popularity as an alternative “going-public vehicle” for startups and entrepreneurial firms. Though SPACs or “blank check companies” have existed for decades, the financial maneuver to use them for going-public by mainstream underwriters and VC-backed private firms is more recent and dramatically took off in early 2020 with the onset of the Covid-19 pandemic and the worry about its impact on financial markets. Though the public visibility of SPACs in popular news media was obvious in 2020, the current financial structure of the SPAC and the presence of mainstream bankers, investors, and VCs in SPAC deals had gradually started to take place since 2018. Given the uncertainty and the volatility of the public markets over the last two decades and the typical time required for a firm to go public through a traditional IPO process, private firms have been much more likely to be acquired than to go public during this time: see, for example, Gao et al. (2013), who provide evidence on the dramatic reduction in IPOs in the United States over the last two decades.<sup>1</sup> However, the obvious drawback of an acquisition is that the founders lost the decision-making control of their products and their technology to the acquiring firm. Thus, for those entrepreneurs who wanted to retain that control, the primary approach was to remain private for as long as possible and then attempt to go public via an IPO or a direct listing.<sup>2</sup> It is in this environment, that SPACs provided an alternative way for such private firms to transition to the public market, by significantly shortening the “time to market” relative to a traditional IPO.

The status quo of the soft IPO market and the inclination of large number of startups to remain private for as long as possible was shaken up by the Covid-19 pandemic. In early 2020, the pandemic greatly accelerated the pace of technology adoption across various facets of our life

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<sup>1</sup>According to the National Venture Capital Association (NVCA), there were more exits by venture capitalists through acquisitions than by IPOs in each of the last 20 years. On any given year the number of exists via acquisitions were 5x to 8x more than that of IPOs. The volume of VC-backed IPOs however started picking up from 2017. The NVCA reports that acquisitions constituted 85% of all exits of venture-backed firms in 2010 and gradually came down to 68.5% in 2020.

<sup>2</sup>In September 2019, about 100 private companies, along with partners from venture firms Andreessen Horowitz, Menlo Ventures, Bessemer Venture Partners, GGV and others, attended a private event to discuss alternative ways to go public rather than through the traditional PO. The sentiment of the meeting was nicely summed up by Brad Gerstner, who sponsored the SPAC, Altimeter Growth Capital: “The days of handing low-priced securities, diluting the company to a bunch of long-only shops in Boston and New York, I think those days are over.” Incidentally, in mid-April 2021, Altimeter Growth Corp struck the largest SPAC merger deal agreeing to take Grab public in a \$40B deal. Adapted from Bloomberg, available here: <https://www.bloomberg.com/news/articles/2019-09-30/silicon-valley-calls-summit-to-disrupt-ipos-bankers-not-welcome>



and in the economy, which created an unusually “hot” market for such startups, particularly for high-tech private firms, as it scaled up the product market opportunities rapidly, thus creating a need for large capital infusions and thereby a viable opportunity for them to go public (see, Chemmanur et al., 2010; Clementi, 2002; and Spiegel and Tookes, 2012). In addition to the need for capital infusion, such rapid technology adoption in the economy, also led to faster product commoditization and therefore reduced the benefits to firms for remaining private, causing them to go public earlier in their lives (Spiegel and Tookes, 2020). Similarly, in such situations, where there is rapid commoditization within a particular industry, even firms that may otherwise not require an immediate capital infusion, may go public, given similar decisions of going public by their product market competitors. This will result in IPO waves in industries which experience such rapid changes in adoption of newer products and technologies (Chemmanur and He, 2011).<sup>3</sup>

In some sense, SPACs provide certain features in the going public process that may be more attractive for private firms. First, in SPAC transactions, awareness from investment bank analysts on an upcoming deal is immediate and therefore might generate trading momentum earlier than in a traditional IPO. It is worth mentioning here, that throughout the paper, we will refer to the “SPAC” period and the “de-SPAC” period to correspond to before and after the initial business combination completion, respectively. Specifically, in contrast to a traditional IPO, since de-SPAC transactions are viewed as an acquisition on the part of the publicly traded SPAC entity, analysts are immediately allowed to review the proposed acquisition deal and financial projections of the target firm, and provide their analysis to prospective investors. As a result, once the target company’s stock starts trading publicly, it benefits from analyst reports on its growth projections and valuation metrics, which may help stimulate liquidity of such firms in the public markets, relative to others who go public following a traditional IPO, where the quiet period is binding following the IPO. Second, the direct costs in a SPAC are typically less compared to a traditional IPO and generally

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<sup>3</sup>This is easily seen in many industries. The EV market is one that has drawn a lot of attention recently. The market leader, Tesla, went public in June 2010, but since then the industry was really not advanced enough for any of its competitors to go public - even though there are several firms that are working on the EV and related technology. However, in 2020 the industry took a leap forward with Nikola, Hyliion, Lordstown Motors, and Lucid Motors, all announcing their intent to go public within months of each other, leading to a classic “going public” wave in the EV industry. Anecdotal evidence suggests that a high degree of product commoditization and transitioning to the public markets to maintain parity with their competitors seems to have driven this decision for the firms. Incidentally, all of these firms decided to go public by merging with a SPAC, making it possible to go public within a short period of time of each other.

borne by the SPAC entity sponsors as opposed to the company.<sup>4</sup> Finally, while a direct listing has many of the same advantages as going public through a de-SPAC acquisition, it may not be appropriate for smaller private firms and those where the technology and industry structure is more uncertain. A direct listing also does not allow capital infusion at the listing until a recent regulation change in late 2020, requiring a separate VC or PE deal to be completed either prior to going public or for the firm to issue equity through a seasoned equity offering right after going public - in both cases the costs involved are significantly greater, thus making it less attractive for smaller private firms.

Our study contributes to several strands of literature in empirical finance. First, this paper adds to a growing body of research on the current generation of SPACs - in particular, we are the first paper to empirically document the trading activity of investors immediately following an IPO and a SPAC transaction. In the latter case, we show how investors behave in the periods prior to the announcement of a SPAC deal, following the announcement of a deal but before the target firm starts trading, and finally after the private firm's equity trades in the public market. Second, we further the going public literature by comparing SPAC and IPO as private firms' exit options, in terms of ex ante characteristics of firms and market performance. Lastly, the paper contributes to the literature on individual investor behavior with novel evidence on retail trading activities of SPACs in 2018-2020, which is the most recent period when SPACs have grown rapidly in terms of market activity and recognition by the general public.

Our research complements three contemporaneous papers on SPACs. While we discuss the broader (and earlier) literature in the following section, it is worth mentioning here, how our paper differs from these three other contemporaneous papers, namely, Bai et al. (2020), Klausner et al. (2020), Gahng et al. (2021). Our findings add to the collective understanding of SPACs from these papers by providing new evidence on investor trading activity following SPACs and IPOs. This is the first paper to provide such evidence. Unlike the other papers, our paper also provides evidence on the going public choice between an IPO and a SPAC merger by analyzing the private firm and market specific financial characteristics, prior to going public. While we also document several common return and market outcomes following SPACs, our analysis is benchmarked against the

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<sup>4</sup>In reality though, one would argue that these costs are internalized and passed back to the SPAC target firm, through the valuation that it receives from the sponsors.

equivalent outcomes following a traditional IPO, therefore providing a clear relative comparison with the status quo. Finally, our sample size and coverage of SPACs is also significantly larger than in these contemporaneous papers, as we have relied on manually collecting data from several SEC filings to supplement the existing databases for our analysis, particularly for the current period after 2018.

Our analysis in this paper, can be summarized as follows: We compare firms going public via SPAC versus traditional IPO in terms of (1) ex ante firm characteristics before going public, (2) common share market performance, and (3) retail investor interest. We use a long sample period from 2003-2020 and compare several sub-periods during this time to illustrate the differences and the evolution of SPAC characteristics.

Our results show that firms going public via SPAC tend to be younger, more likely to have received VC financing, have lower current ratio, and less likely to be in a high-tech industry. Between the SPAC IPO and six months after de-SPAC, SPAC common shares have the highest liquidity in the period between target announcement and the initial business combination completion - incidentally, this is also the period when the downside risk is mitigated by the redemption feature offered by SPACs. In 2018-2020, the average buy-and-hold return on SPAC common shares outperformed the market during the SPAC period, and is comparable in magnitude to the buy-and-hold return of IPO shares purchased at the IPO offer price.<sup>5</sup> However, after the de-SPAC merger completion, the run-up in share price does not sustain. In the de-SPAC period, the decline in share price is larger in magnitude for riskier firms, characterized by low revenues prior to going public. Compared to IPO firms, de-SPAC firm shares on average generate larger losses after going public. Retail interest in SPAC investment is weak before target announcement, but rise quickly and strongly after the merger target is identified. SPAC popularity among retail investors is primarily driven by stock price volatility and trading volume, and less so by target firm characteristics or SPAC IPO features. Compared to IPO firms, de-SPAC firms are more likely to attract investors when their stock has

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<sup>5</sup>It is critical to note however, that barring the very recent move by Robinhood and SoFi to allow retail investors to bid in IPOs, historically this has been a market purely for institutional investors. It is unheard of retail investors receiving an IPO allocation at the offer price. On the other hand, it is relatively easy for retail investors to invest in shares of firms going public via a SPAC merger by simply buying shares of the SPAC. Thus, an appropriate comparison would be to compare the buy-and-hold returns to investors in SPACs relative to the return in IPOs bought on the first trading day. In this case the SPAC returns outperforms the IPO returns for the first 30 days from the first day of trading, but however reverses thereafter for longer durations.

high volatility.

The rest of the paper is organized as follows: Section 2 discusses related literature. Section 3 describes sample selection, data, and institutional background. Section 4 presents empirical evidence on firms' choices on the going public decision and the market performance metrics of SPAC common shares, in comparison to IPOs when relevant. Section 5 analyzes retail investments in SPACs. Section 6 concludes.

## **2 Related Literature**

The finance literature has studied SPAC since the initial boom of SPAC activities in the 2000's. Shachmurove and Vulcanovic (2017) provides a literature review on academic research in SPACs as of 2017, in which the data samples typically end in early 2010s. SPACs in recent years are distinct from those studied in earlier research. As detailed in the institutional background in Section 3, several regulatory changes encouraged the growth of the SPAC market since 2010 and the instrument has become well accepted by financial intermediaries and firms only recently.

Following the rapid resurgence of SPACs since 2018, a growing number of studies revisit SPACs and incorporate the latest data. Klausner et al. (2020) analyze the factors that affect post-merger returns of all 47 SPACs that merged between January 2019 and June 2020. They caution the high dilution to public shareholders' equity due to sponsor promote, public share redemption, and warrant issuance. Gahng et al. (2021) analyze the return on common shares and warrants of exchange-traded SPACs since 2010. They also document that sponsors and underwriters forfeit shares and warrants, or structure earn-out compensation in order to induce business combination, using a sample of 20 recent SPAC merger transactions in late 2020. Bai et al. (2020) develop a theoretical framework to explain the cyclicity of SPAC waves. Empirically, they show that SPAC activity correlates with positive sentiment in the equity markets. They argue that SPAC targets tend to be riskier firms that cannot easily go public through traditional IPO due to underwriter liabilities concerns. SPAC wave pattern is also found to be negatively correlated with market uncertainty index (VIX) and time-varying risk aversion (variance risk premium) between 2003 and 2019 (Blomkvist and Vulcanovic, 2020).

This paper provides several unique perspectives not addressed by other recent studies on SPACs, including the firm choice between SPAC and IPO, retail investing in SPACs, event-window

based market performance analysis, and the financing activities of deSPAC firms after the initial business combination. We incorporate a long sample period from 2003-2020 to compare and contrast several sub-samples that represent different stages of the modern history of the SPAC market.

“Going public” is one of the most important milestone events for firms. Traditionally, after the decision to exit, firms may choose to have an IPO or be acquired by another firm. In recent years, going public via SPACs has evolved into a standard option in private firms’ choice set of an “exit”. This paper extends the going-public literature by analyzing SPACs as another exit option for private companies. We provide empirical evidence on the relationship between ex ante firm characteristics and its choice of the going public method. We also compare the market performance and retail investor interest between deSPAC and IPO firms.

We draw from the rich theoretical literature on the going-public decision to select relevant firm characteristics used in the empirical analyses. For example, we control for company age, industry, capital intensity, liquidity and solvency, and innovation and R&D. These features are considered relevant in several models on the going-public choice. Chemmanur and Fulghieri (1999) model the going-public decision in an environment of asymmetric information, which implies that larger and more capital intensive firms, with riskier cash flows, and those operating in industries characterized by lower information production costs are more likely to go public. Bhattacharya and Ritter (1983) and Maksimovic and Pichler (2001) argue that the decision to go public emerges from the tradeoff between the costs of the firm of releasing confidential information (helpful to competitors) at the time of IPO versus the benefits arising from raising capital at a cheaper rate in the public equity markets. These theories imply that firms with greater existing market share, and those operating in industries characterized by a lower degree of competition, are more likely to go public. Spiegel and Tookes (2007) develop a model of the relationship between product market innovation, competition, and the public versus-private financing decision in an infinite-horizon model. Their model predicts that firms will finance projects with the greatest revenue-generating ability privately, and will then go to the public markets only when more modest innovations remain.

Related to the private firm exit choice, existing empirical research has studied the choice between going public via traditional IPO and being acquired. A few papers provide U.S.-based evidence. Using a sample of firms exiting between 1995 and 2004, Poulsen and Stegemoller (2008)

study the choice between sellouts to publicly held acquirers and IPOs for firms moving from private to public ownership. Brau et al. (2012) examine how various industry, stock market, and deal characteristics affect a private firm’s choice between an IPO and an acquisition by a public acquirer. Chemmanur et al. (2018) use unique U.S. census data to conduct comprehensive analysis of the three-way choice of private firms between IPOs, acquisitions by private or public acquirers, and remaining private. This paper brings in discussion on the additional option of going public via SPACs, which is mechanically a combination of an IPO and an acquisition.

Finally, this paper relates to research on the behavior of individual investors. Commission-free online trading has made stock investment highly accessible to U.S. retail investors in recent years. A rapid rise in retail investing activities has been observed in 2020, while SPAC activities reached all time high at the same time. We study how retail investors have approached SPACs as an investment opportunity using users holding data of Robinhood users in 2018-2020. We provide empirical evidence on the timing and main drivers of retail investments in SPACs. This paper joins several other recent papers on retail investor behavior and investment returns using the Robintrack dataset (see, e.g., Barber et al., 2021; Ben-David et al., 2021; Pagano et al., 2020; Moss et al., 2020; and Welch, 2020).

### **3 Sample Selection, Data, and Institutional Background**

In this paper, we study U.S. SPACs traded on major exchanges (AMEX, NYSE, Nasdaq) and draw comparisons between firms that went public via SPAC versus traditional IPO. Our data sources include Thomson Reuters SDC Platinum, FactSet, Compustat, SEC filings, CRSP, Robintrack, Crunchbase, and VentureXpert.

#### **3.1 SPAC and Traditional IPO Sample**

We identify the universe of SPACs listed on U.S. markets from 2003-2020 using SDC New Issues Database and FactSet, from which we also collect standard information associated with an IPO, such as CUSIP, issue date, offer price, exchange, total proceeds, and underwriter identity. We then use the SDC M&A Database and FactSet to determine whether a SPAC has identified merger target(s) and completed its lifecycle. We categorize a SPAC’s lifecycle status into “merger complete”, “closed”, or “active”. A SPAC has the “merger complete” status if it has completed the initial

business combination. A “closed” SPAC ceases to exist as a blank check company through either liquidation or change of corporate charter as approved by shareholders. An “active” SPAC is still searching for a target or in the process of closing the initial business combination. Any inconsistency between SDC and FactSet are reconciled by information from SEC filings. For SPACs with “merger complete” status by the end of 2020, we collect identifiers of the deSPAC firm, target announcement date, deSPAC completion date, and details of the SPAC IPO structure.

Figure 1 reports the SPAC IPO activities from 2003 through 2020, including annual SPAC IPO counts, total proceeds, and average proceeds. SPAC IPOs activity increased leading up to the financial crisis of 2007-2008, and declined thereafter. SPAC IPOs again accelerated from 2015 and reached all-time high in terms of counts and proceeds in 2020. The growth of SPACs continues into 2021. Figure 2 shows the SPAC lifecycle status as of December 2020 by SPAC IPO year. Over time, the proportion of SPACs with a successful initial business combination has increased. This partially attributes to relaxation of exchange listing rules related to maximum redemption and shareholder approval of the business combination, which we provide more details in the recent history section below.

With a focus on comparisons between going public via SPAC vs. traditional IPO, many of our analyses are performed on the subset of SPACs with a “merger complete” status as of the end of 2020. Further, we exclude SPACs traded on OTC or pink sheet during the SPAC period and/or immediately after the initial business combination. 189 SPACs with completed merger remain after this exclusion, among which the first deSPAC event took place in 2006. For this subset of 189 SPACs, we manually collect data on the SPAC IPO offering structure from SEC filings (Table 1). Several key features of SPAC IPO structures have evolved over time. Since 2010, it has become standard practice to place at least 100% of SPAC IPO proceeds in trust account for public share redemption, despite the exchange requirement of 90% minimum. In order to cover the expenses of the SPAC IPO, such as upfront underwriting fees and legal expenses while maintaining the trust reserve level, sponsors contribute additional capital in the form of private placement purchases concurrent to the SPAC IPO. The sponsor capital/SPAC proceeds ratio is on average 3%. Until the recent few years, specialty underwriters such as EarlyBirdCapital have had a large market share in SPAC IPO underwriting. SPAC underwriting gross spread is paid in two

stages, commonly 2% at SPAC IPO and 3.5% at business combination, where the deferred 3.5% are forfeited if a SPAC liquidates. The fee deferral helps sponsors to reduce initial capital investment at the IPO by reducing upfront expenses. A SPAC unit typically consists of one common share and warrants/rights. The warrant incentive has declined over time both in terms of warrant quantity and exercise price. Since 2013, each unit consists of warrant that’s convertible to half of a common share or less on average. Until 2010, warrant exercise price is commonly below the unit IPO offer price, and this has become rare thereafter.

The sample of traditional IPO is constructed from SDC New Issues Database. We start with all U.S. Common Stock IPO offerings from 2006 to 2020, then exclude offerings with offer price below \$5, ADRs, units, REITs, closed-end funds, banks, ETFs, SPACs, direct listings, and stocks not included in CRSP such as OTC issues.<sup>6</sup> Figure 4 summarizes the yearly count of operating firms going public on major exchanges in the U.S. through SPAC and traditional IPO between 2006 and 2020. Over time, SPAC merger has become a more common method to go public for operating companies. In Figure 4, we also show the number of firms that remain listed as a public company as of May 18, 2021 by the going public year and the going public method. Unreported t-tests by year indicate that the delisting proportion are statistically similar between the deSPAC and IPO sample in all but two years. The average time from going public to delisting among inactive firms is 5.4 years for the IPO sample and 4.0 years for the deSPAC sample. As the deSPAC sample is skewed towards recent years, time will tell whether the proportions will diverge. This statistic does not distinguish firms that exit the public market through voluntary versus involuntary delisting.

### **3.2 Firm Fundamental Data**

For operating firms going public in 2015-2020 via either SPAC or IPO, we collect company financial and employment data in the last full fiscal year prior to going public, company age when going public, and VC backing status. Table 2 lists the variables and summarizes the group means. Compustat is the main source for financial data, industry SIC codes, and employment data. However, Compustat data is not a consistent source of SPAC target company information prior to the deSPAC merger. In addition to missing data, Compustat may incorrectly record SPAC period statistics under the deSPAC firms’ identifiers. We screen for these cases and manually collect the data from

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<sup>6</sup>Table 17 in appendix reports summary statistics of the IPO sample.



the 10-K filings when necessary.<sup>7</sup> Company age of the IPO sample is based on the founding dates of Loughran and Ritter (2004). VC backing status of the IPO sample is based on SDC New Issues Database. Company age and VC backing status of the deSPAC sample are collected by the authors using VentureXpert, Crunchbase, and firm websites.

### 3.3 Robinhood Users Holding Data

We use Robinhood users holding data to examine retail investment activity in SPACs between May 2018 and August 2020. The Robinhood data is obtained from Robintrack (<https://robintrack.net/>), an open-access website that scrapes the number of user data through a Robinhood API.<sup>8</sup> Robintrack data report the number of Robinhood user holding a specific ticker on an hourly basis between May 2, 2018 and August 13, 2020.<sup>9</sup> For example, 637,705 users held American Airline (AAL) shares through the brokerage as of 8/13/2020 22:58 UTC.

Robintrack data identifies stock with tickers. We link Robintrack data to CRSP data by ticker and date, then use CRSP identifiers to merge with other data sources. The universe of ticker in Robintrack data includes stocks and ETFs. There are 8,595 unique tickers in the sample. Data period for each ticker varies. Because the holding quantity of each user is unknown, the data is best interpreted as a popularity measure. We analyze popularity and popularity change on a daily frequency. For every ticker, we retain last data point reported each day, based on Eastern Time (the time zone of major U.S. exchanges). Several recent research papers use the same data to study retail investor behavior and investment returns (see, e.g., Barber et al., 2021; Ben-David et al., 2021; Pagano et al., 2020; Moss et al., 2020; and Welch, 2020).

### 3.4 Recent History of the SPAC Market

Here we highlight several important changes in exchange regulations on SPACs and SPAC IPO offering features, which inform our choices of subsamples periods, namely 2003-2009, 2010-2017, and 2018-2020.

The first modern SPAC is known to be Millstream Acquisition Corporation, issued in August

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<sup>7</sup>We use the first 10-K or 10-K/A after the deSPAC merger completion date. Financial data reported in foreign currencies are omitted.

<sup>8</sup>Robinhood shut down this API on August 13, 2020; the data no longer update henceforth.

<sup>9</sup>Data are missing during 2019/01/24-29 and 2020/01/07-15.

2003. Between 2003 and 2008, SPACs were only accepted to list on OTC and American Stock Exchange (now NYSE American). SPAC activities have increased year over year until 2007. Amid the financial crisis, SPAC IPO counts decreased from 66 in 2007 to 17 in 2008 and 1 in 2009; the SPAC fundraising as share of all IPO revenue decreased from 14% in 2007 to 13% in 2008 and 0% in 2009.

Concurrent to the financial crisis, two developments in the SPAC market signify a new generation of SPAC. In 2008, both Nasdaq and NYSE began to permit SPAC listings. These two major exchanges provide better liquidity and broader recognition from investors. Figure 3 shows the listing exchanges of SPAC IPOs by IPO year. Almost all SPAC IPOs have chosen to list on the three major exchanges since 2012.

Moreover, around 2008, SPACs started to widely adopt limitation on public share redemption by blockholders in IPO offerings, where the threshold is 10-20% of shares offered during IPO. Under such restrictions, SPACs reduce the risk of blockholders forcing SPAC management into negotiated sales at the time of merger. Early SPACs without such provisions were frequently targeted by so called “SPAC mafia” hedge funds, who repeatedly used this predictably profitable strategy. Such events result in lower cash at disposal for the merger transaction and negatively impact all remaining stakeholders.

In December 2010, Nasdaq adopted the tender offer rule, which allows SPACs to complete a merger without shareholder approval by offering to buy back public shares at a price pro rata of the trust value. This further increases the likelihood that a merger can be successfully pursued. When the tender rule applies, the *de facto* barrier to business combination is that SPACs must maintain at least \$5 million in net tangible assets; i.e. public redemption cannot exceed the level where less than \$5 million in cash is left. If this requirement is not met, public companies are subject to Rule 419, also known as the “penny stock” rule. Under Rule 419, trading of common stock and warrant are not permitted until completion of business combination, and warrants can be exercised prior to business combination and the proceeds are placed into the trust account. Because Rule 419 will completely alter how SPACs operate, most SPACs specify in the IPO prospectus that business combination will not proceed when public share redemption causes less than \$5 million in cash remains in trust.

Between 2010 and 2017, NYSE listing requirements for SPACs were more restrictive than NASDAQ. In particular, the NYSE listing rules required a shareholder vote for the proposed initial business combination and imposed a maximum 40% redemption requirement. As a result, most SPACs chose to list on Nasdaq between 2010 and 2017. NYSE adopted similar tender offer rules in March 2017. Currently, the listing requirements of Nasdaq and NYSE are almost identical.

#### **4 SPAC vs. IPO: Firm Choice and Market Performances**

In this section, we first examine the choice of going public method between SPAC and traditional IPO by firms based on firm characteristics prior to going public. Regression estimations using post-2015 samples show that firms are more likely to go public via SPACs if they are younger, have received VC financing, have lower current ratios, and not in a high-tech industry. Then we examine the market performance of SPAC common shares, with regard to liquidity, buy-and-hold return, and the Sharpe ratio. We highlight that target announcement during the SPAC period is a critical event, after which trading of SPAC common share is the most active and the share price experience large but unsustainable appreciation in 2018-2020.

##### **4.1 Firm Choice of the Going Public Method**

Alternative going public methods such as SPAC and direct listing have become more common in recent years. By the end of 2020, nearly 200 companies have become listed on a major U.S. exchange through a SPAC merger, and about 6 firms went public in the U.S. through direct listing. A common view is that SPACs are used by firms that cannot successfully complete a traditional IPO. Compared to the traditional IPO process, SPAC is considered to be a speedier method, since roadshows and SEC review are not involved. From a legal liability perspective, SPACs commonly provide financial projections of the target firm to investors prior to merger completion, which is typically not found in a traditional IPO prospectus.<sup>10</sup>

We perform an empirical analysis of firms' choice between SPAC and IPO between 2015-2020, based on firm characteristics prior to going public. For operating companies going public in 2015-2020 via either SPAC or IPO, we collect company financial and employment in the last full fiscal

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<sup>10</sup>The SEC is closely examining the possible regulation arbitrage related to differences in disclosure liability of SPACs compared to IPO. See SEC Public Statement issued on April 8, 2021: "SPACs, IPOs and Liability Risk under the Securities Laws".

year prior to going public, company age and industry SIC code when going public, VC backing status, and IPO/SPAC IPO underwriter rank. Financial statistics are selected to represent firm size, liquidity, profitability, and investment intensity.

We use probit models as the baseline, but also estimate logit and linear probability models to verify the results,

$$\text{Going Public via SPAC}_i = \beta_0 + \mathbf{Firm\ Characteristics}_i \beta_1 + \alpha_1 \text{FF17}_i + \alpha_2 \text{Going Public Year}_i \quad (1)$$

Table 6 lists the definitions of firm characteristics variables. FF17 is a dummy variable for each Fama-French 17 Industry. Going Public Year is a dummy variable for each going public year. Table 4 presents the regression estimates. The estimations are repeated on the 2018-2020 (Panel A) and 2015-2020 sample (Panel B). The results are not sensitive to the sample period choice.

Several results are statistically significant in both sample periods. Firms are more likely to choose SPAC if they have low short-term liquidity, measured by current ratios. Younger and VC-backed firms are more likely to choose SPAC. Firms in a high-tech industry are less likely to choose SPAC, mainly driven by the lack of deSPAC firms in the biotech industry. The underwriter reputation is lower for SPACs. In the past, specialized underwriters have large market shares in SPAC IPO underwriting; this differential may close in the near future as major investment banks step up in SPAC IPO underwriting. Additionally, in 2015-2020, firms with more long-term debt relative to total asset are more likely to choose SPAC. In 2018-2020, firms with more total assets are more likely to choose SPAC. The company size distribution of deSPAC firms has less extreme values on both tails compared to the tradition IPO firms. As the average SPAC IPO proceeds increase over time, SPAC merger target are on average larger than traditional IPO firms.

A common view on SPACs is that this going public method is favored by smaller and riskier firms. The regression estimates partially corroborate this view, in that younger firms and firms with lower liquidity tend to choose SPAC over IPO. This may be related to the valuation certainty and quicker transaction timeline provided by SPACs.

The following sections evaluate the market performances of SPAC common shares with regard to liquidity, buy-and-hold return, and Sharpe ratio. The sample includes 189 SPACs that have completed the deSPAC initial business combination by the end of 2020. When appropriate, we

compare the deSPAC and IPO samples. We define event windows based on two milestones in a SPAC's lifecycle: target announcement during the SPAC period and the deSPAC initial business combination.

## 4.2 Liquidity during SPAC Lifecycle

Table 5 reports the average liquidity measures of SPAC common shares in five event windows. The event windows are (1) SPAC period, before target announcement, (2) SPAC period, after target announcement, (3) deSPAC period, the first 30-Day, (4) deSPAC period, the first 90-day, and (5) deSPAC period, the first 180-day. The averages are computed for three subsamples based on year groups: 2005-2009, 2010-2017, 2018-2020. Year group is based on the SPAC IPO year for the event windows during the SPAC period, and deSPAC year for the event windows during the deSPAC period.

Three liquidity measures are considered: the Amihud illiquidity measure, bid-ask spread, and daily turnover. The three measures show consistent conclusions. Among the three year groups, SPACs are the least liquid during 2010-2017. Across all year groups, SPAC common share is the most liquid during the SPAC period after target announcement (event window 2). We perform mean difference t-tests between the liquidity measures of event window 2 and other event windows. The higher liquidity in event window 2 is statistically significant compared to before target announcement, and 90-day/180-day after deSPAC. This indicates that trading of SPAC common share is particularly active after target announcement until the initial month as the deSPAC company, after which trading activities dwindle.

## 4.3 SPAC Period Buy-and-Hold Return

Then we compute the average buy-and-hold return (BHR) of SPAC common shares, using the same event windows and year-group subsample as the previous section. Table 6 summarizes the event window lengths by years. On average, target announcement takes place 1.3 years after the SPAC IPO and the merger transactions takes about 0.4 years to close. In recent years, both event windows have shortened in length.

Figure 5 summarizes the average buy-and-hold return by event window and year groups. As benchmarks, we also report the average total buy-and-hold return on the CRSP equal-weighted and

value-weighted market indices during matched investment period for each event window observation. Because large intra-day price range is frequently observed on target announcement and deSPAC dates, we use the average of highest and lowest transaction prices to calculate the returns in all buy-and-hold return analyses in the paper. We use SPAC IPO to refer to the first day of SPAC common share trading as recorded by CRSP. SPAC common shares typically start trading 30 to 90 days after the SPAC IPO when unit split becomes possible.

In all three year groups, the buy-and-hold return from the first day of SPAC common share trading to target announcement is positive (Figure 5, Panel (a)). In 2010-2017, the price appreciation is close to 0%, much lower compared to the return of CRSP indices. This year group also corresponds to relatively low liquidity, as shown in Table 5. In 2018-2020, the average SPAC return from SPAC IPO to target announcement is 9%, similar to the market performance in matched investment periods.

The return performance of SPAC common share from target announcement to deSPAC show large differentials across the three year groups(Figure 5, Panel (b)). Between 2005-2017, the return in this event window on average underperforms the CRSP indices. On the other hand, in 2018-2020, the average return during the same event window is 33%, more than 20% higher than CRSP returns in the matched investment period. As shown in the following section, the high return in 2018-2020 is not sustained after the deSPAC event.

#### **4.4 deSPAC Period Buy-and-Hold Return**

This section presents the average buy-and-hold return of SPAC common shares in the deSPAC period, in the first 30-day, 90-day, and 180-day after the deSPAC event (Figure 6). For this calculation, we assume that the investor purchases the common share on the first day of trading as the new deSPAC entity. We use CRSP data through the end of 2020; if the event window extends into 2021, an observation is dropped from the average calculation. The year groups are based on deSPAC year.

In 2006-2017, the buy-and-hold return in the deSPAC period is on average negative and the share price continues to drop over time in the first six months after deSPAC. In 2018-2020, the upward trending price momentum continues in the first 30-day after deSPAC, but reverses in the following months. For investors purchasing shares soon after deSPAC events, the average buy-and-

hold return is -7% in the first 3 months and -11% in the first six months, much lower than the corresponding market performance.

To examine if and how the BHR performance relates to company fundamentals, we calculate the average BHR separately based on the deSPAC firms' annual revenue in the fiscal year prior to going public, for the subsample of deSPAC firms that went public in 2015-2020. In 2018-2020, the average BHR between target announcement and deSPAC is 22% for firms with under \$100 million in annual revenue, and 0% for firms with more than \$100 million in revenue. In the first 30-day after the deSPAC event, the BHR are similar between the two sales size groups, at around 22%. However, in the following three to six months, the share price of firms with under \$100 million in annual revenue experiences a much larger decline. This pattern indicates that, in 2018-2020, investors may have been overly optimistic about deSPAC firms without proven sales record initially after the target announcement, and the market adjusts to a more realistic view of the firms' prospect in a few months after the deSPAC events.

#### **4.5 deSPAC vs. IPO: Buy-and-Hold Return after Going Public**

Next, we compare the return performance of deSPAC and IPO firms. We consider two purchase price possibilities for each group. For deSPAC firms, we compute the BHR from the target announcement date and the first day trading as deSPAC. For IPO firms, we compute the BHR based on the IPO offer price and the midpoint of the IPO Day price range. Figure 8 and Figure 9 summarize the average excess BHR, using CRSP Equal Weighted Index and CRSP Value Weighted Index as benchmark respectively. BHR reported in this section subtract the CRSP index BHR in matched investment period for each observation.

In 2006-2017, the deSPAC average BHR is negative throughout the first six-month after deSPAC. In 2018-2020, through the first 30-day of deSPAC, the average BHR is 35% if an investor purchase shares on the target announcement date, and 17% if share purchase is on the first day of deSPAC firm trading. In the former scenario, the BHR exceeds that of average BHR of IPO shares purchased at IPO offer price. This provides some evidence that SPACs allows broader access to new issues at a more affordable price compared to traditional IPOs. However, in the three to six months following deSPAC events, the average BHR of deSPAC firms become negative in all year groups. BHR for IPO investors who have access to the IPO offer price still experience large positive

returns compared to the benchmarks, while IPO investors who purchase shares in the secondary market on the IPO Day reap much smaller gains.

On a risk-adjusted basis, we compare the average Sharpe ratios between the deSPAC and IPO samples (Figure 10). Both groups have low Sharpe ratios, which is expected as new issues tend to have high volatility. deSPAC firms underperform IPO firms in terms of risk-adjusted buy-and-hold returns, driven by the negative BHR after deSPAC.

#### **4.6 deSPAC Follow-on Offerings**

Firms' going-public decision is driven by their financing needs and the public company disclosure requirement trade-off. We collect data of follow-on offerings of deSPAC firms using the SDC New Issue database by matching with issuer name and CUSIP to deSPAC firms. Within the 189 deSPAC firm sample, 51 deSPAC firms have subsequently issued follow-on equity offering as of February 2021 (Table 8). Among the deSPAC firms with an SEO offering, the average time between deSPAC merger and the first follow-on is 1.4 years. In 2019 and 2020, 14 out of 89 new deSPAC firms have issued SEO, on average within 7 months after the business combination.

The follow-on proceeds has increased over time. The average first follow-on offering proceeds of deSPAC firms that went public in 2020 reaches \$522 million. The over-allotment exercise ratio is on average 12% for the first follow-on, which means that the shares of these firms are in demand. The first follow-on offer price can be compared to the SPAC IPO offer price. In deSPAC business combination, the deSPAC firm's common share is valued at the SPAC IPO price (commonly \$10 per share in recent years). In 2020, the average follow-on offer price is \$19, almost double that of the SPAC share price.

Further we break down the distribution of over-allotment and follow-on offer price by underwriter reputation of the SEO. The certification effect of underwriter is prominent in deSPAC firms' SEO. During the first follow-on, both the over-allotment exercise ratio and the share offer price tend to be higher if the SEO underwriter has higher reputation.

Table 8 summarizes the average buy-and-hold return of the first follow-on offering of deSPAC firms. In 2018-2020, the first follow-on offering of deSPAC firms on average has positive buy-and-hold return in the initial 6 months after the SEO.



## 5 Retail Investment in SPACs

SPACs are often referred to as the “poor man’s private equity fund.” Direct investment in merger and acquisition transactions or younger firms is traditionally only accessible to high net-worth accredited investors through private investment vehicles. SPACs provide a similar investment opportunity to the general public in that anyone can purchase shares of the pooled investment in the secondary market, at possibly close to the offer price. Advocates for SPACs often cite this as a benefit of the instrument in the democratization of investing. Recent studies on SPACs have examined institutional holdings of SPACs using 13-F data (Klausner et al., 2020), but little is known about if and when retail investors purchase SPAC shares.

We use Robinhood users as a sample to examine retail investment interest during SPAC’s lifecycle. The sample in this section includes active SPACs and deSPAC firms that have gone public by the end of 2020. We find that retail investor interest in SPACs is low before target announcement. On the other hand, after target announcement, SPACs rapidly increase in popularity and some SPACs become the most popular tickers among Robinhood users. Popularity among Robinhood users is positively correlated with return volatility and trading volume of SPAC common shares. Before target announcement, SPACs with larger total proceeds tend to be more popular but this effect diminishes when merger details are revealed. Popularity is higher for SPACs whose identified targets are younger and have received VC financing. Between target announcement and the business combination completion, retail investors appear to favor smaller and riskier targets. Compared to the IPO sample, deSPAC firm popularity responds more strongly to lagged volatility, in the positive direction.

### 5.1 Retail Investment Popularity Measures

Robinhood is a phone-app-based zero-commission trading platform widely used by retail investors. The company’s brokerage app launched to the public in 2015 and have acquired 13 million users as of 2020. We analyze data on the number of users holding a specific ticker on Robinhood between May 2, 2018 and August 13, 2020. As discussed, SPACs are not new. However, they have remained largely obscure to the general public until a few well-known business combinations in 2019. This sample covers a pivotal period of the recent SPAC market, from which we can learn about how retail investors have caught up on a new trend in the U.S. equity market.

The Robinhood users holding data is obtained from Robintrack (<https://robintrack.net/>), which is described in detail in Section 3. The data report the number of Robinhood user holding a specific ticker at a given time. Because the holding quantity of each user is unknown, the data is best interpreted as a popularity measure. Our analysis is based on a daily frequency. Warrant and unit trading is not supported by Robinhood during the sample period, hence the analysis on SPAC is pertinent to common share only.

Over time, Robinhood user base has grown from 6 million in 2018 to 13 million in mid-2020.<sup>11</sup> Figure 11 summarizes the distribution of user holdings count on Robinhood and the number of tickers traded on the platform at each month start in the sample period. The number of tickers traded on Robinhood increased from 5,890 in May 2018 to 8,156 in August 2020. The number of users holding at each percentile has grown exponentially over time. In May 2018, at the beginning of the Robintrack data sample period, the most popular ticker has 145,510 users holding, which is about 15% of the same statistic in August 2020. The user count by ticker is highly skewed in cross-section. For example, 928,491 users hold some quantity of the most popular ticker, F (Ford), on 8/1/2020 ; whereas the users holding count is only 1,706 for the median ticker, and 15,534 at the 75th percentile. This skewed pattern holds throughout the sample period.

To compare popularity of stocks in different points of time, we account for the rapid growth of Robinhood's user base during the sample period. Our main measure of popularity is Popularity Percentile Rank ("popularity rank" in short), defined as

$$\text{Popularity Percentile Rank}_{it} = \frac{\text{Descending Rank of Number of User}_{it}}{\text{Total Number of Tickers}_t},$$

for ticker  $i$  and day  $t$ . The Popularity Rank ranges between 0 and 1 and indicates how popular a ticker  $i$  is relative to all other tickers recorded on day  $t$ . The most popular ticker on a certain day has a Popularity Rank of 1.

We link Robintrack data to CRSP data by ticker and date. CRSP provides price, trading volume, daily return, and other security-level market data. Using CRSP identifiers, we merge Robintrack to other data sources. 245 tickers associate with a SPAC (either before or after the

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<sup>11</sup>Robinhood's 13 million users as of May 2020 is more users than Schwab or E-Trade had at the end of 2019 (12.7 and 5.5 million, respectively).

business combination) are in the Robintrack sample, which include 137 SPAC and 108 deSPAC firms.<sup>12</sup> Transition from SPAC to deSPAC is captured for 21 SPACs, where Robintrack data are available for both tickers.<sup>13</sup>

Figure 12 shows the counts and average popularity of SPAC and deSPAC tickers at month start in the data. The number of SPAC and deSPAC firms both trends up over time, which is consistent with increased activity of SPACs. Throughout the sample period, SPACs on average are below median in popularity percentile rank. Until the end of 2019, SPACs on average consistently rank at the bottom 10% of all tickers traded on Robinhood in terms of number of users holding. On the other hand, the average popularity rank of deSPAC firms are between 0.5 and 0.6 throughout the 28-month sample period, which is in line with the average of 0.5 for all tickers. With respect to this statistics, deSPAC firms do not appear to be distinctive from other traded securities.

Several other Robinhood popularity measures are used in later analyses, including Popularity Percentile Rank Change, Users Increase, and Users Growth, whose definitions are detailed below.

To measure change in Popularity Percentile Rank, we define

$$\text{Popularity Percentile Rank Change}_{it} = \text{Popularity Percentile Rank}_{it} - \text{Popularity Percentile Rank}_{i,t-1},$$

for ticker  $i$  and day  $t$ . A positive Popularity Percentile Rank Change means that a ticker is more popular on Robinhood in terms of count of users holding compared to the previous day. The distribution of Popularity Percentile Rank Change is concentrated around 0. In other words, popularity remains stable over time for most observations.

We also compute change in users holding count for every ticker-date observation, in both level change and percentage growth:

$$\text{Users Increase}_{it} = \text{Number of User}_{it} - \text{Number of User}_{i,t-1};$$

$$\text{Users Growth}_{it} = \frac{\text{Number of User}_{it}}{\text{Number of User}_{i,t-1}} - 1.$$

<sup>12</sup>deSPAC count include firms that have changed ticker again after the initial business combination, but continued to associate with the same CRSP PERMNO.

<sup>13</sup>Some SPACs that completed business combination during the sample period is not in the Robintrack data. For example, Social Capital Hedosophia Holdings Corp (IPOA) started trading as Virgin Galactic(SPCE) on October 27, 2019. Robinhood user data on IPOA are unavailable.

## 5.2 Target Announcement Effect on Popularity during SPAC Period

In this section, we analyze how Robinhood retail investors respond to SPAC target announcements in the SPAC period. We show that SPACs are not favored by retail investors before target announcement; however, when a target is identified, SPACs potentially become the most popular stocks among Robinhood investors.

We compare the average popularity percentile rank, number of users holding, user increase, and user growth, before and after a target announcement during the SPAC period, where each ticker-date is one observation. Table 9 summarizes the t-test results. Before target announcements, SPACs are among the least popular tickers on Robinhood, with an average popularity rank of 0.11 (i.e. approximately the bottom 1,000 out of all 8,595 tickers). There is very little change in the number of users holding SPAC during the period before target announcement. On average, the pre-announcement daily user count change is only 3 users, which means that the users holding count of SPACs investor is both low and stable.

On the other hand, SPACs significantly increase in popularity after target announcement. The average user holding count increase from 45 before target announcement to 2,167 after announcement. The percentage growth in user counts is on average 33% per day after target announcement, which means that SPACs continuously attract new investors after the announcement, compounding at 1.33 times each day. While the average SPAC experiences a moderate improvement in popularity rank, the increase is considerable among top movers. For example, the largest change in popularity rank is within one day of Diamondpeak Holding (DPHC)'s announcement of Lordstown Motors (RIDE) as the target. Upon the news, DPHC moves from the 20th percentile to the 94th percentile in terms of users holding count in a single day.

The target announcement effect is also realized very quickly. Among 56 SPACs that announced a target during the sample period, the largest one-day change in popularity rank occurs within 2 days post-announcement for 28 cases (Table 10). The magnitude of one-day increase in popularity rank is also on average higher when the reaction time is shorter.

### 5.3 Effect of Market Variables on Popularity during SPAC Period

We use panel regressions to estimate the relationship between Robinhood popularity and market variables, while controlling for target announcement status,

$$\text{RH Popularity Rank}_{it} = \beta_0 + \mathbf{Market\ Variables}_{it}\beta_1 + \beta_2\text{POST\_ANN}_{it} + \alpha_i + \gamma_i \quad (2)$$

Table 11 reports the linear regression estimation results. The dependent variable is the Robinhood popularity rank. Market variables include market capitalization, lagged return, lagged volatility, trading volume moving average, and the Amihud illiquidity measure.<sup>14</sup> Both ticker ( $\alpha$ ) and year-month ( $\gamma$ ) fixed effects are controlled for. SPAC popularity increases when lagged volatility, trading volume, and liquidity are high. The increase in popularity after SPAC target announcement remains large in magnitude and statistically significant.

### 5.4 Effect SPAC Features on Popularity during SPAC Period

SPAC common shares have similar intrinsic value lower bounds prior to the deSPAC event, as shares can be redeemed at the pro rata trust value. Still, there is variability of SPAC common share popularity among retail investors on Robinhood, even before target announcement. In this section, we investigate how SPAC IPO features affect retail investor interests.

How do retail investors choose which SPACs to invest in? In practice, investors often consider sponsor qualifications and track records to gauge the upside potential of the share price. Because sponsor qualities are highly idiosyncratic and difficult to measure, we use SPAC IPO underwriter reputation as a proxy, under the assumption that reputable underwriter are more likely to work with better sponsors due to either network effects or screening during the due diligence process.

The identity of SPAC IPO lead underwriter(s) is from the SDC New Issue data base. To quantify underwriter reputation, we use the underwriter rank maintained by Loughran and Ritter (2004). The rank is based on a 1 to 9 scale (9 being the most reputable) and first introduced in Carter and Manaster (1990). Underwriter rank tends to be stable over time; however, when there is change in rank over time, we take the time average rank. When there are multiple lead underwriters, we take the rank of the highest-ranking joint bookrunner, following Loughran and

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<sup>14</sup>See variable definitions in Table 6.

Ritter (2004). Underwriter rank greater than 5 is defined as the high rank group in the following analyses.

Table 12 shows the difference in the average popularity measures for high and low underwriter reputation groups. SPACs with higher underwriter reputation are more popular in terms of Robinhood popularity rank and users holding, both before and after target announcement. The target announcement effects are also larger for SPACs with higher-ranked underwriters. This implies that SPACs with more reputable underwriters, and by proxy more reputable sponsors, are more likely to strike merger agreements that attract retail investors.

We further examine the impact of SPAC features on popularity in a regression framework. We consider three SPAC features: underwriter rank, SPAC proceeds raised, and overallotment option exercise ratio at SPAC IPO. Underwriter rank is a proxy of SPAC sponsor quality and potential to negotiate a desirable merger agreement. SPAC IPO proceeds raised measures the size of the investment vehicle. Overallotment option exercise ratio indicates the initial demand of the SPAC IPOs. While the intrinsic values of SPAC units vary from offerings due to the quantities of warrants and rights attached, the valuation of SPAC common share is more straightforward. Before business combination, SPAC common shares are essentially convertible notes, where each share has claim to the pro rata trust value and earns risk-free rate. Demand in SPAC common shares should be based on expectation of the future business combination. Another SPAC feature relevant to the common share valuation is percentage of SPAC proceeds reserved in trust. In recent years, SPAC trust typically has enough reserve to cover at least 100% of public share redemption. Hence, there is little variation in SPAC’s ability to redeem public share in this sample.

In pooled linear regressions, we estimate the effects of underwriter rank, log SPAC IPO proceeds, and IPO overallotment exercise ratio on popularity (Table 13),

$$\text{RH Popularity Rank}_i = \beta_0 + \mathbf{SPAC Features}_i\beta_1 + \mathbf{Market Variables}_i\beta_2 + \gamma_i \quad (3)$$

Control variables include year-month dummy variables ( $\gamma$ ) and all the previously used market variables except for market capitalization, which is inherently correlated with the SPAC proceeds during the SPAC period. The estimation is repeated on three samples:(1) the full sample of SPAC tickers in the Robintrack data (Columns 1-2), (2) “before target announcement” subsample (Columns 3-4),

(3) “after target announcement” subsample (Column 5).

In the full sample, popularity is mainly driven by the market variables and target announcement status. Coefficient estimates on underwriter reputation and SPAC proceeds size are small in magnitude and not statistically significant. Overallotment is negatively correlated with popularity. Before target announcement, SPACs with larger IPO proceeds are more popular, but this effect no longer exists after target announcement. This is sensible because the merger transaction size can be much larger than the initial SPAC proceeds through private placements at the merger. The coefficient estimate on lagged volatility is large and statistically significant in the full sample and the post-announcement period, but not before target announcement.

### **5.5 Effect of Target Company Characteristics on Popularity during SPAC Period**

After the definitive merger agreement, SPAC common share price changes based on the market’s evaluation of the target company and the transaction terms. In this section, we examine whether retail interest in SPACs relates to characteristics of the target companies.

We merge the Robintrack data with the ex ante firm characteristic data. Figure 13 shows the average Robinhood popularity of SPACs by quartiles of target firm total asset, total revenue, and net income in the fiscal year prior to going public. Quartiles are determined by firms that went public through either IPO or deSPAC in 2015-2020. Before target announcement, popularity ranks are low and (unsurprisingly) similar across the fundamental quartiles, which demonstrates that target characteristics cannot be predicted collectively by the crowd. Between target announcement and deSPAC, SPACs whose target have below median asset, sale, and net income are on average more popular. In the deSPAC period, the popularity of the below median firms drop. Previously, we have shown that the buy-and-hold return between target announcement and deSPAC is higher for lower-revenue targets than higher-revenue targets, but quickly reverses after deSPAC (See Figure 7 in Section 4). The popularity comparison across quartiles is another indicative evidence that, for a short period of time, investors are enthusiastic about firms without strong track records in fundamentals, but possibly show high future growth potential.

To control for market variables and SPAC features, we estimate the impact of firm character-

istics on popularity using the post-announcement subsample in pooled linear regressions,

$$\text{RH Popularity Rank}_i = \beta_0 + \text{Target Firm Characteristics}_i\beta_1 + \text{SPAC Features}_i\beta_2 + \text{Market Variables}_i\beta_3 + \gamma_i + \delta_i \quad (4)$$

Table 14 reports regression estimates. The control variables include SPAC IPO features and market variables, year-month dummy variables ( $\gamma$ ), and FF17 industry dummy variables ( $\delta$ ). We first consider the dummy variable "Popularity Above Median" as the dependent variable, where the indicator is equal to 1 if the company's popularity is higher than the median of this subsample (Columns 1-2). A SPAC's popularity is more likely to be above median when its target is larger in size (measured by log total asset or log total sales), younger, and has received VC financing. Target company recent year profitability (measured by net income/asset) does not impact SPAC popularity. Then, we use the popularity rank as the dependent variable (Column 3-4). In this specification, none of the target company characteristics have statistically significant impact on SPAC's popularity among retail investors. The market variables remain as robust determinants of the popularity rank, where coefficient sign/magnitude and statistical significance are similar to previous regressions. Between target announcement and the deSPAC event, SPAC shares are more popular when lagged trading volume, volatility, and liquidity are high.

## 5.6 Popularity after Going Public: deSPAC vs. IPO

Lastly, we compare the RH popularity after firms go public, through either SPAC or IPO. With SPAC being a buzzword in the financial market during the sample period, the deSPAC status is potentially attention-inducing for retail investors. We examine how the going public method may affect retail investor interests in pooled linear regressions using the sample of firms that went public via SPAC or IPO in 2016-2020,

$$\begin{aligned} \text{RH Popularity Rank}_i = & \beta_0 + \beta_1\text{DESPAC}_i + \text{DESPAC*Target Firm Characteristics}_i\beta_2 + \\ & \text{DESPAC*SPAC Features}_i\beta_3 + \text{DESPAC*Market Variables}_i\beta_4 + \\ & \text{Target Firm Characteristics}_i\beta_5 + \text{SPAC Features}_i\beta_6 + \text{Market Variables}_i\beta_7 + \gamma_i + \delta_i \end{aligned} \quad (5)$$

The DESPAC indicator variable is 1 if firm  $i$  went public via SPAC. Then we include target firm characteristics, SPAC features, and market variables, as well as interaction terms with the DESPAC



indicator. Year-month dummy variables ( $\gamma$ ) and FF17 industry dummy variables ( $\delta$ ) are included as control variables. Table 15 reports the estimates. We repeat the estimation on five periods after going public, from the first 90 days to the first 2 years, and the full sample with Robintrack data. In general, the going public method does not affect the RH popularity rank. Overall, retail investors favor newly public firms with lower ex ante profitability (as measured by net income/total assets), higher revenue (sales/total assets), lower R&D expenses (R&D/total assets). Trading volume positively correlate with popularity. Interestingly, return volatility does not increase retail interest for the IPO firms, but does so strongly for the deSPAC firms.

Thus far, return volatility is consistently positively correlated with SPAC popularity during both the post-announcement SPAC period and the deSPAC period. We estimate determinants of return volatility in the same sample and pooled regression framework as in the previous equation,

$$\begin{aligned} \text{Return Volatility}_i = & \beta_0 + \beta_1 \text{L.RH Popularity}_i + \beta_2 \text{DESPAC}_i + \\ & \text{DESPAC*Target Firm Characteristics}_i \beta_3 + \\ & \text{DESPAC*SPAC Features}_i \beta_4 + \text{DESPAC*Market Variables}_i \beta_5 + \\ & \text{Target Firm Characteristics}_i \beta_6 + \text{SPAC Features}_i \beta_7 + \text{Market Variables}_i \beta_8 + \gamma_i + \delta_i \end{aligned} \quad (6)$$

Table 16 reports the estimates. In the first two years of going public, deSPAC firms exhibits higher volatility than the IPO firms. Coefficients on the firm characteristics variable show that, for newly public firms, return volatility is higher when the firm ex ante has lower total assets, higher profitability, lower revenue, is younger and backed by VC capital. Within the first year of going public, lagged Robinhood popularity rank positively correlates with return volatility. This indicates that retail investor attention has the potential to generate significant market movement for shares of newly public companies.

## 6 Conclusion

SPACs have recently become a common alternative to the traditional IPO process for firms to go public. We compare firms going public via SPAC versus traditional IPO in terms of (1) ex ante firm characteristics before going public, (2) common share market performance, and (3) retail investment interest. We use a long sample period from 2003-2020 and compare several sub-samples throughout different stages of the modern history of the SPAC market.

Firms going public via SPAC tend to be younger, more likely to have received VC financing, have lower current ratio, and less likely to be in a high-tech industry. Between SPAC IPO and six months after deSPAC, SPAC common shares have the highest liquidity in the period between target announcement and the initial business combination completion. In 2018-2020, the average buy-and-hold return on SPAC common share outperform the market during the SPAC period, and is comparable in magnitude to the buy-and-hold return of IPO shares purchased at IPO offer price. After the deSPAC merger completion, the run-up in share price does not sustain. In the deSPAC period, the decline in share price is larger in magnitude for riskier firms, characterized by low revenue prior to going public. Compared to IPO firms, deSPAC firm shares on average generate larger losses after going public. Retail interest in SPAC investment is weak before target announcement, but rise quickly and strongly after the merger target is identified. SPAC popularity among retail investors is primarily driven by stock price volatility and trading volume, less so by target firm characteristics or SPAC IPO features. Compared to IPO firms, deSPAC firms are more likely to attract investors when their stock has high volatility.

The SPAC market continues to grow and evolve in real time. In 2021, new SPAC IPO counts and proceeds exceeded the total of 2020 in just three month's time. While new issues have slowed down in the second quarter of 2021, over 300 SPACs are active in the pipeline to bring more companies public. Many private companies will be approached by SPACs seeking to complete a merger before the deadline. Based on evidence through 2020, it is clear that going public via SPAC merger has become a standard option as private firms consider an exit. While the negotiation with SPAC sponsors appears to be a simpler process the traditional IPO, firms should first and foremost determine if they are ready to become a public company, before engaging in a SPAC transaction. We show that SPAC investing is associated with high volatility and risk-seeking behavior of investors in 2018-2020. To support share prices, newly public companies need to carefully manage investor expectation. To the contrary, many SPACs have issued overly optimistic projections at the time of the merger and later fail to meet expectations. We anticipate that, over time, both investors and deSPAC firms will become more rational and realistic, such that the market will develop into a more sustainable state.

Figure 1: SPAC IPO Activities 2003-2020

Figure 1 reports the SPAC IPO activities from 2003 through 2020. The sample includes all SPACs listed in U.S. exchanges, including OTC and Pink Sheets.

The orange bars show the total proceeds of SPAC IPO (in \$ billion) each year. The purple line shows the number of SPAC IPO each year. The gray line shows the average SPAC IPO proceeds (in \$ million) in each year. Proceeds include overallotment.

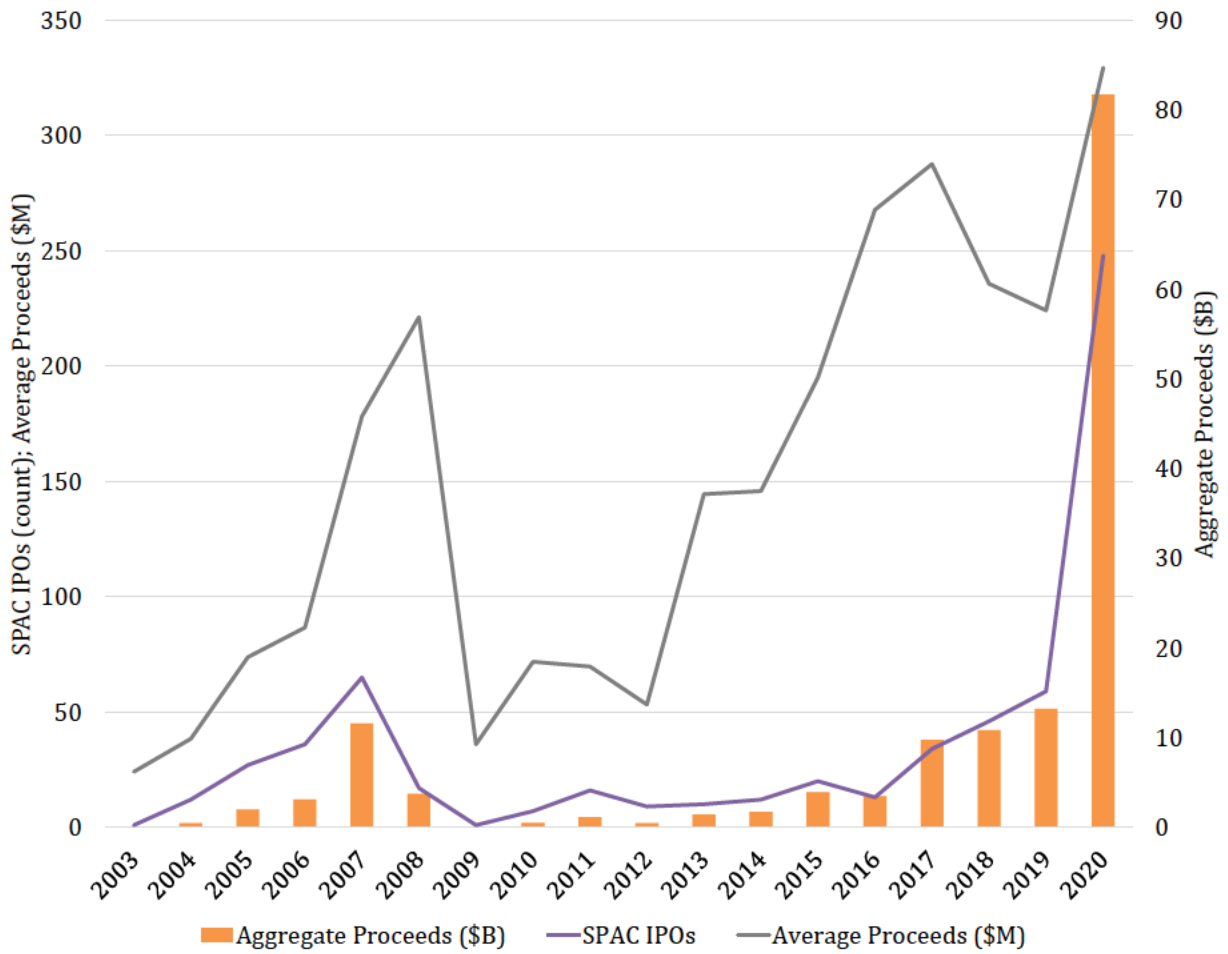


Figure 2: SPAC Lifecycle Status by IPO Year

Figure 2 reports the SPAC status by IPO year. Status is as of December 31, 2020. The sample includes all SPACs listed in U.S. exchanges, including OTC and Pink Sheets. "Merger complete" refers to SPACs that have completed a deSPAC initial business combination. "Closed" refers to SPACs that have liquidated or otherwise ceased to exist as a blank check company. "Active" refers to SPACs that are searching for business combination targets.

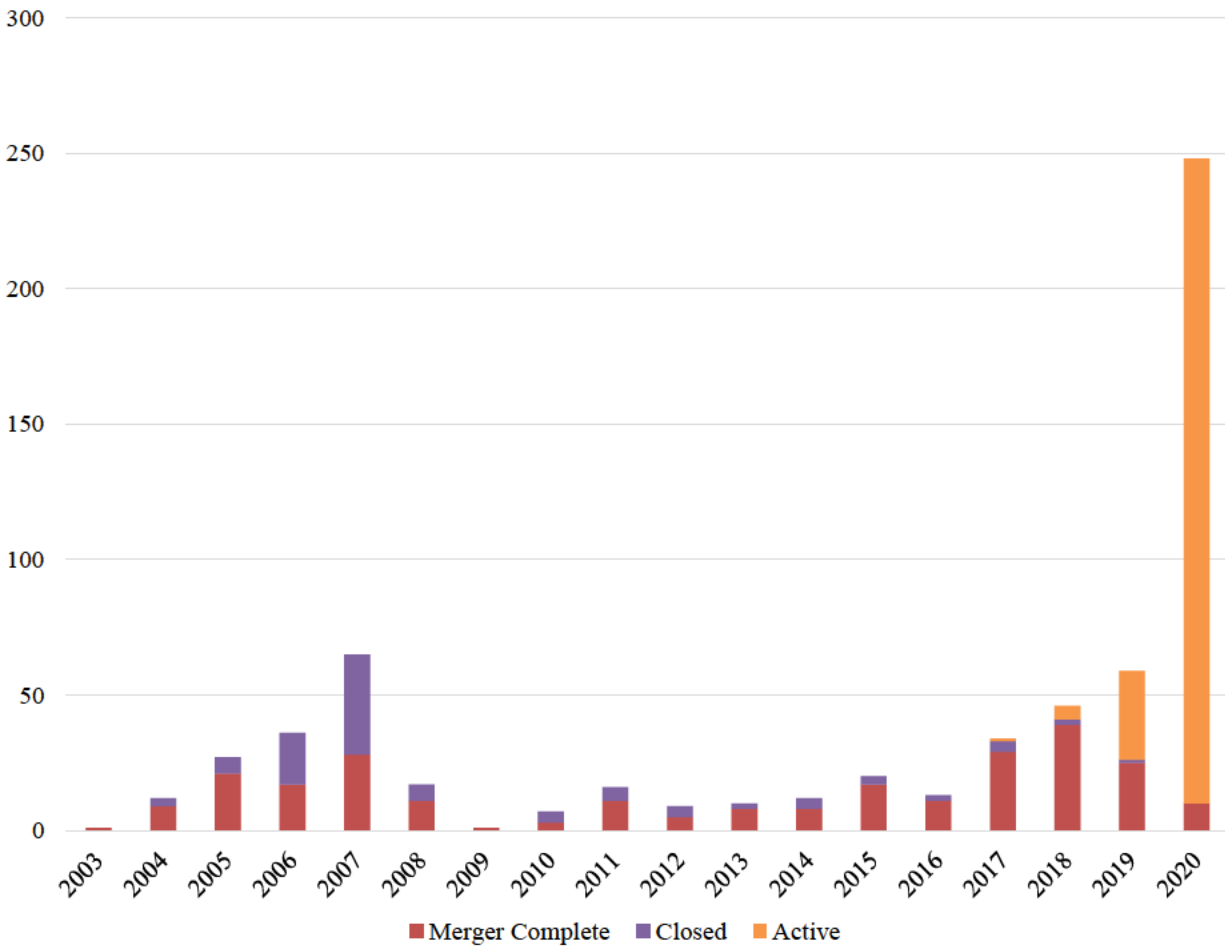


Figure 3: SPAC Listing Exchange by IPO Year

Figure 3 reports the listing exchange of SPACs at the time of SPAC IPO by IPO year. The sample includes all U.S. listed SPACs issued through 2020.

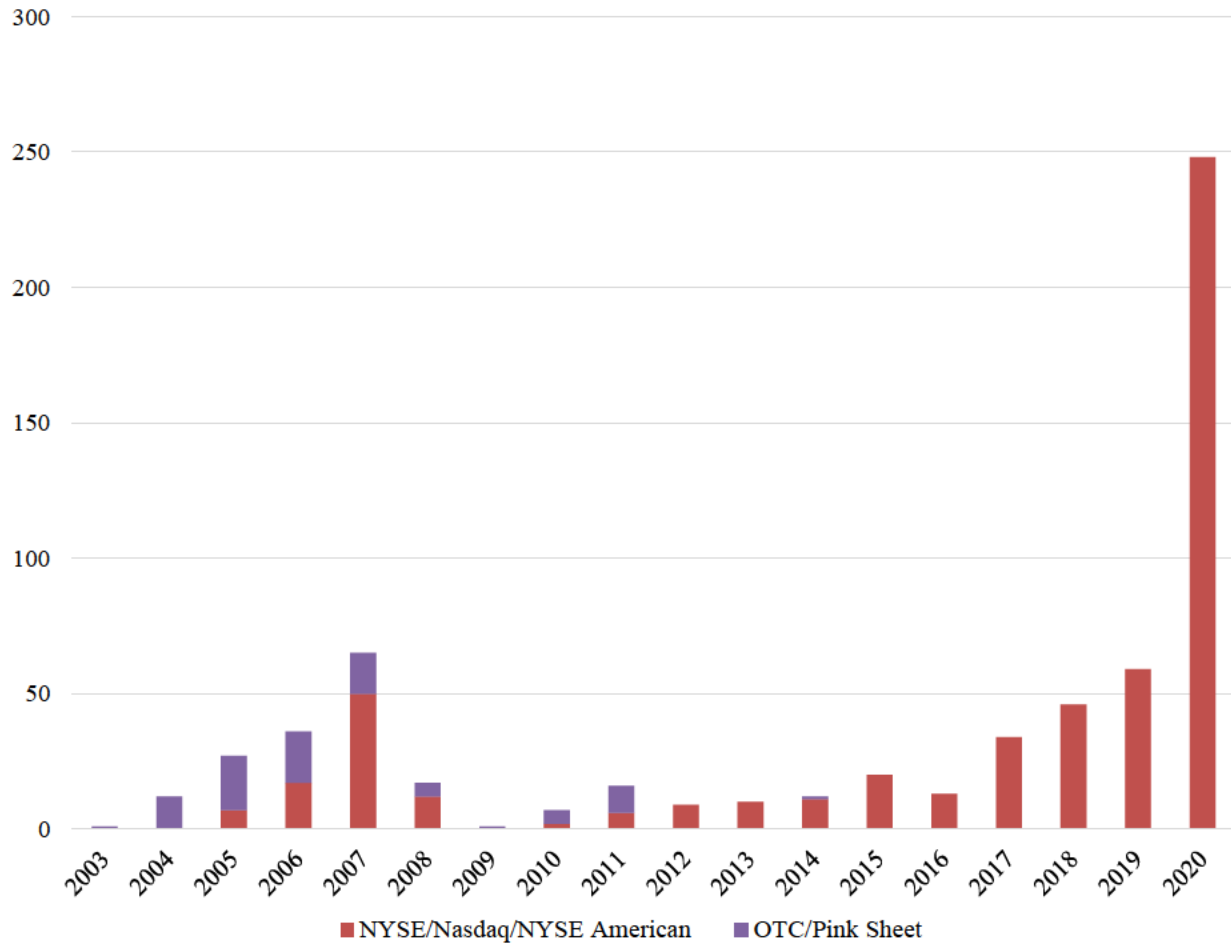


Table 1: Characteristics of SPAC IPOs

Table 1 summarizes the SPAC IPO features for the sample of 189 SPACs traded on major exchanges that have completed the deSPAC initial business combination by the end of 2020.

IPO proceeds include over-allotment exercise option. Sponsor capital refers to the upfront cash investments made by SPAC sponsors, which is intended to cover initial underwriting fees and other operating costs while maintaining trust value. Warrant per unit refers to the number of common shares issuable upon exercise of public warrants included in a SPAC unit. “In-the-money” warrant means that the exercise price of the warrant for one common share is lower than the unit offer price.

Summary Statistics by SPAC IPO Year							
Year	Count	IPO Proceeds (\$M)	Sponsor Capital /Proceeds	Average			Proportion
	by SPAC IPO Year			Underwriter Rank	% of Proceeds in Trust	Warrant Per Unit	Warrant In-the- Money
2005	6	84	0%	4	93%	1.17	100%
2006	6	202	2%	6	95%	1.33	100%
2007	18	279	2%	8	98%	0.97	100%
2008	6	213	3%	7	98%	1.00	100%
2009	0						
2010	1	46	5%	3	101%	1.00	100%
2011	3	104	4%	7	101%	1.00	0%
2012	5	49	6%	6	102%	1.00	40%
2013	7	170	5%	6	102%	0.50	14%
2014	8	160	4%	7	100%	0.44	0%
2015	17	211	4%	7	101%	0.52	0%
2016	11	281	4%	7	101%	0.59	0%
2017	28	283	3%	6	100%	0.51	0%
2018	39	251	3%	6	100%	0.69	0%
2019	24	221	3%	6	100%	0.51	0%
2020	10	343	3%	8	100%	0.38	0%
2005-2009	36	223	2%	7	97%	1.07	100%
2010-2017	80	221	4%	7	101%	0.57	5%
2018-2020	73	254	3%	7	100%	0.59	0%

Figure 4: Number of Firms Going Public via IPOs and SPACs, 2006-2020

Figure 4 reports the yearly count of operating firms going public in the U.S. through traditional IPO and SPAC between 2006 and 2020. The sample includes companies traded on three major exchanges - NYSE, Nasdaq, or AMEX (NYSE American). The IPO sample excludes offerings with offer price below \$5, ADRs, units, REITs, closed-end funds, banks, ETFs, SPACs, direct listings, and stocks not included in CRSP such as OTC issues.

A firm is considered active if the company has a current ticker in the EDGAR database as of May 18, 2021. The average time from going public to delisting among inactive firms is 5.4 years for the IPO sample and 4.0 years for the deSPAC sample.

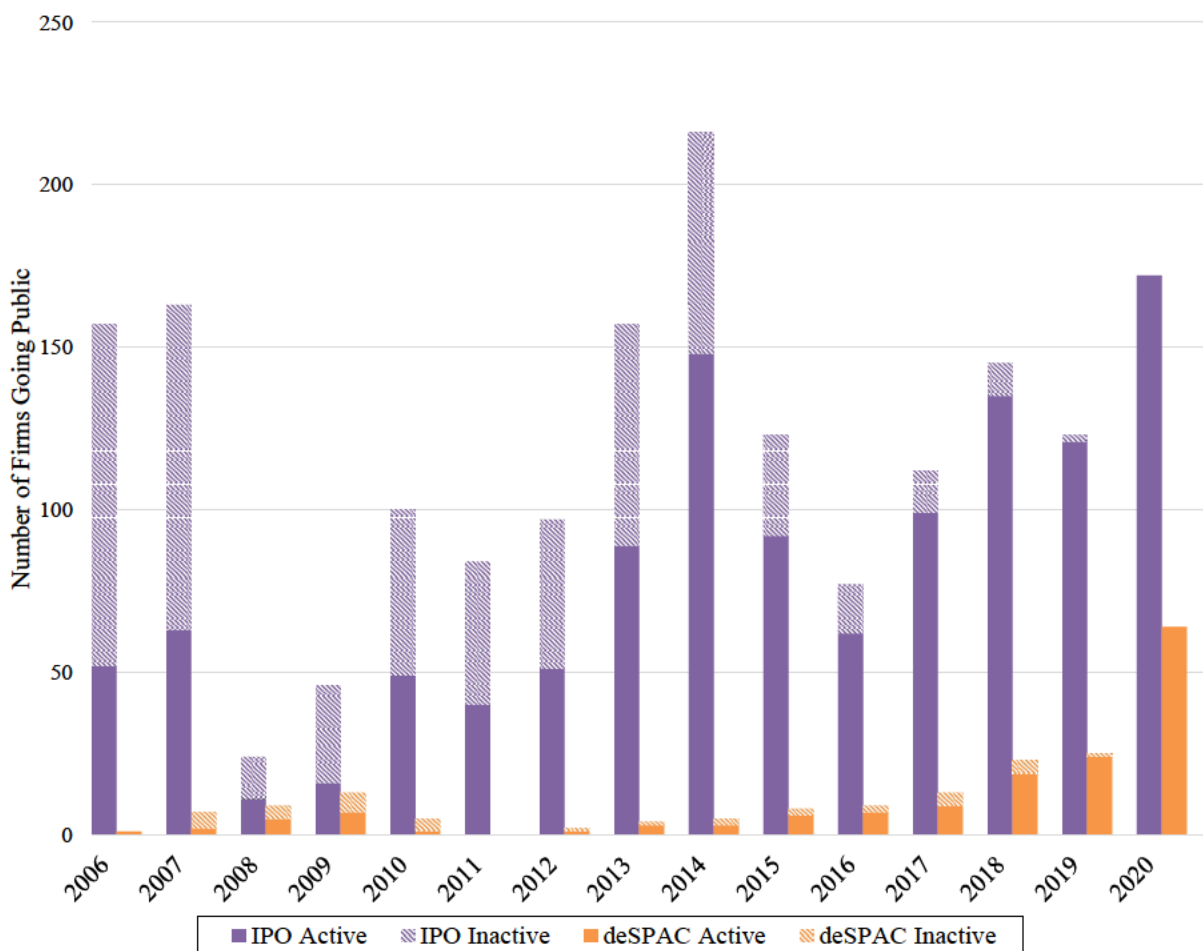


Table 2: Characteristics of IPO and deSPAC Firms Before Going Public, 2015-2020

Table 2 summarizes characteristics of firms that went public via traditional IPO or deSPAC in 2015-2020. Financial data are annual figures in the fiscal year immediately prior to going public. Financial data, industry SIC codes, and employment data are from Compustat and company 10-K filings. Industry categories are based on SIC codes, following definitions in Puri and Zarutskie (2012). Company age of the IPO sample is based on the founding dates of Loughran and Ritter (2004). VC backing status of the IPO sample is based on SDC. Company age and VC backing status of the deSPAC sample are collected by the authors using VentureXpert, Crunchbase, and firm websites.

	IPO			deSPAC			IPO - deSPAC	
	Count	Mean	Std Dev	Count	Mean	Std Dev	Mean Diff	
<b>Firm Characteristics</b>								
Total Assets (\$M)	744	1,268	9,525	129	787	2,767	481	
Total Current Assets (\$M)	709	235	725	118	124	291	111	
Total Liabilities (\$M)	744	1,056	8,797	129	675	2,522	381	
Total Current Liabilities (\$M)	709	175	628	118	123	231	52	
Long-Term Debt (\$M)	741	377	1,623	122	266	722	111	
Total Revenue (\$M)	737	576	2,805	130	390	845	186	
Net Income (\$M)	737	-15	145	129	-20	73	5	
Capital Expenditure (\$M)	737	26	91	129	18	42	8	
R&D Expenditure (\$M)	574	33	90	52	17	28	16	
Current Ratio	709	4.4	7	118	1.9	3	2.5	***
Long-Term Debt/Total Assets	739	0.3	1	121	0.3	0	0.0	
R&D/Total Assets	574	0.8	5	51	1.9	10	-1.1	
Net Income/Total Assets	737	-1.1	8	127	-2.9	29	1.8	
Capx/Total Assets	737	0.04	0	127	0.05	0	-0.01	
Employment (000's)	399	3	15	103	1	3	1	
Company Age	739	15.9	21	142	16.4	22	-0.5	
VC Backed	747	49%		142	37%		11%	**
<b>Industry Distribution</b>								
1 Computer	749	19%		142	17%		2%	
2 Biotech/Medical	749	48%		142	11%		38%	***
3 Electronics	749	2%		142	4%		-1%	
4 Telecom	749	1%		142	1%		0%	
5 Consumer Goods	749	4%		142	4%		0%	
6 Finance	749	6%		142	9%		-3%	
7 Business Services	749	0%		142	1%		-1%	
8 Industrial Goods	749	9%		142	21%		-12%	***
9 Other	749	9%		142	32%		-22%	***
High Tech Industry (1-4)	749	71%		142	32%		38%	***



Table 3: Definitions of Regression Variables

Table 6 describes definitions of regression variables.

Variable Name	Definition
Panel A: Firm Characteristics	
LN_TOTAL_ASSET	ln(total asset in \$M)
LN_SALES	ln(total revenue in \$M)
CURRENT_RATIO	current asset/current liability
LT_DEBT/ASSET	long term debt/total asset
NET_INCOME/ASSET	net income/total asset
CAPX_/ASSET	capital expenditure/total asset
AGE	company age in the going public year
R&D_I	indicator variable that R&D expenditure statistics are available
R&D/ASSET	R&D expenditure/total asset. =0 if R&D_I==0.
HIGH_TECH	=1 if high-tech industry. Dummy variable based on SIC codes, following definitions in Puri and Zarutskie (2012).
VC_BACKED	=1 if company has received VC financing prior to going public
UWRANK	IPO or SPAC IPO bookrunner rank. Rank values follow Loughran and Ritter (2004).
Notes:	
Variables are winsorized at 1% and 99% before log transformation.	
Ratios are winsorized at 1% and 99%.	
Panel B: Market Variables	
LN_MARKET_CAP	ln(market capitalization in \$M)
LN_RET	ln(1+return)
VOLATILITY	standard deviation of the most recent 20-day log returns
LN_MA_VOL	ln(1+20-day moving average trading volume)
LN_AMIHUDD_ILIQ	ln(Amihud illiquidity measure of the calendar month)
POST_ANN.	=1 if after target announcement
Panel C: SPAC IPO Features	
LN_SPAC_PROCEEDS	ln(SPAC IPO Proceeds in \$M)
OVERALLOTMENT	IPO overallocation option exercise ratio



Panel B: 2015-2020		Dependent Variable: Going Public via SPAC=1								
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)	(10)
	Probit	Probit	Probit	Probit	Probit	Probit	Probit	Probit	Logit	Linear
LN_TOTAL_ASSET	-0.026 (-0.65)		-0.054 (-1.36)	-0.059 (-1.46)	-0.055 (-1.35)	0.075 (1.46)	0.088 (1.58)	0.104* (1.95)	0.186* (1.74)	0.014 (1.42)
LN_SALES		0.014 (0.39)								
CURRENT_RATIO	-0.078** (-2.24)	-0.074** (-2.02)	-0.047* (-1.75)	-0.042 (-1.60)	-0.057* (-1.95)	-0.041* (-1.68)	-0.051* (-1.93)	-0.049* (-1.92)	-0.102 (-1.51)	-0.003*** (-2.59)
LT_DEBT/ASSET	0.300** (1.97)	0.268* (1.75)	0.262 (1.63)	0.250 (1.61)	0.325* (1.86)	0.341** (2.21)	0.421** (2.57)	0.374** (2.37)	0.720** (2.34)	0.063** (2.15)
NET_INCOME/ASSET	0.178** (2.14)	0.141** (2.00)	0.091 (1.42)	0.087 (1.34)	0.071 (1.03)	0.112 (1.63)	0.082 (1.04)	0.046 (0.68)	0.169 (1.05)	0.019** (1.97)
CAPX/ASSET	-0.482 (-0.53)	-0.451 (-0.49)	-0.909 (-0.97)	-0.992 (-1.06)	-1.046 (-1.08)	-0.317 (-0.33)	-0.633 (-0.62)	-0.405 (-0.42)	-1.372 (-0.72)	-0.128 (-0.66)
AGE	-0.005 (-1.11)	-0.006 (-1.37)	-0.005 (-1.11)	-0.006 (-1.22)	-0.004 (-0.84)	-0.007 (-1.35)	-0.006 (-1.20)	-0.007 (-1.53)	-0.012 (-1.29)	-0.002 (-1.56)
R&D_I			-1.037*** (-6.00)	-0.872*** (-4.66)	-1.193*** (-6.41)	-0.901*** (-5.08)	-0.926*** (-4.55)	-0.757*** (-4.19)	-1.707*** (-4.11)	-0.221*** (-4.41)
R&D/ASSET			-0.046 (-0.26)	-0.020 (-0.11)	-0.125 (-0.64)	0.072 (0.44)	0.007 (0.04)	0.029 (0.18)	0.072 (0.20)	0.022 (1.02)
HIGH_TECH				-0.551*** (-2.67)			-0.605*** (-2.63)	-0.841*** (-4.48)	-1.081** (-2.46)	-0.162*** (-2.78)
VC_BACKED					0.551*** (3.42)		0.793*** (4.30)	0.787*** (4.46)	1.511*** (4.28)	0.148*** (5.27)
UWRANK						-0.204*** (-4.74)	-0.227*** (-4.96)	-0.226*** (-5.14)	-0.424*** (-4.71)	-0.049*** (-4.99)
FF17 Industry FE	x	x	x	x	x	x	x		x	x
Going Public Year FE	x	x	x	x	x	x	x	x	x	x
Constant	-0.856** (-2.03)	-1.031** (-2.38)	-0.341 (-0.74)	-0.319 (-0.70)	-0.379 (-0.80)	0.511 (1.05)	0.584 (1.17)	0.397 (1.07)	1.028 (1.07)	0.649*** (5.19)
Observations	804	804	804	804	804	793	793	798	793	798
Pseudo $R^2$	0.194	0.194	0.263	0.274	0.283	0.306	0.347	0.311	0.349	
Adjusted $R^2$										0.273

*t* statistics in parentheses

\*  $p < 0.10$ , \*\*  $p < 0.05$ , \*\*\*  $p < 0.01$

Table 5: Average Liquidity Measures During SPACs Lifecycle

Table 5 reports the average liquidity measure of SPACs during five event windows of the SPAC lifecycle. The sample consists of 189 SPACs traded on major exchanges that have completed the deSPAC initial business combination (IBC) through 2020. Mean difference t-tests use period (2) “SPAC period, after target announcement” as the comparison benchmark.

Year	Event Window Mean					Mean Difference Between Event Windows					
	SPAC after Target Ann. (2)	SPAC after Target Ann. (1)	deSPAC 30-Day (3)	deSPAC 90-Day (4)	deSPAC 180-Day (5)	(2)-(1)	(2)-(3)	(2)-(4)	(2)-(5)		
Panel A: Amihud Illiquidity ( $\times 10^5$ )											
2005-2009	3.36	10.22	2.01	7.45	10.92	-6.85 (-3.69)	*** (0.97)	1.35 (-1.29)	-4.08 (-1.29)	-7.6 (-1.21)	
2010-2017	52.41	68.61	70.96	99.82	72.36	-16.20 (-1.07)		-18.55 (-0.48)	-47.41 (-0.99)	-20.0 (-0.56)	
2018-2020	1.28	22.77	10.24	7.65	8.03	-21.49 (-6.30)	*** (-1.41)	-8.96 (-1.50)	-6.36 (-1.50)	-6.8 (-1.34)	
Panel B: Bid-Ask Spread											
2005-2009	0.04	0.06	0.08	0.08	0.08	-0.01 (-2.66)	*** (-4.20)	-0.03 (-4.85)	*** (-4.85)	-0.03 (-4.35)	***
2010-2017	0.23	0.39	0.44	0.49	0.34	-0.16 (-2.51)	** (-1.59)	-0.21 (-1.68)	* (-1.68)	-0.11 (-1.09)	
2018-2020	0.07	0.15	0.77	0.48	0.32	-0.08 (-3.58)	*** (-1.36)	-0.70 (-1.90)	* (-1.90)	-0.25 (-2.28)	**
Panel C: Daily Turnover											
2005-2009	0.011	0.003	0.005	0.004	0.004	0.008 (4.76)	*** (2.69)	0.006 (3.68)	*** (3.68)	0.007 (3.73)	***
2010-2017	0.009	0.002	0.005	0.004	0.004	0.007 (4.17)	*** (1.39)	0.004 (1.78)	* (1.78)	0.005 (1.99)	**
2018-2020	0.039	0.004	0.040	0.016	0.015	0.034 (5.54)	*** (-0.06)	-0.001 (2.67)	*** (2.67)	0.024 (3.11)	***

\*  $p < 0.10$ , \*\*  $p < 0.05$ , \*\*\*  $p < 0.01$ . Mean difference test  $t$  statistics in parentheses. Definitions: For SPAC  $i$  in a holding period of  $T$  trading days,

$$\text{Amihud Illiquidity Measure}_i = \frac{1}{T} \sum_{t=1}^T \frac{|\text{Daily Return}|_t}{\text{Price}_t * (1 + \text{Trading Volume})_t}; \text{Bid-Ask Spread}_i = \frac{1}{T} \sum_{t=1}^T (\text{Ask}_t - \text{Bid}_t); \text{Daily Turnover}_i = \frac{1}{T} \sum_{t=1}^T \frac{\text{Trading Volume}_t}{\text{Shares Outstanding}_t}$$

Table 6: Average Time to Major Events during the SPAC Period

Table 6 summarizes the average time taken by a SPAC to announce merger agreement and complete the initial business combination. The sample is 189 SPACs traded on major exchanges that have completed the deSPAC initial business combination (IBC) through 2020.

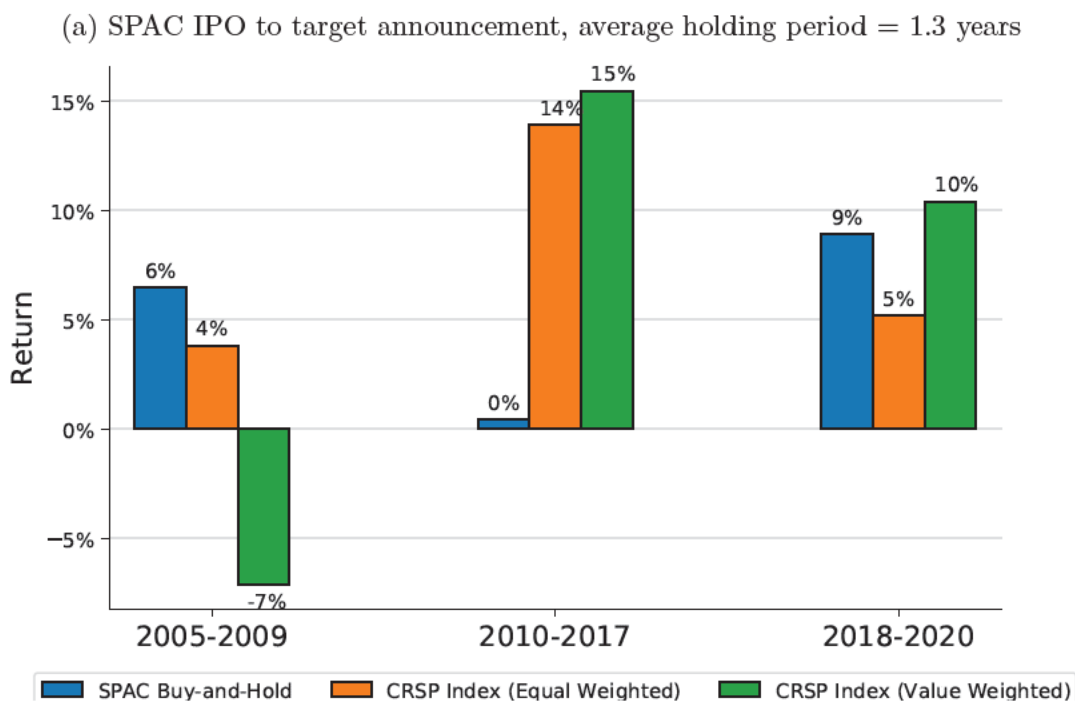
Year	Count		Average Years Between (by SPAC IPO Year)		
	by SPAC IPO Year	by deSPAC Year	SPAC IPO and Target Announcement	Target Announcement and deSPAC	SPAC IPO and deSPAC
2005	6	0	1.5	0.7	2.2
2006	6	1	1.3	0.3	1.5
2007	18	8	1.2	0.4	1.7
2008	6	9	1.3	0.4	1.7
2009	0	13	-	-	-
2010	1	5	1.7	0.3	2.0
2011	3	0	1.4	0.3	1.7
2012	5	2	1.4	0.3	1.7
2013	7	4	1.3	0.3	1.6
2014	8	5	1.4	0.4	1.8
2015	17	8	1.6	0.5	2.1
2016	11	9	1.5	0.4	1.9
2017	28	13	1.3	0.5	1.8
2018	39	23	1.4	0.4	1.8
2019	24	25	1.0	0.4	1.4
2020	10	64	0.4	0.3	0.6
2005-2020	189	189	1.3	0.4	1.7

Figure 5: Common Share Buy-and-Hold Return during the SPAC Period

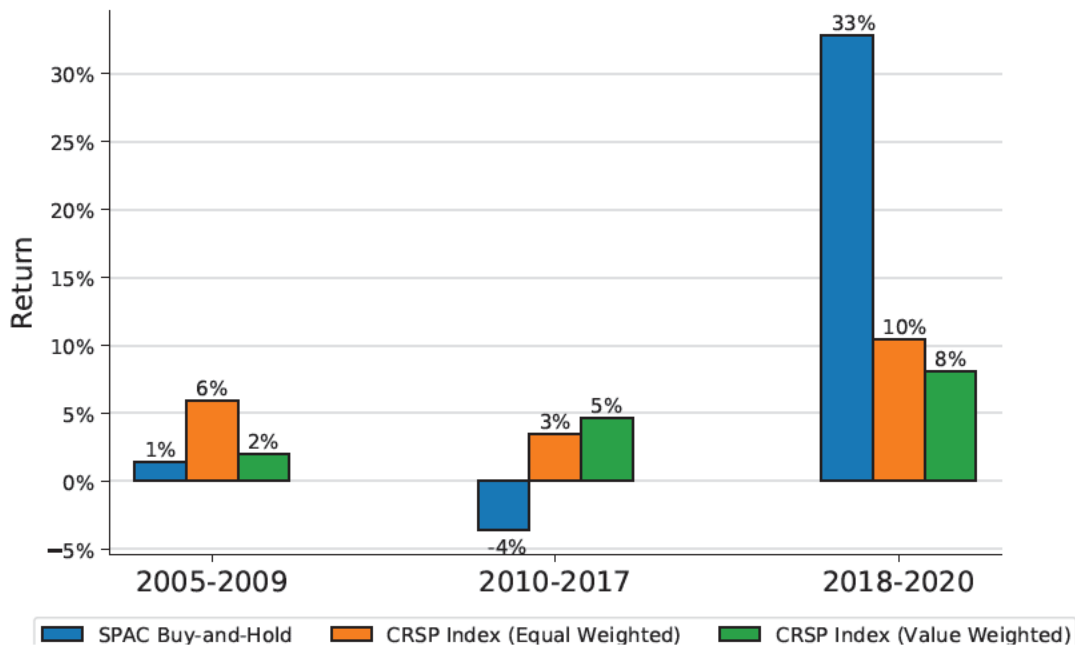
Figure 6 summarizes the SPAC common share average buy-and-hold return during three event windows of the SPAC period. Total return includes capital gain and ordinary dividend. The sample consists of 189 SPACs traded on major exchanges that have completed the deSPAC initial business combination through 2020. Year groups are based on SPAC IPO year.

The buy-and-hold returns are calculated based on the average of the highest and lowest trading prices on the event start and end date. Large intra-day price range is frequently observed on event dates of this sample; hence, intra-day average is more representative than the end-of-day price to estimate the return of investors who trade upon the events.

The SPAC IPO is assumed to be the first day of record in the CRSP database for buy-and-hold return calculation. CRSP data are only available for SPAC common shares. SPAC common shares typically start trading 30 to 90 days after the SPAC IPO when unit split becomes possible. The benchmark CRSP returns are the total return on the CRSP equal-weighted and value-weighted market indices during matched investment period for each event window observation.



(b) Target announcement to deSPAC initial business combination, average holding period = 0.4 years



(c) SPAC IPO to deSPAC initial business combination, average holding period = 1.7 years

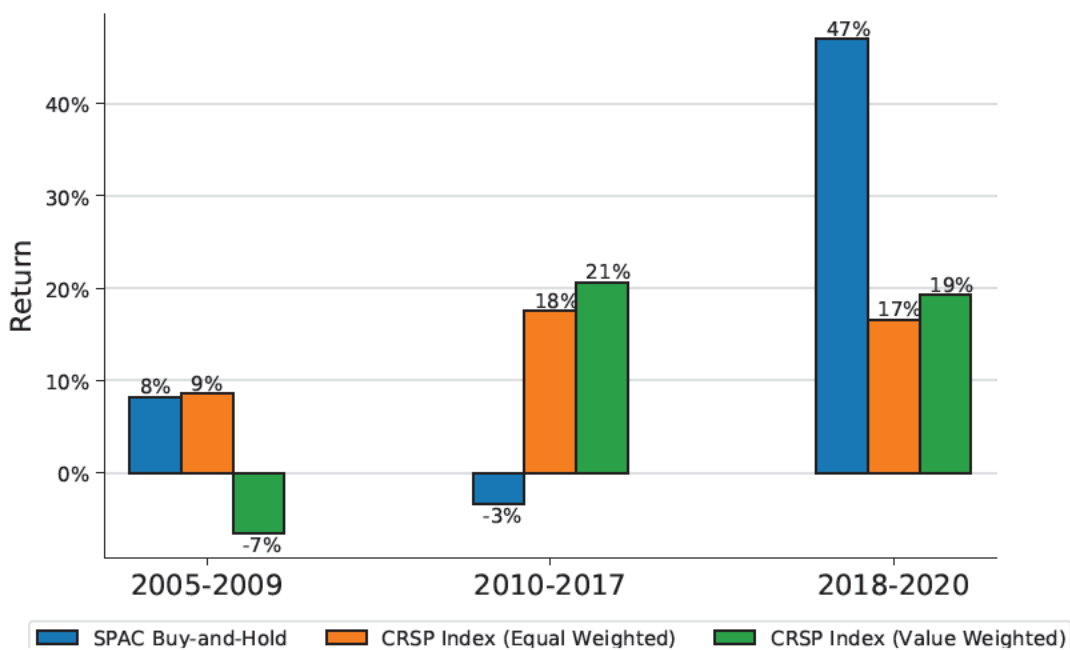
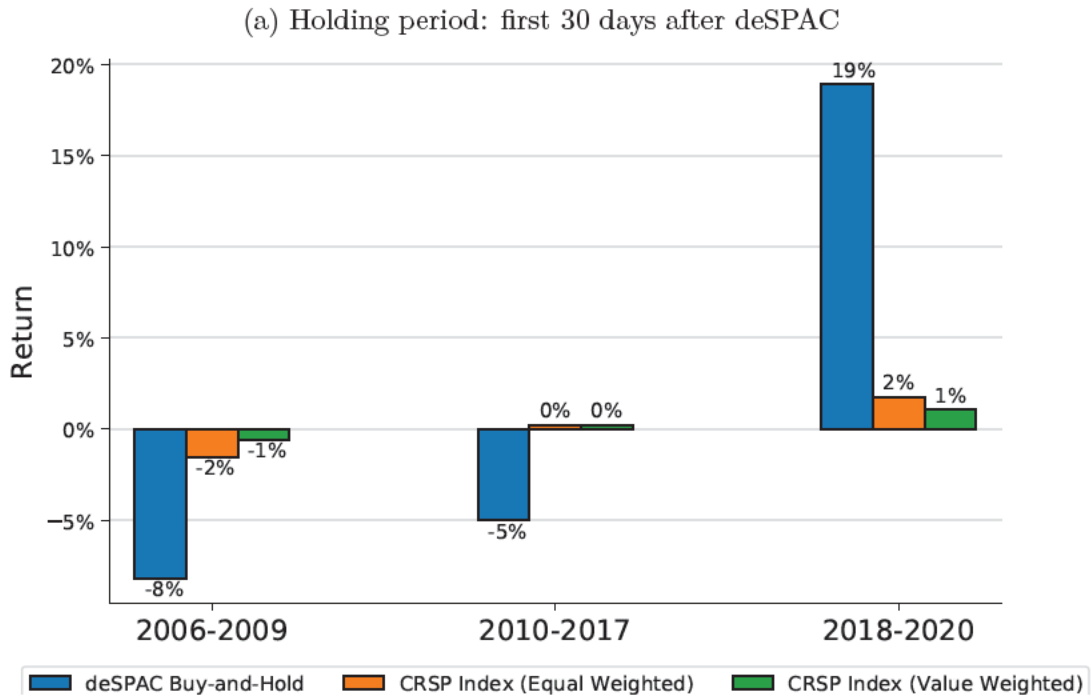


Figure 6: Common Share Buy-and-Hold Return during the deSPAC Period

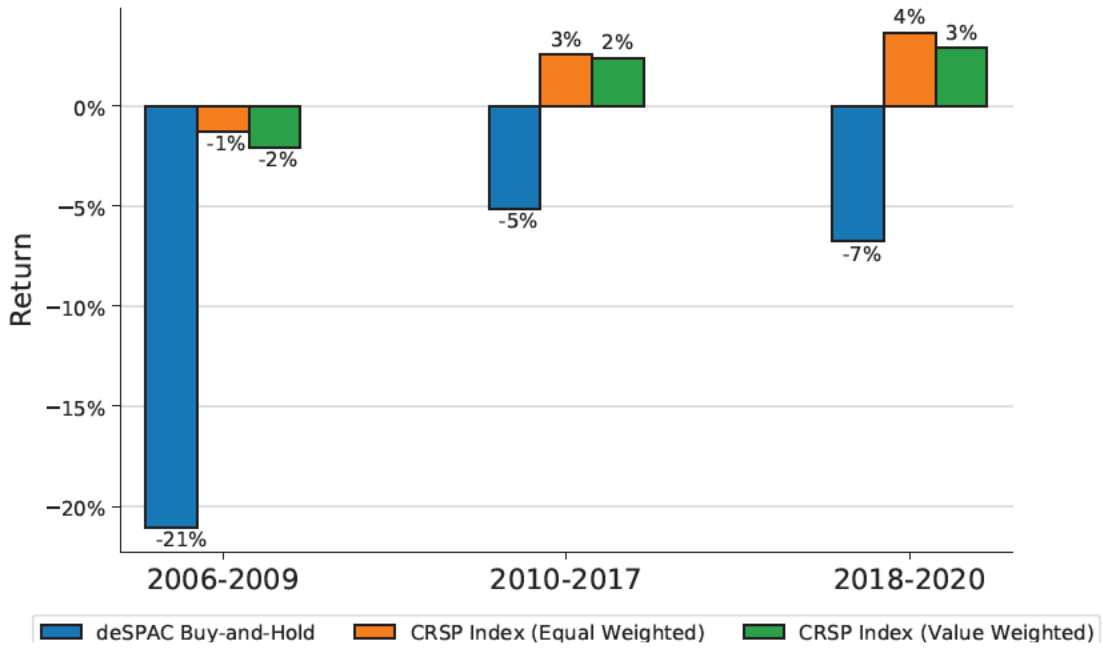
Figure 6 reports the deSPAC common share average buy-and-hold return during the first 30-day, 90-day, and 180-day periods, for investors who purchase the deSPAC company common share on its first day of trading as the new entity and hold through the respective period. Total return includes capital gain and ordinary dividend. The sample includes 189 SPACs traded on major exchanges that have completed the deSPAC initial business combination through 2020. Year groups are based on deSPAC year.

The buy-and-hold returns are calculated with the average of the highest and lowest trading prices on the holding period start and end date. We use CRSP data through the end of 2020; if the event window extends into 2021, an observation is dropped from the average calculation. The benchmark CRSP returns are the total return on the CRSP equal-weighted and value-weighted market indices during matched investment holding period for each event window observation.





(b) Holding period: first 90 days after deSPAC



(c) Holding period: first 180 days after deSPAC

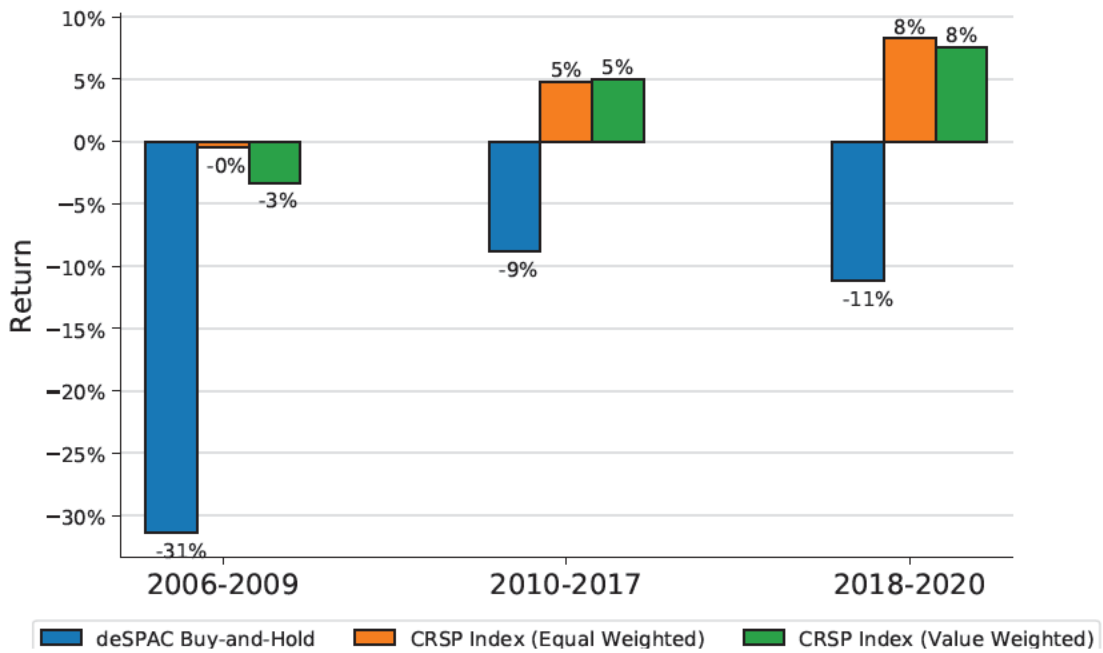
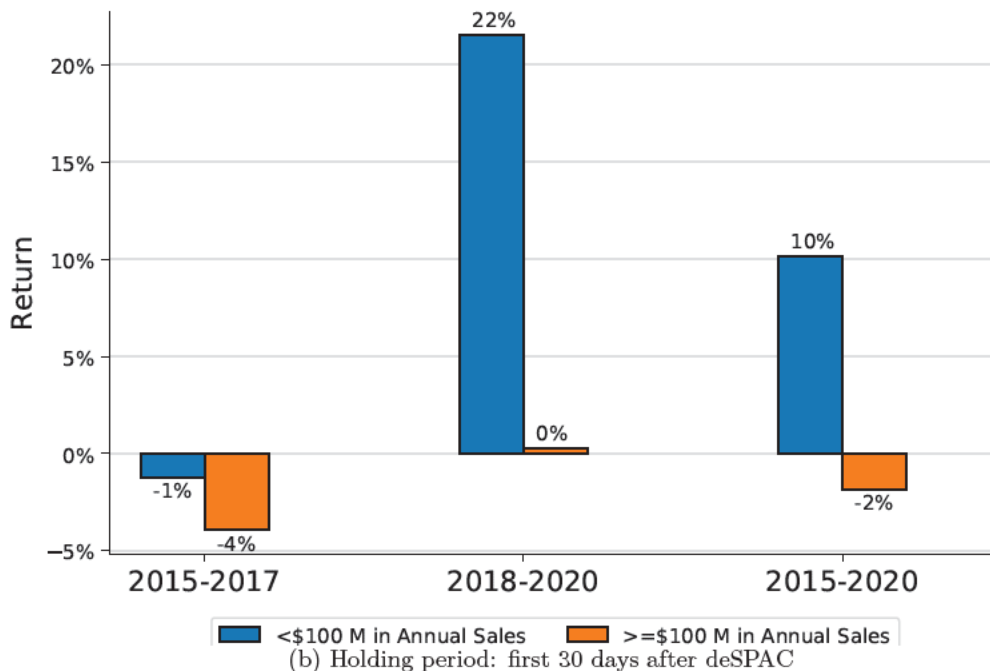


Figure 7: SPAC Common Share Buy-and-Hold Return by Annual Sales

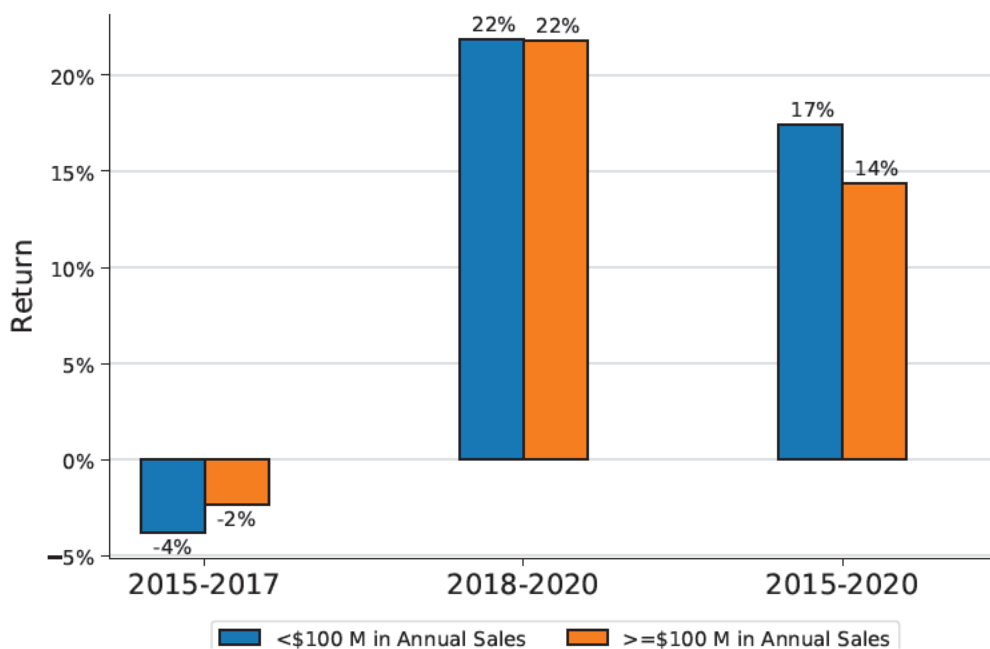
Figure 7 reports the average SPAC buy-and-hold return relative to CRSP equal-weighted index by the annual sales of the target firm in the fiscal year prior to going public. Returns for four holding periods are reported. The sample includes 130 SPACs traded on major exchanges that have completed the deSPAC initial business combination between 2015-2020, for which annual sales data are available from Compustat or 10-K filings. Year groups are based on deSPAC year.

The buy-and-hold returns are calculated with the average of the highest and lowest trading prices on the holding period start and end date. The benchmark CRSP returns are the total return on the CRSP equal-weighted index during matched investment holding period for each event window observation.

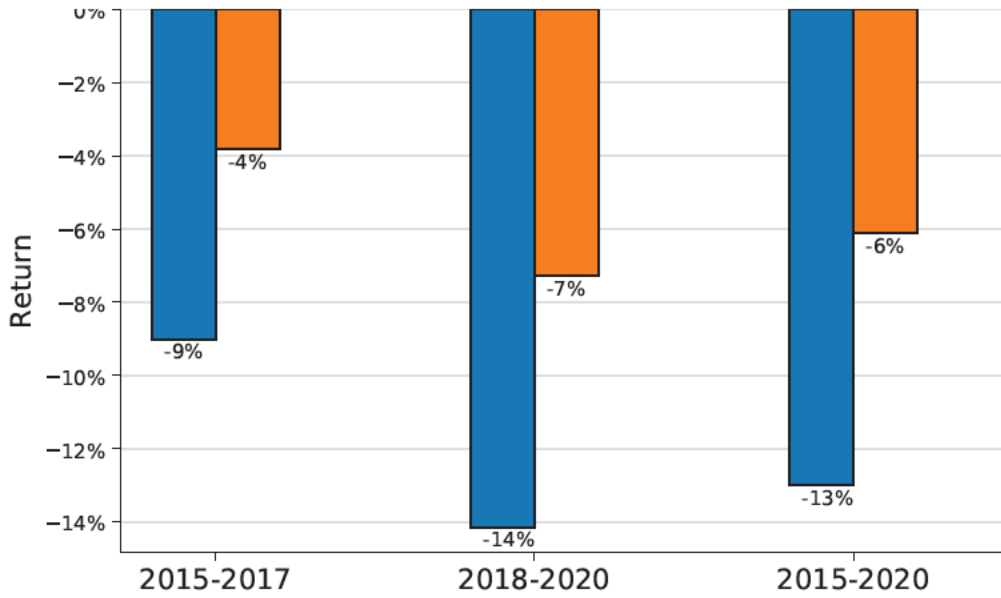
(a) Holding period: Target Announcement to deSPAC



(b) Holding period: first 30 days after deSPAC



(c) Holding period: first 90 days after deSPAC



(d) Holding period: first 180 days after deSPAC

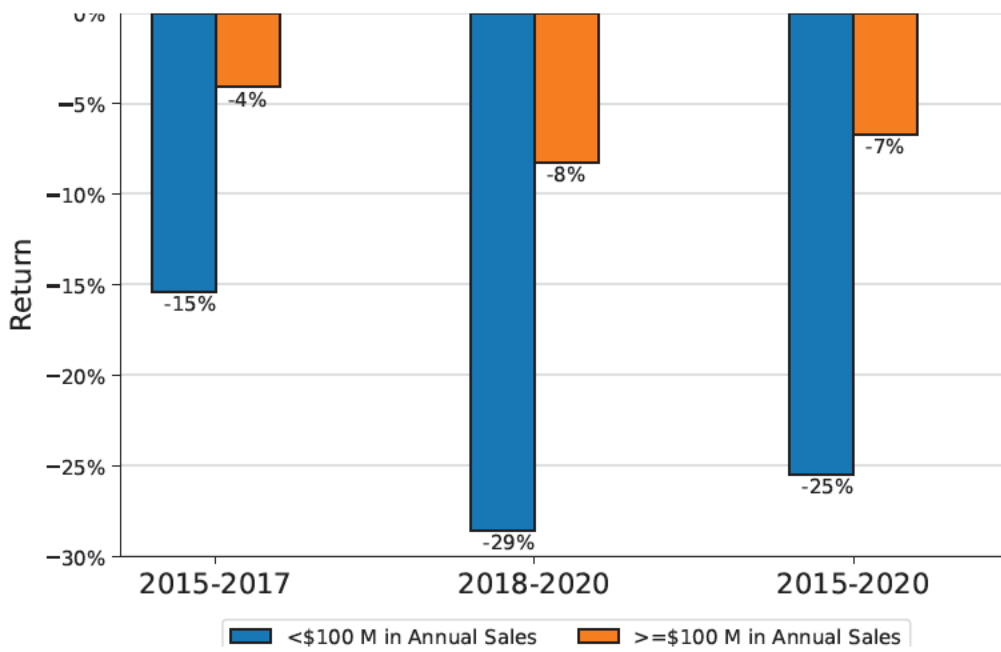
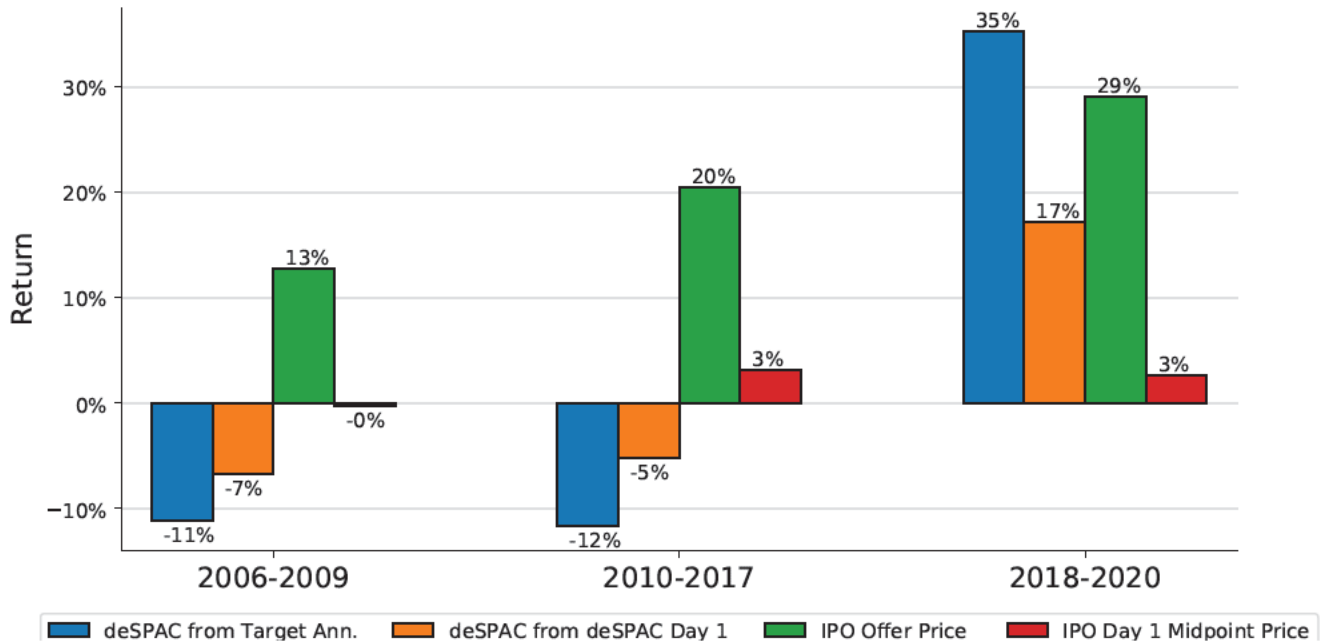


Figure 8: Buy-and-Hold Excess Return of Newly Public Companies Common Stock: SPAC vs. IPO, Relative to CRSP Equal Weighted Index

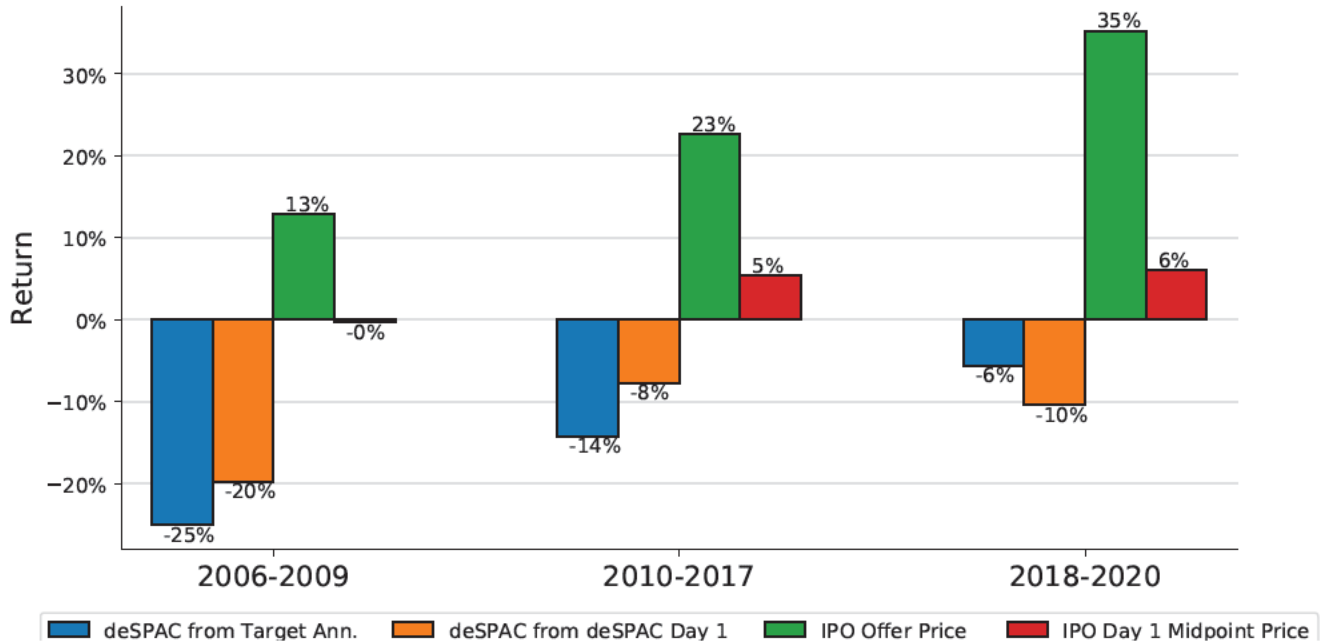
Figure 8 compares the buy-and-hold excess return of newly public stocks of companies that went public via SPAC and traditional IPOs during 2006-2020. The excess return is defined as the SPAC or IPO buy-and-hold return (BHR) minus that of CRSP Value Weighted Index in matched investment period.

The deSPAC sample returns are calculated based on the purchase price on the target announcement date and the first deSPAC trading day. The IPO sample returns are calculated based on both offer price and the average of the highest and lowest trading prices on the first trading day. Year groups are based on deSPAC year for the SPAC sample, and IPO year for the IPO sample.

(a) Holding period: through the first 30 days after going public, excess return relative to CRSP Equal Weighted Index



(b) Holding period: through the first 90 days after going public, excess return relative to CRSP Equal Weighted Index



(c) Holding period: through the first 180 days after going public, excess return relative to CRSP Equal Weighted Index

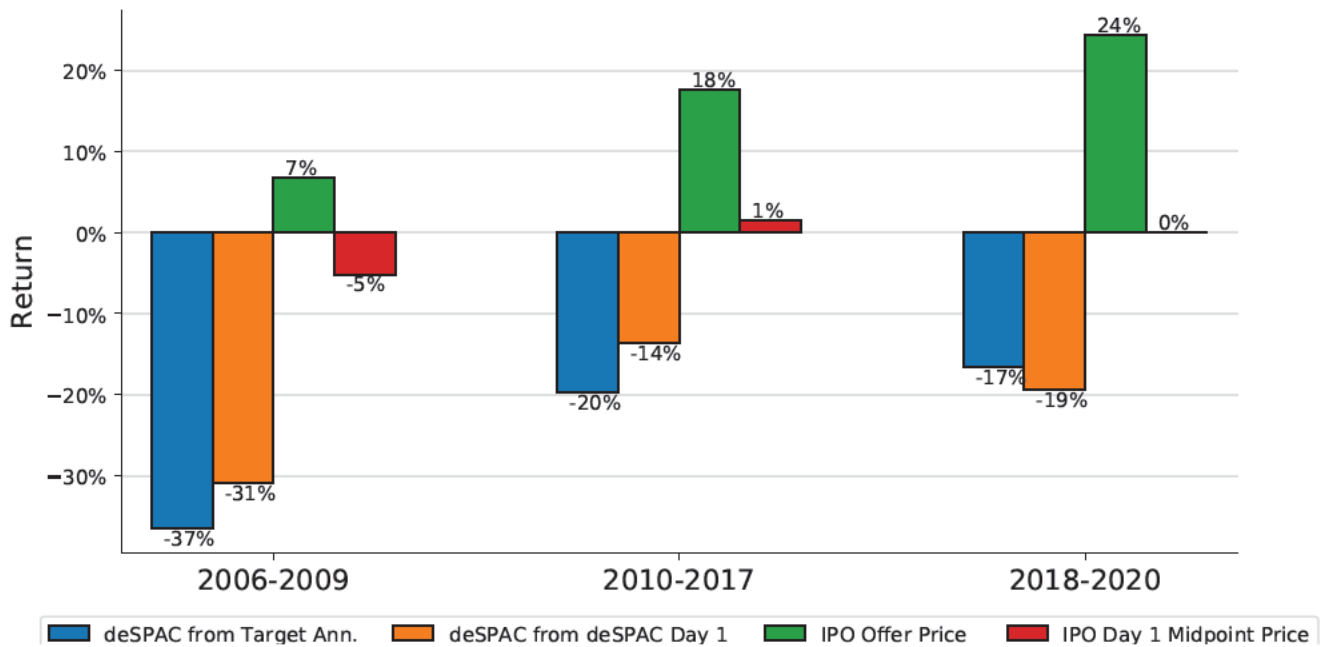
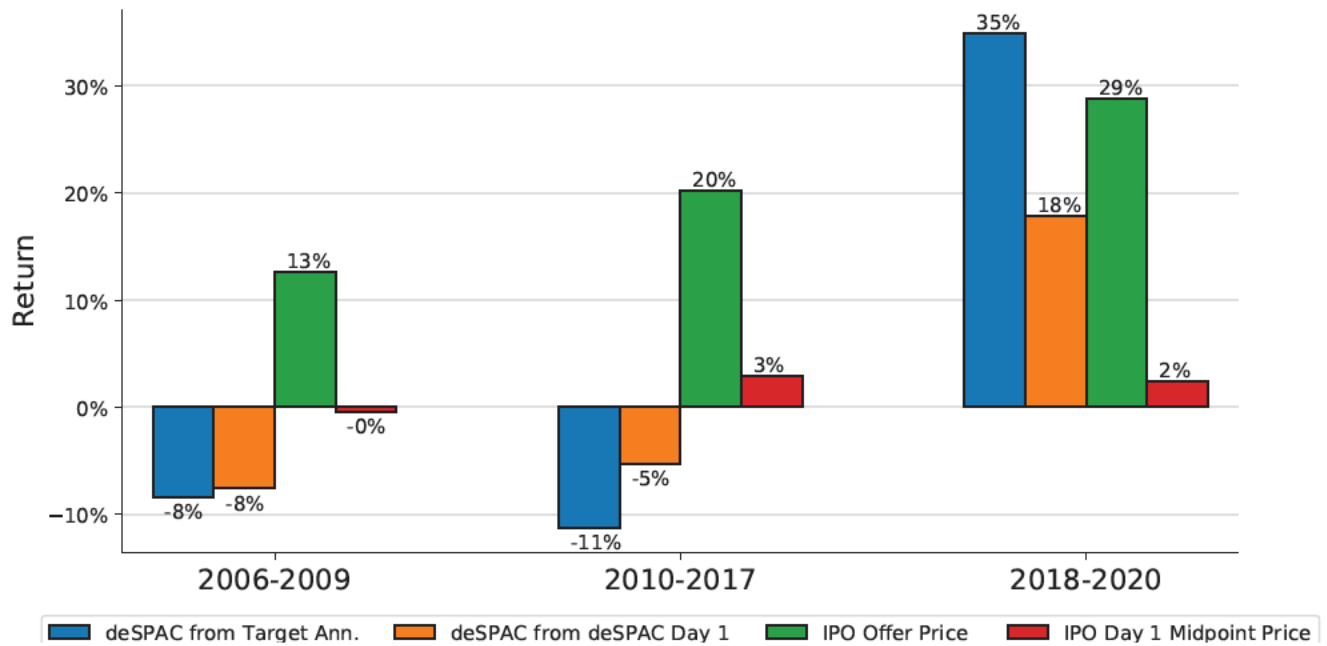


Figure 9: Buy-and-Hold Excess Return of Newly Public Companies Common Stock: SPAC vs. IPO, Relative to CRSP Value Weighted Index

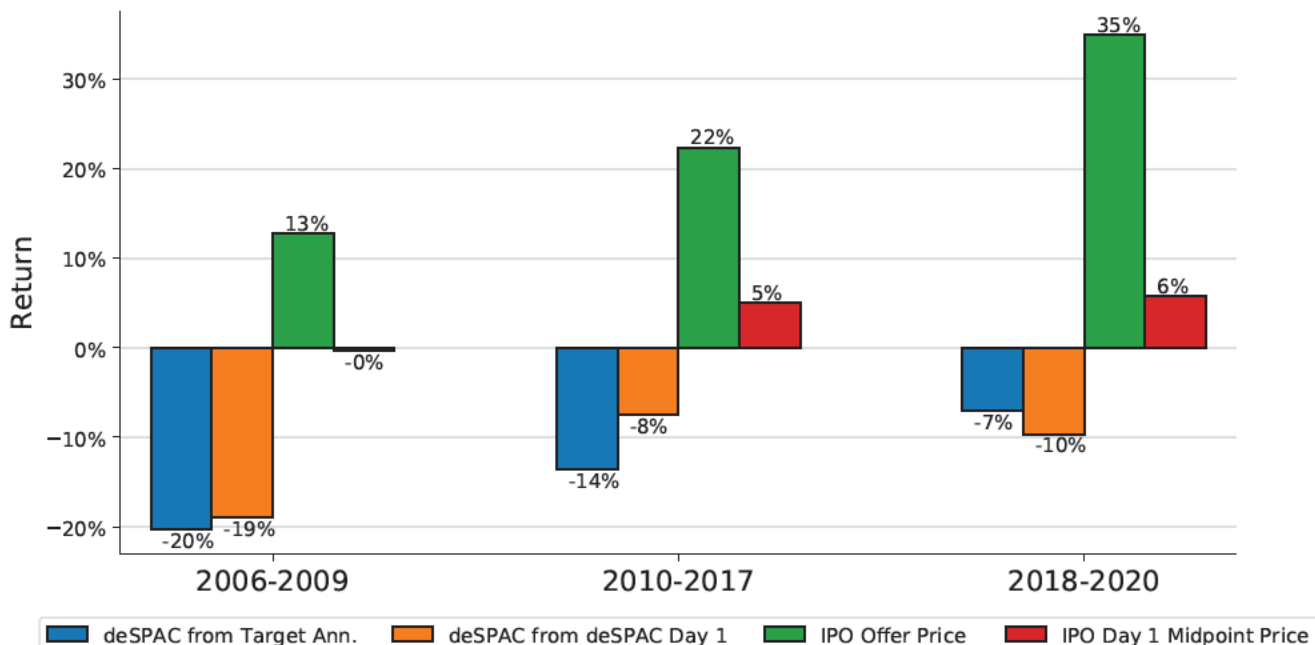
Figure 9 compares the buy-and-hold excess return of newly public stocks of companies that went public via SPAC and traditional IPOs during 2006-2020. The excess return is defined as the SPAC or IPO buy-and-hold return (BHR) minus that of CRSP Value Weighted Index in matched investment period.

The deSPAC sample returns are calculated based on the purchase price on the target announcement date and the first deSPAC trading day. The IPO sample returns are calculated based on both offer price and the average of the highest and lowest trading prices on the first trading day. Year groups are based on deSPAC year for the SPAC sample, and IPO year for the IPO sample.

(a) Holding period: through the first 30 days after going public, excess return relative to CRSP Value Weighted Index



(b) Holding period: through the first 90 days after going public, excess return relative to CRSP Value Weighted Index



(c) Holding period: through the first 180 days after going public, excess return relative to CRSP Value Weighted Index

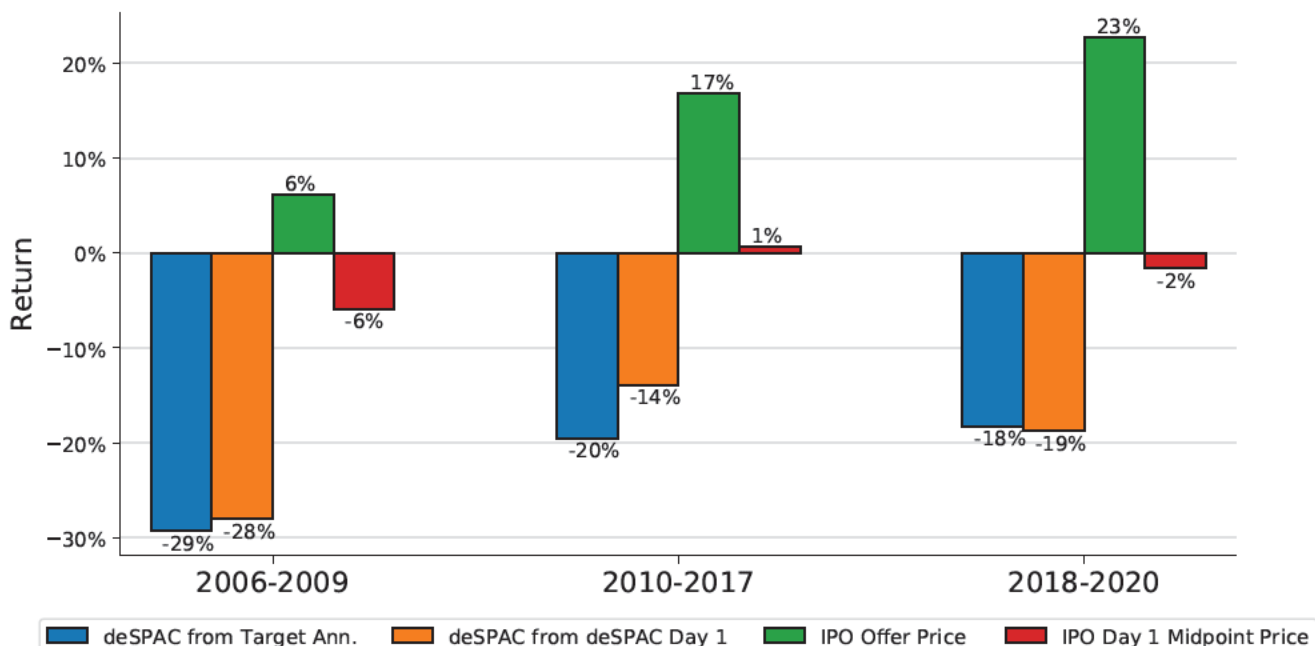
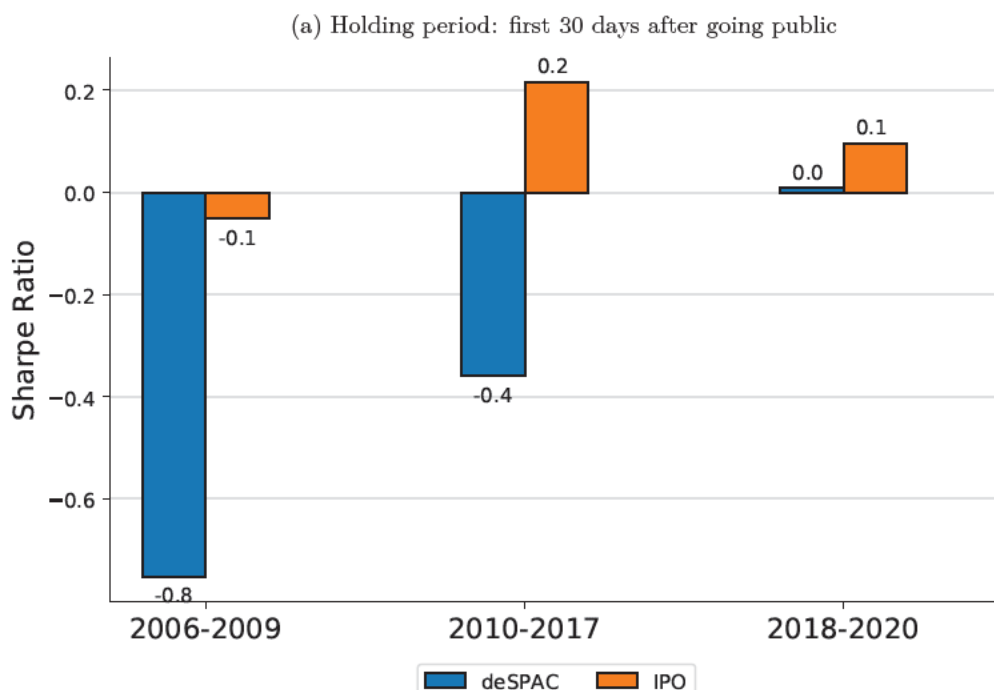


Figure 10: Average Sharpe Ratio of SPACs and IPOs

Figure 10 compares the average Sharpe ratios of deSPAC and IPO buy-and-hold return from the first day of going public. The SPAC sample consists of 189 SPACs traded on major exchanges that have completed the deSPAC initial business combination (IBC) through 2020. The IPO sample consists of traditional IPOs between 2006 and 2020.

$$\text{Sharpe Ratio} = \frac{\text{SPAC or IPO BHR} - R_f}{\sqrt{T} * \text{Holding Period Realized Daily Volatility}},$$

where T is the number of trading days during the holding period;  $R_f$  is the risk free return during the holding period. Year group is based on going public year. IPO return assumes the purchase price to be the midpoint of first-day trading price range.





(b) Holding period: first 90 days after going public

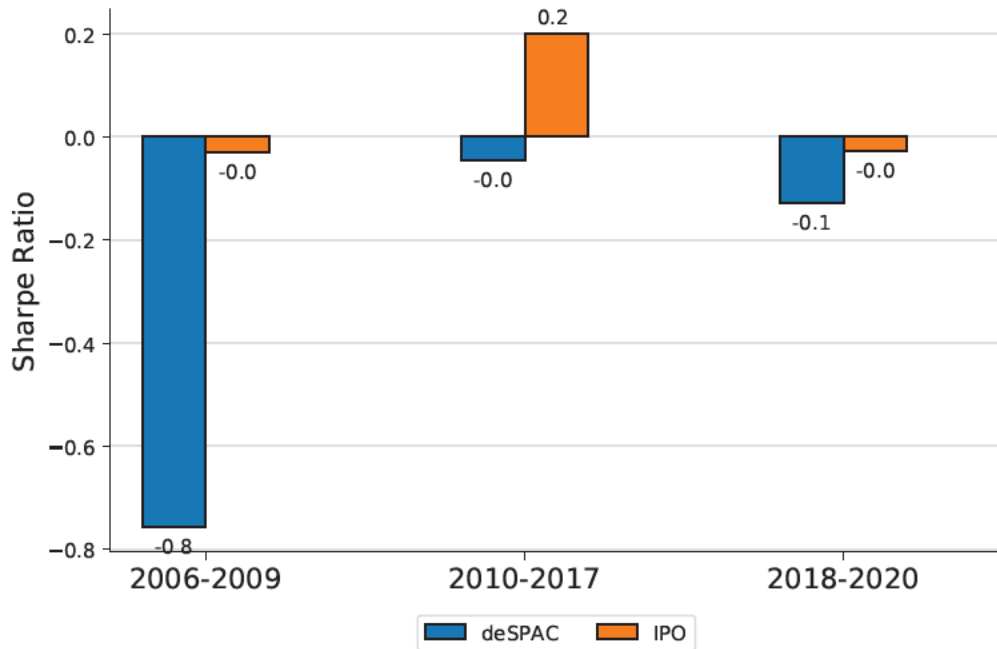
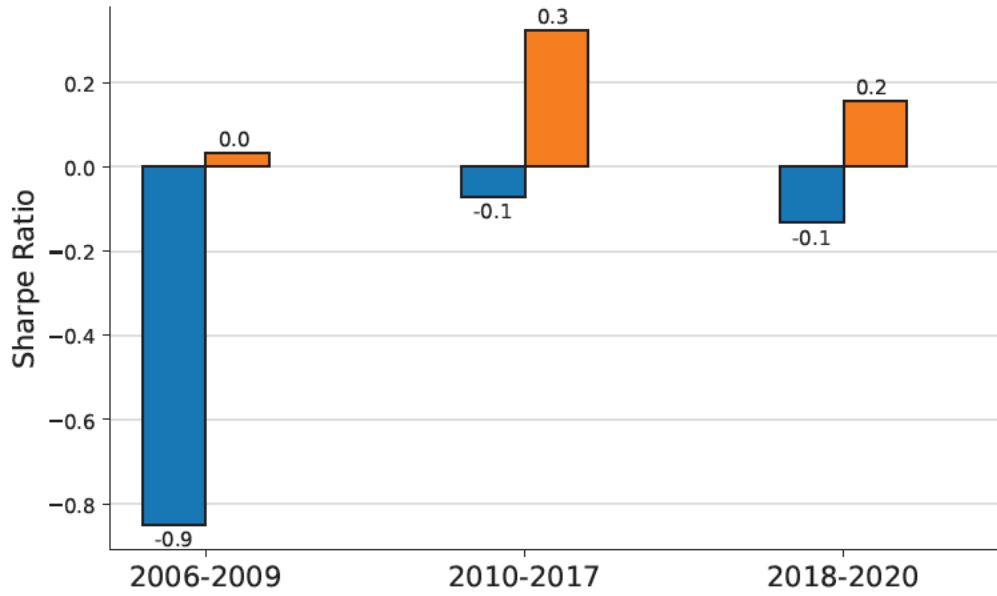


Table 7: Follow-on Equity Offering of deSPAC Firms

Table 7 summarizes the follow-on equity offering activity of deSPAC firms. By February 2021, 51 of the 189 firms that went public via SPAC through 2020 have issued a follow-on equity offering. On average, higher SEO underwriter reputation is related to larger follow-on equity offering size, higher follow-on demand as measured by over-allotment exercise ratio, and higher follow-on offer price. The follow-on offering sample is constructed from the SDC New Issues Database, based on name and CUSIP matching of the deSPAC firms.

1st Follow-On Equity Offering												
Year	deSPAC Merger	Count deSPAC Firm w/ FO	Average Time after deSPAC (Years)	All			High UW Reputation			Low UW Reputation		
				Average		Proportion	Average		Proportion	Average		Proportion
				Proceeds (\$M)	Over-allot.	Offer Price FO>SPAC	Proceeds (\$M)	Over-allot.	Offer Price FO>SPAC	Proceeds (\$M)	Over-allot.	Offer Price FO>SPAC
2006	1	0	-									
2007	7	2	3.9	18	1%	0%				18	1%	0%
2008	10	5	3.7	47	3%	0%	71	8%	0%	30	0%	0%
2009	13	5	1.6	79	6%	20%	79	6%	20%			
2010	5	1	0.5	36	0%	0%	36	0%	0%			
2011	0	0	-									
2012	2	0	-									
2013	4	3	2.2	27	7%	33%	31	11%	50%	21	0%	0%
2014	4	0	-									
2015	8	1	0.3	47	15%	100%	47	15%	100%			
2016	10	5	1.6	167	7%	50%	238	7%	50%	25	8%	50%
2017	13	9	1.4	62	3%	44%	76	4%	57%	14	0%	0%
2018	24	6	0.9	99	7%	33%	97	6%	40%	110	10%	0%
2019	24	9	0.7	161	8%	67%	202	10%	86%	19	0%	0%
2020	64	5	0.4	522	12%	60%	646	15%	75%	25	0%	0%

Definitions:

Low underwriter reputation: Lead underwriter rank <5.

High underwriter reputation: Lead underwriter rank >=5.

Table 8: Average Buy-and-Hold Return of deSPAC Firm Follow-On Equity Offering

Table 8 reports the average buy-and-hold return of the first follow-on equity offering of deSPAC firms. Total return includes capital gain and dividends. Return is calculated assuming purchase price at the follow-on offer price.

deSPAC Year	deSPAC Firm w/ FO	Average Buy-and-Hold Return 1st Follow-On Equity Offering			
		1st Day	30-Day	90-Day	180-Day
2006	0		-	-	-
2007	2	3%	6%	-9%	43%
2008	5	1%	23%	12%	-9%
2009	4	2%	-4%	9%	13%
2010	1	-5%	-27%	-24%	-28%
2011	0		-	-	-
2012	0		-	-	-
2013	3	7%	16%	21%	0%
2014	0		-	-	-
2015	1	2%	-13%	-17%	-24%
2016	6	9%	-5%	-7%	-15%
2017	9	8%	9%	-10%	-7%
2018	6	6%	5%	2%	-4%
2019	9	4%	18%	25%	69%
2020	5	-4%	-3%	9%	27%

Figure 11: Robinhood Users Holding Monthly Statistics

Figure 11 highlights two growth dimensions of the Robinhood platform during the sample period May 2018-August 2020. First, the number of user holding each ticker has increased. Second, the number of ticker present on Robinhood has increased.

On the first day of each month in the sample, we rank all tickers based on the number of users holding. The blue lines show the number of users holding at the mean and various percentiles over time; the axis is on a log scale (see the primary axis on the left). The gray bars show the number of tickers present in the Robintrack data (see the secondary axis on the right). The universe of tickers include common stock and ETFs.

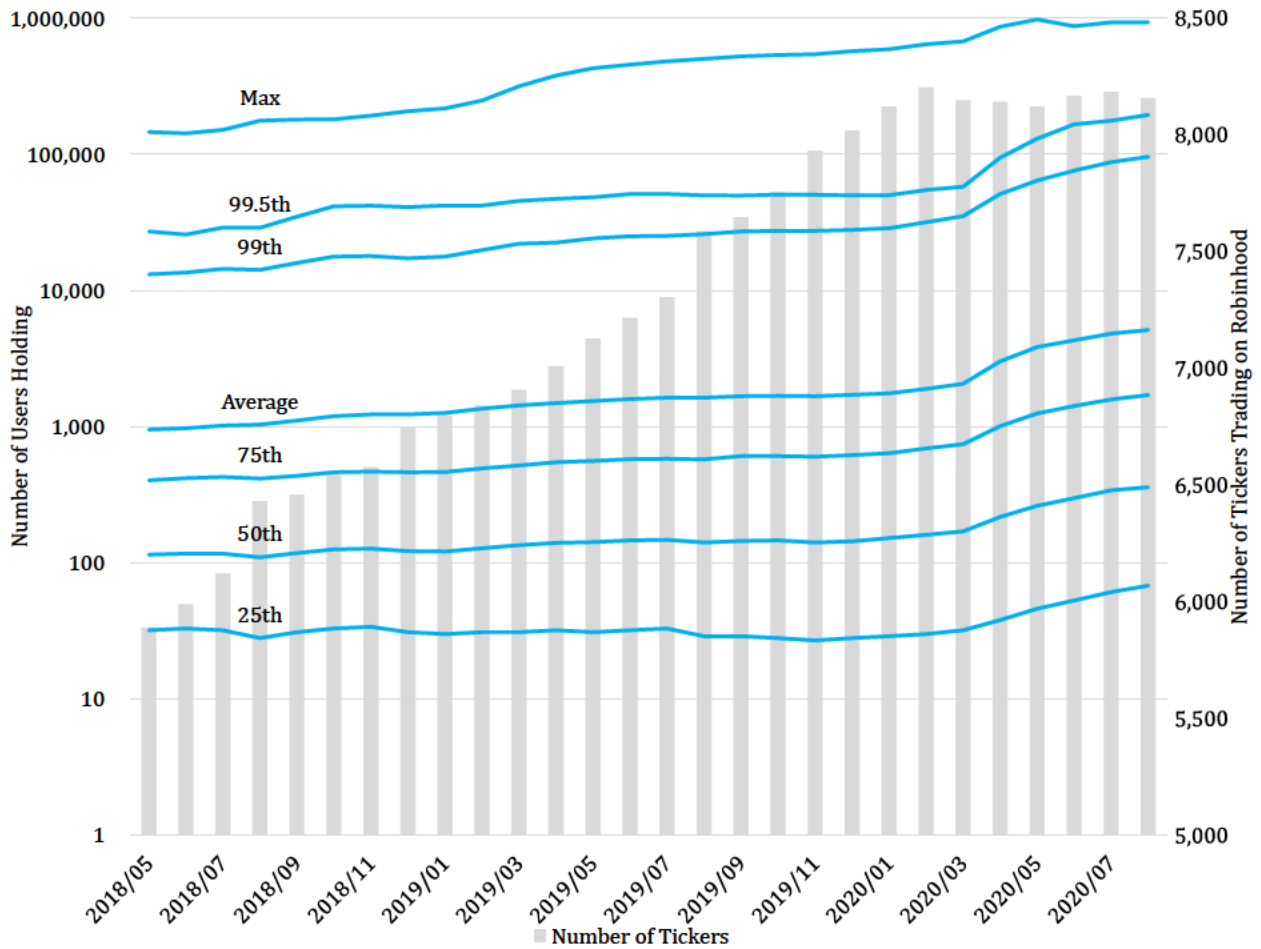


Figure 12: Summary of SPAC-Related Tickers in the Robintrack Sample at Month Start

Figure 12 reports the number of SPAC-related tickers in the Robintrack data as of the start of each month and their average popularity rank. The grey bars report the number of SPAC and deSPAC tickers available each month start in the Robintrack data. The blue lines report the average popularity percentile rank of the SPAC and deSPAC tickers at each month start.

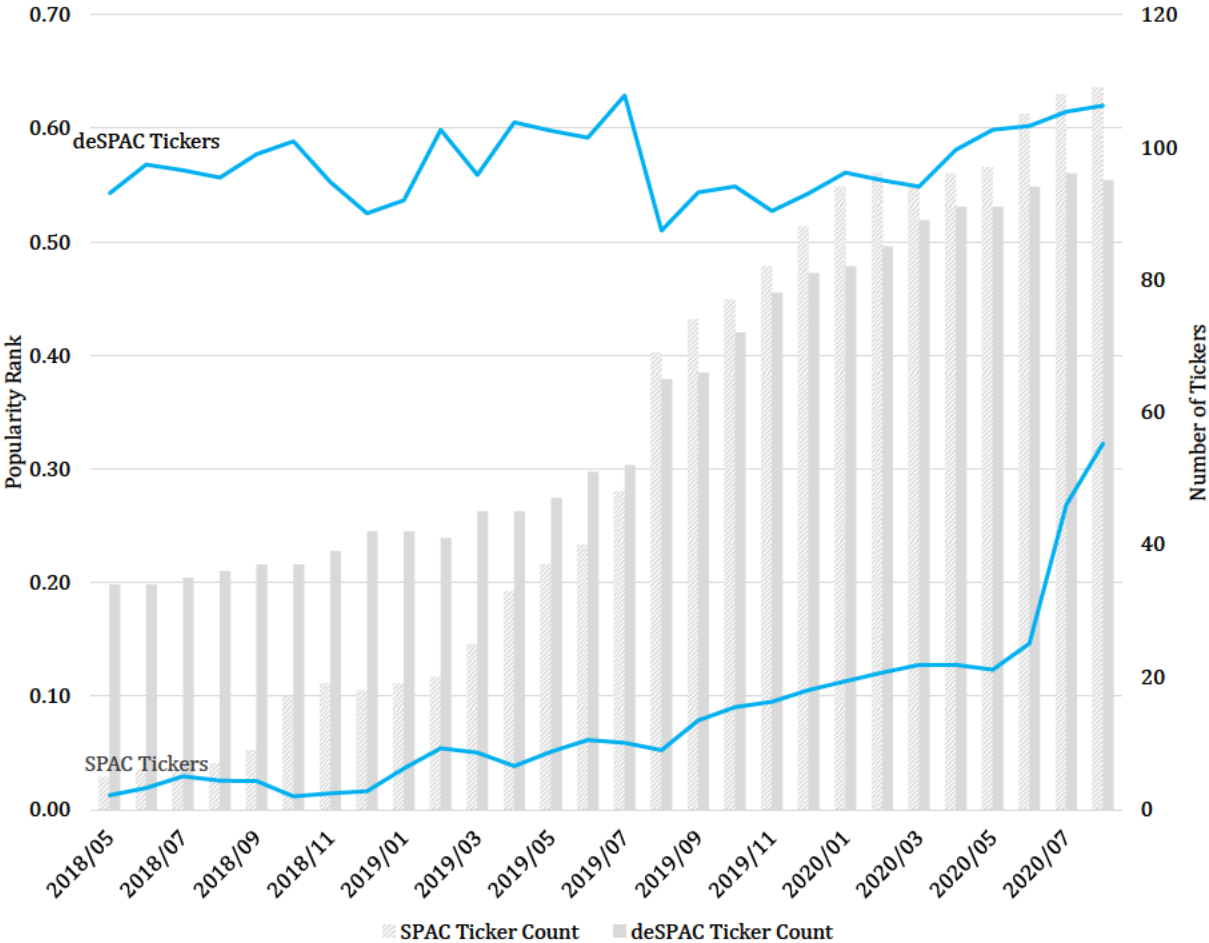


Table 9: Means Tests of SPACs Robintrack Popularity Measures: Before vs. After Target Announcement

Table 9 compares the average popularity of SPACs among Robinhood users before and after the target announcement, during the SPAC period. Four popularity measures are used. The three panels differ in the "before" group. The average time between target announcement and deal completion is 133 days for SPAC mergers through 2020, hence the sample choice for Panel C. User growth rate comparisons exclude observations where users holding<sub>t-1</sub> = 0.

Panel A: Full Sample				
Mean	Target Announcement		After-Before	p-value
	Before	After		
Relative Popularity Rank	0.11	0.27	0.16	0.00
N	23,564	4,036		
Users Holding	45	2,167	2,122	0.00
N	23,564	4,036		
Users Increase (( $\Delta$ ))	3	72	69	0.00
N	23,443	4,029		
Average Users Growth (%)	4%	33%	30%	0.00
N	20,410	3,948		
Panel B: Exclude SPACs without Target Announcement				
Mean	Target Announcement		After-Before	p-value
	Before	After		
Relative Popularity Rank	0.09	0.27	0.179	0.00
N	10,360	4,036		
Users Holding	21	2,167	2,146	0.00
N	10,360	4,036		
Daily Users Increase ( $\Delta$ )	5	72	68	0.00
N	10,314	4,029		
Daily Users Growth (%)	3%	33%	30%	0.00
N	9,027	3,948		
Panel C: Exclude SPACs without Target Announcement "Before" period: up to 133 days prior to deal announcement				
Mean	Target Announcement		After-Before	p-value
	Before	After		
Relative Popularity Rank	0.12	0.27	0.155	0.00
N	3,950	4,036		
Users Holding	43	2,167	2,124	0.00
N	3,950	4,036		
Users Increase (( $\Delta$ ))	12	72	61	0.00
N	3,938	4,029		
Users Growth (%)	5%	33%	28%	0.01
N	3,466	3,948		

Table 10: Reaction Time Distribution of Largest Popularity Change After SPAC Target Announcement

We calculate the daily change in popularity rank for each SPAC after target announcement. Table 10 summarizes when the largest daily change in popularity rank occurs.

The target announcement effect is realized very quickly. Among 56 SPACs that announced a target during the sample period, the largest one-day change in popularity rank occurs within 2 days post-announcement for 28 cases. The magnitude of one-day increase in popularity rank is also on average higher when the reaction time is shorter.

Days Since Announcement	Frequency	%	Cumulative %	Average Largest One-Day Popularity Rank Change
0	8	14%	14%	0.32
1	16	29%	43%	0.26
2	4	7%	50%	0.22
3-10	2	4%	54%	0.04
11-100	11	20%	73%	0.11
101-400	15	27%	100%	0.11
Total	56	100%		

Table 11: Regression Estimation of SPAC Robinhood Popularity and Market Variables Relationship

Table 11 reports the linear regression estimation results of SPAC Robinhood popularity on prices, market capitalization, return volatility, trading volume, and target announcement status. Estimations exclude observations with zero trading volume.

Dependent Variable:	Robinhood Popularity Rank		
	(1)	(2)	(3)
LN_MARKET_CAP	0.0419 (1.22)	0.0329 (1.00)	0.0493 (1.09)
L.LN_RET	0.0752 (0.43)	0.0960 (0.64)	0.122 (0.89)
L.VOLATILITY	2.034** (2.58)	1.779** (2.62)	1.684** (2.41)
L.LN_MA_VOL	0.0197*** (2.79)	0.0142** (2.33)	0.0210** (2.37)
LN_AMIHUUD_ILLIQ	-0.0156*** (-5.58)	-0.0120*** (-4.88)	-0.0119*** (-2.80)
POST_ANN.		0.136*** (3.91)	0.0995*** (3.35)
Ticker Fixed Effects	x	x	x
Month Fixed Effects	x	x	x
Constant	-0.771* (-1.78)	-0.579 (-1.44)	-0.861 (-1.54)
Observations	13,198	13,198	7,152
Adjusted $R^2$	0.752	0.775	0.784
Adjusted $R^2$ (within)	0.327	0.390	0.392

\*  $p < 0.1$ , \*\*  $p < 0.05$ , \*\*\*  $p < 0.01$ .

t statistics are reported in parentheses. Standard errors are double-clustered at the stock and month levels. Column 3 excludes SPACs that have not announced target by the end of the sample period.



Table 12: Means Tests of SPACs Robintrack Popularity Measures: High vs. Low Underwriter Reputation

Table 12 compares Robinhood popularity measures of SPACs with high and low underwriter reputation.

	Underwriter Reputation		High-Low	t-test p-value
	Low	High		
<b>Before Target Announcement</b>				
Average Popularity Rank	0.08	0.12	0.05	0.00
Average Users Holding	16	55	40	0.00
N	6,145	17,419		
<b>After Target Announcement</b>				
Average Relative Popularity Rank	0.20	0.30	0.10	0.00
Average Users Holding	488	2,859	2,372	0.00
N	1,178	2,858		

Definitions:

Low underwriter reputation: Lead underwriter rank <5.

High underwriter reputation: Lead underwriter rank >=5.

Table 13: Regression Estimation of SPAC Robinhood Popularity and SPAC Features

Table 13 reports the linear regression estimation results of Robinhood popularity on SPAC IPO features, conditional on market variables and target announcement status. Estimations exclude observations with zero trading volume.

Columns 1 and 2 uses the full sample of SPAC tickers in the Robintrack data. Column 3 uses the “before target announcement” subsample. Column 4 uses the “before target announcement” subsample, while excluding SPACs without a target announcement by the end of the sample period. Column 5 uses the “after target announcement” subsample.

Dependent Variable:	Robinhood Popularity Rank				
	Full Sample (1)	Full Sample (2)	PA=0 (3)	PA=0 target!=. (4)	PA=1 (5)
UWRANK	0.00312 (0.89)	-0.000877 (-0.19)	-0.00197 (-0.44)	-0.00489 (-0.73)	0.00395 (0.33)
LN_SPAC_PROCEEDS		0.0186 (1.07)	0.0335* (1.84)	0.0480** (2.44)	-0.0350 (-1.26)
OVERALLOTMENT		-0.221* (-1.88)	-0.267** (-2.24)	-0.254 (-1.62)	0.118 (0.41)
POST_ANN.	0.0582** (2.52)	0.0665*** (3.05)			
L.LN_RET	0.0939 (0.47)	0.102 (0.51)	0.0883 (0.66)	0.227 (1.14)	0.225** (2.49)
L.VOLATILITY	2.256*** (2.82)	2.257*** (2.88)	0.798 (1.68)	0.206 (1.20)	3.714*** (7.24)
L.LN_MA_VOL	0.0218*** (3.32)	0.0200** (2.74)	0.00435 (0.76)	0.00312 (0.71)	0.0466*** (4.14)
LN_AMIHUDD_ILLIQ	-0.0201*** (-5.59)	-0.0194*** (-5.22)	-0.0188*** (-3.52)	-0.00334 (-0.96)	-0.0157** (-2.23)
Month Fixed Effect	x	x	x	x	x
Constant	-0.366*** (-4.45)	-0.392*** (-5.13)	-0.276*** (-3.74)	-0.169* (-1.98)	-0.390*** (-3.18)
Observations	13,198	13,198	10,706	4,660	2492
Adjusted $R^2$	0.565	0.571	0.409	0.393	0.731
Adjusted $R^2$ (within)	0.441	0.448	0.254	0.155	0.577

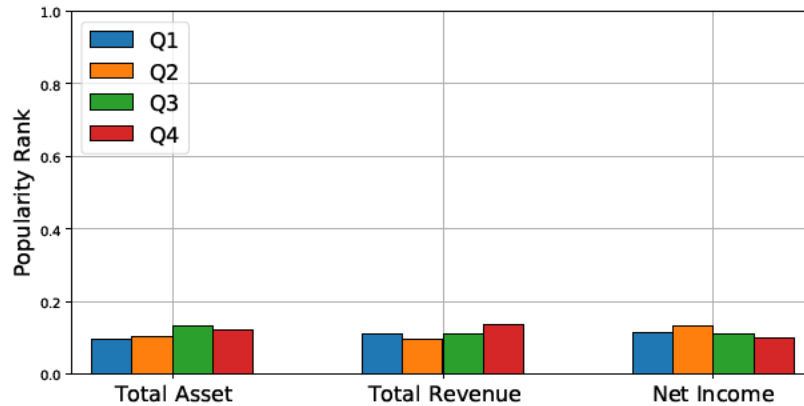
\*  $p < 0.1$ , \*\*  $p < 0.05$ , \*\*\*  $p < 0.01$ .

t statistics are reported in parentheses. Standard errors are double-clustered at the stock and month levels. Exclude observations with zero trading volume.

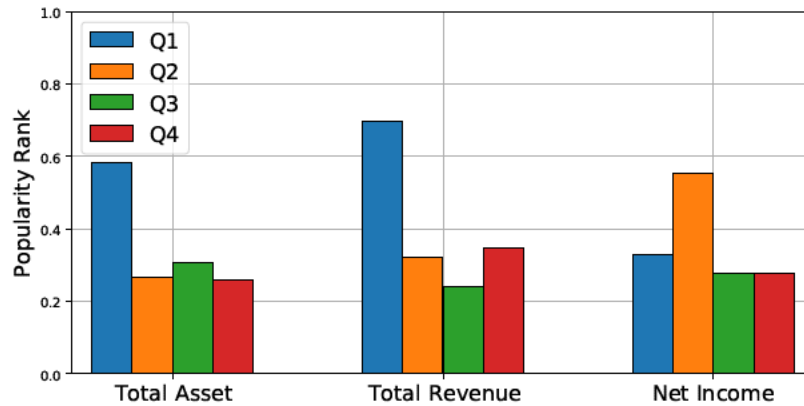
Figure 13: SPAC Average Robinhood Popularity Rank by Target Firm Fundamental Quartiles

Figure 13 shows the average Robinhood popularity of SPACs by quartiles of target firm fundamentals in the fiscal year prior to going public. Quartiles are determined by firms that went public through either IPO or deSPAC in 2015-2020, for which fundamental data are available. The sample includes SPACs with completed business combination through 2020.

(a) Sample period: before target announcement



(b) Sample period: between target announcement and deSPAC



(c) Sample period: the first 180 days since deSPAC

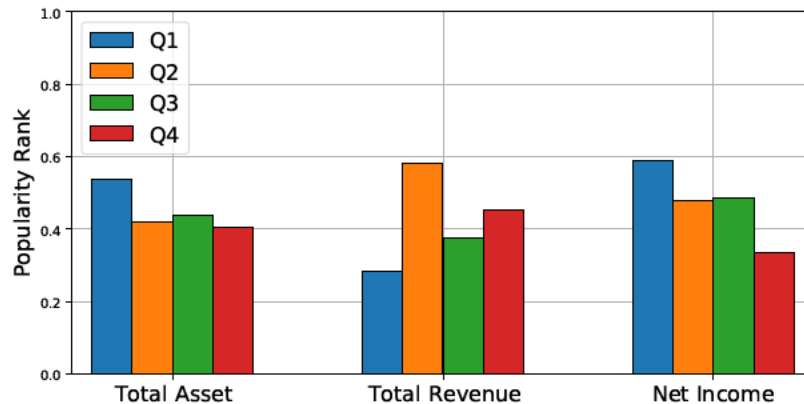


Table 14: Regression Estimation of SPAC Robinhood Popularity on Target Firm Characteristics

Table 14 reports regression results of popularity of SPACs on target firm features during the post-announcement period, controlling for SPAC characteristics and market variables. The sample includes SPACs with completed business combination through 2020.

Dependent Variable:	Popularity Above Median		Popularity Rank	
	(1)	(2)	(3)	(4)
LN_TOTAL_ASSET	0.0711*		-0.0384	
	(1.82)		(-1.45)	
LN_SALES		0.0707*		-0.0346
		(1.99)		(-1.68)
NET_INCOME/ASSET	-0.00492	-0.0957	-0.0553	-0.0364
	(-0.05)	(-1.18)	(-0.62)	(-0.45)
SALES/ASSET	-0.0241		-0.0768	
	(-0.28)		(-1.40)	
R&D_I	-0.245***	-0.252***	-0.0738**	-0.0813**
	(-3.47)	(-3.38)	(-2.32)	(-2.26)
R&D/ASSET	0.0187	-0.231	-0.212	-0.132
	(0.08)	(-1.21)	(-0.96)	(-0.74)
AGE	-0.00704***	-0.00701**	-0.000354	-0.000580
	(-3.15)	(-2.77)	(-0.57)	(-0.65)
HIGH_TECH	0.0584	0.0915	-0.0912	-0.0825
	(0.62)	(0.91)	(-1.17)	(-1.06)
VC_BACKED	0.268***	0.211**	0.0349	0.0346
	(3.02)	(2.24)	(0.51)	(0.53)
UWRANK	-0.0151	-0.0192	-0.00119	0.000185
	(-0.68)	(-0.83)	(-0.09)	(0.01)
OVERALLOTMENT	1.007	0.550	0.568	0.624
	(1.51)	(0.74)	(1.29)	(1.35)
LN_MARKET_CAP	0.0754	0.0670	0.0717	0.0833
	(1.28)	(0.95)	(1.31)	(1.39)
L.LN_RET	0.161	0.142	0.134	0.119
	(1.04)	(0.86)	(1.31)	(1.13)
L.VOLATILITY	3.958***	3.666***	3.352***	3.424***
	(3.84)	(3.23)	(5.53)	(5.43)
L.LN_MA_VOL	-0.00287	-0.00972	0.0391**	0.0366**
	(-0.11)	(-0.36)	(2.75)	(2.61)
LN_AMIHUDD_ILLIQ	-0.0215**	-0.0215***	-0.0112**	-0.0116**
	(-2.84)	(-2.86)	(-2.46)	(-2.69)
FF17 Industry FE	x	x	x	x
Month FE	x	x	x	x
Constant	-1.042*	-0.730	-0.978**	-1.223**
	(-1.99)	(-1.16)	(-2.10)	(-2.09)
Observations	1751	1751	1751	1751
Adjusted $R^2$	0.743	0.734	0.841	0.840
Adjusted $R^2$ (Within)	0.424	0.403	0.539	0.535

\*  $p < 0.10$ , \*\*  $p < 0.05$ , \*\*\*  $p < 0.01$ .  $t$  statistics in parentheses. Standard errors are double-clustered at the stock and month levels. Exclude observations with zero trading volume.

Table 15: Regression Estimation of Robinhood Popularity: IPO vs. deSPAC Firms

Table 15 reports the linear regression estimation results of Robinhood popularity on firm's going public method. The sample includes IPO and deSPAC firms that went public between 2016 and 2020.

Dependent Variable:	Robinhood Popularity Rank				
	Days Since Going Public				
	90 Days	180 Days	1 Year	2 Years	Full Sample
	(1)	(2)	(3)	(4)	(5)
DESPAC	0.326 (1.29)	0.394 (1.49)	0.354 (1.36)	0.492** (2.52)	0.661*** (3.06)
LN_TOTAL_ASSET	0.0110 (1.13)	0.00484 (0.57)	0.000358 (0.05)	-0.00407 (-0.64)	-0.00939 (-1.56)
DESPAC=1 × LN_TOTAL_ASSET	0.0157 (0.57)	0.0213 (0.73)	0.0254 (1.07)	0.0127 (0.65)	0.0174 (0.87)
NET_INCOME/ASSET	-0.0223** (-2.70)	-0.0277*** (-4.50)	-0.0223*** (-3.72)	-0.0189*** (-3.28)	-0.0148** (-2.72)
DESPAC=1 × NET_INCOME/ASSET	-0.0936 (-1.52)	-0.0646 (-1.02)	-0.0452 (-0.77)	-0.0150 (-0.30)	0.0138 (0.26)
SALES/ASSET	0.0363*** (3.73)	0.0407*** (4.42)	0.0287*** (3.20)	0.0214*** (2.77)	0.0180** (2.61)
DESPAC=1 × SALES/ASSET	-0.0104 (-0.29)	0.00912 (0.32)	0.0161 (0.62)	0.0126 (0.76)	-0.00877 (-0.49)
R&D_I=1	0.0147 (0.52)	0.00529 (0.19)	0.00155 (0.06)	0.0160 (0.68)	0.00134 (0.06)
DESPAC=1 × R&D_I=1	-0.0203 (-0.34)	0.0333 (0.53)	0.0604 (1.06)	-0.0130 (-0.33)	0.0109 (0.26)
R&D/ASSET	-0.0192 (-1.18)	-0.0300** (-2.09)	-0.0215 (-1.59)	-0.0239* (-1.95)	-0.0291** (-2.41)
DESPAC=1 × R&D/ASSET	-0.0406 (-0.46)	-0.0209 (-0.25)	-0.00258 (-0.03)	0.0145 (0.19)	0.0513 (0.57)
AGE	-0.000614 (-1.32)	-0.000412 (-0.93)	-0.000508 (-1.16)	-0.000602 (-1.50)	-0.000485 (-1.29)
DESPAC=1 × AGE	-0.00132 (-0.73)	0.0000535 (0.03)	0.00151 (0.96)	0.00122 (1.22)	0.000442 (0.68)
HIGH_TECH=1	0.0667 (1.66)	0.0591 (1.54)	0.0491 (1.33)	0.0631** (2.13)	0.0642** (2.31)
DESPAC=1 × HIGH_TECH=1	-0.0884 (-1.13)	-0.0569 (-0.87)	-0.0654 (-1.16)	-0.0612 (-1.29)	-0.0593 (-1.13)
VC_BACKED=1	0.0126 (0.64)	0.0251 (1.30)	0.0157 (0.89)	0.00993 (0.66)	0.0246* (1.82)
DESPAC=1 × VC_BACKED=1	0.133* (1.79)	0.0826 (1.14)	0.0734 (1.10)	0.0627 (1.25)	0.0504 (1.04)
UWRANK	-0.00643 (-0.99)	-0.00497 (-0.84)	-0.00120 (-0.22)	0.00182 (0.37)	0.00398 (0.97)
DESPAC=1 × UWRANK	-0.0218 (-1.55)	-0.0147 (-1.11)	-0.0184 (-1.47)	-0.00574 (-0.52)	-0.00223 (-0.21)
LN_MARKET_CAP	0.00492 (0.37)	0.0156 (1.29)	-0.00416 (-0.43)	-0.0207** (-2.75)	-0.0211*** (-2.97)
DESPAC=1 × LN_MARKET_CAP	-0.00835 (-0.35)	-0.0188 (-0.83)	-0.0111 (-0.54)	-0.00929 (-0.54)	-0.0305 (-1.62)
LLN_RET	0.00730 (0.22)	-0.0110 (-0.69)	-0.00555 (-0.46)	0.00727 (0.71)	0.00993 (1.08)
DESPAC=1 × LLN_RET	0.0322 (0.74)	0.0489 (1.66)	0.0527 (1.63)	0.0380 (1.58)	0.0626*** (3.02)
L.VOLATILITY	0.580 (1.53)	0.553** (2.48)	0.201 (1.40)	-0.173 (-1.63)	-0.261** (-2.73)
DESPAC=1 × L.VOLATILITY	0.551 (0.97)	0.345 (0.95)	0.608* (1.93)	0.816*** (2.96)	0.880** (2.57)
LLN_MA_VOL	0.101*** (6.87)	0.0848*** (8.34)	0.0935*** (11.01)	0.103*** (14.59)	0.106*** (16.14)
DESPAC=1 × LLN_MA_VOL	-0.0444 (-1.60)	-0.0474** (-2.60)	-0.0482*** (-3.09)	-0.0447*** (-3.06)	-0.0464*** (-3.10)
LN_AMIHUDD_ILLIQ	-0.00469 (-0.36)	-0.00739 (-0.80)	-0.00710 (-1.01)	-0.00686 (-1.14)	-0.00121 (-0.19)
DESPAC=1 × LN_AMIHUDD_ILLIQ	-0.0135 (-0.69)	-0.0132 (-1.07)	-0.00937 (-0.93)	-0.000847 (-0.09)	-0.00737 (-0.66)
FF17 Industry FE	x	x	x	x	x
Month FE	x	x	x	x	x
Constant	-0.922*** (-10.25)	-0.860*** (-10.91)	-0.655*** (-9.41)	-0.535*** (-8.75)	-0.445*** (-7.57)
Observations	9539	23176	50253	99734	155909
Adjusted $R^2$	0.671	0.629	0.625	0.608	0.568
Adjusted $R^2$ (Within)	0.603	0.572	0.569	0.561	0.523

\*  $p < 0.1$ , \*\*  $p < 0.05$ , \*\*\*  $p < 0.01$ .

t statistics are reported in parentheses. Standard errors are double-clustered at the stock and month levels.

Table 16: Regression Estimation of Return Volatility and Robinhood Popularity: IPO vs. deSPAC Firms

Table 16 reports the linear regression estimation results of return volatility on Robinhood popularity, conditional on firm's going public method, firm features, and other market variables. The sample includes IPO and deSPAC firms that went public between 2016 and 2020.

Dependent Variable:	Return Volatility				
	Days Since Going Public				
	90 Days	180 Days	1 Year	2 Years	Full Sample
	(1)	(2)	(3)	(4)	(5)
L.RH Popularity Percentile Rank	0.0313*** (3.41)	0.0220*** (3.44)	0.0120*** (2.90)	0.000920 (0.29)	-0.00199 (-0.56)
DESPAC	0.109*** (3.50)	0.101*** (4.05)	0.0534** (2.59)	0.0354* (1.92)	0.0296 (1.54)
LN_TOTAL_ASSET	-0.00544*** (-3.70)	-0.00501*** (-4.73)	-0.00445*** (-6.79)	-0.00471*** (-8.26)	-0.00441*** (-9.05)
DESPAC=1 × LN_TOTAL_ASSET	0.00220 (0.41)	0.00424 (1.34)	0.00258 (1.12)	0.00182 (1.10)	0.00251 (1.61)
NET_INCOME/ASSET	0.00300** (2.19)	0.00256* (1.85)	0.00198** (2.32)	0.00246*** (4.87)	0.00235*** (4.99)
DESPAC=1 × NET_INCOME/ASSET	0.00866 (0.81)	0.00457 (0.76)	0.00199 (0.53)	-0.000819 (-0.22)	-0.00200 (-0.59)
SALES/ASSET	-0.00587*** (-3.96)	-0.00554*** (-4.44)	-0.00464*** (-5.11)	-0.00403*** (-4.96)	-0.00268*** (-4.36)
DESPAC=1 × SALES/ASSET	0.00595 (1.42)	0.00446* (1.83)	0.00525*** (3.10)	0.00482*** (3.22)	0.00327** (2.18)
R&D_I=1	-0.00519 (-1.34)	-0.00306 (-0.99)	-0.00264 (-1.40)	-0.00417** (-2.31)	-0.00202 (-1.20)
DESPAC=1 × R&D_I=1	0.00326 (0.36)	0.00351 (0.50)	-0.00137 (-0.29)	0.00123 (0.37)	0.00446 (1.35)
R&D/ASSET	0.000950 (0.44)	0.00291 (1.21)	0.00133 (0.86)	0.00191* (1.91)	0.00177* (2.03)
DESPAC=1 × R&D/ASSET	0.0125 (1.00)	0.0147** (2.06)	0.00924* (1.92)	0.00259 (0.55)	0.00326 (0.86)
AGE	-0.0000860* (-1.75)	-0.000107*** (-2.89)	-0.0000668** (-2.35)	-0.0000417* (-1.73)	-0.0000241 (-1.01)
DESPAC=1 × AGE	0.000176 (0.77)	0.000178 (1.18)	0.000164 (1.36)	0.0000896 (1.34)	0.0000127 (0.25)
HIGH_TECH=1	-0.00367 (-0.54)	0.000733 (0.15)	-0.00260 (-0.84)	-0.00343 (-1.41)	-0.00458* (-1.96)
DESPAC=1 × HIGH_TECH=1	0.0110 (1.07)	0.000844 (0.13)	0.00480 (1.09)	0.00416 (1.11)	0.00107 (0.29)
VC_BACKED=1	0.00487** (2.06)	0.00118 (0.59)	0.00319** (2.11)	0.00403*** (2.78)	0.00225* (1.78)
DESPAC=1 × VC_BACKED=1	0.0156 (1.49)	0.0148* (2.05)	0.00748 (1.55)	0.00161 (0.49)	0.00346 (1.19)
UWRANK	0.000568 (0.44)	0.000845 (0.85)	0.000966 (1.43)	0.000702 (1.66)	0.000793* (2.02)
DESPAC=1 × UWRANK	-0.00283 (-1.16)	-0.00315* (-1.82)	-0.00269** (-2.68)	-0.00224*** (-2.81)	-0.00288*** (-3.60)
LN_MARKET_CAP	-0.0000866 (-0.04)	-0.000576 (-0.39)	-0.000335 (-0.24)	-0.000147 (-0.12)	-0.000126 (-0.12)
DESPAC=1 × LN_MARKET_CAP	-0.00594 (-1.63)	-0.00918*** (-2.83)	-0.00588** (-2.28)	-0.00389 (-1.69)	-0.00237 (-0.98)
L.LN_RET	0.00886 (0.93)	0.0121 (1.52)	0.000581 (0.12)	0.00724 (1.20)	0.00607 (0.86)
DESPAC=1 × L.LN_RET	0.0160 (0.45)	0.0132 (0.49)	0.0189 (0.86)	0.0149 (0.98)	0.0143 (1.17)
L.LN_MA_VOL	0.0143*** (5.39)	0.0167*** (10.67)	0.0153*** (12.78)	0.0166*** (12.12)	0.0166*** (14.17)
DESPAC=1 × L.LN_MA_VOL	-0.00251 (-0.52)	-0.00246 (-0.92)	-0.00145 (-0.63)	-0.000542 (-0.24)	0.000512 (0.27)
LN_AMIHUDD_ILLIQ	0.00974*** (5.29)	0.0107*** (9.77)	0.0105*** (10.50)	0.0102*** (10.77)	0.00991*** (11.21)
DESPAC=1 × LN_AMIHUDD_ILLIQ	0.000416 (0.14)	-0.00190 (-0.93)	-0.00178 (-0.94)	-0.000868 (-0.46)	0.000526 (0.28)
FF17 Industry FE	x	x	x	x	x
Month FE	x	x	x	x	x
Constant	0.0653*** (3.54)	0.0569*** (4.33)	0.0676*** (6.42)	0.0522*** (6.18)	0.0454*** (5.75)
Observations	9742	23379	50457	99938	156113
Adjusted R <sup>2</sup>	0.447	0.445	0.431	0.441	0.456
Adjusted R <sup>2</sup> (Within)	0.329	0.329	0.298	0.305	0.316

\* p<0.1, \*\* p<0.05, \*\*\* p<0.01.

t statistics are reported in parentheses. Standard errors are double-clustered at the stock and month levels.

Table 17: Appendix: IPO Sample Summary Statistics, 2006-2020

Table 17 summarizes the IPO sample used in this study. The sample of IPO is constructed from the SDC New Issues Database and . The sample includes U.S. common stock IPO offerings between 2006 and 2020, excluding offerings with offer price below \$5, ADRs, units, REITs, closed-end funds, banks, ETFs, SPACs, direct listings, and stocks not included in CRSP such as OTC issues. Aggregate proceeds exclude overallocation. Amount left on the table is defined as the first-trading day market close price minus the offer price, multiplied by shares offered.

IPO Year	Count	Aggregate Proceeds (\$ Bil)	Mean First-Day Return Equal Weighted	Aggregate Amount Left on the Table (\$ Bil)
2006	157	27.1	12%	3.3
2007	163	29.9	14%	4.5
2008	24	22.8	15%	5.6
2009	46	13.3	11%	1.5
2010	100	30.1	9%	1.8
2011	84	26.4	13%	3.5
2012	97	31.8	18%	3.1
2013	157	38.1	21%	7.8
2014	216	43.2	15%	5.5
2015	123	22.8	21%	4.4
2016	77	12.4	14%	1.8
2017	112	23.2	14%	3.6
2018	145	33.4	18%	6.6
2019	123	39.5	22%	7.0
2020	171	62.5	39%	29.2

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