

# Bank Lines of Credit in Corporate Finance: An Empirical Analysis

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I empirically examine the factors that determine whether firms use bank lines of credit or cash in corporate liquidity management. I find that bank lines of credit, also known as revolving credit facilities, are a viable liquidity substitute only for firms that maintain high cash flow. In contrast, firms with low cash flow are less likely to obtain a line of credit, and they rely more heavily on cash in their corporate liquidity management. An important channel for this correlation is the use of cash flow-based financial covenants by banks that supply credit lines. I find that firms must maintain high cash flow to remain compliant with covenants, and banks restrict firm access to credit facilities in response to covenant violations. Using the cash-flow sensitivity of cash as a measure of financial constraints, I provide evidence that lack of access to a line of credit is a more statistically powerful measure of financial constraints than traditional measures used in the literature.

Bank lines of credit, or revolving credit facilities, are an instrumental component of corporate liquidity management. The “Liquidity and capital resources” sections of firms’ annual reports emphasize the importance of firms’ access to lines of credit; likewise, research reports by credit rating agencies such as Moody’s and Standard and Poor’s (S&P) detail information on revolving credit facilities when discussing a firm’s default risk. Despite the importance of lines of credit in the provision of liquidity in the economy, the absence of data has limited the existing empirical research on their role in corporate financing decisions. The analysis presented here represents one of the first empirical studies of lines of credit in the ongoing liquidity of public corporations.

While there is an extensive theoretical literature on bank lines of credit (Boot et al. 1987, Holmstrom and Tirole 1998, Martin and Santomero 1997), the extant empirical literature on corporate liquidity focuses mainly

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on the role of cash (Almeida et al. 2004, Faulkender and Wang 2006, Opler et al. 1999). The cash literature finds that cash plays an important liquidity role given that capital market frictions prevent firms from obtaining external sources of finance for valuable projects arising in the future.

The empirical finding that firms rely heavily on internal cash for liquidity is somewhat surprising, given hypotheses developed in the theoretical literature on lines of credit. This literature argues that lines of credit are motivated primarily by capital market frictions, and a committed line of credit overcomes these frictions by ensuring that funds are available for valuable projects. In other words, according to the theoretical literature, lines of credit should resolve precisely the capital market frictions that motivate firms to hold cash as a liquidity buffer. In addition, Kashyap et al. (2002) and Gatev and Strahan (2006) argue that banks are the most efficient liquidity providers in the economy, which also suggests that firms should rely on lines of credit over internal cash. Despite the similarities in the literature on cash and lines of credit, there is a lack of interaction between the two areas of research. The extant literature on cash is largely silent on why firms may use cash in place of lines of credit in corporate liquidity management.

This article attempts to bridge this gap. The central question of my analysis is: What governs the use of cash versus bank lines of credit in corporate liquidity management? I attempt to answer this question using a unique data set with two sets of variables collected directly from annual 10-K SEC filings. First, for the universe of public firms in S&P's Compustat from 1996 through 2003, the data set contains information on whether a firm has access to a line of credit. Second, for a random sample of 300 firms from this universe (1908 firm-year observations), the data set contains information on the size of the line of credit, the portion of the line of credit drawn, and the unused availability. In addition, the data set for the random sample contains information on whether firms are in compliance with or in violation of financial covenants associated with the line of credit. This data set is one of the first to contain detailed information on the use of lines of credit by a large sample of public firms.

I use this data set to explore why firms rely on cash versus lines of credit for liquidity. In the first set of results, I find evidence that maintenance of high cash-flow levels is a key characteristic that governs firms' use of lines of credit relative to cash. Firms with high levels of cash flow rely on lines of credit, whereas firms with low levels of cash flow rely on cash. After controlling for firm industry, size, asset tangibility, seasonal sales patterns, market-to-book ratio, and age, I find that increasing lagged cash flow by 2 standard deviations at the mean increases the likelihood of obtaining a line of credit by almost 0.10 at the mean, or about one-quarter standard deviation.

Using the random sample of 300 firms that contains information on line of credit balances, I focus on the *bank liquidity to total liquidity ratio*. This ratio is defined as the ratio of lines of credit to the sum of lines of credit and cash; it represents the fraction of total liquidity available to the firm provided by bank lines of credit. While some firms may have higher demand for *total* liquidity due to seasonal product markets or better investment opportunities, this ratio isolates the *relative* attractiveness of lines of credit versus cash in corporate liquidity management. I find a positive effect of cash flow on the bank liquidity to total liquidity ratio. More specifically, an increase in lagged cash flow by 2 standard deviations at the mean increases the bank liquidity to total liquidity ratio by almost 0.08 at the mean, or about one-quarter standard deviation. This positive relationship is robust when I isolate the intensive margin and examine only firms that have a line of credit, although the magnitudes and statistical significance are slightly weaker.

This result suggests that maintenance of high cash flows is a critical determinant of whether a firm uses lines of credit versus cash in corporate liquidity management. When I split the sample into firms with high and low probabilities of financial distress as measured by Altman (1968), I find that the positive relationship between the use of lines of credit and lagged cash flow is unique among firms with high financial distress likelihoods. There is no such correlation among firms with low distress likelihoods. In other words, when a firm has a significant probability of financial distress, it more heavily uses lines of credit relative to cash only if it maintains high cash flow.

What explains the positive correlation between cash flow and the use of lines of credit? In the second set of results, I explore the importance of cash flow-based financial covenants on lines of credit. In particular, I find evidence that maintenance of cash flow is critical to avoiding financial covenant violations. Reductions in cash flow are a stronger predictor of financial covenant violations than are changes in a firm's current ratio, net worth, or market-to-book ratio. In addition, I find that when a firm violates a covenant, it loses access to a substantial portion of its line of credit. In terms of magnitudes, a covenant violation is associated with a 15 to 25% drop in the availability of both total and unused lines of credit. It is also associated with a 10 to 20% decrease in the bank liquidity to total liquidity ratio.

This result helps explain why cash flow is an important determinant of a firm's use of lines of credit versus cash in corporate liquidity management. Given that lines of credit are contingent on maintenance of cash flow-based covenants, they represent a poor liquidity substitute for firms with low current or expected cash flows. Firms with low current or expected cash flows maintain cash balances as a liquidity buffer given that lines of credit may not be available when most needed. This result also shows that lines of

credit are not totally committed liquidity insurance. The *contingent* lines of credit that exist in the marketplace are distinct from the *committed* lines of credit that are described in the theoretical literature.

In the third set of results, I provide evidence that access to lines of credit as a measure of financial constraints adds valuable information to traditional measures of constraints used in the literature. Theoretical research suggests that lines of credit are critical in reducing future capital market frictions facing firms, yet they have not been considered in the extant literature on financial constraints. I follow Almeida et al. (2004) and examine the cash-flow sensitivity of cash among firms with and without access to lines of credit. The Almeida et al. (2004) theoretical insight is that firms that face capital market frictions are likely to save cash out of cash flow, whereas firms that do not face frictions should show no systematic pattern of cash savings out of cash flow. They empirically explore the cash-flow sensitivity of cash for constrained versus unconstrained firms, where they use four traditional measures of financial constraints: whether a firm is small, whether a firm has a low payout ratio, whether a firm does not have a corporate credit rating by S&P, and whether a firm does not have a commercial paper rating by S&P.

Instead of relying on these traditional measures, I explore the cash-flow sensitivity of cash using a measure of constraints that relies on access to lines of credit. Theoretical research on credit lines suggests that line of credit access as a measure of financial constraints adds valuable information to traditional measures used in the literature. I define as “unconstrained” firms that have two key characteristics. First, they have a line of credit in every year in which they are in the sample. Second, they maintain cash flows scaled by book assets above the median firm throughout the sample. Firms that do not meet this criteria are designated “constrained.” The empirical results using this definition show that firms without access to a line of credit save cash out of cash flow, whereas firms with access to a line of credit do not save cash out of cash flow. In addition, I show evidence that the line of credit measure is more statistically powerful at explaining the pattern of cash-flow sensitivities of cash than the traditional measures used in the literature. For example, consistent with Almeida et al. (2004), firms without an S&P corporate credit rating or commercial paper rating indeed show a higher sensitivity of cash holdings to cash flow. However, among firms without access to a rating, it is *only* the firms *without* access to a line of credit that show a positive sensitivity. In general, my results show that firms that are small, have low payout ratios, or lack ratings *only* show positive cash-flow sensitivities of cash *if* they lack access to a line of credit.

Overall, these results suggest that banks provide credit lines that are contingent on maintenance of cash flow. Reductions in cash flow lead to covenant violations, which in turn lead to a restriction in the availability

of a line of credit. Lines of credit are therefore a poor liquidity substitute for firms that have low existing or expected cash flows. For these firms, cash is a more reliable source of liquidity. These firms rely more heavily on cash and save more cash out of cash flow.

In addition to these results, this article documents several new facts regarding the use of bank lines of credit by public firms. For example, I find that lines of credit are a very large and important source of corporate finance in the economy. Almost 85% of firms in my sample obtained a line of credit between 1996 and 2003, and the line of credit represents an average of 16% of book assets. I also find that lines of credit are utilized among firms that are completely equity financed; 32% of firm-year observations in which no outstanding debt is recorded on the balance sheet have an available unused line of credit. Firms with access to public debt do not cease using revolving credit facilities: 95% of firm-year observations that have corporate credit rating from S&P also have a bank line of credit, and line of credit borrowings represent 12% of total debt outstanding for these firms.

The rest of this article proceeds as follows. In Section 1, I describe lines of credit, the existing literature, the data, and summary statistics. In Section 2, I describe the theoretical framework that motivates the article. Sections 3 through 5 present the empirical analysis, and Section 6 concludes.

## 1. Description, Existing Research, Data, and Summary Statistics

### 1.1 Description and existing research

A firm that obtains a line of credit receives a nominal amount of debt capacity against which the firm draws funds. Lines of credit, also referred to as revolving credit facilities or loan commitments, are almost always provided by banks or financing companies. They can be provided by one bank or multiple banks through syndication. The used portion of the line of credit is a debt obligation, whereas the unused portion remains off the balance sheet. In terms of pricing, the firm pays a commitment fee that is a percentage of the unused portion, and a predetermined interest rate on any drawn amounts. Pricing and maturity data are not always available directly from annual 10-K SEC filings; in a sample of 11,758 lines of credit obtained by 4011 public firms between 1996 and 2003 in Loan Pricing Corporation's *Dealscan*, the median commitment fee is 25 basis points, the median interest rate on drawn funds is 150 basis points above LIBOR, and the median maturity is 3 years.

Corporations detail lines of credit in their annual 10-K SEC filings. Regulation S-K of the U.S. Securities and Exchange Commission requires firms to discuss explicitly their liquidity, capital resources, and result of operations (Kaplan and Zingales 1997). All firms filing with the SEC therefore provide information on the used and unused portions of bank

lines of credit, and whether they are out of compliance with financial covenants. For example, Lexent Inc., a broadband technology company, details their line of credit in their FY 2000 10-K filing as follows:

At December 31, 2000, the Company had notes payable to banks aggregating \$2.0 million under a \$50 million collateralized revolving credit facility, which expires in November 2003. Borrowings bear interest at the prime rate or at a rate based on LIBOR, at the option of the Company. This credit facility is to be used for general corporate purposes including working capital. As of December 31, 2000, the prime rate was 9.5%.

In the 10-K filing, companies typically detail the existence of a line of credit and its availability in the liquidity and capital resources section under the management discussion, or in the financial footnotes explaining debt obligations.

Although information on credit lines is available in annual 10-K SEC filings, the existing empirical research on bank lines of credit relies on alternative data sources. Ham and Melnik (1987) collect data from a direct survey of 90 corporate treasurers. They find that draw downs on lines of credit are inversely related to interest rate cost and positively related to total sales. Agarwal et al. (2004) examine the use of lines of credit for 712 privately held firms that obtained loans from FleetBoston Financial Corporation. They also find that firms with higher profitability obtain larger credit lines, which is consistent with evidence presented here. Berger and Udell (1995) use data on lines of credit extended to small private businesses and show that firms with longer banking relationships pay lower interest rates and are less likely to pledge collateral. Petersen and Rajan (1997) find that small private businesses without access to bank credit lines rely more heavily on trade credit. Shockley and Thakor (1997) focus on the contract structure of credit lines. While Kaplan and Zingales (1997) and Houston and James (1996) present data on unused lines of credit, they do not explore the relationship between lines of credit and firm characteristics. This article is the first, to my knowledge, to systematically analyze balances of used and unused bank lines of credit at public corporations.

## **1.2 Data**

I begin with a *Compustat* universe that contains non-financial U.S.-based firms with at least four consecutive years between 1996 and 2003 of positive data on total assets (*item 6*), and four consecutive years of non-missing data on total liabilities (*item 181*), total sales (*item 12*), a measure of EBITDA (*item 13*), share price (*item 199*), shares outstanding (*item 25*), preferred stock (*item 10*), deferred taxes (*item 35*), and convertible debt (*item 79*). These data limitations are governed by the necessity of these variables in

constructing basic financial characteristics of the firm. I also require firms to have four consecutive years of book leverage ratios between 0 and 1. This yields a sample of 4604 firms, which I follow from 1996 through 2003 (31,533 firm-year observations). I focus on the 1996 to 2003 period because annual 10-K SEC filings are available electronically for all firms in the years after 1995, which makes the costs of data collection much lower for this time period. I restrict the sample to firms with at least four consecutive years of data because I am particularly interested in how line of credit use evolves for a given firm over time. This sample, which I refer to as the “full” sample, forms the basis for the text searching program described below, which provides information on whether firms have a line of credit.

I then form a smaller data set based on a random sample of 300 firms, which I refer to as the “random” sample. I randomly sample 300 firms from the 4604 firms in the full sample, and I follow them from 1996 through 2003, for a total unbalanced panel of 2180 firm-year observations. The random sample represents 6.5% of the firms in the full sample. The random sample forms the basis for manual examination of annual 10-K SEC filings described below.

In the rest of this section, I describe the data searching process that produces measures of firms’ utilization of lines of credit. It is important to understand that the process is iterative. I first use the full sample of annual 10-K SEC filings to search for certain phrases that are indicative of a firm’s having a line of credit. I then use the random sample and manually read the annual 10-K SEC filings to assess whether the search terms are providing an accurate or inaccurate portrayal of the firms that have bank lines of credit. The search procedure described below is the final version of this iterative process.

For the full sample, I link each firm-year observation from *Compustat* to the electronic version of its annual 10-K SEC filing. I then search each filing for 7 terms: “credit lines,” “credit facility,” “revolving credit agreement,” “bank credit line,” “working capital facility,” “lines of credit,” and “line of credit.” In the initial pass through the text filings, I create a variable that is 1 if the annual 10-K SEC filing has any of the search terms in the document, and 0 if the search terms are not in the 10-K filing.

This search algorithm leads to some errors in classifying whether firms in reality have a bank line of credit or revolving credit facility. In terms of classifying the errors, the null hypothesis is that the firm-year observation in question *does not* have a line of credit. A Type I error occurs when the search program determines that the firm *does* have a bank line of credit (rejection of the null hypothesis) when it in fact *does not* have a bank line of credit. A Type II error occurs when the search program determines that the firm *does not* have a line of credit (acceptance of the null hypothesis) when it in fact *does* have a line of credit.

In order to limit Type I errors, the search program produces a document that, for every firm-year observation, contains the 10 lines of text before and after each occurrence of a search term in the 10-K filing. Using this document, I manually search for obvious Type I errors. More specifically, I find occurrences in which one of the search terms is directly proceeded with “no,” “do not have a,” “not have any,” “retired our,” “terminated our,” and “equity.” I also search for the following phrases that may occur right after one of the search terms: “expired,” “terminated,” and “was terminated.” I manually conduct this search for Type I errors, and make sure that each is in fact a Type I error before reclassifying the firm-year observation as having no line of credit. This collection procedure results in a {0, 1} variable measuring whether or not every firm in the full sample has access to a line of credit.

For the random sample of 300 firms, I manually collect detailed data on used and unused lines of credit from annual 10-K SEC filings. It is important to emphasize that there is no search program used to collect line of credit data for the random sample; they are collected manually. I collect data on whether the firm has access to a line of credit and the used and unused portion of the line of credit. If a line of credit backs up a commercial paper program, any outstanding commercial paper is subtracted from the line of credit, but is not recorded as a used portion of the line of credit. Any balance of the backup line of credit that does not support outstanding commercial paper is recorded as an unused part of the line of credit. This is consistent with the actual reporting done by firms. It is important to note that borrowers with a commercial paper backup line of credit draw down the portion of the line that does not backup outstanding commercial paper. Only 5% of firms in my sample have a commercial paper program, and all results are robust to the complete exclusion of these firms.

One advantage of the random sample data collection is that it allows me to assess directly the errors in the search program used for the full sample. In other words, by comparing the line of credit collection from manual inspection with the search program in the random sample, I am able to assess the number of Type I and Type II errors associated with the search program. I find the search program produces Type I errors in 8.6% of all observations in the random sample, and Type II errors in 2.2% of all firm-year observations in the random sample.<sup>1</sup>

For the random sample, I also collect data on whether or not a firm is in violation of a financial covenant associated with the line of credit;

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<sup>1</sup> The relatively large number of type I errors in the full sample (8.6%) is a disadvantage of the data collection procedure. However, these errors are likely to be random. They are based only on slight variations in language used in the annual reports, and are not systematic in any way. Therefore, these errors should bias coefficients toward 0 in the regression analysis.



these are covenants that require the maintenance of financial ratios.<sup>2</sup> The SEC requires firms to report when they are in violation of a financial covenant: “companies that are, or are reasonably likely to be, in breach of such covenants must disclose material information about that breach and analyze the impact on the company if material (SEC 2003).

Core financial variables are calculated from *Compustat* and are defined as follows. Book debt is short-term debt plus long-term debt (*item 34 + item 9*), all divided by total assets (*item 6*). Balance sheet cash is measured using *item 1*. A measure of asset tangibility is tangible assets (*item 8*) divided by non-cash total assets. The market-to-book ratio is defined as total assets less the book value of equity plus the market value of equity less cash, all divided by non-cash total assets. The book value of equity is defined as the book value of assets (*item 6*) less the book value of total liabilities (*item 181*) and preferred stock (*item 10*) plus deferred taxes (*item 35*). The market value of equity is defined as common shares outstanding (*item 25*) multiplied by share price (*item 199*). The primary measure of cash flow is EBITDA (*item 13*) divided by non-cash total assets. Net worth—cash adjusted is defined as non-cash total assets less total liabilities, divided by non-cash assets.

As further described in Section 3, I scale cash flow, asset tangibility, net worth, and the market-to-book ratio with *non-cash* book assets. I do so because firms are likely to jointly determine cash holdings and line of credit usage. This joint determination leads to a mechanical negative correlation between any measure scaled by total assets and the use of lines of credit. For example, suppose one constructs the measure of tangibility as tangible assets scaled by total assets. Given that cash is included in total assets, and given that firms without access to a line of credit hold higher cash balances, this classification of asset tangibility leads to a mechanical negative correlation with lines of credit. A disadvantage of using non-cash assets in place of total assets to scale cash flow and asset tangibility is that it leads to extreme outliers. In order to reduce the influence of outliers, I Winsorize all financial variables from *Compustat* at the 5th and 95th percentile.

I drop any firm-year observation for which any of the variables constructed above are missing. The final full sample contains 4503 firms (28,447 firm-year observations) and the final random sample contains 300 firms (1908 firm-year observations).

### 1.3 Summary statistics

Table 1 contains summary statistics for the full sample (left panel) and the random sample (right panel). In the full sample, 81.7% of firm-years have

<sup>2</sup> The material adverse change clause (MAC) is also an important feature in bank loan agreements. However, I find little evidence from annual 10-K SEC filings that this clause is invoked with frequency.

**Table 1**  
**Summary statistics.**

Full sample				Random sample			
Variable	Mean	Median	St. Dev.	Variable	Mean	Median	St. Dev.
<i>Line of credit variables</i>				<i>Line of credit variables</i>			
Has line of credit {0,1}	0.817	1.000	0.387	Has line of credit {0,1}	0.748	1.000	0.434
				Total line of credit/assets	0.159	0.112	0.169
				Unused line of credit/assets	0.102	0.069	0.125
				Used line of credit/assets	0.057	0.000	0.097
				Total line/(total line + cash)	0.512	0.569	0.388
				Unused line/(unused line + cash)	0.450	0.455	0.373
				Violation of financial covenant {0,1}	0.080	0.000	0.271
<i>Firm characteristics</i>				<i>Firm characteristics</i>			
Book debt/assets	0.204	0.171	0.19	Book debt/assets	0.205	0.169	0.196
EBITDA/(assets – cash)	0.026	0.125	0.358	EBITDA/(assets – cash)	0.034	0.126	0.353
Tangible assets/(assets – cash)	0.340	0.277	0.239	Tangible assets/(assets – cash)	0.331	0.275	0.226
Net worth, cash adjusted	0.436	0.445	0.238	Net worth, cash adjusted	0.451	0.459	0.234
Assets – cash	1,608	102	11,434	Assets – cash	1,441	116	7682
Market-to-book, cash adjusted	2.944	1.532	3.469	Market-to-book, cash adjusted	2.790	1.498	3.319
Industry sales volatility	0.043	0.035	0.033	Industry sales volatility	0.045	0.036	0.035
Cash-flow volatility	0.091	0.052	0.109	Cash-flow volatility	0.091	0.052	0.109
Not in an S&P index {0,1}	0.696	1.000	0.46	Not in an S&P index {0,1}	0.684	1.000	0.465
Traded over the counter {0,1}	0.123	0.000	0.328	Traded over the counter {0,1}	0.142	0.000	0.349
Firm age (years since IPO)	15	10	13	Firm age (years since IPO)	14	9	12

This table presents summary statistics for two samples of non-financial firms from 1996 through 2003. The left panel describes the full sample of 4503 firms (28,447 firm-year observations), and the right panel describes the random sample of 300 firms (1908 firm-year observations). *Market-to-book, cash adjusted*, is the market value of assets less cash divided by the book value of assets less cash balances. *Net worth, cash adjusted*, is the net worth less cash balances divided by book assets less cash.

a line of credit. As discussed above, this percent is overstated given the Type I errors associated with the search program. A more reliable estimate is 74.8%, which is the percent of firm-years that have a line of credit in the random sample. Overall, 85% of firms in the random sample have a line of credit at some point between 1996 and 2003. Using the hand-collected data on line of credit balances in the random sample, I construct a variety of measures to assess the magnitude of lines of credit in corporate liquidity management. On average, the total line of credit represents 16% of book assets; the unused portion represents 10% of book assets and the used portion represents 6%. Given total debt is 21% of book assets in the random sample, this implies that used lines of credit on average represent more than a quarter of outstanding debt balances among public firms. These statistics suggest that lines of credit are widely used by public firms, and they represent large amounts of used debt and unused debt availability.

In order to assess the importance of lines of credit in corporate liquidity management, I create two measures of the *bank liquidity to total liquidity ratio*. The first measure is the total line of credit balance scaled by the sum of total lines of credit and cash. The second measure is the unused line of credit balance scaled by the sum of unused lines and cash. The second measure captures the fraction of liquidity available to the firm in the form of lines of credit. The first measure takes into account mechanical endogeneity concerns that certain types of firms consistently draw down heavily on existing lines of credit. By either measure, bank liquidity represents about 50% of total liquidity available to firms, which suggests that credit lines are an instrumental component of corporate liquidity.

Table 2 provides additional evidence of the importance of credit lines. Firms from all major industries heavily utilize this financial product. The lowest fraction of firms with lines of credit is in the services industry, where 63% of firm-year observations have a line of credit. The highest use is in both wholesale and retail trade, with over 90% of firm-year observations maintaining a line of credit. The industry patterns most likely reflect a seasonal sales component, something I further explore in the results below. A common measure of whether firms have access to public debt is whether they have a corporate credit rating by S&P (Faulkender and Petersen 2006). Table 2 demonstrates that firms with access to public debt extensively utilize bank lines of credit. Almost 95% of firms with a corporate credit rating have access to a line of credit, and outstanding line of credit balances represent  $(0.245 \times 0.190 =)$  4.7% of total assets. Given that total debt is 38% of total assets for these firms, this implies that 12% of total debt outstanding for firms with a corporate credit rating is in the form of used lines of credit. The evidence suggests that lines of credit are an important part of debt policy even for firms that have access to public debt. Finally, the last two rows of Table 2 show that even firms that have

**Table 2**  
Which firms utilize bank lines of credit?

	Line of credit {0,1}	Some debt {0,1}	Conditional on having line of credit		
			Debt/ assets	Total line/assets	Used line/ total line
<i>Industry</i>					
Agriculture, minerals, construction	0.761	0.862	0.303	0.306	0.490
Manufacturing	0.731	0.809	0.205	0.216	0.267
Transportation, communications, and utilities	0.860	0.907	0.366	0.163	0.338
Trade—wholesale	0.932	0.915	0.256	0.221	0.321
Trade—retail	0.920	0.938	0.286	0.194	0.283
Services	0.630	0.701	0.245	0.204	0.304
<i>Corporate credit rating</i>					
No S&P corporate credit rating	0.697	0.765	0.196	0.220	0.319
S&P corporate credit rating	0.945	0.995	0.384	0.190	0.245
<i>Debt outstanding</i>					
No debt outstanding	0.319	0.000	0.000	0.110	0.000
Debt outstanding	0.847	1.000	0.267	0.221	0.325

This table presents data on the use of lines of credit by firms in the random sample of 300 firms (1908 firm-year observations). It reports cell means for subsamples by industry, by having a corporate credit rating, and by having debt outstanding.

no debt outstanding have access to lines of credit. Conditional on having no debt outstanding, 32% of firm-years have an unused line of credit available, and the unused line of credit represents 11% of total assets.

## 2. Theoretical Motivation

In this section, I motivate the empirical analysis by discussing the existing theoretical research in two areas: cash holdings and bank lines of credit. I focus on how an empirical analysis of lines of credit can help resolve unanswered questions in both of these areas.

Almeida et al. (2004) argue that cash holdings represent a safeguard against the inability to obtain financing when valuable opportunities arise. They build a three-period model in which investment opportunities arrive in the first and second periods. Firms are either financially constrained or unconstrained; firms fall into one of these categories on the basis of the level of cash flows and the value of collateral that the firm can pledge to creditors. In the initial period, unconstrained firms have no reason to save cash out of initial cash flows; they can reduce dividends or raise more external financing in the second period to pursue investment opportunities. Constrained firms, on the other hand, retain a portion of their first-period cash flows to “hedge” against the inability to raise external financing in the second period. The optimal level of saving out of cash flow weighs the cost of reducing investment in the first period with the benefit of more investment in the second period. Constrained firms should therefore save a higher proportion of their initial cash flows relative to unconstrained firms.

Empirical support for this framework is found in Almeida et al. (2004); Faulkender and Wang (2006) (henceforth FW); and Opler et al. (1999) (henceforth OPSW). Almeida et al. (2004) sort their sample on the basis of observable measures of financial constraints (payouts, size, and the existence of third-party credit ratings), and find that more constrained firms save more cash out of cash flow. FW find that shareholders place higher value on an additional dollar of cash within financially constrained firms, where the measures of financial constraints used are similar to those in Almeida et al. (2004). OPSW find that larger firms and those with credit ratings hold less cash.

While the theoretical literature on cash is instructive, there are two shortcomings. First, the cash literature does not provide direct insight into the precise financing constraint that prevents firms from accessing external funds. The theoretical frameworks of Almeida et al. (2004) and FW rely only on a non-specific “limitation in [the] capacity to raise external finance” (Almeida et al., p 1781). They do not take an empirical stand on what the limitation is. Second, this literature does not discuss the importance of bank lines of credit. As demonstrated in Tables 1 and 2, bank lines of credit are widely used by public firms, yet the cash literature does not isolate how lines of credit and cash policy may be jointly determined.

This last shortcoming is especially important given that a large body of theoretical literature on lines of credit argues that this financial product is designed to overcome precisely the types of capital market frictions discussed in the cash literature. These articles include Berkovitch and Greenbaum (1991), Boot et al. (1987), Duan and Yoon (1993), Holmstrom and Tirole (1998), Morgan (1994), and Shockley (1995).

I focus here on two of these articles I believe demonstrate the core intuition of these models. Holmstrom and Tirole (1998) motivate the use of lines of credit by embedding a moral hazard problem within a three-period model in which a liquidity shock is realized in the second period. When the liquidity shock is realized in the second period, the borrower must retain a large enough portion of the third period return to motivate her to be diligent; in other words, there is a standard moral hazard problem that forces the borrower to retain a large stake in the project. Given this agency problem, the first best investment level is unattainable. If the liquidity shock is large enough, the borrower will not be able to obtain funds even if the project is profitable given that she must retain enough of the project return to maintain diligence. In the second best solution, the borrower buys liquidity insurance. One mechanism is a line of credit.<sup>3</sup> In the first

<sup>3</sup> Holmstrom and Tirole (1998) emphasize that the line of credit must be irrevocable, and that the liquidity shock is verifiable. In other words, there is no possibility that borrowers misallocate the funds available under the line of credit. In addition, Holmstrom and Tirole (1998) emphasize that other types of financing arrangements may serve the purpose of a bank line of credit in their model, as long as the arrangement provides unconditional financing.

period, creditors provide a commitment to lend in the second period up to a certain point. When the liquidity shock is realized, the borrower has access to committed funds. In some states of the world, the creditors end up losing money in the second period, but they break even in expectation. This is the intuition of the liquidity insurance in the model.

Boot et al. (1987) also use a basic agency problem to motivate corporate demand for lines of credit. They employ a three-period model with an agency problem, where borrowers select an effort level in the first period and choose whether to invest in the second period. The moral hazard problem arises because the effort decision is unobservable to creditors. In the Boot et al. (1987) model, there is a stochastic interest rate realized in the second period that serves a role similar to that of the liquidity shock in Holmstrom and Tirole (1998). If interest rates are too high in the second period, borrowers anticipate a low expected return from the project and thus choose low effort. In other words, high interest rates in the second period lower the return to effort, which leads managers at borrowing firms to shirk. In the second period, banks fully predict such behavior, and thus ration credit. A line of credit signed in the first period solves this problem by charging an up-front fee and guaranteeing a low rate of interest in the second period. Thus, the line of credit serves as interest rate protection, which can guarantee that borrowers put in high effort initially.

According to the theoretical literature, lines of credit are committed liquidity insurance that should protect firms against future capital market frictions. In addition, Kashyap et al. (2002) find that banks have natural cost advantages in the provision of liquidity given deposit-based financing. Likewise, Gatev and Strahan (2006) find that banks are better liquidity providers given that deposits flow into banks during aggregate financial crises.

The theoretical literature on lines of credit and banks' advantages in liquidity provision suggests that firms should rely fully on bank lines of credit in their liquidity management. This, in some sense, is the null hypothesis that is rejected by the cash literature. Clearly, lines of credit do not provide sufficient liquidity insurance for all firms in the economy. The primary goal of this article is to resolve empirically these two research areas by exploring what determines whether firms use cash or lines of credit in corporate liquidity management. More specifically, I examine whether low cash flow, low asset tangibility, small size, low firm net worth, or low market-to-book ratios make lines of credit difficult to obtain and maintain. This analysis should provide insight into the precise friction that makes lines of credit a poor liquidity substitute for cash for some firms.

There are three additional hypotheses from the line of credit theoretical literature that I examine in the empirical analysis. First, the models assume that basic agency problems due to information asymmetry motivate the

use of lines of credit. In other words, firms in which management actions are less transparent are more likely to use lines of credit. Second, a bank line of credit must provide some degree of “commitment” if it is to improve on spot market financing. If banks can fully renegotiate the line of credit in the interim period, the contract will not improve on spot market financing. In the line of credit models described above, the optimal behavior for the bank in some states of the interim period is to restrict access to the line of credit. The empirical section of this article attempts to quantify the extent to which lines of credit represent unconditional obligations of banks. The third main empirical hypothesis that comes from these models is that it can be difficult for firms to raise capital in spot markets when investment opportunities arrive or change. Lines of credit provide a particularly flexible source of debt financing that can be drawn upon with fewer difficulties. Lines of credit should therefore be used in industries with higher earnings or sales volatility.

### 3. Lines of Credit versus Cash

#### 3.1 Empirical specification

In this section, I conduct estimations to examine which firm characteristics influence the decision to utilize lines of credit as opposed to cash in corporate liquidity management. There are two sets of dependent variables. First, I examine a  $\{0,1\}$  indicator variable for whether the firm has a line of credit. For this dependent variable, I examine the effect of firm characteristics on the probability of having a line of credit using maximum likelihood probit estimation. Second, I examine the bank liquidity to total liquidity ratio, which varies from 0 to 1. For this dependent variable, I examine the effect of firm characteristics on the ratio using linear (OLS) estimation. When I examine the bank liquidity to total liquidity ratio, I also examine the intensive margin for which I isolate the sample to only firms that have a line of credit.<sup>4</sup> I examine the intensive margin independently to demonstrate that cash flow is positively correlated with the bank liquidity to total liquidity ratio, even among firms that have a line of credit. In all regressions, standard errors are clustered at the firm level and all regressions include year and 1-digit SIC industry indicator variables.<sup>5</sup>

<sup>4</sup> Alternatively, I use a maximum likelihood Tobit specification to simultaneously capture the extensive and intensive margin. Results (unreported) are qualitatively similar.

<sup>5</sup> I use 1-digit SIC indicator variables instead of more disaggregated industry indicators because the use of finer industry controls results in the loss of observations in the random sample when conducting the maximum likelihood probit estimation (given perfect prediction in more disaggregated industry groups). The core results of the analysis are either the same or slightly *stronger* when I use 2-digit SIC indicator variables, but this results in a loss of 262 observations (out of 1908) in the probit estimation in the random sample.

The variables in the matrix  $X$  are motivated by the theoretical framework outlined in Section 2. First, I examine firm characteristics likely to be associated with firms facing a high cost of external relative to internal finance. In other words, I attempt to explain why firms may be forced to rely on internal cash as opposed to using a bank line of credit. I use five measures. First, firm cash flow is measured as EBITDA scaled by non-cash total assets. I employ EBITDA because it is the most common measure of cash flow used by commercial banks when setting various types of covenants on lines of credit. All results are robust to a more common measure of cash flow: the sum of EBIAT and depreciation, all scaled by non-cash total assets. Second, asset tangibility is measured as tangible assets scaled by non-cash total assets. Third, firm size is measured as the natural logarithm of non-cash total assets. I also include net worth scaled by non-cash assets and the market-to-book ratio.

I employ non-cash total assets as opposed to total assets to scale variables. As mentioned in Section 1, I scale by non-cash total assets because firms are likely to jointly determine cash balances and their utilization of lines of credit when setting corporate financial policy. The use of cash in the denominator of the variables in the matrix  $X$  will therefore lead to a mechanical negative correlation. For example, a key finding of the article is that firms with low cash flow scaled by non-cash assets are less likely to utilize a line of credit. This negative correlation between cash flow and line of credit utilization is *mechanical* if cash is included in the assets measure used to scale cash flow, as long as cash flows are positive on average. For the same nominal cash flow, firms without a line of credit will hold more cash than those with a line of credit, which will mechanically lead to a lower level of cash flow scaled by total assets for the firms that do not have a line of credit. Consistent with this mechanical bias, the negative correlation between cash flow and the utilization of lines of credit is much stronger if I scale by total assets as opposed to non-cash assets.

In addition to determining what factors force firms to rely on cash instead of lines of credit, the empirical analysis also attempts to quantify the importance of information asymmetry and business variability on the use of lines of credit. I construct measures of information asymmetry that are consistent with measures in Faulkender and Petersen (2006) and Sufi (2007). Firms with equity that is not traded on a major exchange receive less analyst coverage and media attention. Likewise, firms that are not in one of the three main S&P indices (the S&P 500, the S&P Midcap 400, and the S&P Smallcap 600) also receive less attention. I use an indicator variable for whether the firm's equity trades only over the counter and I use an indicator for whether the firm is NOT included in one of the main S&P indices to measure information asymmetry. Older firms are also more likely to be known to capital markets. I include the natural logarithm of 1



+ the years since the firm's IPO as an additional measure of information asymmetry. The year of the firm's IPO is approximated using the first year in Compustat that the firm's share price is available.

I use two variables to measure business variability. First, I include the median within-year standard deviation of sales for all firms in the given firm's 3-digit SIC code industry.<sup>6</sup> I refer to this variable as *seasonality*. Firms in 3-digit industries that show a larger degree of seasonality in sales may desire lines of credit to manage working capital and inventories. I also include a measure of the variability of cash flow, which is based on the measure used in Mackie-Mason (1990). It represents the standard deviation of annual changes in the level of EBITDA over a lagged four-year period, scaled by average non-cash assets in the lagged period.

### 3.2 Results

Columns 1 and 2 of Table 3 report coefficient estimates from maximum likelihood probit estimation relating the probability of having a line of credit to various firm characteristics. Column 1 shows results from the full sample and column 2 shows results from the random sample. Firm cash flow has a strong positive effect on the probability a firm utilizes a line of credit. The coefficient estimate in column 2 suggests that a two-standard-deviation increase in cash flow (0.7) at the mean leads to  $(0.7 \times 0.124 =) 9\%$  increase in the probability of obtaining a line of credit, which is almost a one-quarter-standard-deviation increase. Consistent with evidence that high market-to-book firms use less debt generally, there is a negative correlation between the market-to-book ratio and the probability of having a line of credit. Size is also a strong statistical predictor of the use of bank lines of credit. A two-standard-deviation increase in the natural logarithm of total assets (4) leads to a  $(4 \times 0.052 =) 21\%$  increase in the probability of obtaining a line of credit, which is a more than one-half standard deviation.

In terms of business variability, the coefficient estimates in columns 1 and 2 suggest that 3-digit SIC code within-year sales volatility has a positive effect on the use of lines of credit. The effect of the sales seasonality variable would be even stronger in the absence of the 1-digit SIC industry fixed effects. There is no statistically strong evidence that information asymmetry has an effect on the probability of utilizing a line of credit.

The results in columns 1 and 2 suggest that firms with low levels of cash flow are less likely to utilize bank lines of credit. Figure 1 shows that they use cash instead. It maps both the fraction of firms that have a line of credit

<sup>6</sup> This variable is constructed as follows: I use the entire set of firms with data available in the Compustat quarterly industrial files. For every firm-year, I calculate the standard deviation of the quarterly differences in sales, scaled by average assets over the year. I then obtain the median across all 3-digit SIC industries, for every given year. This variable is then merged onto each firm-year observation with the same 3-digit SIC code. This measure is similar to the earnings variance measure used in Mackie-Mason (1990).

**Table 3**  
**Bank lines of credit and firm characteristics.**

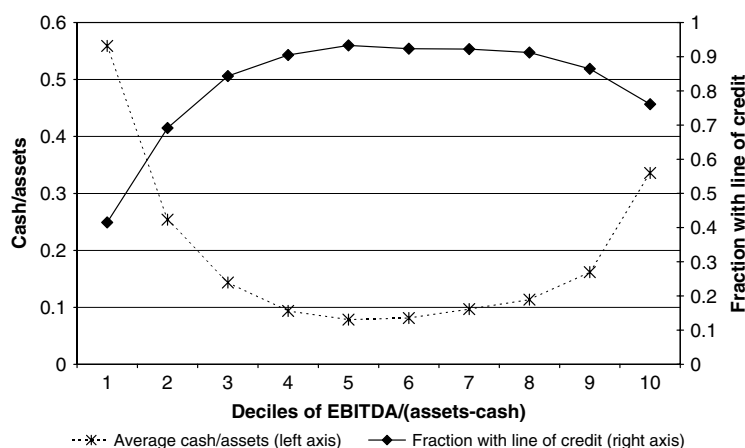
Dependent variable Regression type	Firm has line of credit (0,1) Probit (marginal effects)		Total line/(total line + cash) OLS		Unused line/(unused line + cash) OLS	
Sample	Full (1)	Random (2)	Random (3)	With line of credit (4)	Random (5)	With line of credit (6)
[EBITDA/ (assets – cash)] <sub>t-1</sub>	0.107** (0.011)	0.124* (0.059)	0.109** (0.041)	0.086 (0.056)	0.085* (0.038)	0.120* (0.056)
[Tangible assets/ (assets – cash)] <sub>t-1</sub>	0.015 (0.019)	0.070 (0.084)	0.041 (0.072)	0.014 (0.064)	0.025 (0.068)	–0.006 (0.068)
[Ln(assets – cash)] <sub>t-1</sub>	0.040** (0.003)	0.052** (0.015)	0.054** (0.010)	0.032** (0.009)	0.054** (0.010)	0.041** (0.009)
[Net worth, cash adjusted] <sub>t-1</sub>	–0.094** (0.016)	–0.037 (0.075)	–0.106 (0.059)	–0.136** (0.052)	–0.053 (0.054)	–0.040 (0.054)
[Market-to-book, cash adjusted] <sub>t-1</sub>	–0.013** (0.001)	–0.022** (0.006)	–0.031** (0.004)	–0.037** (0.005)	–0.025** (0.004)	–0.030** (0.006)
[Industry sales volatility] <sub>t-1</sub>	0.674** (0.193)	2.571** (0.885)	0.989** (0.382)	0.081 (0.319)	1.069** (0.401)	0.354 (0.366)
[Cash-flow volatility] <sub>t-1</sub>	–0.020 (0.035)	–0.201 (0.146)	–0.383** (0.125)	–0.374* (0.177)	–0.291* (0.118)	–0.271 (0.181)
[Not in an S&P index{0, 1}]	0.035* (0.013)	0.046 (0.059)	0.065 (0.039)	0.047 (0.034)	0.034 (0.039)	0.026 (0.035)
[Traded over the counter{0, 1}]	–0.014 (0.013)	–0.045 (0.061)	0.077 (0.047)	0.129** (0.036)	0.022 (0.043)	0.029 (0.044)
Ln[Firm age (years since IPO)] <sub>t-1</sub>	–0.003 (0.005)	–0.005 (0.024)	–0.027 (0.018)	–0.034* (0.016)	–0.012 (0.018)	–0.021 (0.016)
Number of observations	28,447	1908	1908	1428	1908	1428
Number of firms	4503	300	300	255	300	255
R <sup>2</sup>	0.20	0.27	0.39	0.31	0.36	0.24

This table presents coefficient estimates from regressions relating the use of a line of credit to various lagged firm characteristics. Columns 1 and 2 report the estimated marginal effects (or effect of going from 0 to 1 for indicator variables) of lagged firm characteristics on the probability of having a line of credit from maximum likelihood probit estimation using the full and random sample, respectively. Columns 3 through 6 report the coefficient estimates from an OLS estimation using the random sample; the estimation relates two different measures of the bank liquidity to total liquidity ratio to lagged firm characteristics. The estimation reported in columns 4 and 6 isolates the intensive margin of the bank liquidity to total liquidity ratio by focusing only on firms that have a line of credit. Regressions include year and 1-digit industry indicator variables; standard errors are clustered at the firm level.

\*\*\* statistically distinct from 0 at the 1 and 5% level, respectively.

and the mean cash scaled by total assets across the cash-flow distribution. There is strong evidence of a negative correlation of line of credit use and cash holdings across the cash-flow distribution, especially at the low end. Moving from the lowest decile to the 5th decile of cash flow leads to a monotonic increase in the probability of obtaining a line of credit from 40% to over 90%. Alternatively, cash balances decline from over 55% of total assets to less than 10% of total assets.

Interestingly, there is an opposite trend at the very high end of the cash-flow distribution in the 9th and 10th decile, where firms on average become less likely to use a line of credit and more likely to hold cash balances. This trend is driven by firms like Microsoft. They are much more likely to be in services industries, much less likely to use debt financing, and



**Figure 1**  
**Use of line of credit versus cash holdings across cash-flow distribution**

The figure maps the mean cash to total assets ratio and the fraction of firms that have a line of credit by lagged cash-flow decile. Going from left to right is going from firms in the lowest cash-flow decile to firms in the highest cash-flow decile.

have much higher market-to-book ratios than firms in the high cash-flow deciles that use lines of credit. The opposite pattern at the high end of the cash-flow distribution is captured by the market-to-book and industry controls in the regression results, which is why they do not influence the positive effect of cash flow on line of credit utilization.

Figure 1 suggests that there is an explicit trade-off across the cash-flow distribution in firms' utilization of bank lines of credit versus cash in corporate liquidity management. Columns 3 through 6 of Table 3 examine the bank liquidity to total liquidity ratio and present coefficient estimates that are consistent with this evidence. Columns 3 and 4 examine total lines of credit scaled by the sum of total lines of credit and cash balances, whereas columns 5 and 6 examine unused lines of credit scaled by the sum of unused lines of credit and cash balances. Consistent with Figure 1, columns 3 through 6 show a positive correlation between lagged cash flow and the bank liquidity to total liquidity ratio. The magnitude of the coefficient in column 3 suggests that moving from the 10th to 90th percentile of the cash-flow distribution leads to a  $(0.7 \times 0.11/0.51 =)$  15% increase in the bank liquidity to total liquidity ratio at the mean. Size also is a strong predictor of the use of lines of credit versus cash by firms. The coefficient estimate on cash-flow volatility is negative and statistically distinct from 0 at the 1% level in column 3, and 5% in column 5. This suggests that firms with high cash-flow volatility rely more on cash in their liquidity management. This evidence is consistent with the notion that firms with high cash-flow volatility may prefer to avoid cash flow-based

financial covenants associated with lines of credit, something I explore further in the next section.

The use of the bank liquidity to total liquidity ratio in Table 3 also helps mitigate concerns that omitted variables correlated with a firm's demand for overall liquidity are influencing the positive correlation between line of credit usage and cash flow. The bank liquidity to total liquidity ratio isolates the relative attractiveness of lines of credit versus cash, while controlling for overall liquidity. Firms with low cash flow may have higher or lower demand for liquidity, but the important result in Table 3 is that they prefer to hold cash balances relative to lines of credit.<sup>7</sup>

One key question is whether the results in columns 3 and 5 are driven uniquely by the extensive margin. In other words, are these results robust when one examines only firms that have a line of credit? Columns 4 and 6 suggest that, conditional on having a line of credit, firms with higher cash flow and larger firms rely more heavily on lines of credit in their liquidity management. The coefficients are weaker, both statistically and in magnitude. However, they suggest that the positive effect of cash flow on the use of lines of credit is robust when isolating only the intensive margin of use.

There is an economically meaningful and statistically robust correlation between a firm's lagged cash flow and its use of a line of credit. Table 4 presents results that help to explain this correlation. In particular, I examine how the effect of cash flow on line of credit utilization varies across firms with high and low probabilities of financial distress. I split firms into high and low distress likelihood groups on the basis of whether they are below or above the sample median *z-score*.<sup>8</sup> I then replicate specifications 1, 2, and 5 from Table 3 with the inclusion of an indicator variable for *low distress likelihood* and an interaction term of this variable with cash flow. In all three specifications, there is a positive correlation between having low distress likelihood and the utilization of a line of credit. In addition, the effect of cash flow on utilization of a line of credit is statistically significantly more negative for firms with low distress likelihood. In other words, the positive effect of cash flow on line of credit utilization is unique among firms with high distress likelihoods, as evidenced by the positive

<sup>7</sup> An alternative concern is that the results in Table 3 are consistent with cash balances increasing as cash flow declines, but line of credit balances remaining unchanged. In unreported results, I show that cash flow has a strong positive effect on line of credit balances scaled by non-cash assets. The evidence suggests that results in Table 3 are not solely driven by changes in cash balances.

<sup>8</sup> I follow Mackie-Mason (1990) and calculate Altman's (1968) *z-score* excluding leverage, given that leverage is a direct function of the proportion of used and unused lines of credit. More specifically, *z-score* is calculated as

$$ZSCORE = 3.3 * \frac{EBIT}{totalassets} + 1.0 * \frac{sales}{totalassets} + 1.4 * \frac{retainedearnings}{totalassets} + 1.2 * \frac{workingcapital}{totalassets}$$

where retained earnings is *item 36* and working capital is *item 179* from Compustat. ZSCORE has a mean of 1.21 and a standard deviation of 2.80 in the random sample.

coefficient on noninteracted cash flow. The evidence in Table 4 suggests that firms with higher distress likelihood prefer cash to lines of credit, unless they have high cash flow. In the next section, I explore how cash flow-based financial covenants may help explain this pattern of results.

#### 4. The Importance of Cash Flow-Based Financial Covenants

There is a robust and economically significant positive correlation between a firm's cash flow and its use of lines of credit versus cash. This correlation is particularly strong among firms with high distress likelihoods. In this section, I explore how cash flow-based financial covenants on lines of credit help explain these results. Violations of financial covenants are an important part of understanding why firms may or may not use lines of credit. Overall, of the 255 firms in the random sample that utilize a line of credit, 90 violate a financial covenant. In other words, 35% of firms that utilize a line of credit commit a covenant violation.

Financial covenants on a line of credit require the maintenance of financial ratios. Financial ratios are specified in the initial contract, and

**Table 4**  
Bank lines of credit and financial distress likelihood.

Dependent Variable	Firm has line of credit {0,1}		Unused line/(unused
Regression type	Probit (marginal effects)		line + cash) OLS
Sample	Full	Random	Random
	(1)	(2)	(3)
[EBITDA/(assets – cash)] <sub><i>t</i>–1</sub>	0.116** (0.012)	0.130* (0.062)	0.081* (0.039)
[Low distress likelihood {0,1}] <sub><i>t</i>–1</sub>	0.089** (0.017)	0.165** (0.055)	0.191** (0.039)
[EBITDA/(assets – cash)] <sub><i>t</i>–1</sub> *	–0.294** (0.074)	–0.447* (0.221)	–0.452** (0.148)
[Low distress likelihood {0,1}] <sub><i>t</i>–1</sub>	0.029 (0.019)	0.109 (0.082)	0.064 (0.067)
[Tangible assets/(assets – cash)] <sub><i>t</i>–1</sub>	0.040** (0.003)	0.052** (0.014)	0.056** (0.009)
[Ln(assets – cash)] <sub><i>t</i>–1</sub>	–0.094** (0.016)	–0.064 (0.074)	–0.083 (0.054)
[Net worth, cash adjusted] <sub><i>t</i>–1</sub>	–0.011** (0.001)	–0.019** (0.005)	–0.021** (0.004)
[Market-to-book, cash adjusted] <sub><i>t</i>–1</sub>	0.606** (0.190)	2.410** (0.868)	0.888* (0.393)
[Industry sales volatility] <sub><i>t</i>–1</sub>	0.008 (0.035)	–0.152 (0.143)	–0.225* (0.112)
Number of observations	28,447	1908	1908
Number of firms	4503	300	300
R <sup>2</sup>	0.20	0.27	0.38

This table presents coefficient estimates from regressions relating the use of a line of credit to various lagged firm characteristics, with a particular focus on how the likelihood of financial distress affects the use of lines of credit. *Low distress likelihood* is an indicator variable that is 0 if the firm has an *Altman's z-score* below the sample median and 1 if the firm has an *Altman's z-score* above the sample median. Regressions include year and 1-digit industry indicator variables, and information asymmetry control variables in Table 3; standard errors are clustered at the firm level.

\*\* \* statistically distinct from 0 at the 1 and 5% level, respectively.

**Table 5**  
**Financial covenants, from LPC's *Dealscan*.**

Loans with financial covenant {0,1}	0.716
Conditional on having financial covenant, loans with:	
Coverage covenant {0,1}	0.700
Debt to cash flow covenant {0,1}	0.485
Net worth covenant {0,1}	0.461
Debt to net worth {0,1}	0.130
Current ratio covenant {0,1}	0.114
Leverage ratio covenant {0,1}	0.201
No cash flow-based covenant {0,1}	0.249

This table presents data from LPC's *Dealscan* on the use of covenants among public corporations. The total sample includes 11,758 loan deals by 4011 public firms. *Coverage covenant* includes fixed charge coverage, interest coverage, cash interest coverage, and debt service coverage covenants. *Net worth covenant* includes net worth and tangible net worth covenants.

the borrower is in default of the loan agreement if a ratio is not satisfied. These violations are referred to as "technical defaults," and the lender has the legal right to accelerate the loan in response to the violation. While most covenant violations are renegotiated, the terms of the loan can change significantly.

Table 5 presents evidence from LPC's *Dealscan* on financial covenants. The sample includes 11,758 sole lender and syndicated lines of credit obtained by non-financial public corporations (firms with a ticker) from 1996 to 2003. Almost 72% of all lines of credit in the sample have information on financial covenants. Coverage covenants, which are written on a measure of cash flow divided by an interest, debt service, or fixed charge expense, are the most common type of financial covenant, occurring on 70% of lines that have financial covenant information. The second most common type of covenant is a debt to cash flow covenant, which is on 49% of lines that have financial covenant information, followed by net worth covenants. Current ratio and leverage ratio-based covenants are less common. In fact, only 25% of lines of credit that have financial covenant information *do not* have a cash flow-based covenant. The evidence from Table 5 suggests that maintenance of cash flow is critical to avoiding noncompliance with financial covenants on lines of credit.<sup>9</sup>

Why do firms violate covenants? Unfortunately, the SEC does not require firms to disclose why they are in violation of covenants. Previous studies using data from the 1980s examine SEC filings and find evidence that net worth and current ratio covenant violations are the most common (Beneish and Press 1993, Chen and Wei 1993). However, both these

<sup>9</sup> In Table 5, I report data on covenants from *Dealscan* because the SEC does not require firms to report their covenants. The SEC does require firms to report whether they are in violation of a covenant. Dichev and Skinner (2001) argue that actual covenant violations represent situations in which firms were unable to obtain an amendment to avoid violation. The violations tracked in my data represent violations that could not be avoided and therefore represent more serious violations than those that can be avoided.

**Table 6**  
The causes and consequences of financial covenant violations.

	Covenant violation <sub><i>t</i></sub> {0,1}	Total line <sub><i>t</i></sub> / assets <sub><i>t-1</i></sub>	Unused line <sub><i>t</i></sub> / assets <sub><i>t-1</i></sub>	[Total line/ [(Total line + cash)] <sub><i>t</i></sub>	[Unused line/ [(unused line + cash)] <sub><i>t</i></sub>
	(1)	(2)	(3)	(5)	(6)
EBITDA <sub><i>t</i></sub> /assets <sub><i>t-1</i></sub>	-0.577** (0.105)	-0.487** (0.112)			
[Debt/assets] <sub><i>t</i></sub>		0.592* (0.246)			
[Net worth/assets] <sub><i>t</i></sub>		-0.007 (0.242)			
Current ratio <sub><i>t</i></sub>		-0.007 (0.010)			
Market-to-book ratio <sub><i>t</i></sub>		-0.003 (0.008)			
Ln(total assets) <sub><i>t</i></sub>		0.033 (0.029)			
Covenant violation <sub><i>t-1</i></sub> {0,1}			-0.041* (0.018)	-0.036** (0.011)	-0.070** (0.027)
EBITDA <sub><i>t-1</i></sub> / assets <sub><i>t-2</i></sub>			0.071 (0.046)	0.062 (0.039)	0.073 (0.054)
[Debt/assets] <sub><i>t-1</i></sub>			0.005 (0.164)	-0.012 (0.109)	0.381* (0.193)
[Net worth/ assets] <sub><i>t-1</i></sub>			0.014 (0.141)	0.016 (0.095)	0.257 (0.169)
Current ratio <sub><i>t-1</i></sub>			-0.001 (0.004)	0.002 (0.004)	-0.022** (0.008)
Market-to-book ratio <sub><i>t-1</i></sub>			0.010* (0.005)	0.002 (0.004)	-0.001 (0.005)
Ln(total assets) <sub><i>t-1</i></sub>			-0.116** (0.025)	-0.087** (0.018)	0.008 (0.025)
Number of observations	1428	1387	1174	1174	1174
Number of firms	255	249	246	246	246
R <sup>2</sup>	0.21	0.24	0.61	0.57	0.73

Columns 1 and 2 present regression coefficients from firm fixed effects regressions relating the probability of a covenant violation to firm characteristics. Columns 3 through 6 present regression coefficients from firm fixed effects regressions relating line of credit balances to a covenant violation in the previous year. The sample used for columns 1 and 2 includes only firms that have a line of credit. The sample used for columns 3 through 6 includes only firms that have a line of credit in the previous year. Regressions include year indicator variables, and standard errors are clustered at the firm level.

\*\*, \* statistically distinct from 0 at the 1 and 5% level, respectively.

studies report that large numbers of violators do not report why the violation occurs (almost 25% in Beneish and Press 1993 and almost 50% in Chen and Wei 1993). Consistent with these earlier studies, a large fraction of firms in the random sample do not report exactly which covenant is violated. Selective reporting by firms may lead to systematic bias if, for example, firms are less likely to disclose information when cash flow-based covenants are violated as opposed to current ratio-based covenant violations. Instead of relying on self-reporting, I conduct a statistical analysis to determine what financial factors have the strongest power in predicting a covenant violation. More specifically, Table 6 reports coefficient estimates from the following firm fixed effects linear specification.

$$Violation_{it} = \alpha_i + \alpha_t + \beta X_{it} + \varepsilon_{it} \quad (1)$$

In this specification,  $X_{it}$  is a matrix that includes measures on which banks write covenants: cash flow, the leverage ratio, net worth, and the current ratio. Given the influence of size on the utilization of lines of credit, I also include firm size in  $X$ . The vector of coefficient estimates of  $\beta$  describes whether reductions in cash flow, reductions in net worth, increases in leverage, or decreases in current ratios lead to credit line covenant violations. The sample for the estimation of Equation (2) includes only firm-years where a line of credit is present, and standard errors are clustered at the firm level.<sup>10</sup>

Column 1 of Table 6 shows that a drop in cash flow leads to a large increase in the probability of violating a covenant. The firm fixed effects coefficient estimate has a  $t$ -statistic above 5, where standard errors are clustered at the firm and there are 255 firms in the sample. The magnitude suggests that a firm that moves from the 90th to the 10th percentile of cash flow scaled by lagged assets has an increase in the probability of default of  $(0.32 \times 0.58 =) 19\%$ . Considering that the mean of the left-hand-side variable is 11% in this sample, this represents an increase in the probability of default of 160% at the mean. Drops in cash flow strongly predict financial covenant violations.

In column 2, I include other measures on which banks place covenants. The only other variable that affects the probability of a covenant violation is the leverage ratio. The magnitude of the coefficient suggests that going from the 10th to the 90th percentile in leverage has an increase in the probability of default of  $(0.47 \times 0.59 =) 27\%$ , which is stronger than the effect of cash flow. However, the effect of leverage is not as statistically reliable, given that the effect is only significant at the 5% level with a  $t$ -statistic of 2.41. The  $t$ -statistic for the negative coefficient on cash flow in column 2 is  $-4.35$ . Other measures on which covenants are placed, including net worth, the current ratio, the market-to-book ratio, and total assets, have no statistically significant effect on the probability of a covenant violation when included with cash flow and the leverage ratio. This evidence is consistent with data from *Dealscan* in Table 5: cash flow and leverage ratios are the most common component of financial covenants, and they are also the most powerful predictors of covenant violations.

What happens to the availability of the line of credit when a firm defaults on its covenants? Figure 2 focuses on the 90 firms in the sample that violate a financial covenant during the sample period. It maps the total, used, and

<sup>10</sup> I also estimate Equation (2) using fixed effects in a maximum likelihood logit specification; these results (unreported) are qualitatively similar to the linear regression results.

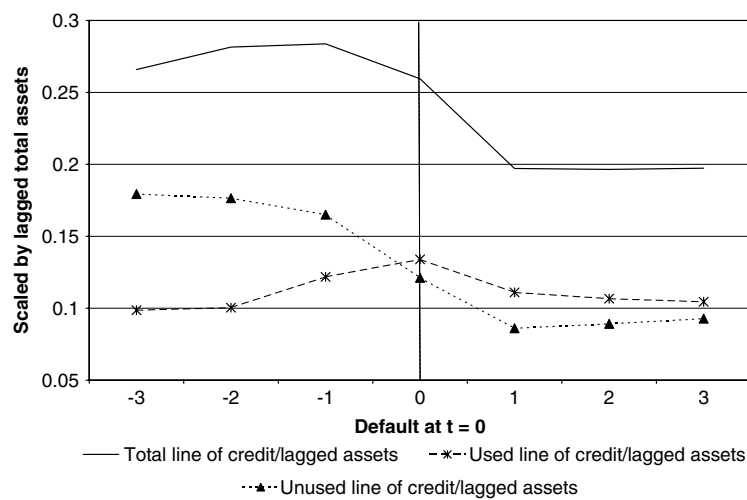


unused lines of credit to lagged total assets ratios relative to the default year. The data are “default-time scaled” so that  $t = 0$  is the year that the firm initially defaults on a financial covenant in its credit agreement. Figure 2 shows that the used portion of the line of credit increases directly before the firm defaults on a covenant. When a firm defaults, there is a subsequent reduction in the availability of the total, used, and unused line of credit at  $t + 1$ . The unconditional means suggest that the bank reduces the availability of the total line of credit from 0.259 to 0.197, or about 25%, in the year after a covenant violation. From the year before the violation to the year after the violation, the unused portion of the line of credit goes from 0.165 to 0.086, a reduction of almost 50%. The results in Figure 2 suggest that the availability under a line of credit is conditional on the maintenance of covenants.

Columns 3 through 6 of Table 6 present estimates showing the effect of a covenant violation on the available balances of lines of credit. More specifically, I estimate the fixed effects equation:

$$Line_{it} = \alpha_i + \alpha_t + \beta X_{i,t-1} + \gamma * Violation_{i,t-1} + \varepsilon_{it}. \quad (2)$$

The sample includes only those firm-year observations in which a line of credit was present at  $t - 1$ . Standard errors are clustered at the firm level. The results in columns 3 and 4 demonstrate that a default at  $t - 1$  leads to a reduction in the total and unused line of credit scaled by lagged assets of 0.041 and 0.036, which is 18 and 26% at the mean of the



**Figure 2**  
The effect of a covenant violation on availability of line of credit

The figure maps the availability of lines of credit around a covenant violation. Time  $t = 0$  is the year in which the covenant violation takes place, and the y-axis represents the availability of the line in years before and after the covenant violation.

left-hand-side variable, respectively. The coefficient estimates in columns 5 and 6 demonstrate a comparable reduction in the bank liquidity to total liquidity ratios of between 10 and 20%. It is important to note that the regression specification controls for all variables that are shown in columns 1 and 2 of Table 6 to predict covenant violations. In other words, even controlling for the financial factors leading to covenant violations, a covenant violation has an independently large and statistically significant effect on the availability of lines of credit.

These findings help explain why firms with low cash flow are less likely to use lines of credit, and why firms with low cash flow and high cash-flow volatility rely more heavily on cash in corporate liquidity management. Firms with low cash flow are more likely to violate financial covenants, and violations of covenants lead to a restriction in the availability of the line of credit.

These results also empirically quantify the degree to which lines of credit represent unconditional liquidity available to the firm in all future states. Lines of credit are not totally unconditional obligations of banks; banks use covenant violations to restrict the availability of the line of credit. The evidence is consistent with Rajan and Winton (1995) and Park (2000), who argue that covenants facilitate bank monitoring. Banks appear to use covenant violations to reassess their lending position with a borrower.

## **5. Access to a Line of Credit and Financial Constraints**

In this section, I examine whether access to a line of credit is a more statistically powerful measure of financial constraints than measures previously used in the literature. In particular, I relate my findings with the empirical findings of Almeida et al. (2004). Their important theoretical insight, described above in Section 2, is that firms facing capital market frictions are more likely to save cash out of cash flow. They empirically examine their model by sorting a firm into financially constrained and unconstrained categories based on four measures: the payout ratio of the firm, the size of the firm, whether the firm has a bond rating, and whether it has a commercial paper rating.<sup>11</sup> In this section, I use access to lines of credit as a measure of financial constraints: a firm is considered unconstrained if it (i) has access to a line of credit in every year of the sample, and (ii) maintains cash flow above the median in every year of the sample. The latter part of the definition reflects the fact that firms with low cash flow are at risk of violating a covenant and losing access to their line

<sup>11</sup> For details, see Almeida et al. (2004), pp. 1789–90. The only discrepancy in my categorization is to split firms into size groups based on whether they are above or below the median in every year. Almeida et al. (2004) split firms into size groups on the basis of whether they are in the top 3 deciles or bottom 3 deciles. I change the definition because there are almost no “unconstrained” firms by the line of credit measure in the smallest 3 size deciles.

**Table 7**  
Correlation with other measures of financial constraints.

	Line of credit	Payout decile	Size decile	Bond rating
Payout ratio decile	0.09			
Size decile	0.19	0.44		
Bond rating	0.18	0.29	0.68	
CP rating	0.26	0.32	0.48	0.52

This table presents correlations between various measures of financial constraints used in the literature. *Line of credit* takes on the value 1 if the firm (i) has access to a line of credit in every year of the sample, and (ii) maintains cash flows above the median firm in every year of the sample. *Bond rating* and *commercial paper rating* take on the value 1 if the firm ever has an S&P corporate credit rating and commercial paper rating through the sample, respectively.

Note: All correlations are statistically distinct from 0 at the 1 percent level.

of credit; therefore, they are likely to face capital market frictions and have a reason to save cash out of cash flow. A firm that does not satisfy either part of the definition is considered constrained. Given that theoretical research emphasizes the importance of bank lines of credit in reducing potential financial constraints and providing liquidity, this categorization is a natural extension of Almeida et al. (2004).

Table 7 presents the unconditional correlations between the measures of financial constraints used in Almeida et al. (2004) and the measure based on access to lines of credit. As the first column demonstrates, the measures are positively correlated. In terms of magnitudes, the credit line access measure is most correlated with the commercial paper rating measure of financial constraints. While there is a positive correlation between having access to a line of credit and other measures of financial constraints, it is smaller than the correlations of the other measures with each other. This suggests that access to a line of credit may provide additional statistical power beyond the traditional measures used in the literature.

Table 8 examines the cash-flow sensitivity of cash for various subsamples based on measures of financial constraints used in Almeida et al. (2004). More specifically, the coefficient estimates presented in Table 8 are the outcome of firm fixed effects regressions relating the difference in cash holdings from  $t - 1$  to  $t$  on cash flow, a measure of investment opportunities ( $Q$ ) and the natural logarithm of total assets, all measured at time  $t$ . The estimations replicate the estimations that generate results reported in Table 3 of Almeida et al. (2004). Each coefficient estimate in Table 8 represents the effect of cash flow on cash holdings from a separate regression for a different subsample.

In Panel A, I focus on the line of credit measure. Column 1 shows that firms that are constrained according to the line of credit measure save cash out of cash flow, whereas firms that are unconstrained do not save cash out of cash flow; the difference in the coefficient estimates is statistically distinct from 0 at the 1% level. The evidence is consistent with the intuition in Almeida et al. (2004): firms with access to lines of credit are less likely to

**Table 8**  
**Availability of bank lines of credit and the cash-flow sensitivity of cash.**

Dependent variable Δ Cash Holdings		Panel A		
		(1)	(2)	(3)
		Splitting constrained firms by:		
			Low cash flow, no line of credit	Low cash flow, line of credit
				High cash flow, no line of credit
1. Lines of credit				
Unconstrained	−0.062 (0.045)			
Constrained	0.084 <sup>**</sup> ,+ (0.010)	0.085 <sup>**</sup> (0.010)	0.063 <sup>**</sup> (0.016)	0.028 (0.097)
		Panel B		
		(1)	(2)	(3)
		Splitting constrained firms by:		
			Constrained, by line of credit measure	Unconstrained, by line of credit measure
2. Payout ratio				
Highest 3 deciles	0.110 <sup>**</sup> (0.025)			
Lowest 3 deciles	0.089 <sup>**</sup> (0.014)		0.090 <sup>**</sup> (0.014)	−0.077 (0.121)
3. Firm size (assets)				
Largest 5 deciles	−0.023 (0.027)			
Smallest 5 deciles	0.096 <sup>**</sup> ,+ (0.012)		0.096 <sup>**</sup> (0.012)	0.014 (0.083)
4. Bond rating				
Has a rating	0.016 (0.027)			
Does not have a rating	0.086 <sup>**</sup> ,+ (0.011)		0.089 <sup>**</sup> (0.011)	−0.042 (0.063)
5. Commercial paper rating				
Has a rating	−0.047 (0.047)			
Does not have a rating	0.082 <sup>**</sup> ,+ (0.010)		0.085 <sup>**</sup> (0.010)	−0.050 (0.052)

This table presents coefficient estimates from regressions relating the change in cash holdings to cash flow. The estimation follows that of Almeida et al. (2004), which they describe in their Equation (8) and Table 3. Each reported coefficient is the effect of cash flow on cash holdings from a separate regression. Panel A splits the sample on the basis of the line of credit measure of financial constraints in which firms that (i) have access to a line of credit in every year of the sample, and (ii) maintain cash flows above the median firm in every year of the sample are considered unconstrained. Panel B examines the measures of financial constraints used in Almeida et al. (2004). All estimations include year and firm fixed effects. Standard errors are heteroskedasticity-robust, clustered at the firm.

<sup>\*\*</sup>, <sup>\*</sup> distinct from 0 at 1 and 5%, respectively; <sup>+</sup> distinct from unconstrained sample at 10% or better.

face a high external cost of capital and therefore save no cash out of cash flow. On the other hand, firms without access to a line of credit are likely to face a high external cost of capital, and therefore save cash out of cash flow.

Firms that are constrained according to the line of credit measure can be constrained for one of three reasons: (i) they do not have a line of credit in

every year in which they are in the sample and have low cash flow, (ii) they have a line of credit in every year but have low cash flow, (iii) they have high cash flow but they do not have a line of credit in every year. Columns 2, 3, and 4 report the cash-flow estimates after splitting the constrained group into these 3 respective subgroups. Firms that have low cash flow and do not have a line of credit in every year of the sample show the largest cash-flow sensitivity of cash, followed by firms that have a line of credit in every year of the sample but have low cash flow. The estimate in column 3 of Panel A suggests that firms that have access to a line of credit and low cash flows continue to save cash out of cash flow despite having access to a line of credit. This is consistent with evidence presented in the previous section on the importance of cash flow-based covenants: firms for whom the line of credit is likely to be revoked due to a financial covenant violation are the same firms that continue to save cash out of cash flow. Firms that have high cash flow but do not have a line of credit show a positive cash-flow sensitivity of cash, but it is not statistically distinct from 0 at a reasonable confidence level.

In column 1 of Panel B, I replicate the procedure using the Almeida et al. (2004) measures of financial constraints. The ordering of the point estimates is consistent with their findings in every category except the payout ratio categorization. Smaller firms, firms without an S&P corporate credit rating, and firms without an S&P commercial paper rating save cash out of cash flow. At the same time, large firms and firms with credit ratings do not save cash out of cash flow. The magnitudes of the coefficients are almost identical for all financial constraints categorizations except the payout ratio.

In columns 2 and 3 of Panel B, I examine whether the availability of lines of credit adds statistical power to the measures used in Almeida et al. (2004). For each of *their* measures of financial constraints, I split the “constrained” sample further on the basis of whether the firm is constrained or unconstrained by the credit line access measure. As the reported estimates in columns 2 and 3 demonstrate, the positive effect of cash flow on cash holdings among *their* constrained firms is driven exclusively by firms that do not have access to lines of credit. In other words, firms that Almeida et al. (2004) classify as constrained but have access to a line of credit show no statistically significant positive relationship between cash flow and cash holdings. The results suggest that firms that have access to a line of credit are not financially constrained, even if they are classified as constrained by the Almeida et al. (2004) measures. Overall, the results in columns 2 and 3 of Panel B suggest that the measures of financial constraints used in Almeida et al. (2004) are accurate measures of financial constraints on average. However, a measure of the availability of bank lines of credit adds important information that improves the cash–cash flow sensitivity estimates.

## 6. Conclusion

Bank lines of credit, or revolving credit facilities, are an instrumental component of corporate liquidity management. Theoretical research on credit lines argues that this financial instrument should resolve future capital market frictions facing firms. Existing research also suggests that banks can provide liquidity to firms more efficiently than reliance on internal cash. However, empirical findings from the cash literature suggest that cash plays an important liquidity role for certain firms in the economy. While these two areas are related, there has been very little interaction between the two. Extant research does not discuss why some firms utilize lines of credit while others rely on cash for liquidity. This article attempts to bridge the gap between these two areas. In particular, I examine the factors that lead firms to utilize bank lines of credit instead of cash in corporate liquidity management.

The principal finding of the article is that firm cash flow is a strong predictor of whether a firm uses bank lines of credit or cash in corporate liquidity management. Firms with low cash flow or high cash-flow volatility rely more heavily on cash. The positive correlation between lagged cash flow and the use of lines of credit is robust to both the extensive margin of use (whether a firm obtains a line of credit) and the intensive margin (conditional on having a line of credit, how large a fraction the credit line is of firm liquidity). Finally, the positive correlation exists only among firms with a high probability of financial distress; in other words, if a firm has high distress likelihood, then high cash flow is critical to obtaining a line of credit.

I document that an important channel for this correlation is banks' use of cash flow-based financial covenants on the lines of credit they supply. Cash flow-based covenants are more common than any other type of covenant. Decreases in cash flow are a strong predictor of covenant violations, and a firm that violates a covenant loses access to 15 to 30% of its line of credit capacity. This finding suggests that lines of credit are a poor liquidity substitute for cash for firms that have existing or expected low cash flow. In turn, these firms rely on cash for liquidity.

Theoretical research suggests that lines of credit help alleviate capital market imperfections facing firms. Using this insight, I propose an alternative measure of financial constraints based on access to a line of credit. I provide evidence that this measure is a more statistically powerful measure of financial constraints than traditional measures used in the literature. More specifically, I show that the cash-flow sensitivity of cash is positive *only* among "constrained" firms using traditional measures such as size, payout ratio, or access to public debt market *if* they do not have access to a line of credit. Access to a line of credit is a theoretically justified and statistically powerful measure of whether a firm faces capital market imperfections.

This article represents one of the first empirical examinations of the use of bank lines of credit among a large sample of public firms. There

are several avenues for future research, two of which I outline here. First, researchers could explore the time series of the data to examine how corporate liquidity varies through business cycles and how it is correlated with alternative measures of liquidity. There are interesting macroeconomic questions regarding how lines of credit may affect credit crunches and the transmission of monetary policy (Holmstrom and Tirole 1998, Morgan 1994). Second, lines of credit may play an instrumental role in firm investment policy. Firms with lines of credit may face lower investment costs and a lower cost of external finance. Using properly identified shocks to the supply of available financing, researchers may be able to examine how credit lines affect investment policy when firms face high costs of raising new external finance.

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