Who Seeks and Who Receives? Implications of Demand for and Access to Financial Capital by Young Firms

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Abstract:

In the aftermath of the latest financial crisis, policy-makers at all levels are concerned about the impact of the crisis on access to financial resources by young firms, particularly as major changes occur in bank-lending practices and uncertainties surround the implementation of financial reform legislation. In this paper, we analyze the types and sources of financing used in young firms over the years 2007 through 2009. We find differential outcomes for firms who applied for loans and received them, those who applied and were denied, and those which did not apply for fear of denial. We explore the factors that mitigate the decision to apply for a loan and the subsequent outcomes of firm survival and growth. Our work provides insights into the relative importance of supply and demand for financing both prior to and subsequent to the financial shocks. We leverage various measures and perceptions to disentangle the decision to seek bank loans from the likelihood of receiving a loan based on credit scores and other objective measures. We find that both tangible and intangible assets, particularly intellectual property, play a significant role in receipt of bank loans in the firms' early years of operation.

Keywords: Entrepreneurial finance, bank loans, financial crisis, credit scoring, behavioral finance

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1 Introduction

The economics and finance literature provide strong evidence that sufficient starting capital is a binding constraint for new firms. Entry into entrepreneurship increases with sudden increase in personal wealth, e.g. via bequest (Cagetti and De Nardi (2006)) or external change in taxation rate (Nanda (2008)), and with increased access to bank financing through deregulation and loosening of branching restrictions (Black and Strahan (2002)). Likewise, absence of funds inhibits entry. For example, Evans and Jovanovic (1989) find that borrowing capacity limits entrepreneurial entry; using the National Longitudinal Survey they estimate that new entrepreneurs are limited by 1.5 times the size of their initial assets in starting a new business. In this paper, we analyze the use of bank loans in young firms during their early years of existence, and how these are altered in the wake of the financial shocks of 2008. We address two crucial questions: 1) what characteristics of startups—and their founders—are related to seeking and receiving bank credit and 2) how did the financial crisis alter or amplify this dynamic.

This paper exploits rich information regarding the types and sources of financing used in new firms over the first six years of operations in the confidential Kauffman Firm Survey (KFS) microdata, a longitudinal panel study of 4,928 businesses that started in 2004. The baseline survey of new businesses has been followed up with fix subsequent annual surveys in an ongoing effort to track the new business trajectories (Ballou, Barton, Desroches, Potter, Reedy, Robb, Shane and Zhao (2008); Reedy and Robb (2009)). Importantly, the most recent survey spanned the financial shocks of 2008 and 2009, which occurred in the fifth and sixth year of operations for the firms in this survey. We probe both material and behavioral drivers of the entrepreneur's decision to apply for bank credit. We then account for the selection bias in the first stage decision to apply in our analysis of ultimate receipt of bank credit. The role of tangible and intangible assets and financial performance prior to the crisis are explored.

Moreover, we explore the particular relevance of these questions to new high-technology firms. In previous work using the KFS data, Winston Smith (2010) provided evidence that banks increase lending to high-technology firms as information asymmetry and inherent uncertainty surrounding the firm are lessened. While high-tech firms account for a relatively small percent of the sample, they are disproportionately likely to contribute to economic growth through employment, revenue, assets, and innovations. In this paper, we specifically address the impact on high-tech firms relative to firms in low-tech industries.

The findings in this paper provide important insights for the growing policy debate concerning the constrained credit and lending for small businesses, as well as assessing the relative significance of constrained supply and dampened demand. The paper concludes with a discussion of the factors that contribute to young firm survival and growth, including the relative importance of intangible assets, such as intellectual property, and its role in in mitigating the consequences of the financial crisis and facilitating economic recovery, which have important implications for economic recovery.

2 Theoretical Framework

2.1 Banks and New Firm Finance

In practice, entrepreneurs rely on a mixture of financing options for new companies, often employing both informal and formal sources of capital for early financing, including bank loans and venture capital (Bhide (2000); Parker (2009); Shane (2008). In the aggregate, the financing of small businesses in the United States has been a roughly equal mixture debt and equity (Berger and Udell (1998); Haynes and Brown (2009); Robb and Robinson (2010)).

Banks are the largest source of external debt and insiders the largest source of equity for small firms in the United States (Haynes and Brown (2009); Robb, Reedy, Ballou, DesRoches, Potter and Zhao (2010). Similar patterns hold in the United Kingdom (Parker (2009)).

Chemmanur, He, and Nandy (2009) find that having outside financing prior to IPO increases the likelihood of going public, and that this result holds for both venture capital and bank financing. Banks overall play a substantial role in new firm formation and growth (Ayyagari, Demirguc-Kunt and Maksimovic (2010); Beck, Demirgüç-Kunt and Maksimovic (2008); Kerr and Nanda (2009)). Black and Strahan (2002) show that deregulation of interstate banking and loosening of branching restrictions fostered increased entrepreneurial activity. In a sample of Italian firms, Benfratello, Schiantarelli, and Sembenelli (2008) find that bank loans facilitate innovation. In a sample of French firms, Landier and Thesmar (2009) find that banks provide entrepreneurs with short-term and long-term debt.

Importantly, debt financing places the risk associated with the new firm squarely with the entrepreneur, who must repay regardless of outcome; equity financing spreads the risk between the entrepreneur and the investor but also dilutes the owners' control (Jensen and Meckling (1976). Equity financing smoothes the financial cost of potential failure, while debt financing amplifies it. However, the entrepreneur may be averse to giving up an equity stake and control, preferring instead to obtain non-dilutive debt financing. For example, recent evidence suggests a growing use of external debt by very young firms, with significant differences among firms in high-technology industries compared with other industries. (CITES?)

2.2 Information asymmetry and uncertainty in new firm financing

Significant principal-agent conundrums arise in financing new ventures due to information asymmetry, moral hazard, and adverse selection. The literature on entrepreneurial finance identifies specific features of contracting and allocation of rights as a crucial mechanism for aligning interests under circumstances of information asymmetry and uncertainty. Adverse selection arises when the entrepreneur knows her own ability while the outside investor does not. Amit, Glosten, and Muller (1990) model the relationship between information asymmetry and

outside financing. Their model suggests that under conditions of information asymmetry outside investors are more likely to be presented with lower quality entrepreneurs, as higher quality entrepreneurs will choose to proceed alone. Analyzing contracts between venture capitalists and portfolio companies, Kaplan and Stromberg (2003) find that cash flow rights and control rights are allocated separately, and are made contingent upon observable performance measures. Contingencies are included for financial performance, non-financial performance, and milestones. Allocation of different types of rights and specification of contingencies are used as complements. They also find a strong preference for convertible preferred stock among the contracts studied. In a separate paper, the same authors further associated specific risks with particular contractual terms (Kaplan and Stromberg (2004)). Specifically, they find that venture capitalists are given greater control rights under conditions of greatest information uncertainty between the entrepreneur and the investor. Moral hazard is introduced both when the entrepreneur's effort cannot be monitored and when the investor's commitment cannot be assured. Schmidt (2003) finds that such double moral hazard situations are ameliorated through the use of convertible securities in venture capital contracts. Casamatta (2003) likewise models the use of convertible bonds and preferred equity in aligning interests in VC contracts. She models an entrepreneur who seeks money and advice and a VC who can provide both, compared to "consultants" able to provide advice but not funding. In this circumstance, both entrepreneur and VC must exert effort, which is unobservable, and incentives are aligned through commensurate allocation of cash flow rights. Hellman (2006) models the role of convertible preferred securities in allocating cash flow rights contingent upon exit strategy, i.e. acquisition or IPO.

One key strand within the literature on entrepreneurial finance addresses the preference for debt versus equity financing. In broad terms, the entrepreneur prefers to maintain control

rights but generally faces capital constraints, while the investor seeks the highest return for a given level of risk. Importantly, debt financing and equity financing (bank loans compared to venture capital) have different implications for entrepreneurs (Schmidt (2003); Ueda (2004)). Plausible theoretical arguments can be made in favor of the entrepreneur preferring debt financing in order to secure adequate funds while retaining full control rights (Berger and Udell (1998)). However, economic theory also suggests the entrepreneur will prefer to smooth risk by avoiding the necessity to make fixed payments if the firm does not do well, suggesting preference of equity over debt financing.

Inherent information asymmetry makes it difficult for banks to evaluate young firms, particularly in high-technology industries due to the lack of tangible assets and concurrent reliance on knowledge assets, as well as technical and market uncertainty. Banks face greater liquidity constraints than VC firm, leading to preference for collateral, transparent valuation, and other lower risk sectors (Berger and Udell (1998); Ueda (2004); Winton and Yerramilli (2008)). The information asymmetries associated with new, high tech firms make traditional bank lenders less likely to lend to these firms. Empirical studies support this finding broadly (Cole (2008); Cosh, Cumming and Hughes (2009)). Hellman, Lindsey, and Puri (2008) find that banks use strategic equity investing, i.e. venture capital, to build subsequent banking relationships. Banks use relationship building and the gathering of "soft" information to mitigate information asymmetry.

3 Data description and summary statistics

3.1 Kauffman Firm Survey (KFS)

A new panel microdataset is used to explore the types and sources of financing used in new firms at the start and over time. The Kauffman Firm Survey (KFS) is a longitudinal panel study of 4,928 businesses founded in 2004 and tracked over their early years of operation. The

panel structure of the data covers the first six years of operation of the firm, and thus the results provide critical insights into how firms are financed in their early years of operation, and the relationship to subsequent outcomes. Detailed data are gleaned on the nature of new business formation activity including internal and external sources of financing, firm size and focus, and data related to the characteristics, experience and human capital and of the entrepreneur (Ballou, Barton, Desroches, Potter, Reedy, Robb, Shane and Zhao (2008); Reedy and Robb (2009)). Data were collected in the baseline survey of new businesses started in 2004, and subsequent annual surveys. The businesses in this sample all came into existence in 2004, with business start defined in terms of state unemployment insurance paid, FICA, Schedule C income reported on personal income tax, EIN, or the presence of legal status.

The KFS data are based on a large sample of a wide spectrum of industries, oversampled on high-technology to insure the inclusion of sufficient numbers in the sample. Firms are oversampled with known weights, allowing econometric approaches to account for clustered errors arising from stratified sampling (Wooldridge (2002)). The breadth of industry inclusion and flexibility in measuring high-technology industries enable us to gather insights into the dynamic relationship between financing and innovation in industries where innovation is most likely.

These data offer additional advantages for examining new firm financing and testing the relationship between initial financing choice and the subsequent innovation trajectory of the firm. First, many studies of new firms are innately subject to survivor bias when studies are limited to firms that receive particular forms of financing, such as venture capital or even angel financing, in which not all firms survive long enough to reach that point, or studies of firms that go public, which must by necessity have survived long enough to issue public securities. In the KFS data firms all began operations in 2004, and thus the sample does not suffer from inherent

survivor bias. The inclusion of a range of industries facilitates comparison between and among distinct types of sectors. Multiple levels of sensitivity to industry differences are considered here. The U.S. Bureau of Labor Statistics and the National Science Foundation provide guidelines for categorizing technology-generating and technology-employing industries (Hecker (2005)).

3.2 Summary statistics

Many of the variables of interest, such as loan application outcomes and fear of being turned down are only available beginning in 2007. Thus, we have this information only for those firms that started in 2004 and survived until at least 2007. In the appendix, we compare the sample characteristics of all firms in 2004, broken out by whether or not they survived until at least 2007, in terms of owner characteristics, firm characteristics, and financial capital investments. While there are some differences, we note that the two groups are remarkably similar.

Summary statistics for the sample we analyze in the multivariate models are presented in Table 1, as well as for non-high-tech and high-tech only firms. Key outcome variables are summarized in Panel A. About 12% of firms in the sample applied for a new loan in each of the three years under observation; the percent was higher for young high-tech firms (15-16%). In the full sample, between 60 and 70% of loan applicants were always approved; the approval rate was lower for high-tech firms (52-60%), especially in the first two years. Interestingly, the percentage always approved rose nearly 10 percentage points over the three period for high tech firms, while the percentage declined more than 10 percentage points for non-high tech firms.

The variables describing behavioral characteristics are of particular interest in this study, as these may be important determinants of demand for and access to credit. These variables are summarized in Panel B. The variable *fear of denial* takes a value of 1 when firm owners decided

not to seek a new loan at some point when they needed credit for fear of having their loan application turned down. High-tech firms have a slightly lower average for this variable, but the percentage rose over the three years for both high tech firms and firms overall. While just over half of all firms had high credit scores, the percentage was more than 60% for high tech firms.

Owner and firm characteristics are summarized in Panel C and Panel D, respectively. Comparing high-tech firms with non-high-tech firms, the table shows that high-tech firms have a lower representation of female owners and have more educated owners with greater industry and startup experience. Relative to to other firms, high-tech firms are more often organized as corporations, and are more likely to be located in metropolitan areas rather than rural areas. Not surprisingly, high tech firms also are more likely to have intellectual property (patents, trademarks, or copyrights) and to have a larger share of employees working in R&D.

3.3 Empirical Approach

Our model draws on standard assumptions in the banking literature (Gorton and Winton (2003)). We model the decision to apply for a bank loan in year *t* as a function of growth prospects and degree of credit/liquidity constraint as well as control variables for industry, firm size, and owner characteristics (Chava and Purnanandam, and Edelstein (1975)). We are further interested in the role of information asymmetry in mediating the loan application and approval process. We proxy for information asymmetry in several ways. Particularly for a new firm, having a credit rating inherently reduces the information asymmetry between loan applicant and lender(Gorton and Winton (2003)). We use the actual information contained in the Dun & Bradstreet credit score to a dummy for high creditworthiness and include thise as a predictor of applying for a loan as well as the loan application outcome. The credit score provides significant information to the lender about the creditworthiness of the applicant, thereby reducing the information asymmetry dramatically.

Asset tangibility also plays a role in bank lending decisions. We follow a growing interest in the role of intellectual property in bank lending decisions (Winston Smith (2010)). We include controls for firm and owner characteristics that have been shown to affect likelihood of bank borrowing in the previous literature. Firm characteristics include industry, legal form of ownership, and team ownership. Owner characteristics include race, ethnicity, gender, and age. We also include measures of the owners' human capital, including education, years of prior industry experience, and prior startup experience.

Finally, we are interested in trying to dissect financial and behavioral aspects of the loan process. To this purpose we characterize *loan demand* as a function of observable financial and performance measures (extent of credit constraint primarily, profit/revenue in previous year) and a function of behavioral characteristics (risk, fear of denial).

The equation for who applies for a loan, S, can be expressed as a function of the following characteristics:

$$S = \beta_0 + \beta_j growth + \beta_k constrained + \beta_l behavioral + \beta_x firm + \beta_y entrepreneur + \varepsilon$$
 (1)

The second stage equation for likelihood of the loan being approved, A, can be expressed as a function of the following characteristics:

$$A = \delta_{0} + \delta_{1} growth + \delta_{2} asymmetry + \delta_{3} growth + \delta_{4} entrepreneur + \mu$$
 (2)

Comparing equation 1 and equation 2 above we see that the decision to apply for a loan can be identified through the measures of credit or liquidity constraint and the likelihood of approval can be identified through the degree of information asymmetry between the entrepreneur and the lender.

In our empirical approach we first estimate separate maximum likelihood probit regressions on the probability of applying for a loan and the probability of receiving a loan. We then carry out two-stage analysis taking into account that individuals first decide whether or not

to apply for a loan and then estimate the likelihood of approval, conditional on the decision to apply. Results are described below and summarized in Tables 2-4.

4 Results

4.1 Probit analysis

Table 2 presents the results of a probit model estimating the determinants of the choice to apply for a new loan. The first column presents the results for the three years of pooled data (2007, 2008, and 2009) for high tech firms only, while columns 2 and 3 present results the pooled data for firms from all industries. Column 3 includes interactions of some of the independent variables with high tech status. Across the board, firms that had negative expectations of receiving a loan were more likely to apply for a new loan. There was no difference between high tech firms and firms overall.

The demographic characteristics that had the largest impact of the choice to apply for a new loan are whether the owner had a college degree (for all firms) and previous startup experience (for high tech firms). Previous access to financing through trade credit is also a significant determinant of loan demand. Having a business credit line was also important, while having a business credit card was important for just high tech firms. Interestingly, credit score does not seem to play a role in the likelihood of applying for a new loan. In terms of firm size, overall employment had a positive relationship in loan applications for high tech firms, while the share of R&D employment had a negative relationship. Assets were positively associated with loan applications for firms overall.

In terms of intellectual property, its presence was negatively associated with loan applications, but it was not statistically significant. However, when the presence was interacted with the flag for being in a high tech industry, the coefficient on the interaction term was negative and statistically significant, indicating that high tech firms with intellectual property

were less likely to apply for loans controlling for other factors. There were no differences between years in the likelihood of applying for a new loan, controlling for other factors.

Table 3 presents the results from a similar set of analyses looking at whether the firm's application for loan was denied or approved. Across all specifications we see that fear of being rejected is a strong predictor of being denied a loan. Firms in high tech industries were also more likely to have their loan applications denied. One unexpected resulted is the negative coefficient on startup experience (p<.01). A possible interpretation of this result is that previous startup experience may have resulted in business closure or failure, which is not captured in the survey but is likely known to banks. Logically, having started a business that failed in the past might lead to lower likelihood of new loan approvals. Interestingly, the coefficient on high credit score is not statistically significant alone, but is positive and highly significant when interacted with the dummy variable for high tech industries, suggesting that the information revealed through verified creditworthiness is particularly valuable in the context of informationally opaque firms, which is often inherent in high-tech industries. Having intellectual property was negatively associated with loan approvals, but the effect was no longer statistically significant when allowed to vary by high tech status. The ratio of R&D employment to total employment and the ratio of insider financing to total financing were also negatively associated with loan approvals. That latter finding might indicate firms are tapping internal resources as well as friends and family to meet their financing needs.

Having a business credit card is associated with lower likelihood of approval, while having a business line of credit is positively related to the likelihood of approval. Taken together, these several variables suggest a nuanced relationship between degree of credit constraint and the types of alternative credit available to a new firm.

Two-stage analysis

In the two-stage analysis we predict the likelihood of loans being approved, conditional on selection into applying for a loan. We estimate a maximum likelihood probit model with a Heckman sample selection correction in the first stage. In all models our first stage selection equation includes a subjective measure and an objective measure. Both the behavioral variable, *fear of denial*, and the existing loan ratio, *bankloan_ratio*, are positive and statistically significant in the first stage selection equation, indicating the appropriateness of these choices.

Controlling for this selection bias reveals additional nuances in our results along several dimensions. Our findings on credit score continue to suggest that the information revealed through verified creditworthiness is particularly valuable for firms in high-tech industries. Yet, the coefficients on intellectual property and the interaction of high-tech and intellectual property are now negative and statistically significant. For high tech firms, having business credit cards, trade credit, and/or business credit lines are negatively associated with having their loan application approved. In addition, their reliance on internal funding is also negatively associated with having their loan application approved. Only credit score was positive and statistically significant in the model with just high tech firms, indicating the potential difficulty of overcoming information asymmetries that are frequently present in these informationally opaque firms.

Firms in the high tech industries appeared to fare worse than non-high tech firms, except in the case of high tech firms that had high credit scores. Firms with higher levels of R&D spending, at least in terms of employment, also tended to fare worse during this period. The coefficients on the year dummies for 2008 and 2009 were negative and statistically significant in column 3, indicating that tightened credit markets continued to persist.

We also see evidence for the contraction of credit markets in the overall financing patterns of these young firms. In Appendix 2, we can see that the ratio of formal debt financing as a share of financial injections is rising each year to a peak of 70% in 2007. This drops dramatically in 2008, as owners end up contributing more of their own funds into the business. The overall level of financial injections also shows a dramatic drop in 2008, but it appears formal debt financing and levels of new financial injections rebound in 2009.

5 Discussion and Policy Implications

From these results we see that a nuanced relationship exists between the young firms' access to financial resources and key characteristics of the entrepreneur, the firm, and capital constraints within the context of a turbulent economy. This work has important implications for policy and policymakers at all levels. In particular, given the role of young firms and entrepreneurs in job creation and economic growth, policymakers need to consider ensure that entrepreneurs and credit worthy firms are able to secure adequate financial resources for growth and success. In addition, this work provides policy makers additional insight into young technology-based firms, which are important contributors to the U.S. economy. Securing funding for new technology-based firms is particularly problematic, however, since many such firms are built upon intellectual capital rather than on physical assets, so it is difficult to determine the value and prospects of the firm. Ensuring these firms have adequate access to financial capital will enable them to continue to drive innovation, growth, and job creation in the U.S. economy.

2007 12.0% 70.7% 0.48 0.02	2007, All 2008 12.5% 66.4% 0.55 0.05	2009 2009 12.1% 60.9% 0.57 0.02		15.1% 58.4% 0.65	2009 15.8% 60.5%	Noi 2007 11.8% 72.3%	n-High Te 2008 12.4%	2009 11.8%
12.0% 70.7% 0.48 0.02	2008 12.5% 66.4% 0.55	12.1% 60.9% 0.57	2007 15.7% 51.6% 0.46	2008 15.1% 58.4%	2009 15.8%	2007 11.8%	2008 12.4%	2009
12.0% 70.7% 0.48 0.02	12.5% 66.4% 0.55	12.1% 60.9% 0.57	15.7% 51.6% 0.46	15.1% 58.4%	15.8%	11.8%	12.4%	
70.7% 0.48 0.02 15.6%	66.4% 0.55	60.9% 0.57	51.6% 0.46	58.4%				11.8%
0.48 0.02 15.6%	0.55	0.57	0.46		60.5%	72.3%	67 10/	
0.02				0.65			67.1%	60.9%
15.6%	0.05	0.02	-0.26		0.70	0.48	0.54	0.56
			+	0.04	0.01	0.04	0.05	0.02
				_	-			
	18.5%	21.0%	13.6%	17.9%	18.2%	15.7%	18.6%	21.2%
52.8%	53.0%	53.0%	61.8%	62.2%	62.2%	52.3%	52.4%	52.4%
8.7%	8.7%	8.7%	9.5%	9.5%	9.5%	8.7%	8.7%	8.7%
30.6%	30.6%	30.6%	15.3%	15.3%	15.3%	31.6%	31.6%	31.6%
42.5	42.5	42.5	42.7	42.7	42.7	42.4	42.4	42.4
44.9	44.9	44.9	44.7	44.7	44.7	44.9	44.9	44.9
48.6%	48.6%	48.6%	72.2%	72.2%	72.2%	47.2%	47.2%	47.2%
43.8%	43.8%	43.8%	60.1%	60.1%	60.1%	42.8%	42.8%	42.8%
42.4%	42.4%	42.4%	47.3%	47.3%	47.3%	42.1%	42.1%	42.1%
61.6%	61.6%	61.6%	70.6%	70.6%	70.6%	61.1%	61.1%	61.1%
51.0%	51.0%	51.0%	50.9%	50.9%	50.9%	51.0%	51.0%	51.0%
49.3%	49.3%	49.3%	53.4%	53.4%	53.4%	49.1%	49.1%	49.1%
5.6%	5.6%	5.6%	100.0%	100.0%	100.0%	0.0%	0.0%	0.0%
52.3%	57.9%	60.7%	52.6%	57.9%	59.8%	52.3%	57.9%	60.7%
21.4%	18.7%	16.9%	28.7%	27.6%	25.8%	21.0%	18.2%	1 6.4%
16.5%	16.5%	16.5%	9.7%	9.7%	9.7%	16.8%	16.8%	16.8%
34.5%	35.1%	34.9%	27.2%	25.3%	23.9%	34.9%	35.8%	35.7%
11.1%	9.1%	8.2%	13.4%	14.3%	14.2%	11.0%	8.8%	7.9%
54.1%	57.3%	57.4%	48.8%	52.4%	49.5%	54.4%	57.6%	57.9%
0.841	0.824	0.797	0.938	1.052	0.947	0.834	0.809	0.787
0.115	0.105	0.099	0.109	0.108	0.117	0.115	0.104	0.098
6.4%	5.8%	6.5%	4.8%	3.2%	7.1%	6.5%	5.9%	6.4%
14.8%	1 5.0%	13.2%	11 .4%	12.7%	14.2%	14.9%	15.1%	13.1%
6.2%	7.0%	6.0%	8.6%	9.0%	8.4%	6.1%	6.9%	5.9%
14.6%	11.6%	10.7%	23.9%	22.6%	22.8%	14.1%	10.9%	10.0%
	4122			540			3582	
	8.7% 30.6% 42.5 44.9 48.6% 43.8% 42.4% 61.6% 51.0% 49.3% 5.6% 52.3% 21.4% 16.5% 34.5% 11.1% 6.41% 0.115 6.4% 14.8% 6.2% 14.6%	8.7% 8.7% 30.6% 42.5 42.5 44.9 44.9 48.6% 43.8% 42.4% 42.4% 61.6% 51.0% 51.0% 49.3% 5.6% 52.3% 57.9% 21.4% 18.7% 16.5% 34.5% 35.1% 11.1% 9.1% 54.1% 57.3% 0.841 0.824 0.115 0.105 6.4% 5.8% 14.8% 15.0% 6.2% 7.0% 14.6% 11.6%	8.7% 8.7% 8.7% 30.6% 30.6% 30.6% 42.5 42.5 42.5 44.9 44.9 44.9 48.6% 48.6% 43.8% 42.4% 42.4% 42.4% 61.6% 61.6% 61.6% 51.0% 51.0% 51.0% 49.3% 49.3% 49.3% 5.6% 5.6% 5.6% 52.3% 57.9% 60.7% 21.4% 18.7% 16.9% 16.5% 16.5% 16.5% 34.5% 35.1% 34.9% 11.1% 9.1% 8.2% 54.1% 57.3% 57.4% 0.841 0.824 0.797 0.115 0.105 0.099 6.4% 5.8% 6.5% 14.8% 15.0% 13.2% 6.2% 7.0% 6.0% 14.6% 11.6% 10.7%	8.7% 8.7% 8.7% 9.5% 30.6% 30.6% 30.6% 15.3% 42.5 42.5 42.5 42.7 44.9 44.9 44.9 44.7 48.6% 48.6% 48.6% 72.2% 43.8% 43.8% 43.8% 60.1% 42.4% 42.4% 42.4% 47.3% 61.6% 61.6% 61.6% 70.6% 51.0% 51.0% 50.9% 49.3% 49.3% 49.3% 53.4% 5.6% 5.6% 5.6% 100.0% 52.3% 57.9% 60.7% 52.6% 21.4% 18.7% 16.9% 28.7% 16.5% 16.5% 16.5% 9.7% 34.5% 35.1% 34.9% 27.2% 11.1% 9.1% 8.2% 13.4% 54.1% 57.3% 57.4% 48.8% 0.841 0.824 0.797 0.938 0.115 0.105 0.099 0.109 6.4% 5.8% 6.5% 4.8% 14.8% 15.0% 13.2% 11.4% 6.2% 7.0% 6.0% 8.6% 14.6% 11.6% 10.7% 23.9%	8.7% 8.7% 8.7% 9.5% 9.5% 30.6% 30.6% 30.6% 15.3% 15.3% 42.5 42.5 42.7 42.7 44.9 44.9 44.9 44.7 44.7 43.8% 43.8% 43.8% 60.1% 60.1% 42.4% 42.4% 42.4% 47.3% 47.3% 47.3% 61.6% 51.0% 51.0% 51.0% 50.9% 50.9% 49.3% 49.3% 49.3% 53.4% 53.4% 5.6% 5.6% 5.6% 100.0% 100.0% 52.3% 57.9% 60.7% 52.6% 57.9% 16.5% 16.5% 16.5% 9.7% 9.7% 34.5% 35.1% 34.9% 27.2% 25.3% 11.1% 9.1% 8.2% 13.4% 14.3% 54.1% 57.3% 57.4% 48.8% 52.4% 0.841 0.824 0.797 0.938 1.052 0.115 0.105 0.099 0.109 0.108 6.4% 5.8% 6.5% 4.8% 3.2% 14.6% 11.6% 10.7% 23.9% 22.6% 4122 540	8.7% 8.7% 8.7% 9.5% 9.5% 9.5% 30.6% 30.6% 30.6% 15.3% 15.3% 15.3% 15.3% 42.5 42.5 42.5 42.7 42.7 42.7 44.7 44.9 44.9 44.9 44.7 44.7 44.7 43.8% 43.8% 43.8% 60.1% 60.1% 60.1% 60.1% 42.4% 42.4% 42.4% 47.3% 47.3% 47.3% 47.3% 61.6% 51.0% 51.0% 51.0% 50.9% 50.9% 50.9% 49.3% 49.3% 49.3% 53.4% 53.4% 53.4% 55.6% 5.6% 5.6% 100.0% 100.0% 100.0% 52.3% 57.9% 60.7% 52.6% 57.9% 59.8% 21.4% 18.7% 16.9% 28.7% 27.6% 25.8% 16.5% 16.5% 16.5% 16.5% 9.7% 9.7% 9.7% 34.5% 35.1% 34.9% 27.2% 25.3% 23.9% 11.1% 9.1% 8.2% 13.4% 14.3% 14.2% 54.1% 57.3% 57.4% 48.8% 52.4% 49.5% 0.841 0.824 0.797 0.938 1.052 0.947 0.115 0.105 0.099 0.109 0.108 0.117 6.4% 5.8% 6.5% 4.8% 3.2% 7.1% 14.8% 15.0% 13.2% 11.4% 12.7% 14.2% 6.2% 7.0% 6.0% 8.6% 9.0% 8.4% 14.6% 11.6% 10.7% 23.9% 22.6% 22.8%	8.7% 8.7% 9.5% 9.5% 9.5% 8.7% 30.6% 30.6% 30.6% 15.3% 15.3% 15.3% 31.6% 42.5 42.5 42.7 42.7 42.7 42.4 44.9 44.9 44.7 44.7 44.7 44.9 48.6% 48.6% 72.2% 72.2% 72.2% 47.2% 43.8% 43.8% 60.1% 60.1% 60.1% 42.8% 42.4% 42.4% 47.3% 47.3% 47.3% 42.1% 61.6% 61.6% 61.6% 70.6% 70.6% 70.6% 61.1% 51.0% 51.0% 50.9% 50.9% 50.9% 51.0% 49.3% 49.3% 53.4% 53.4% 49.1% 5.6% 5.6% 50.9% 50.9% 50.9% 51.0% 49.3% 49.3% 53.4% 53.4% 53.4% 49.1% 5.6% 5.6% 100.0% 100.0% 100.0% 0.0% 52.3% 57.9% 60.7% 52.6% 57.9% 59.8% 52.	8.7% 8.7% 9.5% 9.5% 9.5% 8.7% 8.7% 30.6% 30.6% 15.3% 15.3% 15.3% 31.6% 31.6% 42.5 42.5 42.7 42.7 42.7 42.4 42.4 44.9 44.9 44.7 44.7 44.7 44.9 44.9 48.6% 48.6% 72.2% 72.2% 72.2% 47.2% 47.2% 43.8% 43.8% 60.1% 60.1% 60.1% 42.8% 42.8% 42.4% 42.4% 47.3% 47.3% 47.3% 42.1% 42.1% 61.6% 61.6% 70.6% 70.6% 70.6% 61.1% 61.1% 61.1% 51.0% 51.0% 51.0% 50.9% 50.9% 50.9% 51.0% 51.0% 49.3% 49.3% 49.3% 53.4% 53.4% 53.4% 49.1% 49.1% 5.6% 5.6% 100.0% 100.0% 100.0% 0.0% 0.0% 52.3% 57.9% 60.7% 52.6% 57.9% 59.8% 52.3% 5

Table 2. New Loan Applications

This table reports coefficient estimates from multinomial probit regressions. The dependent variable in all columns is a dummy variable equal to 1 if the entrepreneur applied for a new loan in the specified calendar year and 0 otherwise. Column 1 includes 2007, 2008, and 2009 for high tech firms only, while Columns 2 and 3 are firms from all industries. Column 3 includes interactions with *high-technology industry*. All columns include entrepreneur, firm, and industry controls. Standard errors (in parentheses) are survey-weighted and heteroskedasticity robust. Significance is denoted as: p < 0.01, p < 0.05, p < 0.10

		High Tech						
VARIABLES		Firms Only		All Firms		All Firms		
					-			
Fear of Denial	_	0.471**	_	0.288***	_	0.281***		
	•	(0.218)	-	(0.0795)		(0.0837)		
High Tech Industry				0.173		0.0406		
				(0.118)		(0.221)		
High TechX Fear of Denial						0.302		
						(0.232)		
Intellectual Property		-0.139		-0.0563		-0.0187		
		(0.176)		(0.0812)		(0.0868)		
High Tech X Intellectual Property						-0.403*		
					•	(0.214)		
College Degree or Higher		0.144		0.179**		0.191***		
		(0.215)		(0.0730)		(0.0732)		
Previous Startup Experience		0.533***		0.0644		0.0565		
		(0.184)		(0.0687)		(0.0691)		
High Credit Score		0.286		0.0253		0.00529		
		(0.194)	-	(0.0699)		(0.0731)		
High Tech X High Credit Score						0.290		
						(0.217)		
R&D Employment		-0.557**	•	0.0479		0.0450		
		(0.266)		(0.0959)		(0.0969)		
Trade Credit		0.588***		0.414***		0.410***		
		(0.182)		(0.0732)		(0.0737)		
Log of Employment		0.296**		0.0632	F.	0.0671		
5	•	(0.123)		(0.0410)		(0.0412)		
Log of Assets		-0.263		0.243**		0.216**		
G		(0.293)		(0.0967)	P*	(0.0978)		
Has Business Credit Card		0.563***		0.0838	F .	0.0872		
	F	(0.189)	•	(0.0699)	F.	(0.0701)		
Has Business Credit Line	F	0.299		0.776***		0.775***		
		(0.211)	F	(0.0790)		(0.0794)		
Year Dummy: 2008	•	0.113		-0.0489	E .	-0.0585		
	•	(0.207)	•	(0.0800)		(0.0806)		
Year Dummy: 2009		0.104	•	-0.0141		-0.0111		
rear Barring. 2000	F	(0.203)		(0.0797)		(0.0800)		
Constant		-4.455**		-1.459***		-1.396***		
Conocant	•	(1.737)		(0.494)		(0.498)		
Owner Demographic Controls		Υ		Y		Y		
Firm Char. Controls		Ϋ́		Ϋ́		Ý		
Industry Fixed Effects		Ϋ́		Ý		Ϋ́		
Observations	r	469	*	3280	r	3259		

Table 3. Loan Application Always Approved

This table reports coefficient estimates from multinomial probit regressions. The dependent variable in all columns is a dummy variable equal to 1 if the entrepreneur's loan applications were always approved and 0 if any applications were denied. Column 1 includes high-tech firms only for 2007, 2008, and 2009. Columns 2 and 3 are for all firms. Column 3 includes interactions with high-technology industry. All columns include entrepreneur, firm, and industry controls. Standard errors (in parentheses) are survey-weighted and heteroskedasticity robust. Significance is denoted as p < 0.01, p < 0.05, p < 0.10

High Tech Firms									
VARIABLES Only All Firms All Firms									
Fear of Denial		-1.165***		-2.170***		-2.221***			
	•	(0.386)	•	(0.187)	•	(0.209)			
High Tech		,		-0.259		-1.225**			
				(0.272)		(0.490)			
High Tech X Fear of Denial						0.781*			
					•	(0.456)			
Previous Startup Experience	•	0.491		-0.351**		-0.395**			
		(0.451)	•	(0.163)		(0.166)			
High Credit Score		0.963**	•	0.0634		-0.0404			
		(0.393)	•	(0.177)		(0.195)			
High Tech X High Credit Score						1.348***			
						(0.484)			
Intellectual Property		-0.254		-0.320*		-0.328			
		(0.432)		(0.182)		(0.201)			
High Tech * Intellectual Property					-	-0.479			
						(0.429)			
R&D Employment		-0.0677		-0.406**		-0.409*			
		(0.579)	•	(0.203)		(0.210)			
Trade Credit		-0.730		0.126	F	0.109			
		(0.451)	•	(0.170)		(0.171)			
Log of Employment		0.0924	•	-0.0112		0.000337			
		(0.206)		(0.101)		(0.104)			
Log of Assets		-0.913*	•	-0.0923		-0.0849			
		(0.500)		(0.144)		(0.147)			
Inside Financing/Total Financial Capital		-3.960***		-0.905***		-0.912***			
		(1.412)		(0.341)		(0.345)			
Has Business Credit Card		-0.529*		-0.313*		-0.316*			
		(0.311)	•	(0.168)		(0.171)			
Has Business Credit Line		-0.535		0.413**		0.413**			
	-	(0.335)	•	(0.167)		(0.169)			
Year Dummy: 2008		0.192		-0.332*		-0.374*			
		(0.393)	•	(0.197)		(0.201)			
Year Dummy: 2009		0.0601		-0.311		-0.324			
		(0.395)	•	(0.198)		(0.203)			
Constant		10.12*		1.043		1.303			
		(5.342)	•	(1.374)	•	(1.409)			
Owner Demographic Controls		Υ		Υ		Υ			
Firm Char. Controls		Υ		Υ		Υ			
Industry Fixed Effects		Υ		Υ		Υ			
Observations	r	101	r	606	r	602			

Table 4. Loan Application Approval (2 stage)

This table reports coefficient estimates from two-stage Heckprobit regressions with selection correction. The dependent variable in in the first stage is a dummy variable equal to 1 if the entrepreneur applied for a new loan and 0 otherwise. The dependent variable in second stage regression is a dummy variable equal to 1 if the loan was approved. Column 1 is years 2007, 2008, and 2009 for high tech firms only. Column 2 and 3 are for all firms. Column 3 includes interactions with *high-technology industry*. All columns include entrepreneur, firm, and industry controls. Standard errors (in parentheses) are survey-weighted and heteroskedasticity robust. Significance is denoted as: p < 0.01, p < 0.05, p < 0.10

High Tech Firms										
VARIABLES .	Only	All Firms	All Firms							
First Stage Results	,									
Fear of denial	0.705***	0.636***	0.633***	_						
	(0.155)	(0.0580) ₁	(0.0580)							
Bank Loans/Total Financial Capital	0.708*	0.399***	(0.394***	1						
	(0.412)	(0.0760)	(0.0757)	-						
Second Stage Results	_	·	_							
High Tech Industry		-0.0778	-0.236**							
		(0.0718)	(0.108)	1						
Previous Startup Experience	0.0807	-0.0914**		-						
· · · · ·	(0.131)	(0.0407)	(0.0402)	7						
High Credit Score	0.209*	-0.0169		1						
	(0.114)	(0.0410)	(0.0434)	4						
High-Tech X High Credit Score	<u> </u>	· ' -	0.450***	1						
			— (0.109)	1						
Intellectual Property	-0.143	-0.105**	-0.0910*	4						
· · · · · · · · · · · · · · · · · · ·	(0.137)	(0.0471)	(0.0501)	1						
High Tech X Intellectual Prop.		i i	-0.235**	1						
			(0.106)							
R&D Employment	0.0248	-0.127***	-0.126***	7						
	(0.220)	(0.0469)		٦						
Have Trade Credit	-0.183*	0.0277	0.0245							
	(0.100)	(0.0428)	(0.0420)]						
Log of Employment	-0.00783	-0.0110	-0.0110	٦						
	(0.0609)	(0.0250)	(0.0246)	_						
Log of Assets	-0.270	0.00466	0.00624]						
·	(0.202)	(0.0374)	(0.0381)]						
Insider financing/Total Financial Capital	-0.746**	-0.310***	-0.304***	_						
	(0.324)	(0.101)	(0.100)]						
Has business credit card	0.171*	-0.0981**	0.0940**]						
	(0.0973)	(0.0413)	(0.0412)	-						
Has business credit line	0.173*	0.106***	0.110***							
	_ (0.0943)	(0.0394)	(0.0391)	- 1						
2008 Year Dummy	0.0320	-0.0751	0.0803*	4						
	(0.0917)	(0.0472)	(0.0472)	Ţ						
2009 Year Dummy	0.00760	-0.121**	-0.122**	Ţ						
	(0.158)	(0.0480)	(0.0480)	_						
athro	1.156	1.506***	1.516***	Ţ						
	_ (1.162)	(0.129)	(0.130)	4						
Insigma	-0.734	-0.355***	-0.366***							
	(0.530)	(0.0552)	(0.0556)	_						
Owner Demographic Controls	Y	Y	Y	—						
Firm Chareristic Controls	Y	T Y	Ŷ	ı						
Industry Fixed Effects	Y	Υ	Y	_						
Observations	605	4415	4411	_						

Appendix 1: Sample (All Firms, Analysis Sample (those that survived until at least 2007), Closed by 2007)

Baseline Data (2004)

Owner Characteristics	All Firms	Survivors	Closures		
Average Hours Worked (week)	42.4	43.0	41.1		
Previous Industry Work Experience	11.8	12.6	10.4		
Previous Startup Experience	42.5%	43.1%	41.5%		
Number of Previous Startups	1.0	1.0	1.1		
Owner Age	44.9	45.3	44.4		
Black	8.8%	7.9%	11.1%		
Asian	4.2%	4.3%	3.4%		
Hispanic	5.2%	4.8%	6.2%		
Other	2.3%	1.8%	2.9%		
White	79.7%	81.2%	76.2%		
Female	30.5%	29.5%	32.5%		
High School Graduate or Less	13.5%	12.1%	16.4%		
Some College	36.4%	35.0%	37.6%		
College Degree	30.2%	31.6%	29.7%		
Post-Grad Education	17.5%	18.4%	14.7%		
Firm Characteristics					
Incoporated	58.4%	57.8%	61.4%		
Home Based	49.4%	50.3%	47.6%		
Comparative Advantage	62.6%	64.1%	60.9%		
Intellectual Property	19.3%	20.4%	16.9%		
Employer Firm	38.3%	37.5%	39.9%		
Employment	1.858	1.772	2.105		
High Credit Score	12.0%	13.3%	8.9%		
Multi-Owner Firm	34.8%	33.6%	37.8%		
Startup Financing					
Owner Equity	28.2%	26.2%	32.8%		
Insider Equity	1.7%	1.3%	3.1%		
Outsider Equity	15.7%	18.0%	8.0%		
Owner Debt	4.2%	3.9%	5.3%		
Insider Debt	5.8%	5.1%	7.7%		
Outsider Debt	44.4%	45.4%	43.1%		
Total Startup Capital	\$ 115,835	\$ 122,983	\$ 101,184		

Source: Kauffman Firm Survey Microdata

Appendix 2: Startup Capital and Subsequent New Financial Injections (2004-2009)

<u></u>													
•		2004		2005		2006		2007		2008		2009	
Owner Equity	\$	32,612	\$	16,728	\$	12,858	\$	10,304	\$	10,218	\$	8,676	
Insider Equity	\$	1,929	\$	1,539	\$	846	\$	577	\$	551	\$	833	
Outsider Equity	\$	18,232	\$	20,097	\$	16,308	\$	11,522	\$	5,477	\$	10,371	
Owner Debt	_ \$	4,884	\$	4,595	\$	4,058	\$	4,173	\$	4,675	\$	3,034	
Insider Debt	\$	6,704	\$	5,847	\$	5,346	\$	4,815	\$	3,386	\$	10,118	
Outsider Debt	\$	51,474	\$	47,430	\$	54,405	\$	73,480	\$	47,435	\$	75,605	
Total Financial	\$	115,835	\$	96,235	\$	93,821	\$	104,870	\$	71,741	\$	108,636	
										_			
Owner Equity		28.2%		17.4%		13.7%		9.8%		14.2%	,	8.0%	
Insider Equity		1.7%		1.6%	0.9%		0.5%		0.8%		0.8%		
Outsider Equity		15.7%		20.9%		17.4%		11.0%		7.6%		9.5%	
Owner Debt		4.2%		4.8%	4.3%		4.0%		6.5%		2.8%		
Insider Debt		5.8%	6.1%		5.7%		4.6%		4.7%		9.3%		
Outsider Debt		44.4%	49.3%		58.0%		70.1%		66.1%		69.6%		
Total Financial		100.0%	2	100.0%		100.0%		100.0%		100.0%		100.0%	

Source: KFS microdata

References

- Amit, Raphael, Lawrence Glosten, and Eitan Muller, 1990, Entrepreneurial ability, venture investments, and risk sharing, *Management Science* 36, 1233-1246.
- Ayyagari, Meghana, Asli Demirguc-Kunt, and Vojislav Maksimovic, 2010, Formal versus informal finance: Evidence from china, *Rev. Financ. Stud.*
- Ballou, Janice, Tom Barton, David Desroches, Frank Potter, E. J. Reedy, Alicia Robb, Scott A. Shane, and Zhanyun Zhao, 2008, The kauffman firm survey: Results from the baseline and first follow-up surveys, http://ssrn.com/abstract=1098173
- Beck, Thorsten, Asli Demirgüç-Kunt, and Vojislav Maksimovic, 2008, Financing patterns around the world: Are small firms different?, *Journal of Financial Economics* 89, 467-487
- Benfratello, Luigi, Fabio Schiantarelli, and Alessandro Sembenelli, 2008, Banks and innovation: Microeconometric evidence on italian firms, *Journal of Financial Economics* 90.
- Berger, Allen N., and Gregory F. Udell, 1998, The economics of small business finance: The roles of private equity and debt markets in the financial growth cycle, *Journal of Banking & Finance* 22, 613-673.
- Bhide, Amar V., 2000. *The origin and evolution of new businesses* (Oxford University Press, Oxford).
- Black, Sandra E., and Philip E. Strahan, 2002, Entrepreneurship and bank credit availability, *Journal of Finance* 57, 2807-2833.
- Cagetti, Marco, and Mariacristina De Nardi, 2006, Entrepreneurship, frictions, and wealth, *Journal of Political Economy* 114, 835-870.
- Casamatta, Catherine, 2003, Financing and advising: Optimal financial contracts with venture capitalists, *The Journal of Finance* 58, 2059-2085.
- Chava, Sudheer, and Amiyatosh Purnanandam, The effect of banking crisis on bank-dependent borrowers, *Journal of Financial Economics* In Press, Corrected Proof.
- Chemmanur, Thomas J., Shan He, and Debarshi K. Nandy, 2009, The going-public decision and the product market, *Rev. Financ. Stud.*
- Cole, Rebel A., 2008, What do we know about the capital structure of privately held firms? Evidence from the surveys of small business finance, http://ssrn.com/paper=1013085.
- Cosh, Andy, Douglas J. Cumming, and Alan Hughes, 2009, Outside entrepreneurial capital, *Economic Journal* 119, 1494-1533.
- Edelstein, Robert H., 1975, Improving the selection of credit risks: An analysis of a commercial bank minority lending program, *The Journal of Finance* 30, 37-55.

- Evans, David S., and Boyan Jovanovic, 1989, An estimated model of entrepreneurial choice under liquidity constraints, *Journal of Political Economy* 97, 808-827.
- Gorton, Gary, and Andrew Winton, 2003, Chapter 8 financial intermediation, in M. Harris G.M. Constantinides, and R. M. Stulz, eds.: *Handbook of the economics of finance* (Elsevier).
- Haynes, George, and Scott R. Brown, 2009, An examination of financial patterns using the survey of small business finance, Small Business in Focus: Finance, A Compendium of Research by the Small Business Administration's Office of Advocacy (Washington, DC).
- Hecker, Daniel, 2005, High-tech employment: A naics based update, *Monthly Labor Review* 57-72.
- Hellmann, Thomas, 2006, Ipos, acquisitions, and the use of convertible securities in venture capital, *Journal of Financial Economics* 81, 649-679.
- Hellmann, Thomas, Laura Lindsey, and Manju Puri, 2008, Building relationships early: Banks in venture capital, *Rev. Financ. Stud.* 21, 513-541.
- Jensen, Michael C., and William H. Meckling, 1976, Theory of the firm: Managerial behavior, agency costs and ownership structure, *Journal of Financial Economics* 3, 305-360.
- Kaplan, Steven N., and Per Stromberg, 2003, Financial contracting theory meets the real world: An empirical analysis of venture capital contracts, *The Review of Economic Studies* 70, 281-315.
- Kaplan, Steven N., and Per Stromberg, 2004, Characteristics, contracts, and actions: Evidence from venture capitalist analyses, *The Journal of Finance* 59, 2177-2210.
- Kerr, William, and Ramana Nanda, 2009, Banking deregulations, financing constraints, and firm entry size, *National Bureau of Economic Research Working Paper Series* No. 15499.
- Landier, Augustin, and David Thesmar, 2009, Financial contracting with optimistic entrepreneurs, *Review of Financial Studies* 22, 117-150.
- Nanda, Ramana, 2008, Cost of external finance and selection into entrepreneurship, (SSRN).
- Parker, Simon C., 2009. *Economics of entrepreneurship* (Cambridge University Press, Cambridge).
- Reedy, E. J., and Alicia Robb, 2009, Comparisons between the kauffman firm survey and other business data
- Robb, Alicia, E. J. Reedy, Janice Ballou, David DesRoches, Frank Potter, and Zhanyun Zhao, 2010, An overview of the kauffman firm survey: Results from the 2004-2008 data, (Ewing Marion Kauffman Foundation).
- Schmidt, Klaus M., 2003, Convertible securities and venture capital finance, *Journal of Finance* 58, 1139-1166.

- Shane, Scott, 2008. The illusions of entrepreneurship: The costly myths that entrepreneurs, investors, and policy makers live by (Yale University Press, New Haven).
- Ueda, Masako, 2004, Banks versus venture capital: Project evaluation, screening, and expropriation, *Journal of Finance* 59, 601-621.
- Winston Smith, Sheryl, 2010, Beg, borrow, and deal? Entrepreneurship and financing in new firm innovation, *Ewing Marion Kauffman Foundation Research Paper Series* http://ssrn.com/paper=1573685.
- Winton, Andrew, and Vijay Yerramilli, 2008, Entrepreneurial finance: Banks versus venture capital, *Journal of Financial Economics* 88, 51-79.
- Wooldridge, Jeffrey M., 2002. *Econometric analysis of cross section and panel data* (MIT Press, Cambridge, MA).