Proksch, Dorian; Stranz, Wiebke; Pinkwart, Andreas; Schefczyk, Michael

Article
Risk management in the venture capital industry:
Managing risk in portfolio companies

The Journal of Entrepreneurial Finance (JEF)

Provided in Cooperation with:
The Academy of Entrepreneurial Finance (AEF), Los Angeles, CA, USA

Suggested Citation: Proksch, Dorian; Stranz, Wiebke; Pinkwart, Andreas; Schefczyk, Michael (2016) : Risk management in the venture capital industry: Managing risk in portfolio companies, The Journal of Entrepreneurial Finance (JEF), ISSN 1551-9570, Montrose, CA, The Academy of Entrepreneurial Finance (AEF), Vol. 18, Iss. 2, pp. 1-33

This Version is available at:
http://hdl.handle.net/10419/197553

Terms of use:
Documents in EconStor may be saved and copied for your personal and scholarly purposes. You are not to copy documents for public or commercial purposes, to exhibit the documents publicly, to make them publicly available on the internet, or to distribute or otherwise use the documents in public.

If the documents have been made available under an Open Content Licence (especially Creative Commons Licences), you may exercise further usage rights as specified in the indicated licence.

https://creativecommons.org/licenses/by-nc/4.0/
Risk management in the venture capital industry: Managing risk in portfolio companies

Dorian Proksch  
*HHL Leipzig Graduate School of Management*

Wiebke Stranz  
*HHL Leipzig Graduate School of Management*

Andreas Pinkwart  
*HHL Leipzig Graduate School of Management*

Michael Schefczyk  
*Technical University of Dresden*

Follow this and additional works at: [https://digitalcommons.pepperdine.edu/jef](https://digitalcommons.pepperdine.edu/jef)  
Part of the [Entrepreneurial and Small Business Operations Commons](https://digitalcommons.pepperdine.edu/jef) and the [Finance and Financial Management Commons](https://digitalcommons.pepperdine.edu/jef)

Recommended Citation  
Available at: [https://digitalcommons.pepperdine.edu/jef/vol18/iss2/1](https://digitalcommons.pepperdine.edu/jef/vol18/iss2/1)

This Article is brought to you for free and open access by the Graziadio School of Business and Management at Pepperdine Digital Commons. It has been accepted for inclusion in The Journal of Entrepreneurial Finance by an authorized editor of Pepperdine Digital Commons. For more information, please contact josias.bartram@pepperdine.edu, anna.speth@pepperdine.edu.
Risk management in the venture capital industry: Managing risk in portfolio companies

Cover Page Footnote
We thank the nine venture capital funds from Germany which supported our research by allowing us to analyze their documents and interview the investment managers. We are grateful for their trust and support. Further, we thank the 'Wissenschaftsförderung der Sparkassen-Finanzgruppe e.V.' as well as the 'NRW.Bank' for their continuous support of our project. In addition, we would like to thank Arnd Wiedemann and Volker Stein from the University of Siegen for their value feedback at the 3rd Annual Conference on Risk Governance.
Risk management in the venture capital industry: Managing risk in portfolio companies

Dorian Proksch
HHL Leipzig Graduate School of Management

Wiebke Stranz
HHL Leipzig Graduate School of Management

Andreas Pinkwart
HHL Leipzig Graduate School of Management

Michael Schefczyk
Technical University of Dresden

ABSTRACT

Managing risk is one of the main activities of venture capital companies. Despite the fact that this topic is of high practical relevance, only little research was published on risk management performed by venture capital companies in their ventures. Hence, we conducted a structured literature review which was the basis for developing five hypotheses concerning measures to decrease failure risk in venture capital-backed ventures. We tested these hypotheses with an empirical data set of 93 venture capital-backed ventures in Germany using original deal data from nine different venture capital funds using a structural equation model. We showed that the experience and the skills of the corresponding investment manager have a significant negative impact on the failure risk of a venture. Investment manager’s experience and skills were measured by the working and founding experience, the technology expertise and the network size. Hence, the results emphasize the importance of the selection of the investment manager for risk management in venture capital investments.

Keywords: Risk management, failure risk, venture capital, new-technology-based firms

1 We thank the nine venture capital funds from Germany which supported our research by allowing us to analyze their documents and interview the investment managers. We are grateful for their trust and support. Further, we thank the ‘Wissenschaftsförderung der Sparkassen-Finanzgruppe e.V.’ as well as the ‘NRW.Bank’ for their continuous support of our project. In addition, we would like to thank Arnd Wiedemann and Volker Stein from the University of Siegen for their value feedback at the 3rd Annual Conference on Risk Governance.

Copyright © 2016 Pepperdine Digital Commons and the Academy of Entrepreneurial Finance. All rights reserved. ISSN: 2373-1761.
I. Introduction

Risk management can add value and is necessary in all types of companies to secure long-term stability (Frooth et al., 1993; Mackay and Moeller, 2007). Nevertheless, the topic of risk management is still in its infancy as articles are mainly published in finance and accounting, but less in management or entrepreneurship journals (Bromiley et al., 2015). Especially venture capital (VC) investments are well known as high risk investments since venture capital companies (VCCs) invest in ventures with a high growth but also high risk potential (LiPuma and Park, 2014). Young entrepreneurial firms face the challenge of “liability of newness” resulting in particular difficulties, e.g. shorter expected life, and a greater risk of failure (Ang, 1992; Coleman, 2004).

In the investment decision making process, VCC are often faced with uncertainty about the future performance of the venture and the adverse selection problem. The reason for that is that VCC have to rely on information about the venture supplied by the entrepreneur (Tourani-Rad and England, 2003). A comparative study by Zacharakis and Meyer (2000) showed that VC investments fail at a rate of 35 to 55 per cent. Young and entrepreneurial firms are an essential part of the German economy and an important source for innovation in order to stay competitive on a global basis. Hence, research on comprehensive risk management for the VC industry is of great practical importance to improve the practices how German VCCs pursue risk management which might reduce the risk of failure of their ventures.

Risk management pursued in VC-backed ventures is only moderately researched in academic literature (Yoshikawa et al., 2004; Tan et al., 2008). Previous studies either focus on single types of risk, e.g. macro-risk (Ning et al., 2015) or liquidity risk (Cumming et al., 2005) or on specific types of risk management measures, e.g. syndication (Wang et al., 2012), (Hopp, 2010) or financial contracting and incentive mechanisms (Tan et al., 2008). Studies analyzing comprehensive sets of risk management measures applied by VCCs ventures are limited (see e.g. Kut et al., 2006; Kut et al., 2007; Kut and Smolarski, 2006 and Smolarski et al., 2005). These studies used comparable methods and similar samples leading to a lack of new findings (Dimov and Murray, 2008; Milavo and Fernhaber, 2009). However, risk management is one of the core competencies of VCC and therefore a highly relevant topic in practice. One reason for the limited amount of studies in this field might be the lack of reliable data. VC-backed ventures are private companies and only limited subject to the
duty of publishing company data and financial statements (Bygrave, 2006; Neergaard and Ulhoi, 2006).

To analyze risk management measures and their impact on the failure risk of VC-backed ventures we pursued the following research strategy: We conducted a structured literature review to develop five hypotheses on risk management measures applied by VCC in ventures, i.e. the assessment and evaluation of new ventures, contracts, investment manager’s experience and skills, governance mechanisms and management support. These hypotheses were tested with a structural equation model using an empirical data set of 93 VC-backed ventures in Germany from nine different VC funds.

As risk management received relatively little attention in entrepreneurship literature (Pinkwart, 2002) and is an important research topic, but largely unsystematic and not easy to diversify (Manigart et al., 2002), we add to literature and practice as follows:

1. We used a rare data set with in-depth quantitative and qualitative data from nine public and private VC funds combining data obtained from a survey with investment managers and original deal documents like business plans, investment committee papers, reporting and annual statements.

2. We provide an analysis of the Germany VC market which was rather moderately studied in literature before. Thereby, we shed light into the risk management practices of German VC funds.

3. The results from our structural equation model imply that particularly investment manager’s experience and skills have a statistically significant impact to reduce failure risk in VC-backed ventures. This finding supports Hopp and Lukas (2014) who were among the first showing that investment managers can have technological, industry, financial and managerial experience and leadership skills which might be crucial for the success and failure ventures. Furthermore, governance mechanisms, e.g. milestones and reporting, were heavily applied by all VCC. However, contrary to other studies like Bengtsson and Sensoy (2011) and Tan et al. (2008) we cannot show that governance mechanisms have a significant effect on reducing the risk of venture’s failure.

II. Literature review and hypothesis development

Risk can be defined as the probability and severity of adverse effects (Aven, 2011). Therefore, risk management is crucial to manage the uncertainty of risks. A
sound risk management is characterized as proactive, aligned and economic including the identification, estimation, evaluation, treatment and monitoring of possible negative influences on performance (Hain, 2011). VCC are financial intermediaries investing foremost in ventures bridging the gap created by the shortage of appropriate financing for small and entrepreneurial firms (LiPuma and Park, 2014; Okpala, 2012). By investing in ventures VCC bear high risk due to information asymmetries between the investor and entrepreneur known as the principal agent problem (LiPuma and Park, 2014). Hence, VCC apply different types of risk management measures to reduce the risk of the investment. Risk management in VC-backed ventures was sparsely analyzed in academic literature (Yoshikawa et al., 2004; Tan et al., 2008). The studies of Kut et al. (2006), Kut et al. (2007), Smolarski et al. (2005) investigated how buy-out and VC funds in Europe overall, in India, UK, France and Germany manage risk in the pre-screening phase of the investment, in existing ventures, the portfolio risk and macro risk considering a comprehensive set of risk management measures. These studies showed first attempts to analyze a set of risk management measures. Nevertheless, the studies are subject to several limitations especially due to partially small samples.

We conducted a structured literature review to study the current state of academic literature on the topic of risk management in VC-backed ventures. First, we analyzed all entrepreneurship journals ranked in the 55th edition of the Journal Quality List edited by Prof. Anne-Wil Harzing from 2005 to 2015 regarding the keywords “risk”, “risk management”, “venture capital” and “failure”. We identified thirteen relevant studies. Second, we searched in the EBSCOhost Online Research Databases for the abovementioned keywords in the titles and abstracts of all types of academic journals from 2005 to 2015. Overall, we identified 17 relevant papers (see Appendix A, Table 4). The samples of the different studies vary greatly in size and data collection method. A considerable number of studies are of explorative nature due to partly small sample sizes. This implies that this field of research is relatively unexplored offering room for further research. The majority of papers used data from the US and in some cases parts of Europe or Asia. Only few studies were conducted in Germany.

A. Risk types in VC-backed ventures

VC investments are subject to several risks. Our structured literature review showed that academic scholars investigated agency risk, financial or liquidity risk, technology risk, market risk, human resources risk, internationalization risk and macro risk. In the following, the different types of risks are described.

Agency risk is often stated as the major risk for VCC due potential problems of adverse selection and moral hazard between entrepreneurs and VCC (Bengtsson and
Sensoy, 2011; Lu et al., 2006; Tan et al., 2008). The theory was developed by William Meckling, Eugene Fama and Michael Jensen depicting the conflict of interest between the principal and the agent, in the case of VC founders or managers of the venture and the VCC (Fama and Jensen, 1983; Jensen and Meckling, 1976). Mechanisms like financial contracting, milestones, gradual provision of capital and active involvement in the board are applied by venture capitalists to overcome the information asymmetry between the VCCs and the entrepreneur (Bengtsson and Sensoy, 2011; Lu et al., 2006; Tan et al., 2008).

**Liquidity or financial risks** are partially used as synonyms in academic literature. Kut et al. (2006), Kut et al. (2007), Smolarski et al. (2005) classified financial risk in their analysis on the level of the portfolio and macro economy. Contrary, liquidity risk was analyzed by Cumming et al. (2005) indicating that VCC adjust their investment decisions according to liquidity risk. Liquidity risk refers according to Cumming et al. (2005) to the exit risk for the VCC in IPO markets describing the risk of not being able to reach an exit in a proper way. The study showed that VCC prefer to invest in high-tech and early stage ventures to defer the exit and increases the syndication size (Cumming et al., 2005). In our analysis, we define liquidity or financial risk as the risk of the venture to become illiquid or even bankrupt.

**Technology risk** is often used synonymously as product risk, technological risk or technical risk in academic literature. Assessing the technology or product risk is crucial risk for VCC before investing in ventures due to the fact that technologies and products are often not market-ready. Technology due diligences, syndication and the opinions of investment managers with industry experience are used to overcome the risk associated with technologies and products (Kut et al., 2006; Kut et al., 2007; Smolarski et al., 2005; Wang et al., 2012).

**Market risk** is mainly related to the commercialization of a new technology (Wang et al., 2012). Ventures often lack the marketing capabilities necessary to take the technology to market (Wang et al., 2012). VCC apply due diligences to assess the market risk as a central part of the investment decision process (Lu et al., 2006). Furthermore, VCC utilize their network and skills to foster the market introduction. According to Kaplan and Strömberg (2002) major market risks are market size and growth, competition and barriers to entry and the likelihood of customer adoption. However, the results indicated that competition, market size and customer adoption risks mentioned at a moderate rate of 40, 31 and 22 per cent in the investment documents (Kaplan and Strömberg, 2002).

**Human resources risks** are risks associated with the quality and capabilities of the management of the venture. This was analyzed by the studies of Kut et al. (2006), Kut
et al. (2007) and Smolarski et al. (2005). In these studies human resources risk was measured by the lack of management performance and the lack of management focus. To mitigate the risk related to the management, VCC can verify the track record of the management team and can invest in management teams which are previously known (Kut et al., 2007). Kaplan and Strömberg (2002) showed that risks associated with the management were mentioned in 61% of the analyses. It was documented that the CEO is a “difficult” person, that the management lacks in financial planning, the management is not able to focus or that the management is young and inexperienced (Kaplan and Strömberg, 2002). In addition, a further risky issue for VCC is an incomplete management team (Kaplan and Strömberg, 2002). Overall, the results indicate that risks associated with human capital are of high relevance for VCC.

LiPuma and Park studied the special topic of internationalization risk using longitudinal data of 962 invested rounds in 334 VC-backed ventures (LiPuma and Park, 2014). Variables for risk management were round size, round interval and round syndication. Compared to solely domestic ventures, VCC use smaller syndicates and provide smaller and less frequent rounds of capital for ventures which internationalize opportunistically (LiPuma and Park, 2014).

Volatility and macroeconomic drivers, namely macro risk, affect VC investments by the total amount, by the number of deals, and by the average amount per deal (Ning et al., 2015). Types of macro risk can be inflation risk, business-cycle risk, interest rate risk and foreign exchange rate risk (Kut et al., 2006). Therefore, in times of high macro risk VCC adapt their risk preferences and investment strategies by investing in fewer deals with a smaller average amount per deal, raising their investments in later investment stages and injecting a lower percent of cash in the first several financing sequences (Ning et al., 2015).

Failure risk as one of the most severe risks for VCC was not contained in the results of our structured literature review. In a further search we explored that only very few researchers studied this topic empirically (Dimov and De Clercq, 2006). Therefore, we focused in our analysis on this type of risk since failure risk consists partially of the above mentioned risk types according to insolvency literature (Carter and van Auken, 2006), (Davila et al., 2003), (Headd, 2003), (Pinkwart et al., 2015), (Pleschak, 2002), (Schilling, 2002), namely liquidity risk, market risk, human resources risk and technology risk. Therefore we include these risk factors as variables in our model to describe failure risk.
B. Risk management in VC-backed firms

The literature review has shown that VCC apply the following risk management measures: 1) assessment and evaluation of new ventures (Kut et al., 2007; Lu et al., 2006), 2) governance (Bengtsson and Sensoy, 2011; Tan et al., 2008) and 3) contracting (Bengtsson and Sensoy, 2011; Kut et al., 2007; Tan et al., 2008). In the course of the interviews with practitioners we identified two further influencing factors to reduce risk in VC investments, investment manager’s experience and skills as well as management support. There might be some interactions between the different risk management measures, e.g. governance mechanisms and management support. In the context of support functions of VCC governance mechanisms are a part of the support functions. However, in this context we separated governance mechanisms from management support due to the fact that governance mechanisms belong to the most important risk management measures in VC deals.

Assessing and evaluating potential new investments are the first steps of risk management VCC can apply in the investment process. Kut et al. (2007) showed that risk management in evaluating new investments is a well-developed area in practice in the VC industry (Kut et al., 2007). VCC have a variety of tools to assess and evaluate potential investment targets regarding risk and return, e.g. performing different types of due diligences like financial, product, market, customer, legal, competitor analysis internally and externally and analyzing audited financial statements (Kut et al., 2007; Kut and Smolarski, 2006; Lu et al., 2006; Smolarski et al., 2005). Information asymmetries can for example be resolved through the overall coherence of the business plan and the VCC’s own due diligence report according to Tournani-Rad and England (2003). VCC can also check the risk associated with the management by verifying the track record of the management team and board members and performing criminal background checks (Kut et al., 2007; Kut and Smolarski, 2006; Smolarski et al., 2005). Further measures to be conducted before an investment decision is made can be the consideration of synergies with existing ventures and the risk preferences of the investors of the fund (Kut et al., 2007; Kut and Smolarski, 2006; Smolarski et al., 2005). We assume that a better assessment of the risk before investing might lead to a lower failure rate of VC-backed venture.

Hypotheses 1: A high effort in assessing and evaluating the investment is negatively related to failure risk of VC-backed ventures.

Financial contracting can be used by VCC as a protection against downside risk (Bengtsson and Sensoy, 2011; Kut et al., 2007; Tan et al., 2008), but also to generate value in portfolio companies by mitigating the agency problem with financial contracts (Kaplan and Strömberg, 2002). Financial contracting is one measure next to active
involvement (Kut et al., 2007) and direct monitoring to reduce information asymmetry, motivational and financial problems (Bengtsson and Sensoy, 2011). VCC apply financial contracting mechanisms like liquidation preference, anti-dilution rights, cumulative dividends, redemption rights, participation rights and pay-to-play provisions according to Bengtsson and Sensoy (2011) and Tan et al. (2008). Syndication is a common measure in the VC industry to team up for assessing and investing collaboratively ventures to share the risk (Hopp, 2010; Hopp and Lukas, 2014; Smolarski et al., 2005; Wang et al., 2012). Staged financing is a useful control mechanism for VCC to gather information and monitor the progress of the venture having the option to inject further capital when milestones are reached and periodically abandon the venture (Kut and Smolarski, 2006; Tan et al., 2008). Adding to this, Bengtsson and Sensoy (2011) identified that good governance abilities can be a substitute for measures of financial contracting. Therefore, we state that a high use of contracting mechanisms might lower the risk of failure of VC-backed firms:

**Hypothesis 2:** An extensive use of contracting mechanisms is negatively related to the failure risk of VC-backed ventures.

VCC are known as active investors in their ventures. Risk management mechanisms related to governance like milestones, reporting and an active involvement in the board are applied by VCC to reduce agency risk. This risk type describes the interest conflict in the relationship between the investor and the entrepreneur. A considerable amount of studies investigated how VCC use control and incentive mechanisms to enhance the firm’s performance and receive higher returns. Contrary, only a few studies focused on this topic to reduce downside or failure risk (Bengtsson and Sensoy, 2011; Tan et al., 2008). According to Tan et al. (2008) governance risk management measures can be distinguished in either control mechanisms like monitoring (e.g. reporting, frequency of interaction, convertible securities), staged investments, which we allocated to financial contracting, and the allocation of ownership and control rights or incentive mechanisms (Tan et al., 2008). Shares of stock rights of the entrepreneur and employee stock options are incentive mechanisms to reduce agency risk. The greater VCC’s monitoring abilities, the more effective is the monitoring at constraining the entrepreneur’s behavior (Bengtsson and Sensoy, 2011). From practice, we know that all VCC use control mechanism, hence we assume:

**Hypothesis 3:** The extensive use of governance mechanisms are negatively related to the failure risk of VC-backed ventures.

Investment managers are responsible for assessing new ventures and investment decisions in the pre-investment phase as well as the management of existing ventures in the post-investment phase, i.e. communication, meetings, controlling and supporting
the venture. Investment managers can have technological, industry, financial and managerial experience and leadership skills which might be crucial for the success and failure ventures (Hopp and Lukas, 2014). According to Hopp and Lukas (2014) more experienced investment managers control their investments less often than less experiences investors. Furthermore, more industry experience allows less frequent and intense evaluation (Hopp and Lukas, 2014). Yazdipour and Constand (2010) argued that researchers cannot ignore the human/managerial/decision-making side in failure prediction. Hence, they suggest in human decisions about the making or breaking of a private company a shifts from the commerce/operational (effect) side of failure analysis to the human/managerial/decision making (cause) side of it (Yazdipour and Constand, 2010). We assume that an experienced investment manager can be better in assessing risk and using countermeasures which can lead to a lower failure risk of VC-backed firms:

**Hypothesis 4:** The degree of investment manager’s experience is negatively related to the failure risk of VC-backed ventures.

A variety of studies proved that VCC add value to their portfolio companies (Alperovych and Hübner, 2013; Manigart et al., 2002; Sapienza, 1992; Sapienza et al., 1996) by applying different types of value added services like financials, governance, strategy, operational improvements and human capital improvements (Bottazzi et al., 2002; Cumming et al., 2005; Guo and Jiang, 2013; Tang et al., 2014; Timmons and Bygrave, 1986). We transferred the positive effects from management support to the literature of risk management in the VC industry. Hence, we assume that management support can have an impact on the failure risk of VCC’s portfolio companies:

**Hypothesis 5:** The extensive use of management support provided by VCC is negatively related to the failure risk of VCC’s portfolio companies.

### III. Data and method

#### A. Sample

Our sample consists of 93 VC-backed firms collected from nine different public and private public partnership VC funds in Germany. Considering the statistics of the Bundesverband Deutscher Kapitalbeteiligungsgesellschaften, which recorded 433 seed investments in Germany from 2005 to 2010, our sample covers 21.5% of the seed investments in this time frame in Germany. We conducted a survey with the corresponding investment managers. In addition, we had access to the original deal documents including the business plans, investment committee papers, reporting and
annual statements of the investments. That enabled us to collect in-depth quantitative and qualitative data.

Considering our data set, the VC-backed firms are on average 4.6 years old at the time of data collection. In the seed round the firms received on average 784,487 Euros as investment and in the series A round 1,202,948 Euros (see table 1). The firms in our data set are technology-based firms as they operate in the industries information technology and automation (39 %), life science (34%), material science (9 %), energy (5 %), communication (4 %) and others (9 %).

Table 1. Overview of our data set

<table>
<thead>
<tr>
<th>Variable</th>
<th>Mean</th>
<th>Median</th>
<th>Std. Dev</th>
</tr>
</thead>
<tbody>
<tr>
<td>Age of ventures</td>
<td>4.59 years</td>
<td>5</td>
<td>2.09</td>
</tr>
<tr>
<td>Size of founders team</td>
<td>2.85 founders</td>
<td>3</td>
<td>1.13</td>
</tr>
<tr>
<td>Number of founding rounds</td>
<td>1.98 rounds</td>
<td>2</td>
<td>0.89</td>
</tr>
<tr>
<td>Investment sum Seed</td>
<td>784,487 Euro</td>
<td>644,109 Euros</td>
<td>519,577</td>
</tr>
<tr>
<td>Investment sum Series A</td>
<td>1,202,948 Euros</td>
<td>816,287 Euros</td>
<td>1,179,085</td>
</tr>
<tr>
<td>Number of investors Seed</td>
<td>2.55</td>
<td>2</td>
<td>1.98</td>
</tr>
<tr>
<td>Number of investors Series A</td>
<td>3.94</td>
<td>3</td>
<td>2.54</td>
</tr>
</tbody>
</table>

B. Measures and variables

We used a structural equation model approach to build and test our model because failure risk can hardly be measured directly. Hence, we used a set of proxy variables. We built a partial least squares (PLS) model because of its suitability for proxy variables and the lack of existing scales in this field of research (Ainudding et al., 2007; Henseler et al., 2009). Furthermore, the fit of PLS models compared to covariance based methods for sample sizes smaller than 100 also attributed to our choice (Fornell and Bookstein, 1982). Not all of our items follow normal distribution. Hence, they would have been omitted once using a covariance based approach. In PLS models items do not have to follow a certain distribution (Hulland, 1999). In addition, we use variables measured with a 5-point Likert scale in our model. PLS models support the use of nominal, ordinal and interval scaled data (Fornell and Bookstein, 1982, Nitzl, 2010; see also Menzar and Nigh, 1995, Brinckmann et al., 2011). We
decided to use a reflective measurement model for the outer constructs of the risk management measures, the control variables and the construct of business failure as well as for the inner construct for two reasons. Reflective measurement models have defined reliability test criteria and are well researched (Roy and Tarafdar, 2012). Further, our indicator variables strongly correlate within our construct.

1. Dependent variables

Measuring failure risk of a company is difficult. Therefore, we measure failure risk by proxy variables, namely human resource risk, technology risk, financial risk and market risk based on the literature of bankruptcy and insolvency (Pinkwart et al., 2015). Pinkwart et al. (2005) showed that 80 per cent of the reasons for failure include a lack of management companies or management companies. Other studies confirmed human resources as an important reason for business failure (Carter and van Auken, 2006; Headd, 2003). A further cause of failure is risk related to the technology of a venture (Schilling, 2002). These companies are dependent on developing their technology to a working and marked-proved product. If ventures do not succeed in reaching the market readiness in a timely manner development costs can grow in outstanding way which ventures often cannot afford (Pleschak et al., 2002). Difficulties in getting a follow-up financing, miscalculation for the capital need and bad planning are among the most common reasons for business failure, namely financial risk (Davila et al., 2003; Headd, 2003; Pleschak et al., 2002; Thornhill and Amit, 2003). New ventures often need too long to break even or even fail because of the lack of financial resources. A further reason for the failure of companies can be found in the area of the market. Problems with the market entry or in marketing and sales are among the most common reasons of failure (Wagner, 1994; Dowling and Drumm, 2002; Pleschak, 2002). This can be explained by a lack of experience in marketing and sales as well as an overoptimistic planning (Hall, 1992; Thornhill and Amit, 2003). In addition, new companies often rely on a few key clients leading to a strong dependency from these customers (Brüderl et al., 1996; Guggemoos, 2012). We measured the five above mentioned risk types by the assessment of the supervising investment managers on a scale from 1 to 5 (1: very low risk, 5: very high risk).

2. Independent variables

As mentioned in chapter II. B. we identified five groups of risk management measures applied by VCC, i.e. assessment and evaluation of new ventures, contracting, governance, investment managers’ experience and skills and management support. Each group was measured by different items since VCC use several risk management measures for each group comprehensively in practice.
We used different items to measure the degree of assessment. We first looked if an external assessment of the company was done. From the VCC documents we knew that often external companies are hired to evaluate e.g. technology, market and legal risks. Further, we looked at the intellectual property protection. If the technology is protected by e.g. patents or registered designs the market and technology risks might be lower. In addition, we measured if the VCC relied on their network in assessing the technology and the competencies of the founders.

Contracts handle different aspects of risks between the entrepreneur and the VCC. An important item is liquidation preference. A high liquidation preference lowers the risk of VCC as it minimizes possible losses. We analyzed how strongly a liquidation preference was used. Further, we measured the number of syndication partners. Syndication is a possibility to share risk with other investors. Further, we looked at the investment sum. If the investment sum is lower it might increase the risk of failure in terms of liquidity. In addition we intended to measure if staged financing was used. However, due to the fact that this was the case for all our cases we did not include this item in our model.

Governance mechanisms like reporting and milestones are useful to assess risks continuously. To measure governance we included five proxy variables in our model. At first we looked if milestones were used and monitored. VCC often use milestones to bind founders to certain goals. If milestones are reached, founders receive the full investment sum. In addition, we looked at reporting. From expert interviews we knew that successful companies report regularly. If the company does not perform as expected, reporting rates might decrease. We therefore measured how heavily VCC rely on reporting. Furthermore, we included personal exchange in our model as it indicates a high interaction between founders and investment managers. Fourthly, we included the variable information through network in our model. According to principal agent theory a conflict exists between founders and VCC due to information asymmetry. Therefore, if VCC use their network to lower information asymmetries risks might be reduced. Lastly, we investigated at the shares of the founders. If the founders still have a high share of equity they might be more motivated financially and incentivized even if they lost decision rights due to the contract with the VCC.

Investment manager’s experience and skills might have a significant influence on the failure risk of ventures. We described the experience and skills of the investment manager by five variables. First, we looked at the working experience. More working experience might make it easier to deal in business environment. Second, we assessed the founding experience. Third, we analyzed the expertise with the field of technology. Forth, we assessed the business skills of the investment manager. An investment
The manager has to have a profound understanding of business to be able to evaluate the development of the ventures. Lastly, we analyzed the network size. With a superior network, the investment manager has more possibilities to get additional knowledge and support for areas he is not an expert on.

We measured the degree of management support by six variables. Firstly, we looked at the support by the VCC using own competencies. Bringing in their experience in the company might lead to better development of the portfolio company. Further, we looked at sales support. Young companies often fail because of a lack of sales activities. A support in the area could possibly lead to a lower risk of failure. Thirdly, we analyzed support with technology. For new ventures technology is a crucial success factor. Fourthly, we examined strategic support. A strong strategy is often an indicator for successful firms. In addition, we looked at support in follow-up financing. For new ventures it is critical to raise additional financing in a timely manner to avoid illiquidity and bankruptcy. Lastly, we measured the use of network in general to lower the risk of the venture after the investment took place. Networks might be useful to get new customers or consultants for solving issues.

3. Control variables

We controlled for age and industry. The risk of failure might be higher when companies are younger. Albach (1987) suggested that for most companies the highly probable chance of failure ends after five years. In addition, some industries might have higher failure rate than others.

IV. Results

A. Descriptive statistics

The results of the descriptive statistics are presented in appendix B (see Table 5). The failure risk was measured on a 5-point Likert scale (1: very low, 5 very high). The statistics show that liquidity risk has the highest value of 3.652 at the lowest standard deviation of 1.152. Technology risk was rated on average at 2.711 depicting the lowest failure of risk measures, but at the highest standard deviation of 1.455.

The descriptive statistics for the five groups of risk management present that governance mechanisms like milestones (mean = 4.247) or reporting (mean= 4.355) at a standard deviation of below 0.8 were deployed consistently high by the VCC in our sample. The same applies for risk management like obtaining references of founders (mean=4.086), liquidation preference (mean= 4.096) or support in follow up financing (mean= 4.065).
B. Results of structural equation model

The resulting path model is shown in table 2. The degree of assessment and evaluating new ventures has no significant influence on the failure risk. The t-statistic is not significant on a 95 per cent level for this construct. Therefore, we rejected H1.

Looking at contracting, we found no significance due to low t-statistics. Hence, we rejected H2.

Governance is not significant considering the low t-statistics. Therefore, we rejected H3.

Our results show a high impact of investment manager’s experience and skills on failure risk of VC-backed firms. This is indicated by the high factor loading as well as the high value of the t-statistics. The connotation of the loading is negative stating that a high experience leads to low risk. Thus, we accepted H4.

Management support is highly significant. Therefore, we accept H5. However the sign is surprisingly positive.

Table 2. Reliability measure of the PLS model

Significance of * 90 % level, ** 95 % level, *** 99 % level

<table>
<thead>
<tr>
<th>Construct</th>
<th>Loadings</th>
<th>t-Statistics</th>
<th>Cronbach’s alpha</th>
<th>Composite reliability</th>
<th>AVE</th>
<th>f²</th>
<th>q²</th>
</tr>
</thead>
<tbody>
<tr>
<td>Assessment and evaluation of new ventures</td>
<td>0.157</td>
<td>1.559</td>
<td>1</td>
<td>1</td>
<td>1</td>
<td>0.030</td>
<td>0.012</td>
</tr>
<tr>
<td>Contracting</td>
<td>0.194</td>
<td>1.239</td>
<td>1</td>
<td>1</td>
<td>1</td>
<td>0.039</td>
<td>0.006</td>
</tr>
<tr>
<td>Investment Manager’s experience and skills</td>
<td>-0.273</td>
<td>2.046**</td>
<td>0.682</td>
<td>0.806</td>
<td>0.511</td>
<td>0.088</td>
<td>0.014</td>
</tr>
<tr>
<td>Governance mechanisms</td>
<td>-0.224</td>
<td>1.300</td>
<td>0.705</td>
<td>0.818</td>
<td>0.604</td>
<td>0.037</td>
<td>0.012</td>
</tr>
<tr>
<td>Management support</td>
<td>0.451</td>
<td>3.219**</td>
<td>1</td>
<td>1</td>
<td>1</td>
<td>0.241</td>
<td>0.05</td>
</tr>
<tr>
<td>Failure risk</td>
<td>-</td>
<td>-</td>
<td>0.715</td>
<td>0.823</td>
<td>0.540</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

The control variables, age and industry, had no significant effect. We therefore removed them from the final model.
C. Results of inner models

In the following, we looked at the indicator variables of the single constructs. The indicator loadings and t-statistics are shown in table 3. Looking at the construct assessment and evaluating ventures the only significant variable was expert assessment. IP protection, references for technology and references for founders are not significant. Analyzing the construct contracting only liquidation preference is significant. We omitted the items syndication partner, investment sum and shares of investors. All items except of business skills were relevant when we looked at the construct investment manager’s experience and skills. In the construct governance mechanisms all variables were significant.

Looking at the construct business failure all variables were significant and had a high factor loading. The factor loading was above 0.6 for all our variables which is an acceptable value (Hair et al., 2013). This showed the validity of our approach to measure failure risk by using the four most important risks of bankruptcy.

Table 3. Loadings and t-statistics of the items
Significance of * 90 % level, ** 95 % level, *** 99 % level

<table>
<thead>
<tr>
<th>Item</th>
<th>Loadings</th>
<th>t-Statistics</th>
<th>Item</th>
<th>Loadings</th>
<th>t-Statistics</th>
</tr>
</thead>
<tbody>
<tr>
<td>Assessment and evaluation of new ventures</td>
<td></td>
<td></td>
<td>Management support</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Evaluation</td>
<td>1</td>
<td>-</td>
<td>Use of Contacts</td>
<td>1</td>
<td>-</td>
</tr>
<tr>
<td>Contracting</td>
<td></td>
<td></td>
<td>Failure</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Liquidation</td>
<td>1</td>
<td>-</td>
<td>HR risk</td>
<td>0.624</td>
<td>3.695**</td>
</tr>
<tr>
<td>Investment Manager’s experience and skills</td>
<td></td>
<td></td>
<td>Liquidity risk</td>
<td>0.775</td>
<td>9.426**</td>
</tr>
<tr>
<td>Working experience</td>
<td>0.797</td>
<td>2.258**</td>
<td>Market risk</td>
<td>0.701</td>
<td>5.126**</td>
</tr>
<tr>
<td>Founding experience</td>
<td>0.705</td>
<td>2.957**</td>
<td>Technology risk</td>
<td>0.823</td>
<td>7.904**</td>
</tr>
<tr>
<td>Network size</td>
<td>0.623</td>
<td>1.976**</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Technology expertise</td>
<td>0.722</td>
<td>2.066**</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Governance mechanisms</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
</tbody>
</table>
D. Reliability measures

We followed the framework of Hair et al. (2013) to assess the reliability of the PLS model. We therefore looked separately at the structural model and the measurement model.

1. Reliability measures of structural model

The $R^2$ of our model was 0.282 which is an acceptable value (Huber et al., 2007; Nitzl, 2010). The $Q^2$ value was greater than zero and therefore indicates a predictive relevance of the model (Henseler et al., 2009). The effect size of the constructs contracting, investment manager’s experience and skills, governance and support were above 0.02 showing a weak effect. The effect size of the construct assessment and evaluation of new ventures was below 0.02. This is not surprising as the t-test is not significant and the factor loading is below 0.2. The predictive relevance for the construct investment manager’s experience and skills is weak stated by a value above 0.02. The value for predictive relevance for the constructs is below 0.02 indicating a low predictive relevance. We choose to include the constructs in the model due to the explorative nature of the study.

2. Reliability measures of inner model

Indicators with a loading below 0.4 were stepwise removed so that only indicators with a standardized indicator loading above this value were included in the model. This is an acceptable approach according to Henseler et al. (2009) and Hair et al. (2013). All indicators are significant on a 95 per cent level determined by the t-statistics. The internal consistency reliability was measured by Cronbach’s alpha and composite reliability. The value for Cronbach’s alpha was above 0.6, which is permissible (Nunnally and Bernstein, 1994). The composite reliability was above 0.7 for all the constructs (see table 2), which is an acceptable value (Henseler et al., 2009; Hair et al., 2013). The average variance accepted (AVE) was used to measure the convergent validity. This approach is widely accepted in literature (Fornell and Larcker, 1981; Hair et al., 2013). All the measures were above 0.5 showing an excellent value. We used Fornell-Larcker criterion results, cross loadings and heterotrait-monotrait

<table>
<thead>
<tr>
<th>Milestones</th>
<th>0.883</th>
<th>2.926**</th>
</tr>
</thead>
<tbody>
<tr>
<td>Information</td>
<td>0.663</td>
<td>2.231**</td>
</tr>
<tr>
<td>through network</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Reporting</td>
<td>0.770</td>
<td>2.828**</td>
</tr>
</tbody>
</table>
ratio of correlations (HTMT) to test for discriminant validity (Fornell and Larcker, 1981; Ringle et al., 2015). The model passed all three tests as described in the appendix C.

V. Discussion and conclusion

Based on a structured literature review, the analysis of qualitative data of nine VC funds and an empirical analysis using a structural equation model we studied five groups of risk management measures VCCs can partake in their ventures to reduce failure risk. We empirically tested the relevance for each group of risk management measures. As a result, we show which risk management measures have an influence on business failure of VC-backed ventures.

First, the assessment of the investment prior the decision had no significant influence to reduce the failure risk in VCC’s ventures in the model. Therefore, we cannot support the studies of Kut et al. (2007), Kut and Smolarski (2006), Lu et al. (2006), Smolarski et al. (2005) and Tourani-Rad and England (2003) showing the significant relevance of assessment and evaluation in the pre-investment phase. One reason might be that this is often seen as the most important part in the investment decision process. The usage of different assessment methods was high for all cases in the sample (see descriptive statistics in A1), which confirms Kut et al. (2007) that risk management in evaluating new investments is a well-developed area in practice in the VC industry. However, the difference between the usages within the ventures might not be very high resulting in no significant influence.

Second, the construct contracting is not significant. Hence, we cannot support the results of Bengtsson and Sensoy (2011), Kut et al. (2007) and Tan et al. (2008) that financial contracting can be used by VCC as a protection against downside risk. A reason for that could be that VCC use similar formats of contracting, which also can be seen in the descriptive statistics (A1). In addition, all VCC used staged financing, syndication and milestone with each venture. This implies no significant differences across the cases in the sample.

Third, the results show that governance mechanisms are not significant to reduce failure risk in the model. The descriptive statistics showed that governance mechanisms are extensively used in all ventures supporting no significance of the statistical results (see Appendix A). Considering this result, we cannot support Bengtsson and Sensoy (2011) and Tan et al. (2008) who found significant evidence for the importance of governance mechanisms in VC risk management.
The construct investment manager’s experience and skills as a risk management measure in VC-backed ventures are significant, which was rarely discussed in literature before. This finding continues the discussion of Hopp and Lukas (2014) who were among the first showing that investment managers have various competencies, skills and experiences which might be crucial for the success and failure ventures. Also, the study of Yazdipour and Constand (2010) highlighted the importance of human capital in failure prediction of private firms. If the investment manager is more experienced the VC-backed ventures have less failure risk as the investment manager might be able to uncover possible problematic issues earlier and use the right countermeasures. In addition, we found that also VC-backed ventures supported by investment managers with founding experience have a lesser risk of failure.

Last, we found that management support is significant. Counter-intuitively, the connotation is negative. The extensive use of management support leads to a higher risk. This might be a chicken-and-egg problem. Possibly, investors only extensively support their portfolio companies when risk is already high which might be too late to save the company. To test this assumption we created a PLS model to analyze the influence of business failure on the degree of management support. We found that a high chance of business failure has a positive impact on the degree of management support as described by the use of VCC’s network and bringing external consultants into the portfolio company. Therefore, we can assume that this result might be explained by a chicken-egg problem. Considering this problem in the study, we recommend further investigation on the use of the risk management measures in the VC industry. It would be interesting to analyze if risk mitigation measures are only used when a risk occurred or also in a preemptive way.

VI. Limitations and implications

A. Limitations

Like most empirical studies the research is subject to several limitations. First, we could not assess all risk management measures identified in literature. A holistic model including further risk management measures could lead to additional results.

Secondly, we used a self-assessment of the investment manager for their experience in a survey. This might introduce a possible bias. However, the survey covered a variety of different areas of VC financing wherefore it was not clear for the investment managers that a connection between their experience and the risk was made.
Thirdly, we focused on German technology start-ups from public and private VC funds. It is unclear if the results can be generalized to other countries and all types of new ventures. Therefore, similar studies, e.g. using data in the US or Asia, might uncover similarities and differences between risk management measures across countries.

The data set consists of a higher share of public as well as public private partnership funds which also could include a possible bias as public funds might pursue a different investment and risk management strategy as private funds.

The quantitative approach does not allow to further study changes in the perceived failure risk over time. A qualitative approach to explicitly study the development of the risks in different investment stages could further yield to new results.

In addition, the use of PLS does not allow to control for endogeneity effects which is also discussed in current literature (Ronko and Evermann, 2013; Henseler et al., 2014).

The results might be partially biased due to the fact that our sample includes VC investments from 2005 - 2010, i.e. during the financial crisis. It might be possible that risks were higher during that time because of the economic downturn and the restricted capital situation.

B. Implications

Our research has several implications for the literature and practice.

In terms of the literature, the analysis has shown that the research stream of risk management in VC investments is rather underdeveloped, but nevertheless of great practical importance, for VCCs. We tested the effectiveness of different risk management measures on lowering the risk of business failure in new ventures. Thereby, we showed the importance of risk management on the probability of failure. With this article we aim to encourage discussions on and analyses of this field of literature to shed more light on VCC risk management practices. The results indicate the relevance of the investment manager in risk management in VC investments. Continuing this discussion, a possible research question could be which experiences, skills and knowledge as well as what kind of interaction between founders lowers venture’s failure risk. Another research direction might be a cross-country analysis as there are several differences, e.g. legal, between European and US VCCs. The German law for asset management companies like VCC prohibits active involvement of the VCC in the portfolio firm. VCC are only allowed to provide advice which also impacts their risk management practices. Further one, a mixed method approach including, for
example, surveys, verbal protocol analysis, and content analysis might be favorable to explore aspects of formal and informal risk assessment in VC investments. Furthermore, since our sample is limited to early stage VC funds, further investigations into different fund stages might be of interest to explore the differences across investment stages.

In terms of practice, we showed that all VCC in our sample pursue comparable risk management measures for the assessment and evaluation of new ventures, in contracting as well as in governance issues. Looking at the descriptive statistics we observed that particularly governance mechanisms, liquidation preferences and partially assessment and evaluation measures are extensively applied by VCC in their ventures in our data set. Nevertheless, these mechanisms do not show a significant influence on failure risk, which might be explained by the fact that they act like hygiene factors. Our study provides empirical evidence for the great importance of investment manager’s experience and skills which could be understand as the motivator of the analyzed risk management. Considering our empirical results, LPs and VC funds should therefore rely on highly experienced employees managing ventures. The results suggest that VCC have to invest in their human capital to improve the skills and knowledge of their investment managers as well as the working environment and conditions to hire the best investment managers. In that course, an exchange between more experienced and younger investment managers triggered by the VCC might be a possibility to achieve a knowledge transfer.

REFERENCES


Fornell C. and Larcker D., Structural equation models with unobservable variables and measurement error: Algebra and statistics, Journal of Marketing Research, 1981, 18, 382–389.


APPENDIX A:

Table 4. Overview of academic work on risk management in VC-backed ventures from 2005-2015

<table>
<thead>
<tr>
<th>Reference</th>
<th>Sample</th>
<th>Data collection method</th>
<th>Data analysis method</th>
</tr>
</thead>
<tbody>
<tr>
<td>Bamford and Douthett (2013)</td>
<td>Initial public offerings (IPO) n = 545</td>
<td>Investors Daily Digest and Barron’s</td>
<td>Descriptive statistics Logistic regression OLS estimation</td>
</tr>
<tr>
<td>Bengtsson and Sensoy (2010)</td>
<td>Private partnership VCs n = 646</td>
<td>Private Consulting firm VCExperts</td>
<td>Descriptive statistics Correlations Kruskal-Wallis test Probit regression OLS regression Heckman-Sorensen Index</td>
</tr>
<tr>
<td>Cumming et al. (2005)</td>
<td>Investment rounds n = 18,774</td>
<td>VentureExpert</td>
<td>Descriptive statistics Logit regression Poisson regression</td>
</tr>
<tr>
<td>Study</td>
<td>Sample Description</td>
<td>Data Source</td>
<td>Analysis Method</td>
</tr>
<tr>
<td>-------</td>
<td>--------------------</td>
<td>-------------</td>
<td>-----------------</td>
</tr>
</tbody>
</table>
| Hopp (2010) | Capital contributions<br>
n = 2,373 (961 ventures and 437 VCCs) | Thomson Venture Economics | Comparison of proportions and means tests<br>Correlations |
| Hopp and Lukas (2014) | VC investments<br>
n = 2,373 in Germany | Public sources and Thomson Venture Economics | Descriptive statistics<br>Correlations<br>Logistic regression |
| Kut et al. (2007) | Venture capital and buy-out funds<br>
n = 142 | Survey | Descriptive statistics<br>Mann-Whitney test<br>Pearson chi-square test |
| Kut et al. (2006) | Venture capital and buy-out funds<br>
n = 142 | Survey | Descriptive statistics<br>Mann-Whitney test<br>Pearson chi-square test<br>Logit regression<br>OLS regression |
| Kut and Smolarski (2006) | Private equity funds<br>
n = 33 from Germany and France <br>n = 21 from India | Survey | Descriptive statistics<br>Mann Whitney test<br>Pearson Chi-square test |
| LiPuma and Park (2014) | Invested rounds<br>
n = 962<br>VC-backed technology companies<br>
n = 334 | InfoUSA’s CorpTech data | Descriptive statistics<br>Pearson Correlations<br>GLS regression<br>Poisson regression |
| Lu et al. (2006) | VC firms<br>
n = 34 | Questionnaire survey<br>EDB and AVCJ | Descriptive statistics<br>Mann-Whitney-Wilcoxon Correlations<br>Two-sided Pearson chi-square<br>Hierarchical regression |
| Maula et al. (2009) | Technology-based firms<br>
n = 91 | Venture Economics Database Survey | Descriptive statistics<br>Correlations<br>Standardized factor loadings<br>Goodness of fit statistics |
| Ning et al. (2015) | Venture investments and deals<br>
n = 68 | Money Tree Report from PriceWaterhouseCoopers/ National Venture Capital Association using data from Thomson Reuters | Descriptive statistics<br>Correlations<br>Multiple regression models |
Payne et al. (2009) | VC firms
n = 26
VC investors
n = 52 | Survey | Descriptive statistics
Varimax Rotated Factor Loadings
Correlations
GLS regression
Multivariate regression

n = 584 | Jay Ritter’s VentureXpert | Descriptive statistics
Tobit regression
OLS regression
Variance-covariance matrix

Smolarski et al. (2005) | Private equity funds
n = 32 from UK
n = 21 from India | Survey | Descriptive statistics
Mann-Whitney test
Pearson chi-square test

Tan et al. (2008) | VC firms
n = 53 | Survey | Descriptive statistics

Wang et al. (2012) | VC-backed companies
n = 1,757 (772 reporting sales information)
Financing rounds
n = 5,896 (1,757 VC-backed companies) | VentureXpert Alliances database | Descriptive statistics
Correlations
Regression analysis (OLS, negative binomial model)

**APPENDIX B:**

*Table 5. Descriptive statistics*

<table>
<thead>
<tr>
<th></th>
<th>Mean</th>
<th>Std. Dev</th>
<th>Scale</th>
<th>Data Source</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Failure risk</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>HR risk</td>
<td>3.200</td>
<td>1.317</td>
<td>Rating from 1 to 5</td>
<td>Survey with investment managers</td>
</tr>
<tr>
<td>Liquidity risk</td>
<td>3.652</td>
<td>1.152</td>
<td>Rating from 1 to 5</td>
<td>Survey with investment managers</td>
</tr>
<tr>
<td>Market risk</td>
<td>3.311</td>
<td>1.304</td>
<td>Rating from 1 to 5</td>
<td>Survey with investment managers</td>
</tr>
<tr>
<td>Technology risk</td>
<td>2.711</td>
<td>1.455</td>
<td>Rating from 1 to 5</td>
<td>Survey with investment managers</td>
</tr>
<tr>
<td><strong>Assessment and evaluation</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Expert assessment</td>
<td>3.795</td>
<td>0.915</td>
<td>Rating from 1 to 5</td>
<td>Survey with investment managers</td>
</tr>
<tr>
<td>IP protection</td>
<td>3.435</td>
<td>1.424</td>
<td>Rating from 1 to 5</td>
<td>Survey with investment managers</td>
</tr>
<tr>
<td>References of technology</td>
<td>3.806</td>
<td>1.002</td>
<td>Rating from 1 to 5</td>
<td>Survey with investment managers</td>
</tr>
<tr>
<td>References of founders</td>
<td>4.086</td>
<td>0.686</td>
<td>Rating from 1 to 5</td>
<td>Survey with investment managers</td>
</tr>
<tr>
<td>------------------------</td>
<td>-------</td>
<td>-------</td>
<td>-------------------</td>
<td>-------------------------------</td>
</tr>
</tbody>
</table>

**Contracting**

<table>
<thead>
<tr>
<th>Liquidation preference</th>
<th>4.096</th>
<th>0.990</th>
<th>Rating from 1 to 5</th>
<th>Survey with investment managers</th>
</tr>
</thead>
<tbody>
<tr>
<td>Number of syndication partners</td>
<td>2.568</td>
<td>1.975</td>
<td>Metric</td>
<td>Term sheet</td>
</tr>
<tr>
<td>Investment sum</td>
<td>436,169</td>
<td>206,874</td>
<td>Metric (Euros)</td>
<td>Investment committee papers</td>
</tr>
</tbody>
</table>

**Investment manager experience and skills**

<table>
<thead>
<tr>
<th>Working experience</th>
<th>3.311</th>
<th>0.932</th>
<th>Rating from 1 to 5</th>
<th>Survey with investment managers</th>
</tr>
</thead>
<tbody>
<tr>
<td>Founding experience</td>
<td>3.237</td>
<td>0.993</td>
<td>Rating from 1 to 5</td>
<td>Survey with investment managers</td>
</tr>
<tr>
<td>Technology expertise</td>
<td>3.355</td>
<td>0.842</td>
<td>Rating from 1 to 5</td>
<td>Survey with investment managers</td>
</tr>
<tr>
<td>Business skills</td>
<td>3.946</td>
<td>0.578</td>
<td>Rating from 1 to 5</td>
<td>Survey with investment managers</td>
</tr>
<tr>
<td>Network size</td>
<td>3.720</td>
<td>0.851</td>
<td>Rating from 1 to 5</td>
<td>Survey with investment managers</td>
</tr>
</tbody>
</table>

**Governance**

<table>
<thead>
<tr>
<th>Milestones</th>
<th>4.247</th>
<th>0.789</th>
<th>Rating from 1 to 5</th>
<th>Survey with investment managers</th>
</tr>
</thead>
<tbody>
<tr>
<td>Information through network</td>
<td>4.323</td>
<td>0.710</td>
<td>Rating from 1 to 5</td>
<td>Survey with investment managers</td>
</tr>
<tr>
<td>Reporting</td>
<td>4.355</td>
<td>0.653</td>
<td>Rating from 1 to 5</td>
<td>Survey with investment managers</td>
</tr>
<tr>
<td>Shares of Founder</td>
<td>83.30</td>
<td>8.830</td>
<td>Per cent</td>
<td>Term sheet</td>
</tr>
<tr>
<td>Personal exchange</td>
<td>4.323</td>
<td>0.710</td>
<td>Rating from 1 to 5</td>
<td>Survey with investment managers</td>
</tr>
</tbody>
</table>

**Management support**

<table>
<thead>
<tr>
<th>Support with competence</th>
<th>3.554</th>
<th>0.881</th>
<th>Rating from 1 to 5</th>
<th>Survey with investment managers</th>
</tr>
</thead>
<tbody>
<tr>
<td>Support with sales</td>
<td>2.681</td>
<td>0.987</td>
<td>Rating from 1 to 5</td>
<td>Survey with investment managers</td>
</tr>
<tr>
<td>Support with technology</td>
<td>2.304</td>
<td>1.117</td>
<td>Rating from 1 to 5</td>
<td>Survey with investment managers</td>
</tr>
<tr>
<td>Support with strategy</td>
<td>3.839</td>
<td>0.664</td>
<td>Rating from 1 to 5</td>
<td>Survey with investment managers</td>
</tr>
<tr>
<td>Support with follow-up financing</td>
<td>4.065</td>
<td>1.046</td>
<td>Rating from 1 to 5</td>
<td>Survey with investment managers</td>
</tr>
<tr>
<td>Use of network</td>
<td>3.785</td>
<td>0.900</td>
<td>Rating from 1 to 5</td>
<td>Survey with investment managers</td>
</tr>
<tr>
<td>Industry</td>
<td>Metric</td>
<td>Business Plan</td>
<td></td>
<td></td>
</tr>
<tr>
<td>------------------------------</td>
<td>-------------------------</td>
<td>----------------------------------------------</td>
<td></td>
<td></td>
</tr>
<tr>
<td>IT</td>
<td>0.385</td>
<td>Investment committee papers</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Material Science</td>
<td>0.344</td>
<td>Investment committee papers</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Life Science</td>
<td>0.098</td>
<td>Investment committee papers</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Energy</td>
<td>0.057</td>
<td>Investment committee papers</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Telecommunication</td>
<td>0.041</td>
<td>Investment committee papers</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Other</td>
<td>0.090</td>
<td>Investment committee papers</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

**APPENDIX C:**

In table 6, the Fornell-Lacker criterion is shown. The table shows the latent variable correlation. In the diagonal the square root of the AVE can be found. This value should be higher than all values below and left in the table to pass the discriminant validity. This is the case for our model.

**Table 6. Fornell-Lacker Criterion**

<table>
<thead>
<tr>
<th></th>
<th>Failure risk</th>
<th>Assessment and evaluation</th>
<th>Contracting</th>
<th>Investment manager experience and skills</th>
<th>Governance</th>
<th>Management support</th>
</tr>
</thead>
<tbody>
<tr>
<td>Failure risk</td>
<td>0.735</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Assessment and evaluation</td>
<td>0.225</td>
<td>1.000</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Contracting</td>
<td>0.361</td>
<td>0.214</td>
<td>1.000</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Investment manager experience and skills</td>
<td>-0.183</td>
<td>0.130</td>
<td>-0.158</td>
<td>0.715</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Governance</td>
<td>0.201</td>
<td>0.298</td>
<td>0.435</td>
<td>-0.050</td>
<td>0.777</td>
<td></td>
</tr>
<tr>
<td>Management support</td>
<td>0.383</td>
<td>0.304</td>
<td>-0.416</td>
<td>0.197</td>
<td>0.641</td>
<td>1.000</td>
</tr>
</tbody>
</table>

In table 7, the cross loadings of each item in our PLS model are shown. Each variable should load highest on its corresponding construct. Then, the discriminant validity test is passed. This is the case in our model.
Table 7. Cross-Loadings

<table>
<thead>
<tr>
<th></th>
<th>Failure risk</th>
<th>Assessment and evaluation</th>
<th>Contracting</th>
<th>Investment manager experience and skills</th>
<th>Governance</th>
<th>Management support</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Failure risk</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>HR risk</td>
<td>0.624</td>
<td>0.006</td>
<td>0.019</td>
<td>-0.199</td>
<td>-0.126</td>
<td>0.046</td>
</tr>
<tr>
<td>Liquidity risk</td>
<td>0.775</td>
<td>0.105</td>
<td>0.338</td>
<td>-0.222</td>
<td>0.184</td>
<td>0.245</td>
</tr>
<tr>
<td>Market risk</td>
<td>0.701</td>
<td>0.297</td>
<td>0.353</td>
<td>0.071</td>
<td>0.241</td>
<td>0.339</td>
</tr>
<tr>
<td>Technology risk</td>
<td>0.823</td>
<td>0.212</td>
<td>0.259</td>
<td>-0.202</td>
<td>0.187</td>
<td>0.380</td>
</tr>
<tr>
<td><strong>Assessment and evaluation</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Evaluation</td>
<td>0.234</td>
<td>1.000</td>
<td>0.214</td>
<td>0.130</td>
<td>0.298</td>
<td>0.304</td>
</tr>
<tr>
<td><strong>Contracting</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Liquidation</td>
<td>0.361</td>
<td>0.214</td>
<td>1.000</td>
<td>-0.158</td>
<td>0.435</td>
<td>0.416</td>
</tr>
<tr>
<td><strong>Investment manager experience and skills</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Working experience</td>
<td>-0.137</td>
<td>0.177</td>
<td>-0.080</td>
<td>0.797</td>
<td>0.127</td>
<td>0.290</td>
</tr>
<tr>
<td>Founding experience</td>
<td>-0.159</td>
<td>0.078</td>
<td>-0.256</td>
<td>0.705</td>
<td>-0.292</td>
<td>-0.174</td>
</tr>
<tr>
<td>Network size</td>
<td>-0.110</td>
<td>-0.018</td>
<td>0.007</td>
<td>0.623</td>
<td>0.096</td>
<td>0.291</td>
</tr>
<tr>
<td>Technology expertise</td>
<td>-0.106</td>
<td>0.123</td>
<td>-0.068</td>
<td>0.722</td>
<td>-0.001</td>
<td>0.261</td>
</tr>
<tr>
<td><strong>Governance</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Milestones</td>
<td>0.227</td>
<td>0.161</td>
<td>0.414</td>
<td>-0.056</td>
<td>0.883</td>
<td>0.476</td>
</tr>
<tr>
<td>Information through network</td>
<td>0.098</td>
<td>0.349</td>
<td>0.181</td>
<td>0.015</td>
<td>0.663</td>
<td>0.577</td>
</tr>
<tr>
<td>Reporting</td>
<td>0.114</td>
<td>0.304</td>
<td>0.366</td>
<td>-0.055</td>
<td>0.770</td>
<td>0.541</td>
</tr>
<tr>
<td><strong>Management support</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Use of network</td>
<td>0.383</td>
<td>0.304</td>
<td>0.416</td>
<td>0.197</td>
<td>0.641</td>
<td>1.000</td>
</tr>
</tbody>
</table>
In table 8, the heterortrait-monotrait ratio of correlations (HTMT) is shown. If the HTMT is below 0.900 discriminant validity has been established between two constructs. This is the case for all of our items.

Table 8. Heterortrait-Monotrait Ratio Criterion

<table>
<thead>
<tr>
<th></th>
<th>Failure risk</th>
<th>Assessment and evaluation</th>
<th>Contracting</th>
<th>Investment manager experience and skills</th>
<th>Governance</th>
<th>Management support</th>
</tr>
</thead>
<tbody>
<tr>
<td>Failure risk</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Assessment and evaluation</td>
<td>0.249</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Contracting</td>
<td>0.390</td>
<td>0.214</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Investment manager experience and skills</td>
<td>0.332</td>
<td>0.168</td>
<td>0.174</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Governance</td>
<td>0.342</td>
<td>0.407</td>
<td>0.481</td>
<td>0.271</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Management support</td>
<td>0.407</td>
<td>0.304</td>
<td>0.416</td>
<td>0.430</td>
<td>0.797</td>
<td></td>
</tr>
</tbody>
</table>