

An Exploratory Study on determinants of Investment Behaviour of Venture Capitalist in Bangalore

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Abstract

This research has explored the nature of investment strategies by venture capitalists in Bangalore, mainly in terms of deal generation, geographic proximity, monitoring of deals, exit strategy, and evaluation of deals. It is basically to determine which are the prominent factors influencing their investment decisions along with their priorities. The adopted methodology for conducting this research work is exploratory, based on a sample population of 100 venture capitalists identified through convenience sampling. A structured questionnaire was used to collect data, and the responses were analysed on a five-point Likert scale. Multiple regression techniques were applied in SPSS to test the relationships among variables. The results indicate that deal evaluation and exit vehicles are the most important factors, whereas geographic proximity exerts the least influence, thus contesting conventional assumptions. Structured mechanisms of evaluation and robust exit strategies are the only means towards investment success. These findings have theoretical contributions to investment decision-making literature and practical implications toward venture capitalists, but a sample size and focus on a particular region bring about the need to carry out further research to make the generalizability better.

Keywords: Venture capitalist, Deal Generation, Multiple Regression, SPSS

1.Introduction

Venture capital (VC) is a cornerstone of entrepreneurial growth, offering financial support to innovative startups and high-growth businesses. Bangalore, often regarded as the "Silicon Valley of India," has emerged as a focal point for venture capital investments due to its robust startup ecosystem and technological innovation. Venture capitalists in Bangalore evaluate various factors when making investment decisions, such as the scalability of the business model, market potential, innovation, and the competence of the founding team. Their investment behaviour is crucial in shaping the success of startups in key sectors like technology, healthcare, fintech, and edtech. Understanding these behaviors provides valuable insights into the dynamics of venture capital funding, fostering economic and entrepreneurial growth (Gompers & Lerner, 2001; National Startup Report, 2023).



Bangalore's ecosystem thrives on a unique combination of global connectivity, established industries, and an entrepreneurial culture that attracts venture capital. Venture capitalists in this city often follow a strategic investment approach, balancing risk with potential returns and considering external factors like economic conditions, policy frameworks, and market trends. The interplay of these factors influences their investment behavior, which is pivotal for driving innovation and startup success. This study explores the motivations, challenges, and strategies of venture capitalists in Bangalore, aiming to shed light on their role in India's startup landscape (**Kaplan & Strömberg, 2004; Nasscom Startup Report, 2023**).

The conduct of VCs has a great impact on the developmental paths of startups, especially in rapidly evolving ecosystems like Bangalore. Although this area is recognized as a center for innovation and entrepreneurial activities, there exists a relative dearth of research that investigates the distinct investment trends, criteria for decision-making, and the obstacles encountered by VCs in this locale. These factors are crucial to understand because venture capital investment propels not only the growth of the economy but also innovation in the fields of technology, healthcare, and fintech (**Gompers & Lerner, 2001**). This study fills a significant gap in the literature that would otherwise be lacking, as it sheds light on the motivations and strategies of venture capitalists in Bangalore by providing real-world insights for entrepreneurs, policymakers, and investors.

Moreover, the study is necessary given the increasing competition in Bangalore's startup ecosystem, where VCs play a pivotal role in determining which businesses succeed. Insights into VC behavior can help startups tailor their pitches, align with investor expectations, and secure funding. At a broader level, the findings can inform policymakers about creating a supportive regulatory framework that enhances VC activity and startup growth. Considering Bangalore's status as a leading hub for startups in India, it is vital to understand the venture capital environment that will support its innovation-driven economy (Kaplan & Strömberg, 2004; Nasscom Startup Report, 2023).

2.Literature Review

Li et al. (2019): This paper examines the decision-making criteria of venture capitalists (VCs) in emerging markets. It focuses on scalability and innovation potential. Based on survey-based data from 120 VCs, the study concludes that a startup's growth trajectory and its potential to disrupt the existing market are major factors determining investment decisions. The authors further concluded that macroeconomic stability significantly affects the risk appetite of VCs.



Sharma and Gupta (2020): Sharma and Gupta conducted an in-depth interview with 50 VCs to gauge the competency of the founding team in VC investment. Based on the study, leadership skills, vision, and team cohesiveness were critical determinants that decided funding. The findings conclude that startups with robust leadership are likely to garner confidence from VCs, and hence, there is a positive association with the outcomes of investments.

Jain et al. 2020: This paper analyzes the effects of digital technology on the process of funding venture capital. By surveying 200 VCs, this article established that the application of digital tools has made due diligence more efficient because the financial and operational information is more accessible. Thus, digitalization facilitated the accessibility of financing for startups and accelerated decisions regarding funding for venture capitalists.

Chaturvedi et al. (2021): Chaturvedi carried out case studies on the sectoral biases of venture capitalists based on investments in India. Findings: the technology and health sectors showed significant growth and attracted venture capital with less volatility, hence they found a large percentage of investment from venture capitalists. It also states that sector-specific knowledge was essential to make decisions. Kapoor and Singh (2019):

This qualitative study investigated the motives and risk assessment practices of VCs in high-risk sectors such as fintech. Case study analysis led the researchers to the conclusion that VCs like to invest in startups that show a clear risk mitigation strategy. Market intelligence and previous entrepreneurial success were cited as significant influencers in investment decisions.

Wu et al. (2021):Wu used the quantitative modelling approach to understand how economic policies affect the investment decisions of venture capital firms. The research findings suggest that favourable government policies in terms of tax incentives and leniency in regulation have significantly increased VC participation. The authors found that such policies reduce perceived risk factors, making investments in new ventures relatively attractive. Raj et al. (2022):

This paper focuses on the issue of market saturation in mature sectors and its implications on VC investment behaviour. Based on secondary data from VC funding reports, the researchers deduced that investors were looking at new emerging fields such as green technologies and renewable energy. According to the study, VCs now identify sustainable investment opportunities more in advance.



Patel and Roy (2023): Patel and Roy employed a mixed-methods approach to examine how predictive analytics and AI impacts the investment of venture capital. In conclusion, from the findings by Patel and Roy, data-informed decision-making seems to be integral when making startup performance decisions. Prediction models performed better in gauging prospects of long-run growth and reducing subjective investment decisions. Khan et al. (2021):

This paper has thrown light on the geographical preference of VCs and why cities like Bangalore attract more investments. This study concludes that the availability of skilled labour, digital infrastructure, and networking are the major factors affecting the geographical preferences of VCs. According to Khan et al., urban ecosystems are conducive to the growth of startups; hence, they are highly preferred by venture capitalists.

Verma et al. (2022): Verma analysed venture capital investment behaviour post-pandemic and how global trends affect it. The study found that VCs became more cautious, considering startups with proven revenue models and resilience to economic shocks. Moreover, there is a greater preference for technology and health sectors and startups thereof, as is being seen during the post-pandemic era and due to a surge in demand for digital and healthcare solutions

3. Conceptual Framework: Venture Capitalists' Investment Practices

Various theoretical frameworks guide the decision-making strategies of venture capitalists in terms of investment practices. This paper explores the driving factors of venture capital investments, including deal origination, local proximity, monitoring of deals, exit strategies, and evaluation of deals. These driving factors are construed from various perspectives, oriented specifically around investment decision-making, risk management, and value creation in early-stage companies. The subsequent sections delve into each factor within the framework, drawing on relevant theoretical underpinning:

Deal Generation: This term refers to the process used by venture capitalists to identify potential sources of investment. According to the resource-based view (**Barney**, **1991**), venture capitalists leverage networks, industry-specific knowledge, and reputation to cultivate high-quality investments. The importance of deal flow in determining successful venture capital investing lies in that a larger, more diverse pipeline of deals raises the likelihood of finding high-value opportunities (**Gompers & Lerner**, **2001**). The more involved the VCs are with the networks and collaborations, the better the chances of finding innovative startups.



Geographic Proximity

Such geographical proximity between organizations affects investment behavior, as postulated in the social capital theory by **Coleman (1988)**, which holds that distance reduces trust, hinders communication and the transmission of central information. Venture capitalists tend to invest in companies within areas they are well conversant with since they understand the dynamics of the local market, cultural nuances, and established business networks. The proximity of such geography diminishes information asymmetry, hence lowering perceived risks and allowing for easier decision making (**Lockett et al., 2002**).

Deal Monitoring

Deal monitoring is the management and control of investments post-funding. Agency theory (Jensen & Meckling, 1976) is relevant in this regard as it describes the possibility of agency issues between the principals, which are the investors, and the agents, which are the entrepreneurs. VCs engage in active monitoring of the startup on its activities along the set goals. Effective monitoring minimizes risks and ensures responsibility; it also enhances the potential for successful exits (Sapienza et al., 1996).

Exit Vehicle: Exit strategies are important in determining the return on investment for venture capitalists. According to liquidity theory (**Holmström & Tirole, 1993**), venture capitalists prefer exits that provide immediate and significant returns, such as IPOs or acquisitions. The choice of an exit vehicle is influenced by factors such as prevailing market conditions, the developmental stage of the startup, and the investment phase. An effective exit strategy will ensure the VC can reap the value created by their investment, balancing risk and reward (**Cumming & MacIntosh, 2003**).

Deal Evaluation

Deal evaluation involves assessing possible investments based on a number of criteria, which may include market potential, capabilities of the startup team, and financial projections. It follows then that signalling theory, as coined by **Spence (1973)**, is relevant for understanding how VC firms determine a level of quality through observable signals concerning the experience level of the startup founders, startup past performance, or industry-specific internal endorsements. At times, investments are based solely on the level of positive signalling that suggests successful prospects, as reported in **Chahine & Filatotchev (2008)**.



4.Objective of the Study

1. To explore the investment behaviour of venture capitalists in Bangalore, focusing on key factors such as deal generation, geographic proximity, deal monitoring, exit vehicles, and deal evaluation.

2. To identify the key predictors that significantly affect venture capital investment decisions in the context of Bangalore.

3. To provide insights and recommendations for improving venture capital investment strategies, particularly in the Bangalore startup ecosystem.

5.Hypotheis of the Study

H1: Deal generation has a significant positive effect on investment behaviour.

H2: Geographic proximity has a significant positive effect on investment behaviour.

H3: Deal monitoring has a significant positive effect on investment behaviour.

H4: Exit vehicle has a significant positive effect on investment behaviour.

H5: Deal evaluation has a significant positive effect on investment behaviour.

6.Research Methodology

This research used an exploratory design to determine the investment behavior of the venture capitalists in Bangalore. The study used convenience sampling to generate a sample size of 100 venture capitalists from the pool available, selected according to their ease of access and willingness to take part in the study. Data is collected from a structured questionnaire with a five-point Likert scale, mainly designed to observe various factors responsible for investment decision-making, for example, the generation of a deal, deal monitoring, an exit strategy, geographical proximity, and deal evaluation. The sample size is considered quite adequate for this kind of study, which should be able to carry out more generalizable investigations. A multiple regression analysis examines the relationship of the independent variables with dependent variables (investment decisions). The data is analyzed and insights are derived using SPSS software to identify the key predictors of investment decisions by venture capitalists.

7.Data Analysis and Interpretation

Descriptive statistics of the results suggest that respondents normally regard all variables that affect investment behavior to have a moderate level of significance. Mean values of investment behavior stand at 2.930, deal generation 3.085, geographic proximity 3.000, deal monitoring 3.245, exit vehicle 3.150, and deal evaluation 3.205. All of them are slightly above the midpoint, thus having a positive but not very positive view of their relevance. Standard deviations vary between 0.8784 and 1.1253, thereby indicating that variations in responses were moderate mainly concerning the geography aspect and the evaluation of the deal. This would suggest that though people are broadly on the same side about these issues, differences on the level of value given to these aspects are rather sizeable by respondents.

Table 1: Descriptive Statistics of Investment Behaviour

	Mean	Std. Deviation	Ν
Investment Behaviour	2.930	0.9538	100
Deal Generation	3.085	0.8784	100
Geographic proximity	3.000	1.1253	100
Deal Monitoring	3.245	0.9848	100
Exit Vehicle	3.150	0.9283	100
Deal Evaluation	3.205	1.0811	100

The multicollinearity statistics in Table 2 show the relationships between the independent variables and their potential for multicollinearity in the regression model. The tolerance values range from 0.596 to 0.960, and the VIF values range from 1.042 to 1.679. Values of tolerance greater than 0.1 and VIF less than 10 indicate that there is no multicollinearity among the independent variables, hence the variables are not highly correlated with each other. In fact, the VIF of Deal Generation is 1.325, and the VIF of Geographic Proximity is 1.124, and the VIF of Deal Monitoring is 1.042, all of which are low enough to signify very little correlation. The VIFs for Exit Vehicle (1.679) and Deal Evaluation (1.600) are also high but not over the cut-off value. Therefore, the conclusion is drawn that multicollinearity does not pose any significant problem in this evaluation.

Table 2: Multicollinearity

	Collinearity Statistics		
Independent Variables	Tolerance	VIF	
Deal Generation	0.755	1.325	
Geographic proximity	0.890	1.124	
Deal Monitoring	0.960	1.042	

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Exit Vehicle	0.596	1.679	
Deal Evaluation	0.625	1.600	

Table 3 Summary of the model used in the regression analysis An R value of 0.729 indicates a very strong positive correlation between the independent variables and the dependent variable. This means that the model has a strong predictive relationship. The R Square value of 0.532 means that approximately 53.2% of the variation in the dependent variable is explained by the independent variables included in the model. Overall, the model appears to explain a significant portion of the variance in the dependent variable, with a reasonable degree of prediction accuracy.

Table 3: Model Summary

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Model	R	R Square	Adjusted R Square	Std. Error of the Estimate
1	.729 ^a	0.532	0.520	0.6608

The ANOVA table reveals that the regression model is statistically significant as indicated by a large F-value of 44.106. An F-value represents the ratio of the variance explained by the model to the residual variance, and thus a large value indicates that independent variables have a lot to say about the variability of the dependent variable. The degrees of freedom for the regression are 5, which is the number of predictors, and the residual degrees of freedom is 194, which is the remaining unexplained variance. The p-value of 0.000 is less than the common significance threshold of 0.05, meaning that the model is statistically significant, and the null hypothesis can be rejected, which states no relationship between the independent and dependent variables. This implies that the independent variables in the model have a high influence on the dependent variable.

Table 4: ANOVA for Investment Behaviour

ANOVA							
		Sum of		Mean			
Model		Squares	df	Square	F	Sig.	
1	Regression	96.302	5	19.260	44.106	.000	
	Residual	84.718	194	0.437			
	Total	181.020	199				

Table 5 below shows the Beta coefficients that describe the standardized relationship of each independent variable to the dependent variable. For instance, Deal Evaluation has the highest value of Beta, which stands at 0.439; this means it has the strongest positive influence on the dependent variable, while at the same time,



Geographic Proximity has a value of 0.005, meaning an extremely poor correlation. The t-value is an indicator of the significance of each predictor. For instance, Deal Evaluation has a t-value of 7.059, which is quite high, meaning that the relationship is strong and significant. In contrast, Geographic Proximity has a t-value of 0.103, which means that its effect is not statistically significant. The p-value is used to determine statistical significance; a p-value less than 0.05 typically means that the predictor is significant. For instance, Exit Vehicle shows a p-value of 0.000; hence it is highly significant while, on the other hand, Geographic Proximity has a p-value of 0.918, meaning that it is not significant enough to impact the dependent variable. In the same manner, Deal Monitoring indicates a p-value of 0.029, making it statistically significant; Deal Generation also shows a p-value of 0.043, which also is significant.

		Unstand Coeff	dardized icients	Standardized Coefficients		
Model		В	Std. Error	Beta	t	Sig.
1	(Constant)	0.048	0.244		0.198	0.843
	Deal	0.125	0.061	0.115	2.036	0.043
	Generation					
	Geographic	0.005	0.044	0.005	0.103	0.918
	proximity					
	Deal	0.107	0.049	0.111	2.205	0.029
	Monitoring					
	Exit	0.284	0.065	0.277	4.347	0.000
	Vehicle					
	Deal	0.387	0.055	0.439	7.059	0.000
	Evaluation					

 Table 5: Coefficient Table

8.Conclusion

This research has been conducted based on the investments of venture capitalists in Bangalore regarding the deal generation, geographic closeness, deal oversight, exit strategies, and deal assessment. Findings are that while geographic closeness does not influence the investment practices much, other factors such as deal assessment, exit strategies, deal oversight, and deal generation do have a strong influence, and the strongest among them is deal assessment. These observations highlight the complex and varied aspects inherent in investment decision-making and underscore the importance of systematic assessment frameworks along with proficient exit strategies. Understanding these factors helps venture capitalists refine their approaches to decision-making and improve their overall performance in investments.



The research enriches the literature on investment behaviour by including and validating aspects such as deal evaluation and exit strategy in the context of venture capital. It supports the hypothesis that successful deal evaluation processes are critical to achieving success in investments, consistent with previous studies on decision-making models in venture capitalism. Moreover, the importance of exit strategies to profitability and liquidity is emphasized within the study-a much more elaborate understanding of their impact on the decision to invest. On a related note, the relatively minor role of geographic proximity issues defies conventional wisdom, suggesting that technological changes may have diminished its relevance, and theoretical exploration is also worthwhile.

Despite its worth, this study has some shortcomings. First of all, it was based on a sample of 100 venture capitalists located in Bangalore; thus, this limits the generalizability of the study to other areas or industries. Besides, the cross-sectional design of this study does not capture the time-varying features of investment behavior. Although the study highlights key factors determining investment behavior, it has ignored the other contextual determinants like market trends, industry-specific risks, or government policies. Future studies can overcome the limitations by making use of a longitudinal approach and expanding the scope geographically. They can include some additional factors as well to develop a better understanding of the venture capitalist's behaviour.

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