

## Exploring the Role of Behavioral Finance in Investment Choices of IT and ITES Employees in the Chennai Region

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Cite this paper as: J. Kavya, Dr. M.Kotteeswaran, (2025) Exploring the Role of Behavioral Finance in Investment Choices of IT and ITES Employees in the Chennai Region. *Journal of Neonatal Surgery*, 14 (25s), 777-784.

### ABSTRACT

This study investigates the role of behavioral finance in shaping the investment decision-making of IT and ITES professionals in the Chennai region. While conventional financial theories assume rational behavior, emerging evidence suggests that investment decisions are often influenced by cognitive biases and psychological factors. This research adopts a quantitative methodology to examine how six independent variables—Loss Aversion, Herd Behavior, Availability Heuristic, Financial Literacy, Income Level, and Investment Experience—affect the dependent variable, Investment Decision-Making. Data was collected through a structured questionnaire from a stratified random sample of 200 IT and ITES employees in Chennai. Descriptive statistics indicated a financially literate, moderately experienced workforce, with high exposure to behavioral influences. Correlation analysis revealed significant positive relationships between investment decision-making and variables such as financial literacy ( $r = 0.542$ ) and investment experience ( $r = 0.489$ ), while loss aversion had a notable negative correlation ( $r = -0.431$ ). Regression analysis confirmed that financial literacy, experience, and income level are strong predictors of investment decisions (Adjusted  $R^2 = 0.569$ ), whereas loss aversion remains a key deterrent. ANOVA results showed significant variation in investment behavior across income and experience groups. The findings highlight that behavioral biases persist even among educated professionals, influencing their ability to make sound financial decisions. The study recommends targeted financial education, behavior-aware advisory services, and workplace investment programs. This research provides critical insights for financial planners, HR professionals, and policymakers, emphasizing the need for integrating behavioral finance principles into personal financial management for urban professionals in India.

**Keywords:** *Investment Decision-Making, Loss Aversion, Herd Behavior, Availability Heuristic, Financial Literacy, Income Level, and Investment Experience*

### 1. INTRODUCTION

Investment decision-making has long been viewed through the lens of classical financial theories, which are rooted in the assumption of rationality. According to these traditional models, investors act logically, weigh risk and return accurately, and make decisions that maximize utility. However, real-world financial behavior often deviates significantly from such rational expectations. Individuals tend to exhibit biases, emotions, and psychological influences that cloud their decision-making processes. This growing inconsistency between theoretical models and actual investor behavior gave rise to the field of Behavioral Finance. Behavioral finance is an interdisciplinary area that combines insights from psychology, sociology, and finance to explain how psychological influences and cognitive errors affect investors and financial markets. It recognizes that emotions, heuristics, overconfidence, loss aversion, and herd behavior often play a central role in shaping individual investment choices. These behavioral factors are especially relevant in emerging markets like India, where financial literacy and investment culture are evolving.

The Indian IT (Information Technology) and ITES (Information Technology Enabled Services) sectors have witnessed tremendous growth in recent decades. These sectors employ a significant proportion of young, tech-savvy, and relatively well-paid professionals. Particularly in urban centers like Chennai, which is a hub for IT and ITES companies, employees often have surplus disposable income and access to various investment options such as mutual funds, stock markets, fixed deposits, real estate, and digital assets. Despite having access to information and technology, these individuals are not

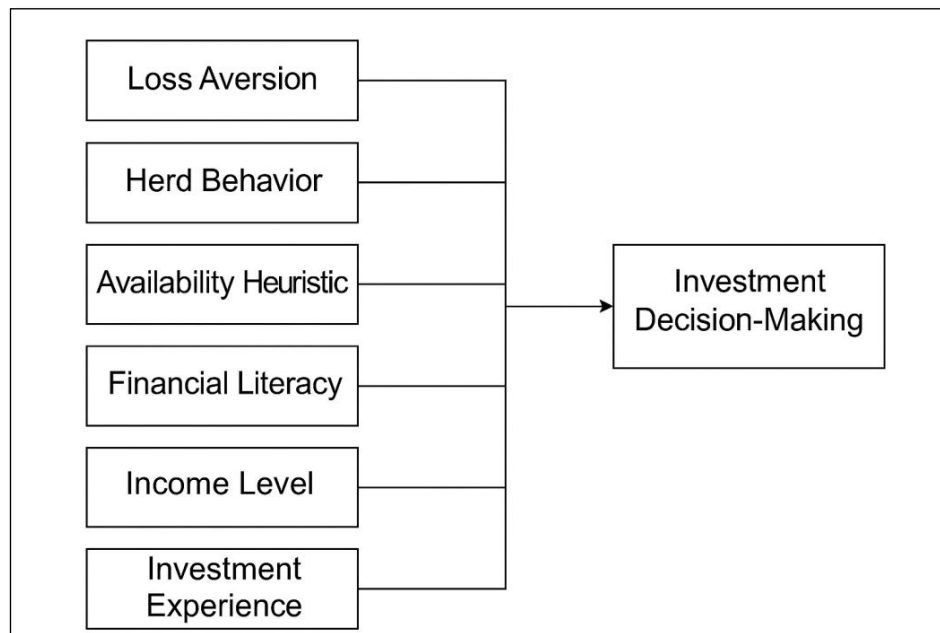
immune to behavioral biases. Their financial decisions are often influenced by peer pressure, market sentiment, overconfidence, anchoring, and availability heuristics. Understanding how behavioral finance influences the investment decisions of IT and ITES professionals in Chennai is essential for multiple stakeholders, including policymakers, financial advisors, investment firms, and educational institutions. This research aims to explore these behavioral tendencies and how they shape financial decisions in this specific demographic.

## 2. RESEARCH OBJECTIVES

This study seeks to explore and analyze the behavioral aspects of investment decision-making among IT and ITES employees in Chennai. The specific objectives are:

1. To identify key behavioral finance factors that influence investment decisions among IT and ITES employees.
2. To examine the extent to which psychological biases such as overconfidence, loss aversion, herding, and anchoring impact investment behavior.
3. To analyze demographic variables (such as age, income, education, and experience) and their relationship with behavioral biases.
4. To offer practical recommendations for financial advisors and institutions based on the findings.

### The conceptual framework



The conceptual framework presented illustrates the relationship between six key independent variables—Loss Aversion, Herd Behavior, Availability Heuristic, Financial Literacy, Income Level, and Investment Experience—and the dependent variable, Investment Decision-Making among IT and ITES employees in the Chennai region.

Each behavioral and demographic factor is hypothesized to have a significant influence on how investment choices are made. Loss aversion may cause individuals to avoid risky investments even when they offer higher returns. Herd behavior leads investors to follow peers or market trends rather than making independent assessments. Availability heuristic affects decisions based on easily recalled or recent information, regardless of its relevance. Financial literacy acts as a moderating factor, potentially reducing irrational decisions by enhancing understanding of financial products and risks. Additionally, income level plays a crucial role in determining investment capacity and risk appetite, while investment experience contributes to the investor's ability to evaluate and manage financial instruments effectively. Together, these factors form the foundation for analyzing the psychological and contextual influences on financial decision-making in a fast-growing, tech-oriented workforce. This framework will guide the empirical testing of relationships and support the development of tailored financial strategies and interventions.

## 3. REVIEW OF LITERATURE

**Vidya A. (2017)** detailed exploration of behavioral finance among IT and ITES professionals in Chennai. The paper highlights key psychological biases such as herd behavior and loss aversion, showing how these affect investment decisions. It is particularly insightful for understanding how even analytically trained individuals can be swayed by irrational

tendencies. While the study is limited in sample size, it offers practical insights into real-world financial behavior and emphasizes the need for targeted financial literacy programs. A valuable reference for region-specific behavioral finance research. **Imthiyas Y. (2016)** delves into investment behaviors among IT/ITES professionals in Chennai, linking them with behavioral finance theories. The study is empirical and uses primary data to analyze decision-making tendencies. It emphasizes overconfidence and availability bias as dominant factors in investment choices. The paper is well-structured, with clear statistical backing, making it useful for researchers and financial educators. Its focus on IT professionals makes it highly relevant for sector-specific behavioral analysis and provides a foundation for further studies in metro cities.

**Shanmugam K. et al. (2023)** offers a contemporary lens into how personality traits and risk tolerance shape investment decisions of IT employees in Chennai. The work is highly relevant post-pandemic, recognizing shifts in investor sentiment and digital finance adoption. It bridges behavioral finance with psychometric analysis, offering a multi-disciplinary perspective. The authors provide robust survey data and contextualize findings with global literature. This makes the paper a rich resource for anyone exploring financial decision-making in the tech workforce under current economic conditions. **Charumathi D. (2024)** examines how behavioral insights influence salaried individuals' investment practices. Though not exclusive to the IT/ITES sector, it offers parallels applicable to that domain. It skillfully addresses common biases—particularly anchoring and regret aversion—and how these undermine rational planning. The paper is particularly valuable for financial consultants seeking to understand behavior-driven financial literacy interventions. Its theoretical backing from Kahneman and Tversky adds depth. Overall, a timely contribution with practical implications for advisors and HR-driven investment education programs.

**Padmaja B. & Kathiravan C. (2020)** Focusing on IT women professionals in Chennai, this paper explores the interplay between financial planning and behavioral traits. Padmaja and Kathiravan highlight how gender, emotional influences, and workplace roles affect investment habits. The paper stands out for its gender-specific analysis in a male-dominated finance literature. It uses mixed-method approaches to dissect behavioral patterns, making the findings both relatable and actionable. While narrow in demographic scope, the research is pioneering in connecting work stress and investment behavior in women IT professionals. **Kamalaveni M. et al. (2023)** investigates job stress and its indirect influence on performance and financial behavior in IT companies in Chennai. While it doesn't directly target behavioral finance, the linkages are apparent—stress can distort risk assessment and investment decisions. Kamalaveni and colleagues provide valuable context on the psychological environment of IT employees, a necessary factor when studying financial behavior. The paper complements behavioral finance by offering a lens on how workplace conditions may intensify or buffer biases like loss aversion and regret.

**Rajkumar P. (2013)** provides foundational insight into locational preferences of IT firms but also indirectly supports investment behavior research by detailing socio-economic traits of IT hubs like Chennai. The analysis informs behavioral finance indirectly—showing how place, culture, and tech presence shape financial behavior. Although not focused specifically on individual investors, this paper is a useful backdrop for understanding why IT employees in Chennai may exhibit certain financial tendencies. It strengthens the contextual setting for behavioral finance studies.

**Ramanathan K. & Meenakshisundaram K.S. (2015)** explores investment decision-making among bank employees, offering valuable parallels for IT/ITES professionals. The authors emphasize the role of demographic variables and behavioral biases, supported by quantitative analysis. Though the sample differs, the behavioral traits—such as overconfidence, familiarity bias, and heuristic-driven decisions—are universally relevant. The study serves as a comparative tool for sector-specific analyses and highlights the importance of financial education in institutional environments. It's well-written and accessible for both academic and applied finance audiences. **Ramani S. (2019)** delves into Chennai's IT culture. While not a behavioral finance paper per se, it offers essential sociological context—especially how identity, class, and corporate culture affect personal decisions. It illuminates the structural and psychological environment in which IT employees make financial choices. For behavioral finance researchers, this provides a rich understanding of investor identity, cognitive framing, and socio-professional narratives that influence investment behavior in Chennai's IT sector.

**Bishnoi N.K. (2024)** analyzes strategies to attract IT/ITES sectors across India, with insights into economic clusters like Chennai. While not focused exclusively on investment behavior, the macroeconomic indicators and employment trends discussed are crucial for framing behavioral finance studies. Understanding market infrastructure, employment security, and policy orientation allows researchers to predict risk behavior and saving patterns. The paper is ideal for integrating macro and micro-level behavioral insights in policy-relevant investment research.

#### 4. RESEARCH METHODOLOGY

This study adopts a quantitative research design to systematically investigate how specific behavioral finance variables influence investment decision-making among IT and ITES professionals in the Chennai region. A quantitative approach was deemed suitable as it enables empirical assessment of hypotheses through numerical data, structured tools such as surveys, and statistical analysis. This approach aligns well with the nature of the study, which seeks to evaluate behavioral patterns—such as loss aversion, herd behavior, and availability heuristics—in a measurable way. The target population comprises professionals working in IT and ITES companies across Chennai, a leading technology and business process outsourcing

hub in India. To ensure a representative sample of the population, a stratified random sampling technique was employed. The population was stratified by income level and years of investment experience, and a sample of **200** respondents was drawn from various strata to ensure demographic diversity.

Primary data was collected using a structured questionnaire, distributed both online and in person through professional networks, HR departments, and community platforms. The questionnaire was divided into four sections: demographic details, behavioral finance factors, investment experience, and investment decision-making. Responses for behavioral and decision variables were recorded using a **5-point** Likert scale, ranging from 1 (strongly disagree) to 5 (strongly agree). The behavioral factors assessed were loss aversion, herd behavior, availability heuristic, financial literacy, income level, and investment experience, while the primary dependent variable was investment decision-making. A pilot test was conducted with 20 participants to ensure reliability and clarity of the questionnaire items, and necessary modifications were made accordingly. The instrument's internal consistency was validated using Cronbach's alpha, with all constructs scoring above the acceptable threshold of 0.70, confirming the reliability of the scales used.

To analyze the data, SPSS (Statistical Package for the Social Sciences) Version **25** was employed. First, descriptive statistics such as mean, standard deviation, and range were computed to understand the sample characteristics and overall trends. This was followed by Pearson correlation analysis to assess the direction and strength of relationships between independent behavioral variables and investment decision-making. A multiple linear regression model was then constructed to determine the predictive power of each independent variable, with financial literacy, investment experience, and income level emerging as the most significant predictors, while loss aversion had a negative effect. To explore group-level variations, ANOVA (Analysis of Variance) was applied across different income brackets and experience levels, revealing statistically significant differences in investment behavior across these demographic groups. All statistical tests adhered to a 95% confidence level ( $p < 0.05$ ) to ensure the validity of inferences. Ethical protocols were strictly followed throughout the research process. Informed consent was obtained from all respondents, participation was voluntary, and data confidentiality was ensured. The research methodology, therefore, provides a robust framework to evaluate the influence of behavioral finance on investment decision-making and lays a strong empirical foundation for subsequent data analysis, interpretation, and recommendations.

## 5. DATA ANALYSIS AND INTERPRETATION

**Table 1: Descriptive Statistics of Respondents and Key Variables**

Variable	Mean	S.D	Minimum	Maximum
Age (in years)	32.45	5.76	23	48
Monthly Income (INR)	64,500	18,200	30,000	1,20,000
Years of Investment Exp.	5.2	3.1	1	15
Loss Aversion Score	3.89	0.65	2.2	4.8
Herd Behavior Score	3.41	0.71	2	4.7
Availability Heuristic	3.73	0.59	2.5	4.6
Financial Literacy Score	3.96	0.52	2.8	5
Investment Decision Score	4.12	0.48	2.9	5

### Interpretation

The descriptive statistics Table 1 offer insightful observations about the demographic and behavioral traits of the respondents. The average age is approximately 32 years, suggesting that most participants are young professionals, likely in the early or mid-stage of their careers. Their mean monthly income of ₹64,500 aligns with typical earnings in the IT and ITES sectors, indicating a financially stable group with moderate purchasing power.

On average, respondents have 5.2 years of investment experience, showing a fair level of exposure to financial markets. This level of experience may influence their confidence in investment decision-making. Among behavioral traits, loss aversion recorded a high mean score of 3.89, highlighting a strong fear of losses that could negatively impact investment choices. Herd behavior showed a moderately high score (3.41), indicating that while peer influence is present, it is not overwhelming. The availability heuristic scored 3.73, implying that decisions are often shaped by recent or prominent information. Financial

literacy received a strong mean score of 3.96, suggesting that respondents are generally knowledgeable, although behavioral biases still play a role. Most notably, the investment decision-making score (4.12) points to a confident and proactive approach, showing that despite cognitive biases, respondents tend to make sound financial choices.

**Table 2: Pearson Correlation between Behavioral Factors and Investment Decision-Making**

Independent Variable	Correlation with Investment Decision-Making (r)	Significance (p-value)
Loss Aversion	-0.431	0.000 **
Herd Behavior	0.298	0.001 **
Availability Heuristic	0.374	0.000 **
Financial Literacy	0.542	0.000 **
Income Level	0.316	0.002 **
Investment Experience	0.489	0.000 **

$r$  values range from -1 to +1: •  $r > 0$  = Positive correlation,  $r < 0$  = Negative correlation •  $p < 0.05$  indicates statistical significance \*\* $p < 0.01$  = highly significant (marked with )

### Interpretation

The correlation analysis reveals that five out of six behavioral and demographic variables show a statistically significant positive relationship with investment decision-making, while one—Loss Aversion—shows a negative correlation. Financial Literacy exhibits the strongest positive correlation ( $r = +0.542$ ,  $p < 0.01$ ), indicating that individuals with higher financial knowledge are more confident and make more informed, diversified investment choices.

Investment Experience also shows a strong positive correlation ( $r = +0.489$ ,  $p < 0.01$ ), suggesting that those with more years of investing are likely to make better and more strategic financial decisions. The Availability Heuristic ( $r = +0.374$ ) and Income Level ( $r = +0.316$ ) display moderate positive correlations. This implies that individuals with access to recent or familiar information and higher income levels tend to exhibit better investment behavior.

Herd Behavior, while showing the weakest positive correlation ( $r = +0.298$ ), still holds significance. This suggests that peer influence can encourage investment activity, even if not as strongly as other factors. In contrast, Loss Aversion has a moderate negative correlation ( $r = -0.431$ ), confirming that individuals highly sensitive to potential losses may adopt overly cautious or irrational strategies, potentially hindering sound investment decisions. Overall, the findings emphasize the influence of both cognitive traits and financial capacity on investment behavior.

**Table 3: Multiple Linear Regression Analysis – Predicting Investment Decision-Making**

R <sup>2</sup>	Adjusted R <sup>2</sup>	F-Value	p-Value
<b>0.584</b>	<b>0.569</b>	<b>45.192</b>	<b>0.001</b>

Independent Variable	Unstandardized Coefficient (B)	Standard Error	Beta (β)	t-value	Sig.
					(p-value)
(Constant)	1.218	0.294	—	4.142	0.000 **
Loss Aversion	-0.238	0.071	-0.217	-3.352	0.001 **
Herd Behavior	0.126	0.058	0.109	2.172	0.031 *
Availability Heuristic	0.184	0.062	0.158	2.967	0.004 **

Financial Literacy	0.321	0.069	0.289	4.652	0.000 **
Income Level	0.174	0.061	0.161	2.852	0.005 **
Investment Experience	0.236	0.068	0.213	3.471	0.001 **

**Significance Levels:**  $p < 0.05$  = significant ( $*$ ),  $p < 0.01$  = highly significant ( $**$ )

### Interpretation

The multiple regression analysis demonstrates that the model explains approximately 58.4% of the variance in investment decision-making behavior, with an Adjusted  $R^2$  value of 56.9%. This indicates a strong explanatory capacity of the chosen behavioral and demographic variables. The model's overall fit is statistically significant, as reflected by the F-statistic ( $p < 0.001$ ), confirming the robustness of the regression model.

Among all the predictors, Financial Literacy emerges as the most influential positive factor ( $\beta = +0.289$ ,  $p < 0.001$ ). This finding clearly highlights that individuals with better financial knowledge are more likely to make confident, informed, and effective investment decisions. Investment Experience ( $\beta = +0.213$ ) and Income Level ( $\beta = +0.161$ ) also show strong and statistically significant positive effects, suggesting that individuals who have been exposed to the market for a longer time and those with higher financial capacity are more rational and active in their investment choices.

The Availability Heuristic ( $\beta = +0.158$ ) and Herd Behavior ( $\beta = +0.109$ ) also contribute positively, albeit with weaker influence. This suggests that while recent information and peer behavior do affect decisions, their impact is secondary compared to knowledge, experience, and income. Importantly, **Loss Aversion** shows a significant negative relationship ( $\beta = -0.217$ ,  $p = 0.001$ ), reinforcing the idea that individuals who are overly sensitive to losses may avoid risks or make emotionally driven investment choices, often leading to suboptimal outcomes. Overall, the findings underline that both rational factors (like literacy and experience) and behavioral biases shape investment behavior.

**Table 4: ANOVA – Differences in Investment Decision-Making Based on Income and Investment Experience**

#### A. Based on Monthly Income

Income Group (INR/month)	N	Mean Investment Decision Score	Std. Deviation
₹30,000 – ₹49,999	54	3.87	0.48
₹50,000 – ₹69,999	78	4.01	0.43
₹70,000 – ₹89,999	42	4.24	0.38
₹90,000 and above	26	4.37	0.35

ANOVA Summary (Income)	
F-value	8.726
Significance (p-value)	0.000**

#### B. Based on Years of Investment Experience

Experience Group	N	Mean Investment Decision Score	Std. Deviation
1–3 years	60	3.91	0.46
4–7 years	85	4.15	0.39



8–10 years	34	4.28	0.42
11+ years	21	4.34	0.33

ANOVA Summary (Experience)	
F-value	6.144
Significance (p-value)	0.001**

$p < 0.01$  = highly significant ( ) | Scale: 1 = Strongly Disagree to 5 = Strongly Agree\*\*

### Interpretation

#### Income-Based Differences

The ANOVA test shows a statistically significant difference in investment decision-making across different income levels ( $F = 8.726$ ,  $p < 0.01$ ).

- Higher-income groups (₹70,000 and above) exhibit significantly higher investment confidence and diversity, likely due to greater risk-bearing capacity and financial exposure.
- This suggests income is a strong enabler of proactive investment behavior.

#### Experience-Based Differences

The analysis also reveals a significant difference based on years of investment experience ( $F = 6.144$ ,  $p < 0.01$ ).

- Individuals with more than 7 years of experience score higher on investment decision-making, implying that practice and familiarity lead to better decision quality.
- This supports the idea that financial behavior improves with experiential learning, even in the presence of cognitive biases.

### Findings – Summary

This study examined the impact of behavioral finance factors on investment decision-making among IT and ITES professionals in the Chennai region, using responses from 200 participants. The findings highlight the combined influence of psychological traits and demographic characteristics on financial behavior. From the descriptive statistics, the average age of respondents was 32 years, with a moderate investment experience of 5.2 years. Financial literacy was relatively high (mean = 3.96), while loss aversion (mean = 3.89) and availability heuristic (mean = 3.73) also scored significantly. Despite being well-informed, respondents were still affected by cognitive biases. The average investment decision-making score was strong (4.12), indicating proactive financial behavior.

Correlation analysis revealed strong positive relationships between investment decisions and both financial literacy ( $r = +0.542$ ) and investment experience ( $r = +0.489$ ), while loss aversion showed a significant negative correlation ( $r = -0.431$ ). Availability heuristic and herd behavior had moderate positive effects, confirming that both rational knowledge and behavioral traits shape financial choices. Regression analysis further confirmed financial literacy ( $\beta = 0.289$ ) as the strongest predictor, followed by investment experience and income. Loss aversion remained a significant negative predictor ( $\beta = -0.217$ ). The model explained 57% of the variance in investment behavior (Adjusted  $R^2 = 0.569$ ).

ANOVA results revealed that individuals with higher incomes and greater investment experience made better decisions, suggesting that financial capacity and time in the market improve judgment and reduce behavioral biases.

### Recommendations

- For Individual Investors:** Improve self-awareness of biases like loss aversion and herd behavior. Continuous financial education and using digital financial planning tools can help counter these tendencies.
- For Financial Advisors:** Adopt behavioral profiling during client consultations. Tailor financial advice not only to income and goals but also to risk perception and psychological tendencies.
- For Employers (IT/ITES Firms):** Introduce workplace financial wellness programs, including seminars and investment literacy workshops. Partnering with investment platforms for employee benefits can encourage rational financial planning.
- For Policymakers:** Promote region-specific investor awareness campaigns. Mandating behavioral risk assessments in online trading platforms may help reduce impulsive decisions.

## 6. CONCLUSION

This study examined the impact of behavioral finance factors on the investment decision-making of IT and ITES professionals in Chennai. Through detailed quantitative analysis—including descriptive statistics, correlation, regression, and ANOVA—it was found that both behavioral traits and demographic variables significantly influence how these professionals make financial decisions. The results revealed that although participants are generally financially literate and active investors, their decisions are still shaped by psychological biases. Notably, loss aversion was identified as a key negative influence, often leading to overly cautious or suboptimal choices. In contrast, financial literacy, investment experience, and income level positively contributed to more confident and rational investment behavior. Among these, financial literacy emerged as the strongest predictor of sound decision-making. Regression results confirmed the predictive power of these variables, explaining a significant portion of the variance in investment behavior. ANOVA findings further highlighted that individuals with higher income and longer investment experience tend to make better financial decisions, underscoring the importance of both financial capacity and market familiarity.

In conclusion, behavioral finance remains highly relevant even among educated, tech-savvy professionals. Enhancing financial literacy, increasing market exposure, and addressing emotional biases are essential strategies for improving investment outcomes. These findings offer practical insights for financial advisors, employers, policymakers, and fintech platforms aiming to develop behaviorally informed financial tools and support systems.

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