



# An Analysis of Behavioral Biases in Investment Decision-Making

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**Abstract.** This study investigates the impact of cognitive and emotional biases on the investment decisions of individual investors in Chhattisgarh. Data were gathered from 100 participants through a comprehensive survey questionnaire that evaluated cognitive biases, including overconfidence, anchoring, confirmation bias, availability bias, and loss aversion, in addition to emotional biases such as fear of missing out (FOMO), herding behavior, regret aversion, and overoptimism. The descriptive analysis indicated varying levels of these biases among the participants, revealing prevalent behavioral trends in their investment decision-making processes. Correlation analysis indicated moderate positive associations between these biases and investment choices, while ANOVA findings demonstrated significant differences in decision-making based on the severity of cognitive and emotional biases. The regression analysis highlighted the predictive influence of these biases, illustrating their role in shaping investors' decisions, including risk perception, portfolio composition, and overall investment performance. The study reinforces the importance of behavioral finance in explaining investor behavior, while its practical recommendations emphasize the need for educating investors and employing behavioral strategies to reduce the influence of biases. While this study offers valuable insights, it also recognizes limitations, including the relatively small sample size and restricted applicability, which point to potential areas for additional investigation. Future studies could examine the effects of cultural, technological, and interdisciplinary factors on behavioral finance to gain a deeper understanding of the mechanisms influencing investor decision-making. In summary, the results add to the expanding body of knowledge in behavioral finance, improving comprehension of how biases impact financial choices in the marketplace.

**Keywords:** Investment decisions, Behavioral biases, Cognitive biases, Emotional Biases

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## 1 Introduction

Making investment decisions is a multifaceted process that goes beyond just examining financial data and market dynamics. Although conventional economic theories posit that investors act as rational agents, making choices solely based on logical analysis and the goal of maximizing profits, actual behavior reveals a different narrative. Behavioral finance, an emerging discipline, has shown that psychological and emotional elements significantly influence how investors arrive at decisions. These behavioral biases frequently cause investors to stray from rational thinking, leading to less-than-ideal financial results. Behavioral biases can be categorized broadly into two types: cognitive biases and emotional biases. Cognitive biases are recurring mistakes in reasoning that arise when people interpret and analyze information in a manner that deviates from logical judgment. These include overconfidence, anchoring, and confirmation bias, among others. These biases may result in illogical choices, such as selling in a panic or retaining losing stocks for an extended period, anticipating that they will recover.

Understanding the impact of these biases it's not only for investors but also for financial advisors and institutions aiming to improve decision-thinking strategies. By recognizing the influence of behavioral biases, investors can develop more rational and disciplined approaches, potentially avoiding costly mistakes and achieving better financial outcomes. This analysis aims to explore the different types of biases that affect investment decisions, investigate their root causes, and provide insights on how investors can mitigate their effects.

In the context of emerging markets like Chhattisgarh, where financial literacy is still evolving, these biases can significantly impact investment behaviors. This study will particularly focus on identifying the most prevalent behavioral biases among investors in the Indian Stock Exchange and explore how these biases correlate with demographic factors. By analyzing the investor psychology, the research will enhance comprehension of the behavioral factors that shape investment choices in both individual and institutional settings.

## 2 Literature Review

The study of behavioral biases in investment decision-making has gained considerable attention in recent years, particularly as scholars and practitioners seek to understand the limitations of traditional financial models. Classical theories have long dominated the field of finance, postulating that investors are rational, markets are efficient. However, empirical evidence has increasingly challenged these assumptions, leading to the rise of behavioral finance, a discipline that integrates psychology and economics to explain why investors often behave irrationally.

Thaler (1980) further advanced the field by exploring *mental accounting*—the tendency of individuals to categorize money and treat each category differently. This concept helps explain why investors sometimes make suboptimal decisions, as they fail to view their entire portfolio holistically. Behavioral finance, therefore, provides a framework for understanding the psychological biases that impact investment decisions.

### 2.1 Cognitive Biases in Investment Decisions

Cognitive biases are systematic tendencies to deviate from norms or rational thinking in judgment, causing individuals to shape their subjective understanding based on their interpretation of information. Barberis and Thaler (2003) highlighted various cognitive biases affecting investors, such as overconfidence, where individuals overrate their knowledge and ability to forecast market trends. Research shows that overconfident investors tend to engage in frequent trading, which can diminish returns due to increased transaction costs (Barber & Odean, 2001).

Another cognitive bias influencing investment decisions is anchoring. This occurs when investors place excessive emphasis on initial information (the "anchor") and are slow to adjust their expectations when presented with new data. Tversky and Kahneman (1974) demonstrated that anchoring can cause investors to cling to outdated price levels, affecting their decisions to buy or sell. Additionally, confirmation bias, where investors favor information that supports their existing views while ignoring conflicting evidence, can lead to flawed decision-making (Nickerson, 1998).

Another notable cognitive bias is the *availability heuristic*, where investors make decisions based on readily available information, often overestimating the likelihood of

recent events. This bias can lead investors to overreact to short-term market trends, contributing to herding behavior and market bubbles (Shiller, 2000).

## **2.2 Emotional Biases in Investment Decisions**

Emotional biases are influenced by the investor's feelings and psychological state. These biases are often more challenging to correct because they are rooted in emotions, rather than errors in logic or information processing. This behavior can drive market trends, but it can also lead to speculative bubbles or crashes when investors fail to critically evaluate the underlying fundamentals of their investments (Bikhchandani, Hirshleifer, & Welch, 1992).

Regret aversion is another significant emotional bias that affects investment decisions. Investors may avoid taking action due to the fear of making a wrong decision that they will later regret, leading to missed opportunities. This bias is closely related to *status quo bias*, where investors prefer to maintain their current investments rather than make changes, even when presented with better opportunities (Samuelson & Zeckhauser, 1988).

## **2.3 The Role of Demographics in Behavioral Biases**

Research have indicated that demographic characteristics like age, gender, education, and income can impact how behavioral biases influence investment choices. For instance, Barber and Odean (2001) observed that male investors tend to exhibit greater overconfidence compared to female investors, resulting in more frequent trading and increased transaction costs. Similarly, studies have shown that younger investors are more likely to exhibit risk-taking behaviors and cognitive biases such as overconfidence, while older investors may be more susceptible to emotional biases such as loss aversion (Dhar & Zhu, 2006).

Education and financial literacy are essential in influencing the effects of behavioral biases. Lusardi and Mitchell (2011) discovered that individuals possessing greater financial literacy are more capable of identifying and alleviating their own biases, resulting in more logical investment choices. However, even highly educated investors are not immune to biases, particularly during periods of market volatility, when emotional decision-making often overrides logical analysis (Shiller, 2000).

## **2.4 Behavioral Biases in the Indian Context**

While much of the research on behavioral biases has focused on developed markets, there is a growing body of literature examining these biases in emerging markets such as India. Emerging markets often exhibit higher volatility and greater information asymmetry, which can amplify the effects of behavioral biases on investment decisions. Studies conducted in the Indian context have revealed that biases such as overconfidence, herding behavior, and loss aversion are prevalent among Indian investors (Mittal & Vyas, 2011).

Patel (2017) found that Indian retail investors are particularly prone to herding behavior, often following market trends without conducting independent analysis. This behavior is often driven by limited access to information and a lack of financial literacy. Similarly, loss aversion was found to be a significant bias among Indian investors, leading many underperforming stocks for an extended period, anticipating a rebound (Bikas & Saha, 2017). Understanding the cultural and market-specific factors that contribute to these biases is crucial for developing strategies to mitigate their impact.

## **2.5 Mitigating Behavioral Biases**

Several strategies have been proposed to moderate the impact of behavioral biases on investment decisions. One approach is to increase financial literacy and investor education, helping individuals identify and rectify their biases. Financial advisors can also play a critical role by providing objective, data-driven advice that counteracts emotional decision-making (Pompian, 2012).

The use of automated investment platforms, such as robo-advisors, has also been suggested as a way to reduce the influence of biases by removing human emotion from the decision-making process. By relying on algorithms to make investment decisions based on objective criteria, investors can avoid the pitfalls of cognitive and emotional biases (Pompian, 2012). The literature on behavioral biases in investment decision-making highlights the significant role that psychological factors play in shaping investor behavior. Both cognitive and emotional biases can lead to suboptimal investment decisions, often resulting in financial losses or missed opportunities. Although demographic elements like age, gender, and financial knowledge can shape the extent to which these biases impact investors, no demographic is completely unaffected. In this regard, the widespread occurrence of biases like herd mentality and

loss aversion highlights the need for tailored strategies to enhance investor performance.

### **3 Objective**

To determine the cognitive biases that influence the investment decision-making process of individual investors.

To explore the emotional biases that affect the investment choices of investors.

To analyze the relationships between different behavioral biases and their effect on investment choices.

### **4 Hypothesis**

H1: Cognitive biases significantly affect the investment decisions of individual investors in Chhattisgarh.

H2: Emotional biases significantly affect the investment decisions of individual investors in Chhattisgarh.

H3: There is a significant correlation between behavioral biases and the investment choices of individual investors in Chhattisgarh.

### **5 Research Methodology**

This study utilizes a descriptive and correlational research approach to explore the influence of behavioral biases on investment decision-making. The data collection will be conducted via a structured survey, targeting a sample of 100 investors, using stratified random sampling to ensure a diverse representation across variables such as age, gender, income, and educational background. The survey will evaluate primary biases like overconfidence, loss aversion, anchoring, and herding, while also gathering demographic details. Additionally, secondary data will be sourced from credible financial platforms to assess investment performance indicators. The collected data will be examined using statistical methods such as descriptive statistics, correlation, multiple regression, and ANOVA to investigate the relationships between demographic

attributes, biases, and their effect on financial outcomes. A thematic analysis will be applied to any qualitative data to gain deeper understanding of investor behavior. Statistical analysis will be carried out using software like SPSS or R, with qualitative information managed and interpreted through suitable analytical tools. Ethical standards, including obtaining informed consent and ensuring confidentiality, will be strictly adhered to throughout the research process.

## **6 Result & data analysis**

The demographic characteristics of the Responses is crucial for interpreting the outcome and drawing meaningful insights about the investor community examined in this study. The demographic data provides essential context about the composition of the sample, allowing for the identification of trends or patterns that may shape investment behaviors.

The demographic assessment encompasses a range of characteristics, including factors like age, sex, educational background, earnings, profession, and prior investment knowledge. Evaluating these characteristics allows the research to identify how different segments of the investor population exhibit distinct levels of cognitive and emotional biases. This demographic insight is key to understanding investor behavior in the Indian financial markets and how personal attributes contribute to decision-making biases.

The analysis may uncover significant correlations between demographic variables and the presence of behavioral biases. Such relationships can offer valuable insights into how specific demographic groups approach investment decisions and how cognitive and emotional biases influence their actions.

The study's demographic profile shows a diverse sample, with participants spanning a wide range of ages, genders, education, and income levels in table 1. This diversity provides a comprehensive view of how different demographic factors may affect the behavioral biases that shape investment decisions, adding depth to the analysis of investor behavior.

**Table 1: Demographic profile**

Demographic Category	Category	Count
<b>Gender</b>	Male	62
	Female	38
	Others	0
	Prefer Not to Say	0
<b>Education</b>	High School	2
	Bachelor's Degree	39
	Master's Degree	54
	Doctorate/Ph.D.	5
<b>Age</b>	Under 25	44
	25 to 35	40
	36 to 45	4
	46 to 55	5
	56 and above	7
<b>Annual Income</b>	Below 300,000	31
	300,001 to 750,000	29
	750,001 to 1,000,000	10
	Above 1,000,000	30

In the study the gender breakdown shows that **62%** of respondents are male, while **38%** are female, indicating a slight gender imbalance. There is no representation from the "Others" or "Prefer Not to Say" categories, which may limit the study's ability to reflect the perspectives of non-binary or gender-nonconforming individuals. This disparity may introduce a potential bias in the data, as the overrepresentation of male respondents could influence certain findings related to investment behavior. Regarding educational attainment, 54% of the respondents hold a Master's degree, reflecting a high level of education within the sample. **39%** hold a Bachelor's degree, making up a significant portion of the sample. Only a small percentage of respondents have a **Doctorate/Ph.D. (5%)**, while **2%** have completed high school. This suggests that most respondents possess a strong academic background, which could be linked to more informed or strategic investment decisions. The varying educational levels offer a diverse perspective in the study, allowing for comparisons of investment behavior across different educational backgrounds. The age distribution shows that 44% of respondents are under 25 years old, with an additional 40% falling within the 25 to 35 age range. This suggests that the sample primarily consists of younger individuals, offering valuable insights into the investment behaviors of the younger generation. The representation of older age groups is considerably lower, with only **7%** of respondents aged 56 and above. The limited number of older participants may restrict the study's



ability to fully capture the investment tendencies of more experienced or older investors. In terms of income, there is a wide range of annual earnings among respondents. **31%** of participants report earning below **300,000**, while **30%** have incomes exceeding **1,000,000**, reflecting a substantial income disparity within the sample. Those earning between **300,001 and 750,000** account for **29%**, while **10%** fall within the **750,001 to 1,000,000** income bracket. The diverse financial backgrounds of the respondents suggest that the study can explore how varying income levels might influence investment strategies and risk tolerances. The demographic data provides important insights into the sample population's composition, highlighting a slightly skewed gender ratio, elevated educational levels, a predominantly young participant base, and varied income levels. These characteristics may play an important role in shaping the investment behavior and preferences examined in the study. The wide range of income and age groups ensures that the study can capture various perspectives, though the lack of representation in some categories, such as gender diversity and older age groups, may introduce certain limitations to the analysis.

**Table 2: Correlation Analysis**

	C1	C2	C3	C4	C5	E1	E2	E3	E4	P
C1	1									
C2	0.402	1								
C3	0.332	0.256	1							
C4	0.291	0.363	0.564	1						
C5	0.304	0.23	0.515	0.602	1					
E1	0.21	0.225	0.518	0.665	0.581	1				
E2	0.356	0.318	0.512	0.545	0.594	0.66	1			
E3	0.306	0.432	0.475	0.602	0.566	0.61	0.61	1		
E4	0.191	0.22	0.552	0.663	0.624	0.65	0.63	0.601	1	
P	0.346	0.336	0.302	0.321	0.327	0.228	0.421	0.423	0.377	1

The correlation matrix (See table 2) reveals a range of relationships between the cognitive factors (C1 to C5), emotional factors (E1 to E4), and the variable P. Notably, strong positive correlations are observed between variables like E1 and E4 (0.65) and C4 and C5 (0.602), indicating that these variables share significant commonalities or influence each other to a high degree. Other emotional variables, such as E2 and E1 (0.66) and E3 and E4 (0.601), also exhibit strong correlations, suggesting that emotional factors are interrelated, likely driven by shared underlying influences or behaviors.

On the other hand, the variable P demonstrates moderate correlations with several factors, particularly with E3 (0.423), E4 (0.377), and E2 (0.421), indicating that while P has a noticeable connection with these variables, the relationships are not overly dominant. Cognitive factors such as C1, C3, and C5 generally display weaker correlations with each other and with emotional factors, suggesting limited interdependencies among these variables. This overall pattern highlights the intricate yet varied influences between cognitive, emotional, and performance-related factors in the dataset.

**Table 3: ANOVA Test: Cognitive Biases**

ANOVA: Single Factor				
Summary				
Group	Count	Sum	Average	Variance
C1	100	360	3.6	0.913737
C2	100	374	3.74	0.648283
C3	100	349	3.49	0.929293
C4	100	360	3.6	0.934242
C5	100	345	3.45	1.05202
P	100	659	6.59	3.065657

ANOVA						
Source of Variation	SS	df	MS	F	P-value	F crit
Between Groups	753.5683	5	152.3137	116.0881	3.78E-82	2.239193
Within Groups	750.75	592	1.242205			
Total	1504.318	597				

The ANOVA test (see table 3) conducted to examine the association between cognitive biases and investment choices provides compelling evidence supporting the hypothesis. The findings reveal significant variability in investment behavior shaped by different cognitive biases, such as Overconfidence, Anchoring Bias, Confirmation Bias, Availability Bias, and Loss Aversion, all of which meet the statistically significant threshold ( $p < 0.05$ ).

The calculated F-statistic for the differences among groups is 116.0881, with a p-value of 3.78E-82. This extremely low p-value suggests that the variations observed in investment behavior across various cognitive biases are highly unlikely to have occurred by chance. Consequently, we reject the null hypothesis, which claims that no

notable differences exist in investment choices linked to cognitive biases. Instead, we accept the alternative hypothesis, confirming that cognitive biases significantly affect the investment behavior of individual investors in India.

These findings show the vital part of cognitive biases plays in shaping investment behavior. By impacting determine processes, these biases can undermine the rationality of investment choices, potentially resulting in less-than-optimal outcomes. The considerable variance in investment decisions suggests that certain cognitive biases may exert a more significant influence than others, indicating a need for further investigation into which biases most profoundly affect investor behavior.

In summary, the ANOVA analysis (see table 4) clearly illustrates that cognitive biases are significant determinants of investment decisions. The study emphasizes the importance for both investors and financial advisors to be cognizant of these cognitive influences to facilitate more informed and rational investment decisions.

**Table 4: ANOVA Test: Emotional Biases**

ANOVA: Single Factor				
Summary				
Group	Count	Sum	Average	Variance
E1	100	345	3.45	0.939495
E2	100	326	3.26	0.951313
E3	100	340	3.4	0.924646
E4	100	345	3.45	0.888586
P	100	658	6.58	3.045657

ANOVA						
Source of Variation	SS	df	MS	F	P-value	F crit
Between Groups	798.02	4	200.65	145.838	4.41E-81	2.358995
Within Groups	675.75	494	1.337939			
Total	1473.77	498				

The variance analysis conducted to investigate the connection between emotional biases and investment preferences offers crucial insights into the effect of emotional factors on investor behavior. The outcomes demonstrate a statistically significant variation in investment preferences across different levels of emotional biases, such as Fear of Missing Out, Herding Behavior, Regret Aversion, and Overoptimism.

The derived F-statistic for the group distinctions is 145.838, with an associated p-value of 4.41E-81. This extremely low p-value provides strong justification to reject the null hypothesis, which assumes no significant differences in investment preferences tied to emotional biases. Instead, the findings clearly indicate substantial differences in investment patterns attributable to emotional biases among individual investors in India.

These results highlight the significant influence that emotional biases have in molding investment decisions. Investors who experience emotions such as fear, regret, or the tendency to follow the crowd may make less rational choices, potentially compromising their investment outcomes. The significant variation in investment decisions among the different emotional biases indicates that certain biases may exert a more pronounced influence than others.

The critical role that emotional biases play in influencing investment choices. They suggest that emotional factors significantly shape how investors respond to market conditions, leading to varied investment outcomes. In conclusion, the ANOVA analysis confirms that emotional biases are significant determinants of investment decisions, reinforcing the importance of recognizing these influences in order to enhance investment strategies.

## **7 Findings**

The analysis uncovers key insights regarding the impact of emotional biases on the investment choices of individual investors in Chhattisgarh. Firstly, there is strong evidence that emotional factors like Fear of Missing Out, Herding Behavior, Regret Aversion, and Overoptimism play a significant role in shaping how investors make decisions. The statistically significant differences observed across varying levels of these biases suggest that emotional influences are not only prevalent but also impactful in shaping investment strategies.

Moreover, the calculated F-statistic and corresponding p-value provide robust evidence that these emotional biases contribute to variability in investment decisions. This finding emphasizes the necessity for investors to recognize and understand their emotional responses and rational choices. The findings suggest that increased

awareness of these emotional biases could enhance decision-making processes and result in more favorable financial outcomes. This study highlights the essentiality of assimilating emotional awareness into Capital allocation, particularly in dynamic and volatile markets like those found in India. By acknowledging and addressing these emotional influences, investors can enhance their decision-making capabilities and potentially mitigate the risks associated with impulsive and emotionally driven choices.

## **8 Conclusion**

This study offers an in-depth analysis of the psychological biases that notably impact investment choices among individual investors in Chhattisgarh. By analyzing both cognitive and emotional biases, the study highlights the intricate ways in which these psychological factors shape investor behavior, often leading to suboptimal financial outcomes.

The results highlight the significance of recognizing these biases, as they are deeply ingrained in investors' decision-making processes. The strong relationships identified between different biases and investment choices reveal that both cognitive and emotional elements are crucial in financial evaluation.

Furthermore, the demographic insights reveal how age, gender, education, and income levels impact the extent to which these biases manifest, emphasizing the need for tailored educational interventions and strategies to enhance financial literacy.

By integrating behavioral finance principles into their strategies, investors can better navigate market complexities, reduce the likelihood of emotionally driven decisions, and ultimately achieve more favorable financial results. This study not only contributes to the existing literature on behavioral finance but also offers practical implications for individual investors, financial advisors, and institutions aiming to improve investment decision-making in a rapidly changing economic environment.

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