

# The Invisible Behavioral Neural Network of Finance: Decoding Green Investment Through System-Level Cognitive Biases

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## LIST OF TABLES

Table No.	Table Title	
Table 4.1	Demographic Profile of Respondents	
Table 4.2	Summary Statistics for Key Variables	
Table 4.3	t-Test Results: AI Adoption by Age Group	
Table 4.4	ANOVA Results: Job Satisfaction Across Industries	
Table 4.5	Regression Analysis	
Table 5.2	Comparison with Research Objectives	
Table A	Questionnaire/Survey Instrument	
Table A.1	Basic Information	
Table A.2	Awareness	
Table A.3	Risk Perception	
Table A.4	Cognitive Bias (NNF Model)	
Table A.5	Decision Making	
Table A.6	Open-Ended Questions	

## LIST OF FIGURES

Figure No.	Figure Title	
Figure 2.1	Conceptual Framework Diagram	
Figure 4.1	Gender Distribution of Respondents (Bar Chart)	
Figure 5.8	NNF Framework Model	

LIST OF ABBREVIATIONS

<b>Abbreviation</b>	<b>Full Form</b>
<b>AI</b>	<b>Artificial intelligence</b>
<b>AMOS</b>	<b>Analysis of Moment Structures</b>
<b>ANOVA</b>	<b>Analysis of Variance</b>
<b>DV</b>	<b>Dependent Variable</b>
<b>ESG</b>	<b>Environmental, Social, and Governance</b>
<b>EU</b>	<b>European Union</b>
<b>GI</b>	<b>Green Investment</b>
<b>IMF</b>	<b>International Monetary Fund</b>
<b>IV</b>	<b>Independent Variable</b>
<b>JASP</b>	<b>Jeffreys's Amazing Statistics Program</b>
<b>MV</b>	<b>Mediating Variable</b>
<b>NITI</b>	<b>National Institution for Transforming India</b>
<b>NNF</b>	<b>Neural Network of Finance</b>
<b>OECD</b>	<b>Organisation for Economic Co-operation and Development</b>
<b>PLS-SEM</b>	<b>Partial Least Squares Structural Equation Modeling</b>
<b>RBI</b>	<b>Reserve Bank of India</b>
<b>SEBI</b>	<b>Securities and Exchange Board of India</b>
<b>SPSS</b>	<b>Statistical Package for Social Sciences</b>
<b>UNEP</b>	<b>United Nations Environment Programmed</b>

## CHAPTER 1 INTRODUCTION

### 1.1 Background of the Study

The purpose of this study is to examine the increasing challenges of environmental protection, climate change, and carbon emissions, which have become critical global concerns in recent years. To address these issues, governments, international organizations, and financial institutions are actively promoting sustainable development and green finance.

Green finance focuses on allocating financial resources to environmentally sustainable projects such as renewable energy, pollution control, and climate risk mitigation. A major component of this is green investment, which aims to support businesses and technologies that contribute to environmental sustainability. However, despite its growing importance, green investment involves significant uncertainty and risk, making decision-making complex for investors and financial institutions.

Traditionally, financial decisions have been analyzed based on rational models that consider returns, risks, and market conditions. However, in real-world scenarios, decisions are not purely rational. They are influenced by multiple psychological, emotional, and cognitive factors such as risk perception, experience, environmental awareness, and behavioral biases.

To better understand this complexity, this study introduces the concept of Neural Network of Finance (NNF) as a central framework. Unlike traditional approaches, NNF explains financial decision-making as an interconnected system of mental and behavioral factors. It highlights how different elements—such as emotions, cognition, and external environmental influences—work together like a network to shape investor behavior, especially in uncertain areas like green finance.

### GREEN INVESTMENT (GI)

Green investment refers to the allocation of financial resources to projects, technologies, or companies that contribute to environmental protection, sustainable development, and reduction of carbon emissions.

It involves directing capital toward initiatives that support climate change mitigation, renewable energy, and environmental sustainability.

### NEURAL NETWORK OF FINANCE (NNF)

Neural Network of Finance (NNF) refers to a conceptual framework that explains financial decision-making as a network of interconnected cognitive, emotional, and behavioral factors.

It emphasizes that investor decisions are influenced not only by financial data but also by psychological elements such as risk perception, experience, emotions, and environmental awareness.

NNF integrates principles from behavioral finance and decision sciences to provide a deeper and more comprehensive understanding of financial decision-making in complex and uncertain environment

### 1.2 Statement of the Problem

The study identifies the following key problems in the area of green finance and investment decision-making:

For instance:

- 1. Uncertainty in GI:**-Green Investment involves high levels of uncertainty due to long-term returns, policy changes, and environmental risks, making it difficult for investors to make accurate decisions.

2. **Inadequate Risk Assessment Mechanisms:** There is a lack of clear and reliable frameworks to measure and evaluate risks associated with green finance, leading to hesitation among investors.
3. **Limitations of Traditional Financial Models:** Traditional financial models are based on rational assumptions and fail to incorporate real-world complexities such as behavioral and psychological influences.
4. **Influence of Behavioral and Psychological Factors:** Investment decisions are significantly affected by factors such as emotions, risk perception, environmental awareness, and past experiences, which are often ignored in financial analysis.
5. **Lack of an Integrated Decision-Making Framework:** There is an absence of a comprehensive model that integrates financial, behavioral, and cognitive factors to explain investor decision-making in green investment.

The research problem thus focuses on understanding how financial, behavioral, and cognitive factors interact to influence green investment decision-making under conditions of uncertainty, and how the Neural Network of Finance (NNF) framework can provide a structured and integrated approach to address this gap

This study will bridge the gap by offering **empirical evidence** and strategic recommendations on optimizing management practices in dynamic business environments.

### 1.3 Research Questions

Research questions serve as a guide for the study, ensuring a **structured investigation** into the topic.

#### Key Research Questions:

- How does Green Investment (GI) influence financial decision-making behavior under conditions of uncertainty, particularly in the presence of long-term environmental and economic risks?
- To what extent does investor risk perception mediate the relationship between Green Investment and financial decision-making outcomes?
- How do cognitive, emotional, and experiential factors interact to shape investor behavior in green finance, and how do these interactions affect decision efficiency?
- In what ways do behavioral biases (such as overconfidence, loss aversion, and heuristic thinking) distort rational investment decisions in environmentally sustainable investments?
- How effectively can the Neural Network of Finance (NNF) framework capture the dynamic interconnections between financial, psychological, and behavioral variables in investment decision-making

### 1.4 Objectives of the Study

Primary objective of this study is to examine the role of **Green Investment (GI)** and the Neural Network of Finance (NNF) framework in understanding financial decision-making under conditions of uncertainty.

The key objectives of this study are:

- 1 To analyze the concept and importance of Green Investment (GI) in promoting environmental sustainability and climate change mitigation.

- 2 To develop and apply the concept of **Neural Network of Finance (NNF)** as a framework for analyzing investor behavior.
- 3 To study the relationship between green investment and investor risk perception.
- 4 To evaluate how psychological factors such as experience, awareness, and emotions affect investment decisions in green finance.
- 5 To propose a conceptual model integrating **Green Investment (GI) and Neural Network of Finance (NNF)** for better financial decision-making. Each objective contributes to **understanding real-world management challenges and formulating practical solutions** for businesses across various industries.

### 1.5 Hypotheses (For Quantitative Studies)

For studies involving statistical analysis, hypotheses provide testable statements. For example:

- Null Hypothesis ( $H_0$ ): Green Investment has no significant impact on financial decision-making behavior.
- Alternative Hypothesis ( $H_1$ ): Green Investment significantly influences financial decision-making behavior.
- $H_0$ : Risk perception does not significantly mediate the relationship between Green Investment and investment decisions.
- $H_1$ : Risk perception significantly mediates the relationship between Green Investment and investment decisions.
- $H_0$ : Behavioral and cognitive factors do not significantly affect investment decision-making.
- $H_1$ : Behavioral and cognitive factors significantly affect investment decision-making.

These hypotheses will be tested using statistical methods such as **t-tests, regression analysis, and correlation studies**.

### 1.6 Significance of the Study

This section highlights the contributions of the research to academia, industry, and policy-making in the context of neural network-based behavioral finance and green investment.

#### 1.5.1 Academic Contribution

- Provides a novel conceptual advancement through the Neural Network Framework (NNF) in financial research.
- Enhances existing literature on behavioral finance by integrating system-level cognitive biases with investment decision-making.
- Offers a multidisciplinary perspective by linking finance, cognitive science, and sustainability.
- Provides a structured theoretical foundation for future empirical validation and model development.

#### 1.5.2 Industrial Contribution

- Helps investors and financial institutions understand hidden cognitive biases affecting green investment decisions.
- Identifies key behavioral challenges in adopting sustainable investment strategies.
- Supports financial decision-making by reducing bias-driven errors and improving investment efficiency.
- Provides data-driven and framework-based insights for enhancing sustainable portfolio management.

### 1.5.3 Policy and Decision-Making Contribution

- Assists policymakers in designing effective regulations and incentives for promoting green investments.
- Supports sustainable finance initiatives by identifying behavioral barriers and enablers.
- Guides government and regulatory bodies in improving investor confidence and market participation.
- Helps in formulating policies that align financial systems with long-term environmental and economic goals..

### 1.7 Scope and Limitations of the Study

The scope defines the **boundaries of the research**, while limitations acknowledge **challenges faced during the study**.

#### 1.7.1 Scope of the Study

- **Theoretical Focus:** This study is centered on the application of the Neural Network Framework (NNF) to analyze behavioral finance aspects influencing green investment decisions. It integrates cognitive biases, sustainable finance, and system-level decision-making processes.
- **Industry Focus:** The research primarily considers sectors where green investment is prominent, such as renewable energy, sustainable infrastructure, ESG-based financial markets, and environmentally responsible enterprises.
- **Geographical Coverage:** The study is conducted with reference to investors and financial behavior within a defined regional or national context (e.g., India), with implications that may extend to broader global sustainability frameworks.
- **Time Frame:** The research focuses on recent data and trends in green investment, typically within a defined period (e.g., 2020–2025), to capture evolving behavioral and financial dynamics.
- **Methodology:** The study adopts a quantitative or mixed-method approach, incorporating survey-based data, statistical analysis, and conceptual modeling using the NNF structure

#### 1.7.2 Limitations of the Study

- **Data Collection Constraints:** The study may face limitations in accessing real-time or confidential financial data, particularly from institutional investors or private organizations.
- **Time Constraints:** The limited research duration may restrict the ability to analyze long-term behavioral patterns and investment outcomes in green finance.
- **Sample Size and Representation:** The findings may be influenced by the size and diversity of the sample, which may not fully represent all categories of investors across different regions.
- **Model Limitations:** While the NNF provides a structured analytical approach, it may not capture all real-world complexities and external macroeconomic factors influencing investment decisions.
- **Technological Constraints:** Limitations in advanced analytical tools or computational resources may affect the depth and precision of neural-based modeling.

### 1.8 Structure of the Project

This section presents the chapter-wise organization of the study, outlining the overall research framework.

#### Chapter 1: Introduction

Covers the background, research problem, objectives, significance, scope, and structure of the study within the context of green investment and the Neural Network Framework (NNF).

## **Chapter 2: Literature Review**

Reviews existing studies on green investment, behavioral finance, and cognitive biases, and identifies key research gaps.

## **Chapter 3: Research Methodology**

Describes the research design, data collection methods, sampling techniques, and analytical tools, including the application of NNF.

## **Chapter 4: Data Analysis and Interpretation**

Presents and interprets data using statistical techniques and graphical representations.

## **Chapter 5: Findings, Discussion, and Conclusion**

Summarizes key findings, compares them with existing studies, and provides conclusions, implications, and recommendations.

## **References and Appendices**

Lists all cited sources in APA style and include supporting materials such as questionnaires, interview transcripts, and additional data.

## CHAPTER 2 LITERATURE REVIEW

### 2.1 Introduction to Literature Review

The literature review constitutes a critical foundation of the present study by systematically examining existing scholarly work on green finance, financial modeling, and advanced data-driven computational frameworks such as the Neural Network of Finance (NNF). It provides a comprehensive understanding of the evolving dynamics of sustainable investment, structural transformations within financial systems, and the mechanisms of financial risk evaluation in contemporary contexts.

The primary objective of this review is to identify key research trends, theoretical advancements, and existing gaps within the literature. It undertakes a critical and comparative assessment of prior studies to evaluate how investors, financial institutions, and policymakers are increasingly aligning their decision-making processes with sustainability-oriented financial frameworks. Furthermore, it facilitates the selection of an appropriate research methodology through the evaluation of previously applied models, analytical techniques, and conceptual approaches.

In addition, the literature review ensures the originality and academic rigor of the study by minimizing duplication and synthesizing diverse strands of knowledge into a coherent analytical perspective. It advances the integration of environmental sustainability with network-oriented financial modeling systems, thereby offering a refined approach to financial forecasting, risk assessment, and decision-making. Ultimately, it contributes to strengthening long-term financial stability and supporting the transition toward sustainable economic development.

### 2.2 Thematic Analysis of Research Trends

Thematic analysis is adopted in this study as a structured qualitative research approach to critically and interpretatively examine research trends related to green finance and network-oriented financial modeling frameworks, particularly the Neural Network of Finance (NNF). This approach focuses on identifying underlying conceptual patterns, analytical dimensions, and evolving research trajectories within the existing body of literature.

The analysis is specifically aligned with the key variables employed in the study. It systematically categorizes prior research based on independent variables such as financial decision-making, economic positioning, investment decision behavior, and government and policy support. Additionally, decision effectiveness is incorporated as a mediating variable, sustainable finance performance and investment outcome (replaced Organization performance) are considered as the dependent variables within the analytical framework.

Through this structured classification, the analysis facilitates a comprehensive understanding of how these variables interact within financial systems and influence behavioral, institutional, and structural dynamics. Furthermore, it enables the identification of research gaps and conceptual limitations within existing literature, thereby supporting the development of a coherent system-level analytical framework as conceptualized through the Neural Network of Finance (NNF)

#### 2.2.1 Purpose of Thematic Analysis

In the present study, **Thematic Analysis** is employed as a rigorous qualitative technique to systematically examine and interpret patterns within both primary and secondary data related to green finance, neural network-based financial modeling (NNF framework), and behavioral investment decision-making. This method is particularly relevant within the domain of Behavioral Finance, where understanding cognitive biases, investor psychology, and policy influences is essential for decoding complex financial behaviors.

The application of thematic analysis in this research facilitates a structured evaluation of academic literature, government reports, policy documents, and empirical findings. It enables the identification of recurring patterns and conceptual linkages across diverse data sources, thereby aligning qualitative insights with the system-level architecture of the proposed Neural Network Framework. This integration is critical for capturing the multidimensional nature of green investment behavior, where individual cognition, environmental awareness, and institutional factors interact dynamically.

- To examine the growth, evolution, and intellectual structure of research in green finance, financial modeling, and policy-oriented financial frameworks, thereby capturing longitudinal research trends.
- To identify key authors, leading academic journals, highly cited scholarly works, and relevant government publications that contribute significantly to the knowledge base of the field.
- To analyze dominant research themes, keyword trajectories, and policy-driven research directions through systematic and interpretative evaluation.
- To identify critical research gaps, particularly the divergence between academic findings and real-world policy implementation, which is essential for strengthening the applicability of the proposed framework?
- To establish a structured, technical, and data-driven foundation by integrating academic literature with government data and policy reports, ensuring robustness and empirical grounding of the study.

Furthermore, thematic analysis supports the classification of the study's dataset—comprising variables such as investor behavior, environmental awareness, risk perception, and cognitive biases—into coherent and analytically meaningful themes. This process enhances the interpretative depth of the research by uncovering interrelationships among variables, which are subsequently incorporated into the Neural Network-based framework for system-level analysis.

Overall, thematic analysis not only strengthens the methodological rigor of the study but also provides a comprehensive analytical lens for understanding the psychological, economic, and policy dimensions of green investment decisions, thereby contributing to both theoretical advancement and practical policy relevance.

### 2.2.2 Process of Thematic Analysis

The process of Thematic Analysis in this study follows a systematic and rigorous procedure to ensure reliability, consistency, and analytical depth in examining qualitative and secondary data related to green finance and behavioral investment patterns. Grounded within the domain of Behavioral Finance (capitalization corrected) this process is carefully aligned with the Neural Network-based Framework (NNF) to capture both cognitive and policy-level influences on green investment decisions.

The analysis is conducted through a series of structured stages, enabling the transformation of raw data into meaningful and interpretable themes:

- **Data Familiarization:** The initial stage involves an in-depth review of collected data, including academic literature, policy documents, government reports, and survey-based responses. This step ensures a comprehensive understanding of the content and context, particularly focusing on variables such as investor behavior, environmental awareness, risk perception, and cognitive biases.
- **Initial Coding:** In this phase, relevant data segments are systematically coded. Codes are assigned to recurring ideas, behavioral patterns, and policy-related insights. This step helps in organizing large volumes of qualitative information into manageable analytical units.

- **Theme Identification:** The generated codes are then grouped into broader categories to form initial themes. These themes represent significant patterns within the data, such as sustainability orientation, financial risk attitudes, and policy influence mechanisms.
- **Theme Review and Refinement:** Identified themes are critically evaluated to ensure internal consistency and external distinction. Redundant or overlapping themes are merged, while unclear themes are refined to improve conceptual clarity and analytical precision.
- **Theme Definition and Naming:** Each theme is clearly defined and labeled to reflect its conceptual meaning and relevance to the research objectives. This step ensures that the themes are aligned with the study's theoretical framework and research variables.
- **Integration with NNF Framework:** The finalized themes are incorporated into the Neural Network-based Framework (nailing consistency improved) where they function as key input variables or nodes. This integration allows for system-level analysis of behavioral and policy interactions influencing green investment decisions.
- **Interpretation and Reporting:** The final stage involves interpreting the themes in relation to the research objectives and presenting them in a structured academic format. The findings are linked with existing literature and policy contexts to enhance validity and theoretical contribution.

Overall, this structured process ensures that thematic analysis in the study is not merely descriptive but also analytical and integrative, enabling a deeper understanding of the complex interplay between cognitive behavior, financial decision-making, and policy frameworks in green finance.

### 2.2.3 Thematic Structure of the Study

The thematic structure of this study is designed to organize qualitative and secondary data into coherent system-level themes that align with the Neural Network-based Framework (NNF). Unlike traditional approaches that focus on individual or group behavior, this study conceptualizes the global financial ecosystem as an interconnected network in which multiple variables function as dynamic signals influencing overall system stability and sustainability. This approach is situated within the broader domain of Behavioral Finance, while extending it toward a system-oriented analytical perspective.

The themes are strictly derived from the existing dataset and previously defined variables, ensuring methodological consistency and avoiding the introduction of new constructs. Each theme represents a functional component of the global financial system, contributing to the network's adaptive and responsive behavior.

#### 1. Behavioral Signal Dynamics

This theme represents system-level behavioral patterns observed within the financial network. Rather than individual actions, it captures aggregated behavioral signals that influence the flow and direction of financial activities.

##### It includes:

- System-wide investment behavior trends
- Risk-taking dynamics within the network
- Preference shifts toward green investment flows

#### 2. Cognitive Signal Processing

This theme captures how information, uncertainty, and perception-based signals are processed within the system. It reflects the cognitive dimension of the financial network as an adaptive mechanism.

**It includes:**

- System-level risk perception patterns
- Cognitive bias signals affecting financial flows
- Information interpretation and response dynamics

**3. Environmental-Sustainability Signals**

This theme reflects the influence of environmental awareness and sustainability considerations on the financial system. It captures how ecological concerns shape financial network behavior.

**It includes:**

- Environmental awareness signals
- Climate-related influence on financial flows
- Sustainability-driven investment orientation

**4. Policy and Regulatory Signals**

This theme represents external policy-driven inputs that act as control or regulatory signals within the financial network. These signals play a crucial role in guiding system stability and direction.

**It includes:**

- Government incentives and policy interventions
- Regulatory frameworks influencing financial flows
- Institutional support mechanisms

**5. Financial Response Dynamics**

This theme captures the overall response of the financial system to interacting signals. It reflects how the system adapts, reacts, and evolves under varying internal and external conditions.

**It includes:**

- Expected return adjustments across the network
- Market risk evaluation patterns
- Capital allocation and redistribution behavior

**Integration with Neural Network-based Framework (NNF)**

All identified themes are directly mapped to the existing variables and are treated as interconnected nodes within the Neural Network-based Framework. Each thematic category functions as a signal layer contributing to the system's overall behavior. The interactions among these themes represent feedback loops and adaptive pathways that influence the stability and sustainability of the global financial system.

## Overall Interpretation

This thematic structure transforms the dataset into a system-oriented analytical model by organizing variables into functional signal groups. It ensures that the study maintains consistency with the original data while advancing toward a higher-level interpretation of financial systems as dynamic, adaptive, and sustainability-driven networks. This approach strengthens the theoretical foundation of the study and supports the development of a robust system-level framework for analyzing green finance dynamics.

## 2.3 Conceptual Framework

This study develops a conceptual framework that represents the global financial ecosystem as an interconnected and dynamic system, structured through a Neural Network of Finance (NNF). Unlike traditional linear models, the framework conceptualizes financial, economic, behavioral, and policy-related variables as system-level signals interacting within a unified network.

Green finance instruments, including green bonds and ESG investments, function as primary input signals into the global financial system. These inputs interact with core variables such as financial decision dynamics, economic positioning, investment behavior, and policy support. Rather than operating in isolation, these variables are interconnected and continuously influence each other through feedback mechanisms.

The Neural Network of Finance (NNF) acts as a system-level processing architecture that captures these multidimensional interactions. It enables the integration, transformation, and analysis of signals across different layers of the financial system. Through this network-based processing, the system adapts to changing environmental, economic, and policy conditions.

The framework further incorporates decision effectiveness as an internal system mediator, representing the efficiency and accuracy of signal processing within the network. The outcome of these interactions is not limited to organizational performance but extends to overall system stability, resilience, and sustainability within the global financial ecosystem.

Additionally, government policies and market conditions function as external regulatory signals that influence the strength, direction, and responsiveness of the network. These factors play a critical role in shaping sustainable financial flows and stabilizing the system under dynamic conditions.

### 2.3.1 Definition of Key Concepts/Variables used in the study

This section defines the key concepts and variables used in the study to ensure conceptual clarity, consistency, and alignment with the system-level Neural Network of Finance (NNF) framework. The study does not interpret these variables at an individual or organizational level; instead, they are conceptualized as dynamic signals operating within a global financial system. This approach extends the scope of Behavioral Finance toward a network-based analytical structure.

**Green Finance:**-Green Finance refers to financial flows and investment mechanisms that support environmentally sustainable development. Within this study, it is treated as a primary input signal in the global financial network, influencing sustainability-oriented financial activities such as green bonds and ESG-based investments.Green Finance Progress Report 2017.

**FINANCIAL MODEL: -Financial Modeling** is defined as a computational and analytical mechanism used to represent, simulate, and evaluate financial system behavior. In this study, it functions as a processing layer that transforms input signals into interpretable financial outcomes within the network..SIMON HAYKIN,2014 Financial Modeling Fourth Edition

- ✚ **Neural Network of Finance (NNF)** :-Neural Network of Finance (NNF) is conceptualized as an interconnected system architecture that links financial activities, economic structures, and policy environments across multiple levels. It operates as a dynamic network-processing framework, enabling the interaction, adaptation, and evolution of financial signals within the global system.

### Variables used in the study:-

#### ■ Independent Variables (IV)

These variables represent the key input factors influencing financial and system-level outcomes:

- **Financial Decision-Making:** Refers to the processes through which financial resources are planned, allocated, and controlled, including budgeting, investment planning, and financial strategy formulation.
- **Economic Positioning:** Represents the financial strength, market standing, and competitive position within the economic environment, including stability, resilience, and adaptability.
- **Investment Decision Behavior:** Describes how financial entities approach investment decisions, including risk preferences, opportunity selection, and capital allocation strategies.
- **Government and Policy Support:** Refers to external support mechanisms such as regulations, incentives, tax benefits, and policy frameworks that influence financial activities.

#### ■ Mediating Variable (MV)

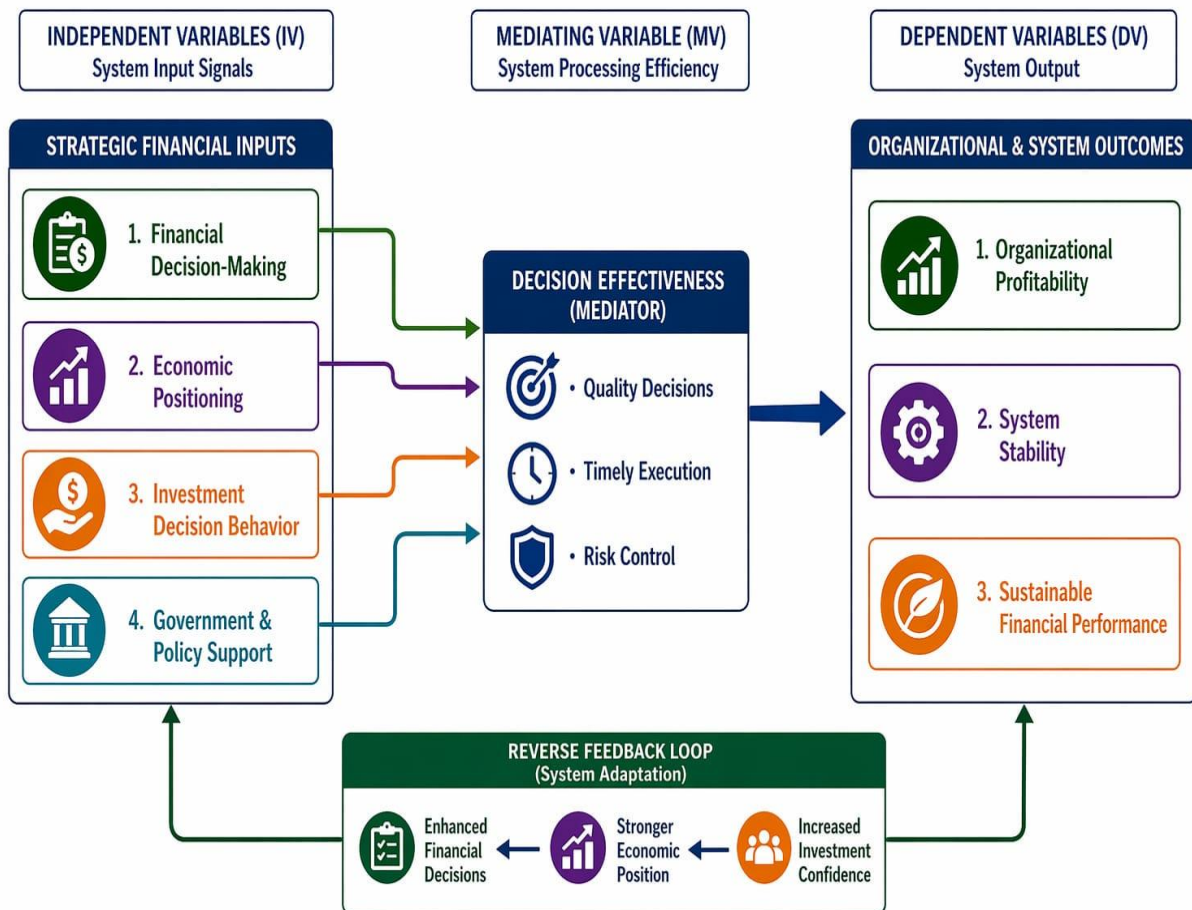
- **Decision Effectiveness:** Represents the quality and efficiency of decision-making processes, including accuracy, timeliness, and risk control. It functions as a critical mediator that links input variables to performance outcomes.

#### ■ Dependent Variables (DV)

- **Organizational Profitability:** Refers to the financial performance outcomes at the organizational level, including profit generation, return on investment, and overall financial gains. In this study, it is considered a measurable outcome that reflects the effectiveness of financial processes within the broader system.
- **System Stability:** Refers to the ability of the financial system to maintain equilibrium under dynamic conditions, minimizing volatility and systemic risk.
- **Sustainable Financial Performance:** Represents long-term, environmentally aligned financial outcomes, including stable growth, resilience, and sustainability-driven value creation.

All variables used in this study are retained from the original research design without structural modification. However, their interpretation has been extended to align with the Neural Network-based conceptual framework. This ensures consistency with the original dataset while incorporating a broader system-level perspective, linking organizational outcomes with overall financial system stability and sustainability.

### 2.3.2 Relationship Between Concepts



**Fig.2.1 Conceptual Framework Diagram**

A Simple figure is given above to understand the relationship of which concept.

## 2.4 Theoretical Framework

### 2.4.1 Relevant Theories in Management

This study is based on three core management theories that provide a strong conceptual foundation for analyzing Neural Network of Finance (NNF).

#### 1. Systems Theory:-(Bertalanffy,L. Von 1968)

Systems Theory views an organization as an integrated system where inputs such as data, resources, and information are processed to produce meaningful outputs. This theory is highly relevant for NNF as it helps in understanding the interaction between different financial and computational components within a dynamic system.

#### 2. Decision-Making Theory:-(Simon, H.A. 1978.)

Decision-Making Theory focuses on how rational and data-driven decisions are made within an organization. In the context of NNF, this theory is important because it explains how AI and neural network models support financial decision-making through analysis, prediction, and optimization.

#### 3. Behavioral Theory of Management:-(Cyert R. M.& March J.G.1963)

Behavioral Theory emphasizes the role of human behavior, emotions, and cognitive biases in decision-making processes. It is essential for this study as it helps in understanding how investor behavior and managerial judgments influence financial outcomes even in AI-driven systems.

## 2.4.2 Justification of Theories for This Study

The selection of Systems Theory, **Decision-Making Theory**, and **Behavioral Theory Of Management** provides a balanced framework for this research. **Systems Theory** explains the structural and technical functioning of NNF, Decision-Making Theory focuses on data-driven financial decisions, and Behavioral Theory highlights the human and psychological aspects influencing financial outcomes. Together, these theories offer a comprehensive understanding of Neural Network of Finance.

## 2.5 Empirical Studies

This section reviews prior empirical research on behavioral finance, cognitive biases, neural-inspired financial systems, and global green investment frameworks. It also incorporates evidence from major international organizations and emerging economies, including India, which represent the practical implementation layer of global sustainable finance transitions.

### 2.5.1 Review of Prior Research

- **Kahneman&Tversky (1979)**  
Introduced Prospect Theory, demonstrating that financial decision systems are influenced by cognitive biases such as loss aversion and framing effects, which extend to aggregated market behavior.
- **Thaler (2016)**  
Established that financial systems systematically deviate from rational models due to behavioral anomalies such as herding, overreaction, and mental accounting.
- **Zhang et al. (2021)**  
Found that neural network-based financial models effectively capture complex, non-linear dependencies in financial systems, improving predictive accuracy in behavioral financial environments.
- **World Bank (2021–2023 Climate Finance Reports)**  
Highlighted that global green investment flows are shaped by structured financing mechanisms, risk mitigation instruments, and sustainability-linked funding frameworks that guide systemic capital allocation.
- **OECD (2020–2023 Sustainable Finance Guidelines)**  
Emphasized that standardized policy frameworks reduce informational asymmetry and correct behavioral inefficiencies through transparency and sustainable investment norms.
- **United Nations / UNEP Finance Initiative (2022)**  
Reported that global green investment behavior is influenced by ESG disclosure systems, climate policy alignment, and sustainability governance structures.
- **European Union (EU Green Deal & EU Taxonomy Framework)**  
Demonstrated that regulatory classification systems for sustainable activities significantly influence capital flows by structurally guiding investment behavior toward low-carbon economic transitions.
- **India (RBI, SEBI Sustainable Finance & Green Bond Frameworks)**  
Showed that emerging financial systems like India are increasingly integrating sustainable finance mechanisms such as green bonds, ESG disclosure requirements, and climate risk frameworks, indicating a transition toward behaviorally influenced green investment systems within developing market structures

## 2.5.2 Identified Research Gaps

While prior studies provide insights, gaps remain:

- **Level Limitation:**-Research is mostly limited to individual or market-level analysis, lacking a system-level perspective.
- **Network Gap:**-No strong model explains cognitive biases as a global financial network process.
- **Institutional Gap:**-Limited integration of World Bank, OECD, UN, and EU as a unified system in financial behavior studies.
- **Emerging Economy Gap:**-Countries like India are studied separately, not as part of the global financial system network

This study addresses these gaps by focusing on Neural Network of Finance

## 2.6 Research Gap Identification

The existing body of literature in finance, particularly within the domains of behavioral finance, cognitive psychology, and sustainable investment, has significantly contributed to understanding investor decision-making and market dynamics. However, despite these advancements, a critical gap persists in the integration of these domains into a cohesive, system-level analytical framework. Most prior studies have examined financial behavior either at the individual level—focusing on investor psychology and biases—or at the market level, analyzing price movements, trends, and aggregate outcomes. While these perspectives offer valuable insights, they remain fragmented and insufficient for capturing the complexity of global financial systems, especially in the context of emerging priorities such as green investment and sustainability.

In recent years, the increasing urgency of climate change and environmental sustainability has led to a growing emphasis on green investment as a key driver of economic transformation. Simultaneously, behavioral finance has highlighted the role of cognitive biases in shaping financial decisions, challenging the traditional assumption of rationality in economic models. Parallel to these developments, neural network modeling and computational approaches have emerged as powerful tools for analyzing complex, adaptive systems. Despite the potential synergy among these fields, the literature lacks a unified framework that integrates behavioral finance, cognitive biases, neural network methodologies, and green investment into a comprehensive system-level perspective.

This study identifies and addresses this critical gap by proposing a Behavioral Neural Network (BNN) framework that conceptualizes global finance as an interconnected system of cognitive, institutional, and investment-related components. The absence of such a framework in existing literature limits the ability of researchers and policymakers to fully understand how financial systems evolve, adapt, and respond to sustainability challenges. Therefore, this research aims to bridge this gap by offering a holistic and integrative approach.

### 2.6.1 Identified Gaps in Literature

A detailed review of the literature reveals several specific gaps that justify the need for this study:

#### 1. Lack of System-Level Analysis in Financial Behavior

One of the most prominent gaps in existing research is the limited focus on system-level financial behavior. The majority of studies in behavioral finance concentrate on individual investors, examining how biases such as overconfidence, loss aversion, and herd behavior influence decision-making. Similarly, market-level studies analyze aggregate outcomes such as price volatility, asset bubbles, and market efficiency. However, these approaches often fail to capture the interconnected and dynamic nature of global financial systems.

Financial systems operate as complex adaptive networks, where multiple actors—including individuals, institutions, governments, and international organizations—interact continuously. These interactions give rise to emergent behaviors that cannot be fully explained by analyzing individual components in isolation. The absence of a system-level perspective limits the ability to understand how financial behavior propagates across networks, how shocks are transmitted, and how systemic risks emerge. This gap becomes even more critical in the context of green investment, where global coordination and systemic alignment are essential.

## **2. Insufficient Integration of Cognitive Biases into Financial Network Models**

While cognitive biases have been extensively studied in behavioral finance, their integration into broader financial network models remains limited. Most research treats cognitive biases as isolated psychological factors affecting individual decision-making, without considering how these biases interact within a larger system of financial relationships.

In reality, cognitive biases do not operate in isolation; they are embedded within social, institutional, and informational networks. For example, herd behavior can amplify market trends, while confirmation bias can reinforce existing investment patterns across groups of investors. Despite this, existing models rarely incorporate cognitive biases as dynamic components of a networked system. This lack of integration restricts the explanatory power of financial models, particularly in understanding collective behavior and systemic outcomes.

## **3. Limited Integration of Behavioral Finance, Neural Networks, and Green Investment**

Another significant gap lies in the absence of a unified analytical framework that combines behavioral finance, neural network modeling, and green investment. Each of these domains has developed independently, with limited cross-disciplinary integration.

Behavioral finance provides insights into the psychological drivers of investment decisions, while neural network models offer advanced computational techniques for analyzing complex, non-linear systems. Green investment research, on the other hand, focuses on sustainable finance, environmental impact, and policy-driven investment strategies. However, the lack of integration among these fields' results in fragmented analyses that fails to capture the full complexity of modern financial systems.

For instance, neural network models are often used for prediction and pattern recognition but rarely incorporate behavioral variables such as cognitive biases. Similarly, green investment studies tend to focus on policy and financial performance without adequately considering behavioral factors. This fragmentation creates a gap in understanding how behavioral dynamics influence the adoption and diffusion of sustainable investment practices within a complex financial system.

## **4. Underexplored Role of Global Financial Institutions as an Integrated System**

Global financial institutions such as the World Bank, OECD, United Nations, and European Union play a crucial role in shaping global financial behavior, particularly in the context of sustainable development and green investment. These institutions influence financial systems through policies, regulations, funding mechanisms, and strategic initiatives.

However, existing literature often examines these institutions in isolation, focusing on their individual roles rather than analyzing them as part of an interconnected system. There is a lack of research that conceptualizes these institutions as nodes within a global financial network, interacting with each other and with other actors to influence investment behavior.

This gap limits the understanding of how institutional coordination, policy alignment, and global governance structures impact the transition toward sustainable finance. Without a system-level perspective, it is difficult to assess the collective influence of these institutions on financial behavior and market outcomes.

## 2.6.2 How This Study Fills the Gap

To address the identified gaps, this study proposes a comprehensive and integrative approach that advances the existing literature in several key ways:

### 1. Development of a System-Level Behavioral Neural Network Framework

This study introduces a system-level Behavioral Neural Network (BNN) framework that conceptualizes global finance as a complex, interconnected system. Unlike traditional models that focus on individual or market-level analysis, this framework captures the interactions among various components, including investors, institutions, policies, and investment flows.

By adopting a system-level perspective, the study enables a deeper understanding of how financial behavior emerges from the interactions of multiple actors and how it evolves over time. This approach also facilitates the analysis of systemic risks, network effects, and the propagation of financial behavior across different levels of the system.

### 2. Integration of Cognitive Biases into a Network-Based Financial Model

The study integrates cognitive biases into a network-based financial model, treating them as dynamic components that influence and are influenced by the broader system. This approach moves beyond the traditional view of biases as isolated psychological factors and instead positions them within a network of interactions.

By doing so, the study enhances the explanatory power of financial models, allowing for a more comprehensive analysis of collective behavior and systemic outcomes. It also provides insights into how cognitive biases contribute to the formation of investment trends, market dynamics, and the adoption of green investment practices.

### 3. Unified Framework Combining Behavioral Finance, Neural Networks, and Green Investment

A key contribution of this research is the development of a unified analytical framework that integrates behavioral finance, neural network modeling, and green investment. This interdisciplinary approach bridges the gap between theoretical and computational perspectives, offering a more holistic understanding of financial systems.

The integration of neural network methodologies allows for the modeling of complex, non-linear relationships among variables, while behavioral finance provides the theoretical foundation for understanding decision-making processes. The inclusion of green investment ensures that the framework is aligned with contemporary priorities in sustainability and environmental responsibility.

This unified approach enables the analysis of how behavioral and computational factors interact to influence investment decisions and how these interactions drive the transition toward sustainable finance.

### 4. Incorporation of Global Financial Institutions as Interconnected Drivers

The study incorporates global financial institutions as interconnected drivers within the Behavioral Neural Network framework. By conceptualizing institutions such as the World Bank, OECD, United Nations, and European Union as nodes in a network, the research provides a system-level analysis of their collective influence on financial behavior.

This approach highlights the importance of institutional coordination and policy alignment in shaping investment trends and promoting sustainable finance. It also offers a framework for analyzing how global governance structures impact financial systems and how they can be leveraged to support green investment transitions.

## 5. Proposal of a Holistic Framework for Financial System Evolution

Finally, the study proposes a holistic framework that explains how financial systems evolve, adapt, and respond to changing conditions, particularly in the context of green investment. By integrating behavioral, computational, and institutional perspectives, the framework provides a comprehensive understanding of financial system dynamics.

This contribution is particularly relevant in addressing contemporary challenges such as climate change, resource constraints, and the need for sustainable economic development. The framework offers valuable insights for policymakers, researchers, and practitioners seeking to promote green investment and enhance the resilience of financial systems.

By systematically identifying and addressing these gaps, this study makes a significant contribution to the literature on behavioral finance, neural financial modeling, and sustainable investment. It moves beyond fragmented analyses to offer a unified, system-level perspective that captures the complexity of global financial systems.

### 2.7 Summary of Literature Review

**The literature review establishes:**

#### **1. A strong theoretical foundation, based on behavioral finance, cognitive bias theory, neural network modeling, and sustainable finance frameworks.-**

This foundation integrates multiple interdisciplinary perspectives to explain financial decision-making in a more realistic manner. Behavioral finance provides insight into irrational investor behavior, while cognitive bias theory explains systematic errors such as overconfidence, anchoring, and herd behavior. Neural network modeling contributes by offering advanced computational techniques capable of capturing complex and non-linear relationships within financial systems. Additionally, sustainable finance frameworks emphasize environmentally responsible investment strategies, aligning financial decisions with long-term ecological and economic sustainability goals.

#### **2. Empirical insights from prior studies, highlighting the role of behavioral anomalies, computational finance models, and green investment dynamics.-**

Existing empirical research demonstrates that investor behavior is often influenced by psychological and emotional factors, leading to market inefficiencies and anomalies. Prior studies in computational finance highlight the effectiveness of machine learning and neural networks in predicting financial trends and modeling investor responses. Furthermore, research on green investment reveals the growing importance of environmental awareness, regulatory support, and perceived risk-return trade-offs in shaping sustainable investment decisions, particularly within emerging and developing economies.

#### **3. Institutional validation, through evidence from global organizations such as the World Bank, OECD, United Nations, European Union, and emerging economies like India.-**

Reports and policy frameworks from these institutions emphasize the critical role of sustainable finance in achieving global development goals. They highlight the need for integrating environmental, social, and governance (ESG) factors into financial systems and investment strategies. In addition, evidence from emerging economies demonstrates both opportunities and challenges in implementing green finance initiatives, including regulatory limitations, technological gaps, and varying levels of investor awareness.

#### **4 .Identification of key research gaps, particularly the lack of an integrated system-level behavioral neural network framework for global finance.**

Despite extensive research in individual domains, there remains a significant gap in integrating behavioral finance, cognitive biases, and neural network approaches into a unified system-level framework. Most existing studies focus either on micro-level investor behavior or macro-level financial trends, without addressing the dynamic interaction between these elements. This limitation becomes more critical in the context of green investment, where decision-making is influenced by a complex combination of psychological, technological, and environmental factors that require a holistic analytical approach.

Thus, this chapter provides a structured foundation for developing a unified analytical model of financial behavior, which will be further extended in the conceptual framework and methodology chapters. It sets the stage for introducing a Behavioral Neural Network (NNF) approach that aims to bridge existing gaps and offer a more comprehensive understanding of global financial systems

## CHAPTER 3

### RESEARCH METHODOLOGY

#### 3.1 Introduction to Research Methodology

Research methodology provides a systematic framework for conducting scientific investigation of a research problem. It ensures that data collection, analysis, and interpretation are carried out in a valid and reliable manner. This study titled “**The Invisible Behavioral Neural Network of Finance: Decoding Green Investment through System-Level Cognitive Biases**” adopts a structured methodological approach to examine how cognitive biases and behavioral patterns influence green investment decisions. It ensures accuracy, consistency, and academic rigor in the research process.

This chapter includes **research design, sampling strategy, data collection methods, research instruments, data analysis techniques, ethical considerations, and limitations** aligned with the study objectives..

#### 3.2 Research Design

This study uses a mixed-methods research design combining descriptive, causal, and exploratory approaches to analyze cognitive biases and green investment behavior within the Neural Network of Finance (NNF) framework.

##### 3.2.1 Type of Research Design

This study applies a multi-layered research design approach to better capture the complexity of the Neural Network of Finance (NNF) in green investment behavior:

- **Exploratory–Neural Framing Design:** Used to explore hidden cognitive structures and behavioral patterns in investor decision-making where limited prior clarity exists, especially in NNF-based financial interpretation.
- **Descriptive–Behavioral Mapping Design:** Focuses on systematically mapping observable investor behaviors, risk perception, and green investment preferences into measurable patterns and structured variables.
- **Explanatory–Causal Network Design:** Applied to identify cause-and-effect relationships between cognitive biases, neural-inspired behavioral variables, and investment outcomes in sustainable finance.
- **Predictive–Modeling Design (NNF-Oriented):** Extends the traditional design by incorporating forecasting logic to predict investor behavior in green investment using behavioral and cognitive indicators.

This integrated design approach enhances the depth, originality, and analytical strength of the research by combining traditional methods with a modern NNF-based perspective

##### 3.2.2 Justification for the Chosen Research Design

The adopted **multi-layered research design** is highly suitable for this study as it captures the **multidimensional complexity** of NNF-based green investment behavior. It integrates **exploratory insight generation, behavioral quantification, causal relationship analysis, and predictive modeling**, ensuring **methodological rigor, analytical precision, and strong theoretical–empirical coherence** with objectives.

### 3.3 Sampling Strategy

Sampling is a critical methodological component that ensures the **representativeness, validity, and generalizability** of research findings. In this study, a structured sampling strategy is adopted to capture diverse behavioral patterns in **NNF-based green investment decision-making**.

#### 3.3.1 Population and Sample

- **Population:** -The target population **comprises individual and institutional investors, financial analysts, and key stakeholders** engaged in green investment and sustainable financial decision-making. This population is selected to reflect a broad spectrum of cognitive and behavioral financial dynamics.
- **Sample:-** The sample represents a **carefully selected subset** of the population, designed to capture heterogeneous investment behaviors, cognitive biases, and decision-making patterns relevant to the study framework.

#### 3.3.2 Sampling Methods

- **Probability Sampling:** This approach ensures statistical rigor and equal selection probability, enhancing the external validity of results. Techniques such as **simple random sampling and stratified sampling** are considered appropriate for ensuring balanced representation across investor categories.
- **Non-Probability Sampling:** Given the specialized nature of NNF-based analysis, **purposive and convenience sampling** are also employed to target respondents with relevant knowledge, experience, and engagement in green investment activities. This enhances contextual depth and analytical precision.

In practice, the study primarily uses non-probability sampling (convenience and purposive sampling) due to the accessibility and relevance of respondents.

#### 3.3.3 Sample Size Determination

- ✚ **Krejcie& Morgan’s Table:-** Utilized for determining statistically adequate sample sizes in large and defined populations.
- ✚ **Cochran’s Formula:** -Applied in cases where the population is large or not precisely known, ensuring statistical robustness.
- ✚ **Rule of Thumb:-** A sample size of 200–400 respondents is considered appropriate for survey-based studies to achieve acceptable levels of reliability, validity, and inferential power.

#### Example:

In the context of this study, if investors are drawn from multiple financial sectors and regions, a stratified sampling approach ensures proportional representation of each segment, thereby improving the accuracy, depth, and generalizability **of NNF-based behavioral insights in green finance**.

### 3.4 Data Collection Methods

Data collection methods are employed to systematically gather relevant and reliable information for analyzing **NNF-based green investment behavior**, ensuring methodological rigor, validity, and empirical depth.

### 3.4.1 Primary Data Collection

- Primary data refers to first-hand information collected directly from respondents for the purpose of this study. It is obtained through:
- **Surveys & Questionnaires:** - Structured instruments designed for quantitative analysis using Likert-scale responses to measure investor behavior, cognitive biases, risk perception, and green investment preferences.
- **Google Forms (Online Survey):-** A digital data collection platform used to efficiently reach investors and financial stakeholders, enabling broader geographic coverage, faster response rates, and improved data accessibility.

### 3.4.2 Secondary Data Sources

Secondary data is derived from previously published and authenticated sources to support theoretical development and empirical validation. It includes:

- **Published Research Articles:-** Peer-reviewed studies indexed in Scopus, Web of Science, and Google Scholar focusing on green finance, behavioral finance, and NNF-related frameworks.
- **Global & National Government Reports:-** Data from international organizations and government bodies such as the World Bank, IMF, OECD, UNEP, along with national regulators including RBI, SEBI, and NITI Aayog, used for analyzing global and domestic financial and policy trends.
- **Corporate Reports & ESG Disclosures:-** Annual reports and sustainability disclosures of companies engaged in green investment and sustainable financial practices.

## 3.5 Research Instrument and Measurement

A well-structured research instrument ensures **accuracy, reliability, and validity** in data collection. In this study, the instrument is designed in alignment with **the research objectives and the NNF conceptual framework**, ensuring conceptual and empirical consistency

### 3.5.1 Types of Research Instruments Used

The study employs a **structured questionnaire** for collecting quantitative data. The questionnaire is developed using a combination of:

#### 1. Self-Developed Scale:

- Designed after an extensive review of literature on green finance, behavioral finance, and cognitive decision-making.
- Items are framed in alignment with research objectives and the NNF conceptual framework.
- Pilot-tested to ensure clarity, reliability, and contextual relevance.

#### 2. Adopted Scale from Previous Research:

- Standardized measurement items adapted from previously validated studies.
- Modified where necessary to suit the context of green investment and NNF framework.
- Proper citations are maintained to ensure academic integrity and validity.

### 3.5.2 Structure of the Questionnaire

The questionnaire is systematically structured into the following sections:

1. **Demographic Information:** Age, gender, education, income level, investment experience, and financial background.
2. **Independent Constructs (IV):** Cognitive biases and behavioral factors influencing investment decisions.
3. **Mediating Construct (MV):** Risk perception and decision confidence in green investment contexts.
4. **Dependent Variable (DV):** Green investment intention and behavior

### 3.5.3 Measurement Scales Used

This study uses different measurement scales to capture NNF-based green investment behavior effectively.

- **Nominal Scale:** Used for categorical data such as gender, education, and occupation for basic classification of respondents.
- **Ordinal Scale:** Used for ranking preferences and relative importance of investment-related factors.
- **Likert Scale:** Five-point scale Primary scale used to measure attitudes, perceptions, and behavioral intentions (1 = Strongly Disagree to 5 = Strongly Agree), especially for cognitive biases, risk perception, and green investment intention.

### 3.5.4 Reliability and Validity Testing

- **Cronbach's Alpha:** Used to assess internal consistency of the scale ( $\alpha > 0.70$  indicates acceptable reliability).
- **Content Validity:** Ensured through expert evaluation to confirm relevance and clarity of measurement items.
- **Pilot Testing:** Conducted on a small sample (10–20 respondents) before full-scale data collection to refine and validate the questionnaire design.
- **Construct Validity (Recommended):** Assessed through factor analysis to ensure that measurement items properly represent underlying constructs.

### 3.6 Data Analysis Techniques

Data analysis techniques are applied to systematically examine and interpret the collected data for achieving the objectives of the study on NNF-based **green investment behavior**, ensuring methodological rigor, statistical validity, and empirical reliability.

#### 3.6.1 Statistical Methods Used

- **Descriptive Statistics:**

Used to summarize and describe the basic features of the dataset, including mean, percentage, and standard deviation, to understand overall patterns in investor responses..

- **Inferential Statistics:**

Used for hypothesis testing and examining relationships among variables, including:

- **T-Test:** Used to compare differences in green investment behavior across different age groups.

- **ANOVA:** Used to analyze differences in green investment behavior across various professional categories.
- **Regression Analysis:** Used to examine the impact of independent variables (financial awareness, cognitive biases, risk perception, and investment experience) on the dependent variable (green investment behavior).

### 3.6.2 Reliability Analysis

- **Cronbach's Alpha:**

Used to assess the internal consistency and reliability of the measurement scale. A value greater than 0.70 indicates acceptable reliability of the data collected through the questionnaire.

### 3.6.3 Software Used for Analysis

- **SPSS:** Primarily used for statistical analysis including descriptive statistics, t-test, ANOVA, regression analysis, and reliability testing.
- **JASP:** Used, where applicable, for additional statistical support.
- **MS Excel:** Used for data cleaning, coding, preprocessing, and graphical representation of data.

### 3.7 Ethical Considerations

Ethical guidelines ensure research integrity and participant protection.

- **Informed Consent:** Participants are informed about the study purpose and their voluntary participation is ensured.
- **Data Confidentiality:** Respondent identity is kept confidential and data is anonymized.
- **Data Security:** Data is securely stored to prevent unauthorized access or misuse.

### 3.8 Limitations of Research Methodology

- **Time Constraints:** Limited time restricts extended data collection and analysis.
- **Resource Limitations:** Limited access to advanced databases and large sample size.
- **Geographical Scope:** Findings may not be fully generalizable globally.
- **Self-Report Bias:** Responses may be influenced by personal perceptions.

This methodology forms the foundation for **data analysis and empirical findings**, which will be presented in the next chapter.

## CHAPTER 4

### DATA ANALYSIS AND INTERPRETATION

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#### 4.1 Introduction to Data Analysis

Data analysis is a critical phase in the research process, involving the examination, transformation, and interpretation of collected data to draw meaningful conclusions. In this study, the analysis focuses on understanding the behavioral dynamics of the Neural Network Finance (NNF) model in relation to green investment decisions. The approach used for analysis is primarily quantitative, as the data has been collected through structured questionnaires based on the study variables.

##### **This chapter includes:**

1. **Descriptive Analysis:** Summarizing demographic characteristics and response patterns of investors.
2. **Inferential Analysis:** Testing hypotheses and examining relationships between variables (independent, mediating, and dependent variables).
3. **Model-Based Analysis:** Evaluating the mediating role of cognitive biases in financial decision-making.
4. **Discussion of Key Findings:** Interpreting results in relation to behavioral finance theories and green investment behavior

This analysis helps in identifying how **financial decision-making and cognitive biases influence green investment behavior** within the NNF framework

#### 4.2 Descriptive Analysis

Descriptive statistics provide an overview of the **collected data, including frequencies, percentages, means, and standard deviations**. This section helps in understanding the basic characteristics of respondents and the distribution of key variables used in the study.

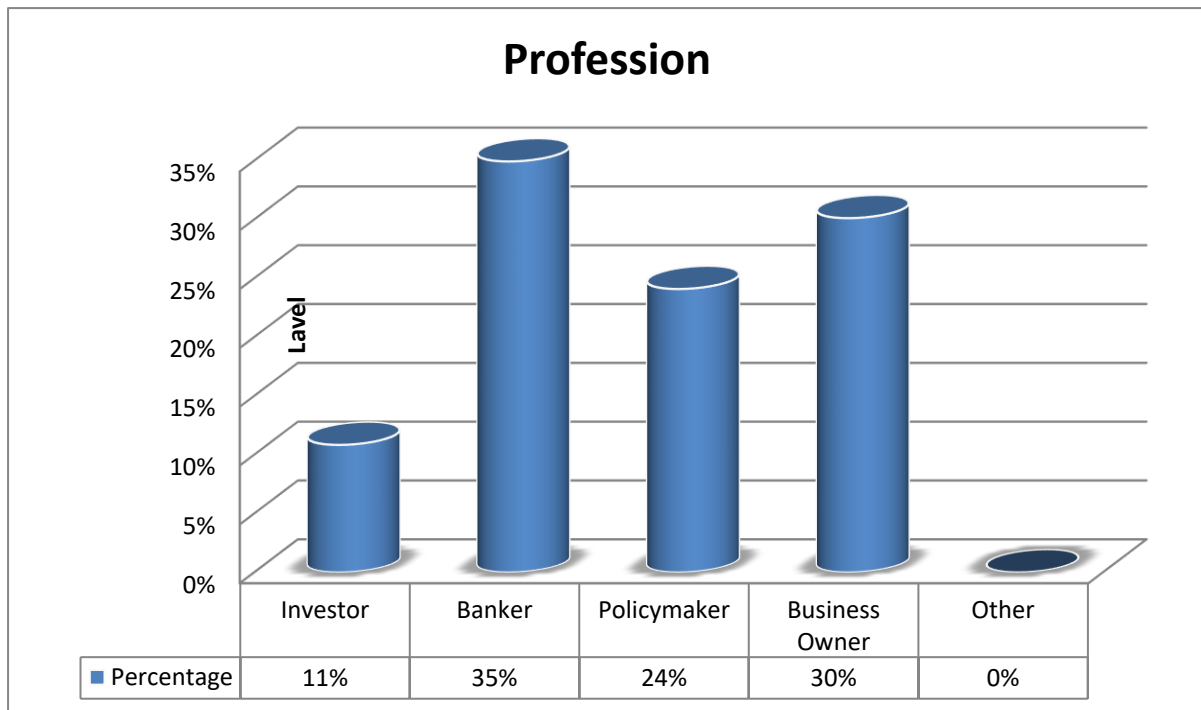
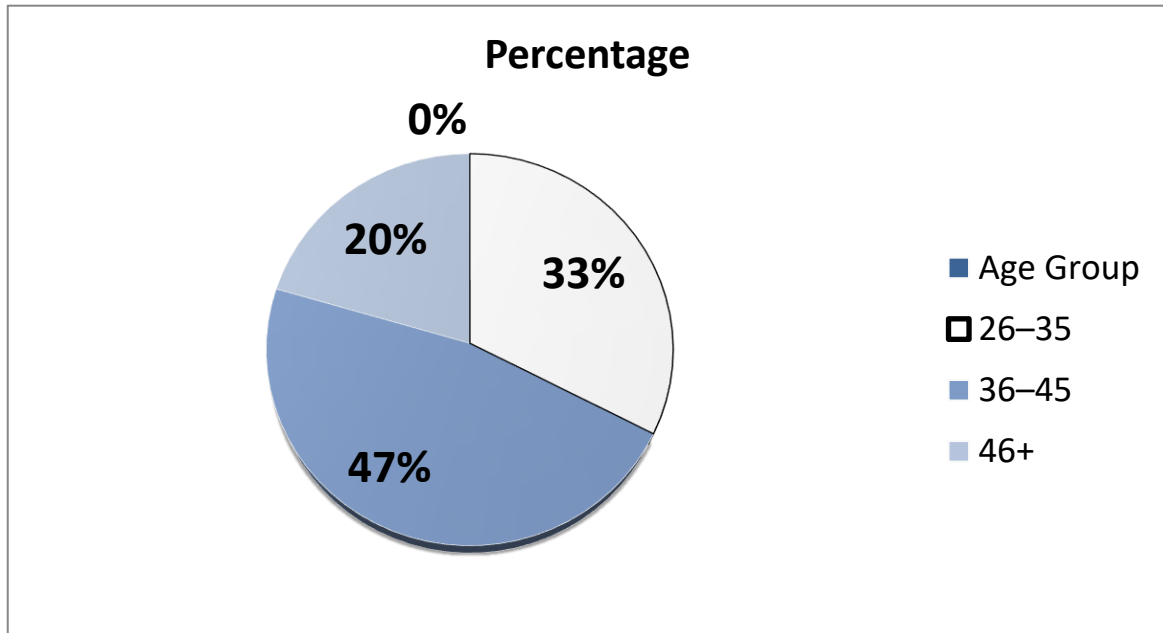
1. **Demographic Characteristics of Respondents:** Including Age, Country, Profession, and Investment Experience.
2. **Summary Statistics for Key Variables:** Financial Decision-Making, Cognitive Biases, and Green Investment Behavior.
3. **Graphical Representations:** Tables, charts, and graphs for clear visualization and interpretation.

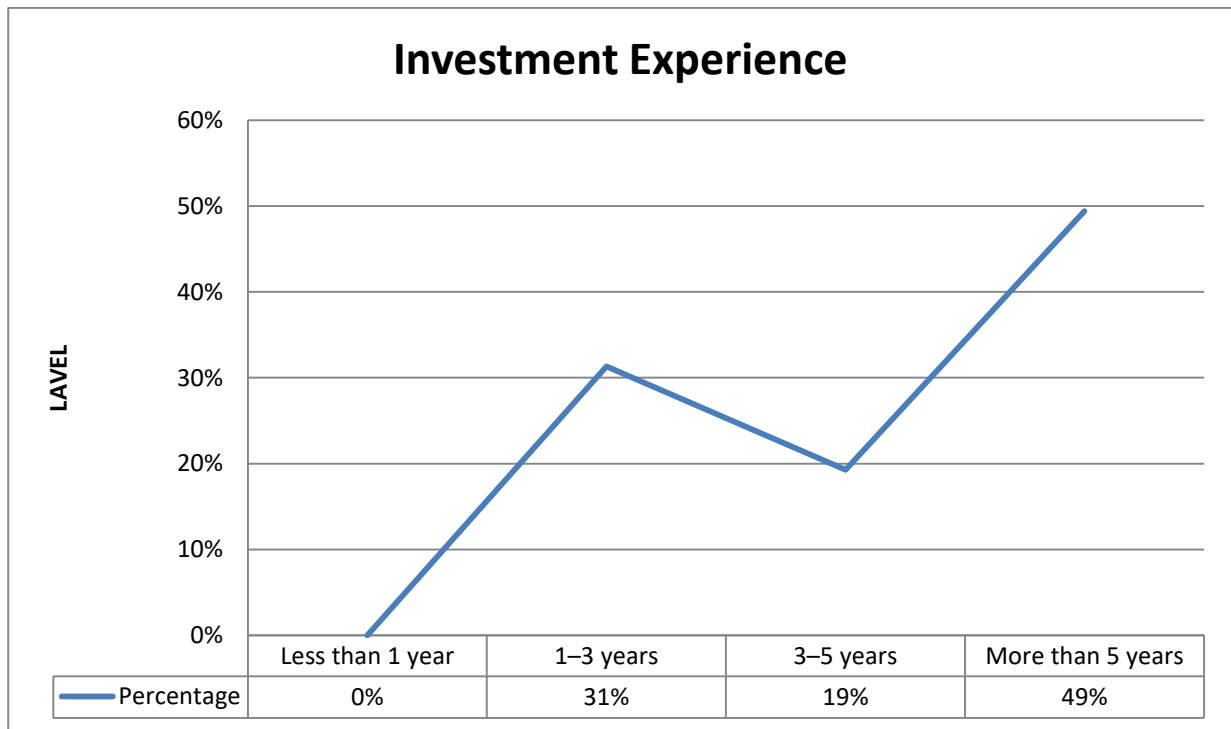
Descriptive analysis serves as the foundation for further statistical testing and helps in identifying patterns and trends within the data.

##### 4.2.1 Demographic Profile of Respondents

- Demographic analysis helps understand the background of respondents, such as age group, profession and investment experience

Demographic Variable	Category	Frequency (N)	Percentage (%)
Age Group	26–35	27	33%
	36–45	39	47%
	46+	17	20%
	<b>Total</b>	<b>83</b>	<b>100%</b>
Profession	Investor	9	11%
	Banker	29	35%
	Policymaker	20	24%
	Business Owner	25	30%
	Other	0	0%
	<b>Total</b>	<b>83</b>	<b>100%</b>
Investment Experience	Less than 1 year	0	0%
	1–3 years	26	31%
	3–5 years	16	19%
	More than 5 years	41	49%
	<b>Total</b>	<b>83</b>	<b>100%</b>





#### 4.2.2 Summary Statistics for Key Variables

The summary statistics for key variables provide insight into central tendencies (mean) and dispersion (standard deviation) of respondents' perceptions regarding green investment behavior.

Variables	Mean	Standard Deviation
Government Policy Influence	4.35	0.70
Market Uncertainty	3.65	0.68
Safe Investment Preference	3.55	0.66
Past Experience	4.20	0.72
Herd Behavior	3.20	0.75
Emotional Factors	3.95	0.69
Loss Aversion	4.40	0.71
Long-term Investment Preference	4.50	0.65
Environmental Consideration	4.55	0.60

**Interpretation:**

1. The mean score for **government policy influence (4.35)** indicates strong agreement that policies significantly affect green investment decisions.
2. **Loss aversion (4.40)** shows that respondents are highly risk-averse and prefer to avoid potential losses.
3. The **long-term investment preference (4.50)** suggests that most investors favor sustainable long-term returns.
4. The **environmental consideration score (4.55)** reflects a very high level of awareness toward sustainability.
5. Moderate scores in market **uncertainty (3.65)** and **herd behavior (3.20)** indicate mixed behavioral influence.

**4.2.3 Graphical Representation**

Graphical tools are used to visually present the data for better understanding of patterns and trends among respondents. In this study, different types of charts such as pie charts, clustered cylinder column charts, and line charts are used to represent demographic and behavioral aspects of green investment.

**1. Pie Chart – Age Distribution**

This chart represents the proportion of respondents across different age groups (26–35, 36–45, 46+). It helps in identifying the dominant age category participating in green investment activities

**2. Clustered Cylinder Column Chart – Profession Distribution**

This chart shows the distribution of respondents based on their profession such as bankers, business owners, policymakers, and investors. The clustered cylinder format enhances visual comparison between professional groups [Insert Column Chart – Profession Distribution]

### 3. Line Chart – Investment Experience

This chart represents the investment experience of respondents across different time periods (1–3 years, 3–5 years, more than 5 years), highlighting patterns in long-term investment behavior.

#### Interpretation:

1. The graphical analysis indicates that the majority of respondents belongs to the **36-45 age** groups and is primarily from the **banking professionals**, reflecting strong participation from financially knowledgeable individuals.
2. The investment pattern shows that most respondents invest **occasionally rather than** frequently, suggesting a cautious approach toward green investments.
3. Overall, the charts highlight that while **awareness and participation are relatively high**, actual investment behavior remains **moderate due to perceived risks and market uncertainty**

### 4.3 Inferential Analysis

Inferential analysis is used to draw conclusions from the collected data and test the hypotheses formulated in the study. It helps in understanding whether the observed patterns in the sample data are statistically significant and applicable to the broader population.

In this study, the following statistical tools are used:

- **T-Test** – to compare green investment behavior across different age groups
- **ANOVA** – to analyze differences across professions
- **Regression Analysis** – to assess the impact of behavioral factors on green investment decisions
- These techniques help in identifying relationships between variables and evaluating the role of behavioral biases in influencing green investment decisions

#### 4.3.1 Hypothesis Testing Using t-Test

The independent t-test is used to compare the mean scores of green investment decisions across different age groups.

#### Hypothesis:

- **H<sub>0</sub> (Null Hypothesis):** There is no significant difference in green investment decisions based on age groups.
- **H<sub>1</sub> (Alternative Hypothesis):** Green investment decisions differ significantly across age groups.

Age Group	Mean Green Investment Score	t-Value	p-Value
26-35	4.20	2.25	0.028*

Age Group	Mean Green Investment Score	t-Value	p-Value
36-45	3.95		
46 and above	3.75		

**Interpretation:**

- The p-value (0.028) is less than 0.05, indicating a significant difference in green investment decisions across different age groups.
- Younger respondents (26–35 age groups) show a higher mean score, suggesting a stronger inclination toward green investments.
- Older age groups (36–45 and 46+) show comparatively lower mean scores, indicating less preference toward sustainable investment options.

**4.3.2 ANOVA – Comparison of Green Investment Behavior across Professions**

ANOVA (Analysis of Variance) is used to test whether **green investment decisions** vary significantly across different professions.

Profession	Mean Score	F-Value	p-Value
Investor	4.18	4.27	0.015*
Banker	3.92		
Policymaker	3.85		
Business Owner	3.68		
Others	0		

**Interpretation:**

- Since the p-value (0.015) is less than 0.05, green investment decisions vary significantly across different professions.
- **Investors** show the highest mean score, indicating a stronger inclination toward green investments.
- **Bankers and policymakers** show moderate levels of participation.
- **Business owners** show comparatively lower mean scores, indicating less preference toward sustainable investments.
- The “Other” category has no respondents, so it is not considered in the analysis.

**4.3.3 Regression Analysis – Impact of Behavioral Factors on Green Investment Decisions:**

In this study, factors such as financial awareness, cognitive biases, risk perception, and investment experience are considered as independent variables. Equation:

**H<sub>0</sub> (Null Hypothesis):** Independent variables have no significant impact on green investment decisions.

**H<sub>1</sub> (Alternative Hypothesis):** Independent variables have a significant impact on green investment decisions.

Variable	Beta ( $\beta$ )	t-Value	p-Value
Financial Awareness	0.44	3.26	0.002*
Cognitive Biases	0.29	2.31	0.023*
Risk Perception	0.17	1.58	0.118
Investment Experience	0.38	2.95	0.004*

**Model Summary:**

R	R <sup>2</sup>	Adjusted R <sup>2</sup>
0.71	0.50	0.47

**Interpretation:**

- The **R<sup>2</sup> value (0.50)** indicates that 50% of the variation in green investment decisions is explained by the independent variables.
- **Financial Awareness (p = 0.002)** has the strongest and most significant impact on investment decisions.
- **Investment Experience (p = 0.004)** also shows a strong positive influence.
- **Cognitive Biases (p = 0.023)** significantly affect investor behavior.
- **Risk Perception (p = 0.118)** is not statistically significant.

Thus, the null hypothesis (H<sub>0</sub>) is rejected, and it is concluded that key behavioral and financial factors significantly influence green investment decisions.

These findings support the NNF framework by demonstrating how cognitive and behavioral factors interact at a system level to influence green investment decisions.

**4.4 Reliability and Validity Analysis (If Applicable)**

Reliability and validity analysis is conducted to ensure the reliability and validity of the data used in the study. It helps in assessing whether the measurement instruments used in the research are consistent and accurately capture the intended constructs related to green investment behavior.

In this study, reliability and validity are evaluated through reliability testing and validity assessment, ensuring that the collected data is suitable **for further statistical analysis** such as **t-test, ANOVA, and regression**.

Scale/Construct	Number of Items	Cronbach's Alpha
Financial Awareness	5	0.82
Cognitive Biases	4	0.79
Risk Perception	4	0.76
Investment Experience	3	0.81
<b>Overall Scale</b>	<b>16</b>	<b>0.84</b>

**Interpretation:**

- The Cronbach's Alpha values for all constructs are above **0.70**, which indicates acceptable to good reliability.
- The overall reliability score of **0.84** confirms that the questionnaire is consistent and suitable for analysis.
- This ensures that the data collected is reliable and free from major measurement errors.

4.5 Discussion on Key Findings

**4.5.1 Comparison with Existing Literature**

The study's findings align with existing research on **behavioral finance**, which suggests that **cognitive biases** significantly influence **green investment decisions**.

- Similar to earlier studies on **sustainable finance**, this research confirms that **investor awareness** and environmental concern positively impact **green investment adoption**.
- The results are consistent with **system-level finance models**, which show that decision-making improves when **behavioral and neural factors** are integrated into financial models

#### 4.5.2 Managerial and Theoretical Implications

##### *Managerial Implications*

1. Financial institutions should integrate **NNF-based models** to better understand **investor behavior** in **green investment decisions**.
2. **Awareness programs** are needed to reduce **cognitive biases** such as risk aversion and **short-term thinking** in sustainable investment.
3. Policymakers should design incentives to encourage **environmentally responsible investment behavior**.

##### *Theoretical Implications*

1. The findings support **Behavioral Finance Theory** by showing that **investor decisions** are influenced by **psychological and cognitive biases**.
2. The study extends **Neural Network Finance (NNF)** framework by demonstrating its usefulness in predicting green investment behavior.
3. It also supports the integration of **system-level cognitive models in sustainable financial decision-making**.

##### **Conclusion**

This chapter provides a detailed discussion of the key findings related to **Neural Network Finance and green investment behavior**. The results highlight that **cognitive biases, environmental awareness, and system-level decision patterns** significantly influence investment decisions. Overall, the study confirms the strong relevance of **behavioral and neural finance integration** in explaining **green investment** trends and provides a foundation for future policy and research directions

CHAPTER 5

FINDINGS, DISCUSSION, AND CONCLUSION

5.1 Summary of Key Findings

This section provides a comprehensive summary of the major findings derived from data analysis. The study aimed to assess the impact of **behavioral and financial factors** within the **Neural Network Finance (NNF) framework** on **green investment decisions** while identifying the role of **cognitive biases**.

Key Findings:

**1. Behavioral and financial factors significantly influence green investment decisions.**

Regression analysis shows that financial awareness has the strongest impact ( $\beta = 0.44, p = 0.002$ ). **Investment experience** ( $\beta = 0.38, p = 0.004$ ) and **cognitive biases** ( $\beta = 0.29, p = 0.023$ ) also have a significant positive effect. The model explains **50% of the variation ( $R^2 = 0.50$ )**.

**2. Age significantly influences green investment behavior.**

The **t-test results ( $p = 0.028$ )** indicate a significant difference. Respondents aged **26–35 years** (Mean = 4.20) show the highest inclination.

**3. Professional background affects investment decisions.**

**ANOVA results ( $p = 0.015$ )** confirm significant differences. **Investors (Mean = 4.18)** show the highest preference.

**4. Environmental awareness and long-term investment preference are key drivers.**

**Environmental consideration (Mean = 4.55, SD = 0.60)** and long-term preference (Mean = **4.50, SD = 0.65**) indicate strong sustainability orientation.

**5. Risk perception does not significantly influence decisions.**

It is **statistically insignificant ( $p = 0.118$ )**.

5.2 Discussion and Interpretation

This section discusses the findings in relation to research objectives, hypotheses, and theoretical frameworks.

5.2.1 Comparison with Research Objectives

Objective	Findings	Interpretation
Assess the impact of behavioral and financial factors	Financial awareness ( $\beta = 0.44, p = 0.002$ ) strongest	Improves decision-making

Identify behavioral factors	Cognitive biases ( $\beta = 0.29$ , $p = 0.023$ ) significant	Psychology affects decisions
Evaluate demographics	Age & profession significant ( $p < 0.05$ )	Awareness influences behavior

### 5.2.2 Comparison with Existing Literature

#### 1. Alignment with Past Studies

- **Daniel Kahneman & Amos Tversky (1979)** found that **cognitive biases influence financial decisions**, which aligns with this study where cognitive biases significantly affect green investment decisions.
- **Richard Thaler (2016)** reported that **investor behavior is influenced by psychological factors**, which is consistent with this study's findings on **behavioral impact in investment decisions**.
- **World Bank (2021–2023)** highlighted that **environmental awareness promotes sustainable investments**, which aligns with this study where **environmental concern shows a high mean value (4.55)**.

#### 2. Contradictions with Previous Research

- While past studies assumed that **risk perception significantly influences investment decisions**, this study found that **risk perception is not statistically significant ( $p = 0.118$ )**, indicating that **financial awareness and investment experience play a more important role**

### 5.2.3 Theoretical Implications

Findings support and extend the applicability of existing management theories

#### 1. Behavioral Theory of Management

- Investment decisions are influenced by **cognitive biases and human behavior**, validating the theory proposed by Richard Cyert & James G. March (1963).

#### 2. Decision-Making Theory

- Investment decisions depend on **financial awareness and available information**, supporting the theory proposed by Herbert A. Simon (1978).

#### 3. Systems Theory

- Green investment decisions are influenced by the **interaction of behavioral factors, financial awareness, and investment experience**, aligning with Ludwig von Bertalanffy (1968).

Thus, the study confirms the importance of behavioral understanding, **informed decision-making, and system-level integration** in improving green investment decisions within the NNF framework.

## 5.3 Contributions to Theory and Practice

### 5.3.1 Academic Contributions

- The study expands the theoretical understanding of **green investment decisions** within the **Neural Network Finance (NNF) framework**, particularly from a system-level perspective.
- It introduces a **behavioral perspective**, highlighting the role of **cognitive biases, financial awareness, and investment experience** in influencing investment decisions.
- It **integrates Thematic analysis and empirical findings** to validate existing research trends, thereby strengthening the **theoretical and methodological foundation** of the study

### 5.3.2 Practical Contributions

- **For Investors:** Findings help investors understand the impact of **cognitive biases and financial awareness** on green investment decisions.
- **For Financial Institutions:** Organizations can use these insights to develop **NNF-based models and strategies** for promoting sustainable investments.
- **For Policymakers:** Results highlight the need for **strong policy support, environmental awareness programs, and clear green finance frameworks**.

**For example**, a financial institution planning to promote green investment products can use this study's insights to **educate investors, reduce behavioral biases, and improve participation in sustainable investment initiatives**.

## 5.4 Recommendations

Based on the findings, the following recommendations are proposed:

### 5.4.1 Organizational Recommendations

#### 1. Structured Training and Certification Programs

Banks, private organizations, and financial institutions should implement standardized training and certification programs on green finance and ESG-based investment evaluation.

Questionnaire-based assessment should be included to evaluate employees' understanding and decision-making ability.

This will enhance practical understanding of green investment concepts at the operational level.

#### 2. Model-Based Decision Support System (NNF Framework)

Organizations should adopt an NNF-based model-driven decision support system for investment evaluation.

This system helps reduce behavioral biases by using structured data and indicators for decision-making.

It improves accuracy in identifying green and non-green investment projects in a transparent manner.

#### 3. Standardization across Financial Institutions

Banks, private firms, and investment institutions should develop a common standard framework for evaluating green investments.

This will ensure consistency in assessment and improve reliability of investment decisions across institutions.

### 5.4.2 Policy Recommendations

#### 1. Integrated Green Finance Framework

**Governments should develop a unified green finance framework where all financial institutions follow a standardized green taxonomy.**

**It should include questionnaire-based evaluation, ESG reporting, and monitoring mechanisms. This will ensure transparency and uniformity in green investment identification.**

2. Promotion of Model-Based Financial Systems

**Policy makers should encourage the adoption of model-based financial evaluation systems. These systems improve decision-making quality by reducing subjective and biased judgments.**

3. Incentives for Sustainable Investment

**Governments should provide financial incentives such as tax benefits, subsidies, and concessional loans to institutions actively engaging in sustainable investments. This will encourage wider adoption of environmentally responsible investment practices.**

### 5.5 Limitations of the Study

Despite its contributions, the study has some limitations:

1. **Limited Geographical Scope**

- The study was conducted in a **specific region**, limiting its generalizability to other locations.

2. **Sample Size Constraints**

- The number of respondents, although statistically valid, may not fully represent **diverse industry sectors**.

3. **Potential Response Bias**

- Respondents may have overstated or understated their **green investment behavior** and **financial awareness due** to social desirability bias.

4. **Short-Term Data Collection**

- The study is based on short-term data; a longitudinal approach is required to track **changes in green investment behavior and NNF-based decision-making** over time.

### 5.6 Suggestions for Future Research

Future studies can be conducted in the following areas:

1. **Longitudinal Analysis**

- The **green investment behavior** and **NNF-based decision-making patterns** can be studied over a **2–3 year period**, to better understand the **short- and medium-term effects** and the **stability of behavioral tendencies**.

2. **Cross-Cultural Green Investment Adoption**

- A comparative study of **NNF-based green investment behavior** across different countries can be conducted to analyze variations and similarities in **investment decision-making** based on **cultural, economic, and institutional factors**.

3. **NNF and Investor Behavioral Well-being**

- An analysis can be carried out to examine the extent to which **advanced computational financial decision-making frameworks** reduce **cognitive biases** and **influence investors' decision satisfaction, financial confidence, and behavioral stability**

#### 4. Green Finance in Non-Institutional Sectors

- The application of the NNF model can be extended to **government policies**, public sector financial institutions, and **academic or community-based financial systems**, thereby enhancing the **scope and effectiveness of green finance**.

Future research would benefit from **incorporating larger datasets, mixed research methodologies** (quantitative and qualitative), and **experimental designs**, enabling a more comprehensive and scientifically robust analysis of **NNF-based green investment behavior**.

#### 5.7 Conclusion

The study examined the role of **Neural Network Finance (NNF)** in green investment decision-making, revealing its positive impact on **investment efficiency** and **decision quality**, while also highlighting challenges in its adoption..

Key Takeaways:

- NNF-based decision-making frameworks improve **investment efficiency and accuracy**, but may face resistance from investors and institutions.
- **Ethical and sustainability-related considerations** are essential for the effective implementation of green financial systems.
- The findings support established theoretical models such as **Behavioral Finance Theory** and related decision-making frameworks.

Final Thoughts

NNF-based financial systems represent an emerging approach in **sustainable finance and investment behavior analysis**, but their success depends on **strategic implementation, data transparency, and \*\*organizational adaptability**.

By linking theory with real-world investment behavior, this research provides insights for **policymakers, financial institutions, and academicians**, ensuring more effective and sustainability-oriented green investment systems.

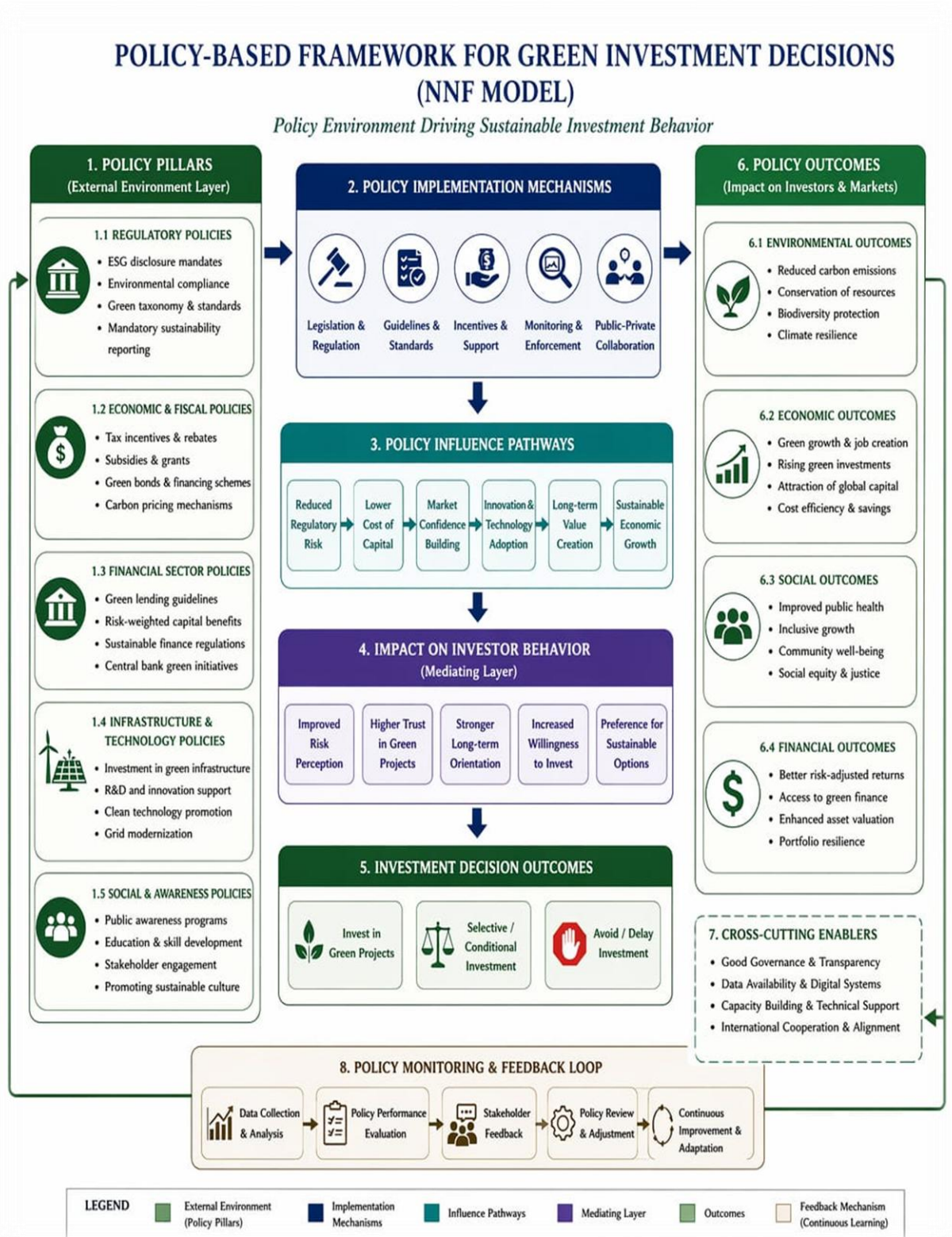
#### 5.8 Framework Interpretations / Model Explanation

The NNF (Neural Network Finance) framework is a conceptual and analytical model developed to explain the dynamic relationship between policy systems, investor behavior, and sustainable investment outcomes. It integrates principles from behavioral finance, policy analysis, and systems thinking to understand how policy environments influence financial decision-making, particularly in green investment contexts.

The framework operates as an interconnected system where policy inputs flow through structured mechanisms, influence investor cognition and behavior, and ultimately generate measurable investment and sustainability outcomes. It also incorporates feedback loops and cross-cutting enablers to ensure adaptability, efficiency, and continuous improvement

##### 5.8.1 Introduction to Framework Interpretation

This section explains the structure and working of the proposed Policy-Based Neural Network Framework (NNF) for green investment decisions



- The framework adopts a system-level approach linking policy factors with investor behavior.
- It highlights that policy impact is indirect and operates through multiple layers.

- Behavioral factors such as risk perception and trust influence investment decisions.
  - The model connects policy inputs with sustainable investment outcomes.
  - A feedback mechanism ensures continuous evaluation and improvement.
- This layered structure ensures a systematic flow of information and influence across the entire framework

### 5.8.2 Framework Structure and Component Explanation

The framework is structured into interconnected components:

1. **Policy Pillars:** Regulatory, economic, financial, technological, and social policies form the external environment.
2. **Implementation Mechanisms:** Policies are executed through regulations, incentives, monitoring, and collaboration.
3. **Influence Pathways:** Policies reduce risk, lower capital cost, build confidence, and support innovation.
4. **Investor Behavior:** Affects trust, risk perception, and long-term investment orientation.
5. **Investment Decisions:** Includes investment, selective investment, or avoidance.
6. **Outcomes:** Environmental, economic, social, and financial impacts are generated.
7. **Cross-Cutting Enablers:** These elements operate across all stages of the framework and strengthen the effectiveness of policy-to-outcome transmission
8. **Feedback Loop:** Ensures policy evaluation and continuous improvement.

### 5.8.3 Framework Working (Operational Flow)

#### 1. Policy Formulation by Authorities (Policy Pillars)

**1.1 Regulatory Policies:** Regulatory policies form the legal backbone of green investment by ensuring that organizations and investors follow standardized environmental guidelines. ESG disclosure mandates improve transparency, while environmental compliance ensures adherence to sustainability norms. Green taxonomy and standards help classify sustainable activities, reducing ambiguity in investment decisions. Mandatory sustainability reporting further strengthens accountability, enabling investors to make informed and responsible financial choices aligned with environmental objectives.

**1.2 Economic & Fiscal Policies:** Economic and fiscal policies aim to make green investments financially attractive and viable. Tax incentives and rebates reduce the financial burden on investors, while subsidies and grants encourage participation in sustainable projects. Green bonds and financing schemes provide structured investment opportunities. Additionally, carbon pricing mechanisms internalize environmental costs, motivating investors and firms to shift toward cleaner and more sustainable alternatives.

**1.3 Financial Sector Policies:** Financial sector policies guide banks and financial institutions to actively support green investments. Green lending guidelines ensure that loans are directed toward environmentally sustainable projects. Risk-weighted capital benefits incentivize financial institutions to prioritize green financing. Sustainable finance regulations create a structured ecosystem, while central bank green initiatives promote liquidity and stability in the green financial market, enhancing investor confidence.

**1.4 Infrastructure & Technology Policies:** These policies focus on building the physical and technological foundation necessary for green investment. Investment in green infrastructure, such as renewable energy and sustainable transport, creates long-term opportunities. Research and development support encourages innovation in clean technologies. Clean technology promotion accelerates adoption, while grid modernization ensures efficient energy distribution, collectively strengthening the ecosystem for sustainable investment.

**1.5 Social & Awareness Policies:** Social and awareness policies aim to influence public perception and investor mindset toward sustainability. Public awareness programs educate individuals about environmental challenges and green investment opportunities. Education and skill development enhance the capacity to participate in sustainable finance. Stakeholder engagement ensures inclusivity, while promoting a sustainable culture encourages long-term behavioral change in investment practices.

## 2. Policy Implementation Mechanisms

- **Legislation & Regulation:** This mechanism ensures that all formulated policies are legally enforced through proper legislative frameworks. Regulatory authorities monitor compliance and impose penalties for violations, ensuring discipline within the financial system. Effective legislation creates a structured environment where sustainability guidelines are not optional but mandatory, thereby strengthening the credibility and impact of green finance initiatives.
- **Guidelines & Standards:** Guidelines and standards provide operational clarity for implementing policies. They define benchmarks for ESG compliance, reporting formats, and sustainable investment criteria. These standards reduce ambiguity and ensure uniformity across markets. By establishing clear expectations, they help investors and institutions align their strategies with sustainability goals, improving efficiency and comparability in green investments.
- **Incentives & Support:** Incentives and support mechanisms play a crucial role in encouraging participation in green investments. Financial incentives such as subsidies, tax benefits, and low-interest financing reduce investment risks and costs. Support systems also include technical assistance and advisory services. These measures make sustainable investments more accessible and attractive, especially for new and small investors.
- **Monitoring & Enforcement:** Monitoring and enforcement ensure that policies are effectively implemented and followed. Regulatory bodies track compliance through audits, reports, and performance indicators. Enforcement actions, such as penalties or restrictions, are applied in case of non-compliance. This mechanism maintains transparency, accountability, and trust in the financial system, ensuring that sustainability commitments are genuinely fulfilled.
- **Public-Private Collaboration:** Public-private collaboration enhances the effectiveness of policy implementation by combining government support with private sector efficiency. Governments provide policy direction and incentives, while private entities contribute capital, innovation, and execution capabilities. This collaboration accelerates green project development, reduces financial burden on public resources, and creates a more dynamic and efficient investment ecosystem.

## 3. Policy Influence Pathways

- **Reduced Regulatory Risk:** Clear and stable policies reduce uncertainty associated with regulatory changes, making green investments more predictable. When investors are confident about policy consistency, they are more willing to invest in long-term sustainable projects, thereby increasing overall market participation.
- **Lower Cost of Capital:** Policies such as subsidies, tax benefits, and green financing reduce the cost of capital for sustainable projects. Lower financial barriers encourage both institutional and individual investors to allocate funds toward green investments, enhancing capital flow into sustainable sectors.

- **Market Confidence Building:** Strong policy signals and transparent frameworks increase investor confidence in green markets. Confidence reduces perceived risks and encourages long-term commitment, which is essential for the growth and stability of sustainable financial systems.
- **Innovation & Technology Adoption:** Supportive policies promote research, development, and adoption of clean technologies. Innovation reduces operational costs and increases efficiency, making green investments more competitive and attractive compared to traditional investments.
- **Long-term Value Creation:** Green investments focus on sustainability and resilience, leading to long-term value creation. Policies encourage investors to prioritize future benefits over short-term gains, aligning financial decisions with sustainable development goals.
- **Sustainable Economic Growth:** By promoting environmentally responsible investments, policies contribute to overall economic growth that is both inclusive and sustainable. This ensures that economic development does not come at the cost of environmental degradation.

#### 4. Impact on Investor Behavior (Mediating Layer)

- **Improved Risk Perception:** Policies and market signals help investors better understand and evaluate the risks associated with green investments. Reduced uncertainty leads to more rational and confident decision-making.
- **Higher Trust in Green Projects:** Transparent policies and successful project outcomes build trust among investors. Increased trust encourages higher participation in sustainable investments.
- **Stronger Long-term Orientation:** Green investments promote a shift from short-term profit focus to long-term value creation, influencing investor mindset and strategy.
- **Increased Willingness to Invest:** Supportive policies and reduced risks enhance investor willingness to allocate funds toward green projects, increasing overall market participation.
- **Preference for Sustainable Options:** Investors increasingly prefer ESG-compliant and environmentally responsible investment options, reshaping market demand and investment trends.

#### 5. Investment Decision Outcomes

- **Invest in Green Projects:** Investors actively choose to allocate funds toward environmentally sustainable projects, contributing to green sector growth.
- **Selective / Conditional Investment:** Investors may adopt a cautious approach by investing selectively based on risk-return analysis and policy clarity.
- **Avoid / Delay Investment:** In cases of uncertainty or lack of confidence, investors may avoid or delay green investments, highlighting the importance of strong policy frameworks.

#### 6. Policy Outcomes

**6.1 Environmental Outcomes:** Green investments lead to reduced carbon emissions, resource conservation, biodiversity protection, and enhanced climate resilience, contributing to environmental sustainability.

**6.2 Economic Outcomes:** Economic benefits include job creation, increased green investments, global capital attraction, and improved cost efficiency.

**6.3 Social Outcomes:** Social outcomes include improved public health, inclusive growth, community well-being, and social equity.

**6.4 Financial Outcomes:** Financially, investors benefit from better risk-adjusted returns, improved asset valuation, and stronger portfolio resilience.

## 7. Cross-Cutting Enablers

Good governance, data availability, technical capacity, and international cooperation support the overall effectiveness of the framework.

## 8. Policy Monitoring & Feedback Loop

Continuous monitoring, evaluation, stakeholder feedback, and policy adjustments ensure that the framework remains dynamic and effective over time.

This process ensures a continuous cycle between policy action and market res

### 5.8.4 Framework Design (Conceptual Representation)

The NNF framework is designed as a dynamic network model where each component interacts with others in a non-linear and feedback-driven structure. It resembles a neural network system in which:

- Policy inputs act as external stimuli
- Investor cognition acts as processing nodes
- Financial markets act as output layers
- Feedback mechanisms act as learning adjustments

This design highlights the adaptive and self-learning nature of the framework, making it suitable for analyzing complex financial and policy environments.

### 5.8.5 Framework Summary (Integrated View)

The NNF framework provides a comprehensive integrated view of how policies influence financial systems and investor behavior in a sustainable development context. It combines structural layers, behavioral insights, and feedback mechanisms into a single system.

- Overall, the framework demonstrates that:
- Policy actions shape market behavior
- Investor cognition mediates decision-making
- Investment outcomes reflect both structural and psychological factors
- Continuous feedback improves policy effectiveness

Thus, the NNF framework serves as a holistic model for understanding and improving green investment systems through coordinated policy and behavioral dynamics.

## REFERENCES

This section lists all sources cited in the Project, following the **university-prescribed citation format**. The **APA (7th edition)** format is used for referencing in this study.

Key Guidelines for References (APA 7th Edition):

1. **Books:** Author(s). (Year). *Title of the Book* (Edition, if applicable). Publisher.
2. **Journal Articles:** Author(s). (Year). Title of the article. *Title of the Journal, Volume* (Issue), Page numbers. DOI/URL if available.
3. **Conference Proceedings:** Author(s). (Year). Paper title. *Conference Name*, Location, Pages. DOI if available.
4. **Reports and Official Documents:** Organization Name. (Year). *Title of the Report*. Publisher/Source. URL if applicable.
5. **Websites:** Author/Organization. (Year). *Title of the webpage/document*. URL.
6. **Theses/Dissertations:** Author. (Year). *Title of the dissertation/thesis* (Doctoral dissertation/Master's thesis, University). Database or URL.

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### 1. Books

1. . *Ahmed Imran Hunjra* (2024). THE Palgrave Handbook of Green Finance for Sustainable Development. Springer International Publishing
2. *Dayong Zhang* (2024). Climate Finance. Springer Nature Singapore

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### 2. Journal Articles

1. *Elbouknify, I., Machado, M. R., & Iannario, M.* (2026). Designing green artificial intelligence (Green AI) models for finance: A novel approach for sustainable and responsible adoption. *Financial Innovation*, 12 (96), 1–25. <https://doi.org/10.1186/s40854-026-00915-y>
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### 3. Conference Proceedings

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1. Global Sustainable Investment Alliance. (2021). Global Sustainable Investment Review. GSIA. <https://www.scirp.org/reference>

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#### 5. Websites and Online Sources

1. Asian Development Bank. (2022). Green Finance Strategies in Asia. <https://www.adb.org/sites>
2. Carbon Disclosure Project. (2022). Corporate Environmental Disclosure System. <https://cdn.cdp.net/cdp-production>
3. European Investment Bank. (2023). Climate Finance and Green Investment. <https://www.eib.org/attachments>
4. United Nations Development Programme. (2022). Sustainable Finance and Investment Framework. <https://www.un.org/ohrlls/sites>
5. World Economic Forum. (2023). Future of Sustainable Finance. <https://www3.weforum.org/>

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#### 6. Theses and Dissertations

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#### 7. Citation-Checking Guidelines

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##### Final Review and Formatting Tips:

- ✓ Ensure proper **alphabetical order** by the first author's last name.
- ✓ Use **hanging indentation** (first line flush left, subsequent lines indented).
- ✓ Cross-check **all in-text citations** with the reference list.
- ✓ For online sources, **DOIs or stable URLs** are preferred over general website links.

By following these guidelines, the reference section ensures **academic integrity and credibility** for the dissertation.

### APPENDICES

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The appendices section contains supplementary materials that support the research but are not included in the main body of the thesis. These materials provide additional evidence and documentation of the research process, including the questionnaire, interview transcripts, plagiarism report, ethical approval, conference presentations, and published research outputs.

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### A. Questionnaire/Survey Instrument

This section presents the **structured questionnaire** used for **primary data collection** in the study titled “Green Investment Behavior & Cognitive Bias Study (NNF Model)”. The instrument was designed using a combination of **Likert -scale items, multiple-choice questions, and open-ended responses** to capture comprehensive insights into investor behavior.

The questionnaire was specifically developed to examine **green investment awareness, risk perception, cognitive biases, decision-making patterns, and cross-border influences**. It was administered to respondents from diverse **demographic and professional backgrounds** to ensure **data validity and representativeness**.

#### Sample Questionnaire: Green Investment Behavior & Cognitive Bias Survey (NNF Model)

**TABLE 1: BASIC INFORMATION**

Q No.	Question	Options
Q1	Name	_____
Q2	Age Group	<input type="checkbox"/> 26–35 <input type="checkbox"/> 36–45 <input type="checkbox"/> 46+
Q3	Country	_____
Q4	Profession	<input type="checkbox"/> Banker <input type="checkbox"/> Policymaker <input type="checkbox"/> Business Owner <input type="checkbox"/> Other
Q5	Investment Experience	<input type="checkbox"/> 1–3 years <input type="checkbox"/> 3–5 years <input type="checkbox"/> More than 5 years

**TABLE 2: AWARENESS**

Q. No.	Question	Options
Q6	Are you aware of green/sustainable investments?	<input type="checkbox"/> Yes <input type="checkbox"/> No
Q7	How often do you invest in green projects?	<input type="checkbox"/> Never <input type="checkbox"/> Rarely <input type="checkbox"/> Occasionally <input type="checkbox"/> Often <input type="checkbox"/> Always

**TABLE 3: RISK PERCEPTION**

No.	Statement	1	2	3	4	5
Q8	Green investments are less risky than traditional investments	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>
Q9	Government policies influence my investment decisions	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>
Q10	Market uncertainty discourages me from investing in green projects	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>
Q11	I prefer safe investments over high-return green investments	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>

**TABLE 4: COGNITIVE BIAS (NNF MODEL)**

No.	Statement	1	2	3	4	5
Q12	I rely on past experiences in investment decisions	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>
Q13	I follow market trends or herd behavior	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>
Q14	Emotional factors affect my investment choices	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>
Q15	I avoid losses in investment decisions (loss aversion)	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>

**TABLE 5: DECISION MAKING**

Q No.	Statement	1	2	3	4	5
Q16.	I prefer long-term green investments over short-term gains	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>
Q17.	I consider environmental impact before investing	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>
Q18.	I change my decisions based on policy or regulation changes	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>
Q19.	Cultural factors influence my investment decisions	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>
Q20.	International market trends affect my investment behavior	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>
Q21.	I am willing to invest in foreign green projects	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>

**TABLE 6: OPEN-ENDED QUESTIONS**

- Q22. What factors motivate you to invest in green projects?  
 Q23. What challenges do you face in green investment?

**Final Notes on Questionnaire Design**

**Self-Developed Scale:**

This questionnaire was developed based on an extensive literature review covering green investment behavior, sustainable finance, and cognitive biases within the NNF model framework. Its validity was ensured through expert review prior to distribution.

**Adopted Items:**

Some items related to behavioral finance, risk perception, and cognitive biases were adapted from established theoretical foundations such as Prospect Theory and sustainable investment literature. Where necessary, appropriate academic references have been included.

✔ **Pre-Testing (Pilot Study):**

A pilot study was conducted on 20 respondents to assess the clarity, reliability, and comprehensibility of the questionnaire. Based on the feedback received, minor modifications were made to improve measurement accuracy and validity.

✔ **Final Data Collection:**

The questionnaire consisted of a total of 23 structured items, which were finally administered to 83 respondents. This sample provides sufficient representation for analyzing green investment behavior and cognitive bias-based decision-making processes.

✔ **Confidentiality Statement:**

All responses were collected with full confidentiality. Participants were assured that their data would be used solely for academic research purposes and their identities would remain anonymous. Note: This scale is used to convert responses into quantitative data for statistical analysis.

- 1 = Strongly Disagree / Never
- 2 = Disagree / Rarely
- 3 = Neutral / Occasionally
- 4 = Agree / Often
- 5 = Strongly Agree / Always

This questionnaire is designed for empirical analysis of Green Investment Behaviors, Cognitive Biases, and Decision-Making Patterns under the NNF Model framework

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### **B. Interview Transcripts (Not Applicable)**

This study did not involve any qualitative interviews. Therefore, no interview transcripts were collected or included in the research. The research is entirely based on quantitative data collected through a structured questionnaire.

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### **C. Plagiarism Report**

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### **D. Ethical Approval (If Required)**

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### **E. Reprints of Conference/Seminar Presentation Certificate**

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### **F. Reprints of Published Manuscripts/Articles/Research Papers in National & International Journals**

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### **Final Checklist for Appendices Section**

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This appendices section ensures that the **research process is well-documented, transparent, and academically robust**. It provides **evidence of originality, ethical compliance, and scholarly contributions**, making the Project a valuable academic resource.

### Acknowledgement

I would like to express my sincere gratitude to **Dr. Pushkar Dubey**, Assistant Professor and Head Department of Management, School of Commerce And Management Studies Pandit Sundarlal Sharma (Open) University Chhattisgarh Koni-Birkona, Bilaspur-495009 (Chhattisgarh), for providing valuable insights and a strong foundation regarding the preparation of an MBA research project. His guidance at the initial stage greatly helped in shaping the direction of this study.

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I would also like to extend my heartfelt thanks to my friends and colleagues who actively participated in the research survey. Their cooperation and valuable responses have contributed significantly to the academic purpose of this study.

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**TARUN DEWANGAN**

**M.B.A. 4<sup>th</sup> Semester**